



Convex Hulls (2D)

O'Rourke, Chapter 3



Announcements

- We are live on Piazza:
<http://piazza.com/jhu/spring2016/600459>
- Code-base was updated last Wednesday
(Fixes a bug where y -values are read in as zero)



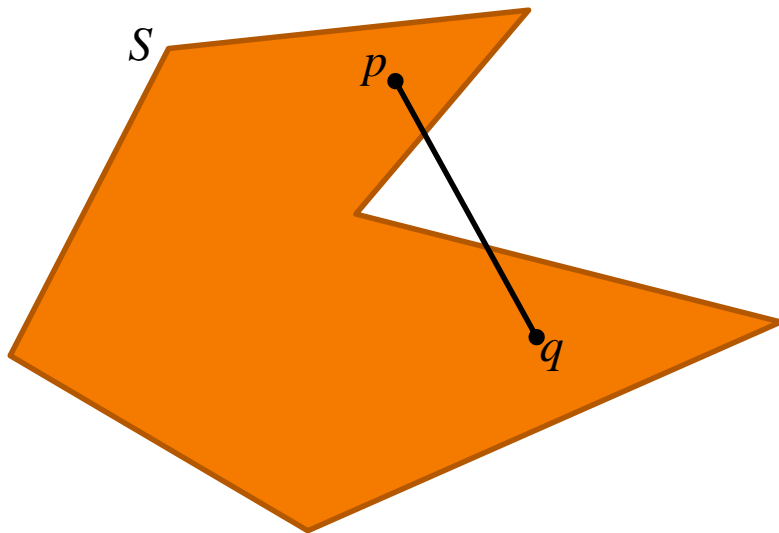
Outline

- Convex Hulls
- Algorithms
 - Naïve Implementation(s)
 - Gift Wrapping
 - Quick Hull
 - Graham's Algorithm
 - Lower bound complexity

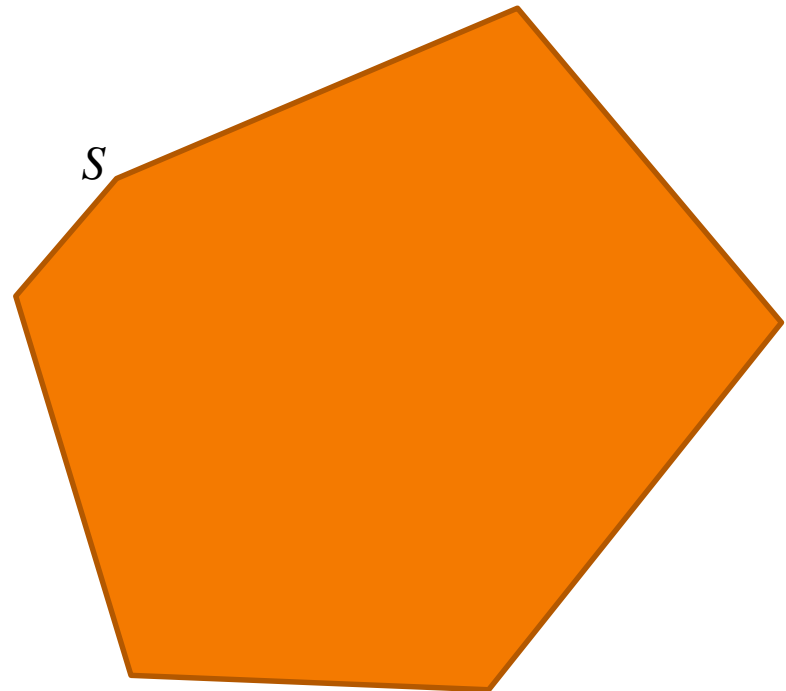


Convexity

A set S is *convex* if for any two points $p, q \in S$ the line segment $\overline{pq} \subset S$.



Not convex

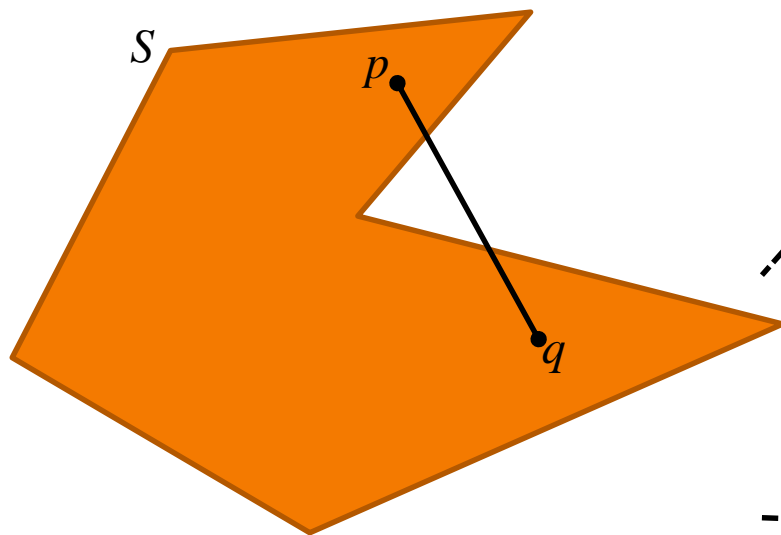


Convex?

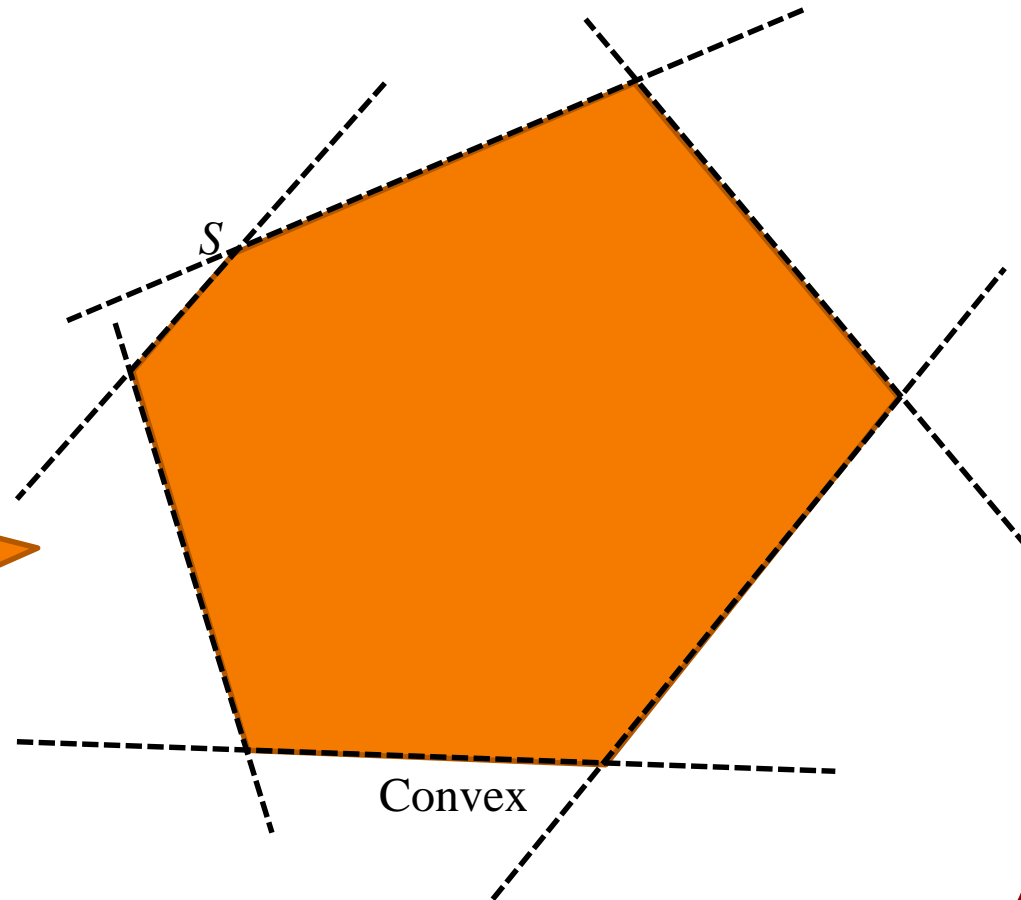


Convexity

A set S is *convex* if it is the intersection of (possibly infinitely many) half-spaces.



Not convex



Convex



Convexity

Given points $\{p_1, \dots, p_n\} \subset \mathbb{R}^d$, a point $q \in \mathbb{R}^d$ is a *convex combination* of the points if q can be expressed as the linear sum:

$$q = \sum_{i=1}^n \alpha_i \cdot p_i$$

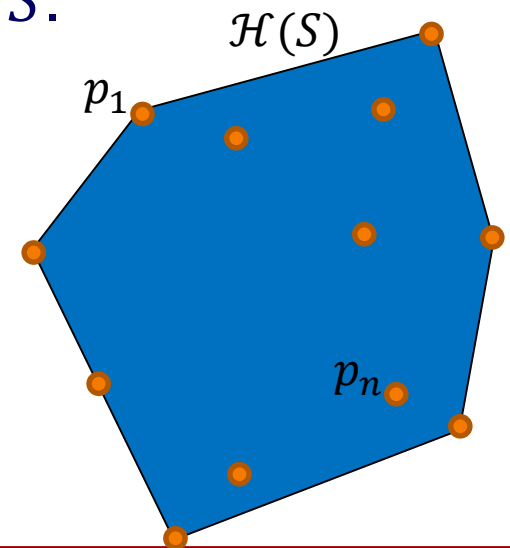
with $\alpha_i \geq 0$ and $\alpha_1 + \dots + \alpha_n = 1$.



Convex Hull

The *convex hull* of a set of points $S \subset \mathbb{R}^d$, denoted $\mathcal{H}(S)$, is the:

- set of all convex combinations of points in S ,
- set of all convex combinations of $d + 1$ points in S ,
- intersection of all convex sets C w/ $S \subset C$,
- intersection of all half-spaces H w/ $S \subset H$,
- smallest convex polygon containing S .



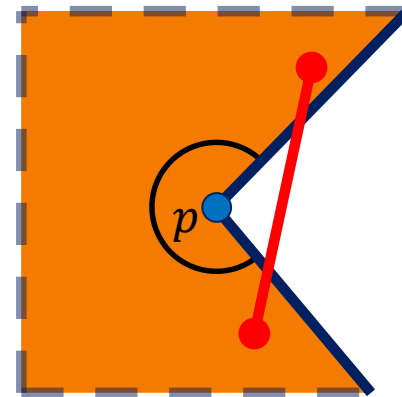


Convex Hull

Note:

If $p \in S \subset \mathbb{R}^2$ and p is a vertex of the convex hull then p must be a convex vertex.

Otherwise, we could create a line segment with vertices inside of the hull but which isn't strictly interior.

