



**Midterm Review**  
Nov 9<sup>th</sup>, 2017

# Algorithm Design and Analysis

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# Midterm!!!



- **Date: 11/16 (Thursday)**
- **Time: 14:20-17:20** (3 hours)
- **Location: R103** (check the seat assignment before entering the room)
- **Content**
  - Recurrence and Asymptotic Analysis
  - Divide and Conquer
  - Dynamic Programming
  - Greedy
- Based on slides, assignments, and some variations (practice via textbook exercises)
- Format: Yes/No, Multiple-Choice, Short Answer, Prove/Explanation
- Easy: ~60%, Medium: ~30%, Hard: ~10%
- Close book

# Algorithm Design & Analysis Process

- 1) Formulate a **problem**
- 2) Develop an **algorithm**
- 3) Prove the **correctness**
- 4) Analyze **running time/space** requirement

**Design Step**

**Analysis Step**

# Algorithm Analysis

- Analysis Skills
  - Prove by contradiction
  - Induction
  - Asymptotic analysis
  - Problem instance
- Algorithm Complexity
  - In the worst case, what is the growth of function an algorithm takes
- Problem Complexity
  - In the worst case, what is the growth of the function the optimal algorithm of the problem takes

# Algorithm Design Strategy

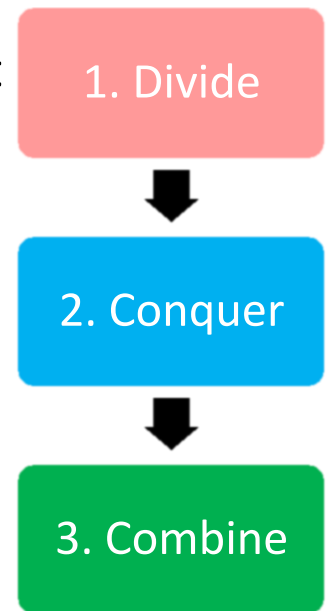
- Do not focus on “specific algorithms”
- But “some strategies” to “design” algorithms
- First Skill: Divide-and-Conquer (各個擊破)
- Second Skill: Dynamic Programming (動態規劃)
- Third Skill: Greedy (貪婪法則)



# Divide-and-Conquer

# What is Divide-and-Conquer?

- Solve a problem recursively
- Apply three steps at each level of the recursion
  1. **Divide** the problem into a number of subproblems that are smaller instances of the same problem (比較小的同樣問題)
  2. **Conquer** the subproblems by solving them recursively  
If the subproblem sizes are *small enough*
    - then solve the subproblems base case
    - else recursively solve itself recursive case
  3. **Combine** the solutions to the subproblems into the solution for the original problem



# How to Solve Recurrence Relations?

1. **Substitution Method** (取代法)
  - Guess a bound and then prove by induction
2. **Recursion-Tree Method** (遞迴樹法)
  - Expand the recurrence into a tree and sum up the cost
3. **Master Method** (套公式大法/大師法)
  - Apply Master Theorem to a specific form of recurrences



# When to Use D&C?

- Analyze the problem about
  - Whether the problem with small inputs can be solved directly
  - Whether subproblem solutions can be combined into the original solution
  - Whether the overall complexity is better than naïve



# Dynamic Programming

# What is Dynamic Programming?

- Dynamic programming, like the divide-and-conquer method, solves problems by combining the solutions to subproblems
  - 用空間換取時間
  - 讓走過的留下痕跡
- “Dynamic”: time-varying
- “Programming”: a *tabular* method

Dynamic Programming: planning over time

# Algorithm Design Paradigms

- Divide-and-Conquer
  - partition the problem into **independent** or **disjoint** subproblems
  - repeatedly solving the common subsubproblems
  - more work than necessary
- Dynamic Programming
  - partition the problem into **dependent** or **overlapping** subproblems
  - avoid recomputation
    - ✓ Top-down with memoization
    - ✓ Bottom-up method

# Dynamic Programming Procedure

- Apply four steps
  1. Characterize the structure of an optimal solution
  2. **Recursively** define the value of an optimal solution
  3. Compute the value of an optimal solution, typically in a **bottom-up** fashion
  4. Construct an optimal solution from computed information

