

Dynamic Programming - Part 2

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Matrix Chain Multiplication

Revisit matrix multiplication

Dot Product

$$\begin{bmatrix} 1 & 5 & 9 & 7 & 3 & 4 \\ 2 & 1 & 9 & 7 & 2 & 6 \\ 9 & 5 & 2 & 2 & 3 & 5 \\ 6 & 6 & 1 & 3 & 1 & 7 \end{bmatrix} \times \begin{bmatrix} 5 & 1 & 3 \\ 9 & 5 & 1 \\ 8 & 7 & 6 \\ 9 & 6 & 8 \\ 8 & 1 & 3 \\ 2 & 2 & 9 \end{bmatrix} = \begin{bmatrix} 217 & 142 & 163 \\ 182 & 126 & 177 \\ 158 & 73 & 114 \\ 141 & 76 & 120 \end{bmatrix}$$

4×6 6×3 4×3

72 Multiplications
in Total!

$(4 \times 6 \times 3)$

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 possible ways (parenthesizations) to compute the product

Matrix-chain Multiplication ...contd

- 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. $((((A_1A_2)A_3)A_4))$

Matrix-chain Multiplication ...contd

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

Algorithm to Multiply 2 Matrices

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 $A \cdot B$

MATRIX-MULTIPLY($A_{p \times q}, B_{q \times r}$)

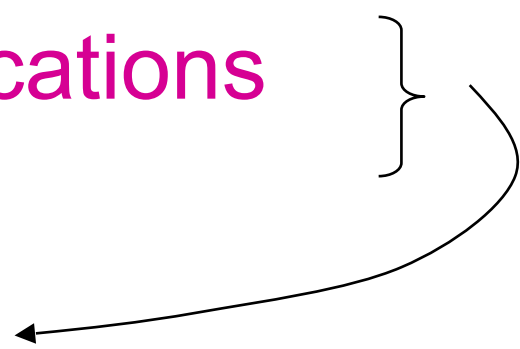
1. **for** $i \leftarrow 1$ **to** p
2. **for** $j \leftarrow 1$ **to** r
3. $C[i, j] \leftarrow 0$
4. **for** $k \leftarrow 1$ **to** q
5. $C[i, j] \leftarrow C[i, j] + A[i, k] \cdot B[k, j]$
6. **return** C

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

Matrix-chain Multiplication ...contd

- 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

Total:
7,500
 - $(A(BC)) = A_{10 \times 100} \cdot E_{100 \times 50}$
 - $BC \Rightarrow 100 \cdot 5 \cdot 50 = 25,000$ scalar multiplications
 - $AE \Rightarrow 10 \cdot 100 \cdot 50 = 50,000$ scalar multiplications

Total:
75,000
- 

Matrix-chain Multiplication ...contd

- 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

Brute force

- For $n \geq 2$, a fully parenthesized matrix product is the product of 2 fully parenthesized matrix subproducts.
- The split can occur between k^{th} and $(k+1)^{\text{th}}$ matrices, for any $k = 1, 2, \dots, n-1$
- So, the recurrence representing the total # of possible parenthesizations is:

$$- P(n) = \begin{cases} 1 & \text{if } n = 1 \\ \sum_{k=1}^{n-1} P(k)P(n-k) & \text{if } n \geq 2 \end{cases}$$

- Solution is tricky -- turns out, it grows as $\Omega\left(\frac{4^n}{n^{3/2}}\right)$
- Or, also true that it grows as $\Omega(2^n)$

Recursion (example of the first level)

- Consider the case multiplying these 4 matrices:
 - A: 2×4
 - B: 4×2
 - C: 2×3
 - D: 3×1
- 1. (A)(BCD) - This is a 2×4 multiplied by a 4×1 ,
 - so $2 \times 4 \times 1 = 8$ multiplications, plus whatever work it will take to multiply (BCD).
- 2. (AB)(CD) - This is a 2×2 multiplied by a 2×1 ,
 - so $2 \times 2 \times 1 = 4$ multiplications, plus whatever work it will take to multiply (AB) and (CD).
- 3. (ABC)(D) - This is a 2×3 multiplied by a 3×1 ,
 - so $2 \times 3 \times 1 = 6$ multiplications, plus whatever work it will take to multiply (ABC).

