

# Lecture 17: Huffman Coding

CLRS- 16.3

## Outline of this Lecture

- Codes and Compression.
- Huffman coding.
- Correctness of the Huffman coding algorithm.

Suppose that we have a 100,000 character data file that we wish to store. The file contains only 6 characters, appearing with the following frequencies:

	<i>a</i>	<i>b</i>	<i>c</i>	<i>d</i>	<i>e</i>	<i>f</i>
Frequency in '000s	45	13	12	16	9	5

A *binary code* encodes each character as a binary string or *codeword*. We would like to find a binary code that encodes the file using as few bits as possible, ie., *compresses it* as much as possible.

In a *fixed-length code* each codeword has the same length. In a *variable-length code* codewords may have different lengths. Here are examples of fixed and variable length codes for our problem (note that a fixed-length code must have at least 3 bits per codeword).

	<i>a</i>	<i>b</i>	<i>c</i>	<i>d</i>	<i>e</i>	<i>f</i>
Freq in '000s	45	13	12	16	9	5
a fixed-length	000	001	010	011	100	101
a variable-length	0	101	100	111	1101	1100

The *fixed length-code* requires **300,000** bits to store the file. The *variable-length code* uses only

$$(5 \cdot 1 + 13 \cdot 3 + 12 \cdot 3 + 16 \cdot 3 + 9 \cdot 4 + 5 \cdot 4) \cdot 1000 = 224,000 \text{ bits,}$$

saving a lot of space! Can we do better?

Note: In what follows a *code* will be a set of codewords, e.g.,

{000, 001, 010, 011, 100, 101}

and {0, 101, 100, 111, 1101, 1100}

## Encoding

Given a code (corresponding to some alphabet  $\Gamma$ ) and a message it is easy to *encode* the message. Just replace the characters by the codewords.

Example:  $\Gamma = \{a, b, c, d\}$

If the code is

$$C_1\{a = 00, b = 01, c = 10, d = 11\}.$$

then **bad** is encoded into **010011**

If the code is

$$C_2 = \{a = 0, b = 110, c = 10, d = 111\}$$

then **bad** is encoded into **1101111**

## Decoding

$$C_1 = \{a = 00, b = 01, c = 10, d = 11\}.$$

$$C_2 = \{a = 0, b = 110, c = 10, d = 111\}.$$

$$C_3 = \{a = 1, b = 110, c = 10, d = 111\}$$

Given an encoded message, *decoding* is the process of turning it back into the original message.

A message is *uniquely decodable* (vis-a-vis a particular code) if it can only be decoded in one way.

For example relative to  $C_1$ , **010011** is uniquely decodable to **bad**.

Relative to  $C_2$  **1101111** is uniquely decodable to **bad**.

But, relative to  $C_3$ , **1101111** is **not** uniquely decipherable since it could have encoded either **bad** or **acad**.

In fact, one can show that every message encoded using  $C_1$  and  $C_2$  are uniquely decipherable. The unique decipherability property is needed in order for a code to be useful.

## Prefix-Codes

Fixed-length codes are always uniquely decipherable (why).

We saw before that these do not always give the best compression so we prefer to use variable length codes.

**Prefix Code:** A code is called a **prefix (free) code** if no codeword is a prefix of another one.

**Example:**  $\{a = 0, b = 110, c = 10, d = 111\}$  is a prefix code.

**Important Fact:** Every message encoded by a prefix free code is uniquely decipherable. Since no codeword is a prefix of any other we can always find the first codeword in a message, peel it off, and continue decoding. Example:

$$01101100 = 01101100 = abba$$

We are therefore interested in finding good (best compression) prefix-free codes.

## Fixed-Length versus Variable-Length Codes

**Problem:** Suppose we want to store messages made up of 4 characters  $a, b, c, d$  with frequencies 60, 5, 30, 5 (percents) respectively. What are the fixed-length codes and prefix-free codes that use the least space?

## Fixed-Length versus Variable-Length Prefix Codes

### Solution:

characters	<i>a</i>	<i>b</i>	<i>c</i>	<i>d</i>
frequency	60	5	30	5
fixed-length code	00	01	10	11
prefix code	0	110	10	111

To store 100 of these characters,

(1) the fixed-length code requires  $100 \times 2 = 200$  bits,

(2) the prefix code uses only

$$60 \times 1 + 5 \times 3 + 30 \times 2 + 5 \times 3 = 150$$

a 25% saving.

