

# SC1007

## Structures and Algorithms

# Dynamic Programming



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# Dynamic Programming

*by Richard Ernest Bellman in 1953*

# What is Dynamic Programming?

- It is not a programming language like C
  - The term “Programming” refers to a tabular method (filling tables)
    - It is applied to optimization problems
      - Other “programming” methods in mathematical optimization are
        - Linear Programming
        - Integer Programming
        - Convex Programming
        - Semidefinite Programming
      - not related to coding
- Applied from system control to economics

# What is Dynamic Programming (DP)?

- It is similar to divide-and-conquer strategy
  - Breaking the big problem into sub-problems
  - Solve the sub-problems recursively
  - Combining the solutions to the sub-problems
- What is the difference between them?
  - DP can be applied when the sub-problems are not independent
    - Every sub-problem is solved once and is saved in a table
  - The problem usually can have multiple optimal solutions
    - DP may just return one of them

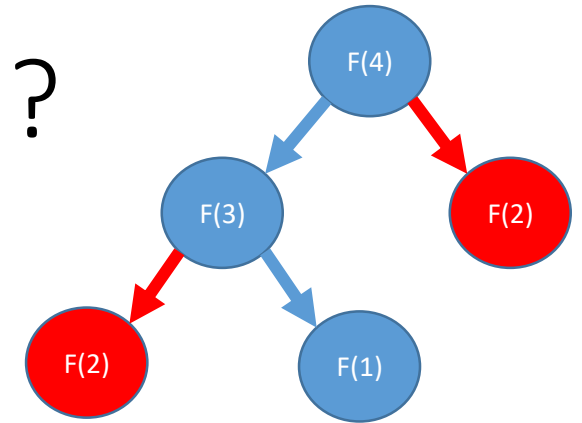
# What is Dynamic Programming (DP)?

- Optimal substructure
  - Combination of optimal solutions to its sub-problems
- Overlapping sub-problems
  - Having the same sub-problems

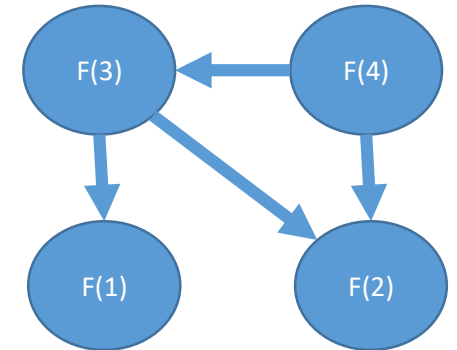
➤ Fibonacci Series:  $F_i = F_{i-1} + F_{i-2}$

- Recursion: problem can be solved recursively
- **Memoization**: Store optimal solutions to sub-problems in table (or memory or cache)  
=> If the sub-problems are independent, DP is not useful!

Dynamic Programming = Recursion + Memoization

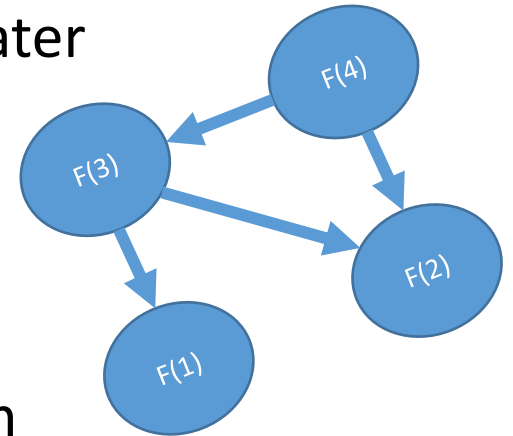


$$\Theta(2^n) \Rightarrow \Theta(n^p)$$



# Dynamic Programming Approaches

- Top-down approach
  - Recursively using the solution to its sub-problems
  - Memoize the solutions to the sub-problems and reuse them later
- Bottom-up approach
  - Figure out the order of calculation
  - Solve the sub-problems to build up solutions to larger problem



