

# CSCI3160 Design and Analysis of Algorithms (2025 Fall)

## Greedy 3: Huffman Codes

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<sup>1</sup>These slides are primarily based on materials prepared by [Prof. Yufei Tao](#) (please refer to [Prof. Tao's version from 2024 Fall](#) for the original content). Some modifications have been made to better align with this year's teaching progress, incorporating student feedback, in-class interactions, and my own teaching style and research perspective.

Given an alphabet  $\Sigma$  (like the English alphabet), an **encoding** is a function that maps each letter in  $\Sigma$  to a binary string, called a **codeword**.

For example, suppose  $\Sigma = \{a, b, c, d, e, f\}$  and consider the following encoding

$$a = 000, b = 001, c = 010, d = 011, e = 100, f = 101.$$

The word “bed” can be encoded as 001100011.

Think: it is interesting to notice that there is no ambiguity when we decode the word “bed.” Is this always true for any encoding? Or is it due to the special design of the above encoding?

We can reduce the length of encoding if letters' usage frequencies are known.

Suppose that, in a document, 10% of the letters are  $a$ , namely, the letter has **frequency** 10%. Similarly, suppose that letters  $b, c, d, e$ , and  $f$  have frequencies 20%, 13%, 9%, 40%, and 8%, respectively.

If we use the encoding  $a = 100$ ,  $b = 111$ ,  $c = 101$ ,  $d = 1101$ ,  $e = 0$ ,  $f = 1100$ , the **average number** of bits per letter is:

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

This is better than using 3 bits per letter.

However, is this the best we can do? What if there is another encoding that cost less on average? E.g.,

$$e = 0, b = 1, c = 00, a = 01, d = 10, f = 11.$$

What is wrong with the encoding  $e = 0, b = 1, c = 00, a = 01, d = 10, f = 11$ ?

- **Ambiguity in decoding!** For example, does the string 10 mean “be” or “d”?

To allow decoding, we enforce the following constraint:

No letter's codeword should be a prefix of another letter's codeword.

An encoding satisfying the constraint is said to be a **prefix code**.

**Example:** The encoding  $a = 100, b = 111, c = 101, d = 1101, e = 0, f = 1100$  is a prefix code. Just for fun, try decoding the following binary string.

10011010100110011100

## The Prefix Coding Problem

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

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

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

The **prefix coding problem**:

- Given an alphabet  $\Sigma$  and the frequency for each letter in it, find a prefix code for  $\Sigma$  with the shortest average length.

# Prefix Codes and Binary Trees

There is an interesting connection between prefix codes and binary trees. We will utilize this connection later.

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

- Every leaf node of  $T$  corresponds to a unique letter in  $\Sigma$ ; every letter in  $\Sigma$  corresponds to a unique leaf node in  $T$ .
- For every internal node of  $T$ , its left edge (if exists) is labeled 0, and its right edge (if exists) is labeled 1.

$T$  generates a prefix code as follows:

- For each letter  $\sigma \in \Sigma$ , generate its codeword by concatenating the bit labels of the edges on the path from the root of  $T$  to  $\sigma$ .

**Think:** Why must the encoding be a prefix code?

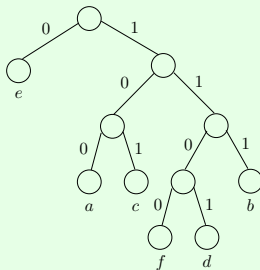
# Prefix Codes and Binary Trees

**Lemma:** Every prefix code is generated by a code tree.

The proof will be left as a regular exercise.

# Prefix Codes and Binary Trees: An Example

**Example:** For our encoding  $a = 100$ ,  $b = 111$ ,  $c = 101$ ,  $d = 1101$ ,  $e = 0$ , and  $f = 1100$ , the code tree is:





# Prefix Codes and Binary Trees

Let  $T$  be the code tree generating a prefix code. Given a letter  $\sigma$  of  $\Sigma$ , its code word length  $len(\sigma)$  is the **level** of its leaf node  $level(\sigma)$  in  $T$  (i.e., the number of edges from the root to node  $\sigma$ ).

