



NUS
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| **Computing**

CS3230

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Computer Science

T07 – Week 8

Dynamic Programming

CS3230 – Design and Analysis of Algorithms

Equality Testing Problem

The key is to break down into the various cases (this I think everyone can do):

- › $A = B$ and decided $A = B$: 100% correct
- › $A = B$ and decided $A \neq B$: 100% correct
- › $A \neq B$ and decided $A = B$: $1 - 1/n$ correct
- › $A \neq B$ and decided $A \neq B$: 100% correct

The trick to break it down into the various cases.

Random Partition of a Graph

Similarly, it is to break down into the various cases. After getting the form, then try to break the math.

- 1 Midterm Exam @ MPSH 1B (80mins): 13-Mar-2025, 14:00-16:00; Arrive at venue by 14:00, Exam starts ~14:10
 - » Info - Up until Randomized Algorithms; No calculators.
 - » Seat Map
 - » Seat Plan
- 2 Anything mentioned in (lectures, tutorials, assignments) would be ok to be quoted; Everything else should be proved before using.
 - » Take note of this as some of you used some beyond the scope of our class (such as Akra-Bazzi, for which that is not accepted without proof.).
- 3 All the best for your midterms!

Key Ideas in Dynamic Programming (DP)

- › **Optimal substructure**: Solve recursively by breaking into subproblems.
- › **Few unique subproblems**: Avoid redundant recomputation.

Two Approaches:

- › **Top-down (Memoization)**: Store computed results to reuse in $O(1)$.
- › **Bottom-up (Tabulation)**: Solve iteratively from base cases.

Both methods improve efficiency by **avoiding redundant** work.

Convex Polygon Triangulation

Minimize the total weight of $n - 2$ triangles in the optimal triangulation, considering:

- › Given a convex polygon with $n \geq 3$ vertices labeled $1, 2, \dots, n$
- › Divide the polygon into $n - 2$ triangles.
- › A triangle (x, y, z) has weight $W(x, y, z)$ (an $O(1)$ black-box function).
- › Multiple triangulations exist.

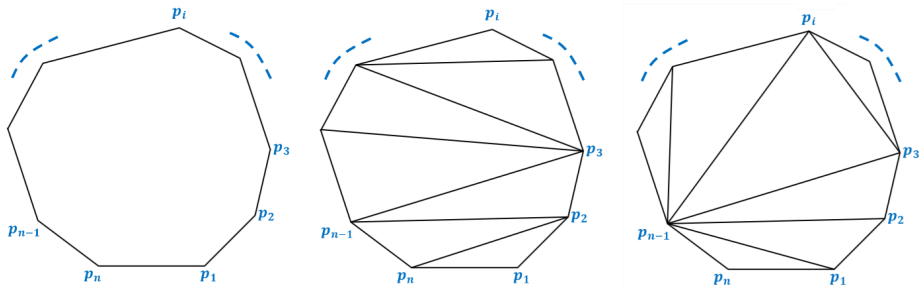


Figure 1: Two triangulation examples (middle, right).

Let $TRI(x, y)$ be a function to triangulate a polygon with minimum weight sum, but we only consider the vertices in the range of $(x, x + 1, x + 2, \dots, y)$. So our problem can be solved by calling $TRI(1, n)$. Your first task is to write a recursive formula of $TRI(x, y)$.

- a. Find the base case of $TRI(x, y)$
- b. Find the recursive case of $TRI(x, y)$

Hint: It calls $TRI(x', y')$ where $x < x'$ or $y' < y$.