**Remark:** We will see later that this is the *optimum* (lowest cost) prefix code.



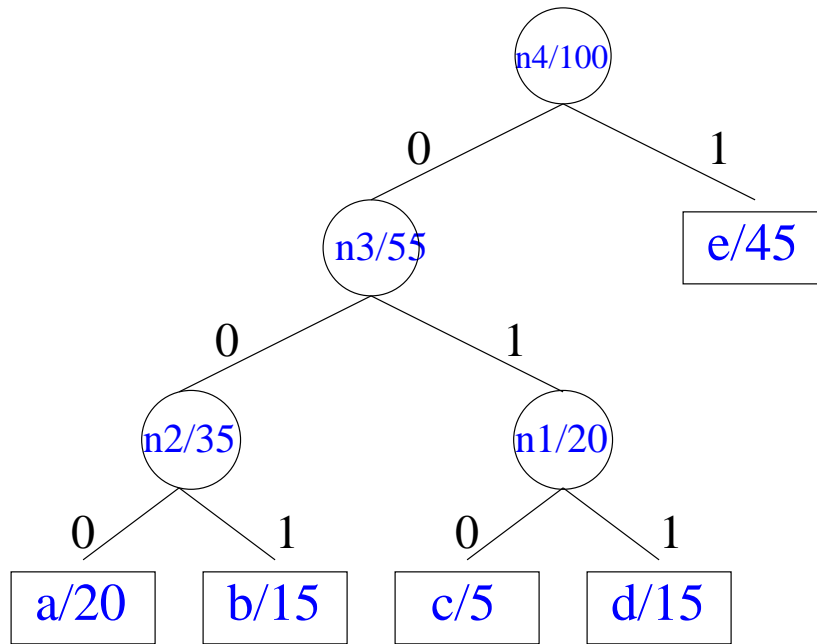
## Optimum Source Coding Problem

**The problem:** Given an **alphabet**  $A = \{a_1, \dots, a_n\}$  with **frequency distribution**  $f(a_i)$  find a binary prefix code  $C$  for  $A$  that minimizes the number of bits

$$B(C) = \sum_{a=1}^n f(a_i) L(c(a_i))$$

needed to encode a message of  $\sum_{a=1}^n f(a)$  characters, where  $c(a_i)$  is the codeword for encoding  $a_i$ , and  $L(c(a_i))$  is the length of the codeword  $c(a_i)$ .

**Remark:** Huffman developed a nice greedy algorithm for solving this problem and producing a minimum-cost (optimum) prefix code. The code that it produces is called a **Huffman code**.



$\{a = 000, b = 001, c = 010, d = 011, e = 1\}$

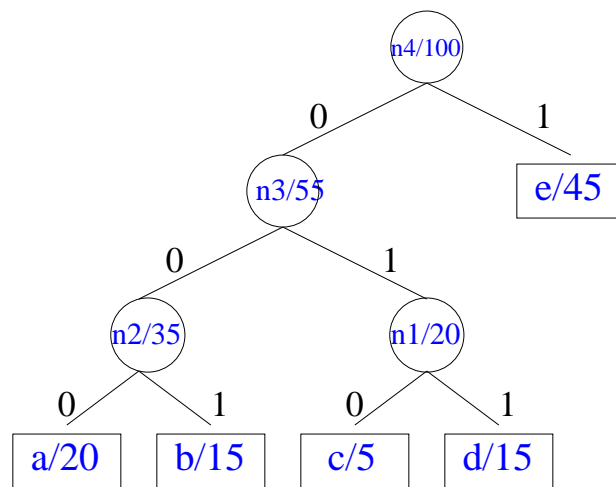
### Correspondence between Binary Trees and prefix codes.

1-1 correspondence between leaves and characters.

Label of leaf is frequency of character.

