

The Greedy Method and Text Compression



The Greedy Method Technique (§ 11.4.2)



- ◆ **The greedy method** is a general algorithm design paradigm, built on the following elements:
 - **configurations**: different choices, collections, or values to find
 - **objective function**: a score assigned to configurations, which we want to either maximize or minimize
- ◆ It works best when applied to problems with the **greedy-choice** property:
 - a globally optimal solution can always be found by a series of local improvements from a starting configuration.

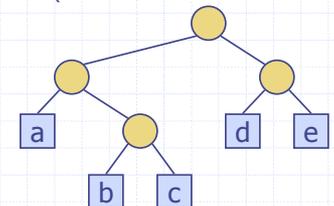
Text Compression (§ 11.4)

- ◆ Given a string X , efficiently encode X into a smaller string Y
 - Saves memory and/or bandwidth
- ◆ A good approach: **Huffman encoding**
 - Compute frequency $f(c)$ for each character c .
 - Encode high-frequency characters with short code words
 - No code word is a prefix for another code
 - Use an optimal encoding tree to determine the code words

Encoding Tree Example

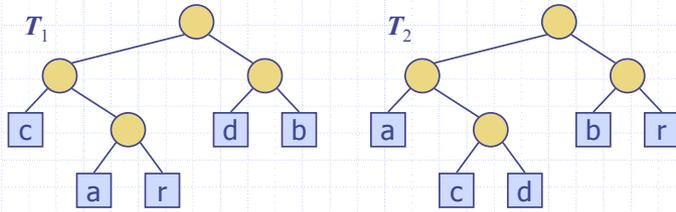
- ◆ A **code** is a mapping of each character of an alphabet to a binary code-word
- ◆ A **prefix code** is a binary code such that no code-word is the prefix of another code-word
- ◆ An **encoding tree** represents a prefix code
 - Each external node stores a character
 - The code word of a character is given by the path from the root to the external node storing the character (0 for a left child and 1 for a right child)

00	010	011	10	11
a	b	c	d	e



Encoding Tree Optimization

- Given a text string X , we want to find a prefix code for the characters of X that yields a small encoding for X
 - Frequent characters should have long code-words
 - Rare characters should have short code-words
- Example
 - $X = \text{abracadabra}$
 - T_1 encodes X into 29 bits
 - T_2 encodes X into 24 bits



Huffman's Algorithm

- Given a string X , Huffman's algorithm constructs a prefix code that minimizes the size of the encoding of X
- It runs in time $O(n + d \log d)$, where n is the size of X and d is the number of distinct characters of X
- A heap-based priority queue is used as an auxiliary structure

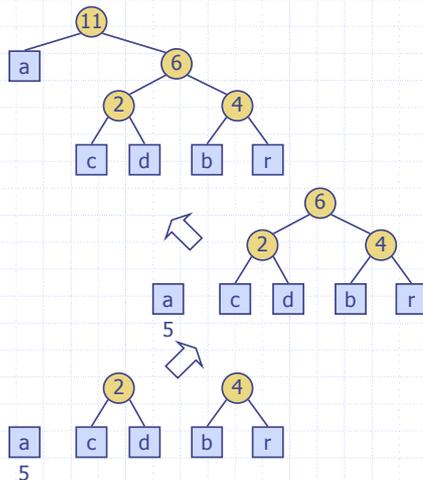
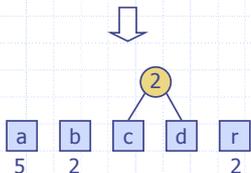
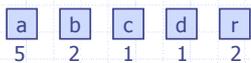
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Algorithm HuffmanEncoding( $X$ )
  Input string  $X$  of size  $n$ 
  Output optimal encoding trie for  $X$ 
   $C \leftarrow \text{distinctCharacters}(X)$ 
   $computeFrequencies(C, X)$ 
   $Q \leftarrow$  new empty heap
  for all  $c \in C$ 
     $T \leftarrow$  new single-node tree storing  $c$ 
     $Q.insert(getFrequency(c), T)$ 
  while  $Q.size() > 1$ 
     $f_1 \leftarrow Q.minKey()$ 
     $T_1 \leftarrow Q.removeMin()$ 
     $f_2 \leftarrow Q.minKey()$ 
     $T_2 \leftarrow Q.removeMin()$ 
     $T \leftarrow join(T_1, T_2)$ 
     $Q.insert(f_1 + f_2, T)$ 
  return  $Q.removeMin()$ 
    