# Fibonacci: Top-down approach

```
Fib(n)
{
    if (n == 0)
        M[0] = 0; return 0;
    if (n == 1)
        M[1] = 1; return 1;

    if (M[n-1] == -1)           //F(n-1) was not calculated
        M[n-1] = Fib(n-1)    //calculate F(n-1) and store in M

    if (M[n-2] == -1)           //F(n-2) was not calculated
        M[n-2] = Fib(n-2)    //calculate F(n-2) and store in M

    M[n] = M[n-1] + M[n-2]
    return M[n];
}
```

Store an array M

0	1	2	3	4	5	6
-1	-1	-1	-1	-1	-1	-1

**Complexity:  $O(n)$**

# Fibonacci: Bottom-up approach

```
Fib(n)
{
    M[0] = 0;
    M[1] = 1;

    int i = 0;
    for (i = 2; i <= n; i++)
        M[i] = M[i-1] + M[i-2];
    return M[n];
}
```

Store an array M

0	1	2	3	4	5	6
-1	-1	-1	-1	-1	-1	-1

**Complexity:  $O(n)$**

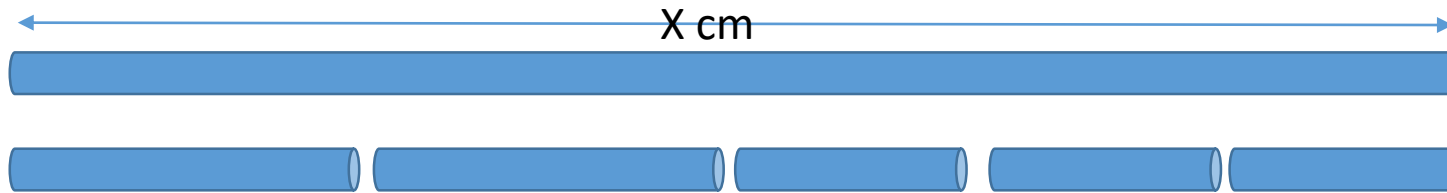


# Examples of DP

- String algorithms like longest common subsequence, longest increasing subsequence, longest common substring etc.
- Graph algorithms like Bellman-Ford algorithm, Floyd's algorithm
- Chain matrix multiplication
- Rod Cutting
- 0/1 Knapsack
- Travelling salesman problem
- Subset Sum

# Rod Cutting Problem

Given a rod of a certain length and price of rod of different lengths, determine the maximum revenue obtainable by cutting up the rod at different lengths based on the prices.



Length cm	1	2	3	4	5	6	7	8	9
Price \$	1	5	8	9	10	17	17	20	24

# Rod Cutting Problem

Length cm	1	2	3	4	5	6	7	8	9
Price \$	1	5	8	9	10	17	17	20	24

If a rod of length 4,

Length of each piece	Total Revenue
4	9
1 + 3	1+8 = 9
1 + 1 + 2	1+1+5 = 7
1 + 1 + 1 + 1	1+1+1+1=4
<b>2 + 2</b>	<b>5+5 =10</b>

From all possible solutions, the maximum revenue is 10 by cutting the rod into two pieces of length 2 each.

