

Solution:

(a) **Idea**

Sort points by x -coordinate in $O(n \log n)$ time. Scan from right to left, keeping a point if its y -coordinate is greater than the maximum y seen so far.

Why It Works

A point $p_i = (x_i, y_i)$ is undominated if no other point has both a larger x and a larger y . After sorting by x , scanning right to left lets us track the maximum y of points with bigger x -coordinates. If y_i exceeds this maximum, p_i isn't dominated by any point to its right, and points to its left have smaller x , so they can't dominate it either. This runs in $O(n \log n)$ time.

Algorithm 1 Find Undominated Points (Sorting and Scanning)

```
1: procedure FIND_UNDOMINATED( $P, n$ )
2:    $P \leftarrow \text{sort}(P, \text{key} = x)$                                 ▷ Sort by  $x$ -coordinate ascending,  $O(n \log n)$ 
3:    $Q \leftarrow []$                                               ▷ Initialize empty list for undominated points
4:    $\text{max\_y} \leftarrow 0$                                           ▷ Track maximum  $y$ -coordinate seen
5:   for  $i = n - 1$  downto 0 do                                ▷ Scan from highest  $x$  to lowest
6:     if  $P[i].y > \text{max\_y}$  then                                ▷ Check if point is undominated
7:        $Q.\text{add}(P[i])$                                           ▷ Add to undominated set
8:        $\text{max\_y} \leftarrow P[i].y$                                 ▷ Update maximum  $y$ 
9:     end if
10:  end for
11:  return  $\text{reverse}(Q)$                                           ▷ Optional: return in increasing  $x$  order
12: end procedure
```

Time Analysis

- Sorting the points by their x -coordinates in ascending order takes $O(n \log n)$ time.
- Scanning from the largest x -coordinate to the smallest takes $O(n)$ time.
- For each point, checking whether it is undominated takes $O(1)$ time by maintaining the current maximal y -coordinate.

Therefore, the total time complexity is

$$O(n \log n)$$

(b) **Argue that for any k there is always a subset of Q that is an optimum solution.**

For any integer k , suppose there exists an optimal solution of size k that includes a point $A \notin Q$. Since A is dominated, there exists a point $B \in Q$ that dominates A . Replacing A with B does not decrease the number of dominated points, as B dominates at least all points that A does.

Thus, for any k , any optimal solution can be transformed into one consisting only of points from Q . \square

Idea

We do preprocessing in advanced

- Compute Q from the method discussed in (a):

$$O(n \log n)$$

- Precompute $D(q_i)$, the set of points dominated by each $q_i \in Q$:

$$O(nm) \text{ total, where } m = |Q| \leq n$$

We define :

$dp[i][k]$ = the maximum number of points in P dominated by a subset of $\{q_0, q_1, \dots, q_i\}$ with exactly k points.

$S[i][k]$ = set of dominated points achieving $dp[i][k]$

Transitions:

- **Do not choose q_i :**

$$dp[i][k] = dp[i-1][k], \quad S[i][k] = S[i-1][k]$$

- **Choose q_i :** Let $D(q_i)$ be the set of points dominated by q_i .

$$dp[i][k] = |S[i-1][k-1] \cup D(q_i)|, \quad S[i][k] = S[i-1][k-1] \cup D(q_i)$$

At the end, we take:

$$\max dp[m][k]$$

where $m = |Q|$.

Algorithm 2 Dynamic Programming for Maximum Dominated Points (Non-Y-Rank)

```

1: Input: Set of points  $P$ , integer  $k$ 
2: Sort  $P$  by decreasing x-coordinate:  $p_1, p_2, \dots, p_n$  ▷  $O(n \log n)$ 
3: Compute  $Q = \{q_0, q_1, \dots, q_{m-1}\}$  (maximal points) ▷  $O(n \log n)$ 
4: for  $i = 0$  to  $m - 1$  do
5:    $D(q_i) = \{p \in P \mid p_x \leq q_{i_x}, p_y \leq q_{i_y}\}$  ▷ Precompute,  $O(nm)$  total
6: end for
7: Initialize  $dp[0][0] = 0, S[0][0] = \emptyset$ 
8:  $dp[0][1] = |D(q_0)|, S[0][1] = D(q_0)$ 
9: for  $k' = 2$  to  $k$  do
10:    $dp[0][k'] = -\infty, S[0][k'] = \emptyset$ 
11: end for
12: for  $i = 1$  to  $m - 1$  do
13:   for  $k' = 0$  to  $k$  do
14:      $dp[i][k'] = dp[i - 1][k']$  ▷ Don't choose  $q_i$ 
15:      $S[i][k'] = S[i - 1][k']$ 
16:     if  $k' \geq 1$  then
17:        $S_{\text{new}} = S[i - 1][k' - 1] \cup D(q_i)$  ▷ Union in  $O(n)$ 
18:        $\text{val} = |S_{\text{new}}|$ 
19:       if  $\text{val} > dp[i][k']$  then
20:          $dp[i][k'] = \text{val}$ 
21:          $S[i][k'] = S_{\text{new}}$ 
22:       end if
23:     end if
24:   end for
25: end for
26: Return:  $\max_{k'=0}^k dp[m - 1][k']$ 

```

Time Analysis

The total time complexity is composed of the following steps:

- **Compute Q :** $O(n \log n)$.
- **Precompute $D(q_i)$:** For each $q_i \in Q$ ($m = |Q| \leq n$), scan all n points to find dominated points in $O(nm)$ total.
- **Dynamic Programming:** For each of the $O(mk)$ states, union sets in $O(n)$ time, totaling $O(mkn)$. With $m, k \leq n$, this is $O(n^3)$.

