

## ***IV Advanced Design and Analysis Techniques***

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# Elements of a Greedy Strategy

1. Determine the optimal substructure of the problem.
2. Develop a recursive solution. Show that if we make the greedy choice, then only one subproblem remains.
3. Prove that it is always safe to make the greedy choice.
4. Develop a recursive algorithm that implements the greedy strategy.
5. Convert the recursive algorithm to an iterative algorithm.

(Steps 3 and 4 can occur in either order.)

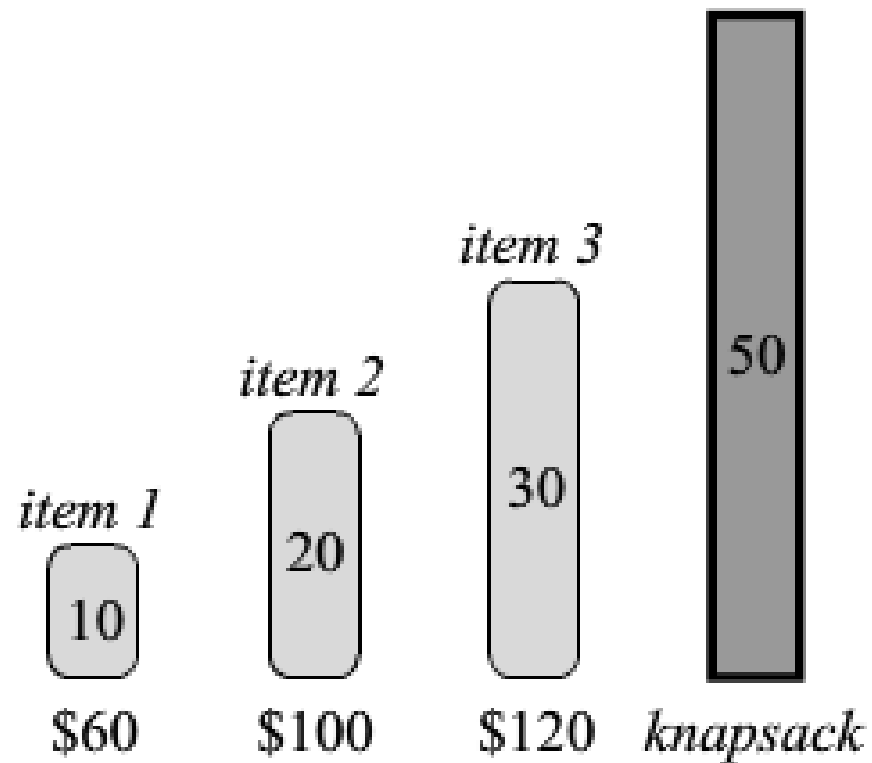
# Greedy Algorithm Design

1. Cast the optimization problem as :
  - One in which we make a choice
  - And are left with one subproblem to solve.
2. Prove that there is always an optimal solution to the original problem that makes the greedy choice
  - The greedy choice is always safe.
3. Demonstrate optimal substructure by
  - Showing that by having made the greedy choice,
  - What remains is a subproblem with the property
    - IF we combine an optimal solution to the subproblem with the greedy choice we have made,
    - THEN we arrive at an optimal solution to the original problem.