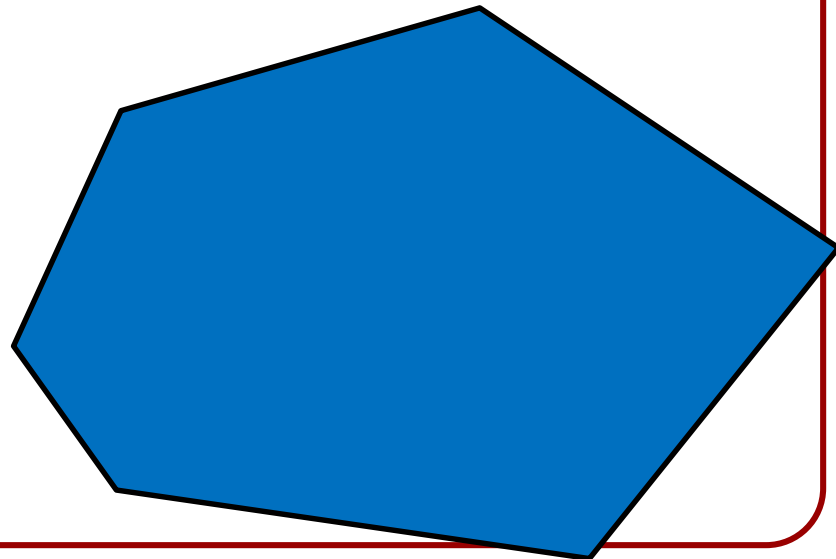




Convex Hull

Claim:

If $P \subset \mathbb{R}^2$ is a polygon whose vertices are all convex, then P is convex.



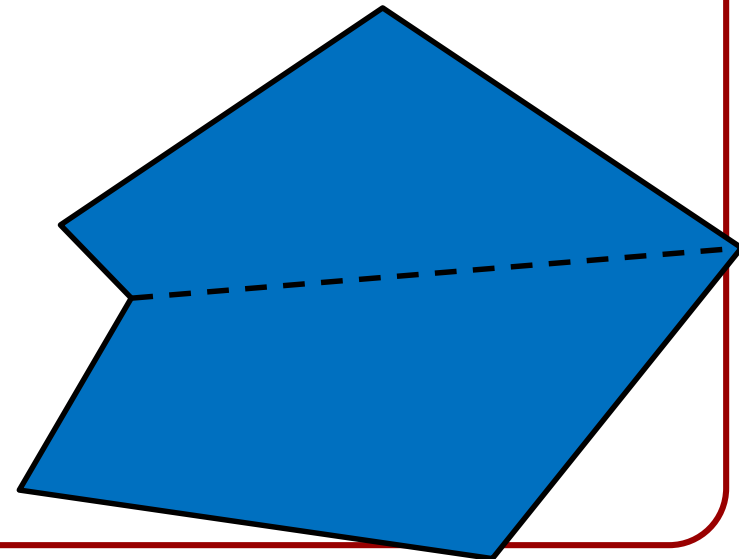


Convex Hull

Proof (by induction):

Otherwise, we could add a diagonal.

⇒ By induction, each half is convex.





Convex Hull

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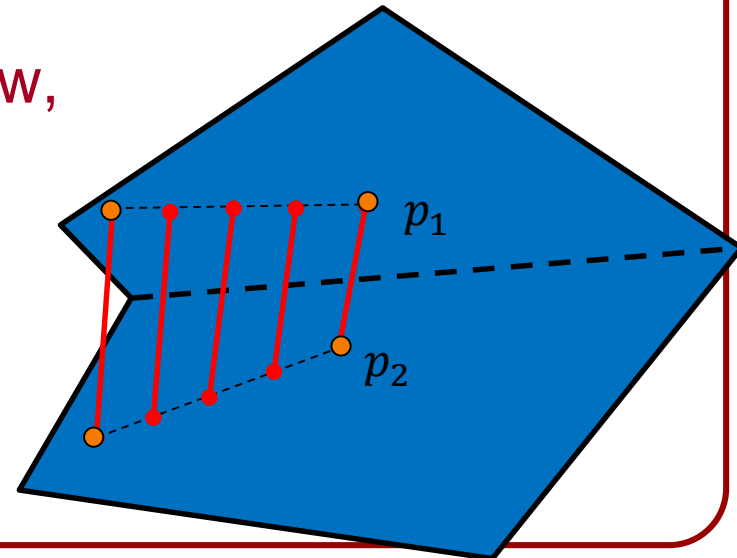
Otherwise, we could add a diagonal.

⇒ If P is not convex there must be a segment between the two parts that exits P .

Choose p_1 and p_2 above/below the diagonal.

Evolve the segment to $\overline{p_1 p_2}$.

Since p_1 and p_2 are above/below, $\overline{p_1 p_2}$ crosses the diagonal and is entirely inside P .





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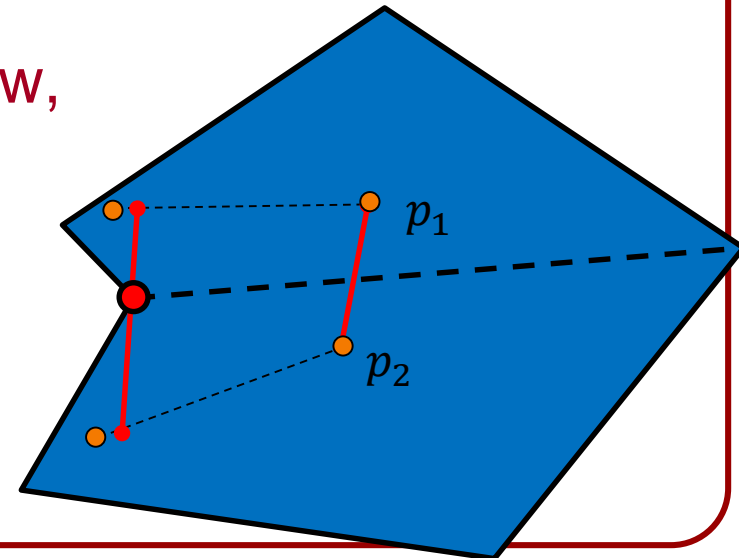
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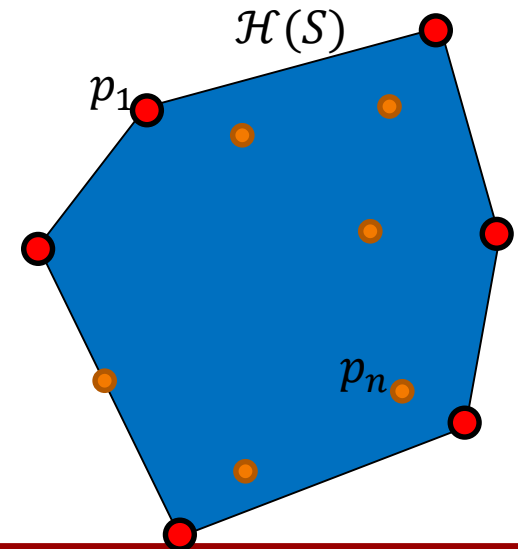
The last point at which the evolving segment is not fully inside must be a reflex vertex.





Convex Hull

The *extreme points* of a set of points $S \subset \mathbb{R}^2$ are the points which are on the convex hull and have interior angle strictly less than π .



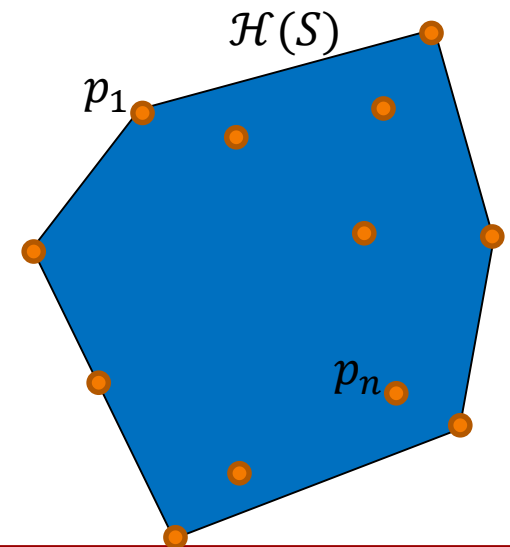


Convex Hull

Goal:

Given a set of points $S = \{p_1, \dots, p_n\} \subset \mathbb{R}^d$, compute the convex hull $\mathcal{H}(S)$ efficiently.

- Do we want all points on the hull or just the extreme ones?
- Do the output vertices need to be sorted or is the set of (extreme) vertices sufficient?



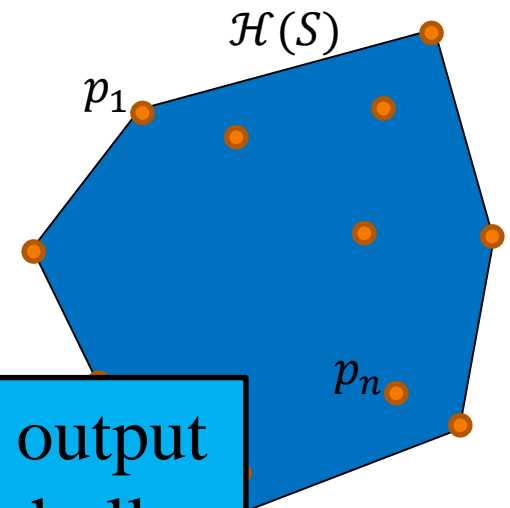


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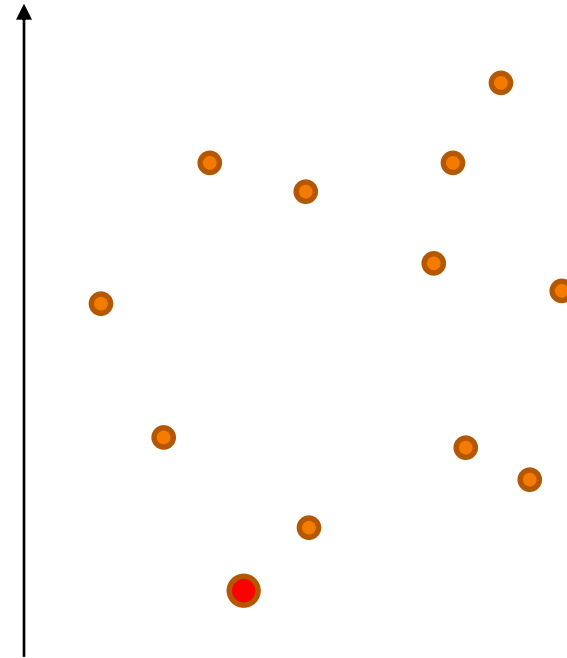
We will focus on the ordered output of the extreme points on the hull.



Convex Hull (2D)

Note:

We can find a hull vertex in linear time by finding the vertex that is extremal w.r.t. to some direction.





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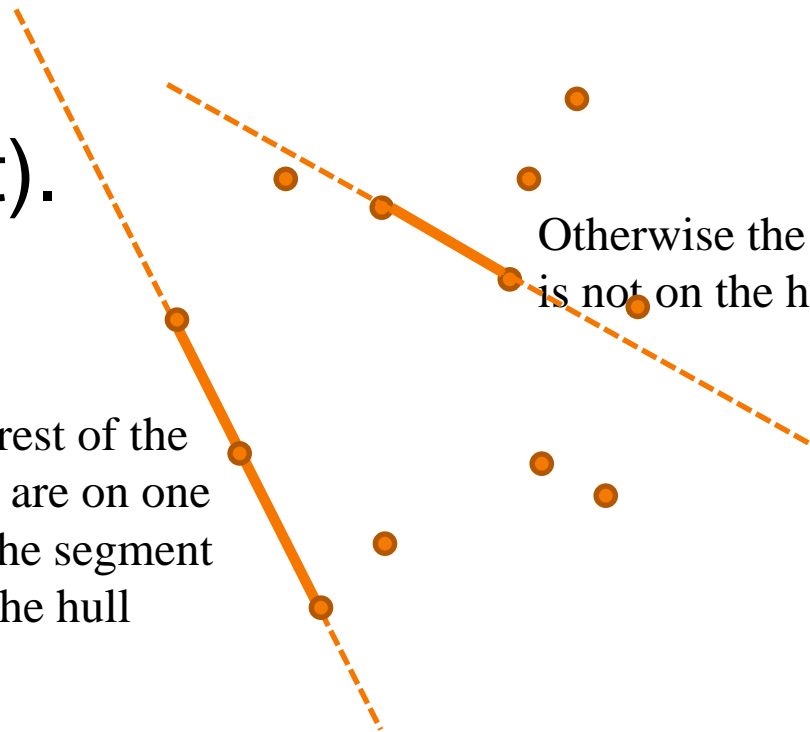
Convex Hull (2D)

Naïve Algorithm:

For each directed edge $e \in S \times S$, check if the half-space to the right of e is empty of points (and there are no points on the line outside the segment).

