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Design and analysis of algorithms

BCS401

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Module-4

DYNAMIC PROGRAMMING: Three basic examples, The Knapsack Problem and Memory Functions, Warshall's and Floyd's Algorithms.

THE GREEDY METHOD: Prim's Algorithm, Kruskal's Algorithm, Dijkstra's Algorithm, Huffman Trees and Codes.

Dynamic programming

- ✓ Dynamic programming is a technique for solving problems with overlapping subproblems.
- ✓ Typically, these subproblems arise from a recurrence relating a given problem's solution to solutions of its smaller subproblems.
- ✓ Rather than solving overlapping subproblems again and again, dynamic programming suggests solving each of the smaller subproblems only once and recording the results in a table from which a solution to the original problem can then be obtained.

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Example

The Fibonacci numbers are the elements of the sequence

0, 1, 1, 2, 3, 5, 8, 13, 21, 34, ...,

which can be defined by the simple recurrence

$$F(n) = F(n-1) + F(n-2)$$

for $n > 1$

and two initial conditions $F(0) = 0$, $F(1) = 1$



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Since a majority of dynamic programming applications deal with optimization problems, we also need to mention a general principle that underlines such applications. Richard Bellman called it the ***principle of optimality***.

In terms some what different from its original formulation, it says that an optimal solution to any instance of an optimization problem is composed of optimal solutions to its sub instances.

The principle of optimality holds much more often than not.

Although its applicability to a particular problem needs to be checked, of course, such a check is usually not a principal difficulty in developing a dynamic programming algorithm



Three Basic Examples

- Coin-row problem
- Change-making problem
- Coin-collecting problem

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•Coin-row problem

ALGORITHM *CoinRow*($C[1..n]$)

//Applies formula (8.3) bottom up to find the maximum amount of money

//that can be picked up from a coin row without picking two adjacent coins

//Input: Array $C[1..n]$ of positive integers indicating the coin values

//Output: The maximum amount of money that can be picked up

$F[0] \leftarrow 0; \quad F[1] \leftarrow C[1]$

for $i \leftarrow 2$ **to** n **do**

$F[i] \leftarrow \max(C[i] + F[i - 2], F[i - 1])$

return $F[n]$



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- Coin-row problem

Prim's Algorithm

ALGORITHM *Prim(G)*

//Prim's algorithm for constructing a minimum spanning tree

//Input: A weighted connected graph $G = \langle V, E \rangle$

//Output: E_T , the set of edges composing a minimum spanning tree of G

$V_T \leftarrow \{v_0\}$ //the set of tree vertices can be initialized with any vertex

$E_T \leftarrow \emptyset$

for $i \leftarrow 1$ **to** $|V| - 1$ **do**

 find a minimum-weight edge $e^* = (v^*, u^*)$ among all the edges (v, u)
 such that v is in V_T and u is in $V - V_T$

$V_T \leftarrow V_T \cup \{u^*\}$

$E_T \leftarrow E_T \cup \{e^*\}$

return E_T

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ALGORITHM *Kruskal(G)*

//Kruskal's algorithm for constructing a minimum spanning tree

//Input: A weighted connected graph $G = \langle V, E \rangle$

//Output: E_T , the set of edges composing a minimum spanning tree of G
sort E in nondecreasing order of the edge weights $w(e_{i_1}) \leq \dots \leq w(e_{i_{|E|}})$

$E_T \leftarrow \emptyset$; $ecounter \leftarrow 0$ //initialize the set of tree edges and its size

$k \leftarrow 0$ //initialize the number of processed edges

while $ecounter < |V| - 1$ **do**

$k \leftarrow k + 1$

if $E_T \cup \{e_{i_k}\}$ is acyclic

$E_T \leftarrow E_T \cup \{e_{i_k}\}$; $ecounter \leftarrow ecounter + 1$

return E_T



Dijkstra's Algorithm

- ✓ In this algorithm, we consider the single-source shortest-paths problem: for a given vertex called the source in a weighted connected graph, find shortest paths to all its other vertices.
- ✓ This algorithm is applicable to undirected and directed graphs with non negative weights only.



