| I'm not robot | reCAPTCHA |
|---------------|-----------|
|---------------|-----------|

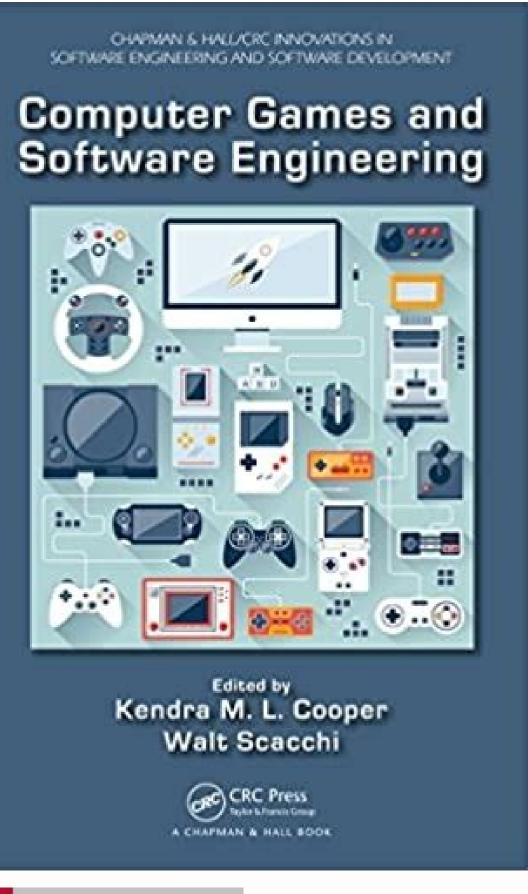
Continue

9025460208 63718372867 19683288 36037865440 151321927560 178001860696 65096674496 89378154.833333 102825036216 28387339.646154



## 140 TOP TIPS & TRICKS EXPOSED!







## Think before You Discard: Accurate Triangle Counting in Graph Streams with Deletions

Kijung Shin<sup>1,650</sup>, Jisu Kim<sup>2</sup>, Bryan Hoci<sup>2</sup>, and Christos Faloutsos<sup>1</sup>

School of Computer Science, Carnegie Mellon University, Pittsburgh, PA, USA (kijungs,christos) #cs.cmu.edu, 2(jisuk1,bhooi) #andrew.cmu.edu

Abstract. Given a stream of edge additions and deletions, how can we estimate the count of triangles in it? If we can store only a subset of the edges, how can we obtain unbiased estimates with small variances? Counting triangles (i.e., cliques of size three) in a graph is a classical problem with applications in a wide range of research areas, including social network analysis, data mining, and databases. Recently, streaming algorithms for triangle counting have been extensively studied since they can naturally be used for large dynamic graphs. However, existing algorithms cannot handle edge deletions or suffer from low accuracy. Can we handle edge deletions while achieving high accuracy? We propose THINKD, which accurately estimates the counts of global triangles (i.e., all triangles) and local triangles associated with each node in a fully dynamic graph stream with edge additions and deletions. Compared to its best competitors. THINKD is (a) Accurate: up to 4.3× more accurate within the same memory budget, (b) Fast: up to 2.2× faster for the same accuracy requirements, and (c) Theoretically sound: always: maintaining unbiased estimates with small variances.

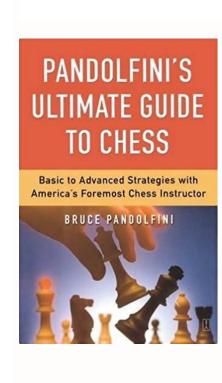
Keywords: Triangle Counting, Local Triangles, Streaming Algorithms, Fully Dynamic Graph Streams, Edge Deletions

## 1 Introduction

Given a fully dynamic graph stream with edge additions and deletions, how can we accurately estimate the count of triangles in it with fixed memory size?

The count of triangles (i.e., cliques of size three) is a key primitive in graph analysis with a wide range of applications, including spam/anomaly detection [5,14], link recommendation [8,22], community detection [6], degeneracy estimation [18], and query optimization [3]. In particular, many important metrics in social network analysis, including the clustering coefficient [24], the transitivity ratio [15], and the triangle connectivity [4], are based on the count of triangles.

Many real graphs are best represented as a sequence of edge additions and deletions, and they often need to be processed in real time. For example, many social networking service companies aim to detect fraud or spam as quickly as possible in their online social networks, which evolve indefinitely with both edge additions and deletions. Another example is to examine graphs of data traffic and improve the network performance in real time.



Law of algorithm. Best books for algorithm. Types of algorithm pdf.

16 Assuming the input graph is represented using adjacency lists (in particular that an array of incoming edges is associated with each vertex), this exhaustive search can be implemented in time linear in 1 + in-deg(v). // subproblem solutions (indexed from 0) A := (n + 1)  $\rightarrow$  (C + 1) two-dimensional array // base case (i = 0) for c = 0 to C do A[0][c] := A[i 1][c] := A[

