

# Lecture 5: Divide & Conquer 1

## 2-D Maxima & Closest Pair

CS 341: Algorithms

Tuesday, Jan 22<sup>nd</sup> 2019

# Outline For Today

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1. 2-D Maxima
2. Closest Pair

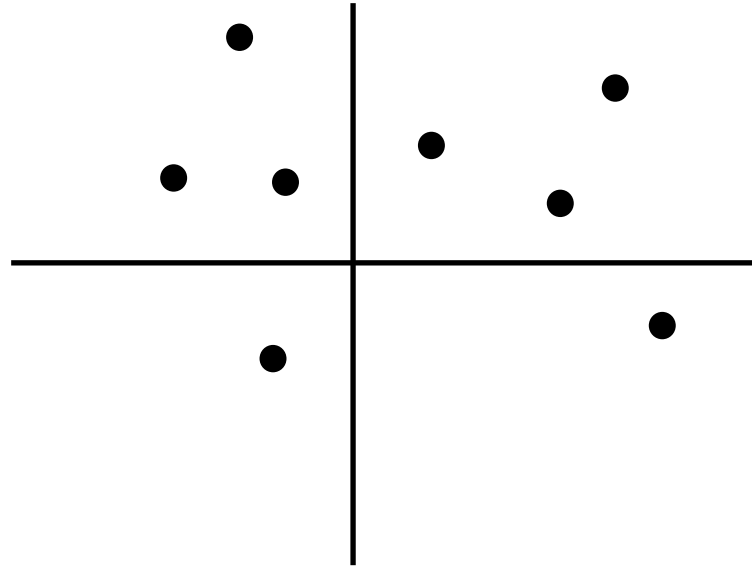
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1. 2-D Maxima
2. Closest Pair

# 2-D Maxima

◆ Input: Set P of n 2-D points



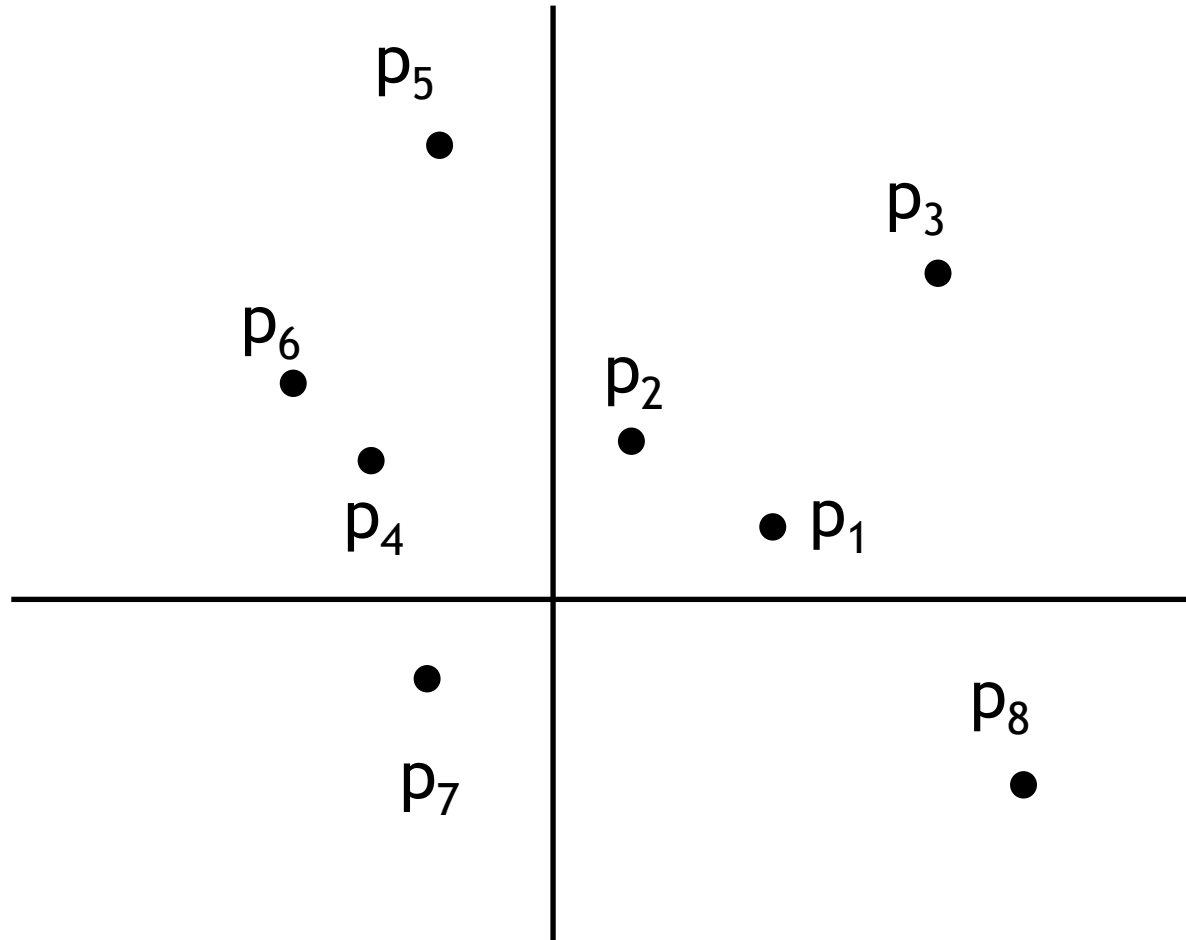
◆ Output: All *maximal* points

◆ Dfn 1: Point p *dominates* point q iff

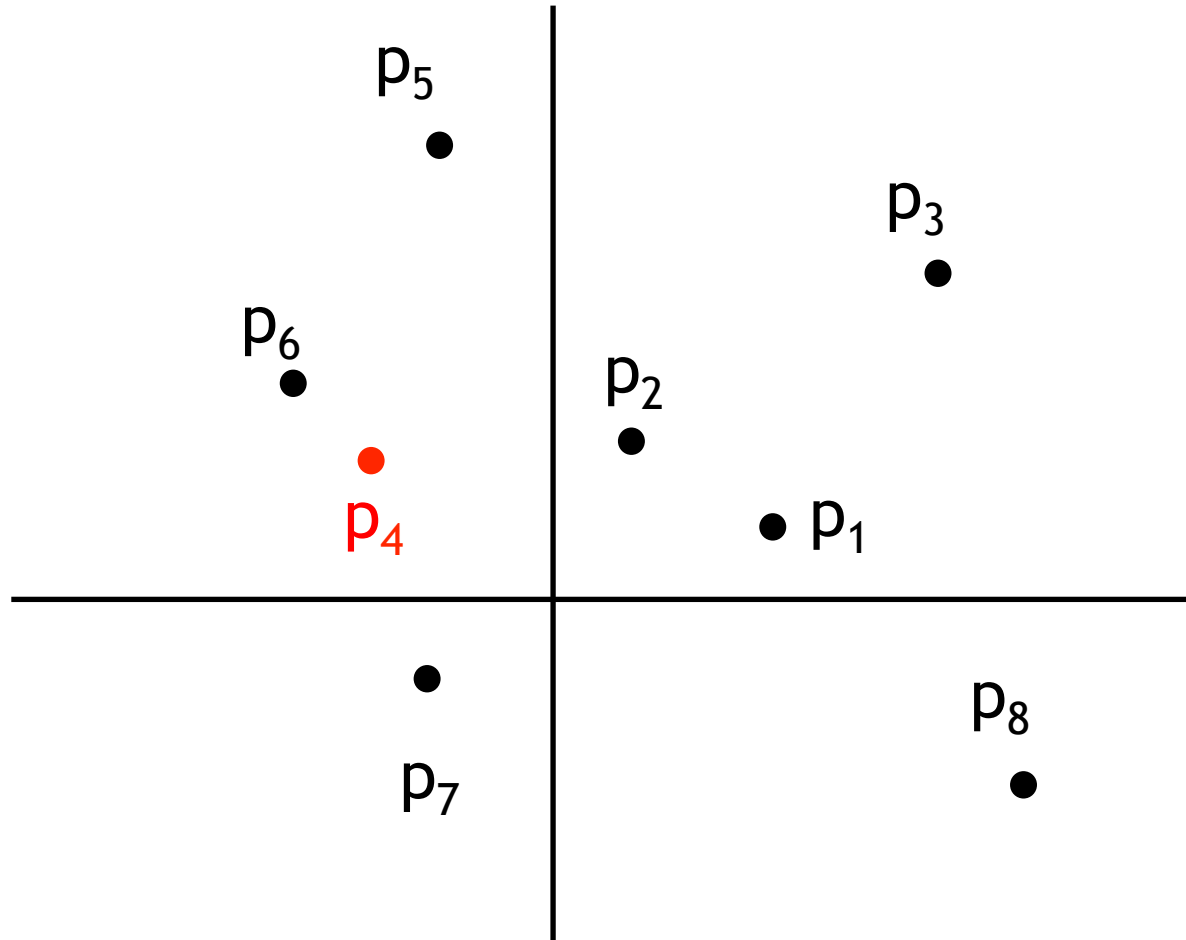
- $p.x > q.x$  AND  $p.y > q.y$

◆ Dfn 2: Point p is *maximal* if no point dominates it

# Example

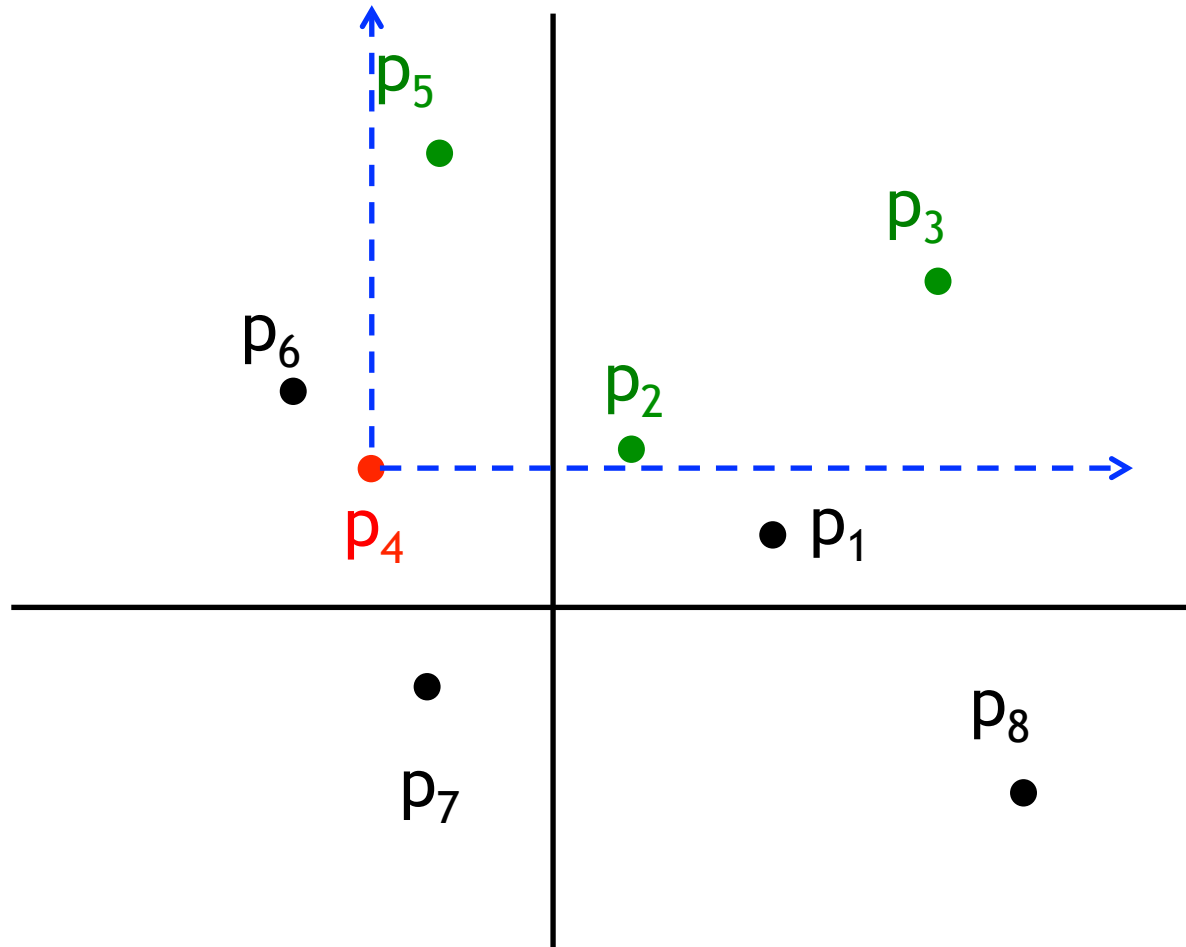


# Example



Q: Is  $p_4$  maximal?

