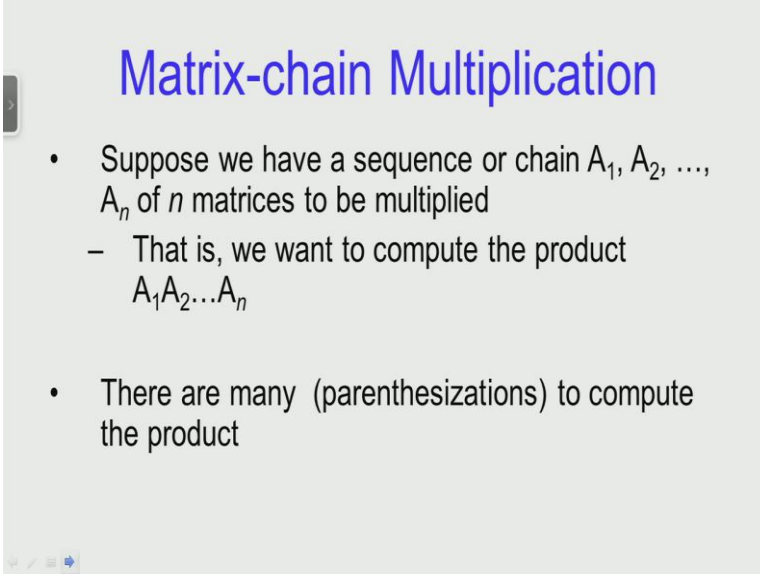


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**Lecture – 57**  
**Dynamic Programming**

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**Matrix-chain Multiplication**

- Suppose we have a sequence or chain  $A_1, A_2, \dots, A_n$  of  $n$  matrices to be multiplied
  - That is, we want to compute the product  $A_1 A_2 \dots A_n$
- There are many (parenthesizations) to compute the product

Today we are going to study a second problem which is solved using the method of dynamic programming; this problem is called the matrix chain multiplication. The problem has that we are given a sequence of matrices that can be multiplied. So,  $A_1, A_2$  up to  $A_n$  are  $n$  matrices, which are given. And our aim is to compute the product of these  $n$  matrices. Let us assume that the product can be performed, in other words the orders of the matrices are appropriately given. There are many parenthesizations to compute the product. What is the parenthesization? The parenthesization is an allocation of parentheses around the expression that has been given. So, that the expression can be evaluated has specified by the parenthesization.

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**Example**

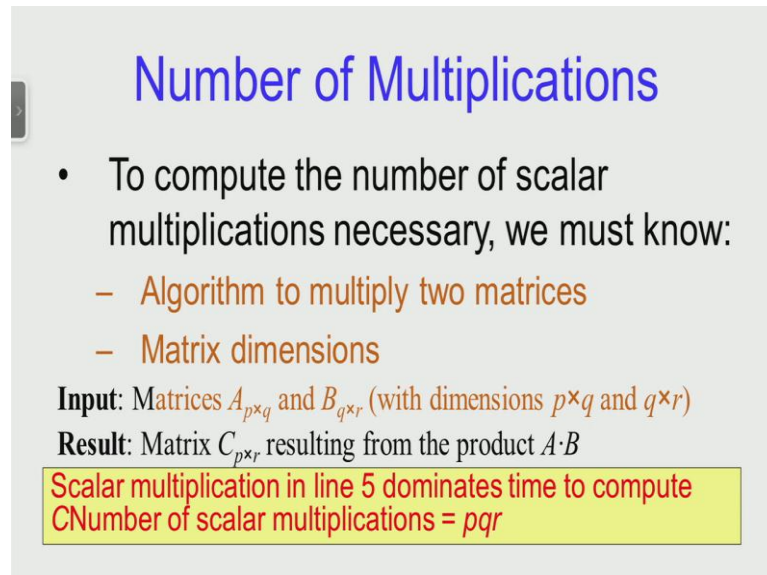
- Example: consider the chain  $A_1, A_2, A_3, A_4$  of 4 matrices
  - Let us compute the product  $A_1A_2A_3A_4$
- There are 5 possible ways:
  1.  $(A_1(A_2(A_3A_4)))$
  2.  $(A_1((A_2A_3)A_4))$
  3.  $((A_1A_2)(A_3A_4))$
  4.  $((A_1(A_2A_3))A_4)$
  5.  $((A_1A_2)A_3)A_4$