# Key Properties for Greedy

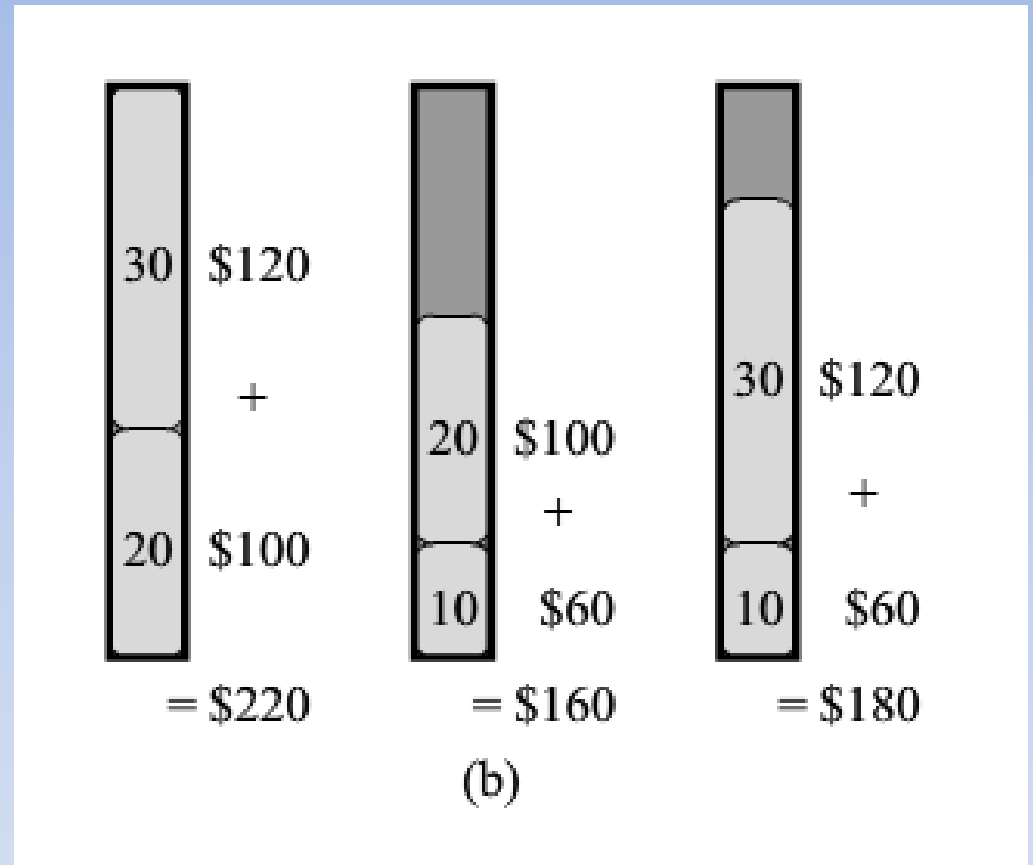
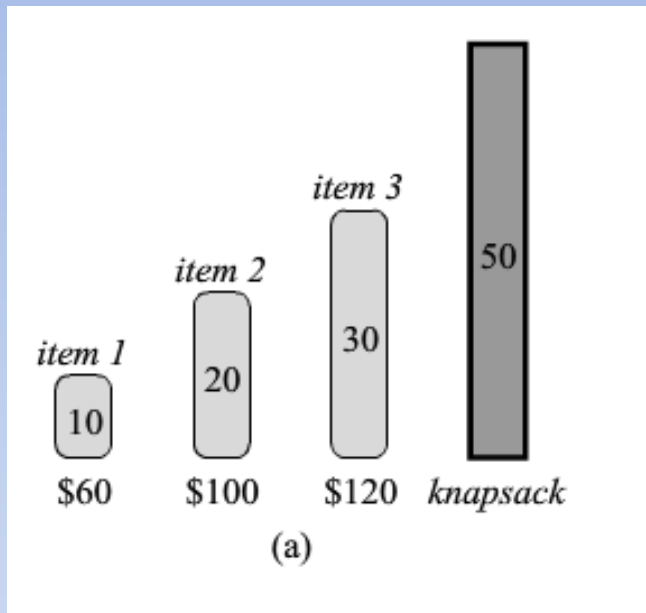
- Greedy-Choice Property:
  - We can assemble a globally optimal solution by making locally optimal (greedy) choices.
- Optimal Substructure:
  - A problem exhibits optimal substructure if an optimal solution to the problem contains within it optimal solutions to subproblems.

