# CS4102 Algorithms Spring 2019

### Warm up

Given access to unlimited quantities of pennies, nickels dimes, and quarters, (worth value 1, 5, 10, 25 respectively), provide an algorithm which gives change for a given value x using the fewest number of coins.











# Change Making

43 cents













# Change Making Algorithm

- Given: target value x, list of coins  $C=[c_1,\dots,c_n]$  (in this case  $C=[1,\!5,\!10,\!25]$ )
- Repeatedly select the largest coin less than the remaining target value:

$$\begin{aligned} & \text{while}(x>0) \\ & \text{let } c = \max(c_i \in \{c_1, ..., c_n\} \mid c_i \leq x) \\ & \text{print } c \\ & x = x - c \end{aligned}$$

# Why does this always work?

- If x < 5, then pennies only
  - 5 pennies can be exchanged for a nickel Only case Greedy uses pennies!
- If  $5 \le x < 10$  we must have a nickel -2 nickels can be exchanged for a dime Only case Greedy uses nickels!
- If  $10 \le x < 25$  we must have at least 1 dime -3 dimes can be exchanged for a quarter and a nickel Only case Greedy uses dimes!
- nickel Only case Greedy uses dimes! If  $x \ge 25$  we must have at least 1 quarter

Only case Greedy uses quarters

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- Dynamic Programming
- Gerrymandering
- Greedy Algorithms
- Choice Function
- Change Making

**CLRS Readings** 

- Chapter 15
- Chapter 16

## Homeworks

- Homework 5 due Wednesday March 27 at 11pm
  - Seam Carving!
  - Dynamic Programming (implementation)
  - Java or Python
- Homework 6 out tonight, due Wednesday April 3 at 11pm
  - Dynamic Programming and Greedy Algorithms
  - Written (using Latex!)

# **Dynamic Programming**

- Requires Optimal Substructure
- $-\,\mbox{Solution}$  to larger problem contains the solutions to smaller ones
- Idea:
  - 1. Identify recursive structure of the problem
    - What is the "last thing" done?
  - 2. Select a good order for solving subproblems

    - "Top Down": Solve each recursively
       "Bottom Up": Iteratively solve smallest to largest
  - 3. Save solution to each subproblem in memory

# Generic Top-Down Dynamic Programming Soln

mem = {}
def myDPalgo(problem):
 if mem[problem] not blank:
 return mem[problem return mem[problem] if baseCase(problem): solution = solve(problem)
mem[problem] = solution
return solution for subproblem of problem: subsolutions.append(myDPalgo(subproblem)) solution = OptimalSubstructure(subsolutions) mem[problem] = solution return solution

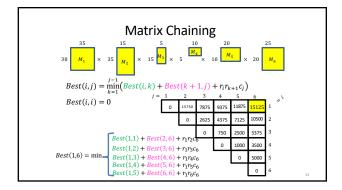
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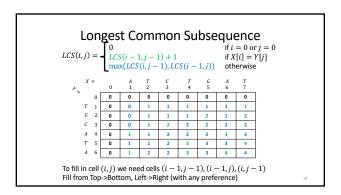
# DP Algorithms so far

- $2 \times n$  domino tiling (Fibonacci)
- Log cutting
- Matrix Chaining
- Longest Common Subsequence
- Seam Carving

# Domino Tiling Tile(n): Initialize Memory M M[0] = 0 M[1] = 0 for i = 0 to n: M[i] = M[i-1] + M[i-2] return M[n]

# 







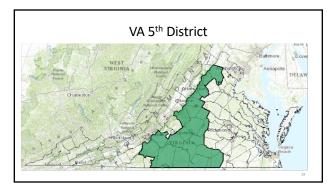
# Gerrymandering

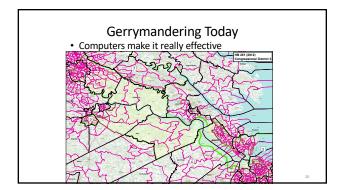
- Manipulating electoral district boundaries to favor one political party over others
- Coined in an 1812 Political cartoon
- Governor Gerry signed a bill that redistricted Massachusetts to benefit his Democratic-Republican Party