Let us now consider this example, where there are 4 matrices -  $A_1, A_2, A_3$ , and  $A_4$ ; and aim is to compute the product of these 4 matrices. There are 5 possible parenthesizations, let us inspect these each of parenthesization, this is the first one. We know that in any expression which is parenthesized, the innermost parenthesization is evaluated first. Therefore, in this parenthesization the first one.  $A_3$  and  $A_4$  are multiplied first, the result is then multiplied with  $A_2$  and the result is then multiplied with  $A_1$ . In the next parenthesization the inner most parenthesis is the one that encloses the product  $A_2, A_3$  which is computed first. The result is then computed is the result is then multiplied with  $A_4$ , the result of this matrix multiplication is then multiplied with  $A_1$ .

The other parenthesis this is more interesting than the first two, because it is different observe that there are two inner most parenthesis; one parenthesis contains the matrix product  $A_1 A_2$ , and the second inner most parenthesis contains the matrix product  $A_3 A_4$ , and again using expression evaluation rules, the leftmost innermost parenthesis is evaluated first. Therefore, the expression involving the product  $A_1 A_2$  is evaluated first, then the expression involving the product  $A_3 A_4$  is evaluated, then the two results are multiplied and this is given us a result of the matrix product of the four matrices. The following two are symmetric to the second and first respectively, and the explanation is

similar. What distinguishes these 5 different parenthesisations? Let us just see that.

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**Number of Multiplications**

- To compute the number of scalar multiplications necessary, we must know:
  - Algorithm to multiply two matrices
  - Matrix dimensions

**Input:** Matrices  $A_{p \times q}$  and  $B_{q \times r}$  (with dimensions  $p \times q$  and  $q \times r$ )  
**Result:** Matrix  $C_{p \times r}$  resulting from the product  $A \cdot B$

Scalar multiplication in line 5 dominates time to compute  
Number of scalar multiplications =  $pqr$

What distinguishes them is the total number of multiplications. Our aim is to now count the total number of scalar multiplications which are necessary. To do this, let us understand the number of multiplications required to multiply two matrices, in this case let us assume that the matrix dimensions are given, the two matrices are A which is a p by q matrix, and B which is a q by r matrix. We know that the result is a p by r matrix, and let us call this result matrix, the value the matrix C. It is clear that the total number of matrix multiplication that need to be performed is p q r. It is clear the total number of matrix multiplications that need to be performed is p multiplied by q multiplied by r. How to this affect? The behavior of chain matrix multiplication.

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### Chain Multiplication Effort

- Example: Consider three matrices  $A_{10 \times 100}$ ,  $B_{100 \times 5}$ , and  $C_{5 \times 50}$
- There are 2 ways to parenthesize
  - $((AB)C) = D_{10 \times 5} \cdot C_{5 \times 50}$ 
    - $AB \Rightarrow 10 \cdot 100 \cdot 5 = 5,000$  scalar multiplications
    - $DC \Rightarrow 10 \cdot 5 \cdot 50 = 2,500$  scalar multiplications
  - $(A(BC)) = A_{10 \times 100} \cdot E_{100 \times 50}$ 
    - $BC \Rightarrow 100 \cdot 5 \cdot 50 = 25,000$  scalar multiplications
    - $AE \Rightarrow 10 \cdot 100 \cdot 50 = 50,000$  scalar multiplications

} Total:  
7,500

Total:  
75,000

For this let us consider the following example, where we consider three matrices; A B and C. A is a order 10 by 100, B is a order 100 by 5, and C is a order 5 by 50. Clearly the three matrices can be multiplied that is the product A B C can be computed. There are two ways of parenthesizing this, this is the first way where the product AB is computed first the result is multiplied with C. Let us assume that the product AB is called D, we know that it is a 10 by 5 matrix, and C is a 5 by 50 matrix the multiplication AB takes 5000 scalar multiplications.

