Closest Pairs

COMP 3711H - HKUST Version of 9/11/2016 M. J. Golin

Outline

- Introduction
- The Gridding Lemma
- An $O(n \log n)$ sweep line algorithm
- An O(n) randomized algorithm
 - Hashing a point set
 - The algorithm

Introduction

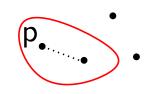
- Most algorithms seen so far have assumed that items have
 1-dimensional weights or keys, i.e., they can be ordered.
- There is often a need, e.g., graphics, data bases, to manipulate multi-dimensional data. The area of algorithms that treats geometric (multi-dimensional) data is called computational geometry.
- In this set of slides, we will see two algorithms for solving the 2-dimensional closest-pair problem.
- The first algorithm will take $O(n \log n)$ worst case time to find the closest pair among n points. The second algorithm will be randomized and only require O(n) average time.

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 - Set $d(p, P) = \min_{p' \in P \{p\}} d(p, p')$ to be distance of nearest point to p.

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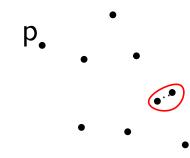


ullet distance between p and p' is

$$d(p, p') = \sqrt{(p.x - p'.x)^2 + (p.y - p'.y)^2},$$

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- distance between p and p' is $d(p,p') = \sqrt{(p.x-p'.x)^2+(p.y-p'.y)^2}$,
 - Set $d(p, P) = \min_{p' \in P \{p\}} d(p, p')$ to be distance of nearest point to p.
 - Set $\delta(P) = \min_{p \in P} d(p, P)$ to be the closest pair distance in P.



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 - Set $\delta(P) = \min_{p \in P} d(p, P)$ to be the closest pair distance in P.
 - The Closest-Pair Problem is to find $p, p' \in P$ such that $d(p, p') = \delta(P)$

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The Gridding Lemma:

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Let P=\{p_1,p_2,\ldots,p_n\} be a set of points and \delta=\delta(P) their closest-pair distance. Grid the plane into \frac{\delta}{2}\times\frac{\delta}{2} boxes using horizontal and vertical lines. Then
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- (a) No box holds more than 1 point
- (b) If $d(p, p') = \delta$ then the boxes containing p and p'

are within two horizontal and two vertical boxes of each other.

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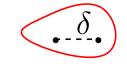
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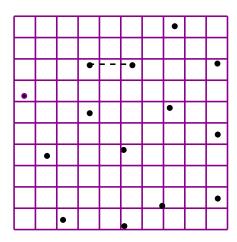
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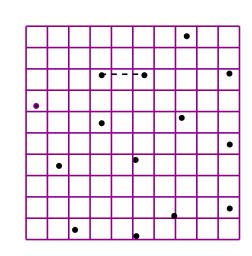
- (a) No box holds more than 1 point
- (b) If $d(p, p') = \delta$ then the boxes containing p and p'

are within two horizontal and two vertical boxes of each other.

Proof:

- (a) if two points are in same box they can be at most $\delta/\sqrt{2} < \delta$ away from each other, contradicting def of δ .
- (b) if two points p,p^\prime are more than two boxes horizontally (or vertically) away from each other then

$$|p.x - p'.x| > \delta$$
 ($|p.y - p'.y| > \delta$), so $d(p, p') > \delta$ contradicting definition of δ .



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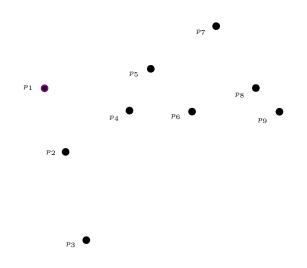
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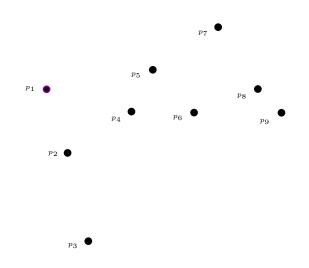
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Sort the points by \boldsymbol{x} coordinate.



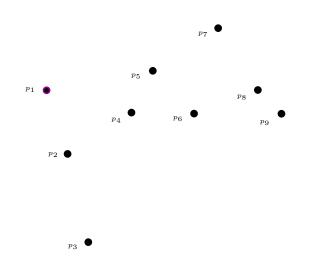
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Sort the points by x coordinate.

