

Computational Geometry

Lecture #4 — Closest Pairs II

Recap:

Master Theorem

“Median of Medians” Algorithm

Closest pair — Bounds

Closest Pair — Divide and conquer

Master Theorem

Analysis of recursive algorithms

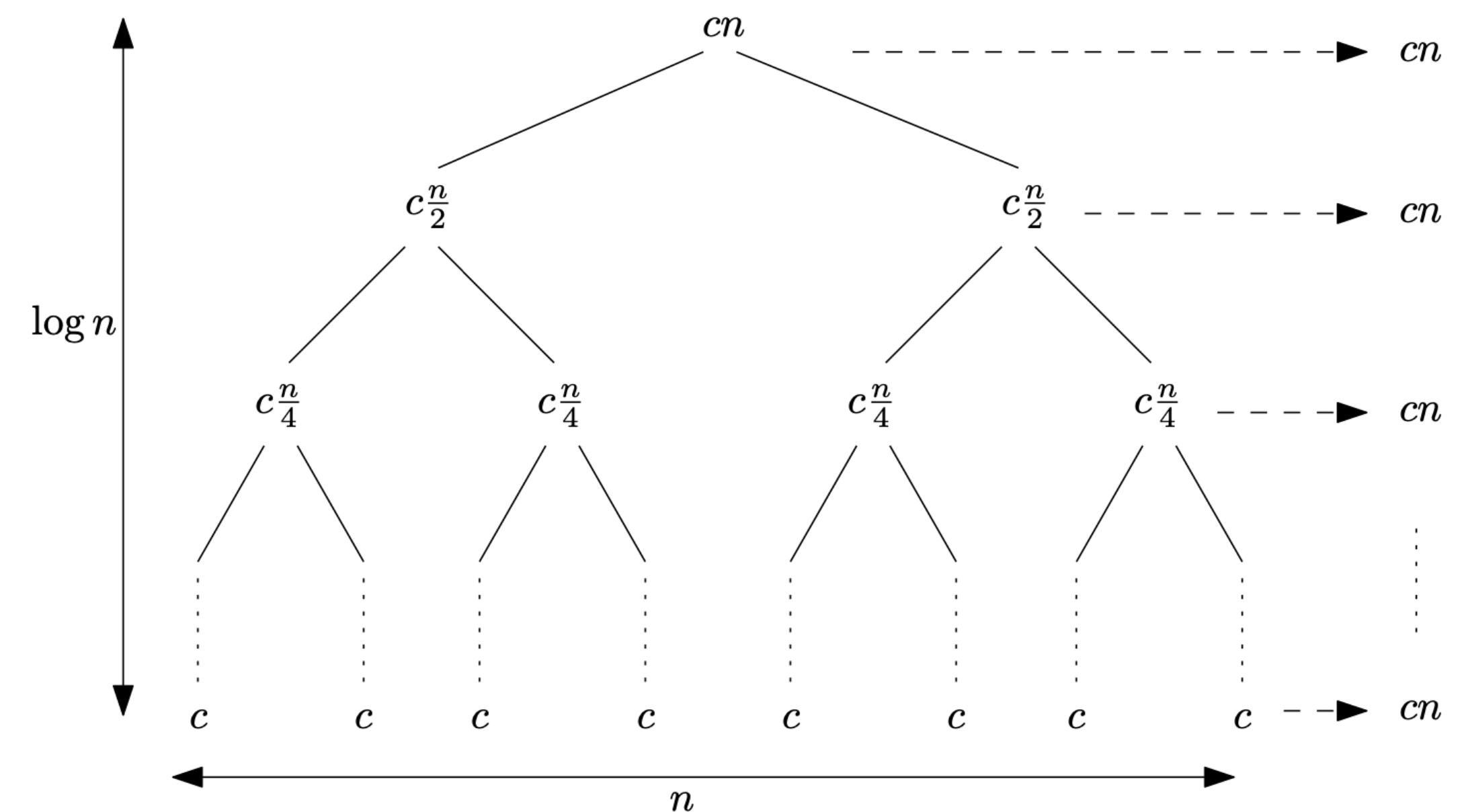
- Runtime based on subproblem size reduction before recursion, in sum:
 - “less than original” $\Rightarrow \Theta(n^k)$
 - “same as original” $\Rightarrow \Theta(n^k \log n)$
 - “more than original” $\Rightarrow \Theta(n^c)$
- k determined by cost of “merge” step
- e.g., Mergesort has $\Theta(n \log n)$

Theorem 3.3 (Master Theorem) Let $T : \mathbb{N} \rightarrow \mathbb{R}$ with

$$T(n) = \sum_{i=1}^m T(\alpha_i \cdot n) + \Theta(n^k),$$

where $\alpha_i \in \mathbb{R}$ with $0 < \alpha_i < 1$, $m \in \mathbb{N}$ and $k \in \mathbb{R}$. Then

$$T(n) \in \begin{cases} \Theta(n^k) & \text{for } \sum_{i=1}^m \alpha_i^k < 1 \\ \Theta(n^k \log(n)) & \text{for } \sum_{i=1}^m \alpha_i^k = 1 \\ \Theta(n^c) & \text{with } \sum_{i=1}^m \alpha_i^c = 1 \text{ for } \sum_{i=1}^m \alpha_i^k > 1 \end{cases}$$



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“Median of Medians” Algorithm

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“Median of Medians” Algorithm

Medians [Blum, Floyd, Pratt, Rivest, Tarjan 1973]

Theorem 3.4: A median of n numbers can be computed in $\mathcal{O}(n)$.

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Theorem 3.4: A median of n numbers can be computed in $\mathcal{O}(n)$.

- ▶ Group the numbers into sets of 5.

1	6	11	17	3
22	18	16	5	9
10	21	2	12	15
13	4	20	19	7
24	25	8	14	23

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- ▶ Use the *median of medians* as a pivot.

≥ n/4 numbers

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3	2	1	5	4
7	8	10	12	6
9	11	13	14	18
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$$T\left(\frac{3n}{4}\right)$$

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Theorem 3.4: A median of n numbers can be computed in $\mathcal{O}(n)$.

$$T(n) = T\left(\frac{n}{5}\right) + T\left(\frac{3n}{4}\right) + \Theta(n)$$

$$\sum_{i=1}^m \alpha_i^k = \frac{1}{5} + \frac{3}{4} = \frac{19}{20} < 1.$$

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Recap:

Closest pair — Bounds

Closest Pair — Divide and conquer

Lower bound for the Closest pair Problem

Via SORTING and ELEMENT UNIQUENESS

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Relevant Paper: Ben-Or, "Lower Bounds For Algebraic Computation Trees"

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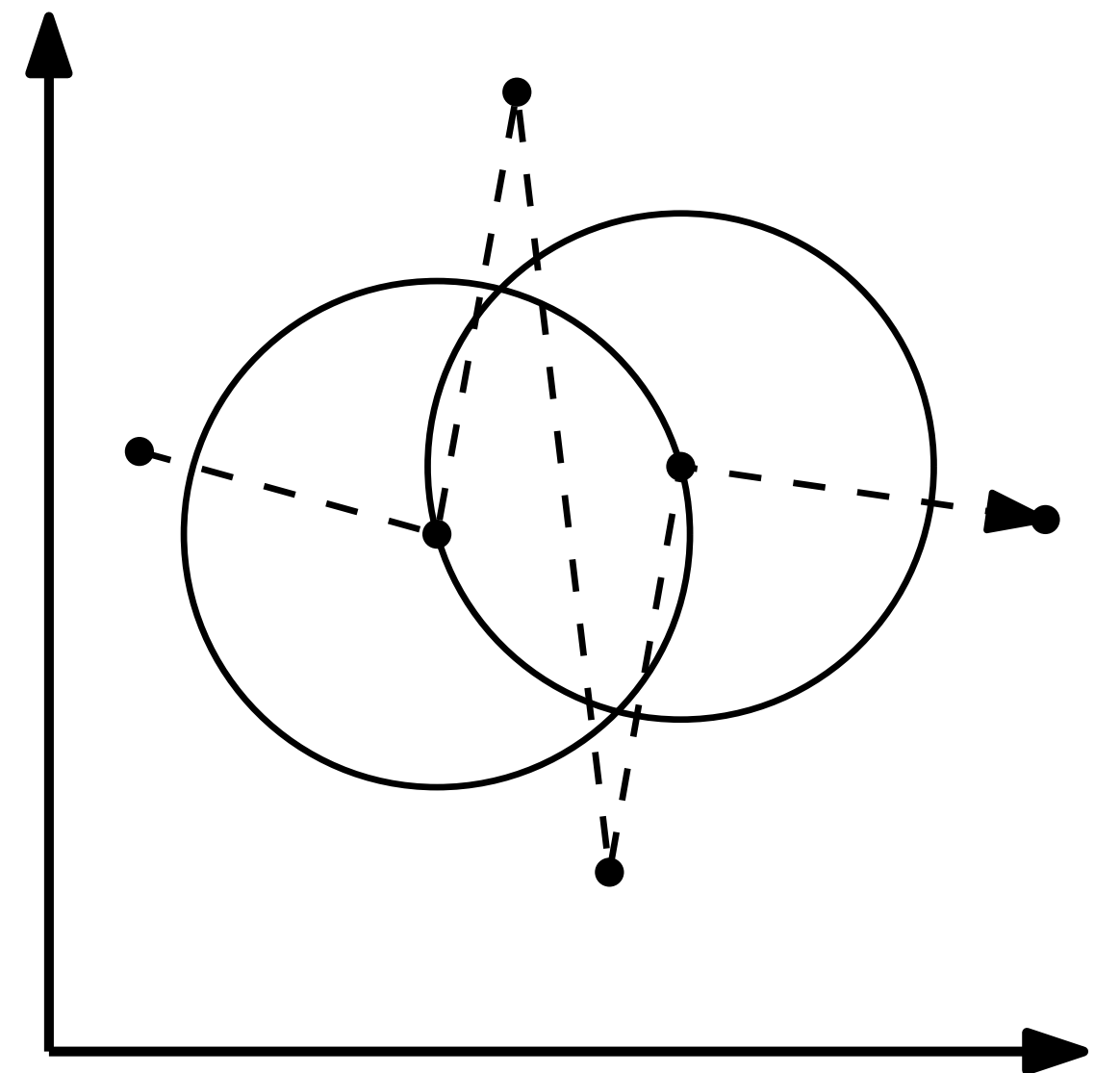
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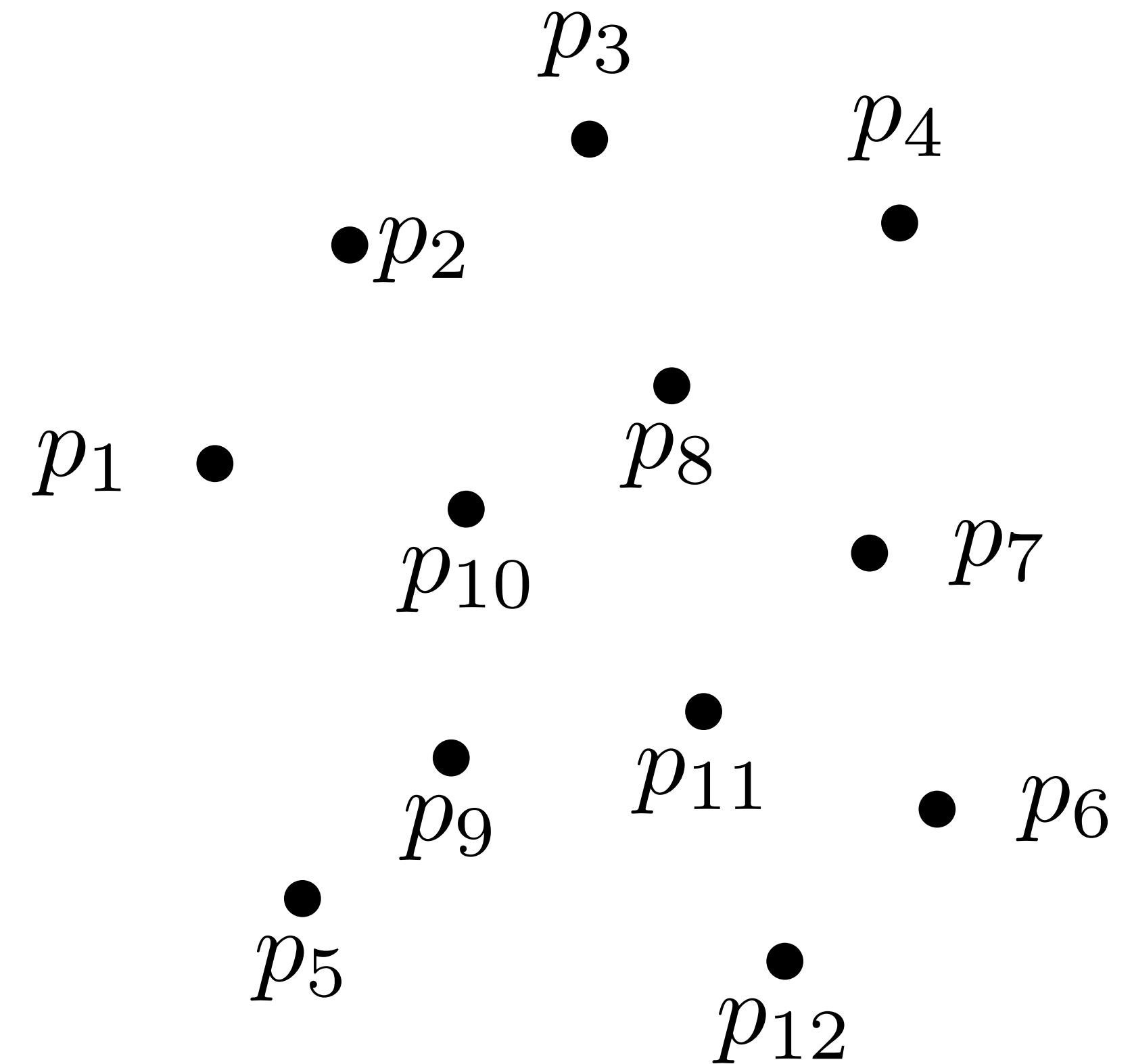
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Bentley & Shamos, 1976

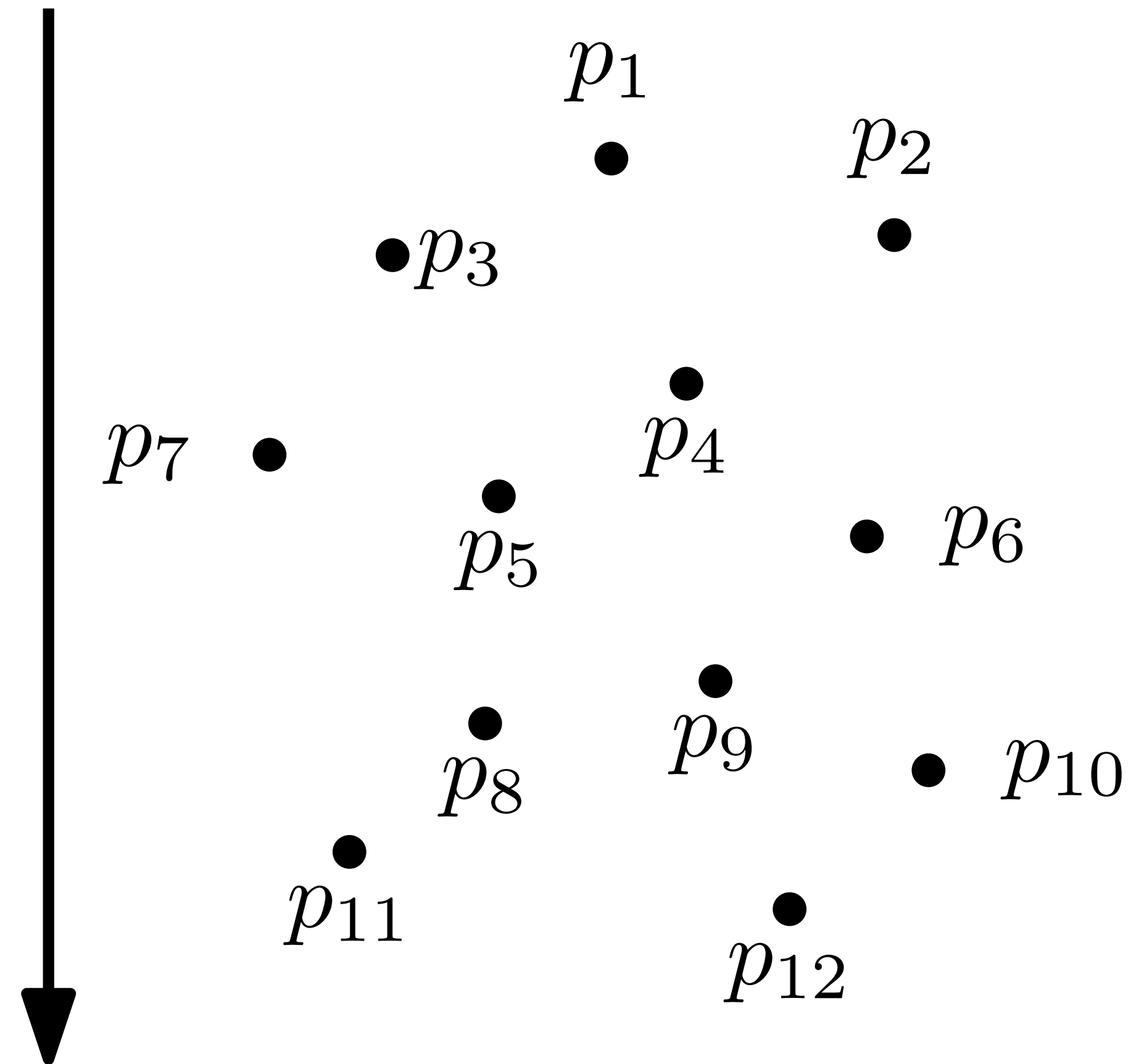
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Closest Pair — Divide and conquer

