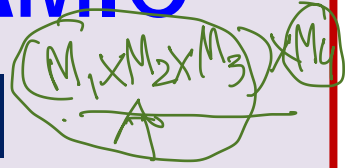
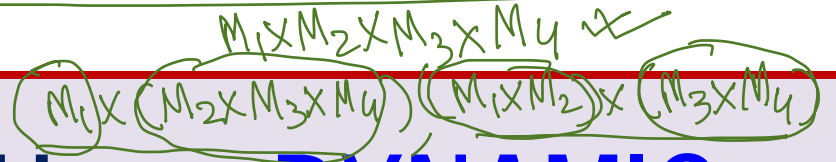


$M_1 \times M_2 \times M_3 \times \dots \times M_n$

$m(i, j) =$  optimal cost of multiplying Matrices  $M_i \times M_{i+1} \times \dots \times M_j$

$m(i, j) =$  BASE:  $\begin{cases} 0 & \text{if } i=j \end{cases}$   
 $M_i = r_i, c_i$



# ALGORITHM DESIGN USING DYNAMIC PROGRAMMING METHOD: I

if  $j = i+1$ ,  $r_i * c_i * c_j$

Recursive:  $- [r_i, c_k] [r_{k+1}, c_j]$

$m_{ij} = \begin{cases} m_{ik} + m_{k+1, j} \end{cases}$



Feb 7, 2022

$M_{i, j_1}, M_{i, j_2}, M_{i, j_3}$

1. Identical Subproblem Recognition
- Remember (Memoization)
- Reuse ✗



MINIMUM  $M + r_i * c_k * c_j$   
 $k = i \text{ to } j$

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Indian Institute of Technology Kharagpur

# Overview of Algorithm Design

## 1. Initial Solution

- a. Recursive Definition – A set of Solutions
- b. Inductive Proof of Correctness
- c. Analysis Using Recurrence Relations

## 2. Exploration of Possibilities

- a. Decomposition or Unfolding of the Recursion Tree
- b. Examination of Structures formed
- c. Re-composition Properties

## 3. Choice of Solution & Complexity Analysis

- a. ~~Balancing the Split~~, Choosing Paths,
- b. **Identical Sub-problems** *memoization*

## 4. Data Structures & Complexity Analysis

- a. Remembering Past Computation for Future
- b. Space Complexity

## 5. Final Algorithm & Complexity Analysis

- a. Traversal of the Recursion Tree
- b. Pruning

## 6. Implementation

- a. Available Memory, Time, Quality of Solution, etc

## 1. Core Methods

- a. Divide and Conquer
- b. Greedy Algorithms
- c. **Dynamic Programming**
- d. Branch-and-Bound
- e. Analysis using Recurrences
- f. Advanced Data Structuring

## 2. Important Problems to be addressed

- a. Sorting and Searching
- b. Strings and Patterns
- c. Trees and Graphs
- d. Combinatorial Optimization

## 3. Complexity & Advanced Topics

- a. Time and Space Complexity
- b. Lower Bounds
- c. Polynomial Time, NP-Hard
- d. Parallelizability, Randomization

# Basics of Dynamic Programming Method

$$n C_r = n-1 C_{r-1} + n-1 C_r$$

PASCAL'S Triangle



1. Recursive Decomposition  
 ↳ optimal sub-structure

optimization in nature

2. HANDLING IDENTICAL SUB-PROBLEMS ✓

3. MEMOIZATION & REUSE ✓ ✓

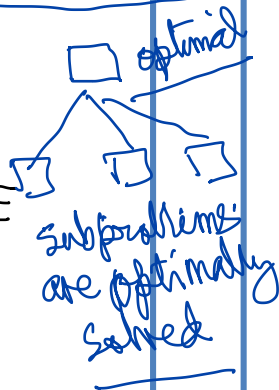
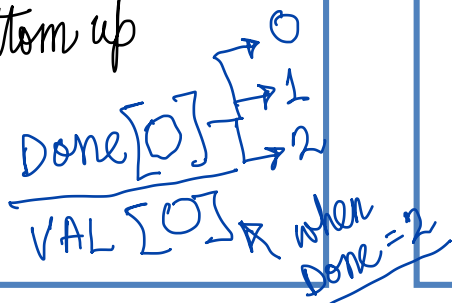
4. Evaluation