Hence:

$$\text{avg length} = \sum_{\sigma \in \Sigma} freq(\sigma) \cdot len(\sigma) = \sum_{\sigma \in \Sigma} freq(\sigma) \cdot level(\sigma) = \text{avg height of } T$$

**Goal (restated):** Find a code tree on  $\Sigma$  with the smallest average height.

# Huffman Coding: An Anecdote I

In 1951, David A. Huffman was a graduate student in electrical engineering at MIT. In a course on information theory taught by Robert M. Fano, students were given a choice for their term project:

- Either write a term paper on the topic of data encoding, or develop a better method of encoding information.

Huffman initially struggled with the problem of finding an optimal way to encode symbols based on their frequency. He spent weeks looking for a solution but was about to give up and start writing the paper instead.

Then, in a moment of insight, he realized that he could construct an optimal prefix code greedily, by building a binary tree from the bottom up, always combining the two least frequent symbols. This method became what we now know as **Huffman coding**.

## Huffman Coding: An Anecdote II

Huffman submitted his algorithm instead of the paper—and his professor, Fano, immediately recognized its significance. Interestingly, Fano himself had been working on a similar problem and had developed **Shannon–Fano** coding, which is less efficient than Huffman’s method.

David Huffman’s homework assignment not only earned him an “A,” it also became one of the most important and widely used algorithms in data compression, used in formats like JPEG, MP3, and ZIP files.

You can access Huffman’s original paper titled [A Method for the Construction of Minimum-Redundancy Codes](#).

# Huffman Coding

## Huffman's Algorithm

We now introduce a simple algorithm for solving the prefix coding problem.

Let  $n = |\Sigma|$ . In the beginning, create a set  $S$  of  $n$  stand-alone leaves, each corresponding to a distinct letter in  $\Sigma$ .

We also define the notion of “frequency” for leaves: If a leaf is corresponding to a letter  $\sigma$ , define the **frequency** of this leaf to be  $\text{freq}(\sigma)$ .

Then, repeat until  $|S| = 1$ :

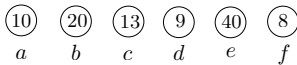
- 1 Remove from  $S$  two nodes  $u_1$  and  $u_2$  with the smallest frequencies.
- 2 Create a node  $v$  with  $u_1$  and  $u_2$  as the children. Set the **frequency** of  $v$  to be the frequency sum of  $u_1$  and  $u_2$ .
- 3 Add  $v$  to  $S$ .

When  $|S| = 1$ , we have obtained a code tree. The prefix code derived from this tree is a **Huffman code**.

### Example

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

Initially,  $S$  has 6 nodes:

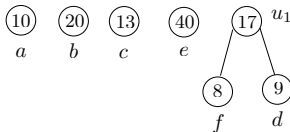


The number in each circle represents frequency (e.g., 10 means 10%).

In short, Huffman's algorithm is to keep merging the two trees with the lowest frequency into a new tree (whose frequency is the sum of the two sub-trees), until there is only one tree left.

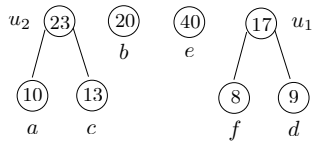
### Example

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



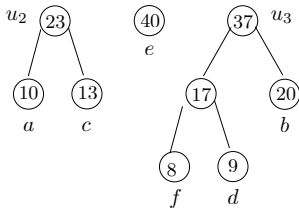
### Example

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



### Example

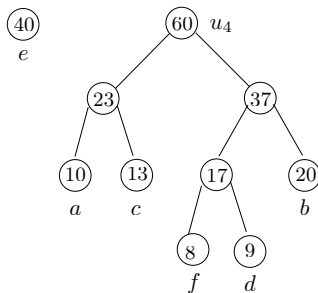
Merge the two nodes with the smallest frequencies 17 and 20. Now  $S$  has 3 nodes  $\{e, u_2, u_3\}$ :





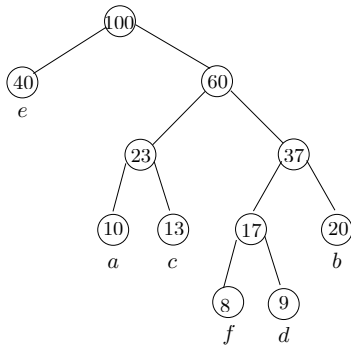
### Example

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 is easy to implement the algorithm in  $O(n \log n)$  time (exercise).