If the rest of the points are on one side, the segment is on the hull

Otherwise the segment is not on the hull





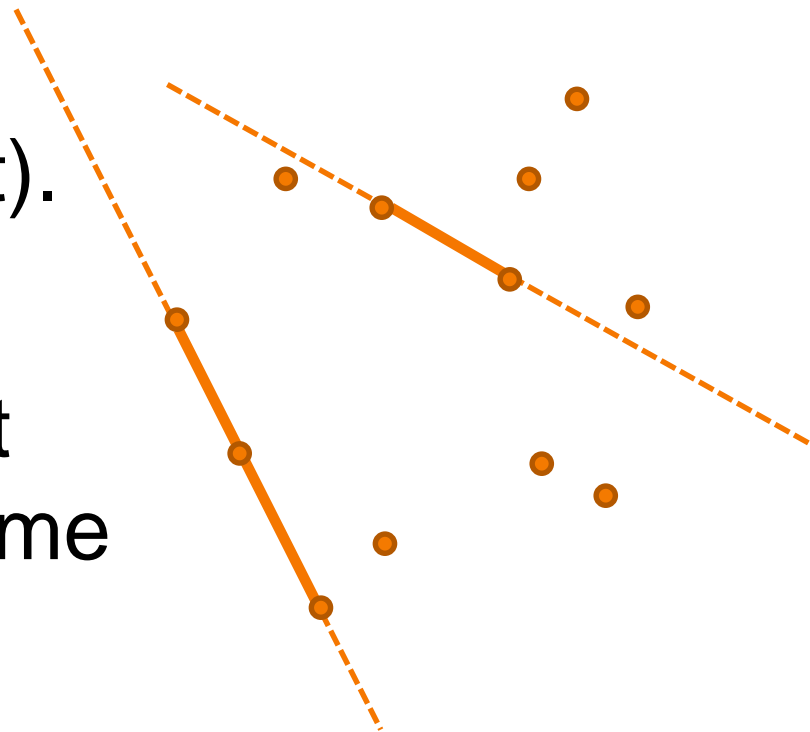
Convex Hull (2D)

Naïve Algorithm $O(n^3)$:

For each directed edge $e \in S \times S$, check if half-space to the right of e is empty of points (and there are no points on the line outside the segment).

Note:

The output is the set of (unordered) extreme points on the hull.





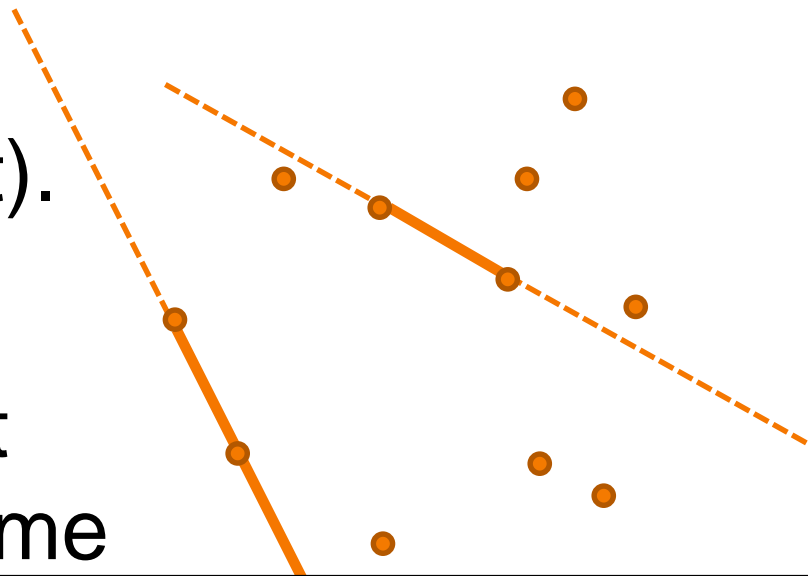
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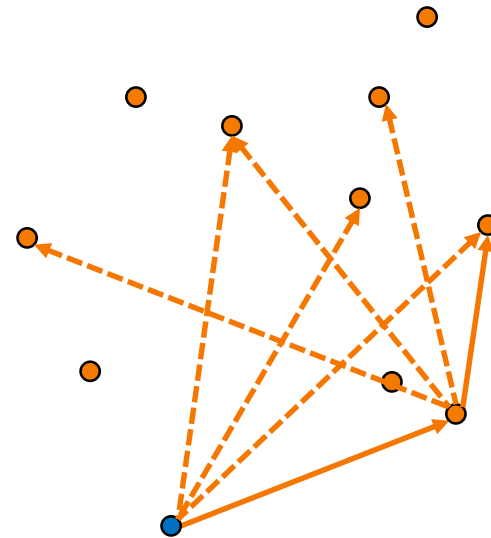
If we want the ordered points, we can stitch the edges together in $\leq O(n^2)$ time in a post-processing step.



Convex Hull (2D)

Naïve Algorithm++:

Grow the hull by starting at a hull vertex and searching for the next edge on the hull by trying all possible edges and testing if they are on the hull.





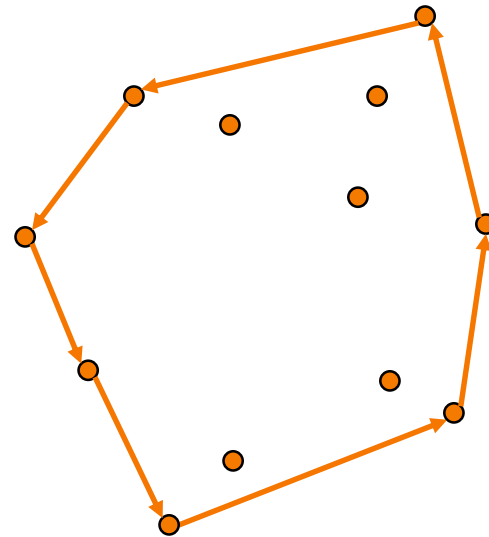
Convex Hull (2D)

Naïve Algorithm++ $O(n^2h)^*$:

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Note:

By explicitly forcing the output to be sorted, we end up with a faster algorithm.



* h is the number of points on the hull.



Convex Hull (2D)

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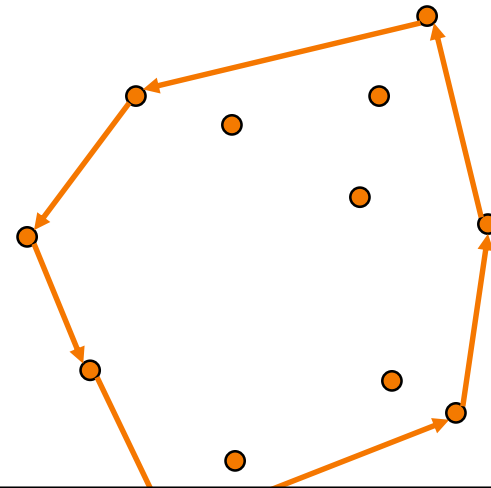
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This implementation is *output sensitive*.



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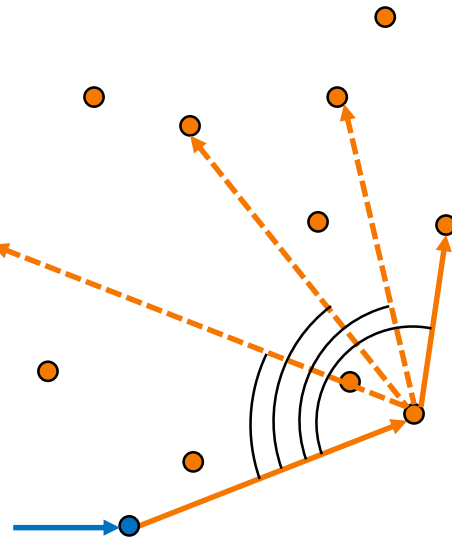
Convex Hull (2D)

Note:

The next edge on the hull is the one making the largest angle. (If two points make the same angle, ignore the closer one.)

Gift Wrapping:

Grow by finding the edge making the largest angle.





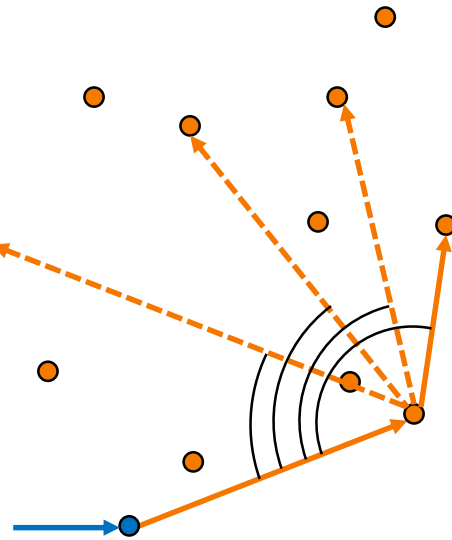
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Gift Wrapping $O(nh)$:

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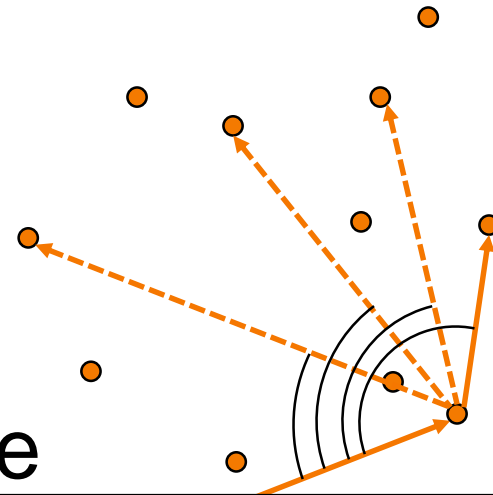




Convex Hull (2D)

Note:

The next edge on the hull is the one making the largest angle. (If two points make the same angle, ignore the closer one.)



Gift Wrapping $O(nh)$:

Grow by finding the edge

make A similar approach makes it possible to find a hull edge in linear time.



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Convex Hull (2D)

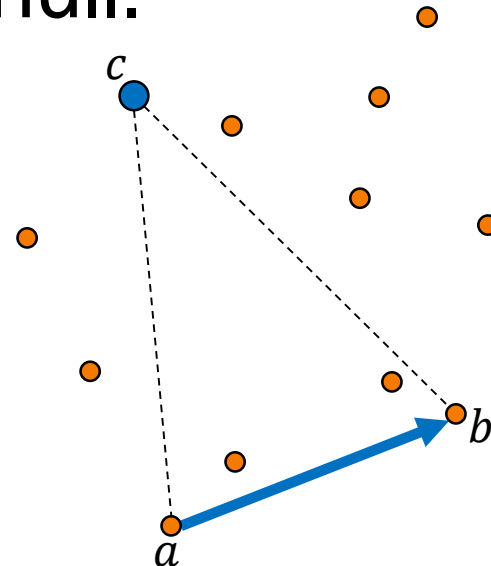
Observation:

Given a hull edge (a, b) , we can find the point c furthest from the edge in linear time.

