

SC1007

Data Structures and Algorithms

Week 11: Dynamic Programming



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Fibonacci sequence

- Let's consider the calculation of **Fibonacci** numbers:

$$F(n) = F(n-1) + F(n-2)$$

with seed values $F(0) = 0$, $F(1) = 1$

- The sequence look like:

0, 1, 1, 2, 3, 5, 8, 13, 21, 34, 55, 89, 144, ...

Fibonacci sequence: recursive algorithm

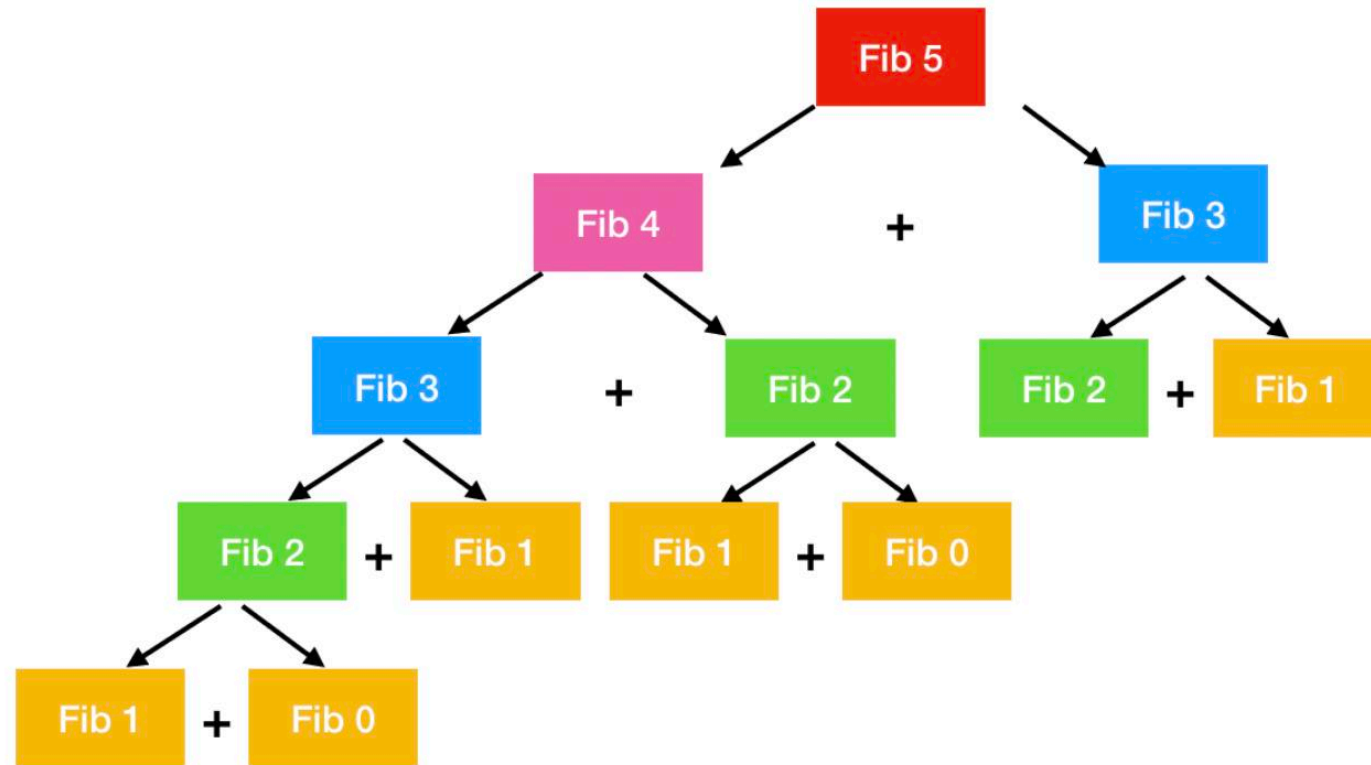
```
Fib(n)
{
    if (n == 0)
        return 0;

    if (n == 1)
        return 1;

    return Fib(n-1) + Fib(n-2);
}
```

It has a serious issue!

Recursion tree



What's the problem?

Many subproblems are overlapping: a lot of recomputation

Complexity: $O(2^n)$

What is Dynamic Programming (DP)?

Dynamic Programming = Recursion + Memoization

- Recursion: problem can be solved recursively
- **Memoization**: Store optimal solutions to sub-problems in table (or memory or cache)

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
- If the sub-problems are independent, DP is not useful!

