

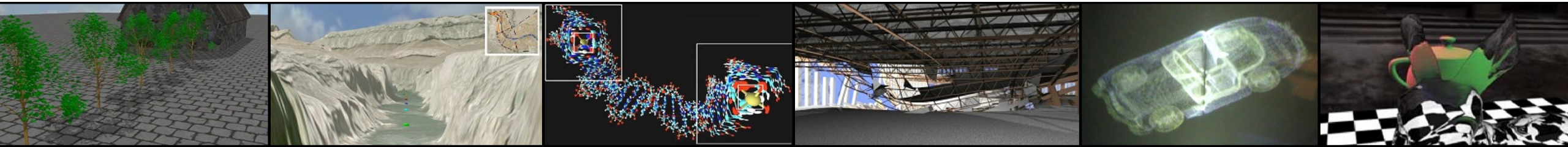
# COT 4521: INTRODUCTION TO COMPUTATIONAL GEOMETRY

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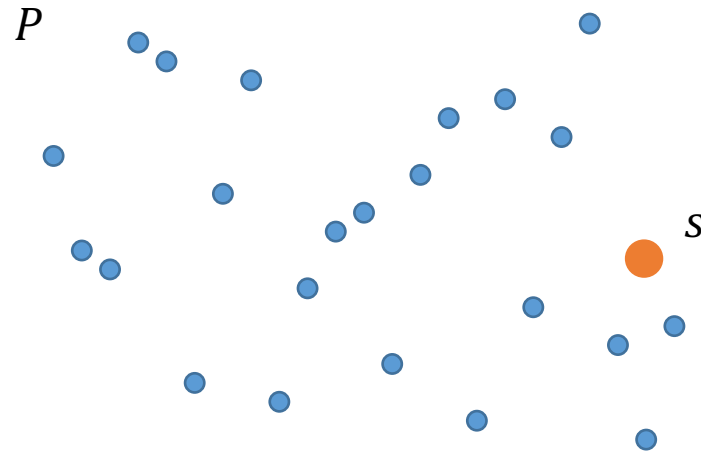
## Searches

Paul Rosen  
Assistant Professor  
University of South Florida

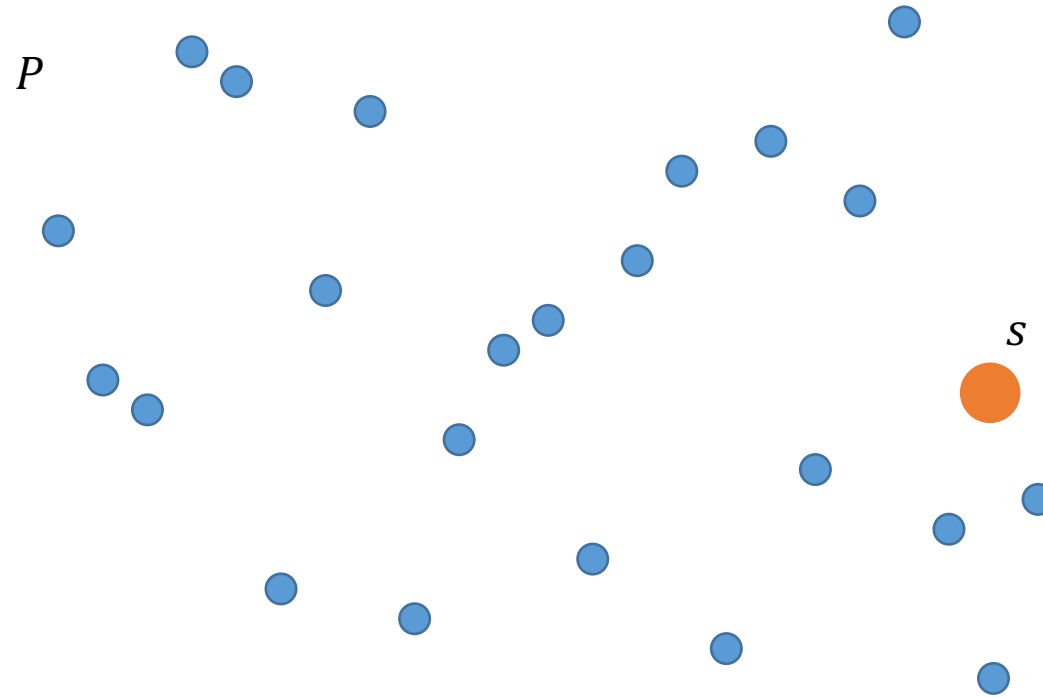


# POINT SEARCHES

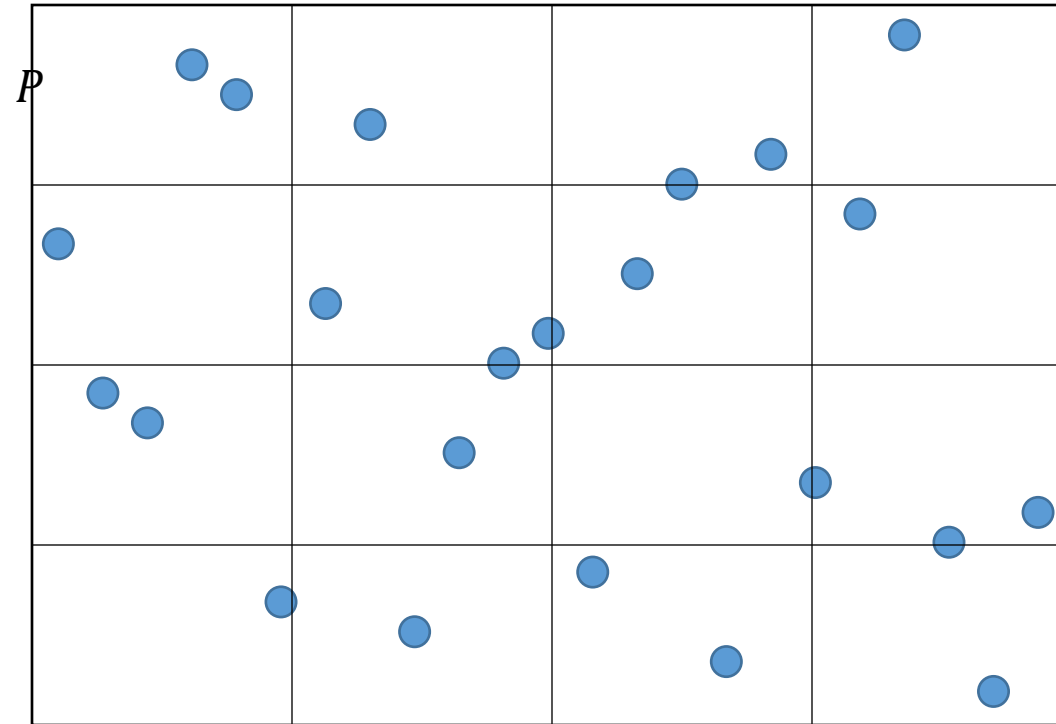
- PROBLEM DEFINITION: GIVEN A SET OF POINTS  $P$  IN  $R^d$  PROVIDE A DATA STRUCTURE THAT GIVEN AN INPUT SEARCH LOCATION  $s$  RETURNS THE  $k$  NEAREST POINTS
  - For simplicity we will primarily consider  $d = 2$  and  $k = 1$