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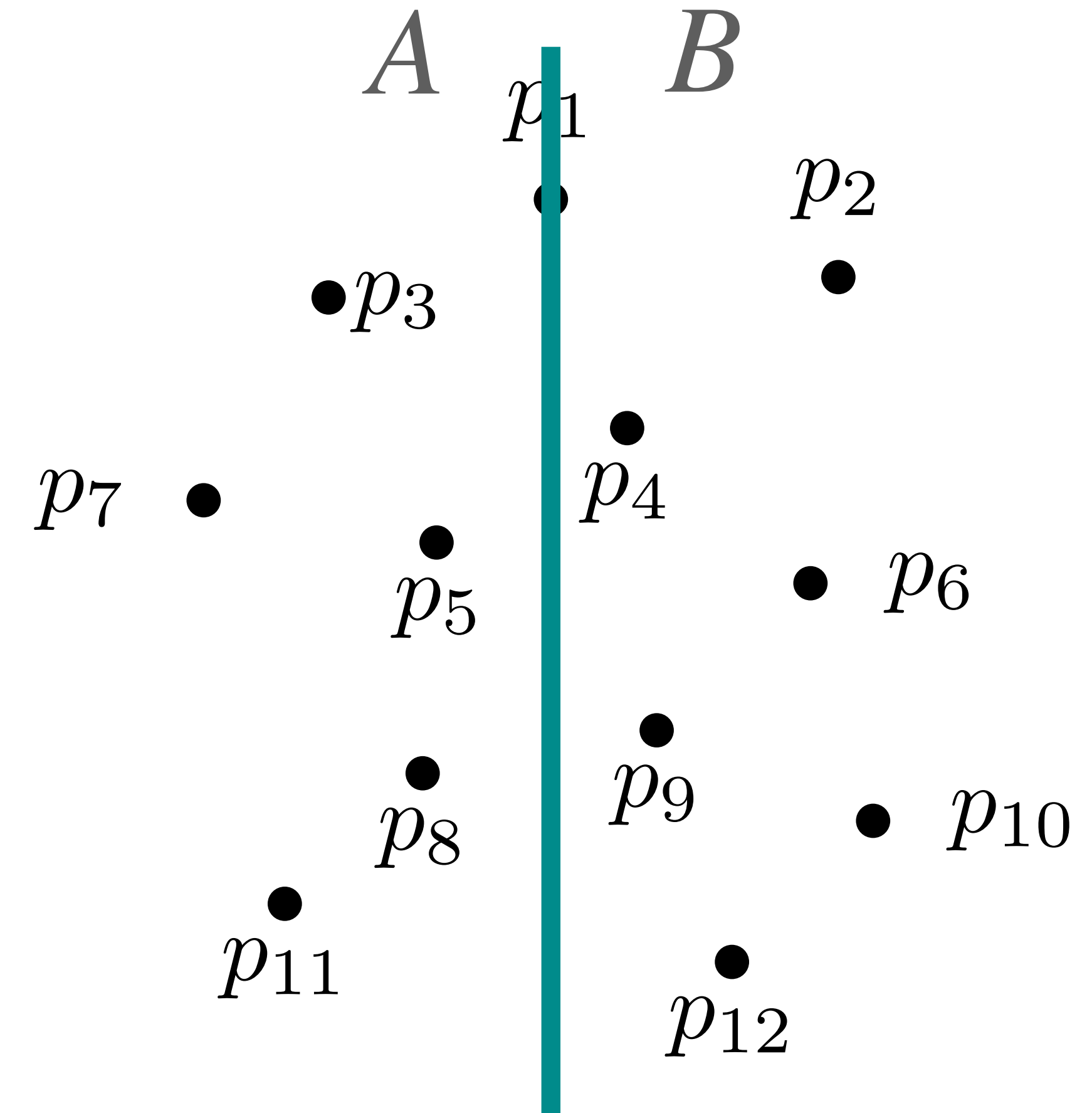
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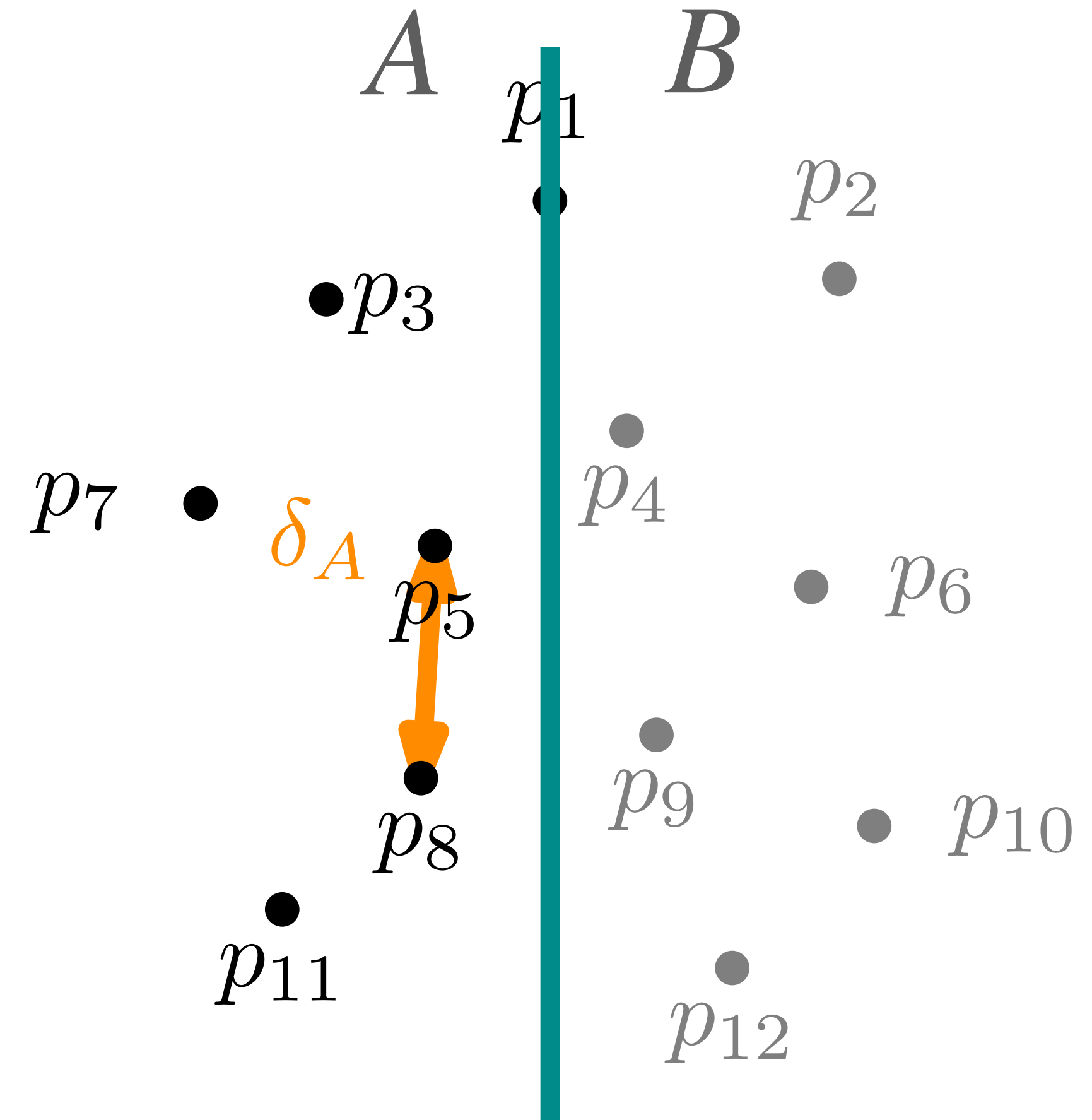
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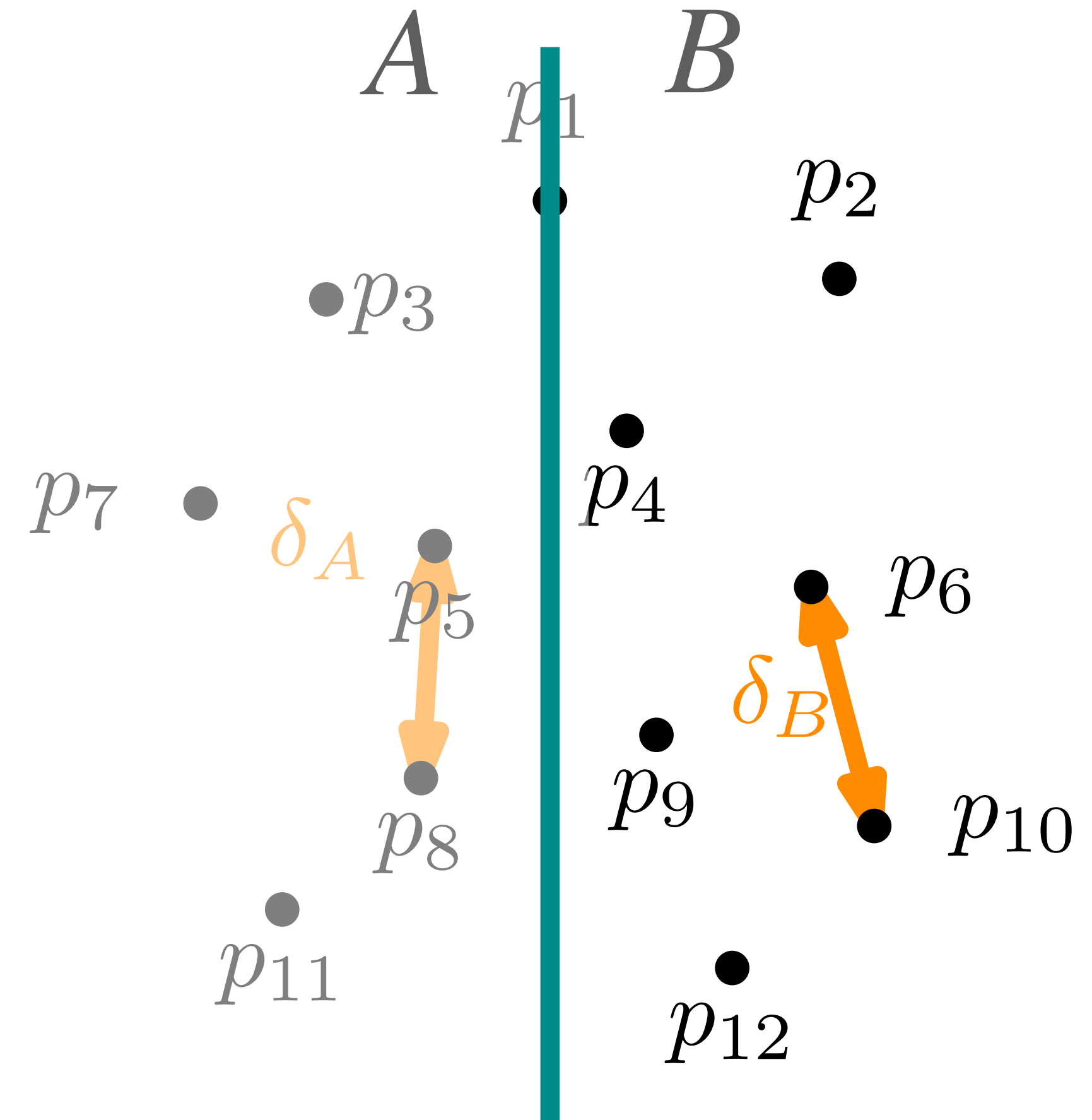
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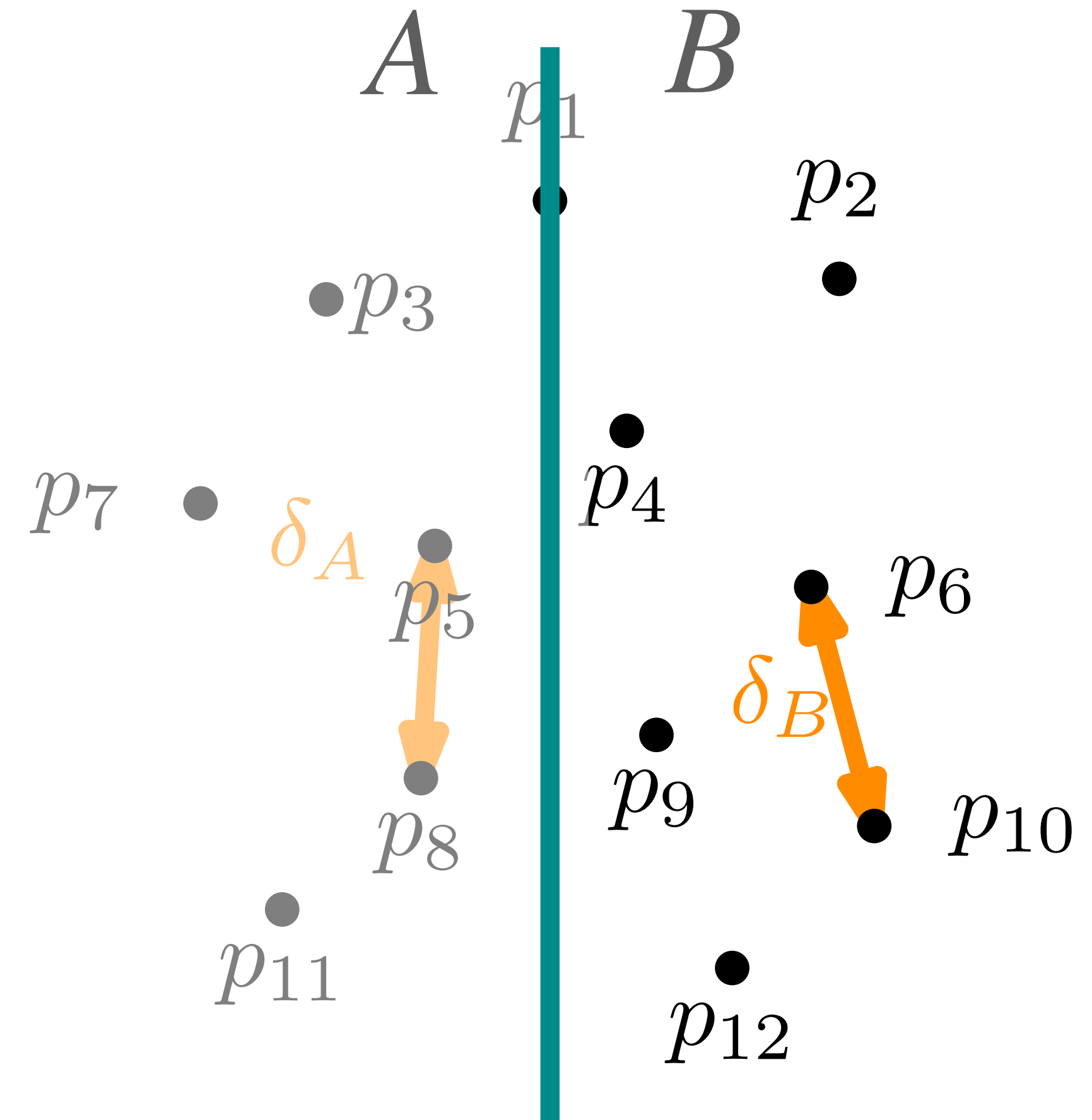
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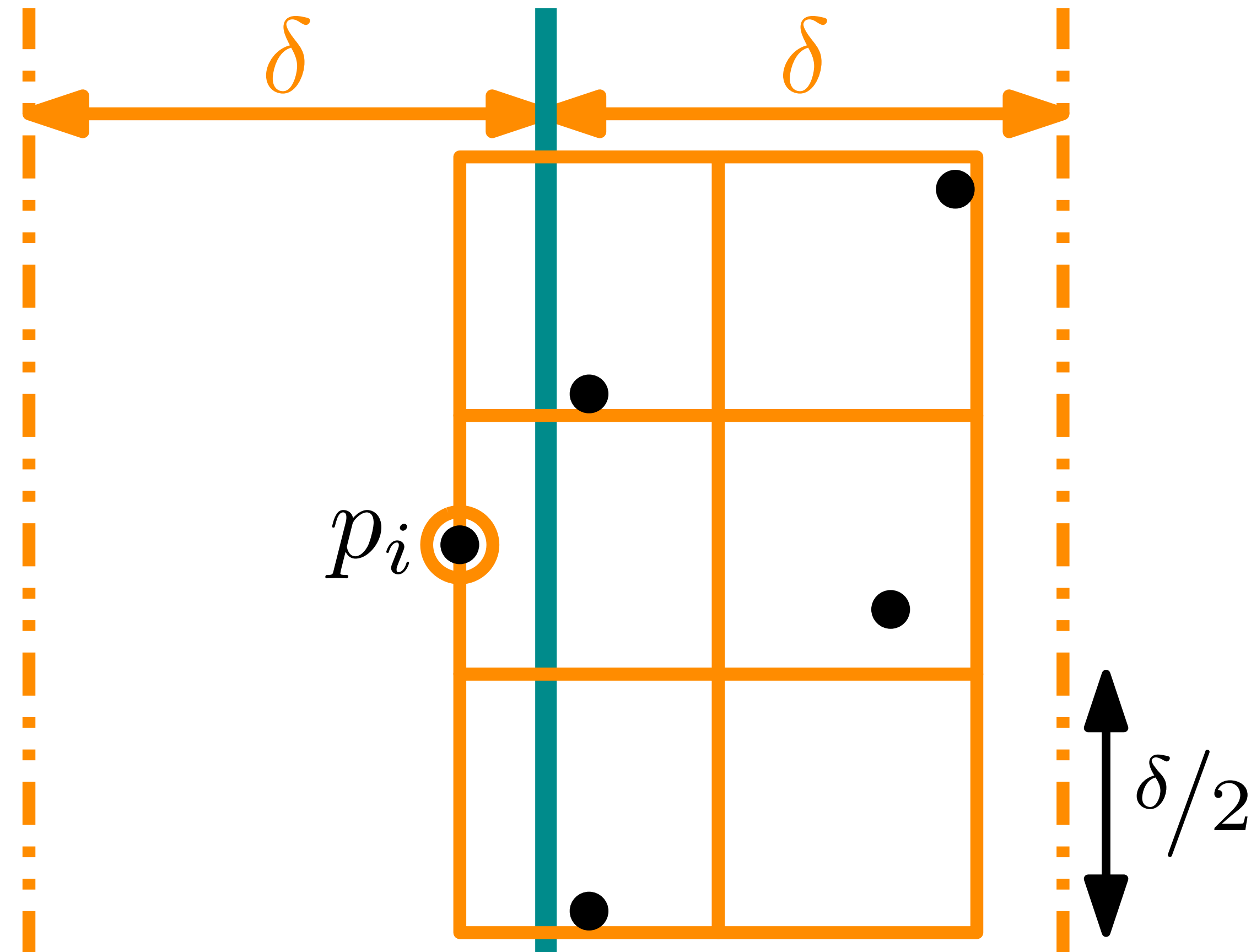
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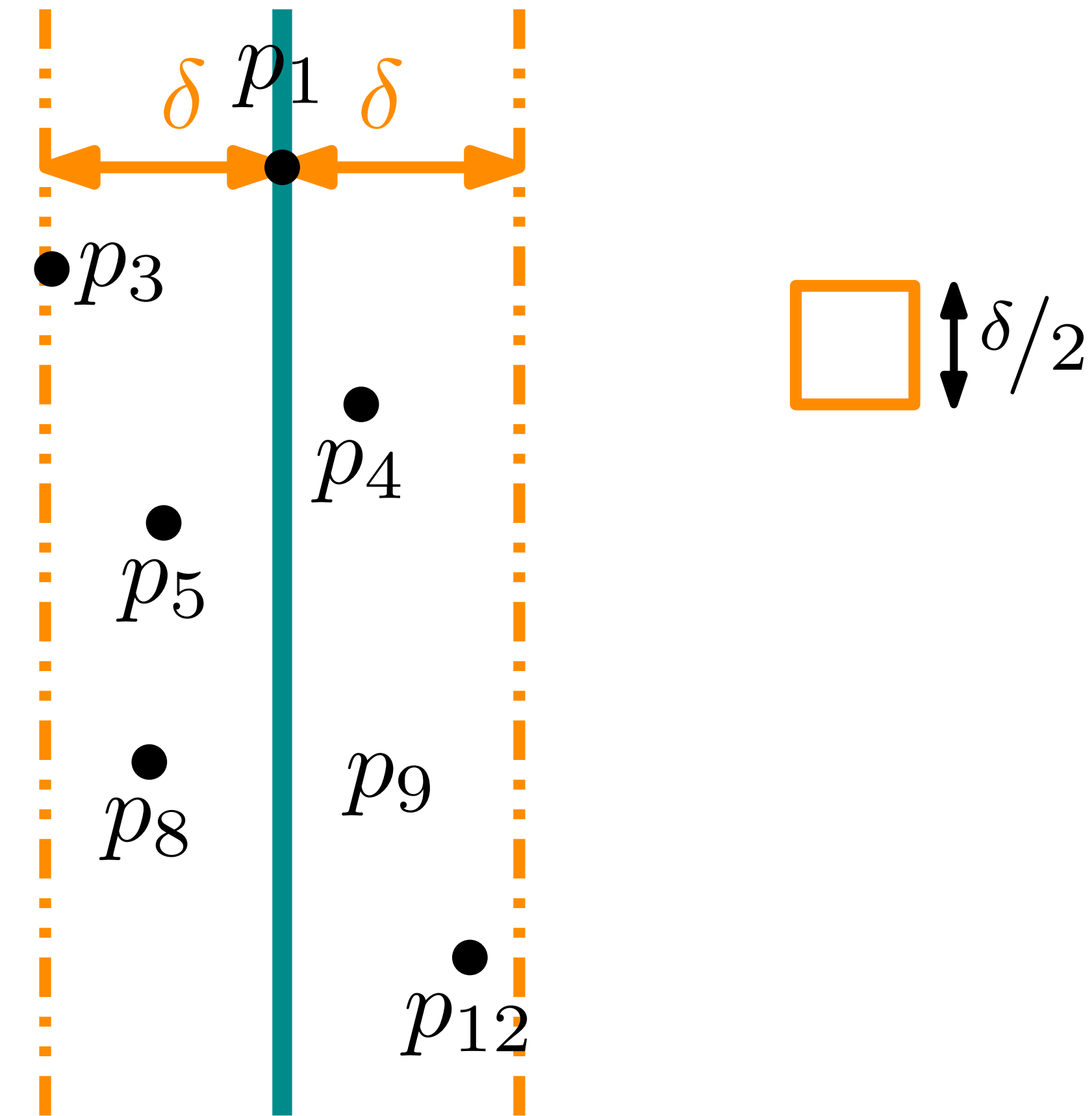
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4. ... merge by checking pairs in $A \times B$?
- Candidate pairs in $A \times B$ must have distance at most δ to the **median line**.
 - Point pairs in A (B) have distance $\geq \delta$.
 - Therefore: For each point in $p_i \in A$, there can be only constantly many points in B that are closer than δ !



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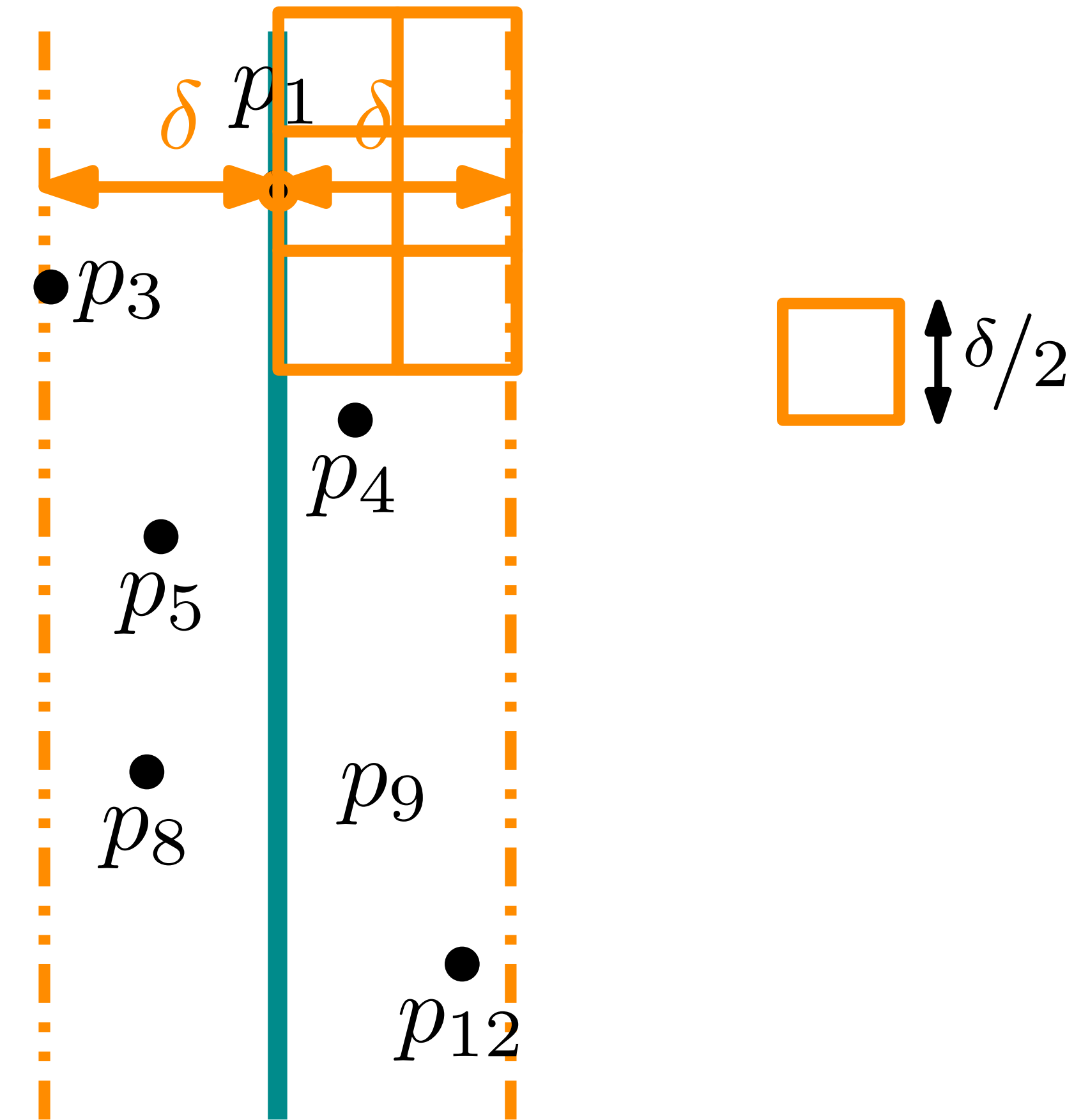
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 - Linear scan along one side, y -maximal to y -minimal coordinates.
 - Keep track of candidates on the other side of the median line.
 - This idea generalizes to higher dimensions!



Closest Pair — Divide and conquer

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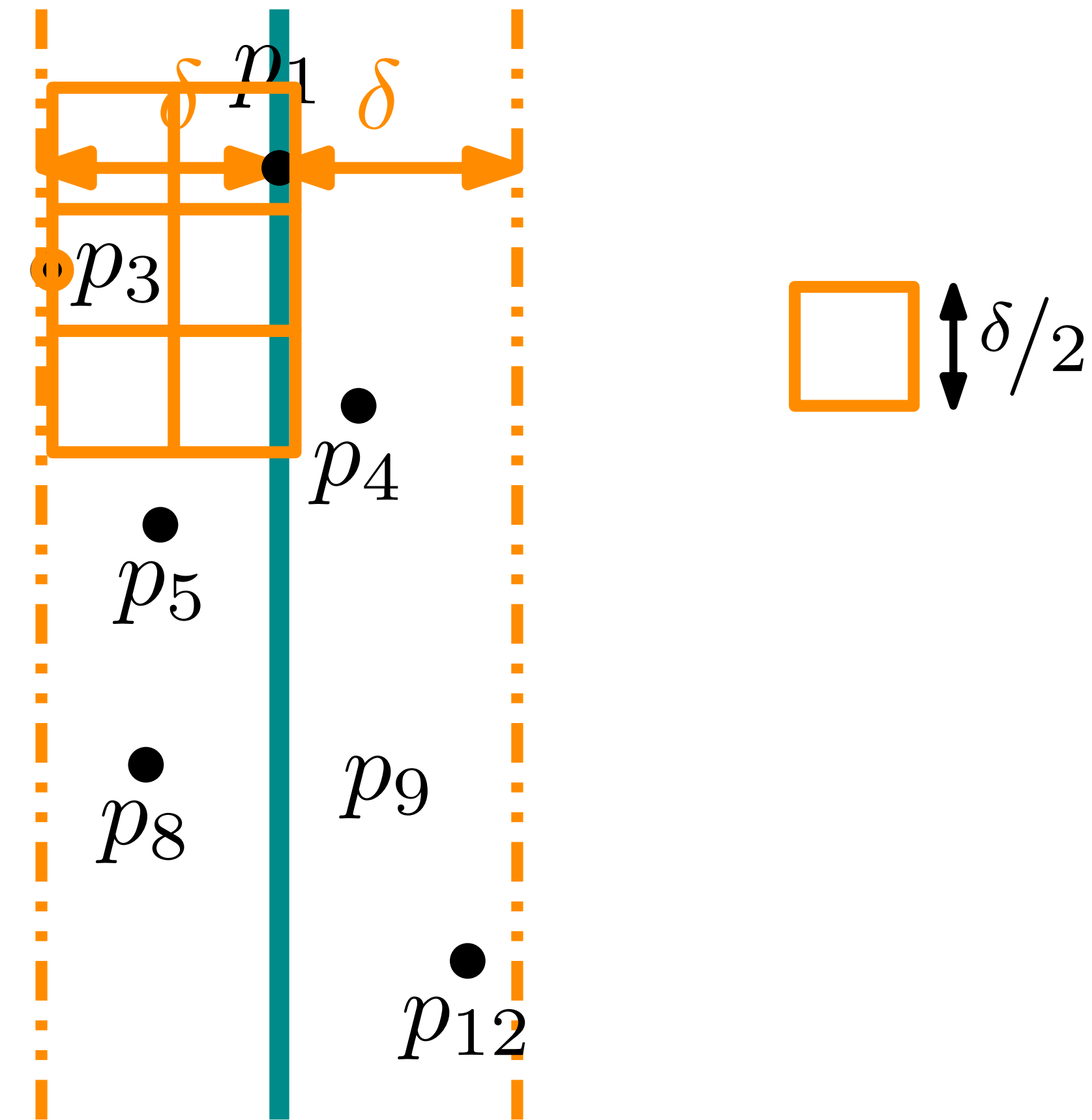
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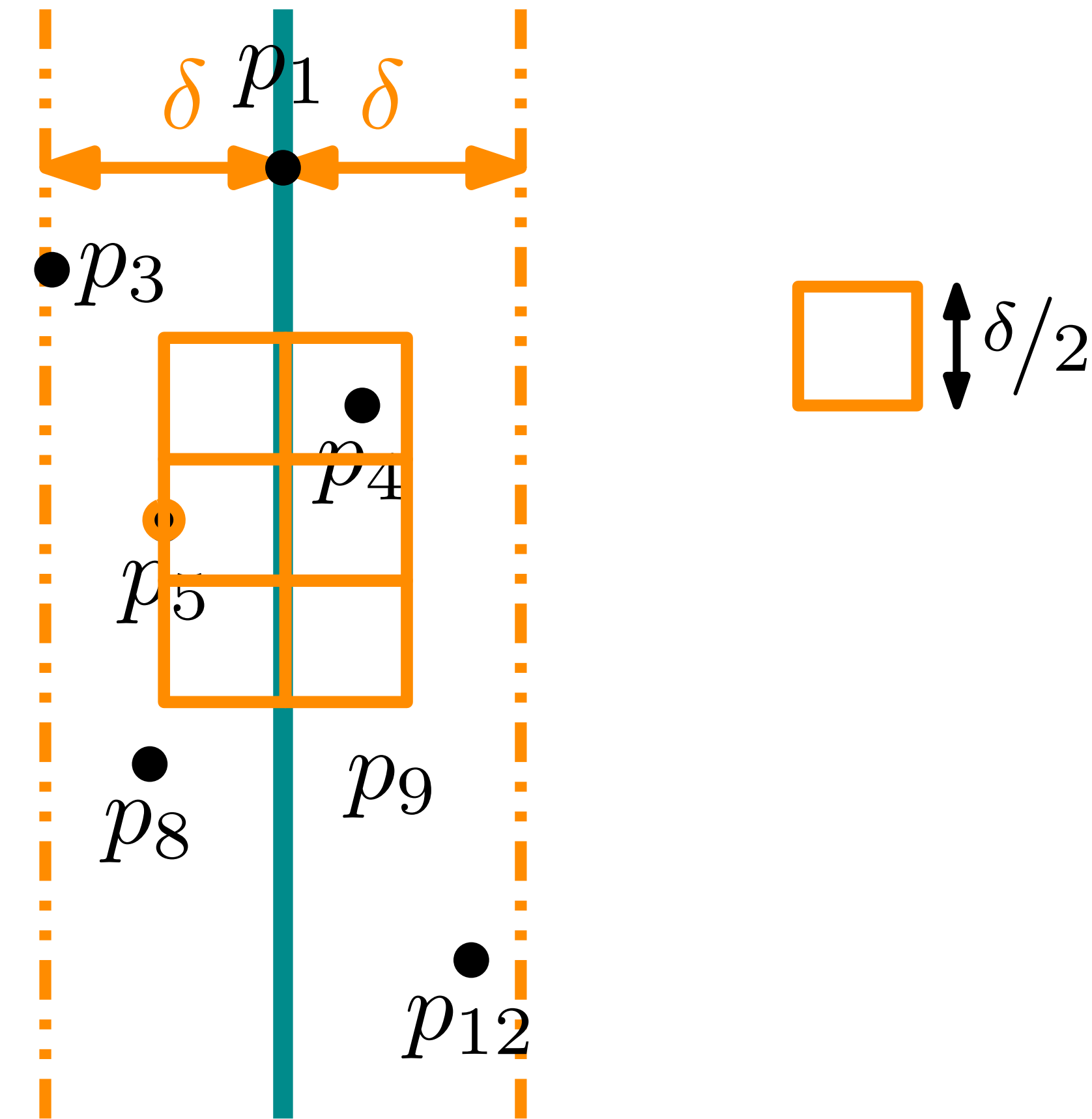
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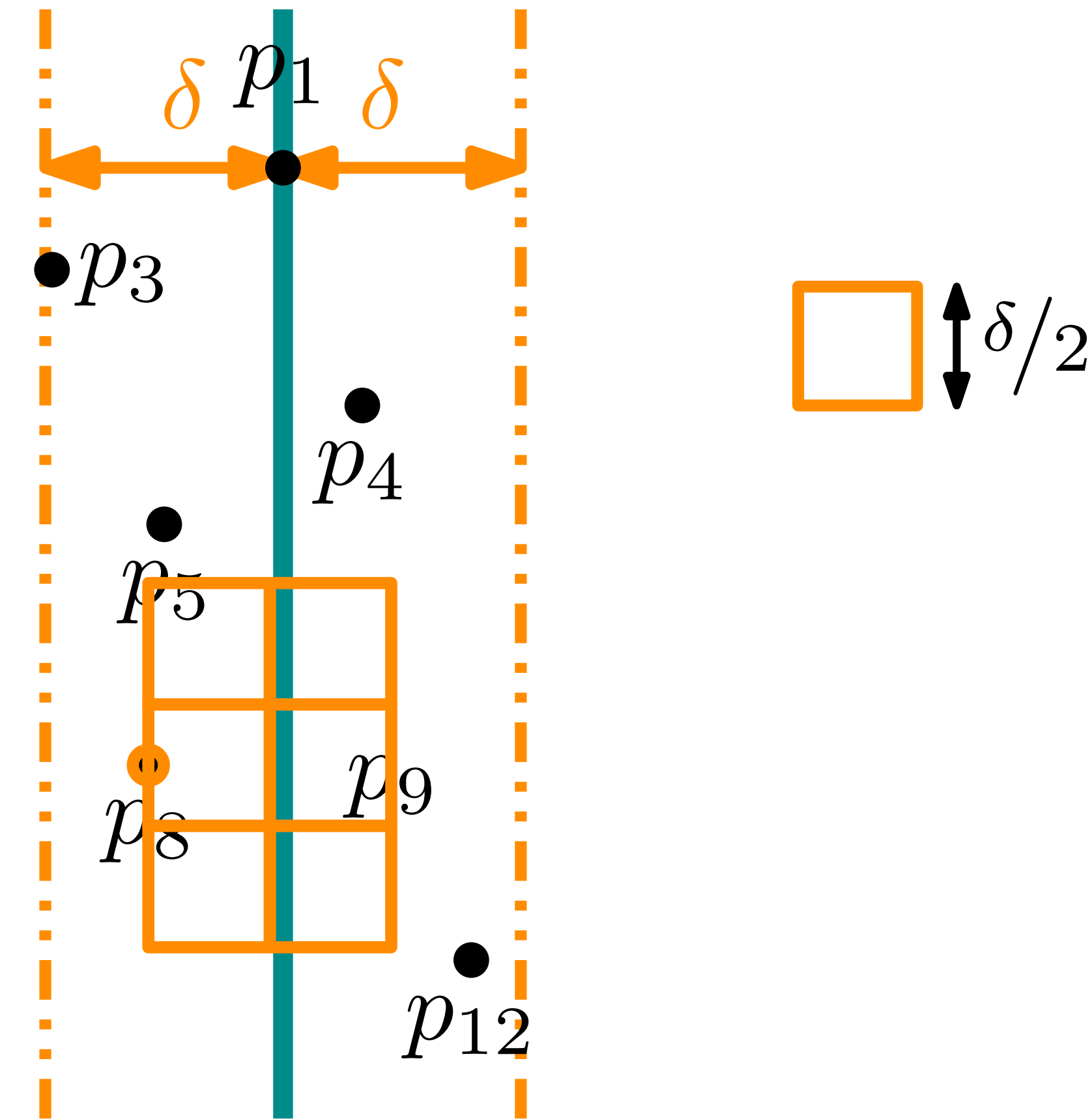
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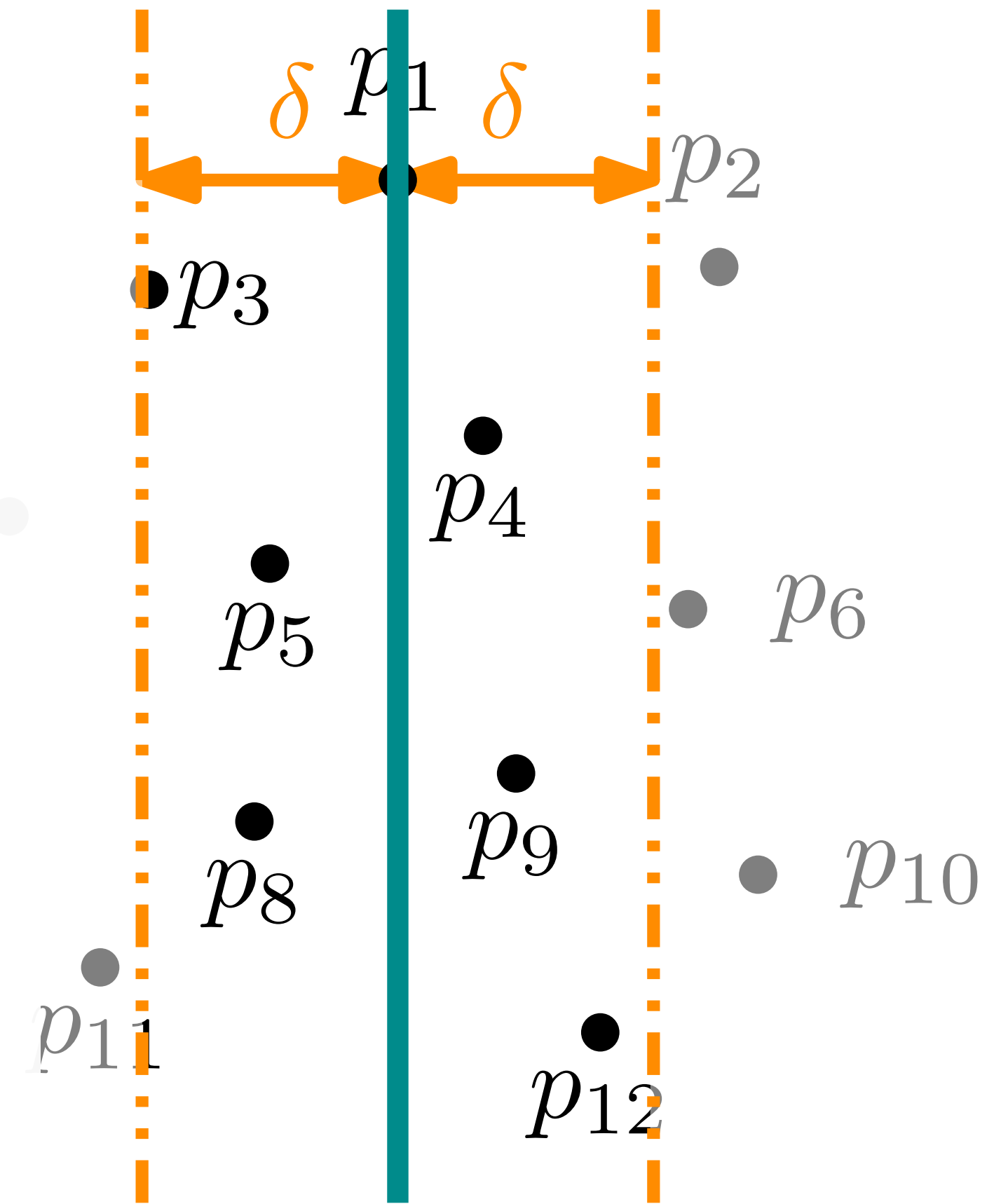


