



# 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:  $((AB)(CD)) \neq ((BA)(DC))$

# 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 ?

$$A = [ a_{00} \ a_{01} \ a_{02}$$

$$\quad a_{10} \ a_{11} \ a_{12} ]$$

$$\quad a_{20} \ a_{21} \ a_{22} ]$$

$$B = [ b_{00} \ b_{01} \ b_{02} \ b_{03}$$

$$\quad b_{10} \ b_{11} \ b_{12} \ b_{13} ]$$

$$\quad b_{20} \ b_{21} \ b_{22} \ b_{23} ]$$

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 2x10 matrix
  - Let B be a 10x50 matrix
  - Let C be a 50x20 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*  $(A \cdot (B \cdot ((CD) \cdot (EF))))$  is optimal
  - Then  $(B \cdot ((CD) \cdot (EF)))$  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: 2x4
  - B: 4x2
  - C: 2x3
  - D: 3x1
- 1. (A)(BCD) - This is a 2x4 multiplied by a 4x1,
  - so  $2 \times 4 \times 1 = 8$  multiplications, plus whatever work it will take to multiply (BCD).
- 2. (AB)(CD) - This is a 2x2 multiplied by a 2x1,
  - 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 treell, null_type,lessll, treell_log,tree_order_statistics_no
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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