# POINT SEARCHES : IDEAS?



# GRIDS



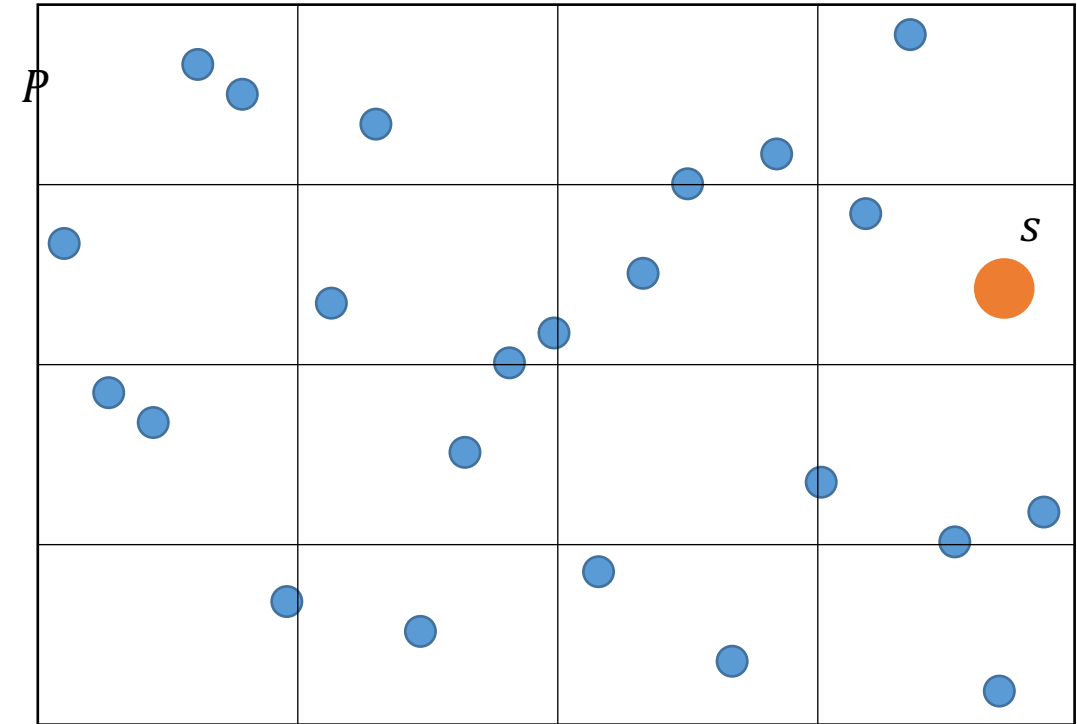
# GRID ALGORITHMS

- CONSTRUCTION
  - Find extrema
  - Divide space by predetermined number of size of interval
  - Place points into grid cells



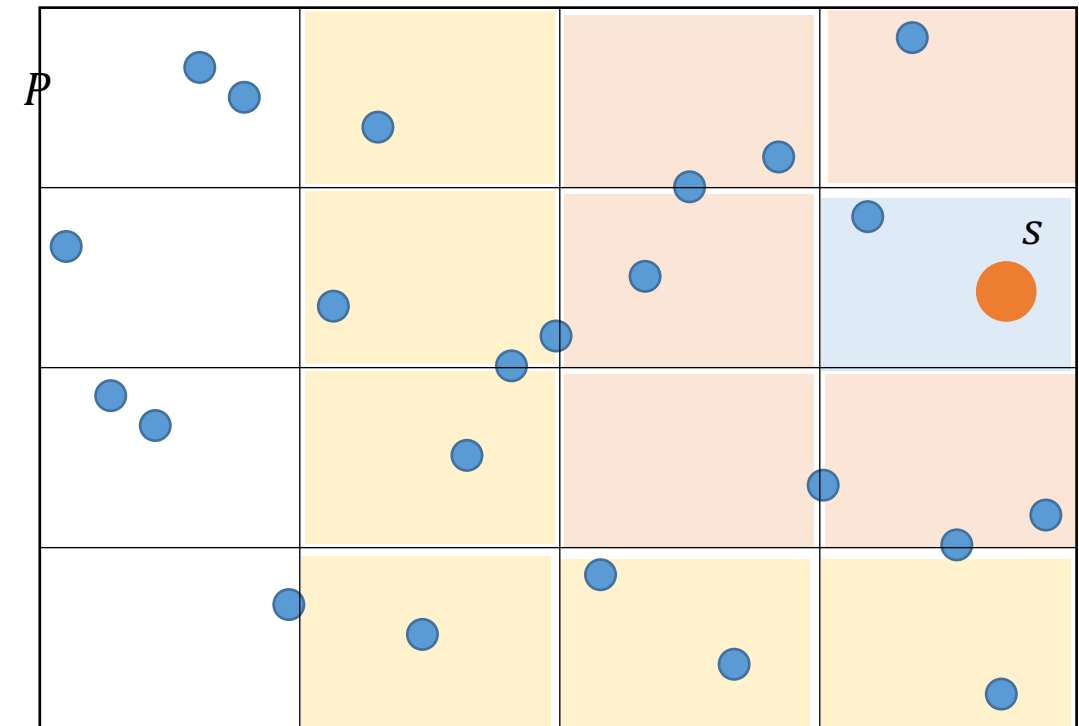
# GRID ALGORITHMS

- SEARCH
  - Given a search location
  - Identify closest grid cell
  - Find closest point inside of cell with distance  $d$
  - Perform the same search with neighboring cells in all directions until cell distance is  $> d$



# GRID ALGORITHMS

- SEARCHING FOR THE CLOSEST POINT  $CP$
- SEARCH RING WHILE
$$d(Ring) < d(CP)$$
- SEARCH CELL IF
$$d(Cell) < d(CP)$$
- FOR EVERY POINT  $p_i$  IN THE CELL
  - If  $d(p_i) < d(CP)$ ,  $CP = p_i$



# FINDING DISTANCE

- POINT-POINT DISTANCE?
- POINT-CELL DISTANCE?
- POINT-RING DISTANCE?





# GRIDS

- CONSTRUCTION
  - $O(n)$
- SEARCH
  - Best Case:  $O(1)$
  - Worst Case:  $O(n)$
  - Average case: 2D  $O\left(\frac{n}{r^2}\right) = O(1)$ , generally  $O\left(\frac{n}{r^d}\right) = O(1)$
- SPACE
  - In 2D  $O(r^2 + n)$ , generally  $O(r^d + n)$

