CS4102 Algorithms Spring 2020 - Horton's Slides

Dynamic Programming, Part Deux

- Our mid-term is coming!
- Spring Break's also coming!
- You'll make it!

 Hang in there!!!

Midterm

- Wednesday, March 4 in class
 - SDAC: Please schedule with SDAC for Wednesday
 - Mostly in-class with a (required) take-home portion
 - Take-home "bonus" If you do better on take-home than on its "starter" question on the in-class, you can earn back half the difference.
- Practice Midterm and Solutions on Collab
- Review Session on Panopto
- More office hours by me! See Piazza

Today's Keywords

- Dynamic Programming
- Longest Common Subsequence
- Seam Carving

CLRS Readings

Chapter 15

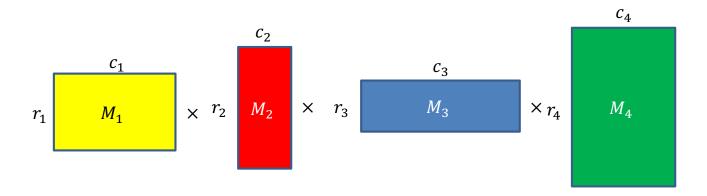
- Section 15.1, Log/Rod cutting, optimal substructure property
 - Note: r_i in book is called Cut() or C[] in our slides. We use their example.
- Section 15.3, More on elements of DP, including optimal substructure property
- -Section 15.2, matrix-chain multiplication
- -Section 15.4, longest common subsequence

Dynamic Programming

- Requires Optimal Substructure
 - Solution to larger problem contains the solutions to smaller ones
- Avoid extra work due to overlapping subproblems
- Idea:
 - 1. Identify the recursive structure of the problem
 - What is the "last thing" done?
 - 2. Save the solution to each subproblem in memory
 - 3. Select a good order for solving subproblems
 - "Top Down": Solve each recursively
 - "Bottom Up": Iteratively solve smallest to largest

Matrix Chaining

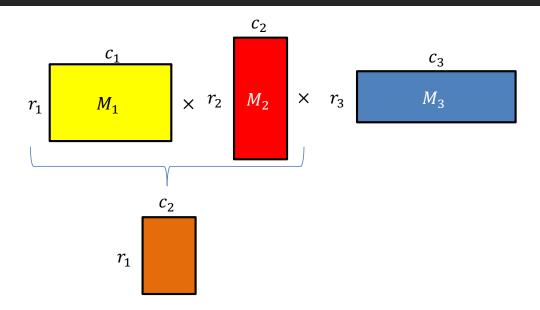
• Given a sequence of Matrices $(M_1, ..., M_n)$, what is the most efficient way to multiply them?



Order Matters!

$$c_1 = r_2$$

$$c_2 = r_3$$

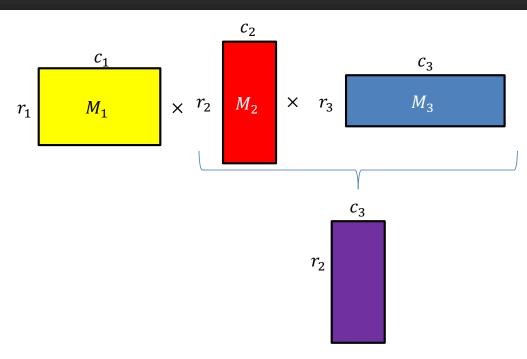


- $(\underline{M_1} \times \underline{M_2}) \times \underline{M_3}$
 - uses $(c_1 \cdot r_1 \cdot c_2) + c_2 \cdot r_1 \cdot c_3$ operations

Order Matters!

$$c_1 = r_2$$

$$c_2 = r_3$$



- $M_1 \times (M_2 \times M_3)$
 - uses $c_1 \cdot r_1 \cdot c_3 + (c_2 \cdot r_2 \cdot c_3)$ operations

Order Matters!

