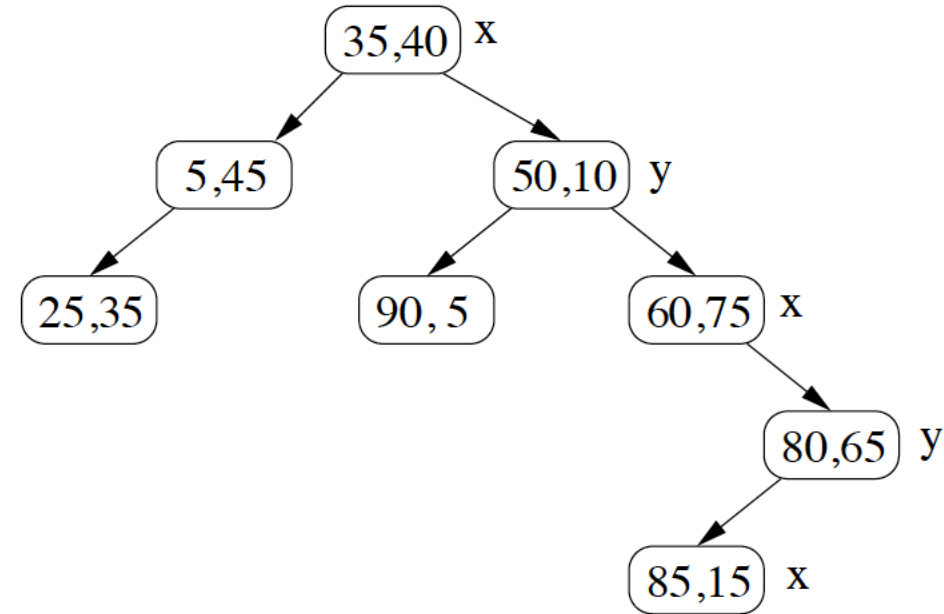
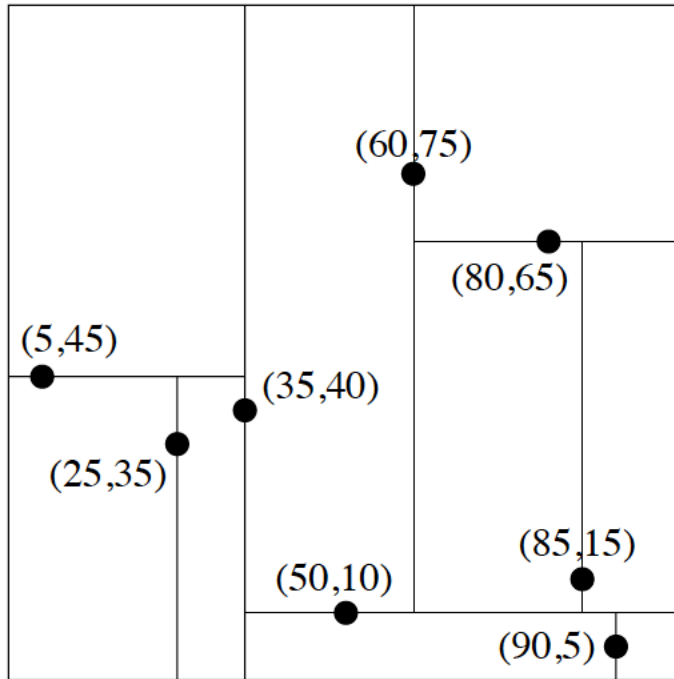


# Point / Point-Region Quadtrees

CMSC 420

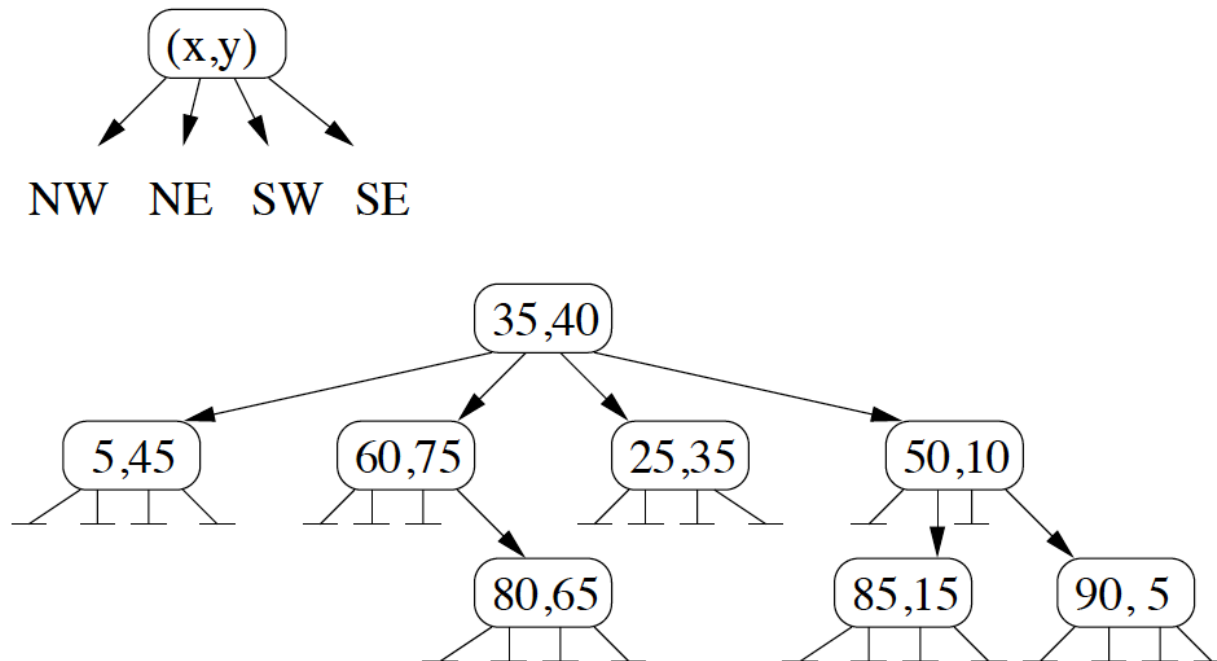
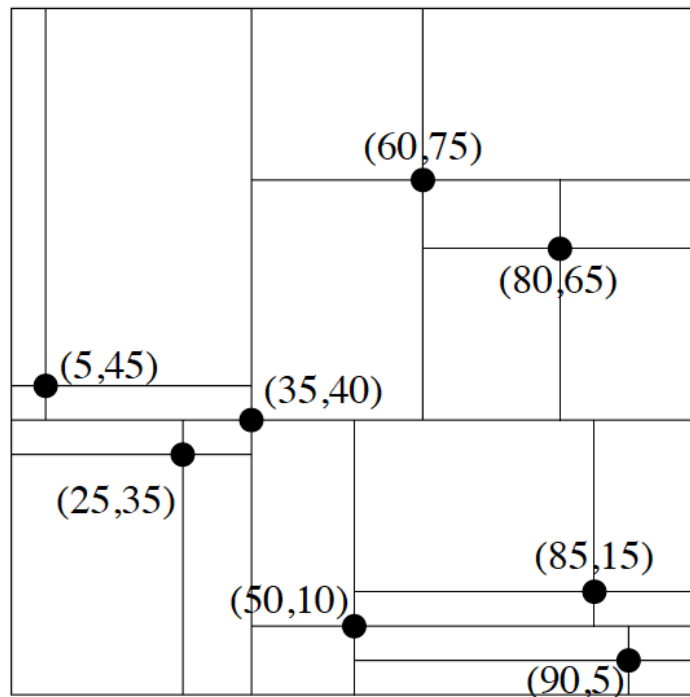
# Reminder: KD-Tree Structure

- **KD-Trees** are **binary trees** fully defined by the data points.



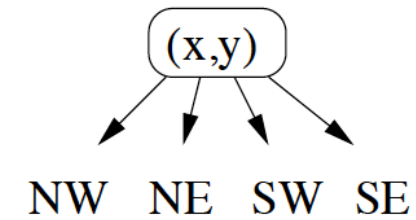
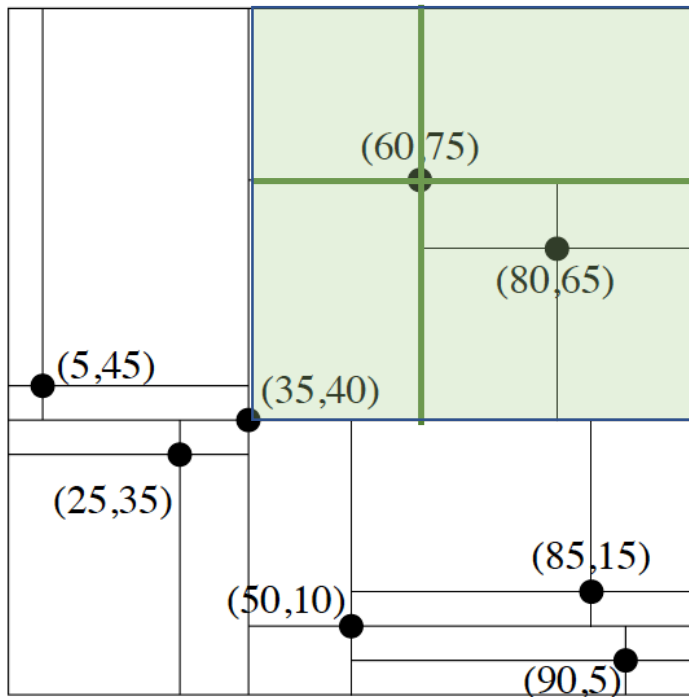
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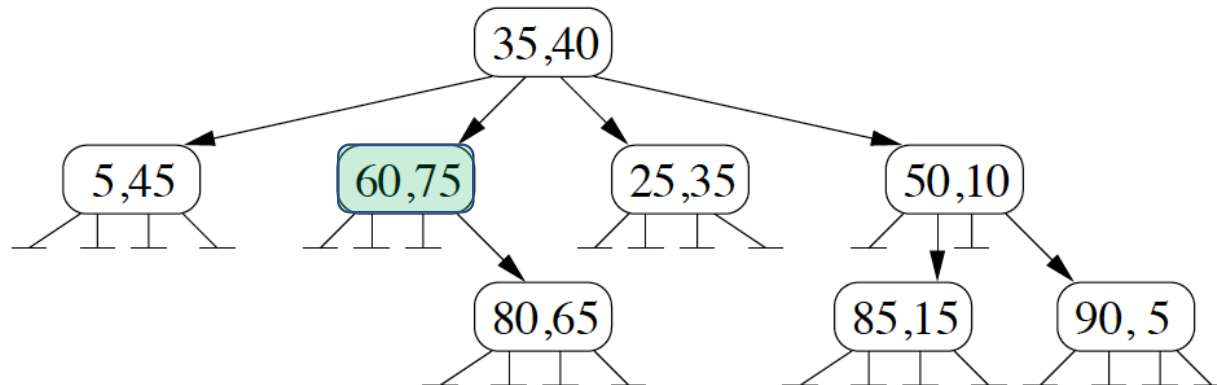


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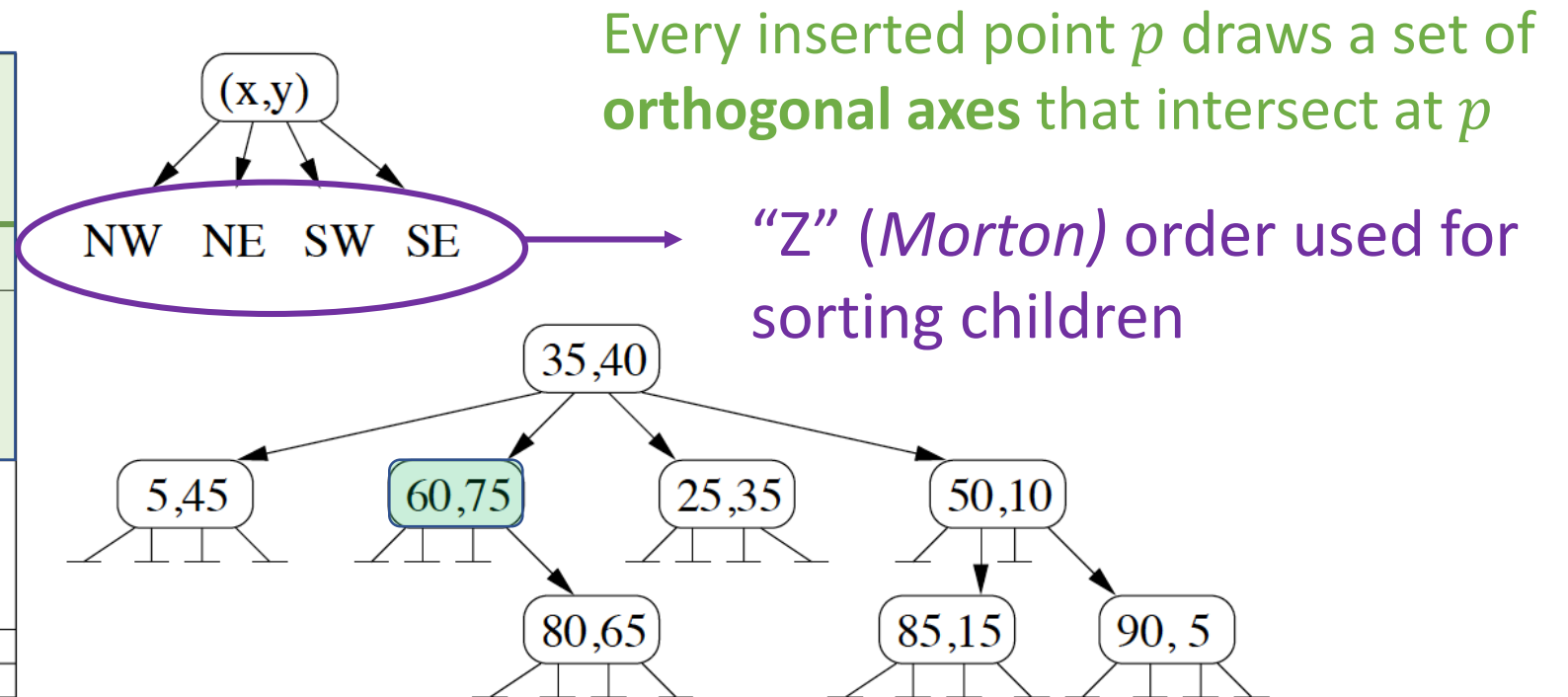
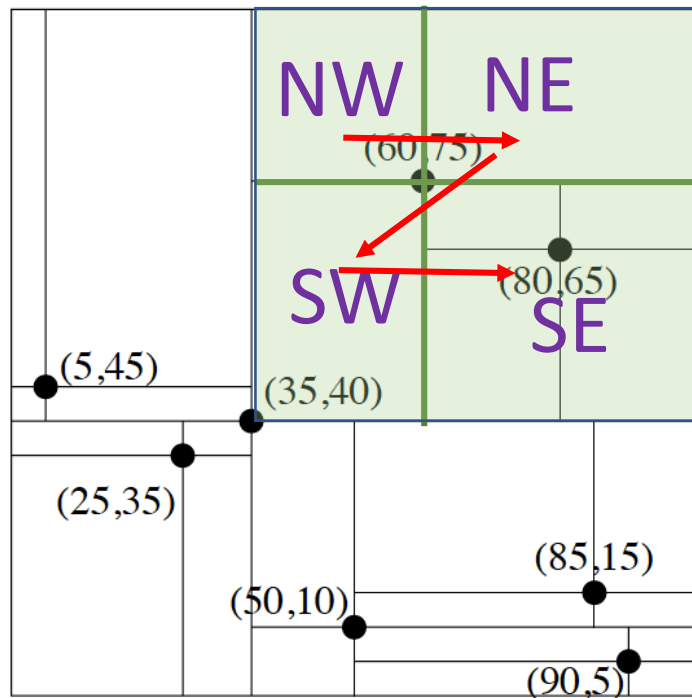


Every inserted point  $p$  draws a set of **orthogonal axes** that intersect at  $p$



# Reminder: KD-Tree Structure

- **Point-Quadrees** are a straightforward **generalization from binary to 4-ary trees!**



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Something  
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Remember: this means *uniform distribution over inputs...*

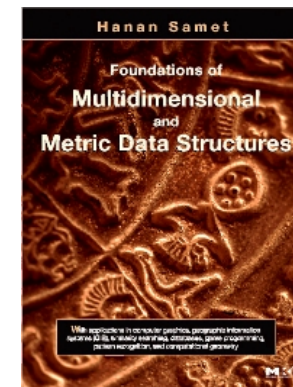


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# Deletion

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- Find the point to be deleted, delete it
- And *re-insert all the points rooted in the relevant subtree* ☹️
- Hanan's book has a better, but *much more complicated* algorithm for Point QuadTree deletion
  - If interested, *ask Jason to post pseudocode on Piazza*



# Range, $m$ -NN, etc...

- Algorithms for **range** and  **$m$ -nearest neighbors** are a straightforward generalization of those in KD-Trees.
- You can still **prune away subtrees** and do **branch-and-bound** and all that cool stuff.

# Main issue with point quadtrees

- Fanout increases **exponentially** with dimension  $d$ .
- For a  $d$  —dimensional space, the Point Quadtree's node has  $2^d$  children.
- In practice, point quadtrees are used only up to 3 dimensions (*oct-trees*).
- KD-Trees are preferred because of their tractable fan-out and deletion efficiency.

# Data-based vs space-based representations

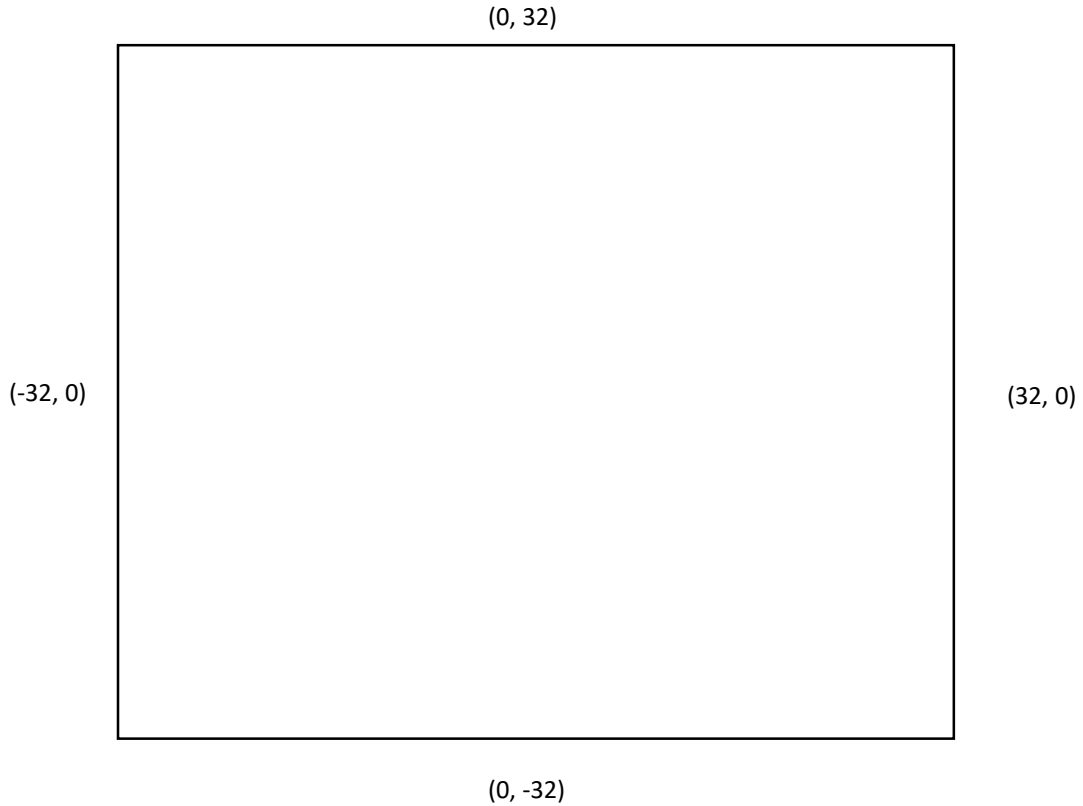
- Point QuadTrees and KD-Trees have the common characteristic that they **split based on data points**.
- This is **not the only way**: we can also build spatial data structures based on **spatial decompositions**!
- Examples: **Region Quadtrees**, **Point-Region (P-R) QuadTrees**
  - We will talk about **PR Quadtrees** this semester and post some old slides on **Region QuadTrees** for your enlightenment

# Main idea

- We will assume a large enough “bounding square” of dimensions  $2^{k-1} \times 2^{k-1}$  for some  $k \in \mathbb{Z}$ .
- Our points are **agoraphobic**: they will need their own private space.
  - To that end, **every time we insert a point, we will subdivide the half-planes by 4** (or 2, if you’re only looking at the edge) **until all the points are separated!**
- This will lead to a tree of fanout 4 with 3 kinds of nodes:
  - White nodes (no point contained)
  - **Black** nodes (point contained)
  - Gray (*mixed*) nodes (with at least one non- White **child node**).

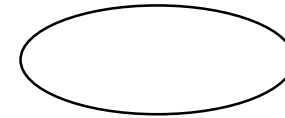
# Example

Space



Suppose we select  $k = 6 \Rightarrow 2^k = 64...$

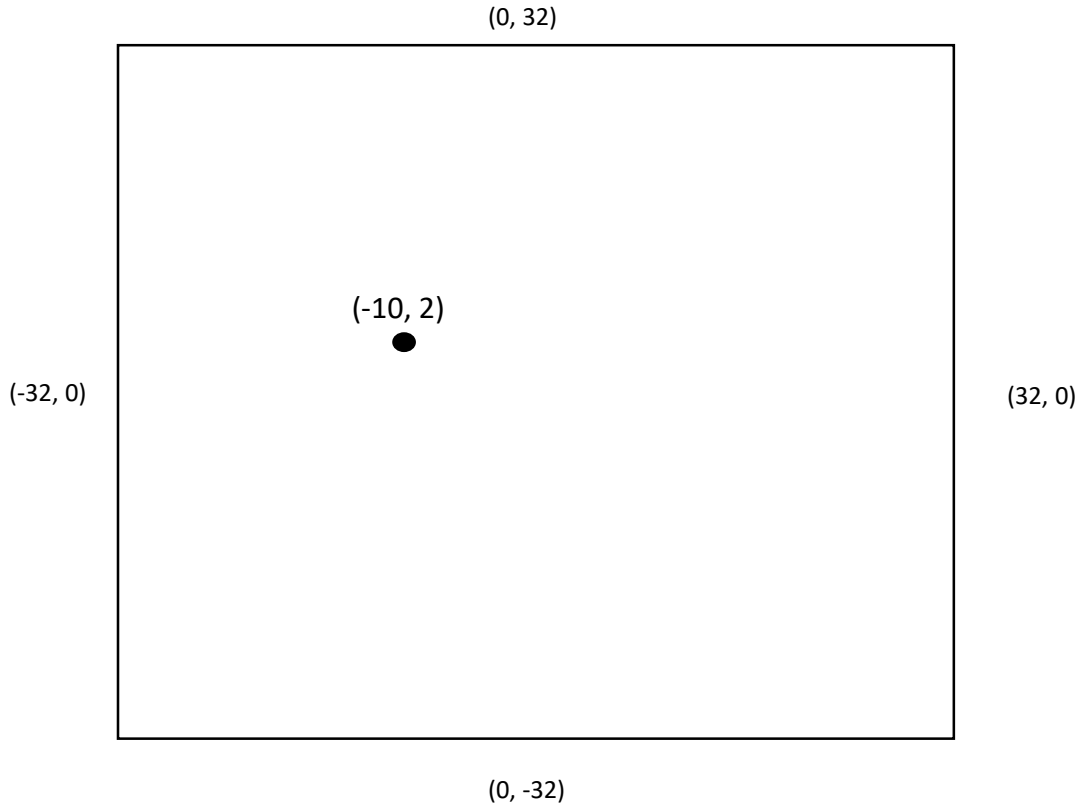
PR-QuadTree



No stored points yet: All we have is a white node!

# Example

Space



PR-QuadTree

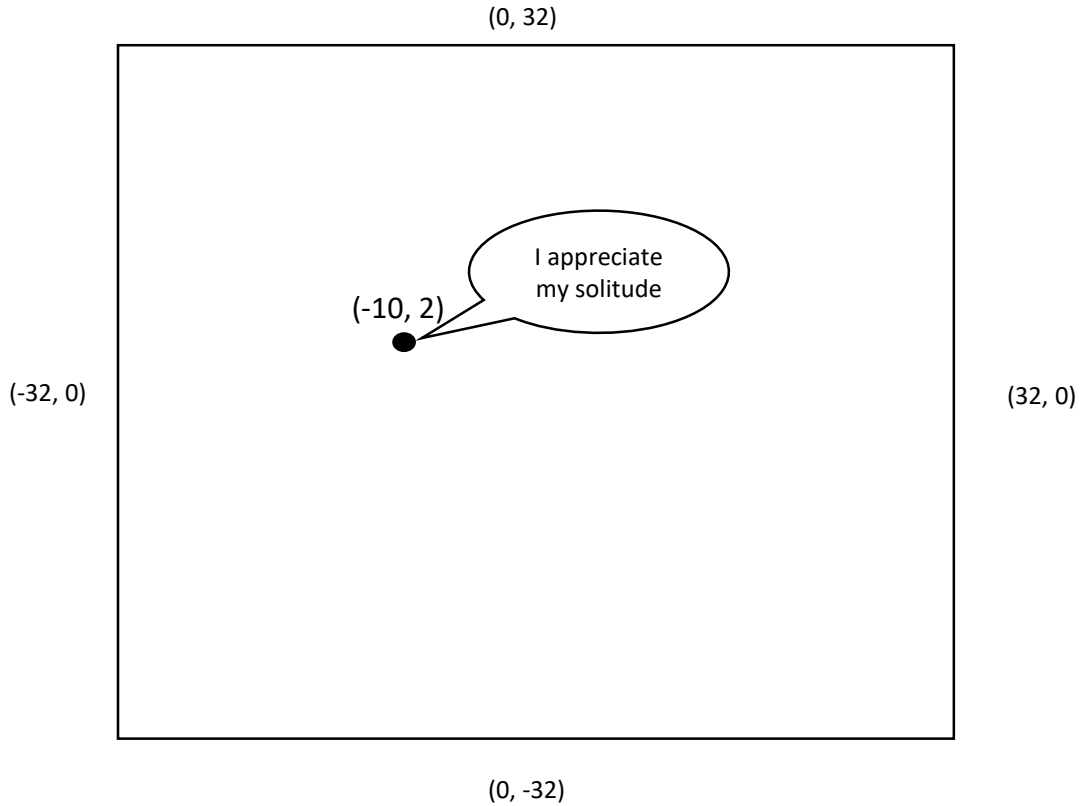


Stored a single point: Entire  
quadtree is now a **black**  
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# Example

Space



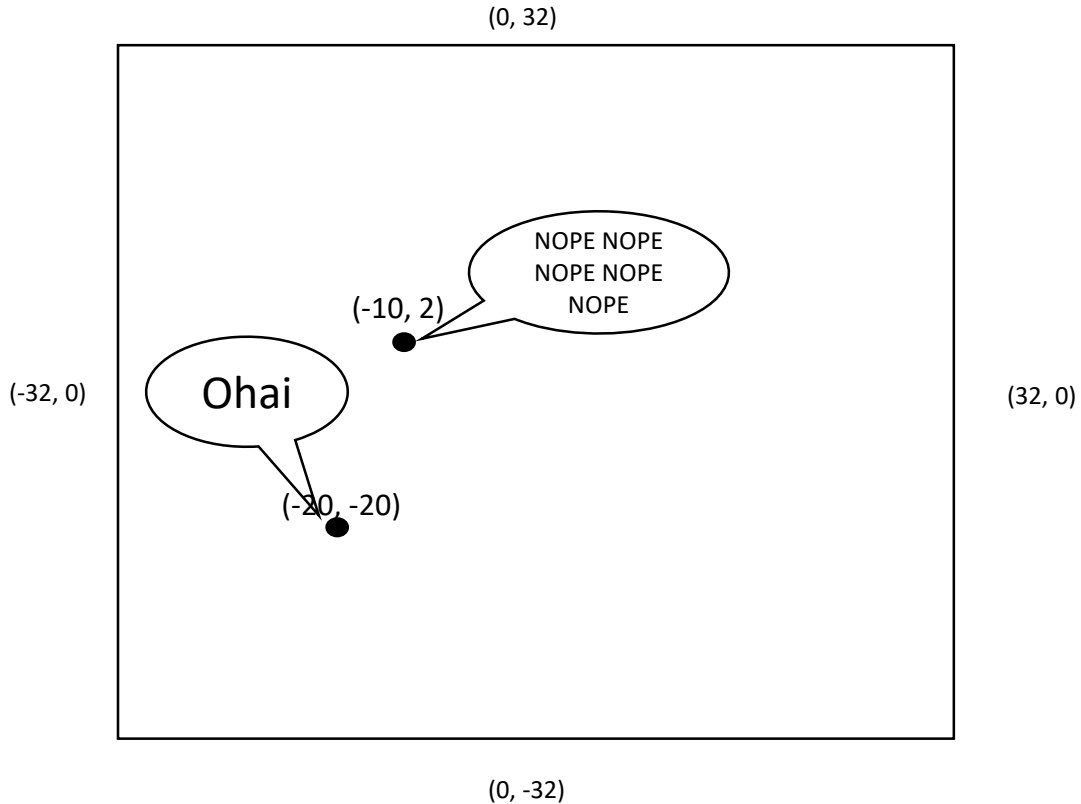
PR-QuadTree



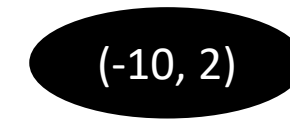
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## Space



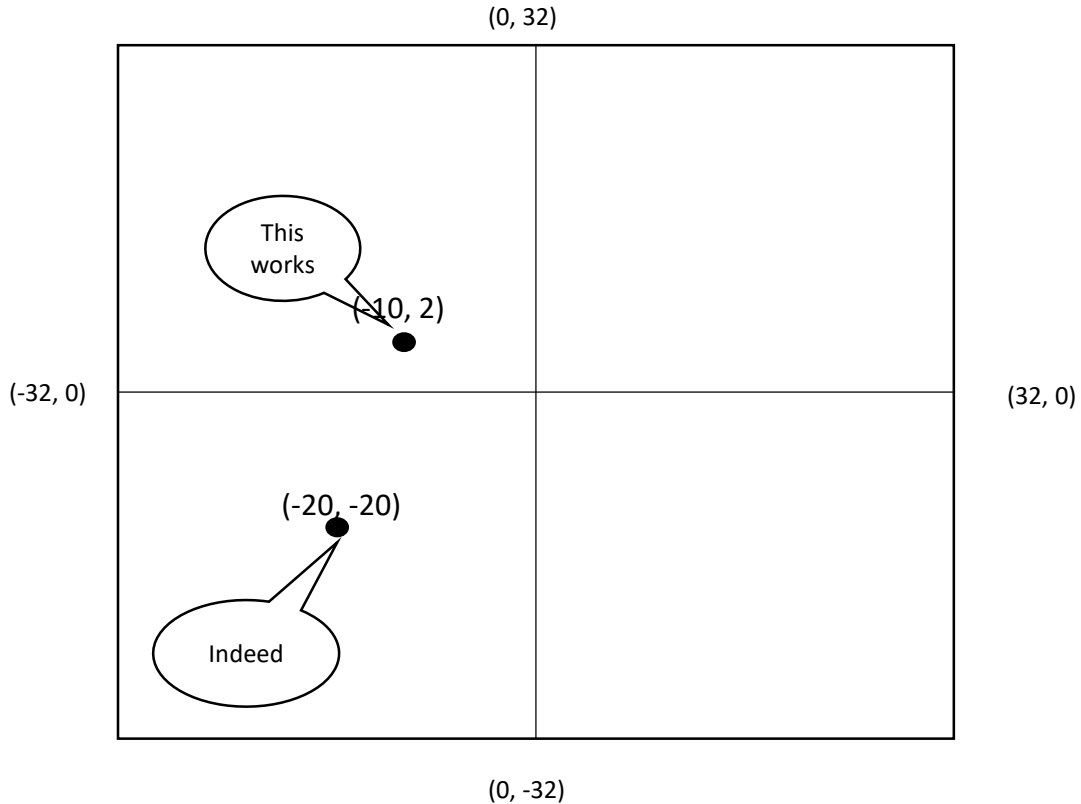
## PR-QuadTree



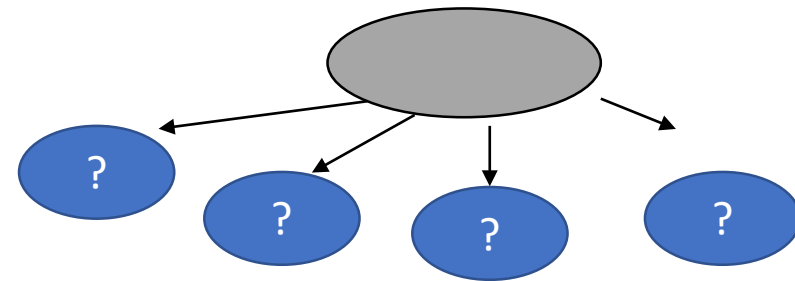
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- Split into gray node and 4 children.

# Example

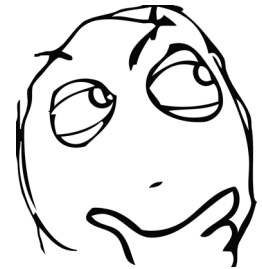
## Space



## PR-QuadTree



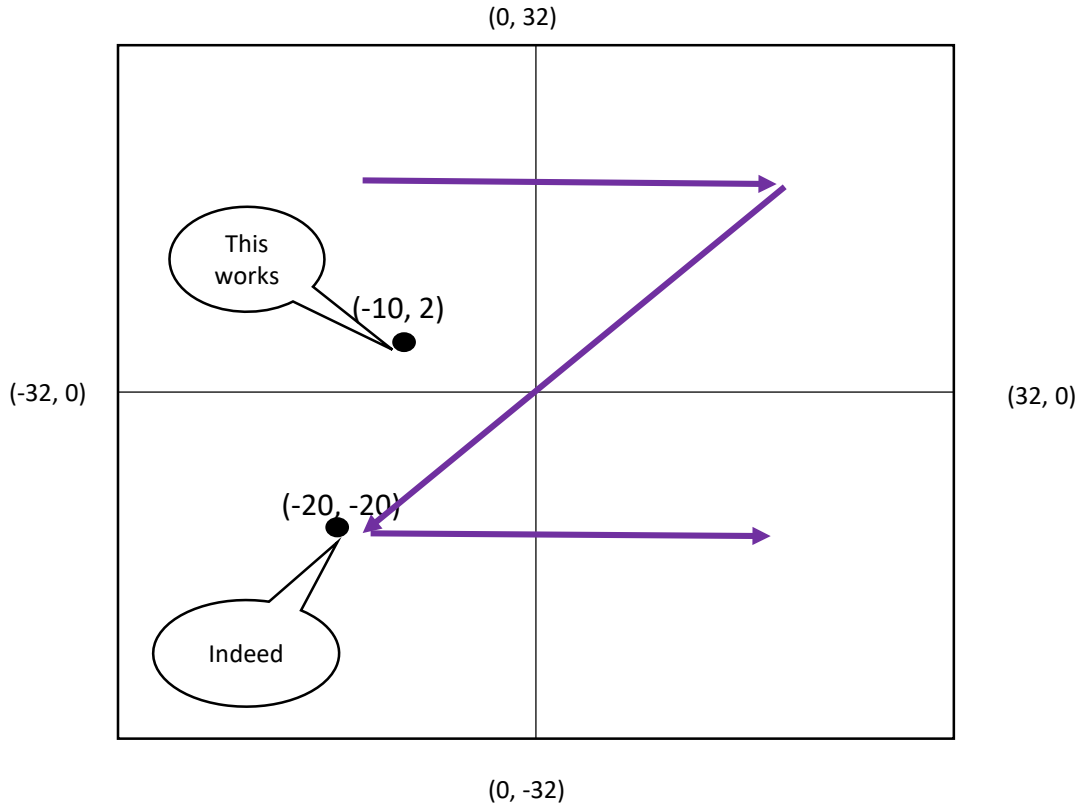
Colors and contents of these nodes?



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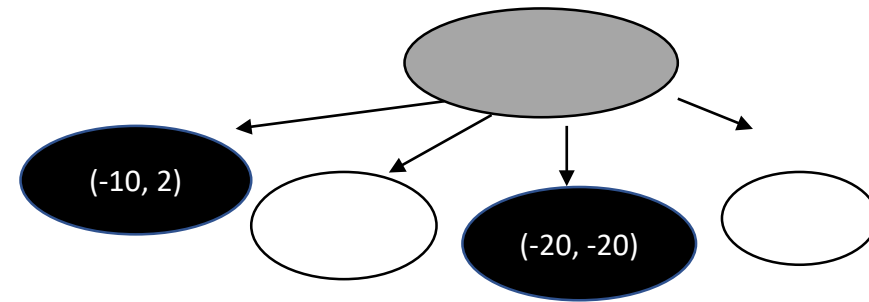
# Example

## Space



*We still do Morton (“Z”) order for traversing the children!*

## PR-QuadTree



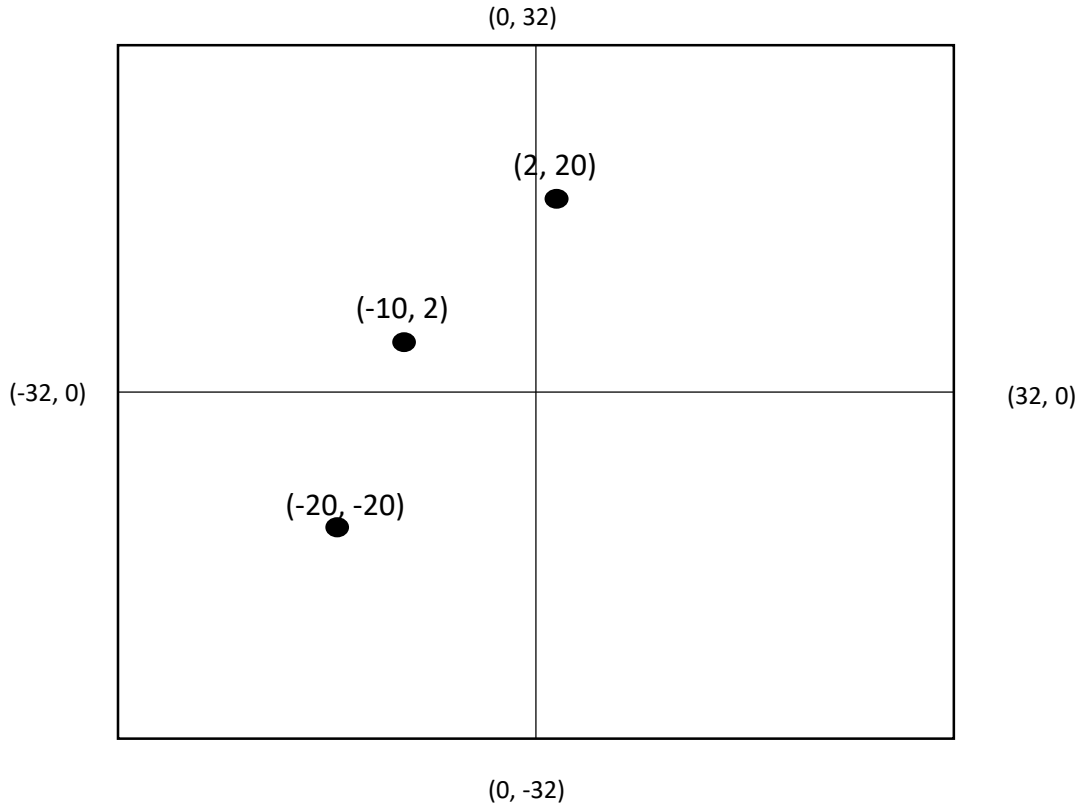
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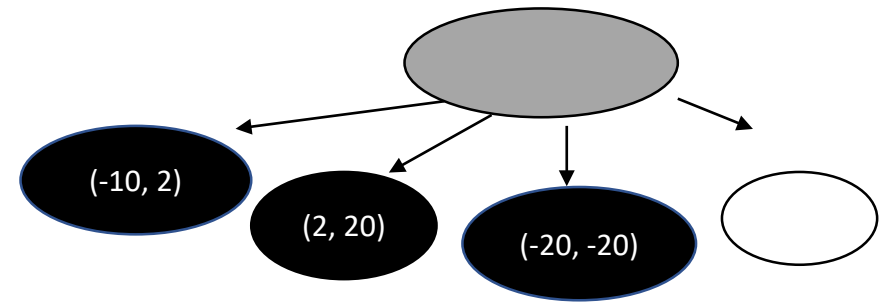
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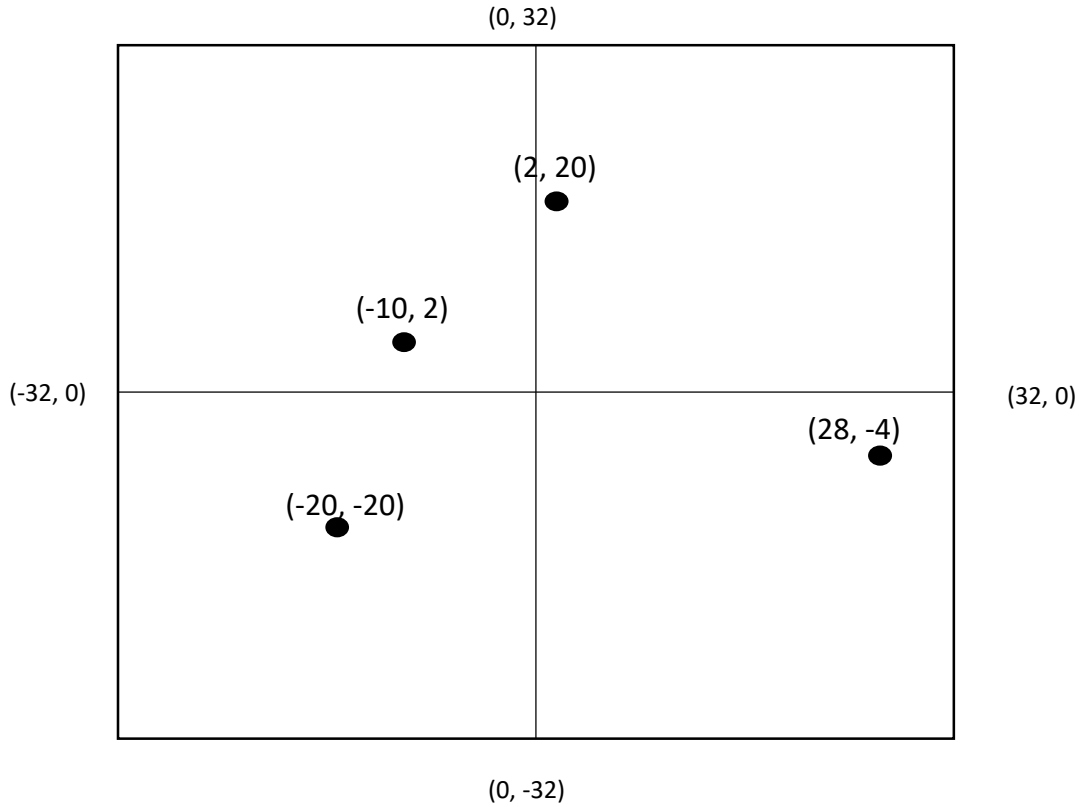


PR-QuadTree

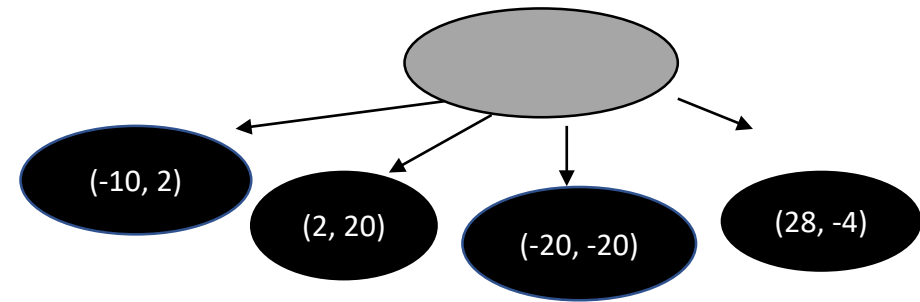


# Example

## Space

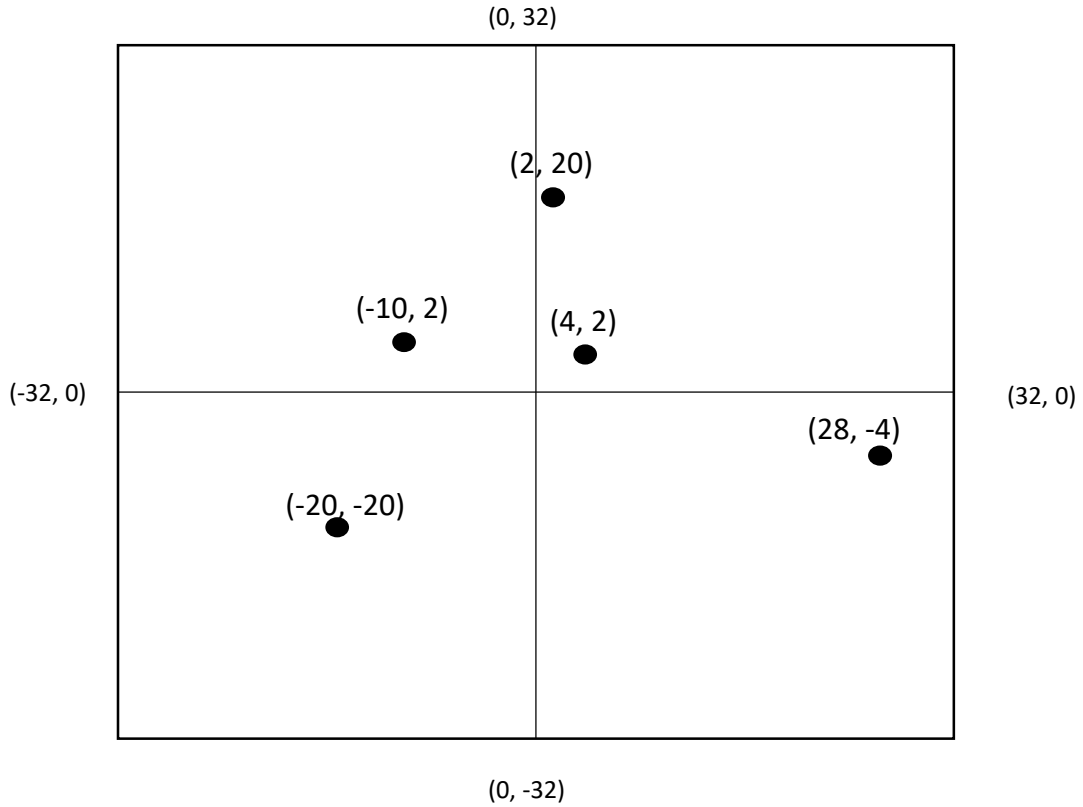


## PR-QuadTree

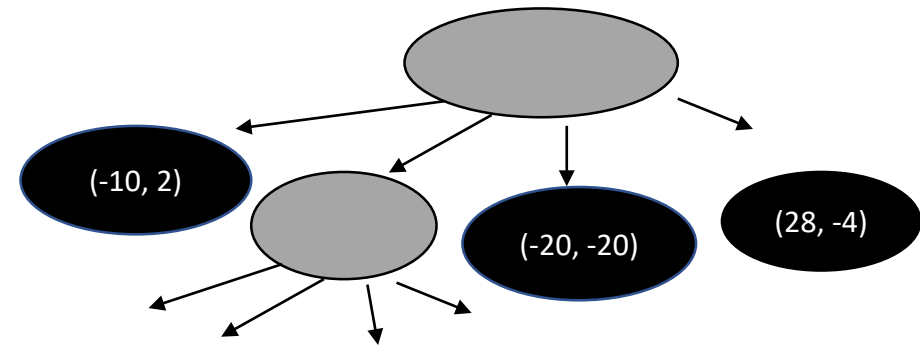


# Example

## Space

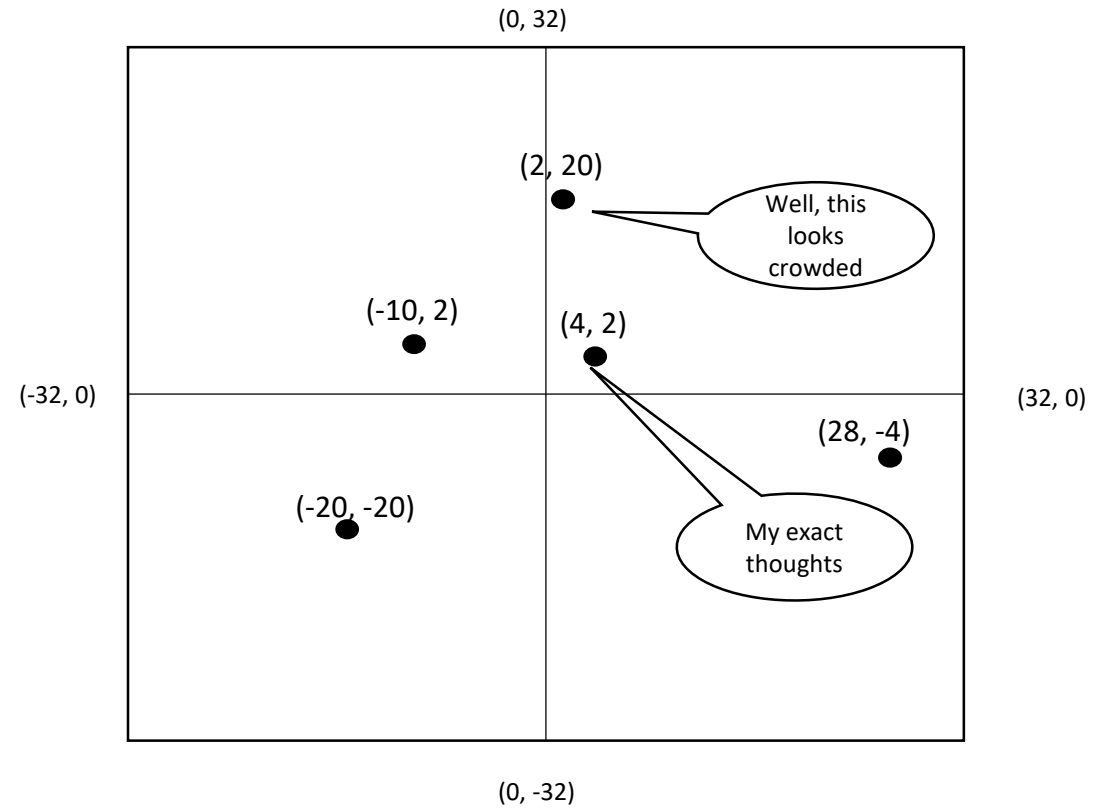


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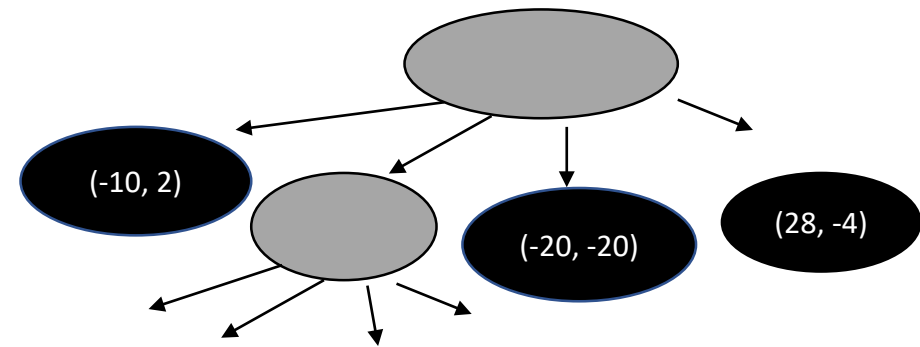


# Example

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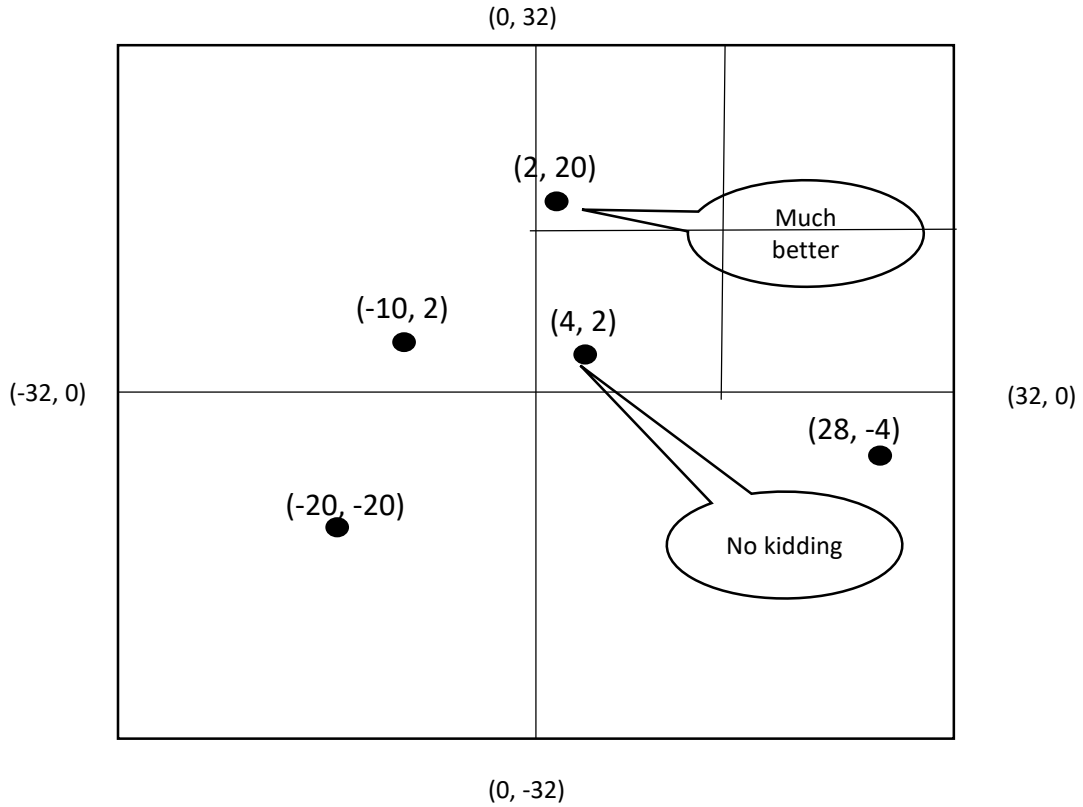
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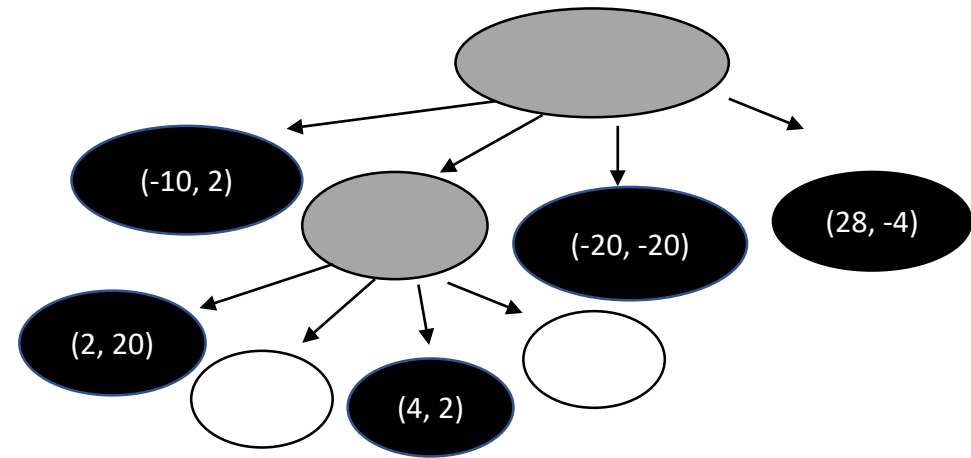


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## Space

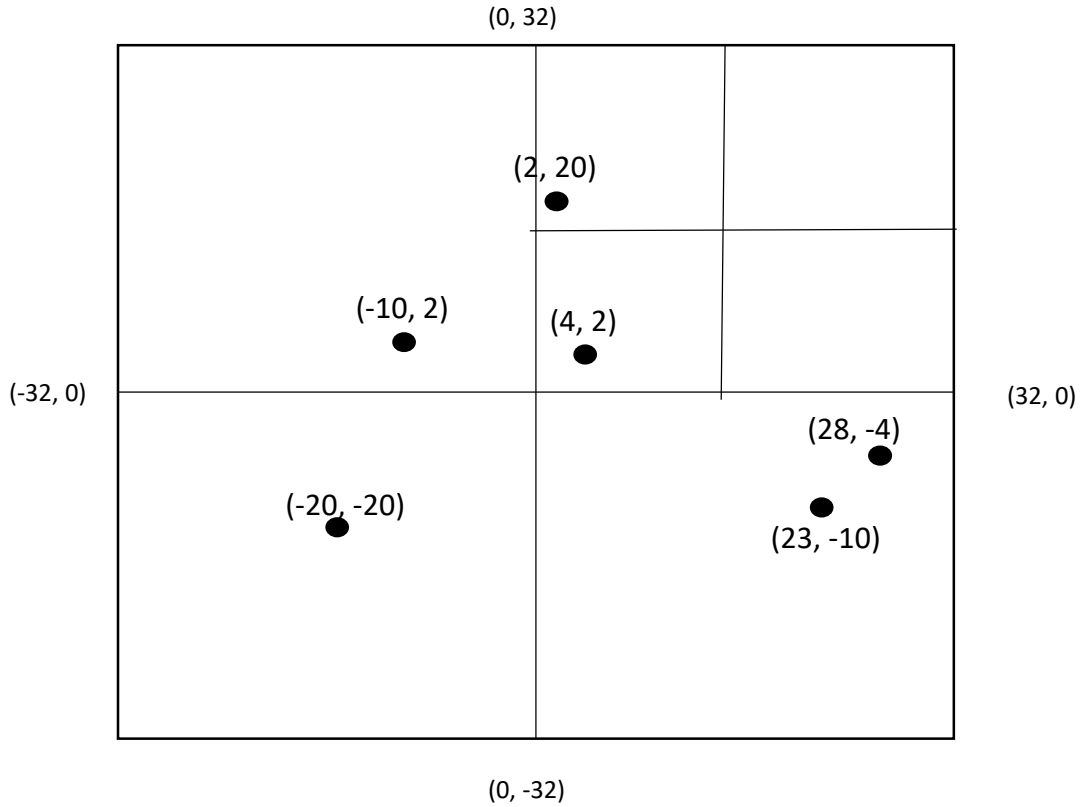


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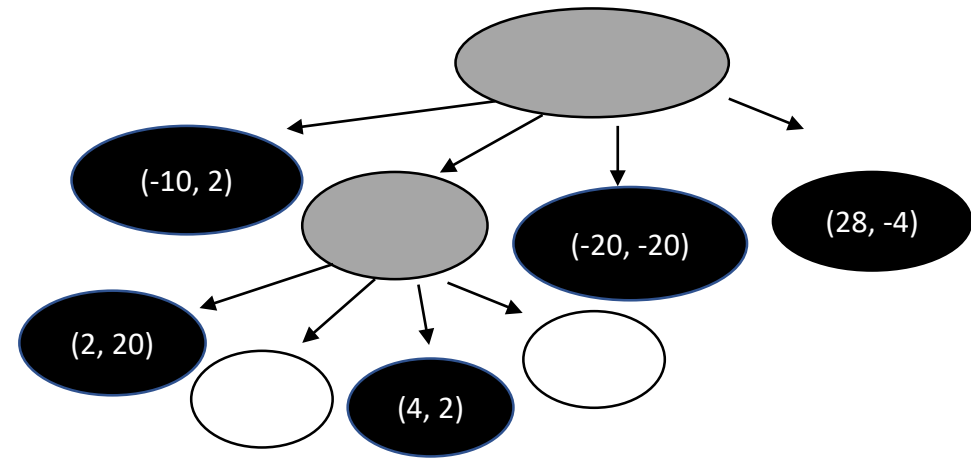


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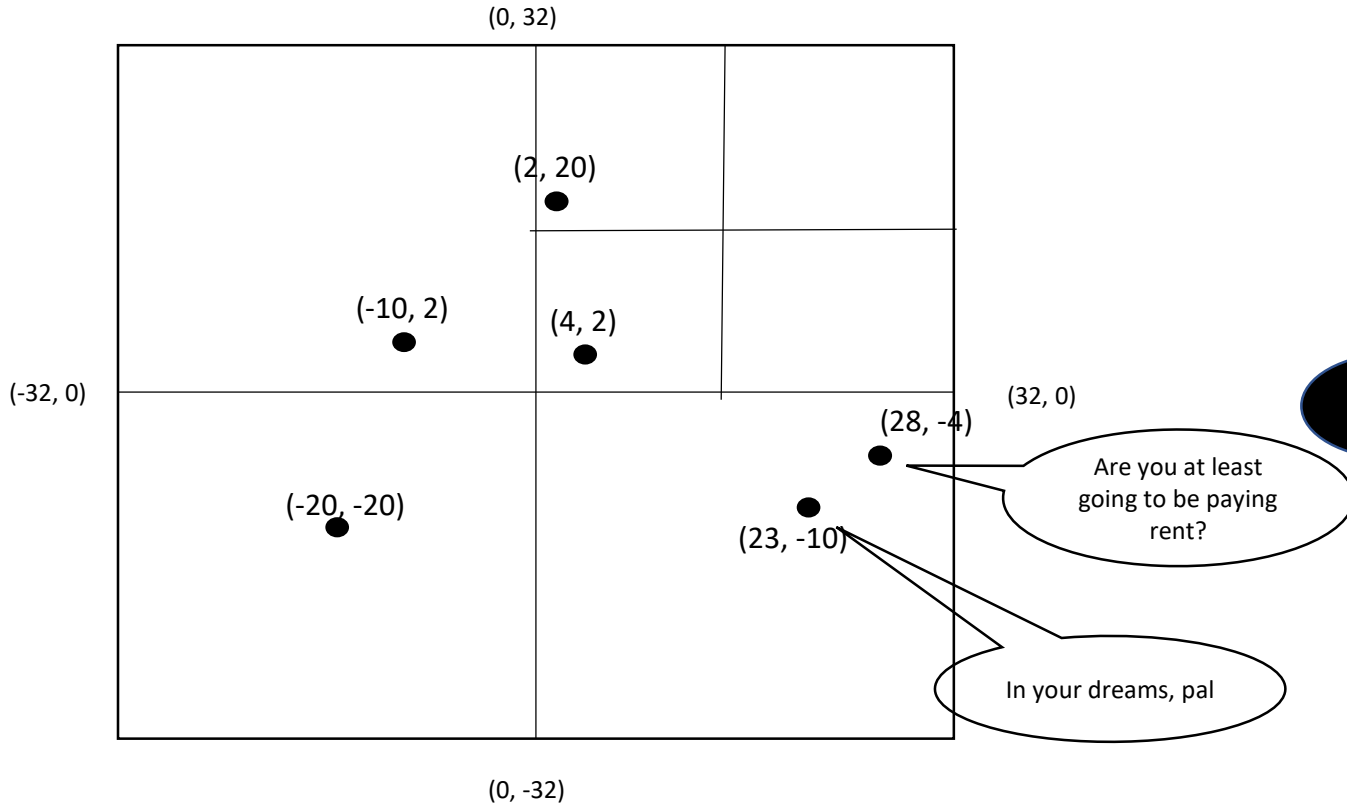


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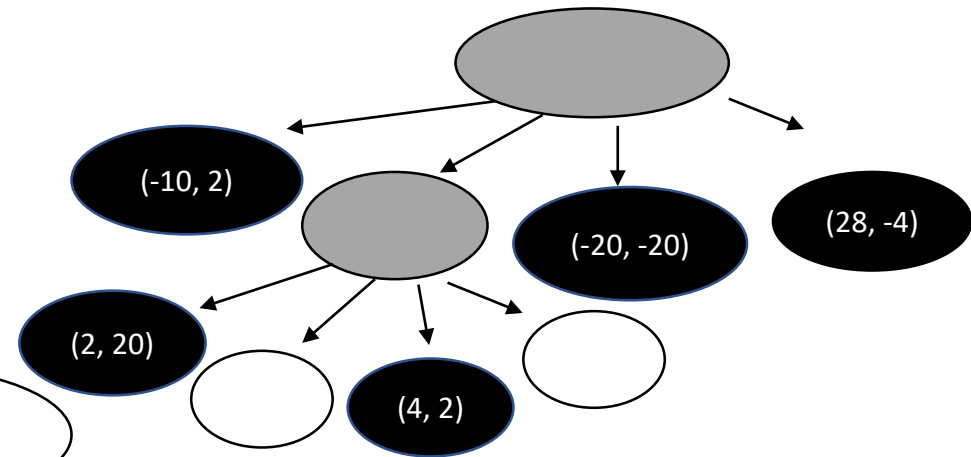


# Example

## Space

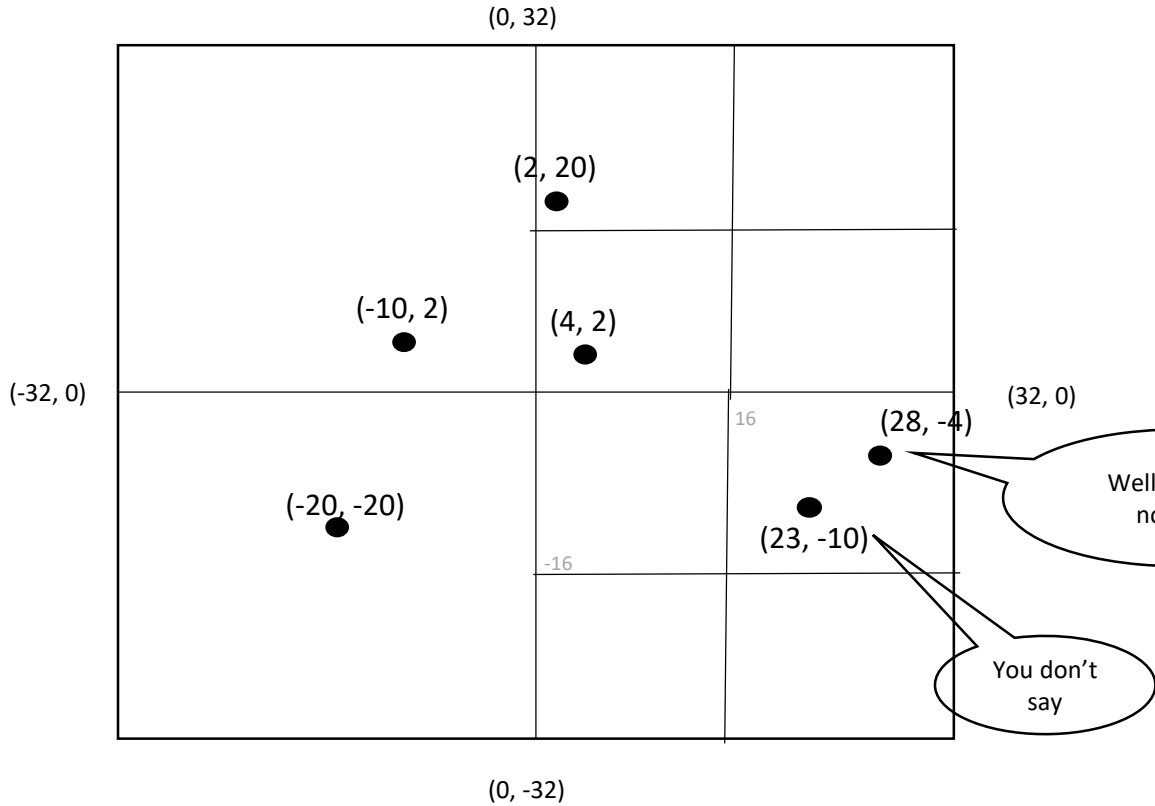


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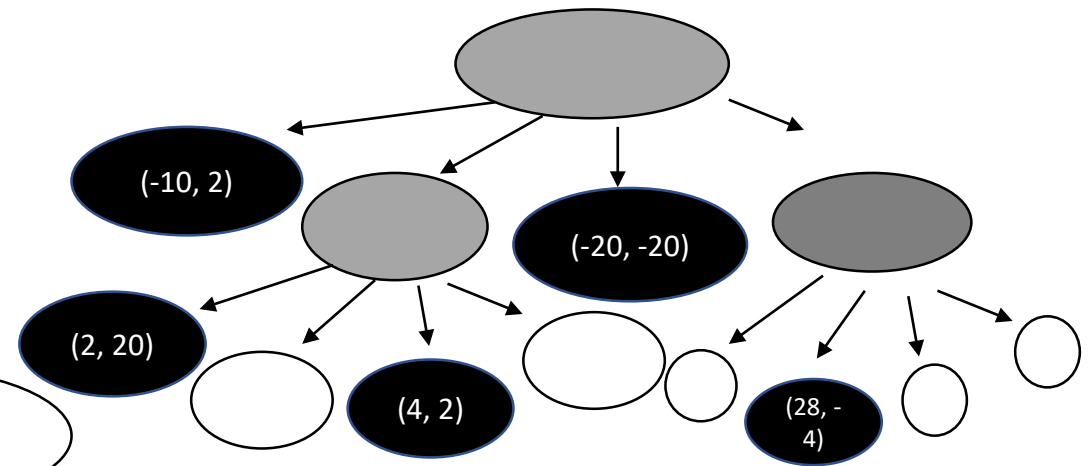


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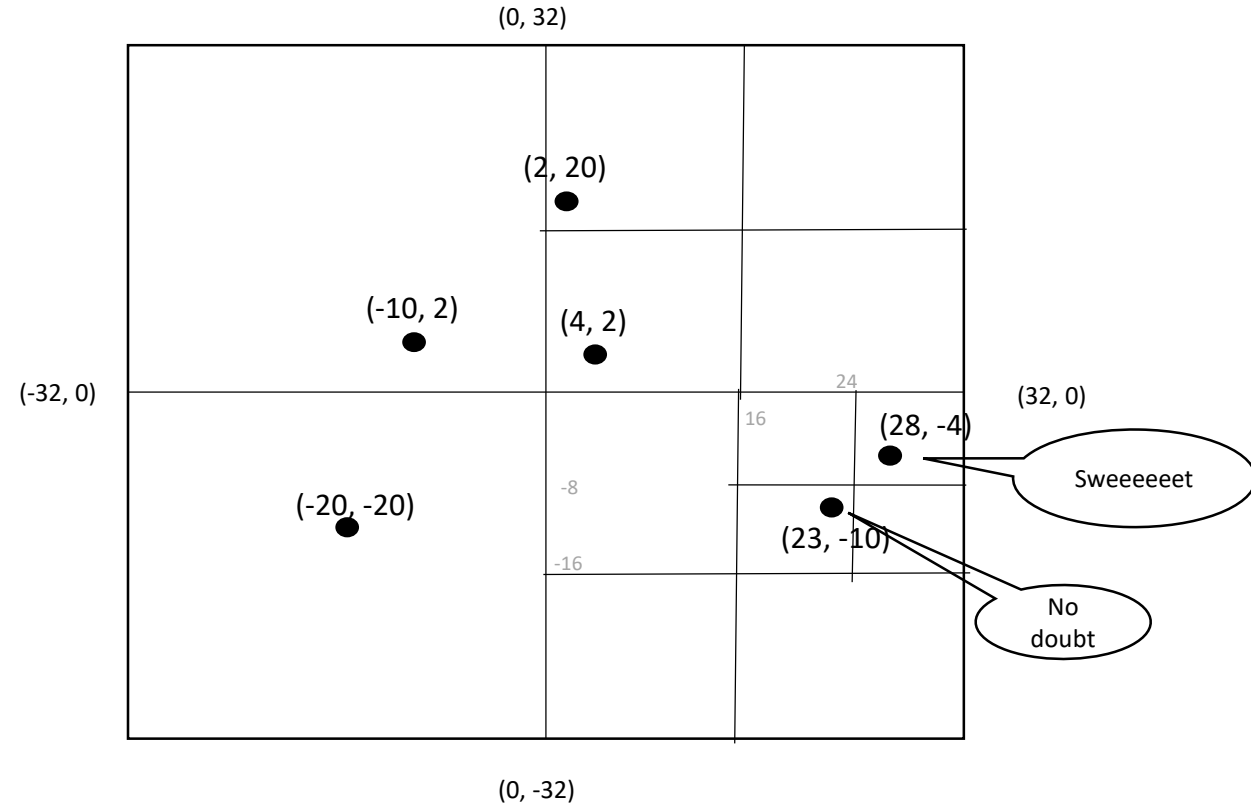


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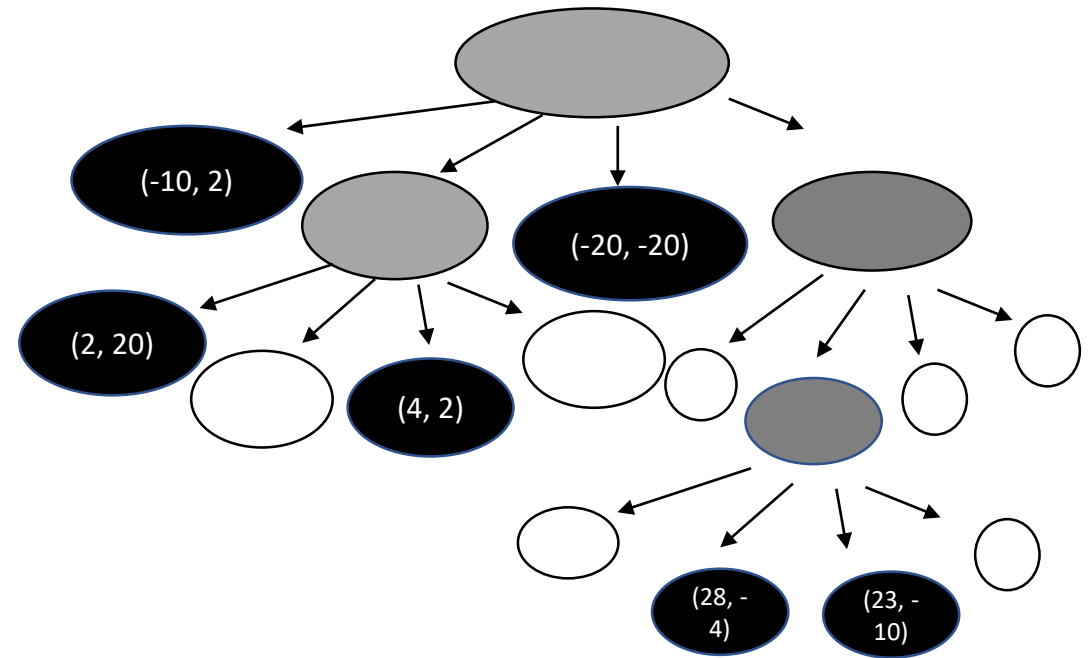


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## PR-QuadTree



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    - **Black** nodes can now store  $b$  points.
  2. **Split at positions other than powers of 2** to better fit your data
    - This is the approach of **“loose”** PR-QuadTrees
    - We will not talk about those, at least not in lecture.

# Exercise!

- Insert the following points into an **initially empty** PR-QuadTree.
  - Parameters:
    - $k = 5$
    - $b = 2$
    - Center of cartesian space is  $(0, 0)$
  - Points:  $(20, 30)$ ,  $(-20, 30)$ ,  $(5, -5)$ ,  $(-8, -2)$ ,  $(-9, -19)$ ,  $(1, 13)$ ,  $(1, 16)$ ,  $(5, 9)$ ,  $(5, 5)$

# Conjectures

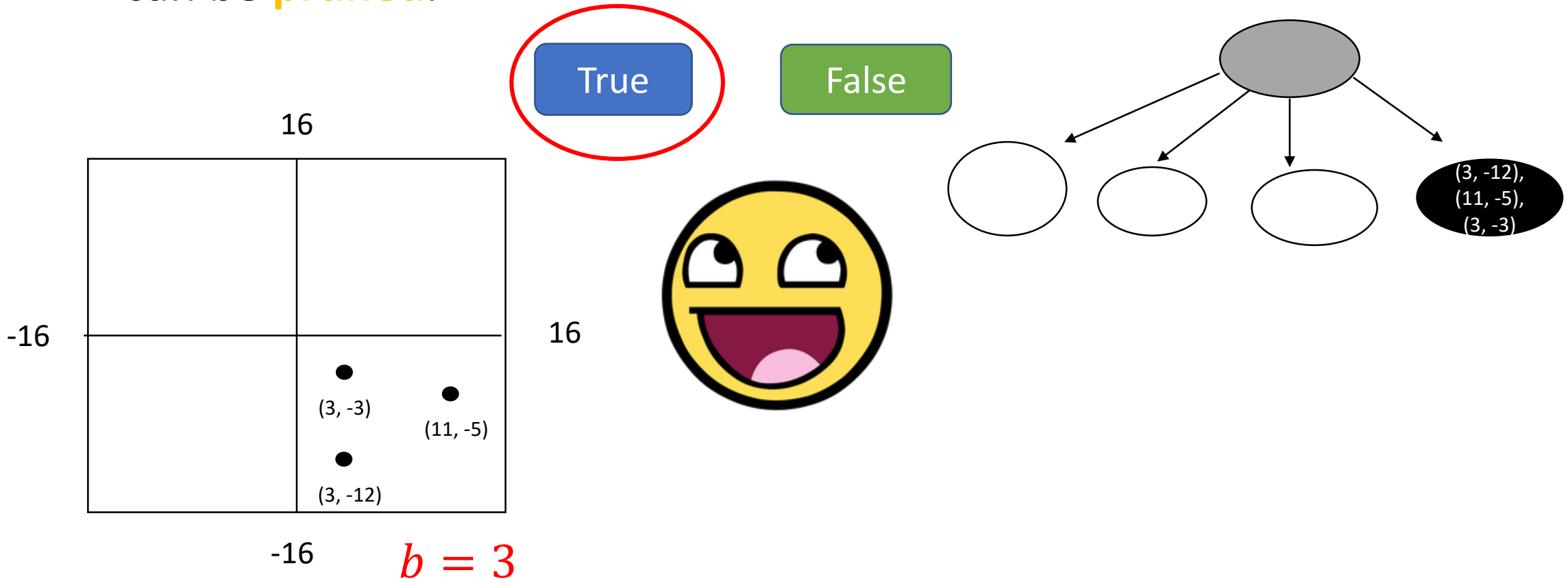
- Gray nodes with 3 white children and 1 black child are **useless** and can be **pruned**.

True

False

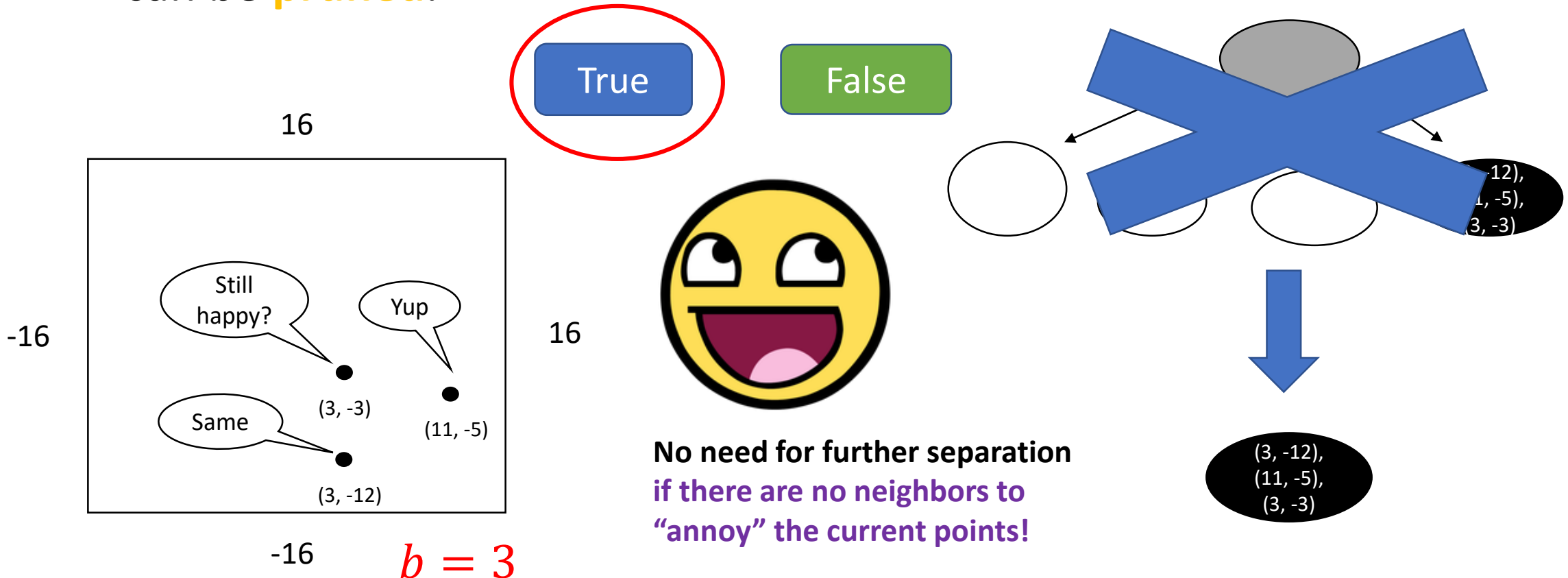
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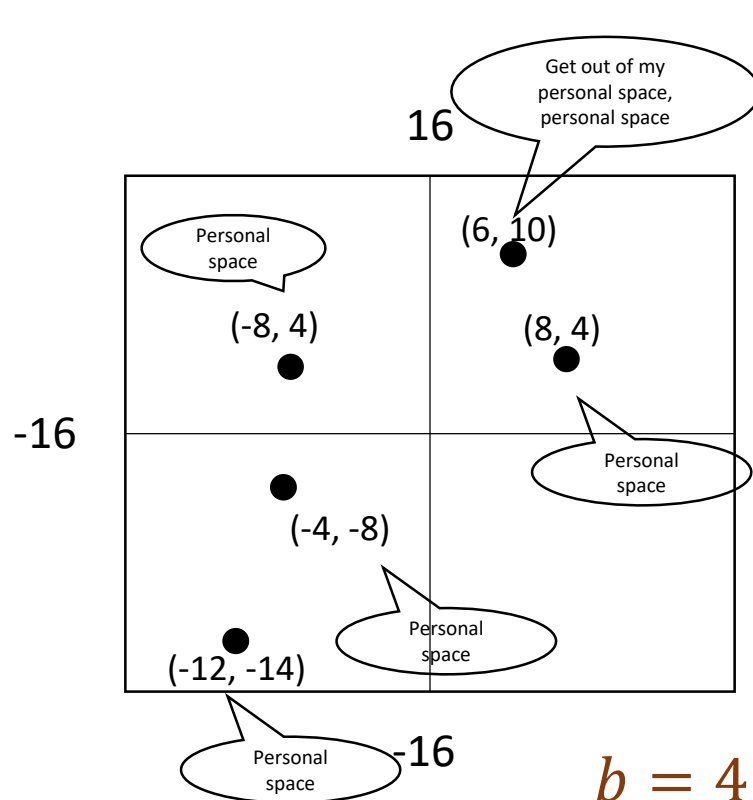
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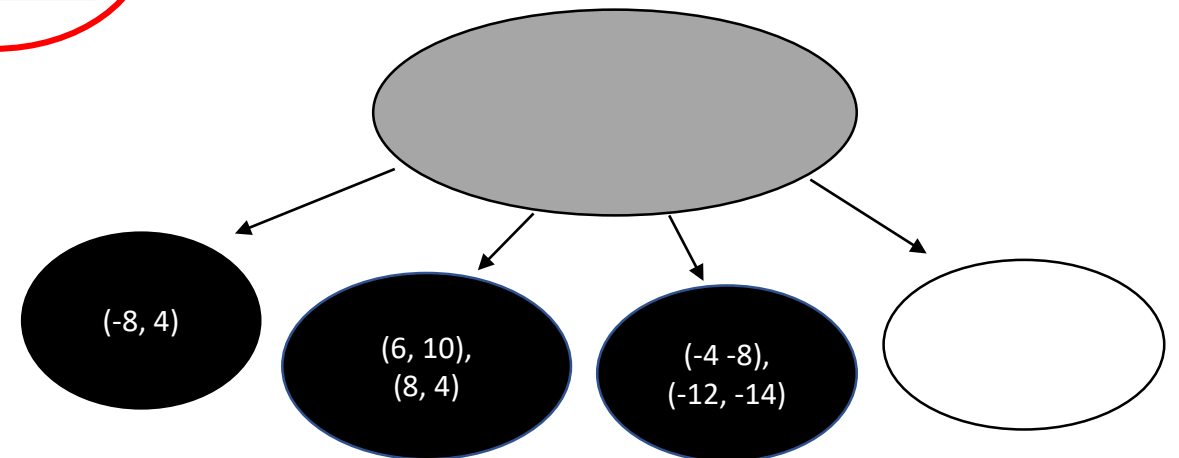
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True

False



What if merging them all into a new black node surpasses parameter  $b$ ?

$$b = 4$$

# Conjectures

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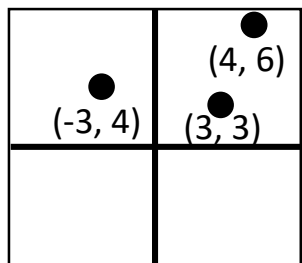
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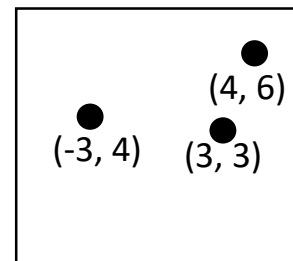
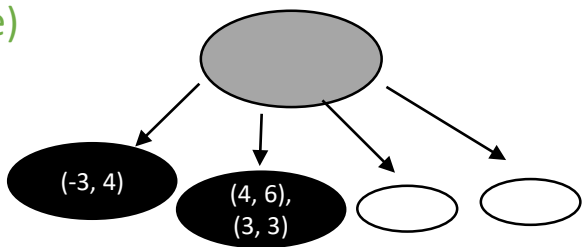
True

False

$d = 3$  ( $2^3 \times 2^3$  space)



$b = 3$



# Conjectures

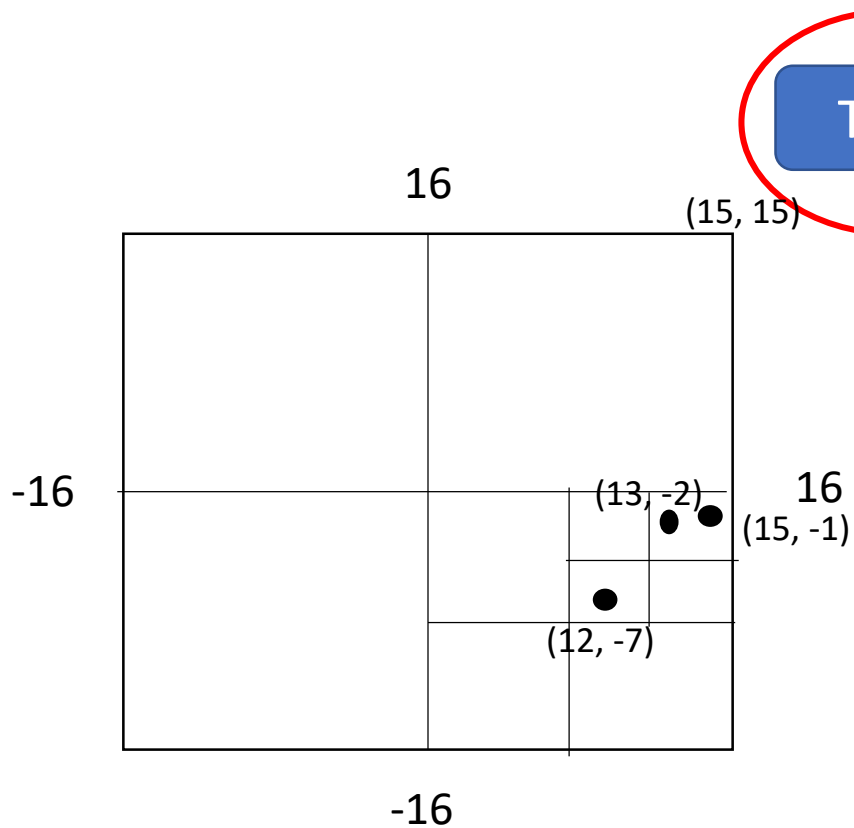
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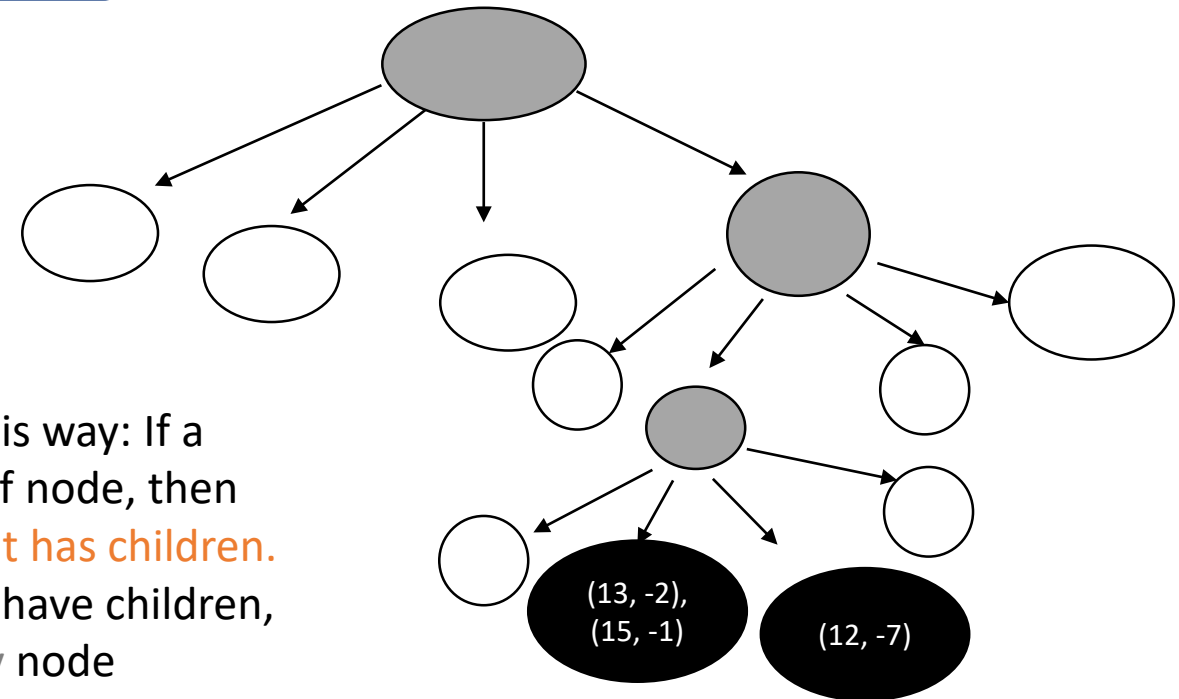
# Conjectures

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True

False



Think about it this way: If a node is not a leaf node, then this means that **it has children**. In order for it to have children, it **must** be a grey node

# Deletion...

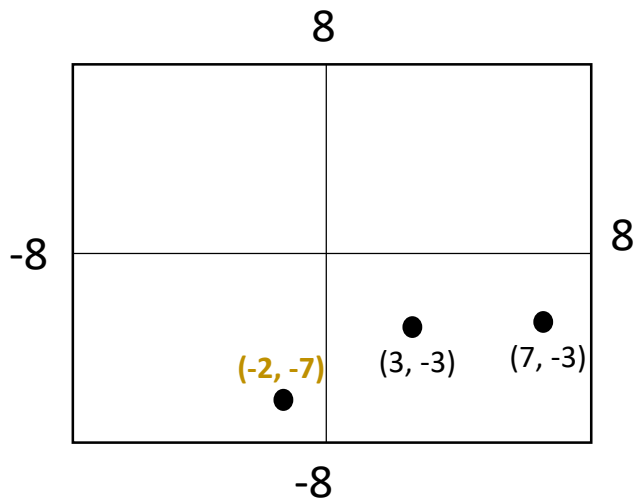
- Insertion of a  $(b + 1)$ th point into a **black node** *splitted* the node into a gray node with 4 children...
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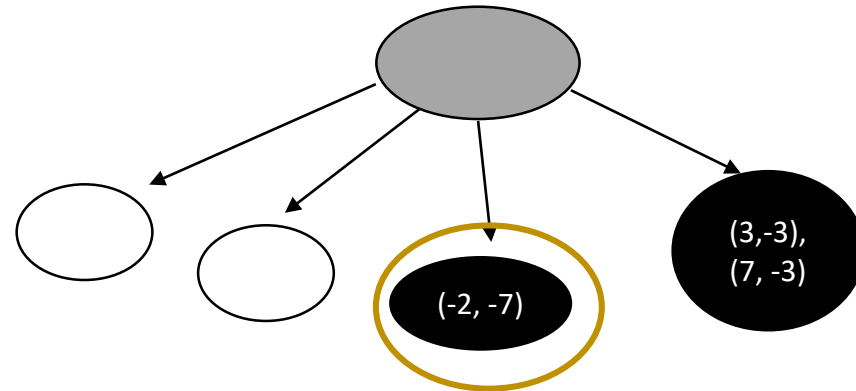
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- E.g deletion of  $(-2, -7)$  below

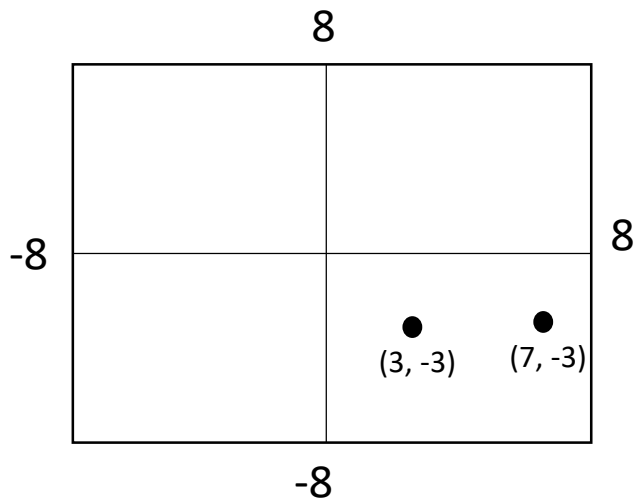


$$b = 2$$

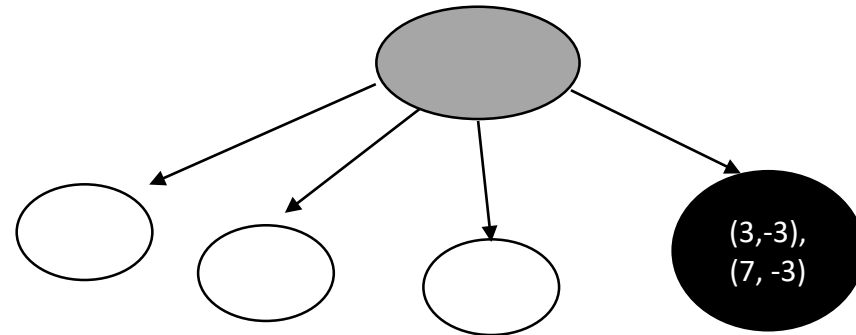


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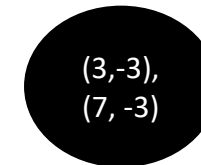
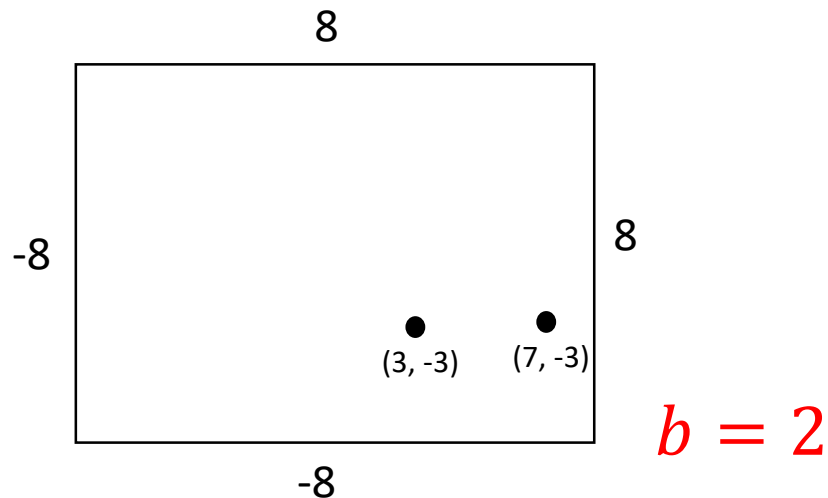


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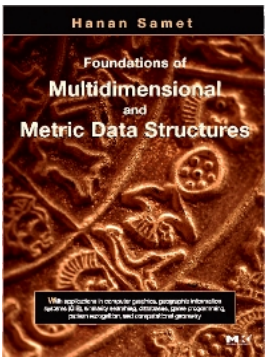


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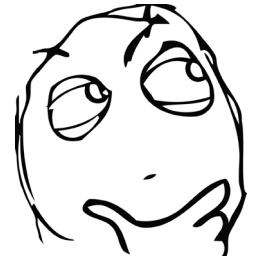


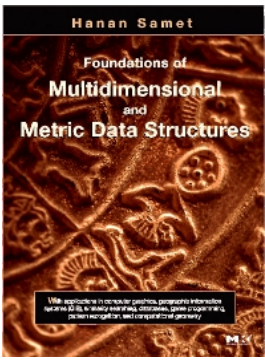




# “Trie – Based” quadtrees?

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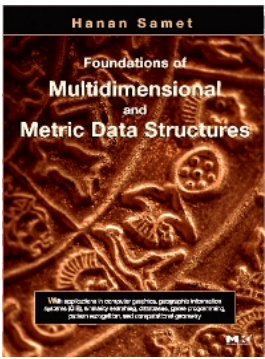




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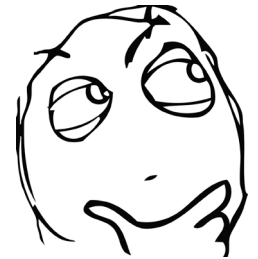
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# “Trie – Based” quadtrees?

- Many resources, including *the bible*, call quadtrees like the PR “trie-based” quadtrees.
- For a long time, Jason could not understand *why*.
- **Let’s see if you can!**
- True or False: The worst-case time complexity of *search* depends on the number  $n$  of points stored in the PR-Quadtree.



True

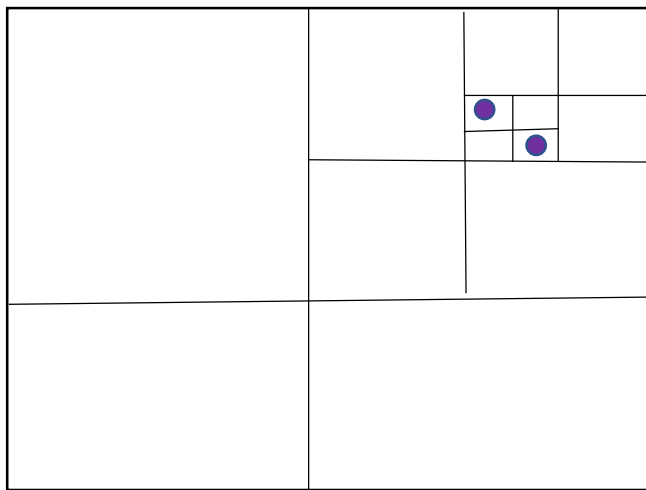
(Give me a  $\mathcal{O}(f(n))$ )

False

(What does it depend on, then?)

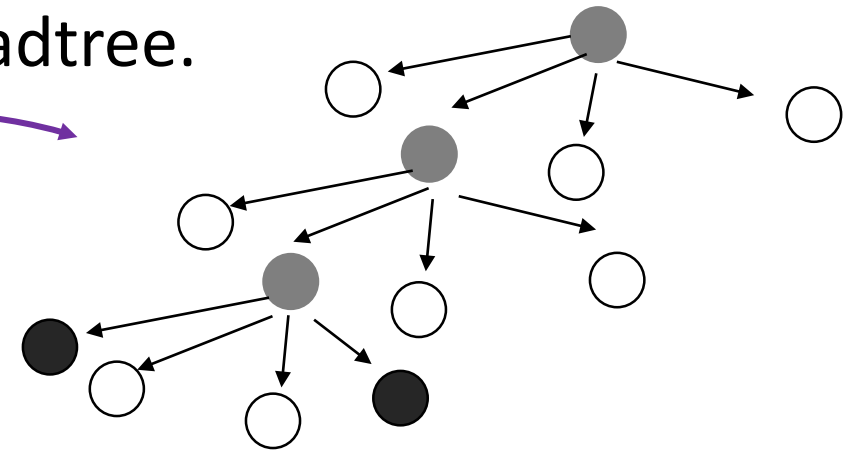
# “Trie – Based” quadtrees?

- Many resources, including *the bible*, call quadtrees like the PR “trie-based” quadtrees.
- For a long time, Jason could not understand *why*.
- **Let’s see if you can!**
- True or False: The worst-case time complexity of *search* depends on the number  $n$  of points stored in the PR-Quadtree.



True  
(Give me a  
 $\mathcal{O}(f(n))$ )

**False**  
(What does it  
depend on,  
then?)



Worst case produced by the minimum distance between two points!

# Spatial Queries in PR-Quadtrees

1. *Descending phase:* Still **aggressively descending** towards anchor point, to cover as many points as we can in a short amount of time.

