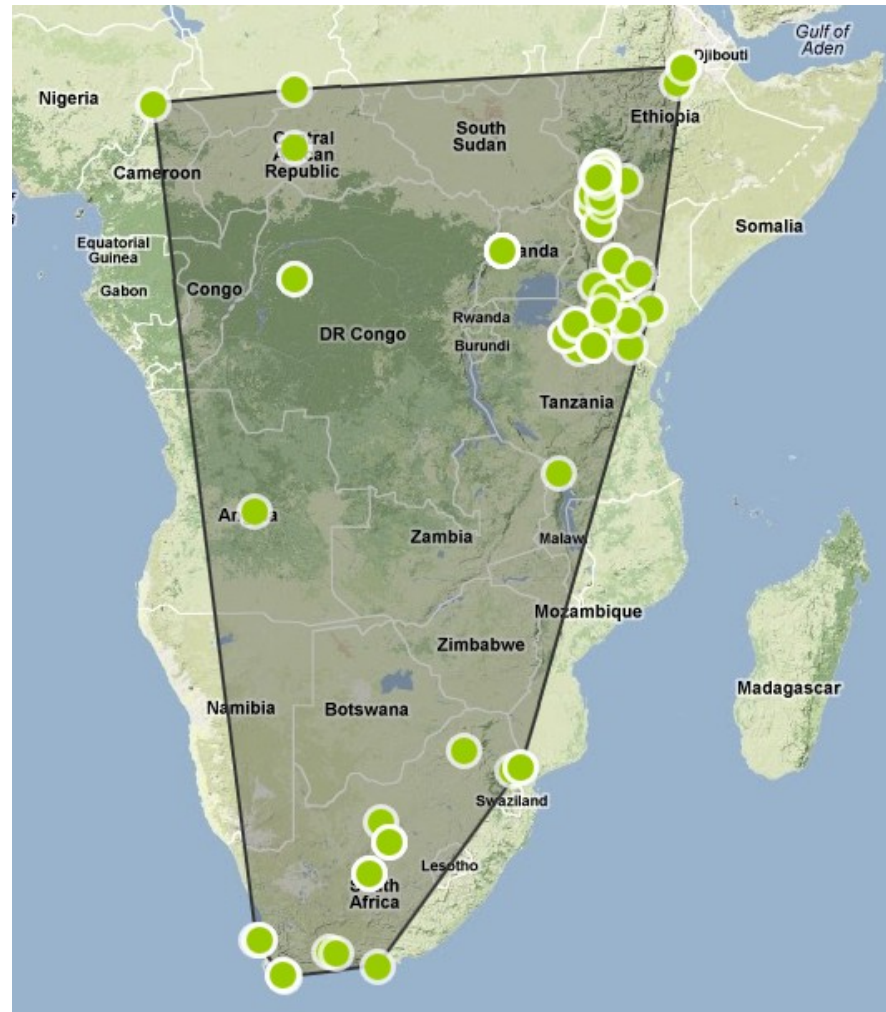


Convex Hulls

Geometric Algorithms

Wildlife analysis: extent of occurrence

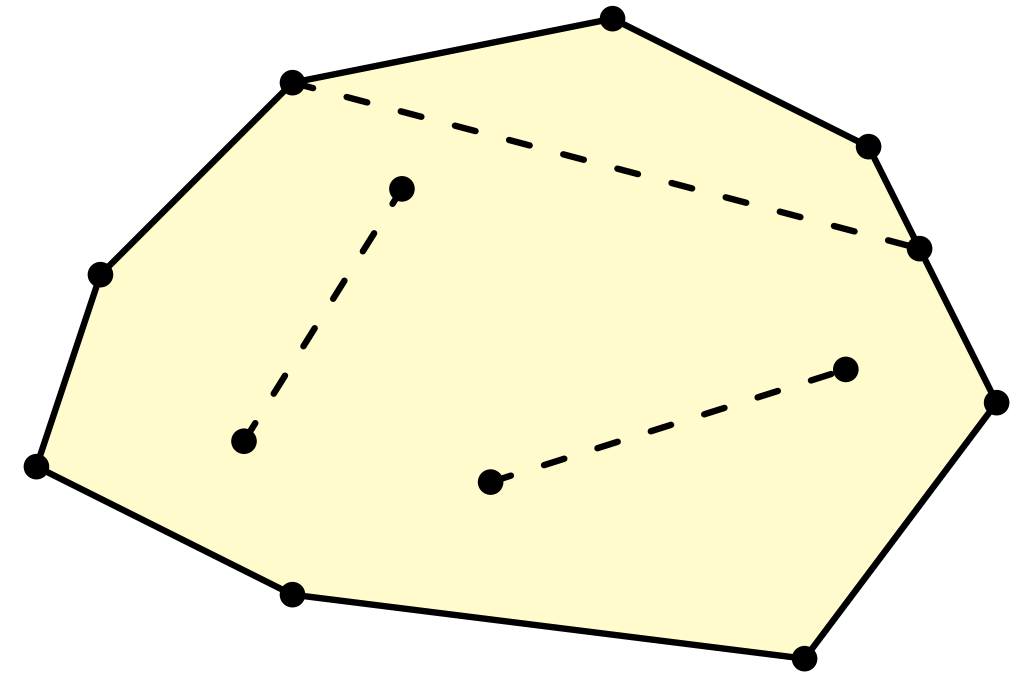


<http://geocat.kew.org/>
<http://www.iucnredlist.org/>

Black Rhino (*Diceros bicornis*)

Convexity

Definition: A shape or set is **convex** if for any two points that are part of the shape, the whole connecting line segment is also part of the shape

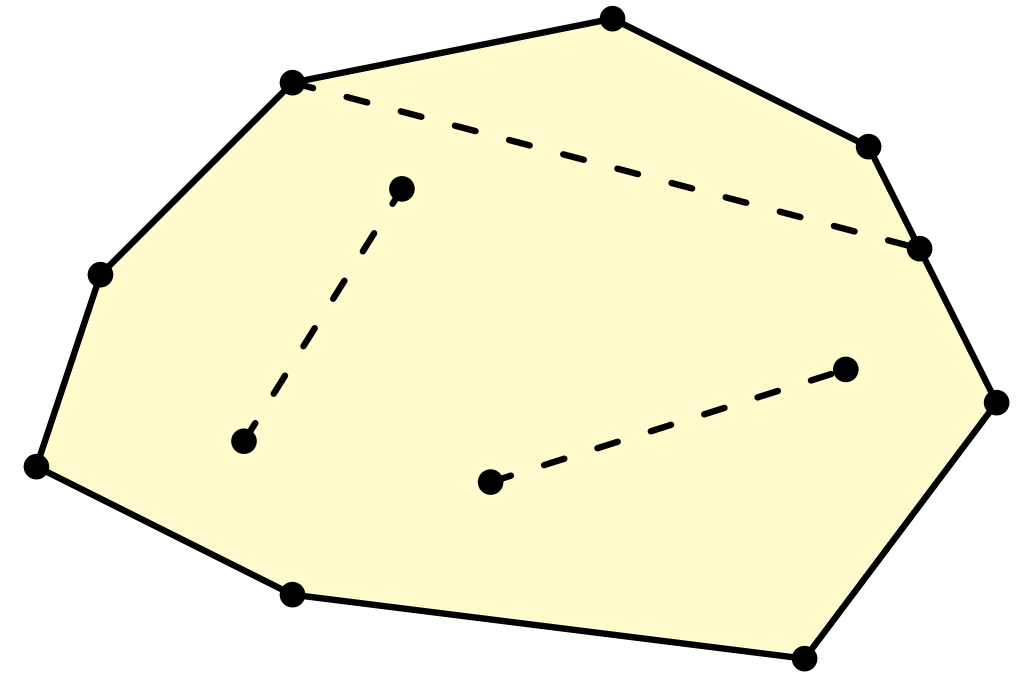


Convexity

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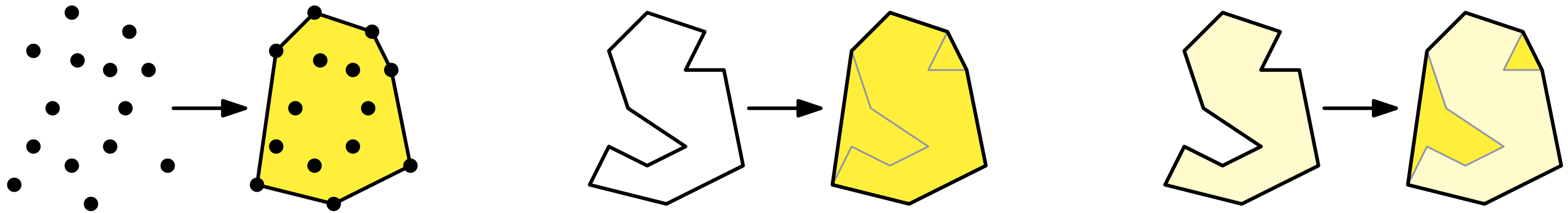
Question: Which of the following shapes are convex?

- point
- line segment
- line
- circle
- disk
- quadrant



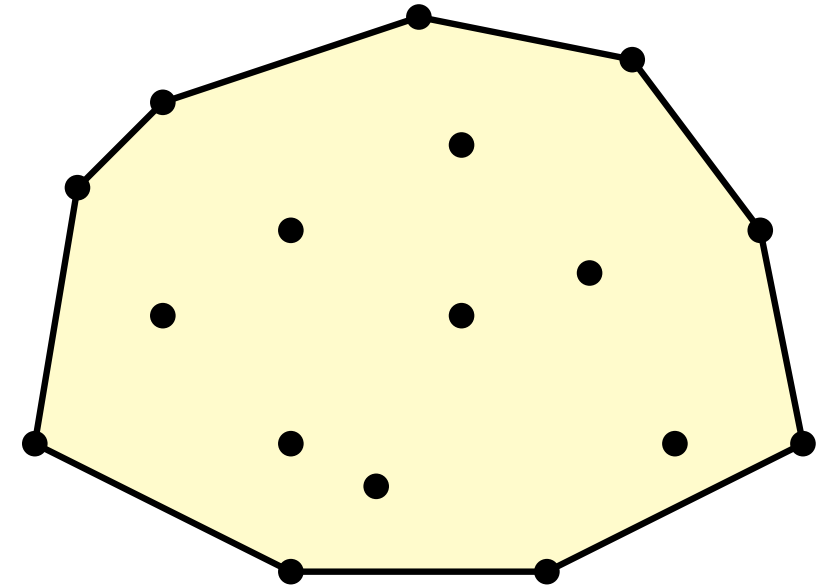
Convex hull

Definition: For any subset of the plane (set of points, polygonal chain, simple polygon), its **convex hull** is the smallest convex set that contains that subset



Convex hull problem

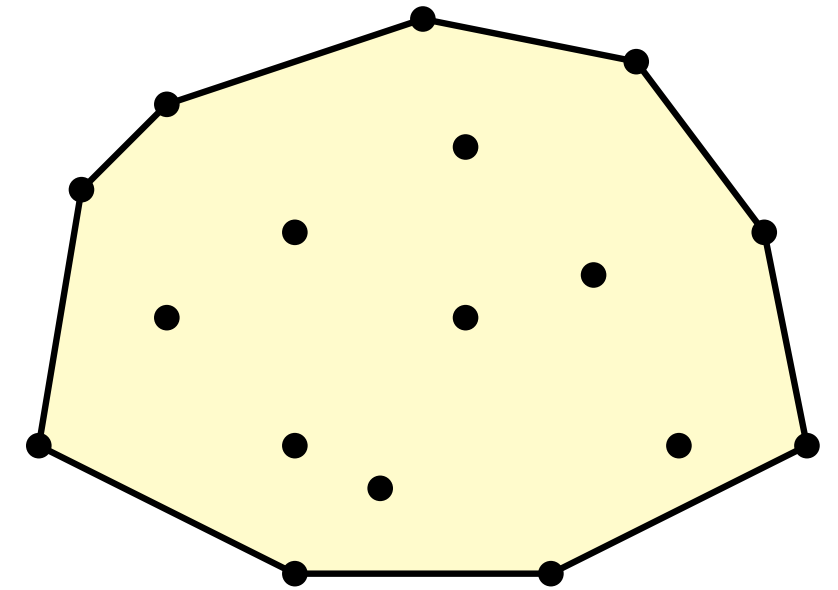
Problem: Give an algorithm that computes the convex hull of any given set of n points in the plane efficiently



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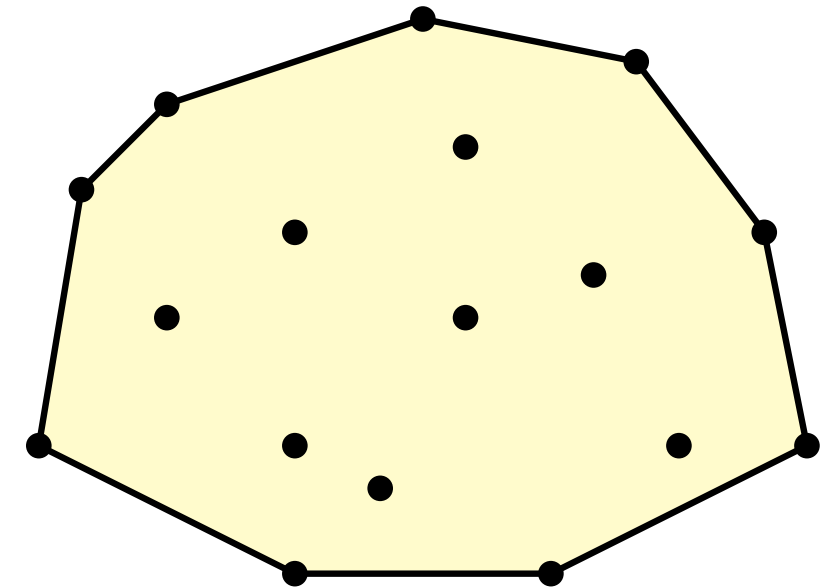
- **input** has $2n$ coordinates, so $O(n)$ size



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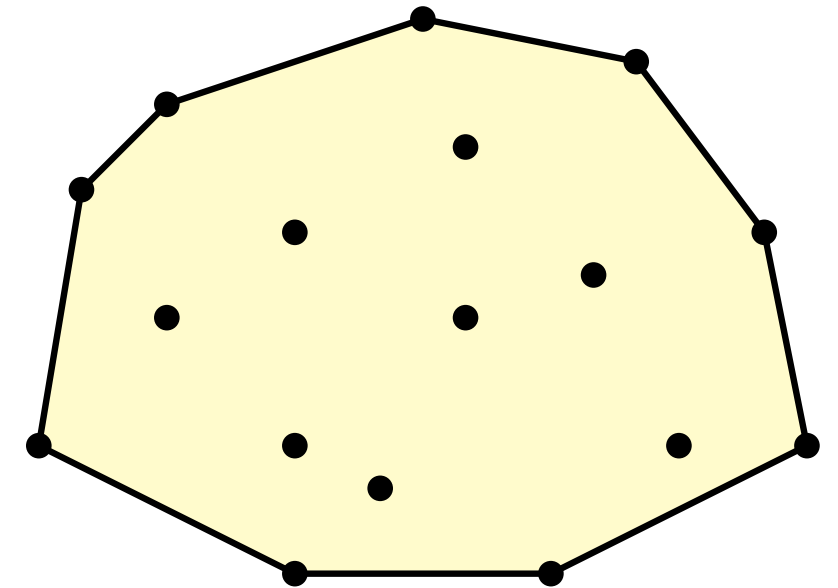
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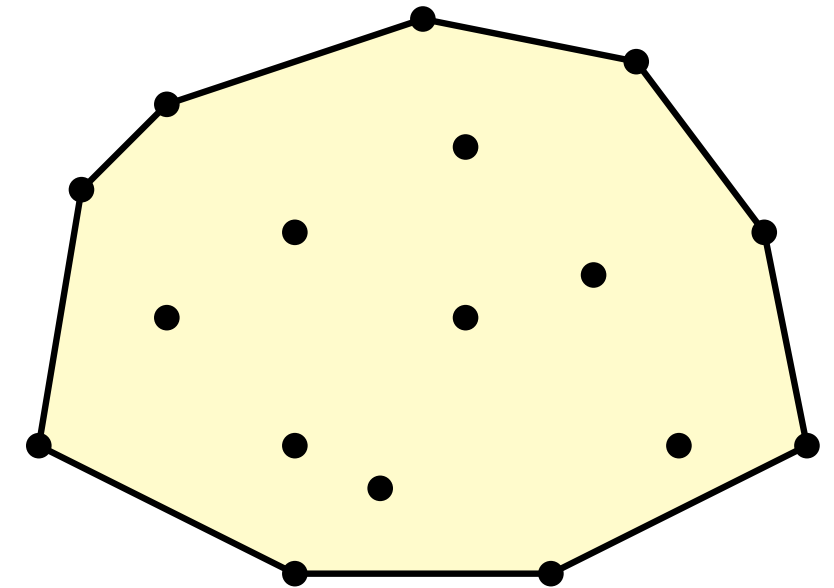
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 - assume the n points are distinct



Convex hull problem

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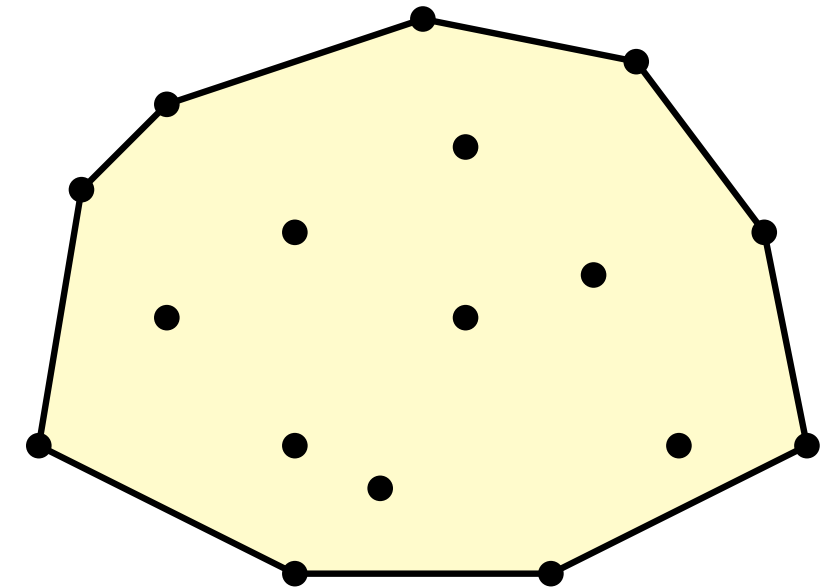
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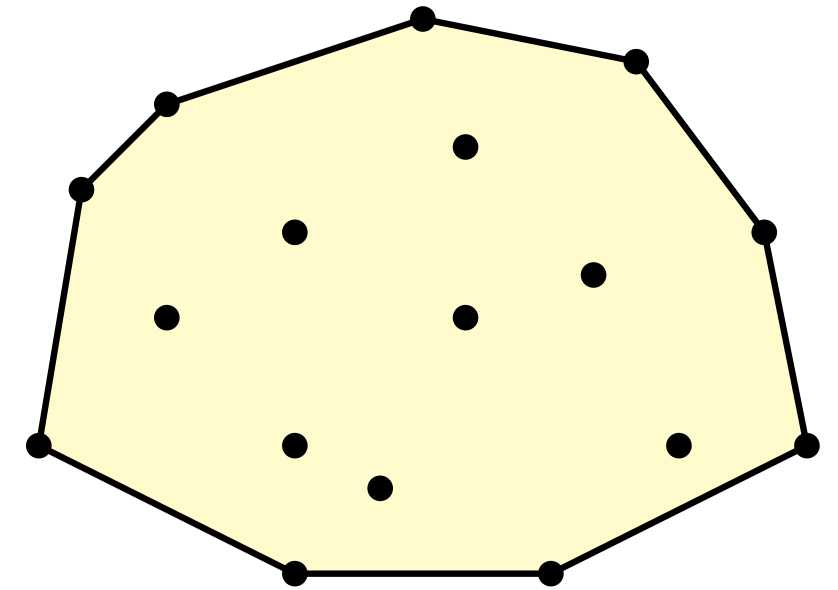
- **input** has $2n$ coordinates, so $O(n)$ size
- **output?**
 - assume the n points are distinct
 - output has at least 2 and at most n points
 - output size is between $O(1)$ and $O(n)$



Convex hull problem: questions

Problem: Give an algorithm that computes the convex hull of any given set of n points in the plane efficiently

Question: Is there an algorithm computing a convex hull faster than $O(n)$ time in the worst case?

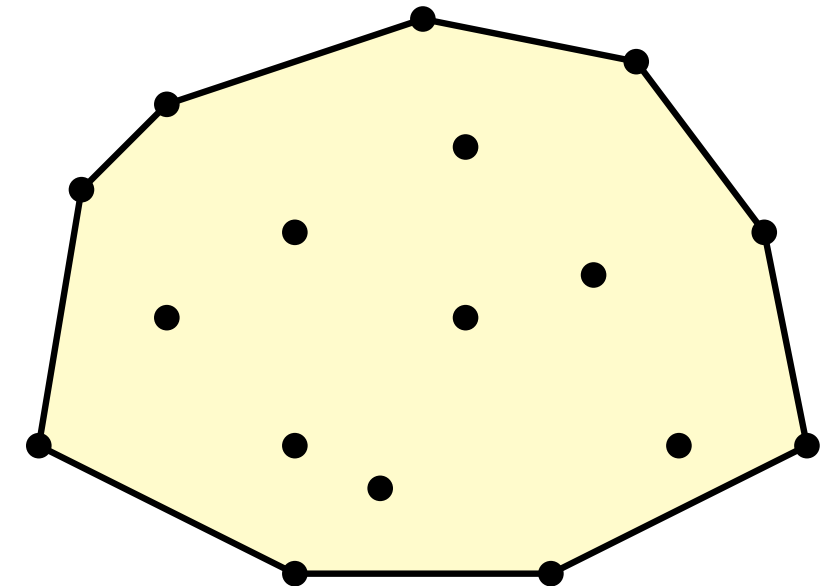


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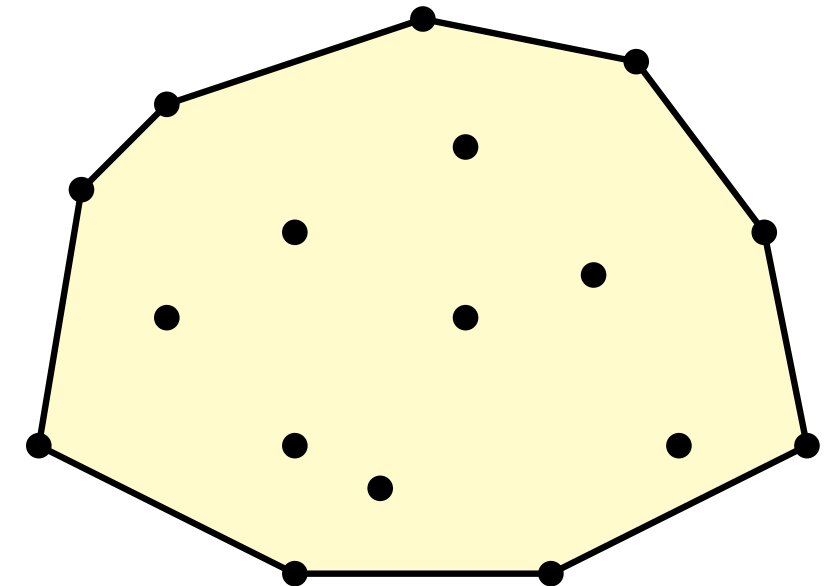
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Question: Is there any hope of finding an $O(n)$ time algorithm?

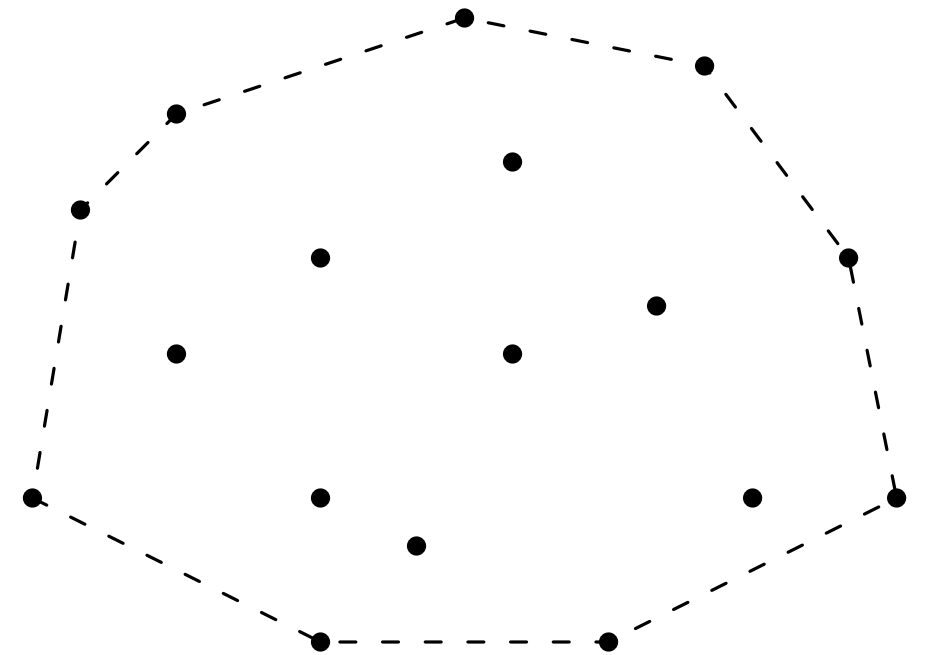


Developing an algorithm

To develop an algorithm, draw many **sketches** to gain **insight**, make various **observations**, find useful **properties**

Convex hull algorithm

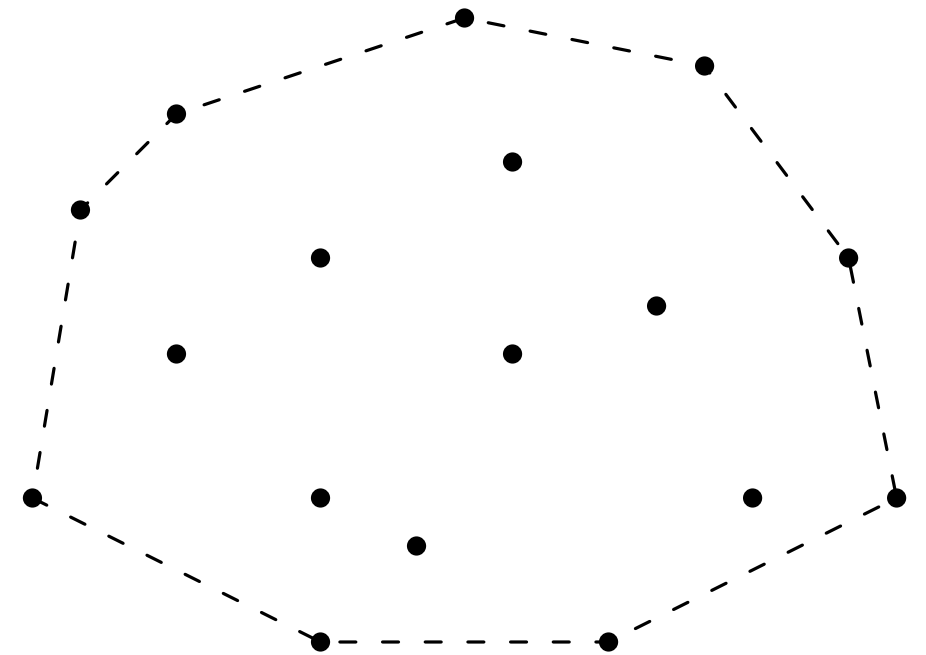
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Property: The vertices of the convex hull are always points from the input

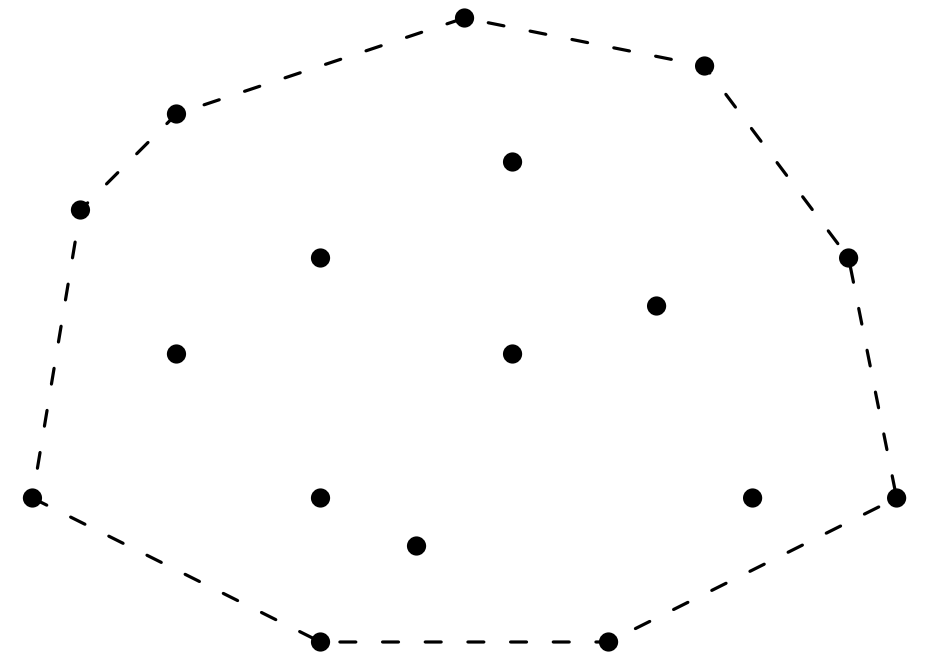


Convex hull algorithm

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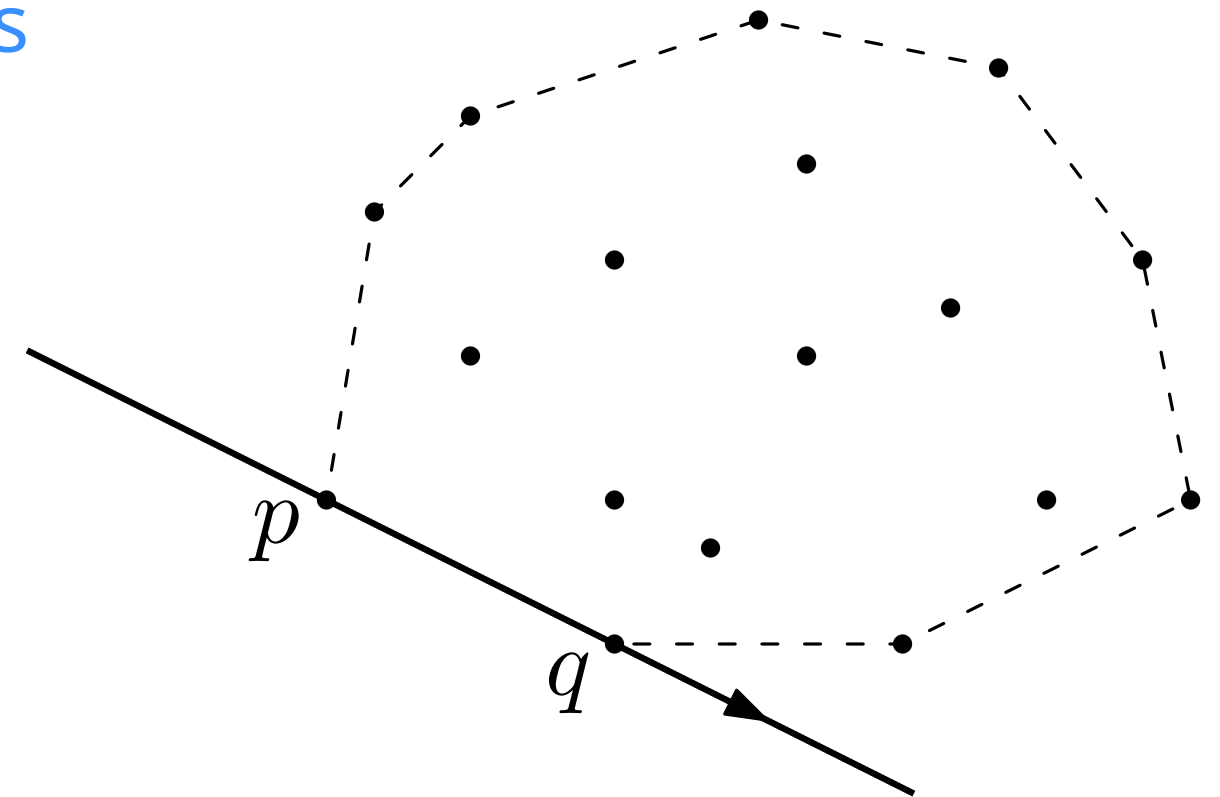


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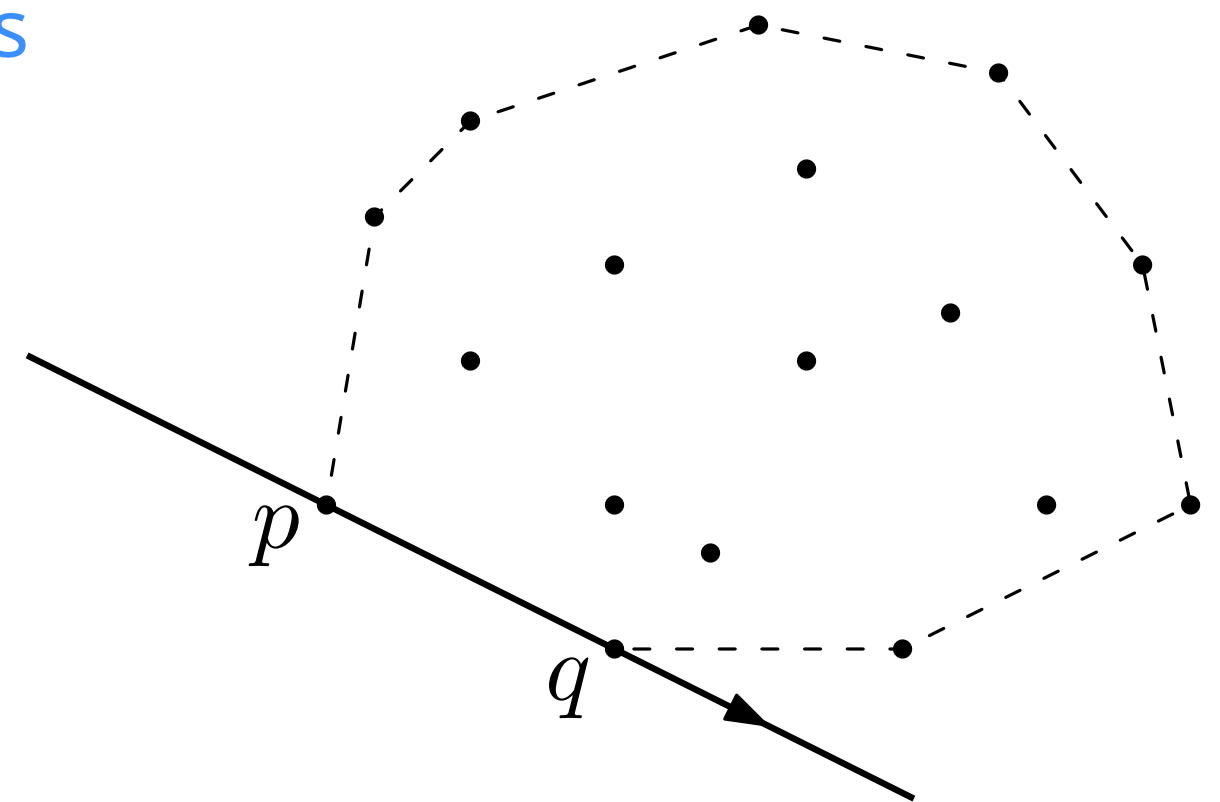


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all points lie left of the directed line from p to q (if the edge from p to q is a CCW convex hull edge)

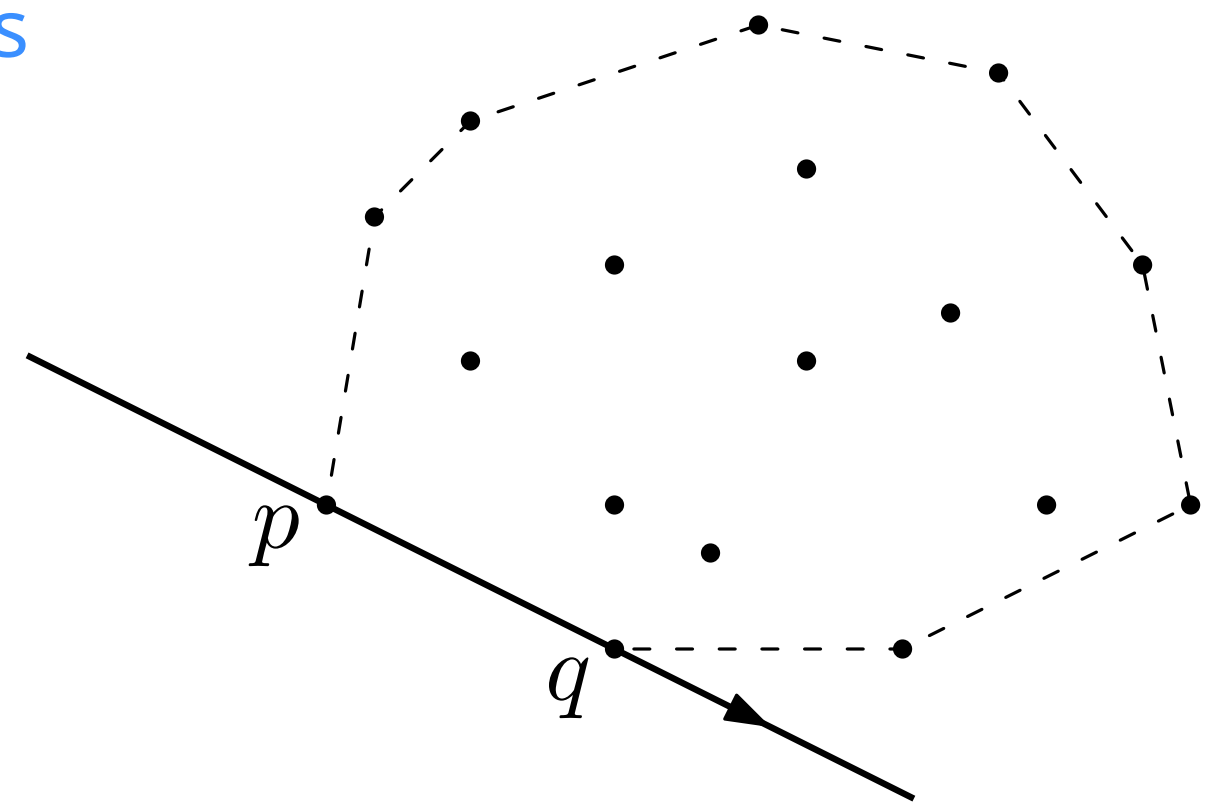
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Convex hull algorithm

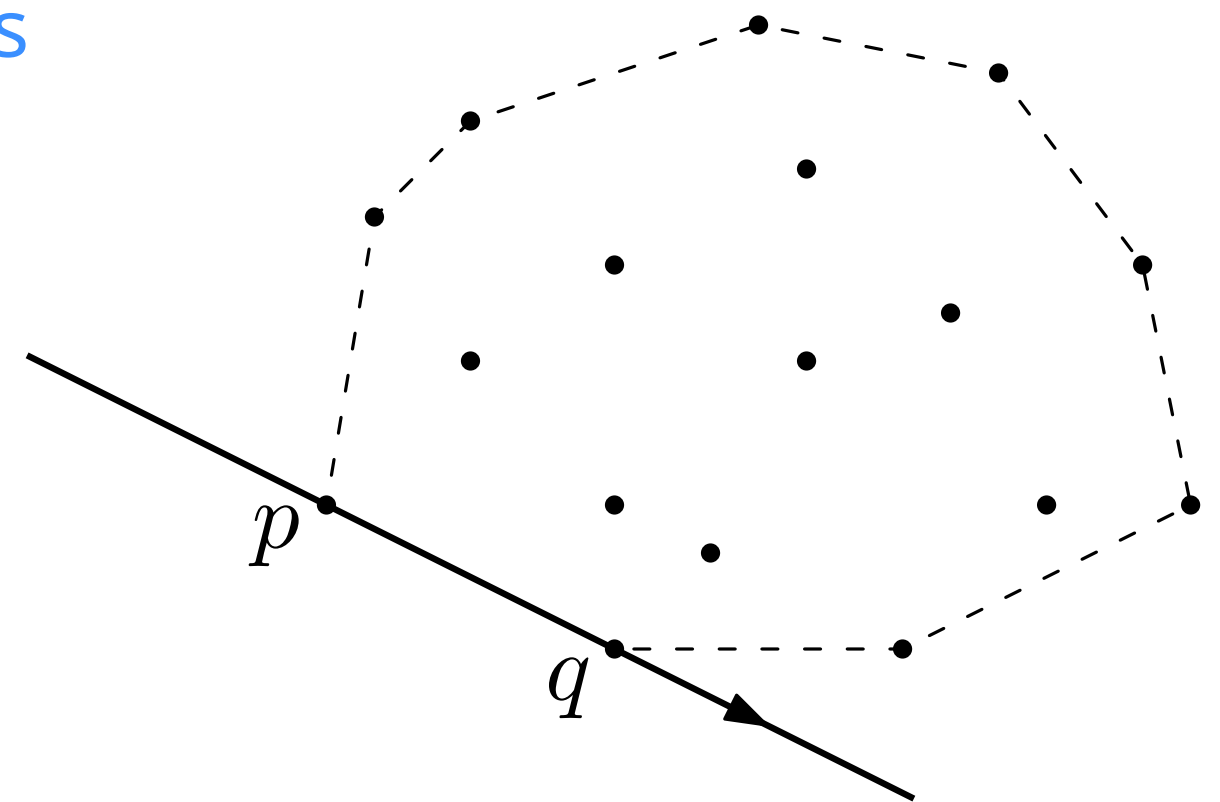
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Simple algorithm?



all points lie left of the directed line from p to q (if the edge from p to q is a CCW convex hull edge)

Convex hull algorithm

Algorithm SLOWCONVEXHULL(P)

Input: set P of distinct points in the plane

Output: list L with vertices of $CH(P)$ in counter-clockwise order

- 1: $E \leftarrow \emptyset$
- 2: **for all** ordered pairs $(p, q) \in P \times P$ with $p \neq q$ **do**
- 3: $valid \leftarrow true$
- 4: **for all** points $r \in P$ not equal to p or q **do**
- 5: **if** r lies right of the directed line from p to q **then**
- 6: $valid \leftarrow false$
- 7: **if** $valid$ **then**
- 8: add directed edge \overrightarrow{pq} to E
- 9: from the set E of edges construct a list L of vertices of $CH(P)$, sorted in counter-clockwise order

Questions to keep in mind

Question: How must line 5 (if r lies right of \overline{pq} ...) be interpreted to make the algorithm correct, i.e., how do we handle degeneracies?