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2. Compute an x -median of \mathcal{P} , pivot. $\Theta(n)$
We get disjoint sets $A \cup B = \mathcal{P}$.
3. Compute δ_A and δ_B recursively. $2 \cdot T\left(\frac{n}{2}\right)$
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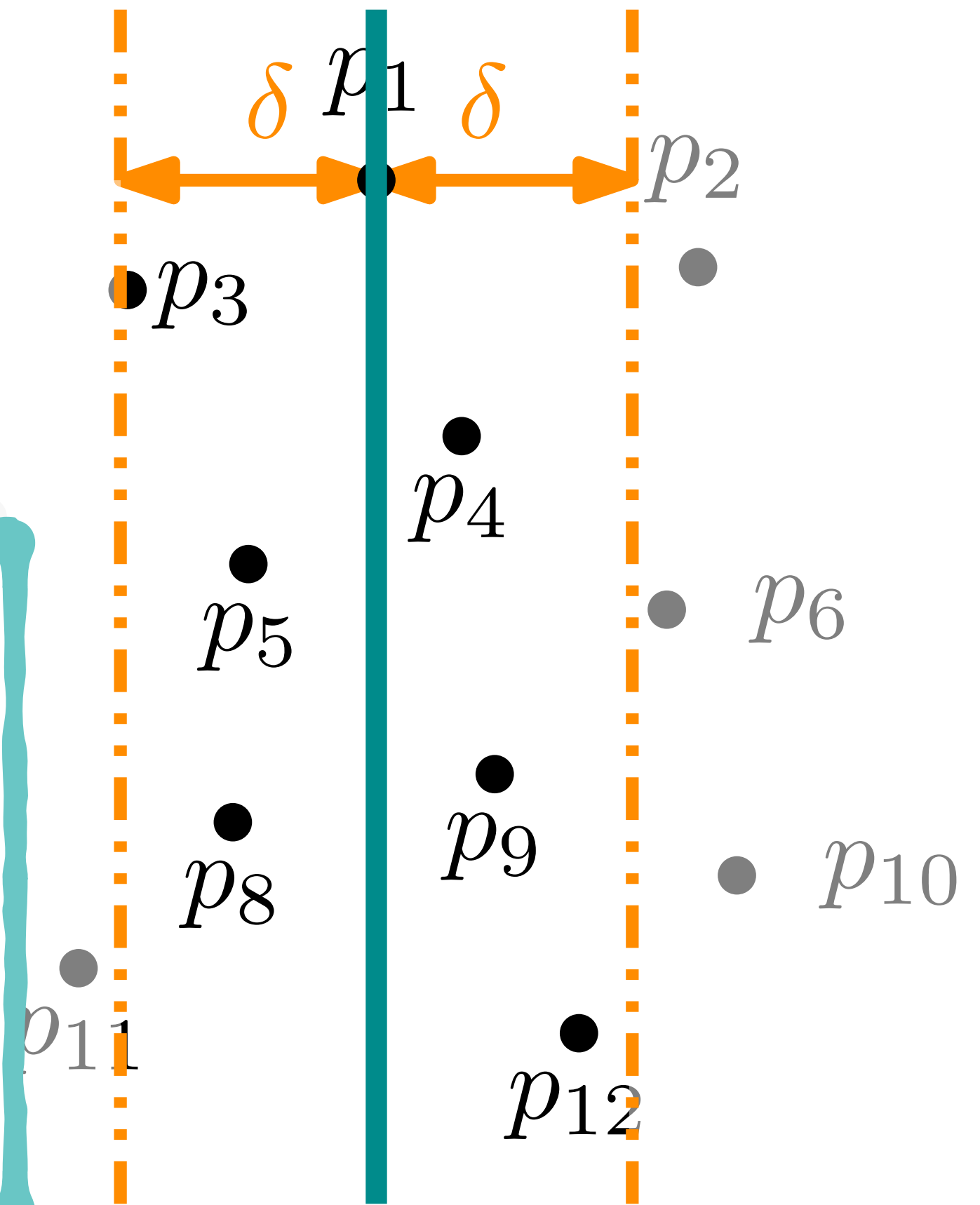
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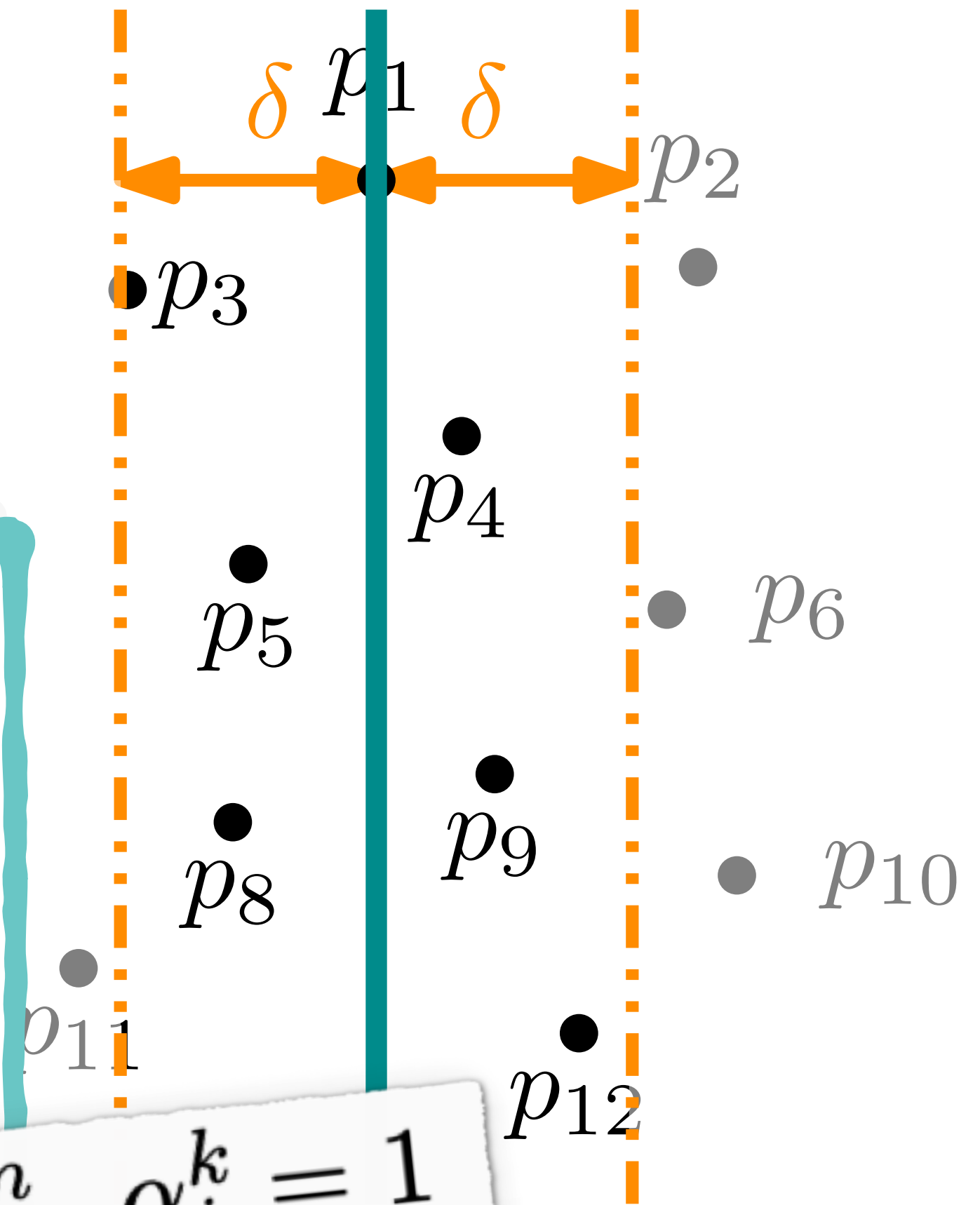
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$$\Theta(n^k \log(n)) \text{ for } \sum_{i=1}^m \alpha_i^k = 1$$

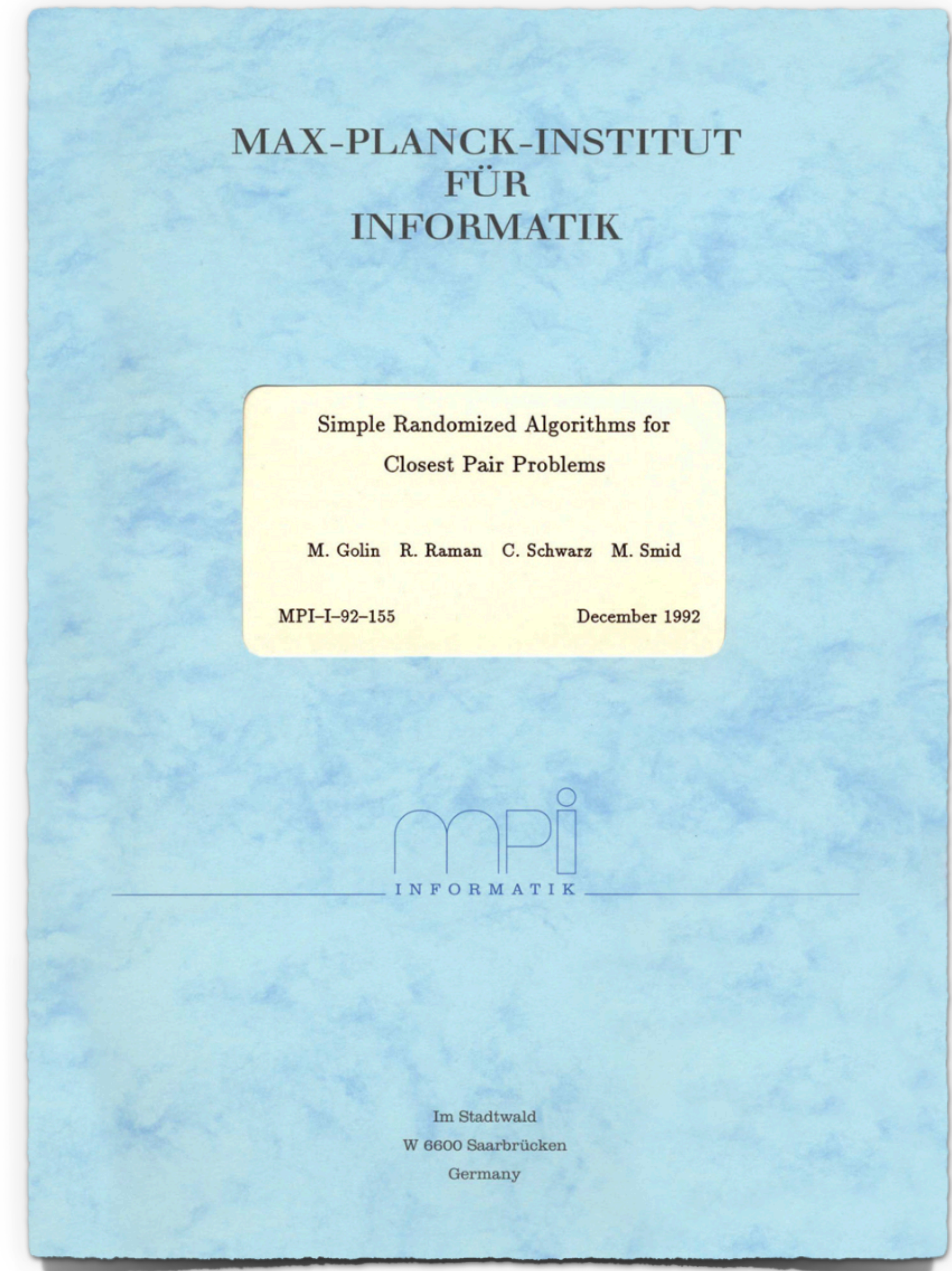
$$T(n)$$



Closest Pair — Randomised Incremental Construction (RIC)