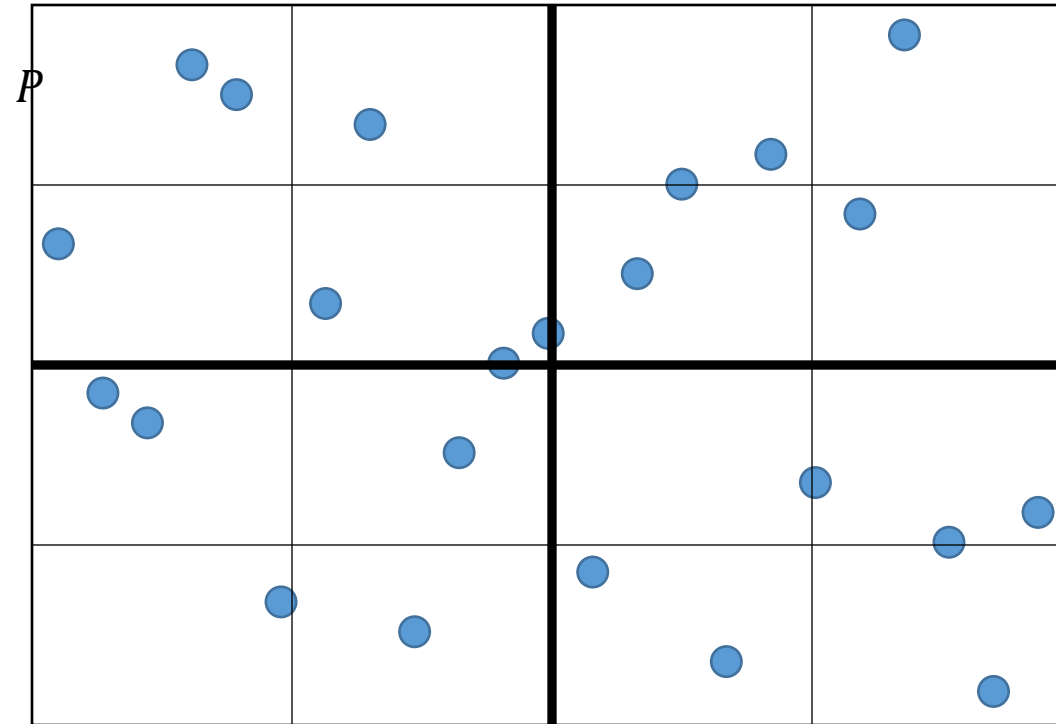


# GRIDS

- **EXAMPLES OF FAILURE CASES?**

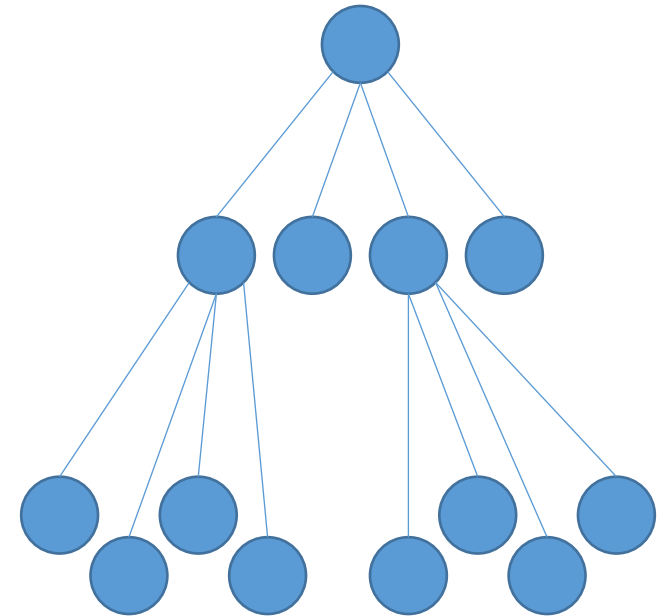


# QUADTREES



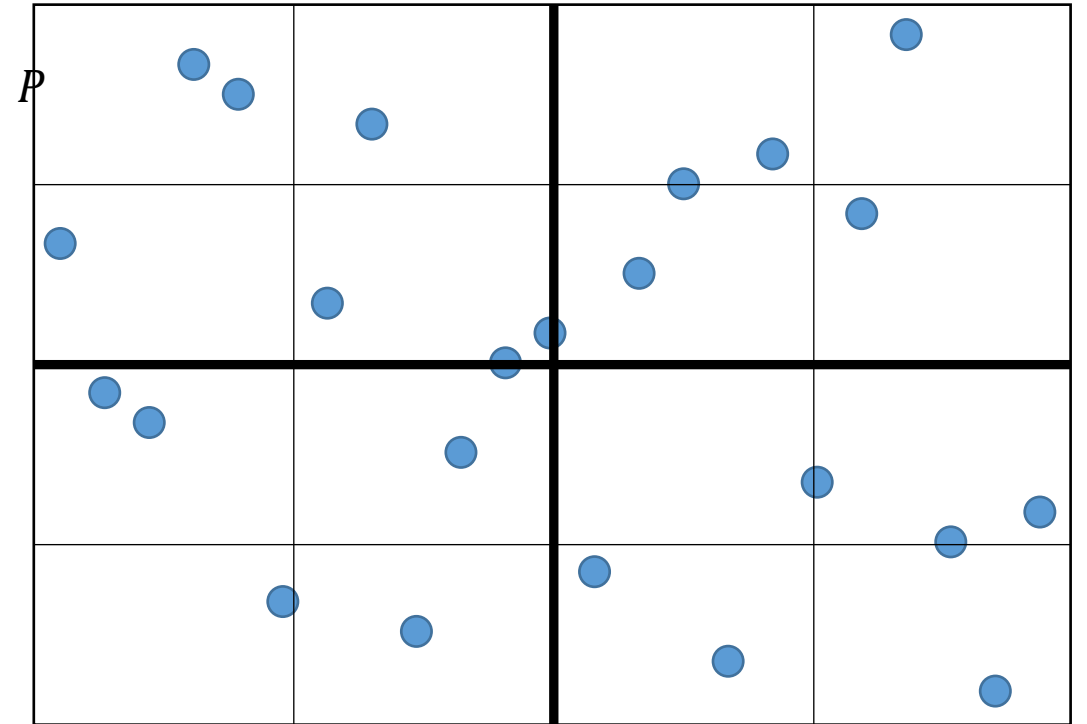
# QUADTREE ALGORITHMS

- CONSTRUCTION
  - Find extrema
  - Divide space in half both horizontally and vertically
    - Ideas?
  - Place points into appropriate quadrant
  - Recurse until termination condition
    - Ideas?



# QUADTREE

- SEARCH
  - Given a search location
  - Recursively identify closest leaf
  - Find closest point inside of leaf with distance  $d$
  - Perform the same search with neighboring leaves



# QUADTREE

- CONSTRUCTION
    - $O(n \log n)$
  - SEARCH
    - Average case:  $O(\log n + k)$ 
      - where  $k$  is the size of the leaf nodes
    - Worst Case:  $O(n)$
  - SPACE: IN  $O(n)$
- \*Octree—3D version of quadtree

