

# Greedy Technique

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# Greedy Technique

The greedy approach suggests constructing a solution through a sequence of steps, each expanding a partially constructed solution obtained so far, until a complete solution to the problem is reached. On each step, the choice made must be:

# Greedy Technique ..contd

- **feasible**: it has to satisfy the problem's constraints
- **locally optimal**: it has to be the best local choice among all feasible choices available on that step
- **irrevocable**: once decision was made, it cannot be changed on subsequent steps of the algorithm



# Principle of Greedy Technique

Principle: A sequence of locally optimal choices will yield a globally optimal solution to the entire problem

However, a sequence of locally optimal choices does not always yield a globally optimal solution

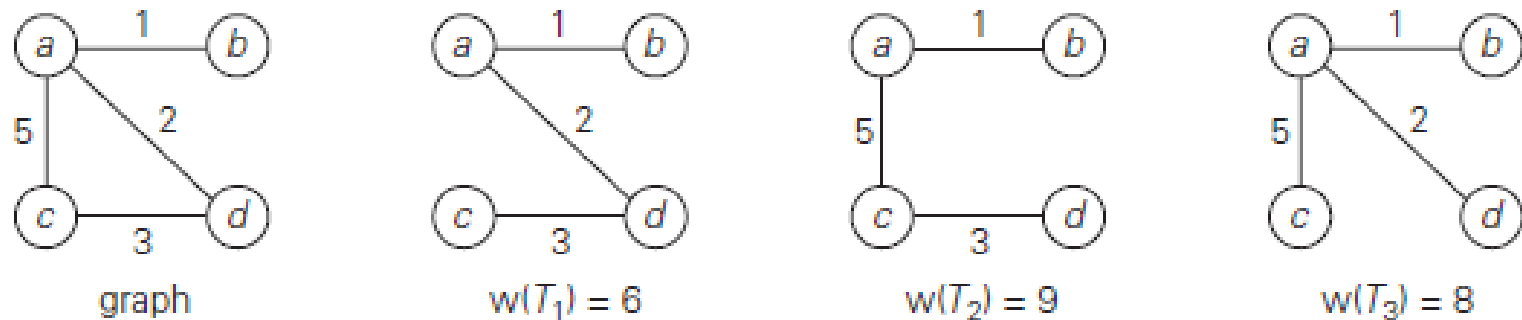
# Minimum Spanning Tree (MST) Problem

- Given  $n$  points, connect them in the cheapest possible way so that there will be a path between every pair of points.
- The solution is called the minimum spanning tree (MST).
- The MST is a tree connecting all points and has the minimum cost (also called length, or weight).

# Minimum Spanning Tree (MST) Problem

- The minimum spanning tree problem is the problem of finding a minimum spanning tree (MST) for a given weighted connected undirected graph.
- The total number of possible spanning trees for  $n$  vertices (nodes, points, or cities) is  $n^{n-2}$  (it grows exponentially).
- Thus, it is impossible to employ the brute-force strategy to solve the MST problem. It is solved by the greedy method.

