CS 240 - Data Structures and Data Management

Module 8: Range-Searching - Enriched

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Based on lecture notes by many previous cs240 instructors

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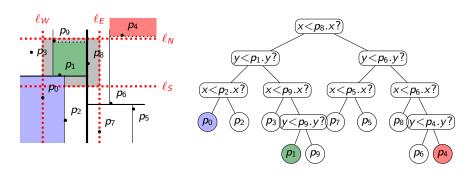
Outline

Boundary nodes in kd-trees

2 3-sided range search

Recall: Q(n) are the boundary-nodes (blue).

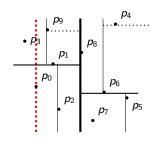
Goal: $Q(n) \in O(\sqrt{n})$.

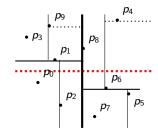


Observation: If v is a boundary-node, then its associated region intersects one of the lines $\ell_W, \ell_N, \ell_E, \ell_S$ that support the query-rectangle.

$$Q(n,\ell) := \max_{\text{kd-trees with } n \text{ points}}$$

number of associated regions that intersect a given line $\boldsymbol{\ell}$



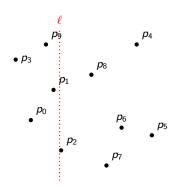


This is independent of ℓ (shift points), so only consider whether ℓ is horizontal or vertical $\rightsquigarrow Q_{\nu}(n), Q_{h}(n)$

$$Q(n) \leq Q(n,\ell_W) + Q(n,\ell_N) + Q(n,\ell_E) + Q(n,\ell_S)$$

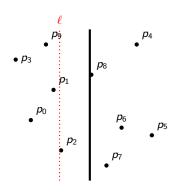
$$\leq 2Q_V(n) + 2Q_h(n)$$

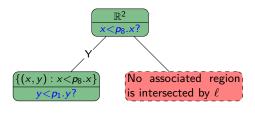
Goal:
$$Q_{\nu}(n) \leq 2Q_{\nu}(n/4) + 2$$
.



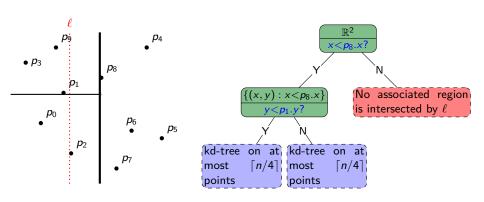


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- $Q(n) \le 2Q_{\nu}(n) + 2Q_{h}(n) \in O(\sqrt{n})$

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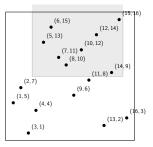
Theorem: In a range-query in a kd-tree (of points in general position) there are $O(\sqrt{n})$ boundary-nodes.

- So range-search takes $O(\sqrt{n} + s)$ time.
- Note: It is *crucial* that we have $\approx n/4$ points in each grand-child of the root.

3-sided range search

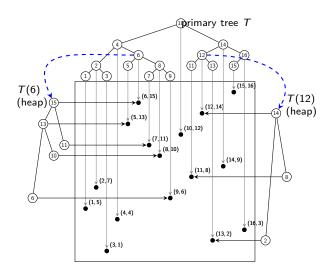
Consider a special kind of range-search:

3sidedRangeSearch(x_1, x_2, y'): return (x, y) with $x_1 \le x \le x_2$ and $y \ge y'$.

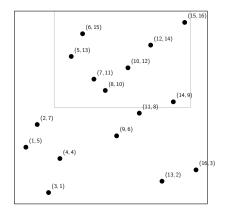


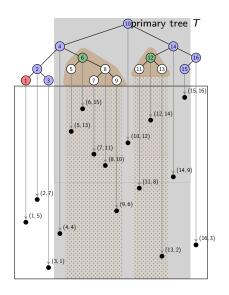
Can we adapt previous ideas to achieve O(n) space and fast range-search time?

Idea 1: Associated heaps

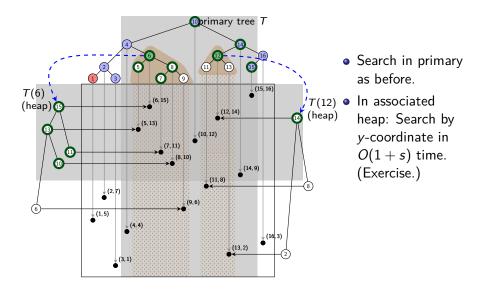


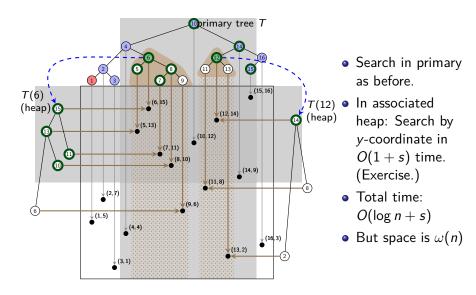
- Primary tree: balanced binary search tree.
- Associated tree: binary heap.
- Space: $\Theta(n \log n)$.
- Range-search time?





