

Dynamic Programming

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Dynamic Programming

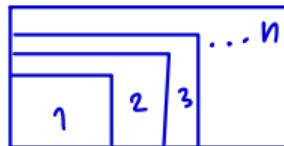
Dynamic Programming

Definition

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

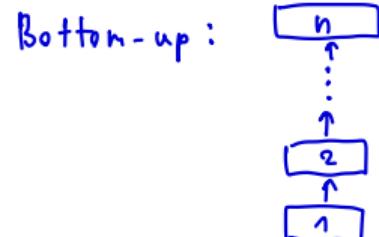
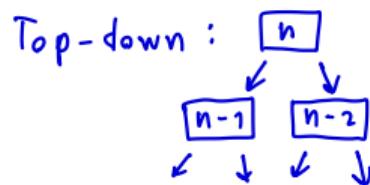
- Introduced by Richard Bellman in the 1950s as a method for optimizing multistage decision processes.
- The term "programming" refers to "planning" rather than computer coding.

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.



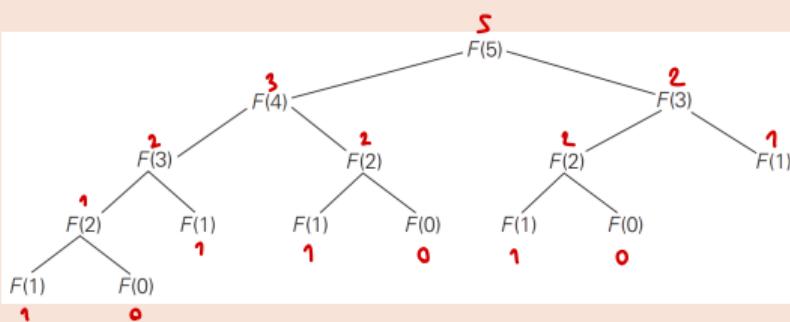
Example - Fibonacci Sequence

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

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

Top-down

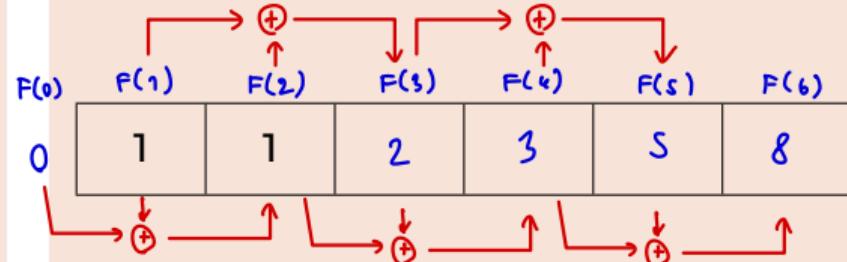
Use recurrence directly



15 Ažj

Bottom-up

Use dynamic programming



4 Ažj

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 → n cities → feasible solution $n!$

DP → $n!$ 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 .
- Define $F(n)$ as the maximum amount collectible from the first n coins.

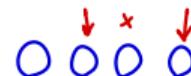
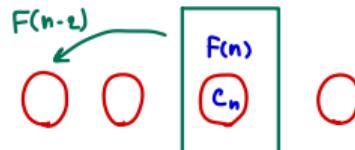
$$F(n) = \max\{c_n + F(n - 2), F(n - 1)\} \quad \text{for } n > 1$$

ເພື່ອມີຄວາມສົບສັນດິນໃຫຍ່ ໃຫຍ່ມີຄວາມສົບສັນດິນໃຫຍ່

- 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

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

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					

Basic case

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

1

$$F[0] = 0, F(1) = S$$

$$F[2] = \max \{ C_0 + F(0), F(1) \} > S$$

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			

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

$\downarrow C_3$ $\downarrow F(1)$
 \downarrow \downarrow
 $\downarrow F(2)$

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

$$F[4] = \max \{C_4 + F(2), F(3)\} = 15$$

$\downarrow C_4$ $\downarrow F(2)$
 \downarrow \downarrow
 $\downarrow F(3)$

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	

$$F[5] = \max \left\{ \begin{matrix} C(5) & F(3) \\ \downarrow & \downarrow \\ 6+7 & 15 \\ \uparrow & \\ F(4) \end{matrix} \right\} = 15$$

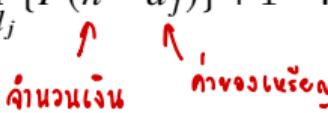
index	0	1	2	3	4	5	6
C		5	1	2	10	6	2
F	0	5	5	7	15	15	17

$$F[6] = \max \left\{ \begin{matrix} C(6) & F(4) \\ \downarrow & \downarrow \\ 2+15 & 15 \\ \uparrow & \\ F(5) \end{matrix} \right\} = 17$$

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 .

