

## Brute-force Algorithms & Greedy Algorithms

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### Brute-force Algorithms

Def'n: Solves a problem in the most simple, direct, or obvious way

- Not distinguished by structure or form
- Pros
  - Often simple to implement
- Cons
  - May do more work than necessary
  - May be efficient (but typically is not)

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### Greedy Algorithms

Def'n: Algorithm that makes sequence of decisions, and never reconsiders decisions that have been made

- Pros
  - May run significantly faster than brute-force
- Cons
  - May not lead to correct/optimal solution

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## Example: Counting Change

### Problem Def'n:

- Cashier has collection of 'coins' of various denominations
- Goal is to return a specified sum using the smallest number of coins

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## Example: Counting Change

### Mathematical Def'n:

#### ■ $n$ coins:

$P = \{p_1, p_2, p_3, \dots, p_n\}$  with value  $D = \{d_1, d_2, d_3, \dots, d_n\}$

- can have repetition (two dimes, three pennies)
- $S$  is a subset of  $P$

$S \subseteq P$ , such that  $s_i = 1$  if  $p_i \in S$ ,  $s_i = 0$  if  $p_i \notin S$

#### ■ $A$ : sum to be returned

- Goal: minimize  $\sum s_i$ , such that  $\sum d_i = A$

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## Brute-force Approach

#### ■ Try all subsets of $P$

- since there are  $n$  coins, there are  $2^n$  possible subsets
- enumerate all possible subsets
- check if a subset equals  $A$ 
  - called 'feasible solution' set
  - $O(n)$
- pick subset that minimizes  $\sum s_i$ 
  - called 'objective function'
  - $O(n)$

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## Brute-force Approach

- Best Case
  - $\Omega(n2^n)$
- Worst Case
  - $O(n2^n)$

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## Greedy Approach

- Go from largest to smallest denomination
  - Return largest coin  $p_i$  from  $P$ , such that  $d_i \leq A$
  - $A = A - d_i$
  - Find next largest coin ...

if money is sorted (by value), then  
algorithm is  $O(n)$

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## Does Greedy Always Work?

Consider  $A = 20$   
and  $D = \{1, 1, 1, 1, 1, 10, 10, 15\}$

Greedy returns 6 coins  
Optimal is 2 coins

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## Text Processing

- Brute-force Pattern Matching
- Improved Pattern Matching
  - Boyer-Moore Algorithm
  - *not really brute-force*
  - *not really greedy either*

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## Pattern Matching

- T: text string of length  $n$
- P: pattern string of length  $m$
- Question: Is P a substring of T?
- Answer: starting index of match or indication that  $P \notin T$

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## Pattern Matching: Pseudocode

```
Algorithm BruteForceMatch (T,P)
  Input: character string T of length n
  and character string P of length m
  Output: integer -1 if  $P \notin T$ , integer i
  (start location of P in T) if  $P \in T$ 
  for i <- 0 to n-m do
    j <- 0
    while (j < m and T[i+j] = P[j]) do
      j <- j + 1
    if j = m then
      return i
  return -1
```

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## Pattern Matching: Complexity

```
for i <- 0 to n-m do           O( )
  j <- 0                       O( )
  while (j < m and T[i+j] = P[j]) do O( )
    j <- j + 1                 O( )
  if j = m then                O( )
    return i                  O( )
return -1                      O( )
```

Worst case complexity:  $O( )$

Best case complexity:  $\Omega( )$

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## Better Pattern Matching: Boyer Moore Algorithm

Two Improvements:

- Looking Glass Heuristic
  - When testing P against T, begin at P[m-1]
- Character Jump Heuristic
  - Mismatch  $T[i] = c$  with  $P[j]$ 
    - if  $c \notin P$ , then shift P past  $T[i]$
    - else if last(c) to left of  $P[j]$  then
      - shift P to align last(c) with  $T[i]$
    - else shift P to right by one

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## Summary: Brute & Greedy

- Brute-force:
  - solve problem in simplest way
  - generate entire solution set, pick best
  - will give optimal solution with (typically) poor efficiency
- Greedy:
  - make local, best decision, and don't look back
  - may give optimal solution with (typically) 'better' efficiency
  - depends upon 'greedy-choice property'
    - global optimum found by series of local optimum choices

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