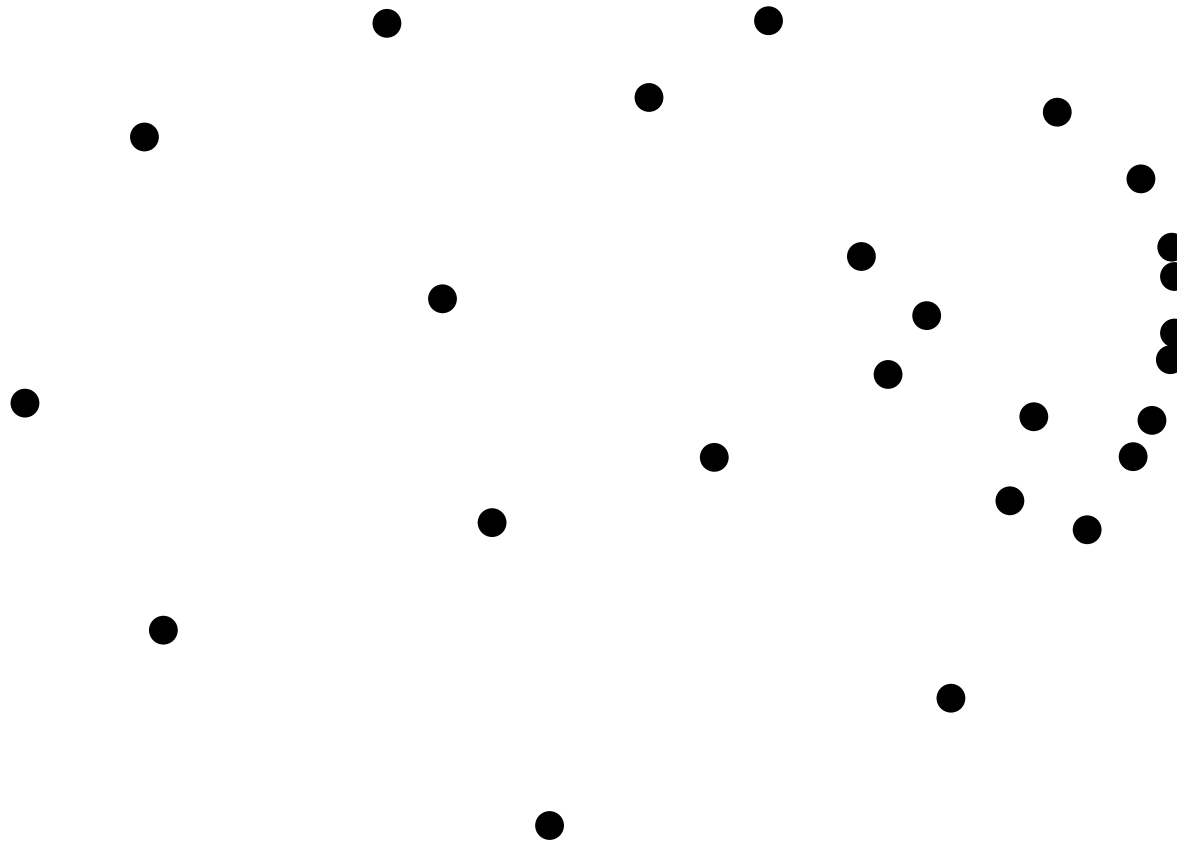
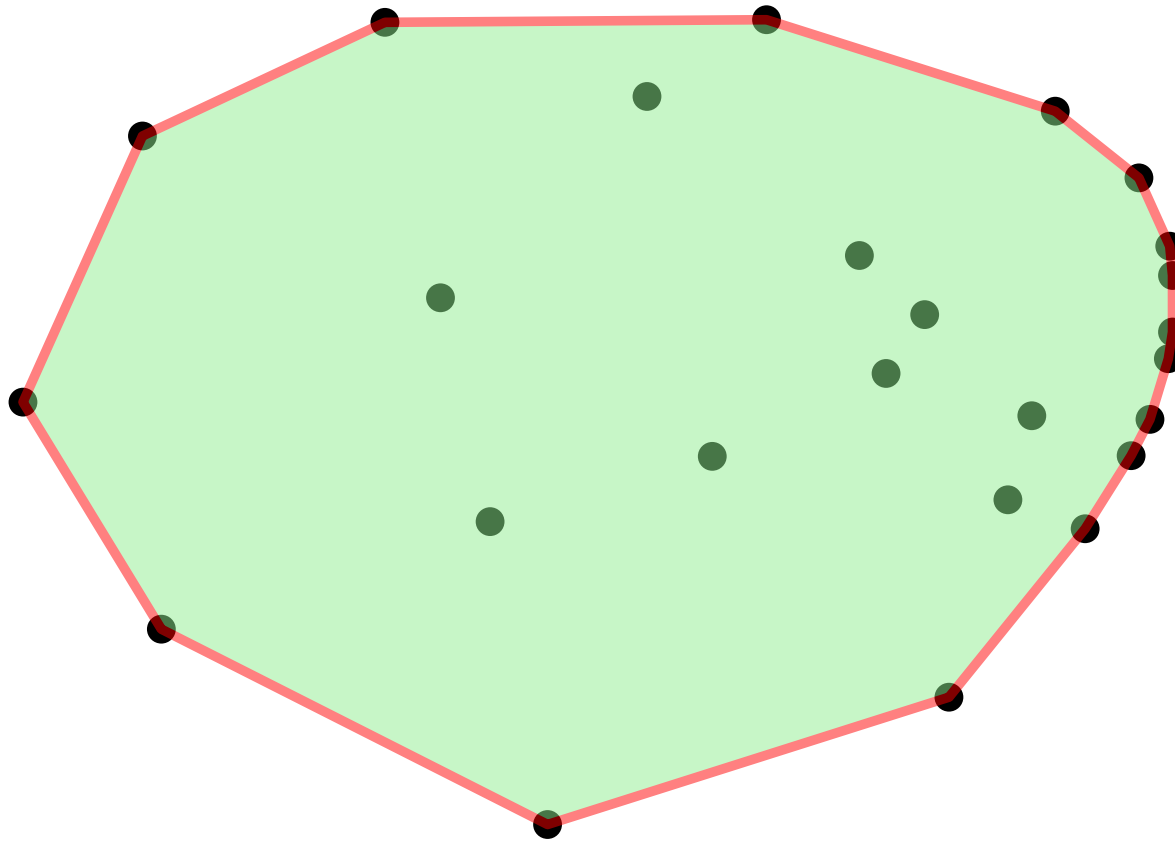


Convex hull

Convex hull



Convex hull



Convex hull

- Definition, extremal point
- Jarvis algorithm
- Orientation predicate
- Buggy degenerate example
- Real RAM model and general position hypothesis
- Graham algorithm
- Lower bound
- Other results
- Higher dimensions

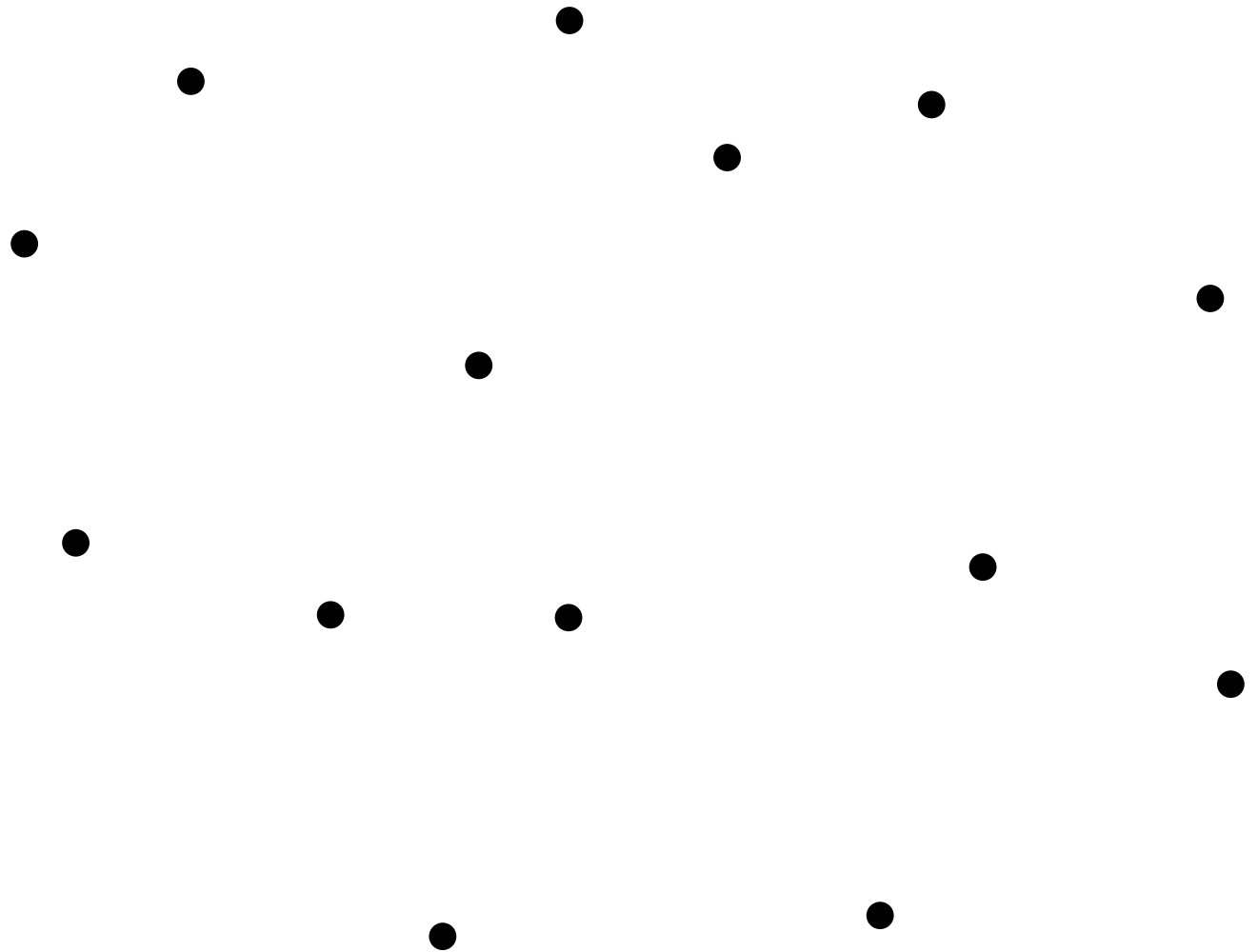
Convex hull

Definition, extremal point

Convex hull

Definition, extremal point

Set of points

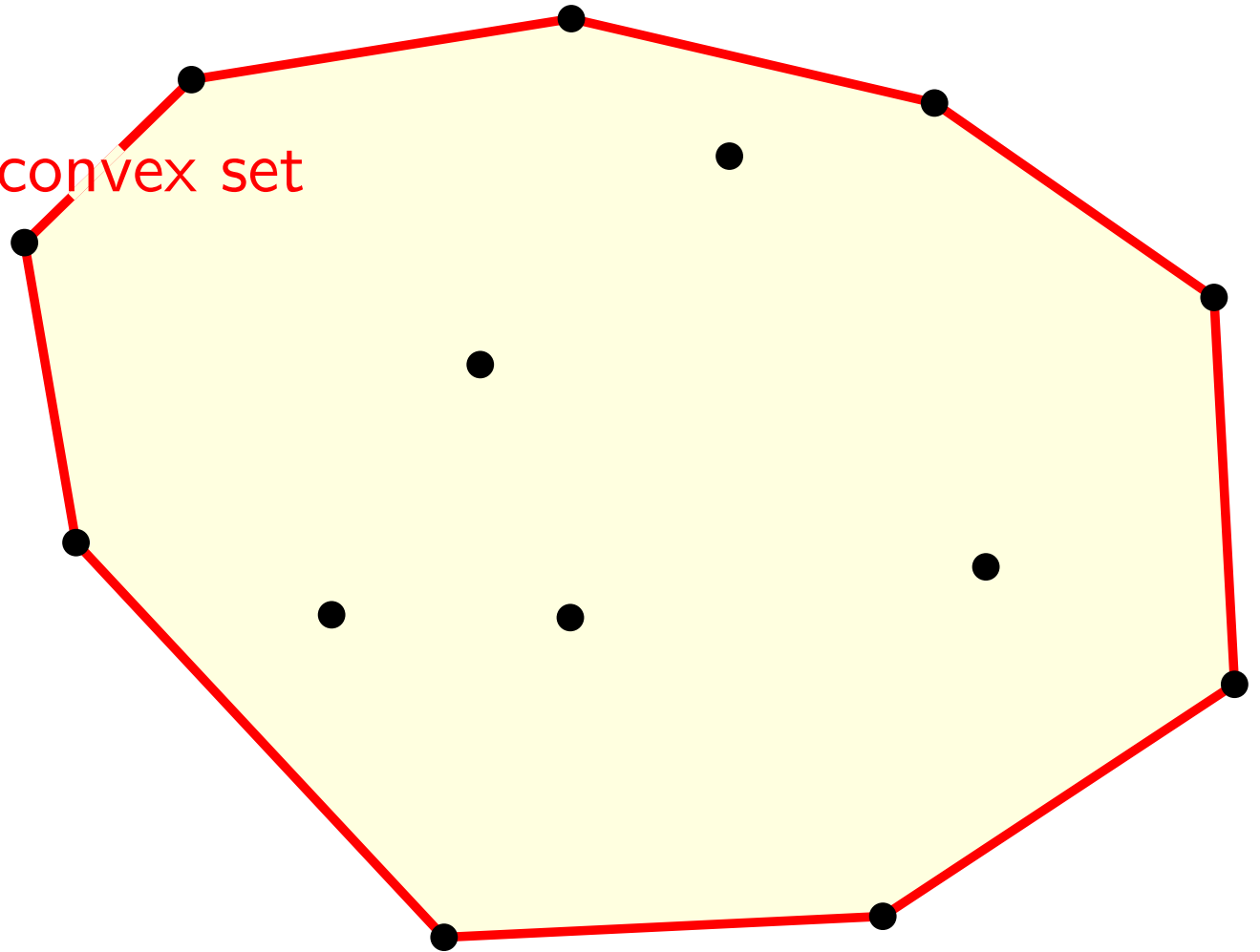


Convex hull

Definition, extremal point

Set of points

Smallest enclosing convex set



Convex hull

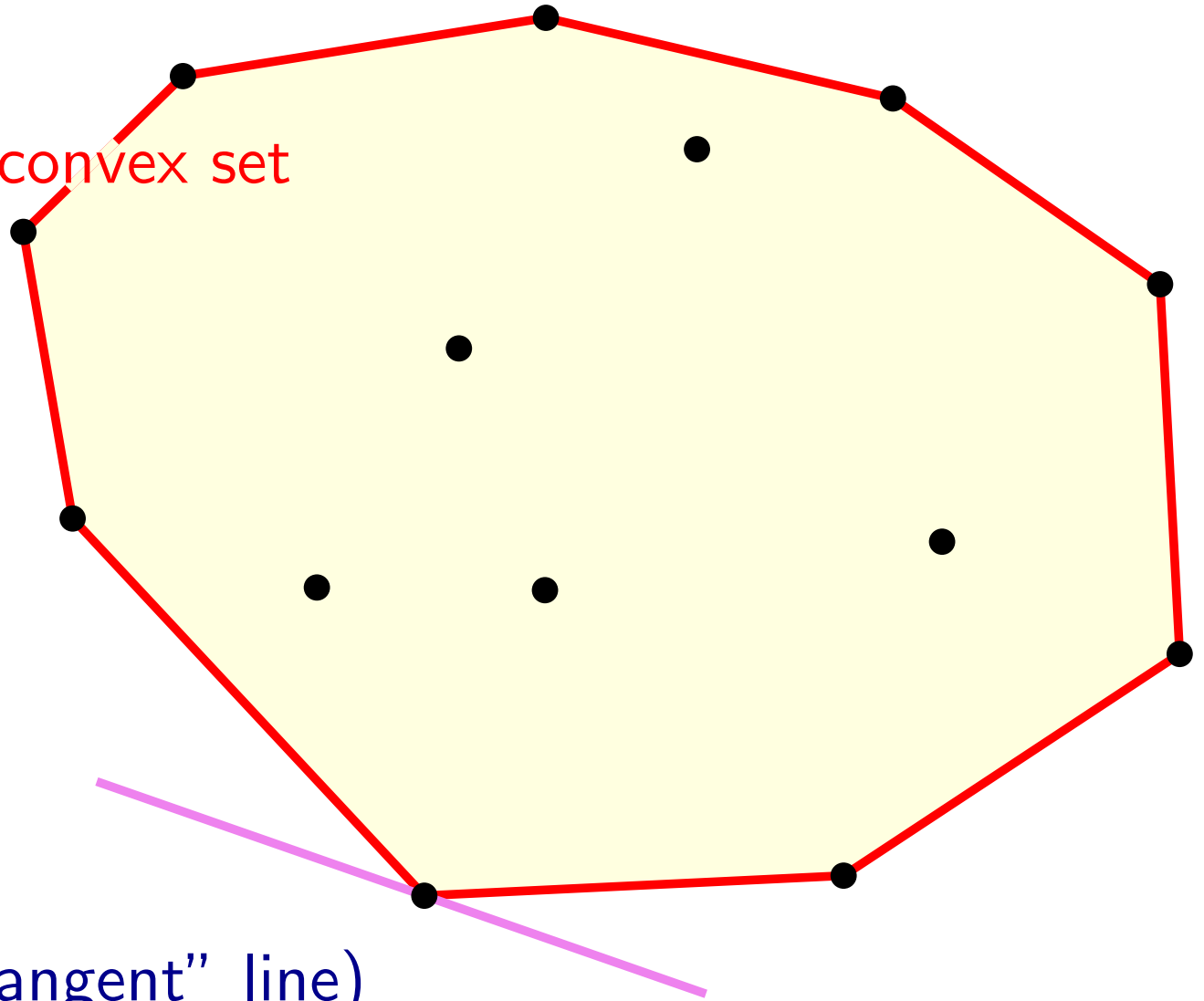
Definition, extremal point

Set of points

Smallest enclosing convex set

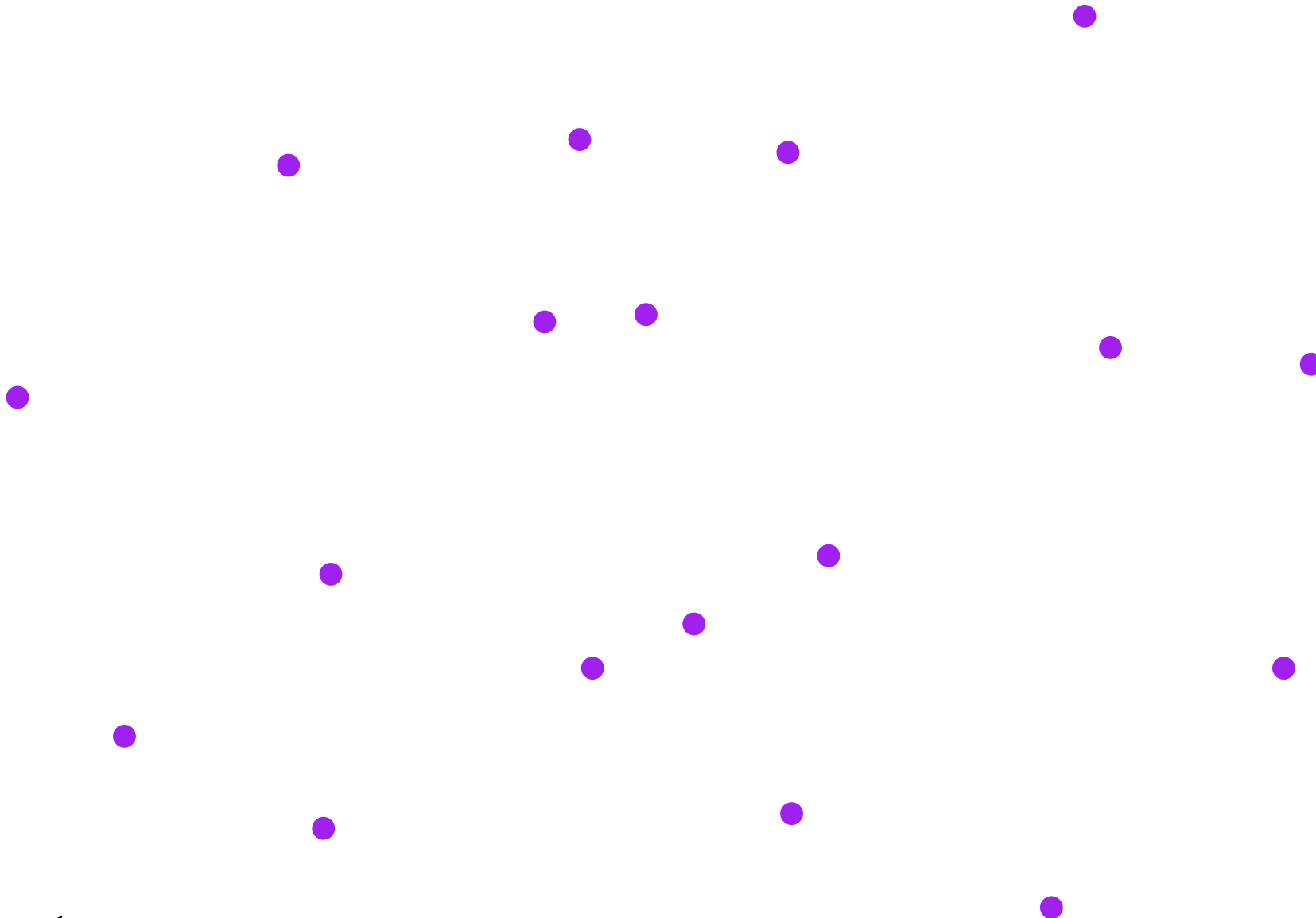
Extremal point

Supporting line ("tangent" line)



