

Memoizing

Dynamic Programming in a recursive (top-down) framework

- Whenever you compute the value of a subproblem

f becomes mf plus MEMO table

```
mf(x)
  if MEMO[x] defined then
  else
  return
```

Replace all calls to f by calls to mf

- includes
- but not

Greedy Algorithm

A series of locally optimum choices produces a globally optimum solution

What is needed:

- Greedy choice property — of all the optimal solutions, at least one
- Optimal substructure — as with

Greedy Choice

- Suppose $Sched$ is a maximal legal schedule that
- Let A_i be the activity in $Sched$ with
- Then $A_i.finish$
- So $sched' =$ is also a maximum legal schedule.

Optimal Substructure

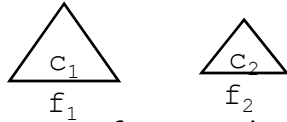
- Assume $Sched$ is maximal legal schedule for
- Let A_i be the first activity with
- Then $Sched - \{A_i\}$ is a maximal legal schedule for

CS 410/584, Algorithm Design & Analysis: Lecture 2

Algorithm Design & Analysis

Greedy Algorithm for Code Generation

Start with n separate 1-character codes
Merge two codes with lowest frequencies



Frequency of merged code is

Until

Lecture 2

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Algorithm Design & Analysis

Huffman Code Example

7	3	4	3	1	2	2	6
◇	A	E	H	N	R	S	T

7	3	4	3			2	6
◇	A	E	H			S	T

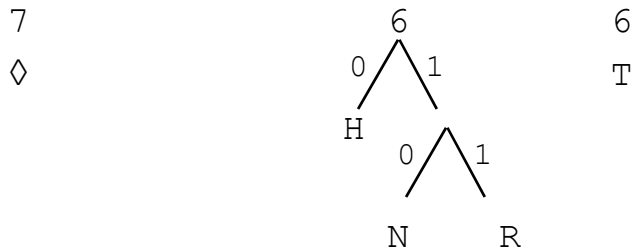
7		4	3	3		6
◇		E	H	0	1	T
				N	R	

Lecture 2

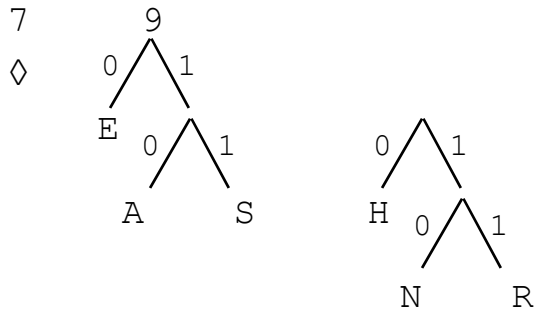
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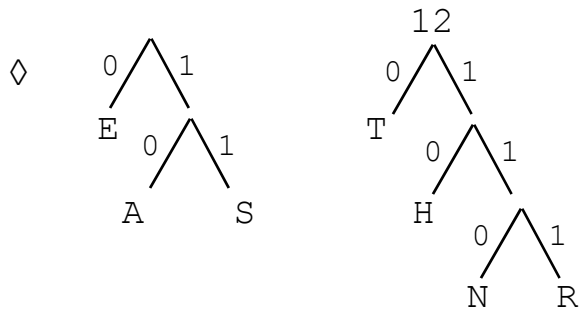
Huffman Code Example 2



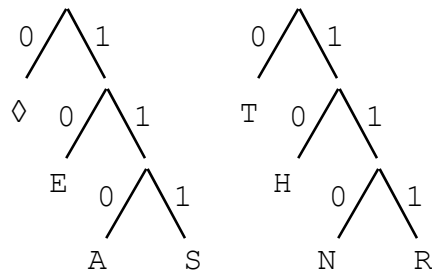
Huffman Code Example 3



Huffman Code Example 4



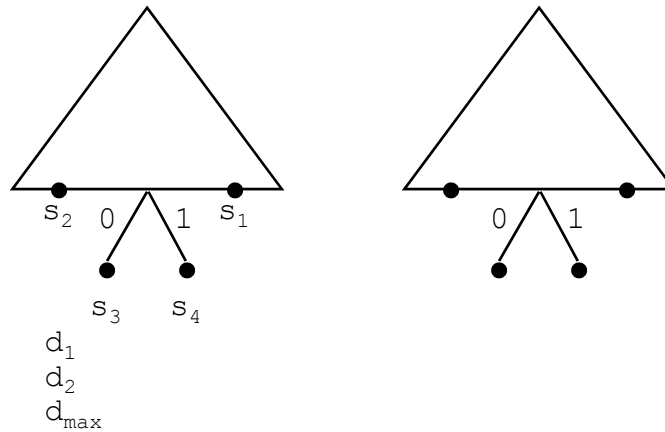
Huffman Code Example 5



Why does it work?

Greedy Choice

• Greedy choice



Greedy Choice 2

$$f_3 \cdot d_{\max} + f_4 \cdot d_{\max} + f_1 \cdot d_1 + f_2 \cdot d_2$$

$$\cdot d_{\max} + \cdot d_{\max} + \cdot d_1 + \cdot d_2$$

s_1-s_3 switch:

old $(f_3 \cdot d_{\max} + f_1 \cdot d_1)$

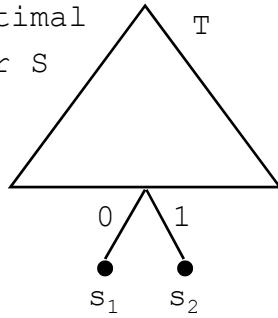
new $(f_1 \cdot d_{\max} + f_3 \cdot d_1)$

subtract

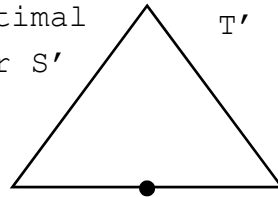
Optimal Substructure

Let S be the set of symbols

optimal
for S



optimal
for S'



$$\text{cost}(T) = \text{cost}(T')$$