Set
$$P_i = \{p_1, \dots, p_i\}.$$

Note
$$\delta(P_i) = \min(\delta(P_{i-1}), d(p_i, P_{i-1}))$$

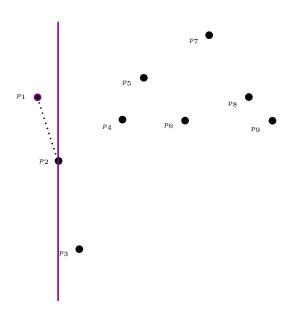


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At end,
$$\delta = \delta(P_n)$$



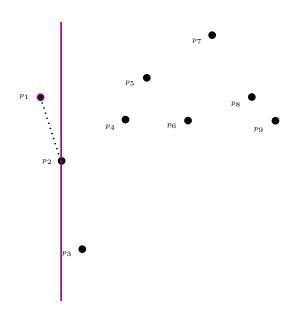
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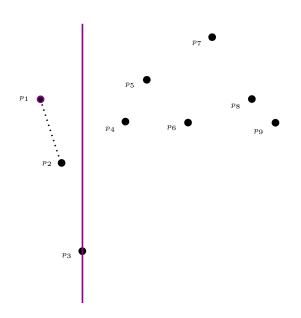
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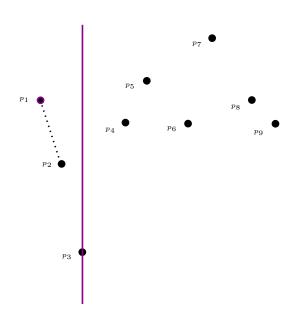
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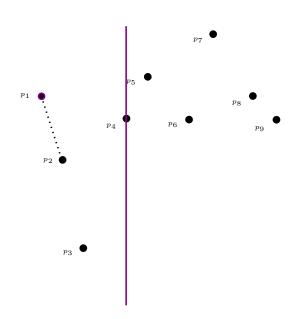
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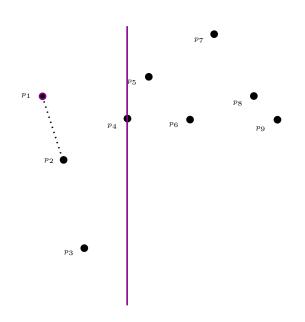
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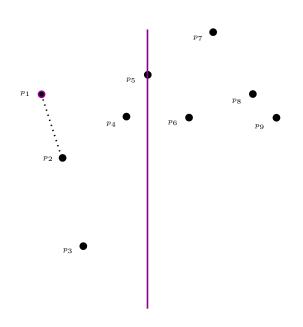
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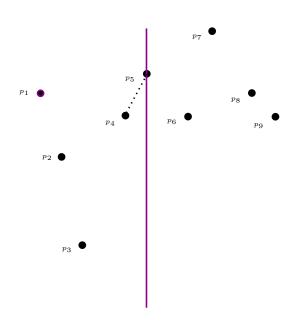
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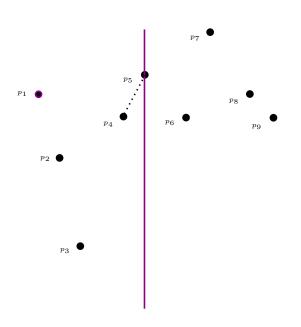
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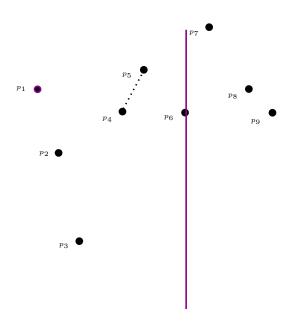
$$\delta(P_5) = d(p_5, p_4)$$