Recursive formula

$$m(i, j) = \begin{cases} 0 & \text{If } i = j \\ \min_{i \leq k < j} \{m(i, k) + m(k + 1, j) + p_{i-1} \cdot p_k \cdot p_j\} & \text{If } i < j \end{cases}$$

Pseudocode

```
MATRIX-CHAIN( $i, j$ )  
  IF  $i = j$  THEN return 0  
   $m = \infty$   
  FOR  $k = i$  TO  $j - 1$  DO  
     $q = \text{MATRIX-CHAIN}(i, k) + \text{MATRIX-CHAIN}(k + 1, j) + p_{i-1} \cdot p_k \cdot p_j$   
    IF  $q < m$  THEN  $m = q$   
  OD  
  Return  $m$   
END MATRIX-CHAIN  
  
Return MATRIX-CHAIN(1,  $n$ )
```

Recurrence relation

$$\begin{aligned}T(n) &= \sum_{k=1}^{n-1} (T(k) + T(n-k) + O(1)) \\&= 2 \cdot \sum_{k=1}^{n-1} T(k) + O(n) \\&\geq 2 \cdot T(n-1) \\&\geq 2 \cdot 2 \cdot T(n-2) \\&\geq 2 \cdot 2 \cdot 2 \dots \\&= 2^n\end{aligned}$$

Explaining the deduction applied to reach 2nd step

If $n = 7$, for $k = 1, 2, \dots, 6$ we get the following

$$1 + (7-1) = (1+6)$$

$$2 + (7-2) = (2+5)$$

$$3 + (7-3) = (3+4)$$

$$4 + (7-4) = (4+3)$$

$$5 + (7-5) = (5+2)$$

$$6 + (7-6) = (6+1)$$

On the right hand side it is simply $(n-1) + (n-1)$ i.e.,
 $2(n-1)$

Recursion to DP through memorisation - trivial

```
MATRIX-CHAIN( $i, j$ )  
  IF  $T[i][j] < \infty$  THEN return  $T[i][j]$   
  IF  $i = j$  THEN  $T[i][j] = 0$ , return 0  
   $m = \infty$   
  FOR  $k = i$  to  $j - 1$  DO  
     $q = \text{MATRIX-CHAIN}(i, k) + \text{MATRIX-CHAIN}(k + 1, j) + p_{i-1} \cdot p_k \cdot p_j$   
    IF  $q < m$  THEN  $m = q$   
  OD  
   $T[i][j] = m$   
  return  $m$   
END MATRIX-CHAIN  
  
return MATRIX-CHAIN(1,  $n$ )
```


Runtime complexity

- Quadratic due to the size of T i.e., the memo

Putting brackets optimally

$$\begin{array}{ccccccccc} A_1 & \times & A_2 & \times & A_3 & \times & A_4 & \times & A_5 \\ 4 \times 10 & & 10 \times 3 & & 3 \times 12 & & 12 \times 20 & & 20 \times 7 \end{array}$$

Goal: Find the optimal way to multiply these matrices to perform the fewest multiplications.

Naïve Approach: Try them all, and pick the most optimal one.

Running time: $\Omega(4^n/n^{3/2})$ - 4^n dominates! Exponential

Substructure Optimality

There is a better way! Dynamic Programming!

Step 1: Check if the problem has Optimal Substructure

If we have an optimal solution for $A_{i...j}$

Assume the solution has the following parentheses:

$$(\underline{A_{i...k}})(\underline{A_{k+1...j}})$$

If there is a better way to multiply $(A_{i...k})$, then we would have a more optimal solution.

Overlapping subproblems



Recursive formula

Now we want to try out a bunch of values for 'k' in order to see what the best one is:

$$M[i,j] = M[i,k] + M[k+1,j] + p_{i-1}p_kp_j$$

100 200 2x3x4

Since we don't know what k is, we try this range of k:

$$\begin{matrix} 100 & 200 \\ (A_{i\dots k}) & (A_{k+1\dots j}) \\ 2 \times 3 & 3 \times 4 \end{matrix}$$

The minimum returned value is our solution!

$$i \leq k < j$$

Our Final Recursive Formula:

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

An instance

Matrix Chain Multiplication

We want to start with $i = j$, then $i < j$ starting with a spread of 1, working our way up

i \ j	1	2	3	4	5
1	0	120			
2	x	0	360		
3	x	x	0		
4	x	x	x	0	
5	x	x	x	x	0

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

$$\begin{matrix} A_1 & \times & A_2 & \times & A_3 & \times & A_4 & \times & A_5 \\ 4 \times 10 & 10 \times 3 & 3 \times 12 & 12 \times 20 & 20 \times 7 \\ p_0 & p_1 & p_1 & p_2 & p_2 & p_3 & p_3 & p_4 & p_4 & p_5 \end{matrix}$$

$$M[1,2] = \min_{1 \leq k < 2} \{M[1,1] + M[1+1,2] + p_0p_1p_2\}$$

$$M[1,2] = \min_{1 \leq k < 2} \{0 + 0 + 4 \times 10 \times 3\}$$

$$M[1,2] = 120$$

$$M[2,3] = \min_{2 \leq k < 3} \{M[2,2] + M[2+1,3] + p_1p_2p_3\}$$

$$M[2,3] = \min_{2 \leq k < 3} \{0 + 0 + 10 \times 3 \times 12\}$$

$$M[2,3] = 360$$

$$M[3,4] = \min_{3 \leq k < 4} \{M[3,3] + M[3+1,4] + p_2p_3p_4\}$$

$$M[3,4] = \min_{3 \leq k < 4} \{0 + 0 + 3 \times 12 \times 20\}$$

An instance

Matrix Chain Multiplication

We want to start with $i = j$, then $i < j$ starting with a spread of 1, working our way up

i \ j	1	2	3	4	5
1	0	120			
2	x	0	360		
3	x	x	0	720	
4	x	x	x	0	1680
5	x	x	x	x	0

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

$$\begin{array}{ccccccccc} A_1 & \otimes & A_2 & \otimes & A_3 & \otimes & A_4 & \otimes & A_5 \\ 4 \times 10 & & 10 \times 3 & & 3 \times 12 & & 12 \times 20 & & 20 \times 7 \\ p_0 & p_1 & p_1 & p_2 & p_2 & p_3 & p_3 & p_4 & p_4 & p_5 \end{array}$$

$$M[1,3] = \min_{1 \leq k < 3}$$

$$k=1$$

$$= M[1,1] + M[1+1,3] + p_0p_1p_3$$

$$= 0 + 360 + 4 \times 10 \times 12$$

$$= 840$$

$$k=2$$

$$= M[1,2] + M[2+1,3] + p_0p_2p_3$$

$$= 120 + 0 + 4 \times 3 \times 12$$

$$= 264 \leftarrow$$

An instance

i \ j	1	2	3	4	5
1	0	120	264	1080	1344
2	x	0	360	1320	1350
3	x	x	0	720	1140
4	x	x	x	0	1680
5	x	x	x	x	0

Trace the solution

$$\begin{array}{c}
 (A_1 \otimes A_2) ((A_3 \otimes A_4) A_5) \\
 \begin{array}{ccccc}
 4 \times 10 & 10 \times 3 & 3 \times 12 & 12 \times 20 & 20 \times 7 \\
 p_0 \ p_1 & p_1 \ p_2 & p_2 \ p_3 & p_3 \ p_4 & p_4 \ p_5
 \end{array}
 \end{array}$$

$k=2$
 $M[1,5] = \underline{M[1,2]} + \underline{M[3,5]} + p_0 p_2 p_5$

$k=4$ ←
 $M[3,5] = M[3,4] + M[5,5] + p_2 p_4 p_5$

Edit Distance

Edit distance

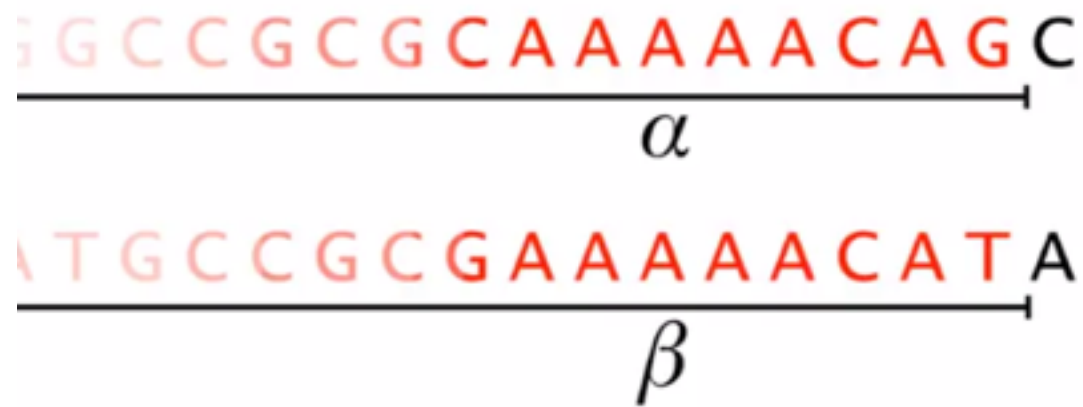
For X, Y where $|X| = |Y|$, *hamming distance* = minimum # substitutions needed to turn one into the other

