

# Geometric Measures on Imprecise Points in Higher Dimensions

Hein Kruger

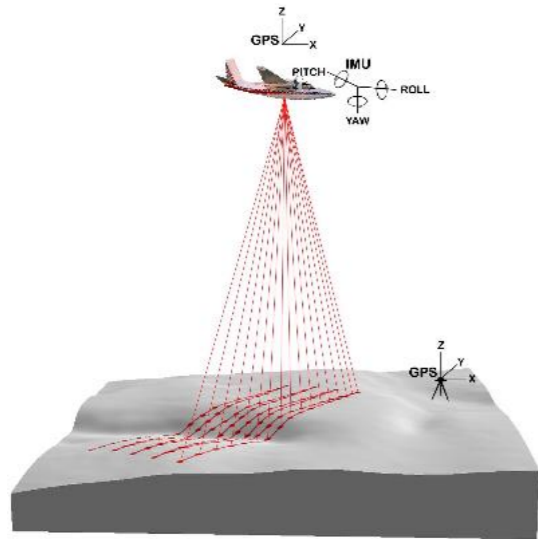
Maarten Löffler

Utrecht University

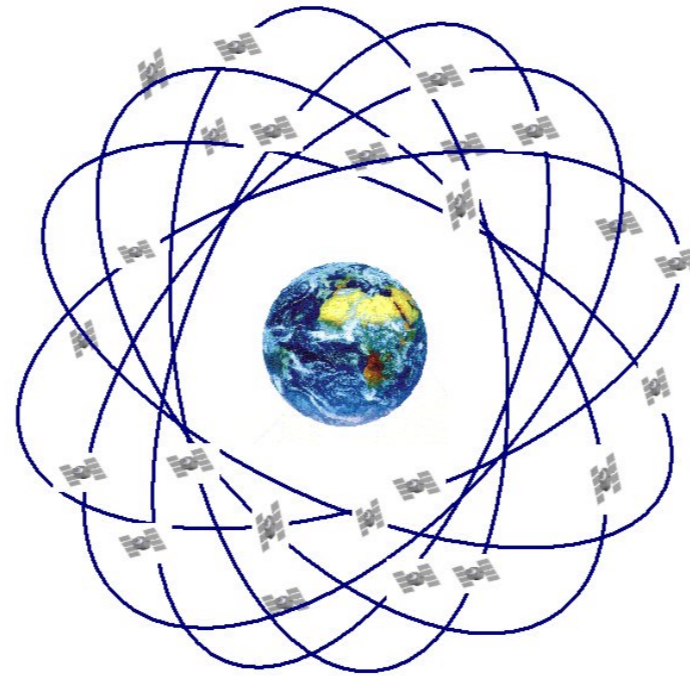
What is this  
about?

Data is often  
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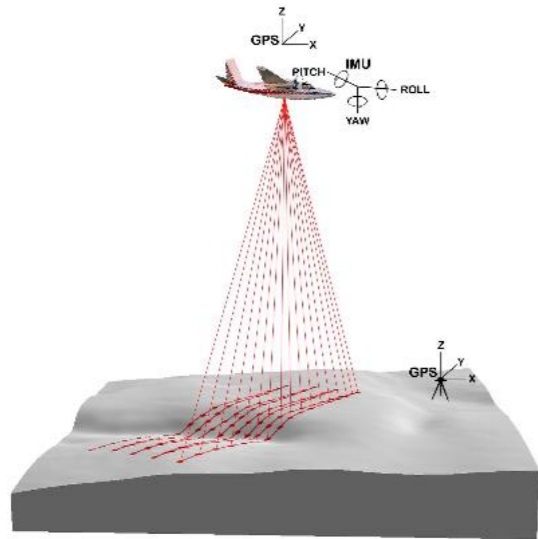
LIDAR



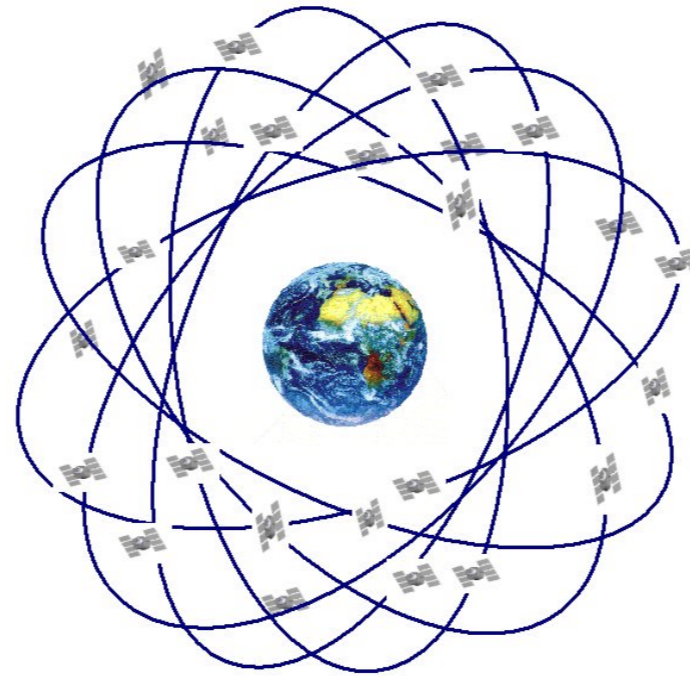
GPS

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Data often stored in some *digital* format.



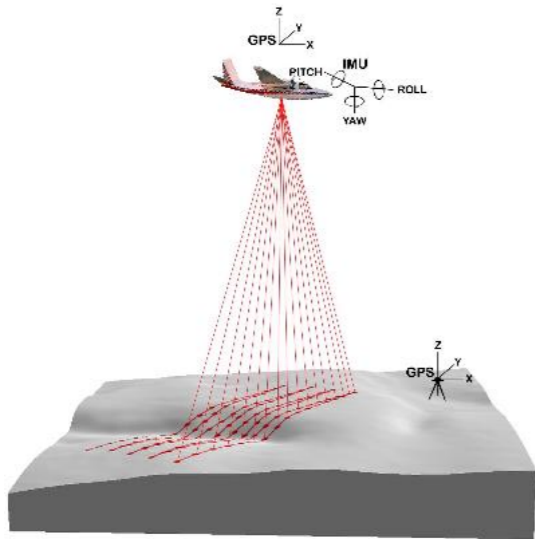
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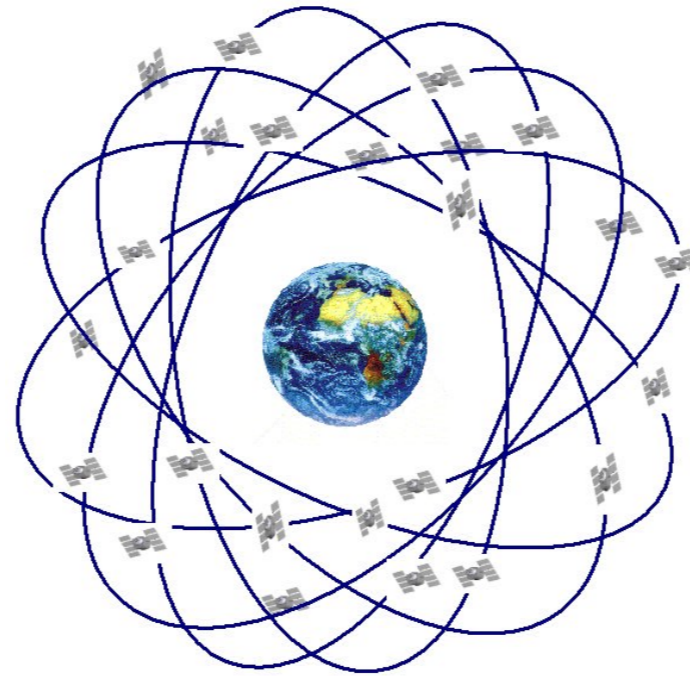
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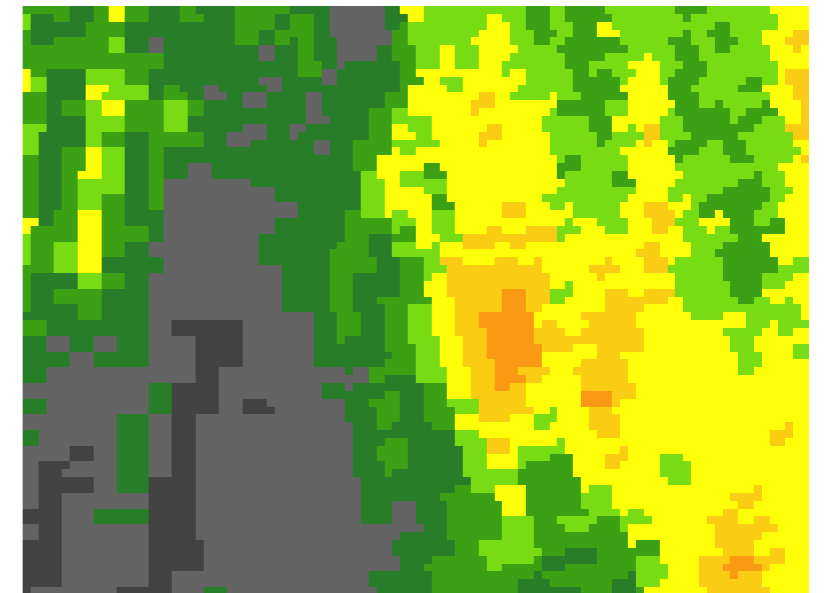
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LIDAR



GPS



Grids

Let's define  
*imprecise*  
*points.*

Computational geometry deals with problems on *precisely* specified *points* in  $\mathbb{R}^2$ .





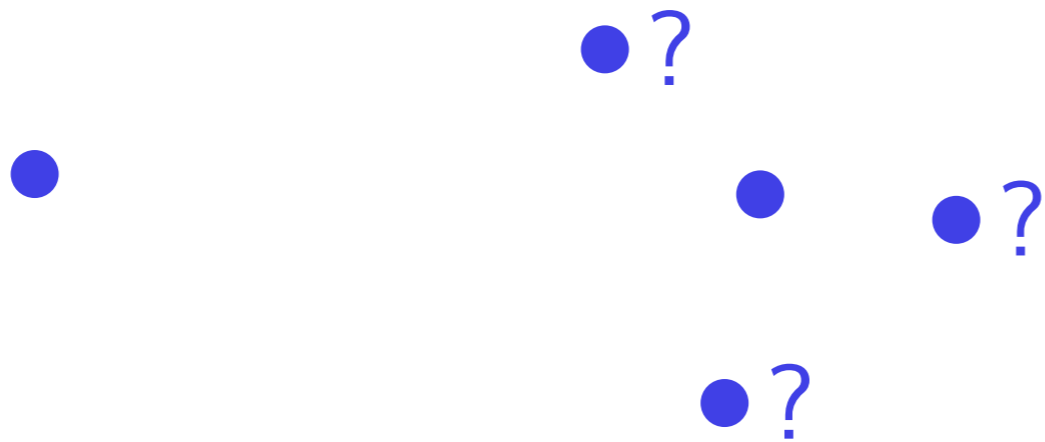
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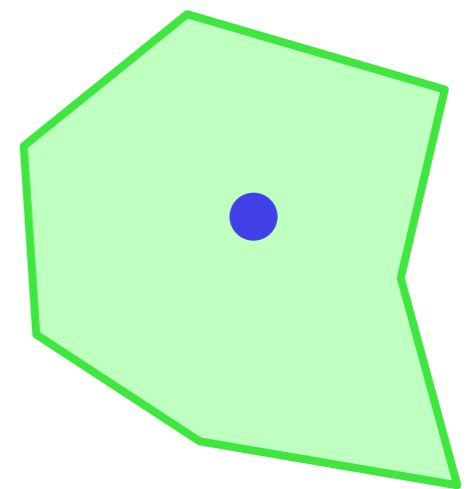
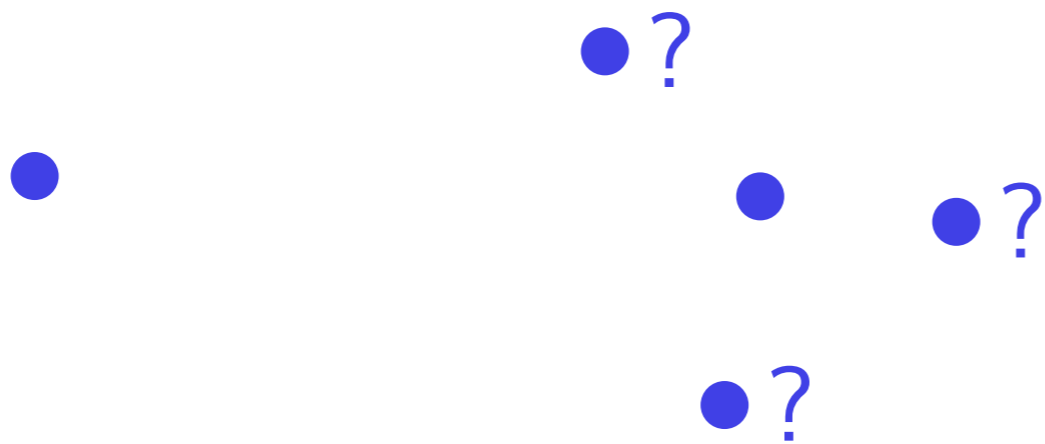
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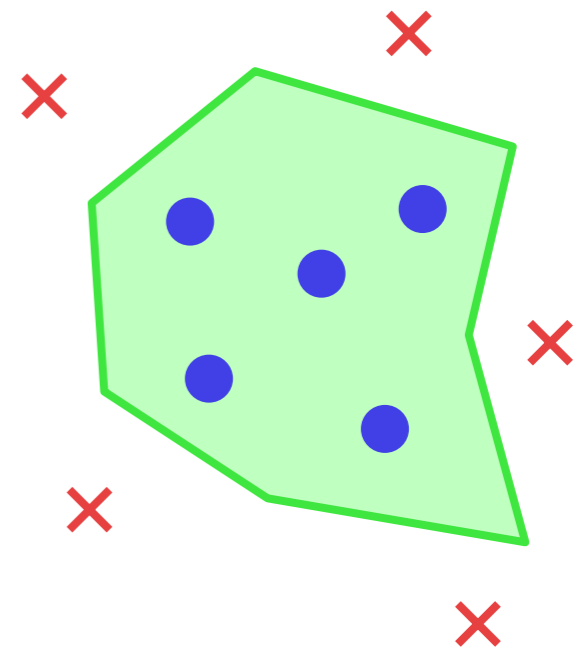
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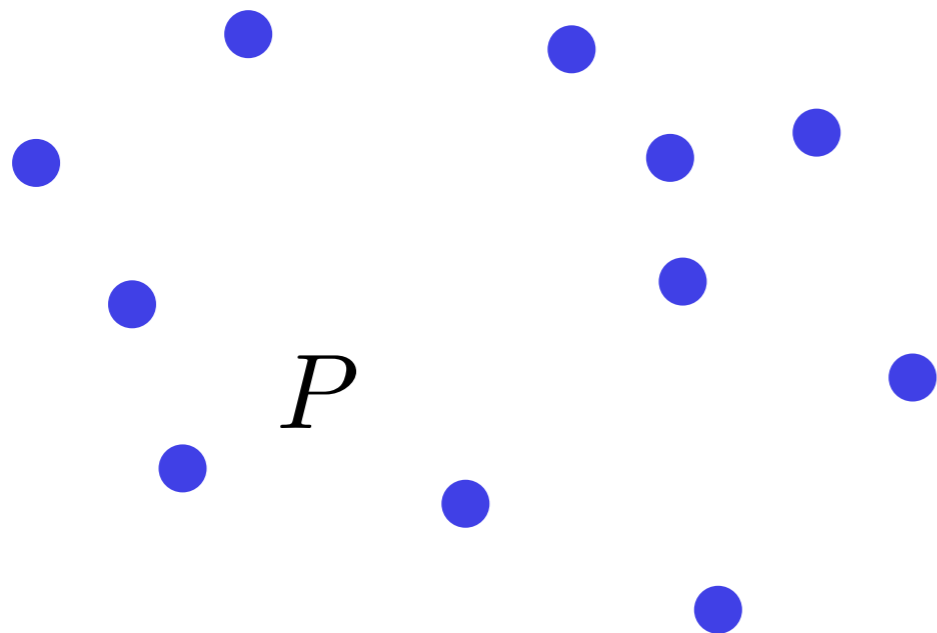
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What can we  
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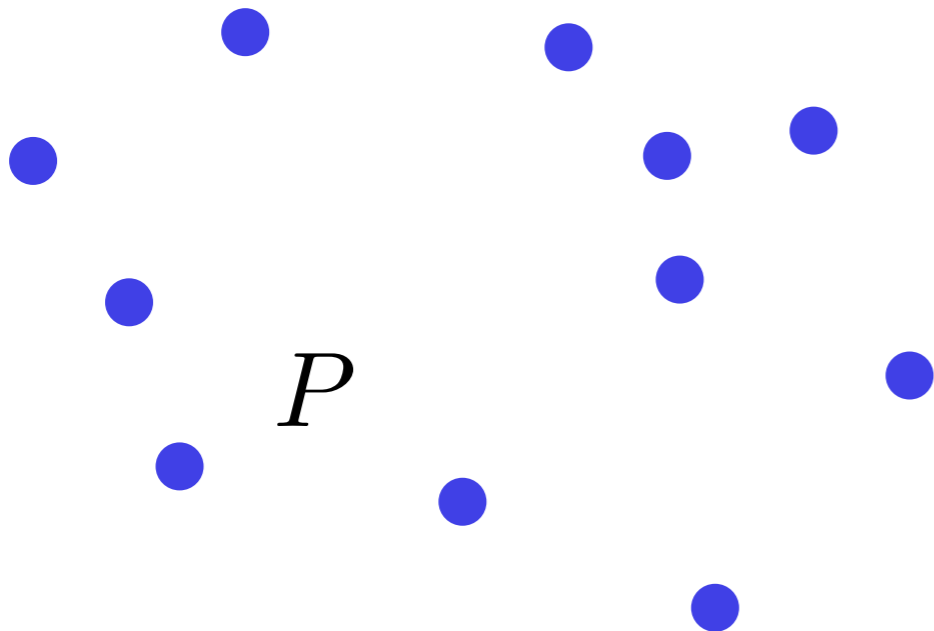
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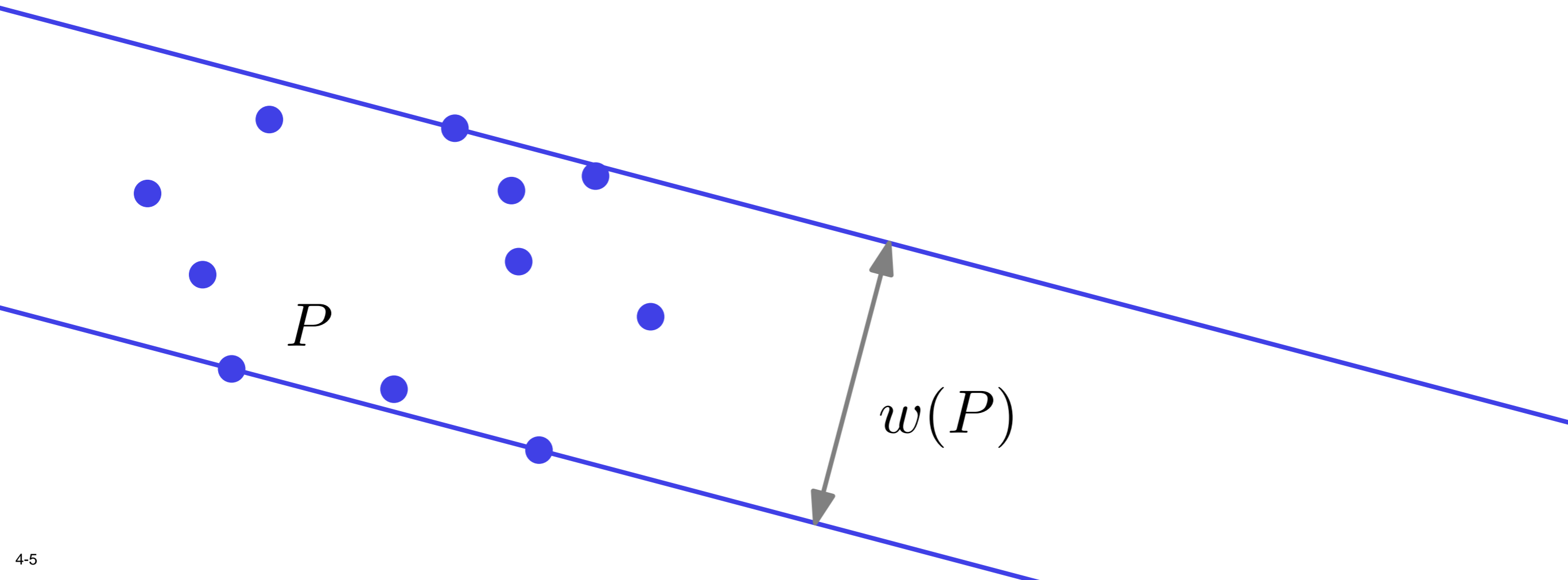
For example, the *width* of  $P$ , or the *diameter* of  $P$ .





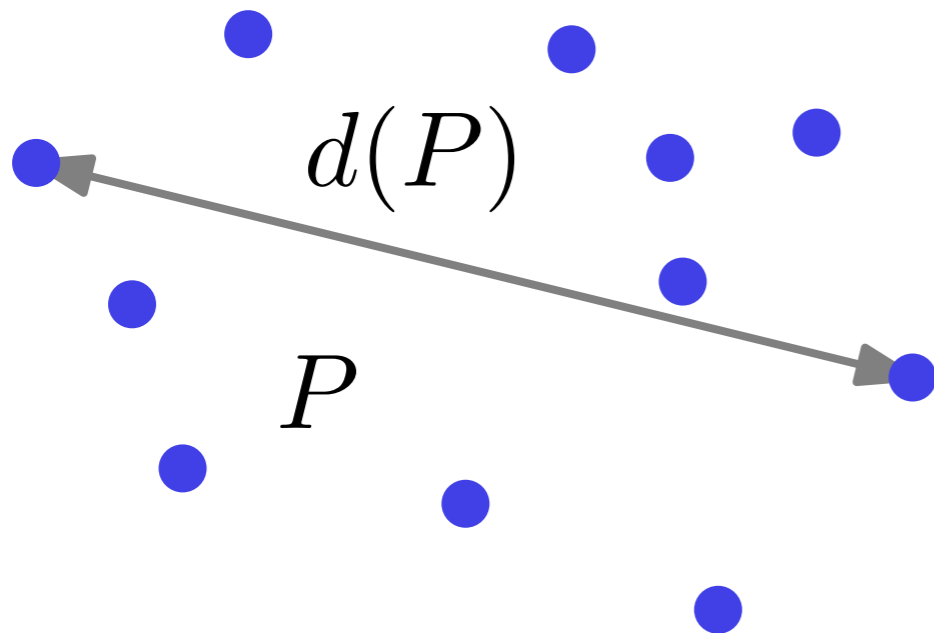
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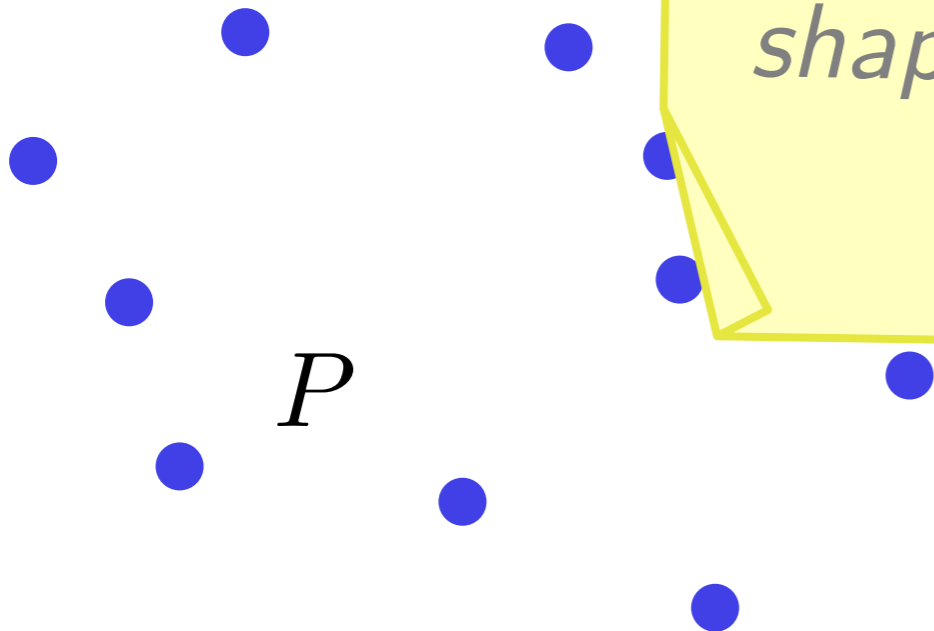
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These measures tell something about the *shape* of  $P$ .

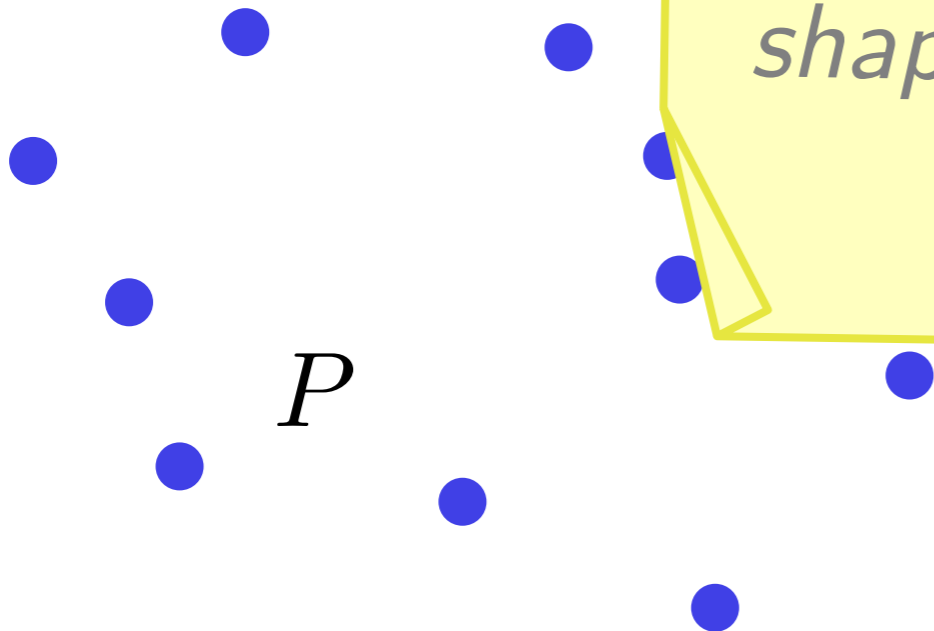


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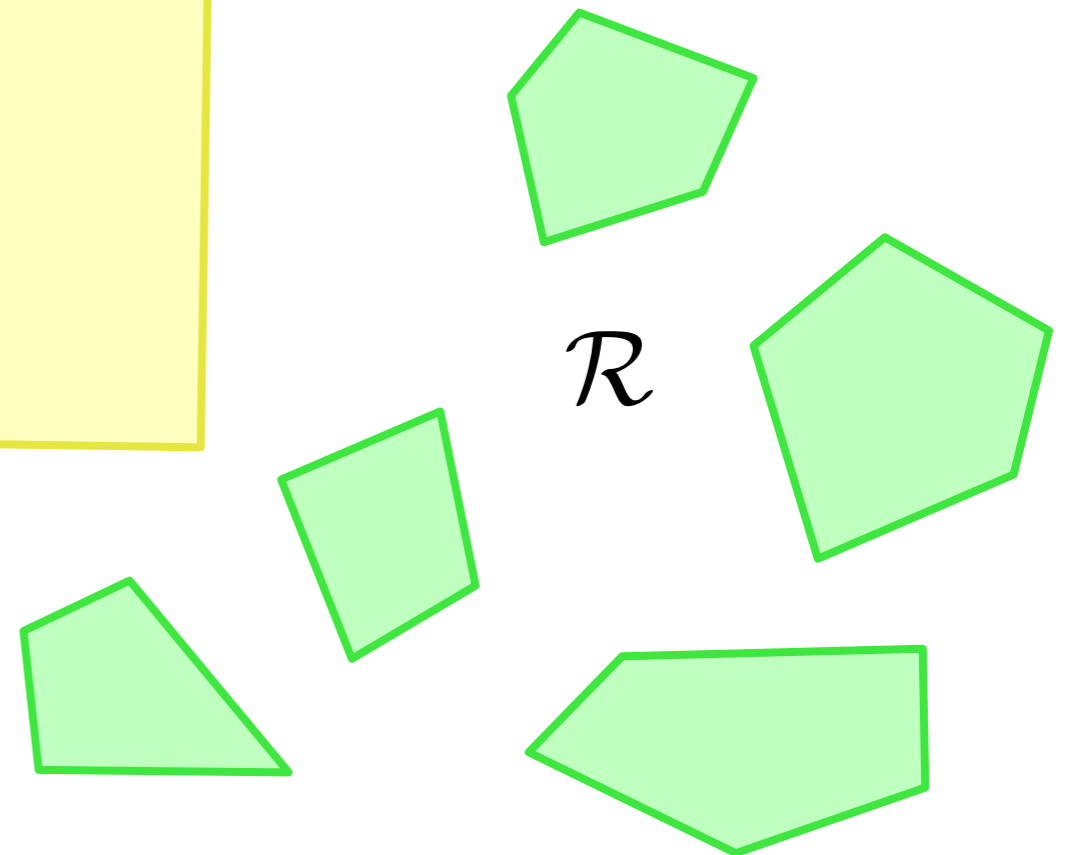
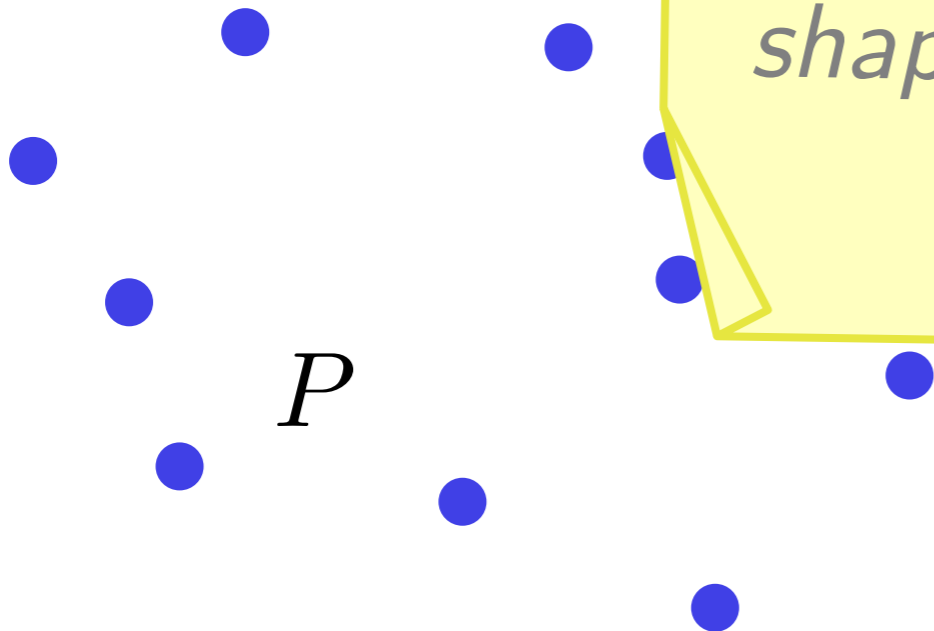