Randomised Incremental Construction

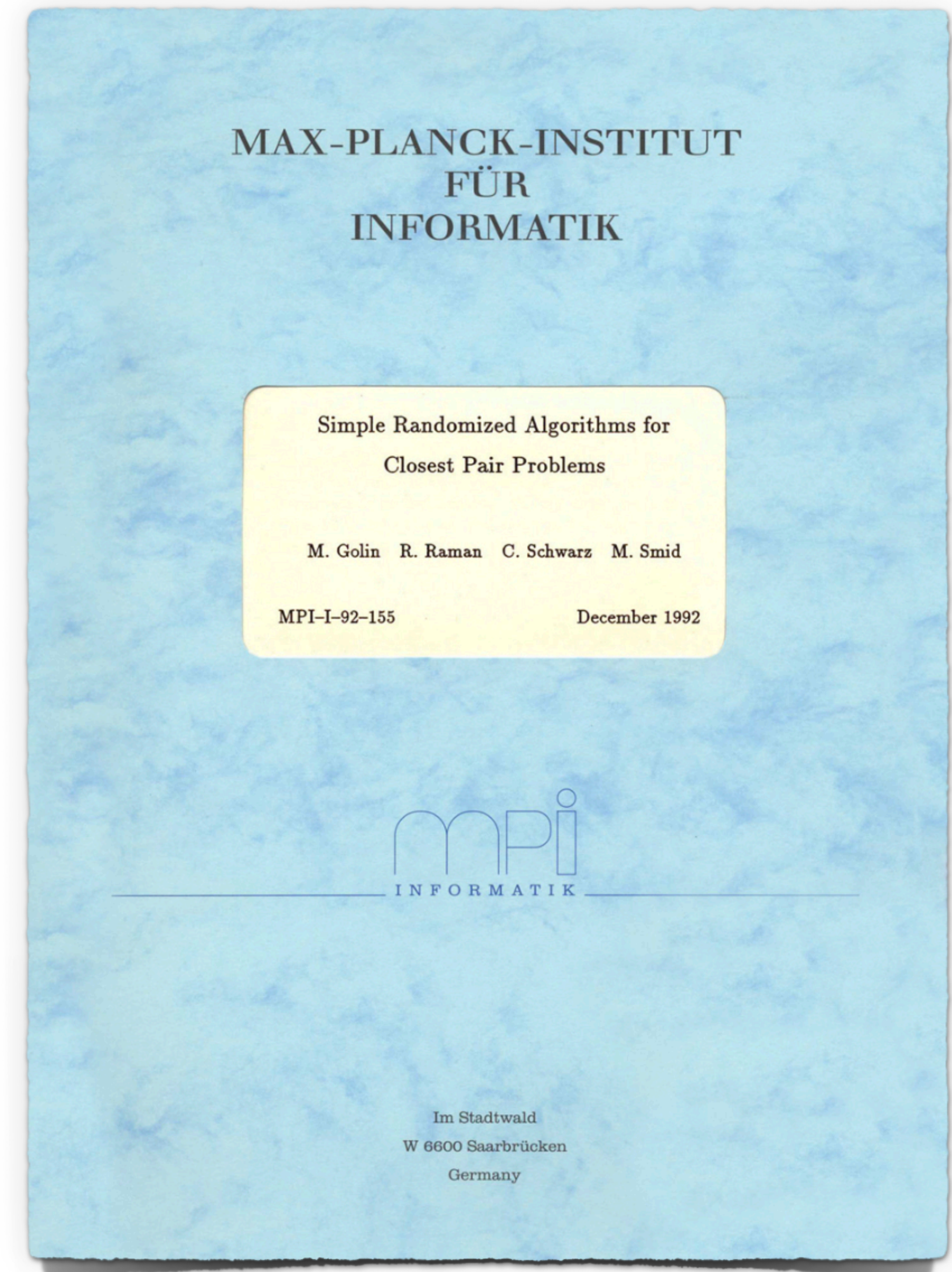
Golin, Raman, Schwarz, Smid 1992/1995



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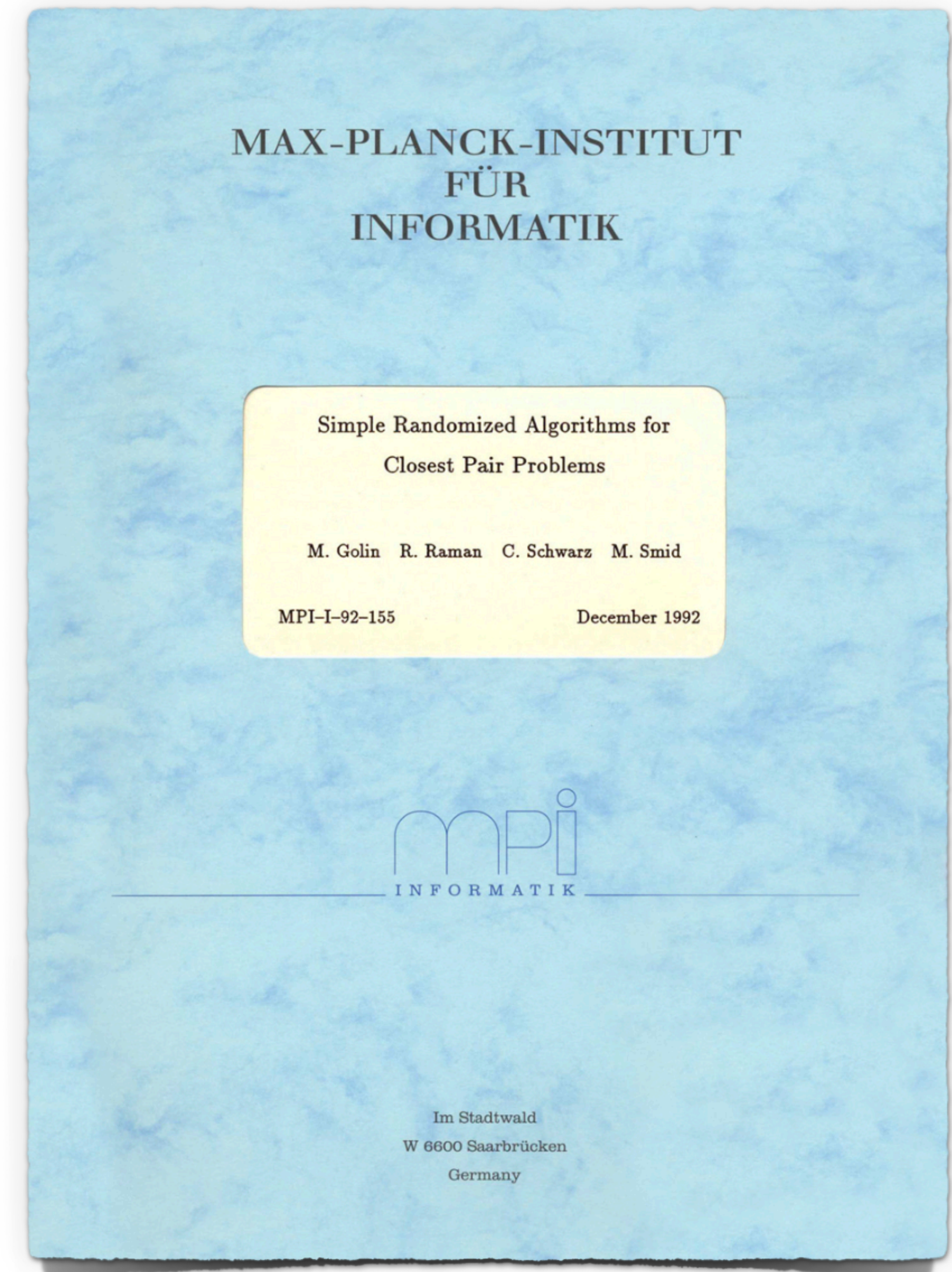
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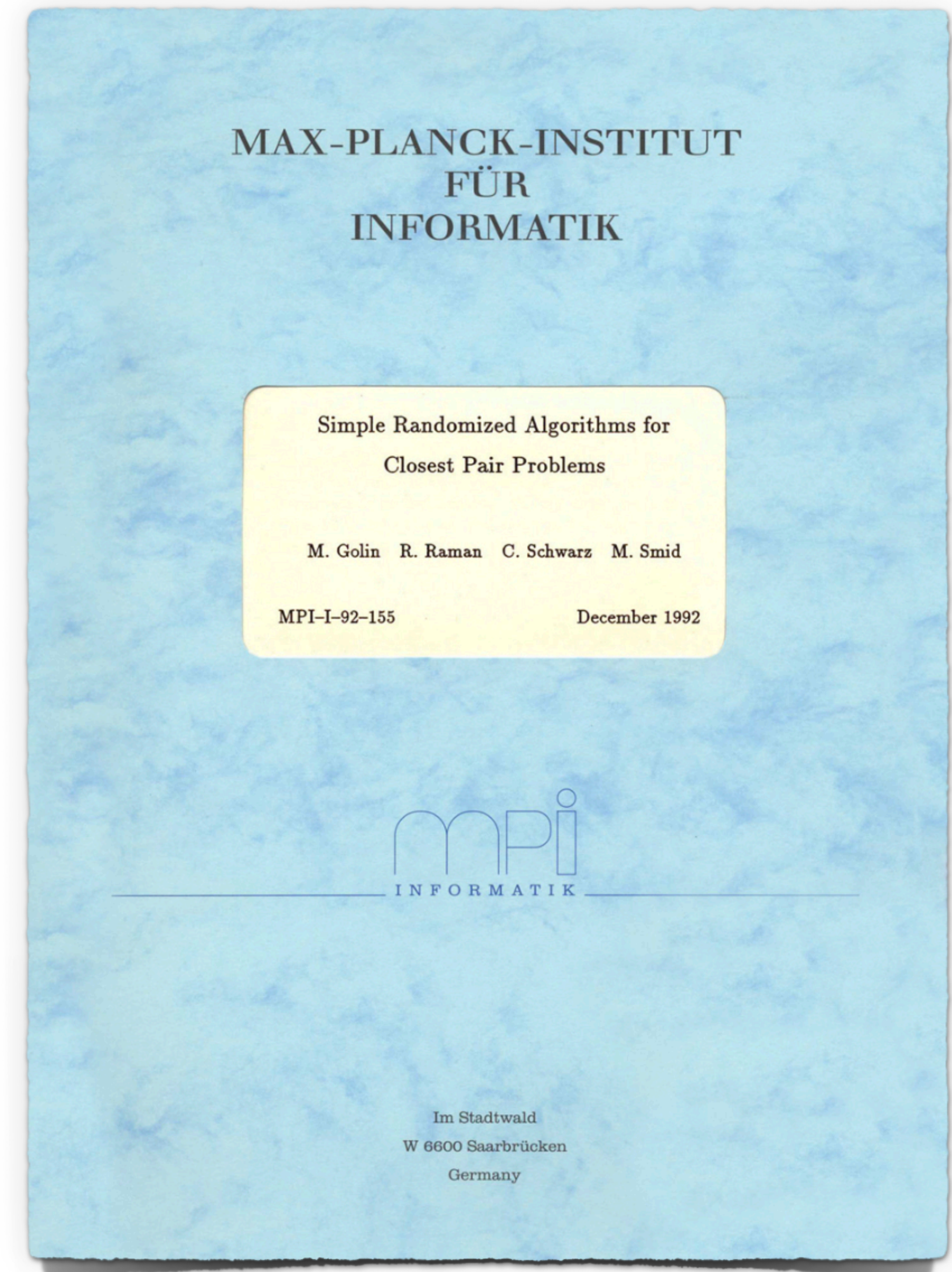
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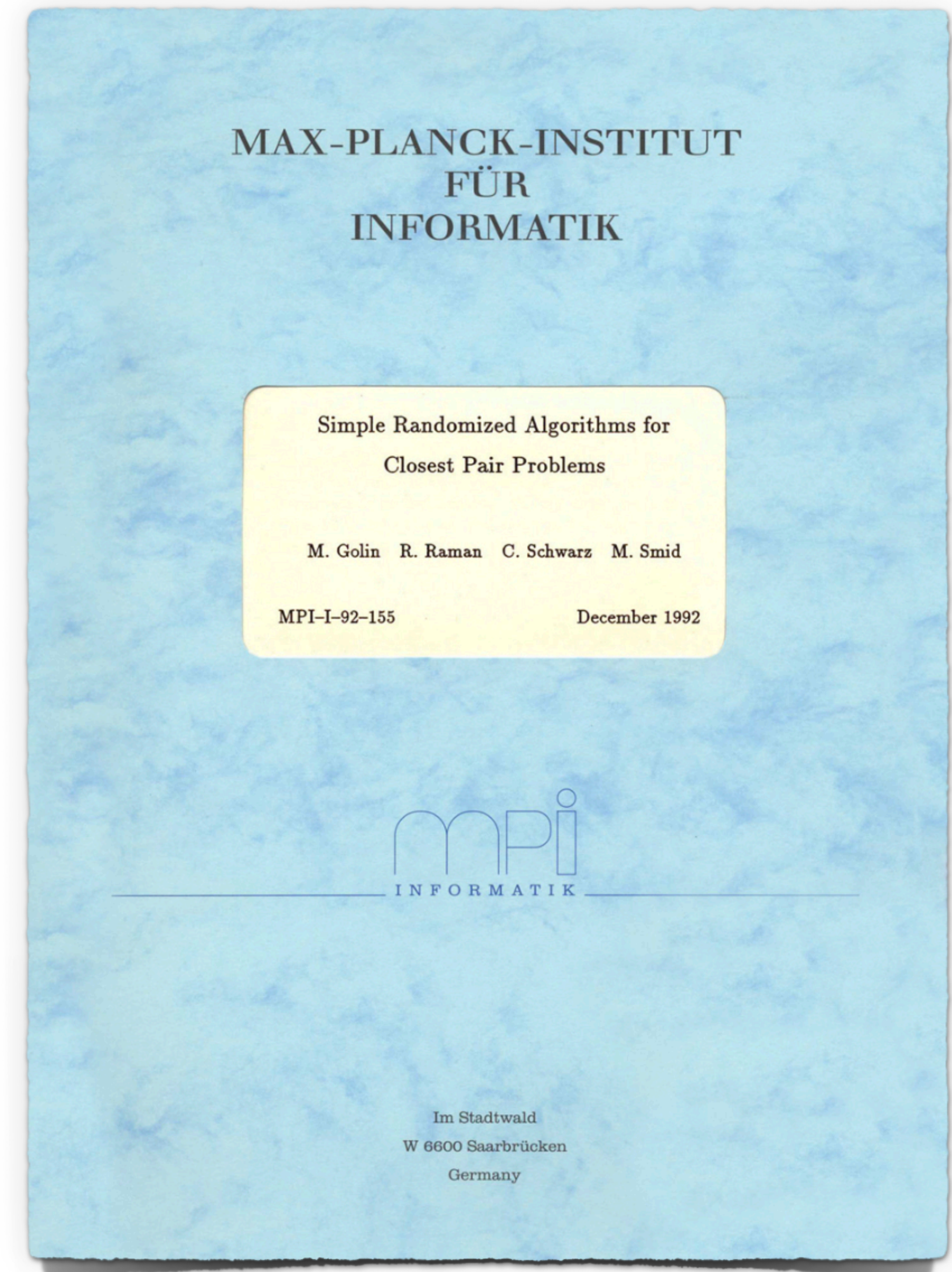
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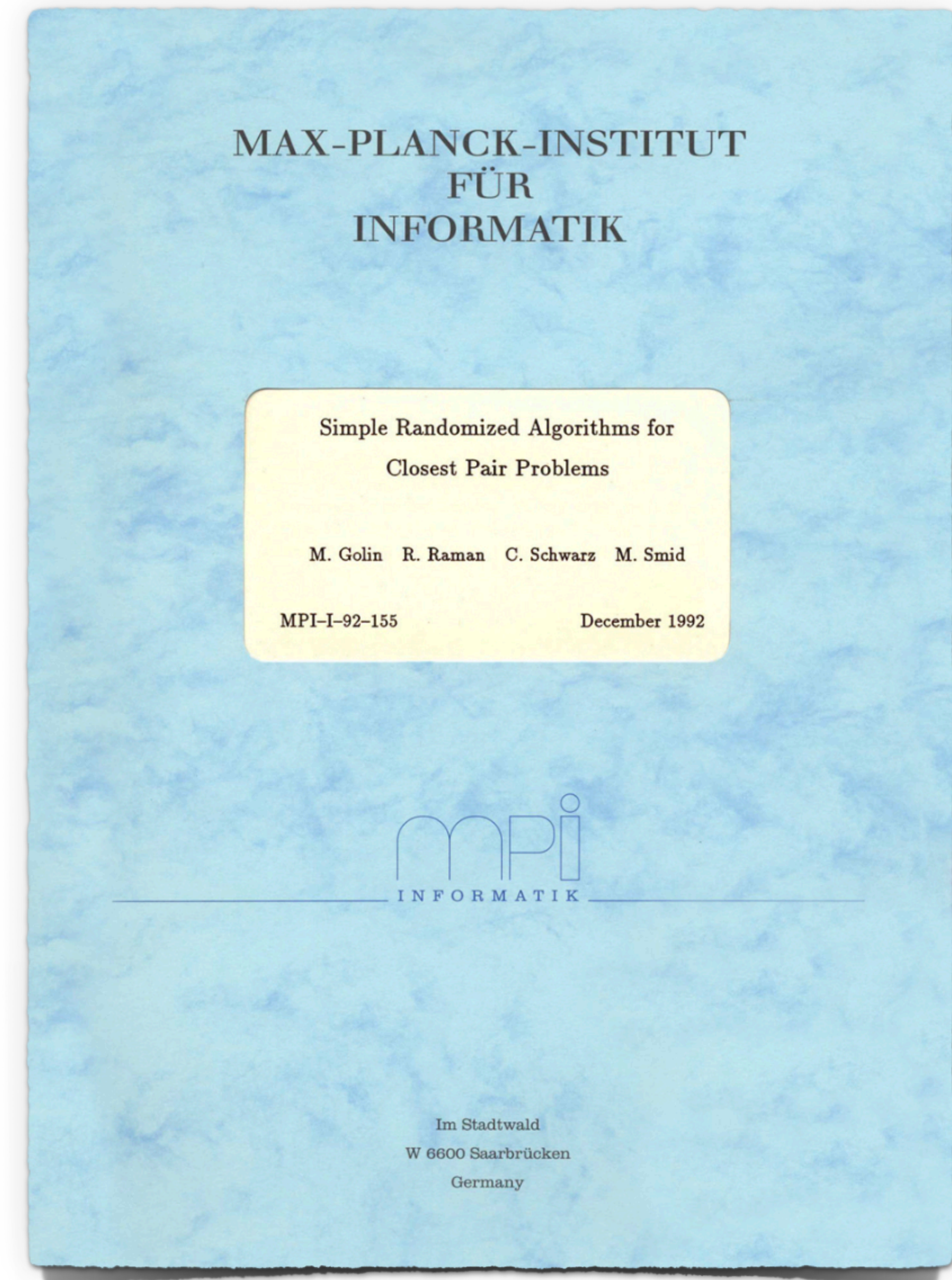
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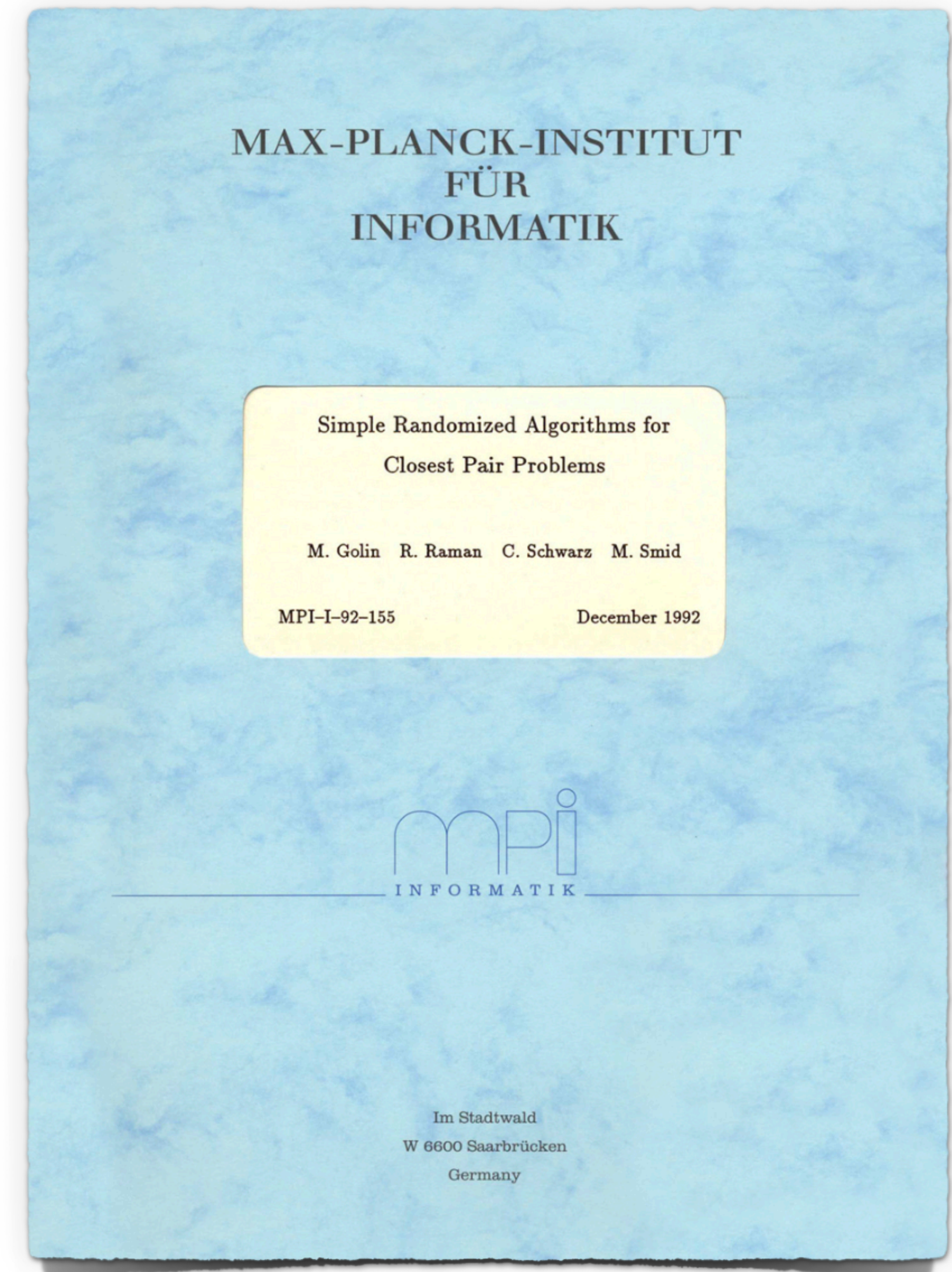
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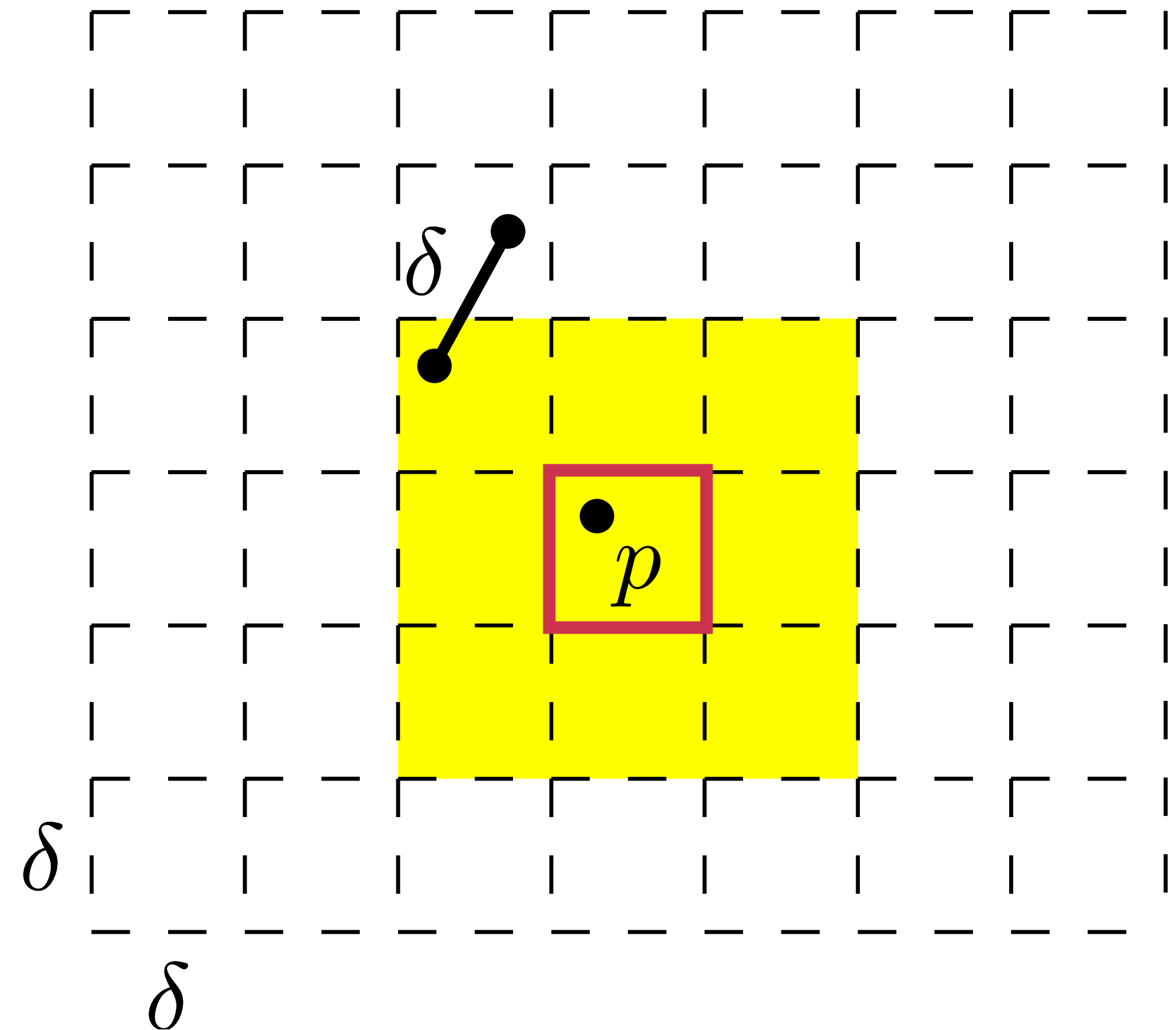
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- Unsorted sequences \Rightarrow No structure.



Randomised Incremental Construction

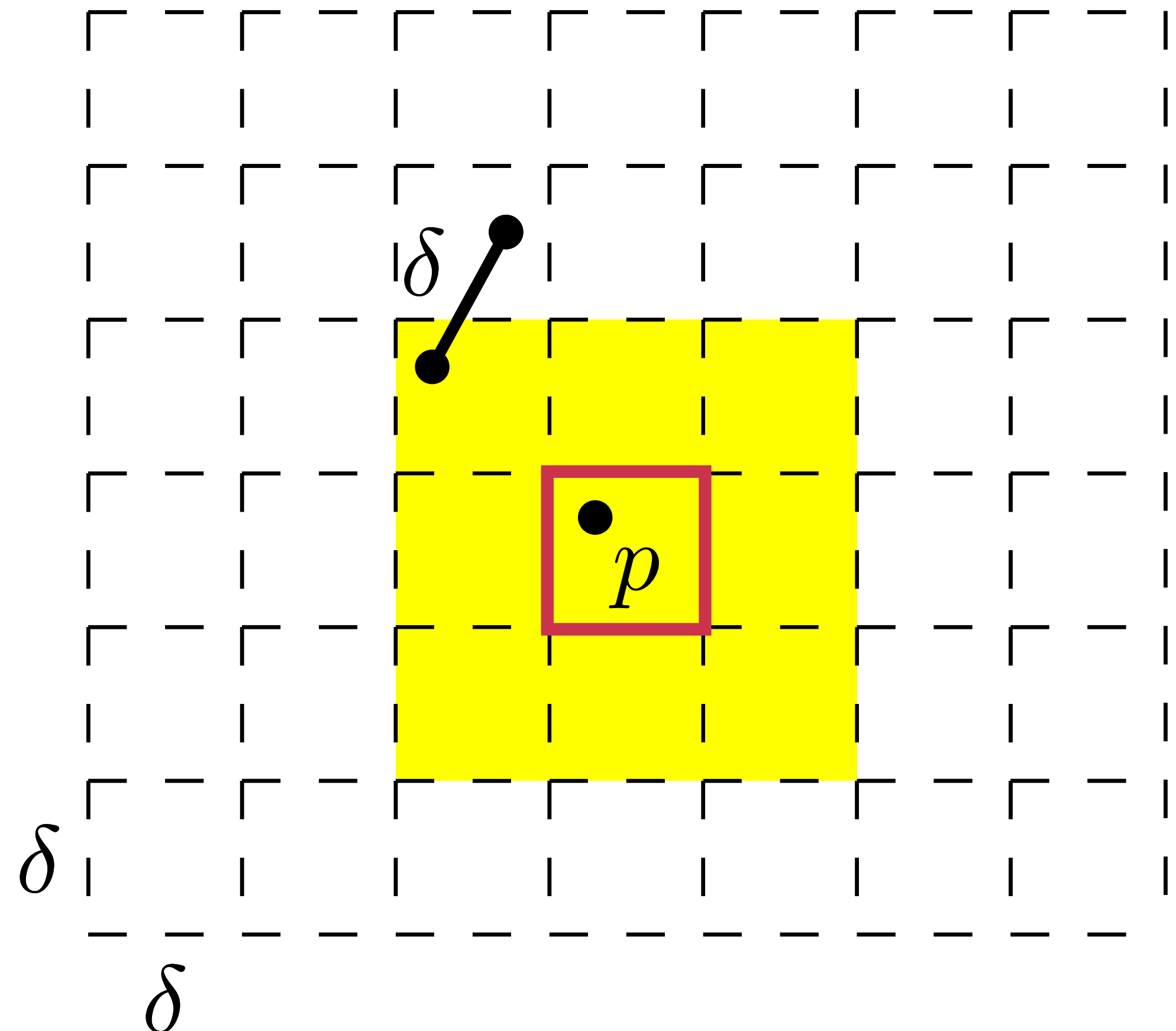
Golin, Raman, Schwarz, Smid 1992/1995



Randomised Incremental Construction

Golin, Raman, Schwarz, Smid 1992/1995

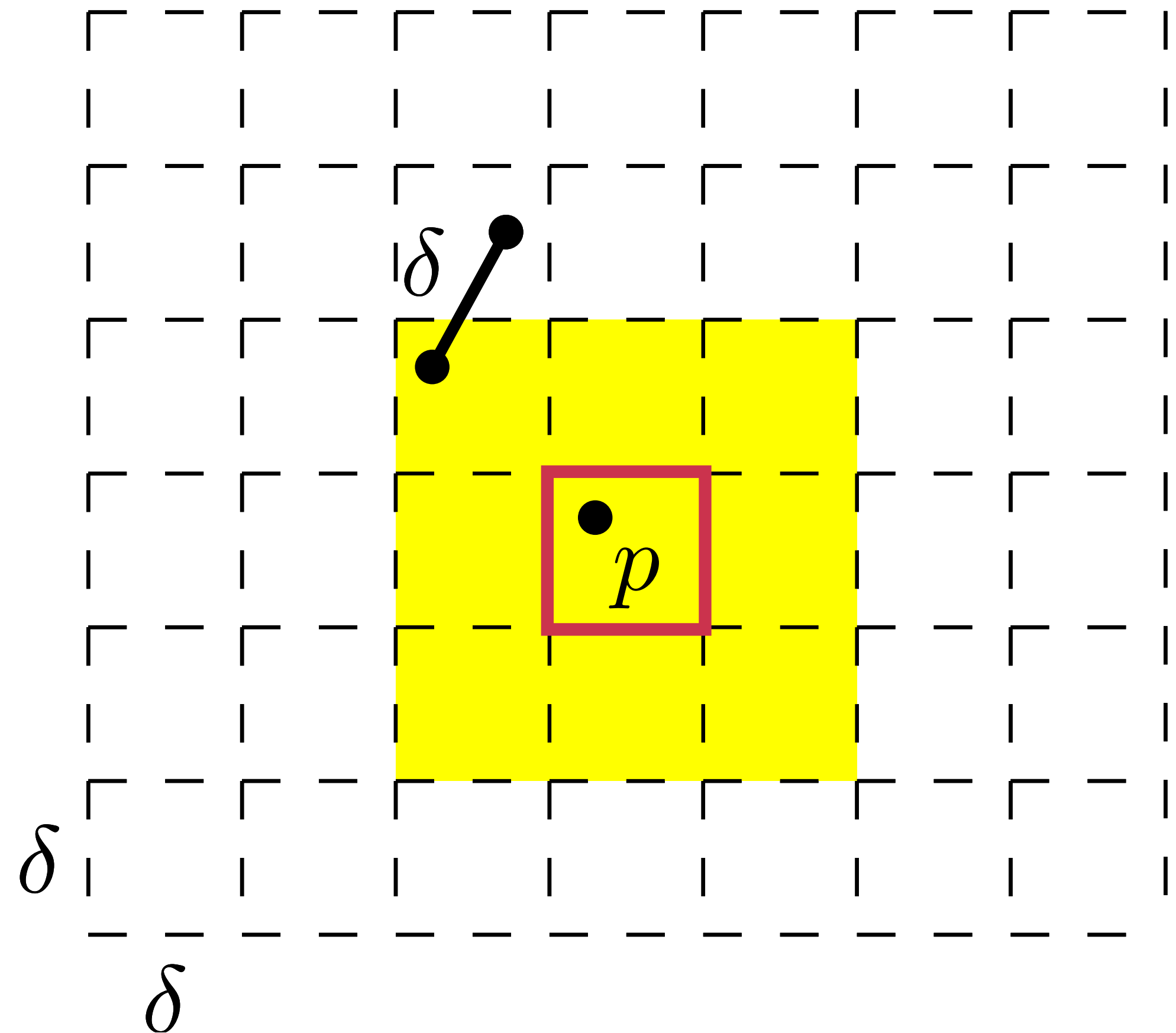
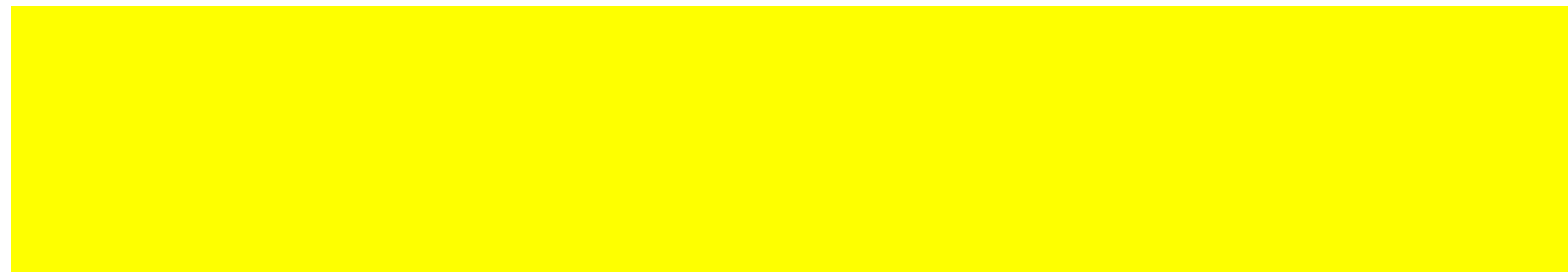
- Cover \mathbb{R}^2 with an infinite $\delta \times \delta$ grid G_δ .
- Cells are identified by column x , row y : $G_\delta[x, y]$.
- W.l.o.g., the cell $G_\delta[0,0]$ is located at $(0,0)$.