$$c_1 = r_2$$

$$c_2 = r_3$$

- $(M_1 \times M_2) \times M_3$
 - uses $(c_1 \cdot r_1 \cdot c_2) + c_2 \cdot r_1 \cdot c_3$ operations
 - $-(10 \cdot 7 \cdot 20) + 20 \cdot 7 \cdot 8 = 2520$
- $M_1 \times (M_2 \times M_3)$
 - uses $c_1 \cdot r_1 \cdot c_3 + (c_2 \cdot r_2 \cdot c_3)$ operations
 - $-10 \cdot 7 \cdot 8 + (20 \cdot 10 \cdot 8) = 2160$

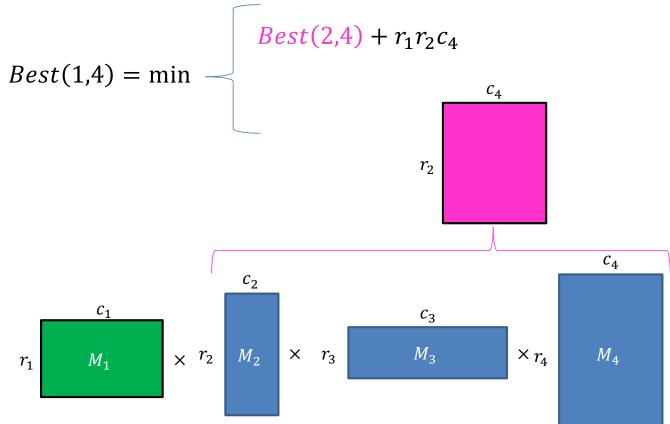
$$M_1 = 7 \times 10$$

 $M_2 = 10 \times 20$
 $M_3 = 20 \times 8$
 $c_1 = 10$
 $c_2 = 20$
 $c_3 = 8$
 $r_1 = 7$
 $r_2 = 10$
 $r_3 = 20$

Dynamic Programming

- Requires Optimal Substructure
 - Solution to larger problem contains the solutions to smaller ones
- Avoid extra work due to overlapping subproblems
- Idea:
 - 1. Identify the recursive structure of the problem
 - What is the "last thing" done?
 - 2. Save the solution to each subproblem in memory
 - 3. Select a good order for solving subproblems
 - "Top Down": Solve each recursively
 - "Bottom Up": Iteratively solve smallest to largest

 $Best(1, n) = \text{cheapest way to multiply together } M_1 \text{ through } M_n$



 $Best(1, n) = \text{cheapest way to multiply together } M_1 \text{ through } M_n$

$$Best(1,4) = \min \begin{array}{c|c} Best(2,4) + r_1 r_2 c_4 \\ Best(1,2) + Best(3,4) + r_1 r_3 c_4 \end{array}$$

$$\begin{array}{c|c} c_2 \\ r_1 \\ \hline \end{array}$$

$$\begin{array}{c|c} c_2 \\ \hline \end{array}$$

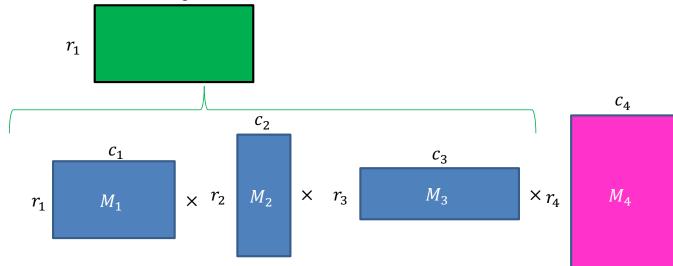
$$\begin{array}{c|c} c_4 \\ \hline \end{array}$$

$$\begin{array}{c|c} c_3 \\ \hline \end{array}$$

$$\begin{array}{c|c} c_4 \\ \hline \end{array}$$

 $Best(1, n) = \text{cheapest way to multiply together } M_1 \text{ through } M_n$

$$Best(1,4) = \min \begin{cases} Best(2,4) + r_1r_2c_4 \\ Best(1,2) + Best(3,4) + r_1r_3c_4 \\ Best(1,3) + r_1r_4c_4 \end{cases}$$