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  - It **doesn't** make sense if:
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# Spatial Queries in PR-Quadtrees

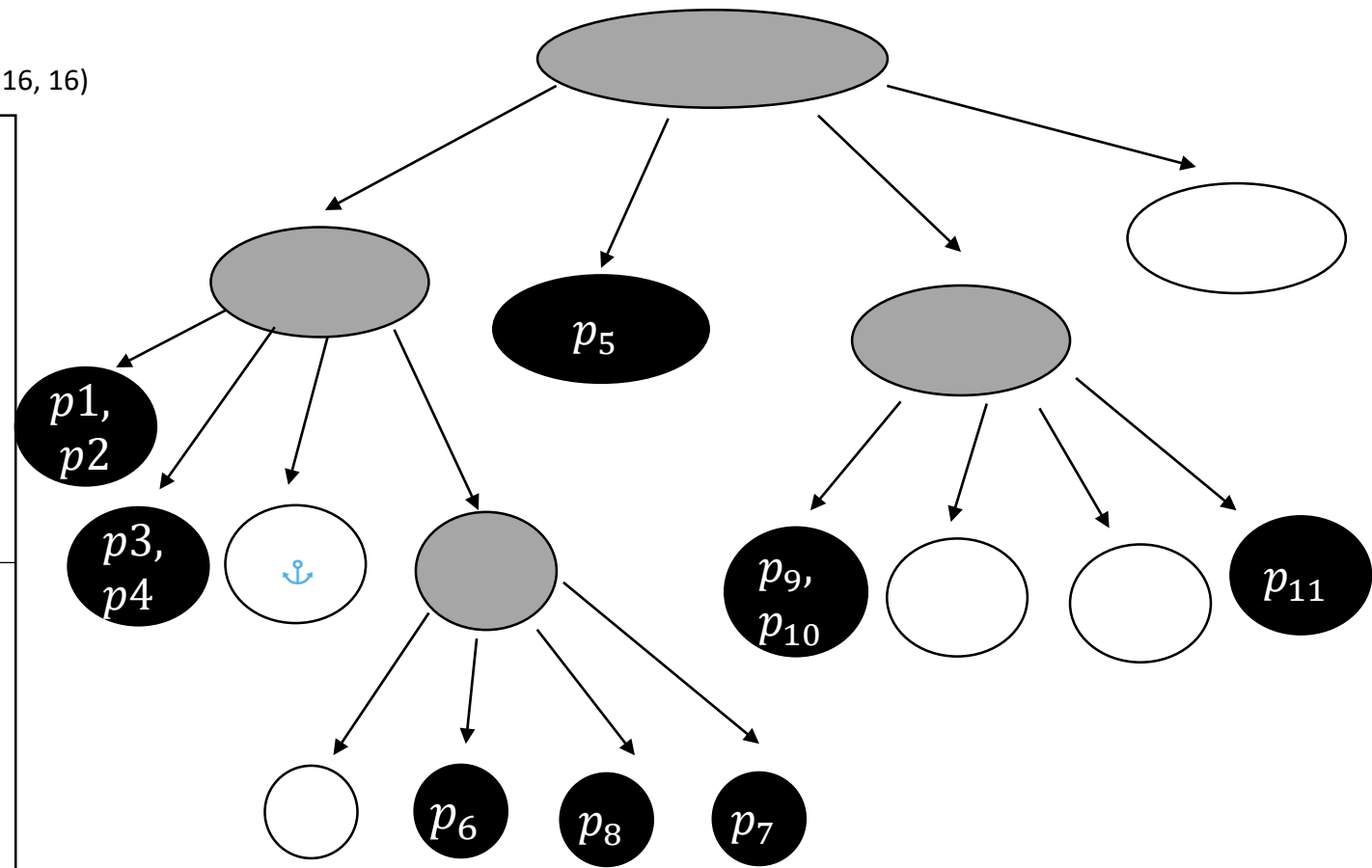
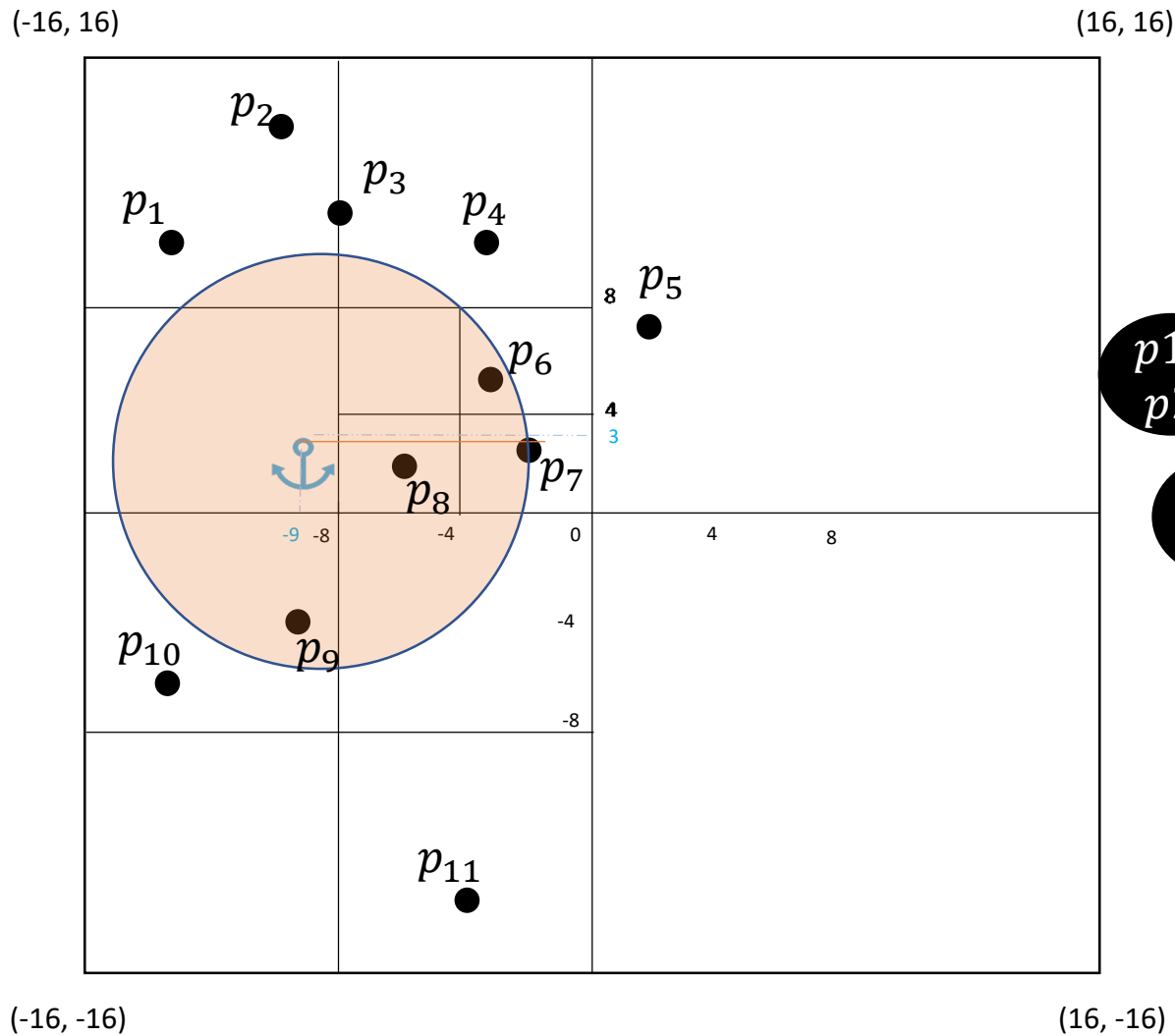
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  - In PR-QuadTrees, we have three different other subtrees to consider!
    - *How should we proceed?*





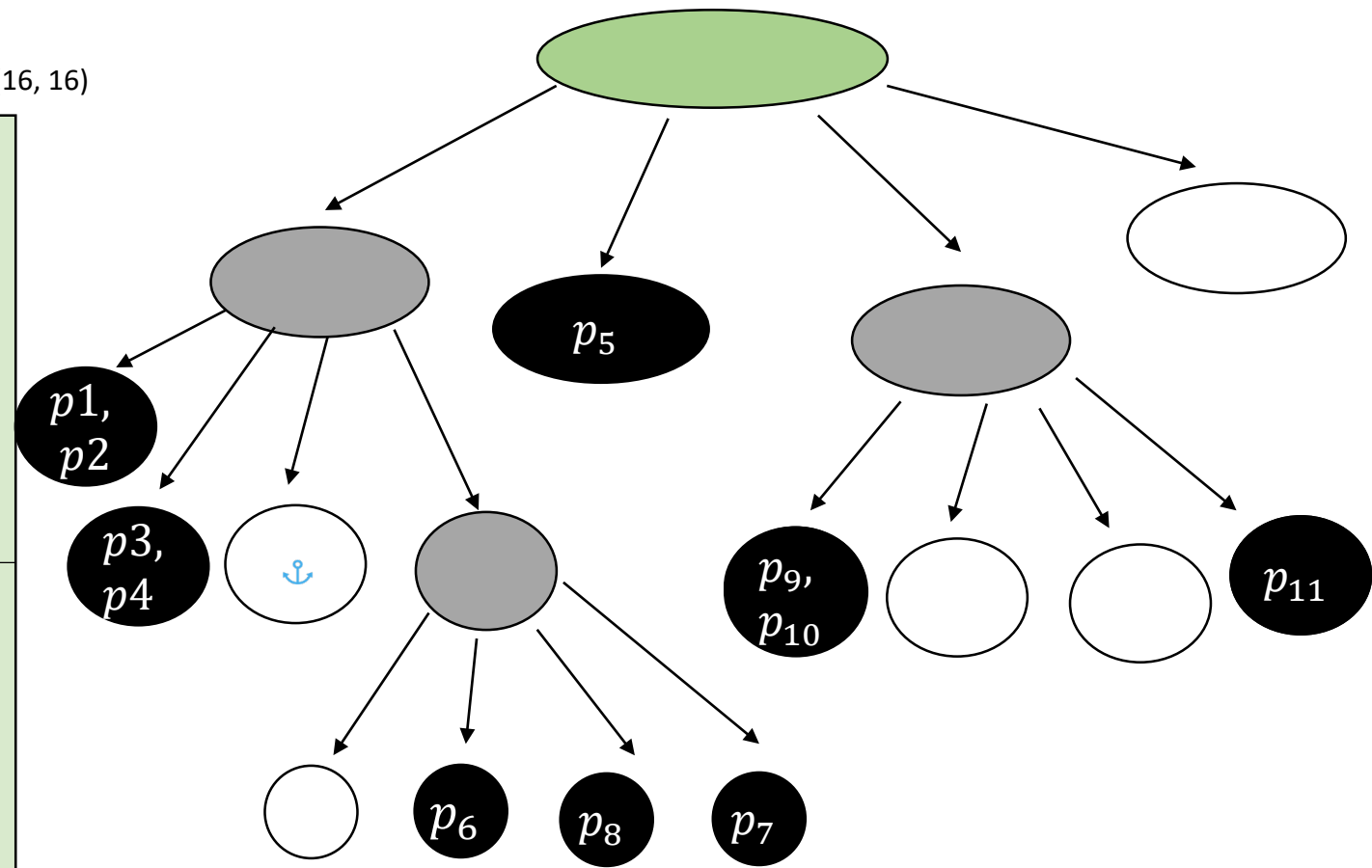
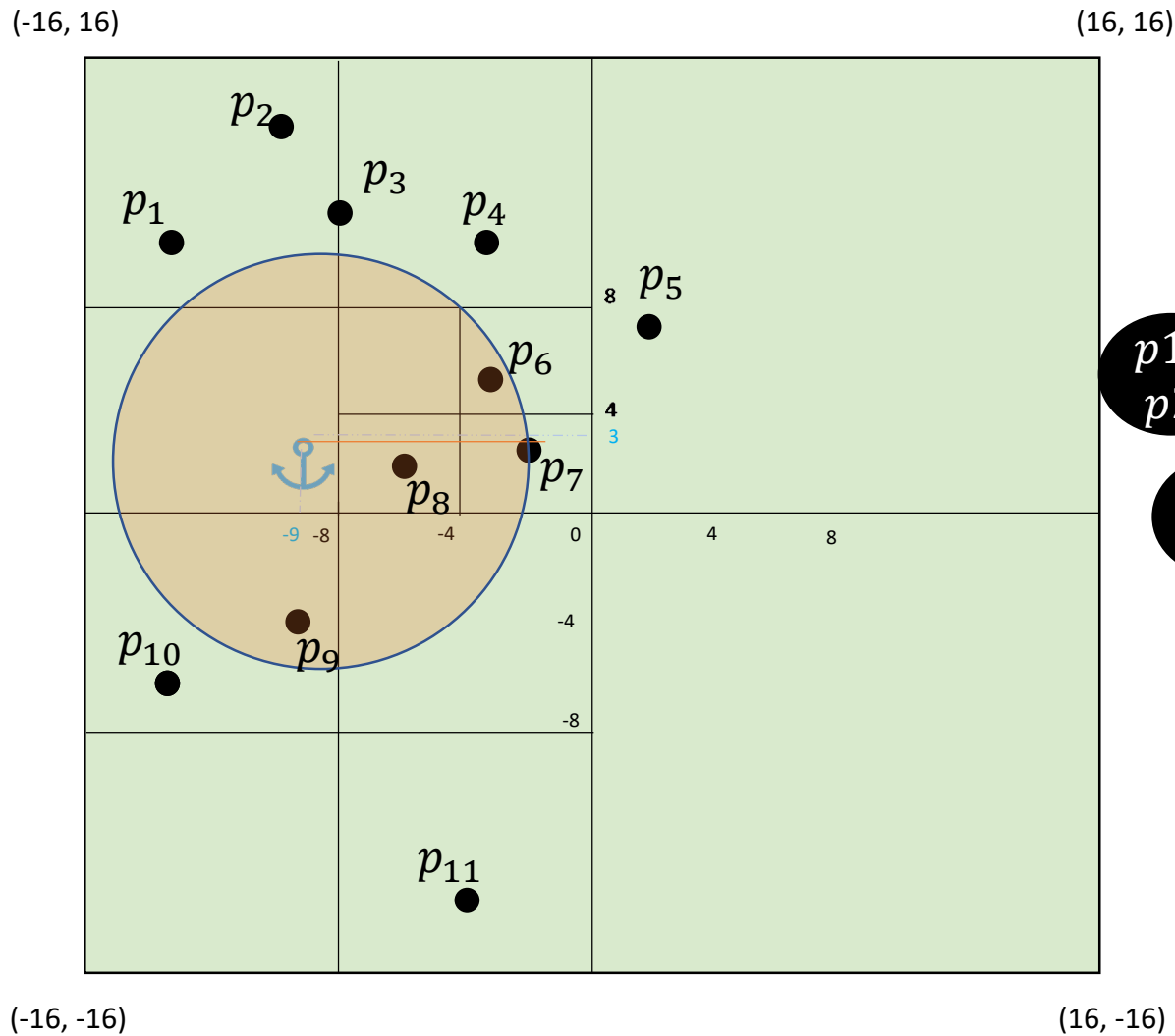
# Example of range query in PR-QuadTree

anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



# Example of range query in PR-QuadTree

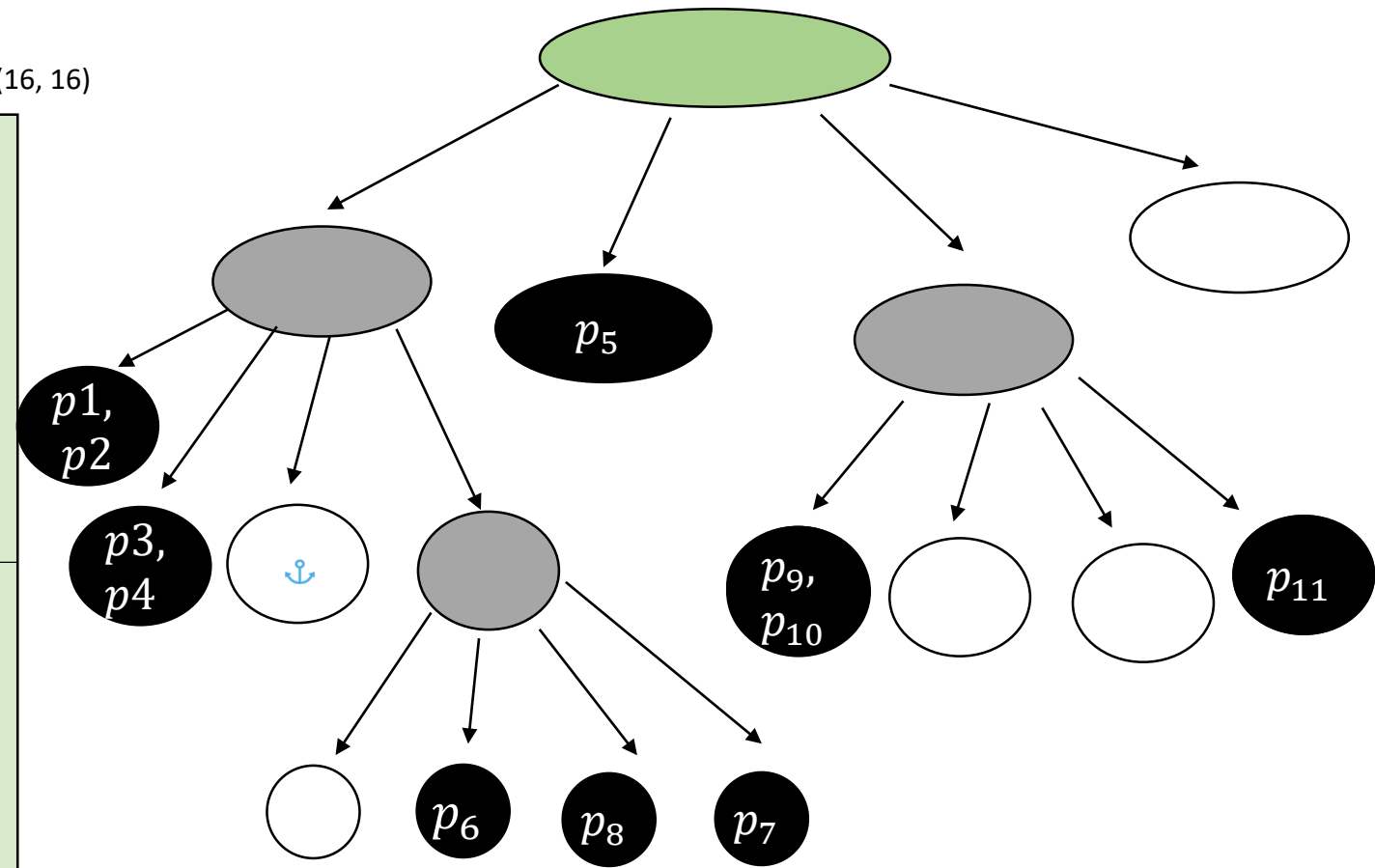
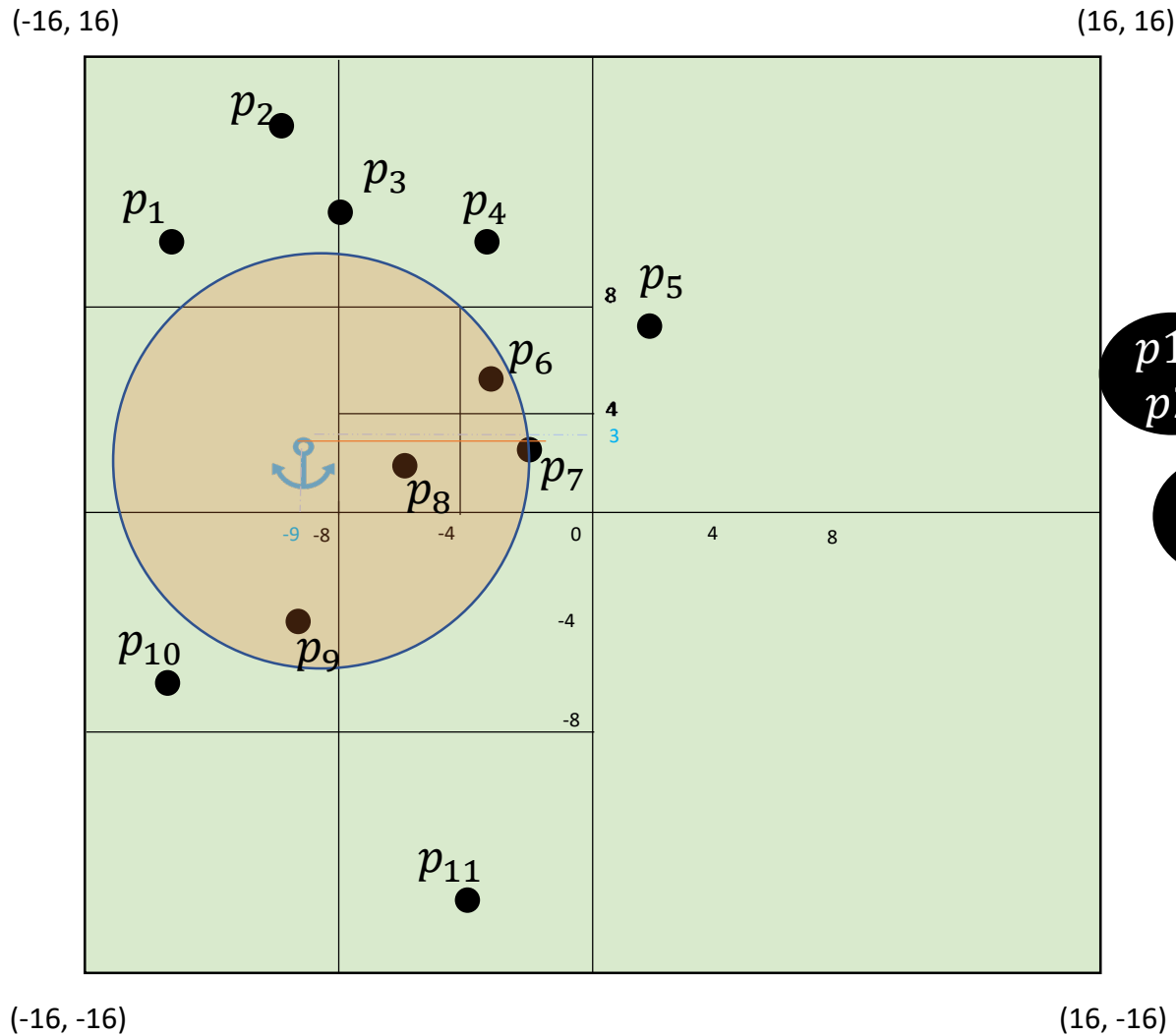
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



Visit root

# Example of range query in PR-QuadTree

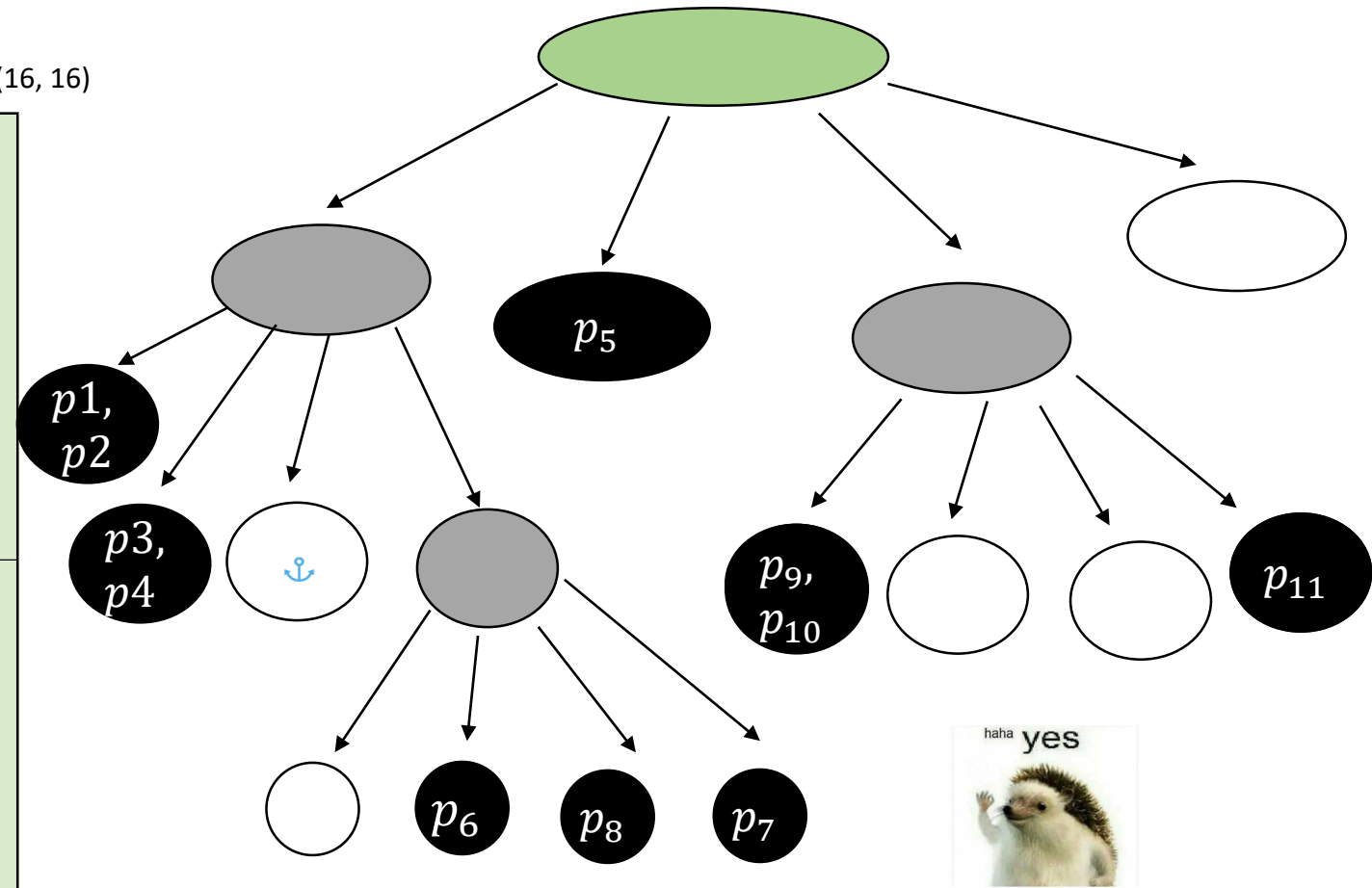
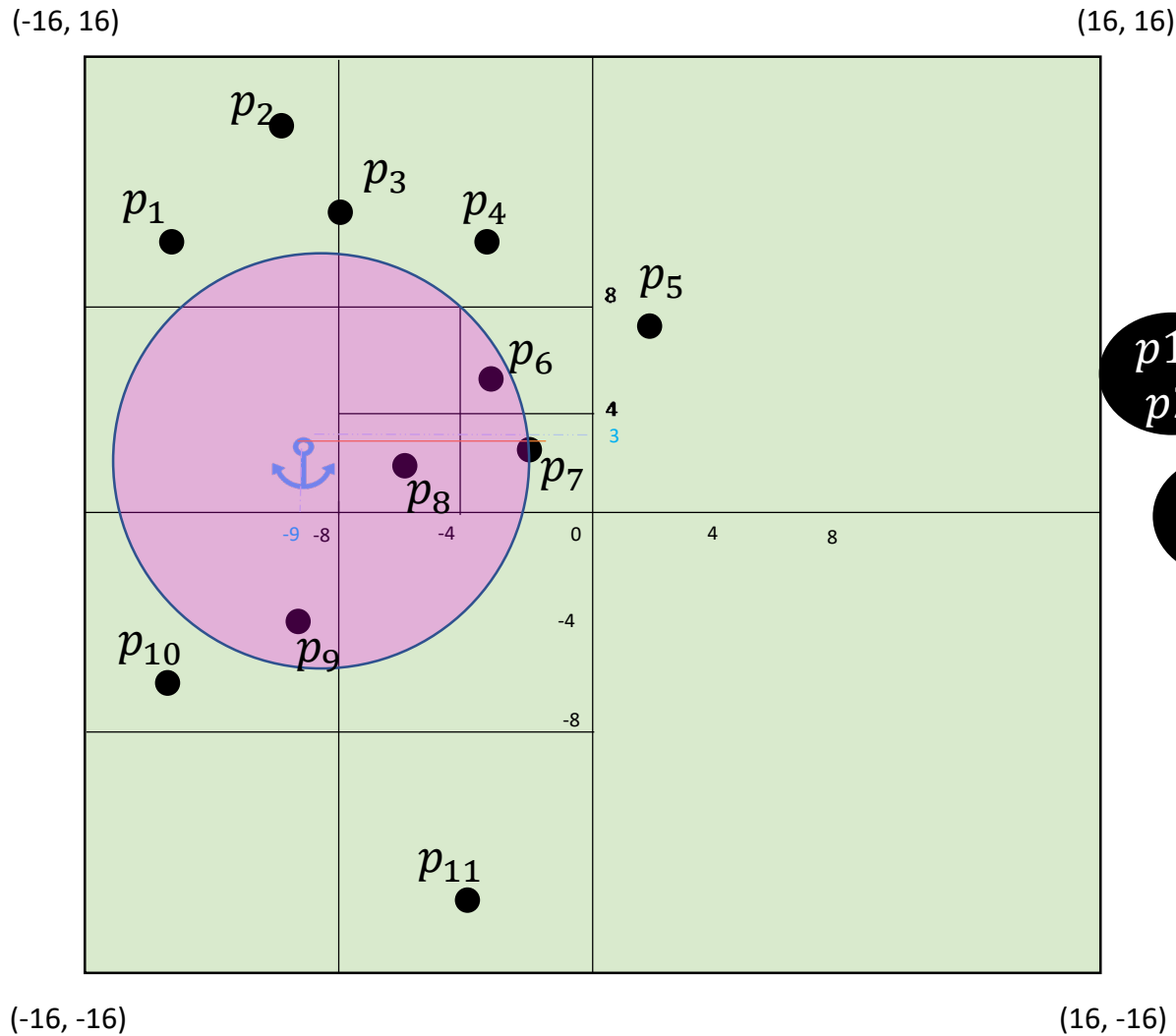
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Is there a **non-zero area intersection** between the **quadrant spanned by the root** and the **range**?

# Example of range query in PR-QuadTree

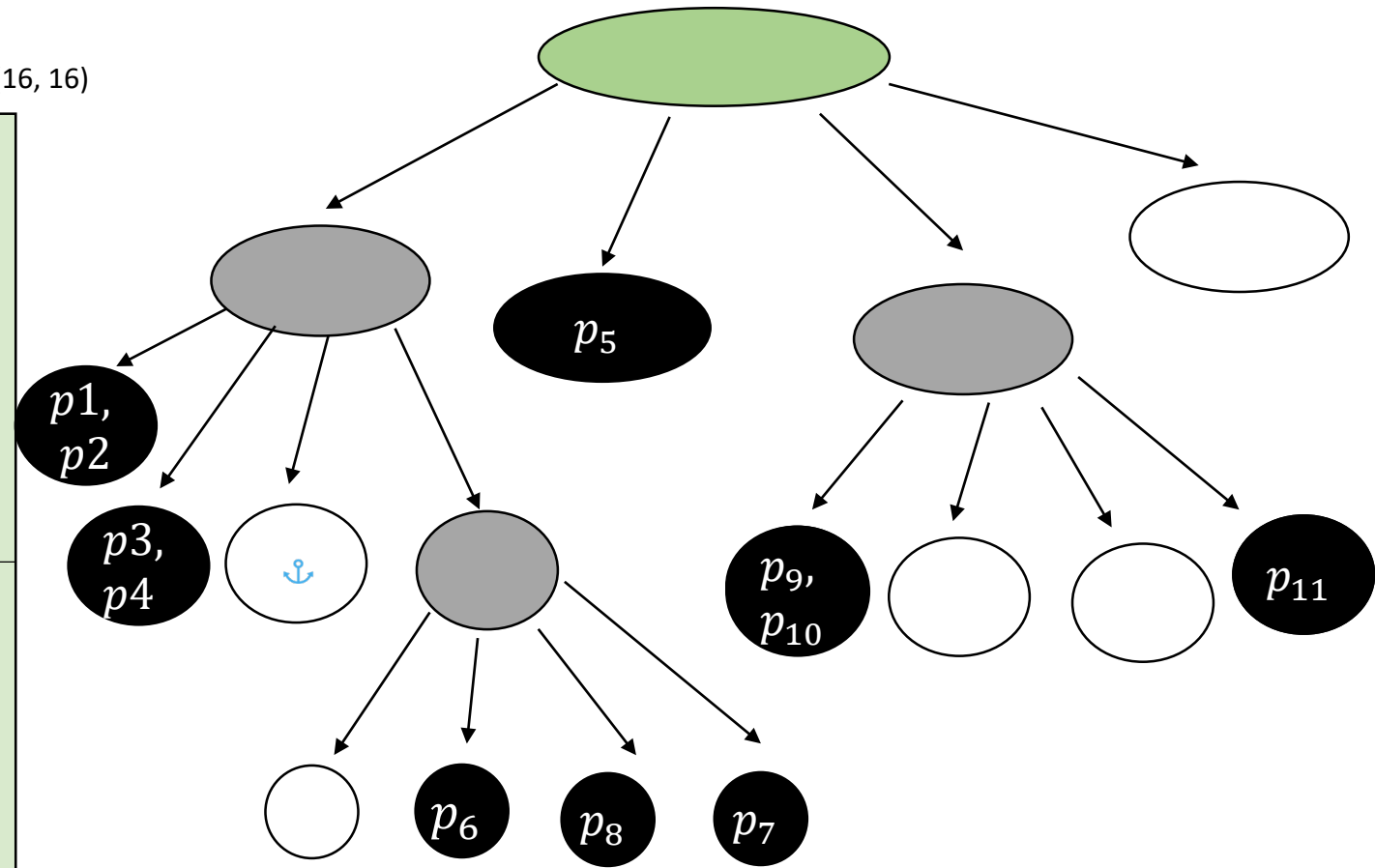
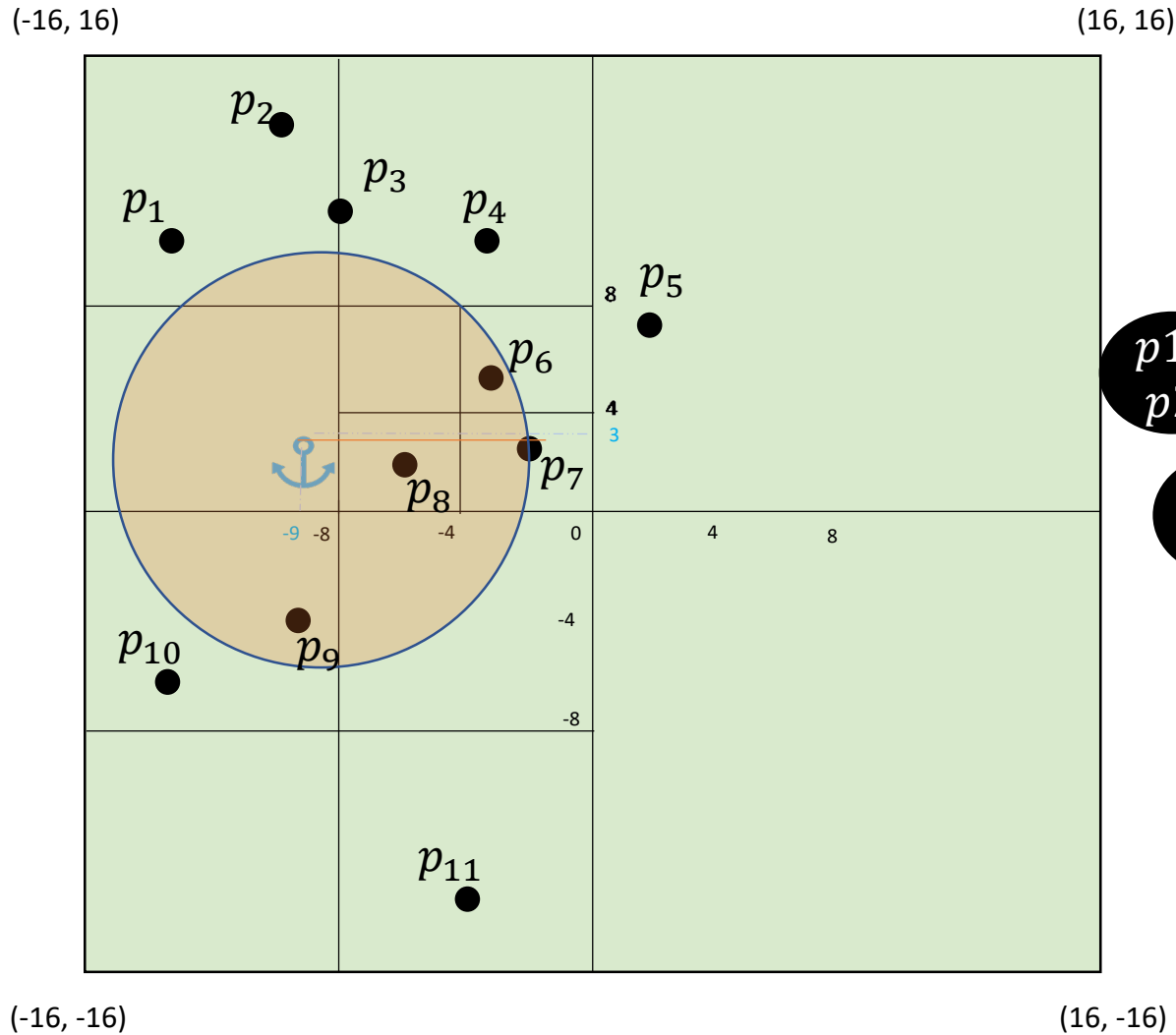
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



Is there a **non-zero area intersection** between the **quadrant spanned by the root** and the **range**? **YUP!**

# Example of range query in PR-QuadTree

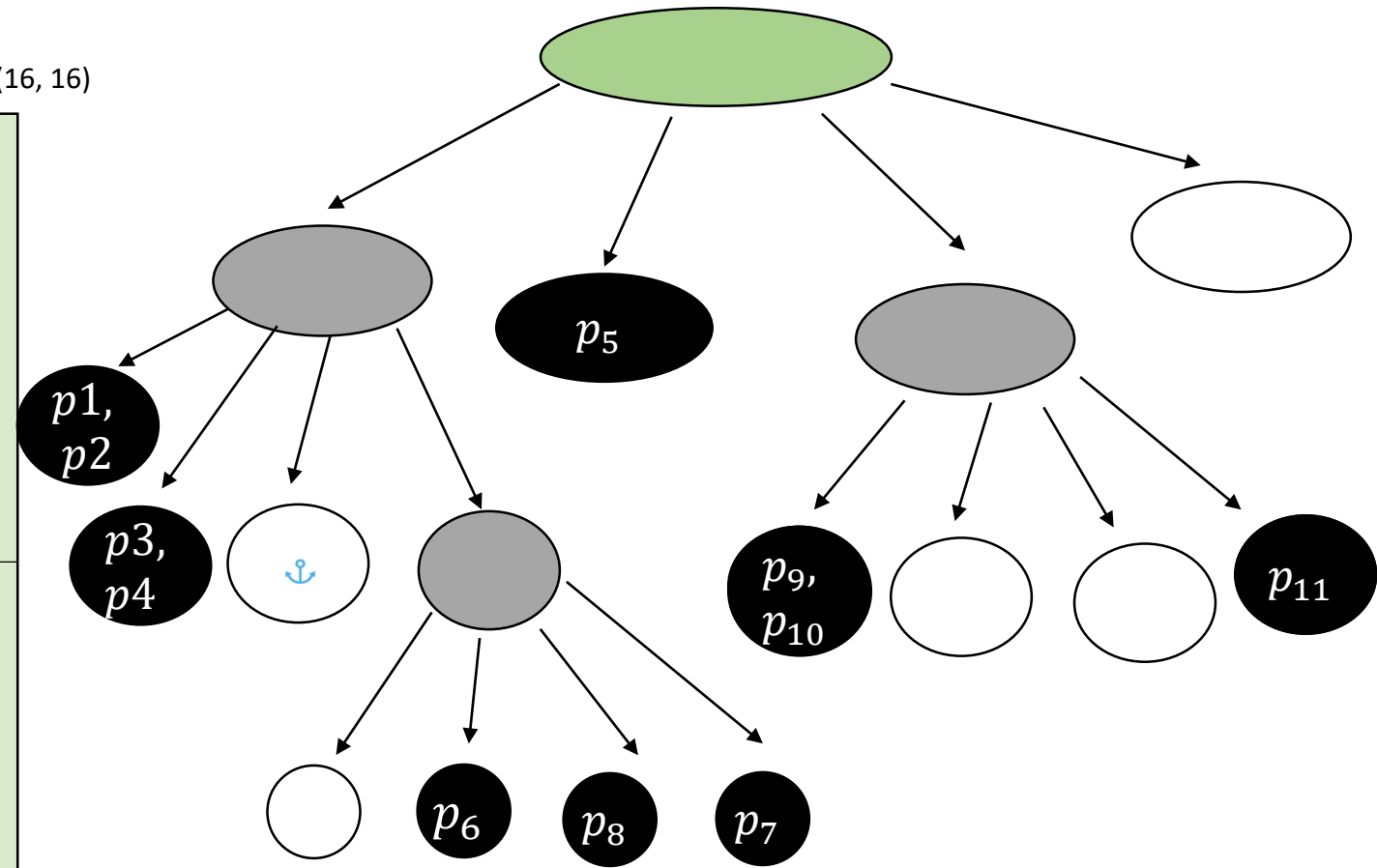
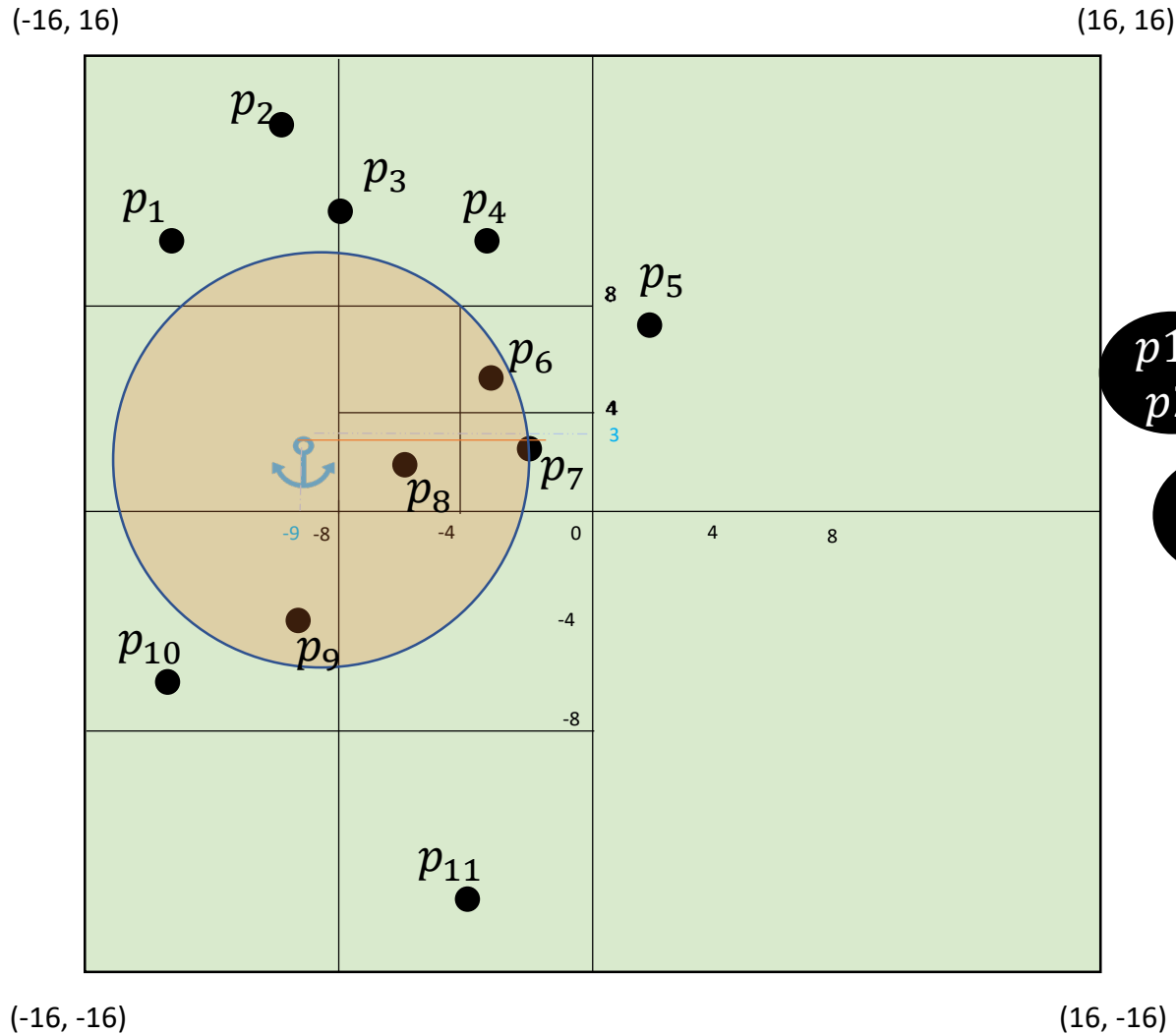
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



So it **makes sense to stay in the tree**, and we don't have to return.

# Example of range query in PR-QuadTree

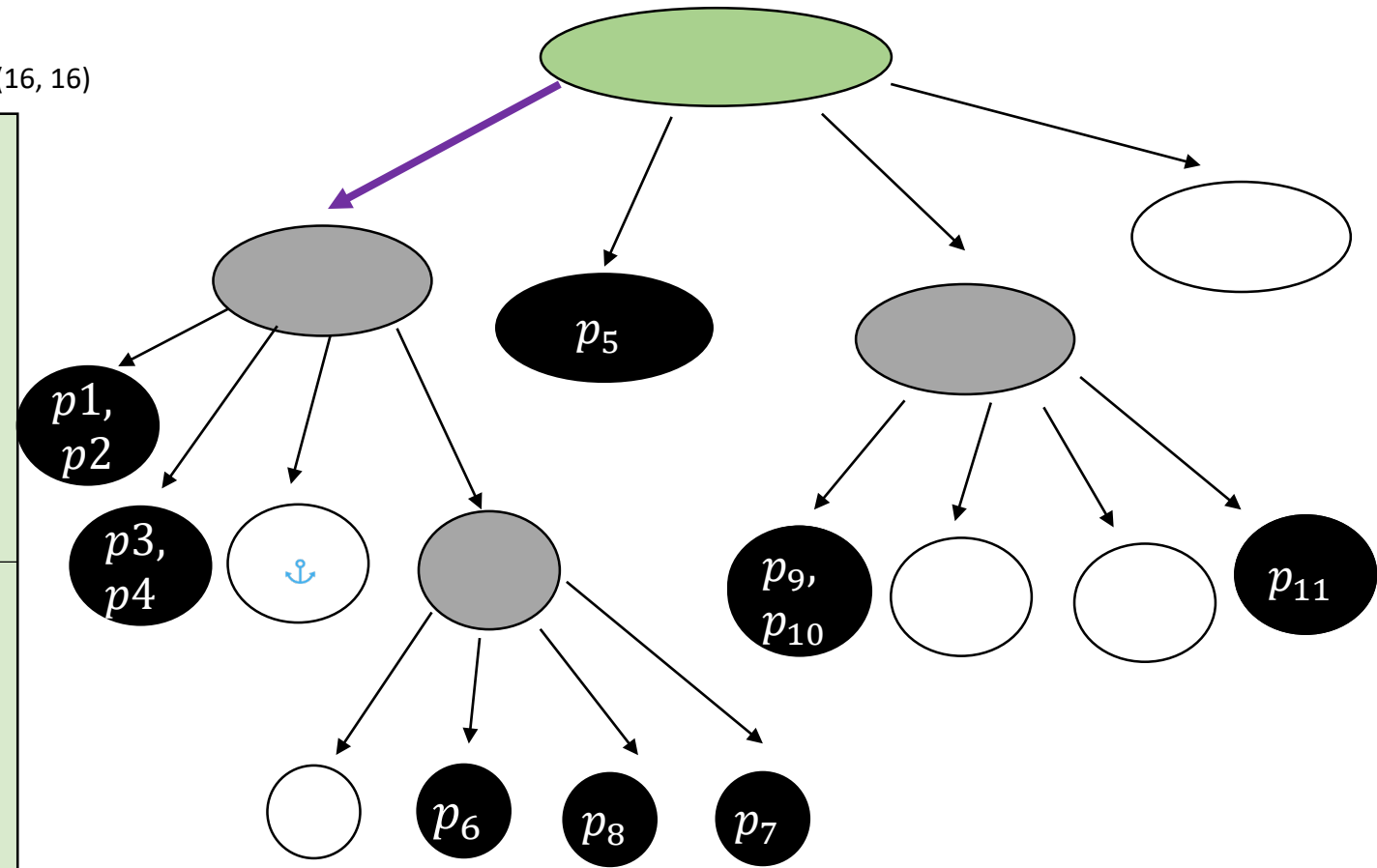
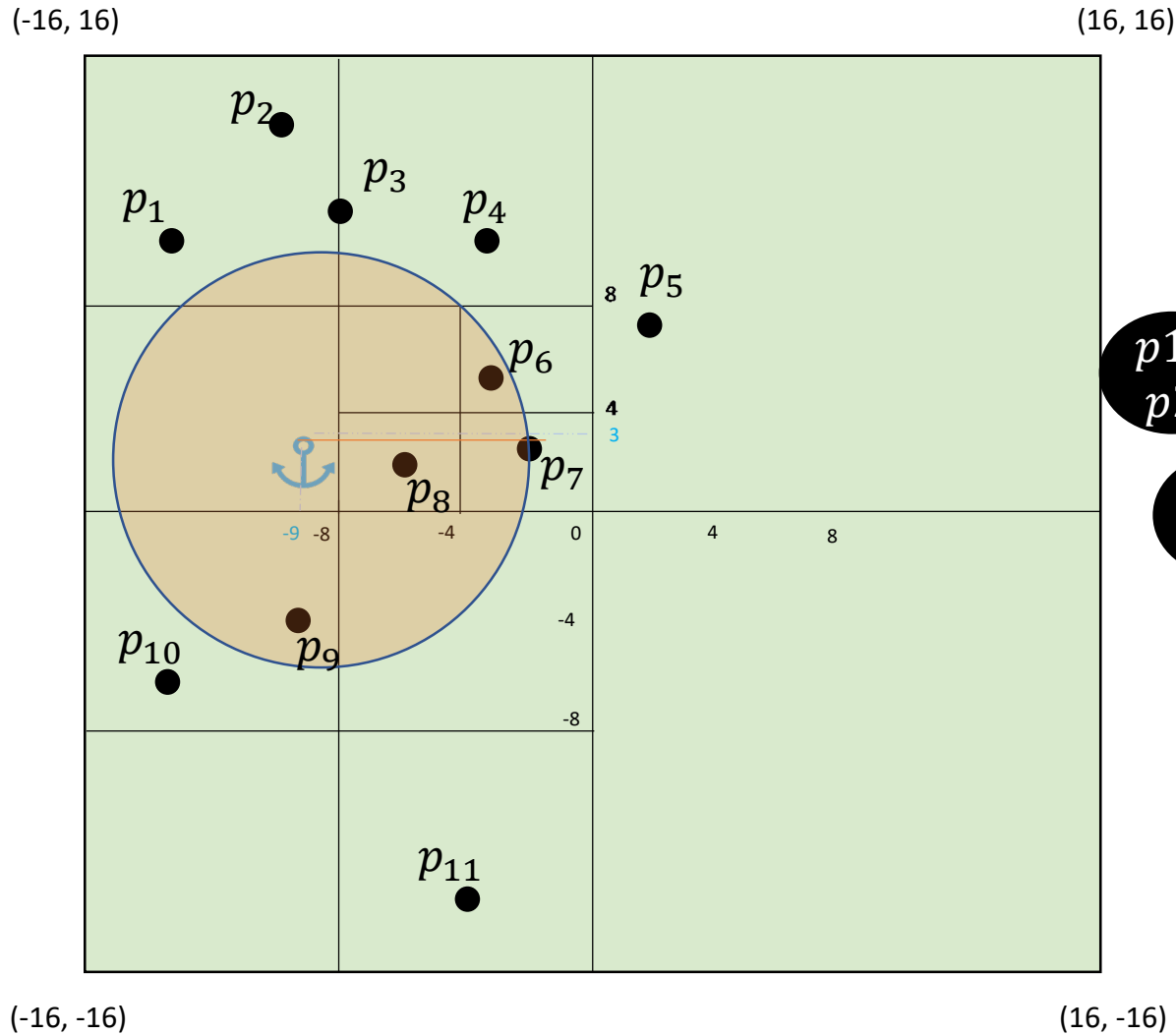
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



This analysis might seem **stupid** to do at the root of the tree, but is exactly the process that needs to happen in every subtree!

# Example of range query in PR-QuadTree

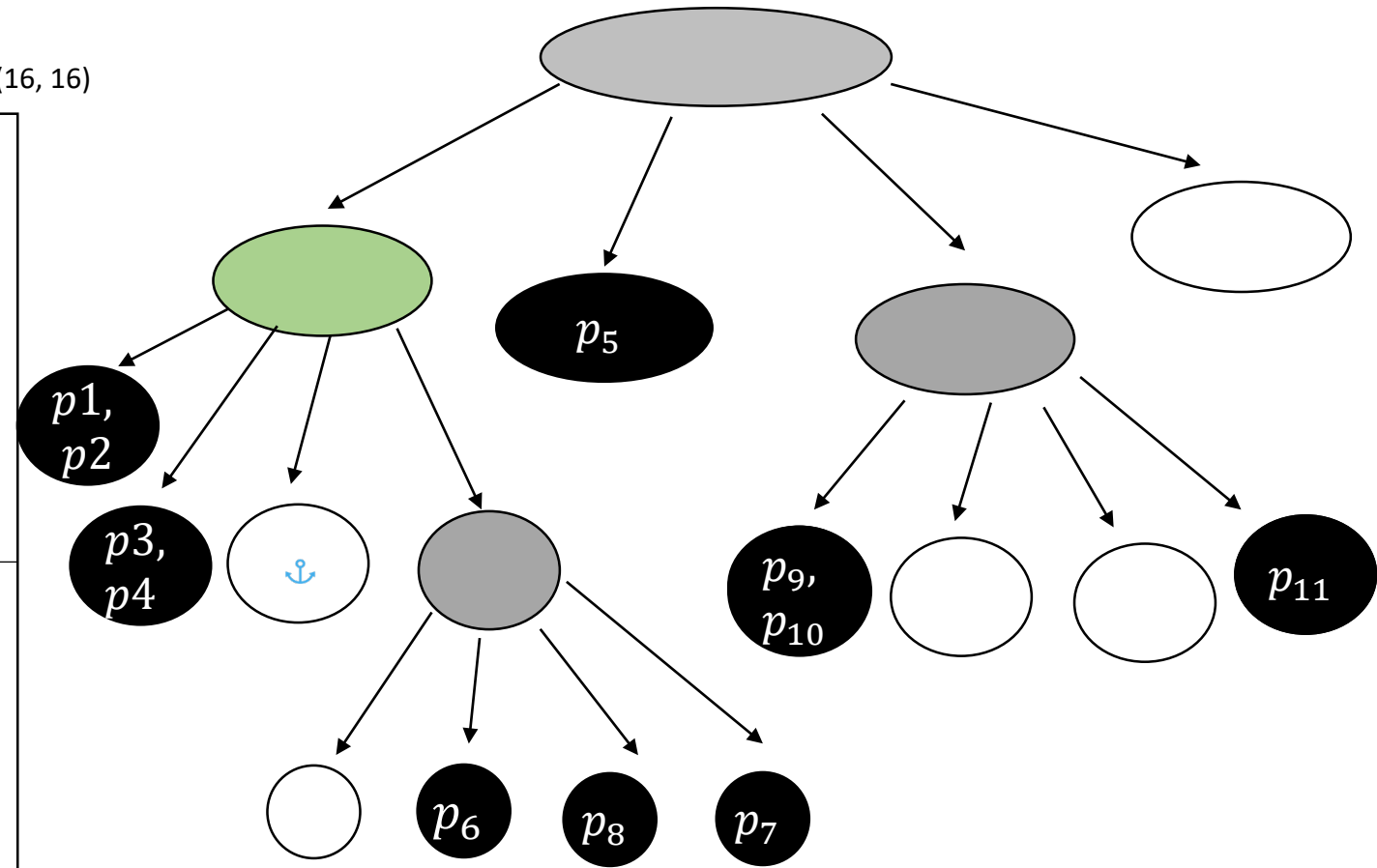
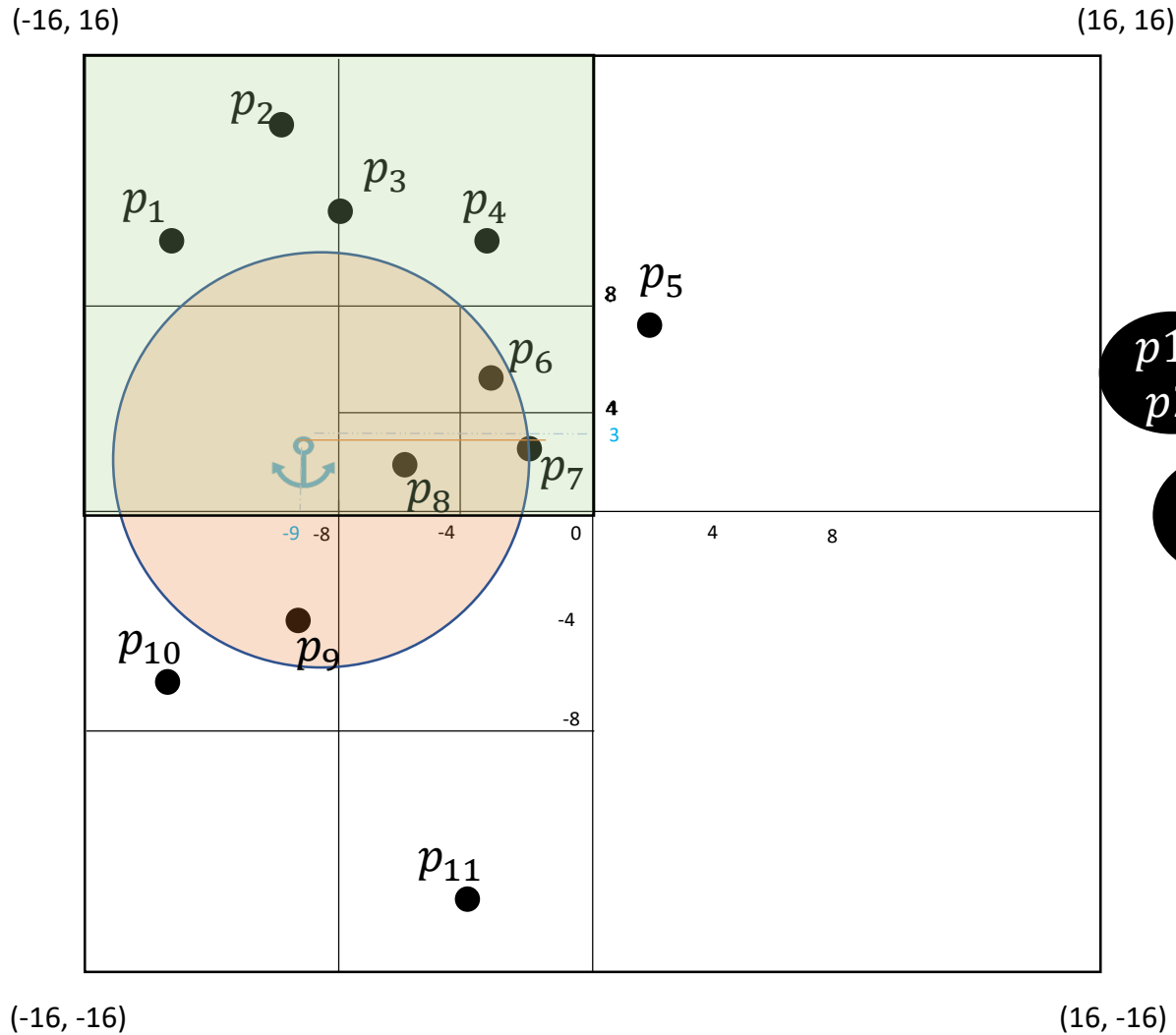
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



**Greedly** approach anchor point to heuristically gather as many points as you can fast. Visit first (NW) subtree.

# Example of range query in PR-QuadTree

anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$

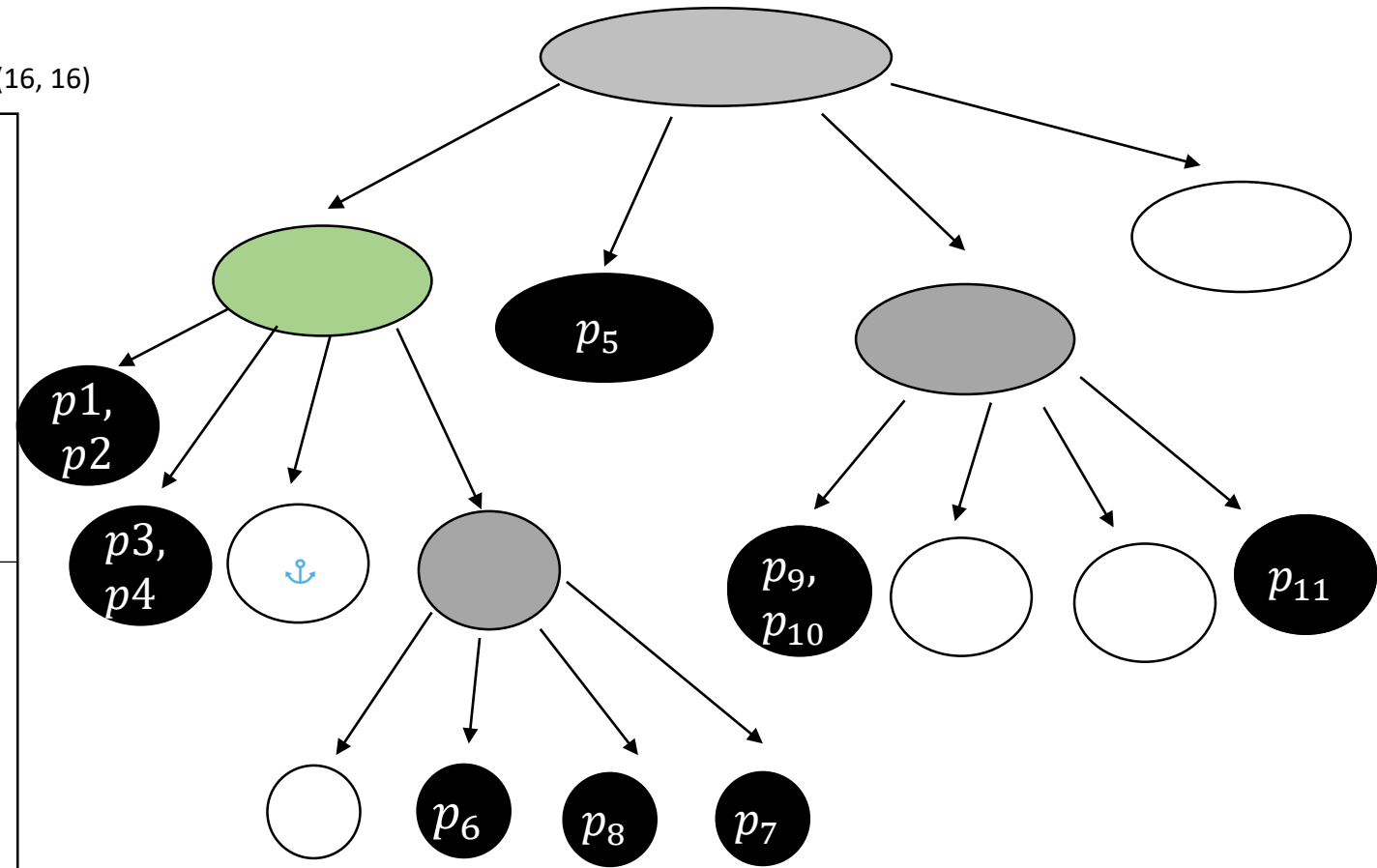
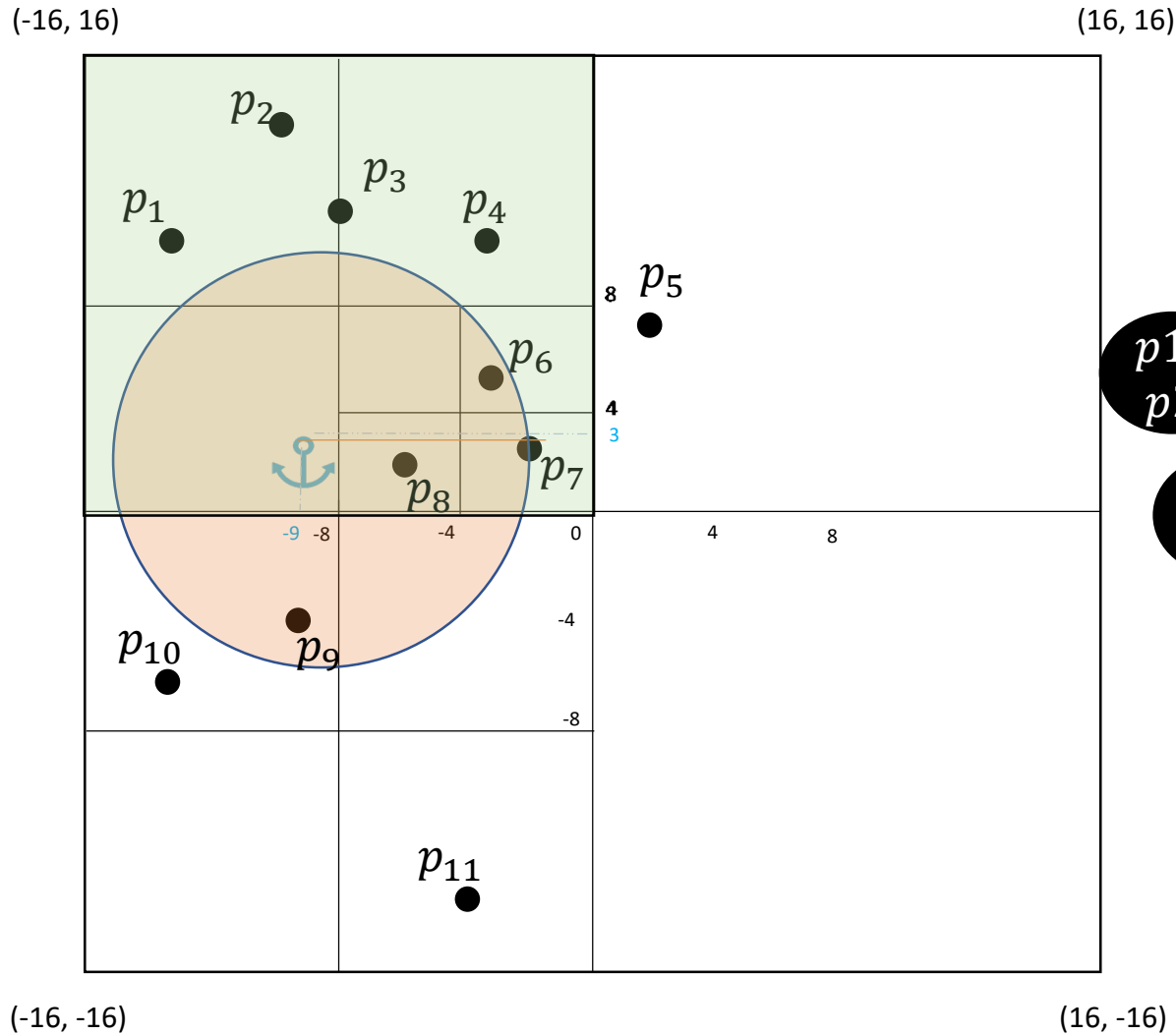


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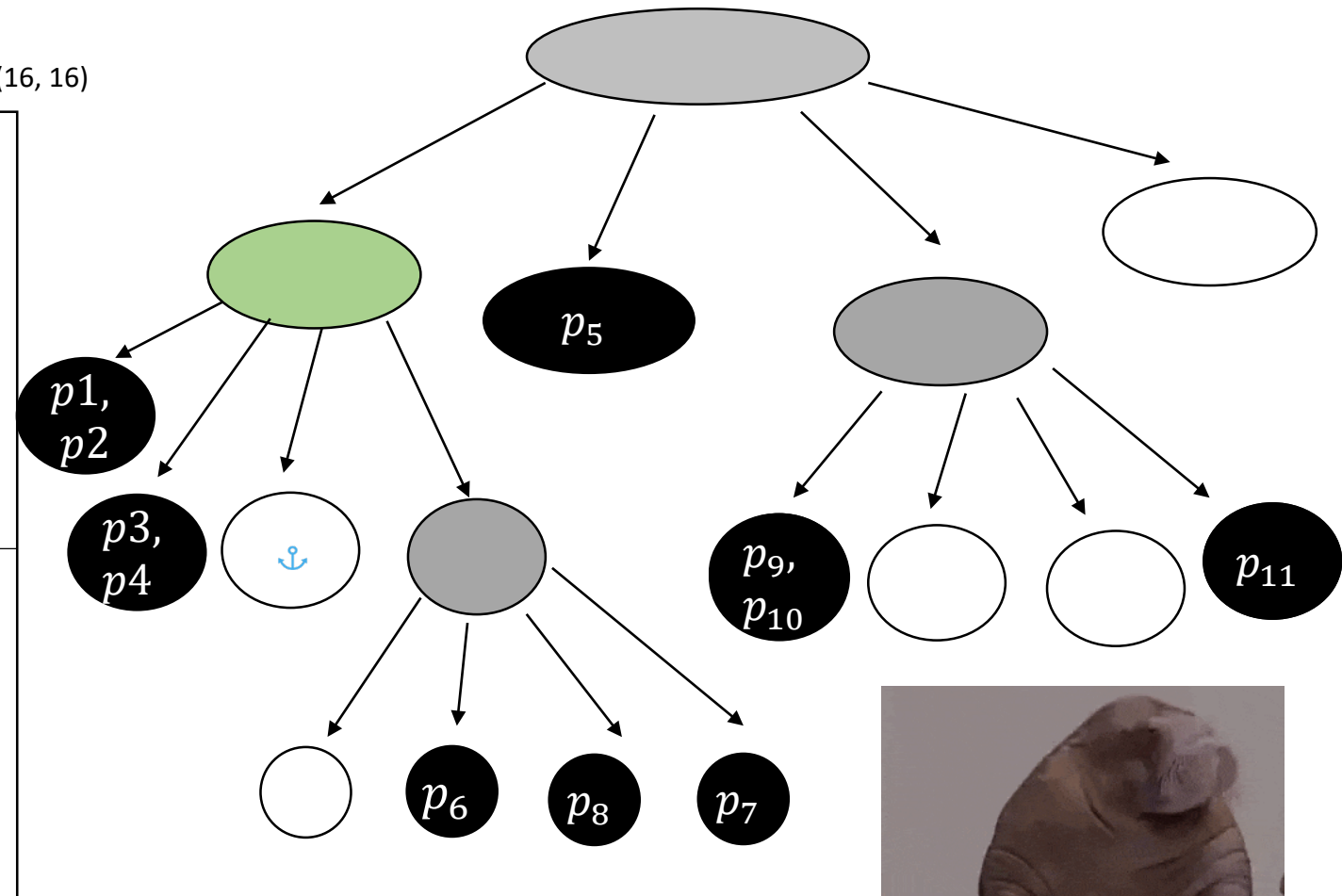
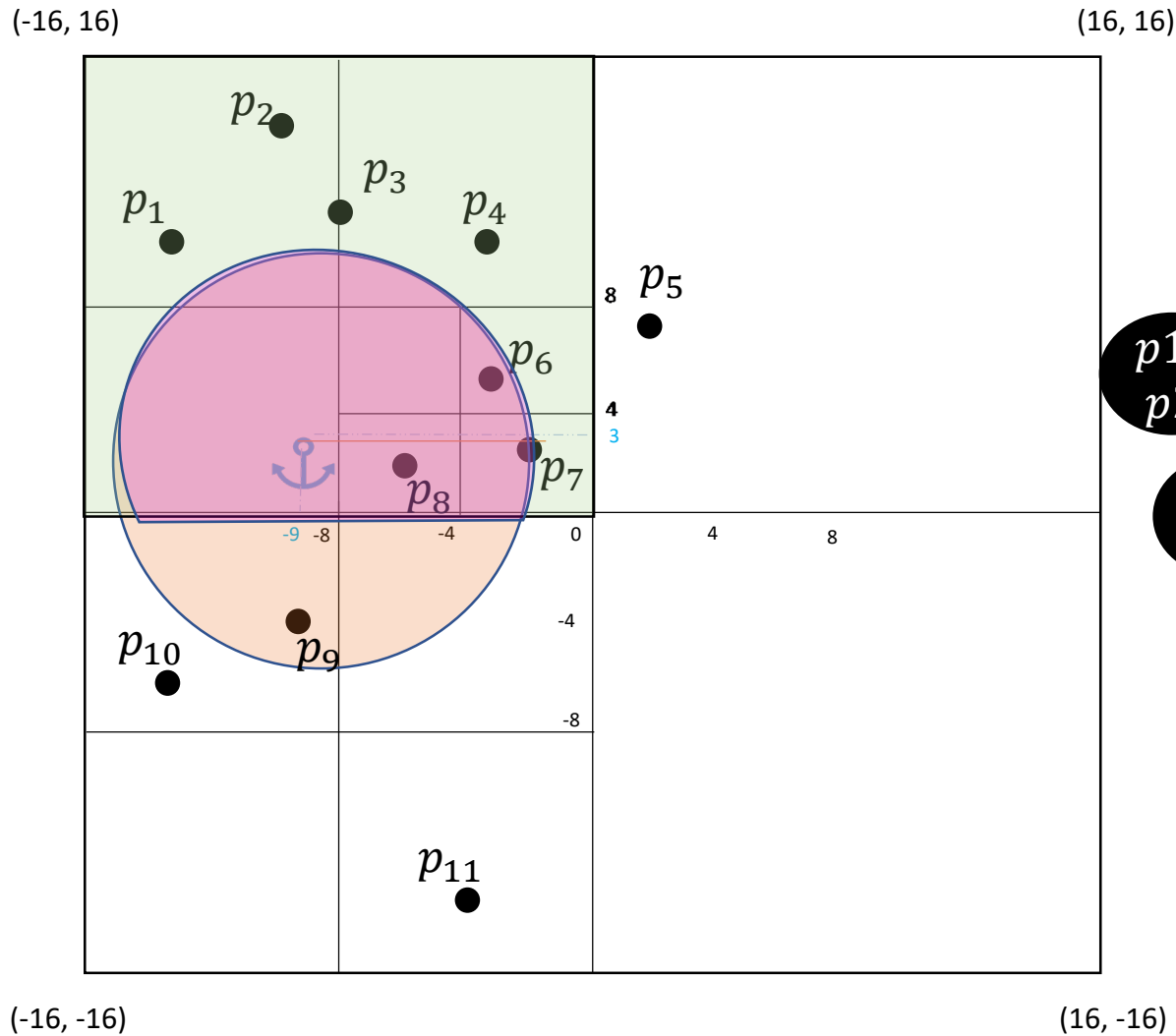
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



Is there a **non-zero intersection** between range and subtree?

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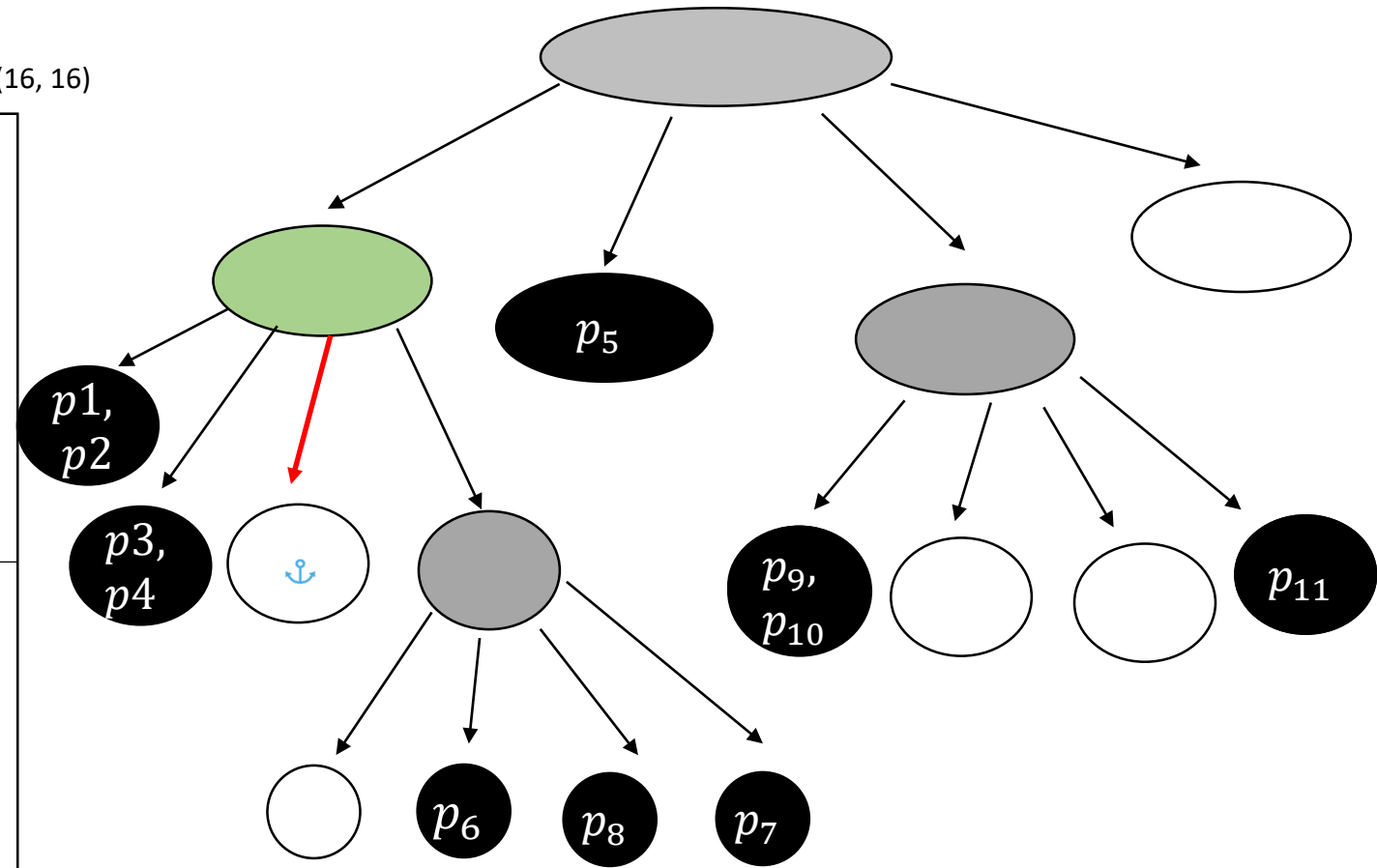
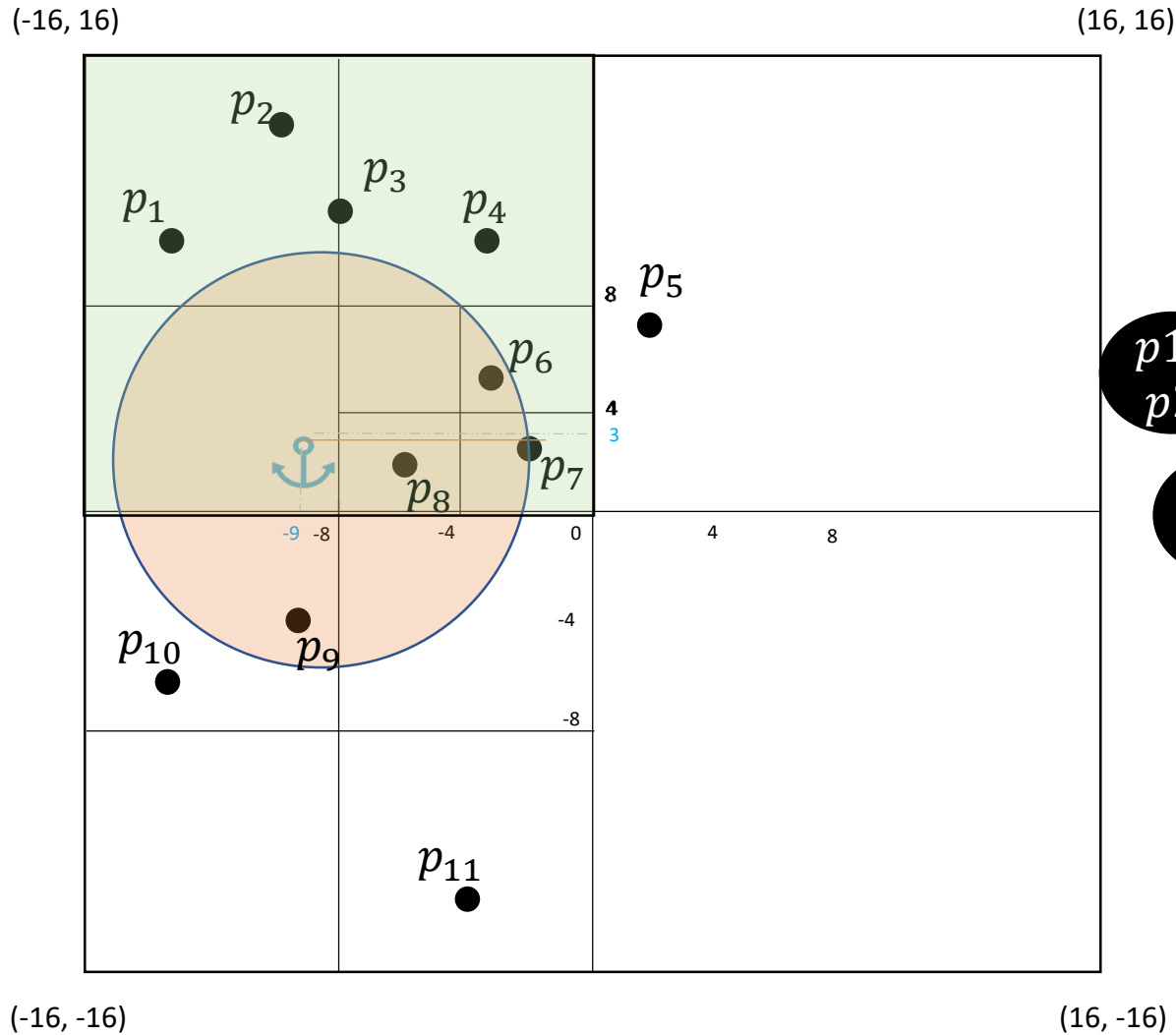
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



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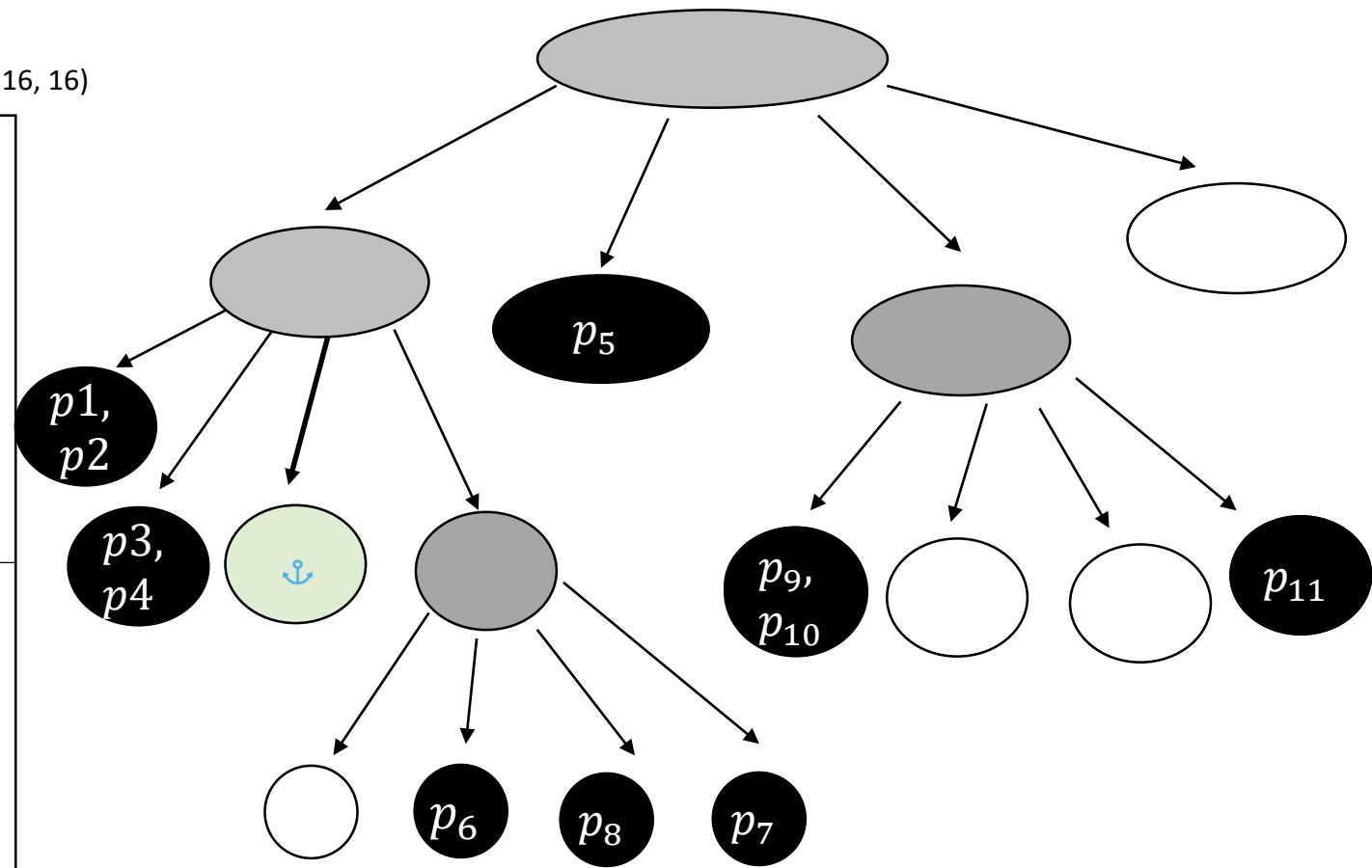
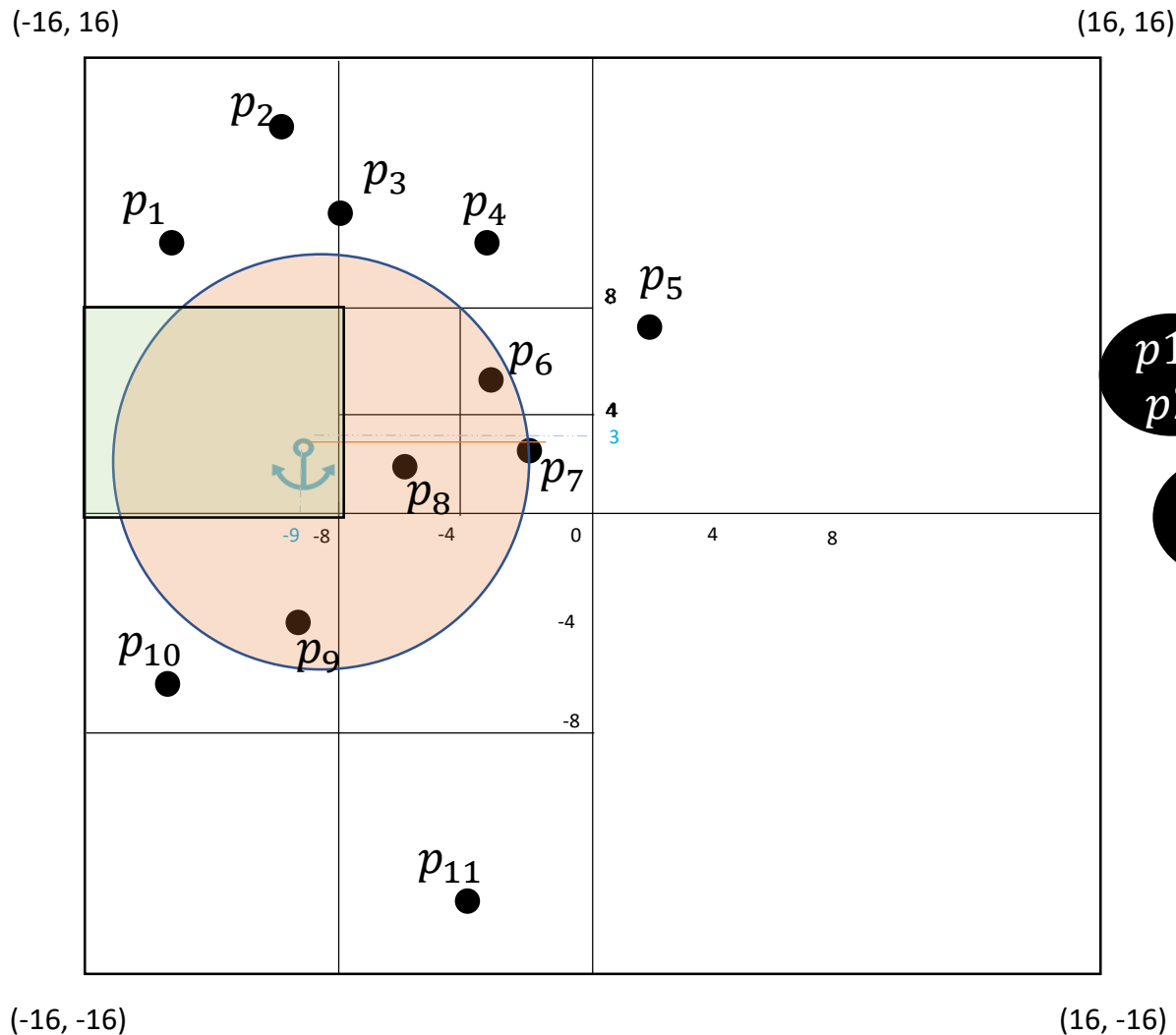
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



**Aggressively** approach the anchor

# Example of range query in PR-QuadTree

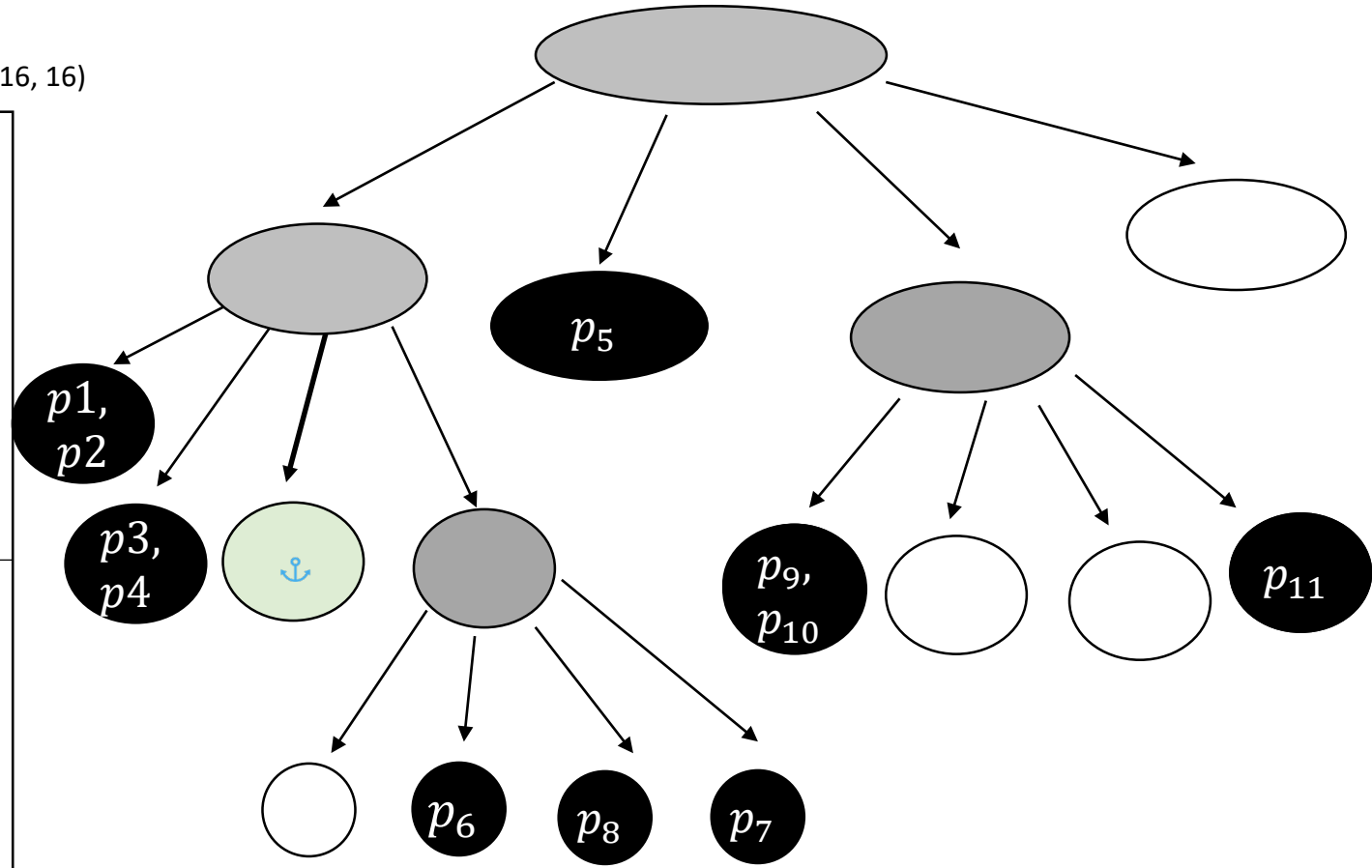
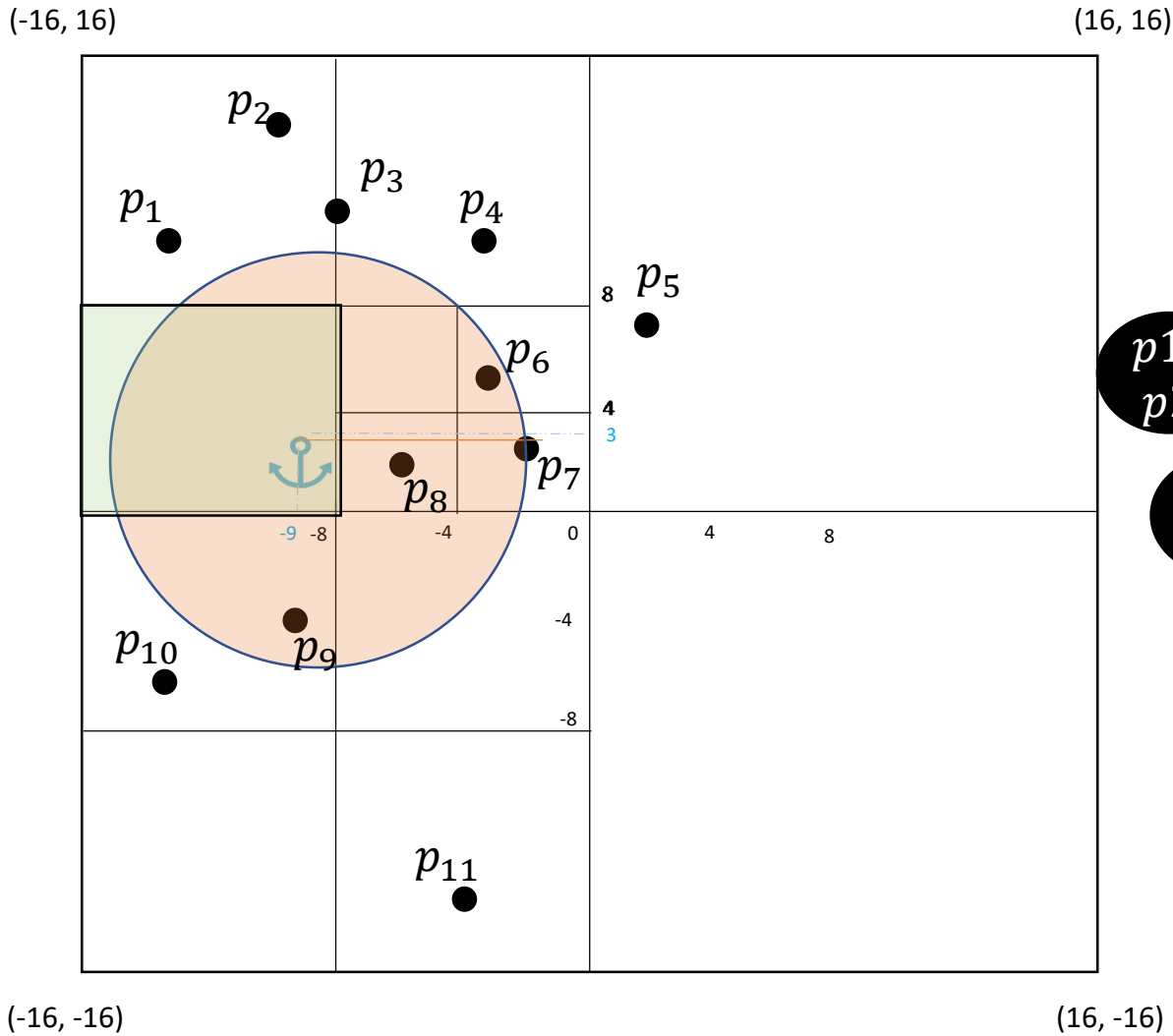
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



**Aggressively** approach the anchor.

# Example of range query in PR-QuadTree

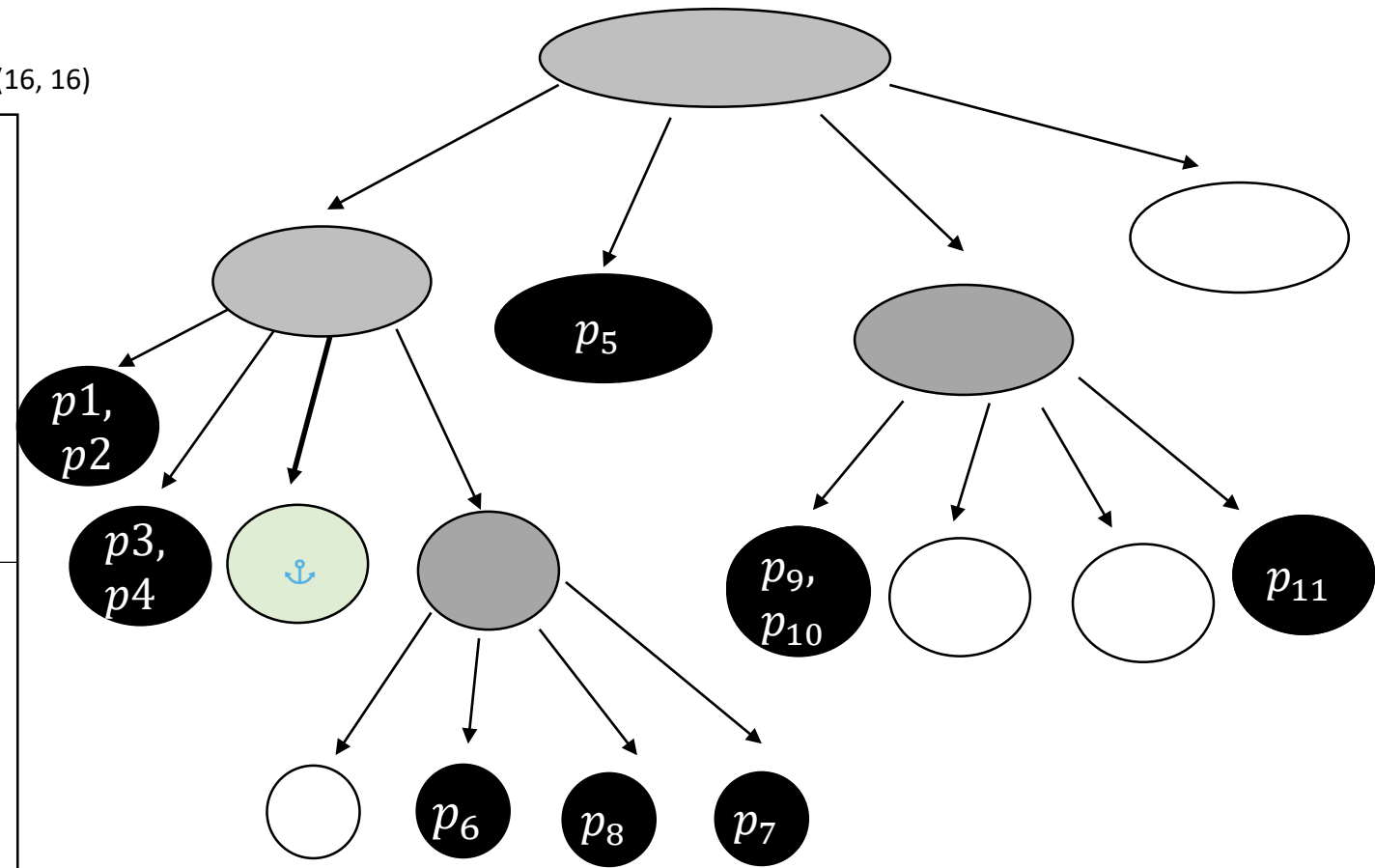
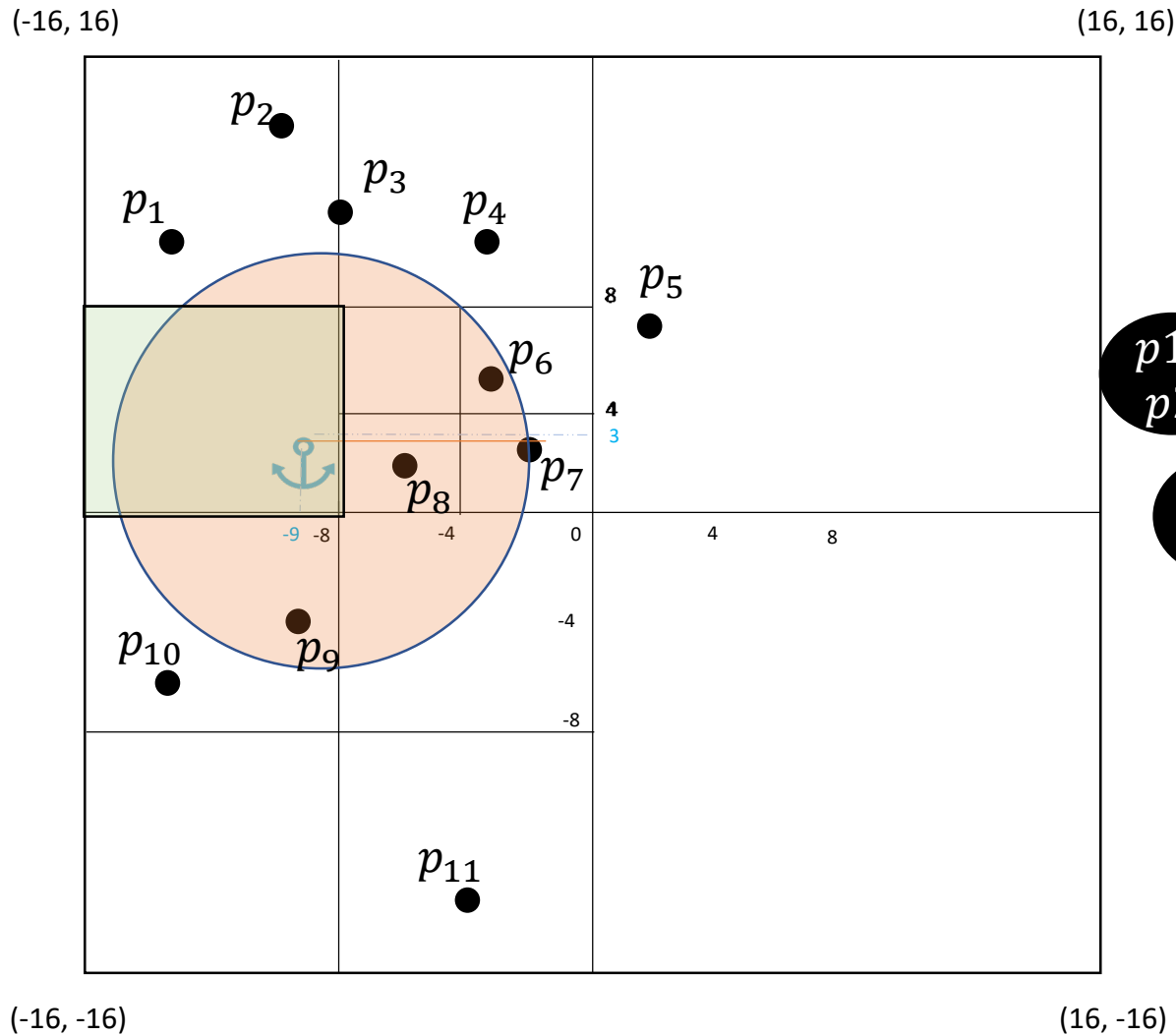
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



While there exists a non-zero intersection, the current node is also a `white` node, so he can't quite contribute to the solution!

# Example of range query in PR-QuadTree

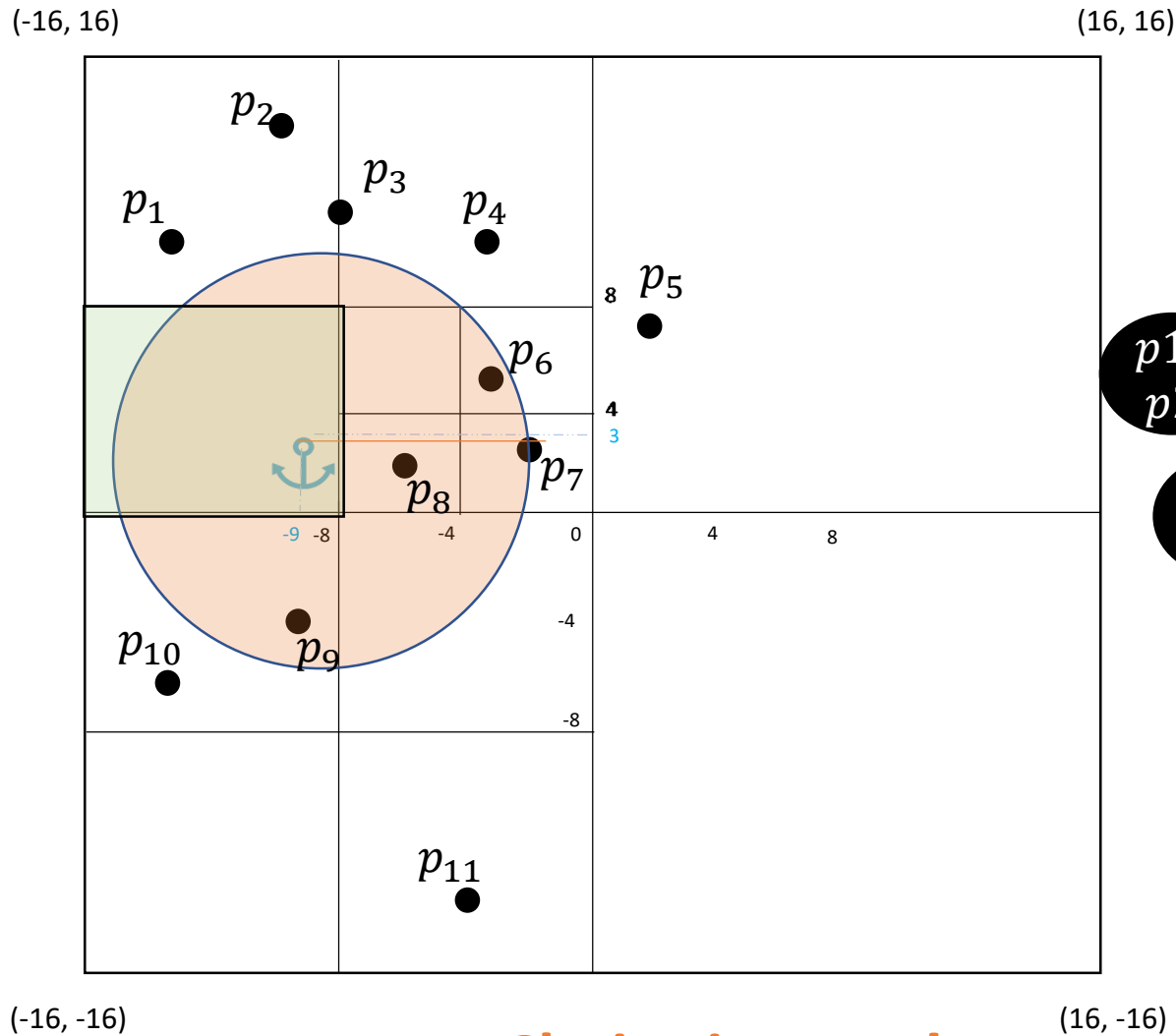
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



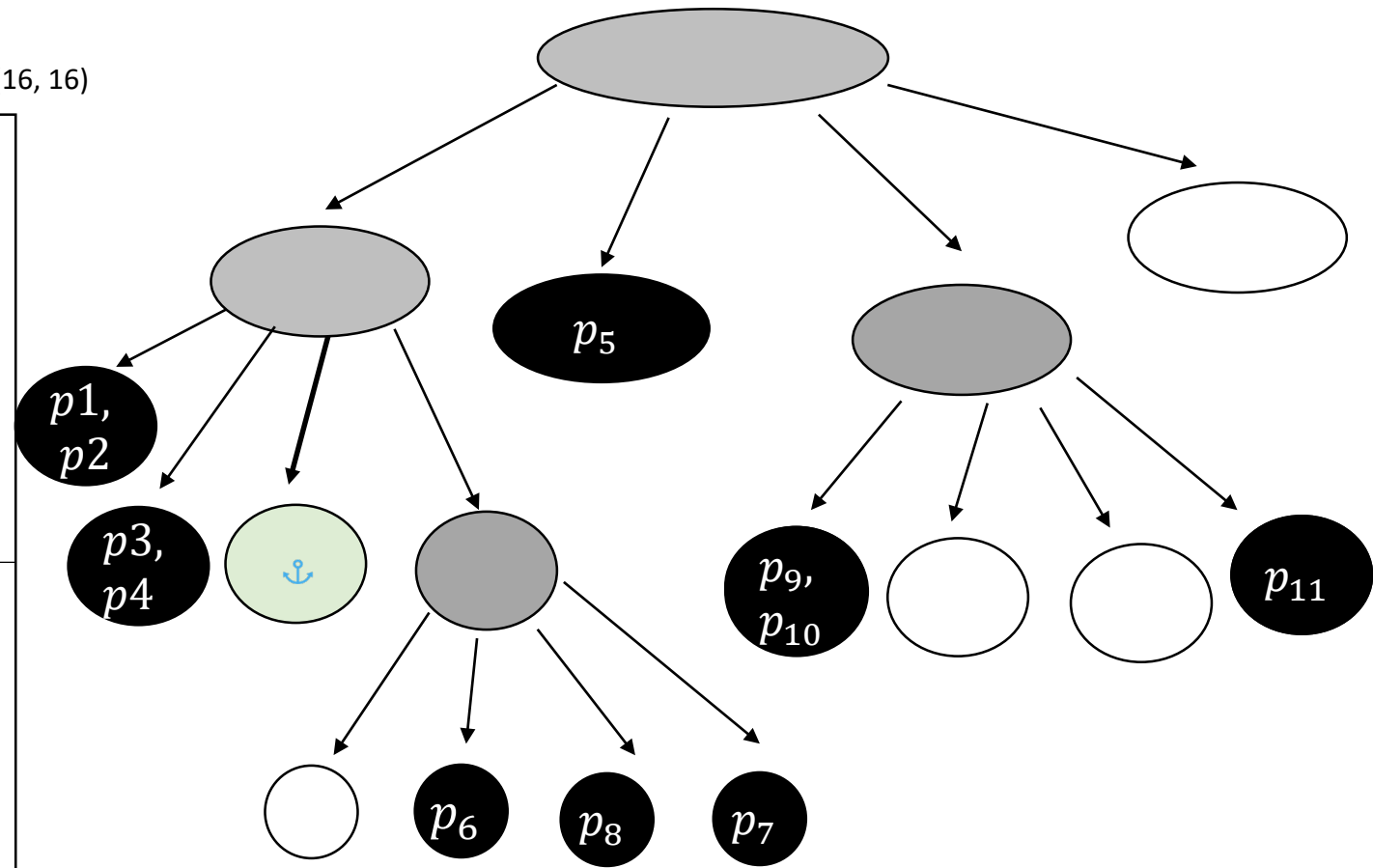
Implementation detail: do I actually **visit** a white node first? Do I check for intersection first to have a common implementation?