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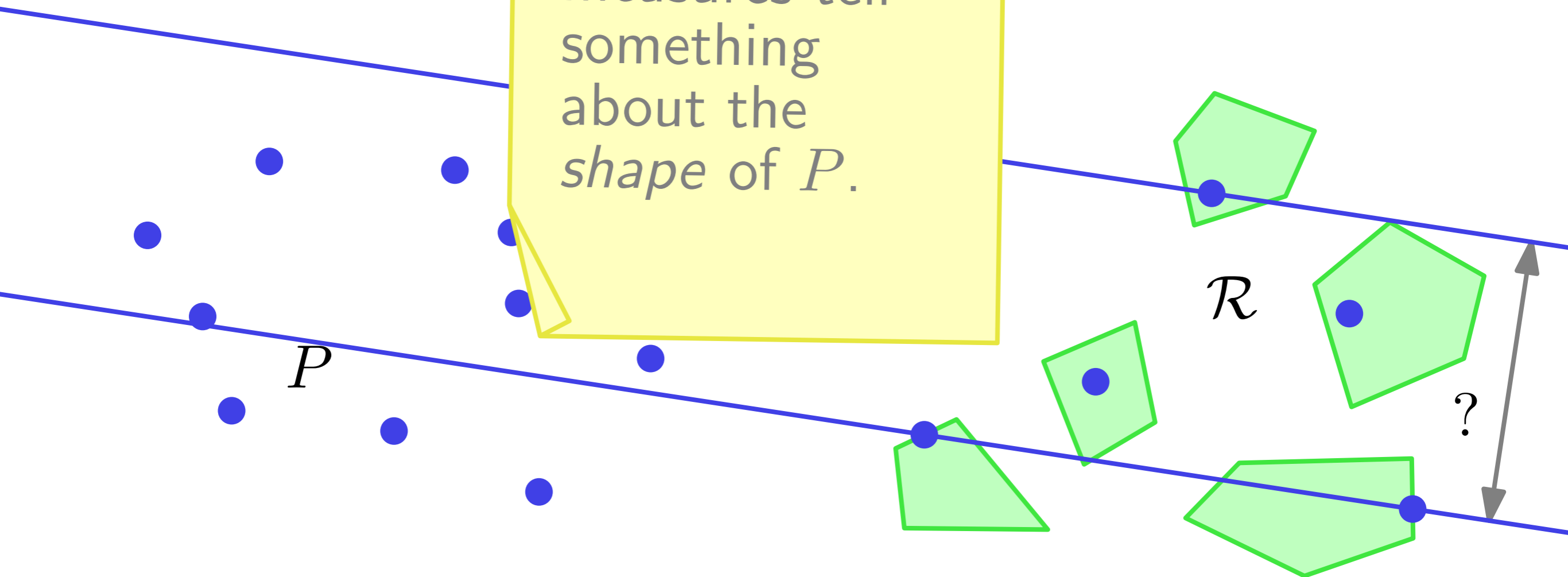


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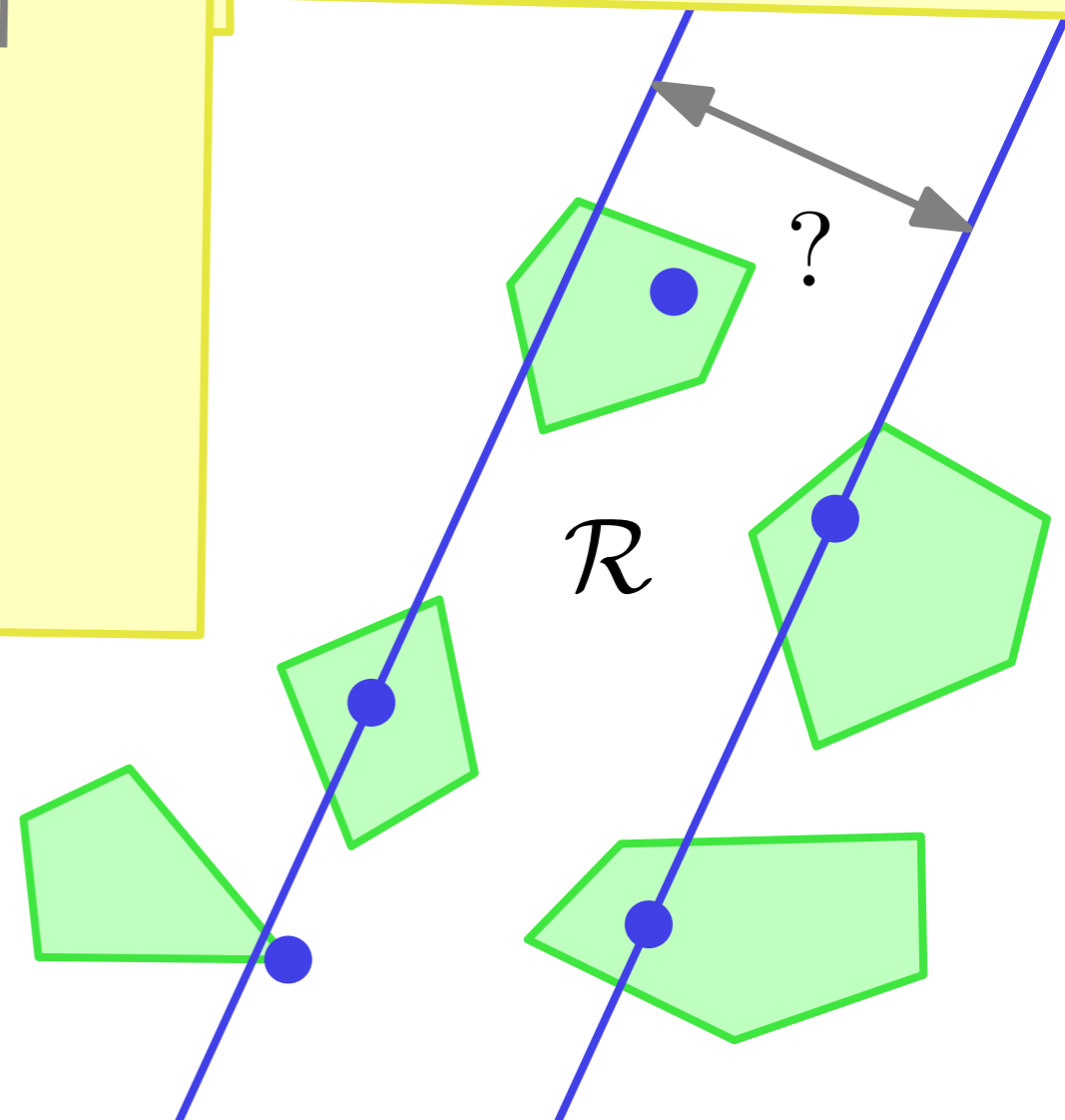
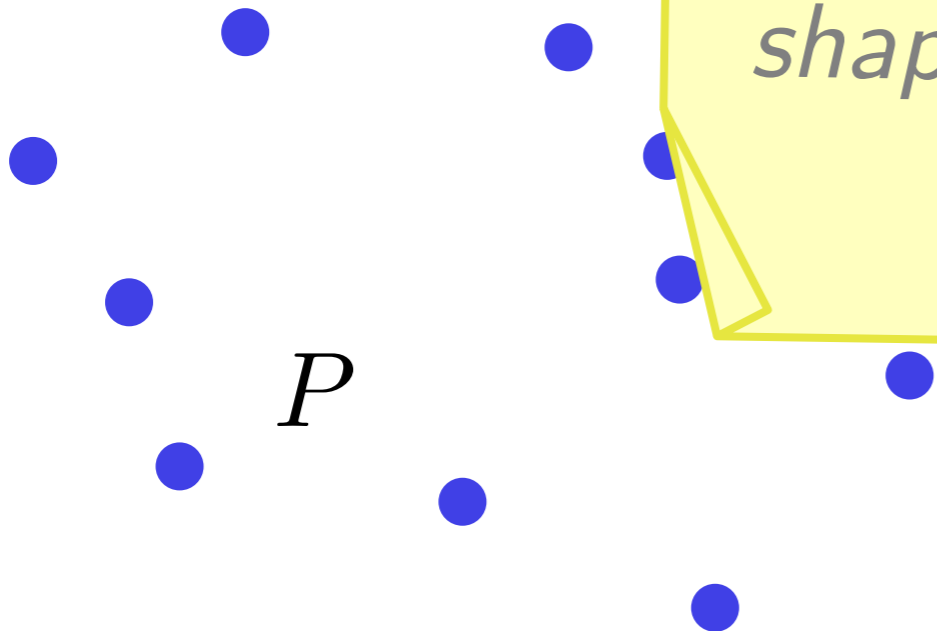


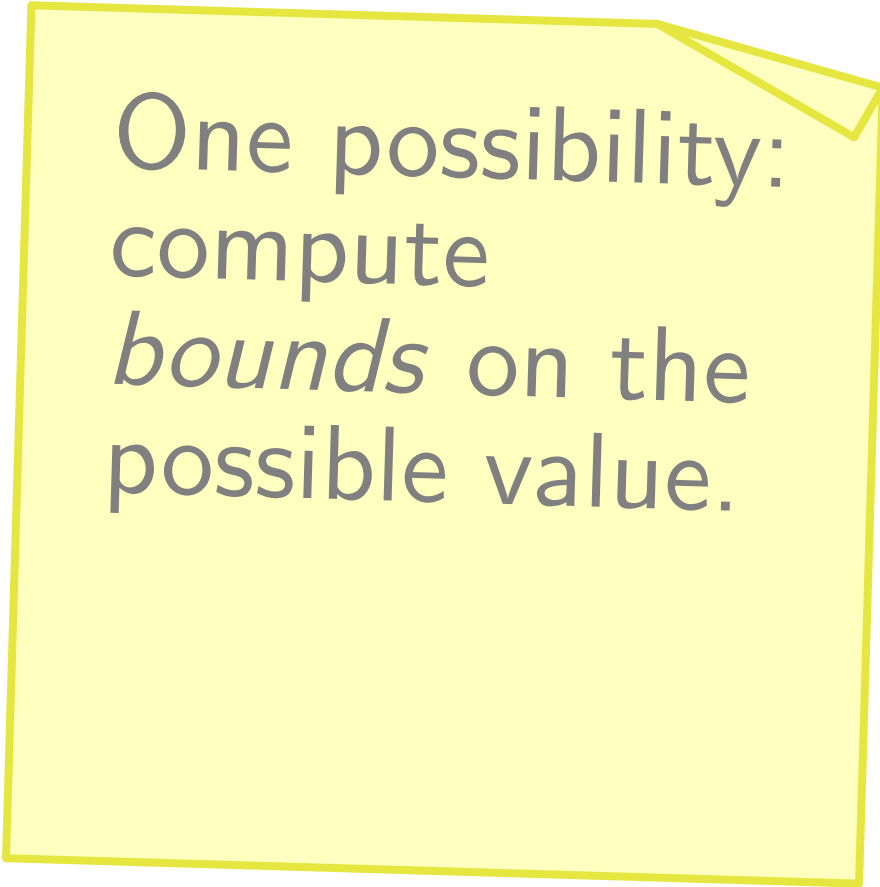
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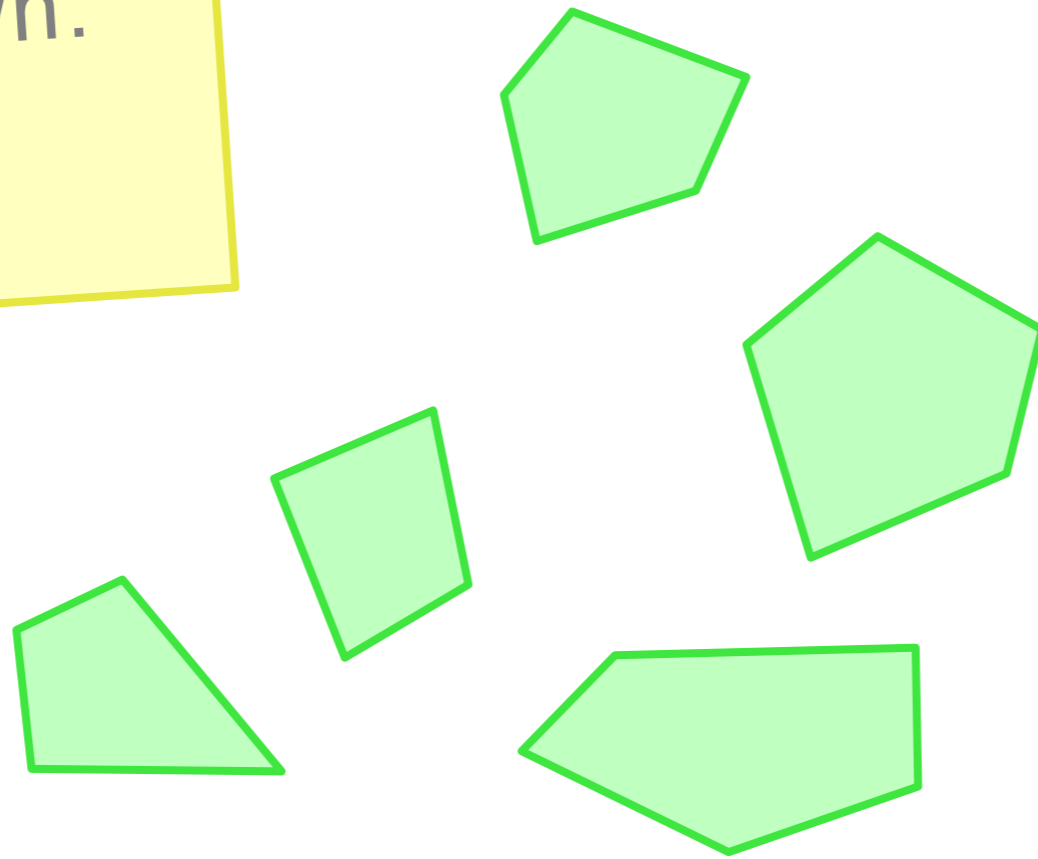


One possibility:  
compute  
*bounds* on the  
possible value.



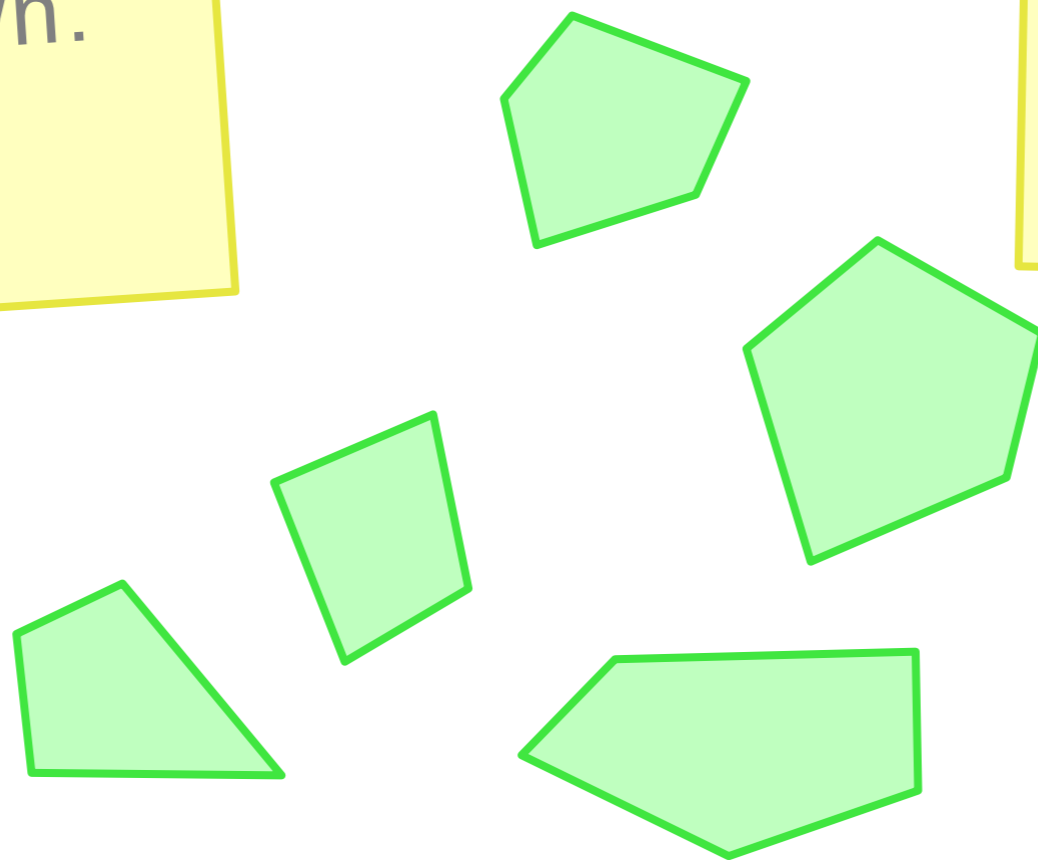
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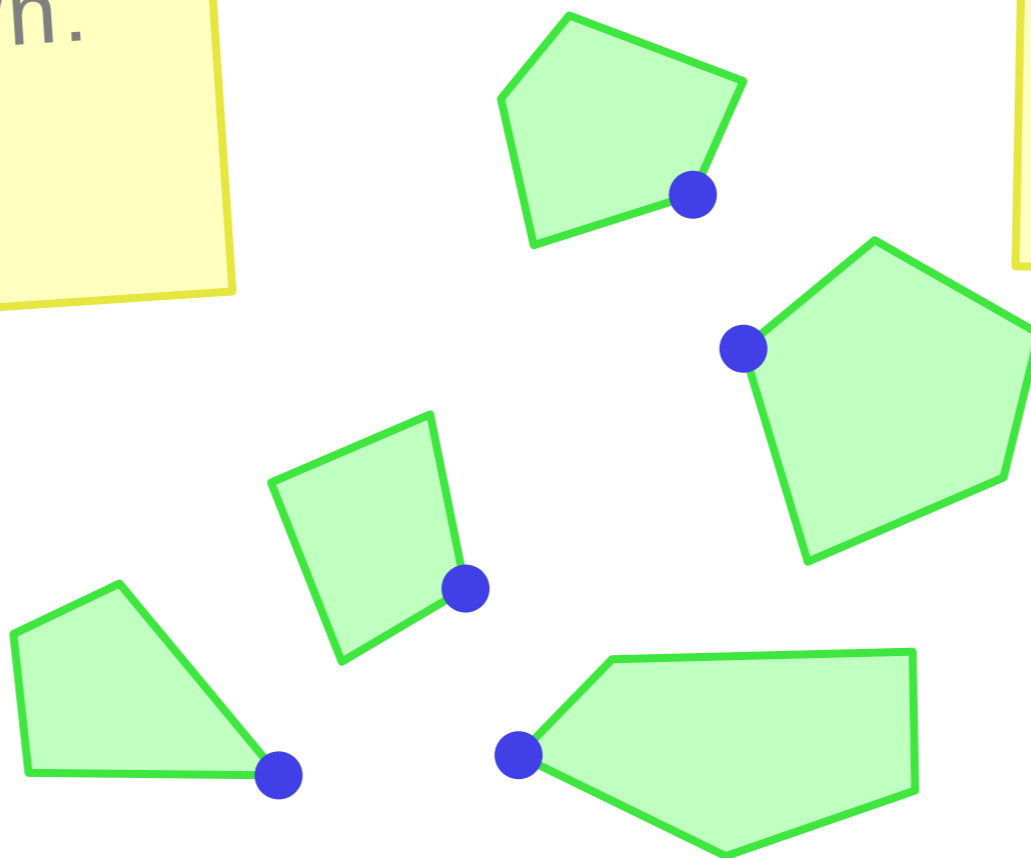
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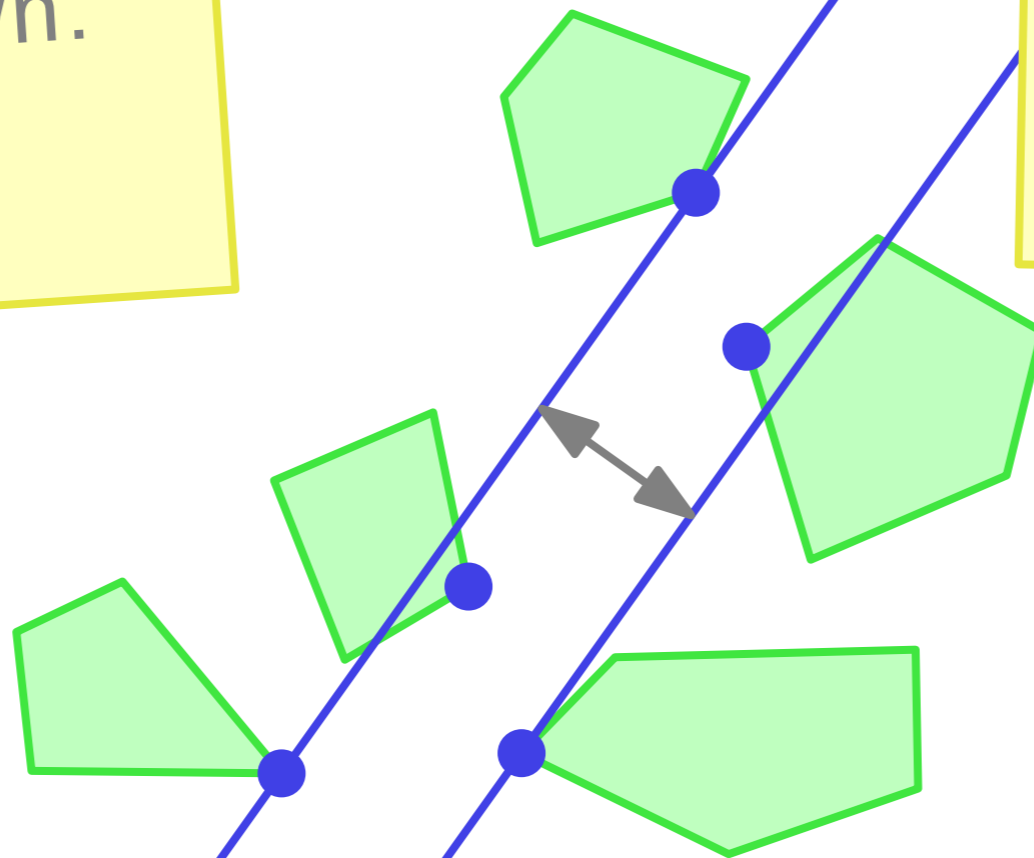
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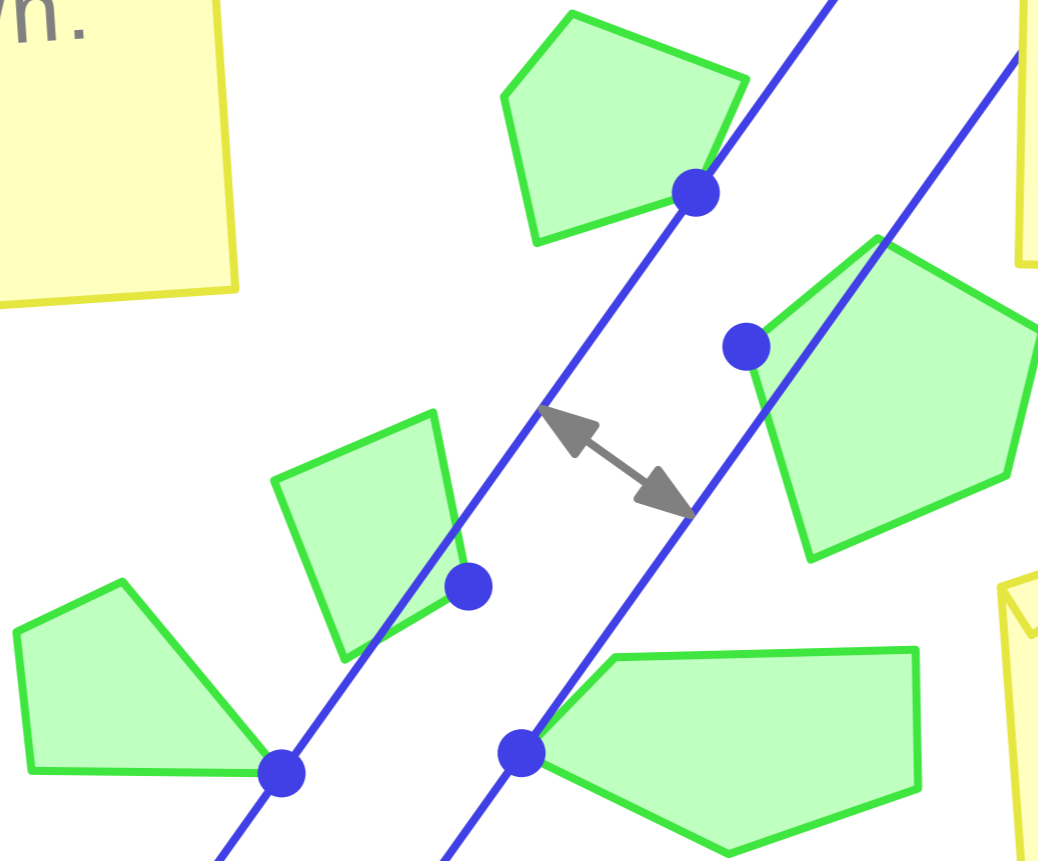
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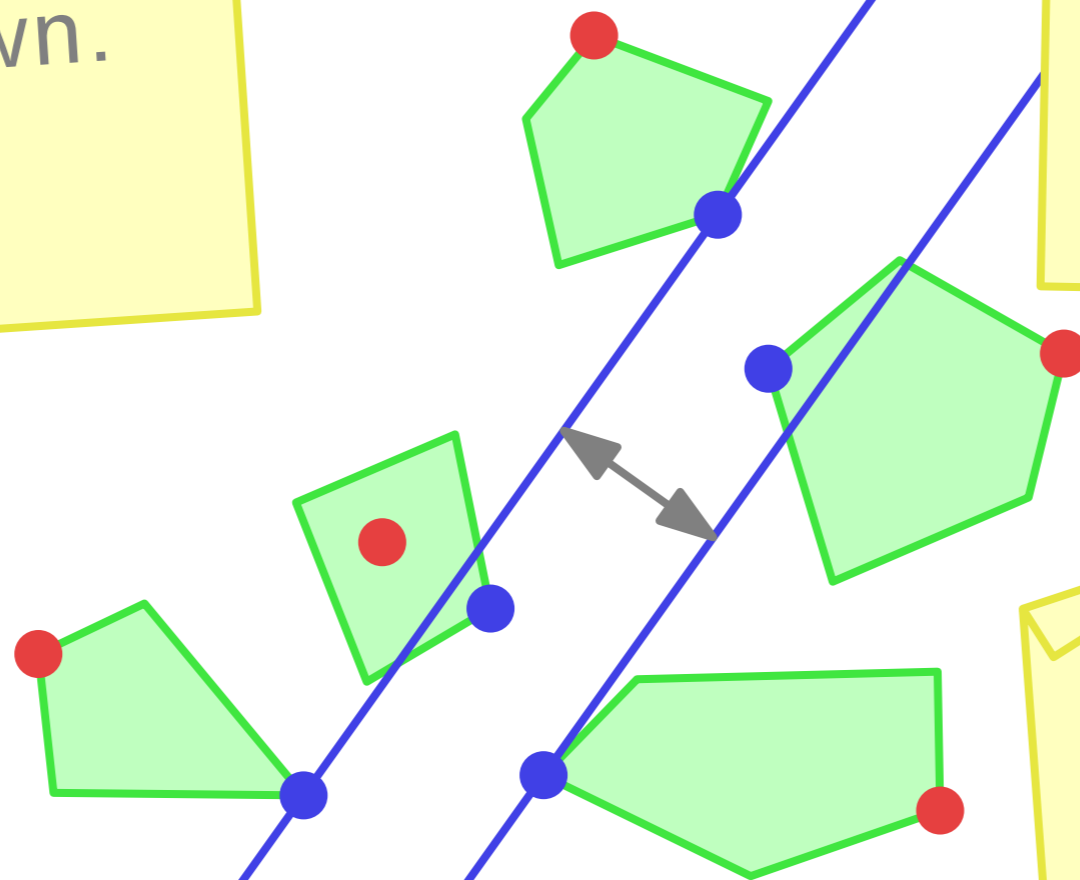
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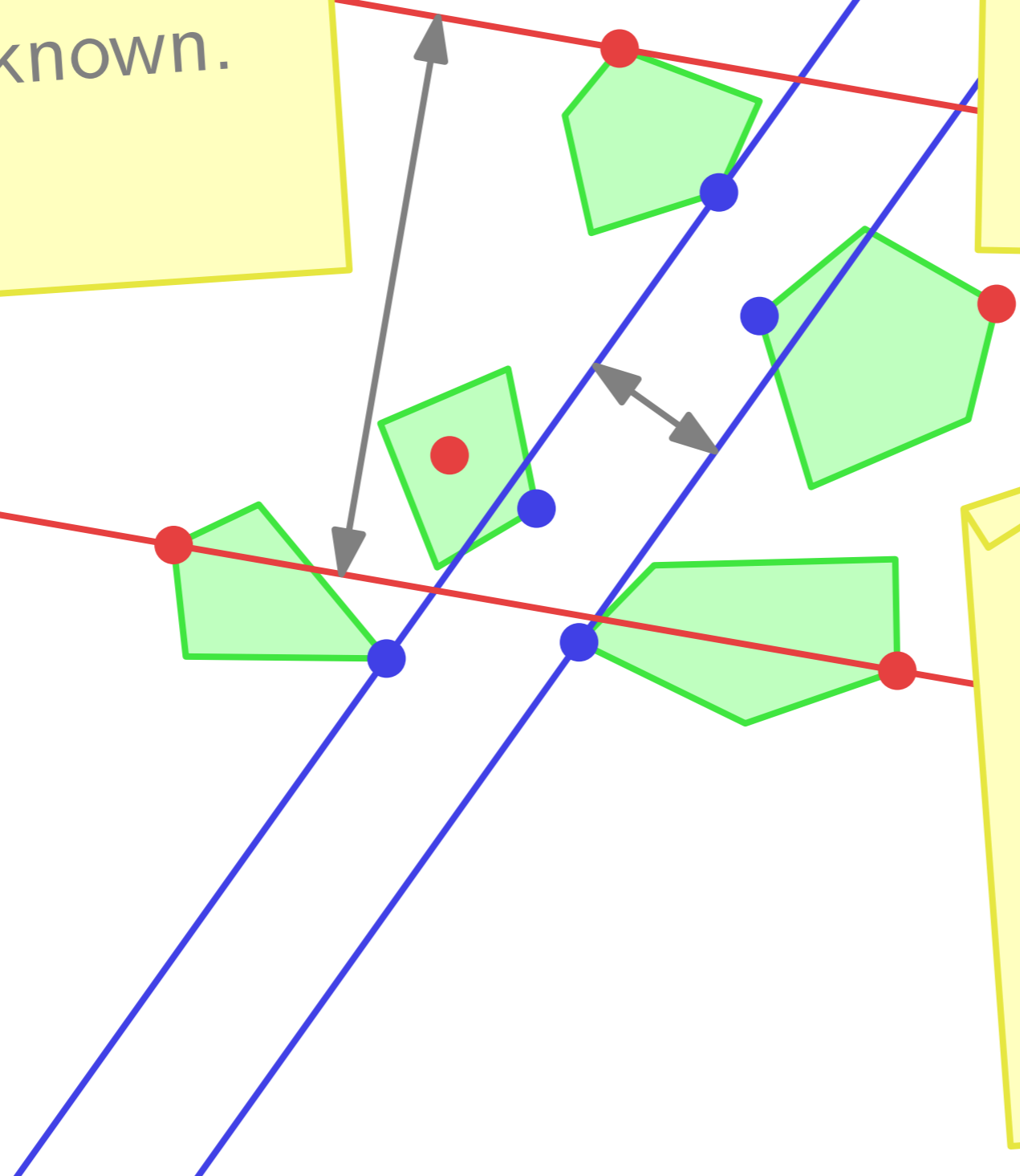