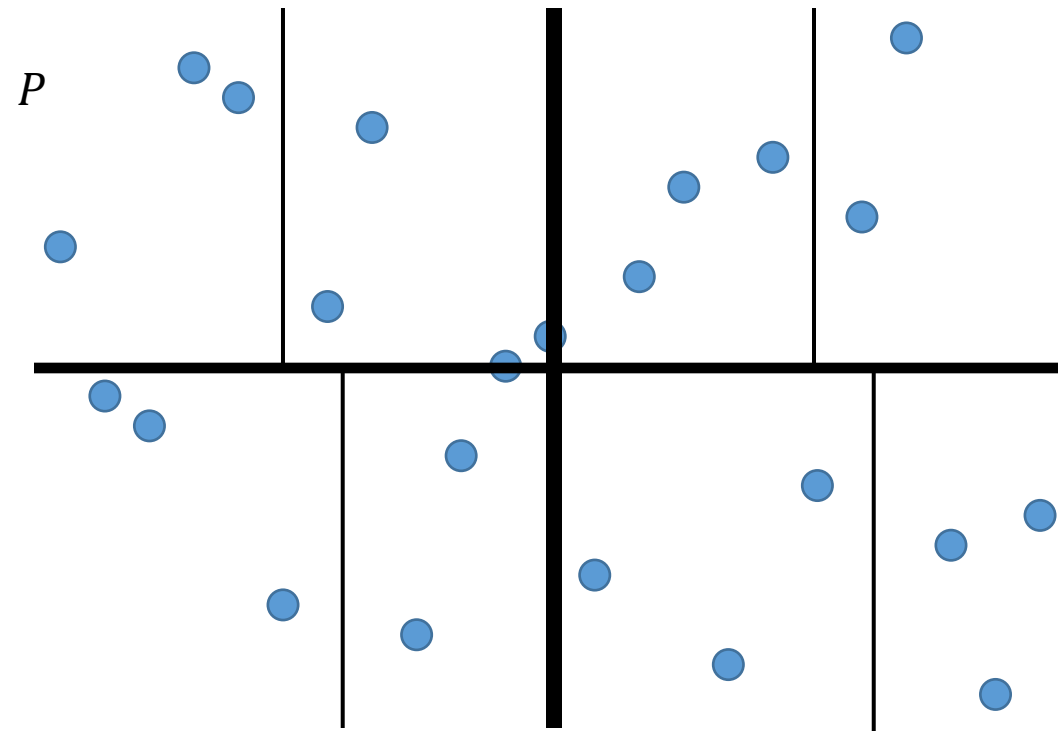


# QUADTREES

- **EXAMPLES OF FAILURE CASES?**



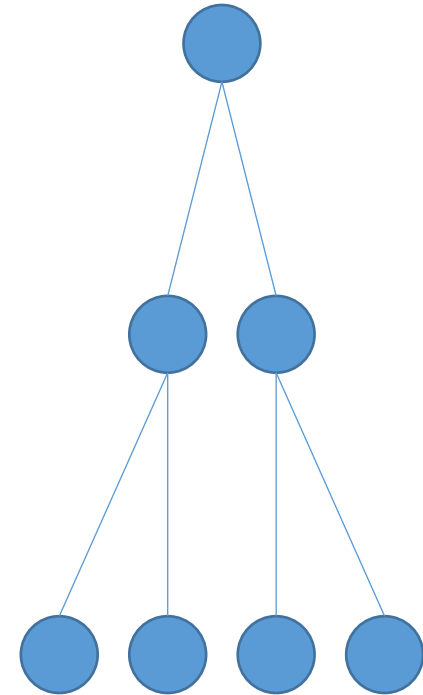
# KD-TREE





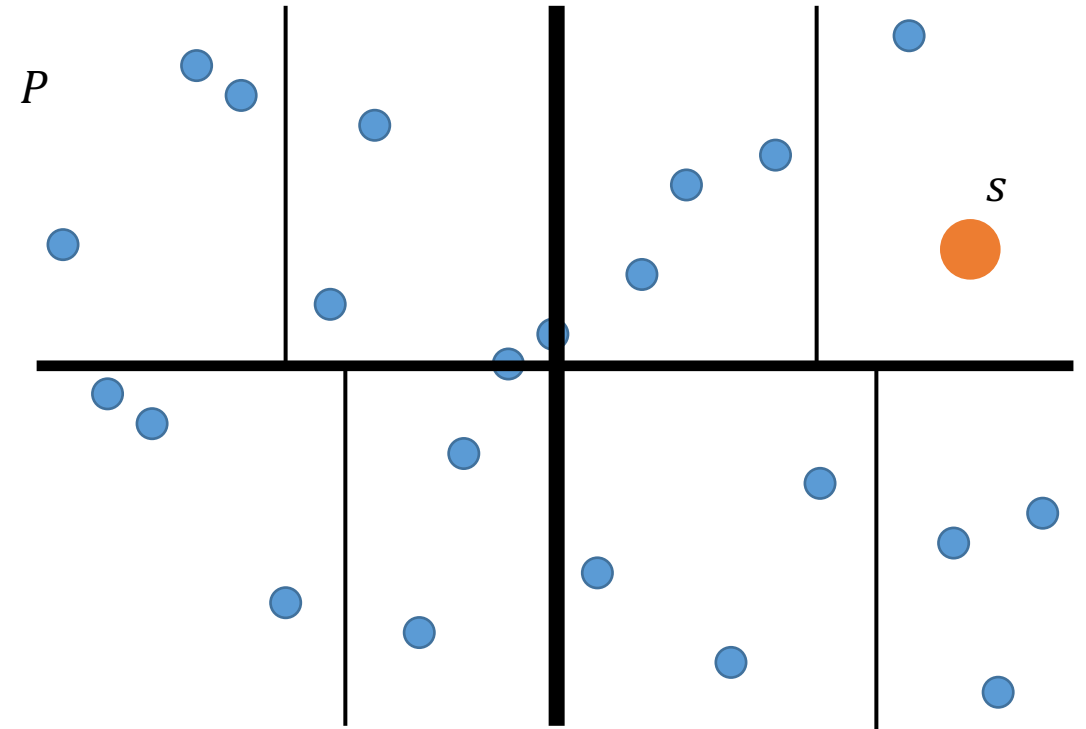
# KD-TREE ALGORITHMS

- CONSTRUCTION
  - Find extrema
  - Divide space in half
    - Ideas?
  - Place points into appropriate half
  - Recurse until termination condition
    - Ideas?

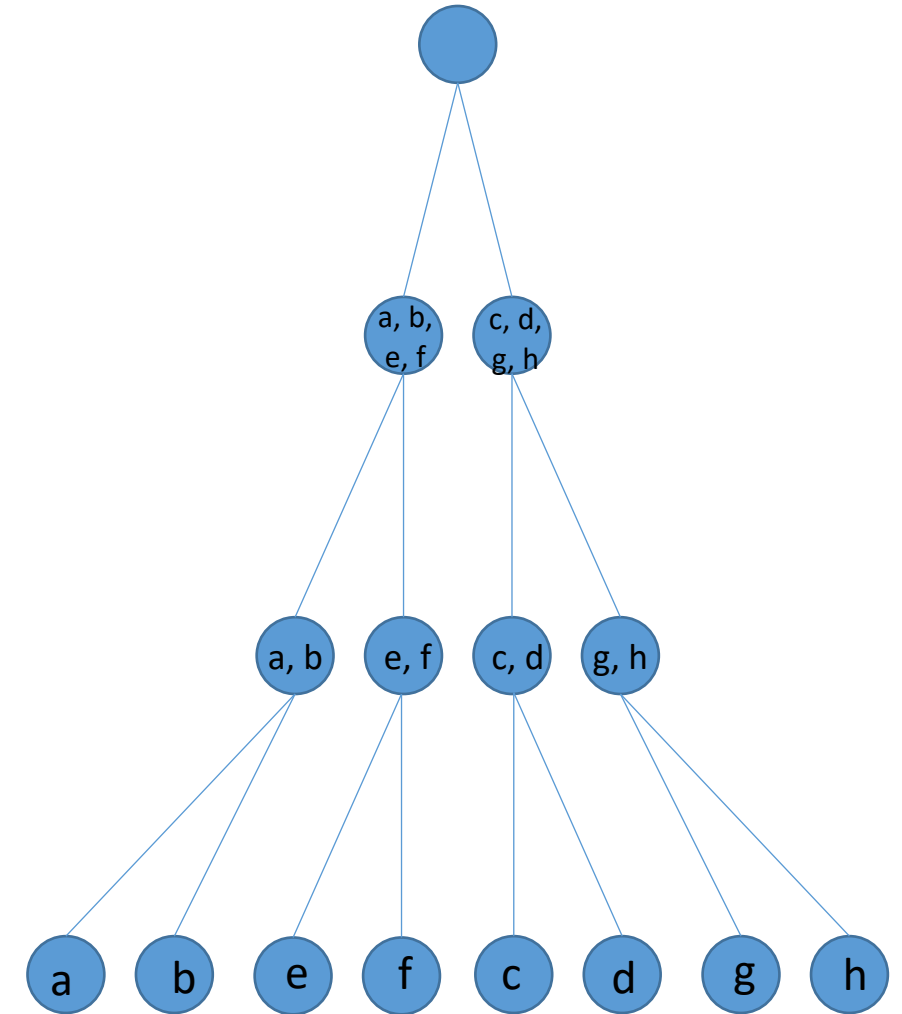
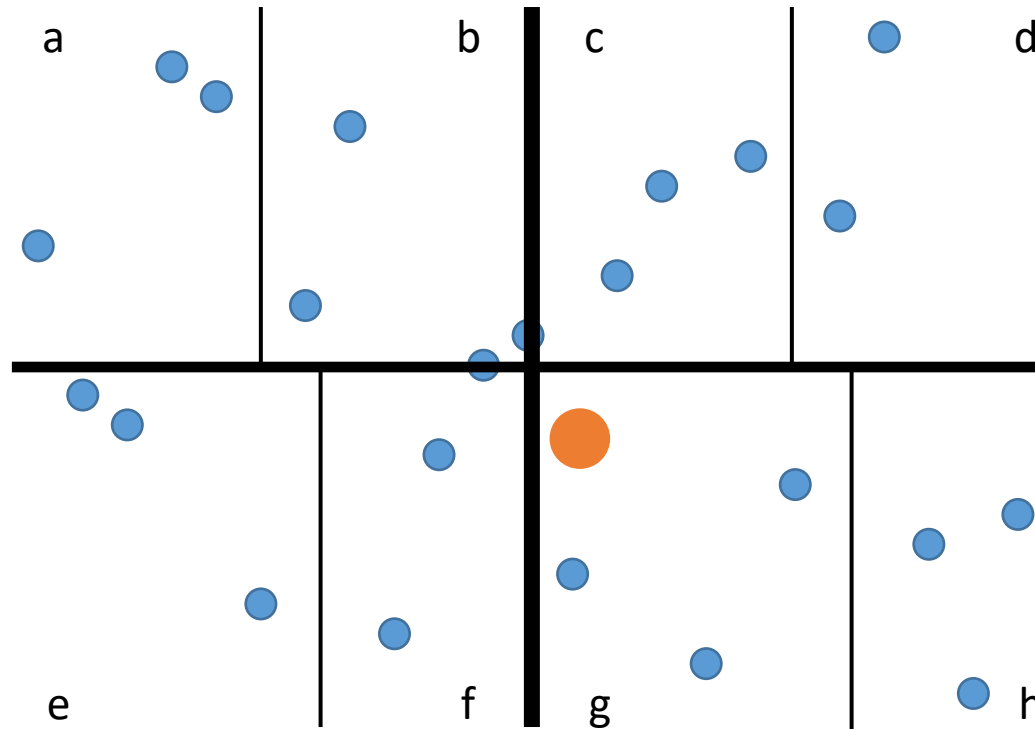


# KD-TREE ALGORITHMS

- SEARCH
  - Given a search location
  - Recursively identify closest leaf
  - Find closest point inside of leaf with distance  $d$
  - Perform the same search with neighboring leaves



# KD-TREE SEARCH EXAMPLE



# KD-TREE

- CONSTRUCTION:
  - $O(n \log n)$
- SEARCH
  - Average case:  $O(\log n + k)$ , where  $k$  is the size of the leaf nodes
  - Worst Case:  $O(n)$
- SPACE:
  - $O(n)$