Question: How **efficient** is the algorithm?

Question: Is the algorithm **robust** against rounding errors?

Note: Robustness is a huge issue when implementing geometric algorithms

Convex hull algorithm 2

Another approach: incremental, from left to right

Convex hull algorithm 2

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Let's first compute the **upper boundary** of the convex hull this way (property: on the upper hull, points appear in x -order)

Convex hull algorithm 2

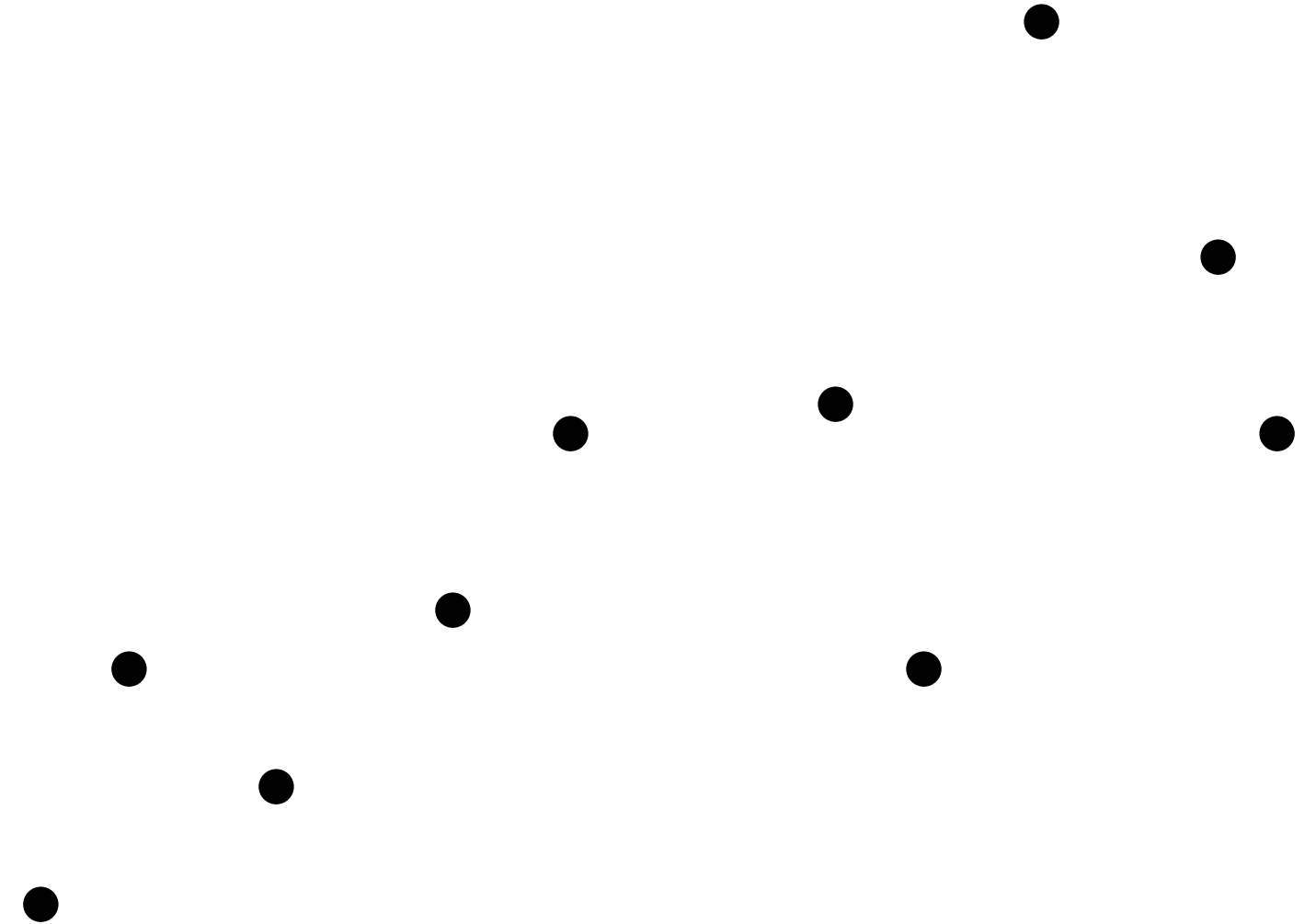
Another approach: incremental, from left to right

Let's first compute the **upper boundary** of the convex hull this way (property: on the upper hull, points appear in x -order)

Main idea: Sort the points from left to right (= by x -coordinate). Then insert the points in this order, and maintain the upper hull so far

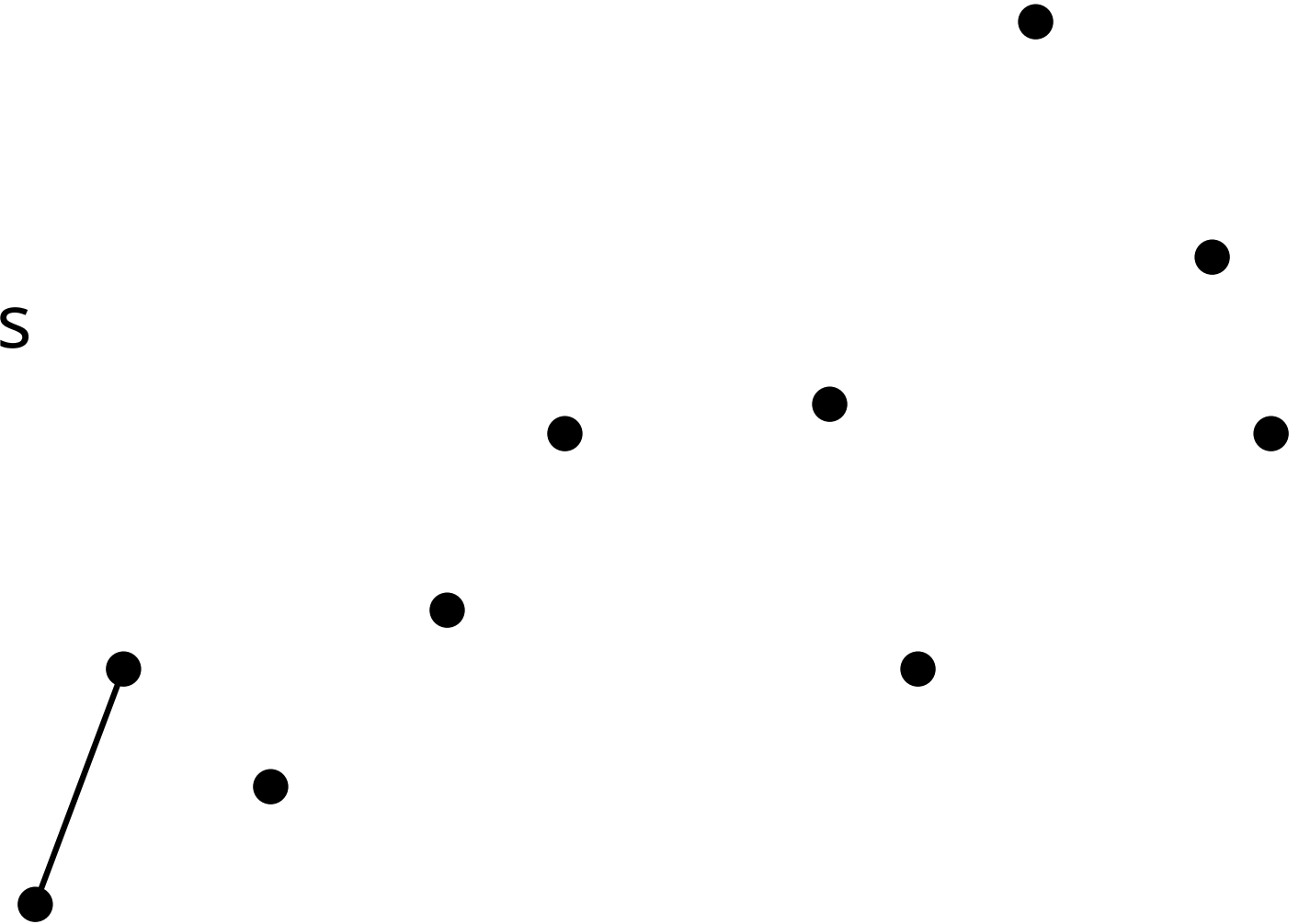
Convex hull algorithm 2

Observation: from left to right, there are only right turns on the upper hull



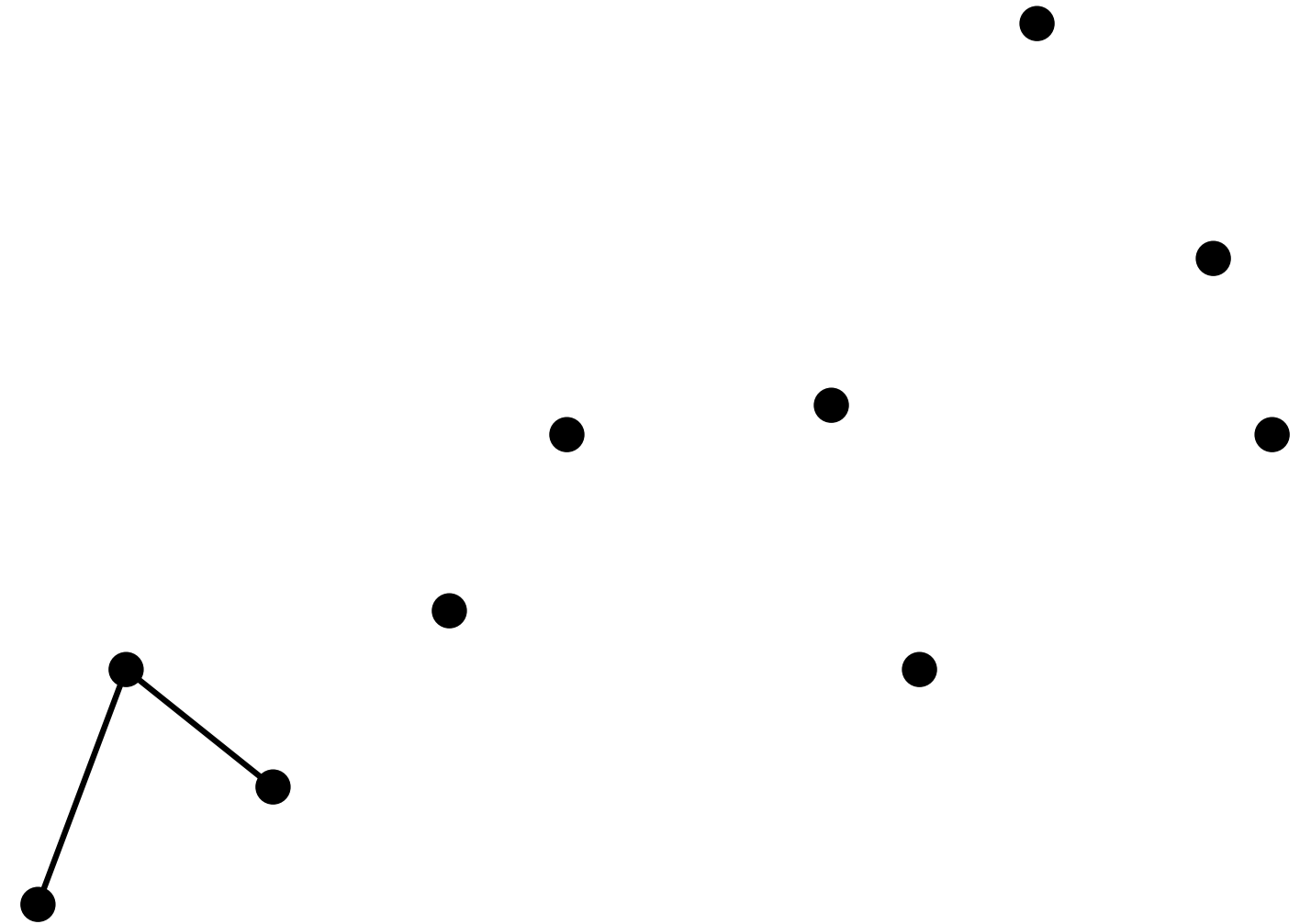
Convex hull algorithm 2

Initialize by inserting the leftmost two points



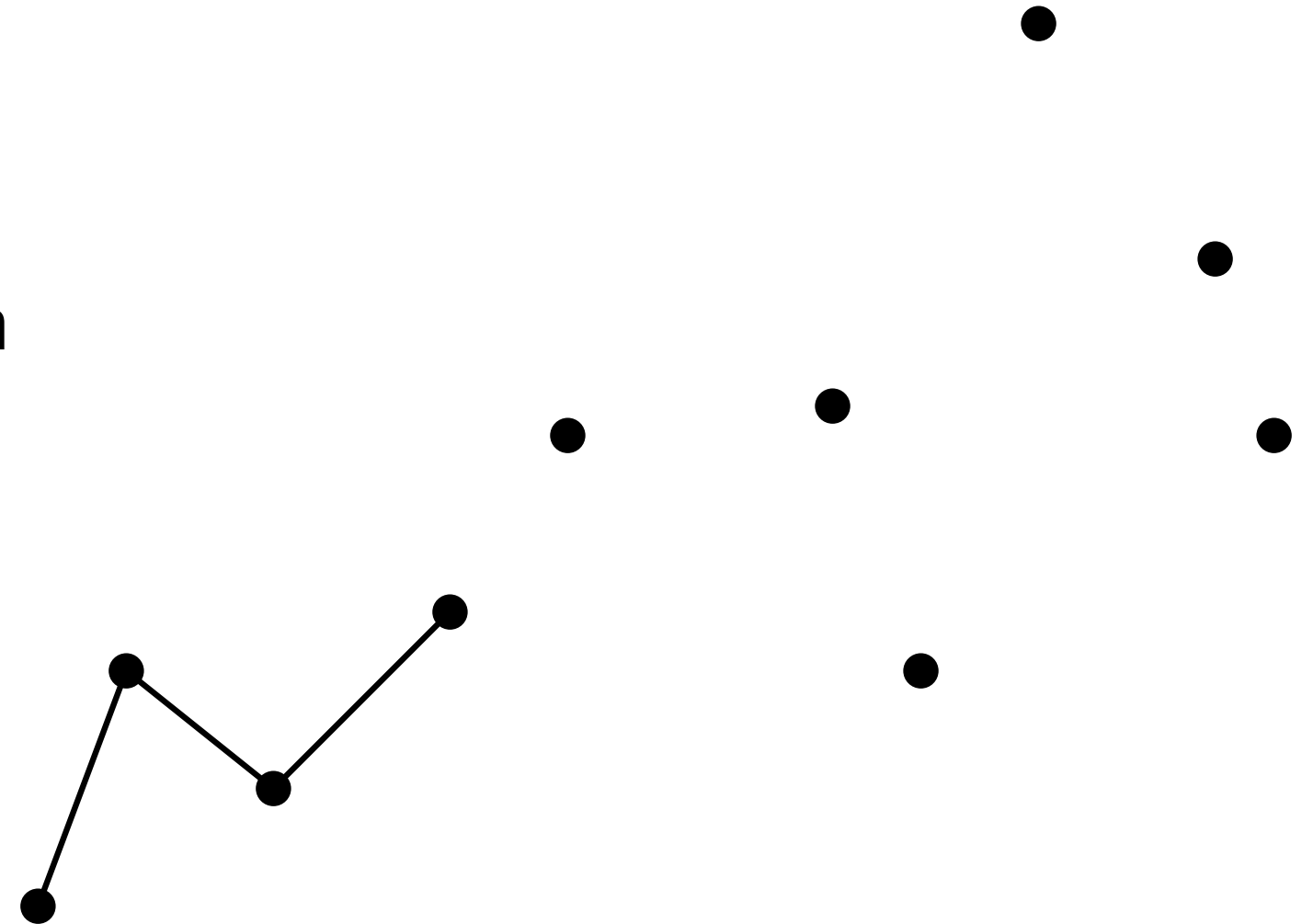
Convex hull algorithm 2

When we consider the third point there will be a right turn at the previous point, so we add it



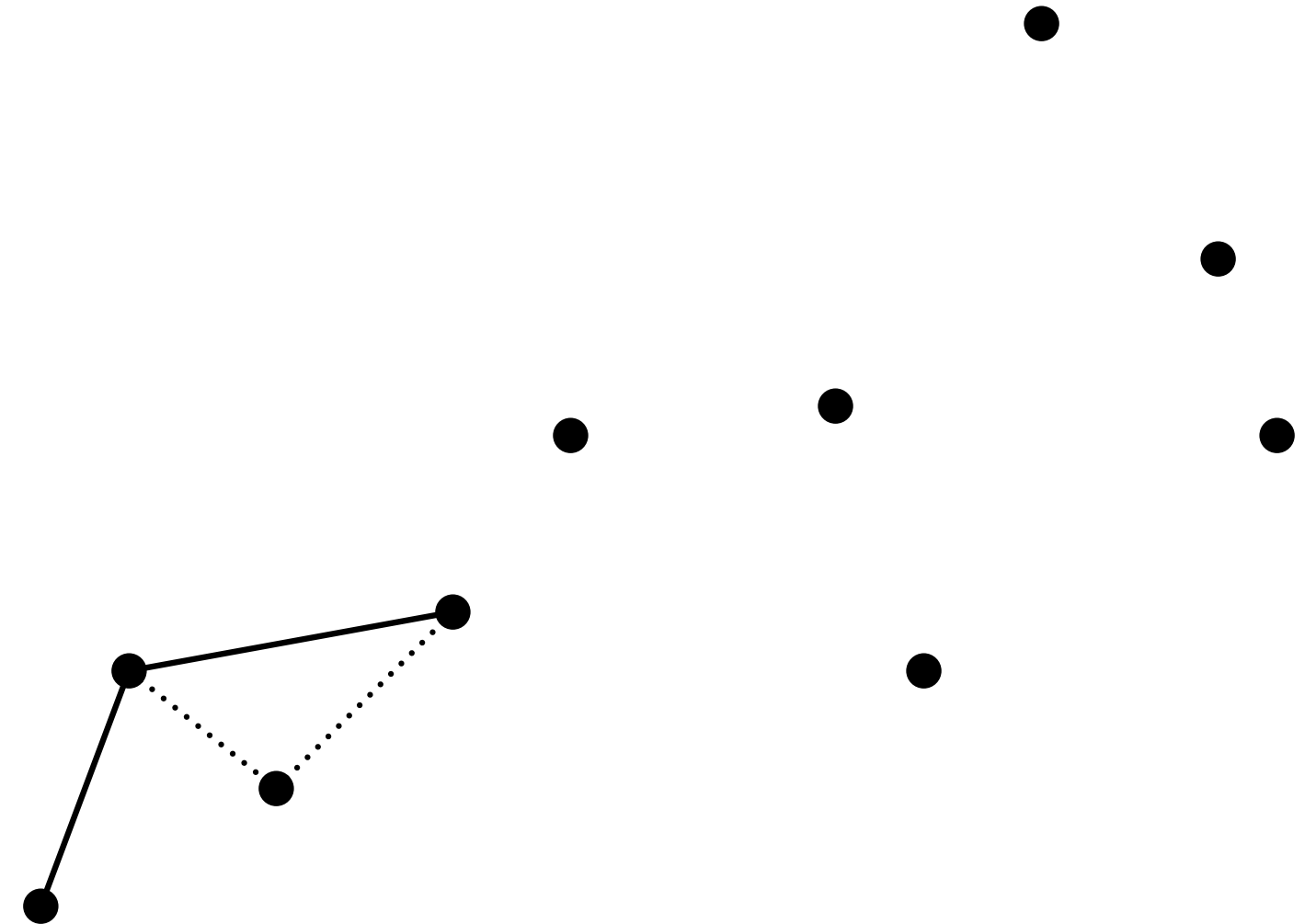
Convex hull algorithm 2

When we consider the fourth point we get a left turn at the third point



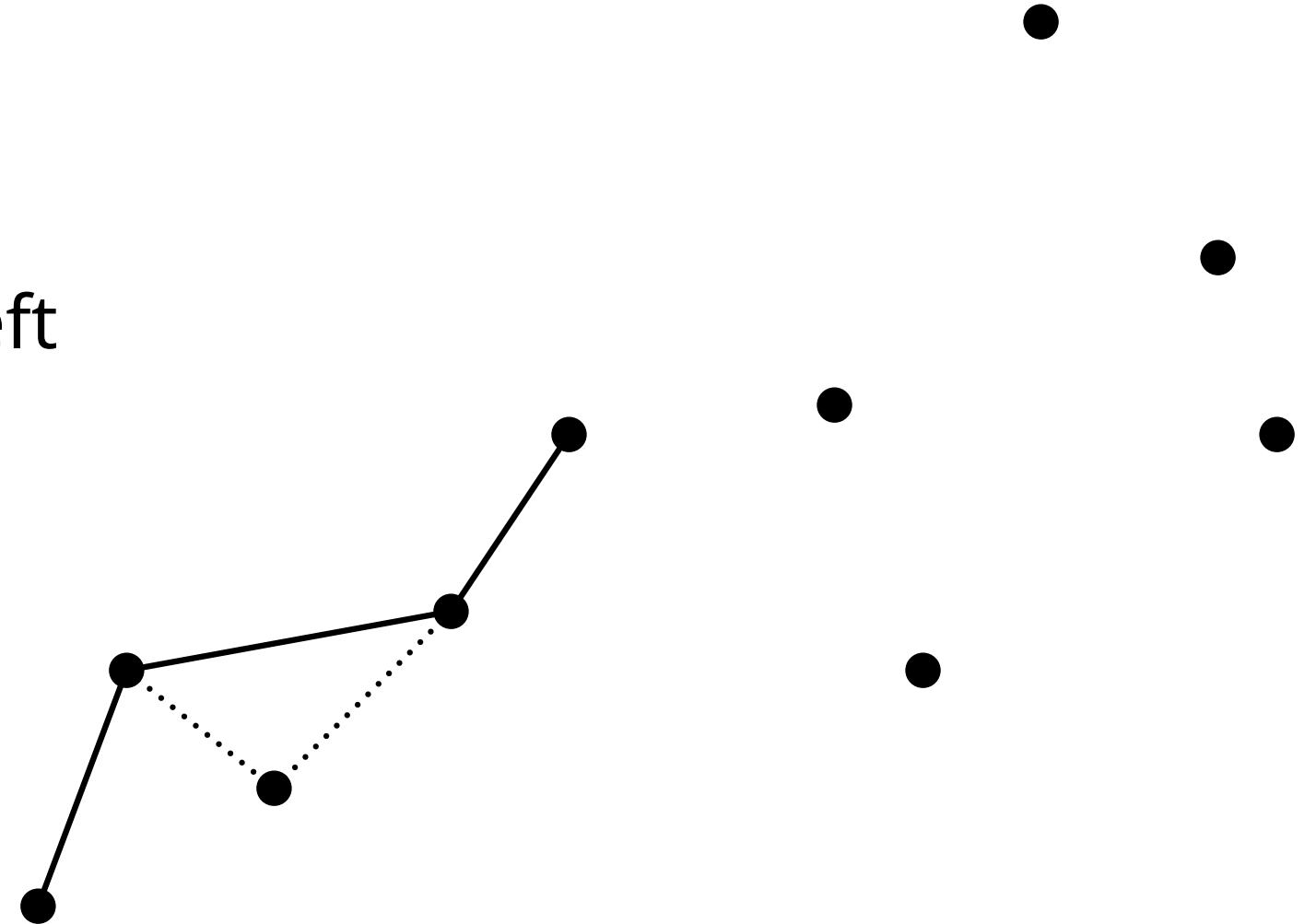
Convex hull algorithm 2

...so we remove the third point from the upper hull when we add the fourth



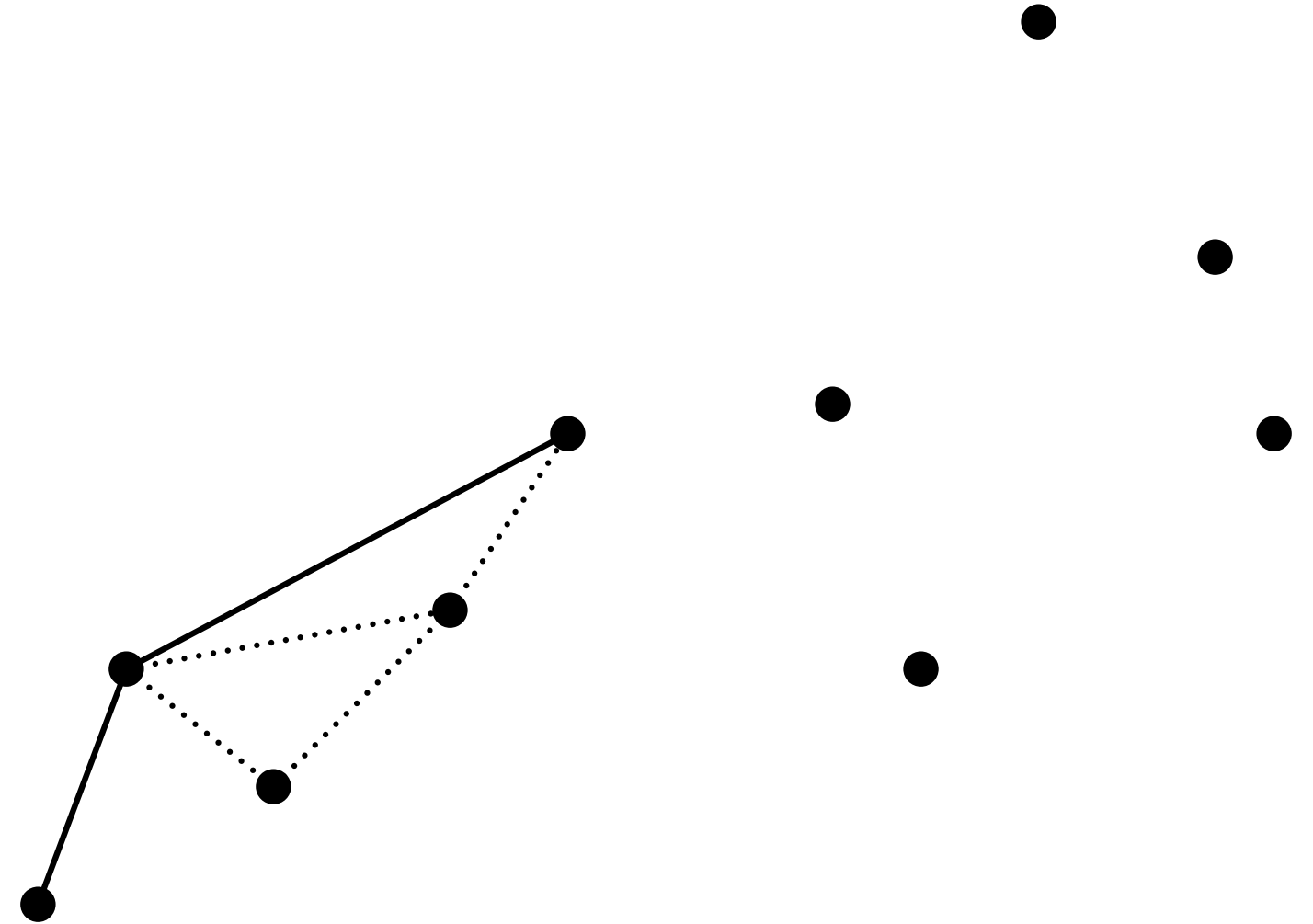
Convex hull algorithm 2

When we consider the fifth point we get a left turn at the fourth point



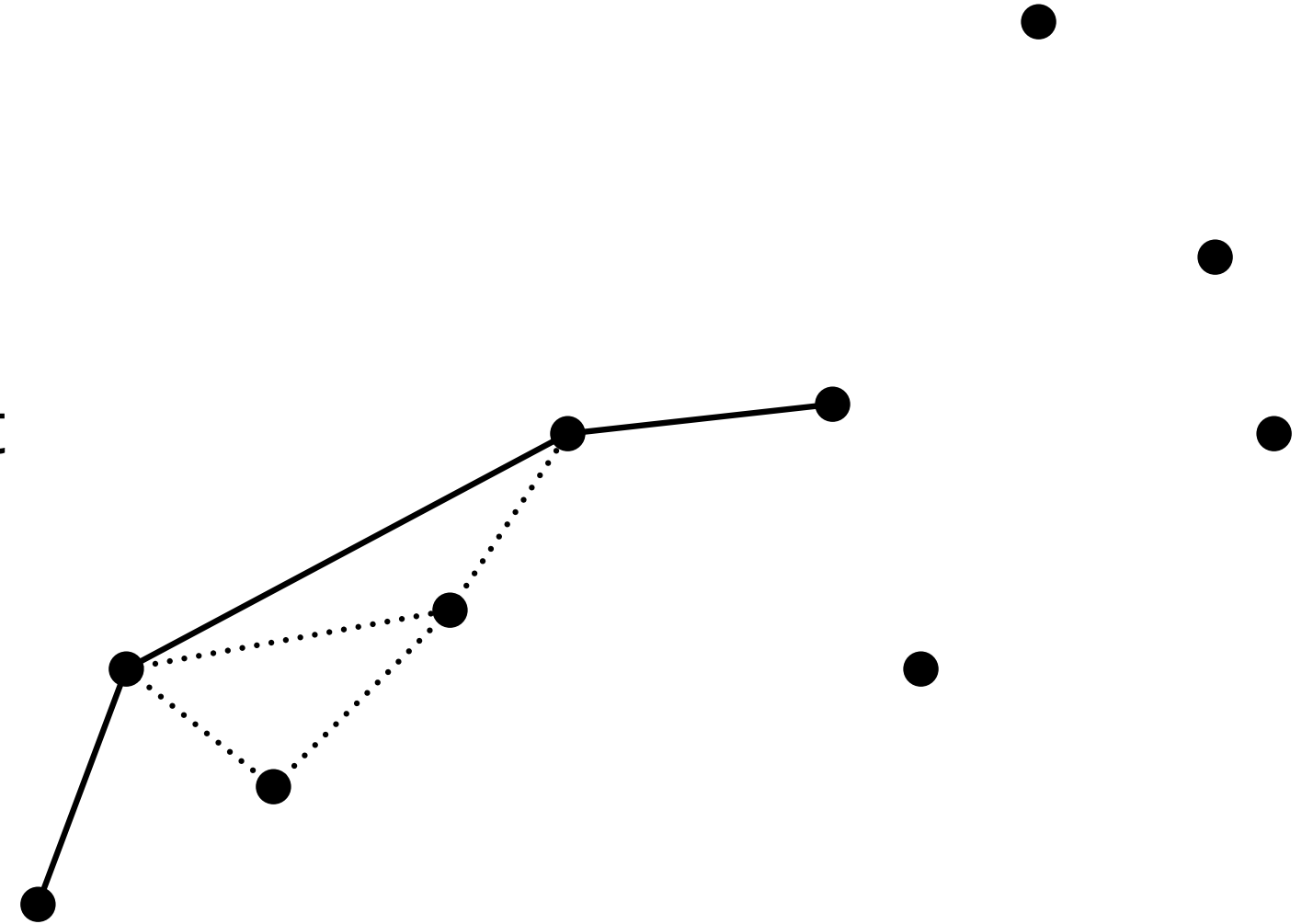
Convex hull algorithm 2

...so we remove the fourth point when we add the fifth



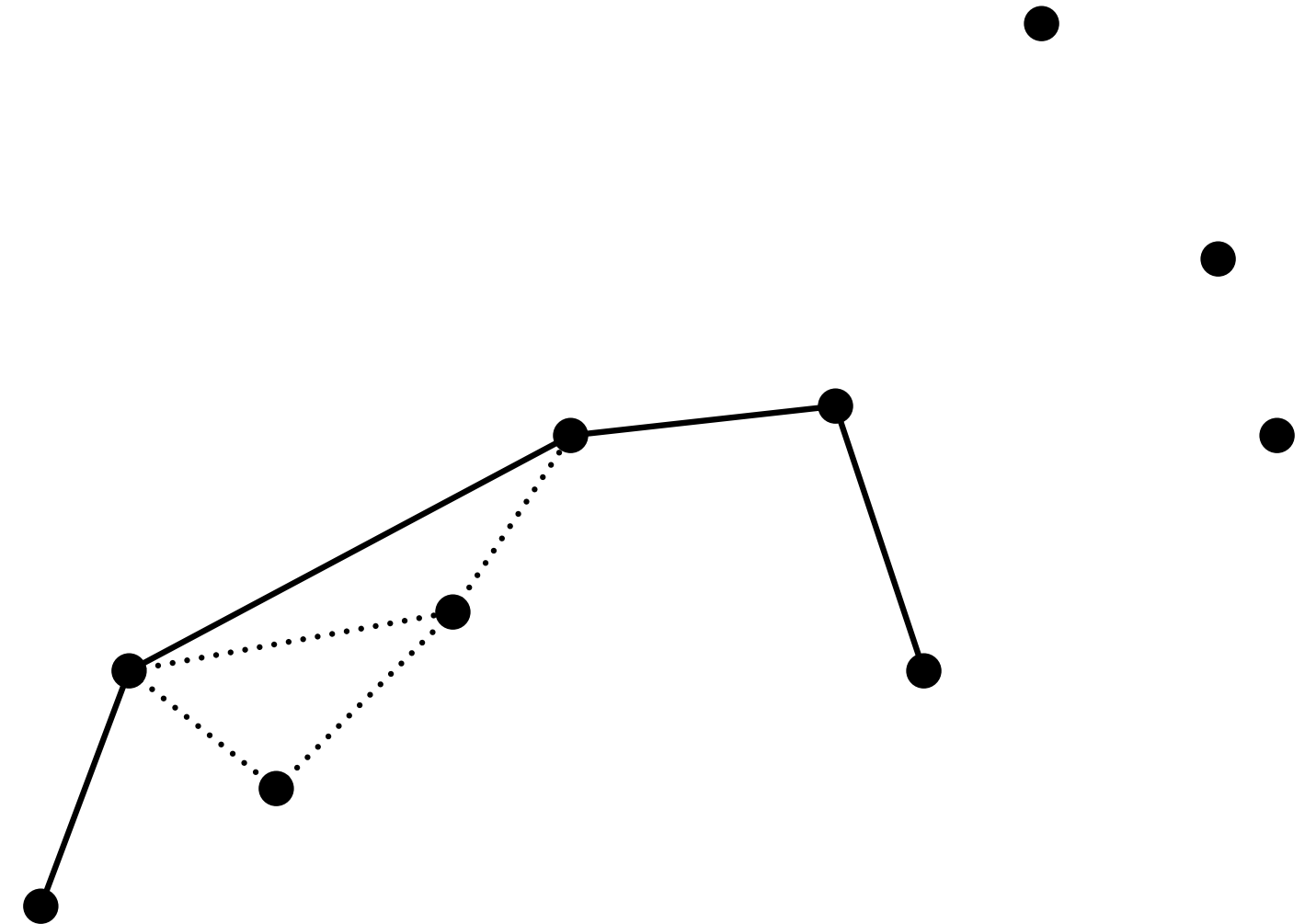
Convex hull algorithm 2

When we consider the sixth point we get a right turn at the fifth point, so we just add it



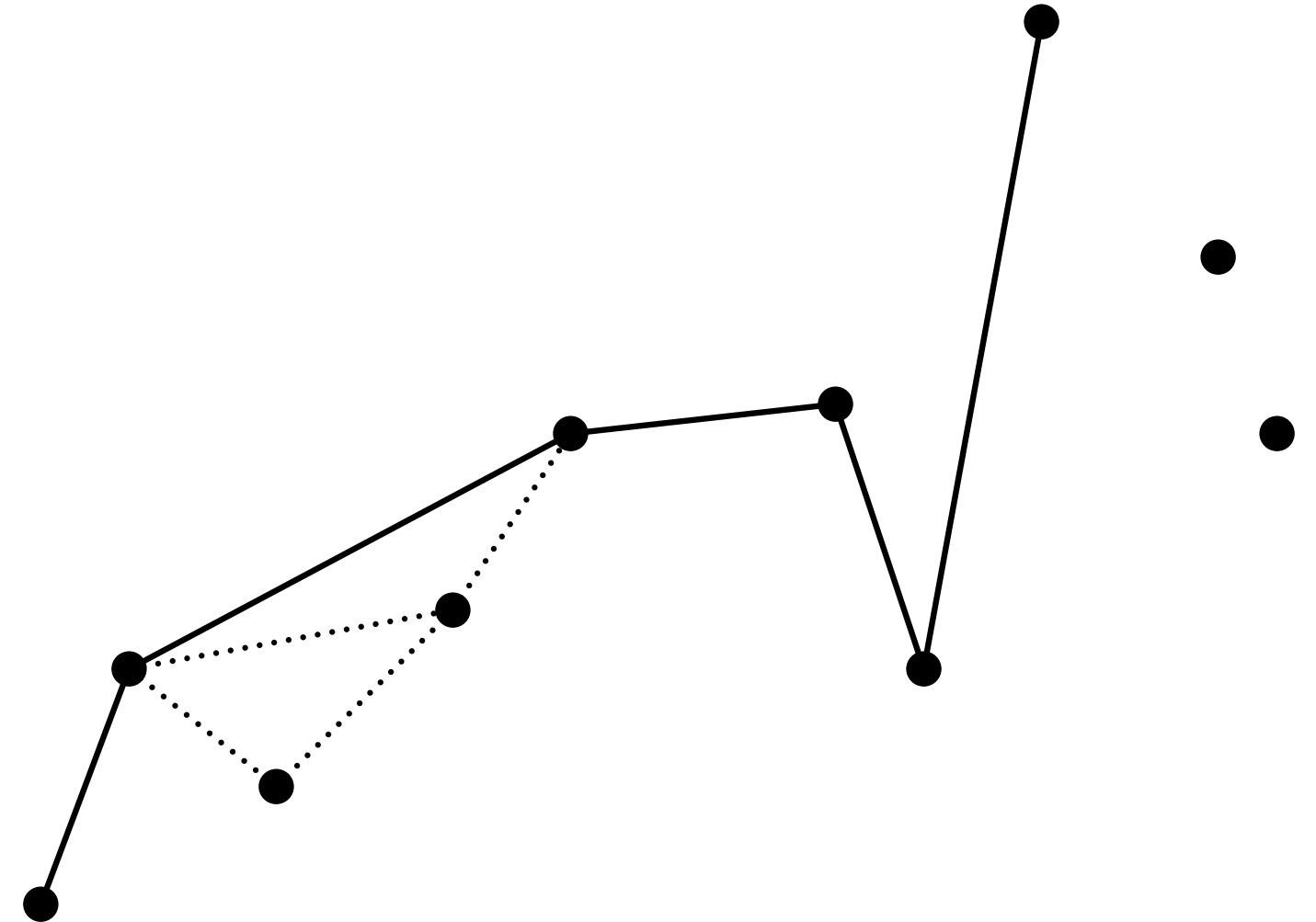
Convex hull algorithm 2

We also just add the seventh point



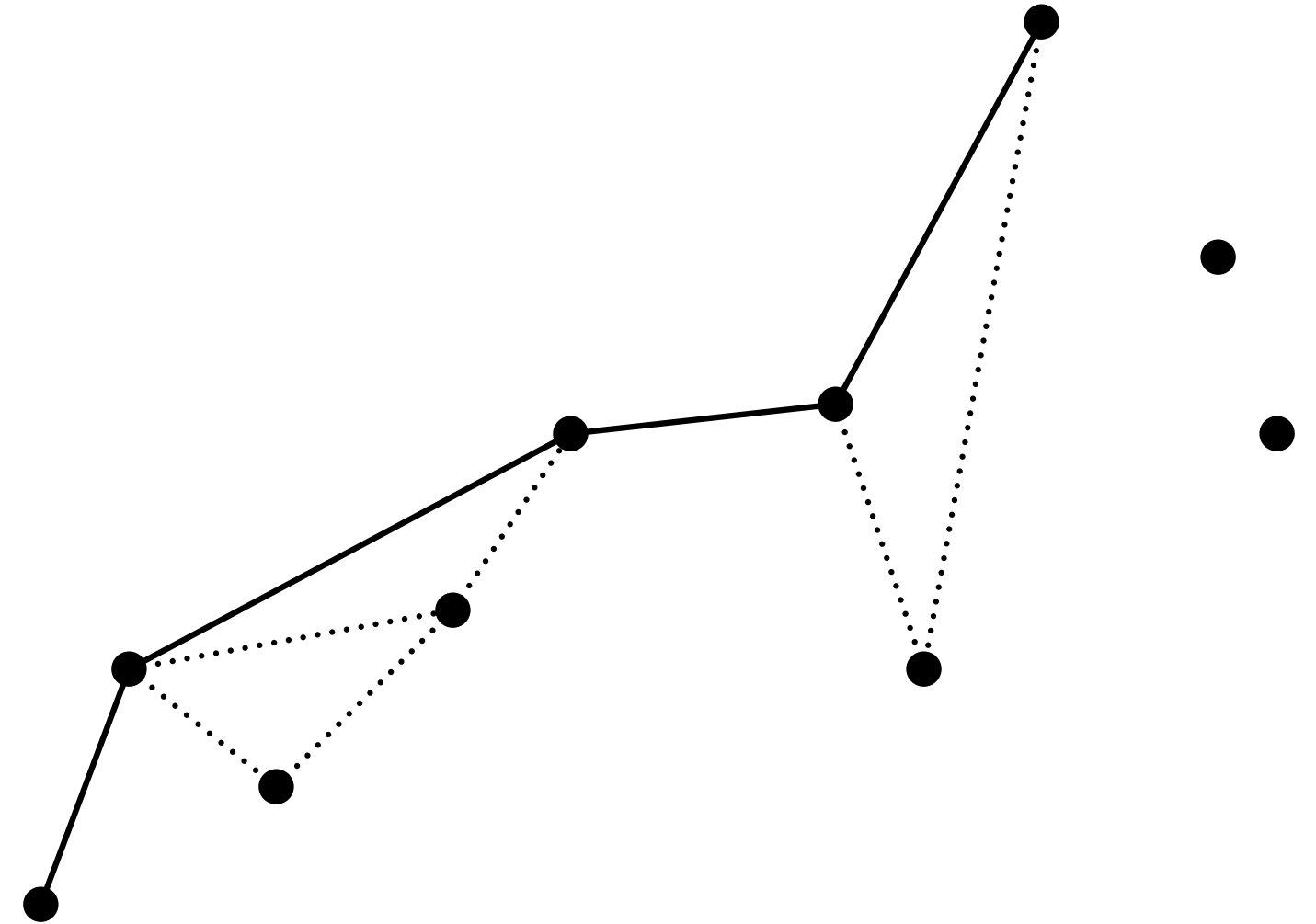
Convex hull algorithm 2

When considering the eighth point...