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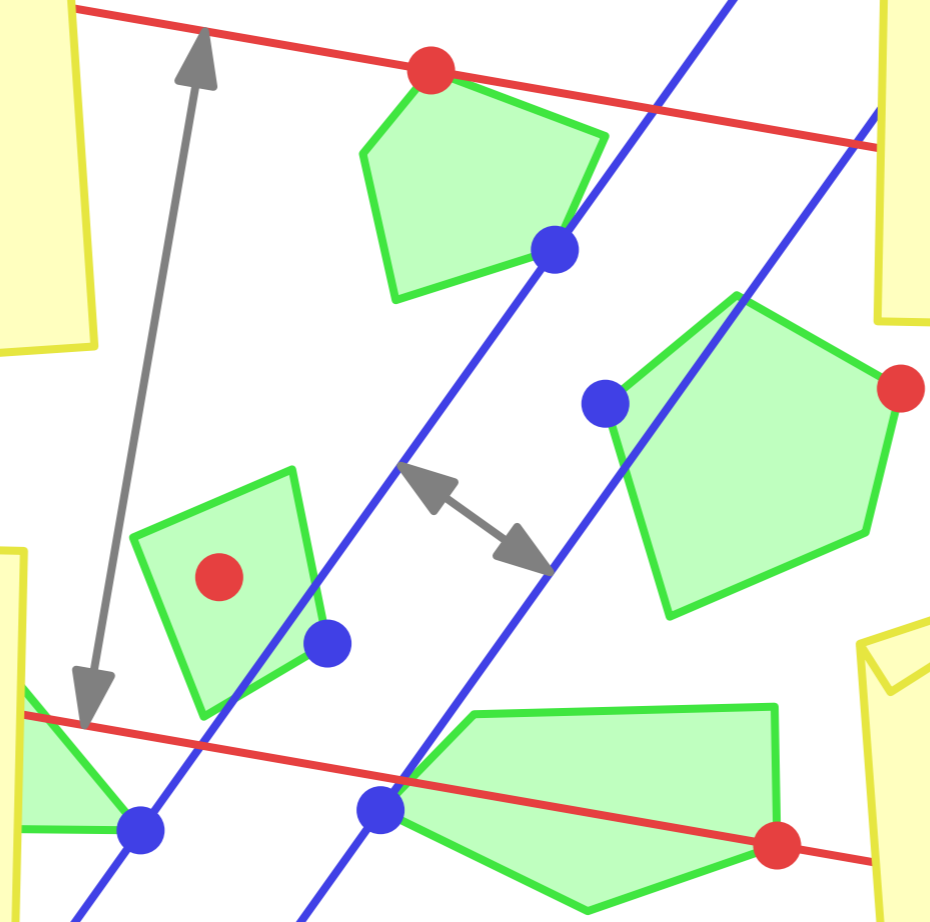


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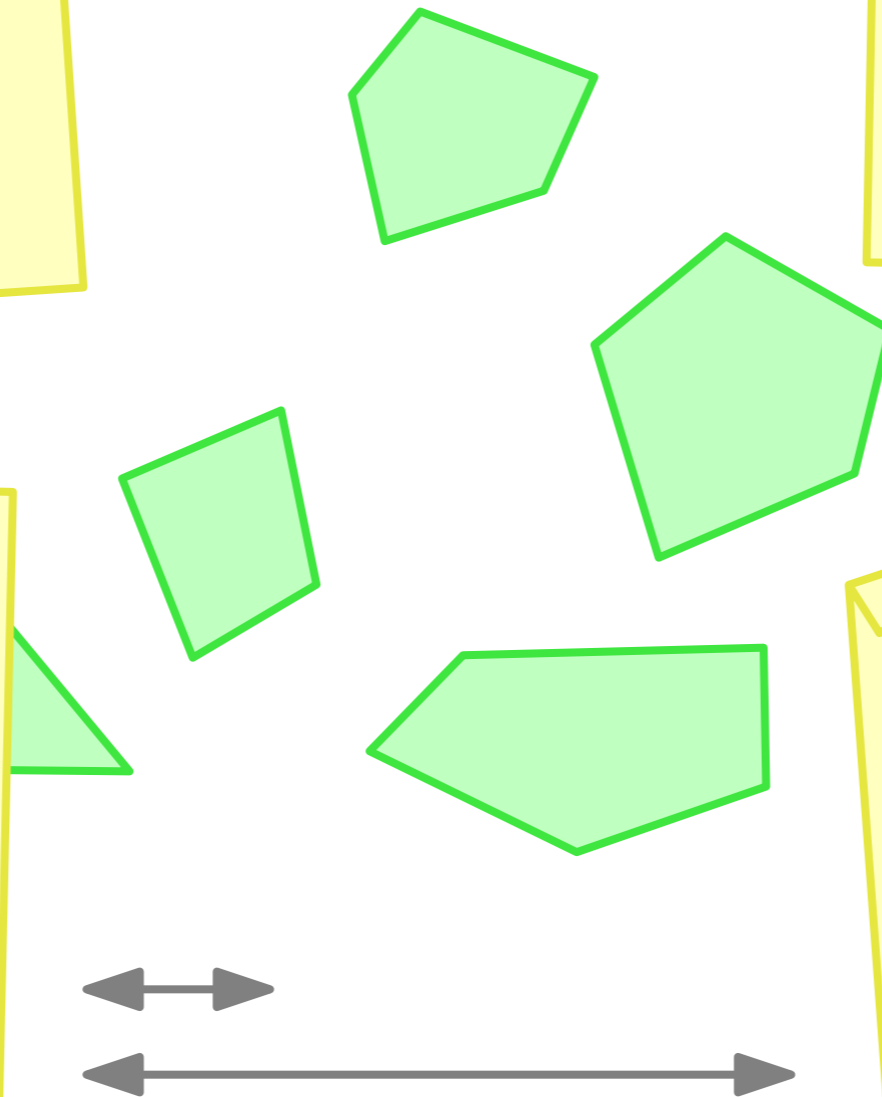


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What results of  
this kind are  
known?

Bounds for  
diameter and  
convex hull  
area in time  
 $O(n \log n)$ .

[Nagai &  
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Times range  
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[van Kreveld &  
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Bounds for diameter and convex hull area in time  $O(n \log n)$ .

[Nagai & Tokura, 2000]

Tight bounds for the same measures are harder to get, depending on imprecision level.

All these results only consider the planar case.

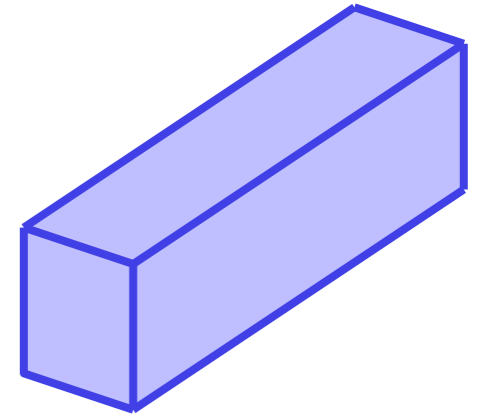
However, tight bounds are *tight*: there may not be any point sets that achieve them.

Times range from  $O(n \log n)$  to  $O(n^{13})$  or even NP-hard. [van Kreveld & ML, 2006]



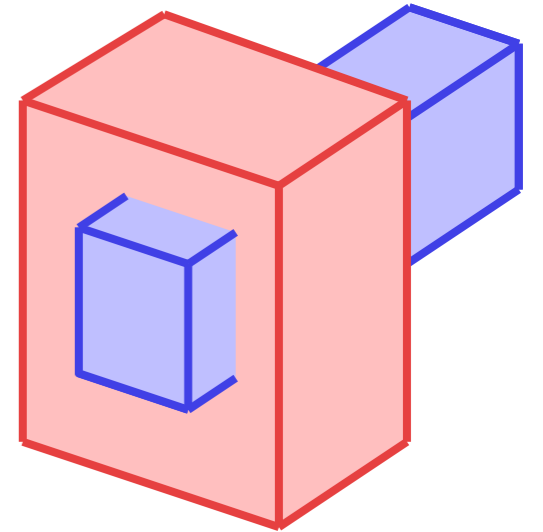
Can similar  
results be  
obtained in  
higher  
dimensions?

We show how  
to compute the  
smallest AABB  
in  $\mathbb{R}^3$  in  $O(n^7)$   
time.



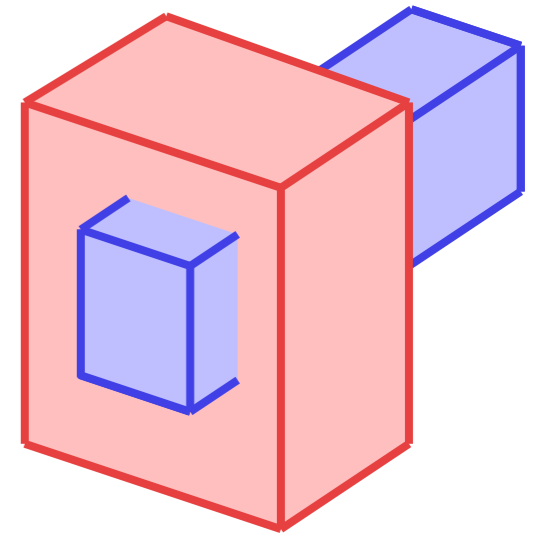
We show how to compute the smallest AABB in  $\mathbb{R}^3$  in  $O(n^7)$  time.

We can also compute the largest AABB in  $\mathbb{R}^d$  in  $O(d^2n + f(d))$  time.

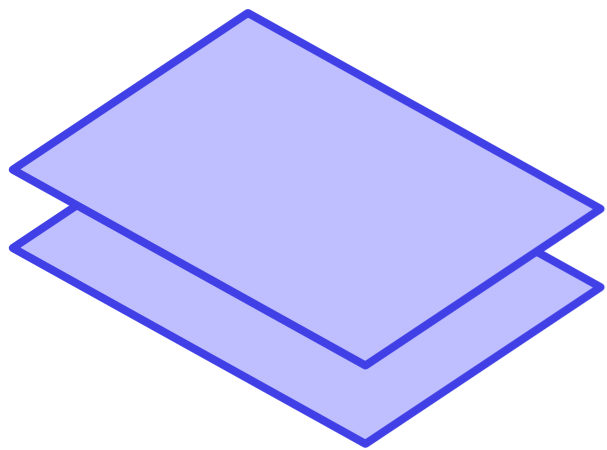


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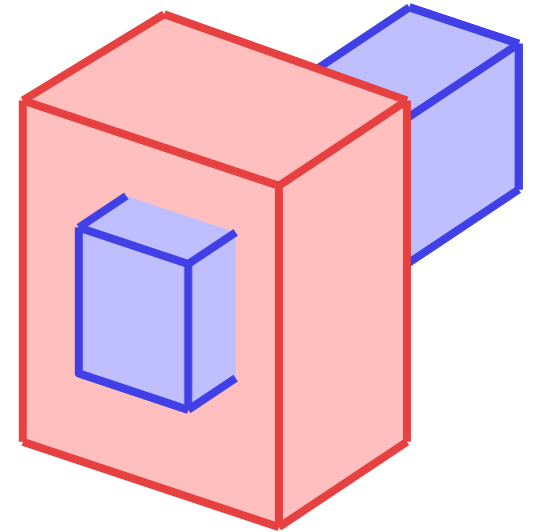


Next, we can compute the smallest width in  $\mathbb{R}^d$  in  $O(n^{2d-1})$  time.



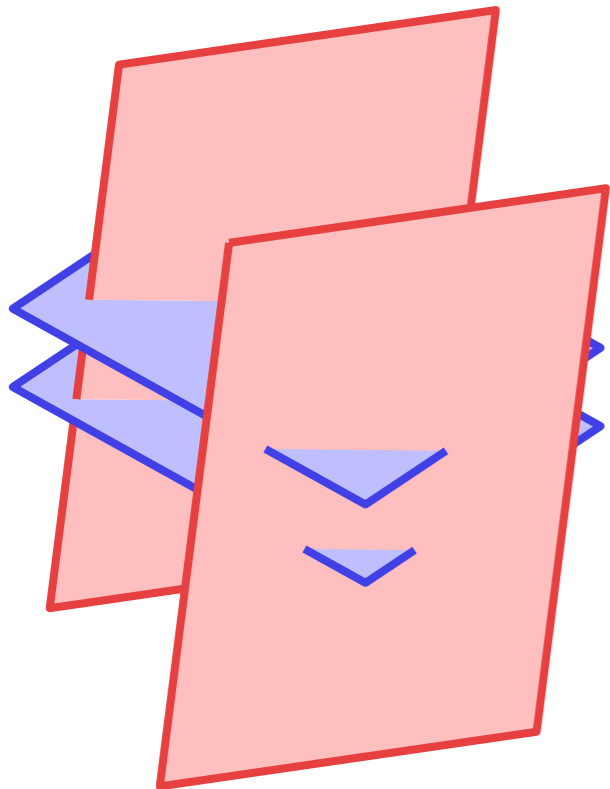
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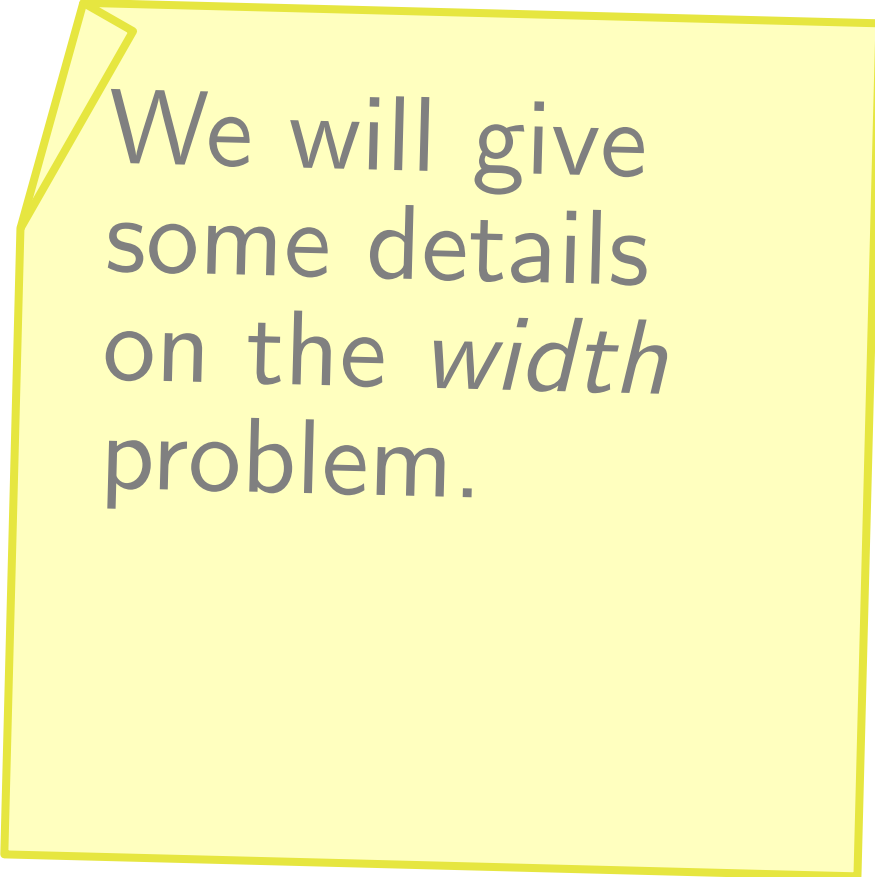
We can also compute the largest AABB in  $\mathbb{R}^d$  in  $O(d^2n + f(d))$  time.



Next, we can compute the smallest width in  $\mathbb{R}^d$  in  $O(n^{2d-1})$  time.

Finally, we show that computing the largest width in  $\mathbb{R}^d$  is NP-hard.

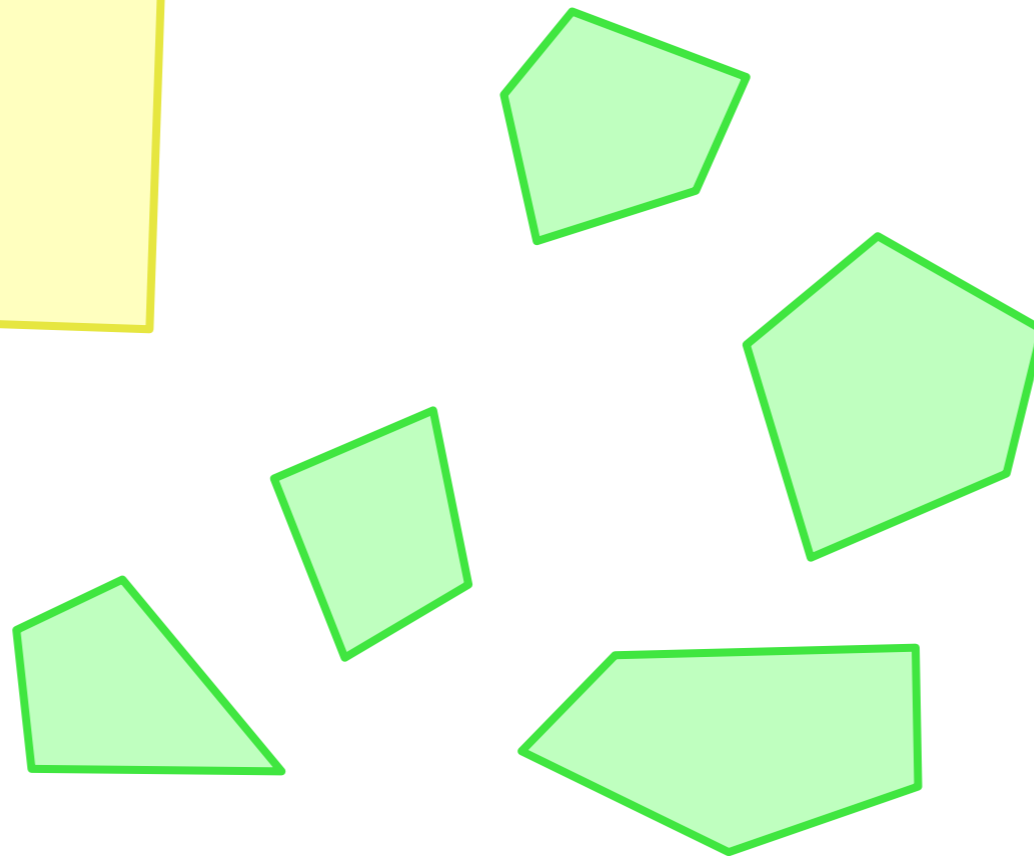




We will give  
some details  
on the *width*  
problem.

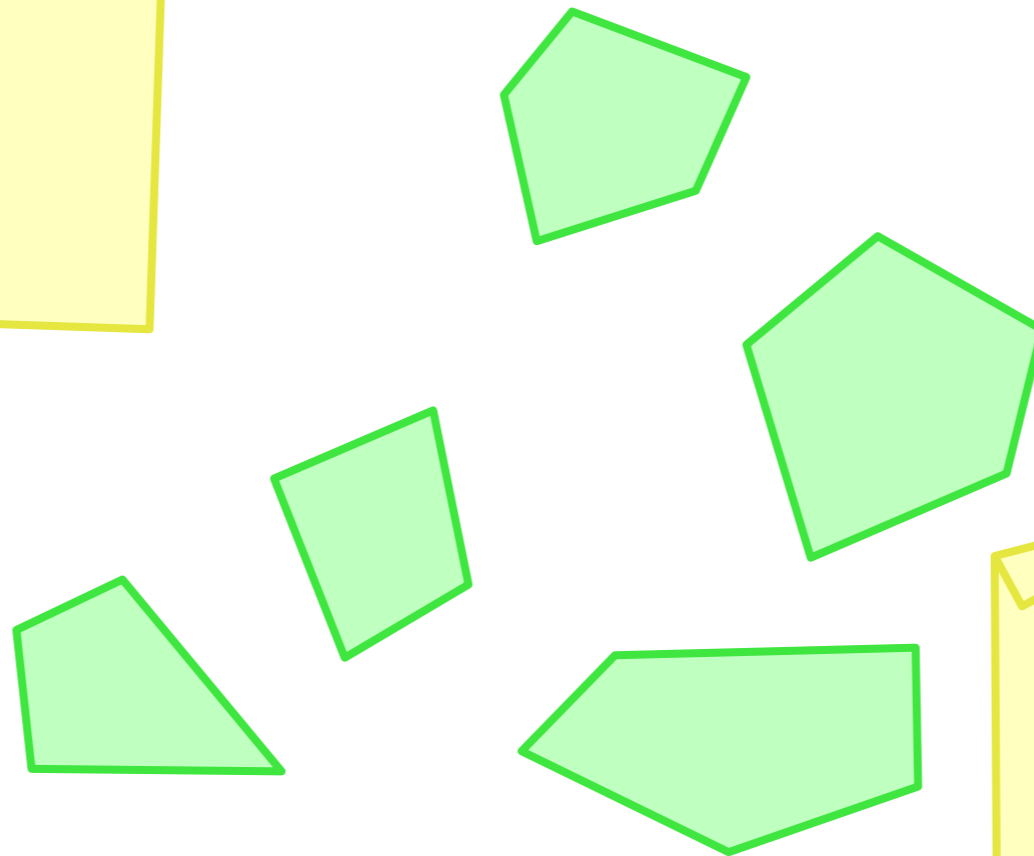
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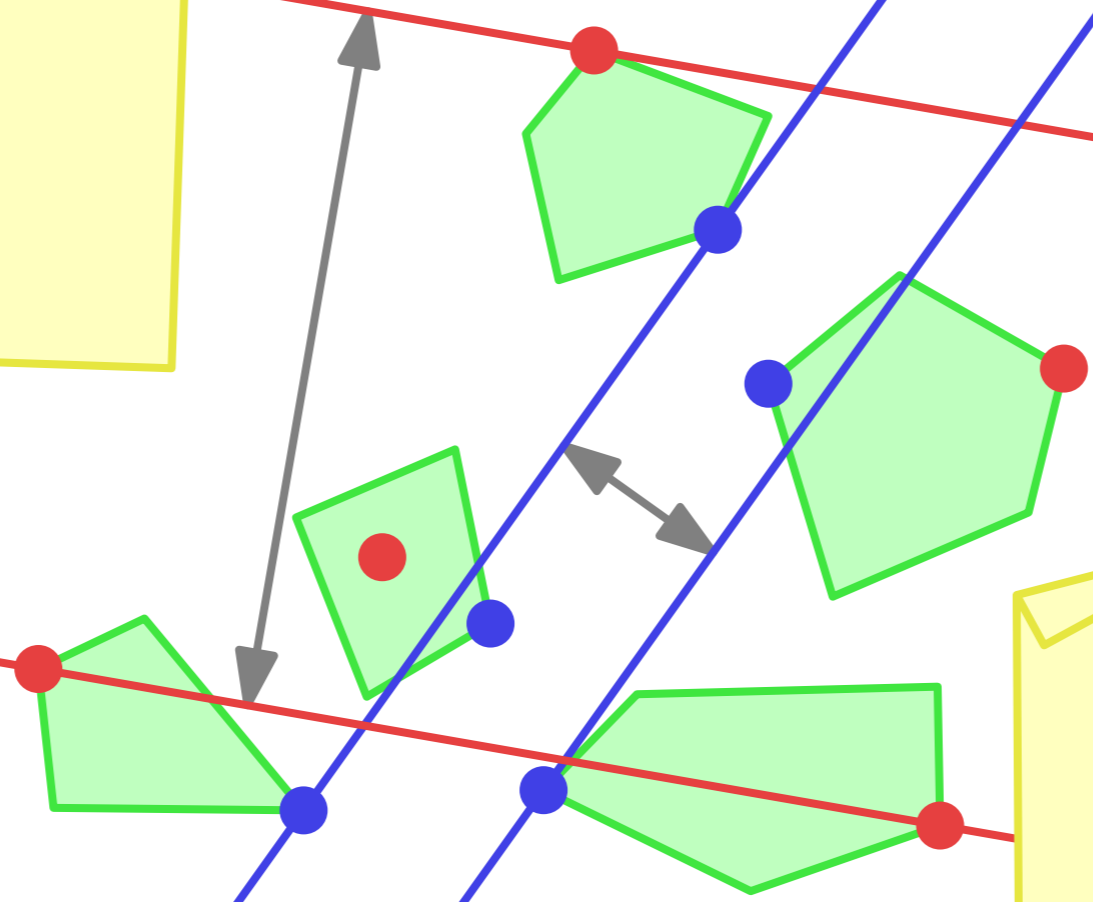


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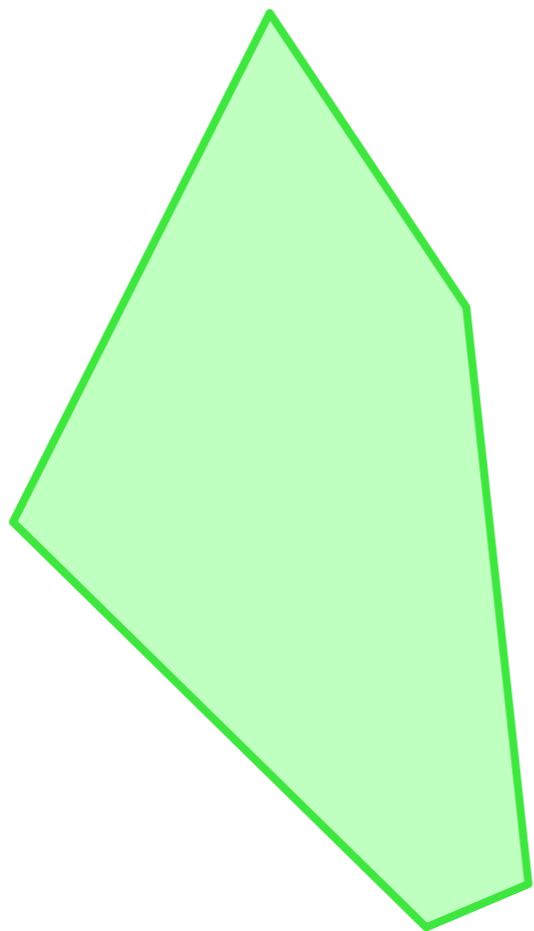
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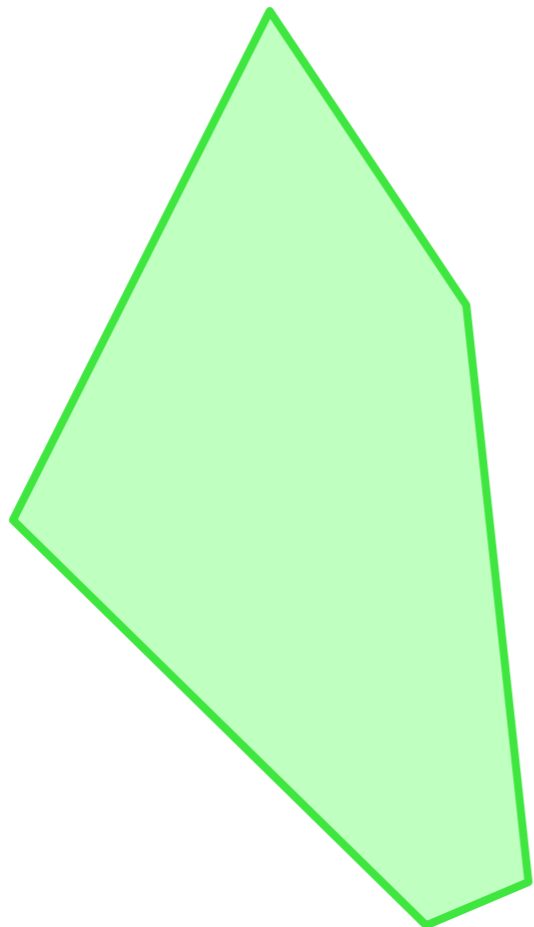
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Let's start with  
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technicalities.