1. The point c is on the hull.

2. The triangle Δabc partitions the input into three regions:

- I. Points inside Δabc .
- II. Points to the right of \overrightarrow{bc} .
- III. Points to the right of \overrightarrow{ca} .





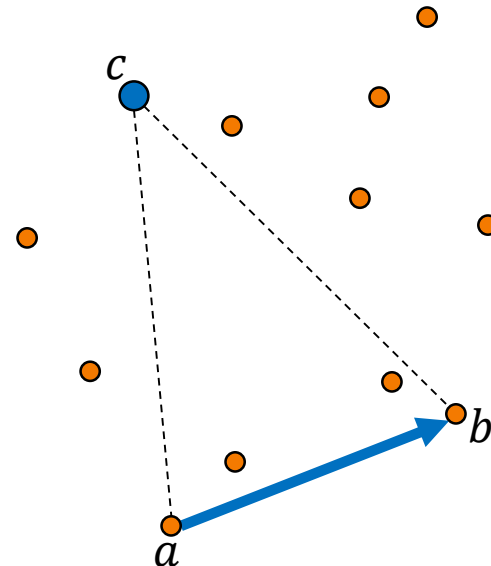
Convex Hull (2D)

Observation:

Given a hull edge (a, b) , we can find the point c furthest from the edge in linear time.

⇒ Divide-and-conquer:

- Discard points inside Δabc
- Separately compute the half-hulls to the right of \vec{bc} and the right of \vec{ca} .
- Merge the two hulls.





Convex Hull (2D)

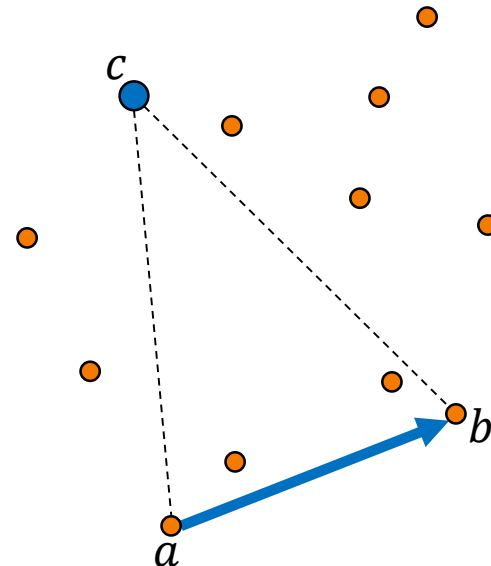
Observation:

Given a point c for

We don't require \overrightarrow{ab} to be a hull edge.
As long as it's a hull diagonal merging is easy.

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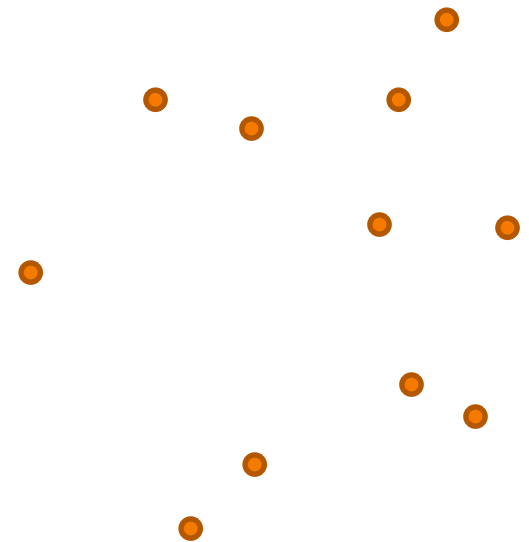




Convex Hull (2D)

QuickHull($S \subset \mathbb{R}^2$)

- $(a, b) \leftarrow \text{HorizontalExtrema}(S)$
- $A \leftarrow \text{RightOf}(S , \overrightarrow{ab})$
- $B \leftarrow \text{RightOf}(S , \overrightarrow{ba})$
- $Q_A \leftarrow \text{QuickHalfHull}(A , \overrightarrow{ab})$
- $Q_B \leftarrow \text{QuickHalfHull}(B , \overrightarrow{ba})$
- return $\{a\} \cup Q_A \cup \{b\} \cup Q_B$

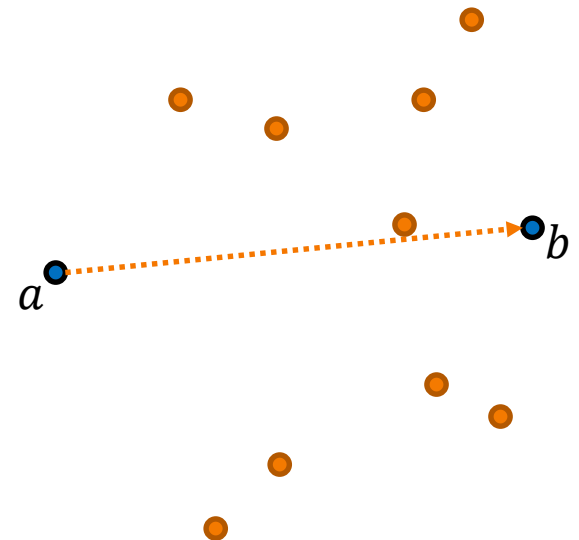




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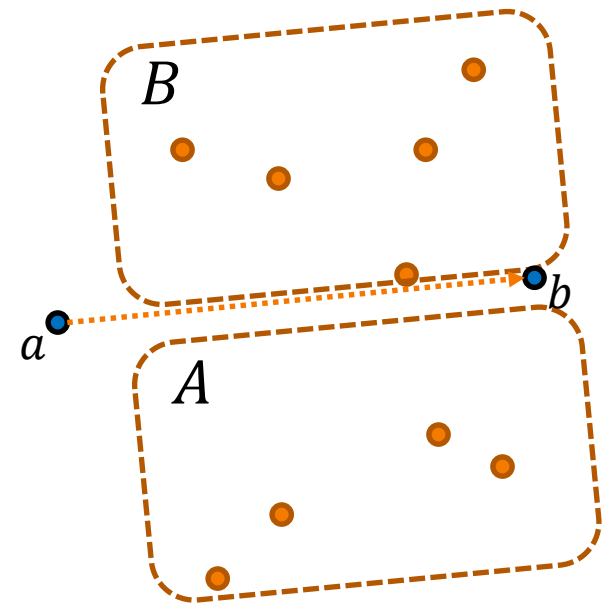




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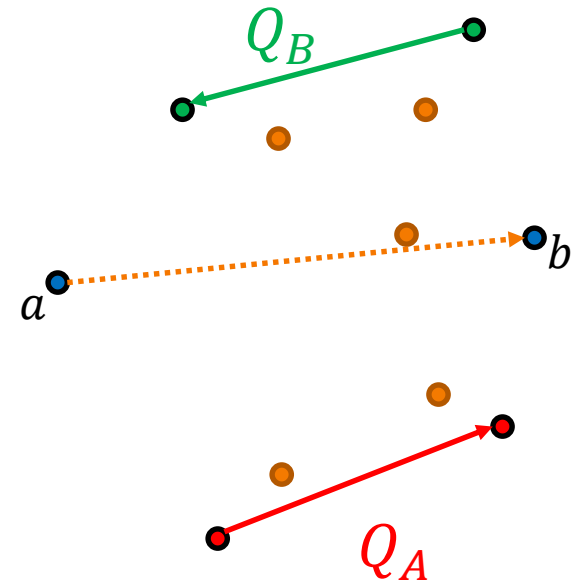




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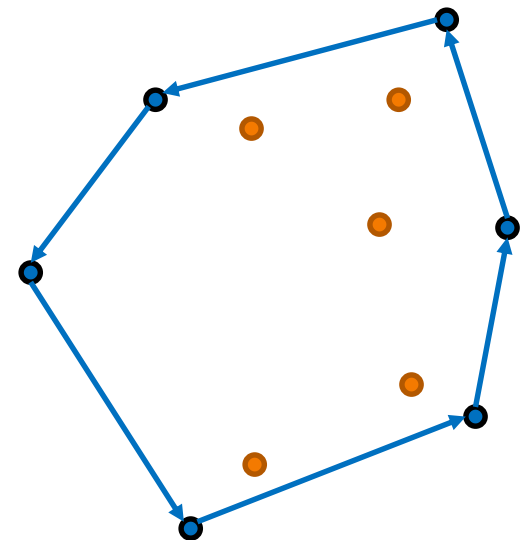




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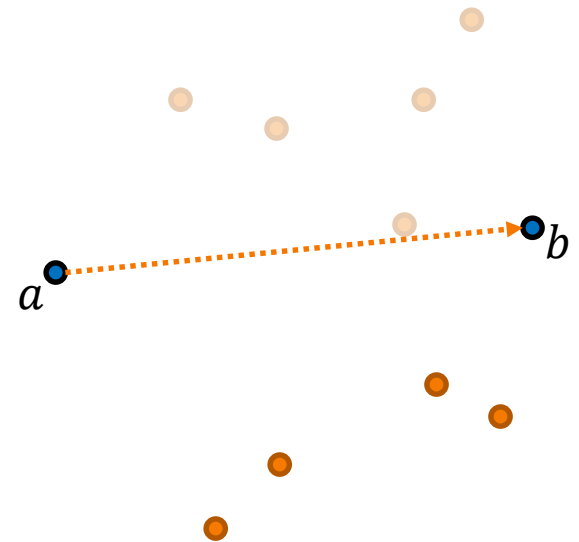


Convex Hull (2D)

(Recursion Level 0)

QuickHalfHull($S \subset \mathbb{R}^2$, $\overrightarrow{ab} \in S \times S$)

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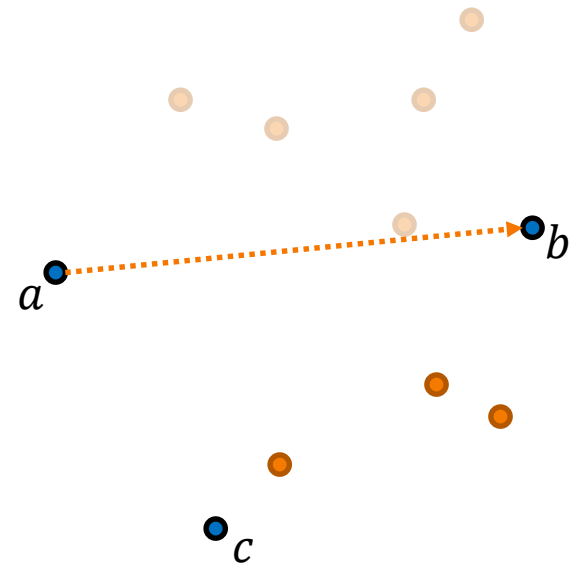