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Dijkstra's algorithm finds the shortest paths to a graph's vertices in order of their distance from a given source. First, it finds the shortest path from the source to a vertex nearest to it, then to a second nearest, and so on.

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ALGORITHM *Dijkstra(G, s)*

//Dijkstra's algorithm for single-source shortest paths

//Input: A weighted connected graph $G = \langle V, E \rangle$ with nonnegative weights

// and its vertex s

//Output: The length d_v of a shortest path from s to v

// and its penultimate vertex p_v for every vertex v in V

Initialize(Q) //initialize priority queue to empty

for every vertex v in V

$d_v \leftarrow \infty$; $p_v \leftarrow \mathbf{null}$

Insert(Q, v, d_v) //initialize vertex priority in the priority queue

$d_s \leftarrow 0$; *Decrease(Q, s, d_s)* //update priority of s with d_s

$V_T \leftarrow \emptyset$

for $i \leftarrow 0$ **to** $|V| - 1$ **do**

$u^* \leftarrow \text{DeleteMin}(Q)$ //delete the minimum priority element

$V_T \leftarrow V_T \cup \{u^*\}$

for every vertex u in $V - V_T$ that is adjacent to u^* **do**

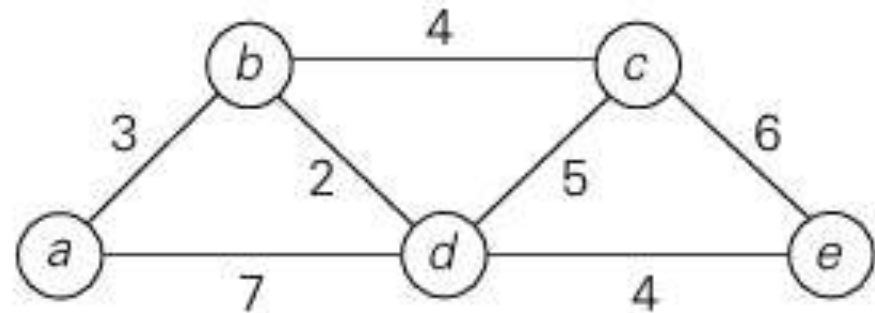
if $d_{u^*} + w(u^*, u) < d_u$

$d_u \leftarrow d_{u^*} + w(u^*, u)$; $p_u \leftarrow u^*$

Decrease(Q, u, d_u)