Convex hull

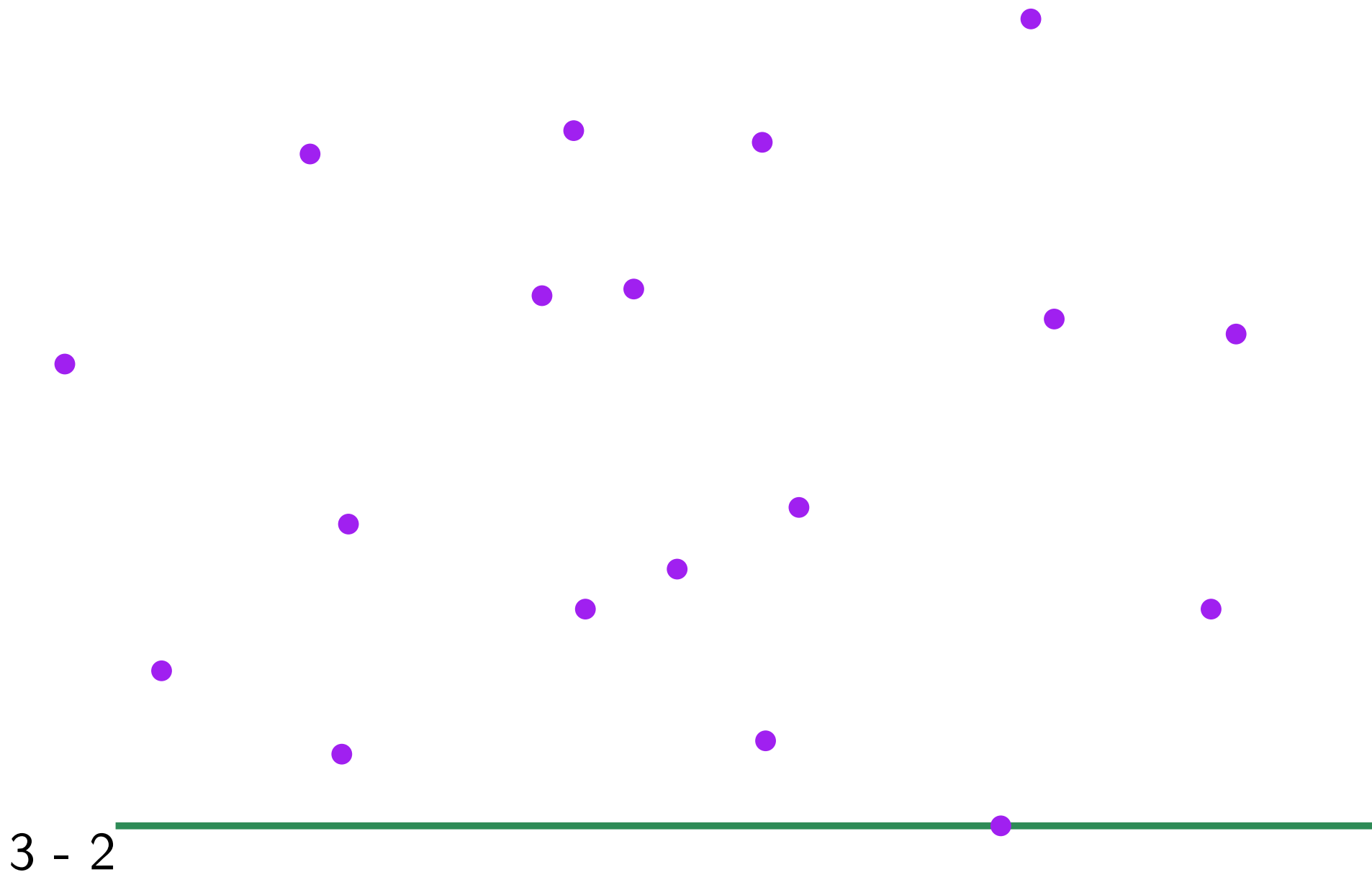
Jarvis algorithm



Convex hull

lowest point is extremal

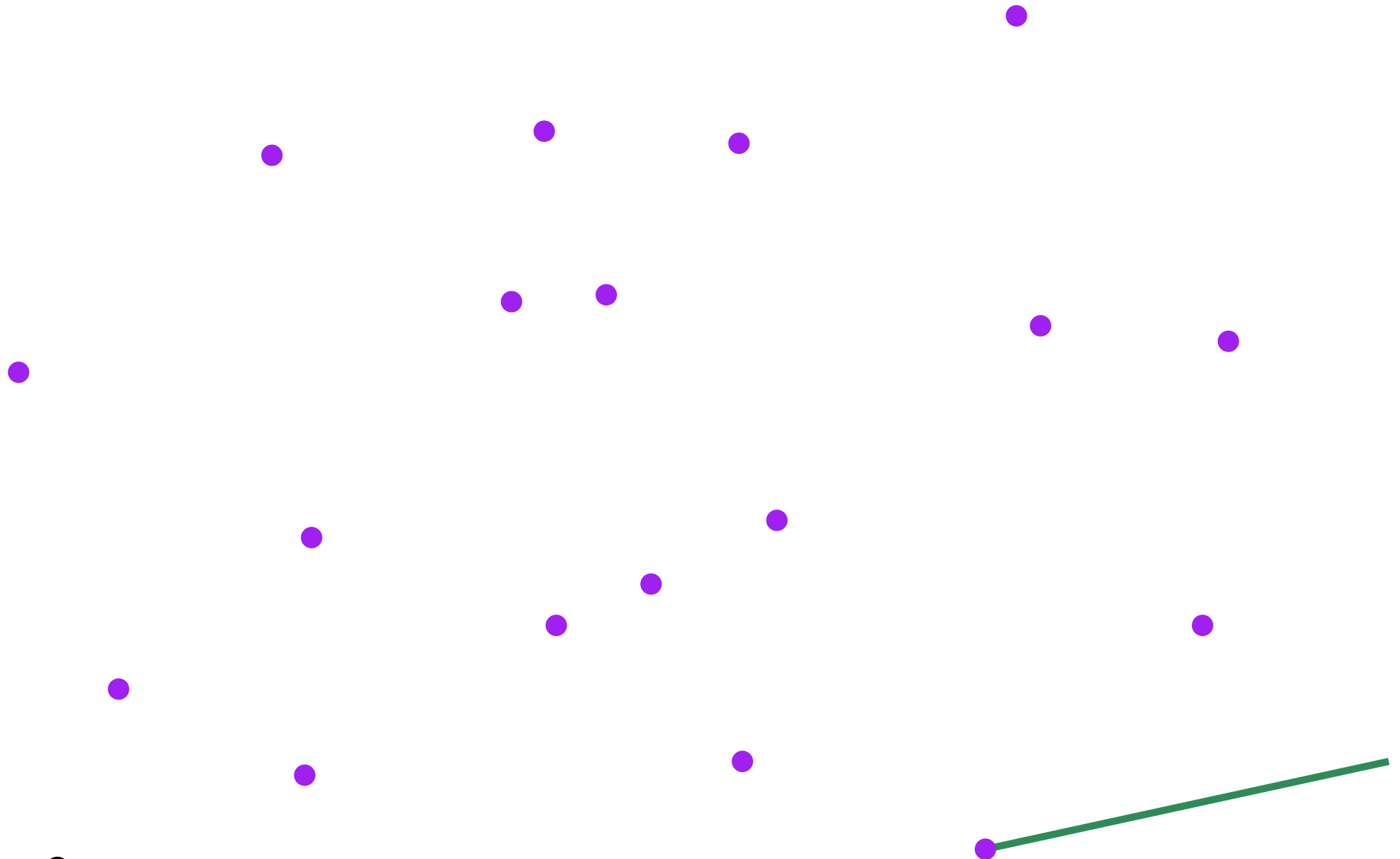
Jarvis algorithm



Convex hull

rotate

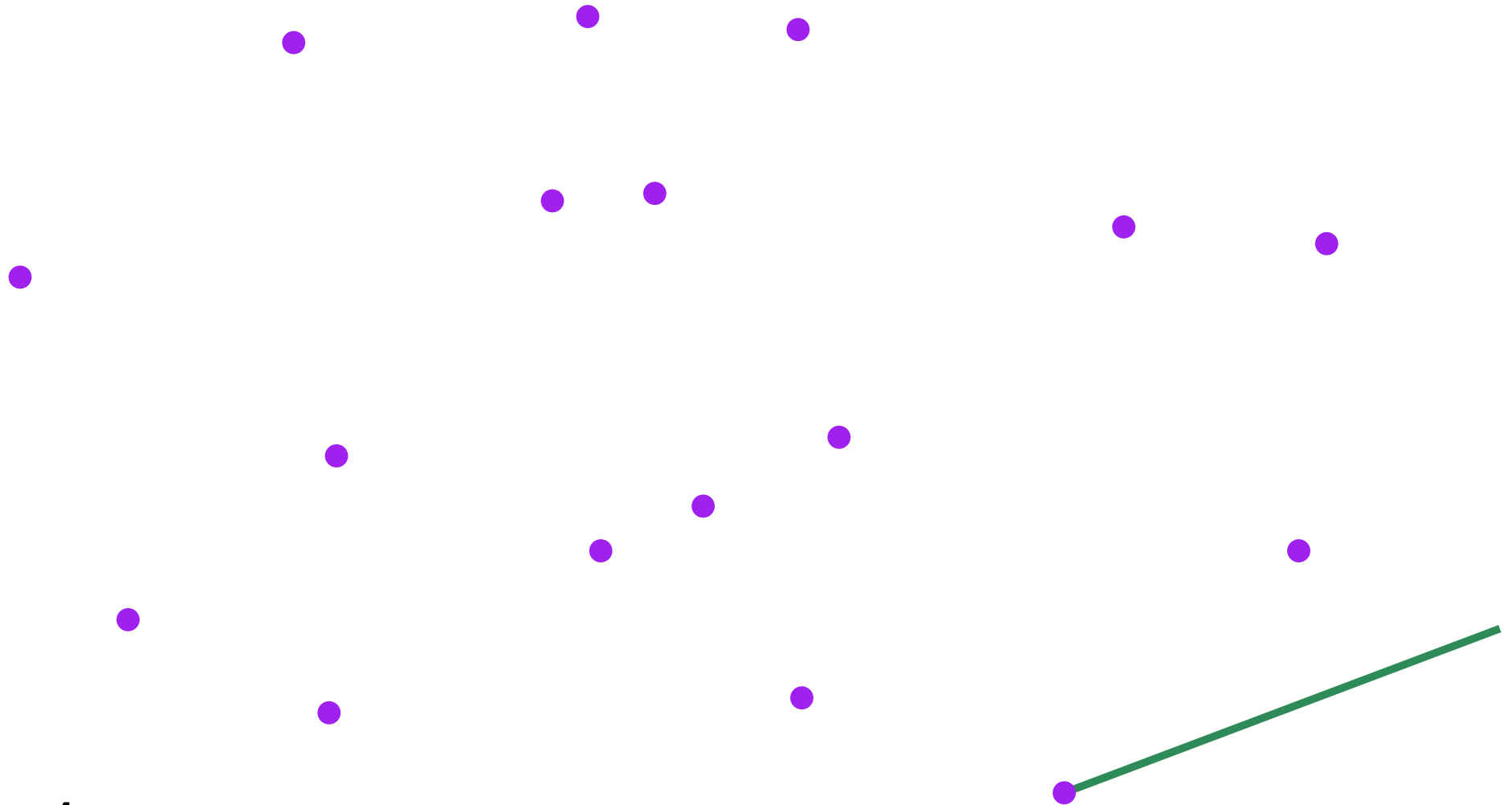
Jarvis algorithm



Convex hull

rotate

Jarvis algorithm

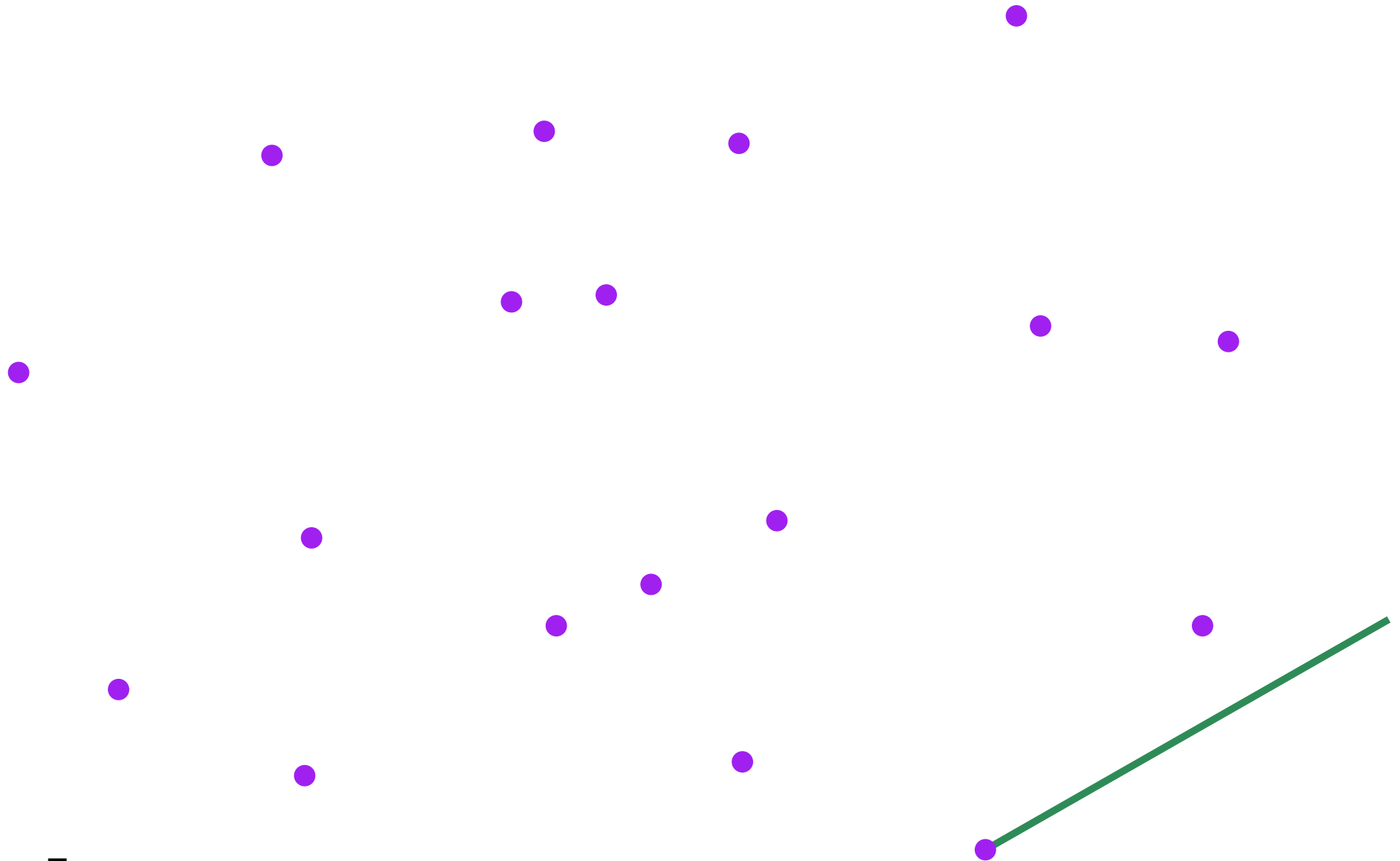


3 - 4

Convex hull

rotate

Jarvis algorithm

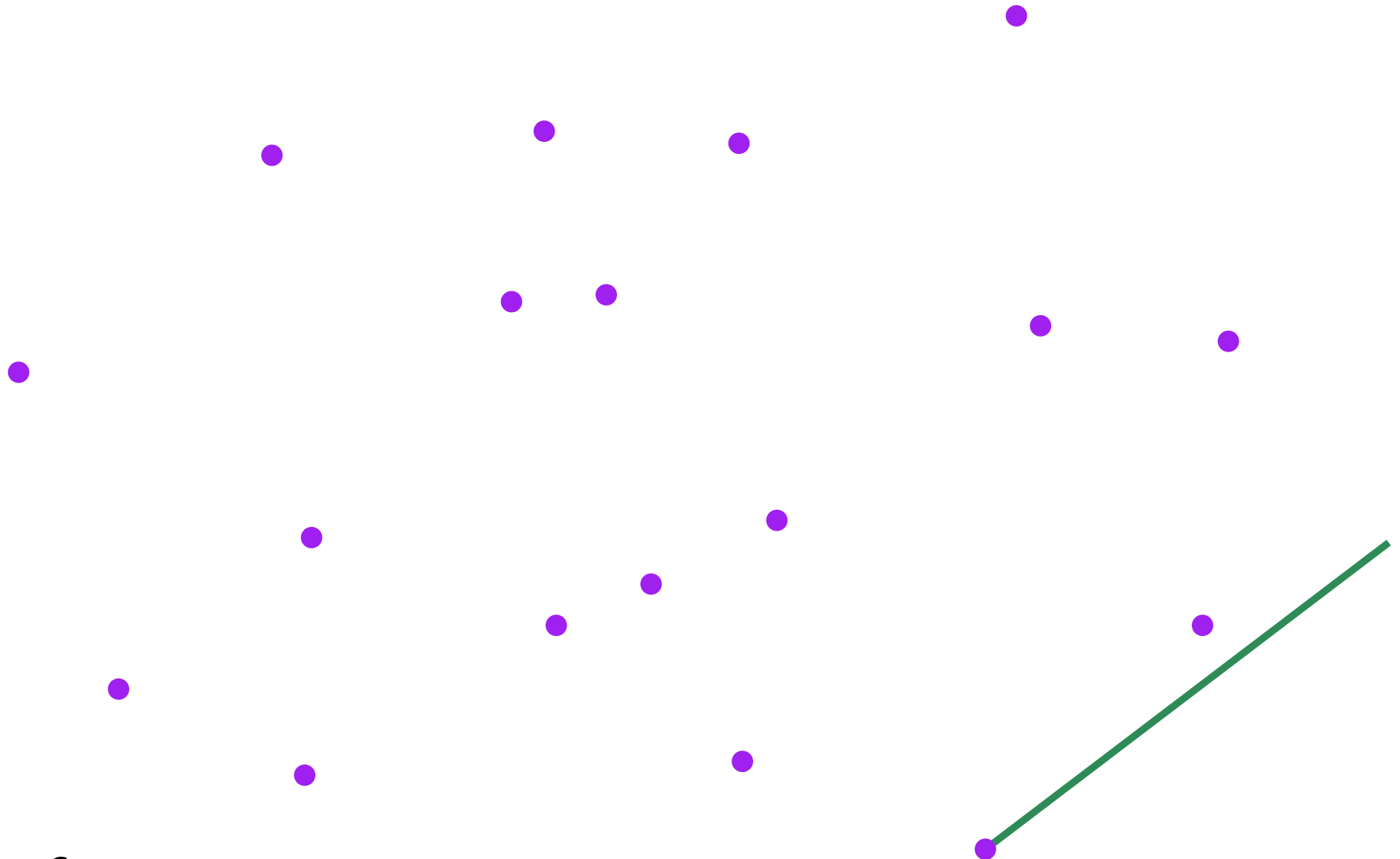


3 - 5

Convex hull

rotate

Jarvis algorithm

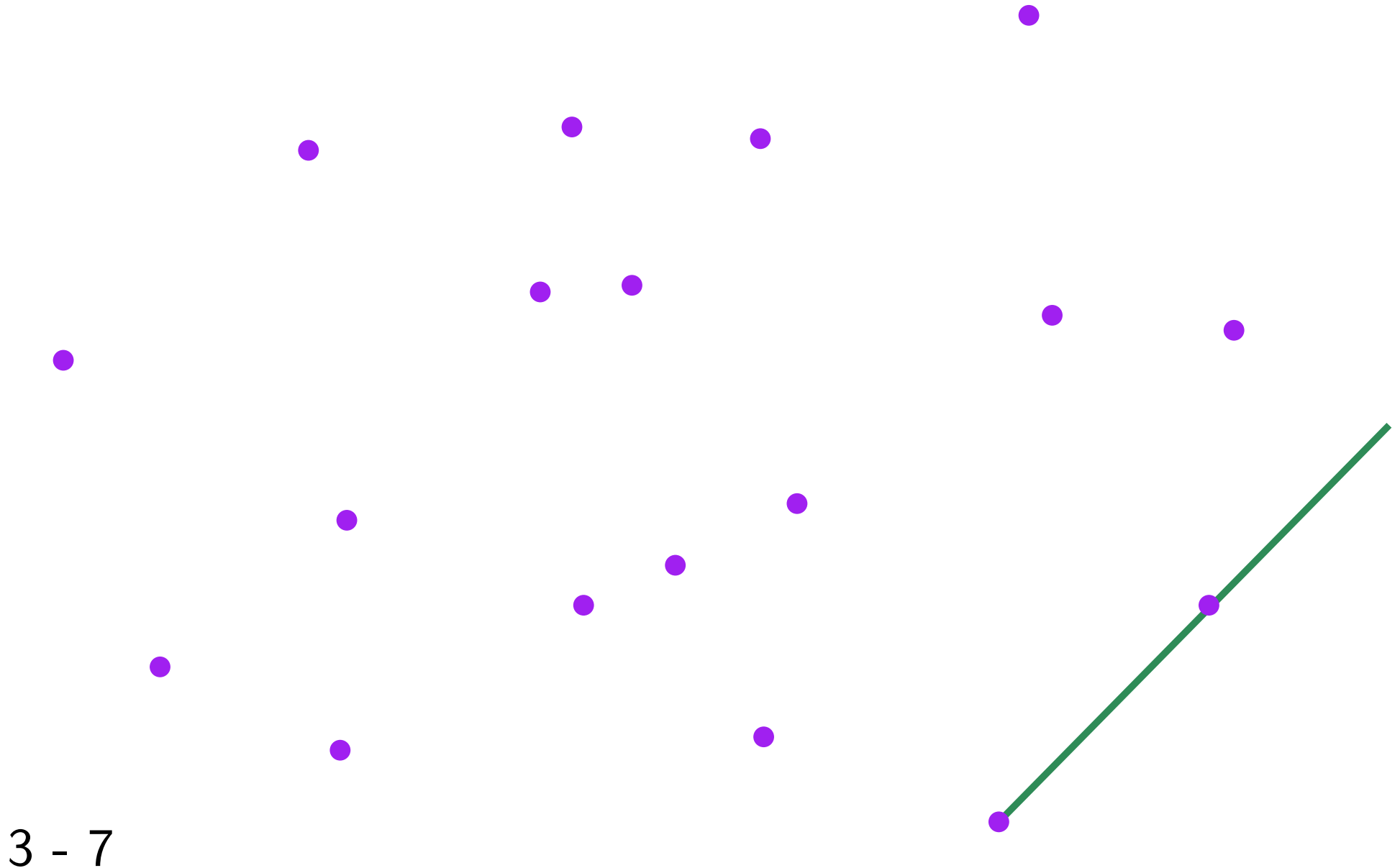


3 - 6

Convex hull

next vertex found

Jarvis algorithm

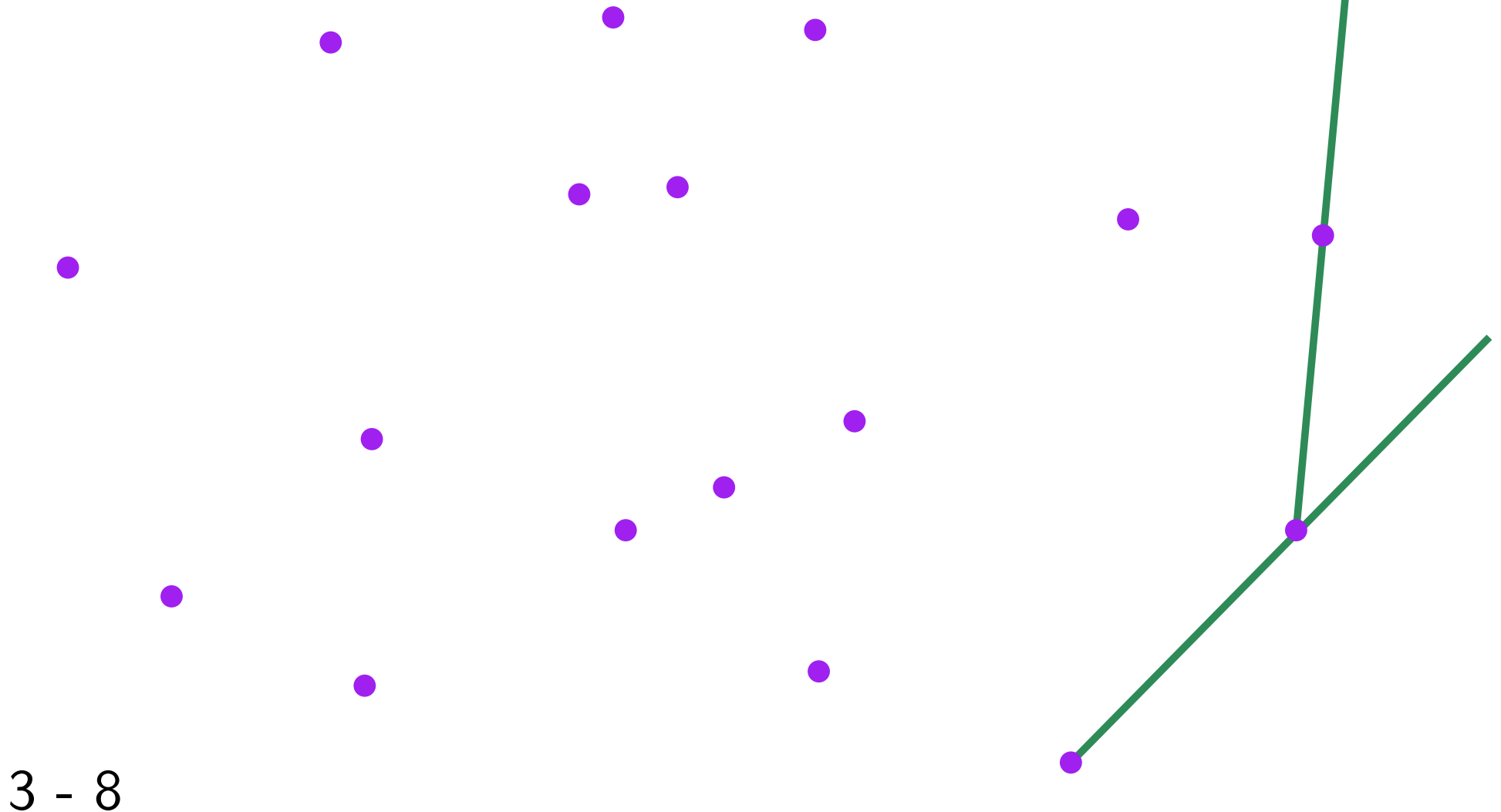


3 - 7

Convex hull

next vertex found
and next one

Jarvis algorithm

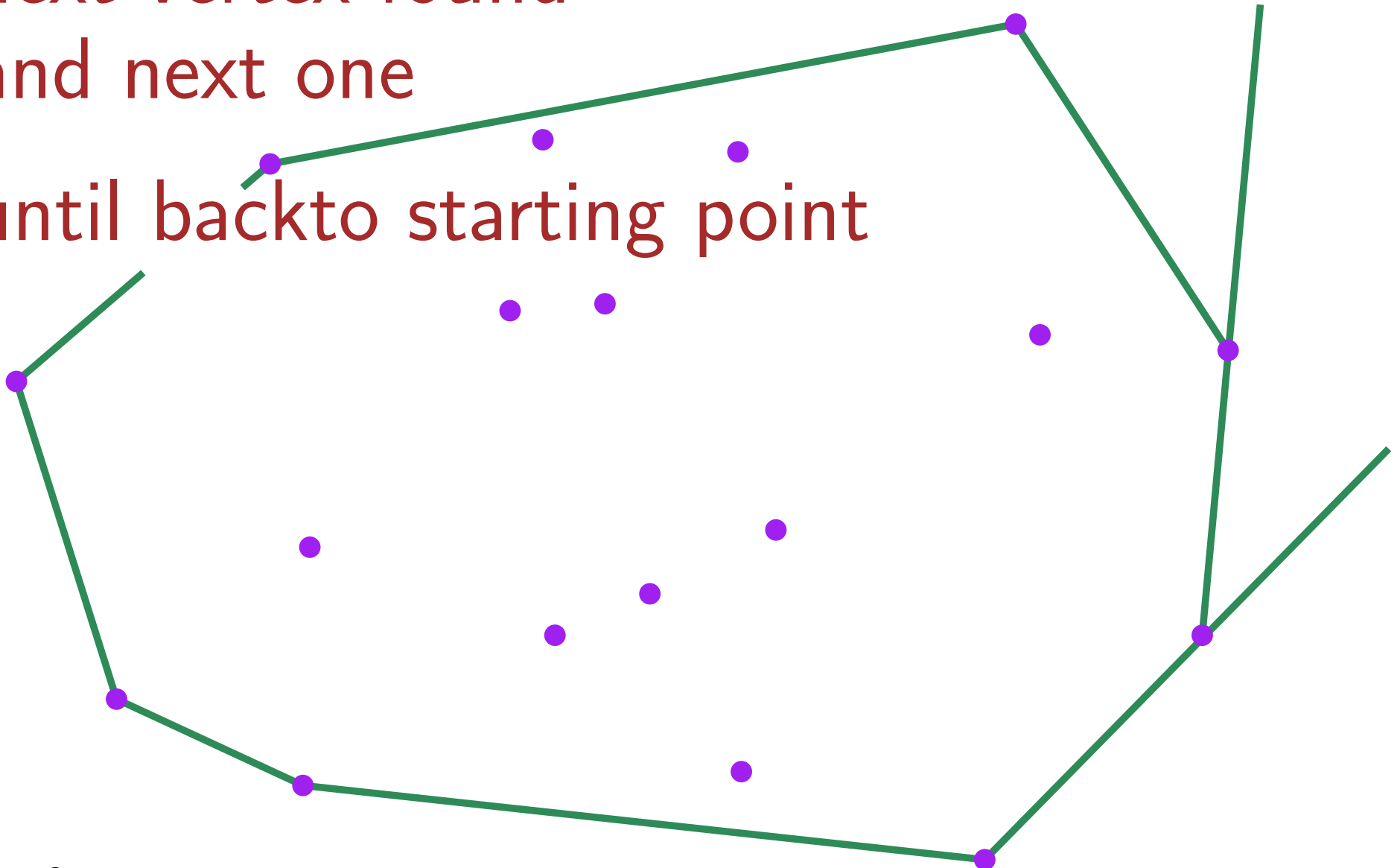


Convex hull

Jarvis algorithm

next vertex found
and next one

until back to starting point



Convex hull

Jarvis algorithm

Input : point set S

$u =$ lowest point in S ; $min = \infty$

For each $w \in S \setminus \{u\}$

 if $angle(ux, uw) < min$

 then $min = angle(ux, uw)$; $v = w$;

$u.next = v$;

Do

$S = S \setminus \{v\}$

 For each $w \in S$

$min = \infty$

 if $angle(v.pred\ v, vw) < min$

 then $min = angle(v.pred\ v, vw)$; $v.next = w$;

$v = v.next$;

While $v \neq u$

Convex hull

Complexity ?

Jarvis algorithm

Input : point set S

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While $v \neq u$

$O(n)$

Convex hull

Complexity ?

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then $min = angle(v.pred\ v, vw)$; $v.next = w$;

$v = v.next$;

While $v \neq u$

$O(n)$

Convex hull

Complexity ?

Jarvis algorithm

Input : point set S

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For each $w \in S \setminus \{u\}$

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Do

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 For each $w \in S$

$min = \infty$

 if $angle(v.pred\ v, vw) < min$

 then $min = angle(v.pred\ v, vw)$; $v.next = w$;

$v = v.next$;

While $v \neq u$

$O(n^2)$

Convex hull

Complexity ?

Jarvis algorithm

Input : point set S

$u =$ lowest point in S ; $min = \infty$

For each $w \in S \setminus \{u\}$

 if $angle(ux, uw) < min$

 then $min = angle(ux, uw)$; $v = w$;

$u.next = v$;

Do

$S = S \setminus \{v\}$

 For each $w \in S$

$min = \infty$

 if $angle(v.pred\ v, vw) < min$

 then $min = angle(v.pred\ v, vw)$; $v.next = w$;

$v = v.next$;

While $v \neq u$

$O(n^2)$

$O(nh)$

Convex hull

Input : point set S

$u =$ lowest point in S ; $min = \infty$

For each $w \in S \setminus \{u\}$

if $angle(ux, uw) < min$

then $min = angle(ux, uw)$; $v = w$;

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Do

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For each $w \in S$

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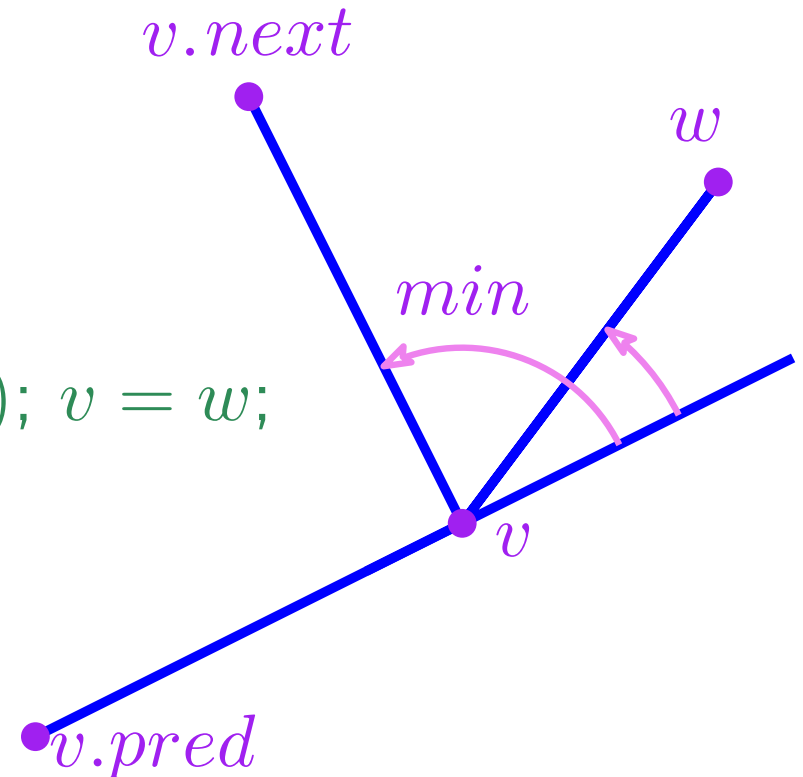
if $angle(v.pred\ v, vw) < min$

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$v = v.next$;

While $v \neq u$

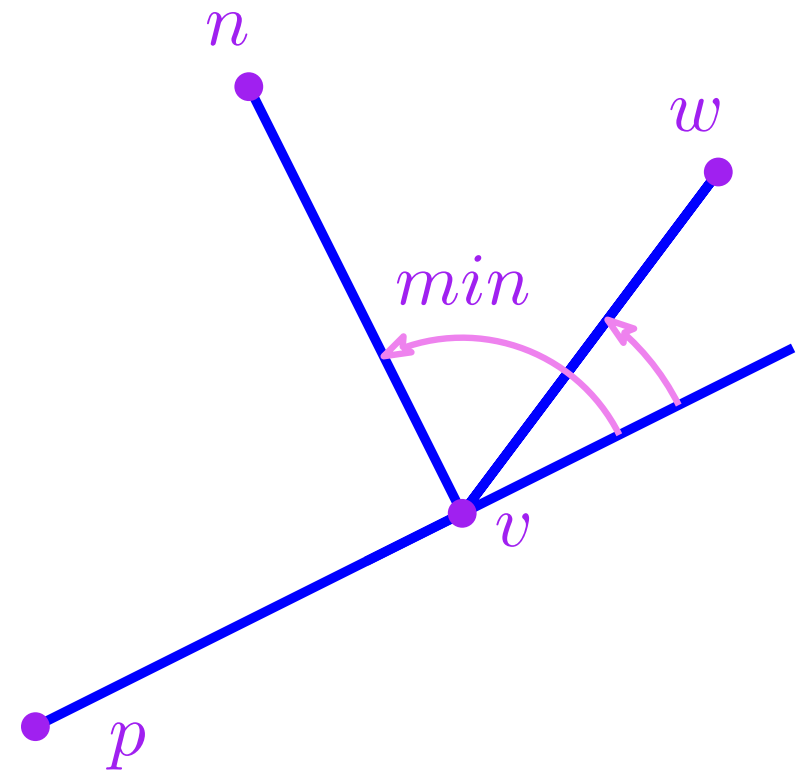
Orientation predicate



Convex hull

if $\text{angle}(pv, vw) < \text{min}$

Orientation predicate



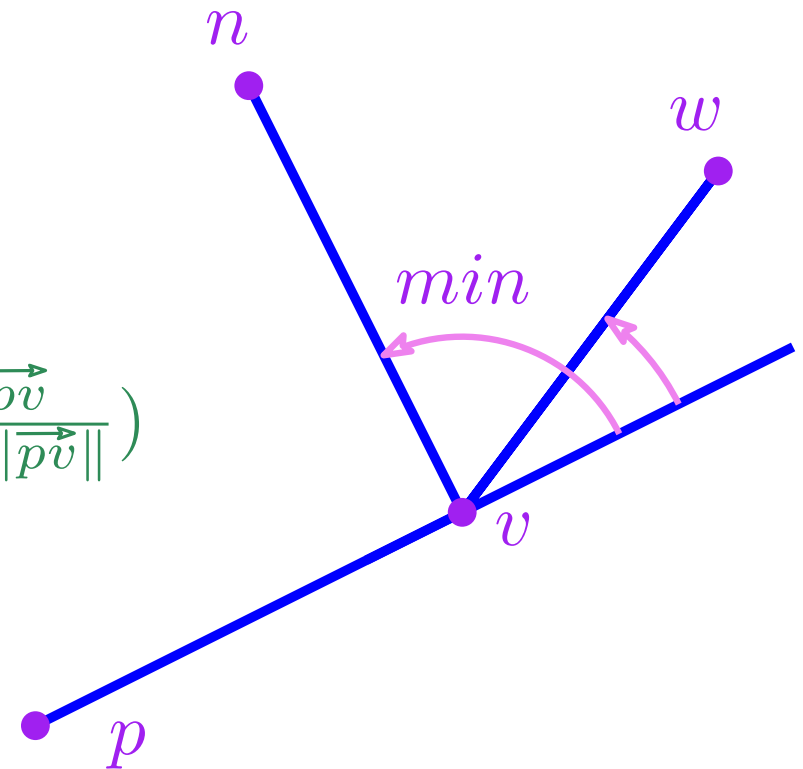
Convex hull

if $\text{angle}(pv, vw) < \text{min}$

$$\text{angle}(pv, vw) = \arccos\left(\frac{\vec{vw} \cdot \vec{pv}}{\|\vec{vw}\| \cdot \|\vec{pv}\|}\right)$$



Orientation predicate

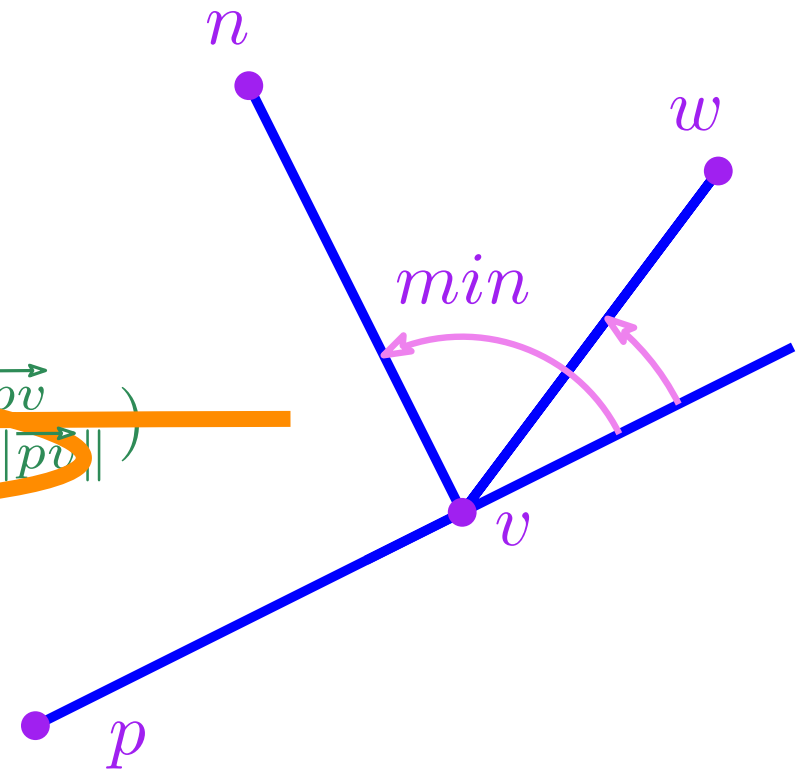


Convex hull

if $\text{angle}(pv, vw) < \text{min}$

Orientation predicate

$$\text{angle}(pv, vw) = \arccos\left(\frac{\vec{vw} \cdot \vec{pv}}{\|\vec{vw}\| \cdot \|\vec{pv}\|}\right)$$



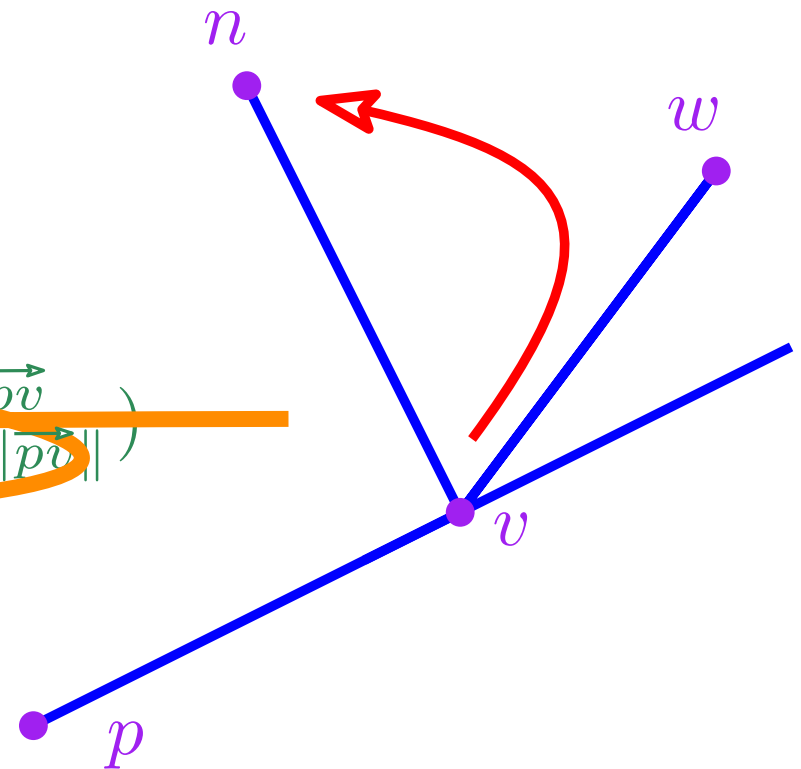
Convex hull

if $\text{angle}(pv, vw) < \min$

$$\text{angle}(pv, vw) = \arccos\left(\frac{\vec{vw} \cdot \vec{pv}}{\|\vec{vw}\| \cdot \|\vec{pv}\|}\right)$$

if vwn turn left

Orientation predicate

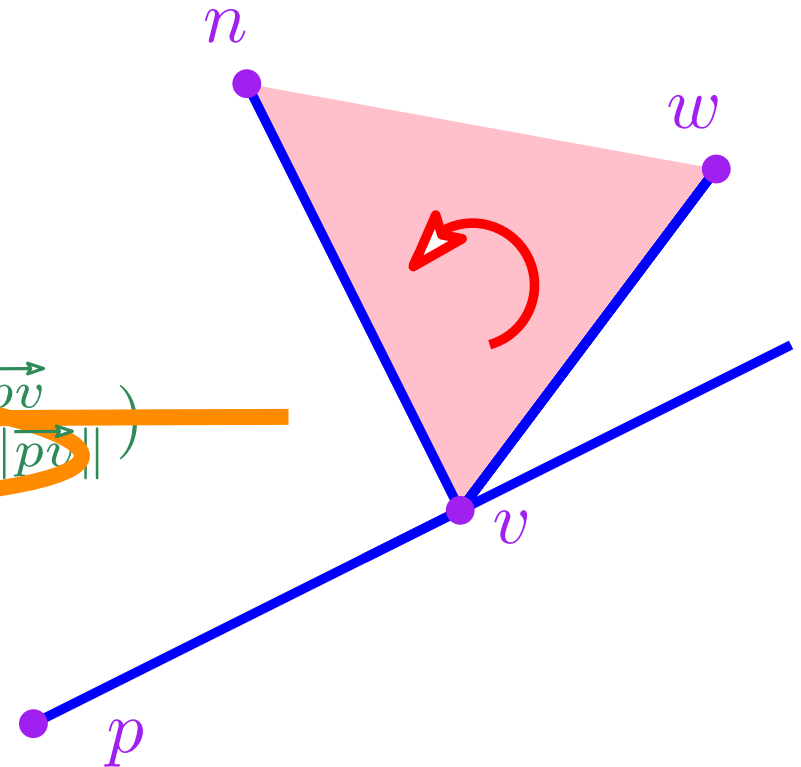


Convex hull

if $\text{angle}(pv, vw) < \min$

$$\text{angle}(pv, vw) = \arccos\left(\frac{\vec{vw} \cdot \vec{pv}}{\|\vec{vw}\| \cdot \|\vec{pv}\|}\right)$$

Orientation predicate



if vwn turn left

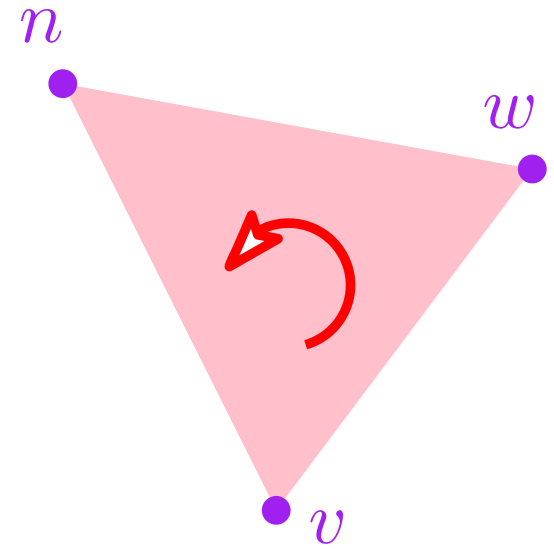
if triangle vwn counterclockwise (ccw)

if triangle vwn positively oriented

Convex hull

$vwn + ?$

Orientation predicate

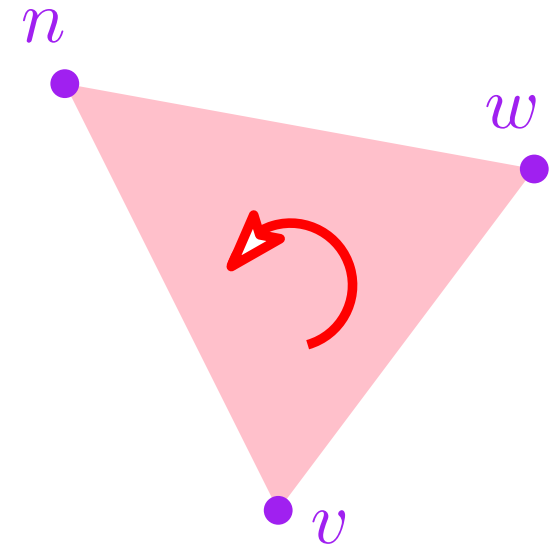


Convex hull

$vwn + ?$

$$\begin{vmatrix} x_w - x_v & x_n - x_v \\ y_w - y_v & y_n - y_v \end{vmatrix} = \begin{vmatrix} 1 & 1 & 1 \\ x_v & x_w & x_n \\ y_v & y_w & y_n \end{vmatrix} > 0$$

Orientation predicate



Convex hull

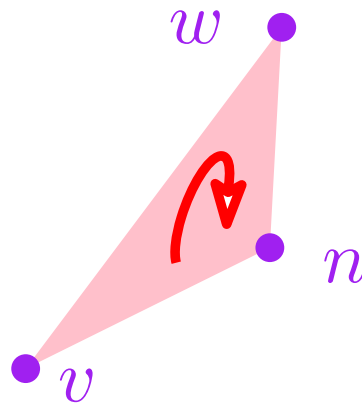
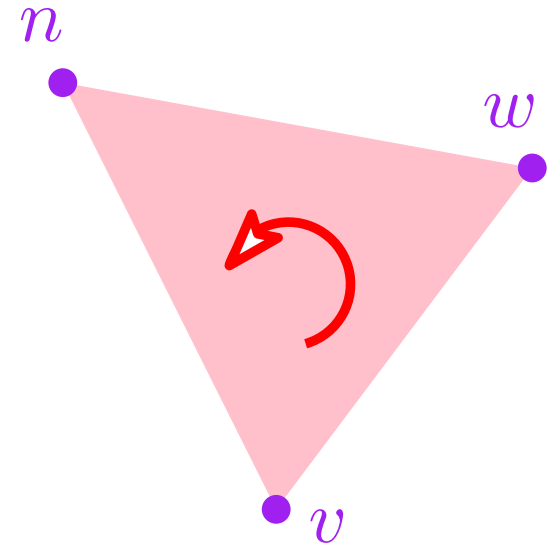
Orientation predicate

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$$\begin{vmatrix} x_w - x_v & x_n - x_v \\ y_w - y_v & y_n - y_v \end{vmatrix} = \begin{vmatrix} 1 & 1 & 1 \\ x_v & x_w & x_n \\ y_v & y_w & y_n \end{vmatrix} > 0$$

$vwn - ?$

$$\begin{vmatrix} 1 & 1 & 1 \\ x_v & x_w & x_n \\ y_v & y_w & y_n \end{vmatrix} < 0$$

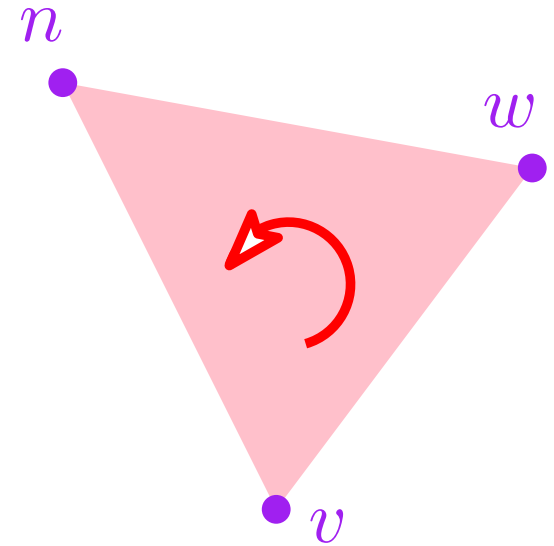


Convex hull

Orientation predicate

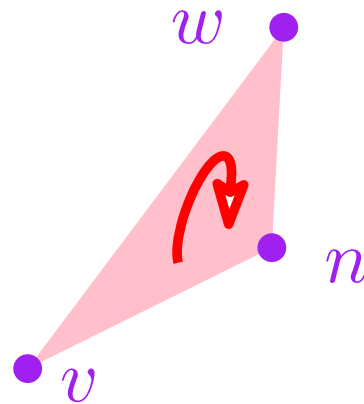
$vwn + ?$

$$\begin{vmatrix} x_w - x_v & x_n - x_v \\ y_w - y_v & y_n - y_v \end{vmatrix} = \begin{vmatrix} 1 & 1 & 1 \\ x_v & x_w & x_n \\ y_v & y_w & y_n \end{vmatrix} > 0$$



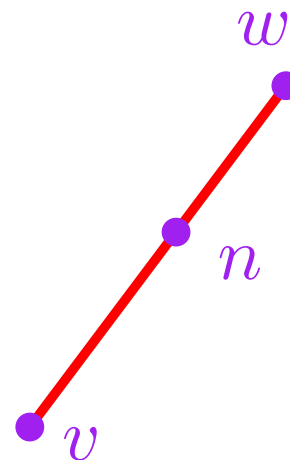
$vwn - ?$

$$\begin{vmatrix} 1 & 1 & 1 \\ x_v & x_w & x_n \\ y_v & y_w & y_n \end{vmatrix} < 0$$



$vwn 0 ?$

$$\begin{vmatrix} 1 & 1 & 1 \\ x_v & x_w & x_n \\ y_v & y_w & y_n \end{vmatrix} = 0$$



Convex hull

Orientation predicate

$vwn + ?$

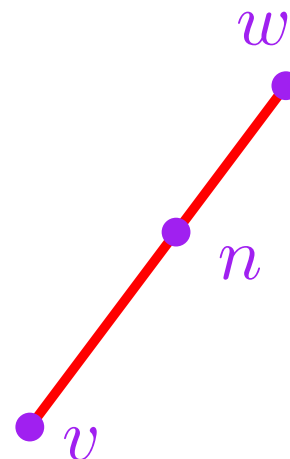
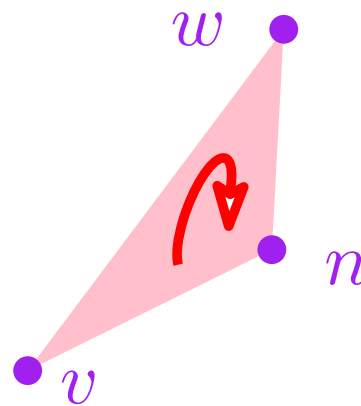
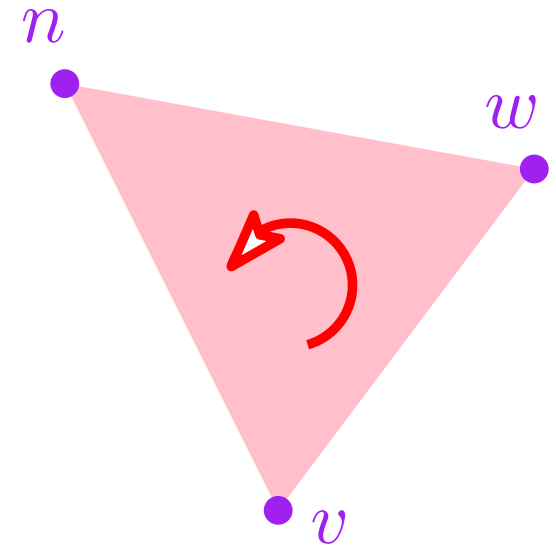
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$vwn - ?$

$$\begin{vmatrix} 1 & 1 & 1 \\ x_v & x_w & x_n \\ y_v & y_w & y_n \end{vmatrix} < 0$$