A) Top-down

B) Iterative Bottom up

5. Data Structures

gcd(2, 8)



a) Fibonacci (Pingala)

b) Matrix Chain Multiplication

c) String Related

- Longest Common Subsequence

- Sequence Alignment

- NLP related problems

d) Matrix operations

e) Graph Algorithms

f) coins / knapsack

g) optimal BST

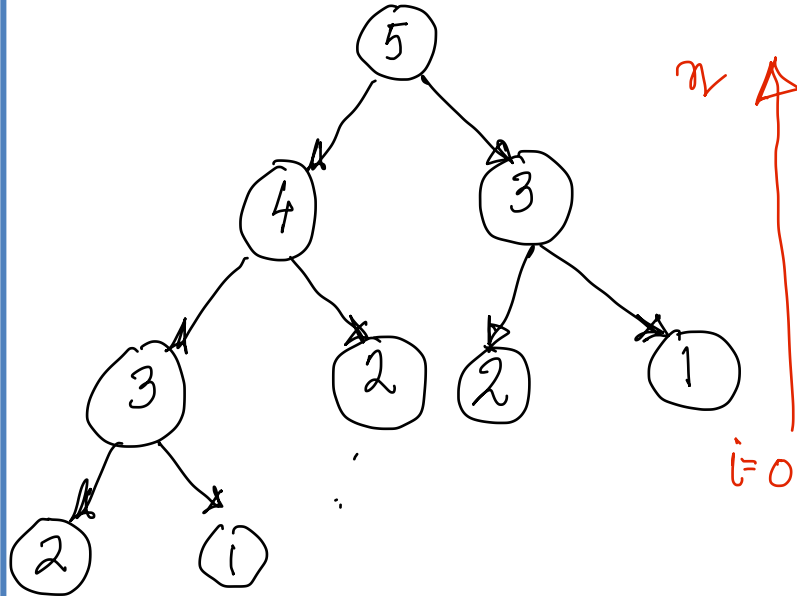
n!, coin, knapsack

# Revising Fibonacci-like Structures

$$f(n) = f(n-1) + f(n-2), n \geq 2$$

$$= 0, n=0$$

$$= 1, n=1$$



$F[]$ ,  $Done[]$

$Done[0] = Done[1] = 1$

Top-down

All other  $Done[i] = 0$

$F[0] = 0$   $F[1] = 1$

eval- $f(n)$

{ if ( $Done[n]$ ) return ( $F[n]$ )

$m = eval-f(n-1) + eval-f(n-2)$

$Done[n] = 1$

Bottom up

$F[n] = m$

return( $F[n]$ )

}

$F[0] = 0, F[1] = 1$   
 for  $i = 2$  to  $n$  do  
 $F[i] = F[i-1] + F[i-2]$

only 2 add'l variables

# Fibonacci-like Structures (cntd.)

$$f(n) = r(n) \text{ if } c(n) \text{ is true} \\ = f(g(n)) + f(h(n)) \\ \text{if } c(n) \text{ is false}$$

where  $c(n)$ ,  $r(n)$ ,  $g(n)$ ,  
 $h(n)$  are not recursive and  
can be computed deterministically