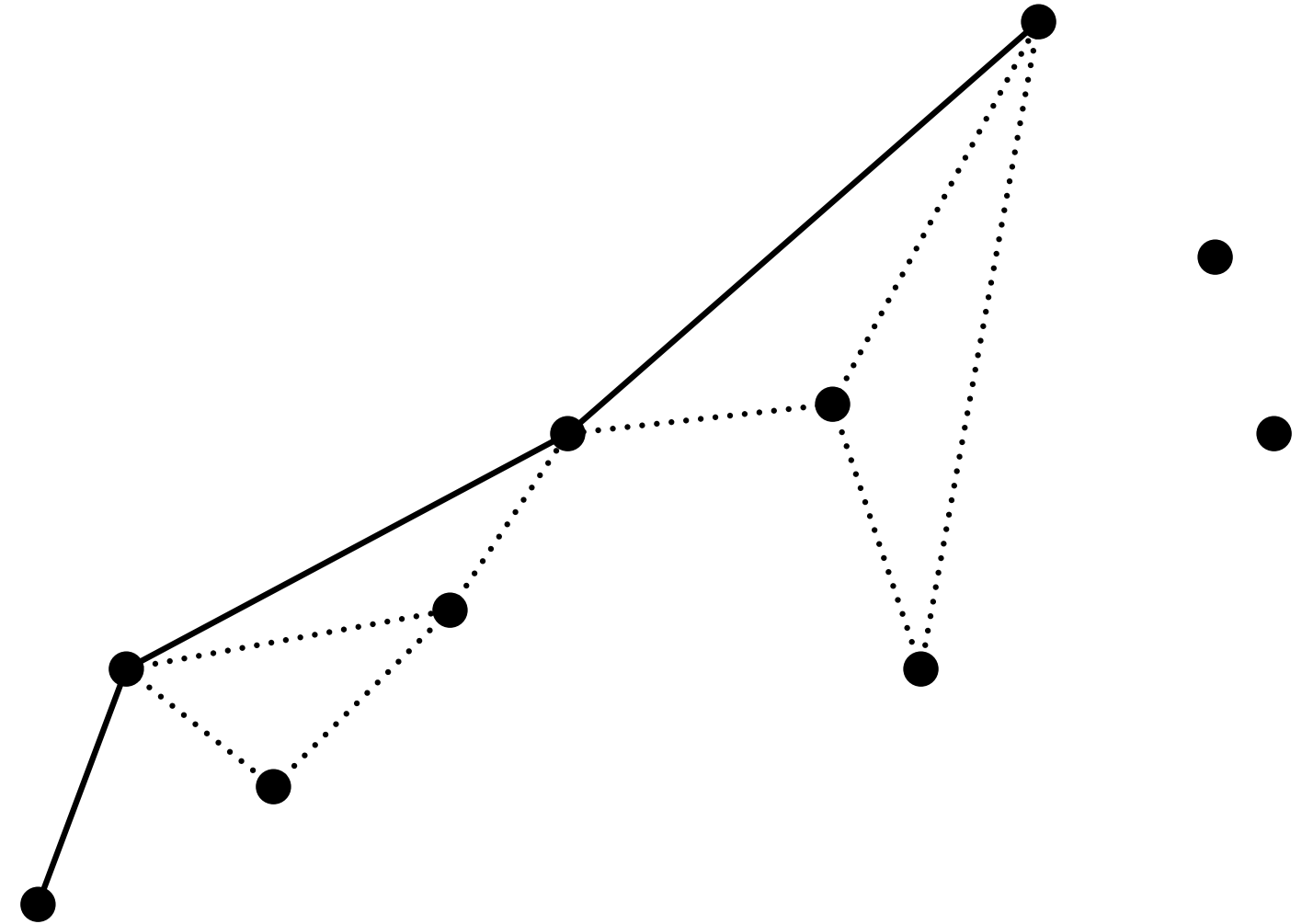
Convex hull algorithm 2

...we remove the seventh point



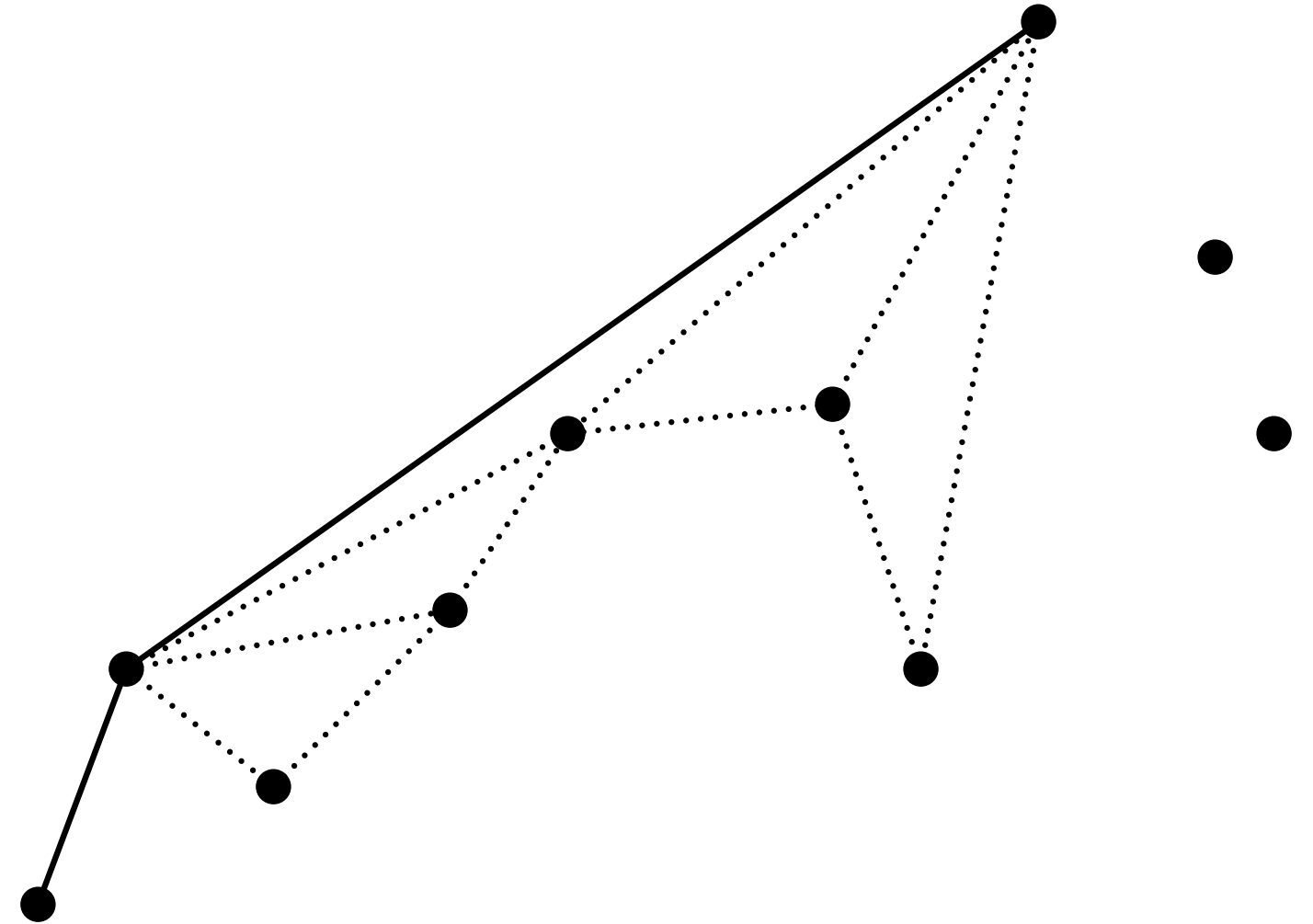
Convex hull algorithm 2

...and also the sixth point



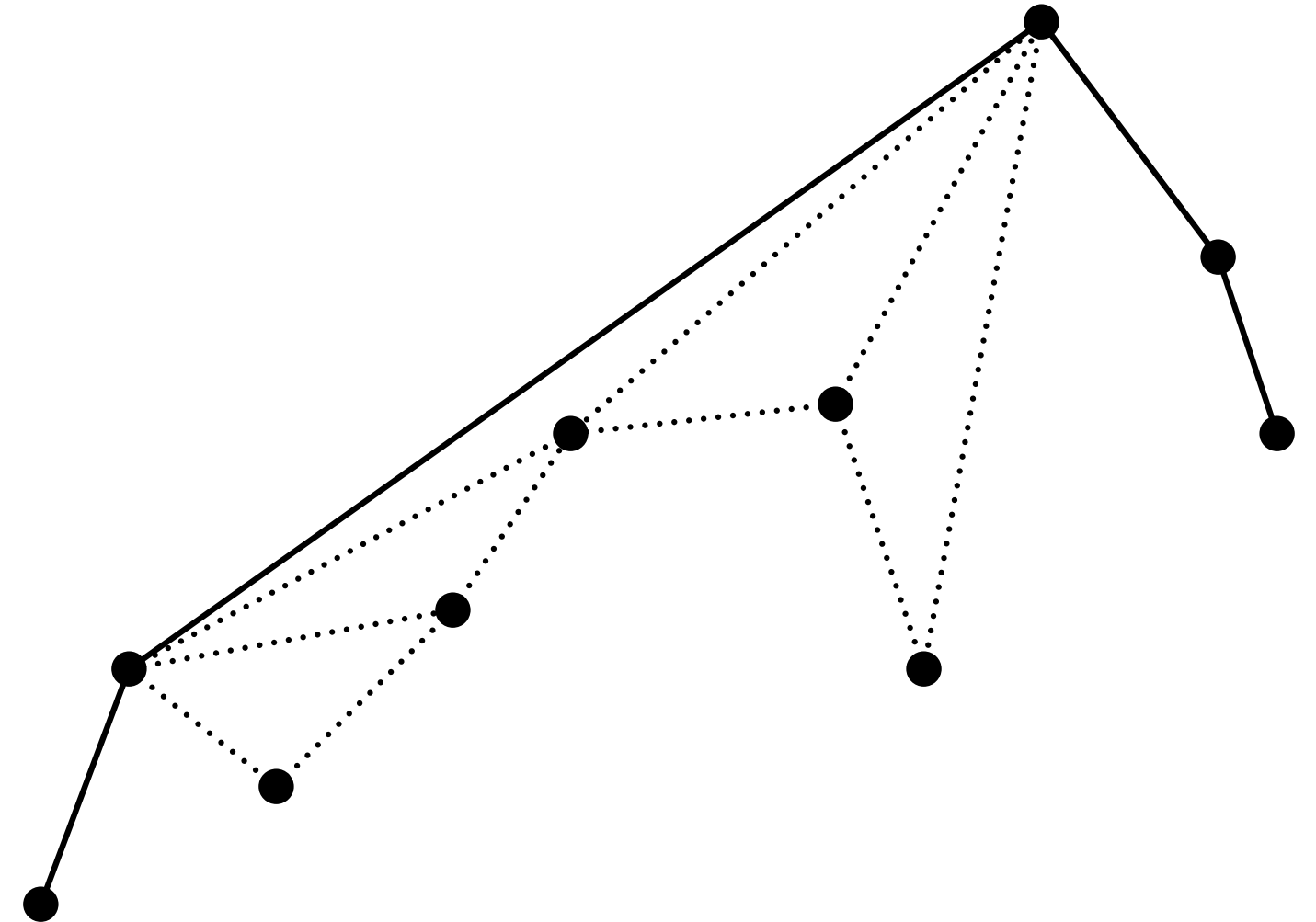
Convex hull algorithm 2

...and also the fifth point



Convex hull algorithm 2

after two more steps we get the upper hull



Convex hull algorithm 2: Graham Scan

Algorithm GRAHAMSCAN(P)

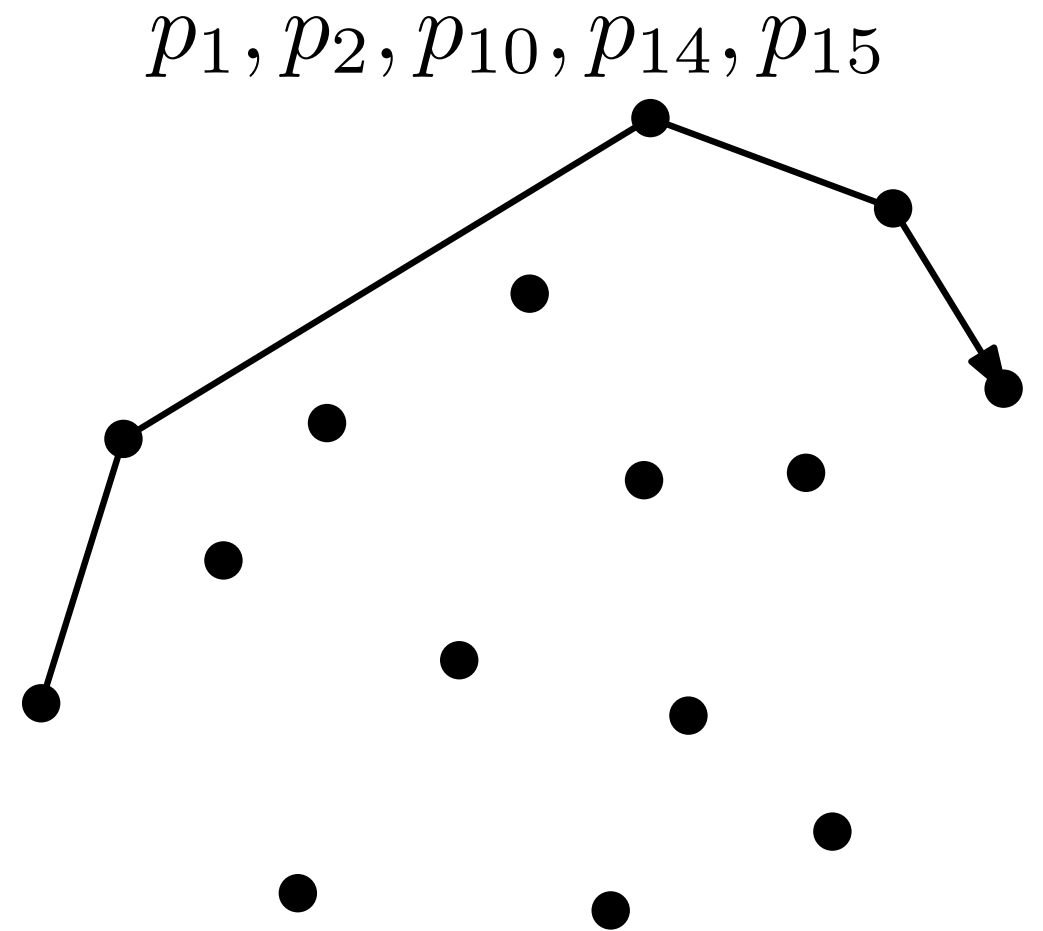
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Convex hull algorithm 2: Graham Scan

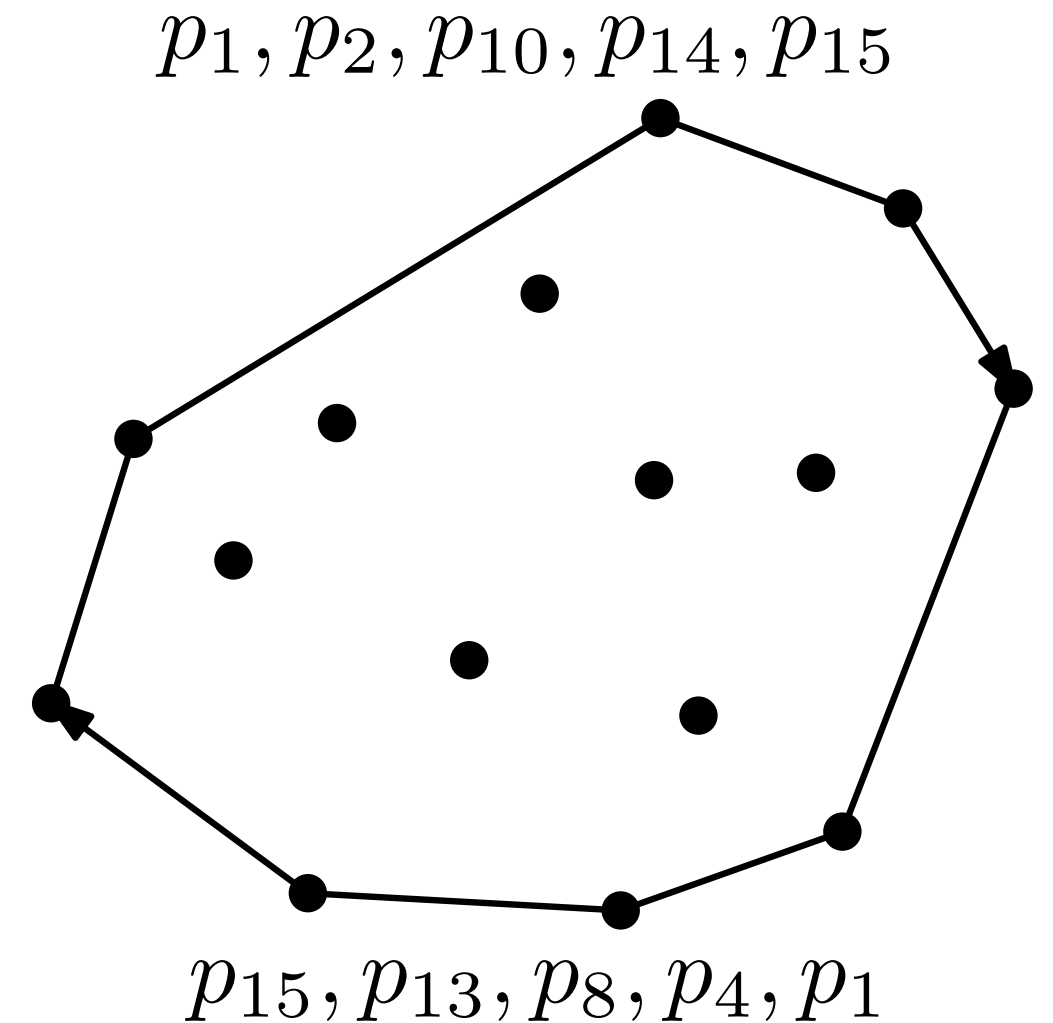
We have computed the upper convex hull



Convex hull algorithm 2: Graham Scan

We have computed the upper convex hull

Then we do the same for the lower convex hull,
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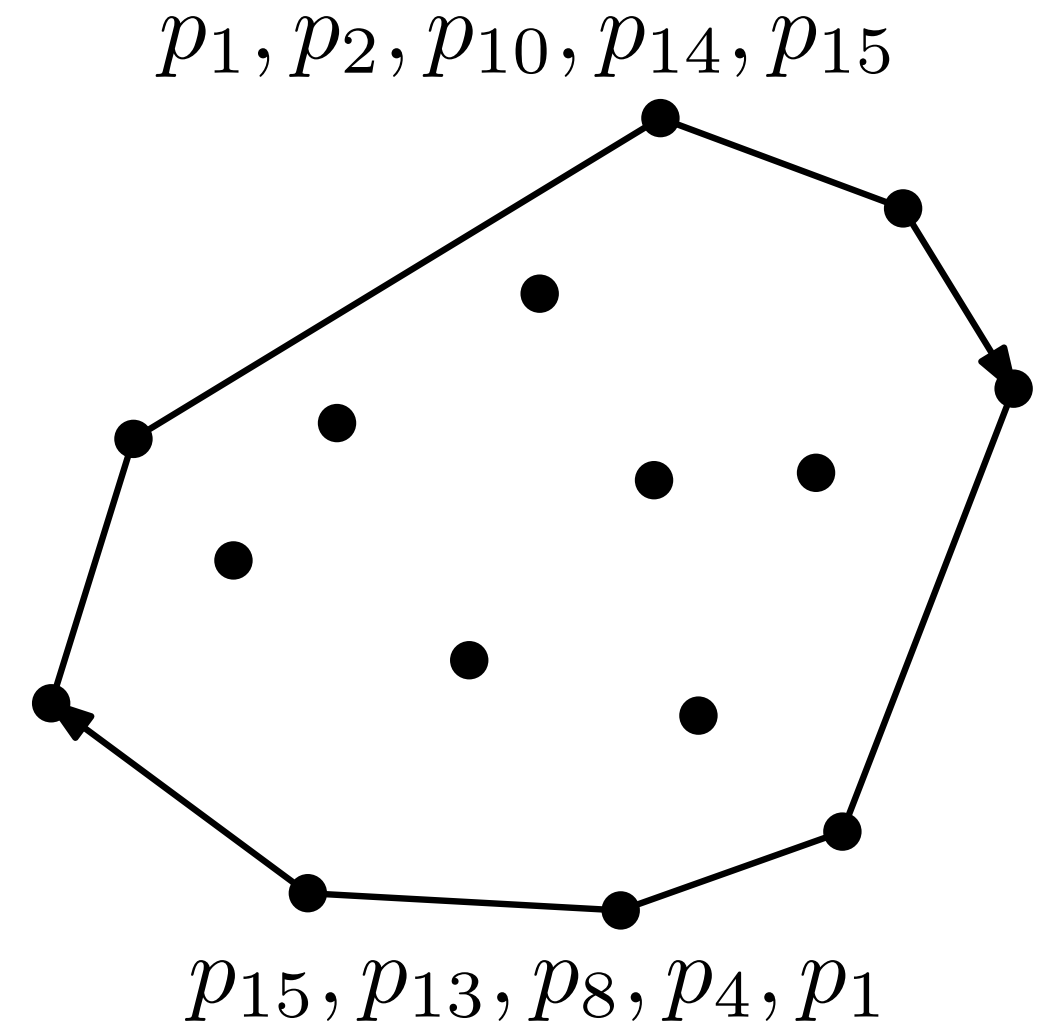


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We remove the first and last points of the lower
convex hull

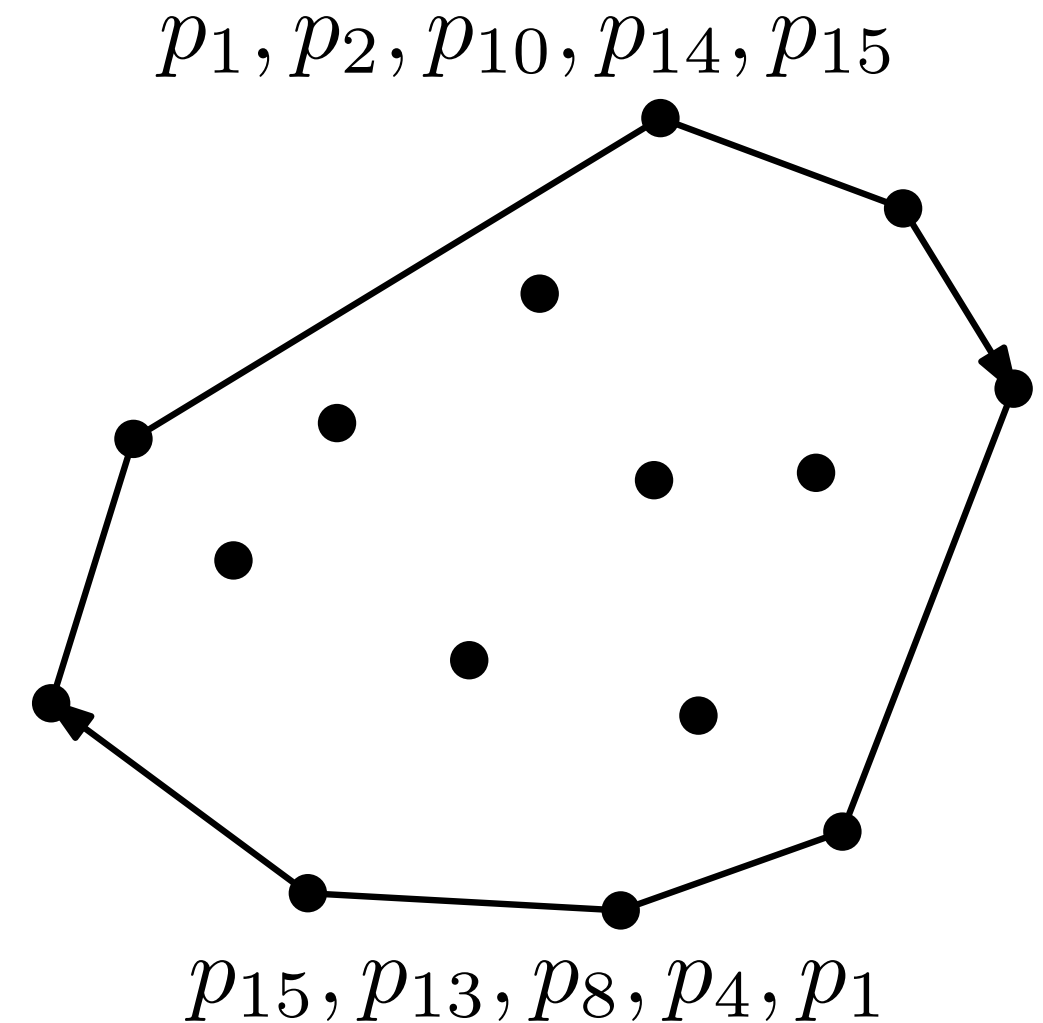


Convex hull algorithm 2: Graham Scan

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Then we do the same for the lower convex hull,
from right to left

We remove the first and last points of the lower
convex hull ...and concatenate the two lists in one



Algorithm analysis

Algorithm analysis generally has two components

- proof of correctness
- efficiency analysis, proof of running time

Correctness

Are the **general observations** on which the algorithm is based correct?

Does the algorithm handle **degenerate cases** correctly?

Correctness

Are the **general observations** on which the algorithm is based correct?

Does the algorithm handle **degenerate cases** correctly?

Here:

- Does the sorted order matter if two or more points have the same x -coordinate?
- What happens if there are three or more collinear points, in particular on the convex hull?

Efficiency

For **each line** of pseudocode identify

- how much time it takes
- how many times it is executed once

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Sometimes there are **global arguments** why an algorithm is more efficient than it seems at first

Efficiency

Algorithm GRAHAMSCAN(P)

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Total time:

$$O(n \log n) + O(1) + \sum_{i=3}^n (O(1) + k_i \cdot O(1)) =$$

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$$O(n \log n) + O(1) + \sum_{i=3}^n (O(1) + k_i \cdot O(1)) = O(n \log n) + \sum_{i=3}^n O(1 + k_i)$$

Efficiency: attempt 1

Total time: $O(n \log n) + \sum_{i=3}^n O(1 + k_i)$

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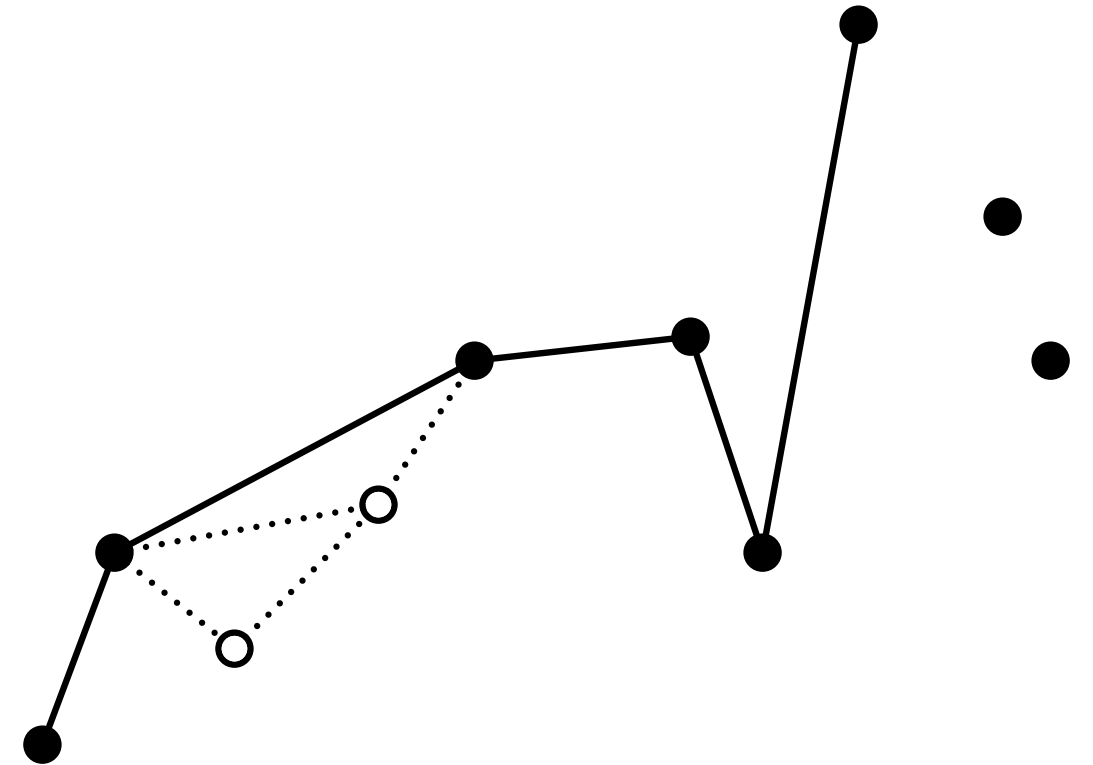
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Question: Is this analysis tight?

Sometimes there are **global arguments** why an algorithm is more efficient than it seems at first

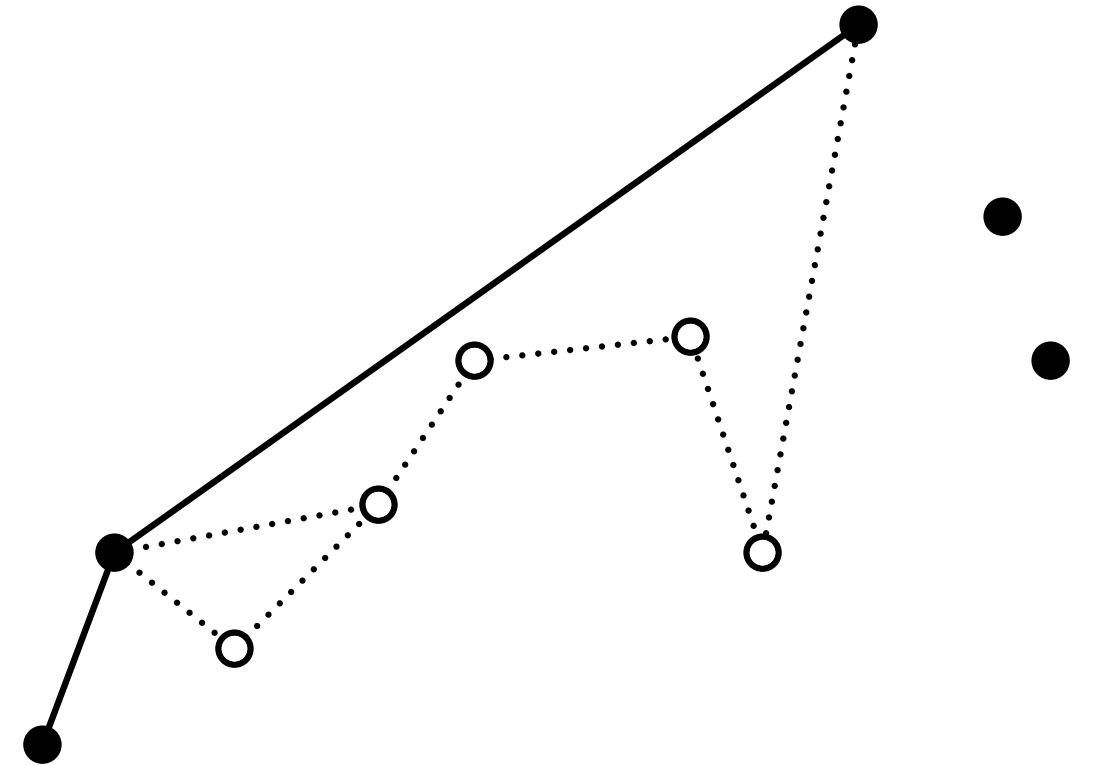
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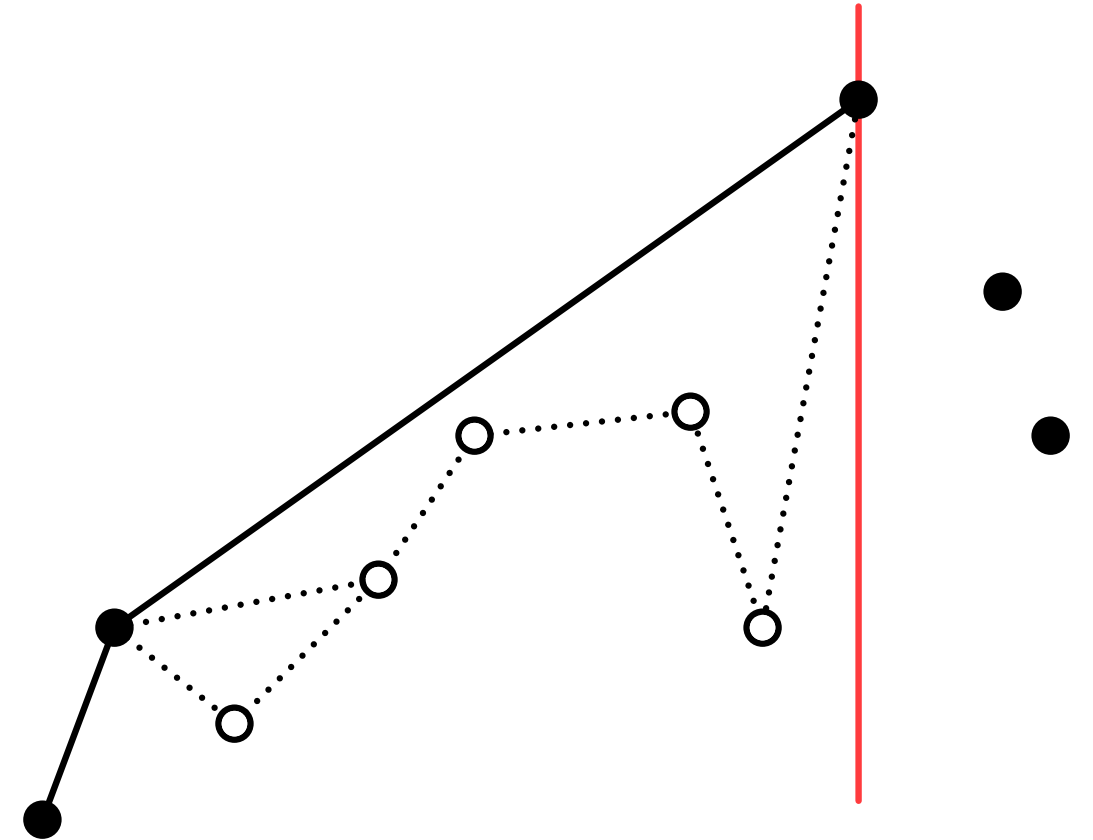
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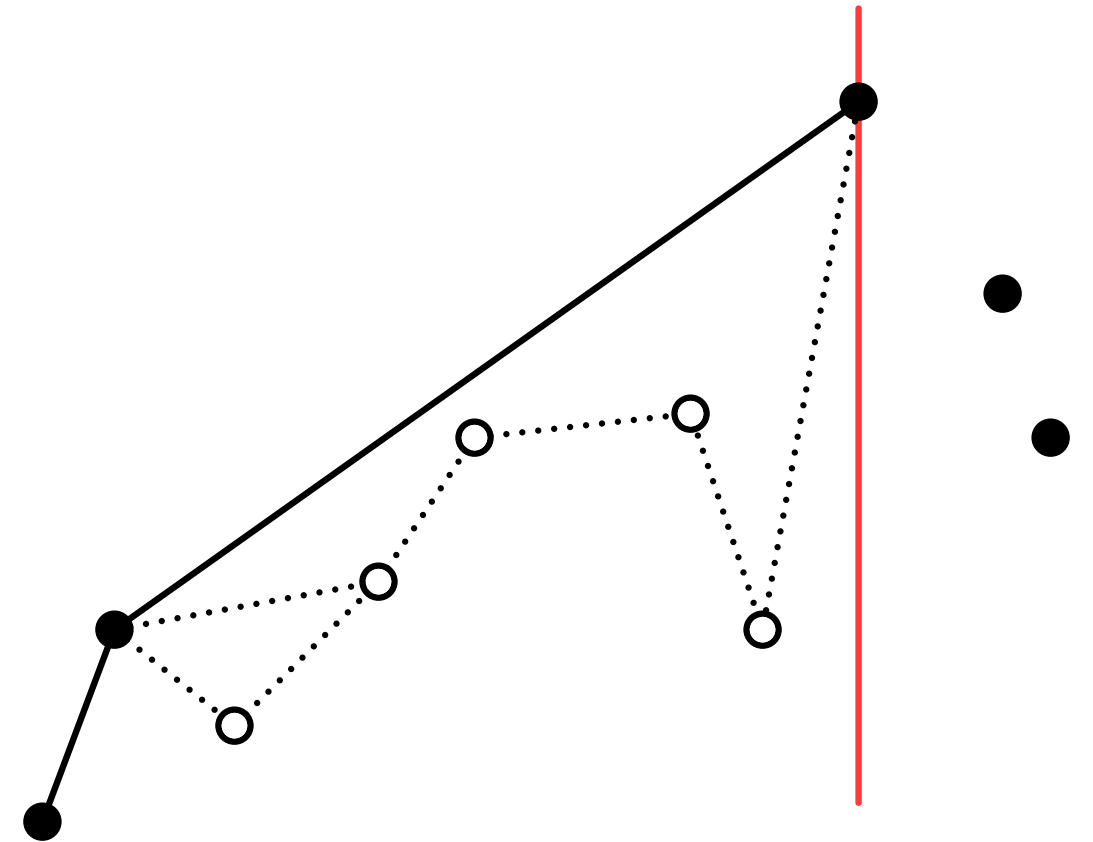
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Global argument: each point can be removed only once from the upper hull