# Knapsack Problem



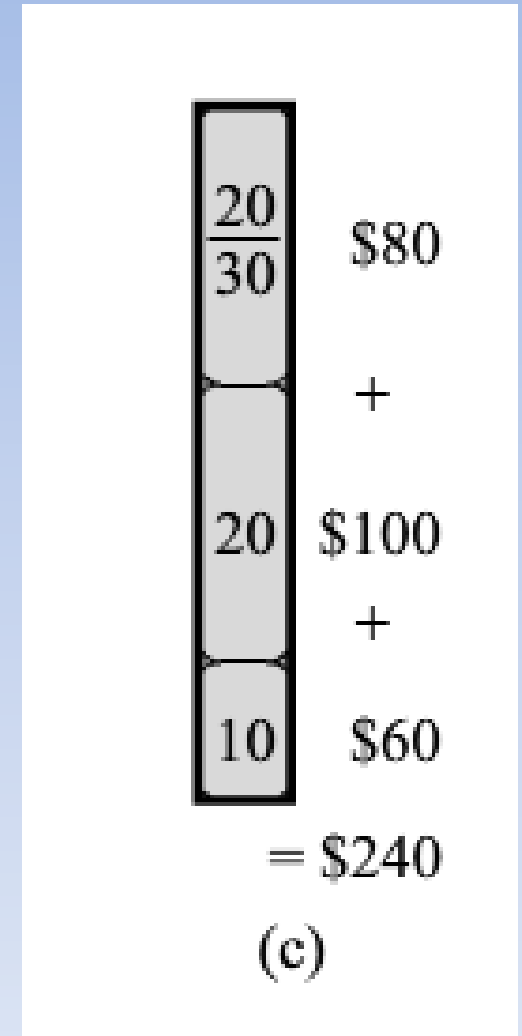
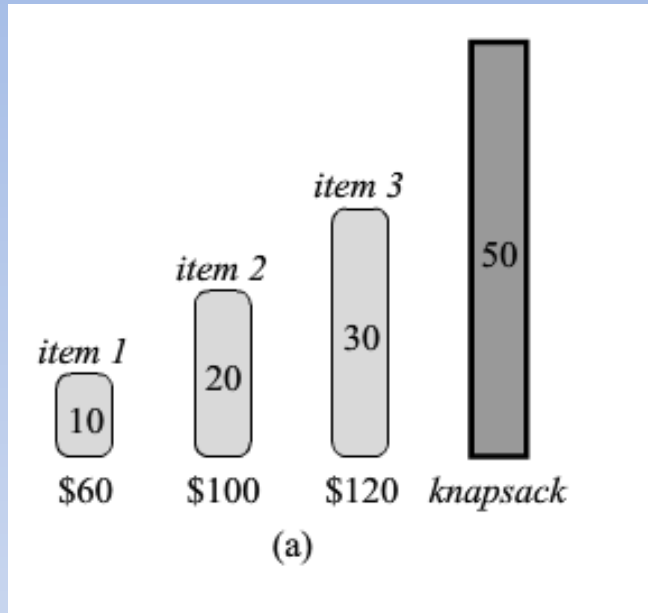
(a)

# 0-1 Knapsack Problem



- Item 1
  - Per/lb Most Expensive
- Optimal Solution does not use Item 1

# Fractional Knapsack



# Huffman Codes

## Ch#16.3

- Compress data well
- Used frequently because of simplicity
- Also combined with other methods.



# Example Problem & Huffman Codes

- 100,000-Character File
- 6 different characters appear
- 'a' appears 45,000 times.
- Need to design a Binary Character Code.

	a	b	c	d	e	f
Frequency (in thousands)	45	13	12	16	9	5
Fixed-length codeword	000	001	010	011	100	101
Variable-length codeword	0	101	100	111	1101	1100

**Figure 16.3** A character-coding problem. A data file of 100,000 characters contains only the characters a–f, with the frequencies indicated. If we assign each character a 3-bit codeword, we can encode the file in 300,000 bits. Using the variable-length code shown, we can encode the file in only 224,000 bits.

- Fixed length code requires 300,000 bits!
  - Can we do better!!
- Variable length code provide uses only 224,000 bits.
- How can we develop a variable length code?

# Prefix Codes

- No code word is also a prefix of any other code word.
  - A=0
  - B=101
  - C=100
  - D=111
  - E=1101
  - F=1100

# Decoding Prefix Codes

- Code String: 001011101
- Code:
  - A=0,
  - B=101, C=100, D=111
  - E=1101, F=1100
- Uniquely Decodes As:
  - 0 => A
  - 0 => A
  - 101 => B
  - 1101 => E

