

CPE 231 Algorithms

# Dynamic Programming

Dr. Taweechai Nuntawisuttiwong



# Contents

- 1 Dynamic Programming
- 2 Three Basic Examples
- 3 The Knapsack Problem and Memory Functions

# Dynamic Programming

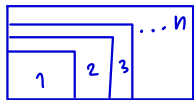
# Dynamic Programming

## Definition

Dynamic Programming (DP) is a technique for solving problems with overlapping subproblems and optimal substructure.

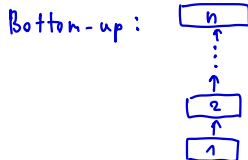
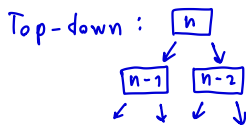
- Introduced by Richard Bellman in the 1950s as a method for optimizing multistage decision processes.
- The term "programming" refers to "planning" rather than computer coding.

# Key Concepts of Dynamic Programming



DP  $\neq$  recursive fn.  
(direct)

- **Overlapping Subproblems**: Solving the same subproblems repeatedly in naive recursive solutions.
- **Memoization**: Storing solutions to subproblems to avoid redundant calculations. recursive but not direct recursive
- Two Approaches:
  - **Top-down** (Memoization): Recursive approach with caching.
  - **Bottom-up** (Tabulation): Iterative approach by solving all smaller subproblems first.



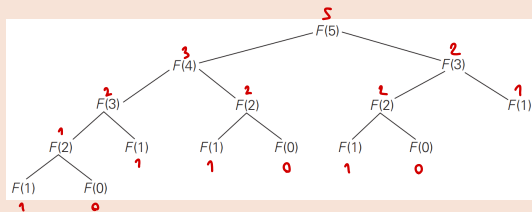
# Example - Fibonacci Sequence

Fibonacci Recurrence:  $F(n) = F(n-1) + F(n-2)$ ,  $n > 1$ .

Initial Conditions:  $F(0) = \frac{1}{0}$ ,  $F(1) = 1$ .

Top-down

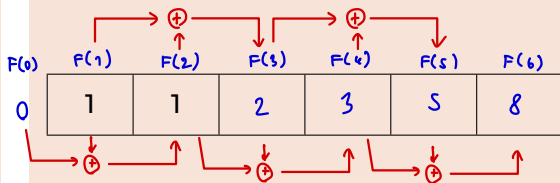
Use recurrence directly



15 A's

Bottom-Up

Use dynamic programming



4 A's

# Applications and Principle of Optimality

- **Principle of Optimality:** Solutions to subproblems contribute to the overall optimal solution.
- **Applications:**
  - Optimization problems like shortest paths in graphs, knapsack, matrix chain multiplication.
  - Real-world examples: Text formatting, image resizing, and engineering optimizations.
- Dynamic Programming is a powerful strategy for solving complex optimization problems efficiently.

shortest path, TSP, VRP



TSP  $\rightarrow$   $n$  cities  $\rightarrow$  feasible solution  $n!$

DP  $\rightarrow$   $n!$  memory unit

# Three Basic Examples



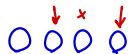
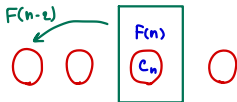
# Coin-Row Problem

- Maximize the amount collected from a row of coins without picking two adjacent coins.
  - Given a row of  $n$  coins, each with a positive integer value  $c_i$ . มีค่าเท่าไรก็ได้
- Define  $F(n)$  as the maximum amount collectible from the first  $n$  coins.

$$F(n) = \max\{c_n + F(n-2), F(n-1)\} \quad \text{for } n > 1$$

หยิบเหรียญที่  $n$  ไปรวมกับค่าของเหรียญก่อนหน้า 2 ครั้ง

- Base cases:  $F(0) = 0, F(1) = c_1$ .
- With last coin: Add  $c_n$  and skip the adjacent coin.
- Without last coin: Take the solution from  $F(n-1)$ .
- Complexity
  - Time:  $O(n)$
  - Space:  $O(n)$



# Coin-Row Problem

## **ALGORITHM** *CoinRow*( $C[1..n]$ )

//Applies formula (8.3) bottom up to find the maximum amount of money  
//that can be picked up from a coin row without picking two adjacent coins

//Input: Array  $C[1..n]$  of positive integers indicating the coin values

//Output: The maximum amount of money that can be picked up

$F[0] \leftarrow 0; \quad F[1] \leftarrow C[1]$

**for**  $i \leftarrow 2$  **to**  $n$  **do**

$F[i] \leftarrow \max(C[i] + F[i - 2], F[i - 1])$

**return**  $F[n]$

# Coin-Row Problem

Example: Coin values [5, 1, 2, 10, 6, 2]

index	0	1	2	3	4	5	6
C		5	1	2	10	6	2
F	0	5					

Base case

index	0	1	2	3	4	5	6
C	⊕	5	1	2	10	6	2
F	0	5	5				

1

$$F[0] = 0, F[1] = 5$$

$$F[2] = \max \{ \overset{C_2}{1} + \overset{F(0)}{0}, \overset{F(1)}{5} \} = 5$$