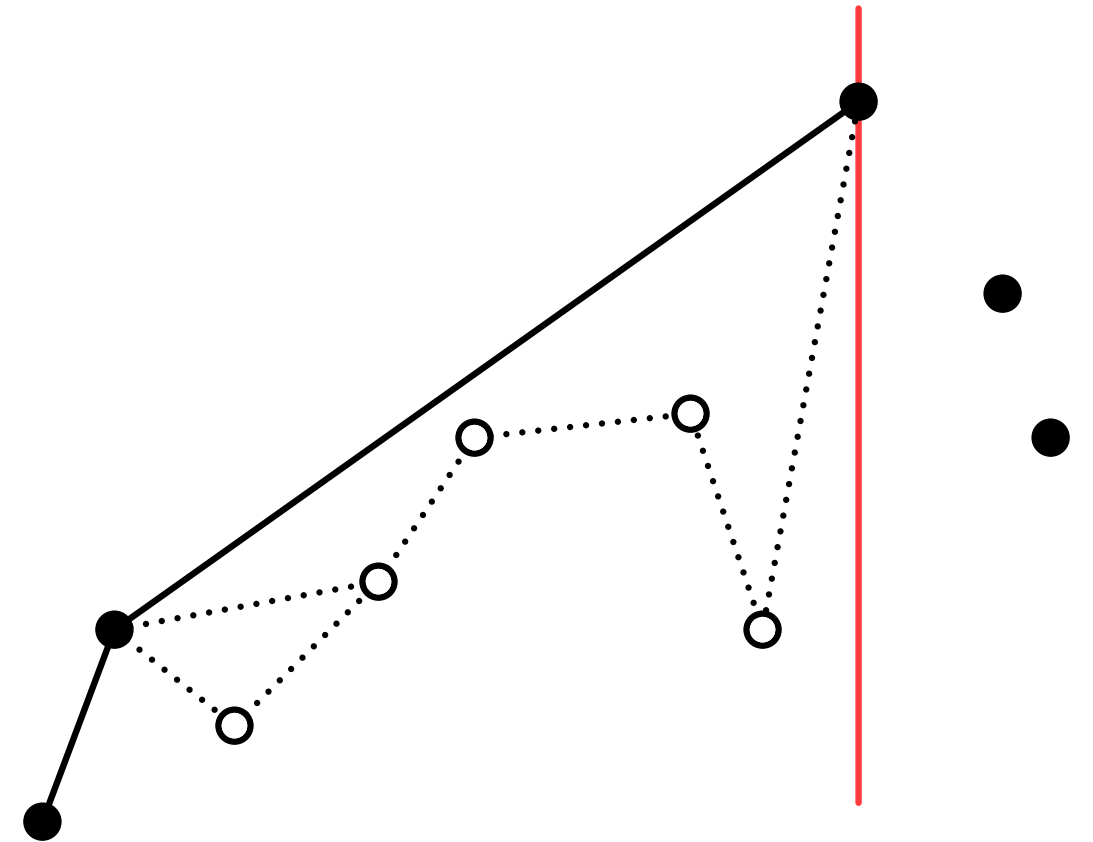


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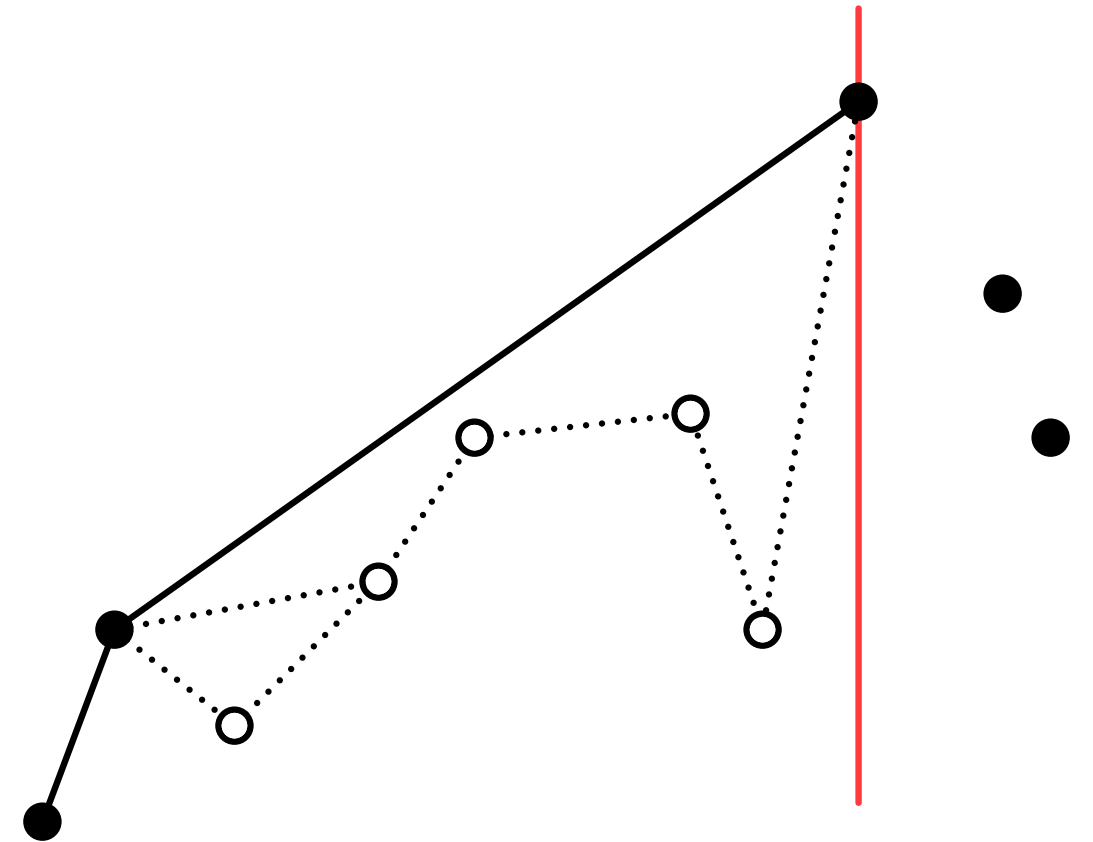
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Hence,

$$O(n \log n) + \sum_{i=3}^n O(1 + k_i) = O(n \log n) + O(n) = O(n \log n)$$



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The convex hull of a set of n points in the plane can be computed in $O(n \log n)$ time, and this is worst-case optimal.

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Question: Can we do better?

$O(nh)$ if the convex hull has h vertices?

$O(n \log h)$? **Yes we can!**

An output-sensitive algorithm

Idea: start on the convex hull and wrap around the convex hull

An output-sensitive algorithm

Algorithm GIFTWRAPPING(P)

