

Can we do better?

Range Trees:

- Space is $O(n \log^{d-1} n)$

- Query time:

Counting: $O(\log^d n)$

Reporting: $O(k + \log^d n)$

→ In \mathbb{R}^2 : $\log^2 n$ much better than \sqrt{n} for large n

→ Range trees are more limited

Layering: Combining search structures

- Suppose you want to answer a composite query w. multiple criteria:

- Medical data: Count subjects

Age range: $a_{l_0} \leq \text{age} \leq a_{h_1}$

Weight range: $w_{l_0} \leq \text{weight} \leq w_{h_1}$

- Design a data structure for each criterion individually


- Layer these structures together to answer full query

→ Multi-Layer Data Structures

Recap:

- kd-Tree: General-purpose data structure for pts in \mathbb{R}^d

- Orthogonal range query:

Count/report pts in axis-aligned rect.  Ans = 4

- kd-Tree: Counting: $O(\sqrt{n})$ time

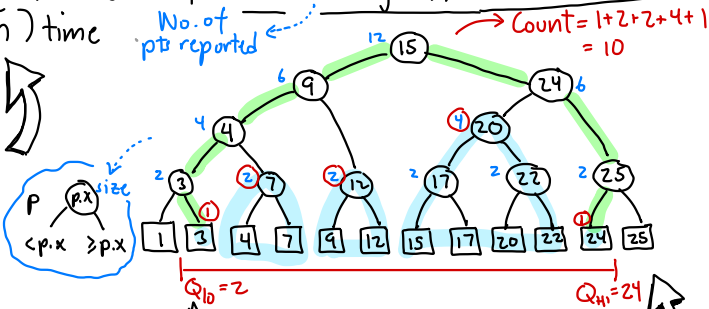
Report: $O(k + \sqrt{n})$ time

No. of pts reported

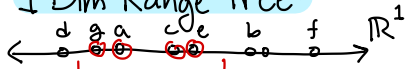
Call this a 1-Dim Range Tree:

Claim: A 1-Dim range tree with n pts has space $O(n)$ and answers 1-D range count/rept queries in time $O(\log n)$ (or $O(k + \log n)$)

Range Trees I



1-Dim Range Tree:



Approach:

- Balanced BST (eg. AVL, RB, ...)

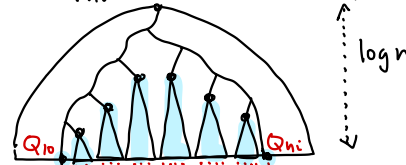
- Assume extended tree

- Each node p stores no. of entries in subtree: $p.size$

Canonical Subsets:

- Goal: Express answer as disjoint union of subsets

- Method: Search for $Q_{l_0} + Q_{h_1}$ + take maximal subtrees



Recursive helper:

int range1Dx(Node p,
Intv Q=[Q_{lo}, Q_{hi}], Intv C=[x_o, x_i])
initial call: range1Dx(root, Q, C_o)

More details:

Given a 1-D range tree T:

- Let Q=[Q_{lo}, Q_{hi}] be query interval

- For each node p, define interval cell C=[x_o, x_i] s.t. all pts of p's subtree lie in C

- Root cell: C_o=[-∞, +∞]



Cases:

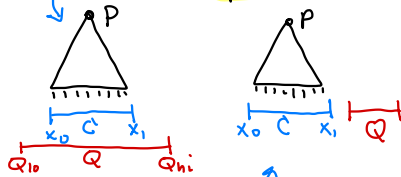
p is external:

- if p.pt.x ∈ Q → 1 else → 0

p is internal:

- C ⊆ Q ⇒ all of p's pts lie within query

→ return p.size

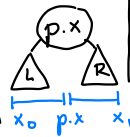


- C is disjoint from Q ⇒ none of p's pts lie in Q

→ return 0

- Else partial overlap

→ Recurse on p's children + trim the cell



Range Trees II

```
int range1Dx(Node p,
Intv Q, Intv C=[xo, xi]) {
if(p is external) return 1
return p.pt.x ∈ Q ? 1 : 0
else if (C ⊆ Q) return p.size
else if (Q + C disjoint) return 0
else return:
    range1Dx(p.left, Q, [xo, p.x])
    + range1Dx(p.right, Q, [p.x, xi])
}
```

x-range:

S(p)

y-range:

p.aux

S'(p)

2-D Range Searching:

- "layer" a range tree for x with range tree for y

- For each node p ∈ 1D-x tree, let S(p) = set of pts in p's subtree

- Def: p.aux: A 1D-y tree for S(p)

Analysis:

Lemma: Given a 1-D range tree with n pts, given any interval Q, can compute O(log n) subtrees whose union is answer to query.

Thm: Given 1-D range tree...

can answer range queries in time O(log n) ... (+k to report)

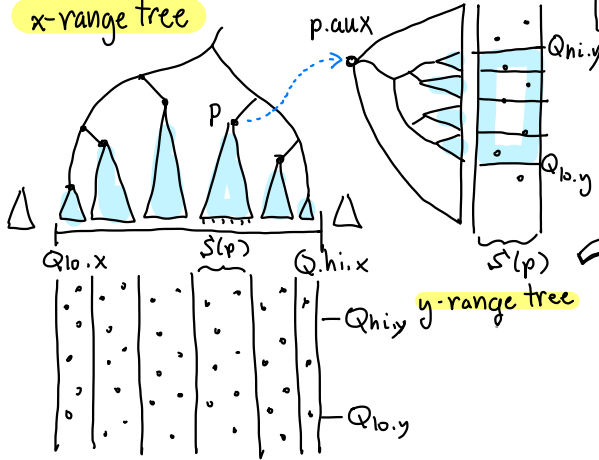
Answering Queries?

Given query range

$$Q = [Q_{lo.x}, Q_{hi.x}] \times [Q_{lo.y}, Q_{hi.y}]$$

- Run range1D_x to find all subtrees that contribute
- For each such node p, run range1D_y on p.aux
- Return sum of all result

x-range tree



Intuition: The x-layer finds subtrees p contained in x-range + each aux tree filters based on y.

2D Range Tree:

- Construct 1D range tree based on x coords for all pts
- For each node p:
 - Let $S(p)$ be pts of p's tree
 - Build 1D range tree for $S(p)$ based on y \rightarrow p.aux
- Final structure is union of x-tree + (n-1) y-trees

Range Trees III

```
int range2D(Node p, Rect Q, Intv C=[x0, x1]) {
    if (p is external) return p.pt ∈ Q ? 1 : 0
    else if (Q.x contains C) { // C ⊆ Q's x-projection
        [y0, y1] = [-∞, +∞] // init y-cell
        return range1Dy(p.aux, Q, [y0, y1])
    } else if (Q.x is disjoint of C) return 0
    else // partial x-overlap
        return range2D(p.left, Q, [x0, p.x])
        + range2D(p.right, Q, [p.x, x1])
}
```

Higher Dimensions?

- In d-dim space, we create d-layers
- Each recurses one dim lower until we reach 1-d search
- Time is the product:

$$\log n \cdot \log n \cdot \dots \log n = O(\log^d n)$$

Analysis: The 1D x search takes of $O(\log n)$ time + generates $O(\log n)$ calls to 1Dy search
 \Rightarrow Total: $O(\log n \cdot \log n) = O(\log^2 n)$

Analysis:

Invoked $O(\log n)$ times - once per maximal subtree

Invoked $O(\log n)$ times - once for each ancestor of max subtree