

Peter Caday & Rishi Gupta:

THEORY

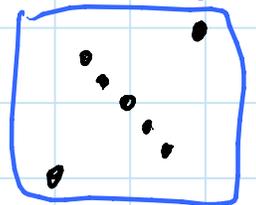
dynamic optimality in BSTs

- cf. Lectures 1 & 2
- alternating Greedy: (from Demaine et al.)
apply \square -Greedy, \square -Greedy, repeat
- Conjecture: never more than 6 iterations!
- simulated annealing algorithm to find bad examples easily finds 6, but not '7
 - troubled by lots of local minima...
- Conjecture: # pts. in step 3

$$< \frac{1}{3} (\# \text{ pts. in steps 1 \& 2})$$

$$\Rightarrow \text{geometric}$$
- either would imply alternating greedy is $O(1)$ -competitive
- Q: is Greedy = $O(\text{unified})$?
 - true, with constant 2, up to $n=10$ & $n=11$ permutations
- Q: is unified = $O(\text{Signed Greedy})$?
 - worst cases computed up to $n=10$, $n=11$ permutations ~ maybe a pattern
 - constant ≈ 1.3

worst case



Cai GoGwilt & Matthew Hwang:

THEORY

- density DSs for Integrated Circuit design
- "dummy fill" common in an IC layer to make density uniform
- problem: given point set & rectangle $\begin{matrix} \square & H \\ \text{(size)} & W \end{matrix}$
 - query: max. weight offset of rectangle
 - update: insert point (no deletions)

- 1D: $O(1)$ query, $O(n^\epsilon)$? insert
 - more interesting than expected
- 2D: $O(1)$ query, $O(n^{1+\epsilon})$ insert

- OPEN: objects are rectangles instead of pts
- OPEN: weighted points
 - update = change weight of point
- in general VLSI seems a rich area for data structural problems

Reid Kleckner:

EXPERIMENT

hopscotch hashing in Python?

- hopscotch hashing [Herlihy, Shavit, Tzafrir 2008]
 - good caching behavior
 - behave well under high load
 - can be made concurrent (maybe avoid GIL)
- Python uses dictionaries everywhere
- Python dictionaries [Tim Peters] use quadratic probing with low load factor
- hopscotch hashing:
 - linear probing with constant cutoff H
 - insertion skips around to move elements out of the way...
 - analysis unclear
 - not dealing with clustering? (parking lot)
 - good performance results
- new results:
 - great for random ints. (like in paper)
 - 0.5% improvement for repeated lookups
 - more for unique keys
 - bad for Python's `hash()` ~ tends to cluster (especially if user gives crappy `--hash--`)
 - even with multiplication method on top?!

Jacob Steinhart & Ye Wang:

THEORY

- parallel functional DSs
- idea: functional \Rightarrow easier to do concurrent access
- want results to correspond to some linearization of ops.
- $k = \# \text{ ops/unit time} \approx \text{parallelism}$
- priority queues:
 - binary heap with path copying
 - $O(\lg n)$ memory/update
 - modification to support parallel inserts with some delay ($O(\lg k)$ merging...)
before they take effect
 - maybe parallel deletes too
- BSTs:
 - do updates bottom up
 - still in progress

Andrew Winslow: kinetic 3D hull

THEORY

(cf. Lecture 5 on kinetic DSs)

- adapt [Bosch 1997] optimal kinetic 2D hull to 3D ~ nontrivial but seems possible
- really kinetic upper/lower envelope
- sketch: binary divide & conquer
at each node, merge & make certificates
- challenges: (3D vs. 2D)
 - vertices & faces can have high degree
 - hull isn't ordered
- "nuke them with "general position" on steroids"
(for now anyway ~ future work to avoid this)
 - $O(1)$ size faces! & deg.-3 vertices
- some new certificates, some adapted
- same performance as 2D
but severe restriction on point sets