$\mathbb{R}^d$

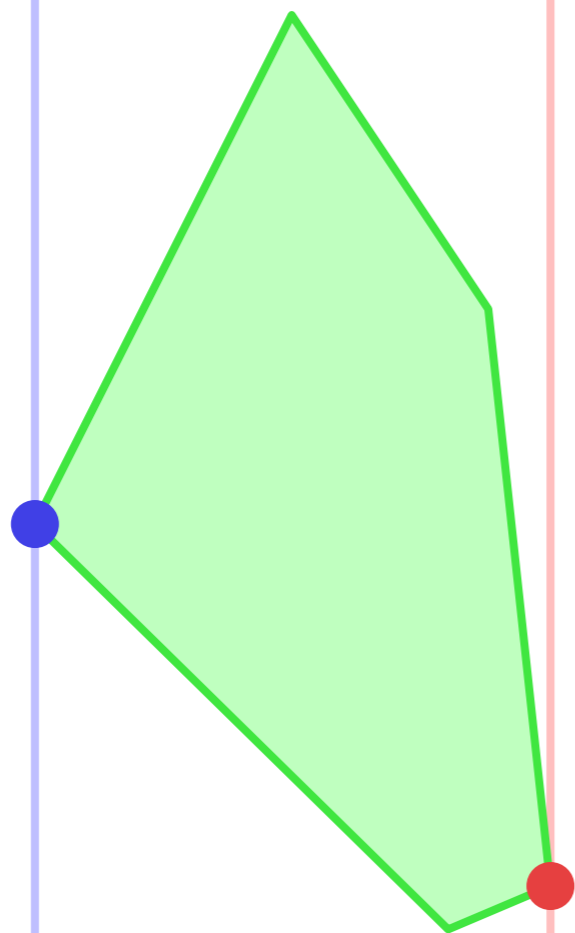


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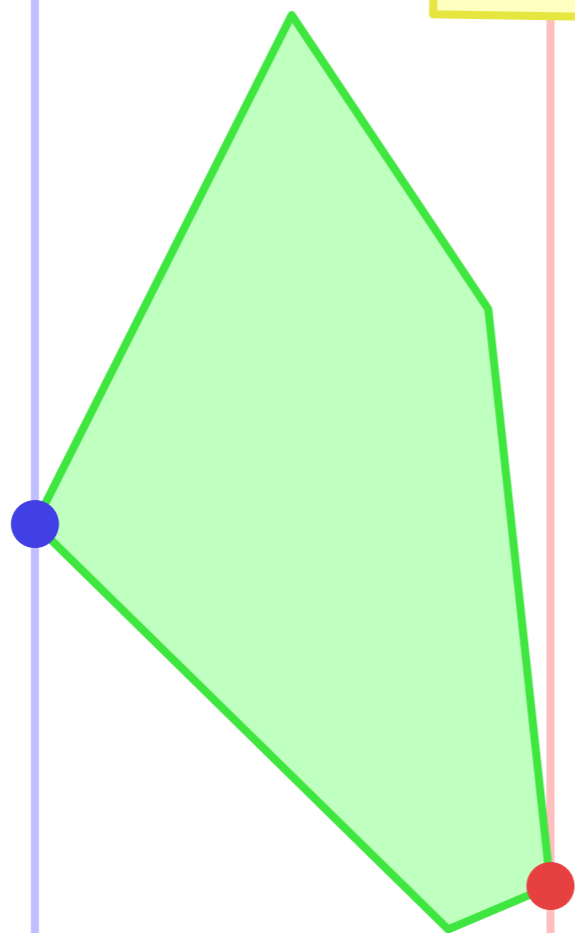
$\mathbb{R}^d$

$\mathcal{S}^{d-1}$



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The sphere of directions is subdivided depending on which points define it.



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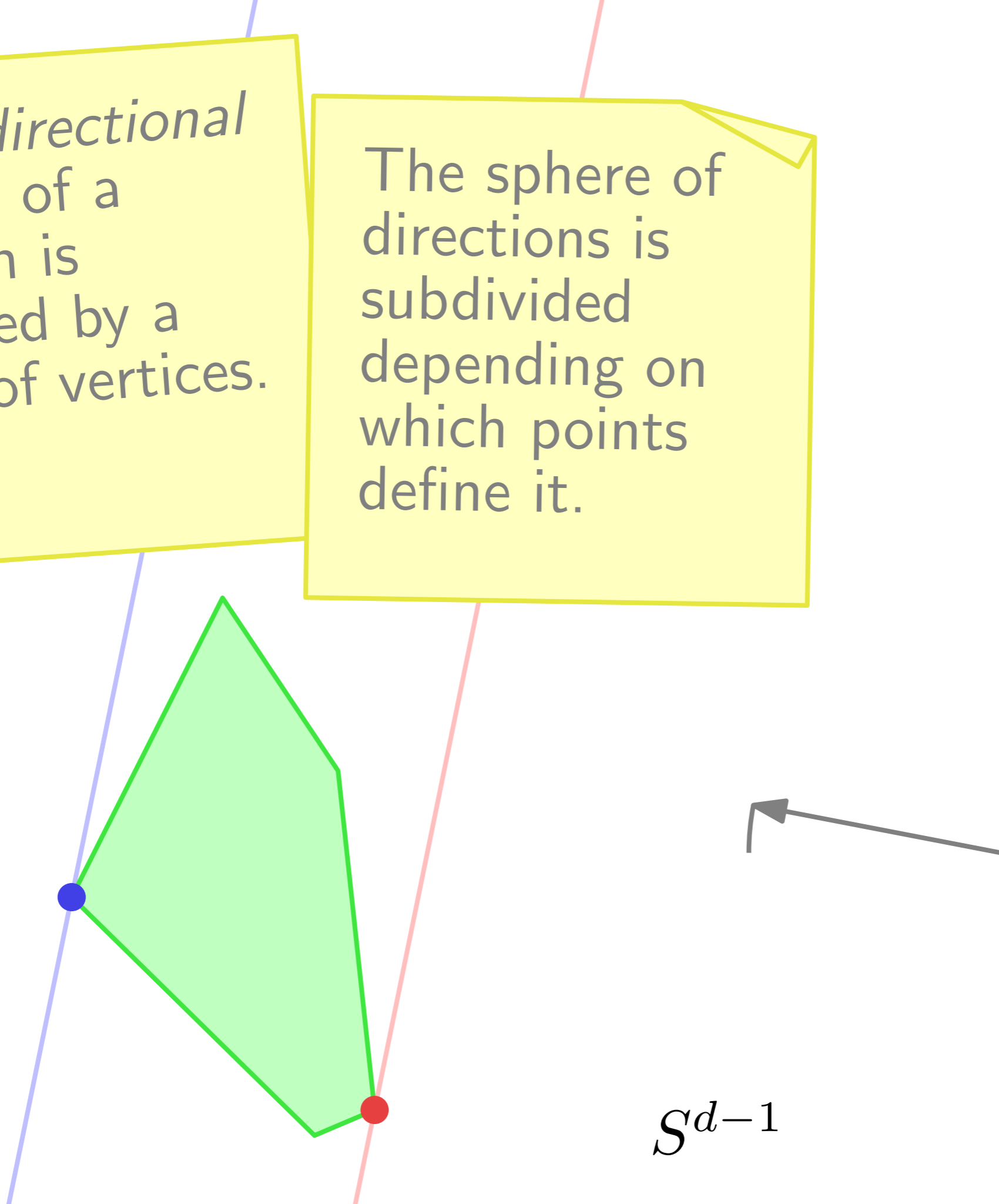
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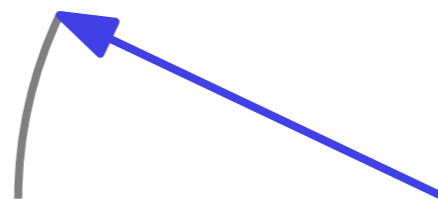


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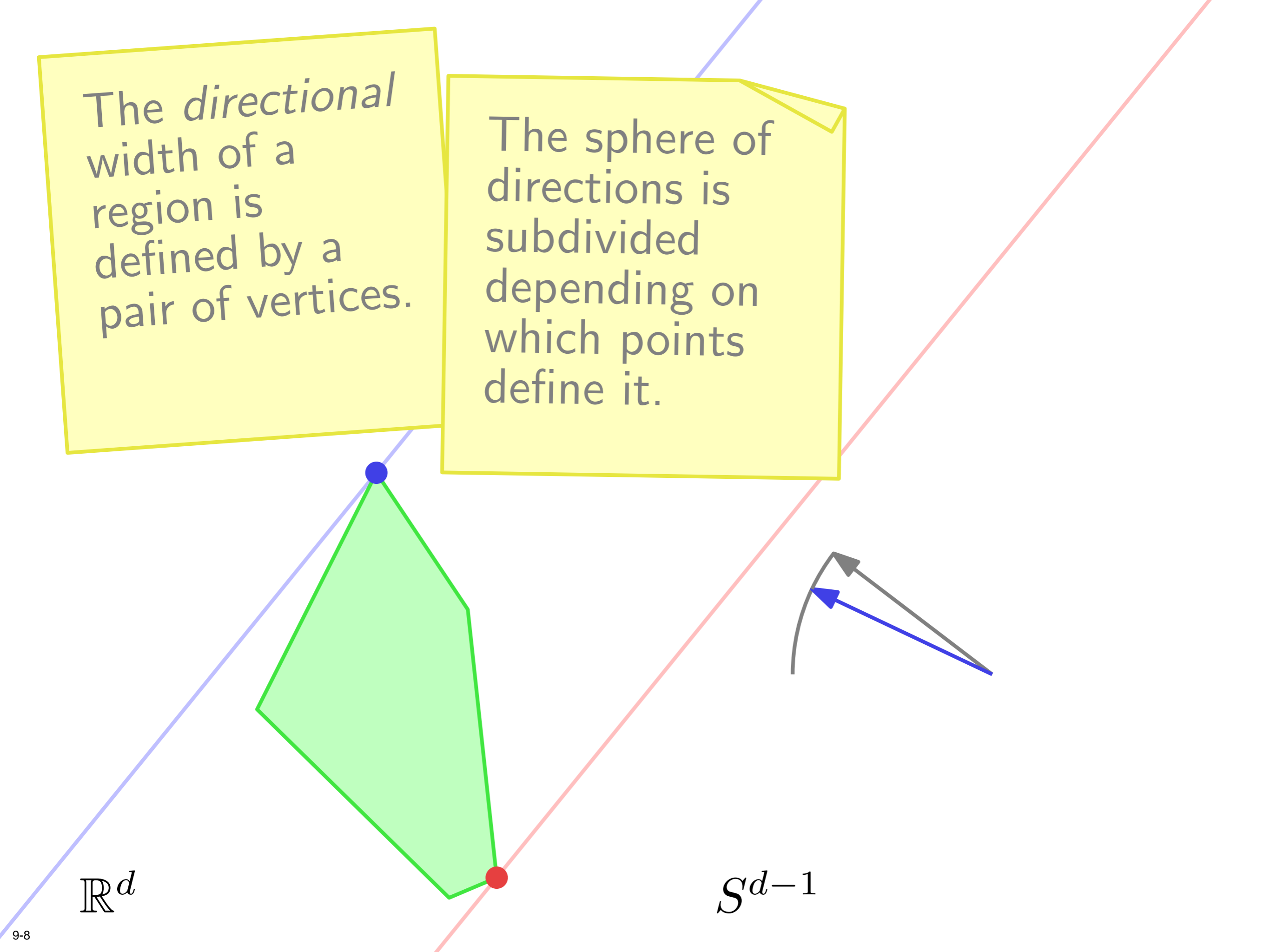
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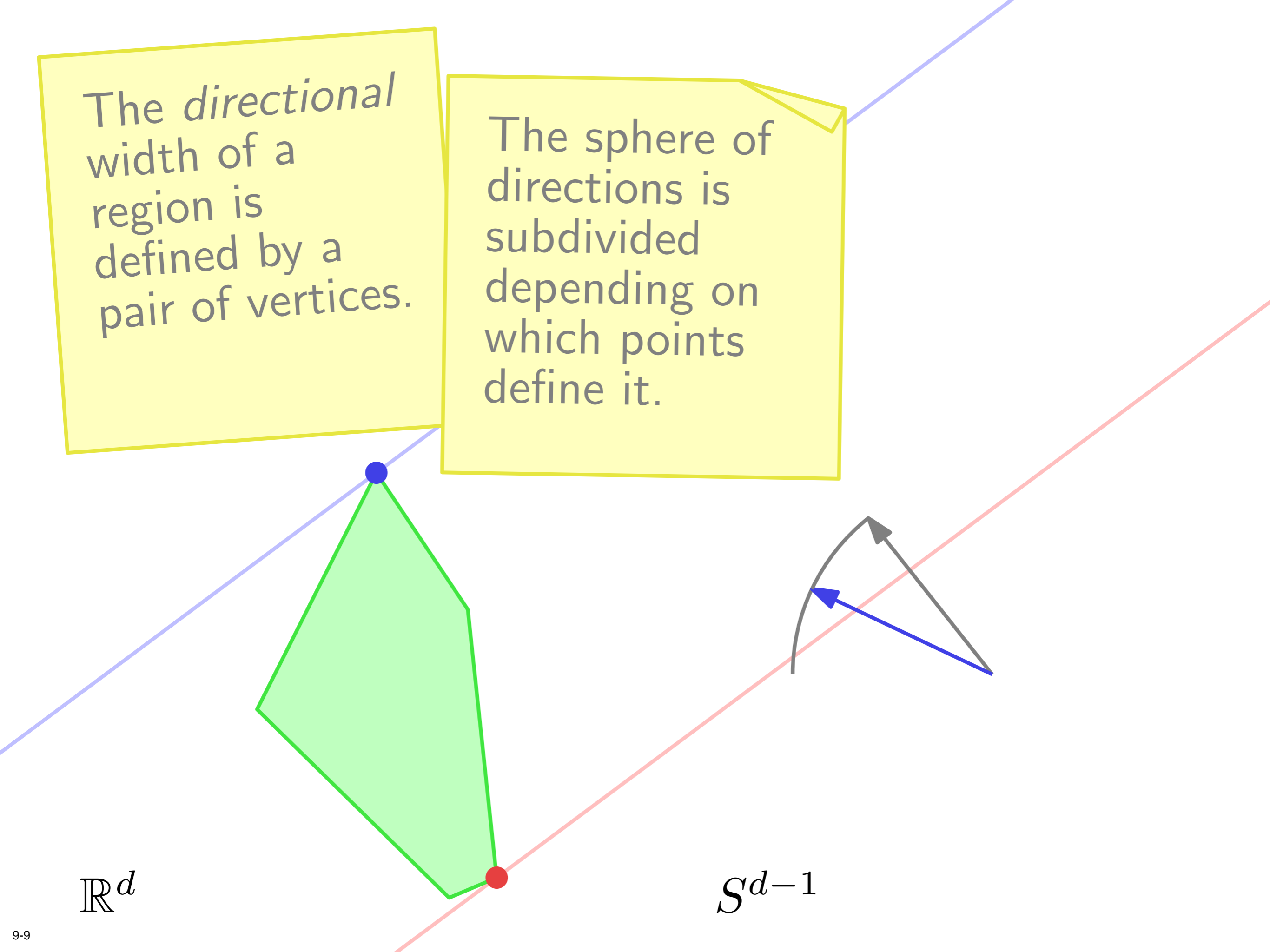


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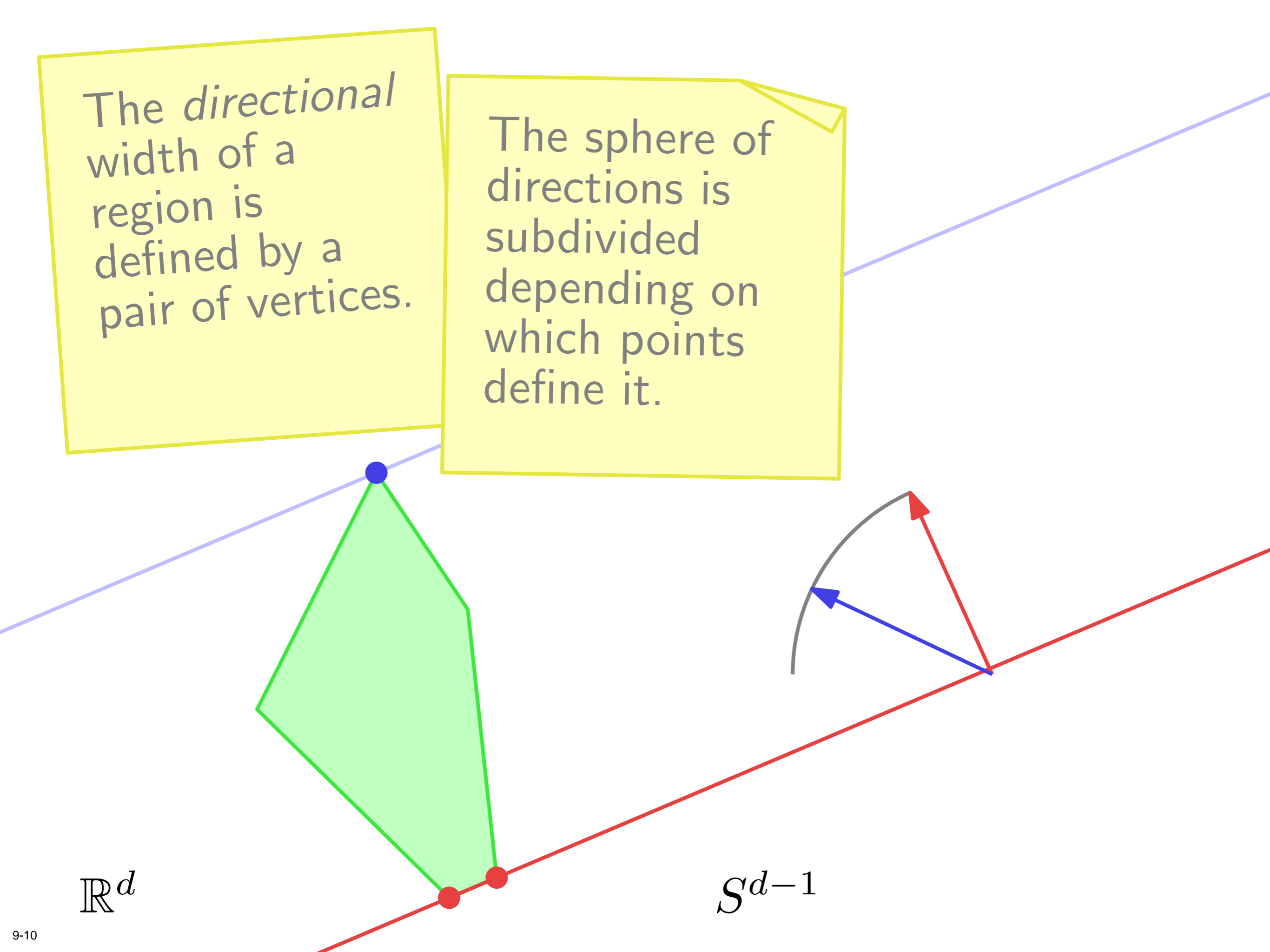


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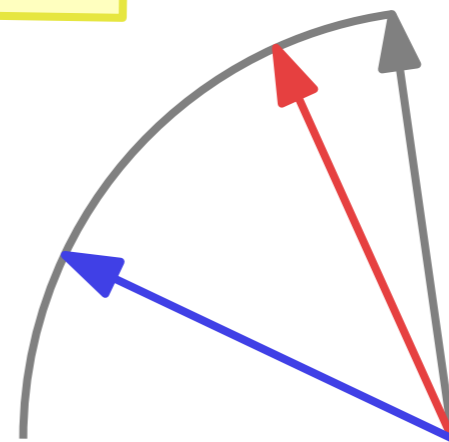
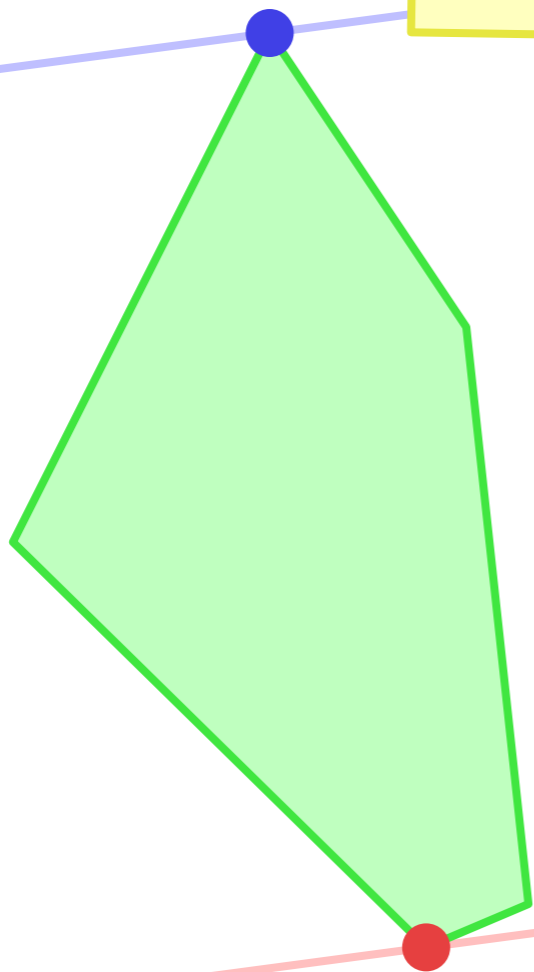


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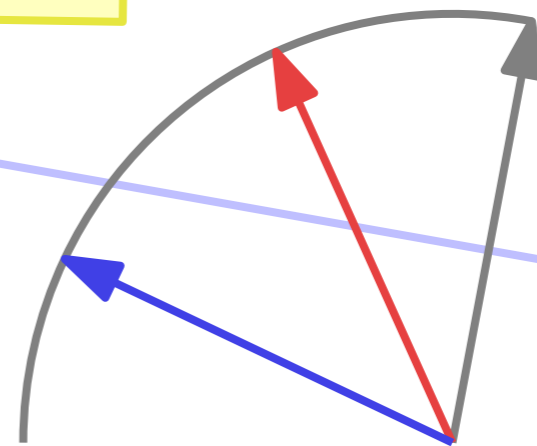
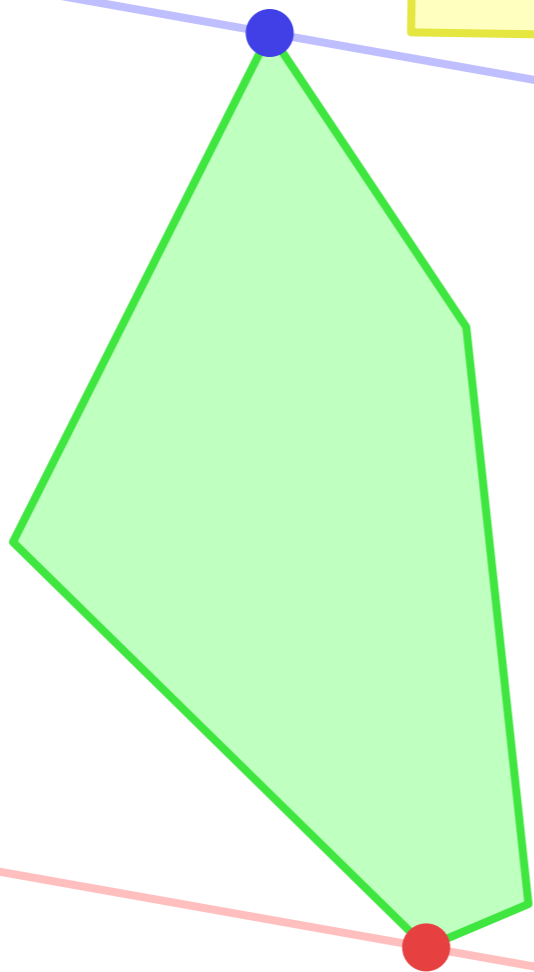


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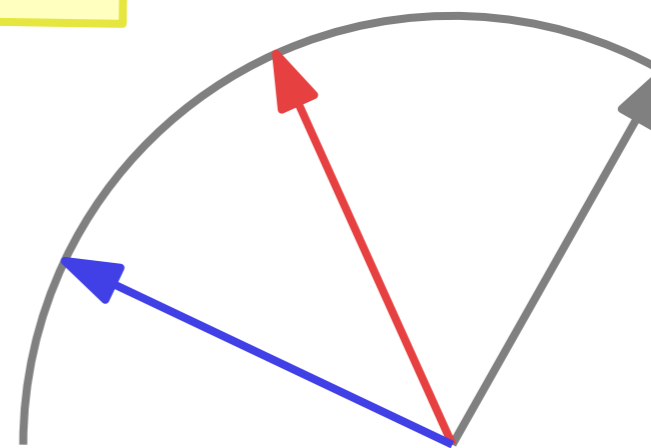
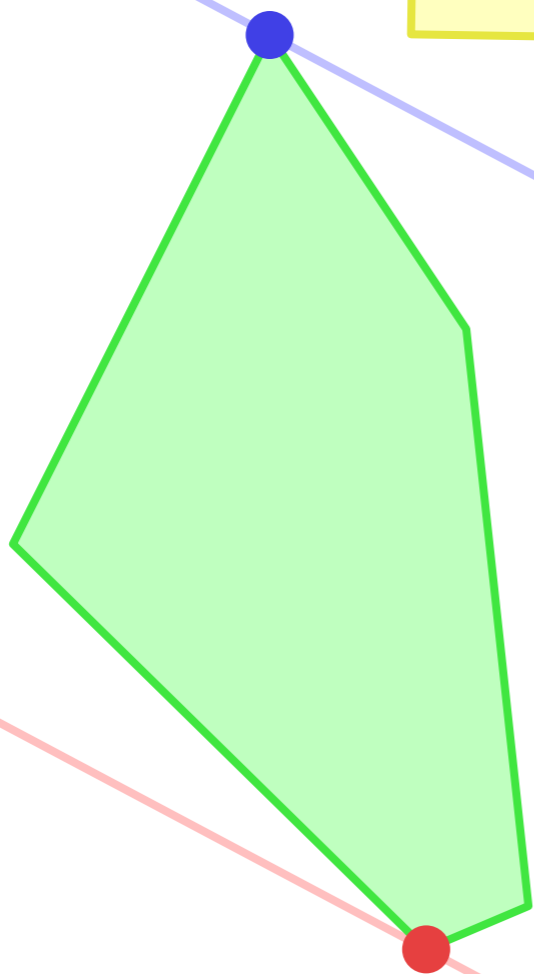


$\mathbb{R}^d$

$S^{d-1}$

The *directional* width of a region is defined by a pair of vertices.