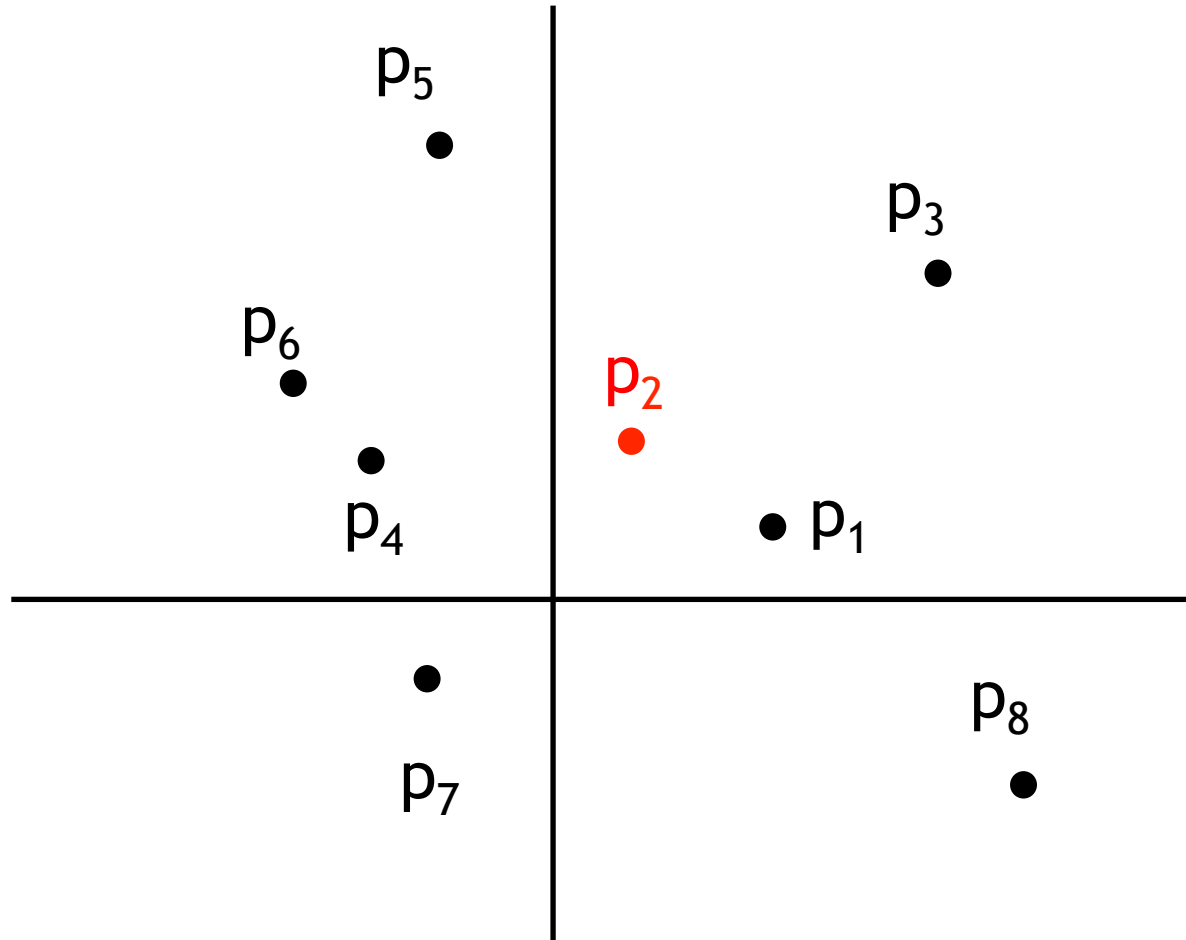
# Example



Q: Is  $p_4$  maximal?

A: No.  $p_2$ ,  $p_3$ , and  $p_5$  dominate  $p_4$ .

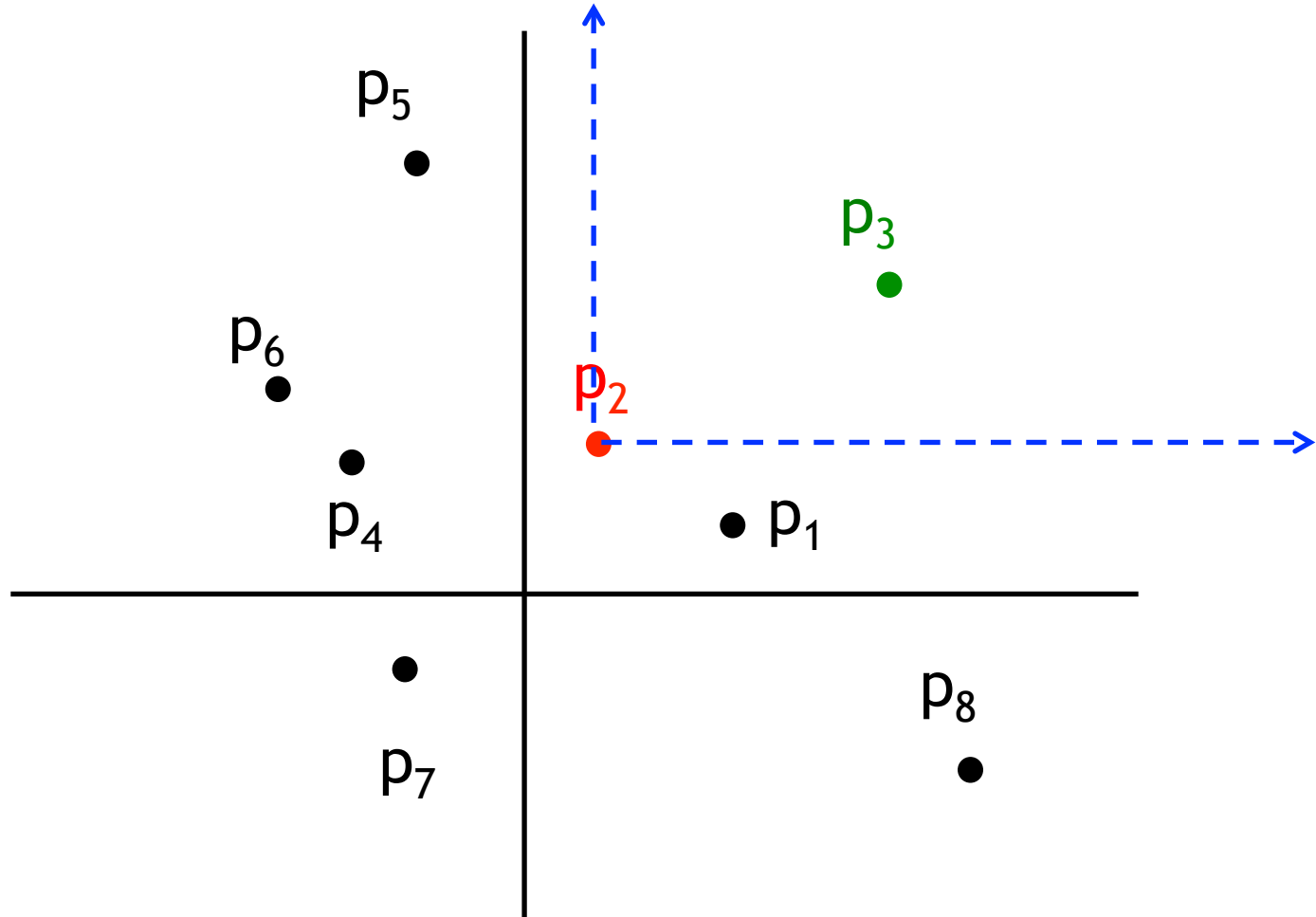
# Example



Q: Is  $p_2$  maximal?



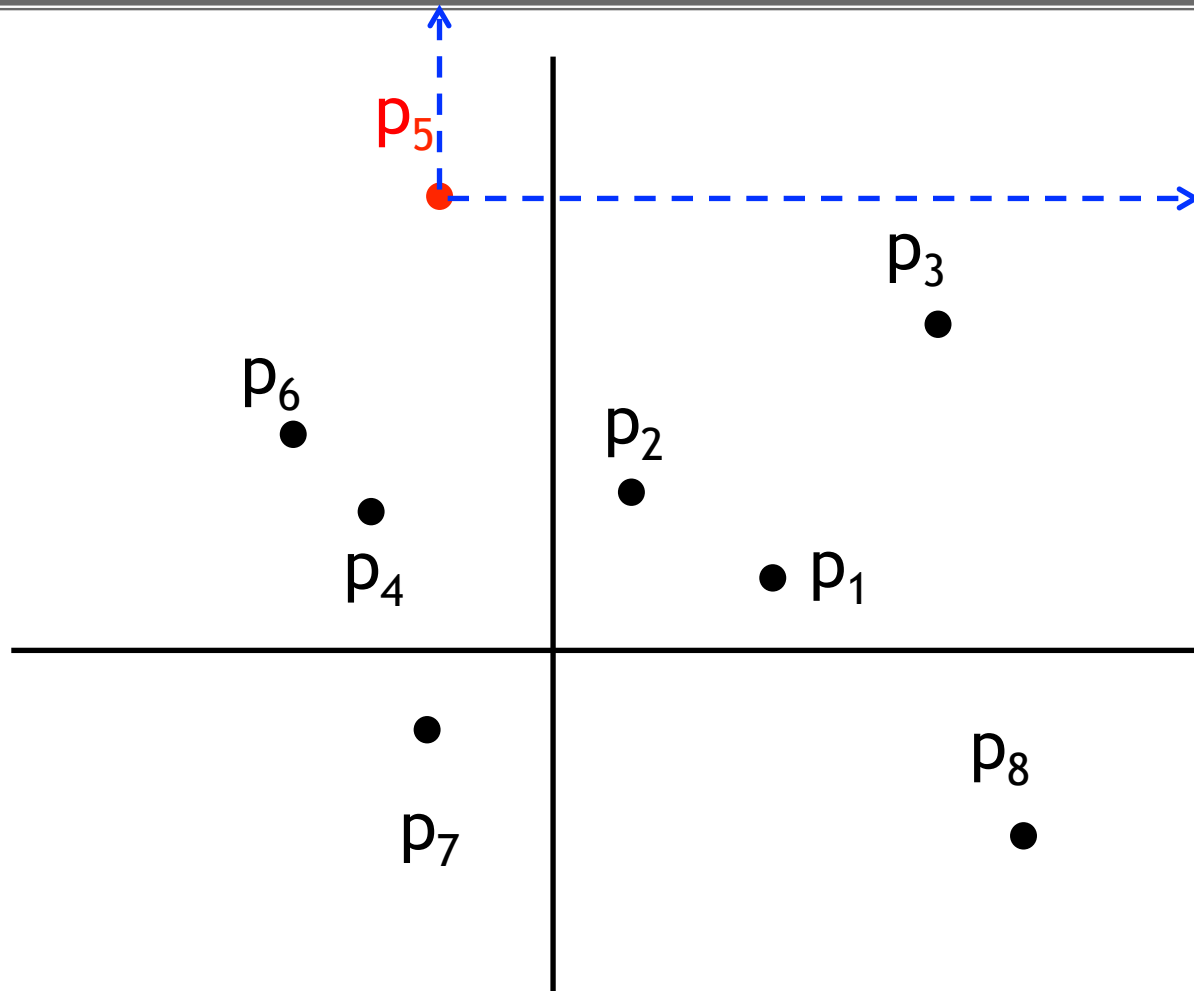
# Example



Q: Is  $p_2$  maximal?

A: No.  $p_3$  dominates  $p_4$ .

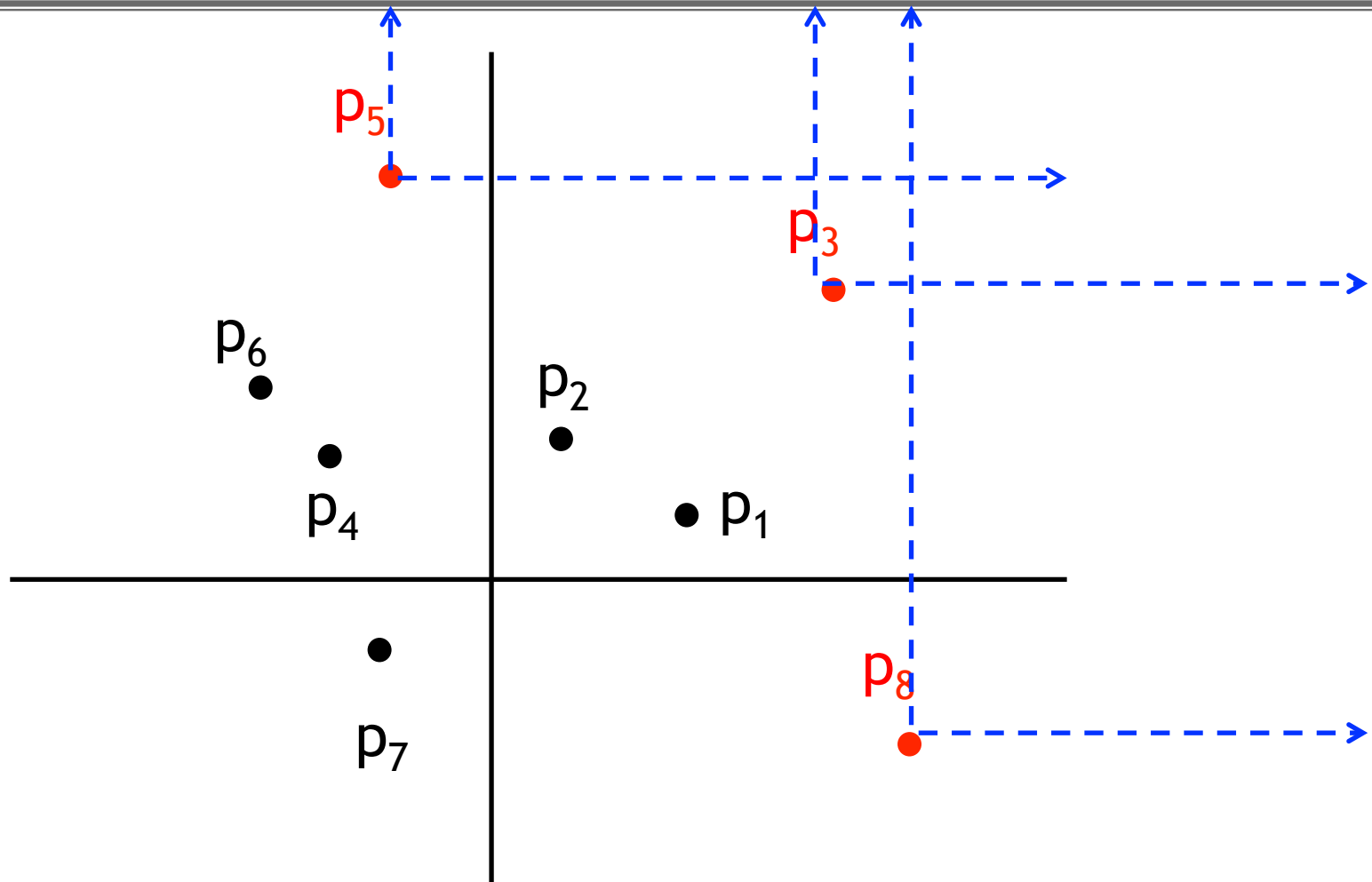
# Example



Q: Is  $p_5$  maximal?