$vwn 0 ?$

$$\begin{vmatrix} 1 & 1 & 1 \\ x_v & x_w & x_n \\ y_v & y_w & y_n \end{vmatrix} = 0$$



degenerate case

Convex hull

Orientation predicate

$vwn + ?$

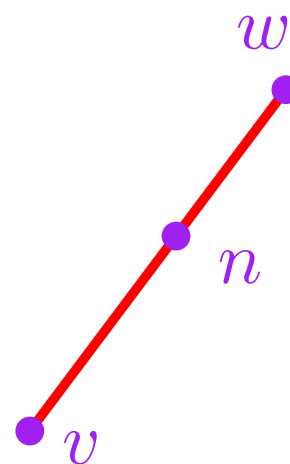
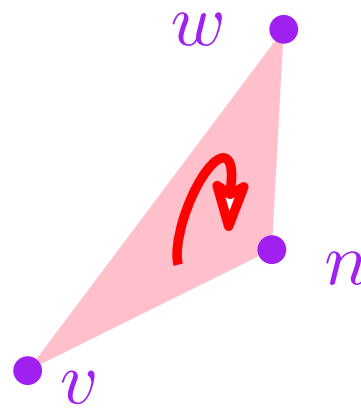
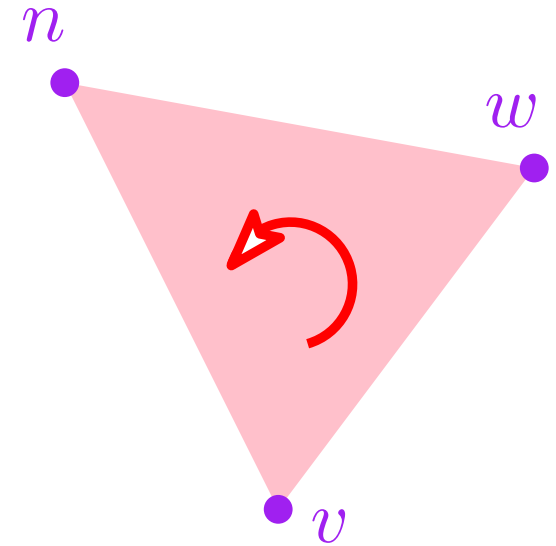
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$$\begin{vmatrix} 1 & 1 & 1 \\ x_v & x_w & x_n \\ y_v & y_w & y_n \end{vmatrix} = 0$$



rounding errors



Convex hull

Rounding errors possible

Orientation predicate

$$p = \left(\frac{1}{2} + x.u, \frac{1}{2} + y.u\right)$$

$$0 \leq x, y \leq 256, u = 2^{-53}$$

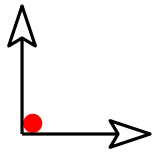
$$q = (12, 12)$$

$$r = (24, 24)$$

Teaser robustness lecture

orientation(p, q, r)

evaluated with double



Convex hull

Rounding errors possible

Orientation predicate

$$p = \left(\frac{1}{2} + x.u, \frac{1}{2} + y.u\right)$$

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Teaser robustness lecture

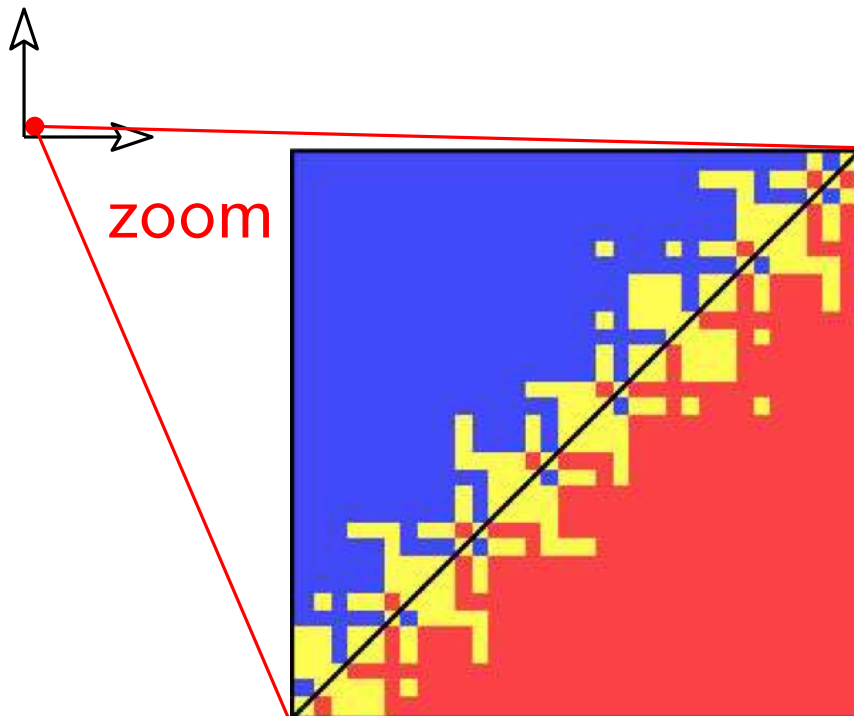
$orientation(p, q, r)$

evaluated with double

$$\leq 0$$

$$0$$

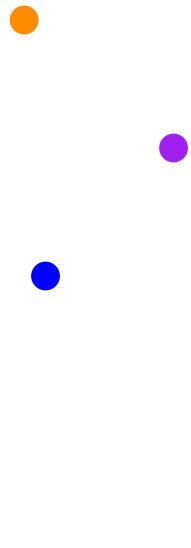
$$\geq 0$$



Convex hull

Teaser robustness lecture

Buggy degenerate example
(single precision)



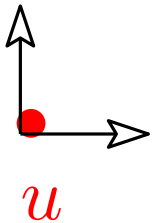
$$w_1 = (12, 12)$$

$$w_2 = (24, 24)$$

$$w_3 = (30, 30.000001)$$

$$w_4 = (23, 36)$$

$$w_5 = (0.5000029, 0.5000027)$$



Convex hull

Teaser robustness lecture

Buggy degenerate example
(single precision)

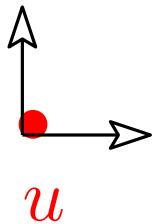
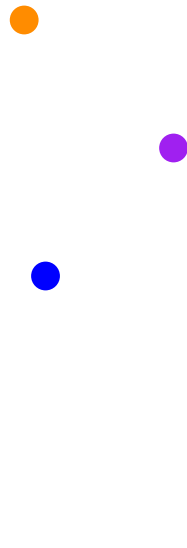
$$w_1 = (12, 12)$$

$$w_2 = (24, 24)$$

$$w_3 = (30, 30.0000001)$$

$$w_4 = (23, 36)$$

$$w_5 = (0.50000029, 0.50000027)$$



Input : point set S

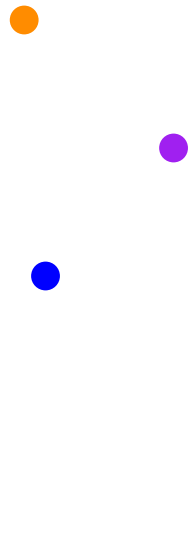
$u = v =$ lowest point in S ;

Jarvis

Convex hull

Teaser robustness lecture

Buggy degenerate example
(single precision)



$$w_1 = (12, 12)$$

$$w_2 = (24, 24)$$

$$w_3 = (30, 30.000001)$$

$$w_4 = (23, 36)$$

$$w_5 = (0.5000029, 0.5000027)$$

Do

$n = \text{first in } S;$

For each $w \in S$

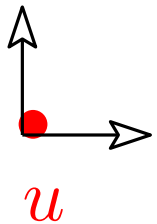
 if vwn positive

 then $n = w;$

$v.next = n; v = n;$

$S = S \setminus \{v\}$

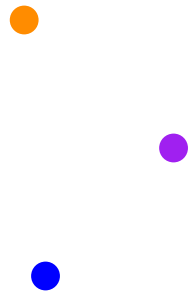
While $v \neq u$



Convex hull

Teaser robustness lecture

Buggy degenerate example
(single precision)



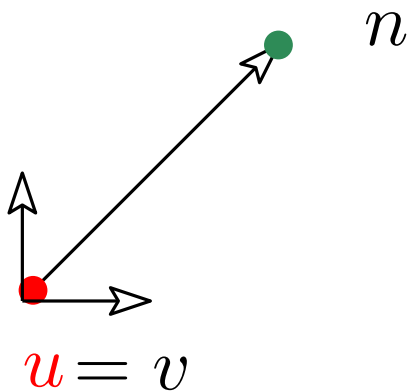
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$$w_2 = (24, 24)$$

$$w_3 = (30, 30.0000001)$$

$$w_4 = (23, 36)$$

$$w_5 = (0.50000029, 0.50000027)$$



Do

$n = \text{first in } S;$

For each $w \in S$

if vwn positive

then $n = w;$

$v.next = n; v = n;$

$S = S \setminus \{v\}$

While $v \neq u$

Convex hull

Teaser robustness lecture

Buggy degenerate example
(single precision)

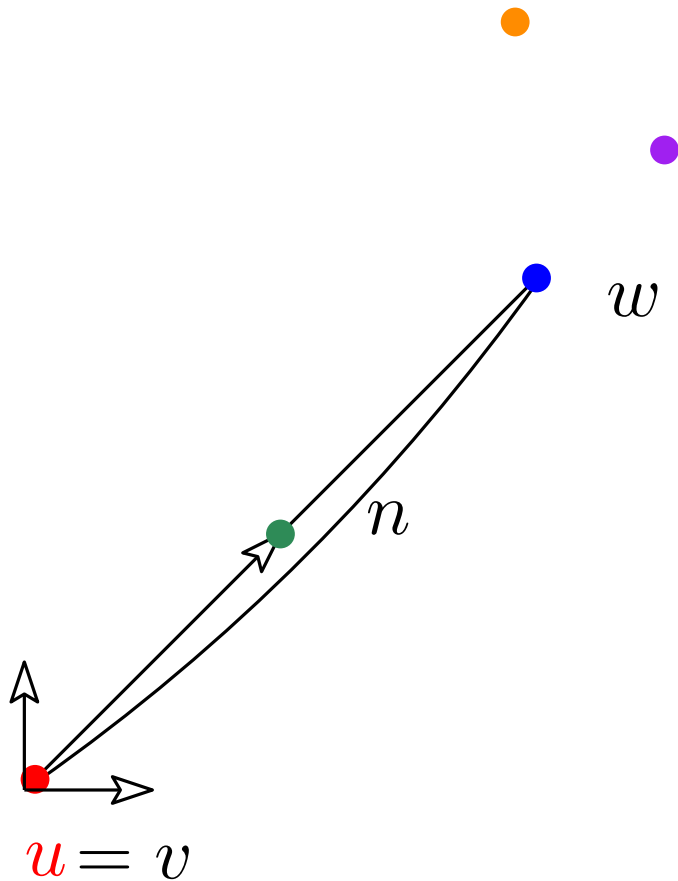
$$w_1 = (12, 12)$$

$$w_2 = (24, 24)$$

$$w_3 = (30, 30.0000001)$$

$$w_4 = (23, 36)$$

$$w_5 = (0.50000029, 0.50000027)$$



Do

$n = \text{first in } S;$

For each $w \in S$

if vwn positive

then $n = w;$

$v.next = n; v = n;$

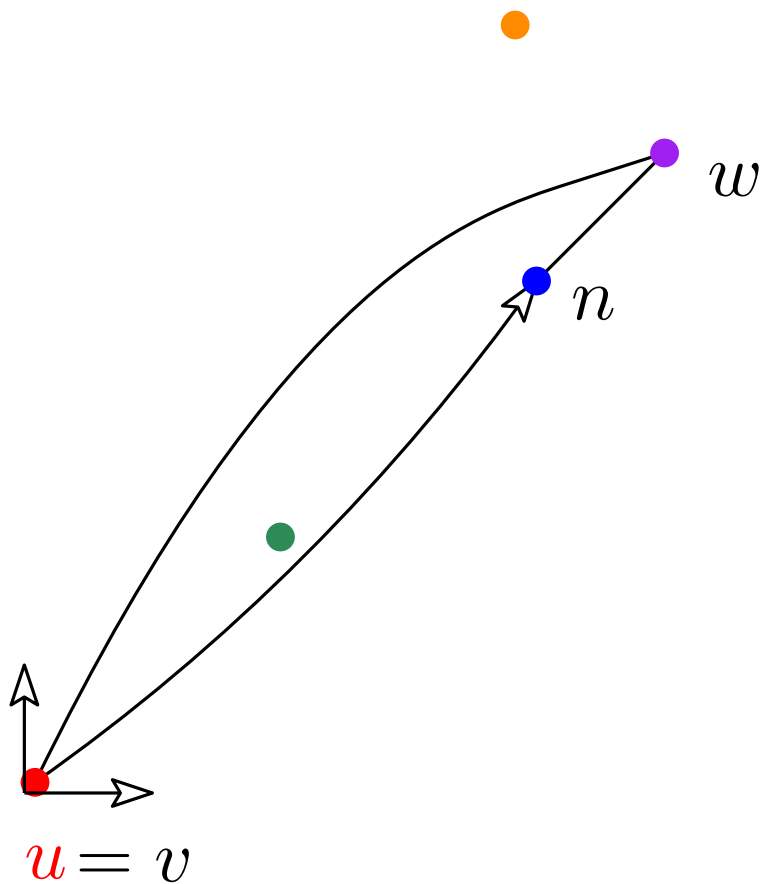
$S = S \setminus \{v\}$

While $v \neq u$

Convex hull

Teaser robustness lecture

Buggy degenerate example
(single precision)



$$w_1 = (12, 12)$$

$$w_2 = (24, 24)$$

$$w_3 = (30, 30.0000001)$$

$$w_4 = (23, 36)$$

$$w_5 = (0.50000029, 0.50000027)$$

Do

$n = \text{first in } S;$

For each $w \in S$

if vwn positive

then $n = w;$

$v.next = n; v = n;$

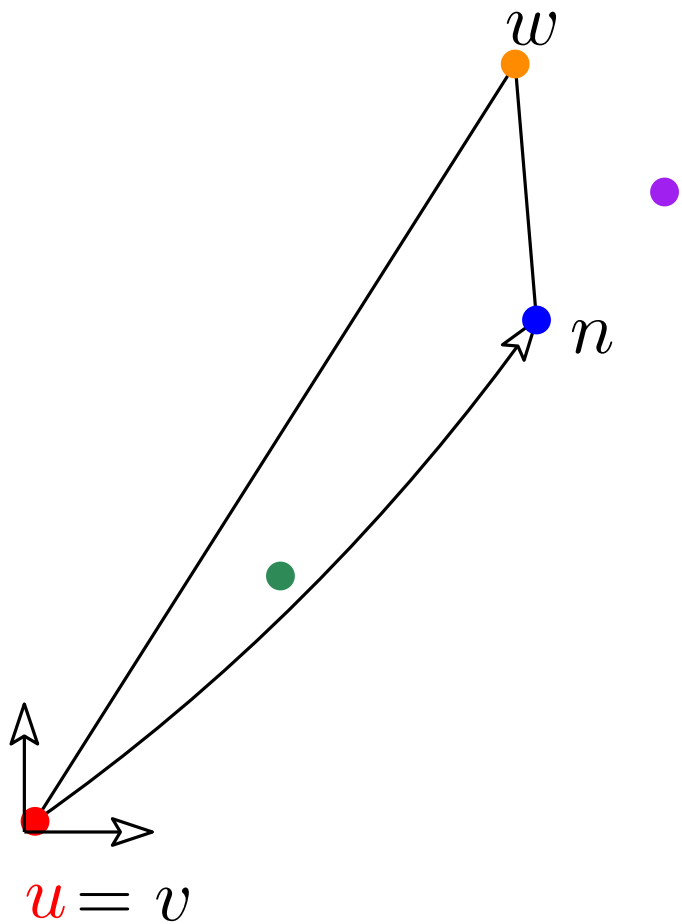
$S = S \setminus \{v\}$

While $v \neq u$

Convex hull

Teaser robustness lecture

Buggy degenerate example
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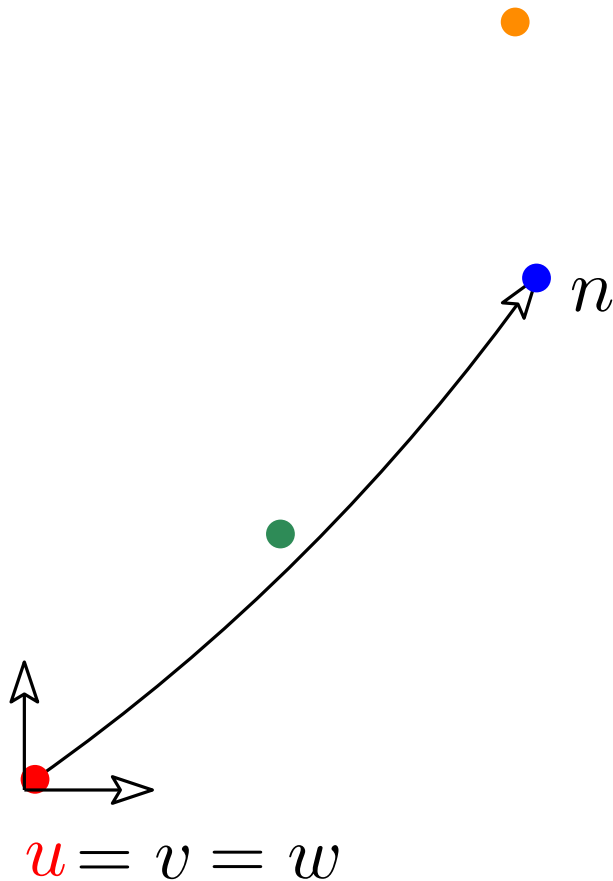
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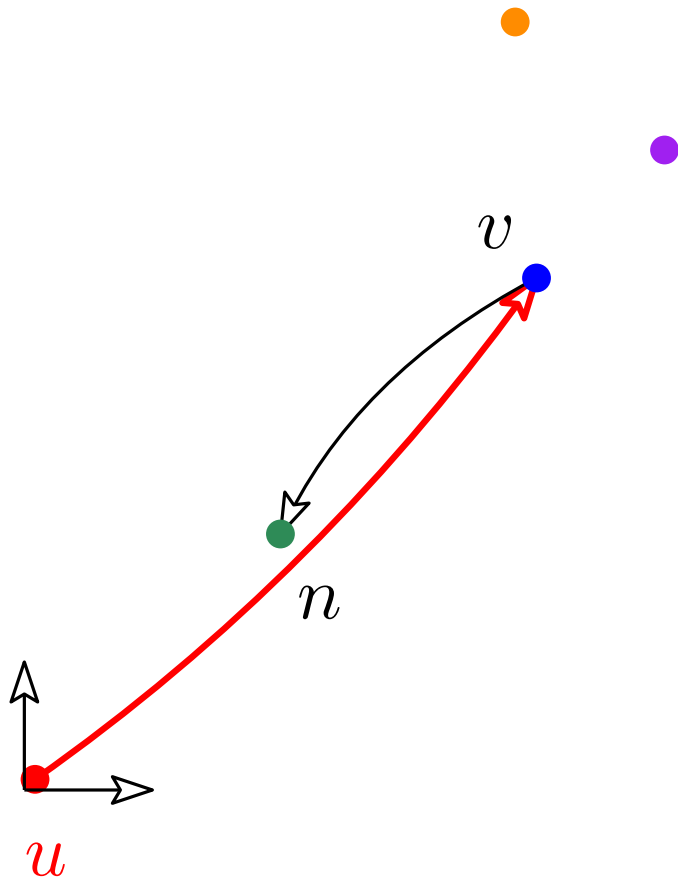
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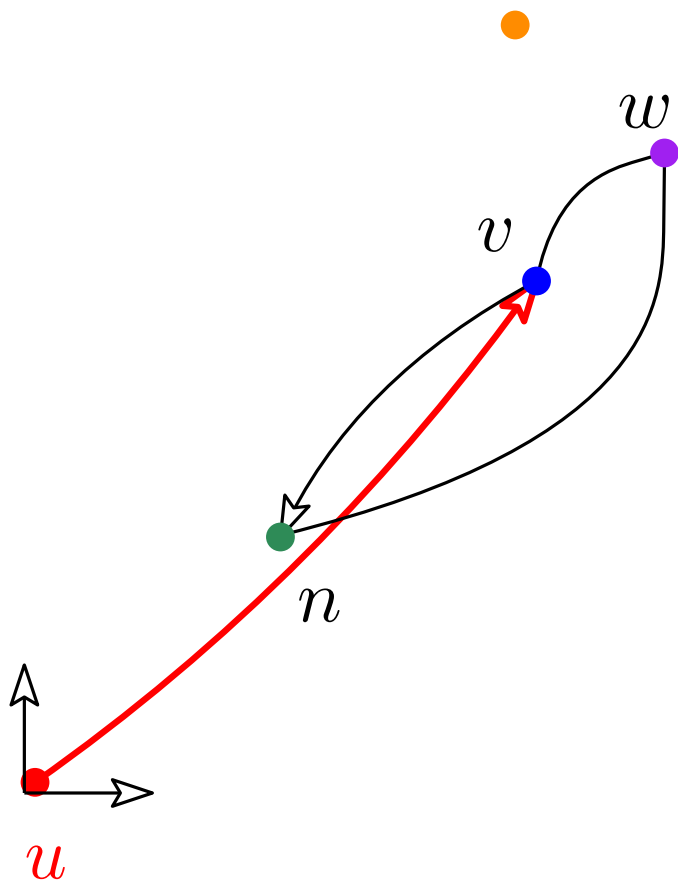
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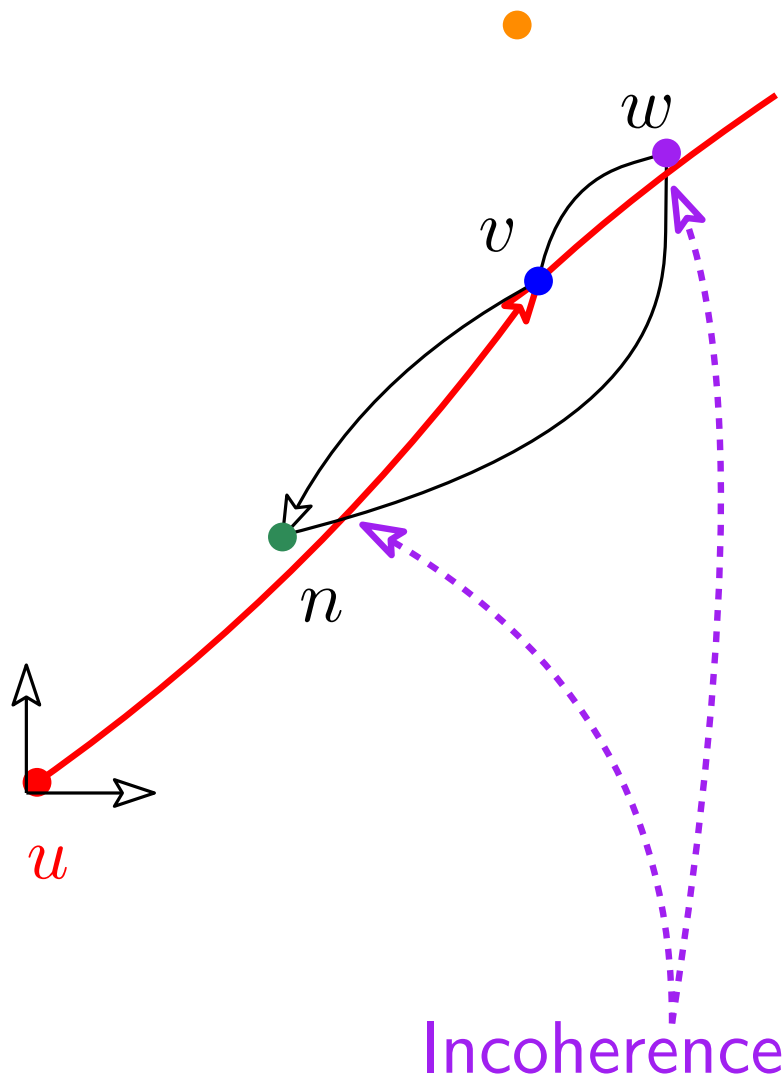
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Convex hull

Teaser robustness lecture

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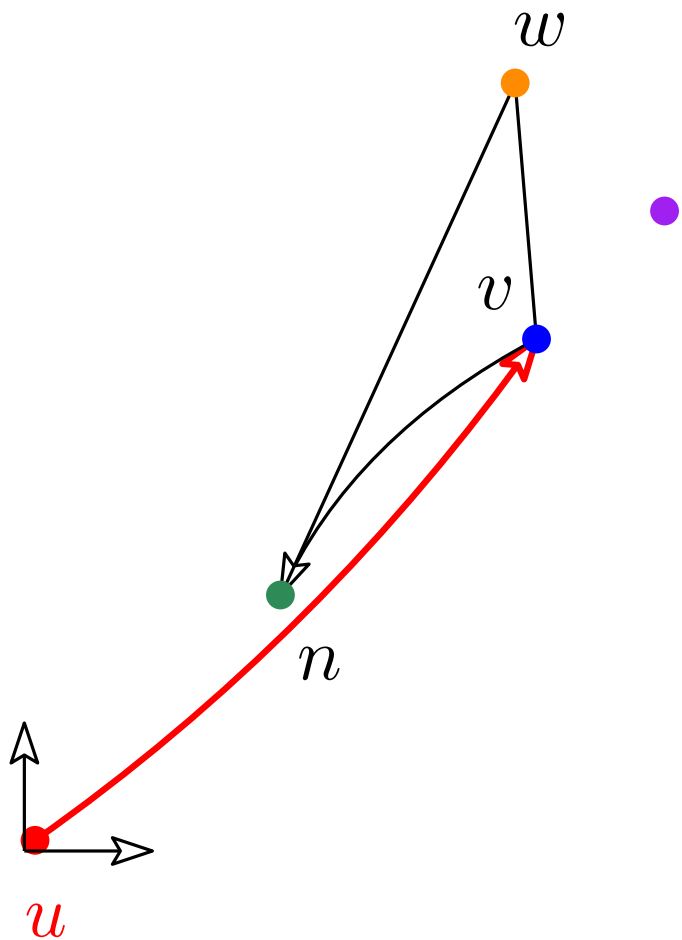
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Convex hull

Teaser robustness lecture

Buggy degenerate example
(single precision)



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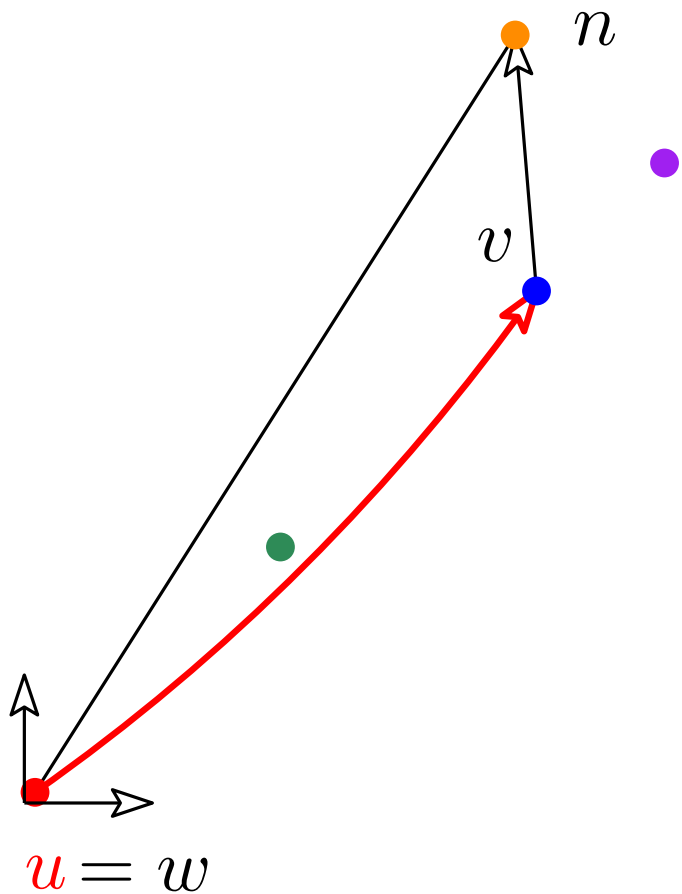
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Convex hull

Teaser robustness lecture

Buggy degenerate example
(single precision)



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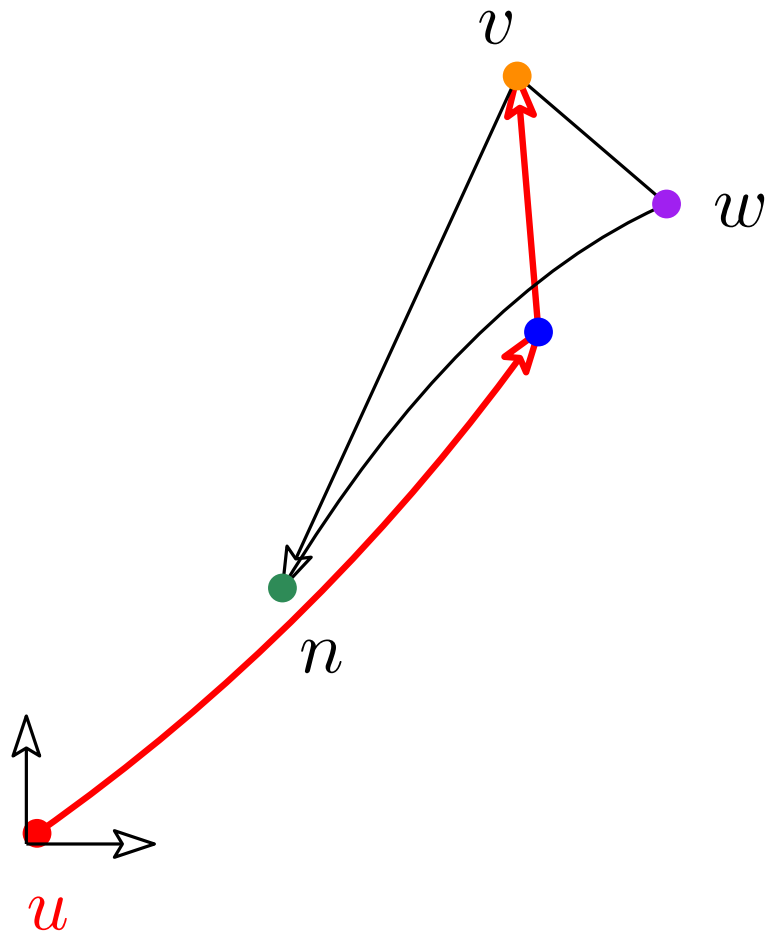
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While $v \neq u$

Convex hull

Teaser robustness lecture

Buggy degenerate example
(single precision)



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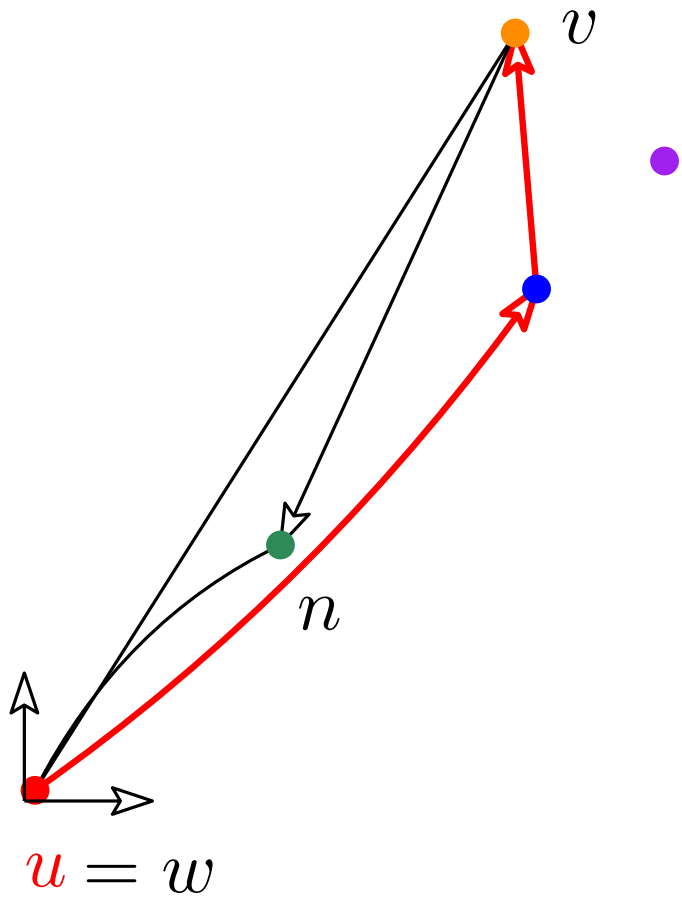
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While $v \neq u$

Convex hull

Teaser robustness lecture

Buggy degenerate example
(single precision)



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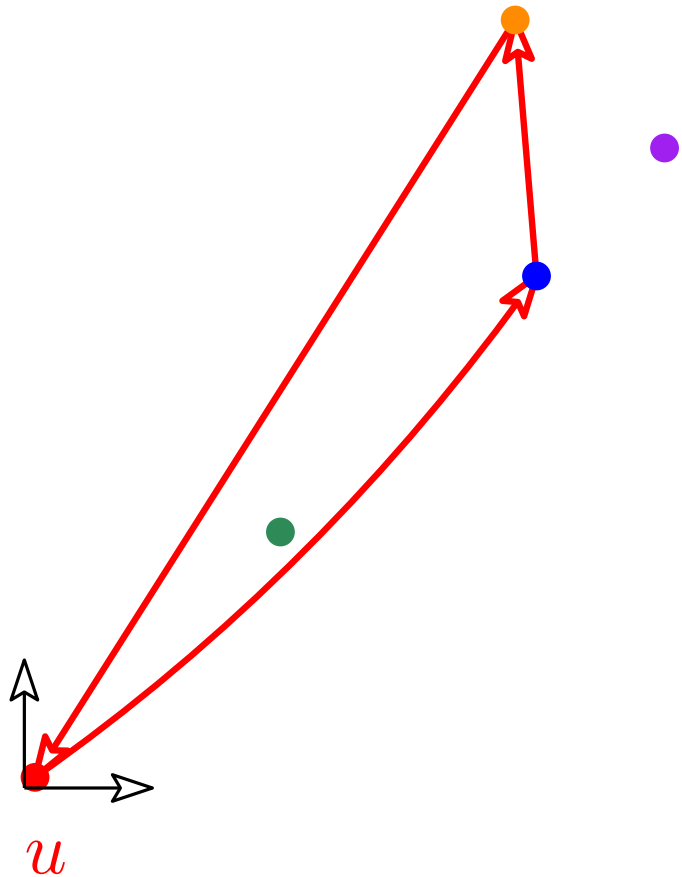
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While $v \neq u$

Convex hull

Teaser robustness lecture

Buggy degenerate example
(single precision)



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$u = w_5 = (0.50000029, 0.50000027)$

Do

```
 $n = \text{first in } S;$   
For each  $w \in S$   
    if  $vwn$  positive  
        then  $n = w;$   
 $v.next = n; v = n;$   
 $S = S \setminus \{v\}$ 
```

While $v \neq u$

Convex hull

Teaser robustness lecture

Buggy degenerate example
(single precision)

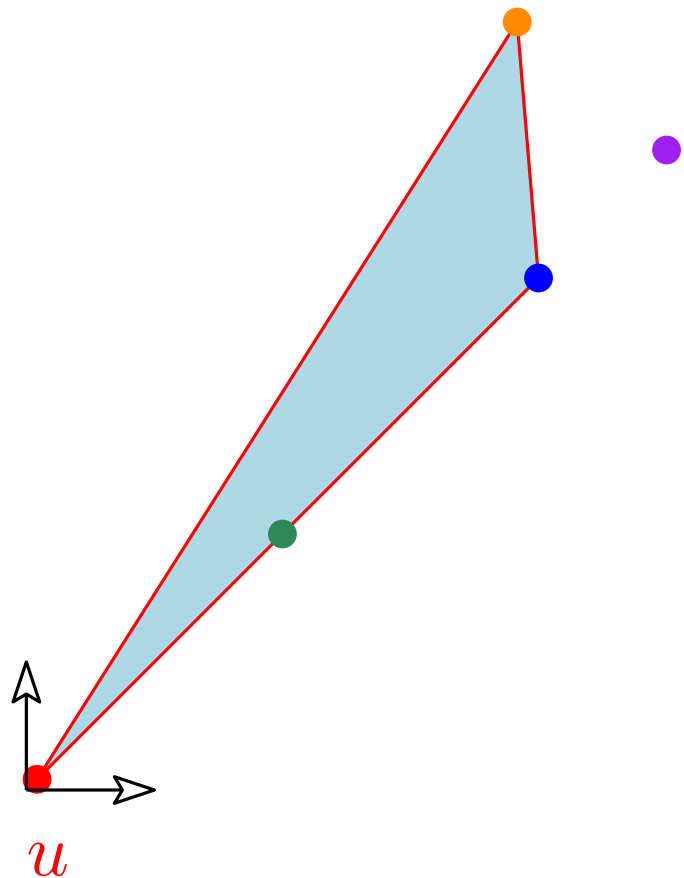
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$$w_2 = (24, 24)$$

$$w_3 = (30, 30.000001)$$

$$w_4 = (23, 36)$$

$$w_5 = (0.5000029, 0.5000027)$$



Result is really wrong

Convex hull

Real RAM model and
general position hypothesis

Real Random Access Memory model

Assume exact computation on real numbers

constant time for single operations: $+$, $-$, $\sqrt{\quad}$, $\sin \dots$

Convex hull

Real RAM model and
general position hypothesis

Real Random Access Memory model

Assume exact computation on real numbers

constant time for single operations: $+$, $-$, $\sqrt{}$, $\sin \dots$

General position hypotheses

Predicate: $\text{Input} \mapsto \{-1, 0, 1\}$

Convex hull

Real RAM model and
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Assume exact computation on real numbers