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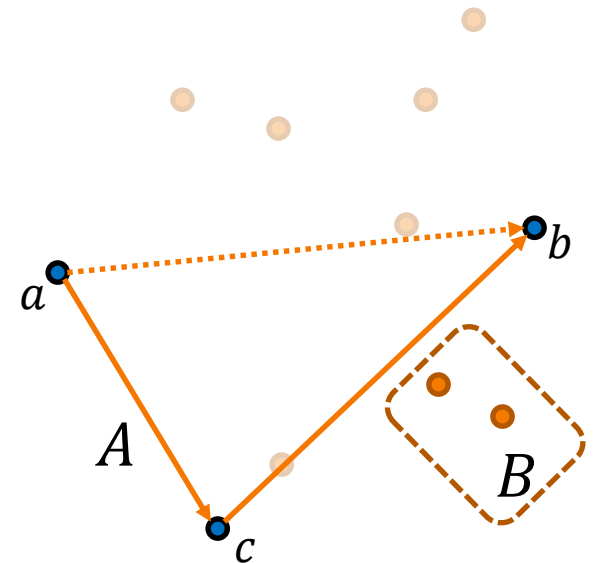


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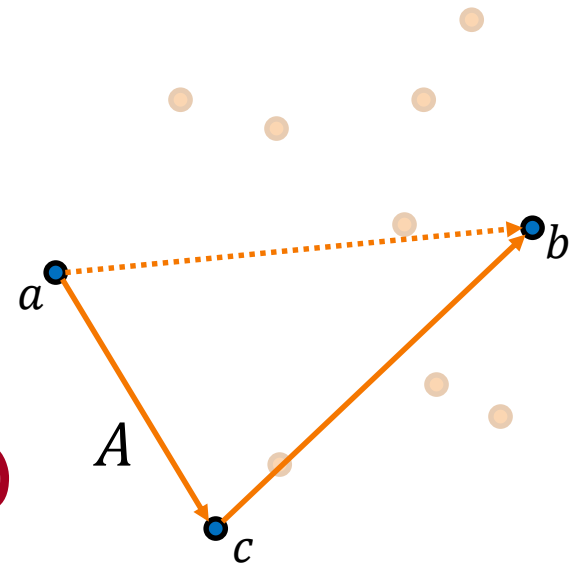


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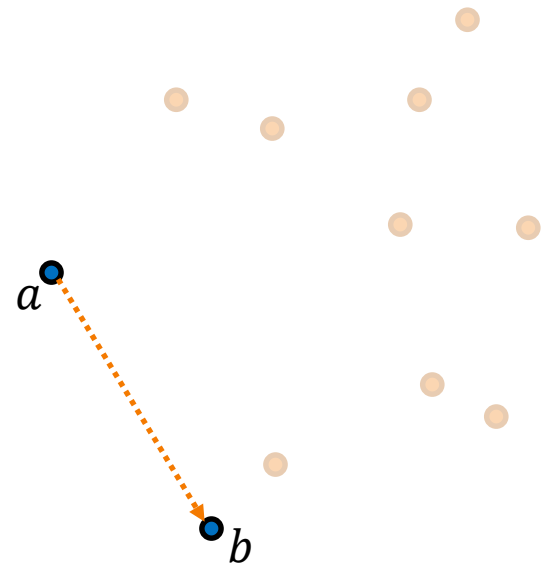


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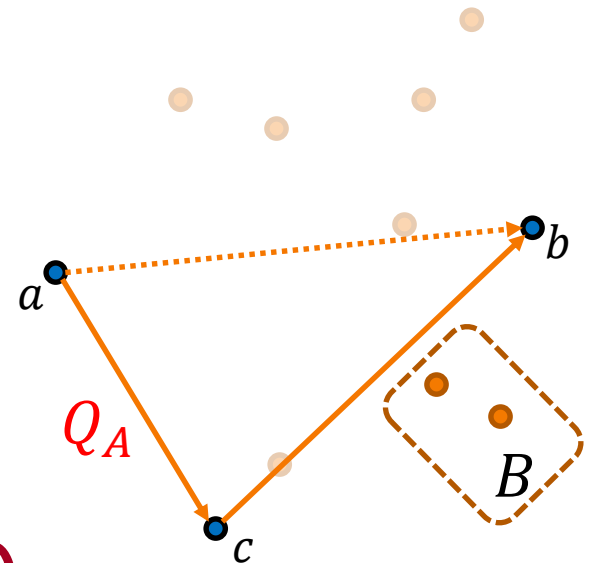


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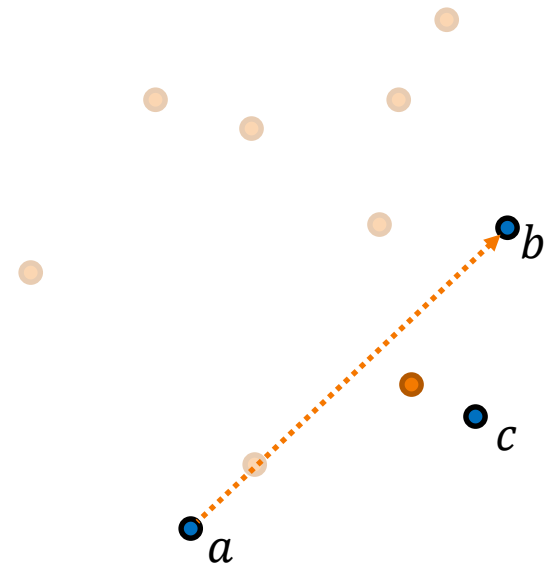


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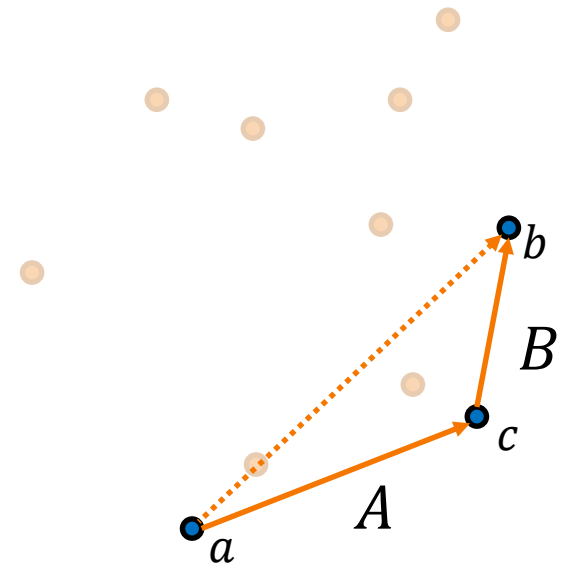


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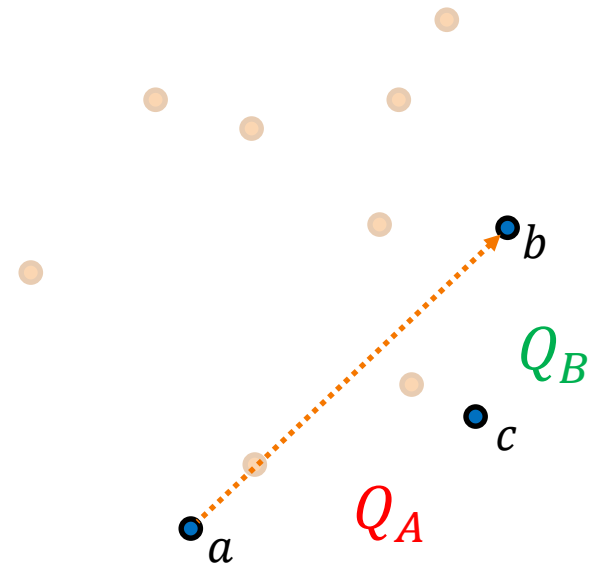


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 - » $c \leftarrow \text{Furthest}(S , \overrightarrow{ab})$
 - » $A \leftarrow \text{RightOf}(S , \overrightarrow{ac})$
 - » $B \leftarrow \text{RightOf}(S , \overrightarrow{cb})$
 - » $Q_A \leftarrow \text{QuickHalfHull}(A , \overrightarrow{ac})$
 - » $Q_B \leftarrow \text{QuickHalfHull}(B , \overrightarrow{cb})$
 - » return $Q_A \cup \{c\} \cup Q_B$



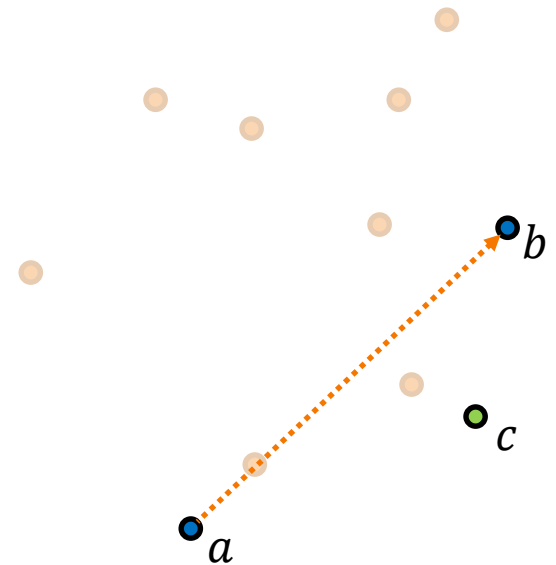


Convex Hull (2D)

(Recursion Level 1)

QuickHalfHull($S \subset \mathbb{R}^2$, $\overrightarrow{ab} \in S \times S$)

- if($S == \emptyset$) return \emptyset
- else
 - » $c \leftarrow \text{Furthest}(S , \overrightarrow{ab})$
 - » $A \leftarrow \text{RightOf}(S , \overrightarrow{ac})$
 - » $B \leftarrow \text{RightOf}(S , \overrightarrow{cb})$
 - » $Q_A \leftarrow \text{QuickHalfHull}(A , \overrightarrow{ac})$
 - » $Q_B \leftarrow \text{QuickHalfHull}(B , \overrightarrow{cb})$
 - » **return** $Q_A \cup \{c\} \cup Q_B$





Convex Hull (2D)

(Recursion Level 0)

$\text{QuickHalfHull}(S \subset \mathbb{R}^2, \overrightarrow{ab} \in S \times S)$

- if($S == \emptyset$) return \emptyset
- else

» $c \leftarrow \text{Furthest}(S, \overrightarrow{ab})$

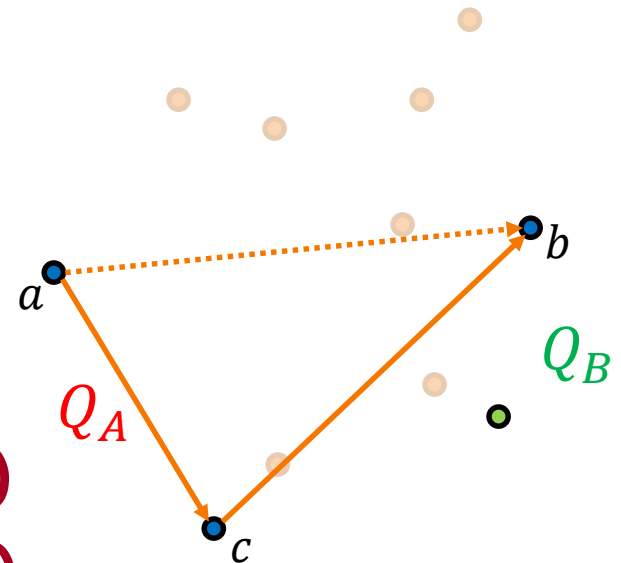
» $A \leftarrow \text{RightOf}(S, \overrightarrow{ac})$

