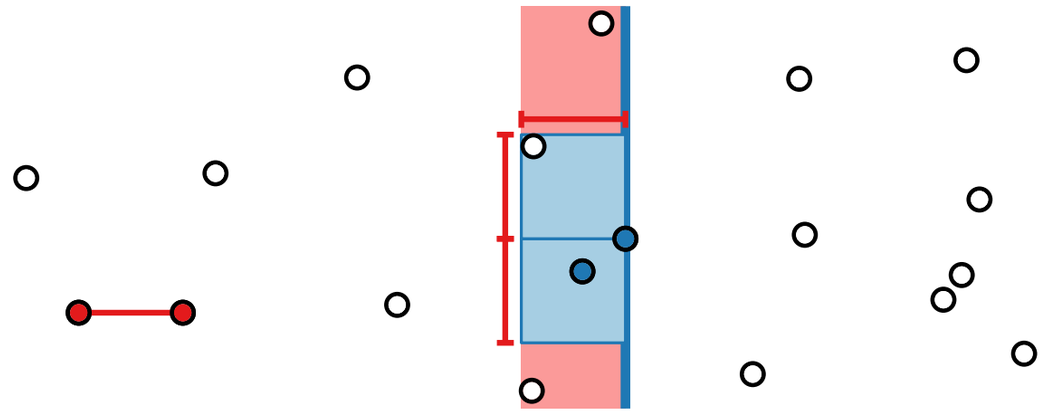
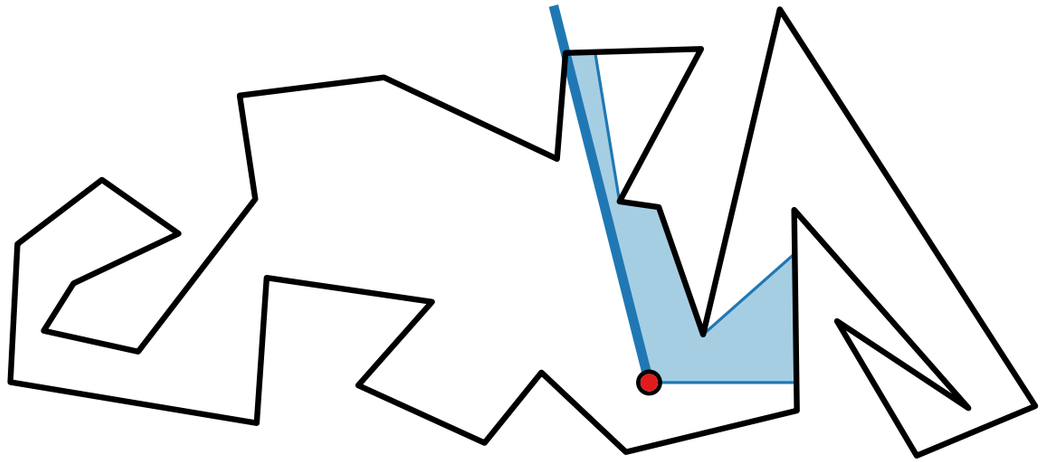
Julius-Maximilians-
UNIVERSITÄT
WÜRZBURG

Advanced Algorithms

Computational Geometry

Sweep-Line Algorithms





Alexander Wolff

Summer term 2026

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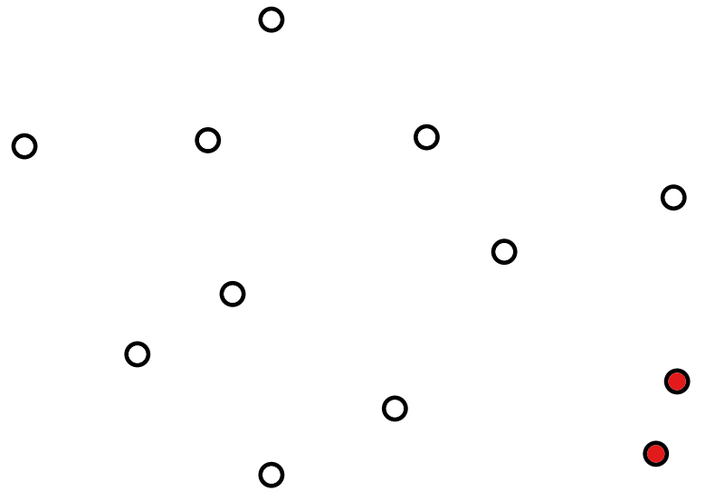
- Closest Pair



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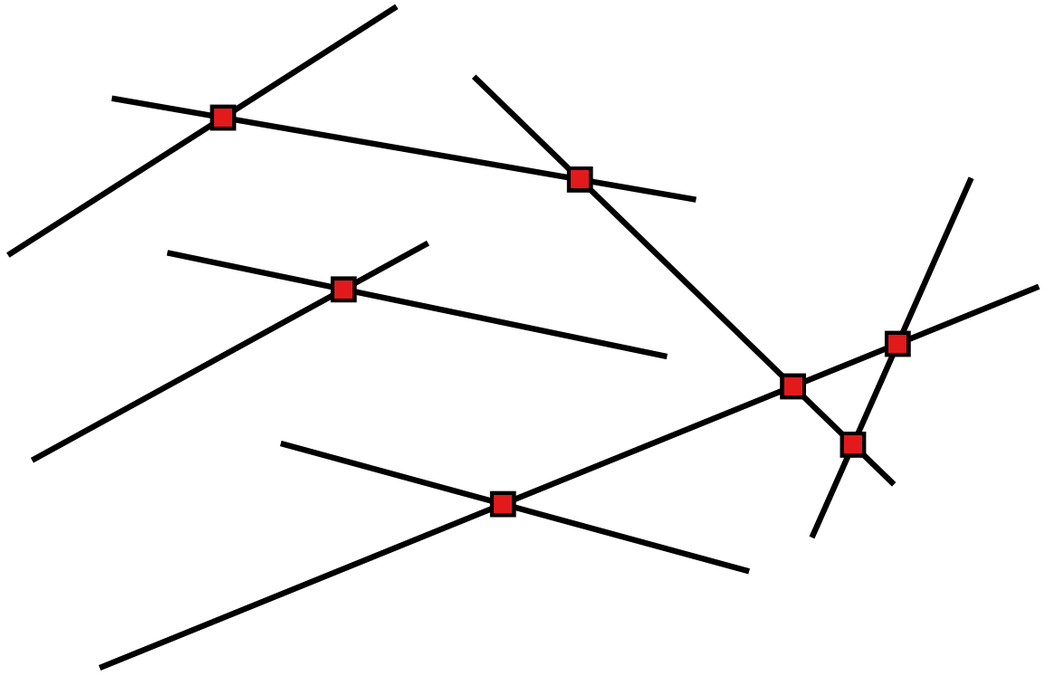

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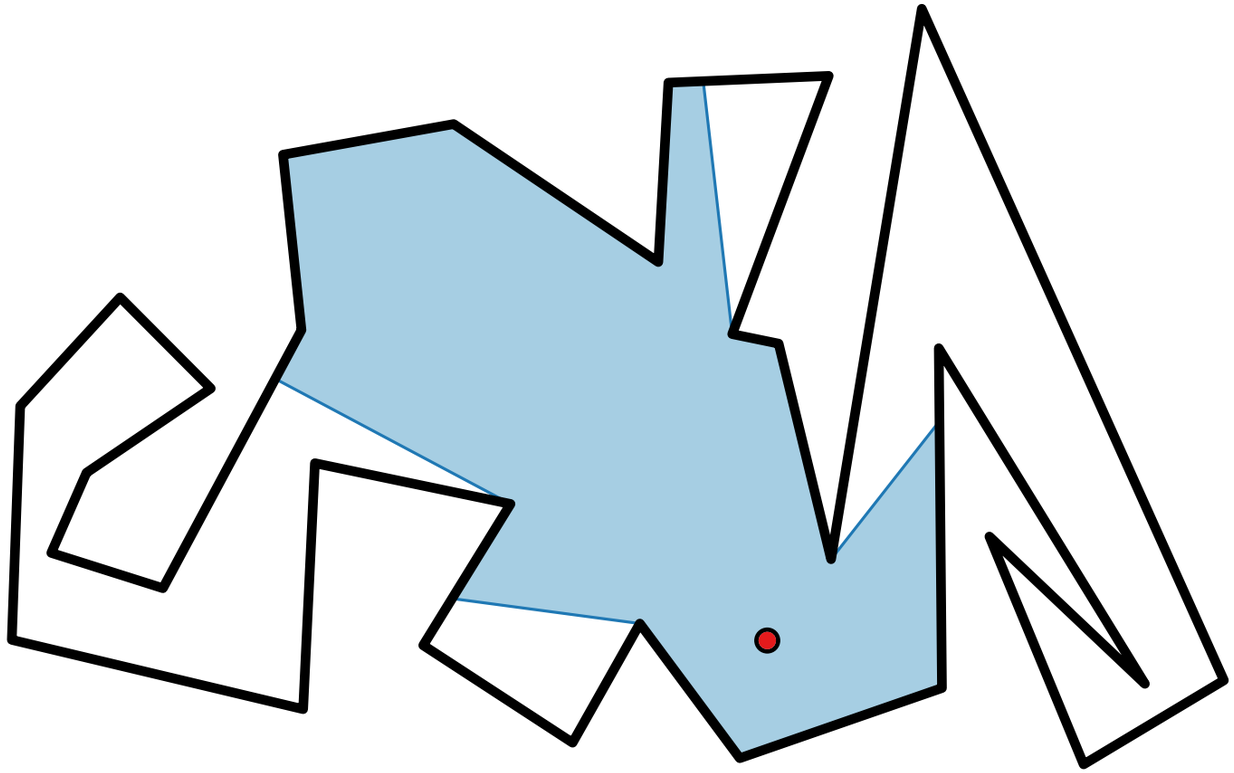

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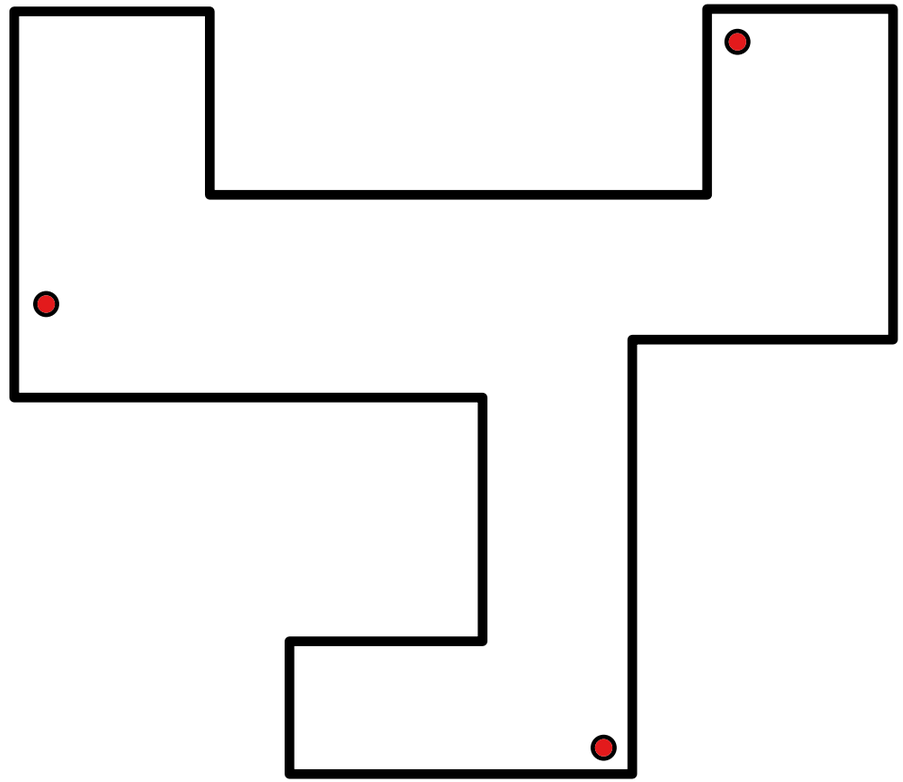

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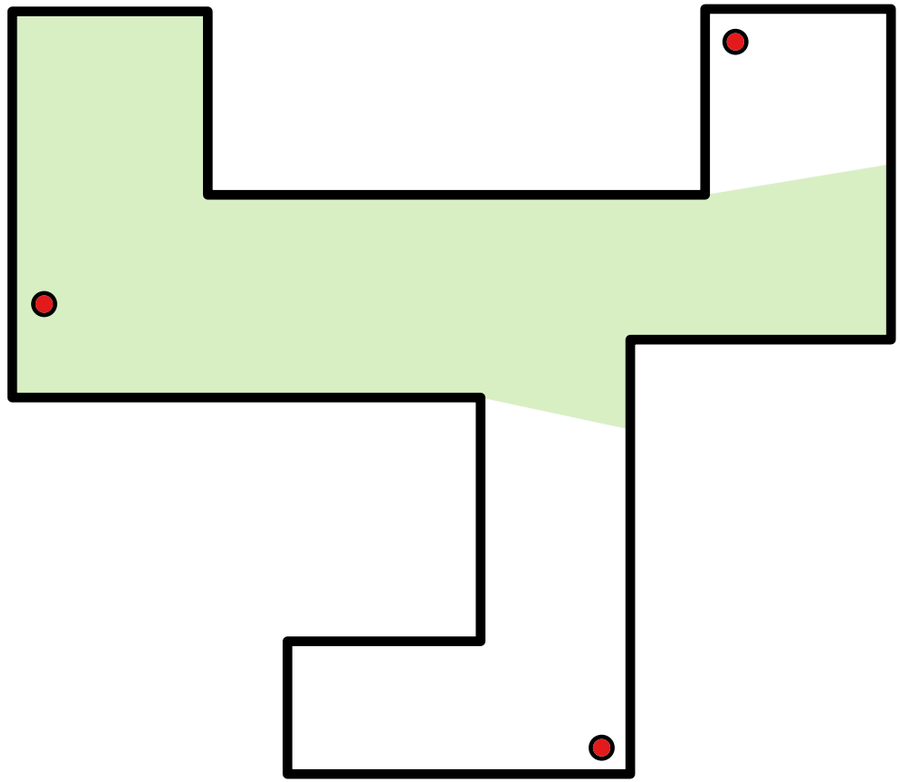

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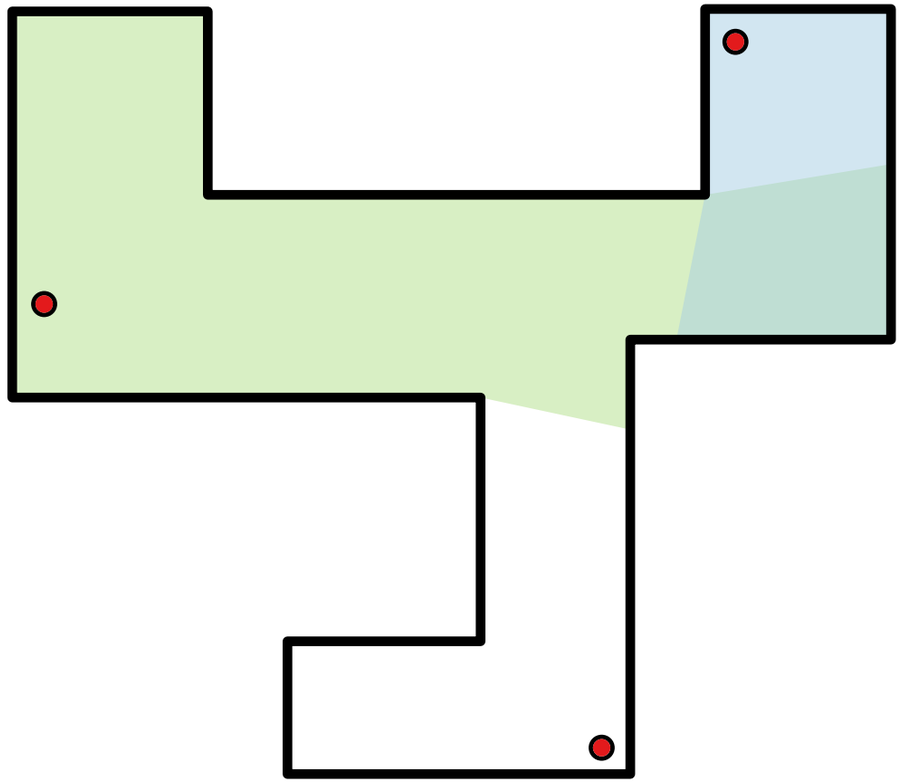

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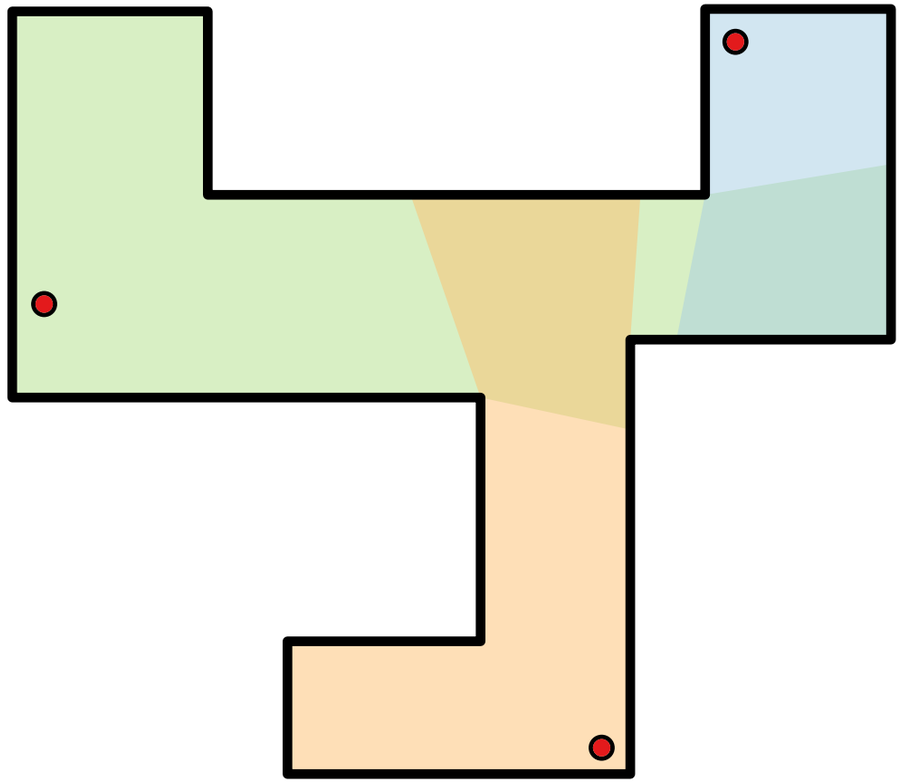

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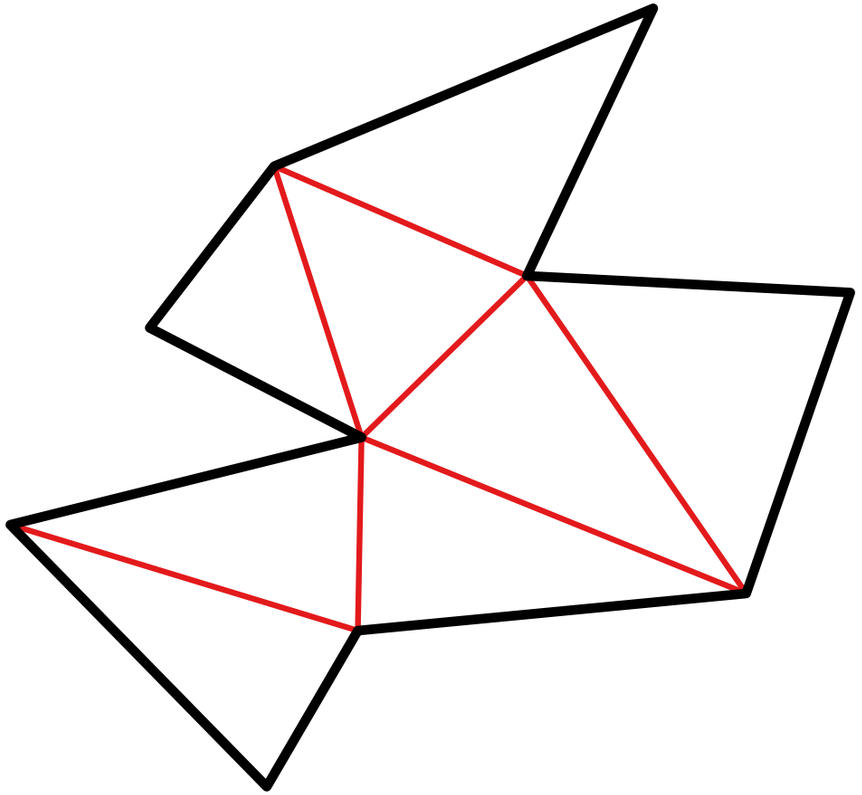

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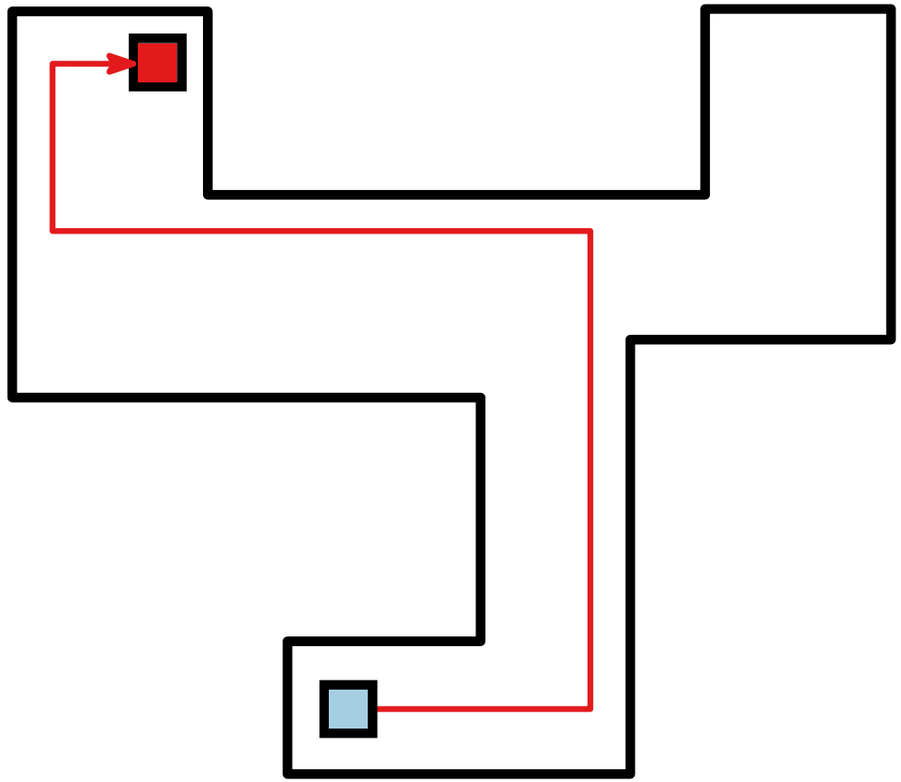

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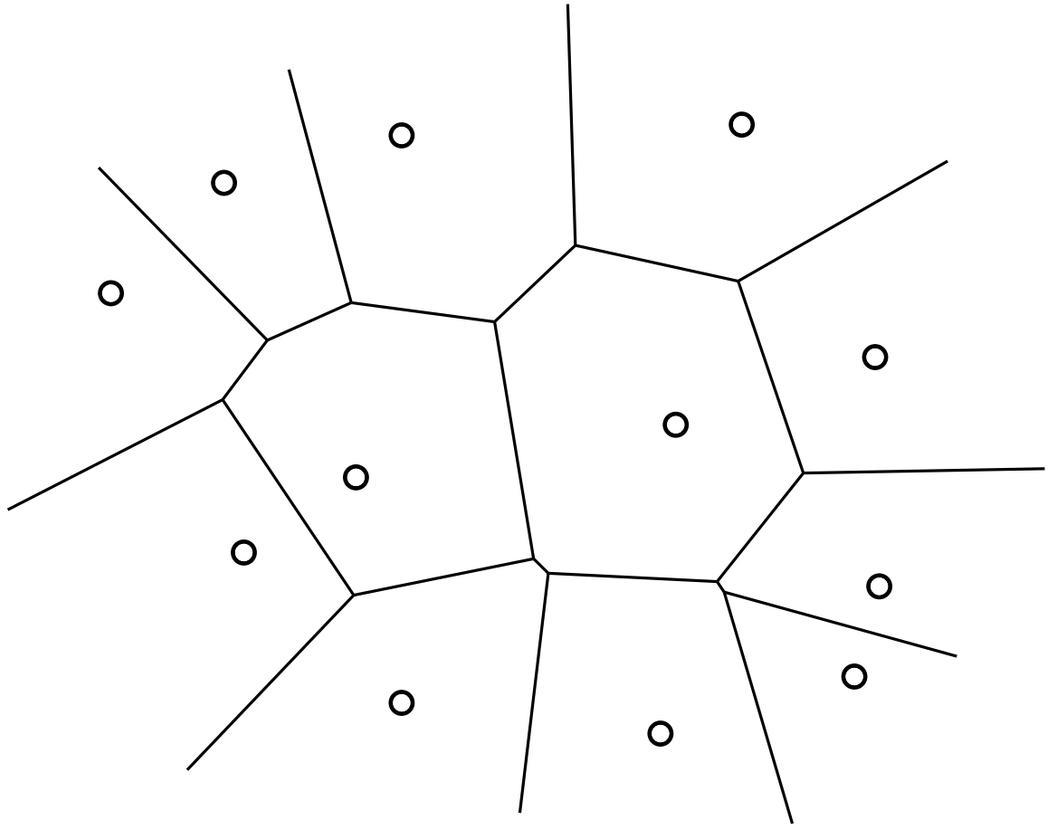

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Some problems:

- Closest Pair
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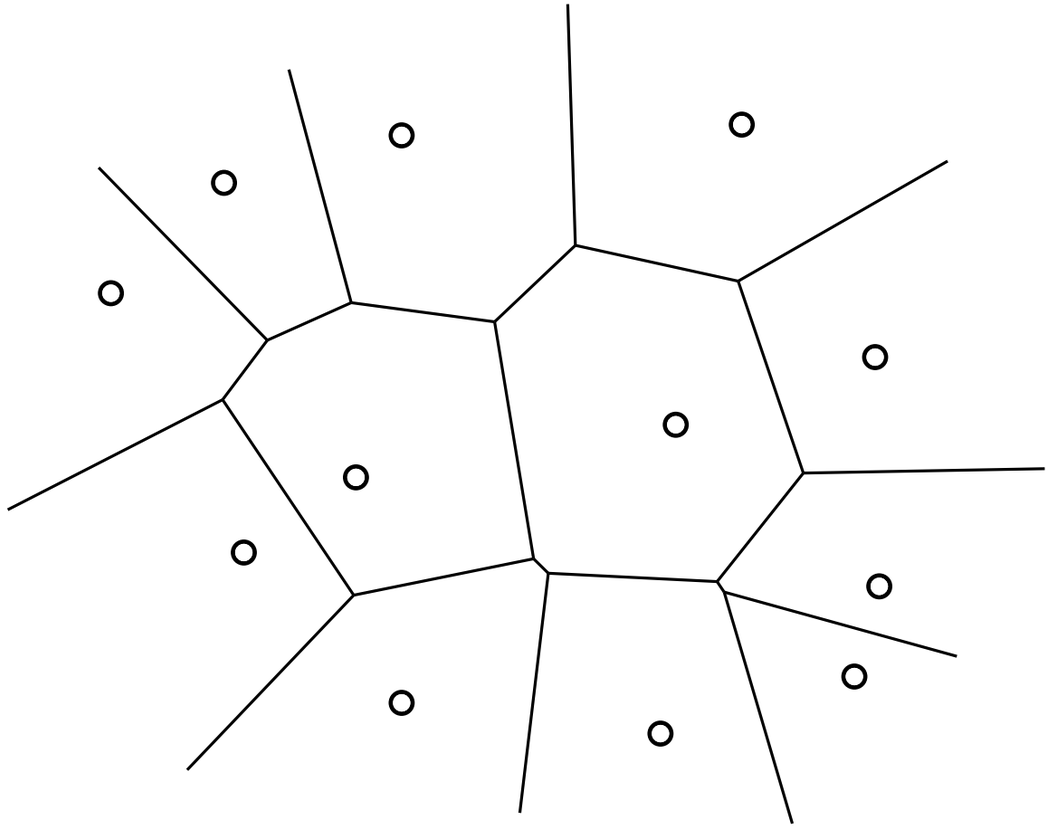
- Triangulating a polygon
- Motion planning
- Finding the closest post office
- and many more.