Answer

$$TRI(x, y) = \begin{cases} 0, & \text{if } y - x = 1 \\ \min_{k \in [x+1, y-1]} [TRI(x, k) + W(x, k, y) + TRI(k, y)]. & \text{otherwise} \end{cases}$$

- a. Base Case:** Cannot triangulate a line (adjacent vertices x and y).
- b. Recursive Case:** Try all triangulations in any order in the recurrence:
 - Subproblems $TRI(x, k)$ and $TRI(k, y)$
 - Triangle (x, k, y) with weight $W(x, k, y)$

Illustration

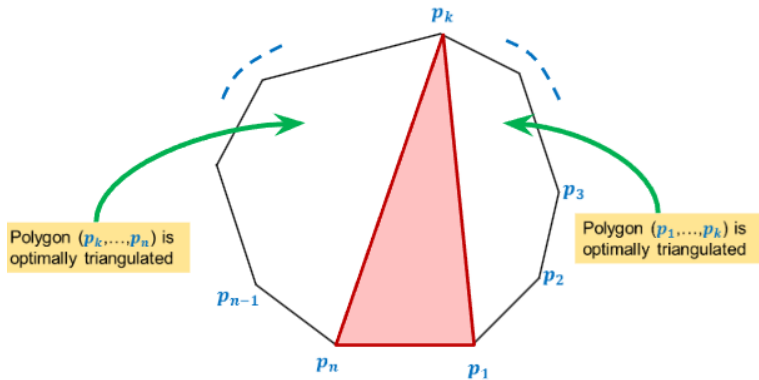


Figure 2: Optimal substructure

What is the time complexity of this recursive formula $TRI(1, n)$, if implemented verbatim.

- a. $O(n^2)$
- b. $O(n^3)$
- c. $O(3^n)$

Answer

Let $T(n)$ be the worst-case running time of $TRI(1, n)$.

$$T(2) = c, \quad \text{when } y - x = 1.$$

Expanding the recurrence for $T(n), T(n - 1)$:

$$\begin{aligned} T(n) &= (T(2) + T(n - 1) + c) + (T(3) + T(n - 2) + c) \\ &\quad + \dots + (T(n - 2) + T(3) + c) + (T(n - 1) + T(2) + c) \\ T(n - 1) &= (T(2) + T(n - 2) + c) + (T(3) + T(n - 3) + c) \\ &\quad + \dots + (T(n - 2) + T(2) + c) \end{aligned}$$

Subtracting $T(n - 1)$ from $T(n)$:

$$\begin{aligned} T(n) - T(n - 1) &= 2T(n - 1) + c \\ \implies T(n) &= 3T(n - 1) + c \\ \implies T(n) &\approx 3^n \in O(3^n). \end{aligned}$$

Which one is the correct explanation regarding the findings from (Q2)?

- a. It has 3^n non-overlapping subproblems, and each call runs in $\Theta(1)$.
- b. It has n^2 non-overlapping subproblems, and each call runs in $\Theta\left(\frac{3^n}{n^2}\right)$.
- c. It has n^2 subproblems, but there are many overlaps.

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Answer

It has n^2 subproblems with significant overlap, making a Dynamic Programming solution necessary for efficiency.

Design a Dynamic Programming (DP) solution for **Convex Polygon Triangulation** problem.

- a. Using Top-Down DP
- b. Using Bottom-Up DP

Answer

Using Top-Down DP

Use a 2D *memo* table of size $n \times n$ ($O(n^2)$ space).

Algorithm

- 1 $TRI(x, y)$ is previously computed: return $memo[x][y]$
- 2 otherwise recursively solve $O(n)$ subproblems:
 - a. Compute the min for $x < k < y$: $TRI(x, k) + W(x, k, y) + TRI(k, y)$
 - b. Store it in $memo[x][y]$

Analysis

- › $O(n^2)$ different subproblems
- › each sub-problem is only computed once in $O(n)$
- › so the total time complexity is $O(n^2 \times n) = O(n^3)$.

Illustration's Weights

Different weight functions can be used. Standard implementations typically define a triangle's weight as its perimeter, the sum of its side lengths. For illustration, we randomly assign weights to each triangle.

