

- ① • The algorithm uses a divide-and-conquer approach to efficiently find the minimum distance between any two drones on a 2D plane.
- It sorts the drones based on their x-coordinates to simplify the problem.
  - It recursively divides the set of drones into two halves, solves the problem for each half, and then combines the results.
  - The crucial step is finding the minimum distance between drones in the two halves and checking for drones close to the dividing line.

procedure find\_closest\_pair(drones):

$n = \text{drones.length}$

if  $n \leq 3$ : // If there are fewer than 3 drones, brute force check (base case)

return minimum\_distance( calculate\_distance(drones[i], drones[j]) for  $i$  in range(n) for  $j$  in range(i+1, n))

end if

drones.sort(key=lambda drone: drone.x) // Sort drones based on x-coordinates

// Recursively find the closest pair in the left and right halves

mid =  $n / 2$

left\_half = drones[:mid]

right\_half = drones[mid:]

min\_left = find\_closest\_pair(left\_half)

min\_right = find\_closest\_pair(right\_half)

min\_distance = minimum(min\_left, min\_right) // Minimum distance among the left and right halves

// Check for drones with x-coordinates close to the middle

strip = [drone for drone in drones if absolute\_value(drone.x - drones[mid].x) < min\_distance]

strip.sort(key=lambda drone: drone.y) // Sort the strip based on y-coordinates

min\_strip = minimum\_distance( calculate\_distance(strip[i], strip[j]) for  $i$  in range(strip.length) for  $j$  in range(i+1, strip.length))

return minimum(min\_distance, min\_strip)

end

procedure calculate\_distance(drone1, drone2):

return square\_root(((drone1.x - drone2.x)\*\*2 + (drone1.y - drone2.y)\*\*2))

end

procedure closest\_pair\_of\_drones(drones):

drones.sort(key=lambda drone: drone.x)

return find\_closest\_pair(drones)

end

Time Complexity:

• Divide and Conquer part:  $T(n) = 2 \cdot T(n/2) + O(n) \rightarrow \begin{matrix} a=2 \\ b=2 \end{matrix} \begin{matrix} f(n)=n \\ d=1 \end{matrix} \rightarrow a=b^d \rightarrow O(n \log n)$

• Sorting processing: The initial sorting step of the drones based on x-coordinates takes  $O(n \log n)$  time, where  $n$  is the number of drones.

Therefore, the overall time complexity of the given algorithm is  $O(n \log n)$

②

procedure secure\_perimeter(sensors):

if sensors.length == 0:

return 0

else if sensors.length == 1:

return 1

end if

sensors.sort(key=lambda sensor: sensor.x) // Sort sensors based on x-coordinates

mid = sensors.length / 2

left\_half = sensors[:mid]

right\_half = sensors[mid:]

// Recursively find the minimum sensors for each half

left\_min = secure\_perimeter(left\_half)

right\_min = secure\_perimeter(right\_half)

// Find the sensors on the dividing line

mid\_x = sensors[mid].x

dividing\_line = [sensor for sensor in sensors if abs(sensor.x - mid\_x) < radius]

dividing\_line.sort(key=lambda sensor: sensor.y) // Sort the dividing line based on y-coordinates

min\_line = min\_sensors\_dividing\_line(dividing\_line) // Calculate the minimum sensors for the dividing line

return left\_min + right\_min + min\_line

end

procedure min\_sensors\_dividing\_line(dividing\_line):

min\_sensors = 1

rightmost\_sensor = dividing\_line[0]

for sensor in dividing\_line[1:]:

if sensor.y - radius > rightmost\_sensor.y:

min\_sensors += 1

rightmost\_sensor = sensor

end if

end for

return min\_sensors

end

• The algorithm aims to find the minimum number of sensors needed to establish a secure perimeter around a campsite.

• It uses a divide-and-conquer approach, sorting sensors based on their coordinates.