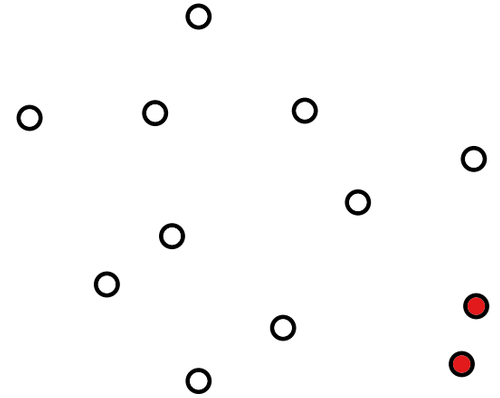


We offer an entire course on Computational Geometry in the winter term!

Closest Pair

Given: (multi-)set of points $P \subseteq \mathbb{R}^2$.

Task: Find a pair $(p, q) \in P^2$ of distinct elements whose Euclidean distance $\|p - q\|$ is minimum.



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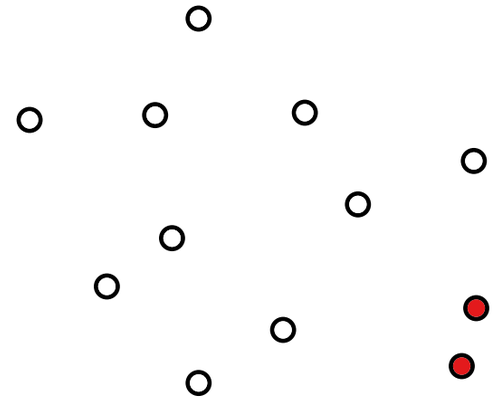
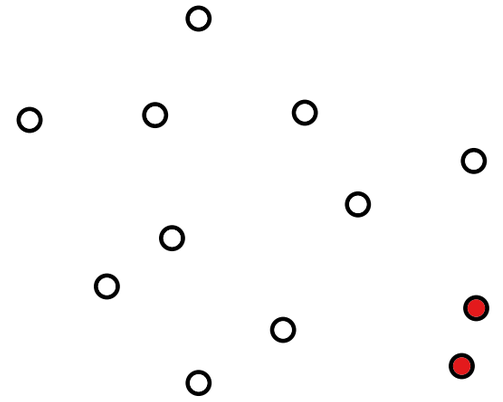


Deterministic algorithms:

Brute-force	$\mathcal{O}(n^2)$

Closest Pair



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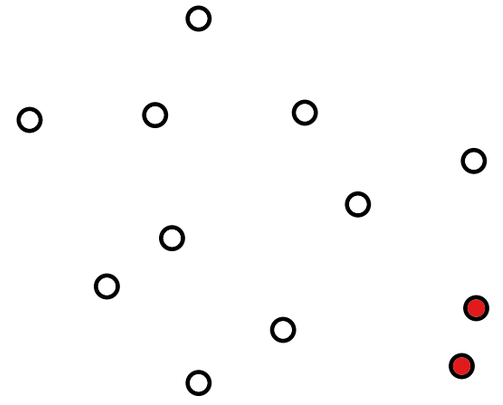

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Randomized algorithm:

Randomized incremental construction	$\mathcal{O}(n)$	(expected runtime)

later in this course!

A Randomized Incremental Algorithm for Closest Pair

Define $P_i = \{p_1, p_2, \ldots, p_i\}$ and let δ_i be the dist… losest pair in P_i.

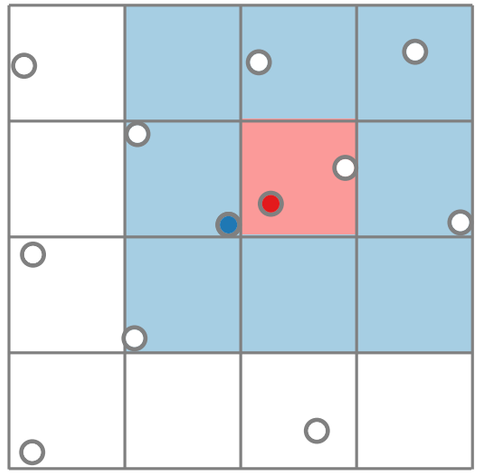
Idea: $\delta_2 = ||p_1, p_2||$. Compute $\delta_3, \delta_4, \ldots, \delta$… the points iteratively.

Suppose we have already determin…

Consider a square grid with c… $\delta_{i-1} \times \delta_{i-1}$.

Add the point p_i. If δ_i… p_i must be part of each closest pai…

Moreover, p… the cell of p_i or one of the adjacen…



(simple) exercise

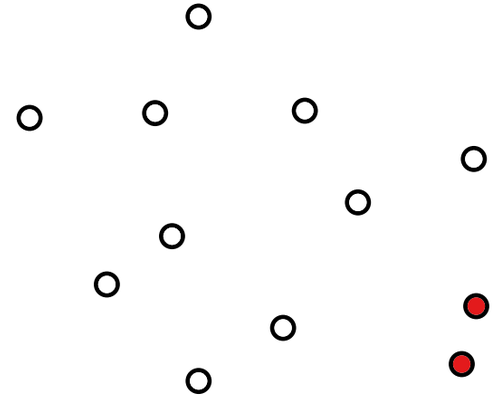
Ea… cells contains at most $\mathcal{O}(1)$ points of P_{i-1} ($\Leftarrow$ packing argument).

The co…inates of the cell of p_i can be determined in $\mathcal{O}(1)$ time assuming the floor function can be computed in $\mathcal{O}(1)$ time.

$\Rightarrow$ The test $\delta_i < \delta_{i-1}$ can be performed in $\mathcal{O}(1)$ time assuming P_{i-1} is stored in a suitable dictionary for the nonempty cells (implementable via dynamic perfect hashing).

Upcoming lecture on randomized algorithms

Closest Pair



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Task: Find a pair of distinct elements $p_a, p_b \in P$ such that the Euclidean distance $||p_a - p_b||$ is minimum.

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Brute-force	$\mathcal{O}(n^2)$	
Divide and conquer (recall from ADS)	$\mathcal{O}(n \log n)$	(optimal)

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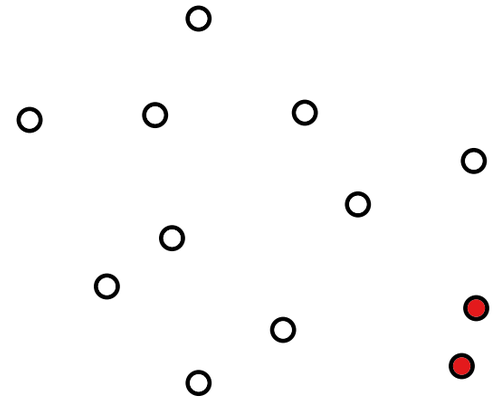

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Deterministic algorithms:

Brute-force	$\mathcal{O}(n^2)$		
Divide and conquer (recall from ADS)	$\mathcal{O}(n \log n)$	(optimal)	
Sweep line	$\mathcal{O}(n \log n)$	(optimal)	**now!**

Randomized algorithm:

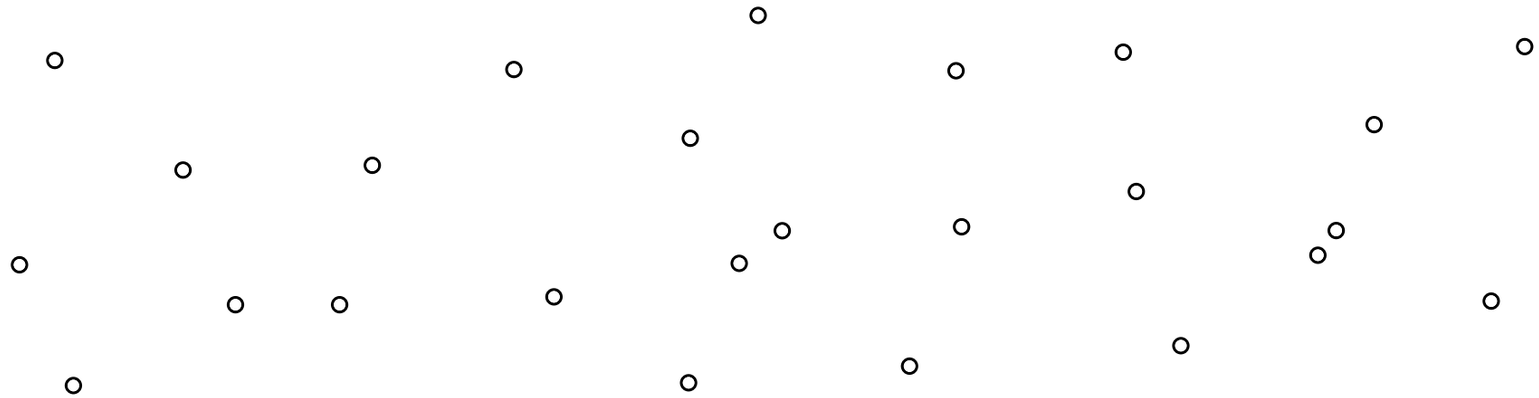
Randomized incremental construction	$\mathcal{O}(n)$	(expected runtime)	

later in this course!

A Sweep Line Approach for Closest Pair

Assumption: The points in P have pairwise distinct x-coordinates.

Idea: Sweep the plane from left to right with a vertical line ℓ (the **sweep line**).



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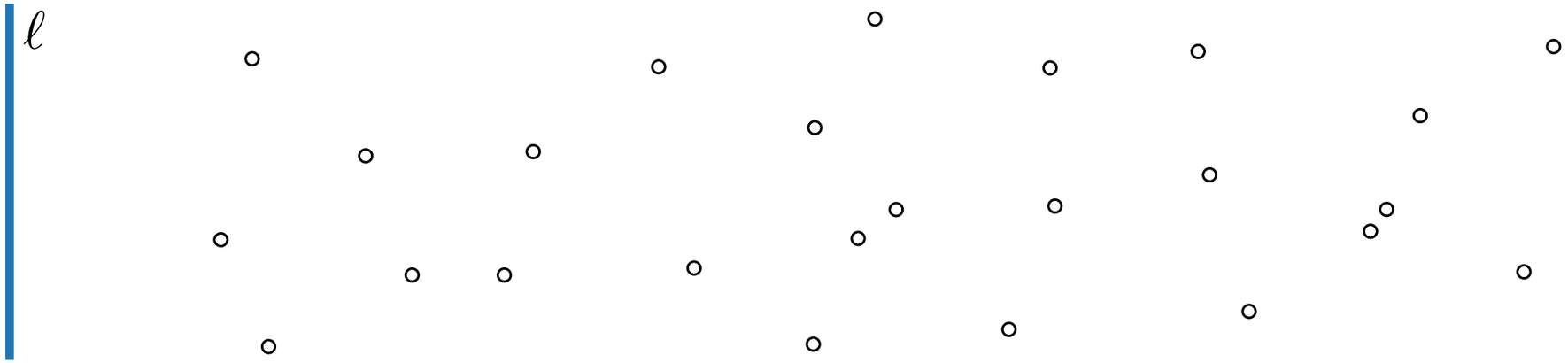

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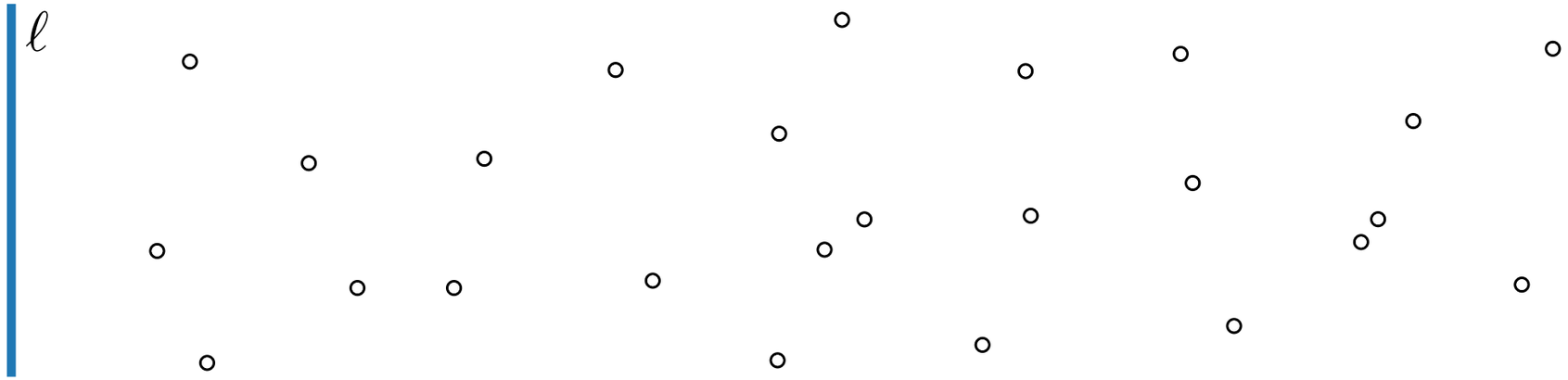

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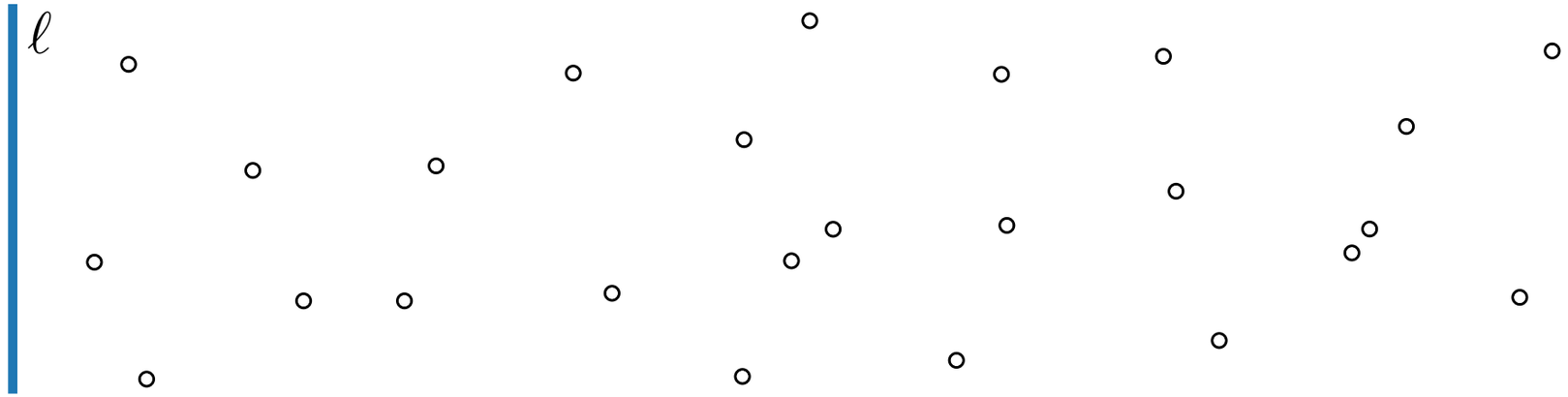

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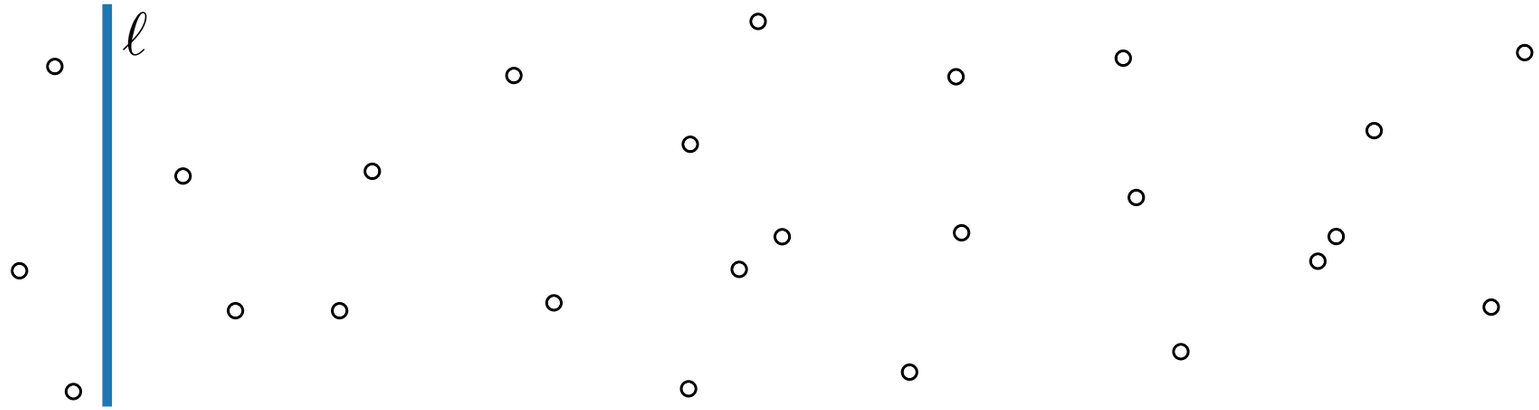

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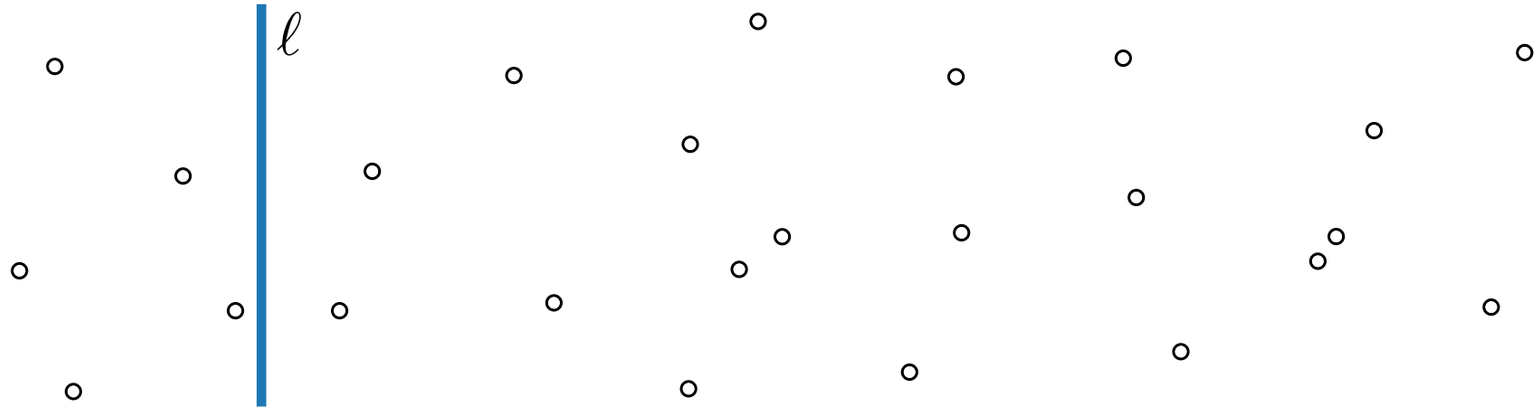

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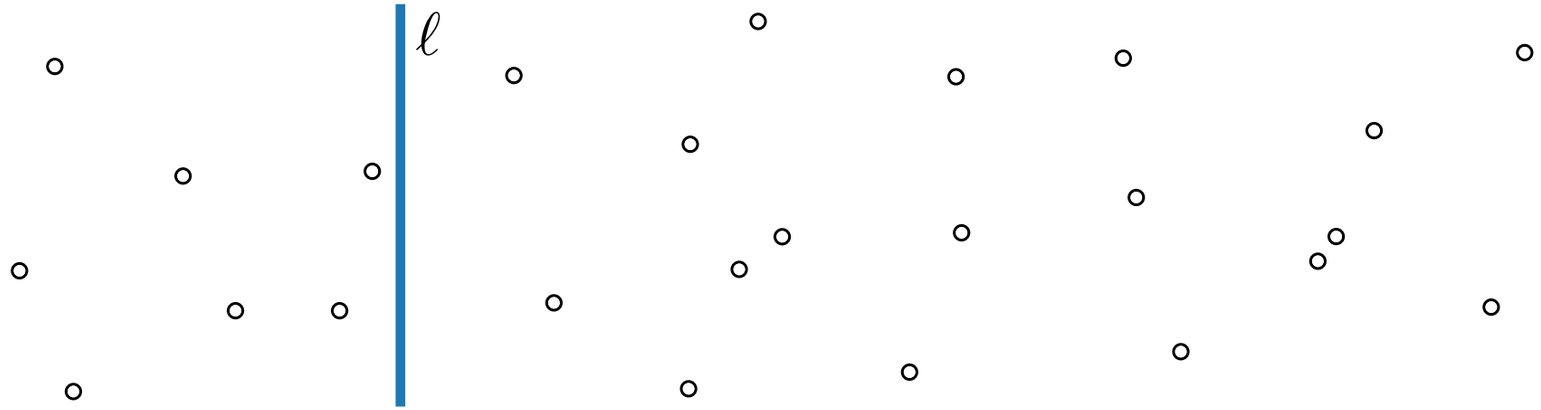