# Example



**FIGURE 9.2** Graph and its spanning trees, with  $T_1$  being the minimum spanning tree.

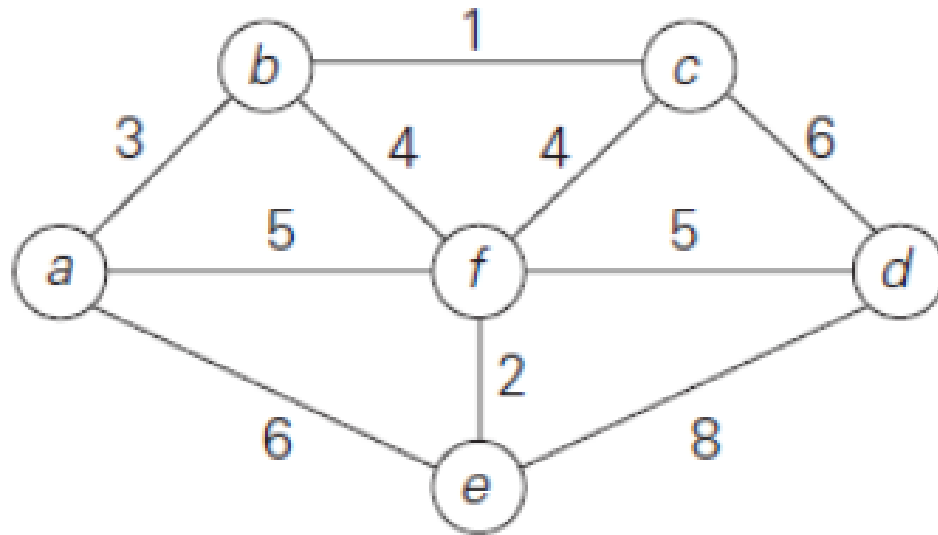
# Algorithms for obtaining MST

- 1) Prim's Algorithm
- 2) Kruskal's Algorithm



# Prim's Algorithm

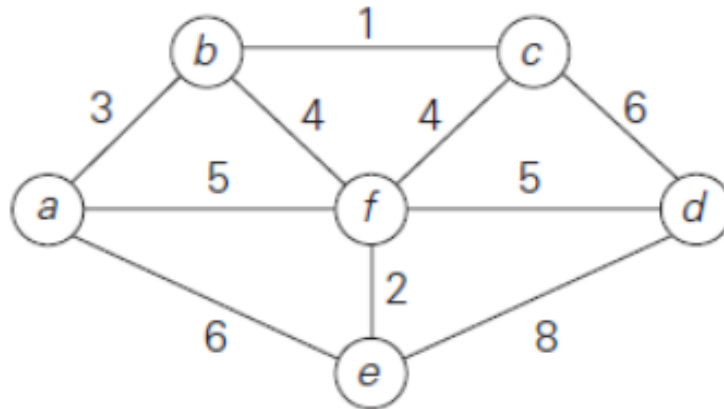
- Consider the following weighted, connected, undirected graph



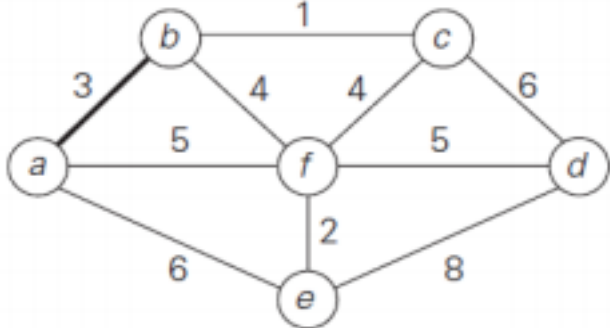
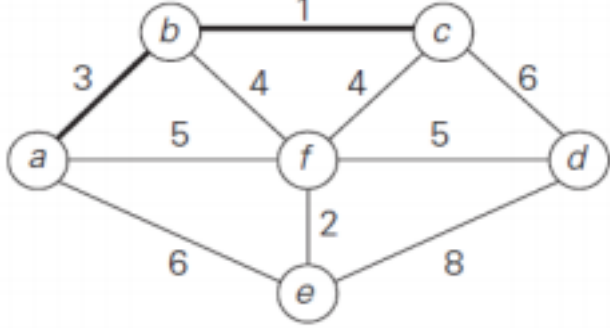
# Prim's Algorithm

- Attach two labels to a vertex:
  - the name of the nearest tree vertex, and
  - the weight of the corresponding edge.
- Example: select a as starting vertex

- a(-, -)
- b(a, 3)
- c(-,  $\infty$ )



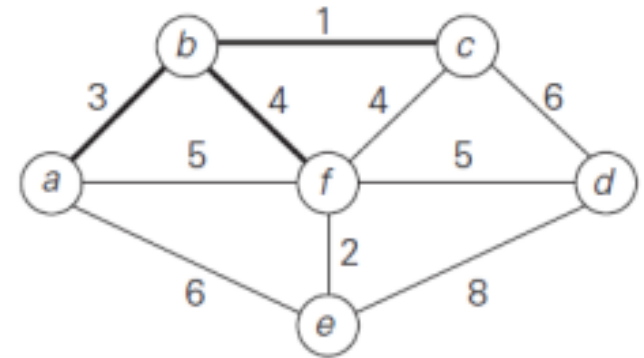
# Prim's Algorithm

Tree vertices	Remaining vertices	Illustration
$a(-, -)$	$\mathbf{b(a, 3)}$ $c(-, \infty)$ $d(-, \infty)$ $e(a, 6)$ $f(a, 5)$	
$\mathbf{b(a, 3)}$	$\mathbf{c(b, 1)}$ $d(-, \infty)$ $e(a, 6)$ $f(b, 4)$	