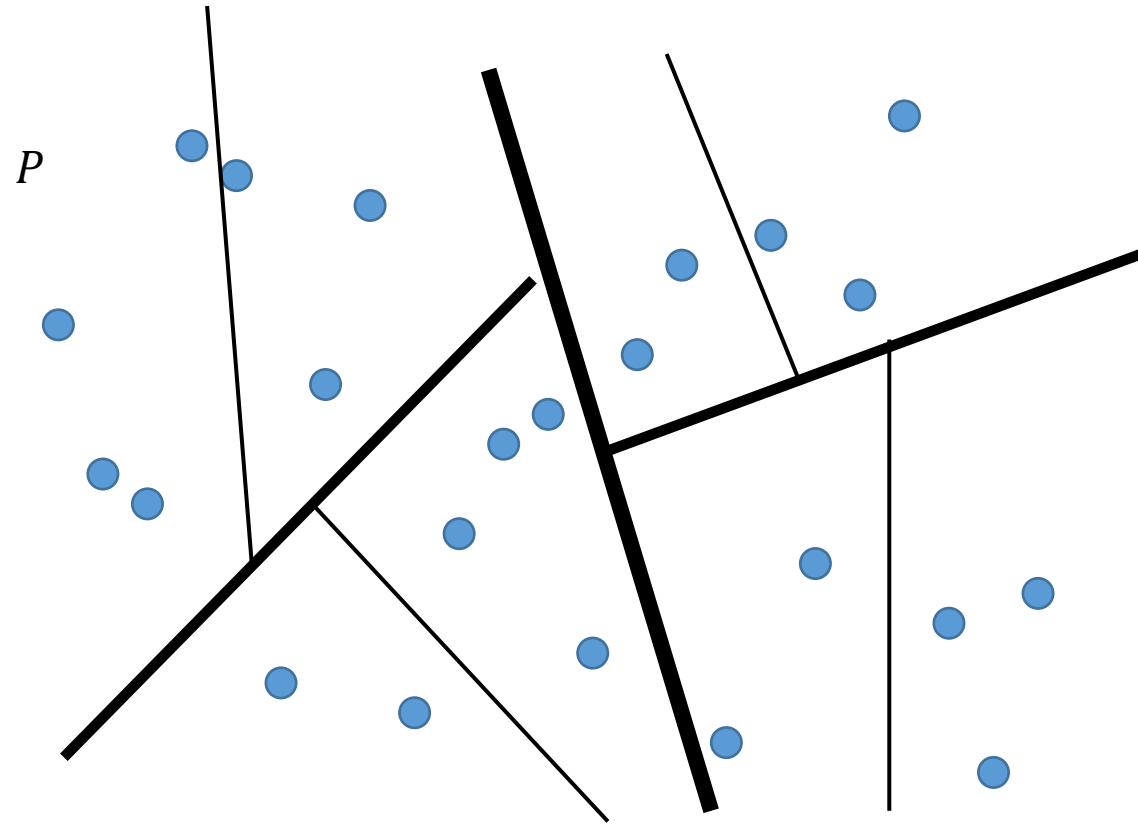


# KD-TREE

- EXAMPLES OF FAILURE CASES?

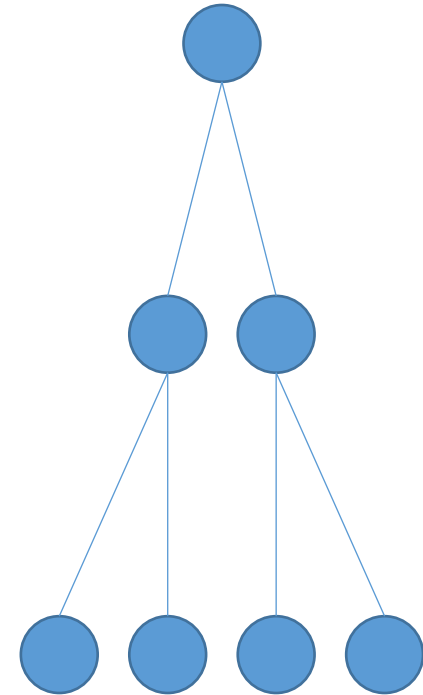


# BINARY SPACE PARTITION (BSP)



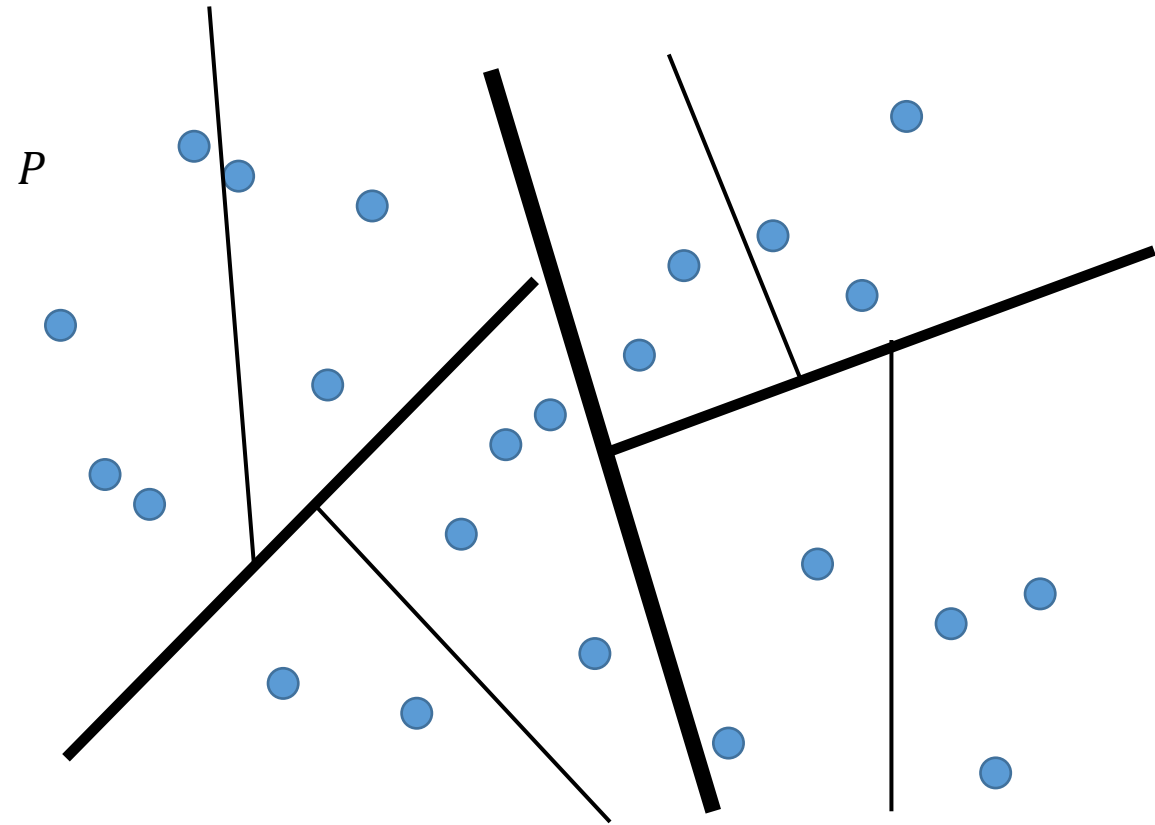
# BSP ALGORITHMS

- CONSTRUCTION
  - Find extrema
  - Divide space in half
  - Place points into appropriate half space
  - Recurse until termination condition



# BSP ALGORITHMS

- SEARCH
  - Given a search location
  - Recursively identify closest leaf
  - Find closest point inside of leaf with distance  $d$
  - Perform the same search with neighboring leaves





# BINARY SPACE PARTITION

- CONSTRUCTION
  - $O(n \log n)$
- SEARCH
  - Average case:  $O(\log n + k)$ 
    - where  $k$  is the size of the leaf nodes
  - Worst Case:  $O(n)$
- SPACE:
  - $O(n)$



# BINARY SPACE PARTITION

- EXAMPLES OF FAILURE CASES?



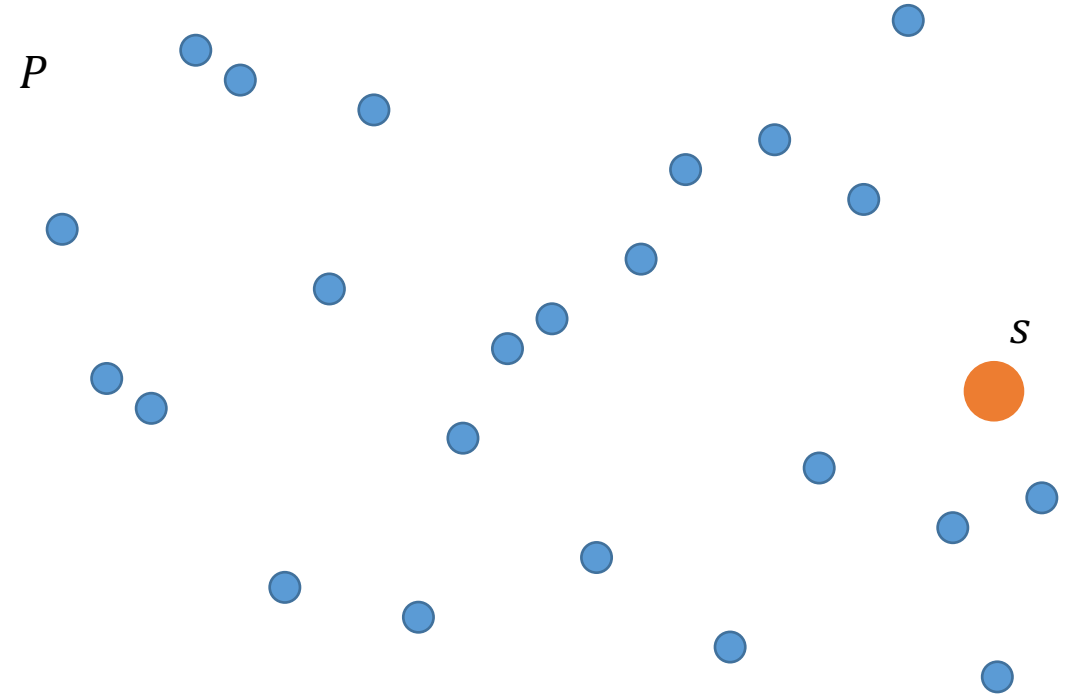
# TASKS

- CLOSEST POINT SEARCH
  - Task covered thus far
- K-NEAREST NEIGHBORS SEARCH (NEXT)
- RANGE SEARCH
- CLUSTERING (WE'LL TALK ABOUT THIS NEXT LECTURE)