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


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- 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						

n	0	1	2	3	4	5	6
F	0	1					

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

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

$$F[0] = 0$$

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

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

$$\begin{aligned} F[3] &= \min \{ F(3 - d_j) \} + 1 = \min \{ F(0), F(2) \} + 1 \\ &= 1 \end{aligned}$$

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

$$F[4] = \min \left\{ F(4 - 1) \atop^{1,3,4} \right\} + 1 = \min \{ F(0), F(1), F(3) \} + 1 \\ = F(0) + 1 = 1$$

$$F[5] = \min \left\{ F(5 - 1) \atop^{1,3,4} \right\} + 1 = \min \{ F(1), F(2), F(4) \} + 1 \\ = F(4) + 1 = 2$$

$$F[6] = \min \left\{ F(6 - 1) \atop^{1,3,4} \right\} + 1 = \min \{ F(2), F(3), F(5) \} + 1 \\ = F(3) + 1 = 2$$

Coin-Collecting Problem

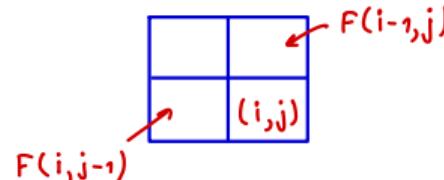
- Collect the maximum number of coins on an $n \times m$ grid by moving only right or down.
 - A grid with coins in specific cells (either 0 or 1).
- Define $F(i, j)$ as the maximum coins collectible to reach cell (i, j) .

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

- 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: $O(nm)$



	0	0	0	
	0	0	0	
				F

Coin-Collecting Problem

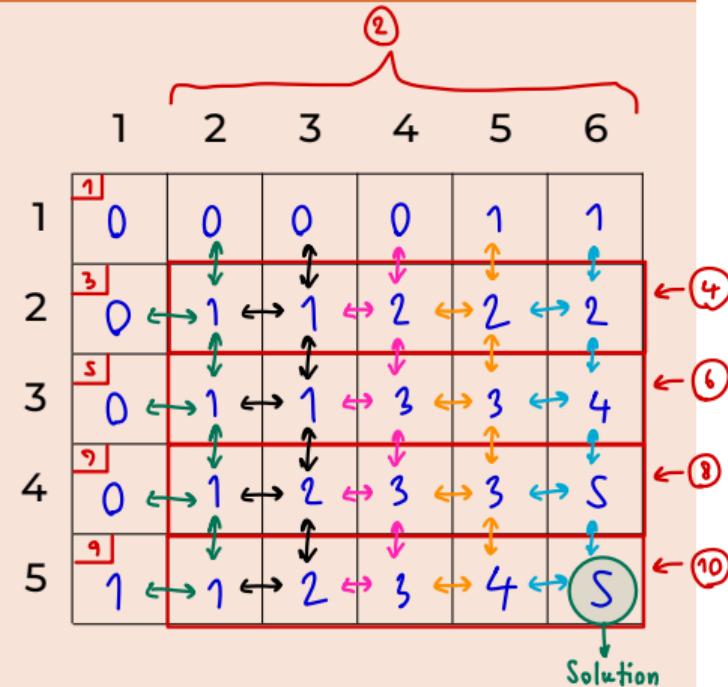
ALGORITHM *RobotCoinCollection(C[1..n, 1..m])*

```
//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
				●	
	●		●		
		●			●
		●			●
5	●			●	



Coin-Collecting Problem

Example:

Tracing :

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

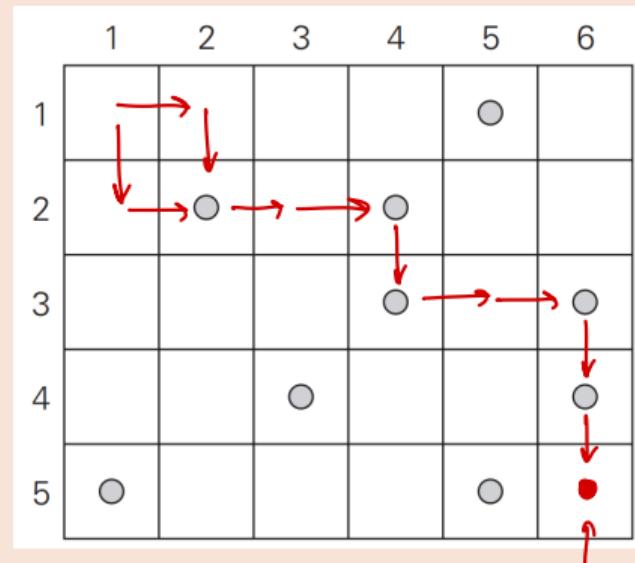
$v_{i,j} > v_{i-1,j}$ ↓
 $\rightarrow (i,j)$

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

$v_{i,j} < v_{i-1,j}$ ← → (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 .

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

ယင်းနှင့် မြတ်သွေးနိုင်သူများ
ယင်းနှင့် မြတ်သွေးနိုင်ဘဲမြတ်သွေးနိုင်ဘူး

- 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:

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

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

capacity j

i	0	1	2	3	4	5
w ₁ = 2 v ₁ = 12	0	0	0	0	0	0
w ₂ = 1 v ₂ = 10	1	0	12	12	12	12
w ₃ = 3 v ₃ = 20	2	0	12	22	22	22
w ₄ = 2 v ₄ = 15	3	0	12	22	30	32
	0	10	15	25	30	39

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:

