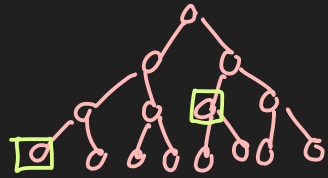


Dynamic Programming

Divide & Conquer + Lookup Table *ส่วนไหน, เป็น array*

Overview




แก้แล้วไม่ทำซ้ำ (Dynamic Programming)

- The key idea of **Divide & Conquer** is to **break** a problem into smaller sub-problems and **combine** the result of those subproblems
- Some Problem can be divided into subproblems that is **overlapping**, i.e., same subproblem that happens **more than once**
 - If we use general **D&C**, each **copies** of the same subproblem will **be solved repeatedly**, wasting time
 - **Dynamic Programming** is a technique that **use a look up table** to store result of each sub-problem and immediately **use it** if any subproblem is required **multiple times**

overlapping
sub problem

Fibonacci Number

Fibonacci Number

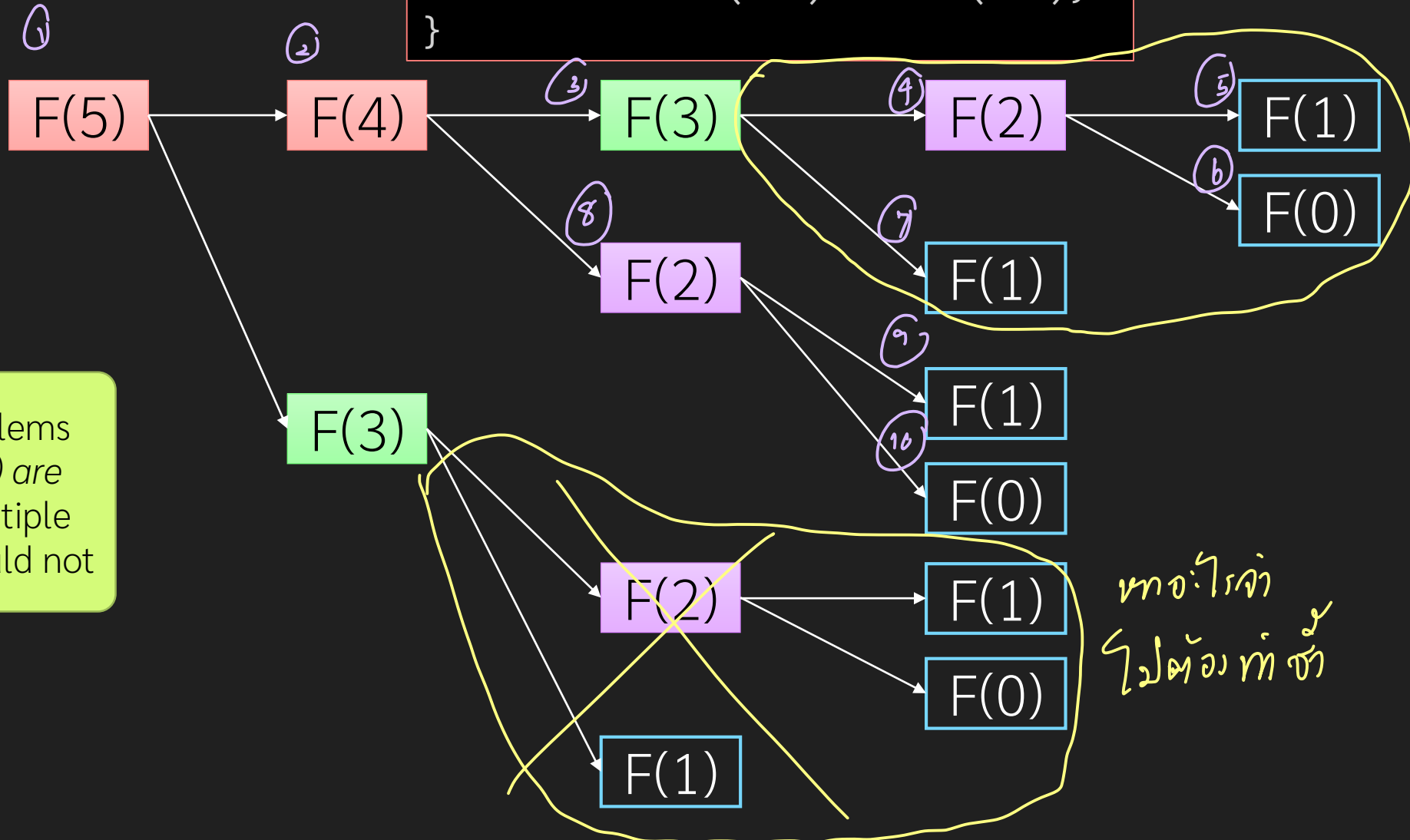
- **Problem:** compute $F(N)$, the Fibonacci function of N
- **Input:**
 - An integer $N \geq 0$
- **Output:**
 - $F(N)$, according to 

$$F(N) = \begin{cases} F(N-1) + F(N-2) & ; n > 1 \\ 1 & ; n = 1 \\ 0 & ; n = 0 \end{cases}$$

Can be solve directly using Divide & Conquer

Recursion Tree

```
int fibo(int n) {  
    if (n == 1 || n == 0)  
        return n;  
    if (n >= 2)  
        return fibo(n-1) + fibo(n-2);  
}
```



Some subproblems
($F(3)$ and $F(2)$) are
computed multiple
times, they should not

บางโหนด
ไม่จำเป็นต้องทำซ้ำ

Memoization: Simplest form of Dynamic Programming

- Top-Down approach
- Remember what have been done, if the subproblem is needed again, use the remembered result

```
ResultType DC(Problem p) {  
    if (p is trivial) {  
        solve p directly  
        return the result  
    } else {  
  
        divide p into  $p_1, p_2, \dots, p_n$   
        for (i = 1 to n)  
             $r_i = DC(p_i)$   
        combine  $r_1, r_2, \dots, r_n$  into r  
  
        return r  
    }  
}
```

```
ResultType DP(Problem p) {  
    if (p is trivial) {  
        solve p directly  
        return the result  
    } else {  
        if p is solved  
            return table.lookup(p);  
        divide p into  $p_1, p_2, \dots, p_n$   
        for (i = 1 to n)  
             $r_i = DP(p_i)$   
        combine  $r_1, r_2, \dots, r_n$  into r  
        table.save(p, r);  
        return r  
    }  
}
```

Fibonacci: Top-Down DP

↑: make global

- `table` is an array[1..n] initialized by 0

```
int fibo_memo(int n) {  
    if (n == 1 || n == 0) ↑: make global table[n] = 0  
        return n;  
  
    if (n >= 2) {  
        if (table[n] > 0) {  
            return table[n]; use  
        }  
        int value = fibo_memo(n-1) + fibo_memo(n-2);  
        table[n] = value; remember  
        return value;  
    }  
}
```

Exercise

- Draw recursion tree when we call `fibonacci_memo(7)` = 13

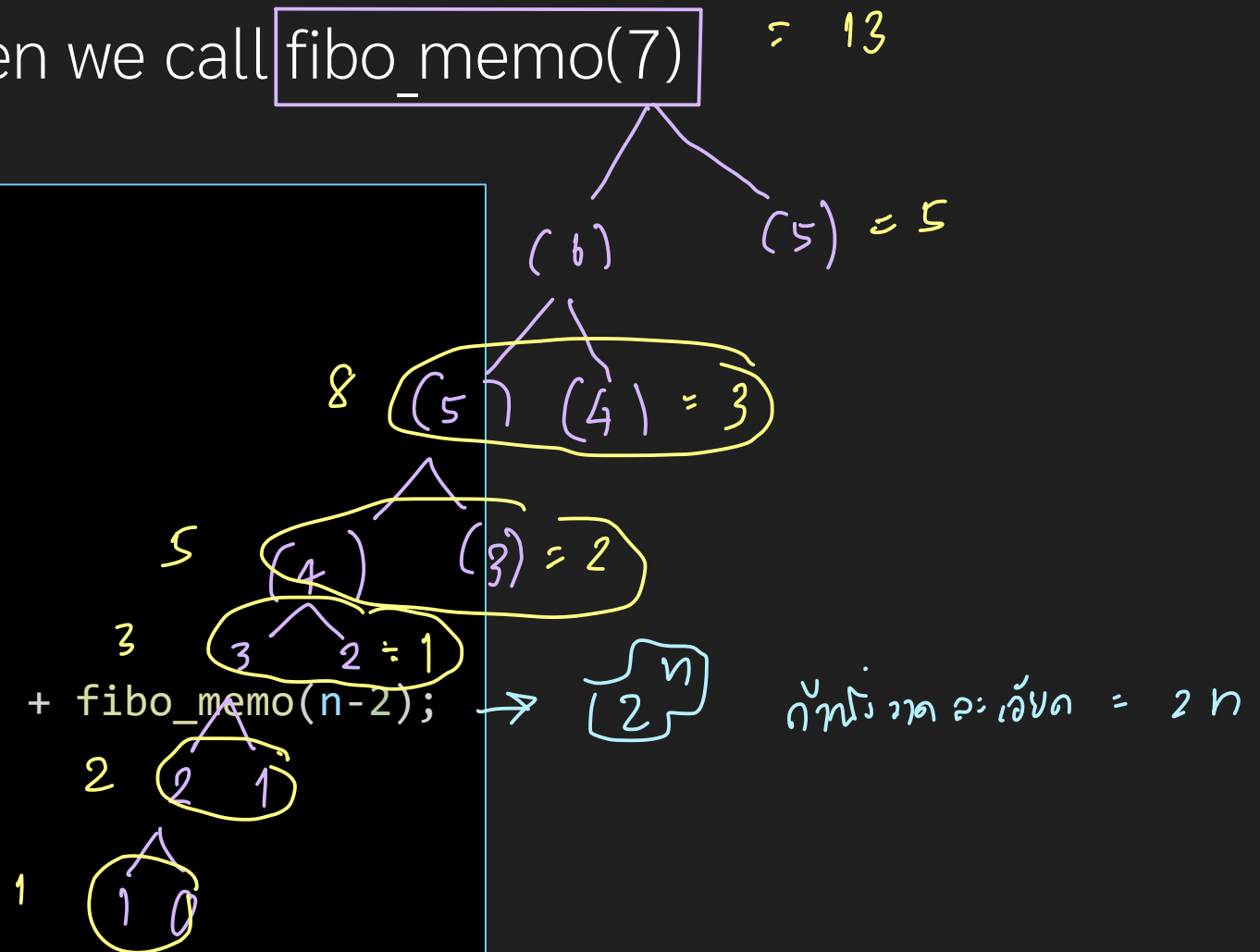
//table is a global variable

```
int fibonacci_memo(int n) {  
    if (n == 1 || n == 0)  
        return n;
```

```
    if (n >= 2) {  
        if (table[n] > 0) {  
            return table[n];  
        }  
    }
```

```
    int value = fibonacci_memo(n-1) + fibonacci_memo(n-2);  
    table[n] = value;  
    return value;
```

```
}
```



Top-down recursion DC 100%

Bottom-up dynamic programming

- Instead of relying on recursion to discover repetition of subproblems, we analyze the recursion directly and **build table constructively** from smaller subproblems
 - The **initial subproblems** are the ones from **trivial case** of Divide & Conquer recurrent relation
- **Benefit:** no-recursion, better runtime performance, (usually) easier to analyze
 - ข้อดี: ไร้การซ้ำ
 - ข้อเสีย: เขียนตารางเก็บค่าที่เยอะ
 - วิเคราะห์ loop
- **Drawback:** sometime, we build unnecessary sub-problem

Fibonacci: Bottom-Up DP

- From the definition of $F(N)$, we know that
 - $F(n)$ **needs** to know $F(n-1)$ and $F(n-2)$
 - In other words, if we know $F(n-1)$ and $F(n-2)$, then **we can construct $F(N)$**
- Initial Condition:
 - $F(0) = 0, F(1) = 1$
 - i.e., `table[0] = 0; table[1] = 1;`
`table[2] : [1] + [0]`
- From the recurrent
 - `table[3] = table[2] + table[1]`
 - `table[4] = table[3] + table[2]`
 - ...

Fibonacci: Bottom Up

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

$O(n)$

Step 1

0	1				
---	---	--	--	--	--

Step 2

0	1	1							
---	---	---	--	--	--	--	--	--	--

Step 3

0	1	1	2						
---	---	---	---	--	--	--	--	--	--

Step 4

0	1	1	2	3	5	8	13		
0	1	2	3	4	5	6	7		

สามารถเขียน recursive ได้

Optimized version of Bottom-Up Fibo

- From bottom up approach, we know that we only need two prior Fibonacci numbers ($F(n-1)$ and $F(n-2)$) to compute the current Fibonacci number ($F(n)$)
 - There is no need to lookup for $F(n-3)$, $F(n-4)$, ... if we know $F(n-1)$, and $F(n-2)$
 - Hence, no need to use entire table
 - Just remember two previous Fibonacci number

```
def fibo(n)
  if (n == 0 || n == 1)
    return n
  f2 = 0
  f1 = 1
  for i from 2 to n
    #calculate current
    f = f2 + f1
    #prepare f1 and f2 for next round
    f2 = f1
    f1 = f
  end
  return f
end
```

* จำไว้ 2 ตัว (ข้อ bottom up)

Binomial Coefficient

choose r things from n things

Example 2: Binomial Coefficient

- $C_{n,r}$ = how to choose r things from n things
 - We have a closed form solution
 - $C_{n,r} = n! / (r!(n-r)!)$
 - We also have recurrence relation of $C_{n,r}$
 - $C_{n,r} = C_{n-1,r} + C_{n-1,r-1}$
 $= 1$; $r = 0$
 $= 1$; $r = n$
- What is the subproblem?
- Do we have overlapping subproblem?

← ทัดในภา
งนี้เหมือนกัน

- Input:
 - Two integer r and n ($0 \leq r \leq n$)
- Output:
 - $C_{n,r}$

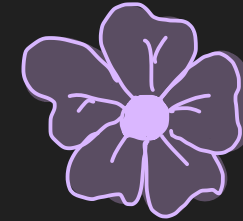
Binomial Coefficient

- Each subproblem is represented by 2 numbers, r and n
 - Hence, the table should be 2D

```
int bino_naive(int n,int r) {  
    if (r == n) return 1;  
    if (r == 0) return 1;  
  
    int result = bino_naive(n-1,r) + bino_naive(n-1,r-1);  
    return result;  
}
```

Binomial Coefficient: Top-Down (Memoization)

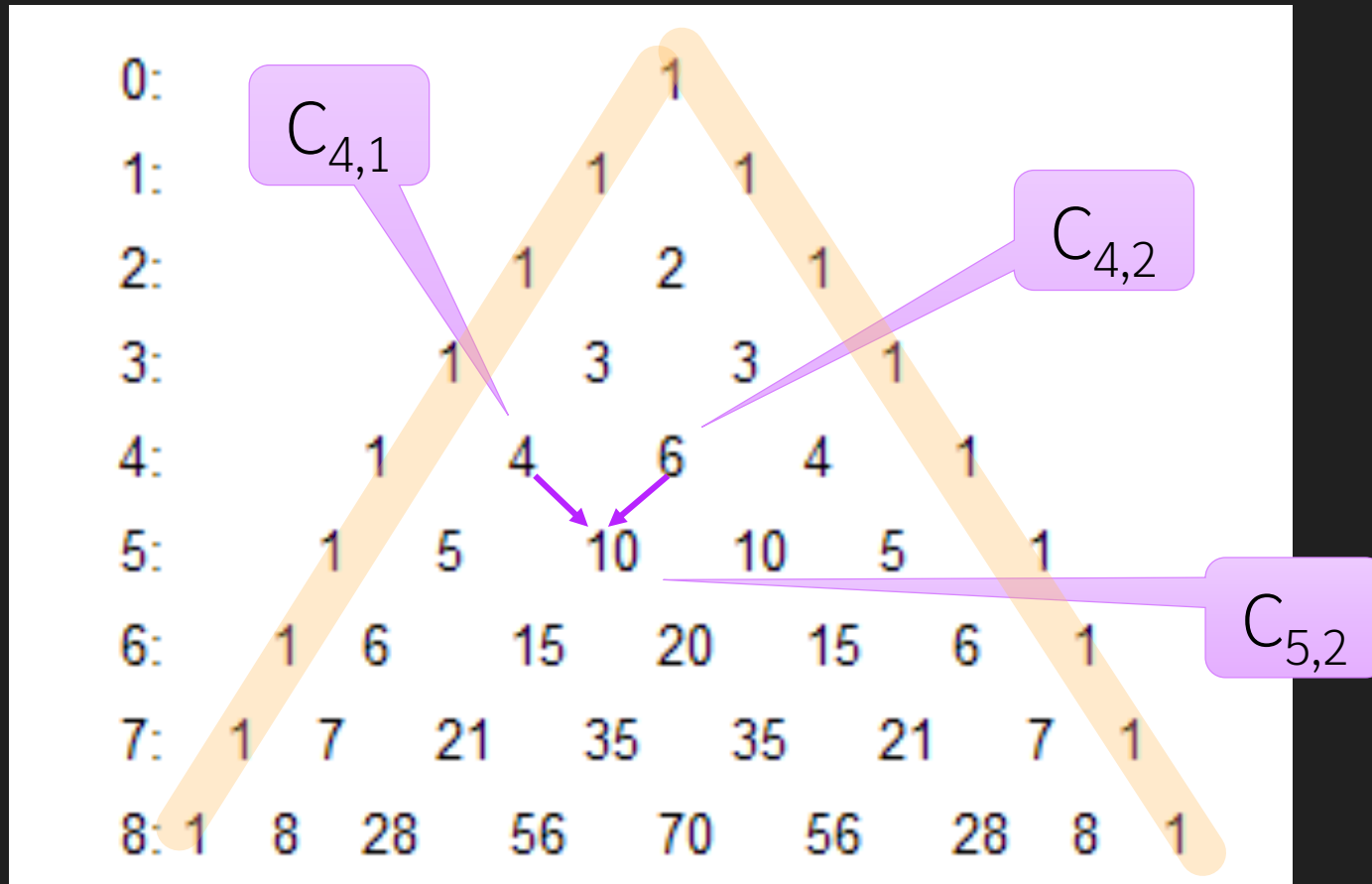
- `table[0..n][0..n]` is initialized by `-1`



```
int bino_memoize(int n,int r) {  
    if (r == n) return 1;  
    if (r == 0) return 1;  
  
    if (table[n][r] != -1)  
        return table[n][r];  
  
    int result = bino_memoize(n-1,r) + bino_memoize(n-1,r-1);  
    table[n][r] = result;  
  
    return result;  
}
```


Binomial Coefficient: Bottom Up

- Pascal triangle is a by-hand bottom-up DP of Binomial Coeff.



Binomial Coefficient: Bottom Up

ונטבל' → ונח + כדמכנ'ל'נ'נ'

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

$r = 3$ $\text{so } r = n$
 $= 1$

```

int bino_DP(int n, int r) {
    for (int i = 0; i <= n; i++) {
        table[i][0] = 1;
        table[i][i] = 1;
    }
    for (int i = 1; i <= n; i++) {
        for (int j = 1; j < min(i, r); j++) {
            table[i][j] = table[i-1][j] + table[i-1][j-1];
        }
    }
    return table[n][r];
}
    
```

initial condition

$\sigma(n^2)$

Question

- Is it possible to fill the table in different order?
- Does previous code solve subproblem that we does not need?
- If yes, how to avoid?

Yes

pbc

$$O(n^3) \rightarrow O(n^2)$$

$$\downarrow$$
$$O(n \lg n) \longrightarrow O(n)$$

Maximum Subarray Sum

Revisiting

The problem

- Given array $A[1..n]$ of numbers, may contain negative number
 - Find a non-empty subarray $A[p..q]$ such that the summation of the values in the subarray is maximum
- Input:
 - $A[1..n]$
- Output:
 - p and q , where $1 \leq p \leq q \leq n$ and summation of $A[p..q]$ is maximum
- Example:
 - $A = [1, 4, 2, 3]$ output: 1 and 4
 - $A = [-2, -1, -3, -5]$ output: 2 and 2
 - $A = [2, 3, -6, 4, -2, 3, -5, -4, 3]$ output: 4 and 6

$$T(n) = T(n-1) + \text{work}$$

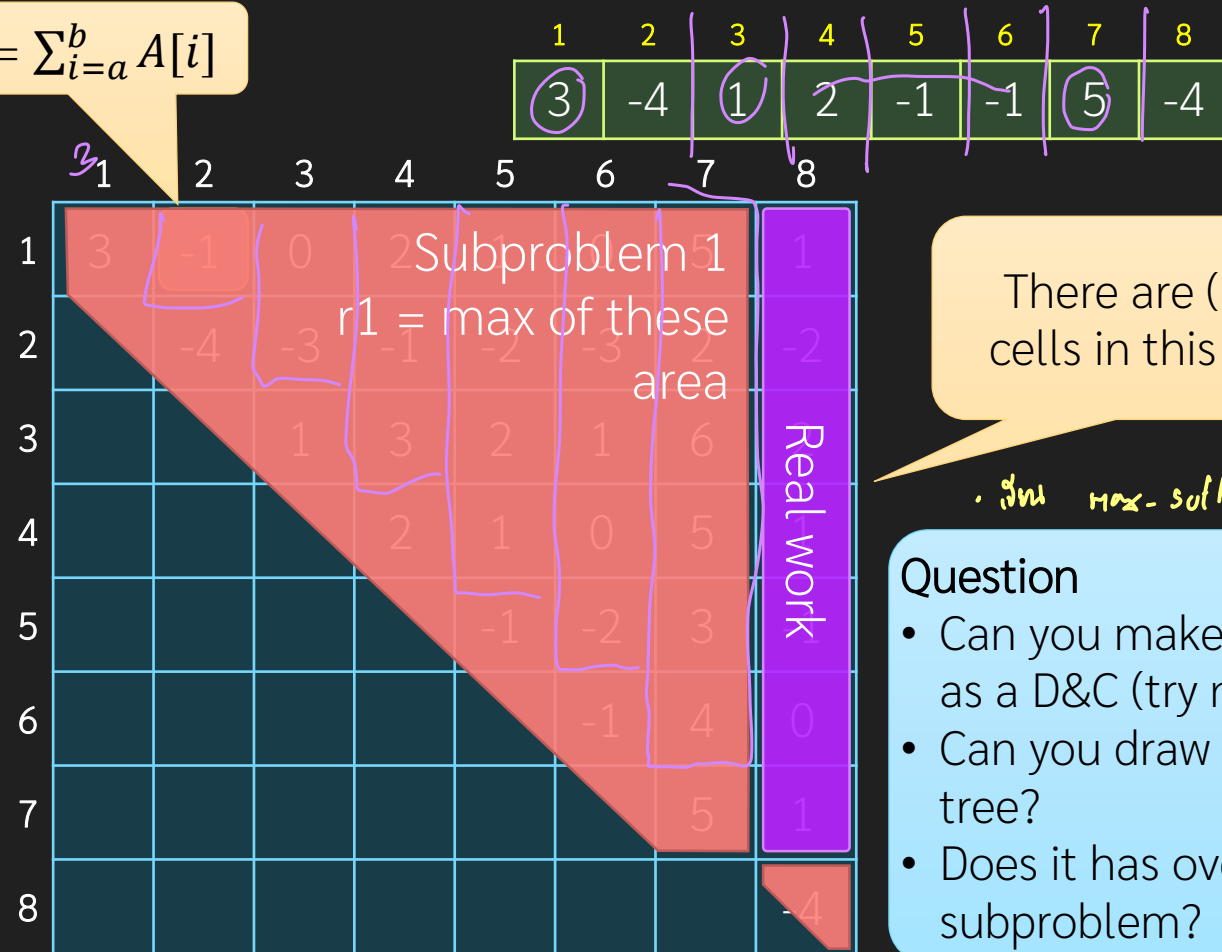
D&C by n-1

- Instead of dividing n into 2 of $n/2$ as previously done, we divide by $n-1$ and 1
 - The real work is solved by another D&C

```
def mss(A, stop)
    if (stop == 1)
        return A[1]
    r1 = mss(A, stop-1)
    r2 = A[stop]
    r3 = max_suffix(A, stop-1) + A[stop]
    return max(r1, r2, r3)
end
```

$$\text{max_suffix}(A, m) = \max_{1 \leq k \leq m} \sum_{i=k}^m A[i]$$

$$B[a][b] = \sum_{i=a}^b A[i]$$



There are $(n-1)$ cells in this area

• Can max-suffix D&C?

Question

- Can you make max_suffix as a D&C (try $n \rightarrow n-1$)
- Can you draw a recursion tree?
- Does it has overlapping subproblem?

```
def max_suffix(A, stop)
    if stop == 1
        return A[1]
    return max(A[stop],
               A[stop] + max_suffix(A, stop-1))
end
```

2	3	4	5	6

count: 0

max - sum(A, m, idx, count, max)

if (idx == m) return A[m]

count += 1

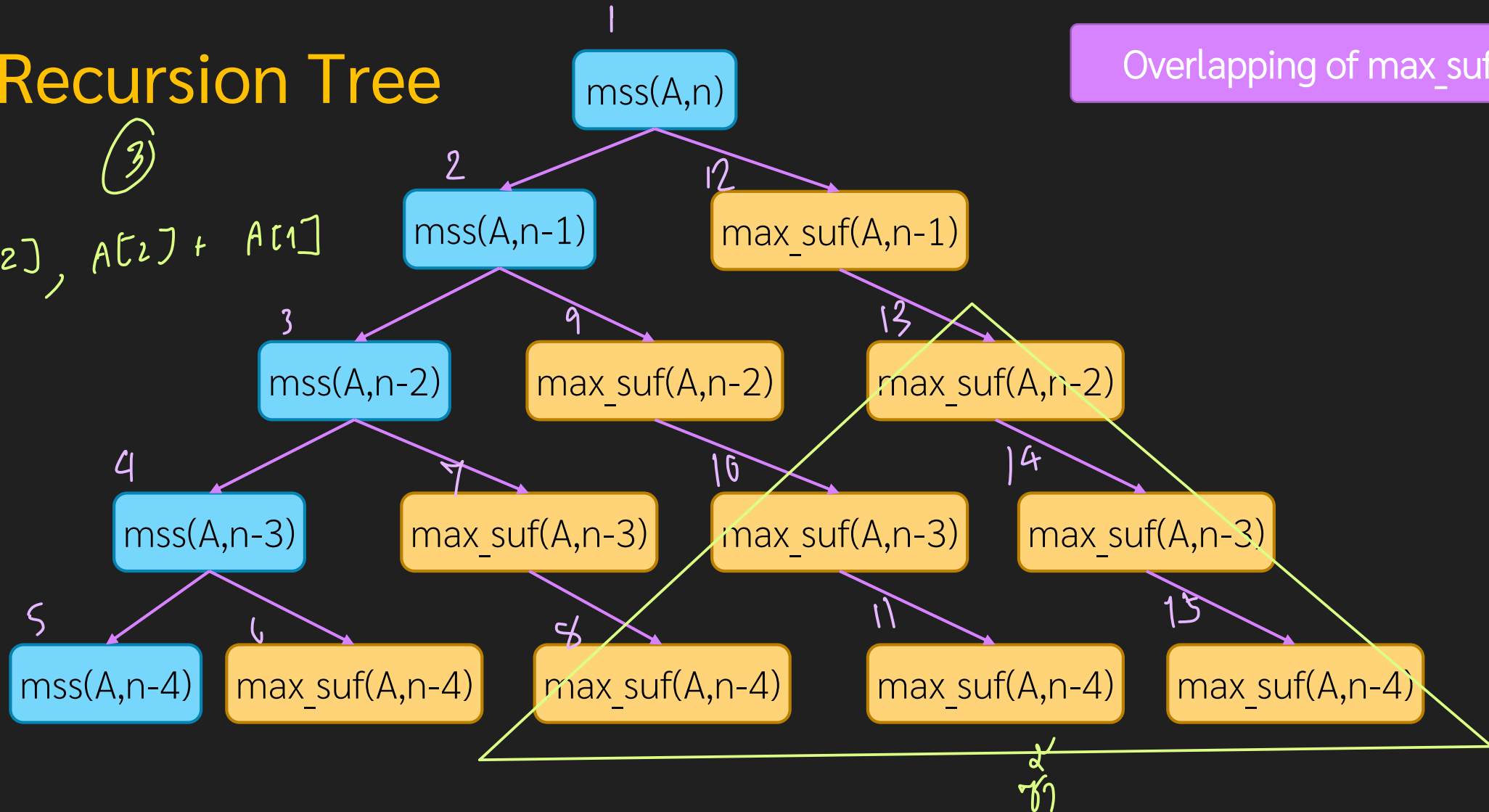
A[idx]

Recursion Tree

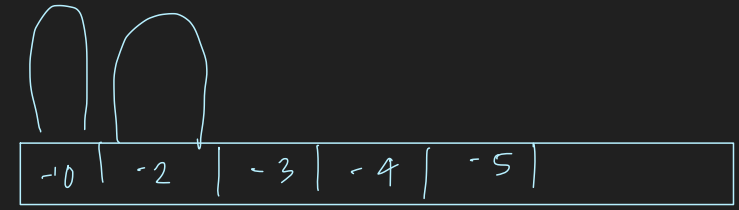
Overlapping of max_suf

③

$$\max(A[2], A[2] + A[1])$$



MSS with Dynamic Programming



```
def mss(A, stop)
  if (stop == 1)
    return A[1]
  r1 = mss(A, stop-1)
  r2 = A[stop]
  r3 = max_suffix(A, stop-1) + A[stop]
  return max(r1, r2, r3)
end
```

```
def max_suffix(A[1..n], stop, table[1..n], done[1..n])
  if stop == 1
    return A[1]
  if (done[stop])
    return table[stop]

  table[stop] =
    max(A[stop],
        A[stop] + max_suffix(A, stop-1))
  done[stop] = true
  return table[stop]
end
```

- Memoization (top-down) approach
- Since the value of `max_suffix` can be negative, we need another table to determine whether this subproblem is already solved
 - `done[1..n]` is initialized as `false`

Kadane's Algorithm

```
def kadane(A[1..n])  
    suf = A[1]  
    mss = A[1]  
    for i from 2 to n  
        suf = max(A[i], suf+A[i])  
        mss = max(mss, suf)  
    return mss  
end
```



- Calculate both **mss** and **suf** on the fly
- Original problem was proposed by **Ulf Grenander** in 1977
 - Originally 2D problem, convert to 1D to gain insight
- **$O(n \log n)$** D&C proposed by **Michael Shamos**
- **Joseph Born Kadane** heard the problem in a seminar and propose **$O(n)$**

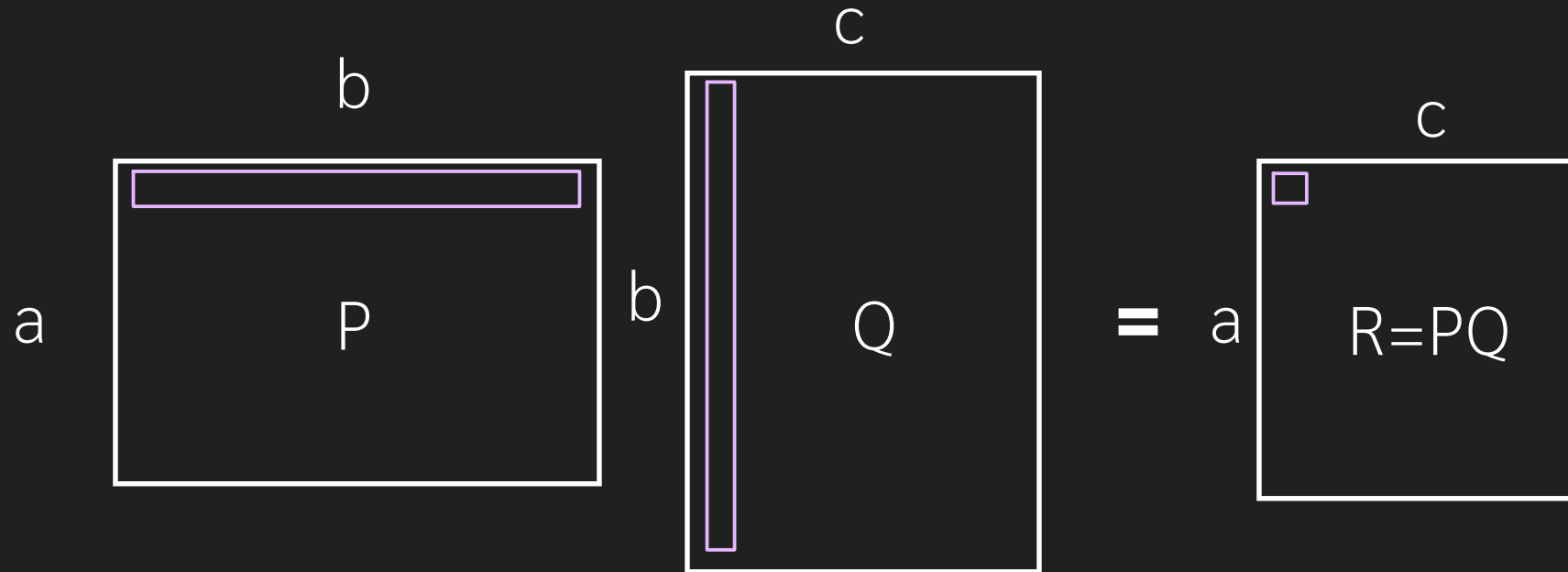
MCM



Matrix Chain Multiplication

Non-trivial bottom-up

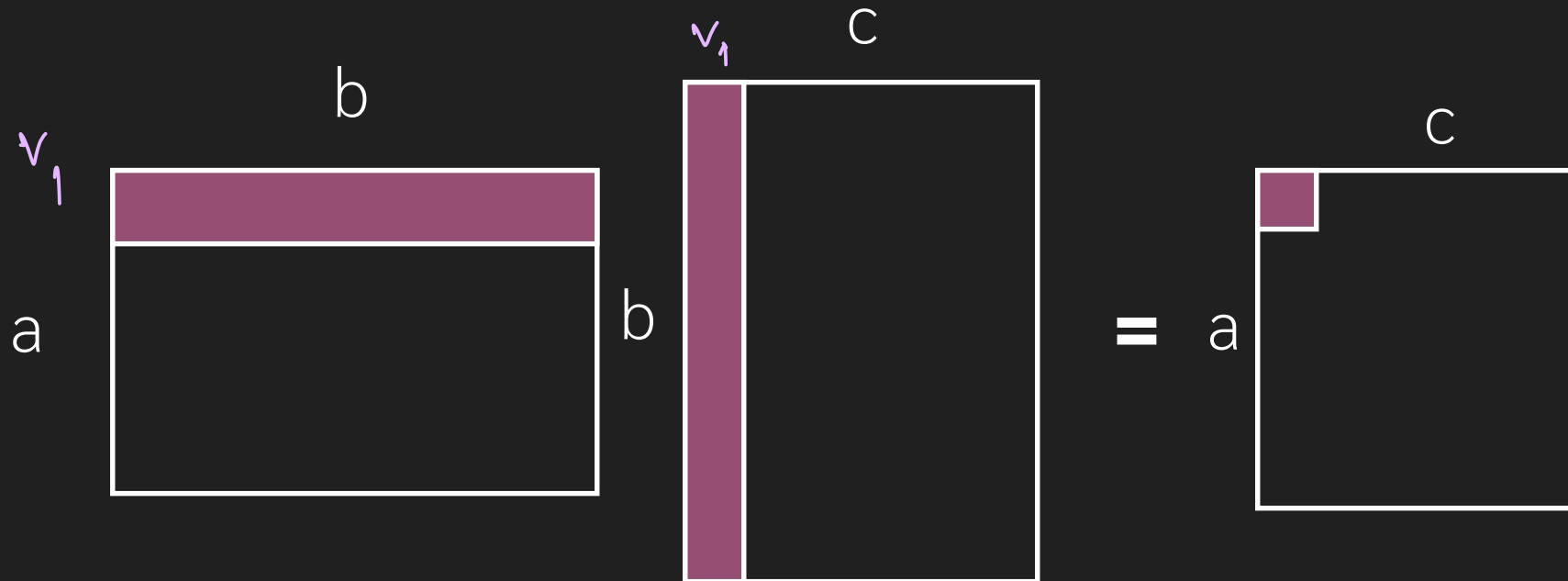
Matrix Multiplication



P = matrix with a rows and b columns

Q = matrix with b rows and c columns

Multiplying the Matrix



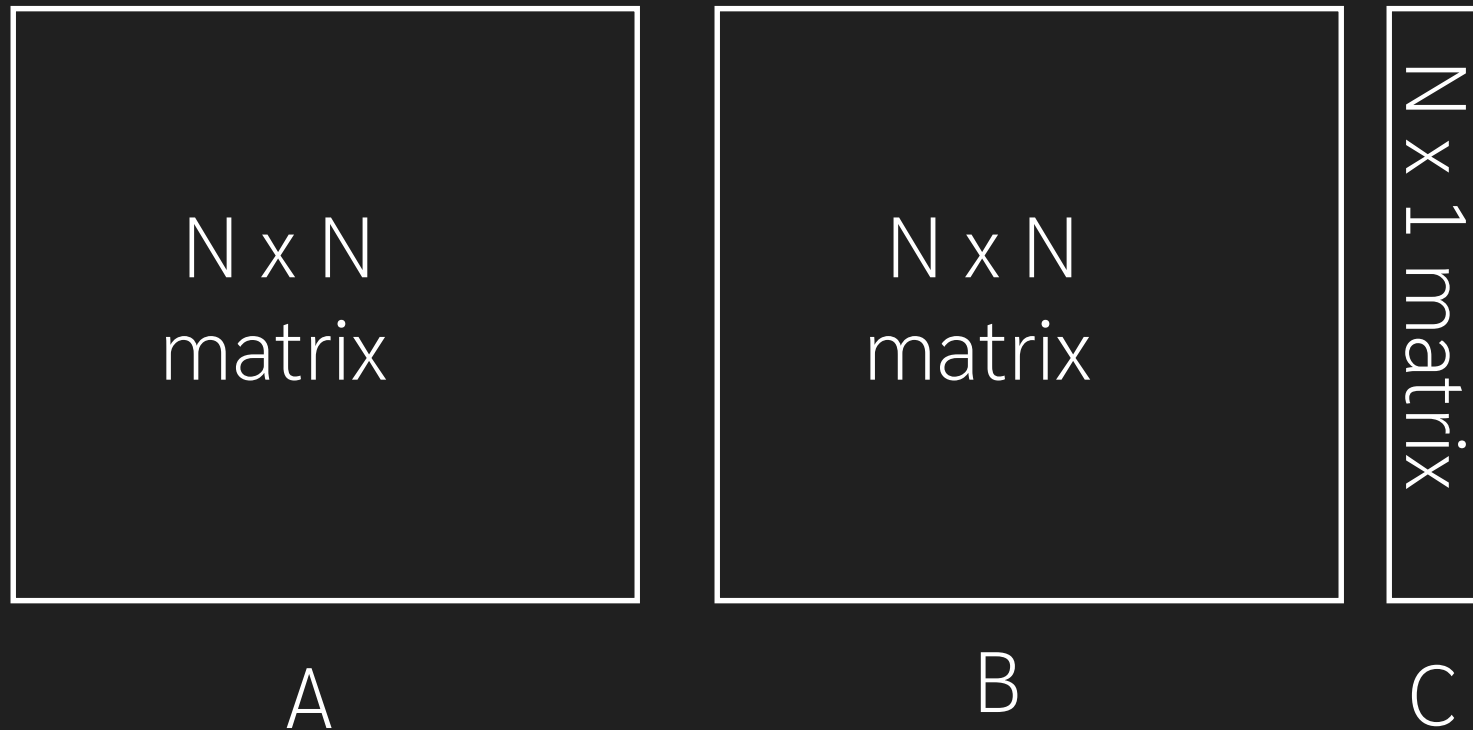
Time used = $\Theta(abc)$

Naïve Method

```
for (i = 1; i <= a; i++) {  
    for (j = 1; j <= c; j++) {  
        sum = 0;  
        for (k = 1; k <= b; k++) {  
            sum += P[i][k] * Q[k][j];  
        }  
        R[i][j] = sum;  
    }  
}
```

$O(abc)$

Matrix Chain Multiplication

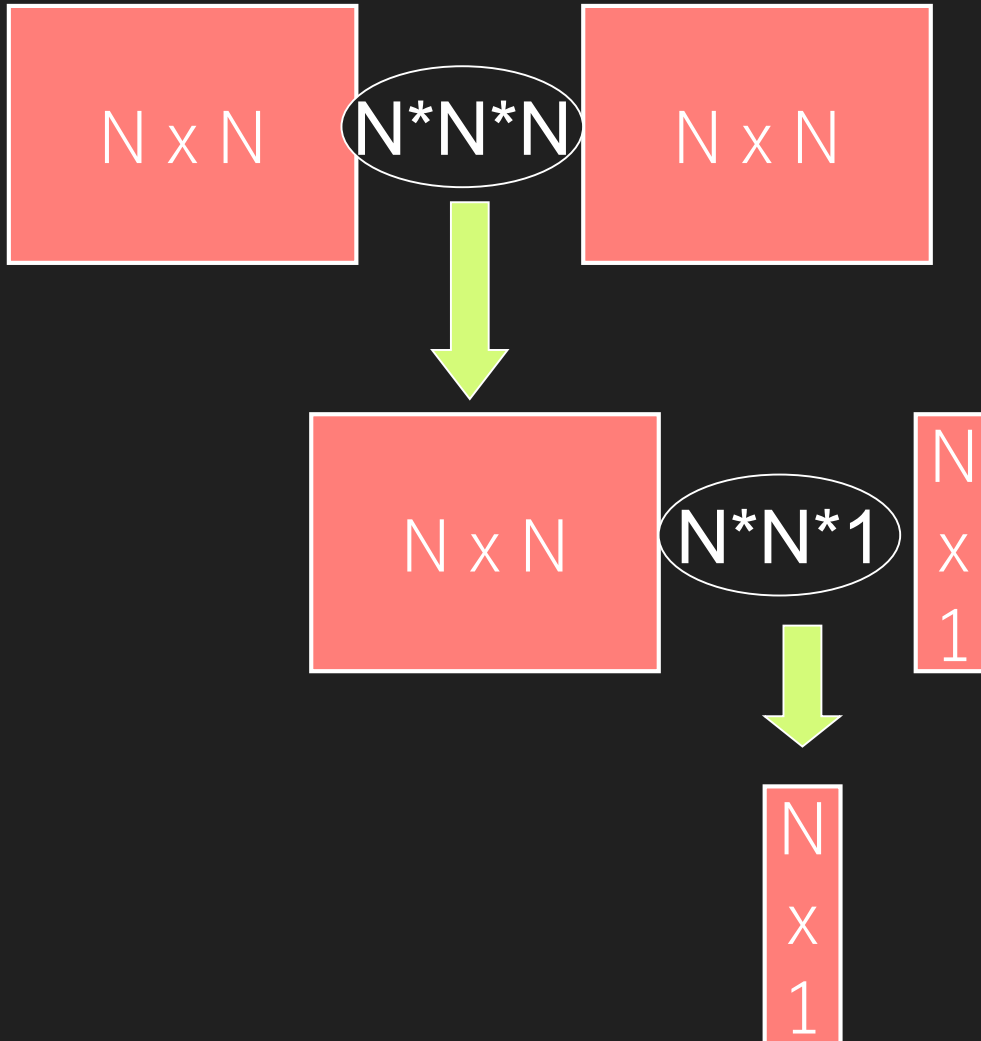


How to compute ABC ?

Matrix Multiplication

- $ABC = (AB)C = A(BC)$ ๒ วิธีคำนวณเหมือนกัน
- $(AB)C$ differs from $A(BC)$?
 - Same result, different efficiency
- What is the cost of $(AB)C$?
- What is the cost of $A(BC)$?

$$\overset{n}{\underbrace{(AB)}_n} \overset{n}{\underbrace{C}_{n^1}}$$

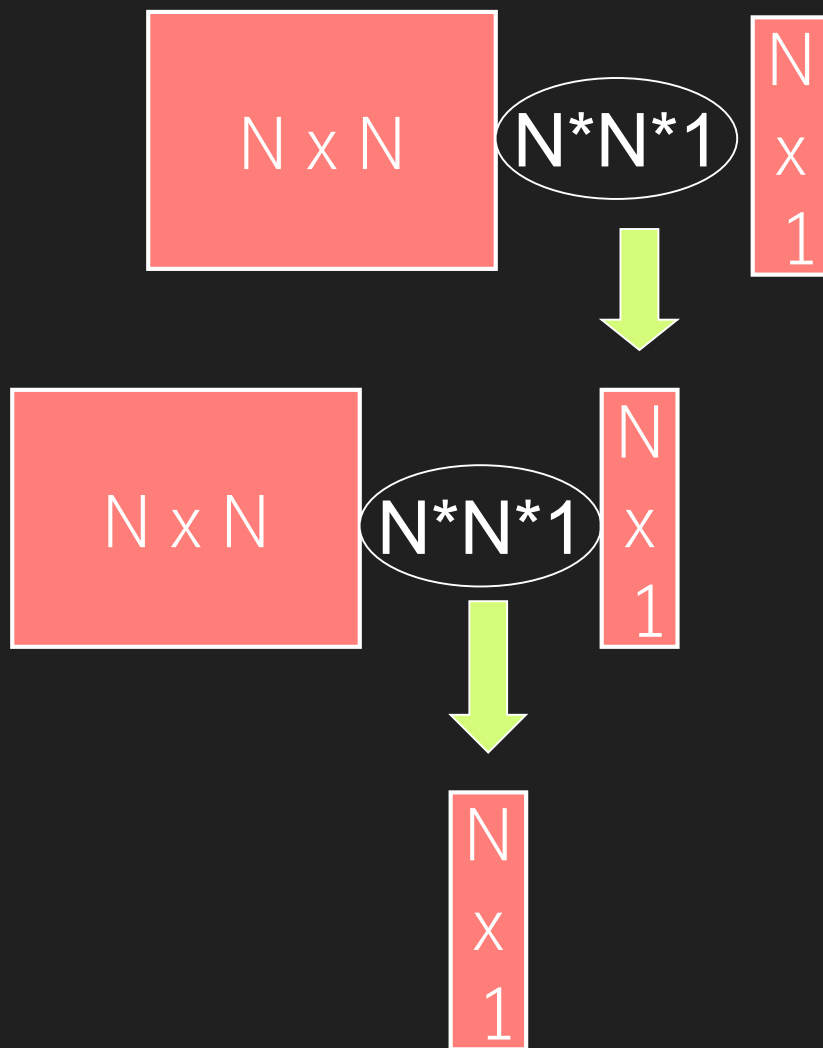


- Total = $N^3 + N^2$

A(BC)

- Total = $2N^2$

কাজের মোটঃ



Handwritten notes in green ink: two horizontal ovals, a vertical oval, and two small circles.

The Problem

- Input:
 - $a_1, a_2, a_3, \dots, a_n$
- Output:
 - The order of multiplication
 - How to parenthesize the chain
 - How many multiplication is needed
- Example Instance:
 - Input: 10 10 10 1

$$\begin{array}{cccc}
 & & 10 & 10 & 10 & 1 \\
 10 & 10 & & & & \\
 & & (10)(10)(1) \\
 & & (10)(10)
 \end{array}$$

These represents the size of the $n-1$ matrices $B_1 \dots B_{n-1}$

$$\begin{array}{ll}
 a_1 \times a_2 & B_1 \\
 a_2 \times a_3 & B_2 \\
 a_3 \times a_4 & B_3 \\
 \dots & \\
 a_{n-1} \times a_n & B_{n-1}
 \end{array}$$

Output: $(B_1(B_2B_3))$

200

More Example

INPUT

- $a_1 \quad a_2 \quad a_3 \quad a_4 \quad a_5 \quad a_6$
- $10 \times 5 \times 1 \times 5 \times 10 \times 2$
 $B_1 \quad B_2 \quad B_3 \quad B_4 \quad B_5$

Possible Output

$$((B_1 B_2)(B_3 B_4))B_5$$

$$(B_1 B_2)((B_3 B_4)B_5)$$

$$(B_1((B_2 B_3)B_4))B_5$$

And much more...

Consider the Output

What do

มีจุดที่เหมือนกัน

$$(B_1 B_2) (B_3 B_4) B_5$$

$a \quad b \quad c$

$$(B_1 B_2) (B_3 (B_4 B_5))$$

$a \quad b \quad c$

have in
common?

What do

$$((B_1 B_2) (B_3 B_4)) B_5$$

$a \quad b \quad c$

$$(((B_1 B_2) B_3) B_4) B_5$$

$a \quad b \quad c$

have in
common?

① — $a \cdot b \cdot c + \dots$

② — \dots

B_1, B_2
 $B_3 \dots B_5$

③ $a \cdot b \cdot c + \dots$

B_1, B_2
 $B_3 \dots B_5$

Solving $B_1 B_2 B_3 B_4 \dots B_{n-1}$

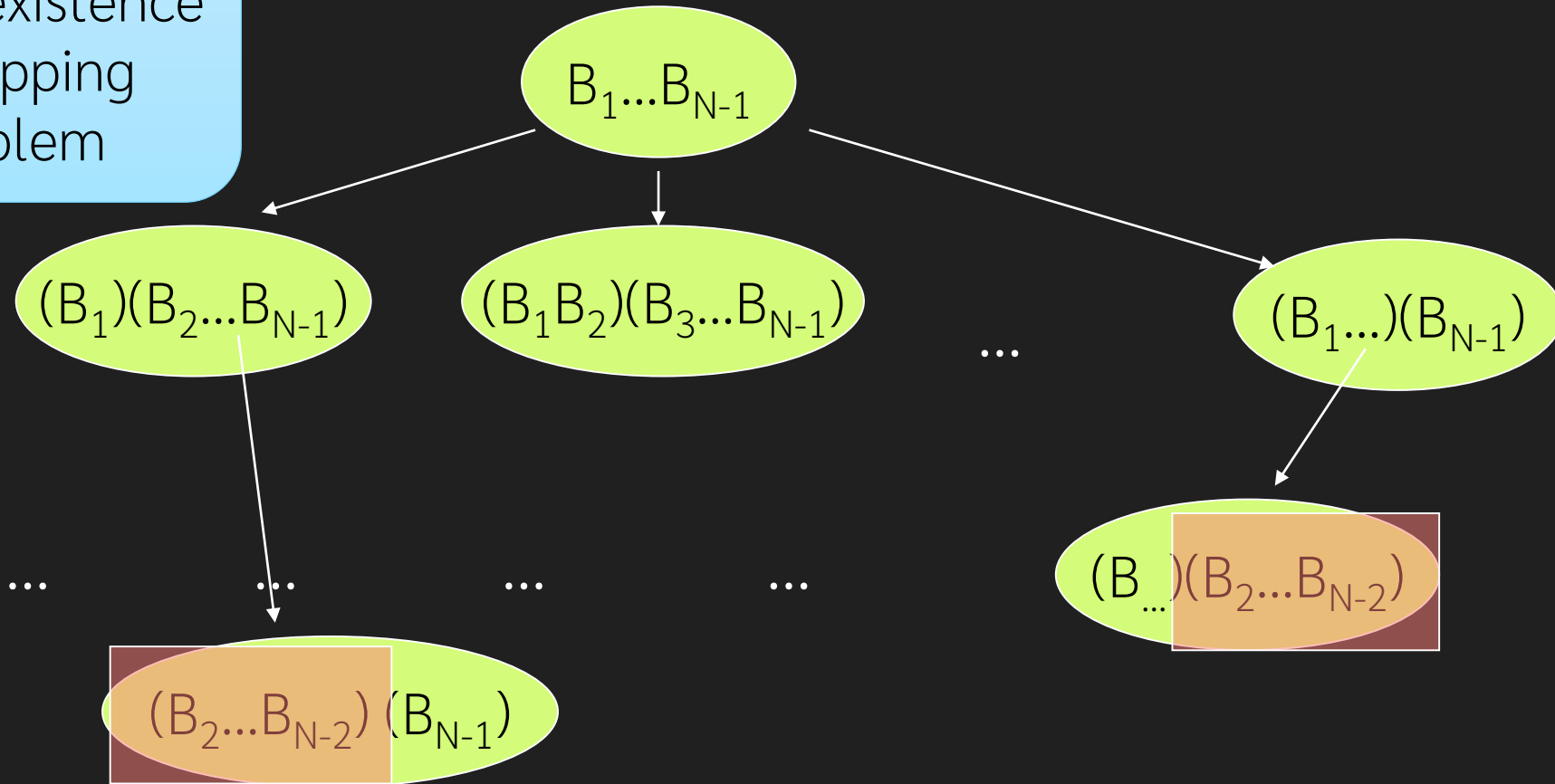
Min cost of

(1) B_1 $(B_2 B_3 \dots B_{n-1})$
 (2) $(B_1 B_2)$ $(B_3 B_4 \dots B_{n-1})$
 (3) $(B_1 B_2 B_3)$ $(B_4 B_5 \dots B_{n-1})$
 ...
 (n-2) $(B_1 B_2 B_3 B_4 \dots)$ B_{n-1}

- Each options ((1)..(n-2)) has 1 or 2 subproblems
- Sub problem is described by indices of left and right matrix
 - Needs 2 integers to describe a subproblem
- No overlapping subproblem (yet)