In general:

```
Best(i,j) = \text{cheapest way to multiply together } M_i \text{ through } M_j
Best(i,j) = \min_{k=i}^{j-1} \left(Best(i,k) + Best(k+1,j) + r_i r_{k+1} c_j\right)
Best(i,i) = 0
Best(2,n) + r_1 r_2 c_n
Best(1,2) + Best(3,n) + r_1 r_3 c_n
Best(1,3) + Best(4,n) + r_1 r_4 c_n
Best(1,4) + Best(5,n) + r_1 r_5 c_n
...
Best(1,n-1) + r_1 r_n c_n
```

Dynamic Programming

- Requires Optimal Substructure
 - Solution to larger problem contains the solutions to smaller ones
- Avoid extra work due to overlapping subproblems
- Idea:
 - 1. Identify the recursive structure of the problem
 - What is the "last thing" done?
 - 2. Save the solution to each subproblem in memory
 - 3. Select a good order for solving subproblems
 - "Top Down": Solve each recursively
 - "Bottom Up": Iteratively solve smallest to largest

2. Save Subsolutions in Memory

In general:

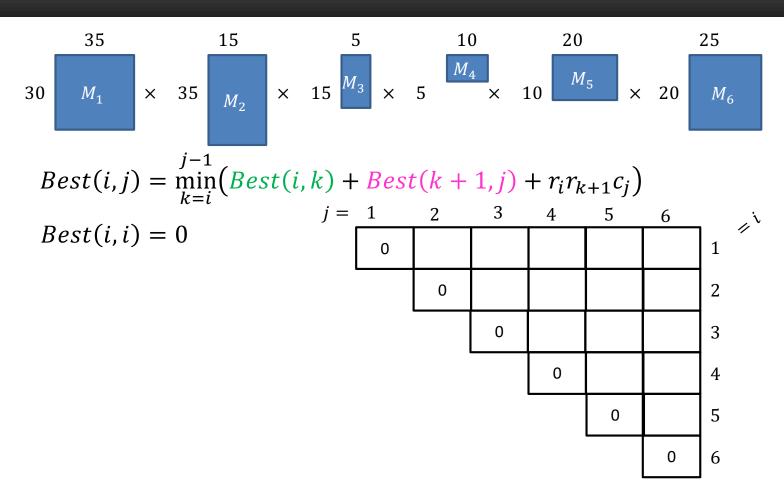
```
Best(i,j) = \text{cheapest way to multiply together } M_i \text{ through } M_j
Best(i,j) = \min_{k=i}^{j-1} \left( Best(i,k) + Best(k+1,j) + r_i r_{k+1} c_j \right)
Best(i,i) = 0
Best(i,i) = 0
Best(2,n) + r_1 r_2 c_n
Best(1,2) + Best(3,n) + r_1 r_3 c_n
Best(1,3) + Best(4,n) + r_1 r_4 c_n
Best(1,4) + Best(5,n) + r_1 r_5 c_n
Best(1,n-1) + r_1 r_n c_n
```