# Prim's Algorithm

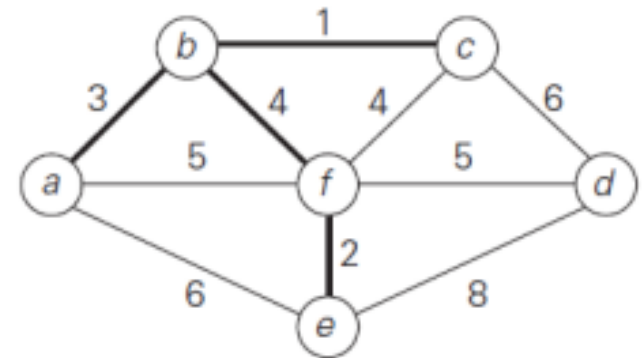
$c(b, 1)$

$d(c, 6)$   $e(a, 6)$   **$f(b, 4)$**



$f(b, 4)$

$d(f, 5)$   **$e(f, 2)$**

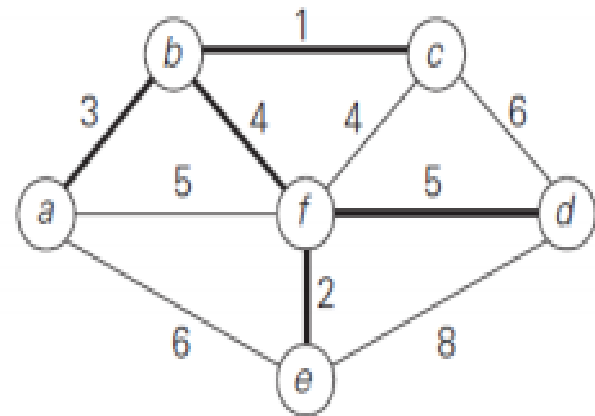


# Prim's Algorithm

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$e(f, 2)$

$d(f, 5)$



$d(f, 5)$

# 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**  $E_T$

# Time Complexity

- In particular, if a graph is represented by its weight matrix and the priority queue is implemented as an unordered array, the algorithm's running time will be in  $\Theta(|V|^2)$
- If a graph is represented by its adjacency lists and the priority queue is implemented as a min-heap, the running time of the algorithm is in  $O(|E| \log |V|)$ .



Thank You



# Dijkstra's Algorithm for Single Source Shortest Path Problem using Greedy Technique

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# *Single Source Shortest Path Problem*

*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.*

- Best-known algorithm for the single-source shortest-paths problem, called *Dijkstra's algorithm*
- *This algorithm is applicable to undirected and directed graphs with nonnegative weights only*

# Dijkstra's algorithm

- 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.
- In general, before its  $i$ th iteration commences, the algorithm has already identified the shortest paths to  $i - 1$  other vertices nearest to the source.
- These vertices, the source, and the edges of the shortest paths leading to them from the source form a subtree  $T_i$  of the given graph

# Dijkstra's algorithm

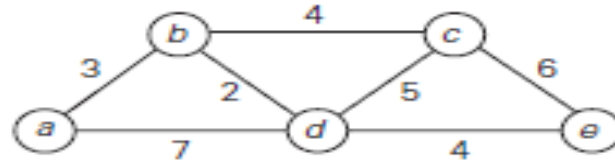
Among vertices not already in the tree, it finds vertex  $u$  with the smallest sum

$$d_v + w(v,u)$$

where

- $v$  is a vertex for which shortest path has been already found on preceding iterations (such vertices form a tree)
- $d_v$  is the length of the shortest path from source to  $v$
- $w(v,u)$  is the length (weight) of edge from  $v$  to  $u$

