

# Greedy 3: Huffman Codes

Yufei Tao

Department of Computer Science and Engineering  
Chinese University of Hong Kong

## Coding

Suppose that we have an alphabet  $\Sigma$  (like the English alphabet). The goal of coding is to map each alphabet to a binary string—called a **codeword**—so that they can be transmitted electronically.

For example, suppose  $\Sigma = \{a, b, c, d, e, f\}$ . Assume that we agree on  $a = 000$ ,  $b = 001$ ,  $c = 010$ ,  $d = 011$ ,  $e = 100$ , and  $f = 101$ . Then, a letter such as “bed” will be encoded as 001100011.

We can, however, achieve better coding efficiency (i.e., producing shorter digital documents) if the **frequencies** of the letters are known. In general, more frequent letters should be encoded with less bits. The next slide shows an example.

### Example

Suppose we know that the frequencies of  $a, b, c, d, e, f$  are 0.1, 0.2, 0.13, 0.09, 0.4, 0.08, respectively.

If we encode each letter with 3 digits, then the average number of digits per letter is obviously 3.

However, if we adopt the encoding of  $a = 100$ ,  $b = 111$ ,  $c = 101$ ,  $d = 1101$ ,  $e = 0$ ,  $f = 1100$ , the average number of digits per letter is:

$$3 \cdot 0.1 + 3 \cdot 0.2 + 3 \cdot 0.13 + 4 \cdot 0.09 + 1 \cdot 0.4 + 4 \cdot 0.08 = 2.37.$$

So in the long run, the new encoding is expected to save  $1 - (2.37/3) = 21\%$  of bits!

## Example

You probably would ask: why not just encode the letters as:  
 $e = 0, b = 1, c = 00, a = 01, d = 10, f = 11$ —namely, encode the next frequent letter using as few bits as possible?

The answer is: you **cannot decode** a document unambiguously! For example, consider the string 10: how do you know whether this is two letters “be”, or just one letter “d”?

This issue arises because the codeword of a letter happens to be a **prefix** of the codeword of another letter. We, therefore, should prevent this, which has led to an important class of codes in coding theory: the **prefix codes** (actually “prefix-free” codes would have been more appropriate, but the name “prefix codes” has become a standard).

### Example

Consider once again our earlier encoding:  $a = 100$ ,  $b = 111$ ,  $c = 101$ ,  $d = 1101$ ,  $e = 0$ ,  $f = 1100$ . Observe that the encoding is “prefix free”, and hence, allows unambiguous decoding.

For example, what does the following binary string say?

10011010100110011011001101

## The Prefix Coding Problem

An encoding of the letters in an alphabet  $\Sigma$  is a **prefix code** if no codeword is a prefix of another codeword.

For each letter  $\sigma \in \Sigma$ , let  $\text{freq}(\sigma)$  denote the **frequency** of  $\sigma$ . Also, denote by  $l(\sigma)$  the number of bits in the codeword of  $\sigma$ .

Given an encoding, its **average length** is calculated as

$$\sum_{\sigma \in \Sigma} \text{freq}(\sigma) \cdot l(\sigma).$$

The objective of the **prefix coding problem** is to find a prefix code for  $\Sigma$  that has the smallest average length.

## A Binary Tree View

Let us start to attack the prefix coding problem. A key observation is that **every prefix code can be represented as a binary tree.**

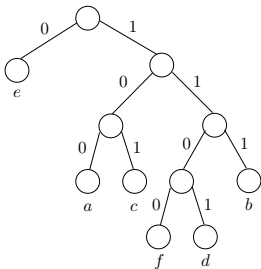
A **code tree** on  $\Sigma$  as a binary tree  $T$  satisfying:

- Every leaf node of  $T$  is labeled with a distinct letter in  $\Sigma$ ; conversely, every letter in  $\Sigma$  is the label of a distinct leaf node in  $T$ .
- For every internal node of  $T$ , its left edge (if exists) is labeled with 0, and its right edge (if exists) with 1.

The codeword of a letter  $\sigma \in \Sigma$  can be obtained by concatenating the bit labels of the edges on the path from the root of  $T$  to the leaf  $\sigma$ .

### Example

Consider once again our earlier encoding:  $a = 100$ ,  $b = 111$ ,  $c = 101$ ,  $d = 1101$ ,  $e = 0$ ,  $f = 1100$ . The following is the corresponding code tree:



**Think:** Why must every letter be at the leaf? (Hint: prefix free)



## Average Length from the Binary Tree

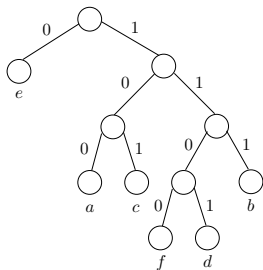
Let  $T$  be the code tree corresponding to a prefix code.