A: Yes

# Example

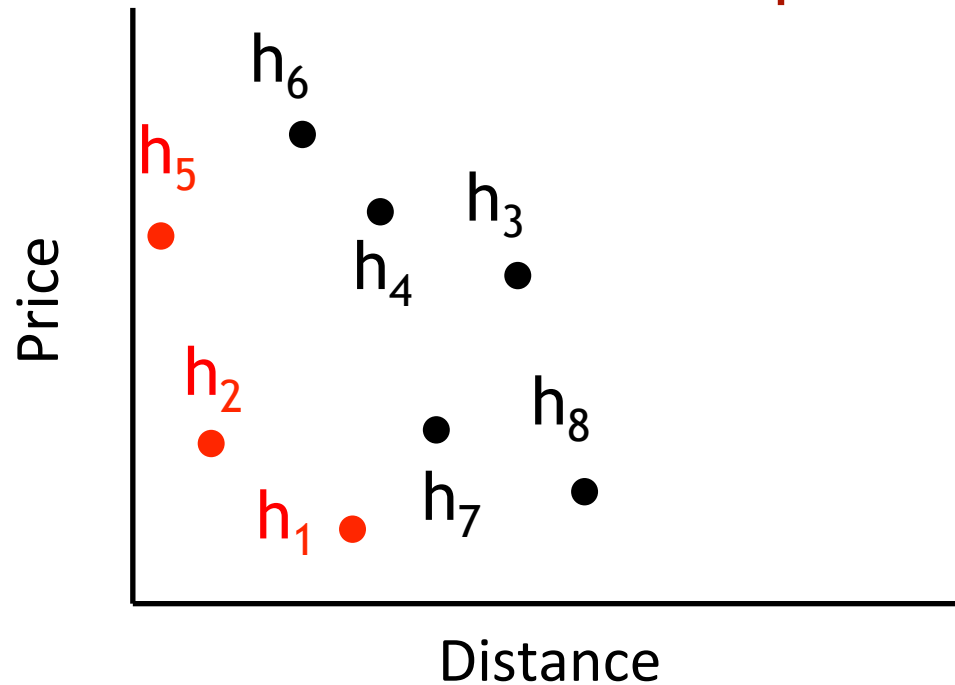


$p_3$ ,  $p_5$ , and  $p_7$  are maximal  
other points are not

# Applications

## ◆ Databases: Skyline queries

- Example “Skyline query”: *minima*
- Find “minimal” hotels in a price & distance graph