The sphere of directions is subdivided depending on which points define it.



$\mathbb{R}^d$

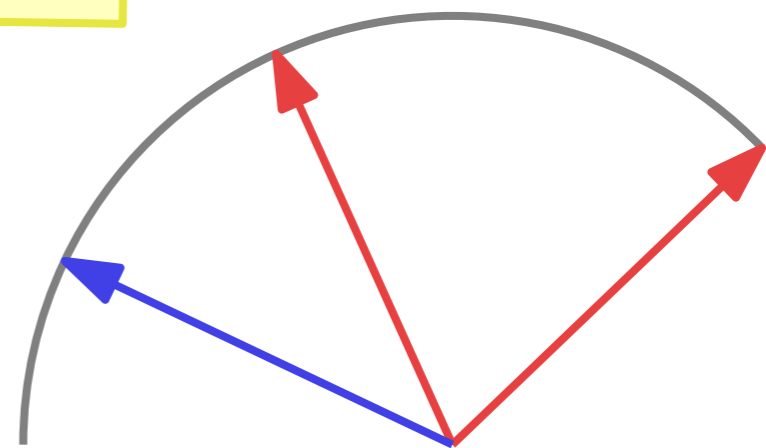
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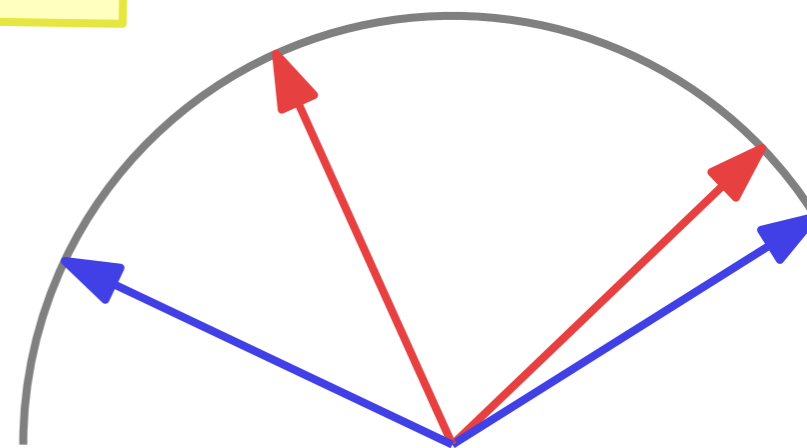
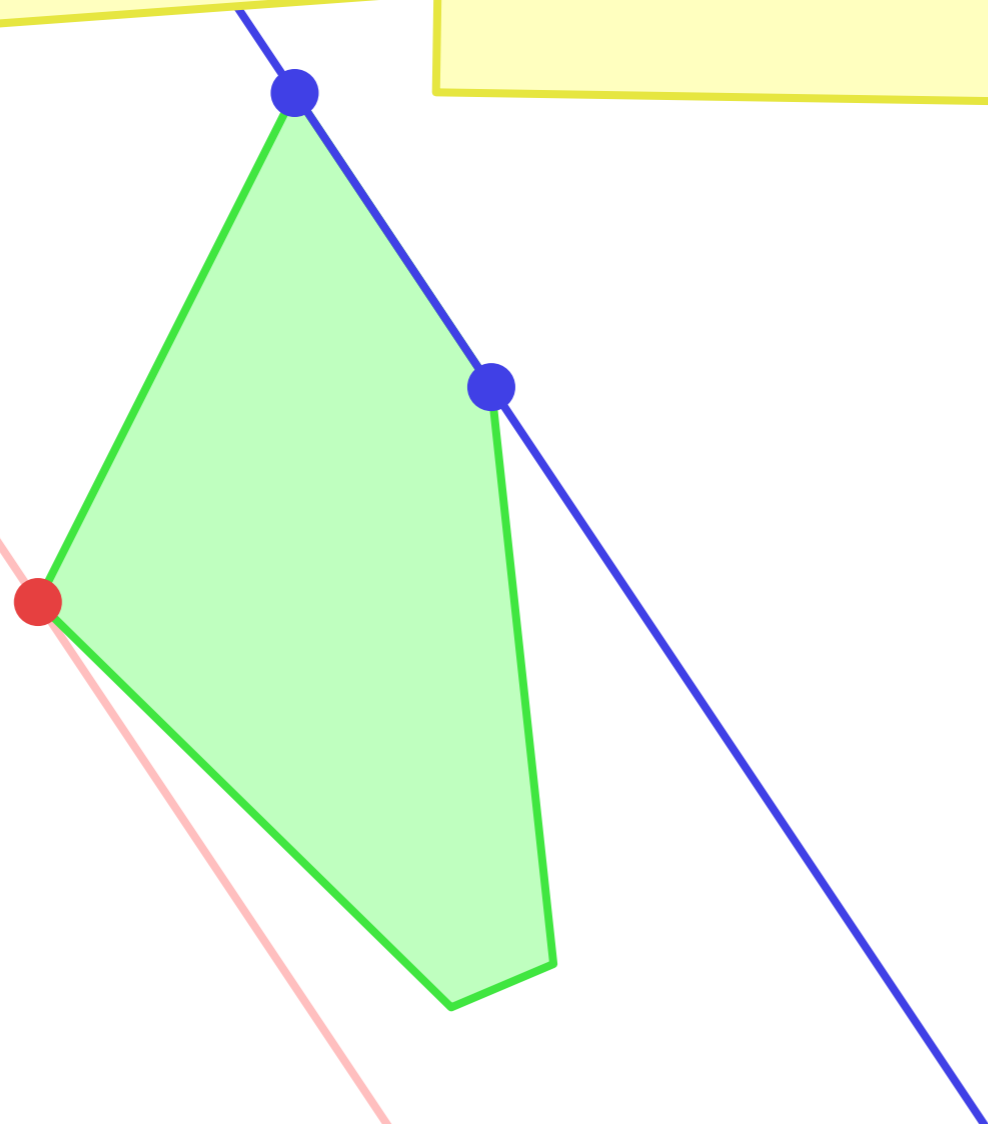
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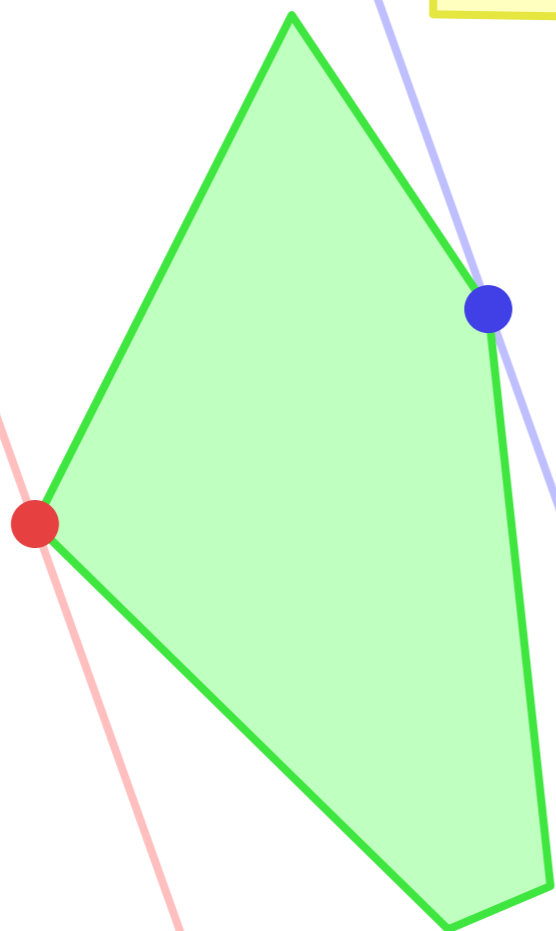


$\mathbb{R}^d$

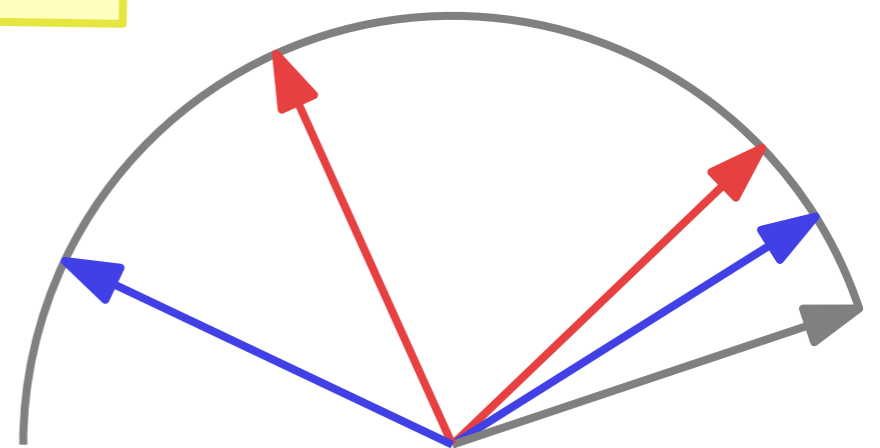
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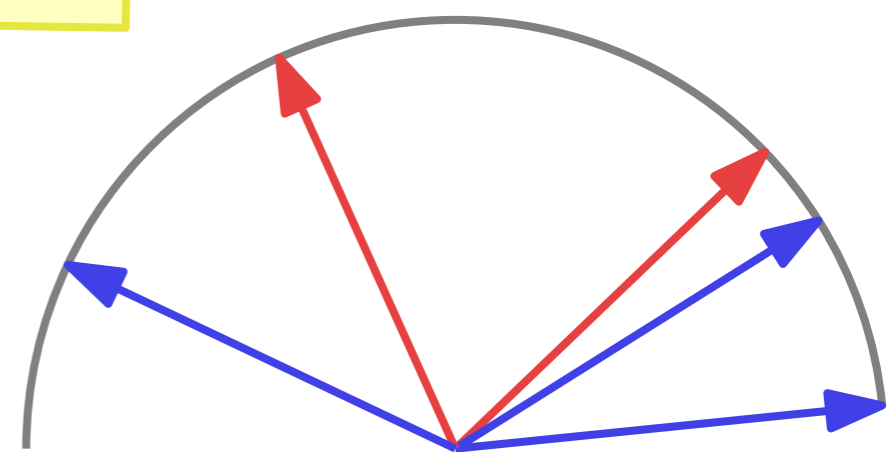
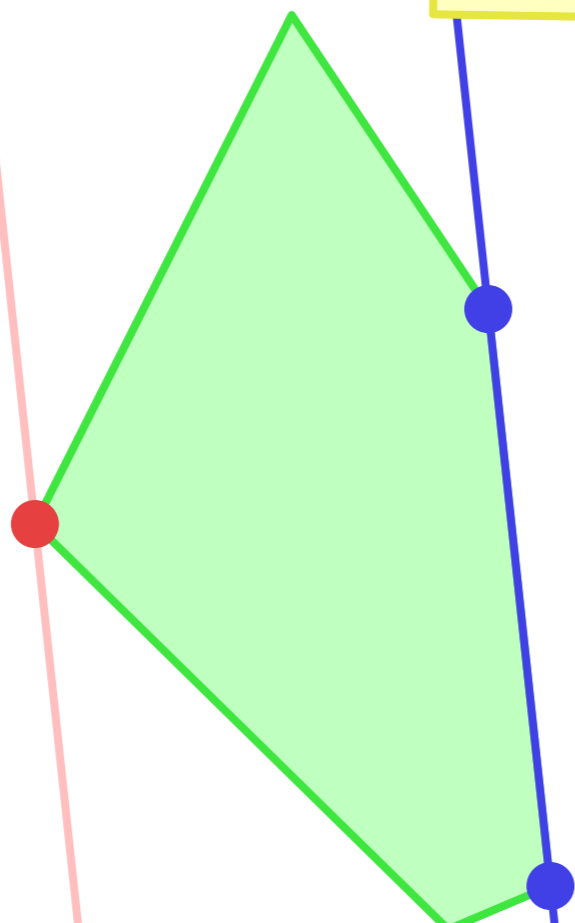
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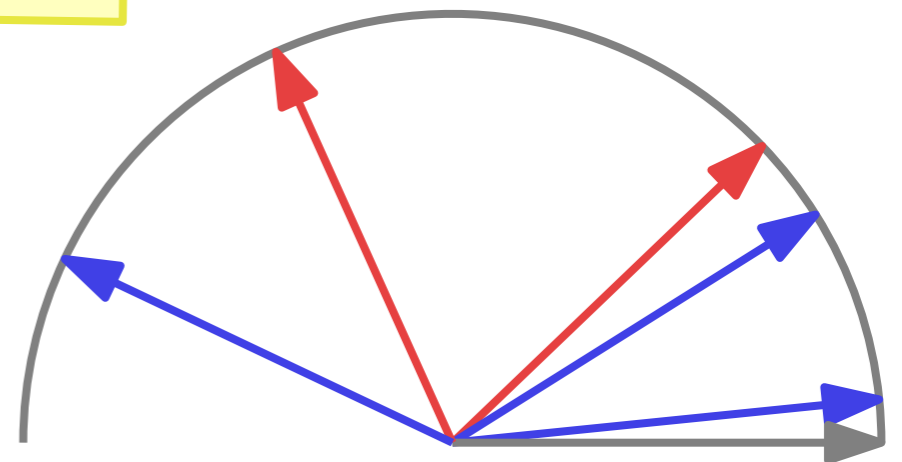
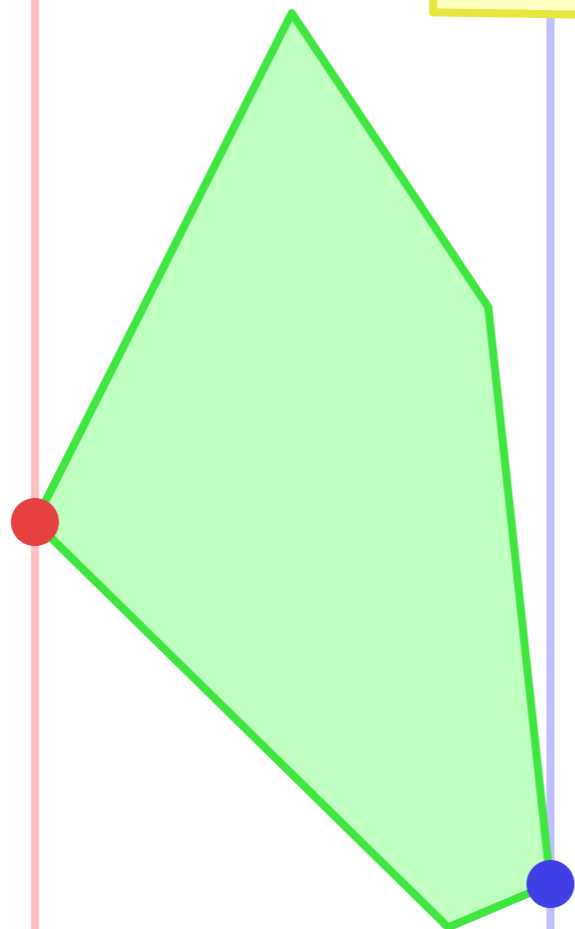


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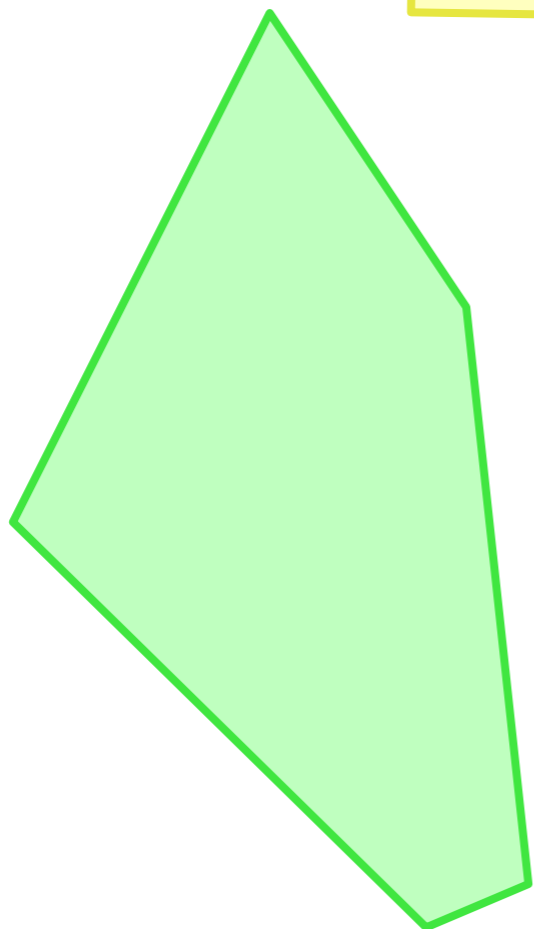


$\mathbb{R}^d$

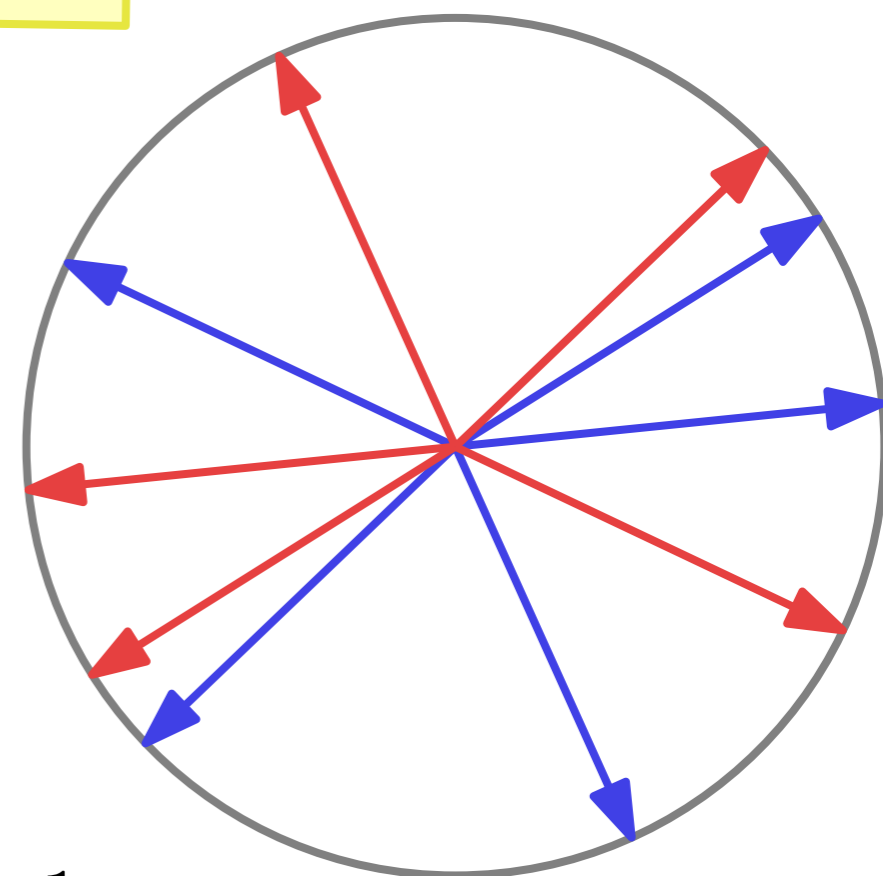
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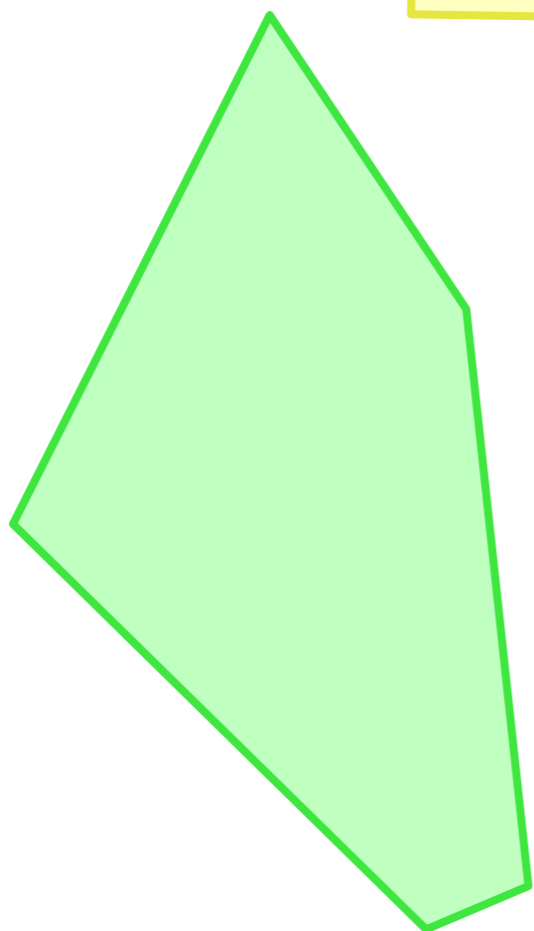
$\mathbb{R}^d$



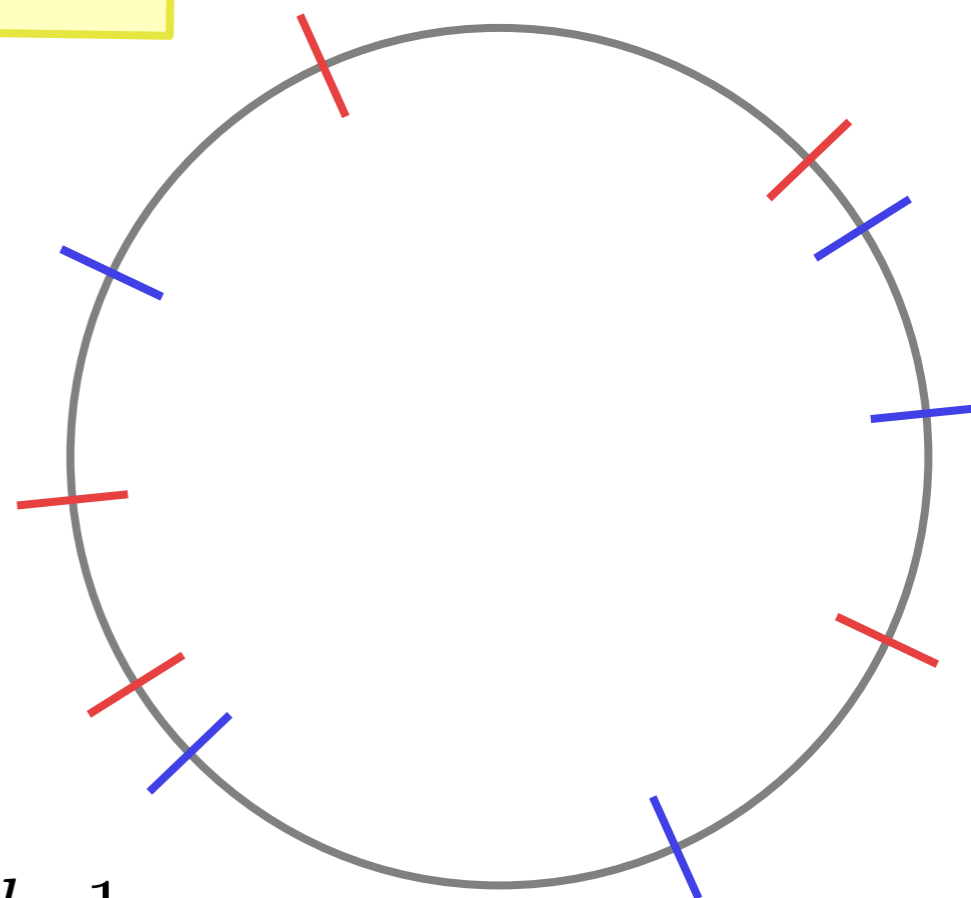
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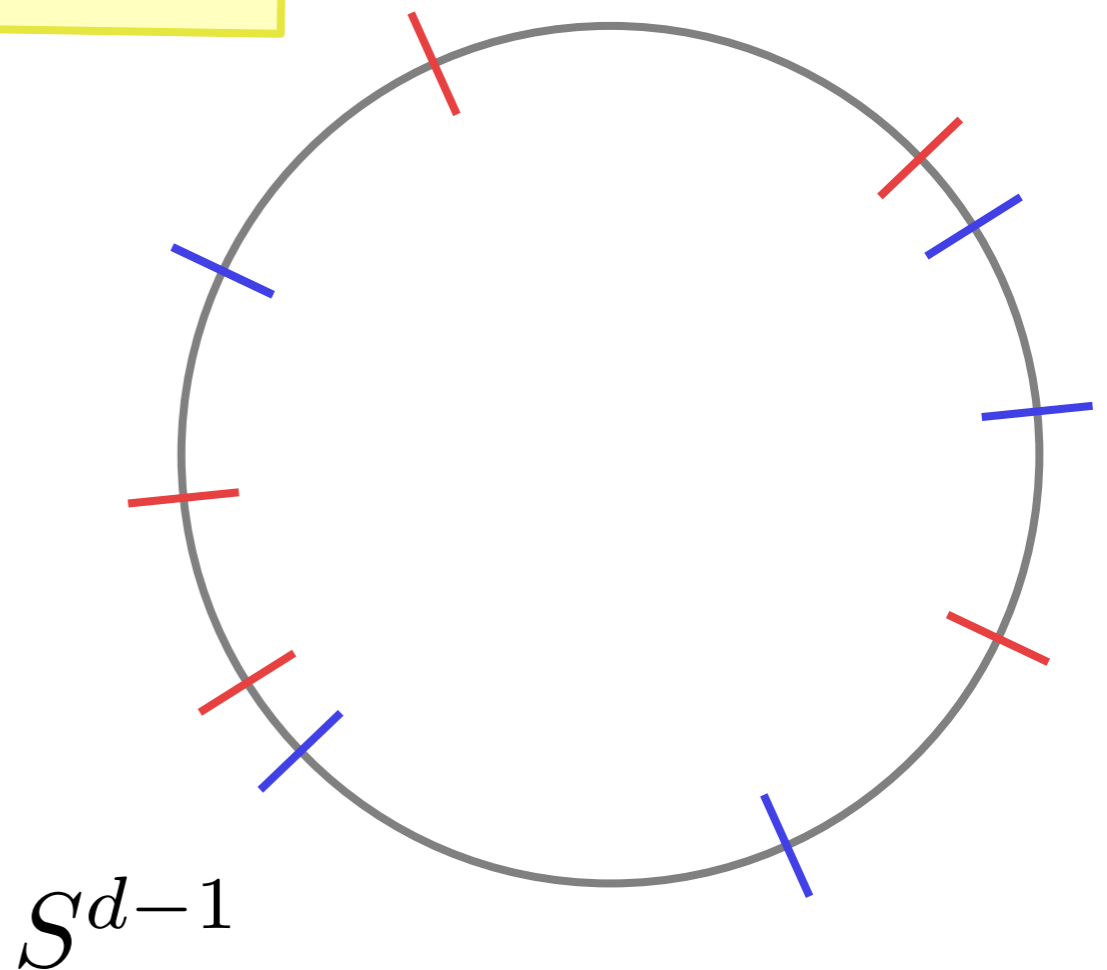
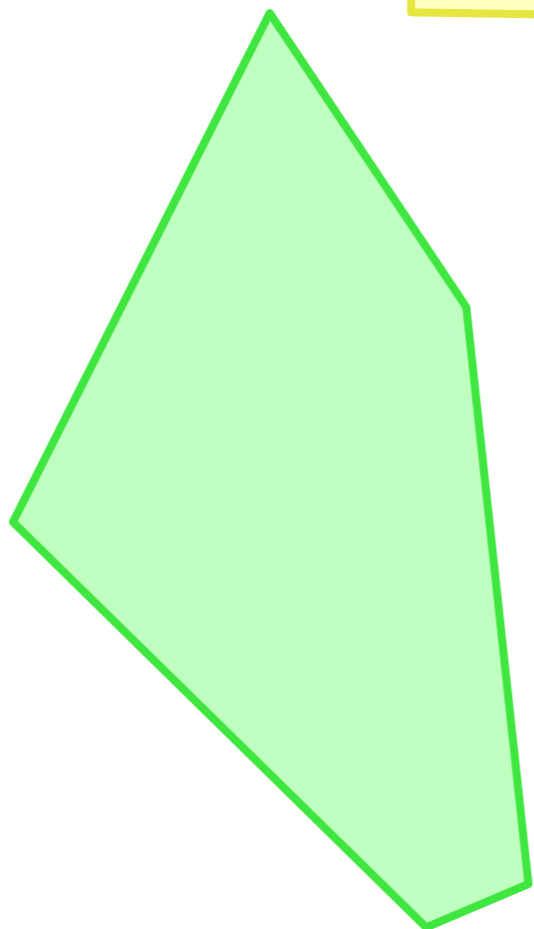


$S^{d-1}$

The *directional* width of a region is defined by a pair of vertices.

The sphere of directions is subdivided depending on which points define it.

In  $d$  dimensions, this subdivision may have complexity  $O(d^2)$ .