» $B \leftarrow \text{RightOf}(S, \overrightarrow{cb})$

» $Q_A \leftarrow \text{QuickHalfHull}(A, \overrightarrow{ac})$

» $Q_B \leftarrow \text{QuickHalfHull}(B, \overrightarrow{cb})$

» return $Q_A \cup \{c\} \cup Q_B$



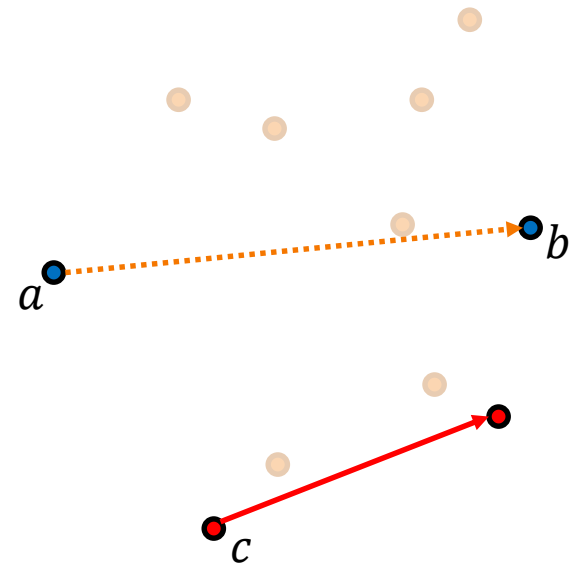


Convex Hull (2D)

(Recursion Level 0)

QuickHalfHull($S \subset \mathbb{R}^2$, $\overrightarrow{ab} \in S \times S$)

- if($S == \emptyset$) return \emptyset
- else
 - » $c \leftarrow \text{Furthest}(S , \overrightarrow{ab})$
 - » $A \leftarrow \text{RightOf}(S , \overrightarrow{ac})$
 - » $B \leftarrow \text{RightOf}(S , \overrightarrow{cb})$
 - » $Q_A \leftarrow \text{QuickHalfHull}(A , \overrightarrow{ac})$
 - » $Q_B \leftarrow \text{QuickHalfHull}(B , \overrightarrow{cb})$
 - » **return** $Q_A \cup \{c\} \cup Q_B$



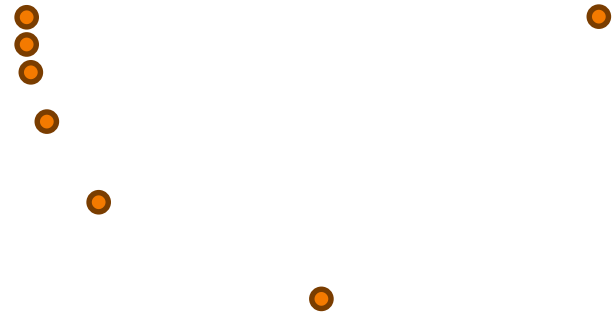


Convex Hull (2D)

QuickHull Complexity:

Like QuickSort:

- In the worst case, the complexity can be $O(n^2)$.



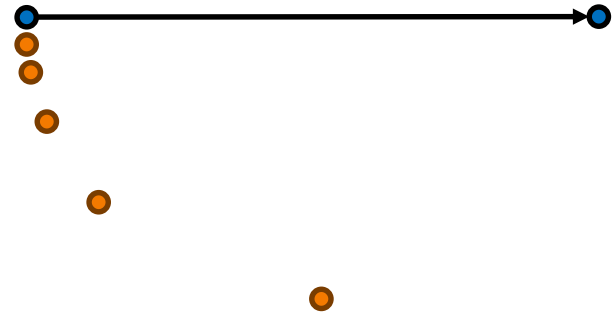


Convex Hull (2D)

QuickHull Complexity:

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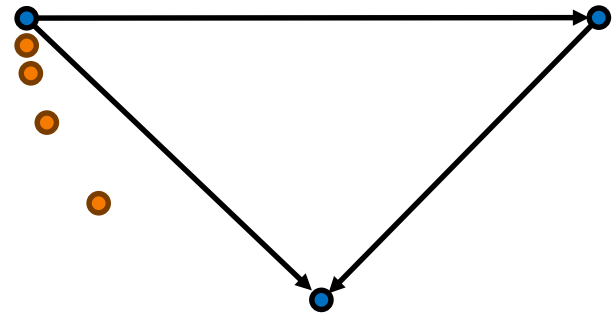


Convex Hull (2D)

QuickHull Complexity:

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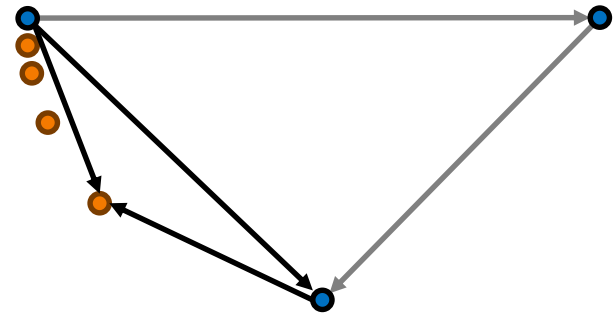


Convex Hull (2D)

QuickHull Complexity:

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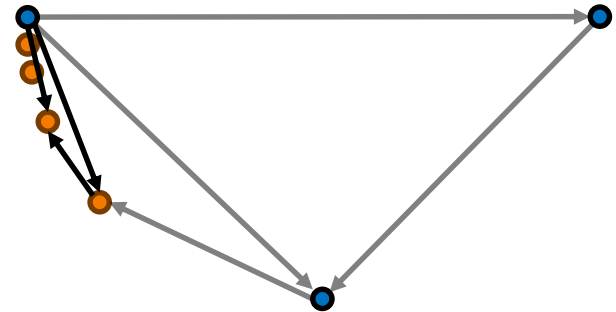


Convex Hull (2D)

QuickHull Complexity:

Like QuickSort:

- In the worst case, the complexity can be $O(n^2)$.





Convex Hull (2D)

QuickHull Complexity:

Like QuickSort:

- In the worst case, the complexity can be $O(n^2)$.
- In practice it is $O(n \log n)$.
- The implementation is output sensitive.

Does it extend to higher dimensions?



Outline

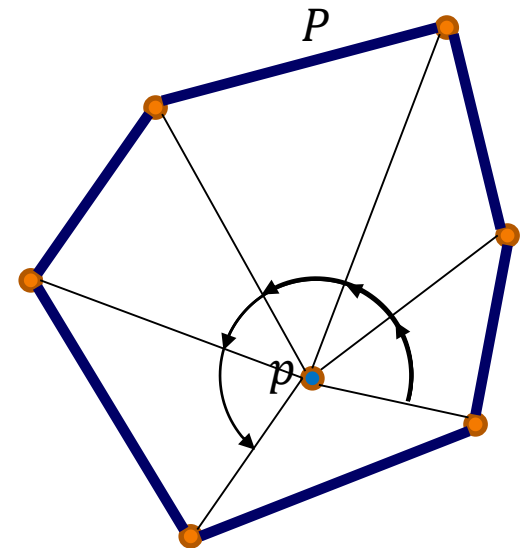
- Convex Hulls
- Algorithms
 - Naïve Implementation(s)
 - Gift Wrapping
 - Quick Hull
 - **Graham's Algorithm**
 - Lower bound complexity



Convex Hull (2D)

Graham's Observation:

If $P \subset \mathbb{R}^2$ is a convex polygon and $p \in P$ is a point in the interior of the polygon, then the angle of the line segments between p and the ordered vertices of P is monotonic.





Convex Hull (2D)

Graham's Observation:

WLOG assume p and v_i lie on a vertical line with p below v_i .

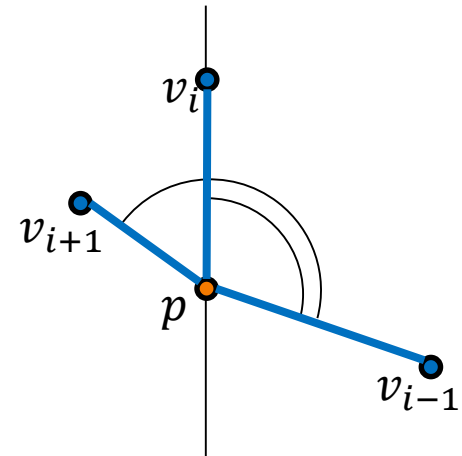
Since the polygon is convex, p is to the left of $\overrightarrow{v_i v_{i+1}}$.

$\Rightarrow v_{i+1}$ is to the left of the vertical.

Since the polygon is convex, p is to the left of $\overrightarrow{v_{i-1} v_i}$.

$\Rightarrow v_{i-1}$ is to the right of the vertical.

\Rightarrow The angles $\angle p v_{i-1}$, $\angle p v_i$, $\angle p v_{i+1}$ increase monotonically.





Convex Hull (2D)

GrahamScan($S \subset \mathbb{R}^2$)

- $p \leftarrow \text{PointInHull}(S)$
- $H \leftarrow \text{SortByAngle}(S , p)$
- while(RemoveReflexVertex(H)) { }
- return H



Convex Hull (2D)

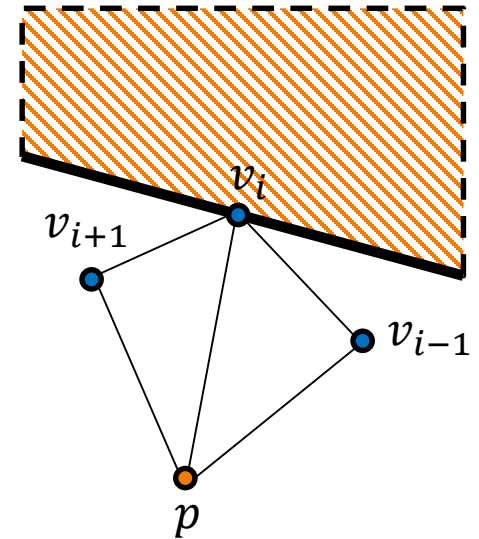
GrahamScan($S \subset \mathbb{R}^2$)

- $p \leftarrow \text{PointInHull}(S)$
- $H \leftarrow \text{SortByAngle}(S , p)$
- while(RemoveReflexVertex(H)) { }
- return H

Note:

At every iteration, the vertices of H are sorted by angle relative to p .

⇒ Hull vertices can never be removed because the angle between the previous and next vertex is always convex.





Convex Hull (2D)

GrahamScan($S \subset \mathbb{R}^2$)

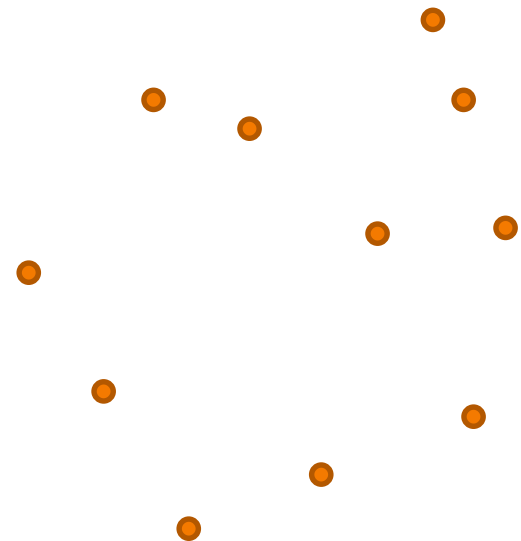
- $p \leftarrow \text{PointInHull}(S)$
- $H \leftarrow \text{SortByAngle}(S , p)$
- while(RemoveReflexVertex(H)) { }
- return H

Correctness:

- The output polygon has only convex vertices.
⇒ It's convex.
⇒ $H \subset \mathcal{H}(S)$.
- All hull vertices are in H .
⇒ $\mathcal{H}(S) \subset H$.