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General position hypotheses

Predicate: Input $\mapsto \{-1, 0, 1\}$

2D convex hull: no three points colinear

Convex hull

Real RAM model and
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Real Random Access Memory model

Assume exact computation on real numbers

constant time for single operations: $+$, $-$, $\sqrt{\quad}$, $\sin \dots$

General position hypotheses

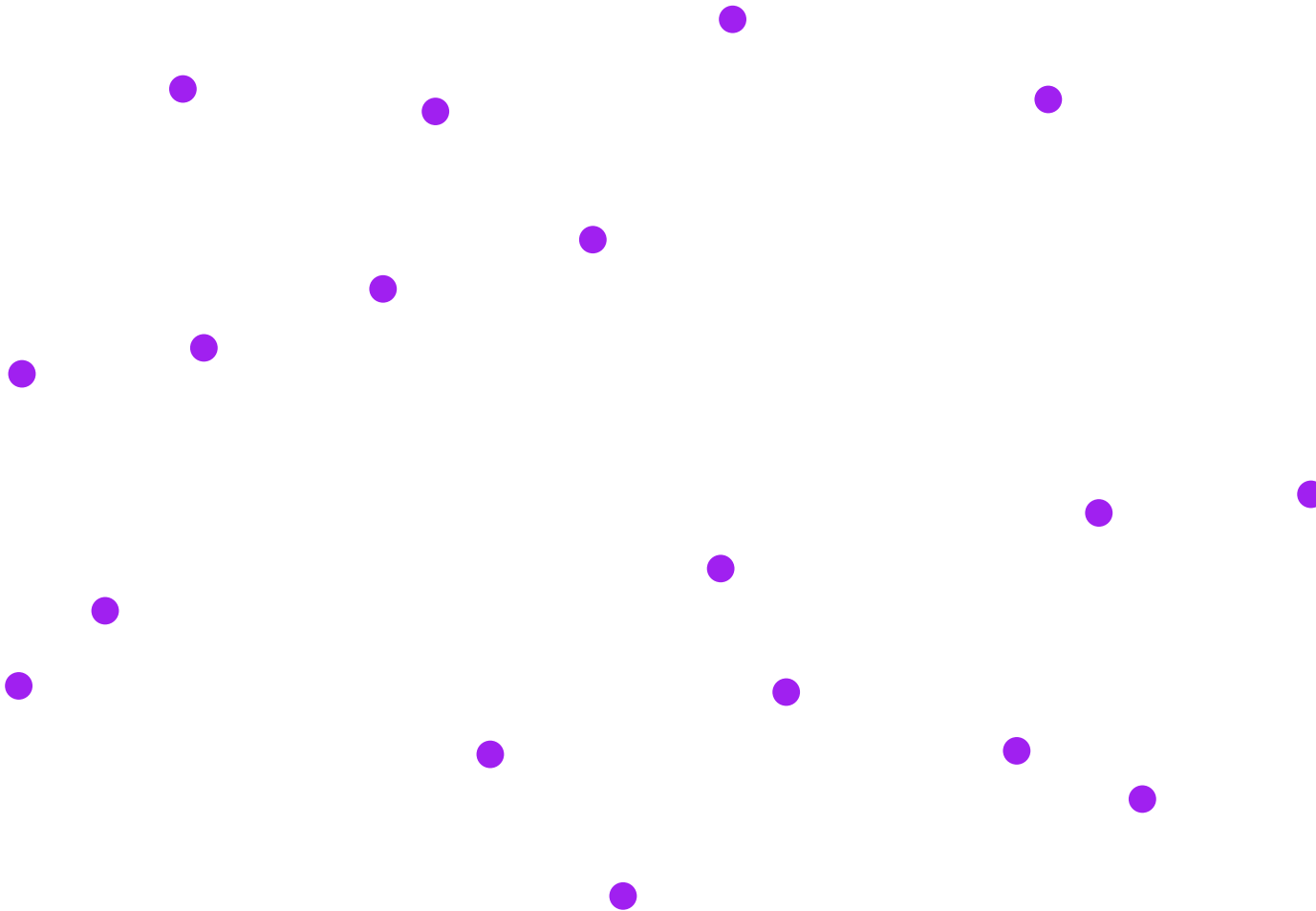
Predicate: Input $\mapsto \{-1, 0, 1\}$

2D convex hull: no three points colinear

possibly: no 2 points with same x

Convex hull

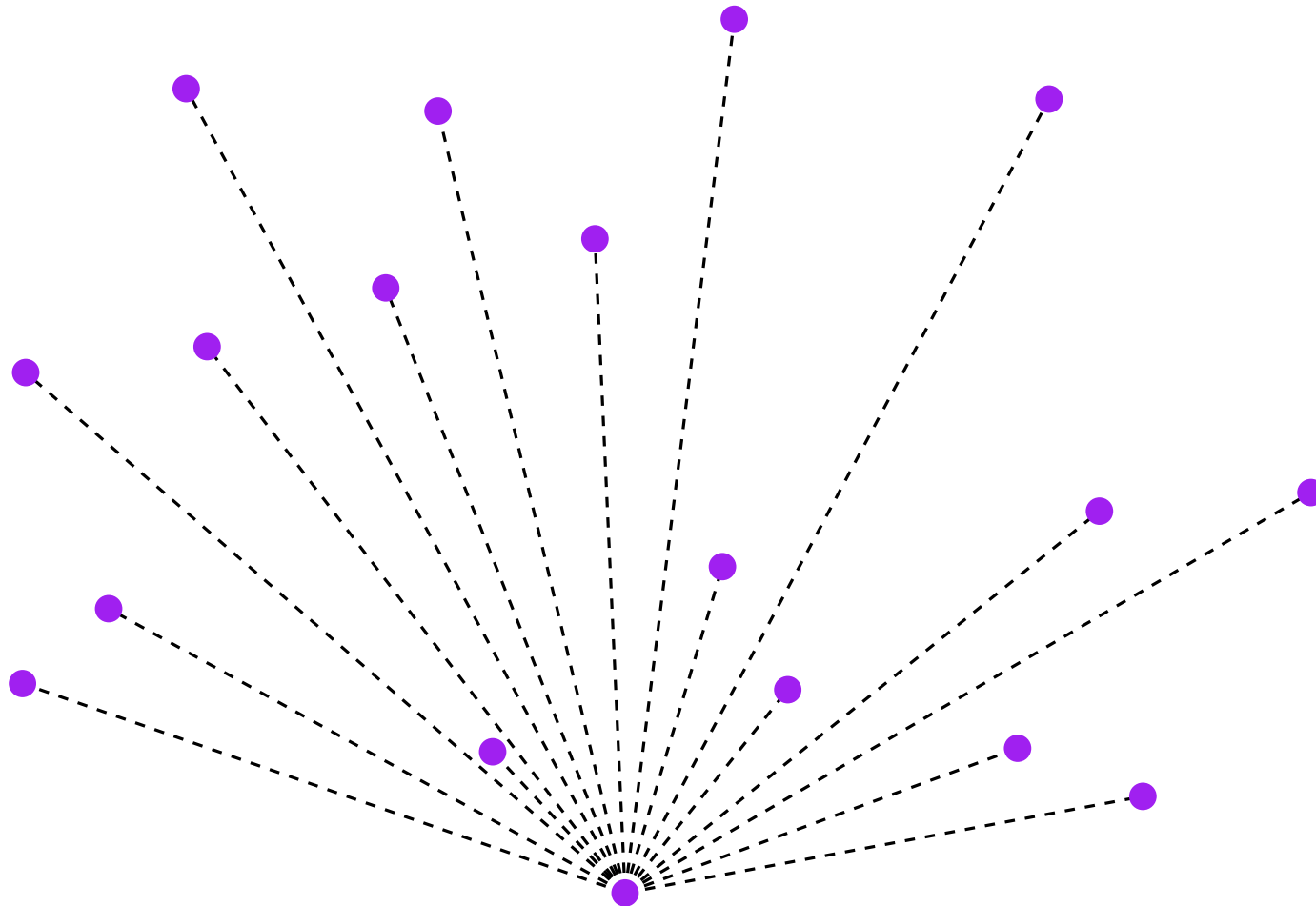
Graham algorithm



Convex hull

Graham algorithm

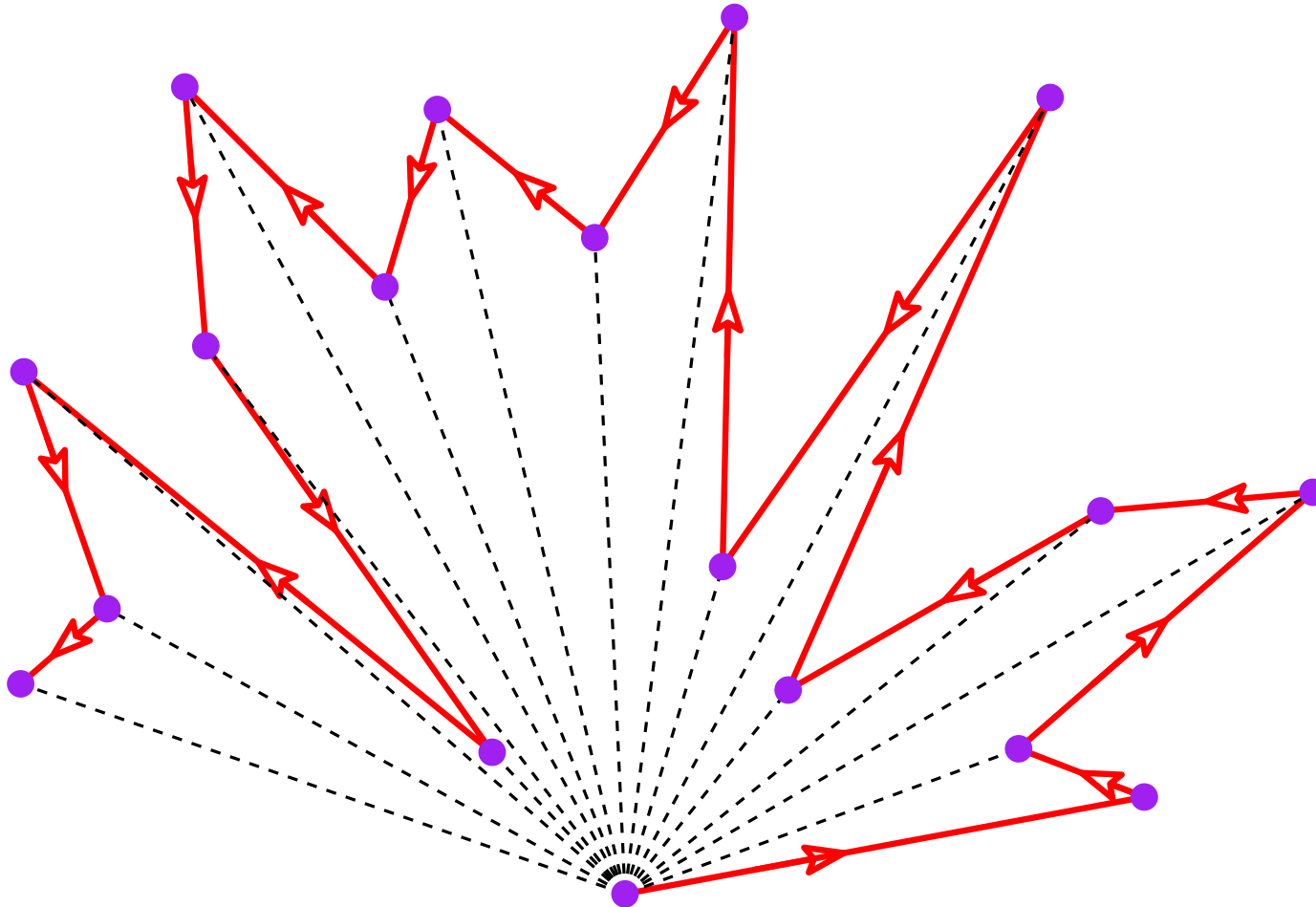
sort around a point (e.g. lowest point)



Convex hull

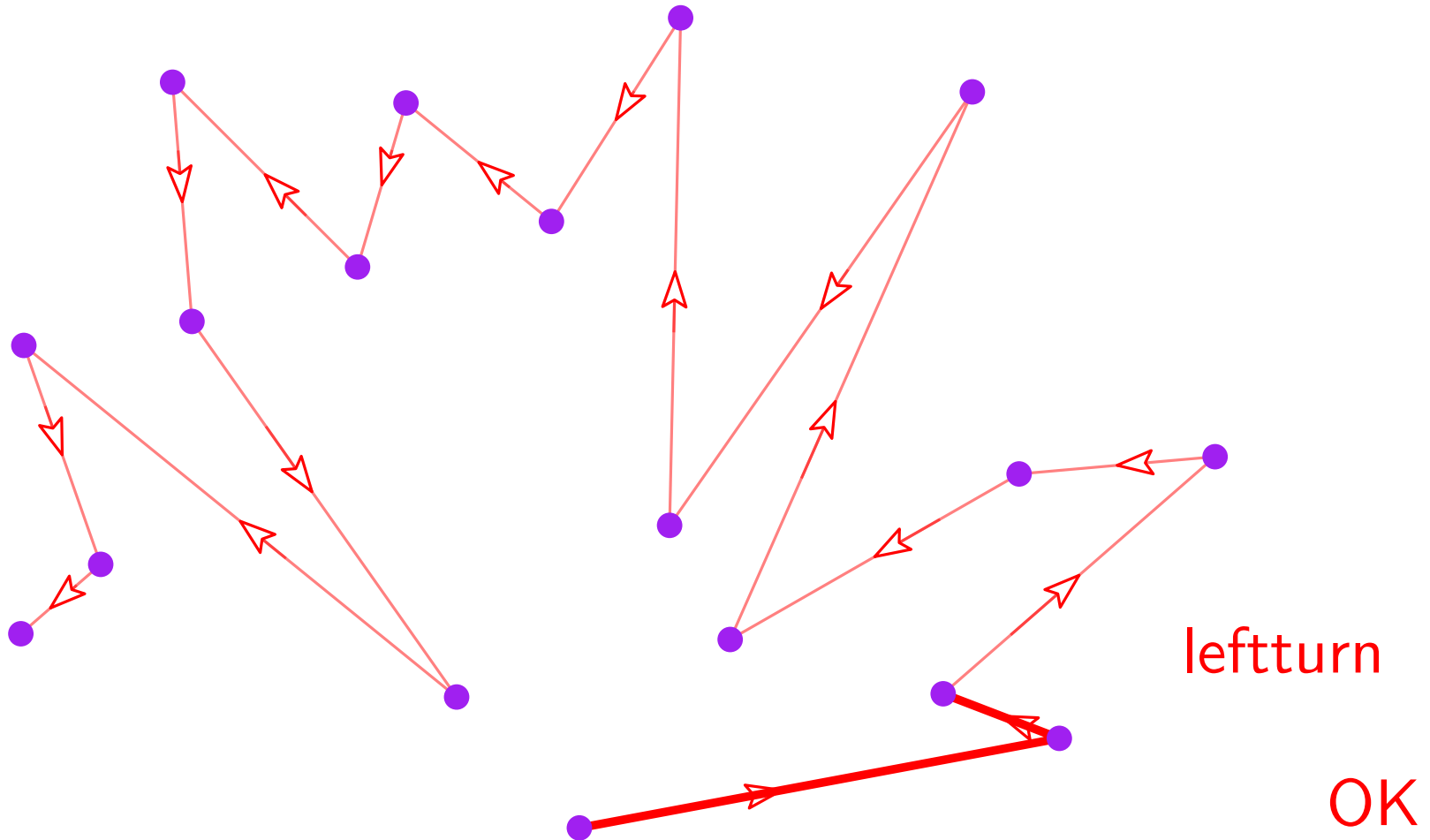
Graham algorithm

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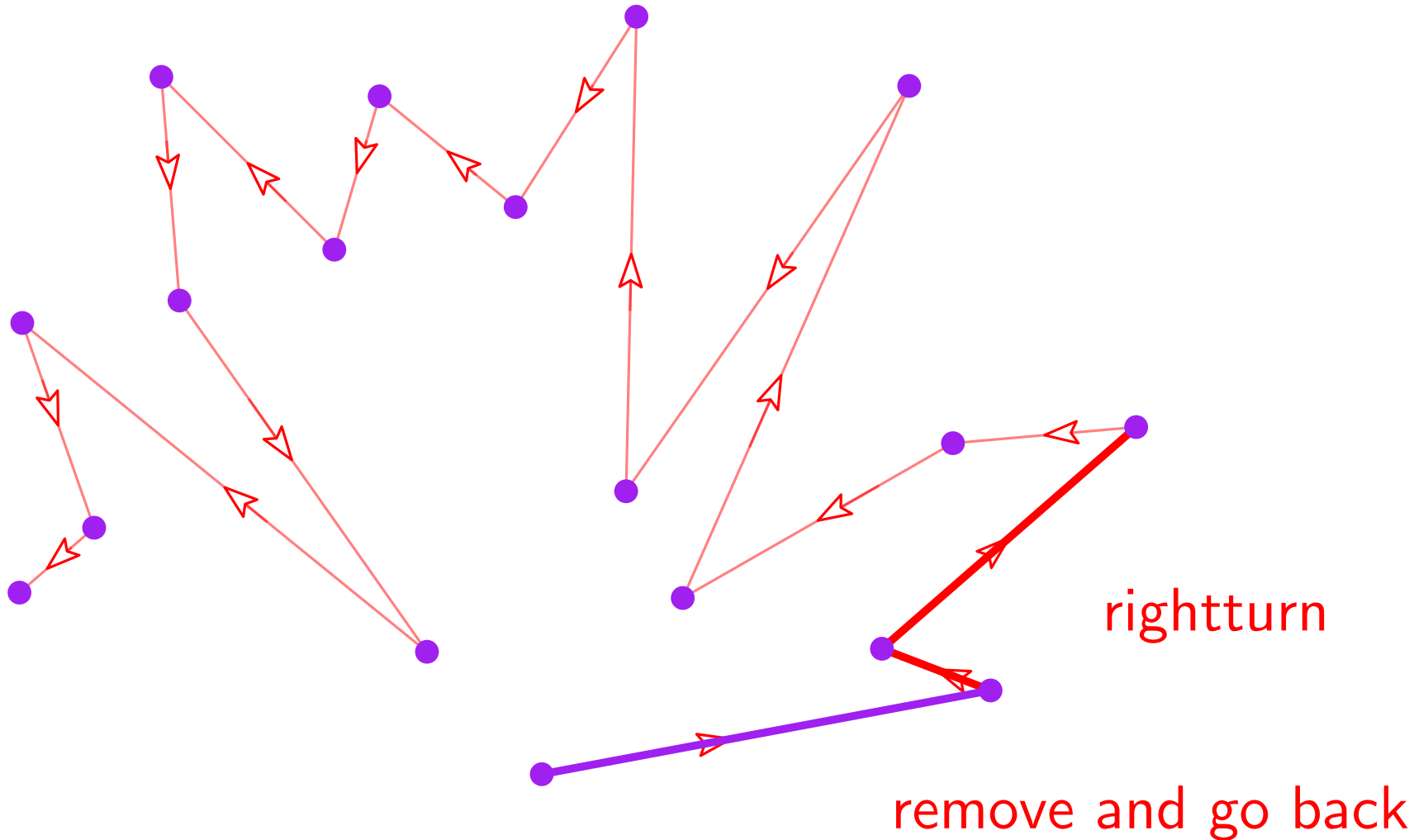
Convex hull

