

LECTURE 10

Friday January 27, 2017

*based on notes by Denis Pankratov*Dynamic Programming (*continue*)

Chain Matrix Multiplication

Define: cost of multiplyling two matrices A_1 of dimension $n \times p$ and A_2 of dimension $p \times m$ as npm .**Remember:** Matrix multiplication is associative!

$$(A \times B) \times C = A \times (B \times C)$$

Overall cost of matrix multiplication (of n matrices) depends on parenthesization.Example: Given the following matrices w/ their respective... dimensions

Matrix name	Dimension
A	50×20
B	20×1
C	1×10
D	10×100

We have to come up with a way to optimize the multiplication of these 4 matrices. The key is working with the values 50, 20, 1, 10, 100 in the most efficient way.

Keep in mind that:

Parenthesis	Overall Cost
$A \times ((B \times C) \times D)$	$20 \times 1 \times 10 + 20 \times 10 \times 100 + 50 \times 20 \times 100 = 120,200$
$(A \times (B \times C)) \times D$	$20 \times 1 \times 10 + 50 \times 20 \times 10 + 50 \times 10 \times 100 = 60,200$
$(A \times B) \times (C \times D)$	$50 \times 20 \times 1 + 1 \times 10 \times 100 + 50 \times 1 \times 100 = 7,000$

Input: D - array of $n + 1$ positive integers (0-based indexing), which represents: $D[0] \times D[1]$ – dimension of A_1 $D[1] \times D[2]$ – dimension of A_2 \vdots

$$D[n-1] \times D[n] - \text{dimension of } A_n$$

Output: Optimal (*as in, with minimum overall cost*) parenthesization.

Rough Approach: Assume that we already have an optimal solution of the form

$$(A_1 \times A_2 \times \dots \times A_k) \times (A_{k+1} \times \dots \times A_n)$$

So that the first interval has a dimension of $D[0] \times D[k]$, while the second interval has a dimension of $D[k] \times D[n]$. k is used to split the left from the right part of our multiplication.

Semantic array:

$C[i, j]$ = minimum overall cost to multiply matrices $A_i \times \dots \times A_j$, $D[i-1], D[i], \dots, D[j]$.

Computational array:

$i < j, C[i, j] = \min_{i \leq k < j} \{C[i, k] + C[k+1, j] + D[i-1] \cdot D[k] \cdot D[j]\}$.

Base case: $C[i, i] = 0$

Correctness (equivalence):

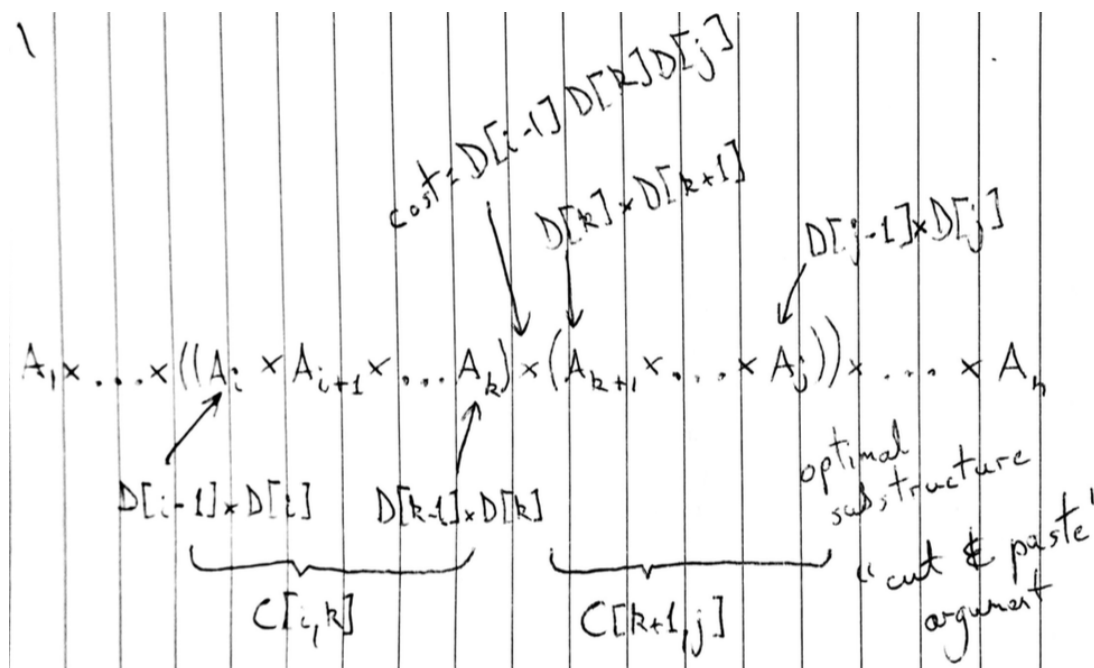


Figure 1: Rough visual representation of correctness

Algorithm: Pseudocode to compute the *value* of optimal

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1 def CMM(D, n):
2   initialize C array of size n * n (1-based indexing)