# Decoding Prefix Codes

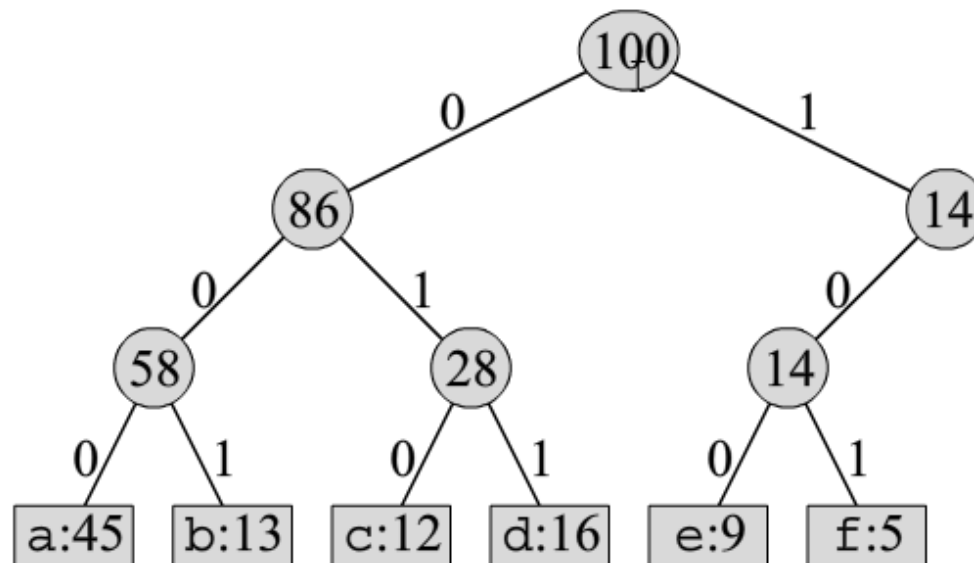
- Code String: 001011101
- Code:
  - A=0,
  - B=101, C=100, D=111
  - E=1101, F=1100
- Uniquely Decodes As:
  - 0 => A
  - 0 => A
  - 101 => B
  - 1101 => E
- Need an Efficient Data Structure to support decoding!

# Prefix Codes w/ Binary Tree

- Code 1:
  - A=000, B=001, C=010,
  - D=011 E=100, F=101

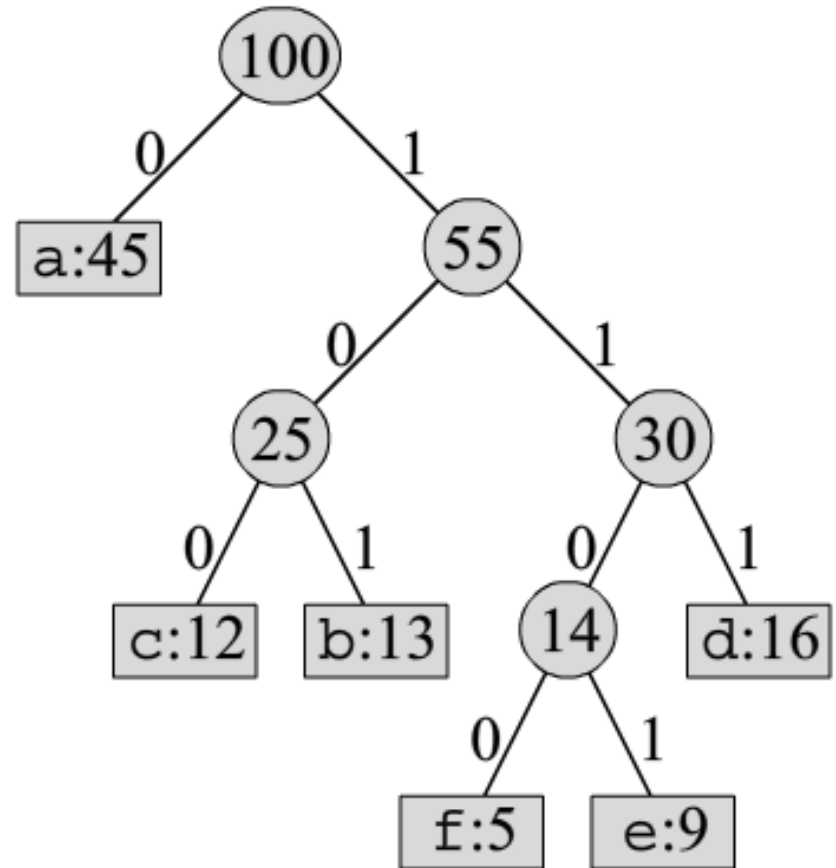
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*Chapter 16 Greedy Algorithms*



# Prefix Codes w/ Binary Tree

- Code:
  - A=0,
  - B=101, C=100, D=111
  - E=1101, F=1100



# Huffman Algorithm

HUFFMAN( $C$ )

```
1   $n = |C|$ 
2   $Q = C$ 
3  for  $i = 1$  to  $n - 1$ 
4      allocate a new node  $z$ 
5       $z.left = x = \text{EXTRACT-MIN}(Q)$ 
6       $z.right = y = \text{EXTRACT-MIN}(Q)$ 
7       $z.freq = x.freq + y.freq$ 
8       $\text{INSERT}(Q, z)$ 
9  return  $\text{EXTRACT-MIN}(Q)$     // return the root of the tree
```