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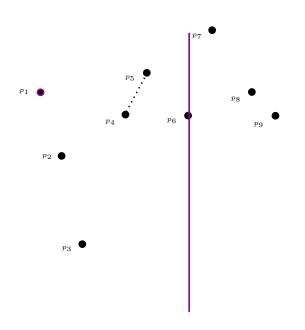
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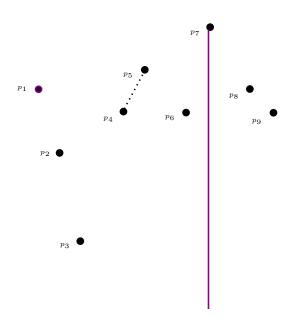
$$\delta(P_6) = d(p_5, p_4)$$

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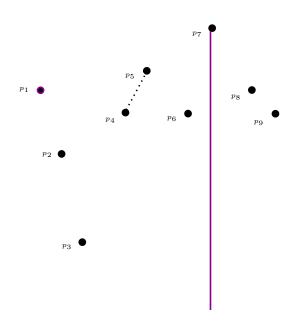
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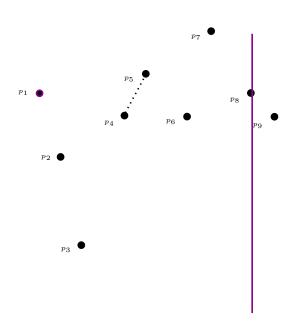
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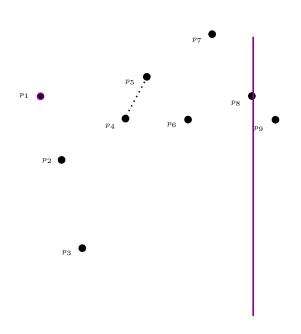
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Sort the points by x coordinate.

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Sort the points by x coordinate.

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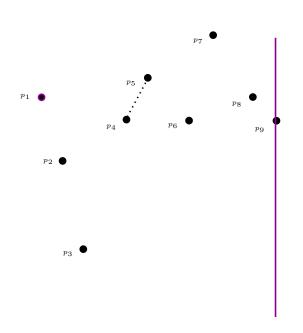
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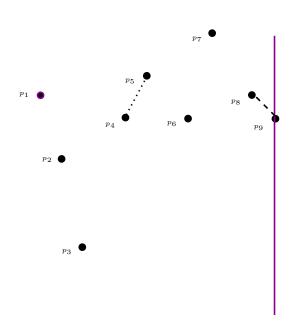
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A Sweep Algorithm



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Algorithm sweeps a vertical line through points, i.e., p_2, p_3, \ldots, p_n , at each step using equation to calculate $\delta(P_i)$.

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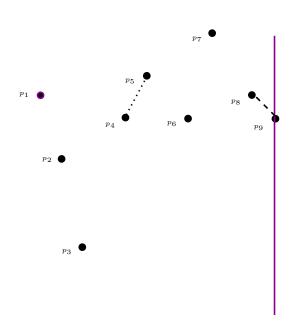
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A Sweep Algorithm



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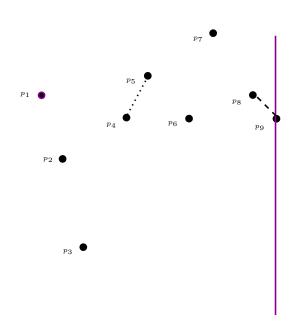
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$$\delta(P_i) = \min(\delta(P_{i-1}), d(p_i, P_{i-1}))$$

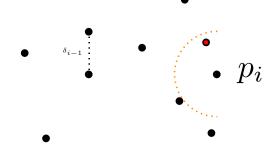
Set $\delta_i = \delta(P_i)$ and $d_i = d(p_i, P_{i-1})$.

If
$$d_i < \delta_{i-1}$$
 then $d_i = d(p_i, p_j)$ for some $j < i$ with
$$|p_i.x - p_j.x| \le d_i = \delta_i < \delta_{i-1}$$