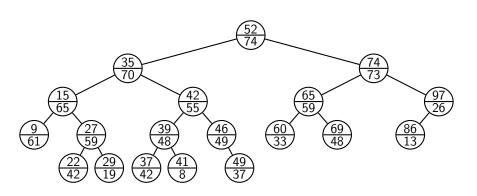
 Search in primary as before.





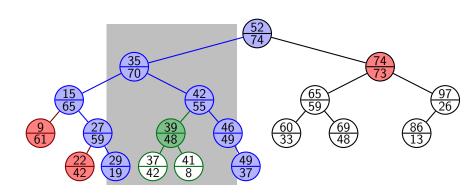
Idea 2: Treaps

Recall: Treap = binary search tree (with respect to keys) + heap (with respect to priorities)



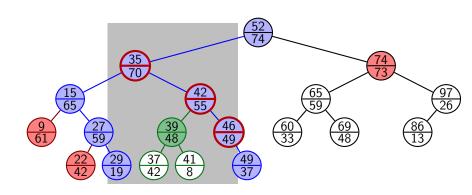
Idea: Use x-coordinate as key, y-coordinate as priority. Space: $\Theta(n)$.

Treap::3-sided-range-search(T, 28, 47, 36):



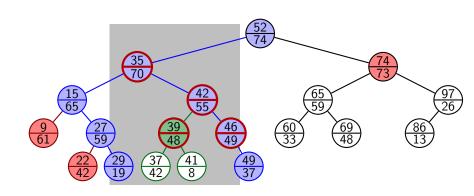
• BST::range-search(x_1, x_2) to get boundary and topmost inside nodes.

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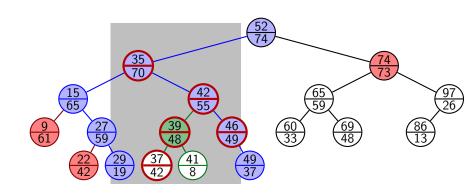
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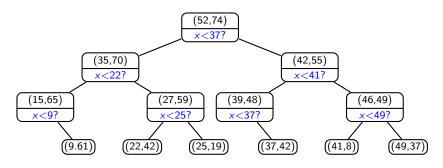
Run-time for 3-sided range search in treaps:

- BST::range-search(x_1, x_2) O(height) since we do not report points.
- Testing boundary-nodes: O(height)
- Testing heap: $O(1 + s_v)$ per topmost inside-node v
- $\Rightarrow O(height + s)$ run-time, O(n) space

But: No guarantees on the height of the treap (not even in expectation) since we cannot choose priorities.

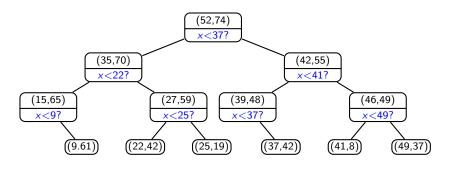
Idea 3: Priority search trees

- Design a new data structure
- Keep good aspects of treap (store *y*-coordinates in heap-order)
- Keep good aspects of kd-tree (split in half by x-coordinate)



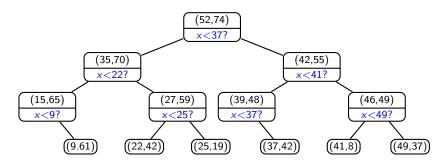
Key idea: The x-coordinate stored for splitting can be *different* from the x-coordinate of the stored point.

Idea 3: Priority search trees



- every node v stores a point $p_v = (x_v, y_v)$,
 - y_v is the maximum y-coordinate in subtree (heap-property!)
- every non-leaf v stores an x-coordinate x'_v (split-line)
 - Every point p in left subtree has $p.x < x'_v$
 - Every point p in right subtree has $p.x \ge x'_{\nu}$
- x'_{ν} is chosen so that tree is balanced \Rightarrow height $O(\log n)$.

Idea 3: Priority search trees



- Construction: $O(n \log n)$ time (exercise)
- search: $O(\log n)$ time
 - Get search-path by following split-lines, check all nodes on path
- insert, delete: Re-balancing is difficult, but can be done (no details).
- 3-sided range search: As in treaps, but height now $O(\log n)$.
 - Run-time $O(\log n + s)$

3-sided range search summary

- Idea 1: Scapegoat tree + associated heaps $O(\log n + s)$ time for range search, but $\omega(n)$ space.
- Idea 2: Treaps O(n) space, but range search takes O(height + s), could be slow
- Idea 3: Priority search tree O(n) space, $O(\log n + s)$ time for range search.

Sometimes it pays to design purpose-built data structures.