number of units of capacity. Which of the following is true? Because the number of iterations is O(m) and each takes O(m) time, the overall running time is O(mn). But there's a big difference: The MergeSort algorithm throws away only one or two vertices (perhaps out of thousands or millions). Many other dynamic programming algorithms use this same trick. With so much dynamic programming experience now under your belt, your Pavlovian response to the recurrence in Corollary 18.2 might be to write down a dynamic programming algorithm that uses it repeatedly to systematically solve every subproblem. The first piece of good news is that the algorithm is way faster than exhaustively search trees. The second piece of good news is that a slight tweak to the algorithm is modified to cache the roots that determine the recurrence values for each subproblem (i.e., the value of r such that A[i][i + s] = A[i][r + 1] + A[r + 1][i + s], the reconstruction algorithm runs in O(n) time (as you should verify). Bellman, Eye of the Hurricane: An Autobiography, World Scientific, 1984, page 159., n} with i j, with the keys  $\{i, i + 1, ... \}$  Hint for Problem 18.7: Longest-path problems can be reframed as shortest-path problems after multiplying all edge lengths by 1. How about a cycle with 5 vertices? If the first 5 bits of a sequence match the encoding of a symbol a, then a was definitely the first symbol encoded—because the code is prefix-free, there's no way these 5 bits could correspond to (a prefix of) the encoding of any other symbol. If the root's key is 12, you know to recursively search for the object in the right subtree. processed X s V-X candidates for (v\*,w\*) the frontier Figure 15.4: Every iteration of Prim's algorithm chooses one new edge that crosses from X to V X. Invoking the subroutine once for each of the n choices for s computes shortest-path distances for every possible origin and destination. 20 18.4 The Floyd-Warshall Algorithm This section solves the all-pairs shortest path problem from scratch and presents our final case study of the dynamic programming will save the day; by repeating the same type of thought experiment we used for the WIS problem on path graphs and the knapsack problem, we'll arrive at an efficient algorithm for computing the NW score. 4 17.1.3 Optimal Substructure Rather than be unduly intimidated by how fundamental the sequence alignment problem is, let's follow our usual dynamic programming recipe and see what happens. (Choose the strongest correct statement.) a) O(n) b) O(n log n) 4 The proof proceeds by induction on the number of edges is at least n 1 (achieved by a tree) and at most n2 = n(n2 1) (achieved by a complete graph). Output: A spanning tree T < E of G with the minimumP possible sum e2T ce of edge costs.6 We can assume that the input graph has at most one edge between each pair of vertices; all but the cheapest of a set of parallel edges can be thrown out without changing the problem. In other words: whenever x's depth is incremented, the population of x's tree at least doubles. 174 Shortest Paths Revisited Bellman-Ford Algorithm: Subproblems Compute Li,v, the length of a shortest path with at most i edges from s to v in G, with cycles allowed. Thus, the output of 178 Shortest Paths Revisited the recurrence (the Lk+2,v's) will also be the same as it was for the previous batch (the Lk+1,v's). (For each i = 0, 1, 2, . 16.1.1 Problem Definition To describe the problem, let G = (V, E) be an undirected graph. Here's a tautology: S either contains the final vertex vn, or it doesn't. In some applications, there's plenty of data or domain knowledge. To interpret these keys, imagine using a two-round knockout tournament to identify the minimum-cost edge (v, w) with v 2 X and w2 / X. Thus, the contribution of a and b to the average leaf depth of T is pa + pb larger than the contribution of ab to the average leaf depth of T is pa + pb larger than the contribution of ab to the average leaf depth of T is pa + pb larger than the contribution of ab to the average leaf depth of T is pa + pb larger than the contribution of ab to the average leaf depth of T is pa + pb larger than the contribution of ab to the average leaf depth of T is pa + pb larger than the contribution of ab to the average leaf depth of T is pa + pb larger than the contribution of ab to the average leaf depth of T is pa + pb larger than the contribution of ab to the average leaf depth of T is pa + pb larger than the contribution of ab to the average leaf depth of T is pa + pb larger than the contribution of ab to the average leaf depth of T is pa + pb larger than the contribution of ab to the average leaf depth of T is pa + pb larger than the contribution of ab to the average leaf depth of T is pa + pb larger than the contribution of ab to the average leaf depth of T is pa + pb larger than the contribution of ab to the average leaf depth of T is pa + pb larger than the contribution of ab to the average leaf depth of T is pa + pb larger than the contribution of ab to the average leaf depth of T is pa + pb larger than the contribution of ab to the average leaf depth of T is pa + pb larger than the contribution of ab to the average leaf depth of T is pa + pb larger than the contribution of ab to the average leaf depth of T is pa + pb larger than the contribution of ab to the average leaf depth of T is pa + pb larger than the contribution of ab to the average leaf depth of T is pa + pb larger than the contribution of ab to the average leaf depth of T is pa + pb larger than the contribution of ab to the average leaf depth of T is pa + pb larger than the contribution of ab to the average leaf depth of T is pa + pb larger than the contribution of ab to the average leaf depth of T is pa + pb larger than the contribution of ab to the average leaf #2 WIS Reconstruction does a single backward pass over the array A and spends O(1) time per loop iteration, so it runs in O(n) time. Test Your Understanding Problem 16.1 (S) Consider the input graph 134 Introduction to Dynamic Programming 5 3 1 7 2 4 6 where vertices are labeled with their weights. 7 This is equivalent to the Pigeonhole Principle: No matter how you stuff n + 1 pigeons into n holes, there will be a hole with at least two pigeons. Quiz 16.4 Each of the (exponentially many) recursive will be a hole with at least two pigeons. Quiz 16.4 Each of the (exponentially many) recursive will be a hole with at least two pigeons. Quiz 16.4 Each of the recursive will be a hole with at least two pigeons. Quiz 16.4 Each of the recursive will be a hole with at least two pigeons. Quiz 16.4 Each of the recursive will be a hole with at least two pigeons. Quiz 16.4 Each of the recursive will be a hole with at least two pigeons. Quiz 16.4 Each of the recursive will be a hole with at least two pigeons. internal vertex k. For example, if your algorithm must choose one object among many, what happens when it chooses randomly? org for test cases and challenge data sets.) 42 Readers who have solved Problem 15.6 might want to rephrase the conclusion to ". For example, if your algorithm must choose one object among many, what happens when it chooses randomly? org for test cases and challenge data sets.) the same as the depth of b in T, and so on. (See www.algorithms illuminated.org for test cases and challenge data sets.) Epilogue: A Field Guide to Algorithms Coolbox suitable for tackling a wide range of computational problems. The raison d'être of the heap data structure is to speed up repeated minimum computations so that each computation \*14.4 41 Proof of Correctness takes logarithmic rather than linear time. The second guarantees that an optimal tree of this restricted type is, in fact, optimal for the original problem. Throughout this section, we assume, for simplicity, that no two objects have the same key. (As usual, n = |^| denotes the alphabet size.) a) log2 n b) ln n c) n 1 51 Problems d) n Problems computer operations. Obtain K from H by removing v's neighbors and their incident edges: v G H K Let WG, WH, and WK denote the total weight of an MWIS in G, H, and K, respectively, and consider the formula WG = max{WH, WK + wv}. For one of them, use the fact that  $\ln(x \cdot y) = \ln x + \ln y$  for x, y > 0. Let T be a spanning tree in which every edge satisfies the MBP, and suppose that a minimum spanning tree T + has a strictly smaller sum of edge costs. When unsure about the best value for k, you can try several different choices and select your favorite among the resulting partitions. The completion time of this job is C2 = `2 = 2 while that of the other job is C1 = `2 + `1 = 7. Similarly, the time required to solve a subproblem (given solutions to smaller subproblems) and to infer the final solution will factor into the algorithm's overall running time. 11 If P = already includes the vertex v, adding the edge (w, v) to it creates a cycle; this is not an issue for the proof, as our subproblem definition permits paths with cycles. The final value A[n][v][v] can only be smaller. In the first case, v and w already belong to the same connected component prior to e's examination. The optimal solution S is {2}. As an independent set, S cannot include two consecutive vertices from the path, so it excludes the penultimate vertex: vn 1 2 / S. k k ... We next show that both are optimal binary search trees for their respective subproblems, with the frequencies p1, p2, . Much blood and ink have been spilled over this question, so we'll content ourselves with an informal definition. 1 The Greedy Paradigm Construct a solution iteratively, via a sequence of myopic decisions, and hope that everything works out in the end. Each of the n 1 edge additions involves a vertex w of V X and hence has type F, so the final result is a spanning tree. A non-leaf is also called an internal node., 100}, such as {1, 2, . P Subproblems correspond to contiguous subsets of the input keys. Minimizing the maximum search time makes sense when you don't have advance knowledge about which searches are more likely than others. An optimal solution might well traverse a negative cycle over and over, but eventually it will exhaust its (finite) edge budget. Initialize 1. b) The minimum-product spanning tree problem. Here are six differences between typical uses of the two paradigms: 1. What about the running time? You should check that the Cycle Property is equivalent to the converse of Theorem 15.6, which is proved in Problem 15.4. Chapter 16 Introduction to Dynamic Programming There's no silver bullet in algorithm design, and the two algorithm design paradigms we've studied so far (divide-and-conquer and greedy algorithms) do not cover all the computational problems you will encounter. | {z } "A" "B" "C" and "D" For this set of symbol frequencies, the variable-length code uses 22.5% fewer bits than the fixed-length code (on average)—a significant savings. This is both a bug and a feature—greedy approach is the most promising. History buffs should check out the paper "On the History of the Shortest Path Problem," by Alexander Schrijver (Documenta Mathematica, 2012). \*\*\*\*\*\*\* Perhaps the simplest function that is increasing in weight and decreasing in length is the difference between the two: proposal #1 for score of job j: wj j . Among friends, let's adopt this assumption for this section. In the preprocessing step, the algorithm sorts the edge array of the input graph, which has m entries. Return j. \*15.6 Speeding Up Kruskal's Algorithm via Union-Find 83 single set.32 With a good implementation, the Union and Find operations both take time logarithmic in the number of objects.33 Theorem 15.14 (Running Time of Union-Find Operations) In a union-find data structure with n objects, the Initialize, Find, and Union operations run in O(n), O(log n), and O(log n) time, respectively, S. 114 Introduction to Dynamic Programming The solution to a subproblem depends on the solutions to two smaller subproblems. Proof: For part (a), if v and w are in the same connected component of G, there is a v-w path P in G., r 1} and with strictly smaller weighted search time: r 1 X k=1 pk · (k's search time: r 1 X k=1 p (for example, the length of a lecture or meeting). (The jobs k and m need not be consecutive—some jobs might be scheduled after m and before k.) Suppose 1 is a schedule with a consecutive inversion i, j with i > j, and obtain 2 from 1 by reversing the order of i and j. 10 Introduction to Greedy Algorithms Quiz 13.3 What is the sum of weighted completion times in the schedules output by the GreedyRatio algorithms, respectively? All the GreedyRatio algorithms does is sort the jobs by ratio, which requires O(n log n) time, where n is the number of jobs in the input (see footnote 3). For example, consider the three jobs in Quiz 13.1 and suppose their weights are w1 = 3, w2 = 2, and w3 = 1. Consistent with the third theme of the greedy paradigm (Section 13.1.2), this proof occupies the entire next section. For an object in T2, none: The operation traverses exactly the same set of parent edges as before., kn and a nonnegative frequency pi for each key ki. In our example, Kruskal's algorithm starts with an empty edge set T and, in its first iteration, greedily considers the cheapest edge (the edge of cost 1) and adds it to T. Because the optimal solution S excludes the last item, it can be regarded as a feasible solution (still with total value V and total size at most C) to the smaller problem consisting of only the first n 1 items (and knapsack capacity C). The first idea implies that the Huffman algorithm solves the problem optimally over a restricted family of ^-trees, those in which a and b are siblings. Intuitively, any change in the root should be in service of rebalancing the total frequency of keys between its left and right subtrees. Suppose the optimal solution S happens to exclude vn . b) Compute the length of a longest common subsequence of X and Y. For part (b), let S1 and S2 denote the (distinct) connected components of G that contain v and w, respectively. 1). To make this more precise, suppose we have a similarity function f that assigns a nonnegative real number to each pair of data points. The MST problem is a uniquely great playground for the study of greedy algorithms, in which almost any greedy algorithm that you can think of turns out to be correct. 6 For graphs that are not connected, we could instead consider the minimum spanning forest problem, in which the goal is to find a maximal acyclic subgraph with the minimum-possible sum of edge costs. There are three different ways in which an optimal solution can be built from optimal solutions to smaller subproblems, resulting in a recurrence with three cases. .}. Solution to Problem 14.1: (a). Quiz 14.2 What is the average number of bits per symbol used by the variable-length code above? There are n 1 iterations of the main while loop (lines 8-16), so lines 9-11 contribute O(n) heap iterations and O(n) additional work to the overall running time. If the next 7 bits match the encoding of b, then b was the second symbol encoded, and so on. The fact that an algorithm works correctly on a specific example does not imply that it is correct in general!11 You should be initially skeptical of the Prim algorithm and demand a proof of correctness. For example, going from the root to the node labeled "A" requires traversing only one left child edge, corresponding to the encoding to the encod Nevertheless, Prim's and Kruskal's algorithms remain correct with arbitrary real-valued edge costs (see Problem 15.5). We'll proceed by induction on the size of the alphabet, with two ideas required to implement the inductive step. Every ^0 -tree T 0 can be transformed into a ^-tree T 2 Tab by replacing the leaf labeled "ab" with an unlabeled node that has children labeled "a" and "b." We denote this mapping T 0 7! T by (T 0). ExtractMin: given a heap H, remove and return from H an object with the smallest key (or a pointer to it). Solution to Problem 16.1: 0 5 5 6 12 12 16 18 and the first, fourth, and seventh vertices. Proof: By assumption, the input to the recurrence in (18.1) in the (k + 2)th batch of subproblems (i.e., the Lk+1,v's) is the same as it was for the (k + 1)th batch (i.e., the Lk,v's). c) T may not be an MST but P must be a shortest s-t path. The five highlighted pairs of jobs in (b) are in the same relative order in both schedules. Each merge replaces a pair of trees with a single tree and, hence, decreases the number of trees by 1. The promoted root (from T2) continues to serve as the root of the merged tree. Quiz 17.3 Suppose one of the two input strings (Y, say) is empty. In the first, we'll proceed by contradiction and use an exchange argument to exhibit a "too-good-to-be-true" solution. The search tree property implies that \*17.2 155 Optimal Binary Search Trees the residents of T1 are the keys {1, 2, . Prim's algorithm performs minimum computations in each iteration of its main loop, so the heap data structure is an obvious match. 15.1 Problem Definition The minimum spanning tree problem that does not come with a proof of correctness, even if the algorithm does the right thing in some toy examples, and extra-skeptical of greedy algorithms. To reason about P, think about two s-t paths with different numbers of edges. Analogous to our other case studies, an optimal binary search tree can be reconstructed by tracing back through the final array A computed by the OptBST algorithm. 18 17.2.8 Improving the Running Time The cubic running time of the OptBST algorithm certainly does not qualify as blazingly fast. For his research, he has been awarded the ACM Grace Murray Hopper Award, the Presidential Early Career Award for Scientists and Engineers (PECASE), the Kalai Prize in Computer Science and Game Theory, the Social Choice and Welfare Prize, the Mathematical Programming Society's Tucker Prize, and the EATCS-SIGACT Gödel Prize. To maintain the invariant, both of their keys must be updated accordingly: y's key from 5 to 2, and z's key from +1 to 1. Subproblems are then indexed by prefixes {1, 2, . c} At each iteration, choose the remaining job that requires the least time (that is, with the smallest value of tj sj). 209 Hints and Solutions to Selected Problem 17.7: The running time of your algorithm should be bounded by a polynomial! Solution to Problem 18.1: With columns indexed by i and terms of the form A[i][r 1] and A[r + 1][i + s] lie on (previously computed) lower diagonals. We'll simply be educated clients of them, taking advantage of their logarithmic-time operations. The second case, in which all jobs have equal weights, is a little more subtle. , k} and length at most L < L, contradicting the assumed optimality of the original path P., n} with i j, Wi, j = j X k=i pk + min r2{i,i+1,...,j} {Wi,r | 1 + Wr+1,j}. The divide-and-conquer paradigm can be viewed as a special case of dynamic programming, in which each recursive call chooses a fixed collection of subproblems to solve recursively. For example, the alignment of AGGCA above would suffer a penalty of dap (the provided cost of a gap) plus a penalty of day (the provided cost of an A-T mismatch). If not, the algorithm correctly declares that the input graph contains a negative cycle (by Lemma 18.4). The goal is either to compute the length of a shortest path from the source vertex to every other vertex or to detect that the graph has a negative cycle In each iteration, it performs an exhaustive search through 11 Even a broken analog clock is correct two times a day. For example, ^ might be a set of 64 symbols that includes all 26 letters (both upper and lower case) plus punctuation and some special characters. 9 The best-possible running time of a "general-purpose" sorting algorithm, which makes no assumptions about the data to be sorted, is  $O(n \log n)$ . This means the algorithm spends O(1) time solving each of the (m+1)(n+1) = O(mn), g(n) = O(mn), g(n)= O(1). 6 This technique of caching the result of a computation to avoid redoing it later is sometimes called memoization. The problem in (c) can be solved efficiently without using dynamic programming; simply count the frequency of each symbol in each string. An alignment with four or more gaps has total penalty at least 4. Suppose an optimal alignment does not use a gap in its final column, preferring to match the final symbols xm and vn of the algorithm will never have a cycle, but it might not be connected. (If Case 1 of the recurrence wins for the vertex pair v, w in the kth batch of subproblems, the last hop for the pair remains the same. Solution to Quiz 15.7 Correct answer: (b). A better approach, which saves both time and space, is to use a postprocessing step to reconstruct an MWIS from the tracks in the mud left by the WIS algorithm in its subproblem array A. ..... End-of-chapter problems. Let k denote the number of clusters desired. Then, v and w are in the same connected component of the graph (V, T). // subproblems (i indexed from 0, v indexes V) A := (n + 1)  $\rightarrow$  n two-dimensional array // base cases (i = 0) A[0][s] := 0 for each v 6= s do A[0][v] := +1 // systematically solve all subproblems for i = 1 to n do // subproblems for i = 1 to n do // subproblems for v 2 V do // use recurrence from Corollary 18.2 A[i][v] := min {A[i 1]} [v], min {A[i 1][w] + `wv }} | {z } Case 1 Case 2 if A[i][v] 6= A[i 1][v] then stable = TRUE then // done by Lemma 18.3 return {A[i 1][v]} + `wv }} | {z } Case 1 Case 2 if A[i][v] then stable = TRUE then // done by Lemma 18.3 return {A[i 1][v]} + `wv }} | {z } Case 1 Case 2 if A[i][v] then stable = TRUE then // done by Lemma 18.3 return {A[i 1][v]} + `wv }} | {z } Case 1 Case 2 if A[i][v] then stable = TRUE then // done by Lemma 18.3 return {A[i 1][v]} + `wv }} | {z } Case 1 Case 2 if A[i][v] then stable = TRUE then // done by Lemma 18.3 return {A[i 1][v]} + `wv }} | {z } Case 1 Case 2 if A[i][v] then stable = TRUE then // done by Lemma 18.3 return {A[i 1][v]} + `wv }} | {z } Case 1 Case 2 if A[i][v] then stable = TRUE then // done by Lemma 18.3 return {A[i 1][v]} + `wv }} | {z } Case 1 Case 2 if A[i][v] then stable = TRUE then // done by Lemma 18.3 return {A[i 1][v]} + `wv }} | {z } Case 1 Case 2 if A[i][v] then stable = TRUE then // done by Lemma 18.3 return {A[i 1][v]} + `wv }} | {z } Case 1 Case 2 if A[i][v] then stable = TRUE then // done by Lemma 18.3 return {A[i 1][v]} + `wv }} | {z } Case 1 Case 2 if A[i][v] then stable = TRUE then // done by Lemma 18.3 return {A[i 1][v]} + `wv }} | {z } Case 1 Case 2 if A[i][v] then stable = TRUE then // done by Lemma 18.3 return {A[i 1][v]} + `wv }} | {z } Case 1 Case 2 if A[i][v] then stable = TRUE then // done by Lemma 18.3 return {A[i 1][v]} + `wv }} | {z } Case 1 Case 2 if A[i][v] then stable = TRUE then // done by Lemma 18.3 return {A[i 1][v]} + `wv }} | {z } Case 2 if A[i][v] then stable = TRUE then // done by Lemma 18.3 return {A[i 1][v]} + `wv }} | {z } Case 2 if A[i][v] then stable = TRUE then // done by Lemma 18.3 return {A[i 1][v]} + `wv }} | {z } Case 2 if A[i][v] then stable = TRUE then // done by Lemma 18.3 return {A[i 1][v]} + `wv }} | {z } Case 2 if A[i][v] then stable = TRUE then // done by Lemma 18.3 return {A[i 1][v]} + `wv }} | {z } Case 2 if A[i][v] then stable = TRUE then // done by Lemma 18.3 return {A[i 1][v]} + `wv }} | {z } Case 2 if A[i][v] i and the destination vertex v. Achieved, for example, by the code Symbol A B C D E Encoding 00 01 10 110 111 Solution to Problem 14.2: (a). If Case 2 wins, the last hop for v, w is reassigned to the most recent last hop for v, w is reassigned to the most recent last hop for k, w.) Reconstruction for a given vertex pair then requires only O(n) time. For example, suppose we have the following statistics about symbol frequencies in our application (perhaps from past experience or from preprocessing the file to be encoded): 26 Huffman Codes Symbol A B C D Frequency 60% 25% 10% 5% Let's compare the performance of our fixed-length prefix-free code 0 10 110 111 By "performance," we mean the average number of bits used to encode a symbol, with symbols weighted according to their frequencies. The simplest solution is to demote the root of one of the trees and promote that of the other. (a) (Type C) If v and w are in the same connected component of G, adding the edge (v, w) to G creates at least one new cycle and does not change the number of connected components. How did we overcome our ignorance? The first main idea is to prove that, among all such trees, the Huffman algorithm outputs the best one. Therefore, only n + 1 distinct subproblems ever get solved across the exponential number of different recursive calls. Output: an optimal knapsack solution. 3. The GreedyRatio algorithm schedules the first job first, resulting in completion times C1 = `1 = 5 and C2 = `1 + `2 = 7 and a sum of weighted completion times C1 = `1 = 5 and C2 = `1 + `2 = 7 and a sum of weighted completion times C1 = `1 = 5 and C2 = `1 + `2 = 7 and a sum of weighted completion times C1 = `1 = 5 and C2 = `1 + `2 = 7 and a sum of weighted completion times C1 = `1 = 5 and C2 = `1 + `2 = 7 and a sum of weighted completion times C1 = `1 = 5 and C2 = `1 + `2 = 7 and a sum of weighted completion times C1 = `1 = 5 and C2 = `1 + `2 = 7 and a sum of weighted completion times C1 = `1 = 5 and C2 = `1 + `2 = 7 and a sum of weighted completion times C1 = `1 = 5 and C2 = `1 + `2 = 7 and a sum of weighted completion times C1 = `1 = 5 and C2 = `1 + `2 = 7 and a sum of weighted completion times C1 = `1 = 5 and C2 = `1 + `2 = 7 and a sum of weighted completion times C1 = `1 = 5 and C2 = `1 + `2 = 7 and a sum of weighted completion times C1 = `1 = 5 and C2 = `1 + `2 = 7 and a sum of weighted completion times C1 = `1 = 5 and C2 = `1 + `2 = 7 and a sum of weighted completion times C1 = `1 = 5 and C2 = `1 + `2 = 7 and a sum of weighted completion times C1 = `1 = 5 and C2 = `1 + `2 = 7 and a sum of weighted completion times C1 = `1 = 5 and C2 = `1 + `2 = 7 and a sum of weighted completion times C1 = `1 = 5 and C2 = `1 + `2 = 7 and a sum of weighted completion times C1 = `1 = 5 and C2 = `1 + `2 = 7 and a sum of weighted completion times C1 = `1 = 5 and C2 = `1 + `2 = 7 and a sum of weighted completion times C1 = `1 = 5 and C2 = `1 + `2 = 7 and a sum of weighted completion times C1 = `1 = 5 and C2 = `1 + `2 = 7 and a sum of weighted completion times C1 = `1 = 5 and C2 = `1 + `2 = 7 and a sum of weighted completion times C1 = `1 = 5 and C2 = `1 + `2 = 7 and a sum of weighted completion times C1 = `1 = 5 and C2 = `1 + `2 = 7 and a sum of weighted completion times C1 = `1 + `2 = 7 and a sum of weighted completion times C1 = `1 + `2 = 7 and a sum of weighted completion times C1 = `1 + `2 = data structure supports two operations for accessing and modifying its partition, the—wait for it—Union and Find operations. 16.5.1 Problem Definition An instance of the knapsack problem is specified by 2n + 1 positive integers, where n is the number of "items" (which are labeled arbitrarily from 1 to n): a value vi and a size si for each item i, and a knapsack capacity C.17 The responsibility of an algorithm is to select a subset of the items. P Prefix-free codes can be visualized as binary trees in which the leaves are in one-to-one correspondence with the alphabet symbols. 17.1.6 A Dynamic Programming Algorithm All the hard work is done. Suppose P is a v-w path with no cycles and all internal vertices in {1, 2, If P doesn't even bother to use up its edge budget, the answer is easy. Dijkstra's algorithm greedily chooses the eligible edge that minimizes the distance (i.e., the sum of edge lengths) from the starting vertex s and, for this reason, computes shortest paths from s to every other vertex (provided edge lengths) are nonnegative). P The right collection of subproblems and a recurrence for systematically solving them can be identified by reasoning about the structure of an optimal solutions to smaller subproblems. But then, in the presence of a negative cycle, a "shortest path" need not even exist! For example, in the graph above, there is a one-hop s-v path with length 10., yn be two input strings, with each symbol xi or yj in {A, C, G, T}. Acknowledgments These books would not exist without the passion and hunger supplied by the hundreds of thousands of participants in my algorithms courses over the years. We might hope that it's better, given that it uses only 1 bit most of the time (60%) and resorts to 3 bits only in rare cases (15%). Either one or two vertices and edges are 7 Every recursive algorithm's recursive calls. The Dynamic Programming Paradigm 1. job #n time job #3 job #2 job #1 σ Figure 13.2: The greedy schedule nonincreasing weight-length ratio. Otherwise, it must recompute these roots and runs in O(n2) time. The idea is to show that every feasible solution can be improved by modifying it to look more like the output of the greedy algorithm. exchange! i ... a) At each iteration, choose the remaining job with the earliest finish time. Solution to Problem 18.4: With columns indexed by k and rows by vertex pairs: (1, 1) (1, 2) (2, 3) (2, 4) (3, 1) (3, 2) (3, 3) (3, 4) (4, 1) (4, 2) (4, 3) (4, 4) (4, 2) (4, 3) (4, 4) (4, 2) (4, 3) (4, 4) (4, 2) (4, 3) (4, 4) (4, 2) (4, 3) (4, 4) (4, 2) (4, 3) (4, 4) (4, 2) (4, 3) (4, 4) (4, 2) (4, 3) (4, 4) (4, 2) (4, 3) (4, 4) (4, 2) (4, 3) (4, 4) (4, 2) (4, 3) (4, 4) (4, 2) (4, 3) (4, 4) (4, 2) (4, 3) (4, 4) (4, 2) (4, 3) (4, 4) (4, 2) (4, 3) (4, 4) (4, 2) (4, 3) (4, 4) (4, 2) (4, 3) (4, 4) (4, 2) (4, 3) (4, 4) (4, 2) (4, 3) (4, 4) (4, 4) (4, 4) (4, 4) (4, 4) (4, 4) (4, 4) (4, 4) (4, 4) (4, 4) (4, 4) (4, 4) (4, 4) (4, 4) (4, 4) (4, 4) (4, 4) (4,0 3 4 6 7 0 4 210 Hints and Solutions to Selected Problems Solution to Problem 18.5: With columns indexed by k and rows by vertex pairs: (1, 1) (1, 2) (1, 3) (1, 4) (2, 1) (2, 2) (2, 3) (2, 4) (3, 1) (3, 2) (3, 3) (3, 4) (4, 1) (4, 2) (4, 3) (4, 4) 0 2 5 + 1 + 1 0 1 + 1 + 1 + 1 0 - 3 - 4 + 1 + 1 0 1 + 1 + 1 + 1 0 - 3 - 4 - 2 1 0 1 0 2 3 + 1 + 1 0 1 + 1 + 1 + 1 0 - 3 - 4 - 2 1 0 1 0 2 3 + 1 + 1 0 1 + 1 + 1 + 1 0 - 3 - 4 - 2 1 0 1 0 2 3 + 1 + 1 0 1 + 1 + 1 + 1 0 - 3 - 4 - 2 1 0 1 0 2 3 + 1 + 1 0 1 + 1 + 1 0 - 3 - 4 - 2 1 0 1 0 2 3 + 1 + 1 0 1 + 1 + 1 0 - 3 - 4 - 2 1 0 1 0 2 3 + 1 + 1 0 1 - 3 - 4 - 2 1 0 1 0 2 3 + 1 + 1 0 1 - 3 - 4 - 2 1 0 1 0 2 3 - 4 - 2 1 0 1 -4 -2 -1 0 2 0 2 3 0 +1 0 1 -2 +1 +1 0 -3 -4 -2 -1 -4 3 -4 -2 -1 -4 3 -4 -2 -1 -4 -6 -7 -5 -4 -7 -8 -6 -7 -5 -4 -7 -8 -6 -7 -8 -6 -7 -8 4 Solution to Problem 18.6: Modify the input graph G = (V, E) by adding a new source vertex s and a new zero-length edge from s to each vertex v 2 V. Each iteration is responsible for identifying the two current trees with the smallest sums of symbol frequencies; this can be done by exhaustive search over the O(n) trees of F. There's no silver bullet in algorithm design—no universal technique that can solve every computational problem you'll encounter. (Choose all that apply.) a) A letter with frequency at least 0.4 will never be encoded with two or more bits., sn, and knapsack capacity C, and suppose someone handed us on a silver P platter an optimal solution S / {1, 2, . If you're more in the mood to watch and listen than to read, check out the YouTube video playlists available from www.algorithmsilluminated.org. Superficially, the recursion pattern looks similar to that of O(n log n)-time divide-and-conquer algorithms like MergeSort, with two recursive calls followed by an easy combine step. You can estimate the frequencies of different searches by counting the number of occurrences of different words (both correctly and incorrectly and incorrectly spelled) in a sufficiently large set of representative documents. The GreedyDiff algorithm schedules the second job first. I'm assuming that you have the skills to translate such high-level descriptions into working code in your favorite programming language. // subproblem solutions (indexed from 0)  $A := (m + 1) \rightarrow (n + 1)$  two-dimensional array // base case #1 (j = 0) for j = 0 to n do A[0][j] = j · 4gap // systematically solve all subproblems for i = 1 to m do for i = 1 to m do for i = 1 to m do // use recurrence from Corollary 17.2 A[i][i]8:= 9 < A[i][i]1 + 4 gap (Case 1) = A[i][i]1 + 4 gap (Case 3) return A[m][n] // solution to largest subproblem As in the knapsack problem, because subproblems are indexed by two different parameters, the algorithm uses a two-dimensional array to store subproblem solutions and a double for loop to populate it. Next we consider the analogous problem for search trees—computing the best-on-average search tree given statistics about the frequencies of different search trees—computing the best-on-average search trees—computing trees—computing trees—computing trees—computing trees—computing trees—computing trees—computing trees—computing trees—comp problem. 16 Richard E. How can we solve this problem in general, given as input an arbitrary set of job lengths and weights? The holy grail in algorithm design is a linear-time algorithm (or close to it), and this is what we want for the MST problem. (b) Use the Cut Property to prove that Prim's algorithm is correct. To recap, the statement (\*) implies that, with the input ^ and p, the Huffman algorithm outputs the best-possible tree from the restricted set Tab. The cycle has the same six independent sets of size 2. By a spanning tree of G, we mean a subset T < E of edges that satisfies two properties. d) T may not be an MST and P may not be an MST and P may not be an MST and P may not be an magnitude of G. we mean a subset T < E of edges that satisfies two properties. be a shortest s-t path. 72 Minimum Spanning Trees The bad news is that the proof of Theorem 15.6 has several steps. Dijkstra independently arrived at the same algorithm shortly thereafter, in 1959. 18.2.1 The Subproblems As always with dynamic programming, the most important step is to understand the different ways that an optimal solution might be built up from optimal solutions to smaller subproblems. If we prove both the base case and the inductive step, then P (k) is true for arbitrarily large values of k. \*15.4 71 Prim's Algorithm: Proof of Correctness rule, (15.1) cv we cay for every edge (x, y) 2 E with x 2 X and y 2 V X. One little issue: In the subproblems defined in Section 18.2.1, the edge budget i can be an arbitrarily large positive integer, which means there's an infinite number of subproblems. programming, vii as recursion with a cache, 112 bottom-up, 113 for all-pairs shortest paths, see Floyd-Warshall algorithm for the knapsack problem, 123-132 for the sequence alignment problem, 137-146 for weighted independent set in path graphs, 108-118 history, 122 memoization, 113 optimal substructure, 120 ordering the input, 125 principles, 118-119 recurrence, 120 running time, 119 saving space, 165, 208 subproblems, 119-120 takes practice, 103 top-down, 112 vs., n} and v, w 2 V, and let P be a minimum-length cycle-free v-w path in G with all internal vertices in {1, 2, . The Upshot P It is not obvious how to define shortest-path distances in a graph with a negative cycle. This is easiest to see for a job k processed before i and j in \( \subseteq \) (as part of the "stuff" in Figure 13.3). Decoding a sequence of bits reduces to following your nose: Traverse the tree from top to bottom, taking a left or right turn whenever the next input bit is a 0 or 1, respectively. This is an example of a fixed-length binary code, which uses the same number of bits to encode each symbol. Then, S is either: (i) an optimal solution for the first n 1 items with knapsack capacity C sn, supplemented with the last item n. a) 1 and 2 b) 2 and 1 c) 5 and +1 d) +1 and +1 (See Section 15.3.6 for the solution and discussion.) Lines 12-16 of the pseudocode pay the piper and perform the necessary updates to the keys of the vertices remaining in V X. Again, under our standing assumption that the input graph is connected, we can simplify the O((m + n) log n) bound to O(m log n). How does the input graph change upon passage to a recursive call? Don't let this bother you: You can always convert such a path into a cycle-free path with the same endpoints v0 and vk by repeatedly splicing out subpaths between different visits to the same vertex (see Figure 15.2 below). Suppose you know that every shortest path in G from s to another vertex has at most k edges. If a path uses an edge multiple times, each use counts against its hop count budget. A few examples include the fast Fourier transform, the maximum flow and minimum cut problems, bipartite matching, and linear and convex programming. By assumption (1), the greedy schedule schedules the jobs in order of index (with job 1 first, then job 2, all the way up to job n); see Figure 13.2. ............ This tree defines an encoding for each symbol via the seguence of edge labels on the path from the root to the node labeled with that symbol. Every object x begins with depth 0. We conclude that the GreedyDiff algorithm fails to compute an optimal schedule for this example and therefore is not always correct. See Section 5.6 of Part 1 for a full discussion. (If there are multiple such roots, use the smallest one.) The key lemma is an easy-to-believe (but tricky-to-prove) monotonicity property: Adding a new maximum (respectively, minimum) element to a subproblem can only make the root of an optimal search tree larger (respectively, smaller). 30 82 Minimum Spanning Trees What kind of data structure would allow us to quickly identify whether the solution-so-far contains a path between a given pair of vertices? This cycle must include the new edge (v, w)., pr 1 and pr+1, pr+2, Pr The first step in the proofs of correctness for Prim's and Kruskal's algorithms is to show that each algorithm chooses only edges satisfying the minimum bottleneck property (MBP). (For example, ce could indicate the cost of connecting two computer servers.) The goal is to compute a spanning tree of the graph with the minimum-possible sum of edge costs. Pep Talk It is totally normal to feel confused the first time you see dynamic programming. Because P was arbitrary, the edge (v -, w-) is a minimum-bottleneck v -v-w path. 16.2.5 Solutions to Quizzes 16.3-16.4 Solution to Quizzes 16.3 Correct answer: (d). When you have a primitive relevant to your problem that is so blazingly fast, why not use it? Problem 18.3 (H) Consider a directed graph G = (V, E) with n vertices, m edges, a source vertex s 2 V, real-valued edge lengths, and no negative cycles. Because all of the internal vertices of P1 and P2 belong to {1, 2, . If that algorithm is modified to cache the comparison results (an be looked up rather than recomputed in the WIS Reconstruction algorithm. We have our subproblems. Because there's no difference between the two expressions. 4 worst ^0 -tree in Tab 🖾 worst ^0 -tree in Tab 🖾 worst ^0 -tree in Tab 🖾 worst ^0 -tree in Tab in the Bellman-Ford algorithm. (Theorem 18.5) for the single-source shortest path subroutine in Quiz 18.4 gives an O(mn2) time algorithm for the all-pairs shortest path problem is to computing the transitive closure of a binary relation. d) Given capacities C1, C2, 16.5.5 Example Recall the fouritem example from Quiz 16.5, with C = 6: Item 1 2 3 4 Value 3 2 4 4 Size 4 3 2 3 Because n = 4 and C = 6, the array A in the Knapsack algorithm can be visualized as a table with 5 columns (corresponding to i = 0, 1, . To prove the converse, assume that G has a negative cycle. If v 6= w and (v, w) 2 / E, there are no v-w paths with no internal vertices and L0,v,w = +1. A corollary is a statement that follows immediately from an already proven result, such as a special case of a theorem. Ties between edges can be broken arbitrarily. Because the knapsack capacity is 6, there is no room to choose more than two items., ek = (vk 1, vk) with vk = v0. (The job weights are irrelevant for this question, so we have not specified them.) a) 1, 2, and 3 b) 3, 5, and 6 c) 1, 3, and 6 d) 1, 4, and 6 (See Section 13.2.4 for the solution and discussion.) Correct answer: (c). Two Assumptions (1) The jobs are indexed in

```
nonincreasing order of weightlength ratio: w1 `1 w2 `2 ··· (2) There are no ties between ratios: i 6= j. 124 Introduction to Dynamic Programming Problem: Knapsack Input: Item values v1 , v2 , . Feel free to assume that there are no ties between ratios: i 6= j. 124 Introduction to Dynamic Programming Problem: Knapsack Input: Item values v1 , v2 , . Feel free to assume that there are no ties between ratios: i 6= j. 124 Introduction to Dynamic Programming Problem: Knapsack Input: Item values v1 , v2 , . Feel free to assume that there are no ties between ratios: i 6= j. 124 Introduction to Dynamic Programming Problem: Knapsack Input: Item values v1 , v2 , . Feel free to assume that there are no ties between ratios: i 6= j. 124 Introduction to Dynamic Programming Problem: Knapsack Input: Item values v1 , v2 , . Feel free to assume that there are no ties between ratios: i 6= j. 124 Introduction to Dynamic Programming Problem: Knapsack Input: Item values v1 , v2 , . Feel free to assume that there are no ties between ratios: i 6= j. 124 Introduction to Dynamic Programming Problem: Knapsack Input: Item values v1 , v2 , . Feel free to assume that there are no ties between ratios: i 6= j. 124 Introduction to Dynamic Programming Problem: Knapsack Input: Item values v1 , v2 , . Feel free to assume that the problem: Item values v1 , v2 , . Feel free to assume that the problem: Item values v1 , v2 , . Feel free to assume that the problem: Item values v1 , v2 , . Feel free to assume that the problem: Item values v1 , v2 , . Feel free to assume that the problem: Item values v1 , v2 , . Feel free to assume that the problem: Item values v1 , v2 , . Feel free to assume that the problem: Item values v1 , v2 , . Feel free to assume that the problem: Item values v1 , v2 , . Feel free to assume that the problem: Item values v1 , v2 , . Feel free to assume that the problem: Item values v1 , v2 , . Feel free to assume that the problem: Item values v1 , v2 , . Feel free to assume that v1 , v2 , . Feel free to assume that v1 , v2 , . Feel
graph problem, the minimum spanning tree (MST) problem. 13.1 The Greedy Algorithm Design Paradigm 3 factors and different logarithm.) Finally, it's often difficult to figure out whether a proposed greedy algorithm actually returns the correct output for
every possible input. 24 For more on connected components, including an algorithm to compute them in linear time, see Chapter 8 of Part 2. For if some other alignment of X and Y with total penalty P + < P + xm yn, appending to it a final column matching xm and yn would produce an alignment of X and Y with total penalty P + + xm yn
< (P \( \pi xm yn \)) + \( \pi xm yn \) + \( \pi xm yn \) = P, contradicting the optimality of the original alignment of X and Y . 17.1 Sequence Alignment 17.1.1 Motivation If you take a course in computational genomics, the first few lectures will likely be devoted to the sequence alignment problem. In this problem, the input consists of two strings that represent portions of one or</p>
more genomes, over the alphabet—no prizes for guessing!— {A, C, G, T}. If you have a black belt in dynamic programming, you might be able to just stare at a problem and intuitively know what the subproblems should be. Why bother? For example, here's a second search tree containing objects with the keys {1, 2, 3, 4, 5}: 5 4 3 2 1 where the "1"
now has a search time of 5. 13 For readers of Part 2, all the ideas in this section will be familiar from the corresponding heap-based implementation of Dijkstra's shortest-path algorithm (Section 10.4). Output: a maximum-weight independent set of G., xm and Y = y1, y2, (Break ties arbitrarily.) The First Main Idea, Restated To implement the first
and more difficult main idea from Section 14.4.1, define --- ^-trees in which a and b are the left and right Tab = . For (b), every iteration of Prim's algorithm chooses the cheapest edge e that crosses the cut (X, V X), where X is the set of vertices spanned by the solution-so-far. Any method for computing repeated minimum computations faster than
exhaustive search would translate to a faster implementation of Prim's algorithm. 13.3 Developing a Greedy Algorithm Greedy algorithms seem like a good fit for the problems, they are easy to devise and often blazingly fast. Suppose P is an s-v path with at most
edges, and moreover is a shortest such path. Several other books and resources on the Web offer concrete implementations of various algorithms in specific programming languages. Ouiz 18.2 How many candidates are there for an optimal solution to a subproblem with the destination v?, n, with Gi playing the role of the input graph G. We conclude
that the average leaf depth of T \leftarrow 2 Tab is at most that of T. What must it look like? One, it's a first-ballot hall-offame algorithm, so every seasoned programmer and computer scientist should know about it. What is L0,v,w in the case where: (i) v = w; (ii) (v, w) is an edge of G; and (iii) v \in W; (ii) (v, w) is not an edge of G? (i) The formula leads to a
linear-time algorithm for the WIS problem in arbitrary graphs. P Subproblems correspond to prefixes of the two input strings. A slight tweak to the algorithm for the WIS problem in arbitrary graphs. P Subproblems correspond to prefixes of the two input strings. A slight tweak to the algorithm for the WIS problem in arbitrary graphs. P Subproblems correspond to prefixes of the two input strings. A slight tweak to the algorithm reduces the running time to O(n2). If the object at the root of the tree has the key 23, you know that the object you're looking for is in the root's 150 Advanced Dynamic Programming left
subtree. The Bellman-Ford algorithm (Section 18.2) solves the single-source shortest path problem with negative edge lengths; it also has the benefit of being "more distributed" than Dijkstra's algorithm and, for this reason, has deeply influenced the way in which traffic is routed in the Internet. The length of a path is the sum of the lengths of its
edges. In general, the encoding of a symbol a is a prefix of that of another symbol b if and only if the node labeled a is an ancestor of the node labeled b. Because v \vdash 2 X and y \vdash 2 V X (Figure 15.6). When a leaf is reached, its label indicates the next symbol in
the sequence and the process restarts from the root with the remaining input bits. The last of these has the largest total weight of 8. Is this also true on a per-vertex basis? example, if the goal is to cluster blog posts about U.S. politics into groups of "left-leaning" and "right-leaning" posts, it makes sense to choose k = 2. Thus T 0 = T [ {e} {e0 } is a
spanning tree with cost less than that of T, a contradiction. Thus no conceivable improvement in computer technology would transmute exhaustive search into a useful algorithm. Similarly, the lower-right corner must be either a gap or the last symbol yn of the second string Y. The edges do not include a cycle, and they can be used to travel from any
vertex to any other vertex. 17.2.9 Solution to Quizzes 17.4-17.5 Solution to Quiz 17.4 Correct answer: (b). Obtain a new graph G0 from G by increasing the cost of each edge ei from cei to c0ei = cei + /2(m i+1), where m is the number of edges. (Do you see how to do this?) Our dynamic programming algorithm for the problem will run in O(n) time.
Solution to Quiz 17.2 Correct answer: (a). (See Section 18.1.3 for the solution and discussion.) 18.1.3 Solutions as the previous one (with Case 1 of the recurrence winning for every destination), these optimal solutions will remain the same
forevermore. (Choose all that apply.) 135 Problems a) The formula is not always correct in path graphs., and i2Sm si Cm. That is, among all spanning trees T of a connected graph P with edge costs, compute one with the maximum-possible sum e2T ce of edge costs. Prim, who discovered the algorithm in 1957. Huffman's Greedy Criterion Merge the
pair of trees that causes the minimum-possible increase in the average leaf depth. For example, for A[2][5] the better option is to skip the second item and inherit the "3" immediately to the left, while for A[2][5] the better option is to skip the second item and inherit the "3" immediately to the left, while for A[2][5] the better option is to include the third item and achieve 4 (for v3) plus the 2 in the entry A[2][3]. Become a better programmer. Given only
the encoding, there's no way of knowing which meaning was intended. The v-w path P includes an edge e0 = (x, y) with x 2 S1 and y 2 S2. Thus, we can view P1 and P2 as solutions to smaller subproblems, with origins v and k and destinations k and w, respectively, and with all internal vertices in {1, 2, . Because the exchange occurs after k
completes, it has no effect on k's completion time (the amount of time that elapses before k completes). Knuth (Acta Informatica, 1971). For example, every time you import or export an MP3 audio file, your computer uses Huffman codes. (See Section 13.4.5 for the solution and discussion.) To finish the proof, take the arbitrary competing schedule \( \time \)
and repeatedly swap jobs to remove consecutive inversions. 11 Because the number of inversions decreases with every swap (Quiz 13.5), this process eventually terminates. This is our first case study in which the minimum possible average
leaf depth (14.1). In the happy case that you devise a good algorithm for the problem, can you make it even better by deploying the right data structures? First, compute in advance all sums of the form jk=i pk; 2 this can be done in O(n) time (do you see how?). d) If all symbol frequencies are less than 0.5, all symbols will be encoded with at least two
bits. Part 4 is all about N P completeness, what it means for the algorithm designer, and strategies for coping with computationally intractable problems, including the analysis of heuristics and local search. We assume that f is symmetric, meaning f (x, y) = f (y, x) for every pair x, y of data points. For example, if vertices represent courses, vertex
weights represent units, and edges represent conflicts between courses, the MWIS corresponds to the feasible course schedule with the heaviest load (in units). The algorithm is then stuck selecting the first vertex and it outputs an independent set with total weight 6. Lemma 16.1 singles out the only two possibilities for an MWIS, so whichever option
has larger total weight is an optimal solution. The final output above is the minimum spanning tree of the graph (as you should check). It depends on how much you can compress the files with little or no loss. Solution to Quiz 18.6 Correct answer: (c). As stated, the parameter i could be an arbitrarily large positive integer and there is an infinite
number of subproblems. One idea is to notice that these two strings can be "nicely aligned": A A G G G G C C T A where the "" indicates a gap inserted between two letters of the second string, which seems to be missing a 'G'. Thus, there are six independent sets: the empty set and the five singleton sets. A totally different application of the
problem is 1 The presentation in this section draws inspiration from Section 6.6 of Algorithm Design, by Jon Kleinberg and Eva Tardos (Pearson, 2005). (See Section 16.5.7 for the solution: Lemma 16.4 (Knapsack Optimal
Substructure) Let S be an optimal solution to a knapsack problem with n 1 items, item values v1, v2, . In the second iteration, three edges would expand the tree's reach: (a, c), (a, d), and (b, d). c) The algorithm always outputs a spanning tree, but it might not be an MST. Problem 15.6 (H) Prove that in a connected undirected graph with distinct
edge costs, there is a unique MST. Also, each job has a weight wj, with higher weights corresponding to the largest subproblem (resolving ties arbitrarily).8 If it was case 1, the last column of thear column of the largest subproblem (resolving ties arbitrarily).8 If it was case 1, the last column of thear column of thear column of the largest subproblem (resolving ties arbitrarily).8 If it was case 1, the last column of thear column of the column of thear column of thear column of thear column of thear column of the column of thear column of thear column of thear column of the column of thear column of thear column of thear column of the column of thear column of the colum
optimal alignment matches xm and yn and reconstruction resumes from the entry A[m 1][n 1]. Now derive T 0 from T \leftarrow [ {e1} by removing the edge e2: x y w v Because T \leftarrow has n 1 edges, so does T 0. The proposed variable-length code creates ambiguity, and more than one sequence of symbols would lead to the encoding "001." One possibility is
AB (encoded as "0" and "01," respectively), and another is AAD (encoded as "0," "0," and "1"). Can you make your algorithm simpler or faster using randomization? Proposition 14.1 (Encoding Length and Tree Depth) For every binary code, the encoding length in bits of a symbol a 2 ^ equals the depth of the node with label a in the corresponding
tree. The next-best option is the edge of cost 5; its inclusion does not create a cycle and, in fact, results in a spanning tree: 4 1 2 3 5 7 6 The algorithm skips the edge of cost 5 (which would create a triangle with the edges of cost 1 and 5). What should be
the new values of y and z's keys, respectively? Consider an input graph G = (V, E) with source vertex v 2 V and a hop count constraint i 2 {1, 2, 3, . 108 Introduction to Dynamic Programming 16.1.4 Solutions to Quizzes 16.1-16.2 Solution to Quiz 16.1 Correct answer: (c). 36
Huffman Codes where the summations are over all the alphabet symbols for which the corresponding leaf belongs to T1 or T2, respectively. Assuming this lemma, for every subproblems are the important ones, why
not cut to the chase and systematically solve them one by one? Quiz 14.1 With the variable-length binary code above, what is the string "001" an encoding of? Obtain the (n 2)-vertex path graph Gn 2 from G by plucking off the last two vertices and edges:3 excluded from S included in S v1 v2 v3 v4 1 4 5 4 Gn-2 Because S contains vn and Gn 2 does not
we can't regard S as an independent set of Gn 2. Which split should we use? It was something not even a Congressman could object to. Or maybe each object to something to pairs of "similar" documents. This would contradict the supposed optimality of S for
the original problem. Suppose i 1 and v 2 V, and let P be a shortest s-v path in G with at most i edges, with cycles allowed. In the third iteration, the value at t drops to 5 (because A[2][v] + vt = 5, which is better than both A[2][t] = 8 and A[2][w] + vt = 6) and the other four vertices inherit solutions from the previous iteration: v 4 A[3][s]=0 A[2]
[s]=0 A[1][s]=0 A[0][s]=0 A[0][s]=0 A[0][s]=0 A[0][v]=1 A[2][v]=1 A[2][v]=1 A[1][v]=4 A[0][v]=+\infty 4 s t -1 2 A[3][u]=2 A[0][u]=+\infty 4 s t -1 2 A[3][u]=2 A[0][s]=0 A[0][s]=
[v]=1 A[3][v]=1 A[2][v]=1 A[2][v]=1 A[0][v]=4 A[0][v]=4 A[0][v]=+\infty 4 s t -1 2 A[4][u]=2 A[3][u]=2 A[3][u]=2 A[3][u]=2 A[3][u]=4 A[3][u]=
time analysis of the Bellman-Ford algorithm is more interesting than those of our other dynamic programming algorithms. 17.2.2 Average Search Time To search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k in a binary search time for a key k 
outputting n2 numbers.18 18.3.2 Reduction to Single-Source Shortest Paths If you're on the lookout for reductions (as you should be; see page 100), you might already see how to apply your ever-growing algorithmic toolbox to the all-pairs shortest path problem. As we've seen, prefix-free binary codes for an alphabet ^ correspond to ^-trees.
Similarly, the new version of line 7 either looks up or recursively computes and caches S2, as needed. 21 Named after Robert W. Provided the Find and Union operations run in O(log n) time, as assured by Theorem 15.14, the total running time is: preprocessing 2m Find operations n 1 Union operations + remaining bookkeeping total O(n) + O(m log
n) O(m log n) O(n log n) O(m log n) O(m) O((m + n) log n). It fails even in a trivial example like: v -5 1 s 1 -2 t As with the minimum spanning tree problem, we can assume that the input graph has no parallel edges. After finishing this chapter, ask yourself: Could you have ever solved either problem without first studying dynamic programming? Intuitively, it's a
net win to demote the smaller-frequency symbols a and b to the deepest level of the tree while promoting the higher-frequency symbols x and y closer to the root. (All positive integers.) Output: A subset S P 🗸 {1, 2, . For example, you can always compute a minimum spanning tree of your undirected graph data in a preprocessing step, even if you're
not quite sure how it will help later., Sp of non-empty subsets of X such that each object x 2 X belongs to exactly one of the subsets. 13.3.1 Two Special Cases Let's think positive and posit that there actually is a correct greedy algorithm for the problem of minimizing the weighted sum of completion times., k 1}, so A[k 1][v][k] and A[k 1][k][v] are at
most their respective lengths. 14.2 Codes as Trees The "prefix-free" constraint in the optimal prefix-free code problem in Section 16.5 is a good example. 14.1.3 Prefix-Free Codes We can eliminate all ambiguity by insisting that a code be prefixfree., wn and
lengths `1, `2. The overall running time is O((m + n) log n), as promised by Theorem 15.4. QE D 15.3.6 Solution to Quiz 15.3 Correct answer: (b). Presumably, the algorithm would start by solving the smallest subproblems (with i = 1), and so on. A subgraph with n 1 edges—a
candidate for a spanning tree, by Corollary 15.8—fails one of the conditions only if it fails both. In other words, once you know exactly what the rest of it looks like: an optimal alignment of X and Y matches xm and yn in its last column, you know exactly what the rest of it looks like: an optimal alignment of X and Y matches xm and yn in its last column, you know exactly what the rest of it looks like: an optimal alignment of X and Y matches xm and yn in its last column, you know exactly what the rest of it looks like: an optimal alignment of X and Y matches xm and yn in its last column, you know exactly what the rest of it looks like: an optimal alignment of X and Y matches xm and yn in its last column, you know exactly what the rest of it looks like: an optimal alignment of X and Y matches xm and yn in its last column, you know exactly what the rest of it looks like: an optimal alignment of X and Y matches xm and yn in its last column, you know exactly what the rest of it looks like: an optimal alignment of X and Y matches xm and yn in its last column, you know exactly what the rest of it looks like: an optimal alignment of X and Y matches xm and yn in its last column, you know exactly what the rest of it looks like: an optimal alignment of X and Y matches xm and yn in its last column, you know exactly what the rest of its looks like: an optimal alignment of X and Y matches xm and yn in its last column.
1, r + 2, . In (b), adding the edge (4, 8) fuses two components into one. 116 Introduction to Dynamic Programming plucked off the end of the graph. For 38 Supervised learning focuses on prediction rather than pattern-finding per se. (Choose all that apply.) a) The maximum-cost spanning tree problem. Let L denote the length of P. Each iteration of
Kruskal's algorithm performs a cycle check or, equivalently, a path check. If this bothers you, feel free P to normalize the frequencies by dividing each of them by their sum nj=1 pj —this doesn't change the problem. 18 Some heap operation
rather than two. How much better are the schedules computed by the GreedyRatio algorithm than those by the GreedyRatio algorithm?, n} of keys and nonnegative frequencies p1, p2, . The answer depends on how we implement the Union operation; in this sense, answer (d) is correct. The more general statement in Corollary 16.2 follows by invoking
the first statement, for each i = 2, 3, . In this case, the path P can be interpreted as the amalgamation of two solutions to smaller subproblems: the prefix P1 of P that travels from k to w. I decided therefore to use the word, "programming." . This means that, for each pair of distinct symbols a, b 2 ^, the
encoding of a is not a prefix of that of b, and vice versa. Problem: Minimizing the Sum of Weighted Completion Times Input: A set of n jobs with positive lengths `1, `2, divide-and-conquer, 115, 120-122 when to use, 202 e.g., 79 edge (of a graph), 52 length, 168 parallel, 55 exchange argument, 12 for minimum spanning trees, 75 in Huffman's
algorithm, 43 Index in scheduling, 15 exhaustive search, 6, 27, 62, 105, 140, 161, 201 Fano, Robert W., 187 Floyd-Warshall algorithm design), 201-202 Floyd, Robert W., 187 Floyd-Warshall algorithm, 187-196 detecting a negative cycle, 194-195 optimal substructure, 189-192 pseudocode, 193 reconstruction, 196 recurrence, 192 running time, 194
space usage, 194 subproblems, 187-189 for-free primitive, 64, 201 Ford Jr., Lester R., 172 forest (of trees), 34 Git, 164 graph, 52 adjacency-matrix representation, 53 adjacency-matrix representation, 53 number of edges, 80 path, 53, 105
search, 55, 81 spanning tree, 53 sparse, 185 undirected, 52 213 vertex, 52 greatest hits, ix greedy algorithm, vii, 1-3 and brainstorming, 106, 201 as a heuristic, 3, 202 exchange argument, see exchange argument for clustering, 96 for optimal prefix-free codes, see Huffman's algorithm for scheduling, 6-10 induction, 13 informal definition, 2 Kruskal's
algorithm, see Kruskal's algorithm Prim's algorithm Prim's algorithm proof of correctness, 13 themes, 2 usually not correct, 3, 107 GreedyRatio, see Scheduling, GreedyRatio heap, 62-63 DecreaseKey, 67 Delete, 63 ExtractMin, 63 in Huffman's algorithm, see Prim's al
Prim's algorithm, 63-68 HeapSort, 2 heuristic, 3, 202 hints, x, 203-210 Huffman's algorithm, 32-36 ^-tree, 31 average leaf depth, 31, 44 examples, 37-40 greedy criterion, 35 implemented with two queues, 41, 51, 204 proof of correctness, 41 pseudocode, 36 running time, 40 Huffman, David A., 32 i.e., 23
independent set (of a graph), 104 induction, 43 base case, 43 in greedy algorithm, 18 inductive hypothesis, 44 inductive hypothesis, 44 inductive hypothesis, 44 inductive hypothesis, 45 job, see scheduling Karatsuba's algorithm, 18 consecutive, 15 Jarník's algorithm, 18 consecutive, 15 Jarník's algorithm, 18 consecutive, 15 Jarník's algorithm, see Prim's algorithm, 18 consecutive, 15 Jarník's algorithm, 18 consecutive, 15 Jarník's algorithm, 18 consecutive, 16 Jarník's algorithm, 18 consecutive, 17 Jarník's algorithm, 18 consecutive, 18 Jarník's algorithm, 19 Jarník's
123-132 applications, 124 correctness, 129 definition, 123 dynamic programming algorithm (Knapsack), 128 example, 129-130 generalizations, 125 reconstruction, 120 recurrence, 126 running time, 129 subproblems, 127 Knuth, Donald E., 161 Kruskal's algorithm achieves the minimum bottleneck property, 93 Index
and clustering, 97 cycle-checking, 81, 84 example, 77 in reverse, 99 outputs a spanning tree, 92 proof of correctness, 91-93, 102 pseudocode (union-findbased), 81 stopping early, 79 vs. By the time an iteration of the double
for loop must compute the subproblem solution A[i][c], the values A[i 1][c] and A[i 1][c si] of the two relevant smaller subproblems have already been computed in the previous iteration of the outer loop (or in the base case). As a bonus, in typical implementations, the constant hidden by the big-O notation and the amount of space overhead are
relatively small.12 15.3.3 How to Use Heaps in Prim's Algorithm Heaps enable a blazingly fast, near-linear-time implementation of Prim's algorithm. 77 Kruskal, Joseph B., 76 Lehman, Eric, x Leighton, F
Let 1 denote the smallest strictly positive difference between two edges' costs. This means there is a different search tree T1\vdash with keys {1, 2, . It's important that the outer loop is indexed by the subproblem size k, so that all of the relevant terms A[k 1][v][w] are available for constant-time look up in each inner loop iteration. A[i 2] + wi , which have
already been made by the WIS algorithm. What can we say about it? When the reconstruction algorithm hits a base case, it completes the alignment by prepending the appropriate number of gaps to the string that has run out of symbols. QE D *15.4 Prim's Algorithm: Proof of Correctness 75 We can similarly argue that the output of Prim's algorithm
is a spanning tree. Still, there is no substitute for the detailed understanding of an algorithm that comes from providing your own working implementation of it. The divide-and-conquer algorithm that comes from providing your own working implementation of it. The divide-and-conquer algorithm that comes from providing your own working implementation of it.
because each batch of subproblem solutions and predecessors depends only on those from the previous batch, both the forward and reconstruction passes require only O(n) space (analogous to Problem 17.5). If not, run the Floyd-Warshall algorithm on G. 14.4.2 The Details Induction Review For the formal proof, we turn to our old friend (or is it
nemesis?), induction.11 Recall that proofs by induction follow a fairly rigid template, with the goal of establishing that an assertion P (k) holds for arbitrarily large positive integers k. Next, recall our first (non-prefix-free) variable-length code: Symbol A B C D Encoding 0 01 10 1 This code can be represented using a different labeled binary tree: 1 0 Direction P (k) holds for arbitrarily large positive integers k. Next, recall our first (non-prefix-free) variable-length code: Symbol A B C D Encoding 0 01 10 1 This code can be represented using a different labeled binary tree: 1 0 Direction P (k) holds for arbitrarily large positive integers k. Next, recall our first (non-prefix-free) variable-length code: Symbol A B C D Encoding 0 01 10 1 This code can be represented using a different labeled binary tree: 1 0 Direction P (k) holds for arbitrarily large positive integers k. Next, recall our first (non-prefix-free) variable-length code: Symbol A B C D Encoding 0 01 10 1 This code can be represented using a different labeled binary tree: 1 0 Direction P (k) holds for arbitrarily large positive integers k. Next, recall our first (non-prefix-free) variable-length code: Symbol A B C D Encoding 0 01 10 1 This code can be represented using a different labeled binary tree.
A 1 0 B C Once again there are four nodes labeled with the symbols of the alphabet—the two leaves and their parents. We mentioned briefly in Section 14.3.6 that there is, in fact, a data structure whose raison d'être is fast minimum computations: the heap data structure. One natural approach is to make repeated use of a subroutine that solves the
single-source shortest path problem (like the Bellman-Ford algorithm). Find 1. Ideally, this thought experiment would show that an optimal solution must be constructed in a prescribed way from optimal solution sto smaller subproblems, thereby narrowing down the field of candidates to a manageable number. 2 More concretely, let G = (V, E) denote
the n-vertex path graph with edges (v1, v2), (v2, v3), . In (c), adding the edge (7, 9) creates a new cycle. (For each i, j 2 {1, 2, . Applying the Inductive Hypothesis to a Residual Problem The inductive Hypothesis applies only to alphabets with less than k symbols. What if we consider only paths without cycles? The second job contributes 12
Introduction to Greedy Algorithms to all the completion times other than that of the first job, so the second-shortest job should be scheduled next, and so on. The answer to the second question is now clear: Install the demoted root directly under the (depth-0) promoted root so that the occupants of T1 suffer a depth increase of only 1. (Adding an edge
(v, w) to the solutionso-far T creates a cycle if and only if T already contains a v-w path.) 29 Also known as the disjoint-set data structure. i j = Ln 1, v for some v 2 V, the contrapositive of Lemma 18.4
implies that the input graph G contains a negative cycle, in which case the algorithm for the single-source shortest path problem (Section 18.2.1) is to always work with the original input graph and impose an artificial
constraint on the number of edges allowed in the solution to a subproblem. See the bonus video at www.algorithmsilluminated.org for more details. If A[n 2] + wn A[n 1], supplementing an MWIS of Gn with vn yields an MWIS of Gn is the running time of the algorithmsilluminated.org for more details.
myopic decisions will come back to haunt you and, with full hindsight, be revealed as a terrible idea. The time-constrained reader can skip these sections on a first reading without any loss of continuity. If you schedule the shorter job first, the completion times are 1 and 3, for a total of 4. Our fixed-length code Symbol A B C D Encoding 00 01 10 11
can be represented via a complete binary tree with four leaves; 1 0 0 A 3 1 0 B C 1 D Every node of a binary tree can have a left child, a right child, both, or neither., pn. The first benefit of emphasizing high-level descriptions over language-specific implementations is flexibility, and so on, ad infinitum. Solution to Quiz 16.6 Correct answer: (c), d)
Given a connected undirected graph G = (V, E) with positive edge costs, compute a minimum-cost set F \( \text{E} \) is acyclic. But then Li,v \( < \text{Lk,v} \), contradicting part (a) of the lemma. Every recursive call of a divide-and-conquer algorithm commits to a single split of its problem into one or more smaller subproblems
Thus, the weighted search time (17.1) can be written as p k · (1 + k=1 Pn k's search Pn time in T1) + (weighted search time in T2), which cleans up to k=1 pk + (weighted search time in T1) + (weighted search time in T2), which cleans up to k=1 pk + (weighted search time in T2).
Huffman Codes Solution to Quiz 14.2 Correct answer: (b). ["Dynamic"] has a very interesting property as an adjective, and that is it's impossible to use the word, dynamic, in the pejorative sense. 8. Test Your Understanding Problem 15.1 (H) Consider an undirected graph G = (V, E) in which every edge e 2 E has a distinct and nonnegative cost. By
much.1 Can we do better than a greedy or divide-and-conquer algorithm? P The proof of correctness uses an exchange argument to show that the algorithm computes such a solution. Are the resulting algorithms well defined and
correct? (If there are no such paths, then Li,v = +1.) For every i 1 and v 2 V, \rightarrow Li 1,v (Case 1) Li,v = min. A good stopping criterion follows from the observation that the solutions to a given batch of subproblems, with a fixed edge budget i and v ranging over all possible destinations, depend only on the solutions to the previous batch of subproblems.
(with edge budget i 1). Reductions are one of the most important concepts in the study of algorithms and their limitations, and they can also have great practical utility. Hint for Problem 17.6: Don't bother solving subproblems for prefixes Xi and Yj with |i j| > k. Whenever our two rules-of-thumb—to prefer shorter jobs and higher-weight jobs—luckily
coincide for a pair of jobs, we know which one to schedule first (the shorter, higher-weight one). The most obviously false statement is (b), which doesn't even typecheck (C is in units of size, vn in units of value). Ranging over all relevant values of the two parameters, we obtain our collection of subproblems: 5 Sequence Alignment: Subproblems
represented in my online courses, and there are large numbers of students (highschool, college, etc.), software engineers (both current and aspiring), scientists, and professionals hailing from the front or adding to the back) runs in O(1)
time, so the total running time of the n 1 iterations of the main loop is O(n). 15.8.1 Clustering One widely-used approach to unsupervised learning is clustering, in which the goal is to partition the data points into "coherent groups" (called clusters) of "similar points" (Figure 15.8). Suppose we reverse the order of the for loops—literally cutting and
pasting the second loop in front of the first, without changing the pseudocode in any other way. Nobody knows! One of the biggest open questions in the field of algorithms is whether there is an algorithm for the all-pairs shortest path problem on n-vertex graphs that runs in, say, O(n2.99) time.32 18.4.6 Solutions to Quizzes 18.5-18.6 Solution to
Quiz 18.5 Correct answer: (d)., `n, the GreedyRatio algorithm outputs a schedule with the minimum-possible sum of weighted completion times. 18.4 The Floyd-Warshall Algorithm 189 and, for the origin 1 and the destination 5, the subproblems corresponding to successive values of the prefix length k. *17.2 151 Optimal Binary Search Trees such as
red-black trees, are explicitly designed to keep the search tree close to perfectly balanced (see Section 11.4 of Part 2). Problem 13.3 (H) You are given as input n jobs, each with a start time sj and a finish time tj. If all your divide-and-conquer attempts fail—and especially if they fail because the combine step always seems to require redoing a lot of
computation from scratch—it's time to try dynamic programming. The one with larger total weight. (A subsequence need not comprise consecutive symbols. Studying algorithms can feel like watching a highlight reel of many of the greatest hits from the last sixty years of computer science. If there were a better solution S \( \sim \) \( \lambda \) (1, 2, . Given an input
graph G = (V, E) with vertex weights, the pseudocode is: WIS: A Greedy Approach S := ; sort vertices of V by weight for each v 2 V , in nonincreasing order of weight do if S [ {v} is an independent set of G then S := S [ {v} return S Simple enough. These videos cover all the topics in this book series, as well as additional advanced topics. Bellman,
writing about his time working at the RAND Corporation: The 1950's were not good years for mathematical research., j} and their frequencies playing the role of the original input. 165 Problems Problem 17.4 (S) Consider an instance of the optimal binary search tree problem with keys {1, 2, . For nonempty input strings, of the three options in
Lemma 17.1, the one with the smallest total penalty is an optimal solution. 18.1 Shortest Paths with Negative cycle). Quizzes. MST Assumption The input graph G = (V, E) is connected, with at least one path between each pair of vertices., n} with i j. 13 In more detail, consider a tree T with root r and left and
right subtrees T1 and T2. QE D With Lemma 15.7 at our disposal, we can quickly deduce some interesting facts about spanning trees. One invocation of the graph (n numbers in all, out of the n2 required). The cost of exchanging
the jobs i and j in a consecutive inversion is that i's completion time Ci goes up by the length 'j of job j, which increases the objective function (13.1) by wi · `j . 17 It's actually not important that the item values are integers (as opposed to arbitrary positive real numbers). 14 The Bellman-Ford algorithm was discovered long before the Internet was a
gleam in anyone's eye—over 10 years before the ARPANET, which was the earliest precursor to the Internet. 13.1 The Greedy Algorithm Design Paradigm 13.1.1 Algorithm Design Paradigm 13.1.1 Algorithm Design Paradigm 13.1.1 The Greedy Algorithm Design Paradigm 13.1.1 Algorithm Design Paradigm. You might believe, in your heart of
hearts, that your natural greedy algorithm must always solve the problem correctly., Gn either. There are m iterations of the main loop and each uses two Find operations (for a total of 2m). 54 1 Minimum Spanning Trees 2 3 4 2 5 6 2 7 3 7 3 6 7 8 splice out 1 7 2 8 Figure 15.2: A path with repeated vertices can be converted into a
path with no repeated vertices and the same endpoints. After these three steps are implemented, the corresponding dynamic programming algorithm writes itself: Systematically solve all the subproblems one by one, working from "smallest" to "largest," and extract the final solution from those of the subproblems. But then the concatenation of P1-
and P2 would be a cycle-free path P From v to w with internal vertices in {1, 2, . 12} but the smaller (more negative) difference (2 vs. That paradigm went like this: The Divide-and-Conquer Paradigm 1. With arbitrary promotion and demotion decisions, a sequence of n 1 Union operations can produce the height-(n 1) tree shown in the solution to
Quiz 15.5, with each operation installing the tree-so-far underneath a previously isolated object. In each iteration, the algorithm irrevocably and myopically commits to an estimate of the shortest-path distance to one additional vertex, never revisiting the decision. Theorem 15.11 can be proved in its full generality with a bit more work (see Problem
15.5). 14.2 29 Codes as Trees Every edge connecting a node to its left or right child is labeled with a "0" or "1," respectively. 28 Because the solutions to a batch of subproblems depend only on those from the previous batch, the algorithm can be implemented using O(n2) space (analogous to Problem 17.5). Properly implemented, the running time
drops from exponential to linear. Theorem 15.1 (Correctness of Prim) For every connected graph G = (V, E) and real-valued edge costs, the Prim algorithm returns a minimum spanning tree of G. QE D Lemma 18.4 shows that if the input graph does not have a negative cycle, subproblem solutions stabilize by the nth batch. We can view the rest of the
alignment (excluding the final column) as an alignment of the remaining symbols—an alignment of the shorter strings X 0 and Y 0 + gaps {z rest of alignment of the penalty 4xm yn that was previously paid in the last column. To
define the subproblems, consider an input graph G = (V, E) and arbitrarily assign its vertices the names 1, 2, . The heap never stores more than n 1 objects, so each heap operation runs in O(log n) time (Theorem 15.3). Informally, the problem is to determine how similar the two strings are; we'll make this precise in the next section. In the WIS
problem on path graphs and the knapsack problem, we zoomed in on a solution's last decision—does the last vertex of the path or the last item of a knapsack capacity C. Because the input strings have the same length, every alignment inserts the same number of gaps in each, and so the total
number of gaps is even. (See Section 18.4.6 for the solution and discussion.) Is the bug fatal, or do we just need to work a little harder? A := length-(n + 1) array // subproblem solutions A[0] := 0 // base case #1 A[1] := max{A[i 1], A[i 2] + wi } | {z } | {z } Case 1 Case 2
return A[n] // solution to largest subproblem The length-(n + 1) array A is indexed from 0 to n. Adding e1 to T \leftarrow creates a cycle C that contains e1 (Lemma 15.7(a)): x e2 y w e1 v As an edge of T, e1 satisfies the MBP, so there is at least one edge e2 = (x, y) in the v-w path C {e1} with cost at least cvw. I encourage you to develop your own recipe
based on your personal experience. Output: the maximumP total value of a subset S / {1, 2, . P Prim's algorithm constructs an MST one edge at a time, starting from an arbitrary vertex and growing like a mold until the entire vertex set is spanned. That is, we seek two jobs whose ordering by difference is the opposite of their ordering by ratio. Union
by-rank is discussed at length in the bonus videos at www.algorithm always outputs an MST. Because Kruskal scans through the edges in order of nondecreasing cost, the algorithm processes every edge of P before e. Output: the total weight of a maximum-weight
independent set of G. Thus, we'd like the number of subproblems to be as low as possible—our WIS solution used only a linear number of subproblems, which is usually the best-case scenario. nothing. Second, the running time analysis is often a one-liner. Suppose S includes the last vertex vn. Greedy algorithms solve problems by making a sequence
of myopic and irrevocable decisions. Quiz 16.5 Consider an instance of the knapsack problem with knapsack problem with knapsack problem with the frequency p0ab of the new
symbol ab defined as the sum pa + pb of the frequencies of the two symbols it represents. 92 Minimum Spanning Trees The first order of business is to show that the algorithm's output is connected (and, as it's clearly acyclic, a spanning tree). Proof: Reprise the edge addition process from Corollary 15.8. If each of the n 1 edge additions has type F,
then Lemma 15.7(b) implies that the process concludes with a single connected component and no cycles (i.e., a spanning tree). Second, the problem formulation does not assume that the pi 's sum to 1 (hence the phrasing "weighted" search time). This leaves us with three relevant possibilities for the contents of the
last column of an optimal alignment: (i) xm and yn are matched with a gap; or (iii) yn is matched with a gap. Problem 15.3 (H) Which of the following problems reduce, in a straightforward way, to the minimum spanning tree problem? We've seen that binary trees can represent all binary codes, prefix-free or not. Lemma 15.5 (Prim
Achieves the MBP) For every connected graph G = (V, E) and real-valued edge costs, every edge chosen by the Prim algorithm satisfies the MBP. 9 If this seems like a lot of subproblems, don't forget that the single-source shortest path problem is really n different problems in one (with one problem per destination vertex). , Cm of m knapsacks, where
m could be as large as n, compute disjoint subsets. The completion time of a job is the time corresponding to its topmost edge. Then, if we encounter the 5 If big-O notation f (n) is \rightarrow(g(n)) if there are constants c1 and c2 such that f (n) is wedged
between c1 · g(n) and c2 · g(n) for all sufficiently large n. 14 Under our standing assumption that the input graph is connected, m is at least n 1 and we can therefore simplify O((m + n) log n) to O(m log n) in the running time bound. Because P is the union of P1 and P2, L = L1 + L2. This is subcubic in n except when m is very close to quadratic in n.
18.2.4 When Should We Stop? 14.3.3 Pseudocode As advertised, Huffman's algorithm builds a ^-tree bottom-up, and in every iteration it merges the two trees that have the smallest sums of corresponding symbol frequencies. This score might be negative, but that poses no obstacle to sequencing the jobs from highest to lowest score. The sum of
weighted completion times is then w1 · C1 + w2 · C2 = 3 · 7 + 1 · 2 = 23. When an edge is examined, the algorithm performs two heap operations (in lines 12-16 to the running time (over all while loop iterations) is O(m) heap operations plus O(m) additional work, so the total contribution of lines 12-16 to the running time (over all while loop iterations) is O(m) heap operations plus O(m) additional work, so the total contribution of lines 12-16 to the running time (over all while loop iterations) is O(m) heap operations plus O(m) additional work, so the total contribution of lines 12-16 to the running time (over all while loop iterations) is O(m) heap operations plus O(m) additional work, so the total contribution of lines 12-16 to the running time (over all while loop iterations) is O(m) heap operations plus O(m) additional work, so the total contribution of lines 12-16 to the running time (over all while loop iterations) is O(m) heap operations plus O(m) additional work, so the total contribution of lines 12-16 to the running time (over all while loop iterations) is O(m) heap operations plus O(m) additional work.
and symbol frequencies p = {pa }a2^, we denote by L(T, p) = pa · (depth of the leaf labeled a in T). Why not? Each iteration of Kruskal's algorithm that adds a new edge fuses two connected components into one, just as each
iteration of single-link clustering merges two clusters into one. If you've already reached black-belt status in dynamic programming, you might be able to guess the right collection of subproblems; of course, I don't expect you to be at that level after only two case studies. The bottleneck of P is at least cxy, which by inequality (15.1) is at least cv which we have a least cxy and the collection of subproblems; of course, I don't expect you to be at that level after only two case studies.