$$\delta(P_i) = \min(\delta(P_{i-1}), d(p_i, P_{i-1}))$$

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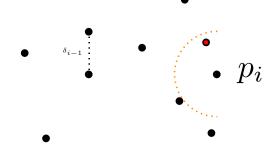
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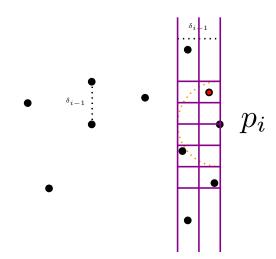
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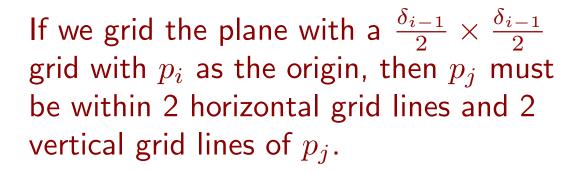
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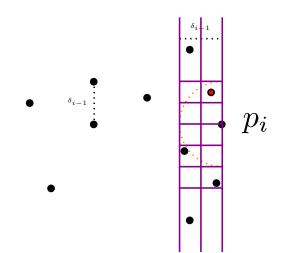


$$\delta(P_i) = \min(\delta(P_{i-1}), d(p_i, P_{i-1}))$$

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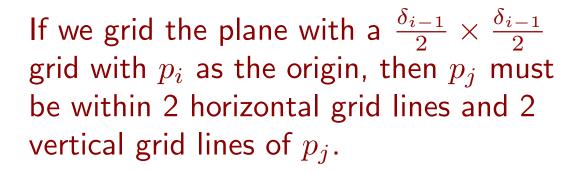


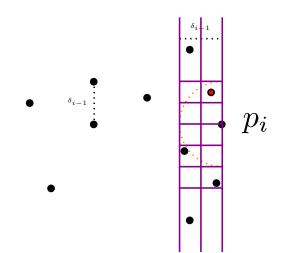


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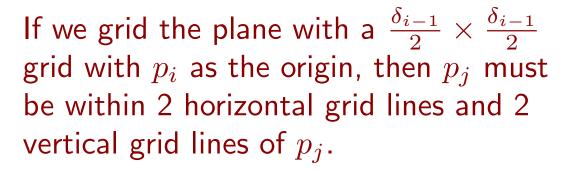
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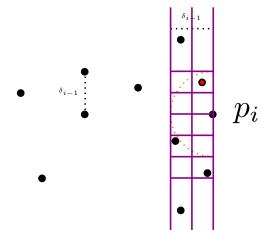


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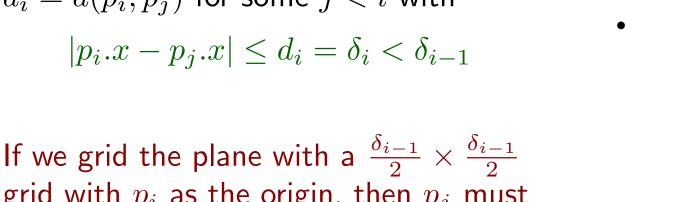


Gridding Lemma tells us there is at most one point per box.



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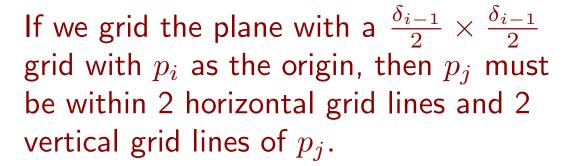
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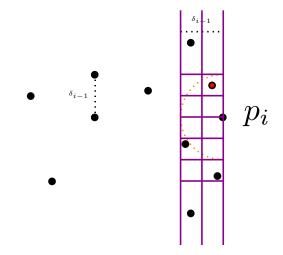
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This implies

if
$$d_i < \delta_{i-1}$$
 then

$$d_i = (p_i, p)$$
 where

(a)
$$p \in S_i$$
 and

(b)
$$p$$
 is one of the 4 points above p_i in S_i

or

p is one of the 4 points below p_i in S_i .

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We could then calculate δ_i in $O(\log n)$ time as follows

- In S_i , find the 4 points above p_i and four points below it This takes $O(\log n)$ time
- Find the distances between these (at most) 8 points and p_i . This takes O(1) time. Call this minimum distance d'(note that there might be no points in S_i . In this case $d' = \infty$)
- Set $\delta_i = \min(\delta_{i-1}, d')$

The Full Sweep Algorithm

```
Let S_i = \{ p \in P_{i-1} : p_i.x - p.x < \delta_{i-1} \}
```

We just developed the following algorithm to calculate $\delta_2, \delta_3, \ldots, \delta_n$

- A Set $\delta_2=d(p_1,p_2)$; Create a y-sorted balanced BST on $S_2=\{p_1\}$
- B For i = 3 to n
 - 1. Update the BST from storing S_{i-1} to storing S_i
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If we can implement step (B)(1) in $O(n \log n)$ total time the entire algorithm then is $O(n \log n)$.