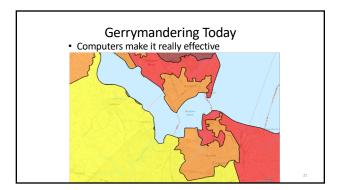


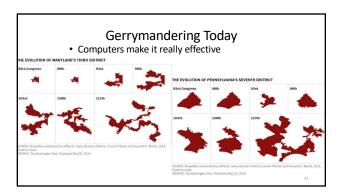
# According to the Supreme Court

- Gerrymandering cannot be used to:
  - Disadvantage racial/ethnic/religious groups
- It can be used to:
  - Disadvantage political parties









# How does it work?

- States are broken into precincts
- All precincts have the same size
- We know voting preferences of each precinct
- Group precincts into districts to maximize the number of districts won by my party







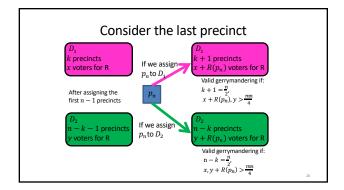
# **Gerrymandering Problem Statement**

- Given:
- A list of precincts:  $p_1, p_2, \dots, p_n$  precincts, n of them Each containing m voters

  Output: n n the votes, but how many Output:  $\begin{array}{c} \text{Output:} \\ -\text{Districts } D_1, D_2 \subset \{p_1, p_2, \dots, p_n\} \\ -\text{Where } |D_1| = |D_2| \\ -R(D_1), R(D_2) > \frac{mn}{4} \quad \frac{1}{2} \cdot \frac{1}{N} \\ \cdot R(D_l) \text{ gives number of "Regular Party" voters in } D_l \\ \cdot R(D_l) > \frac{mn}{4} \text{ means } D_l \text{ is majority "Regular Party"} \\ -\text{"failure" if no such solution is possible} \end{array}$

# **Dynamic Programming**

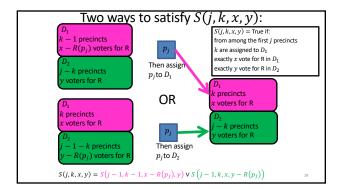
- Requires Optimal Substructure
  - Solution to larger problem contains the solutions to smaller ones
- Idea:
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# **Define Recursive Structure**

S(j,k,x,y)= True if from among the first  ${\it j}$  precincts:  ${\it k}$  are assigned to  ${\it D}_1$  exactly  ${\it x}$  vote for R in  ${\it D}_1$  exactly  ${\it y}$  vote for R in  ${\it D}_2$ 

4D Dynamic Programming!!!



# Final Algorithm $S(j,k,x,y) = S(j-1,k-1,x-R(p_j),y) \vee S\left(j-1,k,x,y-R(p_j)\right)$ Initialize S(0,0,0,0) = True for $j=1,\dots,n$ : for $k=1,\dots,jm$ : for $k=1,\dots,jm$ : for $k=1,\dots,jm$ : exactly k vote for k in k

# Run Time $S(j,k,x,y) = S(j-1,k-1,x-R(p_j),y) \vee S(j-1,k,x,y-R(p_j))$ Initialize S(0,0,0,0) = True n for j = 1,...,n: $\frac{n}{2} \text{ for } k = 1,...,\min(j,\frac{n}{2})$ : nm for y = 0,...,jm: nm for y = 0,...,jm: $S(j,k,x,y) = S(j-1,k-1,x-R(p_j),y)$ $\vee S(j-1,k-1,x-R(p_j),y)$ $\vee S(j-1,k,x,y-R(p_j))$ Search for True entry at $S(n,\frac{n}{r_2},\frac{mn}{4},\frac{mn}{4})$

# $\Theta(n^4m^2)$

- Runtime depends on the *value* of m, not *size* of m
- Run time is exponential in size of input
- Note: Gerrymandering is NP-Complete

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