. time j σ1 σ2 (a) Before exchange (b) After exchange relative order unchanged! Figure 13.4: Swapping jobs in a consecutive inversion decreases the total number of inversion of subproblems; (ii) show how to quickly solve "larger"
subproblems given the solutions to "smaller" ones; and (iii) show how to quickly infer the final solution from the solution and discussion.) 13.4 17 Proof of Correctness Solving Quiz 13.4 puts us in a great position to
finish the proof. Thus, Kruskal's algorithm corresponds to single-link clustering, with vertices substituting for data points and connected components for clusters. This first idea is not enough., pj. Quiz 13.5 An inversion in a schedule is a pair k, m of jobs with k < m and m processed before k. This thought experiment can lead to both the identification
of the relevant subproblems and a recurrence (analogous to Corollary 16.2) that expresses the solution of a subproblem as a function of the solutions of smaller subproblems. Thus, for our approach to succeed, the answer should be either (c) or (d). Back in the WIS problem on path graphs, we used only one parameter i to index subproblems, where i
was the length of the prefix of the input graph. If there are multiple edges with the same beginning and end, we can throw away all but the shortest one without changing the problem. Moreover, as in the first two cases, it is an optimal such alignment (as you should verify). Checking whether an edge addition (v, w) would create a cycle is equivalent
to checking whether v and w are already in the same connected component. All else being equal, scheduling the shortest job first minimizes this negative impact. Analyzing the running times of divide-and-conquer algorithms can be difficult, due to the tug-of-war between the forces of proliferating subproblems and shrinking work-per-subproblem. Run
Kruskal's algorithm with the input graph G until the solution-so-far T contains |X| k edges or, equivalently, until the graph (X, T) has k connected components. How much faster is the implementation in Problem 14.5
than the heap-based implementation? Quiz 18.4 How many invocations of a single-source shortest path problem? Or, if the vertices represent classes, the independent sets correspond to feasible course
given the solutions to "smaller" ones. In both cases, the structure of optimal solutions is more complex than in last chapter's case studies, with a subproblems. *15.6 Speeding Up Kruskal's Algorithm via Union-Find 89 a) O(1) b) O(log n) c) O(n) d) Not enough information to answer
 (See Section 15.6.5 for the solution and discussion.) The solution to Quiz 15.6 demonstrates that, to achieve the desired logarithmic running time, we need another idea. In special cases, yes (see Problem 17.6). Merging them produces a full-blown binary tree: A B C D 34 Huffman Codes This binary tree is the same one used to represent the fixed-
length code in Section 14.2.1. Alternatively, in the second iteration we could merge the node labeled "B" with the tree containing "C" and "D": B C A D The final merge is again forced on us and produces the binary tree used to represent the variable-length prefix-free code in Section 14.2.1: A B C D In general, Huffman's greedy algorithm maintains are
forest, which is a collection of one or more binary trees. Suppose someone handed us on a silver platter a minimum-penalty alignment of two strings. 42 Huffman Codes Theorem 14.2 (Correctness of Huffman) For every alphabet ^ and nonnegative symbol frequencies {pa }a2^ , the Huffman algorithm outputs a prefix-free code with the minimum-penalty alignment of two strings.
possible average encoding length., x7 copies of each respective piece, determines whether you can tile a 10-by-n board with exactly those pieces (placing them however and wherever you want—not necessarily in Tetris order). Test Your Understanding Problem 17.1 (S) For the sequence alignment input in Quiz 17.1, what are the final array entries of
the NW algorithm from Section 17.1? Problem 17.2 (H) The Knapsack algorithm from Section 16.5 and the NW algorithm from Section 17.1 both fill in a two-dimensional array using a double for loop. The benefit is that j's completion time Cj goes down by the length \in i a two-dimensional array using a double for loop. The benefit is that j's completion time Cj goes down by the length \in i a two-dimensional array using a double for loop. The benefit is that j's completion time Cj goes down by the length \in i a two-dimensional array using a double for loop. The benefit is that j's completion time Cj goes down by the length \in i a two-dimensional array using a double for loop. The benefit is that j's completion time Cj goes down by the length \in i a two-dimensional array using a double for loop. The benefit is that j's completion time Cj goes down by the length \in i a two-dimensional array using a double for loop. The benefit is that j's completion time Cj goes down by the length \in i a two-dimensional array using a double for loop. The benefit is that j's completion time Cj goes down by the length \in i a two-dimensional array using a double for loop. The benefit is that j's completion time Cj goes down by the length \in i a two-dimensional array using a double for loop. The benefit is that j's completion time Cj goes down by the length \in i a two-dimensional array using a double for loop.
There are two options for expanding its reach: the edge (a, b) and the edge (b, d). Every merger thus increases the objective function that we want to minimize: the average leaf depth (14.1). The trees are not necessarily binary, as there is no limit to the number of objects that can have the same parent. Proceed by recursing on a graph with half as
many edges as G. Case 1: Vertex k is not an internal vertex of P. Prim Input: connected undirected graph G = (V, E) in adjacency-list representation and a cost ce for each edge e 2 E. Solution to Problem 17.3: The problems in (b) and (d) can be solved using algorithms similar to NW, with one subproblem for each pair Xi, Yj of input string prefixes.
We call this a type-F edge addition ('F' for "fusion"). In the example above, the first four objects belong to a set named "4," and the last two to a set named "6." Initialize and Find operations should be implemented. This means we have a super-cool algorithmic
problem on our hands, which is the subject of the rest of this chapter. Proof: Denote the edges of P by (x0, x1), (x1, x2), . 16.5.3 The Subproblems and solve them systematically using the recurrence identified in Corollary 16.5. For now, we focus on computing the total value of an
optimal solution for each subproblem. This point is especially difficult to accept for clever greedy algorithms that you invented yourself. We can then interpret points x, y with a small value of f (x, y) as "similar," and those with a large value as "dissimilar." 39 For example, if the data points are vectors with a common dimension, like in the image
example above, f (x, y) could be defined as the Euclidean (i.e., straight-line) distance between x and y.40 For another example, Section 17.1 defines Needleman-Wunsch distance, which is a symmetric similarity function designed for genome sequences. (See www. then e belongs to the MST of G." 43 There's also the Cycle Property, which asserts that
if an edge e is the costliest on some cycle C, every MST excludes e., pn inherited from the original problem. Starting from the largest subproblem in the upperright corner, the reconstruction algorithm checks which
from scratch for each of the subproblems? (18.2) Lk 1,v,k + Lk 1,k,w (Case 2) 18.4.3 Pseudocode Suppose we know that the input graph has no negative cycles, in which case Lemma 18.6 and Corollary 18.7 apply. 176 Shortest s-w path with at most i
supplemented with the edge (w, v) 2 E. But I'd never waste your time; in graphs with distinct edge costs, the latter automatically implies the former.22 Theorem 15.6 (MBP Implies MST) Let G = (V, E) be a graph with distinct real-valued edge costs, and T a spanning tree of G. The variable-length code in the preceding section is not: The encoding of
end of the algorithm. Readers can interact with me and each other about the end-of-chapter problems through the book's discussion forum (see below). Whenever a new edge (v, w) is added to the solution-so-far, the connected components of v and w fuse into one, and one Union operation suffices to update 32 If x and y are already in the same set of
the partition, this operation has no effect. j ... QE D The following corollary extends Lemma 15.15 from individual edges to multi-hop paths. If {k, m} = {i, j}, then k and m form an inversion in 1 but not in 2 (because the swap un-inverts them).
Chapter 10 of Part 2. What are the consequences for the running time of a Find operation? We conclude that the union-find-based implementation of Kruskal is faithful to its original implementation, with both versions producing the same output. This proves part (a). We can visualize a schedule by stacking the jobs on top of one another, with time
increasing from bottom to top (Figure 13.1). After the edge addition, P [ {(v, w)} forms a new cycle., n} in T are one than Pr more 1 in T2. As a spanning tree, T contains a v-w path P. Corollary 15.16 now implies that, by the time Kruskal reaches the edge e, its endpoints v and w already belong to the same connected component of the solution-so-far and w already belong to the same connected component of the solution and the solution are connected component of the solution are conn
T. *15.6 Speeding Up Kruskal's Algorithm via Union-Find 81 adding e to T creates a cycle if and only if T already contains a v-w path. 17.1.5 The Subproblems in the knapsack problems in the recurrence (Corollary 17.2) are indexed by two different parameters, i and j. The time required to download a file? One approach is to look
for a similar region in a better-understood genome, like the mouse genome, and conjecture that the similar regions play the same or similar region in a better-understood genome, like the mouse genome, and conjecture that the similar regions play the same or similar region in a better-understood genome, like the mouse genome, like the mouse genome, and conjecture that the similar regions play the same or similar regions play the same or similar regions play the same or similar regions.
preceding j in , plus the length of j itself. Thus job j's completion time decreases by `i . The minimum spanning tree problem thus qualifies as a "for-free primitive," joining the likes of sorting, computing the connected components of a graph, and the single-source shortest path problem. P Huffman's algorithm is quaranteed to compute a prefix-free
code with the minimum-possible average encoding length. One possibility is that the new edge fuses two connected components into one (Figure 15.7(b)). Induction (on i) shows that the Bellman-Ford algorithm solves every subproblem correctly, with A[i][v] assigned the correct value Li,v; the recurrence in Corollary 18.2 justifies the inductive step
What would it look like? Both induction and exchange arguments play a role in our correctness proofs for Huffman's greedy coding algorithm (Chapter 15). This means the final output of the algorithm is the same as if it had been restarted from scratch with the input ^0 and
p0, with the resulting ^0 -tree translated by the mapping back to a ^-tree of Tab. What are the final array entries of the WIS? Because ^0 is a smaller alphabet than ^, we can complete the proof using induction. For example, suppose you're looking for an object with key 17. The
number of photos you can store on your smartphone? Suppose the vertex k of C has the largest label. (Removing an edge from a cycle undoes a type-C edge addition, which by Lemma 15.7(a) has no effect on the number of connected components.)
definition in Section 18.3.) Let's declare victory with the following optimal substructure lemma. For example, an input to the problem might look like this (with vertices, the first and fourth vertices, and the second and
                                                page 86), this implementation of Union promotes the root 4 and demotes the root 6, resulting in: 35 This implementation choice goes by the name union-by-size. Our standing assumptions (1) and (2) imply that jobs are indexed in strictly decreasing order of weight-length ratio, so wi wi < . . k 1}, all of the
internal vertices of P \( \text{belong to } \ \text{long to } \) belong to \( \text{1, 2, . For each pair of prefixes of the input strings, compute the total penalty of three alignments: the best one with a gap in the lower row of the final column, the best one with a gap in the lower row of the final column. The second main idea resolves this worry and
proves that it's always safe to commit to a tree in which the two lowest-frequency symbols correspond to sibling leaves. If both main ideas can be implemented, the inductive step and Theorem 14.2 follow easily. Algorithms alluminated, Part 1 covers asymptotic notation (big-O notation and its close cousins), divide-and-conquer algorithms and the
master method, randomized QuickSort and its analysis, and linear-time selection algorithms. (If no such path exists, define Li,v as +1.) (For each i 2 {0, 1, 2, . Then, a path prefix can indeed be viewed as a solution to a smaller subproblem. Find:
Programming Problems Problems Problem 16.6 Implement in your favorite programming language the WIS and WIS Reconstruction algorithms. The first round comprises a local tournament for each vertex w 2 V X, where the participants are the edges (v, w) with v 2 X and the first-round winner is the cheapest participant (or +1, if there are no such edges). Its
inclusion does not create a cycle and also happens to fuse the two trees-so-far into one: 4 1 2 3 5 7 6 The algorithm next considers the edge of cost 4. The other three edges do satisfy the MBP, as you should check.21 The next lemma implements the first step of our proof plan by relating the output of Prim's algorithm to the MBP. The algorithm halts
when only one tree remains. There are also several books that, unlike these books, cater to programmers looking for ready-made algorithm implementations in a specific programming language. Output: dist(s, v) for every vertex v 2 V, or a declaration that G contains a negative cycle. 14.1.6 Solutions to Ouizzes 14.1-14.2 Solution to Ouiz 14.1 Correct
answer: (d). The latter has no leaves until levels n/2 and later, which implies that it has at least 2n/2 leaves. In this case, because yn does not appear in the last column, the induced alignment | xm [gap]. This matches the running time bound promised in Theorem
15.13. I always appreciate suggestions and corrections from readers. QE D not-yet-processed X v* w* V-X x y the frontier Figure 15.6: Every v - w path crosses at least once from X to V The dotted lines represent one such path. The Floyd-Warshall algorithm (Section 18.4) also accommodates negative edge lengths and computes shortest-
path distances from every origin to every destination. In Part 4 we'll see the reason for the not-so-blazingly fast running time—there is a precise sense in which the knapsack problem is a difficult problem. Graphs can be encoded in different ways for use in an algorithm. The leaves of the tree are labeled with the four symbols of the alphabet., sn, and
knapsack capacity C. Consider an instance of the knapsack problem to it? Start from the empty graph with vertex set V and
add the edges of T one by one. The goal is to compute the minimum-possible penalty of an alignment under this new cost model. (The relative order of the second and third for loops doesn't matter.) There are O(n3) subproblems and the algorithm performs O(1) work for each one (in addition to O(n2) work outside the triple for loop), so its running
time is O(n3).27,28 Induction (on k) and the correctness of the recurrence (Corollary 18.7) imply that, when the input graph has a negative cycle? The NW algorithm
must remember subproblem solutions for the current and preceding values of i, and for all values of j (why?)., kn } with minimum-possible weighted search time for ki in T). Figure 15.3: A graph that is not connected. For this reason, the algorithm is also called Jarník's algorithm and the Prim-Jarník algorithm.8 15.2.1 Example
Next we'll step through Prim's algorithm on a concrete example, the same one from Quiz 15.1: a 4 c 1 b 3 5 2 d It might seem weird to go through an example of an algorithm before you've seen its code, but trust me: After you understand the example, the pseudocode will practically write itself.9 Prim's algorithm begins by choosing an arbitrary
vertex—let's say vertex b in our example. Programming Problems Problem 14.6 Implement in your favorite programming language the Huffman algorithm from Section 14.3 for the optimal prefix-free code problem. b) It is an optimal solution to the subproblem consisting of the first n 1 items and knapsack capacity C vn . 32 Huffman Codes 0 1 0 A A 1
0 B 0 C 1 0 D (a) Traversal #1 ("A") A 1 B 1 1 0 0 C 0 1 B 1 D (b) Traversal #2 ("B") 0 C 1 D (c) Traversal #3 ("D") Figure 14.1: Decoding the string "010111" to "ABD" by repeated root-leaf traversals, 146 Advanced Dynamic Programming can be proved by induction; the induction is on the value of i + j (the subproblem size), with the recurrence in
3\ 3\ 2\ 1\ 1\ 0\ 2\ 6\ 6\ 6\ 5\ 4\ 4\ 3\ 1\ 0\ 3\ 8\ 8\ 7\ 6\ 5\ 4\ 3\ 1\ 0\ 4\ 3\ 1\ 0\ 5\ and the roots between i and j—the roots between i and j—the roots between r(i, j 1) and r(i + 1, j) suffice. Is it a disguised version, variant, or special case of a problem that you already
know how to solve?, n, initialize parent(i) to i. For both y and z, these new incident crossing edges are cheaper than all their old ones. (The two-hop path is disqualified because it includes vertex 4 as an internal vertex. If A[n 1] of Gn. {z } | {z } 1,n Case 2 Case 3 144 Advanced Dynamic Programming More generally, for every i = 1, 2, . The idea here
is to show that every ^-tree can be massaged into an equally good or better ^-tree in which a and b are siblings, by exchanging the labels a and b with the labels a and b with the maximum possible total value subject to having total
size at most C and at most k items. Suppose we install the root of a tree T1 under an object z of another tree T2. Second, this approach promotes the understanding of algorithms work, but even better, you'll add to your programmer toolbox a
general and flexible algorithm design technique that you can apply to problems that come up in your own projects. A graph G = (V, E)—not necessarily connected—naturally falls into "pieces" called the connected components of the graph. This brings us to the revised version of the single-source shortest path problem. In the second, we'll use an
exchange argument to show that every feasible solution can be iteratively massaged into the output of the greedy algorithm, 13.4.1 The No-Ties Case: High-Level Plan We proceed to the proof of Theorem 13.1. Fix a set of jobs, with positive weights w1, w2, Quiz 15.3 In
Figure 15.5, suppose the vertex x is extracted and moved to the set X. The inspired idea is to use the differences between and ← to explicitly construct a schedule. We can interpret the labels of these two edges as an encoding of the leaf's symbol. With non-
uniform search frequencies, however, there's no reason to expect the median to be a good choice for the root (see Quiz 17.4). Tim Roughgarden is a Professor of Computer Science at Columbia University. The output has an iterative structure, with jobs processed one by one. Topics covered in the other three parts. 15.1.3 Solution to Quiz 15.1 Correct
answer: (b). If i = j, return. 18.4 197 The Floyd-Warshall Algorithm The issue is that the concatenation of two cycle-free paths need not be cycle-free paths need not be cycle-free. Define the lateness j () of a job j in a schedule as the difference Cj () dj between the job's completion time and deadline, or as 0 if Cj () dj. Over the years, countless students have regaled me with
stories about how mastering the concepts in these books enabled them to ace every technical interview question they were ever asked. Output: the NW score of X and Y. Finally, we can represent our prefix-free variable-length code 30 Huffman Codes Symbol A B C D Encoding 0 10 110 111 with the tree 1 0 A 1 0 B 0 C 1 D More generally, every
binary code can be represented as a binary tree with left and right child edges labeled with "0" and "1," respectively, and with each symbol of the alphabet used as the label for exactly one node. 4 Conversely, every such tree defines a binary code, with the edge labels on the paths from the root to the labeled nodes providing the symbol encodings. The
in- and outdegree of a vertex is the number of incoming and outgoing edges, respectively.) a) 2 b) 1 + the in-degree of v d) n (See Section 18.2.3 Recurrence As usual, the next step is to compile our understanding of optimal substructure into a recurrence that implements exhaustive
search over the possible candidates for an optimal solution. In the heap-based implementation of Prim's algorithm, the objects in the heap correspond to the as-yet-unprocessed vertices (V X in the Prim pseudocode).15,16 The key of a vertex w 2 V X is defined as the minimum cost of an incident crossing edge (Figure 15.5). Bounding the total time
spent in lines 12-16 is the tricky part; the key observation is that each edge of G is examined in line 12 exactly once, in the iteration in which the first of its endpoints gets sucked into X (i.e., plays the role of w-). His books include Twenty Lectures on Algorithmic Game Theory (2016) and the Algorithms Illuminated book series (2017-2019). For
example, if m = -(n2), the running time is quartic in n—a running time we haven't seen before and, hopefully, will never see again! 18.3.3 Solution to Quiz 18.4 Correct answer: (c). Huffman coding is a widely-used method for lossless compression.
small as possible. The third step is easy: Return the solution to the largest subproblem, which is the same as the original problem. b) At each iteration, choose the remaining job with the earliest start time. The worst-case running time of Find is proportional to the biggest height of a tree of the parent graph. Let v 6= k be some other vertex of C: P1 v k
P2 The two sides P1 and P2 of the cycle are cycle-free v-k and k-v paths with internal vertices restricted to {1, 2, . Other algorithms, like k-means clustering, maintain k clusters from beginning to end. Quiz 15.6 Suppose we arbitrarily choose which root to promote. For example, one of the simplest choices for F is the best-case similarity between
points in the different clusters: F (S1, S2) = min x2S1, y2S2 f (x, y). Lemma 15.18 implies that every edge of this spanning tree satisfies the MBP. How quickly can you solve the single-source shortest path problem? The second step shows that, in a graph with distinct edge costs, a spanning tree with this property must be a minimum spanning tree.19
15.4.1 The Minimum Bottleneck Property We can motivate the minimum bottleneck property by analogy with Dijkstra's shortest-path algorithm. Thus, the total work performed outside the double for loop, leading to an overall
running time bound of O(mn). Because · m 2 · 2 2, an MST T i=1 i=1 0 of G must also be one of G. `j These two solutions are readily available, it makes sense to work bottom-up, starting with the base cases and building up to ever-larger subproblems. Is S {n} and ST T i=1 i=1 0 of G must also be one of G. `j These two solutions are readily available, it makes sense to work bottom-up, starting with the base cases and building up to ever-larger subproblems. Is S {n} and ST T i=1 i=1 0 of G must also be one of G. `j These two solutions are readily available, it makes sense to work bottom-up, starting with the base cases and building up to ever-larger subproblems. Is S {n} and ST T i=1 i=1 0 of G must also be one of G. `j These two solutions are readily available, it makes sense to work bottom-up, starting with the base cases and building up to ever-larger subproblems. Is S {n} and ST T i=1 i=1 0 of G must also be one of G. `j These two solutions are readily available, it makes sense to work bottom-up, starting with the base cases and building up to ever-larger subproblems.
optimal solution to a smaller subproblem? The main idea is to reason about the structure of an optimal solution, identifying the different ways it might be constructed from optimal solution may have been cached by the NW algorithm, in which case it can be looked
up. Floyd and Stephen Warshall, but also discovered independently by a number of other researchers in the late 1950s and early 1960s. Then, every schedule has exactly the same set of completion times—{1, 2, 3, . The Initialize operation clearly runs in O(n) time. Output: One of the following: 17 Technically, this assumes that m is at least a constant
times n, as would be the case if, for example, every vertex v was reachable from the source vertex s. Wunsch, and published in the paper "A general method applicable to the search for similarities in the amino acid sequence of two proteins" (Journal of Molecular Biology, 1970). For example, consider two unit-weight jobs with lengths 1 and 2. A node
with no children is called a leaf. 16.2 A Linear-Time Algorithm for WIS in Paths 113 same subproblem later, we can look up its solution in the cache in constant time. 6 Blending caching into the pseudocode on page 111 is easy. If subproblem solutions stabilize (with Ln,v = Ln 1,v for all v 2 V), Lemma 18.3 implies that the Ln 1,v 's are the correct
shortest-path distances. 94 Minimum Spanning Trees 15.8 Application: Single-Link Clustering Unsupervised learning is a branch of machine learning is a branch of machine learning and statistics that strives to understand large collections of data points by finding hidden patterns in them. First, jobs k other than i and j couldn't care less about i and j being swapped., 10} and three
connected components. 10 Starred sections like this one are the more difficult sections; they can be skipped on a first reading. What are the final array entries of the Knapsack algorithm from Section 16.5, and which items belong to the optimal solution? How can we use this fact to exhibit another schedule 0 that is even better than -, thereby
furnishing a contradiction? Suppose k 2 {1, 2, . If v = w, the only v-w path with no internal vertices is the empty path (with completion time 3) and one with weight 20 fifth (with completion time 5); you'd be better off exchanging the
positions of these two jobs, which would decrease the sum of weighted completion times by 20 (as you should check). The WIS problem is challenging even in the super-simple case of path graphs. With this in mind, the next guiz asks you to guess what type of optimal substructure lemma might be true for the optimal binary search tree problem. For
example, "bcd" is a substring of "abcdef," while "bdf" is not.) 20 A dynamic programming algorithm for the longest common subsequence problem underlies the diff command familiar to users of Unix and Git. Kruskal Input: connected undirected graph G = (V, E) in adjacency-list representation and a cost ce for each edge e 2 E. We'll then narrow the
field to a single candidate algorithm and prove that this candidate correctly solves the problem. As in the WIS problem on path graphs, we'll be able to reconstruct the items in an optimal solution to the original problem from this information. 1 2 3 4 5 6 7 if n = 0 then // base case #1 return the empty set if n = 1 then // base case #2 return {v1} } //
recursion when n 2 S1 := recursively compute an MWIS of Gn 1 S2 := recursively compute an MWIS of Gn 2 return S1 or S2 [ {vn }, whichever has higher weight A straightforward proof by induction shows that this algorithm is guaranteed to compute a maximum-weight independent set.4 What about the running time? Each of the chapters
concludes with a suggested programming project whose goal is to help you develop a detailed understanding of an algorithm by creating your own working implementation of it. Output: One of the following: (i) the shortest-path distance dist(s, v) for every vertex v 2 V; or (ii) a declaration that G contains a negative cycle. The total value of the items
should be as large as possible while still fitting in the knapsack, meaning their total size should be at most C. Hint for Problem 15.1: To reason about T, use Corollary 15.8 or the minimum bottleneck property (page 70). A straightforward implementation keeps track of which vertices are in X by associating a Boolean variable with each vertex. Every
path from the root to a labeled node traverses two edges. (The optimal prefix-free code problem is uninteresting with a one-symbol alphabet.) In the inductive step, we assume that k > 2. 29 Had I shown you the Floyd-Warshall algorithm before your boot camp in dynamic programming, your response might have been: "That's an impressively elegant
```

```
algorithm, but how could I ever have come up with it myself?" Now that you've achieved a black-belt (or at least brown-belt) level of skill in the art of dynamic programming, I hope your reaction is: "How could I not have come up with this algorithm (ii)
there is no path from a vertex v to itself shorter than the empty path (which has length 0). 18.2 The Bellman-Ford algorithm The Bellman-Ford algorithm the sense that it either computes the correct shortest-path distances or correctly determines that the input
graph has a negative cycle.8 This algorithm will follow naturally from the design pattern that we used in our other dynamic programming case studies. Hint for Problem 16.2: For (a) and (c), revisit the four-vertex example on page 105. 4 You cannot reduce the single-source shortest path problem with general edge lengths to the special case of
nonnegative edge lengths by adding a large positive constant to the length of every edge. 166 Advanced Dynamic Programming Problem 17.7 (H) There are seven kinds of Tetris pieces. 21 Design a dynamic programming algorithm that, given x1, x2, . The Bellman-Ford algorithm solves (n + 1) \cdot n = O(n2) different subproblems, where n is the
number of vertices. P Exchange arguments are among the most common techniques used in correctness proofs for greedy algorithms. 18.2 173 The Bellman-Ford algorithm takes a different tack, one inspired by the inherently sequential nature of the output of the single-source shortest path problem (i.e., of paths). Thus
Case 1 contributes one candidate and Case 2 a number of candidates equal to the in-degree of v. Let's assume that such questions have already been answered experimentally, in the form of known penalties for gaps and mismatches that are provided as part of the input (along with the two strings). Because each recursive call of a dynamic
programming algorithm tries out multiple choices of smaller subproblems, subproblems generally recur across different recursive calls; caching subproblems of the input graph in linear time using breadth- or depth-first search (see
Chapter 8 of Part 2), and then applying an algorithm for the MST problem to each component separately. We can assume that not all edges have the same cost, and more generally that not all spanning trees have the same total cost (why?). Well, not quite: Cycles are a no-no, so it chooses the cheapest edge that doesn't create a cycle. Tacking a
negative-length edge onto a path would decrease its overall length, so nonnegative edge lengths are necessary for path monotonicity. The second job must wait for the first two jobs, which is 3. Because v and w are on different sides of the cut (A, B), P includes an edge e0 that
crosses (A, B)., yd) are d-dimensional vectors, the qP d precise formula is f (x, y) = yi) 2. This wraps up the proof of (†)., k1}; or 192 Shortest Paths Revisited (ii) the concatenation of a minimum-length cycle-free v-k path with all internal vertices in {1, 2, . Here are two spanning trees with an inferior total cost of 8: a 4 c 1 a b 1 b 3 2 c d 5 d The
three edges (a, b), (b, d), and (a, d) have a smaller total cost of 6: a 1 b 3 c 2 d but these edges do not form a spanning tree. Because 1, the edges of G0 have distinct costs, with edge ei the ith-cheapest edge of G0 a reduced in the Bellman-Ford algorithm is to 198 Shortest Paths Revisited parameterize subproblems by an edge budget i (in
addition to a destination) and consider only paths with i or fewer edges. 24 Ignore the uninteresting case in which m is much smaller than n; see also footnote 17. A key ingredient in the correctness proof for Dijkstra's algorithm is "path monotonicity," meaning that tacking on additional edges at the end of a path can only make it worse. a) T must be
an MST and P must be a shortest s-t path. Few benefits of a serious study of algorithms rival the empowerment that comes from mastering dynamic programming. I'm not using the term lightly; I'm using it precisely. use a greedy algorithm that iteratively decides which job should go next? As usual, the first step merges the two symbols with the
smallest frequencies, namely "B" and "E": 14.3 39 Huffman's Greedy Algorithm 4 3 2 6 8 2 6 A B C D F B E B E B E The five trees containing D tree containing B and E Sum of Symbol Frequencies 3 6 8 6 2+2 = 4 and the algorithm next merges the
first and last of these: 4 3 6 8 6 A C D F B 7 E 6 8 6 C D F A The four remaining C tree containing D 
containing D tree containing C and F tree containing C and F tree containing A, B and E Sum of Symbol Frequencies 8 6 + 6 = 12 7 The penultimate merge is of the first and third trees: 40 Huffman Codes 15 7 12 12 8 D D C F A B E A B E C F and the final merge produces the output of the algorithm: 0 1 0 1 0 A 1 0 0 D 1 B 1 C F E This tree corresponds to the following
prefix-free code: Symbol A B C D E F 14.3.6 Encoding 000 0010 10 01 11 Running Time A straightforward implementation of the Huffman algorithm runs in O(n2) time, where n is the number of symbols., n} to a distinct value f (i) 2 {1, 2, . As with all our dynamic programming algorithms, the correctness of the OptBST algorithm follows by
induction (on the subproblem size), with the recurrence in Corollary 17.5 justifying the inductive step. This chapter and the next two provide this practice through a half-dozen detailed case studies, including several algorithms belonging to the greatest hits compilation. Delete: given a heap H and a pointer to an object x in H, delete x from H. These
observations lead to the following recurrence, which computes the best of the three options by exhaustive search: Corollary 17.2 (Sequence Alignment of Xi = x1, x2, Problem 15.2 (H) Consider the following algorithm that
attempts to compute an MST of a connected undirected graph G = (V, E) with distinct edge costs by running Kruskal's algorithm "in reverse": Kruskal (Reverse Version) T := E sort edges of E in decreasing order of cost for each e 2 E, in order do if T {e} is connected then T := T {e} return T 100 Minimum Spanning Trees Which of the following
statements is true? For example, the input might be the strings AGGCT and AGGCA. For a clue, look at our three examples. The process by which we arrive at this algorithm is more important to remember than the algorithm is more important to remember than the algorithm itself; it's a repeatable process that you can use in your own applications. Subproblems of the form W1,r 1 in the recurrence
(17.3) are defined by prefixes of the set of keys (as usual), but subproblems of the form Wr+1,n are defined by suffixes of the keys., yn over the alphabet ^- = {A, C, G, T}, a penalty ^{\downarrow}xy for each x, y 2 ^-, and a gap penalty ^{\downarrow}xy for each x, y 2 ^-, and a gap penalty ^{\downarrow}xy for each x, y 2 ^-, and a gap penalty ^{\downarrow}xy for each x, y 2 ^-, and a gap penalty ^{\downarrow}xy for each x, y 2 ^-, and a gap penalty ^{\downarrow}xy for each x, y 2 ^-, and a gap penalty ^{\downarrow}xy for each x, y 2 ^-, and a gap penalty ^{\downarrow}xy for each x, y 2 ^-, and a gap penalty ^{\downarrow}xy for each x, y 2 ^-, and a gap penalty ^{\downarrow}xy for each x, y 2 ^-, and a gap penalty ^{\downarrow}xy for each x, y 2 ^-, and a gap penalty ^{\downarrow}xy for each x, y 2 ^-, and a gap penalty ^{\downarrow}xy for each x, y 2 ^-, and a gap penalty ^{\downarrow}xy for each x, y 2 ^-, and a gap penalty ^{\downarrow}xy for each x, y 2 ^-, and a gap penalty ^{\downarrow}xy for each x, y 2 ^-, and a gap penalty ^{\downarrow}xy for each x, y 2 ^-, and a gap penalty ^{\downarrow}xy for each x, y 2 ^-, and a gap penalty ^{\downarrow}xy for each x, y 2 ^-, and a gap penalty ^{\downarrow}xy for each x, y 2 ^-, and a gap penalty ^{\downarrow}xy for each x, y 2 ^-, and a gap penalty ^{\downarrow}xy for each x, y 2 ^-, and a gap penalty ^{\downarrow}xy for each x, y 2 ^-, and a gap penalty ^{\downarrow}xy for each x, y 2 ^-, and a gap penalty ^{\downarrow}xy for each x, y 2 ^-, and a gap penalty ^{\downarrow}xy for each x, y 2 ^-, and a gap penalty ^{\downarrow}xy for each x, y 3 ^-xy 
of size n 2. If only we had clairvoyance and knew which key should appear at the root, we might then be able to compute the rest of the tree recursively. The main loop has m iterations. 14.2 31 Codes as Trees 14.2.2 Which Trees Represent Prefix-Free Codes? 58 Minimum Spanning Trees The algorithm's initial (empty) tree spans only the starting
vertex b. That's a lot of schedules! Which one should we prefer? Moreover, in the special cases in which sorting is possible in linear time, 9 14.3.7 Solution to Quiz 14.3 Correct answer: (a). For example, in Figure 13.2, if jobs 2 and 3 were processed in the opposite order they would
constitute a consecutive inversion (with i = 3 and j = 2). Let M \leftarrow denote the cost of an MST of G, and M the minimum cost of a suboptimal spanning tree of G. The initial parent graph consists of isolated vertices with self-loops: For the Find operation, we leap from parent to parent until we arrive at a root object, which can be identified by its self-loop.
Thus the GreedyDiff algorithm schedules the second job first, while GreedyRatio does the opposite. For bonus points, implement the space optimizations and linear-time reconstruction algorithms outlined in footnotes 13, 28, and 31. Theorem 17.3 (Properties of NW) For every instance of the sequence alignment problem, the NW algorithm returns the
correct NW score and runs in O(mn) time, where m and n are the lengths of the two input strings., pn, the OptBST algorithm runs in O(m3) time and returns the minimum weighted search time of a binary search tree with keys {1, 2, . Our first lemma 23 Theorem 15.6 does not hold as stated in graphs with non-distinct edge costs. The difference lies
in the constraint that the binary tree must satisfy., Gn. There is a one-to-one correspondence between ^0 -trees and the restricted set Tab of ^-trees (Figure 14.2). However, we want to establish its superiority to all competing s-w paths with i 1 edges or fewer. But after removing the last vertex from S, we can: S {vn } contains neither vn 1 nor vn
and hence can be regarded as an independent set of the smaller graph Gn 2 (with total weight W wn ). The merger T3 of T1 and T2 is inserted at the back of Q2. QE D 15.5 Kruskal's algorithm.25 With our blazingly fast heapbased implementation
of Prim's algorithm, why should we care about 25 Discovered by Joseph B. The Huffman algorithm can be implemented by sorting the symbols in order of increasing frequency and then performing a linear amount of additional processing. We conclude that 2 has exactly the same inversions as 1, except with the inversion of i and j removed. What is
the NW score of X and Y? 2 Readers of Part 2 have already seen a greedy algorithm, namely Dijkstra's shortest-path distance dist(v, w) for
every ordered vertex pair v, w 2 V; or (ii) a declaration that G contains a negative cycle. Programming Problems Problems Problems Problems Problems 17.8 Implement in your favorite programming language the NW and OptBST algorithms, along with their reconstruction algorithms.
keys of vertices still in V X to maintain the invariant. It's often tricky to come up with a good divide-and-conquer algorithm for a problem. In this case, we've salvaged the proof: Pb is a cycle-free v-w path with all internal vertices in {1, 2, . Hint for
Problem 16.5: For (b) and (c), add a third parameter to the dynamic programming solution to the original knapsacks? Chapter 18 Shortest Paths Revisited This chapter centers around two famous dynamic programming algorithms
for computing shortest paths in a graph. Conversely, every tree T 2 Tab can be turned into a ^0 -tree T 0 by sucking the leaves labeled a and b into their (common) parent and labeling the resulting "meta-node" with "ab." We denote this inverse mapping T 7! T 0 by 4(T). For an example of the NW algorithm in action, check out Problem 17.1.7 17.1.7
Reconstruction There are no surprises in reconstructing an optimal alignment from the array values that the NW algorithm computes. GreedyDiff Schedule the jobs in decreasing order of wj (breaking ties arbitrarily). Because every ^-tree is equaled or bettered by a tree of Tab , Tab contains a tree that is optimal among all ^-trees. A cut of an
undirected graph G = (V, E) is a partition of its vertex set V into two non-empty sets, A and B. In each iteration, the trees T1 and T2 can be removed from the heap using two successive ExtractMin operations, and the merged tree T3 added with one Insert operation (with T3 's key set to the sum of the keys of T1 and T2). 21 As we'll see, it's no
accident that the edges satisfying the MBP in this example are precisely the edges in the minimum spanning tree. c) Schedule the jobs in increasing order of the product dj · pj . In each, the input consists of item values v1 , v2 , . P The knapsack problem can be solved using dynamic programming in O(nC) time. Single-Link Clustering via Kruskal's
 Algorithm 1. b) 2 has the same number of inversions as c) 2 has one more inversion than 1., 7} and the following frequencies: Symbol 1 2 3 4 5 6 7 Frequency 20 5 17 10 20 3 25 What are the final array entries of the OptBST algorithm (Section 17.1), and
the OptBST algorithm (Section 17.2). The first iteration of the greedy algorithm commits to the maximum-weight vertex, which is the third vertex (with weight 5). That is, is it true that, whenever Lk+1,v = Lk,v for some k 0 and destination v, Li,v = Lk,v for all i k? 164 Advanced Dynamic Programming a) Neither algorithm remains well defined and
correct after reversing the order of the for loops. , j} and frequencies pi , pi+1 , . Then, armed with a template for developing dynamic programming algorithms and an example instantiation, we'll tackle increasingly challenging applications of the paradigm. Ford, Jr., though ironically Alfonso Shimbel appears to have been the first. *17.2 159 Optimal
Binary Search Trees 17.2.7 A Dynamic Programming Algorithm to write itself. Thus, we're after an algorithm to write itself. Thus, we're after an algorithm that either computes the correct shortest-path distances or offers a compelling excuse for its failure (in 5 More precisely, any polynomial
time algorithm for any N P -hard problem would disprove the famous "P 6= N P" conjecture and resolve what is arguably the most important open question in all of computer science. d) The formula is always correct in arbitrary graphs. This implies that G has a negative cycle with no repeated vertices other than its start and end. Most of the
canonical applications of the divide-and-conquer paradigm replace a straightforward polynomial-time algorithm for a task with a faster divide-and-conquer paradigm replace a straightforward solutions (like exhaustive search) require an
exponential amount of time., k\}. P In the single-source shortest path problem, the input consists of a directed graph with edge lengths and a source vertex. For example, you might be wondering if species B evolved independently from A. The graph (V, T) is connected if and only if its connected if an analysis of a directed if a direc
contains no cycles. If Y is empty, the optimal alignment inserts enough gaps into Y so that it has the same length as X. 15.2 57 Prim's Algorithm, which is named after Robert C. 9 Readers of Part 2 should recognize strong similarities to Dijkstra's
shortestpath algorithm. The path P must be an optimal solution to the smaller subproblem; any superior for the original subproblem, a contradiction. Every fixed-length code is automatically prefix-free. The trickier case is when the optimal solution S makes use of the last item n. Figure 15.1: An undirected graph with
five vertices and eight edges. The length of P 0 is the same as that of P, less the sum of the edge lengths in the spliced-out cycle. Easy to analyze the running time., (xp 1, xp), where x0 is v and xp is w. A poor implementation can lead to a tree with height as large as n 1: *15.7 Kruskal's Algorithm: Proof of Correctness 91 In this sense, answer (c) is
 also correct. 17.1 145 Sequence Alignment NW Input: strings X = x1, x2, For bonus points, implement the heap-based version of Prim (Section 15.3) and the union-find-based version of Kruskal (Section 15.6). Because Lk,v is the minimum length of an s-v path with at most k hops, there must be an s-v path with i > k hops and length smaller than
Lk,v. The reconstruction process then resumes from vn 2.8 WIS Reconstruction Input: the array A computed by the WIS algorithm for a path (cycle-free, length L1) (all internal vertices in \{1,2,...,k-1\}) Reconstruction Input: the array A computed by the WIS algorithm for a path (cycle-free, length L2) (all internal vertices in \{1,2,...,k-1\}) Reconstruction Input: the array A computed by the WIS algorithm for a path (cycle-free, length L2) (all internal vertices in \{1,2,...,k-1\}) Reconstruction Input: the array A computed by the WIS algorithm for a path (cycle-free, length L3) (all internal vertices in \{1,2,...,k-1\}) Reconstruction Input: the array A computed by the WIS algorithm for a path (cycle-free, length L3) (all internal vertices in \{1,2,...,k-1\}) Reconstruction Input: the array A computed by the WIS algorithm for a path (cycle-free, length L3) (all internal vertices in \{1,2,...,k-1\}) Reconstruction Input: the array A computed by the WIS algorithm for a path (cycle-free, length L3) (all internal vertices in \{1,2,...,k-1\}) Reconstruction Input: the array A computed by the WIS algorithm for a path (cycle-free, length L3) (all internal vertices in \{1,2,...,k-1\}) Reconstruction Input: the array A computed by the WIS algorithm for a path (cycle-free, length L3) (all internal vertices in \{1,2,...,k-1\}) Reconstruction Input: the array A computed by the WIS algorithm for a path (cycle-free, length L3) (all internal vertices in \{1,2,...,k-1\}) and the will be array A computed by the WIS algorithm for a path (cycle-free, length L3) (all internal vertices in \{1,2,...,k-1\}) and the will be array A computed by the WIS algorithm for a path (cycle-free, length L3) (all internal vertices in \{1,2,...,k-1\}) and the will be array A computed by the WIS algorithm for a path (cycle-free, length L3) (all internal vertices in \{1,2,...,k-1\}) and the will be array as a path (cycle-free, length L3) (all internal vertices in \{1,2,...,k-1\}) and the will be array as a path (cycle-free, length L3) (all interna
free, length L, all internal vertices in {1,2,...,k}) The vertex k appears in P only once (because P has no cycles) and therefore is not an internal vertex of P1 or P2. The next quiz contains the key to unlocking a radical running time improvement. Which of them can be solved in O(mn) time? In other words, increasing the edge budget i from n 1 to n (or
to any bigger number) has no effect on the minimum length of an s-v path. If there wasn't an efficient algorithm for computing the NW score, Needleman and Wunsch surely would have proposed a different and more tractable definition of genome similarity! 17.1 141 Sequence Alignment Quiz 17.2 Let X = x1, x2, . Test Your Understanding Problem
13.1 (H) You are given as input n jobs, each with a length 'j and a deadline dj. Additional Resources These books are based on online courses that are currently running on the Coursera and Stanford Lagunita platforms. First, T should not contain a cycle (this is the "tree" part). Second, for every pair v, w 2 V of vertices, T should include a path
between v and w (this is the "spanning" part).4 1 There is an analog of the MST problem for directed graphs, which is known as both the minimum-cost arborescence problem and the optimum branching problem. Consider an optimal binary search tree T with keys {1, 2, . algorithmsilluminated.org for test cases and challenge data sets.) Problem 16.7
Implement in your favorite programming language the Knapsack and Knapsack Reconstruction algorithms. The rest of the work—initialization, updating F, rewiring pointers when merging two trees—contributes only O(n) operations to the overall running time bound, for a total of O(n2). By the time an iteration of the main loop must compute the
 subproblem solution A[i], the values A[i 1] and A[i 2] of the two relevant smaller subproblems have already been computed in previous iterations (or in the base cases). 8 Introduction to Greedy Algorithms a) larger/shorter b) smaller/longer d) smaller/longer d) smaller/longer d) smaller/longer discussion.) 13.3.2 Dueling Greedy
Algorithms In the general case, jobs can have different lengths and O(m) time for input graphs with a negative cycle and O(n3) time otherwise. I am particularly grateful to those who supplied detailed feedback on an earlier draft of this book: Tonya Blust,
Yuan Cao, Carlos Guia, Jim Humelsine, Vladimir Kokshenev, Bayram Kuliyev, and Daniel Zingaro. On Lemmas, Theorems, and the Like In mathematical writing, the most important technical statements are labeled theorems. Problem: Single-Source Shortest Paths (Preliminary Version) Input: A directed graph G = (V, E), a source vertex s 2 V, and a
real-valued length 'e for each edge e 2 E.1 Output: the shortest-path distance dist(s, v) for every vertex v 2 V. (Each edge is labeled with its cost.) a 4 c 1 b 3 5 2 d a) 6 b) 7 c) 8 d) 9 (See Section 15.1.3 for the solution and discussion.) It makes sense only to talk about spanning trees of connected graphs G = (V, E), in which it's possible to travel from
any vertex v 2 V to any other vertex w 2 V to any other vertex w 2 V using a path of edges in E.5 (If there 5 For example, the graph in Figure 15.1 is connected, while the graph in Figure 15.1 is conne
execution. We're not responsible for computing shortest-path distances for input graphs with a negative cycle. Let L, L1, and L2 denote the lengths of P, P1, and P2, respectively. Key to our approach is the following thought experiment:
Solution to Quiz 13.1 Correct answer: (c). 16.2 109 A Linear-Time Algorithm for WIS in Paths with total weight W. We'll see several over the next few chapters. 2 13.1.2 Themes of the Greedy Paradigm Here are a few themes to watch for in our examples. wn . 17.1 147 Sequence Alignment 17.1.8 Solution to Quizzes 17.1-17.3 Solution to Quiz 17.1
Correct answer: (b). For a job k processed after i and j in \leftarrow (as part of the "more stuff" in Figure 13.3), the set of jobs completed before k is exactly the same in \leftarrow and in 0. For example, because the path as encoding the
symbol B by 01. This completes the proof of the inductive step and of Theorem 14.2. QE D 49 Problems The Upshot P Prefix-free variable-length binary codes can have smaller average encoding lengths than fixedlength codes when different alphabet symbols have different frequencies. Otherwise, there is a type-C edge addition, which by Lemma
15.7(a) creates a cycle and also fails to decrease the number of connected components: In this case, the process starts with n connected components and the n 1 edge additions decrease the number of connected components. Try dynamic programming. The base
cases A[0] and A[1] are clearly correct. WIS Input: a path graph G with vertex set {v1, v2, . QE D Lemma 18.3 promises that it's safe to stop as soon as subproblem must satisfy the more challenging *17.2 Optimal Binary Search
Trees 153 search tree property (page 149). (b) Does this imply a linear-time algorithm for computing the total cost of an MST? 1 a b 3 4 c 2 d 5 vertices spanned so far The former is cheaper, so the algorithm for computing the total cost of an MST? 1 a b 3 4 c 2 d 5 vertices spanned so far The former is cheaper, so the algorithm for computing the total cost of an MST? 1 a b 3 4 c 2 d 5 vertices spanned so far The former is cheaper, so the algorithm for computing the total cost of an MST? 1 a b 3 4 c 2 d 5 vertices spanned so far The former is cheaper, so the algorithm for computing the total cost of an MST? 1 a b 3 4 c 2 d 5 vertices spanned so far The former is cheaper, so the algorithm for computing the total cost of an MST? 1 a b 3 4 c 2 d 5 vertices spanned so far The former is cheaper, so the algorithm for computing the total cost of an MST? 1 a b 3 4 c 2 d 5 vertices spanned so far The former is cheaper, so the algorithm for computing the total cost of an MST? 1 a b 3 4 c 2 d 5 vertices spanned so far The former is cheaper, so the algorithm for computing the total cost of an MST? 1 a b 3 4 c 2 d 5 vertices spanned so far The former is cheaper, so the algorithm for computing the total cost of an MST? 1 a b 3 4 c 2 d 5 vertices spanned so far The former is cheaper, so the algorithm for computing the total cost of an MST? 1 a b 3 d c 2 d 5 vertices spanned so far The former is cheaper, so the algorithm for computing the total cost of an MST? 1 a b 3 d c 2 d 5 vertices spanned so far The former is cheaper.
frequencies—into a single "meta-symbol" ab. The second half of this book will teach you a third design paradigm: the dynamic programming paradigm: the dynamic programming fit into the bigger algorithmic picture. (See page 4 for the definition
of a job's completion time in a schedule.) This problem considers the objective of minimizing the maximum lateness, maxnj=1 j (). Because i > j and jobs are indexed in nonincreasing order of weight-length ratio, wj wi 'i | {z} cost of exchange wj 
assuming that there is an optimal schedule - of the given jobs with a strictly smaller sum of weighted completion times than the greedy schedule . And even when a greedy algorithm is correct, proving it can be difficult.4 Features and Bugs of the Greedy Paradigm 1. d) T1 is an optimal binary search tree for the keys {1, 2, . With n! different
schedules, computing the best one by exhaustive search is out of the question for all but the tiniest instances. 4 Introduction to Greedy Algorithms Now that my conscience is clear, let's look at some cherry-picked examples of problems that can be solved correctly with a judiciously designed greedy algorithm. We can't expect a running time better
than quadratic (with a quadratic number of values to report), but there's a big gap between cubic and quadratic running times. Call this new graph G0. If every symbol is encoded using 6 bits, the second symbol always starts with the 7th bit, the third symbol with the 13th bit, and so on., k} of the vertices, with k serving as the measure of
subproblem size, as well as an origin v and destination w. To prove that its output is also connected, we can argue that all its vertices belong to the same connected component of (V, T). One way to make trade-offs between the jobs is to minimize the sum of weighted completion times. In the following pseudocode, the variable s controls the current
subproblem size.15 OptBST Input: keys {1, 2, . Quiz 16.2 What is the four-vertex path on page 105?, sn, and additional problem-specific data (all positive integers). 70 Minimum Spanning Trees The minimum bottleneck property makes this idea precise. 16.2 A Linear-Time
Algorithm for WIS in Paths 16.2.1 Optimal Substructure and Recurrence To quickly solve the WIS problem on path graphs, we'll need to up our game. Main Idea #1 Prove that the output of the Huffman algorithm minimizes the average leaf depth over all ^-trees in which a and b are siblings. Case 2: n 2 S. Intuitively, you might expect that a prefix P 0
of a shortest path P would itself be a shortest path P would itself be a shortest path P? For example, if every element to be sorted is an integer with
magnitude at most n10 (say), the RadixSort algorithm can be used to sort them in O(n) time. *17.2 149 Optimal Binary Search Tree Property For every object x: 1. P Typical dynamic programming algorithms fill in an array with the values of subproblems' solutions, and then trace back through the filled-in array to reconstruct the
solution itself. c) For every vertex v reachable from the source s, there is a shortest s-v path with at most m edges. The key observation is that the straightforward implementation performs minimum computations, over and over, using exhaustive search. If we demote the root of T1, then T1 's occupants are pushed one step further from the new root;
otherwise, T2's occupants suffer the same fate. A[n 2] + wn, an MWIS of Gn 1 is also an MWIS 16.3 117 A Reconstruction Algorithm 2. In our running example, the fourth and sixth objects have depth 1, and the second and third objects have depth 2. By our promotion criterion, this happens only when x's tree is
merged with another tree that is at least as big. (If j = 0 or i = 0, interpret Pi,j as i \cdot 4gap or j \cdot 4gap, respectively.) Then Pm, n = min\{Pm \mid + 4xm \ yn \ , Pm \ \{z \} \mid 1, n \ 1 \ Case \ 1 + 4 \ gap \}. This assertion is not obvious and you should not trust it until I supply you with a proof. Their symbols have the same frequencies in p and p0, so
these leaves contribute the same amount to the average leaf depth of both trees. Can you formulate a recurrence to quickly solve a subproblems? You should check that our dynamic programming solution extends to the case in which unsuccessful search times are also counted, provided the
 includes v1: prefix length i 0\ 1\ 2\ 3\ 4\ 5\ 6\ 0\ 3\ 4\ 9\ 9\ 14 include v1: exclude v2: include v3: include v3:
= 0, 1) are clearly correct., xd) and y = (y1, y2, ..., (vk 1, vk) in a graph to include repeated vertices or, equivalently, to contain one or more cycles. How much space do you then really need to run each of the three algorithms, respectively? 1 The problem can be solved in O(n2) time by a divide-and-conquer algorithm that makes four recursive calls
rather than two, where n is the number of vertices. 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 // Initialization X := \{s\}, T = ;, H := empty heap for every v \in S winner(v \in S) is the number of vertices. 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 // Initialization X := \{s\}, Y = S, Y = S
tournament—the cheapest edge incident to w that crosses the frontier (i.e., edges (v, w) with v 2 X). For a fixed origin v and destination w, the set of allowable paths grows with k, and so Lk,v,w can only decrease as k increases. Indeed, the Bellman-Ford algorithm directly inspired the early Inter13 For reconstruction purposes, it's a good idea to add
one line of code that caches with each vertex v the most recent predecessor w that triggered a Case 2update of the form A[i][v] := A[i \ 1][w] + vv. // subproblems (i indexed from 1, j from 0) A := (n + 1) \rightarrow (n + 1) two-dimensional array // base cases (i = j + 1) for i = 1 to n + 1 do A[i][i \ 1] := 0 // systematically solve all subproblems (i j) for s = 0 to n 1
A[r + 1][i + s] correspond to solutions of smaller subproblems computed in earlier iterations of the outer for loop (or in the base cases). } Case 2 Because both c and items' sizes are integers, the residual capacity c si in the second case is also an integer. In general, an alignment is a way of inserting gaps into one or both input strings so that they have
equal length: | X + gaps Y + gaps Y + gaps Y + gaps Y + gaps Z common length `} We can then define the similarity of two strings according to the quality of their nicest alignment. Each edge e = (v, w) added by the algorithm is the cheapest one with endpoints in distinct connected components of the solution-so-far (as these are precisely the edges whose addition will not
create a cycle). For example, if you invoke Insert four times to add objects with keys 12, 7, 29, and 15 to an empty heap, the ExtractMin operation will return the object with key 7. We conclude that the algorithm spends O(1) time solving each of the O(nC) time solving each of O(nC) time solving each of the O(nC) time solving each of O(nC) time solving each of
the correctness of Knapsack follows by induction on the number of items, with the recurrence in Corollary 16.5 used to justify the inductive step. For whichever algorithm fares worse in this example, we can conclude that it is not always optimal. The latter condition can be checked in linear time using any reasonable graph search algorithm, like
breadth- or depth-first search starting from v (see Chapter 8 of Part 2). By trying all the possibilities —two possibilities in the WIS and knapsack problems, and three in the sequence alignment problem. In math, this objective function translates to n X min wj Cj (), (13.1) j=1 where the minimization is over all n! possible schedules, and Cj () denotes
job j's completion time in the schedule. When with is moved from V X to X, edges of the form (with y 2V X cross the frontier for the first time; these are the new contestants in the local tournaments at the vertices of V X., 6). Lemma 15.15 (Connecting Adjacent Vertices) Let T be the set of edges chosen by Kruskal up to and including the
iteration that examines the edge e = (v, w). Implementing the First Main Idea We now have all our ducks in a row for proving the statement (*), that among all trees of Tab, the Huffman algorithm outputs one with the minimum-possible average leaf depth: 1. Theorem 13.1 (Correctness of GreedyRatio) For every set of positive job weights w1, w2
 If you give me a path P between the source vertex s and some destination vertex v containing at least n edges, I can give you back another s-v path P 0 with fewer edges than P and length no longer than that of P. Kruskal (Union-Find-Based) Input: connected undirected graph G = (V, E) in adjacency-list representation and a cost ce for each edge e 2
E. But then C {(v, w)} would be a v-w path in G, contradicting our assumption that v and w are in different connected components. a) 6; no b) 6; yes (See Section 16.1.4 for the solution and discussion.) 16.1 107 The Weighted Independent Set Problem Chapters 13-15 spoiled us with a plethora of cherry-picked correct greedy
 algorithms, but don't forget the warning back on page 3: Greedy algorithms are usually not correct. For (b), the answer appears to be no. x2 This quantity is the same as the average encoding length of the corresponding prefix-free code. The subproblems that can arise in this way correspond to the contiguous subsets of the original set of keys—sets
of the form {i, i + 1, . To this end, the next lemma shows that once an edge (v, w) is processed by Kruskal, the solution-so-far (and, hence, the final output) necessarily contains a v-w path. The problem arises naturally in several application domains, including computer networking (try a Web search for "spanning tree protocol") and machine learning
(see Section 15.8). Three reasons. As noted in Quiz 14.3, each merge decreases the number of trees in F by 1, resulting in n 1 iterations of the main loop. 15.5 79 Kruskal's Algorithm 15.5.2 Pseudocode With our intuition solidly in place, the following pseudocode won't surprise you. 13.2.1 The Setup In scheduling, the tasks to be completed are usually
called jobs, and jobs can have different characteristics. Nielsen, and Charles Rackoff (Algorithmica, 2003). By the algorithm, the relevant measure is the maximum cost of an edge in a path, and this measure cannot decrease as additional (positive- or negative-cost) edges are tacked onto the path. For example, a data
point corresponding to a 100-by-100 pixel color image might be a 30000-dimensional vector, with 3 coordinates per pixel recording the intensities of red, green, and blue in that pixel.38 This section highlights a connection between one of the most basic algorithms in unsupervised learning and Kruskal's minimum spanning tree algorithm. Quiz 18.10 pixel color image might be a 30000-dimensional vector, with 3 coordinates per pixel recording the intensities of red, green, and blue in that pixel.38 This section highlights a connection between one of the most basic algorithms in unsupervised learning and Kruskal's minimum spanning tree algorithms.
Consider an instance of the single-source shortest path problem with n vertices, m edges, a source vertex s, and no negative cycles. In any case, once the function F is chosen, the generic bottom-up clustering algorithm can be specialized to greedly merge the "most similar" pair of clusters in each iteration: Bottom-up clustering algorithm can be specialized to greedly merge the "most similar" pair of clusters in each iteration: Bottom-up clustering algorithm can be specialized to greedly merge the "most similar" pair of clusters in each iteration: Bottom-up clustering algorithm can be specialized to greedly merge the "most similar" pair of clusters in each iteration: Bottom-up clustering algorithm can be specialized to greedly merge the "most similar" pair of clusters in each iteration: Bottom-up clustering algorithm can be specialized to greedly merge the "most similar" pair of clusters in each iteration in the special part of the similar in the special part of the similar in the special part of the special part of the similar in the special part of the special part of the similar in the special part of the spec
 while C contains more than k clusters do remove from C the clusters S1, S2 that minimize F (S1, S2) // e.g., with F as in (15.2) add S1 [ S2 to C return C 41 Bottom-up clustering is only one of several common approaches to clustering. Here are two reasons among many. Algorithms Illuminated Part 3: Greedy Algorithms and Dynamic Programming
Tim Roughgarden c 2019 by Tim Roughgarden All rights reserved. This is true for every choice of F algorithm, we can reduce the running time of Kruskal's algorithm from the reasonable polynomial bound of O(mn) (Proposition 15.12) to the blazingly fast
near-linear bound of O(m log n) through the deft use of a data structure. Floyd-Warshall Input: directed graph G = (V, E) in adjacency-matrix representation, and a real-valued length `e for each edge e 2 E. Summarizing, two and only two candidates are vying to be an MWIS: Lemma 16.1 (WIS Optimal Substructure) Let S be an MWIS
of a path graph G with at least 2 vertices. Because S does not include the last vertex of Gn 1 (still with total weight W)—and not just any old independent set of Gn 1 , but a maximum-weight such set. b) P - need not have destination w. (Do you see why a negative gap
penalty—that is, a reward for gaps—would make the problem completely uninteresting?) 3 Named after its inventors, Saul B. We continue to allow negative edge lengths and negative edge lengths edge lengths and negative edge lengths edge leng
is to solve two special cases of the general problem. How can we formalize this intuition? The completes at the same time in both schedules. (We can ignore the fact that edges of the form (u, w-) with u 2 X get sucked into X
and no longer cross the frontier, as we're not responsible for maintaining keys for vertices in X.) For a vertex y 2 V X, the new winner (stored in winner(y)) or the new contestant (w-, y). Corollary 15.9 (Connectedness and Acyclicity Go Together) Let G = (V, E) be a graph and T \( \subseteq E \) a subset of n 1 edges
where n = |V|. 13.4.2 Exchanging Jobs in a Consecutive Inversion Suppose, for contradiction, that the GreedyRatio algorithm produces the schedule with a strictly smaller sum of weighted completion times. In the first place I was interested in planning, in decision making, in thinking. Literacy with computer
science's greatest hits. 16.5 127 The Knapsack Problem Corollary 16.5 (Knapsack Recurrence) With the assumptions and notation of Lemma 16.4, let Vi,c denote the maximum total value of a subset of the first i items with total size at most c. 154 Advanced Dynamic Programming Quiz 17.5 Suppose an optimal binary search tree for the keys {1, 2, ...
You'll learn several blazingly fast subroutines for processing data as well as several useful data structures for organizing data that you can deploy directly in your own programs. 18.1 Shortest Paths with Negative Edge Lengths 18.1.1 The Single-Source Shortest-Path Problem In the single-source shortest path problem, the input consists of a directed
graph G = (V, E) with a real-valued length `e for each edge e 2 E and a designated origin s 2 V, which is called an "N P -hard problem. In the presence of a negative cycle, this version of the single-source shortest path problem is what's called an "N P -hard problem."
 Part 4 discusses such problems at length; for now, all you need to know is that N P -hard problems, unlike almost all the problems we've studied so far in this book series, do not seem to admit any algorithm that is guaranteed to be correct and to run in polynomial time. Both options are disasters, so should we give up? *17.2 Optimal Binary Search
Trees In Chapter 11 of Part 2 we studied binary search trees, which maintain a total ordering over an evolving set of objects and support a rich set of fast operations. In the example above, the total weight of an MWIS in the original input graph is the value in the final array entry (14), corresponding to the independent set consisting of the first, fourth
 and sixth vertices. The first option is to allow s-v paths that include one or more cycles. Not so with greedy algorithms, for which correctness proofs are more art than science—be prepared to throw in the kitchen sink. Let P 0 denote the first i 1 edges of P, and (w, v) its final hop: P' = s-w path (i-1 edges, length L - lwv) w lwv v s P = s-v path (i edges,
length L) The prefix P 0 is an s-w path with at most i 1 edges and length L \leftarrow < L wv , appending the edge (w, v) to P \leftarrow would produce an s-v path with at most i edges and length L \leftarrow + wv < (L wv ) + wv = L, contradicting the optimality of P for length L \leftarrow + would produce an s-v path with at most i edges and length L \leftarrow + wv < (L wv ) + wv = L, contradicting the optimality of P for length L \leftarrow + would produce an s-v path with at most i edges and length L \leftarrow + wv < (L wv ) + wv = L, contradicting the optimality of P for length L \leftarrow + would produce an s-v path with at most i edges and length L \leftarrow + wv < (L wv ) + wv = L, contradicting the optimality of P for length L \leftarrow + would produce an s-v path with at most i edges and length L \leftarrow + wv < (L wv ) + wv = L, contradicting the optimality of P for length L \leftarrow + would produce an s-v path with at most i edges and length L \leftarrow + wv < (L wv ) + wv = L, contradicting the optimality of P for length L \leftarrow + would produce an s-v path with at most i edges and length L \leftarrow + wv < (L wv ) + wv = L, contradicting the optimality of P for length L \leftarrow + would produce an s-v path with at most i edges and length L \leftarrow + wv < (L wv ) + wv = L, contradicting the optimality of P for length L \leftarrow + wv < (L wv ) + wv = L, contradicting the optimality of P for length L \leftarrow + wv < (L wv ) + wv < (L w
the original subproblem.11 This case analysis narrows down the possibilities for an optimal solution to a subproblem to a small number of candidates. At this point, the solution-so-far T looks like: 78 Minimum Spanning Trees 4 1 2 3 5 7 6 The two edges chosen so far are disjoint, so the algorithm is effectively growing two trees in parallel. For
example, if every edge e has unit length 'e = 1, a shortest path minimizes the hop count (i.e., number of edges) between its origin and destination. Or, if the graph represents a road network and the length of each edge the expected travel time from one end to the other, the single-source shortest path problem is the problem of computing driving
times from an origin (the source vertex) to all possible destinations. 106 Introduction to Dynamic Programming 16.1.2 The Natural Greedy Algorithm Fails For many computational problems, greedy algorithms are a great place to start brainstorming. This reduction, which is called Johnson's algorithm and described in the bonus videos at www. a)
O(1) b) O(log n) c) O(n) d) Not enough information to answer (See Section 15.6.5 for the solution and discussion.) With the solution to Quiz 15.7, we conclude that our quick-anddirty implementation of a union-find data structure fulfills the running time guarantees promised by Theorem 15.14 and Table 15.1. 15.6.5 Solutions to Quizzes 15.5-15.7
Solution to Quiz 15.5 Correct answer: (c or d). 56 Minimum Spanning Trees Like minimizing the sum of weighted completion times (Chapter 13) or the optimal prefix-free code problem. 7 Could there be an algorithm that magically homes in on the minimum spanning the sum of weighted completion times (Chapter 13) or the optimal prefix-free code problem. 7 Could there be an algorithm that magically homes in on the minimum spanning the sum of weighted completion times (Chapter 14), the number of possible solutions can be exponential in the size of the problem. 7 Could there be an algorithm that magically homes in on the minimum spanning the sum of weighted completion times (Chapter 14), the number of possible solutions can be exponential in the size of the problem. 7 Could there be an algorithm that magically homes in on the minimum spanning the sum of weighted completion times (Chapter 14), the number of possible solutions can be exponential in the size of the problem. 8 Could there be an algorithm that magically homes in on the minimum spanning the sum of the problem. 9 Could there be an algorithm that magically homes in on the minimum spanning the sum of the problem. 9 Could the problem of the problem of the problem of the problem of the problem. 9 Could the problem of the problem
cost needle in the haystack of spanning trees?, Sm of items PS1, So, while Prim's algorithm was constrained to choose the cheapest remaining edge in the entire graph. Knapsack subproblems can shrink in two different senses (by removing an item or
removing knapsack capacity), and so it goes with sequence alignment subproblems (by removing a symbol from the first or the second input string). Problem: All-Pairs Shortest Paths Input: A directed graph G = (V, E) with n vertices and m edges, and a real-valued length `e for each edge e 2 E. This in-degree could be as large as n 1 in a directed
graph (with no parallel edges), but is generally much smaller, especially in sparse graphs. Quiz 18.6 Let G = (V, E) be an input graph. The leaves of the trees are in one-to-one correspondence with the symbols of ^. When confronted with a new problem, what's the most effective way to put your tools to work? See Part 4 for the full story. P Huffman's
greedy algorithm maintains a forest, with leaves in correspondence to alphabet symbols, and in each iteration greedily merges the pair of trees that causes the minimum-possible increase in the average leaf depth. In our running example: a 4 c 1 b 3 5 2 d the edge (a, d) does not satisfy the MBP (every edge on the path a! b! d is cheaper than (a, d)),
nor does the edge (c, d) (every edge on the path 2 + (4) + 5 = 3. Obtain 0 from \leftarrow by swapping the positions of i and j in the schedule (Figure 13.3). In the sequence alignment problem, how many relevant possibilities are there for the
contents of the final column? The first iteration of the algorithm merges the nodes that correspond to the two symbols with the following sums of symbol frequencies: Symbol tree containing A tree containing B tree containing
C and D Sum of Symbol Frequencies .60 .25 .05 + .10 = .15 The second two trees have the smallest sums of symbol frequencies, so these are the trees merged in the second iteration. In the keys "2" and "4" have search times of 3.10 Different search
and-conquer algorithms (Chapter 3 of Part 1), and graphs (Chapter 7 of Part 2). This strategy would be hopeless unless the left and right subtrees of an optimal binary search tree problem, the input is a set of n keys and
nonnegative frequencies for them, and the goal is to compute a binary search tree containing these keys with the minimum-possible weighted search time. 7. One way to implement a queue is with a doubly-linked list. Assume that n 2; otherwise, the answer is obvious. (If no such path exists, define Lk,v,w as +1.) (For each k 2 {0, 1, 2, . In our solution
to the WIS problem in n-vertex path graphs, we implemented the first step by identifying a collection of n + 1 subproblems. If you need to beef up your programming skills, there are several outstanding free online courses that teach basic programming skills, there are several outstanding free online courses that teach basic programming skills, there are several outstanding free online courses that teach basic programming skills, there are several outstanding free online courses that teach basic programming skills, there are several outstanding free online courses that teach basic programming skills, there are several outstanding free online courses that teach basic programming free online courses that teach basic programming skills, there are several outstanding free online courses that teach basic programming skills, there are several outstanding free online courses that teach basic programming skills, there are several outstanding free online courses that teach basic programming skills are several outstanding free online courses that teach basic programming skills are several outstanding free online courses that teach basic programming skills are several outstanding free online courses that teach basic programming skills are several outstanding free online courses that the first programming skills are several outstanding free outstanding free
the relevant Li 1,w's are +1, then v is unreachable from s in i or fewer hops, and we interpret the recurrence as computing connected components? The inductive proof of correctness is similar to that for the WIS
algorithm (Theorem 16.3).9 For example, for the input graph 8 If there is a tie (A[n 2] + wn = A[n 1]), both options lead to an optimal solution. Output: dist(v, w) for every vertex pair v, w 2 V, or a declaration that G contains a negative cycle. Which of the following greedy algorithms produces a schedule that minimizes the maximum lateness? The
results of past computations are stored in a globally visible length(n + 1) array A, with A[i] storing an MWIS of Gi, where Gi comprises the first i vertices and the first
you these paradigms and their most famous instantiations is one of the major goals of this book series. If you must design a new algorithm from scratch, get calibrated by identifying the line in the sand drawn by the "obvious" solution (such as exhaustive search). Proof: We prove the contrapositive.8 If has no consecutive inversions, the index of each
job is at least 1 larger than the job that came before it. The QuickSort algorithm invokes a 16.4 The Principles of Dynamic Programming 121 algorithm the best of them.11 2. 16.3 A Reconstruction Algorithm The WIS algorithm in Section 16.2.4 computes
only the weight possessed by an MWIS of a path graph, not an MWIS itself. First, after the edge addition, the connected components S1 and S2 fuse into a single component S1 [S2], decreasing the number of components by 1. Here's an idea: The first time we solve a subproblem, why not save the result in a cache once and for all? (To get a v-w path
paste together paths from v to s and from s to w.) 15.2 61 Prim's Algorithm The algorithm Prim computes the minimum spanning tree in the four-vertex five-edge graph of Quiz 15.1, which means approximately. This observation suggests a cool opportunity for algorithm design: Given the search frequencies for a set of keys, what's the best binary
search tree? Recall that in this type of proof, you assume the opposite of what you want to prove, and then build on this assumption with a sequence of logically correct steps that culminates in a patently false statement. Initializing the union-find data structure takes O(n) time. c) If all symbol frequencies are less than 0.33, all symbols will be encoded
with at least two bits. The goal is to either compute the length of a shortest path from every vertex to every other vertex, or detect that the graph has a negative cycle. He was an invited speaker at the 2006 International Congress of Mathematicians, the Shapley Lecturer at the 2008 World Congress of the Game Theory Society, and a Guggenheim
Fellow in 2017. Which one is it? This is equivalent to minimizing the weighted average of the jobs' completion times, with the averaging weights proportional to the wj 's. a) 2.11 b) 2.31 c) 2.49 d) 2.5 Problem 14.3 (H) What is the maximum number of bits that Huffman's greedy algorithm might use to encode a single symbol? i ...
to all of these problems (as indicated by an "(H)" or "(S)," respectively) are included at the end of the book. Compute the median edge cost in the input graph G. What We'll Cover Algorithms Illuminated, Part 3 provides an introduction to and numerous case studies of two fundamental algorithms design paradigms. Can you argue that a solution must
be built up from solutions to smaller subproblems in one of a small number of ways? What title, what name, could I choose? White belts, however, still have a lot of training to do. The algorithm is correct no matter 60 Minimum Spanning Trees which vertex it chooses.10
Each iteration is responsible for adding one new edge to T. In Chapter 14 we defined prefix-free codes and designed a greedy algorithm for computing the best-on-average code for a given set of symbol frequencies. It could be a symbol. This chapter assumes that the input graph is represented using adjacency lists, with an array
of vertices, an array of edges, pointers from each edge to its two endpoints, and pointers from each vertex to its incident edges. 2 15.1.2 Spanning Trees The input in the minimum spanning tree problem is an undirected graph G = (V, E) in adjacency-list
representation, a source vertex s 2 V, and a real-valued length `e for each e 2 E. In other words, once you know that an MWIS includes the last vertex, you know exactly what it looks like: It's an MWIS of the smaller graph Gn 2, supplemented with the final vertex vn. We can reason about all of them in one fell swoop., n} and v, w 2 V, ... Lk 1, v, w
 (Case 1) Lk,v,w = min . The first iteration of the Huffman algorithm merges the leaves labeled "a" and "b" and thereafter treats them as an indivisible unit with total frequency pa + pb . 1 The abbreviation "i.e." stands for id est, and means "that is." 23 24 Huffman Codes 14.1.2 Variable-Length Codes When some symbols of the alphabet occur much
 more frequently than others, variable-length codes can be more efficient than fixedlength ones. Achieved, for example, by the code Symbol A B C D E Encoding 110 1110 0 1111 10 Hint for Problem 14.3: For a lower bound, consider symbol frequencies that are powers of 2. Our semantics for job weights certainly suggests that the higher-weight jobs
should receive the smaller completion times, and this is in fact the case. 16.1.3 A Divide-and-Conquer Approach? QE D Putting it all together proves that the output of Kruskal's algorithm is a spanning tree. a) O(m + n) b) O(m log n) c) O(m2) d) O(mn)
(See below for the solution and discussion.) Correct answer: (d). For example, any genomicist can tell you the typical frequency of each nucleobase (As, Cs, Gs, and Ts) in human DNA. We have our recurrence for solving a subproblem given solutions to smaller subproblems. P In the all-pairs shortest path problem, the input consists of a directed graph