Randomised Incremental Construction

Golin, Raman, Schwarz, Smid 1992/1995

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- Each (half-open) grid cell $G_\delta[x, y]$ stores all previously considered points of \mathcal{P} inside it.

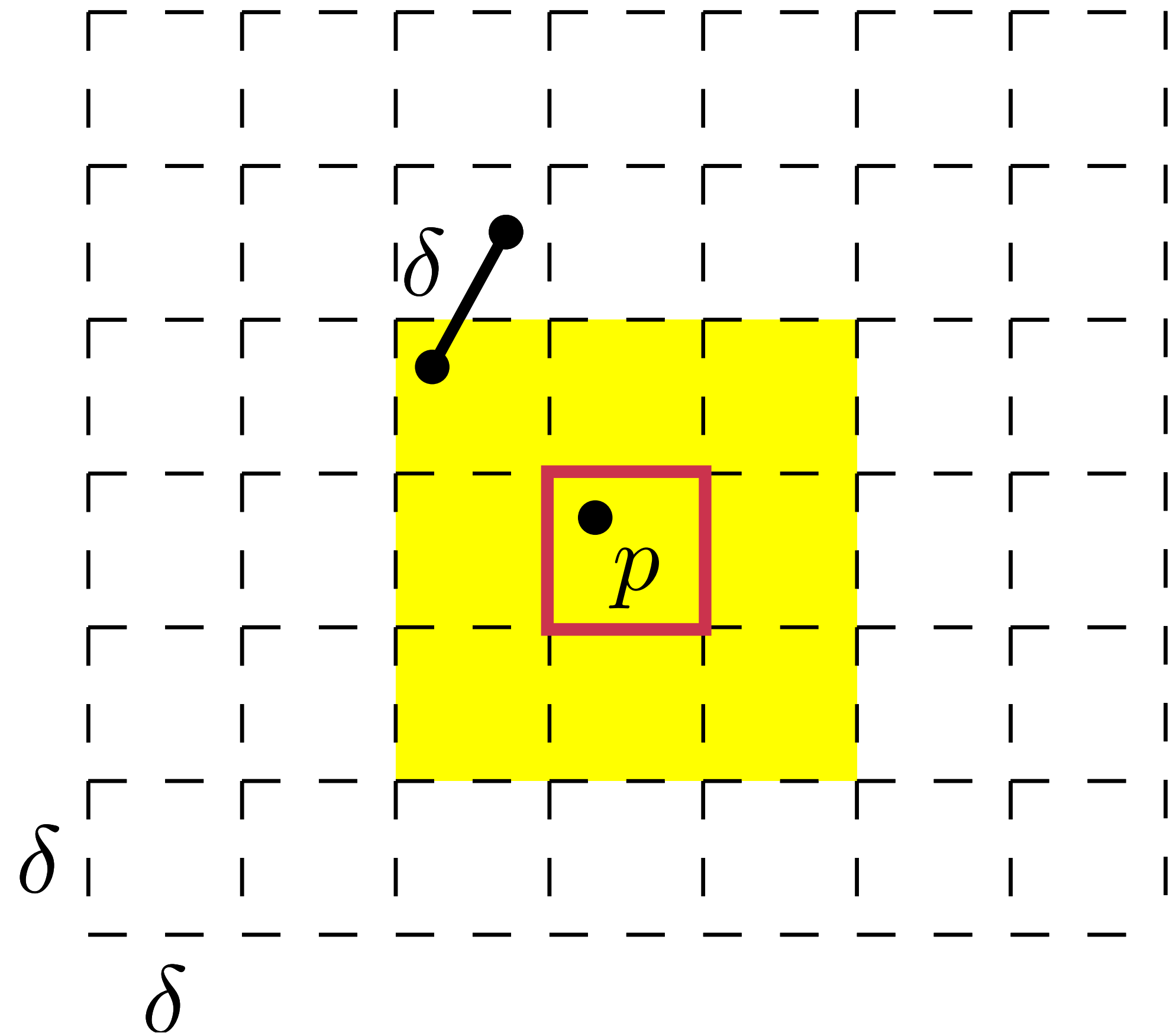


Randomised Incremental Construction

Golin, Raman, Schwarz, Smid 1992/1995

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- The neighbourhood of $G_\delta[x, y]$ refers to:

$$N_\delta(x, y) = \bigcup_{u, v \in [-1, 1]} G_\delta[x - u, y - v].$$



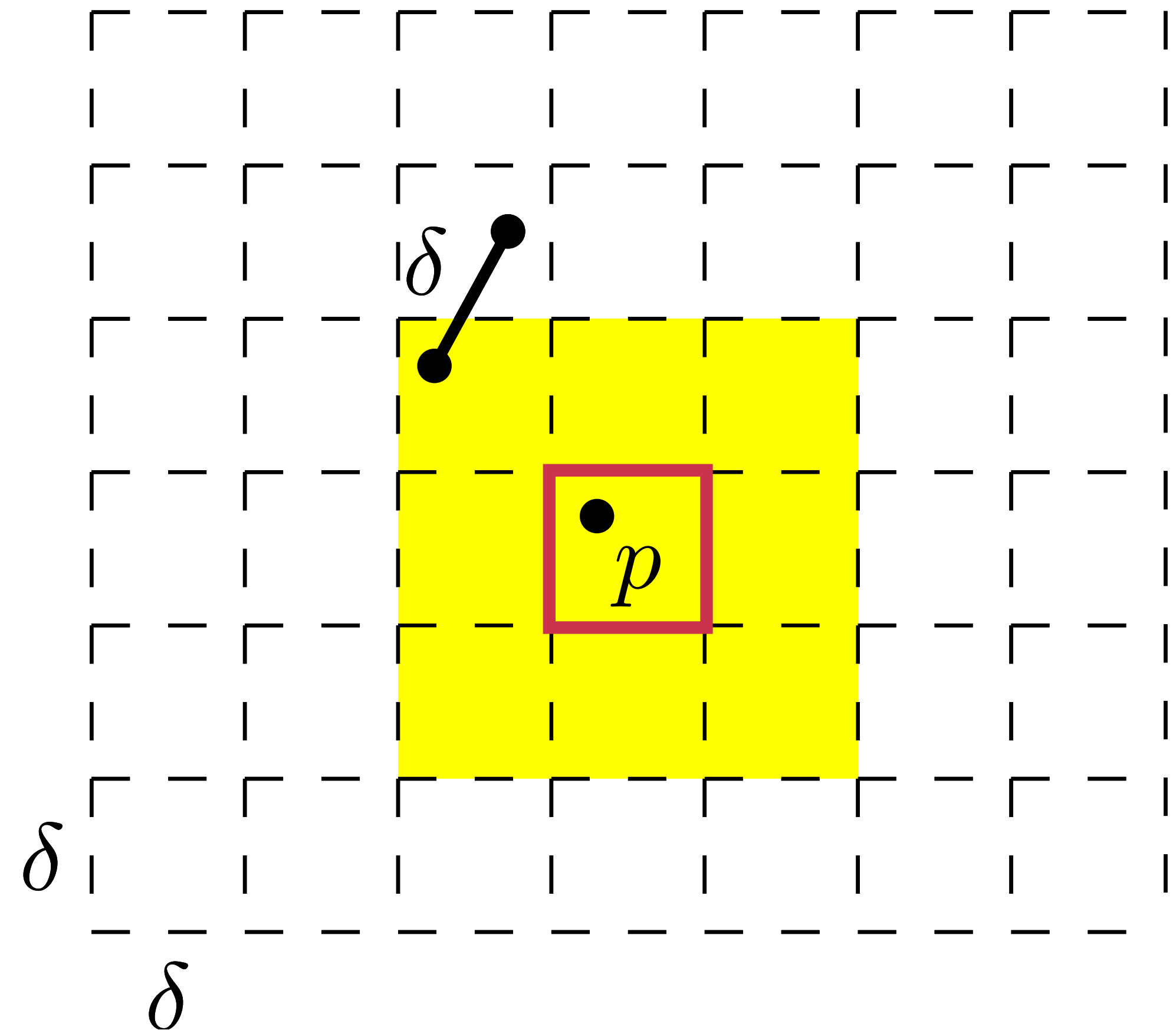
Randomised Incremental Construction

Golin, Raman, Schwarz, Smid 1992/1995

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Lemma. For $p, q \in \mathcal{P}$ such that $p \in G_\delta[x, y]$:

$$d(p, q) \leq \delta \implies q \in N_\delta(x, y)$$


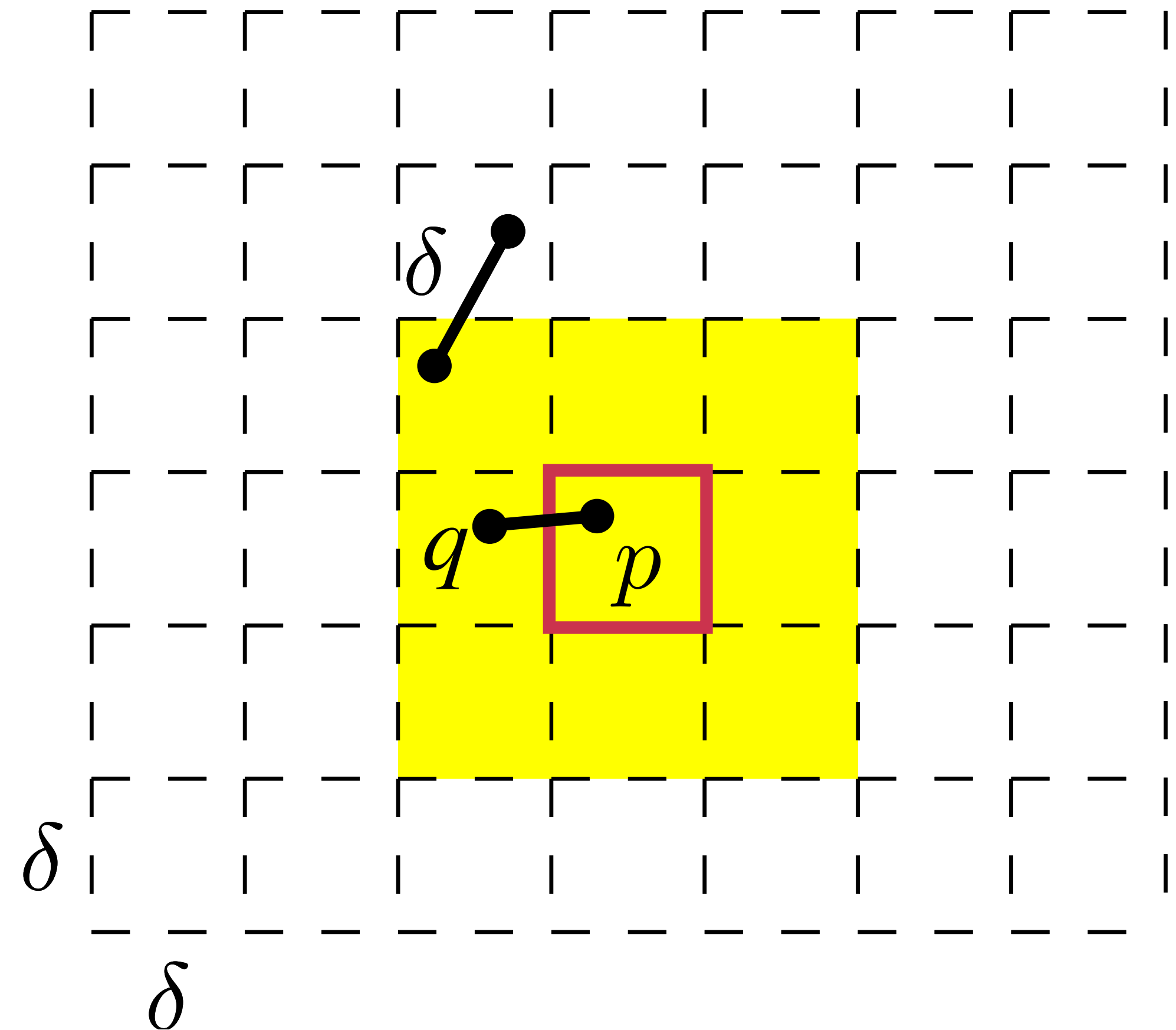
Randomised Incremental Construction

Golin, Raman, Schwarz, Smid 1992/1995

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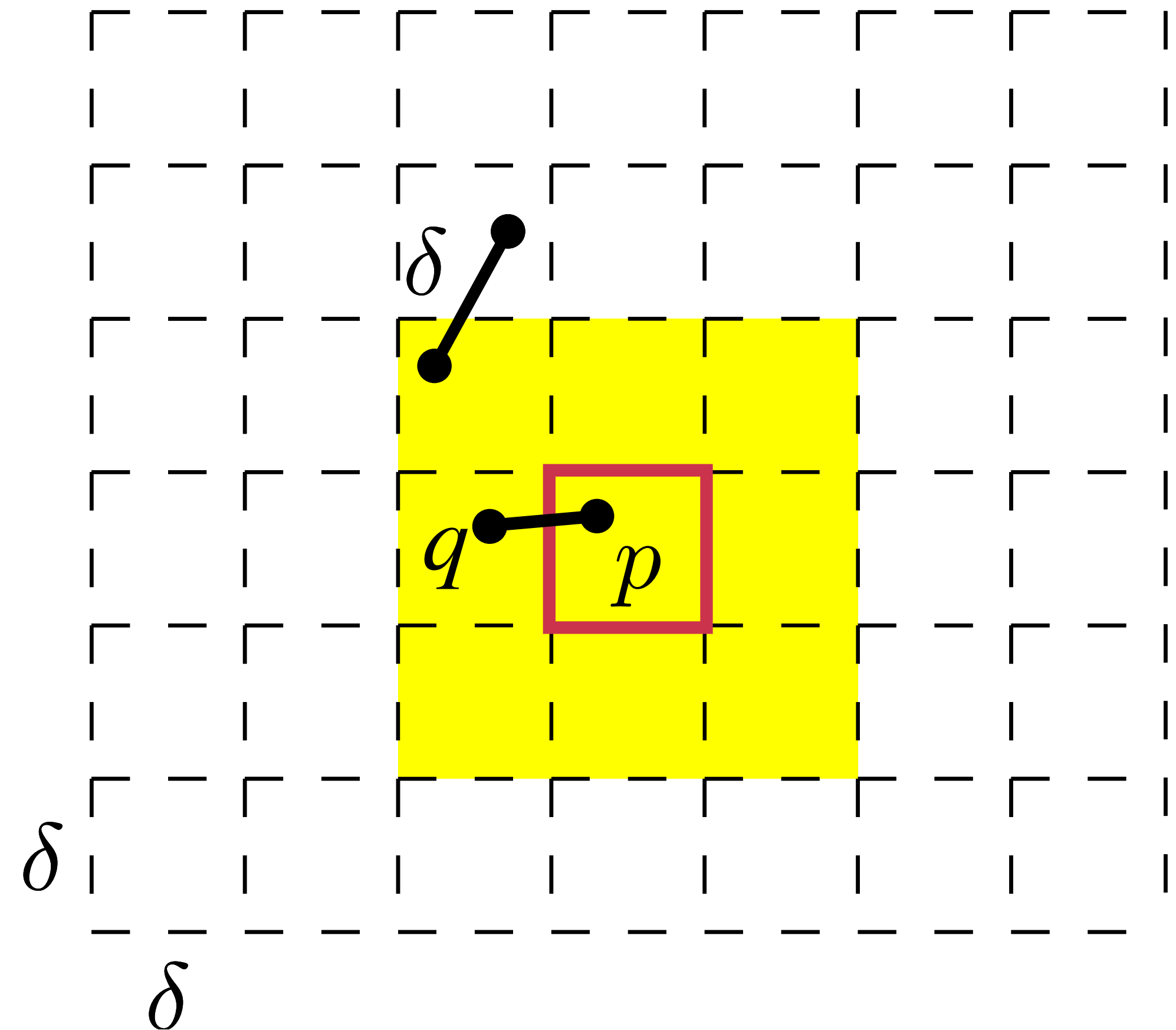
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Randomised Incremental Construction

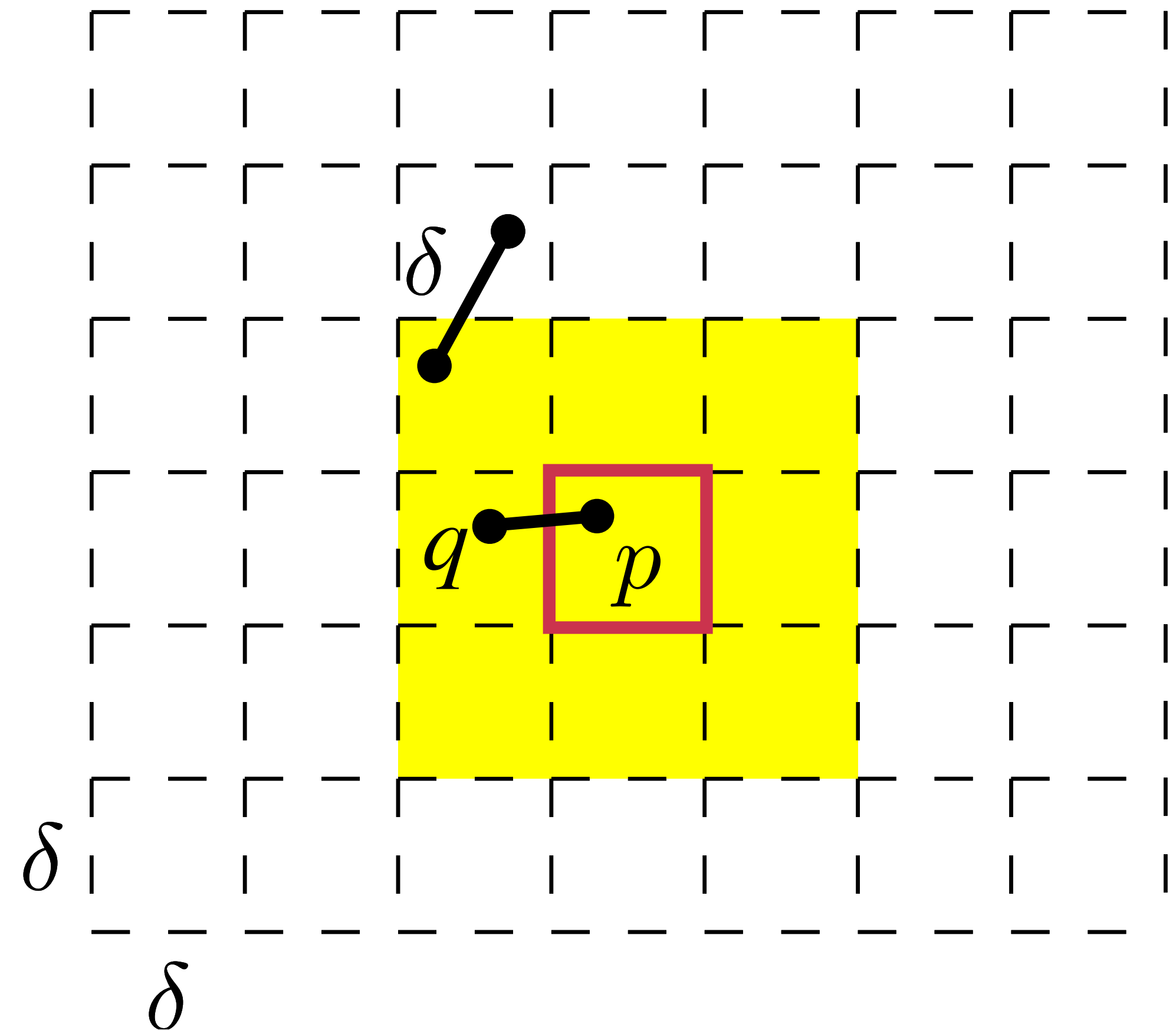
Golin, Raman, Schwarz, Smid 1992/1995



Randomised Incremental Construction

Golin, Raman, Schwarz, Smid 1992/1995

- **Idea:** Sparsity! Keep $|N_\delta(x, y)| \in \mathcal{O}(1)$.
- If each $N_\delta(x, y)$ contains constantly many points of pairwise distance at least δ , ...
- ... then $\mathcal{O}(1)$ comparisons for next $p \in \mathcal{P}$.



Randomised Incremental Construction

Golin, Raman, Schwarz, Smid 1992/1995

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- ... then $\mathcal{O}(1)$ comparisons for next $p \in \mathcal{P}$.

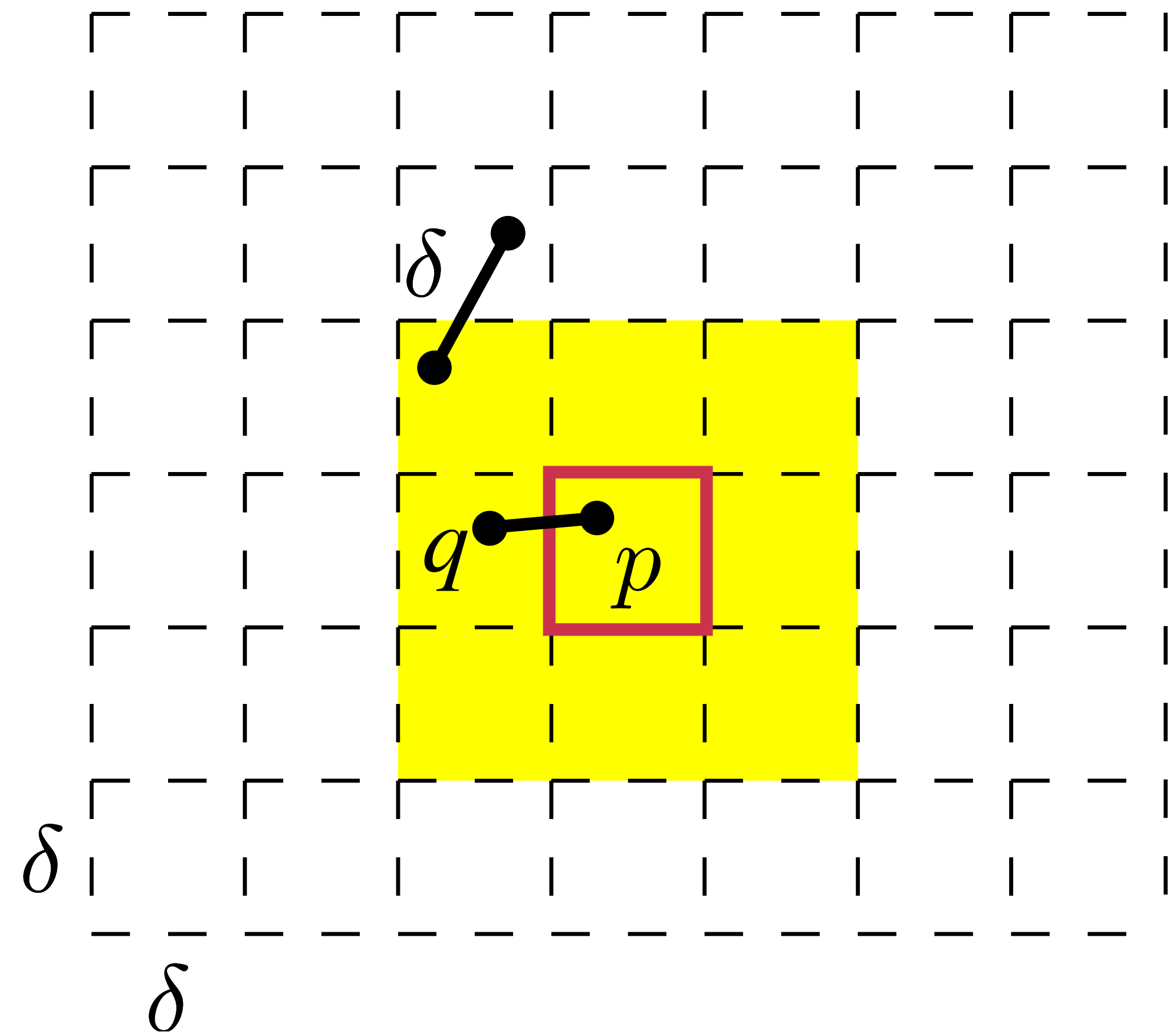
For each $p \in \mathcal{P}$:

Find x, y such that $G_\delta[x, y]$ contains p .

Insert p to $G_\delta[x, y]$.

If p is part of new closest pair $\{p, q\}$:
Update $\delta = d(p, q)$.

Rebuild sparse grid, $G_{\delta'}[x, y]$.