The Problem: Given p_1, p_2, \ldots, p_n sorted by nondecreasing x-coordinate and a sequence $\delta_2 \geq \delta_3 \geq \cdots \geq \delta_{n-1}$, iteratively construct balanced BSTs storing the sets S_2, S_3, \ldots, S_n sorted by y-coordinate. Use only $O(n \log n)$ time in total.

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so S_i is a continuous sequence of points $p_k, p_{k+1}, \ldots, p_{i-1}$.

Furthermore, if $p_k \not\in S_{i-1}$ then $p_k \not\in S_i$, because

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Every point is added once and removed at most once, so we perform O(n) inserts and O(n) deletes from the BST, requiring only $O(n \log n)$ time

Odds & Ends

- This approach of sorting the points by x then updating solution by moving from left to right is a common one in computational geometry. It is called the sweep-line technique.
- It is possible to prove that $O(n \log n)$ is a lower bound for solving this problem in the algebraic decision tree model of computation (a generalization of the comparison tree model we used for sorting). So, our algorithm is optimal in this model.

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- With randomization and the use of the floor function $(\lfloor x \rfloor)$ this can be impoved to an O(n) average case time randomized algorithm.

Outline

- Introduction
- The Gridding Lemma
- An $O(n \log n)$ sweep line algorithm
- An O(n) randomized algorithm
 - Hashing a point set
 - The algorithm

A Randomized Algorithm

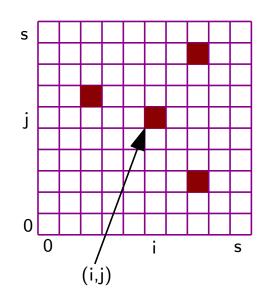
- In this section we will see a randomized algorithm for finding the closest pair in O(n) average case time.
- ullet Input is n two-dimensional points.
- The randomization will come from two places
 - The first step of the algorithm will be to randomly order the input. That is, it will choose one of the n! possible orderings of the n points uniformly at random and label the points as p_1, p_2, \ldots, p_n in that order
 - The only other place randomization is used will be in the use of (universal) hashing
- Before starting the main part of the algorithm we perform an O(n) scan finding x_{\min} , x_{\max} , y_{\min} y_{\max} , the smallest and largest x and y coordinates among all of the points.

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Hashing a Sparse Array

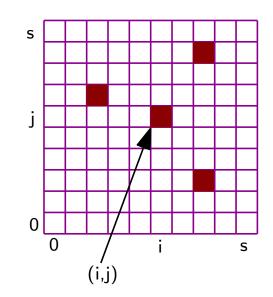
- Given an $s \times s$ array in which at most n items exist (are non-zero).
- We would like to be able to perform the following two operations in O(1) time each.
 - Insert(x): Insert new item x into the array
 - Search(x): If x exists, return it (else return nil)
- In practice, array item is a pointer to a record. Successful search returns pointer to record's data.



Hashing a Sparse Array

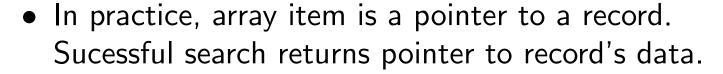
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 Successful search returns pointer to record's data.
- This can be solved using hashing
- Array index is (i,j) with $0 \le i,j \le s$. Can map array indices uniquely to integer $i+js < s^2+1$



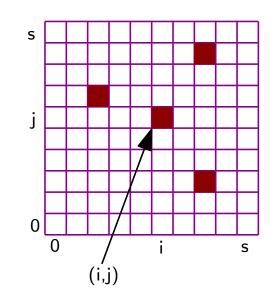
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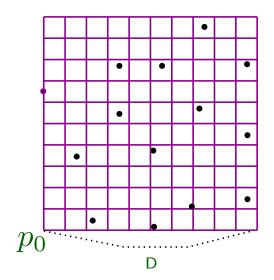