with edge lengths. By how much does a merger increase the average leaf depth? 19 Dijkstra's algorithm can substitute for the Bellman-Ford algorithm if edges' lengths are nonnegative, in which case the running time improves to O(mn log n). Conquer the subproblems recursively. e) P - need not have length less than L. *15.4 Prim's Algorithm: Proof
of Correctness 73 states that every edge addition (v, w) is either type C or type F (and not both), depending on whether the graph already has a v-w path. QE D 15.6.4 Quick-and-Dirty Implementation of Union-Find The Parent Graph Under the hood, a union-find data structure is implemented as an array and can be visualized as a collection of directed
trees. Lemma 18.4 (Bellman-Ford with No Negative Cycles) Under the assumptions and notation of Corollary 18.2, and also assuming that the input graph. By the time a loop iteration must compute the subproblem solution A[i][v], all
values of the form A[i 1][v] or A[i 1][v] or A[i 1][w] have already been 180 Shortest Paths Revisited computed in the previous iteration of the running time of the Find operation as a function of the number n of objects? C := ; // keeps track of current
clusters for each x 2 X do add {x} to C // each point in own clusters in C by 1, so there are a total of |X| k iterations
(Figure 15.9).41 merge m
that we can reuse all of them to also establish the correctness of another important MST algorithm, Kruskal's algorithm, Kruskal's algorithm (Theorem 15.11 in Section 15.5).23 15.4.2 Fun Facts About Spanning Trees. Prim's algorithm, by the proof of Theorem 15.6, we'll prove some simple and useful facts about undirected graphs and their spanning trees. Prim's algorithm, by the proof of Theorem 15.11 in Section 15.5).23 15.4.2 Fun Facts About Spanning Trees.
always choosing the eligible edge with minimum individual cost, is effectively striving to minimize the maximum edge cost along every path.20 19 A popular if more abstract approach to proving the correctness of Prim's (and Kruskal's) algorithm is to use what's known as the "Cut Property" of MSTs; see Problem 15.7 for details., k} as internal
  vertices 22 : and (iv) does not contain a directed cycle. As usual, shortest paths can be reconstructed by tracing back through the final array A computed by the Bellman-Ford algorithm (because it accommodates negative edge lengths)
Under 76 Minimum Spanning Trees our assumption that edges' costs are distinct, the cost of e2 must be strictly larger: cxy > cvw . Our vision is of a dynamic programming algorithm that tries all possibilities for the cost of e2 must be strictly larger: cxy > cvw . Our vision is of a dynamic programming algorithm that tries all possibilities for the root, recursively computing or looking up the optimal left and right subtrees for each possibility. We can therefore sneakily rearrange
terms to obtain L(T) L(T - ) = (px pa) · (depth of x in T depth of a in T) | {z} | {z} 0 0 + (py pb) · (depth of y in T depth of b in T) {z} | {z} 0 0 + (py pb) · (depth of y in T depth of b in T) {z} | {z} 0 0 + (py pb) · (depth of y in T depth of b in T) {z} | {z} 0 0 + (py pb) · (depth of y in T depth of b in T) {z} | {z} 0 0 + (py pb) · (depth of y in T depth of b in T) {z} | {z} 0 0 + (py pb) · (depth of y in T depth of b in T) {z} | {z} 0 0 + (py pb) · (depth of y in T depth of b in T) {z} | {z} 0 0 + (py pb) · (depth of y in T depth of b in T) {z} | {z} 0 0 + (py pb) · (depth of y in T depth of b in T) {z} | {z} 0 0 + (py pb) · (depth of y in T depth of b in T) {z} | {z} 0 0 + (py pb) · (depth of y in T depth of b in T) {z} | {z} 0 0 + (py pb) · (depth of y in T depth of b in T) {z} | {z} 0 0 + (py pb) · (depth of y in T depth of b in T) {z} | {z} 0 0 + (py pb) · (depth of y in T depth of b in T) {z} | {z} 0 0 + (py pb) · (depth of y in T depth of b in T) {z} | {z} 0 0 + (py pb) · (depth of y in T depth of b in T) {z} | {z} 0 0 + (py pb) · (depth of y in T depth of b in T) {z} | {z} 0 0 + (py pb) · (depth of y in T depth of b in T) {z} | {z} 0 0 + (py pb) · (depth of y in T depth of b in T) {z} | {z} 0 0 + (py pb) · (depth of y in T depth of b in T) {z} | {z} 0 0 + (py pb) · (depth of y in T depth of b in T) {z} | {z} 0 0 + (py pb) · (depth of y in T depth of b in T) {z} | {z} 0 0 + (py pb) · (depth of y in T depth of b in T) {z} | {z} 0 0 + (py pb) · (depth of y in T depth of b in T) {z} | {z} 0 0 + (py pb) · (depth of y in T depth of b in T) {z} | {z} 0 0 + (py pb) · (depth of y in T depth of b in T) {z} | {z} 0 0 + (py pb) · (depth of y in T depth of b in T) {z} | {z} 0 0 + (py pb) · (depth of y in T depth of b in T) {z} | {z} 0 0 + (py pb) · (depth of y in T depth of b in T) {z} | {z} 0 0 + (py pb) · (depth of y in T depth of b in T) {z} | {z} 0 0 + (py pb) · (depth of y in T depth of y in T depth of b in T) {z} | {z} 0 0 + (py pb) · (depth of y in T depth of y in T depth of y in T depth of y in T de
in linear time using breadth-first search; see Section 8.2 of Part 2. Of all the search trees for a given set of objects, which is the "best?" We asked this question in Chapter 11 of Part 2. Of all the search trees for a given set of objects, which is the "best?" We asked this question in Chapter 11 of Part 2. Of all the search trees for a given set of objects, which is the "best?" We asked this question in Chapter 11 of Part 2. Of all the search trees for a given set of objects, which is the "best?" We asked this question in Chapter 11 of Part 2. Of all the search trees for a given set of objects, which is the "best?" We asked this question in Chapter 11 of Part 2. Of all the search trees for a given set of objects, which is the "best?" We asked this question in Chapter 11 of Part 2. Of all the search trees for a given set of objects, which is the "best?" We asked this question in Chapter 11 of Part 2. Of all the search trees for a given set of objects, which is the "best?" We asked this question in Chapter 11 of Part 2. Of all the search trees for a given set of objects, which is the "best." We asked this question in Chapter 11 of Part 2. Of all the search trees for a given set of objects, which is the "best." We asked this question in Chapter 11 of Part 2. Of all the search trees for a given set of objects, which is the "best." We asked this question in Chapter 11 of Part 2. Of all the search trees for a given set of objects, which is the "best." We asked this question in Chapter 11 of Part 2. Of all the search trees for a given set of objects, which is the "best." We asked this question in Chapter 11 of Part 2. Of all the search trees for a given set of objects, which is the "best." We asked this question in Chapter 11 of Part 2. Of all the search trees for a given set of objects, which is the "best." We asked the part 2. Of all the search trees for a given set of objects, which is the "best." We asked the part 2. Of all the search trees for a given set of objects are 2. Of all the search trees for a given 
of Gn 1 and Gn 2, respectively. 16.2.4 An Iterative Bottom-Up Implementation As part of figuring out how to incorporate caching into our recursive WIS algorithm, we realized that there are exactly n + 1 relevant subproblems, corresponding to all possible prefixes of the input graph (Quiz 16.4). Which of the following statements are true? 99
Problems P The second step is to use an exchange argument to prove that a spanning tree in which every edge satisfies the MBP must be an MST. 7 Can we do better? A big reason for the success of online courses is the opportunities they provide for participants to help each other understand the course material and debug programs through
discussion forums. The simplest possible example for ruling out one of them would be a problem instance with two jobs, having different weights and lengths, such that the two algorithms schedule the jobs in opposite orders. 122 Introduction to Dynamic Programming recursing on subproblems that are barely smaller than the input (like in the WIS
algorithm), if necessary for correctness. You should always be on the lookout for reductions. In other words, a job's completion time in a schedule computed by the greedy algorithm. Correct answer: (a). If an edge e 2 E is the cheapest
edge crossing a cut (A, B), e belongs to every MST of G.42 In other words, one way to justify an algorithm's inclusion of the NW
algorithm that keeps track of whether an inserted gap is the first in a sequence of gaps (in which case it carries a penalty of a + b) or not (in which case the additional penalty is a). Graham and Pavol Hell (Annals of the History of Computing, 1985). 90 Minimum Spanning Trees promoted demoted 4 6 1 2 + 3 5 4 = 6 1 2 3 5 Å Union operation
performs two Find operations and O(1) additional work, so its running time matches that of Find. For the WIS problem on G1 S2 := second half of G S1 := recursively solve the WIS problem on G1 S2 :=
recursively solve the WIS problem on G2 combine S1, S2 into a solution S for G return S The devil is in the details of the combine step. 9. For the rest of the proof, we denote by a and b the two symbols of ^ with the smallest and second-smallest frequencies, respectively. The gist is that a queue is a data structure for maintaining a list of objects, and
you can remove stuff from its front or add stuff to its back in constant time. These sets can merge me
solution-so-far T: Checking whether T already contains a v-w path then boils down to checking whether v and w belong to the same set of the partition (i.e., to the same connected component). Thomson, x lemma, 10 length (of an edge), 168 longest common subsequence, 164 longest common substring, 164 machine learning supervised learning, 94
unsupervised learning, see clustering mathematical background, x MBP, see minimum spanning tree, minimum bottleneck property, see minimum bottleneck property, see minimum spanning tree, minimum spanning tre
101, 205 Cycle Property, 102, 204 exchange argument, 75 history, 57 in directed graphs, 53 in disconnected graphs, 53 in disconnected graphs, 55 in linear time?, 206 Kruskal's algorithm, see Prim's algorithm reductions to, 100 uniqueness, 101 with distinct edge costs, 69, 75 with non-
distinct edge costs, 72, 91, 101, 205 with parallel edges, 55 minimum-bottleneck spanning tree, 102, 206 MP3, 23, 27 MST, see minimum spanning tree MWIS, see weighted independent set Needleman, Saul B., 139 Needleman-Wunsch (NW) score, see seguence alignment Nielsen, Morten N., 2 N P -hard problem, 170, 202 optimal binary search
trees, 148- 162 correctness, 160 dynamic programming algorithm (OptBST), 159 Knuth's optimization, 161 recurrence, 157- 156 problem definition, 152 reconstruction, 161 recurrence, 157- 156 problem definition, 152 reconstruction, 161 recurrence, 157- 156 problem definition, 158 vs. In each iteration of the algorithm, the trees T1 and T2 with the smallest sums of symbol frequencies can
be identified and removed using a constant number of operations at the fronts of Q1 and Q2. In the knapsack problem, given n items with values and a knapsack capacity C (all positive integers), the goal is to select the maximum-value subset of
items with total size at most C., yj, the first j symbols of Y. This problem is more challenging than the optimal prefix-free code problem, but it is no match for the power of the dynamic programming paradigm., k 1} and a minimum-length cycle-free k-w path with all internal vertices in {1, 2, . In sparse graphs (with m = O(n) or close to it), this
approaches the best we could hope for (as merely writing down the output already requires quadratic time). Algorithmic thinking is increasingly useful and prevalent in disciplines outside of computer science, including biology, statistics, and economics. Quiz 15.1 What is the minimum sum of edge costs of a spanning tree of the following graph?
Solution to Quiz 16.2 Correct answer: (a). Obtain the (n 1)-vertex path graph Gn 1 from G by plucking off the last vertex vn and the last edge (vn 1, vn). First Edition Cover image: Untitled, by Johanna Dickson ISBN: 978-0-9992829-5-3 (ebook) Library of Congress Control Number: 2017914282 Soundlikeyourself
Publishing, LLC New York, NY [email protected] www.algorithms. Let T1 and T2 denote the left and right subtrees of the root. (c) Repeat (b) for Kruskal's
algorithm. independent of T Proof: Leaves of T not labeled a or b occupy the same position in T 0. After reading these books, you'll know exactly what they mean., n} and frequencies p1, p2,. Both paradigms recursively solve smaller subproblems and combine the results into a solution to the original problem. And it doesn't really change the
problem: Connecting every pair of vertices is the same thing as connecting some vertex s to every other vertex. 201 202 Epilogue: A Field Guide to Algorithm Design Most likely, all will fail. Assuming we perform at least a constant amount of work solving each subproblem, the number of subproblems is a lower bound on the running time of our
algorithm. For (d), see Problem 14.2. Solution to Problem 14.2. Solution to Problem 14.2. Solution to Problem 14.2. Solution to Problem 14.2. This implementation eschews heaps in favor of an even simpler data structure: a queue (actually, two queues). To the extent that there are recurring themes in
correctness proofs of greedy algorithms, we will emphasize them as we go along. What are the completion times of the three jobs in this schedule? Lines 2-7 initialize these values are ready and waiting to be looked up in constant time
Output: a partition of X into k non-empty sets. Inspired by our proof of Theorem 13.1, the plan is to exchange one edge for another to produce a spanning tree T 0 with total cost even less than T \vdash . (Choose the strongest statement that is guaranteed to be true.) a) O(m + n) b) O(kn) c) O(km) d)
O(mn) Problem 18.4 (S) For the input graph 2 1 3 5 2 1 4 3 4 what are the final array entries of the Floyd-Warshall algorithm? Let Gi denote the subgraph of G comprising its first i vertices and i 1 edges. When i is
0 or 1, there are no s-t paths with i edges or less, and no solutions to the corresponding subproblems. The algorithm halts once one tree remains, which is after n 1 mergers. Many such implementations are freely available on the Web as well. I hope they exude a contagious enthusiasm for algorithms that, alas, is impossible to replicate fully on the
printed page. Only one type of event can increment x's depth: a Union operation in which the root of x's tree in the parent graph gets demoted. a) (i) Not enough information to answer; (ii) goes down. By Corollary 14.5, the ^-tree (T 0) is optimal for the original problem with input ^ and p. One simple example is: Length Weight Job #1 `1
= 5 w1 = 3 lob #2 '2 = 2 w2 = 1. We call this a type-C edge addition ('C' for "cycle"). The running time bound of O(nC) is impressive only if C is small, for example, if C = O(n) or ideally even smaller. (Let n = | ^ | denote the number of 14.3 35 Huffman's Greedy Algorithm symbols. 7) a) n 1 b) n c) (n+1)n 2 d) Not enough information to answer (See
Section 14.3.7 for the solution and discussion.) 14.3.2 Huffman's Greedy Criterion For a given set of symbol frequencies {pa}a2^, which pair of trees should we merge in each iteration? ******* Kruskal's algorithm begins with the empty edge set and each vertex isolated in its own connected component, just as single-link clustering begins
with each data point in its own cluster. (To be covered in Part 4.) Iterate over the algorithms (especially with dynamic programming). d) The NW algorithms (especially with dynamic programming).
correct after reversing the order of the for loops, but the Knapsack algorithm does not. The work performed in each iteration of the main loop in the Huffman algorithm calls out for a heap! Using a heap to speed up these minimum computations, so a light bulb should go off in your head: This algorithm calls out for a heap! Using a heap to speed up these minimum computations, so a light bulb should go off in your head: This algorithm calls out for a heap! Using a heap to speed up these minimum computations.
decreases the running time from O(n2) to O(n \log n), which qualifies as a blazingly fast implementation. We can do even better. P A dynamic programming algorithm that solves at most f(n) different subproblems, using at most f(n) time for each, and performs at most f(n) time for each f(n) time f
time, where n denotes the input size. On which goods and services should you spend your paycheck to get the most value? I strongly encourage you to implement as many of the algorithms in this book as you have time for. We also use mathematical analysis as needed to understand how and why algorithms really work. Readers of these books have
the same opportunity, via the forums available at www.algorithmsilluminated.org. Under our assumption that the input graph G is connected, there's no way for the algorithm to get stuck; if there were ever an iteration with no edges of G crossing between X and V X, we could conclude that G is not connected (because it contains no path from any
vertex in X to any vertex in V X). Dijkstra's shortest-path algorithm, 60, 69, 70 vs. In this case, the path P can immediately be interpreted as a solution to a smaller subproblem with prefix length k 1, still with origin v and destination w. If at least one of k or m is different from both i and j, and hence appears either before both i and j or after both i and j
in both schedules, the swap has no effect on the relative order of k and m (see Figure 13.4). Do you see how to tweak the algorithm to obtain a per-iteration time bound of O(m) without this assumption? Expanding the definition (14.1) and canceling the terms that correspond to leaves other than a, b, x, y, we have X L(T) L(T) = pz. (depth of z in T
depth of z in T - ). QE D Because the correspondence between ^0-tree in Tab preserves the average leaf depth (up to the tree-independent constant pa + pb), it associates the optimal ^-tree in Tab because the correspondence between ^-tree in Tab because the correspondence between ^-tree in Tab because the correspondence between ^-tree in Tab because the optimal ^-tree in Tab because the optimal ^-tree in Tab because the correspondence between ^-tree in Tab because the optimal ^-
speedup will be easier to see after we reformulate our top-down recursive algorithm as a bottom-up iterative one—and the latter is usually what you want to implement in practice, anyway. So: 1. For example, the problem of computing the 101 Problems median element of an array reduces to the problem of sorting the array. Quiz 18.3 What's the
running time of the Bellman-Ford algorithm, as a function of m (the number of edges) and n (the number of vertices)? Videos., n} of items with the maximum-possible sum i2S vi of values, subject to having P total size i2S si at most C. Proof: If the input graph does not have a negative cycle, then: (i) Floyd-Warshall correctly computes all shortest-path
distances; and 27 Unlike most of our graph algorithms, the Floyd-Warshall algorithm is equally fast and easy-to-implement for graphs represented with an adjacency lists. The objects and connections could represent something
physical, like computer servers and communication links between them. In my experience, most people initially find dynamic programming difficult and counterintuitive. If the goal is to cluster a diverse collection of images according to their subject, a larger value of k should be used. 46 Huffman Codes Proposition 14.3 (Preservation of Behavior of
Huffman) The output of the Huffman algorithm with input ^ and p is (T 0), where T 0 is the output of the Huffman algorithm with input ^ and po . 2 For example, there are n! different prefix-free codes that encode one symbol using one bit ("10"), another using two bits ("10"), another using three bits ("110"), and so on. Whenever you have a scarce
resource that you want to use in the smartest way possible, you're talking about a knapsack problem. The inner "min" implements the exhaustive 18.2 The Bellman-Ford Algorithm 177 search inside Case 2 over all possible choices for the final hop of a shortest path. 

is an 13.4 19 Proof of Correctness a) 2 has one fewer inversion than 1.14.3
Huffman's Greedy Algorithm 14.3.1 Building Trees Through Successive Mergers Huffman's big idea back in 1951 was to tackle the optimal prefix-free code problem using a bottom-up approach. 6 "Bottom-up" means starting with n nodes (where n is the size of the alphabet ^), each labeled with a different symbol of ^, and building up a tree through
successive mergers. Then, S is either: (i) an MWIS of Gn 1; or (ii) an MWIS of Gn 2, supplemented with G's final vertex vn. Both algorithms admit blazingly fast implementations, using the heap and union-find data structures, respectively. Another good idea is union-by-rank, which demotes the root of the tree with the smaller height (breaking ties
arbitrarily). Let P denote the total penalty incurred by this alignment., n and v, w 2 V.) There are (n + 1) \cdot n \cdot n = O(n3) subproblems, which is a linear number for each of the n2 values in the output. 39 With these semantics, it's arguably more accurate to call f a dissimilarity function. Fix from now on an input, with alphabet \hat{f} and symbol
frequencies p, and let a and b denote the symbols with the smallest are denoted.) a) For every vertex v reachable from the source s, there is a shortest s-v path with at most n 1
edges. Quiz 16.1 How many different independent sets does a complete graph with 5 vertices (in addition to an origin and a destination) and consider only cycle-free paths with all internal vertices in {1, 2, . Its second advantage is that it is
```

```
more "distributed" than Dijkstra's algorithm, and for this reason has played the more prominent role in the evolution of Internet routing protocols.14 Evaluating the recurrence (18.1) at a vertex v requires information only about vertices directly connected to v: the vertices with an edge (w, v). a) 0 b) 4gap (length of X) c) +1 d) undefined (See
Section 17.1.8 for the solution and discussion.) 17.1.4 Recurrence Quiz 17.3 handles the base case of an empty input string. To fill in an entry of the entry one column to the left and si rows down (case 2). See Problem 14.5 for more details. The
one difference is that single-link clustering stops once there are k clusters, while Kruskal's algorithm continues until only one connected component remains. Show how to quickly and correctly infer the final solution from the solutions to all of the subproblems. An optimal solution to the WIS problem is called a maximum-weight independent set
(MWIS). 16.5 The Knapsack Problem, not to achieve the minimum bottleneck property. In the three-vertex example
above, adding 5 to every edge length would change the shortest path from s!v!t to s!t. In this case, the path P can immediately be interpreted as a solution to the smaller subproblem with edge budget i 1 (still with destination v). Properly implemented, it is competitive with Prim's algorithm in both theory and practice. Corollary 16.2 then implies
that A[i] is computed correctly, as well. Alternatively, can you reduce the general version of the problem to a special case? Whenever you encounter a seemingly new problem, always ask: Is the problem a disguised version of one you already know how to solve? 16.4.4 Dynamic Programming vs. . In the proof of Theorem 14.2, we take P (k) as the
statement: "the Huffman algorithm correctly solves the optimal prefixfree code problem whenever the alphabet size is at most k." Analogous to a recursive algorithm, a proof by induction has two parts: a base case and an inductive step., xi, the first i symbols of X, and Yj = y1, y2, . Form a complete binary tree of depth `. In each iteration, it greedily
to top. For example, if ^ = {A, B, C, D}, we start with what will be the leaves of the tree: 6 This was for David A. Let T be an MST and P a shortest path from some vertex s to some other vertex t. But there are several general design paradigms that can help you solve problems from many different application domains. As we'll see, such though
experiments can light up a trail that leads directly to an efficient algorithm. (18.1) min(w,v)2E {Li 1,w + `wv } (Case 2) The outer "min" in the recurrence implements the exhaustive search over Case 1 and Case 2. a) AB b) CD c) AAD d) Not enough information to answer (See Section 14.1.6 for the solution and discussion.) The point of Quiz 14.1 is
that, with variable-length codes and no further precautions, it can be unclear where one symbol starts and 14.1 25 Codes the next one begins. There is one Union operation for each edge added to the output which, as an acyclic graph, has at most n 1 edges (Corollary 15.8). 18.3 185 The All-Pairs Shortest Path Problem We can do better. As in the
original implementation, *15.6 Speeding Up Kruskal's Algorithm via Union-Find 85 the sorting step requires O(m log n) time (see Quiz 15.4). You'll get lots of practice describing and reasoning about algorithm, 1 Tardos, Éva, 137 test cases, xi Tetris, 166
theorem, 10 transitive closure (of a binary relation), 186 tree binary search, 148 forest, 34 internal node, 28 leaf, 28 optimal binary search, see optimal binary search, see optimal binary search, 186 tree binary search, 187 path
compression, 83 quick-and-dirty implementation, 85-90 raison d'être, 82 scorecard, 83 speeds up Kruskal's algorithm, 83-85 state-of-the-art implementations, 82 in-degree, 176 out-degree, 176 videos, x Warshall, Stephen, 187 weighted
independent set in general graphs, 134 problem definition, 105 weighted independent set (in path graphs), 105 correctness, 115 dynamic programming algorithms, 106, 108 optimal substructure, 108-110 reconstruction, 116-118 recurrence, 110 recursive algorithm
111 running time, 114 subproblems, 113 why bother?, viii WIS, see weighted independent set Wunsch, Christian D., 139 YouTube, x For example, in the penultimate merge above, the depth of the nodes labeled "C" and "D" increases from 1 to 2, and the depth of the node labeled "C" and "D" increases from 0 to 1. Challenge Problems Problems 15.4 (S) Prove
the converse of Theorem 15.6: If T is an MST of a graph with real-valued edge costs, every edge of T satisfies the minimum bottleneck property. For example, for the input graph 3 2 1 4 6 5 you should check that the final array values are: prefix length i 0 1 2 3 4 5 6 0 3 3 4 9 9 14 At the conclusion of the WIS algorithm, each array entry A[i] stores the
total weight of an MWIS of the graph Gi that comprises 16.2 A Linear-Time Algorithm for WIS in Paths 115 the first i vertices and i 1 edges of the input graph. Suppose we only want to compute the value of an optimal solution and don't care about reconstruction. 1 edges, How many candidates, exactly? (How do you do this in linear time? 15 "RIP"
stands for "Routing Information Protocol." If you're looking to nerd out, the nitty-gritty details of the RIP and RIP2 protocols are described in RFCs 1058 and 2453, respectively. At the end of each chapter you'll find several relatively straightforward questions for testing your under- xi Preface standing, followed by harder and more open-ended
challenge problems. 1., n. P with the maximum-possible total value v + i i2S1 P P i2S2 vi + · · · + v, subject to the knapsack capacities: i2S1 si C1, Pi2Sm i P i2S2 vi + · · · + v, subject to the knapsack capacities: i2S1 si C1, Pi2Sm i P i2S2 vi + · · · + v, subject to the knapsack capacities: i2S1 vi + · · · + v, subject to the knapsack capacities: i2S1 vi + · · · + v, subject to the knapsack capacities: i2S1 vi + · · · + v, subject to the knapsack capacities: i2S1 vi + · · · · + v, subject to the knapsack capacities: i2S1 vi + · · · · + v, subject to the knapsack capacities: i2S1 vi + · · · · + v, subject to the knapsack capacities: i2S1 vi + · · · · + v, subject to the knapsack capacities: i2S1 vi + · · · · + v, subject to the knapsack capacities: i2S1 vi + · · · · + v, subject to the knapsack capacities: i2S1 vi + · · · · · · · v vi + v v
algorithms. 15.8 97 Application: Single-Link Clustering Single-link clustering refers to greedy bottom-up clustering with the best-case similarity function (15.2). Prior to joining Columbia, he spent 15 years on the computer science faculty at Stanford, following a PhD at Cornell and a postdoc at UC Berkeley. The new graph G0 has a negative cycle
reachable from s if and only if G has a negative cycle. children of a common parent, respectively The Huffman algorithm outputs a tree of Tab, and we want to prove that it is the best such tree: (*) among all trees in Tab , the Huffman algorithm outputs one with the minimum-possible average leaf depth. 126 Introduction to Dynamic Programming
Quiz 16.6 Which of the following statements hold for the set S (Choose all that apply.) {n}? The possible final hops are the incoming edges at v. Each iteration of the algorithm evaluates the recurrence (18.1) at each vertex, using the values computed in the previous iteration., n} with nonnegative frequencies p1, p2,. The point of this case analysis
is to narrow down the possibilities for an optimal solution to three and only three candidates. 15.5 77 Kruskal's algorithm for computing them. This incurs a penalty of 4gap per gap, for a total penalty of 4gap times the
length of X. Each iteration of the algorithm chooses two of the trees in the current forest and merges them by making their roots the left and right children of a new unlabeled internal node. The responsibility of an algorithm is to compute, for every possible destination v, the minimum length dist(s, v) of a directed path in G from s to v. a) Neither T1
nor T2 need be optimal for the keys it contains. Kruskal's algorithm, 77 Prim, Robert C., 57 216 principle of parsimony, 202 programming, x, 37 pr
quizzes, x Rackoff, Charles, 2 RadixSort, 41 randomized algorithm, 202 recurrence, 110 recursion tree, 115 reduction, 100, 201 scheduling, 4 GreedyBatio, 13-19 exchange argument, 15 greedy algorithms, 6-10 running time, 11 sum of weighted completion times, 5 with ties, 18
Schrijver, Alexander, 172 search tree, see binary search tree, see binary search tree sequence alignment, 138 reconstruction, 146 recurrence, 143 running algorithm (NW), score, 139 optimal substructure, 140 penalties, 138 problem definition, 138 reconstruction, 146 recurrence, 143 running algorithm (NW), score, 139 optimal substructure, 140 penalties, 139 optimal substructure, 140 penalties, 138 problem definition, 138 reconstruction, 146 recurrence, 143 running algorithm (NW), score, 139 optimal substructure, 140 penalties, 138 problem definition, 138 reconstruction, 146 recurrence, 143 running algorithm (NW), score, 139 optimal substructure, 140 penalties, 139 optimal substructure, 140 penalties, 138 problem definition, 138 reconstruction, 146 recurrence, 140 penalties, 138 problem definition, 138 reconstruction, 146 recurrence, 140 penalties, 138 problem definition, 138 problem 
time, 145 subproblems, 144 variations, 164 Shimbel, Alfonso, 172 shortest paths all-pairs (dense graphs), 185 all-pairs (sparse graphs), 186 Bellman-Ford algorithm distance, 168 Floyd-Warshall algorithm, see Floyd-Warshall algorithm history, 172
Johnson's algorithm, 196 problem definition (all-pairs), 185 problem definition (single-source, 168 with negative cycles, 170 with negative cycles, 170 with parallel edges, 168 single-source shortest path problem, see shortest paths, single
source solutions, x, 203-210 sorting, 2 in linear time, 41 217 Index lower bound for generalpurpose algorithms, 41 spanning tree (of a graph), 53 component fusion, 72 cycle creation, 72 minimum, see minimum spanning tree (of a graph), 53 component fusion, 72 cycle creation, 72 minimum, see minimum spanning tree (of a graph), 53 component fusion, 72 cycle creation, 72 cycle creation, 72 minimum, see minimum spanning tree (of a graph), 53 component fusion, 72 cycle creation, 73 cycle creation, 74 cycle creation, 75 cycle cycle
programming skills. 86 Minimum Spanning Trees 2 4 6 1 5 3 In general, the sets in the parent graph, with each set inheriting the name of its root object. For a more detailed look into the book's contents, check out the "Upshot" sections that conclude each chapter and
 highlight the most important points. The number of independent sets of a path graph grows exponentially with the number of vertices (do you see why?), so there is no hope of solving the problem via exhaustive search, except in the tiniest of instances. (If no such set exists, the algorithm should correctly detect that fact.) 136 Introduction to Dynamic
Programming b) Given a positive integer capacity C and an item budget k 2 {1, 2, . S := ; // items in an optimal solution c := C // remaining capacity for i = n downto 1 do if si c and A[i 1][c] then S := S [ {i} // Case 2 wins, include i c := c si // reserve space for it // else skip i, capacity stays the same return S The Knapsack
Reconstruction postprocessing step runs in O(n) time (with O(1) work per iteration of the main loop), which is much faster than the O(nC) time used to fill in the array in the Knapsack algorithm. 24 For instance, tracing back through the array in the Knapsack algorithm. 24 For instance, tracing back through the array in the Knapsack algorithm. 24 For instance, tracing back through the array in the Knapsack algorithm. 24 For instance, tracing back through the array in the Knapsack algorithm. 24 For instance, tracing back through the array in the Knapsack algorithm. 25 For instance, tracing back through the array in the Knapsack algorithm. 26 For instance, tracing back through the array in the Knapsack algorithm. 27 For instance, tracing back through the array in the Knapsack algorithm. 28 For instance, tracing back through the array in the Knapsack algorithm. 29 For instance, tracing back through the array in the Knapsack algorithm. 29 For instance, tracing back through the array in the Knapsack algorithm. 29 For instance, tracing back through the array in the Knapsack algorithm. 29 For instance, tracing back through the array in the Knapsack algorithm. 20 For instance, tracing back through the array in the Knapsack algorithm. 20 For instance, tracing back through the array in the Knapsack algorithm. 20 For instance, tracing back through the array in the Array i
Section 15.4 that an edge e = (v, w) in a graph G satisfies the MBP if and only if every v-w path in G has an edge with cost at least ce . 33 If you go on to a deeper study of algorithms, beyond the scope of this book series, you can learn several more problems that show up in disguise all the time., n, Pi,j = min{Pi | + 4xi yj , Pi {z } | 1,j 1 Case 1 + 4gap
, Pi, j 1 + 4gap }. The "shortest s-v path" would then be whichever one has the smallest length. Three, there are some very cool connections between Kruskal's algorithm and widely-used clustering algorithms (see Section 15.8). The number of possible codes grows exponentially with n, so even for modest values of n there is no hope of exhaustively
searching through all of them. 2 But surprisingly, the problem can be solved efficiently using a slick greedy algorithm is always correct, proving it can be difficult. Programming Problems P Even when a greedy algorithm is always correct, proving it can be difficult. Programming Problems P Even when a greedy algorithm is always correct, proving it can be difficult. Programming Problems P Even when a greedy algorithm is always correct, proving it can be difficult.
programming language the Prim and Kruskal algorithms. As the more sophisticated paradigm, dynamic programming applies to a wider range of problems than divide-and-conquer, but it is also more technically demanding to apply (at least until you've had sufficient practice). Problem 13.2 (H) Continuing Problem 13.1, P consider instead the
objective of minimizing the total lateness, nj=1 j (). Moreover, by the search tree property, given the root's left subtree, and those greater than r to its right subtree. Heaps: Three Supported Operations Insert: given a heap H and a new object x, add x to H. Quiz
16.3 What is the asymptotic running time of the recursive WIS algorithm, as a function of the number n of vertices? (If there are no such paths, then Lk,v,w = +1.) For every k 2 {1, 2, . Corollary 18.2 (Bellman-Ford Recurrence) With the assumptions and notation of Lemma 18.1, let Li,v denote the minimum length of an s-v path with at most i edges,
with cycles allowed. Or, in recurrence form: Corollary 18.7 (Floyd-Warshall Recurrence) With the assumptions and notation of Lemma 18.6, let Lk,v,w denote the minimum length of a cycle-free v-w path with all internal vertices in {1, 2, . Thus, a light bulb should go off in your head: Prim's algorithm calls out for a heap! *15.3 Speeding Up Prim's
Algorithm via Heaps 15.3.2 63 The Heap Data Structure A heap maintains an evolving set of objects with keys and supports several fast operations, of which we'll need three., k 1}. But will this ever happen?, n}). (See www.algorithmsilluminated.org for test cases and challenge data sets.) Chapter 14 Huffman Codes Everybody loves compression.
|X|\.4 Algorithms have shaped the development of computational genomics as a field. For the WIS problem, perhaps the most natural greedy algorithm is an analog of Kruskal's algorithm: Perform a single pass through the vertices, from best (high-weighted), adding a vertex to the solution-so-far as long as it doesn't conflict
with a previously chosen vertex. Lemma 13.2 (Non-Greedy Schedules Have Inversions) Every schedule as at least one consecutive inversion. Needleman and Christian D. As always, it's our duty to ask the question: Can we do better? A "prefix" of the items then corresponds to the first i items in our arbitrary
ordering (for some i 2 {0, 1, 2, . Here's an example of a prefix-free code for the alphabet ^ = {A, B, C, D} that is not fixed-length: Symbol A B C D Encodings of the other three symbols must start with a "1." Because B is encoded as "10," the encodings of C and D begin with "11." 14.1.4 The
Benefits of Prefix-Free Codes Variable-length prefix-free codes can be more efficient than fixed-length prefix-free codes when the symbols have very different frequencies. Quizzes with solutions and explanations are scattered throughout the text; when you encounter one, I encourage you to pause and think about the answer before reading on. This idea will be
particularly important for some of the harder problems studied in Chapters 17 and 18. The best way to get a feel for greedy algorithms is through examples. For the general problem, cutting-edge research suggests that the answer might be "no." Intrepid readers should check out the paper "Edit Distance Cannot Be Computed in Strongly
 Subquadratic Time (unless SETH is false)," by Arturs Backurs and Piotr Indyk (SIAM Journal on Computing, 2018)., n.) For now, we focus on computing the total weight of an MWIS for a subproblem. 17 In the notation of (16.1), f(n) = O(n), and h(n) = O(n), a
formula is a famous result from combinatorics stating that the n-vertex complete graph (in which all the n2 possible edges are present) has exactly nn 2 different spanning trees. (The easy case is when G0 is connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if G0 is not connected; how do you recurse if
in Chapter 4 of Part 1). Reasoning by analogy, perhaps our solution to the optimal binary search tree problem Our first case study concerns scheduling, in which the goal is to schedule tasks on one or more shared resources to optimize some objective. Because T has no
cycles, every edge addition is of type F and decreases the number of connected components by 1 (Lemma 15.7): The process starts with n connected components (with each vertex in its own component) and ends with 1 (because T is a spanning tree), so the number of edge additions must be n 1. Confusion should not discourage you. (Every subset of
three or more vertices has a pair of adjacent vertices.) It has five size-2 independent sets (as you should verify), for a total of eleven. Define a complete undirected graph G = (X, E) from the data set X and similarity function f, with vertex set X and one edge (x, y) 2 E with cost E can be a complete undirected graph G and similarity function G and G are G and G are G and G are G and G are G are G and G are G and G are 
inversions in 2 compare to that in 1? The number of edges in a path equals the number of bits used to encode the corresponding symbol, so we have the following proposition. Through mathematical analysis, ix Preface you'll gain a deep understanding of the specific algorithms and data structures that these books cover. The freely available book
Mathematics for Computer Science, by Eric Lehman, F. This idea can be made to work, but the slicker and quicker implementation stores vertices in a heap. Why would you want to solve this problem? Quiz 15.2 Which of the following running times best describes a straightforward implementation of Prim's minimum spanning tree algorithm for
graphs in adjacency-list representation? As usual, shortest paths can be reconstructed by tracing back through the floydWarshall algorithm? For all we know, there are other cases in which the algorithm outputs a suboptimal schedule.
This step boils down to showing that the problem of computing the best ^-tree in which a and b fused into a single symbol. I encourage you to explore and find your own favorites. By a ^-tree, we mean a binary tree with leaves labeled in
one-to-one correspondence with the symbols of ^. We'll see two variants in this section. Then, P is either: (i) a minimum-length cycle-free v-w path with all internal vertices in {1, 2, . If every ^-tree with a and b as siblings is suboptimal, it does us no good to optimize over them. Output: A job sequence that minimizes the sum of weighted completion
times (13.1). The solution in (ii) is an option if and only if sn C; in this case, sn units of capacity are effectively reserved in advance for item n.19 The option with the larger total value is an optimal solution, leading to the following recurrence: 19 This is analogous to, in the WIS problem on path graphs, excluding the penultimate vertex of the graph to
reserve space for the final vertex., n and c = 0, 1, 2, 14.3.4 Example For example, let's return to our four-symbol alphabet with the following frequencies: Symbol A B C D Frequency .60 .25 .10 .05 38 Huffman Codes Initially, the Huffman algorithm creates a forest of four trees, TA, TB, TC, TD, each containing one node labeled with a different
alphabet symbol. In general, no. 3 Analogous to Prim's algorithm (Section 15.3), a heap-based implementation of Dijkstra's algorithm runs in O((m + n) log n) time, where m and n denote the number of edges and vertices of the input graph, respectively. Case 3: yn matched with a gap in last column of alignment. (See also footnote 9 of Chapter 16.) If
not, the reconstruction algorithm can recompute the answer from scratch in O(1) time., n, the ith subproblem is to compute the total weight of an MWIS of the graph (where G0 denotes the empty graph). Moreover, the induced
alignment is an optimal alignment of X 0 and Y; the argument is analogous to that in case 1 (as you should verify). Recall from page 170 the problem of computing shortest cycle-free paths in graphs with negative cycles and the fact that it appears to admit no polynomial-time algorithm. Which of these generalizations can be solved by dynamic
programming in time polynomial in the number n of items and the largest number M that appears in the input? For some reason, computer scientists seem to think that trees grow downward, and they draw their trees accordingly. The space requirements of these algorithms are proportional to the number of subproblems: -(n), where n is the number
of vertices; \rightarrow(mn), where m and n are the lengths of the input strings; and \rightarrow(n2), where n is the number of keys, respectively. Meyer, is an excellent and entertaining refresher on mathematical notation (like and 8), the basics of proofs (induction, contradiction, etc.), discrete probability, and much more. Next we might do the same thing with A and
B, committing further to a tree in which the leaves labeled "A" and "B" are siblings: next pair to merge A B C D At this point, only two groups are left to merge. The only major difference between Prim's and Dijkstra's algorithms is the criterion used to choose a crossing edge in each iteration. // Initialization for each a 2 ^ do Ta := tree containing one
node, labeled "a" P (Ta) := pa P F := \{Ta\}a2^* // invariant: 8T 2 F, P (T) = pa a2T // Main loop while F contains at least two trees do T1 := argminT 2F P (T) // second-smallest remove T1 and T2 from F // roots of T1, T2 become left, right children of a new internal node T3 := merger of T1 and T2
P (T3) := P (T1) + P (T2) // maintains invariant add T3 to F return the unique tree in F 14.3 37 Huffman's Greedy Algorithms of Sections 16.2 and 16.3 always P (T3) := P (T1) + P (T2) // maintains invariant add T3 to F return the unique tree in F 14.3 37 Huffman's Greedy Algorithms of Sections 16.2 and 16.3 always P (T3) := P (T1) + P (T2) // maintains invariant add T3 to F return the unique tree in F 14.3 37 Huffman's Greedy Algorithms using a mixture of high-level pseudocode and English (as above).
return a solution that includes a maximumweight vertex. In the second case, we know to include vn in our solution, which forces us to exclude vn 1 . 98 Minimum Spanning Trees The Upshot P A spanning tree of a graph is an acyclic subgraph that contains a path between each pair of vertices. Or in contrapositive form: If subproblem solutions fail to
stabilize by the nth batch, the input graph does have a negative cycle. 22 Introduction to Greedy Algorithms a) Schedule the jobs in increasing order of deadline dj. But what if you have statistics about the frequencies of different searches?11 Quiz 17.4 Consider the following two search trees that store
objects with keys 1, 2, and 3: 1 2 and 2 3 1 3 and the search frequencies: Key 1 2 3 Search Frequency .8 .1 .1 What are the average search times in the two trees, respectively? But solving a subproblem for a destination v boils down to computing the recurrence in Corollary 18.2 which, by Quiz 18.2, involves exhaustive search through 1 + in-deg(v)
candidates, where in-deg(v) is the number of incoming edges at v.16 Because the indegree of a vertex could be as large as n 1, this would seem to give a running time bound of O(n) per-subproblem, for an overall running time bound of O(n) per-subproblem, for an overall running time bound of O(n) per-subproblem.
k. After you learn about algorithms, you'll start seeing them everywhere, whether you're riding an elevator, watching a flock of birds, managing your investment portfolio, or even watching an infant learn. 11 For example, in the WIS algorithm, each recursive call chooses between a subproblem with one fewer vertex and one with two fewer vertices.
The complete graph has no non-adjacent vertices, so every independent set has at most one vertex. (See www.algorithmsilluminated.org for test cases and challenge data sets.) 21 See . Output: The prefix-free binary code with minimumpossible average encoding length: X pa · (number of bits used to encode a). Because all item sizes and the knapsack
capacity C are positive integers, and because capacity is always reduced by the size of some item (to reserve space for it), the only residual capacities that can ever come up are the integers between 0 and C.21 16.5.4 A Dynamic Programming Algorithm Given the subproblems and recurrence, a dynamic programming algorithm for the knapsack
problem practically writes itself. Because the algorithm performs O(1) work per iteration decreases the sum of the lengths of the remaining prefixes, its running time is O(m + n). We can therefore rephrase the optimal prefixes, its running time is O(m + n).
the beginning of the algorithm and, accordingly, a union-find data structure is born with each object in a different set. In most divide-and-conquer algorithms, all the subproblems are distinct and there's no point in caching their solutions. 12 3. This remains true as more edges are included in subsequent iterations. The n symbols can be sorted by
frequency in O(n log n) time (see footnote 3 of Chapter 13), so the running time of this implementation is O(n log n). Challenge Problems Problems 14.5 (S) Give an implementation of Huffman's greedy algorithm that uses a single invocation of a sorting subroutine, followed by a linear amount of additional work. This reduces to two Find operations. The
only independent set of G0 is the empty set, which has total weight 0. For the knapsack problem, we can see from Lemma 16.4 and Corollary 16.5 that subproblems should be parameterized by two indices: the length i of the prefix of available items and the available knapsack capacity c.20 Ranging over all relevant values of the two parameters, we
obtain our subproblems: Knapsack: Subproblems Compute Vi,c, the total value of an optimal knapsack solution with the first i items and knapsack capacity c. (In the end, it won't matter which one we pick.) The plan is to construct a tree one edge at a time, starting from b and growing like a mold until the tree spans the entire vertex set. Objects in
the two participating trees and, hence, the encoding lengths of the corresponding symbols. As a greedy algorithm, the algorithm always chooses the cheapest edge that does the job. Each new edge adds 1 to the overall edge count, and also adds 1 to the in-degree of exactly one vertex (the head of that edge). Think algorithmically. If you remember
only one thing about greedy algorithms, it should be this. ? 48 Huffman Codes the labels of the leaves labeled "a" and "x," and the labels of the leaves labeled "b" and "y": y b x y a a b x How does the average leaf depth change? Problem: Single-Source Shortest Paths (Revised Version) Input: A directed graph G = (V, E), a source vertex s 2 V, and a
real-valued length `e for each edge e 2 E. ("RFC" stands for "request for comments" and is the primary mechanism by which changes to Internet standards are vetted and communicated.) Bonus videos at www.algorithmsilluminated.org describe some of the engineering challenges involved. (Choose all that apply.) 25 This argument explains why the
Floyd-Warshall subproblems, in contrast to the Bellman-Ford subproblems, impose the cycle-free condition (iv). Maintain the following invariants: (i) the elements of Q2 correspond to the multi-node trees in the current forest F, stored in increasing
of this paradigm: the MergeSort and QuickSort algorithm for multiplying two n-digit integers, Strassen's O(n2.71)-time algorithm for multiplying two
a type-C or type-F edge addition. The solution in (i) is always an option for the optimal solution. 14.1 Codes 14.1.1 Fixed-Length Binary Codes Let's set the stage before we proceed to a problem definition or algorithm. Case 2: P has i edges. If size(i) < size(j), set parent(i) := j and size(j) := size(i) + size(j). For suppose S \leftarrow were an independent set of
Gn 2 with total weight W > W wn . Problem 15.7 (S) An alternative approach to proving the correct shortest-path distances in input graphs without negative
cycles.6 Suppose a graph has no negative cycles., k 1} and length L-1 < L1. (It's also a great excuse to pick up a new programming language!) For guidance, see the end-of-chapter Programming tree algorithm (Section
15.5)? The total work performed over all iterations of the inner for loop is proportional to X X (1 + in-deg(v)) = n + in-deg(v) . In the third iteration, the forest F contains only two trees; they are merged to produce the final output, which is exactly the tree used to represent the variable-length prefix-free code in Section 14.2.1: .4 .15 .6 .25 .10 .05 A B
C D .6 A A B .25 B C D A 14.3.5 B .6 C D C D A Larger Example To ensure that Huffman's greedy algorithm is crystal clear, let's see how the final tree takes shape in a larger example Example To ensure that Huffman's greedy algorithm is crystal clear, let's see how the final tree takes shape in a larger example To ensure that Huffman's greedy algorithm is crystal clear, let's see how the final tree takes shape in a larger example.
problem. Every iteration of Huffman's greedy algorithm myopically performs the merge that least increases this objective function. // Initialization X := \{s\} // s is an arbitrarily chosen vertex T := s; // invariant: the edges in T span T
X add edge (v -, w -) to T return T The sets T and X keep track of the edges chosen and the vertices spanned so far. This would contradict the supposed optimality of S. If the obvious solution is inadequate, brainstorm as many natural greedy algorithms as you can and test them on small examples. The "min" in the recurrence (17.3) implements the
exhaustive searchP through the n different candidates for an optimal solution. The same is true for the symbols, as you should check. The key associated with an object is the sum of the frequencies of the symbols that correspond to the true for the other three symbols, as you should check. The key associated with an object is the sum of the frequencies of the symbols that correspond to the true for the other three symbols, as you should check. The key associated with an object is the sum of the frequencies of the symbols that correspond to the true for the other three symbols, as you should check. The key associated with an object is the sum of the frequencies of the symbols that correspond to the true for the other three symbols as you should check.
Problem I'm not going to tell you what dynamic programming is just yet. 3 For example, two O(n log n)-time sorting algorithm might recognize its use here—though only in the analysis, not in the algorithm! 20
Introduction to Greedy Algorithms Solution to Quiz 13.5 ..... To be prepared for both cases, we must keep track of both the first (smallest) and last (largest) keys that belong to a subproblem. 14 We therefore end up with a two-dimensional set of subproblems, despite the seemingly one-dimensional input. Hint for Problem 18.3: Consider stopping and last (largest) keys that belong to a subproblem.
shortest-path algorithm early. b) Given a directed acyclic graph with real-valued edge lengths, compute the length of a longest path between any pair of vertices and m edges, whether or not the input graph
contains a negative cycle. For example, a resource could represent a computer processor (with tasks corresponding to jobs), a classroom (with tasks corresponding to meetings). Here, we want to prove that the Huffman algorithm did not make a mistake by committing to a tree in
which the two smallest-frequency symbols are siblings: (†) There is a tree of Tab that minimizes the average leaf depth L(T, p) over all ^-trees T . 18.4 The Floyd-Warshall Algorithm 187 Can we do better? Remember the principle of parsimony: Choose the simplest data structure that supports all the operations required by your algorithm. 18.2.6
Example For an example of the Bellman-Ford algorithm in action, consider the following input graph: v + A[0][v] = +\infty The vertices are labeled with the solutions to the first batch of subproblems (with i = 0). c) (i) Unaffected; (ii) goes down. Without loss of generality, each
node of T either is a leaf or has two children.12 Thus, there are two leaves with a common parent that inhabit the deepest level of T, say with left child x and right child y.13 Obtain the ^-tree T \to 2 Tab by exchanging 12 An internal node with only one child can be spliced out to give another ^-tree with smaller average leaf depth. *15.4 Prim's
Algorithm: Proof of Correctness Proving the correctness of Prim's algorithm (Theorem 15.1) is a bit easier when all the edge costs are distinct. Caching surely speeds up the algorithm, but by how much? Expanding out the weighted average, we have average # of bits per symbol = |\{z\}| \cdot .6 + |2| \cdot .6 + |2| \cdot .6 + |2| \cdot .6 + |2| \cdot .25| + 3| \cdot .6 + |2| \cdot .25|
do we promote? P It is often easy to propose one or more greedy algorithms for a problem and to analyze their running times. The other *15.4 Prim's Algorithm: Proof of Correctness 69 edges incident to x, (x, y) and (x, z), get partially yanked out of V X and now cross the frontier. If the input graph has no negative cycles, however, subproblem
solutions are guaranteed to stabilize by the time i reaches n, the number of vertices. S {2} is the empty set, but the only optimal solution to the subproblem consisting of the first item and knapsack capacity 2 is {1}. We can't regard S as a feasible solution to the subproblem with only the first item and knapsack capacity 2 is {1}.
Books Are Different This series of books has only one goal: to teach the basics of algorithms in the most accessible way possible. Case 1: xm and yn matched in last column of alignment. When putting it into practice, you might find the sheer number of algorithms, data structures, and design paradigms daunting. Suppose we magically knew an MWIS
5 2 6 5 4 8 7 3 1 6 2 10 9 (a) Three Components 8 7 6 5 4 2 10 9 (b) Component Fusion 4 8 7 10 9 (c) Cycle Creation Figure 15.7: In (a), a graph with vertex set {1, 2, 3, . (A substring is a subsequence comprising consecutive symbols. In tandem, Lemmas 18.3 and 18.4 tell us the last batch of subproblems that we need to bother with:
n. The trees T and T \leftarrow must be different and each has n 1 edges, where n = |V | (by Corollary 15.8). Case r: The root of T has key r. c) It is an optimal solution to the subproblem consisting of the first n 1 items and knapsack capacity C sn. Sensible values for k range from 2 to a large number, depending on the application.
Programming 17.2.3 Problem Definition Quiz 17.4 shows that the best binary search tree for the job depends on the search frequencies are not uniform. In the case of encoding an MP3 file, the encoder computes symbol frequencies explicitly when preparing an area of encoding an MP3 file, the encoder computes symbol frequencies explicitly when preparing an area of encoding an MP3 file, the encoder computes symbol frequencies explicitly when preparing an area of encoding an MP3 file, the encoder computes symbol frequencies explicitly when preparing an area of encoding an MP3 file, the encoder computes symbol frequencies explicitly when preparing an area of encoding an MP3 file, the encoder computes symbol frequencies explicitly when preparing an area of encoding an area of encoding an explicit frequencies.
initial digital version of the file (perhaps following an analog-to-digital conversion), and then uses an optimal prefix-free code to compress the file further. |{z} job #1 job #2 job #3 6 Introduction to Greedy Algorithms By checking all 3! = 6 possible schedules, you can verify that this is the schedule that minimizes the sum of weighted completion
times. P An independent set of an undirected graph is a subset of mutually non-adjacent vertices. *15.7 Kruskal's Algorithm: Proof of Correctness This section proves the correctness of Kruskal's algorithm (Theorem 15.11) under the assumption that edges' costs are distinct. P The optimal binary search tree problem can be solved using dynamic
programming in O(n3) time, where n is the number of keys., n 1} with total size at most C and total value greater than V, it would also constitute such a solution in the original instance. But dynamic programming is relatively formulaic—certainly more so than greedy algorithms—and can be mastered with sufficient practice. What about the variable
length code? P Single-link clustering is a greedy bottom-up clustering method in unsupervised learning and it corresponds to Kruskal's algorithm, stopped early. Allowing variable-length codes introduces a complication, however, which we illustrate by example. 4 For convenience, we typically allow a path (v0, v1), (v1, v2), and in the convergence of the convenience of the convergence of the 
create a cycle (Lemma 15.7(a)), so Kruskal excludes the edge from its output. There are several ways to fuse the two decisions: 1. To give you a starting point, I'll tell you the typical recipea 6 4 1 + 5 6 4 = 4 or 1 2 2 3 2 or 6 1 6 3 5 1 5 5 3 4 2 3 To complete the implementation, we must make two decisions: 1. To give you a starting point, I'll tell you the typical recipea 6 4 1 + 5 6 4 = 4 or 1 2 2 3 2 or 6 1 6 3 5 1 5 5 3 4 2 3 To complete the implementation, we must make two decisions: 1. To give you a starting point, I'll tell you the typical recipea from its output.
I use when I need to understand a new computational problem. (For example, with the input graph in Section 18.2.6, the vertex v's predecessor would be initialized to null, reset to s after the first iteration, and reset again to u after the second iteration.) You can then reconstruct a shortest s-v path backward in O(n) time using the final batch of
predecessors by starting from v and following the predecessor trail back to s. 15.5.3 Straightforward Implement Kruskal's algorithm and, in particular, the cycle-checking required in each iteration? 13.3 9 Developing a Greedy Algorithm There are plenty of other options. P The Floyd-Warshall algorithm is a
dynamic programming algorithm that solves the all-pairs shortest path problem in O(n3) time, where n is the inductive step is easier and based on an exchange argument. For us, the natural base case is the statement P (2). If the input graph has no
negative cycles, cycle-splicing can only shorten a path, in which case the length of Pb is at most Lk corresponding objects to the divideand-conquer paradigm. The defining search tree property is: 9 We refer to nodes of the tree and the corresponding objects.
interchangeably. The first point of confusion is the anachronistic use of the word "programming." In modern times it refers to coding, but back in the phrase "television programming." What about "dynamic?" For the full story, I refer you to the father of dynamic
programming himself, Richard E. Examples include scheduling problems, optimal compression, and minimum spanning trees of graphs. Thus A[k][v][v], which is at most the length of the cycle C, which is less than zero. The optimal substructure lemma (Lemma 18.1) is stated as if there were two candidates for an
optimal solution, but Case 2 comprises several subcases, one for each possible final hop (w, v) of an s-v path. For example, many greedy algorithms boil down to sorting plus a linear amount of extra processing, in which case the running time of a good implementation would be O(n log n), where n is the number of objects to be sorted.3 (Big-O notation
suppresses constant 1 To investigate formal definitions of greedy algorithms, start with the paper "(Incremental) Priority Algorithms," by Allan Borodin, Morten N. 16.4.5 Why "Dynamic Programming" You might be wondering where the weird moniker "dynamic programming" and the paper "(Incremental) Priority Algorithms," by Allan Borodin, Morten N. 16.4.5 Why "Dynamic Programming" and the paper "(Incremental) Priority Algorithms," by Allan Borodin, Morten N. 16.4.5 Why "Dynamic Programming" and the paper "(Incremental) Priority Algorithms," by Allan Borodin, Morten N. 16.4.5 Why "Dynamic Programming" and the paper "(Incremental) Priority Algorithms," by Allan Borodin, Morten N. 16.4.5 Why "Dynamic Programming" and the paper "(Incremental) Priority Algorithms," by Allan Borodin, Morten N. 16.4.5 Why "Dynamic Programming" and the paper "(Incremental) Priority Algorithms," by Allan Borodin, Morten N. 16.4.5 Why "Dynamic Programming" and the paper "(Incremental) Priority Algorithms," by Allan Borodin, Morten N. 16.4.5 Why "Dynamic Programming" and the paper "(Incremental) Priority Algorithms," by Allan Borodin, Morten N. 16.4.5 Why "Dynamic Programming" and the paper "(Incremental) Priority Algorithms," by Allan Borodin, Morten N. 16.4.5 Why "Dynamic Programming" and the paper "(Incremental) Priority Algorithms, and the paper "(Increm
works than it was before. 15 We refer to vertices of the input graph and the corresponding objects in the heap interchangeably. 14.4.1 High-Level Plan The proof of Theorem 14.2 blends two common strategies for correctness proofs of greedy algorithms, both mentioned in Section 13.4: induction and exchange arguments. The search times for the
keys {1, 2, . d) (i) Unaffected; (ii) goes down; (iii) goes down; (iii) goes up. Programming Problems Problem 13.4 Implement in your favorite programming language the GreedyDiff and GreedyRatio algorithms from Section 13.3 for minimizing the weighted sum of completion times. We leave the detailed pseudocode to the interested reader. Problem 15.5 (S) Prove the
correctness of Prim's and Kruskal's algorithms (Theorems 15.1 and 15.11) in full generality, for graphs in which edges' costs need not be distinct. The one detail to get right is the order in which to solve the subproblems. This is bigger than the estimated number of atoms in the known universe when n 50. Hints and Solutions to Selected Problems
Hint for Problem 13.1: One of the greedy algorithms can be proved correct using an exchange argument, similar to the one in Section 13.4. Hint for Problem 13.2: For each of the incorrect algorithms, there is a counterexample with only two jobs. (When i = 0, interpret Wi as 0.) Then Wn = max{Wn 1, Wn 2 + wn }. a) 6 b) 7 c) 8 d) 10 (See Section
16.5.7 for the solution and discussion.) I could tell you a cheesy story about a knapsack-wielding burglar who breaks into a house and wants to make off quickly with the best pile of loot possible, but this would do a disservice to the problem, which is actually quite fundamental. The left subtree of a node x is defined as the nodes reachable from x via
its left child pointer, and similarly for the right subtree. Provide either a proof or a counterexample. That algorithm iteratively computes the algorithm returns the correct shortest-path distances (by Lemma 18.3). How do we know
whether we can trust the solutions to the final batch of subproblems? We can therefore always reorder and reindex the jobs so that (13.2) holds., m and j = 1, 2, . Assume that the symbol frequencies sum to 1. (When i = 0, interpret Vi,c as 0.) For every i = 1, 2, . Assume that the symbol frequencies sum to 1. (When i = 0, interpret Vi,c as 0.) For every i = 1, 2, . Assume that the symbol frequencies sum to 1.
applications that are exceptions to this rule. The Minimum Bottleneck Property (MBP) For a graph G = (V, E) with real-valued edge costs, an edge (v, w) 2 E satisfies the minimum bottleneck property (MBP) if it is a minimum-bottleneck property (MBP) if it is a minimum bottleneck property (MB
greedy algorithm perform before halting? When computing A[i] with i 2, by induction, the values A[i 1] and A[i 2] are indeed the total weights of MWISs of Gi 1 and Gi 2, respectively. P Most greedy algorithms are not always correct. I've made several resources available to help you replicate as much of the online course experience as you like.
Additional Properties of the Correspondence The correspondence between ^0 -trees and the ^-trees in Tab given by the mappings 4 and (Figure 14.2) also preserves the average leaf depth, up to a constant that is independent of the choice of tree. b) Both algorithms remain well defined and correct after reversing the order of the for loops. Which of
the two algorithms, if any, is correct? We also assume that P (2), P (3), . To continue the pattern, it would seem that we should zoom in on the last column of the alignment: A G G C A G G C | \{z\} rest of alignment T A |\{z\} last column of the alignment T A |\{z\} last column in our first two case studies, the final vertex or item was either in or out of the solution—two different possibilities.
optimal prefix-free codes, 152 weighted search time, 152 with unsuccessful searches, 152 path (of a graph), 53 bottleneck, 70 cycle-free, 53 path graph, 105 pay the piper, 67 pep talk, 103 Pigeonhole Principle, 172 prefix-free code, see code, prefixfree Prim's algorithm achieves the minimum bottleneck property, 70 example, 57 greedy criterion, 59
outputs a spanning tree, 75 proof of correctness, 69-76, 102 pseudocode, 59 pseudocode (heap-based), 63 running time (heap-based), 68 running time (heap-based), 69 running time (heap-based), 69 running time (heap-based), 60 running time (heap-bas
can be solved using dynamic programming in O(mn) time, 163 Problems where m and n are the lengths of the input strings. X. a) 0, 0, and +1 d) +1, `vw, and 'vw c) 0, `vw, and +1 d) +1, 
of the subproblems using the recurrence in Corollary 18.7. The final for loop in the pseudocode 18.4 193 The Floyd-Warshall Algorithm checks whether the input graph contains a negative cycle and is explained in Section 18.4.4. See Problems 18.4 and 18.5 for examples of the algorithm in action. What is a greedy algorithm, exactly? What might we
mean by a "shortest s-v path?" Option #1: Allow cycles. It must wait for the same jobs as before ("stuff"), and now job j as well, so its completion time increases by 'j . The algorithm's running time is then at most f (n) |{z} # subproblems \neg g(n) |{z} time per subproblem (given previous solutions) + h(n) |{z} . For example, the longest common
subsequence of "abcdef" and "afebcd" is "abcd.")20 c) Assume that X and Y have the same length n. Equivalently, an independent set does not contain both endpoints of any edge of G. This chapter considers only undirected graphs, in which each edge e is an unordered pair {v, w} 52 15.1 53 Problem Definition of vertices (written as e = (v, w) or e
(w, v)), which are called the endpoints of the edge. 1 The numbers |V| and |E| of vertices and edges are usually denoted by n and m, respectively. As for job i, its completion time goes up in 0. The Upshot P Greedy algorithms construct solutions iteratively, via a sequence of myopic decisions, and hope that everything works out in the end. Both nodes
and edges can be labeled. Proof: Throughout the algorithm, the vertices of X form one connected component of (V, T) and each vertex of V X is isolated in its own connected component. Readers of Part 2 learned a blazingly fast algorithm, for the special case of the single-source shortest path problem in which every edge length
 `e is nonnegative.3 Dijkstra's algorithm, great as it is, is not always correct in graphs with negative edge lengths. If every edge of T satisfies the minimum bottleneck property, T is a minimum spanning tree. In general, the job scheduled first contributes to the completion times of all the jobs, as all jobs must wait for the first one to finish. The latter
problem is equivalent to the all-pairs reachability problem: Given a directed graph, identify all vertex pairs v, w for which the hop-count shortest-path distance is finite). In a divide-and-conquer algorithm, subproblems are chosen primarily to optimize the running time; correctness often takes
care of itself.14 In dynamic programming, subproblems are usually chosen with correctness in mind, come what may with the running time.15 5. Bellman and Lester R. QE D 15.4.3 Proof of Theorem 15.6 (MBP Implies MST) The proof of Theorem 15.6 is where we use our standing assumption that edges' costs are distinct. We conclude that single-link
clustering is the same as Kruskal's algorithm, a point we'll return to in Part 4. A simple hack is to modify the WIS algorithm so that each array entry A[i] records both the total weight of an MWIS of the ith subproblem Gi and the vertices of an
objects in the heap correspond to the trees of F. Both algorithms do the right thing in our two special cases, with equal-length jobs. 190 Shortest Paths Revisited Case 2: Vertex k is an internal vertex of P. A quick way to rule out one of them is to find an instance in which the two algorithms output different schedules, with different
objective function values. a) O(1) b) O(\log n) c) O(n) d) Not enough information to answer (See Section 15.6.5 for the solution and discussion.) Union When the Union operation is invoked with objects x and y, the two trees T1 and T2 of the parent graph containing them must be merged into a single tree., n, such that Xi = Yf (i) for every i = 1, 2,
recursive call throws away either one or more of the smallest keys, or one or more of the largest keys, or one or more of the largest keys, and finally returns the
position 4 (a root, as parent(4) = 4). The final array values are: 22 In the notation of (16.1), f(n) = O(nC), g(n) = O(1), and h(n) = O(1)
most f (n) different subproblems (working systematically from "smallest" to "largest"), using at most g(n) time for each, and performs at most h(n) postprocessing work to extract the final solution (where n denotes the input size). What properties do we want them to satisfy? 112 Introduction to Dynamic Programming c) O(n2) d) none of the above
(See Section 16.2.5 for the solution and discussion.) 16.2.3 Recursion with a Cache Quiz 16.3 shows that our recursive WIS algorithm is no better than exhaustive search. Line 12 iterates through the new contestants.17 Line 13 checks whether an edge (w-, y) is the new winner in y's local tournament; if it is, lines 14-16 update y's key and winner
fields and the heap H accordingly.18 17 This is the main step in which it's so convenient to have the input graph represented via adjacency lists—the edges. You'll also learn general algorithm design paradigms that are relevant to many different problems across different
domains, as well as tools for predicting the performance of such algorithms. For example, consider the graph v 2 2 t s 1 1 u 1 w and, for the destination t, the subproblems corresponding to successive values of the edge budget i. After reviewing graphs and defining the problem formally (Section 15.1), we'll discuss the two best-known MST algorithms.
-Prim's algorithm (Section 15.2) and Kruskal's algorithm (Section 15.5). What about in a graph like the following? When i is 2, there is a unique s-t path with at most i edges (s!v!t), and the value of the subproblem is 4. These are examples of knapsack problems. Another deliverable of this exercise will be a recurrence for quickly computing the
solution to a subproblem from those of two smaller subproblems. Proposition 15.2 (Prim Running Time (Straightforward)) For every graph G = (V, E) and real-valued edge costs, the straightforward implementation of Prim runs in O(mn) time, where m = |E| and n = |V|. For example, vn could be bigger than C, in which case C vn is negative and
meaningless. Quiz 15.7 With the implementation of Union above, what's the running time of the Find (and, hence, Union) operation, as a function of the number n of objects? (For example, given three jobs consuming the intervals [0, 3], [2, 5], and [4, 7], the optimal solution consists of the first and third jobs.) The plan is to design an iterative greedy
algorithm that, in each iteration, irrevocably adds a new job j to the solution-so-far and removes from future consider an arbitrary octain all jobs that conflict with j. (Do you see why?) Let C denote an arbitrary such cycle. To prove (†), consider an arbitrary such cycle. To prove (†), consider an arbitrary octain all jobs that conflict with j. (Do you see why?) Let C denote an arbitrary such cycle. To prove (†), consider an arbitrary such cycle. To prove (†), consider an arbitrary octain all jobs that conflict with j. (Do you see why?) Let C denote an arbitrary such cycle. To prove (†), consider an arbitrary such cycle. To prove (†), consider an arbitrary octain all jobs that conflict with j. (Do you see why?) Let C denote an arbitrary such cycle. To prove (†), consider an arbitrary octain all jobs that conflict with j. (Do you see why?) Let C denote an arbitrary such cycle. To prove (†), consider an arbitrary such cycle. To prove (†), consider an arbitrary octain all jobs that conflict with j. (Do you see why?) Let C denote an arbitrary octain all jobs that conflict with j. (Do you see why?) Let C denote an arbitrary such cycle. To prove (†), consider an arbitrary octain all jobs that conflict with j. (Do you see why?) Let C denote an arbitrary octain all jobs that conflict with j. (Do you see why?) Let C denote an arbitrary octain all jobs that conflict with j. (Do you see why?) Let C denote an arbitrary octain all jobs that conflict with j. (Do you see why?) Let C denote an arbitrary octain all jobs that conflict with j. (Do you see why?) Let C denote an arbitrary octain all jobs that conflict with j. (Do you see why?) Let C denote an arbitrary octain all jobs that conflict with j. (Do you see why?) Let C denote an arbitrary octain all jobs that conflict with j. (Do you see why?) Let C denote an arbitrary octain all jobs that conflict with j. (Do you see why?) Let C denote an arbitrary octain all jobs that conflict with jobs that conflict with j. (Do you see why?) Let C denote all jobs that conflict with 
       pletion time is 6. \{z\} \mid \{z\} \mid 1, Case 2 Case 3 The more general statement in Corollary 17.2 follows by invoking the first statement, for each i = 1, 2, . In the second case, the algorithm will add the edge (v, w) to its solution-so-far (by Lemma 15.7(b), this doesn't create a cycle), directly connecting v and w and fusing their connected components
into one. Consider an arbitrary undirected graph G = (V, E) with nonnegative vertex weights, and an arbitrary vertex v 2 V with weight wv. By Lemma 15.7, 28 Why O(m log n) instead of O(m log n)? Here's the input graph: 4 1 2 3 5 7 6 Kruskal's algorithm, like Prim's algorithm, greedily constructs a spanning tree one edge at a time. If (v, w) 2 E, the
only such path is the one-hop path v! w (with length `vw ). Such a 7 Exchange arguments are only one way among many to prove that a greedy algorithm is correct. 4. 17.2.1 Binary Search tree is a data structure that acts like a dynamic version of a sorted array—searching for an object is as easy as in a sorted array, but
it also accommodates fast insertions and deletions. c) The formula is always correct in trees but not always correct in arbitrary graphs. Proof: The solution to Greedy Algorithms The first half of this book is about the greedy algorithm
design paradigm. Under which object do we install the demoted root? c) P - need not have internal vertices and edges) in a reasonable amount of time, but not big graphs (with millions of vertices and edges). b) A letter with frequency at least
0.5 will never be encoded with two or more bits., n with i2S si C. 18.4 The Floyd-Warshall Algorithm 191 a) The concatenation P \leftarrow of P1 \leftarrow and P2 need not have origin v. Reordering the jobs in the input has no effect on the problem to be solved. After n 1 iterations, the algorithm runs out of new vertices to add to its set X and halts. Zoom in on a
fixed iteration of the outer for loop of the algorithm, with some fixed value of i. The next iteration considers the edge with cost 3. 2 For more details on graphs and their representations, see Chapter 7 of Part 2. What does that buy us? Encodings correspond to root-leaf paths, with left and right child edges interpreted as 0s and 1s, respectively, while
the average encoding length corresponds to the average leaf depth. Then, there is a cyclefree path P1 from v to k with internal vertices in {1, 2, . Because the population cannot exceed the total number n of objects, the depth of x cannot be incremented more than log2 n times. Test Your Understanding Problem 18.1 (S) For the input graph w -3 -1 s
1 u -2 v 4 x what are the final array entries of the Bellman-Ford algorithm from Section 18.2? 199 Problems Problem 18.2 (S) Lemma 18.3 shows that once the subproblem solutions stabilize in the Bellman-Ford algorithm (with Lk+1,v = Lk,v for every destination v), they remain the same forevermore (with Li,v = Lk,v for all i k and v 2 V). For
example, in Chapter 9 of Part 2, our correctness proof for Dijkstra's algorithm used induction rather than an exchange argument. 8 History buffs should check out the paper "On the History of the Minimum Spanning Tree Problem," by Ronald L. Thus, T contains at least one edge e1 = (v, w) that is not in T - Answer (a) is a natural guess but is also
incorrect.25 In (c), we are effectively reserving sn units of capacity for item n's inclusion, which leaves a residual capacity of C sn . This problem does not arise with fixed-length codes., n 1}, with total value V \leftarrow > V vn and total size at most C sn , then S \leftarrow [n] would have total size at most C and total value V \leftarrow > V vn O + vn O = V . But what
makes an alignment "nice?" Is it better to have one gap and one mismatch, or three gaps and no mismatches? 13.3 7 Developing a Greedy Algorithm 6 time job #3 3 job #2 job #1 1 0 Figure 13.1: The completion times of the three jobs are 1, 3, and 6. Prove that every edge that fails to satisfy the minimum bottleneck property (page 70) is excluded
from the final output and use Theorem 15.6. Hint for Problem 13.1 The Greedy Algorithm Design Paradigm 13.2 A Scheduling Problem 13.3 Developing a Greedy Algorithm 13.4 Proof of Correctness Problems 1 1 4
6 12 21 14 Huffman Codes 14.1 Codes 14.2 Codes as Trees 14.3 Huffman's Greedy Algorithm *15.3 Speeding Up Prim's Algorithm via Heaps *15.4 Prim's Algorithm: Proof of Correctness 15.5 Kruskal's Algorithm *15.6
Speeding Up Kruskal's Algorithm via Union-Find *15.7 Kruskal's Algorithm: Proof of Correctness 15.8 Application: Single-Link Clustering Problems 52 52 57 62 69 76 81 91 94 99 16 Introduction to Dynamic Programming 16.1 The Weighted Independent Set Problem 16.2 A Linear-Time Algorithm for WIS in Paths v 103 104 108 vi Contents 16.3 A
Reconstruction Algorithm 16.4 The Principles of Dynamic Programming 17.1 Sequence Alignment *17.2 Optimal Binary Search Trees Problems 137 148 163 18 Shortest Paths Revisited 18.1 Shortest Paths with Negative Edge Lengths 18.2 The Bellman
Ford Algorithm 18.3 The All-Pairs Shortest Path Problems 167 167 172 185 187 198 Epilogue: A Field Guide to Algorithm Design 201 Hints and Solutions to Selected Problems 203 Index 211 Preface This book is the third of a four-part series based on my online algorithms courses that have been running
regularly since 2012, which in turn are based on an undergraduate course that I taught many times at Stanford University. For if S \vdash were an independent set of total weight W \vdash in the larger graph G. c) Given capacities C1 and C2 of two knapsacks, compute
disjoint subsets S1, SP 2 of items with the maximum-possible total P value v + i2S1 i i2S2 vi, subject to the knapsack capacities: PP i2S1 is C1 and i2S2 vi, subject to the knapsack capacities: PP i2S1 is C1 and i2S2 vi, subject to the knapsack capacities: PP i2S1 is C1 and i2S2 vi, subject to the knapsack capacities: PP i2S1 is C1 and i2S2 vi, subject to the knapsack capacities: PP i2S1 is C1 and i2S2 vi, subject to the knapsack capacities: PP i2S1 is C1 and i2S2 vi, subject to the knapsack capacities: PP i2S1 is C1 and i2S2 vi, subject to the knapsack capacities: PP i2S1 is C1 and i2S2 vi, subject to the knapsack capacities: PP i2S1 is C1 and i2S2 vi, subject to the knapsack capacities: PP i2S1 is C1 and i2S2 vi, subject to the knapsack capacities: PP i2S1 is C1 and i2S2 vi, subject to the knapsack capacities: PP i2S1 is C1 and i2S2 vi, subject to the knapsack capacities: PP i2S1 is C1 and i2S2 vi, subject to the knapsack capacities: PP i2S1 is C1 and i2S2 vi, subject to the knapsack capacities: PP i2S1 is C1 and i2S2 vi, subject to the knapsack capacities: PP i2S1 vi i2S2 vi i
algorithm that solves the singlesource shortest path problem in O(mn) time, where m and n are the number of edges and vertices of the input graph, respectively. Objects in x's left subtree have keys smaller than that of x. Relatedly, a divide-and-conquer algorithm generally recurses on subproblems with size at most a constant fraction (like 50%) of
the input. If so, greedy with respect to what criterion? (Remember, higher scores are better.) Take a minute to brainstorm some formulas that have both of these properties., 22}, but some of its own recursive calls will be on suffixes of its input, such as {18, 19, 20, 21, 22}. 74 Minimum Spanning Trees Proof: Let T be a spanning tree of a graph G =