For X, Y , *edit distance* = minimum # edits (substitutions, insertions, deletions) needed to turn one into the other

If $|X| = |Y|$ what can we say about the relationship between **editDistance**(X, Y) and **hammingDistance**(X, Y)?

$$\text{editDistance}(X, Y) \leq \text{hammingDistance}(X, Y)$$

Substructure optimality



Recurrence

$\alpha \text{ C}$

$\beta \text{ A}$

$$\text{edist}(\alpha \text{ C}, \beta \text{ A}) = \min \begin{cases} \text{edist}(\alpha, \beta) + 1 \\ \text{edist}(\alpha \text{ C}, \beta) + 1 \\ \text{edist}(\alpha, \beta \text{ A}) + 1 \end{cases}$$

Recurrence general

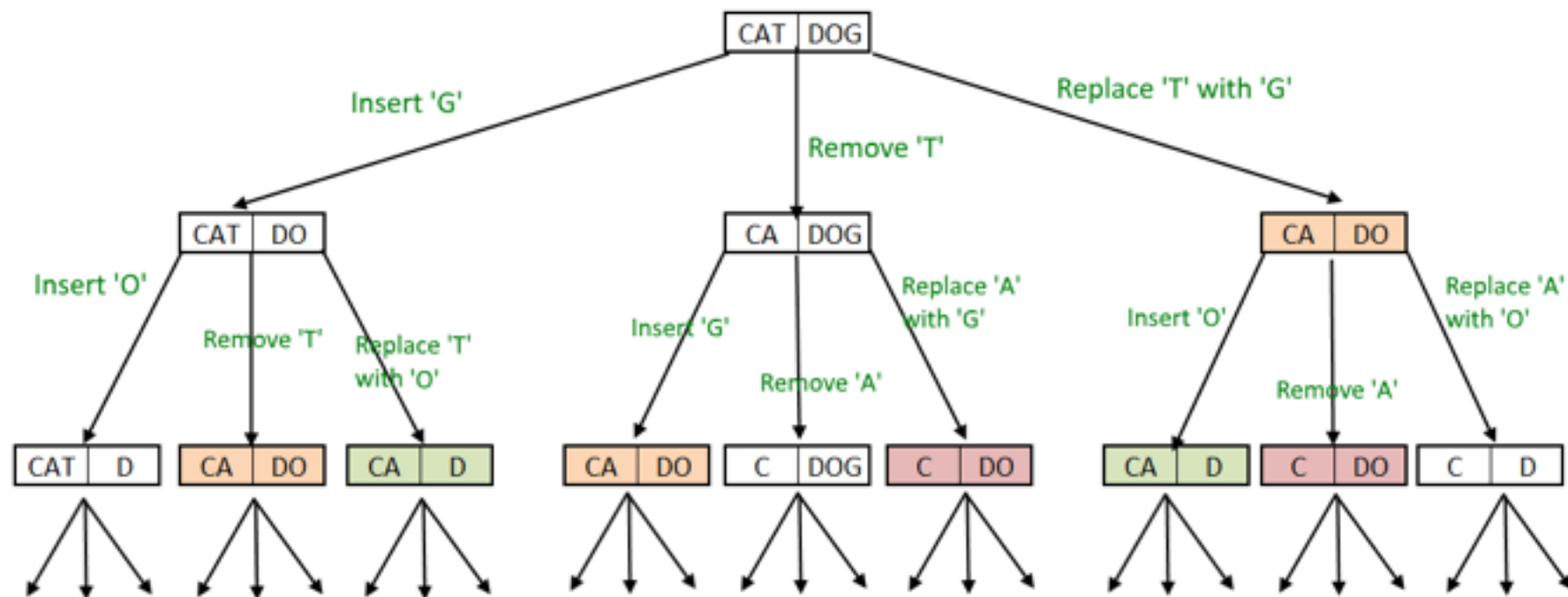
α x

β y

$$\text{edist}(\alpha x, \beta y) = \min \begin{cases} \text{edist}(\alpha, \beta) + \delta(x, y) \\ \text{edist}(\alpha x, \beta) + 1 \\ \text{edist}(\alpha, \beta y) + 1 \end{cases}$$