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- Applying **universal hashing** with $U=s^2+1$ and m=n gives O(1) average case insert and search



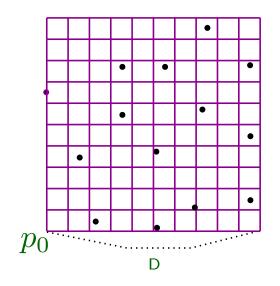
Bucketing: Hashing a Grid

- Let $p_0 = (x_{\min}, y_{\min})$; $D = \max(x_{\max} x_{\min}, y_{\max} y_{\min}).$
- The entire point set is contained in the square of side D with lower-left corner p_0 .
- A grid of size d will denote a horizontal/vertical grid with grid separation d between adjacent parallel lines and p_0 as the origin.



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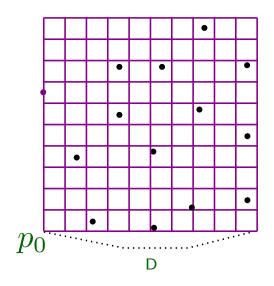
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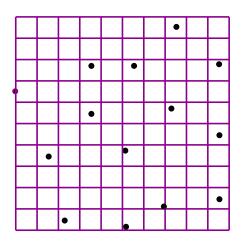
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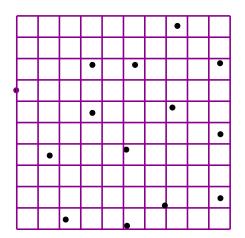
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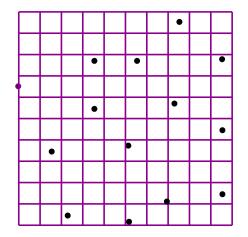
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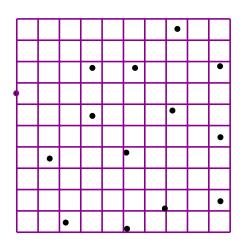
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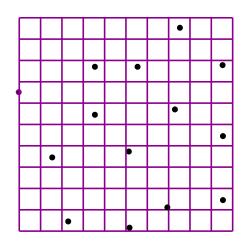


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- Let $P_i = \{p_1, p_2, \dots, p_n\}$, $\delta_i = \delta(P_i)$ and grid size is $d = \delta_i/2$
- Gridding Lemma says that each grid box contains at most 1 point
- ullet \Rightarrow Finding points in a grid box takes average O(1) time

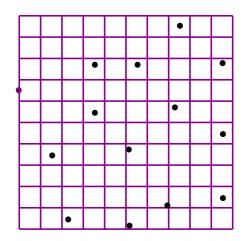
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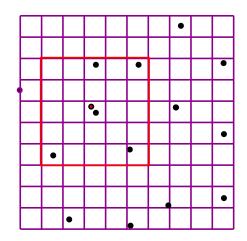
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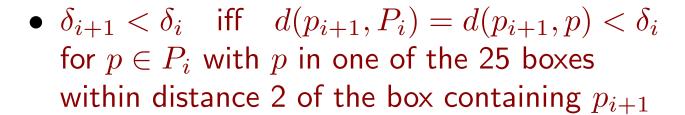
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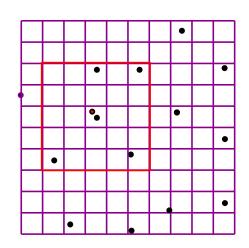


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- $\delta_{i+1} < \delta_i$ iff $d(p_{i+1}, P_i) = d(p_{i+1}, p) < \delta_i$ for $p \in P_i$ with p in one of the 25 boxes within distance 2 of the box containing p_{i+1}