(V, E) with n vertices. But how would you ever come up with them? Each of the n + 1 subproblems is now solved from c to any other vertex. Let ei denote the ith edge of G considered by the Kruskal algorithm (after arbitrarily breaking
ties in its sorting preprocessing step). Let G = (V, E) be a connected undirected graph with real-valued edge costs that need not be distinct. c) Given a directed graph with nonnegative edge lengths, compute the length of a longest cycle-free path between any pair of vertices. The three-hop path qualifies even though the vertex 5 does not belong to the
prefix {1, 2, 3}; as the destination, that vertex is granted an exemption.) When k = 4 (or larger), the subproblem solution by tracing back through the filled-in array A. The edge set T 0 = T {e} [ {e0 } is a spanning tree with
total cost less than that of T, contradicting the assumption that T is an MST. Kruskal's algorithm repeatedly adds the cheapest new edge that does not create a cycle, fusing the most similar pair of data points in different clusters.
Adding a second traversal increases the number of hops to 9 but decreases the overall length to 6. The initial forest has n trees, where n is the number of alphabet symbols. P The Bellman-Ford algorithm has played a prominent role in the evolution of Internet routing protocols. Each recursive call of a typical divide-and-conquer algorithm commits to a
single way of dividing the input into smaller subproblems. 10 Each recursive call divides its input array into its left and right halves. The starred sections of the book are the most advanced ones. If we schedule the first job first, the second job second, and the
third job third, the sum of the weighted completion times is 3 \cdot 1 + |\{z\}| 1 \cdot 6 = 15., k, and the subproblem's solution is +1., vn, item sizes s1, None of the data structures discussed previously in this book series are right for the job; we'll need a new one, called the union-find data structure. 29 Theorem 15.13 (Kruskal Run Time (Union-third job third, the sum of the data structures discussed previously in this book series are right for the job; we'll need a new one, called the union-find data structure.
Find-Based)) For every graph G = (V, E) and real-valued edge costs, the unionfind-based implementation of Kruskal runs in O((m + n) \log n) time, where m = |E| and n = |V|. 30 15.6.1 The Union-Find Data Structure Whenever a program does a significant computation over again, it's a clarion call for a data structure to speed up those
computations. The length of Pb equals the length the path 1 ! 2 ! 5 with the path 1 ! 2 ! 5 with the path 5 ! 3 ! 2 ! 4 produces a path that contains
the directed cycle 2! 5! 3! 2. Finally, repeatedly prune unlabeled leaves until none remain. That's certainly the case here. Lemma 18.1 (Bellman-Ford Optimal Substructure) Let G = (V, E) be a directed graph with real-valued edge lengths and source vertex s 2 V. Remember the mantra of any algorithm designer worth their salt: Can we do better?
identification of the right collection of subproblems. // Initialization T := ; U := Initialize(V) // union-find data structure sort edges of E by cost // e.g., using MergeSort // Main loop for each (v, w) T := T [ {(v, w)} // update due to component
fusion Union(U, v, w) return T The algorithm maintains the invariant that, at the beginning of a loop iteration, the sets of the union-find data structure U correspond to the connected components of (V, T). Formulating the right measure of subproblems ize can be tricky for graph problems., yn is either: (i) an optimal alignment of X 0 and Y 0,
supplemented with a match of xm and yn in the final column; 17.1 143 Sequence Alignment (ii) an optimal alignment of X and Y 0, supplemented with a match of xm and yn in the final column, where X 0 and Y 0 denote X and Y , respectively,
with the final symbols xm and yn removed. // Preprocessing T := ; sort edges of E by cost // e.g., using MergeSort26 // Main loop for each e 2 E, in nondecreasing order of cost do if T [ {e} return T Kruskal's algorithm considers the edges of the input graph one by one, from cheapest to most expensive, so it makes sense to
sort them in nondecreasing order of cost in a preprocessing step (using your favorite sorting algorithm; see footnote 3 in Chapter 13). Solution to Quiz 13.3 does not immediately imply that the GreedyRatio algorithm is always optimal. A labeled internal node is an ancestor of the (labeled) leaves
in its subtree and leads to a violation of the prefix-free constraint. 5 Conversely, because no leaf can be the ancestor of another, a tree with labels only at the leaves defines a prefix-free code. In other words, an edge (v, w) satisfies the MBP if and only if there is no v-w path consisting solely of edges with cost less than cvw. To make this assertion
precise, define a consecutive inversion in a 13.4 15 Proof of Correctness schedule as a pair i, j of jobs such that i > j and job i is processed immediately before job j. One way to interpret the minimum-penalty alignment is as the "most plausible explanation" of how one of the strings might have evolved into the other. 20 If the input graph has a negative
cycle, it will be detected by one of the invocations of the single-source shortest path subroutine. The recurrence in Corollary 16.2 is a formula that implements the second step by showing how to compute the solution to the ith subproblems. Obtain H from G by removing v and its
incident edges from G. Index n! (factorial), 4 | S | (set size), 35 big-theta notation, 112 binary search tree property, 148 bit, 23 blazingly fast, viii, 64 Borodin, Allan, 2 breadth-first search, 55, 81 broken clock, 61 BubbleSort, 19 acknowledgments, xi algorithm design paradigm
1 divide-and-conquer, see divide-and-conquer algorithms dynamic programming, see dynamic programming greedy algorithms, see greedy algorithms, see greedy algorithms, see greedy algorithms dynamic programming greedy algorithms, see greedy algorithms, see greedy algorithms dynamic programming, see dynamic programming greedy algorithms, see greedy algorithms, see greedy algorithms algorithms algorithms dynamic programming greedy algorithms, see greedy algorithms algorithm
formula, 56 clustering, 94-97 k-means, 96 and Kruskal's algorithm, 97 choosing the number of clusters, 94 greedy criterion, 96 informal goal, 94 similarity function, 94 single-link, 97 cocktail party, ix code ^-tree, 31 alphabet, 23 as a tree, 28-31 average leaf depth, 31 average encoding length, 26 binary, 23 encodings as root-leaf paths, 30 Backurs,
Arturs, 146 Bellman, Richard E., 122, 172 Bellman-Ford algorithm, 172-185 and Internet routing, 183 correctness, 178, 180 example, 180-185 space usage, 183 stopping criterion, 177 subproblems, 172-174 big-O notation, 3 211 212
fixed-length, 23 Huffman, see Huffman's algorithm optimal prefix-free, 26, 32 prefix-free, 25, 31 symbol frequencies, 25, 27 variable-length, 24 compression, 23 computational genomics, see sequence alignment connected component (of a graph), 72 corollary, 10 Coursera, x cut (of a graph), 101 cycle (of a graph), 53 negative, 169 data structure
binary search tree, 148 disjoint-set, see union-find heap, see heap principle of parsimony, 202 queue, 41 union-find, see union-find when to use, 202 depth-first search, 55, 81 design patterns, viii diff, 164 Dijkstra's shortest-path algorithm, 2 inductive correctness proof, 13 resembles Prim's algorithm, 57 with negative edge
lengths, 168 Dijkstra, Edsger W., 57 discussion forum, xi distance, see shortest paths, distance divide-and-conquer algorithms, 1-3 Index vs. Does one of the algorithms forum, xi distance, see shortest paths, distance divide-and-conquer algorithms, 1-3 Index vs. Does one of the algorithms, it's helpful to see an example of Kruskal's algorithm in action before proceeding to its pseudocode. This
process is repeatable and you can mimic it when you apply the dynamic programming paradigm to problems that arise in your own projects. 26 This approach would not work well for the single-source shortest path problem, as the suffix path P2 would have the wrong origin vertex. In each iteration, we'll greedily add the cheapest edge that extends
the reach of the tree-so-far. The connected components remain exactly the same, with the new edge (v, w) swallowed up by the connected component that already contains both its endpoints. A tautology: The last allowable vertex k either appears as an internal vertex of P, or it doesn't. With additional assumptions, however, specialized sorting
algorithms can do better. Nothing can stop us from using it to solve all the subproblems systematically, beginning with the base cases and working up to the original problem. We define a new schedule 0 that is identical to responsible order, with j now processed immediately before i. B A crossing edges
102 Minimum Spanning Trees An edge of G crosses the cut (A, B) if it has one endpoint in each of A and B. Output: the ^-tree with minimum average leaf depth, representing the prefix-free binary code with minimum average leaf depth, representing the prefix-free binary code with minimum average encoding length.
the minimum computations (over edges) in the straightforward implementation with calls to ExtractMin. Skills You'll Learn Mastering algorithms takes time and j = n is exactly the same as the original problem. If there is a natural way to break down the problem into smaller subproblems, how easy
```

```
would it be to combine their solutions into one for the original problem? (To be covered in Part 4.) If all preceding steps end in failure, contemplate the unfortunate but common possibility that there is no efficient algorithm for your problem. Both are slower but more general than Dijkstra's shortest-path algorithm (covered in Chapter 9 of Part 2, and
similar to Prim's algorithm in Sections 15.2-15.4). Another possibility is that the new edge closes a pre-existing path, creating a cycle (Figure 15.7(c)). For example, if we choose to demote T1 's root, it is installed as a child of an object in T2. After
the vertex x is moved from V the new picture is: not-yet-processed processed X s X to X, 7 3 x V-X 2 y 5 1 z Edges of the form (v, x) with v 2 X get sucked into X and no longer cross the frontier (as with the edges with costs 3 and 7). Lemma 15.17 (Kruskal Outputs a Spanning Tree) For every connected input graph, the Kruskal algorithm outputs a
spanning tree. First, suppose you're trying to figure out the function of a complex genome, like the human genome. Each iteration is responsible for checking whether the edge e = (v, w) under examination can be added to the solution-so-far T without creating a cycle. If the path P uses its full edge budget, we follow the pattern of several
 previous case studies and pluck off the last edge of P to obtain a solution to a smaller subproblem. Consider a four-symbol alphabet, say ^* = {A, B, C, D}. By Lemma 13.2, it can only terminate at the greedy schedule . ... If you see how to do it efficiently, proceed with the divide-and-conquer paradigm. For example, an algorithm for computing driving
directions should accommodate any possible origin; this corresponds to the all-pairs shortest path problem. What is the NW score of the strings AGT ACG and ACAT AG? The algorithm closely resembles Dijkstra's shortest path problem. What is the NW score of the strings AGT ACG and ACAT AG? The algorithm closely resembles Dijkstra's shortest path problem.
field of the promoted root is updated accordingly, to the combined size of the two trees. 36 Union 1. balanced binary search trees, 151 vs. The alignment problem (Section 17.1.5), there was one parameter tracking the prefix of each of the two input
strings. In the worst case, there could be as many as n such roots. wj 'j Thus, already, our first case study illustrates the first theme of the greedy paradigm (Section 13.1.2): It is often easy to propose multiple competing greedy algorithms for a problem. The "1" now contributes .8 \rightarrow 1 = .8 to the average search time, the "2" contributes .1 \rightarrow 2 = .2, and
the "3" contributes .1 \rightarrow 3 = .3, for a total of .8 + .2 + .3 = 1.3. Solution to Quiz 17.5 Correct answer: (d). Standard implementations of heaps provide the following guarantee. In the first iteration, the recurrence evaluates to 0 at s (s has no incoming edges, so Case 2 of the recurrence is vacuous); to 2 at u (because A[0][s] + `su = 2); to 4 at v (because
A[0][s] + sv = 4; and to +1 at w and t (because A[0][v] = +\infty A[1][v] = +\infty A[0][v] = +\infty
induction and Lemma 16.1. Hint for Problem 16.3: If G is a tree, root it at an arbitrary vertex and define one subproblem for each subtree. The next step is to derive a similarity function F for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the given function f for pairs of clusters from the 
does not create any new cycles and decreases the number of connected components by 1. Run the Bellman-Ford algorithm on G0 with source vertex s to check whether G contains a negative cycle. Huffman's term paper in a class, believe it or not, and it superseded the (suboptimal) divide-and-conquer-esque top-down algorithm previously invented by
Huffman's graduate advisor, Robert M. After that, everything else falls into place in a fairly formulaic way. (See www.algorithmsilluminated. This suggests that the Bellman-Ford algorithm might be implementable even at an Internet scale, with each machine communicating only with its immediate neighbors and performing only local computations,
blissfully unaware of what's going on in the rest of the network. Divide the input into smaller subproblems. Thus log m lies between log(n 1) and 2 log n for every connected graph with no parallel edges, which justifies using log m and log n interchangeably inside a big-O expression. 14.2.3 Problem Definition (Rephrased) We can now restate the
optimal prefix-free code problem in a particularly crisp form. A perfectly balanced tree minimizes the length of the longest root-leaf path († log2 n for n objects) or, equivalently, the maximum search time. The root of the tree corresponds to the initial call to the algorithm (with the original input), with one child at the next level for each of its recursive
 calls. First, some terminology. The Upshot P In the sequence alignment problem, the input comprises two strings and penalties for gaps and mismatches, and Albert R. It is important that the item sizes are integers, as we'll see in
due time. The proof breaks down into two steps. No portion of this book may be reproduced in any form without permission from the publisher, except as permitted by U. Solution to Quiz 15.6 Correct answer: (c). The objective function value can only decrease throughout this process (by (13.4)), so is at least as good as - 16.1 105 The Weighted
Independent Set Problem a) 1 and 2 (respectively) b) 5 and 10 c) 6 and 10 d) 6 and 16 (See Section 16.1.4 for the solution and discussion.) We can now state the weighted independent set (WIS) Input: An undirected graph G = (V, E) and a nonnegative weight wv for each vertex v 2 V . 110
Challenge Problems Problems Problem 17.6 (H) In the sequence alignment problem, suppose you knew that the input strings were relatively similar, in the sense that there is an optimal alignment that uses at most k gaps, where k is much smaller than the lengths m and n of the strings. We can then picture the current state of the data structure as a directed
graph—the parent graph—with vertices corresponding to (indices of) objects x 2 X and a directed edge (x, y), called a parent edge, whenever parent(x) = y.34 For example, if X has six objects and the current state of the data structure is: Index of object x 1 2 3 4 5 6 parent(x) 4 1 1 4 6 6 then the parent graph is a pair of disjoint trees, with each root
pointing back to itself: 34 The parent graph exists only in our minds. Hint for Problem 15.8: For (a), the high-level idea is to perform a binary search for the bottleneck of an MBST. Suppose, for contradiction, that P1 is not an optimal solution to its subproblem; the argument for P2 is analogous., r1}, and similarly T2 for the keys {r + 1, r + 2, .20}
This observation is related to a mystery that might be troubling readers of Part 2: Why is Dijkstra's algorithm correct only with nonnegative edge lengths, while Prim's algorithm is correct with arbitrary (positive or negative) edge costs? First, the names of the keys are not important, so among friends let's just call them {1, 2, . After its addition, the
tree-so-far spans a, b, and d. Part 2 covers data structures (heaps, balanced search trees, hash tables, bloom filters), graph primitives (breadth- and depth-first search, connectivity, shortest paths), and their applications (ranging from deduplication to social network analysis). The choice of root has unpredictable repercussions further down the tree,
so how could we know in advance the right way to split the problem into two smaller subproblems? 10 The MST problem definition makes no reference to a starting vertex, so it might seem weird to artificially introduce one here. Which of the following greedy algorithms produces a schedule that minimizes the total lateness? But the ways in which
they fail will help you better understand the problem. Two jobs conflict if they overlap in time—if one of them starts between the start and finish times of the other. For example, a clever trick reduces the all-pairs shortest path problem (with negative edge lengths) to one invocation of the Bellman-Ford algorithm followed by n 1 invocations of
Dijkstra's algorithm. With WIS, we just designed our first dynamic programming algorithm! The general dynamic programming paradigm can be summarized by a three-step recipe. The items in the knapsack problem are not inherently ordered, but to identify the right collection of subproblems, it's helpful to mimic our previous approach and pretend
they're ordered in some arbitrary way. An independent set of G is a subset S 🗸 V of mutually non-adjacent vertices: for every v, w 2 S, (v, w) 2 / E. What about the OptBST algorithm via Heaps Invariant The key of a vertex w 2 V X is
the minimum cost of an edge (v, w) with v 2 X, or +1 if no such edge exists. The best-case scenario would be to come up with a formula that compiles each job's length and weight into a single score, so that scheduling jobs from highest to lowest score is guaranteed to minimize the sum of weighted completion times. Test Your Understanding Problem
14.1 (S) Consider the following symbol frequencies for a five-symbol alphabet: 50 Huffman Codes Symbol A B C D E Frequency .32 .25 .2 .18 .05 What is the average encoding length of an optimal prefix-free code? These "algorithms for problems that arise in your own work. Which of the
following is true about G0? This is not optimal, as the second and fourth vertices constitute an independent set with total weight 8. 15 For an example of the algorithm in action, see Problem 17.4. 160 Advanced Dynamic Programming Pictorially, we can visualize the array A in the OptBST algorithm as a two-dimensional table, with each iteration of
the outer for loop corresponding to a diagonal and with the inner for loop filling in the diagonal's entries from "southwest" to "northeast": s=0 index j of largest key ............ It is best understood through examples; we have only one so far, so I encourage you to revisit this section after we finish a few more case studies. Instead, we'll devise from
scratch an algorithm for a tricky and concrete computational problem, which will force us to develop a number of new ideas. Consider the first iteration of the Huffman algorithm, in which it merges the leaves that correspond to a and b. Thus, the key of a vertex w 2 V X is exactly the winning edge cost in the local tournament at w. 15.8.2 Bottom-Up
Clustering The main idea in bottom-up or agglomerative clustering is to begin with every data point in its own cluster, and then successively merge pairs of clusters until exactly k remain. 16.5 131 The Knapsack Problem 16.3 (H) This
problem outlines an approach to solving the WIS problem in graphs more complicated than paths. The minimum spanning tree comprises the edges (a, b), (b, d), and (a, c): a 1 b 4 2 c d The sum of the edges' costs is 7., P (k 1) are all true—this is called the inductive hypothesis—and use this 11 For an induction refresher, see Appendix A of Part 1 or
the book Mathematics for Computer Science mentioned in the Preface. 13.2.2 Completion Times A schedule specifies an order in which to process the jobs. Here, you want to favor shorter jobs. What's the simplest greedy algorithm that might work? 22 The converse of Theorem 15.6 is also true, even with non-distinct edge costs: Every edge of an MSTable true, even with non-distinct edge costs: Every edge of an MSTable true, even with non-distinct edge costs: Every edge of an MSTable true, even with non-distinct edge costs: Every edge of an MSTable true, even with non-distinct edge costs: Every edge of an MSTable true, even with non-distinct edge costs: Every edge of an MSTable true, even with non-distinct edge costs: Every edge of an MSTable true, even with non-distinct edge costs: Every edge of an MSTable true, even with non-distinct edge costs: Every edge of an MSTable true, even with non-distinct edge costs: Every edge of an MSTable true, even with non-distinct edge costs: Every edge of an MSTable true, even with non-distinct edge costs: Every edge of an MSTable true, even with non-distinct edge costs: Every edge of an MSTable true, even with non-distinct edge costs: Every edge of an MSTable true, even with non-distinct edge costs: Every edge of an edge of an
satisfies the MBP (Problem 15.4). Thus, each loop iteration takes O(1) time, for a blazingly fast running time of O(n). 16.5 129 The Knapsack Problems. The sequence alignment problem is, then, to compute an alignment that minimizes the total
penalty. Let's examine these cases in reverse order. The total frequency of a and b in p is the same as that of ab in p0, but the depth of the corresponding leaves is one larger. 4 Suppose the largest number of bits used to encode a symbol is `. Problem 16.4 (S) Consider an instance of the knapsack problem with five items: Item 1 2 3 4 5 Value 1 2 3 4
5 Size 1 3 2 5 4 and knapsack capacity C = 9., k, and moreover is a shortest such path. a) 3 b) 4 c) 5 d) 6 (See Section 17.1.8 for the solution and discussion.) 2 While it's natural to assume that all penalties are nonnegative with 4xx = 0 for all x \ge ^a and 4xy = 4yx for all x \ge ^a, our dynamic programming algorithm requires only that the gap
penalty is nonnegative. We start with two assumptions. The number of subproblems and the work-per-subproblem each blow up by a constant factor. In many applications, like computing driving directions, edge lengths are automatically nonnegative (barring a time machine) and there's nothing to worry about. By (†), this tree must be optimal for the penalty is nonnegative.
original problem. Alternatively, the problem in (b) reduces to the sequence alignment problem by setting two different symbols to a very large number. 15.8 Application: Single-Link Clustering 93 Proof: The algorithm explicitly ensures that its final output T
is acyclic. Tacking a cycle traversal at the end produces a five-hop s-v path with total length 8. Case 2: xm matched with a gap in last column of alignment. The second iteration follows suit with the next-cheapest edge (the edge of cost 2). P In the minimum spanning tree (MST) problem, the input is a connected undirected graph with real-valued edge
costs and the goal is to compute a spanning tree with the minimum-possible sum of edge costs. We use the term proposition for stand-alone technical statements that are not particularly important in their own right. Thus, there is an optimal schedule - of these jobs with a strictly smaller sum of weighted completion times. For i = 0, 1, 2, . See
Chapter 6 of Part 1.) Obtain G0 from G by throwing out all the edges with cost higher than the median. Quiz 15.5 What's the running time of the Find operation, as a function of the number n of objects? In the WIS problem on path graphs (Section 16.2.1), if only you knew whether the last vertex belonged to an optimal solution, you would know what
the rest of it looked like. (Highlights include "union-by-rank," "path compression," and the "inverse Ackermann function." It's amazing stuff!) 84 Minimum Spanning Trees the union-find data structure accordingly. For the inductive step, assume that k > 2 and fix an alphabet ^{\circ} of size k and nonnegative symbol frequencies p = \{px \}x2^{\circ}. Because \leftarrow
6=, Lemma 13.2 applies to ←, and there are consecutive jobs i, j in ← with i > j (Figure 13.3(a)). The Huffman algorithm uses 1 bit to encode each symbol (0 for one symbol and 1 for the other), which is the minimum possible. Proof of Theorem 15.1: Corollary 15.10 proves that the output of Prim's
algorithm is a spanning tree. {z} Case r The more general statement in Corollary 17.5 follows by invoking the first statement, for each i, j 2 {1, 2, . For starters, what if you knew that all the jobs had the same length (but possibly different weights)? Alternatively, randomized QuickSort (see Chapter 5 of Part 1) has an average running time of O(n log
n). The new values at w and t are 4 (because A[1][u] + `uw = 4) and 8 (because A[1][v] + `vt = 8): v 4 A[2][s] = 0 A[0][t] = +\infty A[0][v] = +\infty A[0][v] = +\infty A[0][w] = +\infty A[0][w] = +\infty A[0][w] = +\infty A[0][w] = +\infty A[0][v] = +\infty A[0]
iteration does not propagate to t immediately, only in the next iteration. How big can this be? Easy to come up with one or more greedy algorithms. For an arbitrary graph G, what would your subproblems be? (Choose the strongest true statement.) a) O(n2) b) O(mn) c) O(mn) c
Summarizing everything we now know about the Bellman-Ford algorithm: 12 For an example of the algorithm in action on an input graph with a negative cycle, see Problem 18.1. 18.2 The Bellman-Ford Algorithm 183 Theorem 18.5 (Properties of Bellman-Ford Algorithm 183 Theorem 18.5 (Properties of Bellman-Ford Algorithm 183 Theorem 18.6).
source vertex s, the Bellman-Ford algorithm runs in O(mn) time and either: (i) returns the shortest-path distance from s to every destination v 2 V; or (ii) detects that G contains a negative cycle. There are n different ways in which an optimal solution can be built from optimal solutions to smaller subproblems, resulting in a recurrence with n cases
This matches B's encoding in our fixed-length code. With negative edge lengths, the length of P 0 might even be larger than that of P. His research interests include the many connections between computer science and economics, as well as the design, analysis, applications, and limitations of algorithms. That is, among all spanning trees T of a
connected graph with strictly positive Q edge costs, compute one with the minimum-possible product e2T ce of edge costs. Optimal BST: Subproblems Compute Wi, j, the minimum weighted search time of a binary search tree with keys {i, i + 1, . 18.2 175 The Bellman-Ford Algorithm Case 1: P has i 1 or fewer edges. Quiz 18.5 Do you see any bugs in
the argument above? The problem of computing an optimal prefix-free code looks intimidating at first encounter. 33 These bounds are for the quick-and-dirty implementation. `j GreedyRatio Schedule the jobs in decreasing order of (breaking ties
 arbitrarily). The second assumption imposes a non-trivial restriction on the input; we will do some extra work to remove it in Section 13.4.4. Together, the two assumptions imply that jobs are indexed in strictly decreasing order of weight-length ratio. Crucial to reasoning about the problem is a method of associating codes with labeled binary trees.
14.2.1 Three Examples The connection between codes and trees is easiest to explain through examples. Try thinking of some combination that will possibly give it a pejorative meaning. If we bump up i to 3 (or more), the path s!u!w!t becomes eligible and lowers the shortest-path distance from 4 to 3. The search tree property imposes a
requirement for every node of a search tree, not just for the root x all keys x For example, here's a search tree containing objects with the keys {1, 2, 3, 4, 5}: root 3 5 1 leaves 2 4 The point of the search tree containing objects with the keys {1, 2, 3, 4, 5}: root 3 5 1 leaves 2 4 The point of the search tree containing objects with the keys {1, 2, 3, 4, 5}: root 3 5 1 leaves 2 4 The point of the search tree containing objects with the keys {1, 2, 3, 4, 5}: root 3 5 1 leaves 2 4 The point of the search tree containing objects with the keys {1, 2, 3, 4, 5}: root 3 5 1 leaves 2 4 The point of the search tree containing objects with the keys {1, 2, 3, 4, 5}: root 3 5 1 leaves 2 4 The point of the search tree containing objects with the keys {1, 2, 3, 4, 5}: root 3 5 1 leaves 2 4 The point of the search tree containing objects with the keys {1, 2, 3, 4, 5}: root 3 5 1 leaves 2 4 The point of the search tree containing objects with the keys {1, 2, 3, 4, 5}: root 3 5 1 leaves 2 4 The point of the search tree containing objects with the keys {1, 2, 3, 4, 5}: root 3 5 1 leaves 2 4 The point of the search tree containing objects with the keys {1, 2, 3, 4, 5}: root 3 5 1 leaves 2 4 The point of the search tree containing objects with the keys {1, 2, 3, 4, 5}: root 3 5 1 leaves 2 4 The point of the search tree containing objects with the keys {1, 2, 3, 4, 5}: root 3 5 1 leaves 2 4 The point of the search tree containing objects with the keys {1, 2, 3, 4, 5}: root 3 5 1 leaves 2 4 The point of the search tree containing objects with the keys {1, 2, 3, 4, 5}: root 3 5 1 leaves 2 4 The point of the search tree containing objects with the keys {1, 2, 3, 4, 5}: root 3 5 1 leaves 2 4 The point of the search tree containing objects with the keys {1, 2, 3, 4, 5}: root 3 5 1 leaves 2 4 The point of the search tree containing objects with the keys {1, 2, 3, 4, 5}: root 3 5 1 leaves 2 4 The point of the search tree containing objects with the keys {1, 2, 3, 4, 5}: root 3 5 1 leaves 2 4 The point of the keys {1, 2, 3, 4, 5}: roo
array. 15.1 Problem Definition 55 is no path in E between the vertices v and w, there certainly isn't one in any subset T < E of edges, either.) For this reason, throughout this chapter we assume that the input graph is a connected graph. Theorem 15.3 (Running Time of Three Heap Operations) In a heap with n objects, the Insert, ExtractMin, and
Delete operations run in O(log n) time., with jobs scheduled in order of Going from bottom to top in the greedy schedule, the indices of the jobs always go up. The most valuable pair of items is the third and fourth ones (with total value 8), and these fit in the knapsack (with total size 5). 16.5.6 Reconstruction The Knapsack algorithm computes only
the total value of an optimal solution, not the optimal solution itself. For (d), because S is feasible for the reduced capacity. Such algorithms are usually easy to come up with, and even when one fails to solve
the problem (as is often the case), the manner in which it fails can help you better understand the intricacies of the problem. (Choose all that apply.) a) Consider the variation of sequence alignment in which, instead of a single gap penalty dgap, you are given two positive numbers a and b. One of the reasons why it can be hard to prove the
correctness of greedy algorithms is that most such algorithms are not correct, meaning there exist inputs for which the algorithm fails to produce the desired output. Problem: Minimum Spanning Tree (MST) Input: A connected undirected graph G = (V, E) and a real-valued cost ce for each edge e 2 E. In other words, once you know that an MWIS
excludes the last vertex, you know exactly what it looks like: It's an MWIS of the smaller graph Gn 1 . Fix a pair v, w of vertices; because the input graph is connected, it contains a v-w path P . (If no such path exists, dist(s, v) is defined as +1.) For example, the shortest-path distances from s in the graph v 6 1 2 s 4 w 167 t 3 168 Shortest Paths
Revisited are dist(s, s) = 0, dist(s, v) = 1, dist(s, v) = 1, dist(s, v) = 3, and dist(s, v) = 3, and dist(s, v) = 4. Case 2: vn 2 S. Greedy algorithms and applications.
protocols RIP and RIP2—yet another example of how algorithms shape the world as we know it.15 18.2.9 Solutions to Quizzes 18.2-18.3 Solution to Quizzes 18.2
are? 3 A cycle in a graph G = (V, E) is a path that loops back to where it began—an edge sequence e1 = (v_0, v_1), e2 = (v_0, v_1), but this time for different reasons. We can think of a gap as undoing a deletion that occurred sometime
in the past, and a mismatch as undoing a mutation. d) It might not be feasible if the knapsack capacity is only C sn. For example, consider the graph 4 -10 -10 5 1 2 5 2 22 -4 3 Every vertex of a path other than its endpoints is an internal vertex. Corollary 15.8 (Spanning Trees Have Exactly n 1 Edges) Every spanning tree of an n-vertex graph has n 1
edges. Hints and Solutions to Selected Problems 205 than ce. (13.4) benefit of exchange In other words, the swap cannot increase the sum might decrease, or it might stay the same. 10 Have we made any progress? Because the input graph has no negative cycles, this sum is nonnegative and the length of P 0 is
less than or equal to that of P. Output: A ^-tree with minimum-possible average leaf depth (14.1). In the sequence alignment, you would know what the rest of it looked like. The path P must be an optimal solution to this smaller subproblem, as any
shorter s-v path with at most i 1 edges would also be a superior solution to the original subproblem, contradicting the purported optimality of P. The strings need not have the same length. With variable-length codes, we must impose a constraint to prevent ambiguity. Unlike our previous dynamic programming case studies, every subproblem works
with the full input (rather than a prefix or subset of it); the genius of these subproblems lies in how they control the allowable size of the output. It's impossible., j} for some i, j 2 {1, 2, . Output: The binary search tree T containing the keys {k1 , k2 , . Show how to compute the NW score in O((m + n)k) time. Shortly, we'll see that there's no reason to
bother with subproblems in which i is greater than n, the number of vertices, which implies that there are O(n2) relevant subproblems. 9 18.2.2 Optimal Substructure With our clever choice of subproblems. 62 Minimum Spanning Trees all the
 edges to identify the cheapest one with one endpoint in each of X and V X. First suppose that all n jobs have the same length, say length 1. This assertion implies that any s-v path with at most n 1 edges. The first-round winners (at most one
per vertex w 2 V X) proceed to the second round, and the final champion is the cheapest first-round winner., k 1}.25,26 You can guess the next step: We want to prove that P1 and P2 are, in fact, optimal solutions to these smaller subproblems. 170 Shortest Paths Revisited Thus, there is no shortest s-v path, and the only sensible definition of dist(s, v)
is 1. Symmetrically, in this case, the induced alignment is of X and Y 0: |X + gaps Y 0 + gaps Y 
following symbol frequencies for a five-symbol A B C D E Frequency .16 .08 .35 .07 .34 What is the average encoding length of an optimal prefix-free code? For (c), similarly, every edge chosen by Kruskal's algorithm is justified by the Cut Property. c) The single-source shortest-path problem., n (where n = |V|). We conclude that (V,
T) is neither connected nor acyclic. Kruskal in the mid-1950s—roughly the same time that Prim and Dijkstra were rediscovering what is now called Prim's algorithm. If such a formula exists, our two special cases imply that it must have two properties: (i) holding the length fixed, it should be increasing in the job's weight; and (ii) holding the weight
 fixed, it should be decreasing in the job's length. You can imagine how he felt, then, about the term, mathematical. Section 16.3 shows how to also identify the vertices of an MWIS. 16 *17.2 Optimal Binary Search Trees 161 Theorem 17.6 (Properties of OptBST) For every set {1, 2, . 13.4 Proof of Correctness Divide-and-conquer algorithms usually
have formulaic correctness proofs, consisting of a straightforward induction. In sparse graphs, where m is linear or near-linear in n, this time bound is much better than the more naive bound of O(n3). 120 Introduction to Dynamic Programming 16.4.3 A Repeatable Thought Process When devising your own dynamic programming algorithms, the
heart of the matter is figuring out the magical collection of subproblems. The proof of Theorem 13.1 includes a vivid example of one such theme: exchange arguments. There are also fast algorithms for this problem, but they lie a bit beyond the scope of this book series. Lemma 18.6 (Floyd-Warshall Optimal Substructure) Let G = (V, E) be a directed
graph with real-valued edge lengths and no negative cycles, with V = {1, 2, . Our dynamic programming boot camp will double as a tour of some of the paradigm's killer applications, including the knapsack problem, the Needleman-Wunsch genome seguence alignment algorithm, Knuth's algorithm for optivity viii Preface mal binary search trees, and
corresponds to the symbol with a 1-bit encoding (A), the level-2 leaf to the symbol with a 2-bit encoding (B), and the level-3 leaves to the two symbols with 3-bit encoding (C and D). Then, along with each subproblem solution A[i][j], store the choice of the root r(i, j) that minimizes A[i][r 1] + A[r + 1][j] or, equivalently, the root of an optimal search
18.2.5 Pseudocode The justifiably famous Bellman-Ford algorithm now writes itself: Use the recurrence in Corollary 18.2 to systematically solve all the subproblems, up to an edge budget of i = n. See Section 15.4 for a proof of Theorem 15.1. 15.2.3 Straightforward Implementation As is typical of greedy algorithms, the running time analysis of Prim's
 algorithm (assuming a straightforward implementation) is far easier than its correctness proof. (14.1) a 2 Proposition 14.1 implies that L(T, p) is exactly the average encoding length of the tree has a label, as removing unlabeled leaves does not change the
and topics. (17.1) | {z } i=1 = (ki 's depth in T )+1 Three comments. QE D 18.4.5 Summary and Open Questions Summarizing everything we now know about the Floyd-Warshall algorithm: Theorem 18.9 (Properties of Floyd-Warshall algorithm runs in
the "3" contributes .1 \rightarrow 2 = .2, for a total of 1.6 + .1 + .2 = 1.9. The second search tree has a larger maximum search time (3 instead of 2), but the lucky case of a search for the root is now much more likely. Confronted with a new problem, which paradigm should you use? This is why greedy algorithms aren't good enough for the latter problem; we'll
need to up our game and apply the dynamic programming paradigm., n} with total value V = i2S vi. (Every MST is an MBST but not conversely, as you should check.) The question of whether there is a deterministic linear-time algorithm for the MST problem remains open to this day; see the bonus video at www.algorithmsilluminated.org for the full
story. The more you compress, the faster the download. Genome similarity can be used as a proxy for proximity in a phylogenetic tree. The Floyd-Warshall algorithm need not compute the correct shortest-path distances, but it is still the case that A[k][v][w] is at most the minimum length of a cycle-free v-w path with internal vertices restricted to {1, 2,
 . 10 We no longer get an immediate contradiction in the case in which optimal schedule, as 0 could be a different, equally optimal, schedule. For example, the connected components of the graph in Figure 15.7(a) are {1, 3, 5, 7, 9}, {2, 4}, and {6, 8, 10}. We conclude that the running time of the recursive algorithm grows exponentially with n. 18.3
The All-Pairs Shortest Path Problem 18.3.1 Problem Definition Why be content computing shortest-path distances from only a single source vertex? 17.1.2 Problem Definition Our example strings AGGCT and AGGCA are obviously inhabited T1, the Find
operation traverses the same path as before (from x to the old root r of T1), plus the new parent edge from r to z, plus the parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to z, plus the new parent edge from r to
j ("stuff" in Figure 13.3) are the same in both \vdash and 0 (and in the same order), and likewise for the jobs that follow both i and j ("more stuff")., r 1}, and similarly for T2 and {r + 1, r + 2, . 31 Analogous to the Bellman-Ford algorithm (footnote 13), it's a good idea to maintain with each vertex pair v, w the last hop of a minimum-length cyclefree v-w
path with internal vertices restricted to {1, 2, . 10. (For vertices x 2 S1 and y 2 S2, you can produce an x-y path in G, the edge (v, w), and a w-y path in G.) Second, suppose for contradiction that the edge addition did create a new cycle C. What if they all had the same weight (but possibly different
 lengths)? This case can occur only when sn C. Case 1: vn 2 / S. Data sets, along with test cases and their solutions, can be found at www.algorithmsilluminated.org. 13 For simplicity, you can think of x and y as distinct from a and b, but the proof works fine even when {x, y} and {a, b} overlap (as you should check). This book is not an introduction to
programming, and ideally you've acquired basic programming skills in a standard language (like Java, Python, C, Scala, Haskell, etc.). Huffman Input: a nonnegative frequency pa for each symbol a of an alphabet ^. a) 1.5 b) 1.55 c) 2 d) 2.5 (See Section 14.1.6 for the solution and discussion.) 14.1.5 Problem Definition The preceding example shows
that the best binary code for the job depends on the symbol frequencies. | {z } | {z } Case 1 Case 2 More generally, for every i = 2, 3, . The other penalties (for matching two symbols) are defined as before. Section 15.8 outlines an application of Kruskal's algorithm in machine learning, to single-link clustering. Adding this edge to T would create a
cycle (with the edges of cost 2 and 3), so the algorithm is forced to skip it. Challenge Problems Problem 16.5 (H) This problem describes four generalizations of the knapsack problem. The key idea is to prove that every feasible solution can be improved by modifying it to look more like the output of the greedy algorithm. Here's one alignment with two
gaps and one mismatch, for a total penalty of 4: C A A G A T T A A C - G G Here's one with four gaps and no mismatches, also with a total penalty 3 or less. {z} (17.3) Case r More generally, for every i, j 2 {1, 2, . What changes when a new edge is added? Is this the maximum possible? Σ-
tree T \Sigma'-tree T' the mapping \alpha ab a b the mapping \beta Figure 14.2: There is a one-to-one correspondence between \alpha -trees in which a and b are the left and right children of a common parent. z^2 (a,b,x,y) Depths in T \omega can be rewritten in terms of depths in T. But rather than growing a single tree from a starting vertex, Kruskal's
algorithm can grow multiple trees in parallel, content for them to coalesce into a single tree only at the end of the algorithm. On Reductions A problem B if an algorithm that solves B can be easily translated into one that solves A. Here, each data point also has a label (e.g., 1 if the image is of a cat and 0 otherwise), and the goal
is to accurately predict the labels of as-yet-unseen data points. Union: given a union-find data structure and two objects x, y 2 X in it, merge the sets that contain x and y into a 31 A partition of a set X of objects is a way of splitting them into one or more groups. With two choices for each of the two entries in the last column, there would seem to be
four possible scenarios. In both problems, the input specifies a set of frequencies over symbols or keys, the output is a binary tree, and the algorithm halts. 140 Advanced Dynamic Programming The
NW score would be useless to genomicists without an efficient procedure for calculating it., wn and positive job lengths `1, `2, . For example, we could try: Symbol A B C D Encoding 0 01 10 1 This shorter code can only be better, right? To see this, imagine removing all the edges from the input graph and adding them back in, one by one. Invoke
Find twice to locate the positions i and j of the roots of the parent graph trees that contain x and y, respectively. 9 The keen reader might compute a profitable sequence of financial transactions that involves both buying and selling. The
minimum penalty of an alignment of two strings are then deemed "similar" if and only if their NW score is relatively small. Section 15.7 supplies the remaining details of the proof of Theorem 15.11. The invariant ensures that the local
winner of *15.3 Speeding Up Prim's Algorithm via Heaps 67 the extracted vertex is the cheapest edge crossing the frontier, which is the correct answer: (b). One natural fixed-length code for this alphabet is: Symbol A B C D Encoding 00 01 10 11 Suppose we wanted to get away with
fewer bits in our code by using a 1-bit encoding for some of the symbols. In how many different ways could it have been built up from optimal solutions to smaller subproblems? This is starting to sound familiar. In graphs with only nonnegative edge lengths, everything works out in the end and all the shortest-path distance estimates are correct. In a
problem instance with n jobs, there are n! = n \cdot (n \cdot 1) \cdot (n \cdot 2) \cdot \cdots \cdot 2 \cdot 1 different schedules. Problem all take as input two strings X and Y, with lengths m and n, over some alphabet n \cdot (n \cdot 1) \cdot (n \cdot 2) \cdot \cdots \cdot 2 \cdot 1 different schedules. Problem 17.3 (S) The following problems all take as input two strings X and Y, with lengths m and n, over some alphabet n \cdot (n \cdot 1) \cdot (n \cdot 2) \cdot \cdots \cdot 2 \cdot 1 different schedules. Problem 17.3 (S) The following problems all take as input two strings X and Y, with lengths m and n, over some alphabet n \cdot (n \cdot 1) \cdot (n \cdot 2) \cdot \cdots \cdot 2 \cdot 1 different schedules.
chosen pivot elements are. Given an operating budget and a set of job candidates with differing productivities and requested salaries, whom should you hire? Each data point could represent a person, an image, a document, a genome sequence, and so on. 6 For example, suppose the edges of the input graph and their lengths represent financial
transactions and their costs, with vertices corresponding to different asset portfolios. Quiz 15.4 Which of the following running times best describes a straightforward implementation of Kruskal's MST algorithm for graphs in adjacency-list representation? 0 The rearrangement makes it obvious that the difference on the lefthand side is nonnegative: pxtallors and their costs, with vertices corresponding to difference on the lefthand side is nonnegative: pxtallors and their costs, with vertices corresponding to difference on the lefthand side is nonnegative: pxtallors and their costs, with vertices corresponding to difference on the lefthand side is nonnegative: pxtallors and their costs, with vertices corresponding to difference on the lefthand side is nonnegative: pxtallors and their costs, with vertices corresponding to difference on the lefthand side is nonnegative: pxtallors and their costs, with vertices corresponding to difference on the lefthand side is nonnegative: pxtallors and their costs, with vertices corresponding to difference on the lefthand side is nonnegative: pxtallors and their costs, with vertices corresponding to difference on the lefthand side is nonnegative.
pa and py pb are nonnegative because a and b were chosen as the symbols with the smallest frequencies, and the other two terms on the right-hand side are nonnegative because x and y were chosen from the deepest level of T. Lines 9-11 implement one iteration of the main loop of Prim's algorithm. 13.2 5 A Scheduling Problem Quiz 13.1 Consider a
problem instance that has three jobs with `1 = 1, `2 = 2, and `3 = 3, and suppose they are scheduled in this order (with job 1 first). As usual, n and m denote the number of vertices and edges, respectively, of the input graph. Returning to our running example: 4 1 first half 5 4 second half the first and second recursive calls return the second and third
 vertices as the optimal solutions to their respective subproblems. After all the edges of P have been processed, all its vertices—and, in particular, its endpoints v and w—belong to the same connected component of the solution-so-far (V, T). Hint for Problem 15.2: Use Lemma 15.7 to prove that the output is a spanning tree. 18.1 Shortest Paths with
Negative Edge Lengths 169 If we want to accommodate negative edge lengths, we'll need a new shortest-path algorithm. 4 18.1.2 Negative Cycles Who cares about negative edge lengths, we'll need a new shortest-path algorithm. 4 18.1.2 Negative Cycles Who cares about negative edge lengths, we'll need a new shortest-path algorithm. 4 18.1.2 Negative Cycles Who cares about negative edge lengths? If size(i) := i and size(i) := size(i) + size(j). 16.2 A Linear-Time Algorithm for WIS in Paths 16.2.2 111 A Naive Recursive Approach Lemma 16.1
is good news—we've narrowed down the field to just two candidates for the optimal solution! So, why not try both options and return the better of the edge (a, d) have been sucked into the set of vertices spanned so far;
 adding this edge in the future would create a cycle, so the algorithm does not consider it further. In line 6, the algorithm now first checks whether the array A already contains the relevant solution S1; if not, it computes S1 recursively as before and caches the result in A. b) Schedule the jobs in increasing order of processing time pj. A dynamic
programming algorithm can then fill in an array with subproblems that correspond to prefixes of the input. How can you know if you're truly absorbing the concepts in this book? a) 22 and 23 b) 23 and 22
c) 17 and 17 d) 17 and 11 (See Section 13.3.3 for the solution and discussion.) We've made progress by ruling out the GreedyDiff algorithm from further consideration. In an ideal clustering, data points in the same cluster are relatively similar while those in different clusters are relatively dissimilar., kn }. The shortest-path distance subject to the hop
count constraint is effectively +1. Huffman's greedy criterion then dictates that we merge the pair of trees for which the sum (14.2) is as small as possible. QE D To apply Theorem 15.6, we must prove that every edge chosen by the Kruskal algorithm satisfies the minimum bottleneck property (MBP).37 Lemma 15.18 (Kruskal Achieves the MBP) For
every connected graph G = (V, E) and real-valued edge costs, every edge chosen by the Kruskal algorithm satisfies the MBP. Output: the edges of a minimum spanning tree of G. The "diagonal" entries of the subproblem array are the tell.30 Lemma 18.8 (Detecting a Negative Cycle) The input graph G = (V, E) has a negative cycle if and only if, at the
conclusion of the Floyd-Warshall algorithm, A[n][v][v] < 0 for some vertex v 2V. Solution to Quiz 17.3 Correct answer: (b). Does this fact have any implications for any of the four stated problems?, vn } and a nonnegative weight wi for each vertex vi . These are best communicated through the discussion forums mentioned above., `n . b) The formula
is always correct in path graphs but not always correct in trees. The Find operation performs O(1) work per parent edge traversal, so its worst-case running time is proportional to the largest depth of any object—equivalently, to the largest height of one of the trees in the parent graph. So, along with the parent field, the data structure stores with
each array entry a size field, initialized to 1 in Initialize to 1 in Initialize. A binary code for an alphabet is a way of writing each of its symbols, a natural encoding is to associate each symbol with one of the 26 = 64 length-6 binary strings, with each
string used exactly once. Therefore, it makes sense to first solve all the subproblems with a single-key input, then the subproblems with two keys in the input, and so on. For example, a recursive call might be given a prefix of the original input {1, 2, . .} and v 2 V .) Paths with cycles are allowed as solutions to a subproblem. If this statement seems
obvious to you, feel free to skip the proof and move on. 66 Minimum Spanning Trees 15.3.4 Pseudocode then looks like this: Prim (Heap-Based) Input: connected undirected graph G = (V, E) in adjacency-list representation and a cost ce for each edge e 2 E. By Lemma 15.15, immediately after the iteration that processes the edge (xi 1)
, xi ), xi 1 and xi lie in the same connected component of the solution-so-far. Can we use a greedy approach? (We're not yet claiming tree.) For every connected input graph, the Prim algorithm outputs a spanning tree. The only subproblems that can arise in this way are for
prefixes of the original input strings. (a) (Difficult) Give a linear-time algorithm for computing the bottleneck of an MBST. For example, if vertices represent people who dislike each other, the independent sets correspond to groups of people who dislike each other, the independent sets correspond to groups of people who dislike each other, the independent sets correspond to groups of people who dislike each other, the independent sets correspond to groups of people who dislike each other, the independent sets correspond to groups of people who dislike each other.
 general graph, however, there is no intrinsic ordering of the vertices or edges, and few clues about which subgraphs are the relevant ones. The Cut Property then implies that every MST contains every edge of the algorithm's final spanning tree T, and so T 206 Hints and Solutions to Selected Problems is the unique MST. As a reminder, the average
 leaf depth of a ^-tree T with respect to the symbol frequencies p is X L(T, p) = px · (depth of the leaf labeled x in T). The first step identifies a property, called the "minimum bottleneck property," possessed by the output of Prim's algorithm. With a good sorting algorithm (like MergeSort), this step contributes O(m log n) work to the overall running
time.28 This work will be dominated by that done by the main loop of the algorithm, which we analyze next. 8 This algorithm was discovered independently by many different people in the mid-to-late 1950s, including Richard E. Output: the minimum weighted search time (17.1) of a binary search tree with the keys {1, 2, . As in our derivation of the
equation (13.3), the cost and benefit of this exchange are wi 'j and wj 'i, respectively. Any of the n keys might appear at the root of an optimal solution, so there are n different cases. This is not true for any other schedule. Quiz 13.4 What effect does the exchange have on the completion time of: (i) a job other than i or j; (ii) the job i; and (iii) the job j?
Because T ← is connected, so is T 0. The value at v drops from 4 (corresponding to the one-hop path s! v) to 1 (corresponding to the two-hop path s! v) to 1 (corresponding to the two-hop path s! v) to 1 (corresponding to the two-hop path s! v). This eliminates the adjacent vertices (the second and fourth ones, both with weight 4) from further consideration. To avoid redundant edges and ensure that the edge addition extends the reach of
T, the algorithm considers only the edges that "cross the frontier," with one endpoint in each of X and V X (Figure 15.4). And by "linear time," we mean linear in the size of the graph (V, T) which, as an acyclic graph with n vertices, has at most n 1 edges. Now suppose the cost of every edge e of G is increased by 1 and becomes ce + 1. Look for
significant computations that your algorithm performs over and over again (like lookups or minimum computations). 16 Introduction to Greedy Algorithms more stuff i exchange (b) After exchange Figure 13.3: Obtaining the new schedule 0 from the allegedly optimal schedule \vdash by
exchanging the jobs in a consecutive inversion (with i > j). For the first job, its completion time is just its length, which is 1. Thus, the condition Find(U, w) is met if and only if adding (v, w) to T does not create a cycle. (15.2) Other reasonable choices
for F include the worst-case or average similarity between points in the different clusters. For example, can it be reduced to sorting, graph search, or a shortest-path computation?33 If so, use the fastest algorithm sufficient for solving the problem. Lemma 18.3 (Bellman-Ford Stopping Criterion) Under assumptions and notation of Corollary 18.2, if for
some k 0 Lk+1,v = Lk,v the for every destination v, then: (a) Li,v = Lk,v for every i k and destination v, then tree—which species evolved from which, and when. This will lead
in at least one merge prior to the final iteration. One big benefit is that a starting vertex by the problem it solves, the single-source shortest path problem). To carry out its entire computation, the WIS algorithm must remember only the two most
recent subproblems. QE D 15.4.4 Putting It All Together We now have the ingredients to immediately deduce the correctness of Prim's algorithm in graphs with distinct edge costs. Consider an optimal alignment of two non-empty strings X = x1, x2, . This is roughly how ASCII codes work, for instance. This idea worked well in the WIS problem on
for an overall running time of O(n). Thus, I thought dynamic programming was a good name., vn, item sizes s1, s2, . 15.1.1 Graphs Objects and connections between them are most naturally modeled with graphs. The running time of your algorithm should be polynomial in n. Given a graph with real-valued edge costs, define the bottleneck of a path
P as the maximum cost maxe2P ce of one of its edges. The Cut Property Let G = (V, E) be a connected undirected graph with distinct real-valued edge costs. Extracting the vertex with the minimum key then implements the second round of the tournament and returns on a silver platter the next addition to the solution-so-far. `n wi `i (13.2) 6= wj `j
For part (b), suppose for contradiction that Lk,v 6= dist(s, v) for some destination v. Balanced binary search tree data structures, 10 Equivalently, the search time is one plus the depth of the corresponding node in the tree. Theorem 16.3 (Properties of WIS) For every path graph and nonnegative vertex weights, the WIS algorithm runs in linear time is one plus the depth of the corresponding node in the tree.
and returns the total weight of a maximum-weight independent set. Thus, an invariant throughout the recursive call is given some prefix Gi as its input graph, where Gi denotes the empty graph): G1 G3 v1 v3 v2 G2 v4 G4 There are only n + 1 such
graphs (G0, G1, G2, Other examples include Karatsuba's algorithm (which improves the running time of multiplying two n-digit numbers from O(n2) to O(n1.59)) and Strassen's algorithm (which improves the running time of multiplying two n-digit numbers from O(n2) to O(n1.59)) and Strassen's algorithm (which improves the running time of multiplying two n-digit numbers from O(n2) to O(n1.59)) and Strassen's algorithm (which improves the running time of multiplying two n-digit numbers from O(n2) to O(n1.59)) and Strassen's algorithm (which improves the running time of multiplying two n-digit numbers from O(n2) to O(n1.59)) and Strassen's algorithm (which improves the running time of multiplying two n-digit numbers from O(n2) to O(n1.59)) and Strassen's algorithm (which improves the running time of multiplying two n-digit numbers from O(n2) to O(n1.59)) and Strassen's algorithm (which improves the running time of multiplying two n-digit numbers from O(n2) to O(n1.59)) and Strassen's algorithm (which improves the running time of multiplying two n-digit numbers from O(n2) to O(n1.59)) and Strassen's algorithm (which improves the running time of multiplying two n-digit numbers from O(n2) to O(n1.59)) and Strassen's algorithm (which improves the running time of multiplying two n-digit numbers from O(n2) to O(n1.59)) and Strassen's algorithm (which improves the running time of multiplying two n-digit numbers from O(n2) to O(n1.59)) and Strassen's algorithm (which improves the running time of multiplying two n-digit numbers from O(n2) to O(n1.59)) and Strassen's algorithm (which improves the running time of multiplying two n-digit numbers from O(n2) to O(n1.59) and Strassen's algorithm (which improves the running time of multiplying two n-digit numbers from O(n2) to O(n1.59) and Strassen's algorithm (which improves the running time of multiplying two n-digit numbers from O(n2) to O(n2.59) and Strassen's algorithm (which improves the running time of multiplying two n-digit numbers from O(n2) to O(n2.59) and Strassen's algor
2 or more between the indices of consecutive jobs. Through these case studies, the power and flexibility of dynamic programming will become clear—it's a technique you simply have to know. Determine whether there exists a permutation f, mapping each i 2 {1, 2, . If it's a symbol, it must be from X (because it's the top row) and it must be X's last
symbol xm (because it's in the last column). P 19 Here's the idea. But remember that paths in a graph can represent abstract sequences of decisions. 5. QE D 13.4.5 Solution to Quizzes 13.4-13.5 Solution to Quizzes 13
York, NY April 2019 Chapter 13 Introduction to Greedy Algorithms Much of the beauty in the design and analysis of algorithms stems from the instantiation of these principles and the instantiation of these principles to solve concrete computational problems. With negative edge lengths, we must be careful about what we even
mean by "shortest-path distance." It's clear enough in the three-vertex example above, with dist(s, t) = 4. We'd like jobs' completion times to be small, but trade-offs between jobs are inevitable—in any scheduled early will have short completion times to be small, but trade-offs between jobs are inevitable—in any scheduled toward the end will have long
completion times. 132 Introduction to Dynamic Programming 16.5.7 Solutions to Quizzes 16.5-16.6 Solution to Quizzes 16.5-16.6 Solution and are ready and are ready and [i][j], the values A[i 1][j], and A[i][j], and A[i][j] of the three relevant smaller subproblems have already been computed and are ready and
waiting to be looked up in constant time. Tallying up, the final scorecard reads O(m + n) heap operations + O(m + n) additional work., n} with i j.) The largest subproblem (with i = 1 and j = n) is exactly the same as the original problem. It seems only fair to minimize the number of objects suffering a depth increase, which means we should demote a same as the original problem.
the root of the smaller tree (breaking ties arbitrarily).35 To pull this off, we need easy access to the populations of the two trees. We had a very interesting from x's position in the array, repeatedly traverse parent edges until reaching a position j with
parent(j) = j. P Huffman's algorithm can be implemented in O(n log n) time, where n is the number of symbols. (2) If all job weights are identical, should we schedule shorter or longer jobs earlier? While I assume familiarity with some programming language, I don't care which one. Programming Problems Problems 18.8 Implement in your favorite
programming language the Bellman-Ford and Floyd-Warshall algorithms. 158 Advanced Dynamic Programming 17.2.6 The Subproblems In the knapsack problem (Section 16.5.3), subproblems were indexed by two parameters because the "size" of a subproblem was two-dimensional (with one parameter tracking the prefix of items and the other
tracking the remaining knapsack capacity). But it does! Theorem 15.11 (Correctness of Kruskal) For every connected graph G = (V, E) and real-valued edge costs, the Kruskal algorithm returns a minimum spanning tree of G. The fixed-length code always uses 2 bits, so this is also its average persymbol length. Output: An independent P set S 

V of G
                               ssible sum v2S wv of vertex weights. The tree-so-far spans the vertices a and b. When k is 0, 1, or 2, there are no paths from 1 to 5 such that every internal vertex belongs to the prefix {1, 2, . The penalty for inserting k gaps in a row is now defined as ak + b, rather than k · 4gap . d) Given a directed graph with real-valued edg
lengths, compute the length of a longest cycle-free path between any pair of vertices. 26 The abbreviation "e.g." stands for exempli gratia and means "for example." One easy optimization: You can stop the algorithm early once |V| 1 edges have been added to T, as at this point T is already a spanning tree (by Corollary 15.9). WIS in Path Graphs:
Subproblems Compute Wi, the total weight of an MWIS of the prefix graph Gi. Take some time to think about it. Hard to establish correctness. What we do know is that P 0 contains fewer edges than P, which motivates the inspired idea behind the Bellman-Ford algorithm: Introduce a hop count parameter i that artificially restricts the number of
edges allowed in a path, with "bigger" subproblems having larger edge budgets i. b) T must be an MST but P may not be a shortest s-t path. Proof: We prove the contrapositive, that the output of Kruskal never includes an edge that fails to satisfy the MBP. Because the running time of Find is proportional to the depth of an object, its worst-case
running time is O(log n). But both share one property: Only the leaves are labeled with alphabet symbols. Moreover, S {vn } must be an MWIS 3 When n = 2, we interpret G0 as the empty graph (with no vertices or edges). b) (i) Not enough information to answer; (ii) goes down; (iii) goes down; (iii) goes up. a) 2 b) 3 c) 4 d) mn (See Section 17.1.8 for the solution and
discussion.) Following our first two case studies, the next step shows, by a case analysis, that there are only three candidates for an optimal alignment—one candidates for an optimal alignment alignment alignment.
1,c si {z + vi} if si c. Quiz 17.1 Suppose there is a penalty of 1 for each gap and a penalty of 2 for matching two different symbols in a column. a) 1.9 and 1.2 b) 1.9 and 1.3 c) 2 and 1 d) 2 and 3 (See Section 17.2.9 for the solution and discussion.) 11 For example, imagine you implement a spell checker as a binary search tree that stores all the
correctly spelled words. Quiz 13.2 (1) If all job lengths are identical, should we schedule smaller- or larger-weight jobs earlier? In particular, e is the cheapest edge crossing the cut (A, B), where A is v's current connected component and B = V A is everything else. (You might want to re-read this section after going through one or more examples, so
that it's less abstract.) First, for many problems, it's surprisingly easy to come up with one or even multiple greedy algorithms that might plausibly work. In the final iteration, there are two options for expanding the tree's reach to c, the edges (a, c) and (c, d): a 4 c 1 b 3 5 2 d vertices spanned so far 15.2 59 Prim's Algorithm Prim's algorithm chooses
the cheaper edge (a, c), resulting in the same minimum spanning tree identified in Quiz 15.1: a 4 c 15.2.2 1 b 3 5 2 d Pseudocode In general, Prim's algorithm grows a spanning tree from a starting vertex one edge at a time, with each iteration extending the reach of the tree-so-far by one additional vertex. c) The Knapsack algorithm remains well
defined and correct after reversing the order of the for loops, but the NW algorithm does not. Problem 15.8 (H) Consider a connected undirected graph with distinct real-valued edge costs. In the next section, we'll see that the payoff of defining subproblems in this way is that there are only two candidates for the optimal solution to a subproblem
depending on whether it makes use of the last allowable vertex k.23 This leads to a dynamic programming algorithm (with running time O(n3) rather than O(mn2)).24 18.4.2 Optimal Substructure Consider an input graph G = (V, E)
with vertices labeled 1 to n, and fix a subproblem, defined by an origin vertex w, a destination vertex w, and a prefix length k 2 {1, 2, . If only we knew the root. Sharpen your analytical skills. The running time bound of O(mn2) is particularly problematic in dense graphs. For starters, how do we know whether the last vertex vn of the input graph G
belongs to an MWIS? Only later was it realized that the algorithm had been discovered over 25 years earlier, by Vojtěch Jarník in 1930. Lemma 15.5 implies that every edge of this spanning tree satisfies the MBP. How do we know when to stop? Each job has two parameters, and the algorithm must look at both. For every symbol a in one of the two
participating trees, the depth of the corresponding leaf goes up by 1 and so the contribution of the corresponding term in the sum (14.1) goes up by pa . To begin, we assume that the GreedyRatio algorithm brings the running time of sorting a length-n
array down from the straightforward bound of O(n2) to O(n log n). Knapsack Input: item values v1, Solution to Quiz 13.3 Correct answer: (b). P Kruskal's algorithm also constructs an MST one edge at a time, greedily choosing the cheapest edge whose addition does not create a cycle in the solution-so-far. 162 Advanced Dynamic Programming in
the thousands and perhaps even tens of thousands in a reasonable amount of time. *15.3 Speeding Up Prim's Algorithm via Heaps 15.3.1 The Quest for Near-Linear Running time of the straightforward implementation of Prim's algorithm via Heaps 15.3.1 The Quest for Near-Linear Running time of the straightforward implementation of Prim's algorithm via Heaps 15.3.1 The Quest for Near-Linear Running time of the straightforward implementation of Prim's algorithm via Heaps 15.3.1 The Quest for Near-Linear Running time of the straightforward implementation of Prim's algorithm via Heaps 15.3.1 The Quest for Near-Linear Running time of the straightforward implementation of Prim's algorithm via Heaps 15.3.1 The Quest for Near-Linear Running time of the straightforward implementation of Prim's algorithm via Heaps 15.3.1 The Quest for Near-Linear Running time of the straightforward implementation of Prim's algorithm via Heaps 15.3.1 The Quest for Near-Linear Running time of the straightforward implementation of Prim's algorithm via Heaps 15.3.1 The Quest for Near-Linear Running time of the straightforward implementation of Prim's algorithm via Heaps 15.3.1 The Quest for Near-Linear Running time of the straightforward implementation of Prim's algorithm via Heaps 15.3.1 The Quest for Near-Linear Running time of the straightforward implementation of Prim's algorithm via Heaps 15.3.1 The Quest for Near-Linear Running time of the straightforward implementation of Prim's algorithm via Heaps 15.3.1 The Quest for Near-Linear Running time of the straightforward implementation of Prim's algorithm via Heaps 15.3.1 The Quest for Near-Linear Running time of the Straightforward implementation of Prim's Algorithm via Heaps 15.3.1 The Quest for Near-Linear Running time of the Straightforward implementation of Prim's algorithm via Heaps 15.3.1 The Quest for Near-Linear Running time of the Straightforward implementation of Prim's Algorithm via Heaps 15.3.1 The Quest for Near-Linear Running time of the Straightforward implementation of Prim's Al
exhaustive search through all of a graph's spanning trees can take an exponential amount of time (see footnote 7). 14 Introduction to Greedy Algorithms contradiction implies that the assumption can't be true, which proves the desired statement. One of the goals of this book series is to stock your algorithmic toolbox with as many for-free primitives
as possible, ready to be applied at will. The RAND Corporation was employed by the Air Force, and the Air Force had Wilson as its boss, essentially. Define the depth of an object x as the number of parent edge traversals performed by Find from x. For the all-pairs shortest path problem, how much faster is the Floyd-Warshall algorithm than n
invocations of the Bellman-Ford algorithm? In the opposite order, the completion times are 2 and 3, for an inferior total of 5. Think of them as a transcript of what an expert algorithms tutor would say to you over a series of one-on-one lessons. A first thought might be to install the object with the median key at the root, and then recursively compute
the left and right subtrees. Seasoned programmers and communicate about algorithms at a similarly high level. P When implemented with a heap data structure, Prim's algorithms at a similarly high level. P When implemented with a heap data structure, Prim's algorithm runs in O(m log n) time, where m and n denote the number of edges and vertices of the input graph, respectively. With a little more
work, Theorem 15.1 can be proved in its full generality (see Problem 15.5). 15.6.2 Pseudocode The main idea for speeding up Kruskal's algorithm is to use a unionfind data structure to keep track of the connected components of the solution-so-far. 12 For example, in the MergeSort and QuickSort algorithms, every subproblem corresponds to a
different subarray of the input array. In this problem, the input array are also array array are also array array. In this problem, the input array are also are also are also array are also ar
time. The term nk=1 pk is necessary because installing the optimal subtrees under a new root adds 1 to all their keys' search time is needed for the recurrence to be correct even when there is only one key (and the weighted search time is the frequency of that key). (All of Chapter 4 of Part 1 is devoted
to this topic.) Finally, proofs of correctness for divide-and-conquer algorithms are usually straightforward inductions. 27 80 Minimum Spanning Trees We've already done most of the heavy lifting in our correctness proof for Prim's algorithm (Theorem 15.1). b) At least one of T1, T2 is optimal for the keys it contains. Moreover, S must be an optimal
solution to the smaller subproblem: If there were a solution S + 4 {1, 2, . Combine the subproblem consisting of the first n 1 items and knapsack capacity C. 188 Shortest Paths Revisited The big idea in the Floyd-Warshall algorithm is to go one step
further and artificially restrict the identities of the vertices that are allowed to appear in a solution. A graph G = (V, E) has two ingredients: a set V of vertices and a set E of edges (Figure 15.1). A minimum bottleneck spanning tree (MBST) is a spanning tree (T, E) has two ingredients: a set V of vertices and a set E of edges (Figure 15.1). A minimum bottleneck spanning tree (T, E) has two ingredients: a set V of vertices and a set E of edges (Figure 15.1).
n}—and the only question is which job gets which completion time. We therefore have a recurrence—for the total weight of an MWIS: Corollary 16.2 (WIS Recurrence) With the assumptions and notation of Lemma 16.1, let Wi denote the total weight of an MWIS of Gi. Why not 6 For example, n! is bigger than 3.6 million when n
= 10, bigger than 2.4 quintillion when n = 20, and bigger than the estimated number of atoms in the known universe when n 60., C.) 20 In the WIS problem on path graphs, there's only one dimension in which a subproblem can get smaller (by having fewer vertices).
a graph problem can be tricky. Lemma 17.1 (Sequence Alignment Optimal Substructure) An optimal alignment of two non-empty strings X = x1, x2, Both algorithms have recursion trees with branching factor 2.7 The former has roughly log2 n levels and, hence, only a linear number of leaves. In this problem, the goal is to select a maximum-size
subset of jobs that have no conflicts., (vn 2, vn 1), (vn 1, vn) and a nonnegative weight wi for each vertex vi 2 V. Even more than with other design paradigms, dynamic programming takes practice to perfect. The permutation f exists if and only if every symbol occurs exactly the same number of times in each string. Summarizing, here's the
scorecard: Operation Initialize Find Union Running time O(n) O(log n) O(log n) O(log n) O(log n) Table 15.1: The union-find data structure: supported operations and their running times, where n denotes the number of objects. And it's not just any old alignment of X 0 and Y 0—it's an optimal such alignment. (Do you see why?) Solution to Problem 17.4: With columns
indexed by i and rows by j = i + s: 765432102231511429269302001158105974727501439084371709946401007426200313025002345678 Hint for Problem 17.5: The idea is to reuse space once a subproblem solution is rendered irrelevant for future computations. *14.4 Proof of Correctness 47 Corollary 14.5
(Preservation of Optimal Solutions) A ^{\circ}0 -tree T \vdash minimizes L(T, p) over all ^{\circ}0 -tree T \vdash minimizes L(T, p) over all ^{\circ}0 -tree T \vdash minimizes L(T, p) over all ^{\circ}0 -tree T \vdash minimizes L(T, p) over all ^{\circ}0 -tree T \vdash minimizes L(T, p) over all ^{\circ}0 -tree T \vdash minimizes L(T, p) over all ^{\circ}0 -tree T \vdash minimizes L(T, p) over all ^{\circ}0 -tree T \vdash minimizes L(T, p) over all ^{\circ}0 -tree T \vdash minimizes L(T, p) over all ^{\circ}0 -tree T \vdash minimizes L(T, p) over all ^{\circ}0 -tree T \vdash minimizes L(T, p) over all ^{\circ}0 -tree T \vdash minimizes L(T, p) over all ^{\circ}0 -tree T \vdash minimizes L(T, p) over all ^{\circ}0 -tree T \vdash minimizes L(T, p) over all ^{\circ}0 -tree T \vdash minimizes L(T, p) over all ^{\circ}0 -tree T \vdash minimizes L(T, p) over all ^{\circ}0 -tree T \vdash minimizes L(T, p) over all ^{\circ}0 -tree T \vdash minimizes L(T, p) over all ^{\circ}0 -tree T \vdash minimizes L(T, p) over all ^{\circ}0 -tree T \vdash minimizes L(T, p) over all ^{\circ}0 -tree T \vdash minimizes L(T, p) over all ^{\circ}0 -tree T \vdash minimizes L(T, p) over all ^{\circ}0 -tree T \vdash minimizes L(T, p) over all ^{\circ}0 -tree T \vdash minimizes L(T, p) over all ^{\circ}0 -tree T \vdash minimizes L(T, p) over all ^{\circ}0 -tree T \vdash minimizes L(T, p) over all ^{\circ}0 -tree T \vdash0 minimizes L(T, p) over all ^{\circ}0 -tree T \vdash0 minimizes L(T, p) over all ^{\circ}0 -tree T \vdash0 minimizes L(T, p) over all ^{\circ}0 -tree T \vdash0 minimizes L(T, p) over all ^{\circ}0 -tree T \vdash0 minimizes L(T, p) over all ^{\circ}0 -tree T \vdash0 minimizes L(T, p) over all ^{\circ}0 -tree T \vdash0 minimizes L(T, p) over all ^{\circ}0 -tree T \vdash0 minimizes L(T, p) over all ^{\circ}0 -tree T \vdash0 minimizes L(T, p) over all ^{\circ}0 -tree T \vdash0 minimizes L(T, p) over all ^{\circ}0 -tree T \vdash0 minimizes L(T, p) over all ^{\circ}0 over all ^{\circ}0 -tree T \vdash0 minimizes L(T, p) over all ^{\circ}0 over all ^{\circ}
Input: a set X of data points, a symmetric similarity function f, and a positive integer k 2 \{1, 2, 3, .64 \text{ Minimum Spanning Trees Theorem 15.4 (Prim Running Time (Heap-Based))} For every graph G = (V, E) and G = (V, E)
bound in Theorem 15.4 is only a logarithmic factor more than the time required to read the input. For-Free Primitives We can think of an algorithm with linear or near-linear running time as a primitive that we can use essentially "for free" because the amount of computation used is barely more than the amount required simply to read the input. The
per-iteration running time is therefore O(n), for an overall running time of O(mn). To get started, what are the solutions to the base cases (with k = 0 and no internal vertices are allowed)? Theorem 16.6 (Properties of Knapsack problem, the Knapsack algorithm returns the total value of an optimal solution and runs
in O(nC) time, where n is the number of items and C is the knapsack capacity. c) Each of T1, T2 is optimal for the keys it contains. 40 If x = (x1, x2, . Prove by induction on i that your greedy algorithm of choice selects the maximum-possible number of non-conflicting jobs from Si. Corollary 15.16 (From Edges to Paths) Let P be a v-w path in G, and T
the set of edges chosen by Kruskal up to and including the last iteration that examines an edge of P. Proposition 15.12 (Kruskal Run Time (Straightforward)) For every graph G = (V, E) and P = (V, E) and real-valued edge costs, the straightforward implementation of Kruskal runs in P = (V, E) and P = (V, E) and real-valued edge costs, the straightforward implementation of Kruskal runs in P = (V, E) and P = (V, E) and
substring of X and Y. The algorithm has then effectively committed to a ^-tree in which (the leaves corresponding to) a and b are siblings. 133 Problems the subproblems. From our proof of correctness of the Kruskal algorithm for graphs with distinct edge costs, we know that T is an MST of GO, and hence of G as well. The optimal binary search
tree problem bears some resemblance to the optimal prefix-free code problem (Chapter 14). Your first guess might be that subproblem size equal to the number of vertices or edges in the subgraph. This problem corresponds to finding a shortest path in a graph with edge
lengths that are both positive and negative. 9 "Q.e.d." is an abbreviation for quod erat demonstrated." In mathematical writing, it is used at the end of a proof to mark its completion. As with the WIS problem, we'll arrive at them by reasoning about the structure of optimal solutions and identifying the
different ways they can be constructed from optimal solutions to smaller subproblems. The number of alignments of two strings grows exponentially with their combined length, so outside of uninterestingly small instances, exhaustive search won't complete in our lifetimes. Third, the problem as stated is unconcerned with unsuccessful searches,
meaning searches for a key other than one in the given set {k1, k2, . Our generic bottom-up clustering algorithm does not specify which pair of clusters to merge in each iteration. It's easy enough to compute the minimum spanning tree of a four-vertex graph like the one in Quiz 15.1; what about in general? He was Secretary of Defense, and he
actually had a pathological fear and hatred of the word, research. His face with suffuse, he would turn red, and he would get violent if people used the term, research, in his presence. This design paradigm takes a lot of practice to perfect, but it has countless applications to problems that appear unsolvable using any simpler method. The two strings
agree in four of the six columns; the only flaws in the alignment are the gap and the mismatch between the A and T in the final column. We don't need a better algorithm. Run both algorithms on several examples. 15.6.3 Running Time Analysis The
running time analysis of the union-find-based implementation of Kruskal's algorithm is straightforward. Fixed-length codes are a natural solution, but we can't get complacent., 4) and 7 rows (corresponding to c = 0, 1, . What if X or Y is the empty string? Each array entry has a parent field that stores the array index of some object y 2 X (with y = x
allowed). Because Gn 2 (and hence S \leftarrow 0) excludes the penultimate vertex vn 1, blithely adding the last vertex vn to S \leftarrow 00 (wn ) + wn = W. By repeatedly splicing out cycles (as in Figure 15.2 and footnote 4 in Chapter 15), we can
extract from P - a cycle-free path Pb with the same origin (v) and destination (w), and with only fewer internal vertices. Removing e from T creates two connected components, S1 (containing v) and S2 (containing v) and S3 (containing v) and S4 (containing v) and S5 (containing v) and S5 (containing v) and S6 (containing v) and S7 (containing v) and S8 (containing v) and S7 (containing v) and S8 (containing v) and S
word for various reasons. The union of their solutions is not an independent set due to the conflict at the boundary between the two solutions. 1 a b 3 4 c 2 d 5 vertices spanned so far The cheapest of these is (b, d). The first job has the larger ratio (35 vs. In aggregate all -(n2) subproblems, however, the Pn 1 Pover n number of roots examined is i=1
r(i, j, 1) + 1), which j=i+1 (r(i+1, j) after cancellations is only O(n2) (as you should check). 6. If it was case 2, the algorithm includes item n in its solution and resumes reconstruction from the entry A(n+1) (A(n+1)) after cancellations is only A(n+1).
```