Graham algorithm



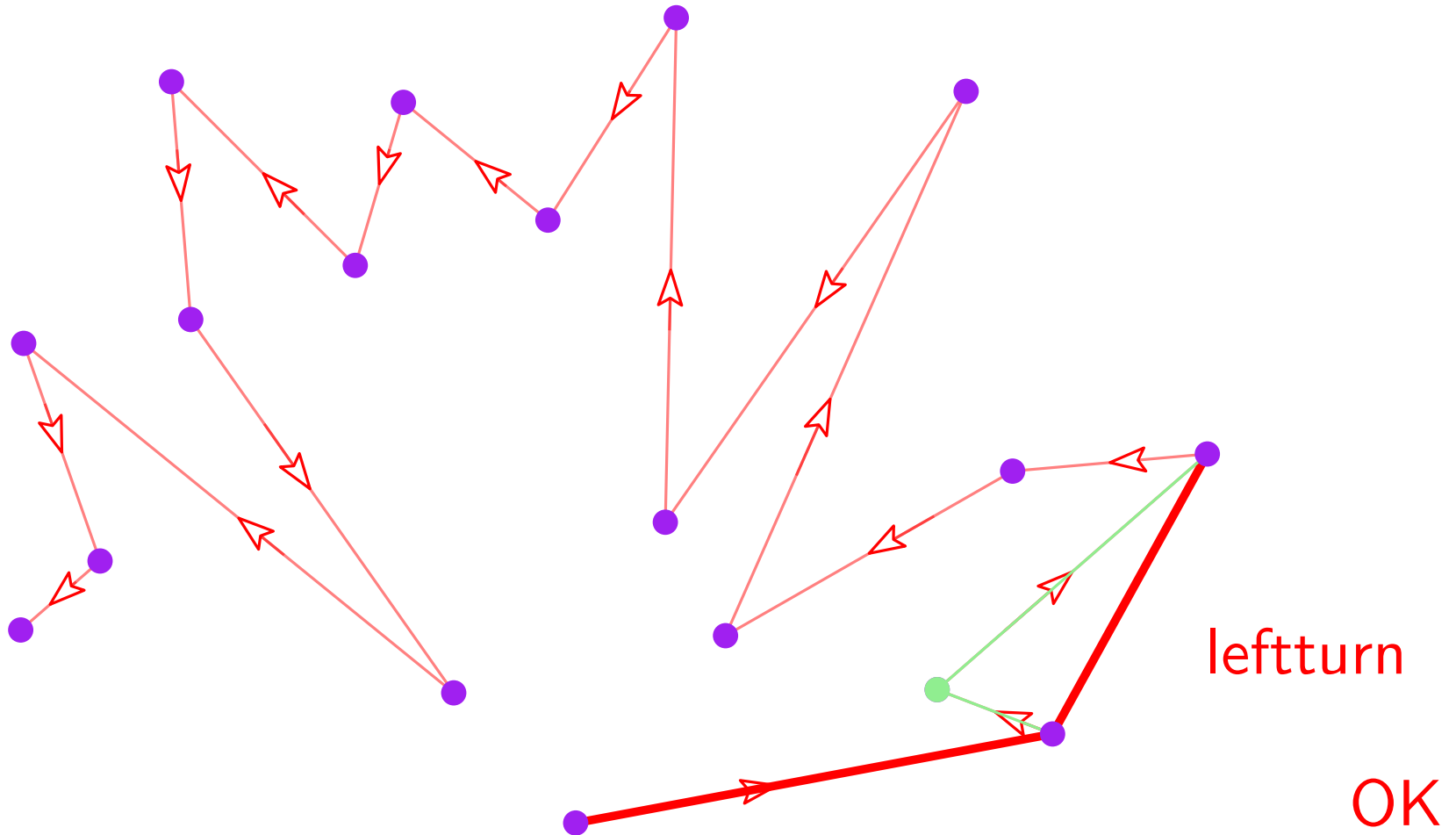
Convex hull

Graham algorithm



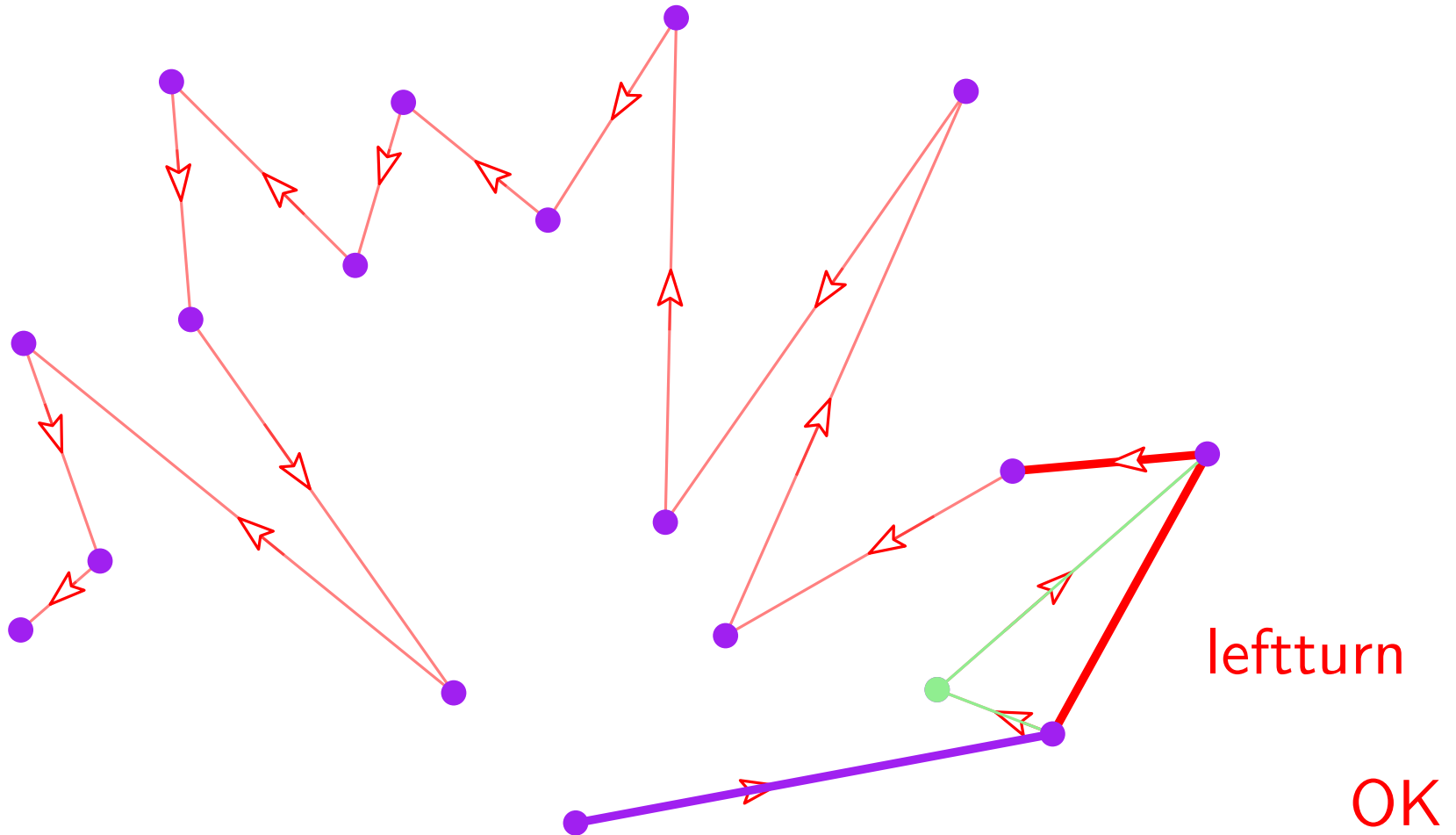
Convex hull

Graham algorithm



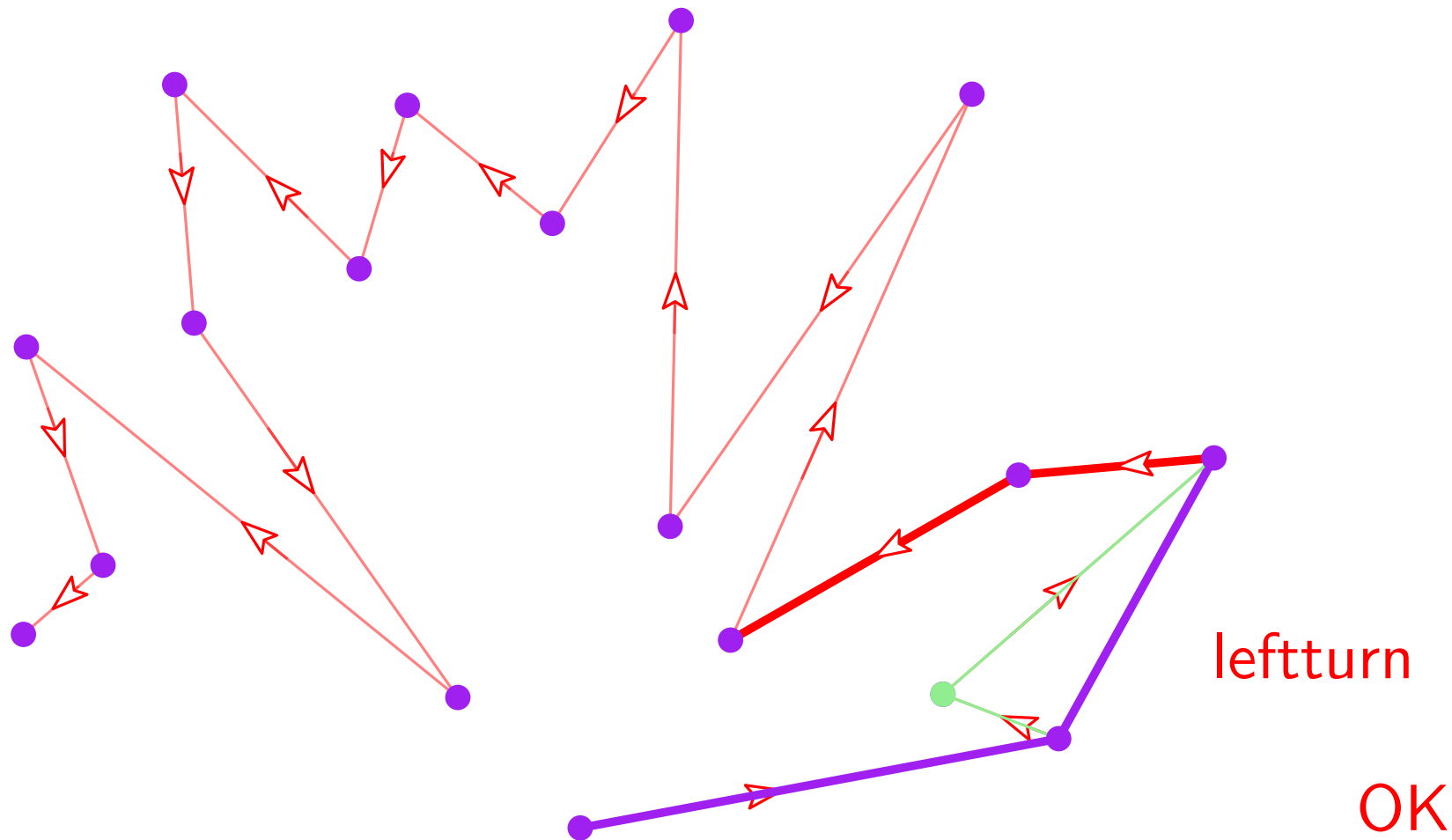
Convex hull

Graham algorithm



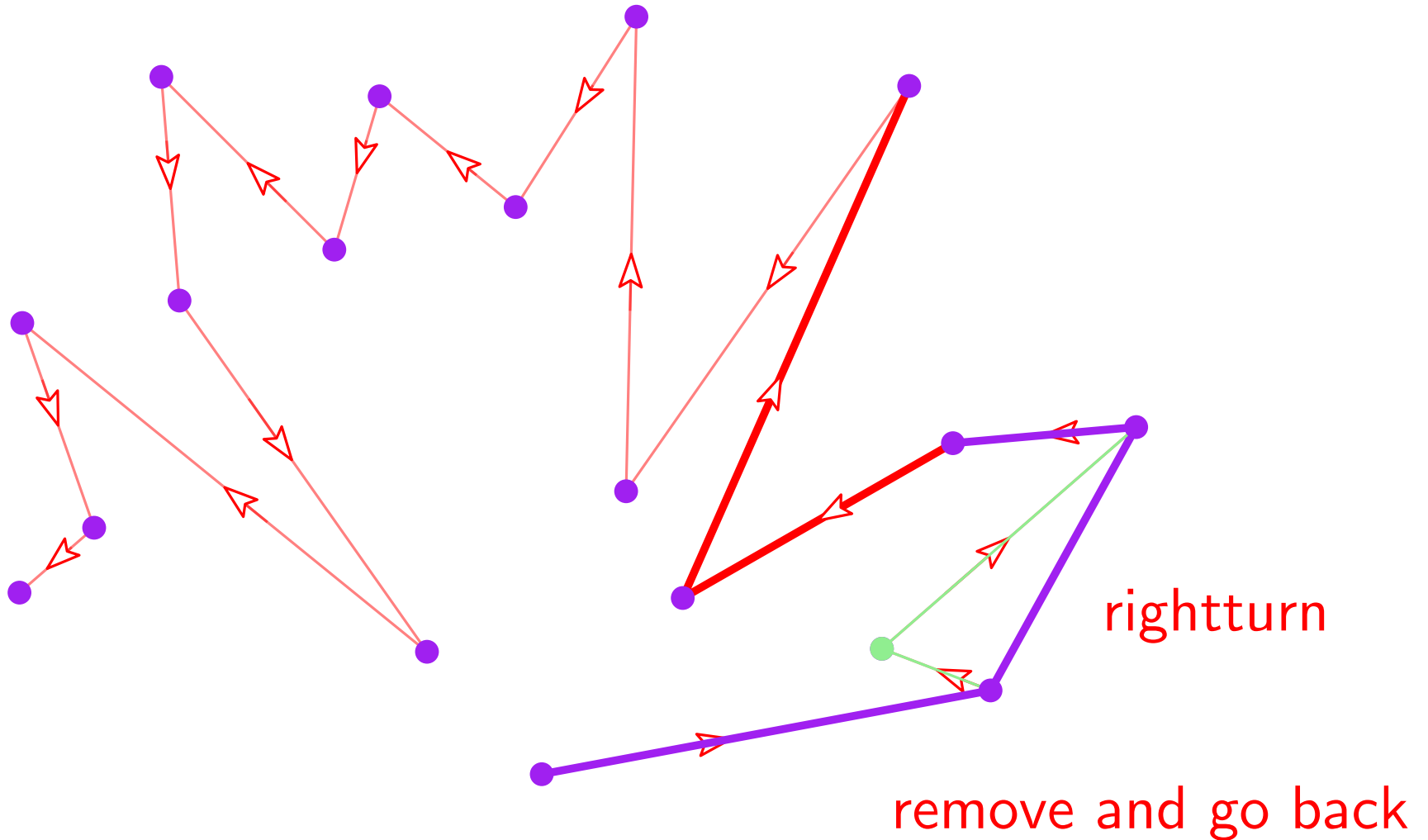
Convex hull

Graham algorithm



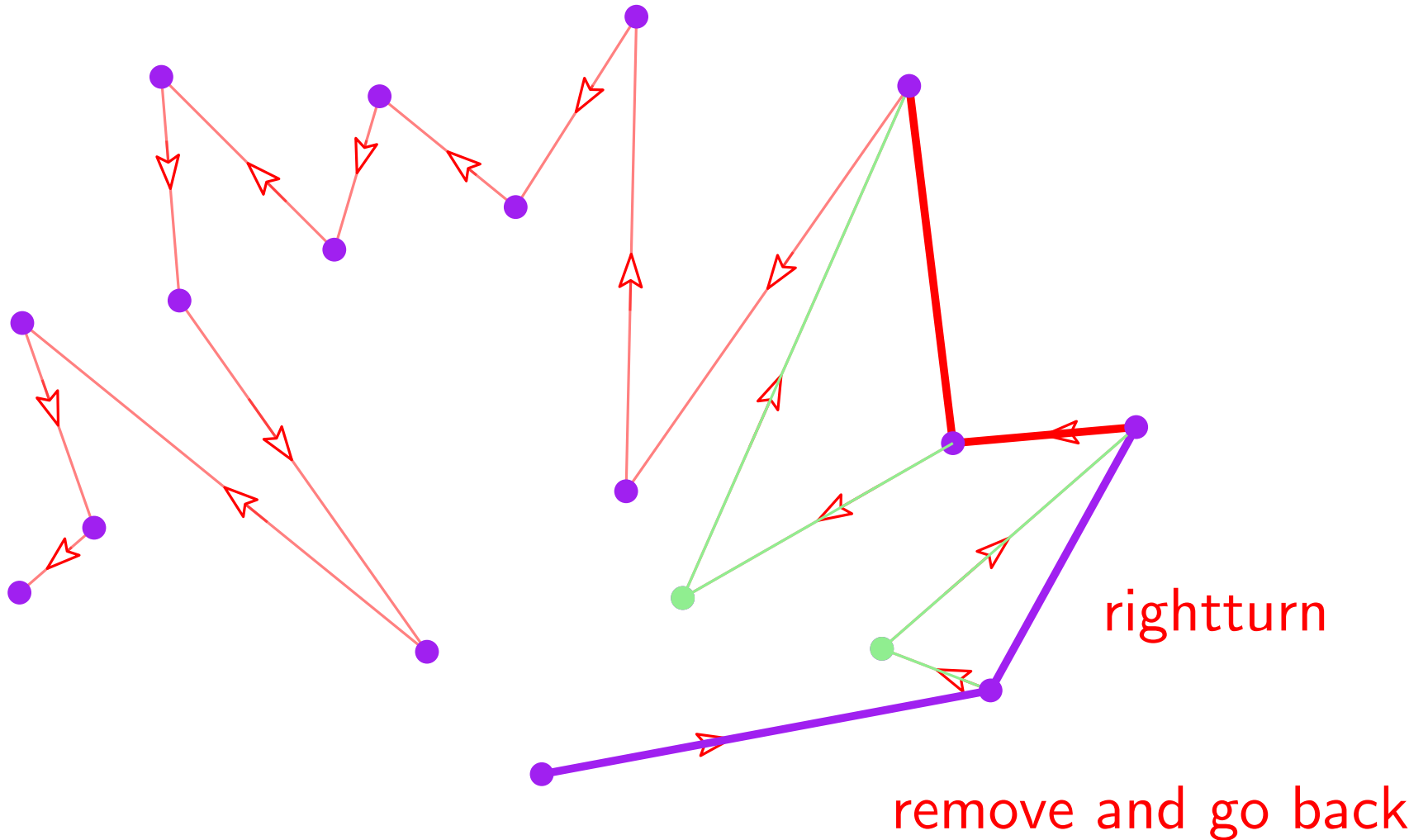
Convex hull

Graham algorithm



Convex hull

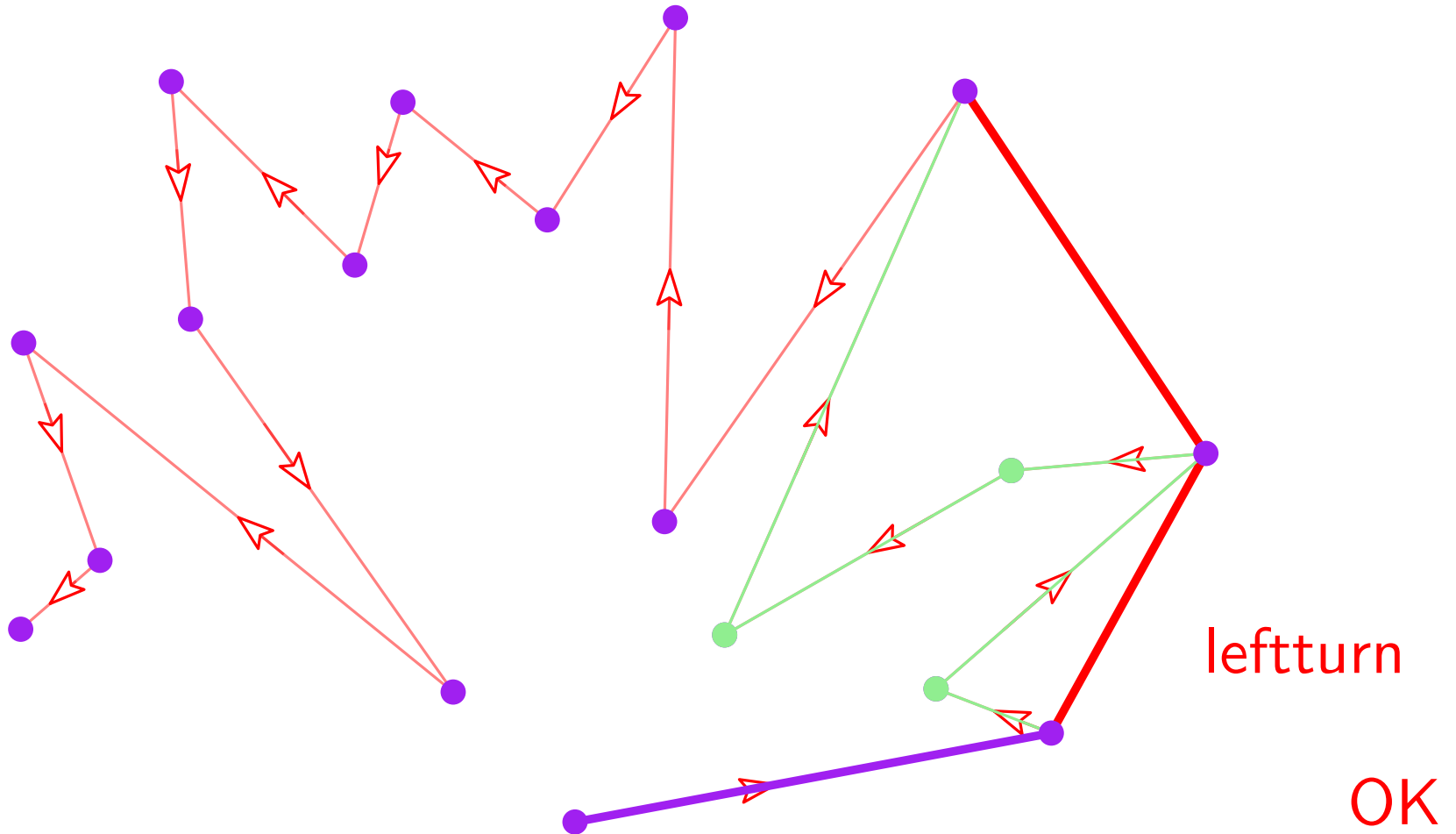
Graham algorithm



10 - 10

Convex hull

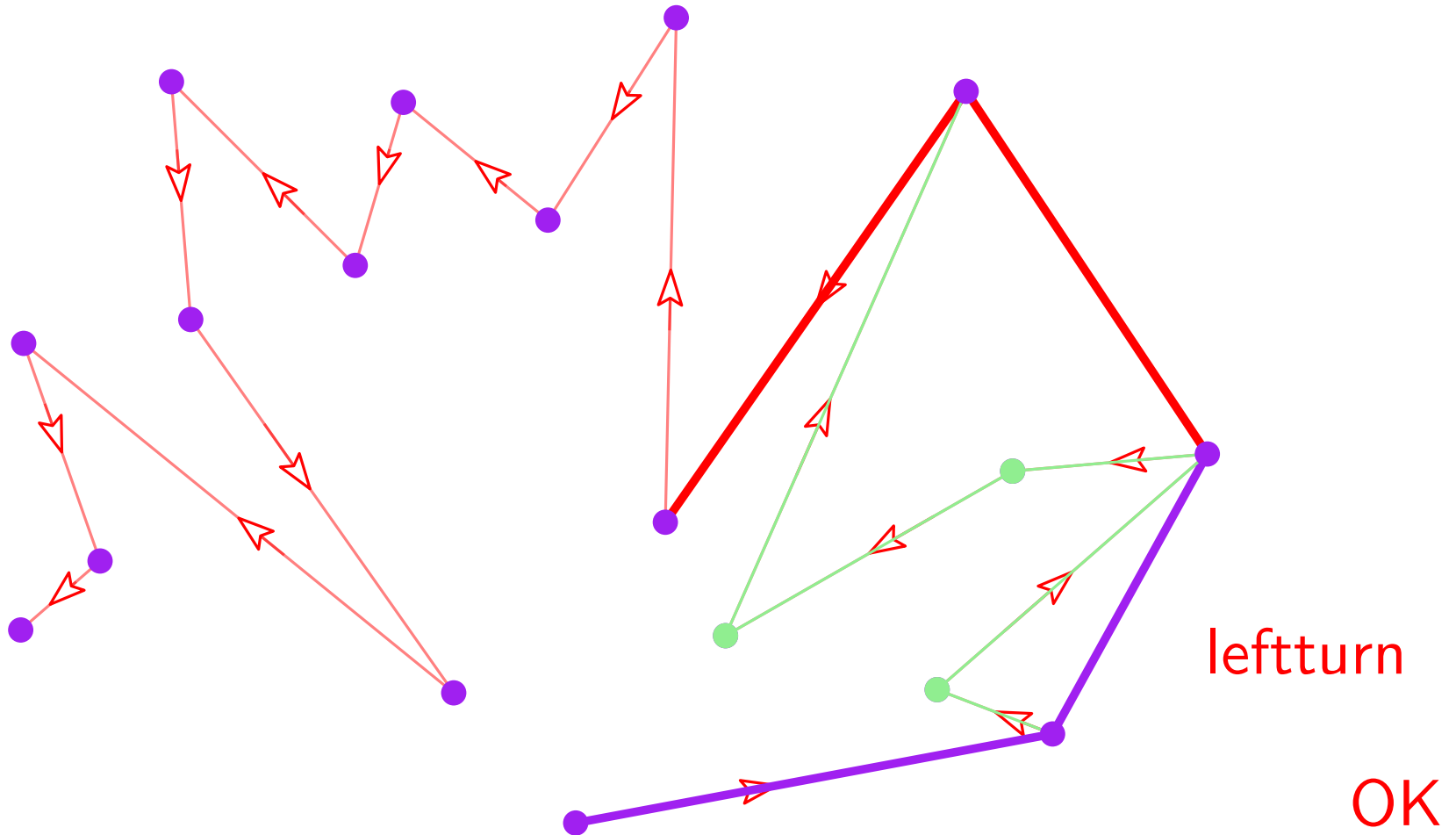
Graham algorithm



10 - 11

Convex hull

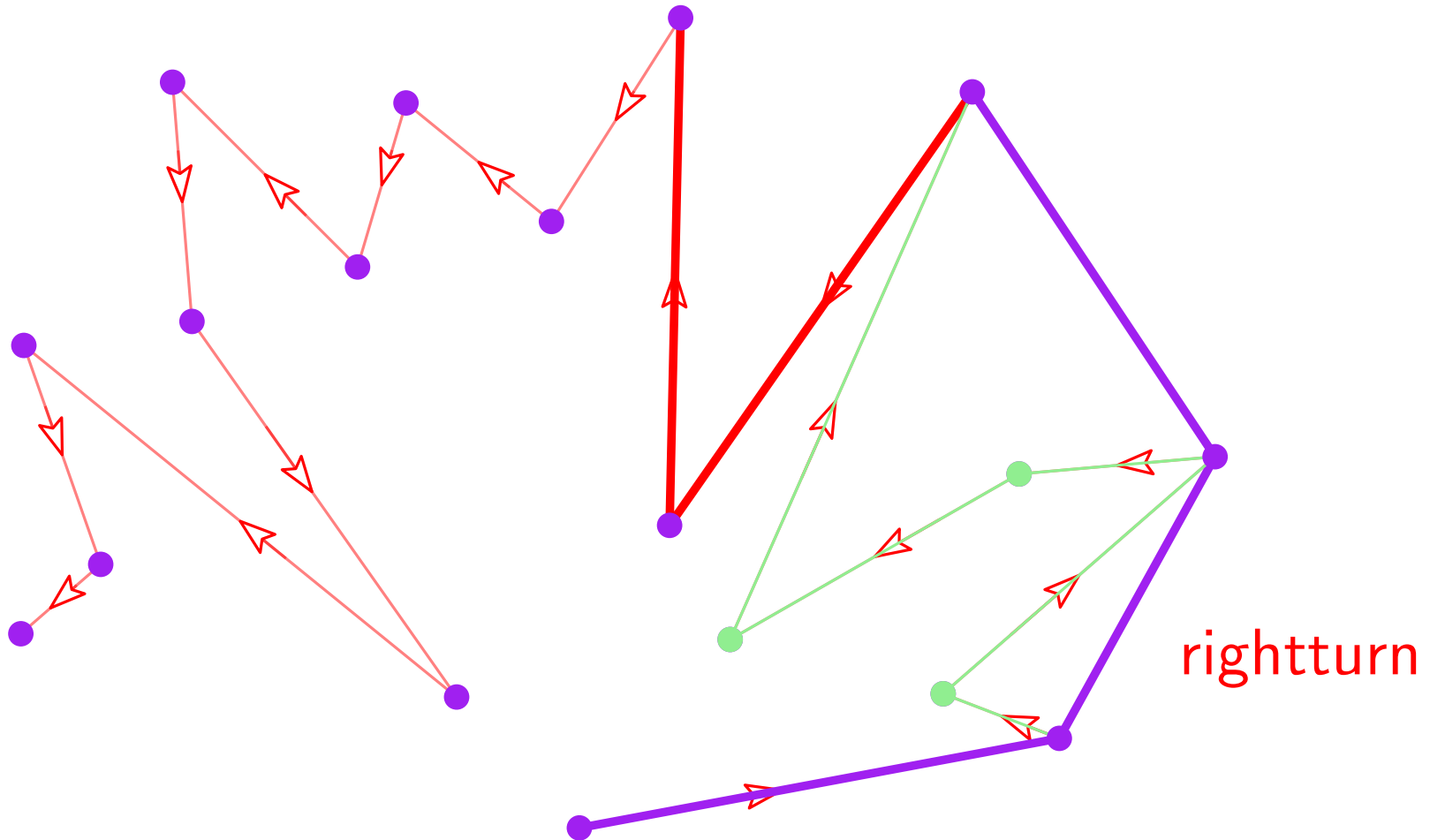
Graham algorithm



10 - 12

Convex hull

Graham algorithm

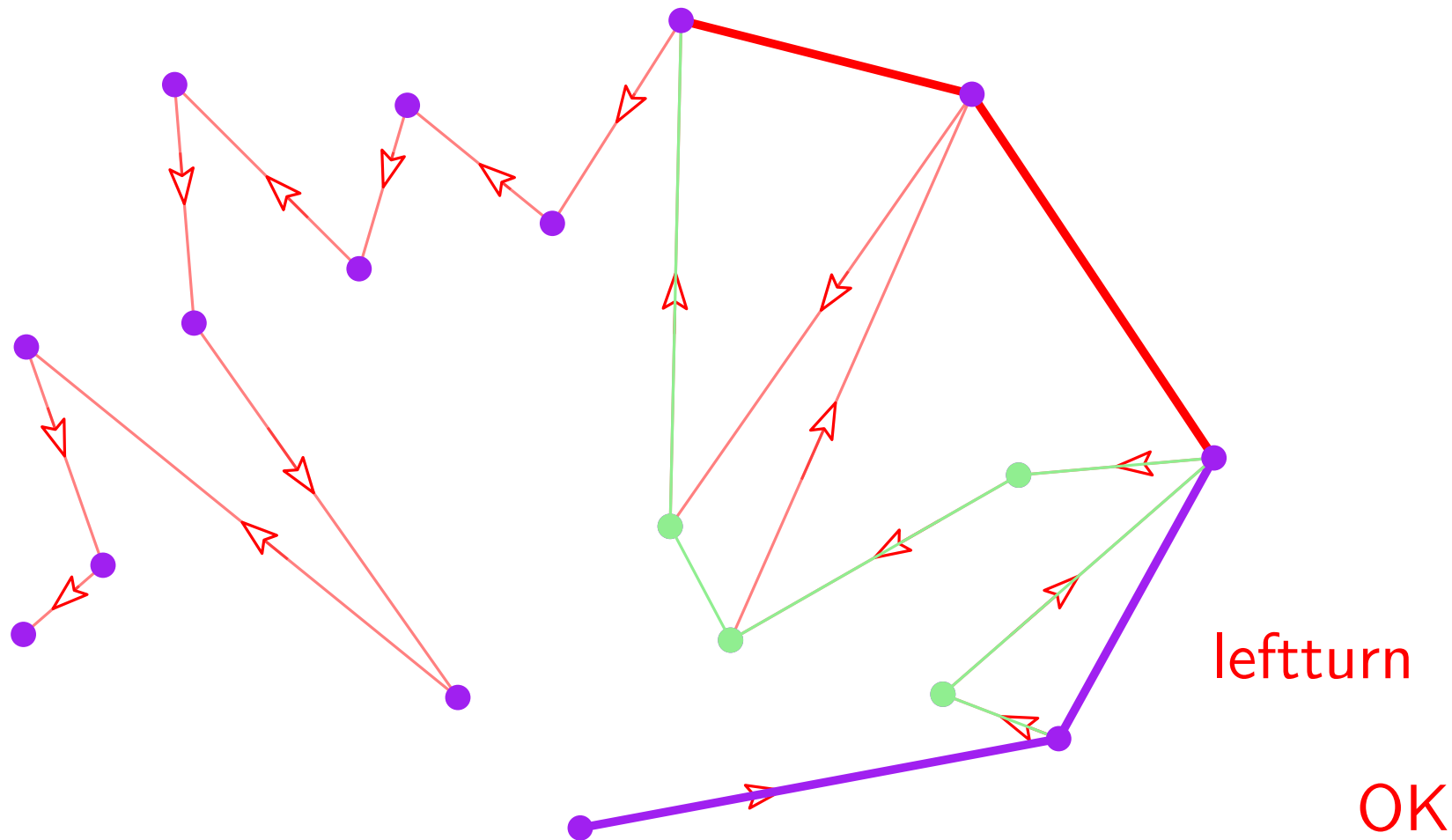


10 - 13

remove and go back

Convex hull

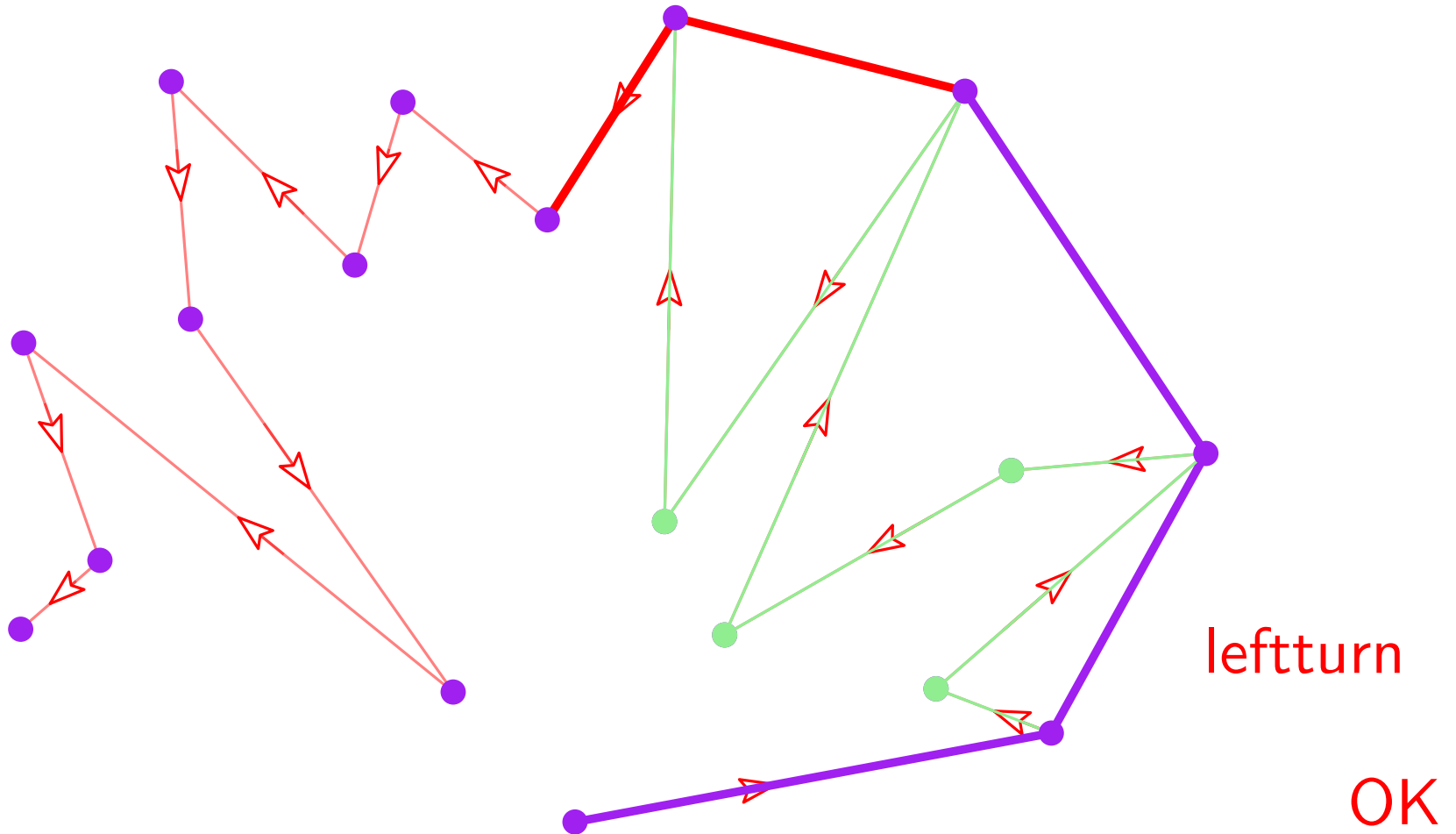
Graham algorithm



10 - 14

Convex hull

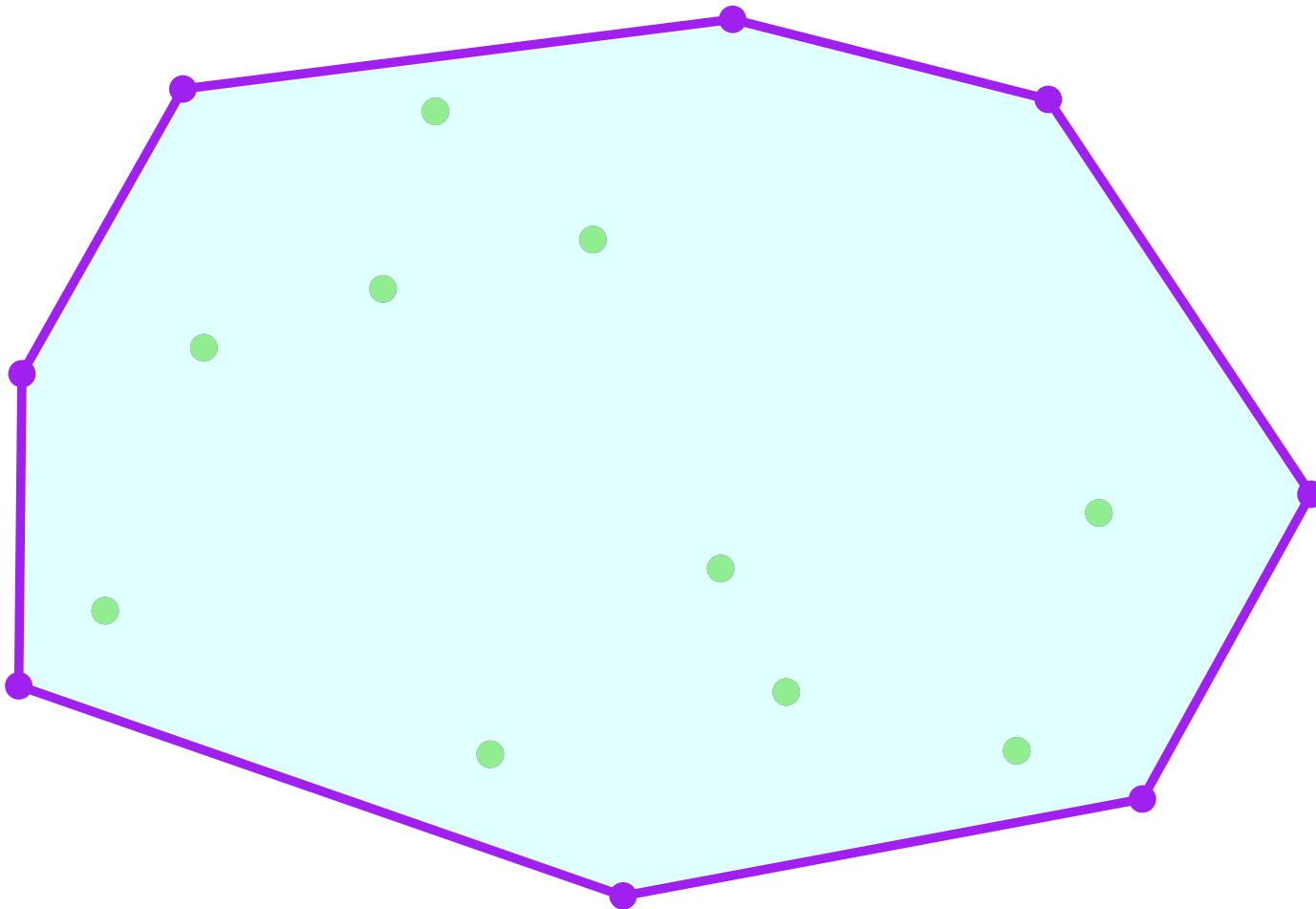
Graham algorithm



10 - 15

Convex hull

Graham algorithm



Convex hull

Complexity

Graham algorithm

Input: point set S

u lowest point of S ;

sort S around u in a circular list including u ;

$v = u$;

while $v.next \neq u$

 if $(v, v.next, v.next.next)$ ccw

$v = v.next$;

 else

$v.next = v.next.next$; $v.next.previous = v$;

 if $v \neq u$ $v = v.previous$;

Convex hull

Complexity

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$O(n)$

Convex hull

Complexity

Graham algorithm

Input: point set S

u lowest point of S ;

sort S around u in a circular list including u ;

$v = u$;

$O(n \log n)$

while $v.next \neq u$

if $(v, v.next, v.next.next)$ ccw

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delete one point
at most n times

Convex hull

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delete one point
at most n times

distance to u decreases
at most n times

Convex hull

Complexity

Graham algorithm

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sort S around u in a circular list including u ;
 $v = u$;

$O(n)$

while $v.next \neq u$

if $(v, v.next, v.next.next)$ ccw

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delete one point
at most n times

distance to u decreases
at most n times

Convex hull

Complexity

Graham algorithm

Input: point set S
 u lowest point of S ;
sort S around u in a circular list including u ;
 $v = u$;

$O(n \log n)$

```
while  $v.next \neq u$ 
  if  $(v, v.next, v.next.next)$  ccw
     $v = v.next$ ;
  else
     $v.next = v.next.next$ ;  $v.next.previous = v$ ;
    if  $v \neq u$   $v = v.previous$ ;
```