$\mathbb{R}^d$

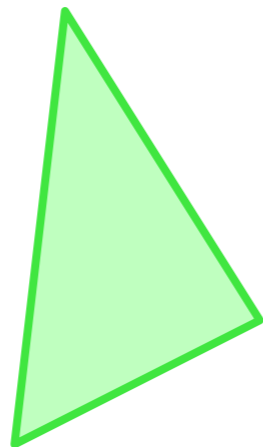
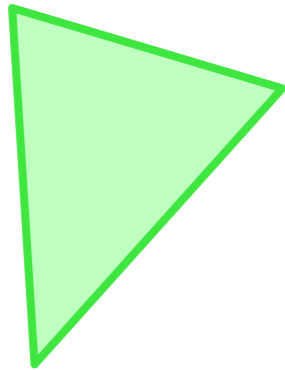
$S^{d-1}$

Now, how do we compute the *smallest* width?

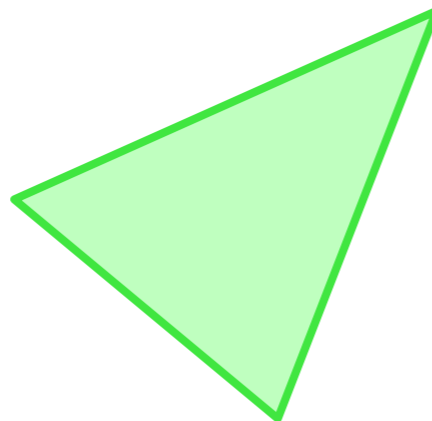
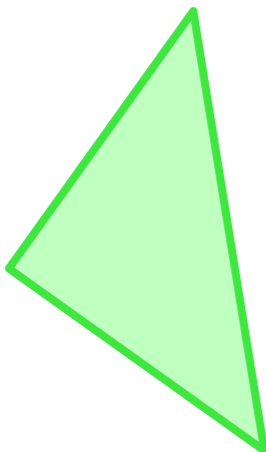


Consider the  
set of regions  
 $\mathcal{R}$ .

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 $\mathcal{R}$ .



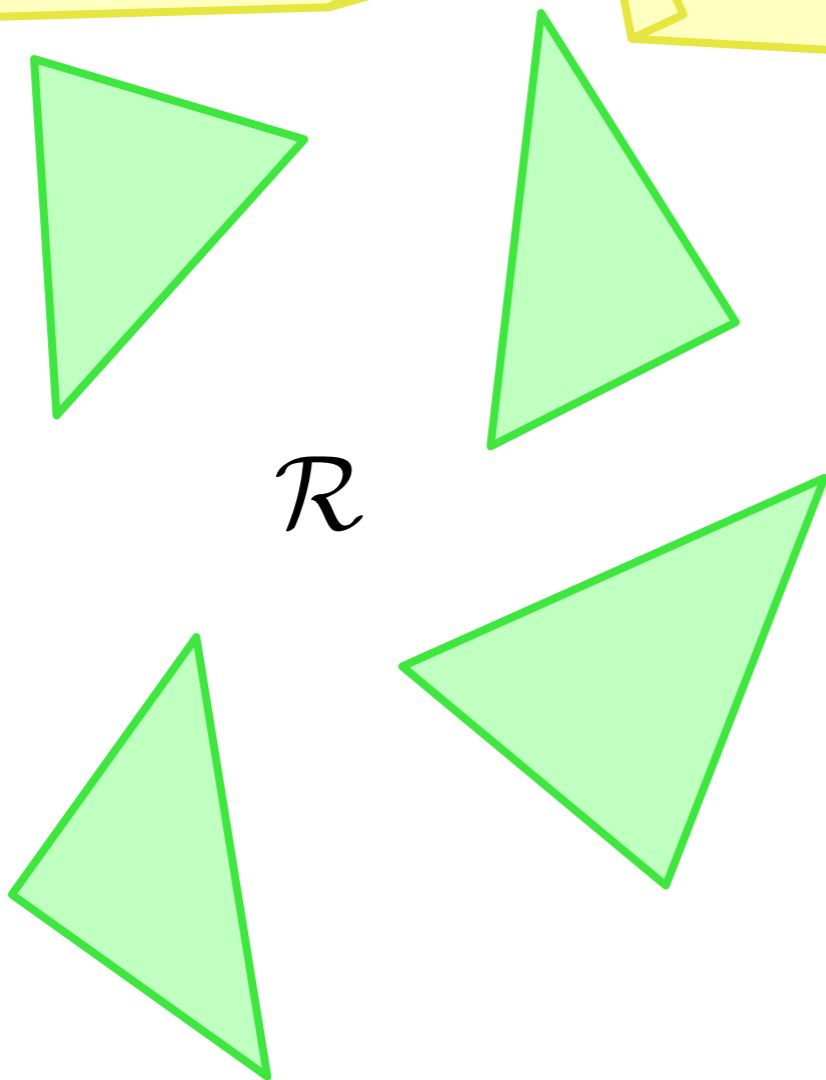
$\mathcal{R}$



$\mathbb{R}^d$

Consider the set of regions  $\mathcal{R}$ .

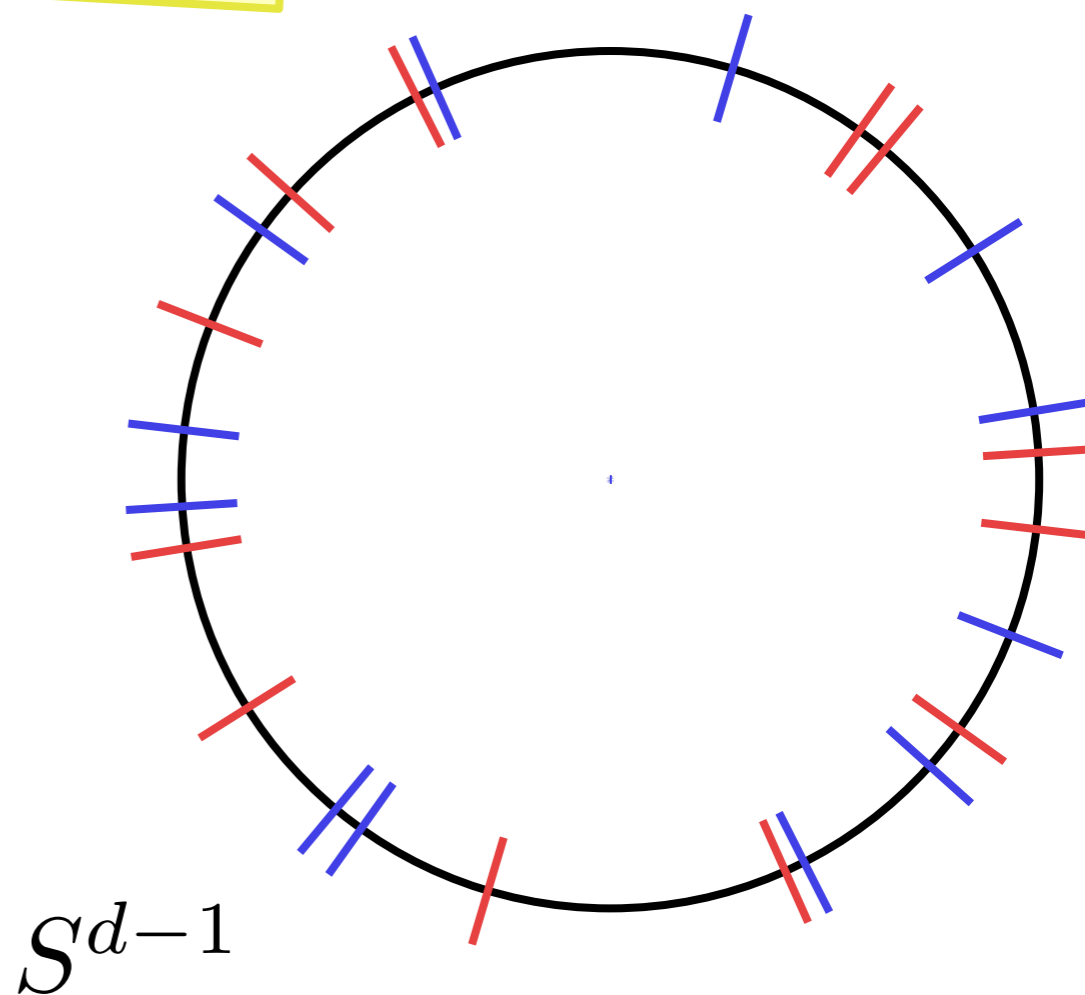
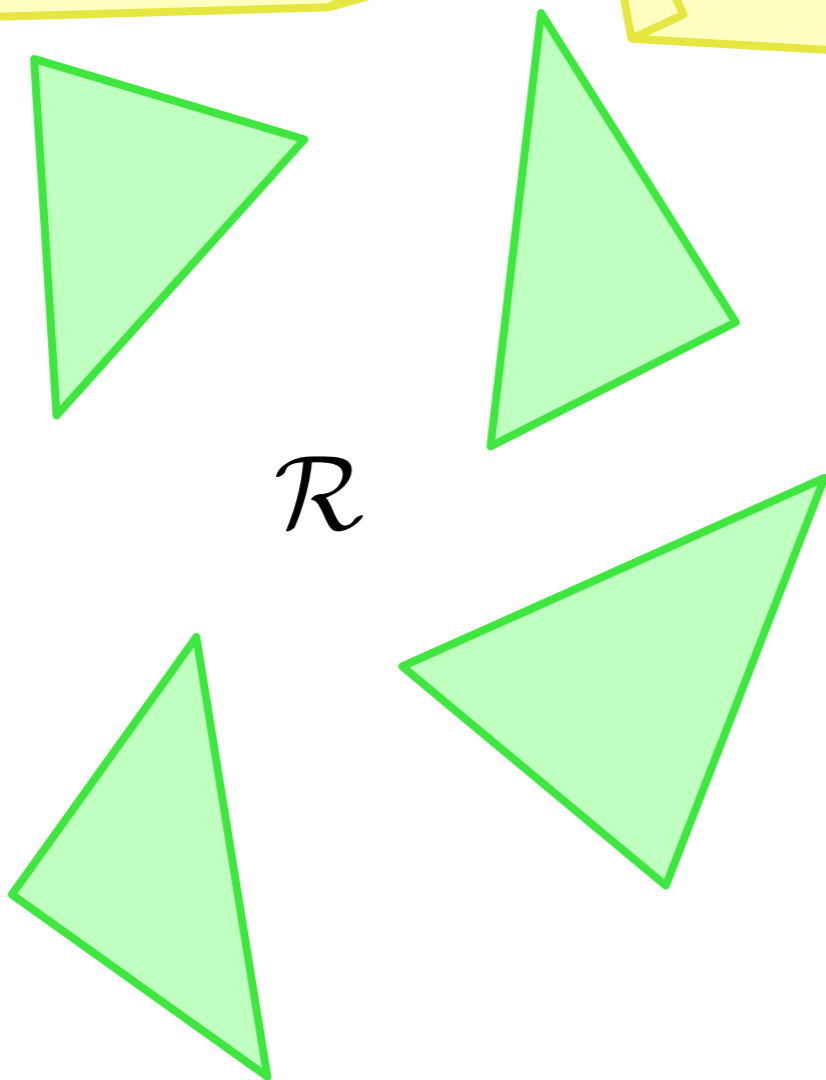
We can compute the overlay of the subdivisions of  $S^{d-1}$  of all regions.



$\mathbb{R}^d$

Consider the set of regions  $\mathcal{R}$ .

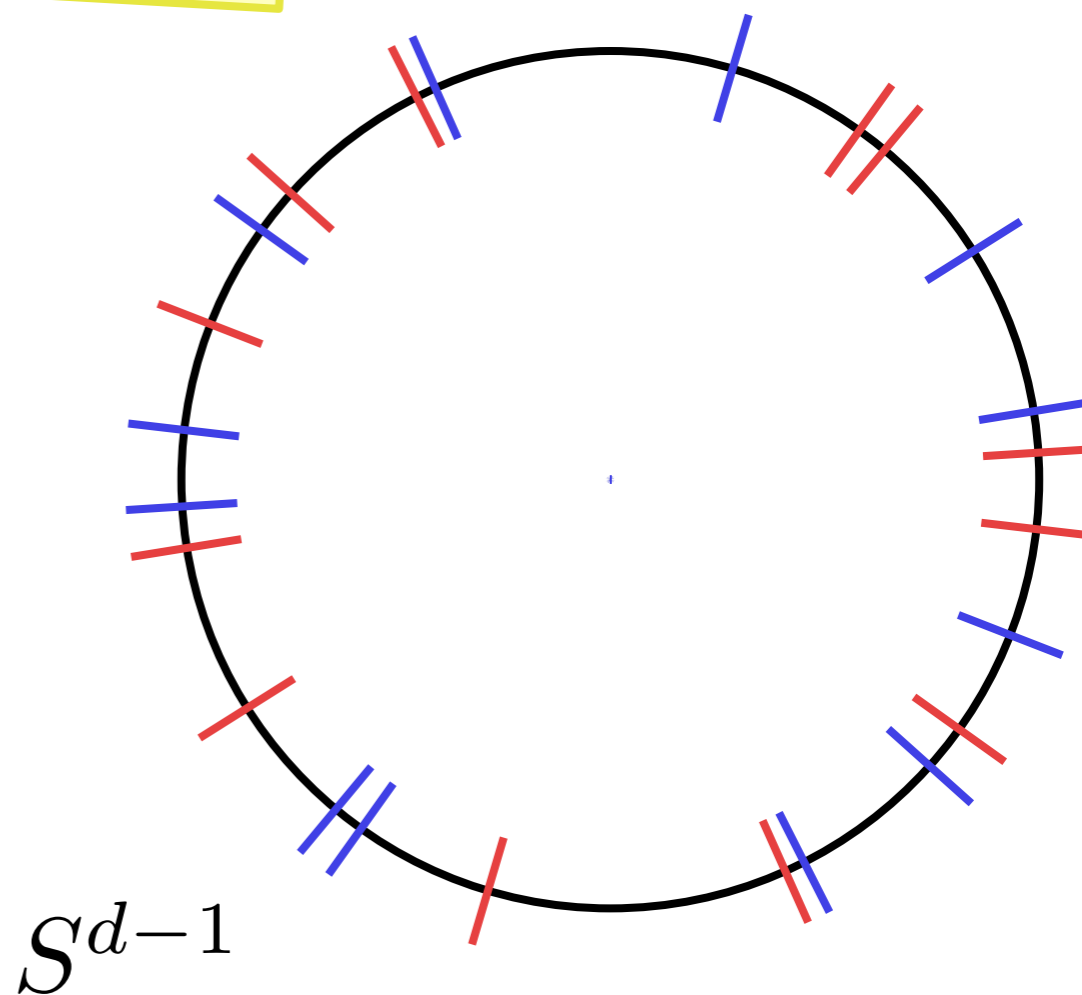
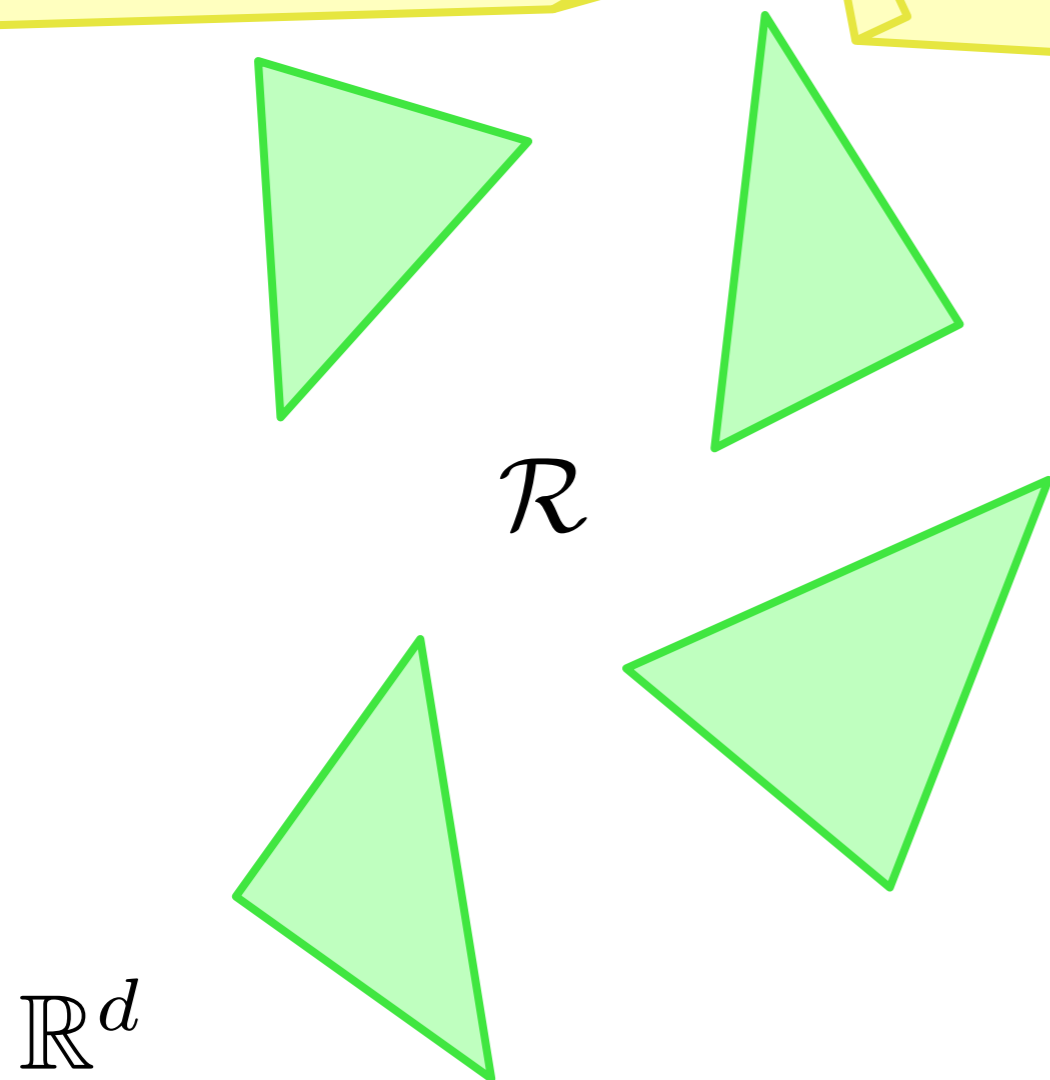
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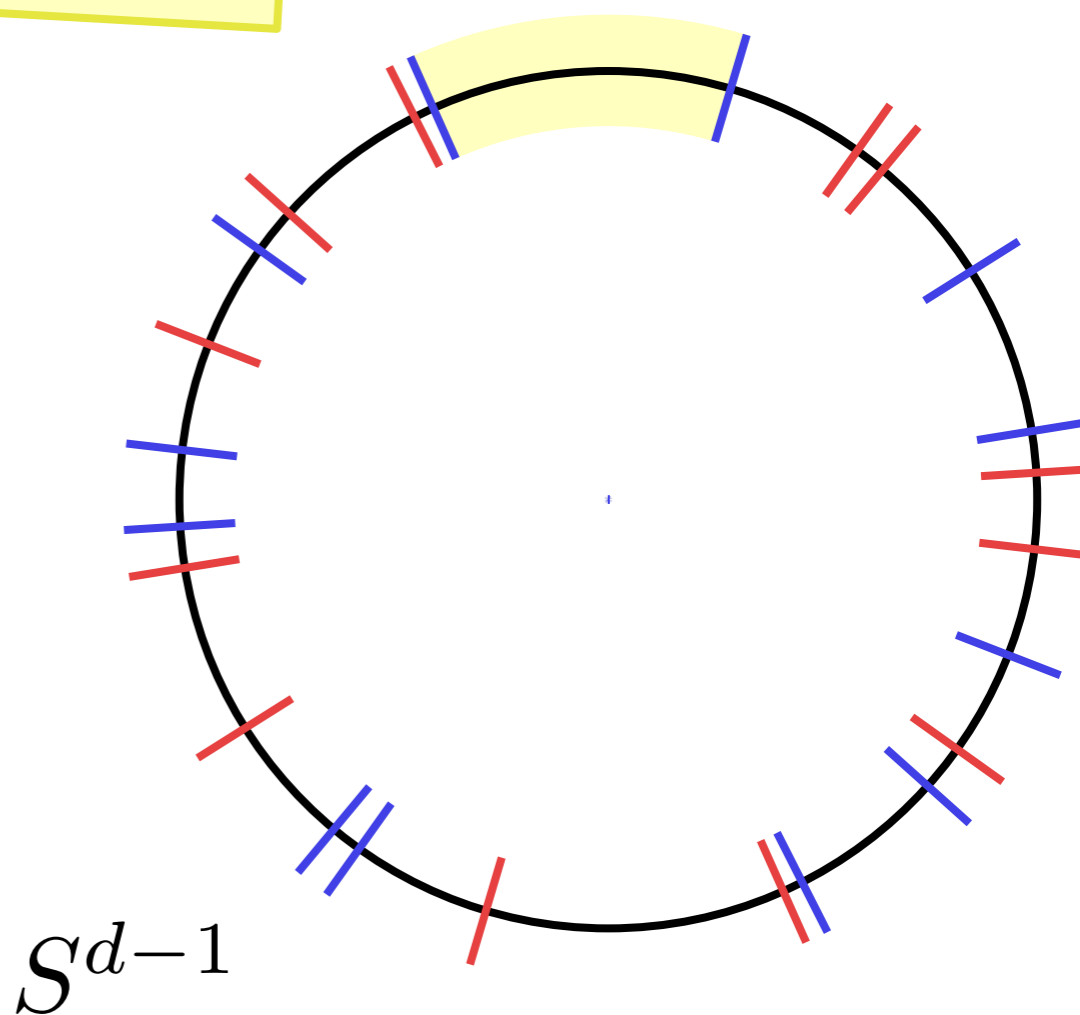
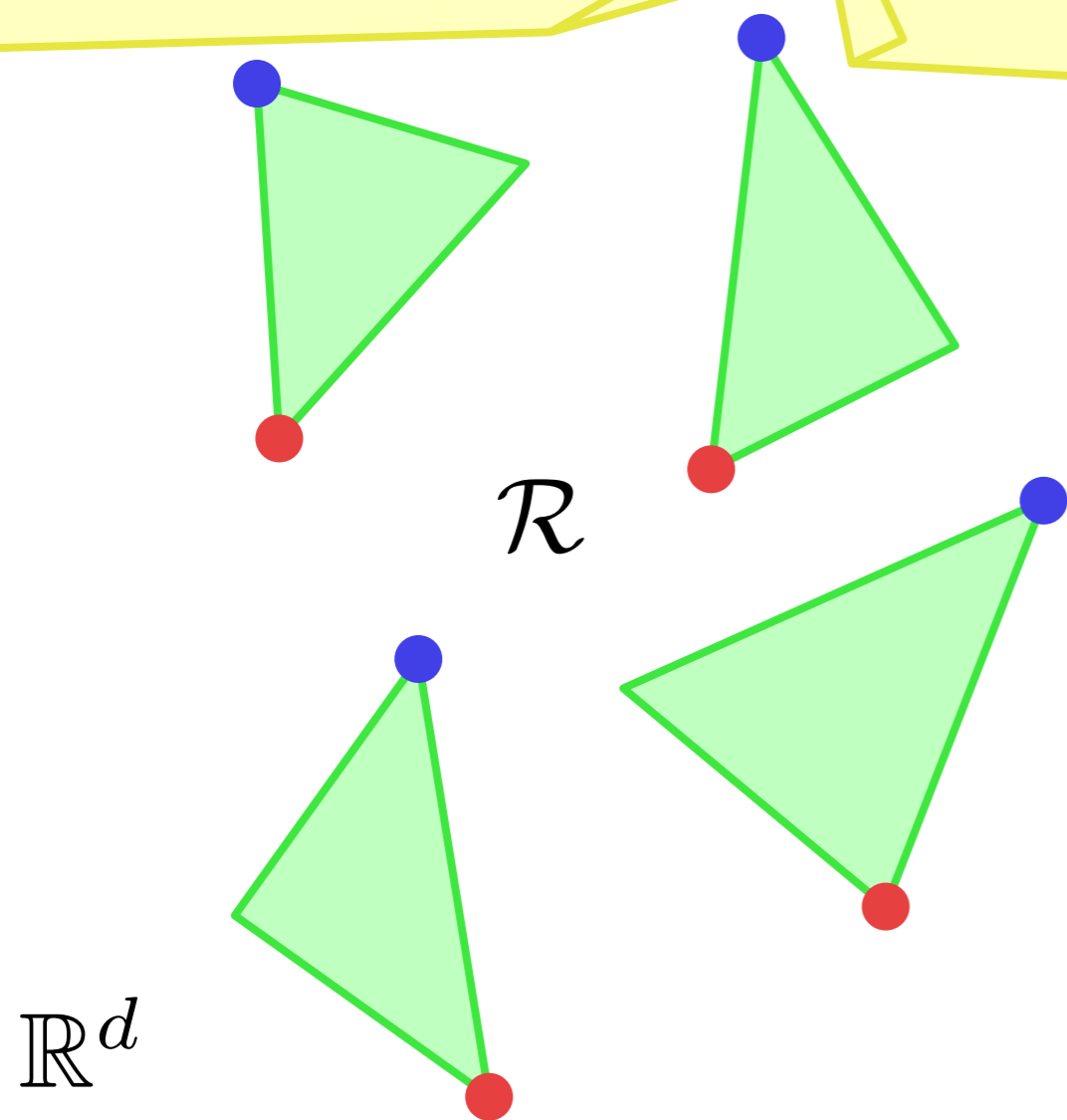
Within each cell, we can now compute the smallest width.



Consider the set of regions  $\mathcal{R}$ .

We can compute the overlay of the subdivisions of  $S^{d-1}$  of all regions.

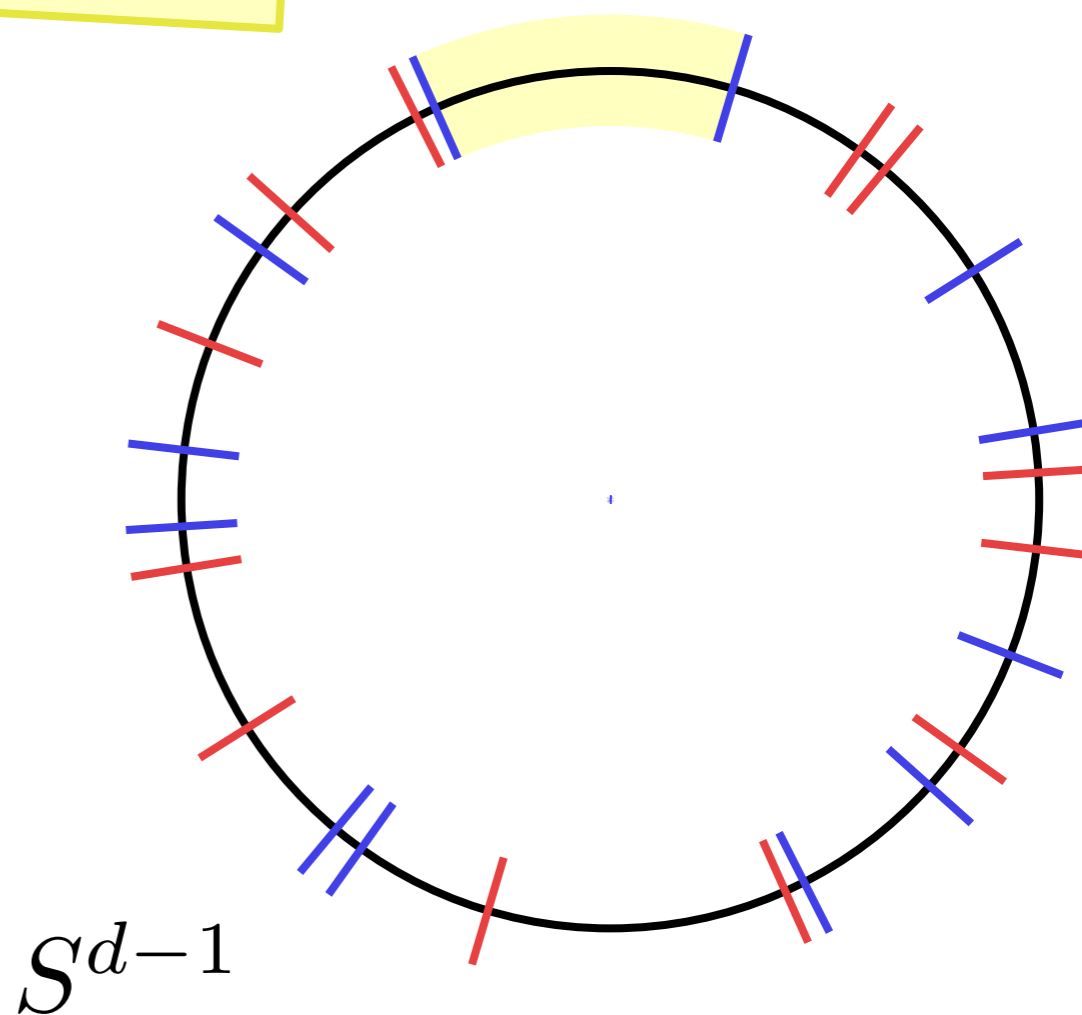
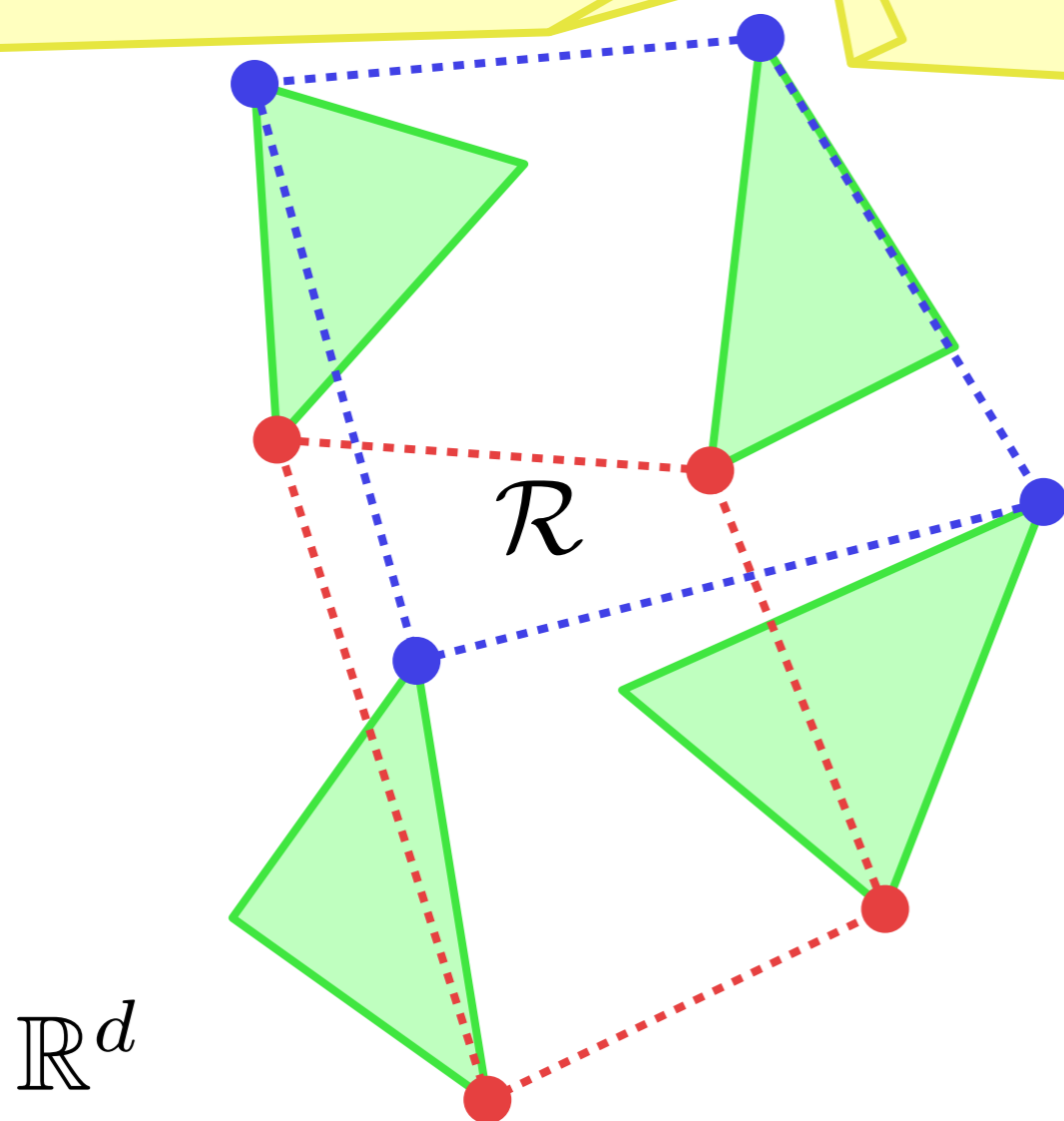
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We can compute the overlay of the subdivisions of  $S^{d-1}$  of all regions.