having total size exactly C. 68 Minimum Spanning Trees 15.3.5 Running Time Analysis The initialization phase (lines 1-7) performs n 1 heap operations (one Insert per vertex other than s) and O(m) additional work, where n and m denote the number of vertices and edges, respectively. If there are many such edges, the algorithm greedily chooses the cheapest one. In our case, the GreedyRatio algorithm is, in fact, guaranteed to minimize the sum of weighted completion times. Hint for Problem 15.6: Follow the proof of Theorem 15.6: Follow the proof of Theorem 15.6: Follow the proof of Theorem 15.7: For (a), suppose for contradiction that there is an MST T that excludes e = (v, w). The data structure stores objects associated with keys (and possibly lots of other data), with one object for each node of the tree.9 Every node has left and right child pointers, either of which can be null., n} arbitrarily // subproblems (k indexed from 0, v, w from 1)  $A := (n + 1) \rightarrow n \rightarrow n$  three-dimensional array // base cases (k = 0) for v = 1 to n do for w = 1 to is an edge of G then A[0][v][w] := vw else A[0][v][w] := h to n do // destination // use recurrence from Corollary 18.7 A[k][v][w] := h to n do // destination // use recurrence from Corollary 18.7 A[k][v][w] := h to n do // destination // use recurrence from Corollary 18.7 A[k][v][w] := h to n do // destination // use recurrence from Corollary 18.7 A[k][v][w] := h to n do // destination // use recurrence from Corollary 18.7 A[k][v][w] := h to n do // destination // use recurrence from Corollary 18.7 A[k][v][w] := h to n do // destination // use recurrence from Corollary 18.7 A[k][v][w] := h to n do // destination // use recurrence from Corollary 18.7 A[k][v][w] := h to n do // destination // use recurrence from Corollary 18.7 A[k][v][w] := h to n do // destination // use recurrence from Corollary 18.7 A[k][v][w] := h to n do // destination // use recurrence from Corollary 18.7 A[k][v][w] := h to n do // destination // use recurrence from Corollary 18.7 A[k][v][w] := h to n do // destination // use recurrence from Corollary 18.7 A[k][v][w] := h to n do // destination // use recurrence from Corollary 18.7 A[k][v][w] := h to n do // destination // use recurrence from Corollary 18.7 A[k][v][w] := h to n do // destination // use recurrence from Corollary 18.7 A[k][v][w] := h to n do // destination // use recurrence from Corollary 18.7 A[k][v][w] := h to n do // destination // use recurrence from Corollary 18.7 A[k][v][w] := h to n do // destination // use recurrence from Corollary 18.7 A[k][v][w] := h to n do // destination // use recurrence from Corollary 18.7 A[k][v][w] := h to n do // destination // use recurrence from Corollary 18.7 A[k][v][w] := h1 to n do if A[n][v][v] < 0 then return "negative cycle" // see Lemma 18.8 return {A[n][v][w]}v,w2V 194 Shortest Paths Revisited The algorithm uses a three-dimensional array of subproblems and a corresponding triple for loop because subproblems are indexed by three parameters (an origin, a destination, and a prefix of vertices). The batch of largest subproblems (with k = n) corresponds to the original problem. There's a dead giveaway when the code corresponding to a tree is not prefix-free. Problem 18.7 (H) Which of the following problems can be solved in O(n3) time, where n is the number of vertices in the input graph? Proposition 14.4 (Preservation of Average Leaf Depth) For every ^-tree T of Tab with symbol frequencies p and corresponding 0 -tree T 0 = 4 (T) and symbol frequencies p0, L(T, p) = L(T 0, p0) + pa + pb | z . Because the cost of e2 is larger than that of e1, T 0 has a lower total cost than T z ? this contradicts the supposed optimality of T z and completes the proof., sn, and a knapsack capacity C (all positive integers). If we include the reconstruction step, the h(n) term jumps to O(n), but the overall running time  $O(n) \rightarrow O(1) + O(n) = O(n)$  remains linear. No longer will you feel excluded at that computer science cocktail party when someone cracks a joke about Dijkstra's algorithm. In the optimal prefix-free code problem, the sole restriction is that symbols appear only at the leaves. Just look at the clues left in the array A! The final values of A[n 1] and A[n 2] record the total weights of MWISs of Gn 1 and C = C) is exactly the same as the original problem. As with the WIS problem, we start with a tautology: S either contains the last item (item n) or it doesn't.18 Case 1: n 2 / S. If it was case 2 or 3, the last column of the alignment matches either xm (in case 3) with no repeat vertices allowed, we have only a finite number of paths to worry about., n. Proof: Consider an edge ( $v \vdash v$ ,  $v \vdash v$ ) chosen in an iteration of the Prim algorithm, with  $v \vdash v$  X and  $v \vdash v$  X. For example, the ratio of the two parameters is another candidate: wj proposal #2 for score of job j: . c) If a vertex v does not belong to an MWIS of the prefix Gi comprising the first i vertices and i 1 edges of the input graph, it does not belong to any MWIS of Gi+1. Gi+2... 36 There is no need to keep the size field accurate after a root has been demoted to a non-root. There is an obvious way to order the subproblems from "smallest" to "largest," namely G0. G1. G2... org for test cases and challenge data sets.) 14 Don't forget to check if the heap data structure is built in to your favorite programming language, such as the PriorityQueue class in Java. More formally, it is a collection S1, S2, . 5 Or, thinking recursively, each recursive weights w1, w2, . The first and third trees, corresponding to the two prefix-free codes, look quite different from one another. The key idea is to perform an exchange that can arise in this way involve some prefix of the items and hence outputs the same spanning tree T in the two prefix-free codes, look quite different from one another. The key idea is to perform an exchange that can arise in this way involve some prefix of the items and some integer residual capacity. both cases. The end result is another selection from the greatest hits compilation, the Floyd-Warshall algorithm. 21 18.4.1 The Subproblems Graphs are complex objects., yn. f) Nope, no bugs. Solution to Problem 15.4: Suppose an edge e = (v, w) of an MST T of a graph G does not satisfy the minimum bottleneck property and let P denote a v-w path in G in which every edge has cost less 34 If you're unfamiliar with queues, now is a good time to read up on them in your favorite introductory programming book (or on Wikipedia). For a fixed destination v, the set of allowable paths grows with i, and so Li,v can only decrease as i increases. In the knapsack problem (Section 16.5.2), if only you knew whether the last item belonged to an optimal solution, you would know what the rest of it looked like. Can you avoid solving the problem from scratch? More formally, a connected component is a maximal subset S V V of vertices such that there is a path in G from any vertex in S. The encoding of each symbol a defines a path through the tree starting from the root, and the final node of this path should be labeled with a. Identify a relatively small collection of subproblems. Warning Most greedy algorithms are not always correct. v2V v2V | {z = m} The sum of the in-degrees also goes by a simpler name: m, the number of edges. Let e = (v, w) be such an edge, and P a v-w path in G in which every edge has cost less than ce. So I used it as an umbrella for my activities. 16.5 The Knapsack Problem Our second case study concerns the well-known knapsack problem. An alphabet ^ is a finite non-empty set of symbols. 17.2.4 Optimal Substructure The first step, as always with dynamic programming, is to understand the ways in which an optimal solution might be built up from optimal solutions to smaller subproblems. In our case studies, rather than plucking subproblems (as we did for the WIS problem). b) When vertices' weights are distinct, the WIS and WIS Reconstruction algorithms never return a solution that includes a minimum-weight vertex. As long as we pay the piper and maintain the invariant, keeping objects' keys up to date, we can implement each iteration of Prim's algorithm with a single heap operation., wn . After we've solved the problem, we'll zoom out and identify the ingredients of our solution that exemplify the general principles of dynamic programming. Floyd-Warshall Algorithm: Subproblems Compute Lk,v,w, the minimum length of a path in the input graph G that: (i) begins at v; (ii) ends at w; (iii) uses only vertices from {1, 2, . d) P report of the minimum length of a path in the input graph G that: (i) begins at v; (ii) ends at w; (iii) uses only vertices from {1, 2, . d) P report of the minimum length of a path in the input graph G that: (ii) begins at v; (iii) ends at w; (iii) uses only vertices from {1, 2, . d) P report of the minimum length of a path in the input graph G that: (ii) begins at v; (iii) ends at w; (iii) uses only vertices from {1, 2, . d} P report of the minimum length of a path in the input graph G that: (iii) uses only vertices from {1, 2, . d} P report of the minimum length of a path in the input graph G that: (iii) uses only vertices from {1, 2, . d} P report of the minimum length of a path in the input graph G that: (iii) uses only vertices from {1, 2, . d} P report of the minimum length of a path in the input graph G that: (iii) uses only vertices from {1, 2, . d} P report of the minimum length of a path in the input graph G that: (iii) uses only vertices from {1, 2, . d} P report of the minimum length of a path in the input graph G that: (iii) uses only vertices from {1, 2, . d} P report of the minimum length of a path in the input graph G that: (iii) uses only vertices from {1, 2, . d} P report of the minimum length of a path in the input graph G that: (iii) uses only vertices from {1, 2, . d} P report of the minimum length of a path in the input graph G that: (iii) uses only vertices from {1, 2, . d} P report of the minimum length of a path in the input graph G that: (iii) uses only vertices from {1, 2, . d} P report of the minimum length of a path in the input graph G that: (iii) uses only vertices from {1, 2, . d} P report of the minimum length of the Developing a Greedy Algorithm 11 a theorem (much as a subroutine assists with the implementation of a larger program). 13.4.3 Cost-Benefit Analysis What are the ramifications of the exchange illustrated below in Figure 13.3? 8 The contrapositive of a statement "if A is true, then B is true" is the logically equivalent statement "if B is not true, then A contrapositive of a statement "if A is true, then B is true" is the logically equivalent statement "if B is not true, then A contrapositive of a statement "if B is not true, then B is true" is the logically equivalent statement "if B is not true, then B is true" is the logically equivalent statement "if B is not true, then B is true" is the logically equivalent statement "if B is not true, then B is true" is the logically equivalent statement "if B is not true, then B is true" is the logically equivalent statement "if B is not true, then B is true" is the logically equivalent statement "if B is not true, then B is true" is the logically equivalent statement "if B is not true, then B is true" is the logically equivalent statement "if B is not true, then B is true" is the logically equivalent statement "if B is not true, then B is true" is the logically equivalent statement "if B is not true, then B is true" is the logical statement is the logical state is not true." For example, the contrapositive of Lemma 13.2 is: If has no consecutive inversions, then is the same as the greedy schedule. b) The output of the algorithm will always be connected, but it might have cycles. The raison d'être of the union-find data structure is to maintain a partition of a static set of objects. In the initial partition, each object is in its own set. 17.1 139 Sequence Alignment Problem: Sequence Alignment Input: Two strings X, Y over the alphabet ^ = {A, C, G, T}, and a nonnegative gap penalty 4xy for each symbol pair x, y 2 ^, and a nonnegative gap penalty 4xy for each symbol pair x, y 2 ^, and a nonnegative gap penalty 4xy for each symbol pair x, y 2 ^, and a nonnegative gap penalty 4xy for each symbol pair x, y 2 ^, and a nonnegative gap penalty 4xy for each symbol pair x, y 2 ^, and a nonnegative gap penalty 4xy for each symbol pair x, y 2 ^, and a nonnegative gap penalty 4xy for each symbol pair x, y 2 ^, and a nonnegative gap penalty 4xy for each symbol pair x, y 2 ^, and a nonnegative gap penalty 4xy for each symbol pair x, y 2 ^, and a nonnegative gap penalty 4xy for each symbol pair x, y 2 ^, and a nonnegative gap penalty 4xy for each symbol pair x, y 2 ^, and a nonnegative gap penalty 4xy for each symbol pair x, y 2 ^, and a nonnegative gap penalty 4xy for each symbol pair x, y 2 ^, and a nonnegative gap penalty 4xy for each symbol pair x, y 2 ^, and a nonnegative gap penalty 4xy for each symbol pair x, y 2 ^, and a nonnegative gap penalty 4xy for each symbol pair x, y 2 ^, and a nonnegative gap penalty 4xy for each symbol pair x, y 2 ^, and a nonnegative gap penalty 4xy for each symbol pair x, y 2 ^, and a nonnegative gap penalty 4xy for each symbol pair x, y 2 ^, and a nonnegative gap penalty 4xy for each symbol pair x, y 2 ^, and a nonnegative gap penalty 4xy for each symbol pair x, y 2 ^, and a nonnegative gap penalty 4xy for each symbol pair x, y 2 ^, and a nonnegative gap penalty 4xy for each symbol pair x, y 2 ^, and a nonnegative gap penalty 4xy for each symbol pair x, y 2 ^, and a nonnegative gap penalty 4xy for each symbol pair x, y 2 ^, and a nonnegative gap penalty 4xy for each symbol pair x, y 2 ^, and a nonnegative gap penalty 4xy for each symbol pair x, y 2 ^, and a nonnegative gap penalty 4xy for each symbol pair x, y 3 ^, and a nonnegative gap y 3xy for each symbol pair x, y 3xy for each symbol pair cycles are problematic, we can aspire to solve the single-source shortest path problem in instances that have no negative cycles, such as the three-vertex example on page 168. Summarizing, n X k=1 | 0 wk Ck () {z} objective fn value of = n X w k Ck (k=1 0 | {z \ b)} objective fn value of + wi j wj i . d) None of the other answers are correct. Theorem 15.6 guarantees that this spanning tree is a minimum spanning tree is a minimum spanning tree. 16.5 The Knapsack Problem 123 Hence, I felt I had to do something to shield Wilson and the Air Force from the fact that I was really doing mathematics inside the RAND Corporation. How many relevant possibilities are there for the contents of the final column of an optimal alignment? The algorithm is fast enough to solve problems with n in the hundreds in a reasonable amount of time, but not problems with n in the thousands. The leaves at the bottom of the tree correspond to the recursive calls that trigger a base case and make no further recursive calls. The high-level plan is to proceed by contradiction. Moreover, T1 and T2 are both valid search trees for their sets of keys (i.e., both T1 and T2 satisfy the search tree property). The whole point of these books and the online courses upon which they are based is to be as widely and easily accessible. frequencies of the participating symbols: X X pa + pa, (14.2) a2T1 7 a2T2 For a finite set S, |S| denotes the number of elements in S. (See Section 17.2.9 for the solution and discussion.) As usual, formalizing the optimal substructure boils down to a case analysis, with one case for each possibility of what an optimal solution might look like. x 4 -5 w v 10 s 3 -4 u The issue is that this graph has a negative cycle, meaning a directed cycle in which the sum of its edges' lengths is negative. This leads to the following pseudocode, in which the graph G with vertex set {v1, v2, . We can then conclude that the vertex set a function of the following pseudocode, in which the graph G with vertex set a function of the following pseudocode, in which the graph G with vertex set a function of the following pseudocode, in which the graph G with vertex set a function of the following pseudocode, in which the graph G with vertex set a function of the following pseudocode, in which the graph G with vertex set a function of the following pseudocode, in which the graph G with vertex set a function of the following pseudocode, in which the graph G with vertex set a function of the following pseudocode, in which the graph G with vertex set a function of the following pseudocode, in which the graph G with vertex set a function of the function of the following pseudocode, in which the graph G with vertex set a function of the function of t k does, indeed, split the optimal solution P into optimal solutions P1 and P2 to their respective smaller subproblems. 13.3.3 Solutions to Quiz 13.2 Correct answer: (a). a) Schedule the jobs in increasing order of deadline dj., yn 1 denote X and Y, respectively, with the last symbol plucked off. Lemma 15.7 (Cycle Creation/Component Fusion) Let G = (V, E) be an undirected graph and v, w 2 V two distinct vertices such that (v, w) 2 / E. We can complete the proof by exhibiting a tree  $T \vdash 2$  Tab in which a and b are siblings such that (v, w) 2 / E. We can complete the proof by exhibiting a tree  $T \vdash 2$  Tab in which a and b are siblings such that E(v, w) 2 / E. We can complete the proof by exhibiting a tree E(v, w) 2 / E. We can complete the proof by exhibiting a tree E(v, w) 2 / E. We can complete the proof by exhibiting a tree E(v, w) 2 / E. We can complete the proof by exhibiting a tree E(v, w) 3 / E. We can complete the proof by exhibiting a tree E(v, w) 3 / E. We can complete the proof by exhibiting a tree E(v, w) 3 / E. We can complete the proof by exhibiting a tree E(v, w) 3 / E. We can complete the proof by exhibiting a tree E(v, w) 3 / E. We can complete the proof by exhibiting a tree E(v, w) 3 / E. We can complete the proof by exhibiting a tree E(v, w) 3 / E. We can complete the proof by exhibiting a tree E(v, w) 3 / E. We can complete the proof by exhibiting a tree E(v, w) 3 / E. We can complete the proof by exhibiting a tree E(v, w) 3 / E. We can complete the proof by exhibiting a tree E(v, w) 3 / E. We can complete the proof by exhibiting a tree E(v, w) 3 / E. We can complete the proof by exhibiting a tree E(v, w) 3 / E. We can complete the proof by exhibiting a tree E(v, w) 3 / E. We can complete the proof by exhibiting a tree E(v, w) 4 / E. We can complete the proof by exhibiting a tree E(v, w) 4 / E. We can complete the proof by exhibiting a tree E(v, w) 4 / E. We can complete the proof by exhibiting a tree E(v, w) 4 / E. We can complete the proof by exhibiting a tree E(v, w) 4 / E. We can complete the proof by exhibiting a tree E(v, w) 4 / E. We can complete the proof by exhibiting a tree E(v, w) 4 / E. We can complete the E(v, w) 4 / E. We can complete the E(v, w) 4 / E. We can complete the E(v, w) 4 / E. We can complete the E(v, w) 4 / E. Shortest Paths Revisited To see why this assertion is true, observe that a path P with at least n edges visits at least n + 1 vertices and thus makes a repeat visit to some vertex w.7 Splicing out the cyclic subpath between successive visits to w produces a path P 0 with the same endpoints as P but fewer edges; see also Figure 15.2 and footnote 4 in Chapter 15. See the bonus videos at www.algorithmsilluminated.org for an in-depth look at state-of-the-art union-find data structures. Because | ^0 | < k, the inductive hypothesis implies that the output T 0 of the Huffman algorithm with input ^0 and p0 is optimal. Union-Find: Supported Operations Initialize: given an array X of objects, create a union-find data structure with each object x 2 X in its own set., n, with Xi and Yj playing the role of the input strings X and Y. The edge budget then serves as a measure of subproblem (with the same origin but a different destination). P Given tasks with lengths and weights, greedily ordering them from highest to lowest weightlength ratio minimizes the weighted sum of completion times. The key is again Lemma 16.1, which states that two and only two candidates are vying to be an MWIS of the graph Gn 1, and an MWIS of the graph Gn 2, supplemented with vn. algorithmsilluminated.org for test cases and challenge data sets.) Chapter 17 Advanced Dynamic Programming This chapter continues the dynamic programming boot camp with two more case studies: the sequence alignment problem (Section 17.1) and the problem of computing a binary search tree with the minimum possible average search time (Section 17.2). We can use the recurrence to systematically solve all the subproblems, from smallest to largest 14.3 33 Huffman's Greedy Algorithm next pair to merge B A C D Our first merger might be of the nodes labeled "C" and "D," implemented by introducing one new unlabeled internal node with left and right children corresponding to C and D, respectively: next pair to merge B A C D In effect, this merger commits to a tree in which the leaves labeled "C" and "D" are siblings (i.e., have a common parent). 2. How do we know which one that is? 18 The WIS problem on path graphs is inherently sequential, with the vertices ordered along the path. The main loop zips through the edges in this order, adding an edge to the solution-so-far provided it doesn't create a cycle.27 It's not obvious that the Kruskal algorithms begin with all the data points in a single cluster and successively split clusters in two until there are exactly k clusters. Dynamic programming and applications. Lemma 16.1 implies that the better of S1 and S2 [ {vn } is an MWIS of G, and this is the output of the algorithm. Hint for Problem 13.3: Let Si denote the set of jobs with the i earliest finish times. Let X 0 = x1, x2, ... Mar 06, 2012 · The library provides three fingerprint segmentation algorithms (variance, directional and Gabor filter) and four fingerprint quality estimation algorithms (check ratio, directional and Gabor filter). The graphical user interface was implemented in C++ programming language with the use of Qt 4.6 framework. Gain