Convex Hull (2D)

GrahamScan($S \subset \mathbb{R}^2$)

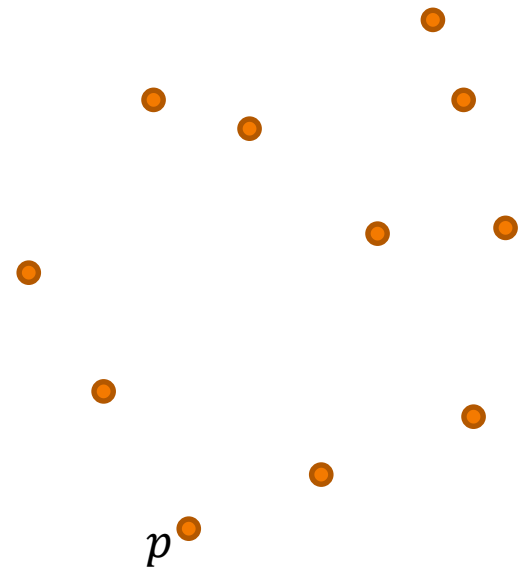




Convex Hull (2D)

GrahamScan($S \subset \mathbb{R}^2$)

◦ $p \leftarrow \text{BottommostRightmost}(S)$

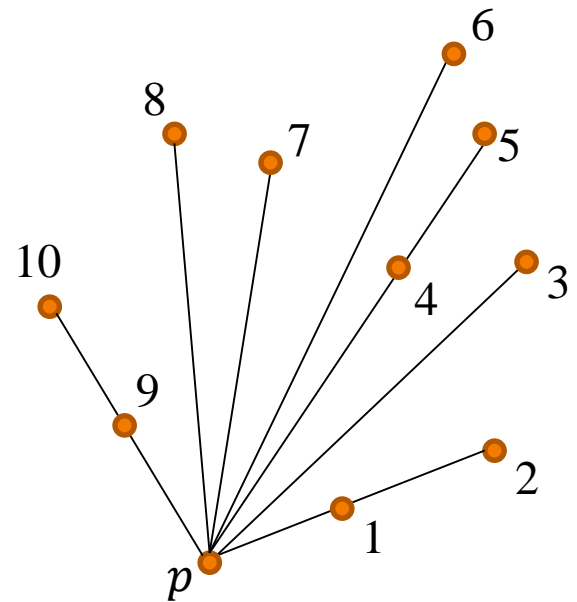




Convex Hull (2D)

GrahamScan($S \subset \mathbb{R}^2$)

- $p \leftarrow \text{BottommostRightmost}(S)$
- $\tilde{S} \leftarrow \text{SortByAngleAndLength}(p , S - \{p\})$

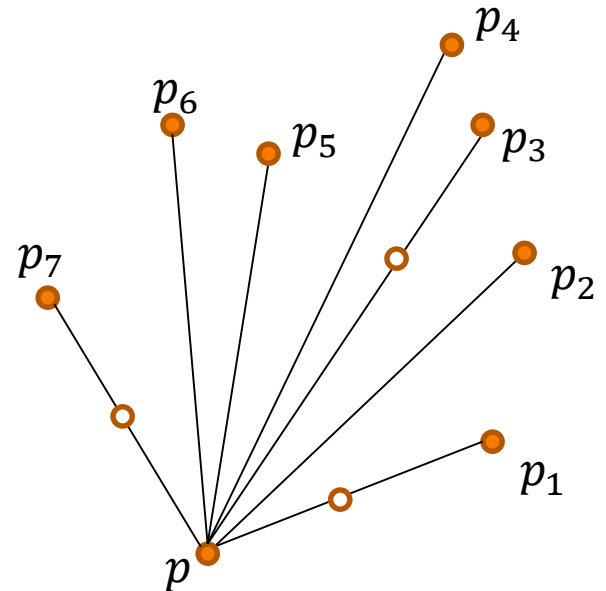




Convex Hull (2D)

GrahamScan($S \subset \mathbb{R}^2$)

- $p \leftarrow \text{BottommostRightmost}(S)$
- $\tilde{S} \leftarrow \text{SortByAngleAndLength}(p , S - \{p\})$
- if($\text{angle}(p_i) == \text{angle}(p_{i+1})$): $\tilde{S} \leftarrow \tilde{S} - \{p_i\}$





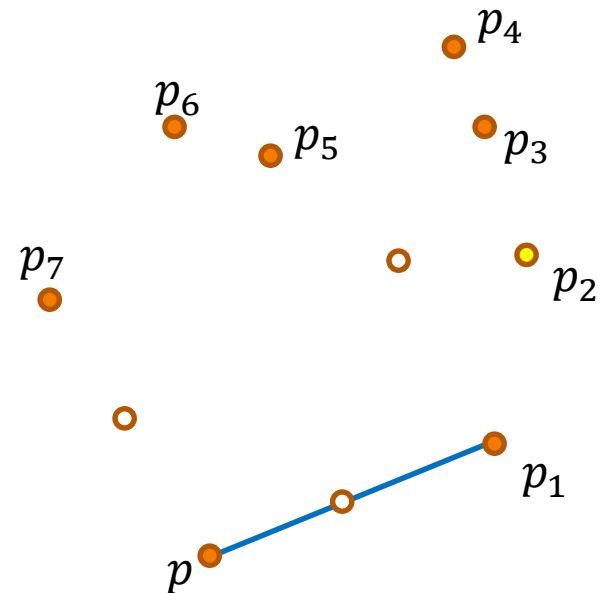
Convex Hull (2D)

GrahamScan($S \subset \mathbb{R}^2$)

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- $\tilde{S} \leftarrow \text{SortByAngleAndLength}(p , S - \{p\})$
- if($\text{angle}(p_i) == \text{angle}(p_{i+1})$): $\tilde{S} \leftarrow \tilde{S} - \{p_i\}$
- $i \leftarrow 2$
- $Q \leftarrow \{p, p_1\}$

Note:

Since p is bottom-(right)-most, vertices are sorted by angle in $(0, \pi]$, and non-extreme points are removed, $\overline{pp_1}$ is on the hull.

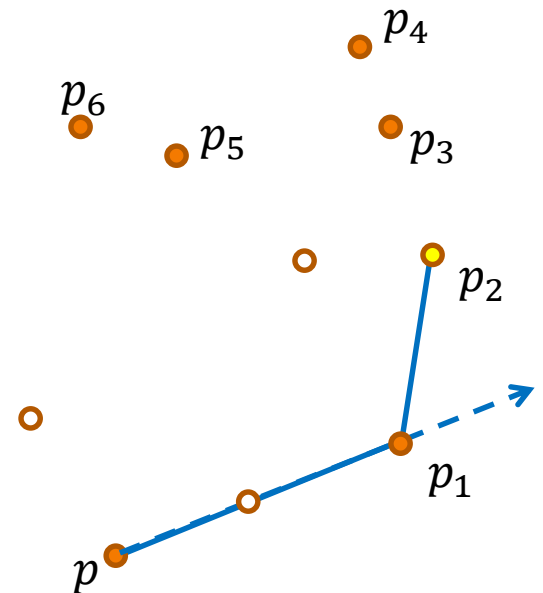




Convex Hull (2D)

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- $i \leftarrow 2$
- $Q \leftarrow \{p, p_1\}$
- while($i < |\tilde{S}|$)
 - » if($\text{Left}(p_i , \text{LastEdge}(Q))$)
 - Push(p_i , Q)
 - $i \leftarrow i + 1$
 - » else
 - Pop(Q)

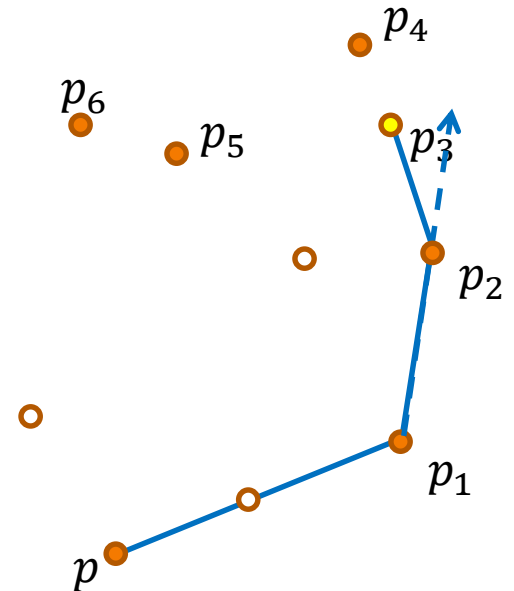




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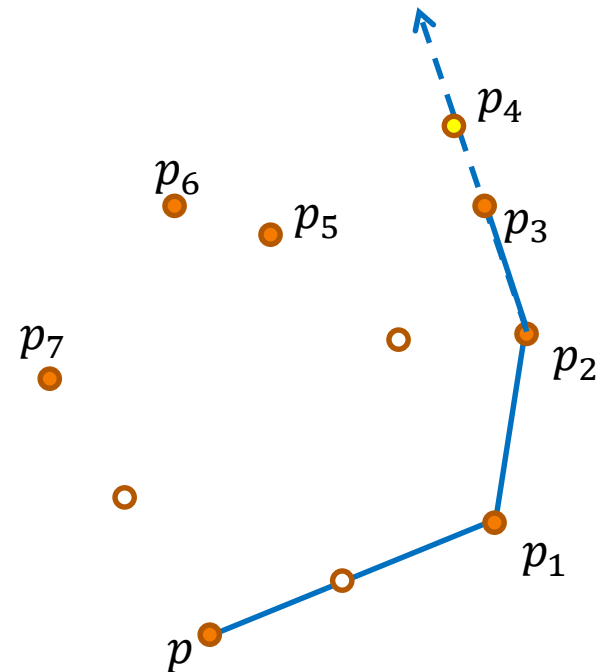




Convex Hull (2D)

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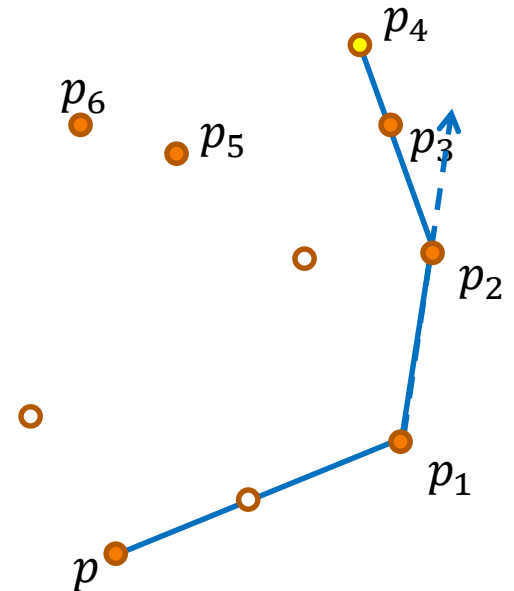




Convex Hull (2D)