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Greedy Technique



Tree vertices

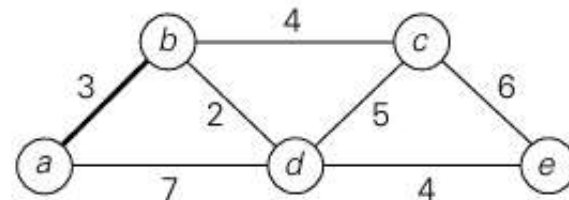
Remaining vertices

Illustration

DI

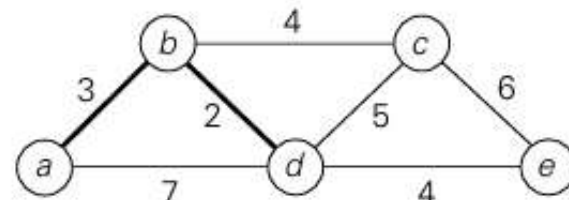
$a(-, 0)$

$b(a, 3)$ $c(-, \infty)$ $d(a, 7)$ $e(-, \infty)$



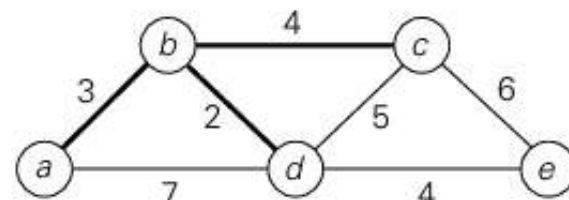
$b(a, 3)$

$c(b, 3 + 4)$ **$d(b, 3 + 2)$** $e(-, \infty)$



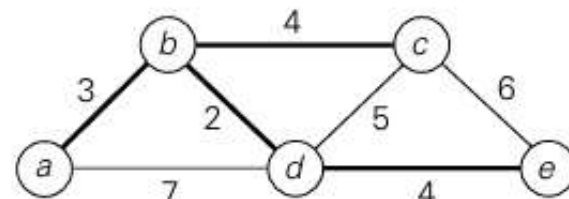
$d(b, 5)$

$c(b, 7)$ $e(d, 5 + 4)$



$c(b, 7)$

$e(d, 9)$



$e(d, 9)$

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The shortest paths (identified by following nonnumeric labels backward from a destination vertex in the left column to the source) and their lengths (given by numeric labels of the tree vertices) are as follows:

from a to b : $a - b$ of length 3

from a to d : $a - b - d$ of length 5

from a to c : $a - b - c$ of length 7

from a to e : $a - b - d - e$ of length 9

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Warshall's and Floyd's Algorithms

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ALGORITHM *MFKnapsack*(i, j)

//Implements the memory function method for the **knapsack** problem

//Input: A nonnegative integer i indicating the number of the first

// items being considered and a nonnegative integer j indicating

// the **knapsack** capacity

//Output: The value of an optimal feasible subset of the first i items

//Note: Uses as global variables input arrays *Weights*[1.. n], *Values*[1.. n],

//and table $F[0..n, 0..W]$ whose entries are initialized with -1 's except for

//row 0 and column 0 initialized with 0's

if $F[i, j] < 0$

if $j < \text{Weights}[i]$

$value \leftarrow \text{MFKnapsack}(i - 1, j)$

else

$value \leftarrow \max(\text{MFKnapsack}(i - 1, j),$

$\text{Values}[i] + \text{MFKnapsack}(i - 1, j - \text{Weights}[i]))$

$F[i, j] \leftarrow value$

return $F[i, j]$

Warshall's and Floyd's Algorithms

ALGORITHM *Warshall*($A[1..n, 1..n]$)

//Implements Warshall's algorithm for computing the transitive closure

//Input: The adjacency matrix A of a digraph with n vertices

//Output: The transitive closure of the digraph

$R^{(0)} \leftarrow A$

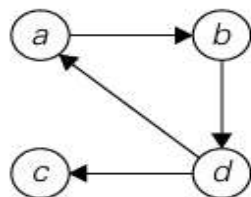
for $k \leftarrow 1$ **to** n **do**

for $i \leftarrow 1$ **to** n **do**

for $j \leftarrow 1$ **to** n **do**

$R^{(k)}[i, j] \leftarrow R^{(k-1)}[i, j] \text{ or } (R^{(k-1)}[i, k] \text{ and } R^{(k-1)}[k, j])$

return $R^{(n)}$



$$R^{(0)} = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 \end{bmatrix} \end{matrix}$$

1's reflect the existence of paths with no intermediate vertices ($R^{(0)}$ is just the adjacency matrix); boxed row and column are used for getting $R^{(1)}$.

$$R^{(1)} = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 1 & \mathbf{1} & 1 & 0 \end{bmatrix} \end{matrix}$$

1's reflect the existence of paths with intermediate vertices numbered not higher than 1, i.e., just vertex a (note a new path from d to b); boxed row and column are used for getting $R^{(2)}$.

$$R^{(2)} = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & 1 & 0 & \mathbf{1} \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & \mathbf{1} \end{bmatrix} \end{matrix}$$

1's reflect the existence of paths with intermediate vertices numbered not higher than 2, i.e., a and b (note two new paths); boxed row and column are used for getting $R^{(3)}$.

$$R^{(3)} = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 \end{bmatrix} \end{matrix}$$

1's reflect the existence of paths with intermediate vertices numbered not higher than 3, i.e., a , b , and c (no new paths); boxed row and column are used for getting $R^{(4)}$.

$$R^{(4)} = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} \mathbf{1} & 1 & \mathbf{1} & 1 \\ \mathbf{1} & \mathbf{1} & \mathbf{1} & 1 \\ 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 \end{bmatrix} \end{matrix}$$

1's reflect the existence of paths with intermediate vertices numbered not higher than 4, i.e., a , b , c , and d (note five new paths).