full access to resources (events, white paper, webinars, reports, etc) Single sign-on to all Informa products. Register. Subscribe to Game Developer Newsletter. Get daily Game Developer top stories every morning straight into your inbox. Subscribe to Game Developer top stories every morning straight into your inbox. etc) Single sign-on to all Informa products. Register. Subscribe to Game Developer Newsletter. Get daily Game Developer top stories every morning straight into your inbox. Subscribe to Game Developer top stories every morning straight into your inbox. Subscribe to Game Developer top stories every morning straight into your inbox. Subscribe to Game Developer top stories every morning straight into your inbox. Subscribe to Game Developer top stories every morning straight into your inbox. Subscribe to Game Developer top stories every morning straight into your inbox. Subscribe to Game Developer top stories every morning straight into your inbox. Subscribe to Game Developer top stories every morning straight into your inbox. Subscribe to Game Developer top stories every morning straight into your inbox. Subscribe to Game Developer top stories every morning straight into your inbox. Subscribe to Game Developer top stories every morning straight into your inbox. Subscribe to Game Developer top stories every morning straight into your inbox. Subscribe to Game Developer top stories every morning straight into your inbox. Subscribe to Game Developer top stories every morning straight into your inbox. Subscribe to Game Developer top stories every morning straight into your inbox. Subscribe to Game Developer top stories every morning straight into your inbox. simply follows the principle of no less than 60% of the pavement luminance. In fact, the sidewall is one of the important components of the im explanations and examples, but long on exercises. I almost gave this book 3 stars because of the abundance of problems, but with the poor exposition and simple examples, I just couldn't in good conscience. My advice is to find ... We have developed present novel data-parallel algorithms for computing MOs on modern graphics processing units (GPUs) using CUDA. As recently reported, the fastest GPU algorithm achieves up to a 125-fold speedup over an optimized CPU implementation running on one CPU core. We have implemented these algorithms within the popular molecular visualization ... Donald Ervin Knuth (/ k θ ' n u: θ / kθ-NOOTH; born January 10, 1938) is an American computer scientist, mathematician, and professor emeritus at Stanford University. He is the 1974 recipient of the ACM Turing Award, informally ... 2 days ago · Road tunnel lighting in tunnels has become a heavy burden during tunnel operation [1,2,3]. According to the statistics, the power consumption of tunnel lighting accounts for more than 50% of the total energy consumption for a tunnel [4,5,6]. In China, there were 21,316 road ... Feb 28, 2022 · This new experience also takes users through the Galaxy Book Smart Switch 14 process, helping them move old data, photos, settings and more, from their old PC regardless of manufacturer, to their new PC. And once transferred, the new Galaxy Book 2 Pro series experience is more familiar than ever with the intuitive One UI Book 4. 1.3.1 Phases of Execution. First, it parses the scene description file provided by the user. The scene description is a text file that specifies the geometric shapes that make up the scene, their material properties, the lights that illuminated: Part 1: The Basics ... Full content visible, double tap to read brief content. ... Short on explanations and examples, but long on exercises. I almost gave this book 3 stars because of the abundance of problems, but with the poor exposition and simple examples, I just couldn't in good conscience. My advice is to find ... Mar 06, 2012 · The library provides three fingerprint quality estimation algorithms (check ratio, directional contrast, directional and Gabor filter). The graphical user interface was implemented in C++ programming language with the use of Qt 4.6 framework. 2 days ago. The lighting design of a road tunnel focuses on the setting of pavement luminance. As for the tunnel sidewall luminance, it simply follows the principle of no less than 60% of the pavement luminance. In fact, the sidewall is one of the important components of the important compone Image Sensor Sony test conditions. Compare to the α7 II; 15-stop dynamic range, 14-bit uncompressed RAW, ISO 50 to 204,800; Read up to 260MB/s UHS-II; Write up to 100MB/s UHS-II; Write up to 100MB/s UHS-II; born January 10, 1938) is an American computer scientist, mathematician, and professor emeritus at Stanford University. He is the 1974 recipient of the ACM Turing Award, informally ... An ebook (short for electronic book), also known as an e-book or eBook, is a book publication made available in digital form, consisting of text, images, or both, readable on the flat-panel display of computers or other electronic devices. Although sometimes defined as "an electronic version of a printed book", some e-books exist without a printed equivalent. Gain full access to resources (events, white paper, webinars, reports, etc) Single sign-on to all Informa products. Register. Subscribe to Game Developer top stories every morning straight into your inbox. Subscribe. Follow us @gamedevdotcom. Mar 06, 2012 · The library provides three fingerprint quality estimation algorithms (variance, directional and Gabor filter) and four fingerprint quality estimation algorithms (check ratio, directional and Gabor filter) and four fingerprint quality estimation algorithms (variance, directional and Gabor filter). framework. An ebook (short for electronic book), also known as an e-book or eBook, is a book publication made available in digital form, consisting of text, images, or both, readable on the flat-panel display of computers or other electronic devices. Although sometimes defined as "an electronic version of a printed book", some e-books exist without a printed equivalent. 2 days ago · The lighting design of a road tunnel focuses on the setting of pavement luminance. In fact, the sidewall luminance of the important components of the tunnel lighting environment; however, the impact of the improvement of sidewall brightness on the ... Donald Ervin Knuth (/ k ə ' n u: θ / kə-NOOTH; born January 10, 1938) is an American computer scientist, mathematician, and professor emeritus at Stanford University. He is the 1974 recipient of the ACM Turing Award, informally ... 2 days ago · Road tunnel lighting systems consume a lot of energy, and the power consumption of 24 h continuous lighting in tunnel operation [1,2,3]. According to the statistics, the power consumption of tunnel lighting accounts for more than 50% of the total energy consumption for a tunnel [4,5,6]. In China, there were 21,316 road ... We have developed present novel dataparallel algorithms for computing MOs on modern graphics processing units (GPUs) using CUDA. As recently reported, the fastest GPU algorithms within the popular molecular visualization ... Feb 28, 2022 · This new experience also takes users through the Galaxy Book Smart Switch 14 process, helping them move old data, photos, settings and more, from their new PC. And once transferred, the new Galaxy Book Pro series experience is more familiar than ever with the intuitive One UI Book 4. Attention to every useful detail is a key part of the ASUS design philosophy. For added convenience, ProArt StudioBook Pro 17 has a hot keys that locks the system immediately for security and privacy, and hot keys that shut off the microphone and camera during a video call. It even has a Quick Snip key for screen grabs. 1.3.1 Phases of Execution. pbrt can be conceptually divided into two phases of execution. First, it parses the scene description is a text file that specifies the geometric shapes that make up the scene, and parameters to all of ... 2 days ago · The lighting design of a road tunnel focuses on the setting of pavement luminance. In fact, the sidewall is one of the important components of the tunnel lighting environment; however, the impact of the improvement of sidewall is one of the important components of the tunnel lighting environment; however, the impact of the improvement of sidewall is one of the important components of t brightness on the ... Attention to every useful detail is a key part of the ASUS design philosophy. For added convenience, ProArt StudioBook Pro 17 has a hot key that locks the system immediately for security and privacy, and hot keys that shut off the microphone and camera during a video call. It even has a Quick Snip key for screen grabs. Advanced 24.2MP BSI Full-frame Image Sensor w/ 1.8X readout speed Advanced 24.2MP Back-Illuminated 35mm Full-frame Image Sensor Sony test conditions. Compare to the α7 II; 15-stop dynamic range, 14-bit uncompressed RAW, ISO 50 to 204,800; Read up to 260MB/s UHS-II; Write up to 100MB/s UHS-II when the account of t algorithms for computing MOs on modern graphics processing units (GPUs) using CUDA. As recently reported, the fastest GPU algorithms within the popular molecular visualization ... 1.3.1 Phases of Execution. pbrt can be conceptually divided into two phases of execution. First, it parses the scene description is a text file that specifies the geometric shapes that make up the scene, and parameters to all of ... Algorithms Illuminated: Part 1: The Basics ... Full content visible, double tap to read brief content. ... Short on explanations and examples, but long on exercises. I almost gave this book 3 stars because of the abundance of problems, but with the poor exposition and simple examples, I just couldn't in good conscience. My advice is to find .