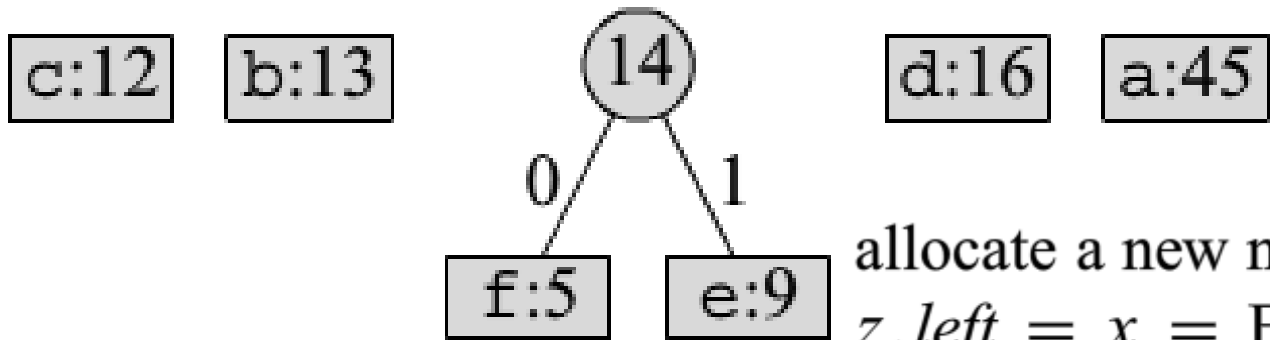
# Huffman Algorithm

## Example

f:5	e:9	c:12	b:13	d:16	a:45
-----	-----	------	------	------	------

- Initial Characters w/ Frequencies

# Huffman Algorithm Example



allocate a new node  $z$

$z.left = x = \text{EXTRACT-MIN}(Q)$

$z.right = y = \text{EXTRACT-MIN}(Q)$

$z.freq = x.freq + y.freq$

$\text{INSERT}(Q, z)$

- First Iteration

- Allocate Node

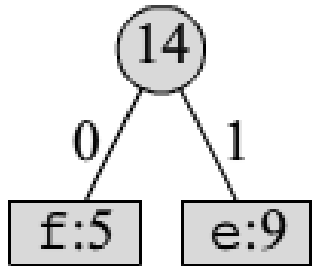
- Extract-Min from  $Q$  for Left: [f:5]

- Extract-Min from  $Q$  for Right: [e:9]

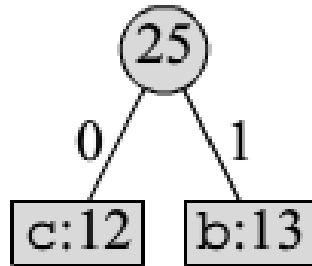
- Sum Frequencies and add New Node to  $Q$

- [14: [f:5], [e:9]]

# Huffman Algorithm Example



d:16



a:45

allocate a new node  $z$

$z.left = x = \text{EXTRACT-MIN}(Q)$

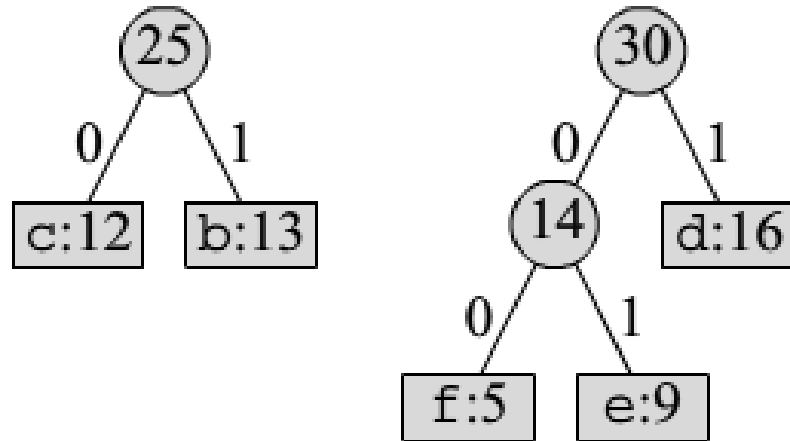
$z.right = y = \text{EXTRACT-MIN}(Q)$

$z.freq = x.freq + y.freq$

$\text{INSERT}(Q, z)$

- Second Iteration
  - Allocate Node
  - Extract-Min from Q for Left: [c:12]
  - Extract-Min from Q for Right: [b:13]
  - Sum Frequencies and add New Node to Q
    - [25: [c:12], [b:13]]