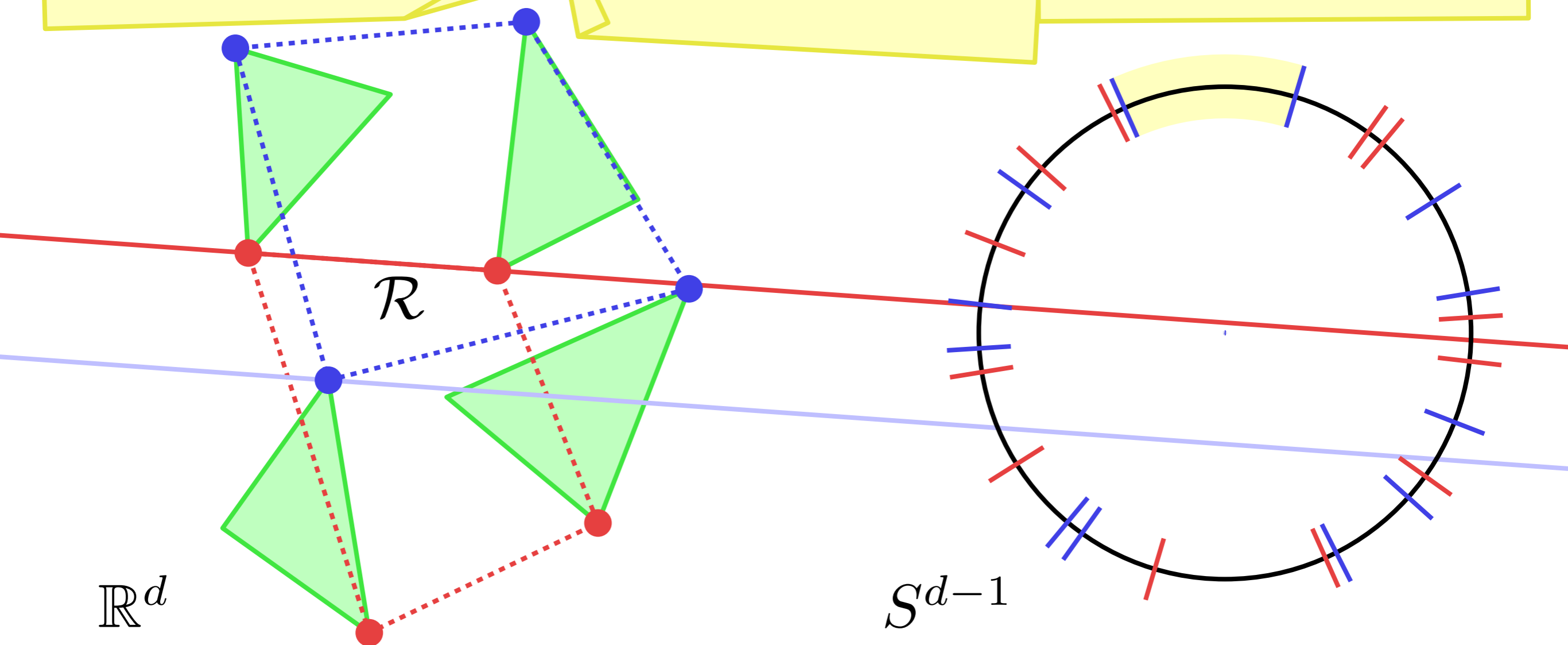
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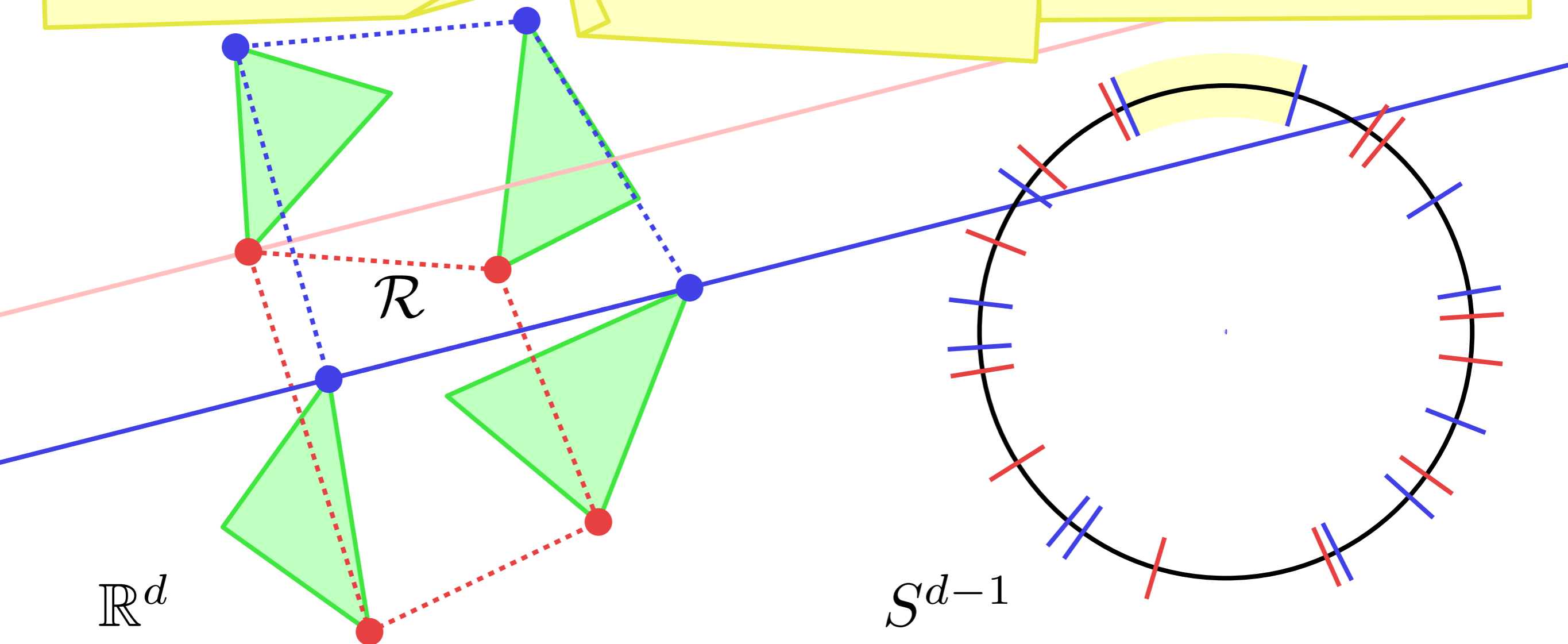




Consider the set of regions  $\mathcal{R}$ .

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Within each cell, we can now compute the smallest width.



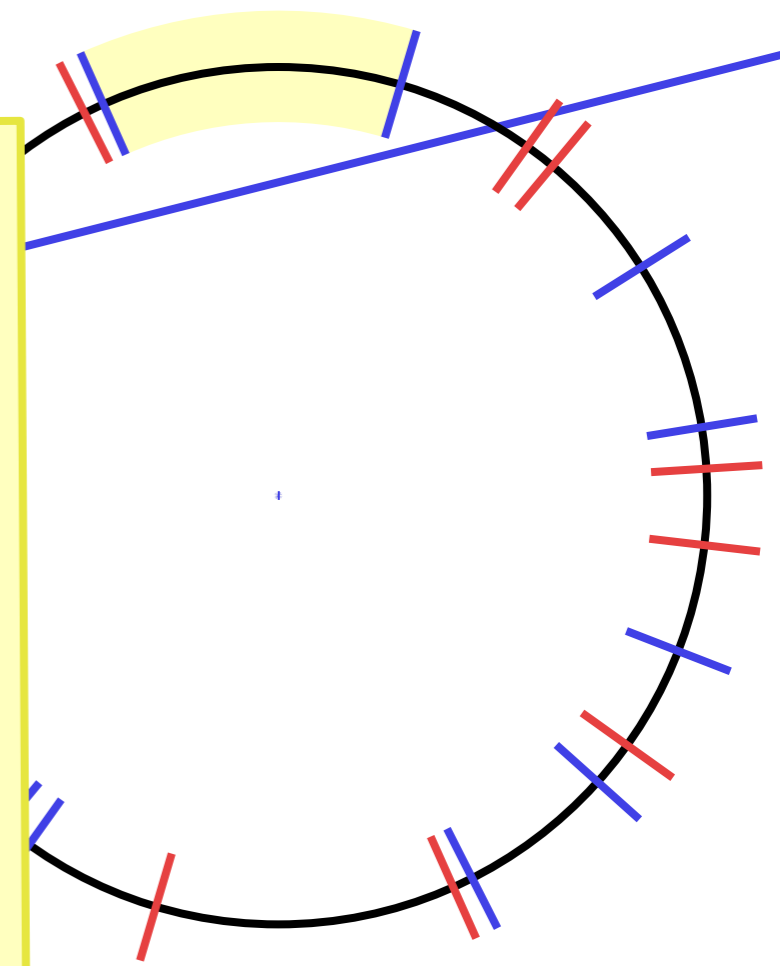
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We can compute the overlay of the subdivisions of  $S^{d-1}$  of all regions.

Within each cell, we can now compute the smallest width.

In total, we spend  $O(n^{2d-1})$  time.

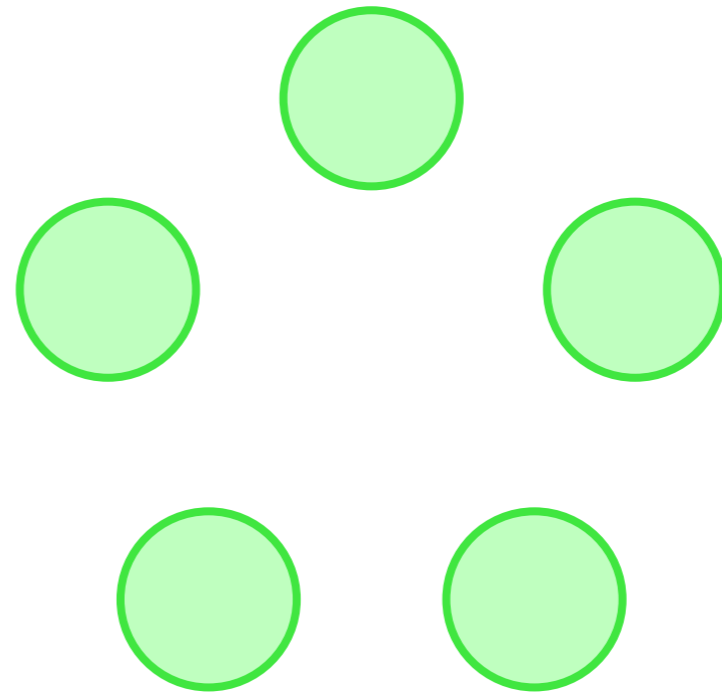
$\mathbb{R}^d$



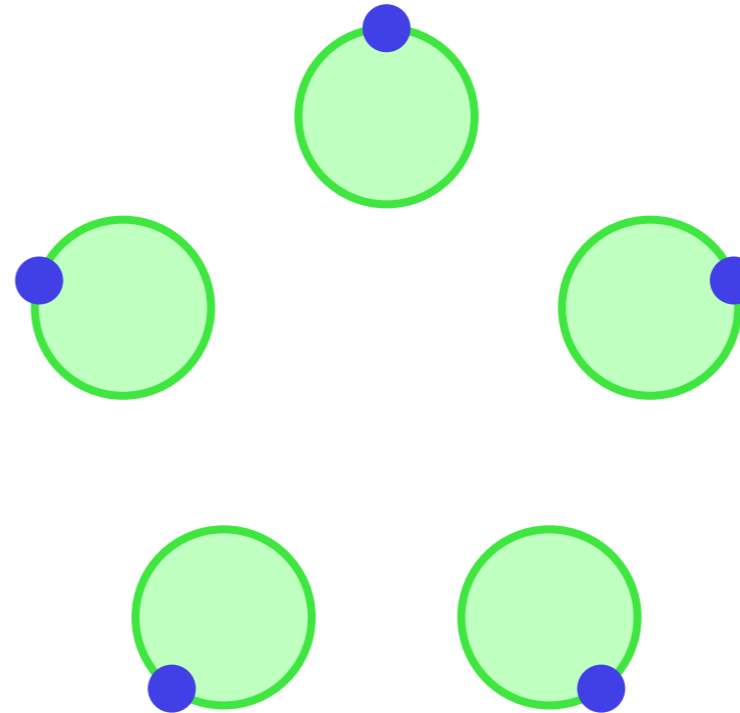
What about  
the *largest*  
width?

The largest width could be determined by many triples of points, keeping each other in balance.

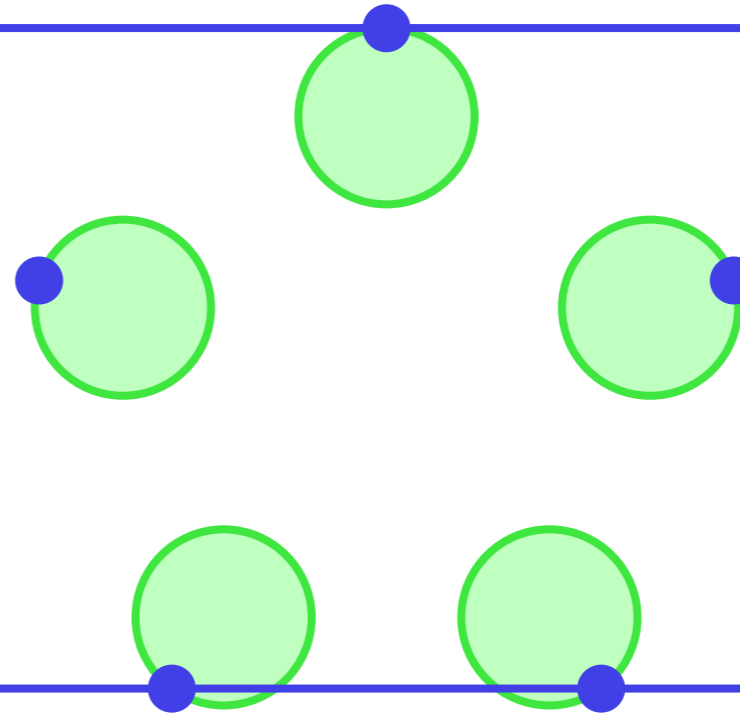
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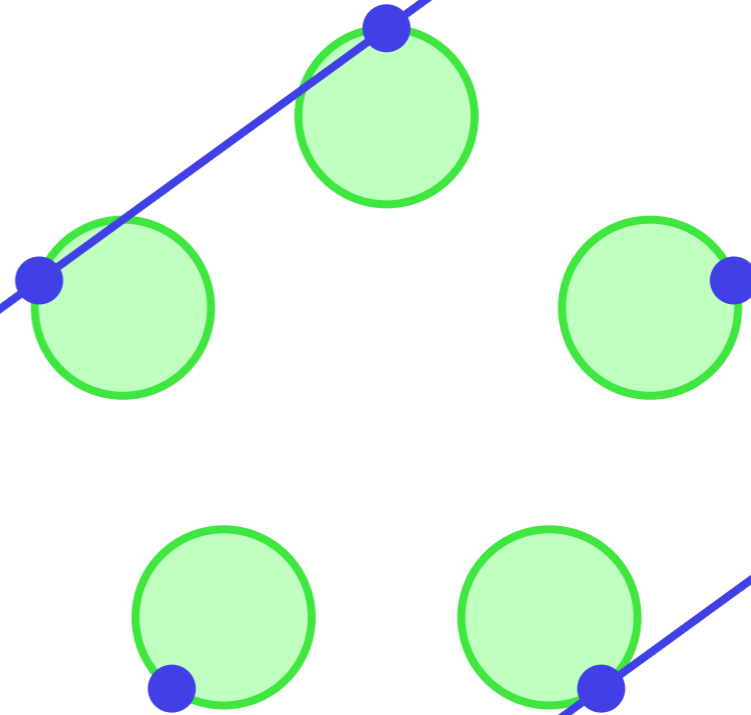
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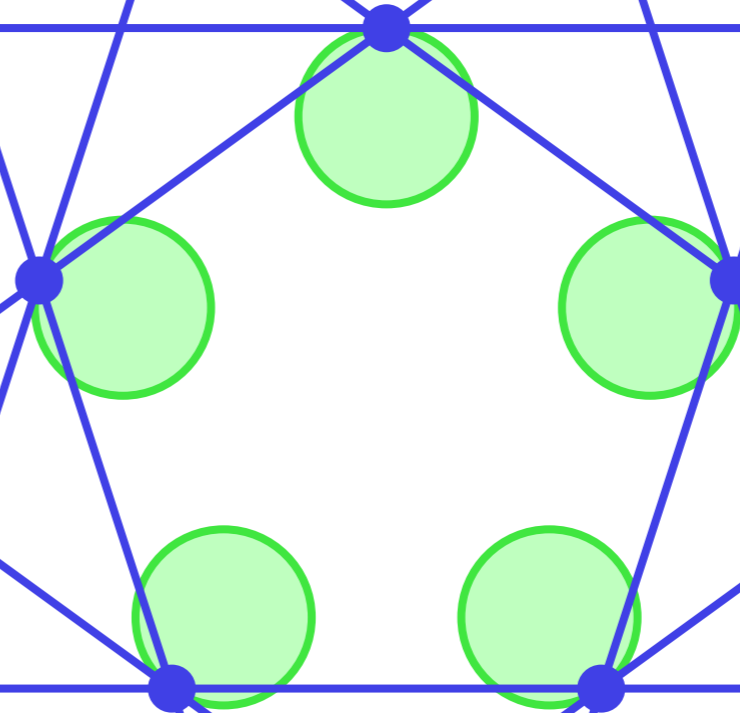


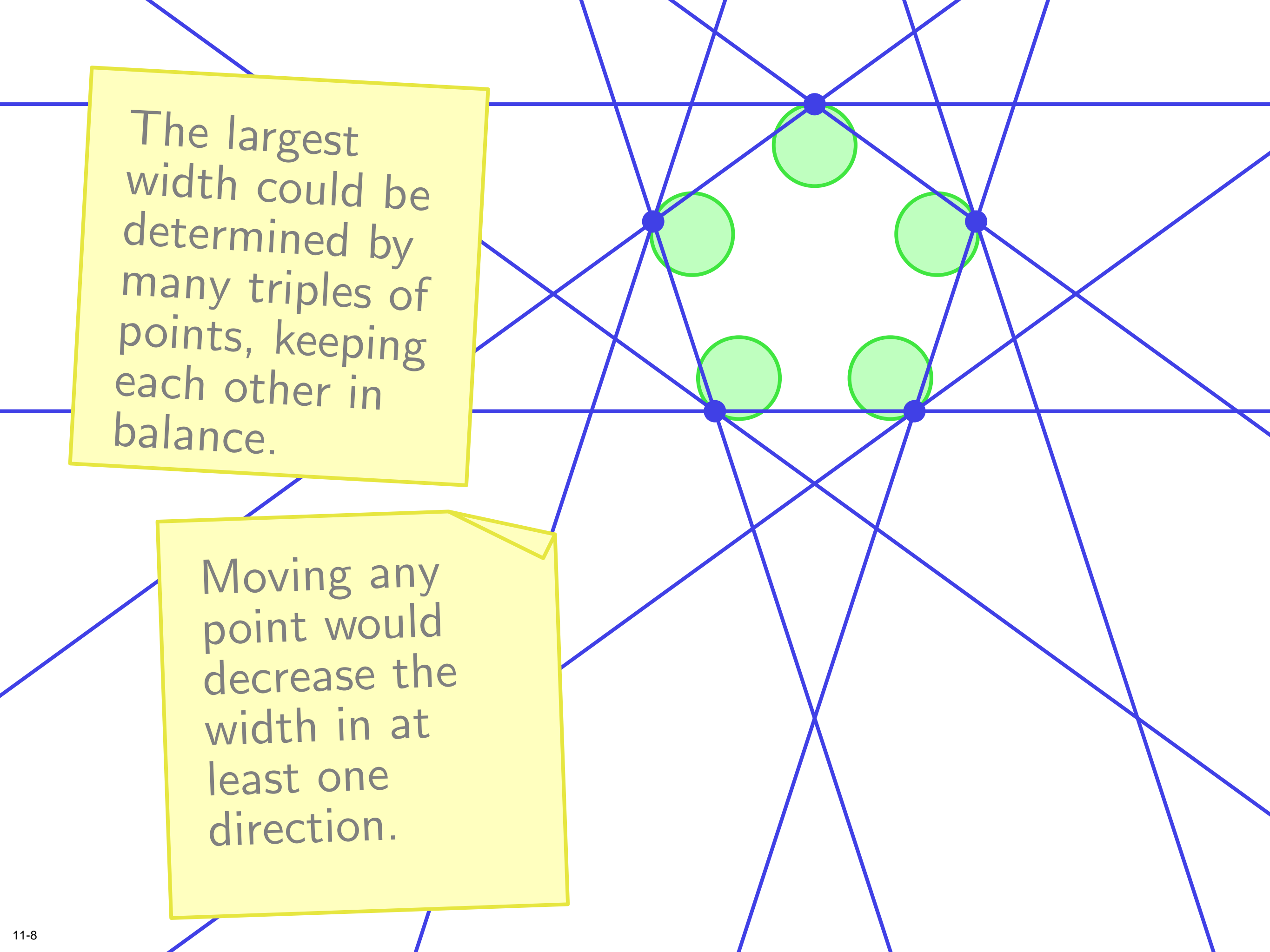
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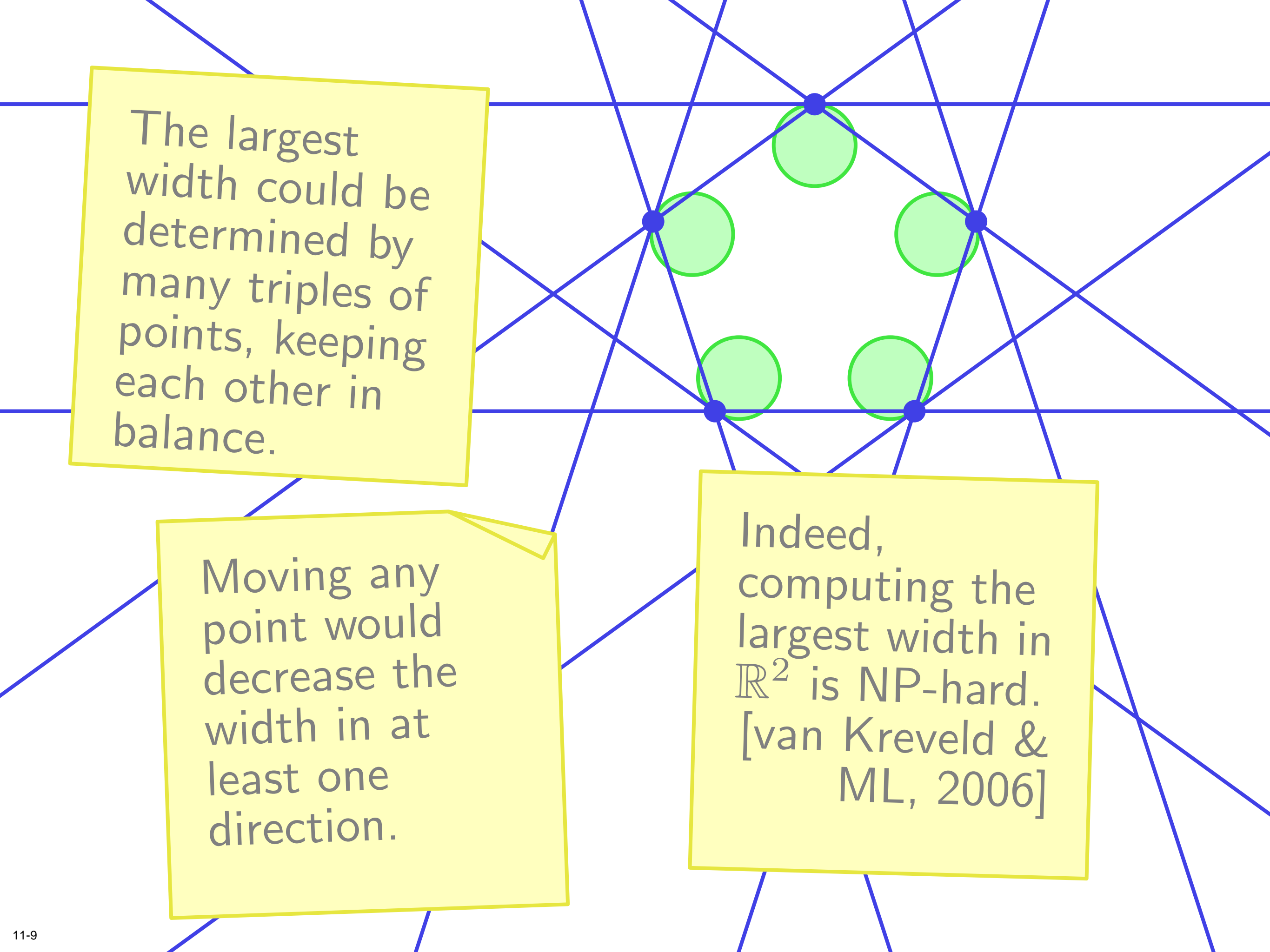
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The largest width could be determined by many triples of points, keeping each other in balance.

Moving any point would decrease the width in at least one direction.



