

ALGORITHM DESIGN USING DYNAMIC PROGRAMMING METHOD: I

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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

divide & conquer
greedy

3. Choice of Solution & Complexity Analysis

- a. Balancing the Split, Choosing Paths
- b. Identical Sub-problems memorization

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

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

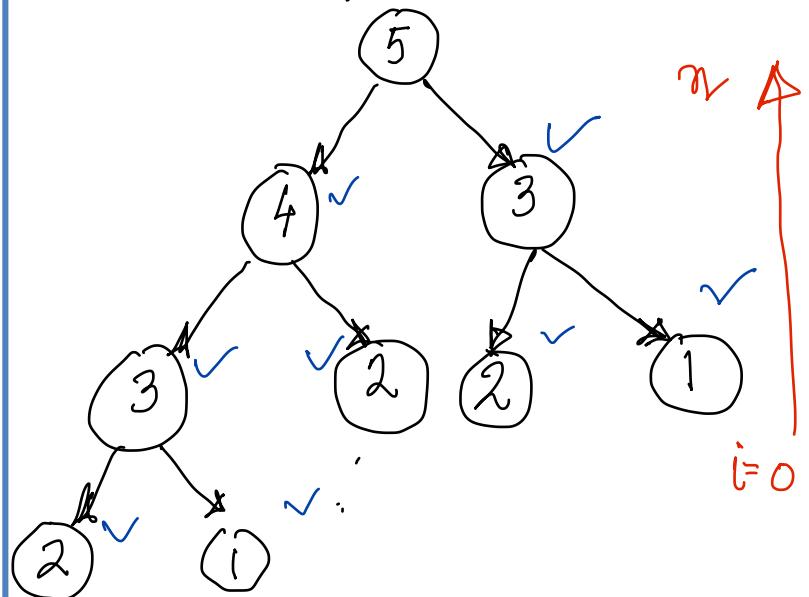
1. Recursive Decomposition optimization
↳ optimal sub-structure
 2. HANDLING IDENTICAL SUB-PROBLEMS }
exponential time complexity
 3. MEMOIZATION & REUSE
Remember → Data Storage
 4. Evaluation
 - A) Top-down ✓ done []
 - B) Iterative Bottom up ↲
 5. Data Structures
↳ preprocessing
- acyclic structure

- 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
- 

Revising Fibonacci-like Structures

$$f(n) = f(n-1) + f(n-2), n \geq 2$$
$$= 0, n=0$$
$$= 1, n=1$$

$n-k$ constant k



$F[i]$, $\text{Done}[i]$

$$\left. \begin{array}{l} \text{Done}[0] = \text{Done}[1] = 1 \\ \text{All other } \text{Done}[i] = 0 \end{array} \right\}$$
$$F[0] = 0 \quad F[1] = 1$$

eval-f(n)

{
 if $\text{Done}[n] = 1$ return ($F[n]$)
 else

$m = \text{eval-f}(n-1) + \text{eval-f}(n-2)$

$\text{Done}[n] = 1$

$F[n] = m$

} 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

Ackermann's Function

Fibonacci-like Structures (cntd.)

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

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

$F[]$

$\text{Done}[i]$

check for cycles

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

Top-down

1

eval-f(n)

if ($\text{Done}[n] = 2$) return ($F[n]$)

Base

if ($c(n) = \text{true}$)

$\text{Done}[n] = 2, F[n] = r(n)$

return ($F[n]$)

3

cycle detection

if ($\text{Done}[n] = 1$) return ("CYCLE")

$\text{Done}[n] = 1$

$x = g(n), y = h(n)$

4

Possible

$z = \text{eval-f}(x) + \text{eval-f}(y)$

$F[n] = z$

$\text{Done}[n] = 2$

return ($F[n]$)

2

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

MATRIX CHAIN MULTIPLICATION Problem

 $M_1 \times N_2 \times \dots \times N_n$

$$(M_1 \times (M_2 \times (M_3 \times M_4))) = ((M_1 \times M_2) \times (M_3 \times M_4)) = (((M_1 \times M_2) \times M_3) \times M_4) = \\ (M_1 \times (M_2 \times M_3)) \times M_4) \leftarrow M_1 \times ((M_2 \times M_3) \times M_4)$$

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 $A \times B = p^*q^*r$

$$\rightarrow C(1) = 0 \quad C(2) = 1$$

$$M_1 \times M_2 \times M_3 \times M_4$$

associative

$$[P_1, P_2] [P_2, P_3] [P_3, P_4]$$

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

$$(M_1 \times M_2) \times M_3 = 10^*30^*5 \text{ for } P = (M_1 \times M_2) \text{ and } 10^*5^*60 \text{ for } P \times M_3. \text{ Total } = 4500$$

$$M_1 \times (M_2 \times M_3) = 30^*5^*60 \text{ for } Q = (M_2 \times M_3) \text{ and } 10^*30^*60 \text{ for } M_1 \times Q. \text{ Total } = 27000$$