Left edge is labelled 0; right edge is labelled 1

Path from root to leaf is codeword associated with character.



$$\{a = 000, b = 001, c = 010, d = 011, e = 1\}$$

Note that  $d(a_i)$ , the depth of leaf  $a_i$  in tree  $T$  is equal to the length,  $L(c(a_i))$  of the codeword in code  $C$  associated with that leaf. So,

$$\sum_{a=1}^n f(a_i)L(c(a_i)) = \sum_{a=1}^n f(a_i)d(a_i).$$

The sum  $\sum_{a=1}^n f(a_i)d(a_i)$  is the *weighted external pathlength* of tree  $T$ .

The Huffman encoding problem is equivalent to the **minimum-weight external pathlength problem**: given weights  $f(a_1), \dots, f(a_n)$ , find tree  $T$  with  $n$  leaves labelled  $a_1, \dots, a_n$  that has minimum weighted external path length.

## Huffman Coding

**Step 1:** Pick two letters  $x, y$  from alphabet  $A$  with the smallest frequencies and create a subtree that has these two characters as leaves. (greedy idea)

Label the root of this subtree as  $z$ .

**Step 2:** Set frequency  $f(z) = f(x) + f(y)$ .

Remove  $x, y$  and add  $z$  creating new alphabet

$$A' = A \cup \{z\} - \{x, y\}.$$

Note that  $|A'| = |A| - 1$ .

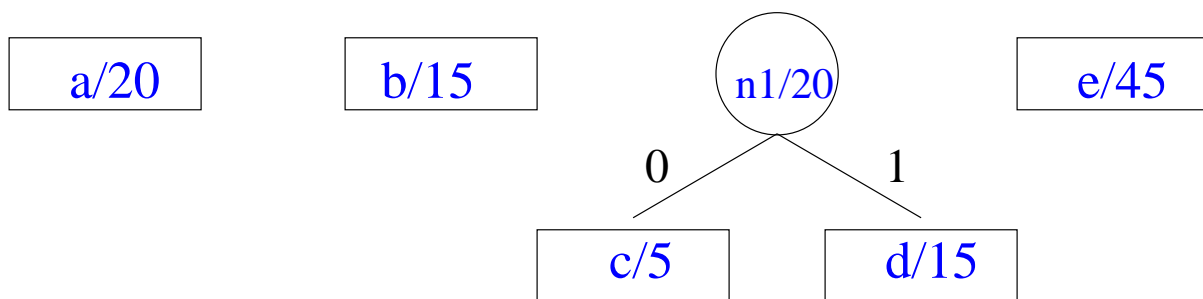
Repeat this procedure, called *merge*, with new alphabet  $A'$  until an alphabet with only one symbol is left.

The resulting tree is the **Huffman code**.

## Example of Huffman Coding

Let  $A = \{a/20, b/15, c/5, d/15, e/45\}$  be the alphabet and its frequency distribution.

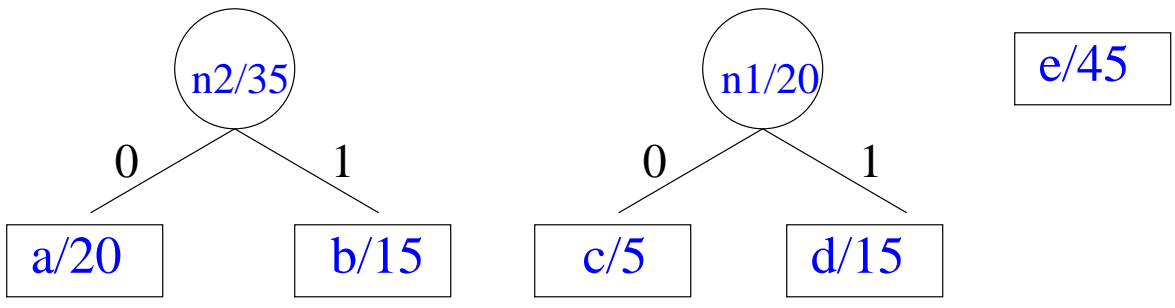
In the first step Huffman coding merges  $c$  and  $d$ .



Alphabet is now  $A_1 = \{a/20, b/15, n1/20, e/45\}$ .