Overlapping Subproblem

Have to dig deeper
to identify existence
of overlapping
subproblem



Deriving the Recurrence Relation for D&C

- $mcm(l, r)$ $| \leq r$
 - The least cost to multiply $B_l \dots B_r$
หรือที่น้อยที่สุดที่ใช่ ให้ได้ผลในผลลัพธ์ที่ถูกต้อง
- The solution is $mcm(1, n-1)$
- Initial Case, when $(r - l) \leq 1$ (one or two matrices)
 - $mcm(x, x) = 0$
 - $mcm(x, x+1) = a[x] * a[x+1] * a[x+2]$

The Recurrence Relation

- Recursion Case

$mcm(l,r) = \min \text{ of}$

min cost of

B_l

$| (B_{l+1} \dots B_r)$

min cost of

$mcm(l, l+1)$

$| (B_{l+2} \dots B_r)$

min cost of

$mcm(l, l+2)$

$| (B_{l+3} \dots B_r)$

...

min cost of

$(B_l \dots B_{r-1})$

$| B_r$

Subproblems

Final multiplication

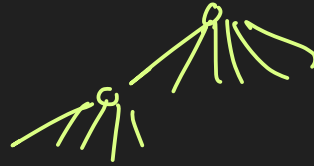
$+ a_l \cdot a_{l+1} \cdot a_{r+1}$

$+ a_l \cdot a_{l+2} \cdot a_{r+1}$

$+ a_l \cdot a_{l+3} \cdot a_{r+1}$

$+ a_l \cdot a_r \cdot a_{r+1}$

Divide & Conquer



```
int mcm(int l,int r) {  
    if (l < r) {  
        minCost = MAX_INT;  
        for (int i = l;i < r;i++) {  
            my_cost = mcm(l,i) + mcm(i+1,r) + (a[l] * a[i+1] * a[r+1]);  
            minCost = min(my_cost,minCost);  
        }  
        return minCost;  
    } else {  
        return 0;  
    }  
}
```

အသုံးပြုမှု top down ၊

Using bottom-up DP

- Design the table
 - $M[i][j]$ = the best solution (min cost) for multiplying $B_i \dots B_j$
 - $M[i][j]$ stores $mcm(i,j)$
 - The solution is at $M[1][n-1]$
- Trivial Case
 - What is $M[x][x]$?
 - No multiplication, $M[x][x] = 0$
- Simple case
 - What is $M[x][x+1]$?
 - $B_x B_{x+1}$
 - Only one solution = $a_x * a_{x+1} * a_{x+2}$

What is $M[i,j]$?

- General case

- What is $M[x][x+k]$?

- $B_x B_{x+1} B_{x+2} \dots B_{x+k}$

min of

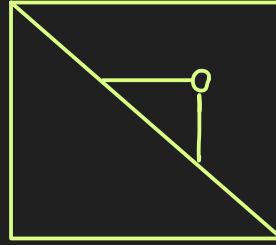
$$\begin{aligned}
 & \left[\begin{array}{l}
 \overset{1}{M[x][x]} + \overset{2}{M[x+1][x+k]} \quad + a_x \cdot a_{x+1} \cdot a_{x+k+1} \\
 \overset{1}{M[x][x+1]} + \overset{3}{M[x+2][x+k]} \quad + a_x \cdot a_{x+2} \cdot a_{x+k+1} \\
 \overset{1}{M[x][x+2]} + \overset{4}{M[x+3][x+k]} \quad + a_x \cdot a_{x+3} \cdot a_{x+k+1} \\
 \dots \\
 \overset{1}{M[x][x+k-1]} + \overset{5}{M[x+k][x+k]} \quad + a_x \cdot a_{x+k} \cdot a_{x+k+1}
 \end{array} \right.
 \end{aligned}$$

Filling the Table

The diagram shows a 6x6 table with a light blue header row and a light blue first column. The remaining cells are white. A yellow box labeled $M[1,1]$ has an arrow pointing to the cell at row 1, column 1. Another yellow box labeled $M[1,6]$ (our solution) has an arrow pointing to the cell at row 1, column 6. A green arrow also points from the $M[1,6]$ box to the cell at row 6, column 6.

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

Filling the Table



Trivial
case

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

Filling the Table

Arbitrary
case

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

The diagram illustrates the process of filling a 6x6 table. The main diagonal cells (1,1), (2,2), (3,3), (4,4), (5,5), and (6,6) are highlighted in light purple and contain the value 0. Red arrows point from these diagonal cells to the cells immediately to their right: from (1,1) to (1,2), from (2,2) to (2,3), from (3,3) to (3,4), from (4,4) to (4,5), from (5,5) to (5,6), and from (6,6) to (6,7). A red square highlights the cell (1,5), which is the source of an arrow pointing to (1,6). This indicates that the value in cell (1,6) depends on the values in cells (1,5) and (1,4).

Filling the Table




Arbitrary
case

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

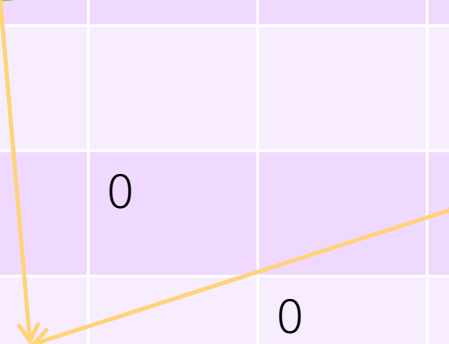
The diagram illustrates the process of filling a 6x6 table. The table has a light purple header and a light blue body. The header row and column are labeled 1 to 6. The body cells are initially empty. The cell at (1,1) contains the value 0. The cell at (2,2) contains the value 0. The cell at (3,3) contains the value 0. The cell at (4,4) contains the value 0. The cell at (5,5) contains the value 0. The cell at (6,6) contains the value 0. A red rectangle is placed in the cell at (1,5). A yellow oval is placed in the cell at (2,6). A green box containing the text "Plus $a_1 \cdot a_2 \cdot a_6$ " is placed in the cell at (4,4). Two orange arrows point from the yellow oval in (2,6) and the green box in (4,4) to the cell at (4,4).

Filling the Table

Arbitrary
case

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

Plus $a_1 \cdot a_3 \cdot a_6$



Filling the Table

Arbitrary
case

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

The diagram illustrates the process of filling a 6x6 table. A green box highlights the expression $\text{Plus } a_1 \cdot a_4 \cdot a_6$ in the cell at row 4, column 3. Two orange ovals are placed in the cells at row 1, column 3 and row 5, column 5. An orange arrow points from the oval at row 1, column 3 to the green box. Another orange arrow points from the oval at row 5, column 5 to the green box. A red rectangle is located in the cell at row 1, column 5. The table has a light purple background with white text for the header and the value 0. The cells containing the ovals and the green box have a light green background.

Filling the Table

Arbitrary
case

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

The diagram illustrates the process of filling a 6x6 table. A green box containing the text "Plus $a_1 \cdot a_5 \cdot a_6$ " is positioned over the cell at row 4, column 3. Two orange arrows originate from this box: one points to the cell at row 1, column 4, and the other points to the cell at row 5, column 5. Both of these target cells contain a "0" and are circled in orange. Additionally, a red rectangle is present in the cell at row 1, column 5.

Filling the Table

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

Filling the Table

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

The diagram illustrates the process of filling a 6x6 table. The table has a header row and header column, both labeled 1 to 6. The diagonal elements (top-left to bottom-right) are all 0. The cells (1,6), (2,5), (3,4), (4,3), (5,2), and (6,1) are filled with values 5, 4, 3, 2, 1, and 0 respectively. Green arrows point from each of these cells to the cell immediately below and to the right of it, showing the sequence of calculations.

Example

- $a_1 \quad a_2 \quad a_3 \quad a_4 \quad a_5 \quad a_6$

- $10 \times 5 \times 1 \times 5 \times 10 \times 2$

$B_1 \quad B_2 \quad B_3 \quad B_4 \quad B_5$



Example

a_1 a_2 a_3 a_4 a_5 a_6
10 x 5 x 1 x 5 x 10 x 2

	1	2	3	4	5
1	0				
2		0			
3			0		
4				0	
5					0

Example



$$a_2 \times a_3 \times a_4$$

$$a_1 \quad a_2 \quad a_3 \quad a_4 \quad a_5 \quad a_6$$

$$10 \times 5 \times 1 \times 5 \times 10 \times 2$$

	1 $a_1 a_2$	2 $a_2 a_3$	3 $a_3 a_4$	4 $a_4 a_5$	5 $a_5 a_6$
1	0	50			
2		0			
3			0		
4				0	
5					0

Example

a_1 a_2 a_3 a_4 a_5 a_6
10 x 5 x 1 x 5 x 10 x 2

	1	2	3	4	5
1	0	50			
2		0	25		
3			0		
4				0	
5					0

Example

a_1 a_2 a_3 a_4 a_5 a_6
10 x 5 x 1 x 5 x 10 x 2

	1	2	3	4	5
1	0	50			
2		0	25		
3			0	50	
4				0	
5					0

Example

a_1 a_2 a_3 a_4 a_5 a_6
10 x 5 x 1 x 5 x 10 x 2

	1	2	3	4	5
1	0	50			
2		0	25		
3			0	50	
4				0	100
5					0

Example

a_1 a_2 a_3 a_4 a_5 a_6
10 x 5 x 1 x 5 x 10 x 2

	1	2	3	4	5
1	0	50			
2		0	25		
3			0	50	
4				0	100
5					0

Example

$$\text{Option 1} = 0 + 25 + 10 \times 5 \times 5 = 275$$

$$\text{Option 2} = 50 + 0 + 10 \times 1 \times 5 = 100$$

↩

$$a_1 \quad a_2 \quad a_3 \quad a_4 \quad a_5 \quad a_6$$

$$10 \times 5 \times 1 \times 5 \times 10 \times 2$$

	1	2	3	4	5
1	0	50			
2		0	25		
3			0	50	
4				0	100
5					0

Example

a_1 a_2 a_3 a_4 a_5 a_6
10 x 5 x 1 x 5 x 10 x 2

	1	2	3	4	5
1	0	50	100 (2) <i>mark 10</i>		
2		0	25		
3			0	50	
4				0	100
5					0

(2) means that
the minimal
solution is by
dividing at B_2

Example

$$\text{Option 1} = 0 + 50 + 5 \times 1 \times 10 = 100$$

$$\text{Option 2} = 25 + 0 + 5 \times 5 \times 10 = 275$$

$$a_1 \quad a_2 \quad a_3 \quad a_4 \quad a_5 \quad a_6$$
$$10 \times 5 \times 1 \times 5 \times 10 \times 2$$

	1	2	3	4	5
1	0	50	100 (2)		
2		0	25		
3			0	50	
4				0	100
5					0

Example

a_1 a_2 a_3 a_4 a_5 a_6
10 x 5 x 1 x 5 x 10 x 2

	1	2	3	4	5
1	0	50	100 (2)		
2		0	25	100 (2)	
3			0	50	70 (4)
4				0	100
5					0

Example

$$\text{Option 1} = 0 + 100 + 10 \times 5 \times 10 = 600$$

$$\text{Option 2} = 50 + 50 + 10 \times 1 \times 10 = 200$$

$$\text{Option 2} = 100 + 0 + 10 \times 5 \times 10 = 600$$

$$a_1 \quad a_2 \quad a_3 \quad a_4 \quad a_5 \quad a_6$$
$$10 \times 5 \times 1 \times 5 \times 10 \times 2$$