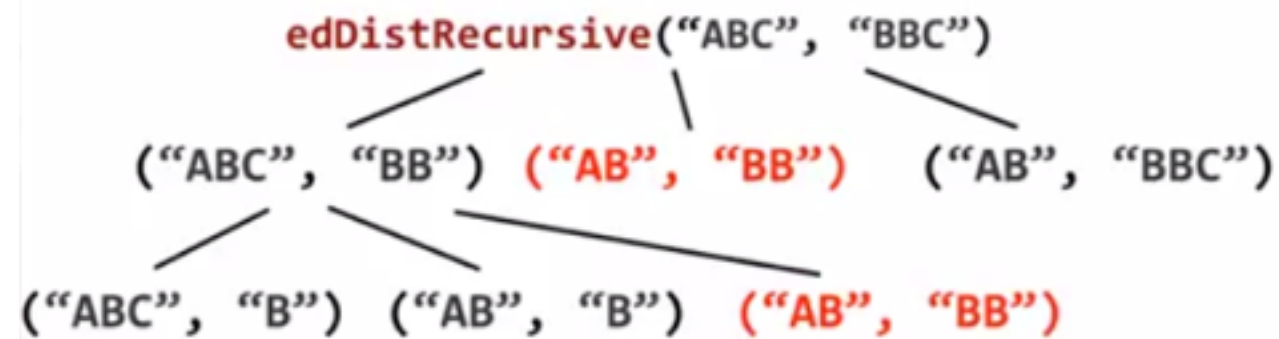
$\delta(x, y) = 0$ if $x = y$, or 1 otherwise

Recursion tree for an example



Recursive function

```
>>> import datetime as d
>>> st = d.datetime.now(); \
... edDistRecursive("Shakespeare", "shake spear"); \
... print (d.datetime.now()-st).total_seconds()
3
31.498284
```



Repetitions

```
n = 0
def edDistRecursive(a, b):
    global n
    if len(a) == 0:
        return len(a)
    if len(b) == 0:
        return len(b)
    if a == 'Shake' and b == 'shake':
        n += 1
    delt = 1 if a[-1] != b[-1] else 0
    return min(edDistRecursive(a[:-1], b[:-1]) + delt,
               edDistRecursive(a[:-1], b) + 1,
               edDistRecursive(a, b[:-1]) + 1)
```

```
>>> edDistRecursive("Shakespeare", "shake spear")
3
>>> n
8989
```

Instance

		Y							
		ε	G	C	T	A	T	A	C
X	ε								
	G								
	C								
	G								
	T								
	A								
	T								
	G								
	C								

$$\boxed{\text{edist}(\alpha x, \beta y)} = \min \begin{cases} \text{edist}(\alpha, \beta) + \delta(x, y) \\ \text{edist}(\alpha x, \beta) + 1 \\ \text{edist}(\alpha, \beta y) + 1 \end{cases}$$

Instance

		Y							
		ε	G	C	T	A	T	A	C
X	ε								
	G								
	C								
	G								
	T								
	A								
	T								
	G								
	C								

$$\text{edist}(\alpha x, \beta y) = \min \begin{cases} \text{edist}(\alpha, \beta) + \delta(x, y) \\ \text{edist}(\alpha x, \beta) + 1 \\ \text{edist}(\alpha, \beta y) + 1 \end{cases}$$

Instance

		Y							
		ε	G	C	T	A	T	A	C
X	ε	0	1	2	3				
	G	1	0	1	2				
	C	2	1	0	1				
	G	3	1	1	1				
	T								
	A								
	T								
	G								
	C								

$$\text{edist}(\alpha x, \beta y) = \min \begin{cases} \text{edist}(\alpha, \beta) + \delta(x, y) = 0 + 1 = 1 \\ \text{edist}(\alpha x, \beta) + 1 = 1 + 1 = 2 \\ \text{edist}(\alpha, \beta y) + 1 = 1 + 1 = 2 \end{cases}$$

End