

CSC279 HW1

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Question 1

Idea

This algorithm computes slopes between each base point and all other points, sorts these slopes to group colinear points, and checks for occurrences of identical slopes. By repeating this process for each base point and applying efficient sorting beforehand, the algorithm ensures all possible lines containing more than two points are identified.

Pseudocode

```
# Helper Functions
def mergeSort(lst):
    # Base Case: If the list has 1 or 0 elements, it is already sorted
    if len(lst) <= 1:
        return lst

    # Find the middle index to divide the list into two halves
    mid = len(lst) // 2
    # Recursively sort both halves
    leftHalf = mergeSort(lst[:mid])
    rightHalf = mergeSort(lst[mid:])

    # Merge the two sorted halves
    return merge(leftHalf, rightHalf)

def merge(left, right):
    # Init. empty list
    sortedlst = []
    # Pointers for left and right halves
```

```

i, j = 0, 0

# Merge elements from both halves in sorted order
while i < len(left) and j < len(right):
    if left[i] <= right[j]:
        sortedlst.append(left[i])
        i += 1
    else:
        sortedlst.append(right[j])
        j += 1

# Add any remaining elements from the left half
while i < len(left):
    sortedlst.append(left[i])
    i += 1

# Add any remaining elements from the right half
while j < len(right):
    sortedlst.append(right[j])
    j += 1

return sortedlst

# Main function
function findLinearWithMoreThanTwoPoints(points):
    n = len(points) - 1

    for i = 0 to n:
        # Let base point be points[i] = (x_i, y_i)
        slopes = emptyList()

        # Calculate slopes with respect to the base point (x_i, y_i)
        for j = 0 to n:
            if i == j:
                # Skip if it's the same point
                continue
            # Get coordinates of points
            x_i, y_i = points[i]
            x_j, y_j = points[j]

            # Calculate slope between points[i] and points[j]
            if x_j == x_i:
                # Special case: vertical line
                slope = INFINITE
            else:
                slope = (y_j - y_i) // (x_j - x_i)

```

```

        # Store the slope
        slopes.append(slope)

    # Sort the slopes using merge sort
    sortedSlopes = mergeSort(slopes)

    # Traverse through sorted slopes to count consecutive occurrences
    counter = 0
    for k = 1 to len(sortedSlopes) - 1:
        if sortedSlopes[k] == sortedSlopes[k-1]:
            counter += 1
            # Found a line with more than two points
            if counter >= 1:
                return True

    # Case where no line with more than two points found
    return False

```

Complexity Analysis

- For each base point (x_i, y_i) , we compute the slope w.r.t. every other point. This takes $O(n)$ for each base point. The overall time complexity is $O(n^2)$.
- Sorting $n - 1$ slopes using “merge sort” takes $O(n \log n)$ time. Noticing that we sort the slope for each of the n base points, the total time for sorting is $O(n \cdot (n \log n)) \Rightarrow O(n^2 \log n)$
- After sorting, counting consecutive occurrences of slopes to check for more than two points on a line takes $O(n)$ time per base point. For all base points, the algorithm takes $O(n^2)$ times.

Overall the time complexity of the algorithm is:

$$O(n^2) + O(n^2 \log n) + O(n^2) \sim O(n^2 \log n)$$

Question 2

Idea

In this algorithm, calculating the cross product (i.e. using CCW algorithm) is effective because it allows us to determine the position of the point $p = (x, y)$ relative to other points. Hence, we can identify which region the target point belongs to. By comparing the point’s position value (NOTE: *value* < 0 indicates point is to the left of the line) , we can use binary search to find the correct part efficiently.

Pseudocode

```
# Helper Function
function CCW(Point_a, Point_b, Point_p):
    (x1, y1) = Point_a
    (x2, y2) = Point_b
    (x, y) = Point_p

    # Compute the cross product to determine orientation
    return (x - x1) * (y2 - y1) - (y - y1) * (x2 - x1)

# Main Function
function findPartition(a, b, x, y):
    Point_p = (x, y)
    low = 0
    high = len(a) - 1

    # Binary search loop
    while low < high:
        mid = (low + high) // 2
        Point_a = (a[mid], 0)
        Point_b = (b[mid], 1)

        if CCW(Point_a, Point_b, Point_p) > 0:
            # Point is to the left of the line, search in the left half
            high = mid
        elif CCW(Point_a, Point_b, Point_p) < 0:
            # Point is to the right of the line, search in the right half
            low = mid + 1
        else:
            # Point is exactly on the line
            return mid

    # point p lies with the interval ({a[low], 0}, {b[low], 1}) AND ({a[high], 0}, {b[high], 1})
    return low
```

Complexity Analysis

- Noting that Binary Search has the complexity $O(\log n)$ since it divides the search space in half each time.
- Calculating CCW from point only costs $O(1)$ time.