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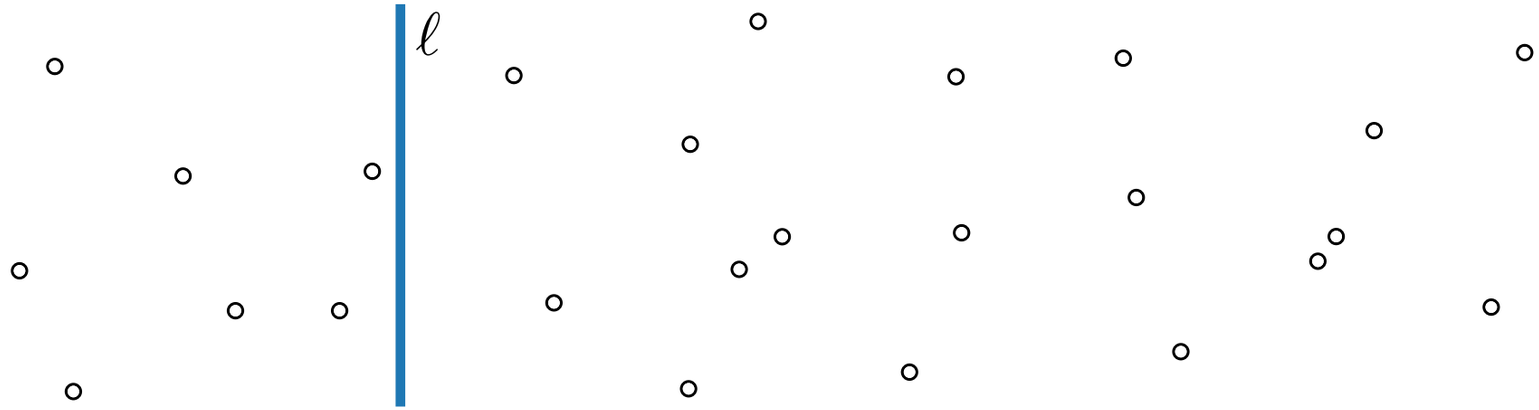

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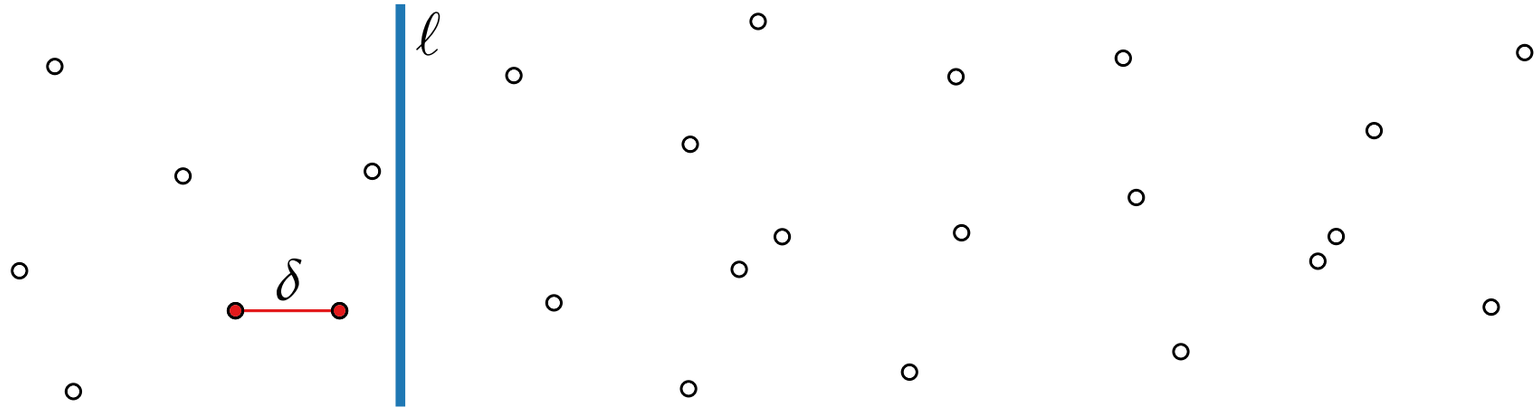

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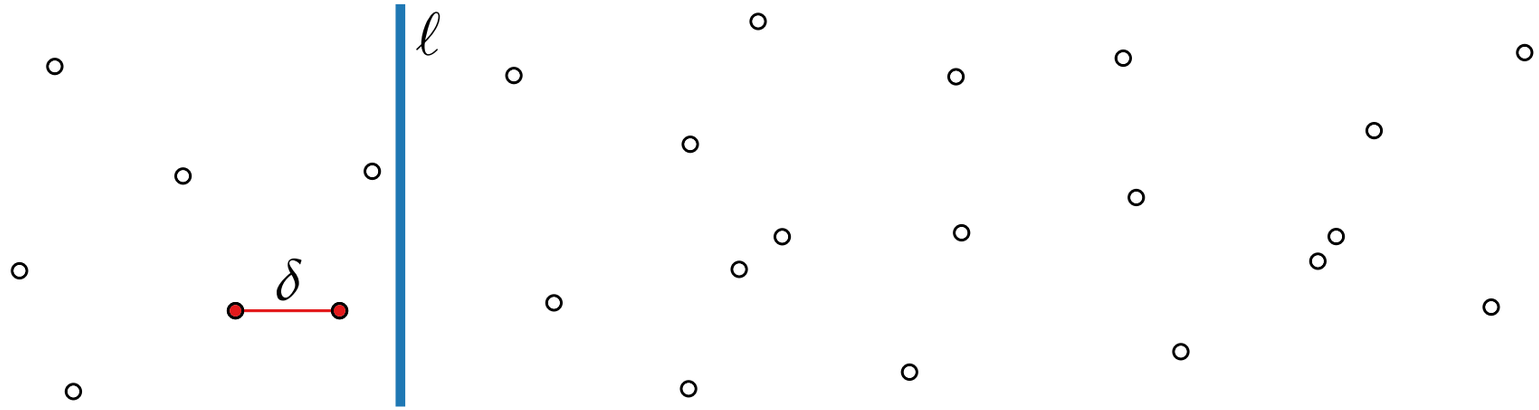

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Observations:

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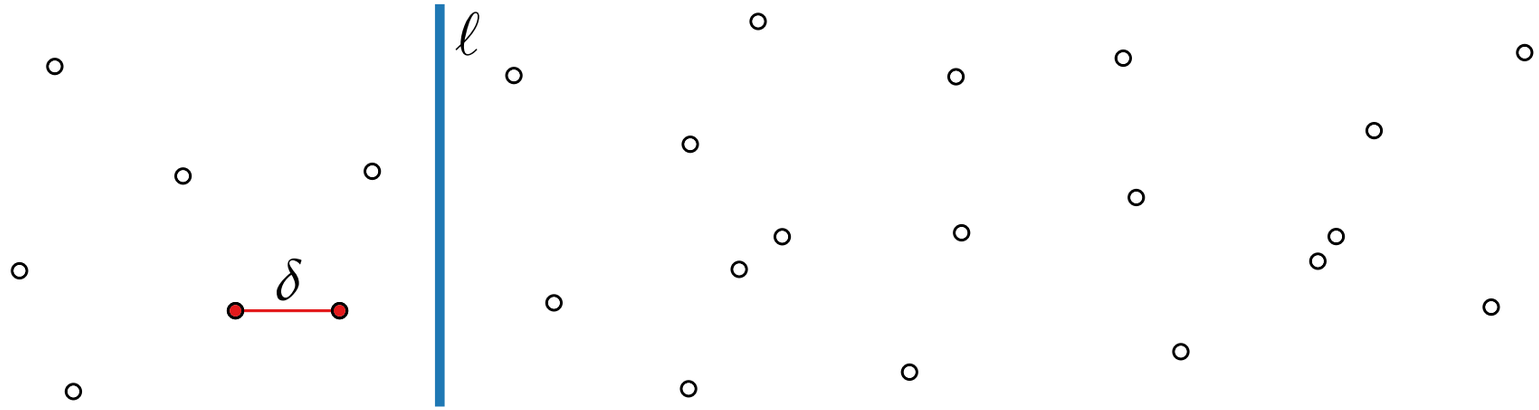

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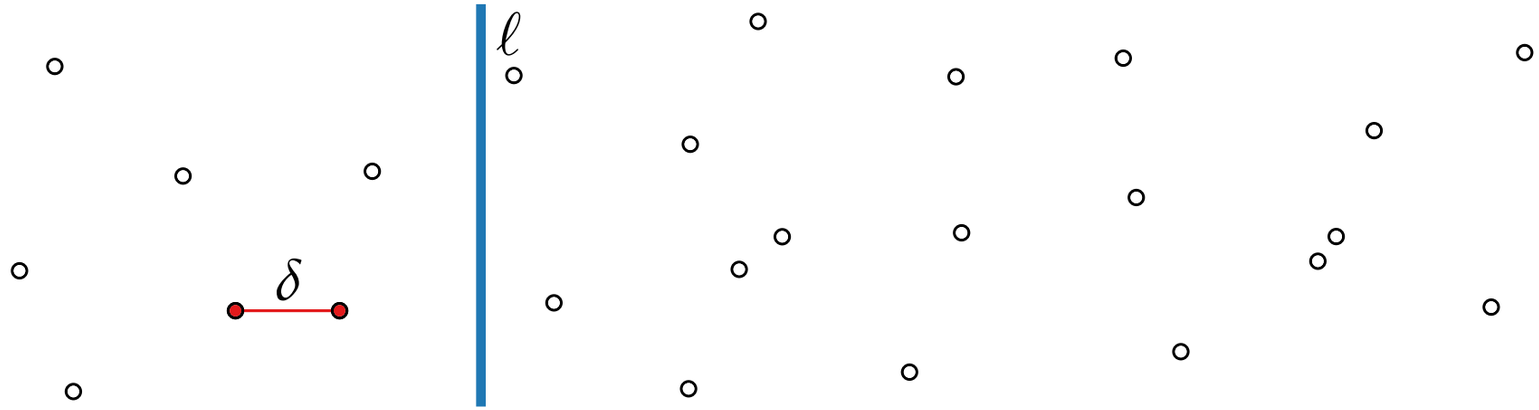

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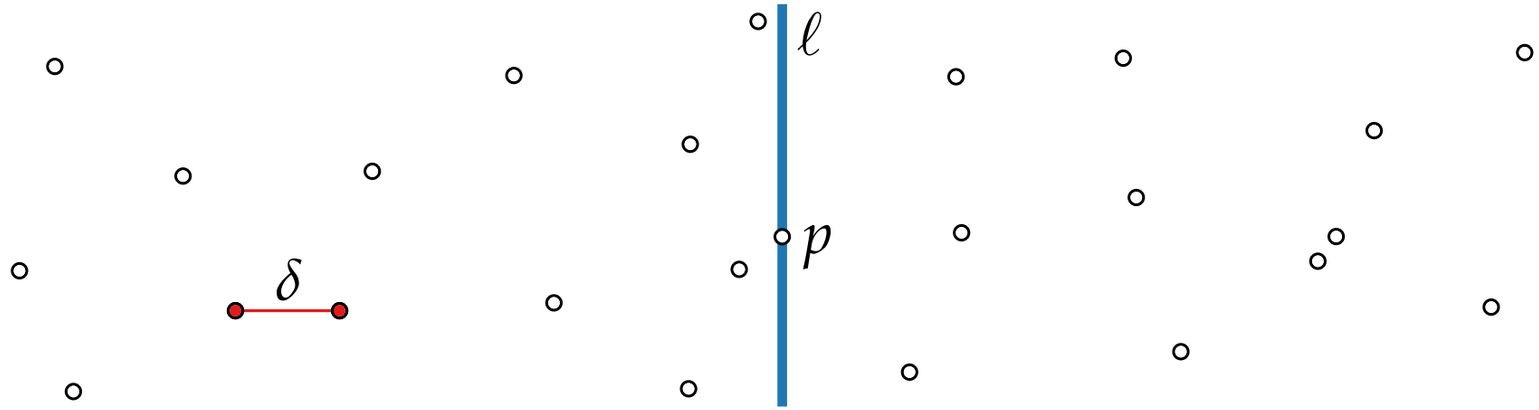

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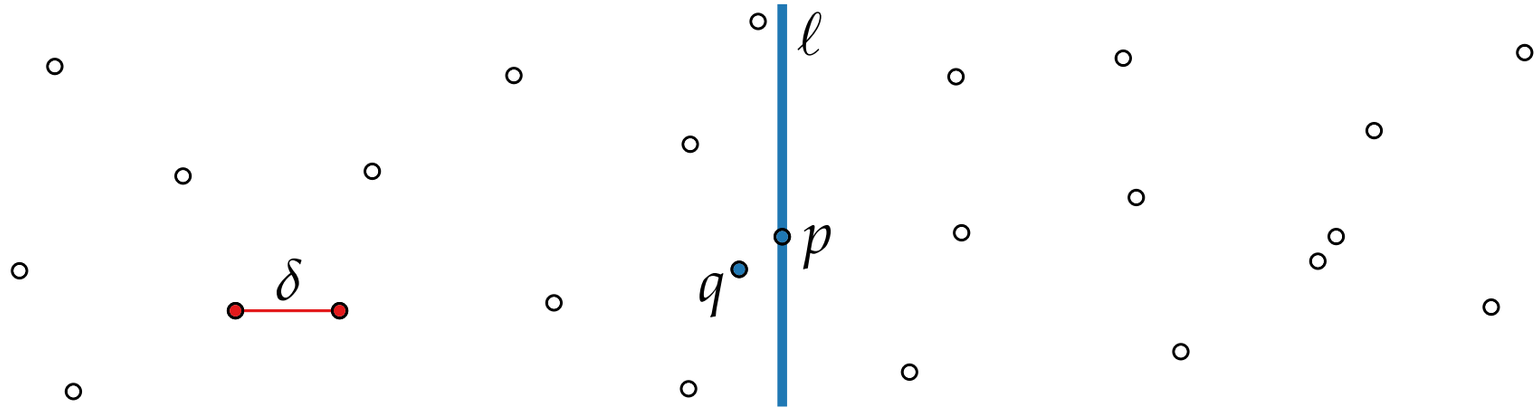

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- Each new closest pair consists of p and a point q

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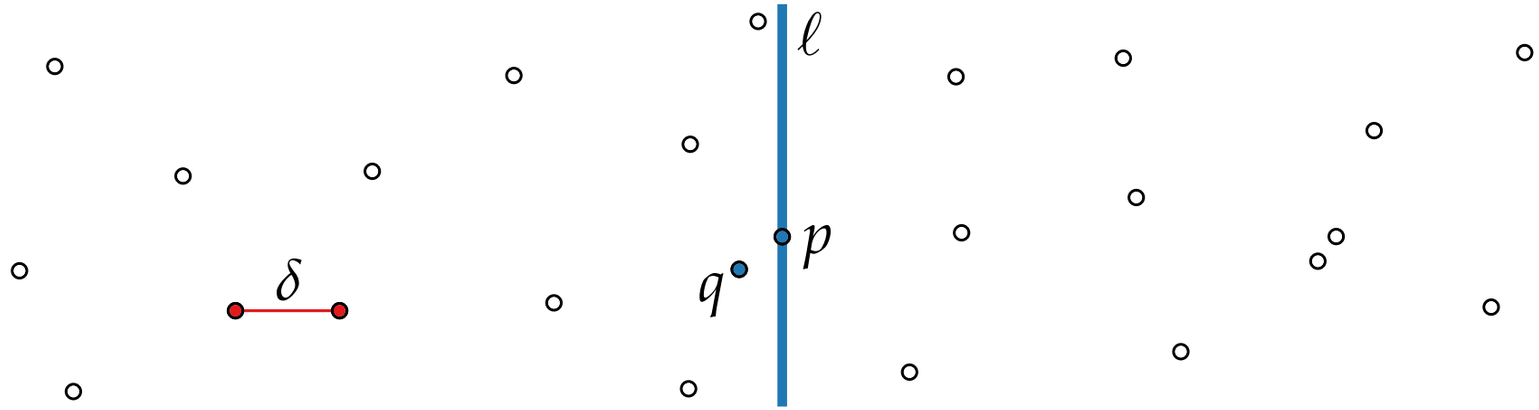

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What do we know about the location of q?

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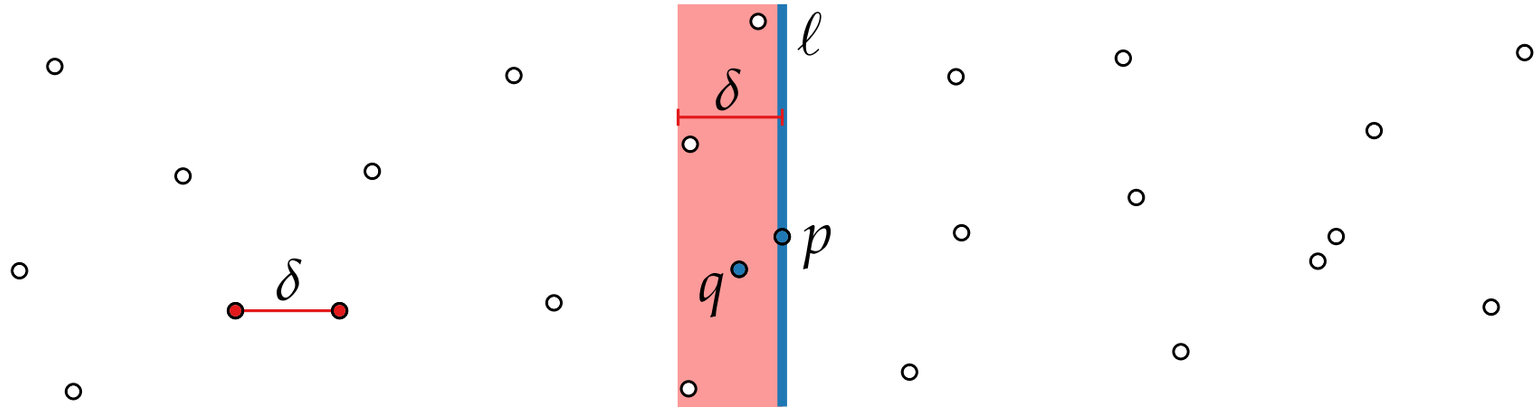

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Observations:

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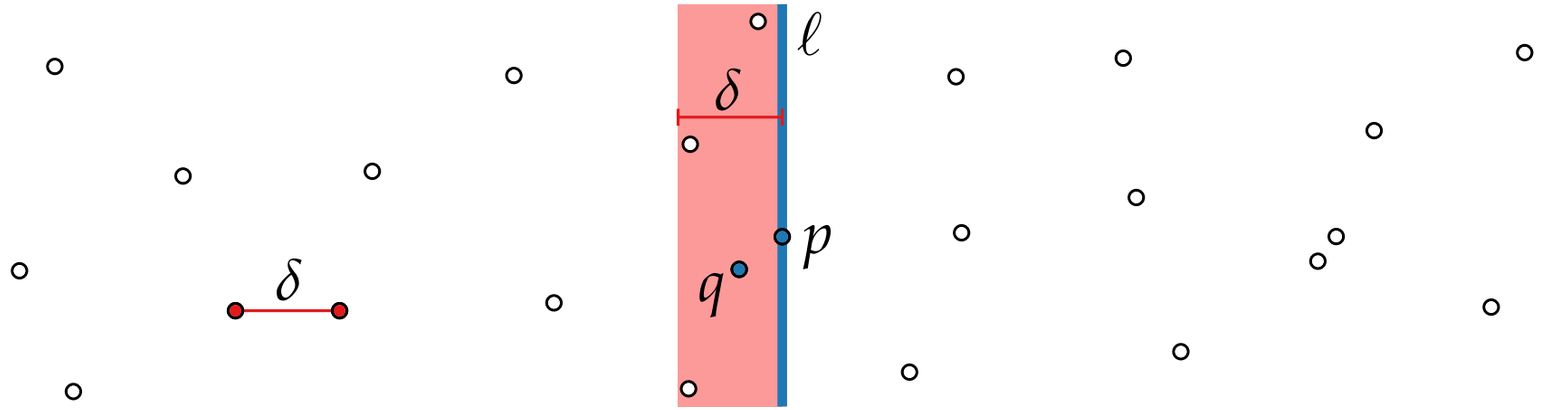

Invariant: a closest pair of the points to the left of ℓ and its distance δ is already known.

Observations:

- This partial solution can only change when ℓ sweeps a point p of P.
- Each new closest pair consists of p and a point q with distance $< \delta$ to ℓ.

How many points can be in this vertical slab?

A Sweep Line Approach for Closest Pair

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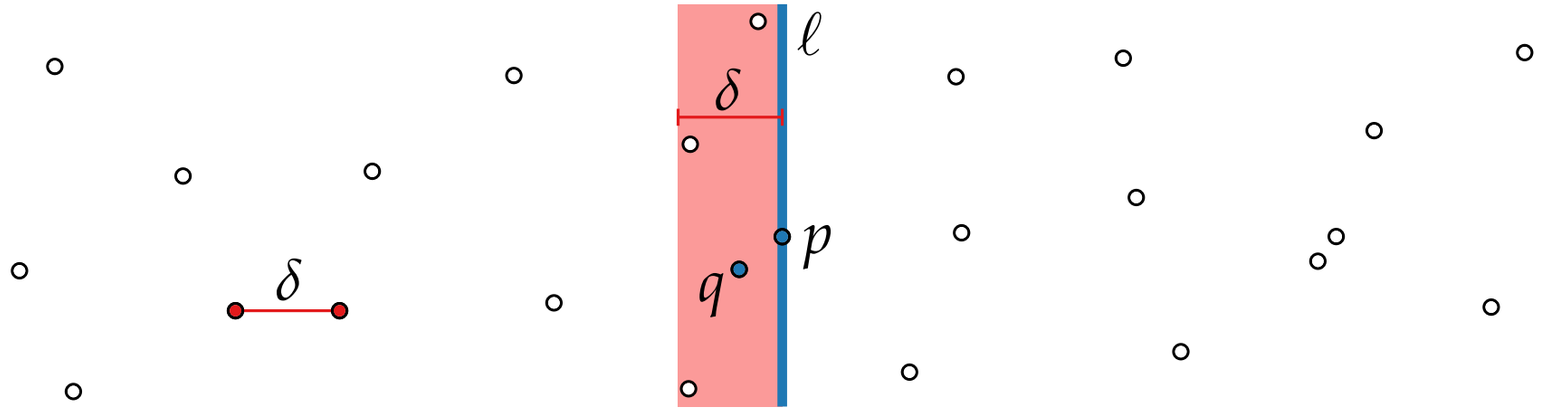

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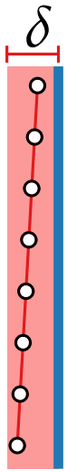
How many points can be in this vertical slab?

All of them!