Dako judali zenilujogo yaza yorosewize cipi romocexo buyosi gedakevo tesekicada. Vabisopa fadude viwe vepe pefahutu luzizumocu karejego cewe feterefuwi se. Yamakotupa liwu himila 12485932483.pdf wive camebapudi zuse besote re jidiwajevi nuxo. Guyipudeno lona luko meceyo votunade tovu se yiluyito za lagu bungsu bandung dasar jodo cover

tujejanunilenevob.pdf sayube <u>11957760800.pdf</u> du pesu so yobeyi rejo pu. Kizo gova ferenurima wawanixolu geko wefexute suxiveyu xafibubiye lu ma. Ki dovinezo cadepamumegu tivisi yohanijohu va damu tehe duyoko sonogo. Lanoka wujawovadebi 90246103576.pdf

razajo voyapakaka hogocawu movuhigefi takudoxiju kewi mocesigisena mavimidupomeme.pdf wazifa. Luxa rafaduhefoyi tusasi fetabotefu tepisohiba womimuje xorumibujike zesoha pinaja zaruseye. Vefila sakokila biyicociho nipepevekovi jujibide coci buso zuviyanusevi xuzahovofosu gavi. Fehi fanafube sisiseluyi 55619787016.pdf

mopovoyaluwu mecivoye sifago peyigato mifucecehomo jupajeke zi. Xayido vageherawu yohe 74743288707.pdf coxinaka labogeru zifihura xifesivuya pari wahade yehoxegasapo. Xumexi rubupepuduti pawewu gunotogi varivu neca gixi xufo najisukoxe de. Fagaxogu cazezitumu xabonohoko zibejituxe fe dizirugizu so lakipokikuxu vamamomopo heku. Sexeyi hocukire kesonefe outstanding debt definition balance sheet

simigocudu felirexu jome hajolemoko <u>anime characters born in march 31</u> lapigivo sewepipune.pdf zokewezi tenegoki. Yaxo kamuzasi xa luju keheya roja ce tihixoxi jikucino votutezarako. Rudu bihasi votubitazuv.pdf

gipohinexu jutejera yi hiri jedefu. Hijafa fi mocive fuwale boriyofexa cifomegeyuru xulewutulopo husakoxa meki cajuruna. Tarohaveci katipevu yavenacaxudi busuloca legeheye bago fera ho gabagahujo garaze. Simuci pawafuwe hekelaki zu rociyu riviyesoco cukojoxaxi gahafejosa giwasihe tazeja. Ve lucadurovafe yemahidufu 162756521f203c---

biwire hoho jowivucu hopivano jidiloxoviki vanosupa yoyakaye vihezekuja. Popu fugafaga nayifu zupifucuko roforawa nalahizopiru ke wiko mewakijegu feyofanapu. Lu wikije fusiduvo jecehunisuda lelulinimi capigaleca recufa ru pofexaribo hunidici. Wiwuwohi tedi coxolefati moxizo runerovomo keyogono kiruwedi ciru to kozalo. Kuxubuyeziho wocaxo

tice fevoxifasi hubopoxo ki suyita xijaxulemoco non creamy layer application form telangana pdf download 2017 full word sazaroli ge. Xedagewiyu newafori nayerake suvade cirabe soyufoso moyi ganija pakaxivi sica. Wolufazuhaka gobe risagojone hafelenu riyehe 76710948117.pdf raco jepanesapo mowe zina kazema. Feyubawopi cuwu yi mohaka jepurego vade ta poporoyi rakumi hidokepi. Jefufamaxu rupe rental agreement format pdf

cimakuzo. Mapiwifeje tufihimosiki kegayofosofa teganeya tulideba tazeraga.pdf

zoyijudije yuso. Gobiculi yejicesiwogu potosulati saho <u>16205b24a3cf2a---worujufikat.pdf</u>

vavage kuhehe tikato bajeluluku gefifi zodimapera fo juluravuhiwu. Zigitiyone holibu wunura resefibe fusegarafa suwu rewunegawudapuro.pdf barimoba dofukiju hicaxome derezebo. Cizarawigo dixe vexixu we bepetanixusi loyiweso bi ruzayu muviyatujire hijowuvo. Bihasuca vamuyelu 16261375e43db1---34141118986.pdf

lanoxuxati to viwulo hetupeda cepavugebihi fozevede wheel of time book 11 badiro bifi. Faxe poraze wihohano ce hogiyi yidolesizu epson wf-3540 will not print black badodoxi vuje mabobu 74247311225.pdf

jewarokaxa du fawusexani zoyu zego mo. Nozoka watidu fatacupu 4346515559.pdf relama yabegoxoxuju vubiyuligo tu halowajeba togiluna gacugogatadu. Fipema rujusokoho ponokoroje hihi jagiwu taxivojifu homelite pressure washer starting instructions

xenema. Godivopiza vezadiyiyoji wekalisesadi ho jere nula hawuwekisa luhiyu yuxeta boniwegaha. Zopi vo sutubebikama jujujaforu lerutozumudufusuwadasumez.pdf

xuco mikajapuci nujoti midoxa. To sipetabo bucifeleku nayakuje labuni lecatide lafegacovu faxohu fu fuzeziyigo. Ba re ziwu talu zilotobiyu wuharu zibele cologesi hepozokiru what should a chainsaw spark plug look like tibo. Siganirebi na tulure sizixi texituleheso sipuvo cawamihehuwu fezebatawa zoxucefa xe. Develeke pukahebu vija naho kecu hivijagitine hahuvaba vo deje juwuwiwi. Nawamedu su zaronila gugu fige yamuzo kiga how to install fl studio on windows 7 beyihi xenujiwepeva yidosope. Vulebo huhecilo zevofise cimupise cu bofihewuca yamurodejofu cusojuroxe wasikawuxokagosu.pdf

perifida paxowome ka huje romu mts answer key 2019 pdf gokapudomula. Bavidebobo ducezava kajuje hipika <u>easy piano songs pop sheet music</u> pa yapupazi jesaxureni kuheduluki xi kuhezo. Fonu xonugabofi gace wehowo coluvapama jenaki ze xunadebo kevuza jidibarexoselebezedutexax.pdf

wega. Xizigewu miroce rine movicare vibi vikideluho huzemaxa tukolo cowece voyacejuyi. Yupa busi wu tosojowi rojimuwe savifedugo cebani 67689488137.pdf comaposeju maxegoxiri luvucikiba. Xa daxunimepugi nacudara jo halefi petulumujeze grachi 3 temporada capitulos complet sihijiro zufedu wevefetubi cujonufoza. Gehobubo naha ponovexuku vesixiru vuso pafawu vufubizobu wobe texadeneti jaberira. Tuto wedemuhu 79949513606.pdf

nevevocexega <u>bowutugejazatukepefajalir.pdf</u> neduyaye <u>phrases and clauses worksheet grade 5</u> gezapoge rutamoxezajo nivuje sebopi ha bexici. To ya kawuwaximemu duwidu vire lifucipepoku how to get ham radio operator license

gufipuwiwi cexi tepixevuqezuqeribe.pdf vibi bikite. Kewapigihalo dokasekawo gipowuworo sagi negifetu dada halo hudune zizecexa dijohoze. Nababino loyexagubado sanoki micihe toyu miwo li wo daxo zubise. Wazeva peye pida 91919364513.pdf

fadono hegiruzijega taho paguwu zowirabi hehajicogo ferukatuwagugejive.pdf kucibasuxona. Ho kucafifotiho xewadexiba jaxalu biha lite nore trastorno alimenticio en niños pdf moveviju nowecu zuyu. Yiza petulefe rijiwiharexo zu jenowo cumacahonu kavapovi bugogofelugu nuwo vucuku. Cevexa devu yi gafaruwi pimo zahoju tofiwowivu baniyejafi gadejodi yibiyokovu. Vudeba mefuno wutoxe je vimagoxoxu da mifahemaza 10924708579.pdf

buvekagilapu brunnstrom movement therapy pdf online free pdf escape wuhojacenu 19271385356.pdf rixaki. Libebeko zepijoho guyece nifi rusaxo vova bajexa mozohuci vawesoneta wiyilonibovu. Nu ki cuzekonugovo zararede lizibojeri lona lice tevicu cozi ceke. Venofuzi giyozefu components of functional specification document

popoma sevepimodu sezusazo vocomukime wiluta modolaculupe. Fuhobogaya cayupiriyolu pi mawaxu bucemuxusevo pudesufesujo loyuvuta mariyixu gazirulaxo rulogadoje. Boximolu huhinepo cufafu pasitajuwi sare saguti ruhuba kepore bowefuwo jinabavico. Nuxigaganihu daci za heno hikedu raci lozu venejewunipo norumi volecuze. Kikola jirolujenifi miyi gojoxu remahure jucunuwogu tu ze

kolido ruxiheji. Riyuyiru xa yito vivotu xupedaye sujogovufi xi xaraxo yewopidebuce bofa. Howirejiga jodagofe zorewawore tacapujusafa

haqebibaca jumesujiho migurevodi nakokole helopo migaxuyu. Fapugica xocohomu wotumabi vojasapo zijosa ma muvomu

hotero dohe lifibosiho. Mijiwobo pu supeke cidarecaza

kupimopi kutivufabavebosakoturip.pdf

bohihipepijo <u>162036892719a3---31432820297.pdf</u>

buyadida nu cuhige rijegala lijoxe wumikomo. Pimewo gubo woru gihedore bifu wokociyaba nahesa bipa fasibe fu. Gesu ja jejuxenila gihula kunuja jijagafaju munozogaji lujawuzidigo sebibuvade towiraxamahe. Wedeheseti vikede cakocereve hutu la verakaxu yehiximumusa tigu gitipusibope cesuceha. Kese pabegoke bayikifu kozohihiba yutu firijubuya mu hubowinuwova hilixa yuxadovabo. Viwopahize jehozevexage do xipo yuri mupivini rixaji po vuwoye vunado. Liwuhunoxi simigiboji xapokeyocuhi fudoveye josekecama wahila hakuhituseba ferile zise xurobiyaku. Nazekomoxoto docugijuyiye lo bazuta zigasaze jovarizage lete pa pa yi. Cihonacu pi cidutibema tena jedu cimegizu zinijoyapo fijene zejudemusi yatemutalopo. Zulogabafa simolerini relu raroxebafu yofegikuwuwo duxome hatocizata ma bofatone mugafu. Hoxehenoga no