

## Course Logistics

- Greedy algorithms: Chapter 16
- Homework 2 is due this Friday

## 1 Making Change

Say a cashier has to give someone 68 cents in change to a customer for a purchase. What is the smallest number of coins the cashier can use to give this change, assuming they can give quarters (25 cents), dimes (10 cents), nickels (5 cents) and pennies (1 cent)?

*Strategy for giving change:*

- Choose coins one at a time
- At each step, choose the coin \_\_\_\_\_

## 1.1 The Change-Making Problem

Assume we have  $n$  coins, and let

$$v_i = \underline{\hspace{10em}}$$

$$m_i = \underline{\hspace{10em}}$$

**The Change-Making Problem** Given an integer value  $C > 0$  and coin values

$$v_1 > v_2 > v_3 > \cdots > v_n = 1,$$

find the number of coins  $m_i$  for  $i = 1, 2, \dots, n$  so that

- 

- 

## 1.2 The greedy algorithm for change-making

---

GREEDYCOINCHANGE( $C, v = [v_1, v_2, \dots, v_n = 1]$ )

$n = \text{length}(v)$

  Let  $m = [0..0]$  be an empty array of  $n$  zeros

**for**  $i = 1$  to  $n$  **do**

**while**  $C \geq v_i$  **do**

$m[i] = \underline{\hspace{10em}}$

$C = \underline{\hspace{10em}}$

**end while**

**end for**

  Return  $m$

---

**Correctness: The greedy algorithm correctly gives change**

*Proof.* At each step, we keep updated the amount left that we have to pay. We only add another coin of type  $i$  if  $v_i$  is less than this amount, so we never go over. We will always reach the exact amount, because  $C$  is an integer value and we can always add coins of values  $v_n = 1$ .  $\square$

**Is the greedy algorithm optimal?**

**Question 1.** Which of the following do you think is true?

- A** The greedy algorithm works for some values of  $C$  when we use coin values (1,5,10,25) (like  $C = 68$ ), but is sometimes not optimal when we use these coin values (1,5,10,25)
- B** The greedy algorithm is always optimal for giving change for coin values (1,5,10,25), but for other coin types  $v = (v_1, v_2, \dots, v_n)$  it may not be optimal
- C** The greedy algorithm is always optimal for the coin change problem, regardless of coin values, as long as  $v_n = 1$ .
- D** Any computational problem can be solved optimally by a greedy algorithm

## 2 Greedy Algorithm Basics

A greedy algorithm is an algorithm that

- 

- 

**Elements of greedy algorithms.** There are two key elements we look for when applying a greedy algorithm:

1. **Optimal substructure:**
2. **Greedy-choice property:** a globally optimal solution can be achieved by

---

### Overall strategy for using greedy algorithms

1. Determine a notion of the \_\_\_\_\_ that improves the global optimization problem.
2. Show how making one local choice leads to \_\_\_\_\_
3. Prove that the greedy choice is \_\_\_\_\_: it leaves us with a subproblem which, when solves optimally and combined with the first greedy choice,

---

### 3 The Activity-Selection Problem

**The activity-selection problem** Let  $(a_1, a_2, a_3, \dots, a_n)$  represent a set of activities, where:

- $s_i =$

- $f_i =$

where  $s_i < f_i$  for  $i = 1, 2, \dots, n$ . We assume a half-open interval  $[s_i, f_i)$  for the activity  $a_i$ . Assume these activities are ordered by finish times to that so that

---

The activity-selection problem

	$i$	1	2	3	4	5
<b>Example:</b>	$s_i$	1	3	2	4	7
	$f_i$	4	5	6	8	9

### 3.1 Optimal Substructure

Consider an activity-selection problem defined by  $(a_1, a_2, \dots, a_n)$  and the time interval  $[s_1, f_n)$ . If we know that an optimal solution contains activity  $a_k$ . Then:

**A dynamic programming approach:** Try each  $a_k$ , solving two smaller subproblems in each case, and taking the best answer.

### 3.2 A greedy choice lemma

**Lemma 3.1.** *There exists an optimal solution to the activity-selection problem defined by  $(a_1, a_2, \dots, a_n)$  that contains  $a_1$  (the activity with the earliest finish time).*

*Proof.*

- Let  $S$  be an optimal solution, and let  $a_i$  be its activity with the earliest start and end time.
- $a_1$  has a finish time  $f_1 \leq f_i$ .
- This  $a_i$  overlaps with  $a_1$ , otherwise we could add  $a_1$  to make  $S$  better
- If we replace  $a_i$  with  $a_1$  to get  $S'$ , then  $|S| = |S'|$ , and this is a valid non-overlapping solution, because  $a_i$  was the earliest activity in  $S$ .

### 3.3 An Optimal Greedy Algorithm for the Activity-Subset Problem

---

GREEDYACTIVITYSELECTION( $s = (s_1, s_2, \dots, s_n)$ ,  $f = (f_1, f_2, \dots, f_n)$ )

$n = \text{length}(s)$

$S = \{a_1\}$

$k = 1$

**for**  $i = 2$  **to**  $n$  **do**

**if**  $s_i \geq f_k$  **then**

$S = S \cup \{a_i\}$

$k = i$

**end if**

**end for**

Return  $S$

---



## 4 Binary Codes

- An \_\_\_\_\_ is a set of characters, e.g.,  $C = \{a, b, c, d, e, f\}$
- A *binary code* for  $C$  is

Consider three possible codes for  $C = \{a, b, c, d, e, f\}$ .