The product DC takes 2500 scalar multiplications, therefore total number of scalar multiplications is 7500. Let us consider the second parenthesizations, where B and C are multiplied first followed by A. So, let the outcome of multiplying B and C be the matrix E which is the 100 by 50 matrix. So, the product B multiplied by C which performs first takes 25000 scalar multiplications. The sub sequence product of A and E takes 50000 scalar multiplications, and therefore we can already see that the first paranthasization uses only 7500 scalar multiplications, but the second parenthesization uses 10 times more number of scalar multiplications, that is it uses 75000 scalar multiplications. Clearly from a efficient C point of you, the first parenthesization is a more preferred parenthesization, then the second parenthesization.

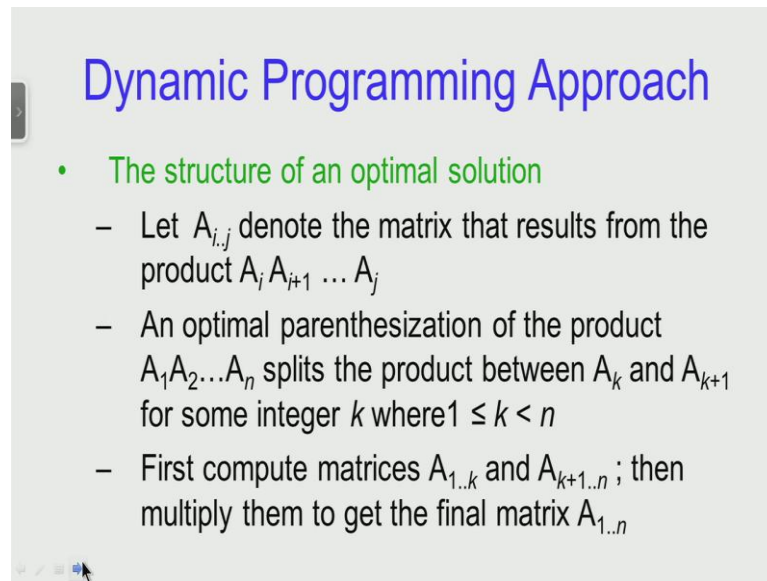
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## A Minimization Problem

- Matrix-chain multiplication problem
  - Given a chain  $A_1, A_2, \dots, A_n$  of  $n$  matrices, where for  $i=1, 2, \dots, n$ , matrix  $A_i$  has dimension  $p_{i-1} \times p_i$
  - Parenthesize the product  $A_1 A_2 \dots A_n$  such that the total number of scalar multiplications is minimized
- Brute force method of exhaustive search takes time exponential in  $n$

This gives rise to a very interesting minimization problem. This minimization problem is called the matrix chain multiplications problem. The input to the matrix chain multiplications problem is a chain of  $n$  matrices;  $A_1, A_2$  to  $A_n$ . For every  $i$  the  $i$ 'th matrix has dimension  $p_{i-1} \times p_i$ ; that is the matrix  $A_i$  has  $p_{i-1}$  rows and  $p_i$  columns. The first matrix  $A_1$  has  $p_0$  rows and  $p_1$  columns. The goal is to parenthesize this chain  $A_1, A_2$  to  $A_n$ . So, that the total number of scalar multiplications is minimized. Recall from the previous example that different parenthesization give rise to different number of scalar multiplications, and our aim is to choose the optimal scalar multiple optimal parenthesization to minimize the total number of scalar multiplications. One natural approach is the brute force method, where we try all possible parenthesizations, I leave this an exercise to the student to calculate how many parenthesization are there for a chain of  $M$  matrices. It is indeed an exponential and  $n$  the exact function is left as an exercise to the student.

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## Dynamic Programming Approach

- The structure of an optimal solution
  - Let  $A_{i..j}$  denote the matrix that results from the product  $A_i A_{i+1} \dots A_j$
  - An optimal parenthesization of the product  $A_1 A_2 \dots A_n$  splits the product between  $A_k$  and  $A_{k+1}$  for some integer  $k$  where  $1 \leq k < n$
  - First compute matrices  $A_{1..k}$  and  $A_{k+1..n}$ ; then multiply them to get the final matrix  $A_{1..n}$