$F[i]$   
 $Done[i]$  —

- 0 if it has not been evaluated
- 1 if evaluation has begun but not completed
- 2 if final value is computed

```
eval-f(n)
{
  if (Done[n] = 2) return (F[n])
  if (c(n) = true)
    { Done[n] = 2, F[n] = r(n)
      return (F[n]) }
  if (Done[n] = 1) return ("CYCLE")
  Done[n] = 1
  x = g(n), y = h(n)
  z = eval-f(x) + eval-f(y)
  F[n] = z
  Done[n] = 2
}
```

$$M_1 \times M_2 \times \dots \times M_n$$

# MATRIX CHAIN MULTIPLICATION Problem

$a+b+c = (a+b)+c, a+(b+c)$   
 $y_1=x_2, y_2=x_3$   
 $M_1 \times M_2 \times M_3$   
 $[x_1, y_1] [x_2, y_2] [x_3, y_3]$   
 Mat Mul  $\rightarrow$  ASSOCIATIVE but NOT COMMUTATIVE

$(M_1 \times (M_2 \times (M_3 \times M_4))) = ((M_1 \times M_2) \times (M_3 \times M_4)) = (((M_1 \times M_2) \times M_3) \times M_4) = (M_1 \times (M_2 \times M_3)) \times M_4$   
 BUT THE NUMBER OF MULTIPLICATIONS TO GET THE ANSWER DIFFER !!

Let A be a [p by q] Matrix and B be a [q by r] Matrix. The number of multiplications needed to compute  $A \times B = p \times q \times r$

$M_1 \times M_2 \times M_3 \Rightarrow ((M_1 \times M_2) \times M_3)$   
 $\rightarrow M_1 \times (M_2 \times M_3)$   
 $A = [60, 6]$   
 $B = [6, 3]$

Thus if M1 be a [10 by 30] Matrix, M2 be a [30 by 5] Matrix and M3 be a [5 by 60] Matrix

Then the number of computations for

$(M_1 \times M_2) \times M_3 = 10 \times 30 \times 5$  for  $P = (M_1 \times M_2)$  and  $10 \times 5 \times 60$  for  $P \times M_3$ . Total = 4500  
 (Handwritten:  $1500 \rightarrow 3000$ )

$M_1 \times (M_2 \times M_3) = 10 \times 30 \times 5$  for  $Q = (M_2 \times M_3)$  and  $10 \times 30 \times 60$  for  $M_1 \times Q$ . Total = 27000  
 (Handwritten:  $9000 \rightarrow 18000$ )

$AB = A \times B [10 \times 3]$   
 $AB[i,j] = 6$   
 $10 \times 30 \times 6$

# Matrix Chain Multiplication: Recursive Definition

$$M_1 \times M_2 \times M_3 \times \dots \times M_n = [P_0, P_1] [P_1, P_2] [P_2, P_3] \dots [P_{n-1}, P_n] \quad [P_0, P_n]$$

$m_{ij}$  = optimal multiplications for multiplying  $M_i \times M_{i+1} \dots \times M_j$

$$m_{ij} = \begin{cases} 0 & \text{if } i=j \\ \min_{k=i}^{j-1} \{ m_{ik} + m_{k+1,j} + p_{i-1} * p_k * p_j \} & \text{if } i < j \end{cases}$$