Convex hull

Lower bound

Problem lower bound is $\Omega(f(n))$

Iff there is **NO** algorithm

solving all size n problems

using less than $Cf(n)$ operations

$\forall n$

C constant independent of n

Sorting

Lower bound

Input: n real (positive) numbers

Sorting

Lower bound

Input: n real (positive) numbers

Output: sorting permutation

Sorting

Lower bound

Input: n real (positive) numbers



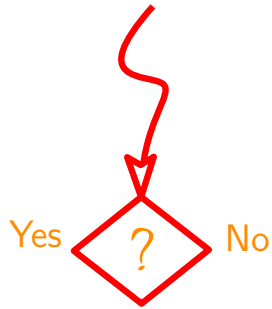
Output: sorting permutation

Monitoring execution

Sorting

Lower bound

Input: n real (positive) numbers



Output: sorting permutation

Monitoring execution

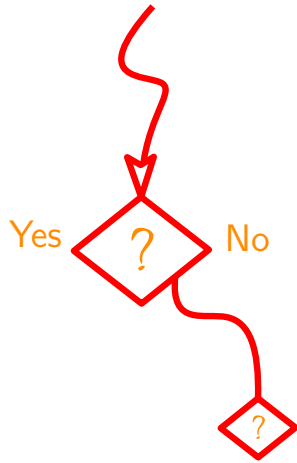
Sorting

Lower bound

Input: n real (positive) numbers

Output: sorting permutation

Monitoring execution



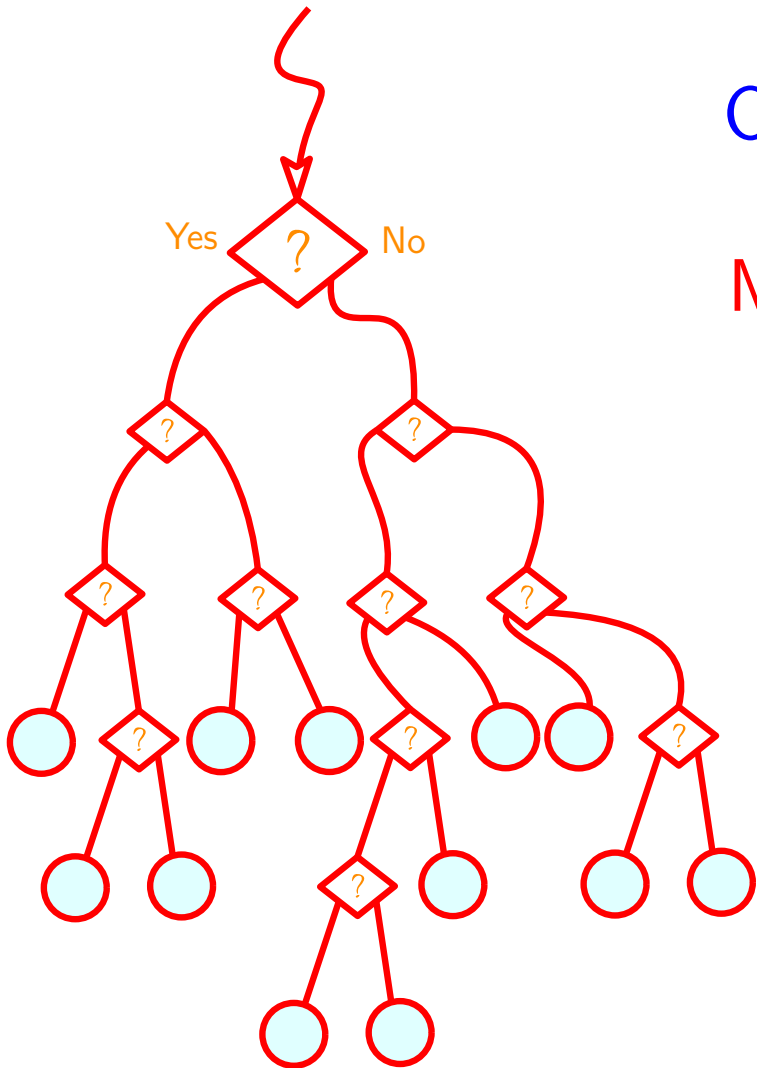
Sorting

Lower bound

Input: n real (positive) numbers

Output: sorting permutation

Monitoring execution



13 - 6

Sorting

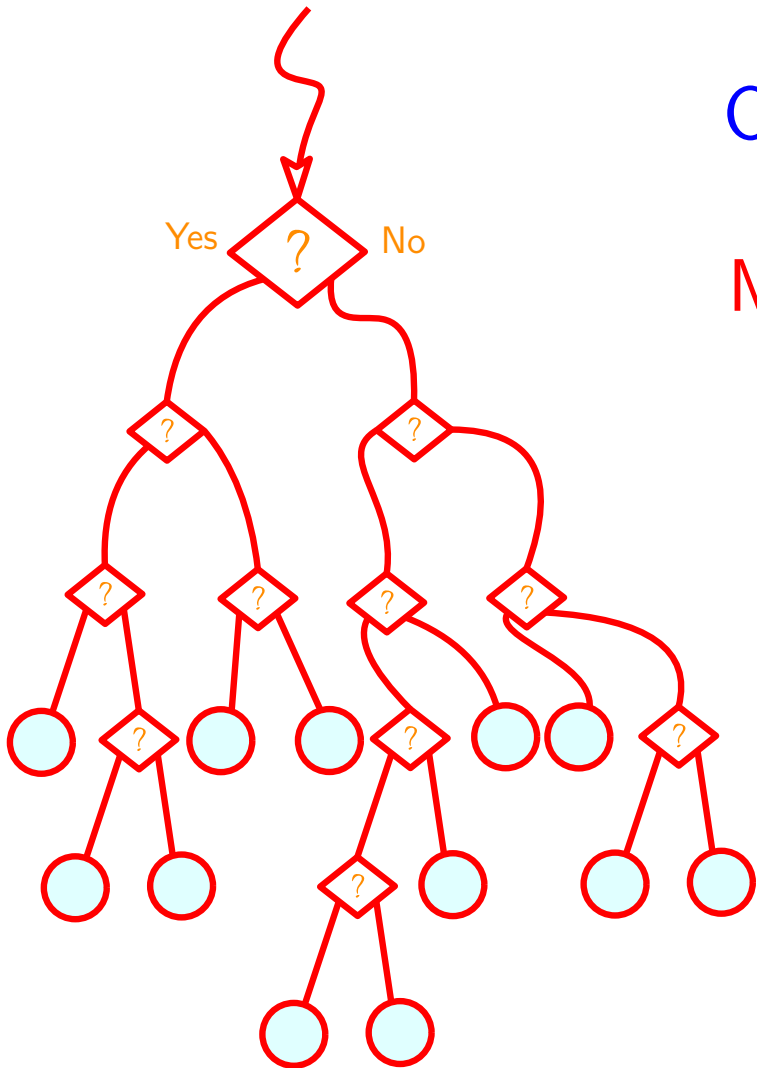
Lower bound

Input: n real (positive) numbers

Output: sorting permutation

Monitoring execution

leaves \geq # permutations



Sorting

Lower bound

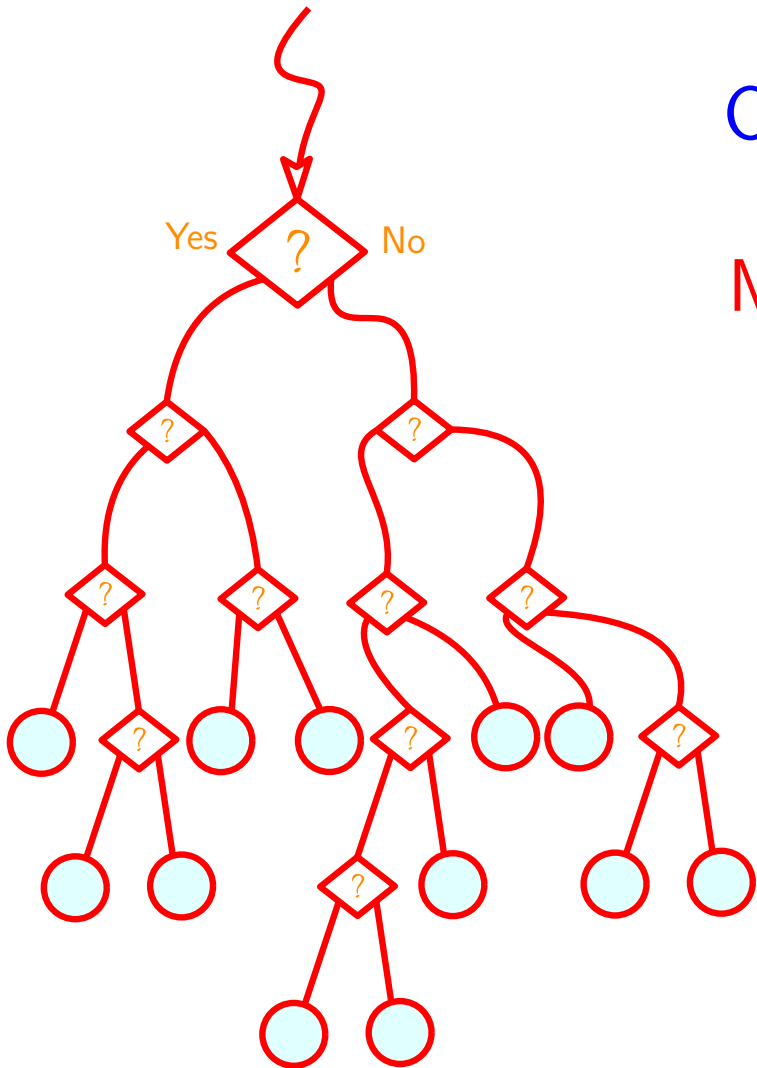
Input: n real (positive) numbers

Output: sorting permutation

Monitoring execution

leaves \geq # permutations

There are $n!$ permutations



Sorting

Lower bound

Input: n real (positive) numbers

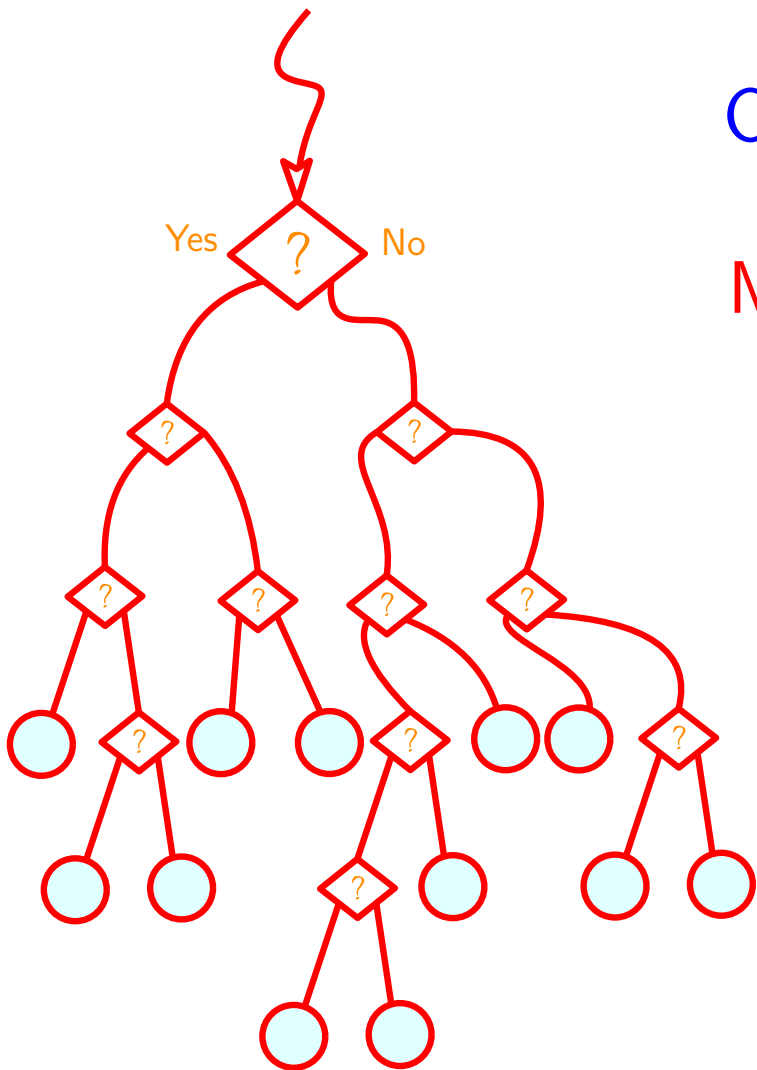
Output: sorting permutation

Monitoring execution

leaves \geq # permutations

There are $n!$ permutations

Tree height is at least \log_2 # leaves



Sorting

Lower bound

Input: n real (positive) numbers

Output: sorting permutation

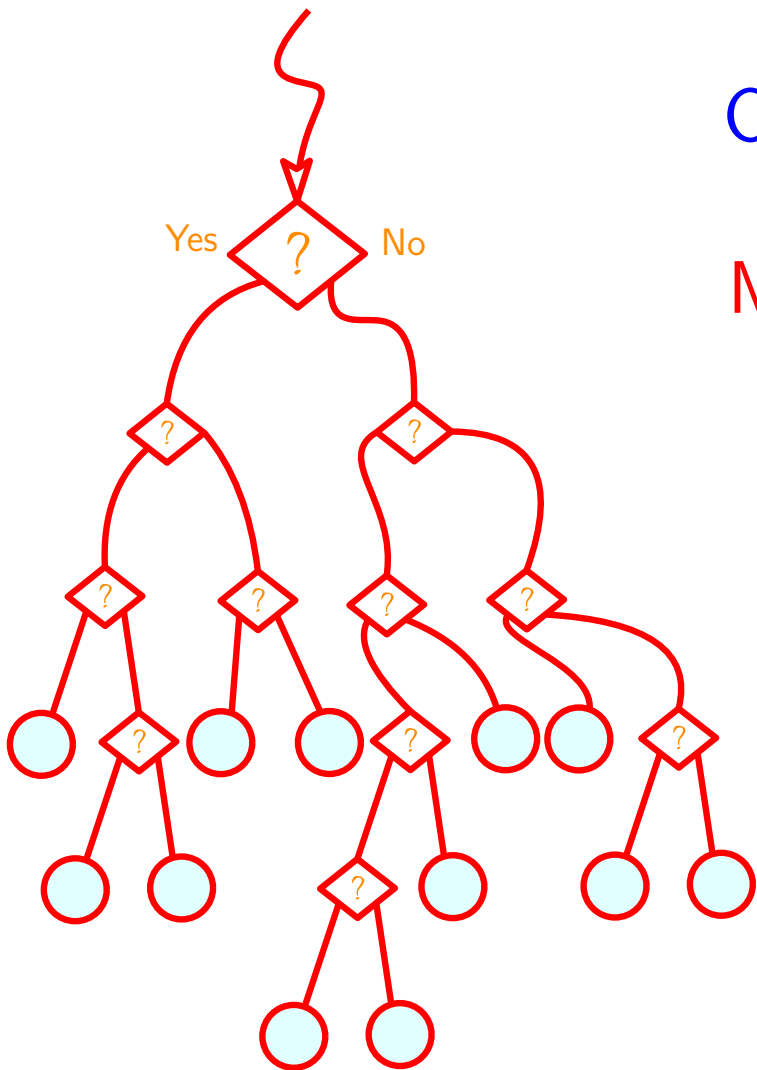
Monitoring execution

leaves \geq # permutations

There are $n!$ permutations

Tree height is at least \log_2 # leaves

comparisons $\leq \log_2 n! \simeq n \log_2 n$



Convex hull

Lower bound

A stupid algorithm for sorting numbers



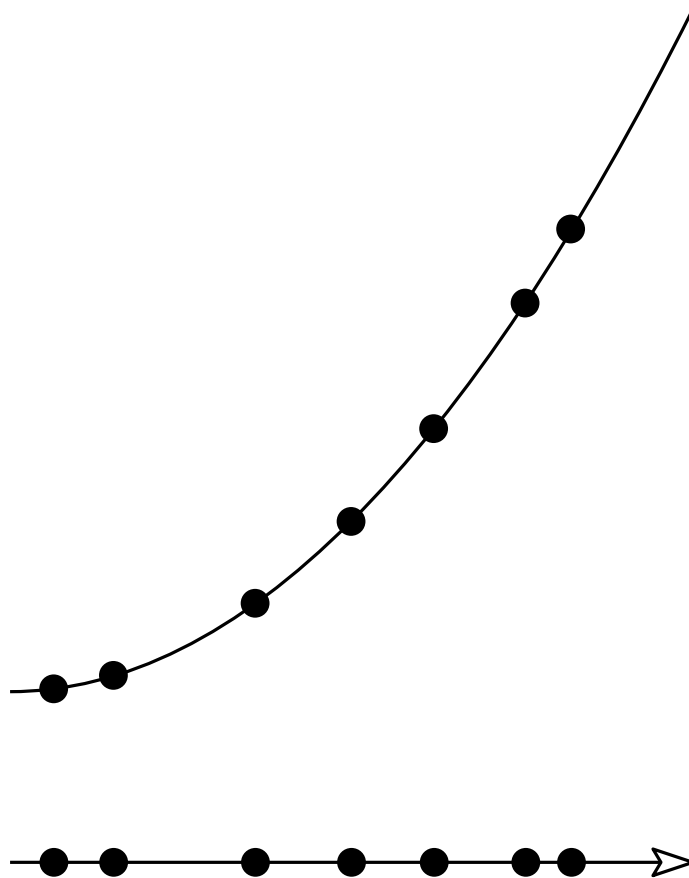
14 - 1

Convex hull

Lower bound

A stupid algorithm for sorting numbers

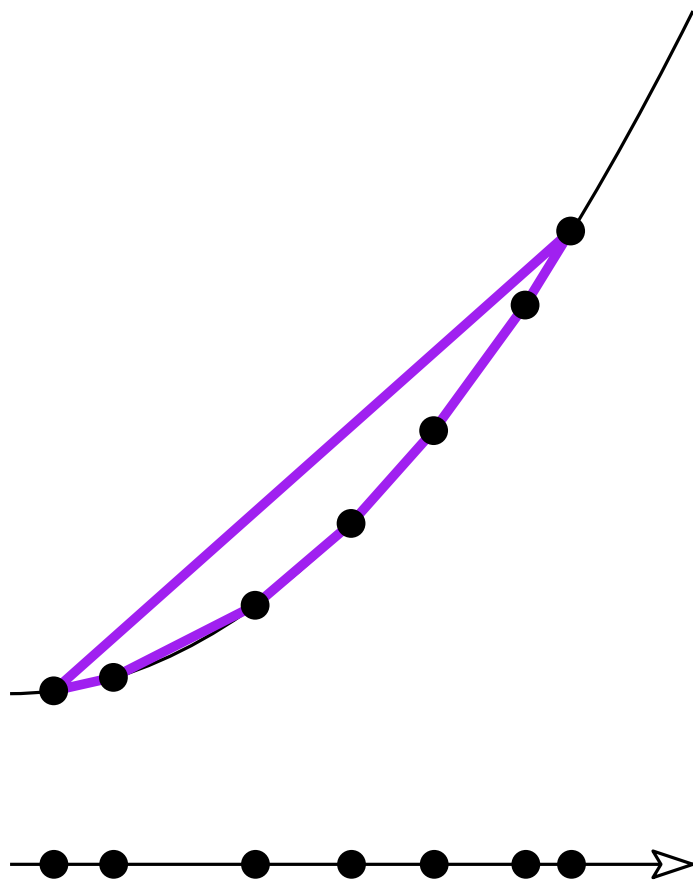
project on parabola



Convex hull

Lower bound

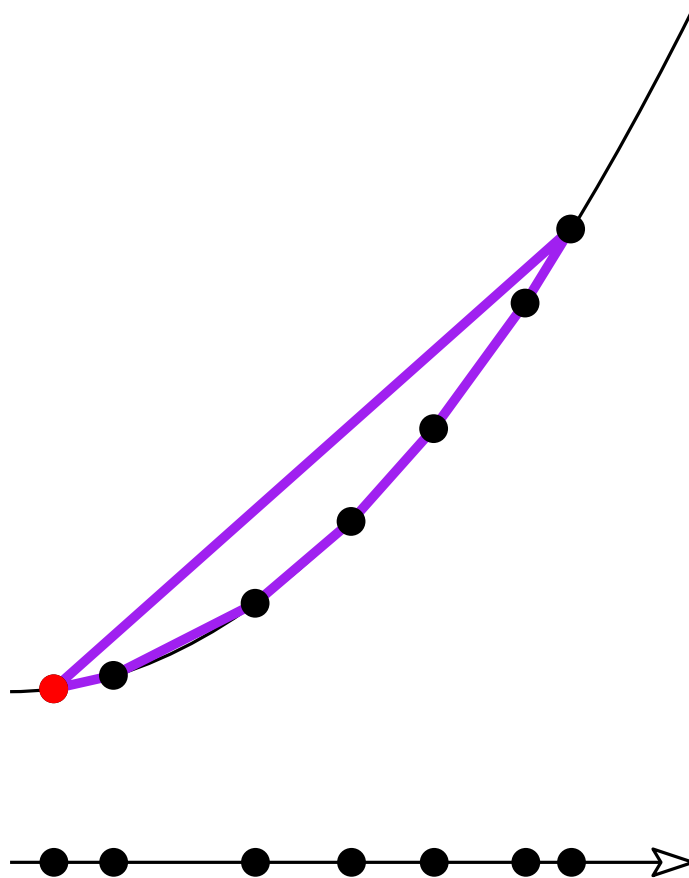
A stupid algorithm for sorting numbers



Convex hull

Lower bound

A stupid algorithm for sorting numbers



project on parabola

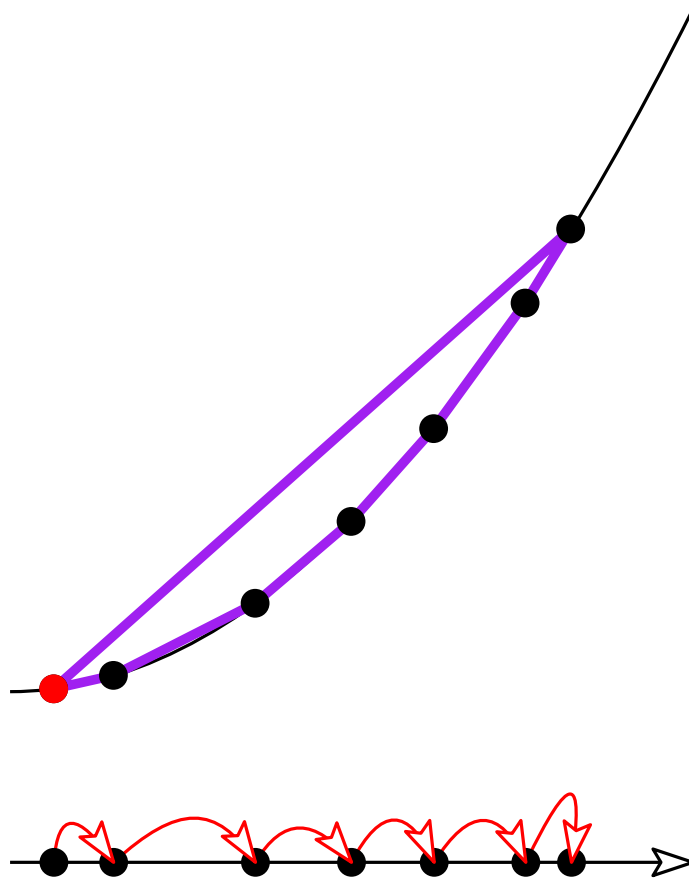
compute convex hull

find lowest point

Convex hull

Lower bound

A stupid algorithm for sorting numbers



project on parabola

compute convex hull

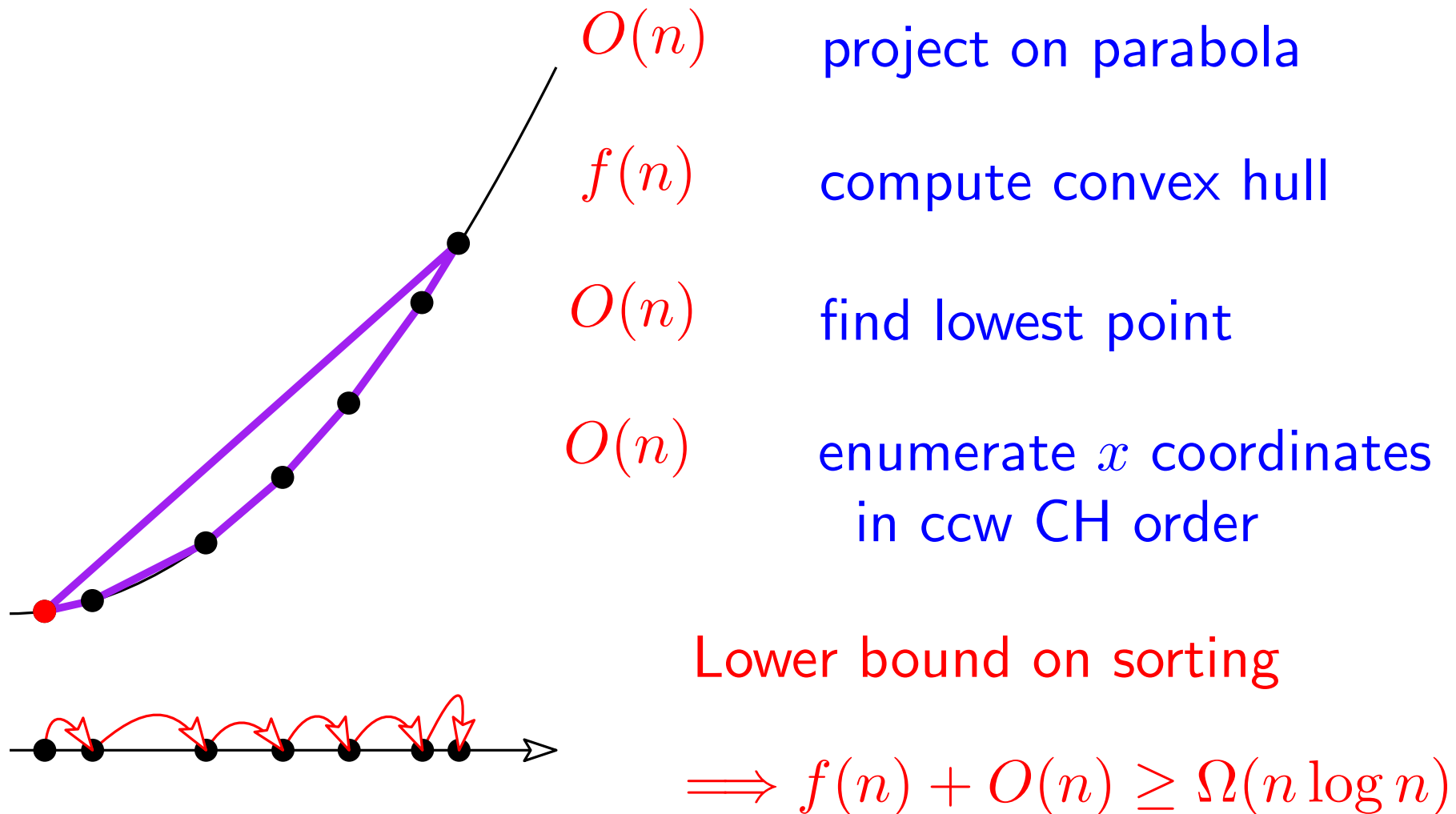
find lowest point

enumerate x coordinates
in ccw CH order

Convex hull

Lower bound

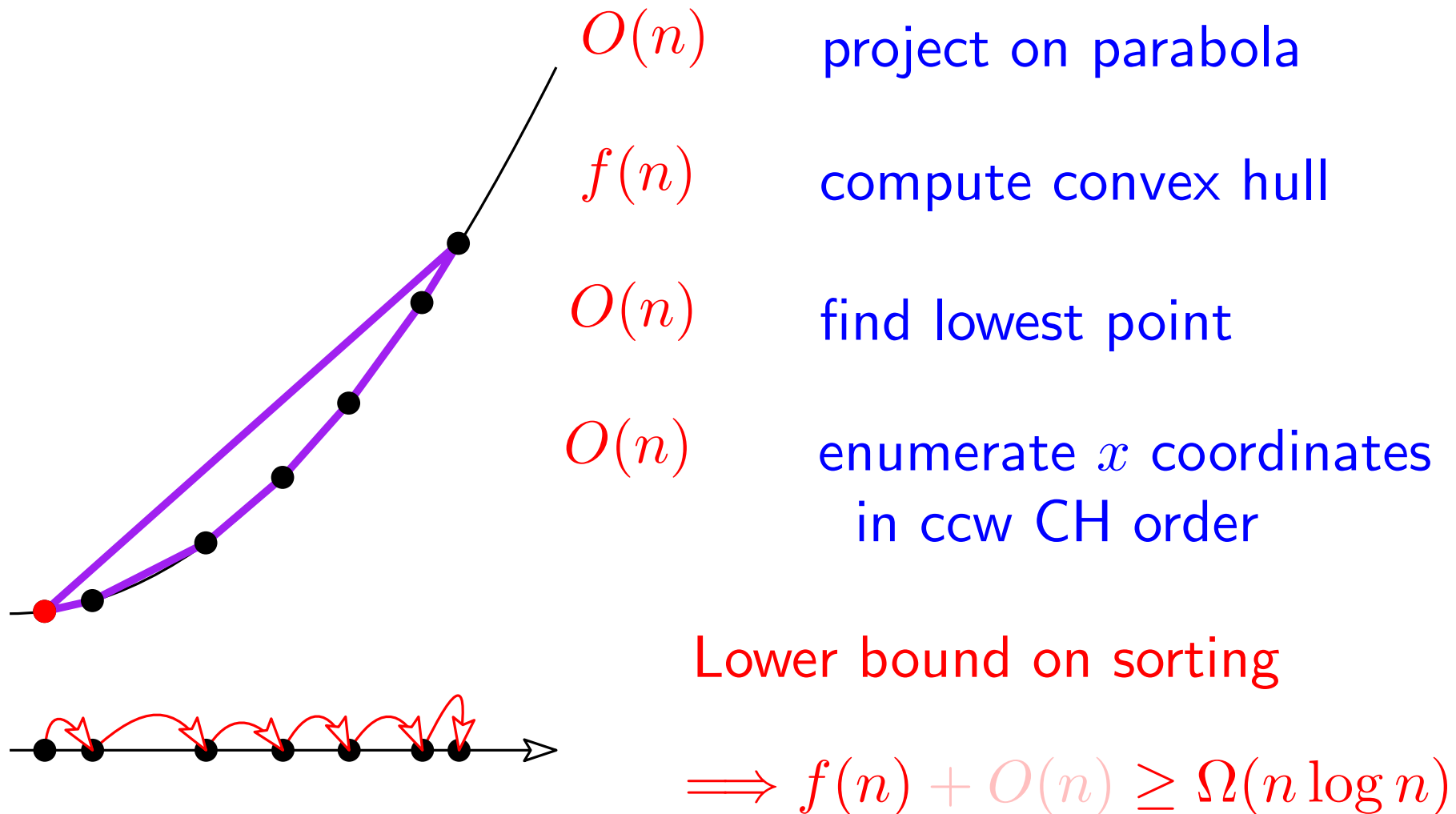
A stupid algorithm for sorting numbers



Convex hull

Lower bound

A stupid algorithm for sorting numbers



Convex hull

Other results

Expected size of the convex hull

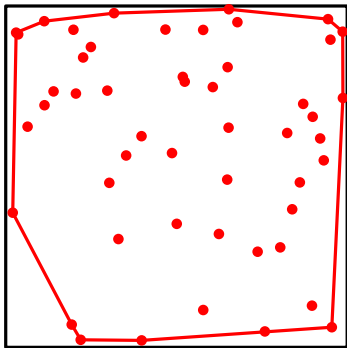
Convex hull

Expected size of the convex hull

n random points in a square

Other results
expected

$\Theta(\log n)$



Convex hull

Expected size of the convex hull

n random points in a square

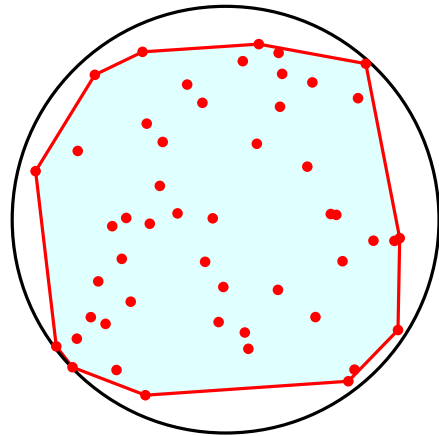
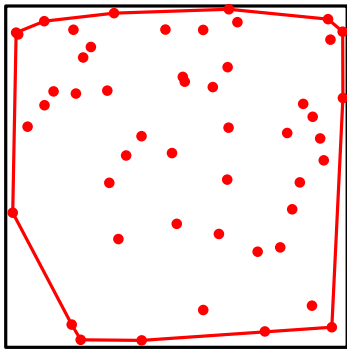
n random points in a disk

Other results

expected

$\Theta(\log n)$

$\Theta(n^{\frac{1}{3}})$



Convex hull

Other results

expected

Expected size of the convex hull

n random points in a square

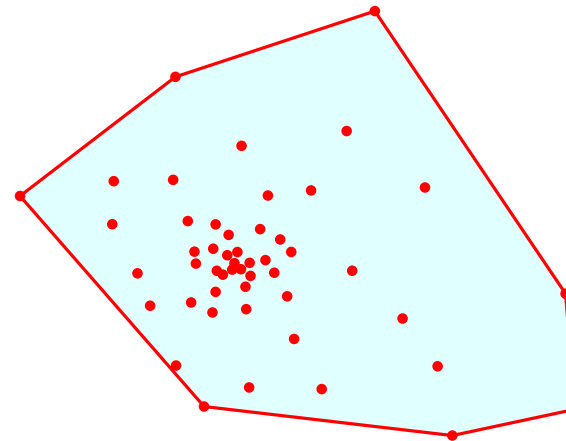
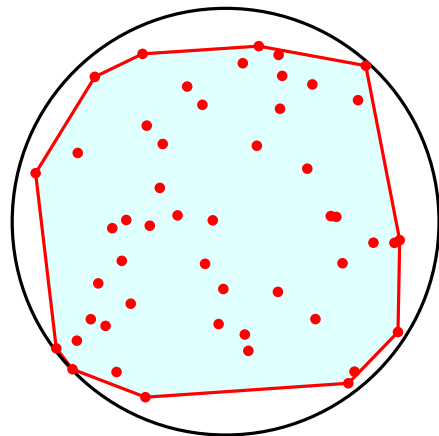
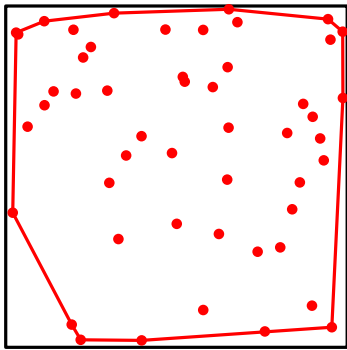
$$\Theta(\log n)$$