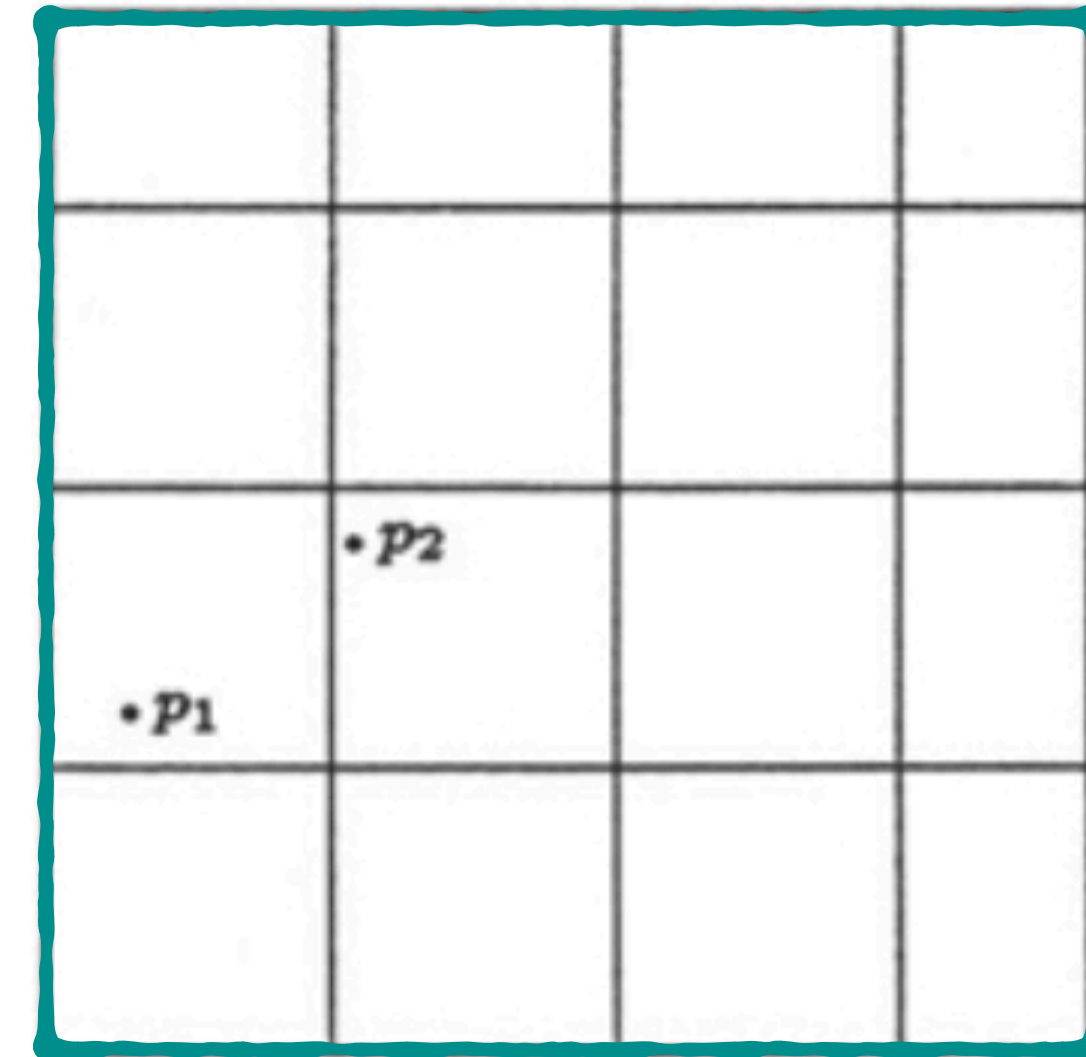
Randomised Incremental Construction

Golin, Raman, Schwarz, Smid 1992/1995

Algorithm $CP(p_1, p_2, \dots, p_n)$

```
(1)  $\delta := d(p_1, p_2); \mathcal{G} := \text{Build}(S_2, \delta);$   
(2) for  $i := 2$  to  $n - 1$  do  
(3)   begin  
(4)    $V := \{\text{Report}(\mathcal{G}, b) : b \text{ is a neighbor of the box containing } p_{i+1}\};$   
(5)    $d := \min_{q \in V} d(p_{i+1}, q);$   
(6)   if  $d \geq \delta$  then  $\text{Insert}(\mathcal{G}, p_{i+1})$   
(7)   else  $\delta := d; \mathcal{G} := \text{Build}(S_{i+1}, \delta);$   
(8)   end;  
(9) return( $\delta$ ).
```

$$\delta(S_2) = d(p_1, p_2)$$



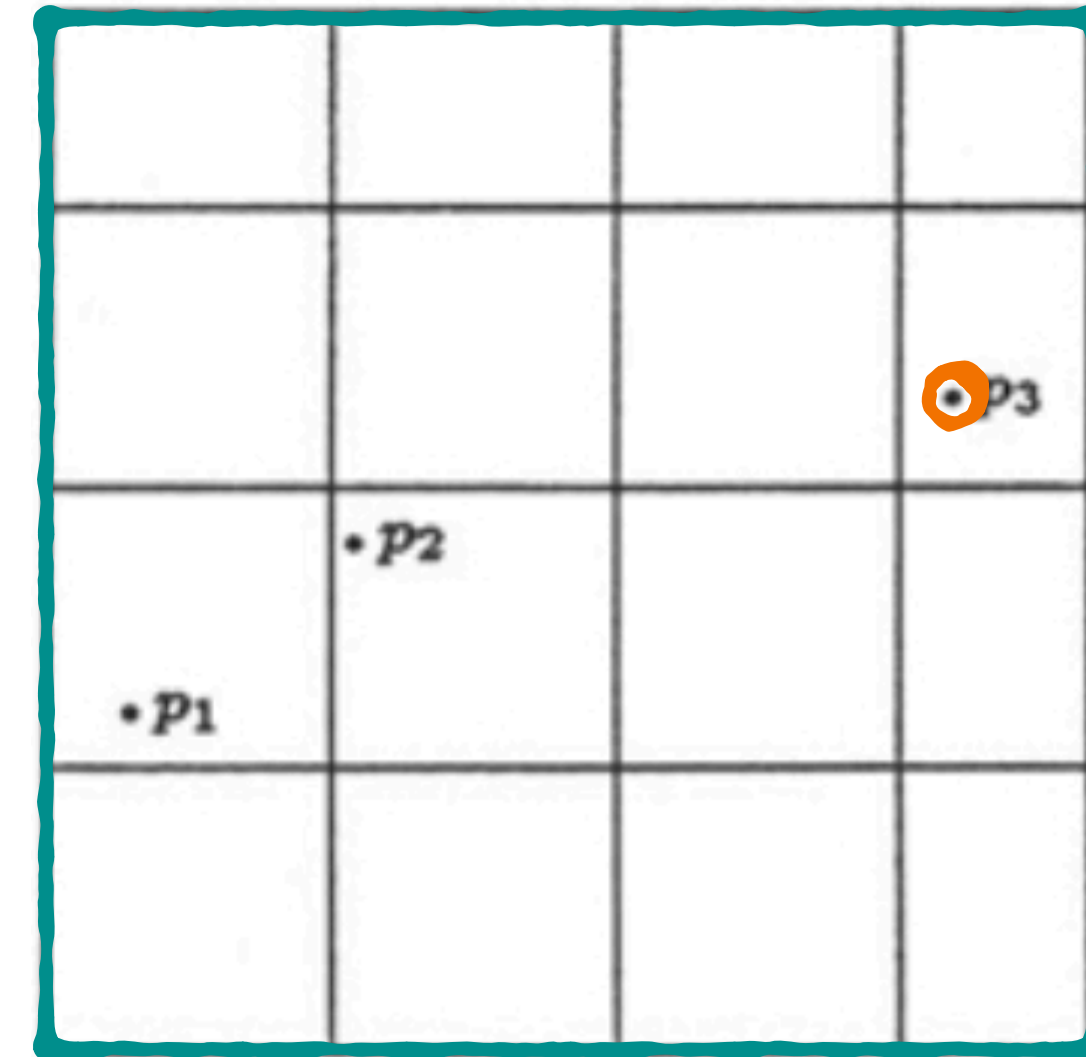
Randomised Incremental Construction

Golin, Raman, Schwarz, Smid 1992/1995

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```

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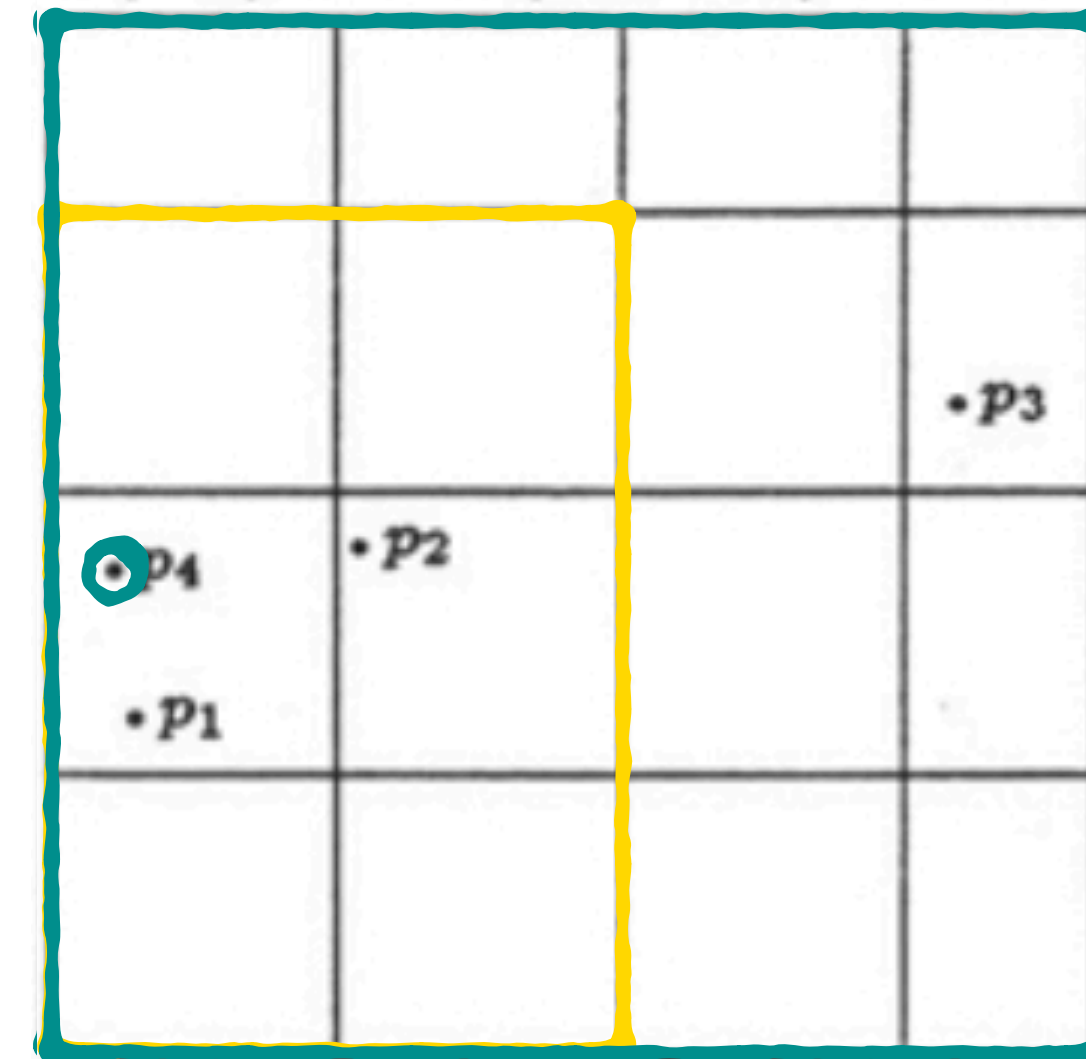
Randomised Incremental Construction

Golin, Raman, Schwarz, Smid 1992/1995

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(1)  $\delta := d(p_1, p_2); \mathcal{G} := Build(S_2, \delta);$   
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(9) return( $\delta$ ).
```

$$\delta(S_3) = d(p_1, p_2)$$



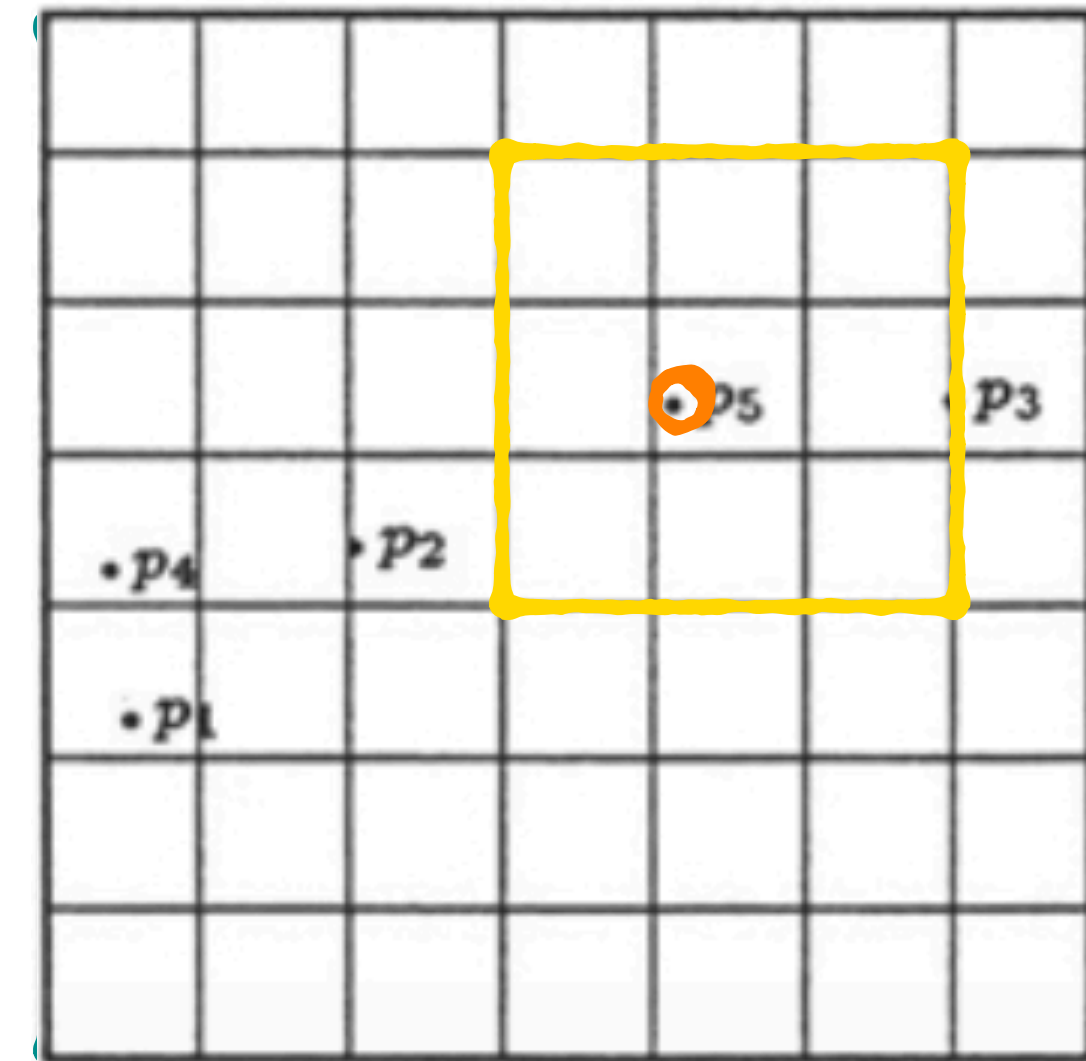
Randomised Incremental Construction

Golin, Raman, Schwarz, Smid 1992/1995

Algorithm $CP(p_1, p_2, \dots, p_n)$

```
(1)  $\delta := d(p_1, p_2); \mathcal{G} := Build(S_2, \delta);$   
(2) for  $i := 2$  to  $n - 1$  do  
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(4)      $V := \{Report(\mathcal{G}, b) : b \text{ is a neighbor of the box containing } p_{i+1}\};$   
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(7)     else  $\delta := d; \mathcal{G} := Build(S_{i+1}, \delta);$   
(8)   end;  
(9) return( $\delta$ ).
```

$$\delta(S_4) = d(p_4, p_1)$$



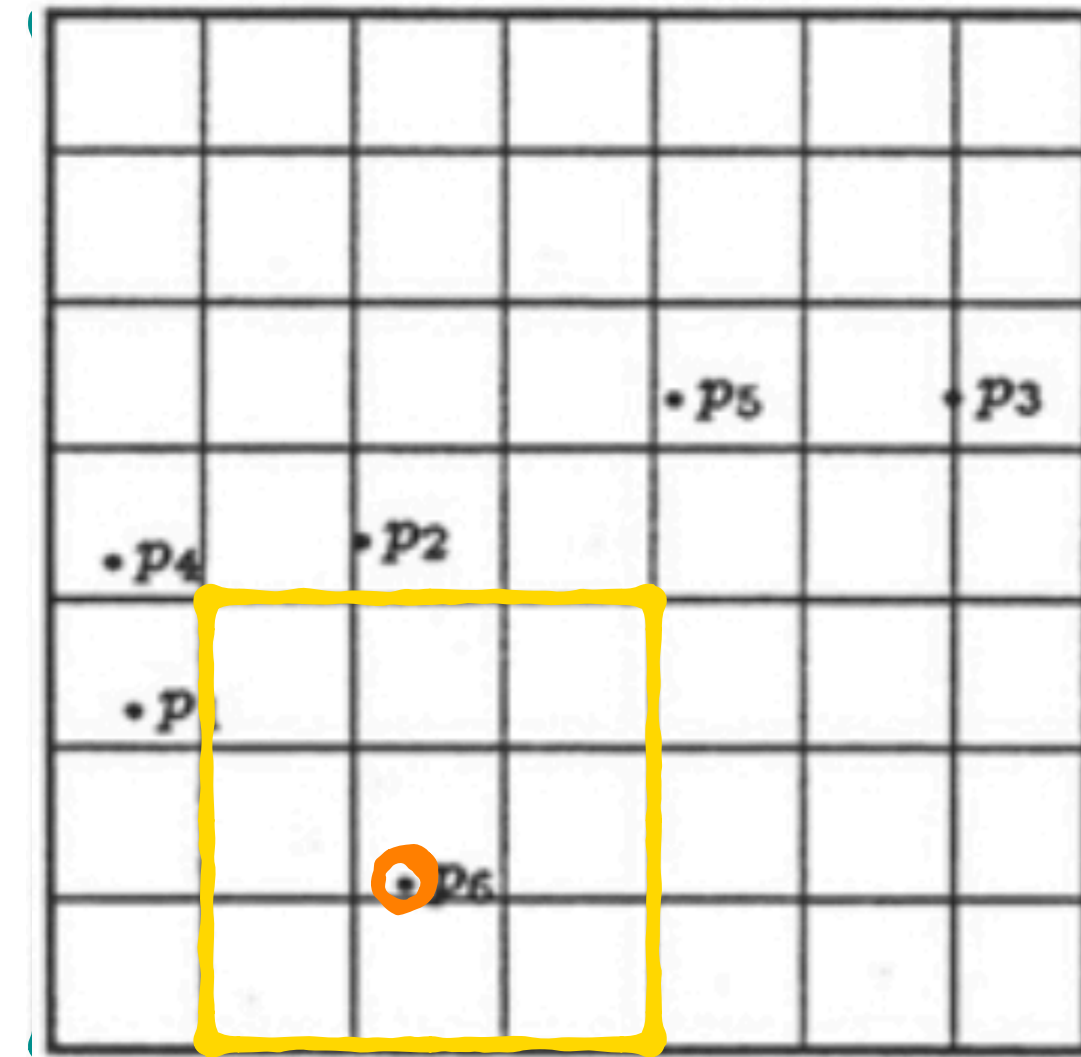
Randomised Incremental Construction

Golin, Raman, Schwarz, Smid 1992/1995

Algorithm $CP(p_1, p_2, \dots, p_n)$

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(7)   else  $\delta := d; \mathcal{G} := Build(S_{i+1}, \delta);$   
(8)   end;  
(9) return( $\delta$ ).
```

$$\delta(S_5) = d(p_4, p_1)$$



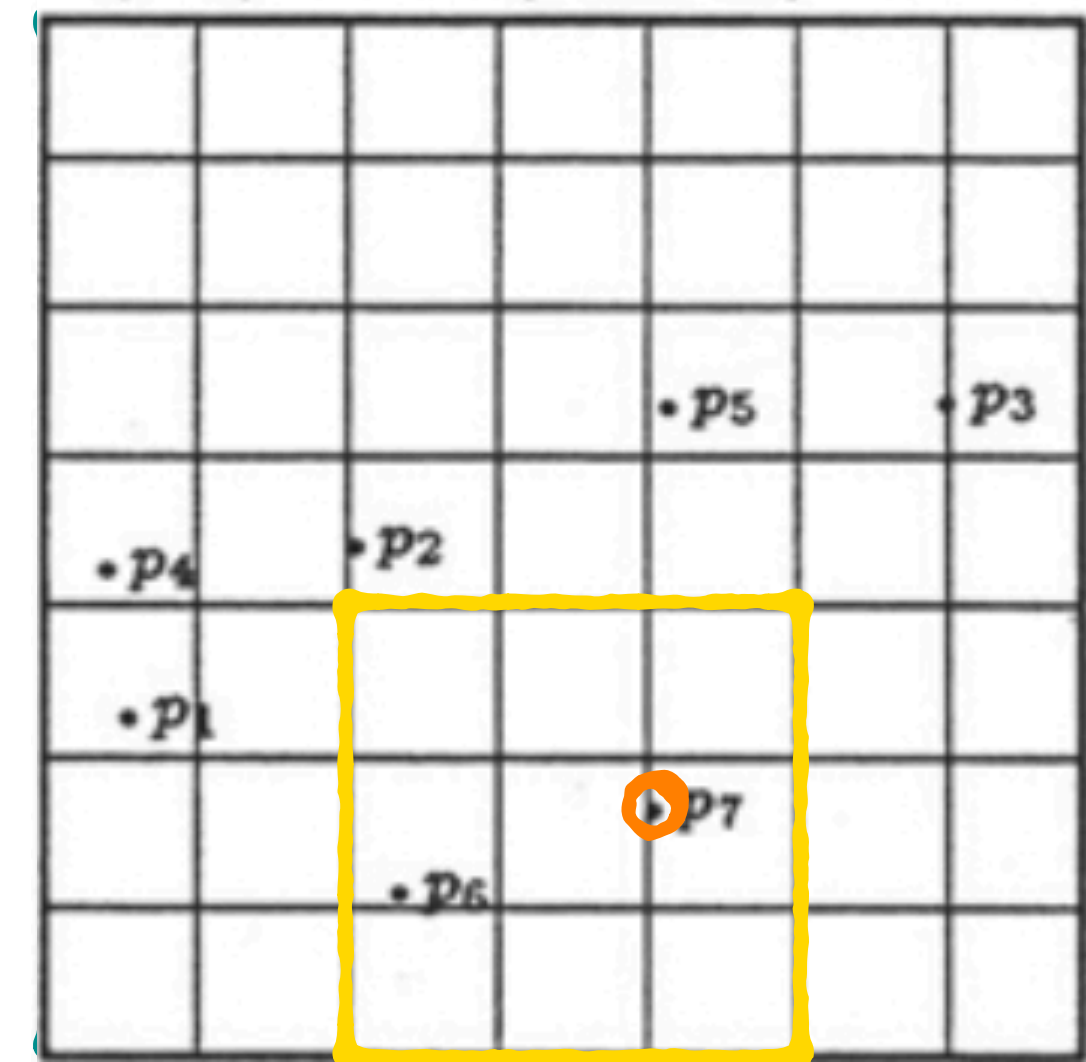
Randomised Incremental Construction

Golin, Raman, Schwarz, Smid 1992/1995

Algorithm $CP(p_1, p_2, \dots, p_n)$

```
(1)  $\delta := d(p_1, p_2); \mathcal{G} := Build(S_2, \delta);$   
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(7)   else  $\delta := d; \mathcal{G} := Build(S_{i+1}, \delta);$   
(8)   end;  
(9) return( $\delta$ ).
```

$$\delta(S_6) = d(p_4, p_1)$$



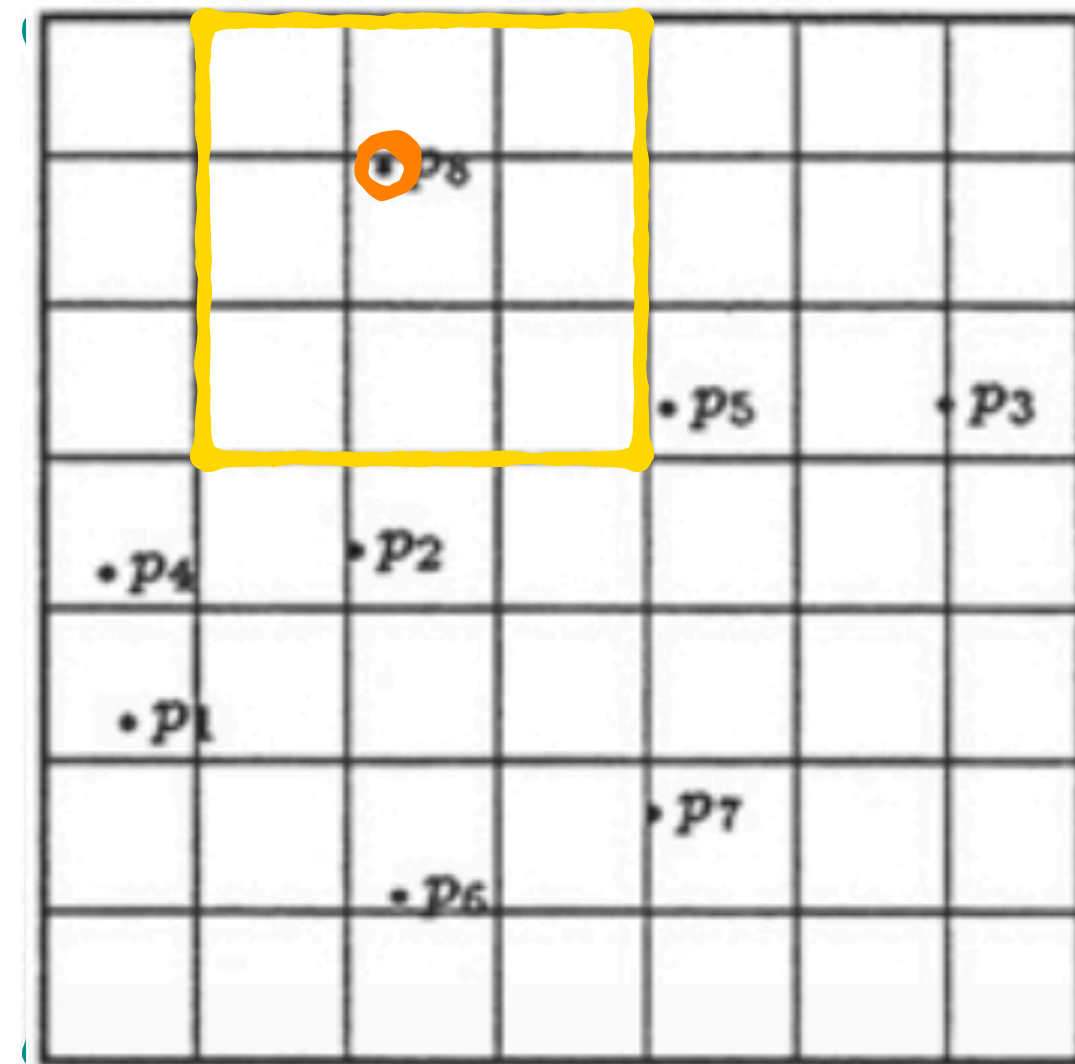
Randomised Incremental Construction

Golin, Raman, Schwarz, Smid 1992/1995

Algorithm $CP(p_1, p_2, \dots, p_n)$

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- (5) $d := \min_{q \in V} d(p_{i+1}, q);$
- (6) if $d \geq \delta$ then $\text{Insert}(\mathcal{G}, p_{i+1})$
- (7) else $\delta := d; \mathcal{G} := \text{Build}(S_{i+1}, \delta);$
- (8) end;
- (9) return(δ).