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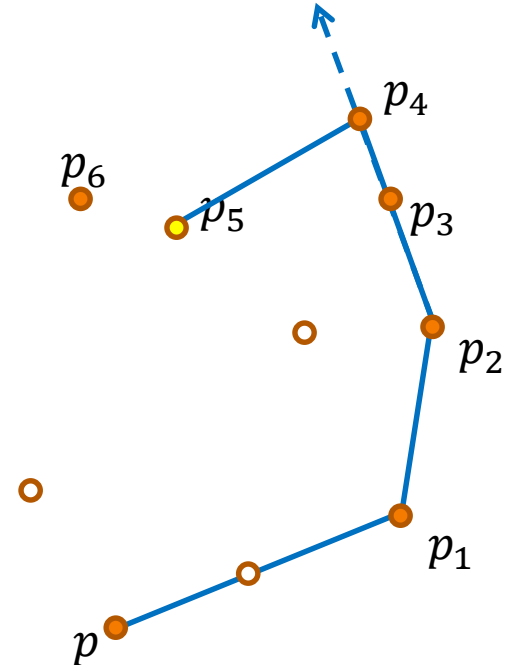




Convex Hull (2D)

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 - Push(p_i , Q)
 - $i \leftarrow i + 1$
 - » else
 - Pop(Q)

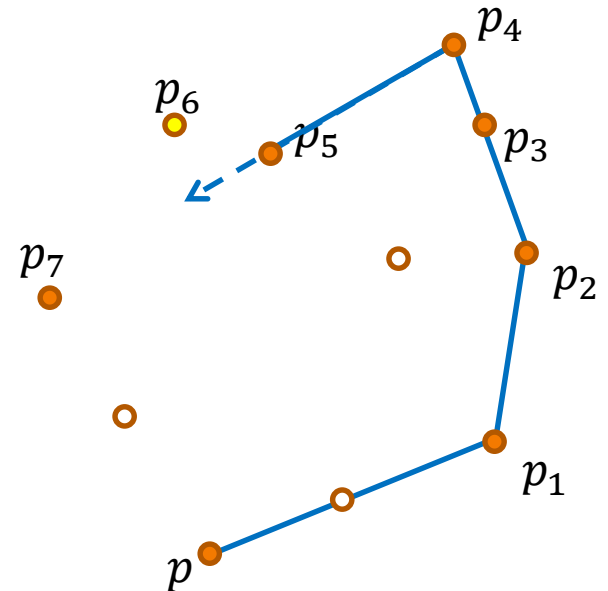




Convex Hull (2D)

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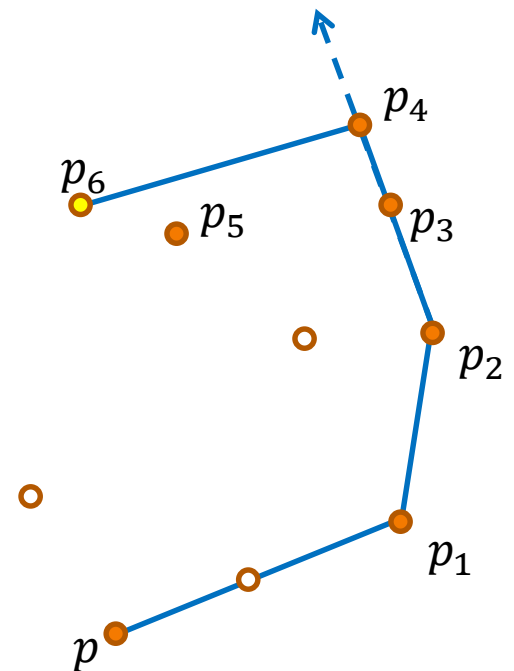




Convex Hull (2D)

GrahamScan($S \subset \mathbb{R}^2$)

- $p \leftarrow \text{BottommostRightmost}(S)$
- $\tilde{S} \leftarrow \text{SortByAngleAndLength}(p , S - \{p\})$
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- $i \leftarrow 2$
- $Q \leftarrow \{p, p_1\}$
- while($i < |\tilde{S}|$)
 - » if($\text{Left}(p_i , \text{LastEdge}(Q))$)
 - Push(p_i , Q)
 - $i \leftarrow i + 1$
 - » else
 - Pop(Q)

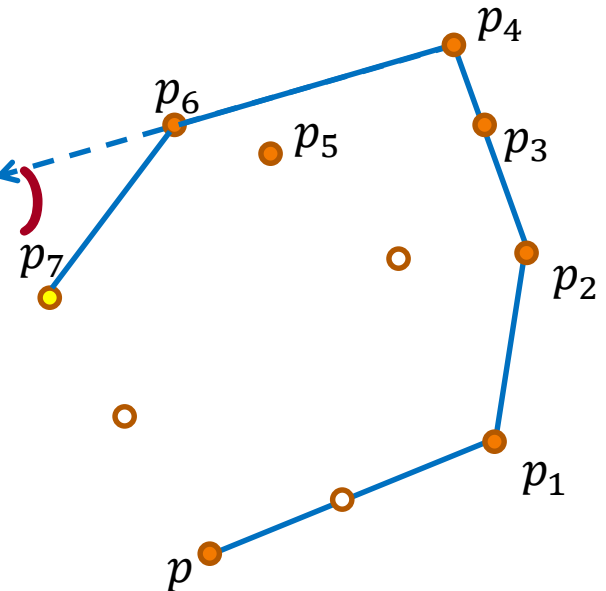




Convex Hull (2D)

GrahamScan($S \subset \mathbb{R}^2$)

- $p \leftarrow \text{BottommostRightmost}(S)$
- $\tilde{S} \leftarrow \text{SortByAngleAndLength}(p , S - \{p\})$
- if($\text{angle}(p_i) == \text{angle}(p_{i+1})$): $\tilde{S} \leftarrow \tilde{S} - \{p_i\}$
- $i \leftarrow 2$
- $Q \leftarrow \{p, p_1\}$
- while($i < |\tilde{S}|$)
 - » if(**Left(p_i , LastEdge(Q))**)
 - **Push(p_i , Q)**
 - $i \leftarrow i + 1$
 - » else
 - **Pop(Q)**

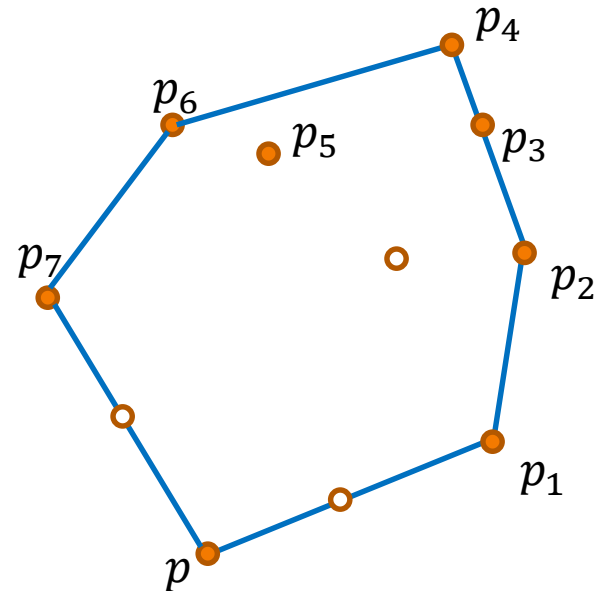




Convex Hull (2D)

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- $i \leftarrow 2$
- $Q \leftarrow \{p, p_1\}$
- while($i < |\tilde{S}|$)
 - » if($\text{Left}(p_i , \text{LastEdge}(Q))$)
 - Push(p_i , Q)
 - $i \leftarrow i + 1$
 - » else
 - Pop(Q)





Convex Hull (2D)

GrahamScan($S \subset \mathbb{R}^2$)

- $p \leftarrow \text{BottommostRightmost}(S)$ $O(n)$
 - $\tilde{S} \leftarrow \text{SortByAngleAndLength}(p , S - \{p\})$ $O(n \log n)$
 - if($\text{angle}(p_i) == \text{angle}(p_{i+1})$): $\tilde{S} \leftarrow \tilde{S} - \{p_i\}$ $O(n)$
 - $i \leftarrow 2$
 - $Q \leftarrow \{p, p_1\}$
 - while($i < |\tilde{S}|$)
 - » if($\text{Left}(p_i , \text{LastEdge}(Q))$)
 - $\text{Push}(p_i , Q)$
 - $i \leftarrow i + 1$
 - » else
 - $\text{Pop}(Q)$
- $O(n)$



Outline

- Convex Hulls
- Algorithms
 - Naïve Implementation(s)
 - Gift Wrapping
 - Quick Hull
 - Graham's Algorithm
 - Lower bound complexity



Lower Bound Complexity

Recall:

Sorting n numbers has lower bound complexity $O(n \log n)$.

Approach:

We will show that computing the 2D hull has the same complexity by reducing sorting to the problem of computing the convex hull.

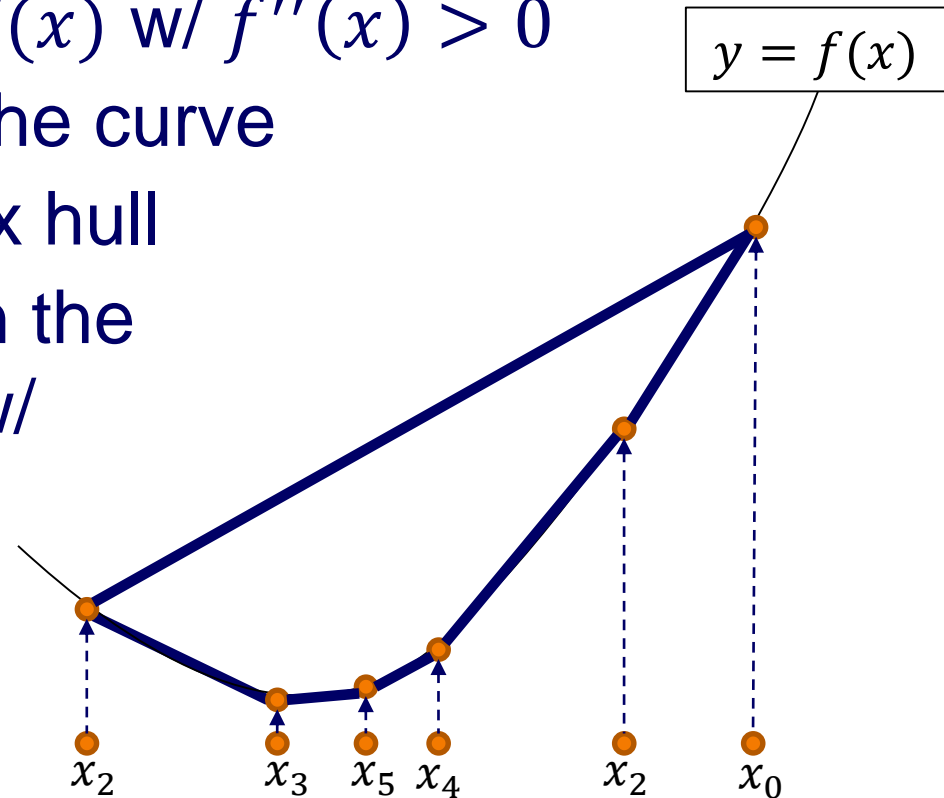


Lower Bound Complexity

Sorting → Convex Hull Reduction (Shamos):

Given a set of points $\{x_i\} \subset \mathbb{R}$:

- Choose a function $f(x)$ w/ $f''(x) > 0$
- Lift the points onto the curve
- Compute the convex hull
- Return the points on the lower hull, starting w/ the left-most.





Lower Bound Complexity

Sorting → Convex Hull Reduction (Shamos):

Given a set of points $\{x_i\} \subset \mathbb{R}^2$:

- The reduction assumes that the hull vertices are output in order.
- Compute the convex hull
- Return the points on the lower hull, starting w/ the left-most.