Input: set P of points in the plane

Output: list L containing vertices of $CH(P)$ in counterclockwise order

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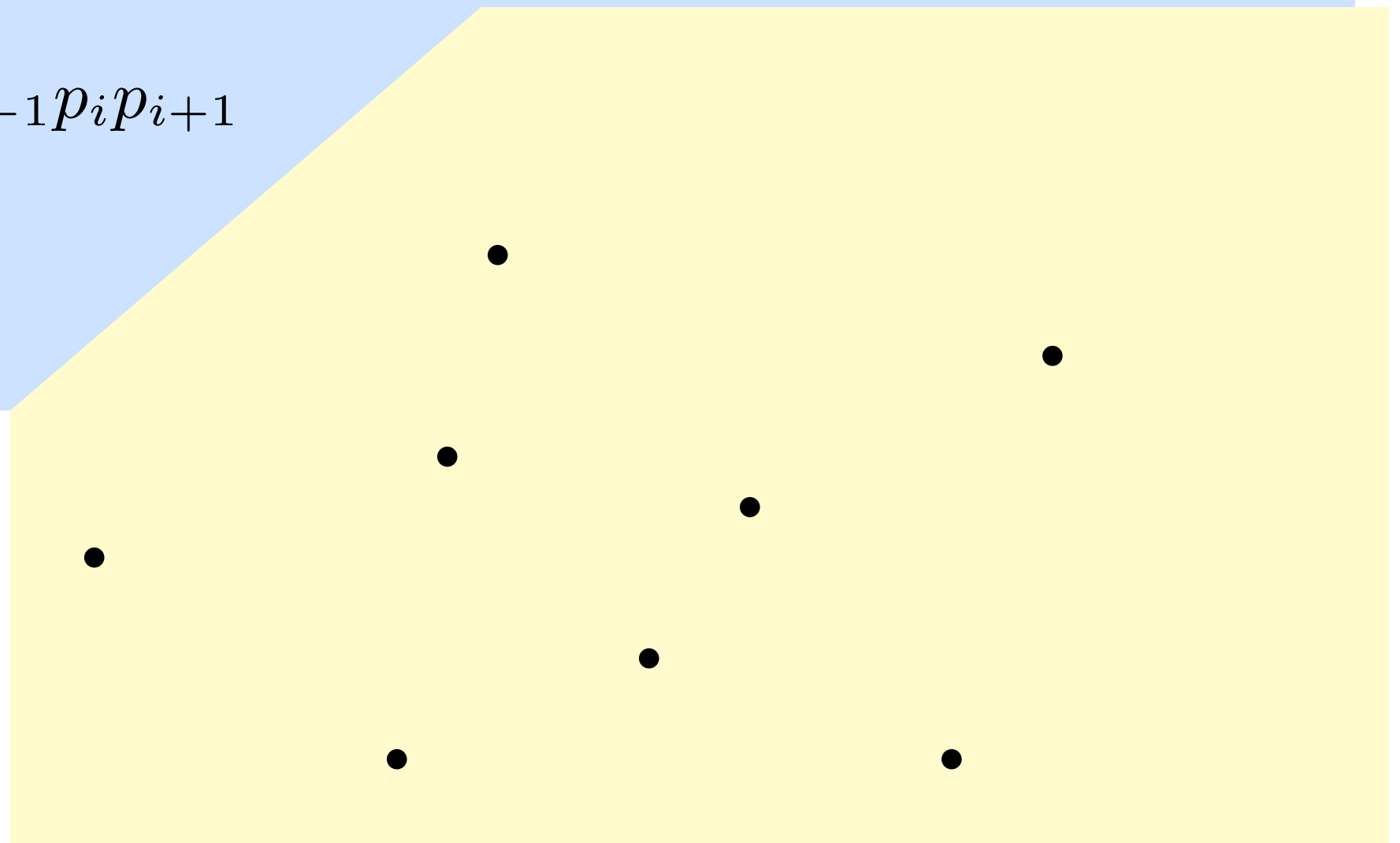
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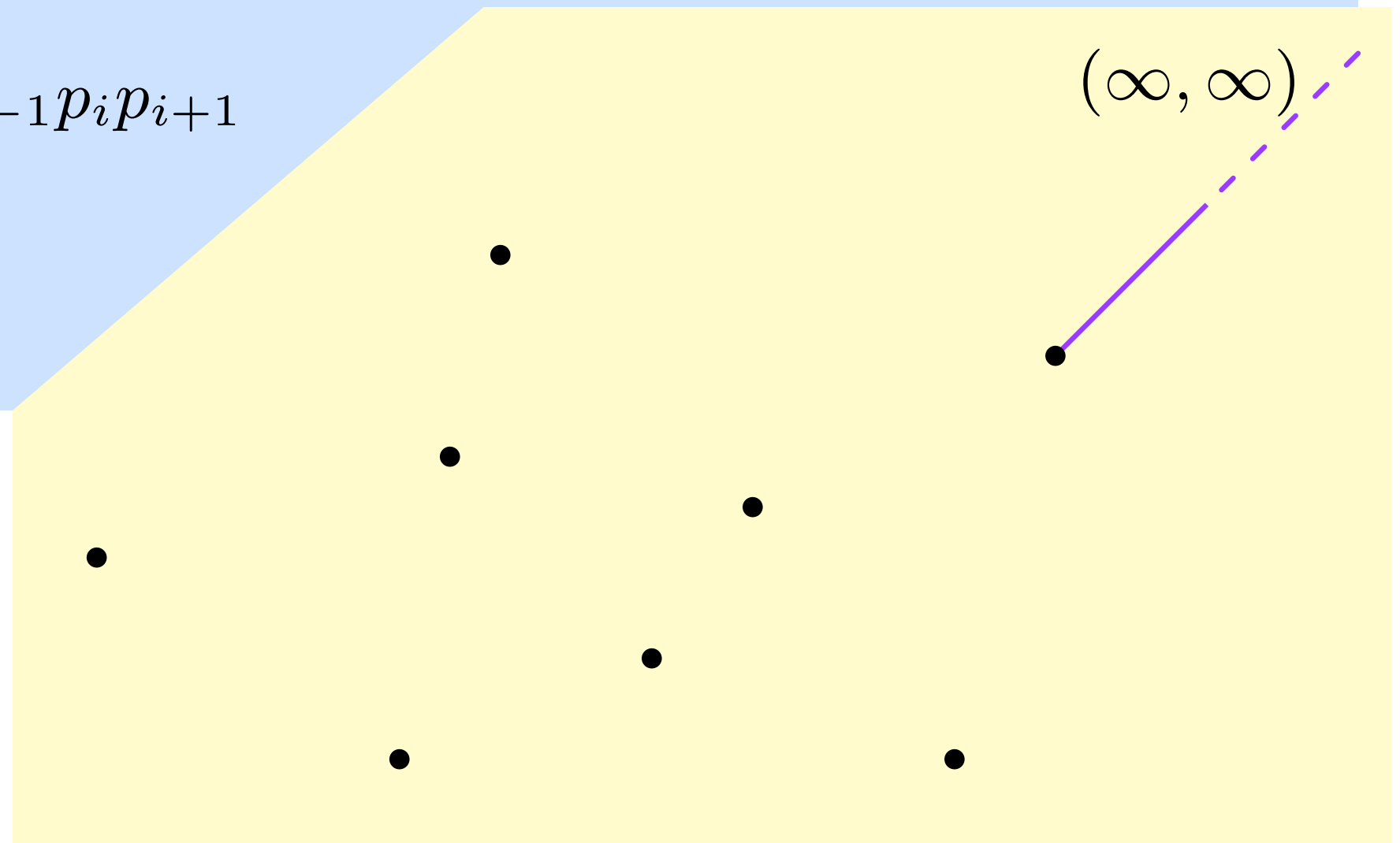
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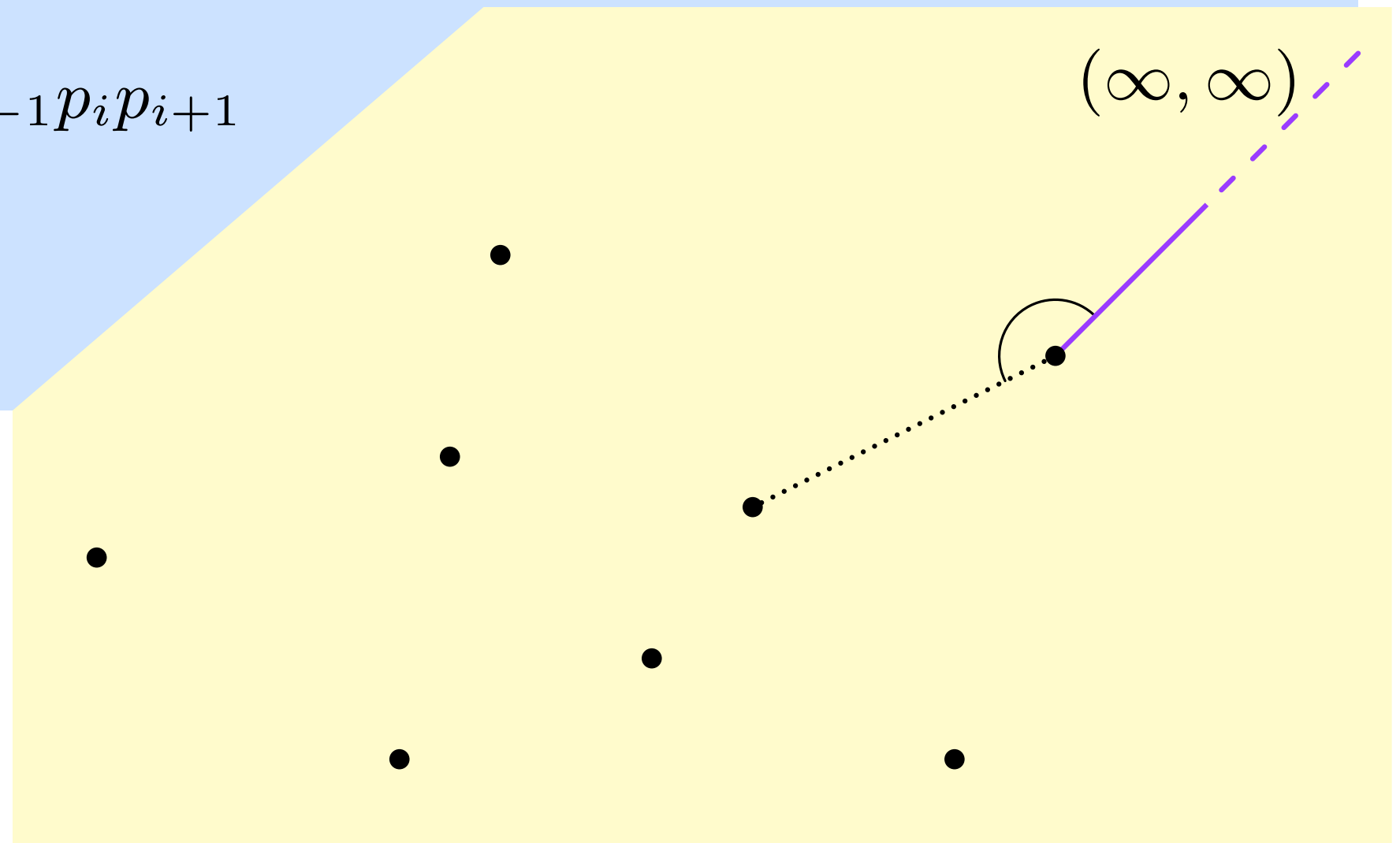
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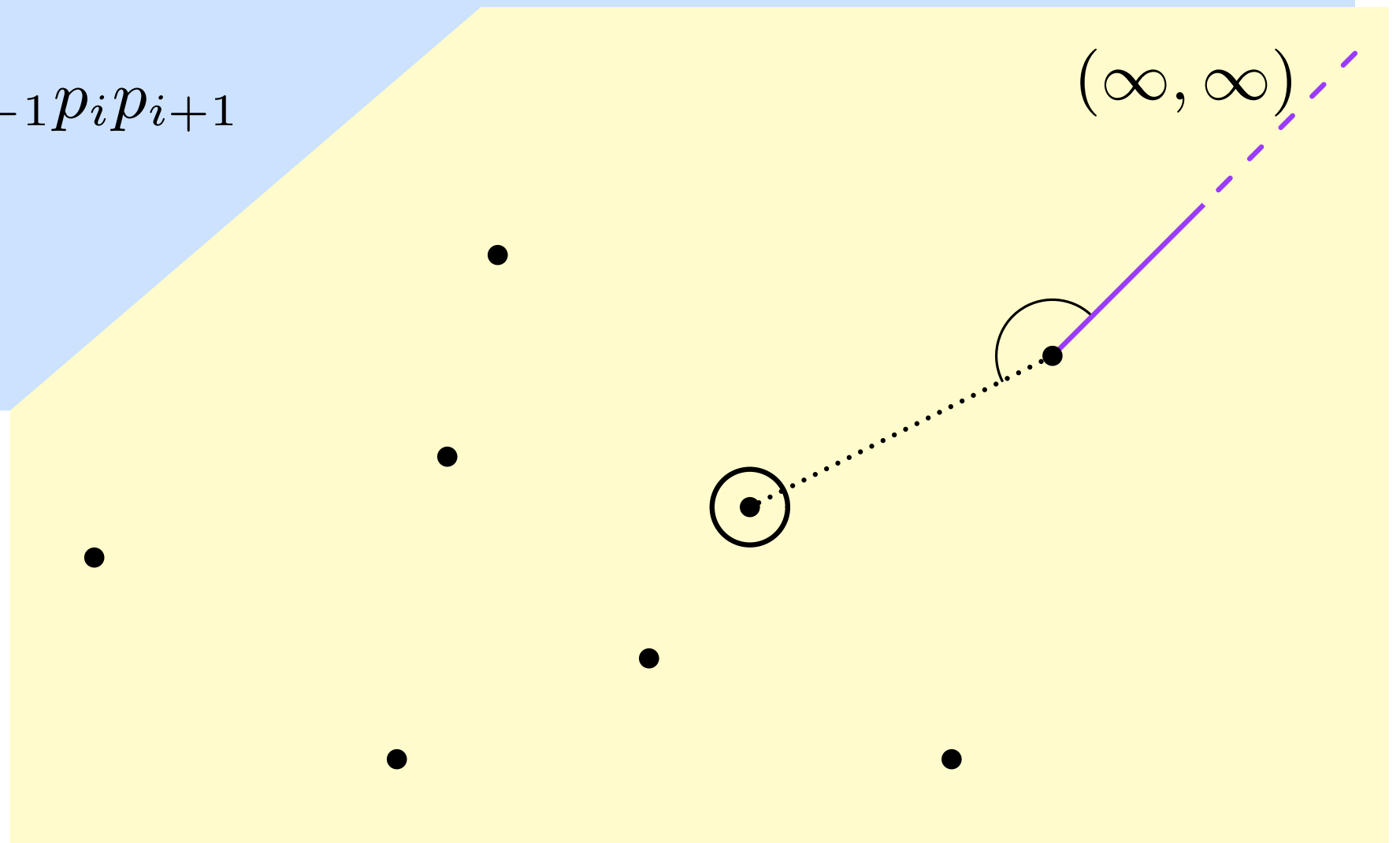
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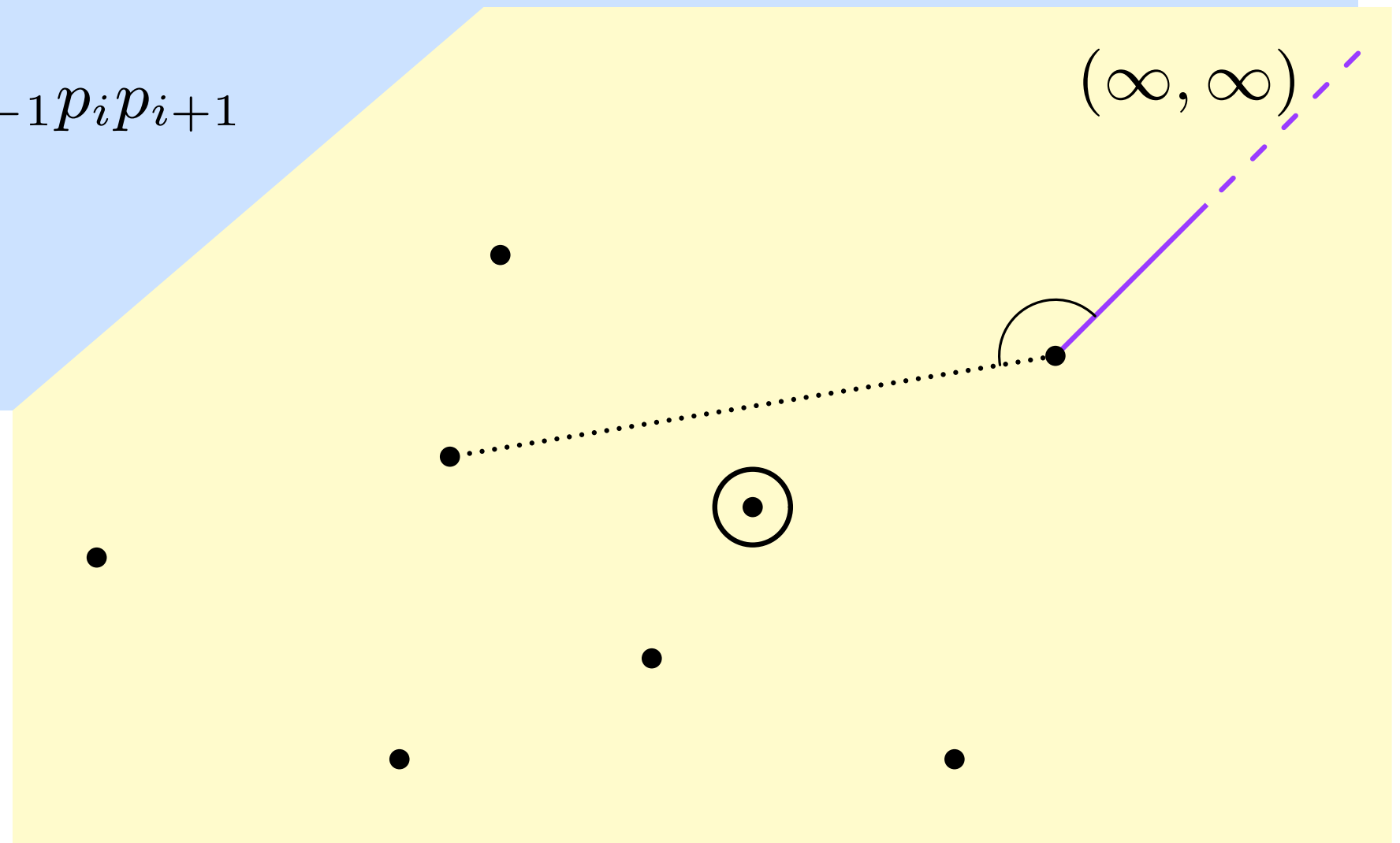
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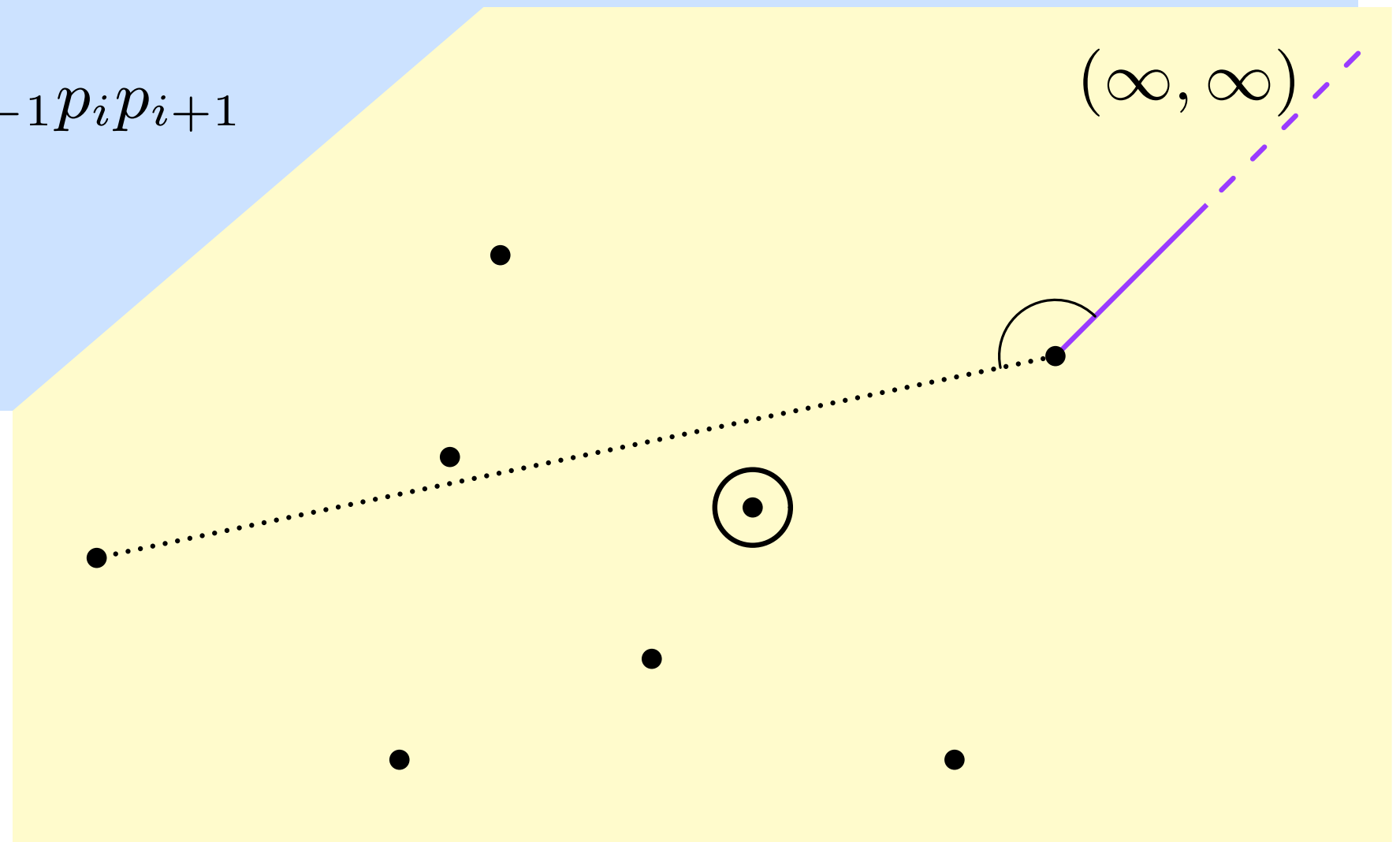
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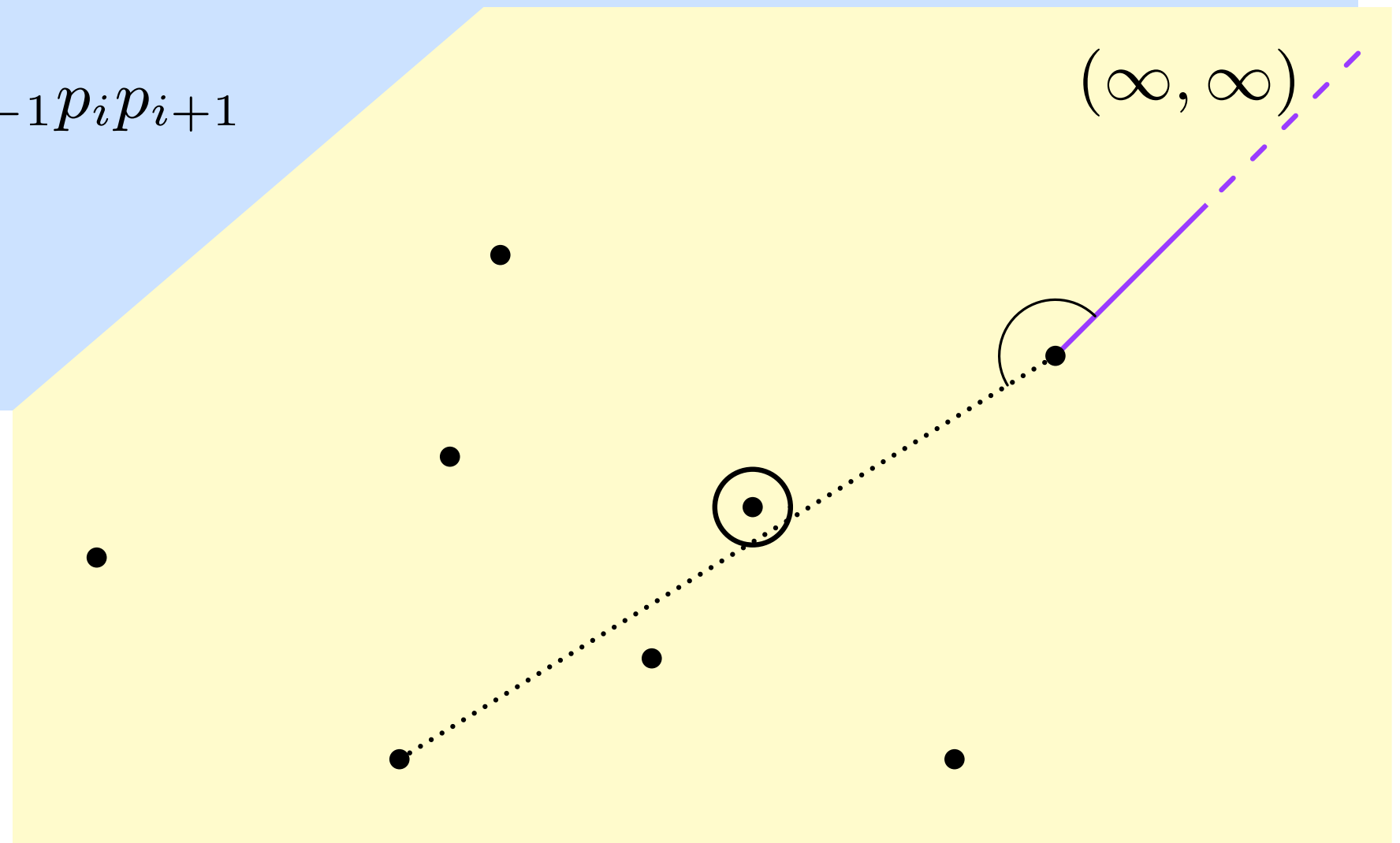
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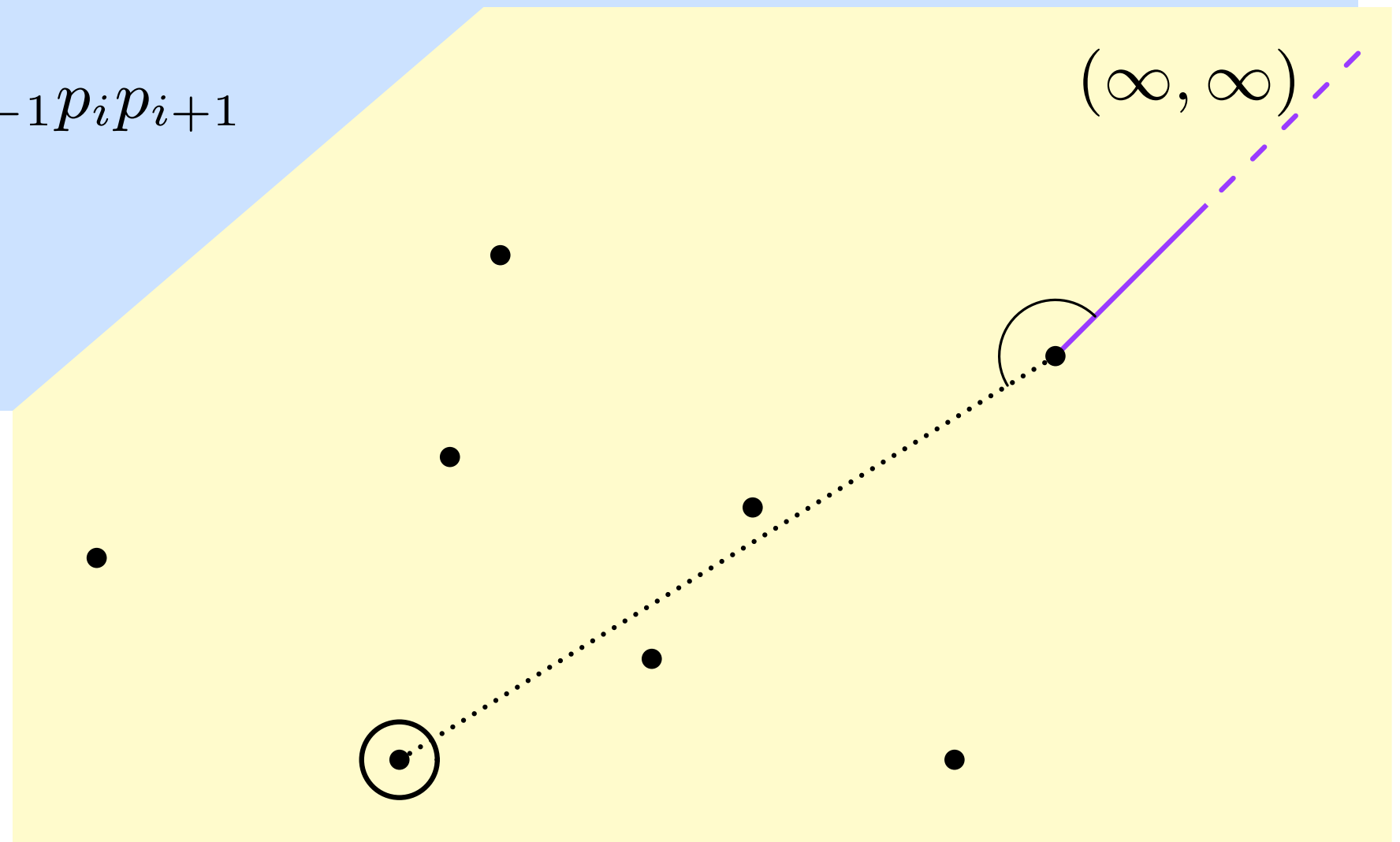
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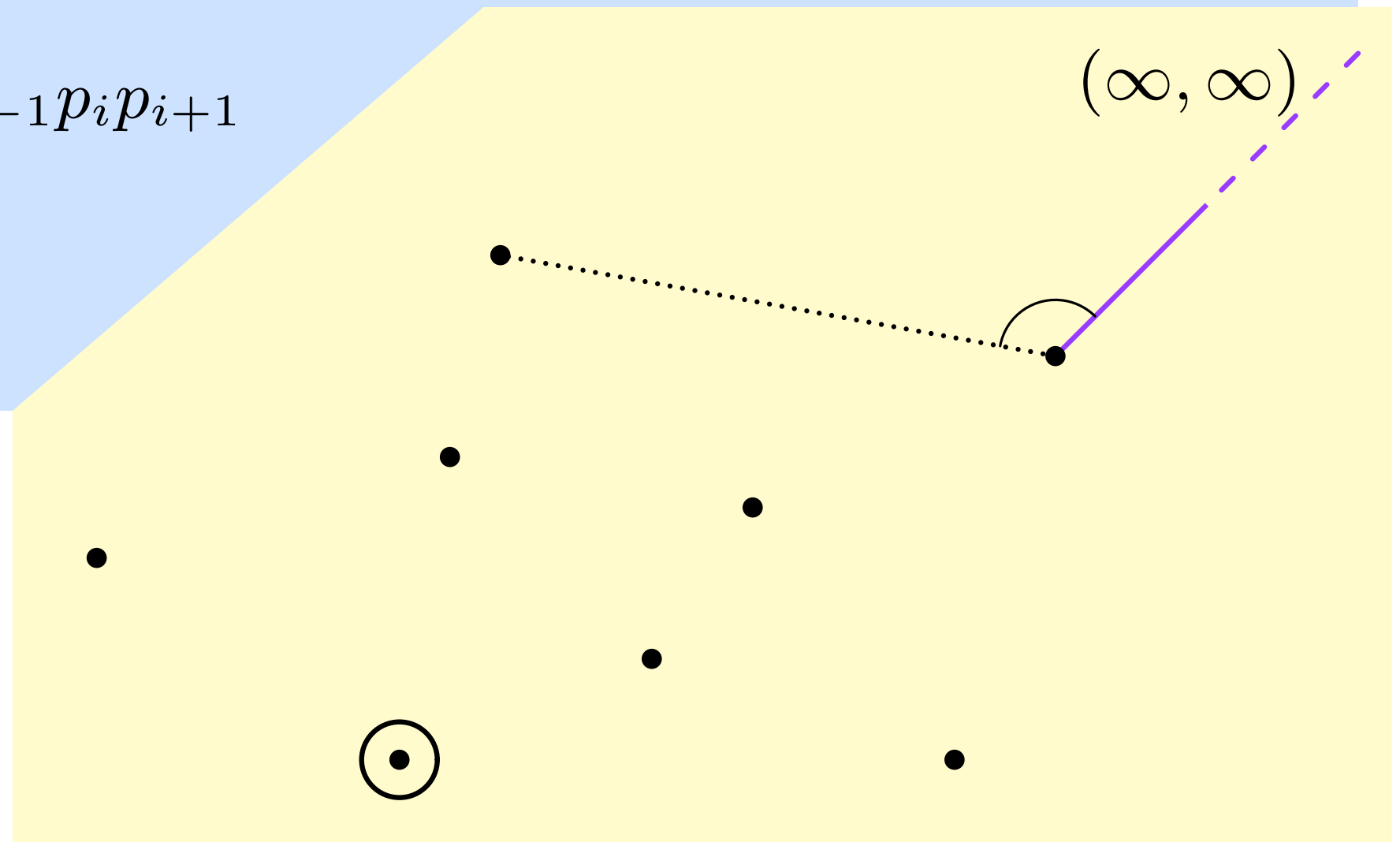
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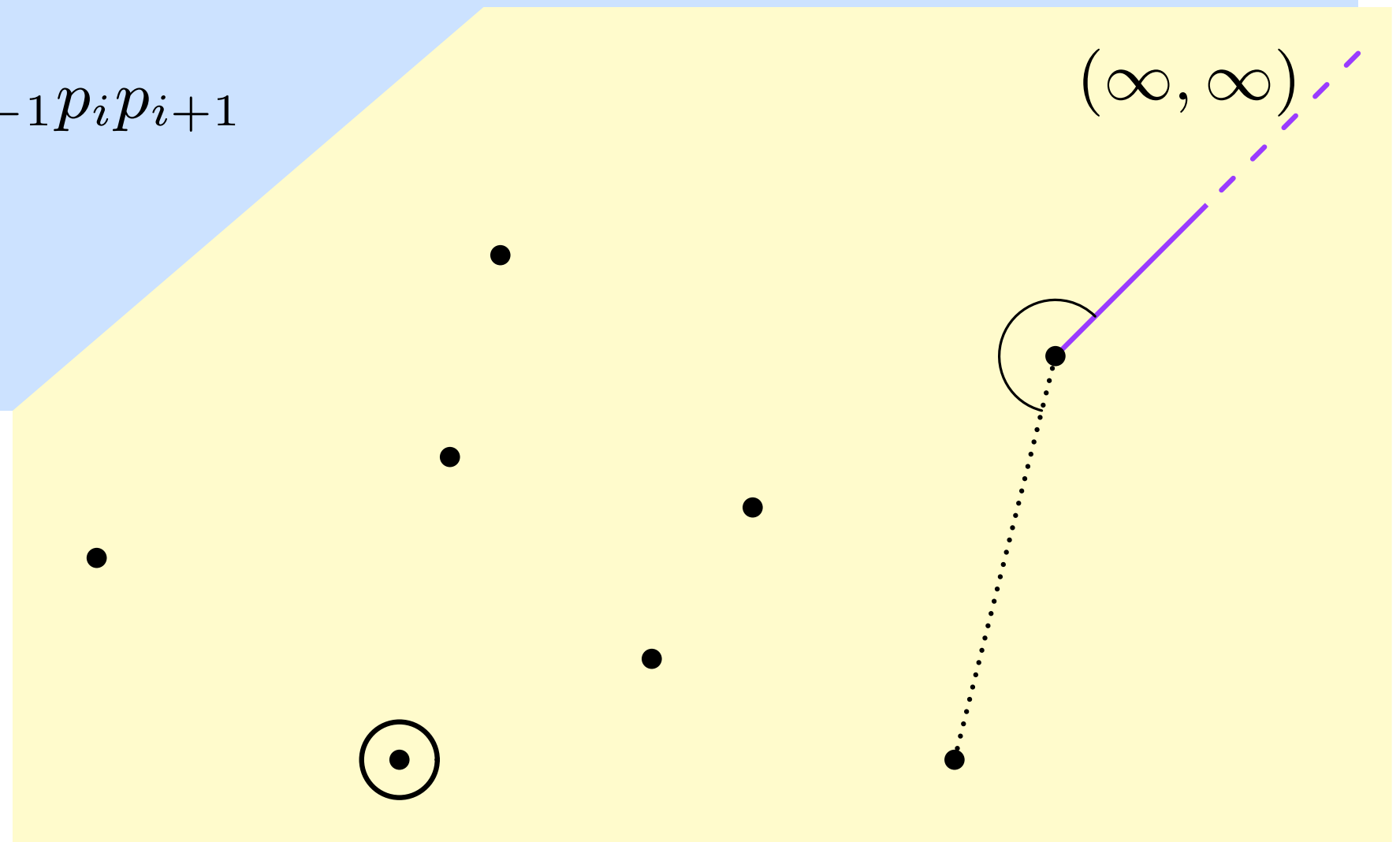
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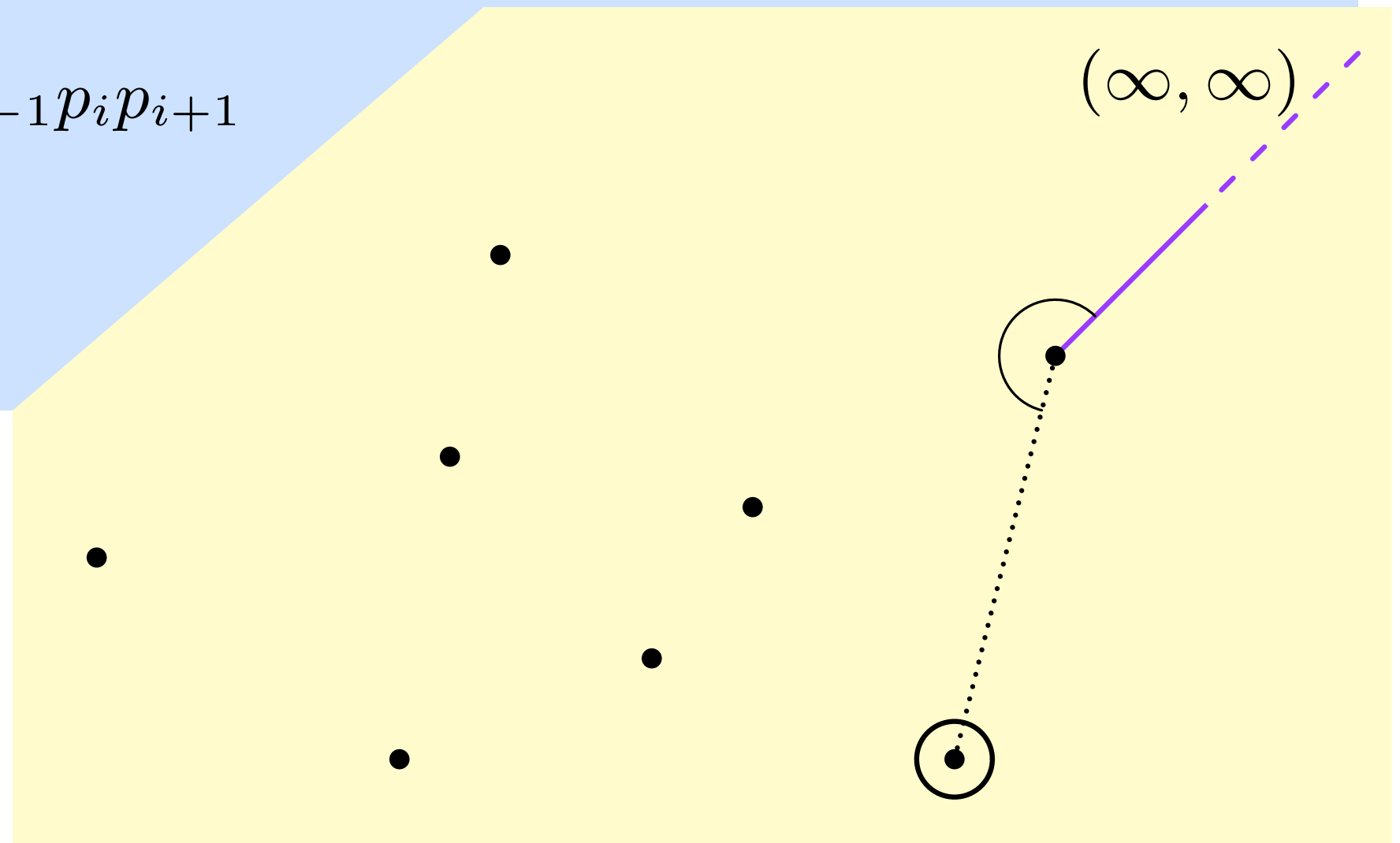
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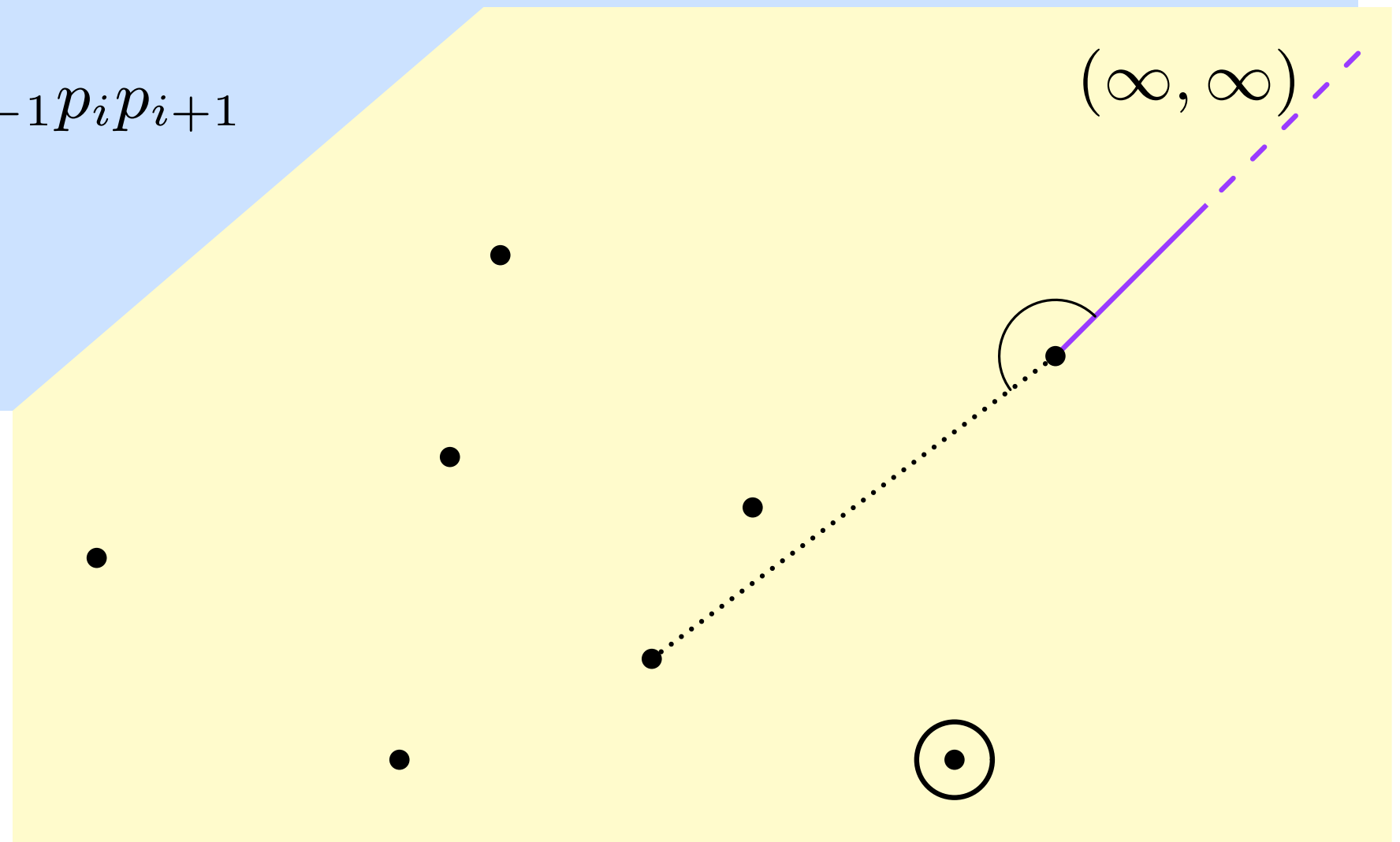
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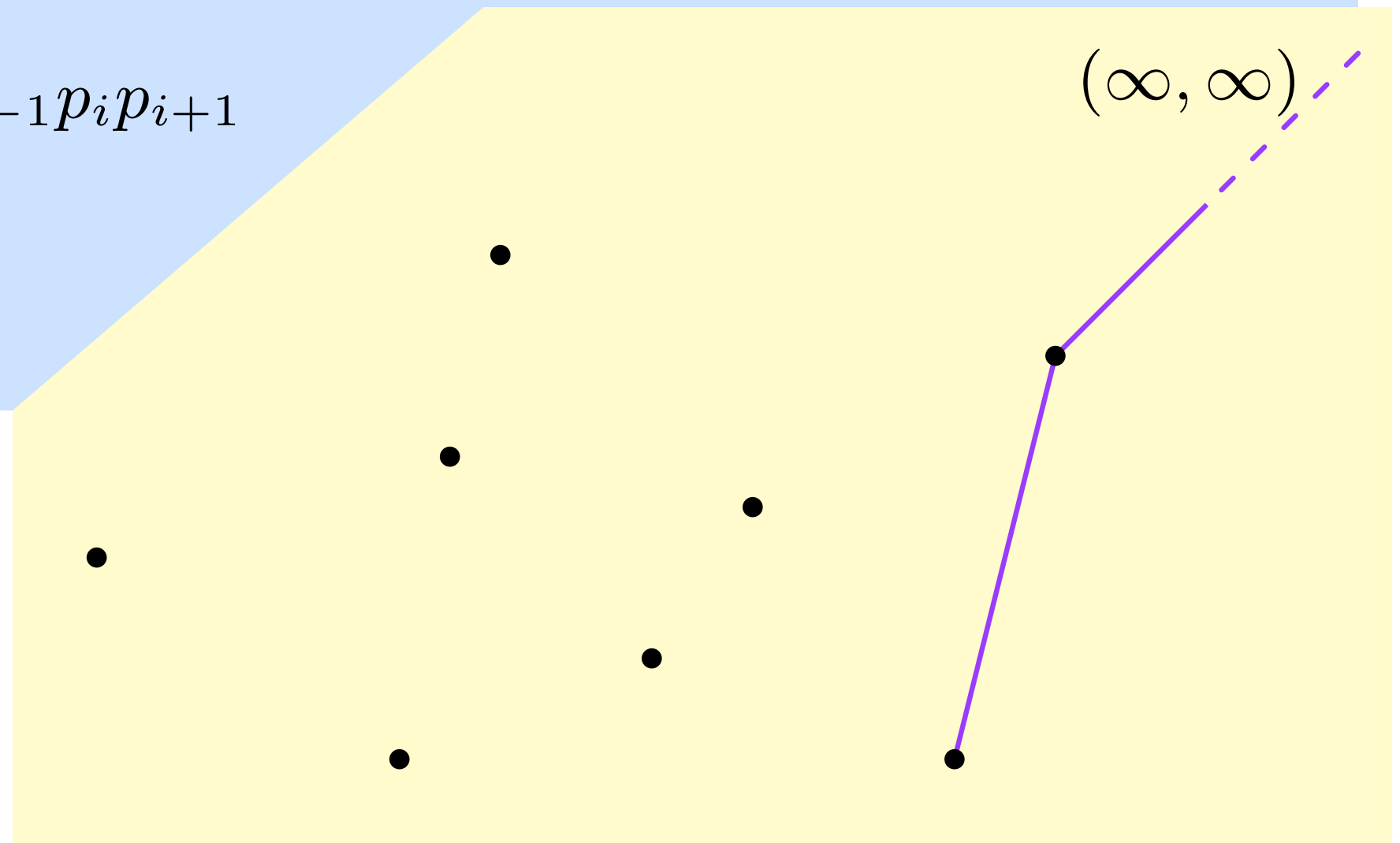
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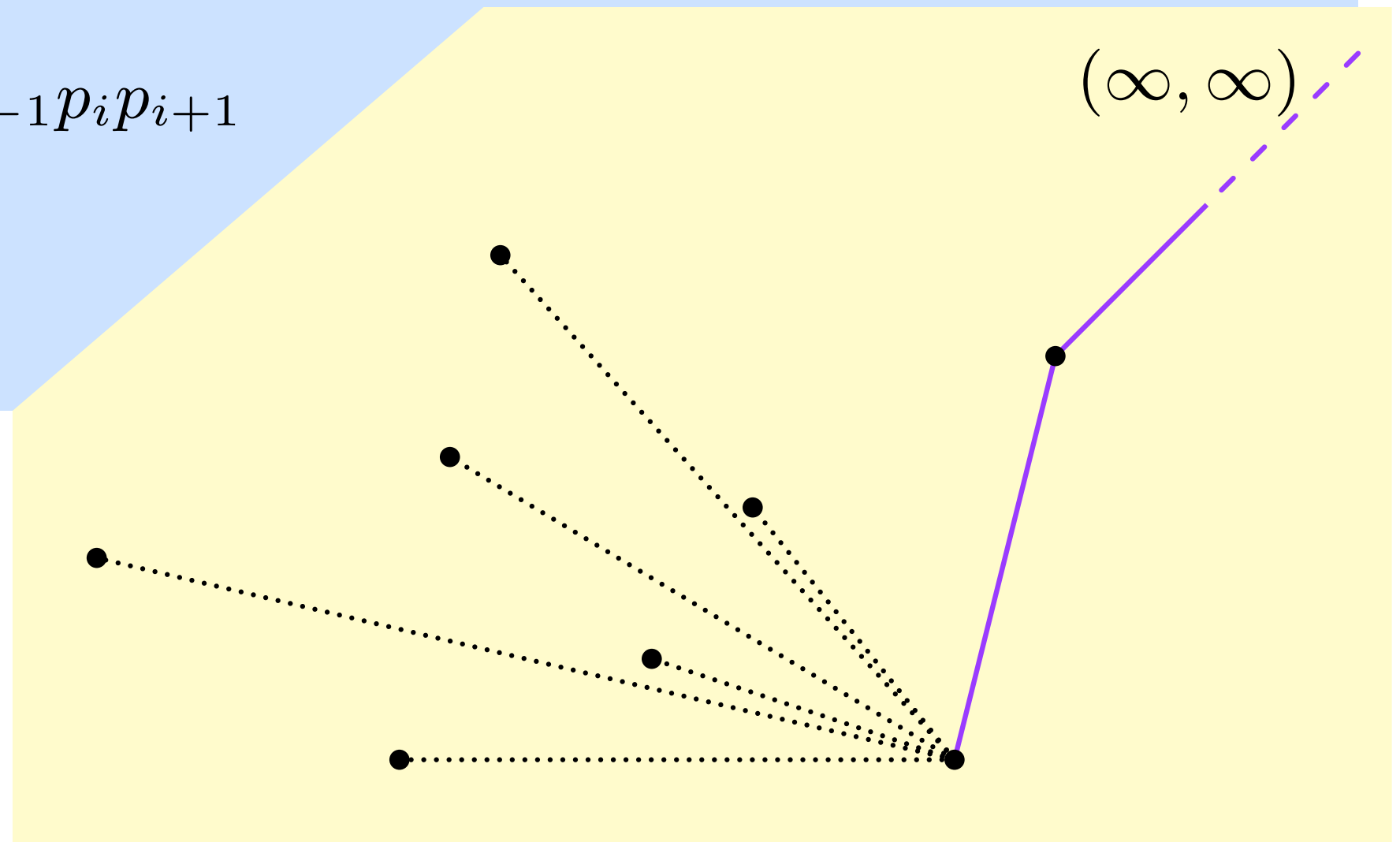
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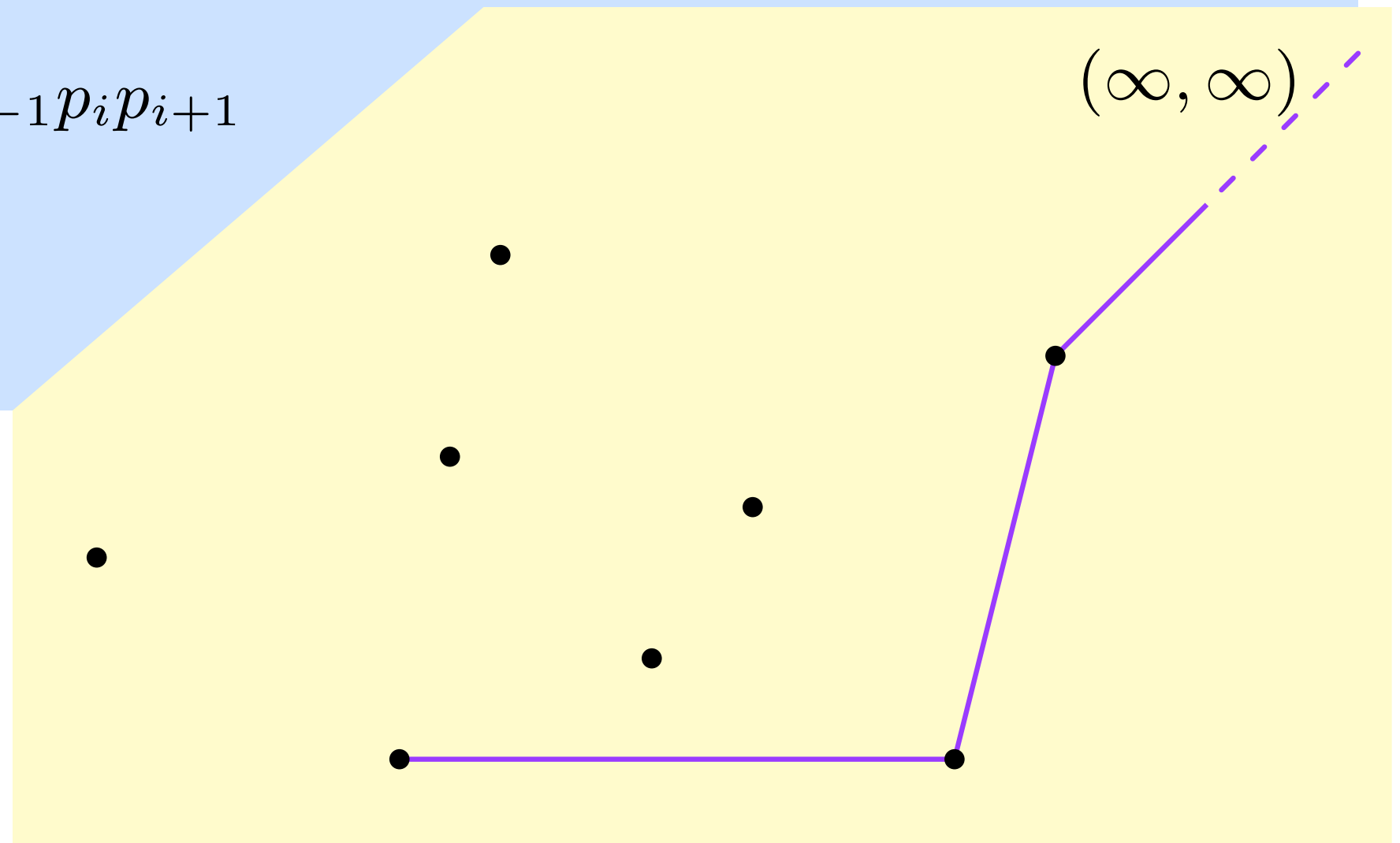
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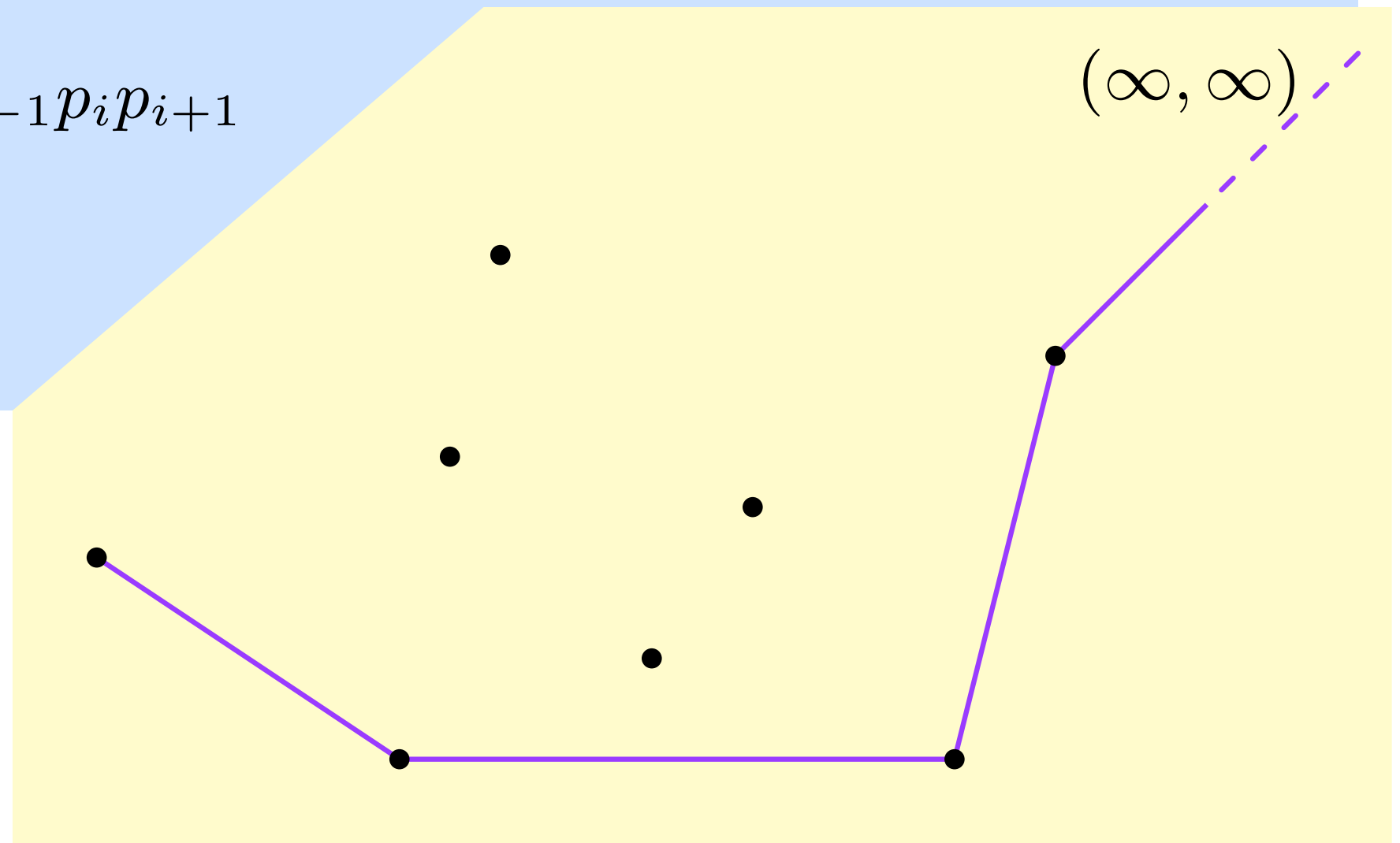
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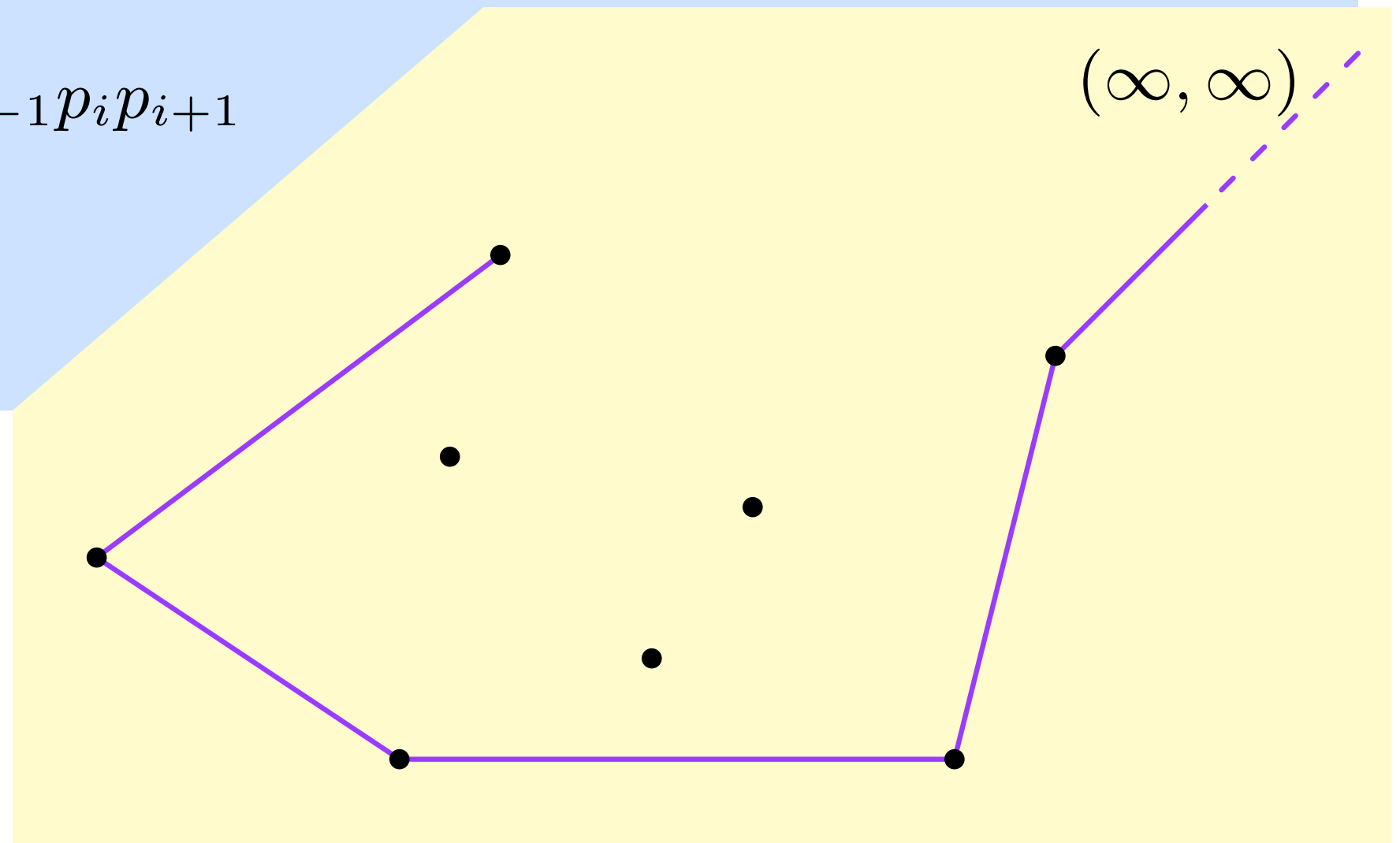
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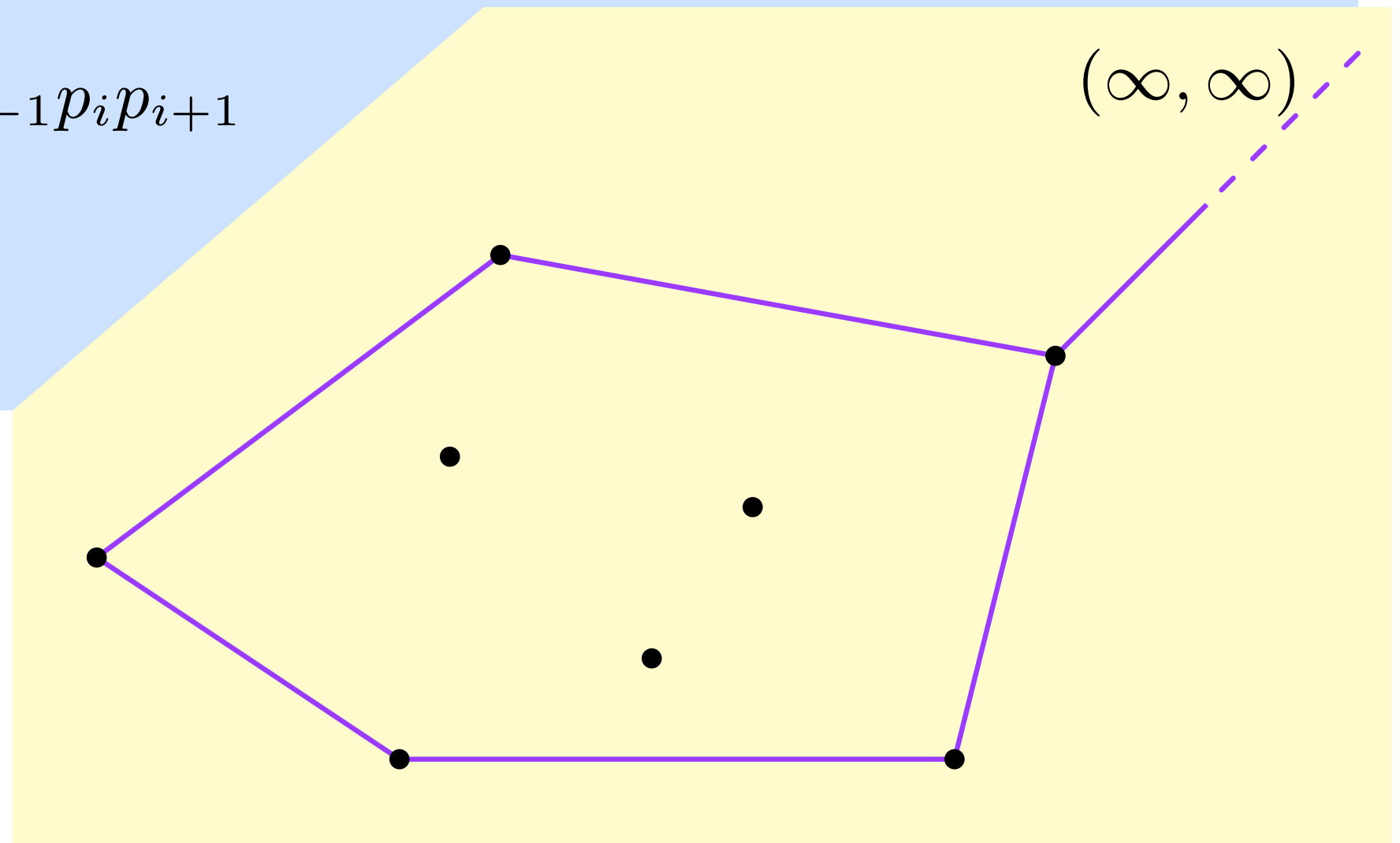
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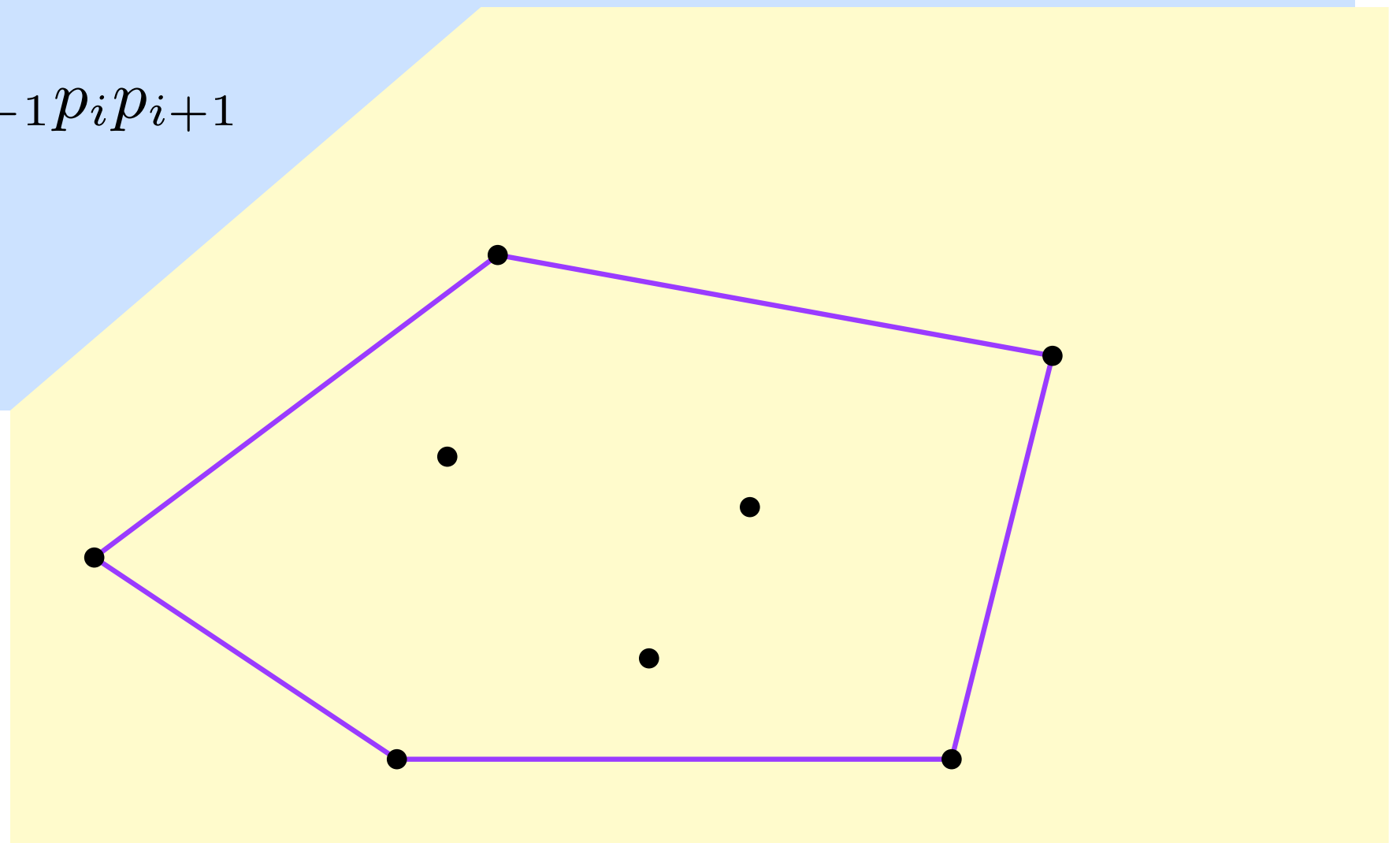
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- maintenance: by construction, all points lie to the right of the line p_i and p_{i+1}

Gift wrapping algorithm: efficiency

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5: **break**

6: insert p_{i+1} into L $O(1)$

} $\times h$

Gift wrapping algorithm: efficiency

Algorithm GIFTWRAPPING(P)

Input: set P of points in the plane

Output: list L containing vertices of $CH(P)$ in counterclockwise order

1:	$p_0 \leftarrow (\infty, \infty), p_1 \leftarrow$ right-most vertex in P , insert p_1 into L	$O(n)$	
2:	while <i>true</i> do		
3:	choose $p_{i+1} \in P$ maximizing $\angle p_{i-1}p_ip_{i+1}$	$O(n)$	} $\times h$
4:	if $p_{i+1} = p_1$ then	$O(1)$	
5:	break		
6:	insert p_{i+1} into L	$O(1)$	

Total time:

$$O(n) + h \cdot (O(n) + O(1) + O(1)) = O(hn)$$

Optimality?

When should we choose which algorithm?

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When should we choose which algorithm?

- many points on $CH(P)$: $O(n \log n)$ [Graham Scan](#)
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Can we do better?