So, now let us use a dynamic programming approach to come up with an algorithm to find the minimum parenthesization. Let us use the dynamic programming approach to come out with an algorithm which will come out with the parenthesization, that uses from the minimum number of scalar multiplications. To do this let us understand the structure of an optimal solution which in this case is a parenthesization. For this we need some simple notation, we use the notation  $A_{i..j}$  to denote the matrix which is a result of the product  $A_i A_{i+1}$  and so on upto  $A_j$ . Let us now observe that in an optimal parenthesization which we do not know which is whatever the algorithm is trying to compute. In an optimal parenthesization, let  $k$  be the index where the product  $A_1, A_2$  to  $A_n$  is split, therefore the approach for the computing the product would first be to compute the matrices  $A_{1..k}$  and  $A_{k+1..n}$ , and then compute the product of these two matrices to get the final matrices  $A_{1..n}$ .

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## Optimal Substructure

- **Key observation:** parenthesizations of the subchains  $A_1A_2\dots A_k$  and  $A_{k+1}A_{k+2}\dots A_n$  must also be optimal if the parenthesization of the chain  $A_1A_2\dots A_n$  is optimal
- That is, the optimal solution to the problem contains within it the optimal solution to subproblems

The key observation that we make about these whole exercise is that if we consider an optimal parenthesization of the change  $A_1, A_2$  to  $A_n$ , then the parenthesization of the sub change  $A_1, A_2$  to  $A_k$  and  $A_{k+1} A_{k+2}$  to  $A_n$  will also be optimal. This is the optimal sub structure, recall that from the previous lecture this is one of the properties of recall from the previous lecture for dynamic programming to be use the problem must have the optimal sub structure. In other words in this case the optimal solution to the parenthesization contains within it the optimal solution to sub problems.

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## Recursive Formulation

- Recursive definition of the value of an optimal solution
  - Let  $m[i, j]$  be the minimum number of scalar multiplications necessary to compute  $A_{i..j}$
  - Minimum cost to compute  $A_{1..n}$  is  $m[1, n]$
  - Suppose the optimal parenthesization of  $A_{i..j}$  splits the product between  $A_k$  and  $A_{k+1}$  for some integer  $k$  where  $i \leq k < j$

So, we will verify the claim that this problem has optimal sub structure while coming that with a recursive formulation of the optimum values. In this case we again introduce a few variables which are necessary for us to compute the minimum number of scalar multiplications. So, we use the two-dimensional array  $m[i, j]$  to denote the minimum number of scalar multiplications necessary to compute  $A_{i..j}$ . We let  $m[i, j]$  denote, let  $m[i, j]$  be the minimum number of scalar multiplications necessary to compute  $A_{i..j}$ . Now we can see with the minimum cost of compute the chain product  $A_1$  to  $A_n$ , recall this is  $A$  subscripted by the range 1 to  $n$  is the value  $m$  of 1 comma  $n$ . Suppose the optimal parenthesization of  $A_{i..j}$  splits the product between  $A_k$  and  $A_{k+1}$  where  $k$  is a number in the range  $i$  to  $j$ . Then we write down a recursive formulation of  $m$  of  $i$  comma  $j$ .



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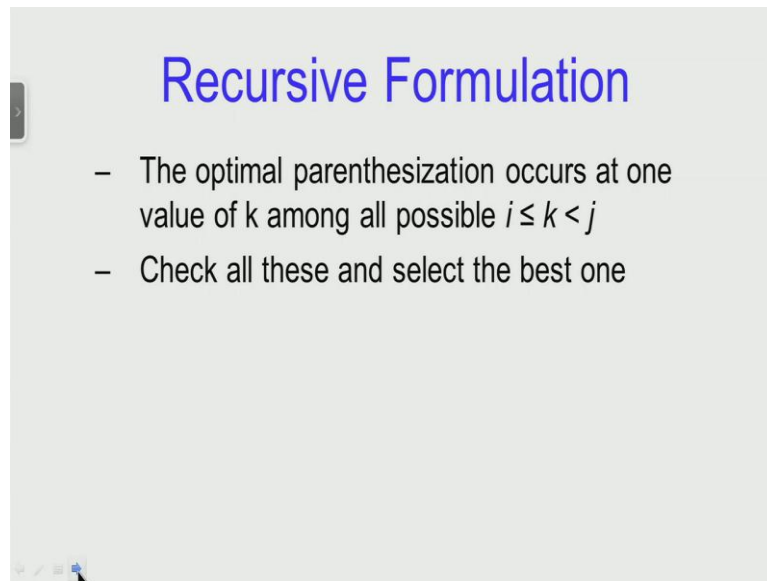
## Recursive Formulation