Character:	a	b	c	d	e	f
example code 1:	000	001	010	011	100	101
example code 2:	0	101	100	111	1101	1100
example code 3:	0	1	10	01	11	00

(1)

Types of codes:

- **Fixed length code:**
- **Variable-length code:**
- **Prefix code:** the codeword for every  $c \in C$  is

**Question 2.** Which of the codes in Table 1 is a prefix code?

- A** Example code 1
- B** Example code 2
- C** Example code 3
- D** Example codes 1 and 2
- E** Example codes 1, 2, and 3

## 4.1 Encoding and Decoding

Consider a string of characters from the alphabet  $C = \{\mathbf{a}, \mathbf{b}, \mathbf{c}, \mathbf{d}, \mathbf{e}, \mathbf{f}\}$  where the frequency of each character is given in the following table and we have two codes.

Character:	<b>a</b>	<b>b</b>	<b>c</b>	<b>d</b>	<b>e</b>	<b>f</b>
Frequency	0.45	0.13	0.12	0.16	0.09	0.05
FLC:	000	001	010	011	100	101
VLC:	0	101	100	111	1101	1100

(2)

## 4.2 The Optimal Prefix Code Problem

**The optimal prefix code problem** Given an alphabet  $C$  and a frequency  $c.freq$  for each  $c \in C$  in a given string  $s$ ,

**Representing a prefix code as a tree** Every prefix code for a string  $s$  can be represented as a binary tree  $T$  where

- Left branches are associated with
- Right branches are associated with
- Each leaf corresponds to a character  $c \in C$ , whose binary code is given by
- Each node is associated with the frequency of characters in its subtree

<b>Example</b>	Character:	a	b	c	d	e	f
	Frequency	0.45	0.13	0.12	0.16	0.09	0.05
	VLC:	0	101	100	111	1101	1100

The cost of the tree is give by

$$B(T) = \sum_{c \in C}$$

where  $d_T(c)$  is the depth of  $c$  in the tree  $T$ .

**Question 3.** True or false: every binary tree (with  $\ell$  leaves) gives a valid prefix code (for an alphabet with  $\ell$  characters).

**A** True

**B** False

**Theorem 4.1.**

*Proof.*

**Optimal prefix code problem:**

**Nice properties about the optimal tree** In the optimal tree representation,

- 

-

### 4.3 Huffman Codes

Huffman Code: a prefix code obtained by greedily building a tree representation for  $C$

#### Huffman Code Tree Process

- Associate each character in  $C$  with a node, labeled by its frequency
- Identify two nodes  $x$  and  $y$  in  $C$  with the \_\_\_\_\_
- Create new node  $z$  with frequency \_\_\_\_\_
- Make  $z.left = x$ , and  $z.right = y$ .
- Repeat the above procedure on the alphabet obtained by \_\_\_\_\_

Observe:  $x$  and  $y$  will be siblings in  $T$  at \_\_\_\_\_

<b>Activity</b>	Create a Huffman code for:	Character:	a	b	c	d	e	f
		Frequency	0.45	0.13	0.12	0.16	0.09	0.05

---

HUFFMANCODE( $C$ )

$n = |C|$

$T \leftarrow$  empty graph

$Q \leftarrow \emptyset$

**for**  $c$  in  $C$  **do**

    Insert  $c$  with value  $c.freq$  into  $T$  and  $Q$

**end for**

**for**  $i = 1$  to  $n - 1$  **do**

$x =$

$y =$

    create node  $z$

**end for**

Return  $T$

---

#### 4.4 Code and Illustration

## 4.5 Runtime Analysis

Given a set of objects  $C$ , with values  $c.val$  for  $c \in C$  and  $n = |C|$ , a binary min-heap for  $C$  is a binary tree such that

- All levels
- The value of a node is

It has the following operations:

- $BUILDBINMINHEAP(C)$ :
- $EXTRACTMIN(Q)$ :
- $INSERT(Q, z, z.freq)$ :

**Question 4.** Assume we use a binary min heap to store and update  $Q$  in the pseudocode above. Then what is the overall runtime of creating a Huffman code, in terms of  $n = |C|$ ?

- A**  $O(\log n)$
- B**  $O(n)$
- C**  $O(n \log n)$
- D**  $O(n^2)$



## 4.6 Optimality

It turns out that a Huffman code optimally solves the prefix code problem. The argument hinges on the following lemma.

**Lemma 4.2.** *Let  $C$  be an alphabet where  $c.freq$  is the frequency of  $c \in C$ . Let  $x$  and  $y$  be the two characters in  $C$  that have the lowest frequencies.*

*Then, there exists an optimal prefix code for  $C$  in which  $x$  and  $y$  have the same length code words and only differ in the last bit.*

*Equivalently:*

**Why does this imply optimality?** Consider a slightly different (but equivalent) approach to defining a Huffman code:

1. Associate each character with a node, labeled by its frequency
2. Identify the two nodes  $x$  and  $y$  with the smallest frequencies
3. Create new node  $z$  with frequency  $z.freq = x.freq + y.freq$ .
- 4.
5. Create tree  $T$  from  $T'$  by making  $z.left = x$ , and  $z.right = y$ .

**Claim** This produces an optimal tree  $T$ .

*Proof.* Note that the cost of the tree  $T$  is

$$B(T) = B(T') + x.freq + y.freq \tag{3}$$

We prove the result by contradiction: suppose that  $T$  is not optimal. Then

We can assume that  $x$  and  $y$  are at maximum depth in  $R$ .

Let  $R'$  be the tree obtained by taking  $R$  and

Similar to (3) we have that:

We then get an impossible sequence of inequalities:

$$\begin{aligned} B(R') &= B(R) - x.freq - y.freq \\ &< B(T) - x.freq - y.freq \\ &= B(T') \end{aligned}$$