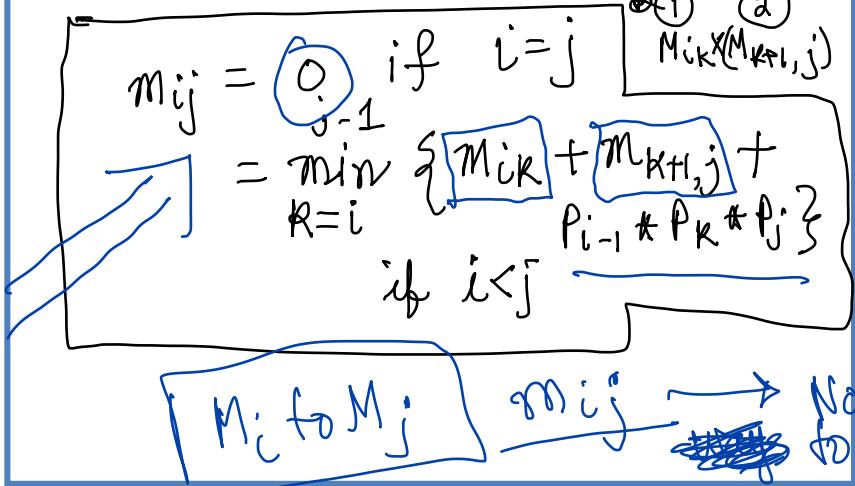
$$C(n) = \sum_{i=1}^{n-1} C(i) * C(n-i)$$

Matrix Chain Multiplication: Recursive Definition

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

$$[P_0, P_1] [P_1, P_2] [P_2, P_3] \dots [P_{n-1}, P_n]$$

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



$$M_1 = 10 \times 30$$

$$M_2 = 30 \times 5$$

$$M_3 = 5 \times 60$$

$$M_4 = 60 \times 4$$

$$p_0 = 10$$

$$p_1 = 30$$

$$p_2 = 5$$

$$p_3 = 60$$

$$p_4 = 4$$

~~computation~~
~~M_i ... M_j~~
~~m_{ik}~~
~~[p_{i-1} p_k] [p_k p_j]~~
~~p_{i-1} * p_k * p_j~~

$$[P_0, P_1] [P_1, P_2] [P_2, P_3]$$

$$M_1 \times M_2 \times M_3$$

$$[P_1, P_2] [P_2, P_3] [P_3, P_4]$$

$$\xrightarrow{M_2} 0$$

$$M_3 \times M_4$$

$$[P_2, P_3] [P_3, P_4]$$

$$P_2 * P_3 * P_4$$

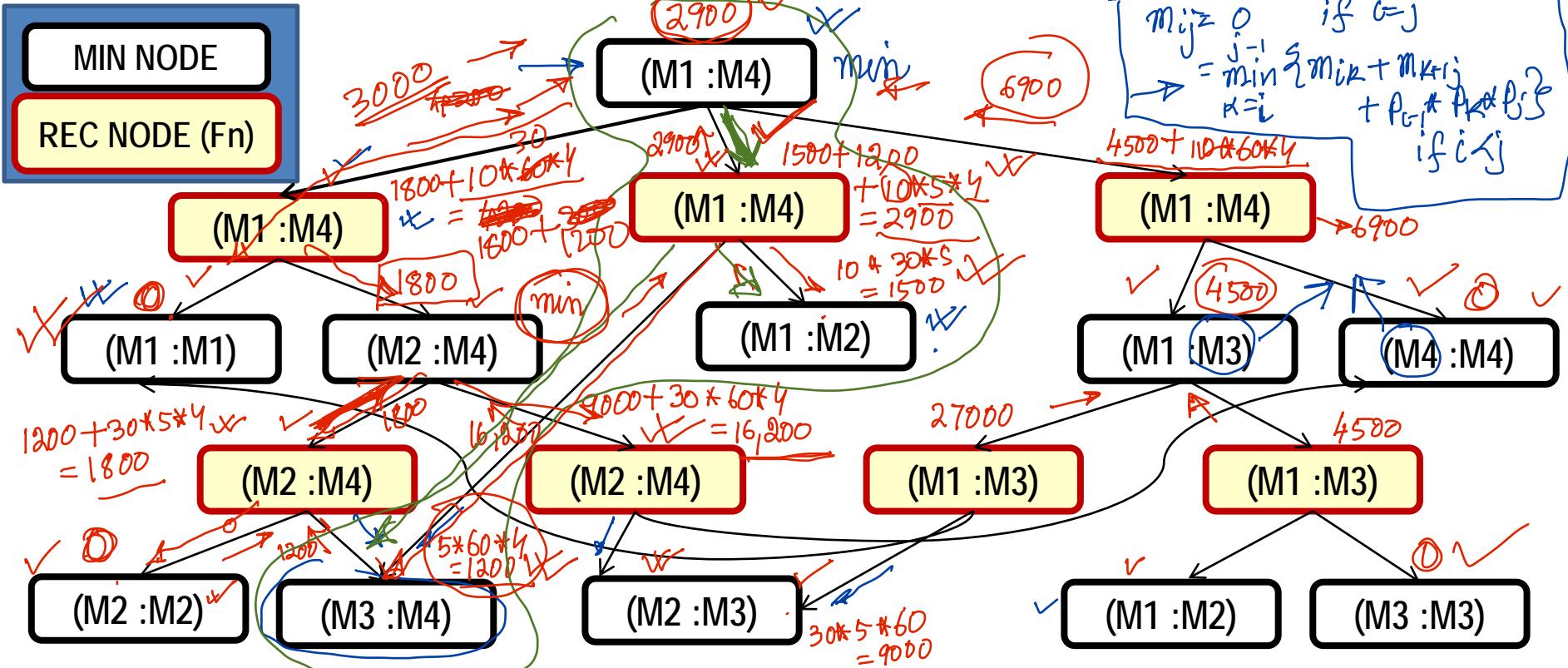
$$(M_i \dots M_k) \times (M_{k+1} \dots M_j)$$