# Example of range query in PR-QuadTree

anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



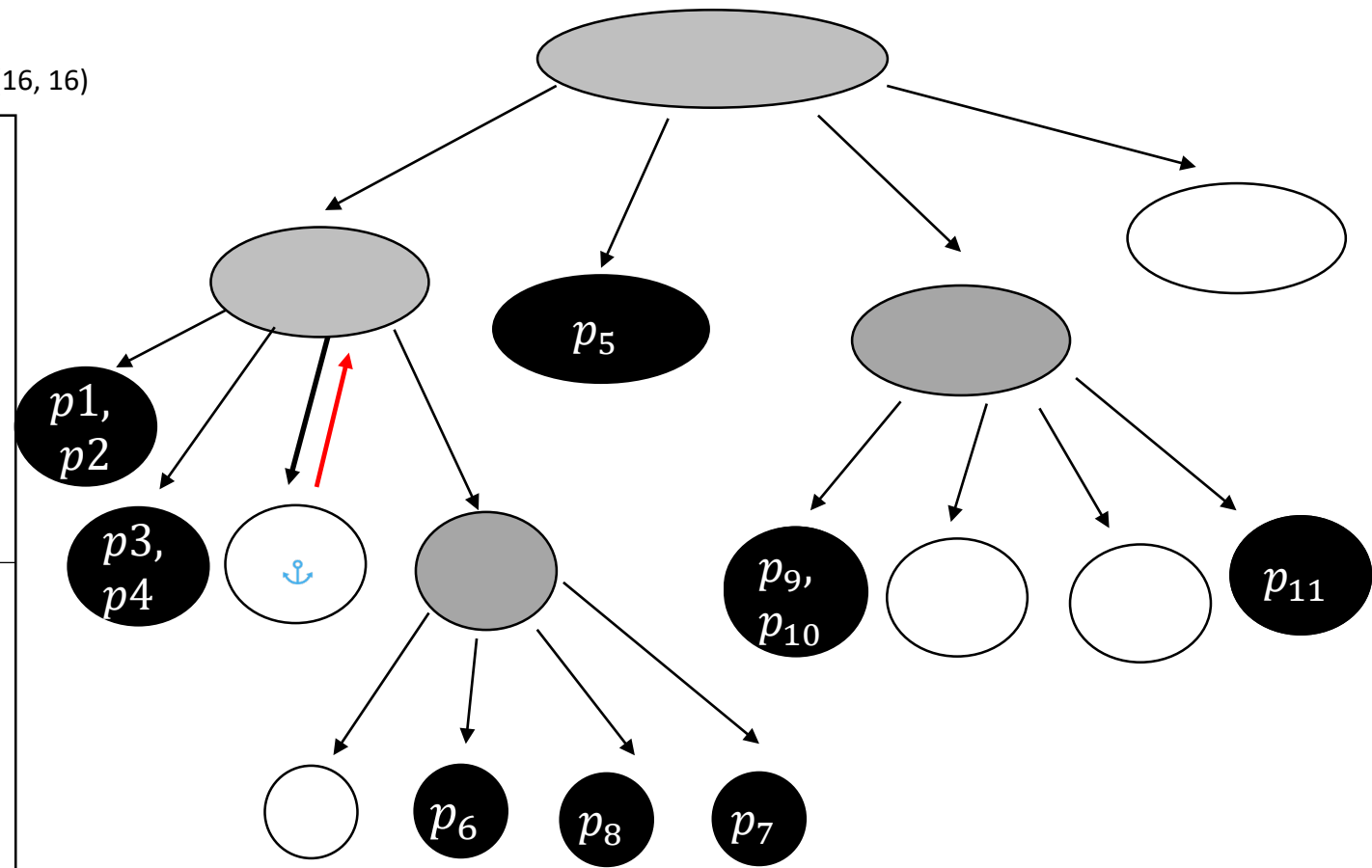
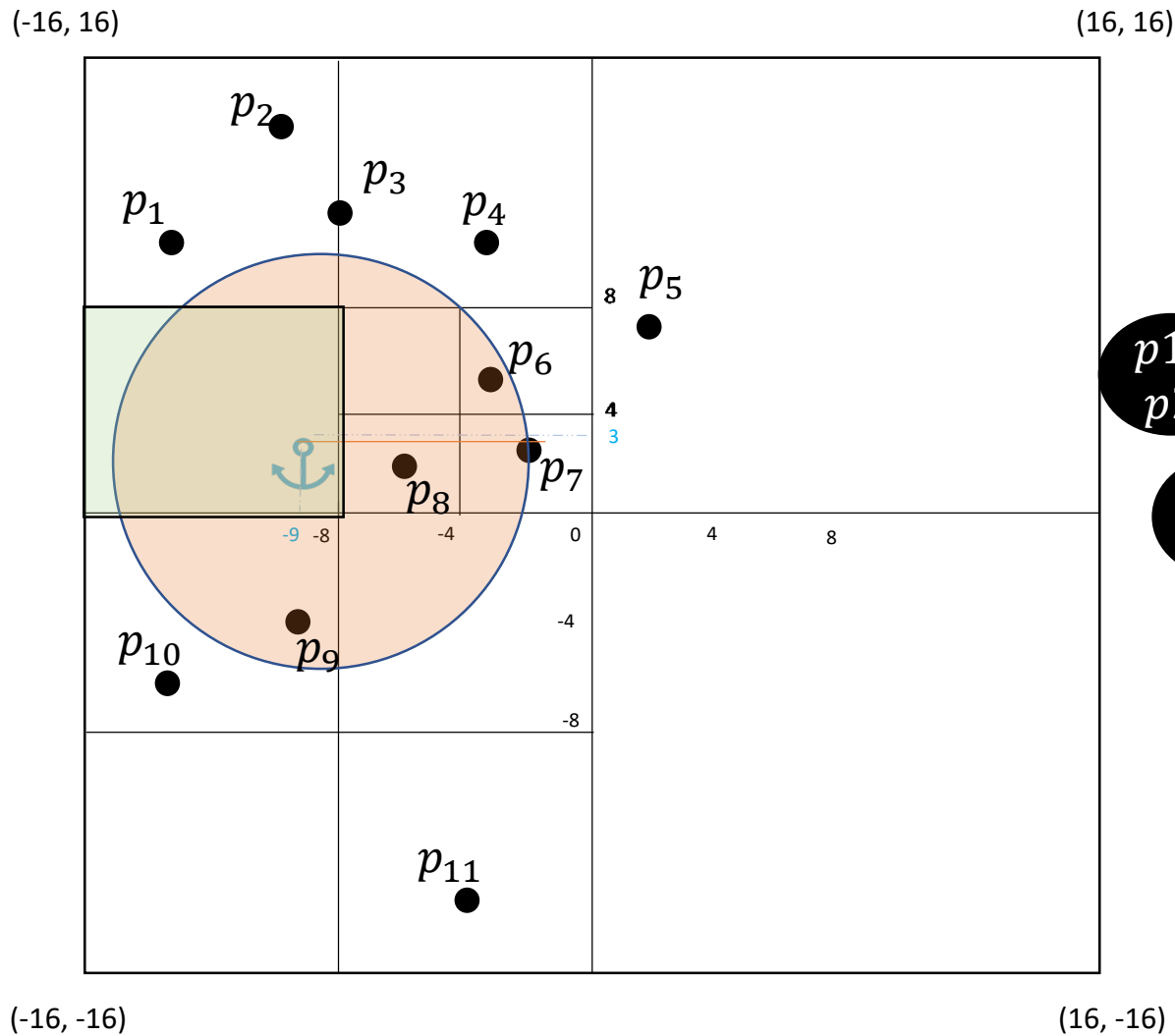
Choice is yours!



Implementation detail: do I actually **visit** a white node first? Do I check for intersection first to have a common implementation?

# Example of range query in PR-QuadTree

anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$

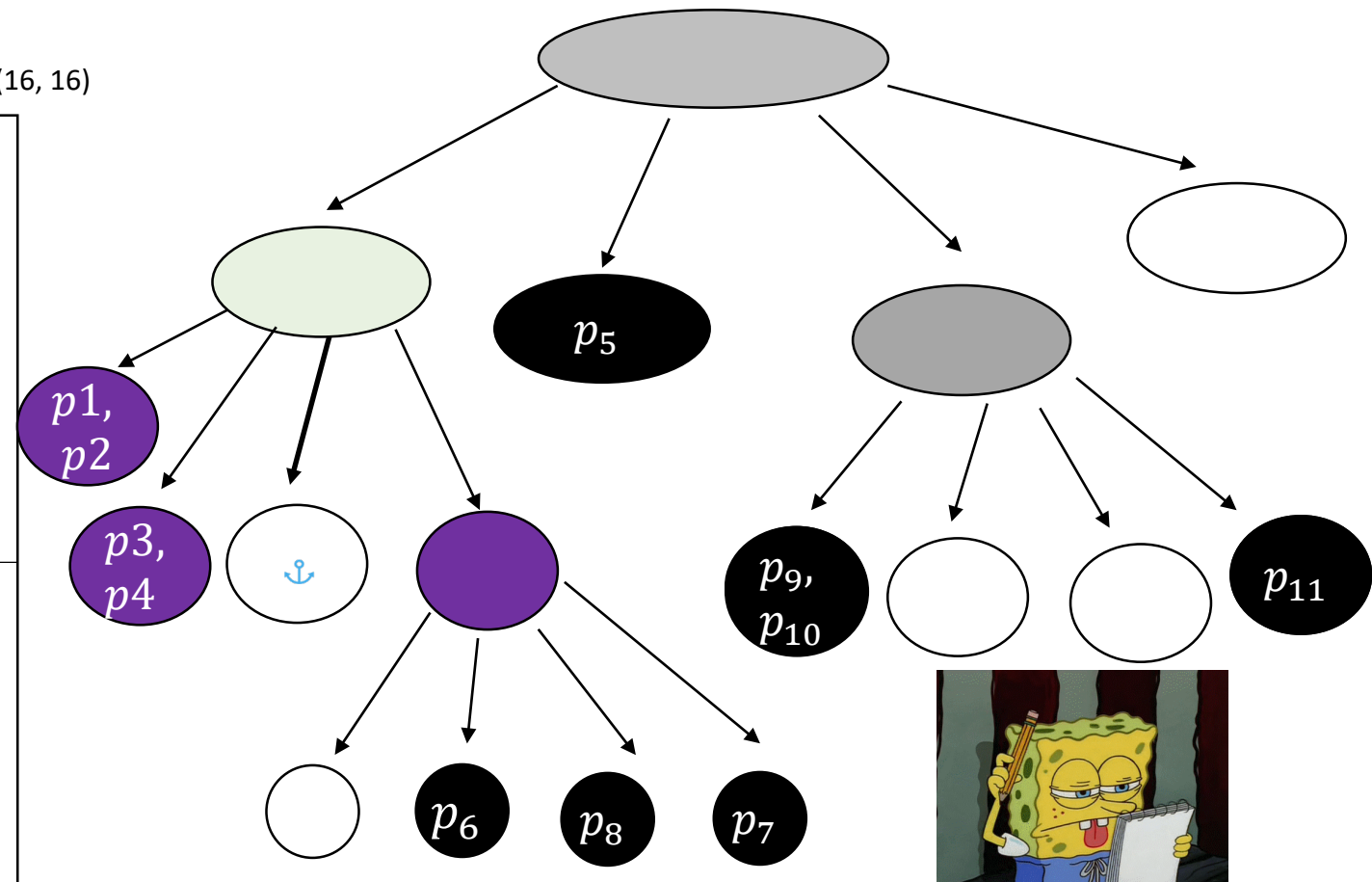
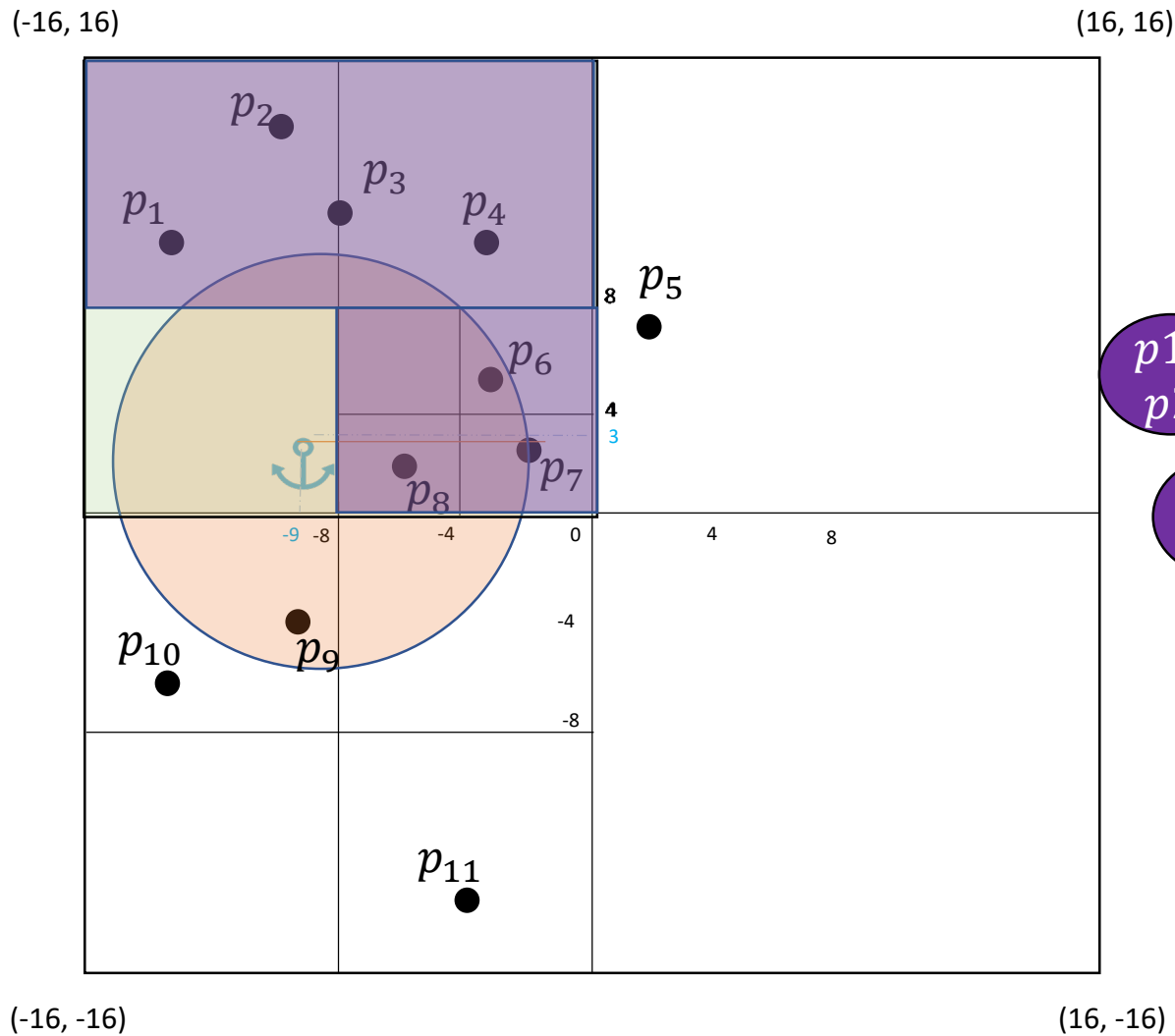


**Backtrack!**



# Example of range query in PR-QuadTree

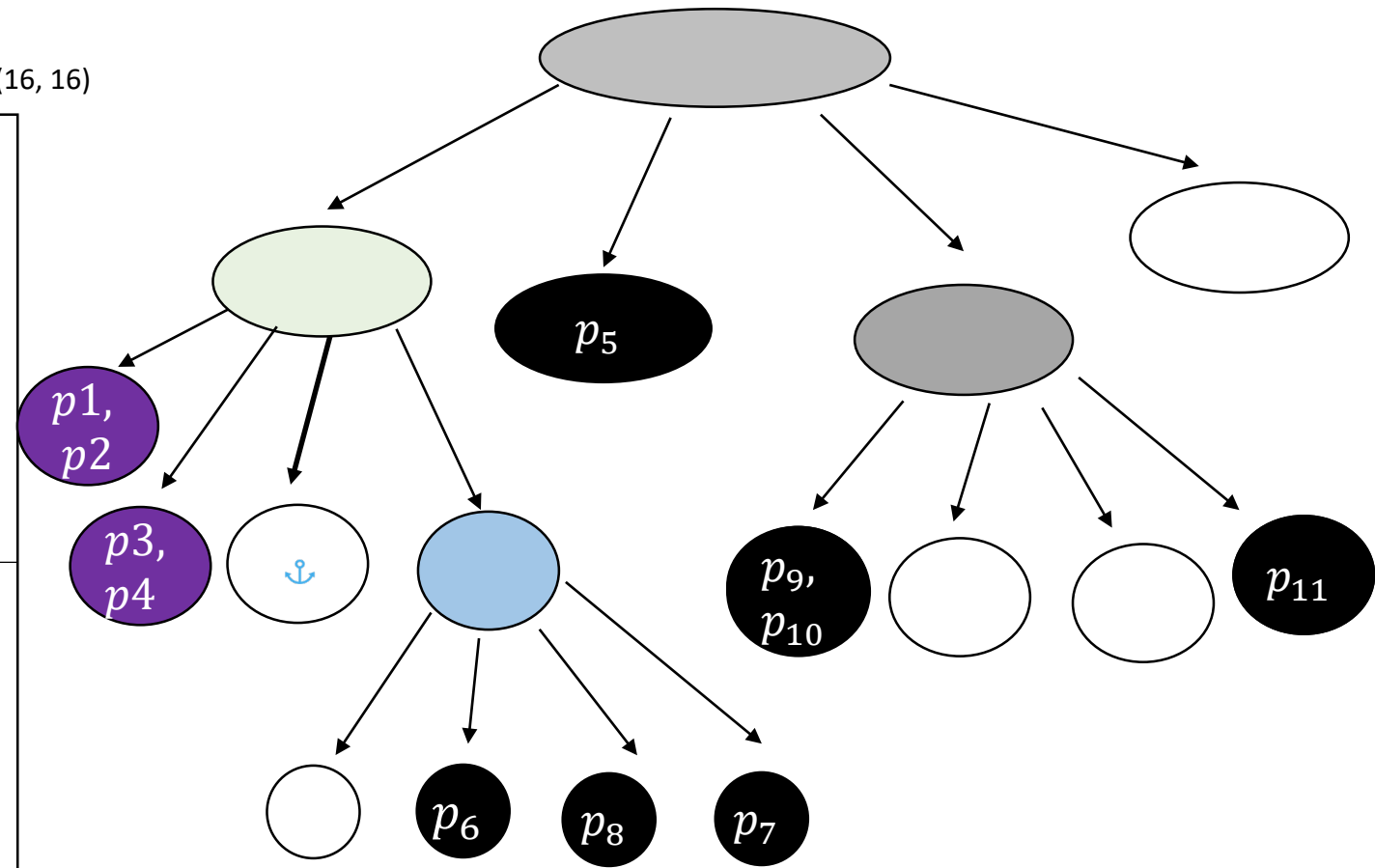
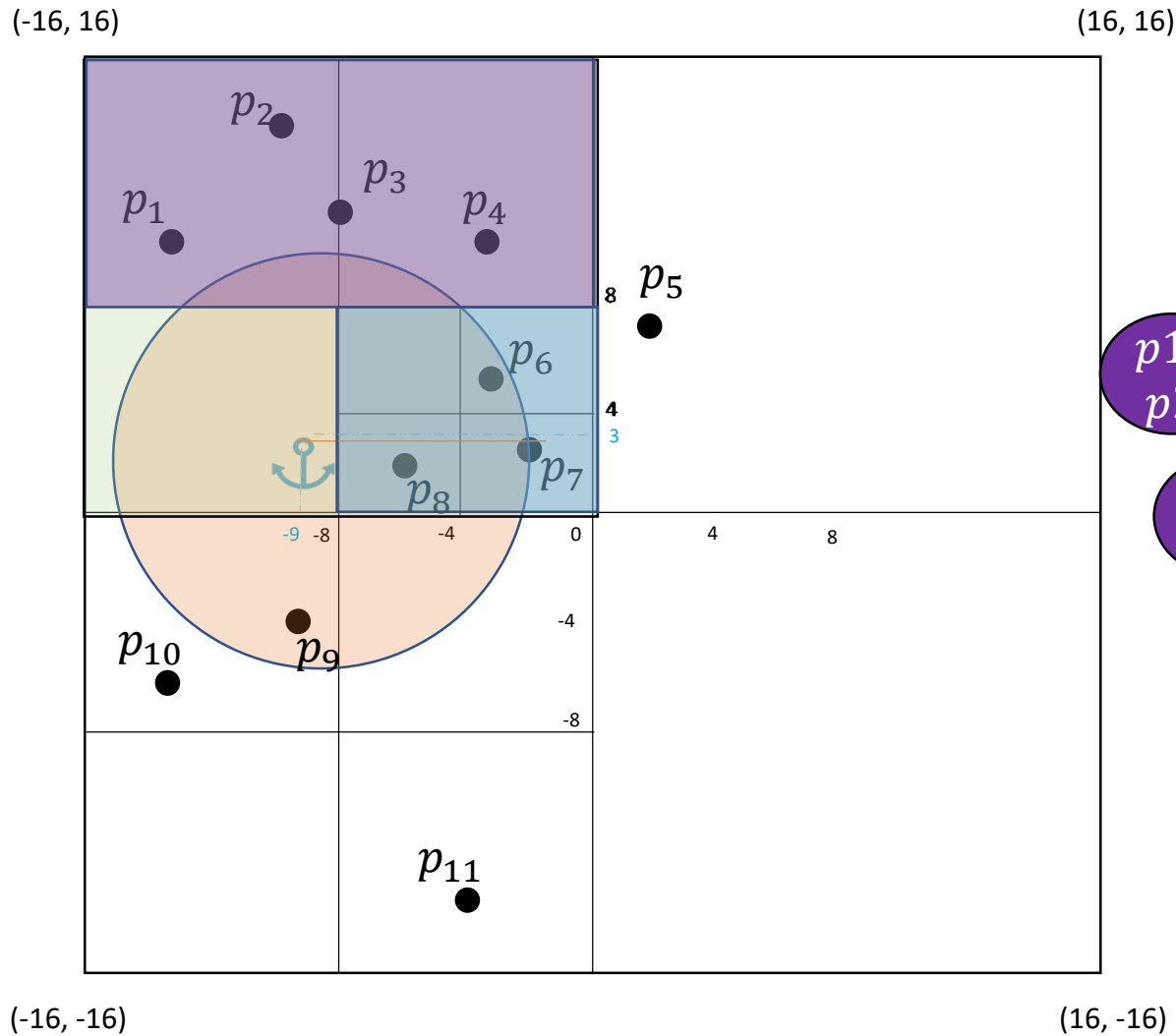
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



*Which other subtree do I visit first?*

# Example of range query in PR-QuadTree

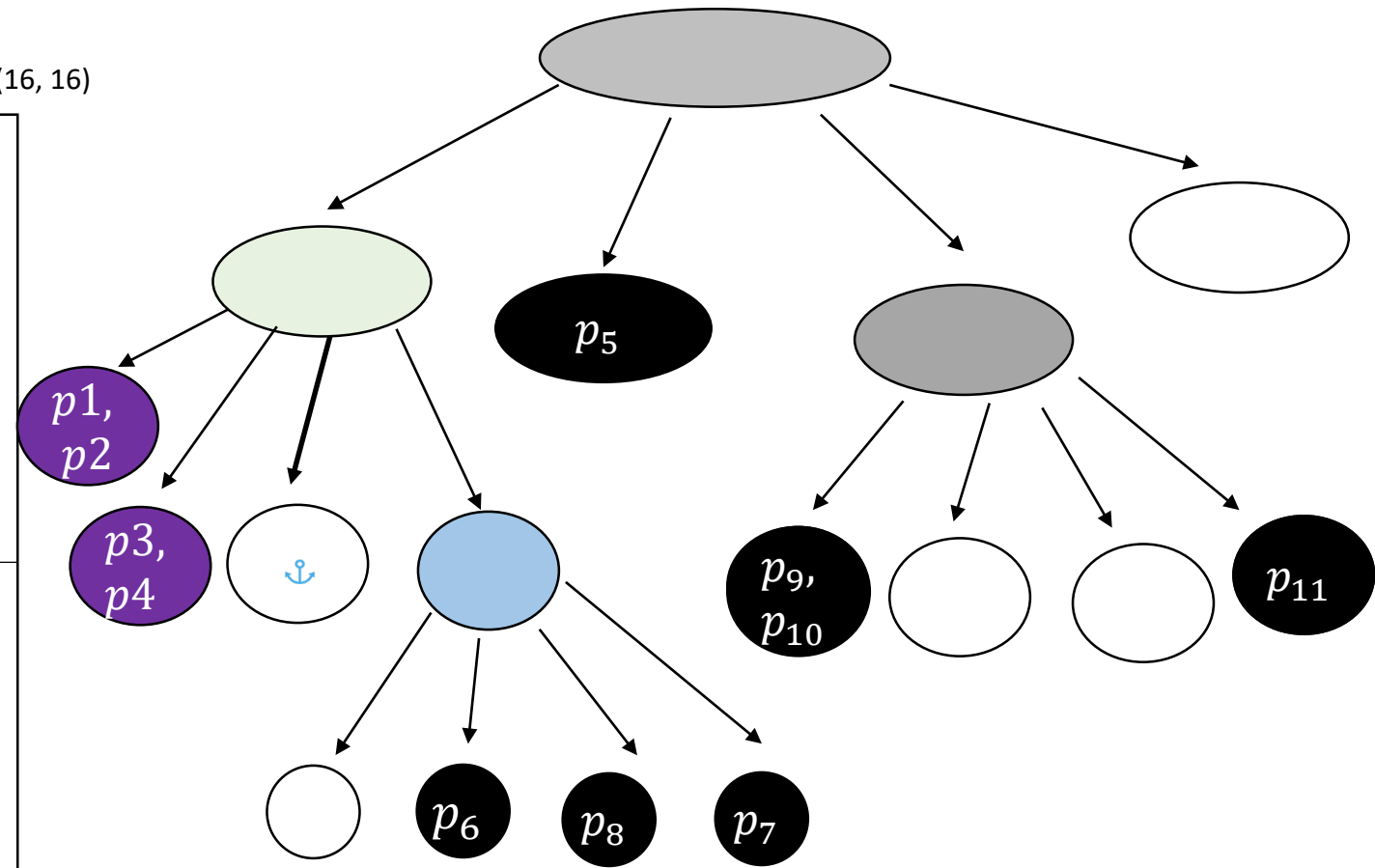
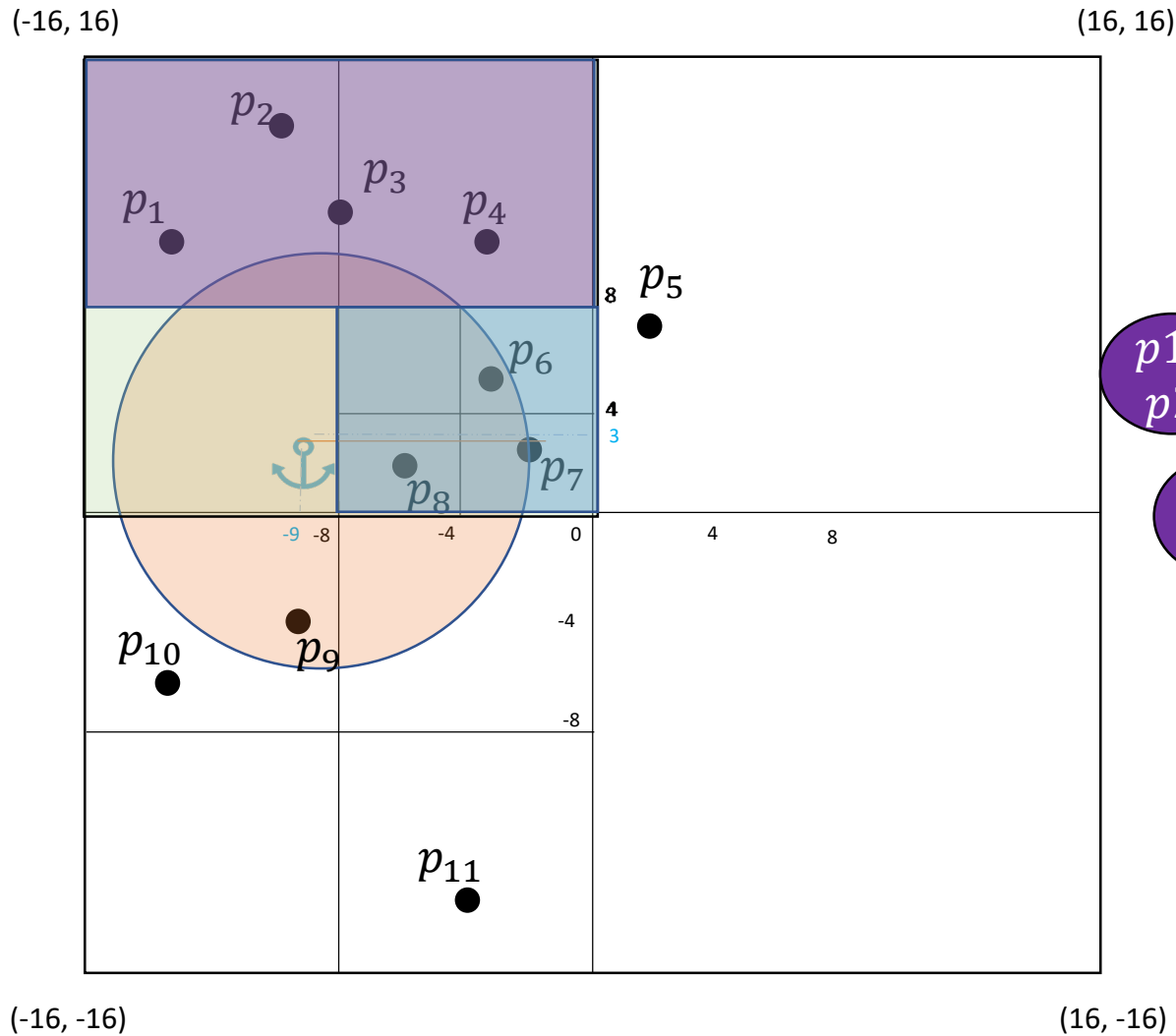
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



Seems as if the **SE child** would bear the most fruit fastest, since it has **the largest area of intersection** with the range...

# Example of range query in PR-QuadTree

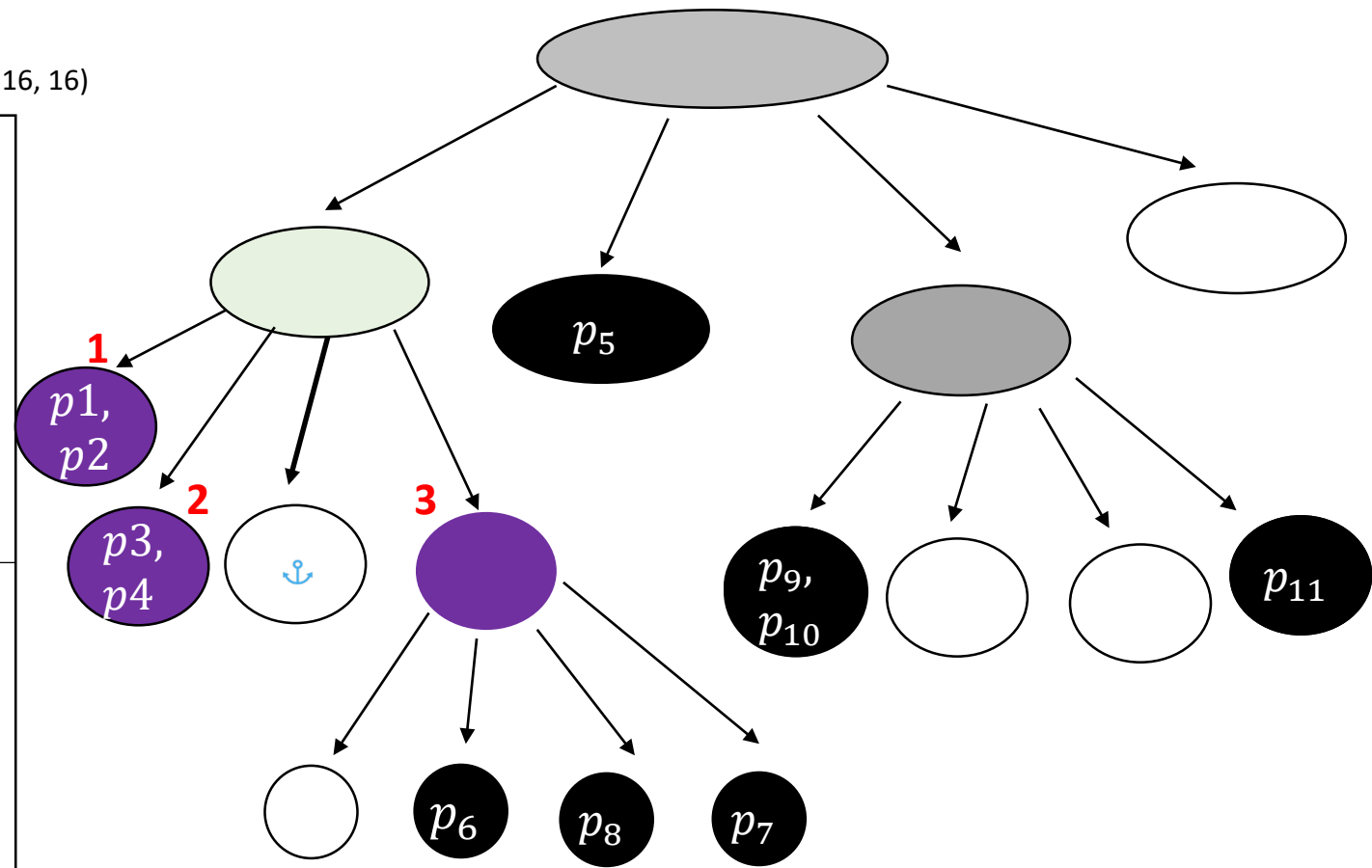
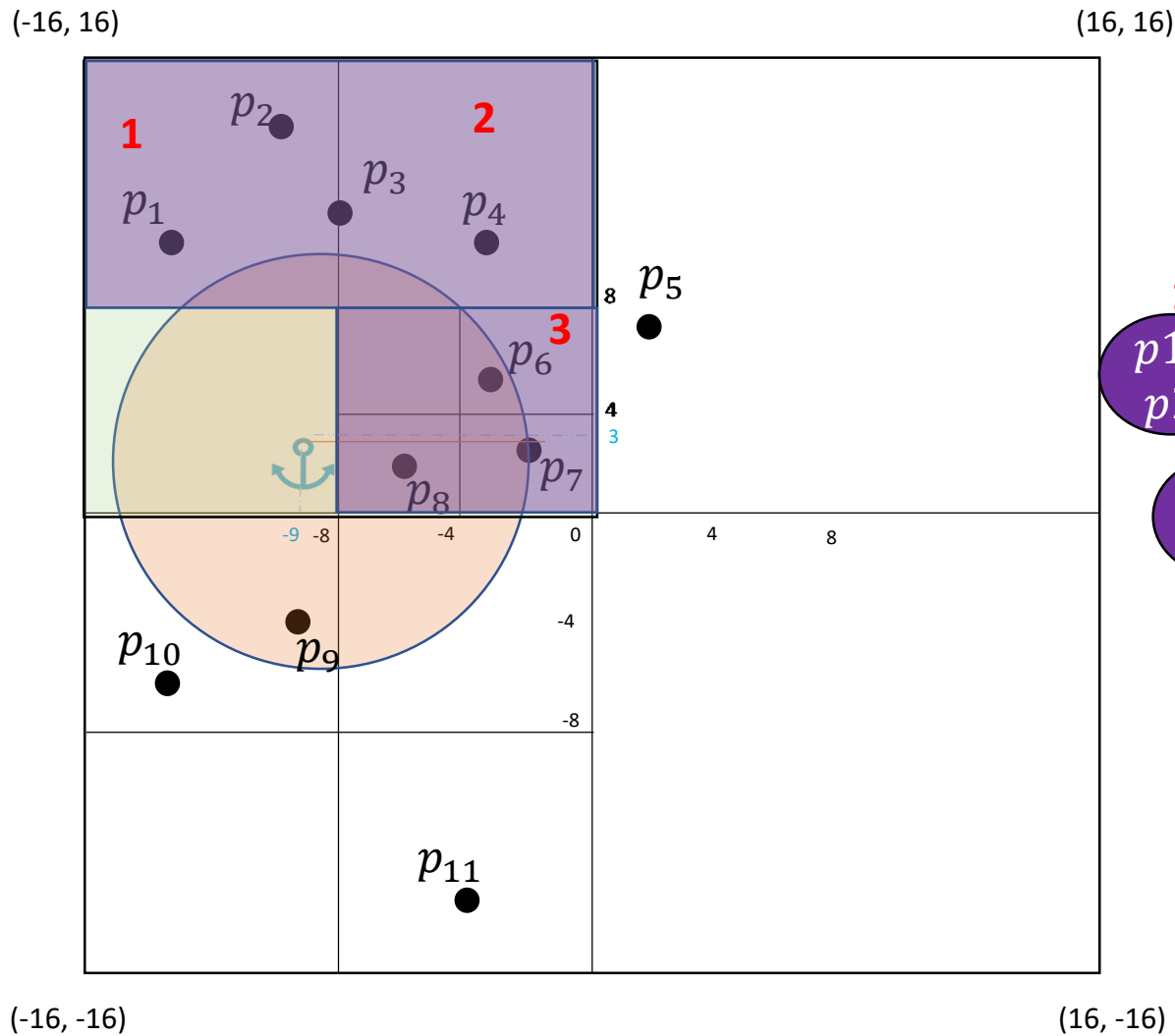
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



**BUT UNDER NO CIRCUMSTANCES** WE REQUIRE THAT YOU CALCULATE THIS AREA....

# Example of range query in PR-QuadTree

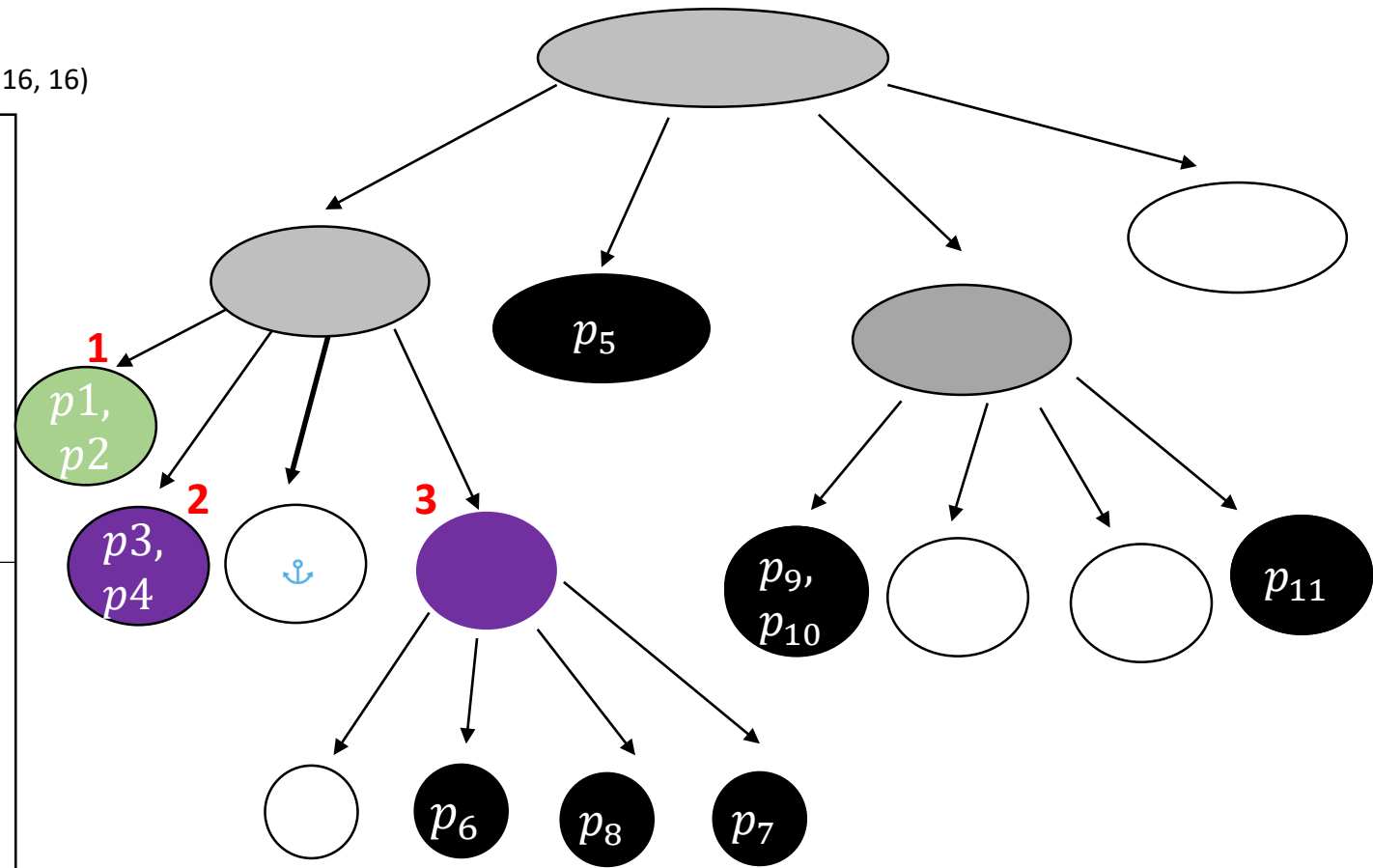
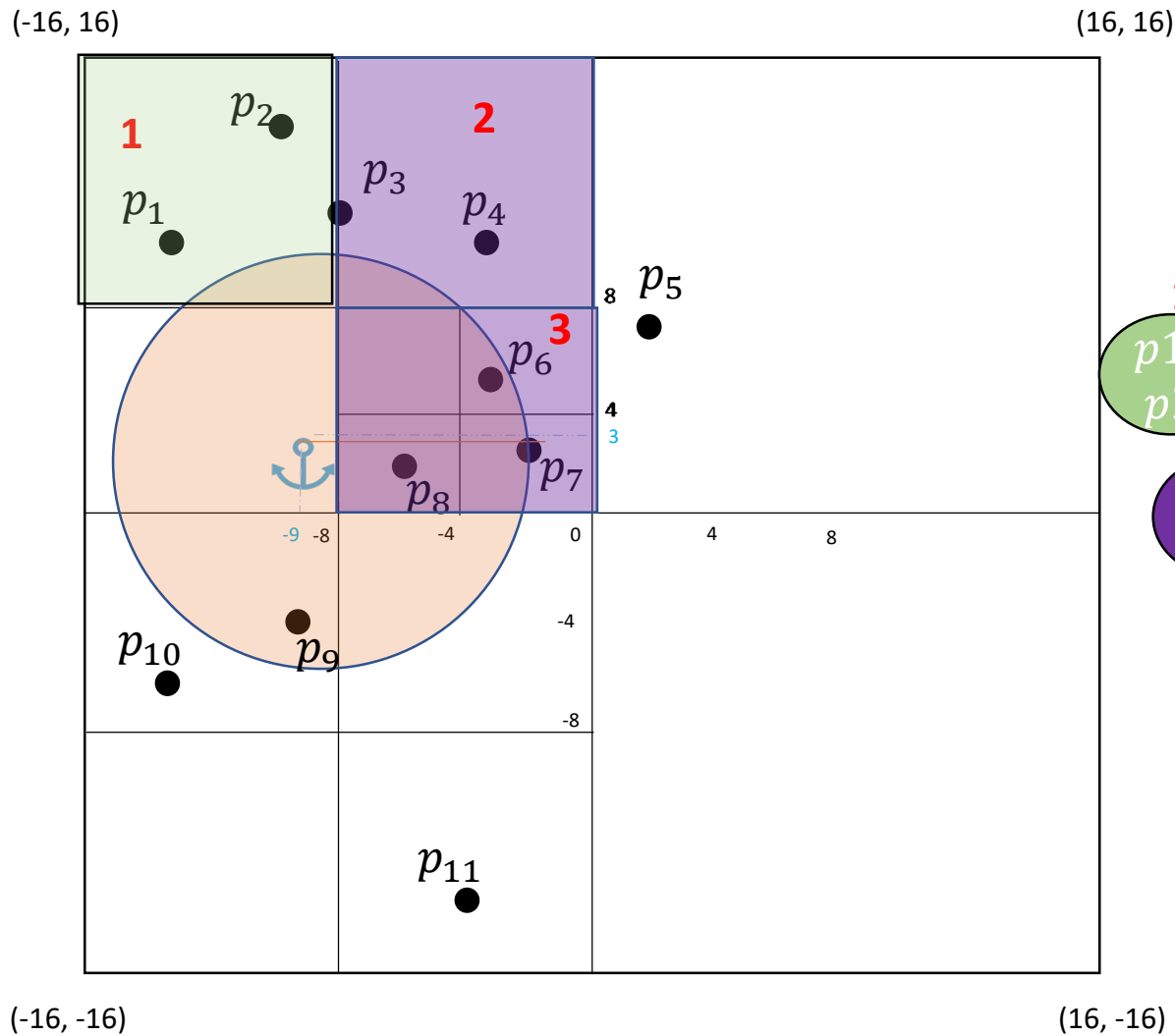
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



*Just do **Z-order** over the remaining children!*

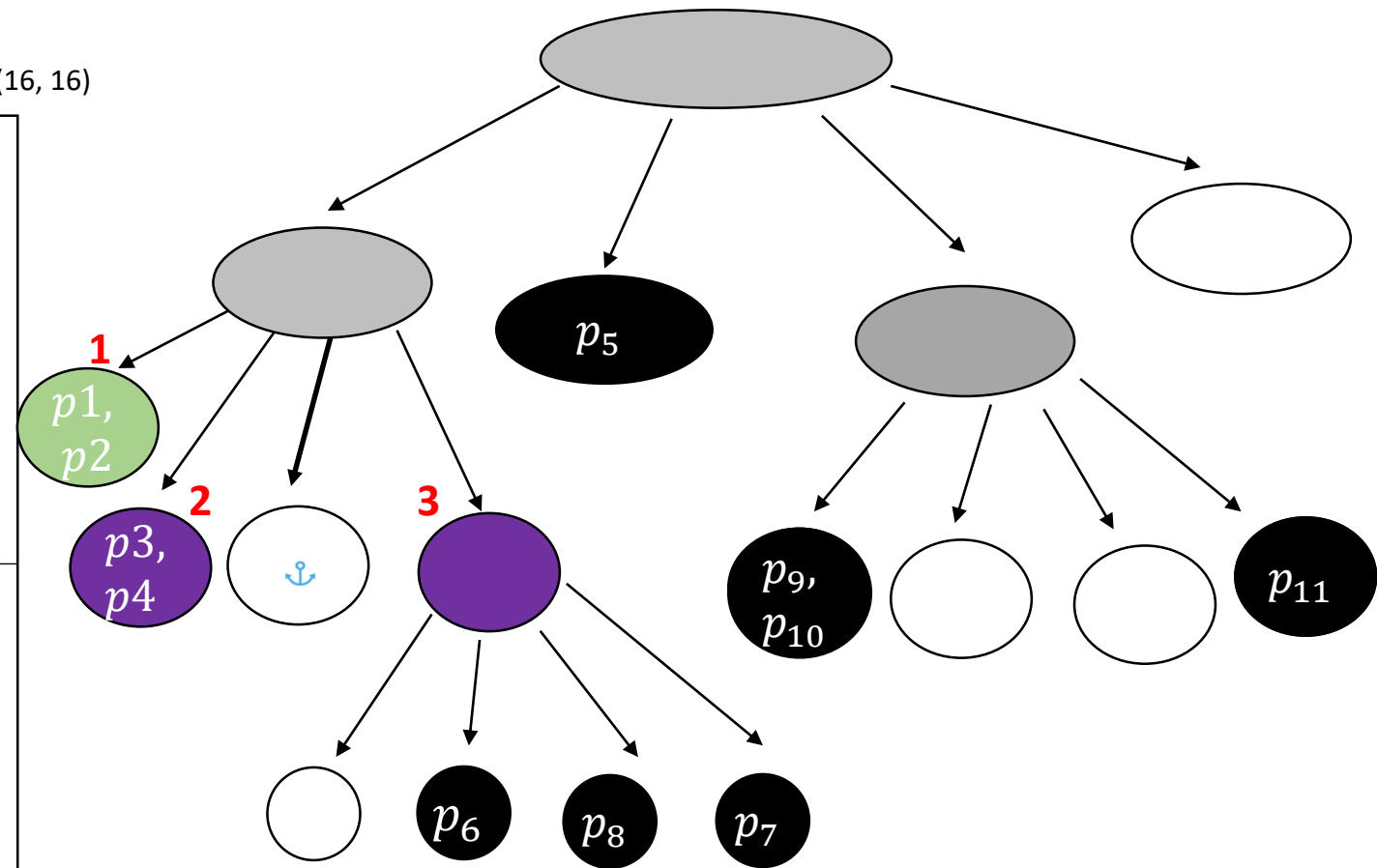
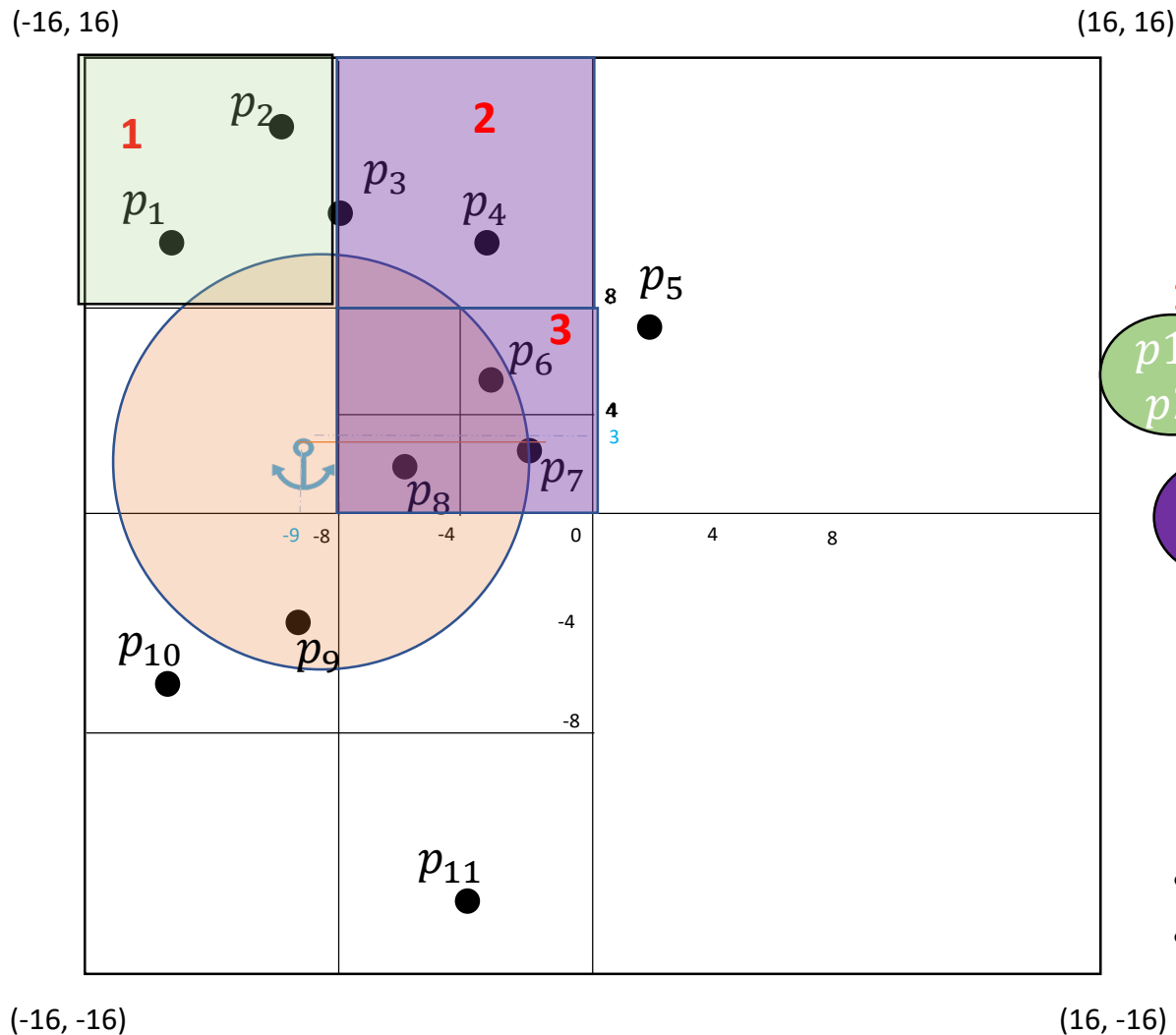
# Example of range query in PR-QuadTree

anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



# Example of range query in PR-QuadTree

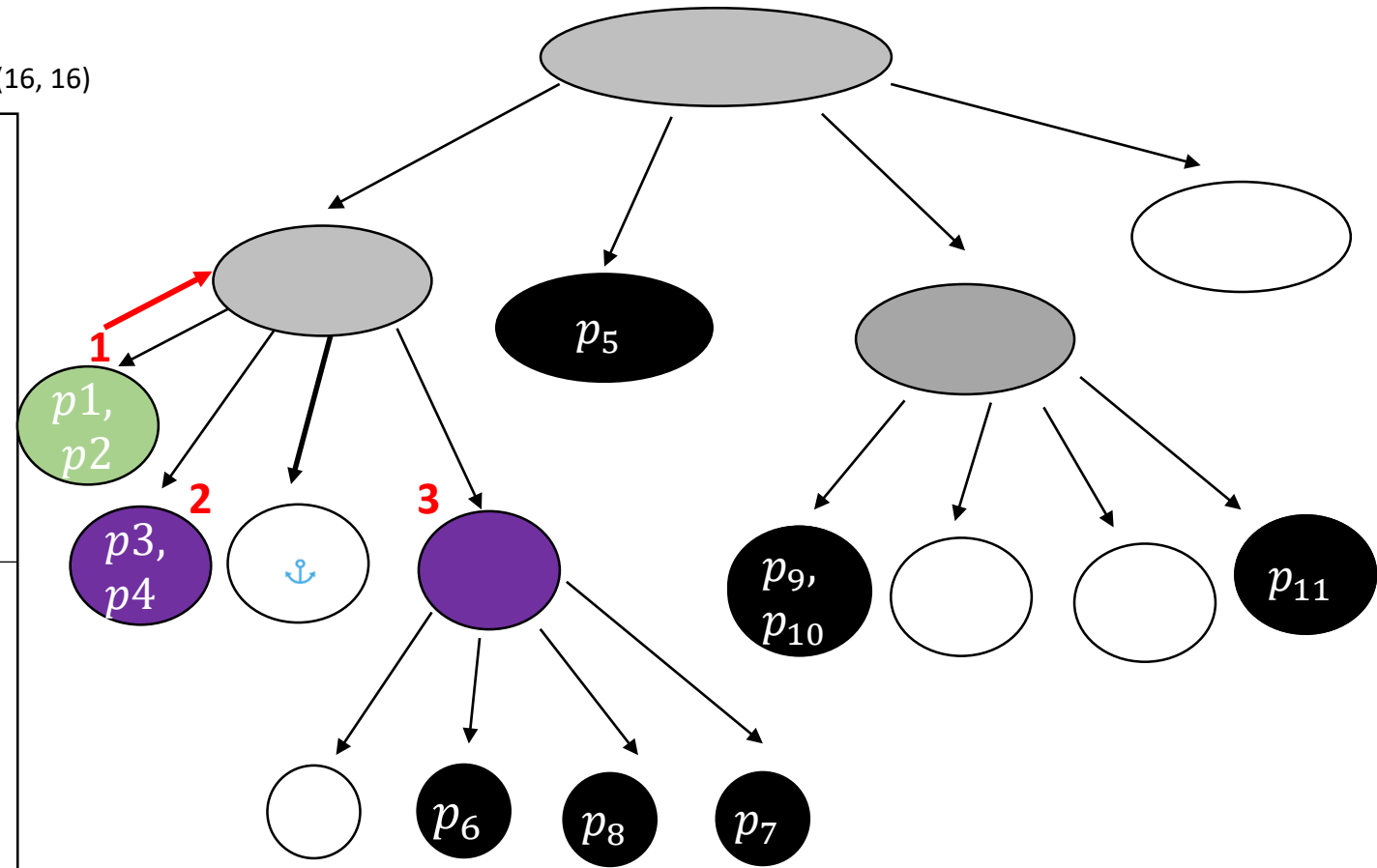
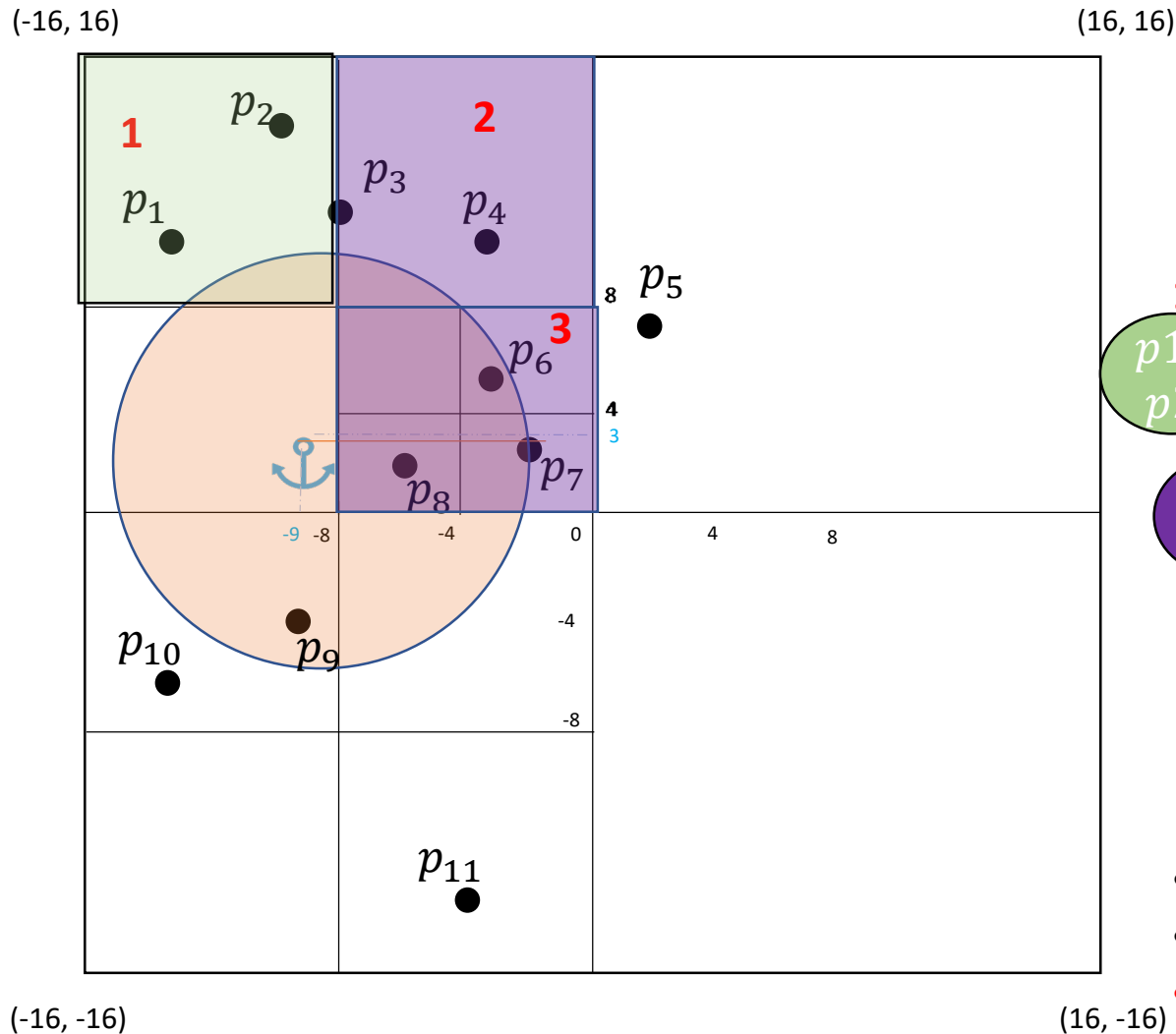
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



- Non-zero intersection, and black node!
- $p_1, p_2$  are not within the range, unfortunately!

# Example of range query in PR-QuadTree

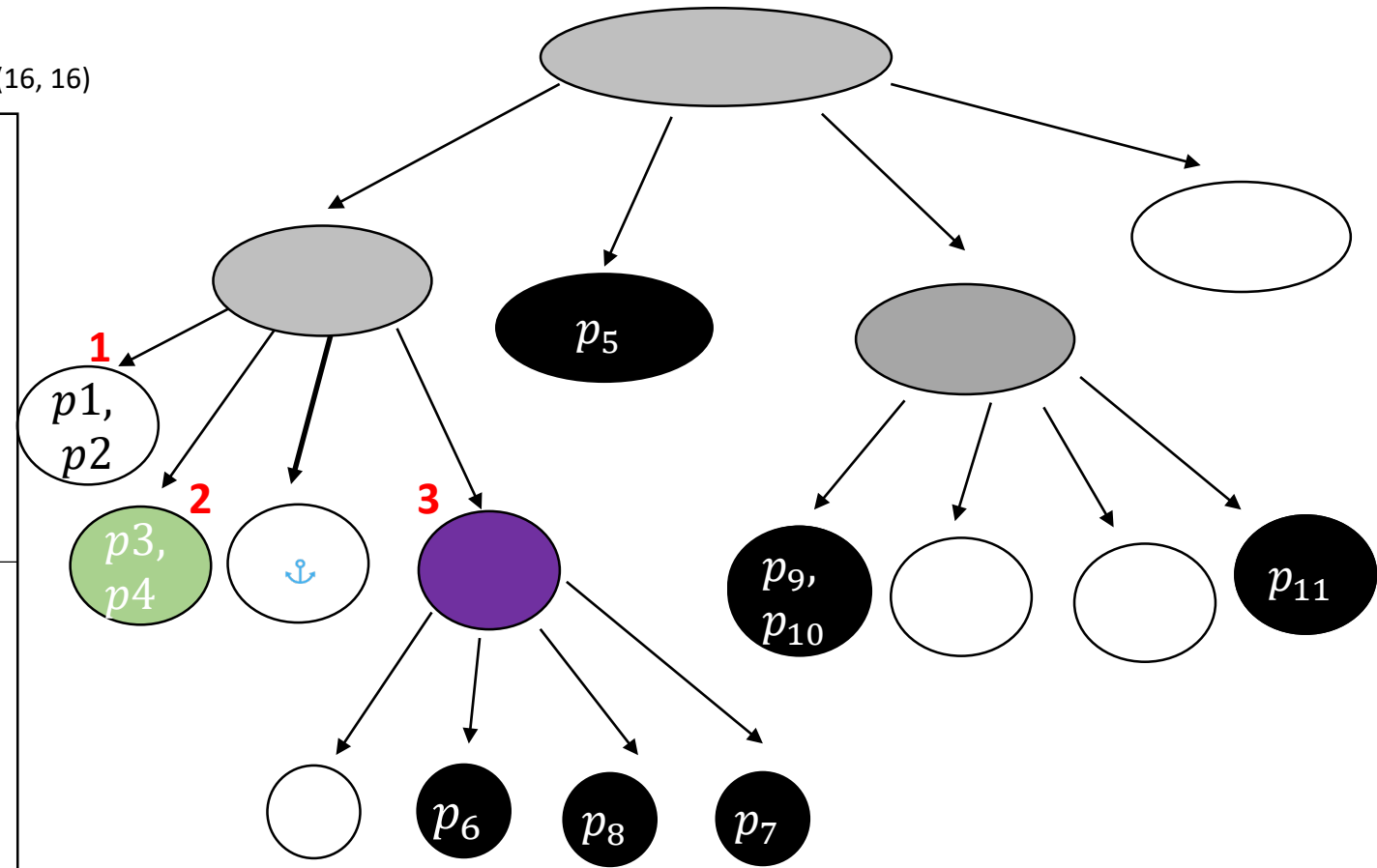
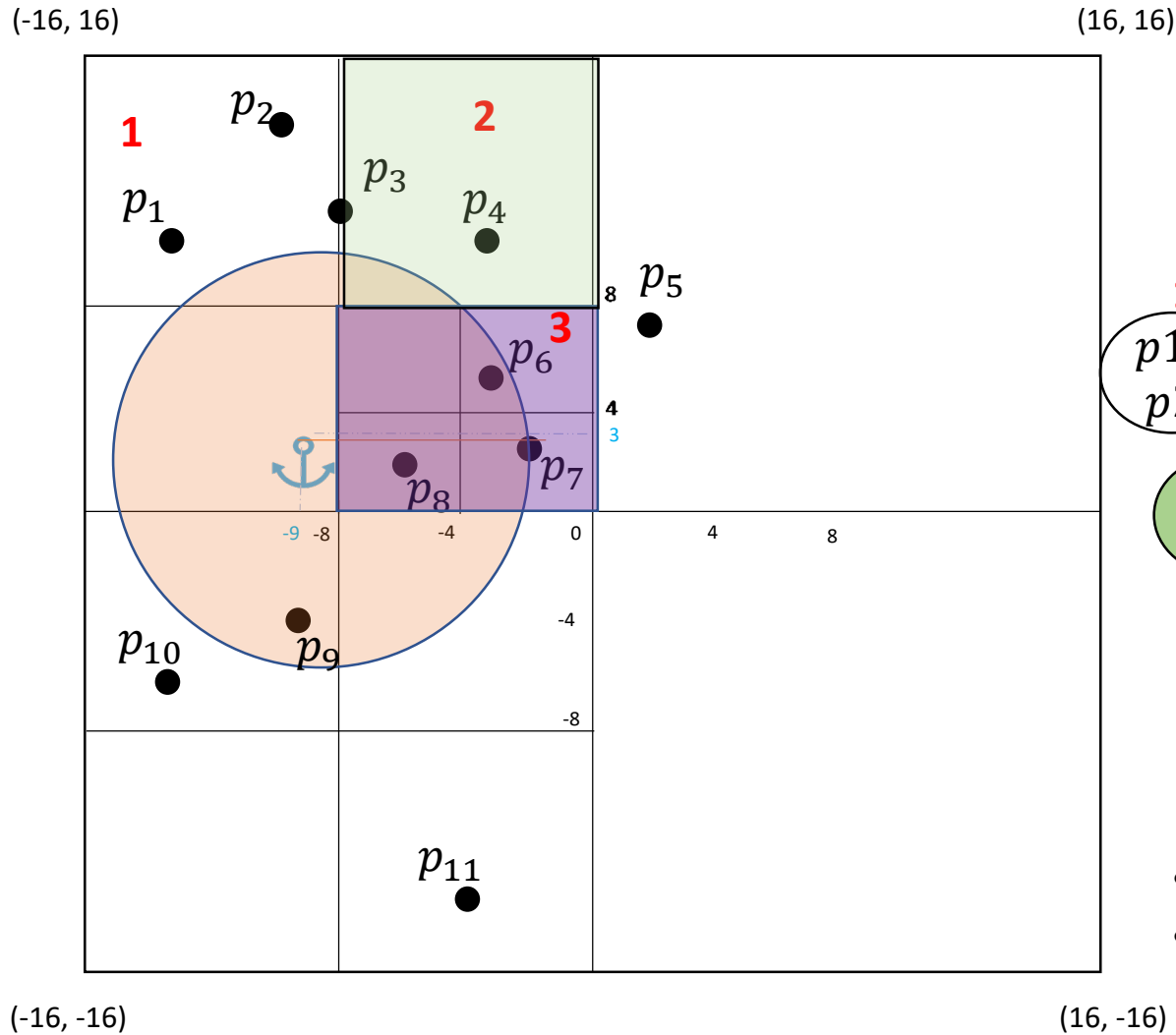
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



- Non-zero intersection, and black node!
- $p_1, p_2$  are not within the range, unfortunately!
- **Backtrack!**

# Example of range query in PR-QuadTree

anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$

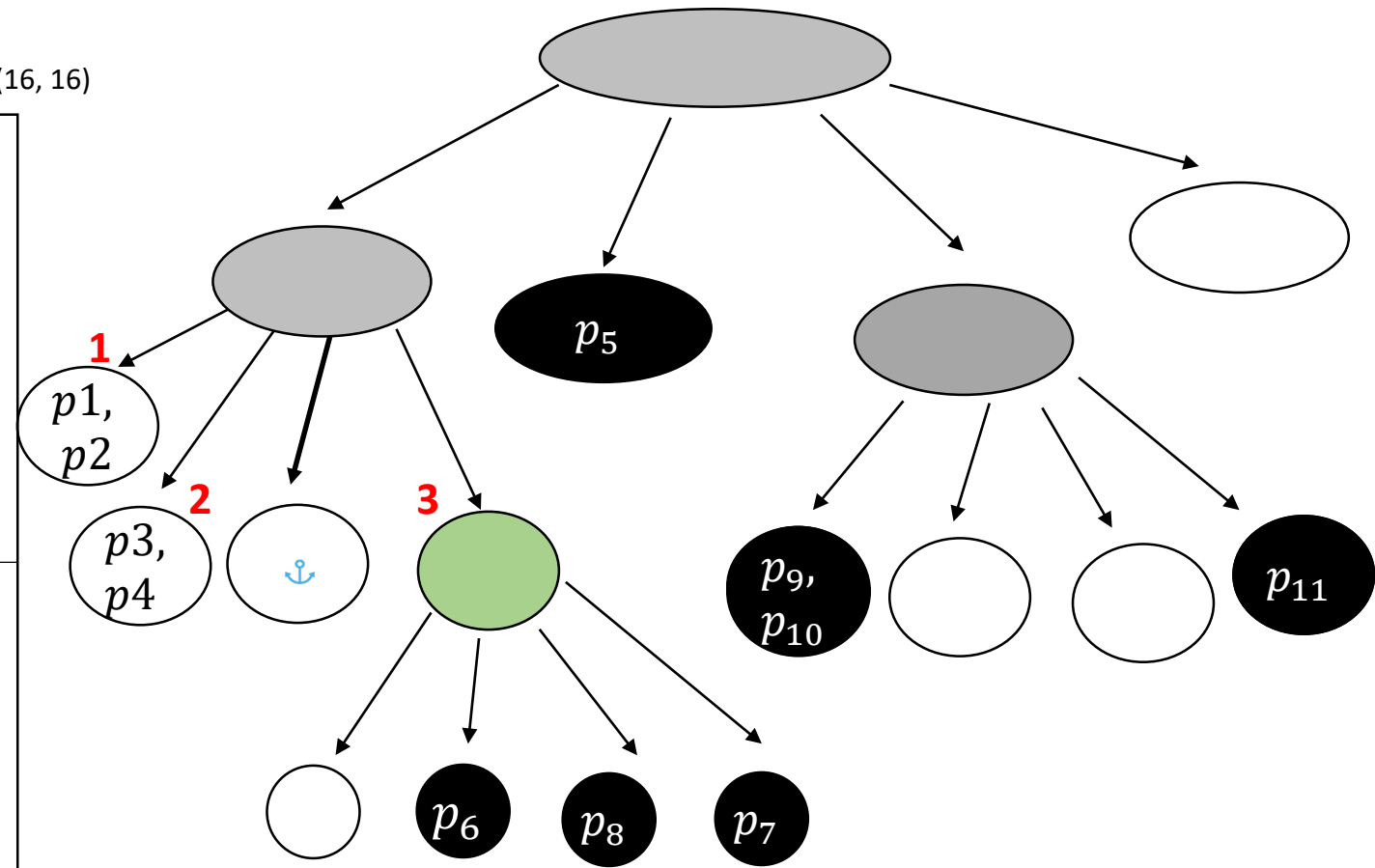
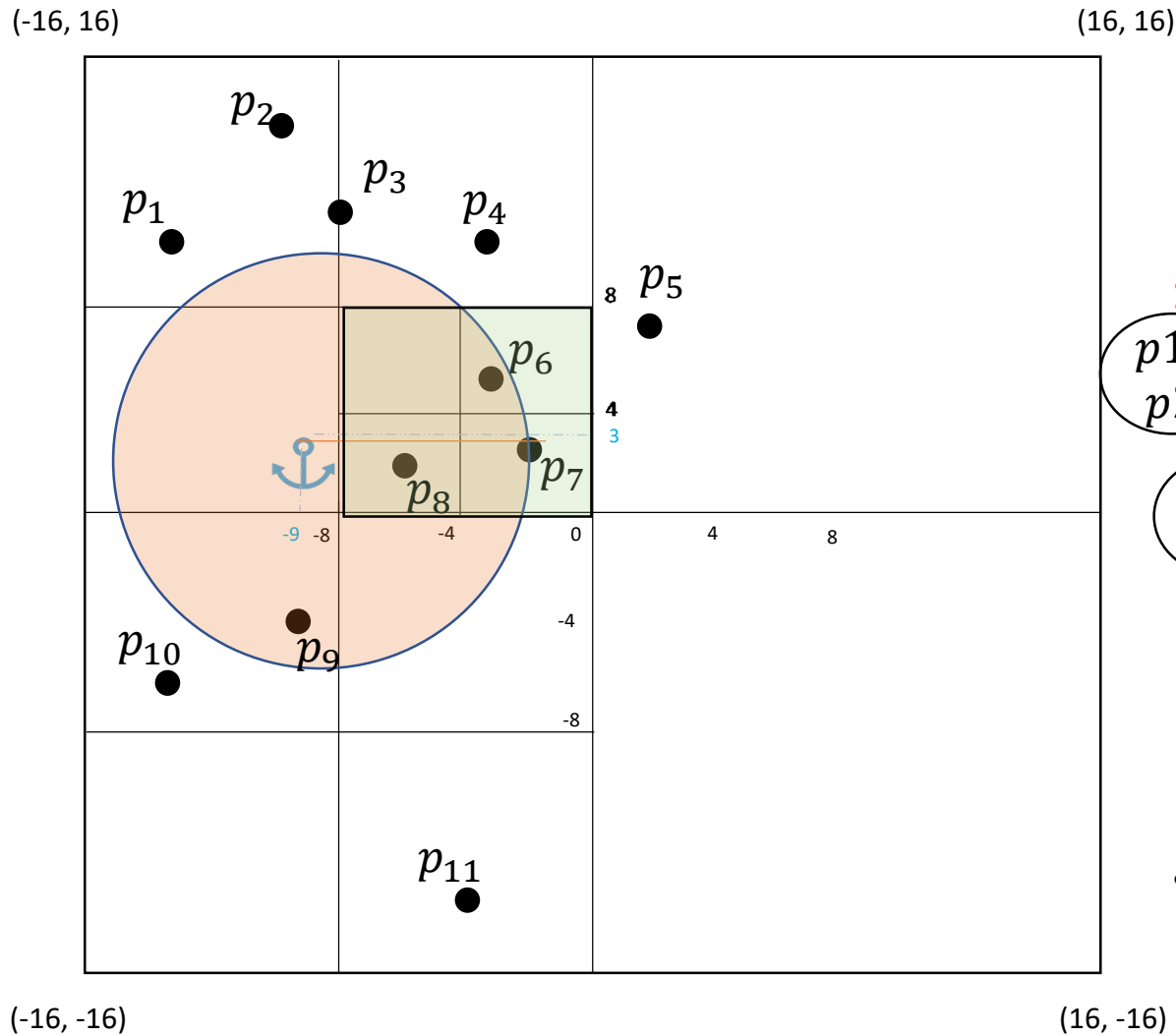


- Black node, non-zero intersection
- $p_3$  and  $p_4$  are too far away ☹️



# Example of range query in PR-QuadTree

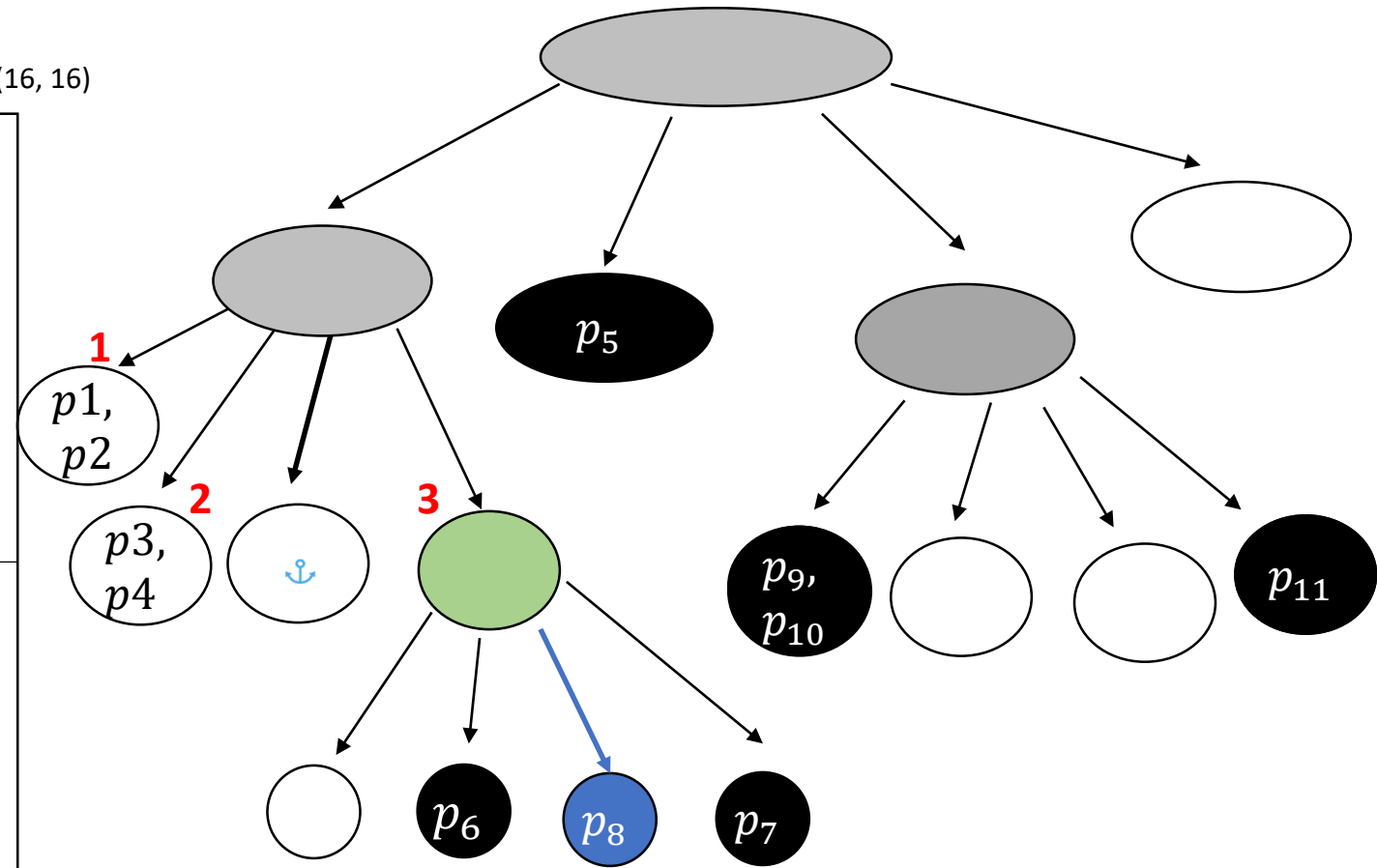
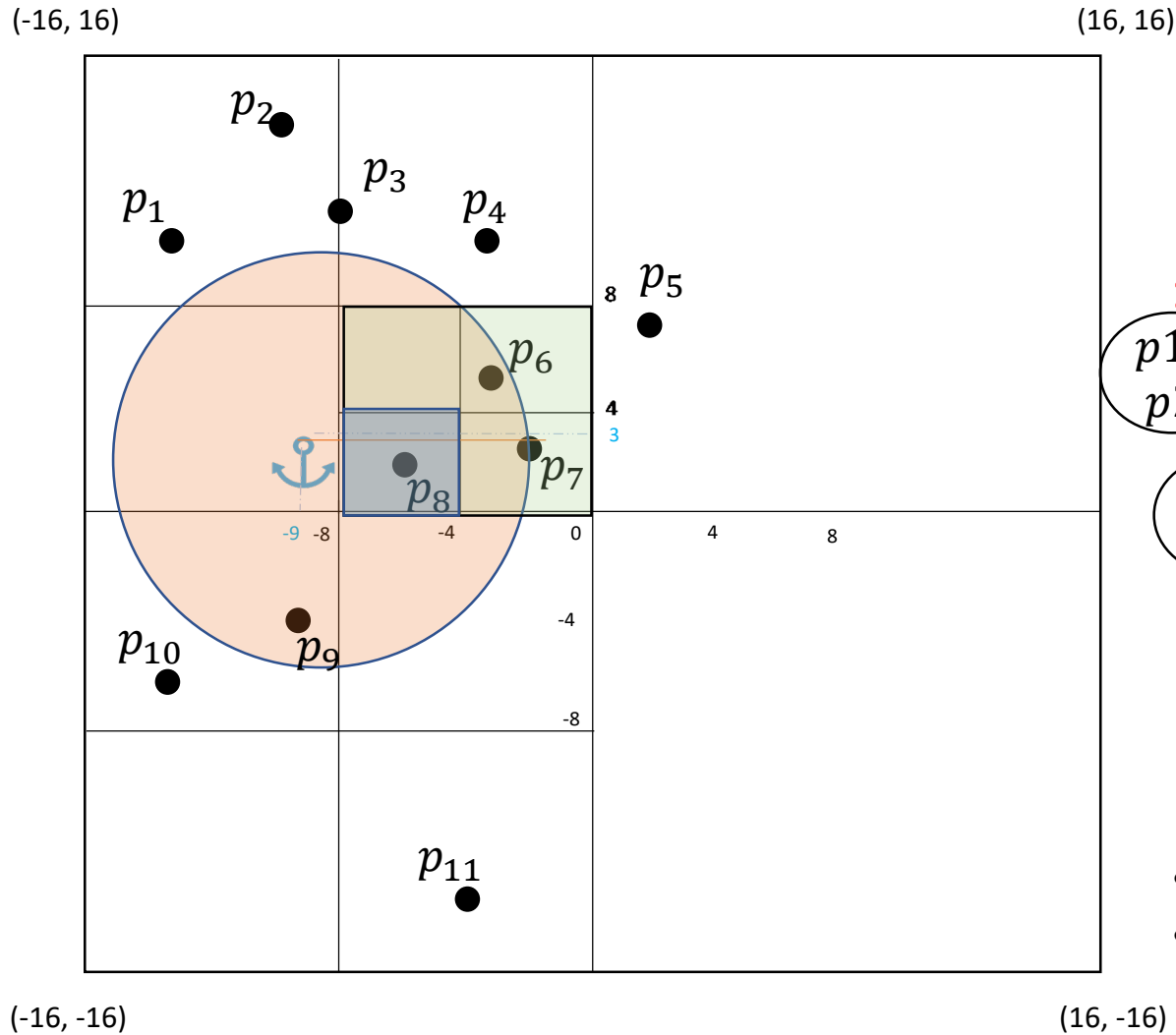
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



- Non-zero intersection, but **gray** node!

# Example of range query in PR-QuadTree

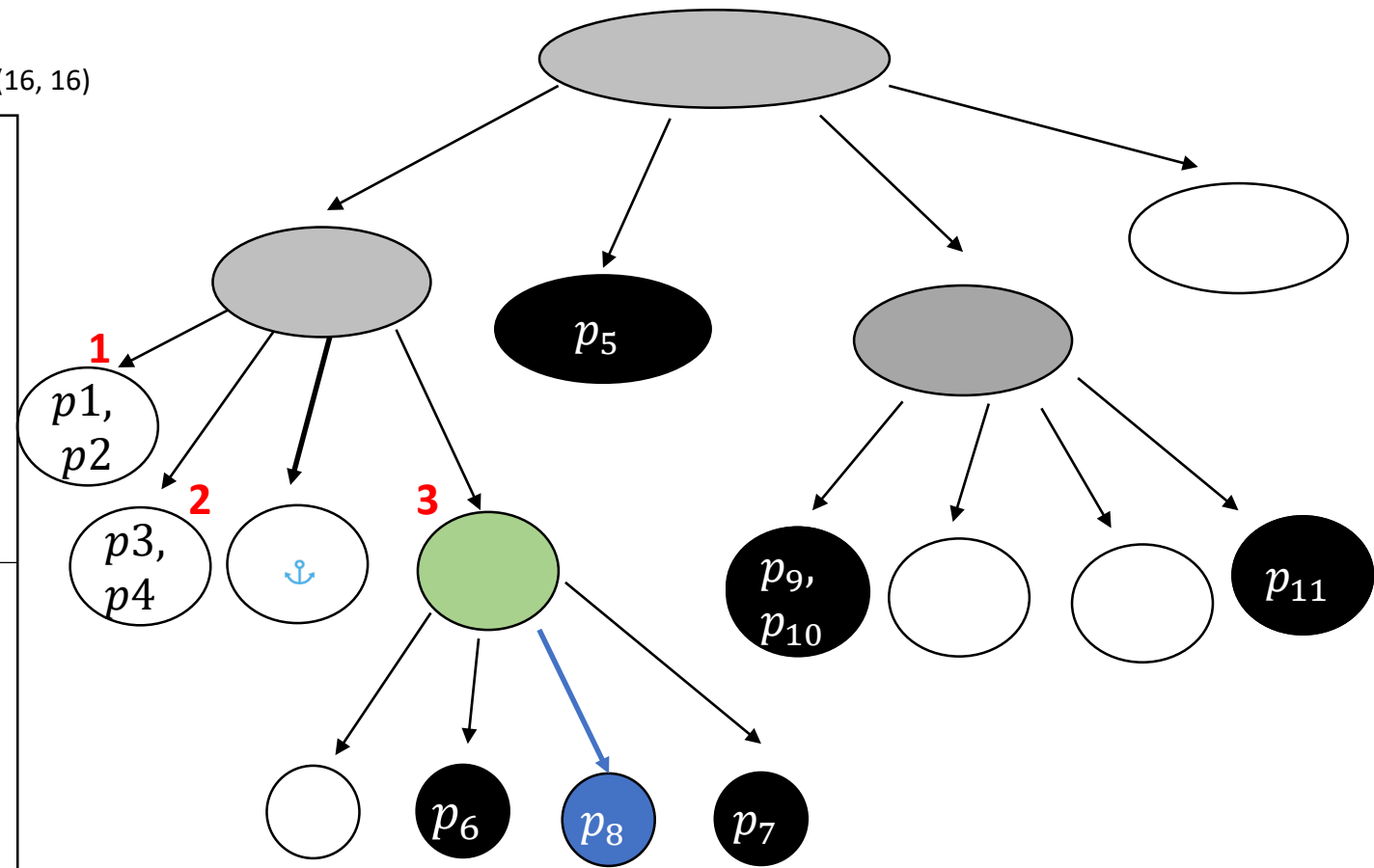
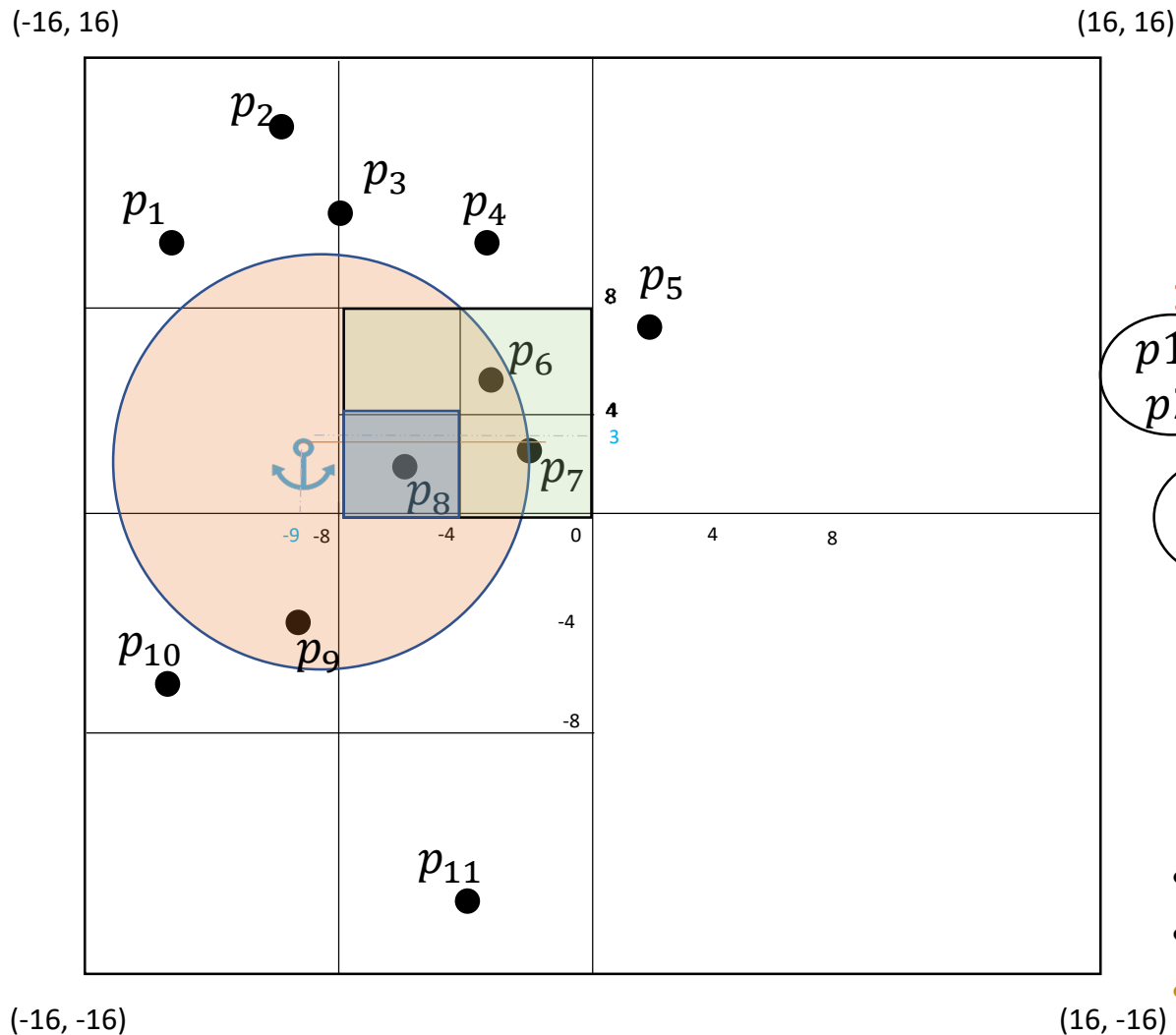
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



- Non-zero intersection, but **gray** node!
- Examine child **closest to anchor** first!

# Example of range query in PR-QuadTree

anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$

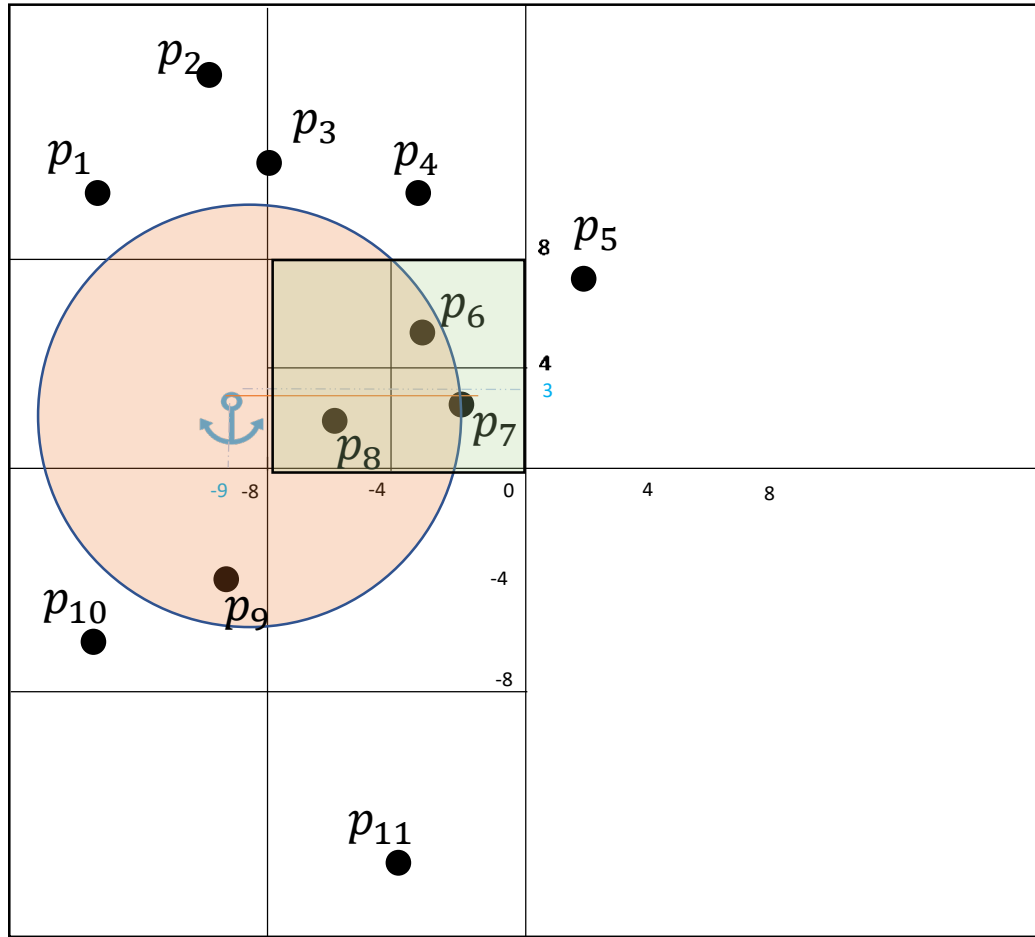


- Non-zero intersection, but **gray** node!
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- $p_8$  reported 😊

# Example of range query in PR-QuadTree

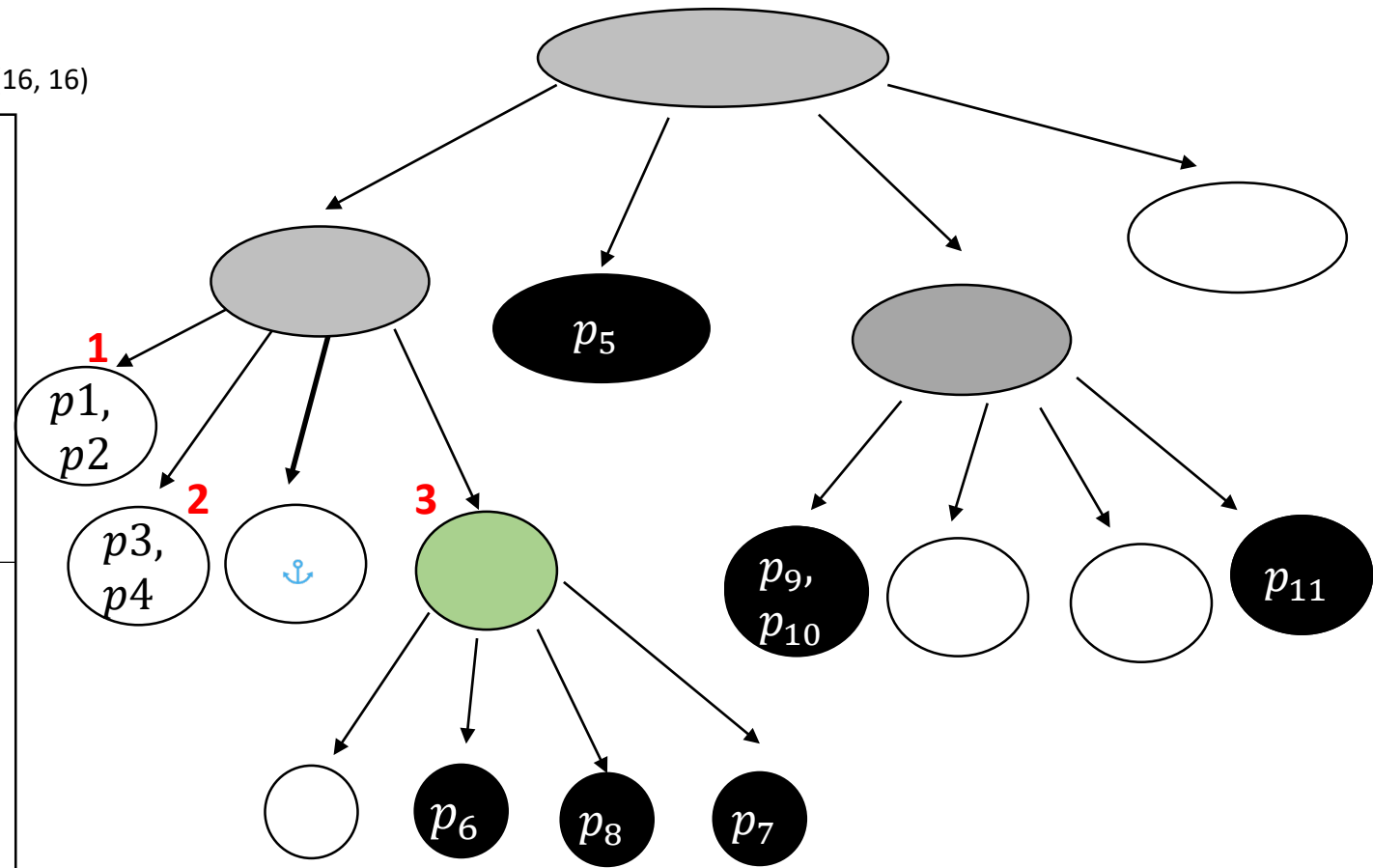
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$

$(-16, 16)$   $(16, 16)$



$(-16, -16)$

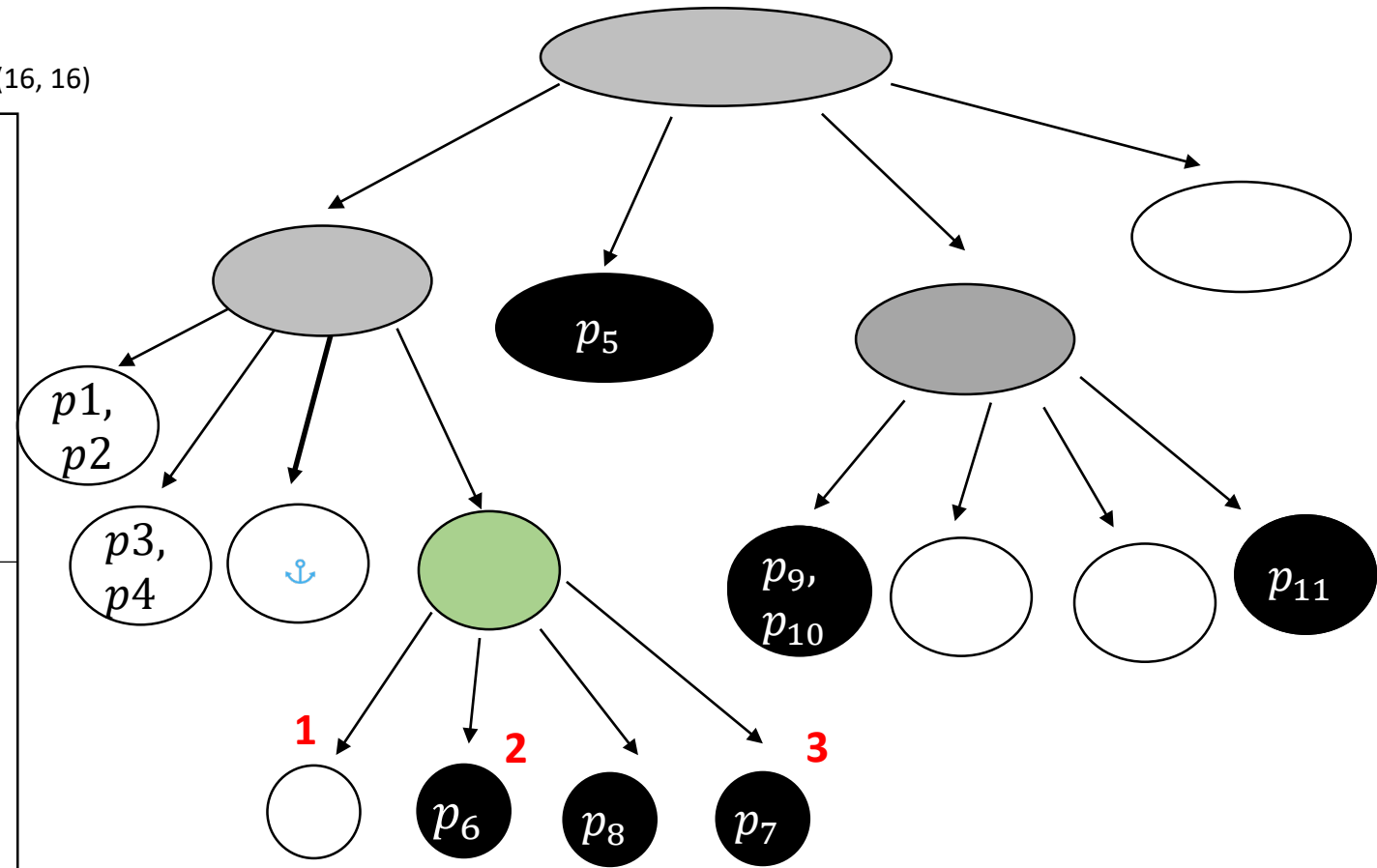
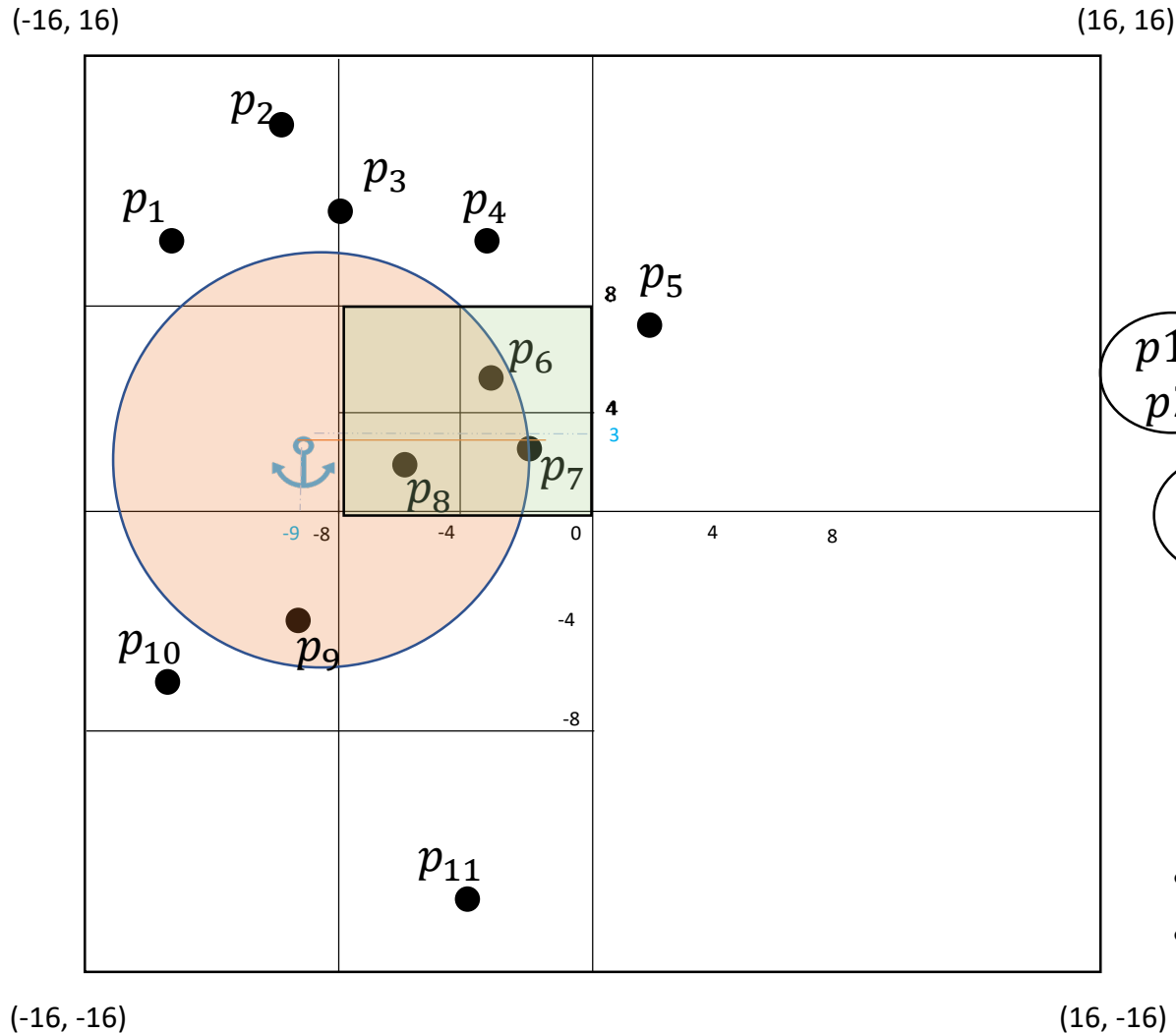
$(16, -16)$



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# Example of range query in PR-QuadTree

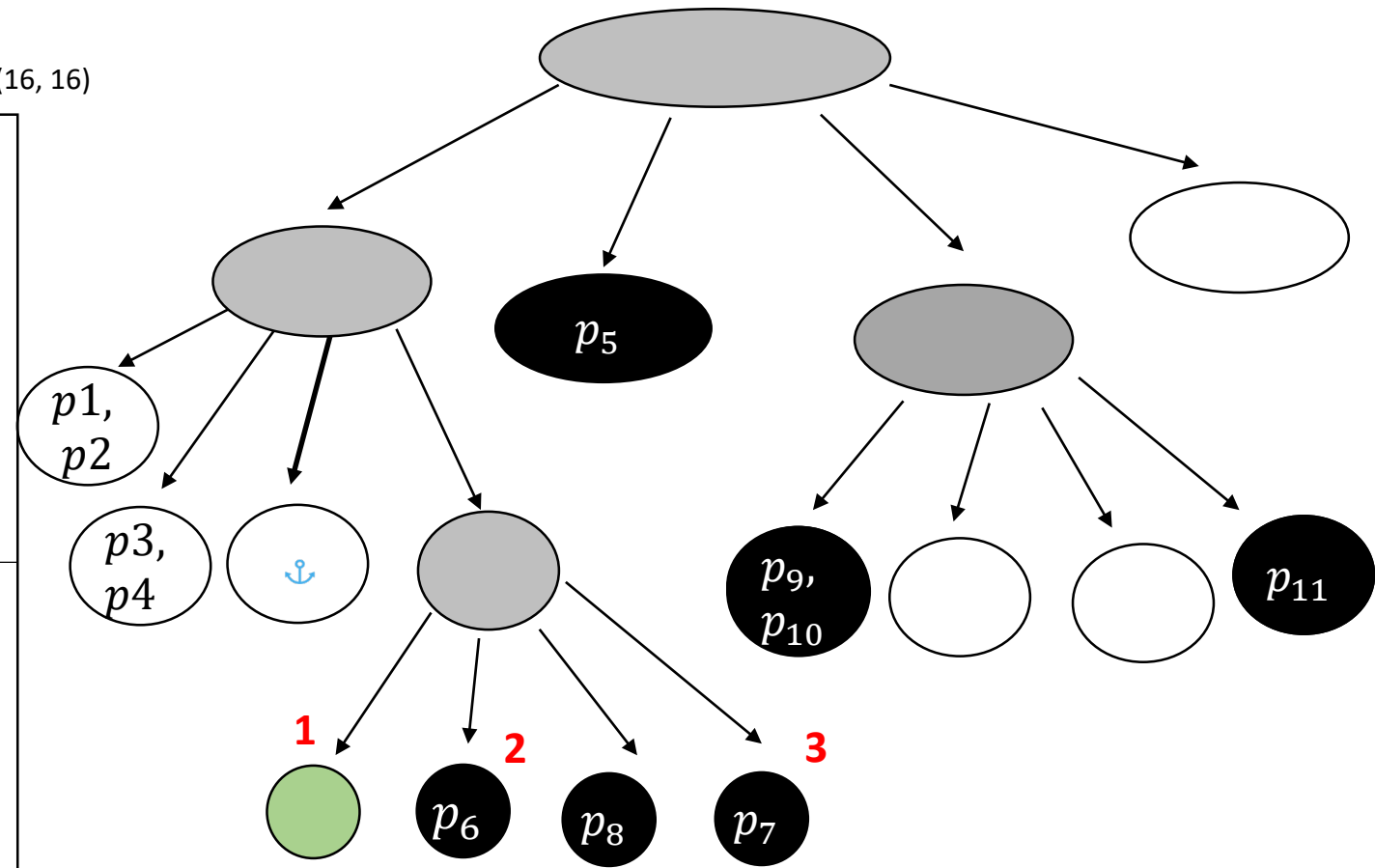
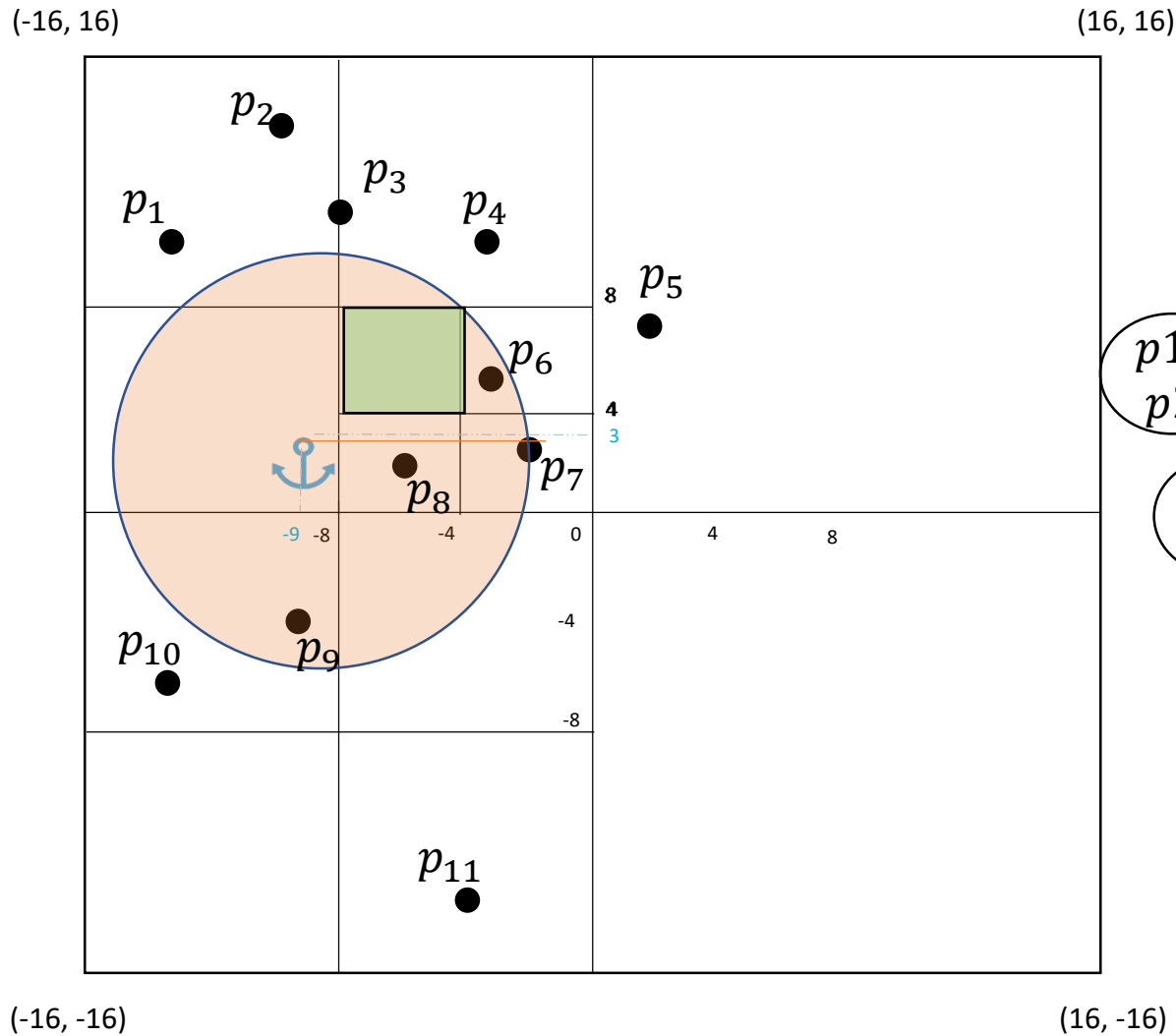
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- Non-zero intersection, but **gray** node!
- Again, do not prioritize children based on area of intersection, but **on Z-order**!