	1	2	3	4	5
1	0	50	100 (2)		
2		0	25	100 (2)	
3			0	50	70 (4)
4				0	100
5					0

Example

a_1 a_2 a_3 a_4 a_5 a_6
10 x 5 x 1 x 5 x 10 x 2

	1	2	3	4	5
1	0	50	100 (2)	200 (2)	
2		0	25	100 (2)	
3			0	50	70 (4)
4				0	100
5					0

Example

a_1 a_2 a_3 a_4 a_5 a_6
10 x 5 x 1 x 5 x 10 x 2

	1	2	3	4	5
1	0	50	100 (2)	200 (2)	
2		0	25	100 (2)	80 (2)
3			0	50	70 (4)
4				0	100
5					0

Example

① բովանդակություն

1122 bottom-up

② อบทินัยมโนปารมิตา
อภินิหารมโน

$$(B_1, B_2) \sim$$
$$a_1 \quad a_2 \quad a_3 \quad a_4 \quad a_5 \quad a_6$$
$$10 \times 5 \times 1 \times 5 \times 10 \times 2$$

	1	2	3	4	5
1	0	50	100 (2)	200 (2)	140 (2)
2		0	25	100 (2)	80 (2)
3			0	50	70 (4)
4				0	100
5					0

Analysis

- There is $O(n^2)$ cell to be filled
 - Each cell has $O(n)$ options
- This totals to $O(n^3)$

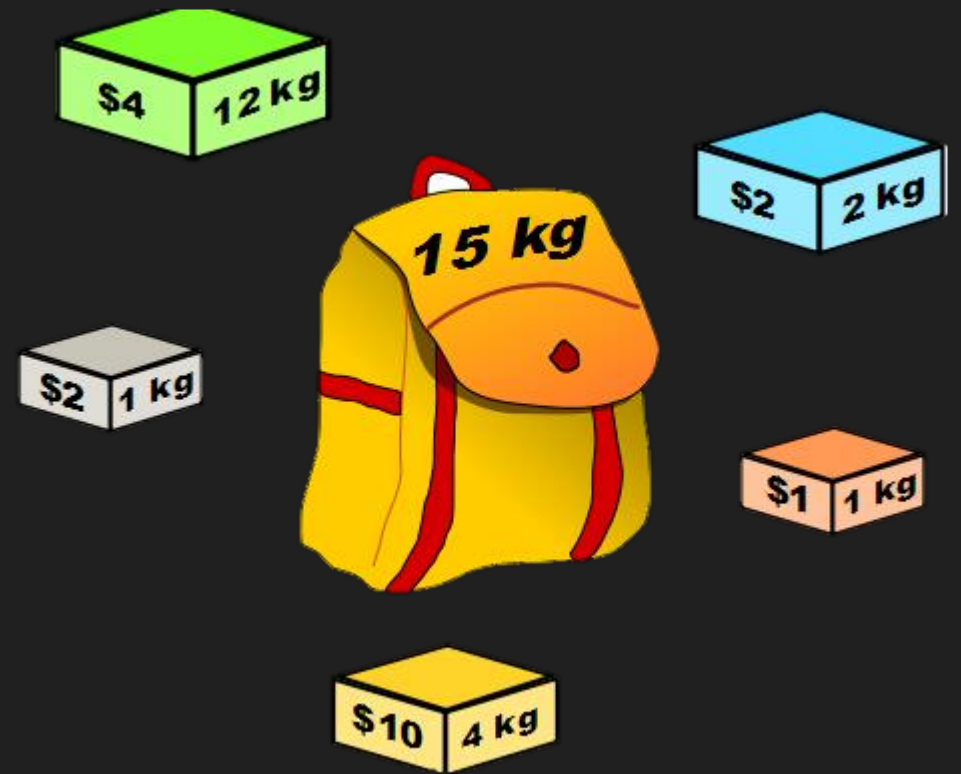
Can you write a code for
Bottom-up DP of Matrix
Chain Multiplication
Problem?

Also can your code build
the actual solution (the
parenthesis of B_i , not
just the minimum cost)

0-1 Knapsack Problem

Knapsack Problem

- Given a sack, able to hold W kg
- Given a list of objects
 - Each has a **value** and a **weight**
- Try to pack the object in the sack so that the total value is maximized



Variation

- Rational Knapsack เลือกชิ้น เป็นชิ้นก็ได้ น้ำหนัก/ราคา
↓
เพื่ออัตราส่วนนี้ เท่ากัน
 - Object is like a gold bar, we can cut it into pieces, each has the same value/weight ratio
- 0-1 Knapsack
 - Object cannot be broken, we have to choose to take (1) or leave (0) the object
 - $W = 50$
 - Objects = (60, 10) (100, 20) (120, 30)
 - Best solution = second and third

The Problem

- Input:

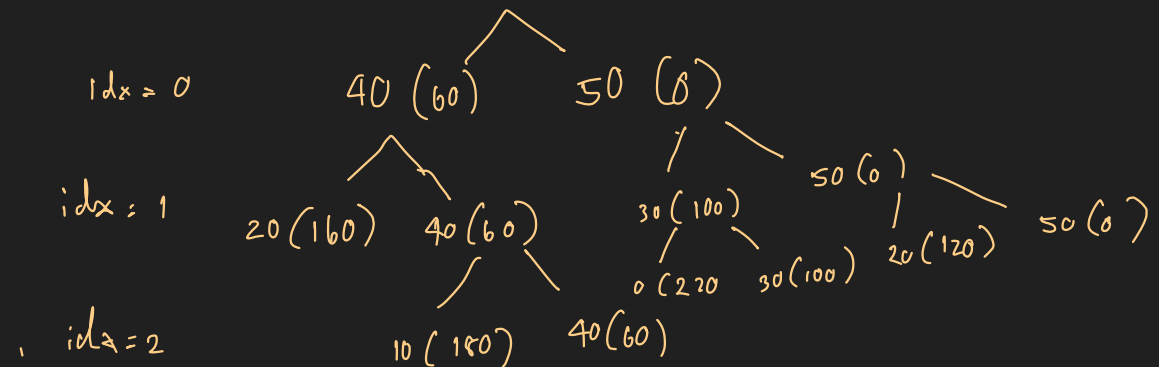
- A number W ^{max}, the capacity of the sack
- n values of weight and price
 - w_i = weight of the i^{th} items
 - p_i = price of the i^{th} item

- Output:

- A subset S of $\{1, 2, 3, \dots, n\}$ such that
 - $\sum_{i \in S} p_i$ is maximum
 - $\sum_{i \in S} w_i \leq W$

- Example Instance

- $W = 50$
- $P_i = 60, 100, 120$ ^{2 20}
- $w_i = 10, 20, 30$ ^{50 7 20 30}
- Best solution = second and third ✓



Naïve approach

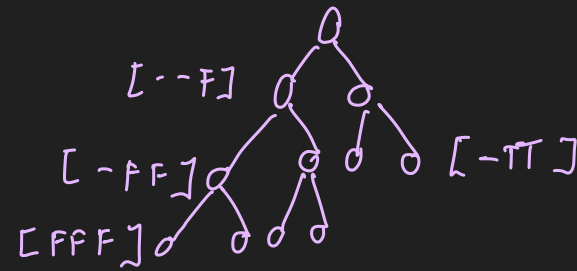
วิธีการแบบที่ ๑๑๑

```
def knapsack(W,w[1..n],p[1..n],idx,pick[1..n])
  if (idx == 0)
    sum_price = 0
    sum_weight = 0
    for i from 1 to n
      if pick[i]
        sum_price += p[i]
        sum_weight += w[i]
        if (sum_weight <= W && sum_price > max)
          max = sum_price

  pick[idx] = false
  knapsack(W,w,p,idx-1,pick)
  pick[idx] = true
  knapsack(W,w,p,idx-1,pick)
end
```

↪ true / false

↪ recursive แล้ว



- Try every possible combination of $\{1,2,3,\dots,n\}$
- Test whether a combination satisfies the weight constraint
 - If so, remember the best one
 - Start with `knapsack(W,w,p,n,[1..n])`
 - `max` is global var
 - $\theta(2^n \cdot n)$
↪ คำนวณค่า

Another Naïve approach

- Keep track of remaining weight, sum the total price along the way
- What is the benefit of this approach?

$knapsack(w, p, n, W)$

↑
เก็บ n.น. ที่เหลืออยู่

```
def knapsack(W,w[1..n],p[1..n],idx,remain)
  if (idx == 0)
    return 0
  if (remain >= w[idx])
    #r1 is that we don't pick item #idx
    r1 = knapsack(W,w,p,idx-1,remain)
    #r2 is that we pick item #idx
    r2 = knapsack(W,w,p,idx-1,remain - w[idx]) + p[idx]
    return max(r1,r2)
  else
    return knapsack(W,w,p,idx-1,remain)
end
```

↓
เลือกของชิ้นที่เท่าไร

↑
ไม่เลือก

↑
เลือก

→ ไม่เลือก w[idx]

The Recurrence Relation

- $K(a,b)$ = the best total price when and only item number (1) to number (a) is considered and the knapsack is of size \underline{b}

- $K(a,b) = 0$ when $\overset{\text{ถ้าไม่มีของในกระเป๋า, ถ้าขนาดกระเป๋าได้ 0}}{\underline{a = 0} \text{ or } \underline{b = 0}}$

- $K(a,b) = K(a-1,b)$ when $\underline{w_a} > b$ น.บ.เกิน ก็ทิ้งไปเลย

- $K(a,b) = \max(K(a-1, b - w_a) + p_a, K(a-1, b))$

- The solution is at $K(n,W)$

The Failed Attempt #1

- Let $K(a)$ be the **best total value** when we consider only item number **1** to number **a** and the weight limit is **W** *W ให้นับใน น.น.*
 - The answer is at $K(n)$
 - By definition, $K(n)$ and $K(n-1)$ and $K(n-2)$... all consider the **same weight limit**
- Let's say that the answer contains item number n
 - Also by definition, it means that $K(n) = K(n-1) + p_n$
 - However, $K(n-1)$ will consider the problem thinking that the weight limit is the same (not reduced by weight of item number n)
 - It is wrong to say that $K(a) = \max(K(a-1) + p_a, K(a-1))$ *เพราะมันไม่นับน้ำหนักของ n.น.ที่เพิ่มเข้ามา*
 - It is not possible to have a recurrence relation that does not consider W

The Failed Attempt #2

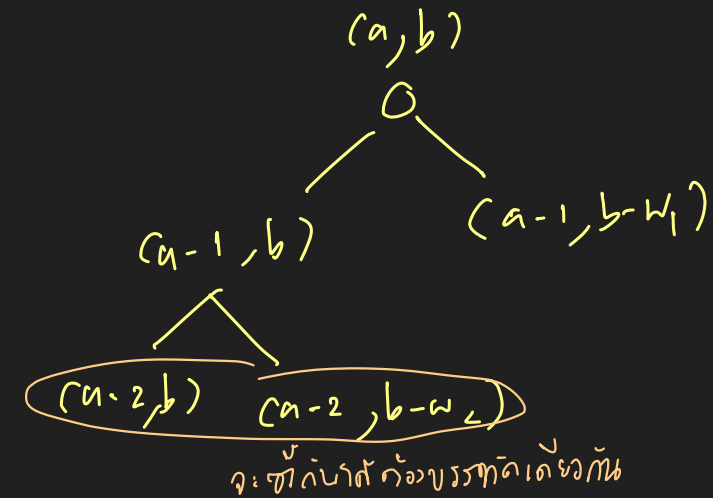
सान์, น.