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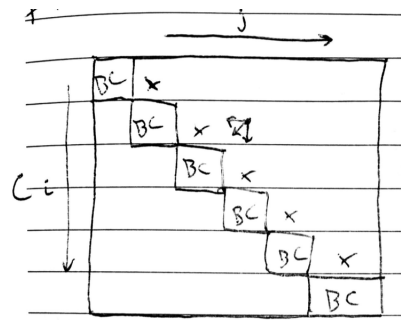


Figure 2: Visual representation of optimal path: fill in the entries by increasing $j - i$

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3
4   for i = 1 to n:
5       C[i, i] = 0
6   for l = 1 to n - 1:
7       for i = 1 to n - l:
8           j = i + l
9           C[i, j] = float('inf')
10          for k = i to j - 1:
11              C[i, j] = min(C[i, j],
12                           C[i, k] + C[k + 1, j] + D[i - 1] * D[k] * D[j])
13   return C[1, n]
```

Exercise: Add book-keeping to compute an optimal parenthesization

Max Independent Set in Trees

Definition: Let $T = (V, E)$ be a tree $S \subseteq V$ is an *independent* set if there are no edges in T between vertices in S . (see Figure 3 below).

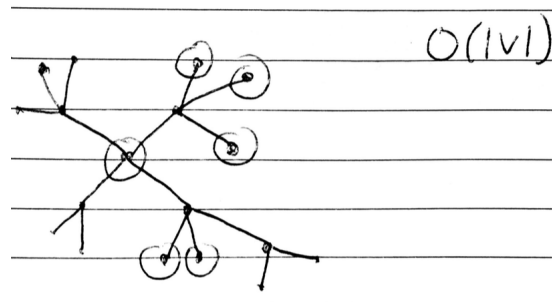


Figure 3: Example of $T = (V, E)$ with target runtime of $\mathcal{O}(|V|)$

Input: $T = (V, E)$ - tree (rooted), r - root

Output: S - max. independent set

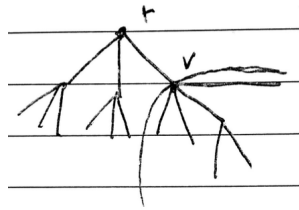


Figure 4: Visualization of r, v and max. independent set

Semantic array:

$\forall v \in V, C[v] = \text{size of max. independent set of the subtree hanging from the vertex } v$

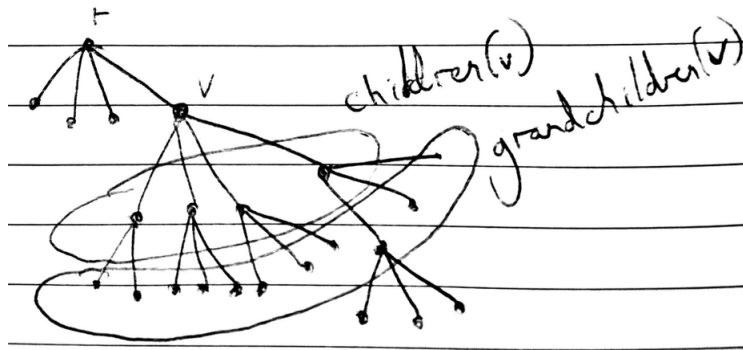


Figure 5: Visualization of v with $\text{CHILDREN}(v)$ and $\text{GRANDCHILDREN}(v)$

Computational array:

$C[v] = \max\{1 + \sum_{u \in \text{GrandChildren}(v)} C[u], \sum_{u \in \text{Children}(v)} C[u]\}$, where the first element in min is $v \in \text{OPT}$, and the second is $v \notin \text{OPT}$.

Exercise:

1. finish proof of correctness
2. *pseudocode*: recursion and memoization

Runtime should be $\mathcal{O}(|V|)$