```

Example

$X = \text{abracadabra}$
Frequencies

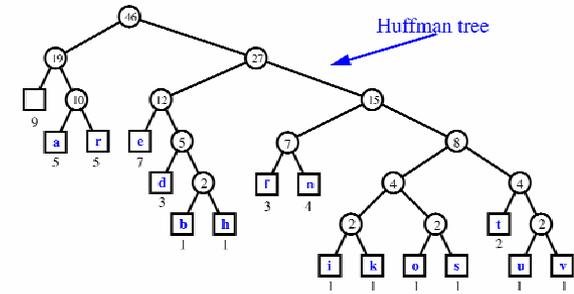
a	b	c	d	r
5	2	1	1	2



Extended Huffman Tree Example

String: **a fast runner need never be afraid of the dark**

Character	a	b	d	e	r	h	i	k	n	o	s	t	u	v
Frequency	9	5	1	3	7	3	1	1	4	1	5	1	2	1



The Fractional Knapsack Problem (not in book)



- Given: A set S of n items, with each item i having
 - b_i - a positive benefit
 - w_i - a positive weight
- Goal: Choose items with maximum total benefit but with weight at most W .
- If we are allowed to take fractional amounts, then this is the **fractional knapsack problem**.
 - In this case, we let x_i denote the amount we take of item i

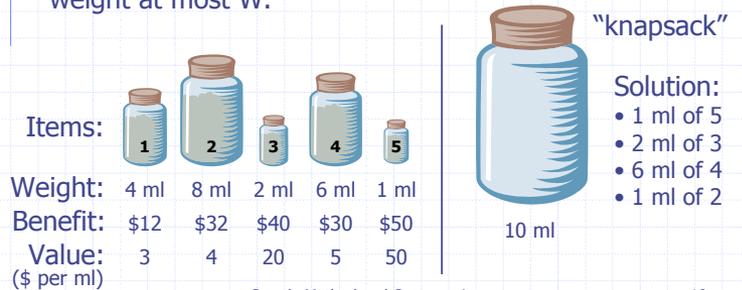
- Objective: maximize
$$\sum_{i \in S} b_i(x_i / w_i)$$

- Constraint:
$$\sum_{i \in S} x_i \leq W$$

Example



- Given: A set S of n items, with each item i having
 - b_i - a positive benefit
 - w_i - a positive weight
- Goal: Choose items with maximum total benefit but with weight at most W .



The Fractional Knapsack Algorithm



- Greedy choice: Keep taking item with highest **value** (benefit to weight ratio)
 - Since $\sum_{i \in S} b_i(x_i / w_i) = \sum_{i \in S} (b_i / w_i)x_i$
 - Run time: $O(n \log n)$. Why?
- Correctness: Suppose there is a better solution
 - there is an item i with higher value than a chosen item j , but $x_i < w_i$, $x_j > 0$ and $v_i < v_j$
 - If we substitute some i with j , we get a better solution
 - How much of i : $\min\{w_i - x_i, x_j\}$
 - Thus, there is no better solution than the greedy one

Algorithm *fractionalKnapsack(S, W)*

Input: set S of items w/ benefit b_i and weight w_i ; max. weight W
Output: amount x_i of each item i to maximize benefit w/ weight at most W

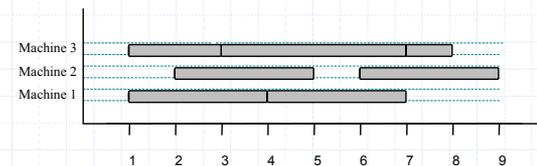
for each item i in S
 $x_i \leftarrow 0$
 $v_i \leftarrow b_i / w_i$ {value}
 $w \leftarrow 0$ {total weight}

while $w < W$
 remove item i w/ highest v_i
 $x_i \leftarrow \min\{w_i, W - w\}$
 $w \leftarrow w + \min\{w_i, W - w\}$

Task Scheduling (not in book)



- Given: a set T of n tasks, each having:
 - A start time, s_i
 - A finish time, f_i (where $s_i < f_i$)
- Goal: Perform all the tasks using a minimum number of "machines."



Task Scheduling Algorithm



- ◆ Greedy choice: consider tasks by their start time and use as few machines as possible with this order.
 - Run time: $O(n \log n)$. Why?
- ◆ Correctness: Suppose there is a better schedule.
 - We can use $k-1$ machines
 - The algorithm uses k
 - Let i be first task scheduled on machine k
 - Machine i must conflict with $k-1$ other tasks
 - But that means there is no non-conflicting schedule using $k-1$ machines

Algorithm *taskSchedule(T)*

Input: set T of tasks w/ start time s_i and finish time f_i

Output: non-conflicting schedule with minimum number of machines

$m \leftarrow 0$ {no. of machines}

while T is not empty

remove task i w/ smallest s_i

if there's a machine j for i then

schedule i on machine j

else

$m \leftarrow m + 1$

schedule i on machine m

Example



- ◆ Given: a set T of n tasks, each having:
 - A start time, s_i
 - A finish time, f_i (where $s_i < f_i$)
 - $[1,4], [1,3], [2,5], [3,7], [4,7], [6,9], [7,8]$ (ordered by start)
- ◆ Goal: Perform all tasks on min. number of machines

