

# Dynamic Programming

①

## 1. Matrix chain Multiplication.

$$\begin{pmatrix} x & x \\ x & x \\ x & x \end{pmatrix} \begin{pmatrix} y & y & y & y \\ y & y & y & y \end{pmatrix} \begin{pmatrix} z \\ z \\ z \\ z \end{pmatrix} = \begin{pmatrix} \vdots \\ \vdots \\ \vdots \end{pmatrix}$$

$3 \times 2 \quad 2 \times 4 \quad 4 \times 1 \quad 3 \times 1$

$$\begin{pmatrix} xy+xy & xy+xy & xy+xy & xy+xy \\ xy+xy & \cdot & \cdot & \cdot \\ xy+xy & \cdot & \cdot & \cdot \end{pmatrix}$$

$3 \cdot 2 \cdot 4 = 24$  multiplications

$$\begin{pmatrix} (xy+xy)z + (xy+xy)z + (xy+xy)z + (xy+xy)z \\ \cdot \\ \cdot \end{pmatrix}$$

$3 \cdot 4 \cdot 1 = 12$  multiplications

Total  
 $24 + 12 = 36$   
But, there  
is another  
order requiring  
 $8 + 6 = 14$  mult.

Matrix chain problem Given matrices  $A_1, A_2, \dots, A_n$  of size  $p_0 \times p_1, p_1 \times p_2, \dots, p_{n-1} \times p_n$ , find a parenthesization that minimizes the number of multiplications.

Claim. There are  $\frac{1}{n} \binom{2n-2}{n-1}$  different parenthesizations of  $n$  matrices. ②

$(( (x x) (x x) ) (x x))$  or  $(( (x (x x)) (x x) ) x) \dots$  etc.

So, it is not efficient to try all parenthesizations.

### A recursive solution

Let  $m_{ij}$  be the min. number of multiplication needed for  $A_i A_{i+1} \dots A_j$ . We assume that  $p \times q$ -matrix times  $q \times r$ -matrix takes  $pqr$  mult.

function  $M(i, j)$

if  $i=j$  then return  $M:=0$

else  $min := \infty$ ;

for  $k := i$  to  $j-1$  do

mult :=  $M(i, k) + M(k+1, j) + p[i-1] * p[k] * p[j]$ ;

if mult < min then min := mult endif

endfor

endif

Assume array

$p[0 \dots n]$  contains

matrix sizes

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## Analysis

$$\begin{aligned} T(n) &= \left( \sum_{k=1}^{n-1} (T(k) + T(n-k)) \right) + O(1) \\ &= 2 \sum_{k=1}^{n-1} T(k) + O(1) \\ &\geq 2 T(n-1) + O(1) \\ &\geq \sum_{i=0}^{n-1} 2^i \\ &= 2^n - 1 \end{aligned}$$

The main reason for this blow-up in the time complexity is that the algorithm repeatedly solves the same problem recursively. (Think about it why).

Solution. Store the results of the subproblems once computed and use them as needed.

# Dynamic Programming Approach.

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The algorithm uses a tables  $M[1...n, 1...n]$  and  $S[1...n, 1...n]$  to store optimal solutions to subproblems. It works bottom-up.

```
procedure MatrixChain(p);
  for i := 1 to n do M[i,i] := 0 endfor
  for l := 2 to n do
    for i := 1 to n-l+1 do
      j := i+l-1; M[i,j] := ∞
      for k := i to j-1 do
        mult := M[i,k] + M[k+1,j]
                + p[i-1] * p[k] * p[j];
        if (mult < M[i,j]) then
          M[i,j] := mult;
          S[i,j] := k
        endif
      endfor
    endfor
  endfor
```

Example .

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$((A_1 (A_2 A_3)) A_4)$

$p = \begin{array}{|c|c|c|c|c|} \hline 0 & 1 & 2 & 3 & 4 \\ \hline 4 & 2 & 5 & 1 & 3 \\ \hline \end{array}$

M:

|   | 1 | 2  | 3  | 4  |
|---|---|----|----|----|
| 1 | 0 | 40 | 18 | 30 |
| 2 |   | 0  | 10 | 16 |
| 3 |   |    | 0  | 15 |
| 4 |   |    |    | 0  |

S:

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

Print-Parent (S, i, j)

if  $i=j$  then print 'A'

else print '('

print-Parent (S, i, S[i,j])

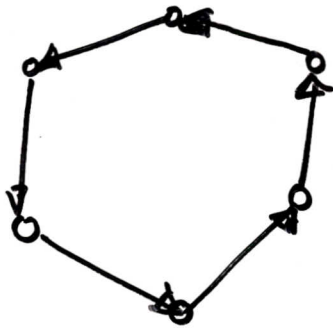
print-Parent (S, S[i,j]+1, j)

print 'j'

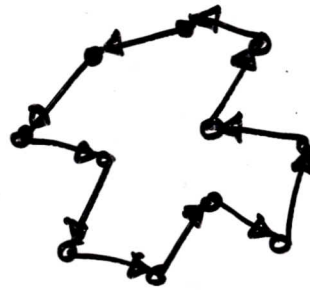
endif

# Polygon Triangulation

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Convex



non-convex

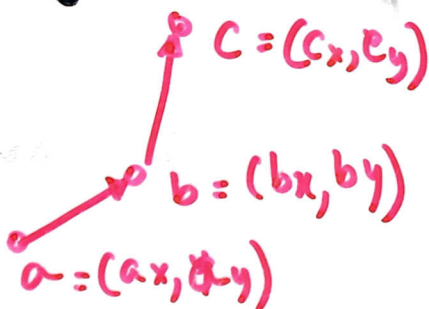
A polygon can be represented with the counterclockwise sequence of its vertices.

type point = record x, y end

polygon = array [0, ..., n-1] of point

var. P : polygon

a. Left-turns. A polygon is convex iff any 3 consecutive points form a left-turn



claim abc is a left-turn  
iff  $\det \begin{bmatrix} a_x & a_y & 1 \\ b_x & b_y & 1 \\ c_x & c_y & 1 \end{bmatrix} > 0$ .