# Huffman Algorithm Example



a:45

allocate a new node  $z$

$z.left = x = \text{EXTRACT-MIN}(Q)$

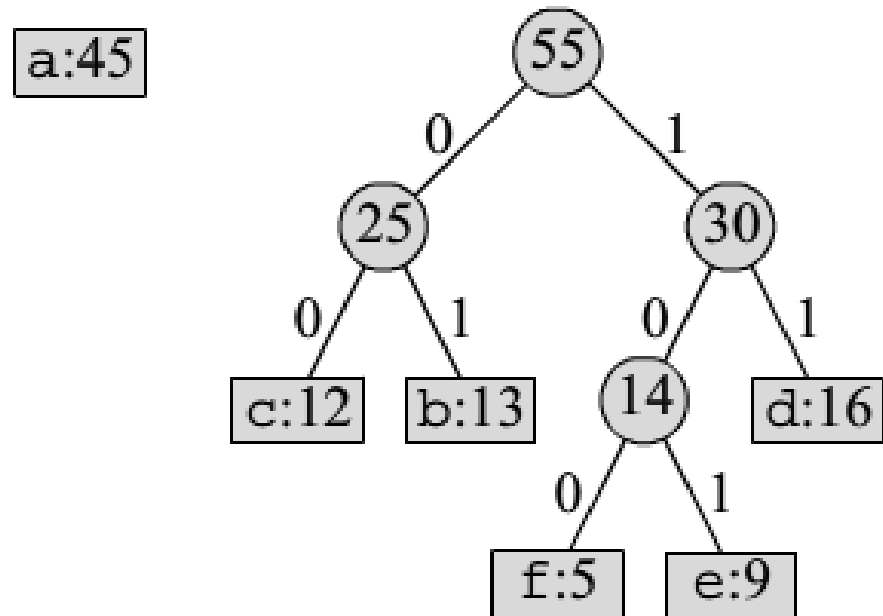
$z.right = y = \text{EXTRACT-MIN}(Q)$

$z.freq = x.freq + y.freq$

$\text{INSERT}(Q, z)$

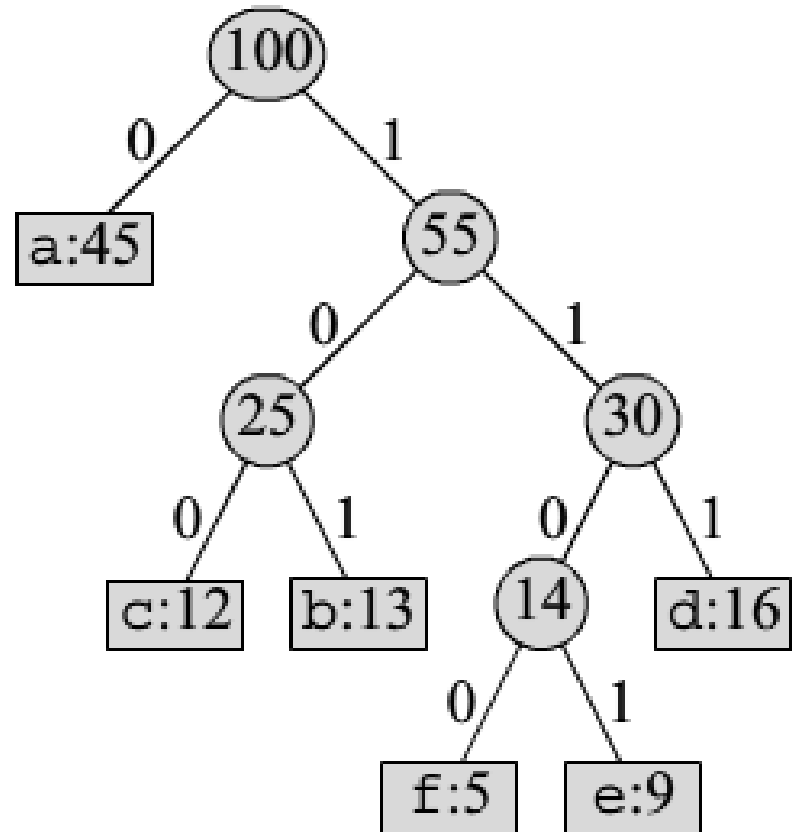
- Third Iteration
  - Allocate Node
  - Extract-Min from Q for Left: [14: [f:5], [e:9]]
  - Extract-Min from Q for Right: [d:16]
  - Sum Frequencies and add New Node to Q
    - [30: [14: [f:5], [e:9]], [d:16]]