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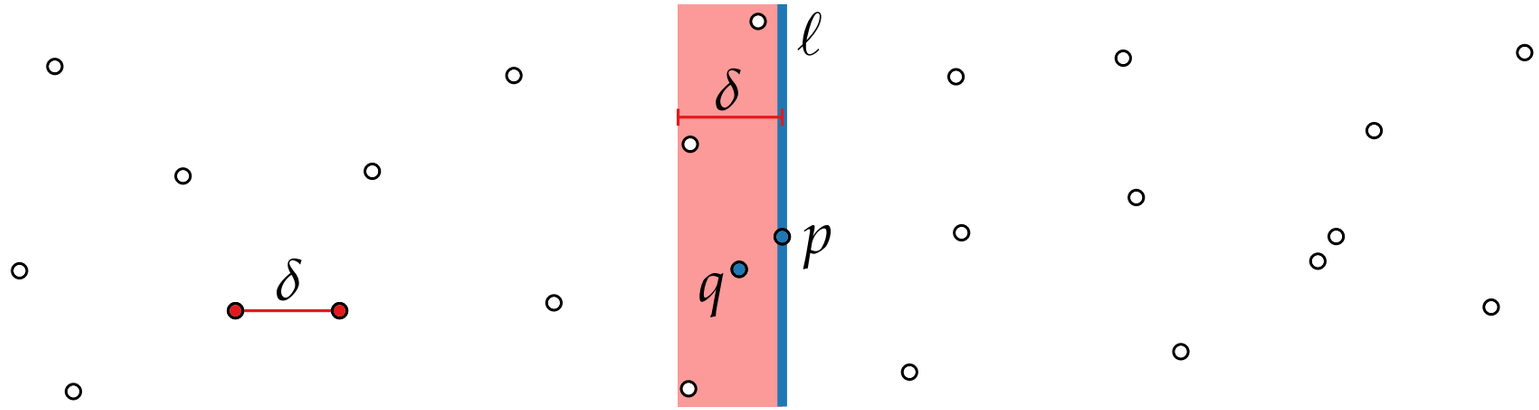

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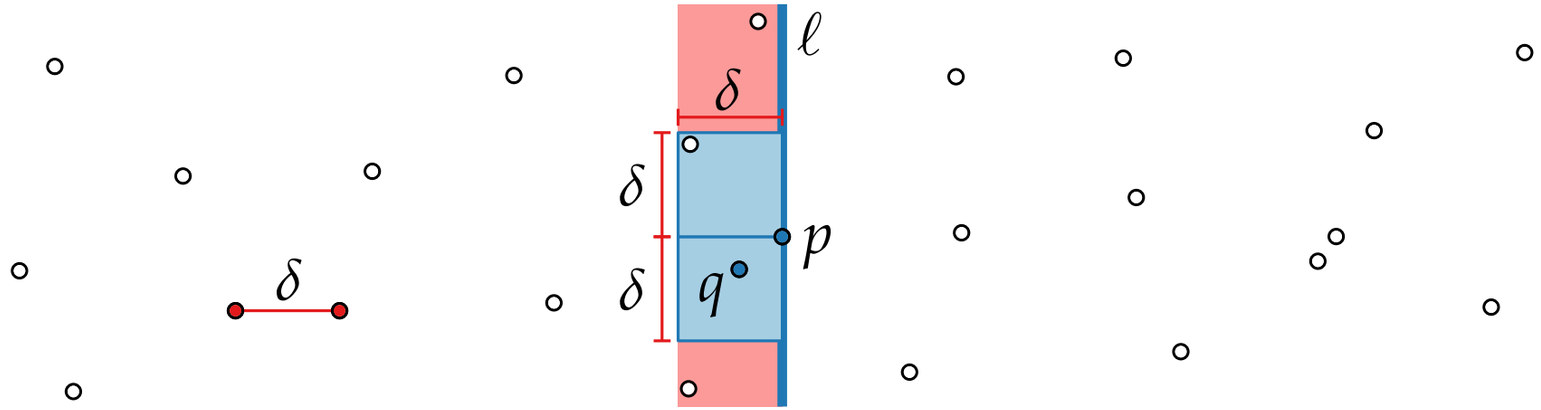

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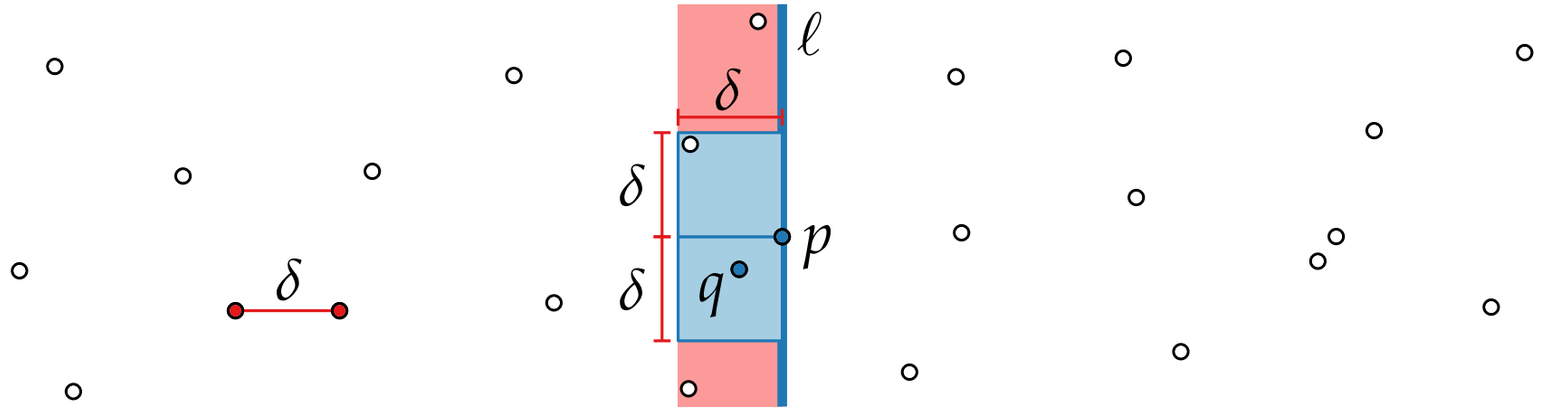

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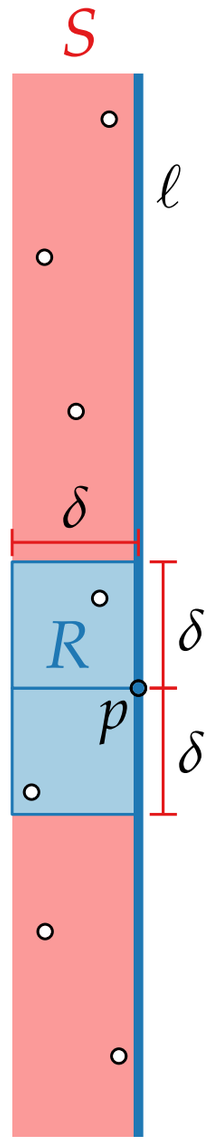
- Each new closest pair consists of p and a point q with distance $< \delta$ to ℓ.
- q needs to be located in a $\delta \times 2\delta$ rectangle R to the left of p.
- R contains $\mathcal{O}(1)$ points of $P \setminus \{p\}$ since their pairwise distance is $\geq \delta$. $\left(\begin{smallmatrix}\text{packing}\\ \text{argument}\end{smallmatrix}\right)$

Computing the Points in R Efficiently

Let S denote the vertical slab of width δ to the left of ℓ.



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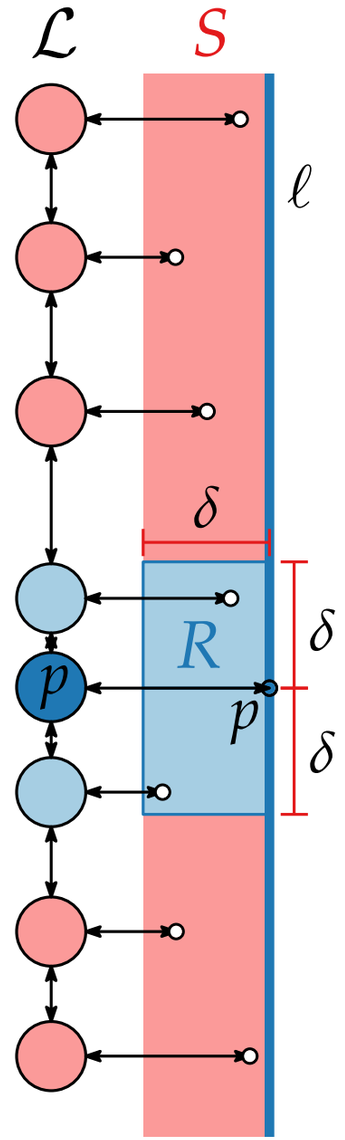
Assume that the points $P \cap S$ are stored in a **linked list** $\mathcal{L}$ sorted according to their y-coordinates.



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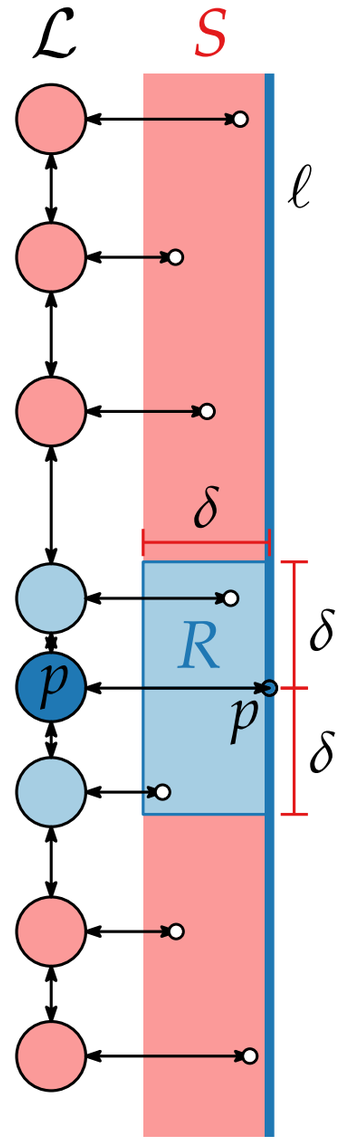

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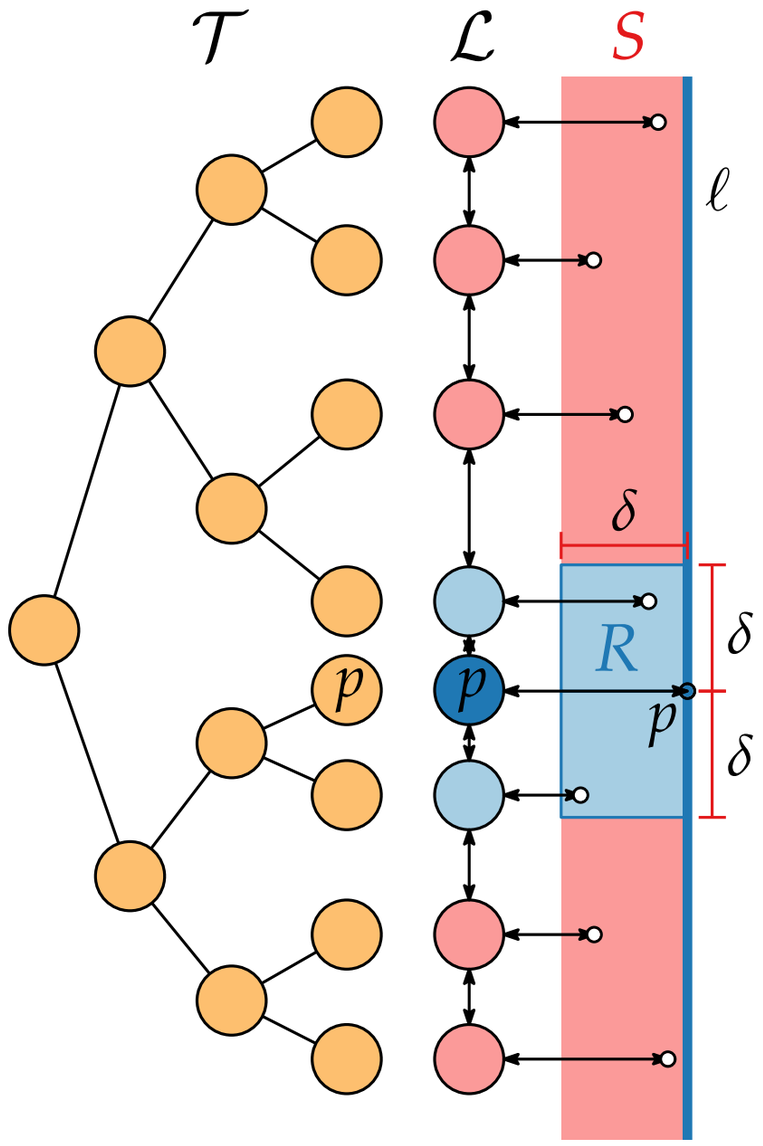

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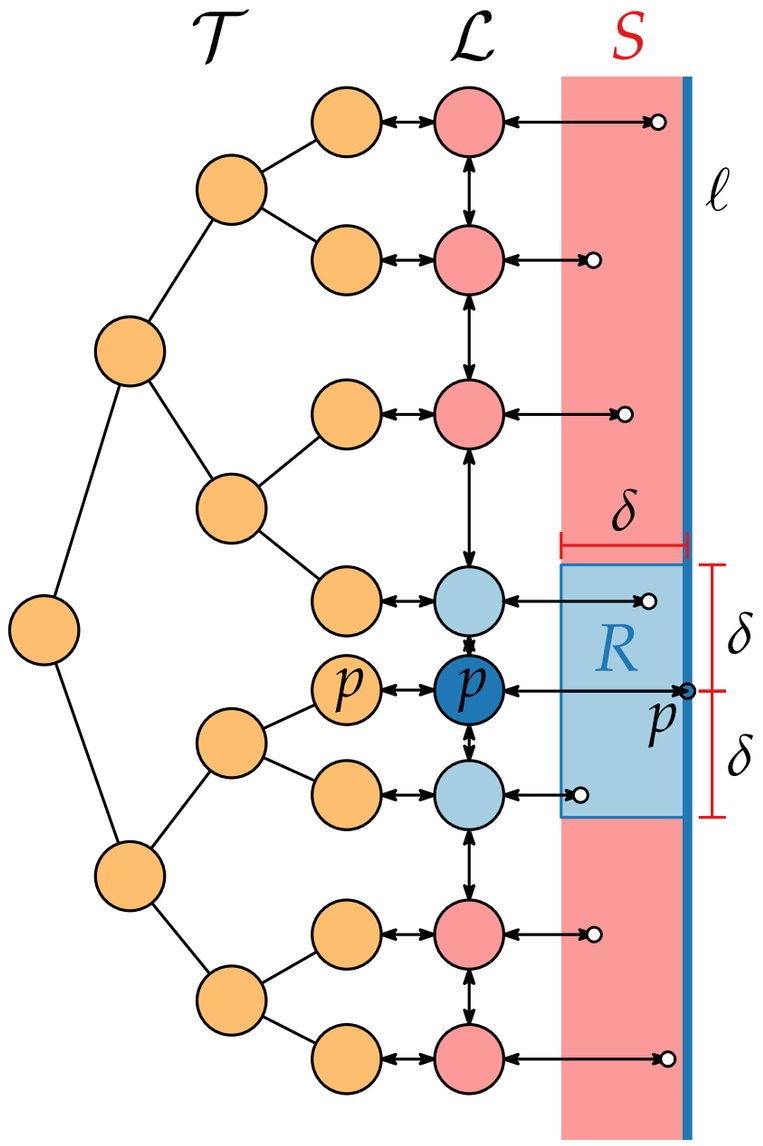

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The corresponding elements in $\mathcal{L}$ and $\mathcal{T}$ are linked.

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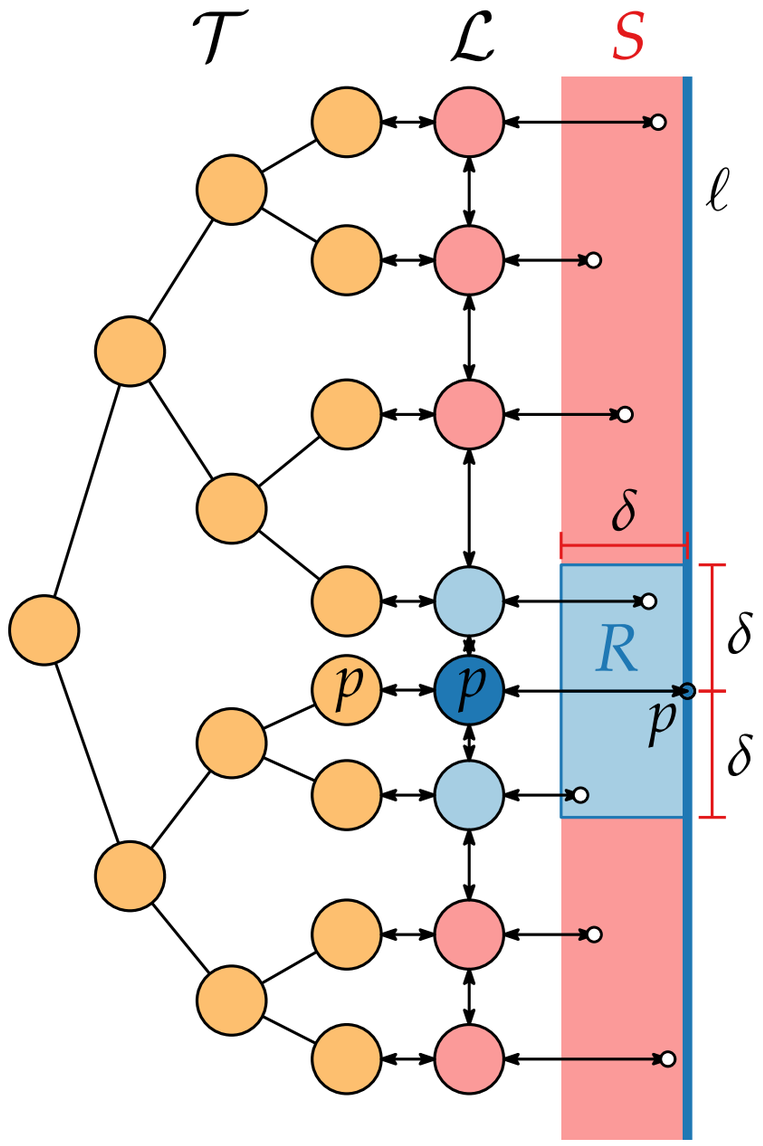

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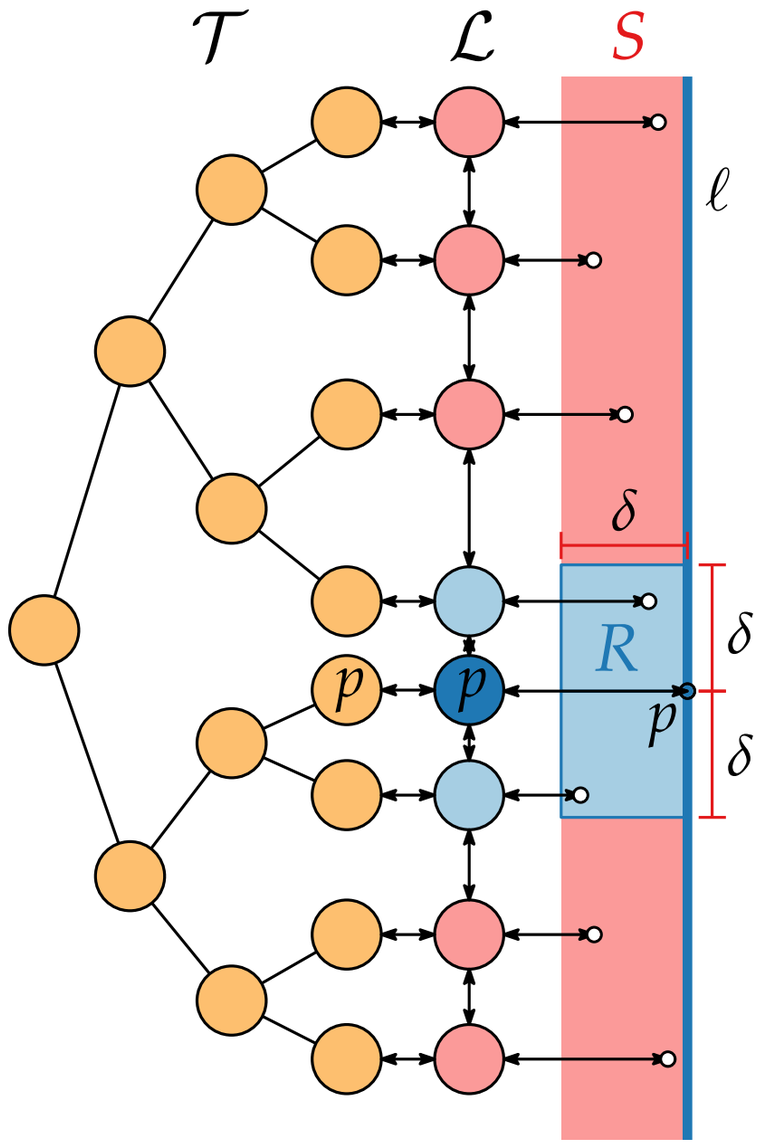

Invariant 2: When we reach a point p, $\mathcal{T}$ and $\mathcal{L}$ contain exactly the points in $P \cap S$.

Algorithm

$p_1, p_2, \ldots, p_n =$ points of P sorted according to their x-coordinates
$P_{\text{min}} =$ nil // current closest pair
$\delta = \infty$ // distance of current closest pair
$k = 1$ // index of the left-most point in $\mathcal{L}$ and $\mathcal{T}$
initialize $\mathcal{L}$ and $\mathcal{T}$ with p_1
for $i = 2, 3, \ldots, n$ **do**
    insert p_i into $\mathcal{L}$ and $\mathcal{T}$
    for $p_j \in [\text{y}(p_i) - \delta,\ \text{y}(p_i) + \delta] \setminus \{p_i\}$ **do**
        if $||p_j - p_i|| < \delta$ **do**
            $P_{\text{min}} = \{p_j, p_i\}; \quad \delta = ||p_j - p_i||$
    while $\text{x}(p_k) < \text{x}(p_{i+1}) - \delta$ **do**
        delete p_k from $\mathcal{L}$ and $\mathcal{T}$
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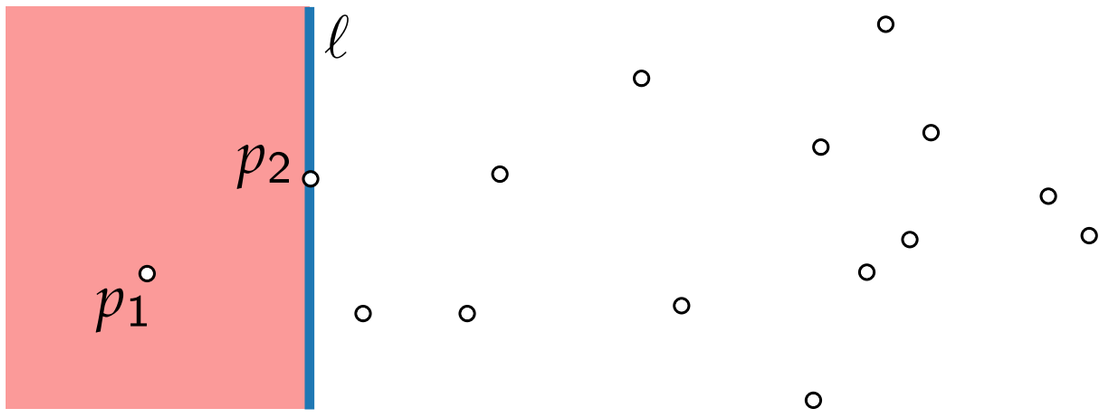

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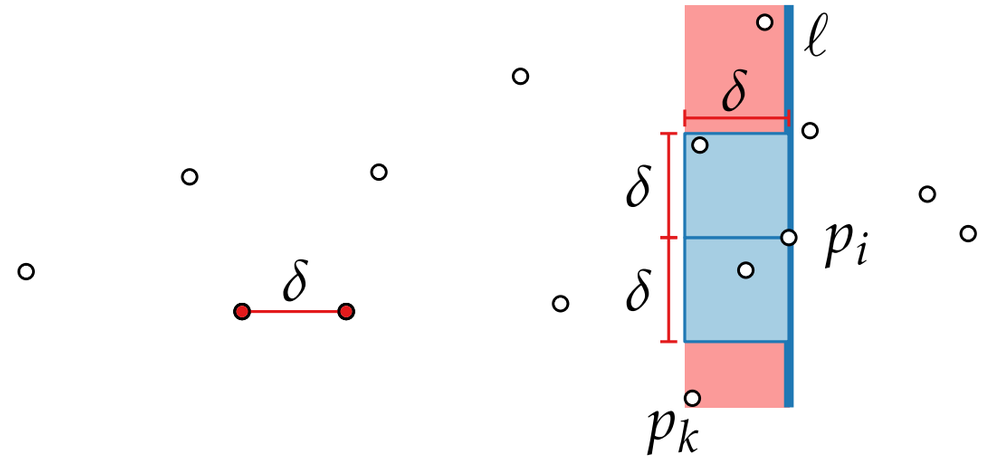

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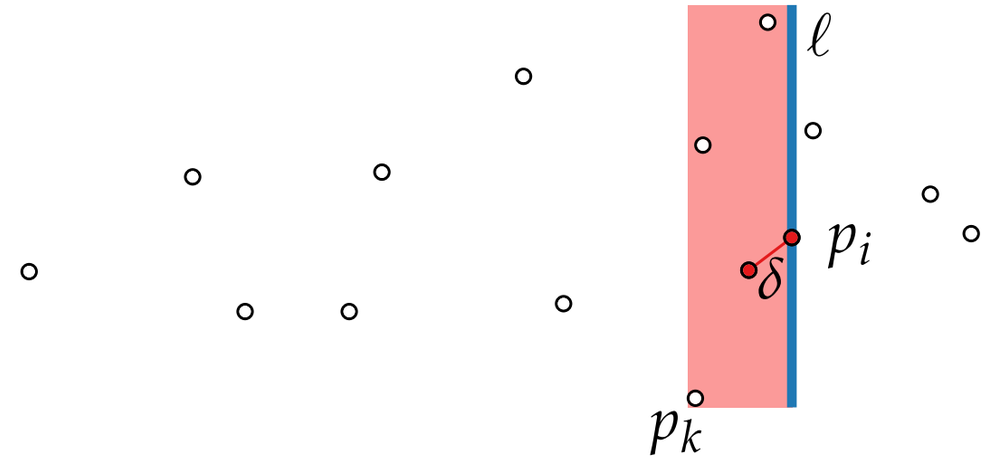

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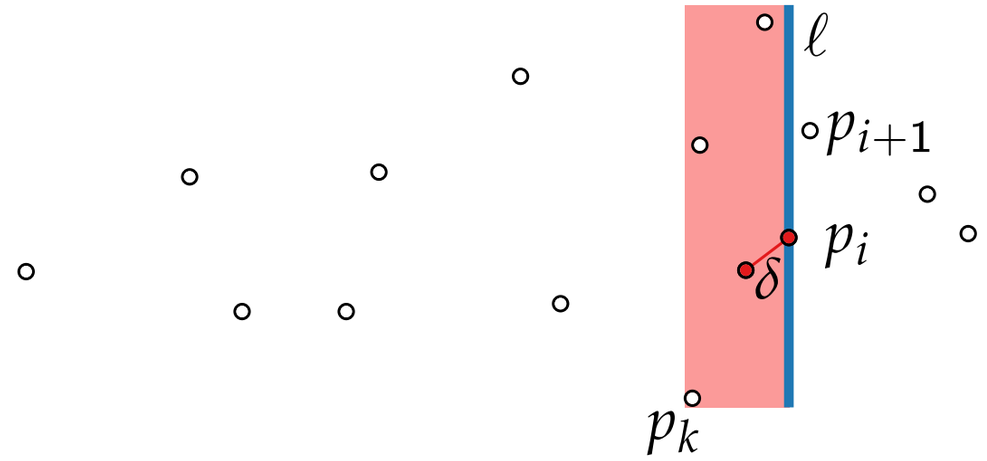