## ◆ Economics: Finding “Pareto Optimal” points

# Alg 1: Brute Force

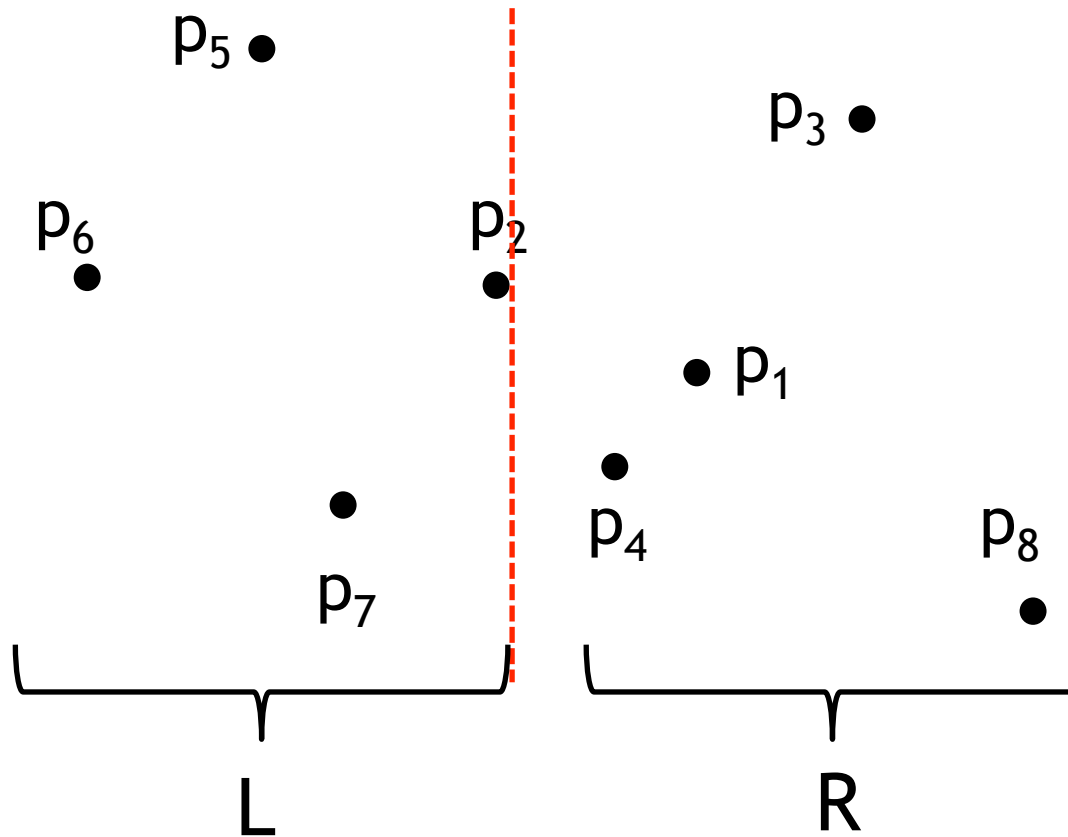
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```
procedure bruteForceMaxima(Set P of n points):  
    M = {}  
    for each p in P:  
        for each q in P:  
            check if p is dominated by q  
        if p is not dominated:  
            M.insert(p);  
return M
```

Runtime:  $O(n^2)$

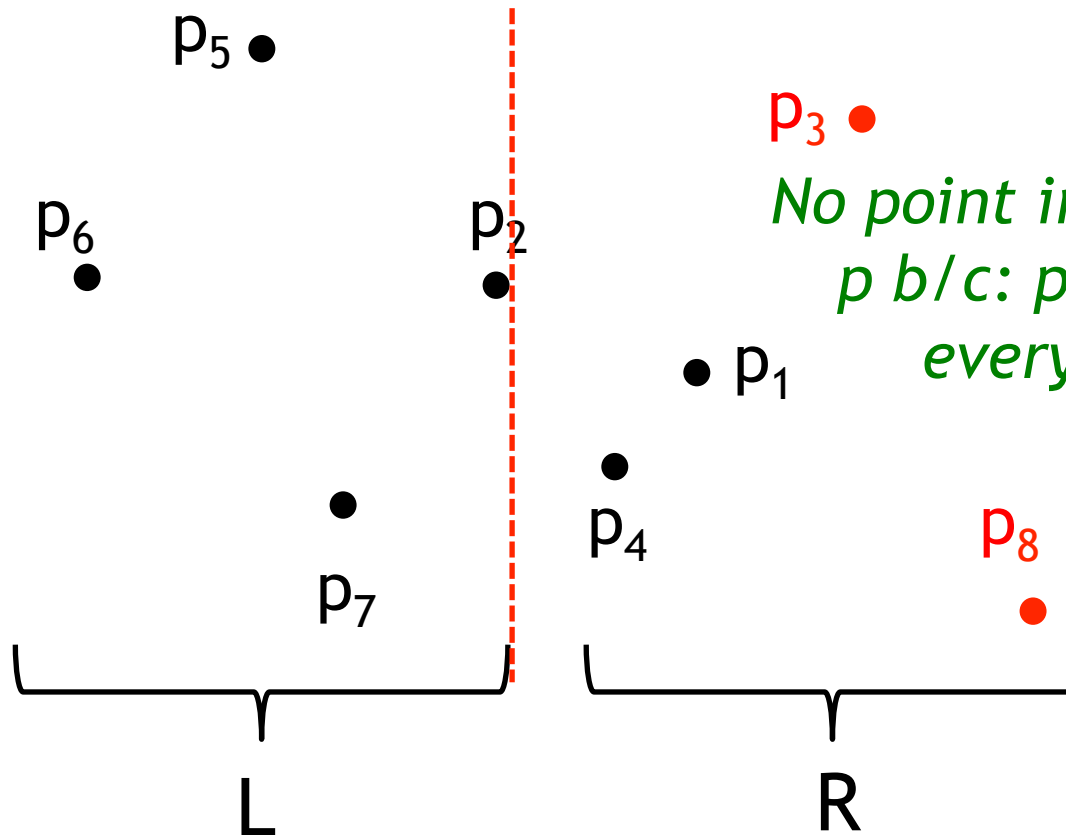
# Can we Divide & Conquer?

◆ Idea: Let's divide  $P$  vertically (on x-axis)



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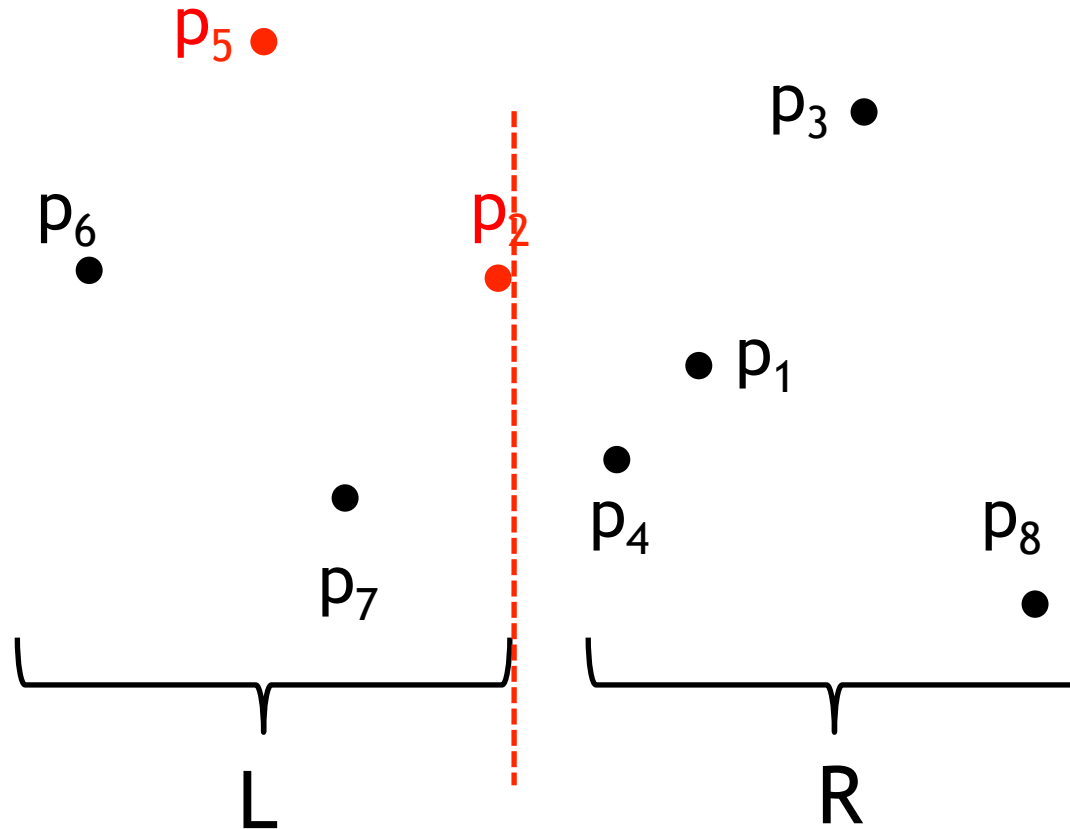


*Why?*  
*No point in  $L$  can dominate  $p$  b/c:  $p.x \geq x$  value of every point in  $L$ .*

*Q1: What can you say about a maximal point  $p$  in  $R$ ?*

*A1: It is maximal in  $P$  as well.*

# Can we Divide & Conquer?



*Q2: What can you say about a maximal point  $q$  in  $L$ ?*

*A2: It's maximal iff  $q.y \geq y$  value of every point in  $R$ .*

*In particular, let  $p^*$  be the point in  $R$  with  $\max y$*

*Then  $q$  is maximal iff  $q.y \geq p^*.y$ ,*



# DC-Maxima

**Procedure** Algorithm(Set P of n points):

Sort P by x values;  $\longrightarrow O(n \log(n))$  work

**return** DCMaxima(P)

**Procedure** DCMaxima (P sorted by x values):

**if** (P.size == 1) **return** P;

L = DCMaxima(P[1...n/2]);

R = DCMaxima(P[n/2+1...n]);

let  $p^*$  be max y valued point in R;  $\longrightarrow O(n)$  work

let M = R;

**for each** q in L:

**if** (q.y  $\geq$   $p^*.y$ )

M.insert(q);

**return** M;

$\longrightarrow O(n)$  work

Total:  $O(n)$  work outside recursive calls

# Runtime Analysis

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Recursive part:  $T(n) = 2T(n/2) + O(n)$

By Master Thm:  $O(n \log(n))$

Total Work:

1. Initial Sorting:  $O(n \log(n))$
2. Recursive part:  $O(n \log(n))$

Total:  $O(n \log(n))$

# Exercise

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After initial sorting by x-axis, design an  $O(n)$  time algorithm (not DC) that gives all of the maximal points.

Note: Total time still  $O(n \log(n))$  b/c of sorting. But post-sorting work is linear instead of  $O(n \log(n))$  of DCMaxima.

Fact:  $\Omega(n \log(n))$  lower bound for comparison based algs.

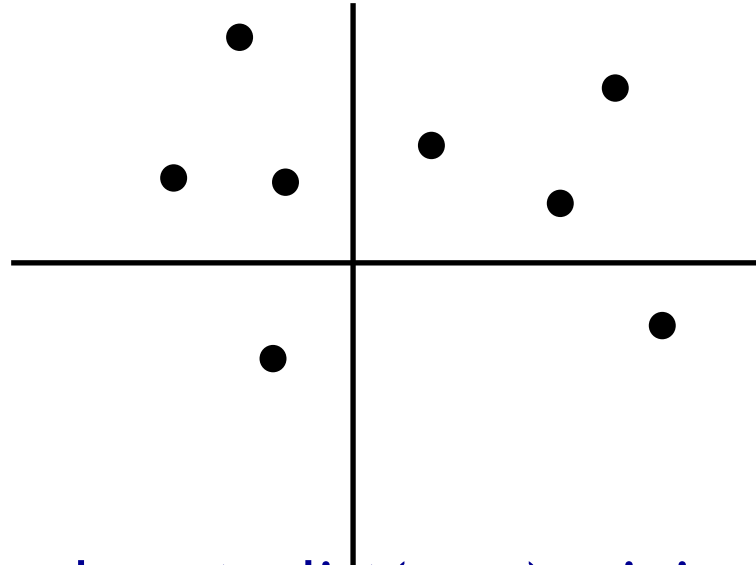
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# Closest Pair Problem

◆ Input: Set P of n 2-D points



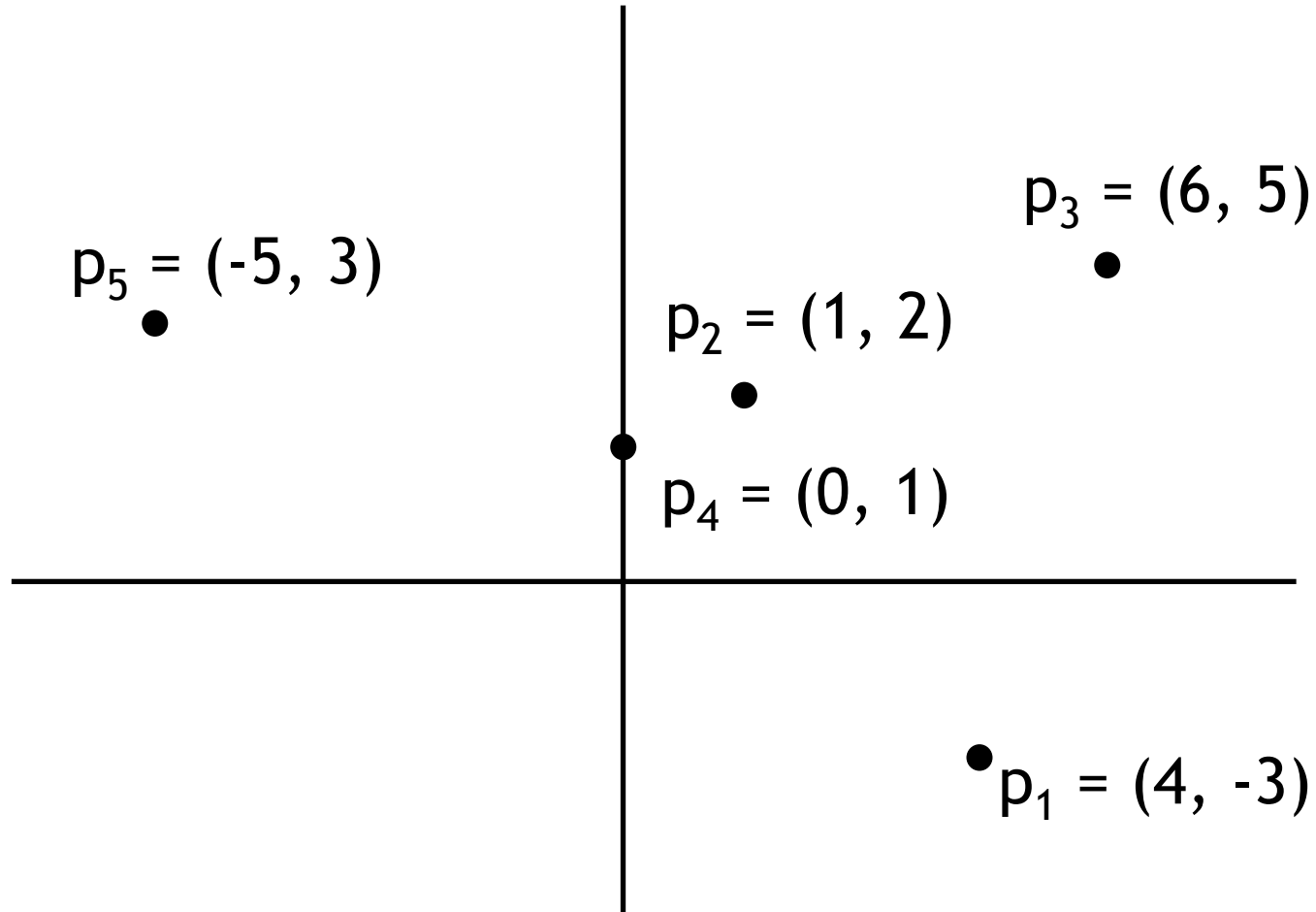
◆ Output: pair p and q s.t.  $\text{dist}(p, q)$  minimum over all pairs

◆ Break ties arbitrarily

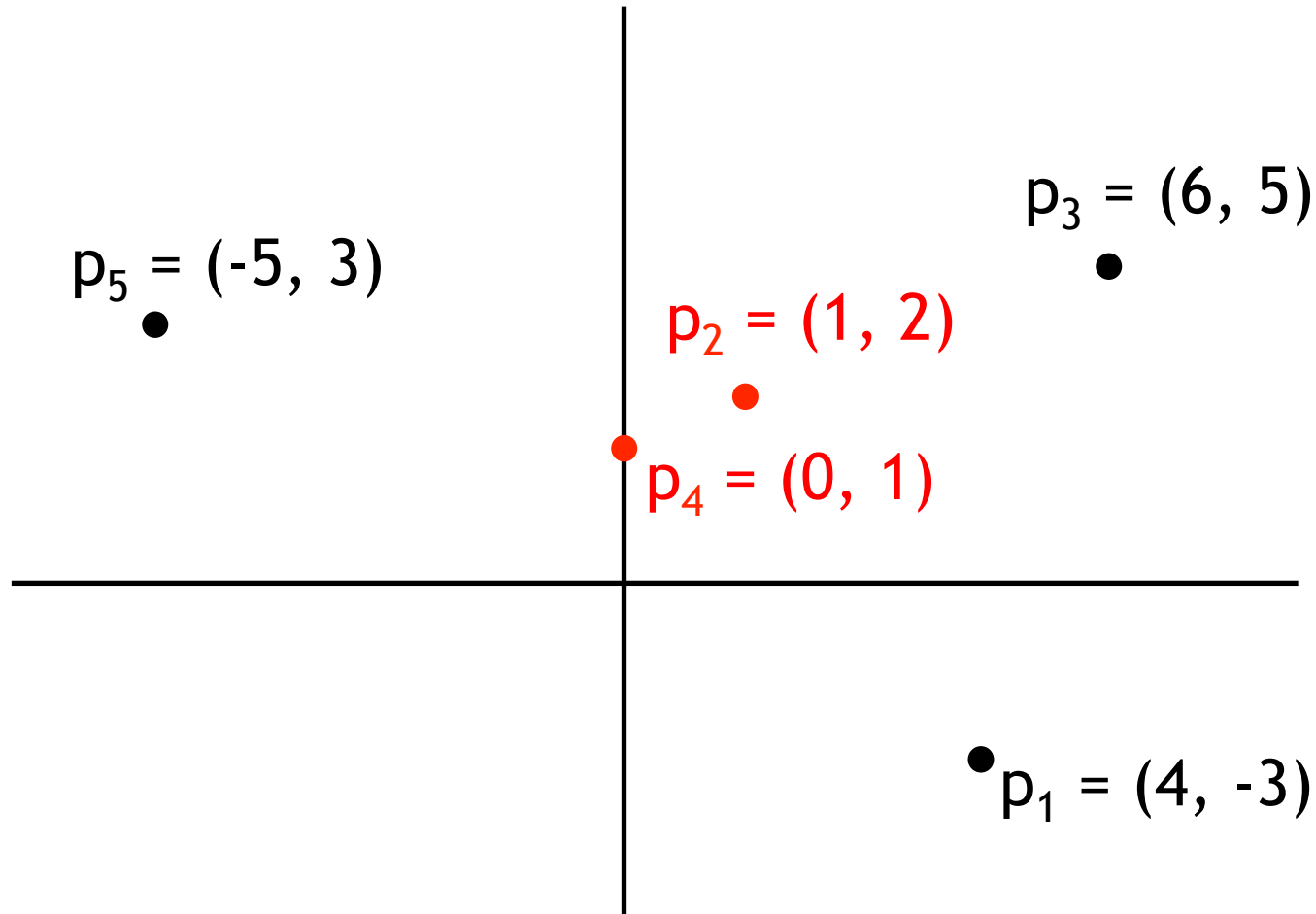
◆  $\text{dist}(p, q)$ : Euclidean distance

$$\text{dist}(p, q) = \sqrt{(p.x - q.x)^2 + (p.y - q.y)^2}$$

# Example



# Example



Closest pair:  $(p_2, p_4)$ ;  $\text{dist}(p_2, p_4)$ :  $\sqrt{1^2 + 1^2} = \sqrt{2}$

# Applications

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- ◆ Very fundamental computational problem
  - Databases
  - Machine Learning
  - Image Processing
  - Computational Geometry



# 1-D Version

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- ◆  $(x_1, x_2, \dots, x_n) = (2.2, 5.8, 1.1, -3.0, 1.2, \dots)$
- ◆ Just sort and scan:
  - ◆ compare each point with the next point in the sorted order
  - ◆ b/c closest pair has to be a consecutive pair
- ◆ Sort:  $O(n \log(n))$  time
- ◆ Scan:  $O(n)$  time
- ◆ Total:  $O(n \log(n))$  time

*Question: Can we do  $(n \log(n))$  in 2-D?*

# Alg 1: Brute Force

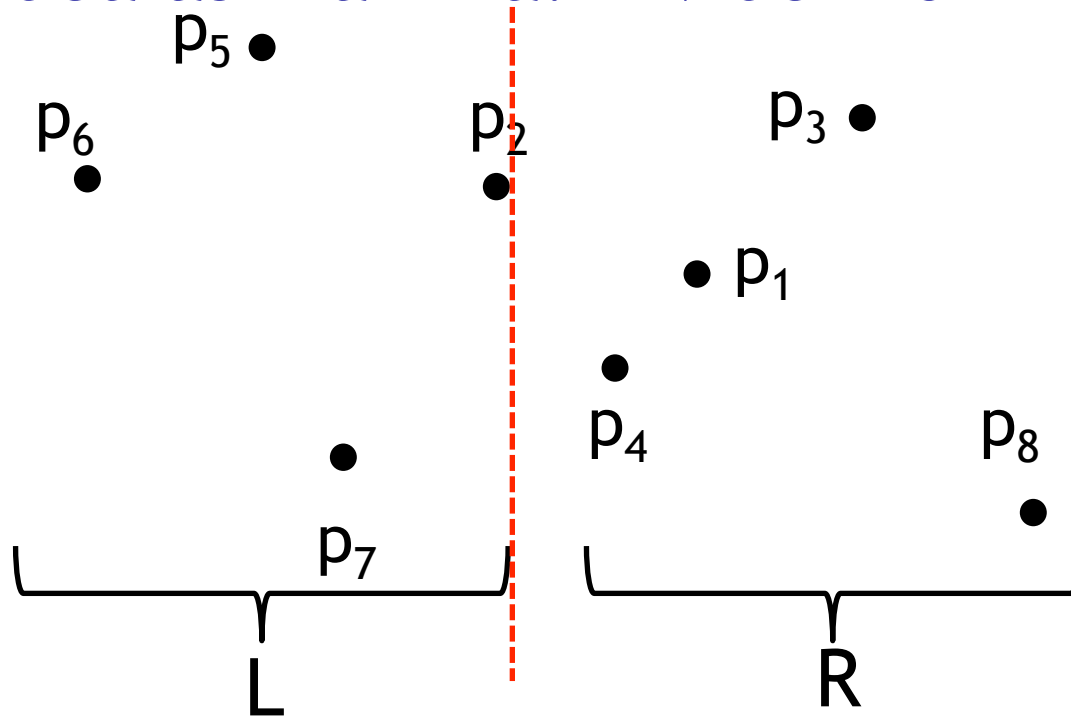
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```
procedure bruteForceCP(Set P of n points):  
    minPair = {}  
    minDist =  $+\infty$   
    for each p in P:  
        for each q in P:  
            if (dist(p, q) < minDist)  
                minPair.insert(p, q);  
                minDist = dist(p, q)  
    return minPair
```