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Can we do better? [yes, by combining both algorithms](#)

Chan's Algorithms



Chan's Algorithms



Chan's Algorithms

Suppose we know h :



Chan's Algorithms

Suppose we know h :

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- 1: partition P into sets P_i with h points each
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Graham Scan

Gift Wrapping

Chan's Algorithms



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Chan's Algorithms



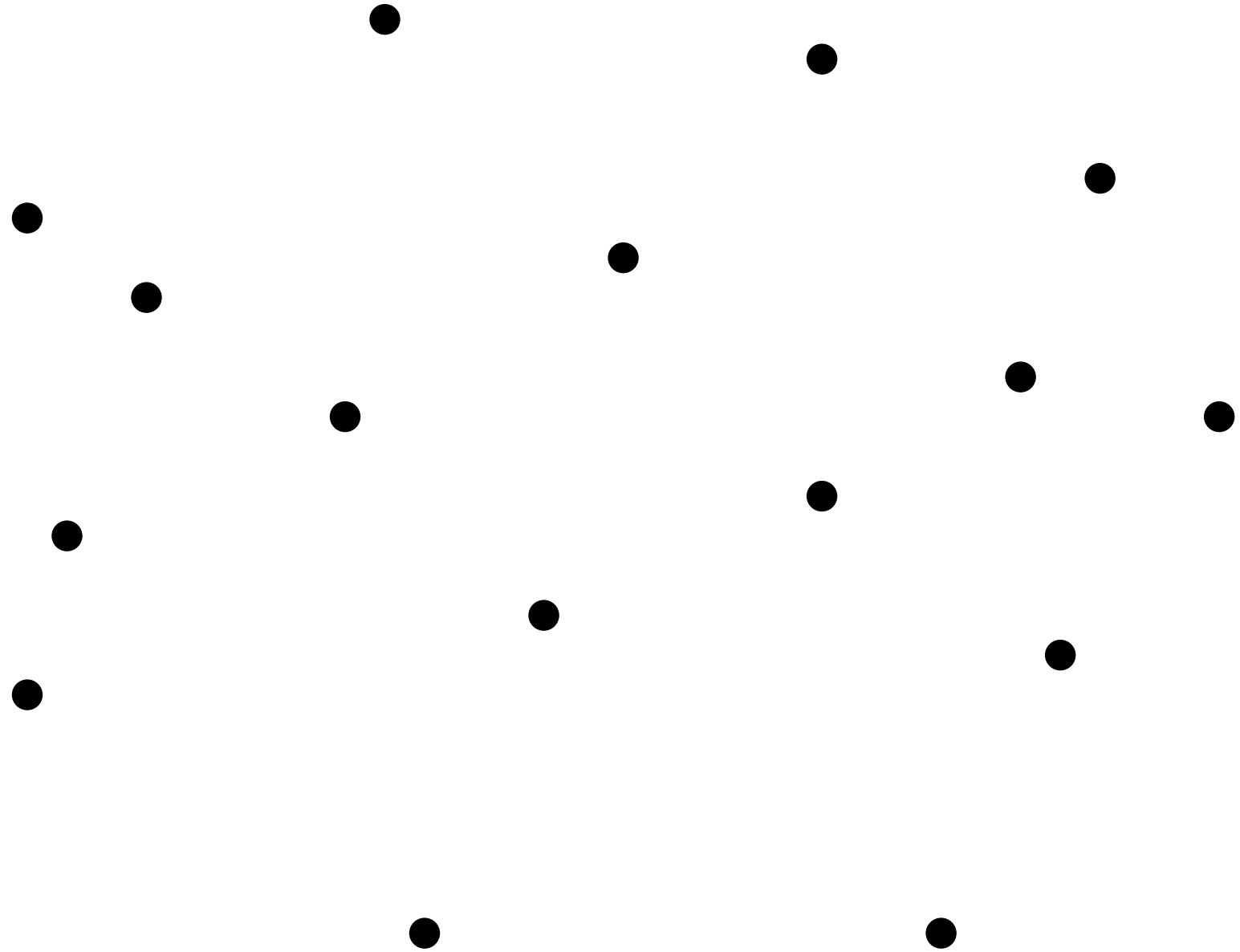
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Total running time: $O(n \log h)$

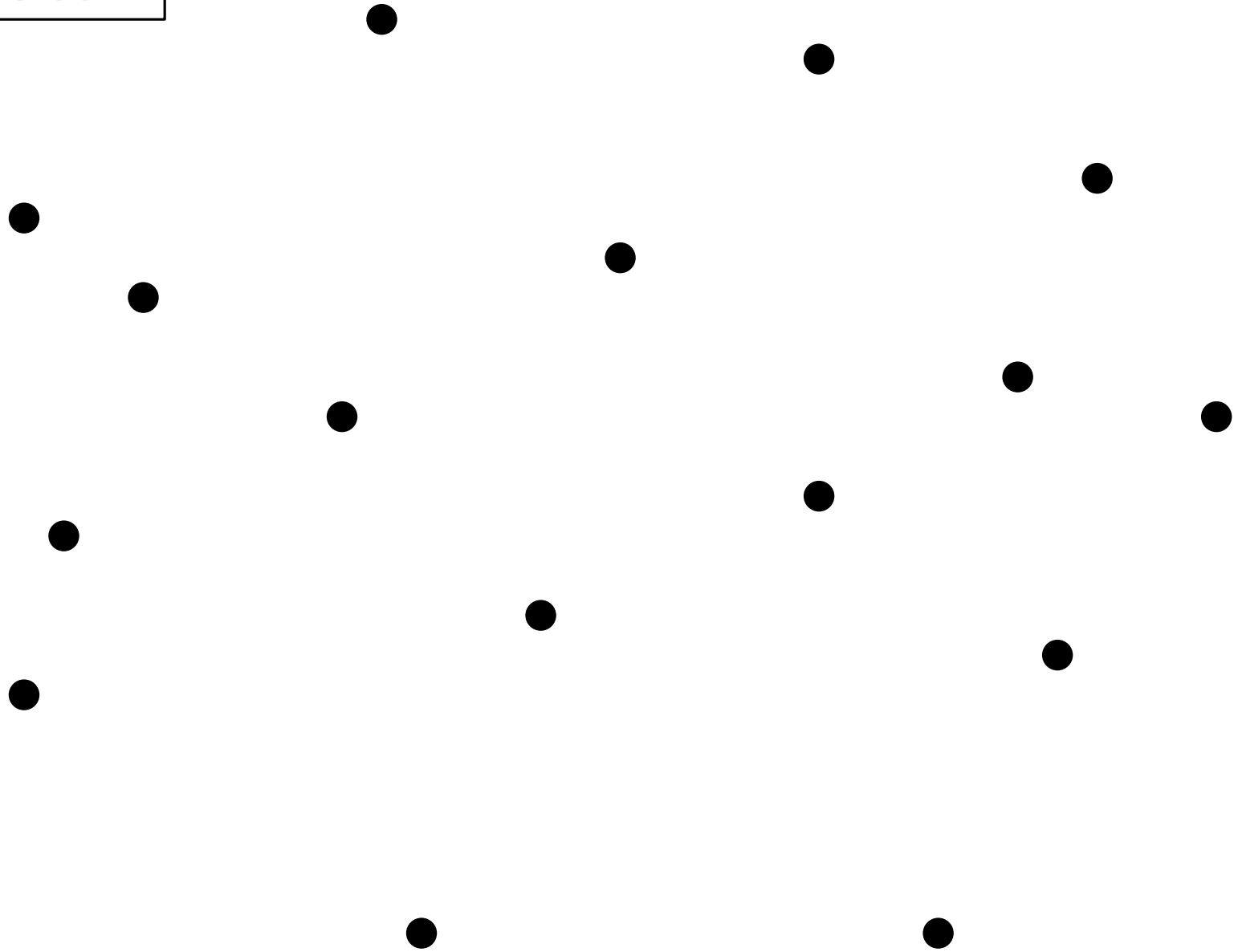
Example



$$n = 16$$

Example

Graham Scan

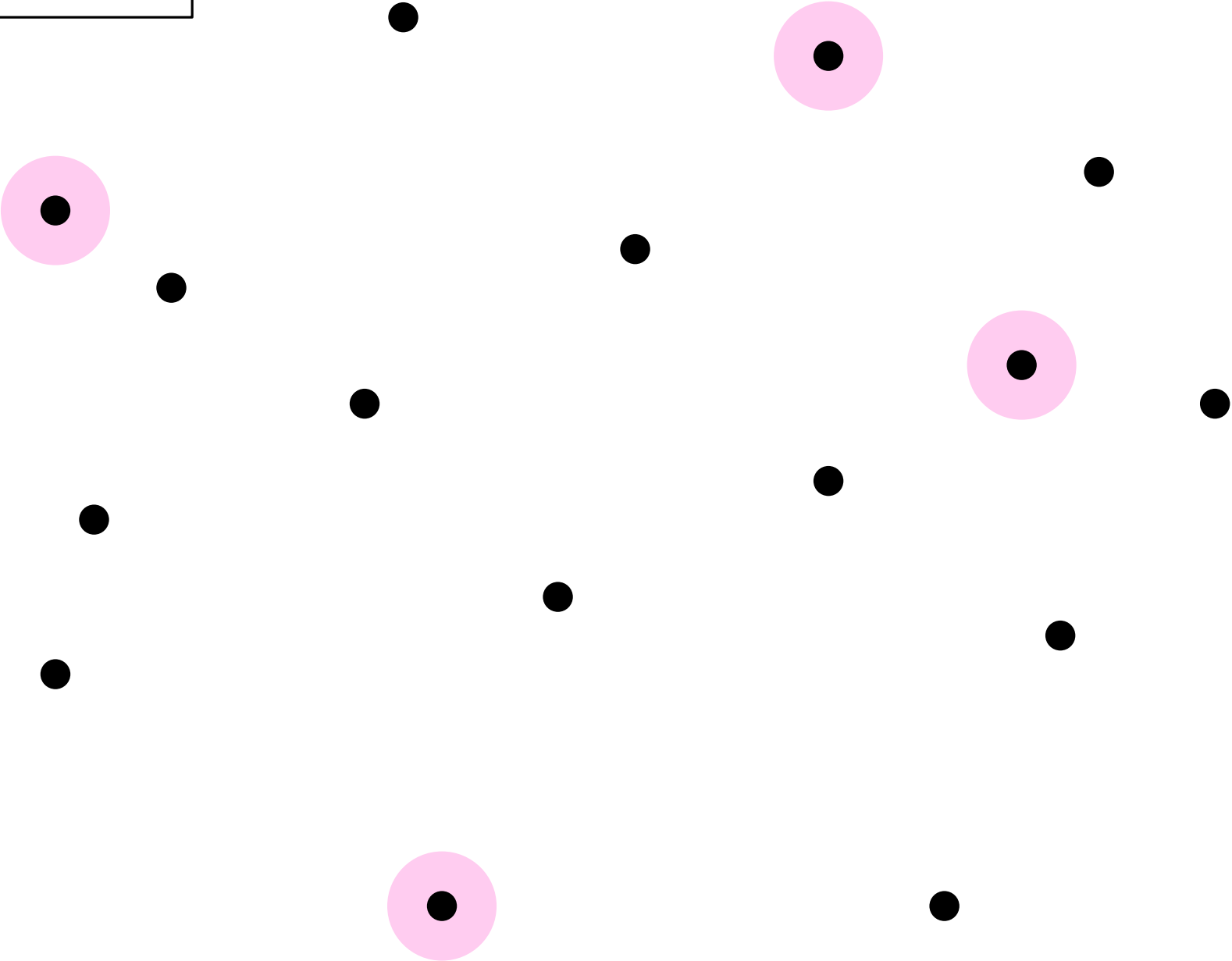


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Example

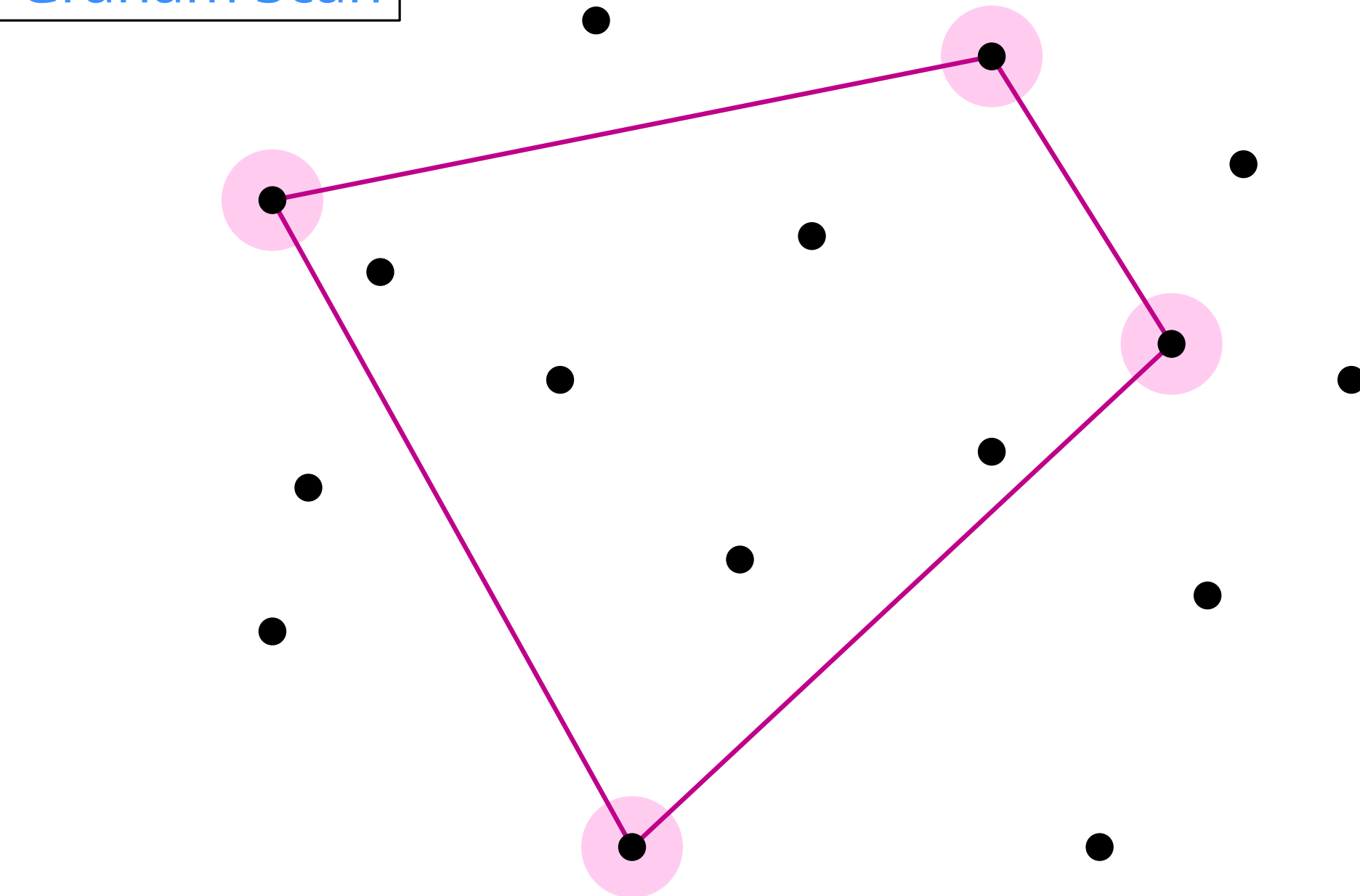
Graham Scan

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Example

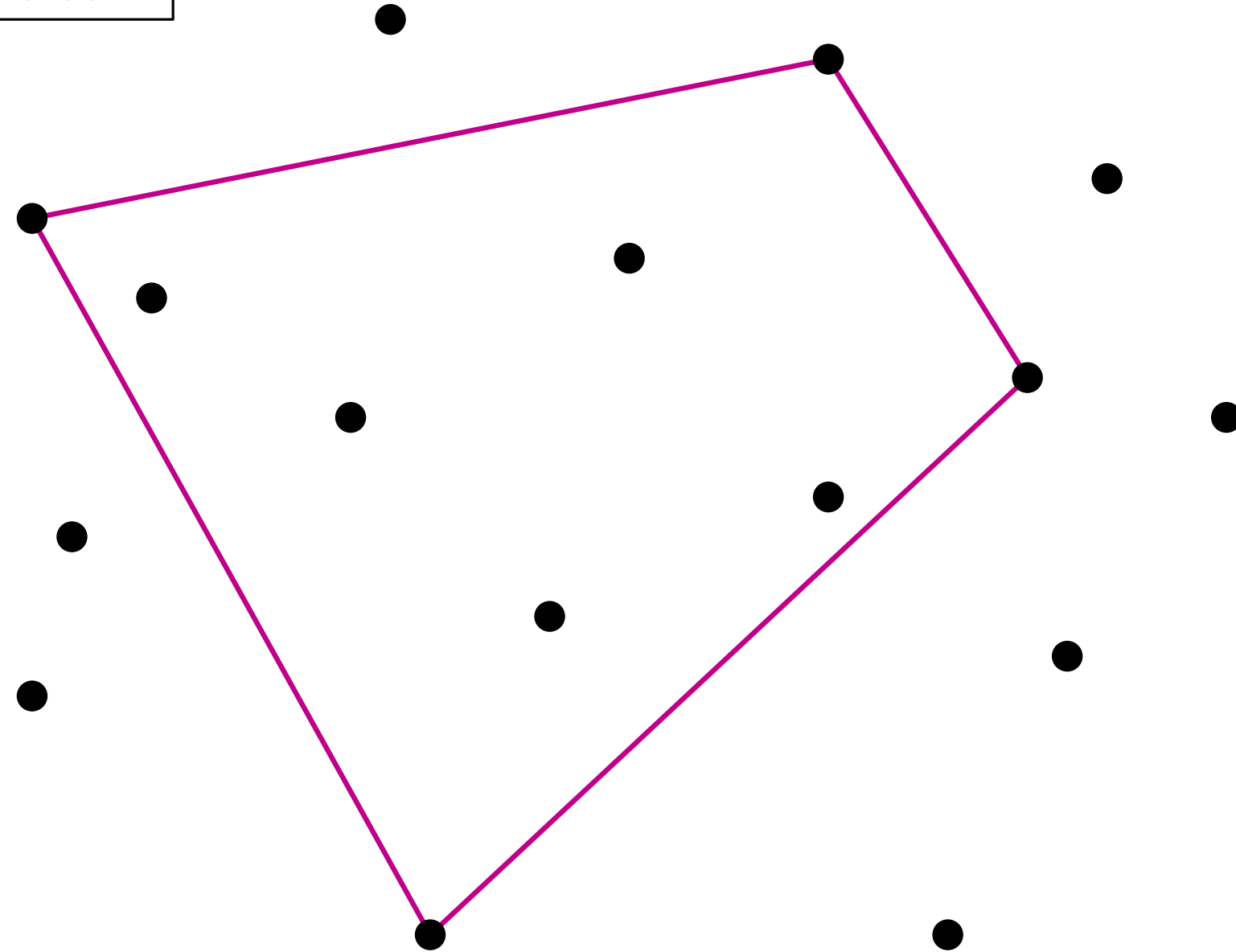
Graham Scan



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Example

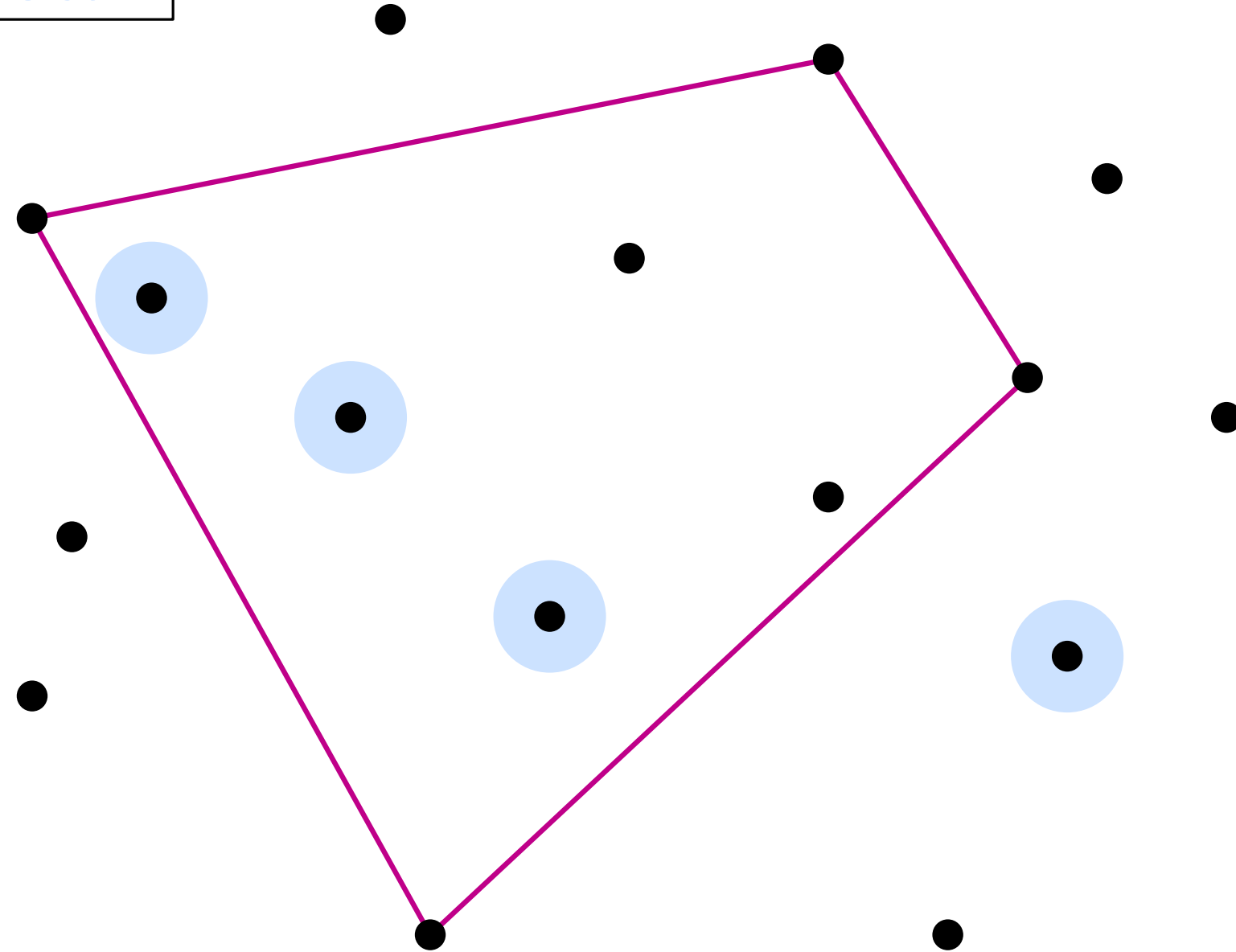
Graham Scan



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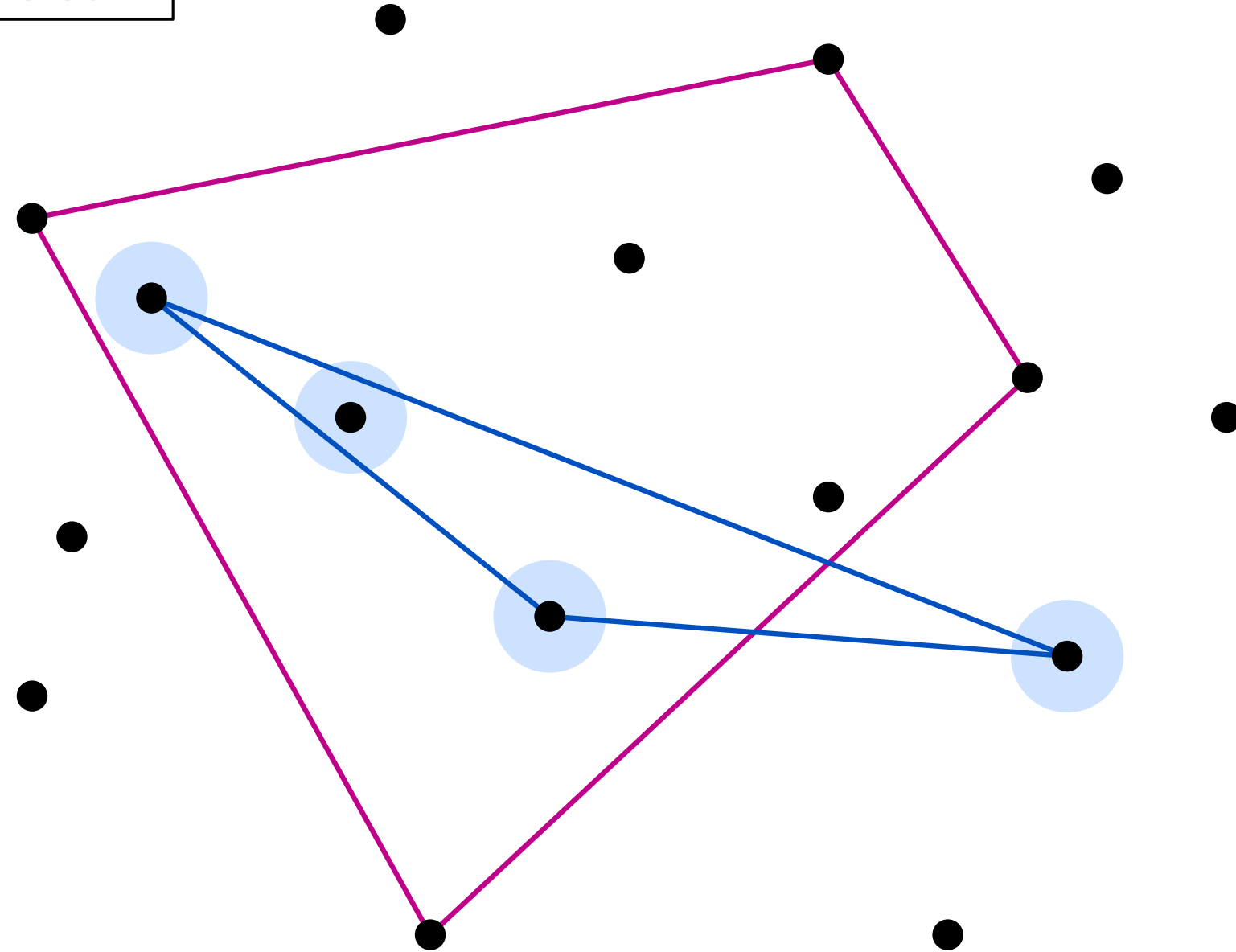
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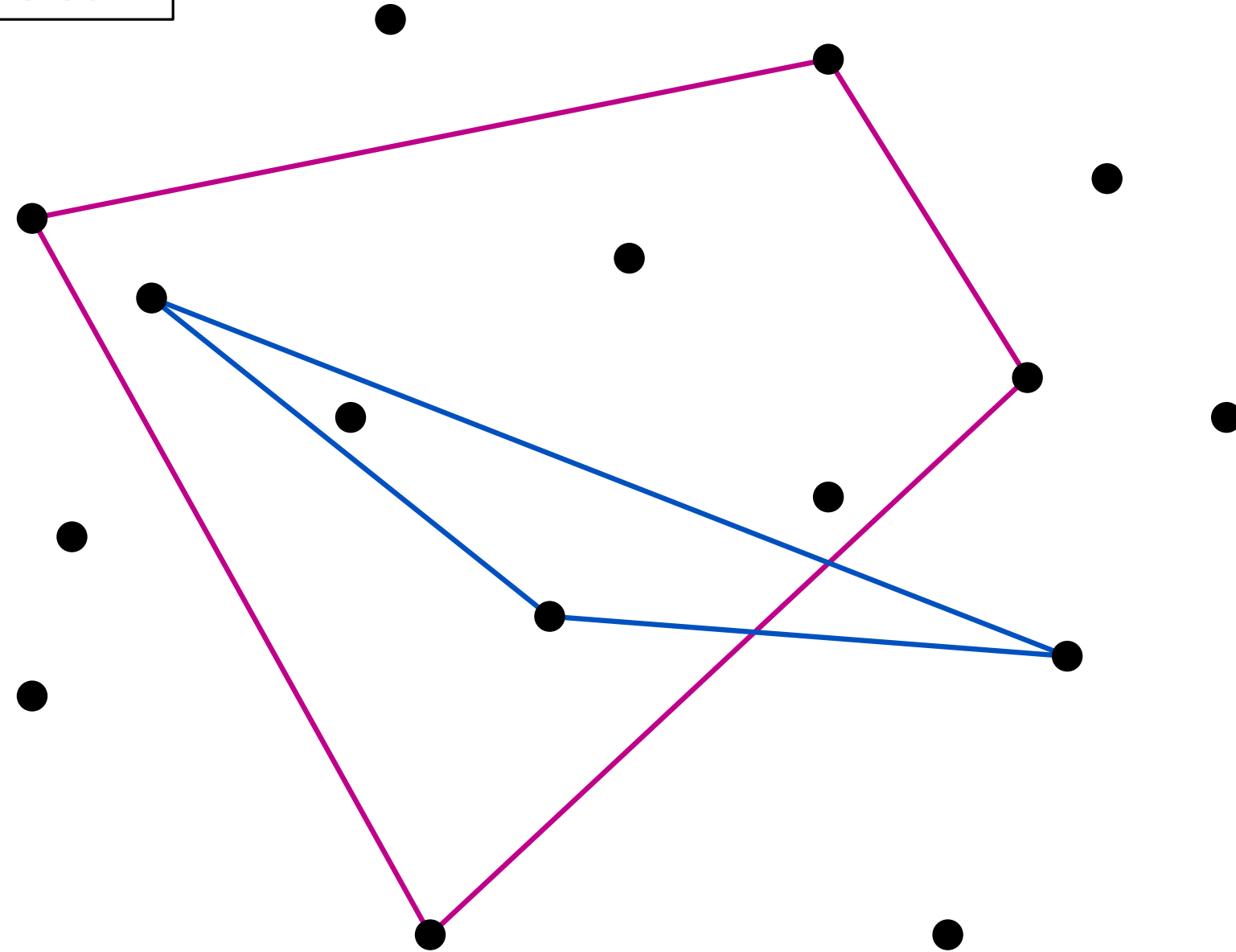
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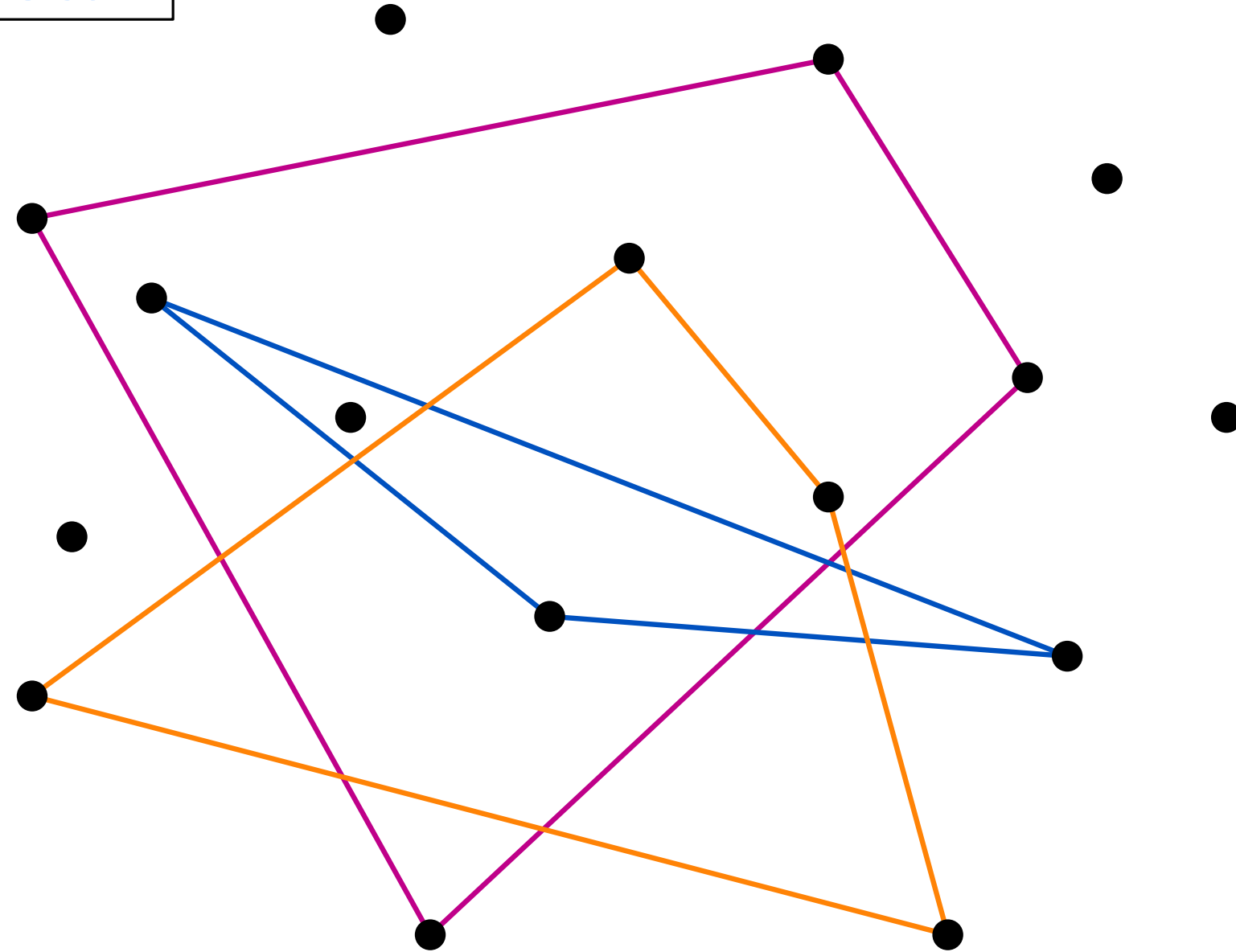
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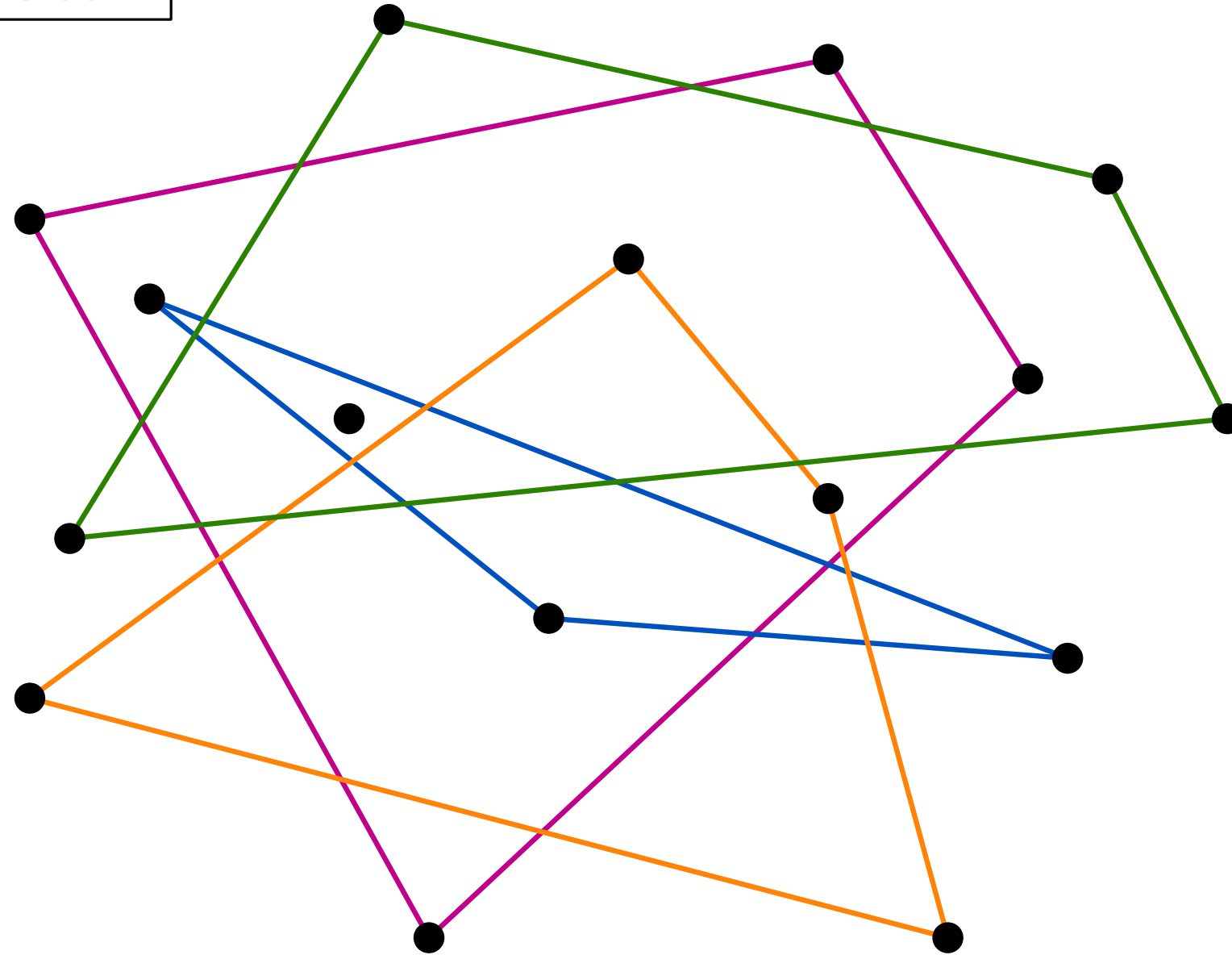
Graham Scan