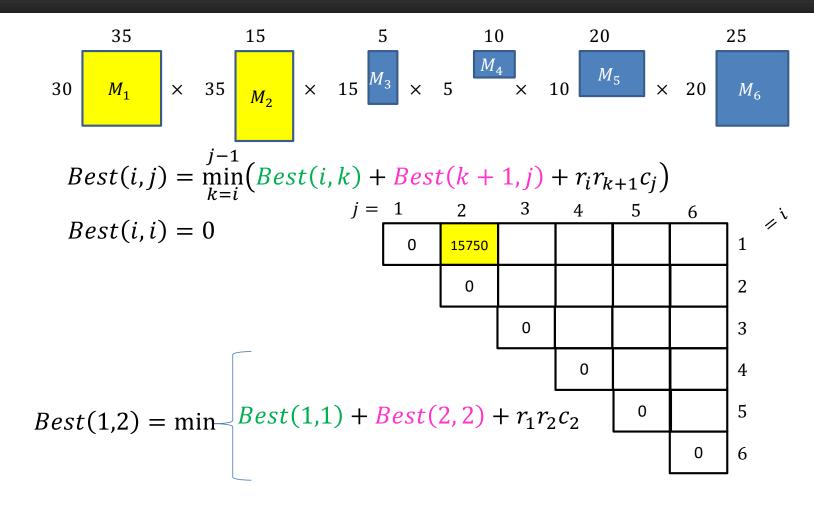
Dynamic Programming

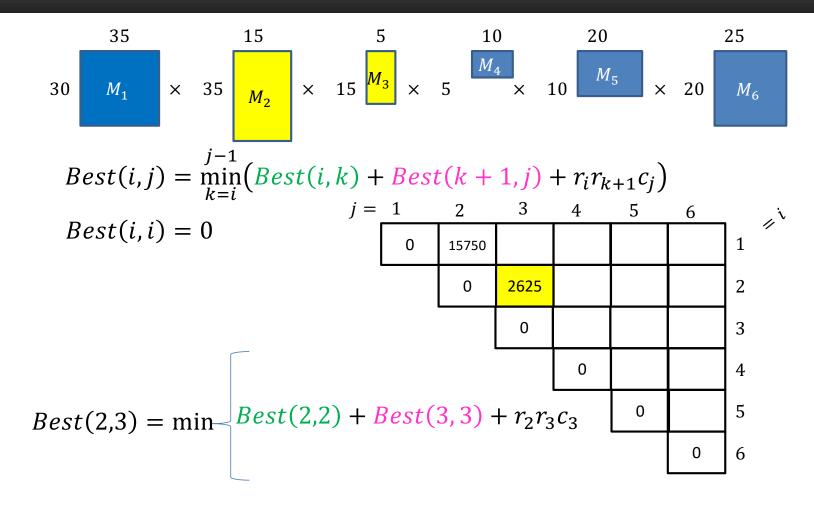
- Requires Optimal Substructure
 - Solution to larger problem contains the solutions to smaller ones
- Avoid extra work due to overlapping subproblems
- Idea:
 - 1. Identify the recursive structure of the problem
 - What is the "last thing" done?
 - 2. Save the solution to each subproblem in memory
 - 3. Select a good order for solving subproblems
 - "Top Down": Solve each recursively
 - "Bottom Up": Iteratively solve smallest to largest

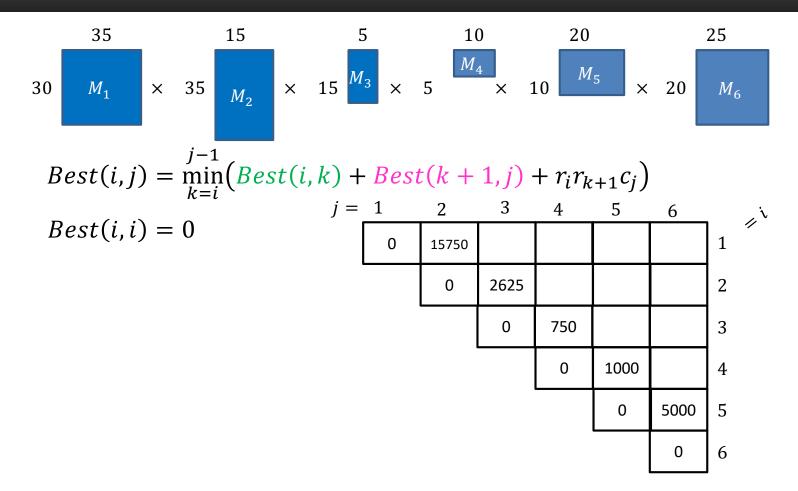
In general:

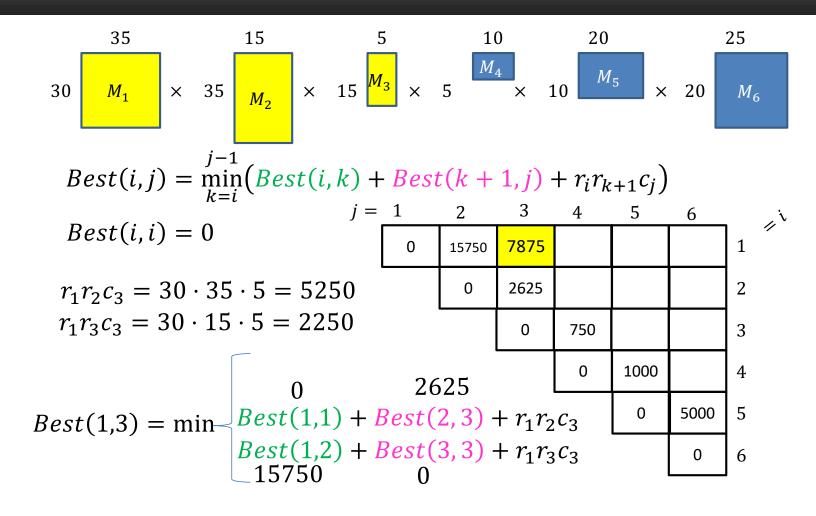
```
Best(i,j) = \text{cheapest way to multiply together } M_i \text{ through } M_j
Best(i,j) = \min_{k=i}^{j-1} \left( Best(i,k) + Best(k+1,j) + r_i r_{k+1} c_j \right)
Best(i,i) = 0
Best(i,i) = 0
Best(2,n) + r_1 r_2 c_n
Best(1,2) + Best(3,n) + r_1 r_3 c_n
Best(1,3) + Best(4,n) + r_1 r_4 c_n
Best(1,4) + Best(5,n) + r_1 r_5 c_n
Best(1,n-1) + r_1 r_n c_n
```

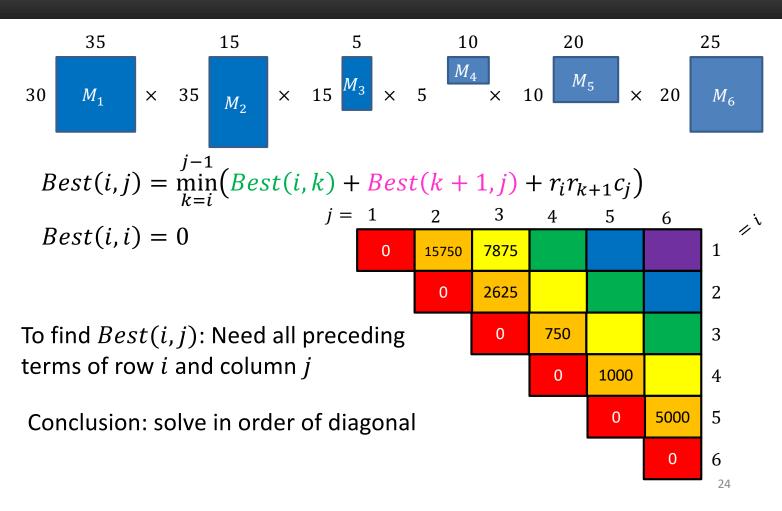












Matrix Chaining

$$Best(i,j) = \min_{k=i}^{j-1} \left(Best(i,k) + Best(k+1,j) + r_i r_{k+1} c_j\right)$$

$$Best(i,i) = 0$$

$$\int_{k=i}^{j-1} \left(Best(i,k) + Best(k+1,j) + r_i r_{k+1} c_j\right)$$

$$\int_{k=i}^{j-1} \left(Best(i,k) + Best(i,k) + Best$$

Run Time

- Initialize Best[i, i] to be all 0s $\Theta(n^2)$ cells in the Array
- 2. Starting at the main diagonal, working to the upper-right, fill in each cell using:
 - 1. Best[i, i] = 0

Each "call" to Best() is a

1.
$$Best[i,i] = 0$$

$$\Theta(n) \text{ options for each cell } O(1) \text{ memory lookup}$$
2. $Best[i,j] = \min_{k=i}^{j-1} \left(Best(i,k) + Best(k+1,j) + r_i r_{k+1} c_j \right)$