n random points in a disk

$$\Theta(n^{\frac{1}{3}})$$

n random points, Gaussian distribution

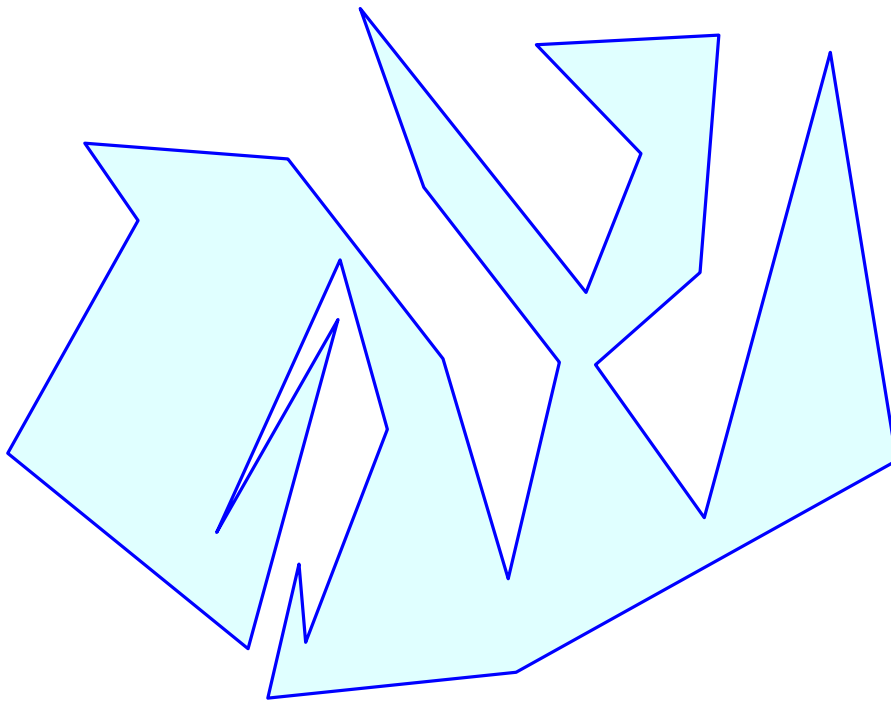
$$\Theta(\sqrt{\log n})$$



Convex hull

Other results

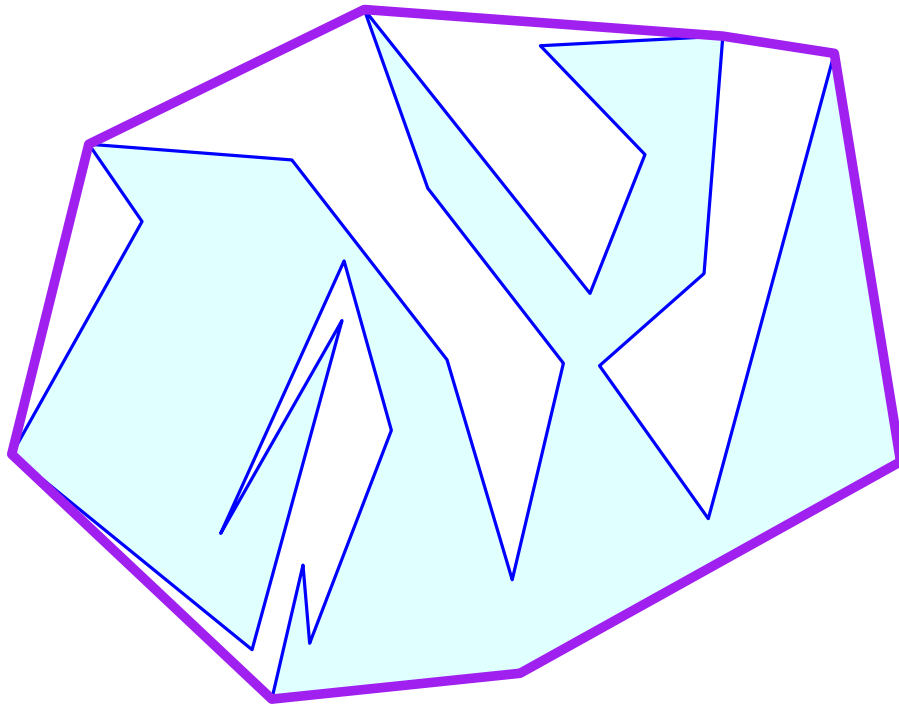
Simple polygon



Convex hull

Other results

Simple polygon

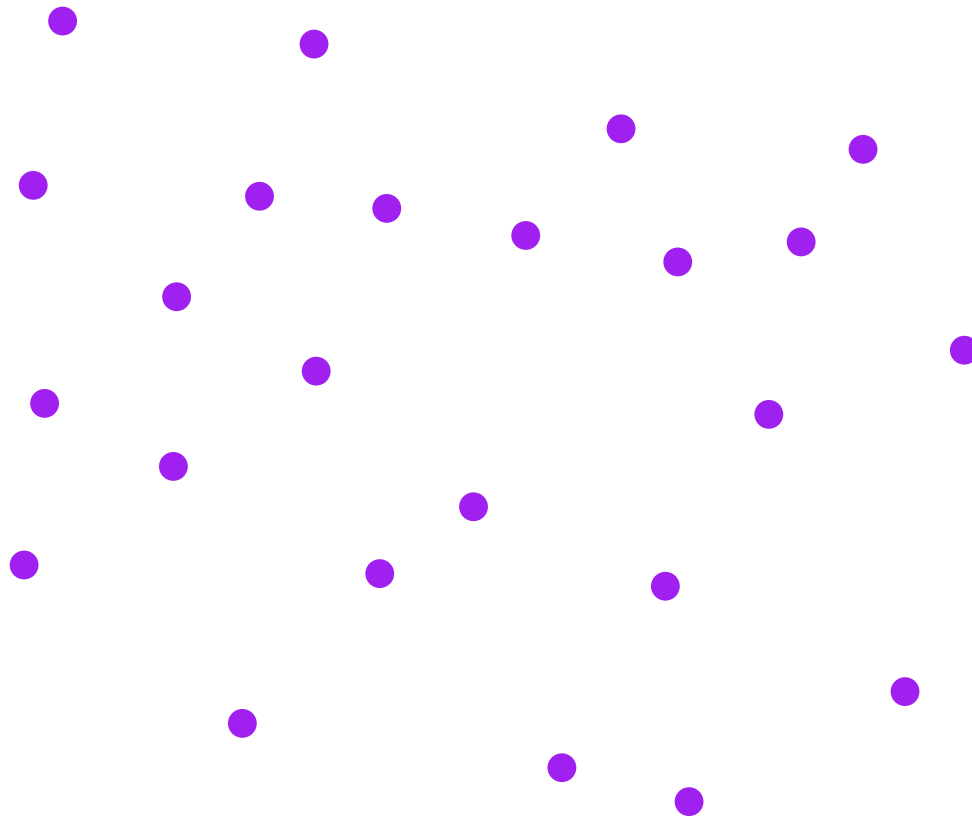


$O(n)$

Convex hull

Other results

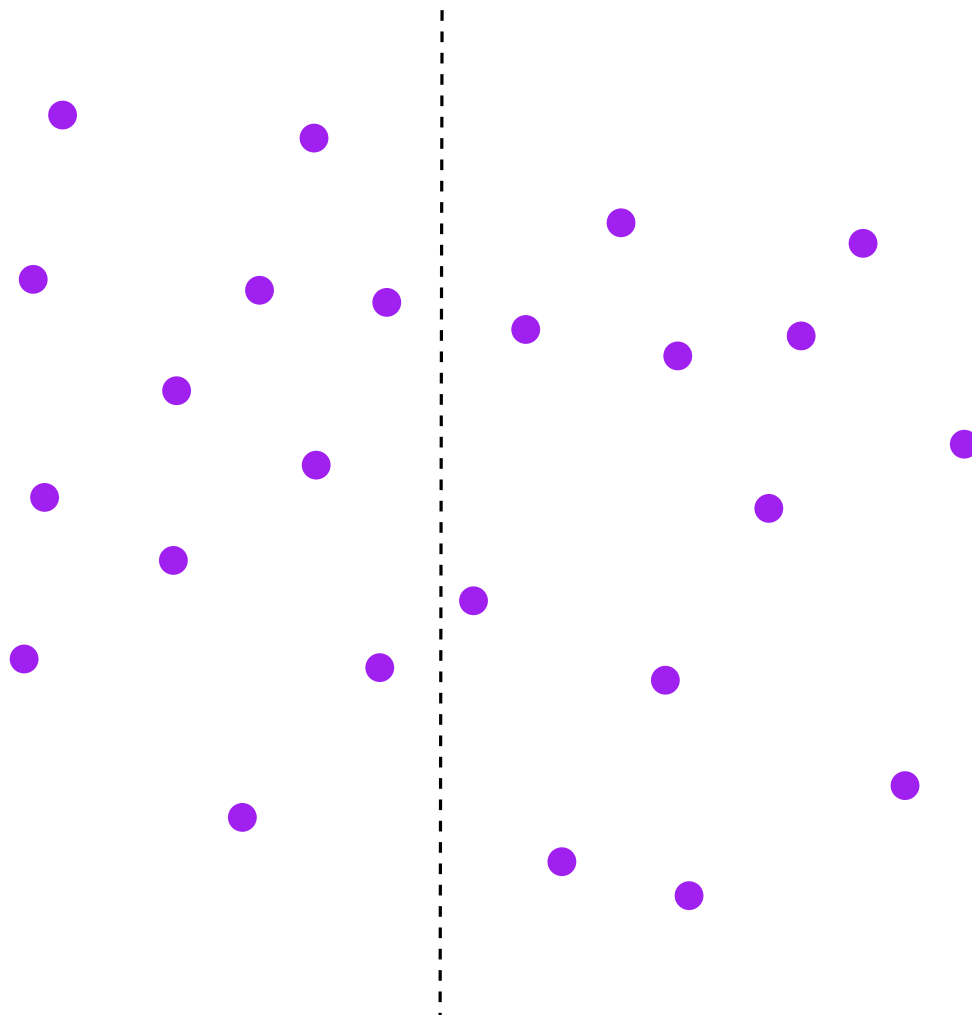
Divide and conquer



Convex hull

Other results

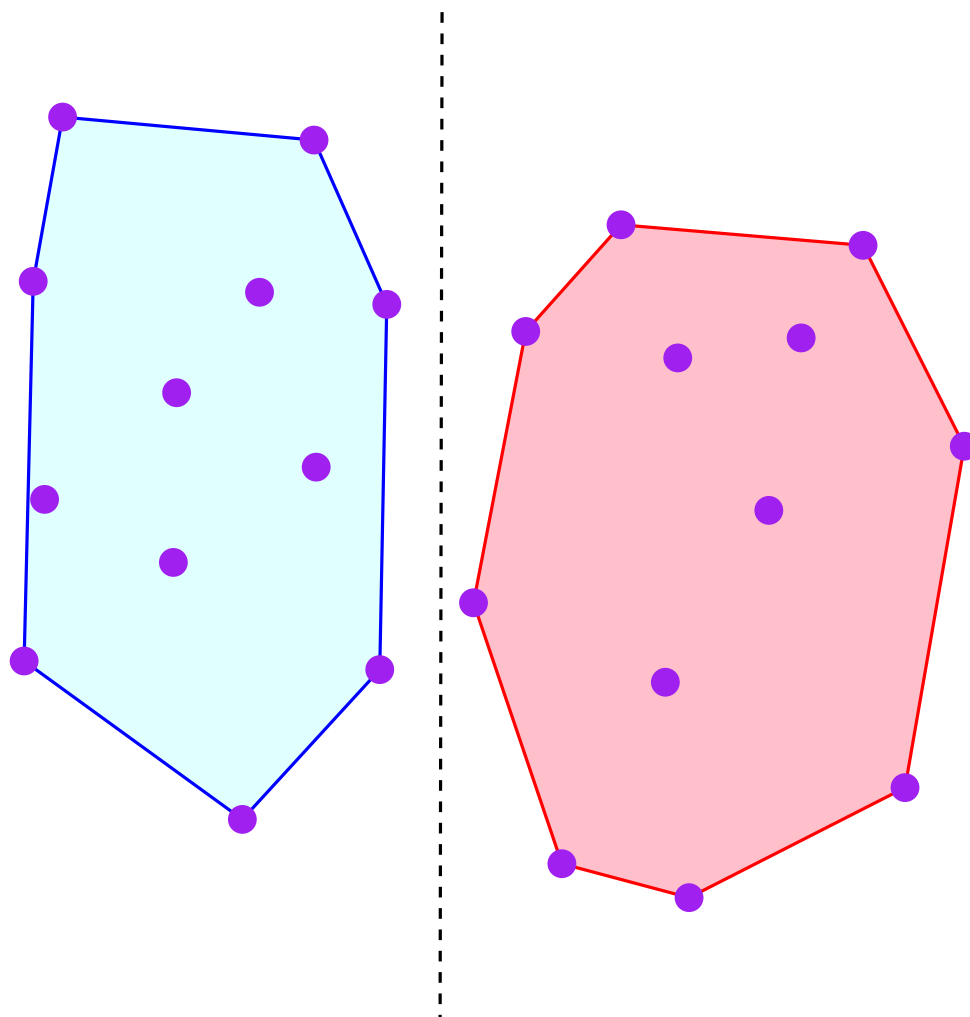
Divide and conquer



Convex hull

Other results

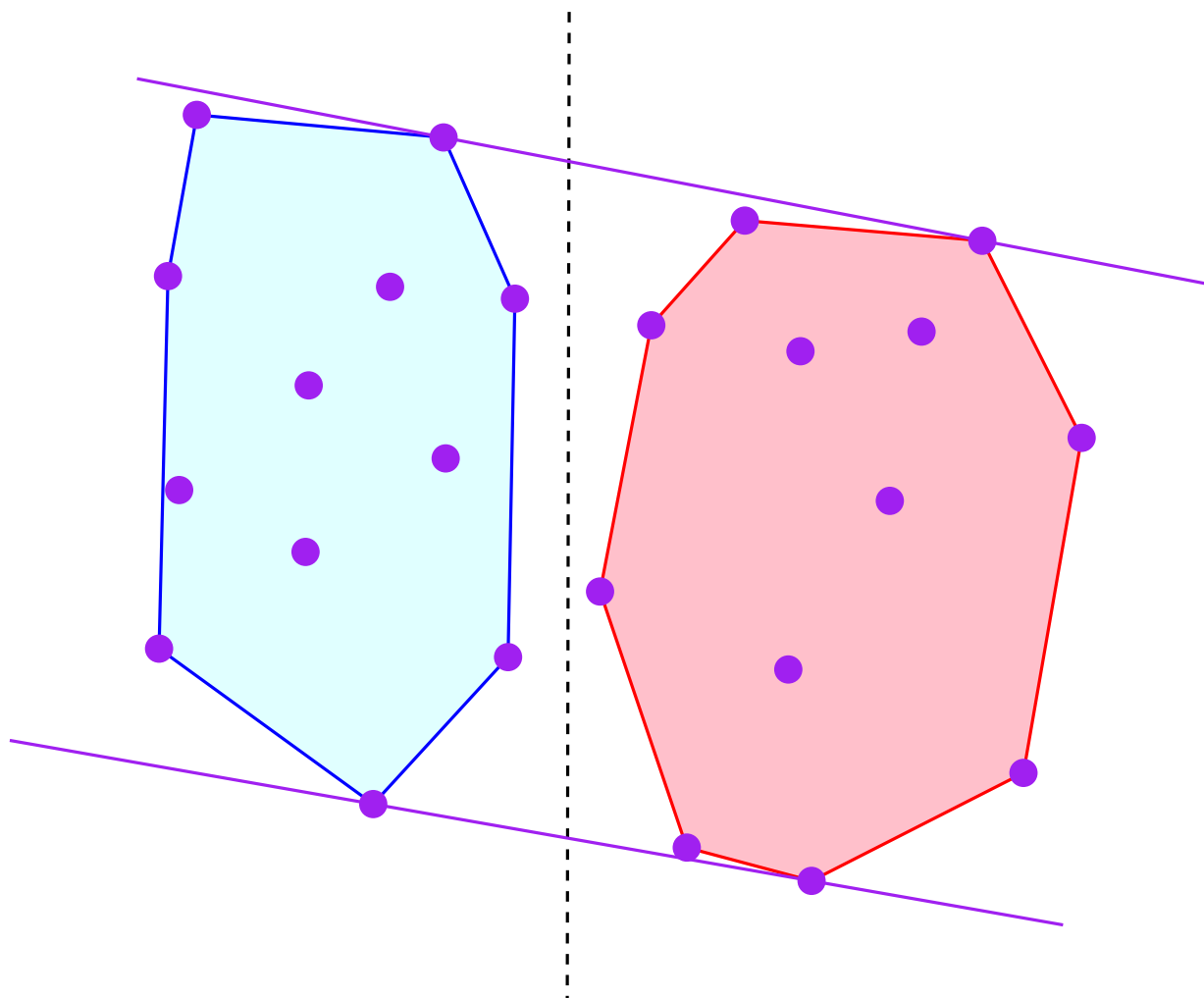
Divide and conquer



Convex hull

Other results

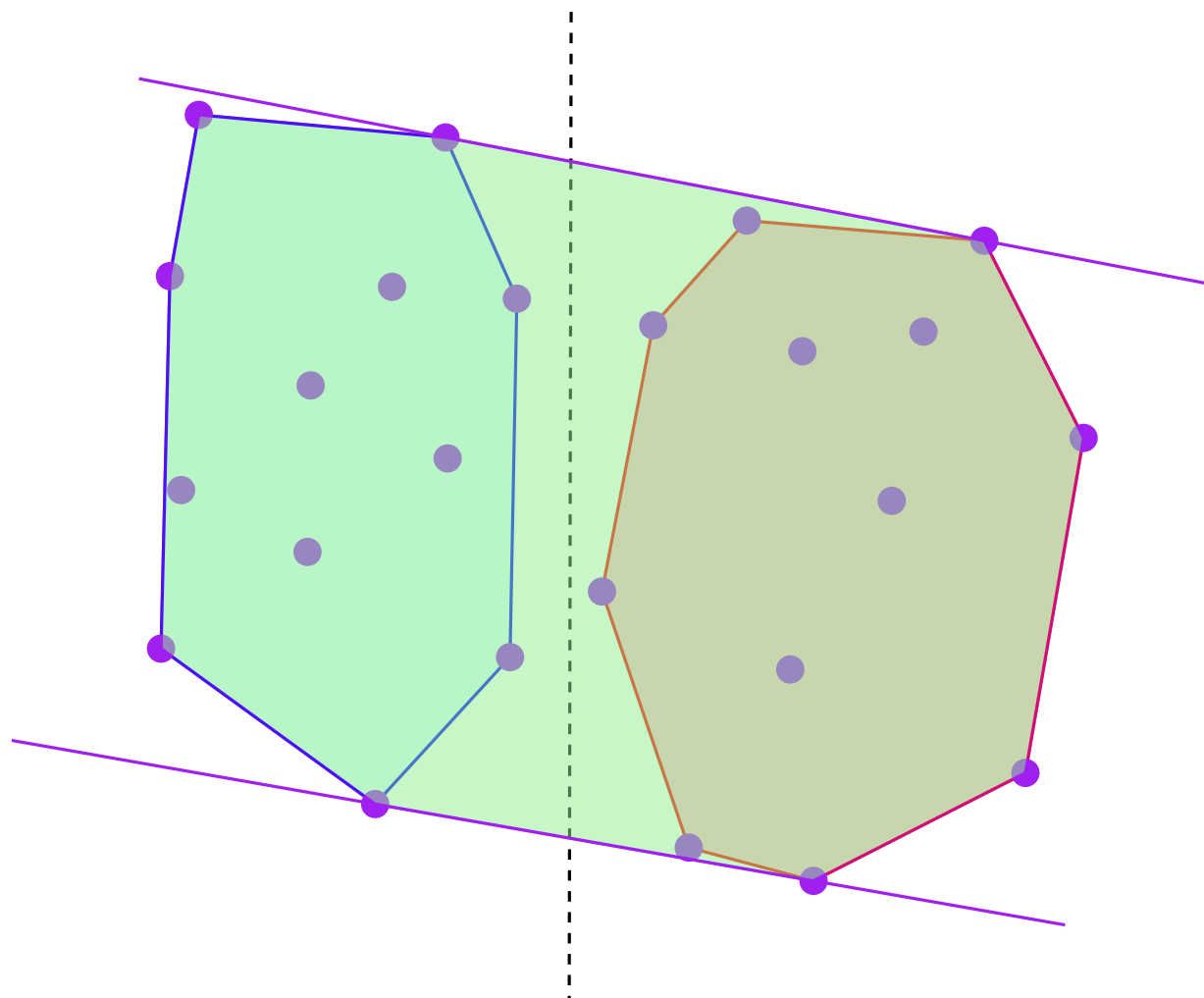
Divide and conquer



Convex hull

Other results

Divide and conquer

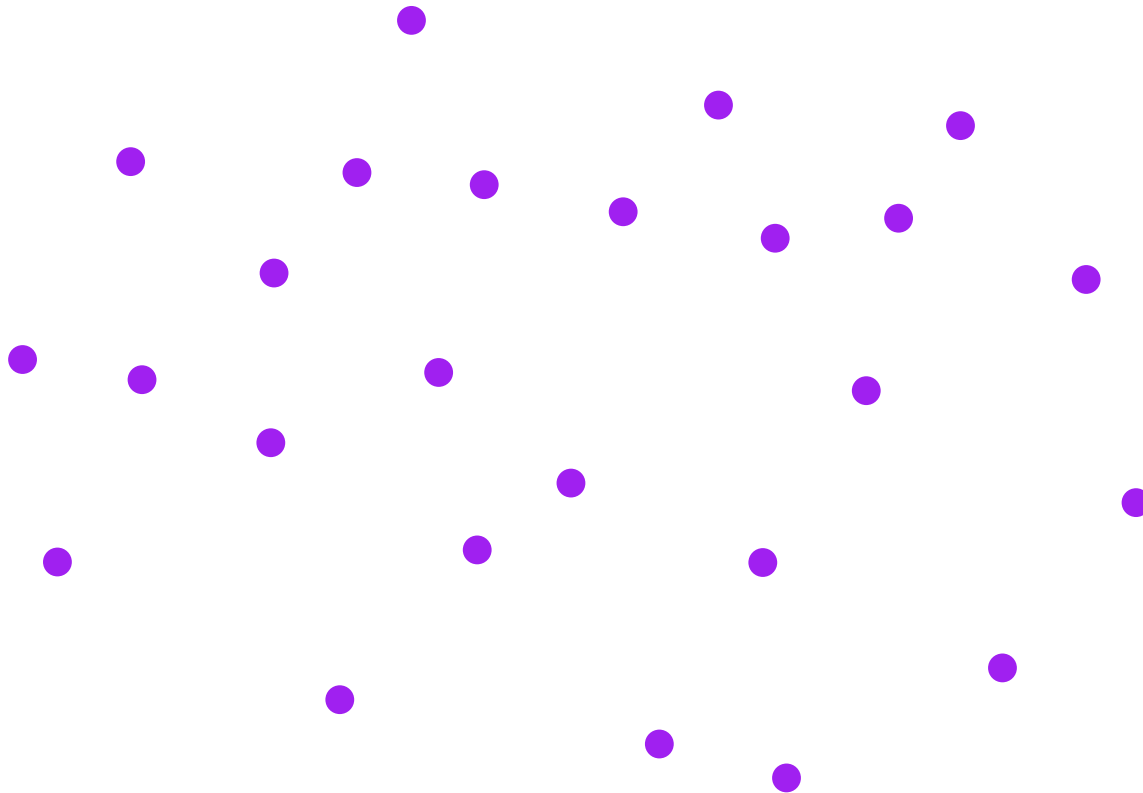


$O(n \log n)$

Convex hull

Other results

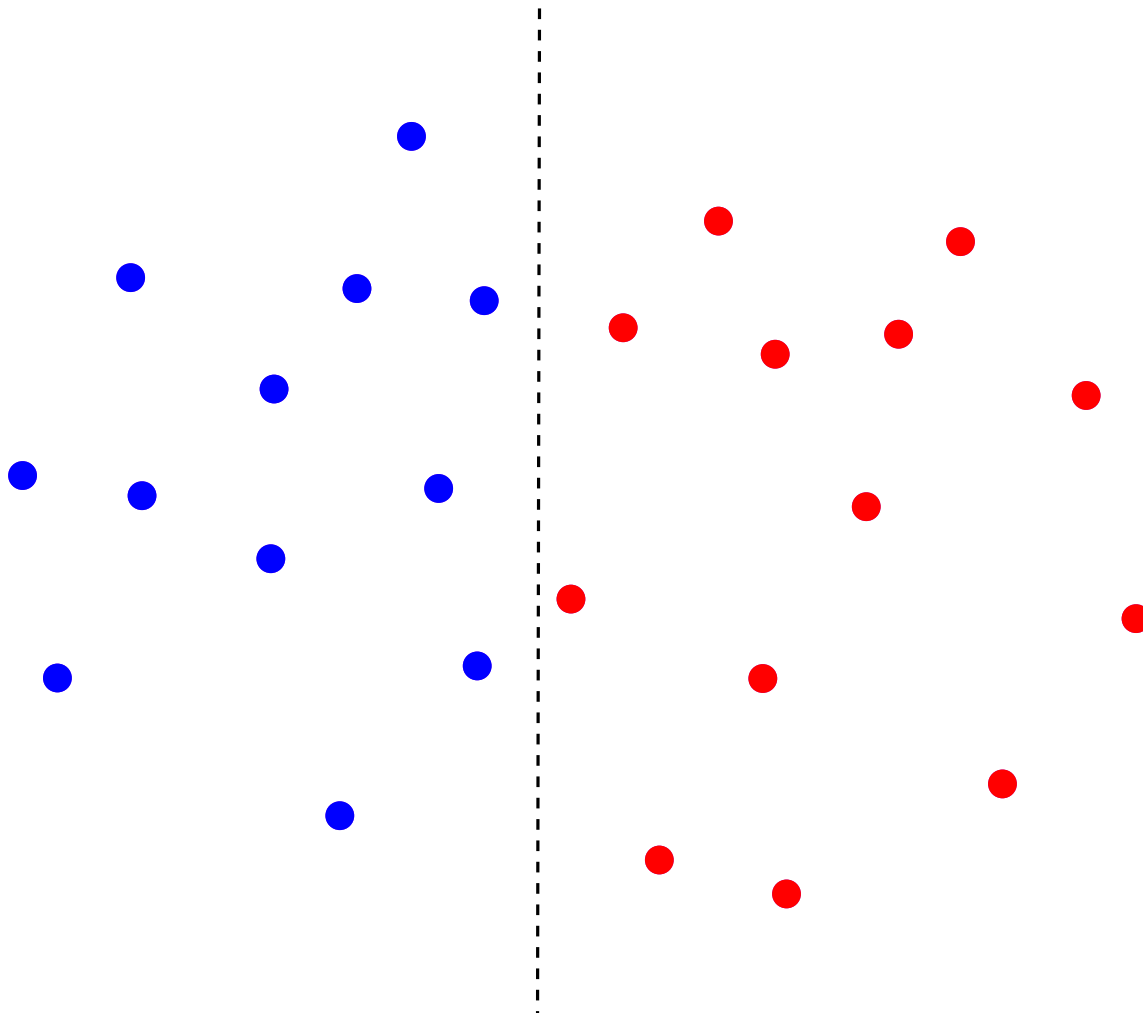
Mariage before conquest



Convex hull

Other results

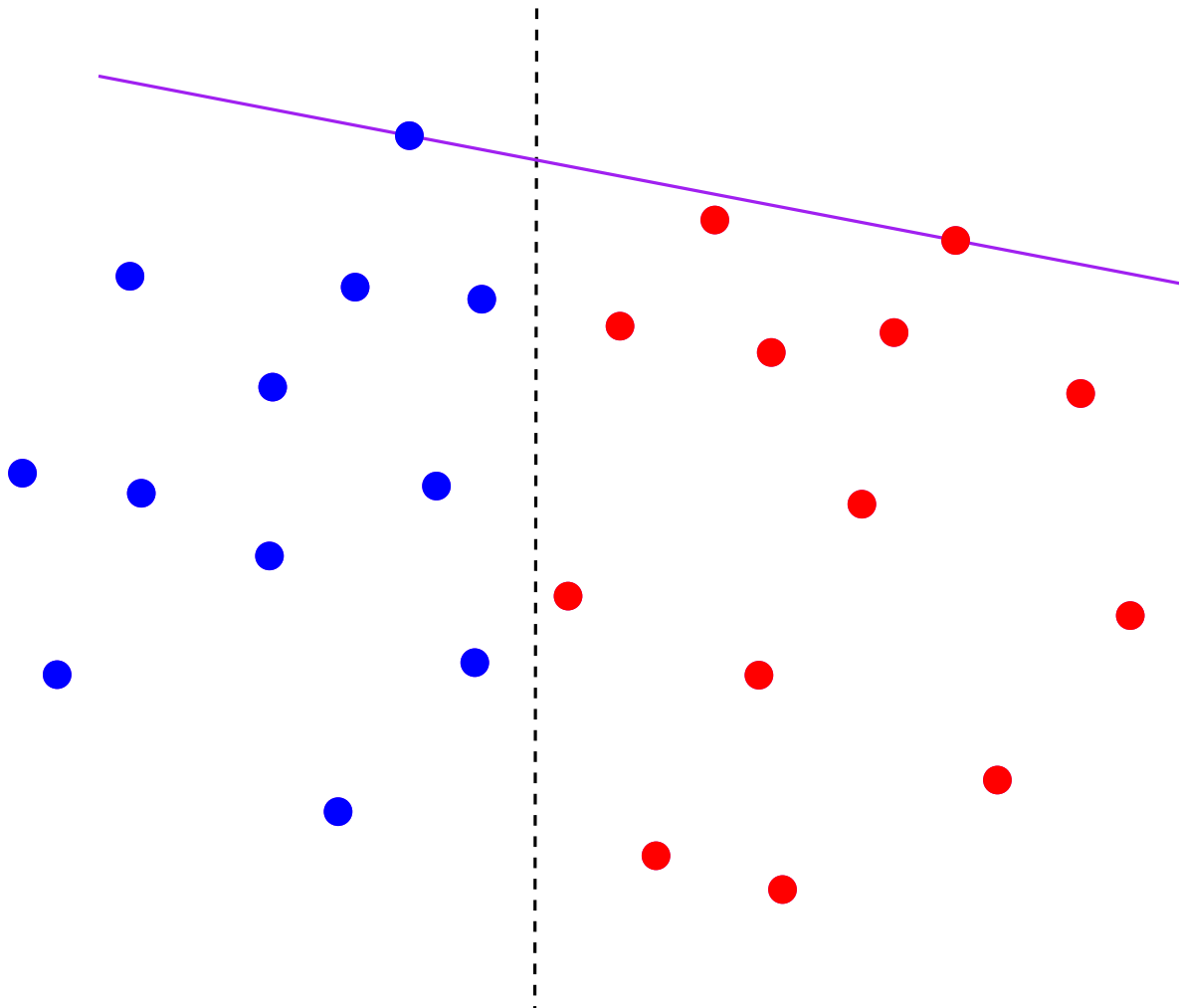
Mariage before conquest



Convex hull

Other results

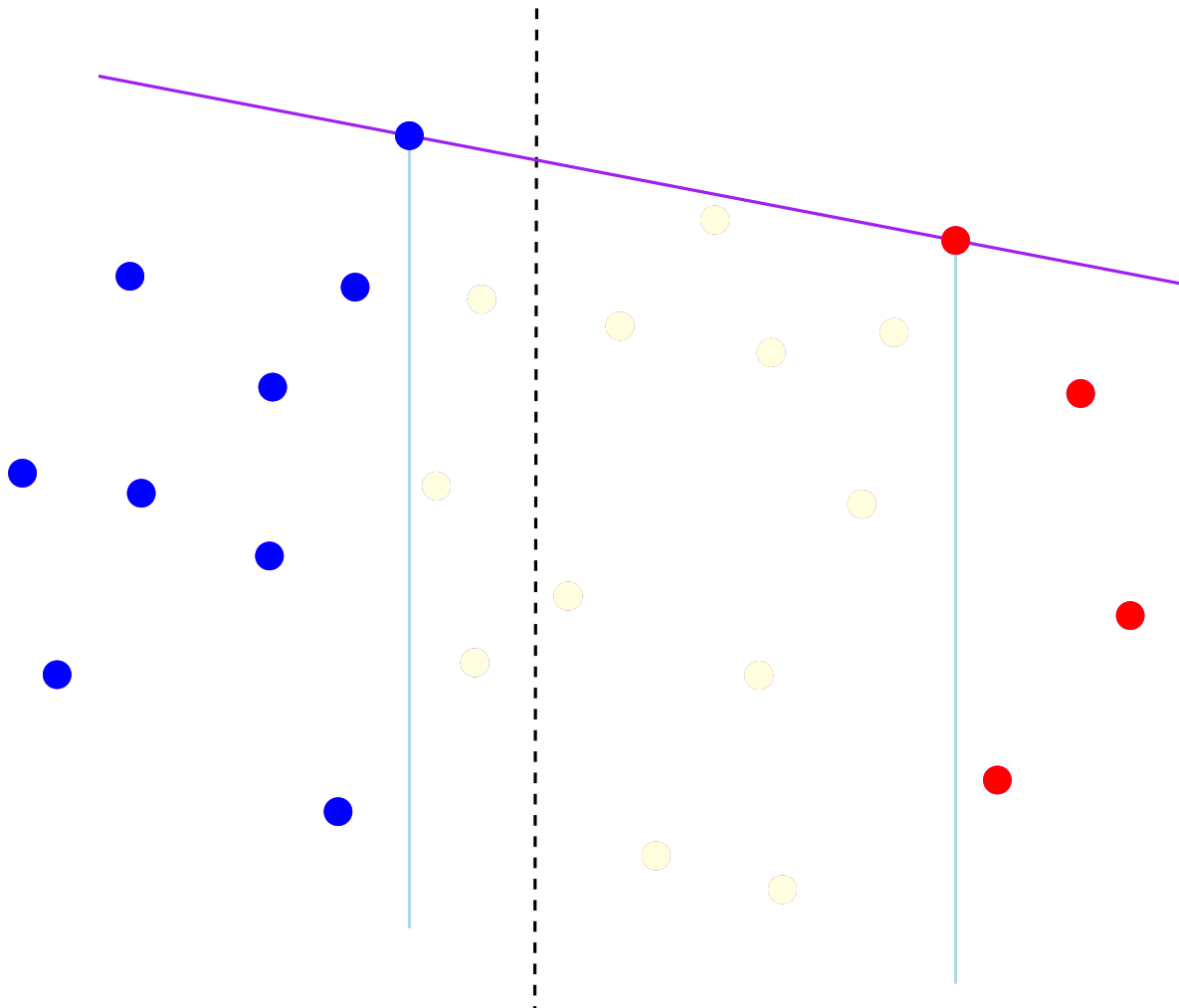
Mariage before conquest



Convex hull

Other results

Mariage before conquest

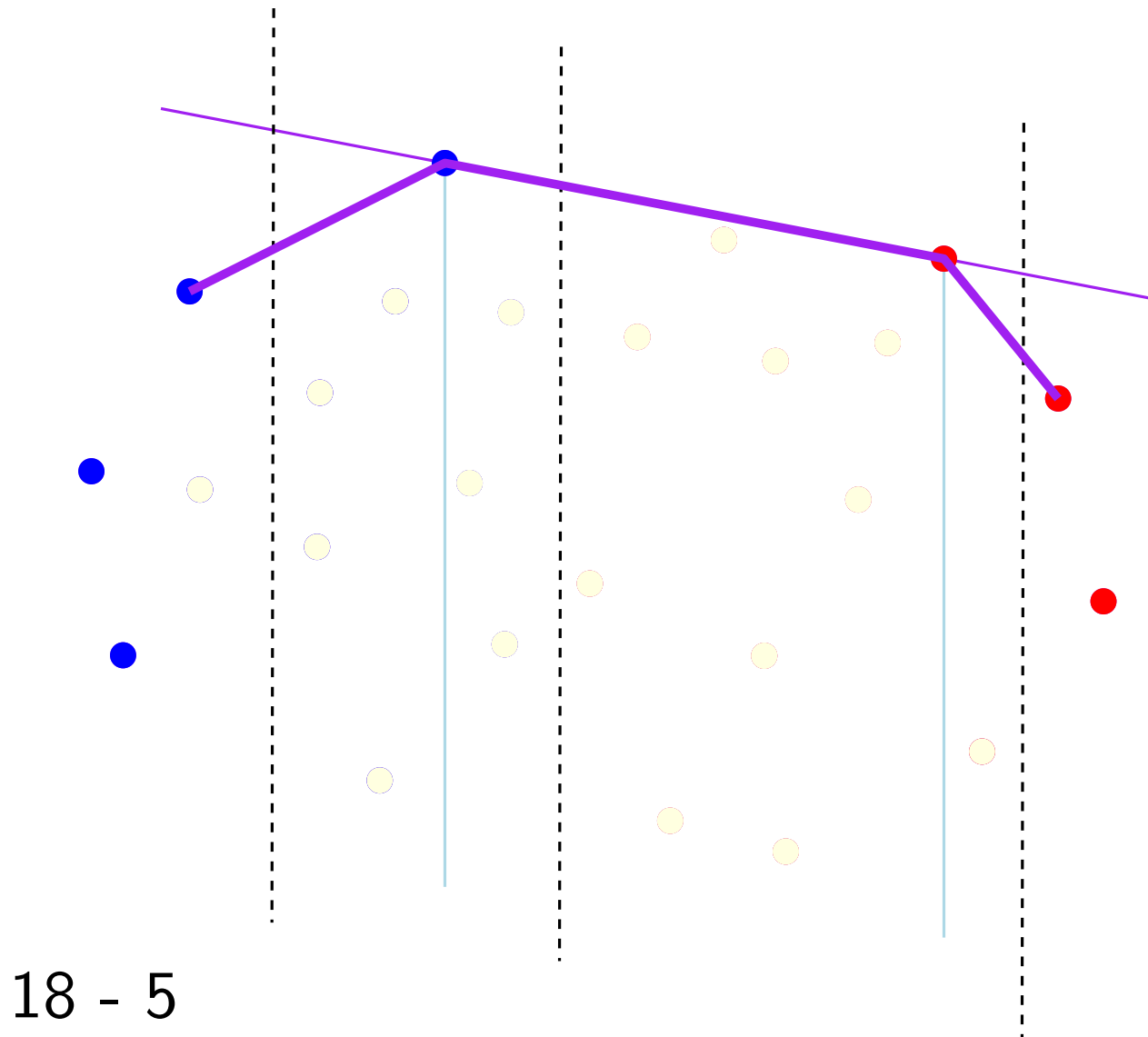


Convex hull

Other results

Mariage before conquest

recursion

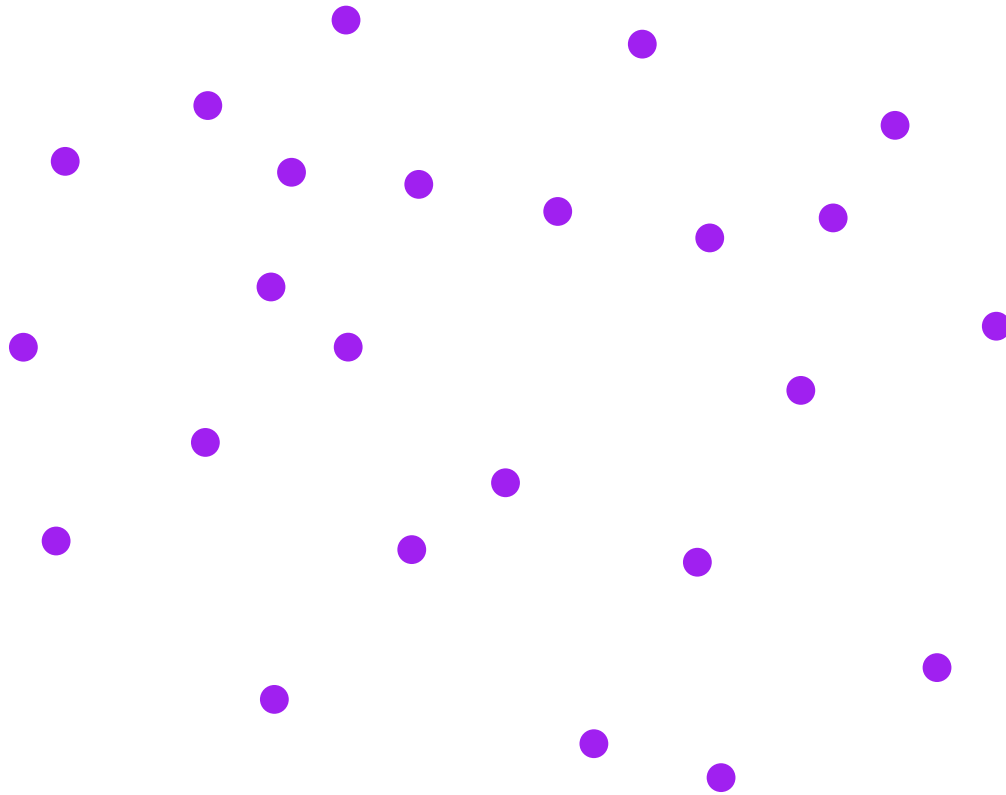


$$O(n \log h)$$

Convex hull

Other results

Quickhull

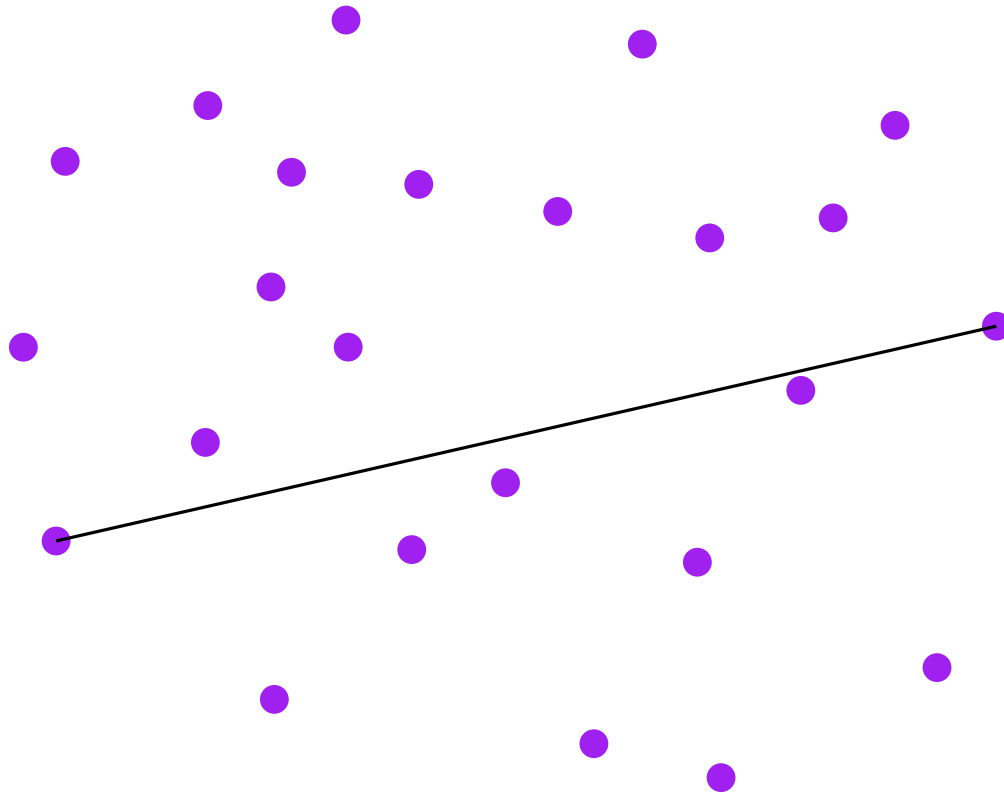


Convex hull

Other results

Quickhull

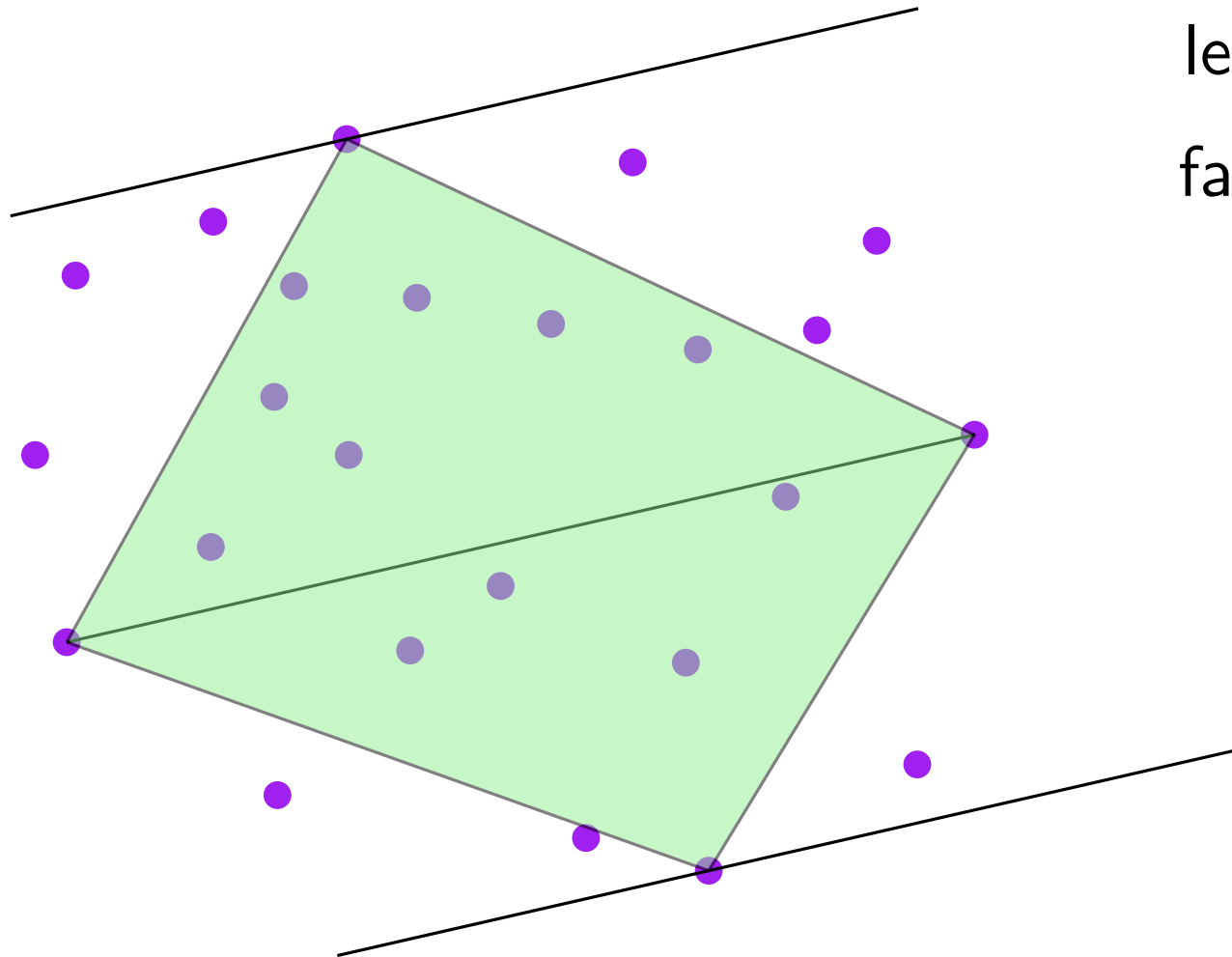
leftmost rightmost



Convex hull

Other results

Quickhull

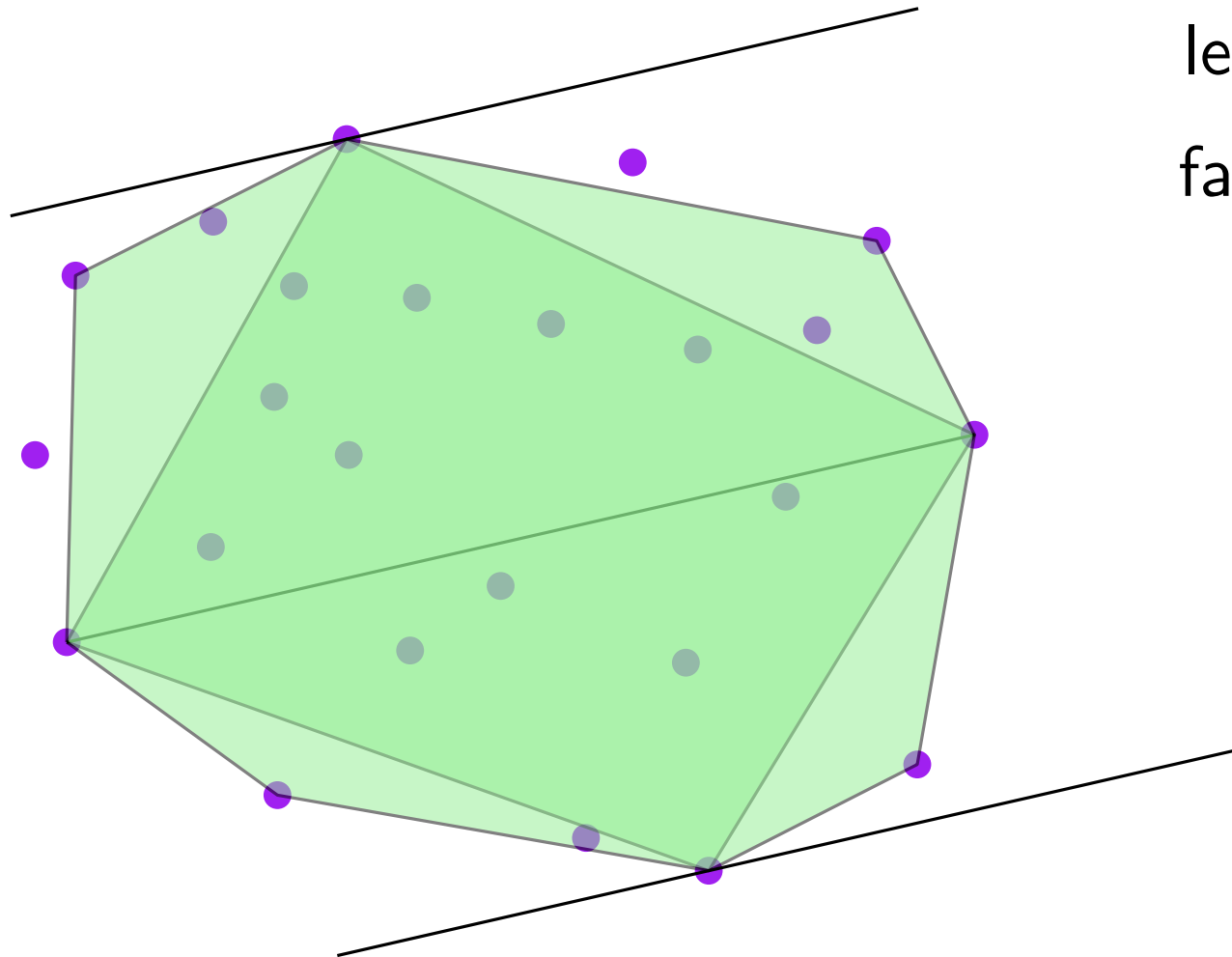


leftmost rightmost
farthest points

Convex hull

Other results

Quickhull

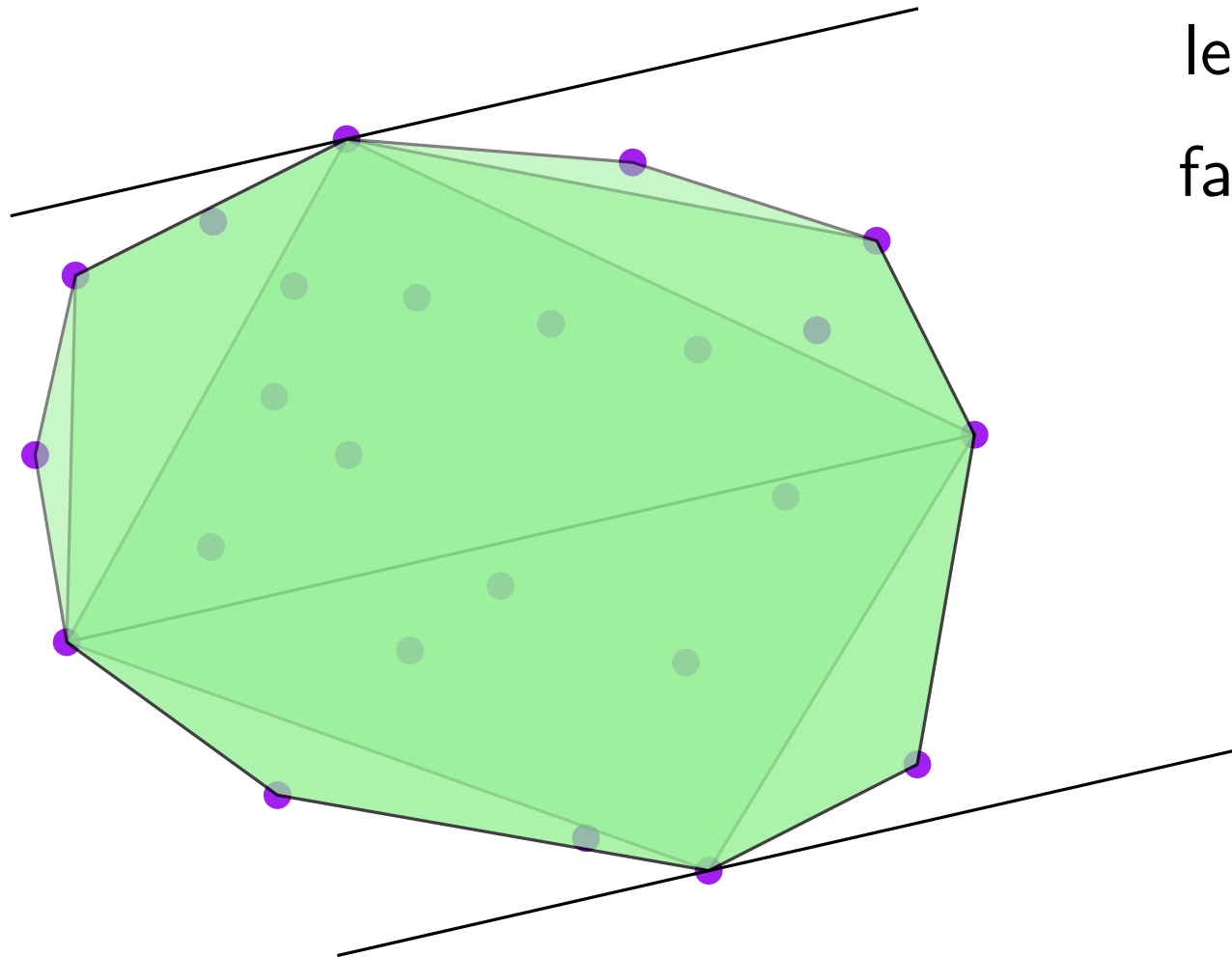


leftmost rightmost
farthest points

Convex hull

Other results

Quickhull

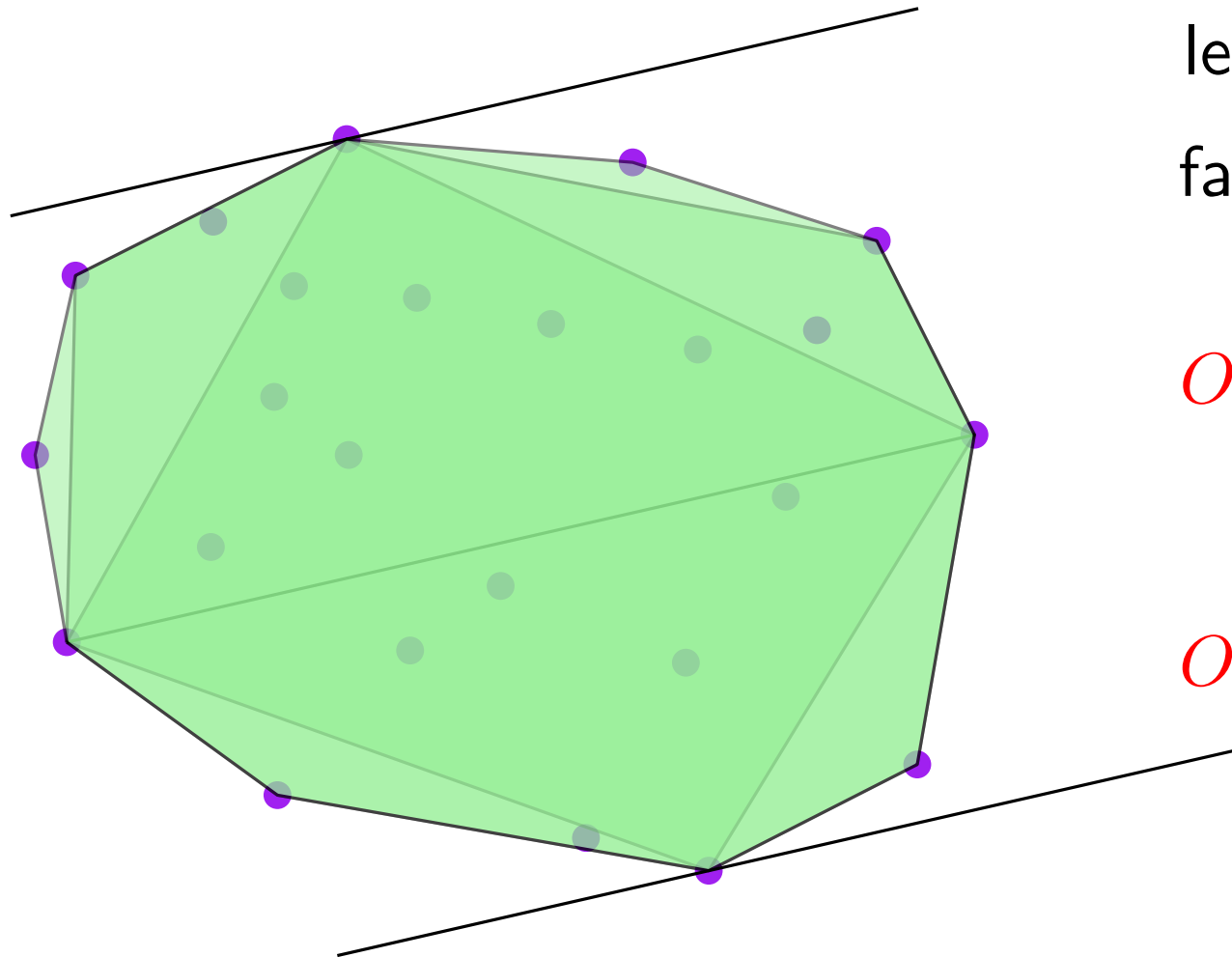


leftmost rightmost
farthest points

Convex hull

Other results

Quickhull



leftmost rightmost
farthest points

$O(n \log n)$

expected

$O(n^2)$

worst-case

Convex hull

Other results

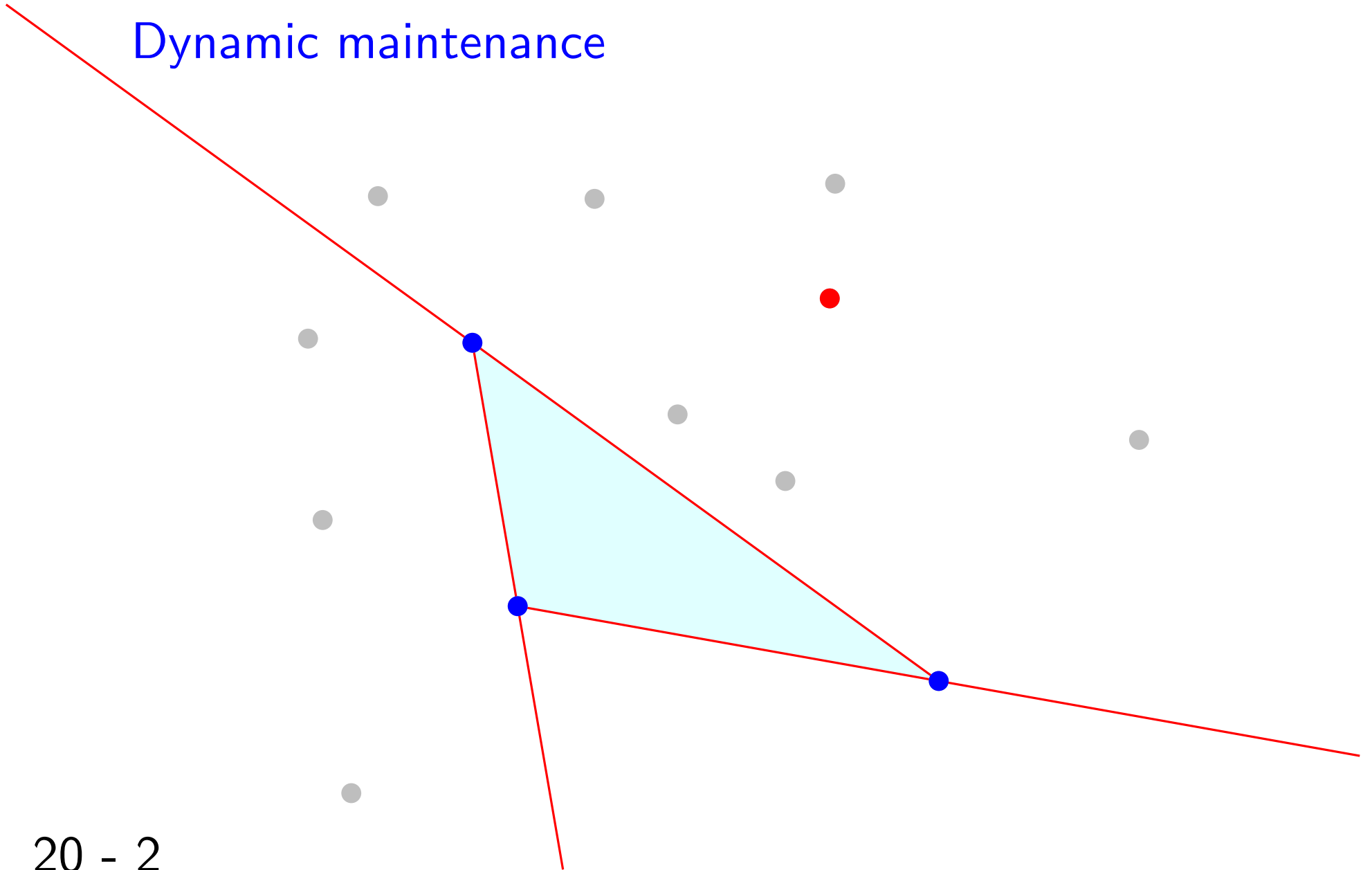
Dynamic maintenance



Convex hull

Other results

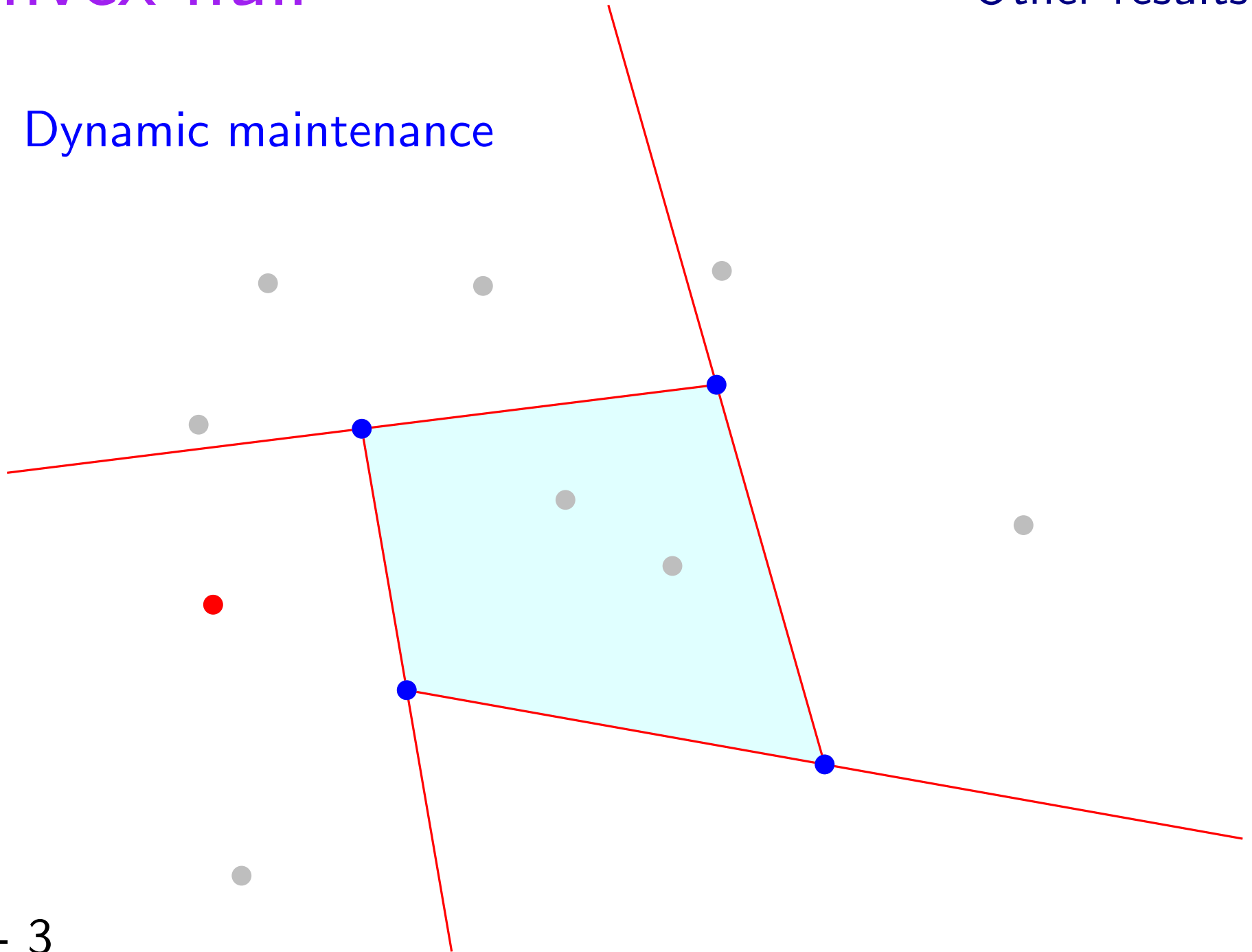
Dynamic maintenance



Convex hull

Other results

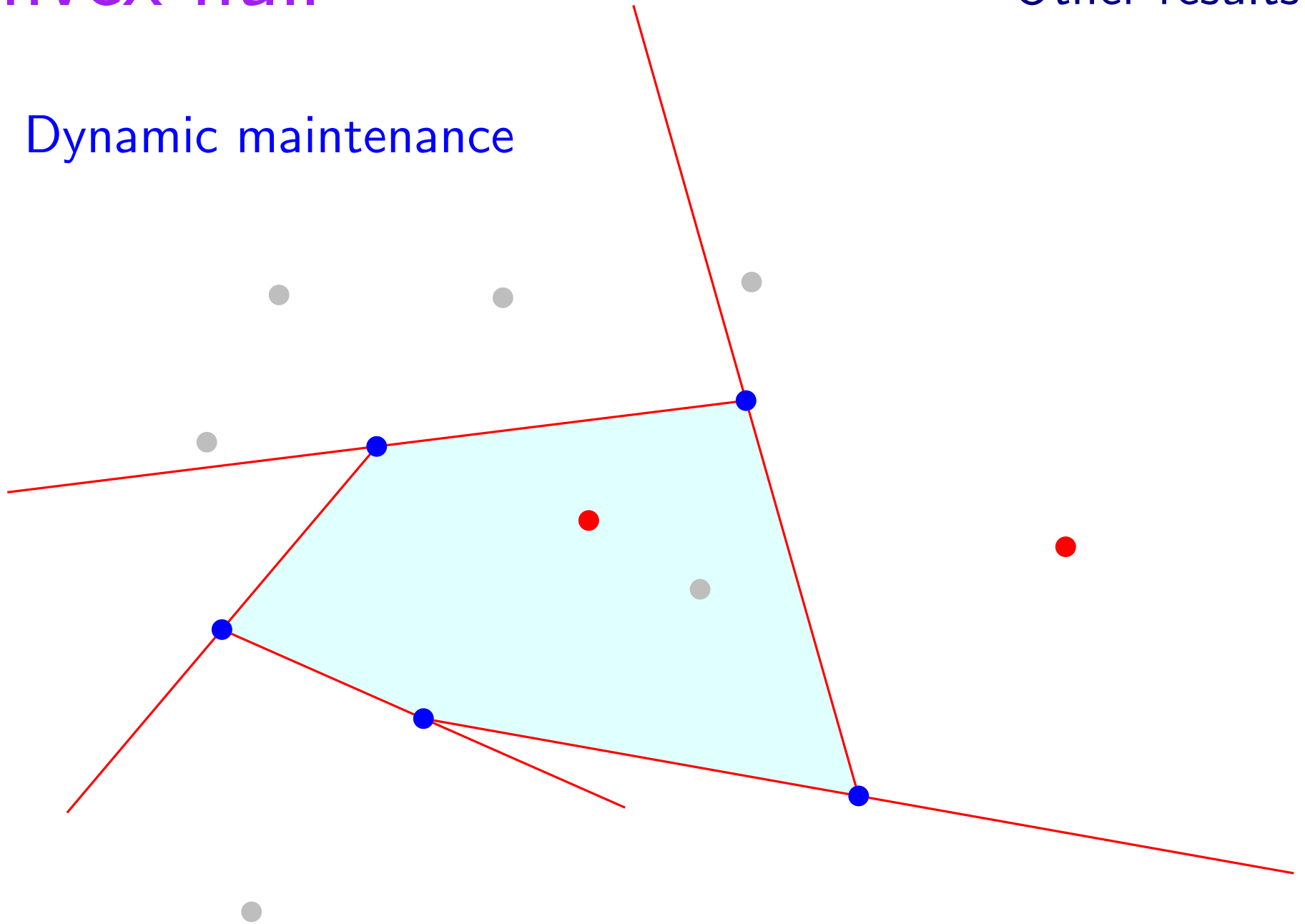
Dynamic maintenance



Convex hull

Other results

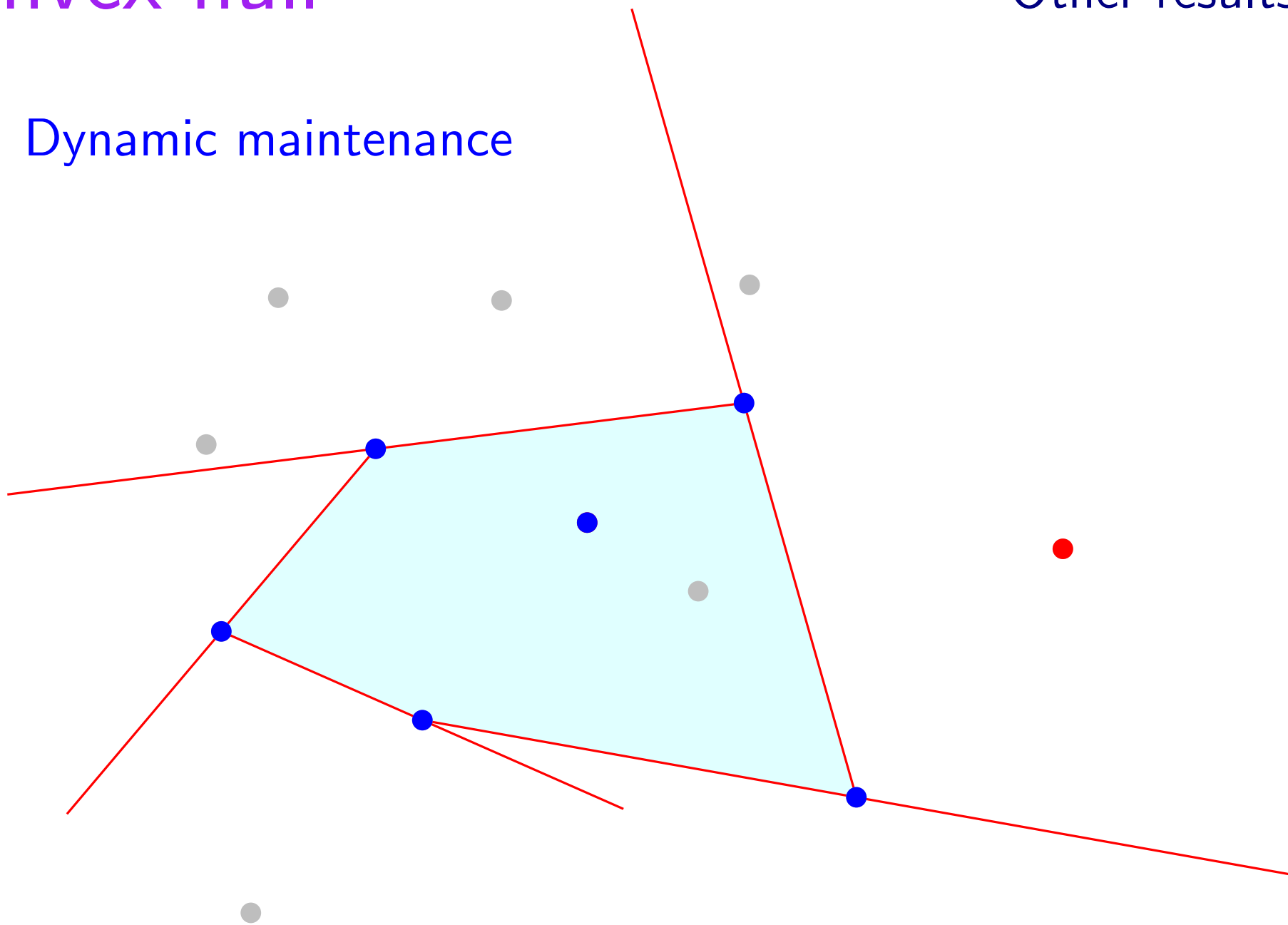
Dynamic maintenance



Convex hull

Other results

Dynamic maintenance

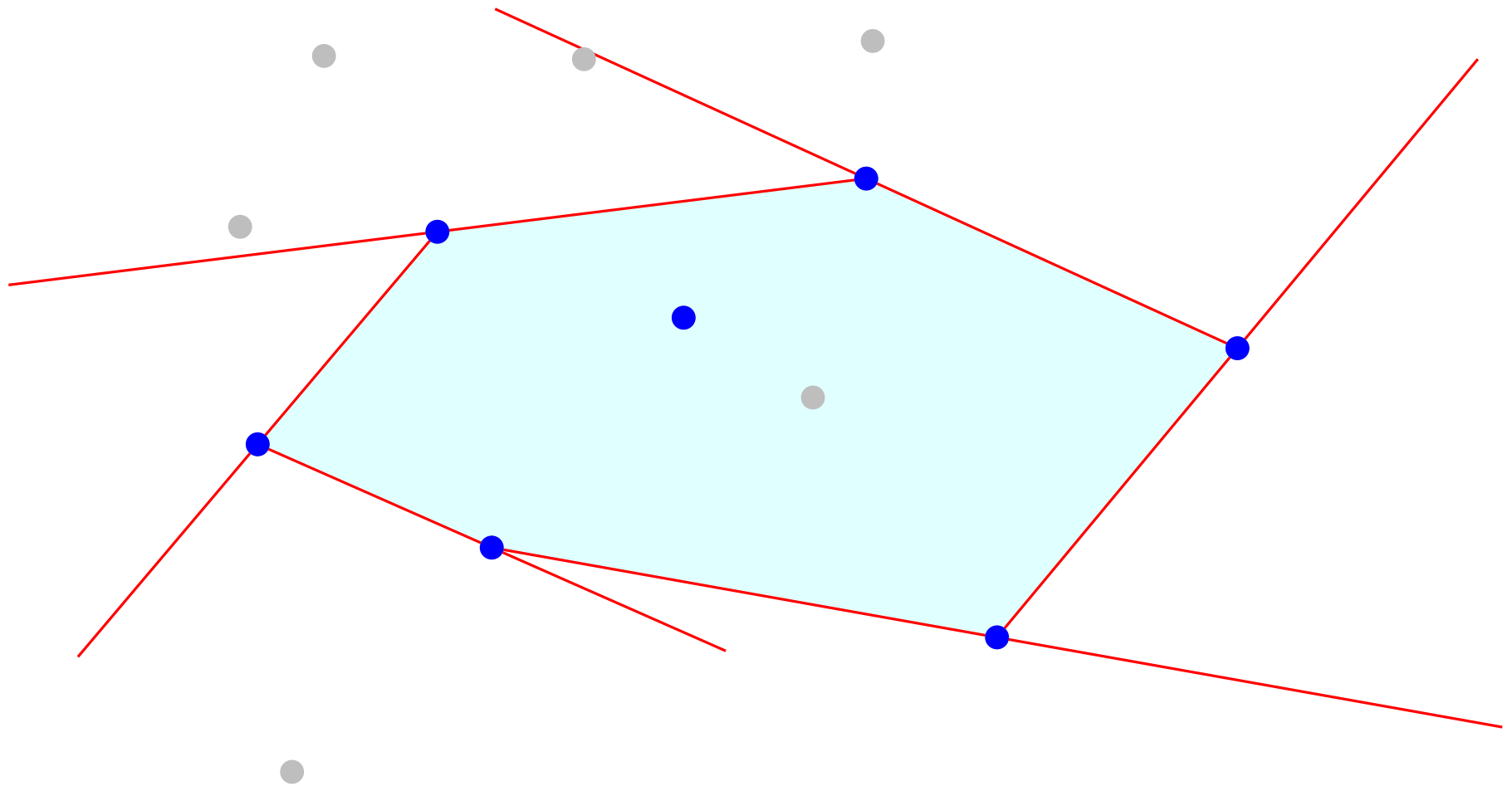


Convex hull

Other results

Dynamic maintenance

$O(\log n)$ per insertion



Convex hull

Three dimensions

Euler relation

Polytope boundary

Vertices

Edges

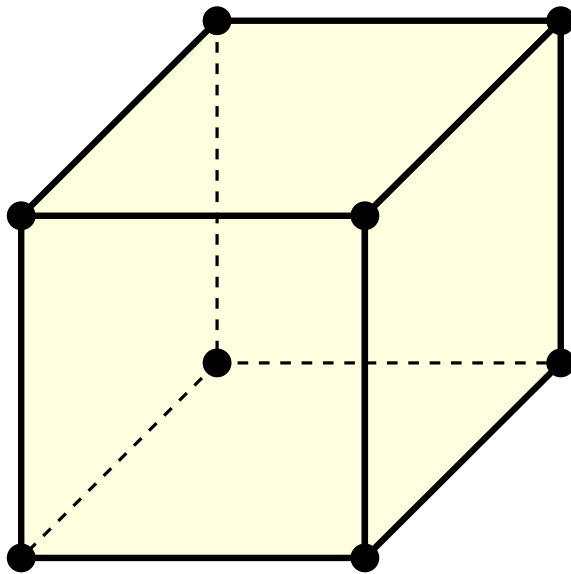
Faces

Convex hull

Three dimensions

Euler relation

Polytope boundary



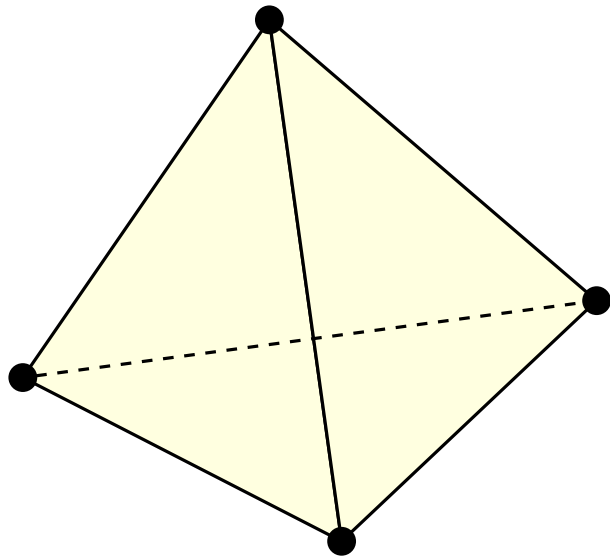
$$\begin{array}{r} \text{Vertices} \\ 8 \end{array} - \begin{array}{r} \text{Edges} \\ 12 \end{array} + \begin{array}{r} \text{Faces} \\ 6 \end{array} = 2$$

