

* Greedy Algorithm ~~or~~ Dynamic Programming

- based on past choices
- optimization can be solved
- cannot prove correctness of greedy
- running time is less compared to divide &

conquer

Most Algo \rightarrow Do not lead to optimal solⁿ, but we can be sure we ~~can~~ ^{are} near to optimal solution

* Divide & Conquer

- Independent of previous choice
- optimization cannot be solved
- Easy to prove correctness
- running time is more

B_3, B_2, B_1, B_0
1 0 1 1

\Rightarrow Knapsack problem ($W = 60\text{kg}$)

Item	Weight	Value	$(V/W)_i$	$(V/W)_i, \text{dex}$
1	5	30	6	$\rightarrow 6 \text{ (1)}$
2	10	40	4	$\rightarrow 4 \text{ (2)}$
3	15	45	3	$\rightarrow 3.6 \text{ (5)}$
4	(22)	77	3.5	$\rightarrow 3.5 \text{ (4)}$
5	25	90	3.6	$\rightarrow 3 \text{ (3)}$

Weight selected $\Rightarrow 5 + 10 + 25 + \left(\frac{60 - 40}{22} \right) \times 77$

Profit $\rightarrow 5 \times 6 + 10 \times 4 + 25 \times 3.6 + \left(\frac{60 - 40}{22} \right) \times 77$

= 230

Capacity = 50

Q	Item	Weight	Value
		10	60
		20	80
		40	100

Weight $\rightarrow 10 + 20 + \left(\frac{50 - 30}{40} \right) \rightarrow$ Here multiply value

Profit $\rightarrow 10 \times 6 + 20 \times 4 + \frac{1}{2} \times 100$

$= 60 + 80 + 50 = 190$

Capacity = 10

Q	Item	Weight	Value	$\left(\frac{\text{Value}}{\text{Weight}} \right)$	$\left(\frac{V}{w} \right)_{\text{desc}}$
	1	4	12	3	50 - (5)
	2	8	32	4	20 - (3)
	3	2	40	20	5 - (2)
	4	6	30	5	4 - (2)
	5	1	50	50	3 - (1)

Weight $= 4 + 2 + 6 + \left(\frac{10 - 9}{8} \right)$

Profit $= 1 \times 50 + 2 \times 20 + 6 \times 5 + \frac{10 - 9}{8} \times 32$

$= 50 + 40 + 30 + 4$

$= 124$

→ Floyd $O(n^3)$

⇒ Dynamic Programming

Similarities betⁿ Dynamic & Greedy approach.

- 1) Both are applicable to the problem which have optimal sub-structure property
- 2) Both have are used for optimisation

Optimal Sub-Structure property :-

→ The optimal solⁿ to the problem always contains optimal solⁿs to the subproblems.

Dynamic	Greedy
→ Time complexity is more	→ Time complexity is less.

⇒ Techniques used to design DP problem

- 1) Tabulation → ~~But~~
- 2) Memoization

Tabulation

Memoization

- Bottom up approach
- Efficient more because we start from basic and will move to final goal

- Top-down approach
- ~~Less~~ Less efficient because we start from advance then will see ^{what} which ~~of~~ needed consequently i.e. decision-making is not perfect

⇒ Matrix - chain multiplication

- Let there is chain of matrices ABCDE... we need to put the parenthesis i.e. multiply in such a way that number of scalar multiplication will be as minimum as possible.
- This process is known as characterization.

⇒ See chap 15 of book.

$$(A(B(C(D E))))$$

~~$$A(B(C(D)E))$$~~

A B C D

→ Formula is $\frac{(2n)!}{(n+1)!n!} = \frac{8!}{5!4!} =$



This is the formula of catalan numbers

~~Let~~ $P(n)$ is the parenthesization for n matrices

$$P(n) = \sum_{k=1}^{n-1} P(k) P(n-k)$$

$$P(4) = P(1)P(3) + P(2)P(2) + P(3)P(1)$$

$$P(3) = P(1)P(2) + P(2)P(1)$$

$$P(2) = P(1)P(1) = 1$$

$$\therefore P(3) = 1 + 1 = 2$$

$$P(4) = 2 + 1 + 2 = 5$$

~~$$P(5)$$~~

$$P(5) =$$

✓

$$m[i, j] = \begin{cases} 0 & \text{if } i=j \\ \min_{i \leq k < j} \{ m[i, k] + m[k+1, j] + P(i-1)P(k)P(j) \} & \text{if } i < j \end{cases}$$

$$m[1, 2] = P(0) \cdot P(1) \cdot P(2) \quad \left(\text{You can cross check by formula} \right)$$

$$m[3, 4] = P(2) \cdot P(3) \cdot P(4) \quad \left(\text{For consecutive numbers} \right)$$

⇒ Matrix chain order (code) from textbook page 375
 (Yaad rakhtono algo exam marks 2 var puchayeta)

Matrix chain order accepts sequence of dimensions as input, also number of matrices is $(\text{total length of dimensions} - 1)$.

$$m[1,3] = \{ m[1,1] + m[2,3] + \cancel{m[1,2]} + p(0)p(1)p(2) \}$$

$$m[2,3] = p(1) \cdot p(2) \cdot p(3)$$

$$\cancel{p(0)} \quad p(1) p(2) p(3) + p(0) p(1) p(3)$$

$$p(0) p(2) p(3)$$

Don't write this term.

