Computational Geometry Homework 3

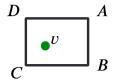
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Problem 1

Implement a full efficient solution to the orthogonal range counting problem for a set S of n points. Test your solution with a driver that allows you to either (a) enter the points interactively, or (b) read them from a file, and query with arbitrary upright rectangles R(a,b,c,d).

For a subset S_i of s is bounded by R, every point $v \in S_i$ must strictly bounded in 4 directions by a permutation of $\{a, b, c, d\}$. As a result this can become a domination problem. Without loss of generality, let a permutation of $\{a, b, c, d\}$ be defined in the following configuration.



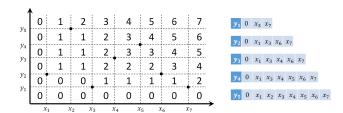
Then for any point $v \in S_i$ v must be dominated by A. Such that, $v \in D(A)$. Where D(A) is the set of points in S dominated by A.

Additionally, it must be true that, $v \notin D(B) \cup D(C) \cup D(D)$.

And so v is in the query rectangle if,

$$v \in D(A) \setminus \{D(B) \cup D(C) \cup D(D)\}\$$

From here note a locus method to preprocess the points in S, by ordering them with respect to increasing y. Then increasing x.



Then, determining if a query point u dominates any point in S, is the same as finding its location in loci, where u dominates everything in its y column with a lower x value.

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        Orthogonal Range Counting
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    def locus_preprocess( points : list ) :
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        y_sort_points = sorted(points, key=lambda v: v.y)
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        loci = {}
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16
        for v in y_sort_points:
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            \#loci[v.y] = [point(0,0)]
18
            loci[v.y] = []
19
            loci[v,y] += sorted(y\_sort\_points[0:y\_sort\_points.index(v)+1], key=lambda q: q.x)
20
21
22
        return loci
23
    def dominates( u : point, locus : dict ) :
25
        keys = locus.keys()
26
        \mathbf{q}_{\mathbf{y}} = 0
27
28
        dominated = []
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30
        for i in keys:
            if i \le u.y:
32
                 q_y = i
33
            else :
34
                 break
35
        if q_y == 0:
37
            return []
38
39
        for v in locus[q_y]:
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             if v.x \le u.x:
                 dominated += [v]
42
43
        return dominated
44
45
    def orthogonal_range_counting( R : rectangle, Locus: dict) :
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        y_max = max([R.a.y, R.b.y, R.c.y, R.d.y])
        x_{max} = \max([R.a.x, R.b.x, R.c.x, R.d.x])
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        y_min = min([R.a.y, R.b.y, R.c.y, R.d.y])
        x_{min} = min([R.a.x, R.b.x, R.c.x, R.d.x])
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        A = point(x_max, y_max)
        B = point(x_max, y_min)
56
57
        C = point(x_min, y_min)
       D = point(x_min, y_max)
58
59
        dom_A = dominates(A, Locus)
        dom_B = dominates(B , Locus)
61
        dom_C = dominates(C , Locus)
dom_D = dominates(D , Locus)
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63
64
        for v in dom_C :
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            dom_B.remove(v)
66
67
        for v in dom_B + dom_D:
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            dom_A.remove(v)
69
70
        return dom_A
71
```

Problem 2

Let $P = \langle p_0, p_1, ..., p_{n-1} \rangle$ with $p_i = (x_i, y_i)$ be a simple polygon stored as a list of vertices in CCW order around ∂P . Vertices are stored using rational coordinates. A chord of P is a segment interior to P whose endpoints belong to ∂P . For example, all diagonals are chords.

Describe a data structure to determine, as fast as possible, the area of the two subpolygons of P induced by a chord C.

- i) When P is convex.
- ii) When P is an arbitrary simple polygon.

Your data structure should require O(n) space. What is the time complexity of your algorithm?

When P is convex.

Let $P = \langle p_0, p_1, ..., p_{n-1} \rangle$, be any convex polygon. Then select p_0 as a root and calculate the area of the triangle created by adjoining p_0 to each other not neighboring vertex through a chord. And so without loss of generality, any one area is found by,

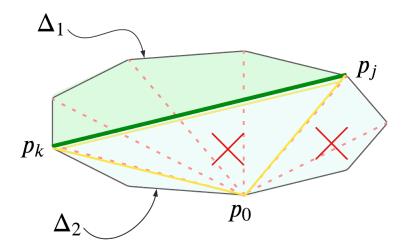
$$\mathcal{A}(p_0, p_i, p_{i-1}) = \frac{1}{2} [(p_1 - p_0) \times (p_{i-1} - p_0)].$$

Now, incrementally store the sum of areas leading to vertex p_i in a hash table.

$$h(i) = \sum_{k < i} \mathcal{A}(p_0, p_k, p_{k-1})$$

And so, the table at index i is the area of the subpolygon $p_i = \langle p_0, p_1, ..., p_i \rangle$.

Now, if a query chord is a diagonal to p_0 , with end points (p_0, p_i) the solution is trivial, the first area is the exact sum in the hash table, h(i) then the second is the total area subtracted by the value in the hash table at h(n) - h(i). Which has constant time. If the chord does not have p_0 as an end-point, has end-points, (p_j, p_k) for $j, k \neq 0$. Then without loss of generality take $j \leq k$, then the area of one-polygon is defined by, $h(k) - h(j) - \mathcal{A}(p_0, p_j, p_k)$. And then of course the area of the second subpolygon is $h(n) - h(k) + h(j) + \mathcal{A}(p_0, p_j, p_k)$.



Here, h(k) is the area of the subpolygon, $< p_0, p_1, ..., p_k >$. And similarly h(j) is the area of the subpolygon $< p_0, p_1, ..., p_j >$. Then the area, Δ_1 is,

$$\Delta_1 = h(k) - h(j) - \mathcal{A}(p_0, p_k, p_j).$$

And,

$$\Delta_2 = h(n) - h(k) + h(j) + \mathcal{A}(p_0, p_k, p_j).$$

Where h(n) is the area of the polygon P.

Solution

And so because the algorithm is a hash table,

Space: O(n), Time: O(c), Preprocessing: O(n)

When P is an arbitrary simple polygon.

Let $P=< p_0, p_1, ..., p_{n-1}>$ by any simple polygon. Then the above method can be generalized to also apply. This is because the area of sub-polygons are in-variant to negative areas in subsections. Or in general, for chord with endpoints, (p_j, p_k) , with out loss of generality let $j \le k$, then the area of the sub-polygon $p_k = (p_0, p_1, ..., p_k)$, is given by the hash table h, h(k), regardless of negative area triangles in the polygon p_k (or has at least one reflex vertex). And so the algorithm holds for any simple polygon.

Solution

$$\Delta_1 = h(k) - h(j) - \mathcal{A}(p_0, p_k, p_j)$$

$$\Delta_2 = h(n) - \Delta_1 = h(n) - h(k) + h(j) + \mathcal{A}(p_0, p_k, p_j)$$

And so because the algorithm is a direct look up hash table,

Space: O(n), Time: O(c), Preprocessing: O(n)