CSC373

Week 3

Huffman Code

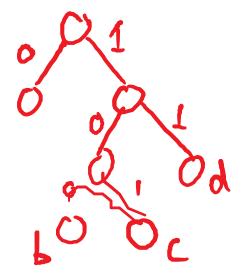
- Variable
 - Length encoding
 - Better approach because every letter has a different frequency of occurrence

Goal: Minimize $ABL(\gamma) = \sum_{x \in S} f(x) \cdot |\gamma(x)|$ S — alphabet f(x) = frequency of occurrence of x $\gamma(x) = \text{encoding of } x$

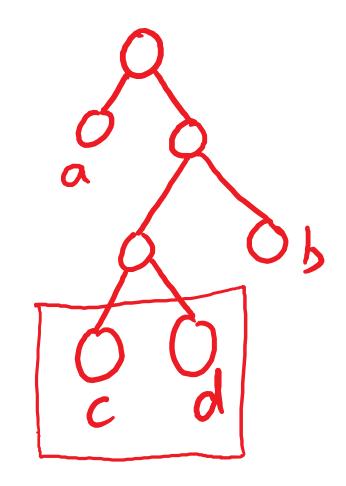
Biggest problem with variable length encoding is ambiguity

Solution: Prefix - code

How to generate prefix codes?



$$S = \{a, b, c, d\}$$
 $f(a) = 0.4$
 $f(b) = 0.3$
 $f(c) = 0.2$
 $f(d) = 0.1$



$$ABL(\chi) = f(0) \cdot 1 + f(1) + f(1) \cdot 3 + f(1) \cdot 3$$

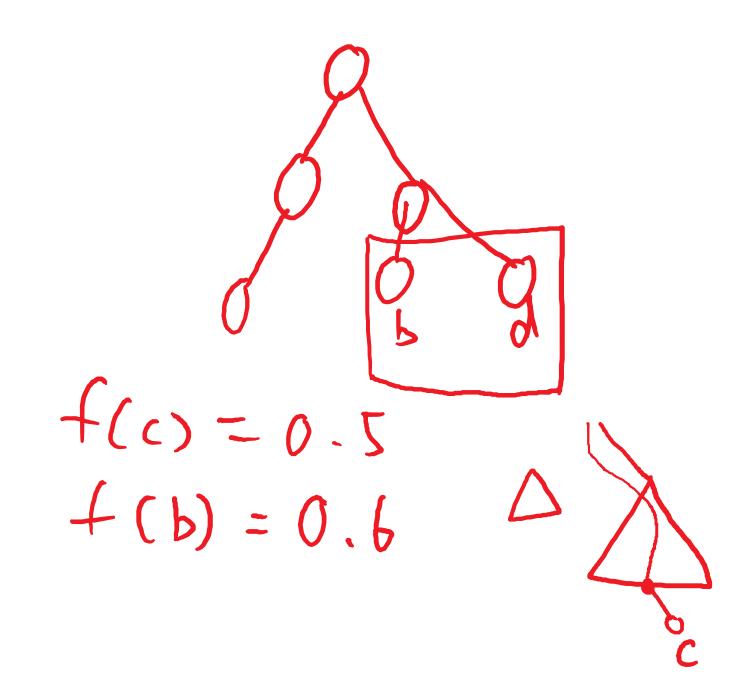
$$ABL(\chi') = f(0) \cdot 3 + f(1) + f(1) \cdot 1 + f(1) \cdot 3$$

$$ABL(\chi') - ABL(\chi) = f(\alpha)(3-1) + f(1) \cdot (1-3)$$

$$= 0.4 \times 2 + 0.2 \times (-2)$$

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Claim: Let x and y be the two least frequent letters. Then there is same optimal tree where x and y are siblings.

- Claim: Let x and y be two least frequent letters. Then there is an optimal tree Θ where x and y are siblings
- Proof: Let T be an optimal tree for (s, f)

Let γ = prefix—code corresponding to T

Let x be the least, and y be the second least

If x and y are siblings, then done.

so x and y are not siblings

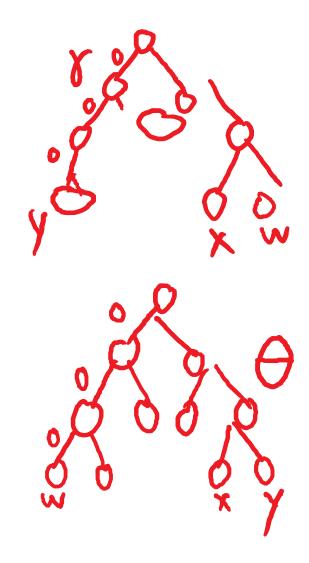
Assume, $|\gamma(x)| \ge |\gamma(y)|$

Let ω be the sibling of x

Interchange ω and y and get a new tree Θ

$$ABL(\Theta) - ABL(T) = f(w) \cdot |\gamma(y)| + f(y) \cdot |\gamma(w)| - f(w) \cdot |\gamma(w)| - f(y) \cdot |\gamma(y)|$$
$$= (f(w) - f(y))(|\gamma(y)| - |\gamma(w)|)$$





