

CS 362, Lecture 7

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Problem:

- We are given a sequence of n matrices, A_1, A_2, \dots, A_n , where for $i = 1, 2, \dots, n$, matrix A_i has dimension p_{i-1} by p_i
- We want to compute the product, $A_1 A_2, \dots, A_n$ as quickly as possible.
- In particular, we want to fully *parenthesize* the expression above so there are no ambiguities about the how the matrices are multiplied
- A product of matrices is *fully parenthesized* if it is either a single matrix, or the product of two fully parenthesized matrix products, surrounded by parentheses

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Today's Outline

- Matrix Multiplication

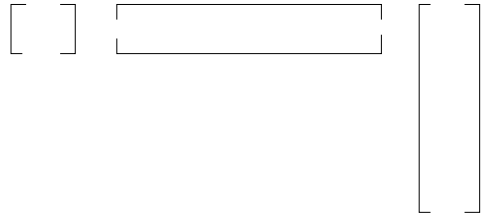
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Parthesizing Matrices

- There are many ways to parenthesize the matrices
- Each way gives the same output (because of associativity of matrix multiplications)
- However the way we parenthesize will effect the *time* to compute the output
- Our Goal: Find a parenthesization which requires the minimal number of scalar multiplications

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Example



- In this example, it's much better to multiply the last two matrices first (this gives us a short, narrow matrix on the right)
- Worse to multiply the first two matrices first (this gives us a short wide matrix on the left)
- In general, our goal is to find ways to always create narrow and short resulting matrices.

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A Problem

- Let $P(n)$ be the number of ways to parenthesize n matrices. Then $P(1) = 1$
- For $n \geq 2$, we know that a fully parenthesized product is the product of two fully parenthesized products, and the split can occur anywhere from $k = 1$ to $k = n - 1$.
- Hence for $n \geq 2$:

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

- In the hw, you will show that the solution to this recurrence is $\Omega(2^n)$

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A Problem

Problem: There can be many ways to parenthesize. E.g.

- $(A_1(A_2(A_3A_4)))$
- $(A_1((A_2A_3)A_4))$
- $((A_1A_2)(A_3A_4))$
- $((A_1(A_2A_3))A_4)$
- $((A_1A_2)(A_3)A_4)$

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The Pattern

Q: Can we develop a DP Solution to this problem?

- **Formulate the problem recursively.** Write down a formula for the whole problem as a simple combination of answers to smaller subproblems
- **Build solutions to your recurrence from the bottom up.** Write an algorithm that starts with the base cases of your recurrence and works its way up to the final solution by considering the intermediate subproblems in the correct order.

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Key Observation

- Let $A_{i..j}$ (for $i \leq j$) be the matrix that results from evaluating the product $A_i A_{i+1} \dots A_j$
- Imagine we are computing $A_{i..j}$
- The last multiplication we do must look like this:

$$A_{i..j} = (A_{i..k}) * (A_{k+1..j})$$

for some k between i and $j - 1$

- Then total cost to compute $A_{i..j}$ is:

cost to compute $A_{i..k}$ +
cost to compute $A_{k+1..j}$ +
cost to multiply $A_{i..k}$ and $A_{k+1..j}$

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Cost to Multiply

- $A_{i..k}$ is a p_{i-1} by p_k matrix
- $A_{k+1..j}$ is a p_k by p_j matrix
- Thus multiplying $A_{i..k}$ and $A_{k+1..j}$ takes $p_{i-1}p_k p_j$ operations
- Hence we have:

$$m(i, j) \leq m(i, k) + m(k + 1, j) + p_{i-1}p_k p_j$$

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Recursive Formulation

- For any integers x, y , let $m(x, y)$ be the minimum cost of computing $A_{x..y}$
- Then for any k between i and $j - 1$,
 $m(i, j) \leq$ optimal cost to compute $A_{i..k}$ +
optimal cost to compute $A_{k+1..j}$ +
cost to multiply $A_{i..k}$ and $A_{k+1..j}$
- In other words:

$$m(i, j) \leq m(i, k) + m(k + 1, j) + \text{cost to multiply } A_{i..k} \text{ and } A_{k+1..j}$$

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Recursive Formulation

- We've shown that $m(i, j) \leq m(i, k) + m(k + 1, j) + p_{i-1}p_k p_j$ for any $k = i, i + 1, \dots, j - 1$
- Further note that the optimal parenthesization must use some value of $k = i, i + 1, \dots, j - 1$. So we need only pick the best
- Thus we have:

$$m(i, j) = 0 \text{ if } i = j$$
$$m(i, j) = \min_{i \leq k < j} \{m(i, k) + m(k + 1, j) + p_{i-1}p_k p_j\}$$

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The Recursive Algorithm

- We now have enough information to write a recursive function to solve the problem
- The recursive solution will have runtime given by the following recurrence:
- $T(1) = 1$,
- $T(n) = 1 + \sum_{k=1}^{n-1} (T(k) + T(n-k) + 1)$
- Unfortunately, the solution to this recurrence is $\Omega(2^n)$ (as shown on p. 346 of the text)