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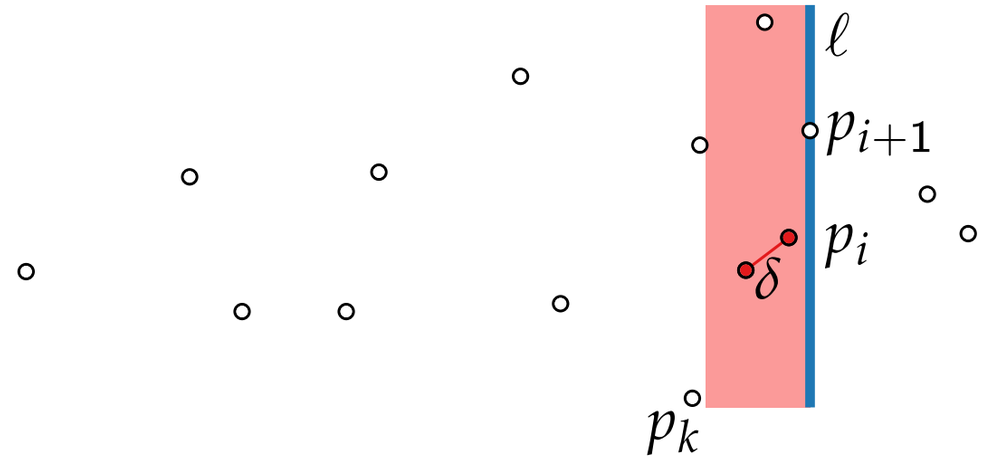

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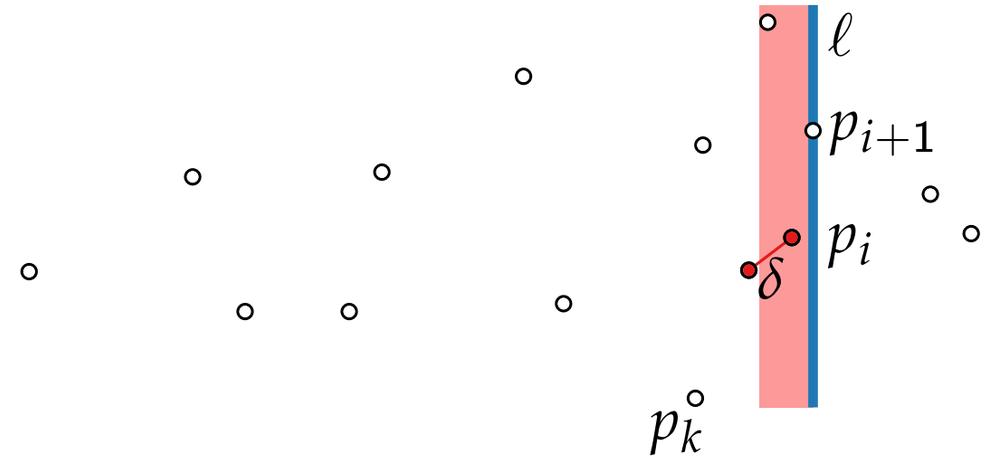

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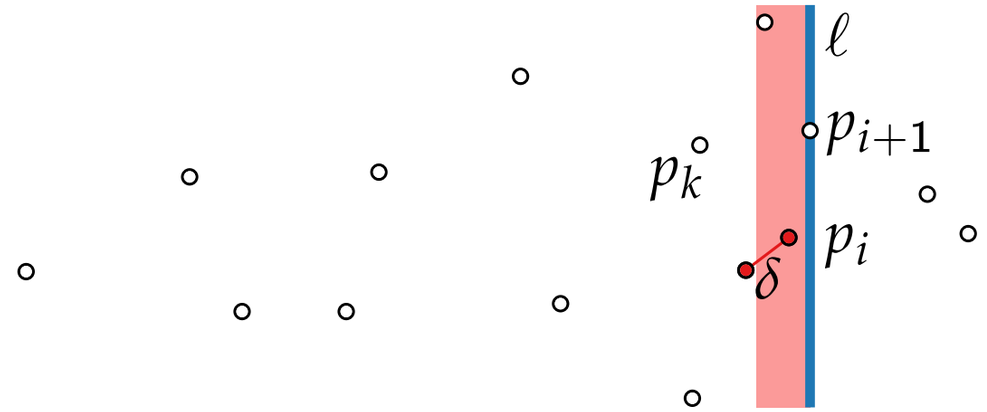

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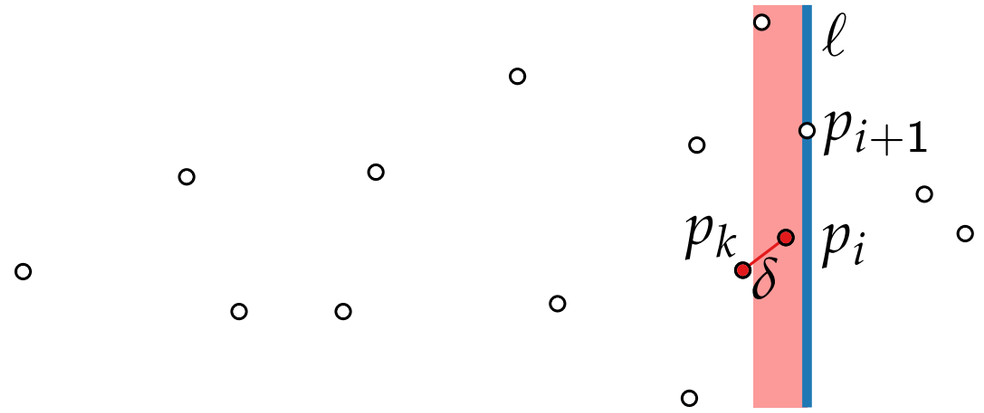

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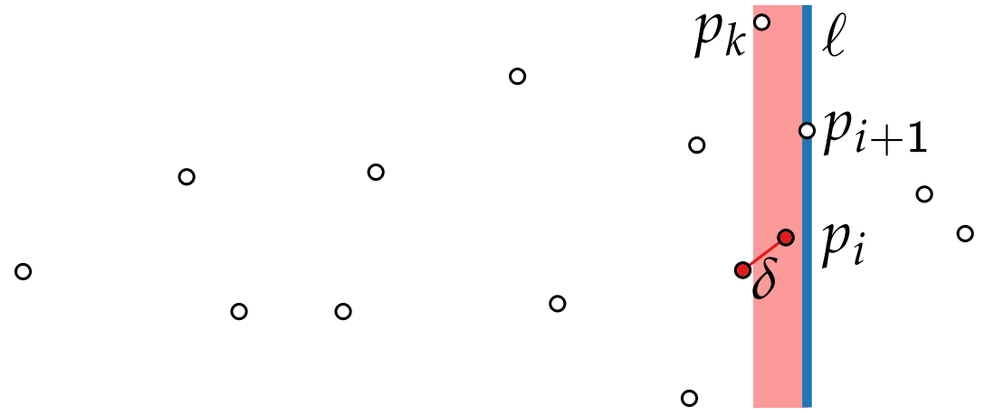

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$\Rightarrow$ **Total runtime:** $\mathcal{O}(n \log n)$

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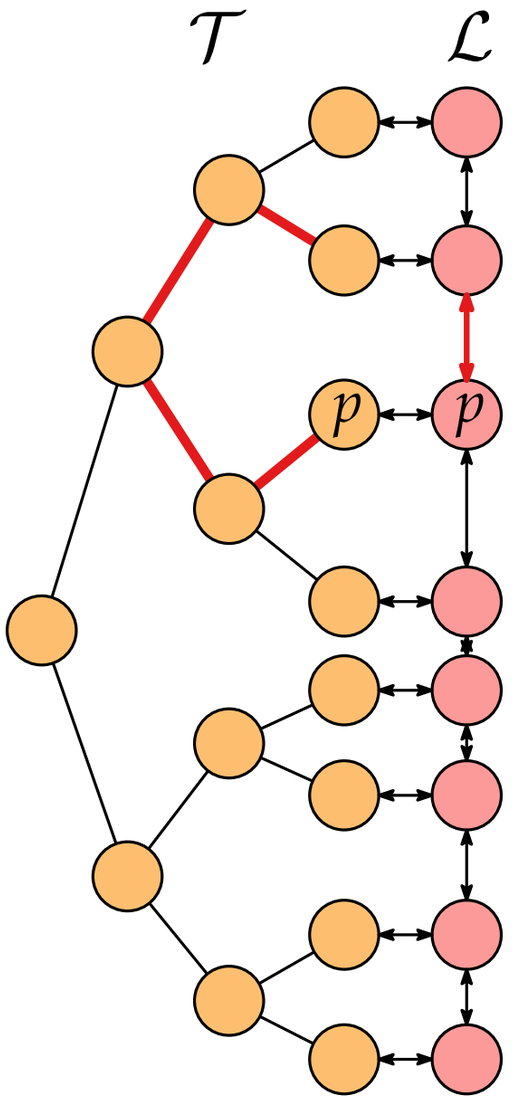

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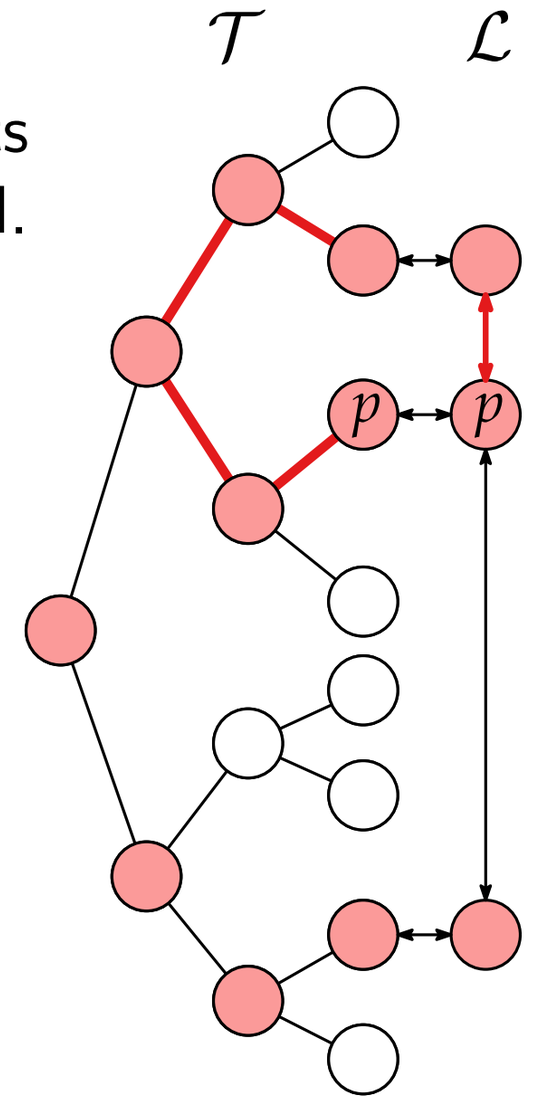

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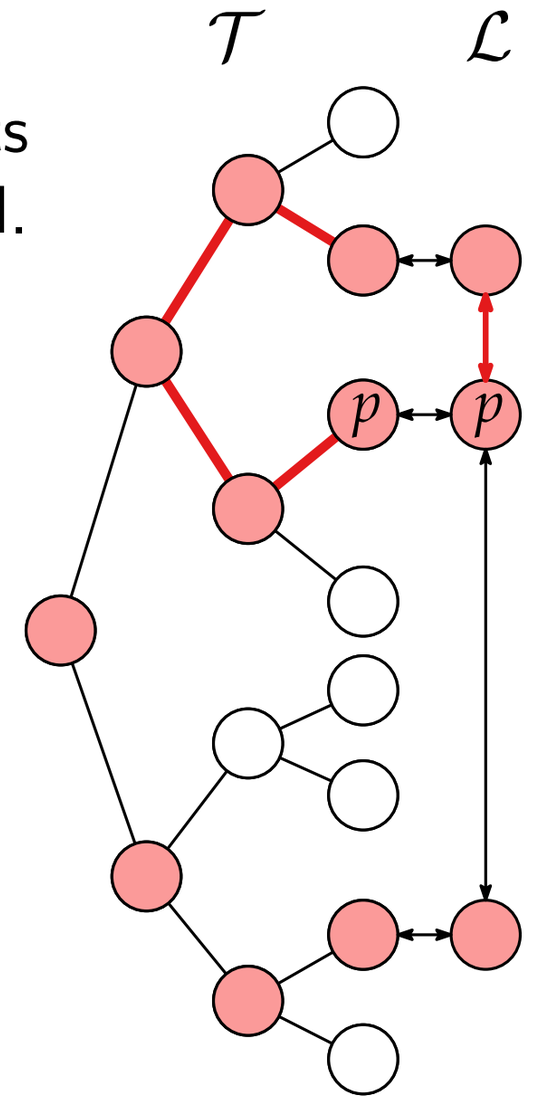

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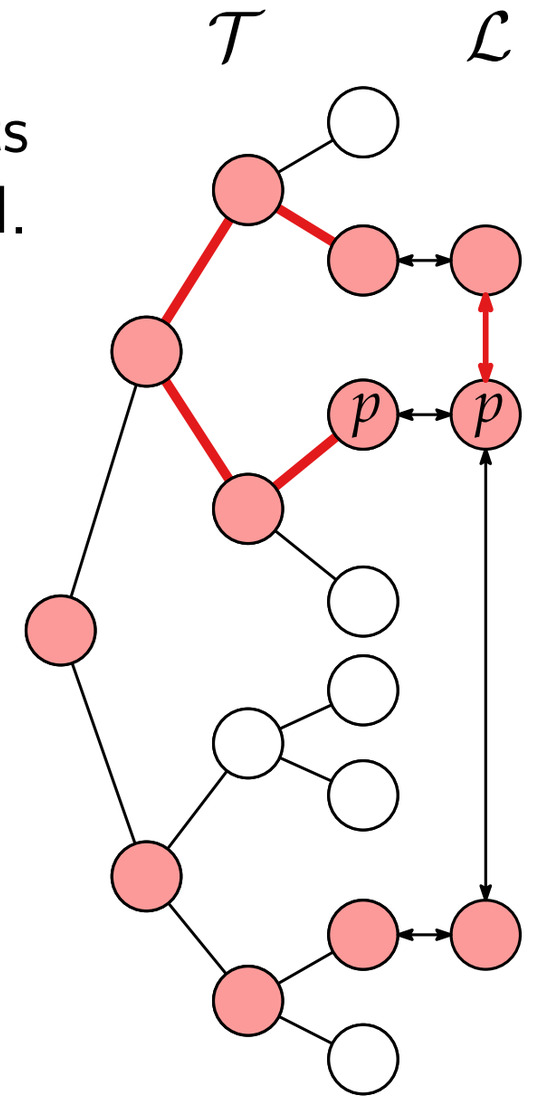
- We assumed that the points in P have pairwise distinct x-coordinates. This situation can be established by rotating P or tilting ℓ slightly.
  Simply visit the points in lexicographical order!



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Main idea: Sweep the plane with a line ℓ while maintaining two invariants:

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The sweep line paradigm is **powerful** and leads to **simple** algorithms for many problems:
computing Voronoi diagrams, crossings in an arrangement of line segments, intersection/union of two polygons, decompositions of polygons, certain triangulations, visibility polygons, ...

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Given: A polygon P with n corners and a point p in its interior.

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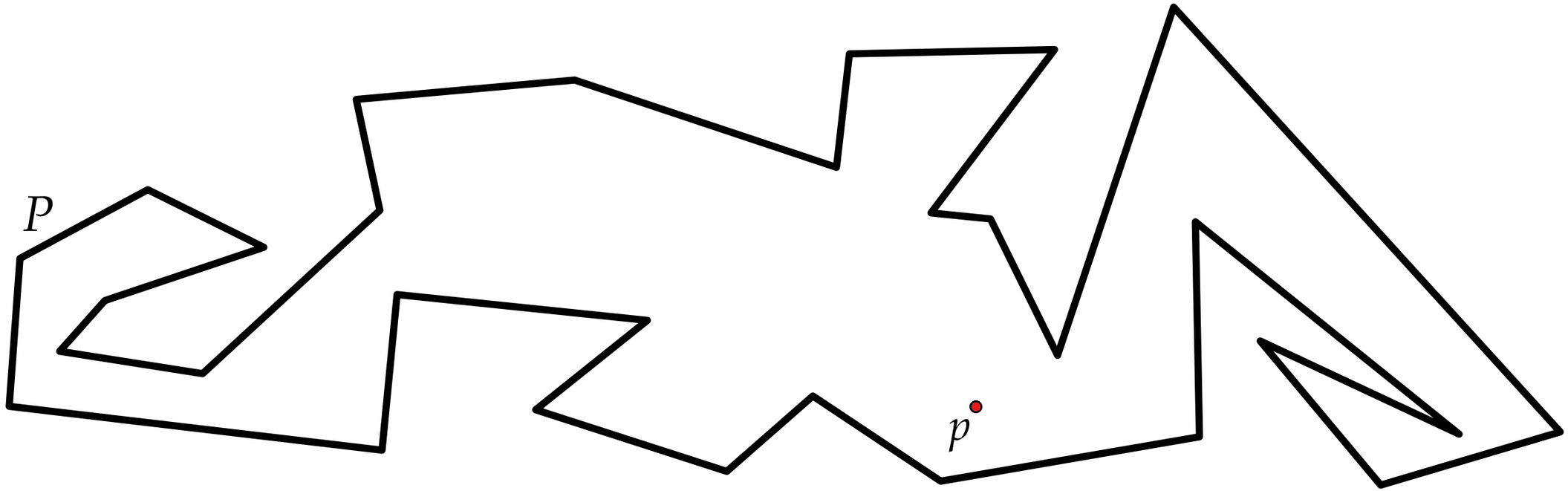

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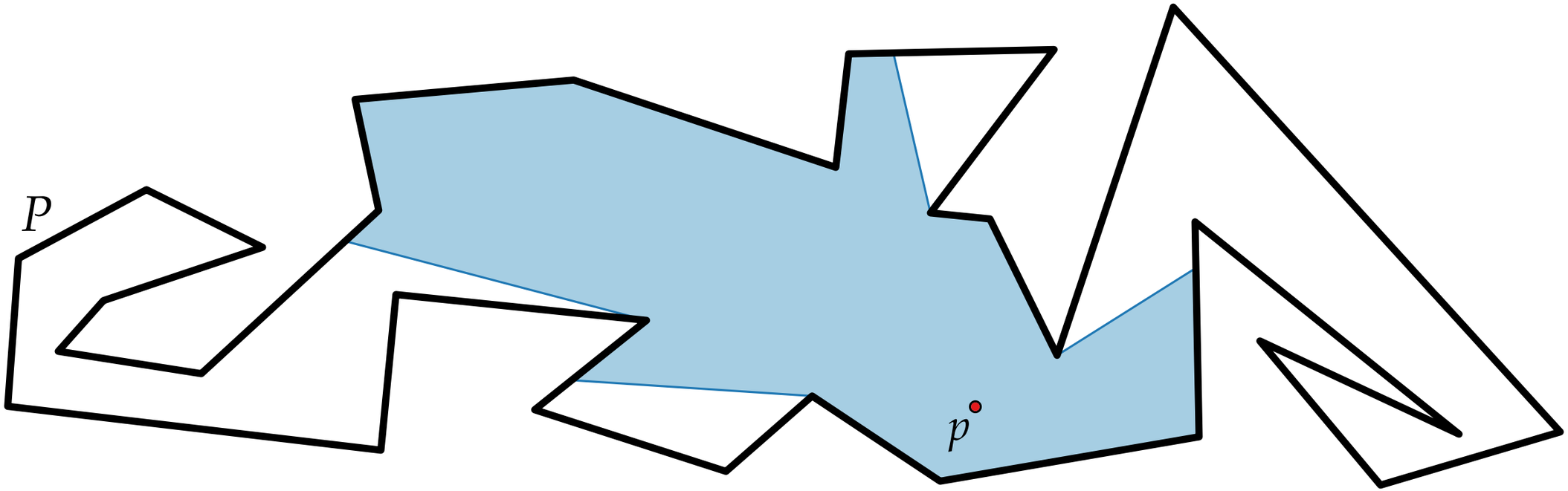

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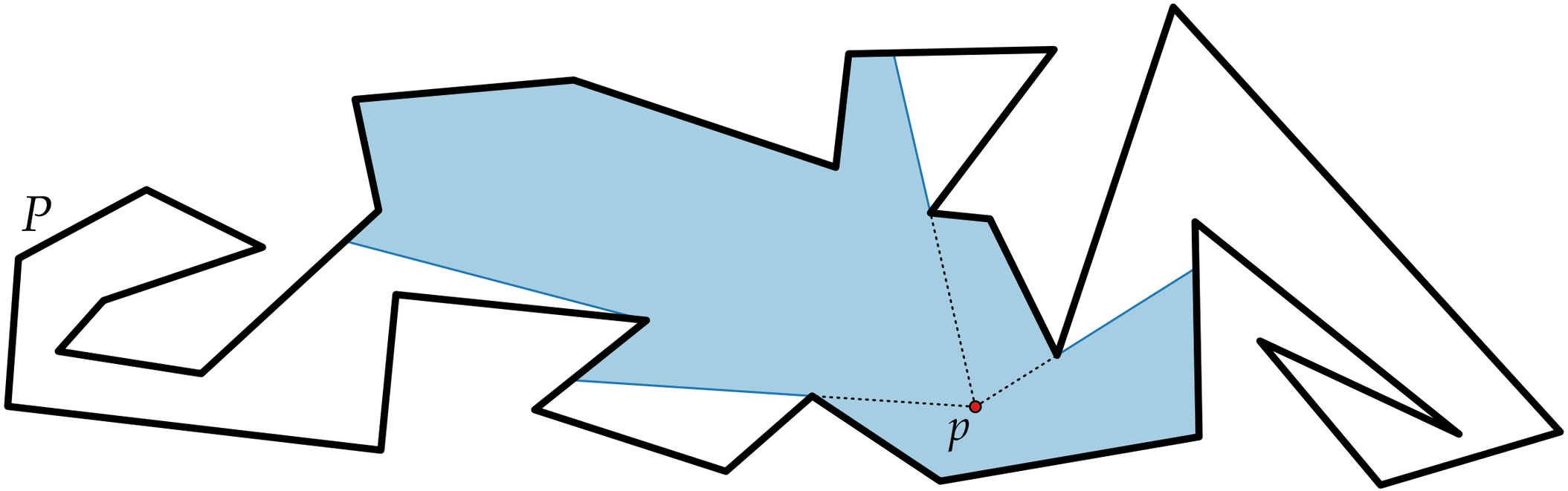

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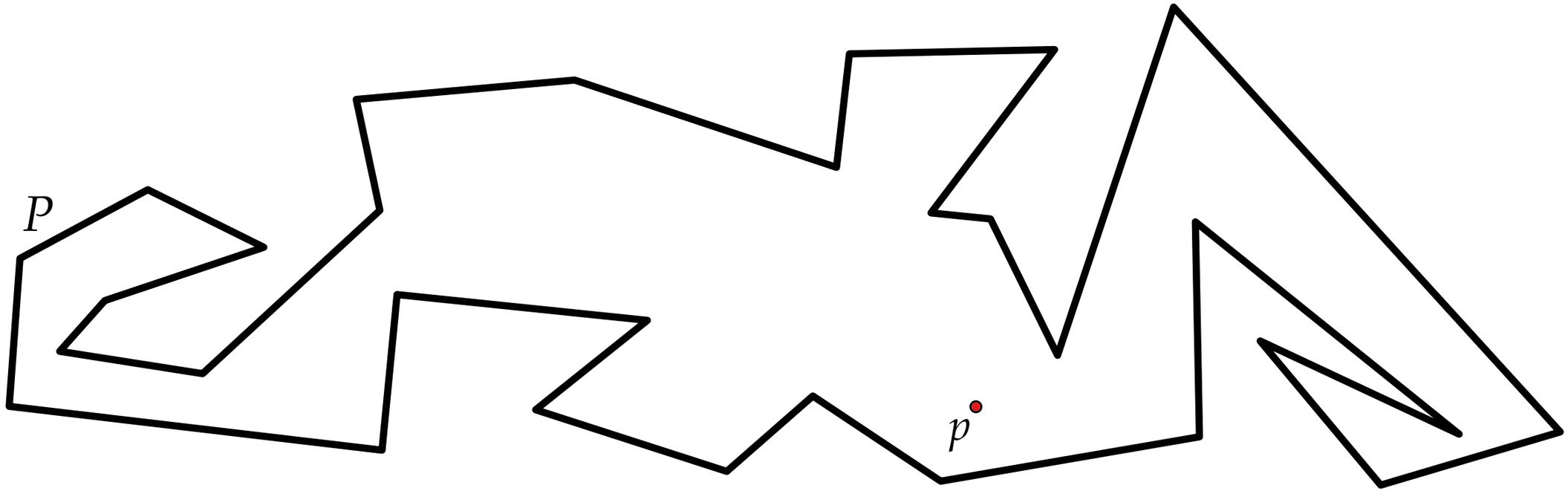

