

Documentation in Number of Boomerangs

Table of Contents

1. [Problem Statement](#)
2. [Intuition](#)
3. [Key Observations](#)
4. [Approach](#)
5. [Edge Cases](#)
6. [Complexity Analysis](#)
 - [Time Complexity](#)
 - [Space Complexity](#)
7. [Alternative Approaches](#)
8. [Test Cases](#)
9. [Final Thoughts](#)

1. Problem Statement

We are given n distinct points in a 2D plane, where points $[i] = [x_i, y_i]$. A **boomerang** is defined as a tuple (i, j, k) such that:

- The distance between (i, j) is equal to the distance between (i, k) .
- The **order** of (i, j, k) matters.

We need to count the total number of boomerangs possible.

Example 1

Input: points = $[[0,0],[1,0],[2,0]]$

Output: 2

Explanation: The valid boomerangs are:

- $[[1,0], [0,0], [2,0]]$
- $[[1,0], [2,0], [0,0]]$

Example 2

Input: points = $\llbracket \llbracket 1, 1 \rrbracket, \llbracket 2, 2 \rrbracket, \llbracket 3, 3 \rrbracket \rrbracket$

Output: 2

2. Intuition

The key observation is that the distance between points is the only factor that matters. If a point i has m points at the same distance, then we can form $m * (m - 1)$ boomerangs by permuting these points.

3. Key Observations

- i. We only need to consider distances and not actual coordinates.
- ii. Instead of storing actual distances (which involve square roots), we can store squared distances to avoid floating-point precision issues.
- iii. Using a hashmap (dictionary) to store distances allows us to efficiently count possible boomerangs in $O(n^2)$.

4. Approach

We iterate through each point i and compute distances to all other points j .

Steps:

- i. Use a hashmap to store the count of each squared distance from point i to all other points.
- ii. For each unique distance d in the hashmap, if there are m points at this distance, they can form $m * (m - 1)$ boomerangs.
- iii. Sum up the total boomerangs.

5. Edge Cases

- i. Single Point ($n = 1$)
 - a. There are no other points to form a boomerang, so the output should be 0.
- ii. All Points are Far Apart
 - a. If no two points have the same distance from any given point, the output should be 0.
- iii. Multiple Points at Same Distance
 - a. If a point has multiple others at the same distance, permutations should be counted correctly.

6. Complexity Analysis

Time Complexity

- Outer loop runs $O(n)$.
- Inner loop runs $O(n)$, computing distances and storing them in a hashmap.
- Counting boomerangs takes $O(n)$.
- Overall Complexity: $O(n^2)$.

Space Complexity

- We use a hashmap to store distances, which in the worst case holds $O(n)$ entries.
- Overall Complexity: $O(n)$.

7. Alternative Approaches

- Brute Force ($O(n^3)$)
 - Try all (i, j, k) combinations and check distances.
 - Not feasible for $n = 500$.
- Optimized Sorting Approach ($O(n \log n)$)
 - Sorting points based on distance before counting.
 - Sorting would be $O(n \log n)$, but hashmap lookup is $O(1)$, so hashmap remains optimal.

8. Test Cases

Basic Cases

```
assert Solution().numberOfBoomerangs([[0,0],[1,0],[2,0]]) == 2
assert Solution().numberOfBoomerangs([[1,1],[2,2],[3,3]]) == 2
assert Solution().numberOfBoomerangs([[1,1]]) == 0
```

Edge Cases

All points at the same location (should return 0)

```
assert Solution().numberOfBoomerangs([[0,0],[0,0],[0,0]]) == 0
```

Large input

```
points = [[i, 0] for i in range(500)]
```

```
assert Solution().numberOfBoomerangs(points) >= 0 # Just to check it runs efficiently
```

9. Final Thoughts

- The hashmap approach provides an optimal solution in $O(n^2)$, which is necessary for large inputs ($n = 500$).
- Using squared distances avoids floating-point precision issues.
- This approach efficiently counts permutations using hashmaps and simple distance calculations.