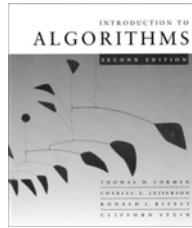




## CS 5633 -- Spring 2008



### Matrix-chain multiplication

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## Matrix-chain multiplication

**Given:** A sequence/chain of  $n$  matrices  $A_1, A_2, \dots, A_n$ , where  $A_i$  is a  $p_{i-1} \times p_i$  matrix

**Task:** Compute their product  $A_1 \cdot A_2 \cdot \dots \cdot A_n$  using the minimum number of scalar multiplications.

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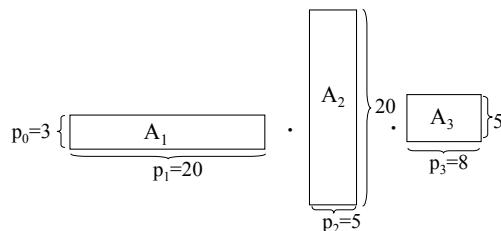
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### Matrix-chain multiplication example

**Example:**  $n=3$ ,  $p_0=3$ ,  $p_1=20$ ,  $p_2=5$ ,  $p_3=8$ .  $A_1$  is a  $3 \times 20$  matrix,  $A_2$  is a  $20 \times 5$  matrix,  $A_3$  is a  $5 \times 2$  matrix. Compute  $A_1 \cdot A_2 \cdot A_3$ .



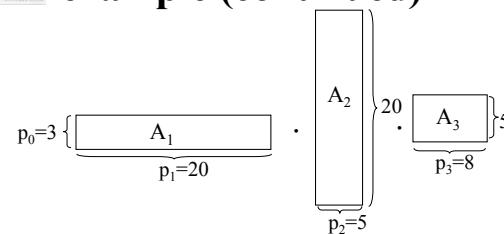
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### Matrix-chain multiplication example (continued)



- Computing  $A_1 \cdot A_2$  takes  $3 \cdot 20 \cdot 5$  multiplications and results in a  $3 \times 5$  matrix.
- Computing  $A_i \cdot A_{i+1}$  takes  $p_{i-1} \cdot p_i \cdot p_{i+1}$  multiplications and results in a  $p_{i-1} \times p_{i+1}$  matrix.

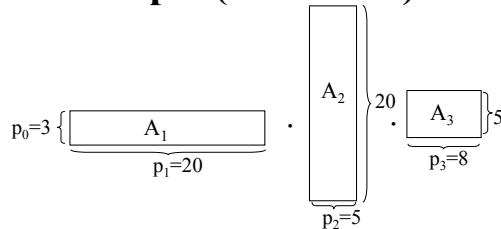
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## Matrix-chain multiplication example (continued)



- Computing  $(A_1 \cdot A_2) \cdot A_3$  takes  $3 \cdot 20 \cdot 5 + 3 \cdot 5 \cdot 8 = 300 + 120 = 420$  multiplications
- Computing  $A_1 \cdot (A_2 \cdot A_3)$  takes  $20 \cdot 5 \cdot 8 + 3 \cdot 20 \cdot 8 = 800 + 480 = 1280$  multiplications

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## Matrix-chain multiplication

**Given:** A sequence/chain of  $n$  matrices  $A_1, A_2, \dots, A_n$ , where  $A_i$  is a  $p_{i-1} \times p_i$  matrix

**Task:** Compute their product  $A_1 \cdot A_2 \cdot \dots \cdot A_n$  using the minimum number of scalar multiplications.

⇒ Find a parenthesization that minimizes the number of multiplications

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## Would greedy work?

1. Parenthesizing like this  $\dots((A_1 \cdot A_2) \cdot A_3) \dots \cdot A_n$  does not work (e.g., reverse our running example).