# Example of range query in PR-QuadTree

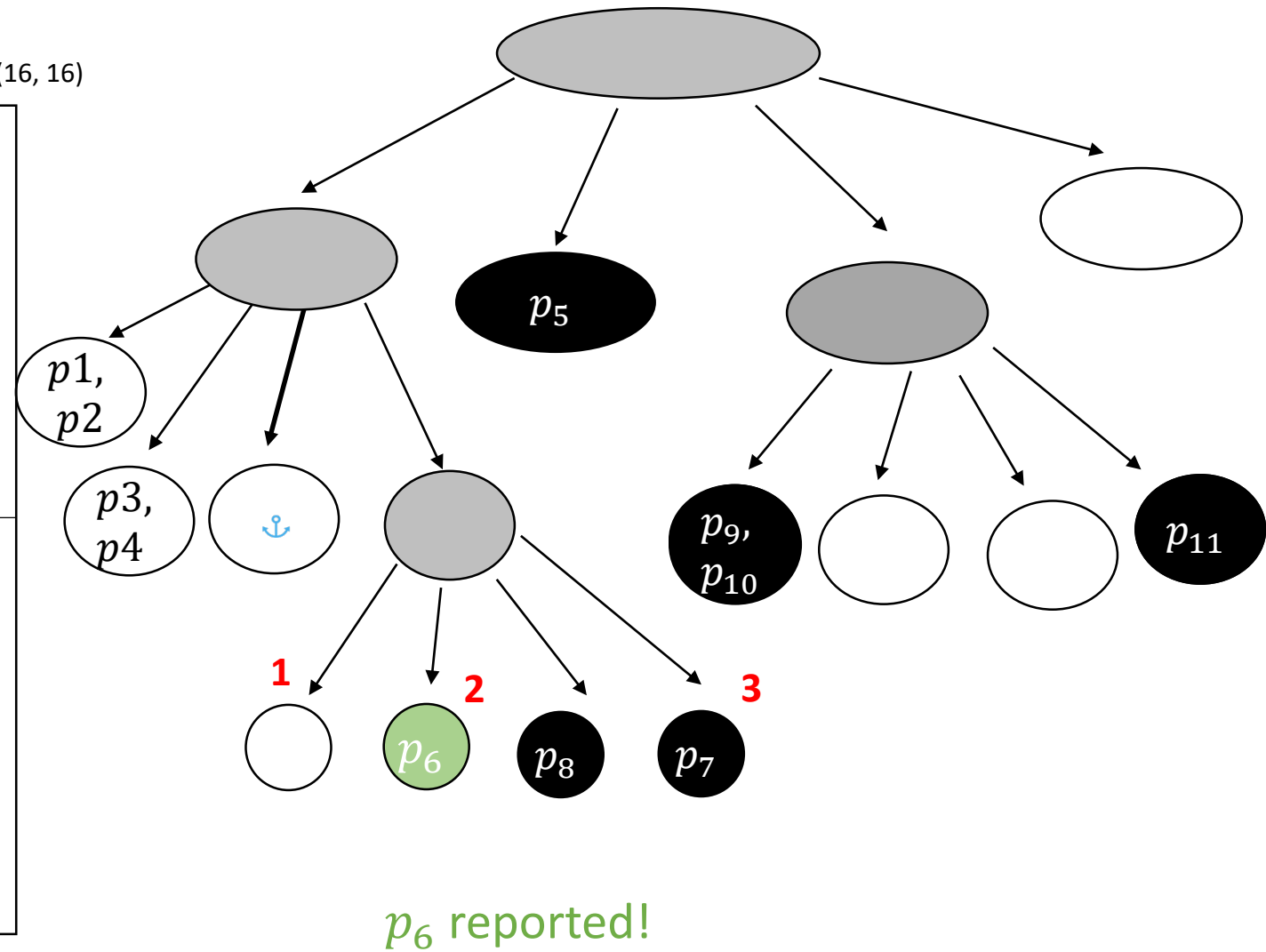
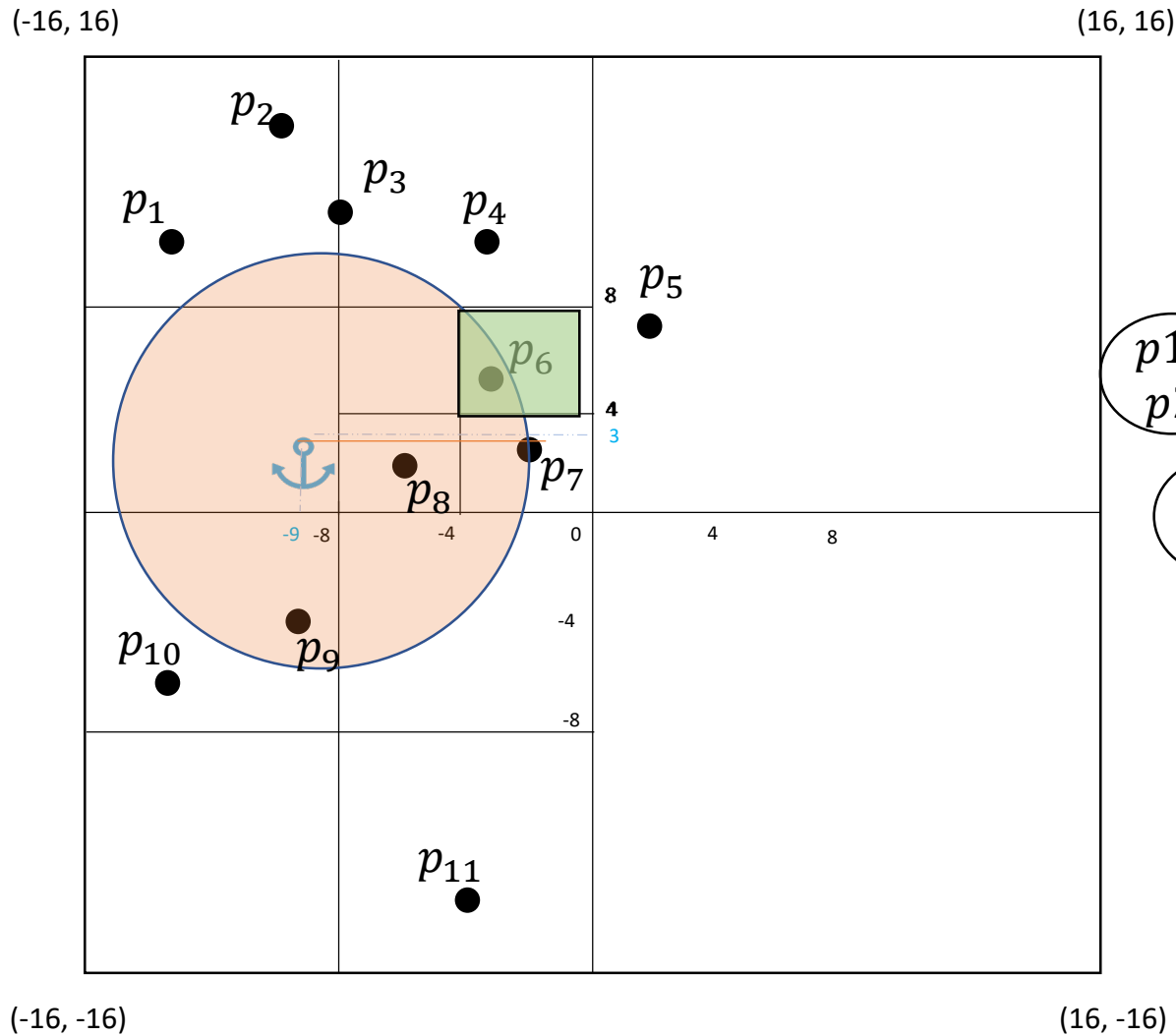
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



White node, nothing to report...

# Example of range query in PR-QuadTree

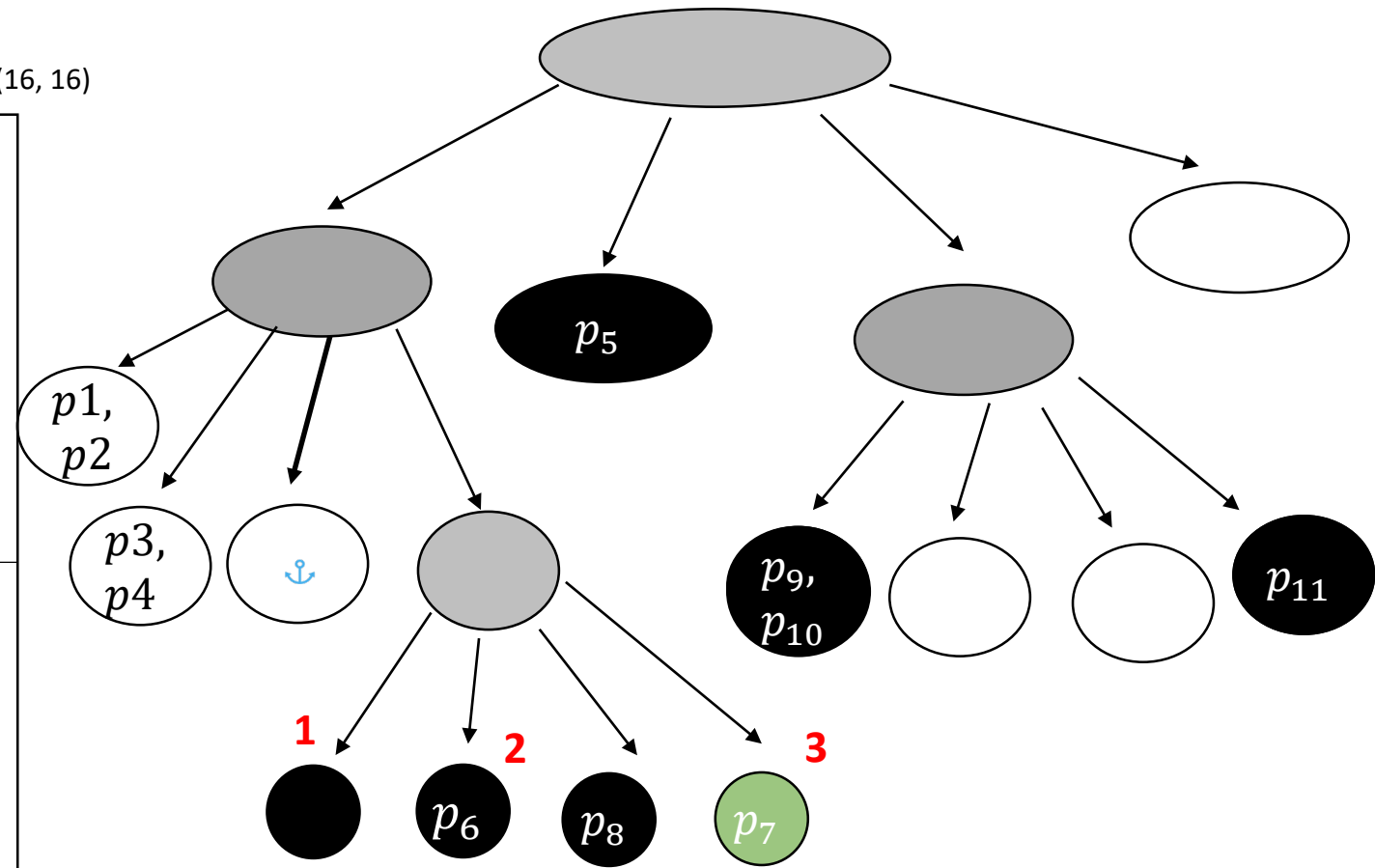
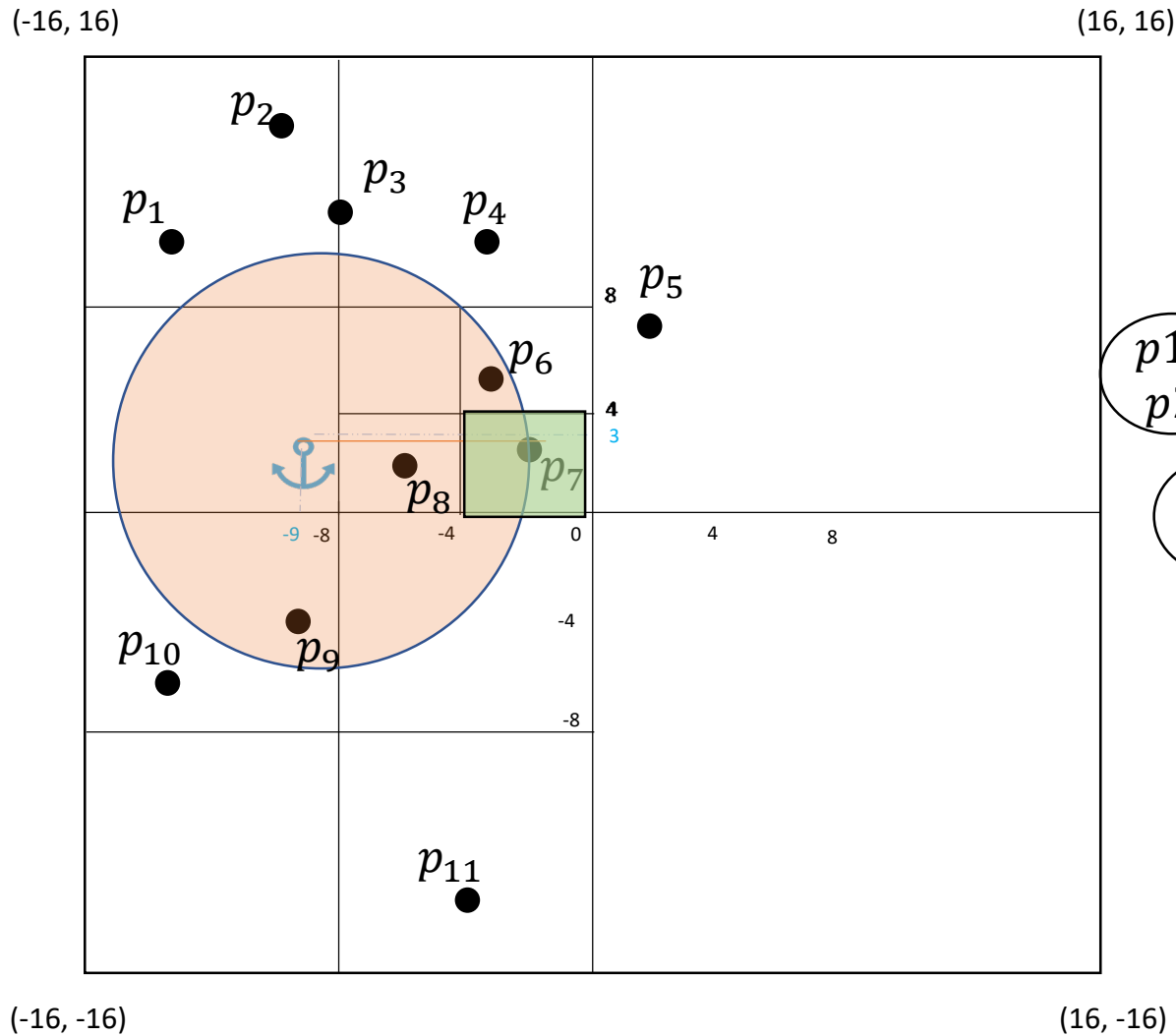
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



$p_6$  reported!

# Example of range query in PR-QuadTree

anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$

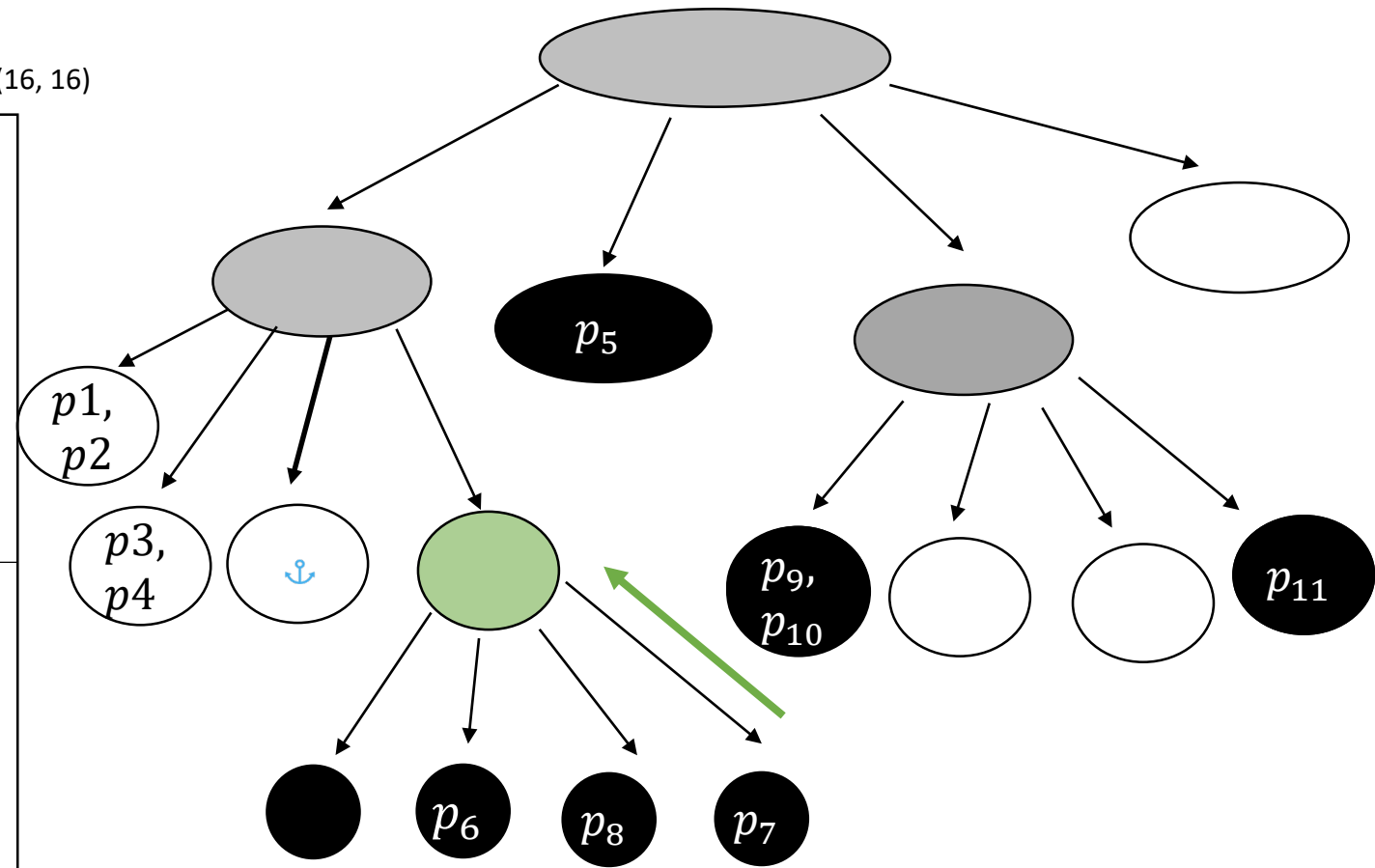
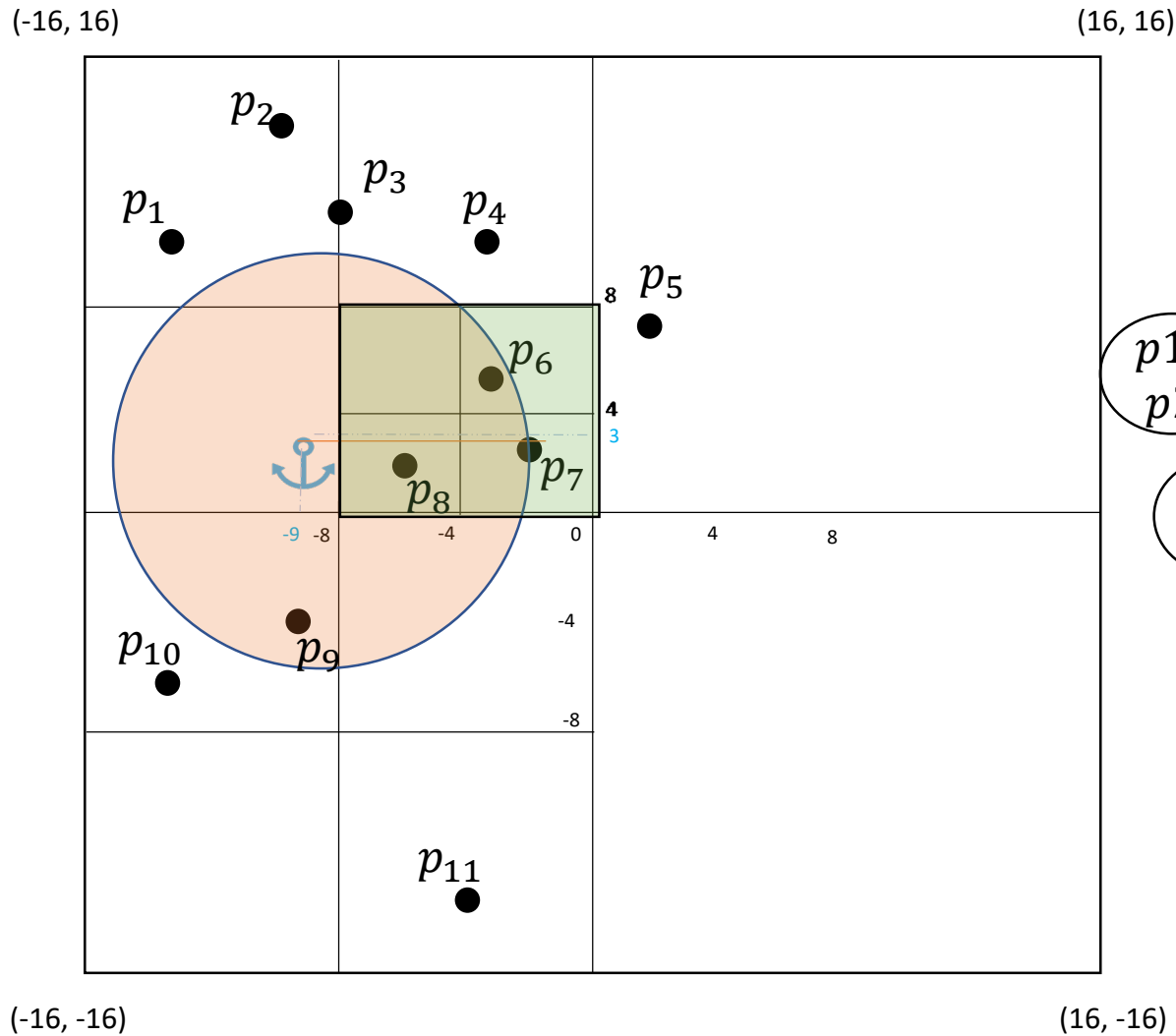


$p_7$  reported! (remember:  
range queries inclusive)



# Example of range query in PR-QuadTree

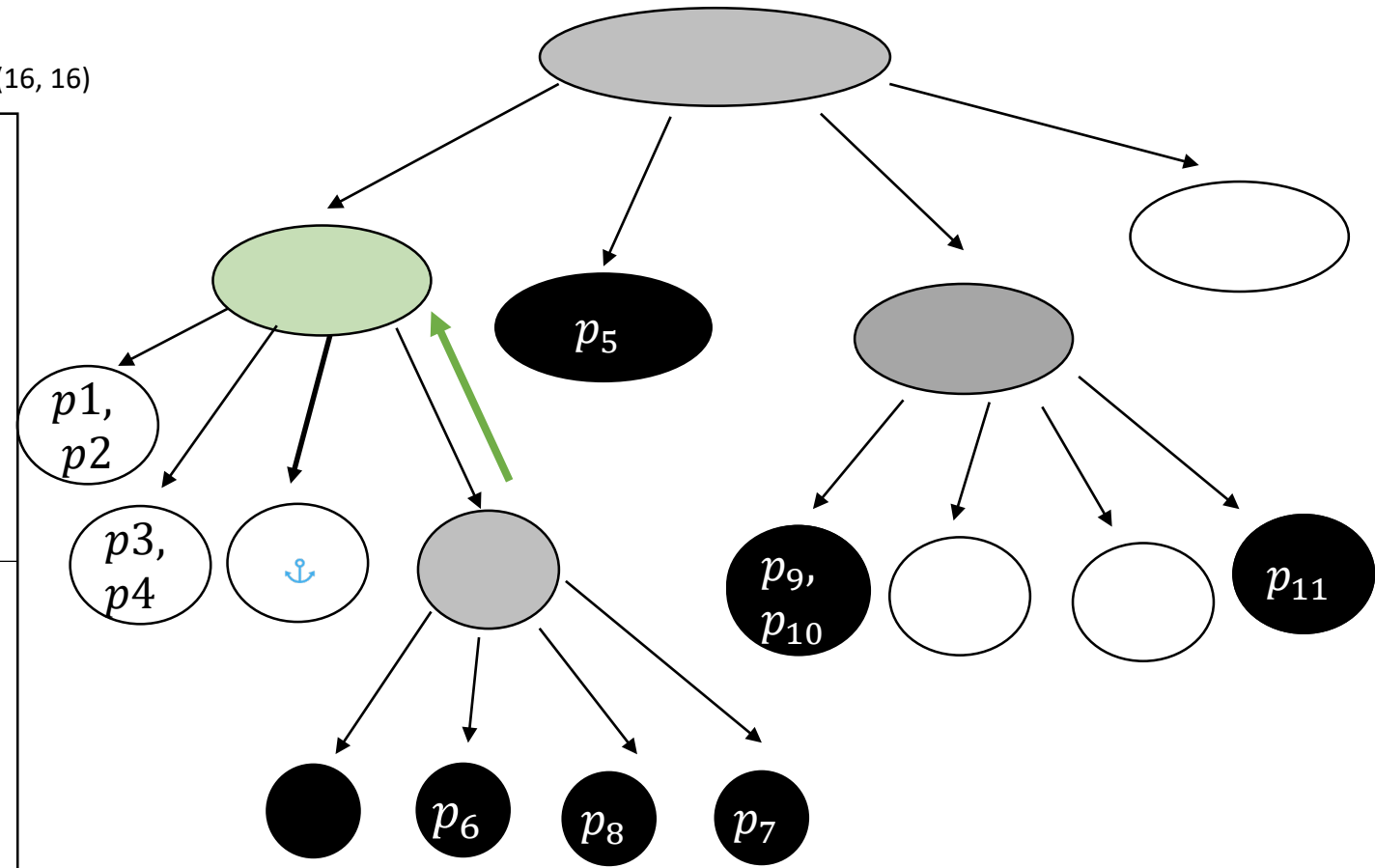
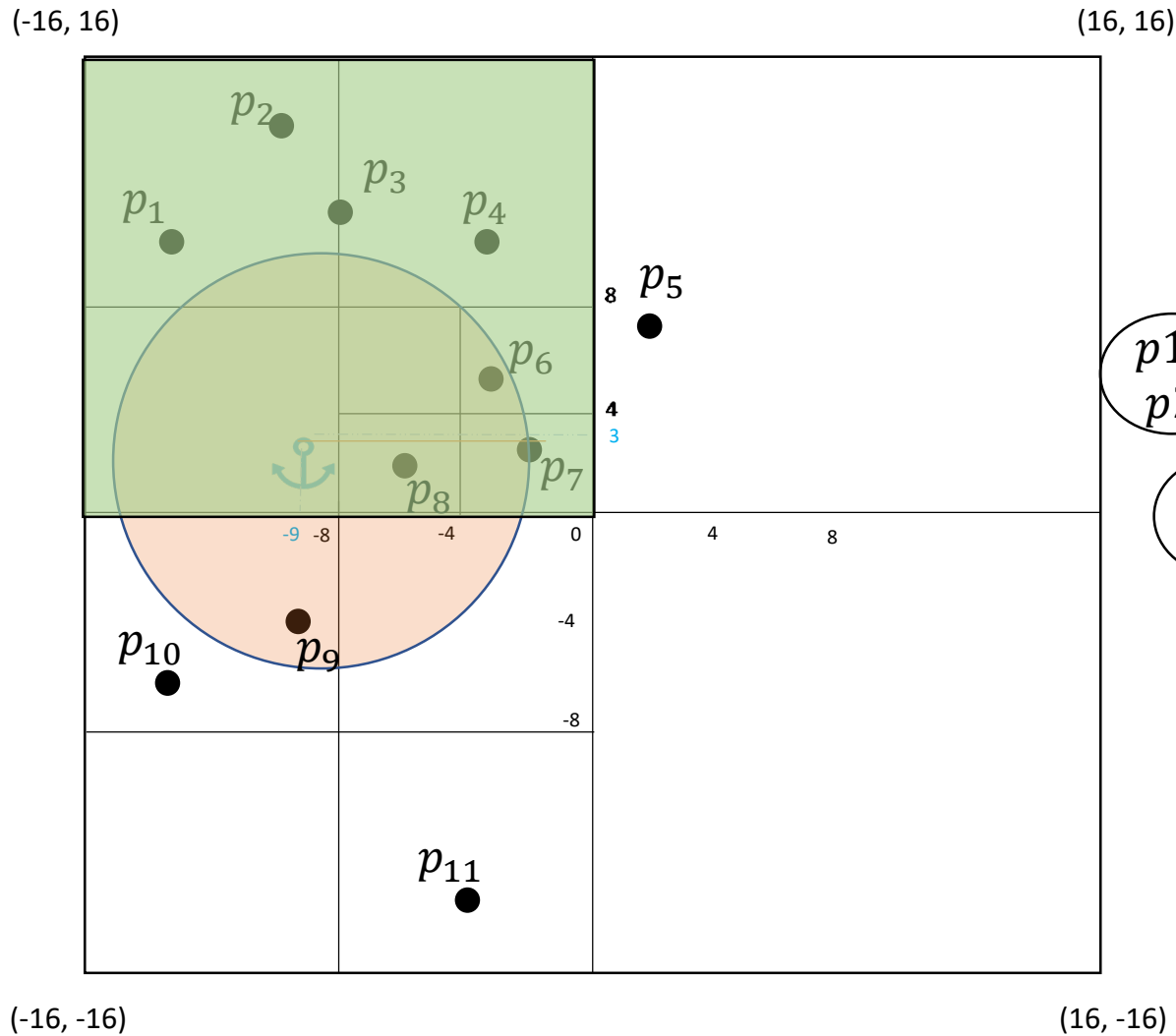
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



Backtracking all the way to the root...

# Example of range query in PR-QuadTree

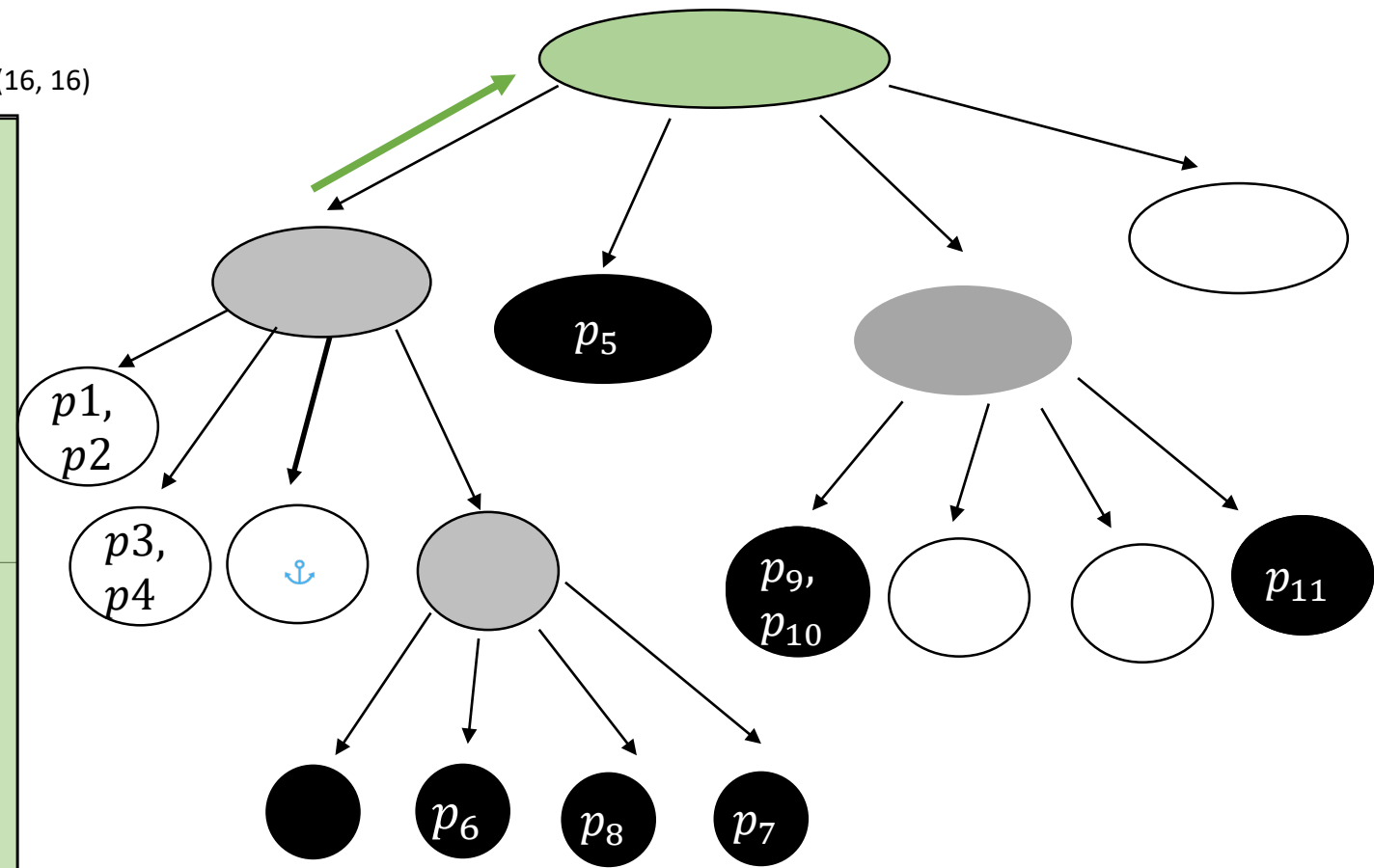
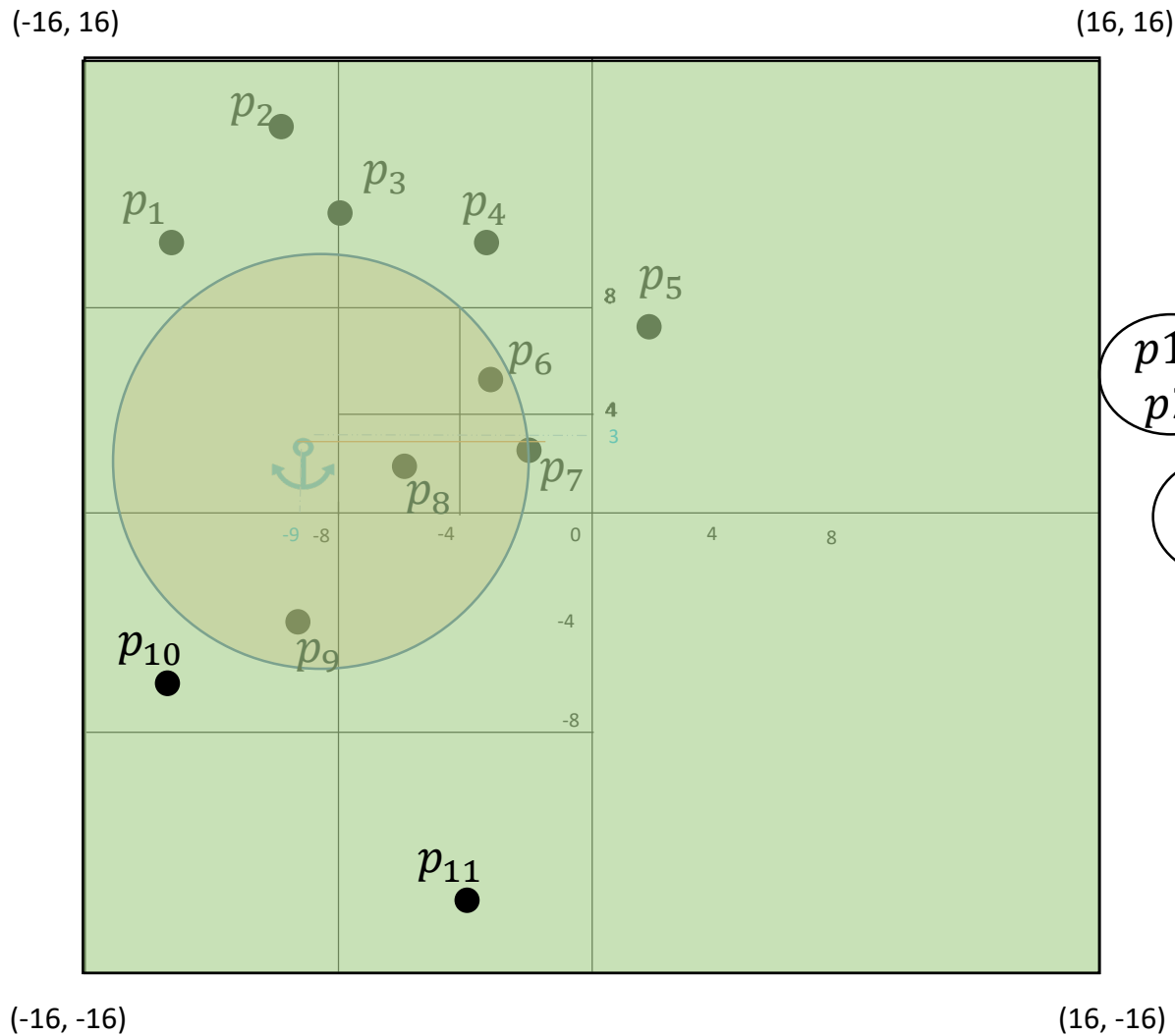
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



Backtracking all the way to the root...

# Example of range query in PR-QuadTree

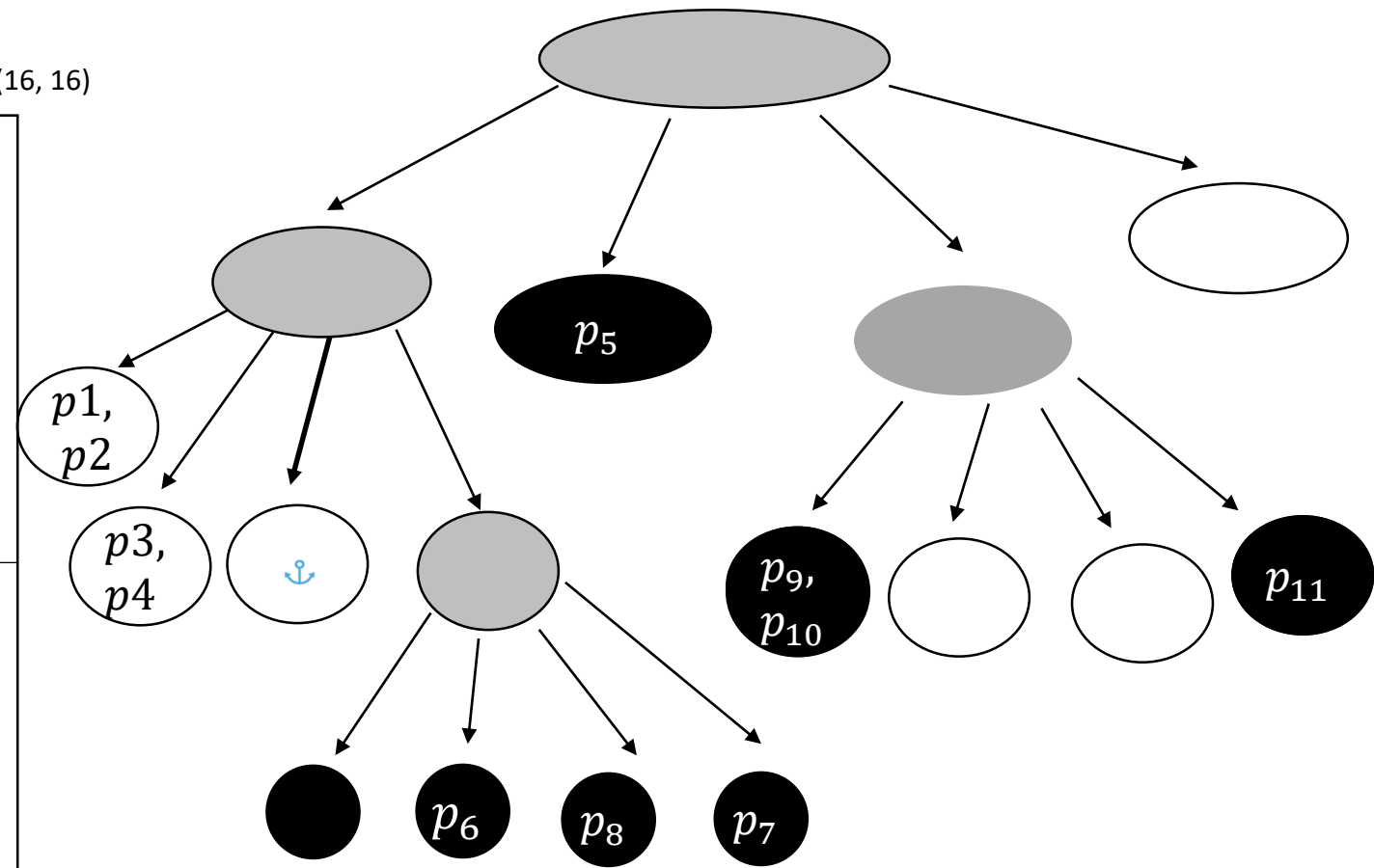
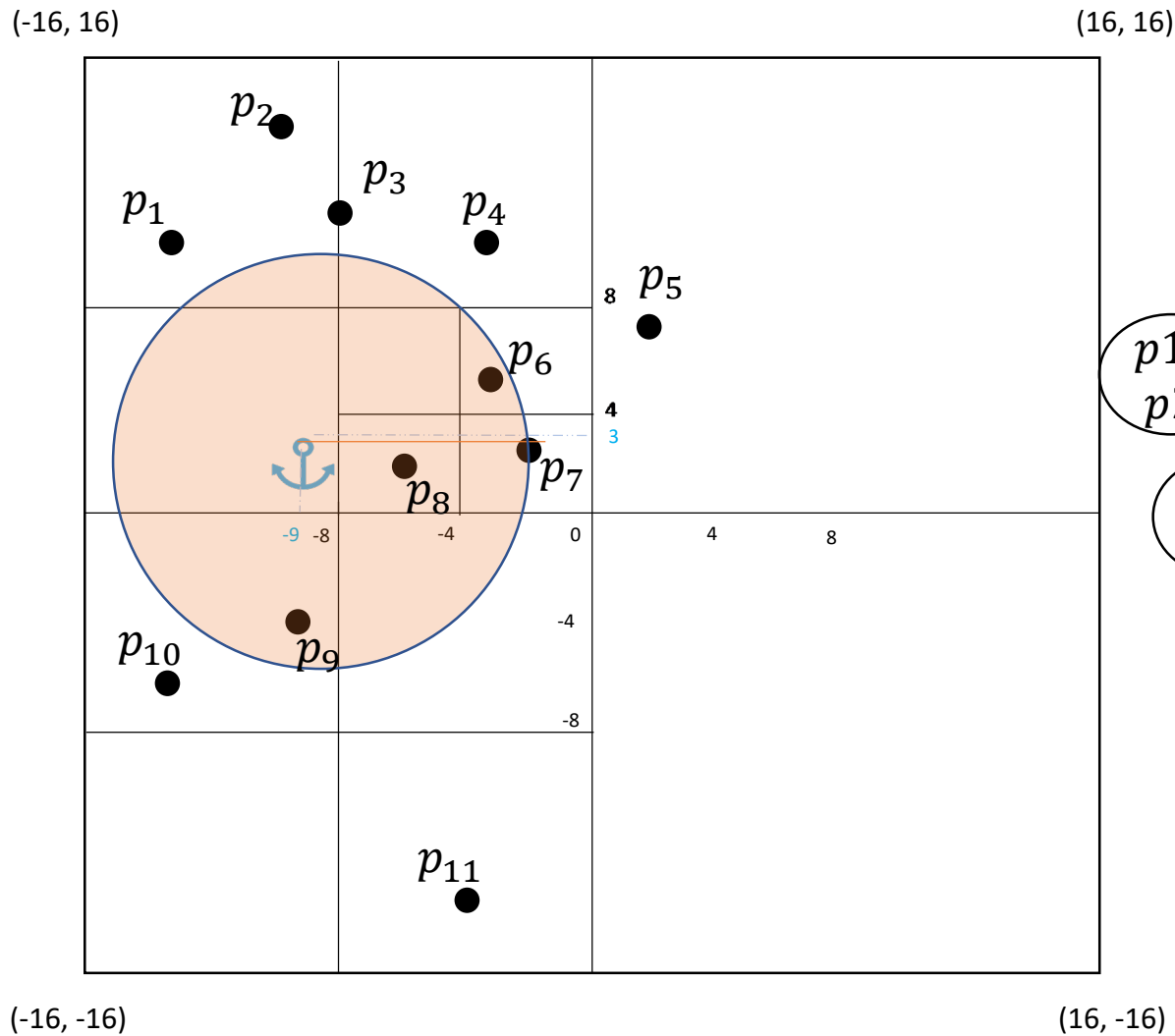
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



Backtracking all the way to the root...

# Example of range query in PR-QuadTree

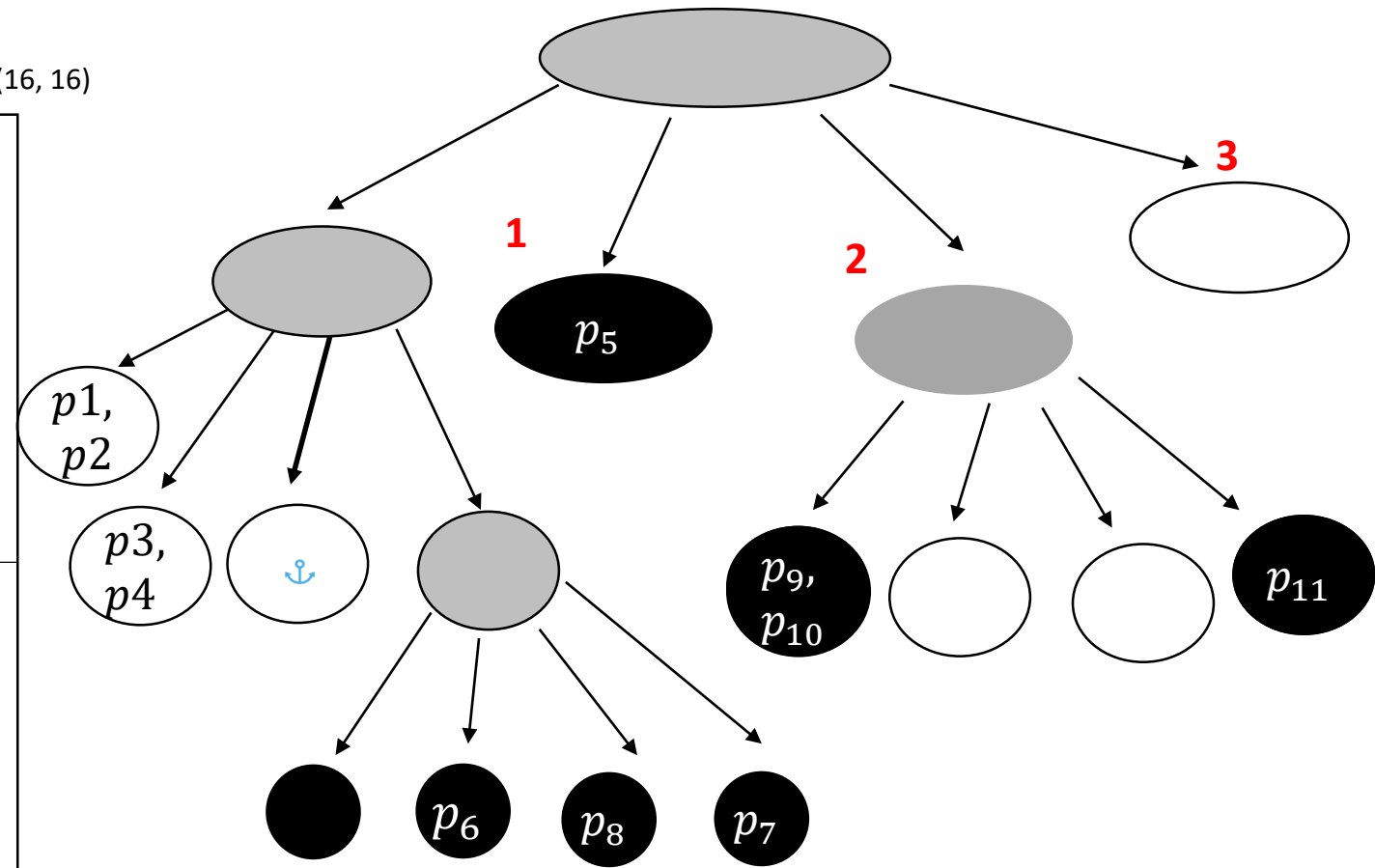
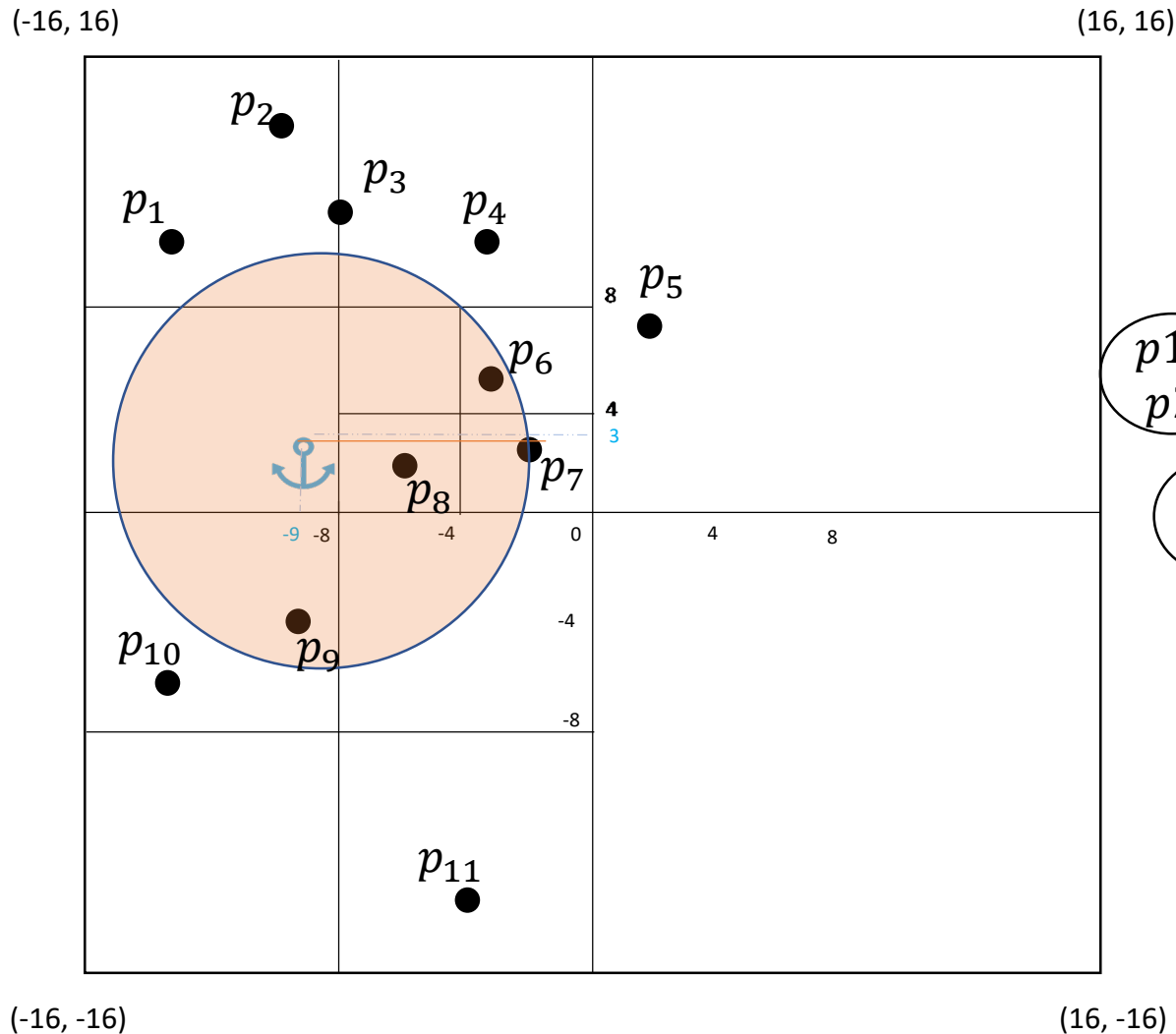
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



Once again, examine  
remaining children in **Z-Order!**

# Example of range query in PR-QuadTree

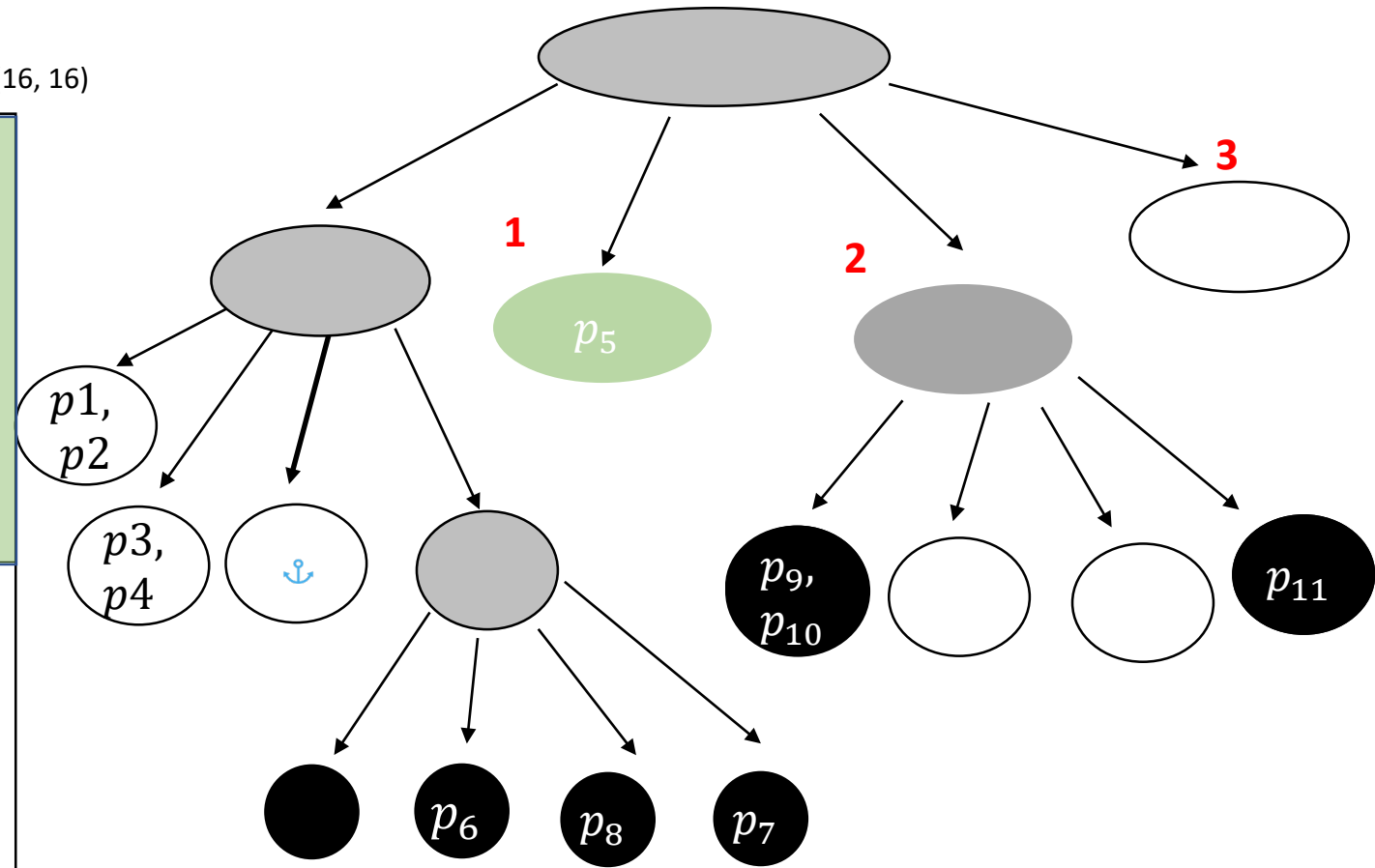
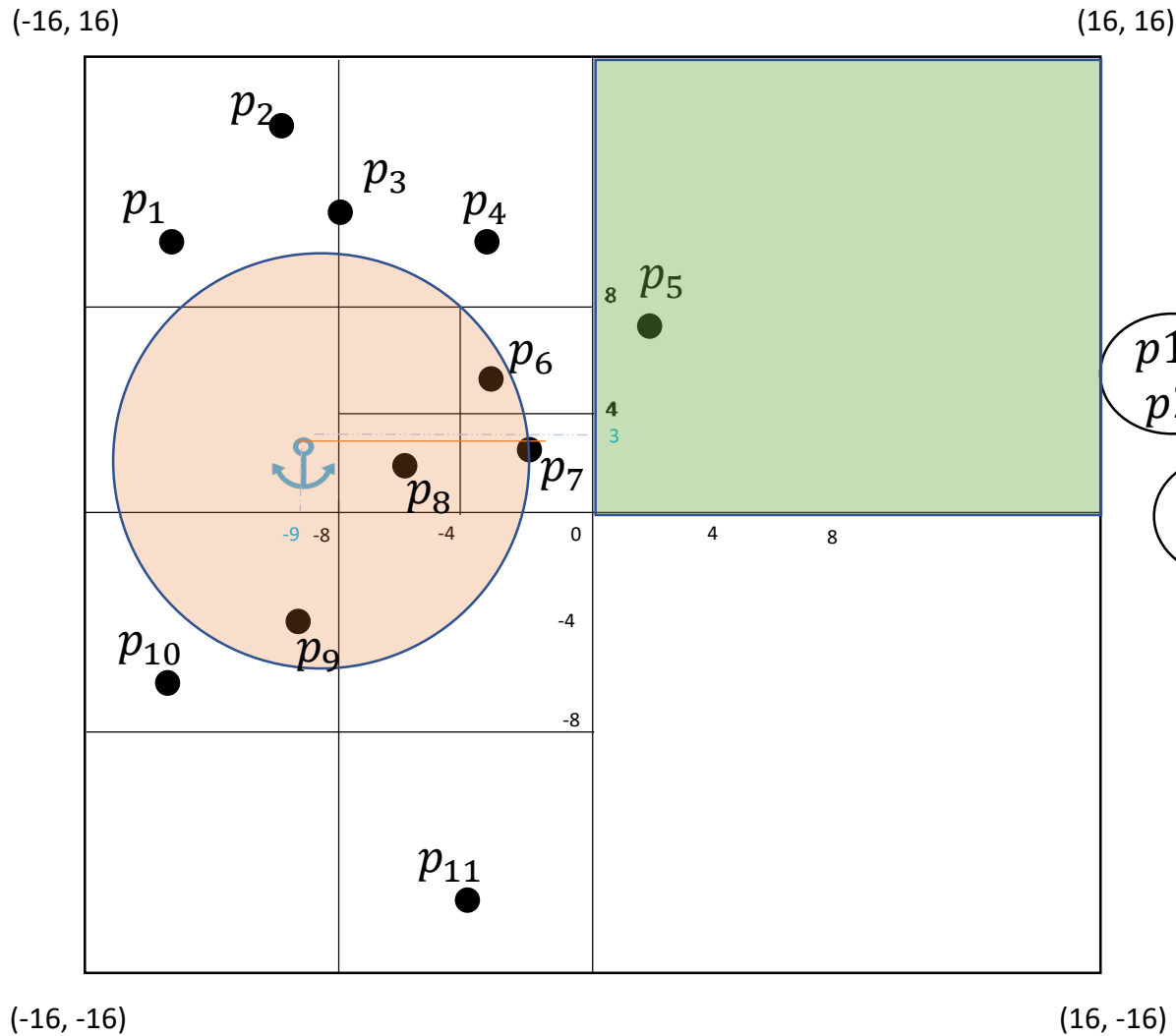
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



Once again, examine remaining children in **Z-Order**!

# Example of range query in PR-QuadTree

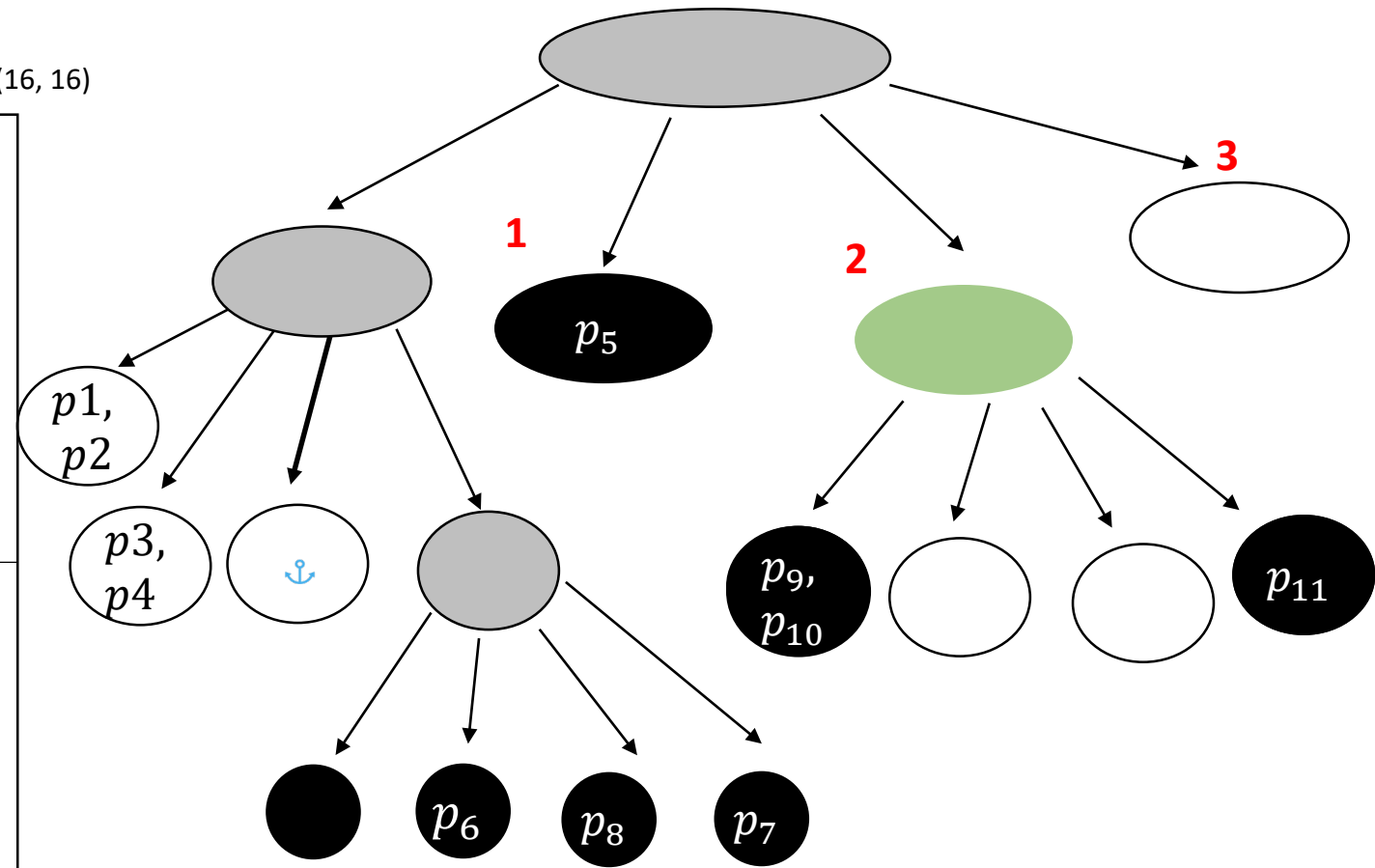
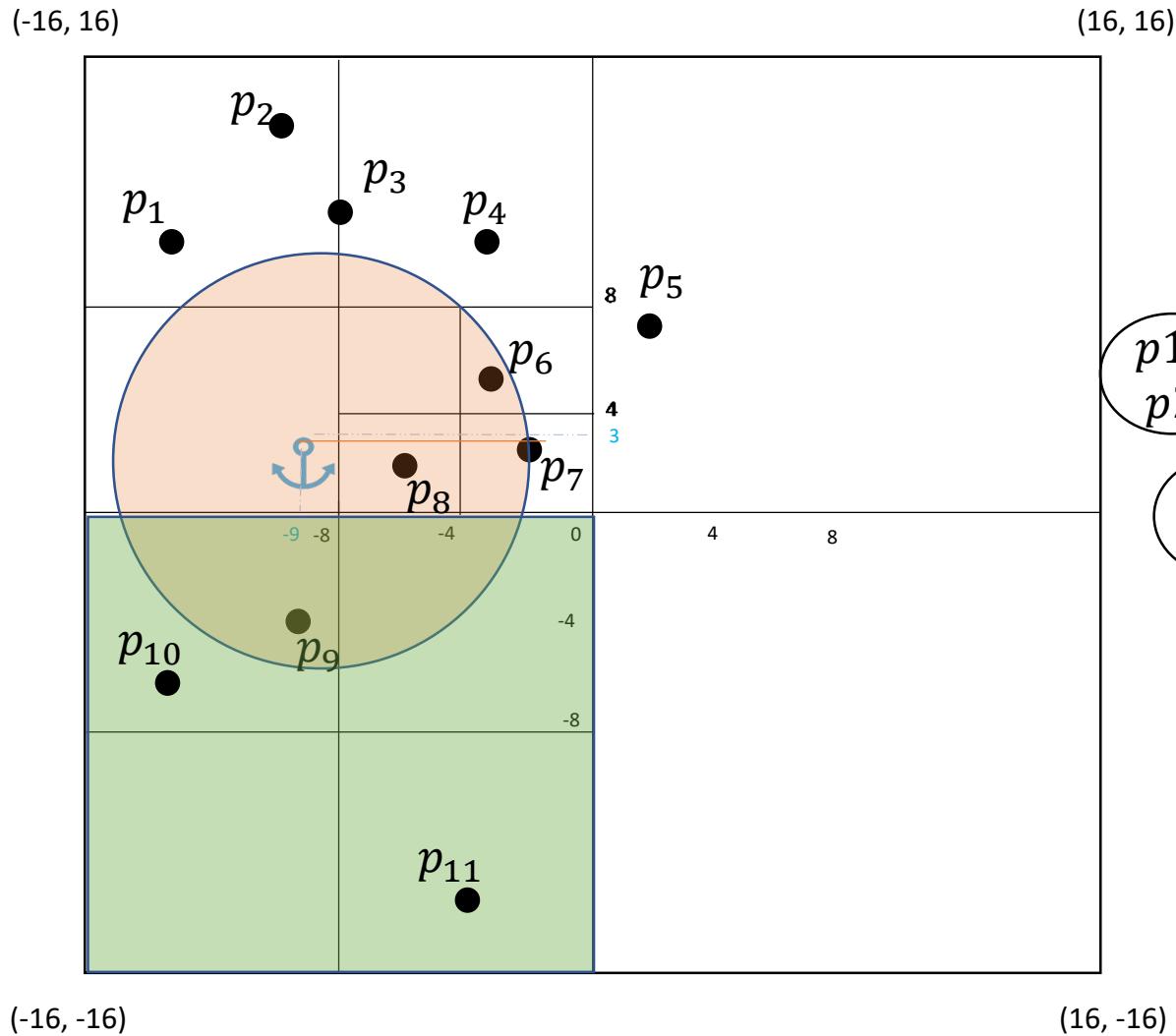
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



**Zero intersection** between currently spanned quadrant and range: Subtree **cannot** contribute to solution set. Immediately backtrack.

# Example of range query in PR-QuadTree

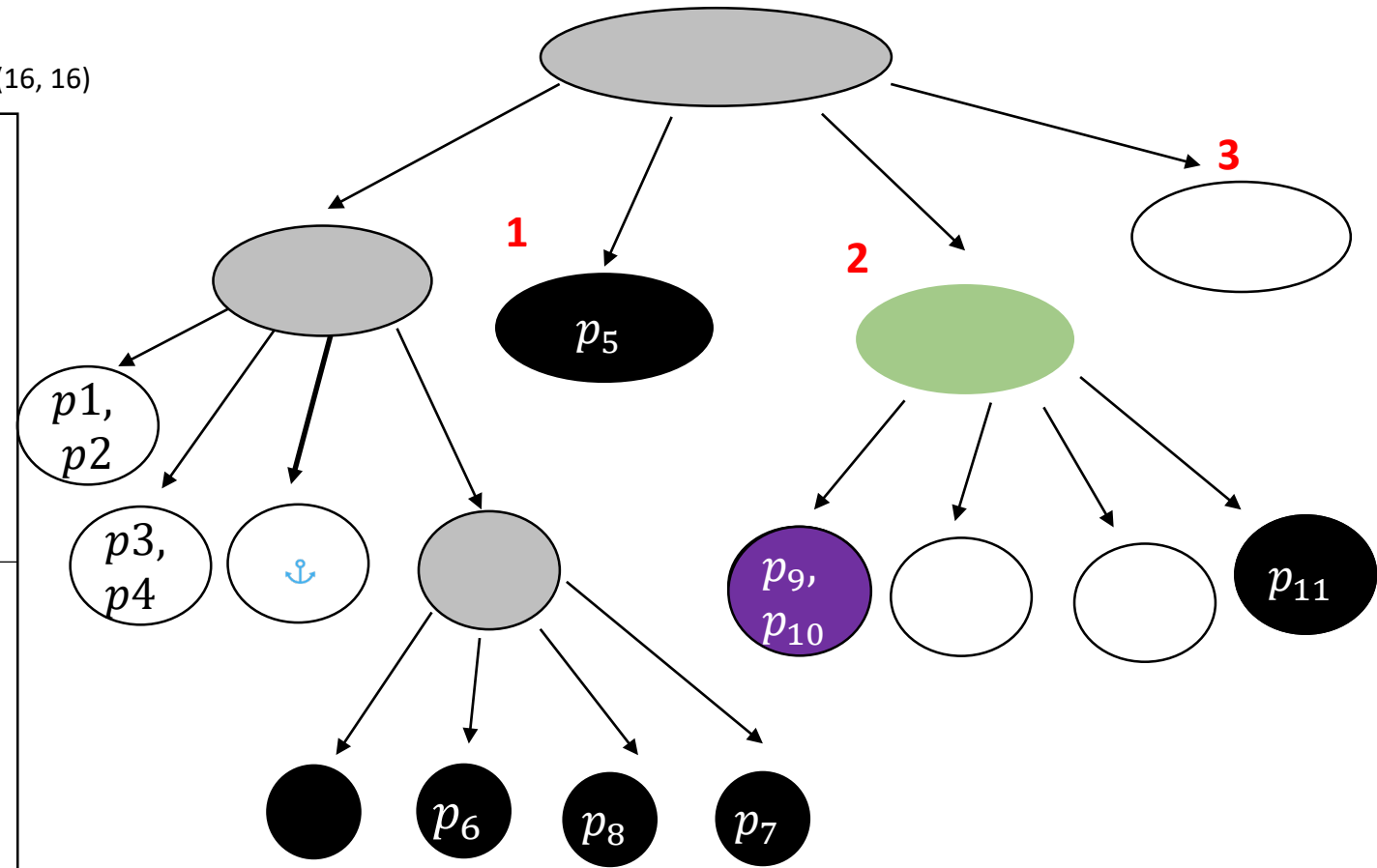
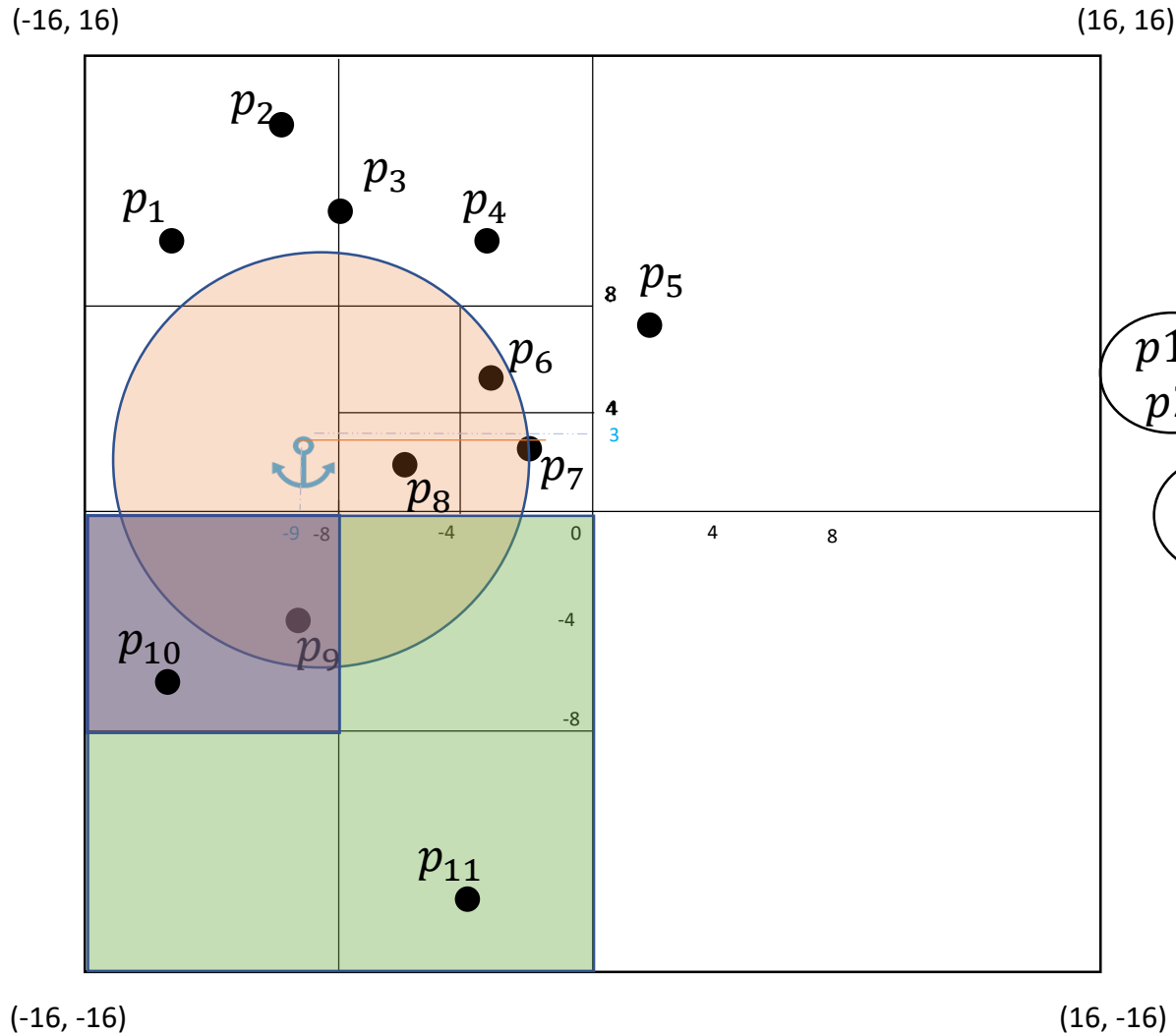
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



- **Non-zero intersection** between currently spanned quadrant and range!
- Also, gray node

# Example of range query in PR-QuadTree

anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$

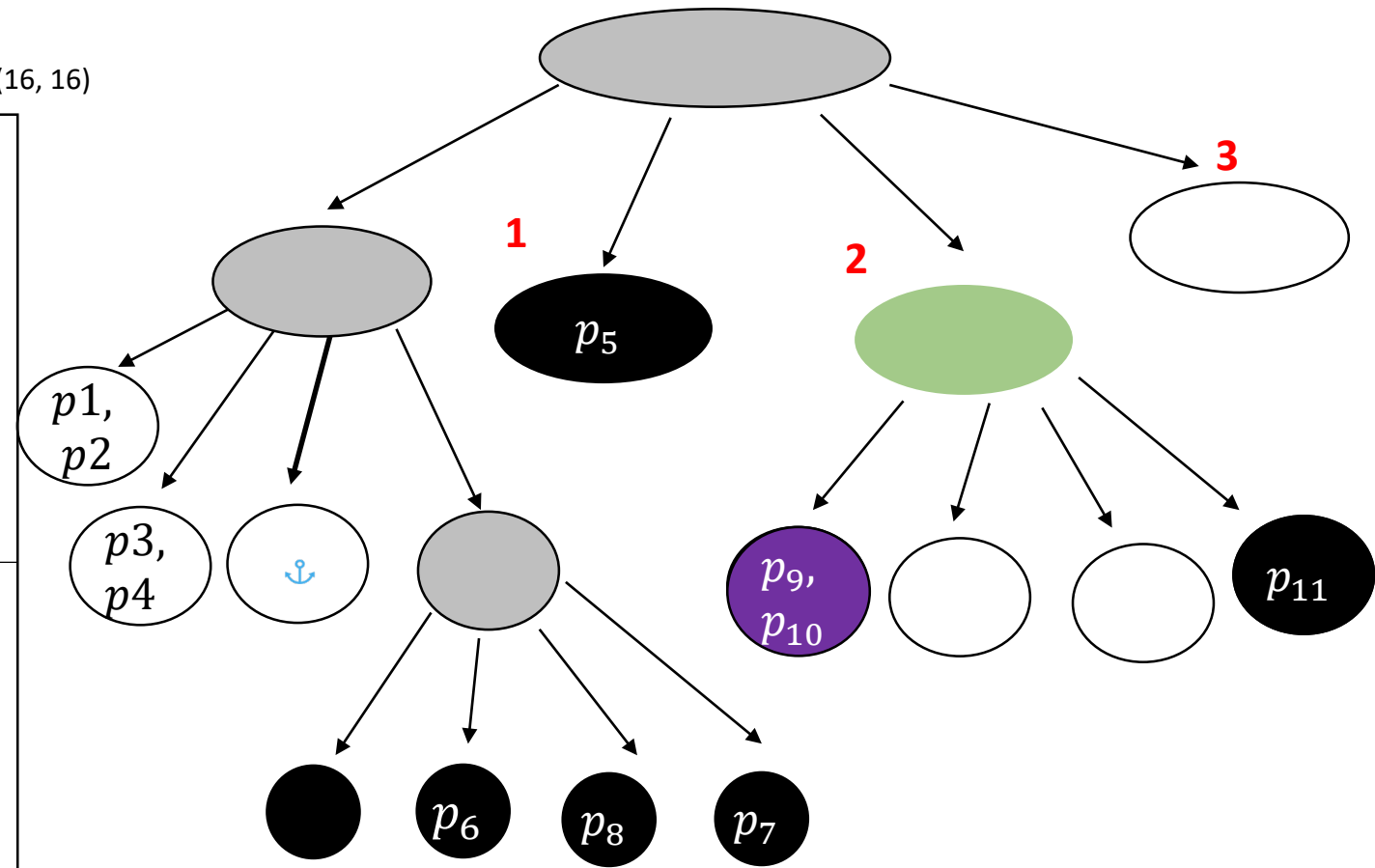
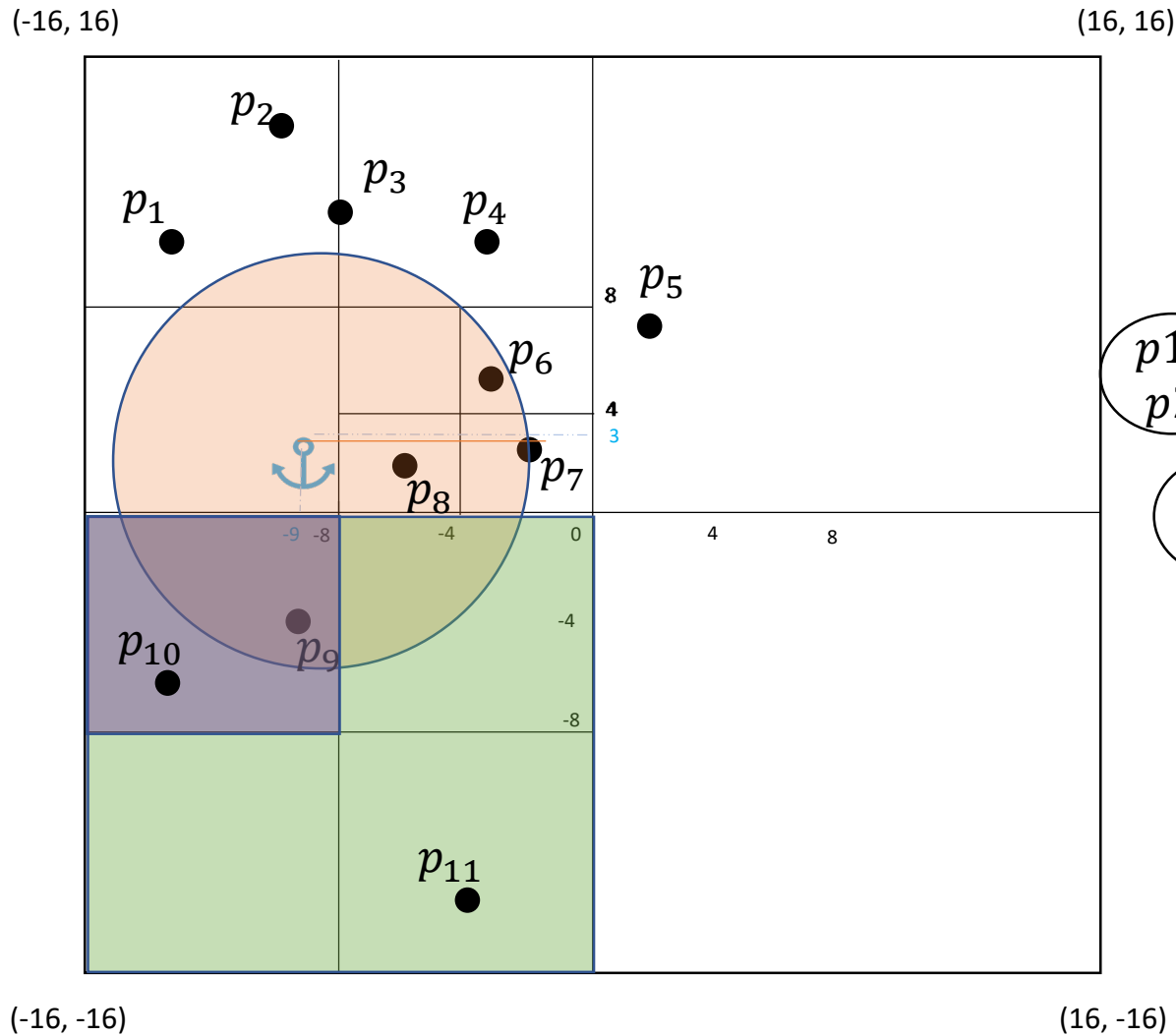


- Recurse to child **closest to anchor** first!
- That happens to be the **NW child**



# Example of range query in PR-QuadTree

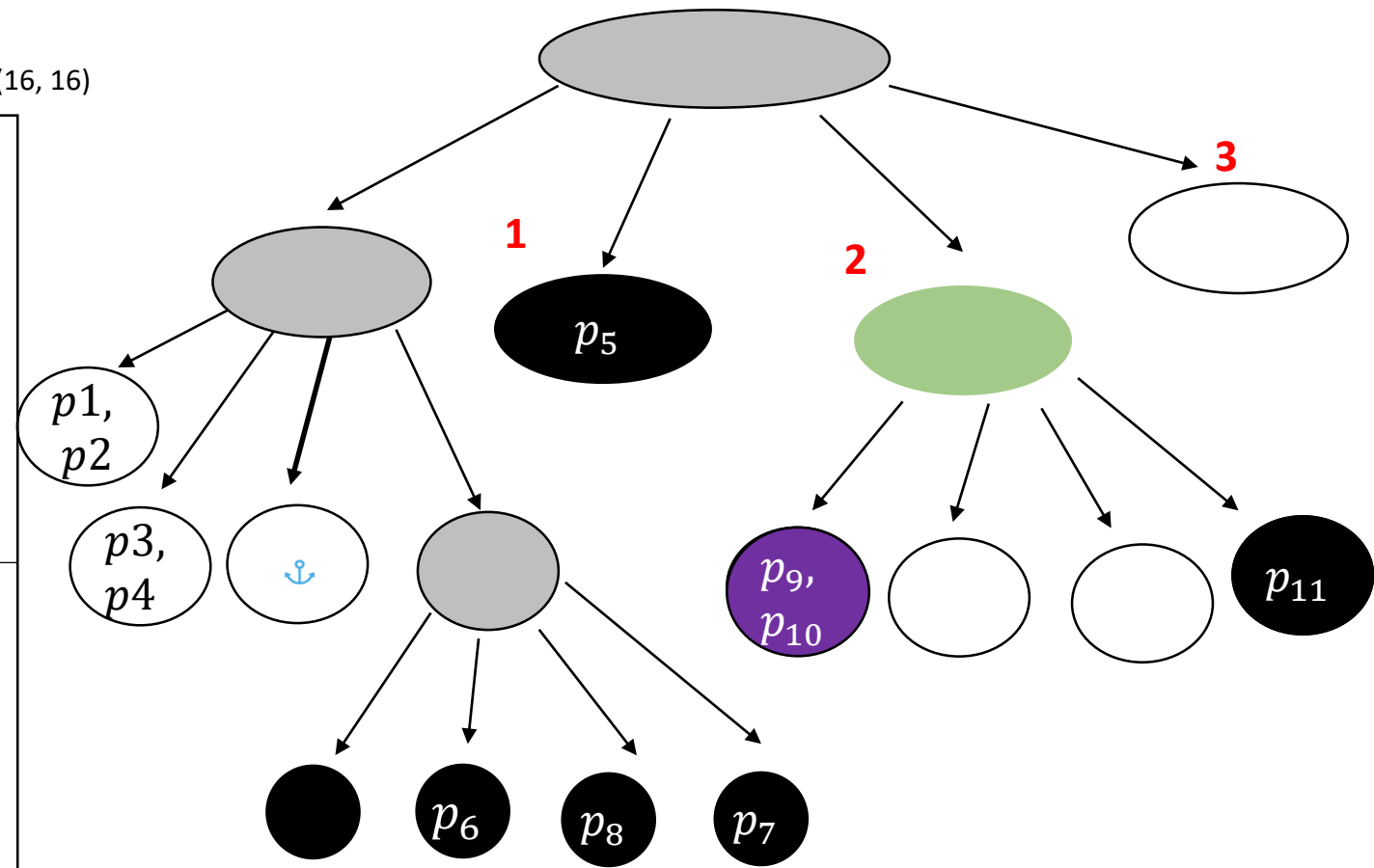
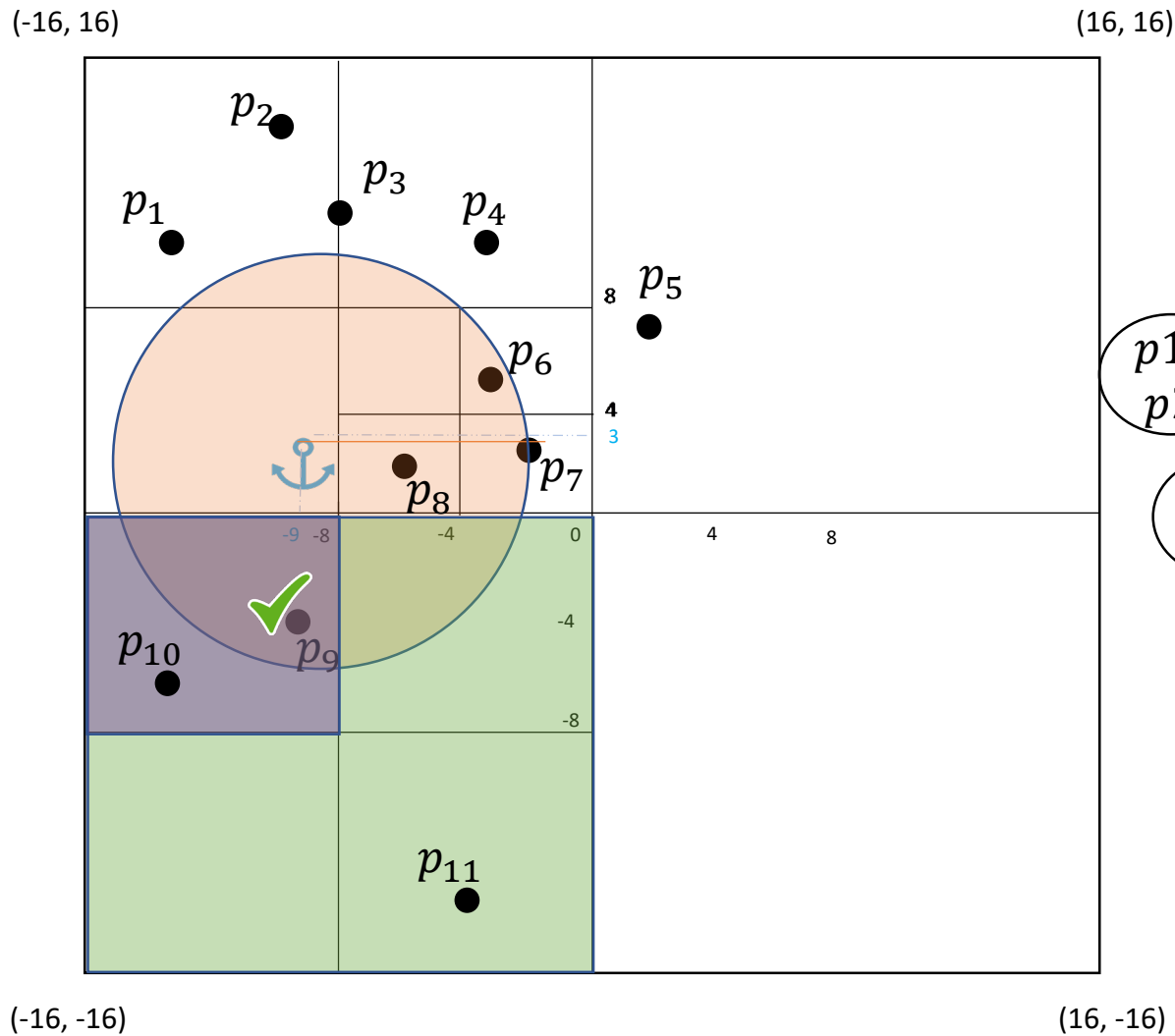
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



- Recurse to child **closest to anchor** first!
- That happens to be the **NW child**
- **Non-zero intersection**, black node

# Example of range query in PR-QuadTree

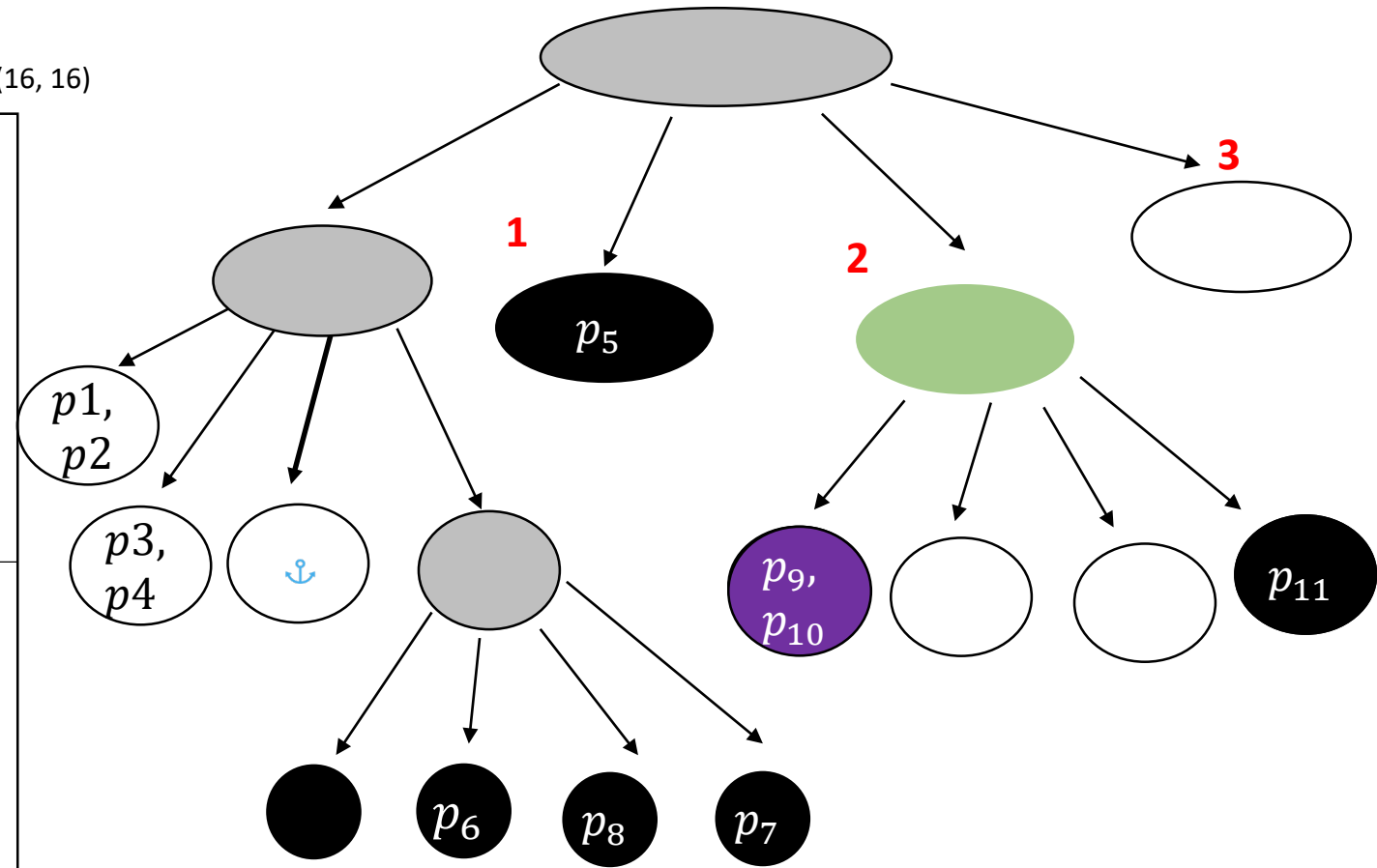
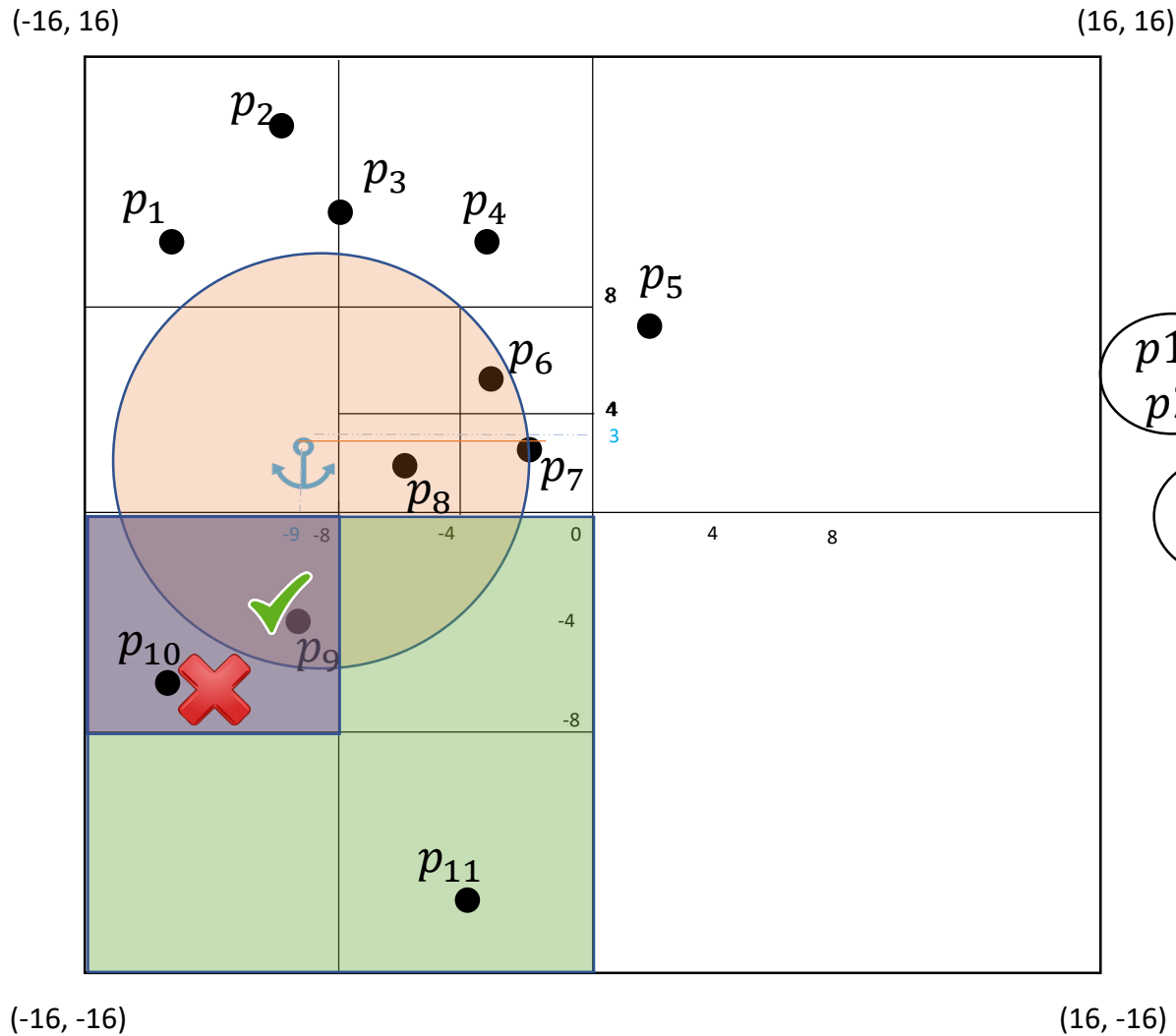
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



