### **Huffman Codes**

- Widely used technique for data compression
- Assume the data to be a sequence of characters
- Looking for an effective way of storing the data
- Binary character code
  - Uniquely represents a character by a binary string

# Fixed-Length Codes

#### E.g.: Data file containing 100,000 characters

	а	b	С	d	е	f
Frequency (thousands)	45	13	12	16	9	5

- 3 bits needed
- a = 000, b = 001, c = 010, d = 011, e = 100, f = 101
- Requires:  $100,000 \cdot 3 = 300,000$  bits

### **Huffman Codes**

#### Idea:

 Use the frequencies of occurrence of characters to build a optimal way of representing each character

	a	b	С	d	е	f
Frequency (thousands)	45	13	12	16	9	5

# Variable-Length Codes

E.g.: Data file containing 100,000 characters

	а	b	С	d	е	f
Frequency (thousands)	45	13	12	16	9	5

- Assign short codewords to frequent characters and long codewords to infrequent characters
- a = 0, b = 101, c = 100, d = 111, e = 1101, f = 1100
- $(45 \cdot 1 + 13 \cdot 3 + 12 \cdot 3 + 16 \cdot 3 + 9 \cdot 4 + 5 \cdot 4) \cdot 1,000$ 
  - = 224,000 bits

### **Prefix Codes**

- Prefix codes:
  - Codes for which no codeword is also a prefix of some other codeword
  - Better name would be "prefix-free codes"
- We can achieve optimal data compression using prefix codes
  - We will restrict our attention to prefix codes

### **Encoding with Binary Character Codes**

### Encoding

 Concatenate the codewords representing each character in the file

#### • E.g.:

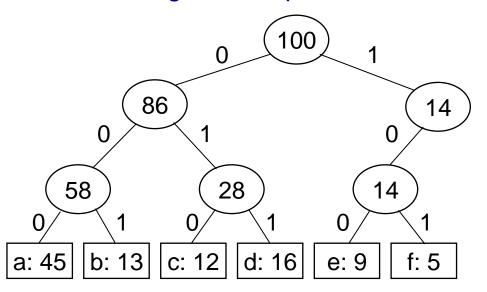
- -a = 0, b = 101, c = 100, d = 111, e = 1101, f = 1100
- $abc = 0 \cdot 101 \cdot 100 = 0101100$

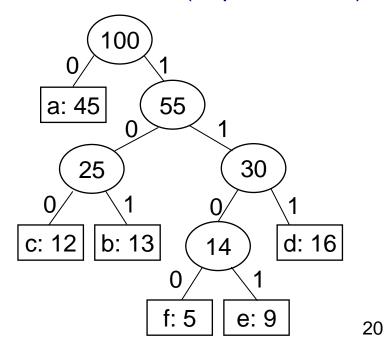
### Decoding with Binary Character Codes

- Prefix codes simplify decoding
  - No codeword is a prefix of another ⇒ the codeword that begins an encoded file is unambiguous
- Approach
  - Identify the initial codeword
  - Translate it back to the original character
  - Repeat the process on the remainder of the file
- *E.g.*:
  - -a = 0, b = 101, c = 100, d = 111, e = 1101, f = 1100
  - $-001011101 = 0.0 \cdot 101 \cdot 1101 = aabe$

# Prefix Code Representation

- Binary tree whose leaves are the given characters
- Binary codeword
  - the path from the root to the character, where 0 means "go to the left child" and 1 means "go to the right child"
- Length of the codeword
  - Length of the path from root to the character leaf (depth of node)





### **Optimal Codes**

- An optimal code is always represented by a full binary tree
  - Every non-leaf has two children
  - Fixed-length code is not optimal, variable-length is
- How many bits are required to encode a file?
  - Let C be the alphabet of characters
  - Let f(c) be the frequency of character c
  - Let d<sub>T</sub>(c) be the depth of c's leaf in the tree T corresponding to a prefix code

$$B(T) = \sum_{c \in C} f(c)d_T(c)$$
 the cost of tree T

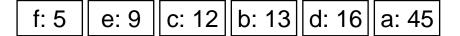
### Constructing a Huffman Code

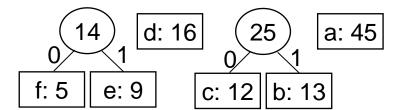
- A greedy algorithm that constructs an optimal prefix code called a Huffman code
- Assume that:
  - C is a set of n characters
  - Each character has a frequency f(c)
  - The tree T is built in a bottom up manner
- Idea:

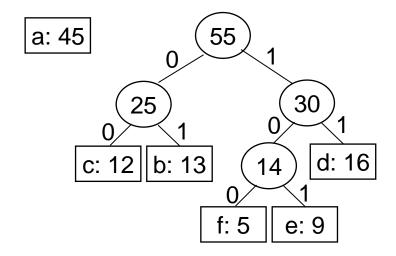
f: 5 e: 9 c: 12 b: 13 d: 16 a: 45

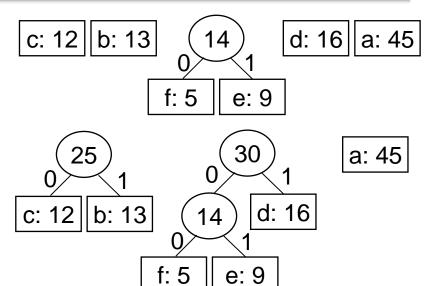
- Start with a set of |C| leaves
- At each step, merge the two least frequent objects: the frequency of the new node = sum of two frequencies
- Use a min-priority queue Q, keyed on f to identify the two least frequent objects

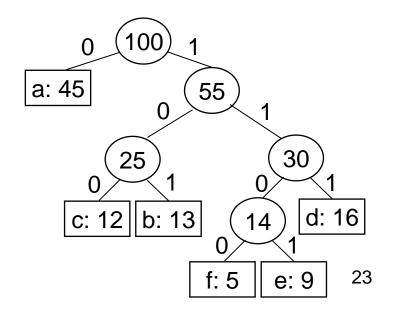
# Example











# Building a Huffman Code

```
Alg.: HUFFMAN(C)
                                    Running time: O(nlgn)
1. n \leftarrow |C|
2. Q \leftarrow C
                                          O(n)
3. for i \leftarrow 1 to n-1
        do allocate a new node z
            left[z] \leftarrow x \leftarrow EXTRACT-MIN(Q)
5.
                                                          O(nlgn)
           right[z] \leftarrow y \leftarrow EXTRACT-MIN(Q)
6.
           f[z] \leftarrow f[x] + f[y]
7.
            INSERT (Q, z)
8.
9. return EXTRACT-MIN(Q)
```