



# More Dynamic Programming

## Matrix Chain Multiplication

# Matrix Chain Multiplication

- Given some matrices to multiply, determine the *best* order to multiply them so you minimize the total number of single element multiplications.
  - i.e. Determine the way the matrices are parenthesized.
- First off, it should be noted that matrix multiplication is associative, but not commutative. But since it is associative, we always have:
- $((AB)(CD)) = (A(B(CD)))$ , or any other grouping as long as the matrices are in the same consecutive order.
- BUT NOT:  $((AB)(CD)) = ((BA)(DC))$

# Matrix Chain Multiplication

- It may appear that the amount of work done won't change if you change the parenthesization of the expression, but we can prove that is not the case!
- Let us use the following example:
  - Let  $A$  be a  $2 \times 10$  matrix
  - Let  $B$  be a  $10 \times 50$  matrix
  - Let  $C$  be a  $50 \times 20$  matrix
- Try  $(AB)C$  vs  $A(BC)$
- But FIRST, let's review some matrix multiplication rules...

# Multiplying two Matrices ( $C = AB$ )

- Dimension of matrix  $A = n * m$
- Dimension of matrix  $B = m * l$
- Dimension of matrix  $C = n * l$  ( how ? )
- Cost of multiplying these two matrices =  $n * m * l$  ,  
how ?

$$A = \begin{bmatrix} a_{00} & a_{01} & a_{02} & & & \\ & a_{10} & a_{11} & a_{12} & & \\ & & a_{20} & a_{21} & a_{22} & & \end{bmatrix} \quad B = \begin{bmatrix} b_{00} & b_{01} & b_{02} & b_{03} & & \\ & b_{10} & b_{11} & b_{12} & b_{13} & \\ & & b_{20} & b_{21} & b_{22} & b_{23} \end{bmatrix}$$

Lets see how many calculations do we make for  $C$

# Matrix Chain Multiplication

- Let's get back to our example: We will show that the way we group matrices when multiplying A, B, C *matters*:
  - Let A be a  $2 \times 10$  matrix
  - Let B be a  $10 \times 50$  matrix
  - Let C be a  $50 \times 20$  matrix
- Consider computing **A(BC)**:
  - # multiplications for  $(BC) = 10 \times 50 \times 20 = 10000$ ,
  - $D = BC$ , dimension of  $D = 10 \times 20$
  - # multiplications for  $A(D) = 2 \times 10 \times 20 = 400$
  - Total multiplications =  $10000 + 400 = 10400$ .
- Consider computing **(AB)C**:
  - # multiplications for  $(AB) = 2 \times 10 \times 50 = 1000$ ,
  - $E = (AB)$ , dimension of  $E = 2 \times 50$
  - # multiplications for  $(E)C = 2 \times 50 \times 20 = 2000$ ,
  - Total multiplications =  $1000 + 2000 = 3000$

# Matrix Chain Multiplication

- Thus, our **goal** today is:
- Given a chain of matrices to multiply, determine the fewest number of multiplications necessary to compute the product, also calculate the final answer matrix

# Matrix Chain Multiplication

- The key to solving this problem is noticing the ***sub-problem optimality condition***:
  - If a particular parenthesization of the whole product is optimal, then any sub-parenthesization in that product is optimal as well.
- ***Say What?***
  - ***If***  $(A (B ((CD) (EF)) ) )$  is optimal
  - Then  $(B ((CD) (EF)) )$  is optimal as well
  - ***Proof on the next slide...***

# Matrix Chain Multiplication

- Assume that we are calculating ABCDEF and that the following parenthesization is optimal:
  - $(A (B ((CD) (EF)) ))$
  - Then it is necessarily the case that
    - $(B ((CD) (EF)) )$
    - is the optimal parenthesization of BCDEF.
- Why is this?
  - Because if it wasn't, and say  $((BC) (DE)) F$  was better, then it would also follow that
    - $(A ((BC) (DE)) F)$  was better than
    - $(A (B ((CD) (EF)) ))$ ,
  - contradicting its optimality!



# Matrix Chain Multiplication

- Our final multiplication will ALWAYS be of the form
  - $(A_0 \cdot A_1 \cdot \dots \cdot A_k) \cdot (A_{k+1} \cdot A_{k+2} \cdot \dots \cdot A_{n-1})$
- In essence, there is exactly one value of  $k$  for which we should "split" our work into two separate cases so that we get an optimal result.
  - Here is a list of the cases to choose from:
    - $(A_0) \cdot (A_1 \cdot A_{k+2} \cdot \dots \cdot A_{n-1})$
    - $(A_0 \cdot A_1) \cdot (A_2 \cdot A_{k+2} \cdot \dots \cdot A_{n-1})$
    - $(A_0 \cdot A_1 \cdot A_2) \cdot (A_3 \cdot A_{k+2} \cdot \dots \cdot A_{n-1})$
    - ...
    - $(A_0 \cdot A_1 \cdot \dots \cdot A_{n-3}) \cdot (A_{n-2} \cdot A_{n-1})$
    - $(A_0 \cdot A_1 \cdot \dots \cdot A_{n-2}) \cdot (A_{n-1})$
- Basically, count the number of multiplications in each of these choices and **pick the minimum**.
  - One other point to notice is that you have to account for the minimum number of multiplications in each of the two products.

# Matrix Chain Multiplication

- Consider the case multiplying these 4 matrices:
  - A:  $2 \times 4$
  - B:  $4 \times 2$
  - C:  $2 \times 3$
  - D:  $3 \times 1$
- 1.  $(A)(BCD)$  - This is a  $2 \times 4$  multiplied by a  $4 \times 1$ ,
  - so  $2 \times 4 \times 1 = 8$  multiplications, plus whatever work it will take to multiply  $(BCD)$ .
- 2.  $(AB)(CD)$  - This is a  $2 \times 2$  multiplied by a  $2 \times 1$ ,
  - so  $2 \times 2 \times 1 = 4$  multiplications, plus whatever work it will take to multiply  $(AB)$  and  $(CD)$ .

# Matrix Chain Multiplication

- **Our recursive formula:**

- $\text{cost}(i, j) = \text{cost}(i, k) + \text{cost}(k + 1, j) + P_{i-1} \cdot P_k \cdot P_j$

- Let's solve it using recursion

```

23 #define ordered_set tree<int, null_type, less<int>, rb_tree_tag, tree_order_statistics_node_adapter>
24 int dp[105][105];
25
26 int MCM(vector<int>&arr,int x,int y)
27 {
28     if(x==y)
29         return 0;
30     if(dp[x][y]!=-1)
31         return dp[x][y];
32     int ans=INT_MAX;
33     for(int i=x;i<y;i++)
34         ans=min(ans,MCM(arr,x,i)+MCM(arr,i+1,y)+arr[i]*arr[x-1]*arr[y]);
35     return dp[x][y]=ans;
36 }
37
38 int main()
39 {
40     boost
41     int t=1;
42     cin>>t;
43     while(t--)
44     {
45         int n;
46         cin>>n;
47         memset(dp,-1,sizeof(dp));
48         vector<int>arr(n);
49         for(int i=0;i<n;i++)
50             cin>>arr[i];
51         cout<<MCM(arr,1,n-1)<<"\n";
52     }
53     return 0;
54 }

```

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