# When to Use DP?

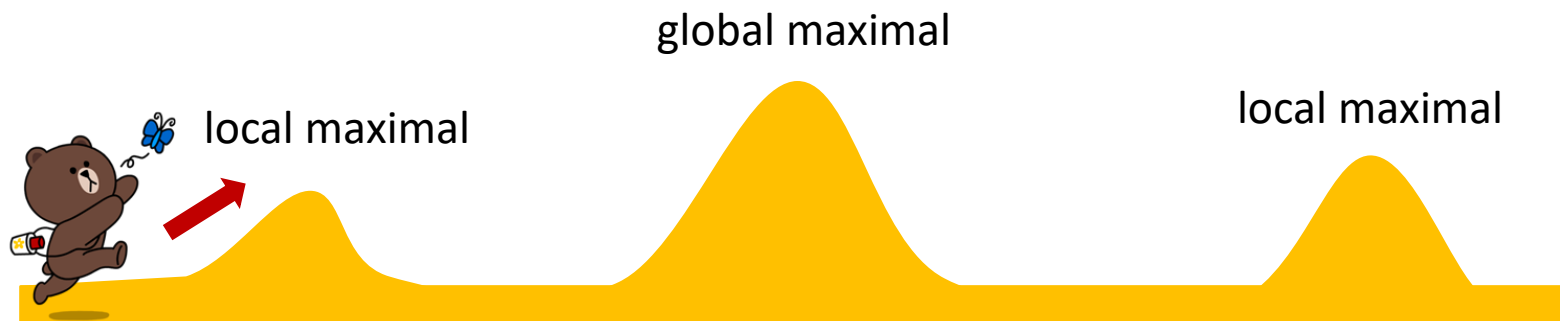
- Analyze the problem about
  - Whether subproblem solutions can combine into the original solution
  - When subproblems are overlapping
  - Whether the problem has optimal substructure
  - Common for optimization problem
- Two ways to avoid recomputation
  - Top-down with memoization
  - Bottom-up method
- Complexity analysis
  - Space for tabular filling
  - Size of the subproblem graph



# Greedy Algorithms

# What is Greedy Algorithms?

- always makes the choice that looks best at the moment
- makes a **locally optimal** choice in the hope that this choice will lead to a **globally optimal** solution
  - not always yield optimal solution; may end up at local optimal

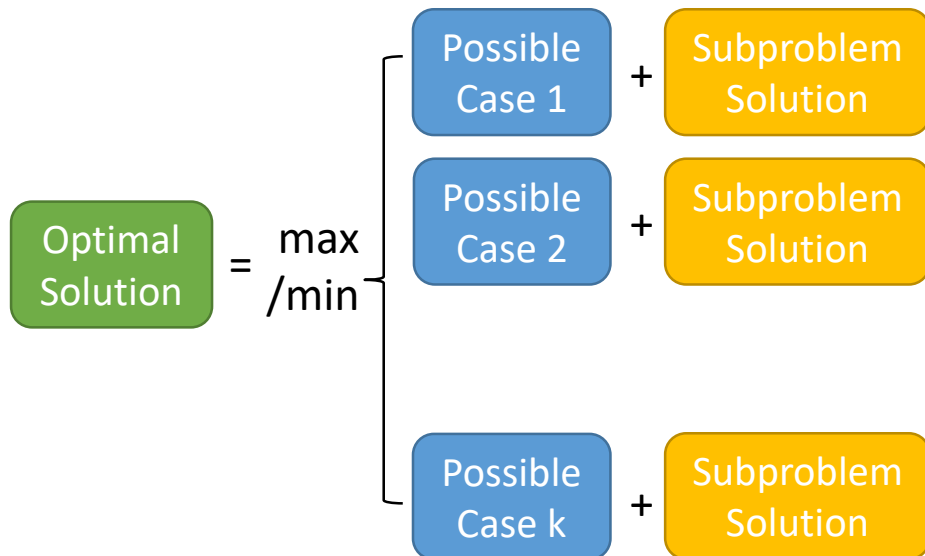




# Algorithm Design Paradigms

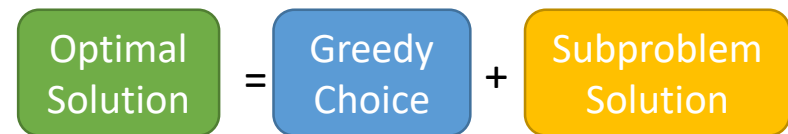
- Dynamic Programming

- has **optimal substructure**
- make an informed choice after getting optimal solutions to subproblems
- dependent** or **overlapping** subproblems



- Greedy Algorithms

- has **optimal substructure**
- make a greedy choice before solving the subproblem
- no overlapping** subproblems
  - ✓ Each round selects only one subproblem
  - ✓ The subproblem size decreases

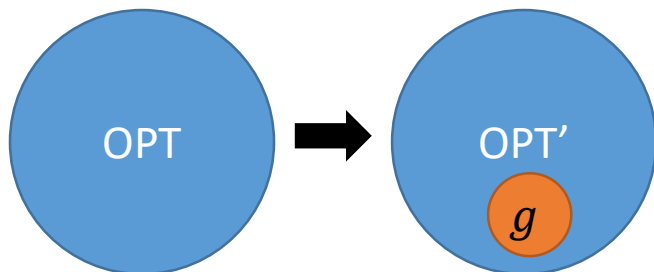


# Greedy Procedure

1. **Cast the optimization problem** as one in which we make a choice and remain one subproblem to solve
2. **Demonstrate the optimal substructure**
  - ✓ Combining an optimal solution to the subproblem via greedy can arrive an optimal solution to the original problem
3. **Prove** that there is always an optimal solution to the original problem that makes the **greedy choice**