## Example of Huffman Coding – Continued

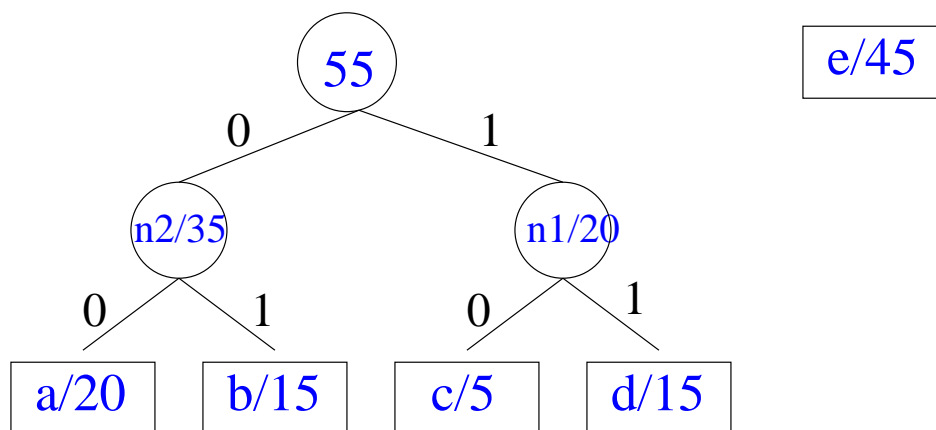
Alphabet is now  $A_1 = \{a/20, b/15, n1/20, e/45\}$ .  
Algorithm merges  $a$  and  $b$   
(could also have merged  $n1$  and  $b$ ).



New alphabet is  $A_2 = \{n2/35, n1/20, e/45\}$ .

## Example of Huffman Coding – Continued

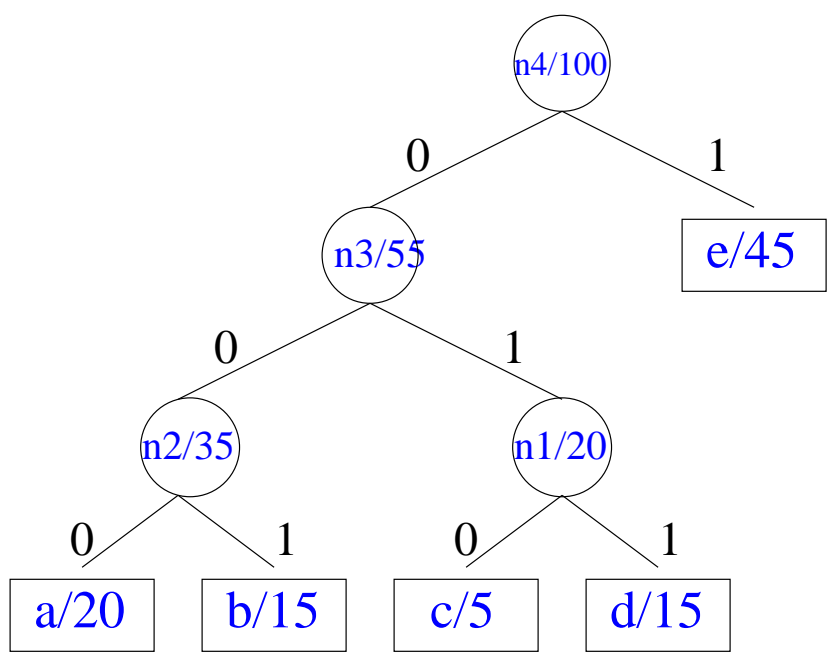
Alphabet is  $A_2 = \{n_2/35, n_1/20, e/45\}$ .  
Algorithm merges  $n_1$  and  $n_2$ .



New alphabet is  $A_3 = \{n_3/55, e/45\}$ .