• Recursively, it calculates the minimum sensors needed for the left and right halves and then determines the minimum for the dividing line.

Time Complexity:

• Sorting sensors based on coordinates:  $O(n \log n)$

• Recursive calls on halves:  $\log n$  levels, and at each level, we do constant work.

Resulting in a time complexity of  $O(n \log n)$

- ③ • The algorithm aligns two DNA sequences with minimum cost, considering insertion, deletion and substitution.
- It uses a dynamic programming table to iteratively calculate the minimum cost for alignment prefixes of the sequence.
  - The table entry at  $dp[i][j]$  represents the minimum cost to align prefixes  $seq1[0...i]$  and  $seq2[0...j]$

Steps:

- 1- Initialize a dynamic programming table with dimensions  $(m+1) \times (n+1)$ .  $m$  and  $n$  are length of sequences.
- 2- Fill the table iteratively, considering the costs of insertion, deletion and substitution.
- 3- The value in the bottom-right cell ( $dp[m][n]$ ) represents the minimum cost for aligning the entire sequences.

procedure align\_dna\_sequence(seq1, seq2):

$m, n = seq1.length, seq2.length$

$dp = \begin{bmatrix} 0 \end{bmatrix}^{(n+1)}$  for  $i$  in range  $(m+1)$  // Initialize the dynamic programming table

// Fill the table iteratively

for  $i$  in range  $(m+1)$ :

for  $j$  in range  $(n+1)$ :

if  $i == 0$ :

$dp[i][j] = j * 3$  // Insertion

else if  $j == 0$ :

$dp[i][j] = i * 3$  // Deletion

else:

insertion =  $dp[i][j-1] + 3$  // Cost of insertion

deletion =  $dp[i-1][j] + 3$  // Cost of deletion

substitution =  $dp[i-1][j-1] + ((seq1[i-1] != seq2[j-1]) ? 3 : 0)$  // Cost of substitution

$dp[i][j] = \min(\text{insertion}, \text{deletion}, \text{substitution})$  // Minimum cost

end if

end for

end for

return  $dp[m][n]$

Time Complexity:

- The time complexity is directly proportional to the product of the lengths of the input sequences.
- Filling the dynamic programming table iteratively:  $O(m \cdot n)$ , where  $m$  and  $n$  are the lengths of the sequences.

$O(m \cdot n)$

④ procedure MaxDiscount (sequence)

$n = \text{sequence.length}$

maxDiscount = array of size  $n$ , initialized to 0

maxDiscount[1] = discount (sequence[1]) // Discount of the first store

for  $i$  from 2 to  $n$ :

maxDiscount[i] = max (maxDiscount[i-1], 0) + discount (sequence[i])

end for

return max(maxDiscount)

'sequence' is the array representing the sequence of stores visited

'discount(store)' gives the discount for a particular store

'maxDiscount[i]' stores the minimum discount achievable up to the  $i$ th store in the sequence.

• The algorithm iterates through the sequence once, updating maxDiscount at each step by considering the maximum discount up to the current store.

Time complexity:

• The single loop contributes to a time complexity of  $O(n)$ , where  $n$  is the length of the sequence.

⑤ • The greedy algorithm selects antennas to maximize coverage without causing interference.

• It sorts antennas based on their coverage endpoints and iteratively selects the antenna with the rightmost endpoint that does not intersect with previously selected antennas.

procedure max\_antennas (antennas):

antennas.sort (key = lambda antenna: antenna.end\_point) // Sort antennas based on coverage points

selected\_antennas = []

current\_coverage =  $-\infty$

for antenna in antennas:

if antenna.start\_point > current\_coverage:

selected\_antennas.append (antenna)

current\_coverage = antenna.end\_point

end if

end for

return selected\_antennas.length

end

Time Complexity:

• Sorting antennas based on coverage endpoints:  $O(n \log n)$

• Iterating through sorted antennas:  $O(n)$

Resulting in a time complexity of  $O(n \log n)$