# Coin-Row Problem

Example: Coin values [5, 1, 2, 10, 6, 2]

index	0	1	2	3	4	5	6
C		5	1	2	10	6	2
F	0	5	5	9			

$$F[3] = \max \{ \overset{C_3 \downarrow}{2} + \overset{F(1) \downarrow}{5}, \underset{F(2) \downarrow}{5} \} = 9$$

index	0	1	2	3	4	5	6
C		5	1	2	10	6	2
F	0	5	5	9	15		

$$F[4] = \max \{ \overset{C_4 \downarrow}{10} + \overset{F(2) \downarrow}{5}, \underset{F(3) \downarrow}{9} \} = 15$$

# Coin-Row Problem

Example: Coin values [5, 1, 2, 10, 6, 2]

index	0	1	2	3	4	5	6
C		5	1	2	10	6	2
F	0	5	5	7	15	15	

index	0	1	2	3	4	5	6
C		5	1	2	10	6	2
F	0	5	5	7	15	15	17

$$F[5] = \max \left\{ \overset{C(5)}{\underset{\downarrow}{6}} + \overset{F(3)}{\underset{\downarrow}{7}}, \underset{\underset{\uparrow F(4)}{15}}{15} \right\} = 15$$

$$F[6] = \max \left\{ \overset{C(6)}{\underset{\downarrow}{2}} + \overset{F(4)}{\underset{\downarrow}{15}}, \underset{\underset{\downarrow F(5)}{15}}{15} \right\} = 17$$

# Change-Making Problem

- Use the minimum number of coins to make a given amount  $n$ .
  - Given a target amount  $n$  and denominations  $d_1 < d_2 < \dots < d_m$ .
- Define  $F(n)$  as the minimum number of coins to make  $n$ .

$$F(n) = \min_{j: n \geq d_j} \{F(n - d_j)\} + 1 \quad \text{for } n > 0,$$

จำนวนเงิน      ค่าของเหรียญ

- Base case:  $F(0) = 0$ .
- Complexity
  - Time:  $O(nm)$
  - Space:  $O(n)$

# Change-Making Problem

**ALGORITHM** *ChangeMaking*( $D[1..m]$ ,  $n$ )

//Applies dynamic programming to find the minimum number of coins  
//of denominations  $d_1 < d_2 < \dots < d_m$  where  $d_1 = 1$  that add up to a  
//given amount  $n$

//Input: Positive integer  $n$  and array  $D[1..m]$  of increasing positive  
//integers indicating the coin denominations where  $D[1] = 1$

//Output: The minimum number of coins that add up to  $n$

$F[0] \leftarrow 0$

**for**  $i \leftarrow 1$  **to**  $n$  **do**

$temp \leftarrow \infty$ ;  $j \leftarrow 1$

**while**  $j \leq m$  **and**  $i \geq D[j]$  **do**

$temp \leftarrow \min(F[i - D[j]], temp)$

$j \leftarrow j + 1$

$F[i] \leftarrow temp + 1$

**return**  $F[n]$

# Change-Making Problem

Example: amount  $n = 6$  and denominations 1, 3, 4.

n	0	1	2	3	4	5	6
F	0						

$$F[0] = 0$$

n	0	1	2	3	4	5	6
F	0	1					

$$F[1] = \min \{ F(1 - \overset{1}{\cancel{1}}) \} + 1 = F(0) + 1 = 1$$

n	0	1	2	3	4	5	6
F	0	1	2				

$$F[2] = \min \{ F(2 - \overset{1}{\cancel{1}}) \} + 1 = F(1) + 1 = 2$$

n	0	1	2	3	4	5	6
F	0	1	2	1			

$$F[3] = \min \{ F(3 - \overset{3,1}{\cancel{1}}) \} + 1 = \min \{ F(0), F(2) \} + 1 = 1$$



# Change-Making Problem

Example: amount  $n = 6$  and denominations 1, 3, 4.

n	0	1	2	3	4	5	6
F	0	1	2	1	1		

n	0	1	2	3	4	5	6
F	0	1	2	1	1	2	

n	0	1	2	3	4	5	6
F	0	1	2	1	1	2	2

$$F[4] = \min_{j \in \{1,3,4\}} \{F(4-j)\} + 1 = \min \{F(0), F(1), F(3)\} + 1 \\ = F(0) + 1 = 1$$

$$F[5] = \min_{j \in \{1,3,4\}} \{F(5-j)\} + 1 = \min \{F(1), F(2), F(4)\} + 1 \\ = F(4) + 1 = 2$$

