## Geometry

Model

- RAM
- operations on reals, including sqrts.
- (why OK)
- line segment intersections
- DISCRETE randomization

Applications:

- graphics of course
- any domain where few variables, many constraints


## Point location in line arrangements

setup:

- $n$ lines in plane
- gives $O\left(n^{2}\right)$ convex regions
- goal: given point, find containing region.
- for convenience, use triangulated $T(L)$
- triangulation introduces $O\left(n^{2}\right)$ segments (planar graph)
- assume all inside a bounding triangle
how about a binary space partition?
- single line splits input into two groups of n-1 rays
- search time (depth) could be $n$

A good algorithm:

- choose $r$ random lines $R$, triangulate
- inside each triangle, some lines.
- good if each triangle has only $a n(\log r) / r$ lines in it
- will show good with prob. $1 / 2$
- recurse in each triangle - halves lines

Lookup method: $O(\log n)$ time.
Proof of good

- As with cut sampling, consider individual "problem" events, show unlikely
- Let $\Delta$ be all triplets of $L$-intersections
- when $\delta \in \Delta$ is bad:
- let $I(\delta)$ be number of lines hitting $\delta$
- let $G(\delta)$ be lines that induce $\delta$ (at most 6 )
- for bad $\delta$, must have all lines of $G(\delta)$ in $R\left(\right.$ call this $B_{1}(\delta)$ ), no lines of $I(\delta)$ in $R$ (call this $B_{2}(\delta)$.
- bound prob. of bad $\delta$ :
- we know

$$
\operatorname{Pr}[\delta] \leq \operatorname{Pr}\left[B_{1}(\delta)\right] \operatorname{Pr}\left[B_{2}(\delta) \mid B_{1}(\delta)\right]
$$

(why not equal?)

- Given $B_{1}(\delta)$, still need $r-|G(\delta)| \geq r-6 \geq r / 2$ drawings (assuming $r>12$ )
- prob. none picked is at most

$$
\left(1-\frac{|I(\delta)|}{n}\right)^{r / 2} \leq e^{-r I(\delta) / 2 n}
$$

- Only care if $I(\delta)>a n(\log r) / r$-large triplets
$-\operatorname{Pr}\left[B_{2}(\delta) \mid B_{1}(\delta)\right] \leq r^{-a / 2}$ for large triplet
- prob. some bad at most

$$
r^{-a / 2} \sum_{\delta} \operatorname{Pr}\left[B_{1}(\delta)\right]
$$

- sum is expected number of large triplets.
- at most $r^{2}$ points in sample
- at most $\left(r^{2}\right)^{3}=r^{6}$ triplets in sample
- expectation at most $r^{6}$
- choose $a>12$, deduce result.

Construction time:

- Recurrence

$$
T(n) \leq n^{2}+c r^{2} T\left(a n \frac{\log r}{r}\right)=O\left(n^{2+\epsilon(r)}\right)
$$

- $\epsilon$ decreasing with $r$
- by choosing large $r$, arbitrarily close to $O\left(n^{2}\right)$


## Randomized incremental construction

Special sampling idea:

- Sample all except one item
- hope final addition makes small or no change

Method:

- process items in order
- average case analysis
- randomize order to achieve average case
- e.g. binary tree for sorting

Randomized incremental sorting

- Less data structure than binary tree
- repeated insert of item into so-far-sorted
- each yet-uninserted item points to "destination interval" in current partition
- bidirectional pointers (interval points back to all contained items)
- when insert $x$ to $I$,
- splits interval $I$
- must update all $I$-pointers to one of two new intervals
- finding easy easy (since back pointers)
- work proportional to size of $I$
- If analyze insertions, bigger intervals more likely to update; lots of quadratic terms.

Backwards analysis

- run algorithm backwards
- at each step, choose random element to un-insert
- find expected work
- works because:
- condition on what first $i$ objects are
- which is $i^{\text {th }}$ is random
- discover didn't actually matter what first $i$ items are.

Apply analysis to Sorting:

- at step $i$, delete random of $i$ sorted elements
- un-update pointers in adjacent intervals
- each pointer has $2 / i$ chance of being un-updated
- expected work $O(n / i)$.
- true whichever are $i$ elements.
- sum over $i$, get $O(n \log n)$
- compare to trouble analyzing insertion
- large intervals more likely to get new insertion
- for some prefixes, must do $n-i$ updates at step $i$.