# Example



Tree vertices	Remaining vertices	Illustration
$a(-, 0)$	<b>b(a, 3)</b> $c(-, \infty)$ $d(a, 7)$ $e(-, \infty)$	
$b(a, 3)$	$c(b, 3 + 4)$ <b>d(b, 3 + 2)</b> $e(-, \infty)$	
$d(b, 5)$	<b>c(b, 7)</b> $e(d, 5 + 4)$	
$c(b, 7)$	<b>e(d, 9)</b>	
$e(d, 9)$		

**FIGURE 9.11** Application of Dijkstra's algorithm. The next closest vertex is shown in bold.

# Dijkstra's algorithm

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

# Dijkstra's Algorithm

## 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  $V$

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

*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*( $Q, u, d_u$ )

# Time Complexity

- The time efficiency of Dijkstra's algorithm depends on the data structures used for implementing the priority queue and for representing an input graph itself
- It is  $(\Theta|V|^2)$  for graphs represented by their weight matrix and the priority queue implemented as an unordered array.
- For graphs represented by their adjacency lists and the priority queue implemented as a min-heap, it is in  $O(|E| \log |V|)$ .



# Dijkstra's Algorithm


- Doesn't work for graphs with negative weights
- Applicable to both undirected and directed graphs

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- Thank You

# Huffman Trees and Codes

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# Huffman Trees and codes

- ▶ A Huffman tree is a binary tree that minimizes the weighted path length from the root to the leaves of predefined weights.
  - ▶ The most important application of Huffman trees is Huffman codes.
  - ▶ A Huffman code is an optimal prefix-free variable-length encoding scheme that assigns bit strings to symbols based on their frequencies in a given text.
  - ▶ This is accomplished by a greedy construction of a binary tree whose leaves represent the alphabet symbols and whose edges are labeled with 0's and 1's.
- 

# Huffman Trees and codes

- ▶ Huffman's encoding is one of the most important file-compression methods.
- ▶ In addition to its simplicity and versatility, it yields an optimal, i.e., minimal-length, encoding

# Encoding Methods

- ▶ Fixed Length Encoding
- ▶ Variable Length Encoding

# Fixed –Length encoding

- ▶ Suppose we have to encode a text that comprises symbols from some  $n$ -symbol alphabet by assigning to each of the text's symbols some sequence of bits called the codeword.
- ▶ For example, we can use a fixed-length encoding that assigns to each symbol a bit string of the same length  $m$  ( $m \geq \log_2 n$ ).
- ▶ This is exactly what the standard ASCII code does

# Fixed Length Codes

Represent data as a sequence of 0's and 1's

Sequence: **BACADAEAFABBAAGA**H

A fixed length code:

**A** 000    **B** 001    **C** 010    **D** 011

**E** 100    **F** 101    **G** 110    **H** 111

Encoding of sequence:

00100001000001100010000010100000100100000000001  
10000111

**The Encoding is  $18 \times 3 = 54$  bits long.**

Can we make the encoding shorter?



# Variable-length encoding

- ▶ *Variable-length encoding, which assigns codewords of different lengths to different symbols*
- ▶ In a prefix code, no codeword is a prefix of a codeword of another symbol.
- ▶ Hence, with such an encoding, we can simply scan a bit string until we get the first group of bits that is a codeword for some symbol, replace these bits by this symbol, and repeat this operation until the bit string's end is reached.

# Variable Length Code–Huffman

- ▶ Make use of frequencies. Frequency of A=8, B=3, others 1
- ▶ A 0      B 100    C 1010    D 1011
- ▶ E 1100   F 1101   G 1110    H 1111
- ▶ Example: BACADAEAFABBAAGH
- ▶ 100010100101101100011010100100000111001111

42 bits (20% shorter)