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Example

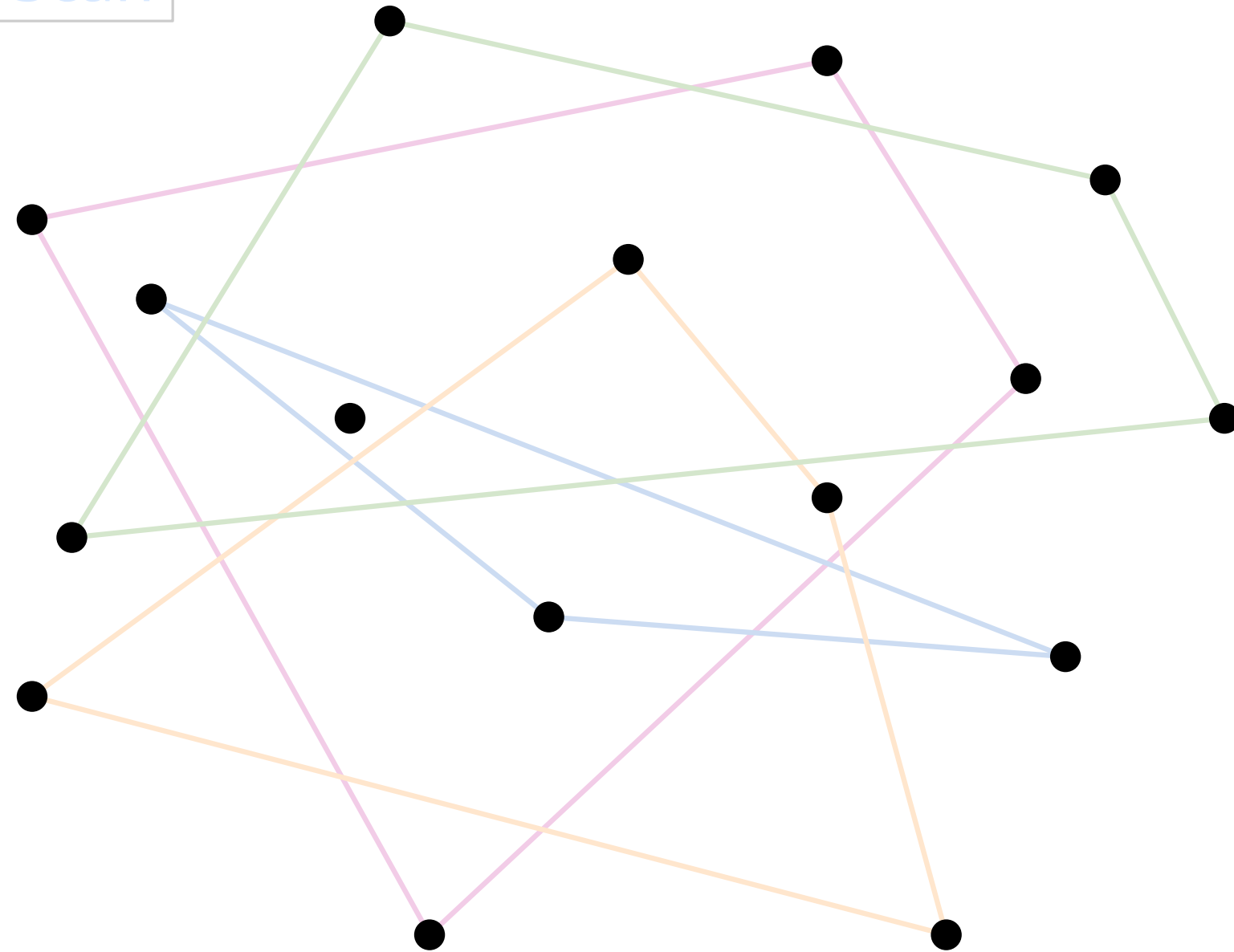
Graham Scan



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Graham Scan

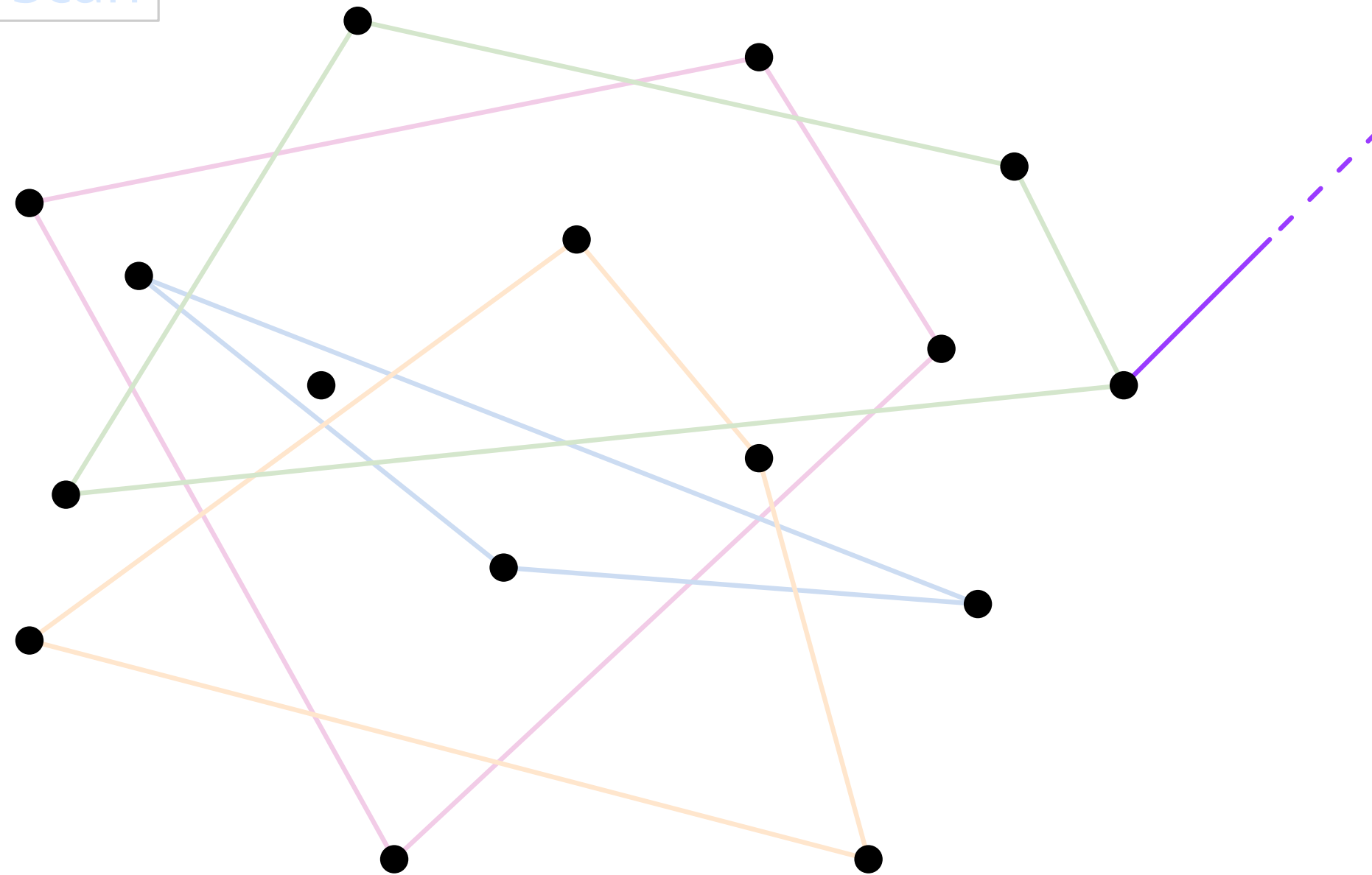


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Gift Wrapping

Example

Graham Scan

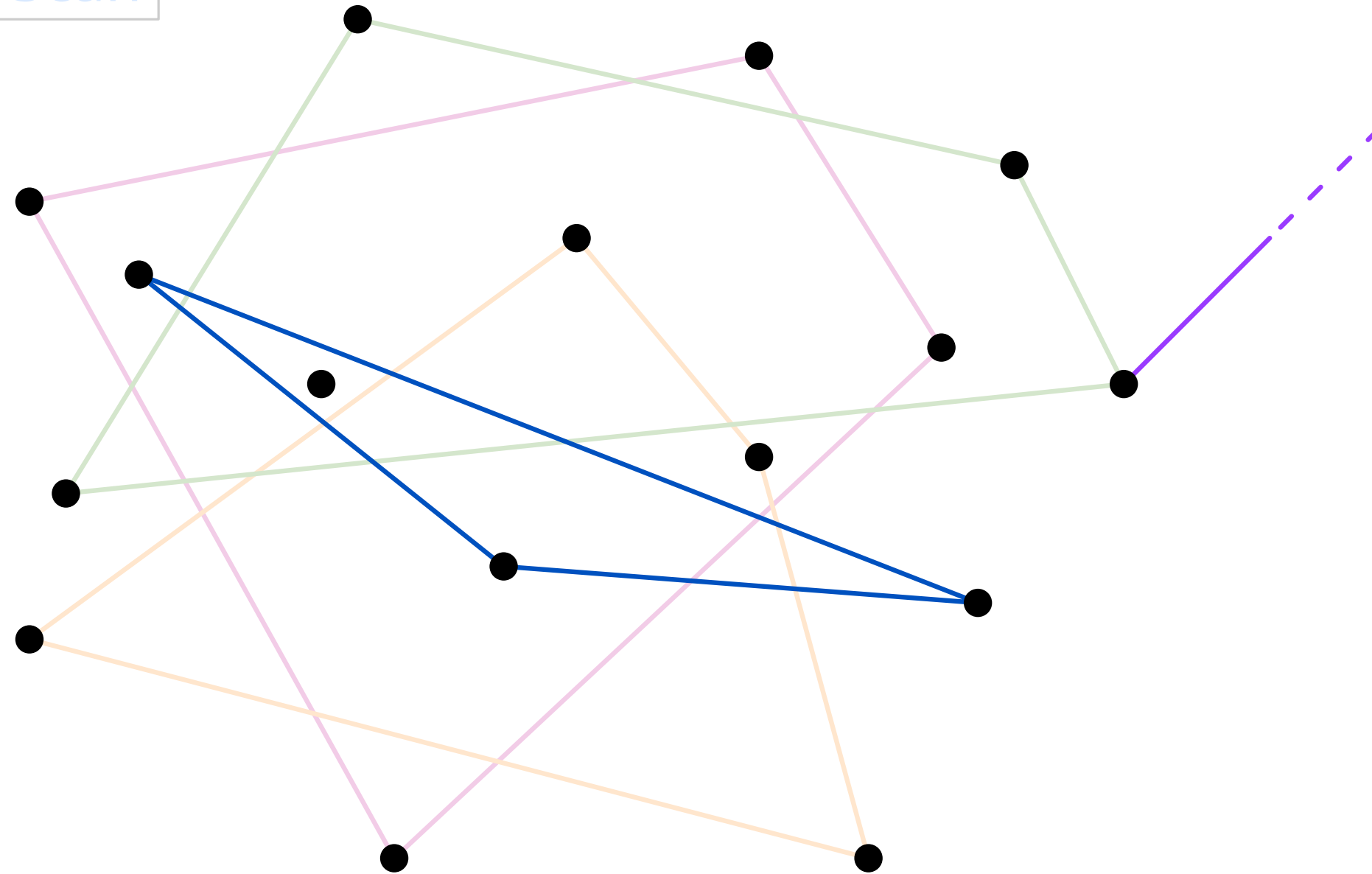


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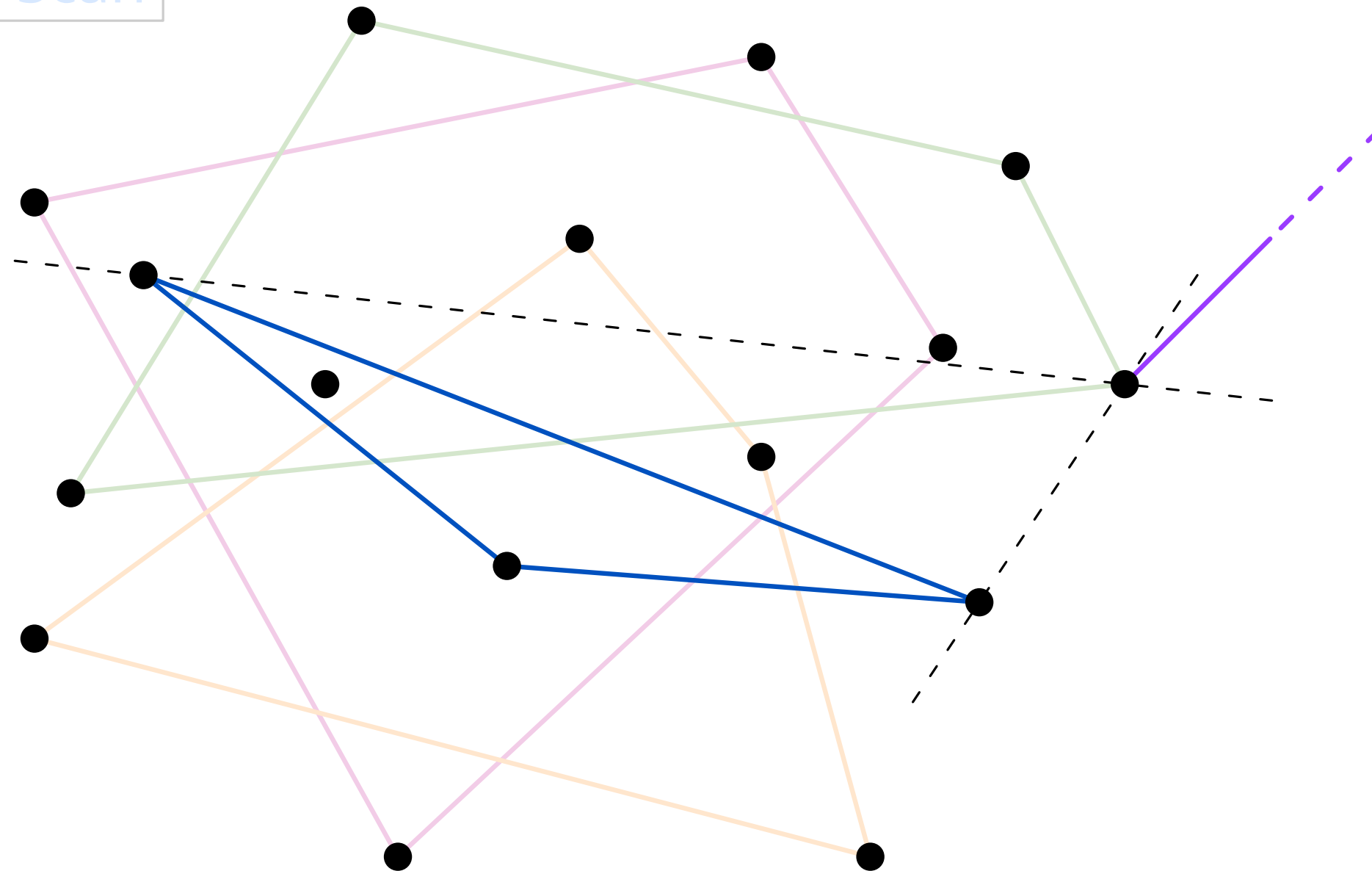


$n = 16$

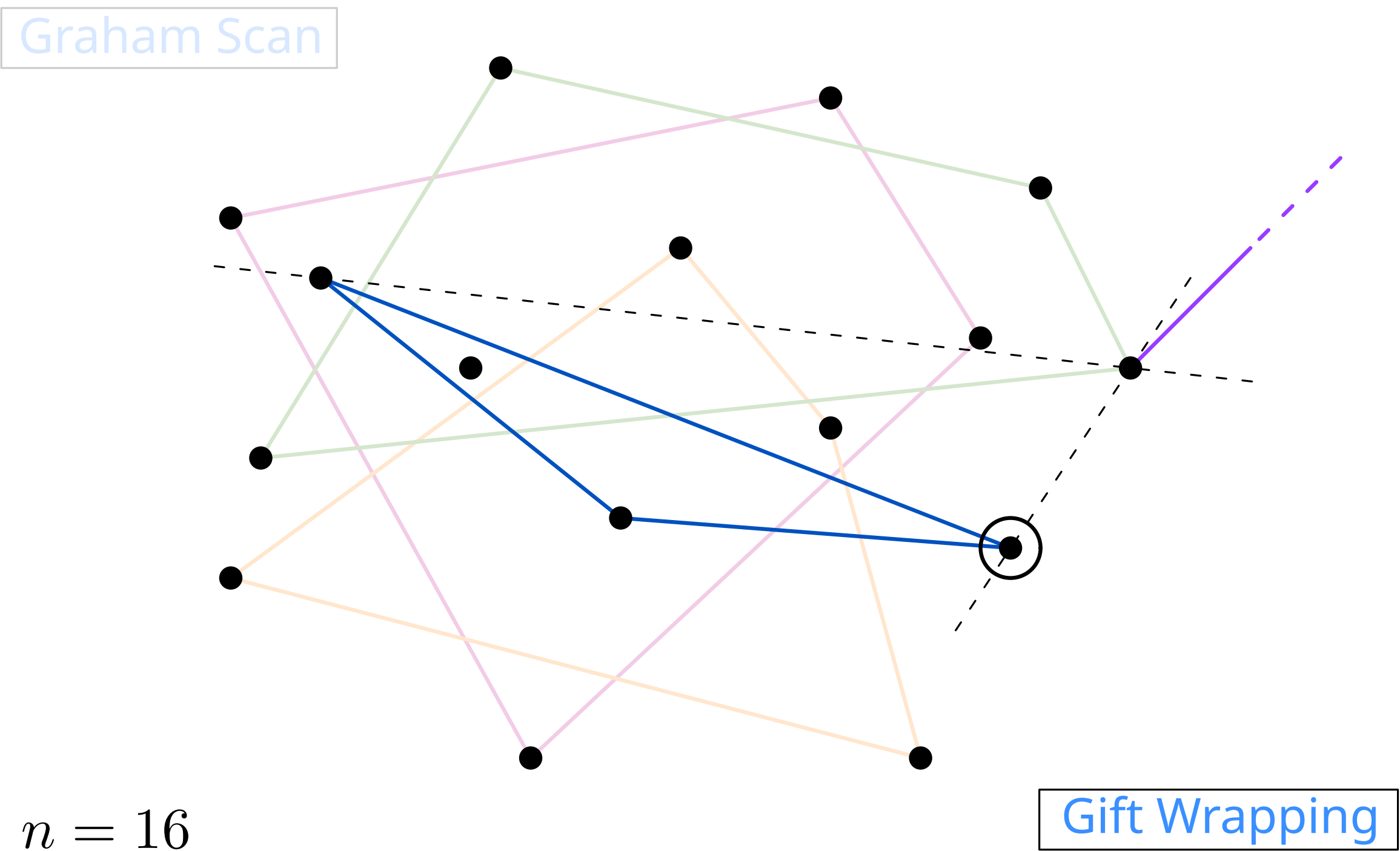
Gift Wrapping

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Graham Scan

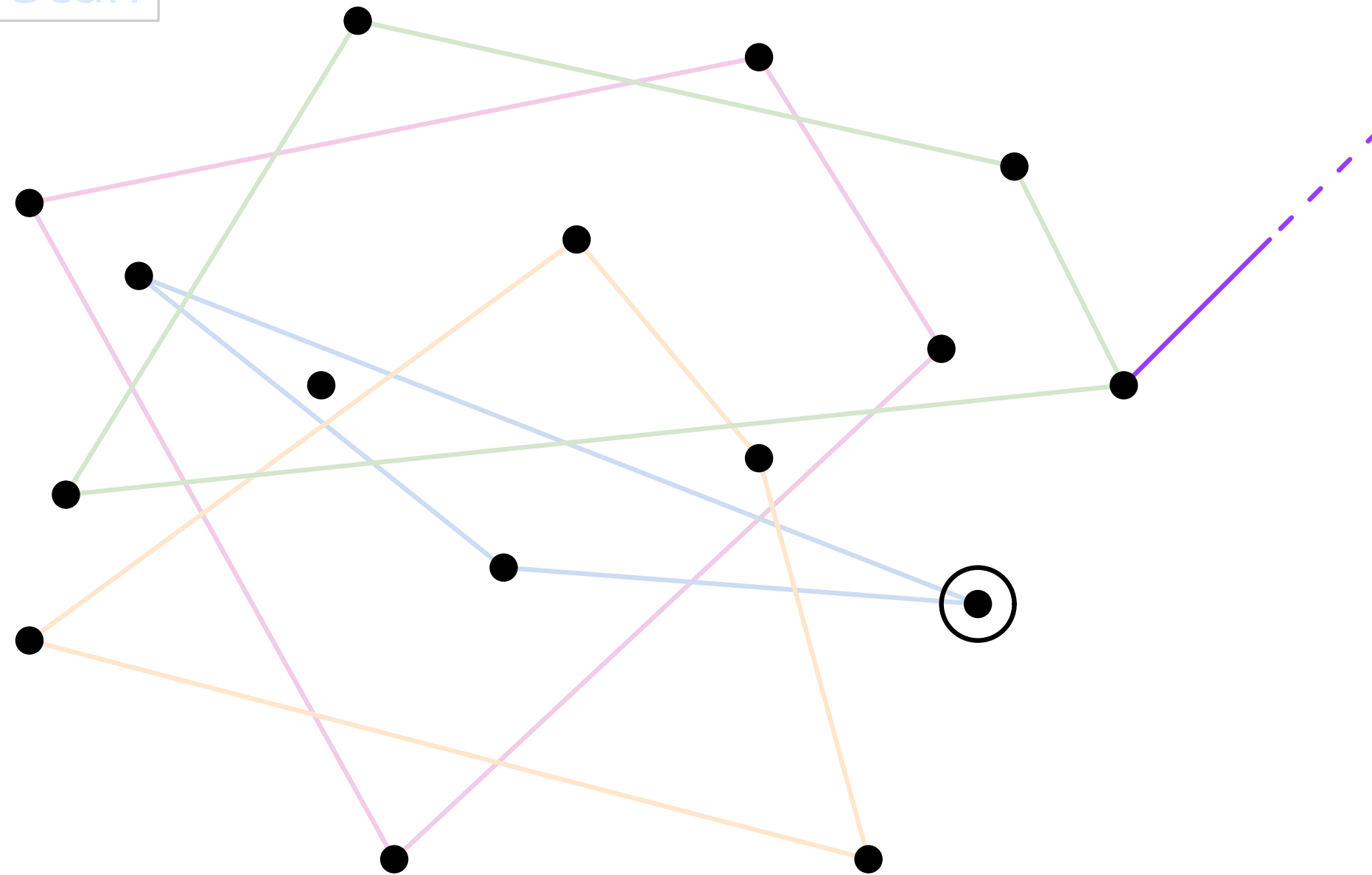


Example



Example

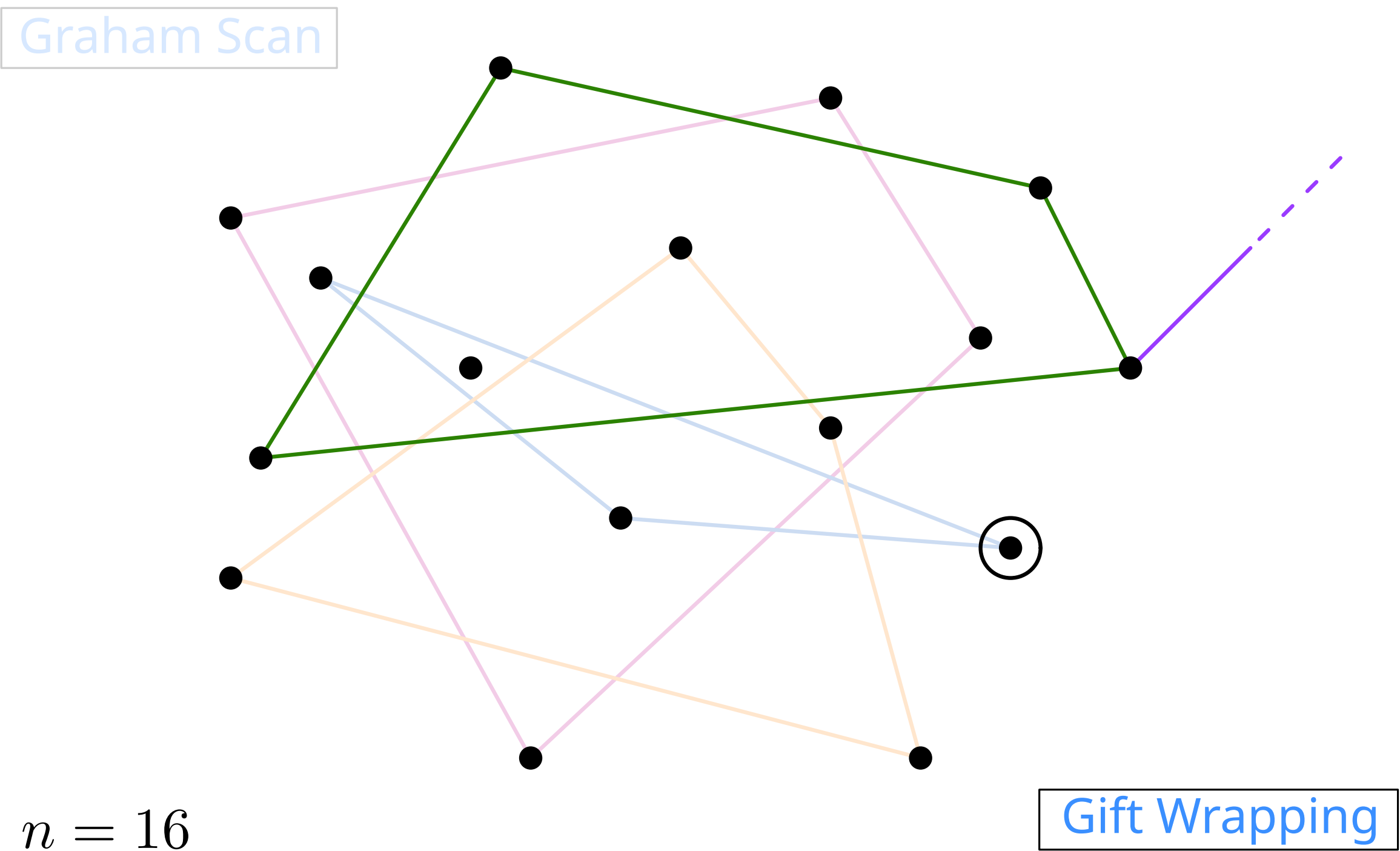
Graham Scan



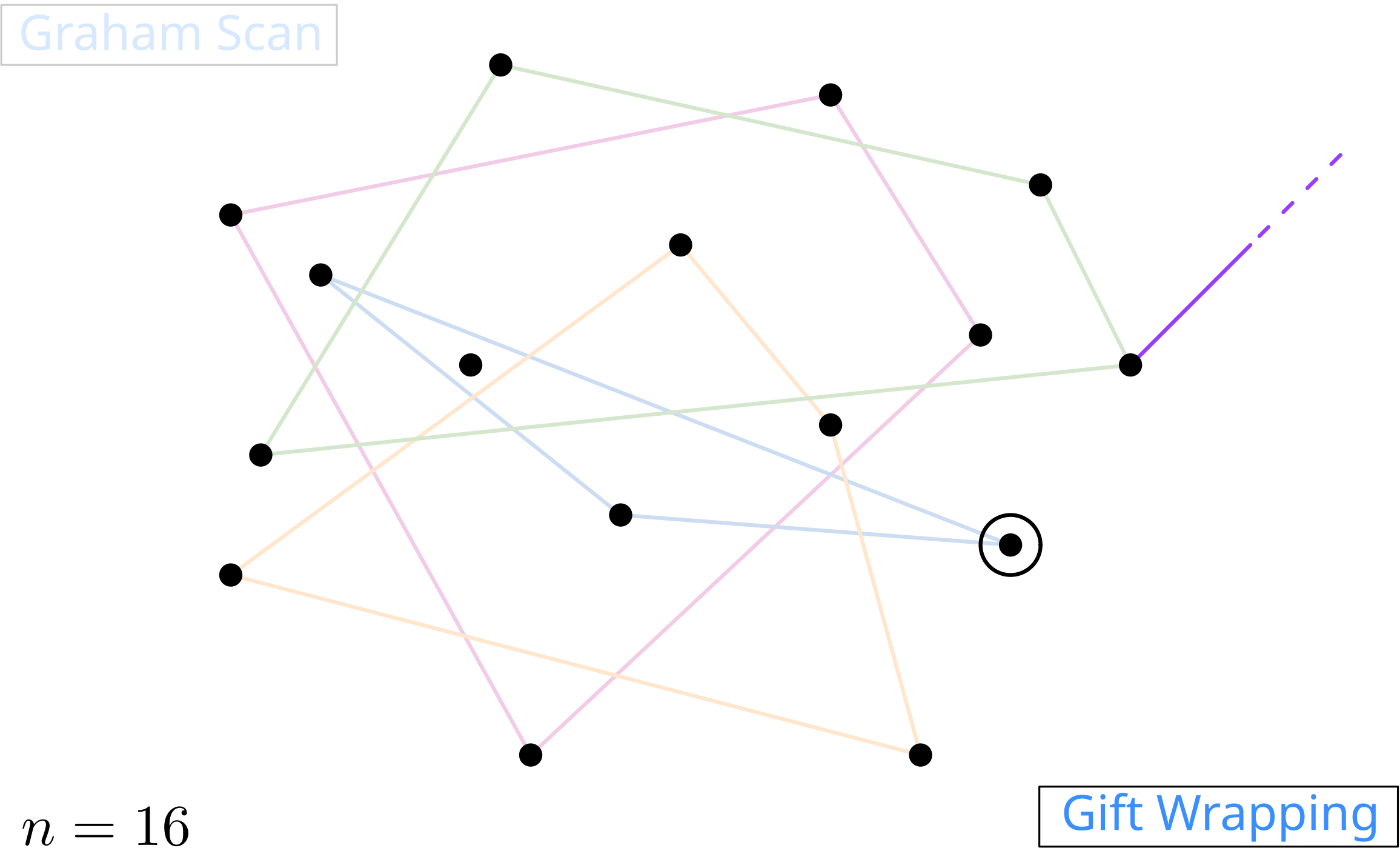
$n = 16$

Gift Wrapping

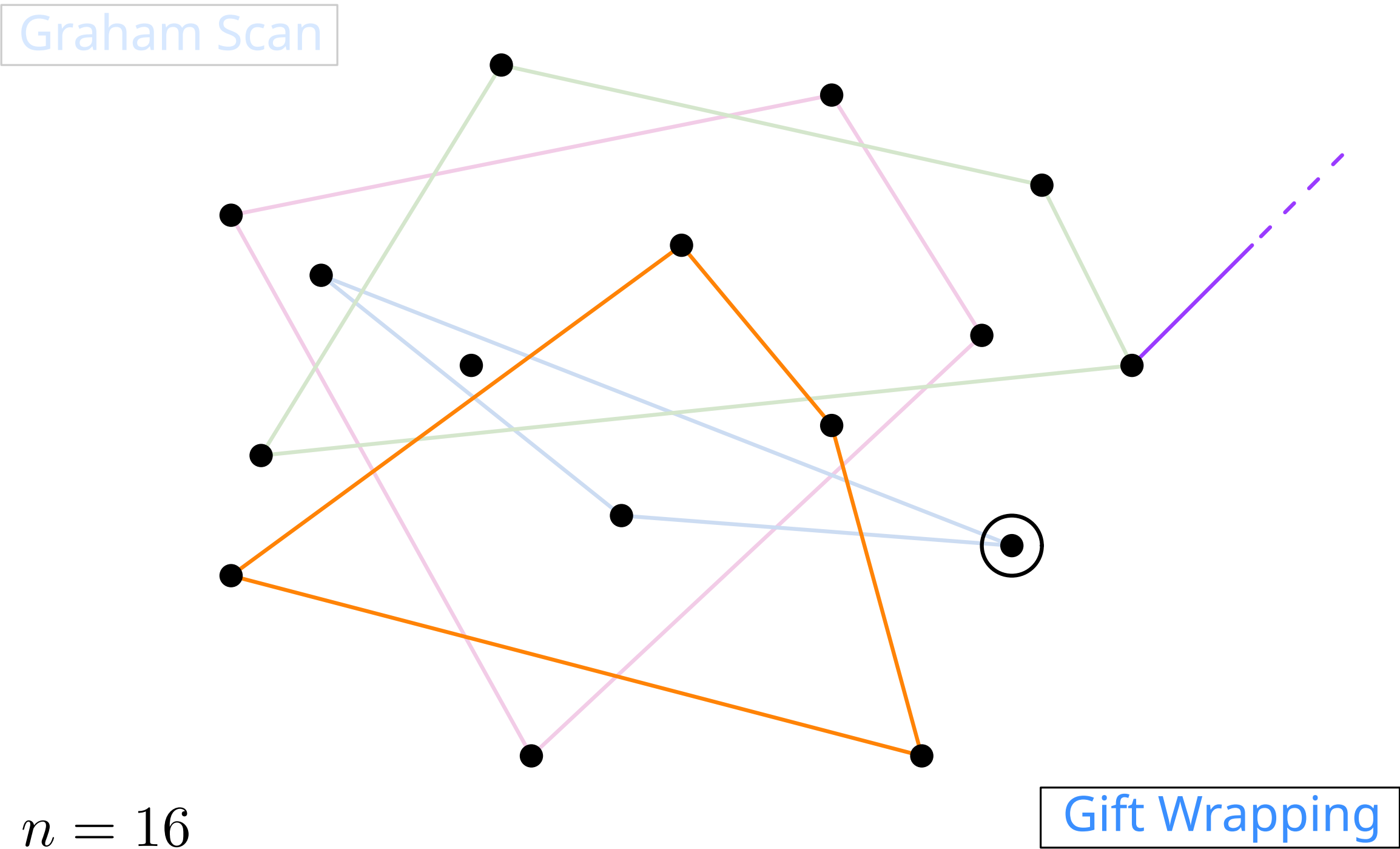
Example



Example



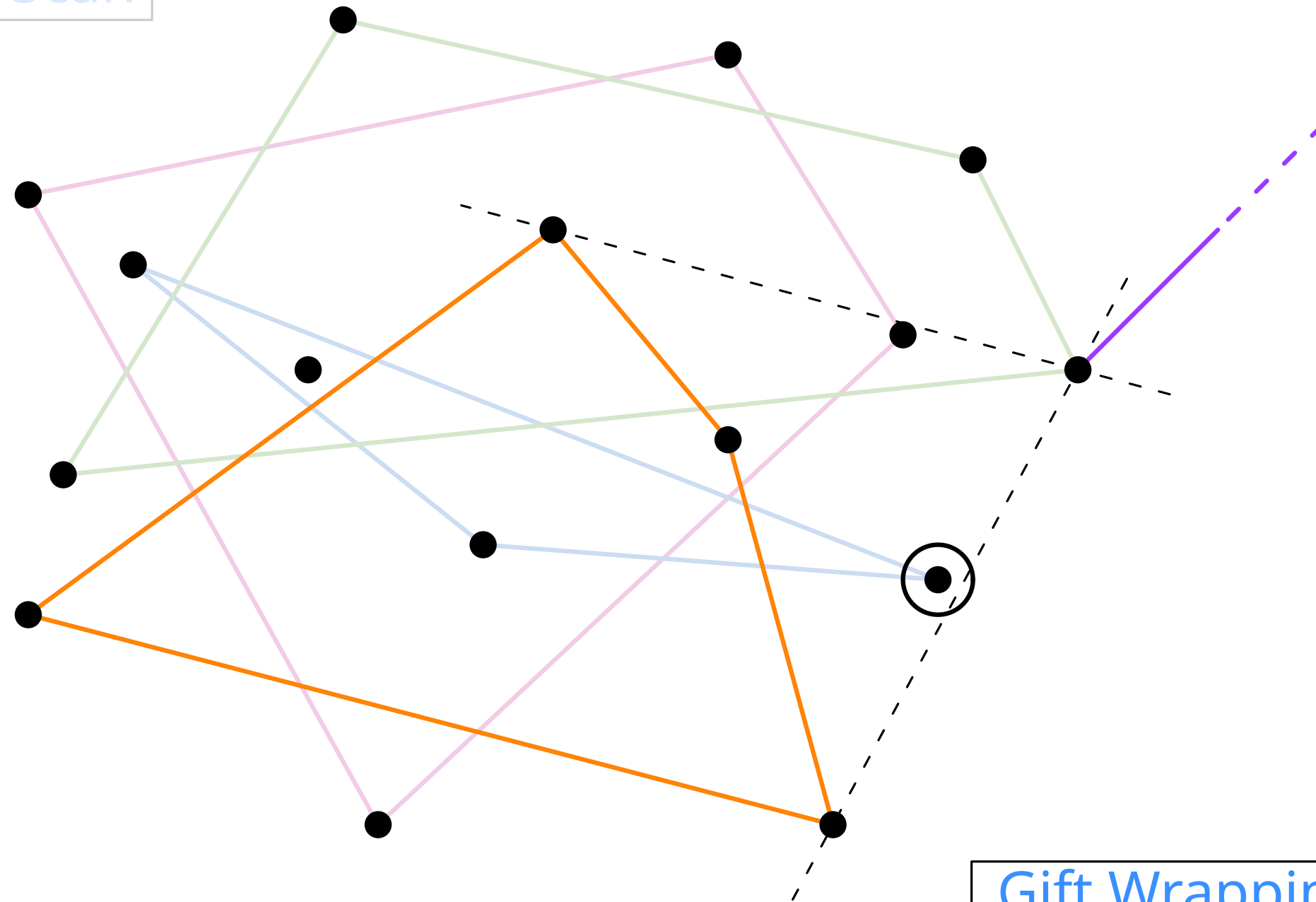
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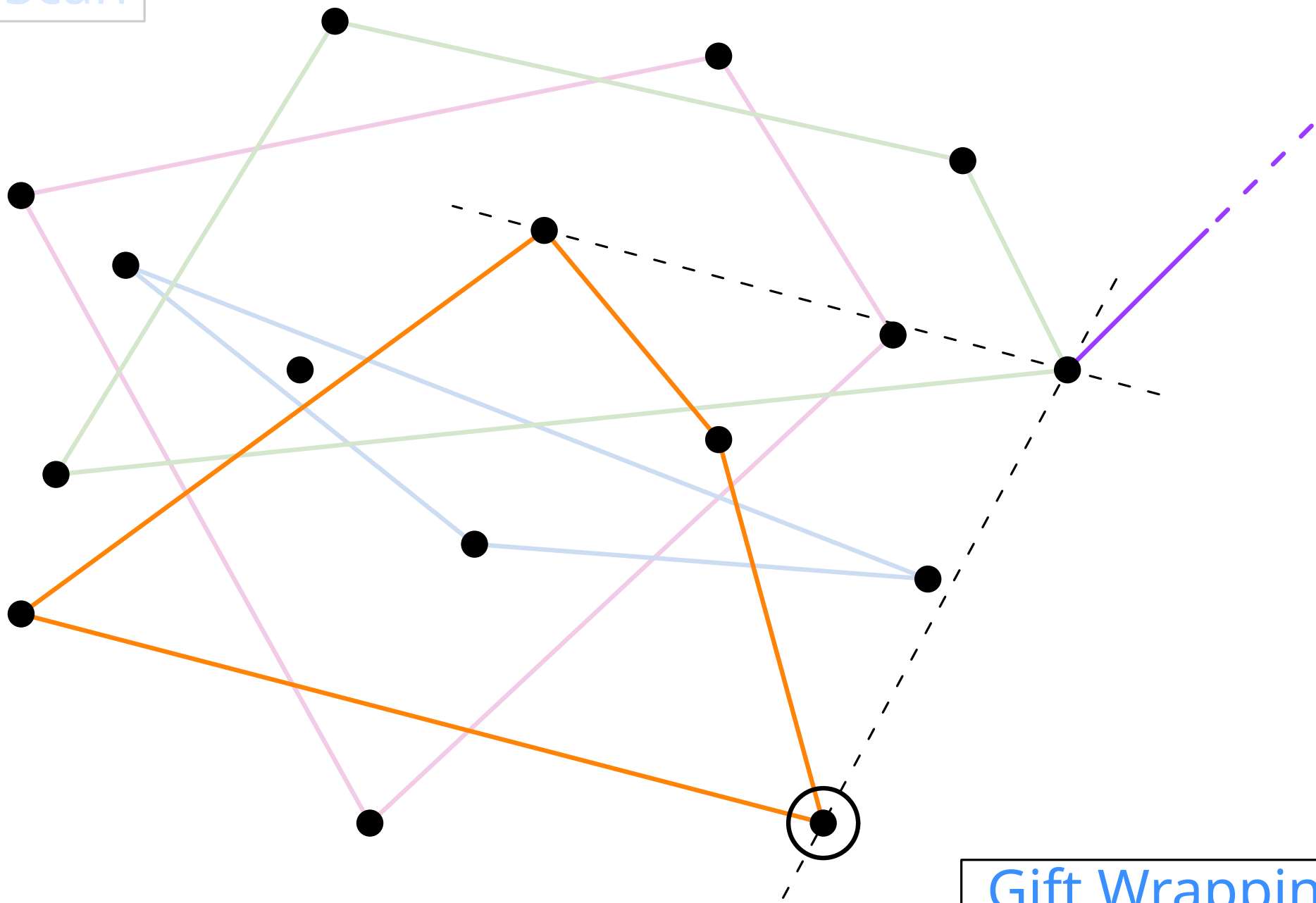
Gift Wrapping