Given a letter  $\sigma$  of  $\Sigma$ , denote by  $d(\sigma)$  its **depth**, which is the **level** of its leaf node in  $T$  (i.e., the number of edges on the path from the root to the leaf).

Clearly, the average length of the prefix code equals

$$\sum_{\sigma \in \Sigma} d(\sigma) \cdot \text{freq}(\sigma).$$

## Example



The depths of  $e, a, c, f, d, b$  are 1, 3, 3, 4, 4, 3, respectively. The average length of the encoding equals

$$\text{freq}(e) \cdot 1 + \text{freq}(a) \cdot 3 + \text{freq}(c) \cdot 3 + \text{freq}(f) \cdot 4 + \text{freq}(d) \cdot 4 + \text{freq}(b) \cdot 3.$$

## Huffman's Algorithm

Next, we will present a surprisingly simple algorithm for solving the prefix coding problem. The algorithm constructs a code tree in a bottom-up manner.

Let  $n = |\Sigma|$ . At the beginning, there are  $n$  stand-alone nodes, each corresponding to a different letter in  $\Sigma$ . If letter  $\sigma$  corresponds to a node  $z$ , set the **frequency** of  $z$  to  $freq(\sigma)$ .

Let  $S$  be the set of these  $n$  nodes.

## Huffman's Algorithm

Then, the algorithm repeats the following until  $S$  has a single node left:

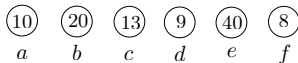
1. Remove from  $S$  two nodes  $u_1, u_2$  with the smallest frequencies.
2. Create a node  $v$  with  $u_1, u_2$  as its children. Set the **frequency** of  $v$  to be the frequency sum of  $u_1$  and  $u_2$ .
3. Insert  $v$  into  $S$ .

When  $S$  has only one node left, we have already obtained the target code tree. The prefix code derived from this code tree is known as a **Huffman code**.

### Example

Consider our earlier example where the frequencies of  $a, b, c, d, e, f$  are 0.1, 0.2, 0.13, 0.09, 0.4, 0.08, respectively.

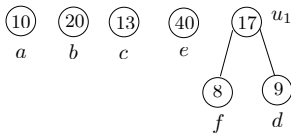
At the beginning,  $S$  has 6 nodes:



The number in each circle represents the frequency of each node (e.g., 10 means 10%).

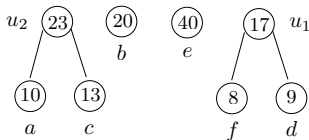
### Example

Merge the two nodes with the smallest frequencies 8 and 9. Now  $S$  has 5 nodes  $\{a, b, c, e, u_1\}$ :



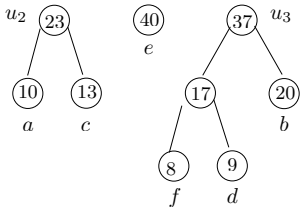
### Example

Merge the two nodes with the smallest frequencies 10 and 13. Now  $S$  has 4 nodes  $\{b, e, u_1, u_2\}$ :



### Example

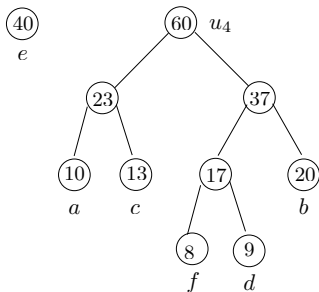
Merge the two nodes with the smallest frequencies 17 and 20. Now  $S$  has 3 nodes  $\{e, u_1, u_3\}$ :





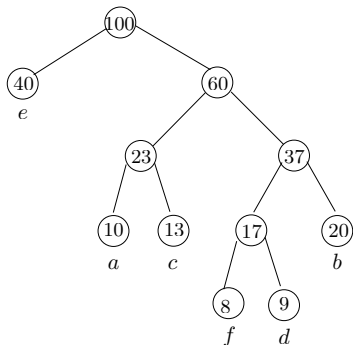
### Example

Merge the two nodes with the smallest frequencies 23 and 37. Now  $S$  has 2 nodes  $\{e, u_4\}$ :



## Example

Merge the two remaining nodes. Now  $S$  has a single node left.



This is the final code tree.

It should be fairly straightforward for you to implement the algorithm in  $O(n \log n)$  time, where  $n = |\Sigma|$ .

Think: Why do we say the algorithm is **greedy**?

Next, we prove that the algorithm indeed gives an **optimal prefix code**, i.e., one that has the smallest average length among all the possible prefix codes.