Backtrack to find the best order

"Remember" which choice of k was the minimum at each cell. Intuitively this was the best place to "split" for that range (i,j).

$$Best(i,j) = \min_{k=i}^{j-1} \left(Best(i,k) + Best(k+1,j) + r_i r_{k+1} c_j\right)$$

$$j = 1 \quad 2 \quad 3 \quad 4 \quad 5 \quad 6$$

$$0 \quad 15750 \quad 7875_1 \quad 9375 \quad 11875 \quad 15125_3 \quad 1$$

$$0 \quad 2625 \quad 4375 \quad 7125 \quad 10500 \quad 2$$

$$0 \quad 750 \quad 2500 \quad 5375 \quad 3$$

$$Best(1,1) + Best(2,6) + r_1 r_2 c_6 \quad 0 \quad 1000 \quad 3500_5 \quad 4$$

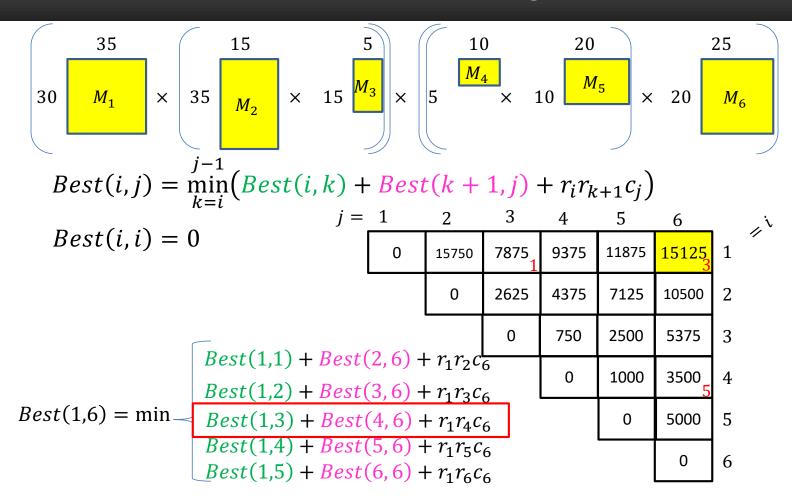
$$Best(1,2) + Best(3,6) + r_1 r_3 c_6 \quad 0 \quad 5000 \quad 5$$

$$Best(1,3) + Best(4,6) + r_1 r_4 c_6 \quad 0 \quad 5000 \quad 5$$

$$Best(1,4) + Best(5,6) + r_1 r_5 c_6 \quad 0 \quad 6$$

$$Best(1,5) + Best(6,6) + r_1 r_6 c_6 \quad 0 \quad 6$$

Matrix Chaining



Storing and Recovering Optimal Solution

- Maintain table Choice[i,j] in addition to Best table
 - Choice[i,j] = k means the best "split" was right after M_k
 - Work backwards from value for whole problem, Choice[1,n]
 - Note: Choice[i,i+1] = i because there are just 2 matrices
- From our example:
 - Choice[1,6] = 3. So $[M_1 M_2 M_3] [M_4 M_5 M_6]$
 - We then need Choice[1,3] = 1. So $[(M_1) (M_2 M_3)]$
 - Also need Choice[4,6] = 5. So $[(M_4 M_5) M_6]$
 - Overall: $[(M_1) (M_2 M_3)] [(M_4 M_5) M_6]$

Dynamic Programming

- Requires Optimal Substructure
 - Solution to larger problem contains the solutions to smaller ones
- Avoid extra work due to overlapping subproblems
- Idea:
 - 1. Identify the recursive structure of the problem
 - What is the "last thing" done?
 - 2. Save the solution to each subproblem in memory
 - 3. Select a good order for solving subproblems
 - "Top Down": Solve each recursively
 - "Bottom Up": Iteratively solve smallest to largest

• Slides on Longest Common Subsequence (LCS) problem moved to next slide set.