# Huffman Algorithm Example



- Fourth Iteration
  - Allocate Node
  - Extract-Min from Q for Left:
    - [25: [c:12], [b:13]]
  - Extract-Min from Q for Right:
    - [30: [14: [f:5], [e:9]], [d:16]]
  - Sum Frequencies and add New Node to Q
    - [55: [25: [c:12], [b:13]], [30: [14: [f:5], [e:9]], [d:16]]]

# Huffman Algorithm Example



- Final Iteration
  - Allocate Node
  - Extract-Min from Q for Left:
    - [a: 45]
  - Extract-Min from Q for Right:
    - [55: [25: [c:12], [b:13]], [30: [14: [f:5], [e:9]], [d:16]]]
  - Sum Frequencies and add New Node to Q
    - [100: [a:45], [55: [25: [c:12], [b:13]], [30: [14: [f:5], [e:9]], [d:16]]]]

# Proof of Correctness

- Lemma 16.2
  - Let  $C$  be an alphabet
    - each character  $c \in C$  has frequency  $c.\text{freq.}$
  - Let  $x$  and  $y$  be two characters in  $C$  having the lowest frequencies.
    - Then there exists an optimal prefix code for  $C$  in which the code words for  $x$  and  $y$  have the same length and differ only in the last bit.

# Proof of Correctness

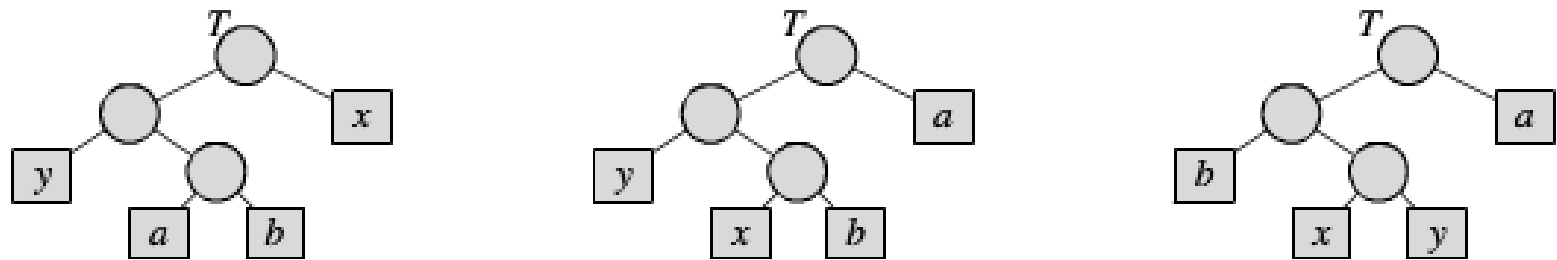
- Proof
  - Take the tree  $T$  representing an arbitrary optimal prefix code
  - Modify  $T$  to make a tree representing another optimal prefix code such that
    - the lowest frequency characters  $x$  and  $y$  appear as sibling leaves of maximum depth in the new tree.
  - If we can construct such a tree, then the code words for  $x$  and  $y$  will have the same length and differ only in the last bit.



# Bits required to encode a file

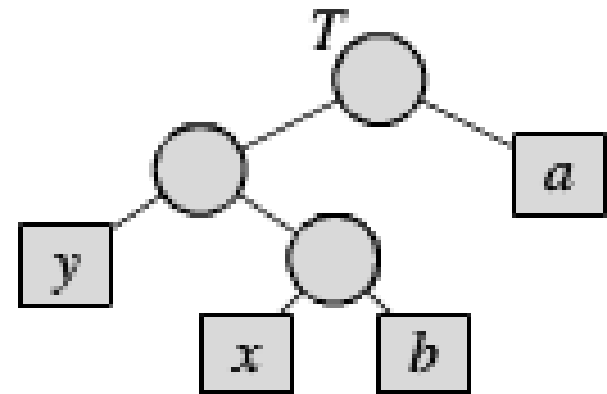
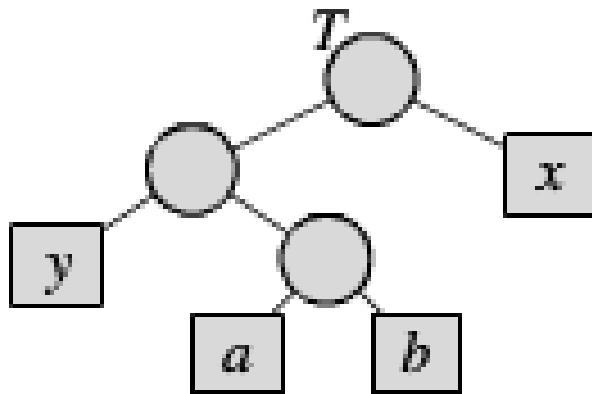
$$B(T) = \sum_{c \in C} c.freq \cdot d_T(c) , \quad (16.4)$$