Runtime:  $O(n^2)$

# Can we Divide & Conquer?

◆ Same idea as maxima: Divide  $P$  on x-axis



*Claim that doesn't require a proof: closest pair  $(p, q)$ :*

- 1.  $(p, q)$  both in  $L$ ;*
- 2.  $(p, q)$  both in  $R$ ; or*
- 3.  $p$  is in  $L$  and  $q$  is in  $R$*

# DC Algorithm Template:

---

**procedure** Algorithm(P of n points):

  sort P by x values

  DC-CP(P)

**procedure** DC-CP(P sorted by x values):

**if** (P.size  $\leq$  3) compare all & return closest;

  pair<sub>L</sub> = DC-CP(P[1,...,n/2])

  pair<sub>R</sub> = DC-CP(P[n/2+1,...,n])

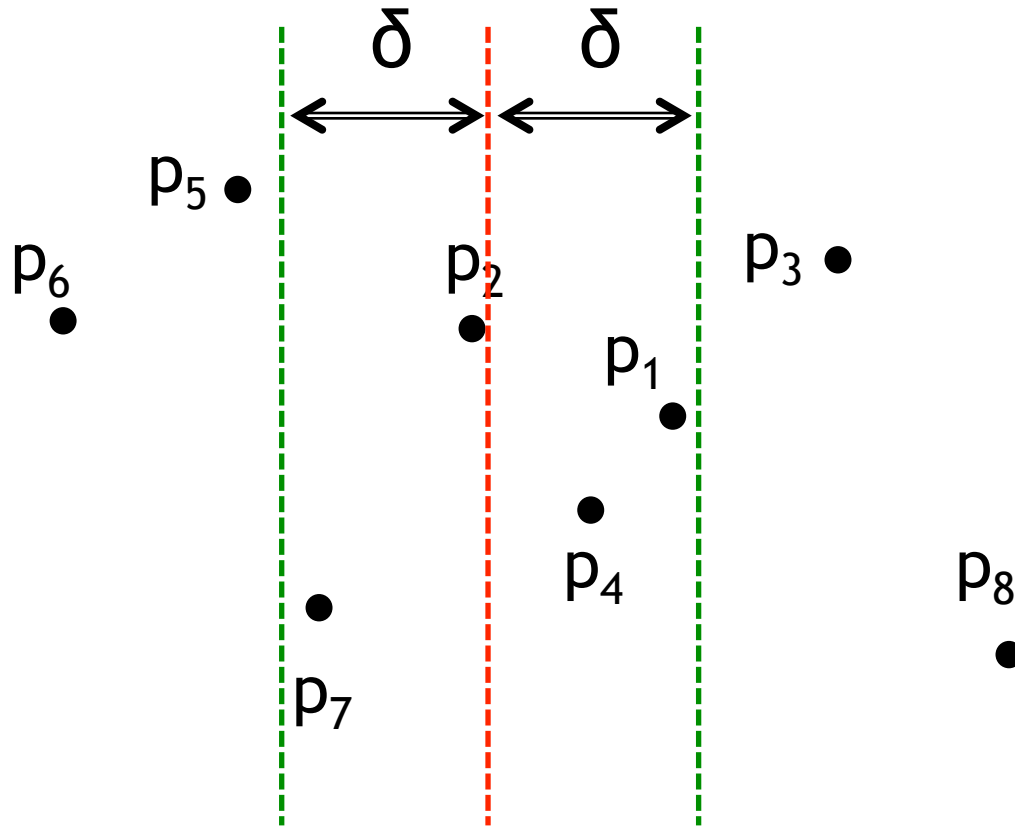
  pair<sub>S</sub> = findMinSpanningPair(P)

**return** min(pair<sub>L</sub>, pair<sub>R</sub>, pair<sub>S</sub>)

*Q: How can we find the spanning pair quickly?*

# Observation 1

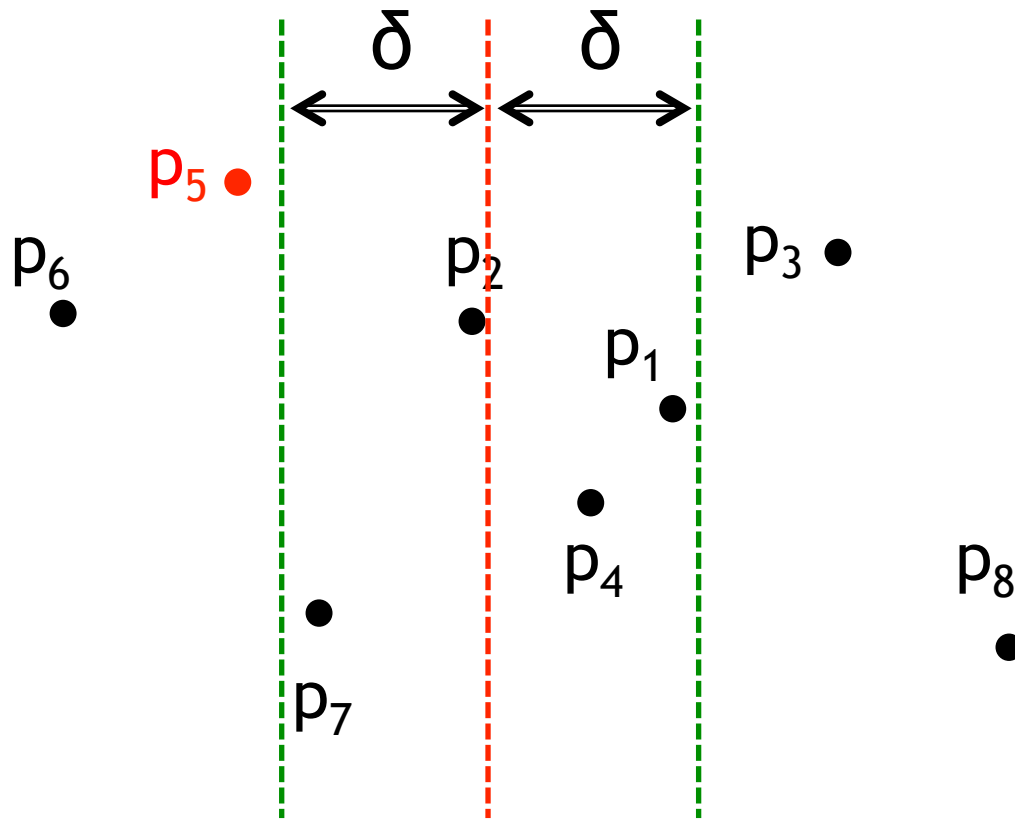
◆ Let  $\delta = \min (\text{dist}(\text{pair}_L), \text{dist}(\text{pair}_R))$



◆ Then  $\text{pair}_s$  (if closest globally) lies in the above  $2\delta$ -wide green strip