2. Recursively parenthesize like this:

$$\underbrace{(A_1 \cdot \dots \cdot A_k)}_{p_0 \times p_k} \cdot \underbrace{(A_{k+1} \cdot \dots \cdot A_n)}_{p_k \times p_n}$$

Find the  $k$  that minimizes  $p_0 \cdot p_k \cdot p_n$ .

Does not work either (example:  $p_0=1, p_1=2, p_2=3, p_3=4$ )

⇒ Try dynamic programming

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## 1) Optimal substructure

Let  $A_{i,j} = A_i \cdot \dots \cdot A_j$  for  $i \leq j$

- Consider an optimal parenthesization for  $A_{i,j}$ . Assume it splits it at  $k$ , so  $A_{i,j} = (A_i \cdot \dots \cdot A_k) \cdot (A_{k+1} \cdot \dots \cdot A_j)$
- Then, the par. of the prefix  $A_i \cdot \dots \cdot A_k$  within the optimal par. of  $A_{i,j}$  must be an optimal par. of  $A_{i,k}$ . (Assume it is not optimal, then there exists a better par. for  $A_{i,k}$ . **Copy and paste** this par. into the par. for  $A_{i,j}$ . This yields a better par. for  $A_{i,j}$ . Contradiction.)

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## 2) Recursive solution

- a) First compute the minimum number of multiplications
- b) Then compute the actual parenthesization

We will concentrate on solving a) now.

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## 2) Recursive solution (cont.)

$m[i,j]$  = minimum number of scalar multiplications to compute  $A_{ij}$

Goal: Compute  $m[1,n]$

$$A_{i,j} = \underbrace{(A_i \cdot \dots \cdot A_k)}_{p_{i-1} \times p_k} \cdot \underbrace{(A_{k+1} \cdot \dots \cdot A_j)}_{p_k \times p_j}$$

Recurrence:

- $m[i,i] = 0$  for  $i=1,2,\dots,n$
- $m[i,j] = \min_{i \leq k < j} (m[i,k]+m[k+1,j] + p_{i-1}p_kp_j)$

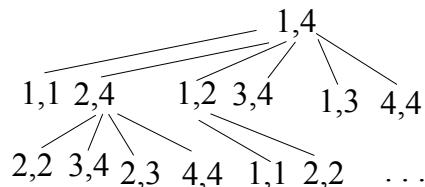
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## Recursion tree



- The runtime of the straight-forward recursive algorithm is  $\Omega(2^n)$
- But only  $\Theta(n^2)$  different subproblems !

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## Dynamic programming

```

MATRIX_CHAIN_DP(p, n):
  for i:=1 to n do m[i,i]=0
  for l:=2 to n do // l is length of chain
    for i:=1 to n-l+1 do
      j:=i+l-1
      m[i,j]=∞
      for k:=i to j-1 do
        q:=m[i,k]+m[k+1,j]+p_{i-1}*p_k*p_j
        if q < m[i,j] then
          m[i,j]:=q
          s[i,j]:=k //index that optimizes m[i,j]
  return m and s;
  
```

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## Dynamic programming

- Use dynamic programming to fill the 2-dimensional  $m[i,j]$ -table
- Bottom-up: Diagonal by diagonal
- For the construction of the optimal parenthesization, use an additional array  $s[i,j]$  that records that value of  $k$  for which the minimum is attained and stored in  $m[i,j]$
- $O(n^3)$  runtime ( $n \times n$  table,  $O(n)$  min-computation per entry),  $O(n^2)$  space
- $m[1,n]$  is the desired value

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## Construction of an optimal parenthesization

```
PRINT_PARENS(s,i,j) // initial call: print_parens(s,1,n)
  if i=j then print "A"i
  else print "("
    PRINT_PARENS(s,i,s[i,j])
    PRINT_PARENS(s,s[i,j]+1,j)
    print ")"
```

Runtime: Recursion tree = binary tree with  $n$  leaves. Spend  $O(1)$  per node.  $O(n)$  total runtime.

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