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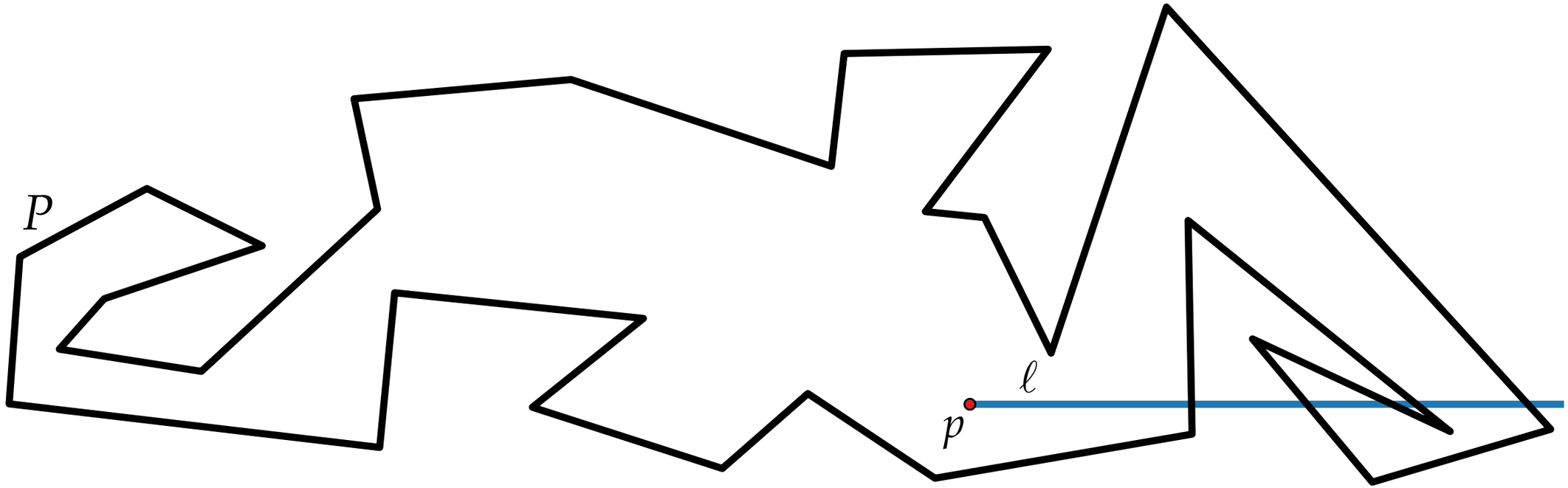

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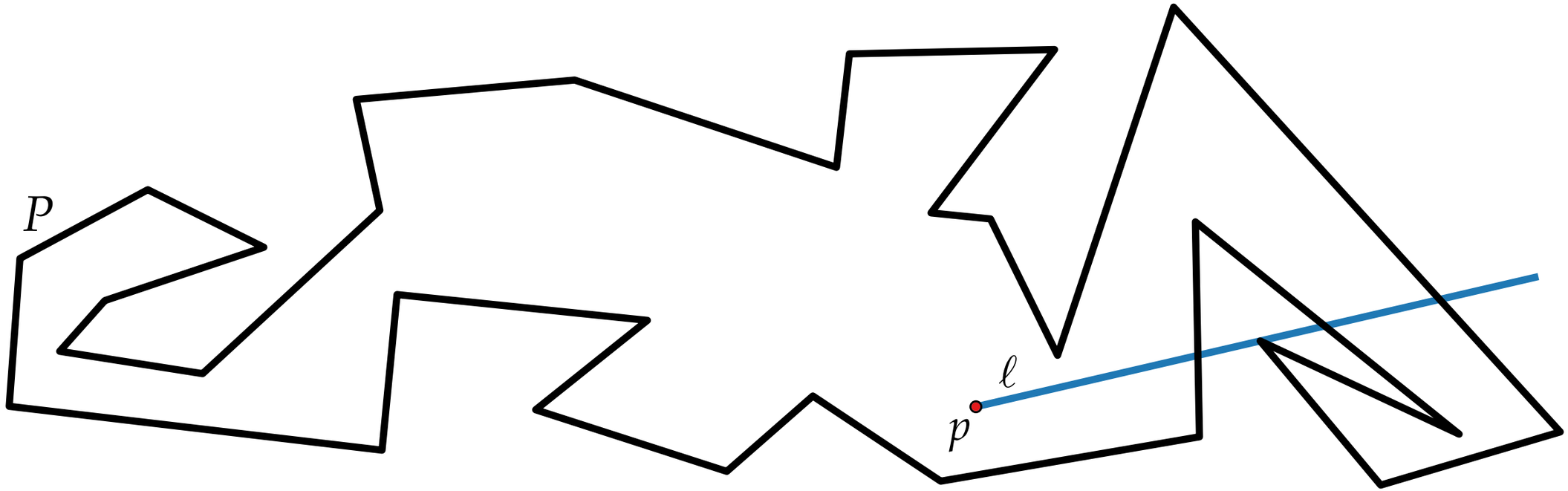

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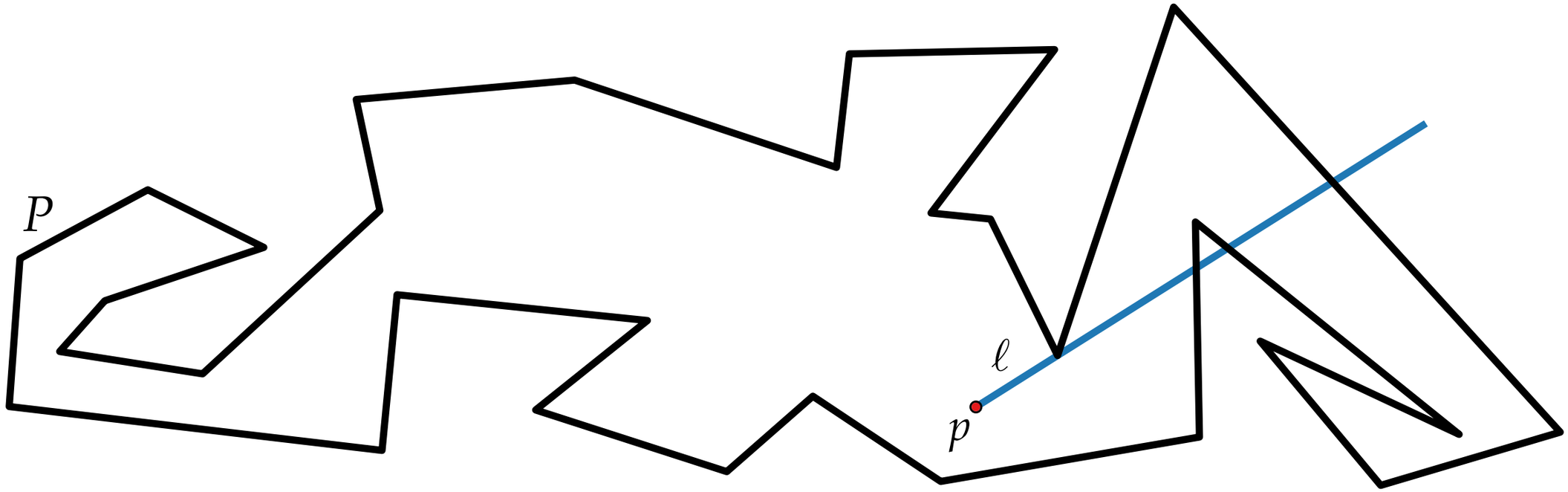

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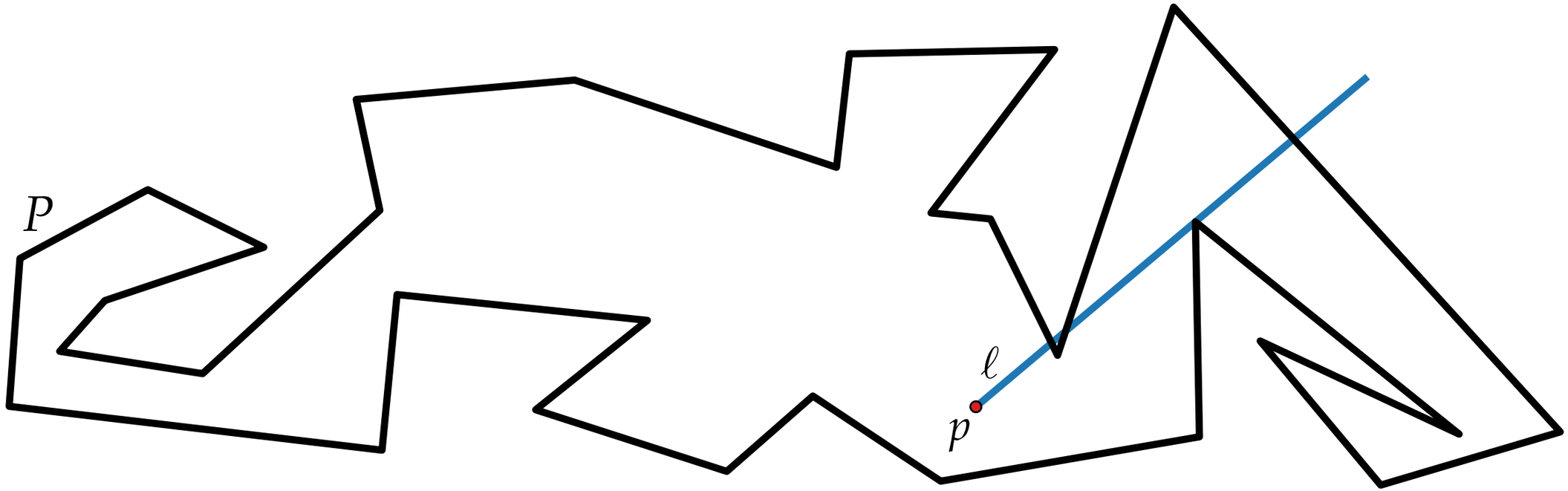

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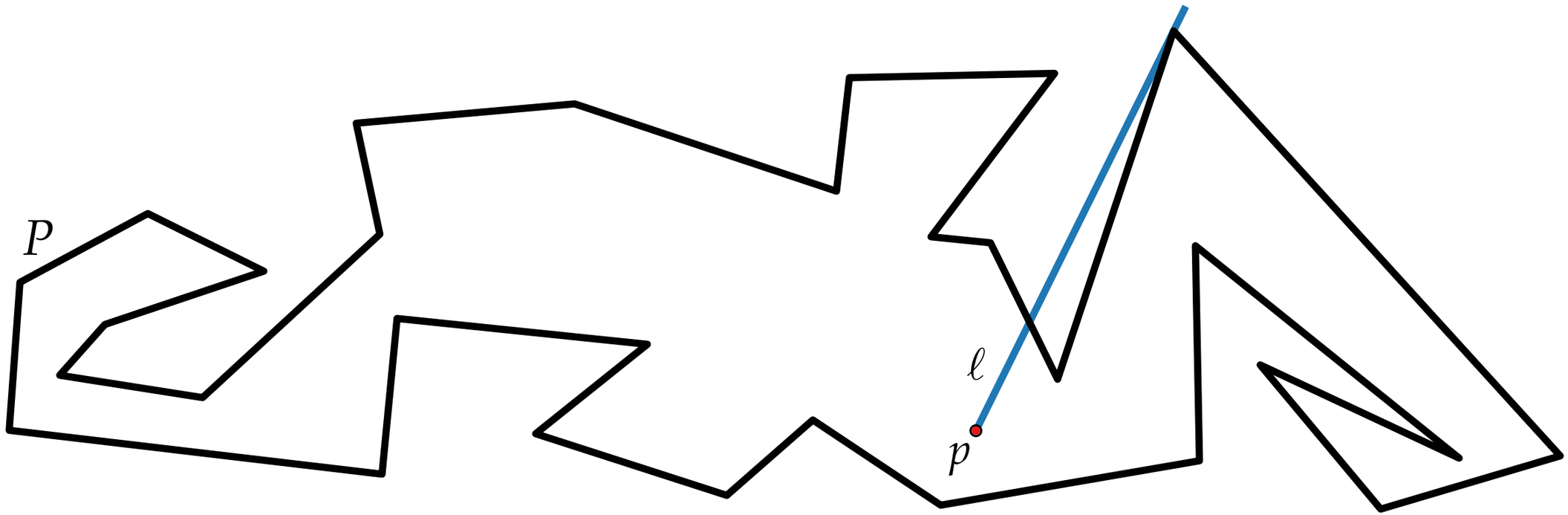

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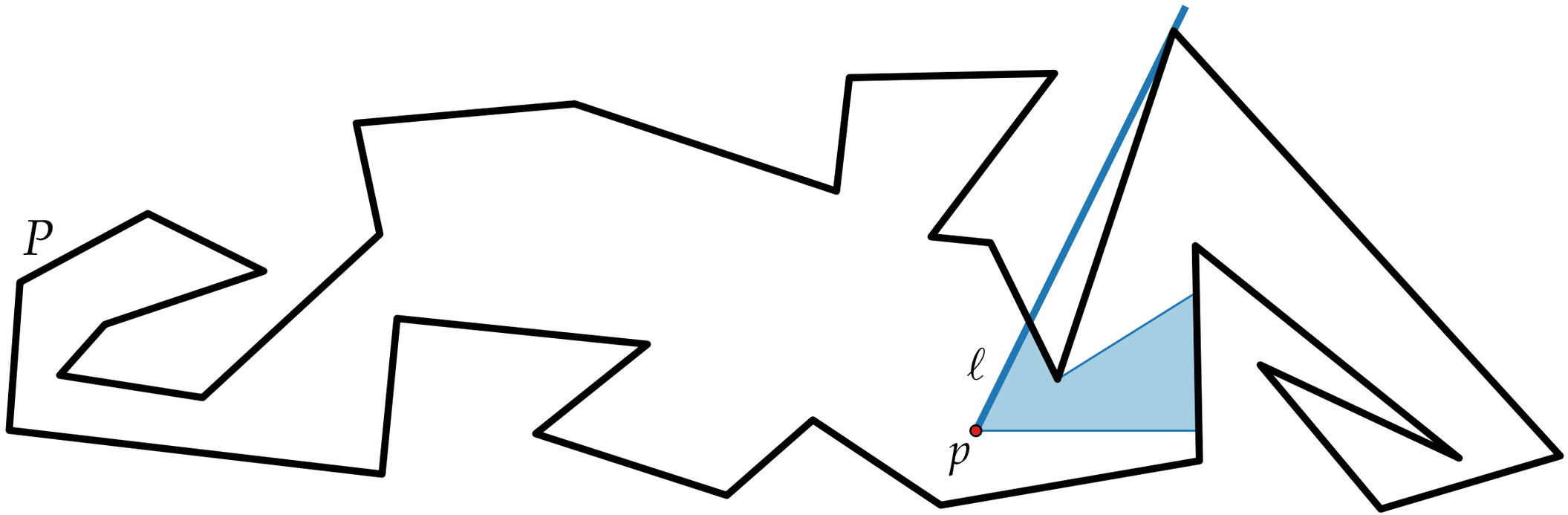

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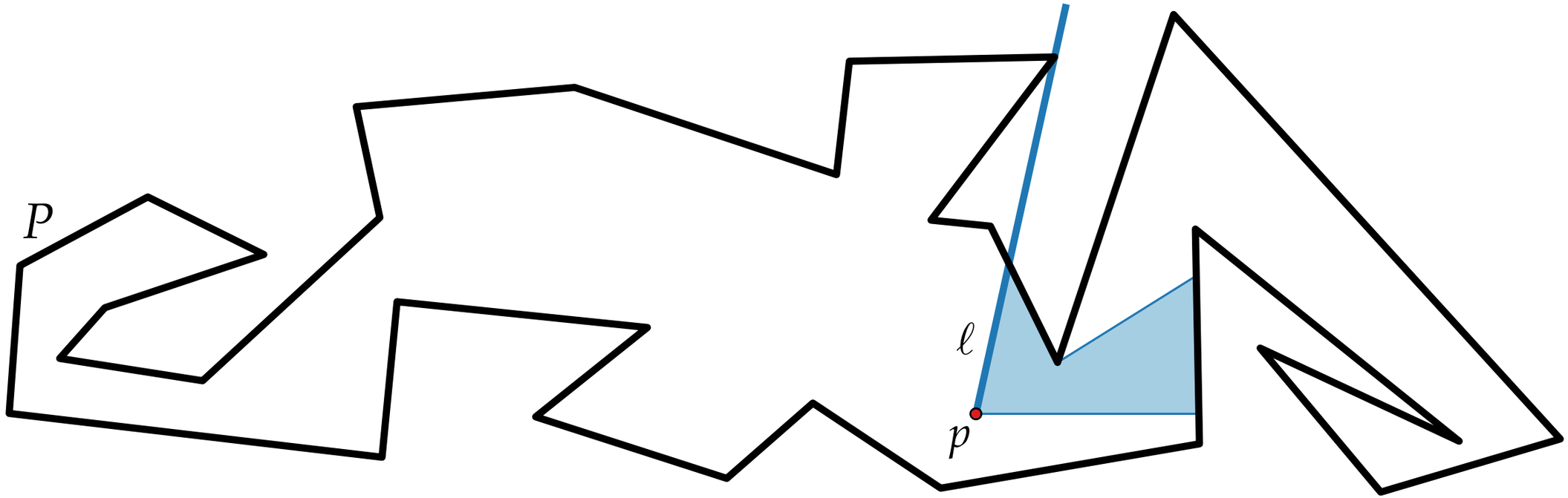

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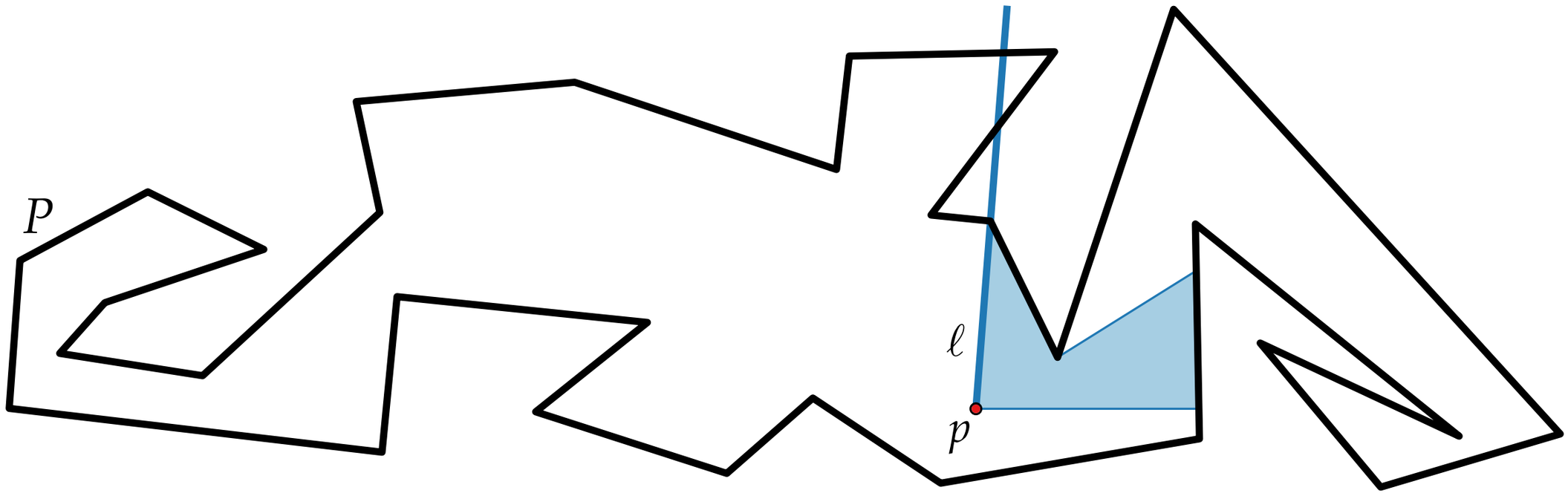

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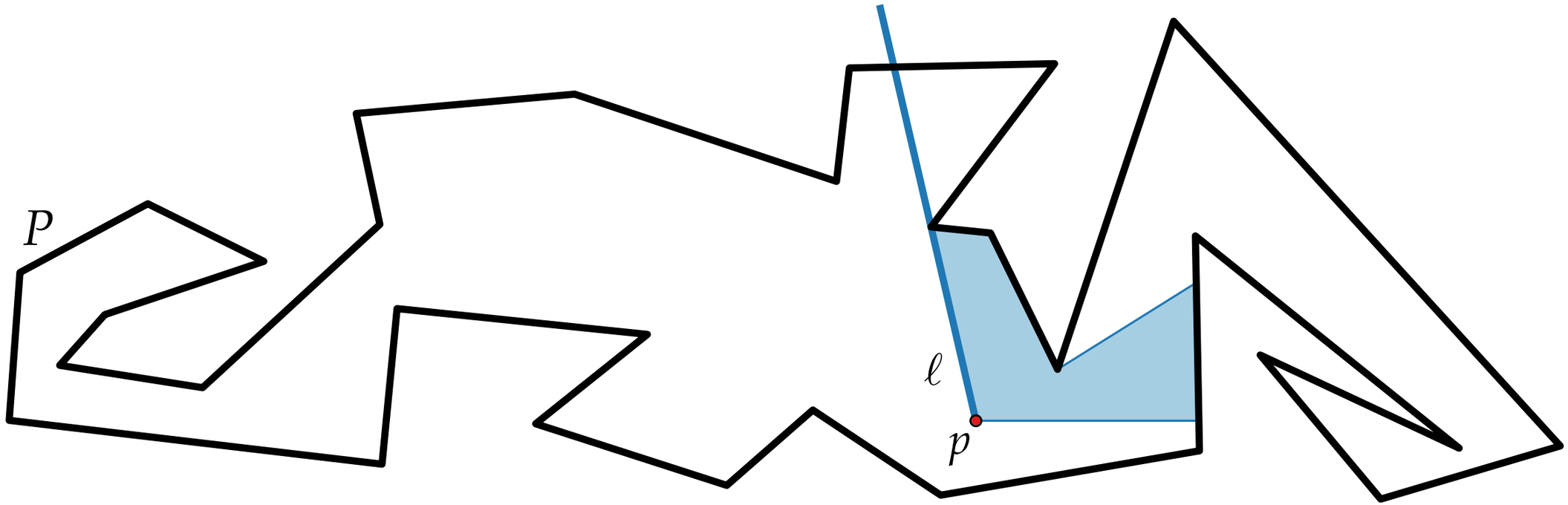

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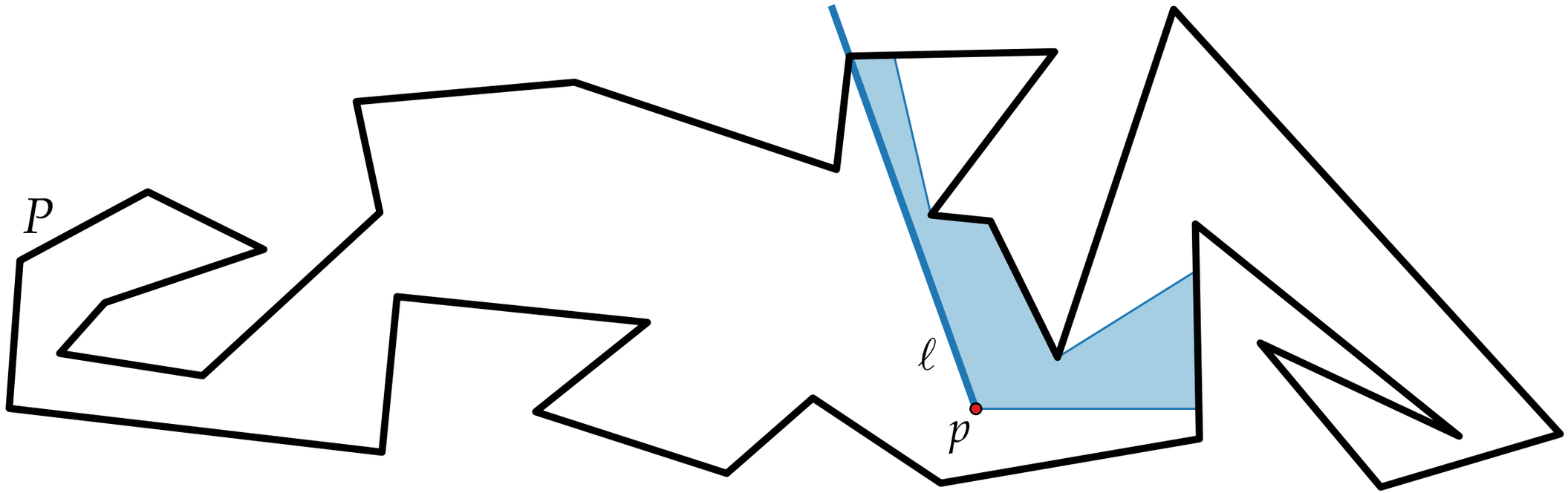

