11. Computational Geometry and Convex Hulls CPSC 535 ~ Spring 2019

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Big Idea: Output Sensitive Algorithm

- ▶ input sensitive: time efficiency is a function of the input e.g. size n, # edges m
- output sensitive: efficiency is also a function of the output size e.g. # items returned
- most relevant when the size of the output could be the bottleneck

Computational Geometry

computational X: interdisciplinary study of computer science with X (computational sociology/epidemiology/physics/finance/etc.)

computational geometry (CG): algorithms, data structures, asymptotic analysis, of geometric objects: points, lines, circles, triangle meshes, etc.

Applications

- computer graphics, user interfaces
- GIS, geographic databases
- scene reconstruction (e.g. LIDAR)
- business operations research (e.g. linear programming, aircraft control)
- manufacturing (e.g. feasibility of assembly, castings)

Putting the Geo in CG

Some general algorithms can actually solve geometric problems efficiently, without any awareness of the geometry.

bounding box problem

input: set of 2D points $P = \{p_1, p_2, \dots, p_n\}$

output: points $tl = (x_l, y_t)$ and $rb = (x_r, y_b)$ such that the rectangle with top-left corner tl and bottom-right corner rb contains P

Naïve, optimal algorithm: $x_l, y_t, x_r, y_b = \min x$, $\min y$, $\max x$, $\max y$ respectively; $\Theta(n)$

Computational geometers are most interested when geometric properties matter.

Line Segment Predicates

We can use arithmetic to answer any of the following predicates (questions) about points p_0, p_1, p_2, p_3 in $\Theta(1)$ time:

- 1. Is line segment $\overline{p_0p_1}$ clockwise from $\overline{p_0p_2}$ around the common endpoint p_0 ?
- 2. If we follow $\overline{p_0p_1}$ and then $\overline{p_1p_2}$, do we turn right or left?
- 3. Do line segments $\overline{p_0p_1}$ and $\overline{p_2p_3}$ intersect?
- ⇒ We may use any of these in pseudocode.

Degeneracy and Non-Degeneracy Assumptions

degenerate object: has the proper shape/type, but the values are a special case that betrays the spirit of the definition

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Example: triangle \equiv three points (p_1, p_2, p_3) degenerate triangle: p_1 = p_2 = p_3; or p_1, p_2, p_3 colinear; etc.
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Non-degeneracy assumption:

- constraint that input to a CG algorithm is not degenerate in specific ways
- simplifies algorithm design
- assume that in practice, some combination of
 - degeneracies do not occur
 - input can be preprocessed to remove degeneracies
 - implementer can modify algorithm to handle degeneracies

Sweep Algorithms

A pattern in CG algorithms:

- line sweep: envision a line "sweeping" through the input
- e.g. a vertical line sweeping left-to-right
- helps us visualize a 2D situation as a 1D situation that changes over time
- like duality, doesn't actually change the problem, but might help us problem-solve
- generalizes to higher dimensions e.g. plane sweep in 3D, hyperplane sweep in any dimension

Convex Hulls

convex hull problem

input: set of $n \ge 3$ points Q

output: CH(Q), the subset of Q that is the set of vertices on the convex hull of Q

Convex hull \equiv boundary of convex polygon enclosing all of Q

Applications

- object intersection in raytracing, video games, GUIs
- drawing implicit regions in GIS
- finding farthest points
- component of other algorithms

Approaches to Convex Hulls

Like the sorting problem, many algorithm patterns work for convex hulls, and there is a rich literature of competitive algorithms.