- $d_t(c)$  = depth of character  $c$  in tree.
- $d_t(c)$  = length of codeword for  $c$ .



**Figure 16.6** An illustration of the key step in the proof of Lemma 16.2. In the optimal tree  $T$ , leaves  $a$  and  $b$  are two siblings of maximum depth. Leaves  $x$  and  $y$  are the two characters with the lowest frequencies; they appear in arbitrary positions in  $T$ . Assuming that  $x \neq b$ , swapping leaves  $a$  and  $x$  produces tree  $T'$ , and then swapping leaves  $b$  and  $y$  produces tree  $T''$ . Since each swap does not increase the cost, the resulting tree  $T''$  is also an optimal tree.

$$\begin{aligned}
 & B(T) - B(T') \\
 &= \sum_{c \in C} c.\text{freq} \cdot d_T(c) - \sum_{c \in C} c.\text{freq} \cdot d_{T'}(c) \\
 &= x.\text{freq} \cdot d_T(x) + a.\text{freq} \cdot d_T(a) - x.\text{freq} \cdot d_{T'}(x) - a.\text{freq} \cdot d_{T'}(a) \\
 &= x.\text{freq} \cdot d_T(x) + a.\text{freq} \cdot d_T(a) - x.\text{freq} \cdot d_T(a) - a.\text{freq} \cdot d_T(x) \\
 &= (a.\text{freq} - x.\text{freq})(d_T(a) - d_T(x)) \\
 &\geq 0,
 \end{aligned}$$



$$\begin{aligned}
 B(T) - B(T') &= \sum_{c \in C} c.freq \cdot d_T(c) - \sum_{c \in C} c.freq \cdot d_{T'}(c) \\
 &= x.freq \cdot d_T(x) + a.freq \cdot d_T(a) - x.freq \cdot d_{T'}(x) - a.freq \cdot d_{T'}(a) \\
 &= x.freq \cdot d_T(x) + a.freq \cdot d_T(a) - x.freq \cdot d_T(a) - a.freq \cdot d_T(x) \\
 &= (a.freq - x.freq)(d_T(a) - d_T(x)) \\
 &\geq 0,
 \end{aligned}$$

- a is at maximum depth
- Since frequency of x is lower than a
  - Swaping x and a CANNOT increase bits required to encode!

# Implications of Lemma 16.2

- Lemma 16.2 implies that building up an optimal tree by mergers can begin with the greedy choice of merging together those two characters of lowest frequency.
- Why is this a greedy choice?
  - We can view the cost of a single merger as being the sum of the frequencies of the two items being merged.

# Lemma 16.3

- Let  $C$  be a given alphabet with frequency  $c.\text{freq}$  defined for each character  $c \in C$ .
- Let  $x$  and  $y$  be two characters in  $C$  with minimum frequency.
  - Let  $C'$  be the alphabet  $C$  with the characters  $x$  and  $y$  removed and a new character  $z$  added, so that  $C' = C - \{x, y\} \cup \{z\}$ .
- Define  $\text{freq}$  for  $C'$  as for  $C$ , except
  - $z.\text{freq} = x.\text{freq} + y.\text{freq}$ .
- Let  $T'$  be any tree representing an optimal prefix code for the alphabet  $C'$ .
  - Then the tree  $T$ , obtained from  $T'$  by replacing the leaf node for  $z$  with an internal node having  $x$  and  $y$  as children, represents an optimal prefix code for the alphabet  $C$ .

# Theorem 16.4

- Procedure HUFFMAN produces an optimal prefix code.
- Proof: Follows immediately from Lemma 16.2 & 16.3