# Proof of Correctness Skills

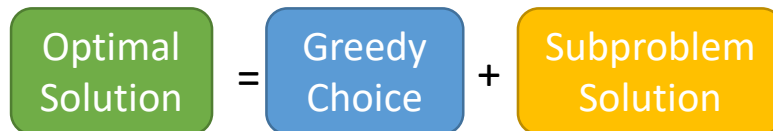
- **Optimal Substructure** : an optimal solution to the problem contains within it optimal solutions to subproblems
- **Greedy-Choice Property** : making locally optimal (greedy) choices leads to a globally optimal solution
  - Show that it exists an optimal solution that “contains” the greedy choice using **exchange argument**
  - For any optimal solution OPT, the greedy choice  $g$  has two cases
    - $g$  is in OPT: done
    - $g$  not in OPT: modify OPT into OPT' s.t. OPT' contains  $g$  and is at least as good as OPT



- ✓ If OPT' is better than OPT, the property is proved by contradiction
- ✓ If OPT' is as good as OPT, then we showed that there exists an optimal solution containing  $g$  by construction

# When to Use Greedy?

- Analyze the problem about
  - Whether the problem has optimal substructure
  - Whether we can make a greedy choice and remain only one subproblem
  - Common for optimization problem





# Exercises

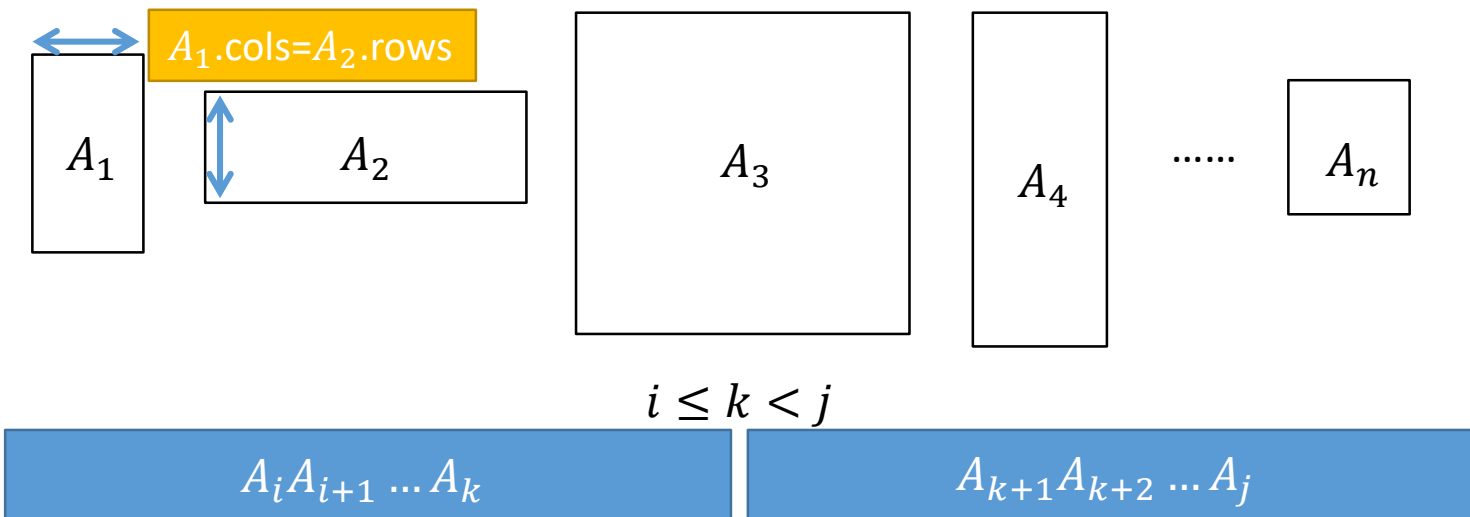
# Short Answer Questions

- True or False: To prove the correctness of a greedy algorithm, we must prove that every optimal solution contains our greedy choice.
- Given the following recurrence relation, provide a valid traversal order to fill the DP table or justify why no valid traversal exists.

$$A(i, j) = F(A(i - 2, j + 1), A(i + 1, j - 2))$$

# Matrix-Chain Multiplication

- Input: a sequence of integers  $l_0, l_1, \dots, l_n$ 
  - $l_{i-1}$  is the number of rows of matrix  $A_i$
  - $l_i$  is the number of columns of matrix  $A_i$
- Output: a order of performing  $n - 1$  matrix multiplications in the **maximum** number of operations to obtain the product of  $A_1 A_2 \dots A_n$



Q: Does optimal substructure still hold?

# Huffman Coding

- In a binary Huffman code, the number of symbols that can be used to form a codeword is 2 (0 and 1). We can use a  $n$ -ary Huffman code, in which the number of symbols that can be used to form a codeword is  $n$ .
- 1) Please prove that the decoding tree which represents an optimal code for a file must be a *full*  $n$ -ary tree, i.e., all non-leaf nodes in the tree have  $n$  children.
- 2) Given the following information for a file, what are the lengths of the codewords for the characters 'a' and 'c' respectively, in a 3-ary Huffman code derived for this file?

Character	a	b	c	d	e	f	g
Frequency	700	400	200	100	1300	2400	100





# Question?

Important announcement will be sent to @ntu.edu.tw mailbox  
& post to the course website

Course Website: <http://ada17.csie.org>

Email: [ada-ta@csie.ntu.edu.tw](mailto:ada-ta@csie.ntu.edu.tw)