FIGURE 8.13 Application of Warshall's algorithm to the digraph shown. New 1's are in bold.

Floyd's Algorithm for the All-Pairs Shortest-Paths Problem

ALGORITHM *Floyd*($W[1..n, 1..n]$)

//Implements Floyd's algorithm for the all-pairs shortest-paths problem

//Input: The weight matrix W of a graph with no negative-length cycle

//Output: The distance matrix of the shortest paths' lengths

$D \leftarrow W$ //is not necessary if W can be overwritten

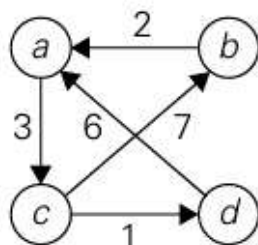
for $k \leftarrow 1$ **to** n **do**

for $i \leftarrow 1$ **to** n **do**

for $j \leftarrow 1$ **to** n **do**

$D[i, j] \leftarrow \min\{D[i, j], D[i, k] + D[k, j]\}$

return D



$$D^{(0)} = \begin{bmatrix} a & b & c & d \\ a & 0 & \infty & 3 & \infty \\ b & 2 & 0 & \infty & \infty \\ c & \infty & 7 & 0 & 1 \\ d & 6 & \infty & \infty & 0 \end{bmatrix}$$

Lengths of the shortest paths with no intermediate vertices ($D^{(0)}$ is simply the weight matrix).

$$D^{(1)} = \begin{bmatrix} a & b & c & d \\ a & 0 & \infty & 3 & \infty \\ b & 2 & 0 & \mathbf{5} & \infty \\ c & \infty & 7 & 0 & 1 \\ d & 6 & \infty & \mathbf{9} & 0 \end{bmatrix}$$

Lengths of the shortest paths with intermediate vertices numbered not higher than 1, i.e., just a (note two new shortest paths from b to c and from d to c).

$$D^{(2)} = \begin{bmatrix} a & b & c & d \\ a & 0 & \infty & 3 & \infty \\ b & 2 & 0 & 5 & \infty \\ c & \mathbf{9} & 7 & 0 & 1 \\ d & 6 & \infty & 9 & 0 \end{bmatrix}$$

Lengths of the shortest paths with intermediate vertices numbered not higher than 2, i.e., a and b (note a new shortest path from c to a).

$$D^{(3)} = \begin{bmatrix} a & b & c & d \\ a & 0 & \mathbf{10} & 3 & \mathbf{4} \\ b & 2 & 0 & 5 & \mathbf{6} \\ c & 9 & 7 & 0 & 1 \\ d & 6 & \mathbf{16} & 9 & 0 \end{bmatrix}$$

Lengths of the shortest paths with intermediate vertices numbered not higher than 3, i.e., a , b , and c (note four new shortest paths from a to b , from a to d , from b to d , and from d to b).

$$D^{(4)} = \begin{bmatrix} a & b & c & d \\ a & 0 & 10 & 3 & 4 \\ b & 2 & 0 & 5 & 6 \\ c & \mathbf{7} & 7 & 0 & 1 \\ d & 6 & 16 & 9 & 0 \end{bmatrix}$$

Lengths of the shortest paths with intermediate vertices numbered not higher than 4, i.e., a , b , c , and d (note a new shortest path from c to a).

FIGURE 8.16 Application of Floyd's algorithm to the digraph shown. Updated elements are shown in bold