

## Outline

- BST Range Search
- K-D Trees and Multi-dimensional Search
- Nearest Neighbor Search

## Binary Search Tree Review

- Invariants:
  - Left nodes are always  $<$  current node.
  - Right nodes are always  $>$  current node value.
- Therefore:
  - At each node, we can split the search space in half, enabling  $O(\log n)$  search times.

## Range Search

- Regular Search:

```
node * query(node * root, int target_val){
    if (target_val == root->value)
        return root;
    if (target_val < root->value)
        return query(root->left, target_val);
    if (target_val > root->value)
        return query(root->right,
target_val);
}
```

## Range Search

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  - What should we return?

## Range Search

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  - We should return a list of matches
  - Recursive or Iterative?

## Range Search

- In order to be able to query a range, what should we do?
  - We should return a list of matches
  - Recursive

## Range Search – Recursive Definition

- Take 5 minutes to develop a recursive definition.
- On a node of value  $x$ :
  - 1. What if  $\min < \max < x$ ?
  - 2. What if  $\min \leq x \leq \max$ ?
  - 3. What if  $x < \min < \max$ ?

## Range Search – Recursive Definition

- 1. What if  $\min < \max < x$ ?
  - $x$  is not in the set, search more to the left if possible.
- 2. What if  $\min \leq x \leq \max$ ?
  - $x$  is in the set, add it, and then search both subtrees if possible.
- 3. What if  $x < \min < \max$ ?
  - $x$  is not in the set, search more to the right if possible.

```
vector<node *> BST::range_query(int start_range, int end_range){
    vector<node *> answers;
    range_query_h(root_node, start_range, end_range, answers);
    return answers;
}

void BST::range_query_h(node * current, int start_range, int end_range,
    vector<node *>& answers){
    if (end_range < current->val){
        range_query_h(current->left, start_range, end_range, answers);
        return;
    }

    if (start_range <= current->val && end_range >= current->val){
        range_query_h(current->left, start_range, end_range, answers);
        answers.push_back(this);
        range_query_h(current->right, start_range, end_range, answers);
        return;
    }

    if (start_range < current->val){
        range_query_h(current->right, start_range, end_range, answers);
        return;
    }
    assert(0); // something really bad happened
}
```

## K-D Trees Review

- Like Binary Search Trees except:
  - Each node has multiple keys.
  - At each branch, you consider a different key to split on.
- Due to similarity, you can easily adapt the algorithms for querying BSTs for use in K-D trees.

## Multi-dimensional Search

- BST version:
  - 1. if goal < current node value, return left
  - 2. if goal == current node value, return the node
  - 3. if goal > current node value, return right
- How to modify this for K-D trees?

## Multi-dimensional Search

- K-D Tree version:
  - 1. if goal[dim] < current node value[dim], dim = (dim+1) % total\_dim, return left
  - 2. if goal[dim] == current node value[dim], return the node
  - 3. if goal[dim] > current node value[dim], dim = (dim+1) % total\_dim, return right
- How to modify this for K-D trees?
  - Keep track of which dimension you are searching.
  - As you keep searching down, be sure to keep track of which dimension the nodes are being split on.

## Nearest Neighbor Search

- Goal:
  - Given a point in a K-dimensional space, find the closest point stored in a data structure.

## Nearest Neighbor Search

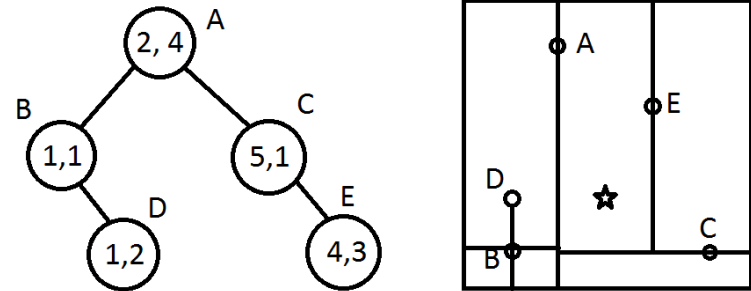
- Goal:
  - Given a point in a K-dimensional space, find the closest point stored in a data structure.
- Use a K-D Tree!

## Nearest Neighbor Search

- Steps:
  - 1. Start with root node and use depth-first search to find where you would insert the node if you were inserting it. Save this as current best
  - 2. Go up one node. If it's better than closest best, it becomes closest best.
  - 3. Check whether there could be any points on the other side of the splitting plane by checking the distance between the target node and the splitting plane.
  - 4. Repeat steps 2-3 until you are at the root node.

## Nearest Neighbor Search

- Given a K-D Tree with the points:
  - (1, 1) (1, 2) (5, 1) (4, 3) (2, 4)
  - Find the point closest to (3, 2) (manhattan space)



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## Nearest Neighbor Search

- 1. Depth-first Search to E.
- 2. E is the best distance = 3.
- 3. Go back to C:
  - C is not closer (3 away). No nodes below C.
- 4. Go back to A.
  - A is not closer. Check if other side of A could be closer.  $\text{Abs}(\text{Goal.x} - \text{A.x}) = 1 < 3$ . Check other side.

## Nearest Neighbor Search

- 1. Depth-first Search to E.
- 2. E is the best distance = 3.
- 3. Go back to C:
  - C is not closer (3 away). No nodes below C.
- 4. Go back to A.
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- 5. DFS to D. D is closer than E (2 away)

## Nearest Neighbor Search

- 3. Go back to C:
  - C is not closer (3 away). No nodes below C.
- 4. Go back to A.
  - A is not closer. Check if other side of A could be closer.  $\text{Abs}(\text{Goal.x} - \text{A.x}) = 1 < 3$ . Check other side.
- 5. DFS to D. D is closer than E (2 away)
- 6. Go back to B.
  - B is not closer. No nodes below B.

## Nearest Neighbor Search

- 4. Go back to A.
  - A is not closer. Check if other side of A could be closer.  $\text{Abs}(\text{Goal}.x - A.x) = 1 < 3$ . Check other side.
- 5. DFS to D. D is closer than E (2 away)
- 6. Go back to B.
  - B is not closer. No nodes below B.
- 7. Go back to the A. Checked both sub-trees, therefore done.