# Naïve Top-down Recursive Approach

**Cut-Rod** ( $p, n$ )

**begin**

**if**  $n == 0$

**return** 0

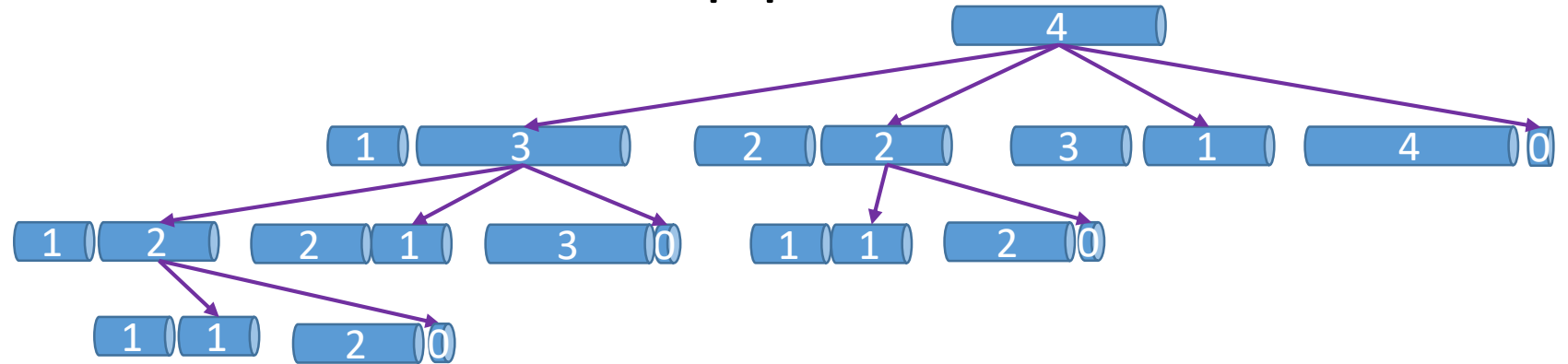
$q \leftarrow -\infty$

**for**  $i = 1$  to  $n$  do

$q \leftarrow \mathbf{max} (q, p[i] + \mathbf{Cut-Rod}(p, n-i) )$

**return**  $q$

**end**



The recursive calls will repeatedly find the revenue for a rod of the same length. Its time complexity is  $\Theta(2^n)$

# Top-down Memoized Approach

- The result of each sub-problem is stored and reused

```
Cut-Rod (p, n)
begin
    r[1, ..., n] ← {0}
    return Mem-Cut-Rod-Aux(p, n, r)
end
```

```
Mem-Cut-Rod-Aux (p, n, r)
begin
    if n==0
        return 0
    if (r[n]>0)
        return r[n]
    else
        q ← -∞
        for i = 1 to n do
            q ← max (q, p[i] + Mem-Cut-Rod-Aux(p, n-i, r))
        r[n] ← q
    return q
end
```

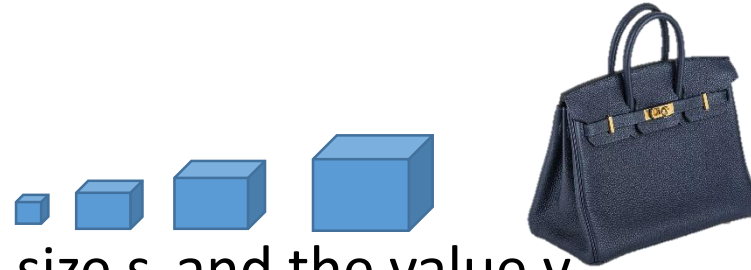
# Bottom-up DP Approach

```
DP-Cut-Rod (p,n)
begin
    r[1, ..., n] ← {0}
    for j = 1 to n do
        for i = 1 to j do
            r[j] ← max (r[j], p[i] + r[j-i])
    return r[n]
end
```

- The bottom-up and top-down versions has the same asymptotic running time,  $\Theta(n^2)$

Length cm	1	2	3	4	5	6	7	8	9
Price \$	1	5	8	9	10	17	17	20	24
Max Rev \$	1	5	8	10	13	17	18	22	25

# 0/1 Knapsack



- Given  $n$  items, where the  $i^{\text{th}}$  item has the size  $s_i$  and the value  $v_i$
- Put these items into a knapsack of capacity  $C$
- *Optimization problem: Find the largest total value of the items that fits in the knapsack*