*Q: Why?*

# Example for Observation 1

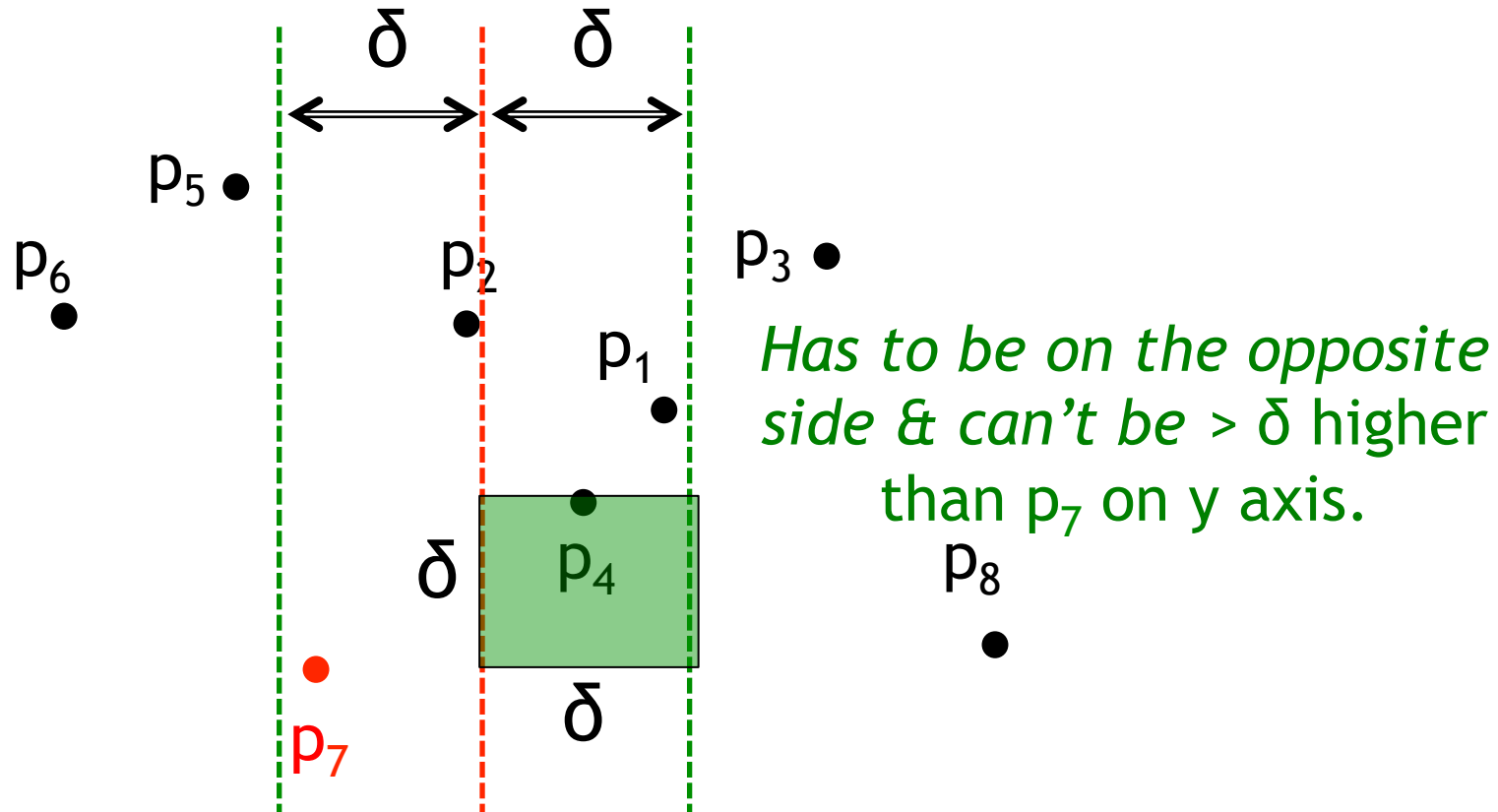


*Q: Can  $p_5$  be part of a globally closest pair<sub>s</sub>?*

*A: No. Any  $p$  in  $R$  is already has  $\text{dist} > \delta$  to  $p_5$ .*

# Observation 2

- ◆ Say,  $p_7$  (the lowest y valued point in strip) is in pair<sub>s</sub>

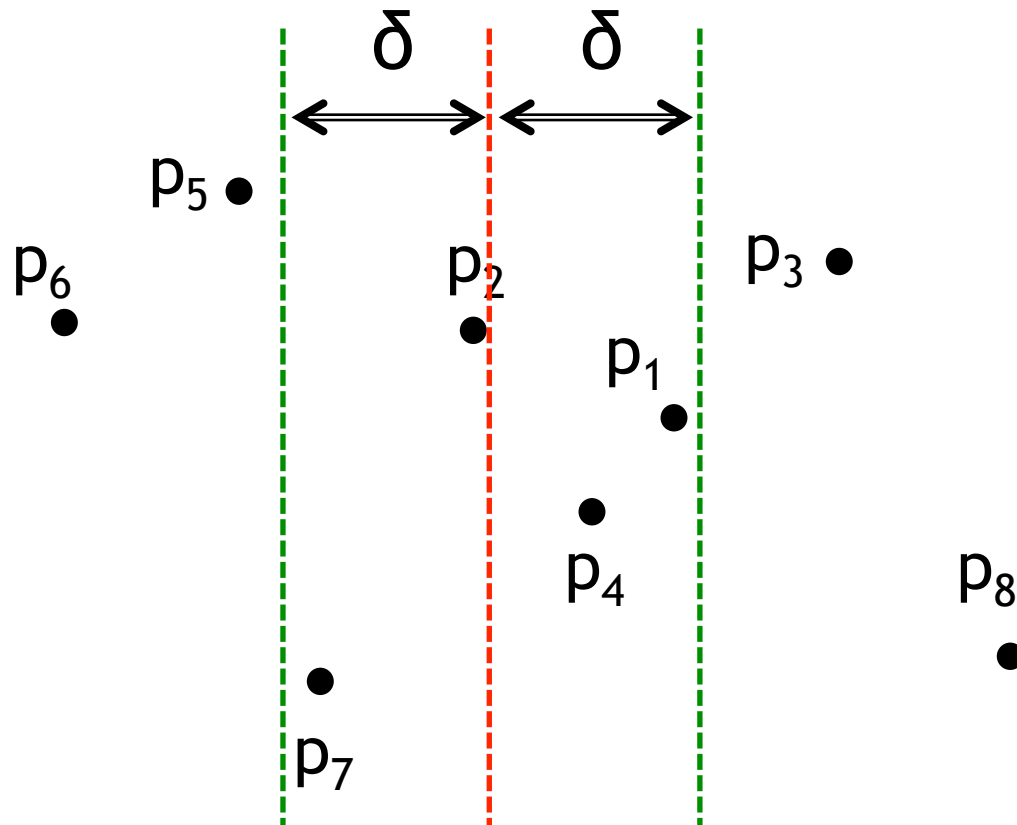


- ◆ Then the other point can only lie in this  $\delta \times \delta$  square.

*Q: Why?*

# Core Idea For Finding Spanning Pair

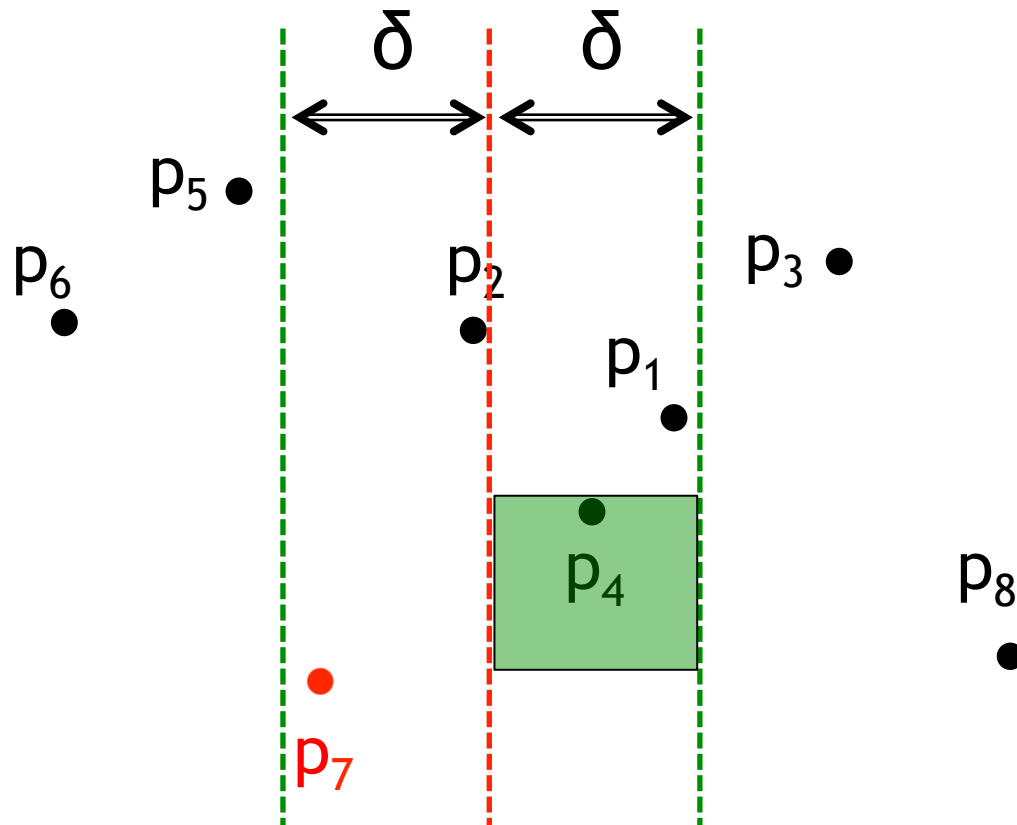
1. Start from lowest y valued point in the strip
2. Search the  $\delta \times \delta$  square points on the opposite side
3. Repeat 1 & 2 this for the next lowest y-valued point
4. So on and so forth...





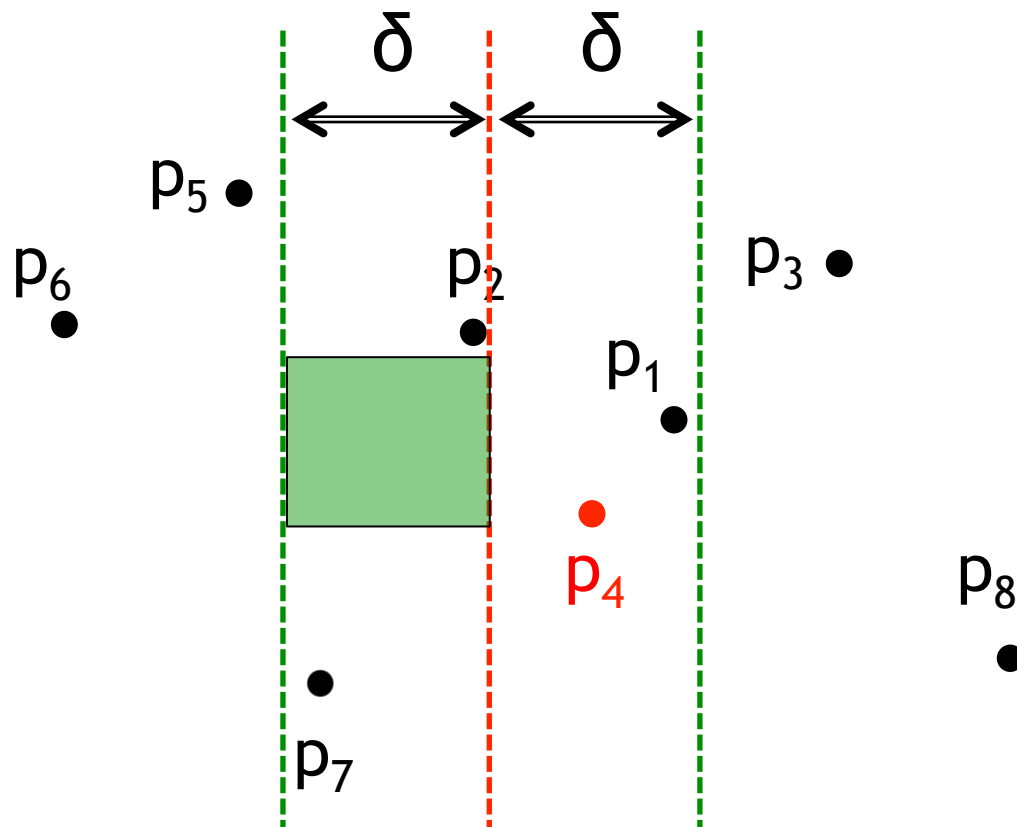
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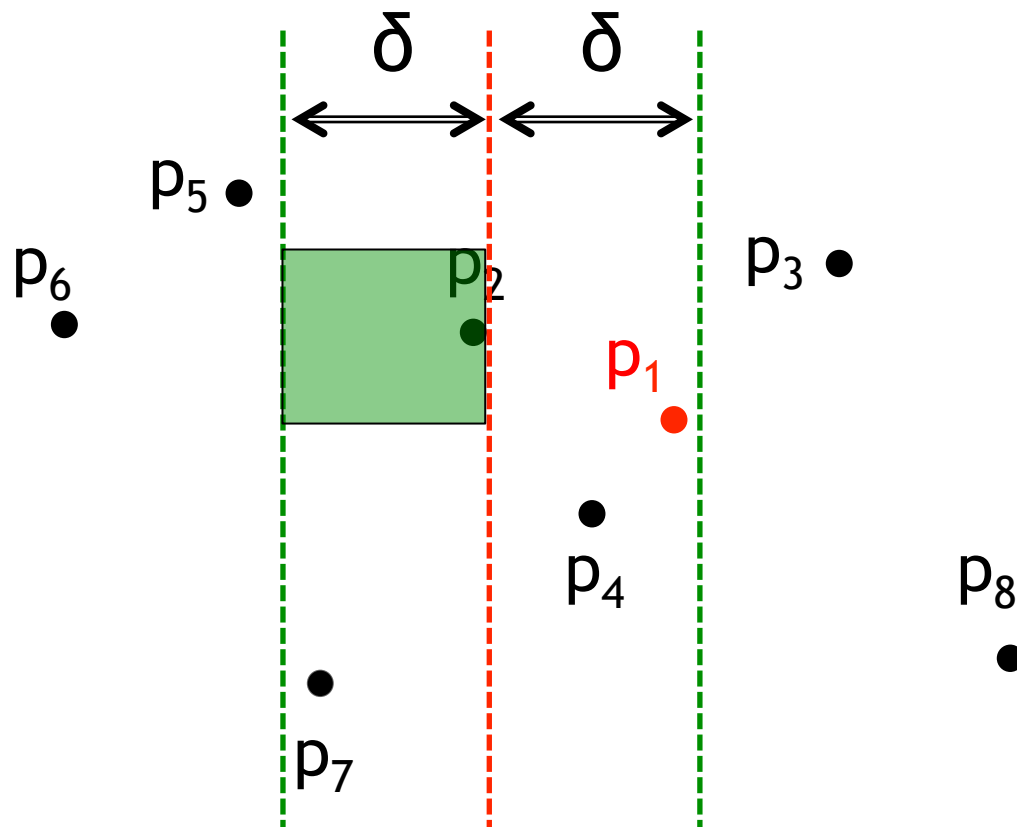
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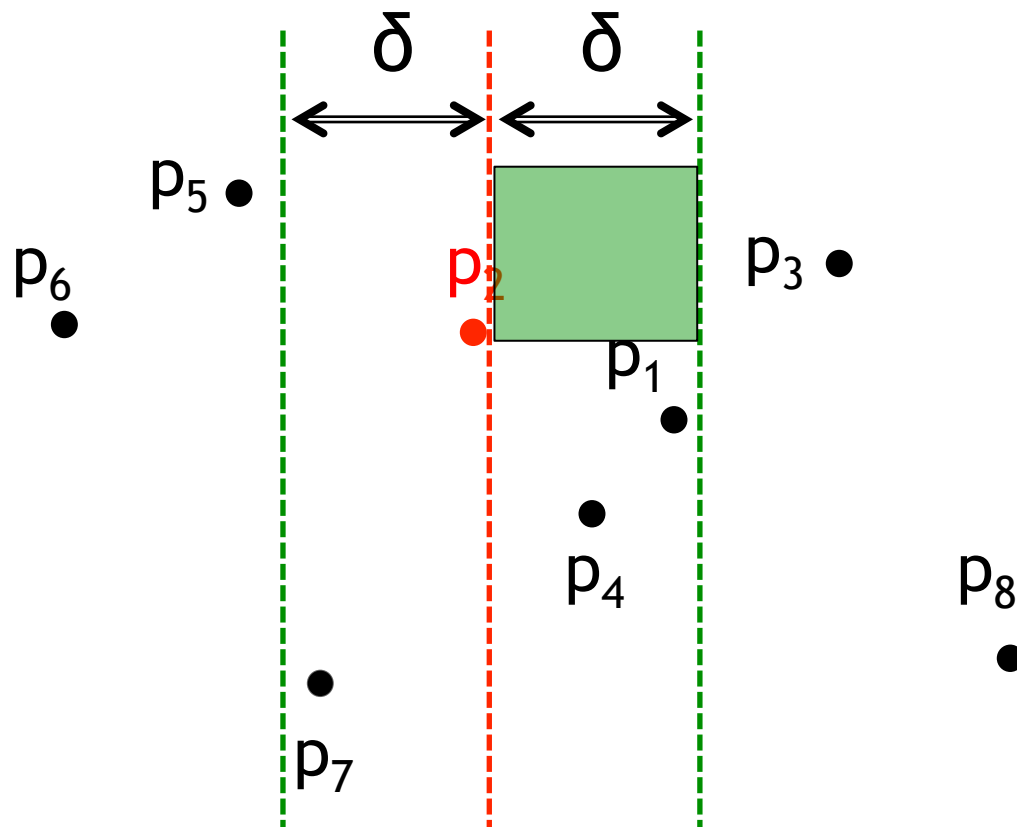
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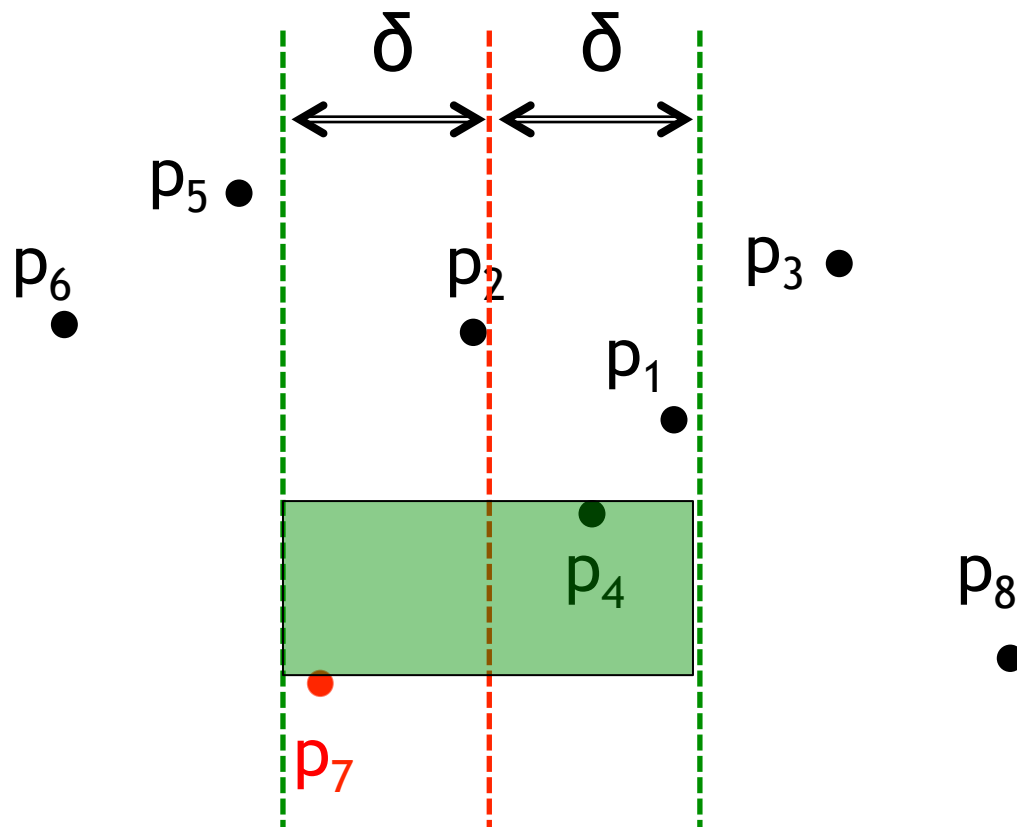
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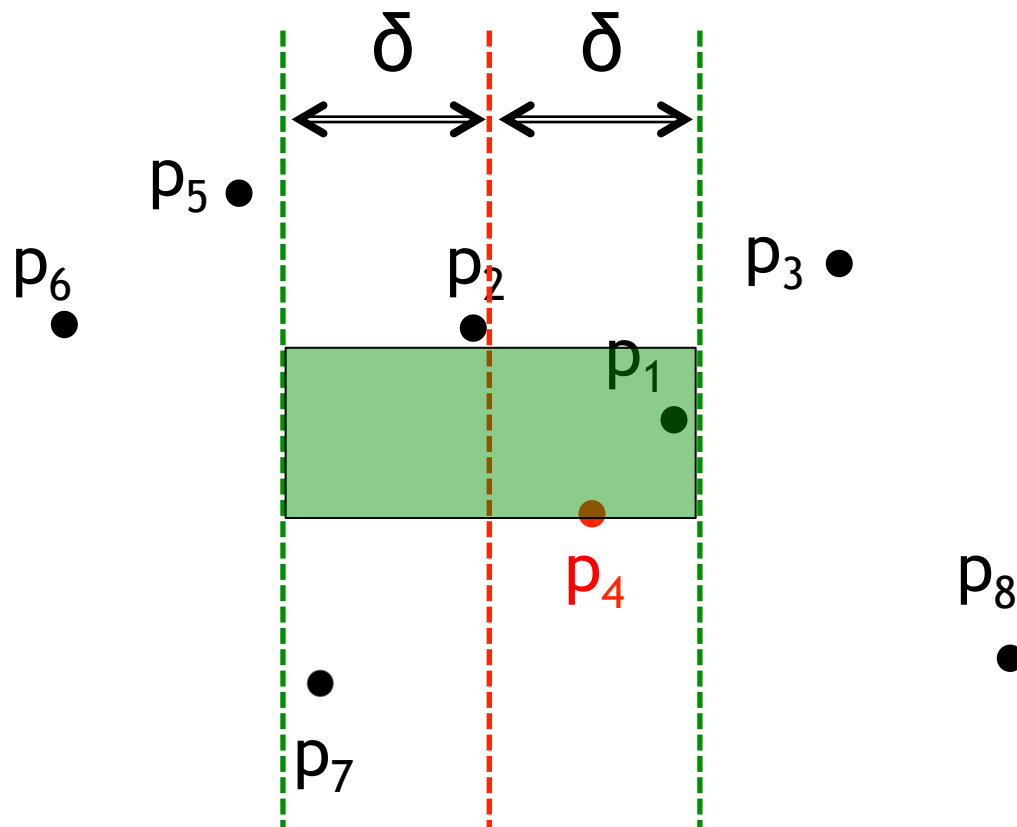
# A More Practical Idea

