

# Lecture 8: Dynamic Programming II

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# Overview

- Longest Common Subsequence
- Minimum Length Triangulation
- Acknowledgements

## Longest Common Subsequence (LCS)

- **Input:** Two strings  $a_1a_2 \cdots a_m$  and  $b_1b_2 \cdots b_n$ , where  $a_i, b_j \in \Sigma$
- **Output:** Largest  $k$  such that there is  $i_1 < i_2 < \cdots < i_k$  and  $j_1 < j_2 < \cdots < j_k$  for which  $a_{i_\ell} = b_{j_\ell}$  for  $\ell \in [k]$
- **Model:** Word RAM model
- **Example:** given two DNA sequences, want to identify common structures.

AAACCGTGAG  
CACCCTAAGCC

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- want to find  $C(1, 1)$

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- $C(i, j)$  = length of LCS of  $a_i \cdots a_m$  and  $b_j \cdots b_n$
- **Optimal Substructure:** if  $A_i = a_i \cdots a_m$ ,  $B_j = b_j \cdots b_n$  are the partial sequences and  $\Gamma(i, j) = c_1 \cdots c_k$  is an LCS of  $A_i, B_j$ , then:
  - 1  $a_i = b_j \Rightarrow c_1 = a_i = b_j$  and  $\Gamma(i+1, j+1)$  is LCS of  $A_{i+1}, B_{j+1}$
  - 2  $a_i \neq b_j$  and  $c_1 \neq a_i$  then  $\Gamma(i, j)$  is LCS of  $A_{i+1}, B_j$
  - 3  $a_i \neq b_j$  and  $c_1 \neq b_j$  then  $\Gamma(i, j)$  is LCS of  $A_i, B_{j+1}$



## Recursion for LCS

- **Optimal Substructure:** if  $A_i = a_i \cdots a_m$ ,  $B_j = b_j \cdots b_n$  are the partial sequences and  $\Gamma(i, j) = c_1 \cdots c_k$  is an LCS of  $A_i, B_j$ , then:
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  - ②  $a_i \neq b_j$  and  $z_1 \neq a_i$  then  $\Gamma(i, j)$  is LCS of  $A_{i+1}, B_j$
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- Based on optimal substructure, we have:

$$C(i, j) = \begin{cases} 0, & \text{if } i > m \text{ or } j > n \\ C(i + 1, j + 1) + 1, & \text{if } a_i = b_j \text{ and } i, j \leq n \\ \max\{C(i + 1, j), C(i, j + 1)\}, & \text{if } a_i \neq b_j \text{ and } i, j \leq n \end{cases}$$

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- Have  $m \cdot n$  subproblems, so bottom up implementation takes  $O(mn)$  time.
- Correctness of solution follows by the correctness of the recurrence.

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## Minimum Length Triangulation

- **Input:**  $n$  points  $P_1, \dots, P_n \in \mathbb{R}^2$  forming a convex  $n$ -gon  $\Gamma$
- **Output:** a triangulation of  $\Gamma$  such that the perimeters of the  $n - 2$  triangles is minimized (output sum of perimeters)
- **Model:** unit cost model
- will assume we can compute distance between two points in  $O(1)$  time.
- hence can assume we have a function  $\Pi$  which computes the perimeter of a triangle.

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- Can we try all possibilities?

Number of triangulations is the  $(n - 2)^{nd}$  Catalan number:

$$\frac{1}{n - 1} \cdot \binom{2n - 4}{n - 2} = \omega(2^n)$$

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- Can we DP it?

Need to find *optimal substructure* first, and then look for *overlapping subproblems*.

## Recurrence Relation

- **Idea:** which triangle will contain edge  $P_nP_1$ ?  
If we choose index  $2 \leq k \leq n - 1$  for the third point of the triangle, we have the following perimeters:
  - 1 Triangle  $P_nP_1P_k$
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- Leads to recurrence:

$$OPT(1, n) = \max_{2 \leq k \leq n-1} \{ \Pi(P_1, P_k, P_n) + OPT(1, k-1) + OPT(k+1, n) \}$$

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- Bottom-up approach, takes  $O(n)$  time to compute  $OPT(i, j)$  if we know optimum for all subproblems. Hence, running time is  $O(n^3)$ .

# Acknowledgement

- Based on [CLRS 2009, Chapter 15] and Prof Lau's notes  
<https://cs.uwaterloo.ca/~lapchi/cs341/notes/L12.pdf>
- Based on Prof. Brown's notes on minimum triangulation

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