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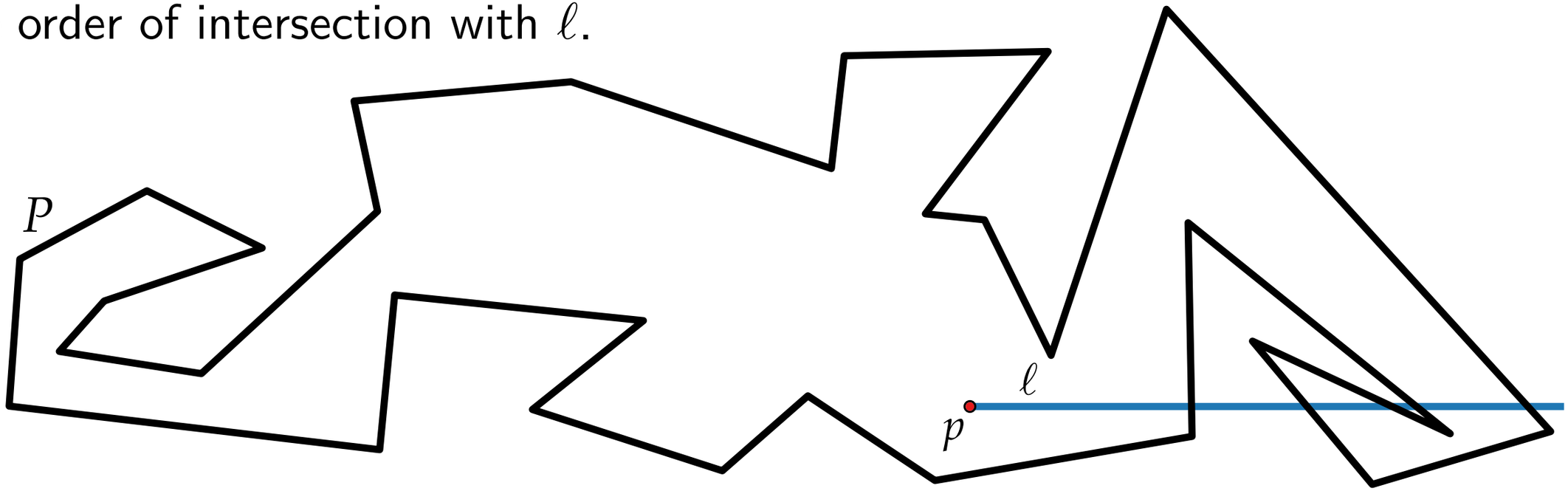

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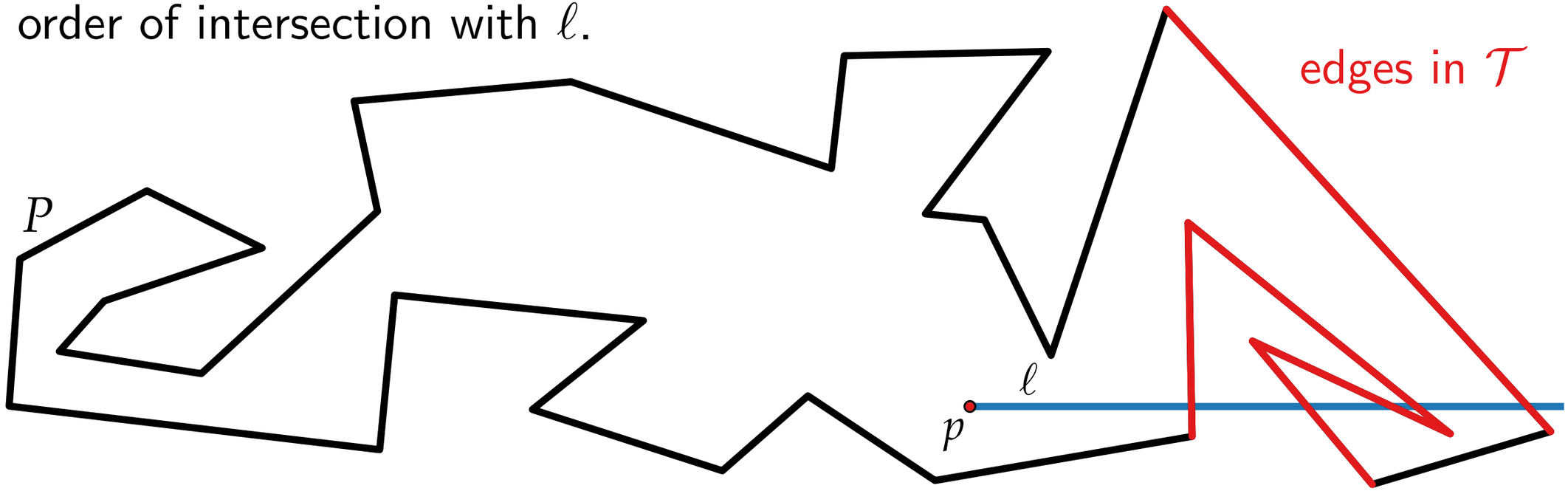

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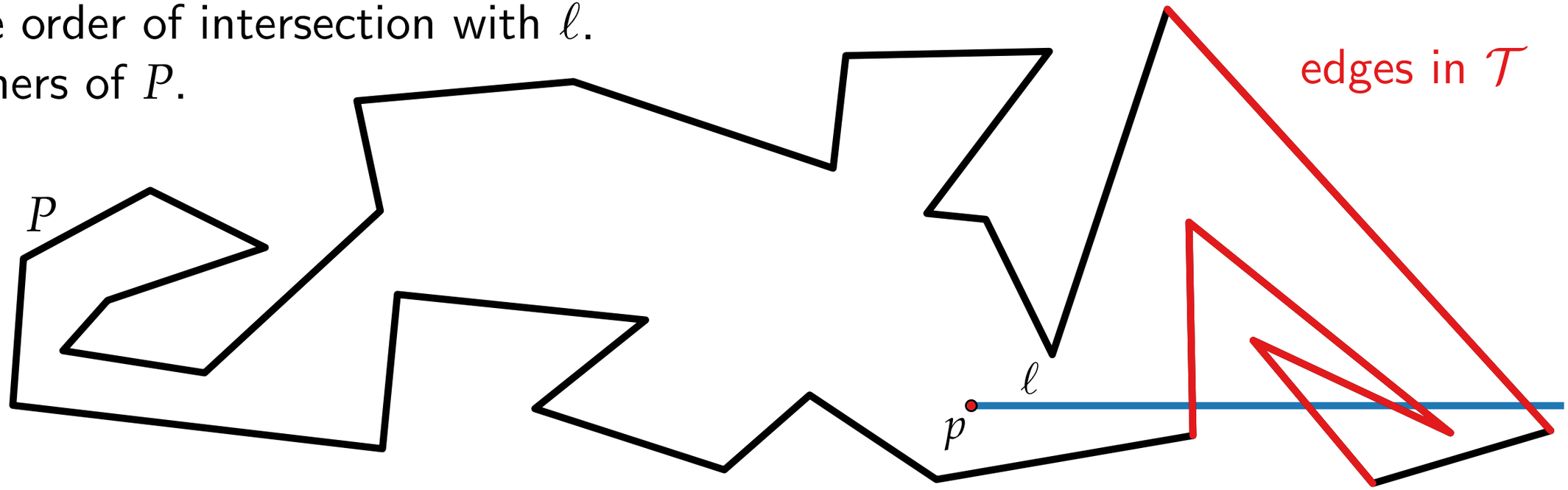

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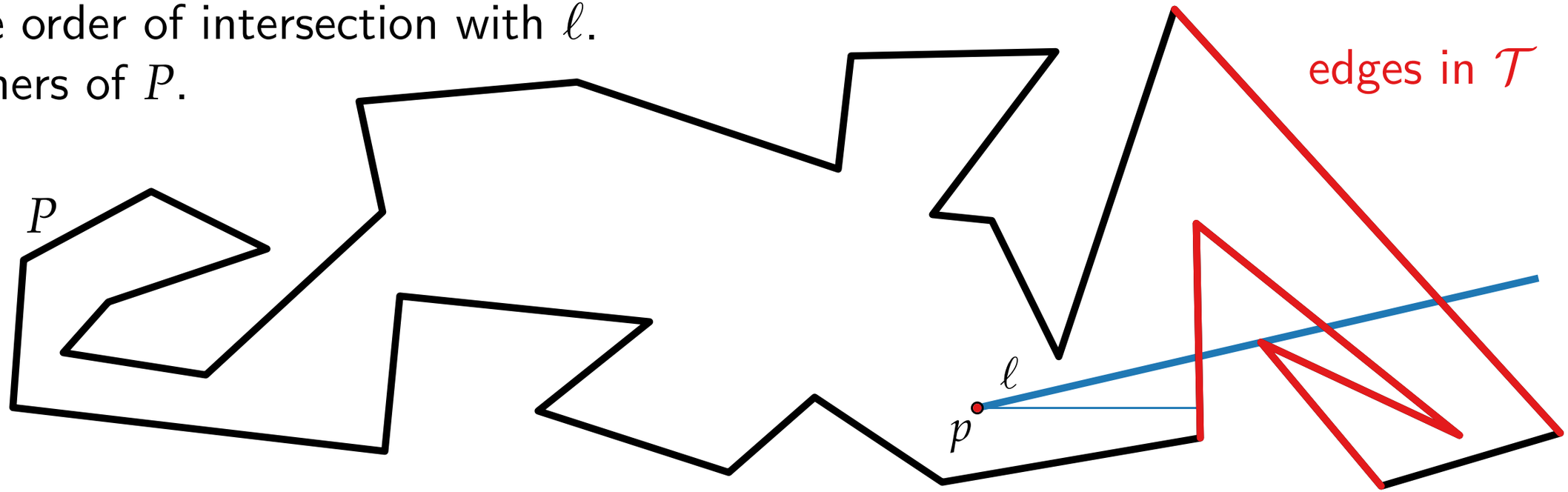

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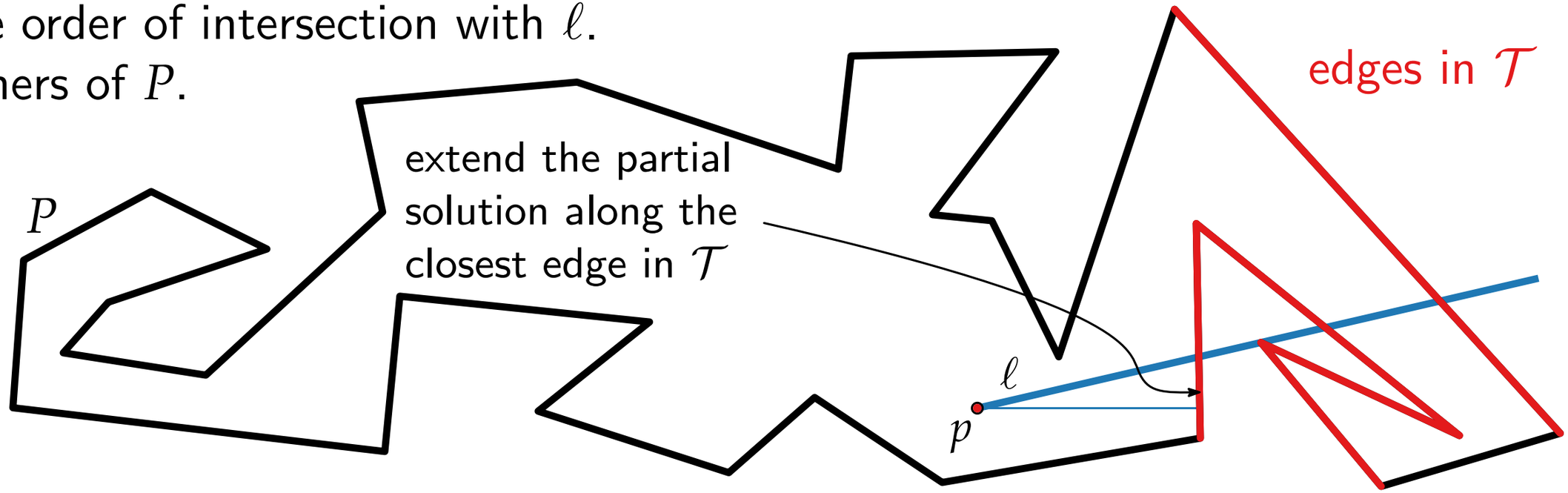

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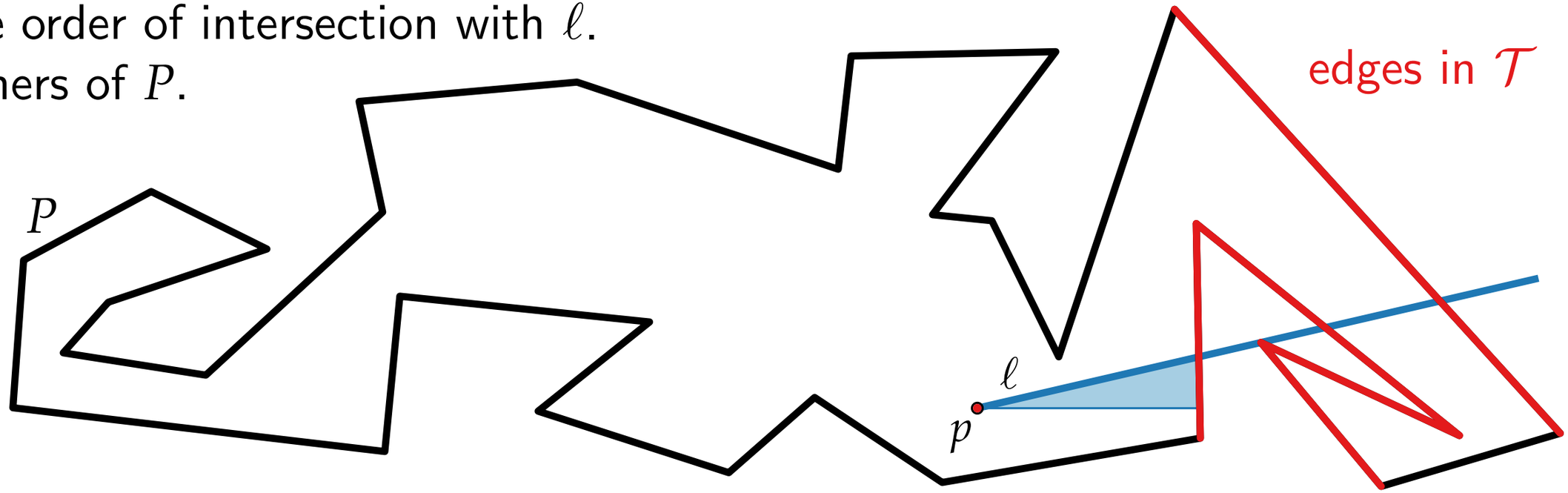

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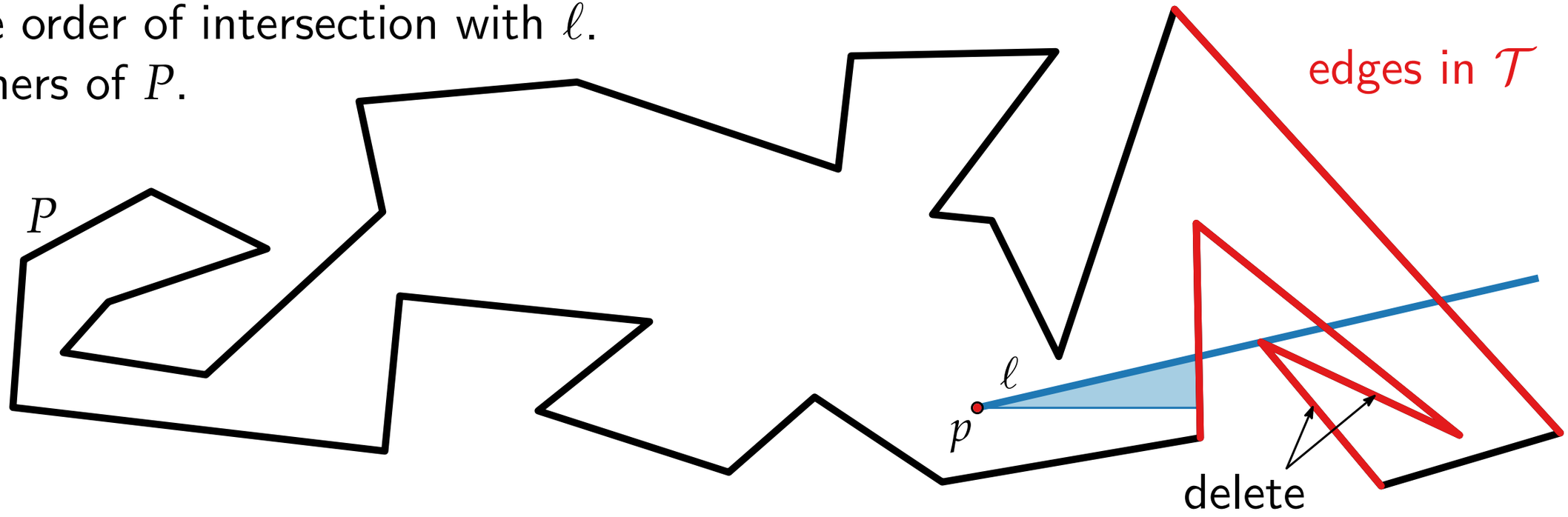

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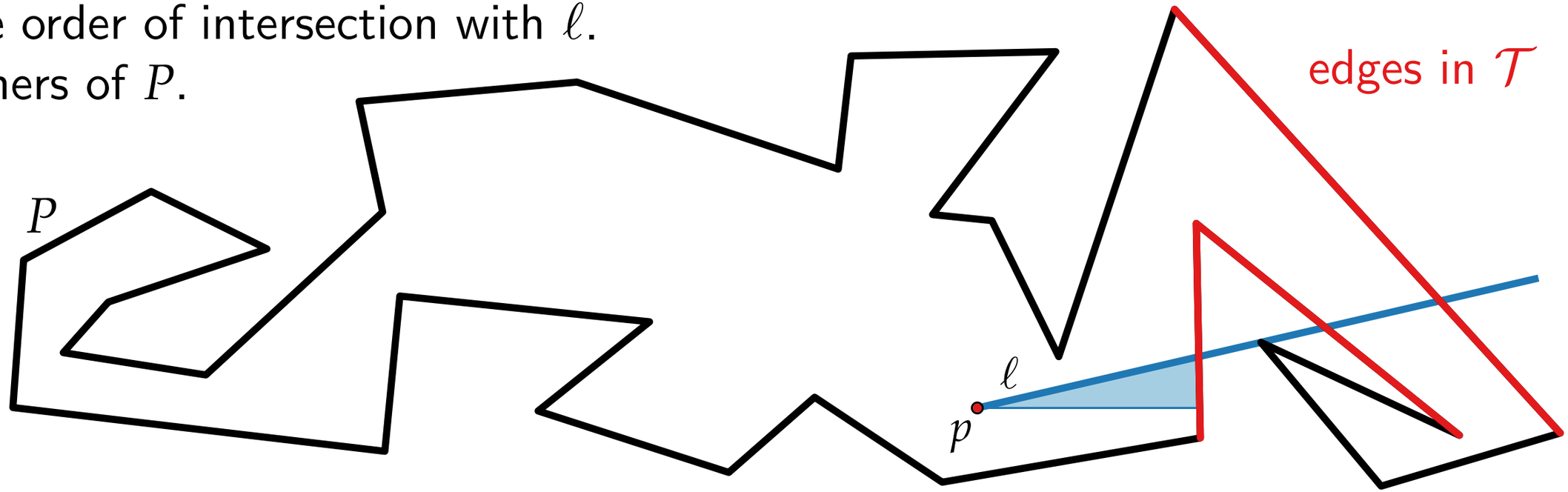

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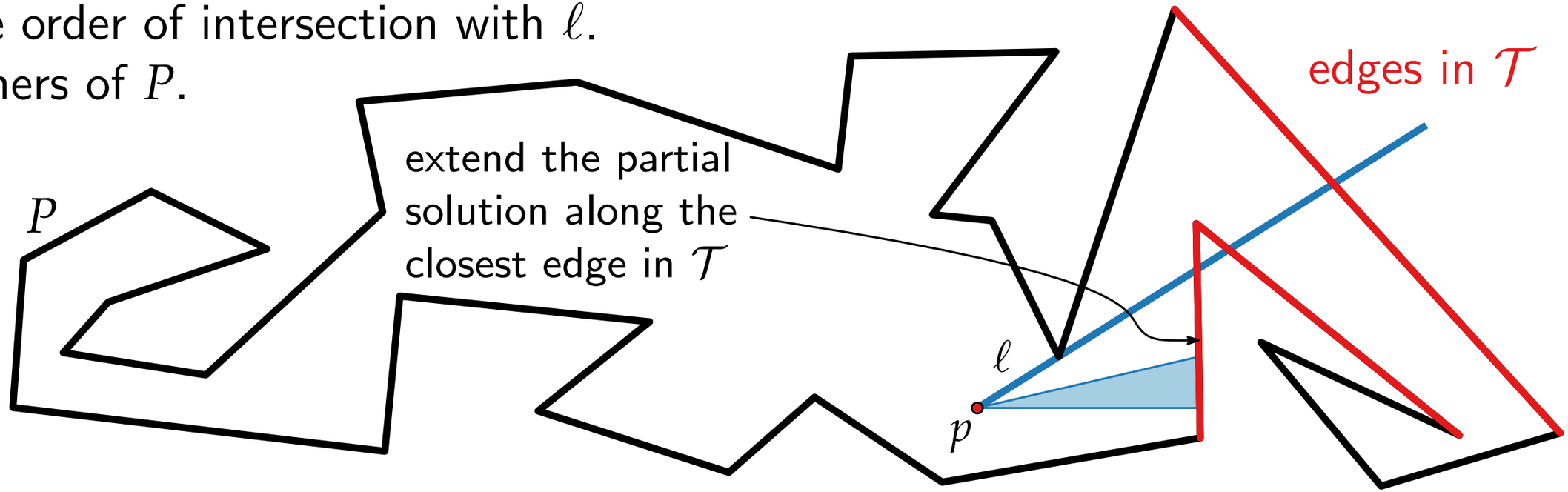

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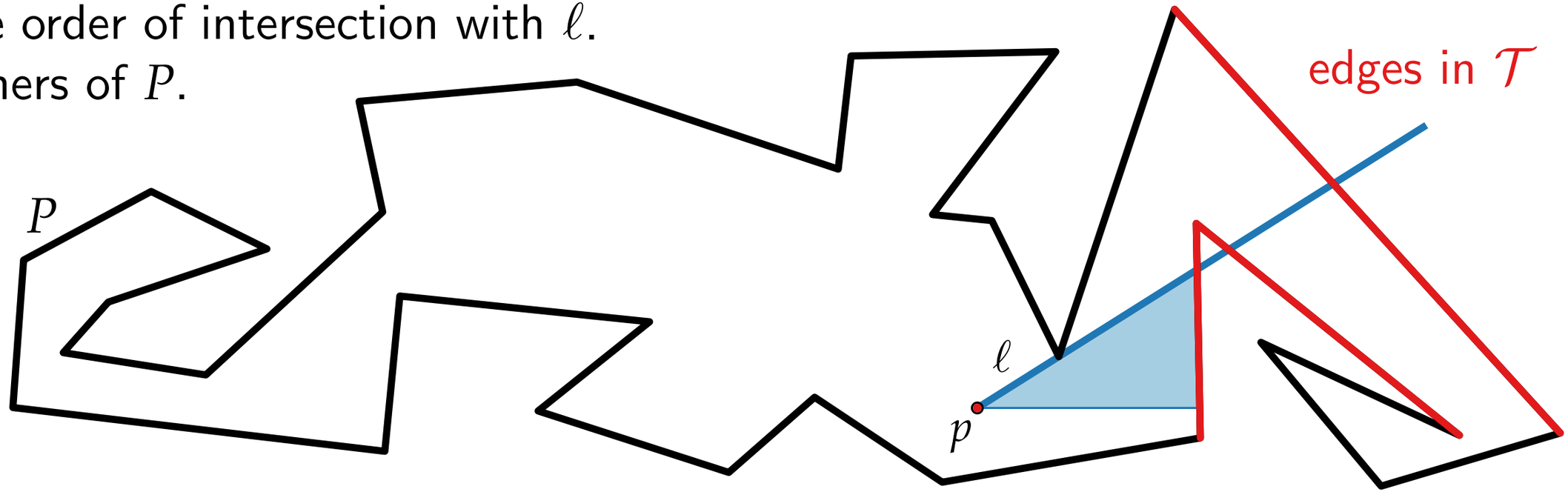