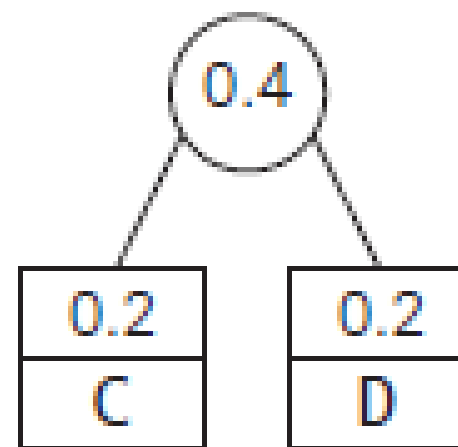
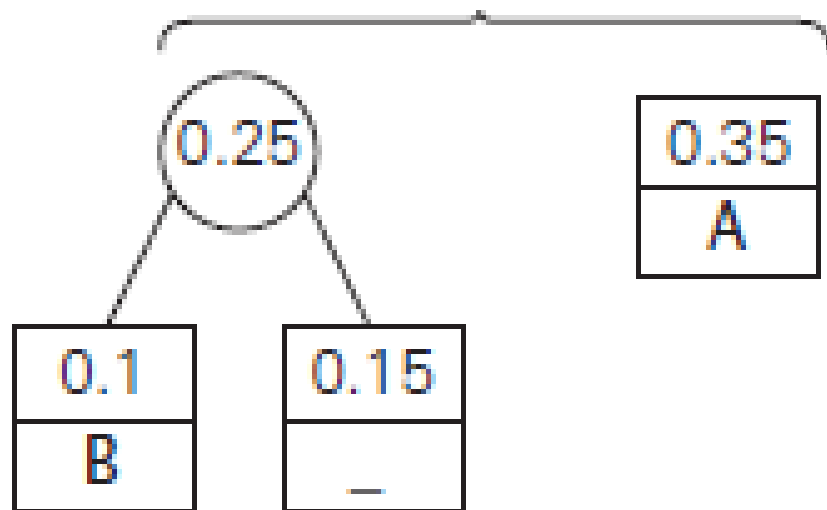
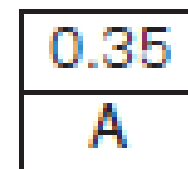
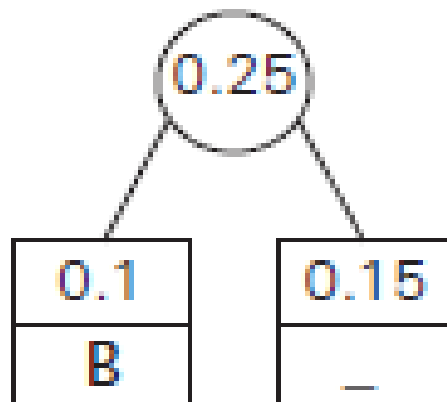
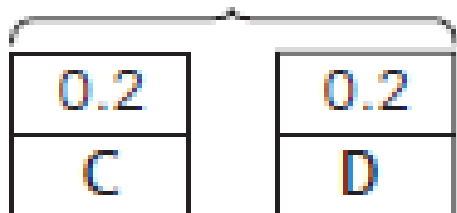
# Huffman's algorithm