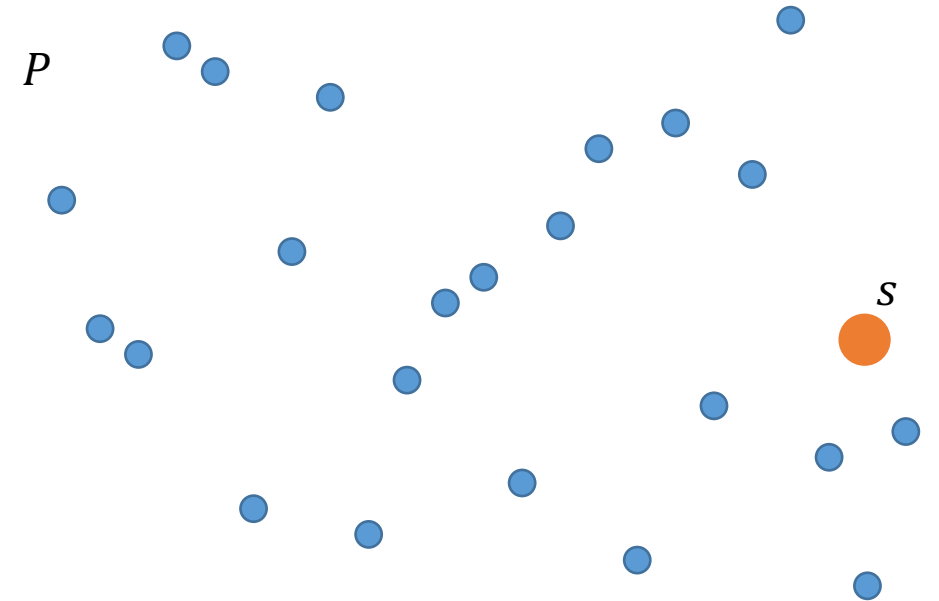
# K-NEAREST NEIGHBOR SEARCH

- PROBLEM: GIVEN A SET OF POINTS  $P$ , FIND THE K-NEAREST NEIGHBORS EFFICIENTLY
- IDEAS?



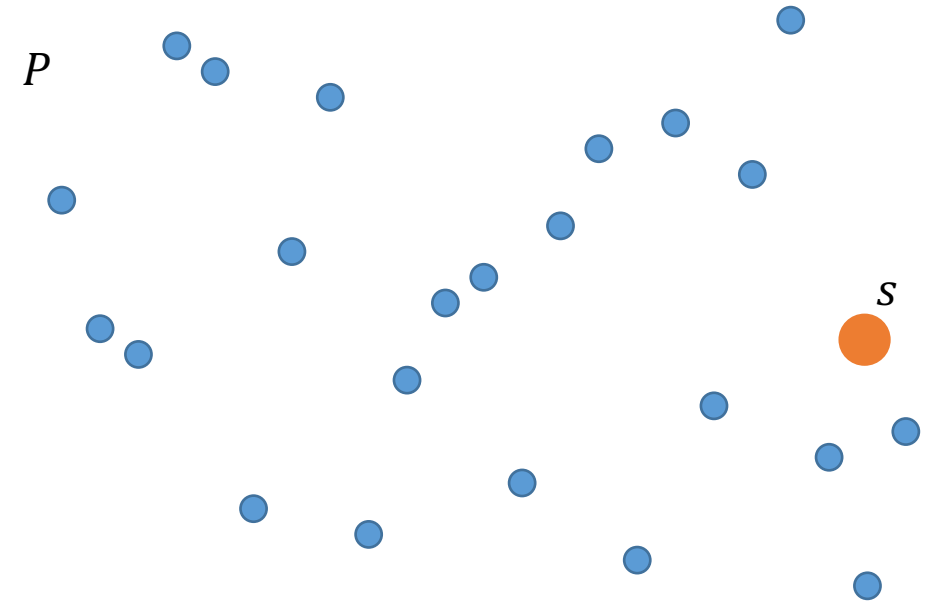
# K-NEAREST NEIGHBOR SEARCH

- USE SPATIAL PARTITIONING OF YOUR CHOICE
- KEEP A LIST OF K LENGTH FOR THE CLOSEST POINTS
- SEARCH PERFORMED SIMILARLY TO CLOSEST POINT SEARCH, EXCEPT THAT OUR STOPPING CONDITION IS ON THE FURTHEST POINT IN THE LIST
- HOW DO WE STORE THE K POINTS MOST EFFICIENTLY?



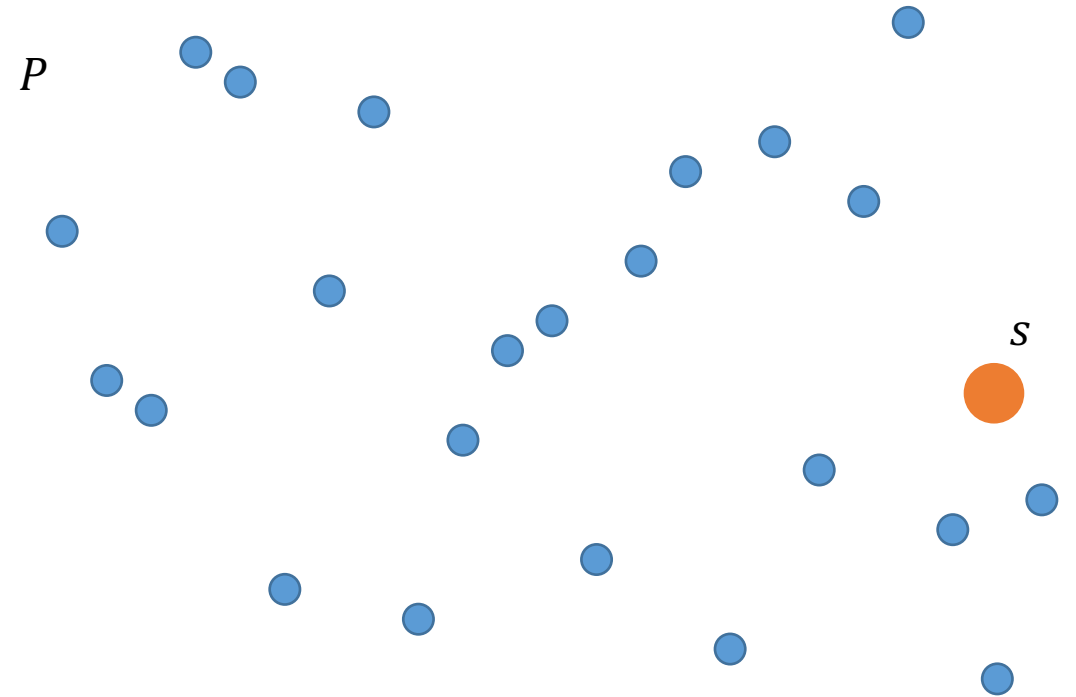
# K-NEAREST NEIGHBOR SEARCH

- MAKE THE LIST EFFICIENT BY KEEPING IT SORTED
  - use a balanced binary tree— $O(\log k)$  insertion costs
  - Or insertion sort— $O(k)$  insertion cost