Table 1: Truncated table of randomly generated weights for animations and illustrations.

x	k	y	$W(x, k, y)$
0	1	2	3
0	1	3	1
0	1	4	7
0	1	5	4
0	2	3	2

Illustration

Show animation of the memoization table: [T07.q4a.gif](#).

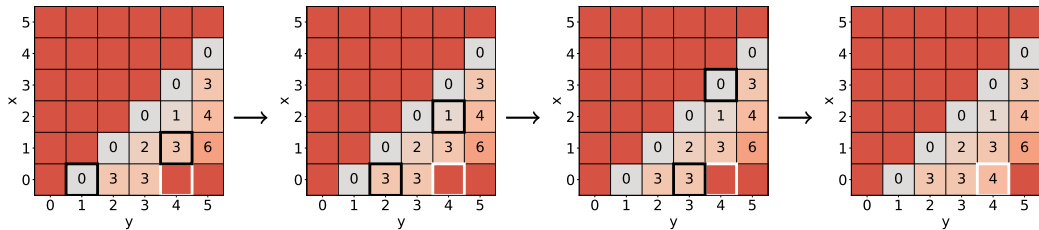


Figure 3: *memo* table: $TRI(0, 4)$ with its subproblems in row $TRI(0, ?)$ and column $TRI(?, 4)$.

Using Bottom-Up DP

Use a 2D *TRI* DP table of size $n \times n$ ($O(n^2)$ space, same as *memo*), but now we must determine the **correct filling order** (**topological order** of the underlying recursion DAG).

Algorithm

- 1 Base Case: For each $x \in [1..n - 1]$, set $TRI[x][x + 1] = 0$. This is **one index away** from the **anti-diagonal** of the $n \times n$ DP table.
- 2 Recursive Case: Fill the table anti-diagonally, starting from **2 indices away** from the anti-diagonal. Each $TRI(x, y)$ needs to compute the min over previously computed values in its row and column, requiring a anti-diagonal filling order.

Analysis

Overall time complexity is $O(n^3)$,

- › which is the same as Top-Down DP approach,
- › Bottom-Up method can benefit from reduced recursion overhead.

Implementation

```
def compute_bottomup(n, w):  
    TRI = [[ -1 ] * n for _ in range(n)] # Initialize the DP table  
    for x in range(n - 1): # Base case, notice the 0-based indexing  
        TRI[x][x + 1] = 0  
  
    # Fill the table anti-diagonally  
    for delta in range(2, n): # Delta is the gap between x and y  
        for x in range(n - delta): # Iterate over all valid x  
            y = x + delta  
            t = float('inf')  
            for k in range(x + 1, y): # min over all x < k < y  
                t = min(t, TRI[x][k] + w(x, k, y) + TRI[k][y])  
            TRI[x][y] = t  
  
    return TRI[0][n - 1]
```

Illustration

Show animation of the DP table: [T07.q4b.gif](#).

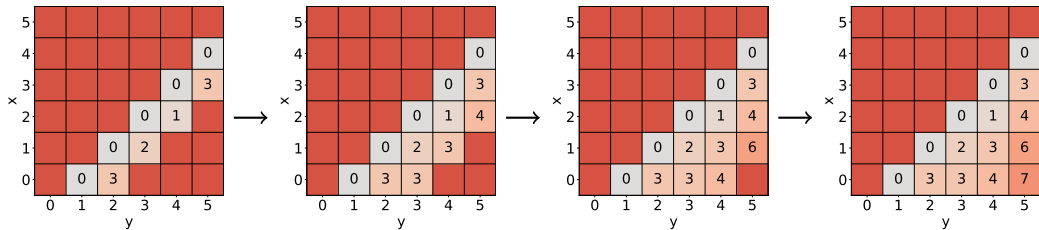


Figure 4: *TRI* DP table: Progressing through each anti-diagonal.

Practical repo: To help you further your understanding, not compulsory; Work for Snack!

- 1 Implement `compute_topdown`.
- 2 Check that you get this output:

```
Top-down == Bottom-up: 7
```