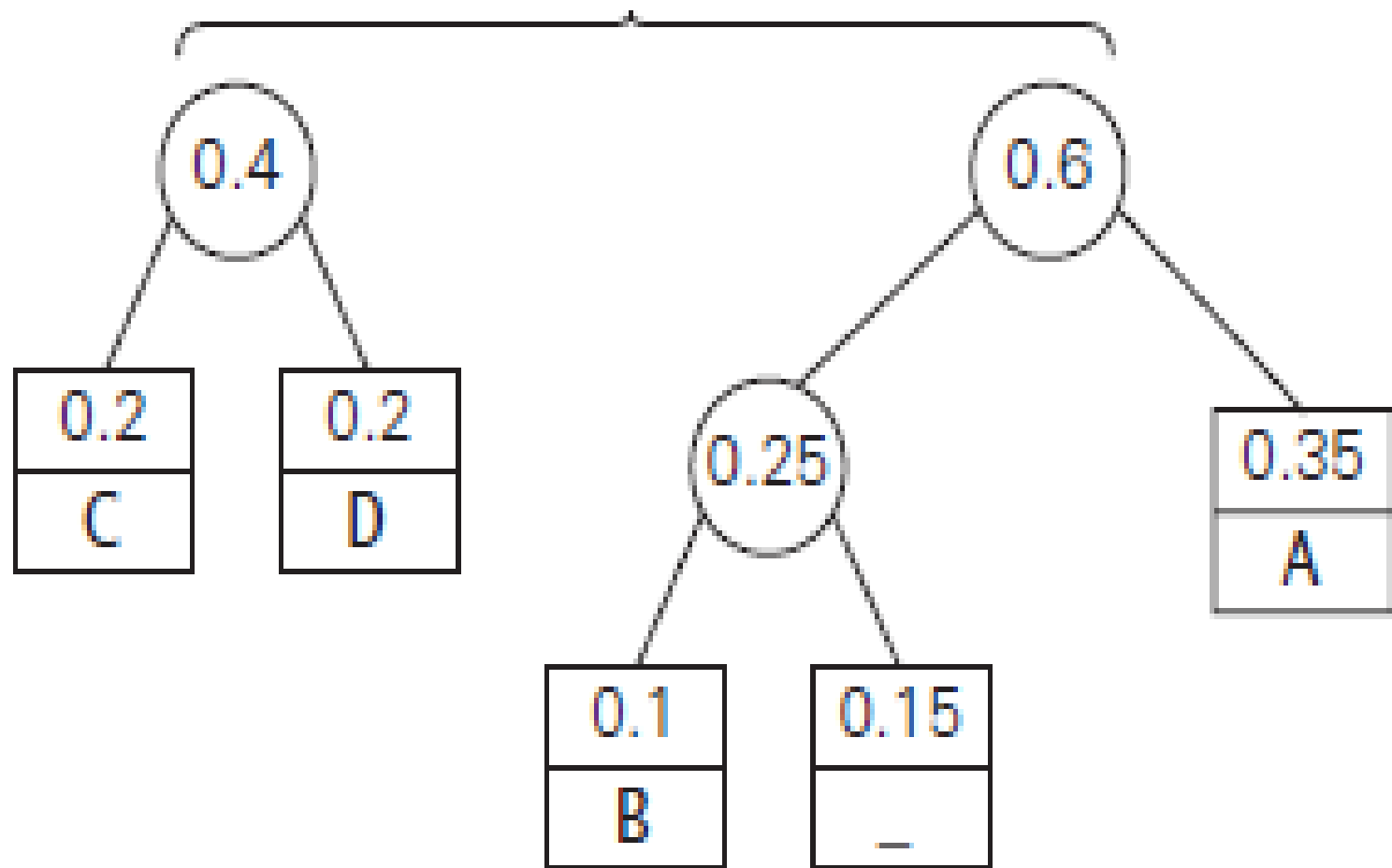
- ▶ **Step 1:** Initialize  $n$  *one-node trees* and label them with the symbols of the alphabet given. Record the frequency of each symbol in its tree's root to indicate the tree's *weight*
- ▶ **Step 2:** Repeat the following operation until a single tree is obtained. Find two trees with the smallest weight. Make them the left and right subtree of a new tree and record the sum of their weights in the root of the new tree as its weight.

A tree constructed by the above algorithm is called a *Huffman tree*. It defines a *Huffman code*