## Example of Huffman Coding – Continued

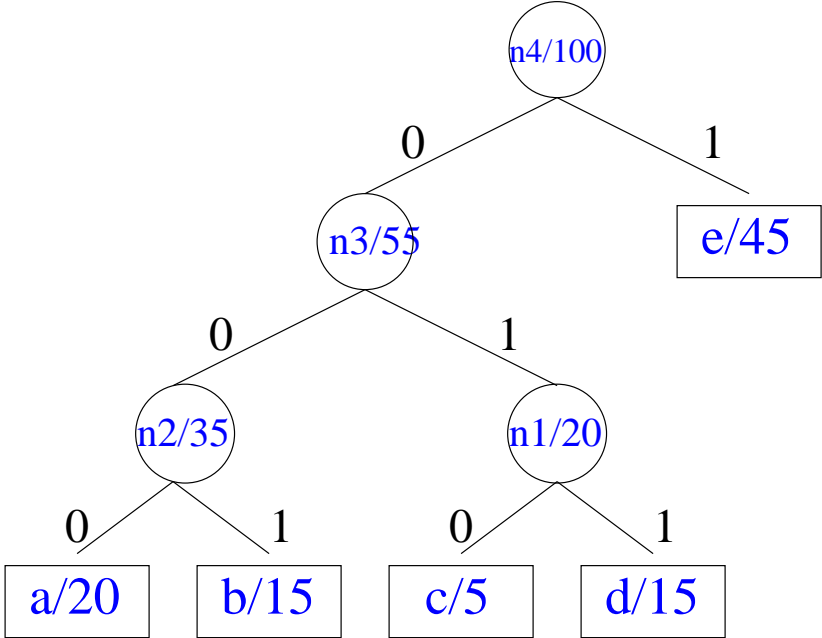
Current alphabet is  $A_3 = \{n_3/55, e/45\}$ .  
Algorithm merges  $e$  and  $n_3$  and finishes.





## Example of Huffman Coding – Continued

Huffman code is obtained from the Huffman tree.



Huffman code is

$a = 000, b = 001, c = 010, d = 011, e = 1.$

This is the optimum (minimum-cost) prefix code for this distribution.

## Huffman Coding Algorithm

Given an alphabet  $A$  with frequency distribution  $\{f(a) : a \in A\}$ . The binary Huffman tree is constructed using a priority queue,  $Q$ , of nodes, with labels (frequencies) as keys.

Huffman( $A$ )

```
{   $n = |A|$ ;  
    $Q = A$ ;           the future leaves  
   for  $i = 1$  to  $n - 1$    Why  $n - 1$ ?  
   {   $z =$  new node;  
       $left[z] = \text{Extract-Min}(Q)$ ;  
       $right[z] = \text{Extract-Min}(Q)$ ;  
       $f[z] = f[left[z]] + f[right[z]]$ ;  
       $\text{Insert}(Q, z)$ ;  
   }  
   return  $\text{Extract-Min}(Q)$   root of the tree  
}
```

Running time is  $O(n \log n)$ , as each priority queue operation takes time  $O(\log n)$ .

## Huffman Codes are Optimal

**Lemma:** Consider the two letters,  $x$  and  $y$  with the **smallest frequencies**. Then is an optimal code tree in which these two letters are sibling leaves in the tree in the lowest level.

**Proof:** Let  $T$  be an optimum prefix code tree, and let  $b$  and  $c$  be two siblings at the maximum depth of the tree (must exist because  $T$  is full). Assume without loss of generality that  $f(b) \leq f(c)$  and  $f(x) \leq f(y)$  (if this is not true, then rename these characters). Since  $x$  and  $y$  have the two smallest frequencies it follows that  $f(x) \leq f(b)$  (they may be equal) and  $f(y) \leq f(c)$  (may be equal). Because  $b$  and  $c$  are at the deepest level of the tree we know that  $d(b) \geq d(x)$  and  $d(c) \geq d(y)$ .

Now switch the positions of  $x$  and  $b$  in the tree resulting in a different tree  $T'$  and see how the cost changes. Since  $T$  is optimum,

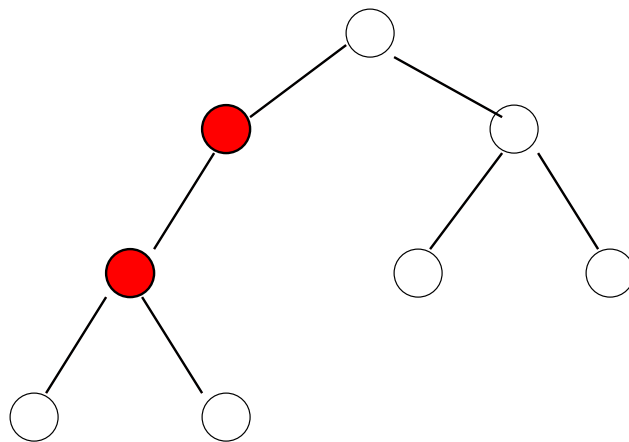
$$\begin{aligned} B(T) &\leq B(T') \\ &= B(T) - f(x)d(x) - f(b)d(b) + f(x)d(b) + f(b)d(x) \\ &= B(T) - (f(b) - f(x))(d(b) - d(x)) \\ &\leq B(T). \end{aligned}$$