$$m_{ik}$$

$$[p_{i-1} p_k] [p_k p_j]$$

$$p_{i-1} * p_k * p_j$$

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 by 30], M2 [30 by 5], M3 [5 by 60], M4 [60 by 4]

$$P_1[10] P_2[30] P_3[5] P_4[60] P_5[4]$$

Matrix Chain Multiplication: Top-Down Evaluation

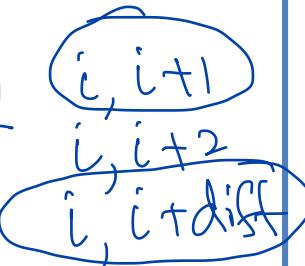
$M[i, j]$, $\text{Done}[i, j] = 0$
 $\text{eval_m}(i, j)$
REUSE
 $\{ \text{if } (\text{Done}[i, j] = 1) \text{ return } (M[i, j]) \}$
BASE
 $\{ \text{if } (i=j) \{ \text{Done}[i, j] = 1; M[i, j] = 0; \text{return } (M[i, j]) \} \}$
RECURSIVE
 $\text{Val} = \alpha$
 $\text{for } (k=i \text{ to } j-1) \quad \checkmark$
 $\{ v_k = \text{eval_m}(i, k) + \text{eval_m}(k+1, j) + p[i-1] * p[k] * p[j] \}$
 $\{ \text{if } (v_k < \text{val}) \text{ val} = v_k \}$
 $(1, n)$

$\text{Done}[i, j] = 1$
 $M[i, j] = \text{val}$
 $\text{return } (M[i, j])$

 $n^2 \rightarrow M[i, j]$
 n
 $O(n^3)$
I can decipher my # acyclic dependency structure
 $(i, i) \quad O(n^2 * n)$
 $(i, i+1) \quad = O(n^3)$
 $(i, i+2) \quad (i, i+3) \dots (i, n)$
Bottom-up iterative algorithm

Matrix Chain Multiplication: Iterative Evaluation

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



	1	2	3	4
1	10	30	5	60
2		2	3	4
3				
4				

$O(n^2 \times n)$

$O(n^3)$

$\frac{(1,2)}{(1,3)}, \frac{(2,3)}{(2,4)}, \dots$

	1	2	3	4
1	0	1500	1500	2900
2		0	9000	1800
3			0	1200
4				0

iterative eval()

{ for ($i = 1$ to n) $M[i, i] = 0$

for ($diff = 1$ to $n-1$)

for ($i = 1$ to $n-diff$)

$j = i + diff$

$M[i, j] = \alpha$

for ($k = i$ to $j-1$)

$q = M[i, k] + M[k+1, j]$
 $+ p[i-1] * p[k] * p[j]$

if ($q < M[i, j]$) $M[i, j] = q$

$O(n \cdot n \cdot n) = O(n^3)$

Summary

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

M

~~Max~~ Min

Max Min

$$L = \{(i, i)\}$$

~~From L~~

$$\text{Let } L = \{(x_1, y_1), (x_2, y_2), \dots, (x_k, y_k)\}$$

Select some (p, q) from L

Select some $(p_1, q_1), (p_2, q_2)$ from L which can be merged (p_3, q_3)

$$\begin{array}{l} p_1 < q_1 \\ p_2 < q_2 \end{array} \quad \begin{array}{l} q_1 = p_2 - 1 \\ P_3 = P_1 \end{array} \quad \begin{array}{l} q_3 = q_2 \\ P_3 = P_1 \end{array}$$

Min

(7, 9)

Thank you

Any Questions?