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Pseudocode

```
Matrix-Chain-Order(int p[]){
    n = p.length - 1;
    for (i=1;i<=n;i++){
        m(i,i) = 0;
    }
    for (l=2;l<=n;l++){ \\l is chain length
        for (i=1;i<=n-l+1;i++){
            j = i+l-1;
            m[i,j] = MAXINT;
            for(k=i;k<=j-1;k++){
                q = m[i,k] + m[k+1,j] + p[i-1]*p[k]*p[j];
                if(q<m[i,j]){
                    m[i,j] = q;
                    s[i,j] = k;
                }
            }
        }
    }
}
```

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DP Algorithm

- Note that we must solve one subproblem for each choice of i and j satisfying $1 \leq i \leq j \leq n$
- This is only $\binom{n}{2} + n = \Theta(n^2)$ subproblems
- The recursive algorithm encounters each subproblem many times in the branches of the recursion tree.
- However, we can just compute these subproblems from the bottom up, storing the results in a table (this is the DP solution)

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Pseudocode

- This code computes both the optimal cost and a parenthesization that achieves that cost
- It uses an m array to store the optimal costs of computing $m(i, j)$. It also uses a s array, where $s(i, j)$ stores the k value which gives $m(i, j)$
- The parenthesization can be recovered from the s array using the pseudocode in the book on p. 338.

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Analysis

- This code has three nested loops, each of which takes on at most $n - 1$ values, and the inner loop takes $O(1)$ time.
- Thus the runtime is $O(n^3)$
- The algorithm also requires $\Theta(n^2)$ space

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Example

- Consider the sequence of three matrices, A_1, A_2, A_3 whose dimensions are given by the sequence 3, 1, 2, 1 (i.e. $p_0 = 3, p_1 = 1, p_2 = 2, p_3 = 1$)
- Let's construct the tables giving the optimal parenthesization
- The (i, j) entry of the first table will give the optimal cost for computing $A_{i..j}$, the (i, j) entry of the second table will give a k value which achieves this optimal cost

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Computations

- $m(1, 1) = m(2, 2) = m(3, 3) = 0$
- $m(1, 2) = p_0 p_1 p_2 = 6$
- $m(2, 3) = p_1 p_2 p_3 = 2$

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Computations

$$\begin{aligned} m(1, 3) &= \min \left\{ \begin{array}{l} m(1, 1) + m(2, 3) + p_0 p_1 p_3, \\ m(1, 2) + m(3, 3) + p_0 p_2 p_3 \end{array} \right\} \\ &= \min \left\{ \begin{array}{l} 0 + 2 + 3, \\ 6 + 0 + 6 \end{array} \right\} \\ &= 5 \end{aligned}$$

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Example, m array

	1	2	3
1	0	6	5
2	-	0	2
3	-	-	0

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Example, s array

	1	2	3
1	-	1	1
2	-	-	2
3	-	-	-

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Example

- Thus an optimal parenthesization is $(A_1(A_2A_3))$
- The cost of this is 5

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Example II

- Consider the sequence of three matrices, A_1, A_2, A_3, A_4 whose dimensions are given by the sequence 3, 1, 2, 1, 2 (i.e. $p_0 = 3, p_1 = 1, p_2 = 2, p_3 = 1, p_4 = 2$)
- Let's construct the tables giving the optimal parenthesization
- The (i, j) entry of the first table will give the optimal cost for computing $A_{i..j}$, the (i, j) entry of the second table will give a k value which achieves this optimal cost

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Example II, m array

	1	2	3	4
1	0	6	5	10
2	-	0	2	4
3	-	-	0	4
4	-	-	-	0

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Example II, s array

	1	2	3	4
1	-	1	1	1
2	-	-	2	3
3	-	-	-	3
4	-	-	-	-

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Example Computation

$$\begin{aligned}
 m(1,4) &= \min \left\{ \begin{array}{l} m(1,1) + m(2,4) + p_0 p_1 p_4, \\ m(1,2) + m(3,4) + p_0 p_2 p_4, \\ m(1,3) + m(4,4) + p_0 p_3 p_4 \end{array} \right\} \\
 &= \min \left\{ \begin{array}{l} 0 + 4 + 6, \\ 6 + 4 + 12, \\ 5 + 0 + 6 \end{array} \right\} \\
 &= 10
 \end{aligned}$$

This minimum is achieved when $k = 1$

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Example II

- Thus an optimal parenthesization is $(A_1((A_2 A_3) A_4))$
- The cost of this is 10

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In-Class Exercise

- Consider the sequence of three matrices, A_1, A_2, A_3 whose dimensions are given by the sequence 1, 2, 1, 2 (i.e. $p_0 = 1, p_1 = 2, p_2 = 1, p_3 = 2$)
- Q1: What are the m array and s array for these inputs?
- Q2: What is the optimal parenthesization?