## Crucial Property 1

**Lemma:** Let  $T$  be a code tree corresponds to an optimal prefix code. Then, every internal node of  $T$  must have two children.

**Proof:** Suppose that the lemma is not true. Then, there is an internal node  $u$  with only one child node  $v$ . Imagine removing  $u$  as follows:

- If  $u$  is the root, simply make  $v$  the new root.
- Otherwise, make  $v$  a child node of the parent of  $u$ .

The above removal generates a new binary tree whose average length is smaller than that of  $T$ , which contradicts the fact that  $T$  is optimal.  $\square$ .

## Crucial Property 2

**Lemma:** Let  $\sigma_1$  and  $\sigma_2$  be two letters in  $\Sigma$  with the lowest frequencies. There exists an optimal prefix code whose code tree has  $\sigma_1$  and  $\sigma_2$  as two sibling leaves at the deepest level.

**Proof:** Take an arbitrary prefix code with binary tree  $T$ . If  $\sigma_1$  and  $\sigma_2$  are indeed sibling leaves at the deepest level, then the claim already holds. Next, we assume that this is not the case.

Suppose  $T$  has height  $h$ . Recall that the **level** of a node is the number of edges on the path from the root to the node. In other words, the deepest leaves are at  $h - 1$ . Take an arbitrary internal node  $p$  at level  $h - 2$ . By the previous lemma,  $p$  must have two leaves (at level  $h - 1$ ). Let  $\sigma'_1$  and  $\sigma'_2$  be the letters corresponding to those leaves.

## Crucial Property 2

**Proof (cont.):** Now swap  $\sigma_1$  with  $\sigma'_1$ , and  $\sigma_2$  with  $\sigma'_2$ , which gives a new binary tree  $T'$ . Note that  $T'$  has  $\sigma_1$  and  $\sigma_2$  as sibling leaves at the deepest level.

How does the average length of  $T'$  compare with that of  $T$ ? As the frequency of  $\sigma_1$  is no higher than that of  $\sigma'_1$ , swapping the two letters can only decrease the average length of the tree (i.e., as we are assigning a shorter codeword to a more frequent letter). Similarly, the other swap can only decrease the average length.

It follows that the average length of  $T'$  is no larger than that of  $T$ , meaning that  $T'$  is optimal as well. □

## Optimality of Huffman Coding

We are now ready to prove:

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

**Proof:** We will prove by induction on the size  $n$  of the alphabet  $\Sigma$ .

**Base Case:**  $n = 2$ . In this case, the algorithm encodes one letter with 0, and the other with 1, which is clearly optimal.

**General Case:** Assuming that the theorem holds for  $n = k - 1$  ( $k \geq 3$ ), next we show that it also holds for  $n = k$ .

## Optimality of Huffman Coding

**Proof (cont.):** Let  $\sigma_1$  and  $\sigma_2$  be two letters with the lowest frequencies. From Property 2, we know that there is an **optimal prefix code** whose code tree  $T$  has  $\sigma_1$  and  $\sigma_2$  as two sibling leaves at the deepest level. Let  $p$  be the parent of  $\sigma_1$  and  $\sigma_2$ .

Construct a new alphabet  $\Sigma'$  that includes all letters in  $\Sigma$ , except  $\sigma_1$  and  $\sigma_2$ , but a **letter  $p$  whose frequency equals  $f(\sigma_1) + f(\sigma_2)$** . Let  $T'$  be the tree obtained by removing leaf nodes  $\sigma_1$  and  $\sigma_2$  from  $T$  (thus making  $p$  a leaf).

Let  $T'$  be the binary tree obtained by Huffman's algorithm on  $\Sigma'$ . Since  $|\Sigma'| = k - 1$ , we know from the inductive assumption that  $T'$  is optimal, meaning that

$$\text{avg length of } T' \leq \text{avg length of } T'$$



## Optimality of Huffman Coding

**Proof (cont.):** Now consider the binary tree  $\mathcal{T}$  produced by Huffman's algorithm on  $\Sigma$ . Clearly,  $\mathcal{T}$  extends  $\mathcal{T}'$  by simply putting  $\sigma_1$  and  $\sigma_2$  as child nodes of  $p$ . Hence:

$$\begin{aligned}\text{avg length of } \mathcal{T} &= \text{avg length of } \mathcal{T}' + f(\sigma_1) + f(\sigma_2) \\ &\leq \text{avg length of } \mathcal{T}' + f(\sigma_1) + f(\sigma_2) \\ &= \text{avg length of } \mathcal{T}.\end{aligned}$$

This indicates that  $\mathcal{T}$  also gives an optimal prefix code. □