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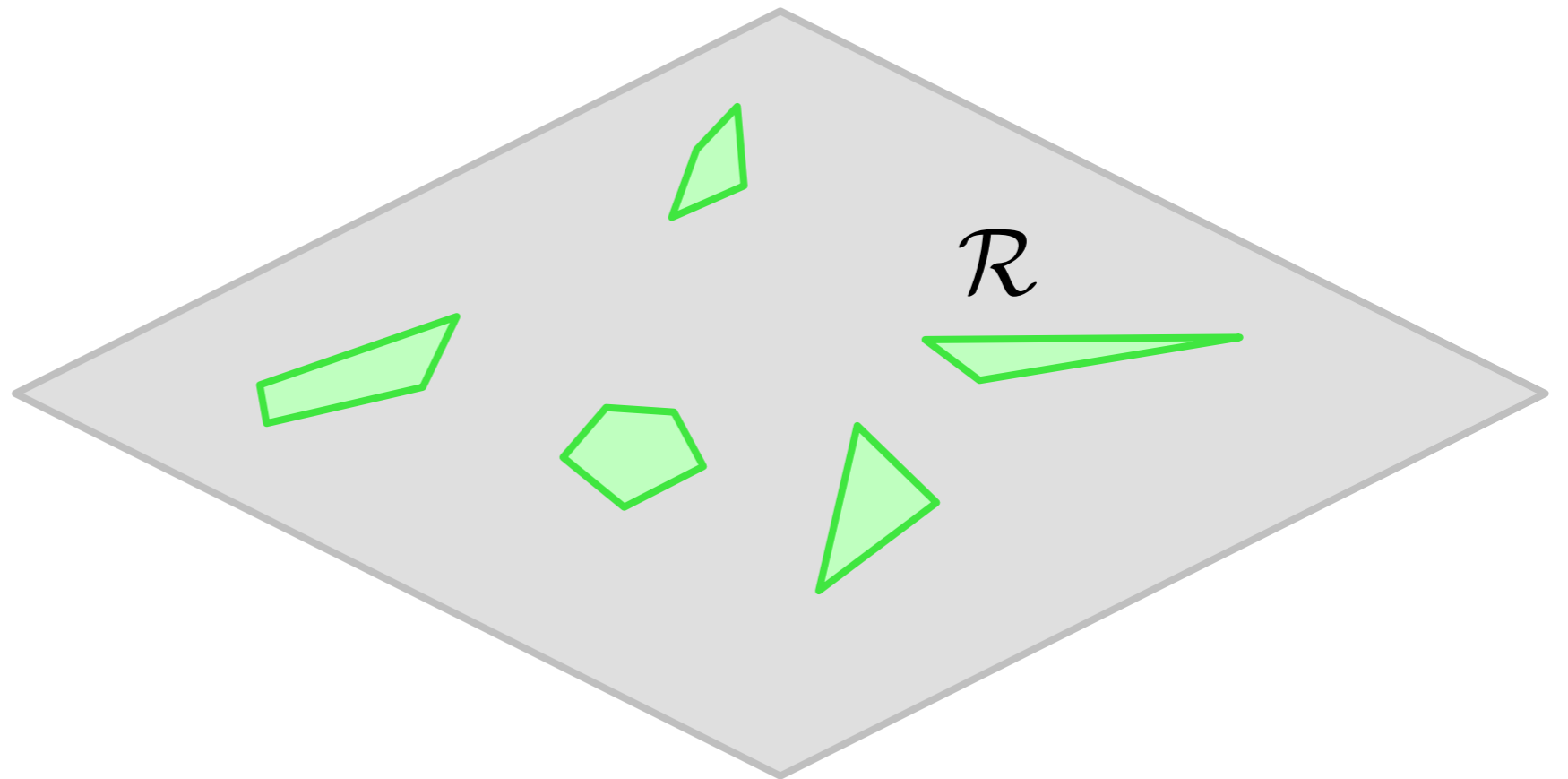
Moving any point would decrease the width in at least one direction.

Indeed, computing the largest width in  $\mathbb{R}^2$  is NP-hard. [van Kreveld & ML, 2006]

We can  
generalise that  
proof to  $\mathbb{R}^d$ .

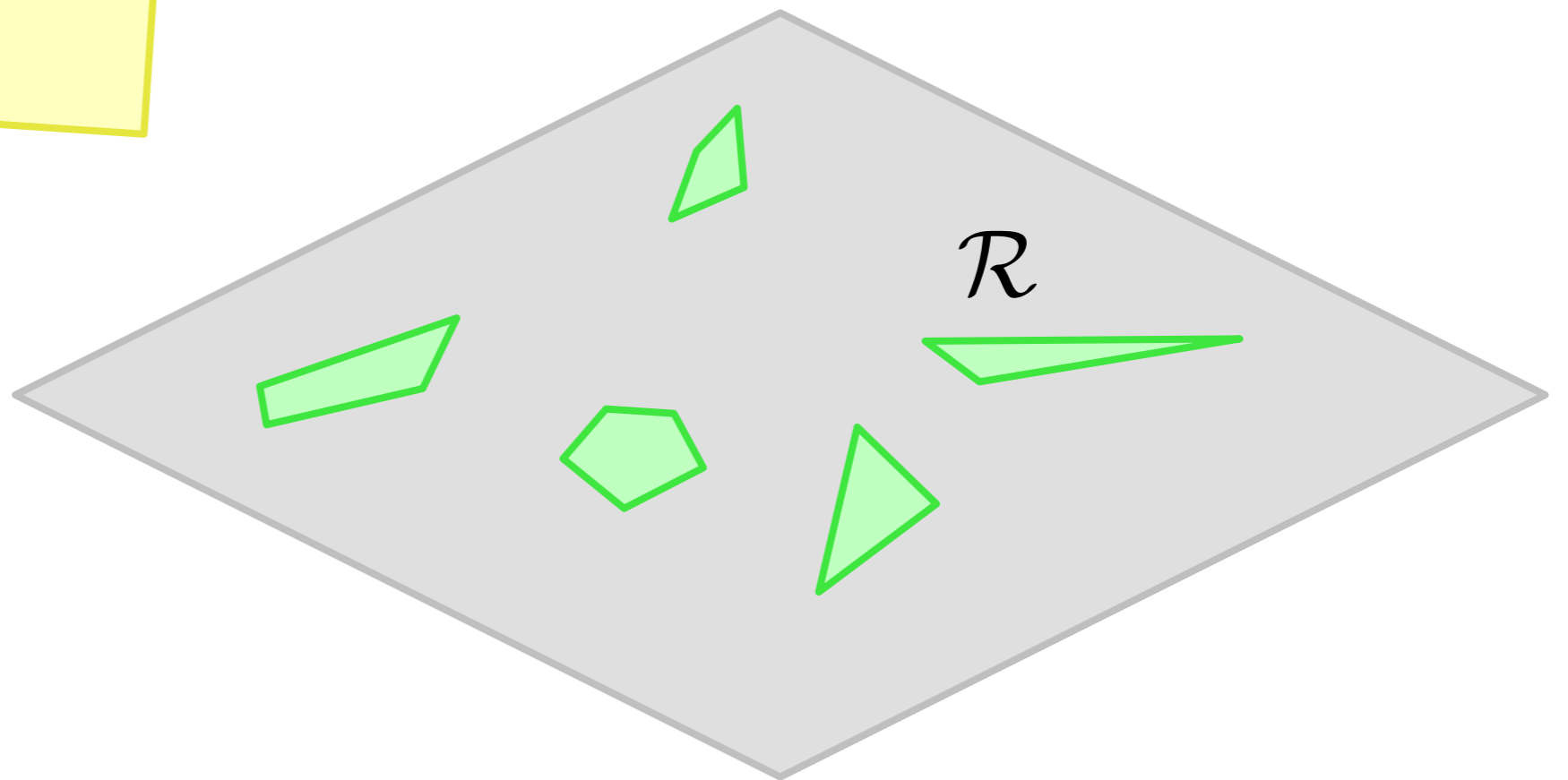
Consider a set  
of regions  $\mathcal{R}$  in  
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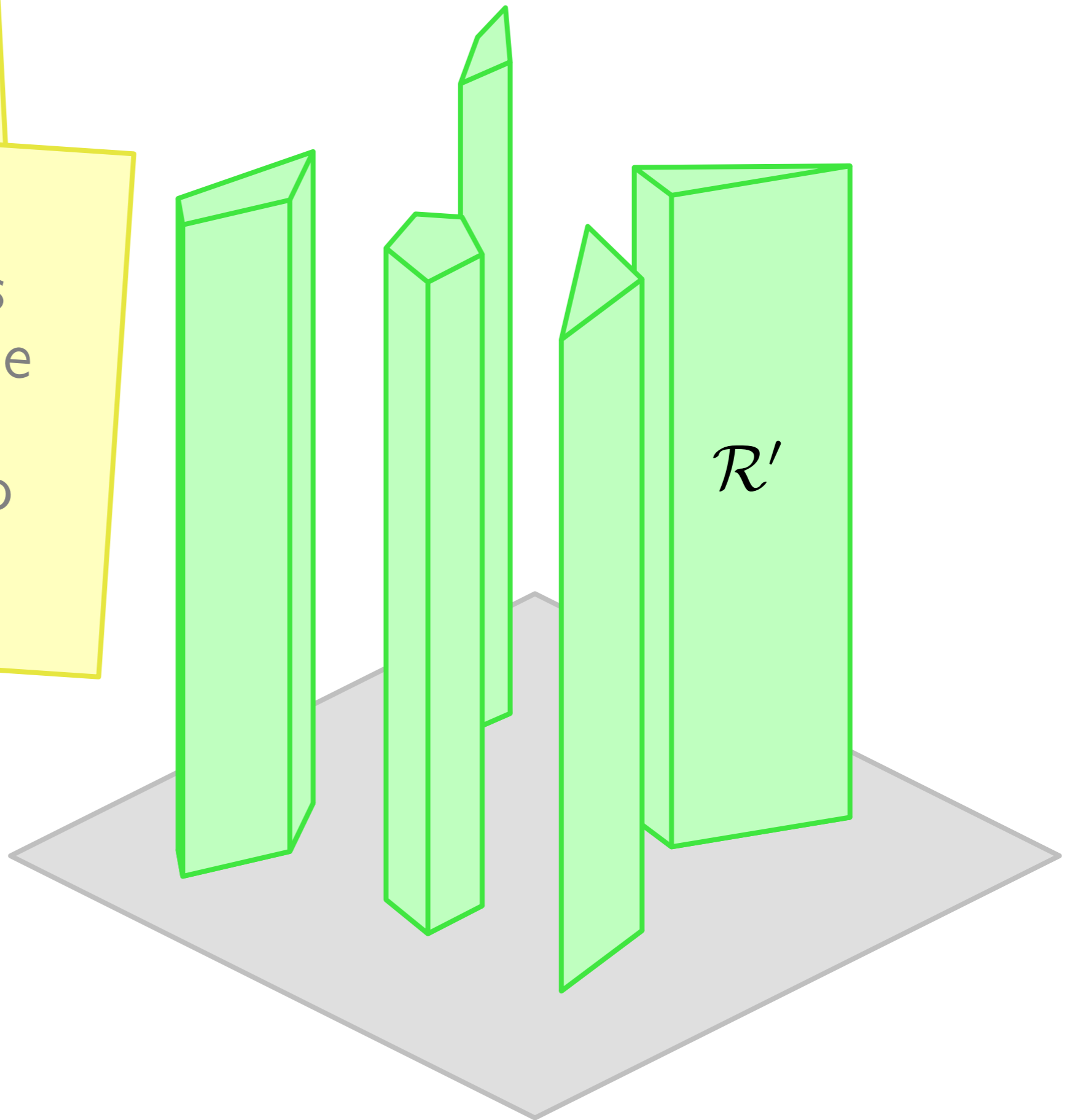
Consider a set  
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the plane

We extend  
these regions  
very far in the  
 $d - 1$  other  
dimensions to  
a set  $\mathcal{R}'$ .



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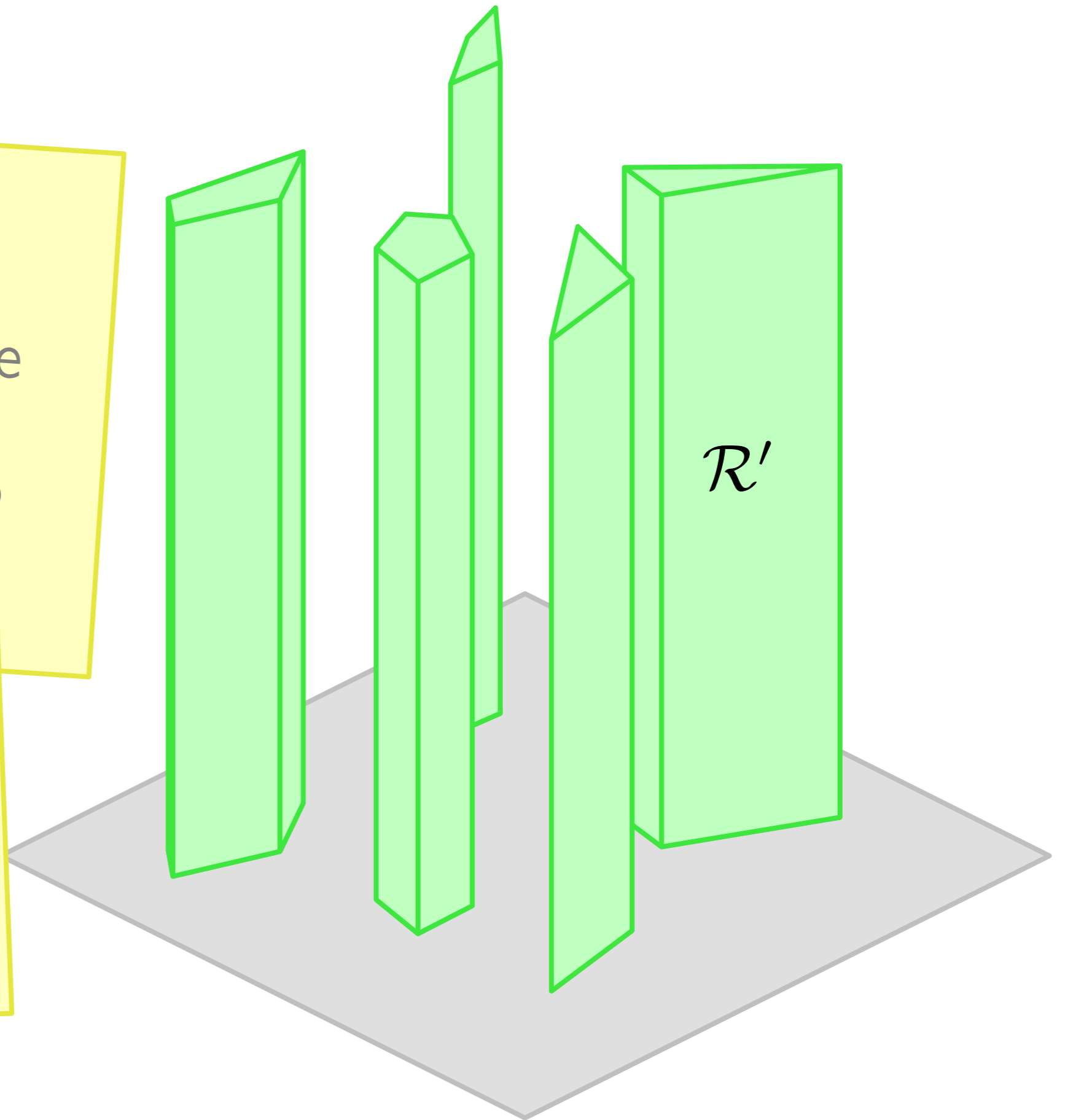




Consider a set of regions  $\mathcal{R}$  in the plane

We extend these regions very far in the  $d - 1$  other dimensions to

Clearly, the width of  $\mathcal{R}'$  cannot be larger than of  $\mathcal{R}$ .

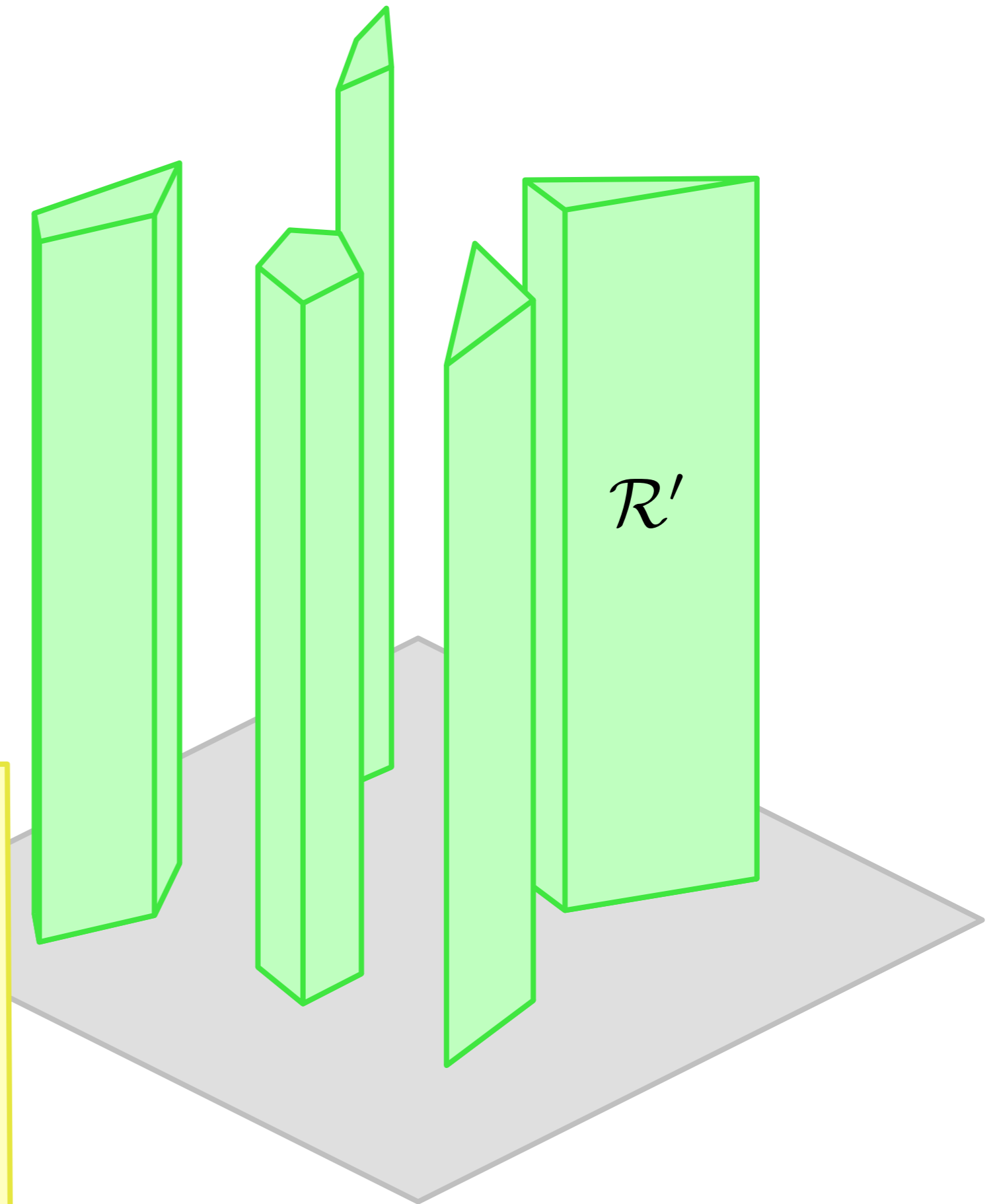


Consider a set of regions  $\mathcal{R}$  in the plane

We extend these regions very far in the  $d - 1$  other dimensions to

Clearly, the width of  $\mathcal{R}'$  cannot be

large. For any  $P$  in  $\mathcal{R}$  we can place points in  $\mathcal{R}'$  such that the width comes arbitrary close.

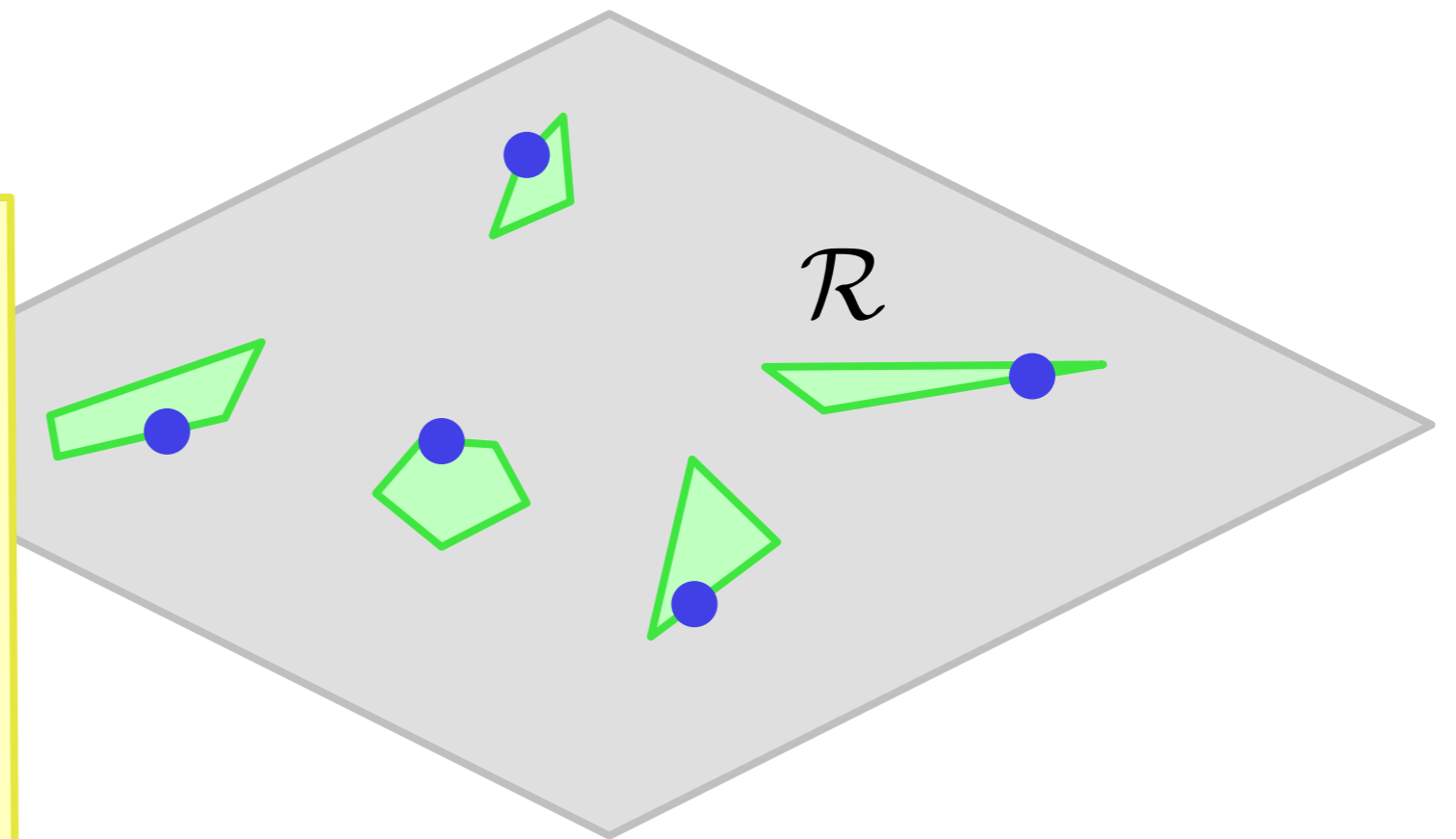


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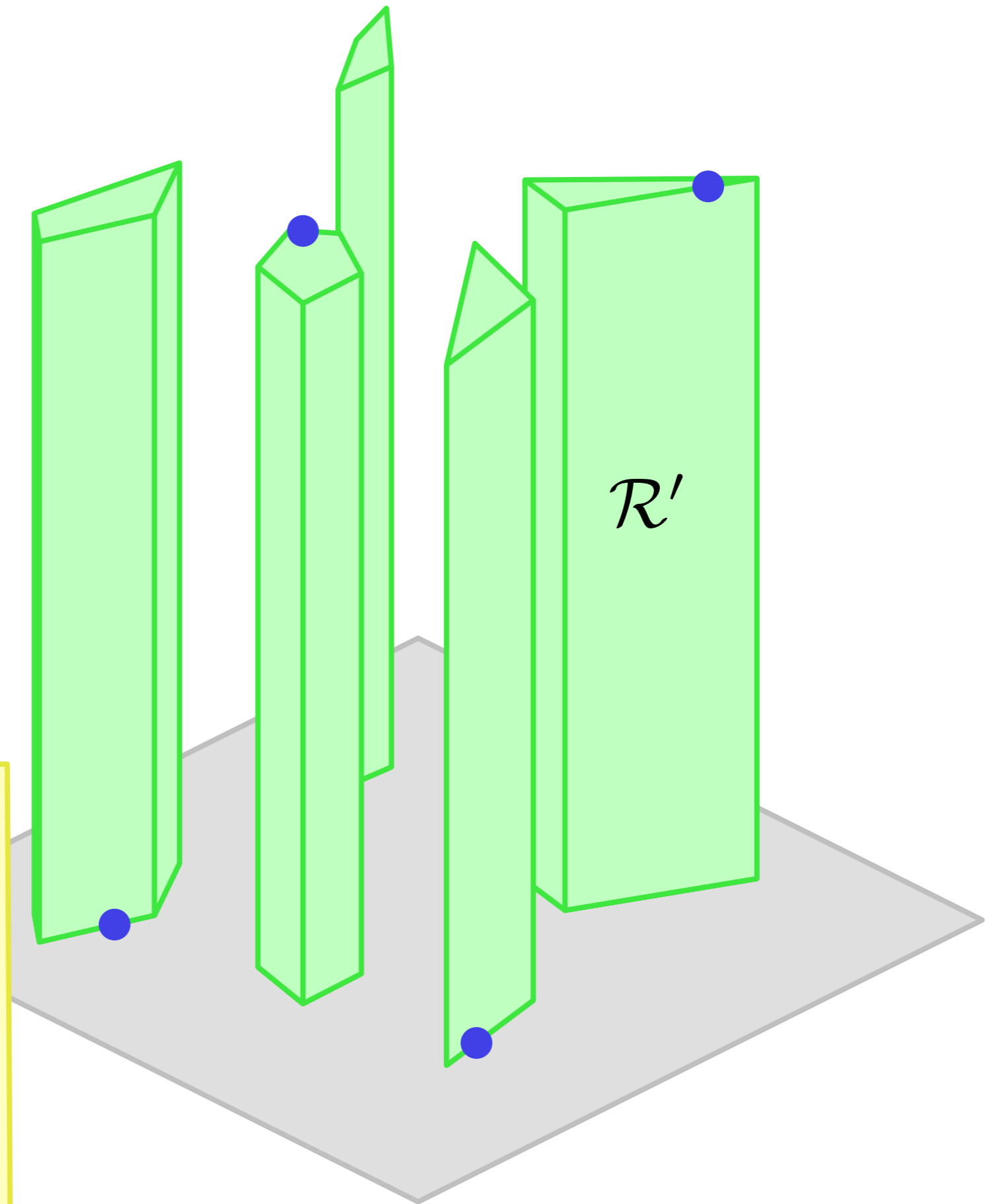


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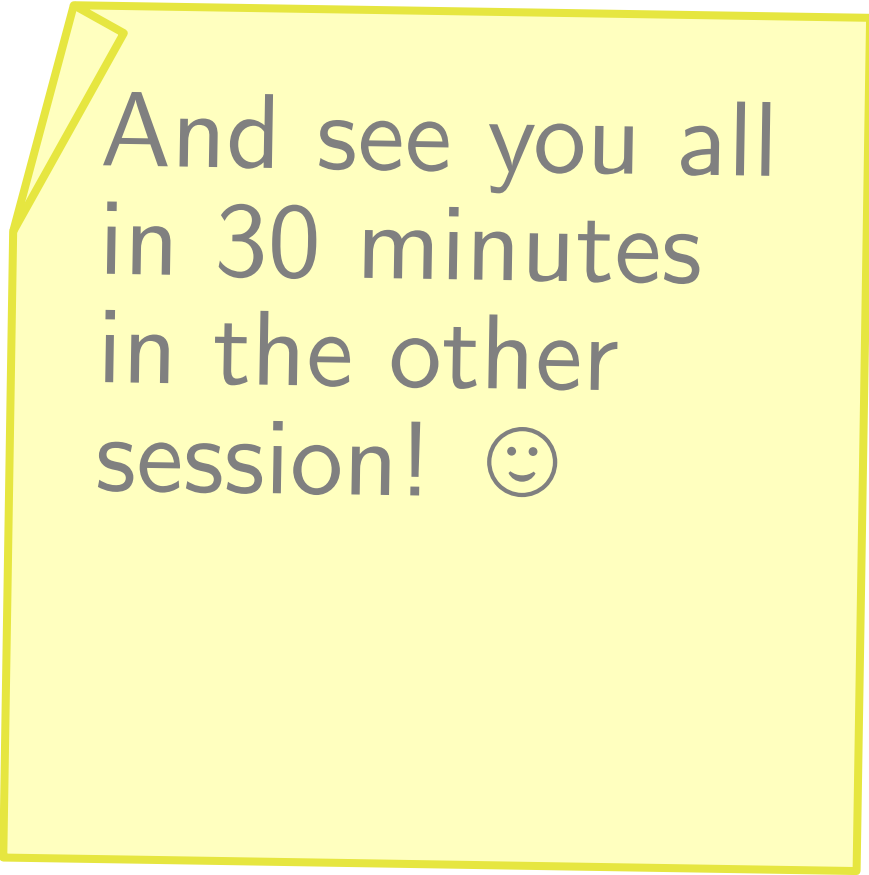
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Thank you for your attention!

# Thank you for your attention!

A yellow sticky note with a folded top-left corner, containing text and a smiley face.

And see you all  
in 30 minutes  
in the other  
session! 😊