Next, we prove that the algorithm gives an **optimal code tree**, i.e., one that minimizes the average height.

# Proof of Correctness

We now proceed to prove the correctness of Huffman's algorithm.

The proof relies on two important facts about prefix codes. In the following, we first discuss about the two facts, and then show the proof.

### Property 1

**Lemma:** In an optimal code tree, every internal node of  $T$  must have two children.

The proof is left as a regular exercise.

## Property 2

**Lemma:** Let  $\sigma_1$  and  $\sigma_2$  be two letters in  $\Sigma$  with the lowest frequencies. There exists an optimal code tree where  $\sigma_1$  and  $\sigma_2$  have the same parent.

**Proof:** If  $\sigma_1$  and  $\sigma_2$  already have the same parent, there is nothing to prove. In the following, we assume they don't.

W.l.o.g., assume  $\text{freq}(\sigma_1) \leq \text{freq}(\sigma_2)$ . Let  $T$  be any optimal code tree in which  $\sigma_1$  and  $\sigma_2$  have different parent. Let  $p$  be an arbitrary **internal node**<sup>2</sup> with the largest level in  $T$ . By Property 1,  $p$  must have two leaves. Let  $x$  and  $y$  be letters corresponding to those leaves such that  $\text{freq}(x) \leq \text{freq}(y)$ . Swap  $\sigma_1$  with  $x$  and  $\sigma_2$  with  $y$ , which gives a new code tree  $T'$ . Note that both  $\sigma_1$  and  $\sigma_2$  are children of  $p$  in  $T'$ .

Convince yourself that the average length of  $T'$  is at most that of  $T$ . Hence,  $T'$  is optimal as well.  $\square$

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<sup>2</sup>Note that leaves are not internal nodes!

# Correctness of Huffman's Algorithm

**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 the theorem's correctness for  $n = k - 1$  where  $k \geq 3$ , next we show that it also holds for  $n = k$ .

**Proof (cont.):** Let  $\sigma_1$  and  $\sigma_2$  be two letters in  $\Sigma$  with the lowest frequencies.

By Property 2, there is an optimal code tree  $T$  on  $\Sigma$  where leaves  $\sigma_1$  and  $\sigma_2$  are the children of the same parent  $p$ .

Let  $T_{huff}$  be the code tree returned by Huffman's algorithm on  $\Sigma$ . Convince yourself that  $\sigma_1$  and  $\sigma_2$  have the same parent  $q$  in  $T_{huff}$ .



**Proof (cont.):** Construct a new alphabet  $\Sigma'$  from  $\Sigma$  by removing  $\sigma_1$  and  $\sigma_2$ , and adding a letter  $\sigma^*$  with frequency  $\text{freq}(\sigma_1) + \text{freq}(\sigma_2)$ .

Let  $T'$  be the tree obtained by removing leaves  $\sigma_1$  and  $\sigma_2$  from  $T$  (thus making  $p$  a leaf).  $T'$  is a code tree on  $\Sigma'$  where  $p$  corresponds to  $\sigma^*$ . Observe:

$$\text{avg height of } T = \text{avg height of } T' + \text{freq}(\sigma_1) + \text{freq}(\sigma_2).$$

Let  $T'_{\text{huff}}$  be the tree obtained by removing leaves  $\sigma_1$  and  $\sigma_2$  from  $T_{\text{huff}}$  (thus making  $q$  a leaf).  $T'_{\text{huff}}$  is a code tree on  $\Sigma'$  where  $q$  corresponds to  $\sigma^*$ .

$$\text{avg height of } T_{\text{huff}} = \text{avg height of } T'_{\text{huff}} + \text{freq}(\sigma_1) + \text{freq}(\sigma_2).$$

**Proof (cont.):**  $T'_{huff}$  is the output of Huffman's algorithm on  $\Sigma'$ .

By our inductive assumption,  $T'_{huff}$  is optimal on  $\Sigma'$ . Thus:

$$\text{avg height of } T'_{huff} \leq \text{avg height of } T'$$

Hence:

$$\text{avg height of } T'_{huff} + \text{freq}(\sigma_1) + \text{freq}(\sigma_2) \leq \text{avg height of } T' + \text{freq}(\sigma_1) + \text{freq}(\sigma_2)$$

which is exactly

$$\text{avg height of } T_{huff} \leq \text{avg height of } T$$

□