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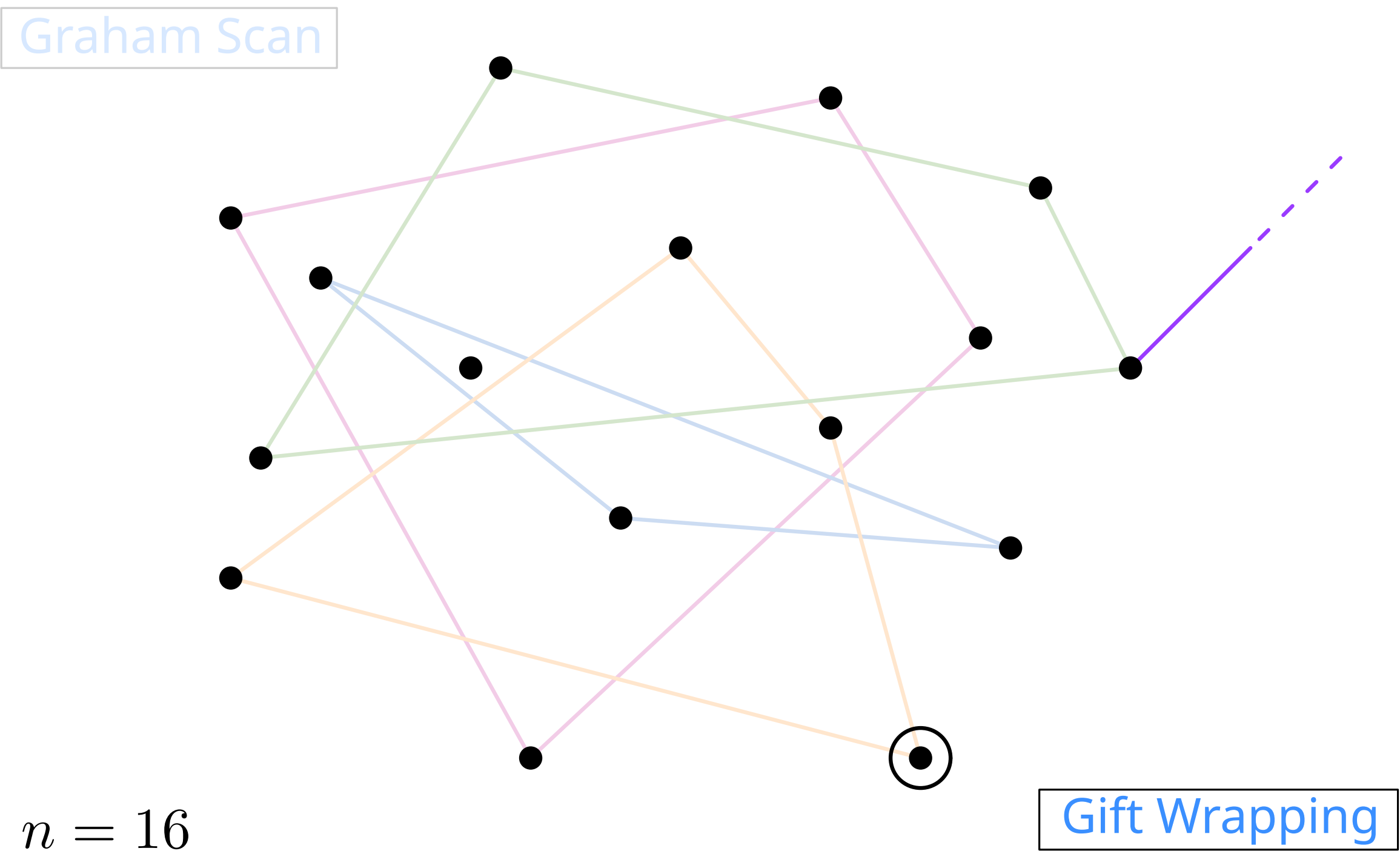
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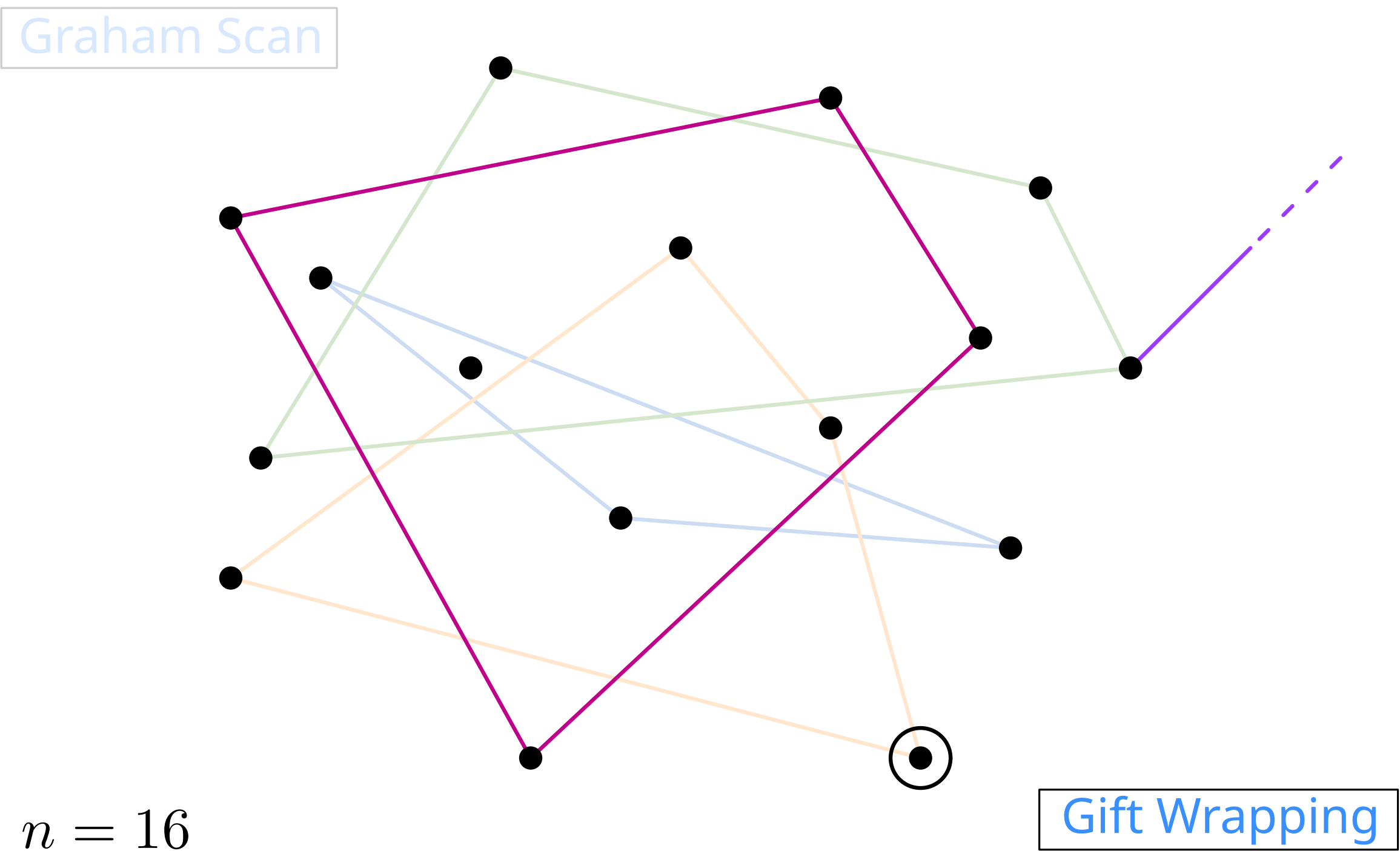
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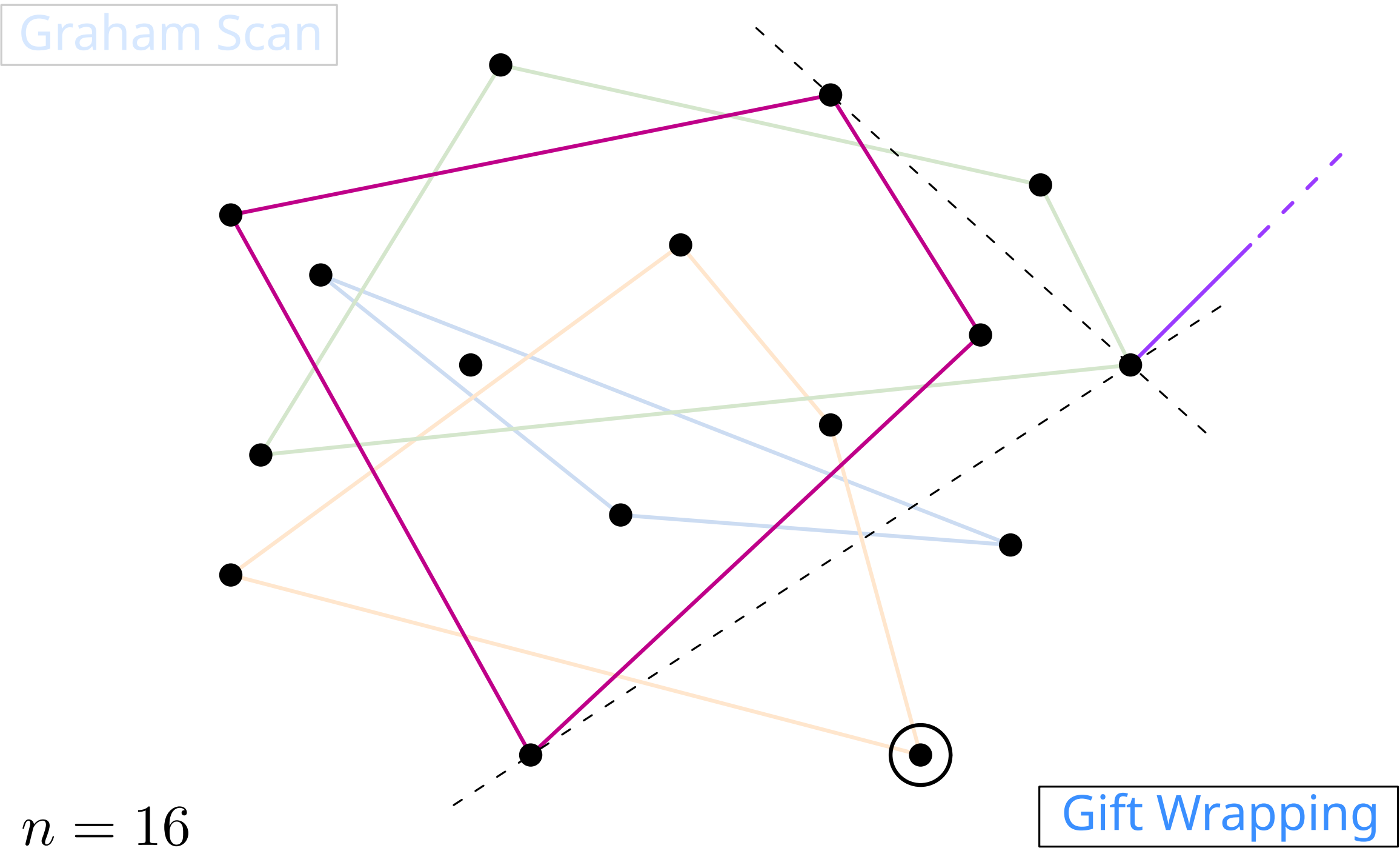
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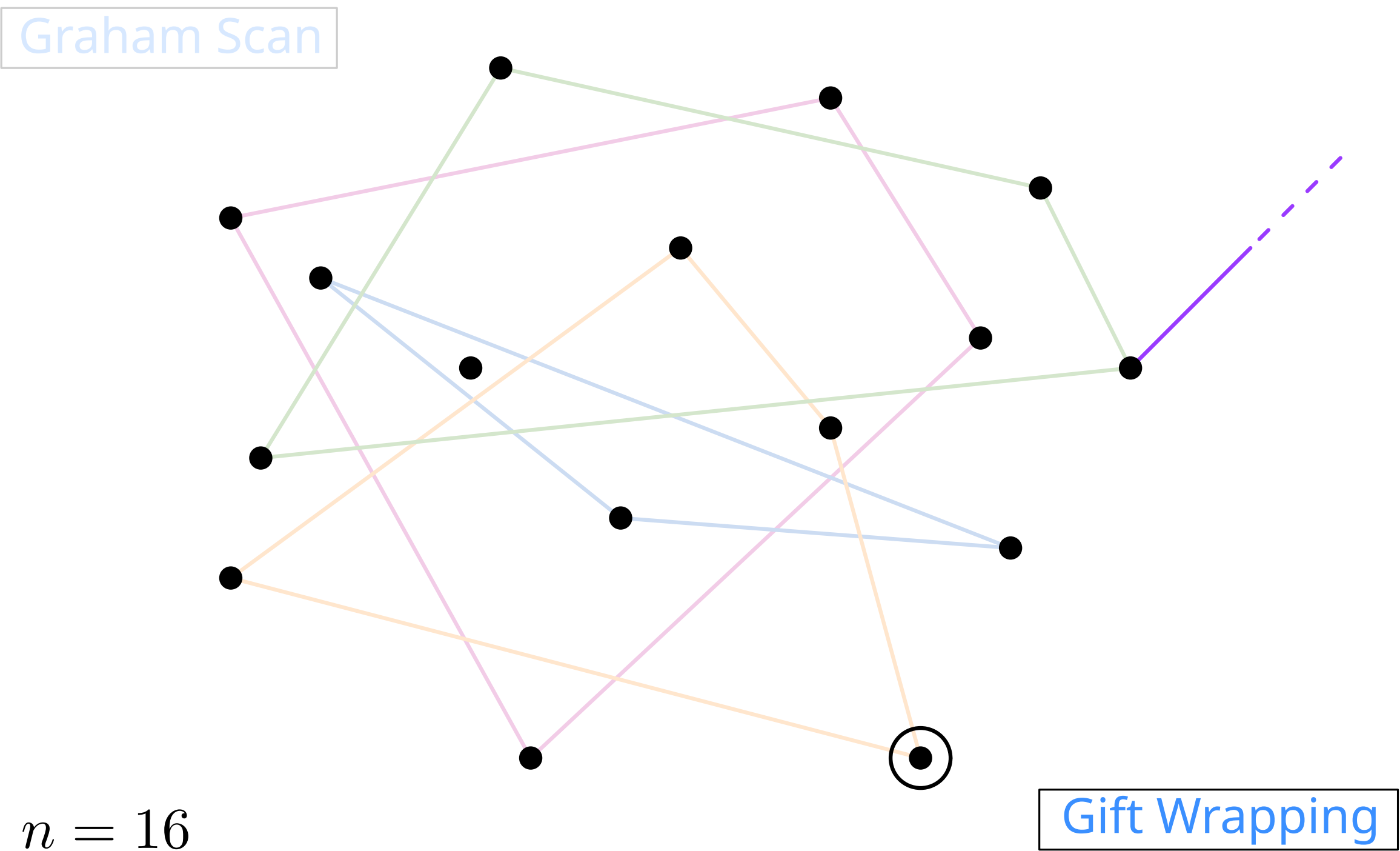
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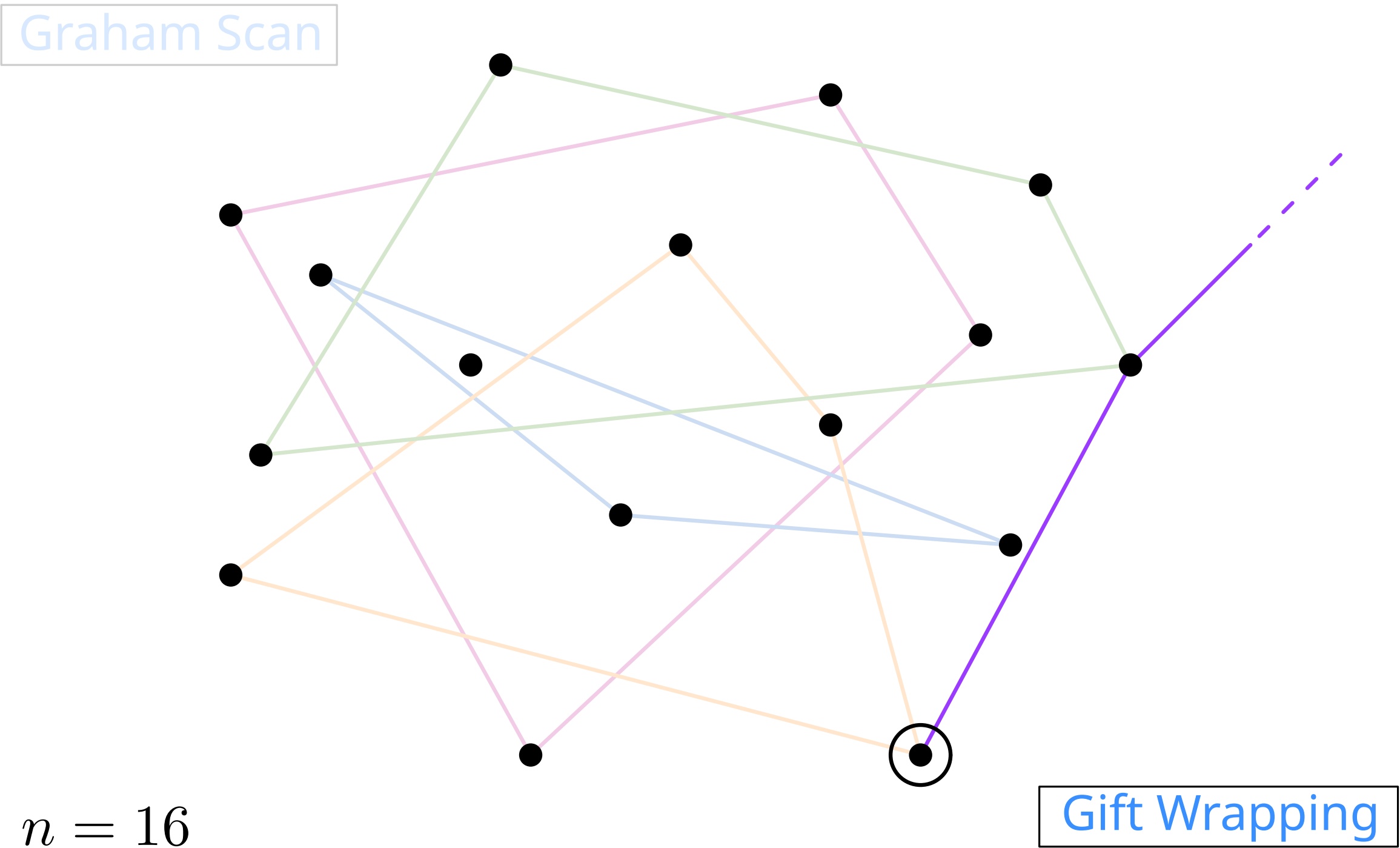
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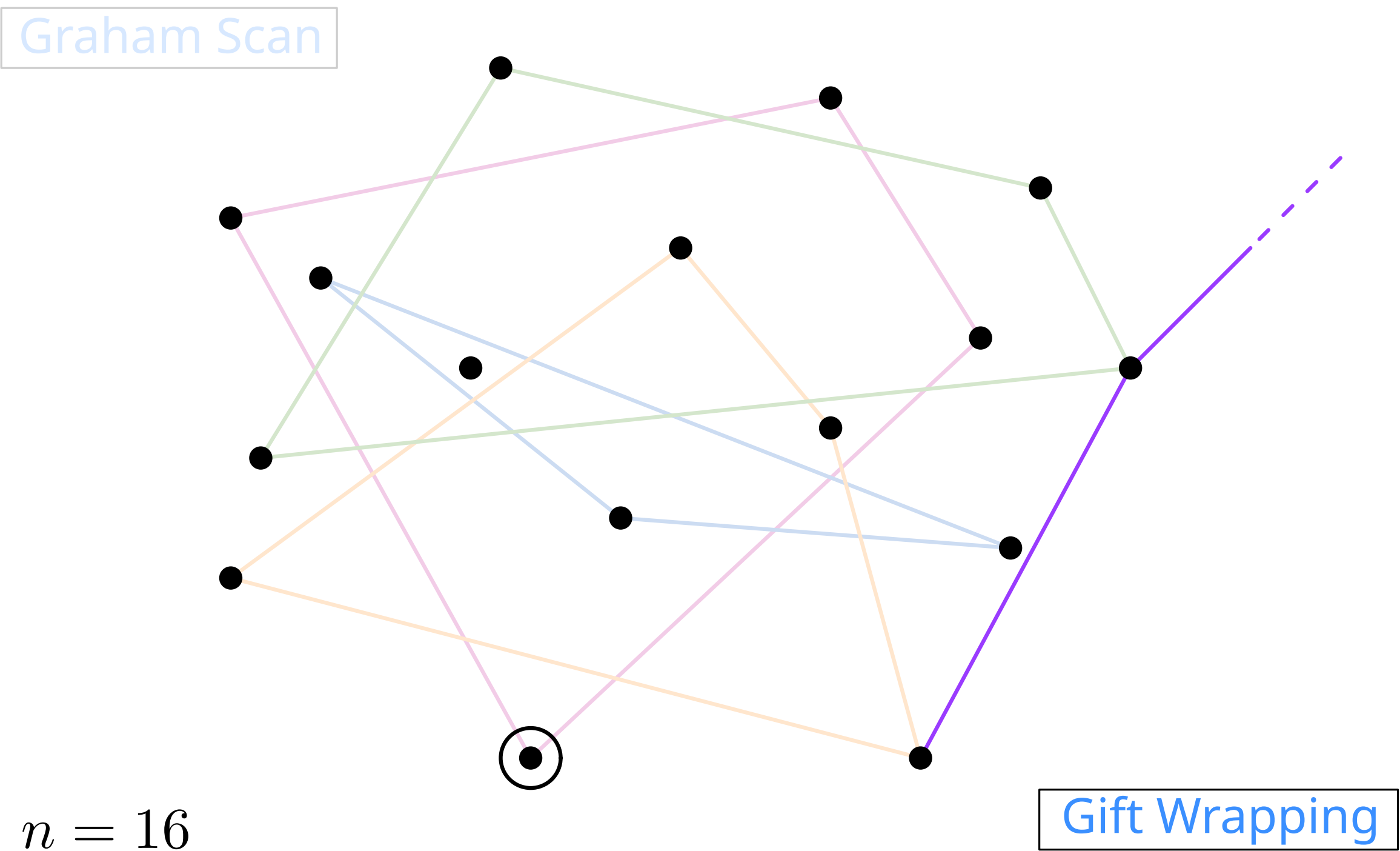
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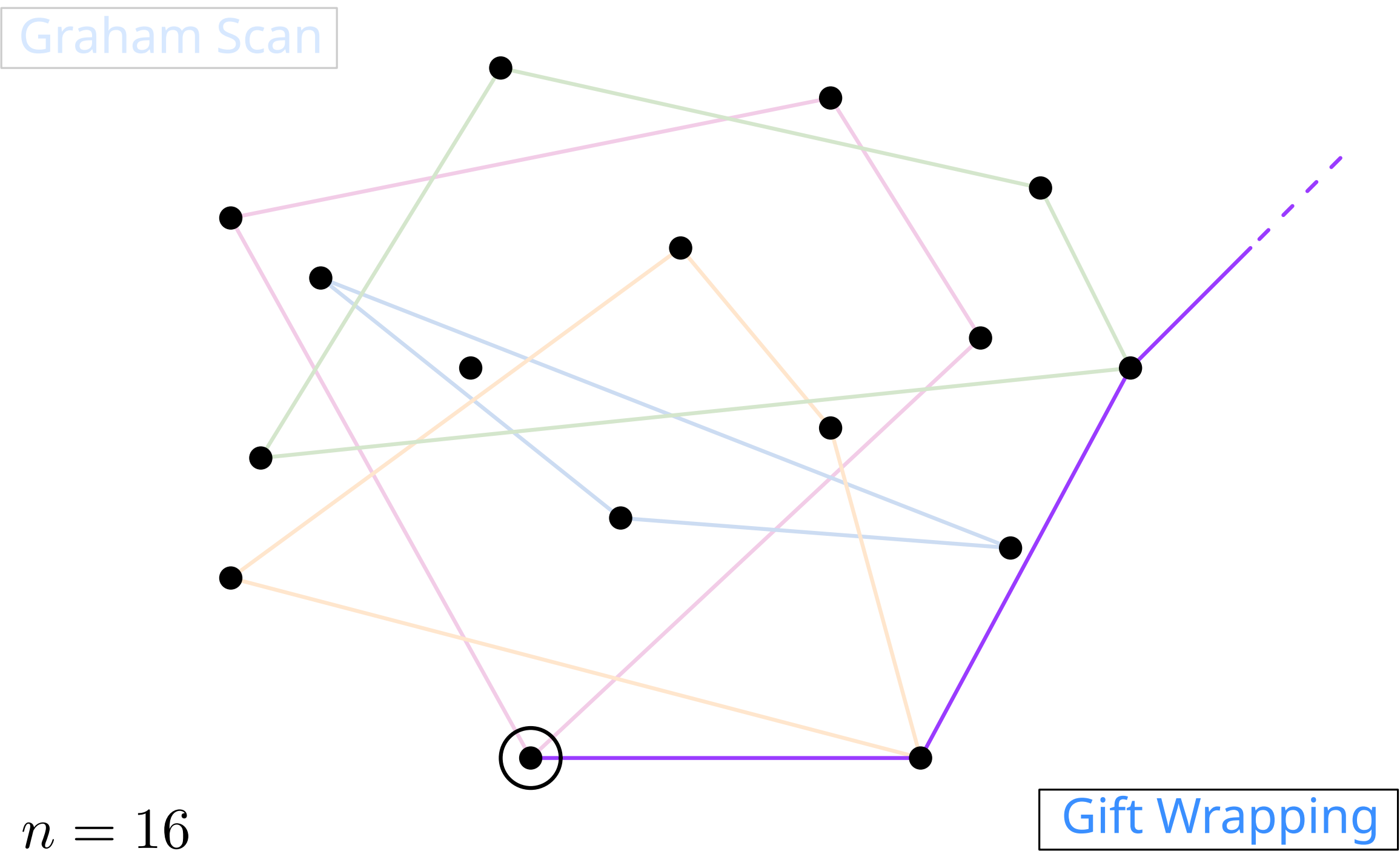
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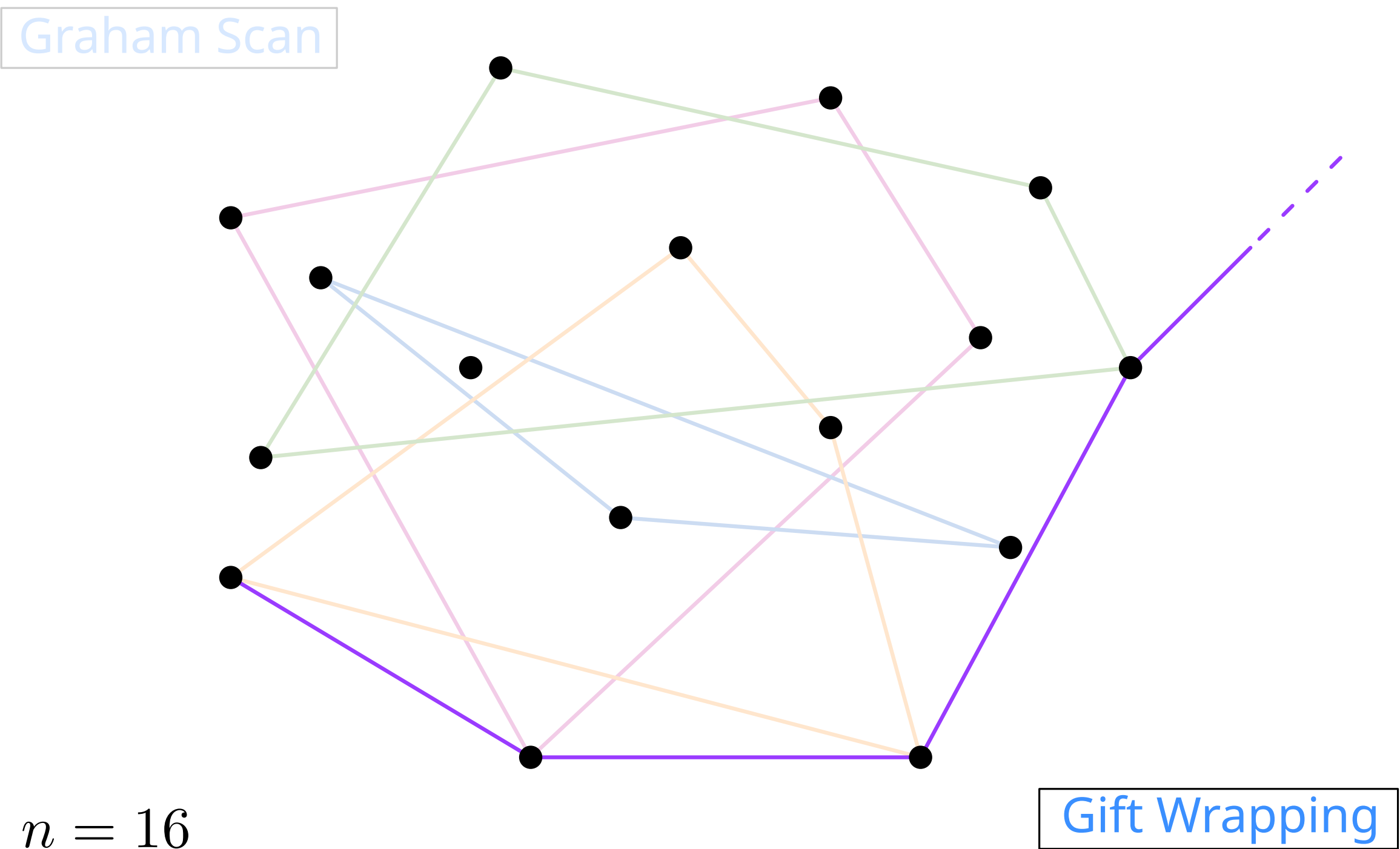
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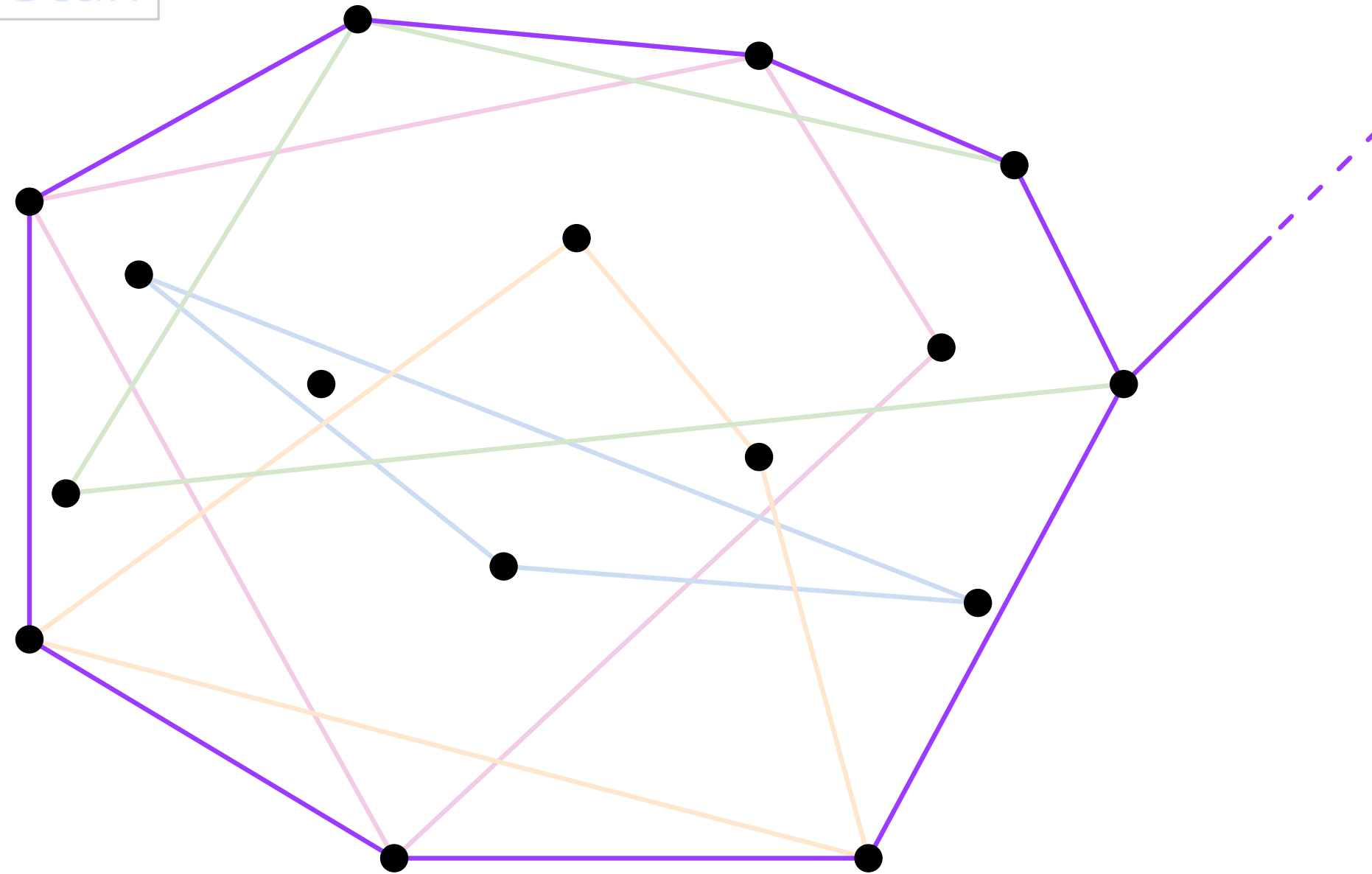


Example



Example

Graham Scan

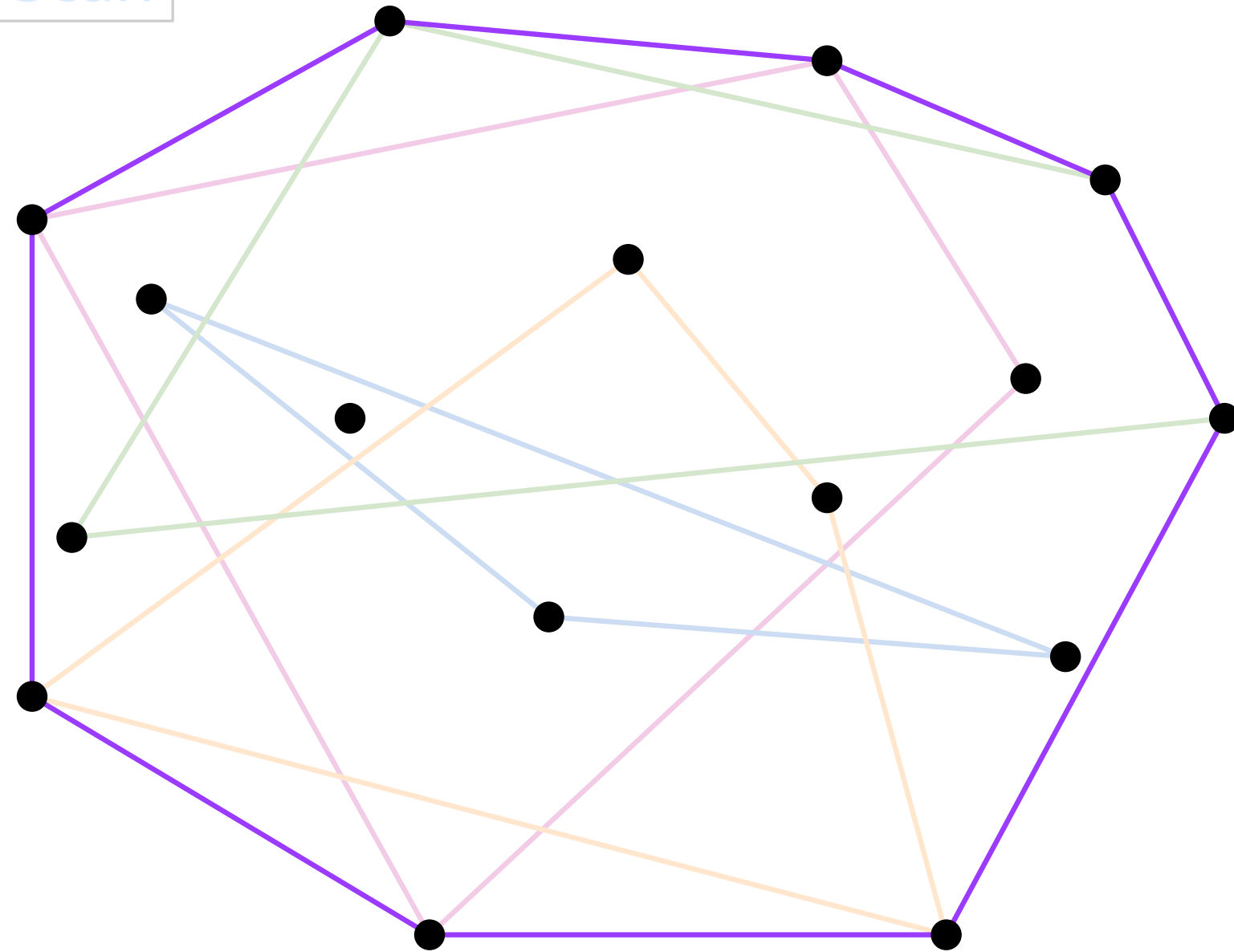


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Chan's Algorithms

But in general we do not know h

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- 2: **for** $i \leftarrow 1$ to n/m **do**
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- 10: **if** $p_{j+1} = p_1$ **then**
- 11: **return** L
- 12: **return** failure

What's up with m ?

Suggestions?

What's up with m ?

Algorithm FULLCHANHULL(P)

```
1: for  $t \leftarrow 0, 1, 2, \dots$  do  
2:    $m = \dots$   
3:    $result \leftarrow \text{CHANHULL}(P, m)$   
4:   if  $result \neq \text{failure}$  then  
5:     break  
6: return  $result$ 
```

What's up with m ?

Algorithm FULLCHANHULL(P)

```
1: for  $t \leftarrow 0, 1, 2, \dots$  do  
2:    $m = \min\{n, 2^{2^t}\}$   
3:    $result \leftarrow \text{CHANHULL}(P, m)$   
4:   if  $result \neq \text{failure}$  then  
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```

What's up with m ?

Algorithm FULLCHANHULL(P)

1: **for** $t \leftarrow 0, 1, 2, \dots$ **do**

2: $m = \min\{n, 2^{2^t}\}$

3: $result \leftarrow \text{CHANHULL}(P, m)$ $O(n \log m) = O(n \log 2^{2^t})$

4: **if** $result \neq \text{failure}$ **then**

5: **break**

6: **return** $result$

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} $\times O(\log \log h)$

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```

} $\times O(\log \log h)$

Running time:

$$\sum_{t=0}^{\log \log h} O(n \log 2^{2^t})$$

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2:    $m = \min\{n, 2^{2^t}\}$   
3:    $result \leftarrow \text{CHANHULL}(P, m)$     $O(n \log m) = O(n \log 2^{2^t})$   
4:   if  $result \neq \text{failure}$  then  
5:     break  
6: return  $result$ 
```

} $\times O(\log \log h)$

Running time:

$$\sum_{t=0}^{\log \log h} O(n \log 2^{2^t}) = O(n) \sum_{t=0}^{\log \log h} O(2^t)$$

What's up with m ?

Algorithm FULLCHANHULL(P)

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} $\times O(\log \log h)$

Running time:

$$\sum_{t=0}^{\log \log h} O(n \log 2^{2^t}) = O(n) \sum_{t=0}^{\log \log h} O(2^t) \leq O(n) \cdot O(2^{\log \log h + 1})$$

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$$= O(n) \cdot O(\log h) = O(n \log h)$$

Summary

Slow Convex Hull $O(n^3)$

Graham Scan $O(n \log n)$

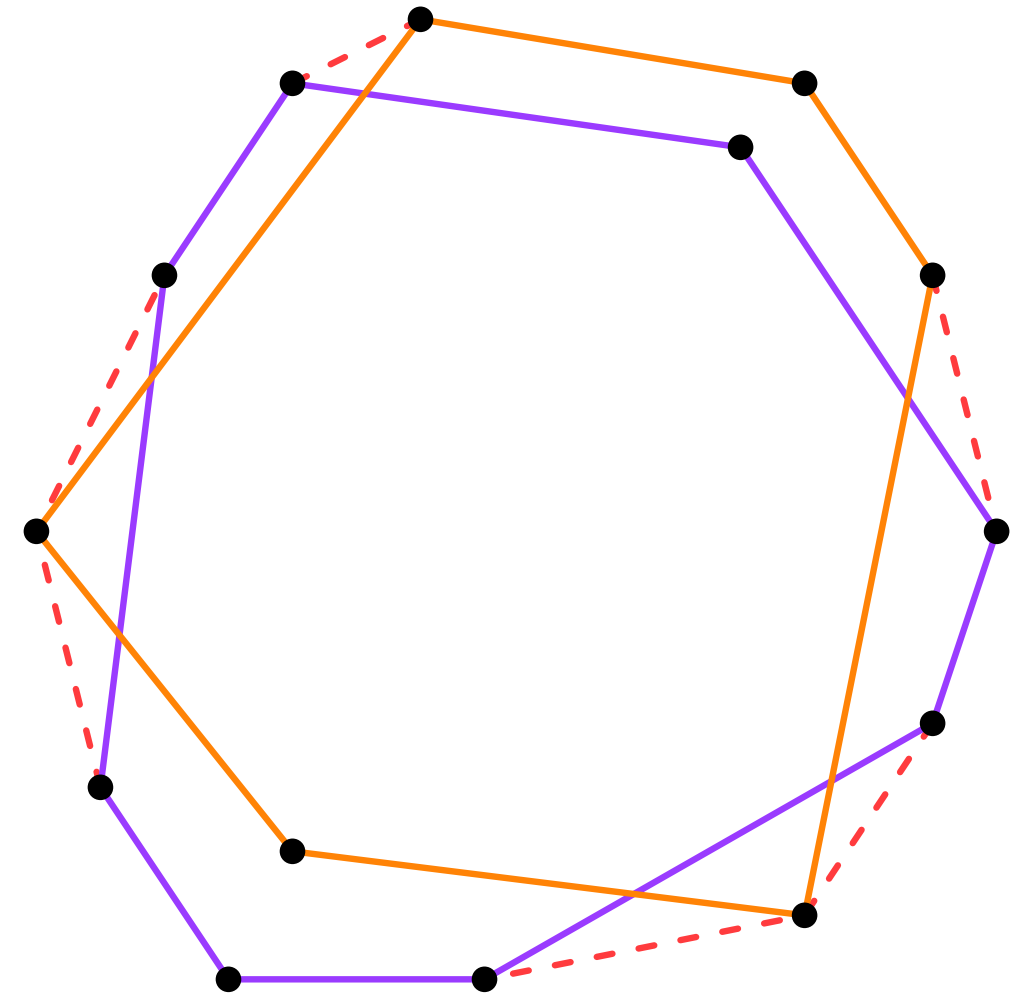
Gift Wrapping $O(nh)$

Chan's Hull $O(n \log h)$

Other approaches: divide and conquer

Split the point set in two halves, compute the convex hulls recursively, and merge

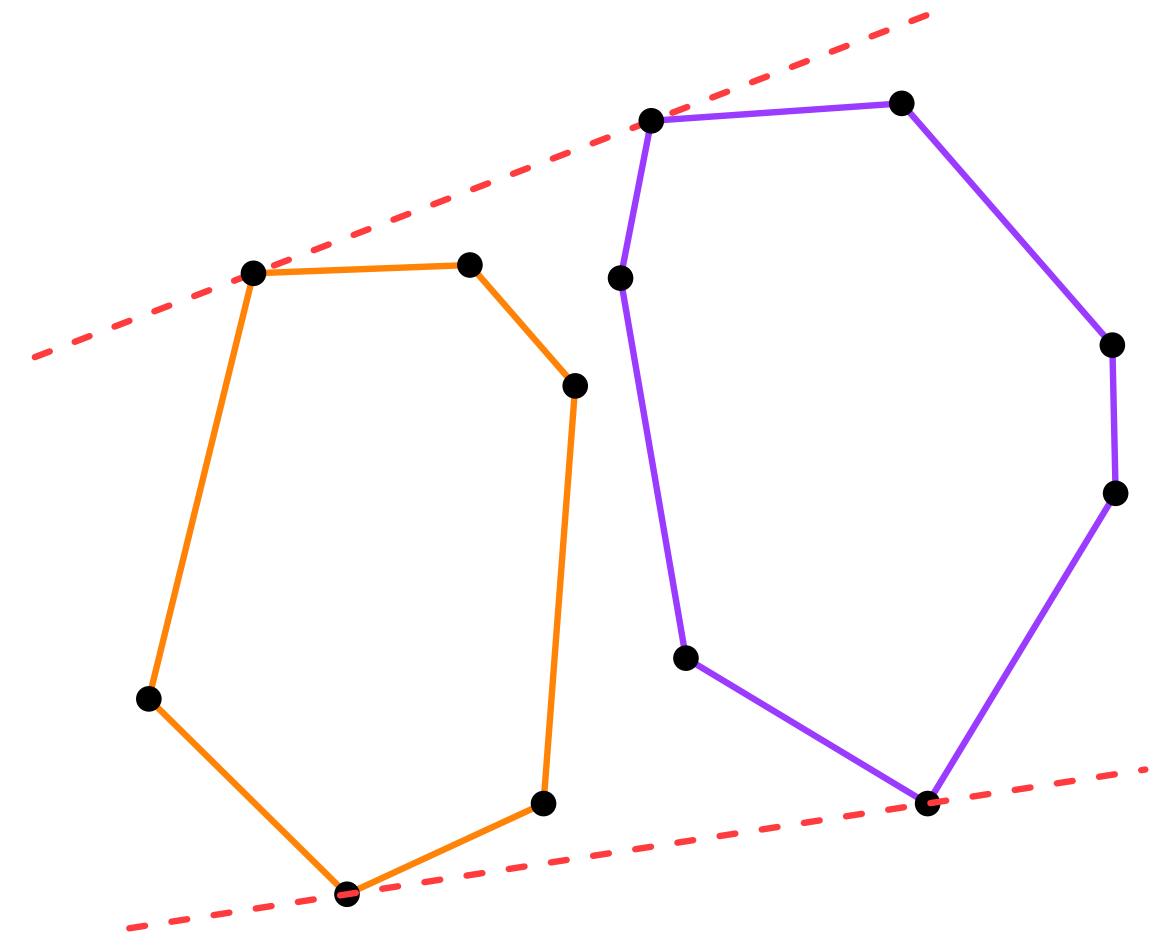
The merge step involves finding “extreme vertices” in every direction



Other approaches: divide and conquer

Alternatively: split the point set in two halves on x -coordinate, compute the convex hulls recursively, and merge

The merge step now comes down to finding two common tangent lines



Convex hulls in 3D

For a 3-dimensional point set, the convex hull is a convex polyhedron

It has vertices (0-dim.), edges (1-dim.), and facets (2-dim.) on its boundary, and a 3-dimensional interior

The boundary is a planar graph, so it has $O(n)$ vertices, edges, and facets



Convex hulls in 4D

For a 4-dimensional point set, the convex hull is a convex polyhedron

It has vertices (0-dim.), edges (1-dim.), 2-facets (2-dim.), and 3-facets (3-dim.) on its boundary, and a 4-dimensional interior

Its boundary can have $\Theta(n^2)$ facets in the worst case!