# Analysis and Design of Algorithms



Algorithms C53230

#### **Tutorial**

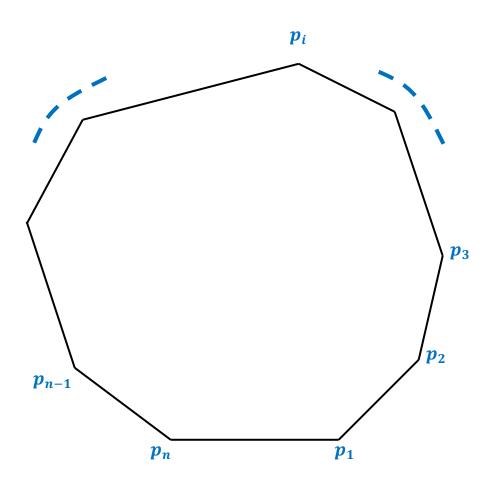
Week 10

## Dynamic Programming National University of Singapore algorithm paradigm (Recap)

- Expressing the solution <u>recursively</u>
- Overall there are only <u>small (maybe polynomial) number of subproblems</u>
- But there is a <u>huge overlap</u> among the subproblems. So the recursive algorithm takes exponential time (solving same subproblem multiple times)
- So we compute the recursive solution <u>iteratively in a bottom-up</u> <u>fashion</u>. This avoids wastage of computation and leads to an efficient implementation

### **A Convex Polygon**





#### Representation:

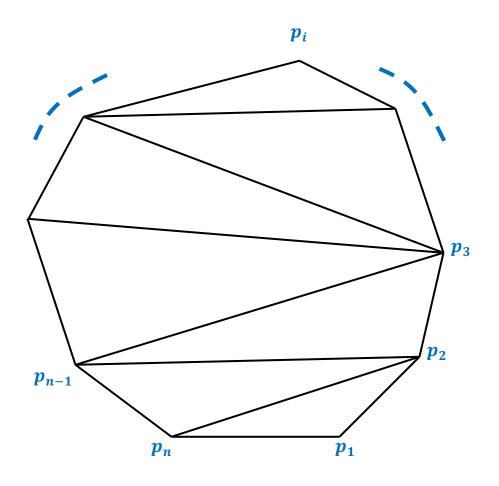
 $< p_1, ..., p_n >$ Stored in an array.

$$< p_i, ..., p_j > :$$

Polygon consisting of points  $p_i$ ,...,  $p_j$ 

## Triangulation of A Convex Polygon





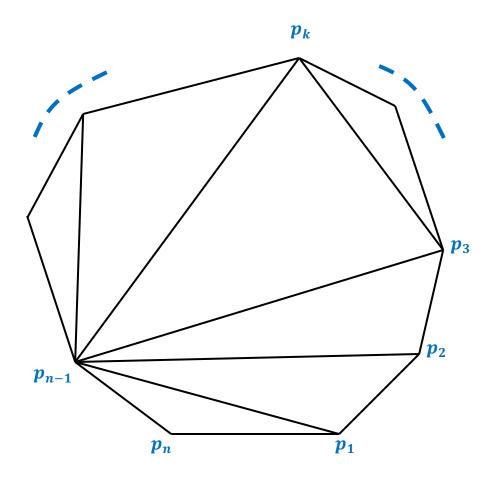
 $\omega(i,j,k)$ : Weight of triangle formed by  $p_i$ ,  $p_j$ ,  $p_k$ .

**Assumption**: It takes O(1) time to compute  $\omega(i,j,k)$ 

Cost of a triangulation: Sum of the weight of n-2 triangles formed.

## **Triangulation of A Convex Polygon**





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**Problem:** Given a convex polygon represented by  $p_1, \ldots, p_n >$ , the objective is to find a triangulation with minimum cost.

Let  $\tau(i,j)$ : cost of an optimal triangulation of polygon  $(p_i,...,p_j)$ 

Write down a recursive formula for the above problem, i.e., express  $\tau(i,j)$  in terms of  $\tau(i',j')$ 's where j'-i' < j-i and  $j' \le j$ ,  $i' \ge i$ .



Consider the following algorithm to find the value of  $\tau(i,j)$ 

```
Find-\tau(i,j)
                                                          What is the running
\{ \mathbf{lf} \ (j=i+1) \}
                                                          time?
     return 0;
                                                           1. 2^{O(j-i)}
Else
                                                          2. O((j-i)^2)
\{t \leftarrow \infty;
                                                          3. O((i-i)^3)
   For (i < k < j)
       temp \leftarrow Find-\tau(i,k) + Find-\tau(k,j) + \omega(i,k,j);
      If (t > temp)
           t \leftarrow temp;
return t;
```



Consider the previous Find $-\tau(1,n)$  algorithm. Which one of the following is/are true.

- 1. Find  $\tau(1, n)$  computes  $2^n$  different sub-problems
- 2. Find- $\tau(1,n)$  computes only at most  $n^2$  different subproblems, but to compute each sub-problem (non-recursively) it takes  $\Omega(\frac{2^n}{n^2})$  time
- 3. Find- $\tau(1, n)$  computes only at most  $n^2$  different sub-problems, but each sub-problem multiple times.

Consider the following algorithm

 $rac{T}{1,n}$ ;

```
Iterative-opt-traingulation (1,n)

\{ for (i = 1 \text{ to } n-1) \mid T[i,i+1] \leftarrow 0; \}
```

```
for (k = i + 1 \text{ to } j - 1)
{
```



Fill the blocks so that the following are true:

- 1. This algorithm finds the value of  $\tau(i,j)$
- 2. This algorithm runs in time  $O(n^3)$  time
- This algorithm computes only at most n<sup>2</sup> different sub-problems, each exactly once