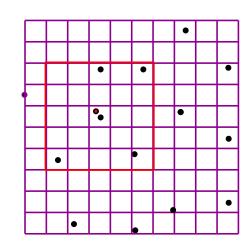
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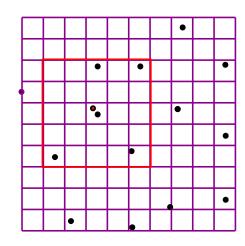
- Using the hash search, these 25 boxes and their contents (at most one point each) can be found in O(1) time. In O(1) further time, find d', the min distance between p_{i+1} and these ≤ 25 points.
- If $d' \geq \delta_i$ then $\delta_{i+1} = \delta_i$ \Rightarrow insert p_{i+1} into the grid in O(1) average time If $d' < \delta_i$ then $\delta_{i+1} = d'$ \Rightarrow Throw away the entire old grid and in O(i) average time, regrid P_{i+1} with the new d = d'/2

• Randomly label the points as p_1, p_2, \ldots, p_n Perform an O(n) scan finding $x_{\min}, x_{\max}, y_{\min}$ y_{\max} , the smallest and largest x and y coordinates among all of the points



- Set $\delta_2 = d(p_1, p_2)$. Hash P_2 into the grid with $d = \delta_2/2$.
- For i=3 to n do
- (A) Find the min distance d' between p_{i+1} and the ≤ 25 points in the 25 grid boxes surrounding p_{i+1}
- (B) If $d' > \delta_i$ then (i) Set $\delta_{i+1} = \delta_i$ and insert p_{i+1} into old grid else $(d' < \delta_i)$
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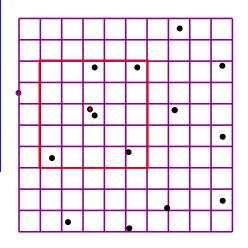


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Average Run Time

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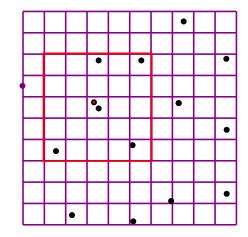


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Average Run Time

1st Step: O(n)

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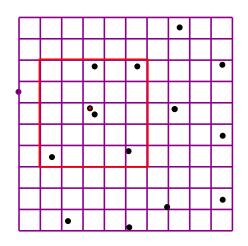
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- Average Run Time

1st Step: O(n)

2nd Step: O(1)

- (B) If $d' > \delta_i$ then (i) Set $\delta_{i+1} = \delta_i$ and insert p_{i+1} into old grid else $(d' < \delta_i)$
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• Randomly label the points as p_1, p_2, \ldots, p_n Perform an O(n) scan finding x_{\min} , x_{\max} , y_{\min} y_{max} , the smallest and largest x and y coordinates among all of the points



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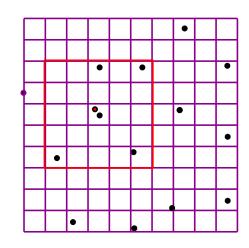
Average Run Time

1st Step: O(n)

2nd Step: O(1)

All 3(A): O(n)

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Average Run Time

1st Step: O(n)

2nd Step: O(1)

All 3(A): O(n)

All 3(B)(i): O(n)

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Average Run Time

1st Step: O(n)

2nd Step: O(1)

All 3(A): O(n)

All 3(B)(i): O(n)

= O(n) + time for all 3(B)(ii)

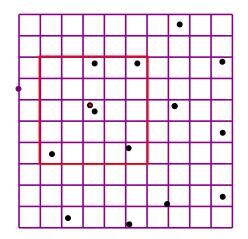
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- Let X_i be amount of work done by 3(B)(ii) at step i+1 $X_i=0$ with Prob $1-\frac{2}{i+1}$; $X_i=O(i+1)$ with Prob $\frac{2}{i+1}$. \Rightarrow expected value $E(X_i)=\frac{2}{i+1}O(i+1)=O(1)$

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- Total work by 3(B)(ii) is $Y = \sum_{i=2}^{n-1} X_i$. Then total average work is $E(Y) = E\left(\sum_{i=2}^{n-1} X_i\right) = \sum_{i=2}^{n-1} E(X_i) = O(n)$

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Average Run Time

1st Step: O(n)

2nd Step: O(1)

All 3(A): O(n)

All 3(B)(i): O(n)

$$= O(n) + time$$
 for all 3(B)(ii)

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Average Run Time

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2nd Step: O(1)

All 3(A): O(n)

All 3(B)(i): O(n)

All 3(B)(ii): O(n)

= O(n)