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 |

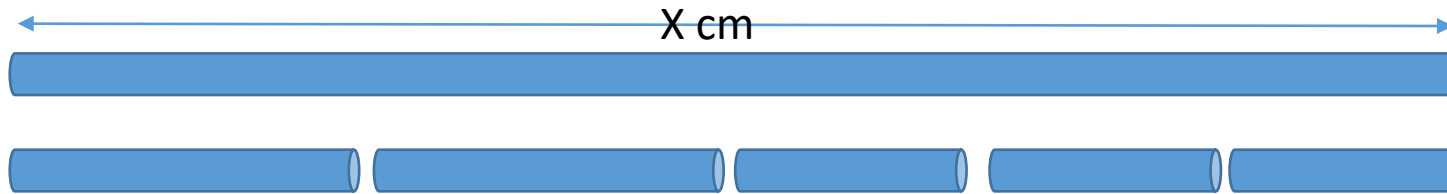
Complexity: O(n)

Other 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

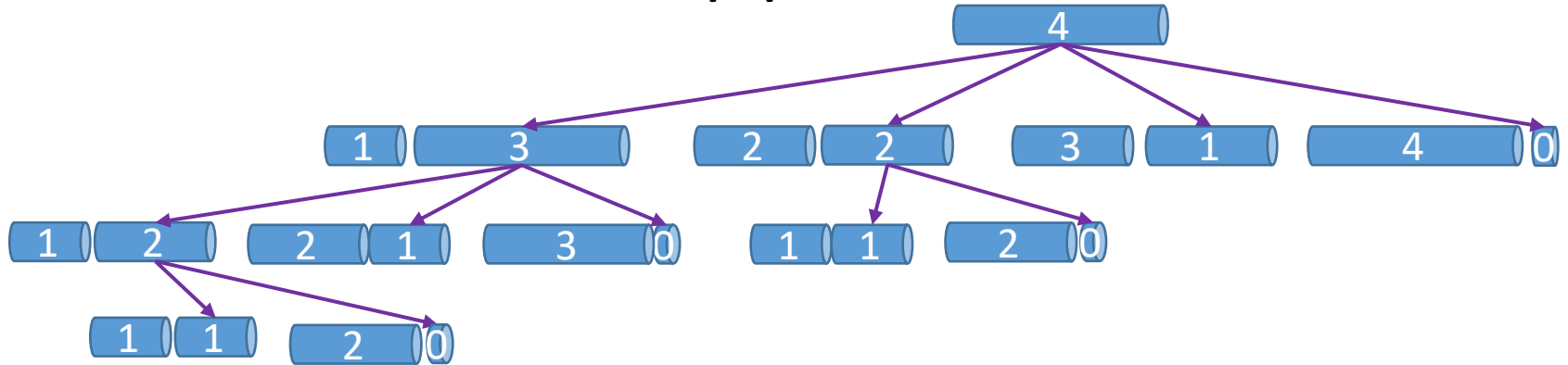
| Length cm | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 |
|-----------|---|---|---|---|----|----|----|----|----|
| Price \$ | 1 | 5 | 8 | 9 | 10 | 17 | 17 | 20 | 24 |

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 |
| 2 + 2 | 5+5 =10 |

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 ← - ∞
  for i = 1 to n do
    q ← max (q, p[i] + 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
```

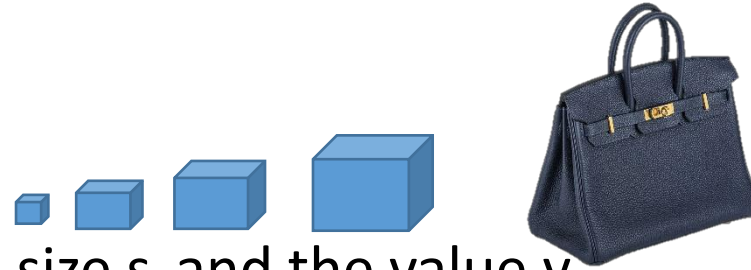
Bottom-up DP Approach

```
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^{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*