$M_i \times (M_{k+1} \dots M_j)$

$$M_1 = 10 \times 30$$

$$M_2 = 30 \times 5$$

$$M_3 = 5 \times 60$$

$$M_4 = 60 \times 4$$

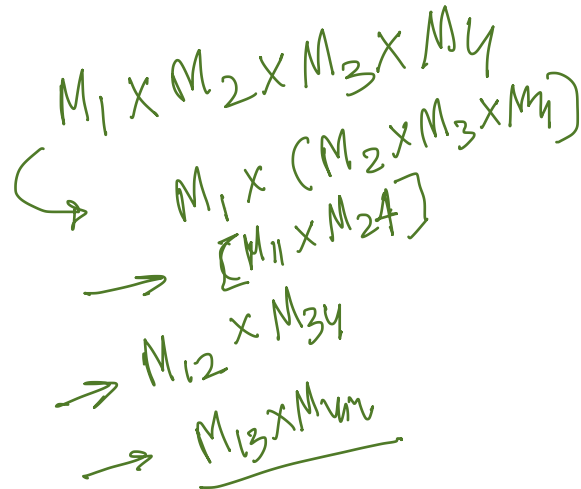
$$p_0 = 10$$

$$p_1 = 30$$

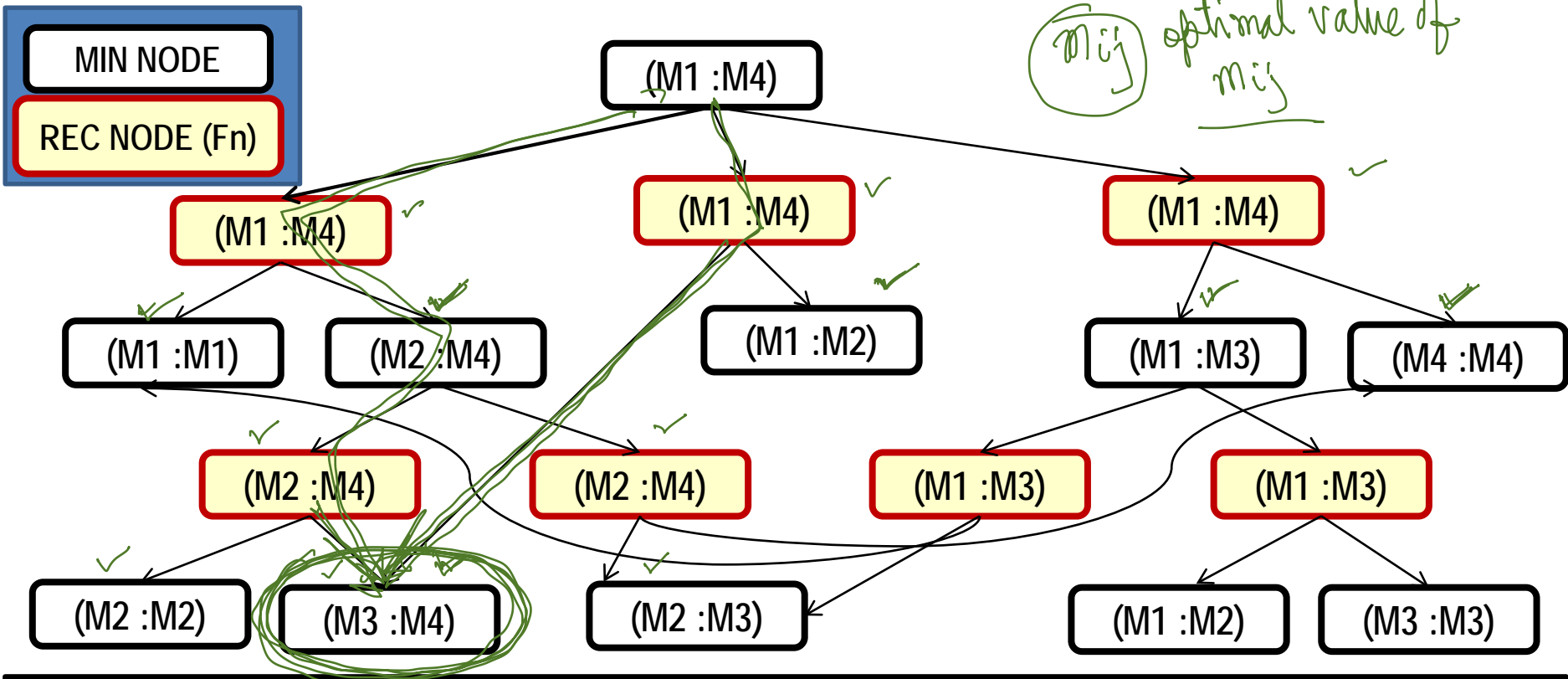
$$p_2 = 5$$

$$p_3 = 60$$

$$p_4 = 4$$