- Let $K(b)$ be the best total value when the weight limit of the sack is b
 - The answer is at $K(W)$
- If the i^{th} item is in the best solution
 - $K(W) = K(W - w_i) + p_i$
- But, we don't really know that the i^{th} item is in the optimal solution
 - So, we try everything
 - $K(W) = \max_{1 \leq i \leq n} (K(W - w_i) + p_i)$
- Is this our algorithm?
 - Yes, if and only if we allow each item to be selected multiple times (that is not true for this problem)

Exercise: Top-Down approach

- Write a top down dynamic programming approach using this recurrence relation

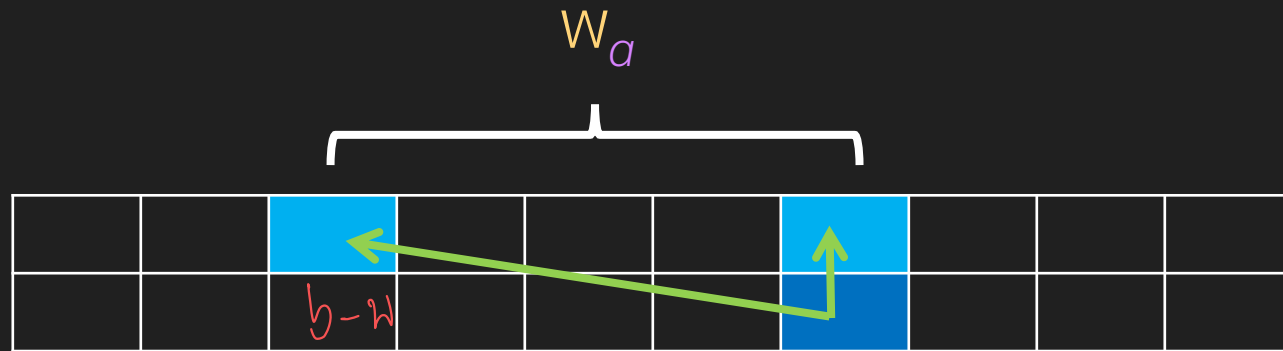
- $K(a,b) = 0$ when $a = 0$ or $b = 0$
- $K(a,b) = K(a-1,b)$ when $w_a > b$
- $K(a,b) = \max(K(a-1,b - w_a) + p_a, K(a-1,b))$



- Which data structure should we use to store result?
 - Should we use 2D array?
 - Should we use associative data structure such as `std::map` or `std::unordered_map`?

The Table for Bottom-Up

- Row = item id (a)
- Col = weight (b)



Normal case ($w_a \leq b$)

$K(a, b)$



$K(a, b)$

Too much weight ($w_a > b$)

Example

$p = \{4, 2, 2, 1, 10\}$

$w = \{12, 2, 1, 1, 4\}$ $W = 15$

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
1	0															
2	0															
3	0															
4	0															
5	0															

$K(a,b) = 0$

when $a = 0$ or $b = 0$

Example

$p = \{4, 2, 2, 1, 10\}$

$w = \{12, 2, 1, 1, 4\}$ $W = 15$

Fill row 1 ($p_1=4$ $w_1=12$)

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	0	0	0	0	0				
2	0															
3	0															
4	0															
5	0															

$$K(a,b) = K(a-1,b)$$

when $w_a > b$

Example

$$p = \{4, 2, 2, 1, 10\}$$

$$w = \{12, 2, 1, 1, 4\} \quad W = 15$$

Fill row 1 ($p_1=4$ $w_1=12$)

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	0	0	0	0	0	4	4	4	4
2	0															
3	0															
4	0															
5	0															

12 +
สี่ คือ แล้ว ว่า

$$K(a,b) = \max(K(a-1, b - w_a) + p_a, K(a-1, b))$$

Example

$$p = \{4, 2, 2, 1, 10\}$$

$$w = \{12, 2, 1, 1, 4\} \quad W = 15$$

Fill row 2 ($p_2=2$ $w_2=2$)

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	0	0	0	0	0	4	4	4	4
2	0	0														
3	0															
4	0															
5	0															

$$K(a,b) = K(a,b-1) \quad \text{when } w_b > a$$

Example

$p = \{4, 2, 2, 1, 10\}$

$w = \{12, 2, 1, 1, 4\}$ $W = 15$

Fill row 2 ($p_2=2$ $w_2=2$)

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	0	0	0	0	0	4	4	4	4
2	0	0														
3	0															
4	0															
5	0															

$$K(a,b) = K(a-1,b)$$

when $w_a > b$

Example

$p = \{4, 2, 2, 1, 10\}$

$w = \{12, 2, 1, 1, 4\}$ $W = 15$

Fill row 2 ($p_2=2$ $w_2=2$)

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	0	0	0	0	0	4	4	4	4
2	0	0	2	2	2	2	2	2	2	2	2	2	4	4	6	6
3	0															
4	0															
5	0															

$$K(a,b) = \max(K(a-1, b - w_a) + p_a, K(a-1, b))$$

Example

$p = \{4, 2, 2, 1, 10\}$

$w = \{12, 2, 1, 1, 4\}$ $W = 15$

Fill row 2 ($p_2=2$ $w_2=2$)

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	0	0	0	0	0	4	4	4	4
2	0	0	2	2	2	2	2	2	2	2	2	2	4	4	6	6
3	0															
4	0															
5	0															

$$K(a,b) = \max(K(a-1, b - w_a) + p_a, K(a-1, b))$$

Example

$p = \{4, 2, 2, 1, 10\}$

$w = \{12, 2, 1, 1, 4\}$ $W = 15$

Fill row 3 ($p_3=2$ $w_3=1$)

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	0	0	0	0	0	4	4	4	4
2	0	0	2	2	2	2	2	2	2	2	2	2	4	4	6	6
3	0	2	2	4	4	4	4	4	4	4	4	4	4	6	6	8
4	0															
5	0															

$$K(a,b) = \max(K(a-1,b - w_a) + p_a, \quad K(a-1,b))$$

Example

$$p = \{4, 2, 2, 1, 10\}$$
$$w = \{12, 2, 1, 1, 4\} \quad W = 15$$

Fill row 3 ($p_3=2$ $w_3=1$)

[illegible]

Example

$$p = \{4, 2, 2, 1, 10\}$$
$$w = \{12, 2, 1, 1, 4\} \quad W = 15$$

Fill row 4 ($p_4=1$ $w_4=1$)

[illegible]

Example

$$p = \{4, 2, 2, 1, 10\}$$

$$w = \{12, 2, 1, 1, 4\} \quad W = 15$$

Fill row 4 ($p_4=1$ $w_4=1$)

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	0	0	0	0	0	4	4	4	4
2	0	0	2	2	2	2	2	2	2	2	2	2	4	4	6	6
3	0	2	2	4	4	4	4	4	4	4	4	4	4	6	6	8
4	0	2	3	4	5	5	5	5	5	5	5	5	5	6	7	8
5	0	← เลือกใช้ค่าจากบน				← ที่บน										

Example

$$p = \{4, 2, 2, 1, 10\}$$

$$w = \{12, 2, 1, 1, 4\} \quad W = 15$$

ค่าที่รับ \rightarrow วนกลับไป ?
 \downarrow

Fill row 5 ($p_5=10$ $w_5=4$)

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	0	0	0	0	0	4	4	4	4
2	0	0	2	2	2	2	2	2	2	2	2	2	4	4	6	6
3	0	2	2	4	4	4	4	4	4	4	4	4	4	6	6	8
4	0	2	3	4	5	5	5	5	5	5	5	5	5	6	7	8
5	0	2	3	4	10	12	13	14	15	15	15	15	15	15	15	15

Example

$p = \{4, 2, 2, 1, 10\}$

$w = \{12, 2, 1, 1, 4\}$ $W = 15$

Fill row 5 ($p_5=10$ $w_5=4$)

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	0	0	0	0	0	4	4	4	4
2	0	0	2	2	2	2	2	2	2	2	2	2	4	4	6	6
3	0	2	2	4	4	4	4	4	4	4	4	4	4	6	6	8
4	0	2	3	4	5	5	5	5	5	5	5	5	5	6	7	8
5	0	2	3	4	10	12	13	14	15	15	15	15	15	15	15	15

Example

$p = \{4, 2, 2, 1, 10\}$

$w = \{12, 2, 1, 1, 4\}$ $W = 15$

Trace the solution backward to get the actual item number
We have item number 5,4,3,2

107 20:9
803 40m

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	0	0	0	0	0	4	4	4	4
2	0	0	2	2	2	2	2	2	2	2	2	2	4	4	6	6
3	0	2	2	4	4	4	4	4	4	4	4	4	4	6	6	8
4	0	2	3	4	5	5	5	5	5	5	5	5	5	6	7	8
5	0	2	3	4	10	12	13	14	15	15	15	15	15	15	15	15

Bottom-Up Code

```
set all K[0][*] = 0 and all K[*][0] = 0
for a = 1 to n
  for b = 1 to W
    if (w[a] > b)
      K[a][b] = K[a - 1][b];
    else
      K[a][b] = max( K[a - 1][b - w[a]] + p[a] ,
                     K[a - 1][b] )
return K[n][W];
```

recursion top down = (2^n)

$O(n(W))$

ไม่ต้องเก็บไว้ก็ได้

Can you write a code that generate the list of actual item that we take?

- Does this code generate too much subproblem?
- Does it generates one that we does not need?
- Is it better to use Top-Down approach?
 - Can you show some instance that Top-Down is better than Bottom-up (this code)

→ สร้าง subproblem ที่เราไม่ต้องการ ให้มันเยอะไป

ไม่ต้องเก็บไว้ก็ได้
มันจะเยอะไป 2^n ?

Analysis

- From Bottom-Up, it is clear that this is $O(Wn)$
- Original generate-all-solution method is $O(2^n)$
- Which one is better
 - In what case that $O(Wn)$ Dynamic Programming will benefit greatly (because there are several overlapping subproblems)

approximation algorithm

Instance