Overall the time complexity of the algorithm is:

$$O(\log n) \cdot O(1) \sim O(\log n)$$

Question 3

For this question, we need to find the Gauss's Area Formula. The correct expression for the area of a simple polygon given its vertices in order is:

$$\frac{1}{2} \left| \sum_{i=0}^n (x_i y_{i+1} - y_i x_{i+1}) \right|,$$

where $x_n = x_0$ and $y_n = y_0$. This is equivalent to the Shoelace formula and correctly computes the polygon's area by summing the signed areas of the trapezoids formed between each edge and the coordinate axes.

- **Expression 1** is correct because it matches Gauss's formula when $x_n = x_0$ and $y_n = y_0$, which properly accounts for the area by summing from 0 to n .
- **Expression 2** is incorrect because it takes the absolute value of each term individually, leading to a potential overestimation of the area.
- **Expression 3** is incorrect since it uses sums and differences of coordinates instead of the cross products needed to compute the polygon's area according to the Shoelace formula. This incorrect formula fails to properly account for the geometric properties and orientation of the polygon.

Question 4

Idea

By leveraging the convex nature of the polygon, the algorithm iteratively narrows down the search range by comparing the x-coordinates of the current midpoint with its neighboring points; and finding the tangent points of the given the polygon from the "left-most" point. We will get three candidates of x-minimum point by these two approaches, and we only need to return the smallest at last.

Pseudocode

```
function findLeftMostPoint(points):  
    n = len(points)  
  
    # Initialize the binary search range  
    low = 0  
    high = n - 1  
    mid = 0  
  
    # Perform binary search  
    while low < high:
```

```

mid = (low + high) // 2

# Compare x-coordinates to find the minimum
if points[mid].x <= points[(mid - 1 + n) % n].x AND points[mid].x <=
points[(mid + 1) % n].x:
    break

elif points[low].x < points[high].x:
    high = mid + 1

else:
    low = mid - 1

# Note: P stands for the given polygon topology, (-INFINITY, -INFINITY) is the ‘‘left-m
tangent_p1, tangent_p2 = find_tangent(P, (-INFINITY, -INFINITY))

# return the point with minimum x-coordinate
return min_x_coordinate(tangent_p1, tangent_p2, points[mid])

```

Complexity Analysis

- By applying binary search in the algorithm, it costs $O(\log n)$ to locate the target point.
- $find_tangent()$ function in class costs $O(\log n)$ time
- x-coordinates comparison calculation only costs $O(1)$.

Overall the time complexity of the algorithm is:

$$(O(\log n) + O(\log n) + O(\log n)) \cdot O(1) \sim O(3 \log n) \sim O(\log n)$$

Extra Credit: Question 5

Idea

This algorithm uses the properties of reflection and perpendicular bisectors. By calculating the midpoint and slope of the perpendicular bisector of a line segment between two points, we can determine the required line that maximized the reflection point pairs.

Pseudocode

```

function find_max_reflecting_line(points):
    n = length(points)
    line_map = new HashMap()

```

```

# Iterate over all pairs of points
for i from 0 to n-1:
    for j from i+1 to n-1:
        # Compute midpoint
        mid_x = (points[i].x + points[j].x) // 2
        mid_y = (points[i].y + points[j].y) // 2
        midpoint = (mid_x, mid_y)

        # Compute perpendicular slope
        dx = points[j].x - points[i].x
        dy = points[j].y - points[i].y

        # Handle vertical line case separately
        if dy != 0:
            slope = -dx // dy
            intercept = mid_y - slope * mid_x
        else:
            # Vertical line case, x = constant (mid_x)
            slope = Infinity
            intercept = mid_x

        # Normalize slope and intercept to avoid precision errors
        normalized_slope = slope if slope != Infinity else None # Handle infinite slope
        normalized_intercept = intercept

        # Use (slope, intercept) tuple as key
        key = (normalized_slope, normalized_intercept)

        # Increment count in hash map or initialize
        if key in line_map:
            line_map[key] += 1
        else:
            line_map[key] = 1

# Find the maximum count and the corresponding line
max_reflections = 0
max_line = None
for key, count in line_map.items():
    if count > max_reflections:
        max_reflections = count
        max_line = key

# Return the line (slope, intercept) that has the maximum count
return max_line

```

Complexity Analysis

- The outer and inner nested for loop and the last parsing loop might cost $O(n^2)$ time.
- The “Mid point” and “Perpendicular slope” calculation only costs $O(1)$.
- Each insertion and lookup operation in the hash map takes $O(\log n)$ time (If we use balanced tree map or similar implementation in the worst case).

Overall, the time complexity of the algorithm is:

$$(O(n^2) \cdot O(1) \cdot O(\log n)) + O(n^2) \sim O(n^2 \log n + n) \sim O(n^2 \log n)$$