- $p_9$  is within the range 😊

# Example of range query in PR-QuadTree

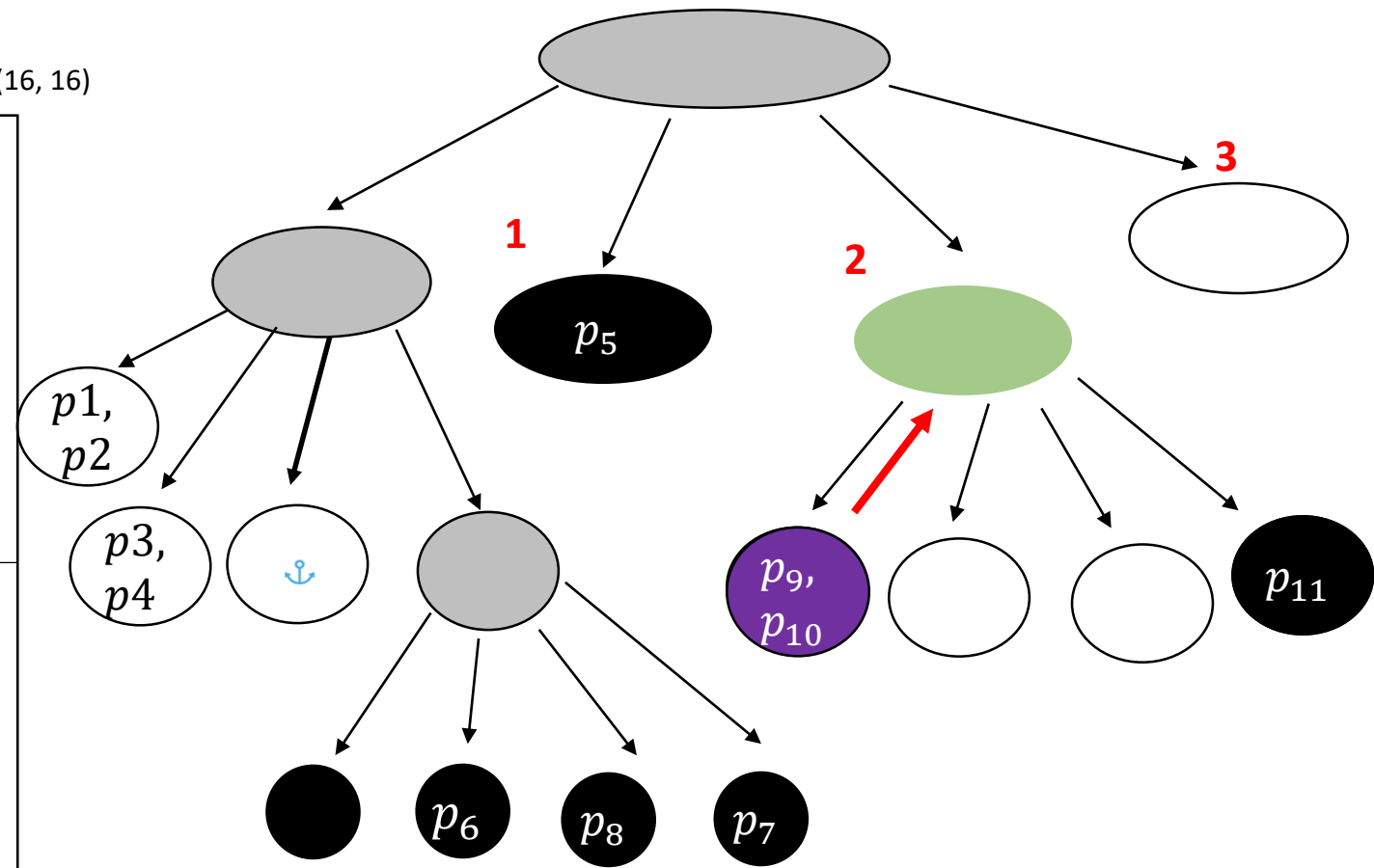
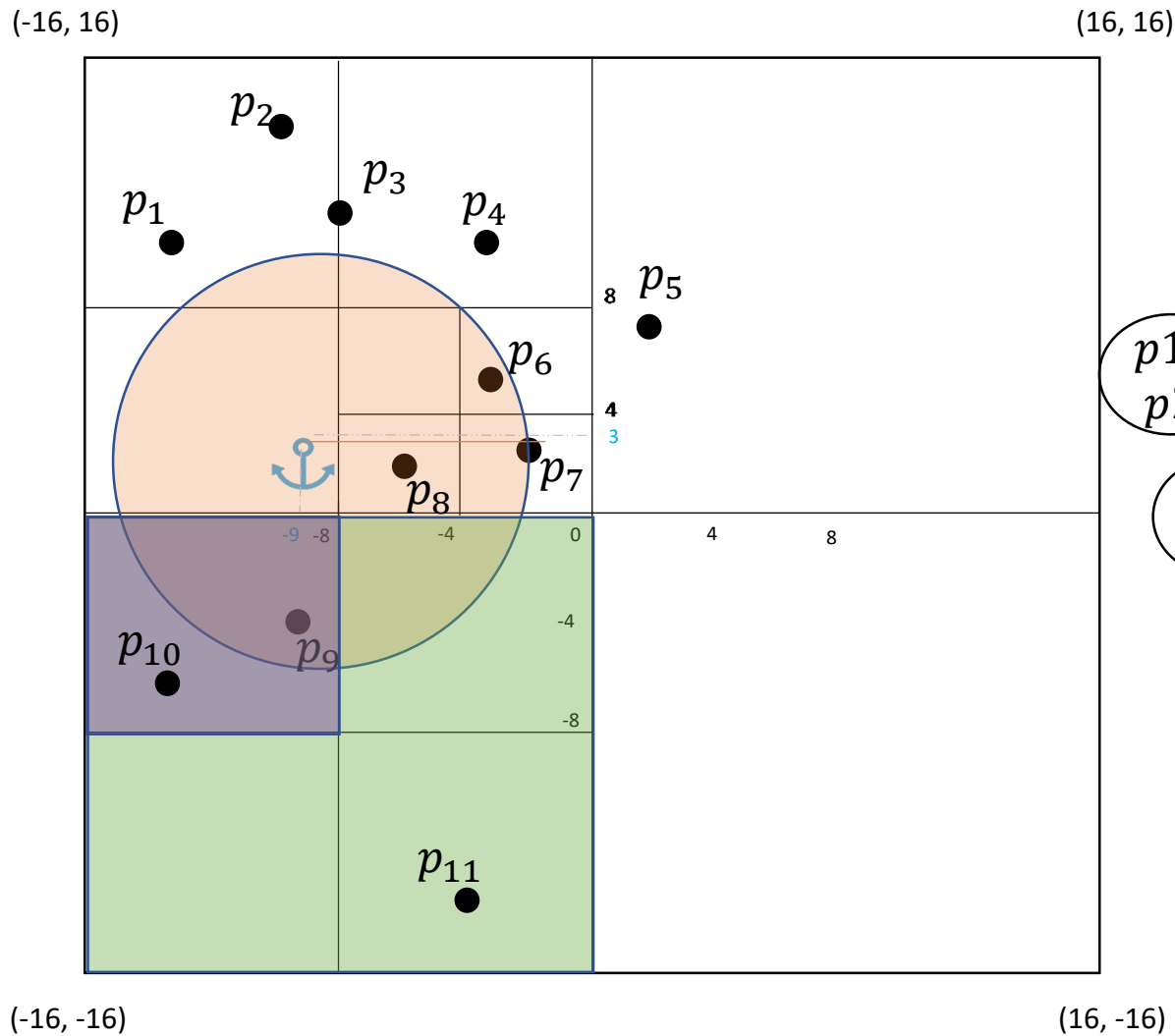
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



- $p_9$  is within the range 😊
- $p_{10}$  is not 😞

# Example of range query in PR-QuadTree

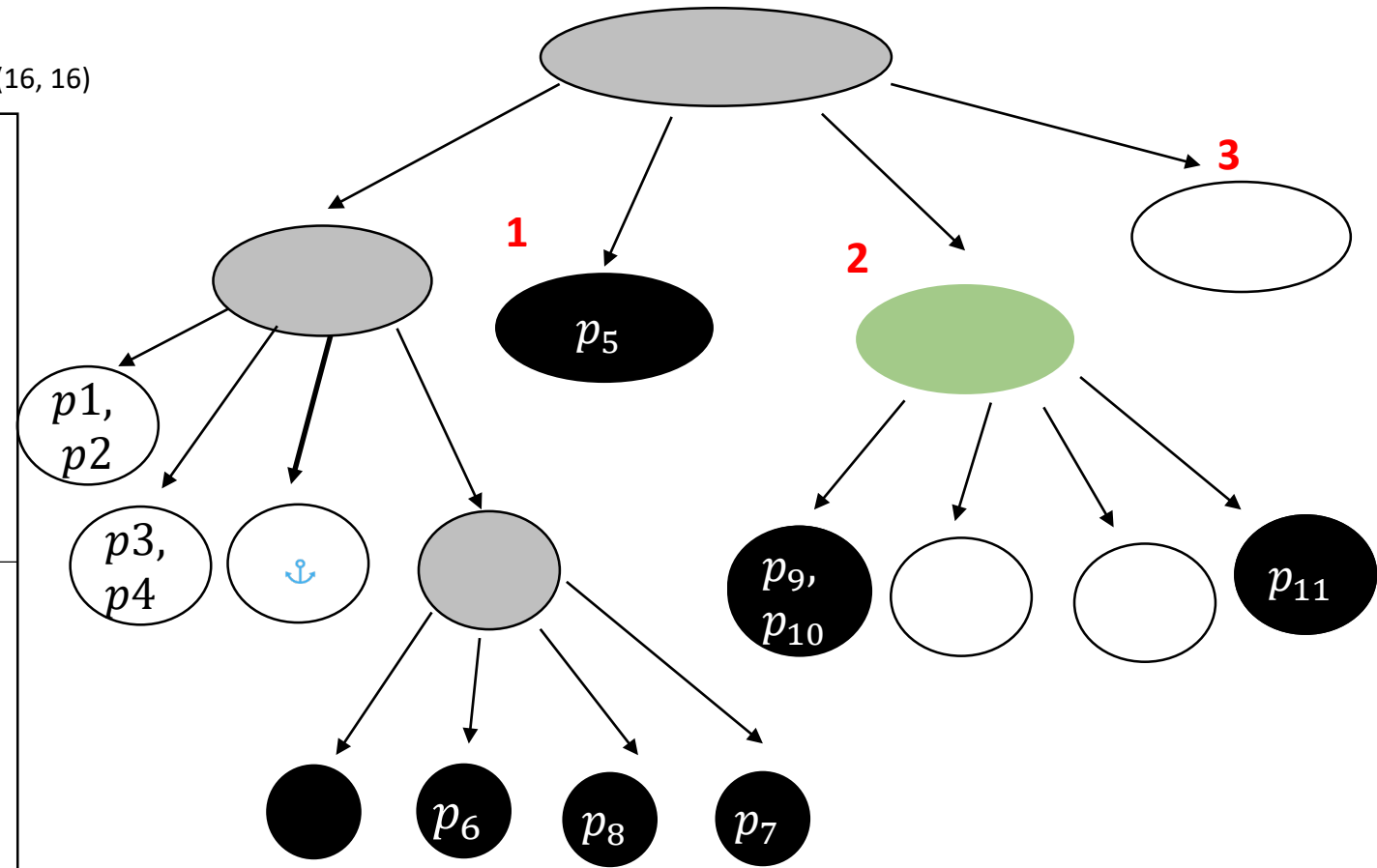
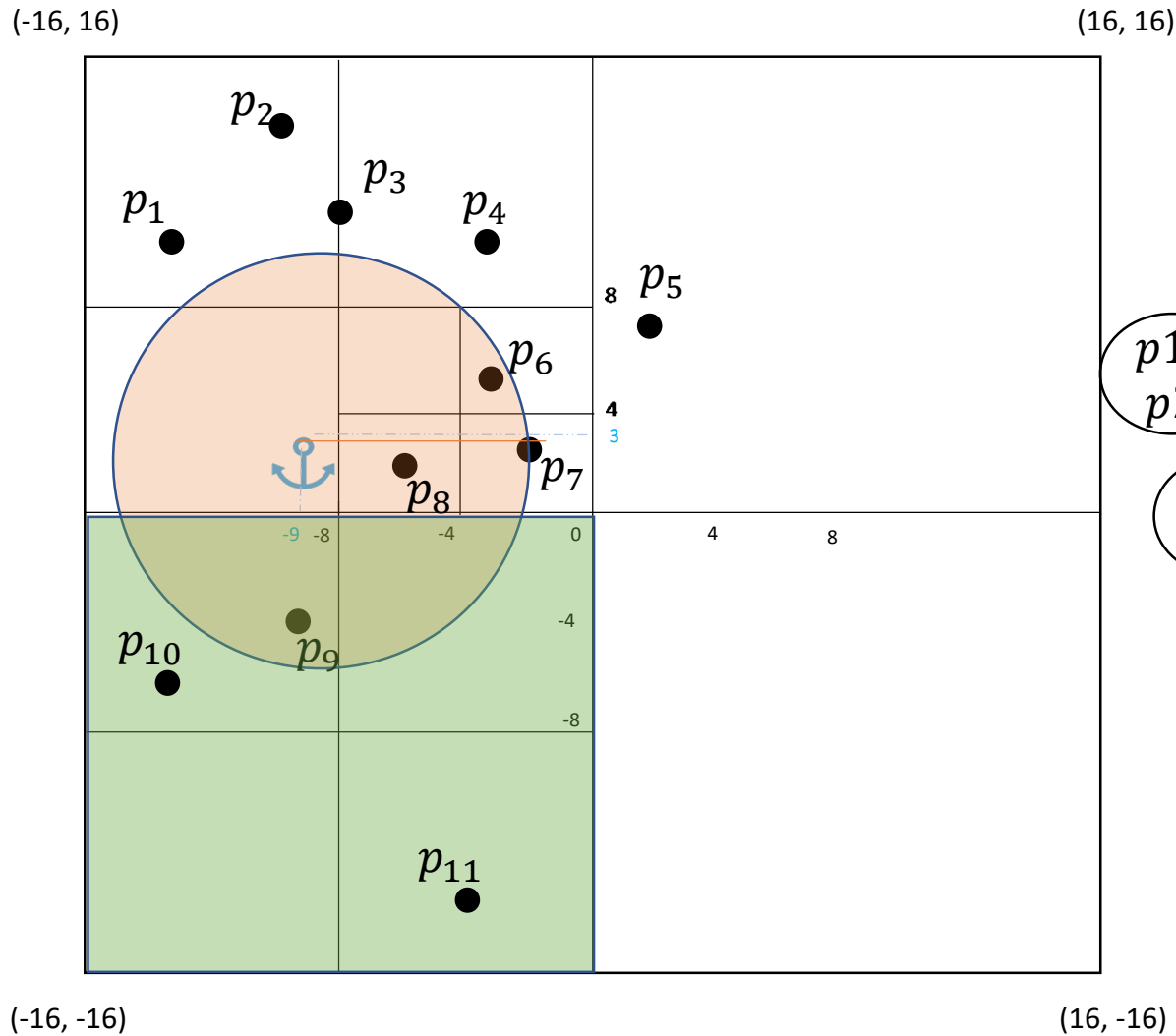
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



• **Backtrack**

# Example of range query in PR-QuadTree

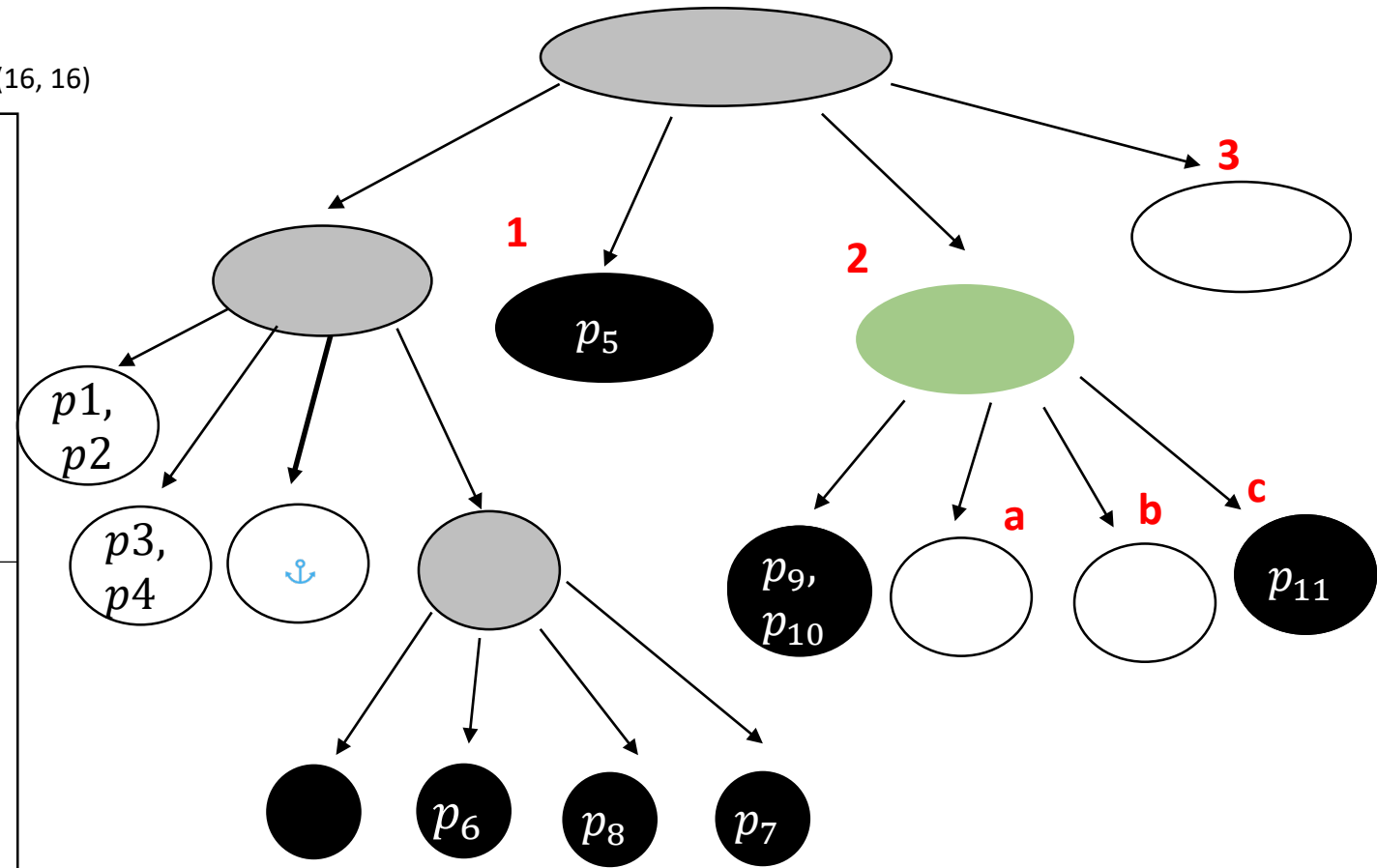
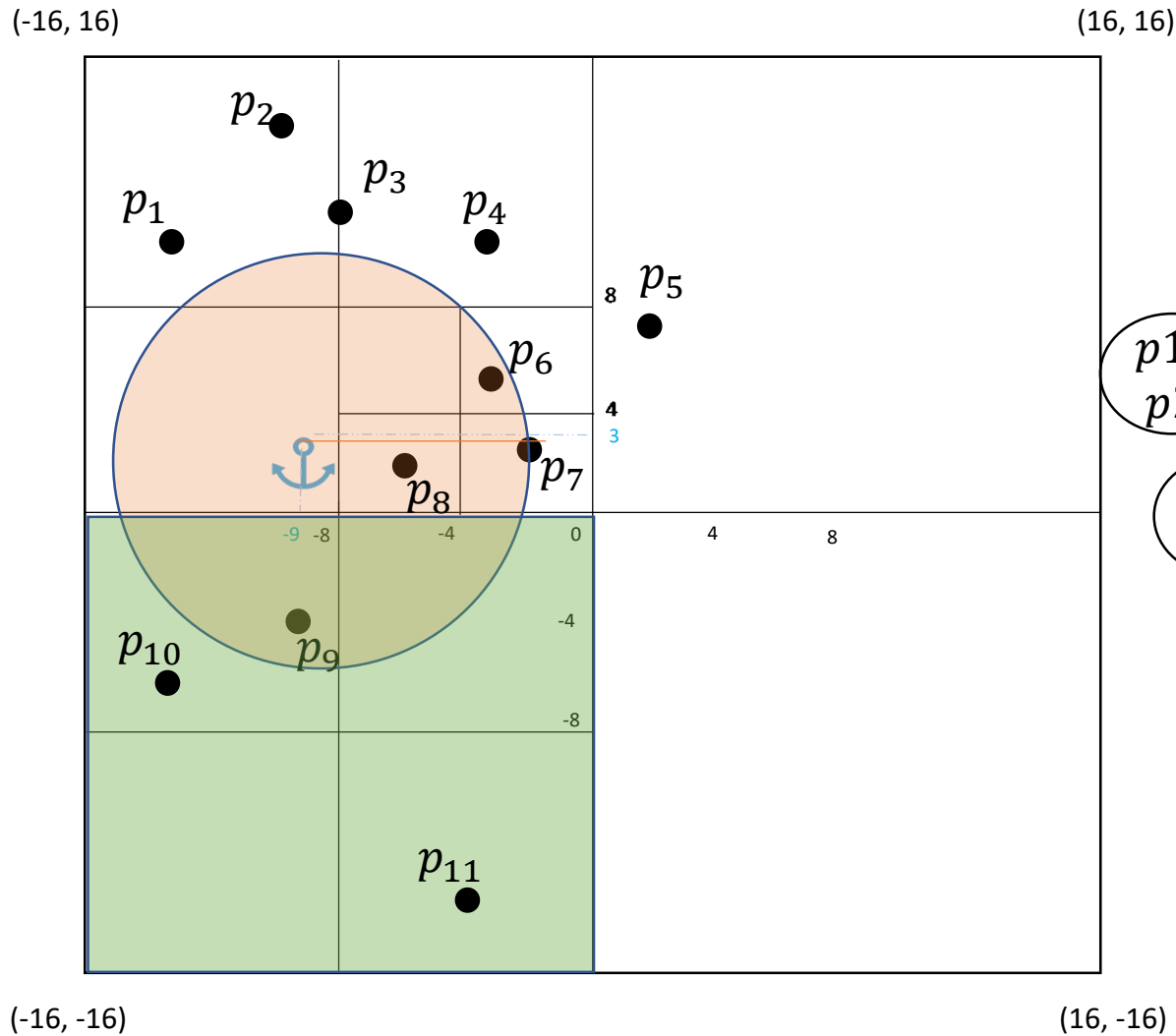
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



- Loop through the rest of the children **in Z-order**

# Example of range query in PR-QuadTree

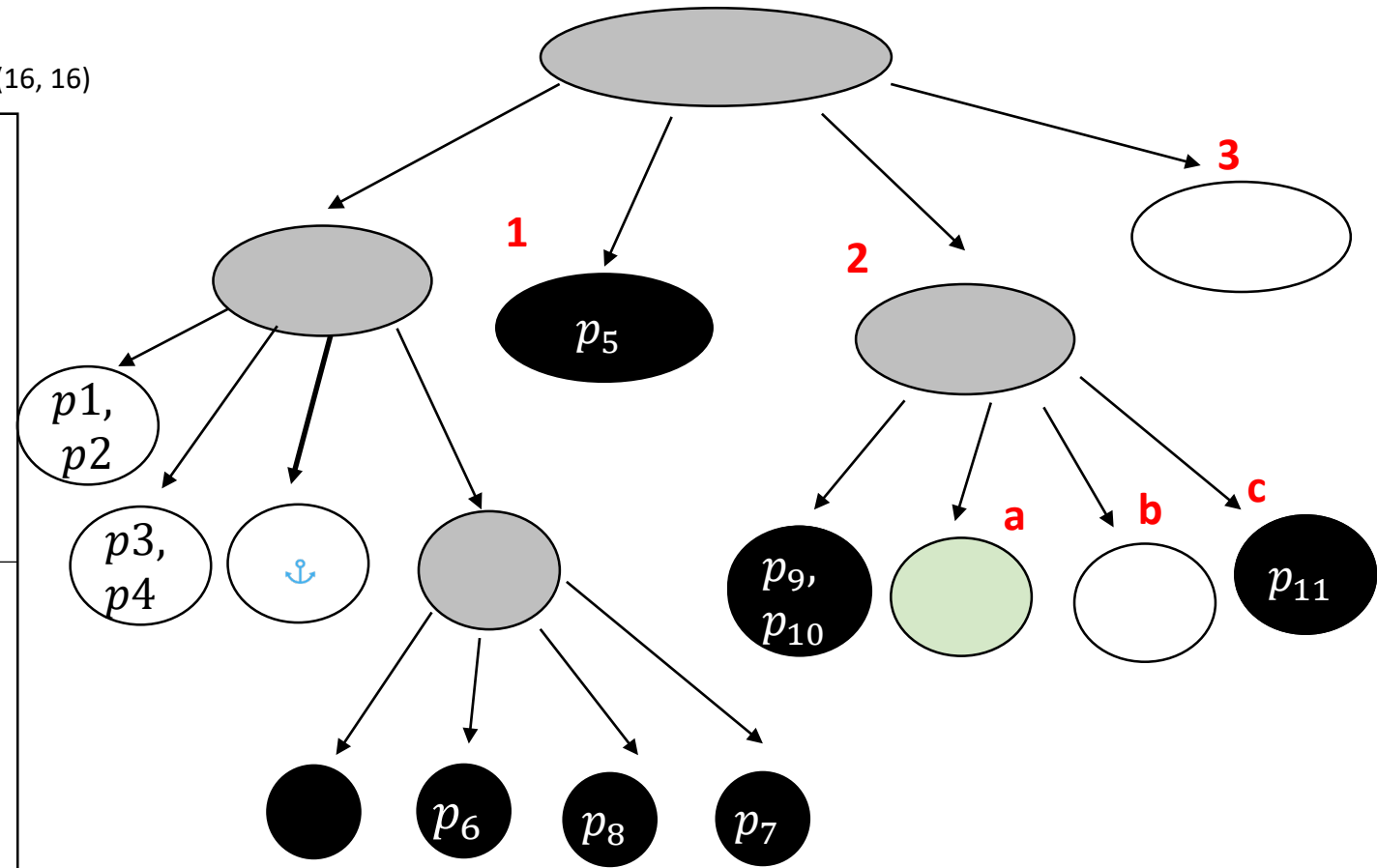
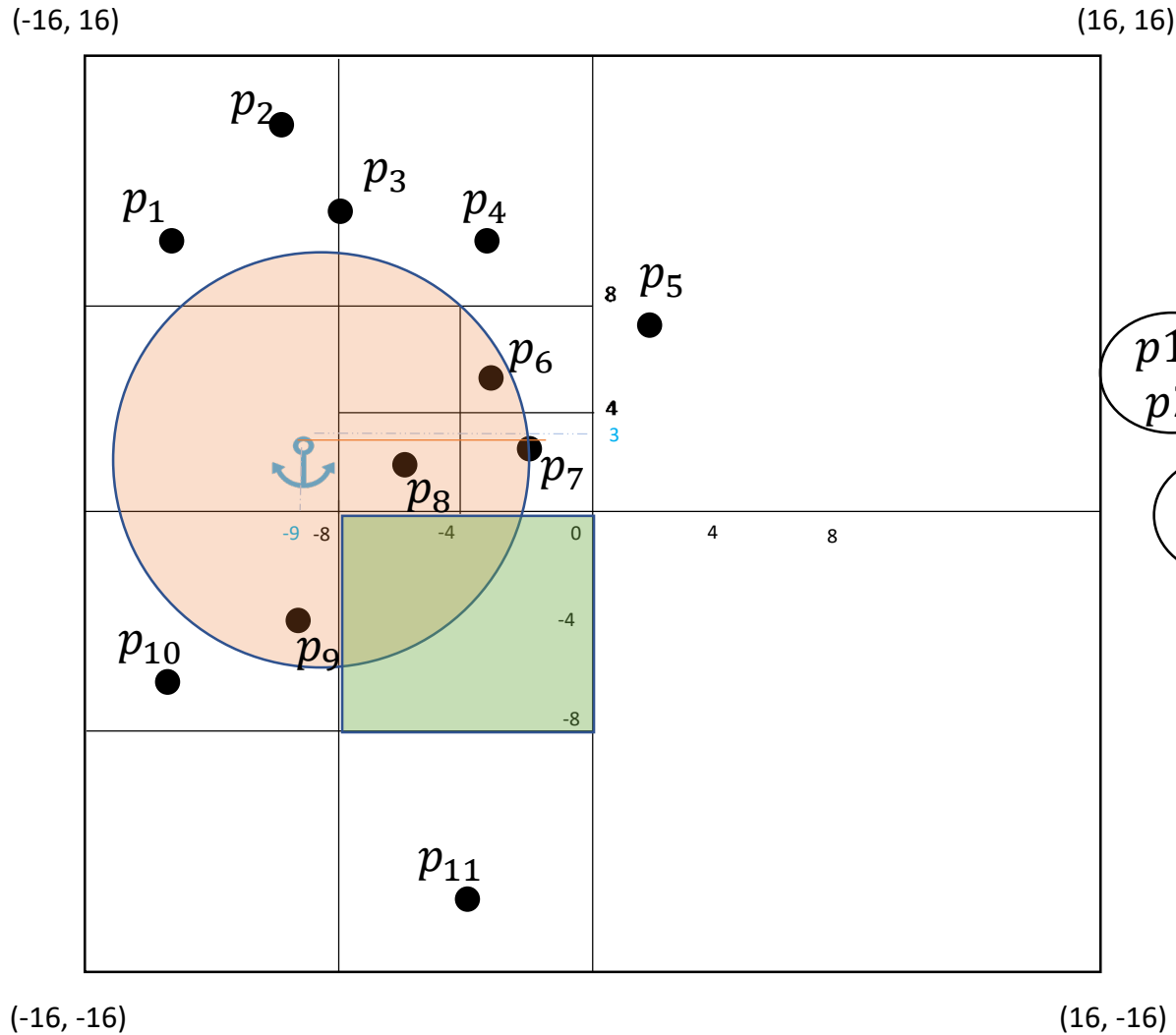
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



- Loop through the rest of the children **in Z-order**

# Example of range query in PR-QuadTree

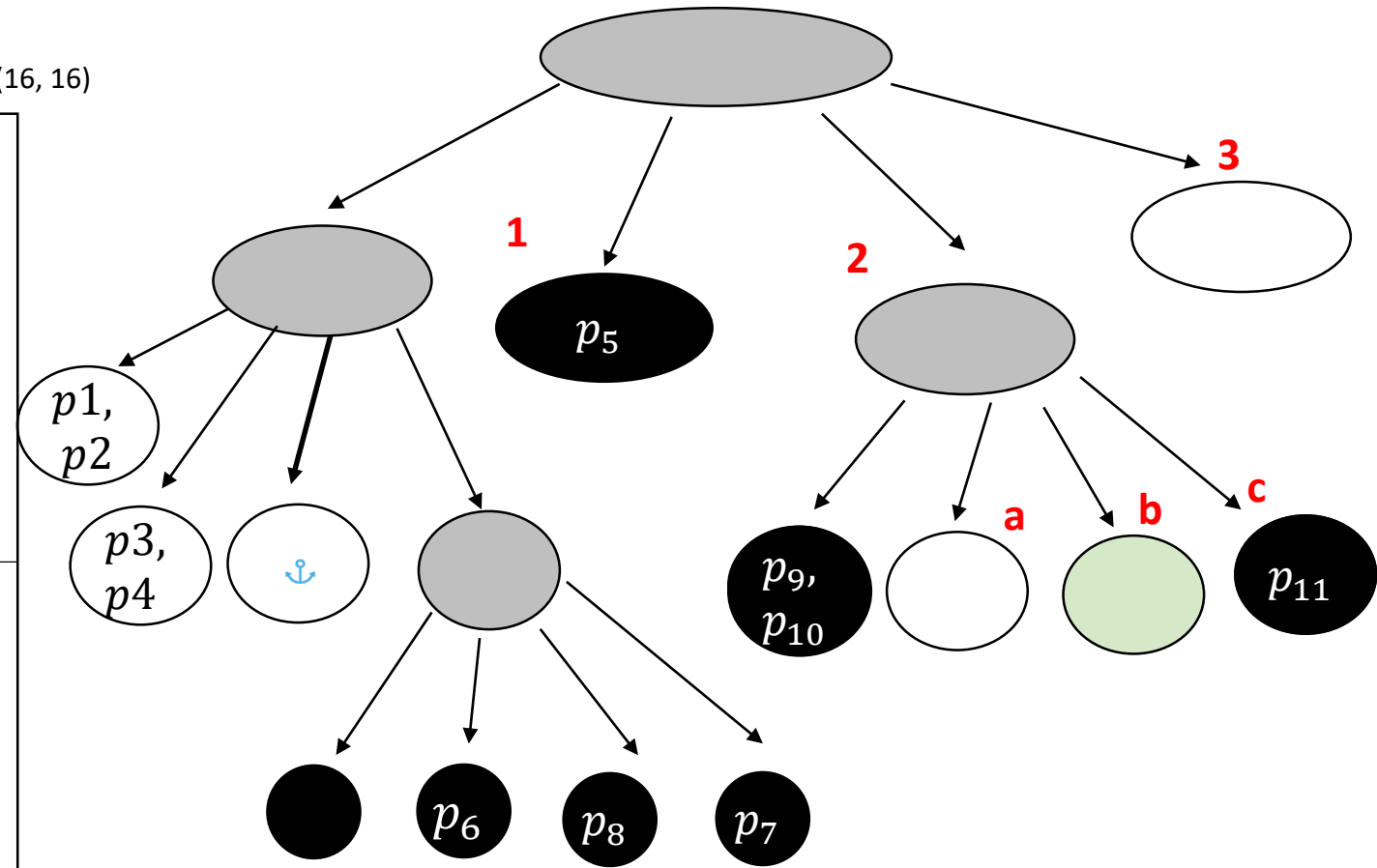
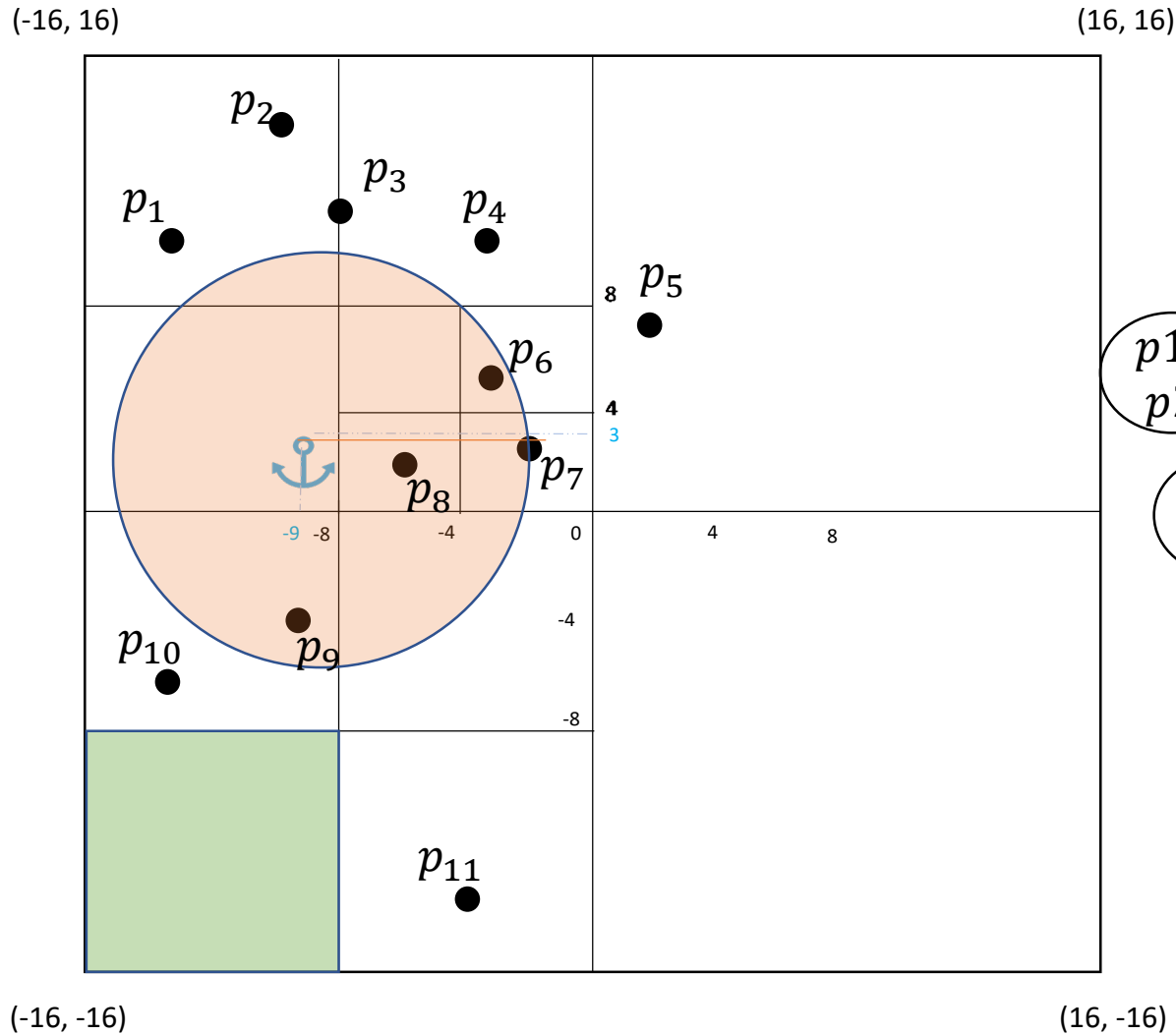
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



- Non-zero intersection, but **white node**; no contribution to solution set.

# Example of range query in PR-QuadTree

anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$

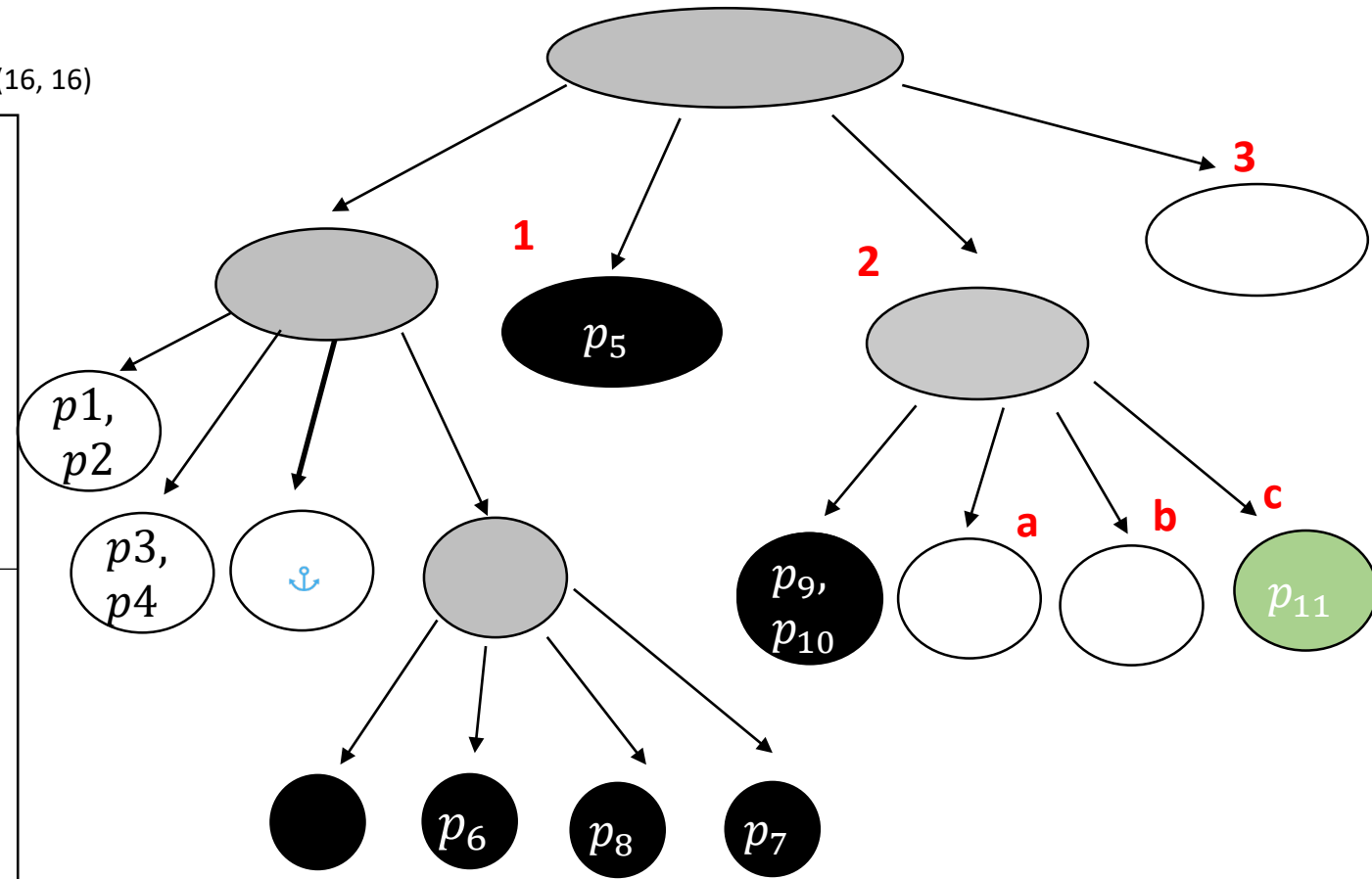
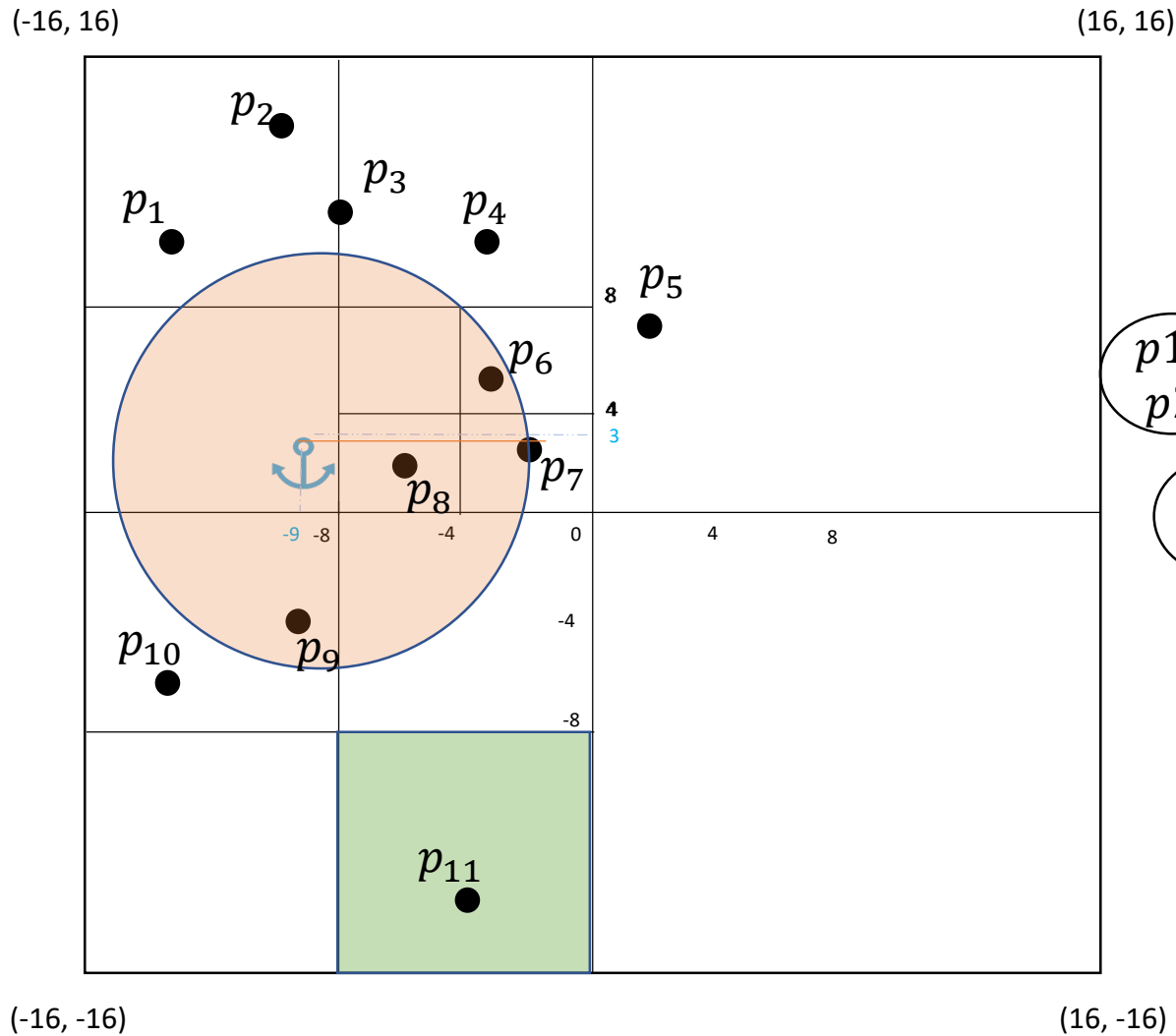


- Zero intersection and white node; no contribution to solution set.



# Example of range query in PR-QuadTree

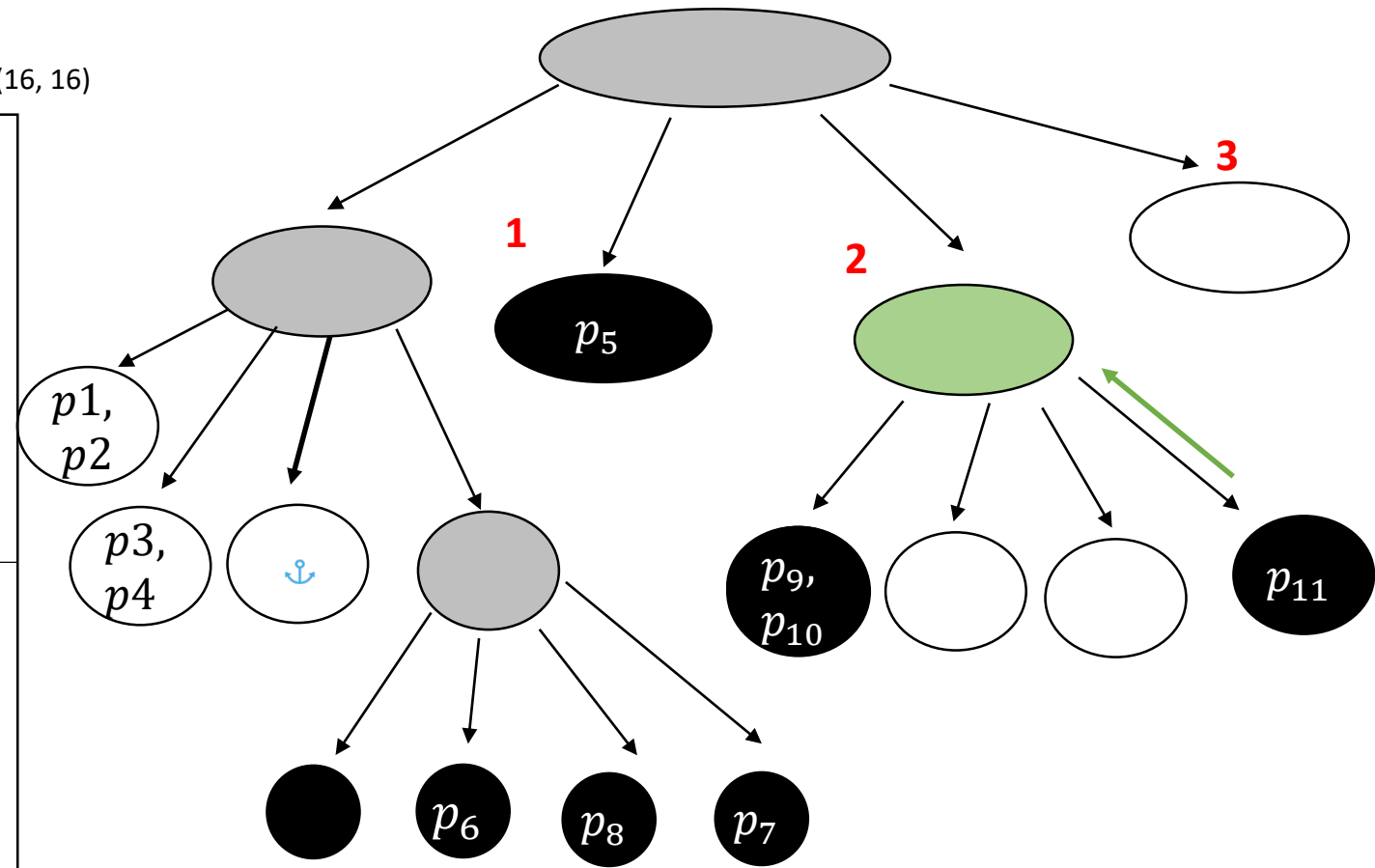
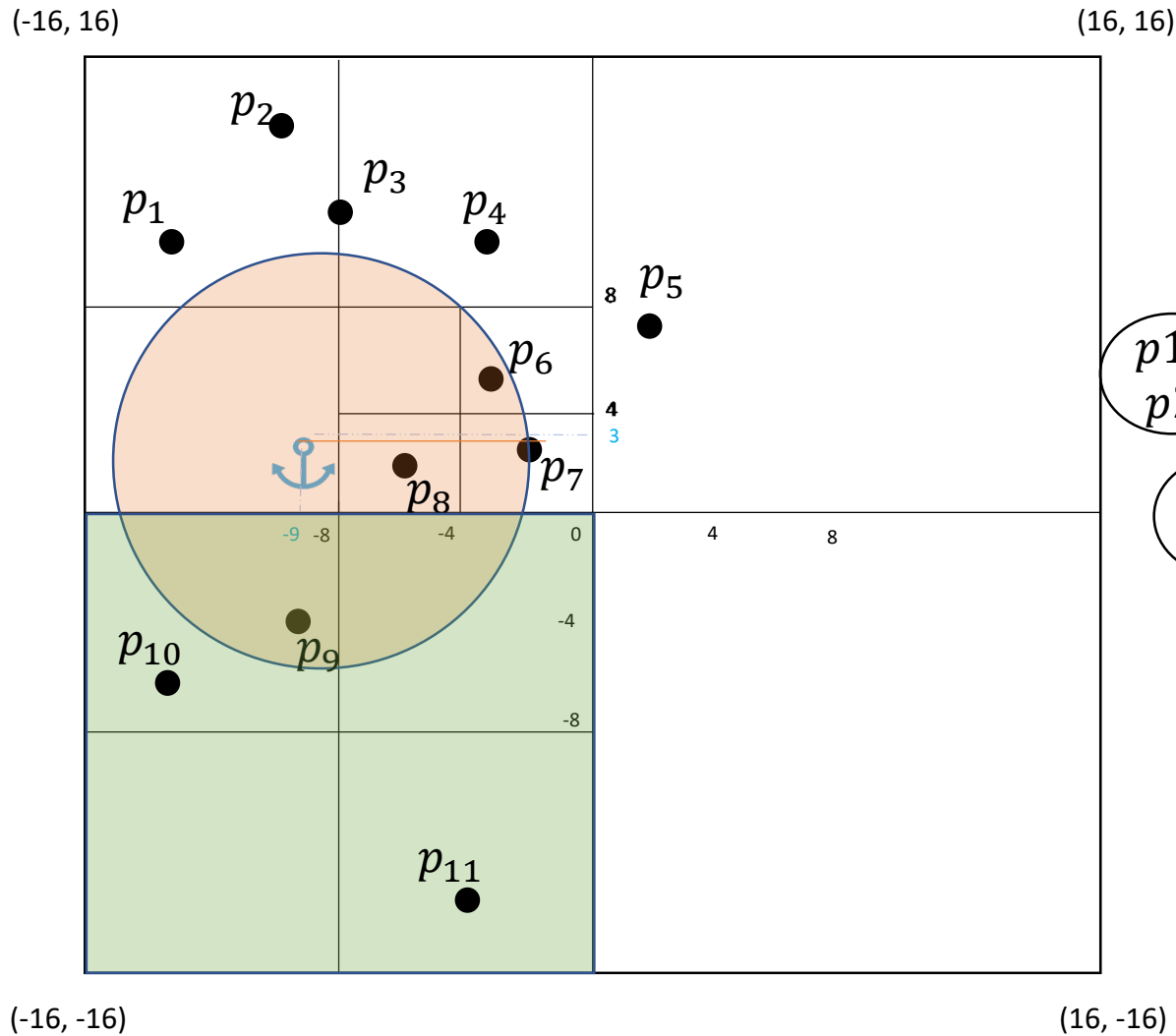
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



- Black node, but **zero intersection** 😞
- **Can't contribute to solution set** 😞

# Example of range query in PR-QuadTree

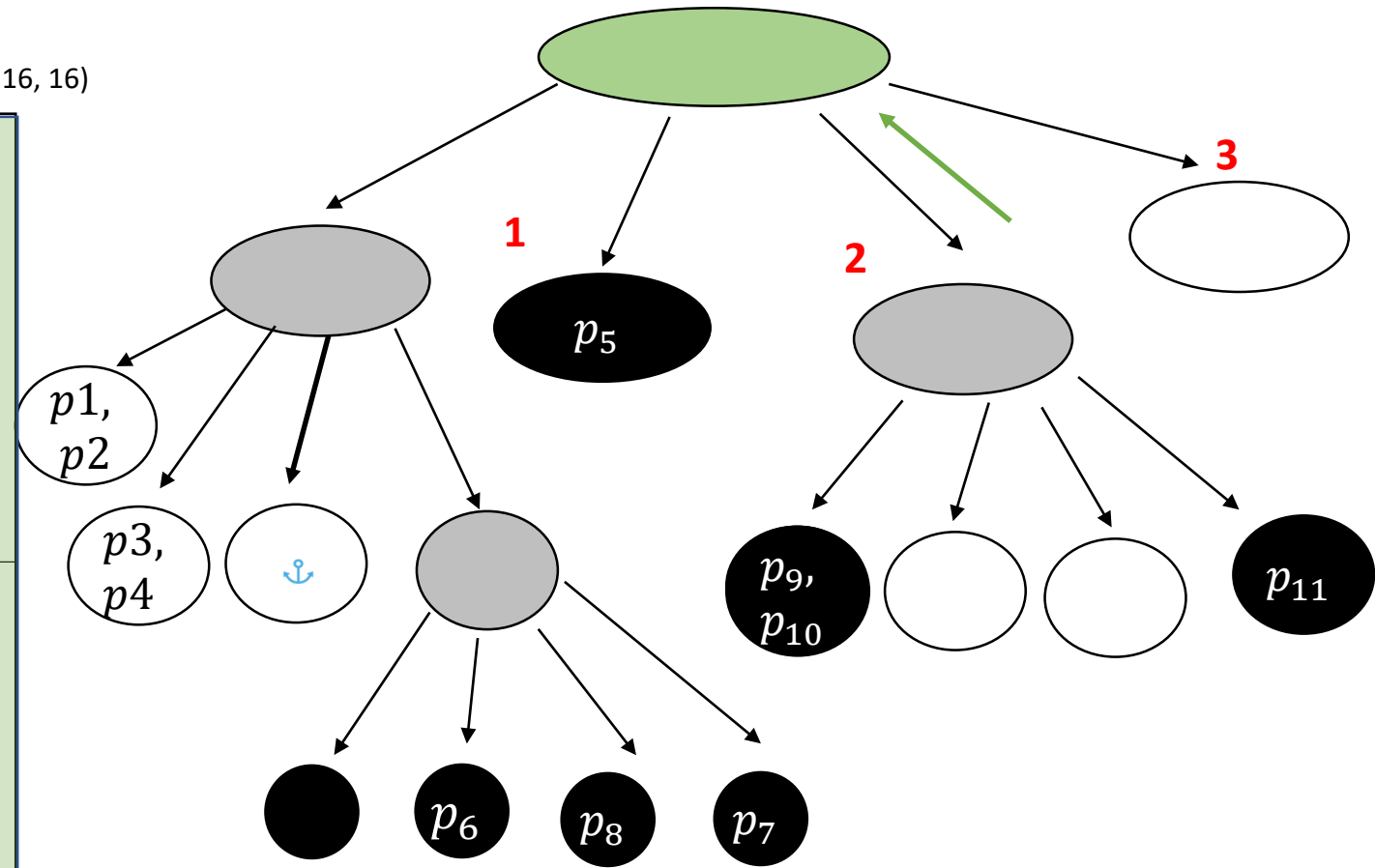
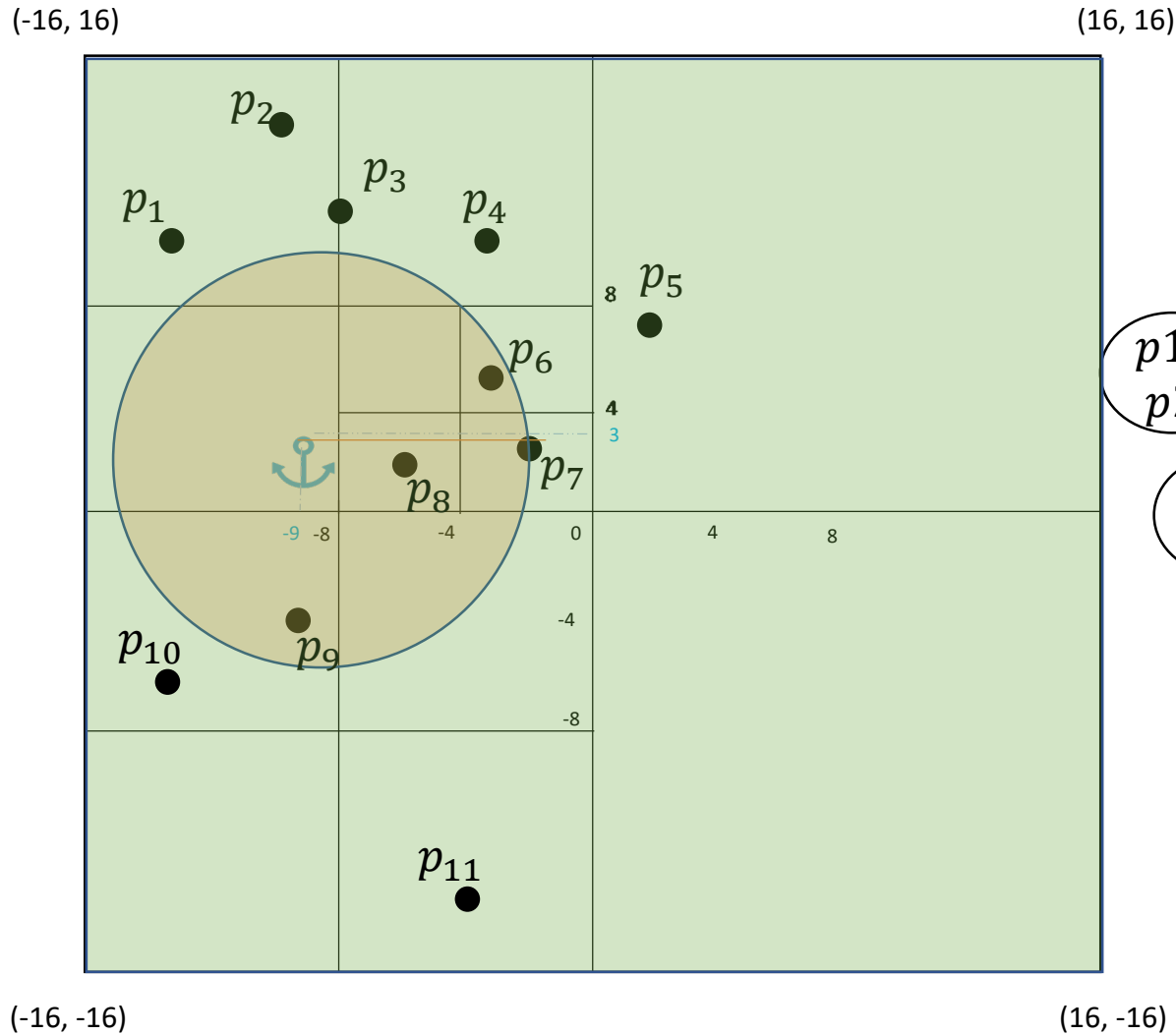
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



- Backtrack to root...

# Example of range query in PR-QuadTree

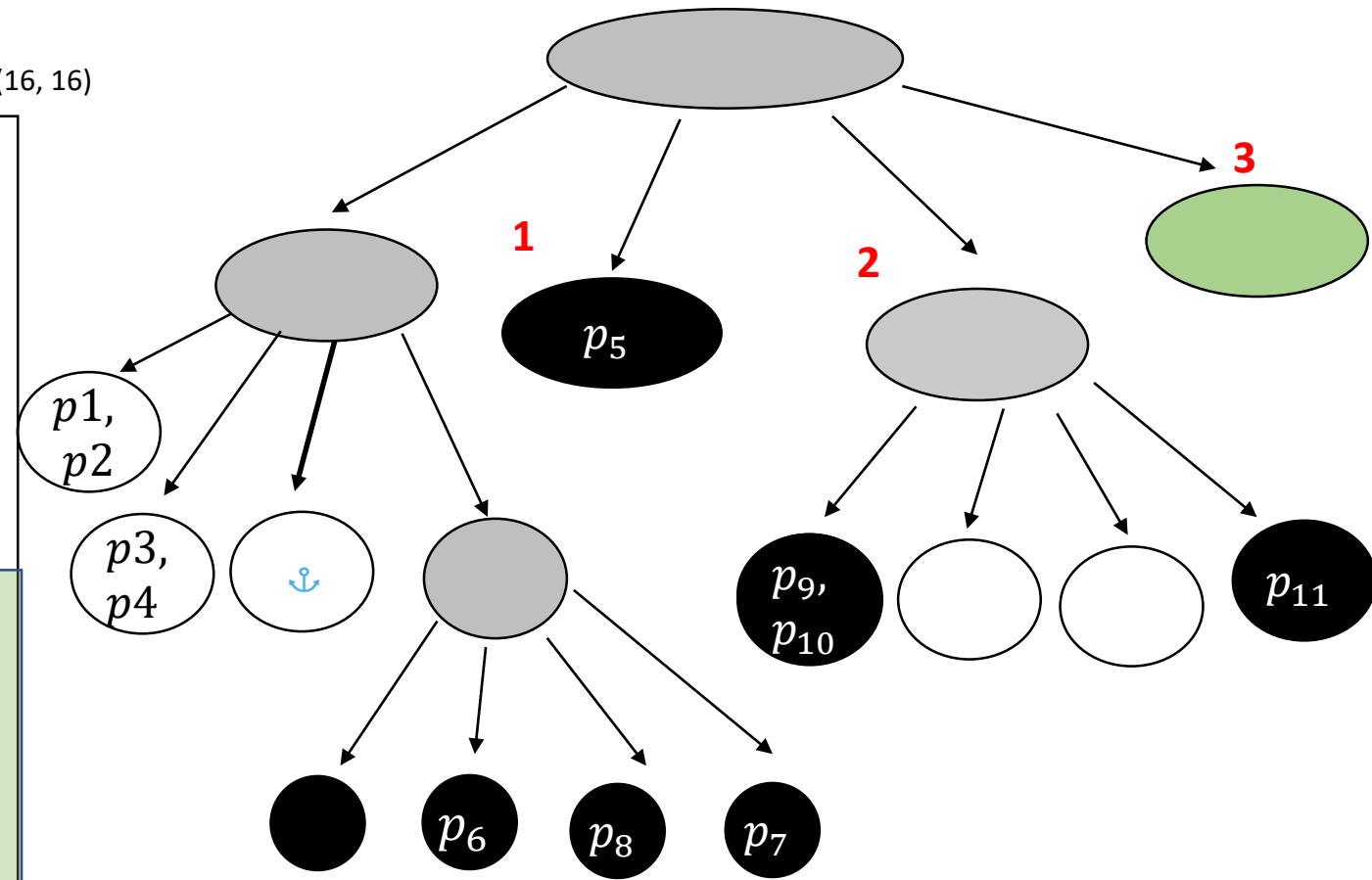
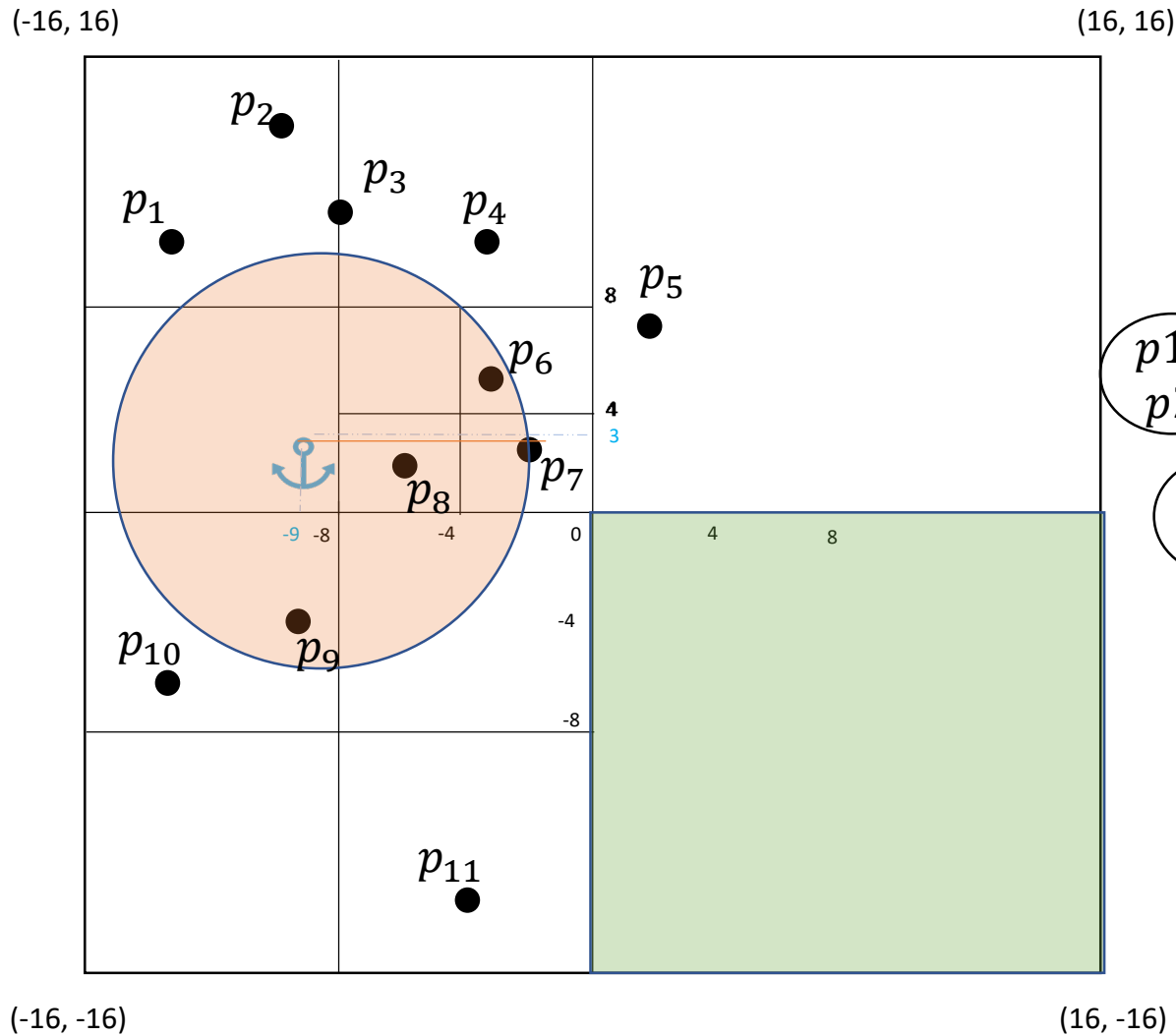
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



- Backtrack to root...

# Example of range query in PR-QuadTree

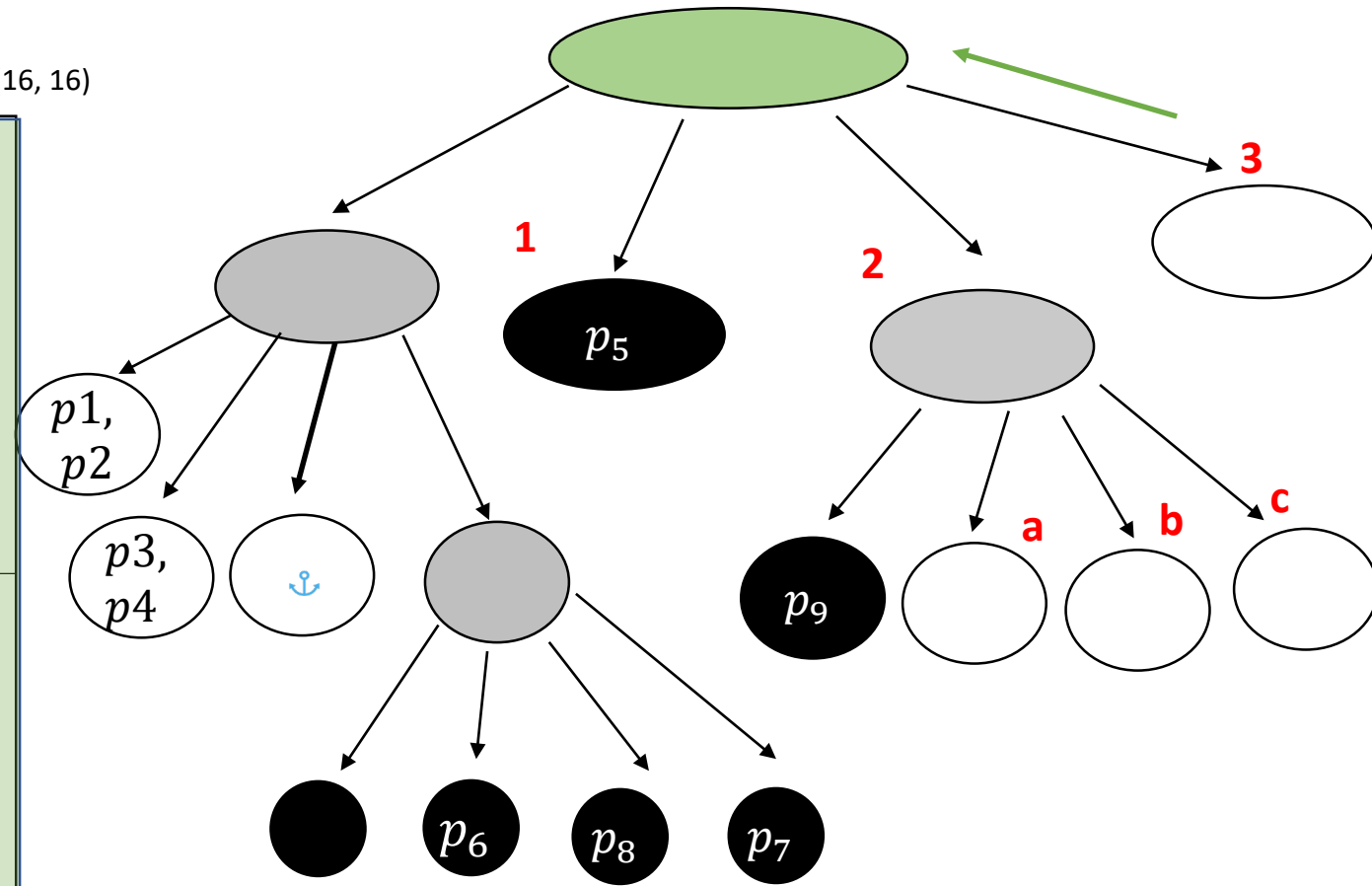
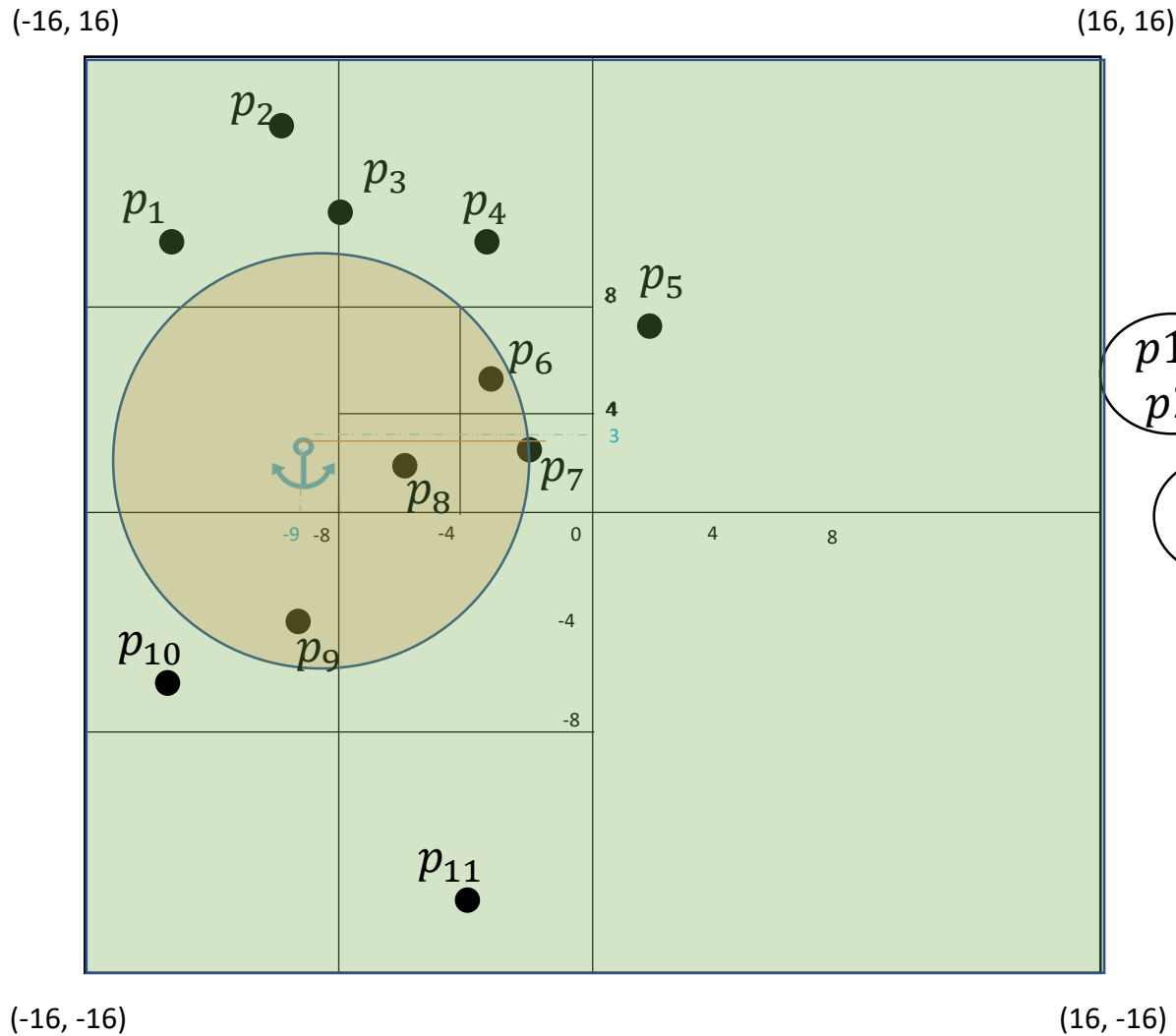
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



- White node **and** beyond range: impossible to contribute to solution set!

# Example of range query in PR-QuadTree

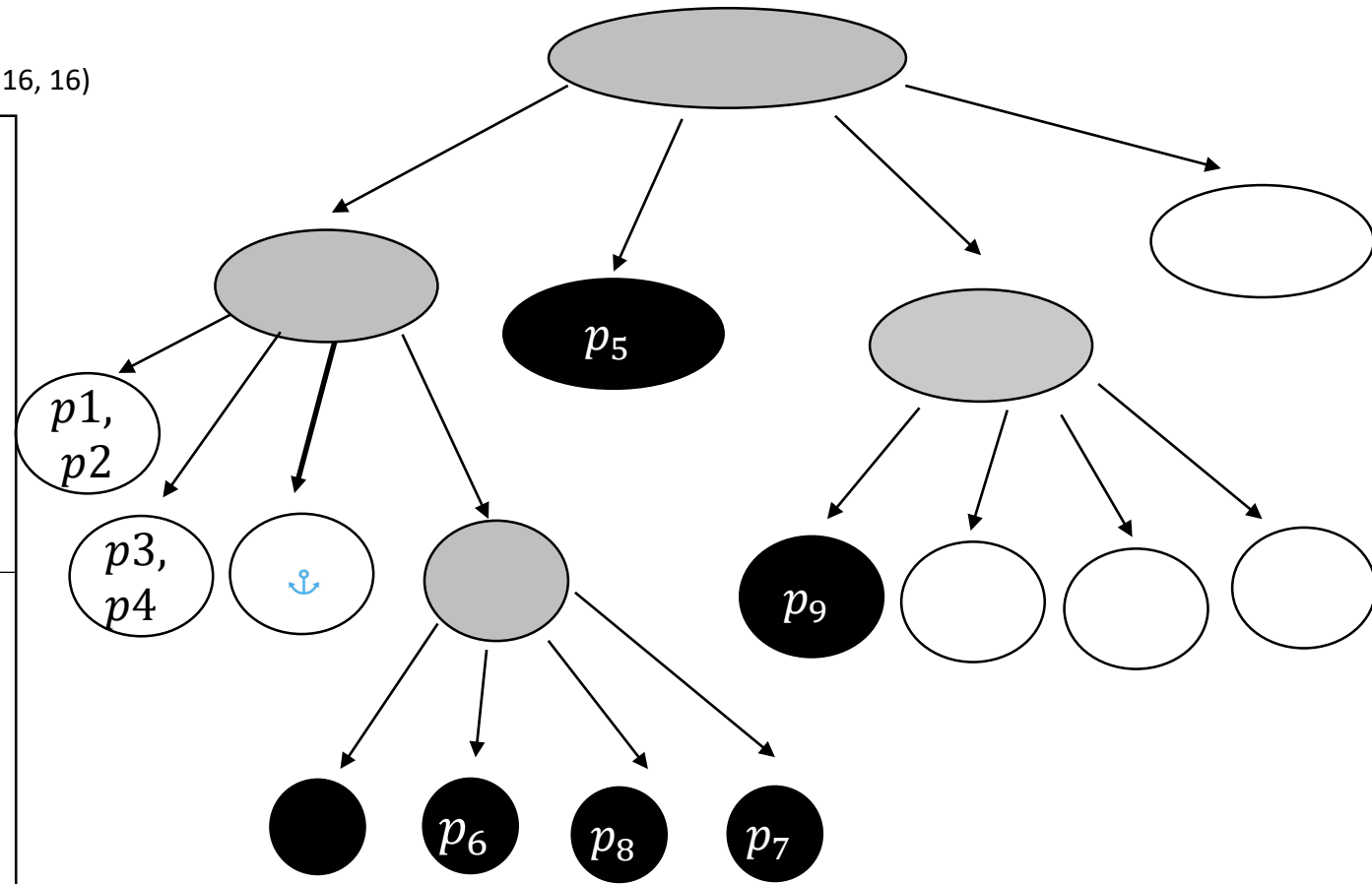
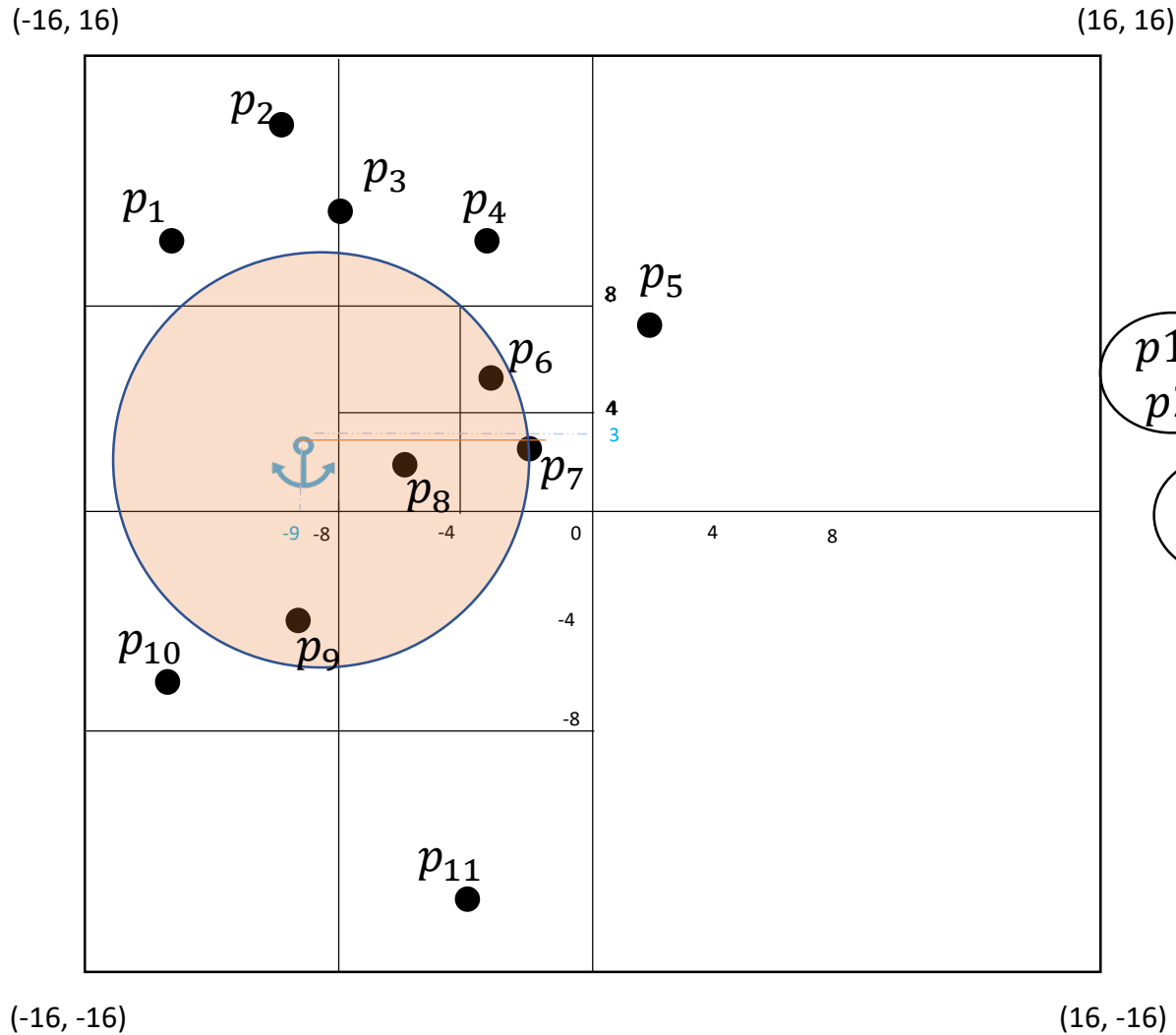
anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$



- White node **and** beyond range: impossible to contribute to solution set!
- **Backtrack....**

# Example of range query in PR-QuadTree

anchor =  $(-9, 3)$ , range  $r = 7$ , Bucket size  $b = 2$

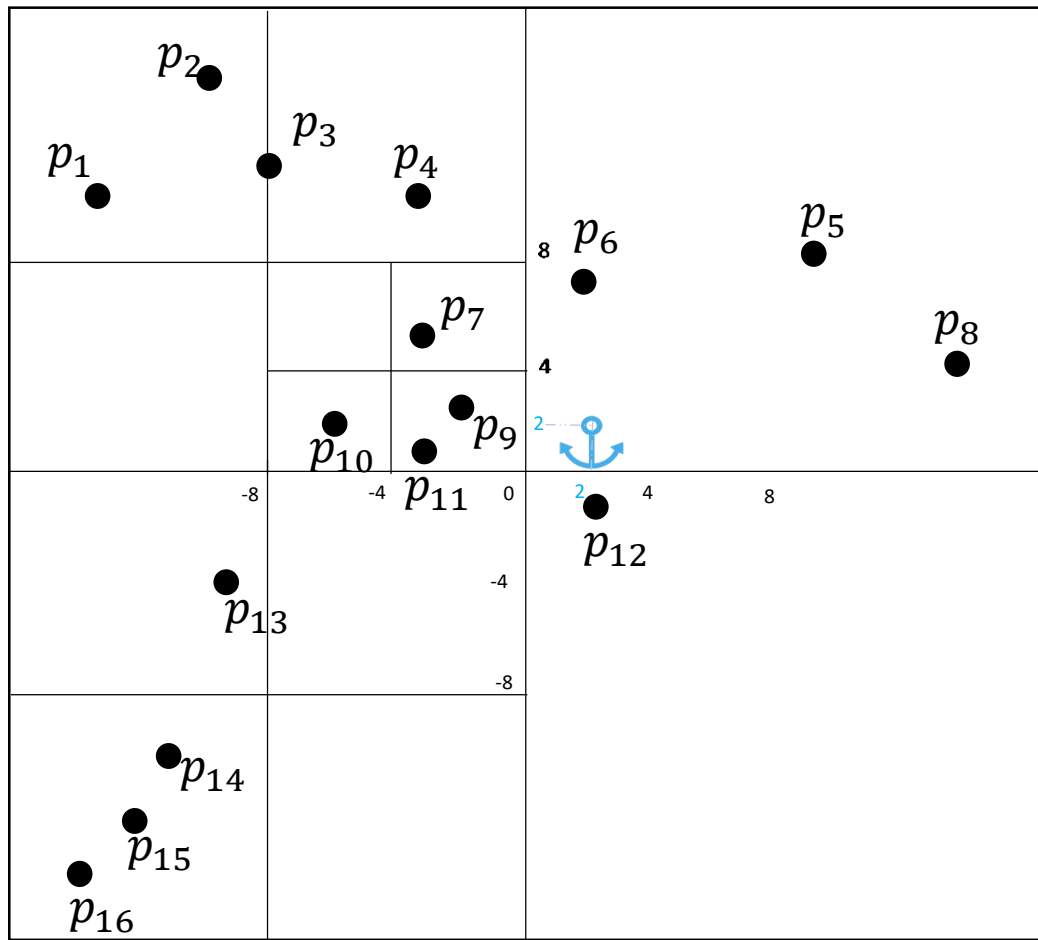


- White node **and** beyond range: impossible to contribute to solution set!
- Backtrack....
- Done! 😊 **Solution set:**  $\{p_6, p_8, p_7, p_9\}$

# Example of $k$ -NN query in PR-QuadTree

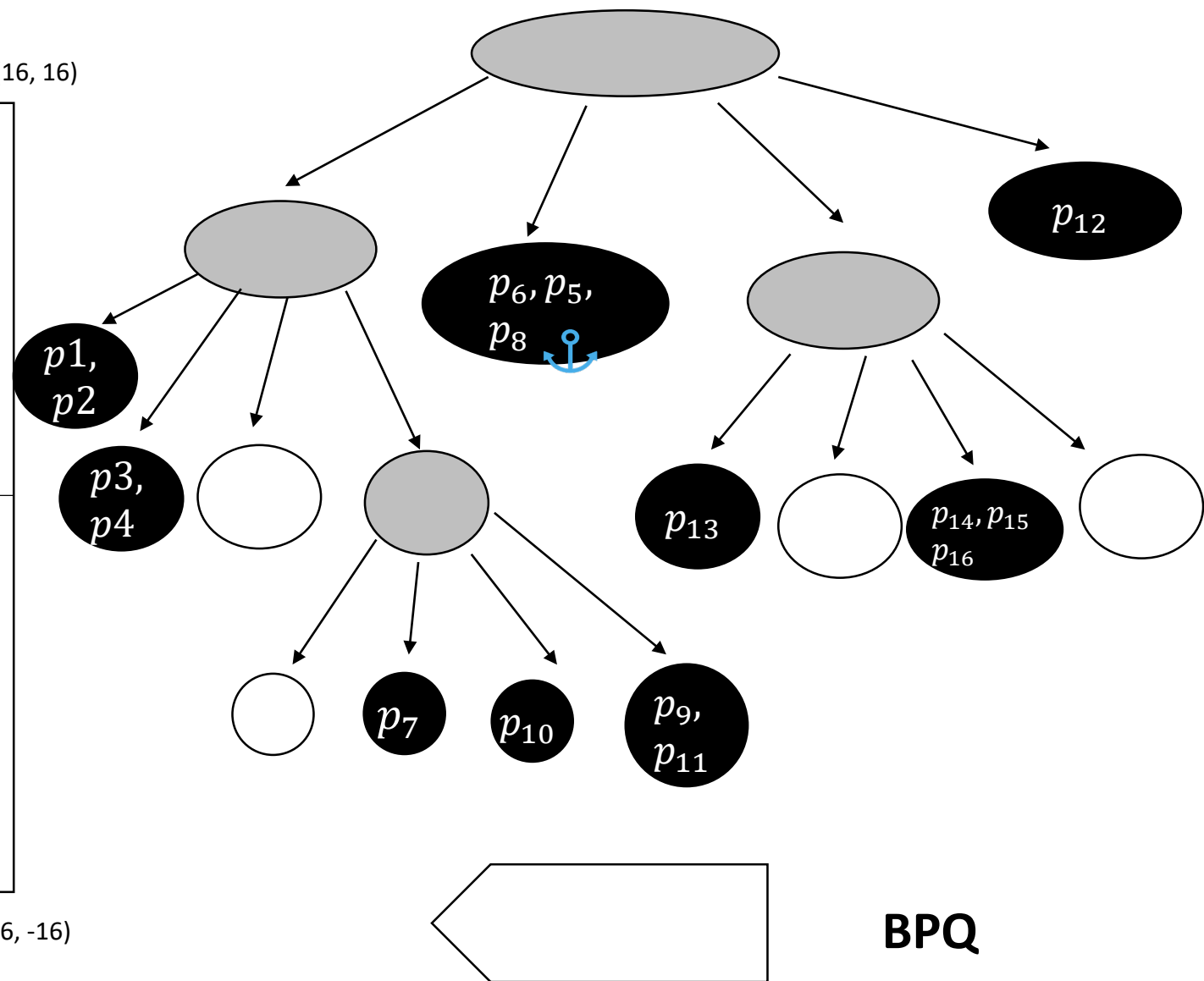
anchor = (2, 2),  $k=2$ , Bucket size  $b = 3$

(-16, 16) (16, 16)



(-16, -16)

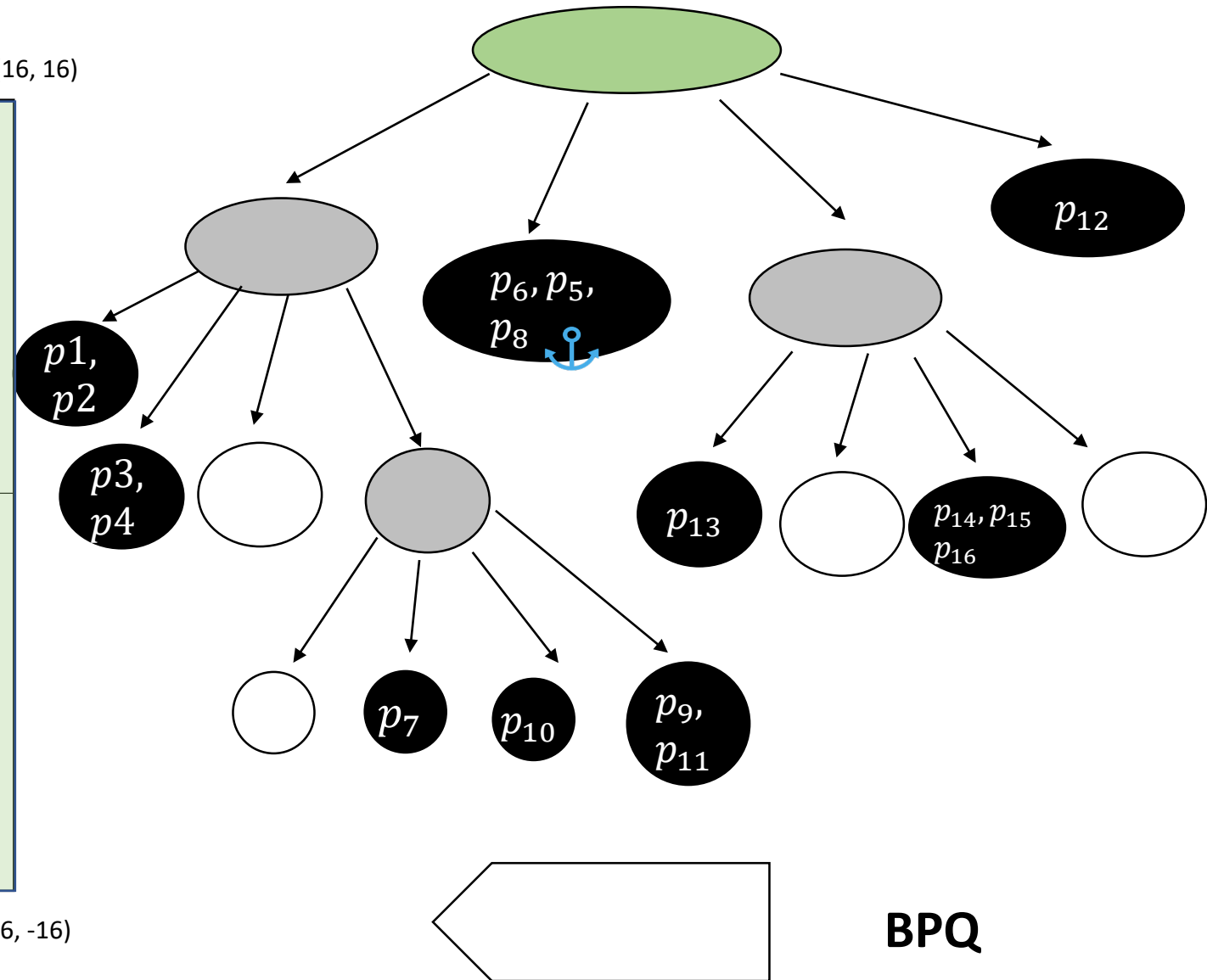
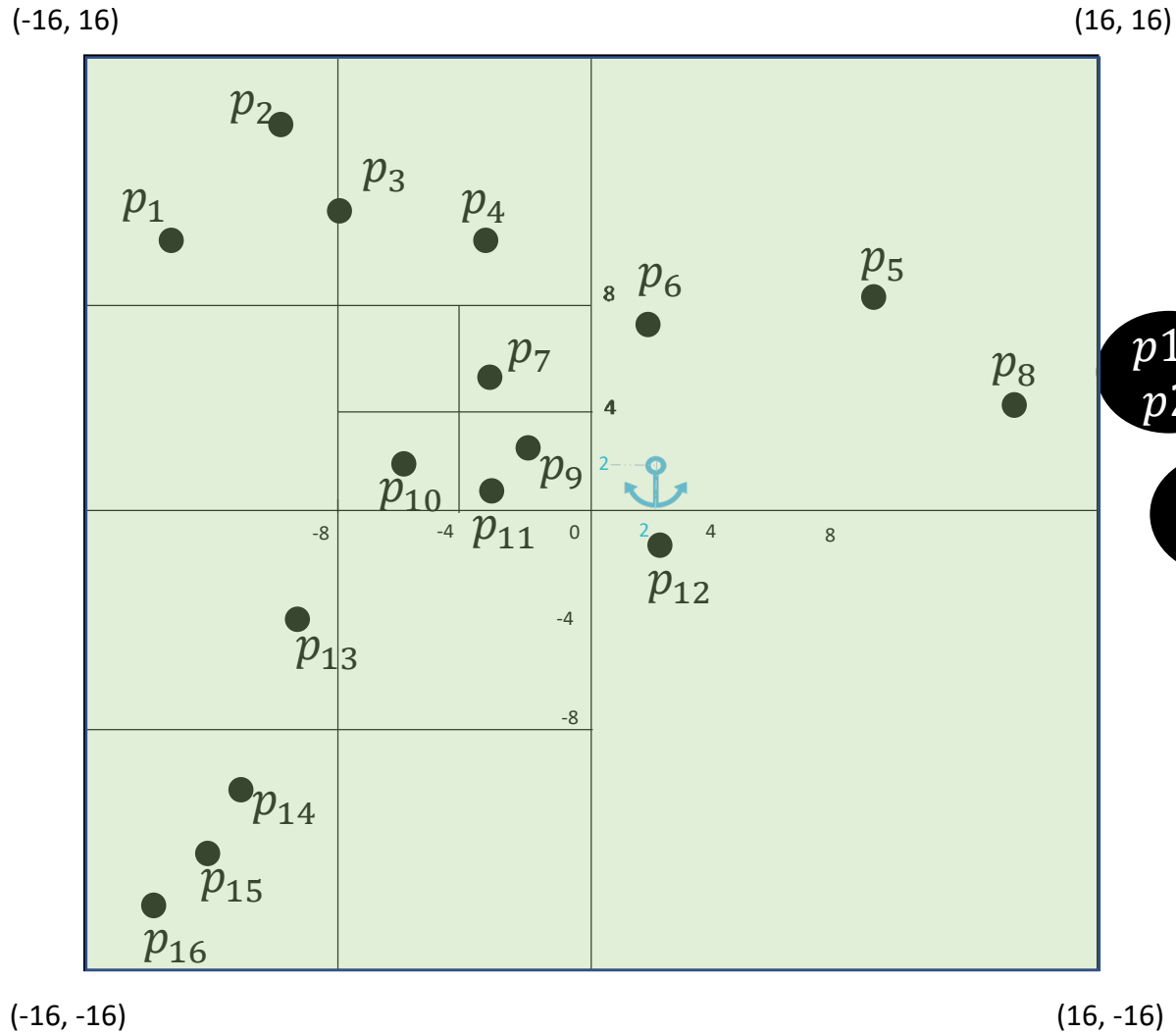
(16, -16)



BPQ

# Example of $k$ -NN query in PR-QuadTree

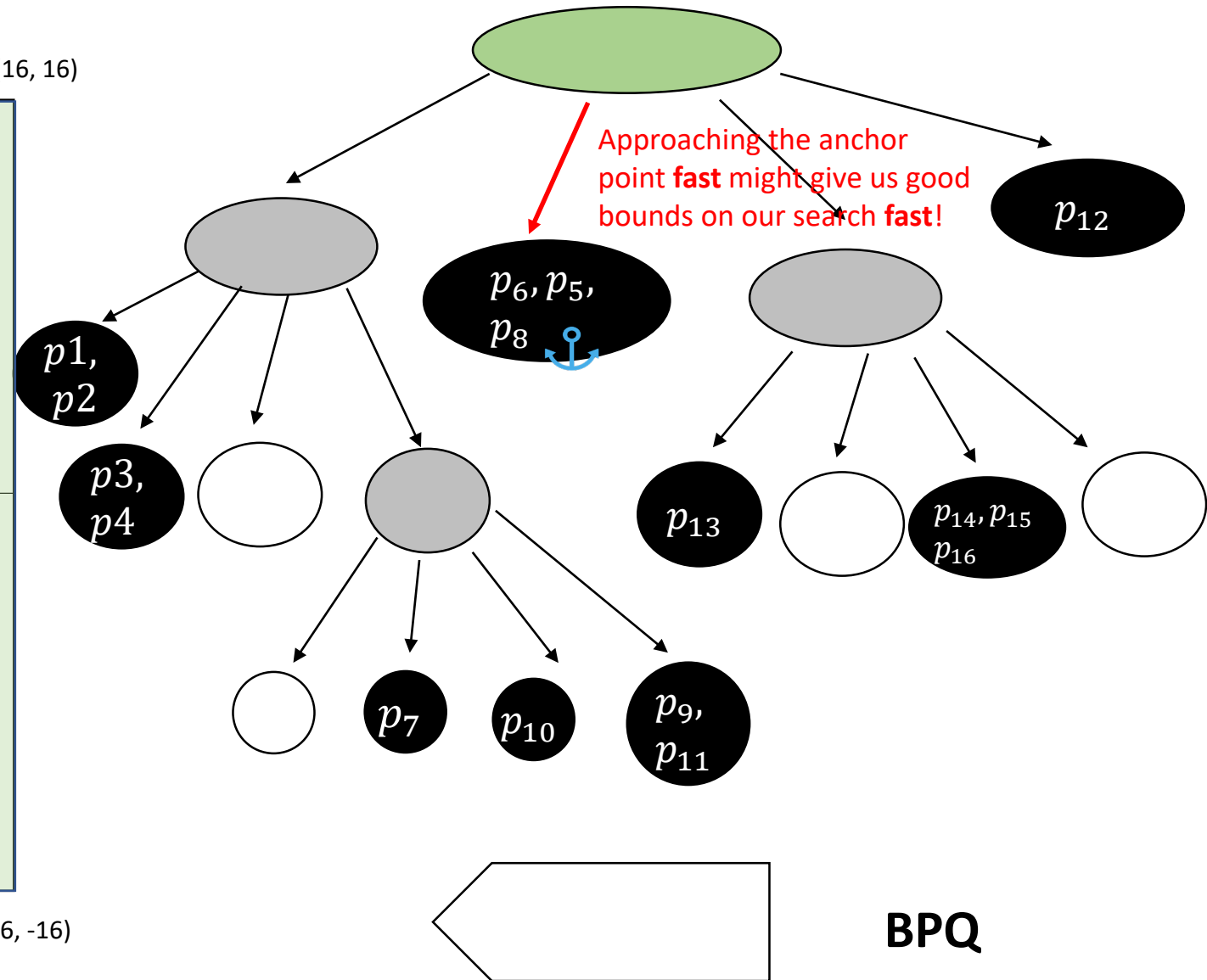
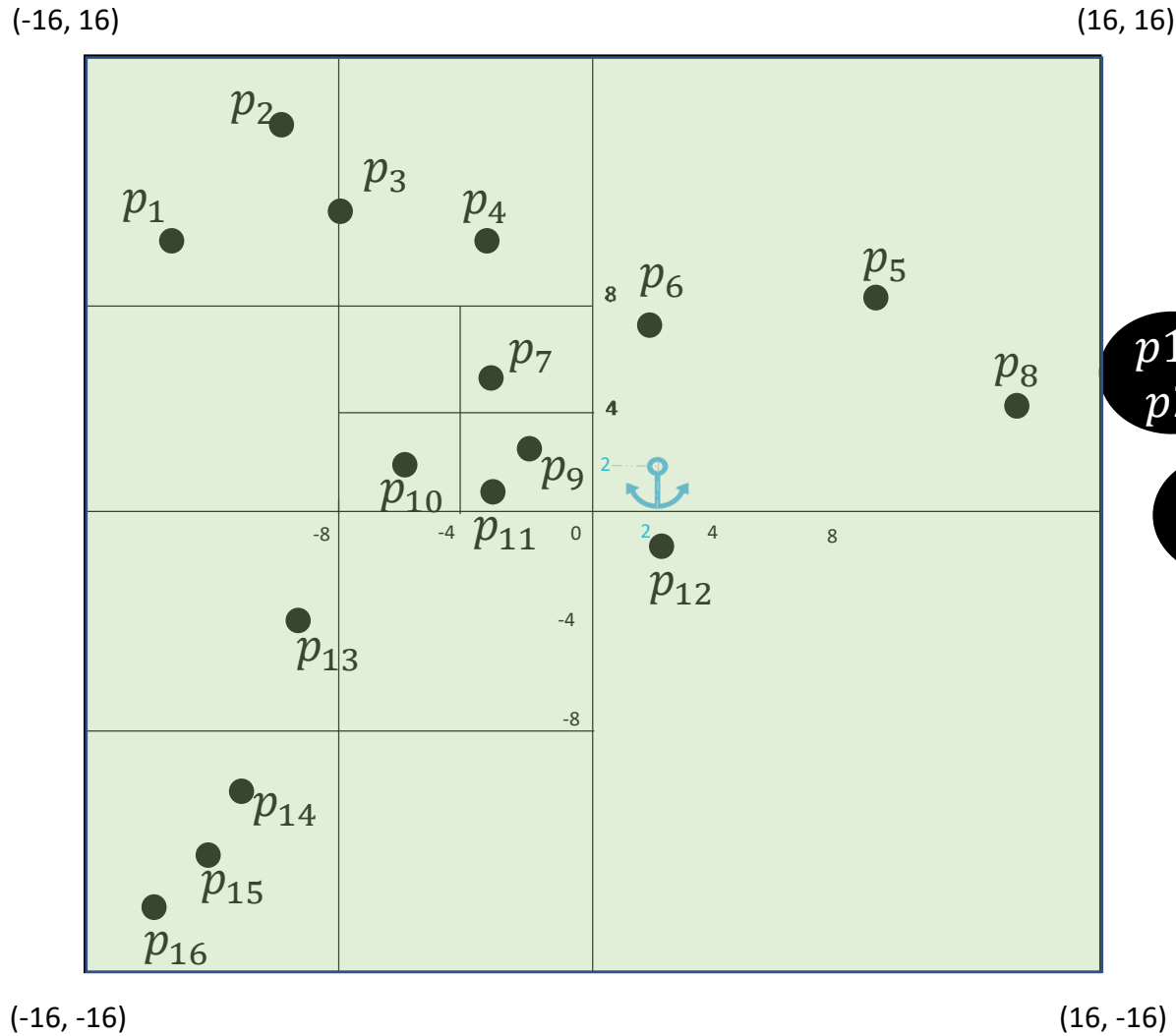
anchor = (2, 2),  $k=2$ , Bucket size  $b = 3$