- $A_{i..j} = (A_i A_{i+1} \dots A_k) \cdot (A_{k+1} A_{k+2} \dots A_j) = A_{i..k} \cdot A_{k+1..j}$
- Cost of computing  $A_{i..j}$  = cost of computing  $A_{i..k}$  + cost of computing  $A_{k+1..j}$  + cost of multiplying  $A_{i..k}$  and  $A_{k+1..j}$
- Cost of multiplying  $A_{i..k}$  and  $A_{k+1..j}$  is  $p_{i-1} p_k p_j$
- $m[i, j] = m[i, k] + m[k+1, j] + p_{i-1} p_k p_j$  for  $i \leq k < j$
- $m[i, i] = 0$  for  $i=1, 2, \dots, n$

So, recursive formulation uses this parenthesization. The matrix  $A_{i..j}$  is obtained by the parenthesization, the matrix  $A_{i..j}$  is obtained by multiplying the matrix chain  $A_i$  to  $A_k$  with the result of the matrix chain  $A_{k+1}$  to  $A_j$ . In other words, this is the product of the two matrices;  $A_{i..k}$  multiplied by  $A_{k+1..j}$ . Therefore, the total cost of computing  $A_{i..j}$  is the cost of computing  $A_{i..k}$  plus the cost of computing  $A_{k+1..j}$  plus the cost of multiplying the two matrices  $A_{i..k}$  and  $A_{k+1..j}$ . Note here that the cost is the total number of scalar multiplications. So, we know that the third term, the cost of multiplying  $A_{i..k}$  and  $A_{k+1..j}$  is  $p_{i-1}$  multiplied by  $p_k$  multiplied by  $p_j$ , this is because the order of the two matrices are  $p_{i-1}$  cross  $p_k$  and  $p_k$  cross  $p_j$ .

So, we specify the recursive now completely, which says that the minimum number of scalar multiplications for the chain, the minimum number of scalar multiplications for multiplying the chain  $A_i$  to  $A_j$  is equal to  $m[i, k] + m[k+1, j] + p_{i-1} p_k p_j$  for  $k$  between  $i$  and  $j$ . And indeed the number of multiplications to compute an empty product is 0; that is  $m[i, i]$  is the cost of multiplying  $A_i$  where there are no multiplications operations involved, therefore this takes to be the value 0.

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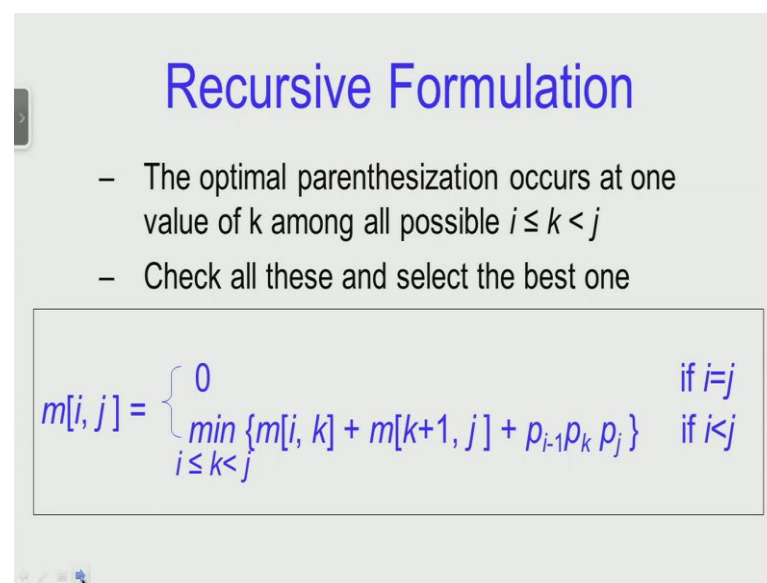


## Recursive Formulation

- The optimal parenthesization occurs at one value of  $k$  among all possible  $i \leq k < j$
- Check all these and select the best one

To complete the recursive of formulation, let us observe that we optimal parenthesization occurs at one of the values of  $k$  between  $i$  and  $j$ . We do not know which one it is, but the algorithmic idea is very simple, we check all the possible values of  $k$  between the range  $i$  and  $j$  and select the one that gives the least value.