Total Time Complexity:

$$O(n \log n) + O(n^2) + O(n^3) = O(n^3)$$

Solution:

- (a) Let $LIPS(i, j)$ denote the length and $Lastchar$ denote the first (or the last) character of the longest interesting palindrome subsequence of $A[i \dots j]$. Use similar idea from Lab, but add condition $A[j] \neq Lastchar(i + 1, j - 1)$ to prevent adjacent characters in w are the same.

This function obeys the following recurrence:

$$(LIPS(i, j), Lastchar(i, j)) = \begin{cases} (0, \phi) & \text{if } i > j, \\ (1, \phi) & \text{if } i = j, \\ (2 + LIPS(i + 1, j - 1), A[i]) & \text{if } i < j, A[i] = A[j], \\ & \text{and } A[j] \neq Lastchar(i + 1, j - 1), \\ \begin{cases} (LIPS(i + 1, j), Lastchar(i + 1, j)) & \text{if } LIPS(i + 1, j) \geq LIPS(i, j - 1) \\ (LIPS(i, j - 1), Lastchar(i, j - 1)) & \text{if } LIPS(i + 1, j) < LIPS(i, j - 1) \end{cases} \end{cases}$$

Algorithm 3 Compute LIPS (Longest Interesting Palindromic Subsequence) Table

```

1: for  $i = n$  down to 1 do
2:    $LIPS(i, i - 1) \leftarrow 0$ 
3:    $LIPS(i, i) \leftarrow 1$ 
4:    $Lastchar(i, i) \leftarrow A[i]$ 
5:   for  $j = i + 1$  to  $n$  do
6:     if  $A[i] = A[j]$  and  $A[j] \neq Lastchar(i + 1, j - 1)$  then
7:        $LIPS(i, j) \leftarrow 2 + LIPS(i + 1, j - 1)$ 
8:        $Lastchar(i, j) \leftarrow A[i]$ 
9:     else
10:      if  $LIPS(i + 1, j) \geq LIPS(i, j - 1)$  then
11:         $LIPS(i, j) \leftarrow LIPS(i + 1, j)$ 
12:         $Lastchar(i, j) \leftarrow Lastchar(i + 1, j)$ 
13:      else
14:         $LIPS(i, j) \leftarrow LIPS(i, j - 1)$ 
15:         $Lastchar(i, j) \leftarrow Lastchar(i, j - 1)$ 
16:      end if
17:    end if
18:  end for
19: end for

```

Time Analysis

The algorithm has two nested loops, each running up to n times, resulting in $O(n^2)$ iterations. Each operation inside the loops takes constant time, so the overall time complexity is:

$$O(n^2)$$

- (b) Let $\text{LCPS}(i, j, x, y)$ denote the Longest Common Palindromic Subsequence (LCPS) for the substring $a[i..j]$ of string a and $b[x..y]$ of string b .

This function obeys the following recurrence:

$$\text{LCPS}(i, j, x, y) = \begin{cases} 0 & \text{if } i > j \text{ or } x > y, \\ 1 & \text{if } i = j \text{ and } x = y \text{ and } a[i] = b[x], \\ 0 & \text{if } i = j \text{ and } x = y \text{ and } a[i] \neq b[x], \\ 2 + \text{LCPS}(i + 1, j - 1, x + 1, y - 1) & \text{if } a[i] = a[j] = b[x] = b[y], \\ \begin{cases} \max(\text{LCPS}(i + 1, j, x, y), \text{LCPS}(i, j - 1, x, y)) & \text{if } a[i] \neq b[x] \text{ or } a[j] \neq b[y], \\ \max(\text{LCPS}(i, j, x + 1, y), \text{LCPS}(i, j, x, y - 1)) & \text{otherwise.} \end{cases} \end{cases}$$

Algorithm 4 Longest Common Palindromic Subsequence (LCPS)

```

function LCPS( $a, b, i, j, x, y$ )
  if  $i > j$  or  $x > y$  then
    return 0
  end if
  if  $i = j$  and  $x = y$  and  $a[i] = b[x]$  then
    return 1
  end if
  if  $a[i] = a[j] = b[x] = b[y]$  then
    return 2 + LCPS( $a, b, i + 1, j - 1, x + 1, y - 1$ )
  else
    return max(LCPS( $a, b, i + 1, j, x, y$ ), LCPS( $a, b, i, j - 1, x, y$ ), LCPS( $a, b, i, j, x + 1, y$ ), LCPS( $a, b, i, j, x, y - 1$ ))
  end if
end function

```

Time Analysis

The recurrence $\text{LCPS}(i, j, x, y)$ involves four indices: i, j for string a , and x, y for string b . Each index ranges from 1 to n , so the total number of possible subproblems is $O(n^4)$.

Since each subproblem is solved in constant time $O(1)$ and there are $O(n^4)$ distinct subproblems, the overall time complexity is:

$$O(n^4)$$

■