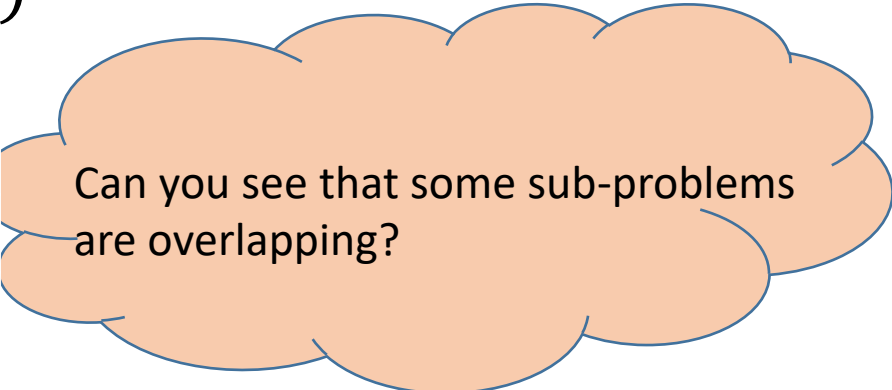
$$\begin{aligned} & \max_x \sum_{i=1}^n v_i x_i \\ & \text{Subject to} \\ & \sum_{i=1}^n s_i x_i \leq C \\ & x_i \in \{0,1\} \quad i = 1, 2, \dots, n \end{aligned}$$

0/1 Knapsack

- Brute-force algorithm
- The i^{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?

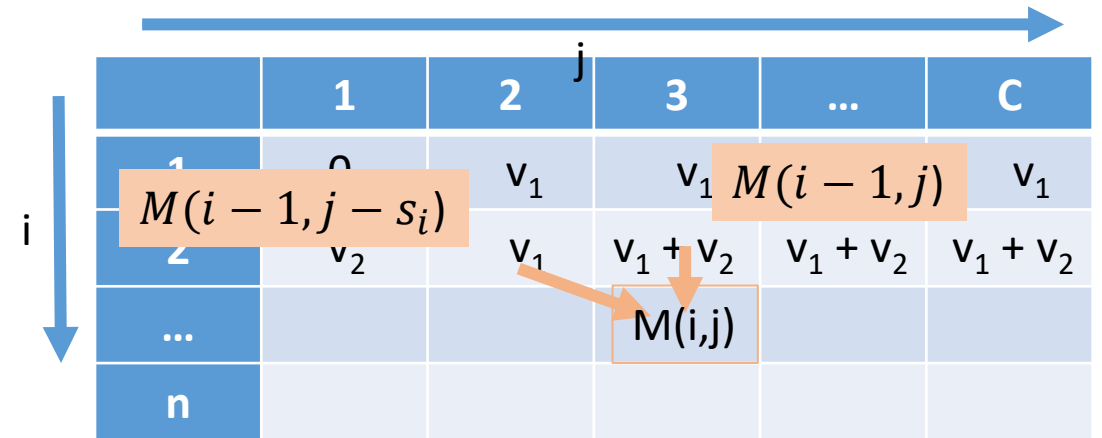
Using DP to solve 0/1 Knapsack

- The recursive formula

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

- $i = 1, \dots, n$

- $j = 1, \dots, C$



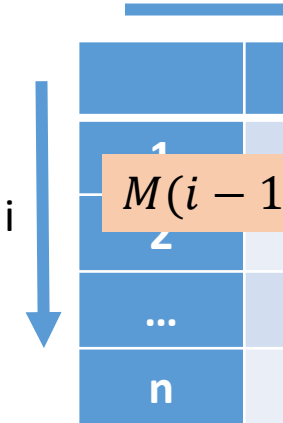
The capacity of knapsack is 5kg. ($C = 5$)

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

Diagram illustrating the DP table structure for the 0/1 Knapsack problem with capacity 5. The table has rows indexed by item i (1 to 4) and columns indexed by capacity j (1 to 5). The value of the optimal solution for item i and capacity j is $M(i, j)$. The diagram shows the recurrence relation: $M(i, j) = \max\{M(i-1, j), M(i-1, j-s_i) + v_i\}$. Arrows indicate the flow of information from the previous row and the previous column to the current cell.

| $i \backslash j$ | 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 | \$37 |

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 $\Theta(nC)$
- 

The diagram illustrates a dynamic programming table for sequence alignment. The table is indexed by i (rows) and j (columns). The row indices are $1, 2, \dots, n$, and the column indices are $1, 2, 3, \dots, C$. The cell $M(i, j)$ is highlighted, and its dependencies are shown as $M(i-1, j)$ and $M(i-1, j-s_i)$.

| | | 1 | 2 | 3 | ... | C |
|-----|-----------------|---|-------|-----------|-------------|-------------|
| 1 | $M(i-1, j-s_i)$ | 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