# MATRIX CHAIN MULTIPLICATION: RECURSIVE STRUCTURE

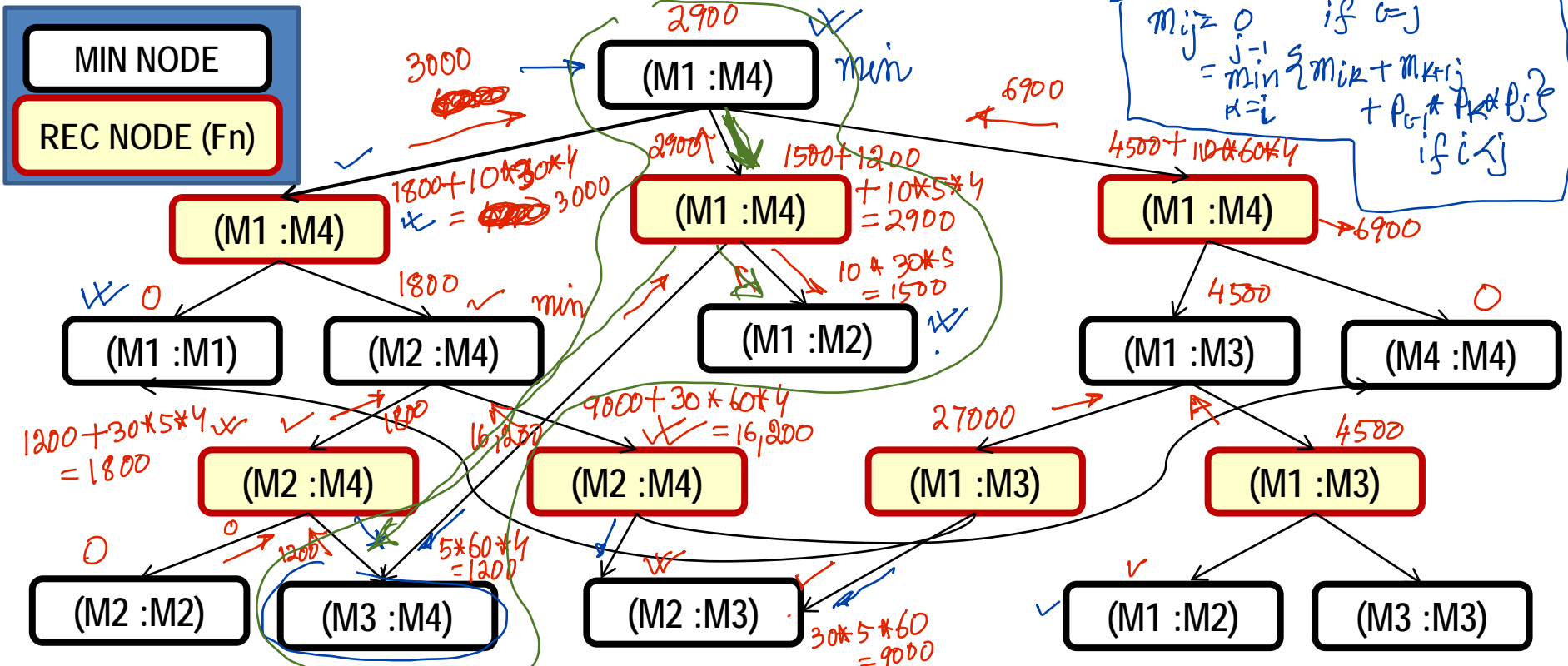


$$(M1 \times (M2 \times (M3 \times M4))) = ((M1 \times M2) \times (M3 \times M4)) = (((M1 \times M2) \times M3) \times M4) = (M1 \times (M2 \times M3)) \times M4$$

$$M1 [10 \text{ by } 30], M2 [30 \text{ by } 5], M3 [5 \text{ by } 60], M4 [60 \text{ by } 4]$$