- ◆ Don't differentiate between same and opposite side
- ◆ Just search the  $2\delta \times \delta$  **above** rectangle each time



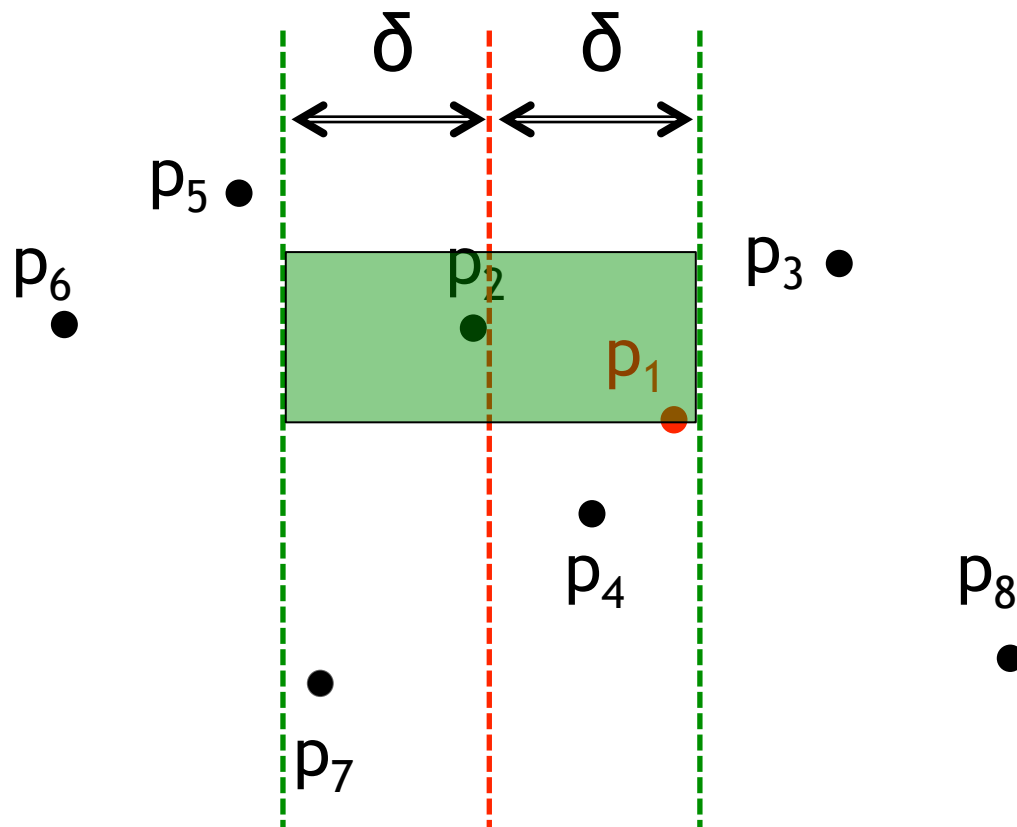
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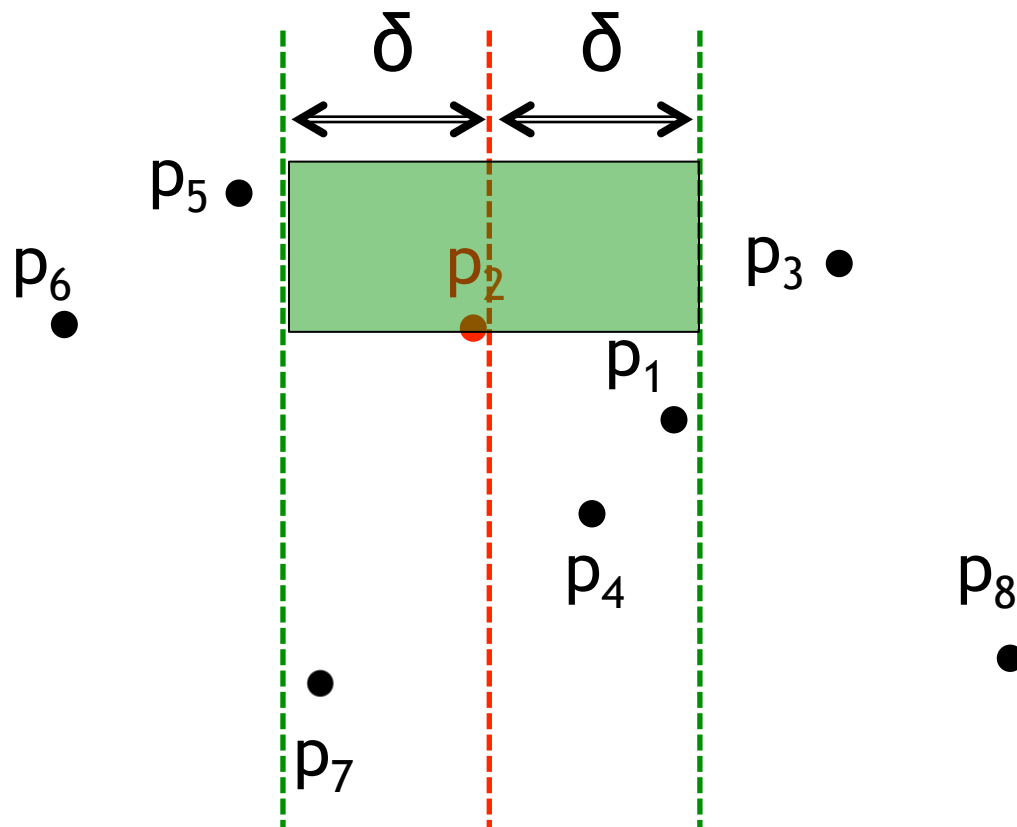
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# DC-CP 1 (1)

---

**procedure** Algorithm(P of n points):

  sort P by x values

  DC-CP1(P)

**procedure** DC-CP1(P sorted by x values):

**if** (P.size  $\leq$  3) compare all & return closest;

  pair<sub>L</sub> = DC-CP1(P[1,...,n/2])

  pair<sub>R</sub> = DC-CP1(P[n/2+1,...,n])

$\delta$  = min(dist(pair<sub>L</sub>, pair<sub>R</sub>))

  pair<sub>S</sub> = findMinSpanningPair( $\delta$ , P)

**return** min(pair<sub>L</sub>, pair<sub>R</sub>, pair<sub>S</sub>)

# DC-CP 1 (2)

```
procedure findMinSpanningPair ( $\delta$ , P):  
  S = select each p in P s.t  $|p_{n/2}.x - p.x| \leq \delta \longrightarrow O(n)$   
  sort(S by increasing y values)  $\longrightarrow O(n \log(n))$   
  minDist =  $+\infty$   
  minPair = null;  
  for i = 1 to S.length:  $\longrightarrow O(n)$   
    j = i+1 (compare S[i] to only the points above S[i])  
    while ( $|S[j].y - S[i].y| \leq \delta$ ):  
      if ( $\text{dist}(S[i], S[j]) < \text{minDist}$ ):  
        minPair = (S[i], S[j]);  
        minDist =  $\text{dist}(S[i], S[j])$   
      j++;  
  return minPair
```