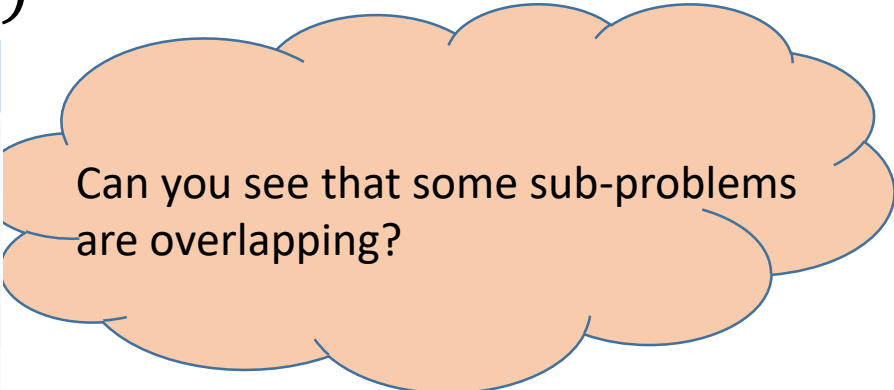
$$\begin{aligned} & \max_x \sum_{i=1}^n v_i x_i \\ & \text{Subject to} \\ & \sum_{i=1}^n s_i x_i \leq C \\ & x_i \in \{0,1\} \quad i = 1, 2, \dots, n \end{aligned}$$

# 0/1 Knapsack

- Brute-force algorithm
- The  $i^{\text{th}}$  item is either included (1) or excluded (0)
- The time complexity of the algorithm is  $\Theta(2^n)$

Item 1	Item 2	Item 3	Value
0	0	0	0
0	0	1	V3
0	1	0	V2
0	1	1	V2+V3
1	0	0	V1
1	0	1	V1+V3
1	1	0	V1+V2
1	1	1	V1+V2+V3

$$\begin{aligned} & \max_x \sum_{i=1}^n v_i x_i \\ & \text{Subject to} \\ & \sum_{i=1}^n s_i x_i \leq C \\ & x_i \in \{0,1\} \quad i = 1, 2, \dots, n \end{aligned}$$



Can you see that some sub-problems are overlapping?



# Using DP to solve 0/1 Knapsack

- The recursive formula

- $$M(i, j) = \max\{ \underbrace{M(i-1, j)}_{\text{ith item is unused}}, \underbrace{M(i-1, j - s_i) + v_i}_{\text{ith item is used}} \}$$

- $i = 1, \dots, n$

- $j = 1, \dots, C$

	1	2	3	...	C
1	0	0	$v_1$	$M(i-1, j)$	$v_1$
2	0	$v_2$	$v_1$	$v_1 + v_2$	$v_1 + v_2$
...	0		$M(i, j)$		
n					

The capacity of knapsack is 5kg. ( $C = 5$ )

Item	Weight	Value
1	2kg	\$12
2	1kg	\$10
3	3kg	\$20
4	2kg	\$15

	1	2	3	4	5
1	\$0	\$12	\$12	\$12	\$12
2	\$10	\$12	\$22	\$22	\$22
3	\$10	\$12	\$22	\$30	\$32
4	\$10	\$15	\$25	\$30	<b>\$37</b>

# Using DP to solve 0/1 Knapsack

- The recursive formula

- $$M(i, j) = \max\{\underbrace{M(i-1, j)}_{\text{ith item is unused}}, \underbrace{M(i-1, j - s_i) + v_i}_{\text{ith item is used}}\}$$

- $i = 1, \dots, n$

- $j = 1, \dots, C$

- Create a n-by-C matrix, M

- All the possible sizes from 1 to C

- Bottom up approach

- Time Complexity is  $\Theta(nC)$

	j →					
	1	2	3	...	C	
1	0	0	$v_1$	$M(i-1, j)$	$v_1$	
2	0	$v_2$	$v_1$	$v_1 + v_2$	$v_1 + v_2$	
...	0		$M(i, j)$			
n						

# Summary

- Dynamic Programming
  - Rod Cutting Problem
  - 0/1 Knapsack Problem