Therefore,  $B(T') = B(T)$ , that is,  $T'$  is an optimum tree. By switching  $y$  with  $c$  we get a new tree  $T''$  which by a similar argument is optimum. The final tree  $T''$  satisfies the statement of the claim.

## An Observation: Full Trees

**Lemma:** The tree for any **optimal prefix code** must be “**full**”, meaning that every internal node has exactly two children.

**Proof:** If some internal node had only one child then we could simply get rid of this node and replace it with its unique child. This would decrease the total cost of the encoding.



## Huffman Codes are Optimal

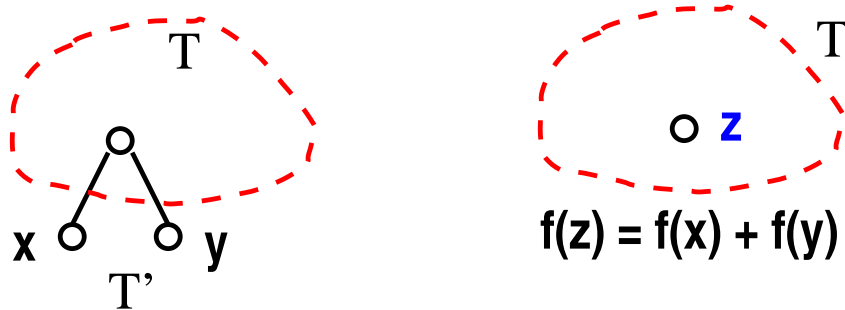
**Theorem:** Huffman's algorithm produces an optimum prefix code tree.

**Proof:** By induction on  $n$ . When  $n = 2$ , obvious.

Assume inductively that with strictly fewer than  $n$  letters, Huffman's algorithm is guaranteed to produce an optimum tree. We want to show this is also true with exactly  $n$  letters.

## Huffman Codes are Optimal

**Proof – continued:** Consider an alphabet  $A'$  of  $n$  letters. Let  $T'$  be an optimum tree for  $A'$  with the two letters of lowest frequency  $x$  and  $y$  as sibling leaves (exists by Lemma). Let  $T$  be the coding tree for  $A = A' \cup \{z\} - \{x, y\}$  ( $n - 1$  leaves) obtained by removing  $x$  and  $y$  and replacing their parent by  $z$ .



$d(x) = d(y) = d(z) - 1$  and  $f(z) = f(x) + f(y)$ ,  
so

$$\begin{aligned} B(T) &= B(T') - f(x)d(x) - f(y)d(y) + f(z)d(z) \\ &= B(T') - (f(x) + f(y)). \end{aligned}$$

## Huffman Codes are Optimal

**Proof – continued:** By the induction hypotheses, the Huffman algorithm gives a Huffman code tree  $H_A$  that is optimal for  $A$ . Let  $H_{A'}$  be the tree obtained by adding  $x$  and  $y$  as children of  $z$  in  $H_A$ . Note that, as in the calculations for  $T$  and  $T'$ , we have:

$$B(H_A) = B(H_{A'}) - (f(x) + f(y)).$$

Also, note that, by the optimality of  $H_A$ , we have that  $B(H_A) \leq B(T)$ , so

$$\begin{aligned} B(H_{A'}) &= B(H_A) + f(x) + f(y) \\ &\leq B(T) + f(x) + f(y) \\ &= B(T'). \end{aligned}$$

Since we started with the assumption that  $T'$  was an optimal tree for  $A'$  this implies that  $H_{A'}$ , which is exactly the tree constructed by the Huffman algorithm, is also an optimal tree for  $A'$  and we are done.