$$\delta(S_7) = d(p_4, p_1)$$



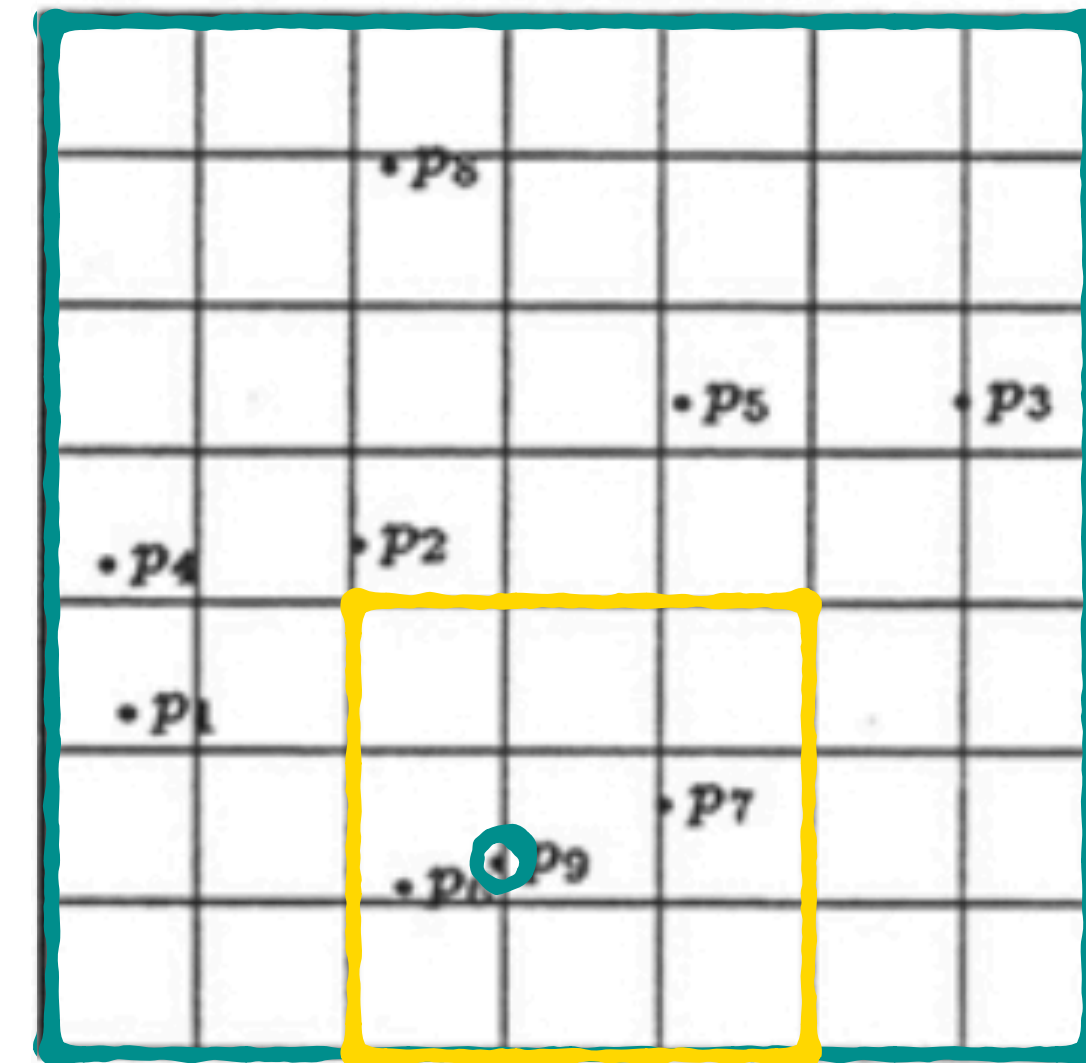
Randomised Incremental Construction

Golin, Raman, Schwarz, Smid 1992/1995

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(7)   else  $\delta := d; \mathcal{G} := Build(S_{i+1}, \delta);$   
(8)   end;  
(9) return( $\delta$ ).
```

$$\delta(S_8) = d(p_4, p_1)$$



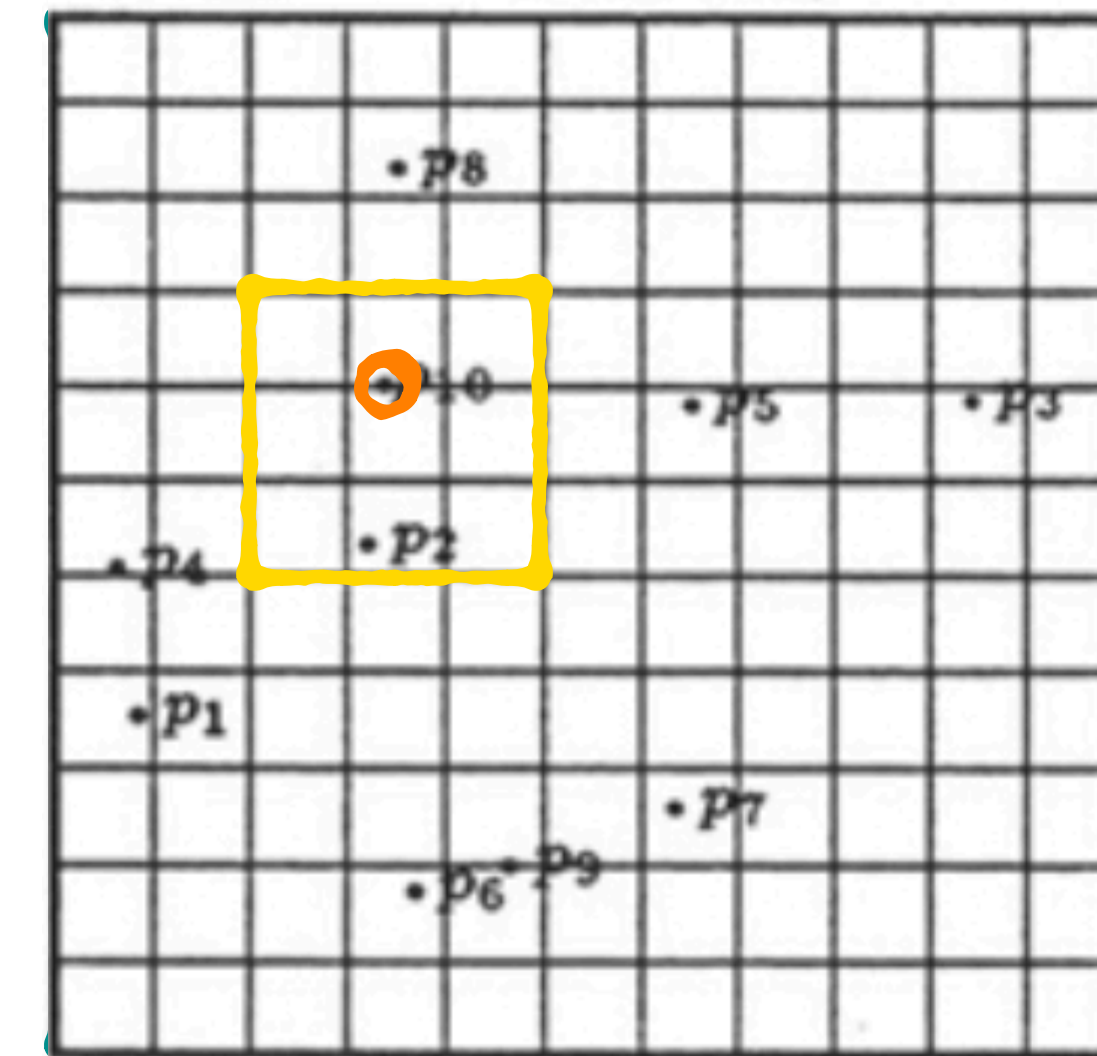
Randomised Incremental Construction

Golin, Raman, Schwarz, Smid 1992/1995

Algorithm $CP(p_1, p_2, \dots, p_n)$

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(8)   end;  
(9) return( $\delta$ ).
```

$$\delta(S_9) = d(p_9, p_6)$$



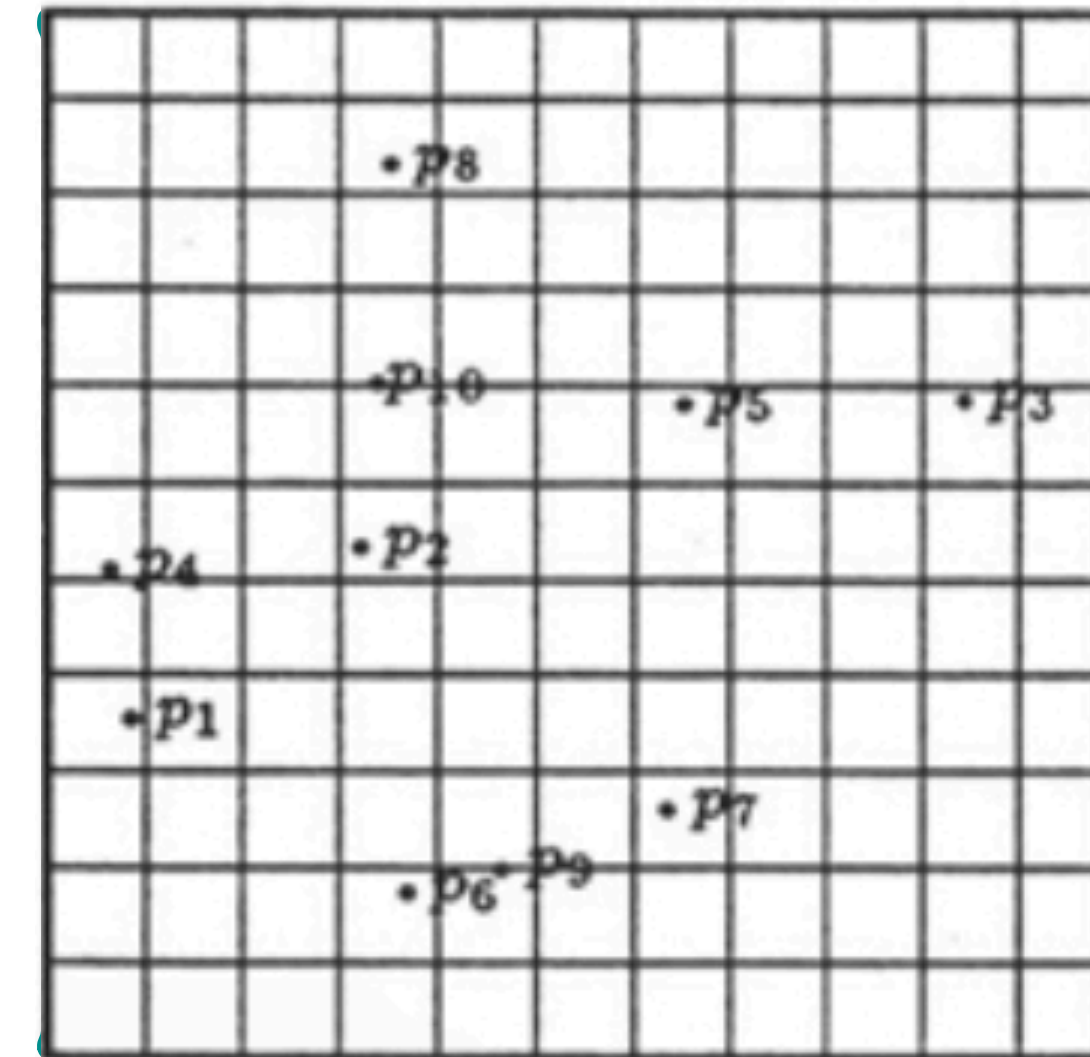
Randomised Incremental Construction

Golin, Raman, Schwarz, Smid 1992/1995

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- (1) $\delta := d(p_1, p_2); \mathcal{G} := Build(S_2, \delta);$
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- (7) else $\delta := d; \mathcal{G} := Build(S_{i+1}, \delta);$
- (8) end;
- (9) return(δ).

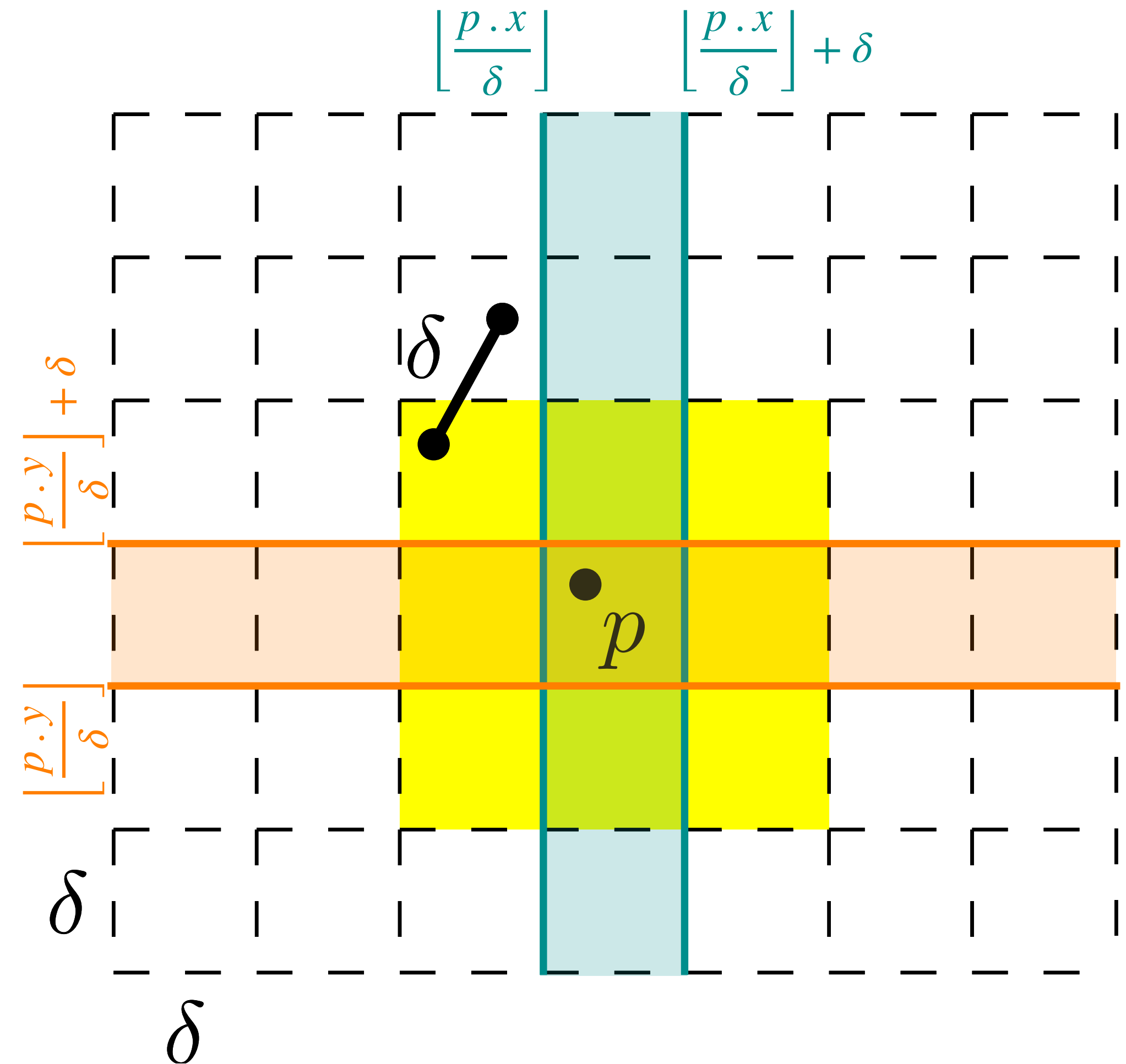
$$\delta(S_{10}) = d(p_9, p_6)$$



Fairly straightforward algorithm :)

Randomised Incremental Construction

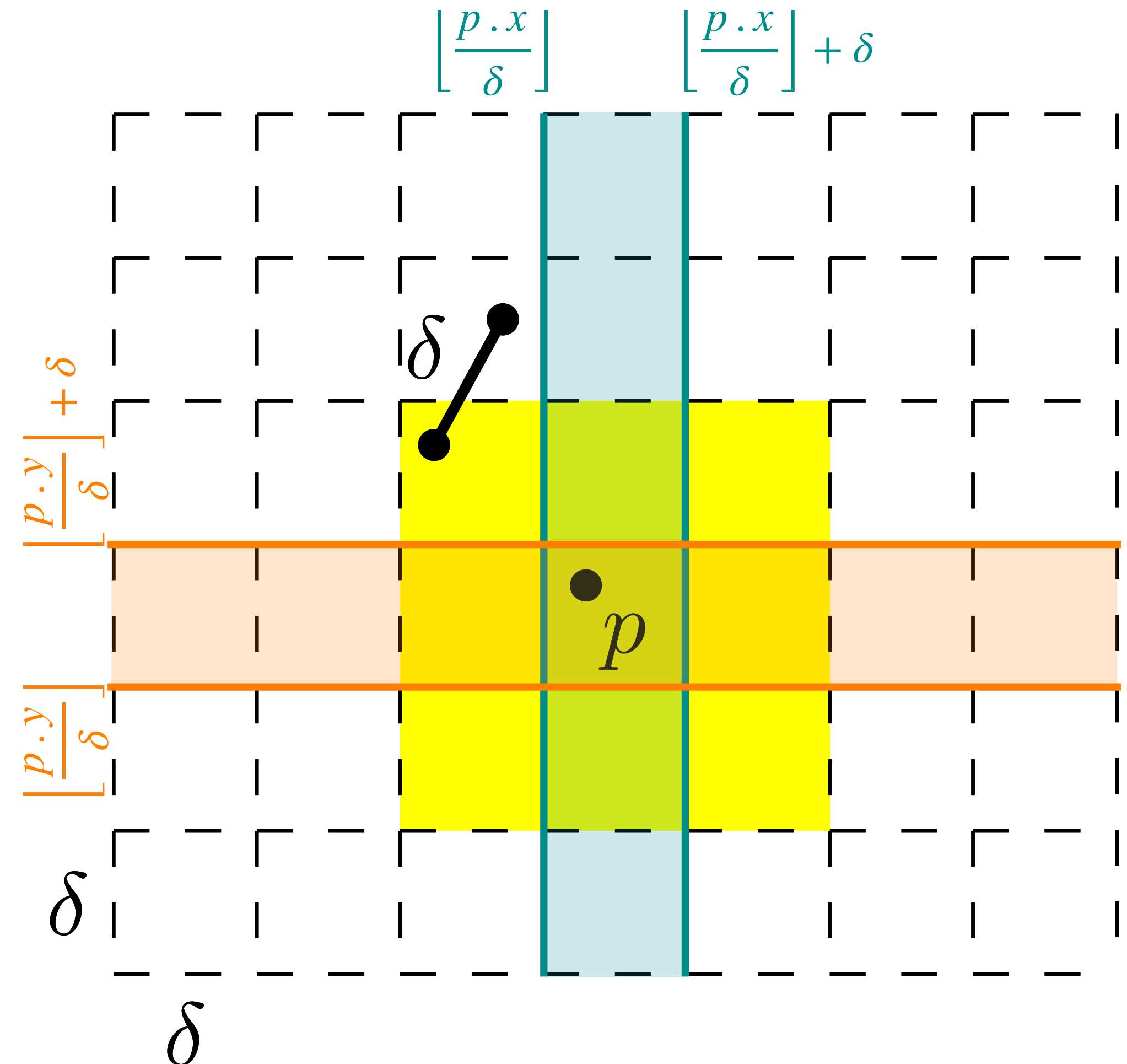
Golin, Raman, Schwarz, Smid 1992/1995



Randomised Incremental Construction

Golin, Raman, Schwarz, Smid 1992/1995

Representation



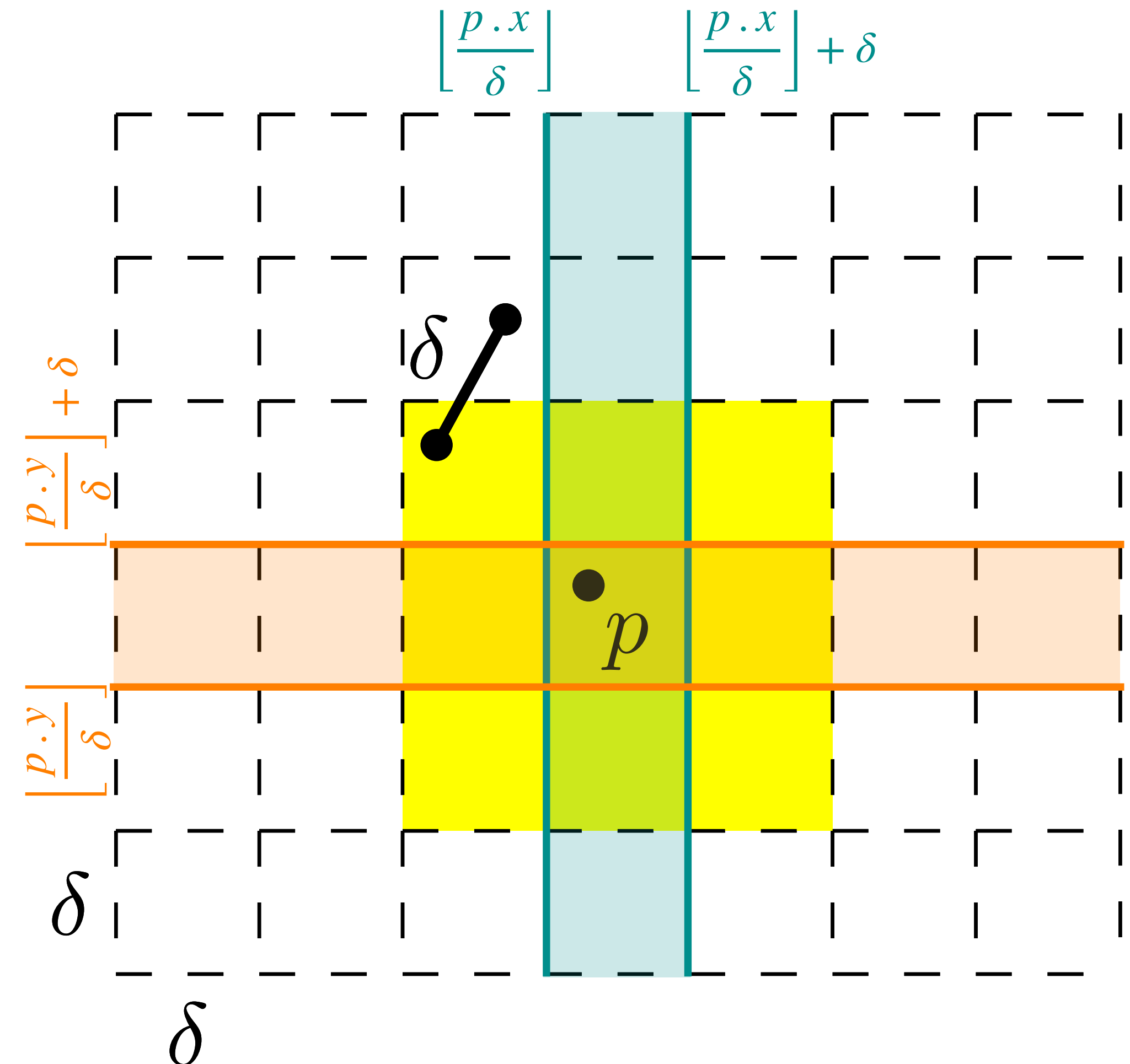
Randomised Incremental Construction

Golin, Raman, Schwarz, Smid 1992/1995

Representation

- Grid cell representation

(x, y)