$$(i \leq k < j-1)$$

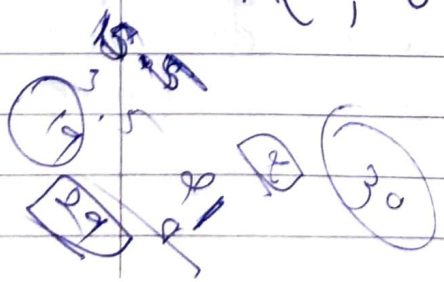
$$m[1,4] = \left\{ \begin{aligned} &\cancel{m[1,1]} + \cancel{m[2,4]} + p_0 p_1 p_4 \\ &m[1,2] + m[3,4] + p_0 p_2 p_4 \\ &m[1,3] + \cancel{m[4,4]} + p_0 p_3 p_4 \end{aligned} \right.$$

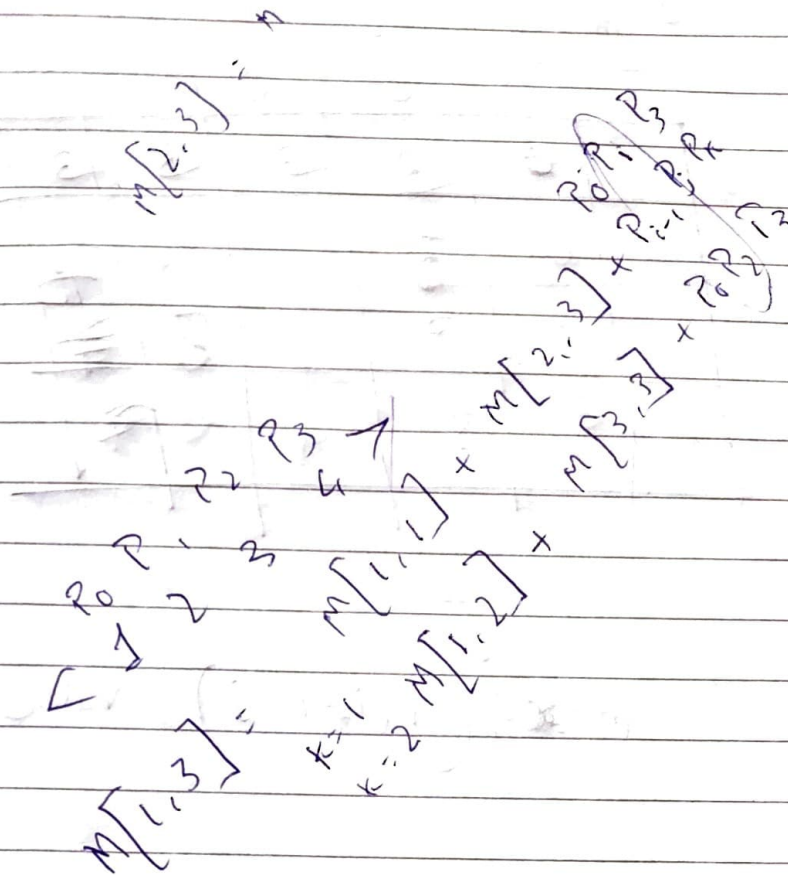
$$m[1,2] = p_0 p_1 p_2$$

$$\cancel{p_1 p_3 p_4}$$

$$m[1,6]$$

$$m[2,4] = \left\{ \begin{aligned} &\cancel{m[2,2]} + \cancel{m[3,4]} + p_1 p_2 p_4 \end{aligned} \right.$$





⇒ 0-1 knapsack problem.

knapsack cap = 5

Item	Weight	Value
1	2	3
2	3	4
3	4	5
4	5	6

Item vs knapsack capacity

$T(1,2)$

$T(0,1)$ Capacity

$T(0,1)$

weight Item i		0	1	2	3	4	5
0		0	0	0	0	0	0
1	2	0	0	0	3	3	3
2	3	0	0	3	4	4	7
3	4	0	0	3	4	5	7
4	5	0	6	3	4	0	7

$$T(i, j) = \max(T(i-1, j), \text{value} + T(i-1, j-\text{weight}))$$

Handwritten calculations and a table:

$T(1, 5) = 4 + T(1, 2)$
 $T(2, 5) = 5 + T(1, 3)$
 $T(3, 5) = 6 + T(1, 4)$
 $T(4, 5) = 7 + T(1, 5)$

weight: 1, 2, 3
value: 10, 15, 40

weight	0	1	2	3	4	5	6
0	0	0	0	0	0	0	0
1	0	10	10	10	10	10	10
2	0	10	15	25	25	25	25
3	0	10	15	40	50	55	65

40

2, 1, 0 = 68

Items $\rightarrow 5$

Weight $\rightarrow 2, 5, 3, 8, 1$

Value $\rightarrow 3, 5, 5, 9, 3$

		j									
		0	1	2	3	4	5	6	7	8	
i	0	0	0	0	0	0	0	0	0	0	
	1	0	0	3	3	3	3	3	3	3	
	2	0	0	3	3	3	5	5	8	8	
	3	0	0	3	5	5	8	8	9	10	
	4	0	0	3	5	5	8	8	8	10	
	5	0	3	3	6	8	8	11	11	13	

			5 3 7 3									
			0	1	2	3	4	5	6	7	8	
0	0	0	0	0	0	0	0	0	0	0	0	
3	1	1	0	3	3	3	3	3	3	3	3	
3	2	2	0	3	3	6	6	6	6	6	6	
5	3	3	0	3	3	6 8	8	8	11	11	11	
5	4	4	0	3	3	6 8	8	8	11	11	11	
9	8	5	0	3	3	6	8	8			14	