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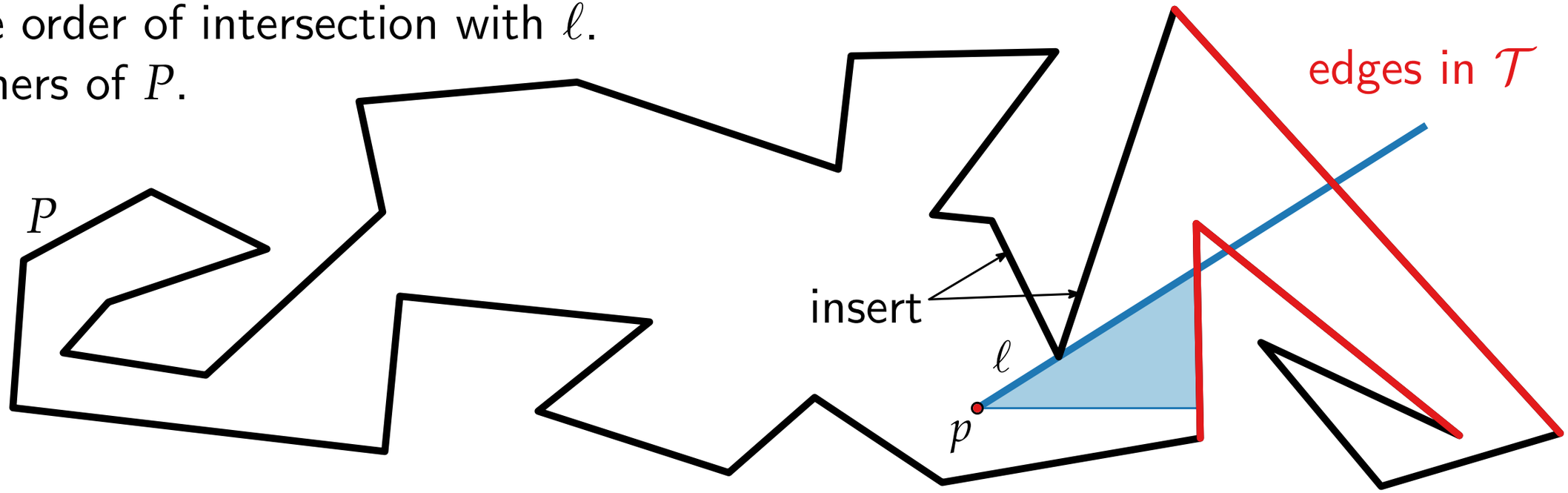

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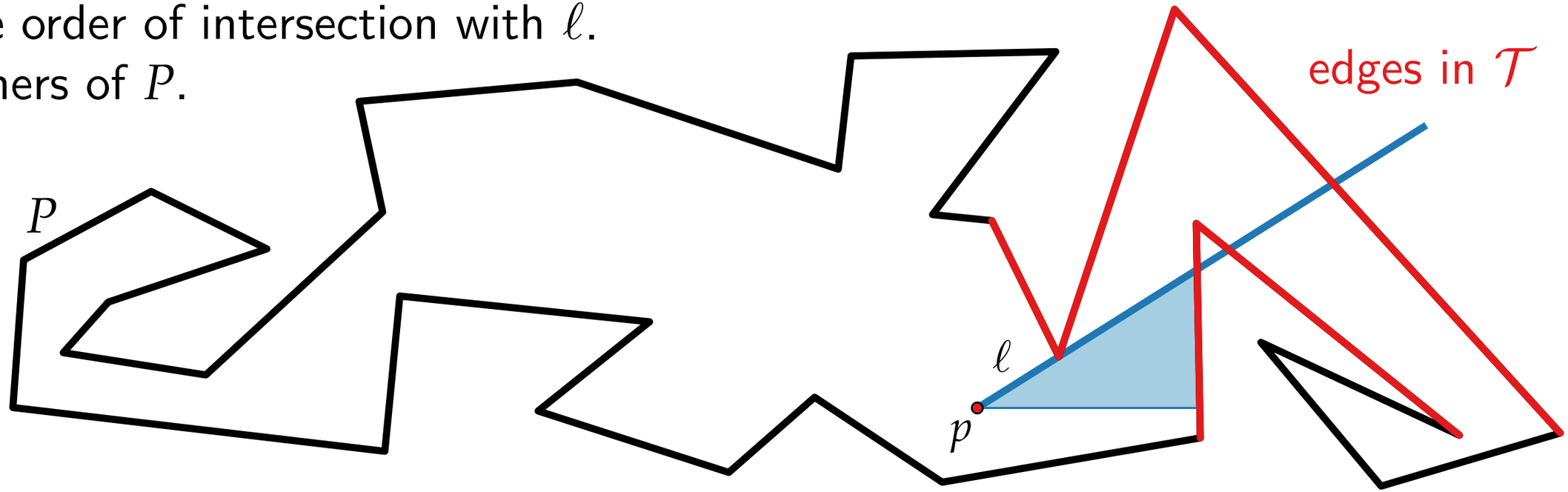

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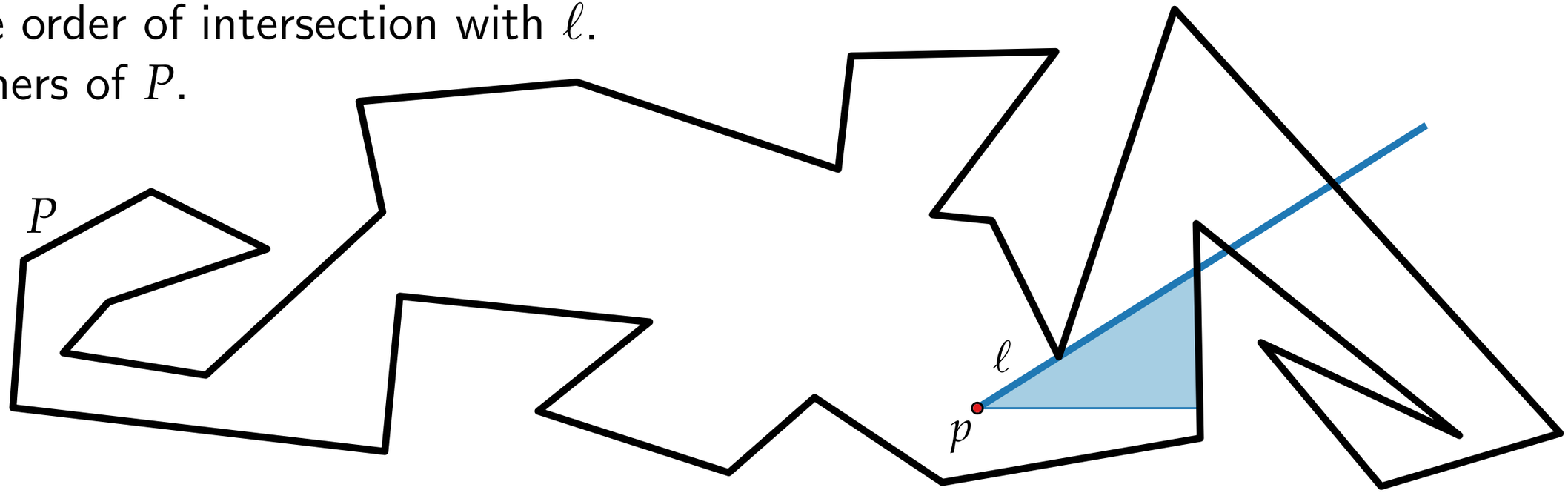

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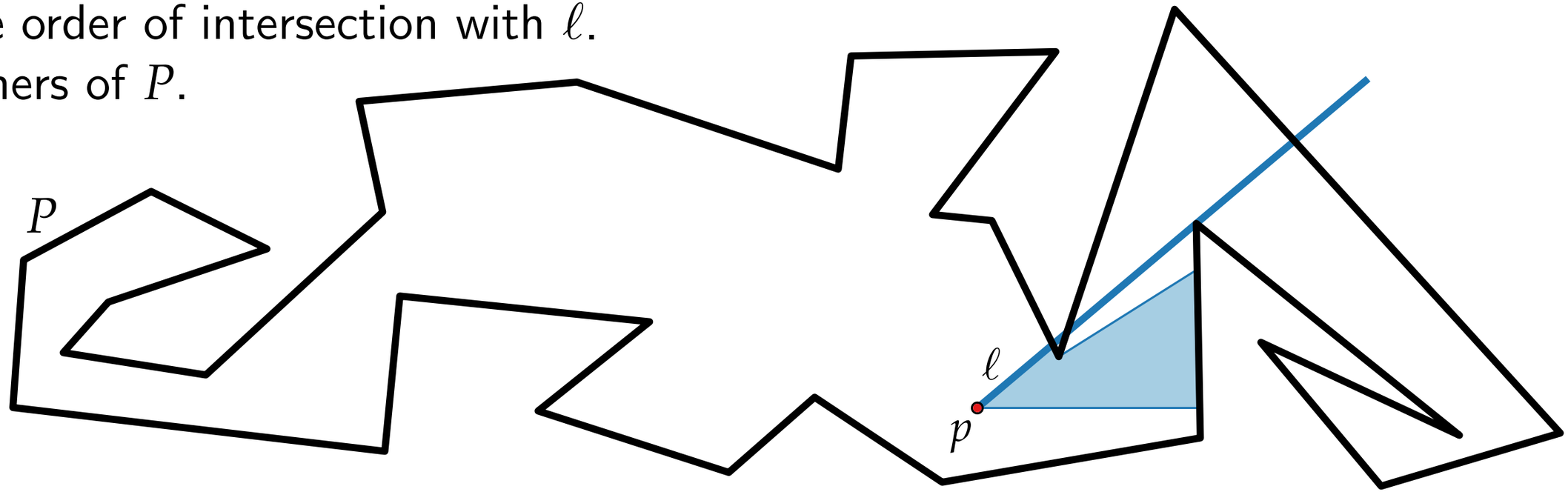

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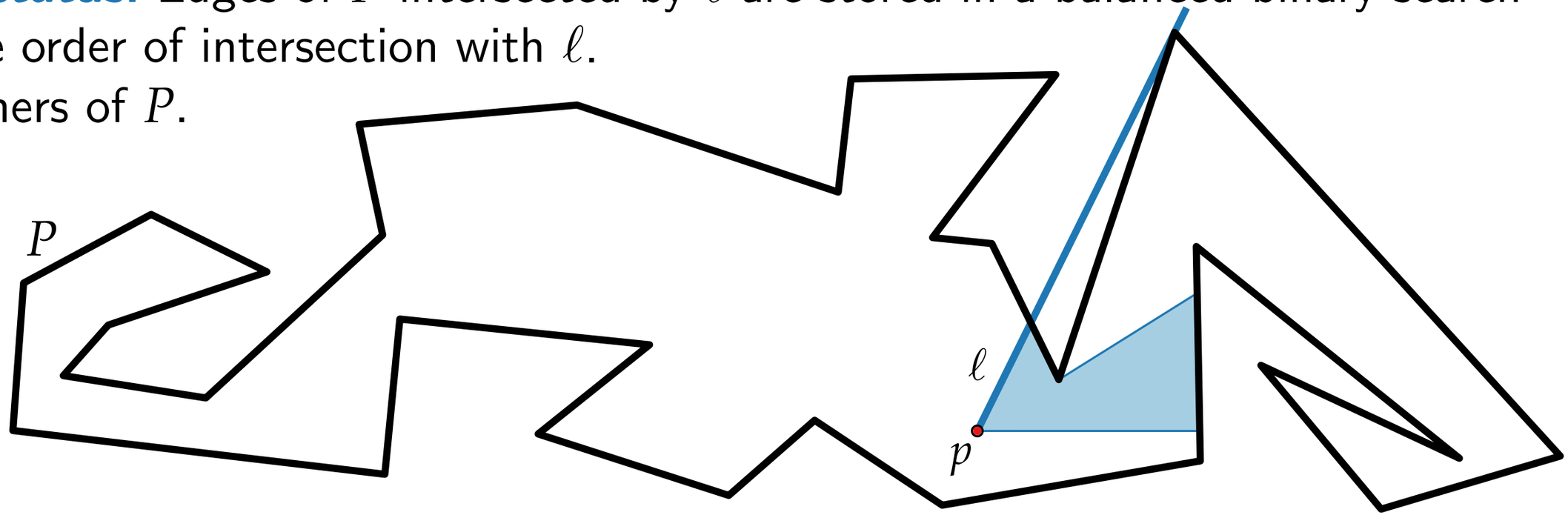

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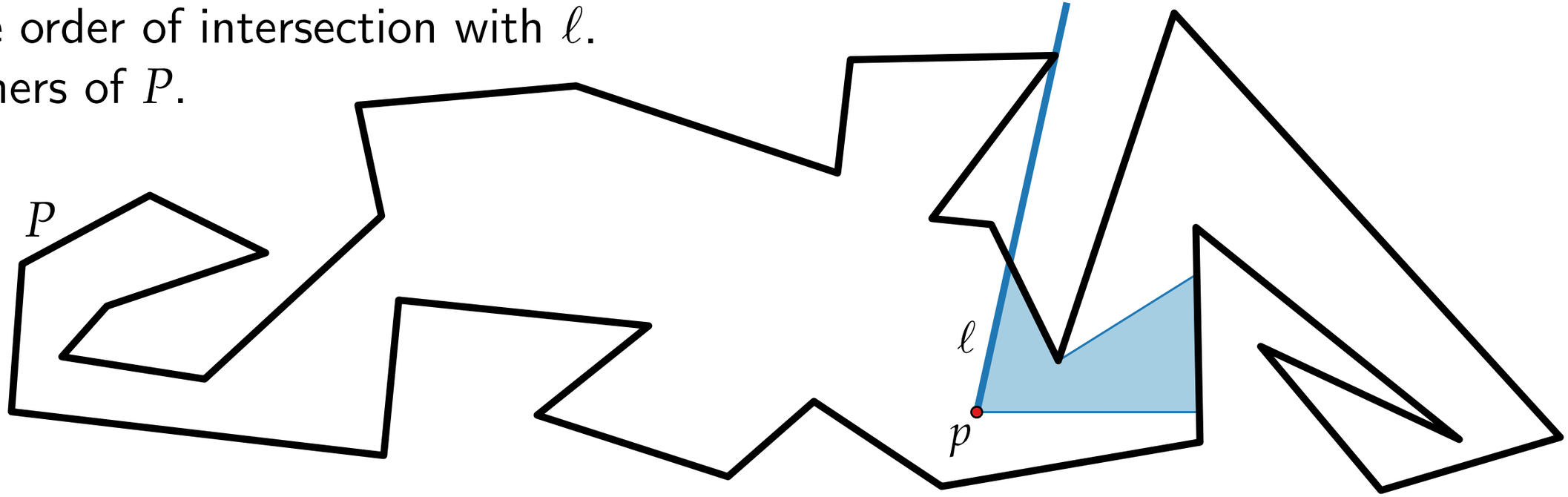

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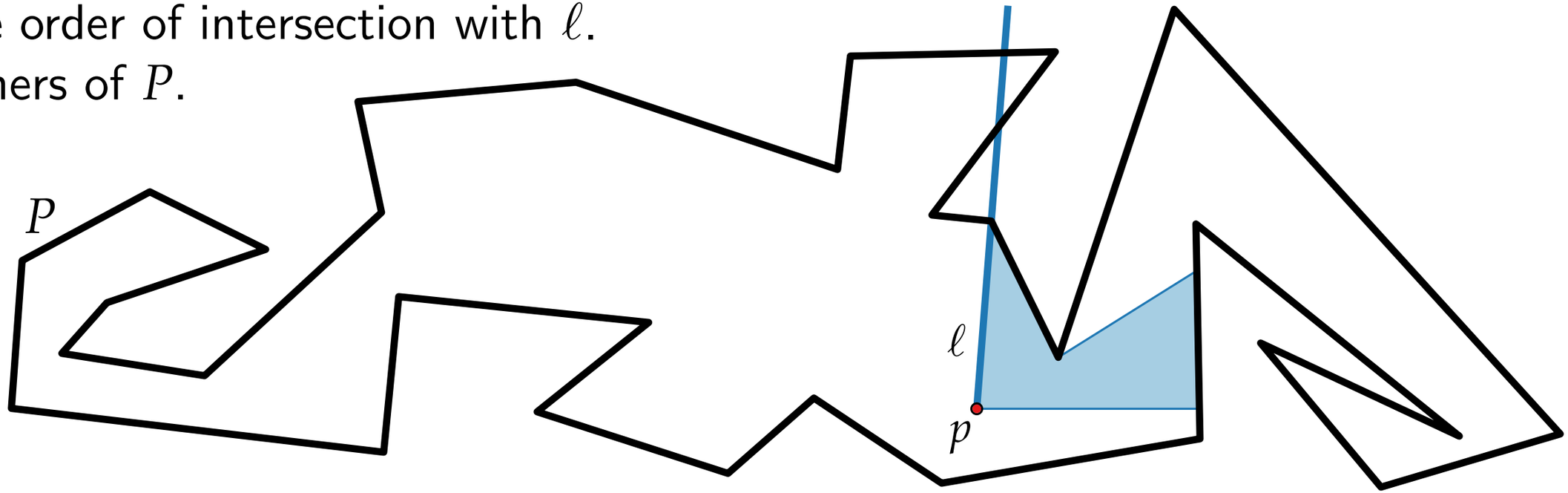

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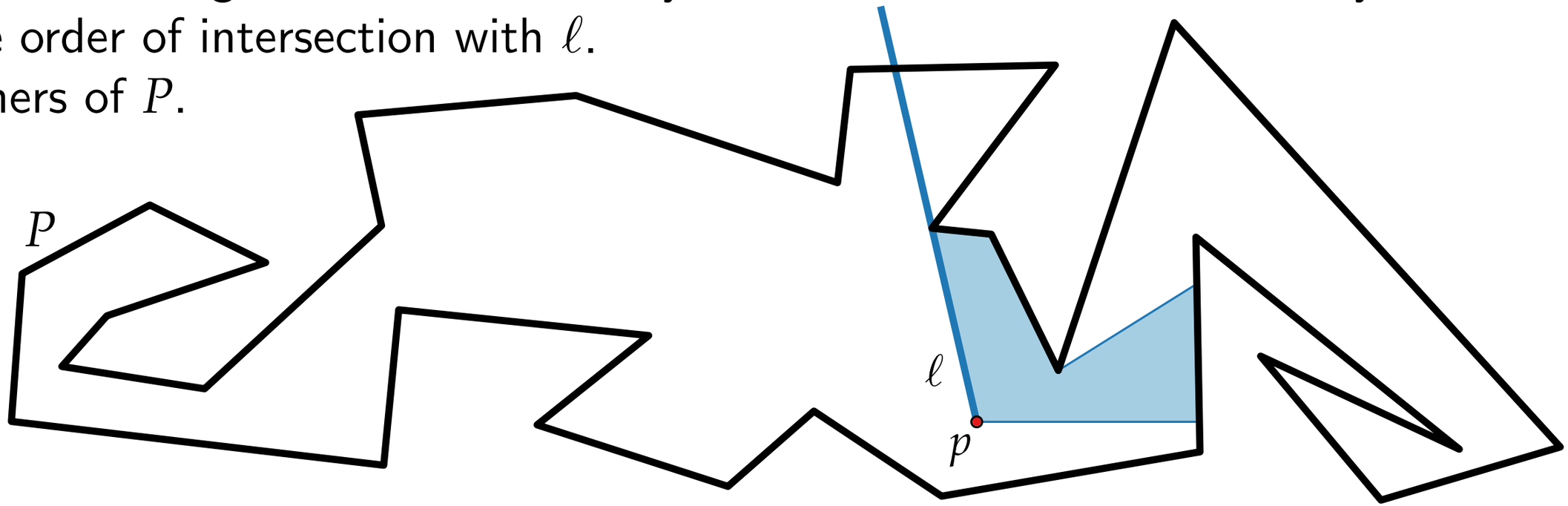

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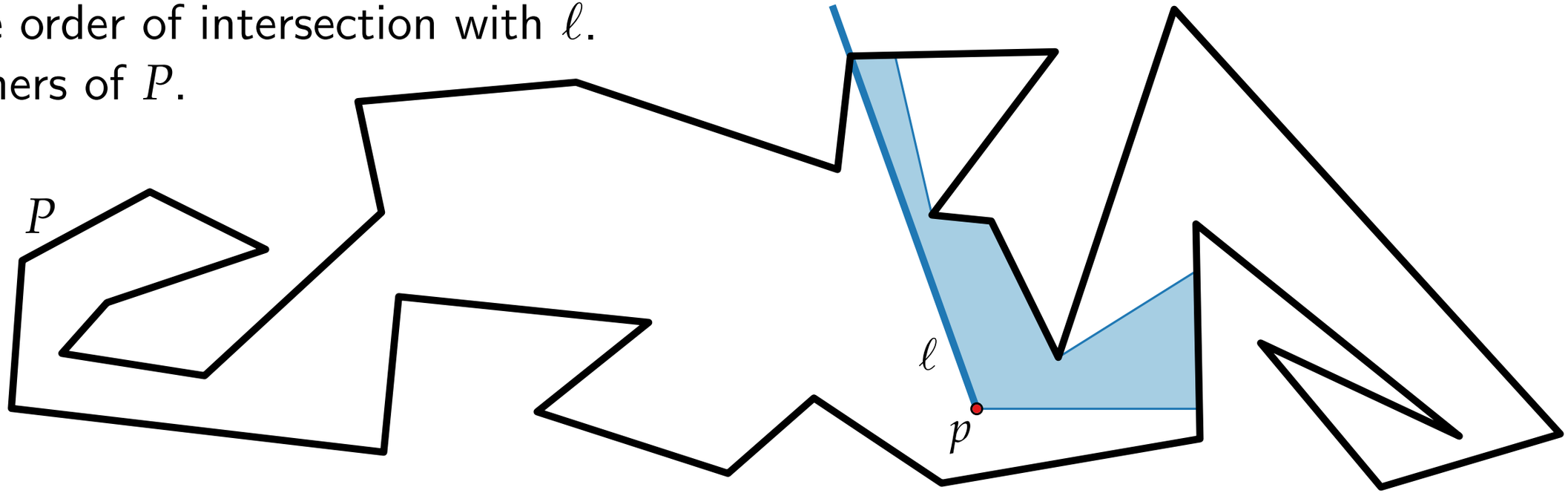

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Total runtime: $\mathcal{O}(n \log n)$

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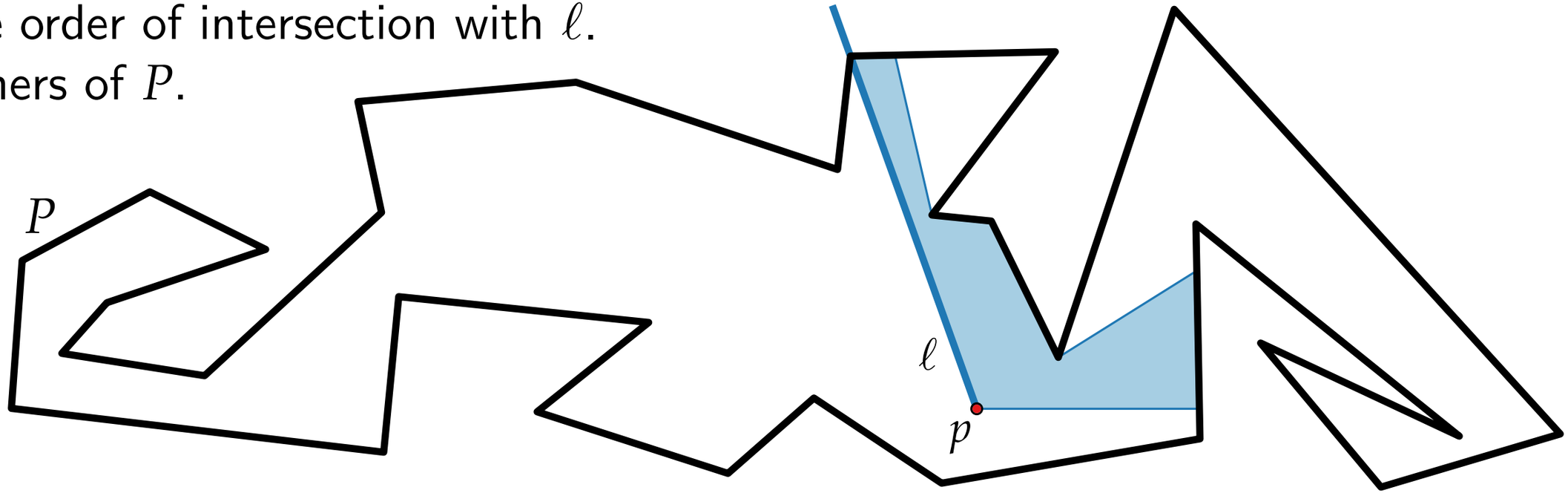

Literature

Rolf Klein. Algorithmische Geometrie: Grundlagen, Methoden, Anwendungen. Springer 2005.

Otfried Cheong, Mark de Berg, Mark Overmars, Marc van Kreveld: Computational Geometry: Algorithms and Applications, 3rd edition, Springer 2008.