Proof (continued)

So,
$$|\gamma(y)| - |\gamma(w)| = |\gamma(y)| - |\gamma(x)| \le 0$$

So,
$$ABL(\Theta) - ABL(T) \le 0$$

Since
$$T$$
 is optimal, $ABL(\Theta) - ABL(T) < 0$
 $\Rightarrow ABL(\Theta) - ABL(T) = 0$
 $\Rightarrow ABL(\Theta) = ABL(T)$
 $\Rightarrow \Theta$ is optimal

Huffman Coding

If *S* has two letters then:

Encode one with 0 and the other with 1

Else:

Let y^* and z^* be the two least frequent letters

Form a new alphabet

$$S' = S \cup \{w\}/\{y^*, z^*\}$$

with $f(w) = f(y^*) + f(z^*)$

"Recursively construct a prefix code γ' for S' with tree T'

Define a prefix code for *S* as follows:

Start with *T*

Take the leaf labeled w in T' and add 2 children below if labeled y^* and z^*

End If

$$f(a) = 0.4$$
 $f(b) = 0.3$
 $f(c) = 0.2$
 $f(d) = 0.1$
 $f(w) = +(c) + +(d) = 0$

$$S = \{a,b,(,d,e\} \\ f(a) = f(b) = f(c) = f(d) = f(e) = 0.2$$

$$S' = \{a,b,c,w\} \\ f(w) = 0.4$$

$$S'' = \{3,c,w\} \\ f(3 = 0.4)$$

$$S'' = (y, w)$$

$$f(y) = 0.6$$

$$C = 000$$

$$C = 001$$

$$C = 001$$

$$C = 01$$

Proof why Huffman Coding works

• Proof by induction on *n*

Base Case:

$$n = 2$$
: $S = \{a, b\}$
 $\gamma(a) = 0$ $\gamma(b) = 1$

Inductive Hypothesis: Assume Huffman coding gives an optimal tree for any alphabeted of size n

Inductive Step: Show it works for alphabet of size n + 1

$$|S| = n + 1$$

Let T be Huffman tree and let Θ be an optimal tree for S

Let y^* and z^* be the two least frequent letters

Replace
$$y^*$$
 and z^* by w

$$S' = S \cup \{w\}/\{y^*, z^*\}$$

$$f(w) = f(y^*) + f(z^*)$$

$$|S'| = n$$

by inductive hypothesis, Huffman coding gives an optimal tree T' for S'

$$ABL(T) = \sum_{x \in S} f(x) \cdot |\gamma(x)|$$

$$= \sum_{x \in S \setminus \{y^*, z^*\}} f(x) |\gamma(x)| + f(y^*) |\gamma(y^*)| + f(z^*) |\gamma(z^*)|$$

$$= \sum_{x \in S \setminus \{y^*, z^*\}} f(x) \cdot |\gamma(x)| + f(z^*) \cdot |\gamma(w)| + 1$$

$$= \sum_{x \in S \setminus \{y^*, z^*\}} f(x) \cdot |\gamma(x)| + f(w) \cdot |\gamma(w)| + 1$$

$$= \sum_{x \in S \setminus \{y^*, z^*\}} f(x) \cdot |\gamma(x)| + f(w)$$

$$= ABL(T') + f(w)$$

$$= ABL(T') + f(y^*) + f(z^*)$$

Proof (continued)

Assume for a contradiction that T is not optimal Let O be an optimal tree for S such that y^* and z^* are sibling in O (by claim proved before)

If you remove y^* and z^* from 0 and replace them by w with $f(w) = f(y^*) + f(z^*)$

Then this new tree O' must be optimal for S'