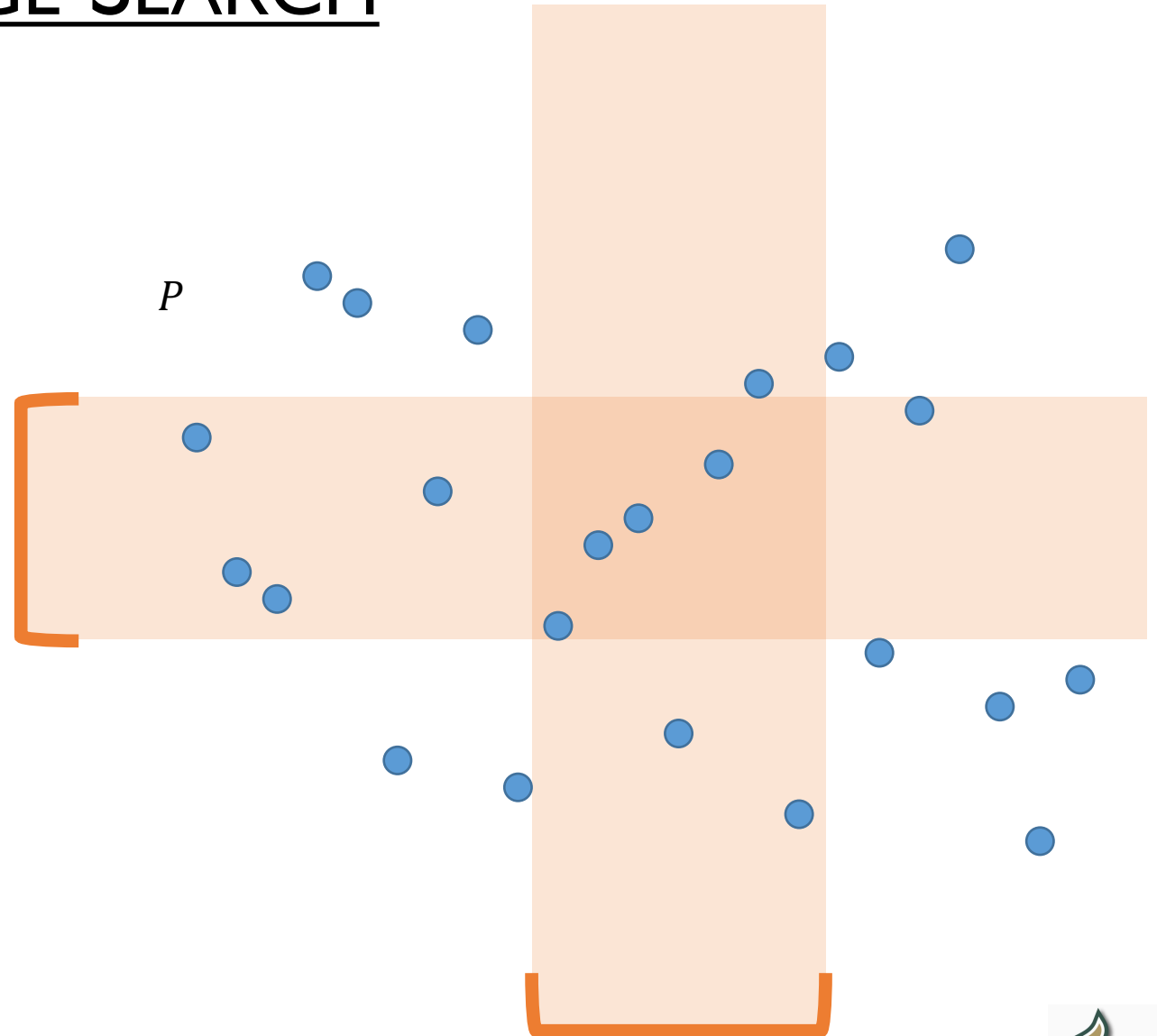
# K-NEAREST NEIGHBOR SEARCH

- PROBLEM: GIVEN A SET OF POINTS  $P$ , FIND THE POINTS WITHIN INTERVALS IN BOTH DIRECTIONS EFFICIENTLY
- IDEAS?



# RANGE SEARCH

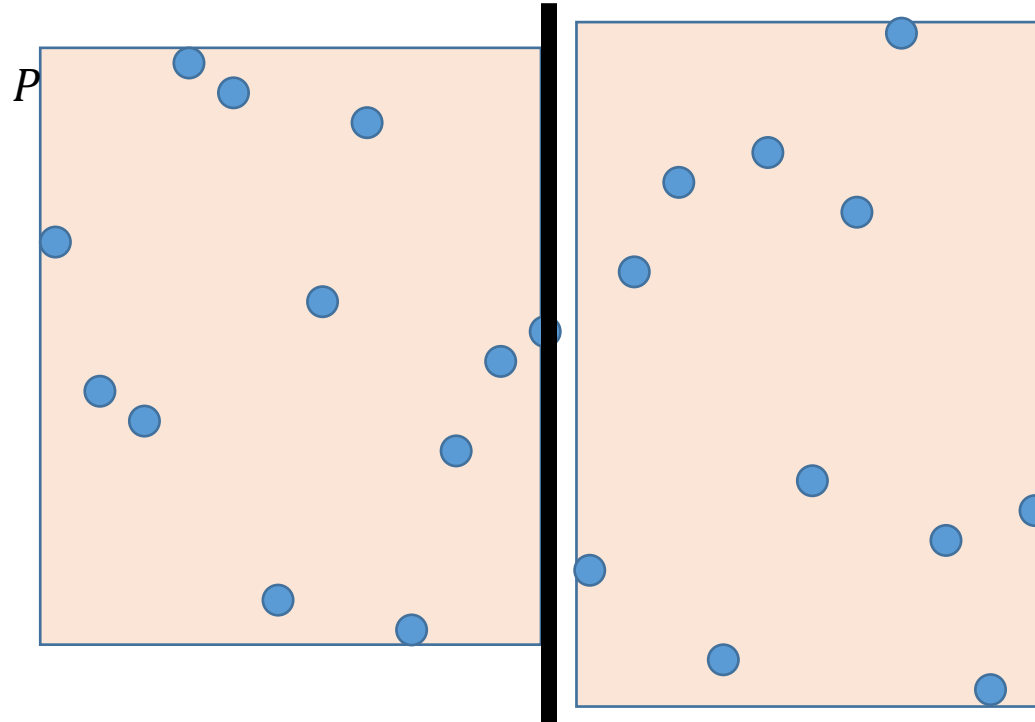
- PROBLEM: GIVEN A SET OF POINTS  $P$ , EFFICIENTLY FIND THE SET OF POINTS WITHIN A SPECIFIED RANGE
- IDEAS?





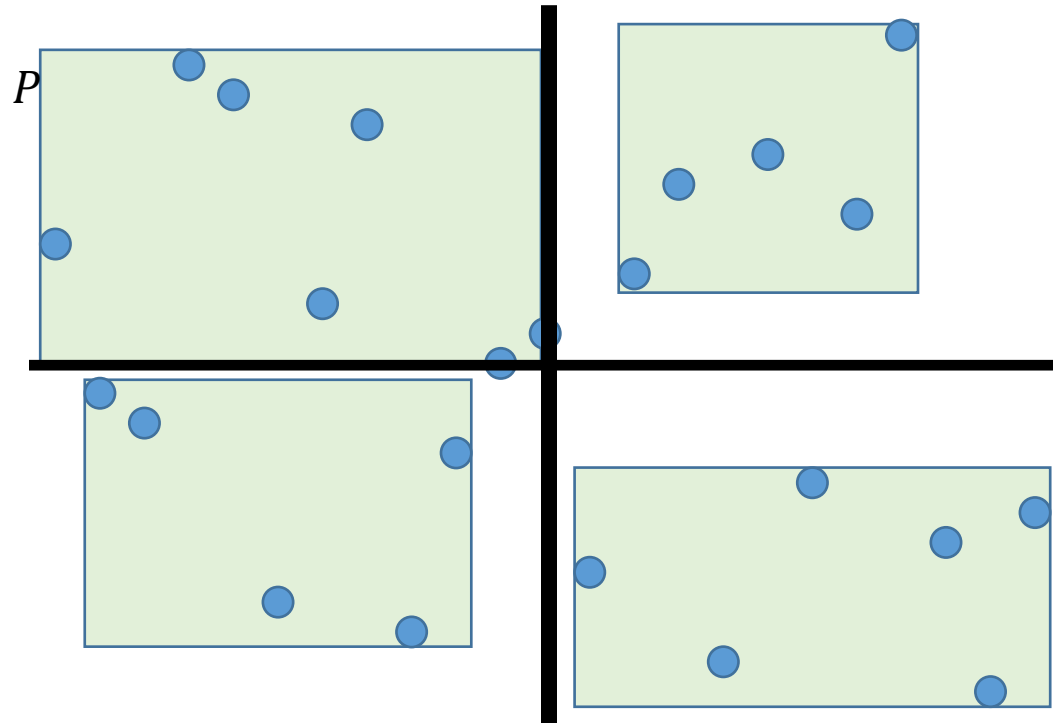
# BOUNDING VOLUME HIERARCHY

- VARIATION ON PREVIOUS APPROACHES THAT STORES THE VOLUME COVERED AT EACH LEVEL OF THE TREE.