Convex hull

Three dimensions

Euler relation

Polytope boundary



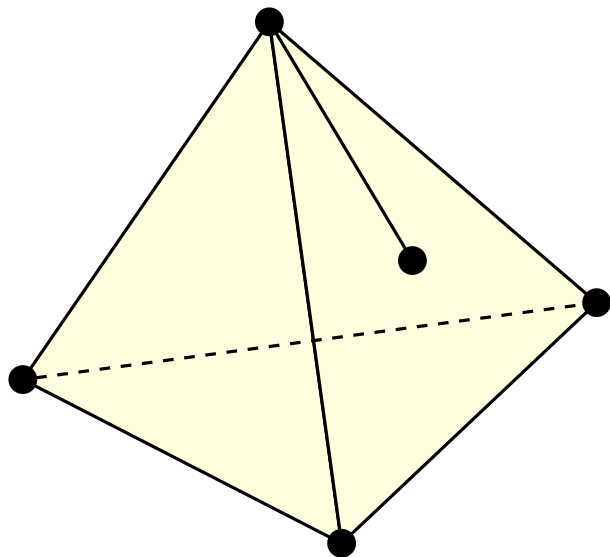
Vertices		Edges		Faces		
8	-	12	+	6	=	2
4	-	6	+	4	=	2

Convex hull

Three dimensions

Euler relation

Polytope boundary



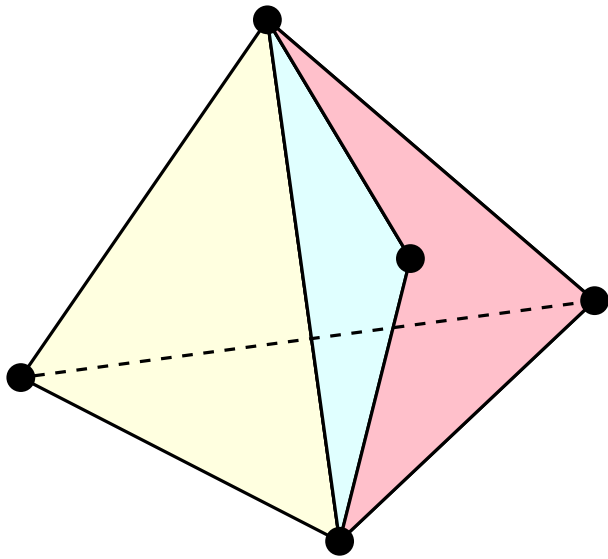
Vertices	Edges	Faces	
8	- 12	+ 6	= 2
4	- 6	+ 4	= 2
+1	- +1	+ 0	= +0

Convex hull

Three dimensions

Euler relation

Polytope boundary



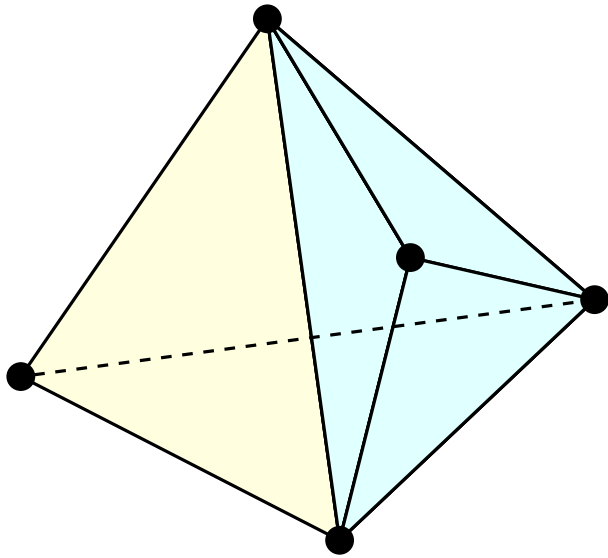
Vertices	Edges	Faces	
8	- 12	+ 6	= 2
4	- 6	+ 4	= 2
+1	-	+1 + 0	= +0
0	-	+1 + +1	= +0

Convex hull

Three dimensions

Euler relation

Polytope boundary



Vertices	Edges	Faces	
8	- 12	+ 6	= 2
4	- 6	+ 4	= 2
+1	-	+1 + 0	= +0
0	-	+1 + +1	= +0

Convex hull

Three dimensions

Euler relation

Polytope boundary

Vertices

Edges

Faces

$$n - e + f = 2$$

Convex hull

Three dimensions

Euler relation

Polytope boundary

Vertices Edges Faces

$$n - e + f = 2$$

triangular faces

$$3f = 2e$$

Convex hull

Three dimensions

Euler relation

Polytope boundary

Vertices Edges Faces

$$n - e + f = 2$$

triangular faces

$$3f = 2e$$

$$f = 2n - 4$$

$$e = 3n - 6$$

Convex hull

Three dimensions

Linear size

$O(n \log n)$ divide and conquer algorithm

$O(nh)$ gift wrapping algorithm

Convex hull

Higher dimensions

Dehn Sommerville relations

$$f_i = \#(\text{faces of dim } i)$$

Euler:

$$f_0 - f_1 + f_2 - \dots - f_{d-1} = (-1)^{d-1} + 1$$

Convex hull

Higher dimensions

Dehn Sommerville relations $f_i = \#(\text{faces of dim } i)$

Euler: $f_0 - f_1 + f_2 - \dots - f_{d-1} = (-1)^{d-1} + 1$

$$\sum_j = k^{d-1} - 1^j \binom{j+1}{k+1} f_j = (-1)^{d-1} f_k$$

$$-1 \leq k \leq d-2 \quad f_{-1} = f_d = 1$$

$\left\lfloor \frac{d+1}{2} \right\rfloor$ independent equations

Convex hull

Higher dimensions

Dehn Sommerville relations

$$f_i = \#(\text{faces of dim } i)$$

If $f_0, f_1, \dots, f_{\lfloor \frac{d-1}{2} \rfloor}$ are known

$f_{\lfloor \frac{d+1}{2} \rfloor}, \dots, f_{d-1}$ can be deduced

Convex hull

Higher dimensions

Dehn Sommerville relations

$$f_i = \#(\text{faces of dim } i)$$

If $f_0, f_1, \dots, f_{\lfloor \frac{d-1}{2} \rfloor}$ are known

$f_{\lfloor \frac{d+1}{2} \rfloor}, \dots, f_{d-1}$ can be deduced

$$f_{\lfloor \frac{d-1}{2} \rfloor} = O(n^{\lfloor \frac{d+1}{2} \rfloor})$$

$$\implies \forall i \quad f_i = O(n^{\lfloor \frac{d+1}{2} \rfloor})$$

Convex hull

Higher dimensions

Dehn Sommerville relations

$$f_i = \#(\text{faces of dim } i)$$

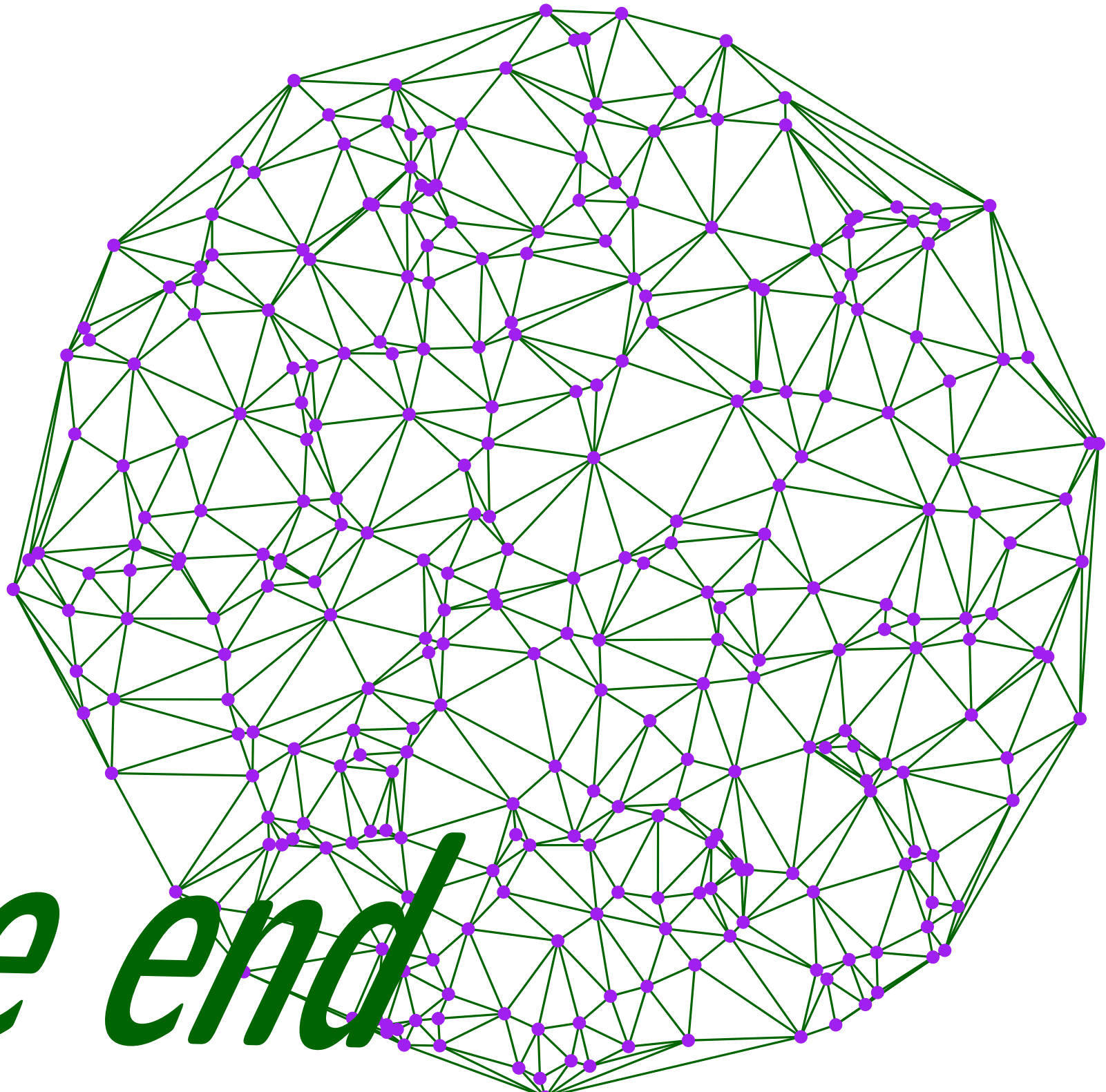
If $f_0, f_1, \dots, f_{\lfloor \frac{d-1}{2} \rfloor}$ are known

$f_{\lfloor \frac{d+1}{2} \rfloor}, \dots, f_{d-1}$ can be deduced

$$f_{\lfloor \frac{d-1}{2} \rfloor} = O(n^{\lfloor \frac{d+1}{2} \rfloor})$$

$$\implies \forall i \quad f_i = O(n^{\lfloor \frac{d+1}{2} \rfloor})$$

\exists an optimal algorithm



The end