$\longrightarrow ?$

*Q: How many times does the while loop execute?*

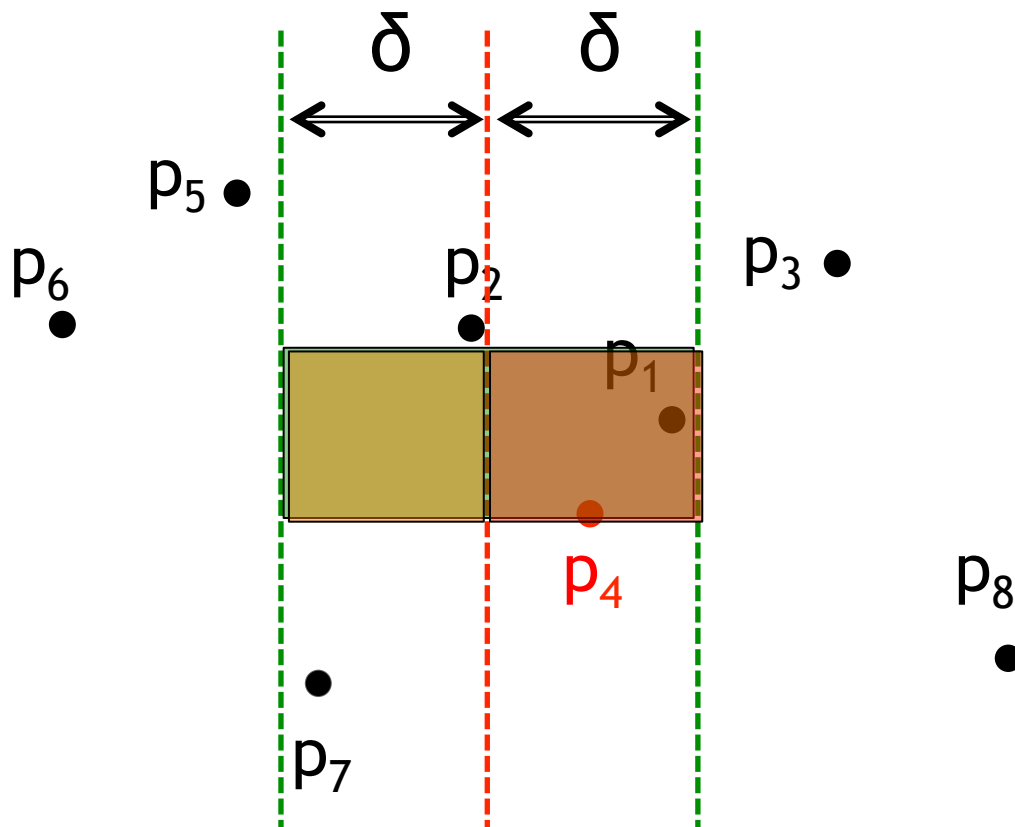
*Claim:  $O(1)$  times*

# For a point $p$ , how many times does while loop execute?

Obs: as many times as there are points in the  $2\delta \times \delta$  rectangle.

Q: How many points can be in a  $2\delta \times \delta$  rectangle?

A: As many as in the left  $\delta \times \delta$  square + right  $\delta \times \delta$  square.



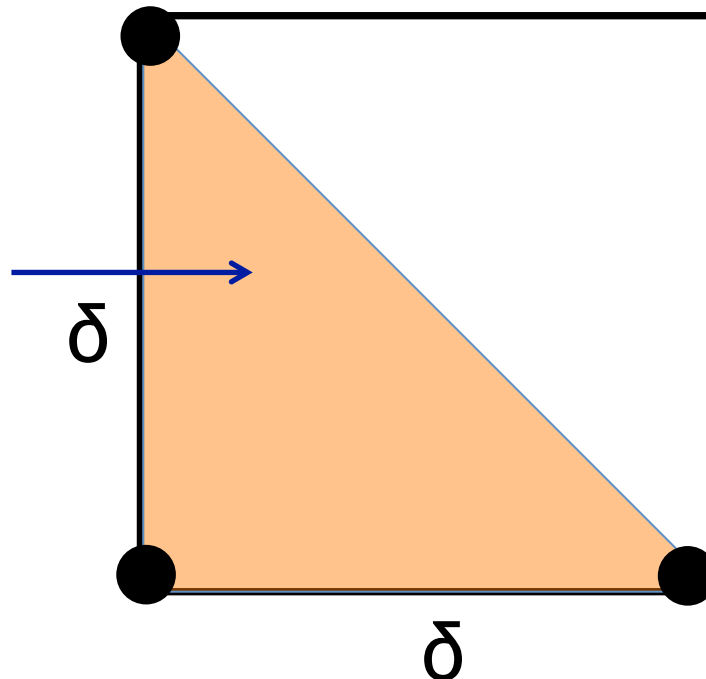
# # Points in a $\delta \times \delta$ Square

Recall: Each point in the square is at least at distance  $\delta$ .

Q1: How many can fit the lower triangle?

A: 3

no other point  
can be inside  
the triangle  
except the  
other two  
corners



# # Points in a $\delta \times \delta$ Square

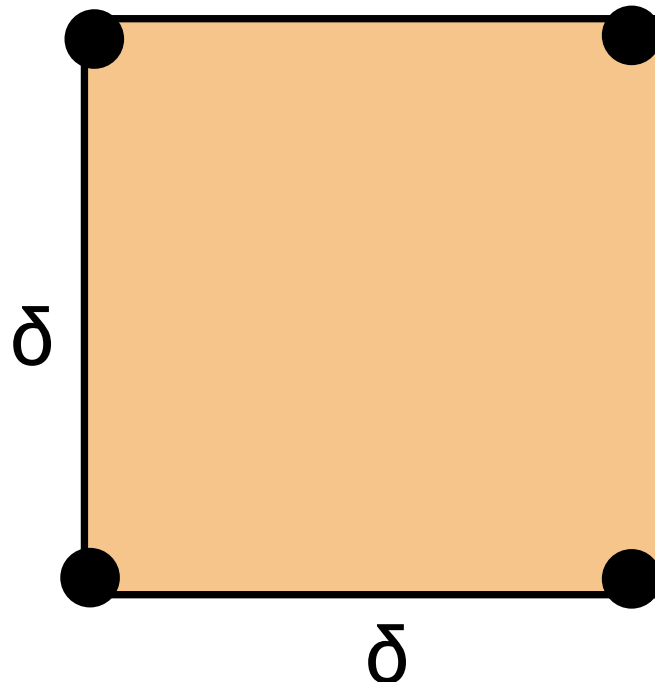
Recall: Each point in the square is at least at distance  $\delta$ .

Q1: How many can fit the lower triangle?

A: 3

Q2: How many can fit the square?

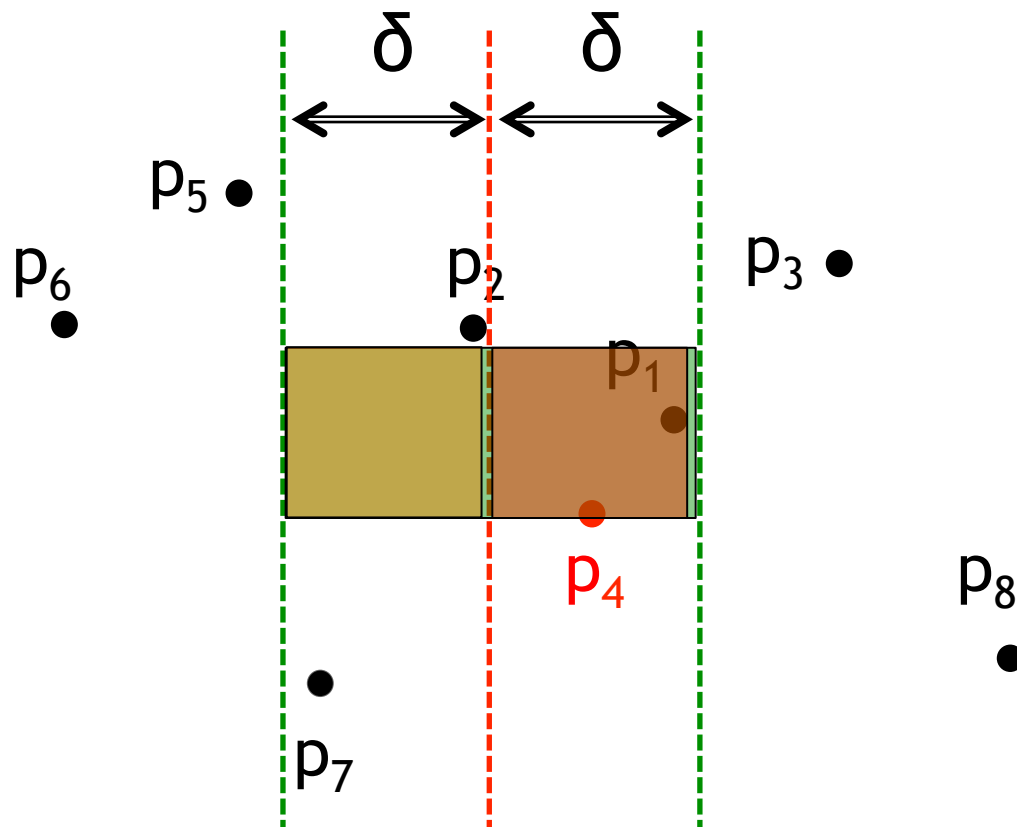
A: 4



# For a point $p$ , how many times does while loop execute?

Obs: as many times as there are points in the  $2\delta \times \delta$  rectangle.

# points in the the  $2\delta \times \delta$  rectangle  $\leq 4 + 4 = 8$



# DC-CP 1 (2)

```
procedure findMinSpanningPair ( $\delta$ , P):  
  S = select each p in P s.t.  $|P[n/2].x - p.x| \leq \delta$  ↗  $O(n)$   
  sort(S by increasing y values) →  $O(n \log(n))$   
  minDist =  $+\infty$   
  minPair = null;  
  for i = 1 to S.length: →  $O(n)$   
    j = i+1  
    while ( $|S[j].y - S[i].y| \leq \delta$ ):  
      if (dist(S[i], S[j]) < minDist) } →  $O(1)$   
        minPair = (S[i], S[j])  
      j++;  
  return minPair
```

*Total:  $O(n \log(n))$*

# DC-CP 1: Runtime Analysis (1)

```
procedure DC-CP1(P sorted by x values):  
  if (P.size  $\leq$  3) compare all & return closest;  
  pairL = DC-CP1(P[1,...,n/2])  
  pairR = DC-CP1(P[n/2+1,...,n])  
   $\delta$  = min(dist(pairL, pairR))  
  pairS = findMinSpanningPair( $\delta$ , P)  
  return min(pairL, pairR, pairS)
```

*Recursive part: Outside Recursive Calls:  $n\log(n)$  work.*

$$T(n) = 2T(n/2) + n\log(n)$$

*Exercise: Show by induction or recursion tree that total work of recursive part is  $O(n\log^2(n))$ .*

*Total Alg Work:  $O(n\log(n)) + O(n\log^2(n)) = O(n\log^2(n))$ .*

*Can improve to  $O(n\log(n))$  by pre-sorting P also by y.*



# Shamos' DC Algorithm (1975) (1)

**procedure** Algorithm(P of n points):

$P_x$  = sort P by x values in increasing order

$P_y$  = sort P by y values in increasing order

DC-Shamos( $P_x$ ,  $P_y$ )

**procedure** DC-Shamos( $P_x$ ,  $P_y$ ):

**if** ( $P_x.size \leq 3$ ) ...;

$P_{yL}$  = select from  $P_y$  points with  $x \leq P_x[n/2].x$

$P_{yR}$  = select from  $P_y$  points with  $x > P_x[n/2].x$

$pair_L$  = DC-Shamos( $P_x[1, \dots, n/2]$ ,  $P_{yL}$ )

$pair_R$  = DC-Shamos( $P_x[n/2+1, \dots, n]$ ,  $P_{yR}$ )

$\delta = \min(\text{dist}(pair_L, pair_R))$

$pair_s$  = findMinSpanningPairShamos( $\delta$ ,  $P_x$ ,  $P_y$ )

**return**  $\min(pair_L, pair_R, pair_s)$

Sorted by  
y already!

# Shamos' DC Algorithm (1975) (2)

Don't need to sort by  $y$ !

Sorted by  $y$  already!

```
procedure findMinSpanningPairShamos( $\delta$ ,  $P_x$ ,  $P_y$ ):  
   $S$  = select each  $p$  in  $P_y$  s.t.  $|P_x[n/2].x - p.x| \leq \delta$   
  minDist =  $+\infty$   
  minPair = null;  
  for  $i = 1$  to  $S.length$ :  $\rightarrow O(n)$   
     $j = i+1$   
    while ( $|S[j].y - S[i].y| \leq \delta$ ):  
      if ( $\text{dist}(S[i], S[j]) < \text{minDist}$ ):  
        minPair = ( $S[i], S[j]$ )  
       $j++$   
  return minPair
```

$\rightarrow O(1)$

*Total:  $O(n)$*

# Runtime Analysis of Shamos' Algorithm

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*Key Idea of Shamos: Avoid sorting by y values in each recursive call by pre-sorting P by y values.*

*Recursive part: Outside Recursive Calls:  $O(n)$  work.*

$$T(n) = 2T(n/2) + O(n)$$

*By Master Thm, total:  $O(n\log(n))$*

*Total Work for Shamos*

$$O(n\log(n)) + O(n\log(n)) = O(n\log(n)).$$