# Existence of a Polygon Triangulation.

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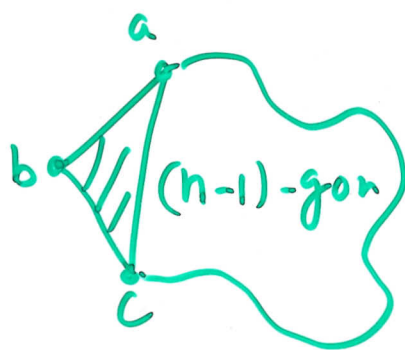
A triangulation is a decomposition of the polygon's interior into triangles whose vertices are the vertices of the polygon.

Claim. Every polygon can be triangulated.

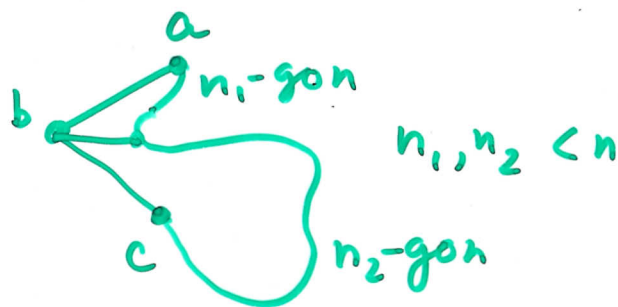
Proof: by Induction.

Let  $b$  be the leftmost point and  $a$  and  $c$  be its predecessor and successor.

Case 1.



Case 2.



The inductive argument can be used to show that the number of triangles is  $n-2$  and the number of chords is  $n-3$ .

$$t(n) = 1 + t(n-1)$$

$$t(n) = t(n_1) + t(n_2)$$

$n_1 + n_2 = n + 2$

$$C(n) = 1 + C(n-1) \text{ and}$$

$$C(n) = 1 + C(n_1) + C(n_2)$$

$n_1 + n_2 = n + 2$

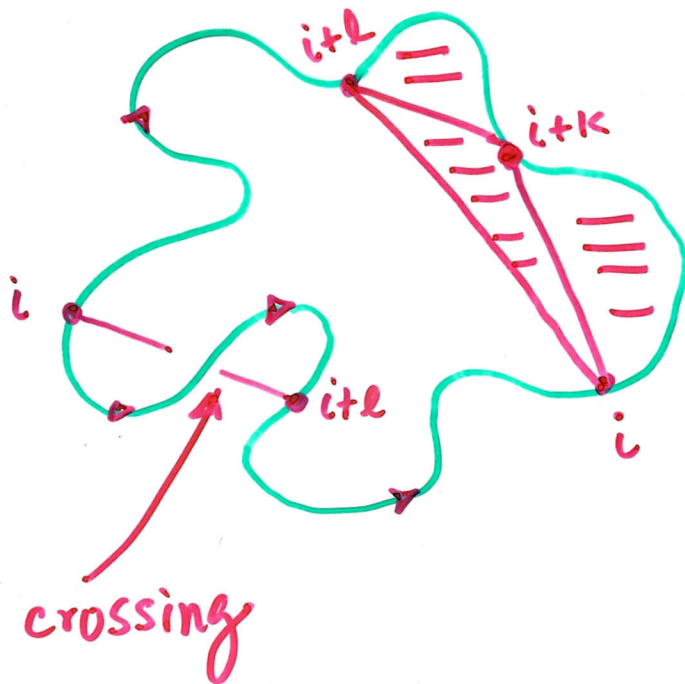
## Construction of Optimal Triangulation.

⑧

It is similar to Matrix chain multiplication.

- Uses arrays  $T[0 \dots n-1, 0 \dots n-1]$  and  $V[0 \dots n-1, 1 \dots n-1]$

when  $T[i, l]$  stores weight of the best triangulation  $P[i, i+l]$  and  $V[i, l]$  stores the vertex  $P[i+k]$  so that  $p[i] p[i+k] p[i+l]$  is a triangle in the optimal triangulation.



$P[i \dots i+l]$  is a polygon iff there is a  $k$  with  $1 \leq k \leq l-1$ , s.t.

$P[i \dots i+k]$  is non-intersecting  
 $P[i+k \dots i+l]$  " " "

$p[i] p[i+k] p[i+l]$  is a  $U$ -fan



Assume  $w(a,b,c)$  is the weight of a triangle  $(a,b,c)$  which is non-negative. ⑨

for  $i := 0$  to  $n-1$  do  $T[i,1] := 0$  endfor;

for  $l := 2$  to  $(n-1)$  do

for  $i := 0$  to  $(n-1)$  do

$T[i,l] := \infty$

for  $k := 1$  to  $l-1$  do

if Left-turn  $(i, i+k, i+l)$  then

$t := T[i,k] + T[i+k, l-k] + w(i, i+k, i+l)$

if  $t < T[i,l]$  then

$T[i,l] := t$ ;  $v[i,l] := i+k$

endif

endif

endfor

endfor

endfor.

$O(n^3)$  time

$O(n^2)$  space.