# Example of $k$ -NN query in PR-QuadTree

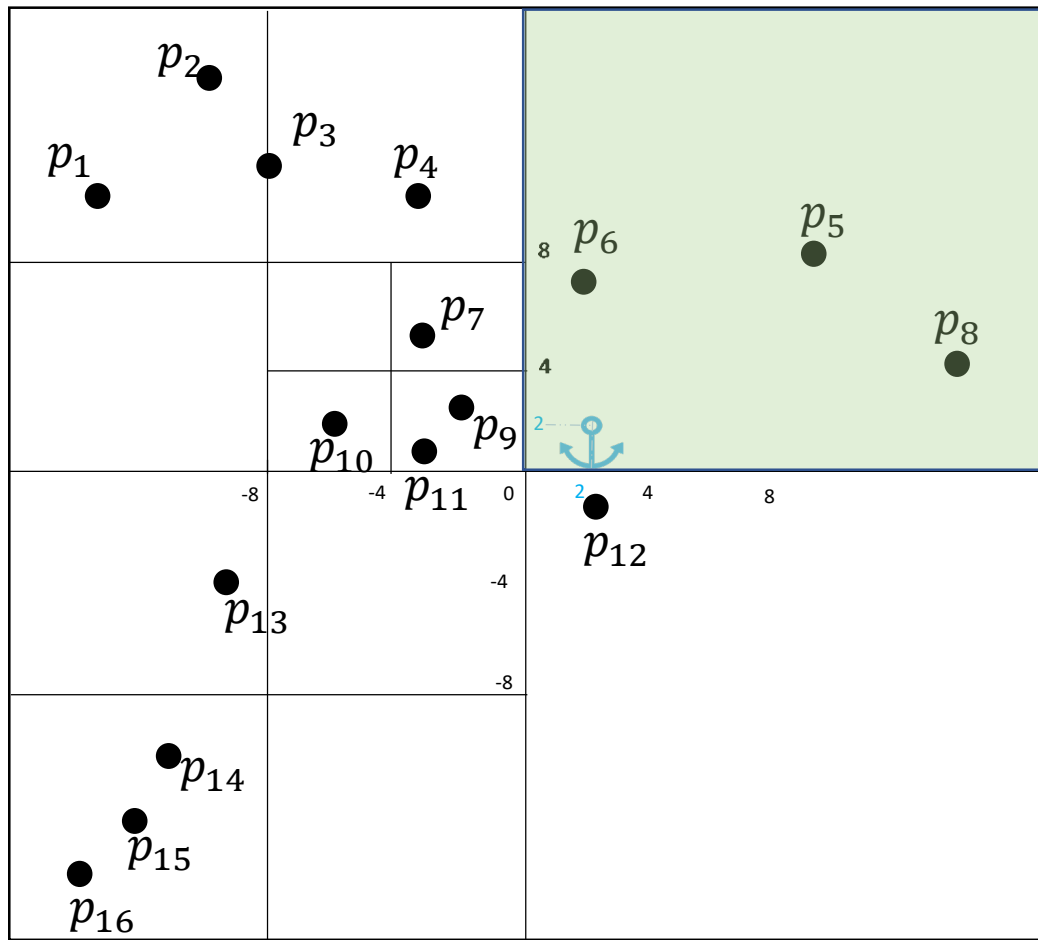
anchor = (2, 2),  $k=2$ , Bucket size  $b = 3$



# Example of $k$ -NN query in PR-QuadTree

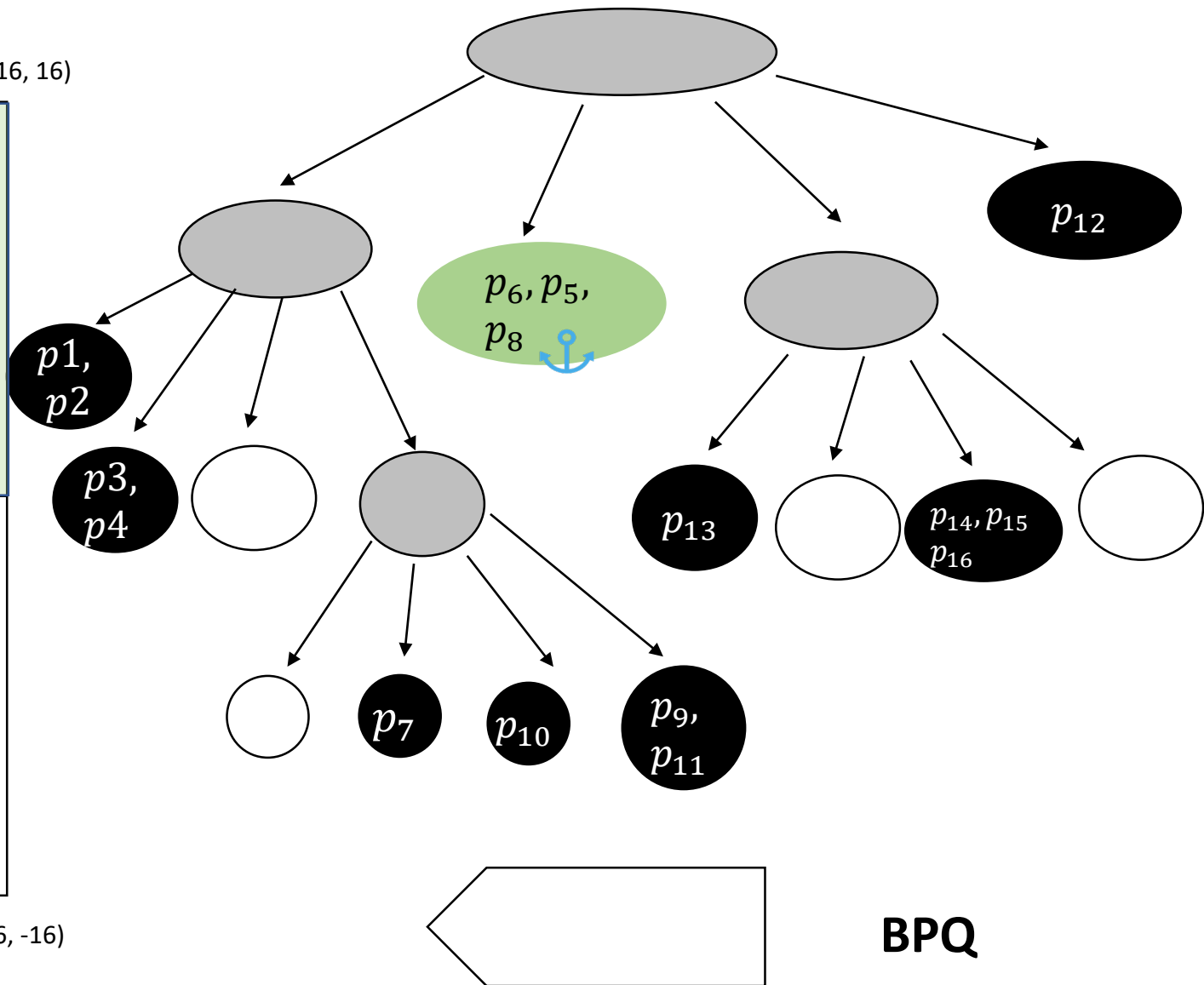
anchor = (2, 2),  $k=2$ , Bucket size  $b = 3$

(-16, 16) (16, 16)



(-16, -16)

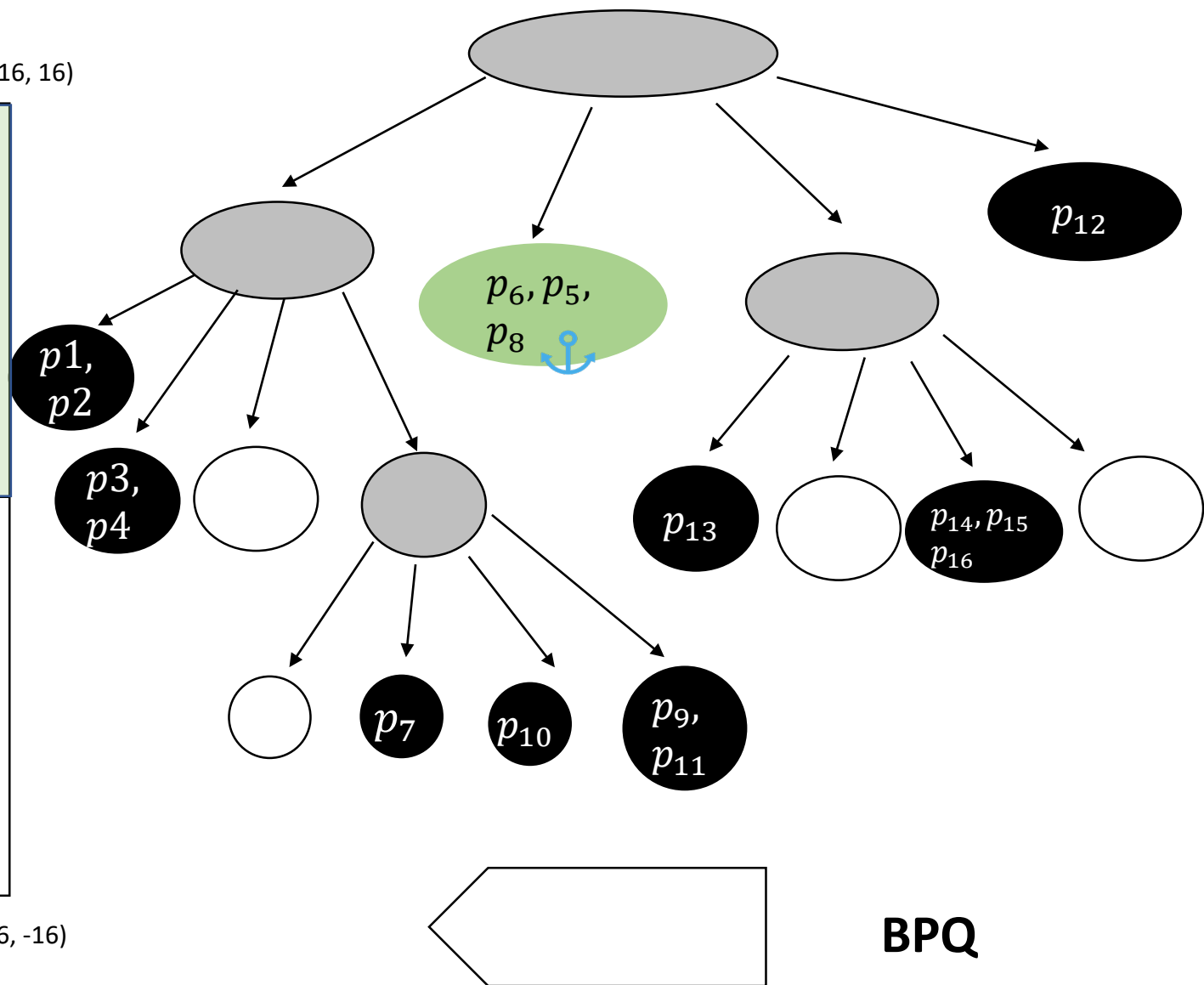
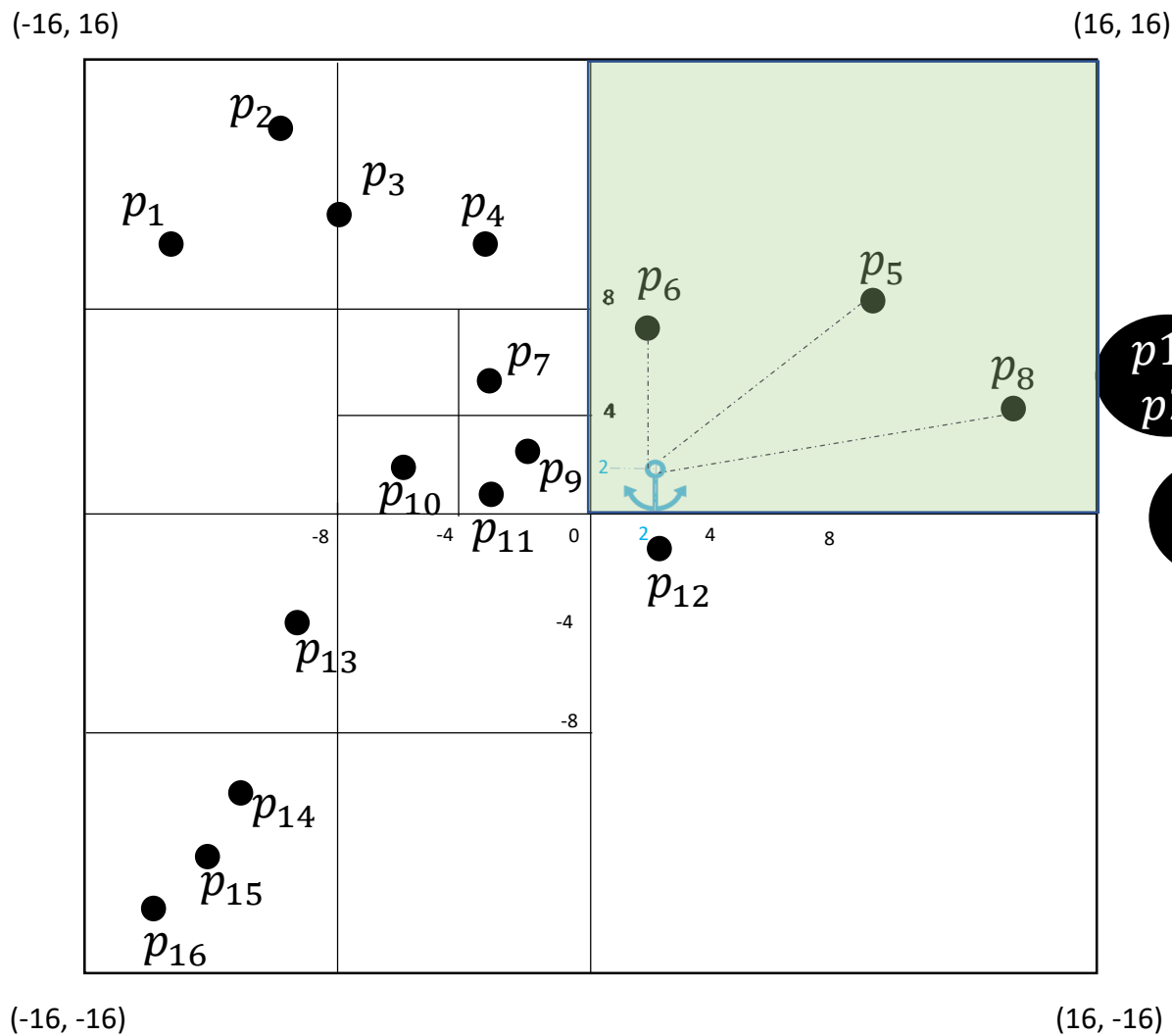
(16, -16)



BPQ

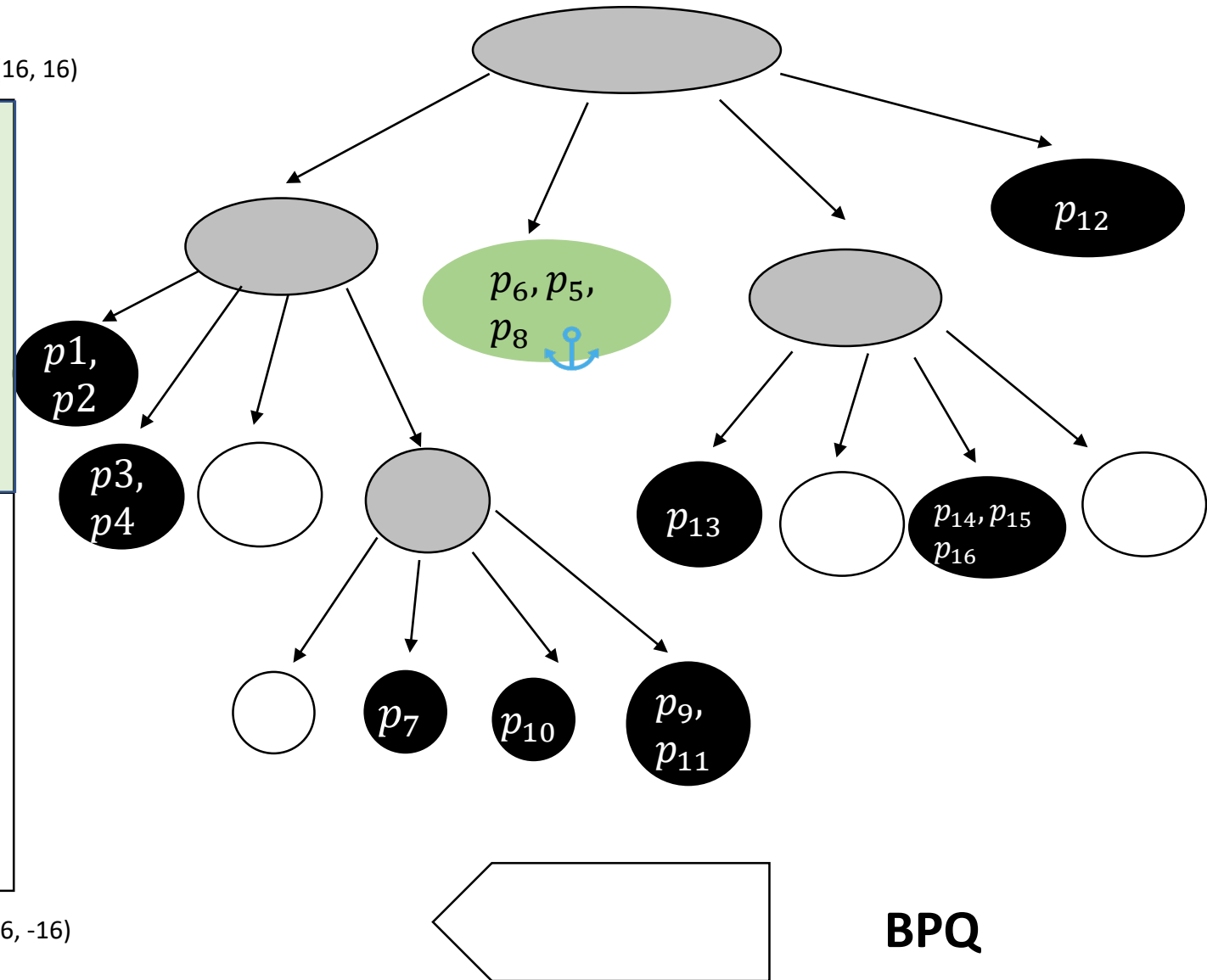
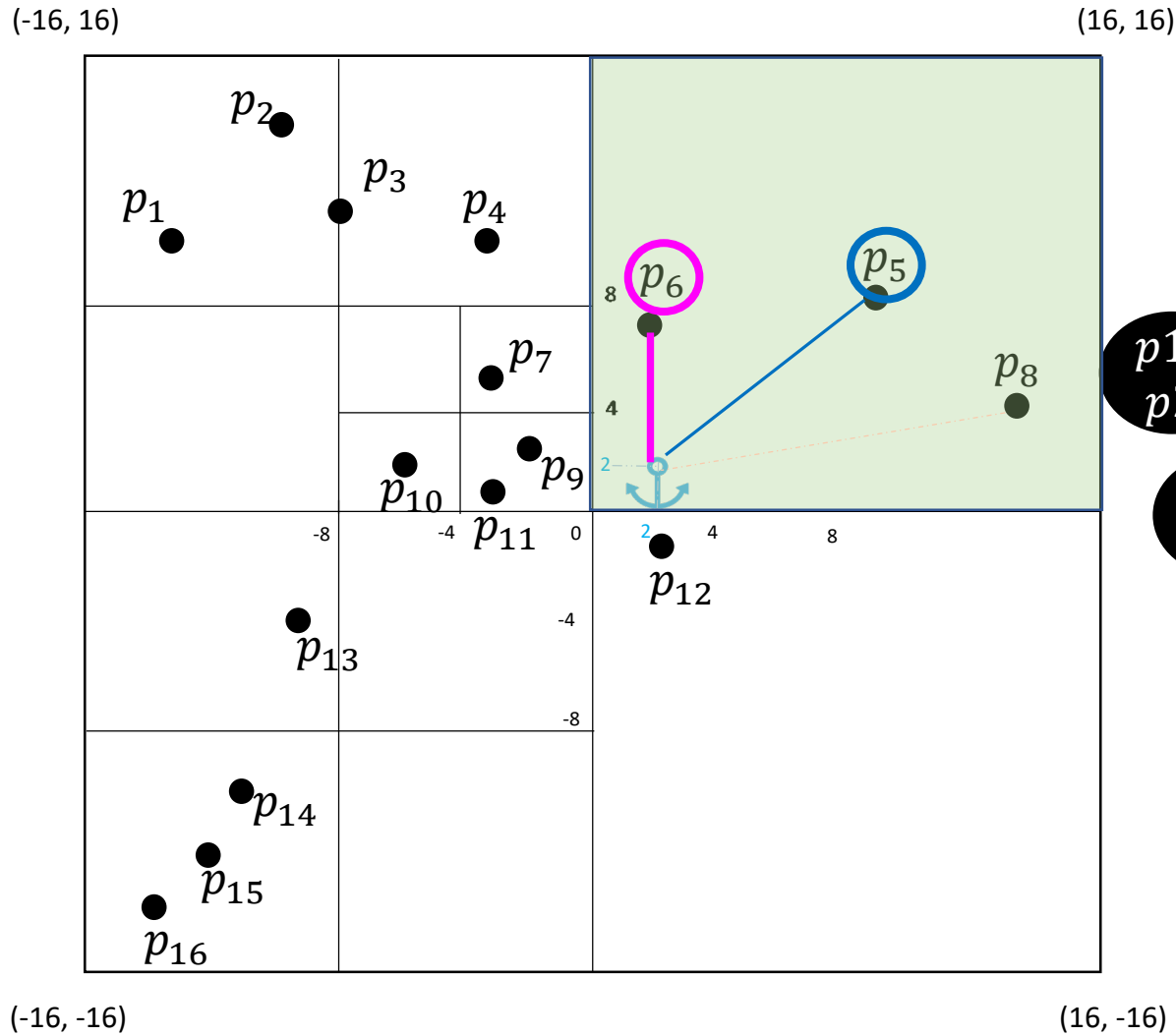
# Example of $k$ -NN query in PR-QuadTree

anchor = (2, 2),  $k=2$ , Bucket size  $b = 3$



# Example of $k$ -NN query in PR-QuadTree

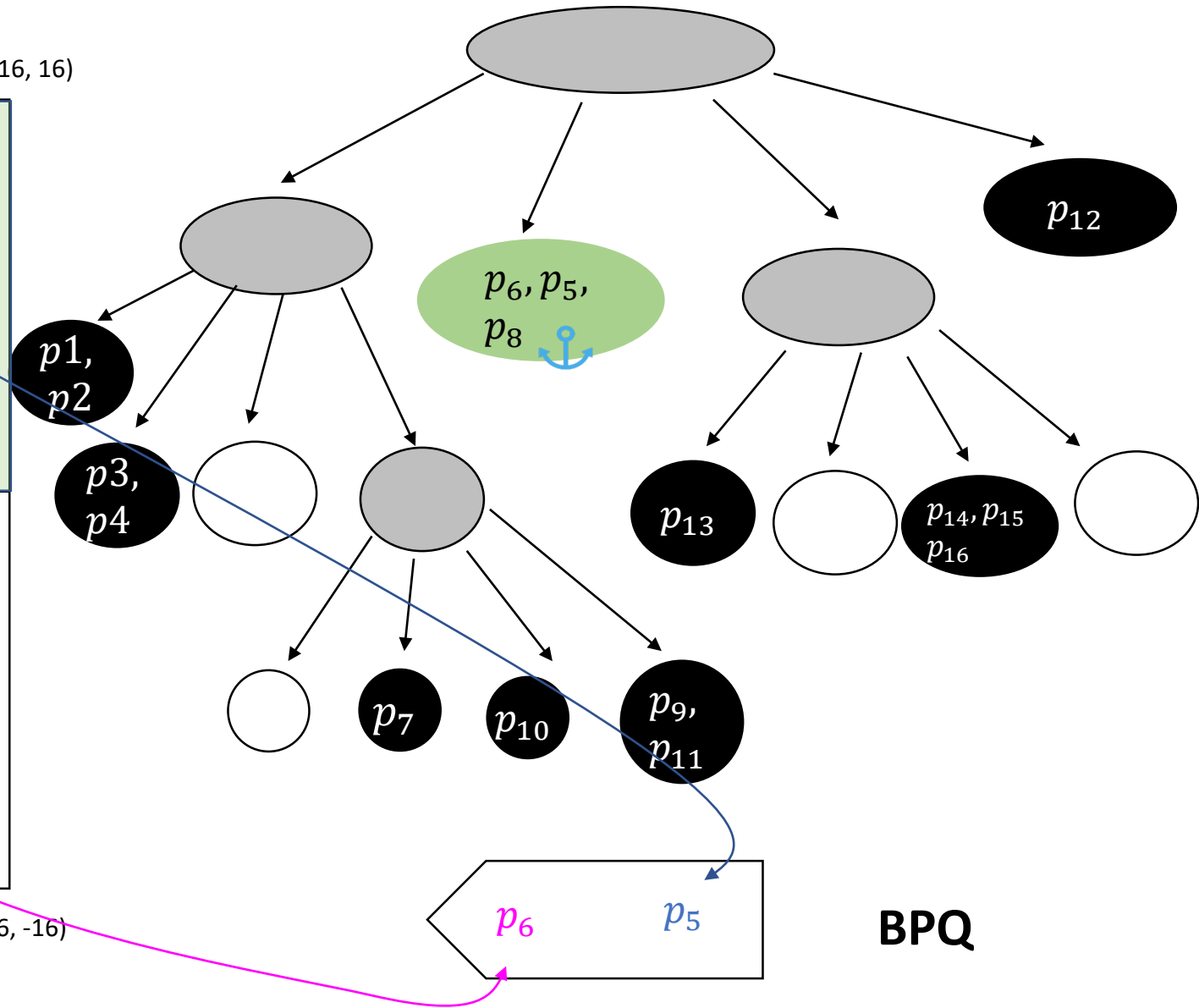
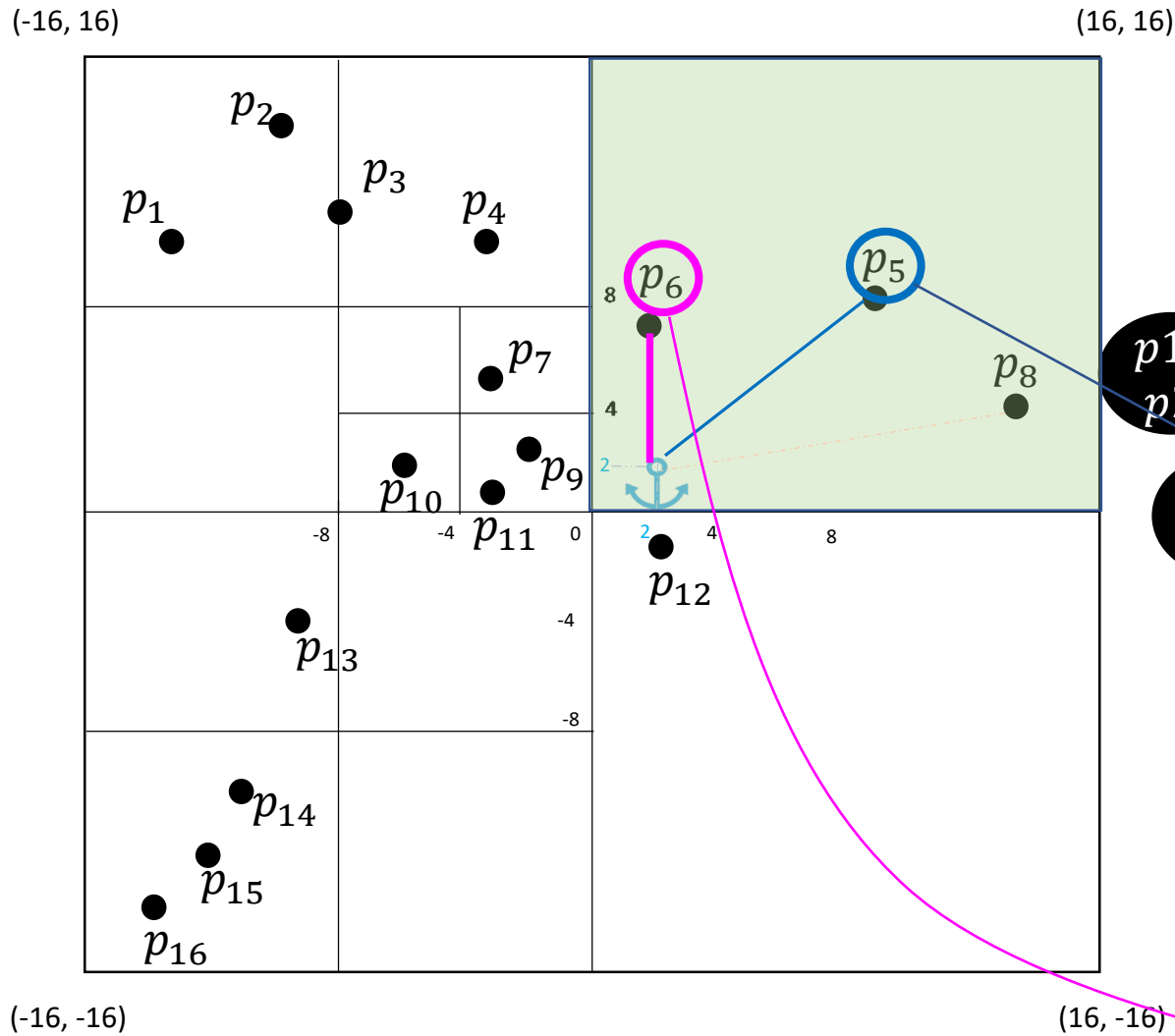
anchor = (2, 2),  $k=2$ , Bucket size  $b = 3$



BPQ

# Example of $k$ -NN query in PR-QuadTree

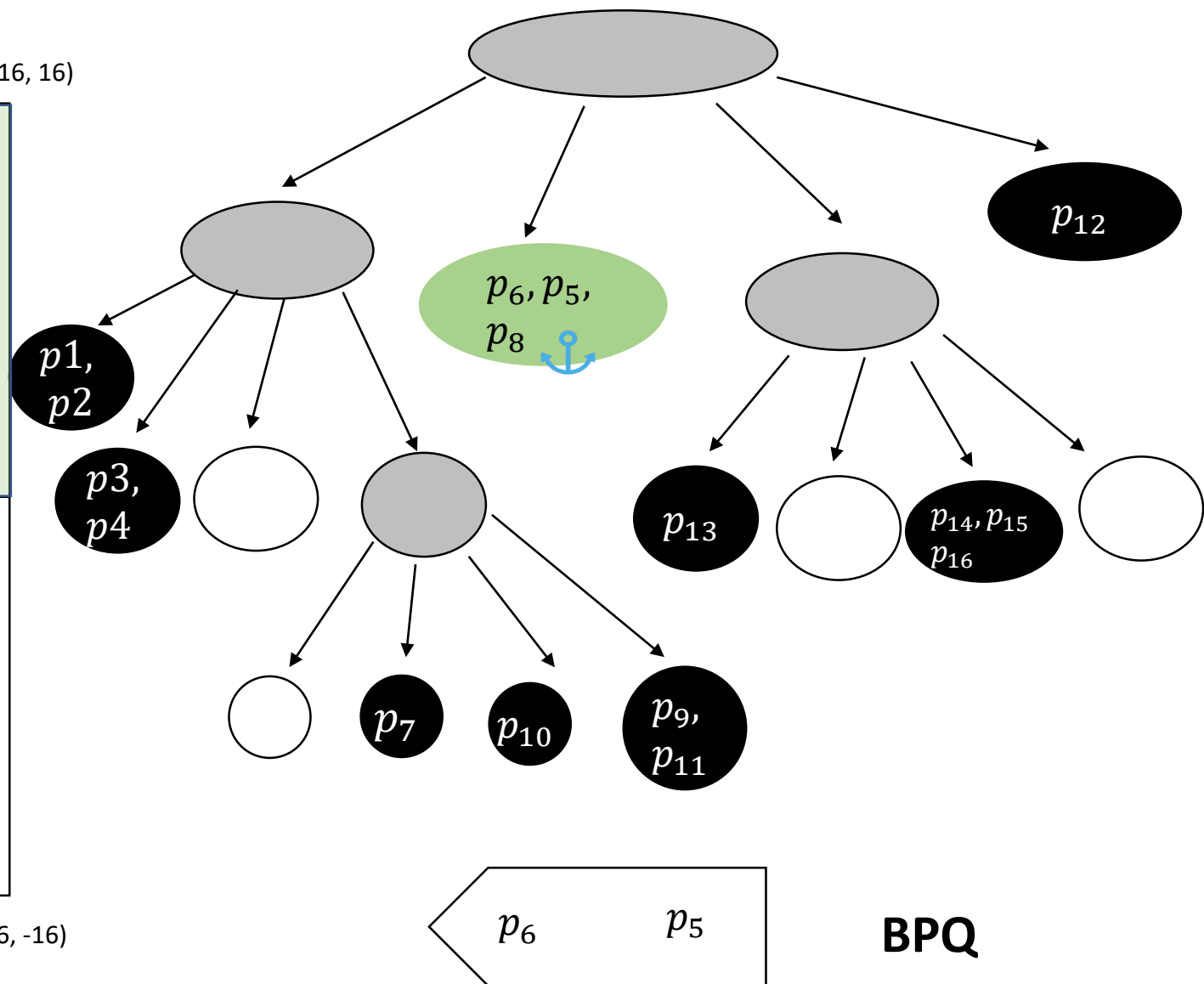
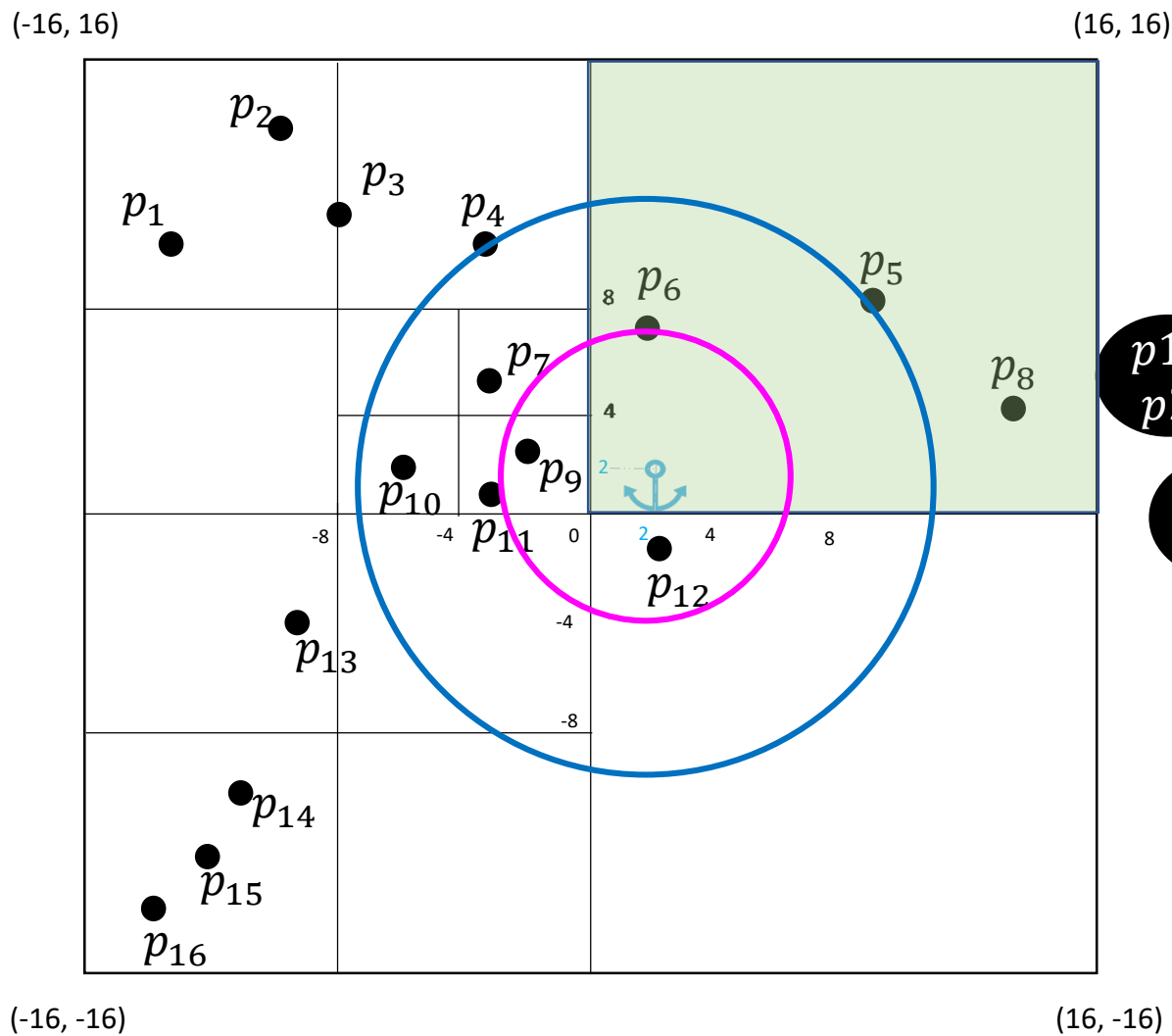
anchor = (2, 2),  $k=2$ , Bucket size  $b = 3$



BPQ

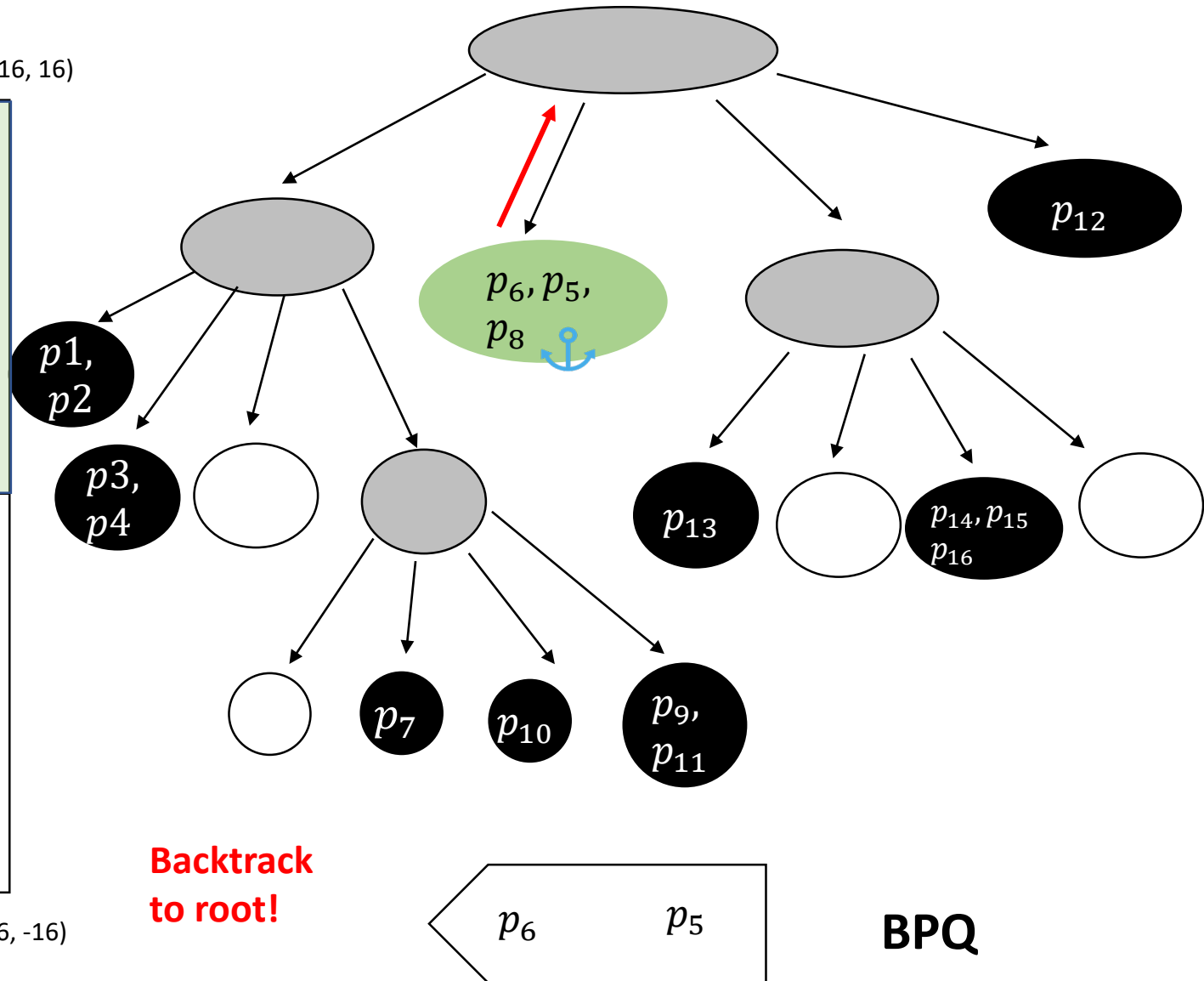
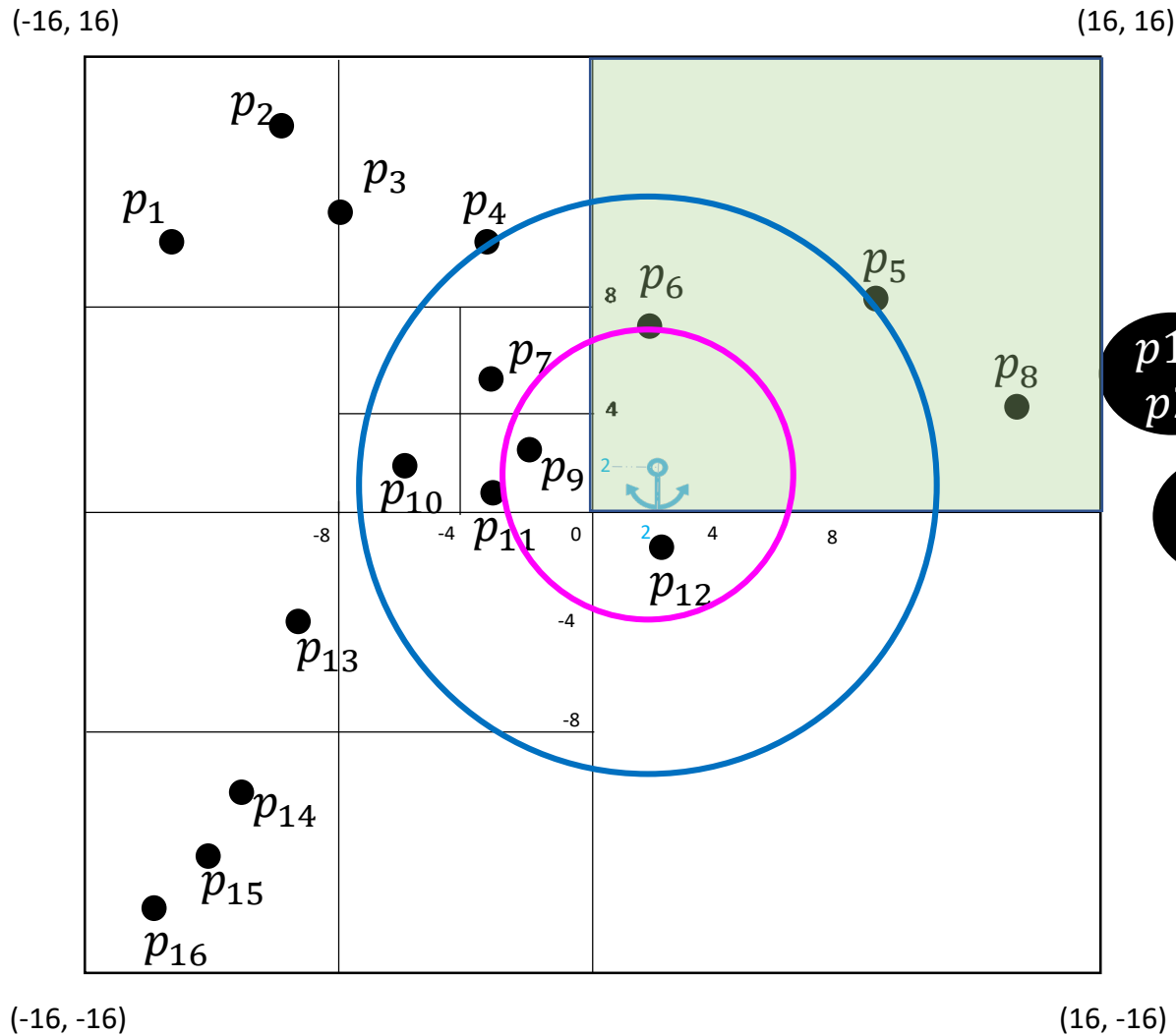
# Example of $k$ -NN query in PR-QuadTree

anchor = (2, 2),  $k=2$ , Bucket size  $b = 3$



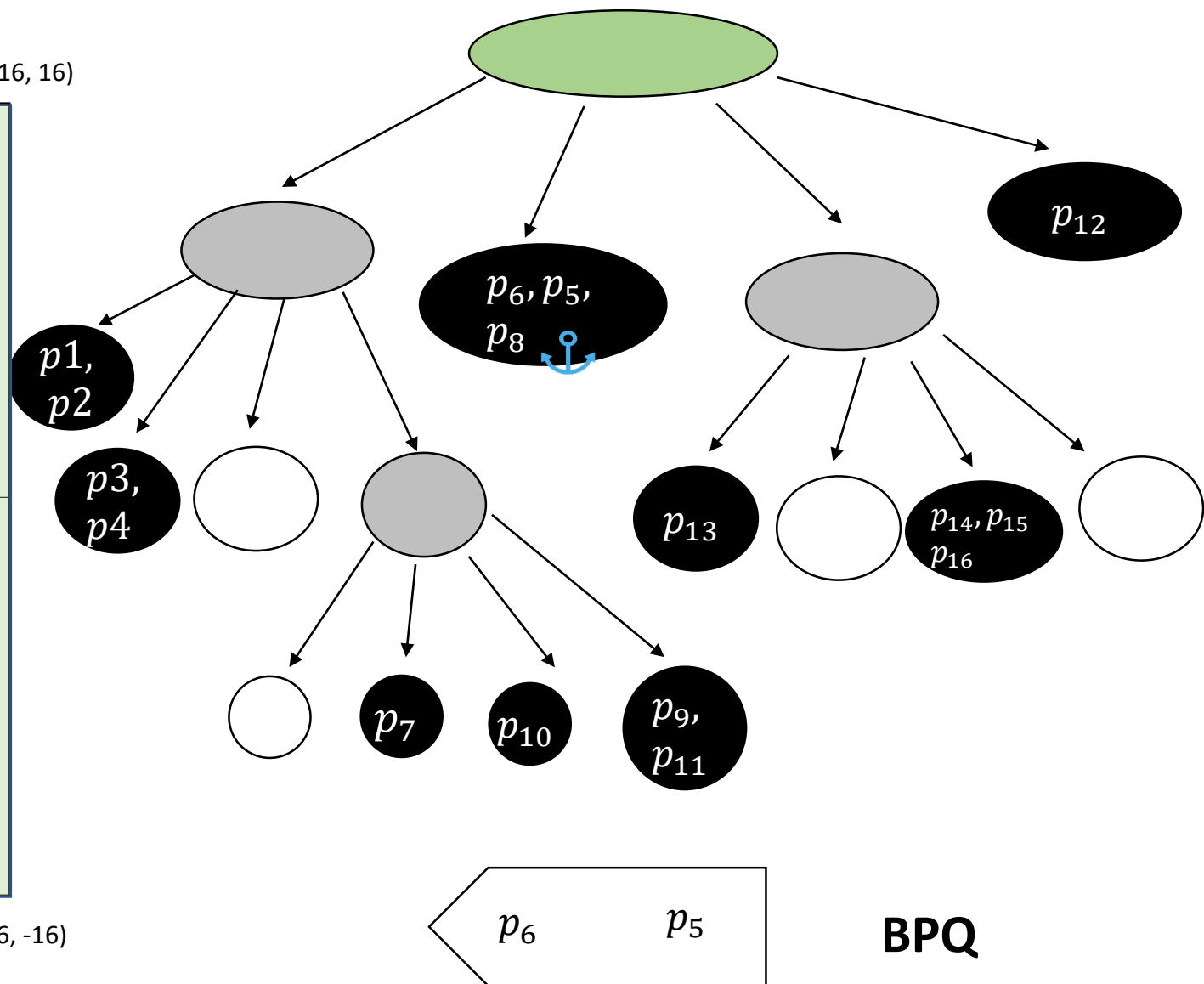
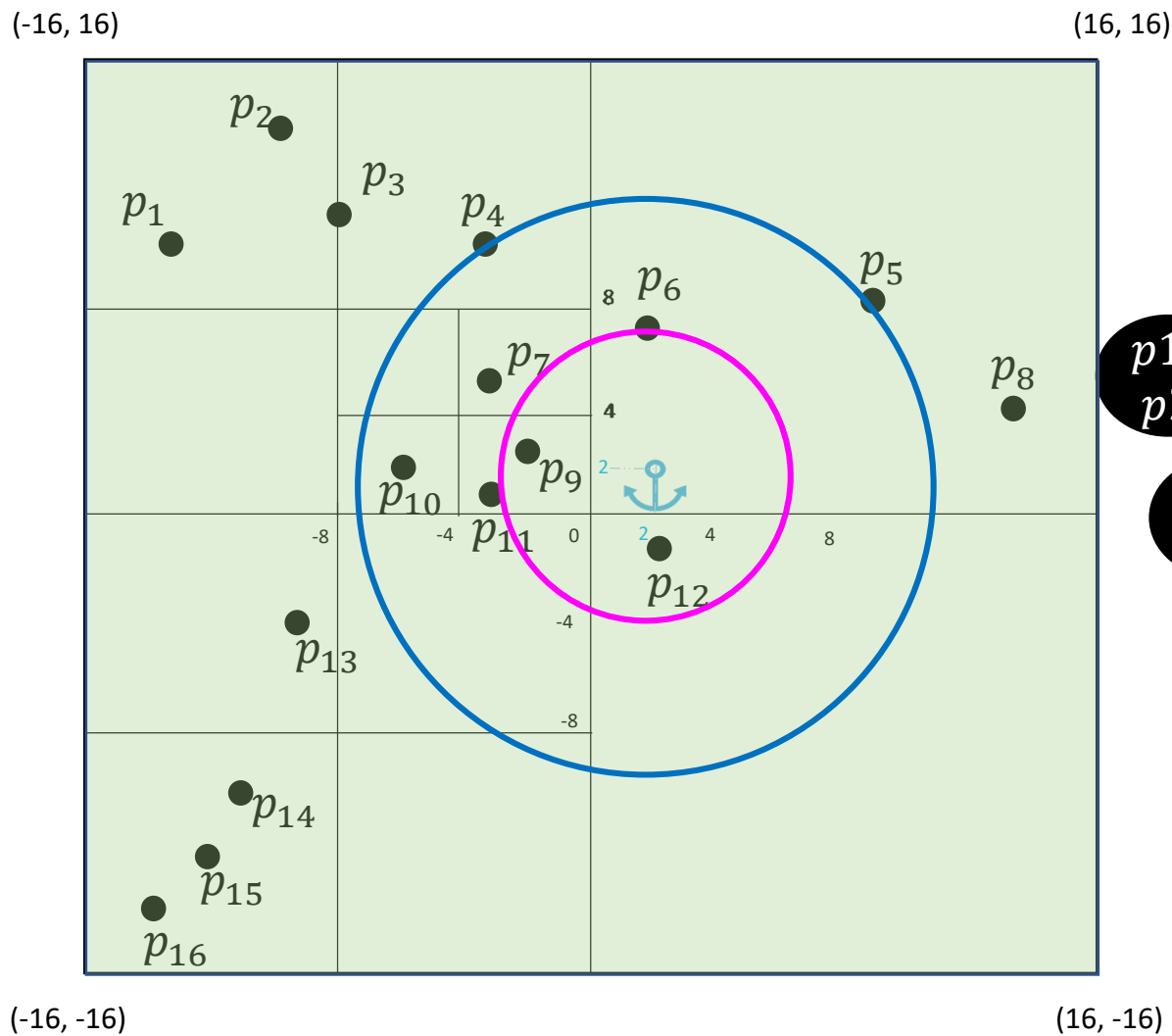
# Example of $k$ -NN query in PR-QuadTree

anchor = (2, 2),  $k=2$ , Bucket size  $b = 3$



# Example of $k$ -NN query in PR-QuadTree

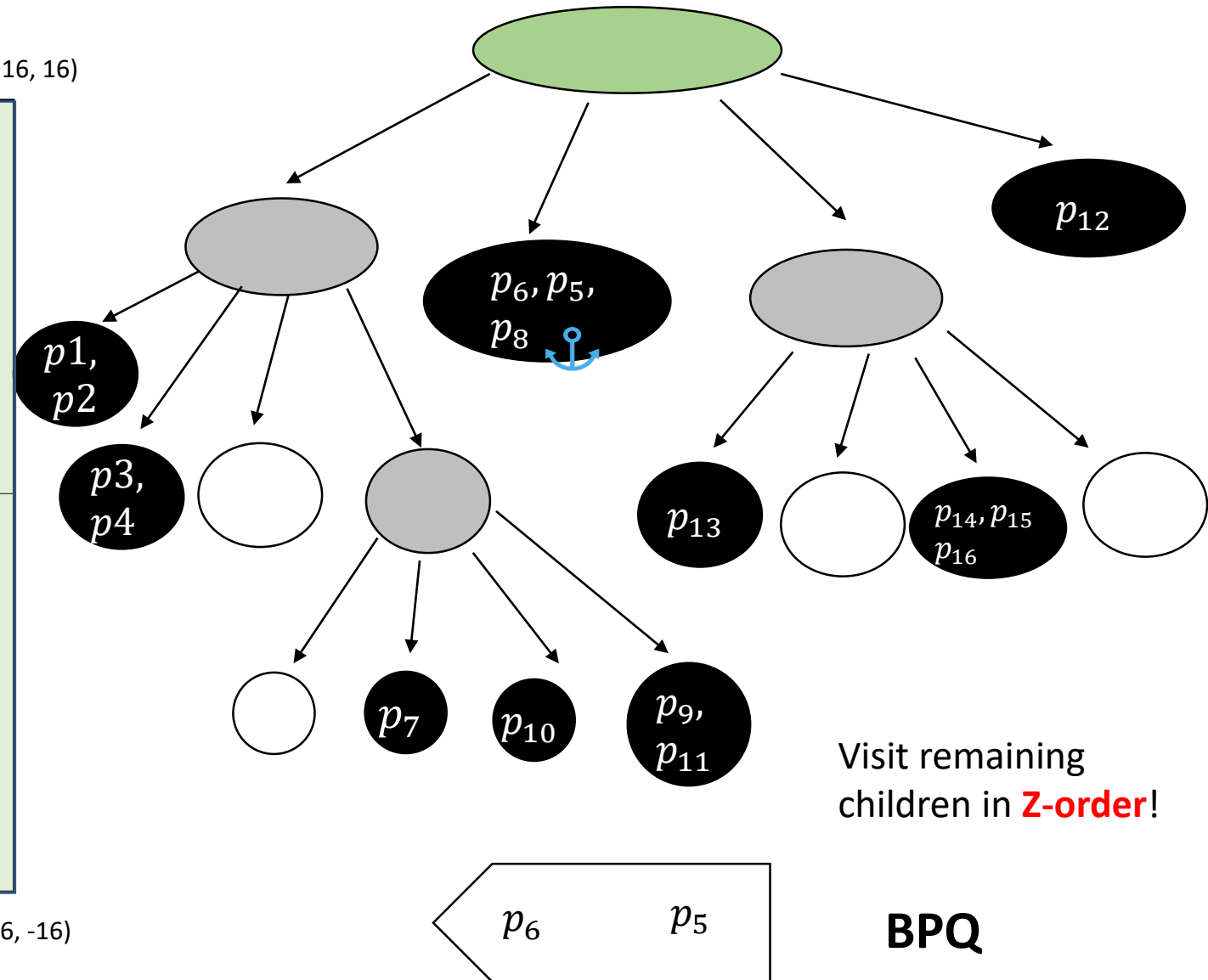
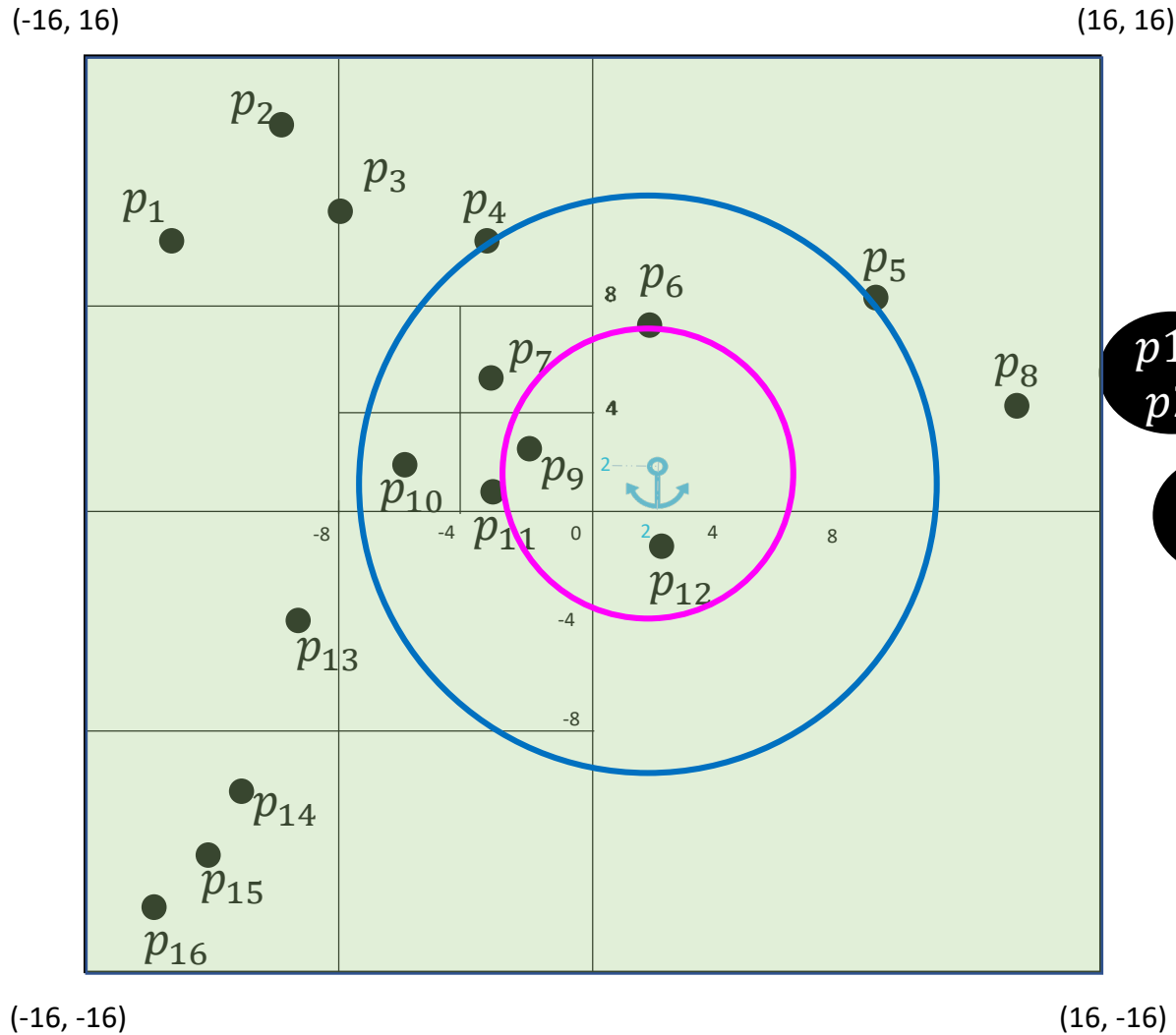
anchor = (2, 2),  $k=2$ , Bucket size  $b = 3$





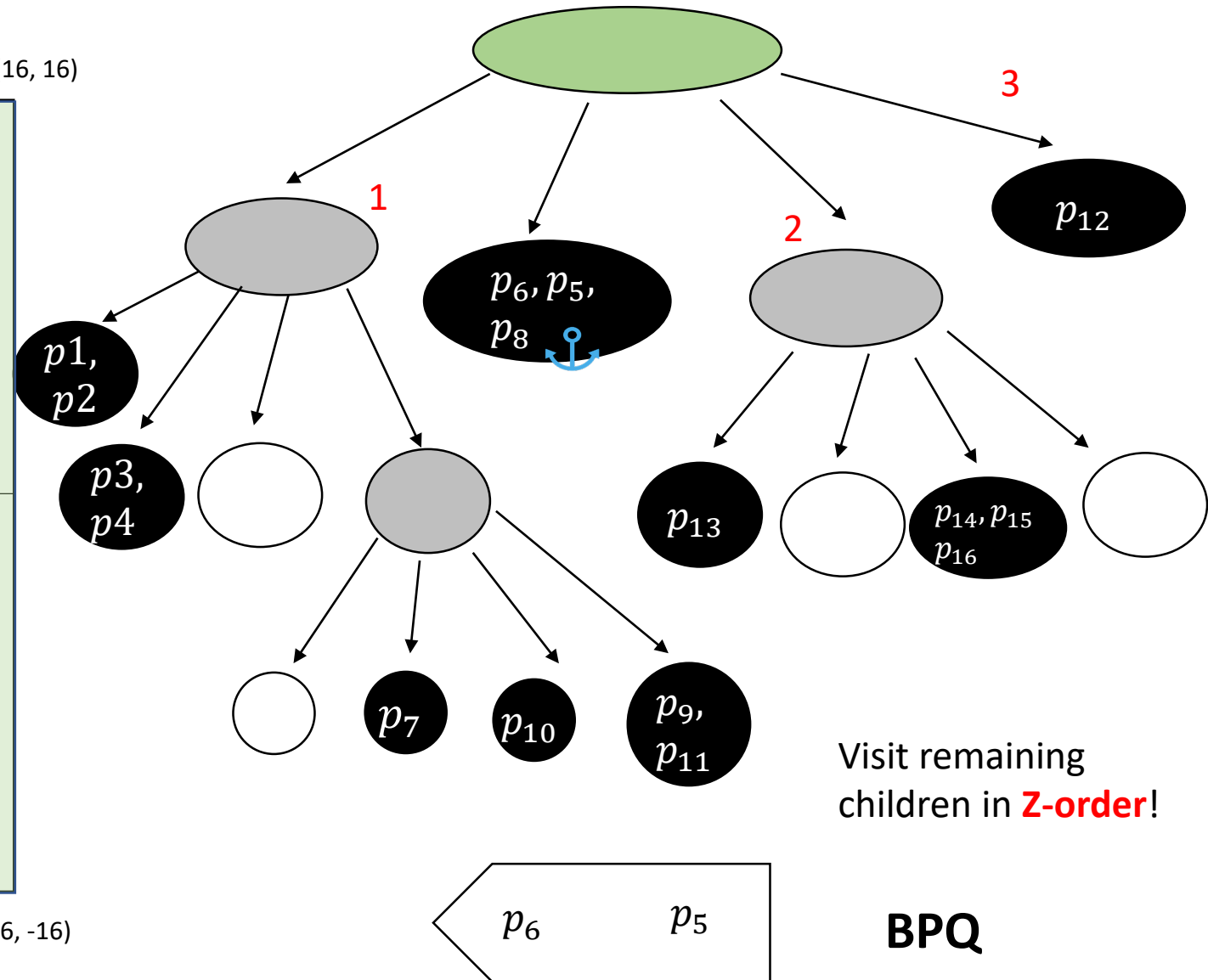
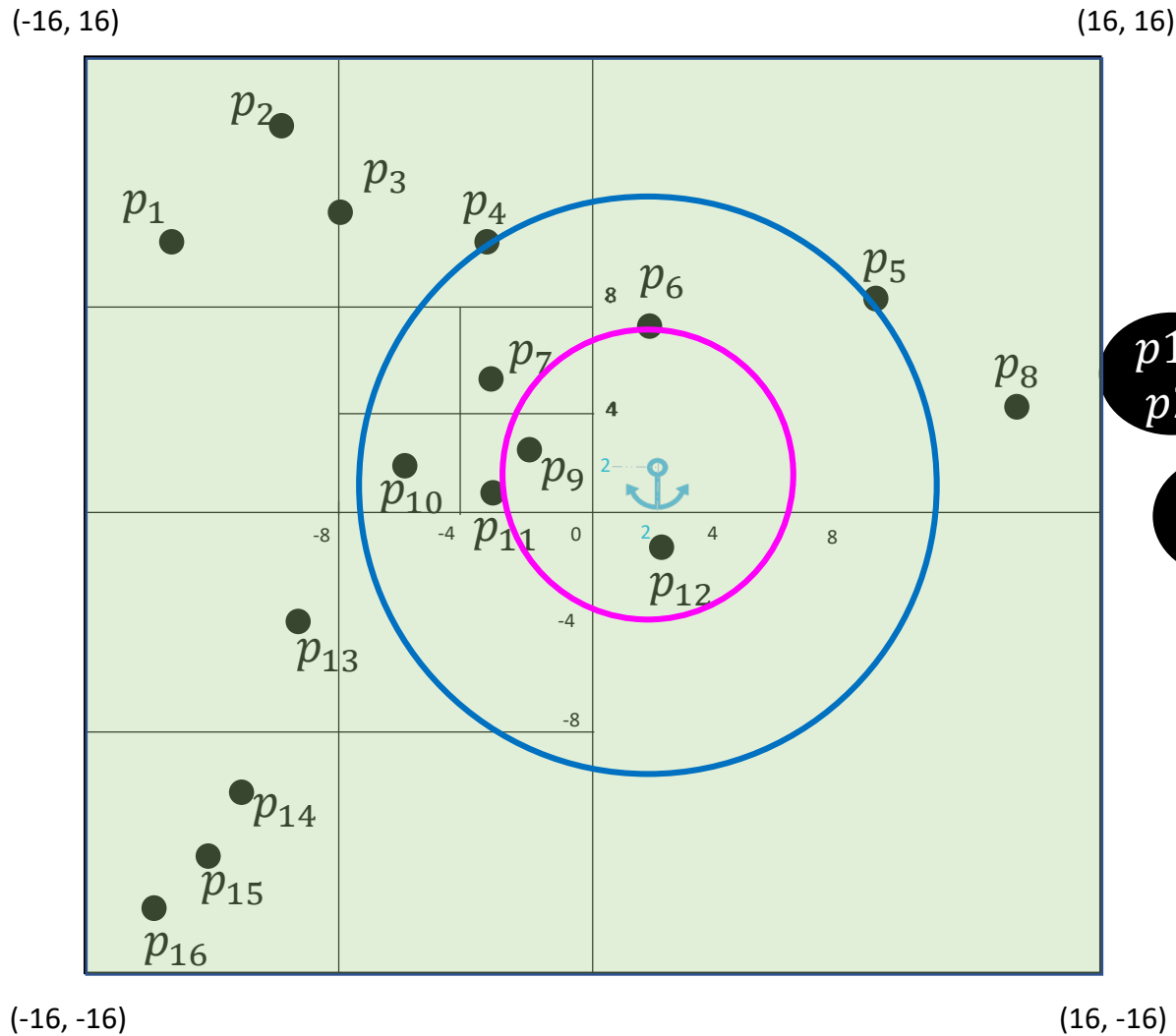
# Example of $k$ -NN query in PR-QuadTree

anchor = (2, 2),  $k=2$ , Bucket size  $b = 3$



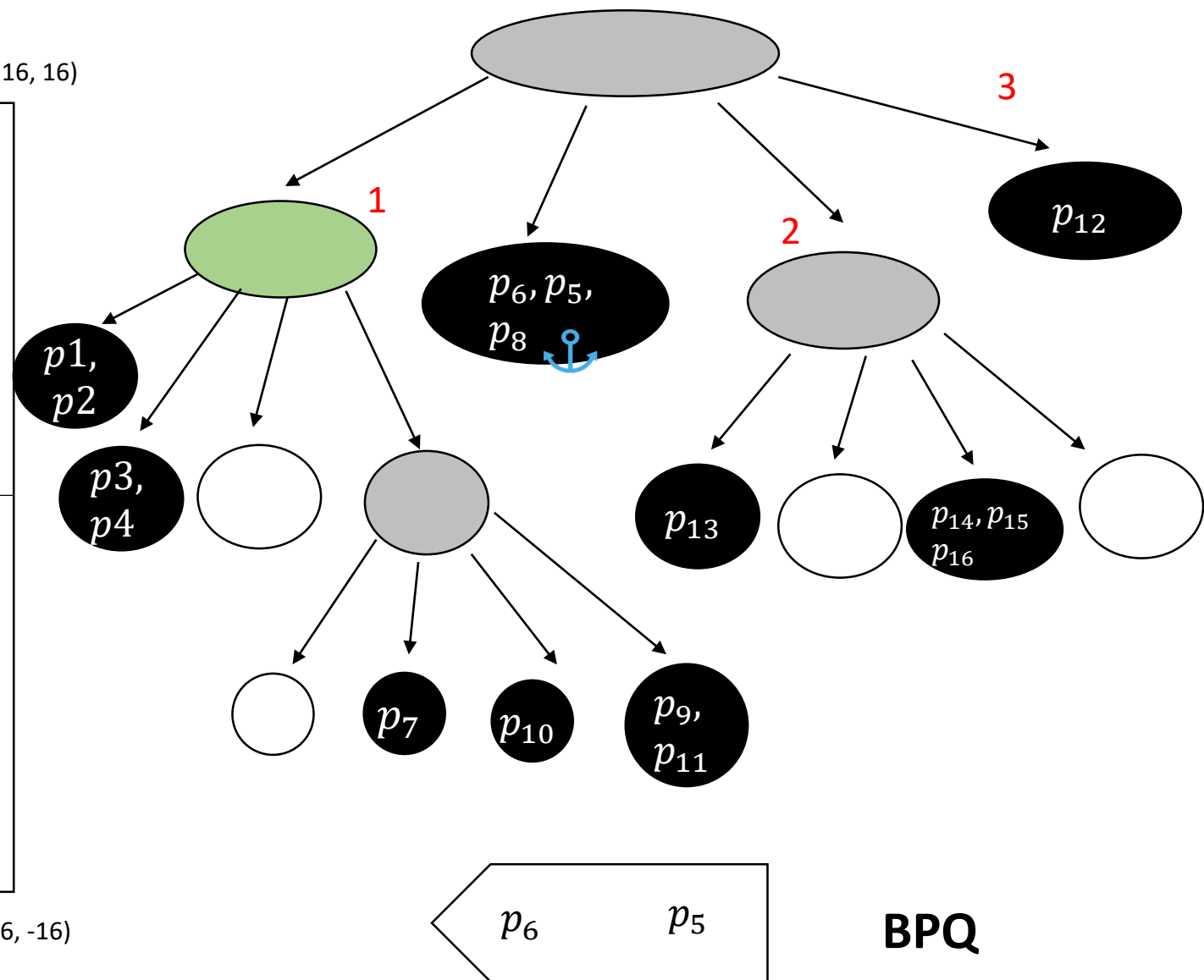
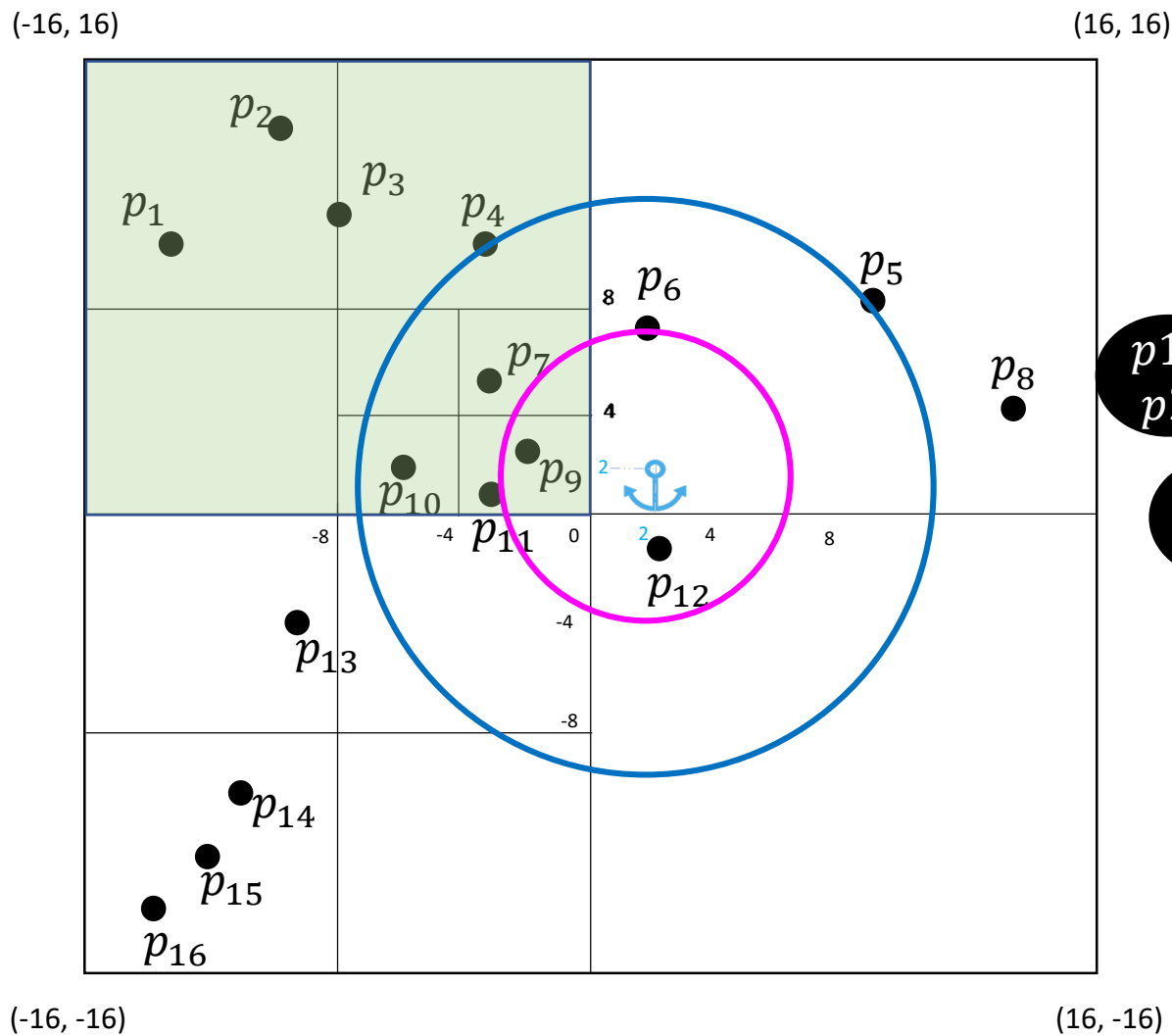
# Example of $k$ -NN query in PR-QuadTree

anchor = (2, 2),  $k=2$ , Bucket size  $b = 3$



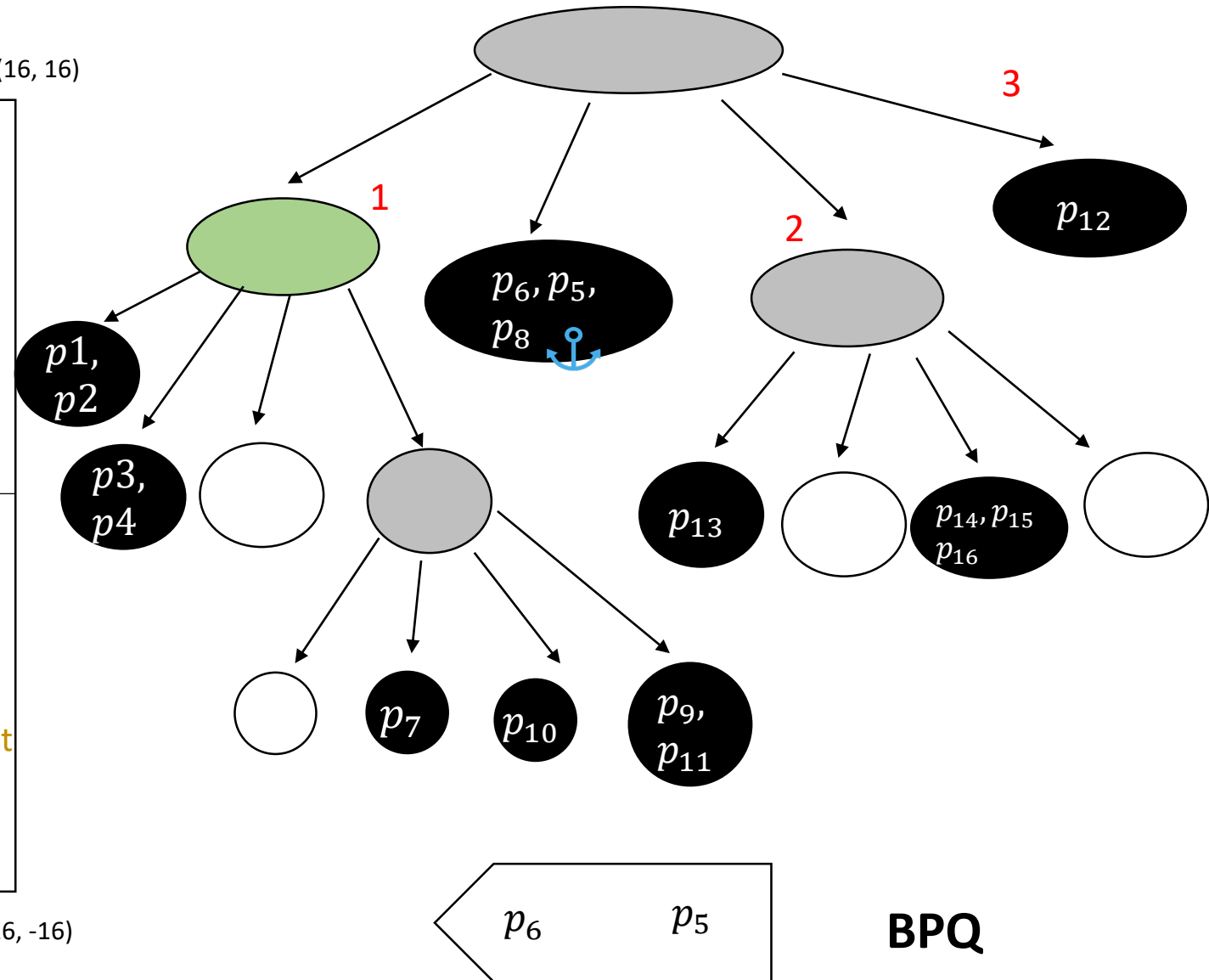
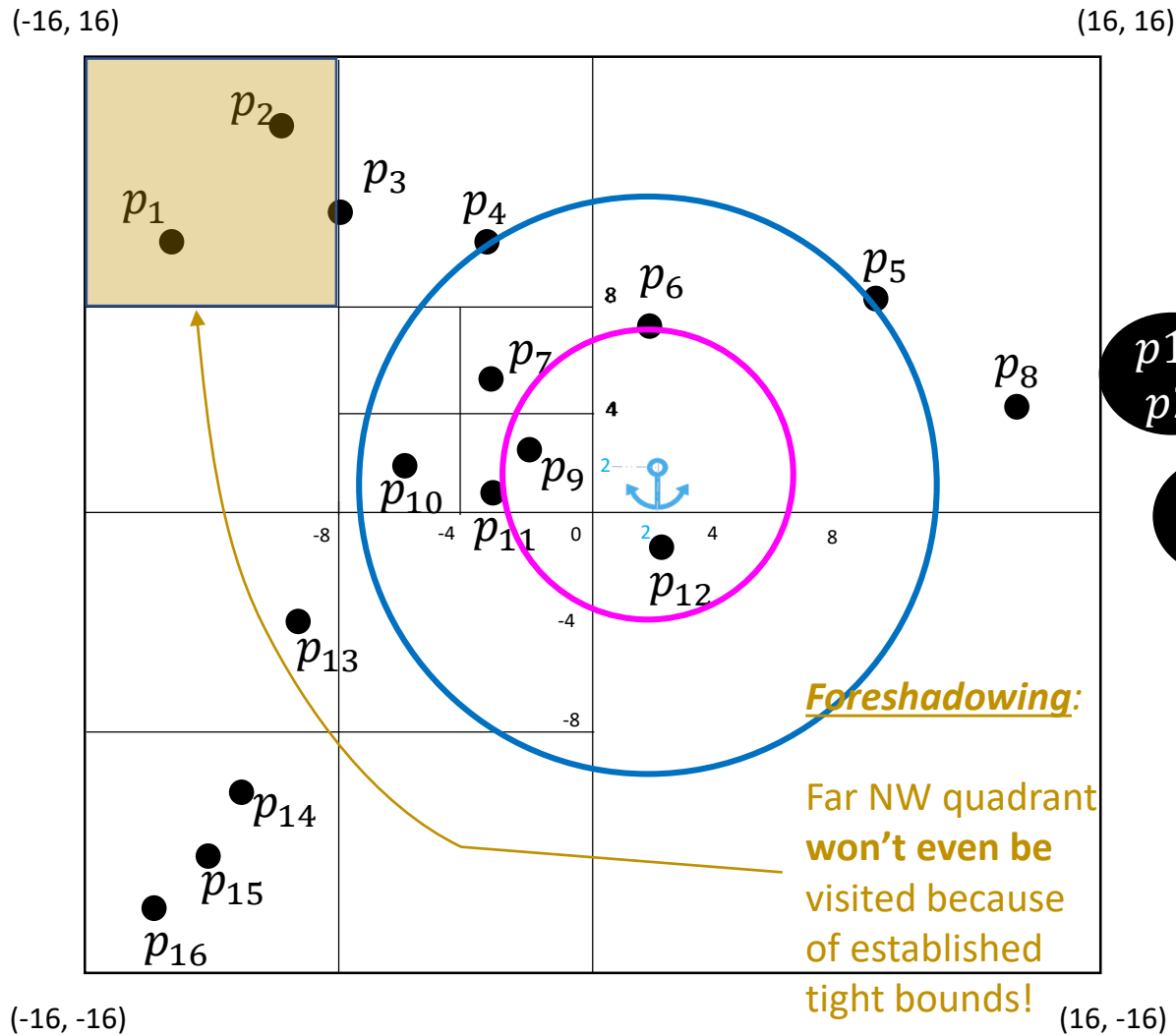
# Example of $k$ -NN query in PR-QuadTree

anchor = (2, 2),  $k=2$ , Bucket size  $b = 3$



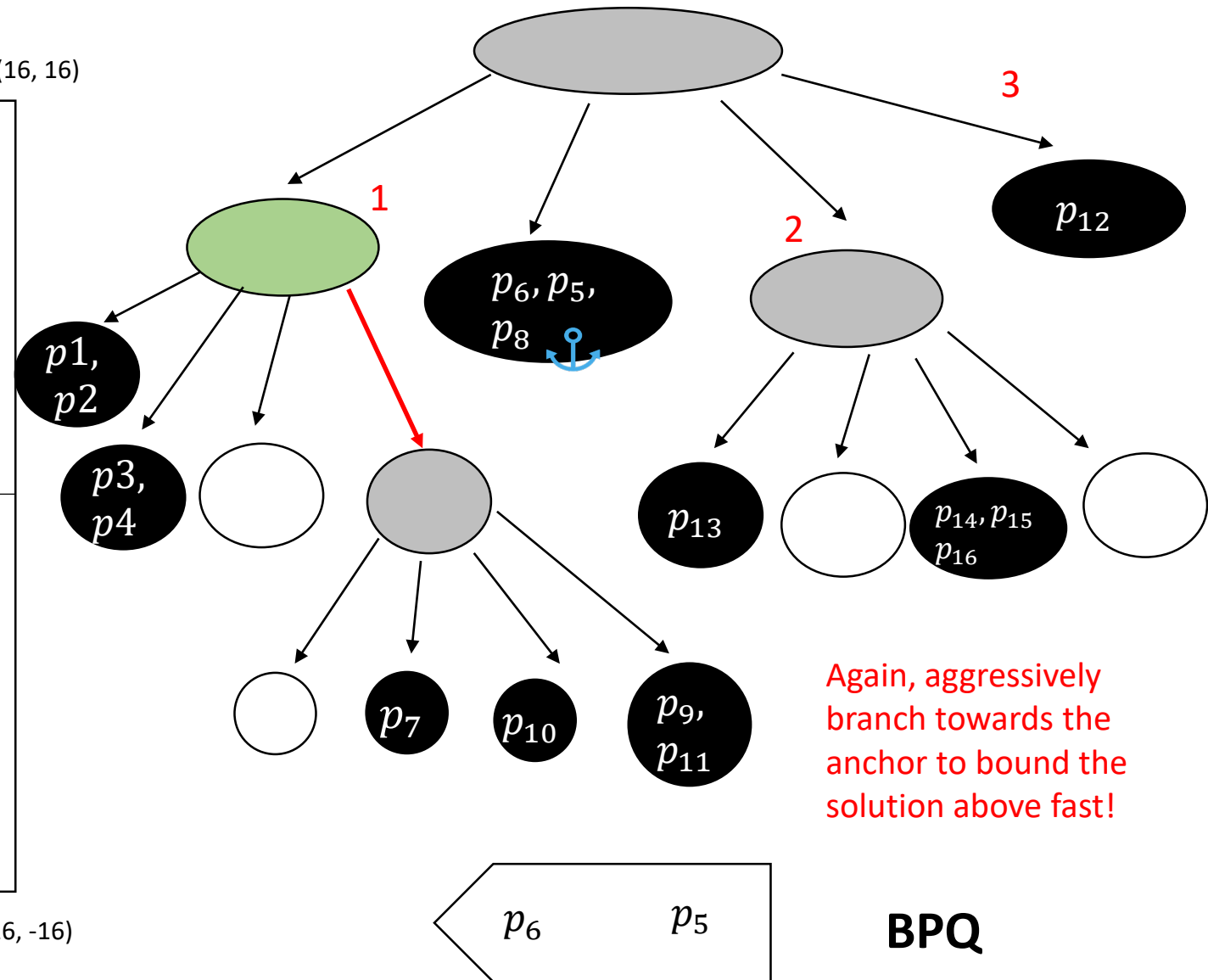
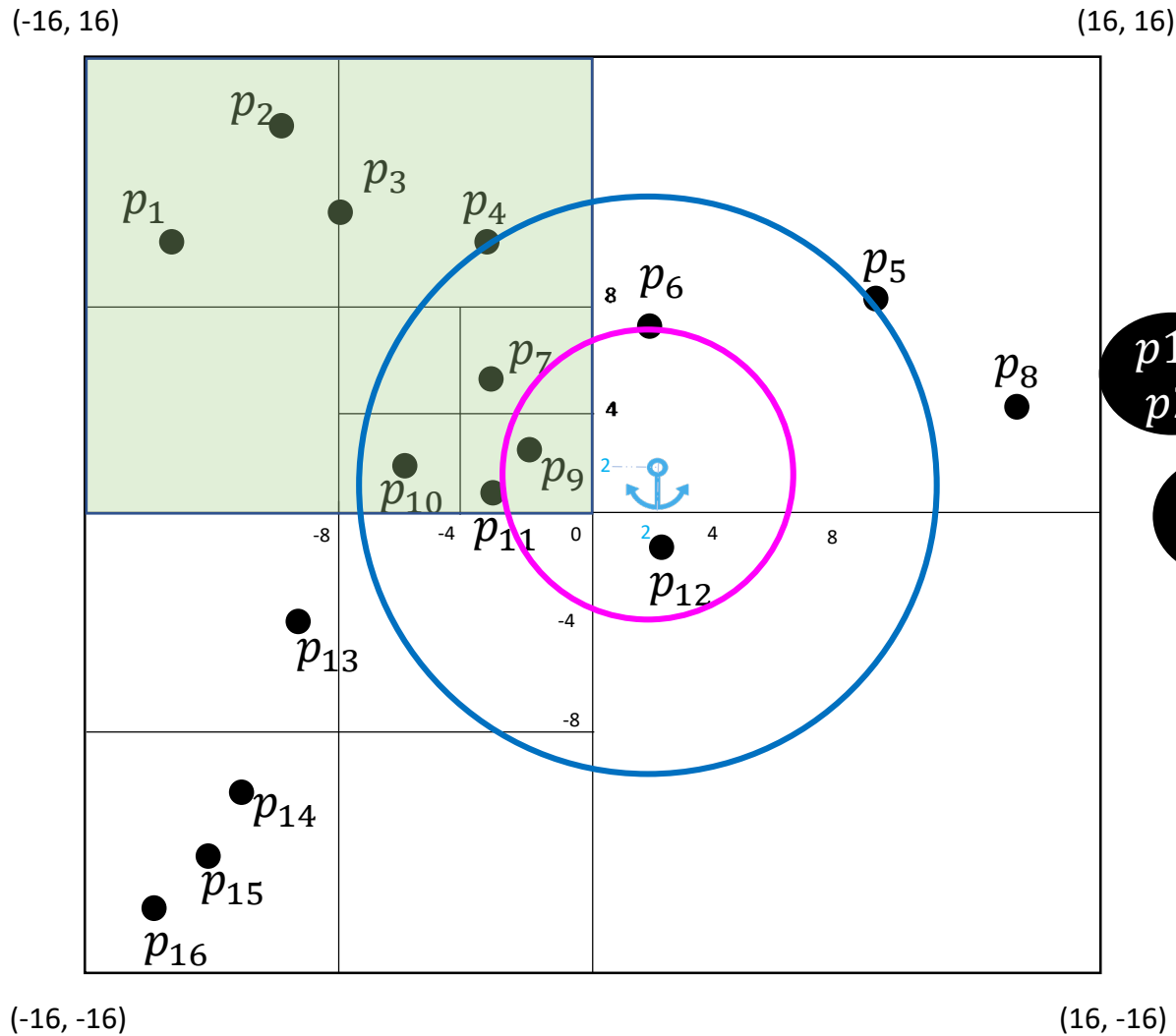
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anchor = (2, 2),  $k=2$ , Bucket size  $b = 3$



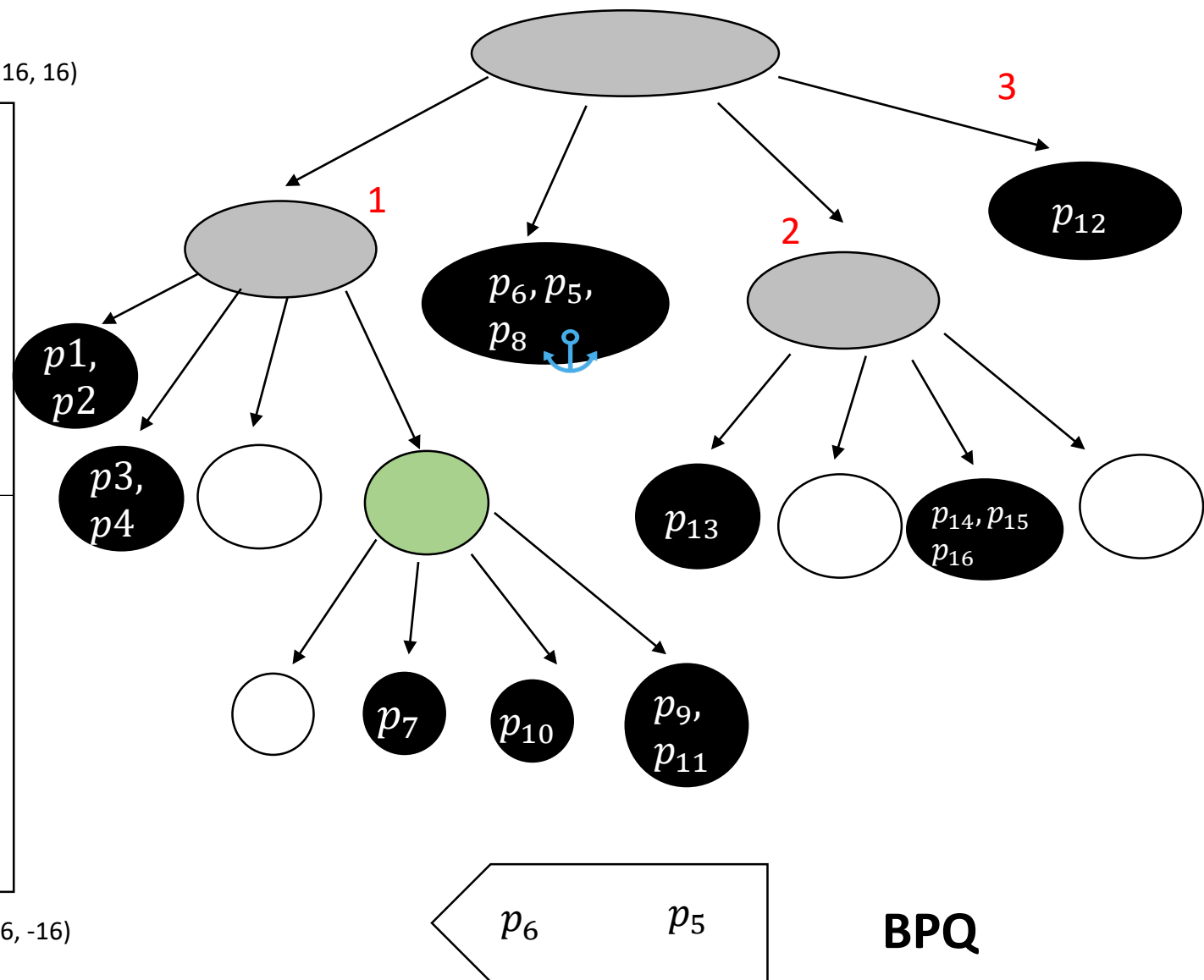
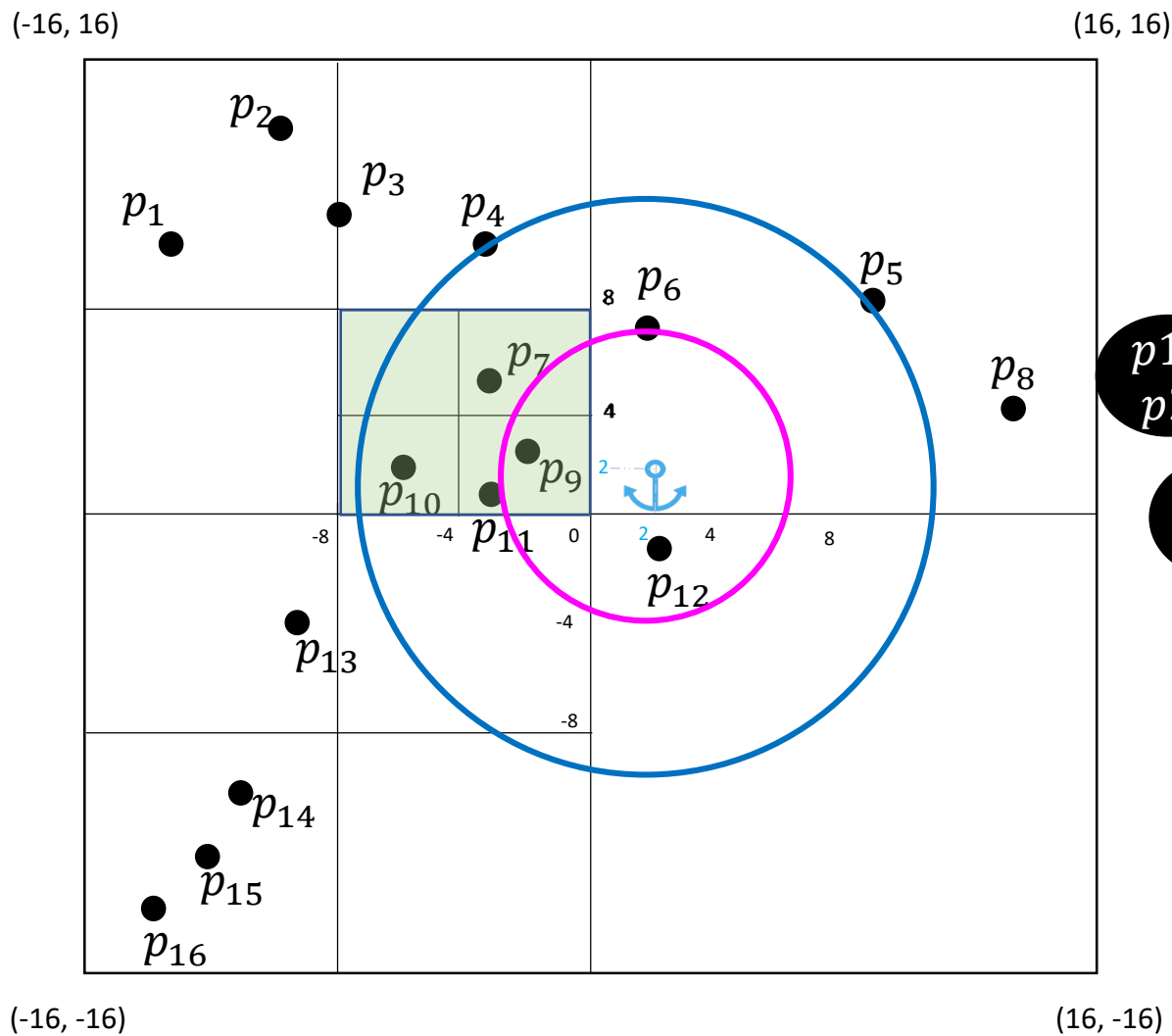
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anchor = (2, 2),  $k=2$ , Bucket size  $b = 3$



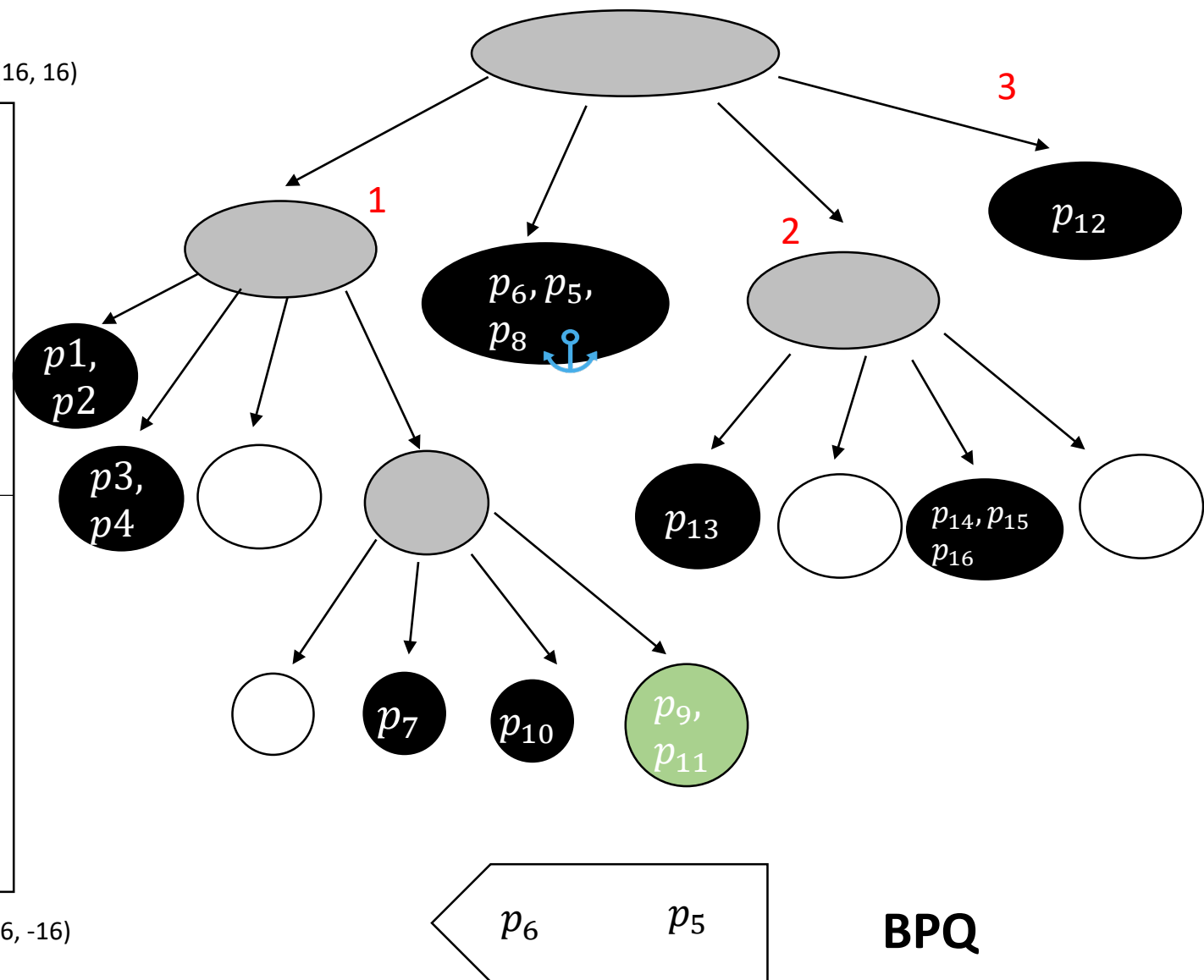
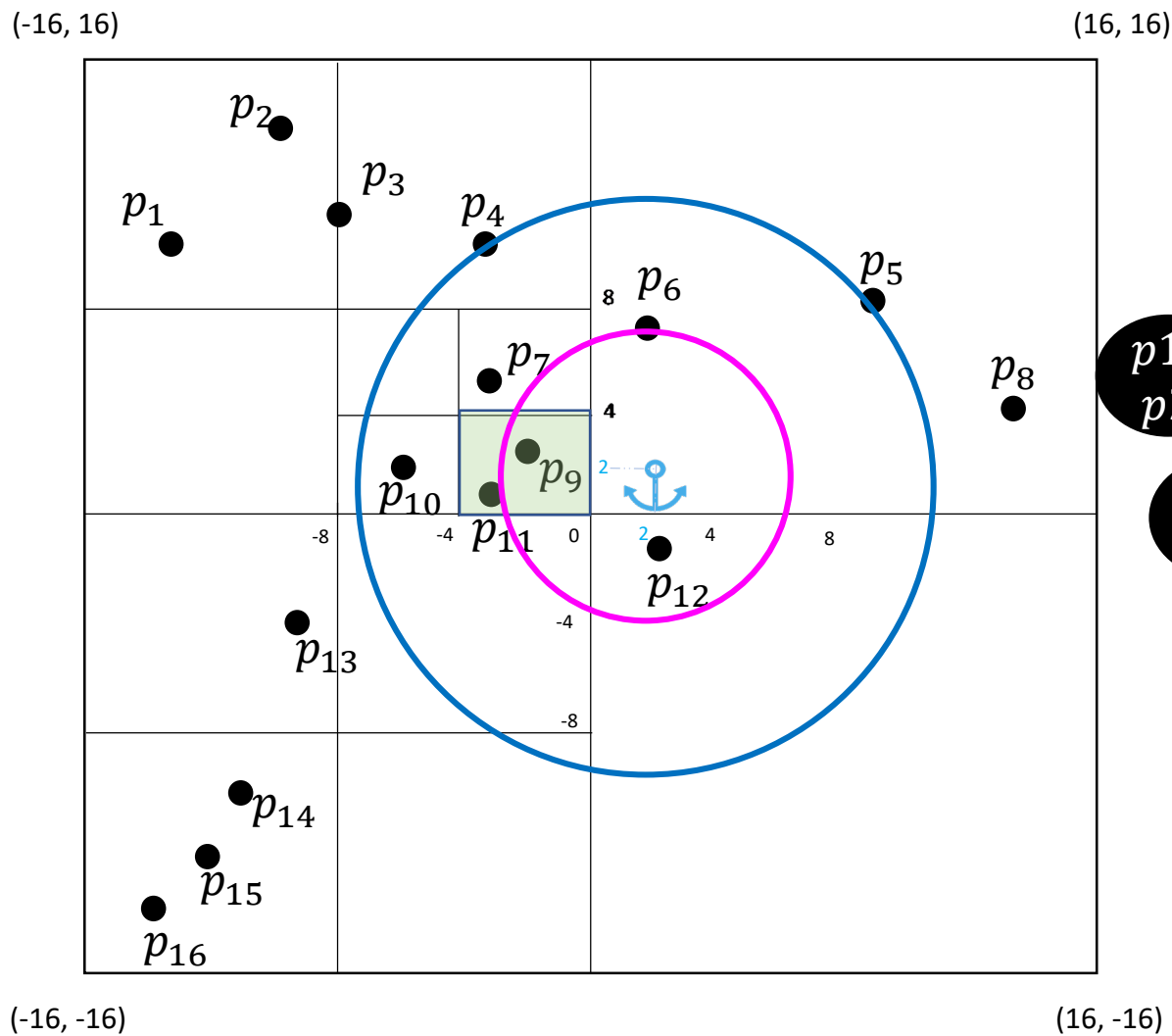
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# Example of $k$ -NN query in PR-QuadTree

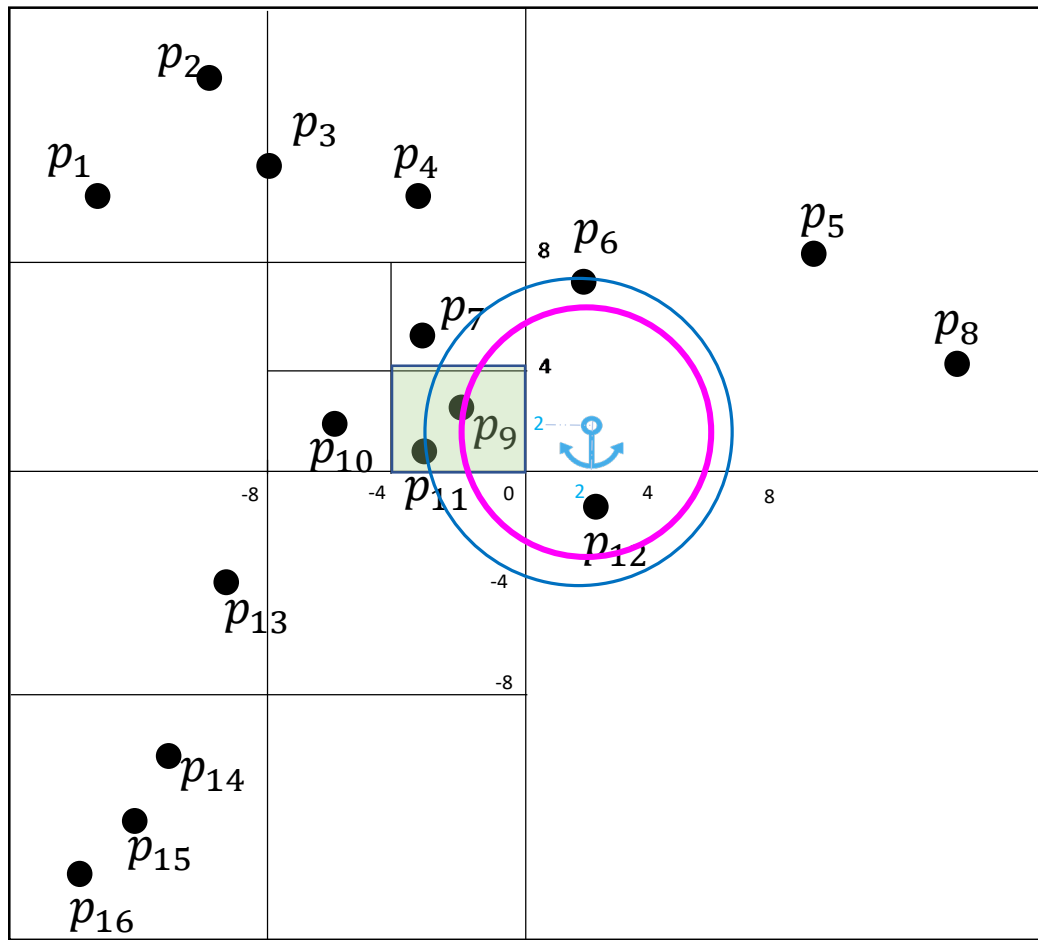
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# Example of $k$ -NN query in PR-QuadTree

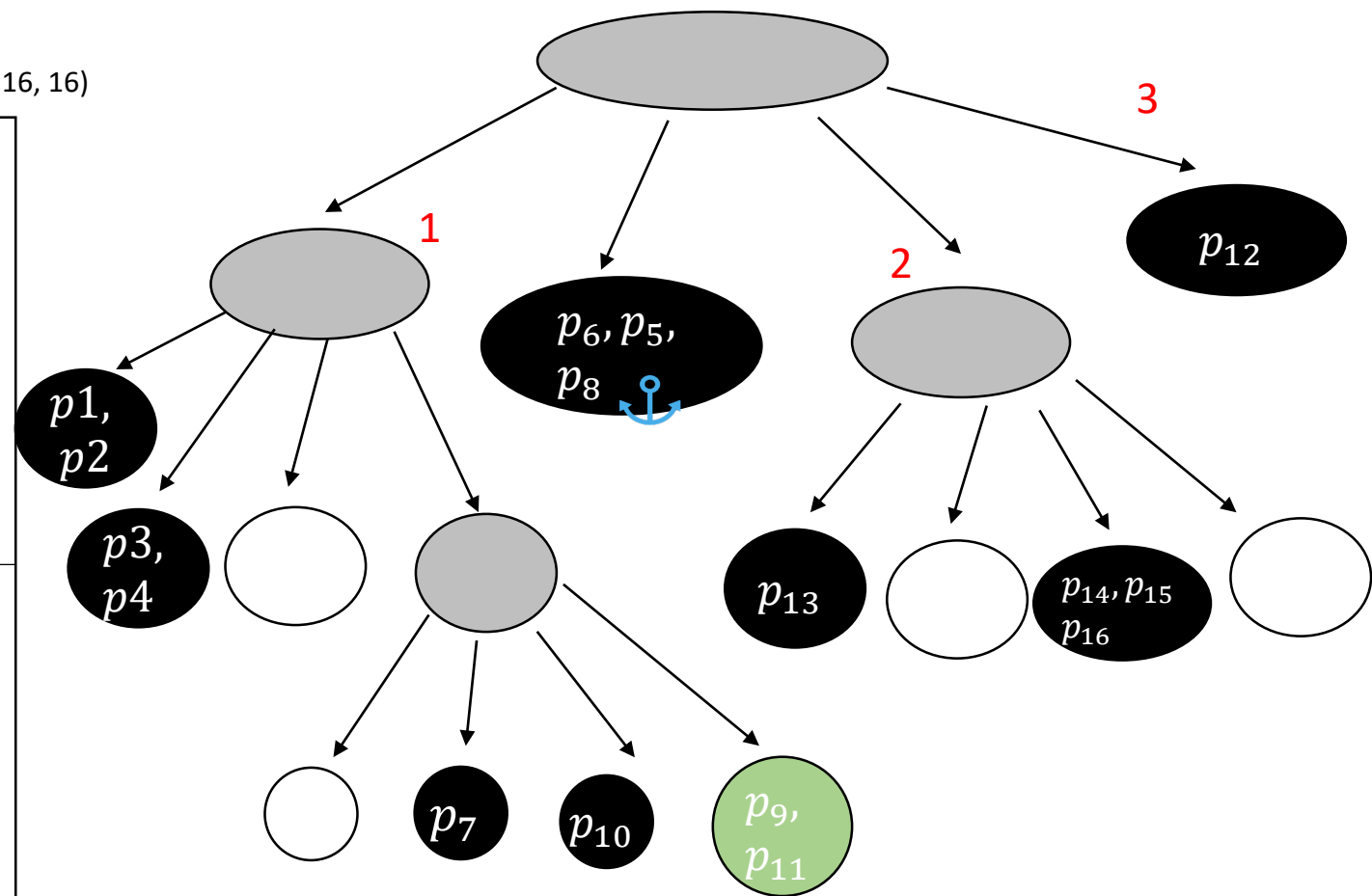
anchor = (2, 2),  $k=2$ , Bucket size  $b = 3$

(-16, 16) (16, 16)

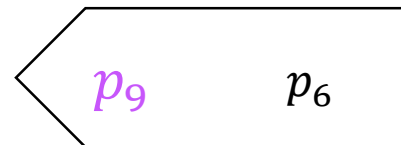


(-16, -16)

(16, -16)



$p_9$  improves upon  $p_6$ !  
New best neighbor!