# MATRIX CHAIN MULTIPLICATION: RECURSIVE STRUCTURE



$(M1 \times (M2 \times (M3 \times M4))) = ((M1 \times M2) \times (M3 \times M4)) = (((M1 \times M2) \times M3) \times M4) = (M1 \times (M2 \times M3)) \times M4$   
 $M1 [10 \text{ by } 30], M2 [30 \text{ by } 5], M3 [5 \text{ by } 60], M4 [60 \text{ by } 4]$

$p[0] = 10 \quad p[1] = 30 \quad p[2] = 5 \quad p[3] = 60 \quad p[4] = 4$

# Matrix Chain Multiplication: Top-Down Evaluation

```

M[i,j], Done[i,j] = 0
eval-m(i,j)
  REUSE { if (Done[i,j] = 1) return (M[i,j])
  if (i=j) { Done[i,j] = 1;
             M[i,j] = 0;
             return (M[i,j]) }
  BASE
  val = ∞
  for (k=i to j-1)
    { v_k = eval-m(i,k) +
          eval-m(k+1,j)
          + (p[i-1]*p[k]*p[j])
    if (v_k < val) val = v_k
  }
  }
  
```

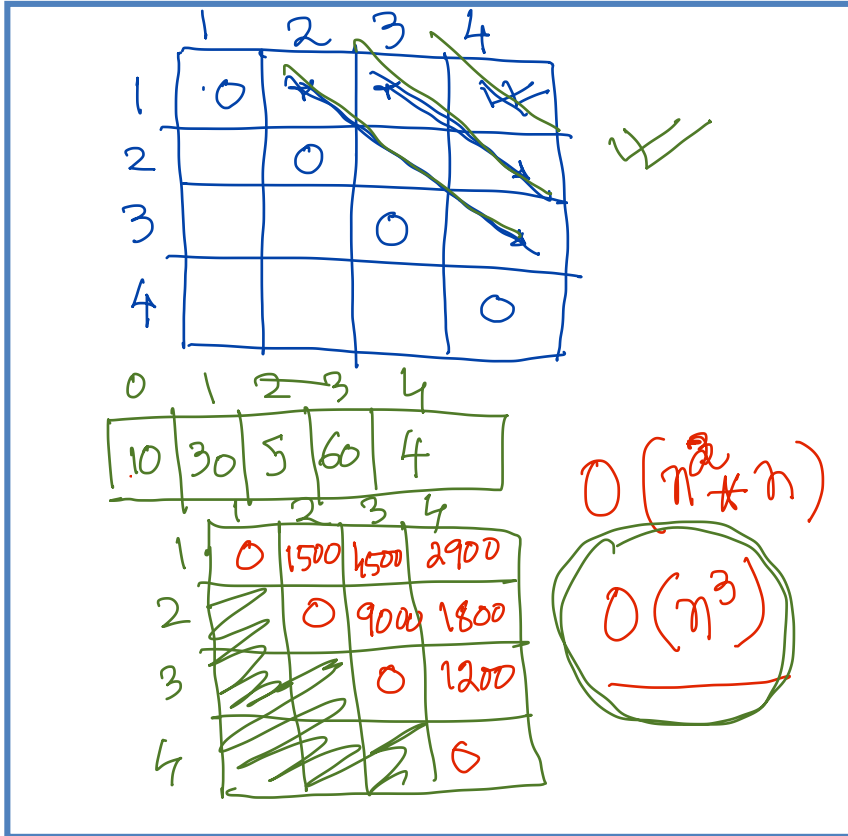
MINIMUM Recombos

```

Done[i,j] = 1 ✓
M[i,j] = val ✓
return (M[i,j]) ✓
}
  