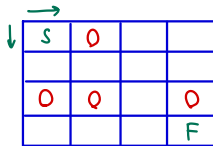
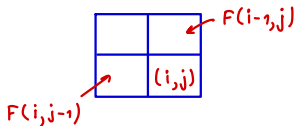
$$F[6] = \min_{j \in \{1,3,4\}} \{F(6-j)\} + 1 = \min \{F(2), F(3), F(5)\} + 1 \\ = F(3) + 1 = 2$$

# Coin-Collecting Problem

- Collect the maximum number of coins on an  $n \times m$  grid by moving only right or down.
  - A grid with coins in specific cells (either 0 or 1).
- Define  $F(i, j)$  as the maximum coins collectible to reach cell  $(i, j)$ .

$$F(i, j) = \max\{F(i-1, j), F(i, j-1)\} + c_{ij} \quad \text{for } 1 \leq i \leq n, 1 \leq j \leq m,$$

- Base cases:  $F(0, j) = 0$  for  $1 \leq j \leq m$ ,  $F(i, 0) = 0$  for  $1 \leq i \leq n$ .
- The robot moves from the cell above or left.
- Complexity
  - Time:  $O(nm)$
  - Space:  $O(nm)$



# Coin-Collecting Problem

**ALGORITHM** *RobotCoinCollection*( $C[1..n, 1..m]$ )

//Applies dynamic programming to compute the largest number of  
//coins a robot can collect on an  $n \times m$  board by starting at (1, 1)

//and moving right and down from upper left to down right corner

//Input: Matrix  $C[1..n, 1..m]$  whose elements are equal to 1 and 0

//for cells with and without a coin, respectively

//Output: Largest number of coins the robot can bring to cell  $(n, m)$

$F[1, 1] \leftarrow C[1, 1];$    **for**  $j \leftarrow 2$  **to**  $m$  **do**  $F[1, j] \leftarrow F[1, j - 1] + C[1, j]$

**for**  $i \leftarrow 2$  **to**  $n$  **do**

$F[i, 1] \leftarrow F[i - 1, 1] + C[i, 1]$

**for**  $j \leftarrow 2$  **to**  $m$  **do**

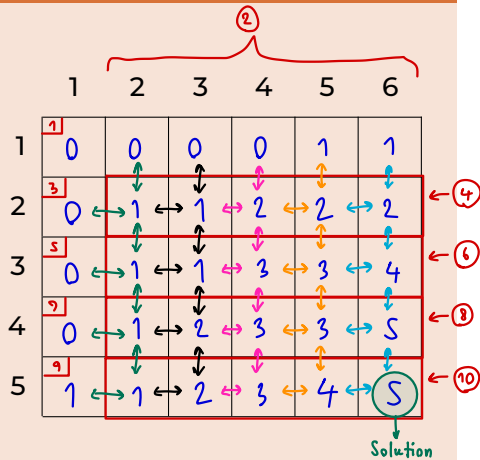
$F[i, j] \leftarrow \max(F[i - 1, j], F[i, j - 1]) + C[i, j]$

**return**  $F[n, m]$

# Coin-Collecting Problem

## Example:

	1	2	3	4	5	6
1					●	
2		●		●		
3				●		●
4			●			●
5	●				●	



# Coin-Collecting Problem

## Example:

Tracing:

1)  $F(i-1, j) > F(i, j-1)$

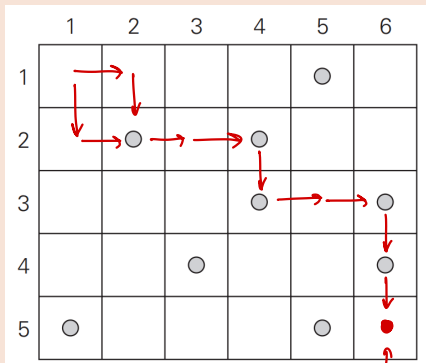
$v_u > \tilde{w}_{i-1}$  ↓  
(i, j)

2)  $F(i-1, j) < F(i, j-1)$

$v_u < \tilde{w}_{i-1}$  → (i, j)

3)  $F(i-1, j) = F(i, j-1)$

↓  
→ (i, j)



Start tracing

# The Knapsack Problem and Memory Functions

# Knapsack Problem

- Find the subset of items that maximizes value without exceeding the capacity.
  - Given  $n$  items with weights  $w_1, w_2, \dots, w_n$  and values  $v_1, v_2, \dots, v_n$ , and a knapsack of capacity  $W$ .
- Define  $F(i, j)$ : Maximum value achievable with the first  $i$  items and capacity  $j$ .