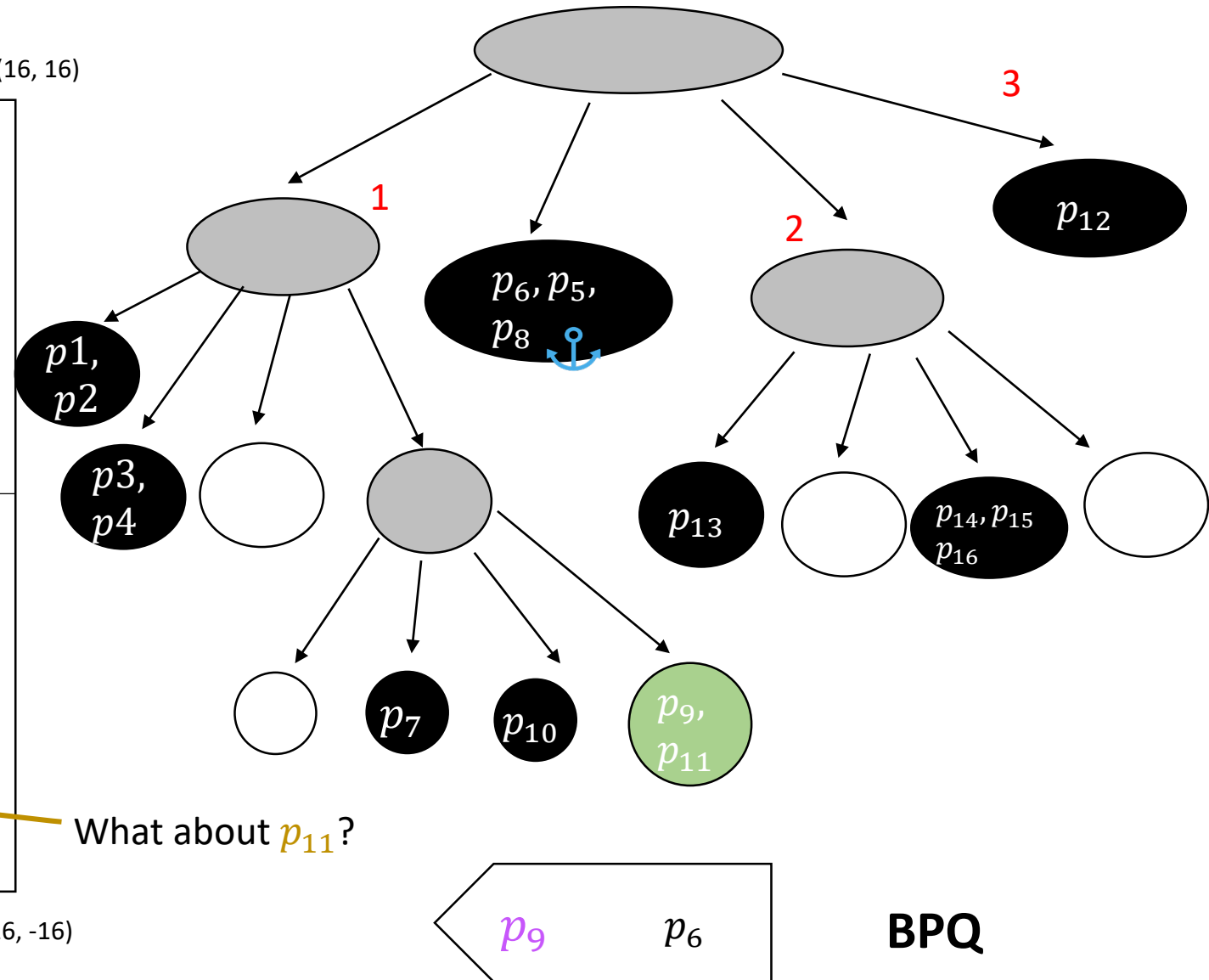
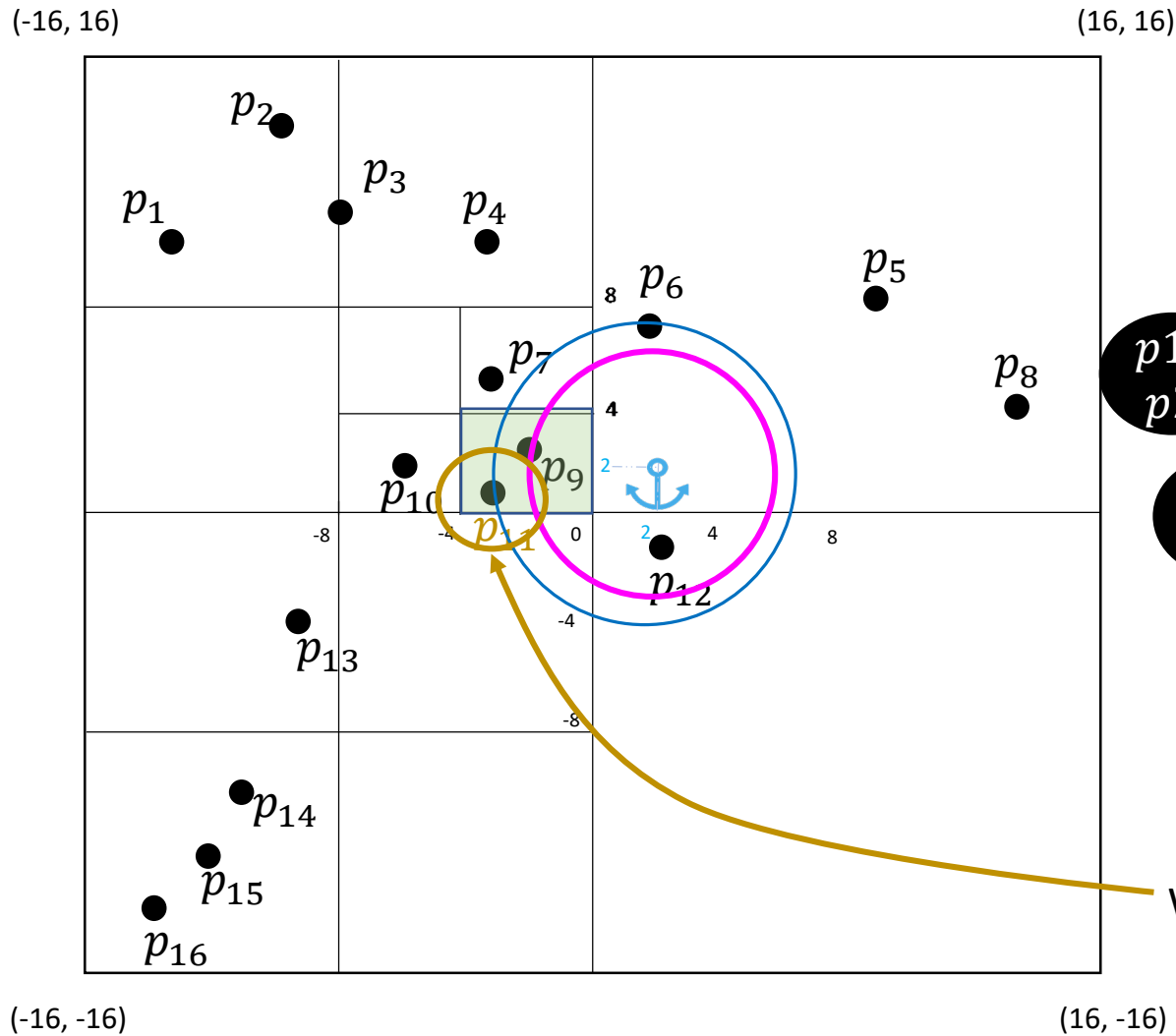


**BPQ**



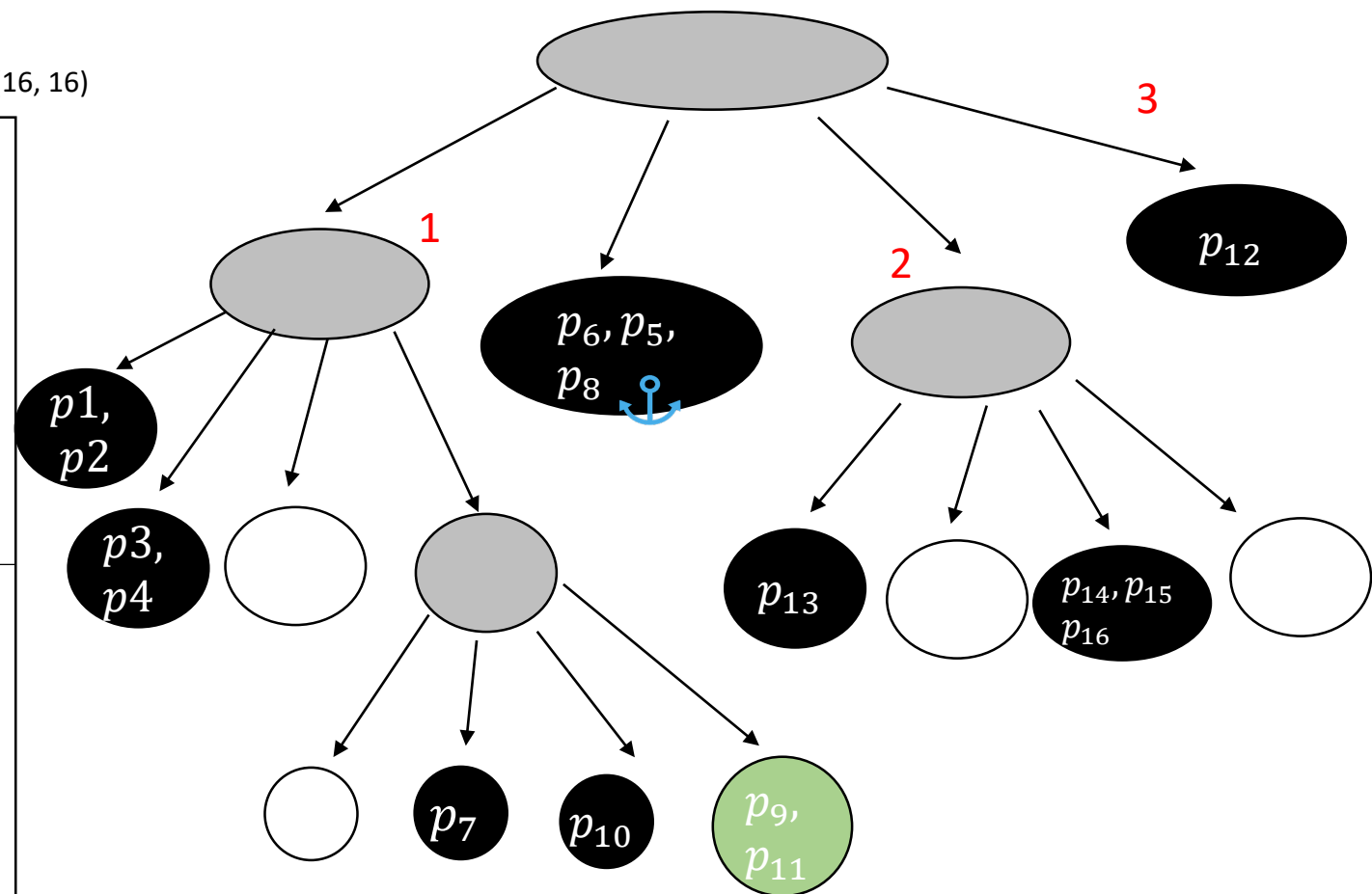
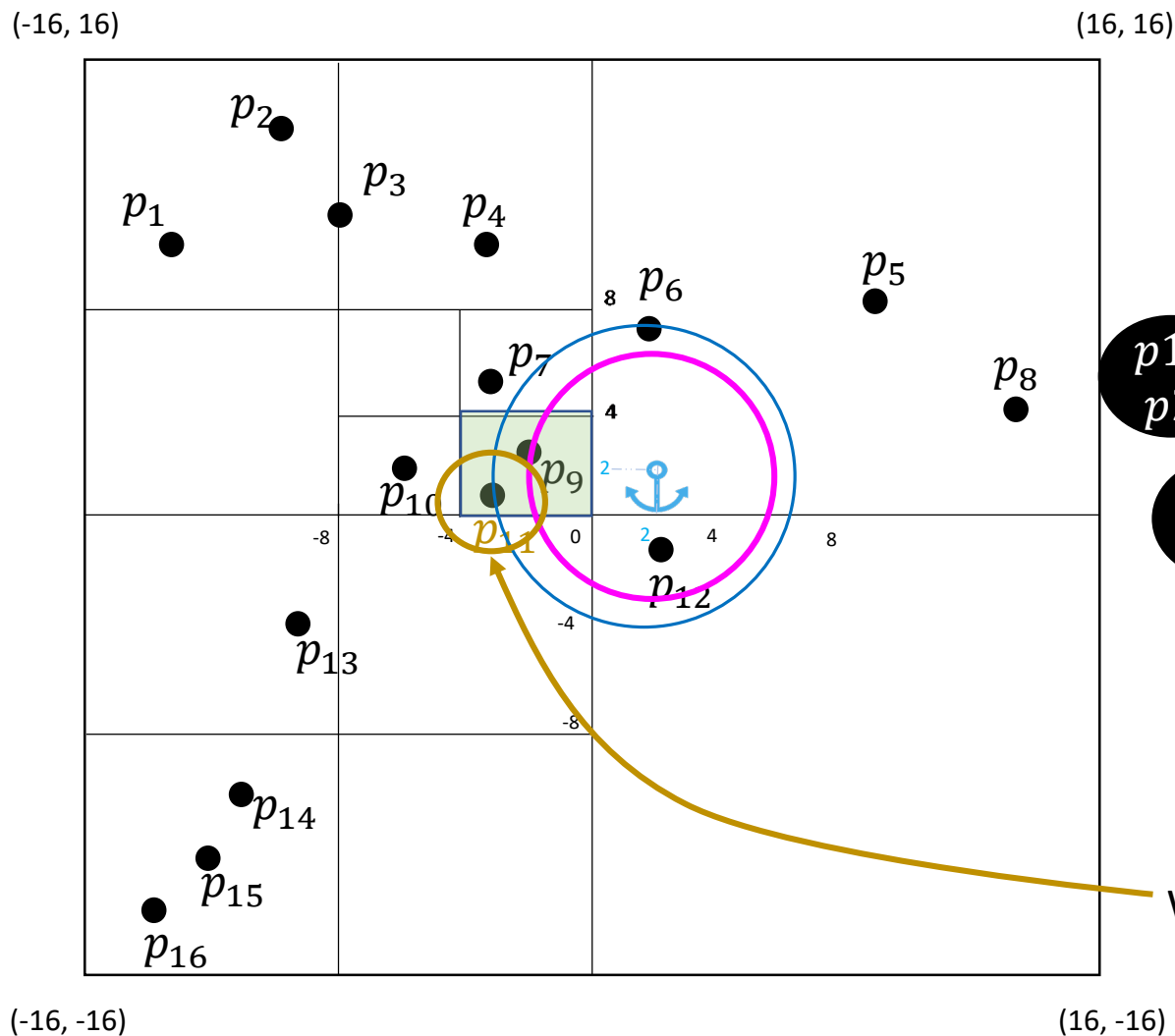
# Example of $k$ -NN query in PR-QuadTree

anchor = (2, 2),  $k=2$ , Bucket size  $b = 3$



# Example of $k$ -NN query in PR-QuadTree

anchor = (2, 2),  $k=2$ , Bucket size  $b = 3$



What about  $p_{11}$ ?

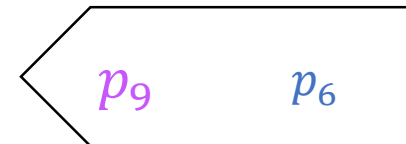
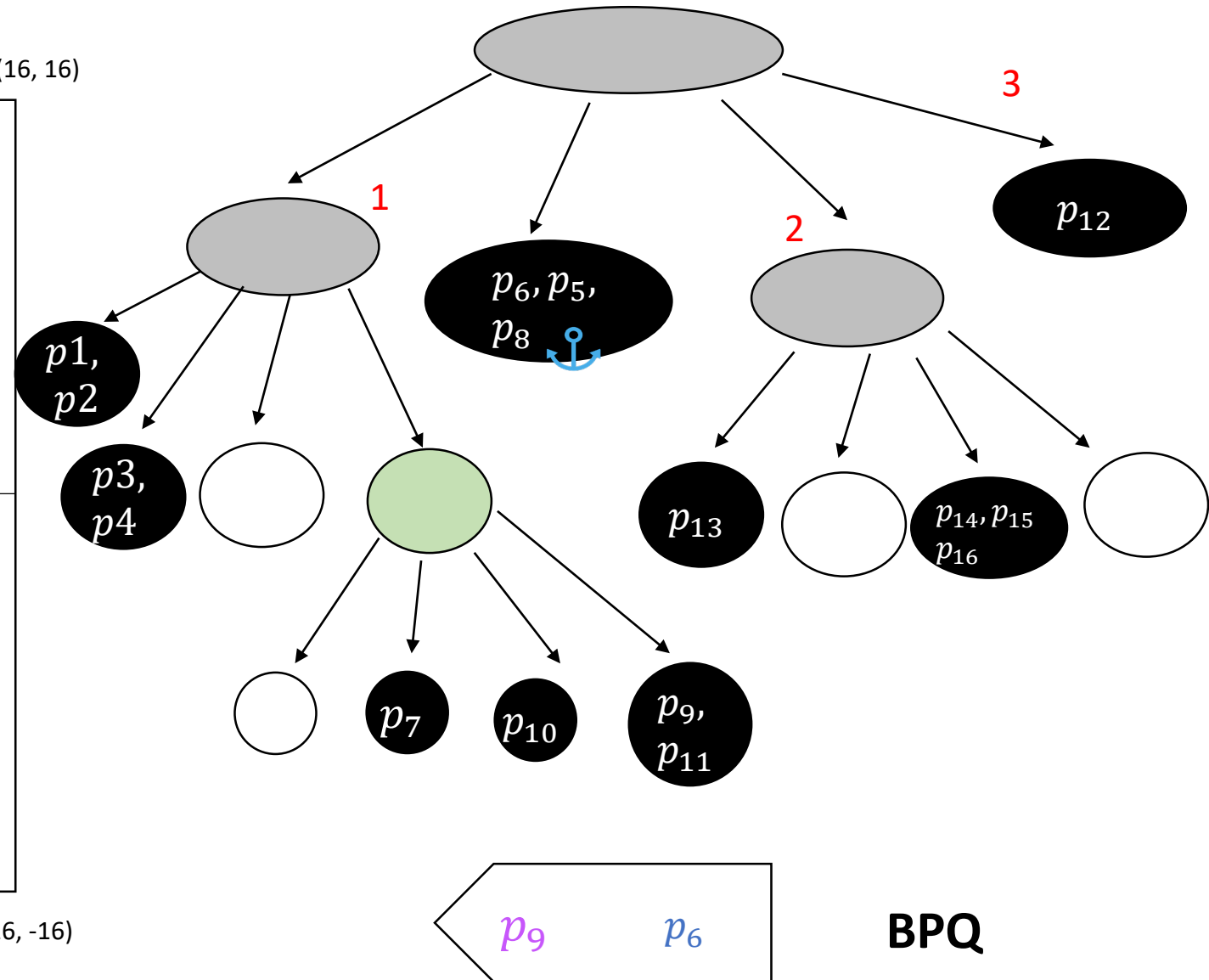
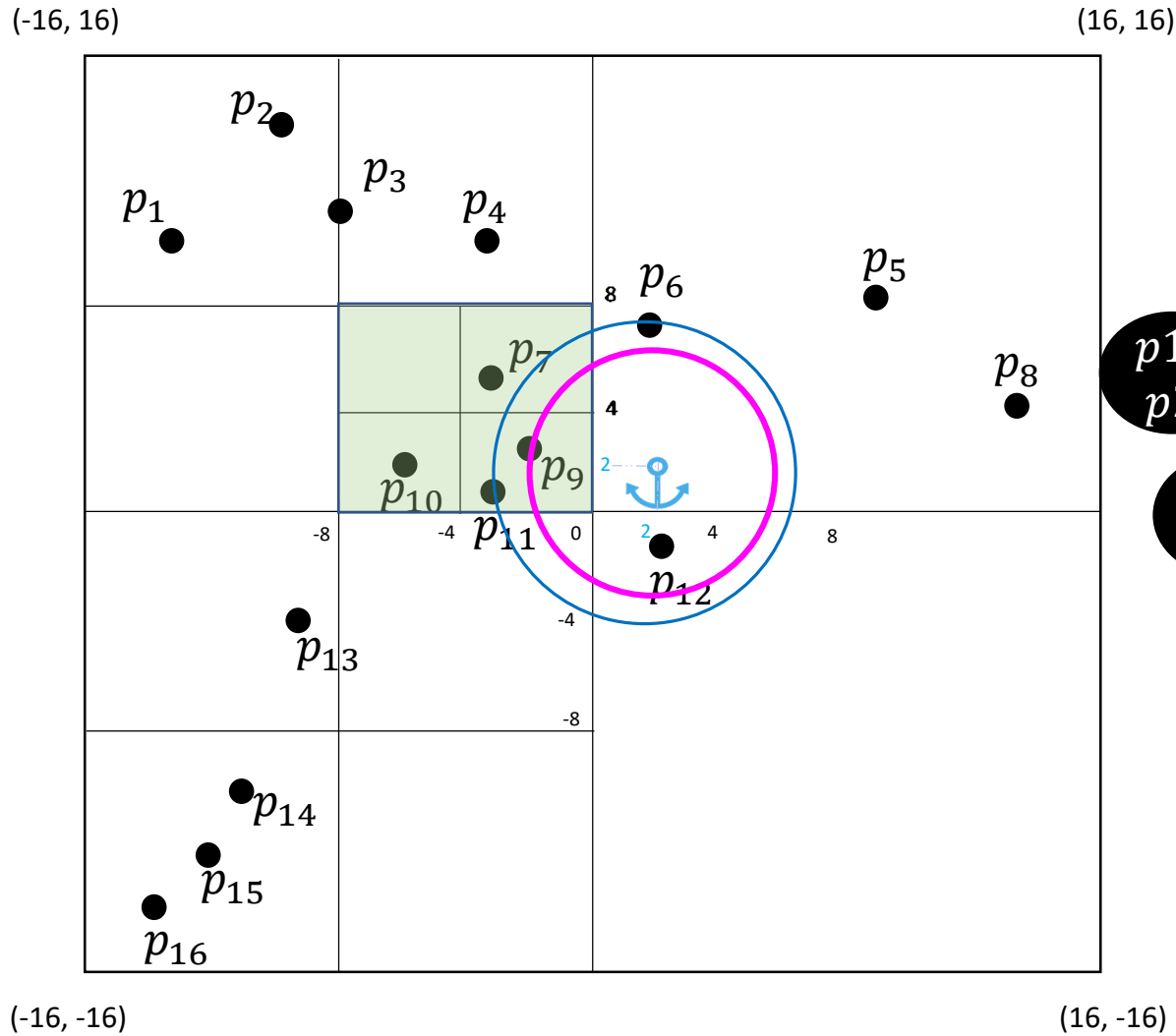


BPQ

Since it has the same "priority" as  $p_6$ , we do **not** insert it!  
 $p_6$  is now our second-best neighbor!

# Example of $k$ -NN query in PR-QuadTree

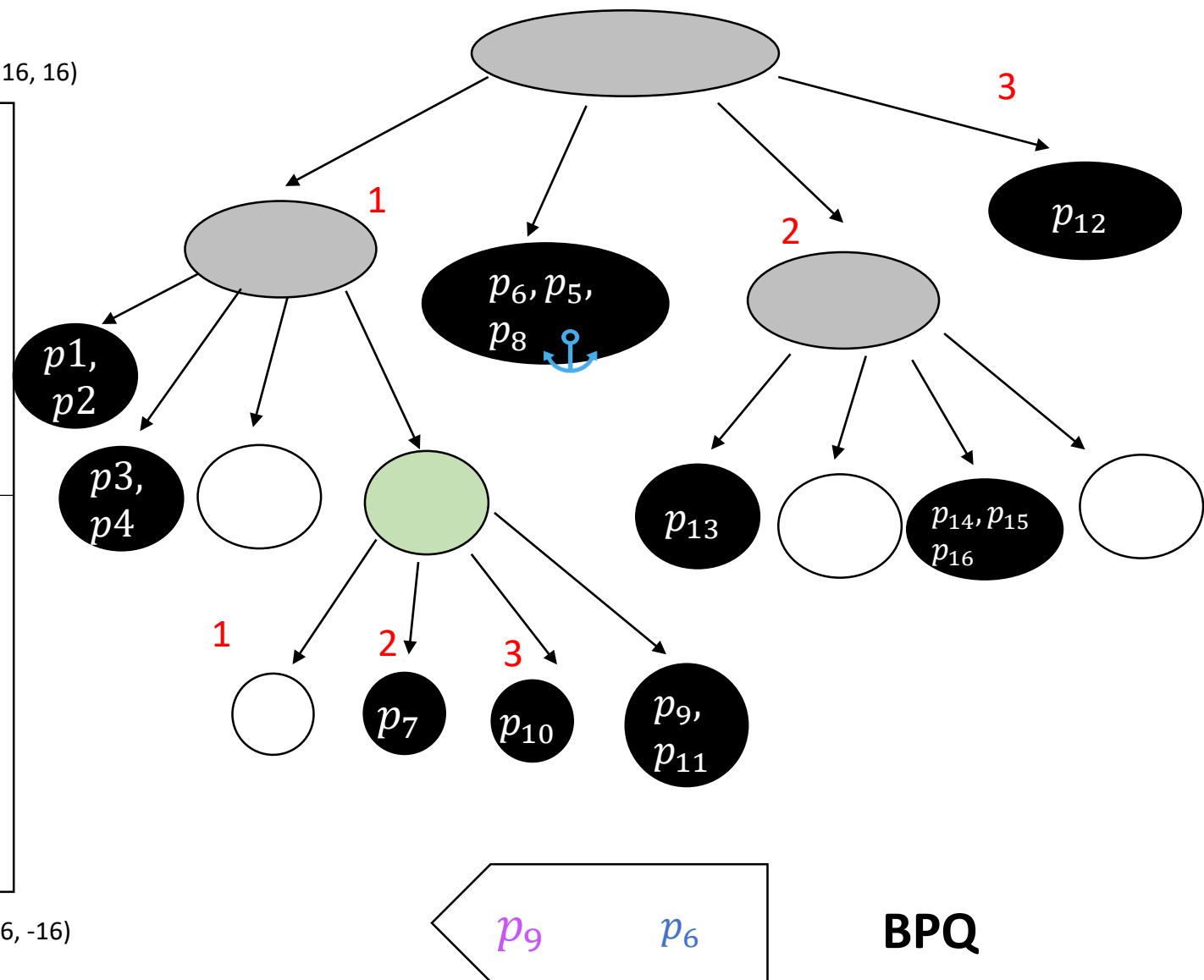
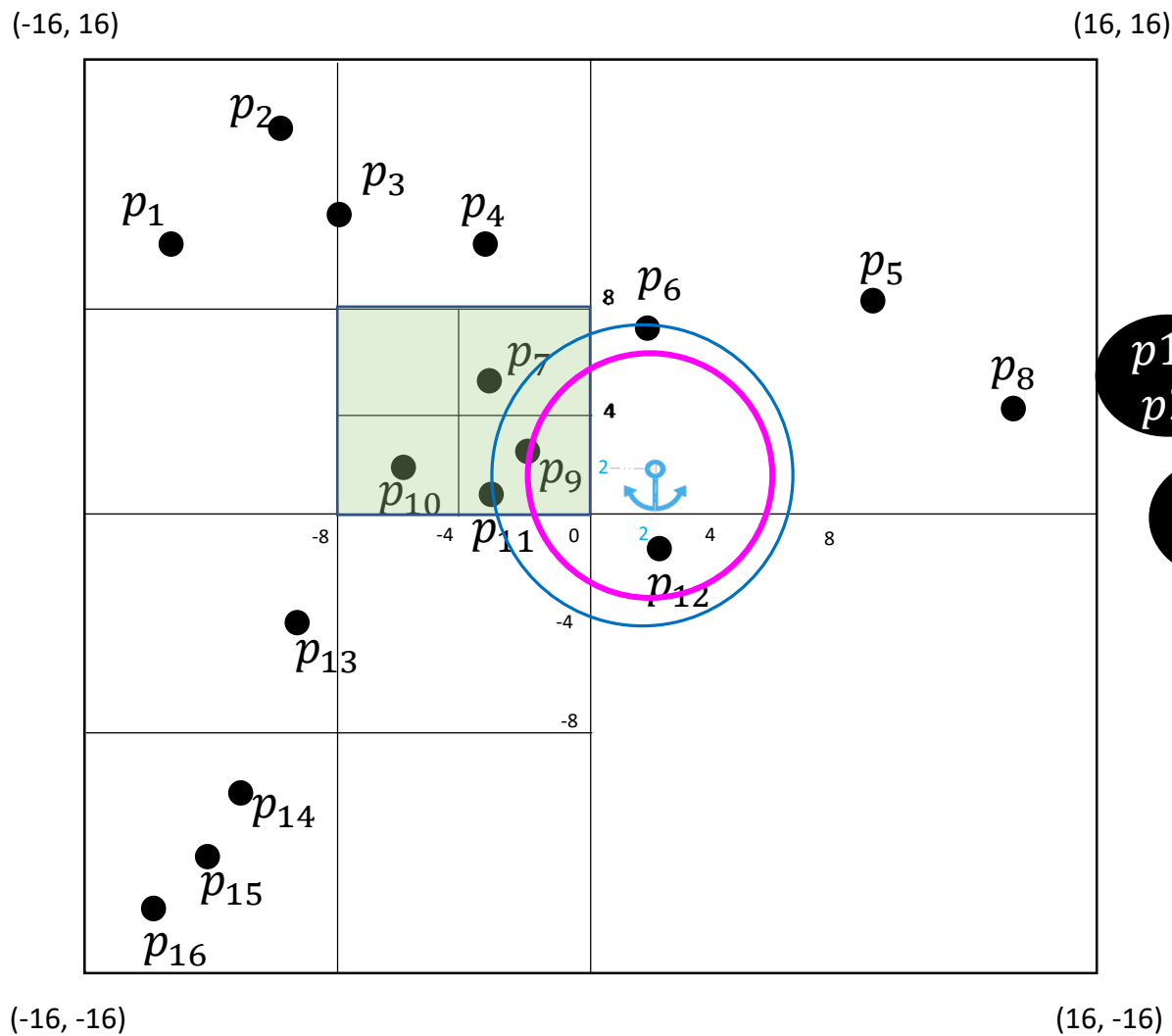
anchor = (2, 2),  $k=2$ , Bucket size  $b = 3$



BPQ

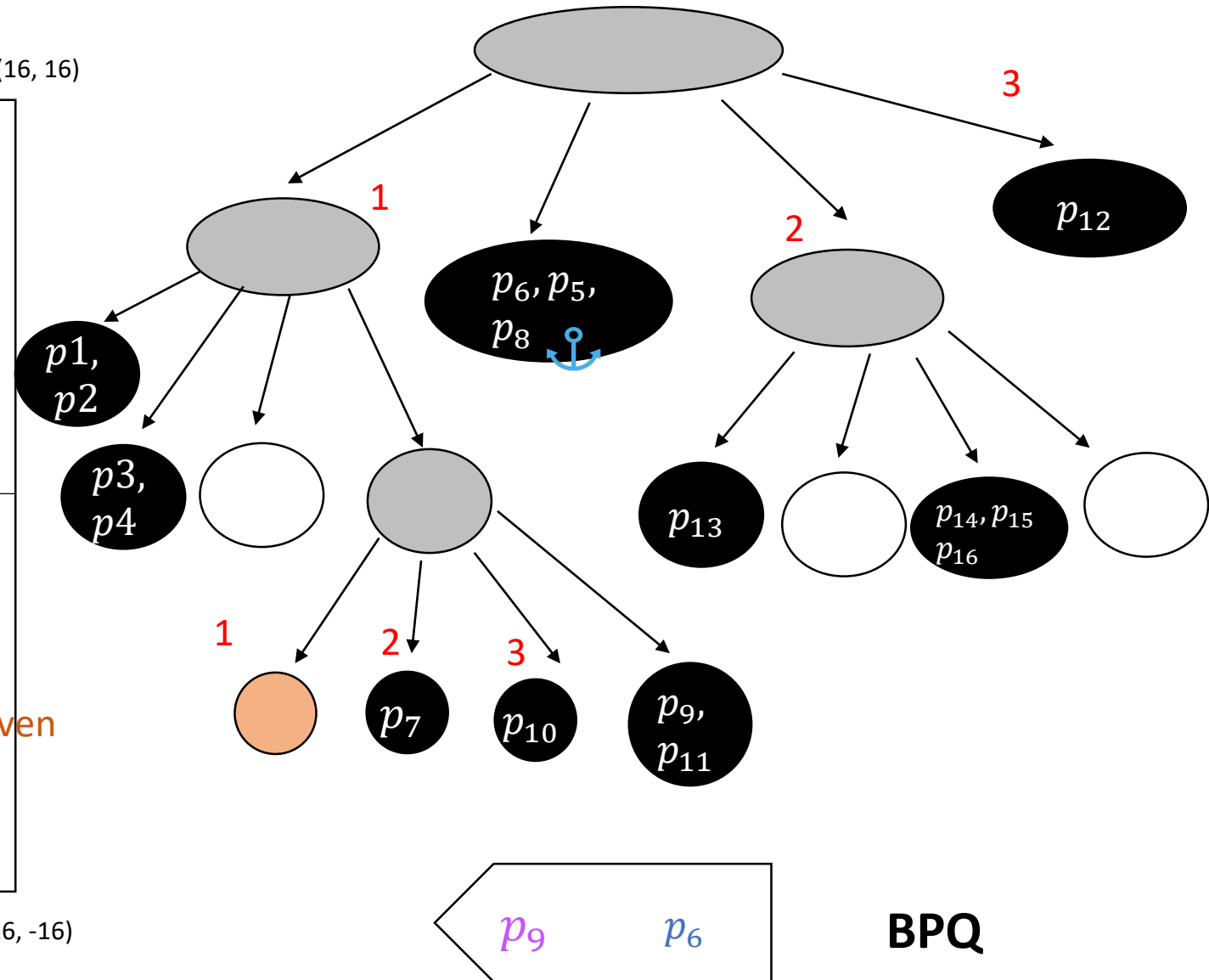
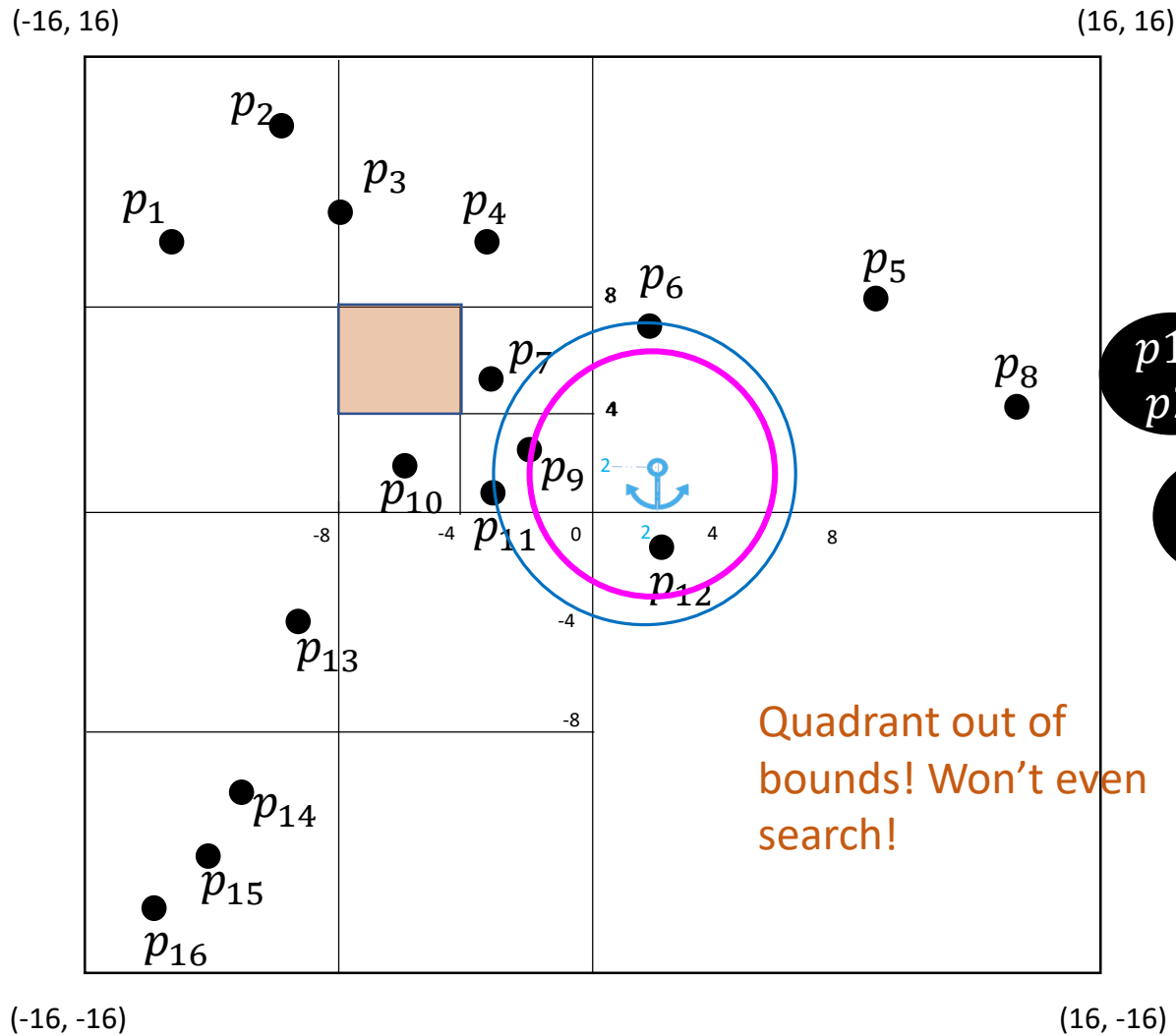
# Example of $k$ -NN query in PR-QuadTree

anchor = (2, 2),  $k=2$ , Bucket size  $b = 3$



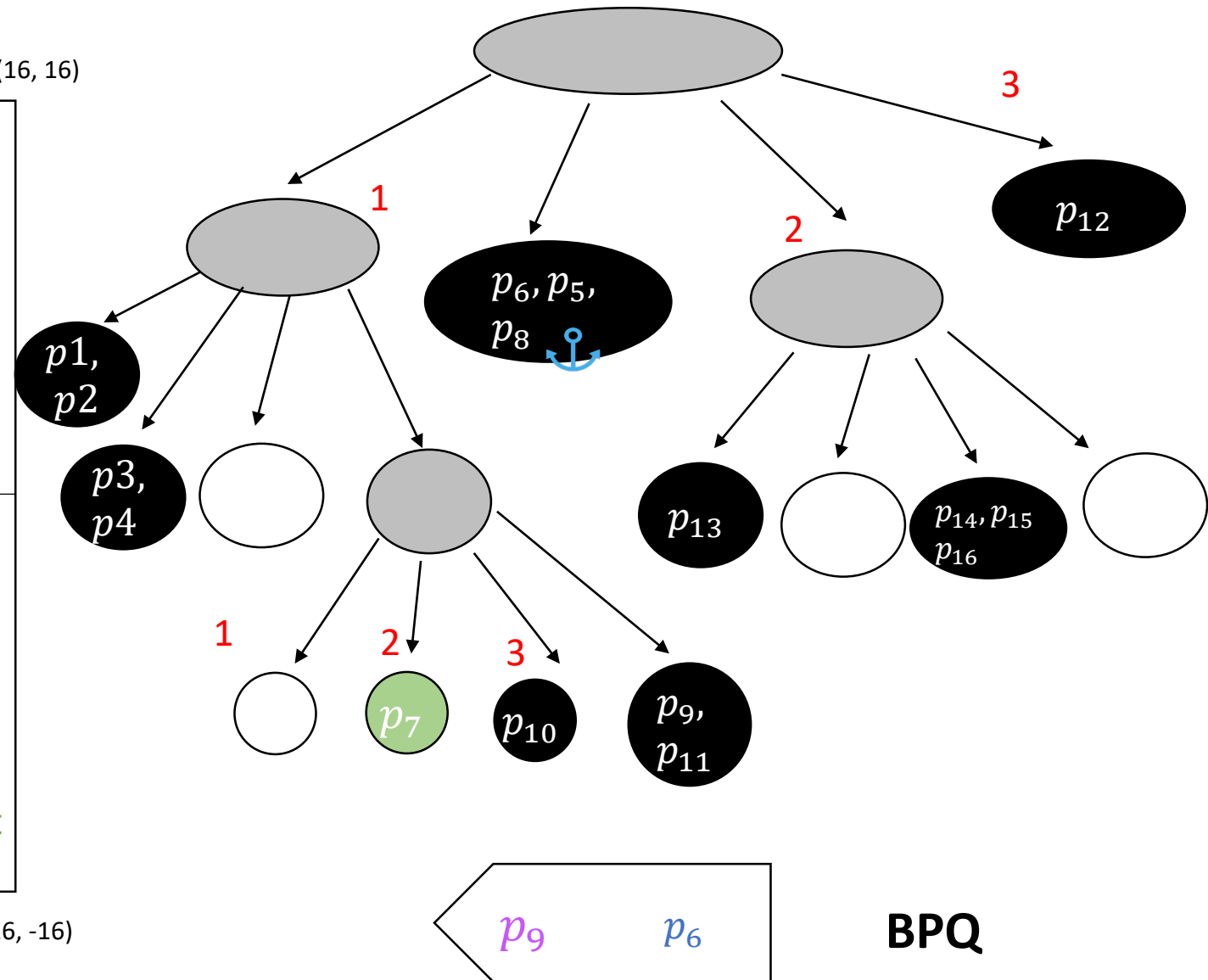
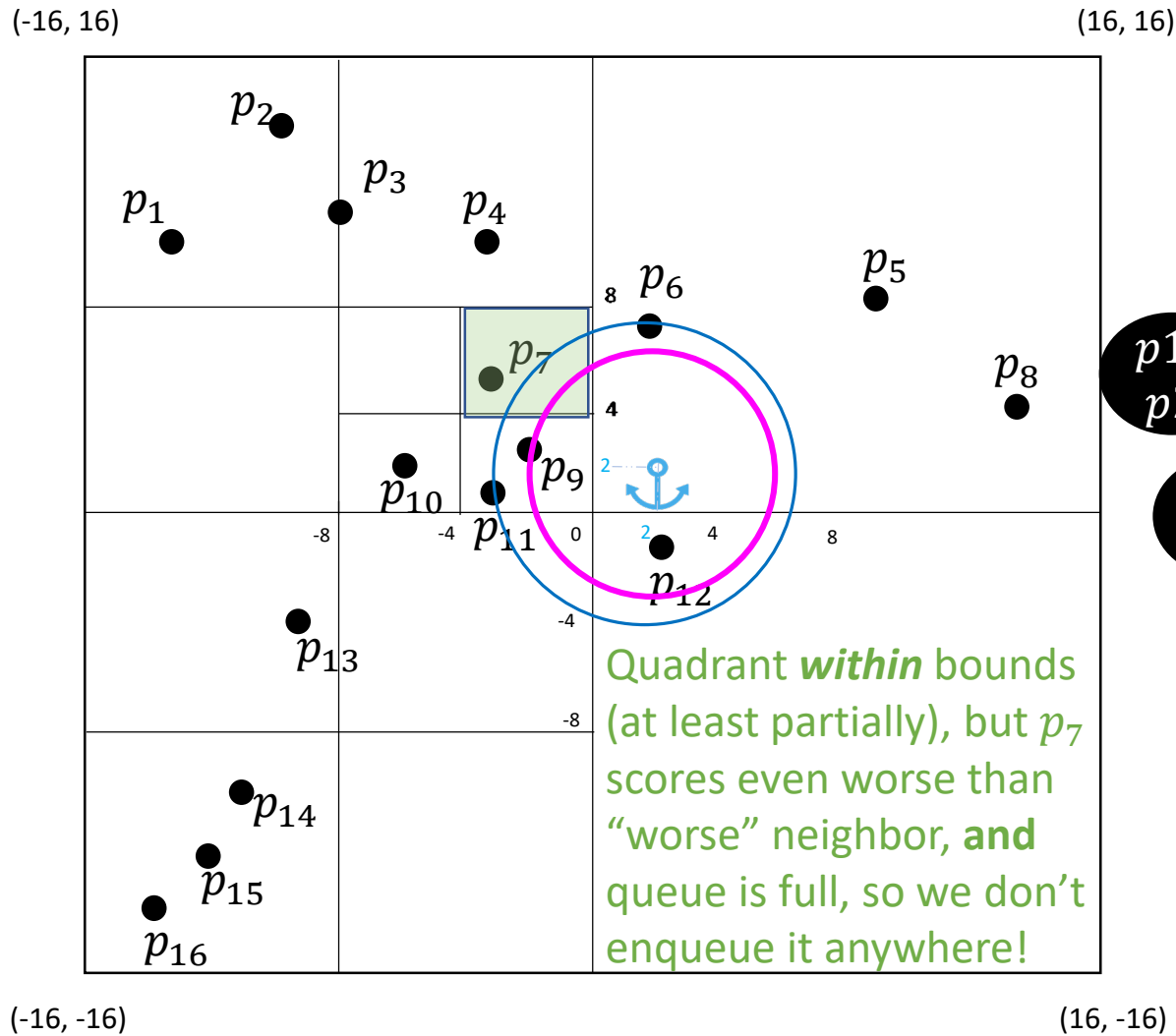
# Example of $k$ -NN query in PR-QuadTree

anchor = (2, 2),  $k=2$ , Bucket size  $b = 3$



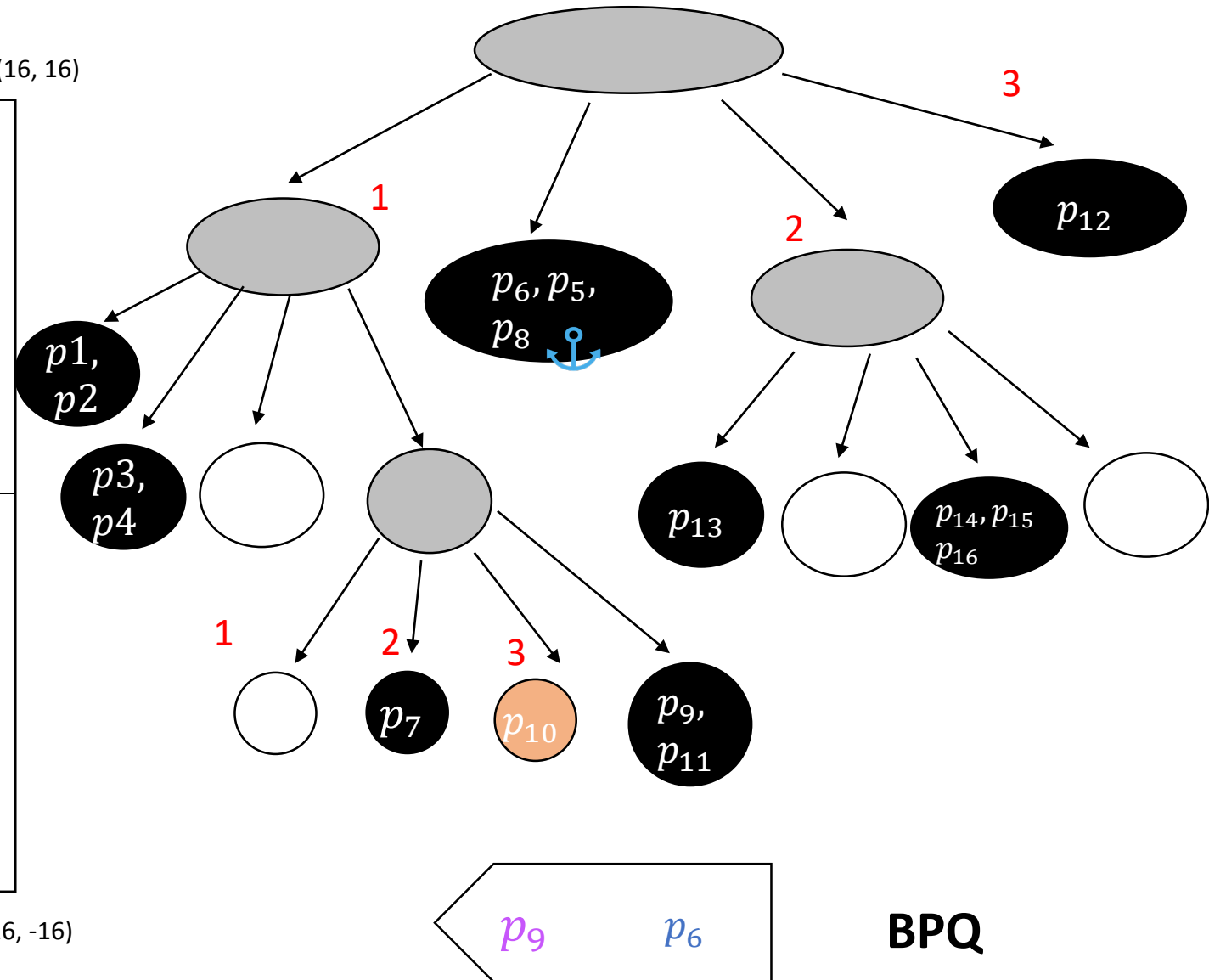
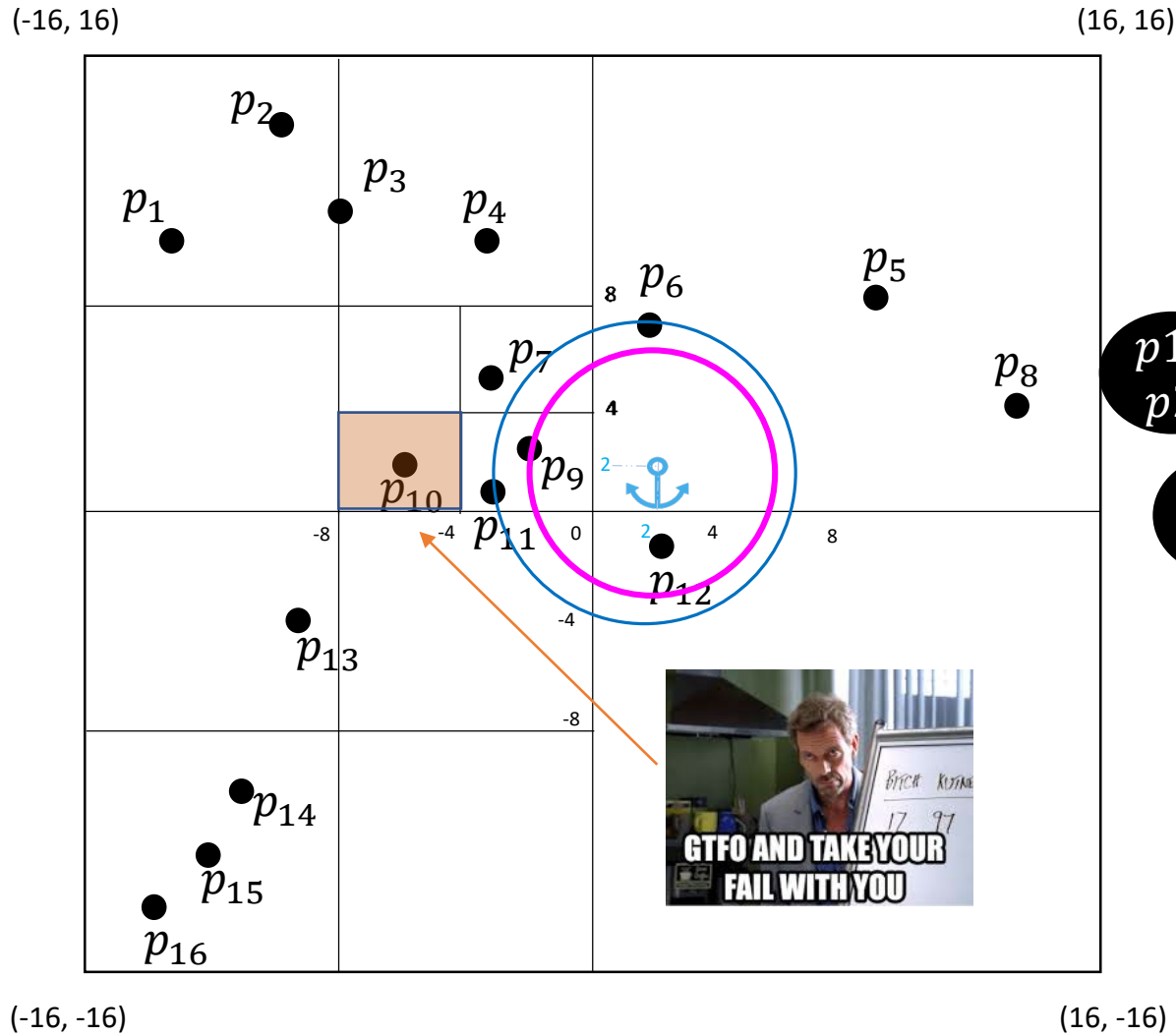
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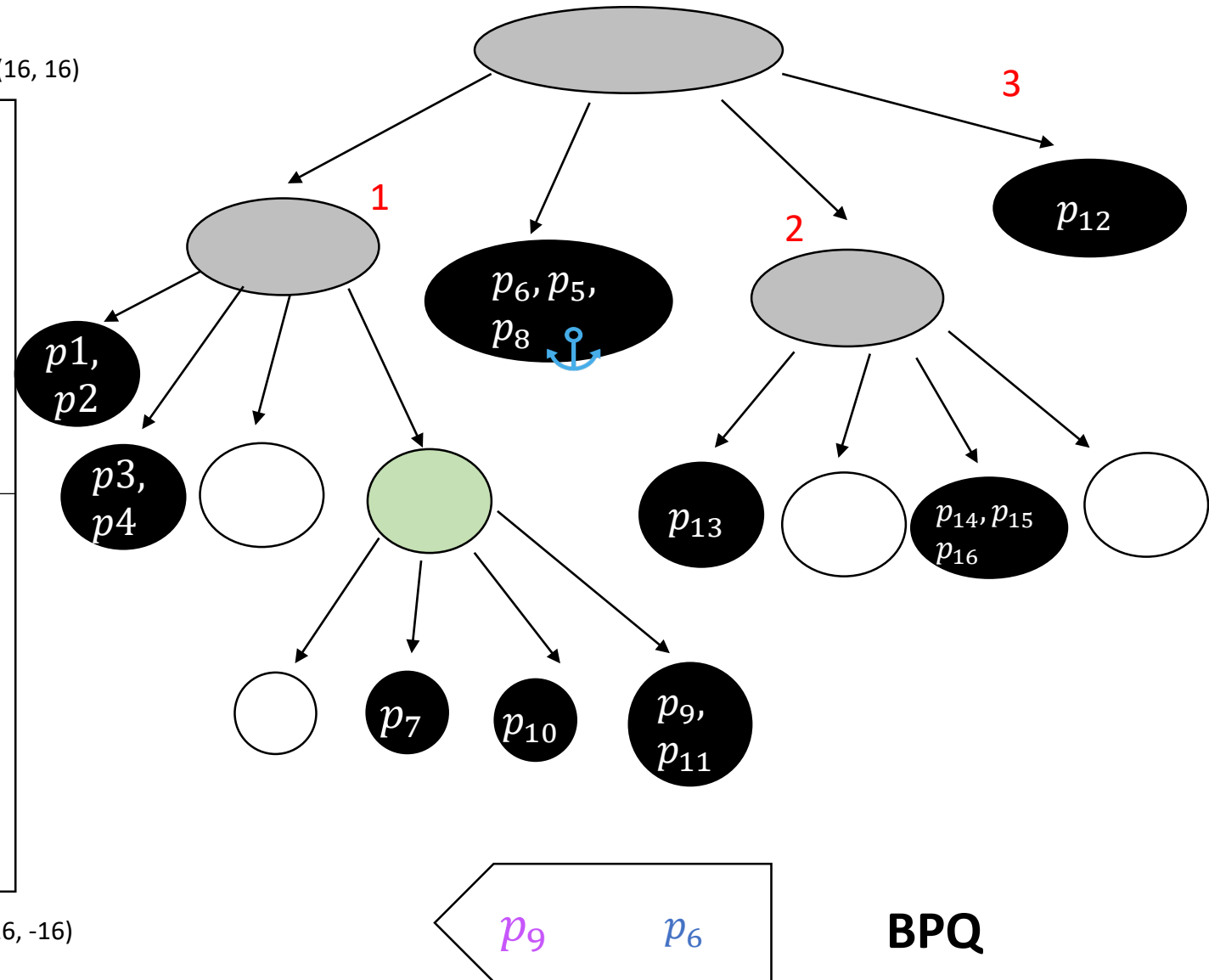
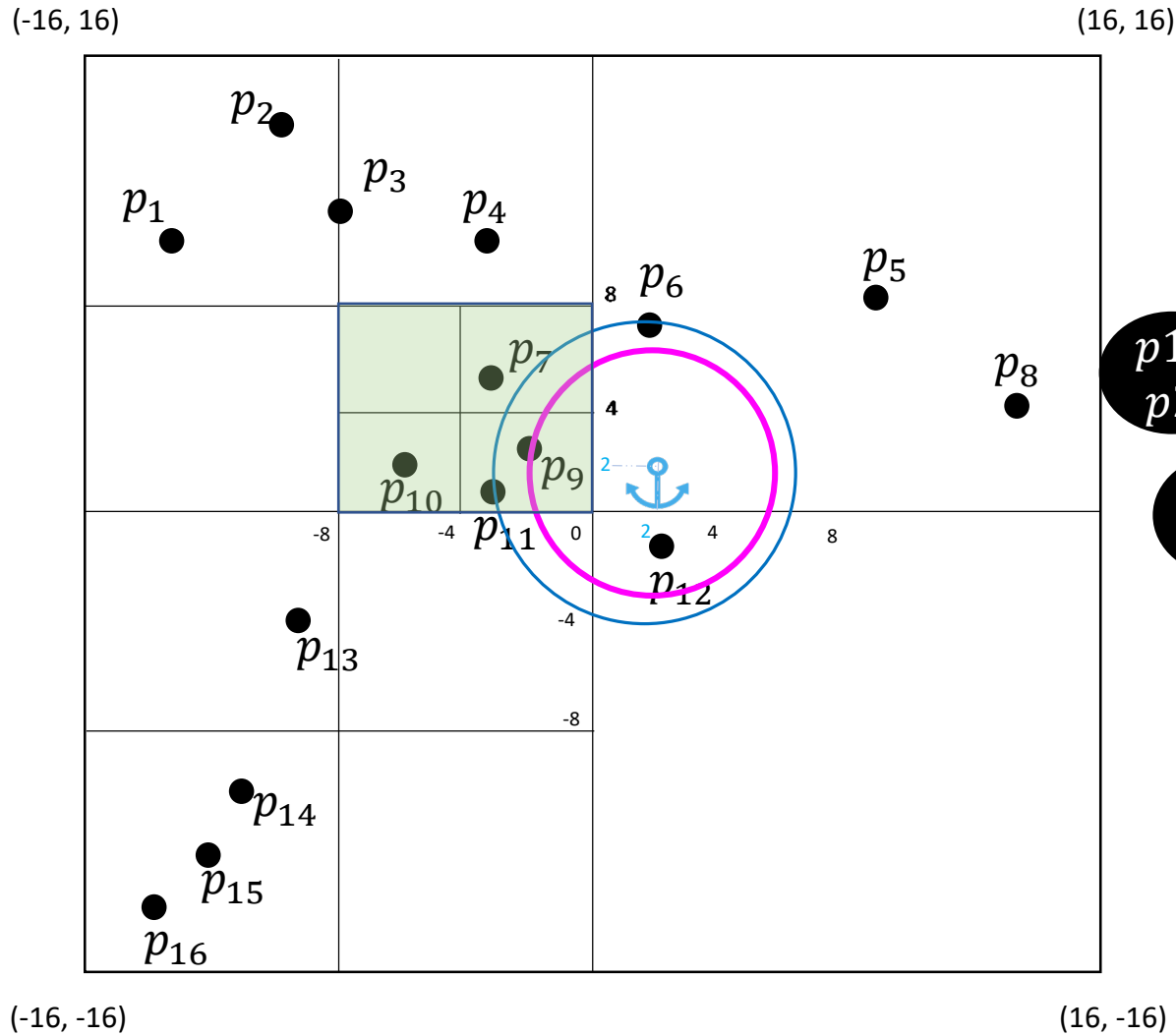
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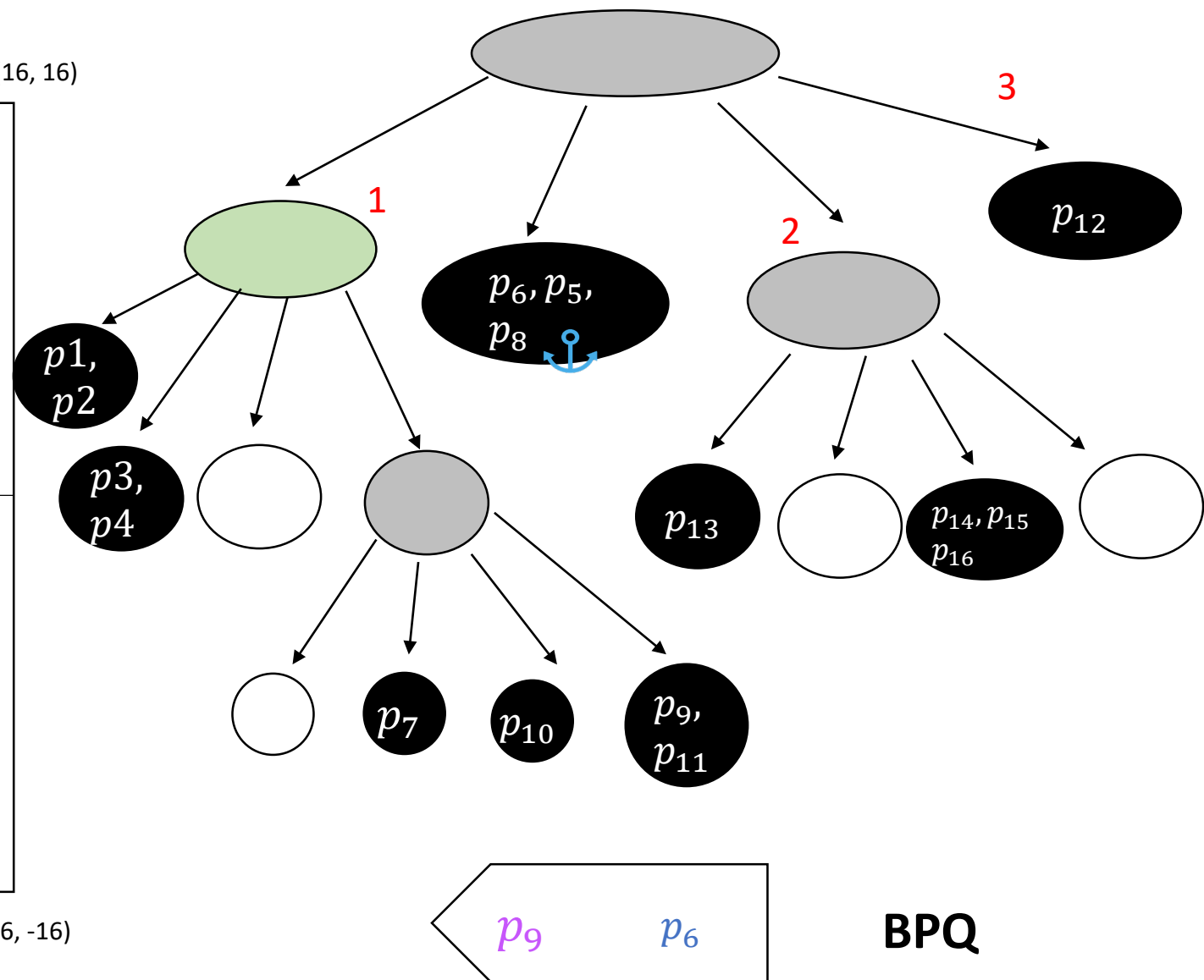
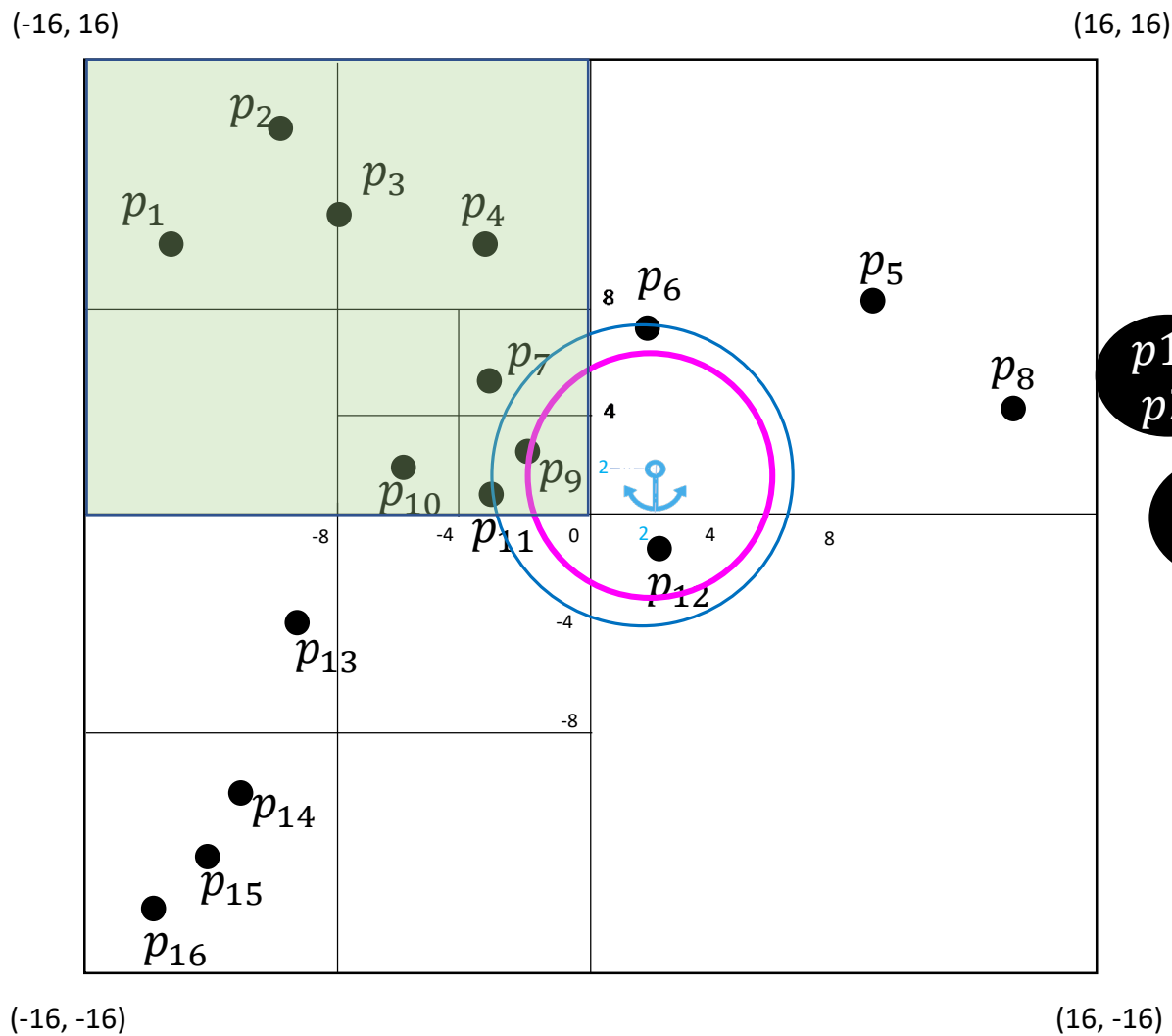
anchor = (2, 2),  $k=2$ , Bucket size  $b = 3$





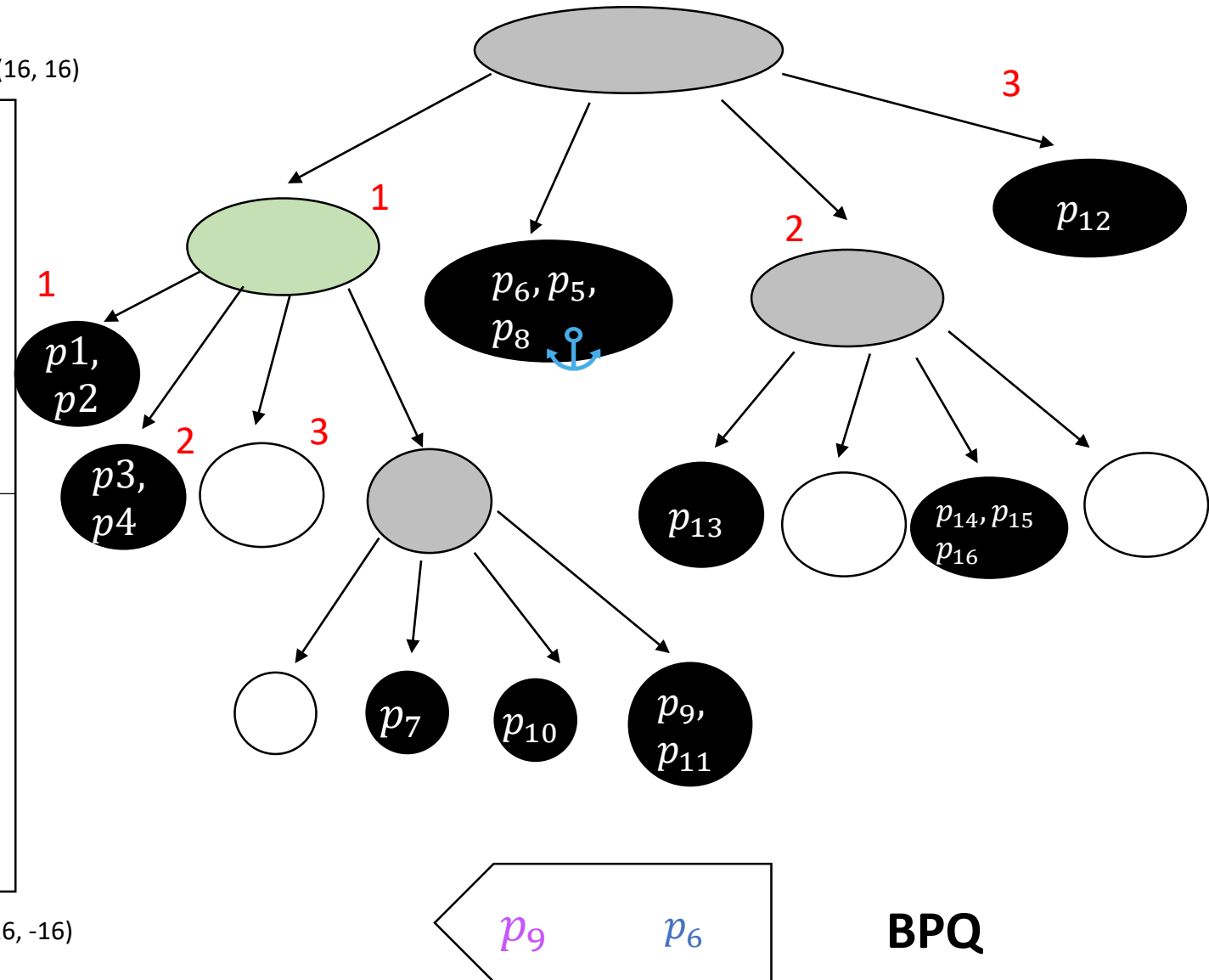
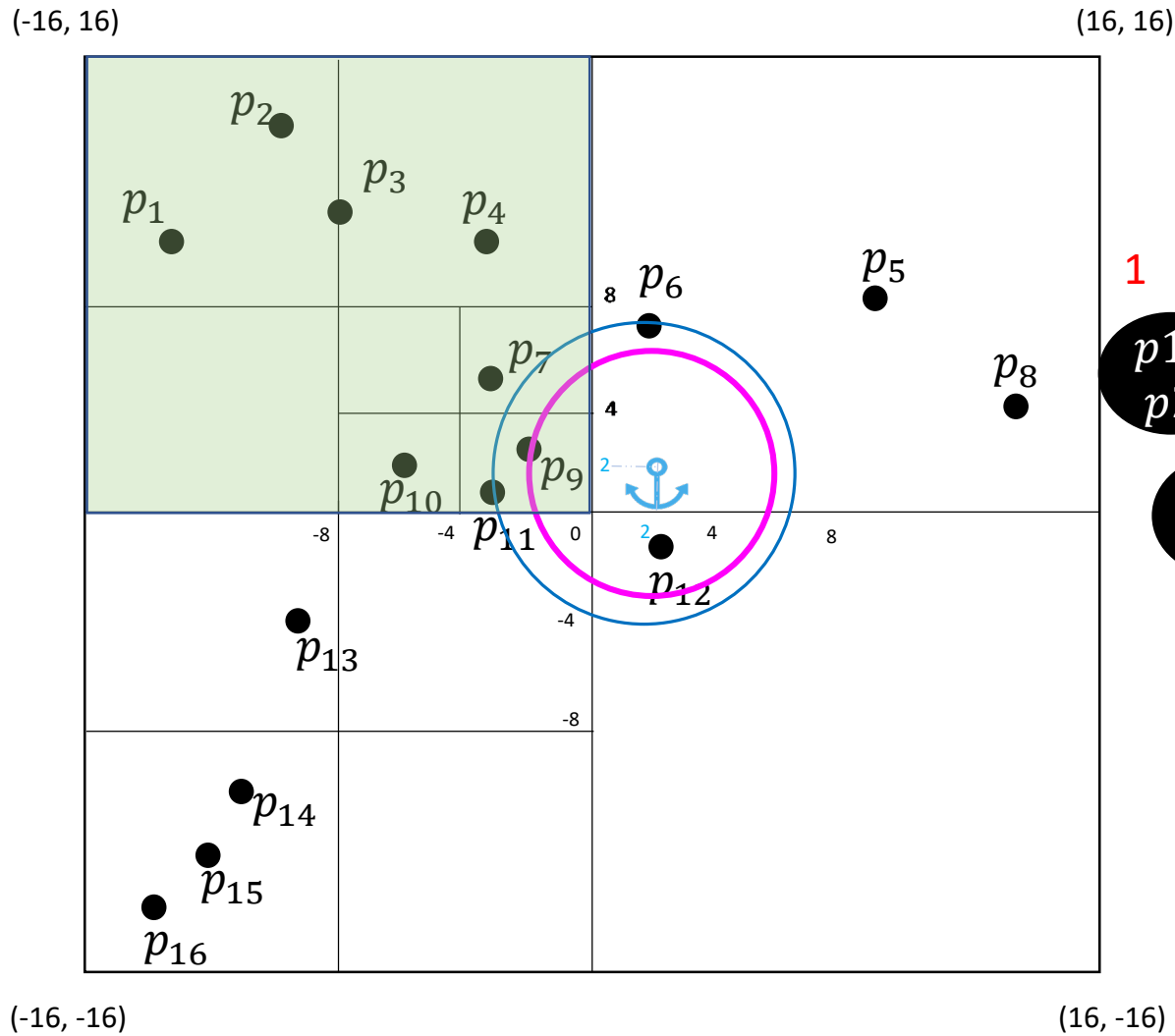
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anchor = (2, 2),  $k=2$ , Bucket size  $b = 3$



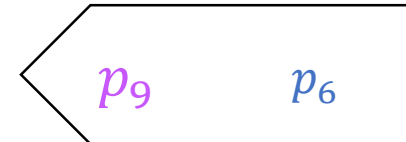
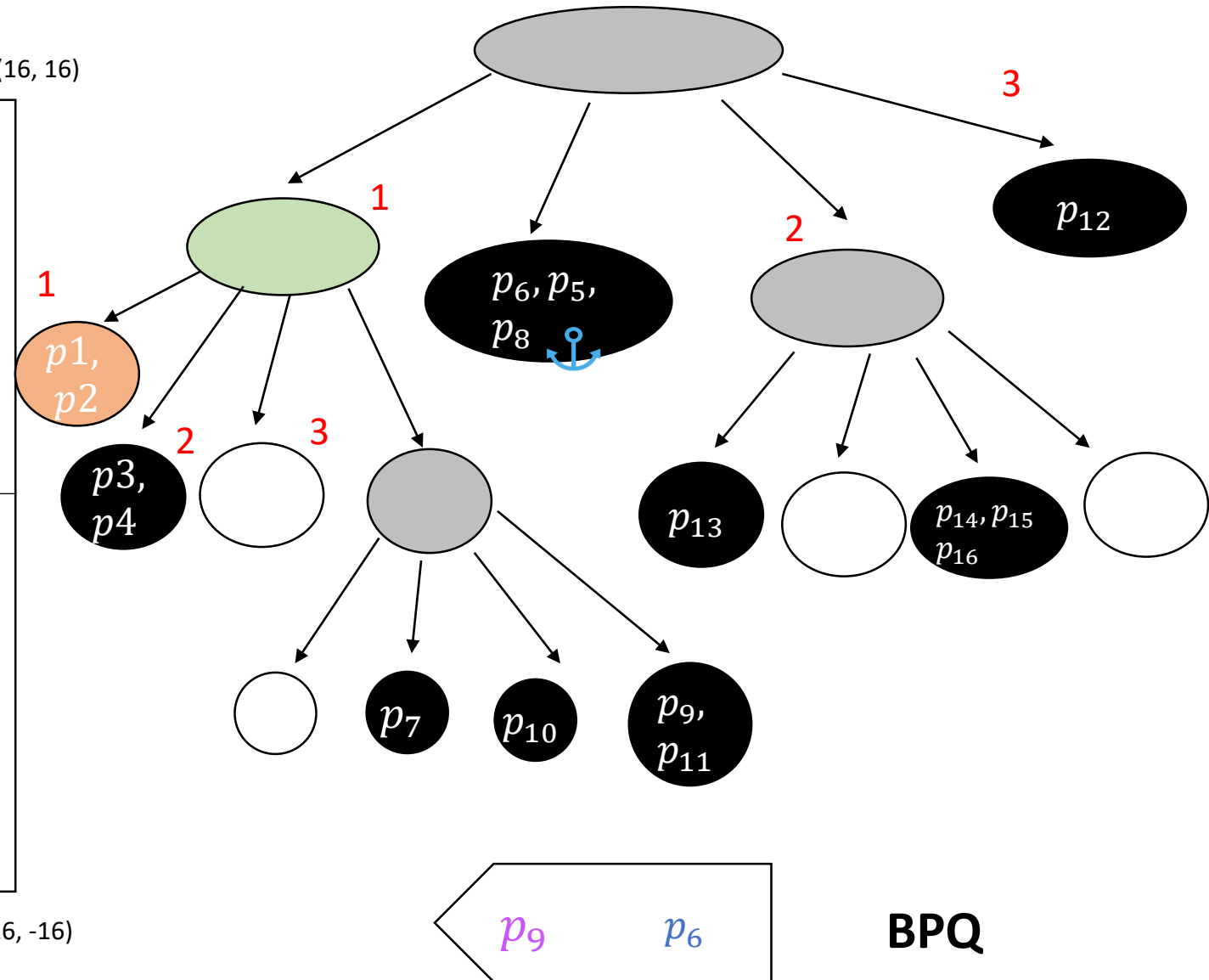
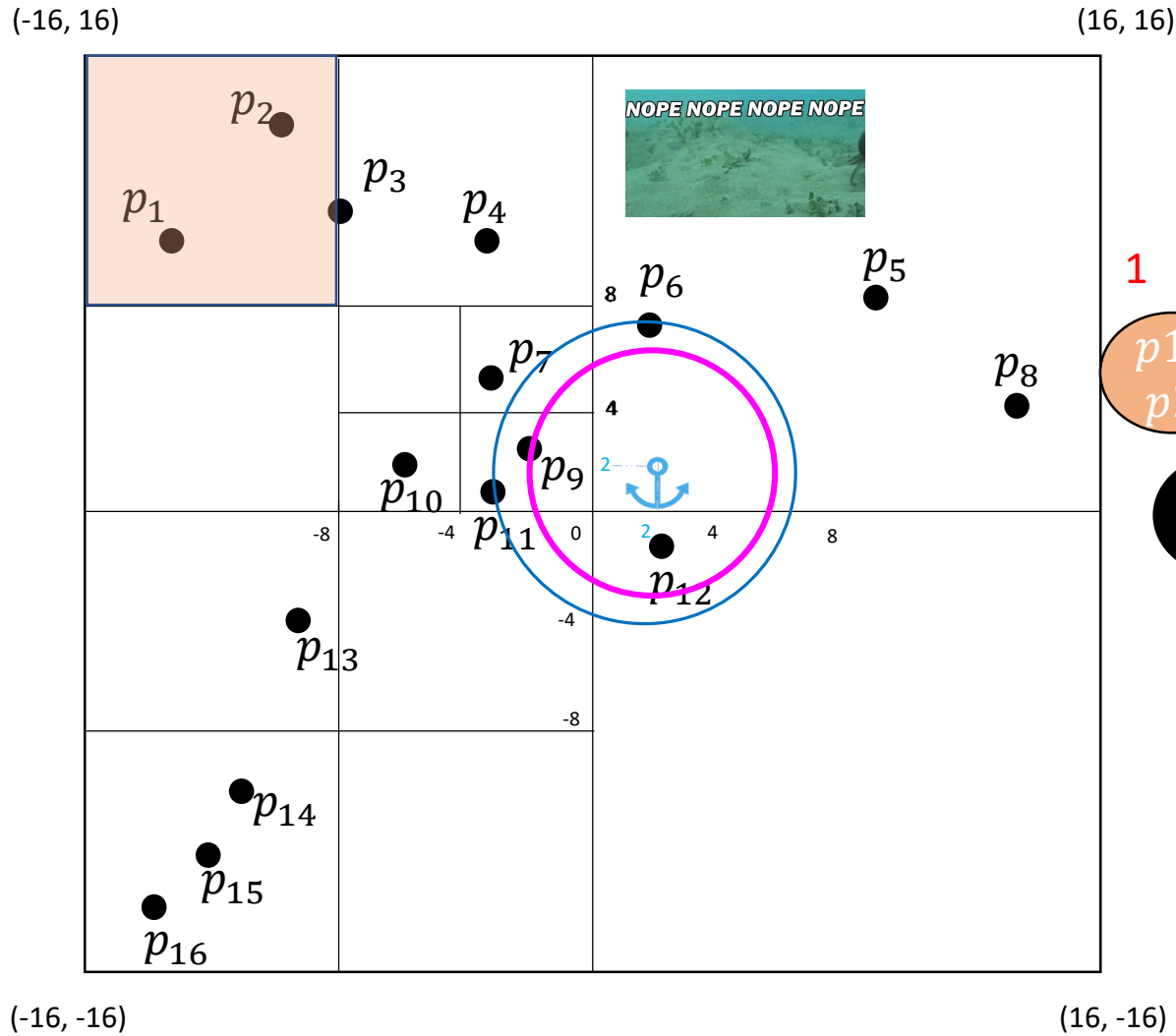
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anchor = (2, 2),  $k=2$ , Bucket size  $b = 3$



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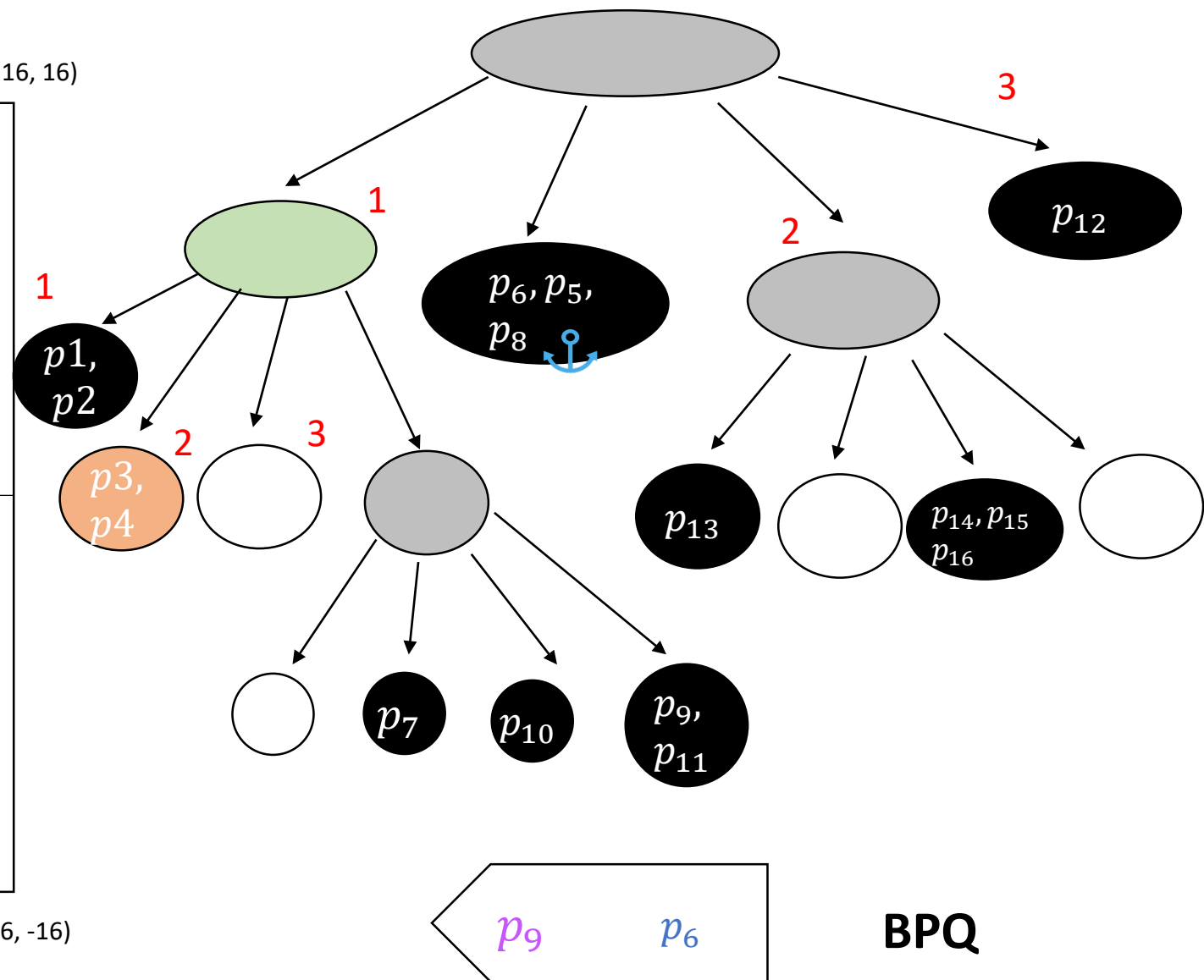
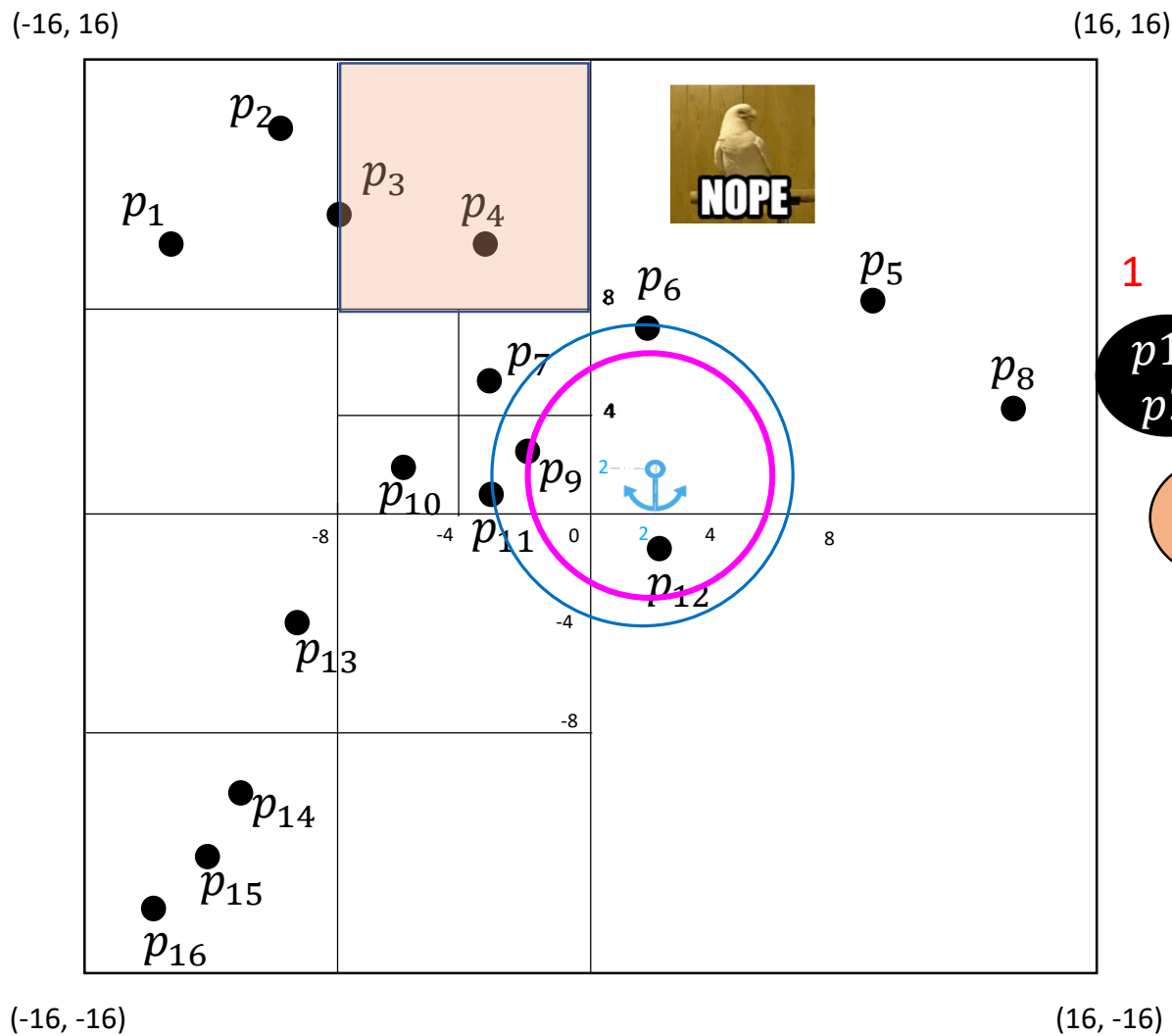
anchor = (2, 2),  $k=2$ , Bucket size  $b = 3$