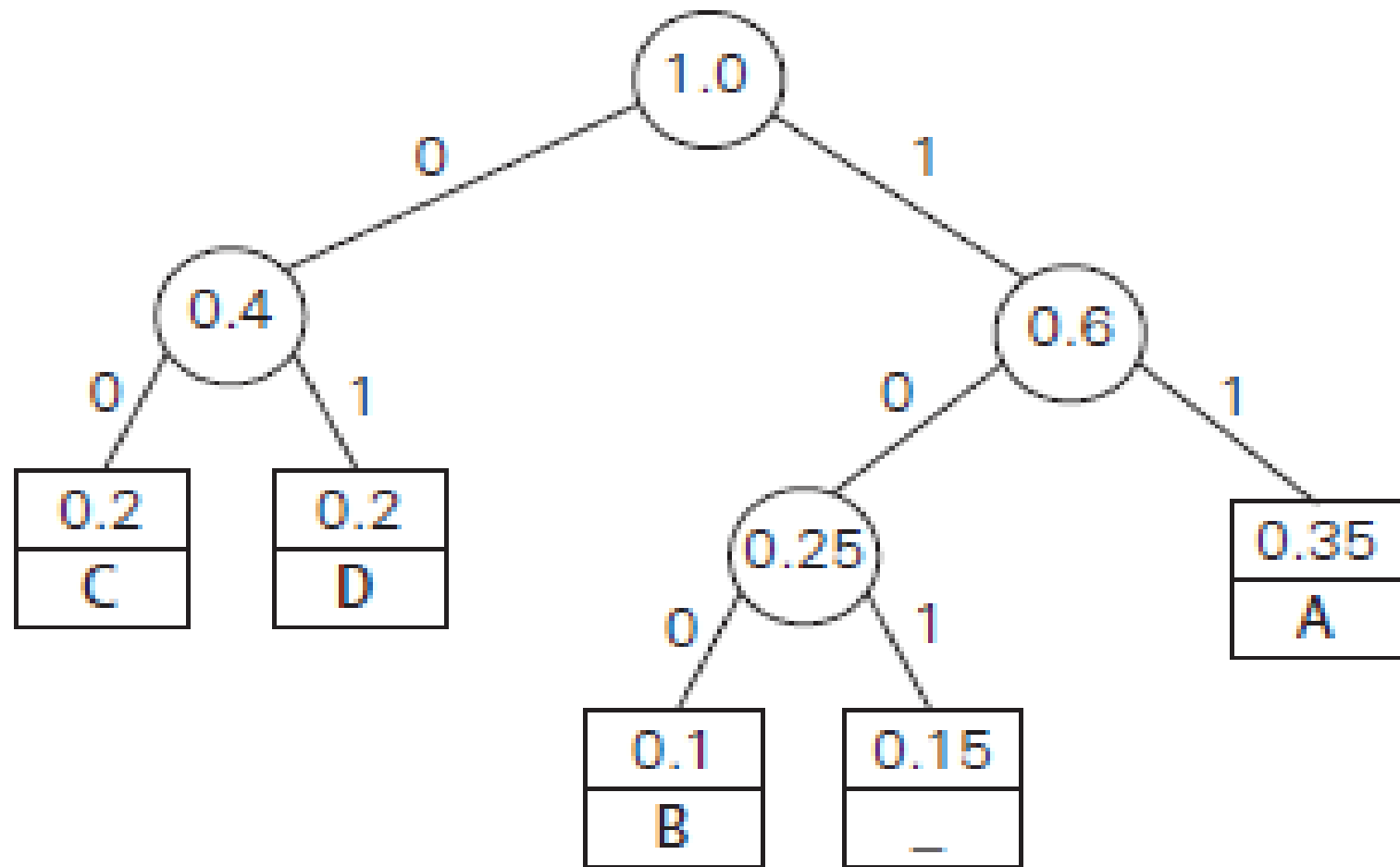
**EXAMPLE** Consider the five-symbol alphabet {A, B, C, D, \_} with the following occurrence frequencies in a text made up of these symbols:

symbol	A	B	C	D	_
frequency	0.35	0.1	0.2	0.2	0.15





# Example of constructing a Huffman coding tree



# Example of constructing a Huffman coding tree

The resulting codewords are as follows:

symbol	A	B	C	D	_
frequency	0.35	0.1	0.2	0.2	0.15
codeword	11	100	00	01	101

- ▶ DAD is encoded as 011101
- ▶ 10011011011101 is decoded as BAD\_AD



# Cost of Tree Corresponding to Prefix Code

- ▶ Given a tree  $T$  corresponding to a prefix code. For each character  $c$  in the alphabet  $C$ ,
  - let  $f(c)$  denote the frequency of  $c$  in the file and
  - let  $d_T(c)$  denote the depth of  $c$ 's leaf in the tree.
  - $d_T(c)$  is also the length of the codeword for character  $c$ .
  - The number of bits required to encode a file is

$$B(T) = \sum_{c \in C} f(c) d_T(c)$$

- which we define as the *cost* of the tree  $T$ .

# Compression Ratio

- ▶ With the occurrence frequencies given and the codeword lengths obtained, the average number of bits per symbol in this code is

$$2 \cdot 0.35 + 3 \cdot 0.1 + 2 \cdot 0.2 + 2 \cdot 0.2 + 3 \cdot 0.15 = 2.25.$$

- ▶ Had we used a fixed-length encoding for the same alphabet, we would have to use at least 3 bits per each symbol
- ▶ Huffman codes have shown that the compression ratio for this scheme typically falls between 20% and 80%, depending on the characteristics of the text being compressed (generally about 25%)

Thank You