- Greedy pattern: line-sweep, update hull as we go
- Divide-and-conquer: divide Q in half, compute convex hulls for each half, merge two convex hulls into one
- Iterative improvement: start with a superset of CH(Q); refine by repeatedly eliminating a constant fraction of the points until only CH(Q) remains

Baseline Algorithm

Observe

- ▶ any two input points define a line ℓ
- when those points are both in CH(Q), remaining n-2 points are all on the same side of ℓ (geometric property)
- \Longrightarrow for each pair of input points p, q, see whether all other points are on the same side of ℓ
- if so include p, q in CH(Q)

Baseline Pseudocode

```
1: function NAIVE-CONVEX-HULL(Q)
 2:
       H = \emptyset
 3:
       for distinct points p, q \in Q do
 4:
           form line \ell intersecting p and q
 5:
           k = \# points above \ell
           if k = (n-2) or k = 0 then
 6:
              H = H \cup \{p, q\}
 7:
 8:
           end if
 9:
       end for
10:
       return H
11: end function
Analysis: \Theta(n^3) time
```

Graham Scan Idea

- greedy pattern, reduction-to-sorting
- Heuristic: when touring the hull in counter-clockwise order, we only make left turns
- right turn = exiting a concavity; middle point not in hull
- → sweep counter-clockwise, keep points that participate in left turns, drop points in the middle of right turns
- alternative kind of line sweep: rotating the line (not left-to-right)

Graham Scan Greedy Heuristic

- $p_1, \ldots, p_m = Q$ sorted into counter-clockwise order, eliminating ties
- stack S of points; contains hull of points visited already
- base case: push first 3 points onto S
 - for any three points p, q, r forming a non-degenerate triangle, $CH(\{p, q, r\}) = \{p, q, r\}$
- inductive case:
 - examine next input point p_i, top of stack t, next-lowest stack point r
 - if $\angle rtp_i$ is not a left turn $\implies t$ not on hull
- Note: need stack data structure w/ accessor to top two elements

Graham Scan Pseudocode

```
1: function GRAHAM-SCAN(Q)
                                                    \triangleright guaranteed |Q| \ge 3
       p_0 = lowest point in Q (break ties by choosing leftmost point)
       p_1 \dots p_m = \text{sort } Q - \{p_0\} into counter-clockwise order, by polar
    angle with p_0; break ties by keeping only the point farthest from p_0
       S = \text{new stack}
4.
5: S.PUSH(p0)
6: S.PUSH(p1)
7: S.PUSH(p2)
8:
       for i from 3 through m do
           while \angle p_i, S. TOP, S. BELOWTOP is non-left turn do
9.
10:
              S.POP()
           end while
11:
           S.PUSH(p_i)
12:
       end for
13:
       return S
14:
15: end function
```

Graham Scan Analysis

- find $p_0: \Theta(n)$
- ▶ sort: $\Theta(n \log n)$
- eliminate tied points: $\Theta(n)$
- ▶ each stack operation is Θ(1)
- ▶ **for** loop repeats *m* < *n* times
- turn angle test, stack operations are $\Theta(1)$
- ▶ $\Rightarrow \Theta(n \log n)$ time
- ▶ dominating term is sort, organizing data structure is arrayed stack ⇒ good constant factors

Jarvis March

Alternative greedy heuristic: moving around the hull counter-clockwise, each step from one vertex to the next is *the input point whose angle is shallowest* ("gift wrapping")

Jarvis march

- 1. Find the lowest and highest points in Q.
- (right chain) Starting from the lowest point, and until we reach the highest point:
 - 2.1 Linear search Q for the next point, minimizing the angle between the two points.
 - 2.2 Add the first point to CH(Q) and move to the second point.
- 3. (left chain) Starting from the highest point, repeat this process until we reach the lowest point.
- 4. Return CH(Q)

Jarvis March Analysis

Preprocessing to find highest/lowest: $\Theta(n)$

Each iteration of the left/right-chain loops identifies one hull point \implies in total they iterate h times, where $h \equiv$ number of points on the hull.

linear search inside the loops takes $\Theta(n)$ time.

 $\therefore \Theta(nh)$ total time.

Faster than Graham scan's $\Theta(n \log n)$ when $h \in o(\log n)$.

Optimal output-sensitive: Chan's algorithm, $\Theta(n \log h)$.

Summary of Convex Hull Algorithms

Algorithm	Time	Main Idea
Graham Scan	$\Theta(n \log n)$	sort, skip right turns
Jarvis March	$\Theta(nh)$	gift-wrapping
Chan's algorithm	$\Theta(n \log h)$	divide w/ Graham, merge w/ Jarvis