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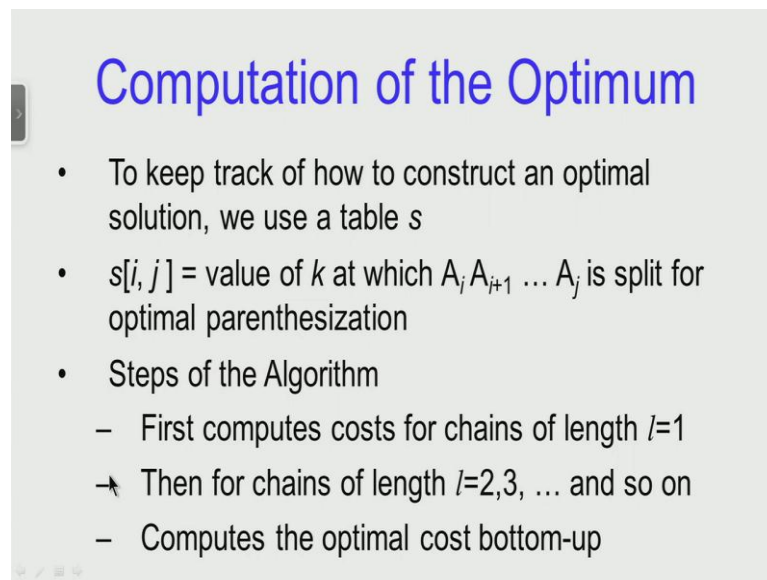
## Recursive Formulation

- The optimal parenthesization occurs at one value of  $k$  among all possible  $i \leq k < j$
- Check all these and select the best one

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

And this specifies completely the recursive formulation of  $m(i, j)$ . If  $i$  and  $j$  are the same, it is 0 because we do not have to perform any multiplication. If  $i$  is not equal to  $j$  and  $i$  is strictly smaller than  $j$  then  $m(i, j)$  we know stores the minimum number of scalar multiplications to multiply the chain product  $A_i$  to  $A_j$ . So, this is obtained by finding the best value of  $k$  by computing  $m(i, k) + m(k+1, j) + p_i \cdot p_{k+1} \cdot p_j$  and choosing the best possible  $k$  that gives the minimum value of  $m(i, j)$ .

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### Computation of the Optimum

- To keep track of how to construct an optimal solution, we use a table  $s$
- $s[i, j]$  = value of  $k$  at which  $A_i A_{k+1} \dots A_j$  is split for optimal parenthesization
- Steps of the Algorithm
  - First computes costs for chains of length  $l=1$
  - Then for chains of length  $l=2, 3, \dots$  and so on
  - Computes the optimal cost bottom-up

This completes the recursive formulation of the minimum that we are interested in. Now we need to convert this recursive formulation into an algorithm, and we have to specify the algorithm and efficient algorithm to compute the minimum. To do this we introduce a second two-dimensional array which we call  $S$ ;  $S$  stands for split and we refer to this two-dimensional array as the split table. The split table tells us where to split a chain. In other words  $S$  of  $i$  comma  $j$  is that value of  $k$  at which the chain  $A_i$  to  $A_j$  is split for an optimal parenthesization. The steps in this algorithm are to compute the minimum number of scalar multiplications for chains of length 1 from there we compute the minimum number of parenthesization for chains of length 2, and 3, and so on. This is the bottom of calculation method of the optimum value of  $m(1, n)$ .