So
$$ABL(O') = ABL(T')$$

$$\Rightarrow ABL(O') + f(y^*) + f(z^*) = ABL(T') + f(y^*) + f(z^*) = ABL(T)$$

$$(I)$$

$$ABL(O)$$

Weighted Interval Scheduling

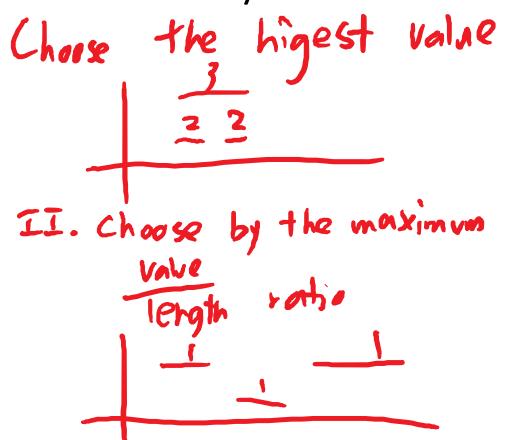
- Single Resource
- Input

$$R = \left\{ \begin{array}{c} 1 \\ 2 \\ (52:13) \\ (2) \\ (3) \\ (4) \\ (4) \\ (5) \\ (4) \\ (5) \\ (4)$$

Goal: Maximize the total value of jobs scheduled



Use Greedy



Use Greedy (continued)

III. Dynamic Programming

Decompose a problem into several subproblems and combine solutions

Goal: Schedule *S*

$$S \subseteq R$$

Question: Does $r_n \in S$?

Answer: It may or may not

Case I: $r_n \in S$

Case II: $r_n \notin S$

- I. Sort all the jobs by their finish time
- II. Define function *P*

P(i) = the largest j < i such that j does not overlap

with *i*

In other words, all jobs between P(i) and i overlap with i

Case I: $r_n \in S$

Any job between $r_{P(n)}$ and r_n is not in S

Next job we can possibly schedule is $r_{P(n)}$

Case II: $r_n \notin S$

Next job we can possibly schedule is r_{n-1}

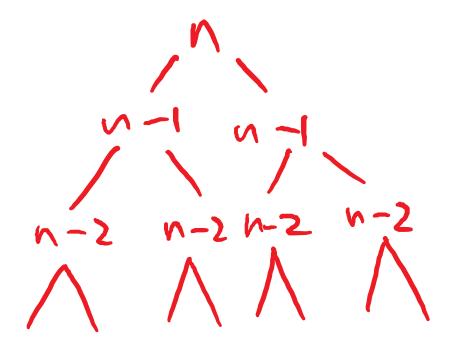
• Subproblems look like solve the same problem on $\{r_1, \dots, r_j\}$ $1 \le j \le n$

```
O_i: optimal solution to subproblems \{r_1, \dots, r_n\}
OPT_i: value of that optimal solution
       =\sum_{i\in S_{\theta_i}}v_i
If r_j \in S_{O_i}: OPT_j = OPT_{P(i)} + v_j
If r_j \notin S_{O_i}: OPT_j = OPT_{j-1}
OPT_i = \max\{OPT_{i-1}, OPT_{P(i)} + v_i\}
If OPT_{i-1} > OPT_{P(i)} + v_i
           then r_j \notin S_{\theta_i}
Else
           r_j \in S_{O_i}
```

```
Compute_WIS(i, P)
if j == 0:
      return 0
Else
      s1 = Compute_WIS((P(i), P) + v_j)
      s2 = Compute_WIS(j-1, P)
      return max(S_1, S_2)
```

P(i) = i - 1

Complexity: $O(2^n)$



Solution: Save the results of any computation of use that as a cache

```
compute_WIS(j, P, M)
 if M[j] is defined
         return M[j]
 else:
         if j == 0 then
                   q = 0
         else
                   q = max(S_1, S_2)
                   (S_1, S_2 as before)
         M[j] = q
         return q
```

Complexity: O(n) \checkmark each pass writes one entry of M and there are only n entries to write

 $OPT_n = compute_WIS(n, p, M)$

```
\begin{aligned} & \text{Compute\_WIS\_Schedule(j, P, M)} \\ & \text{if M[j]} == \text{M[P[j]]} + \text{v\_j} \\ & \text{print v\_j} \\ & \text{q} = \text{P[i]} \\ & \text{else} \\ & \text{q} = \text{j} - 1 \\ & \text{return Compute\_WIS\_Schedule(q, P, M)} \end{aligned}
```

Key concepts

- 1) Optimal substructure
- 2) Recursion relation between the subproblem and the main problem
- Implementing recursion directly will most likely have an exponential growth

Memorization

• Save the results. Complexity drops downs from exponential to (pseudo) polynomial

Divide-and-conquer

 $O(n^2) -> O(n ln n)$ Greedy: $O(n^2) -> O(log_2 7)$

```
Fib(n):

if n == 1 or n == 2

return 1

return Fib(n-1) + Fib(n-2)
```

```
Fib(n, M):
    if n == 1 or n == 2
        return 1
    if M[n] is defined
        return M[n]
    M[n] = Fib(n-1, M) + Fib(n-2, M)
    return M[n]
```

O(n)