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 $O(n^2)$ convex regions
- goal: given point, find containing region.
- for convenience, use triangulated T(L)
- triangulation introduces $O(n^2)$ 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 $an(\log r)/r$ 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 Δ be all triplets of *L*-intersections
- when $\delta \in \Delta$ is bad:
 - let $I(\delta)$ be number of lines hitting δ
 - let $G(\delta)$ be lines that induce δ (at most 6)
 - for bad δ , must have all lines of $G(\delta)$ in R (call this $B_1(\delta)$), no lines of $I(\delta)$ in R (call this $B_2(\delta)$).
- bound prob. of bad δ :
 - we know

$$\Pr[\delta] \le \Pr[B_1(\delta)] \Pr[B_2(\delta) \mid B_1(\delta)]$$

(why not equal?)

- Given $B_1(\delta)$, still need $r |G(\delta)| \ge r 6 \ge r/2$ drawings (assuming r > 12)
- prob. none picked is at most

$$(1 - \frac{|I(\delta)|}{n})^{r/2} \le e^{-rI(\delta)/2n}$$

- Only care if $I(\delta) > an(\log r)/r$ —large triplets
- $-\Pr[B_2(\delta) \mid B_1(\delta)] \le r^{-a/2}$ for large triplet
- prob. some bad at most

$$r^{-a/2} \sum_{\delta} \Pr[B_1(\delta)]$$

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

Construction time:

• Recurrence

$$T(n) \le n^2 + cr^2 T(an\frac{\log r}{r}) = O(n^{2+\epsilon(r)})$$

- ϵ decreasing with r
- by choosing large r, arbitrarily close to $O(n^2)$

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^{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.