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### Algorithm to Compute Optimal Cost

**Input:** Array  $p[0..n]$  containing matrix dimensions and  $n$   
**Result:** Minimum-cost table  $m$  and split table  $s$

```
for  $i \leftarrow 1$  to  $n$ 
   $m[i, i] \leftarrow 0$ 
for  $l \leftarrow 2$  to  $n$ 
  for  $i \leftarrow 1$  to  $n-l+1$ 
     $j \leftarrow i+l-1$ 
     $m[i, j] \leftarrow \infty$ 
    for  $k \leftarrow i$  to  $j-1$ 
       $q \leftarrow m[i, k] + m[k+1, j] + p[i-1] p[k] p[j]$ 
      if  $q < m[i, j]$ 
         $m[i, j] \leftarrow q$ 
         $s[i, j] \leftarrow k$ 
return  $m$  and  $s$ 
```

Takes  $O(n^3)$  time  
Requires  $O(n^2)$  space

This is the algorithm description. There is an initialization phase where the min cost table  $m$  is initialized with 0 for all the diagonal entries, because they do not involve any multiplication. This is followed by three nested iterations to essentially implement the recurrence, and the outer loop iterates over... So, let us consider this algorithmic description to compute the optimal cost. Using this data we will then compute the optimal parenthesization also. The input to these algorithms is an array which is  $n$  plus one element array which contains the dimensions of the matrices. For example, the first element  $p[0]$  and  $p[1]$  give us the information about the dimensions of matrix  $A_1$ , that is  $p[0]$  cross  $p[1]$ , the array entries  $p[1]$  and  $p[2]$  tell us the dimension of the matrix  $A_2$  and so on. The result of this algorithm is we get a min cost table and a split table; these are two arrays that we get. The min cost table stores the value of the minimum cost parenthesization of the chain multiplication involving the chains involving the chain of matrices  $A_i A_{i+1}$  upto  $A_j$ . Similarly the split table the entry  $S$  of  $i, j$  stores the value of the index  $k$  at which the chain  $A_i$  to  $A_j$  is to be split.

The algorithmic as follows it has 4 for loops. The first for loop is an initialization phase where the diagonal entries are all initialized to 0, this is natural because the diagonal entries store the value 0 to denote the fact that there is no matrix multiplication involving the single matrix. The remaining 3 for loops are nested and the intent of these for loops is

to fill the remaining entries in the upper half of the matrix is to fill the entries in the upper half with the matrix, and this is done by filling each diagonal. Observe that there are  $m - 1$  diagonals apart from the principle diagonal of the matrices. The outer loop iterates over the diagonals of the matrix, the outer for loop which is index by the variable  $l$ , iterates over the diagonals.

The next for loop is setup to instantiate each element in the appropriate diagonal and the innermost for loop is the one that evaluates the value of the min cost parenthesization. So, in the second for loop, the initialization of the variable  $j$  to  $i + l - 1$  is the choice of the appropriate element in the  $l$ 'th diagonal. So,  $m[i, j]$  is initialize to the value empty which is the standard think for the minimization problem which takes a positive values  $m[i, j]$  is initialized to the value infinity which is standard practice for minimization problems which takes positive values. The inner most for loop is the loop that implements the recurrence that we have return to formulate the value of  $m[i, j]$ . The way this is done is to iterate over all the possible values of  $k$  starting from  $i$  to  $j - 1$ , and the value  $q$  is computed has  $m[i, k] + m[k + 1, j] + p[i - 1, k] \times p[k, j]$ .

The if statement updates the value of  $m[i, j]$ , if the value of  $q$  is smaller and it also updates the value of the split entry, if the value of  $q$  is smaller than the current value of  $m[i, j]$ . At the end of this whole iteration the matrices  $m$  and  $s$  are computed, and this store the optimum parenthesization for every  $i, j$ , this store the optimum number of scalar multiplications for the chain multiplication involving  $A_i$  to  $A_j$  and the parenthesization information is stored by keeping track of a split value in the matrix  $S$ .

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## Constructing Optimal Solution

- Our algorithm computes the minimum-cost table  $m$  and the split table  $s$
- The optimal solution can be constructed from the split table  $s$ 
  - Each entry  $s[i, j]=k$  shows where to split the product  $A_i A_{i+1} \dots A_j$  for the minimum cost

The split table is used to compute an optimum solution, the algorithm computes first the minimum cost table and the split table  $S$  as we saw in the previous slide. And the optimum solution can be calculated from the split table using the value  $k$ , which is stored in  $S$  of  $i$  comma  $j$ . This enables us to compute  $S$  of  $1$  comma  $n$  recursively.

Thank you.