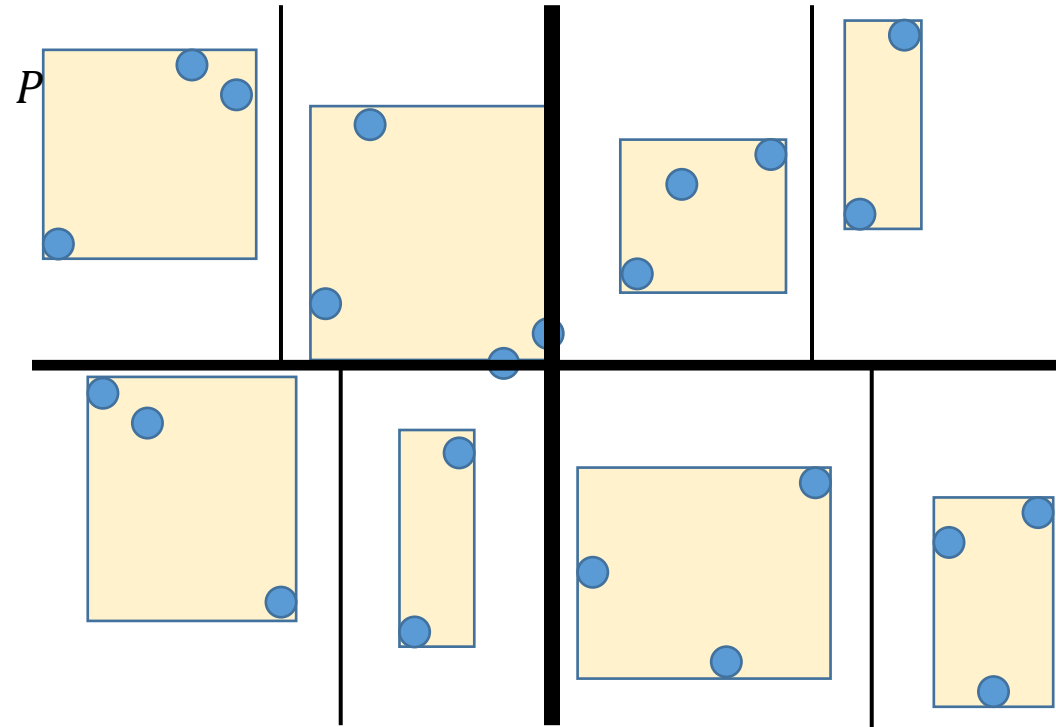
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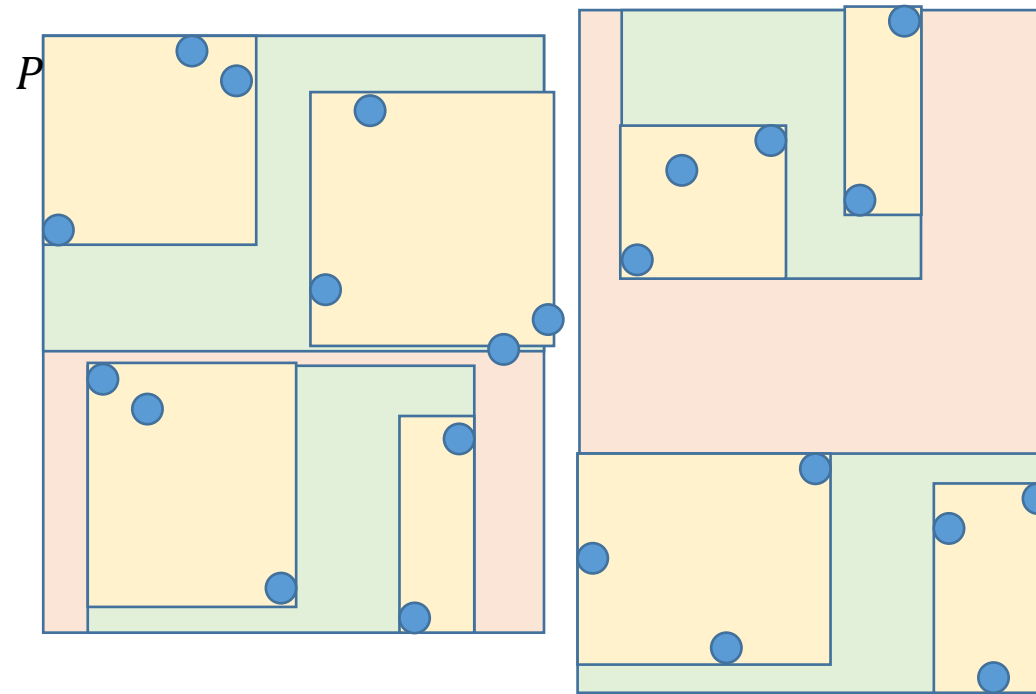
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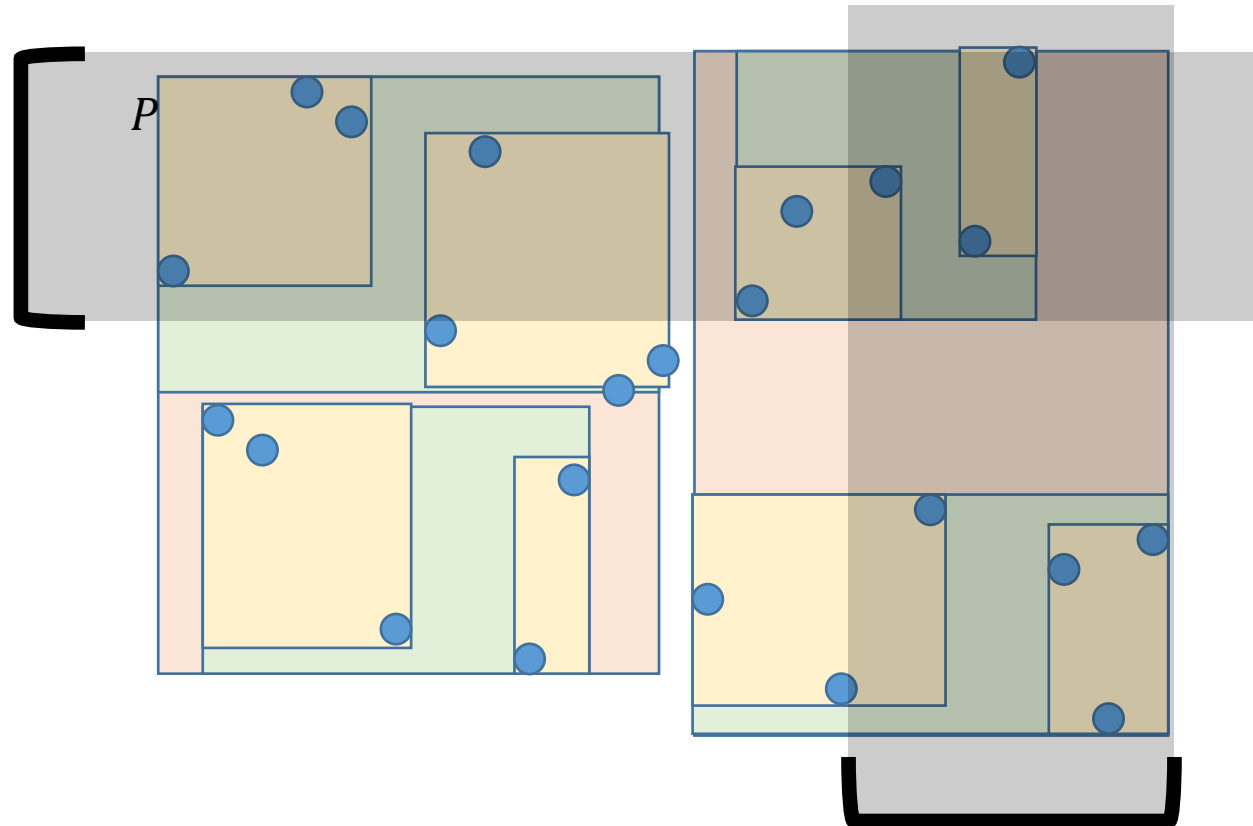
# BOUNDING VOLUME HIERARCHY

- VARIATION ON PREVIOUS APPROACHES THAT STORES THE VOLUME COVERED AT EACH LEVEL OF THE TREE.



# RANGE SEARCH BOUNDING VOLUME HIERARCHY

- VARIATION ON PREVIOUS APPROACHES THAT STORES THE VOLUME COVERED AT EACH LEVEL OF THE TREE.



# BOUNDING VOLUME HIERARCHY

- BUILDING DATA STRUCTURE:  $O(n \log n)$
- SEARCH
  - Average case:  $O(\log n + k)$ 
    - where  $k$  is the size of the leaf nodes
  - Worst Case:  $O(n)$
- SPACE: IN  $O(n)$