```

$n^2 \rightarrow M[i,j]$   
 $n \rightarrow$  BASE  
 $O(n^3)$

# Matrix Chain Multiplication: Iterative Evaluation



iterative eval C)

```

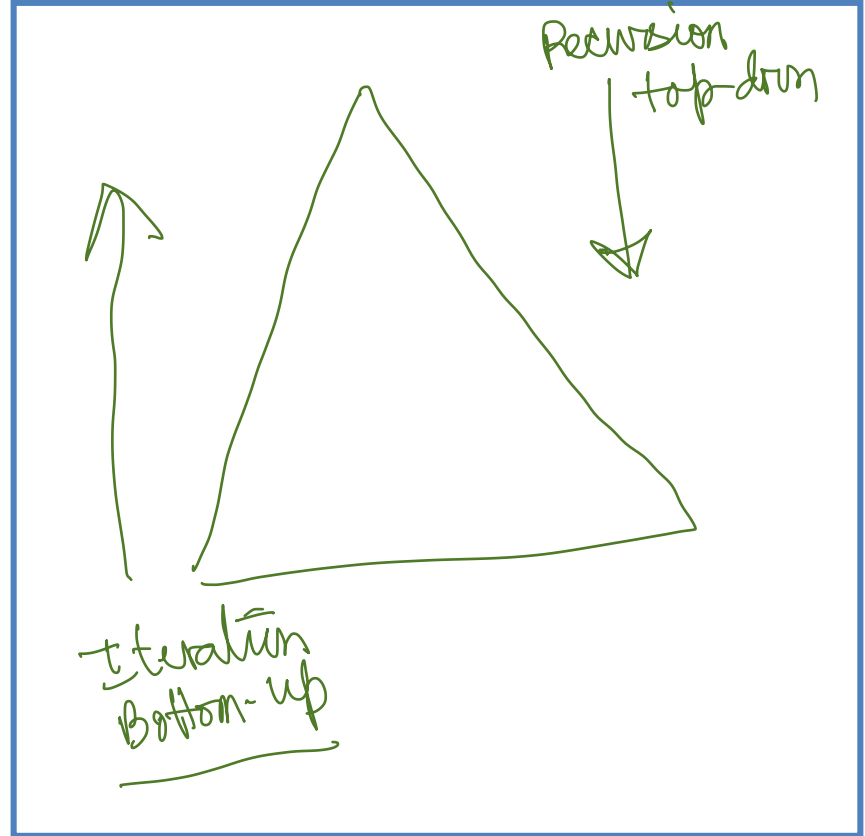
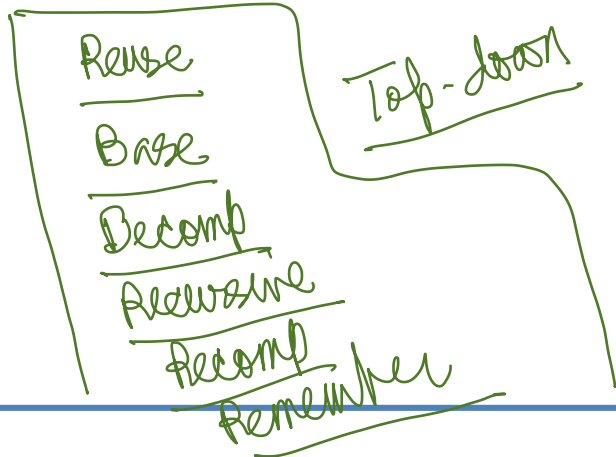
for (i=1 to n) M[i,i]=0 ✓
for (diff=1 to n-1) ✓
  for (i=1 to n-diff) ✓
    j = i + diff
    M[i,j] = ∞
    for (k=i to j-1)
      q = M[i,k] + M[k+1,j]
        + p[i-1] * p[k] * p[j]
      if (q < M[i,j]) M[i,j] = q
    }
  }
}
    
```

$O(n)$

# Summary

1. Recursive Sub-structure
2. Memorization & Reuse
3. Top-down & Iterative (Recursive)

Algorithms



**Thank you**

**Any Questions?**