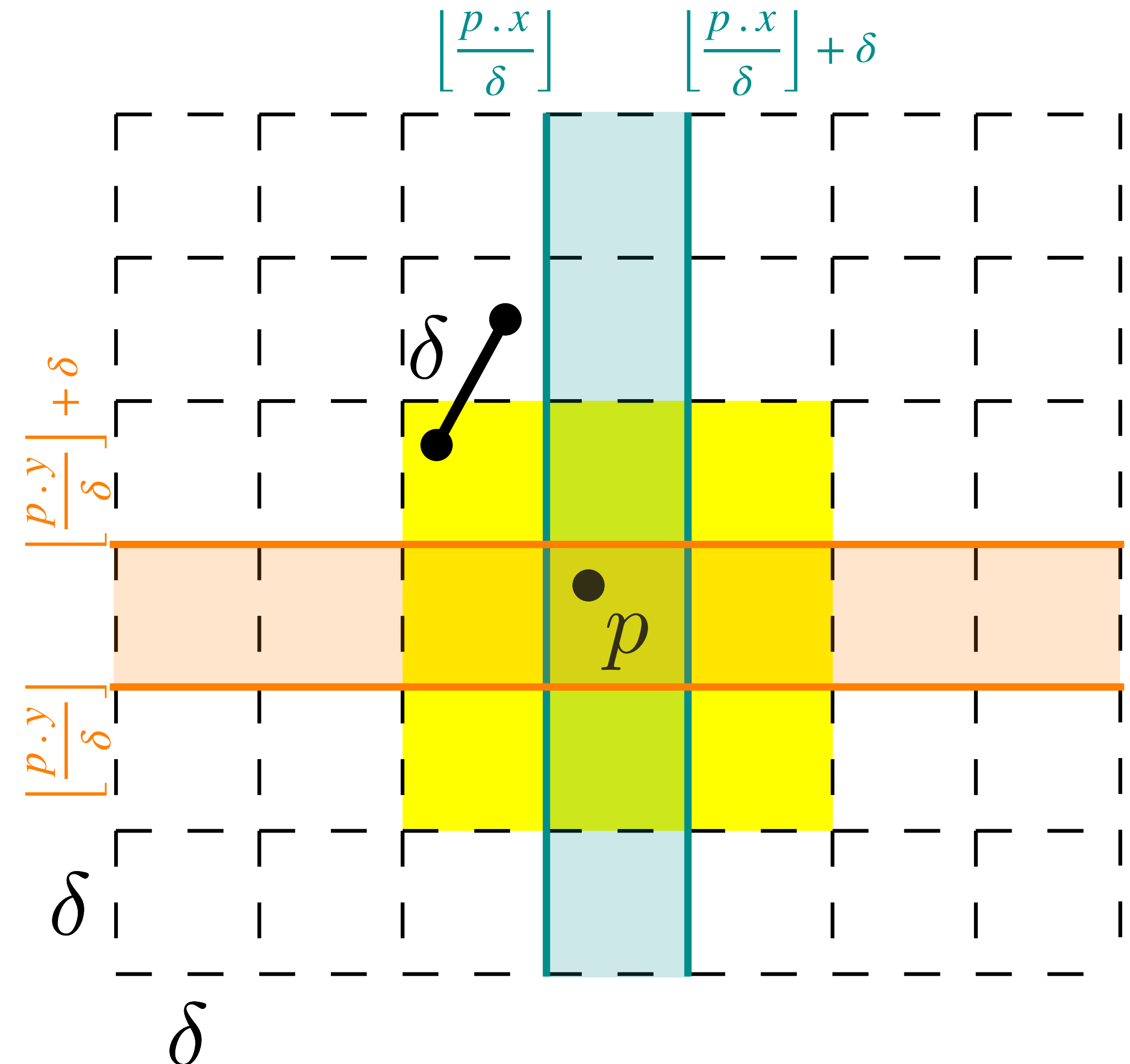


Randomised Incremental Construction

Golin, Raman, Schwarz, Smid 1992/1995

Representation

- Grid cell representation (x, y)
- Grid cell assignment G_δ `OrderedDictionary`



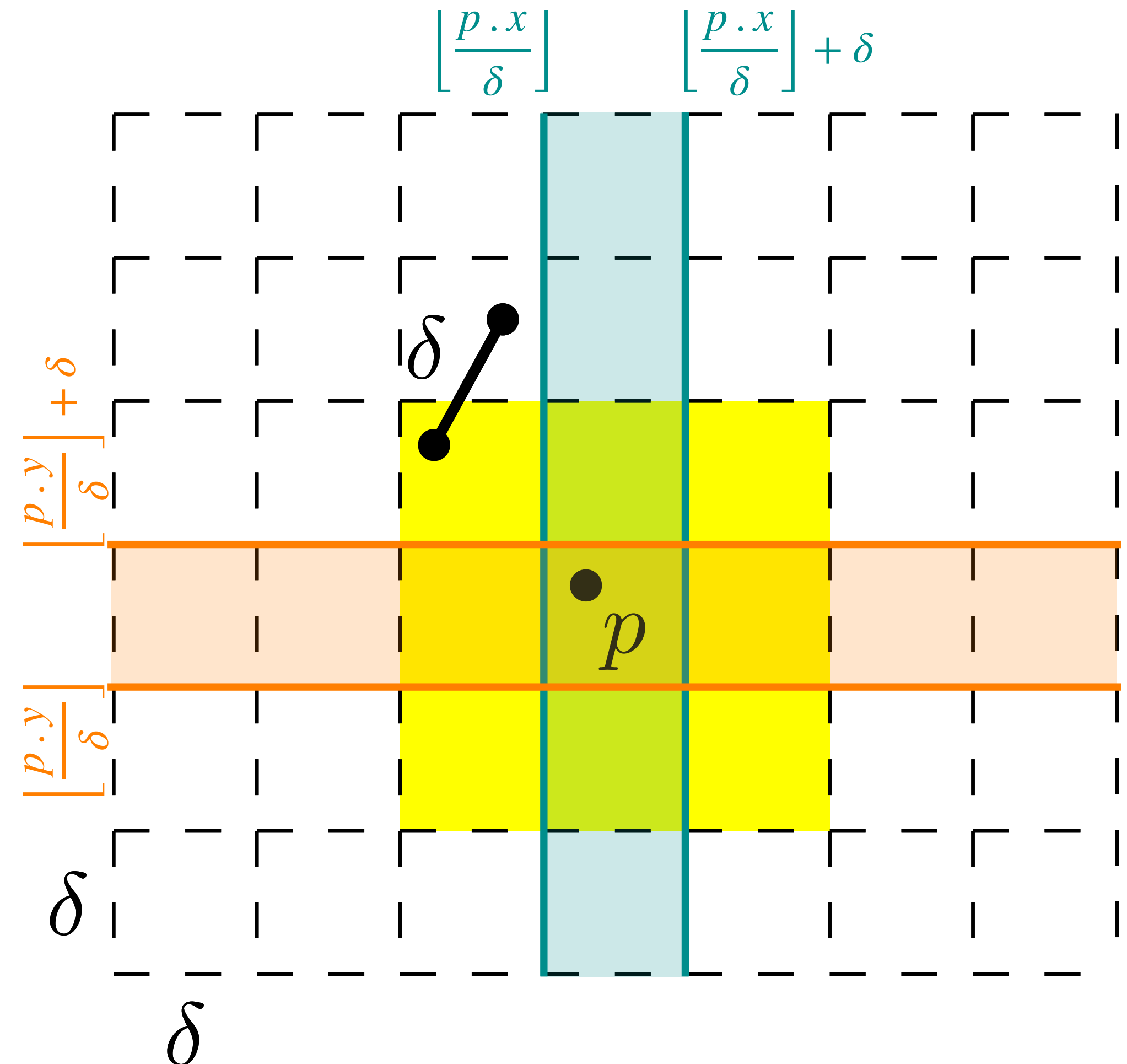
Randomised Incremental Construction

Golin, Raman, Schwarz, Smid 1992/1995

Representation

- Grid cell representation
- Grid cell assignment G_δ
- Identifier/key for $p \in \mathcal{P}$

(x, y)
OrderedDictionary
 $(p.x, p.y)$

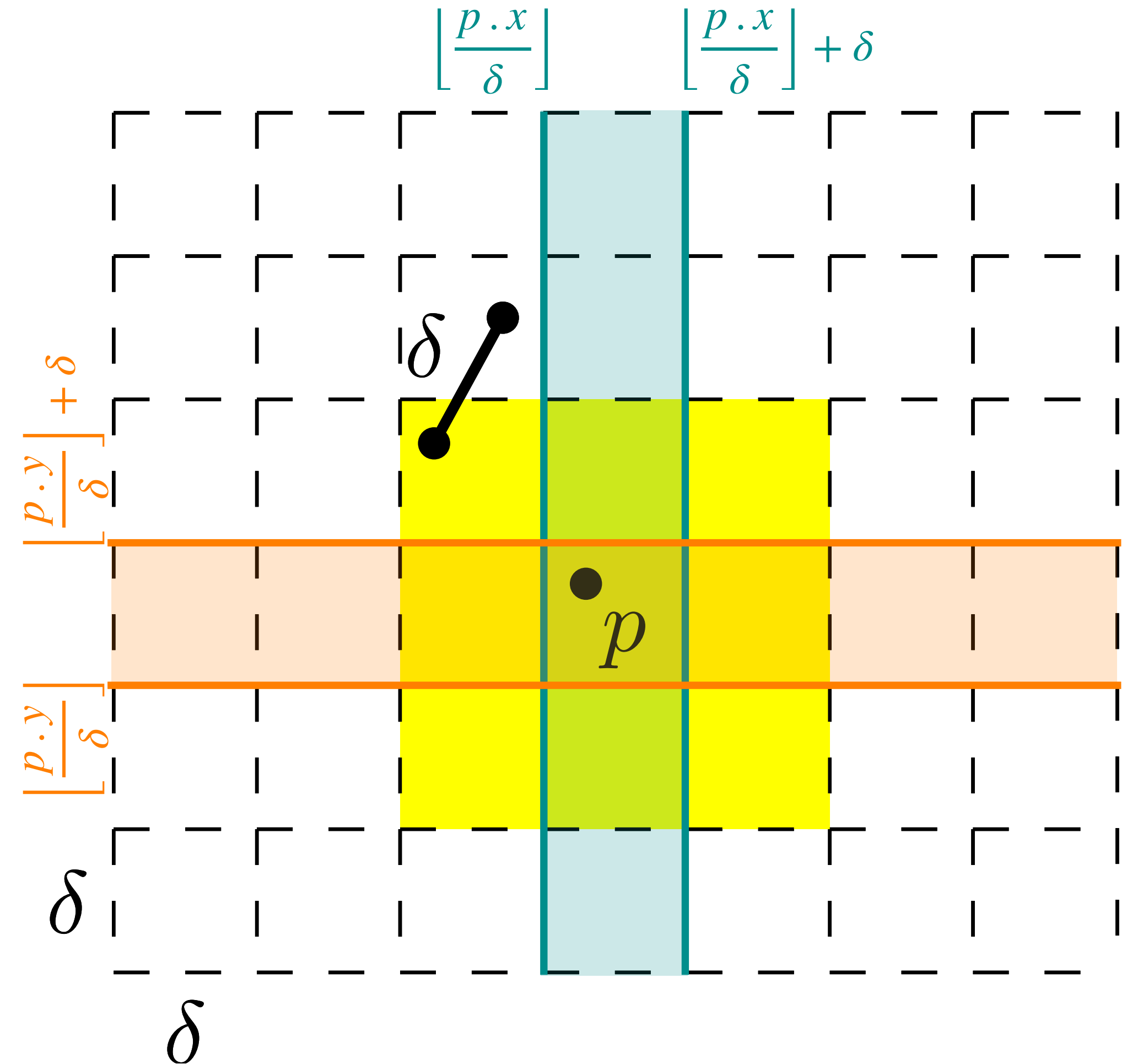


Randomised Incremental Construction

Golin, Raman, Schwarz, Smid 1992/1995

Representation

- Grid cell representation (x, y)
- Grid cell assignment G_δ **OrderedDictionary**
- Identifier/key for $p \in \mathcal{P}$ $(p.x, p.y)$
- $p \in G_\delta[x, y] \Leftrightarrow x = \left\lfloor \frac{p.x}{\delta} \right\rfloor, y = \left\lfloor \frac{p.y}{\delta} \right\rfloor$



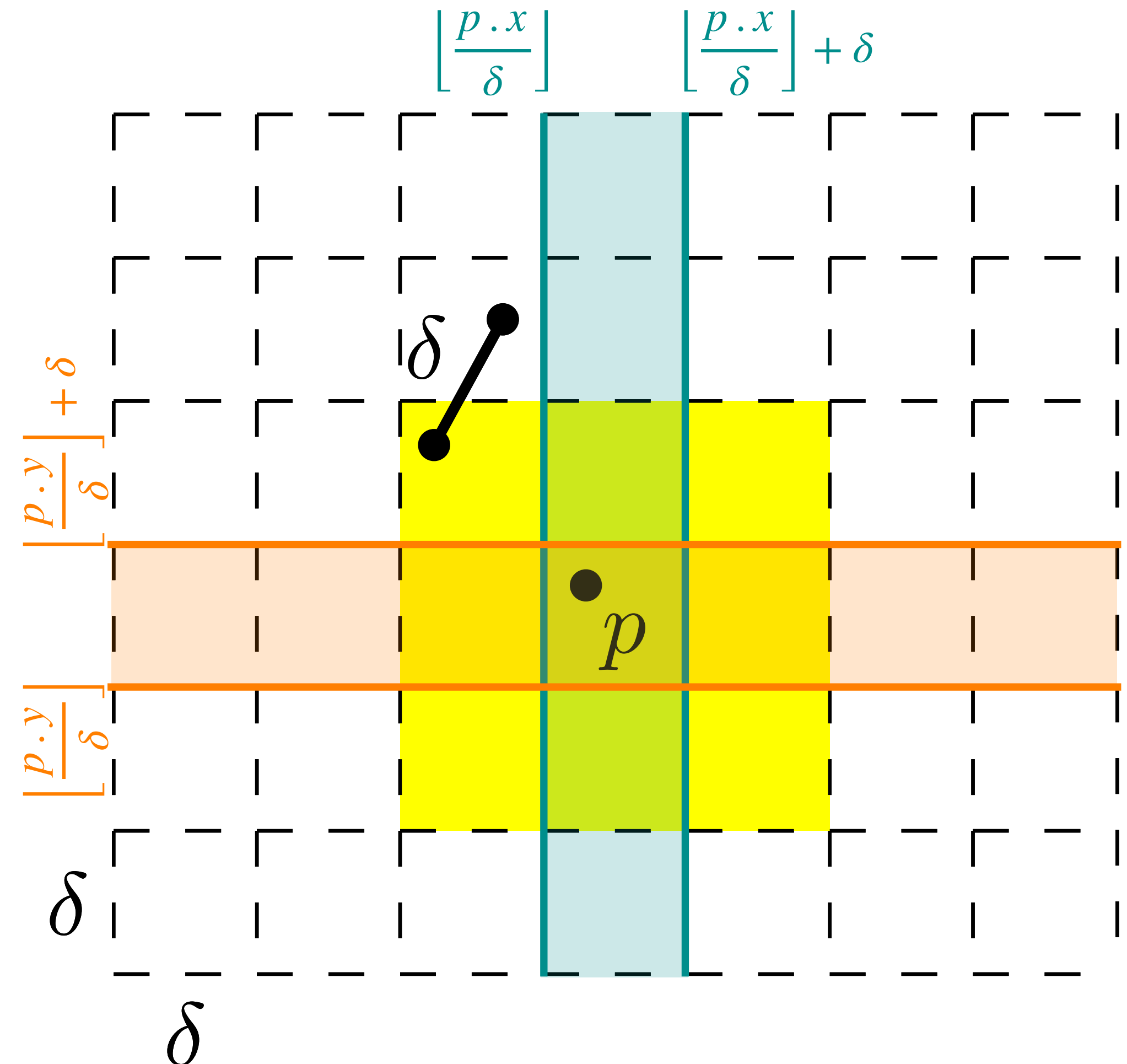
Randomised Incremental Construction

Golin, Raman, Schwarz, Smid 1992/1995

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Implementation



Randomised Incremental Construction

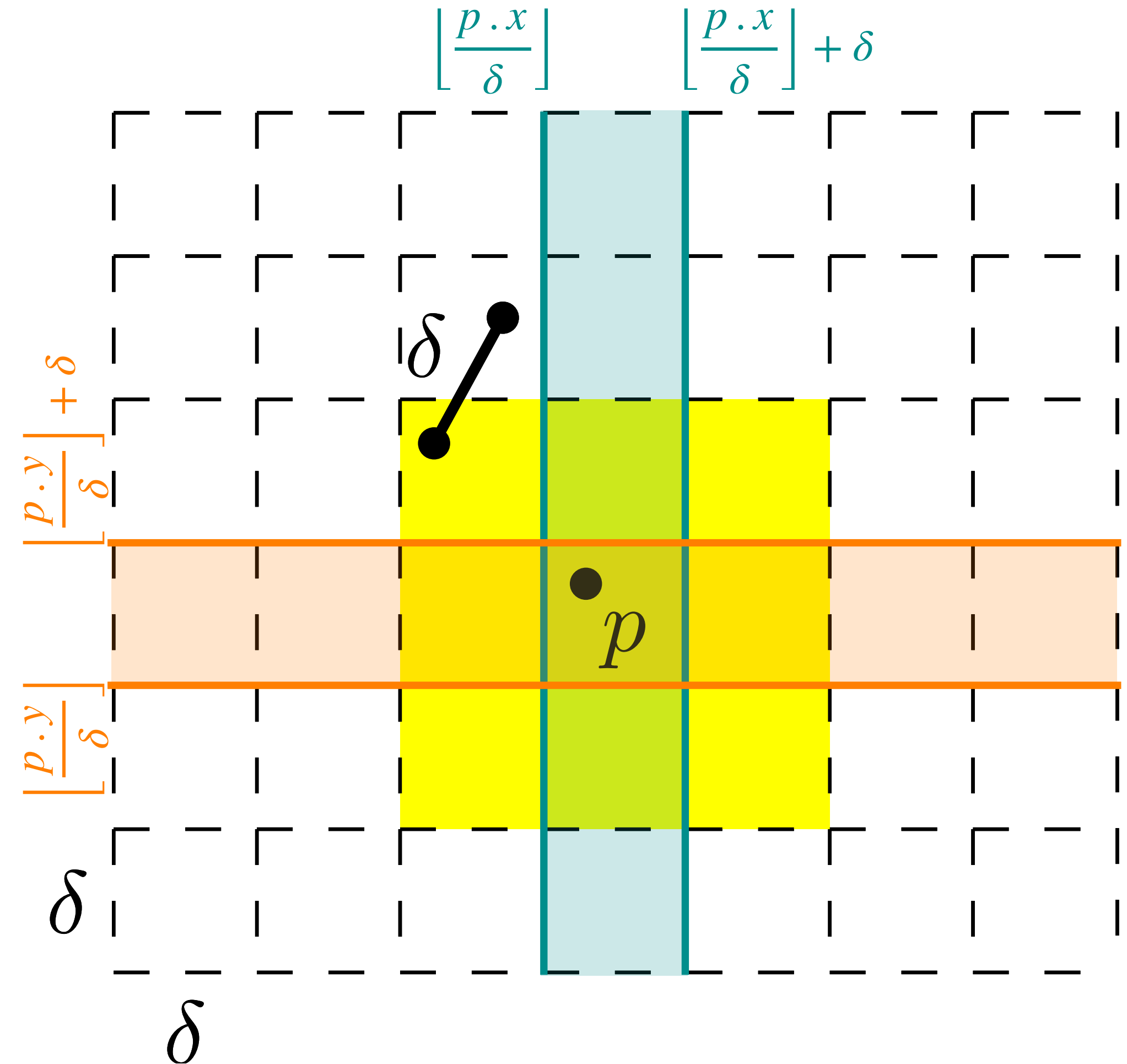
Golin, Raman, Schwarz, Smid 1992/1995

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Implementation

- $T_{build}(n)$ Construction cost



Randomised Incremental Construction

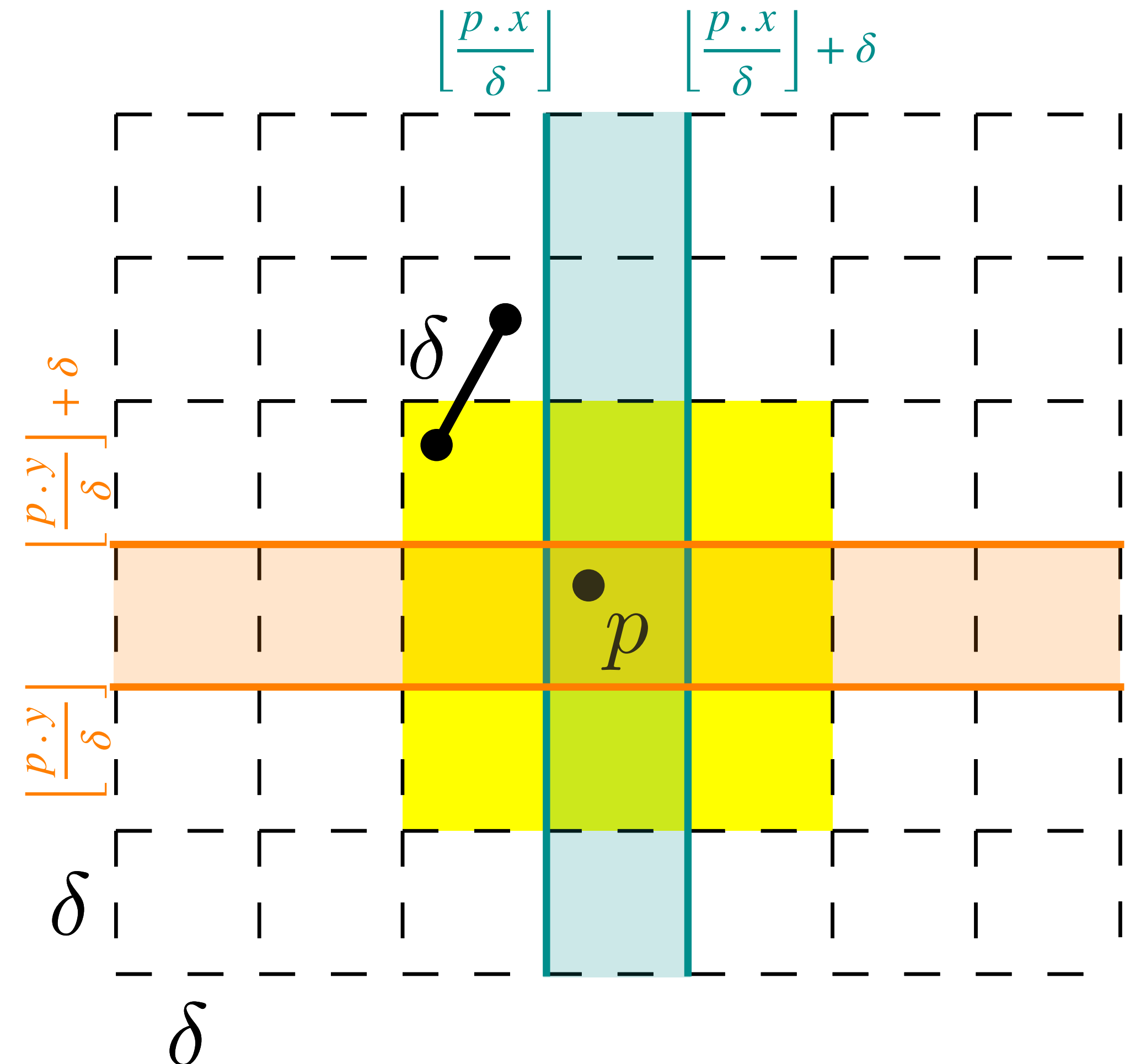
Golin, Raman, Schwarz, Smid 1992/1995

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- Grid cell assignment G_δ **OrderedDictionary**
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Implementation

- $T_{build}(n)$ Construction cost
- $T_{insert}(n)$ Insertion cost



Randomised Incremental Construction

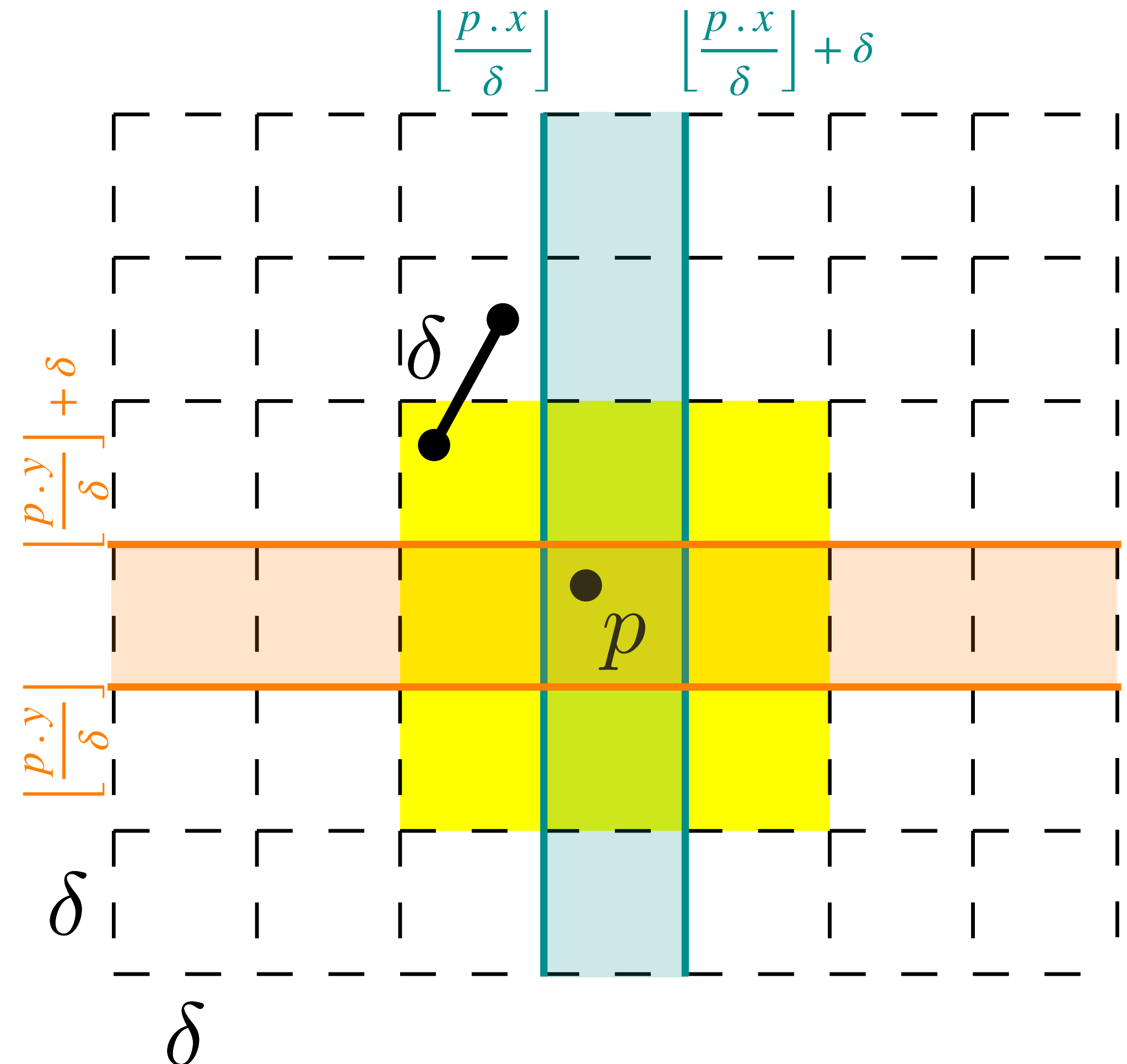
Golin, Raman, Schwarz, Smid 1992/1995

Representation

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- $p \in G_\delta[x, y] \Leftrightarrow x = \left\lfloor \frac{p.x}{\delta} \right\rfloor, y = \left\lfloor \frac{p.y}{\delta} \right\rfloor$

Implementation

- $T_{build}(n)$ Construction cost
- $T_{insert}(n)$ Insertion cost
- $T_{query}(n)$ Query cost



Randomised Incremental Construction

Golin, Raman, Schwarz, Smid 1992/1995

Algorithm $CP(p_1, p_2, \dots, p_n)$

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(8)   end;  
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```

Randomised Incremental Construction

Golin, Raman, Schwarz, Smid 1992/1995

Terminology

Algorithm $CP(p_1, p_2, \dots, p_n)$

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(7)       else  $\delta := d; \mathcal{G} := Build(S_{i+1}, \delta);$   
(8)   end;  
(9) return( $\delta$ ).
```

Randomised Incremental Construction

Golin, Raman, Schwarz, Smid 1992/1995

Terminology

- (p_1, p_2, \dots, p_n)

Permuted sequence

Algorithm $CP(p_1, p_2, \dots, p_n)$

```
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Randomised Incremental Construction

Golin, Raman, Schwarz, Smid 1992/1995

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- Randomised: $X(p_i, \mathcal{P}_{i-1}) = \begin{cases} 1, & \text{if } \delta_i < \delta_{i-1} \\ 0, & \text{otherwise.} \end{cases}$

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Randomised Incremental Construction

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Randomised Incremental Construction

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Lemma 1 *Let p_1, p_2, \dots, p_n be a random permutation of the points of S . Let $S_i := \{p_1, p_2, \dots, p_i\}$. Then $\Pr[\delta(S_{i+1}) < \delta(S_i)] \leq 2/(i+1)$.*

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$$Pr[\{\delta_{i+1} < \delta_i\}] = 0$$

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Impossible — one such pair must have been known.

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Thank you.

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