BPQ

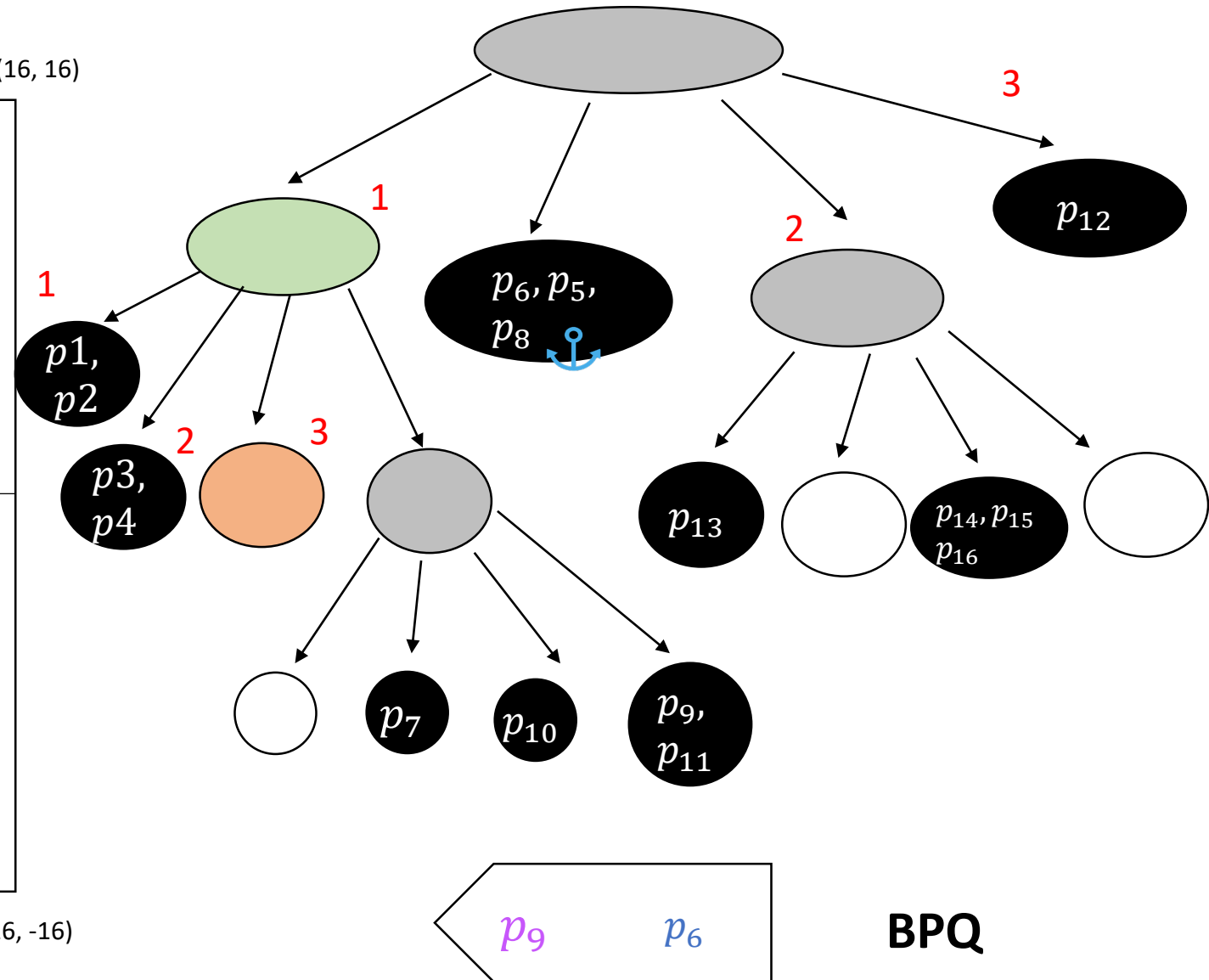
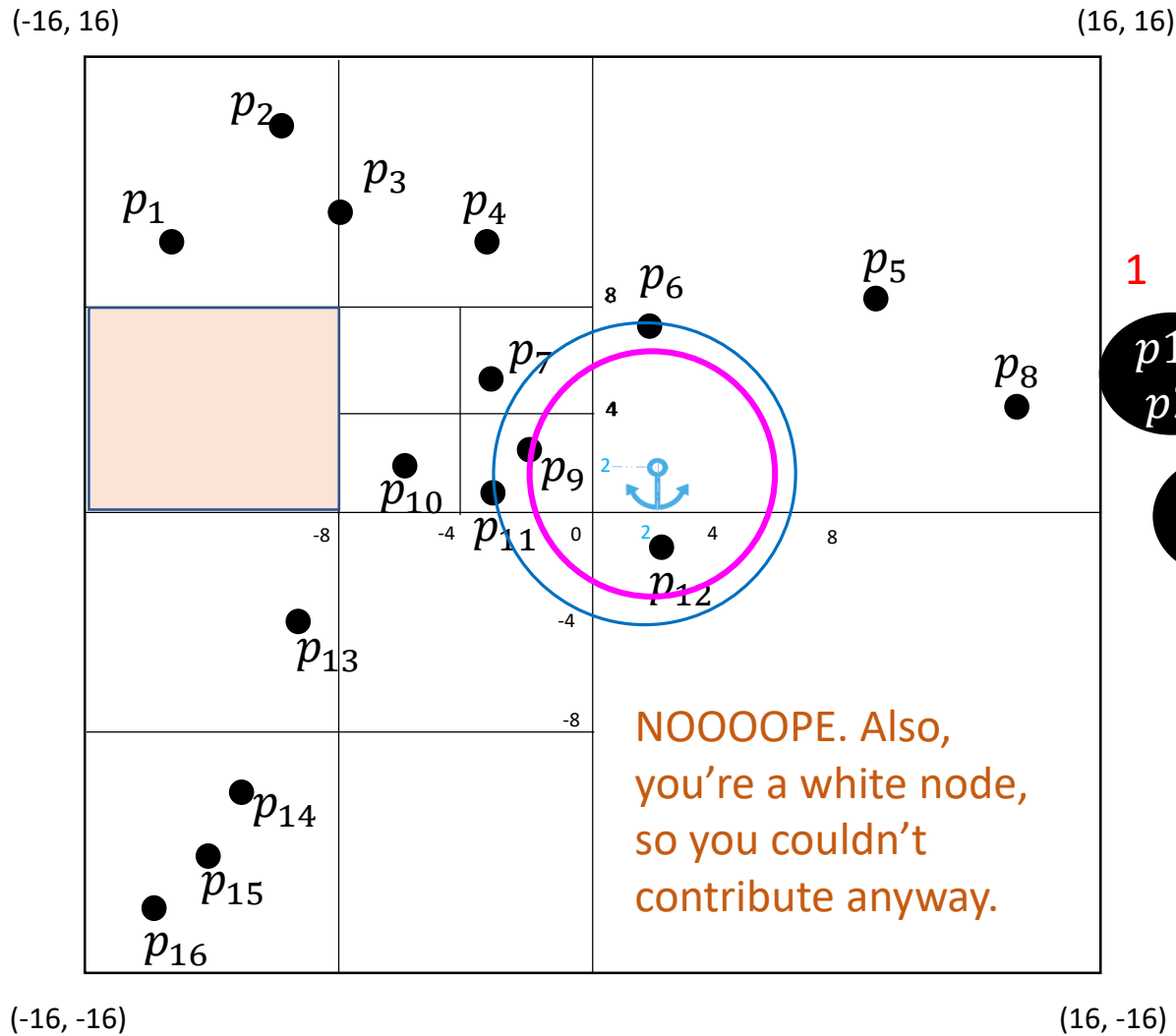
# Example of $k$ -NN query in PR-QuadTree

anchor = (2, 2),  $k=2$ , Bucket size  $b = 3$



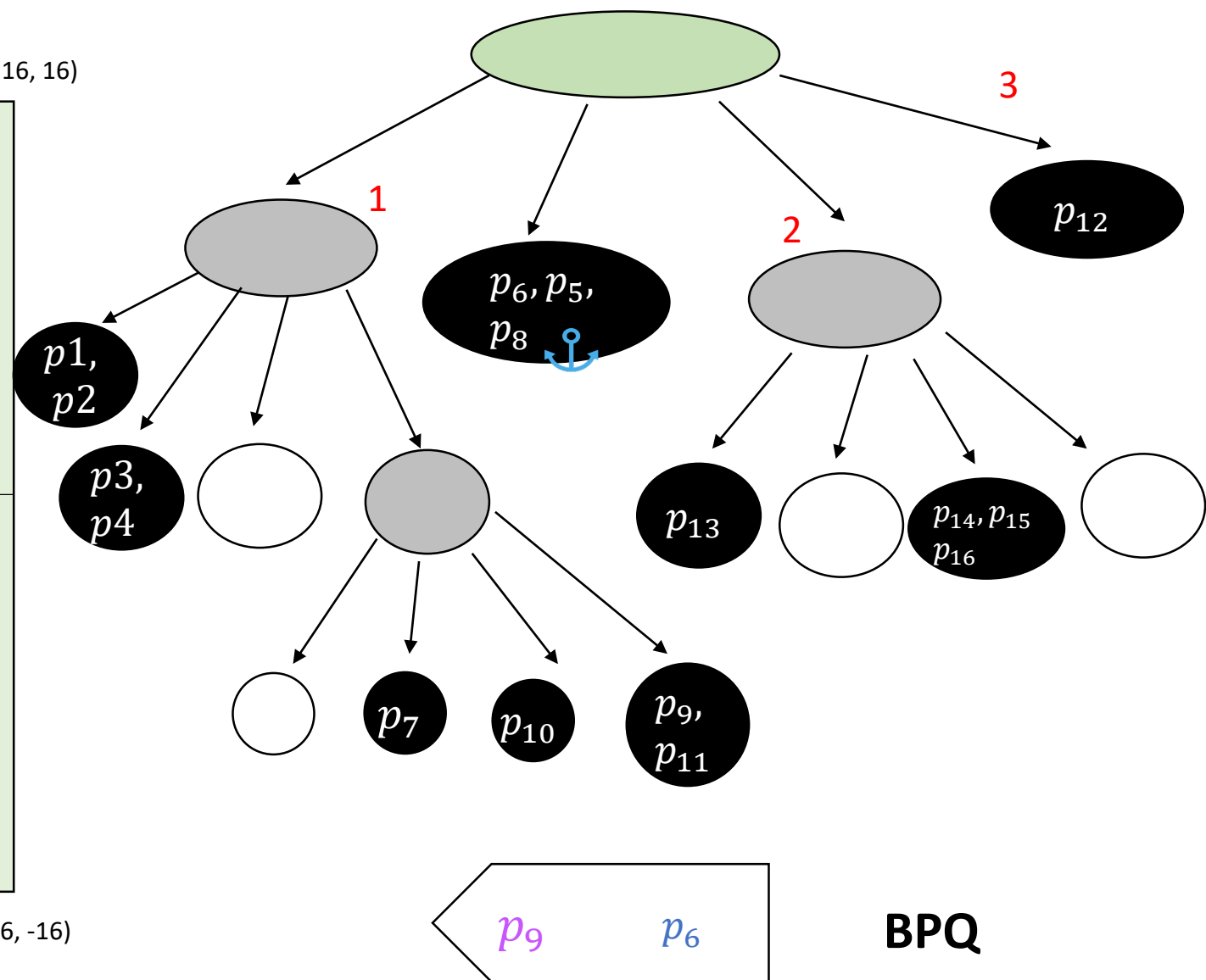
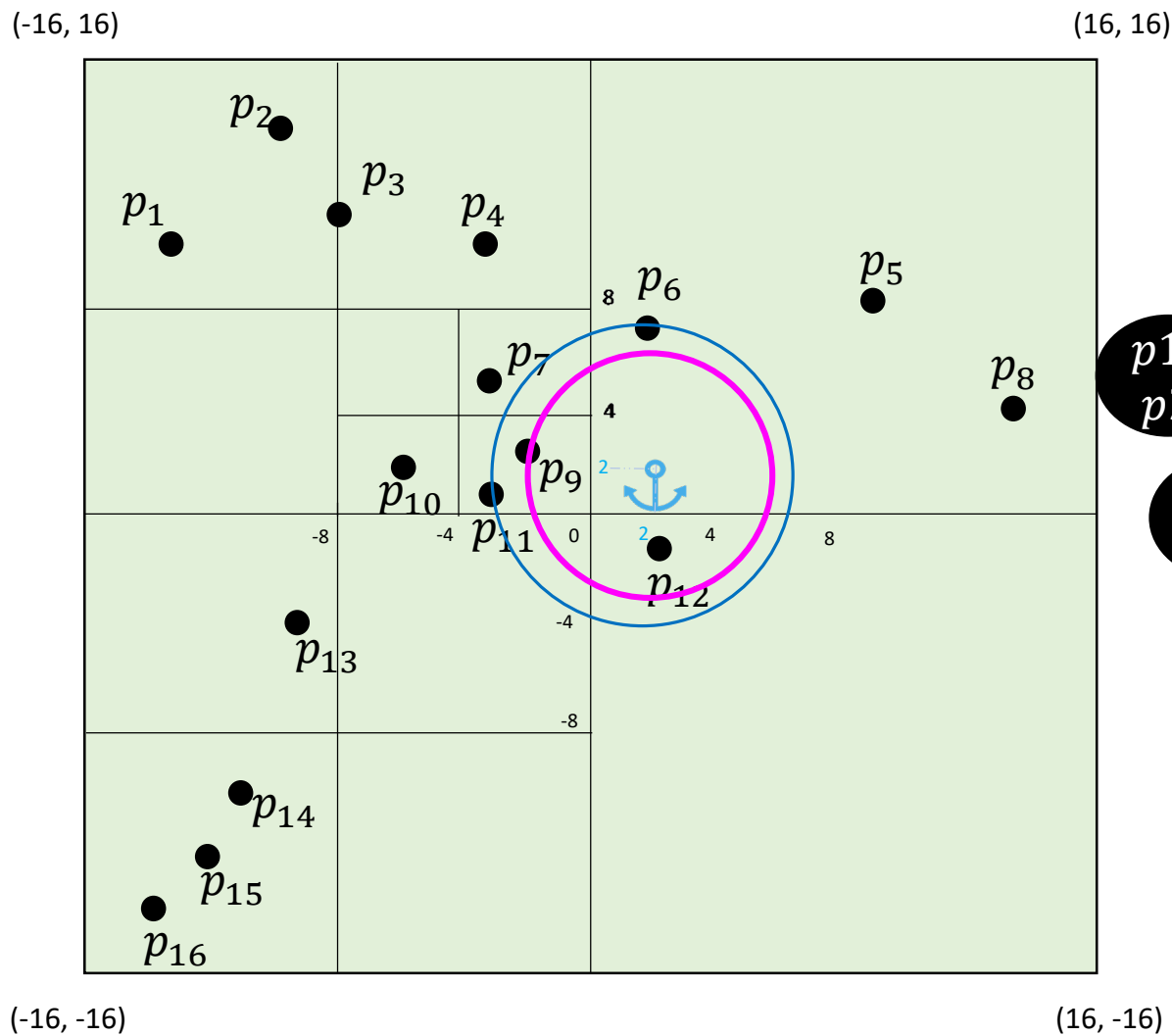
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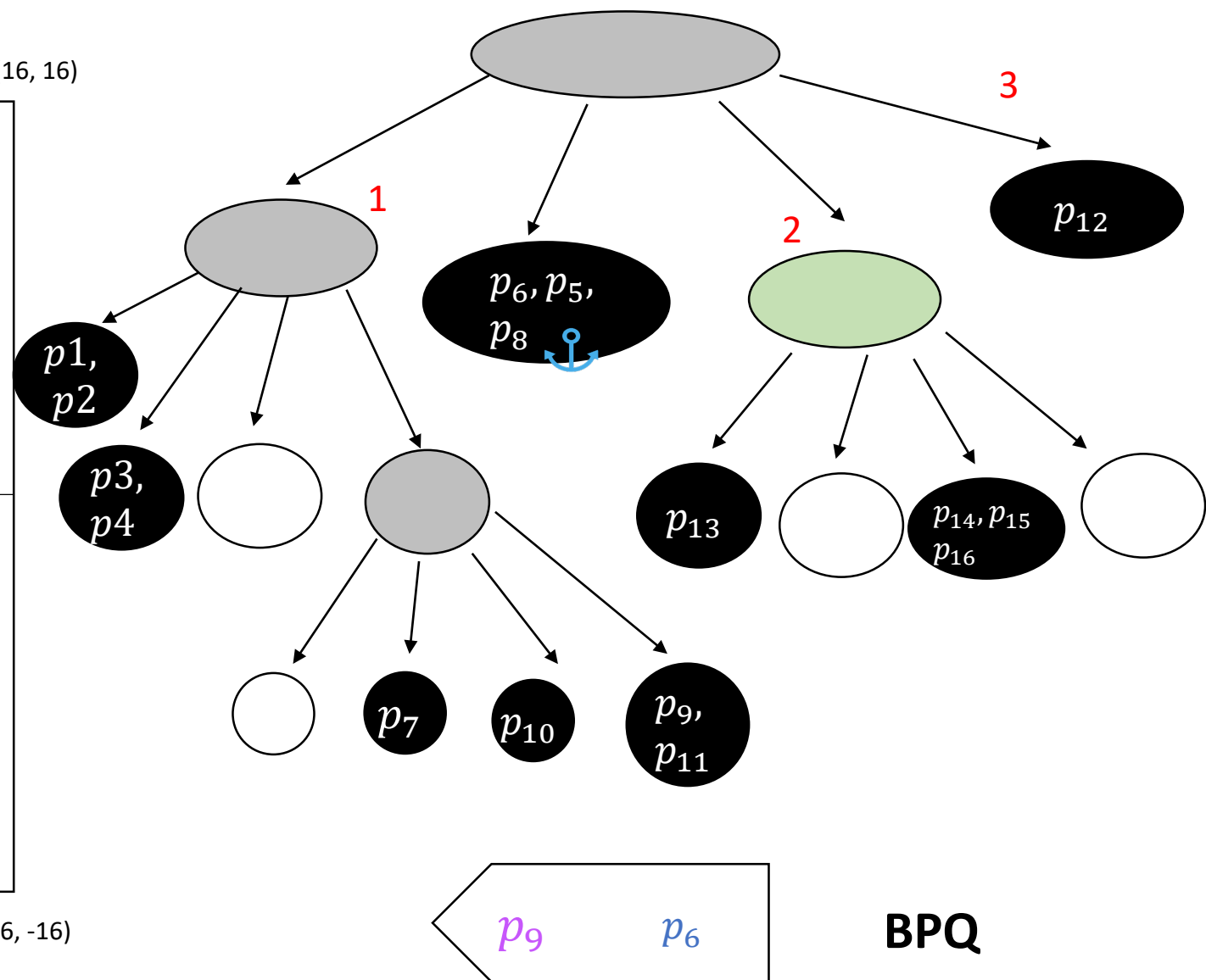
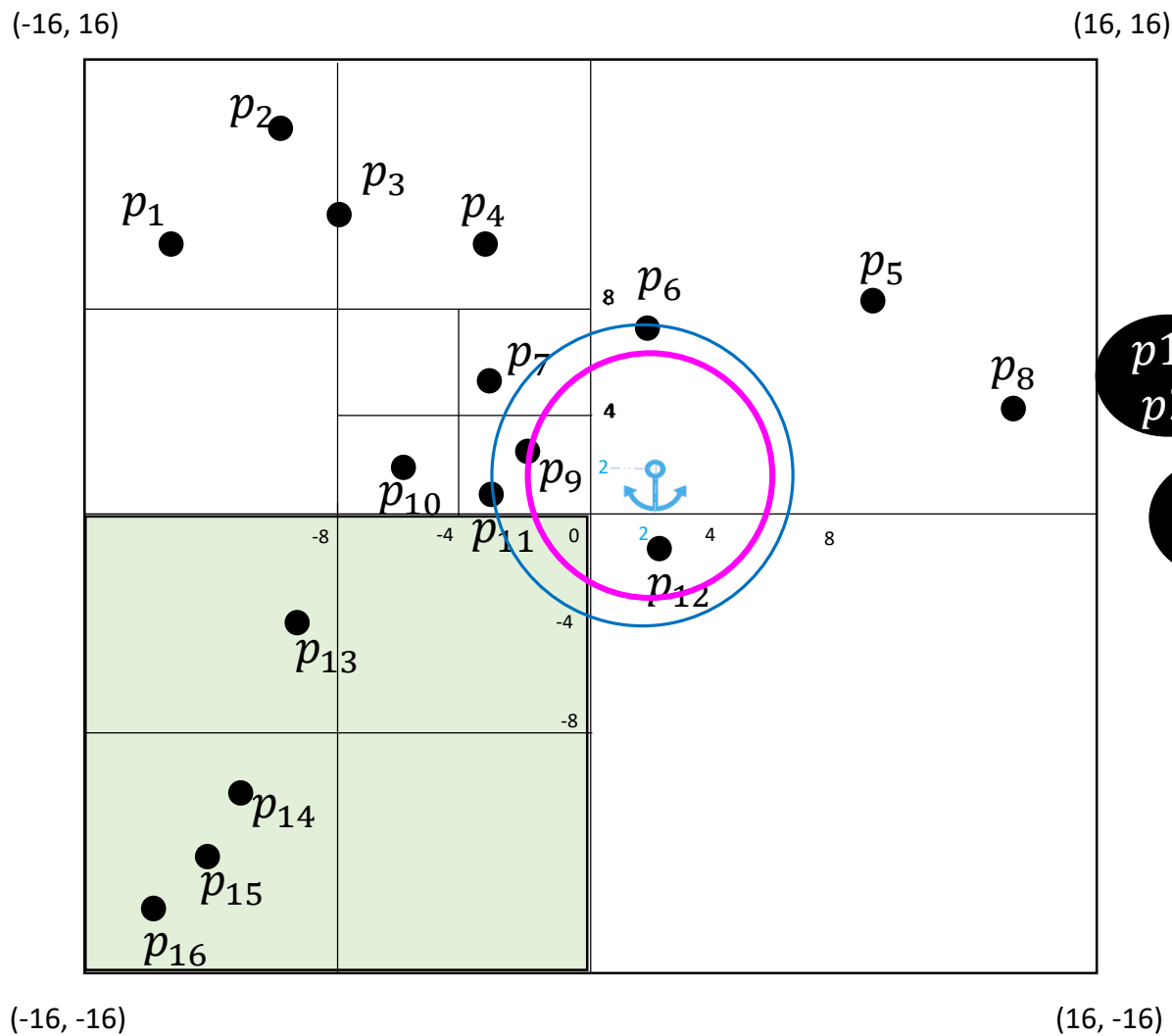
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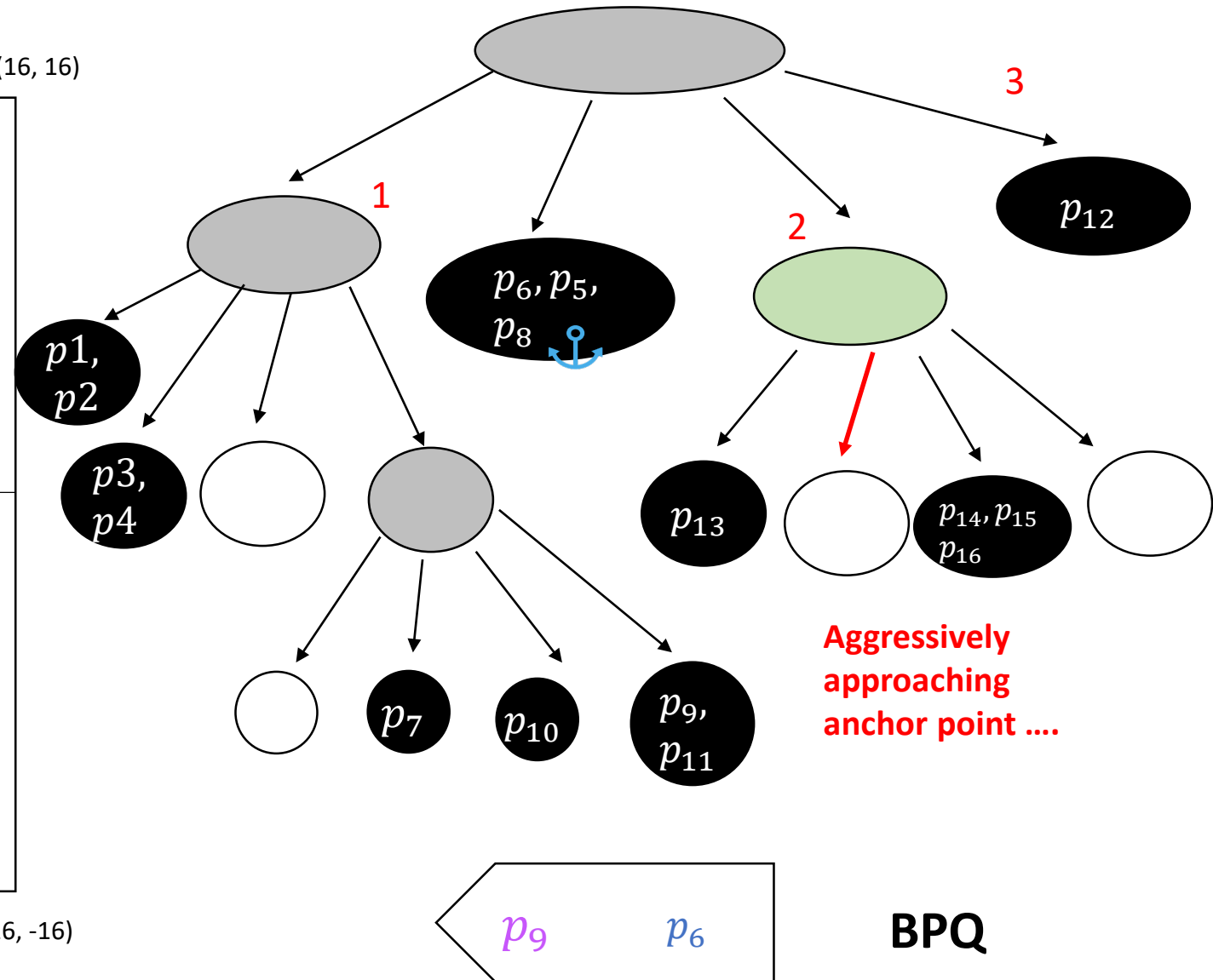
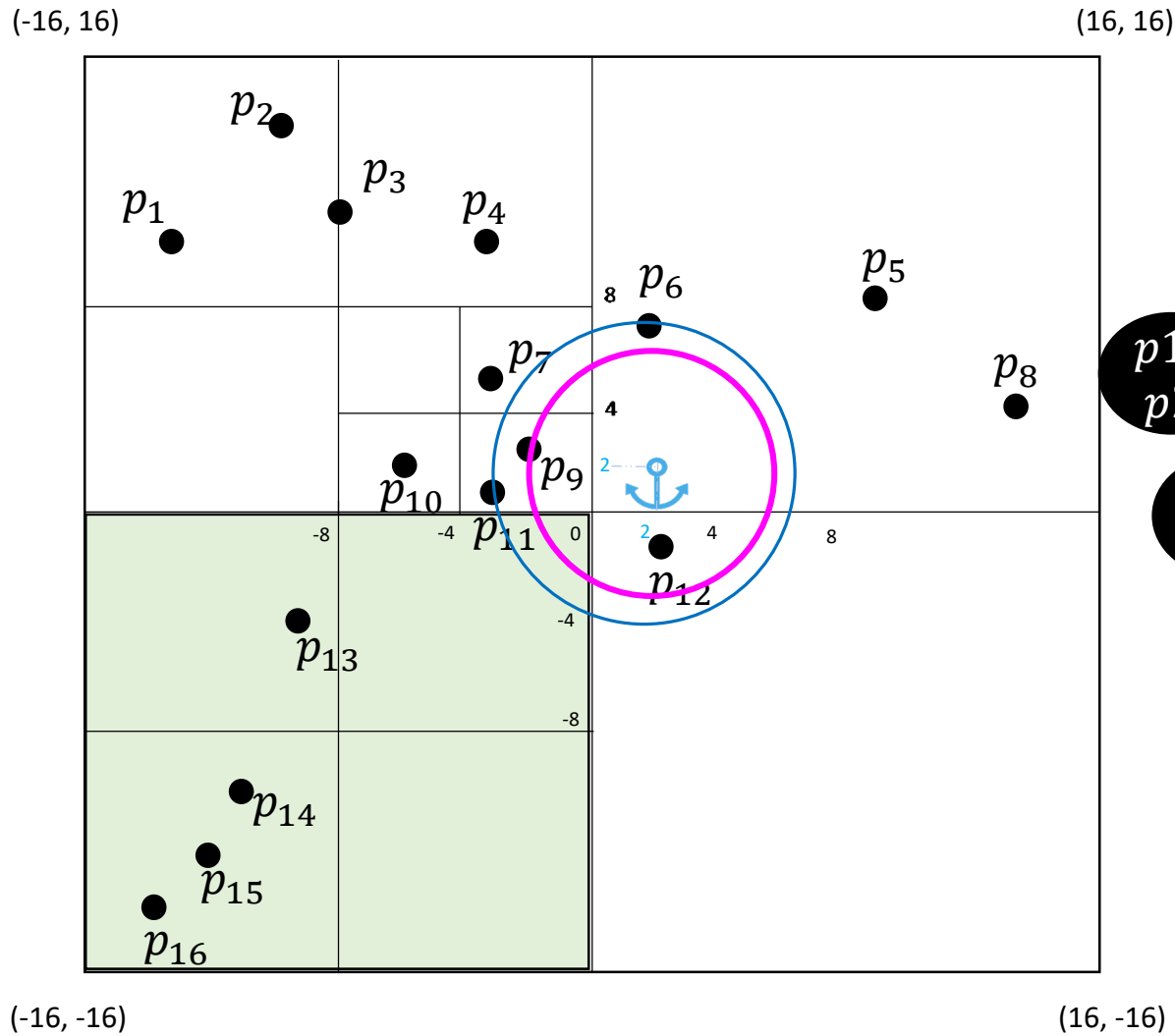
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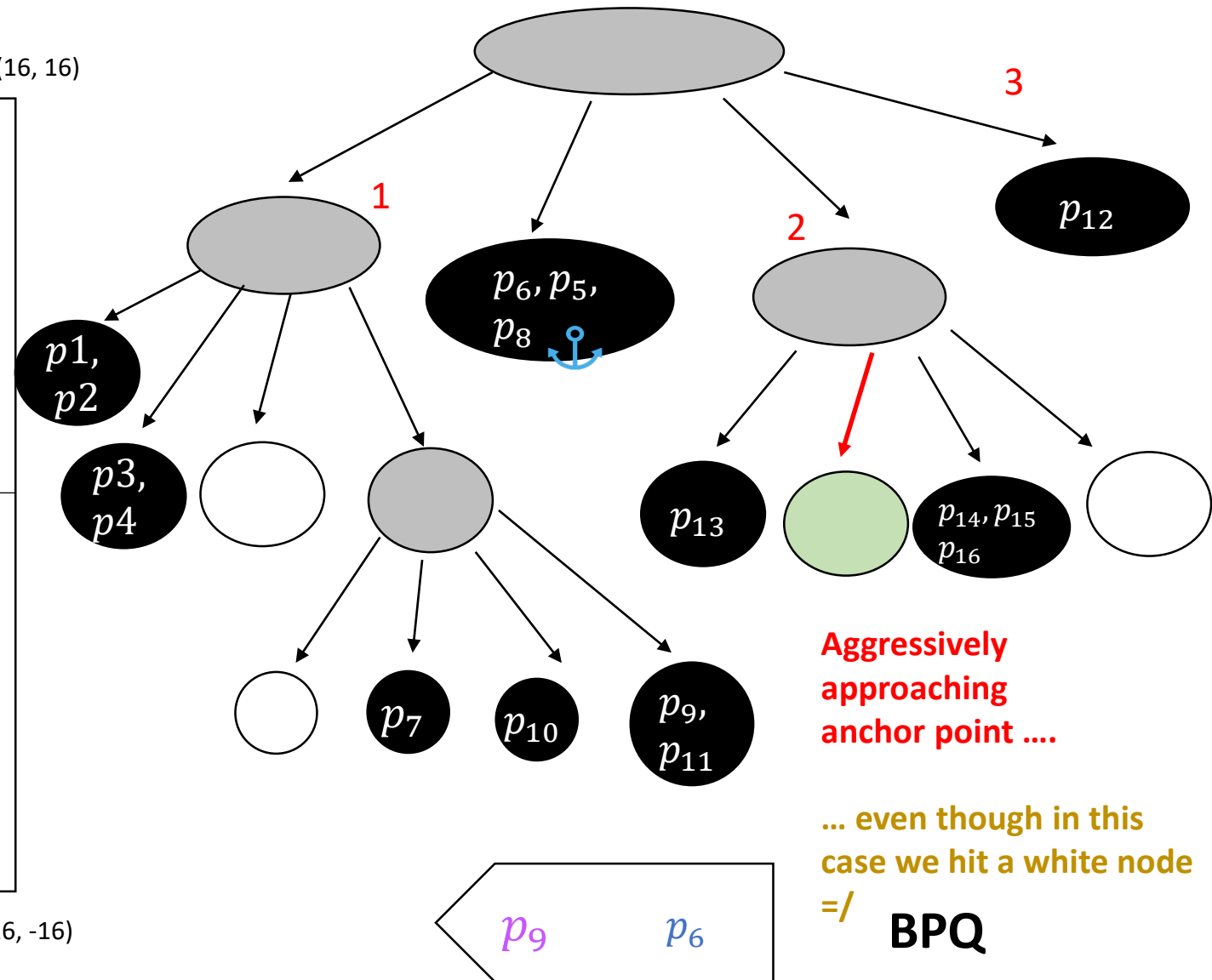
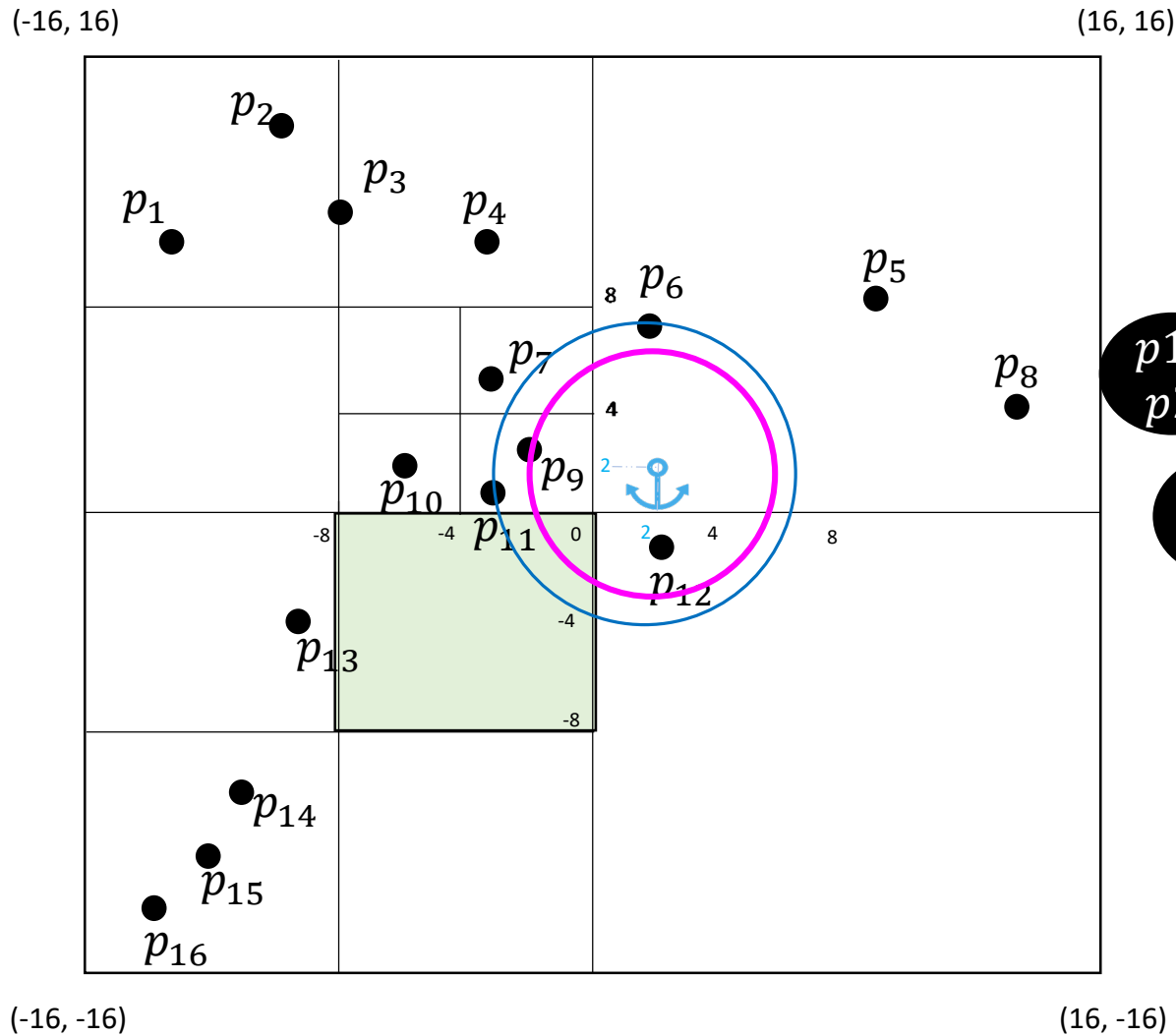
anchor = (2, 2),  $k=2$ , Bucket size  $b = 3$





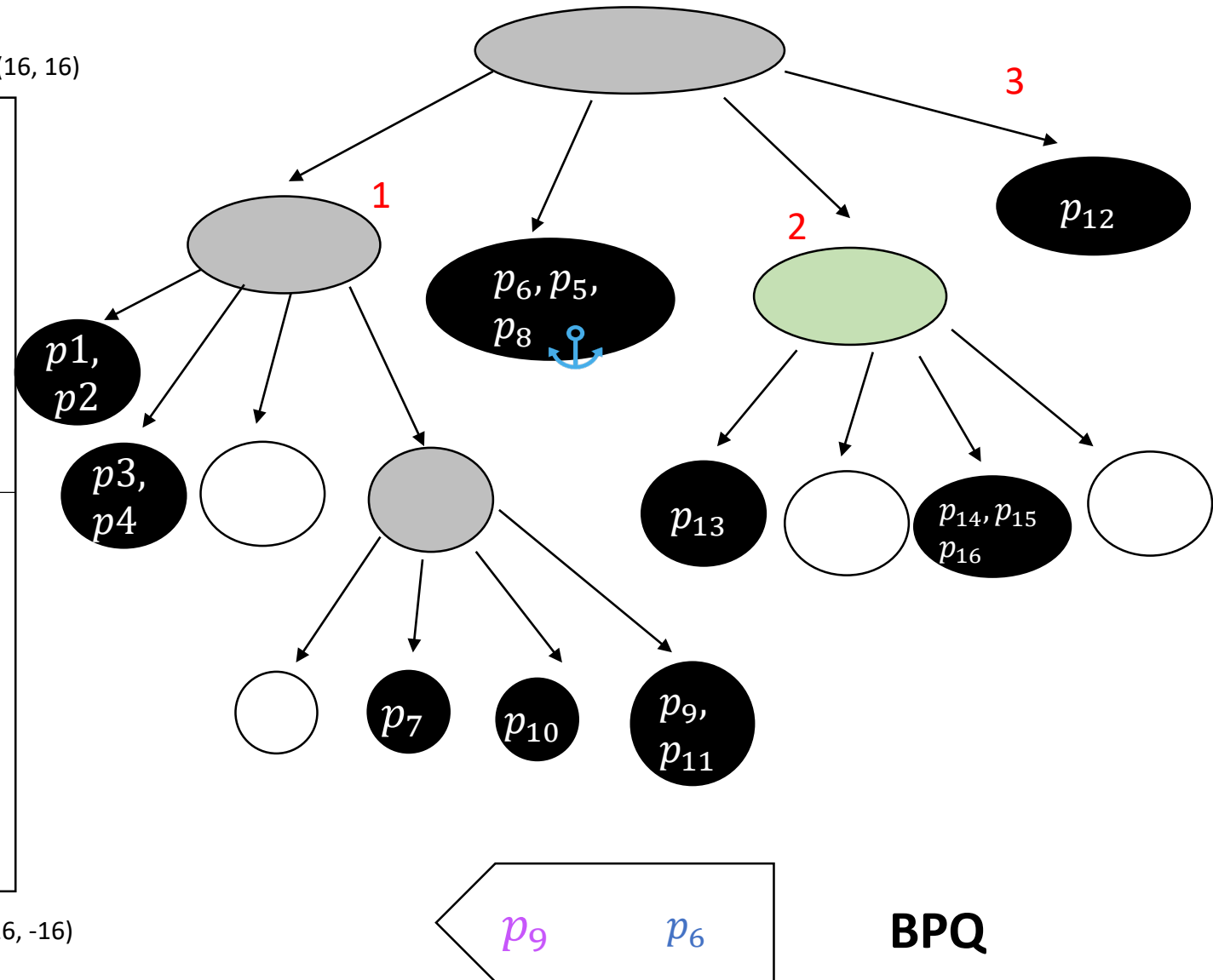
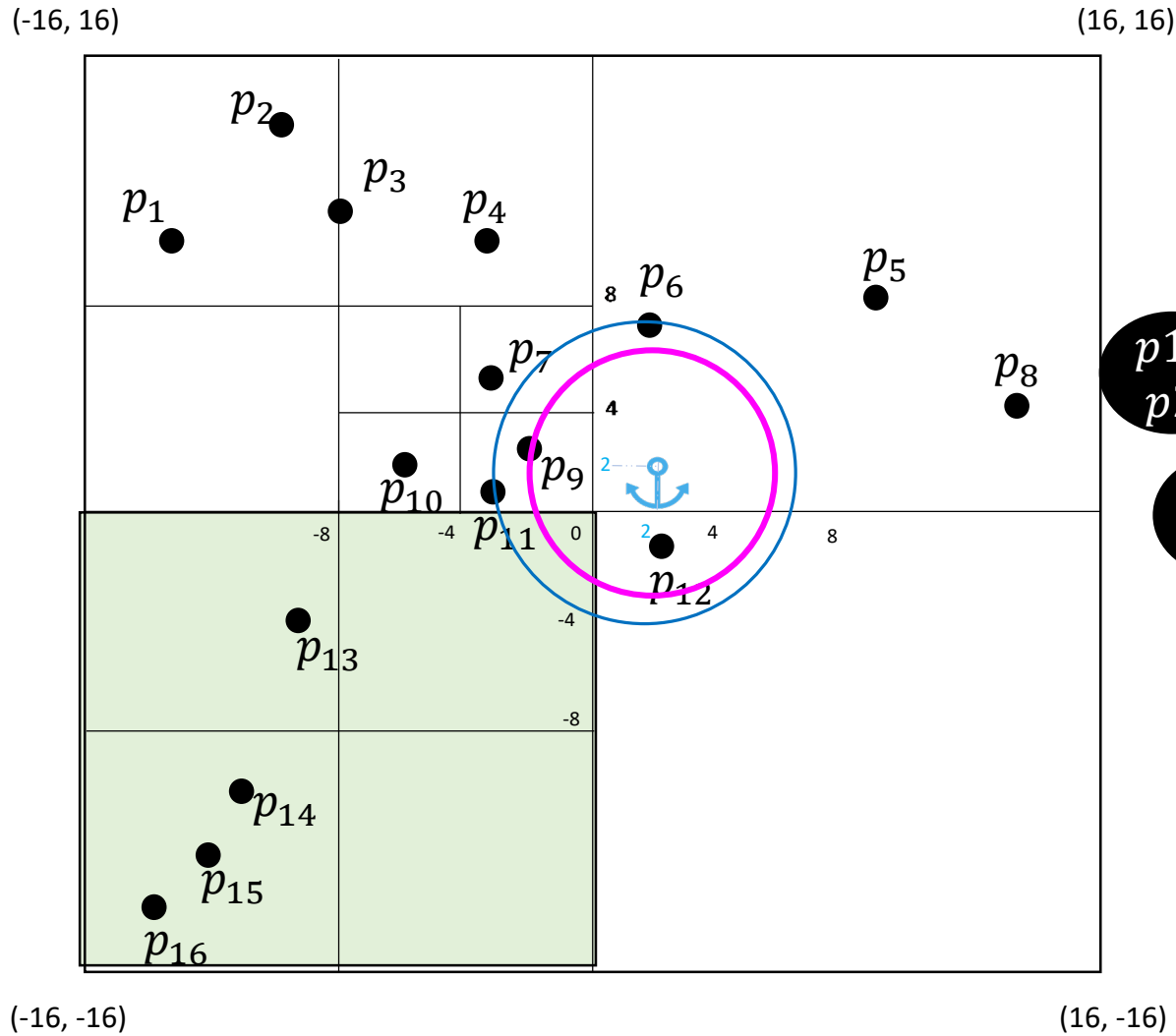
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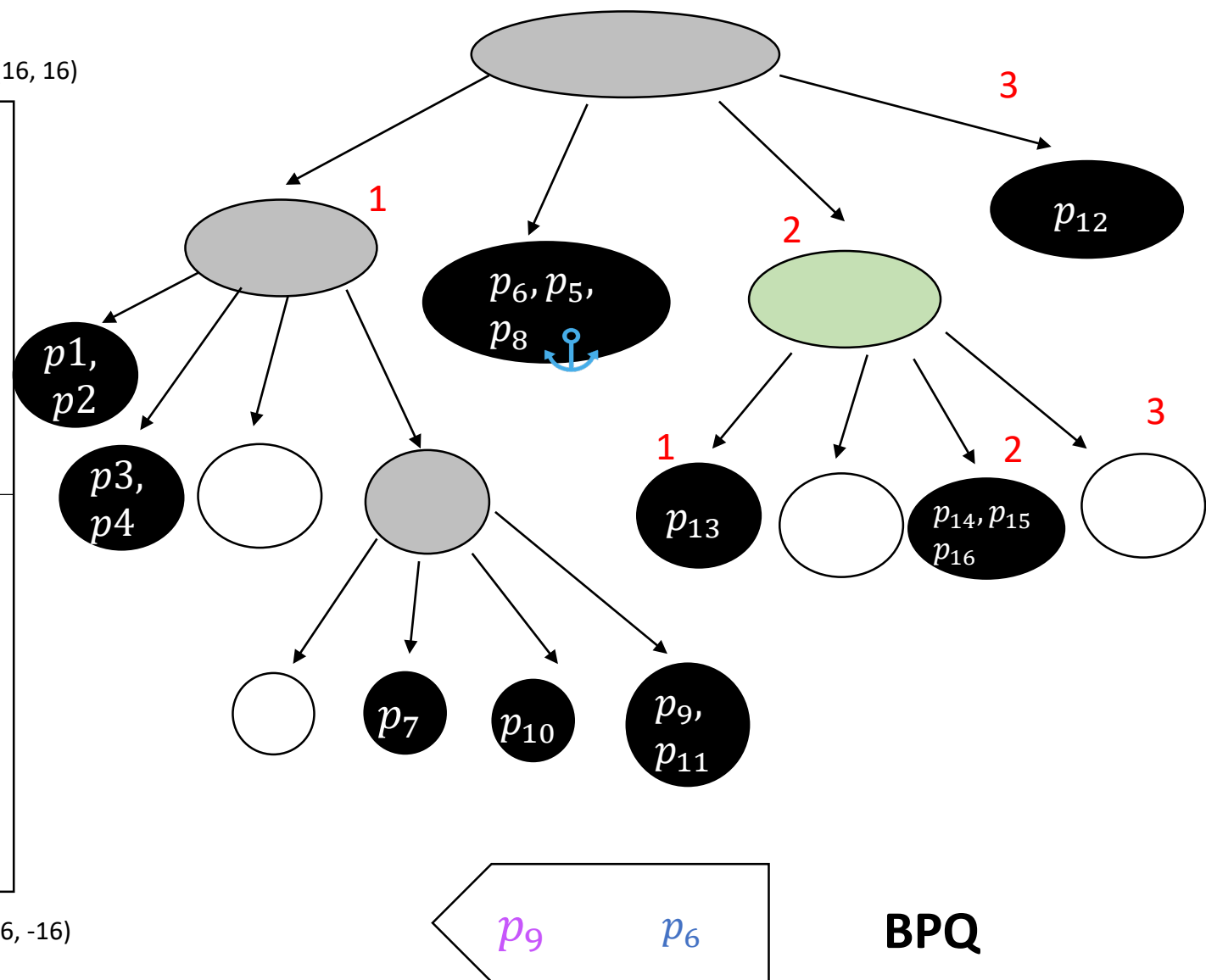
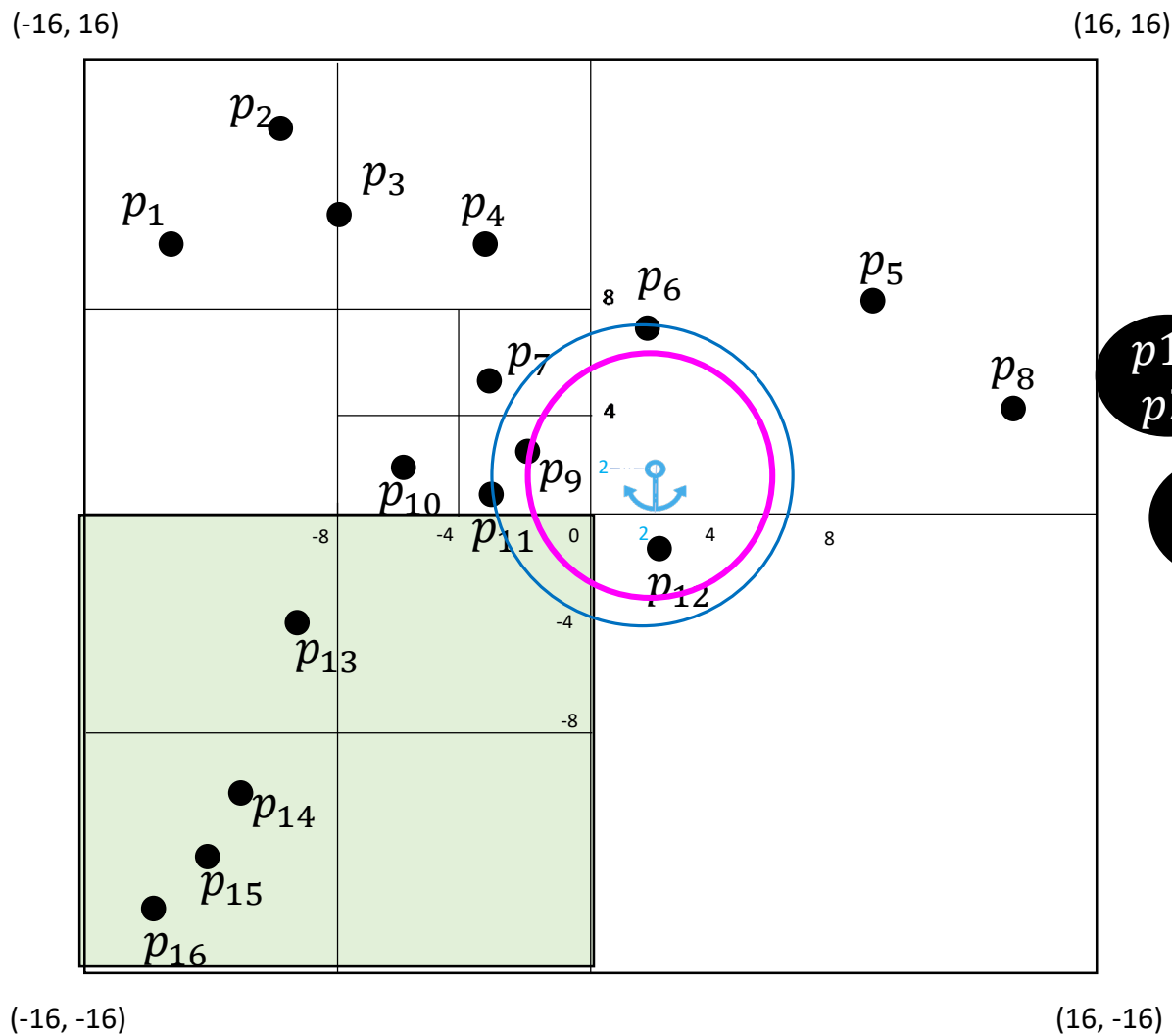
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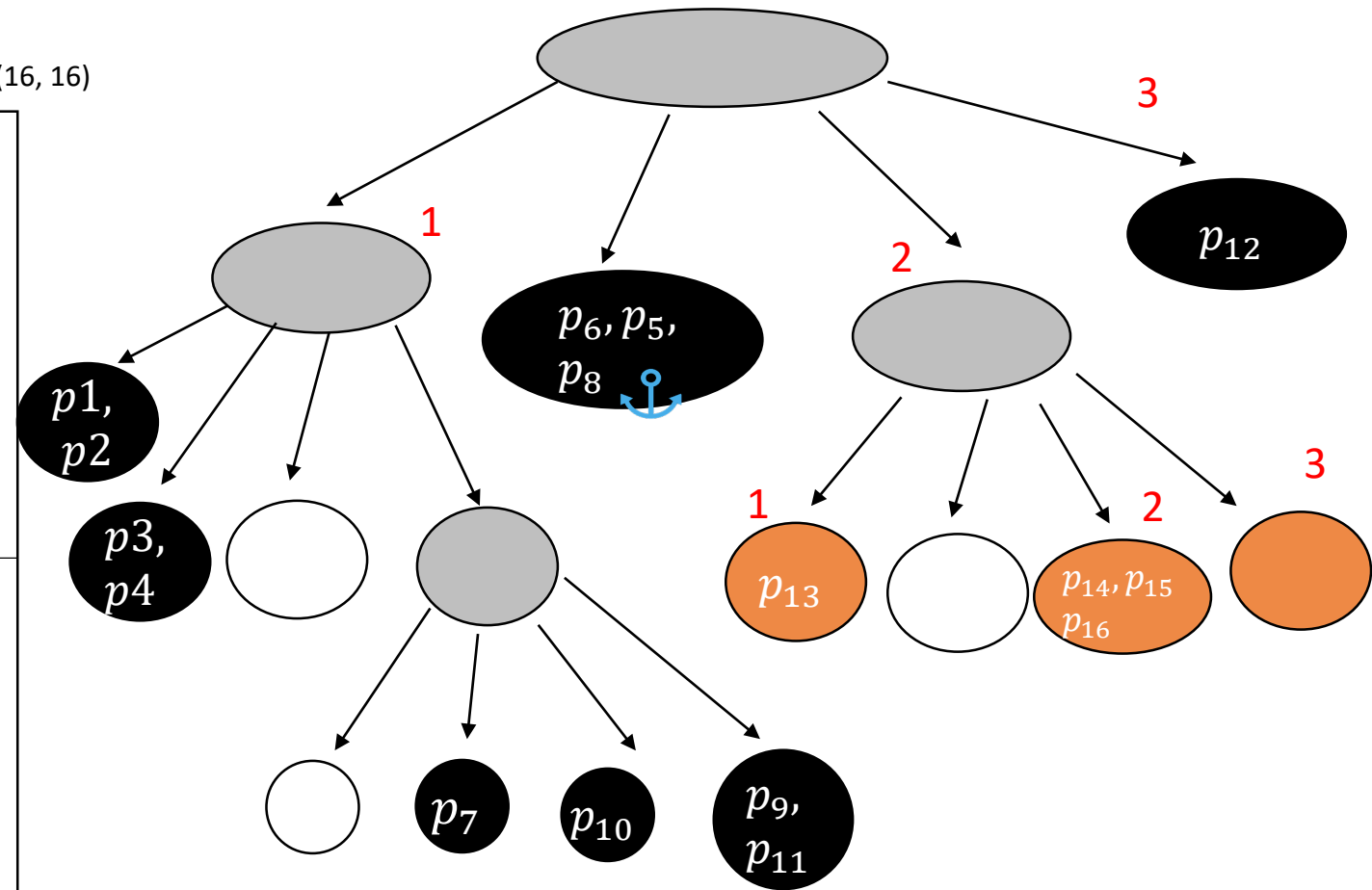
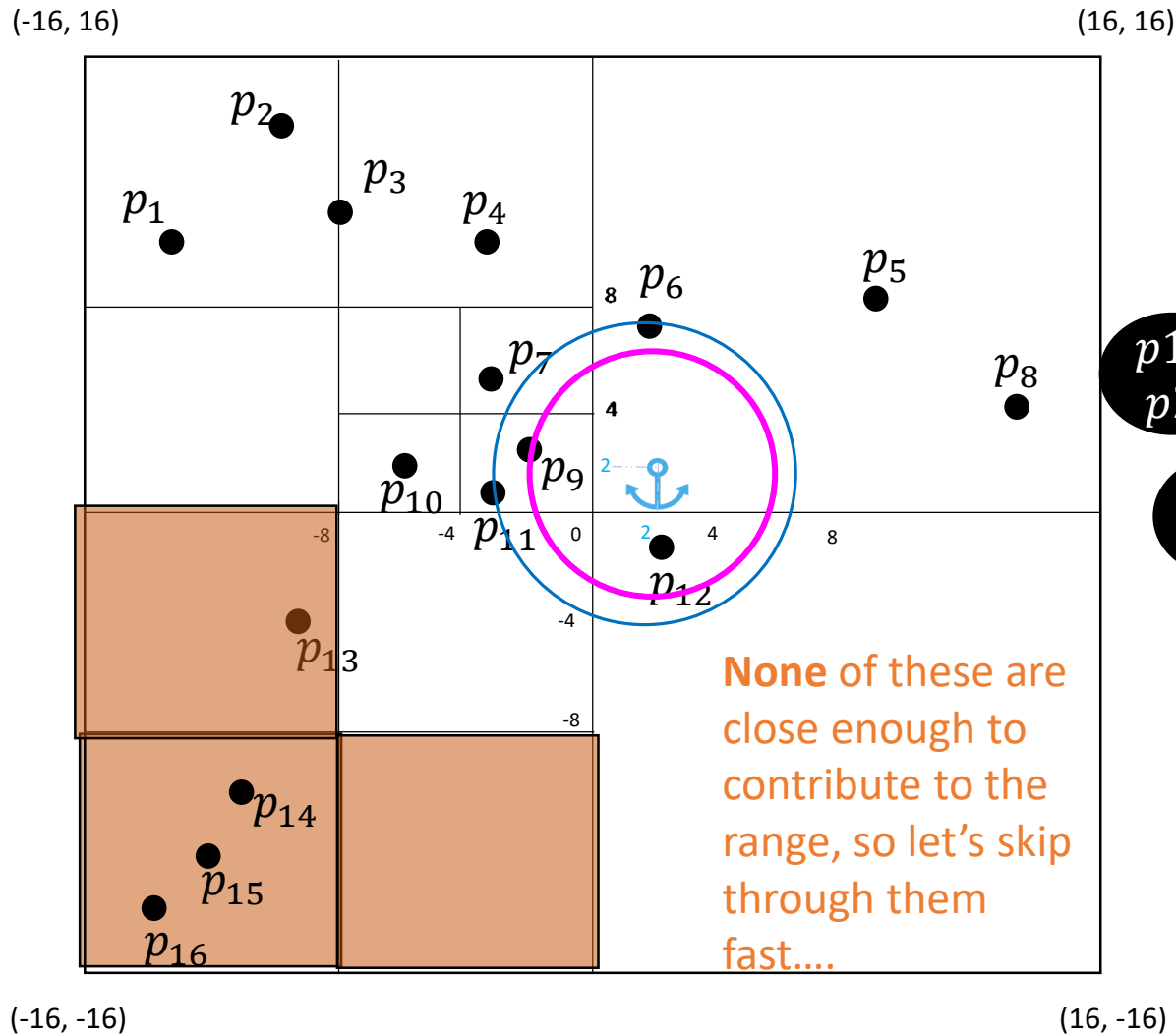
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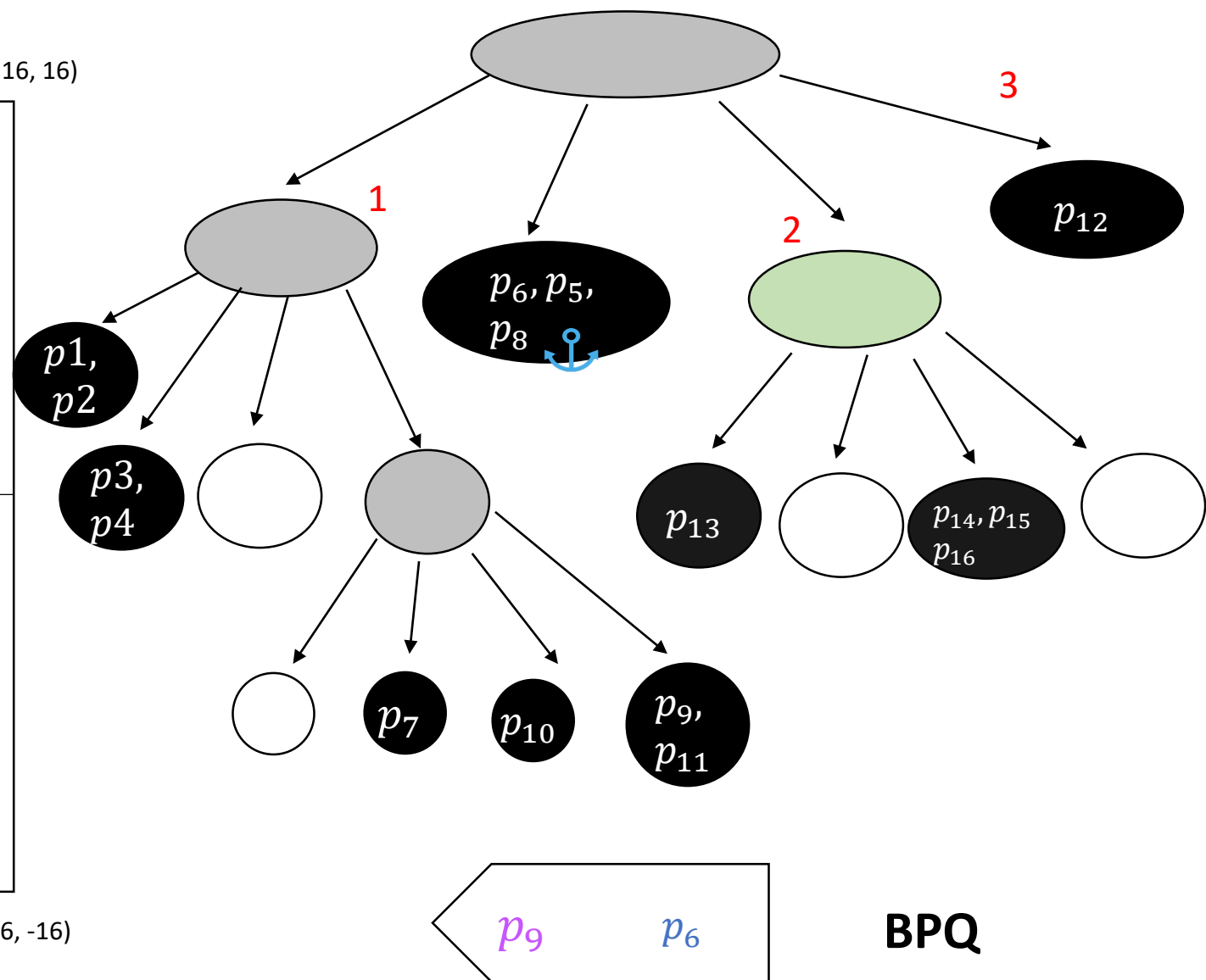
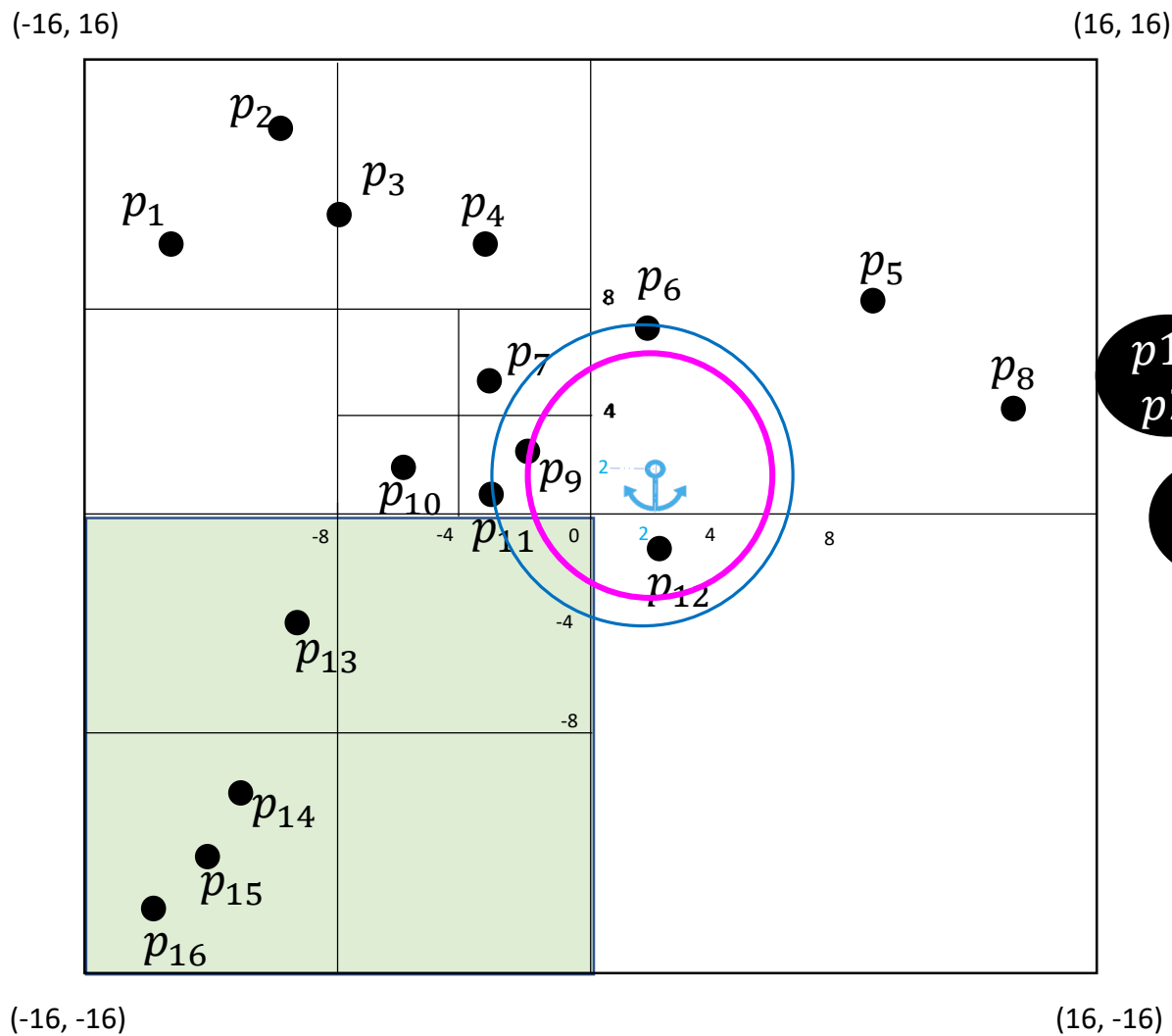
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BPQ

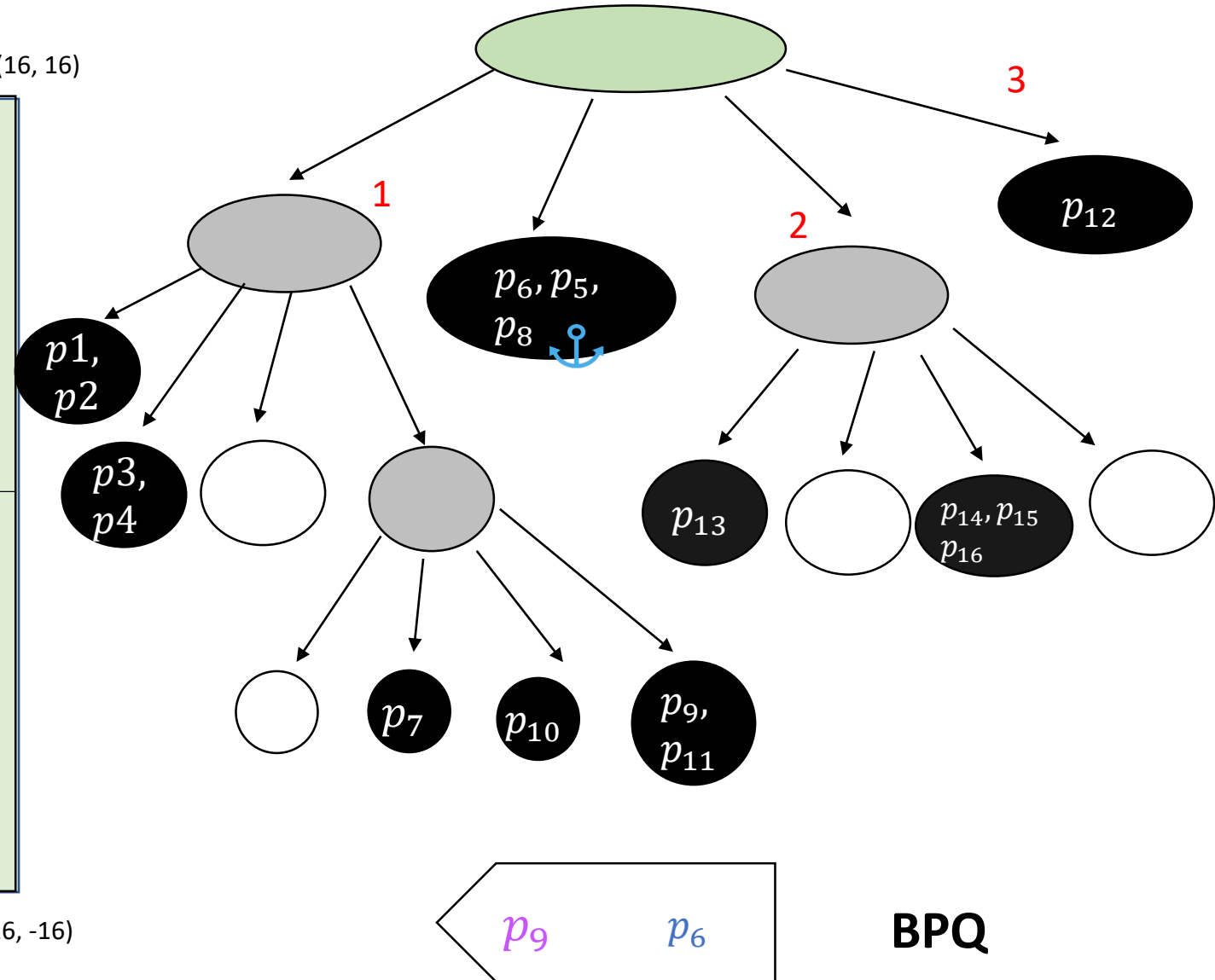
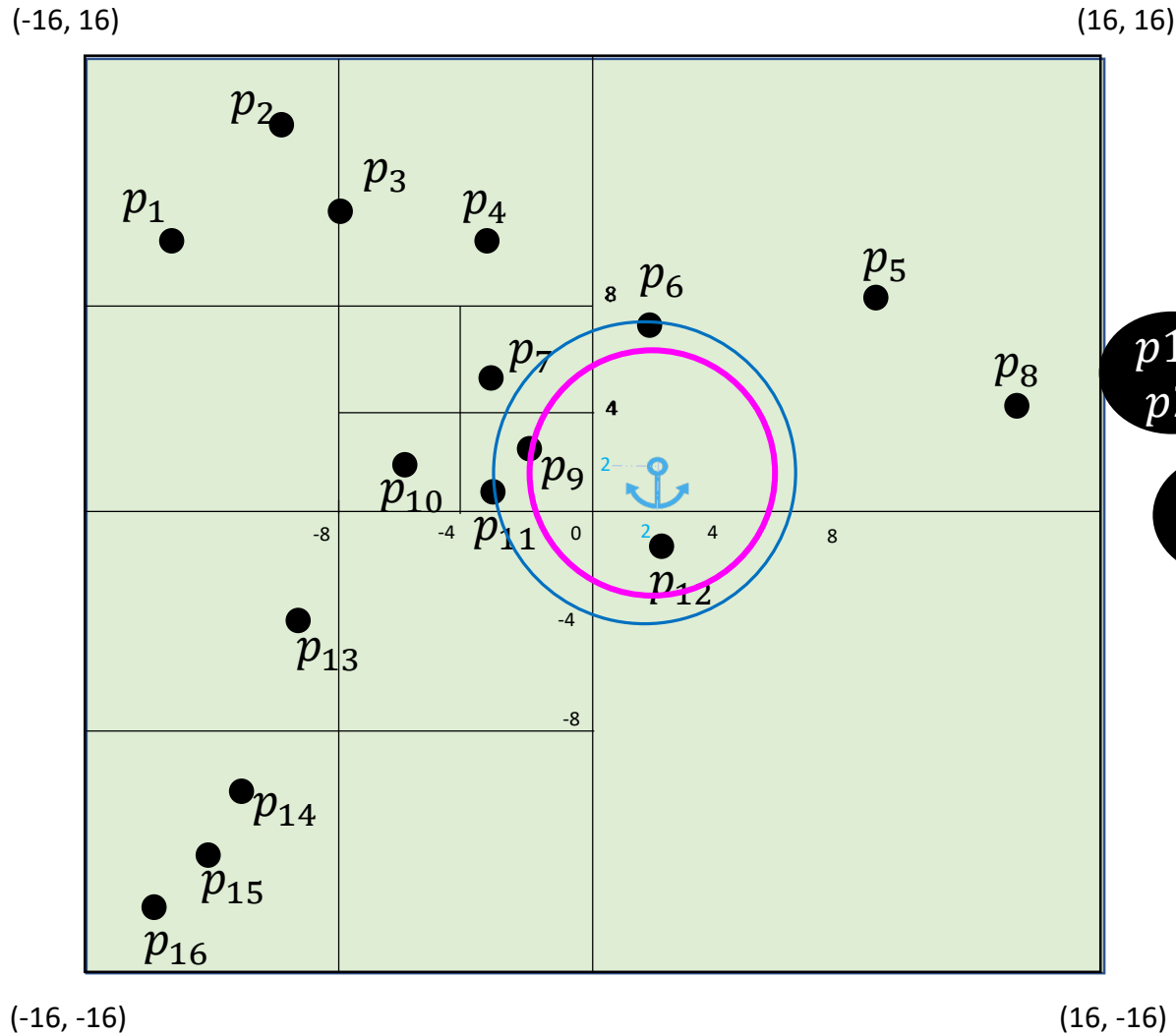
# Example of $k$ -NN query in PR-QuadTree

anchor = (2, 2),  $k=2$ , Bucket size  $b = 3$



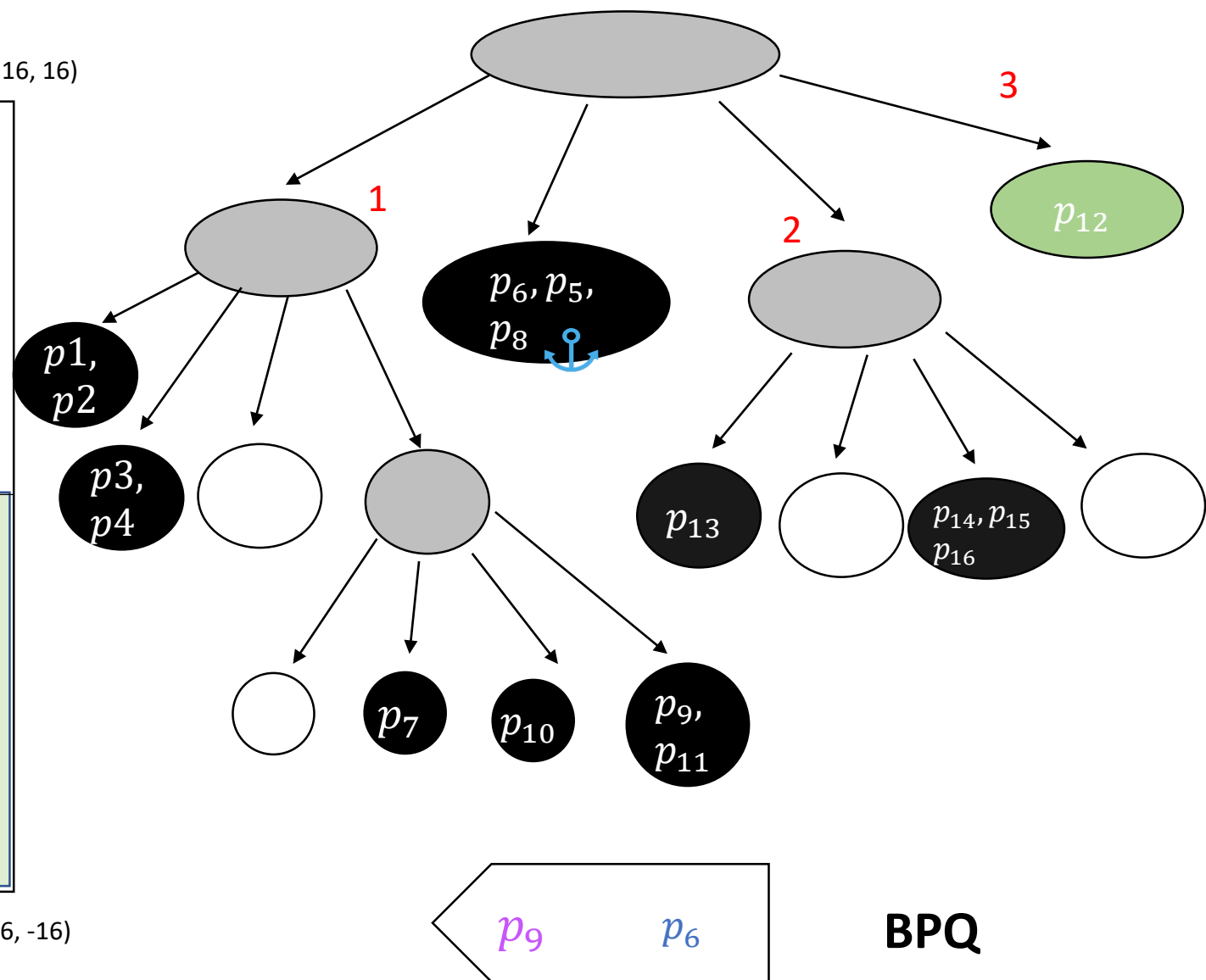
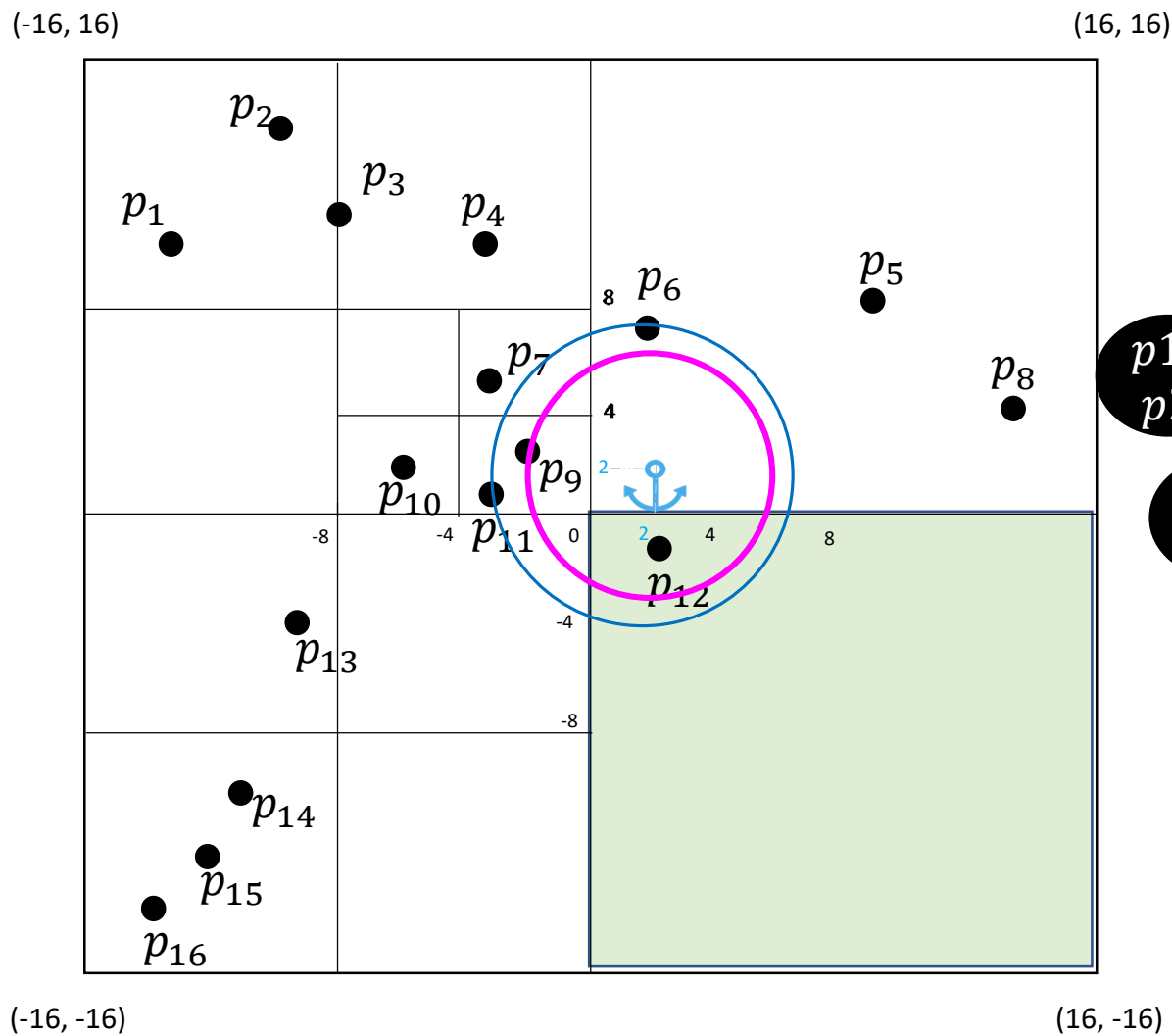
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# Example of $k$ -NN query in PR-QuadTree

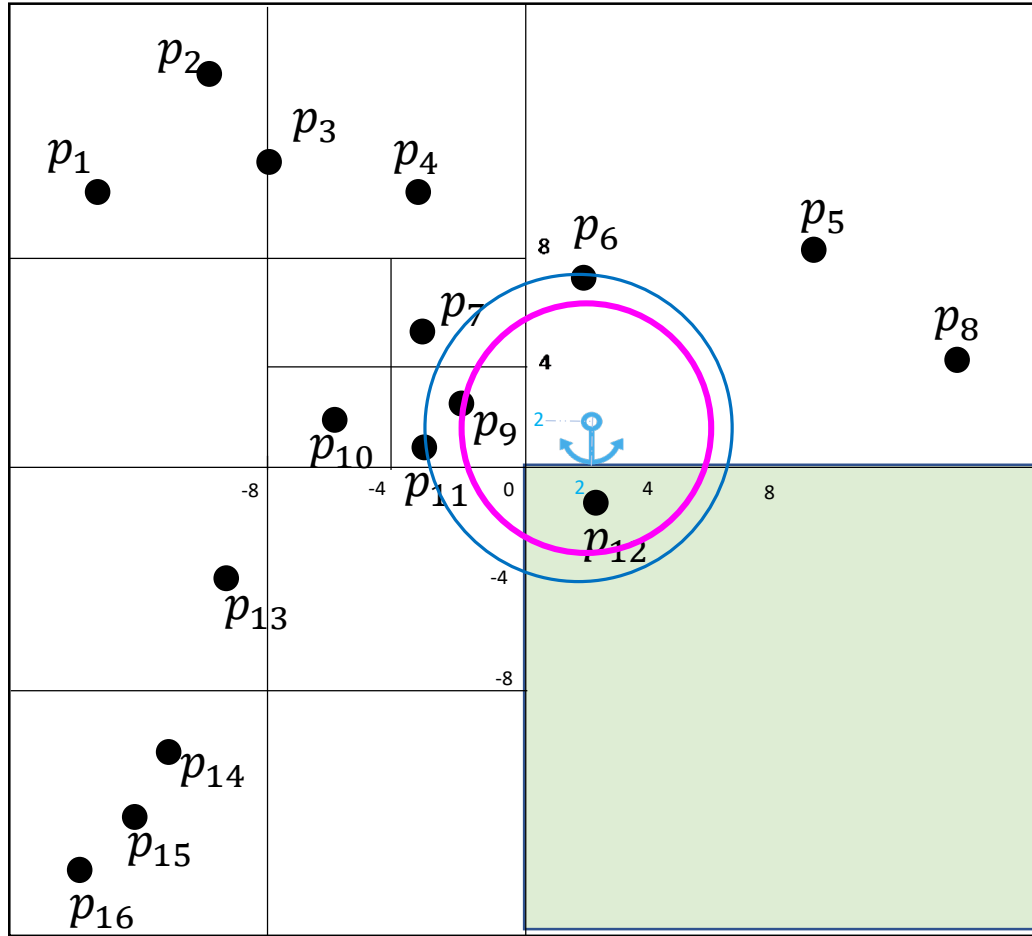
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# Example of $k$ -NN query in PR-QuadTree

anchor = (2, 2),  $k=2$ , Bucket size  $b = 3$

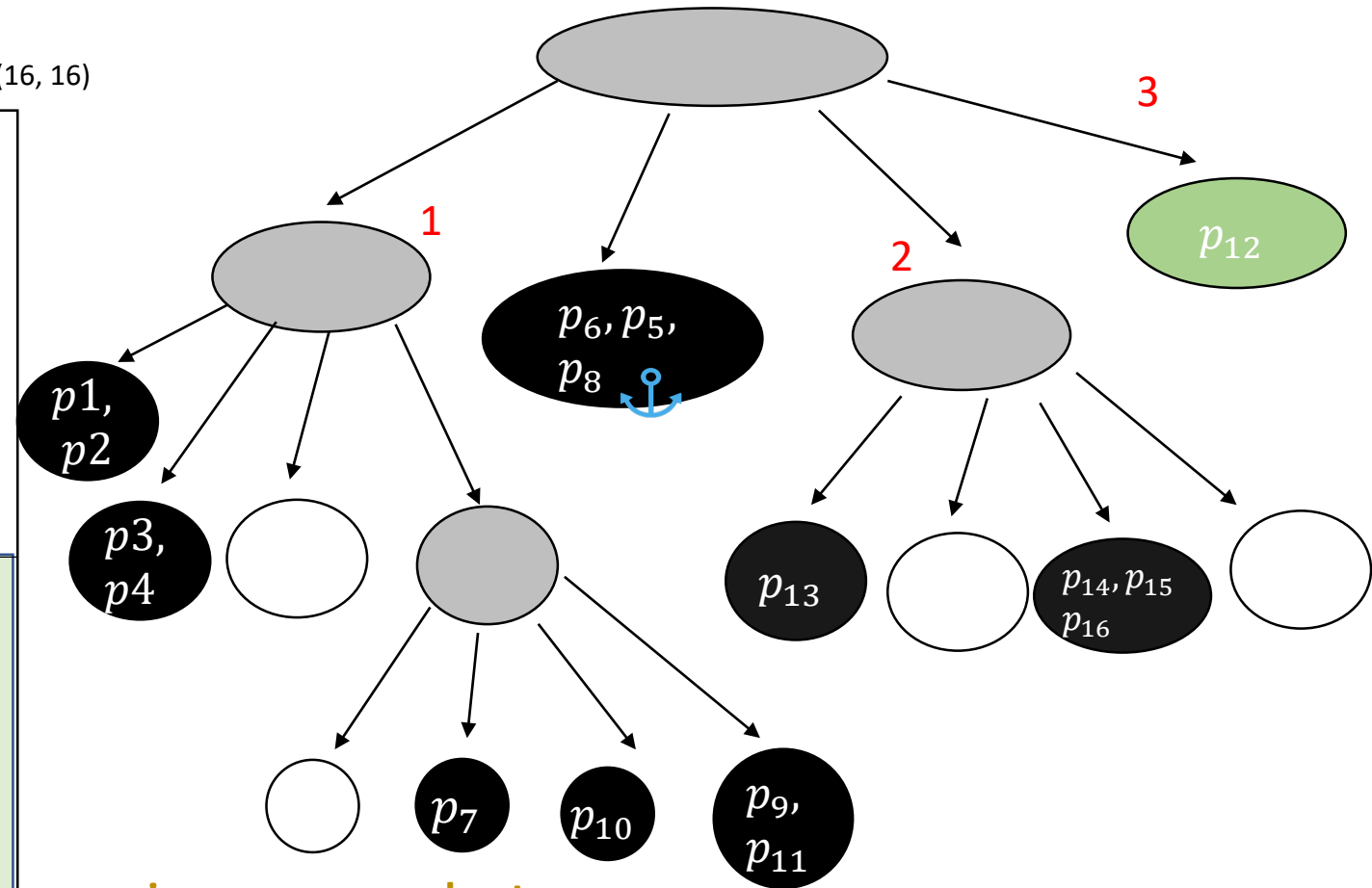
(-16, 16) (16, 16)



(-16, -16)



(16, -16)



$p_{12}$  improves upon best neighbor! Insert into BPQ as new best neighbor, throw away worst neighbor ( $p_6$ )!



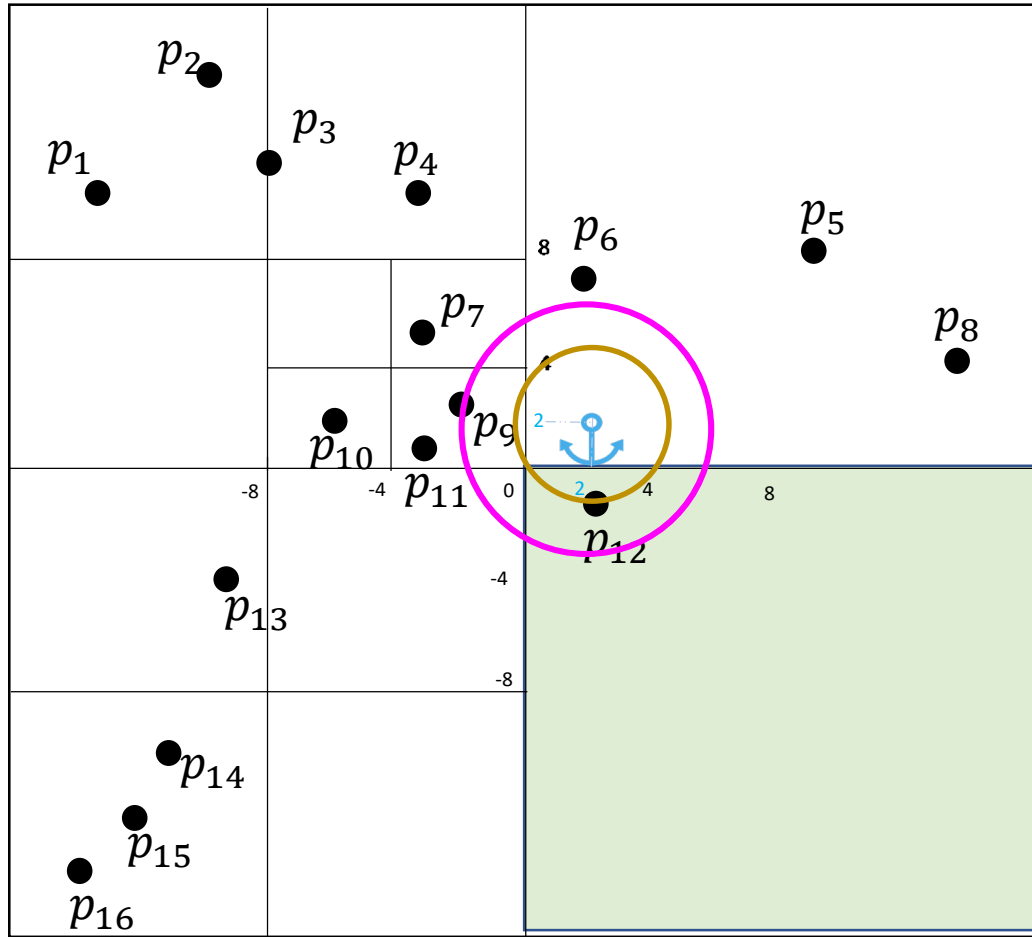
BPQ



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anchor = (2, 2),  $k=2$ , Bucket size  $b = 3$

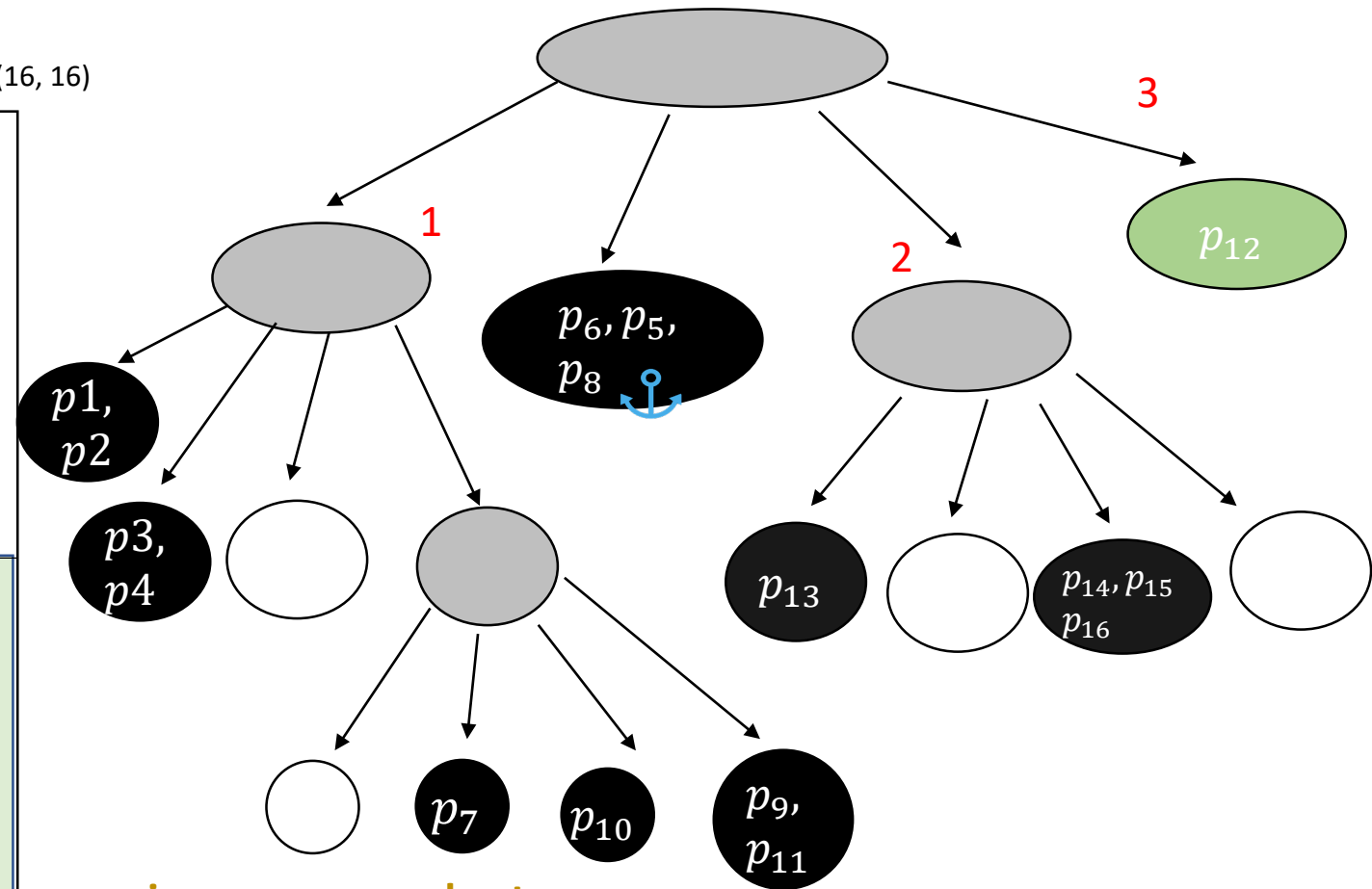
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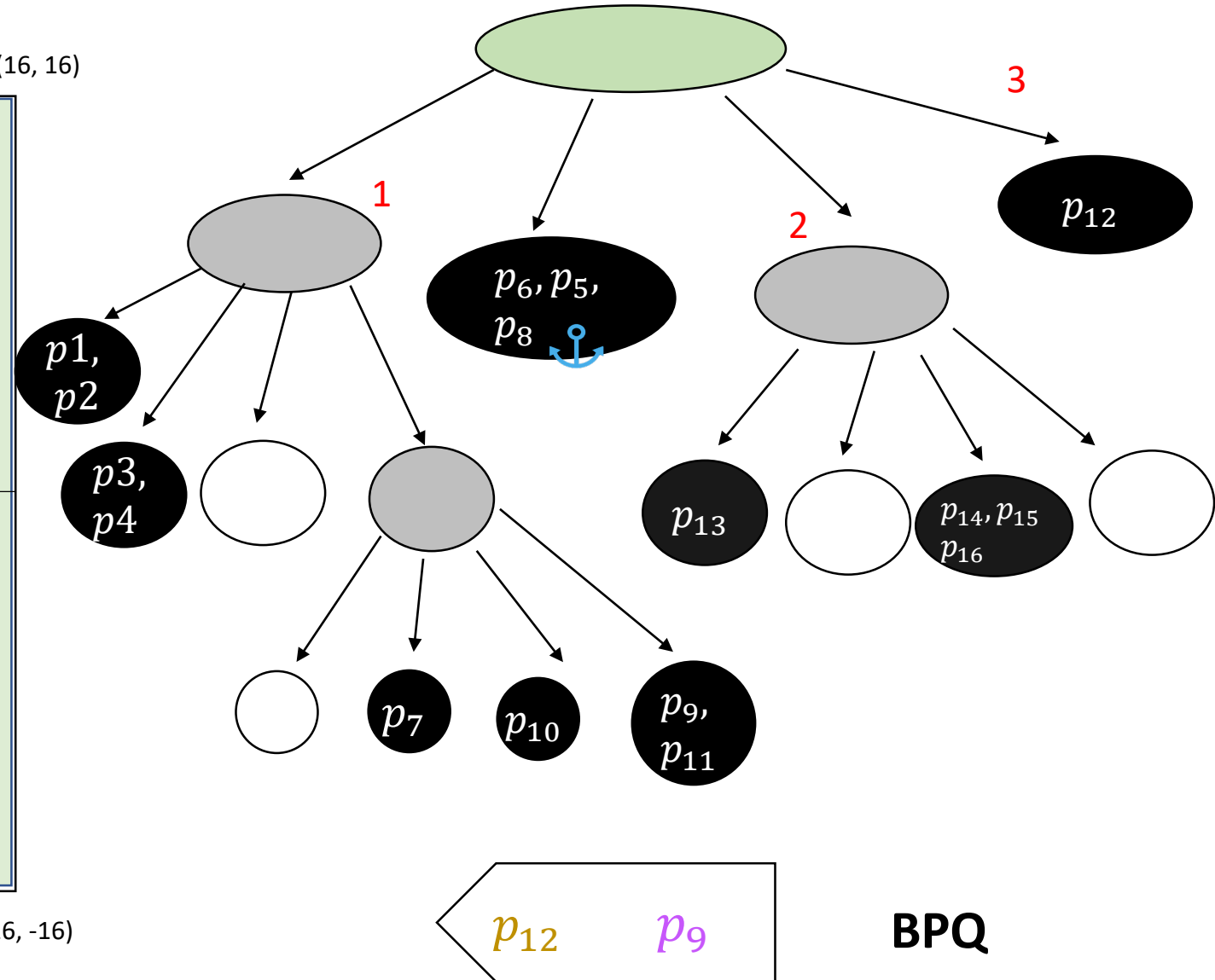
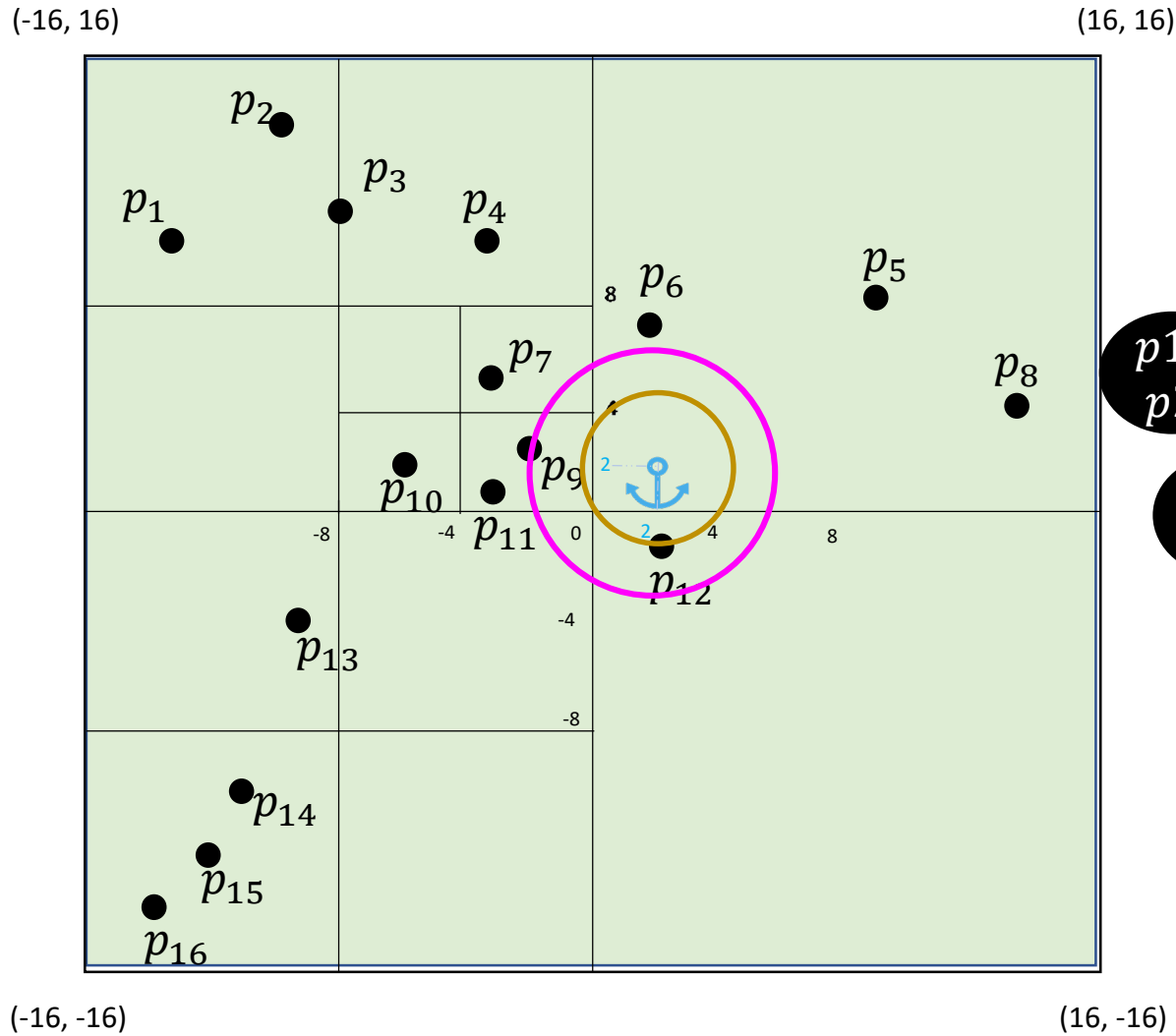
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BPQ

# Example of $k$ -NN query in PR-QuadTree

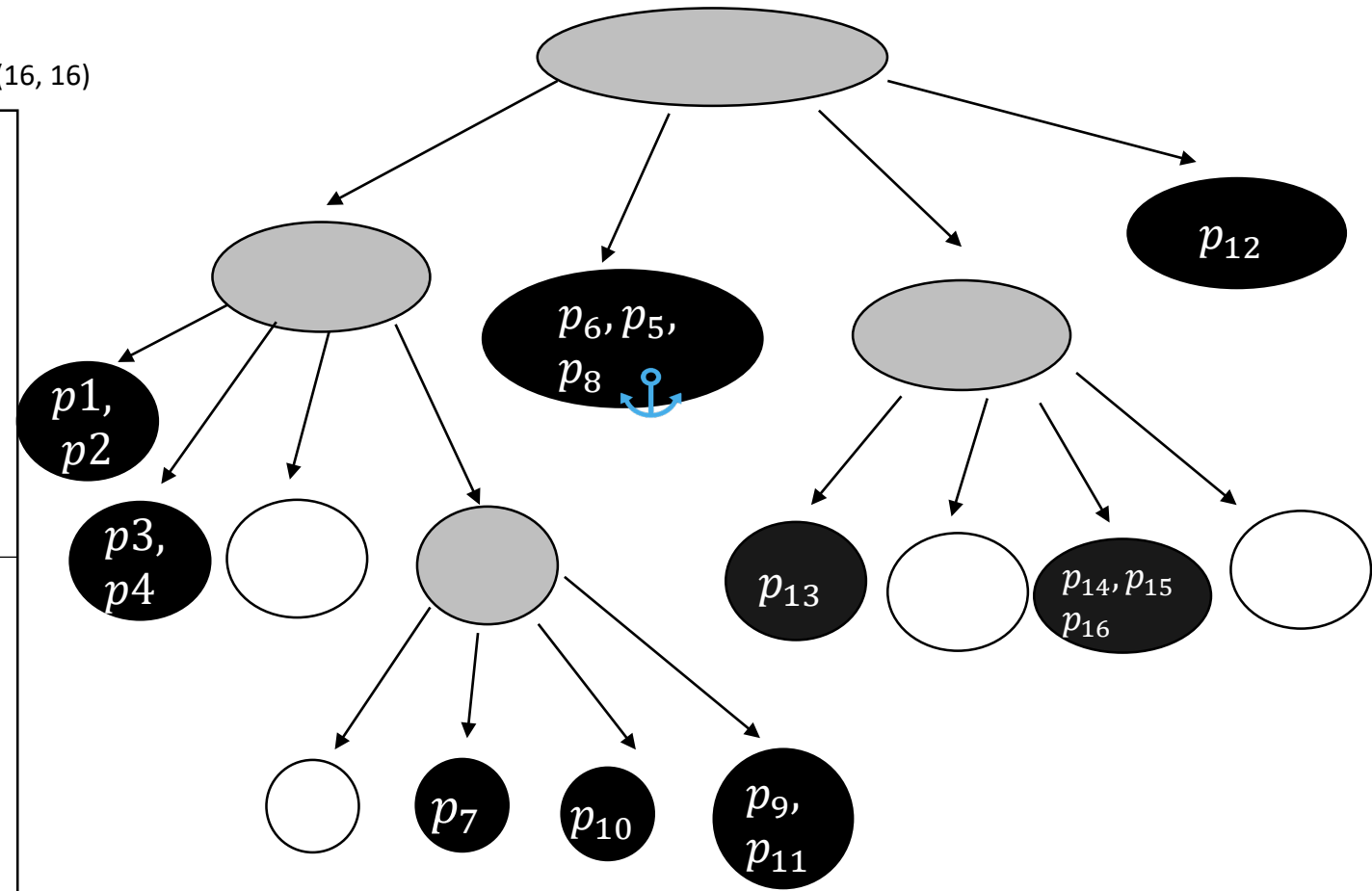
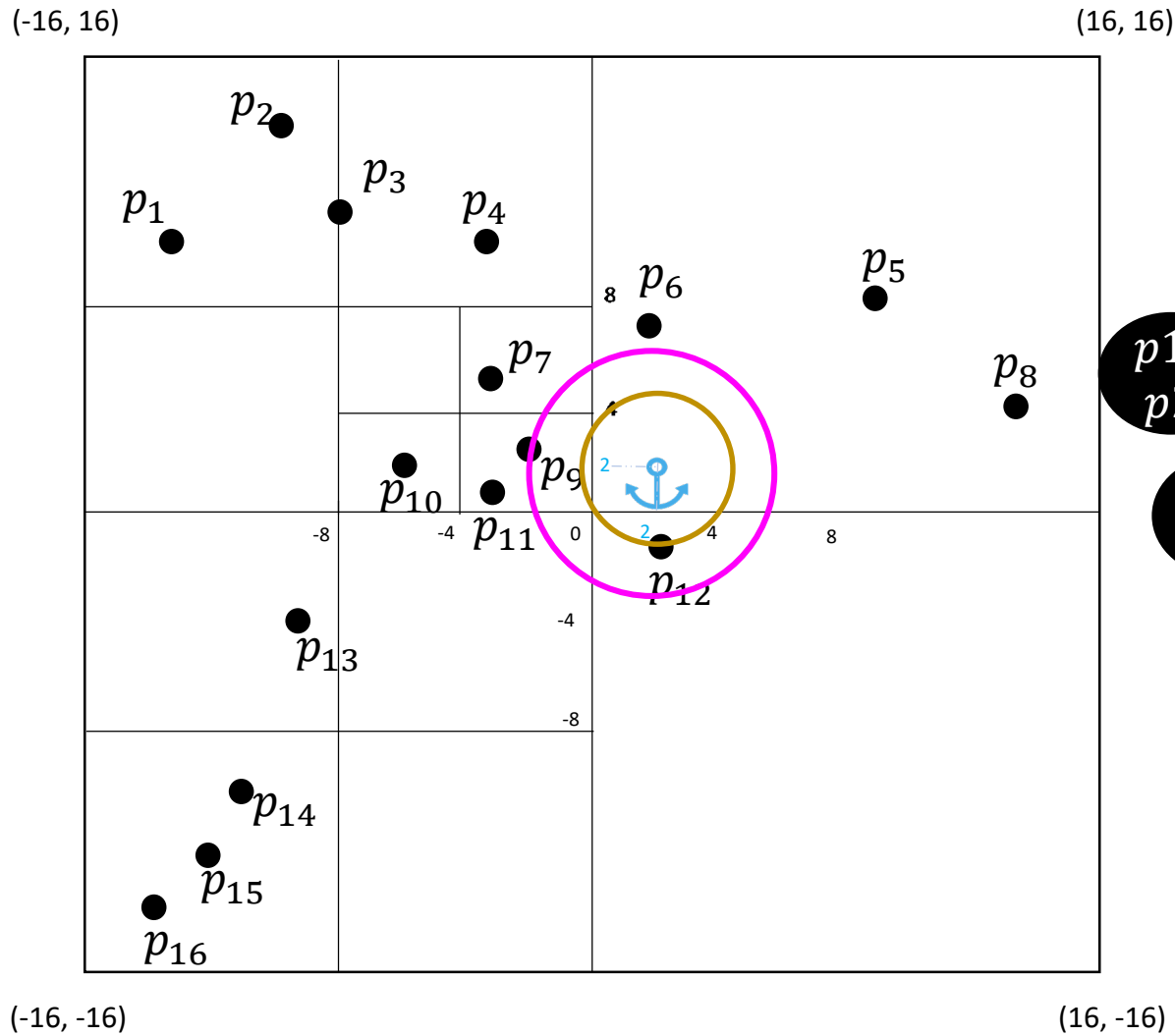
anchor = (2, 2),  $k=2$ , Bucket size  $b = 3$



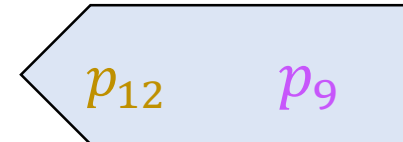
BPQ

# Example of $k$ -NN query in PR-QuadTree

anchor = (2, 2),  $k=2$ , Bucket size  $b = 3$



Done! 😊



BPQ

# One clarification

- You yourselves might want to implement an **improved** version of your project where you prioritize the visiting of the residual sub-children based on either **area of intersection** or **proximity to anchor**.

# One clarification

- You yourselves might want to implement an **improved** version of your project where you prioritize the visiting of the residual sub-children based on either **area of intersection** or **proximity to anchor**.
- You are **definitely allowed** to do this, but we
  - a) **Do not require it**, we instead recommend Z-order to keep things simple, and
  - b) Suggest you implement a **working version of your project first** to secure 100% and then think about submitting an improved version.

# Summary

- In the **descending** phase, you act in exactly the same way as with the KD-Tree in 2 dimensions: aggressively descend towards the point.
  - You get quicker to the anchor point too, since you can move in directions non-orthogonal to axes.

# Summary

- In the **descending** phase, you act in exactly the same way as with the KD-Tree in 2 dimensions: aggressively descend towards the point.
  - You get quicker to the anchor point too, since you can move in directions non-orthogonal to axes.
- In the **backtracking** phase, you should loop over your children **in Z-order**, and check whether each one of them can contribute to the solution set.
  - Check the **protected** method `doesQuadIntersectAnchorRange()` in `PRQuadNode.java`.
  - Do **not** prioritize based on area of intersection or some other complicated geometrical criteria.

# PR-QuadTrees VS KD-Trees

| PR-QuadTrees  |   | KD-Trees  |   |
|---|---|---|---|
| +   | -   | +   | -   |
| $(\forall n \geq 2)[\log_4 n < \log_2 n]$<br>offers better search | Intractable fanout for large $d$<br>because of exponential<br>dependency, even for basic<br>operations like insert, search,<br>delete | For dictionary queries and range,<br>can work quite well even in large<br>$d$ . | Nearest neighbor search can be<br>quite inefficient for large $d$   |
| Faster operations for 2D spaces                                   | Sensitive to relative proximity of<br>points  | Easy to understand and code<br>(Binary trees are KD-Trees with<br>$k = 1$ !)    | Special cases of deletion need<br>care when coding  |
| Insensitive to insertion order                                    | Tricky coding   | ...   | Deletion rather inefficient (calls<br>to <code>findMin()</code> for both subtrees<br>in levels with<br><code>currDim != desiredDim</code> ) |
| ...   | ...   | ...   | Sensitive to insertion order  |