$$F(i, j) = \begin{cases} \max\{F(i-1, j), v_i + F(i-1, j-w_i)\} & \text{if } j - w_i \geq 0, \\ F(i-1, j) & \text{if } j - w_i < 0. \end{cases}$$

Handwritten notes in Thai:

- For  $F(i-1, j)$ : ไม่เพิ่ม (Don't add)
- For  $v_i + F(i-1, j-w_i)$ : เพิ่ม item  $i$  (Add item  $i$ )
- For the condition  $j - w_i \geq 0$ : น้ำหนักไม่เกิน (Weight not exceeded)
- For the condition  $j - w_i < 0$ : น้ำหนักเกิน (Weight exceeded)
- For the second branch  $F(i-1, j)$ : ไม่ใส่ของเพิ่ม (Don't add more items)

- Base cases:  $F(0, j) = 0$  for  $j \geq 0$ ,  $F(i, 0) = 0$  for  $i \geq 0$ .
- Case 1: If the item fits, compare the value including vs. excluding it.
- Case 2: If the item does not fit, skip it.
- Complexity
  - Time:  $O(nW)$
  - Space:  $O(nW)$
  - Time (find the composition of an optimal solution):  $O(n)$

# Knapsack Problem

Example: Consider the instance given by the following data

capacity  $W = 5$ .

item	weight	value
1	2	\$12
2	1	\$10
3	3	\$20
4	2	\$15



# Knapsack Problem

→ solution path

Example:

$$F(i, j) = \begin{cases} \max\{F(i-1, j), v_i + F(i-1, j-w_i)\} & \text{if } j-w_i \geq 0, \\ F(i-1, j) & \text{if } j-w_i < 0. \end{cases}$$

*Handwritten notes:*  
 - Above the first max: "ไม่เลือก item i" (do not choose item i)  
 - Above the second max: "เลือก item i" (choose item i)  
 - Below the first max: "ถ้าเลือกแล้ว" (if chosen)  
 - Below the second max: "ถ้าไม่เลือกแล้ว" (if not chosen)  
 - To the right of the first max: "ถ้าเลือกแล้วไม่เต็ม" (if chosen and not full)  
 - To the right of the second max: "ถ้าเลือกแล้วเต็ม" (if chosen and full)

item :  $S = \{1, 2, 4\}$

capacity  $j$

	i	0	1	2	3	4	5
	0	0	0	0	0	0	0
$w_1 = 2 \quad v_1 = 12$	1	0	0	12	12	12	12
$w_2 = 1 \quad v_2 = 10$	2	0	10	12	22	22	22
$w_3 = 3 \quad v_3 = 20$	3	0	10	12	22	30	32
$w_4 = 2 \quad v_4 = 15$	4	0	10	15	25	30	39

*Handwritten annotations on the table:*  
 - Orange arrows and boxes:  $F(2-1, 2-1) + 10$  (from 0 to 10),  $F(3-1, 2-1)$  (from 12 to 12),  $+10$  (from 10 to 20),  $+20$  (from 10 to 30),  $+15$  (from 10 to 25).  
 - Green arrows and boxes:  $+10$  (from 12 to 22),  $+20$  (from 12 to 32),  $+15$  (from 12 to 27),  $+15$  (from 15 to 30).  
 - The value 39 in the bottom-right cell is boxed in green and labeled "Solution".  
 - To the right of the 39, a red double-headed arrow indicates  $39 > 32$  and  $4 \in S$ .

# Memory Functions

- Reduce unnecessary computations in dynamic programming by combining top-down and bottom-up methods.
- Memory Function Approach: Use a top-down recursion with memoization.
  - Check the table before calculating a subproblem.
  - Compute only if not already solved, then store in the table.
- Avoid solving unneeded subproblems and reduce recursive calls.

# Memory Functions

**ALGORITHM**  $MFKnapsack(i, j)$ 

```
//Implements the memory function method for the knapsack problem
```

//Input: A nonnegative integer  $i$  indicating the number of the first

// items being considered and a nonnegative integer  $j$  indicating

// the knapsack capacity

//Output: The value of an optimal feasible subset of the first  $i$  items

//Note: Uses as global variables input arrays *Weights*[1..n], *Values*[1..n],

//and table  $F[0..n, 0..W]$  whose entries are initialized with  $-1$ 's except for

```
//row 0 and column 0 initialized with 0's
```

**if**  $F[i, j] < 0$ **if**  $j < Weights[i]$ 
$$value \leftarrow MFKnapsack(i - 1, j)$$

else

$$value \leftarrow \max(MFKnapsack(i - 1, j), \\ Values[i] + MFKnapsack(i - 1, j - Weights[i]))$$
$$F[i, j] \leftarrow value$$
**return**  $F[i, j]$

# Memory Functions

## Example:

