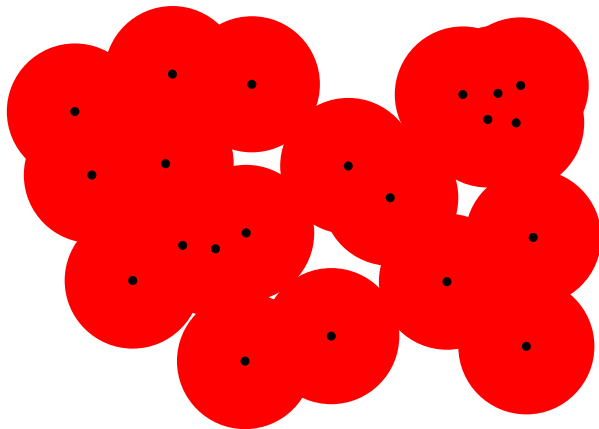


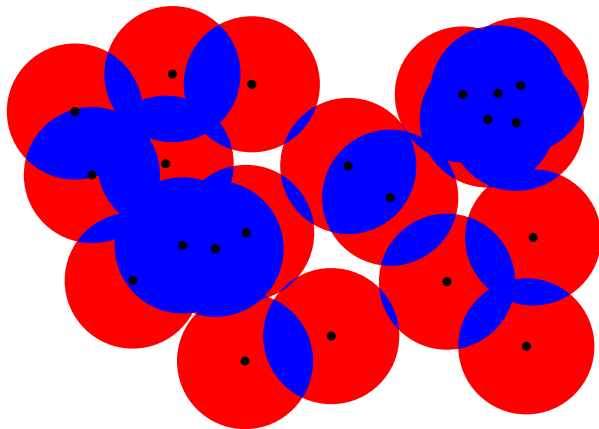
# Approximating $k$ -fold filtrations using weighted Delaunay triangulations

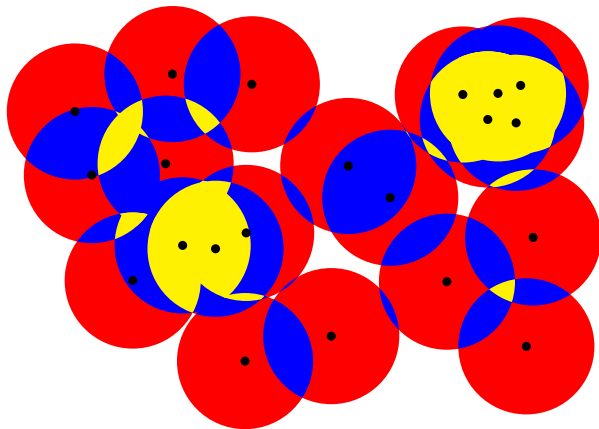
Mickaël Buchet and Michael Kerber

TU Graz

April 4, 2019







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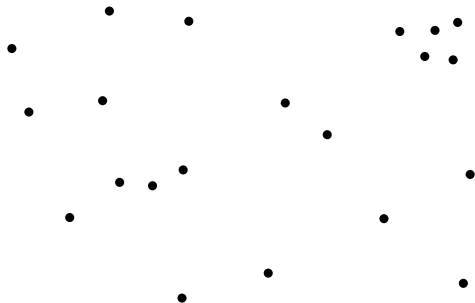
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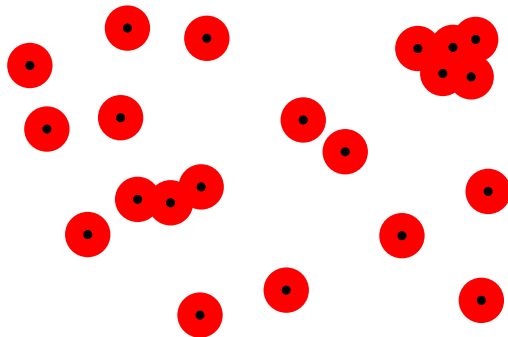
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Can we approximate the result using union of balls?

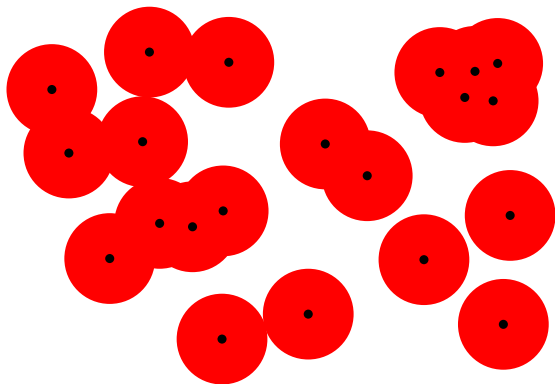
# Starting easy... $k = 1$



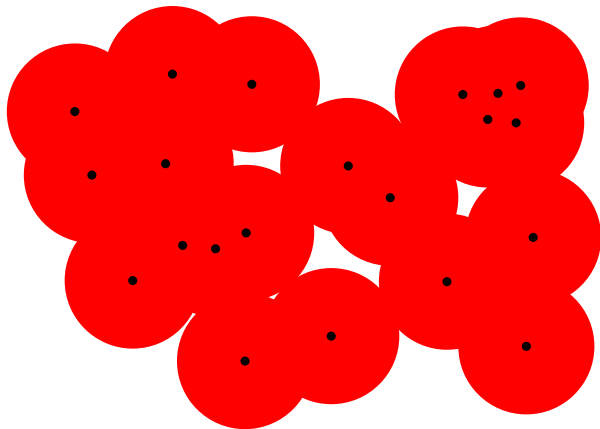
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# Classical data structures

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- Alpha complex.

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Filtration on the Delaunay triangulation



## A functional point of view

$$d(x) = \min_{p \in P} \|x - p\|$$

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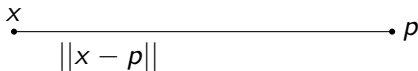
$$d(x) = \sqrt{\min_{p \in P} \|x - p\|^2 + 0}$$

## A functional point of view

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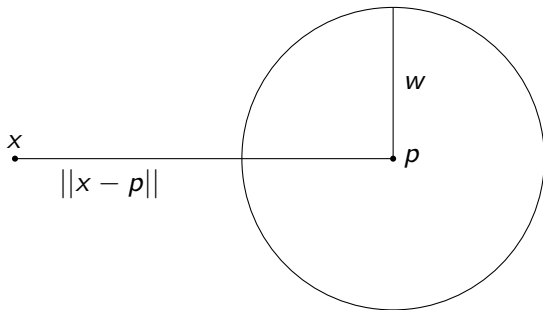
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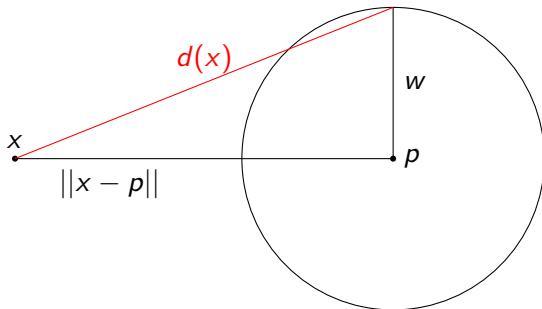
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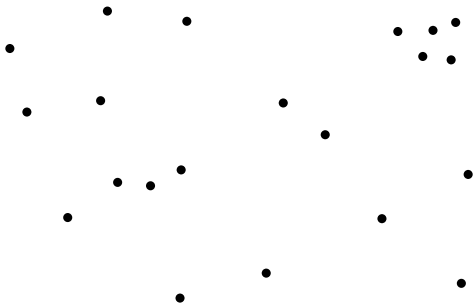


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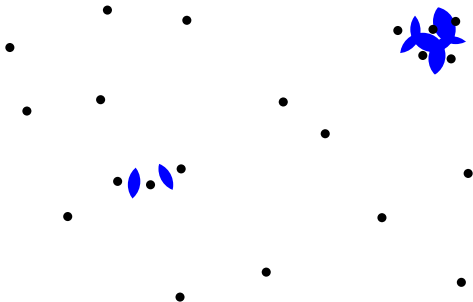
$$d(x) = \sqrt{\min_{p \in P} \|x - p\|^2 + w^2}$$



Now for  $k = 2$

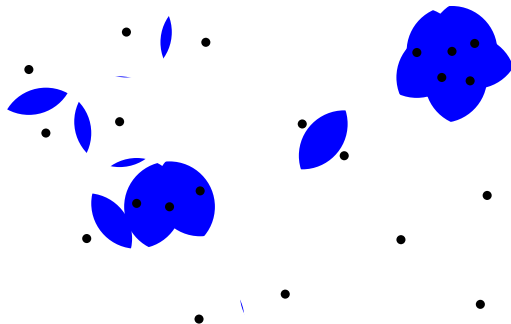


Now for  $k = 2$

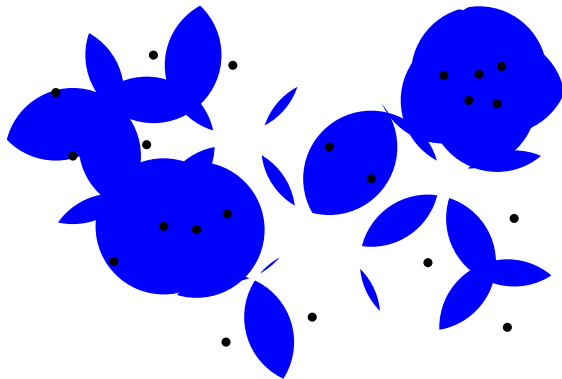




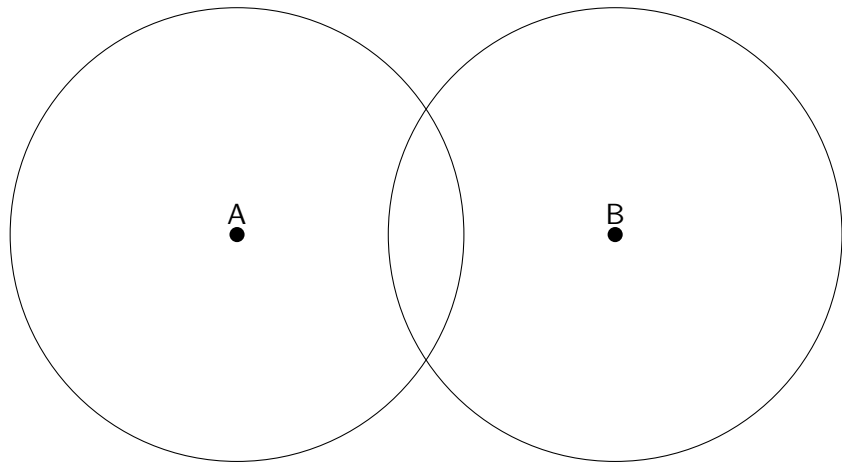
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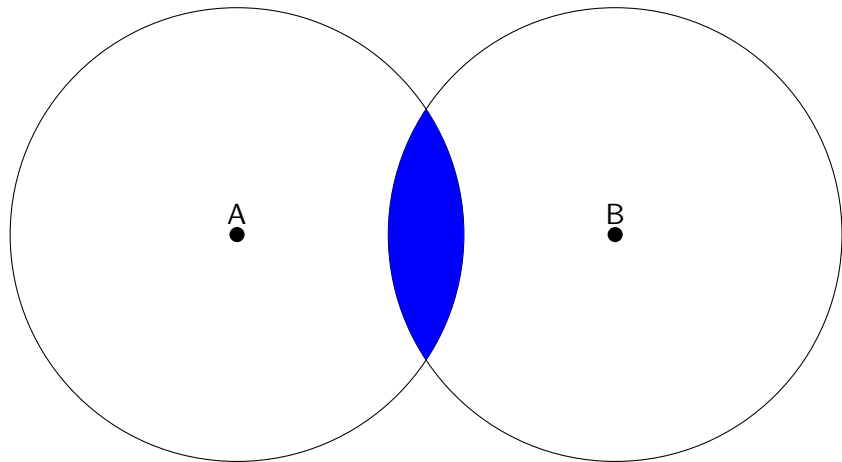
Now for  $k = 2$



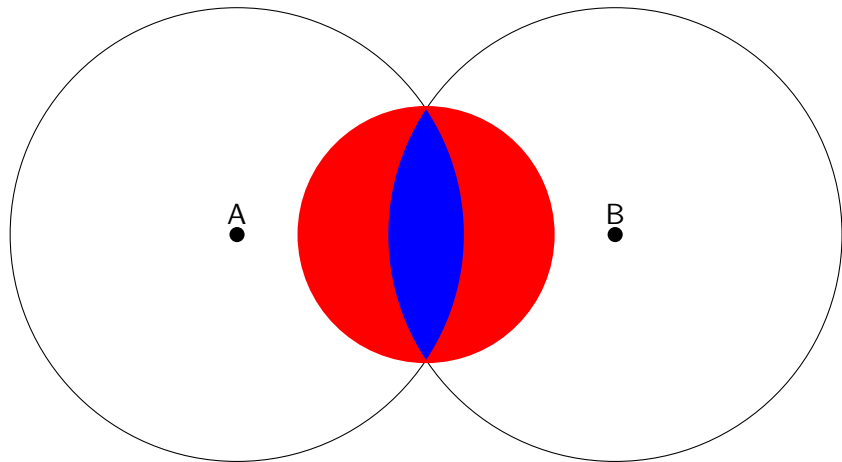
## Zooming on the lense



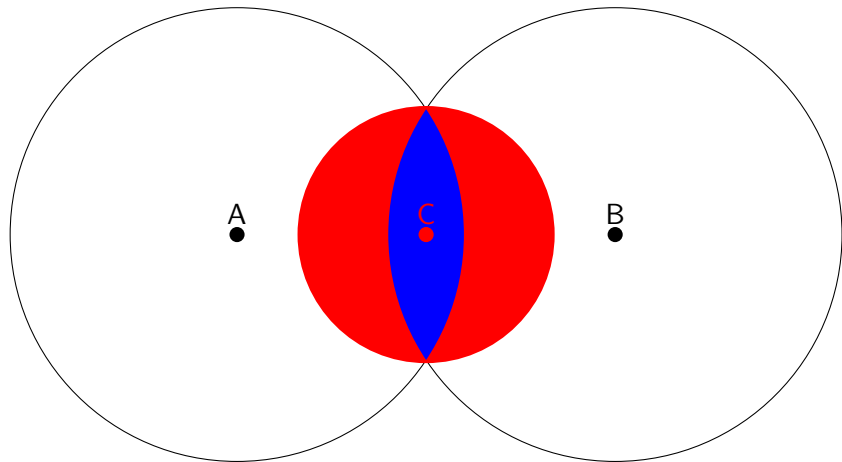
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# Expressing this ball

Using previous work: distance to measure ( $k$ -distance).

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$$f(x) = \sqrt{\|C - x\|^2 + \frac{\|A - B\|^2}{4}}$$



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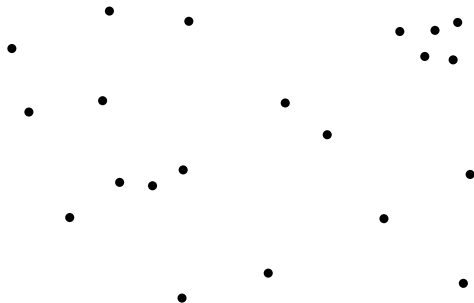
Take the middle point  $C = A + B$  and consider the power distance:

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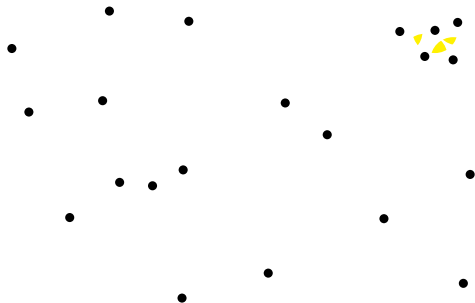
# Manicheism is too simple



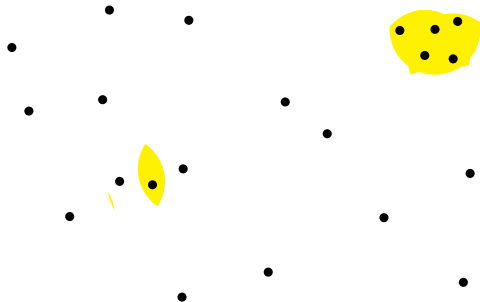
## Going to $k = 3$



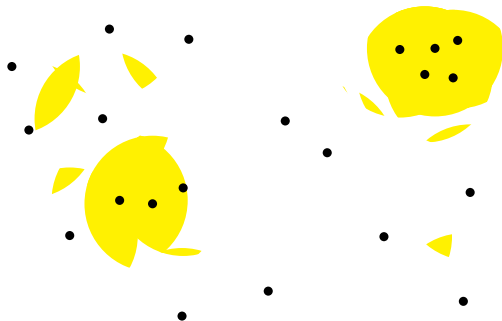
## Going to $k = 3$



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## $k$ -distance and barycentres

- Considering every set of  $k$ -points  $x_1, \dots, x_k$  we build the barycentre  $b$  and compute the weight  $w^2 = \frac{1}{k} \sum \|b - x_i\|^2$ .

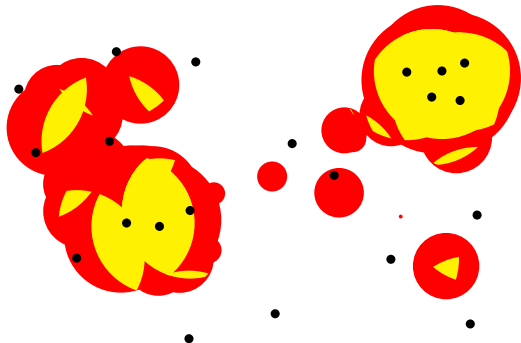
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- Then we look at the sublevel sets of the power distance:

$$f(x) = \min_{x_1, \dots, x_k} \sqrt{\|x - b\|^2 + w^2}$$



# Barycentric over-approximation

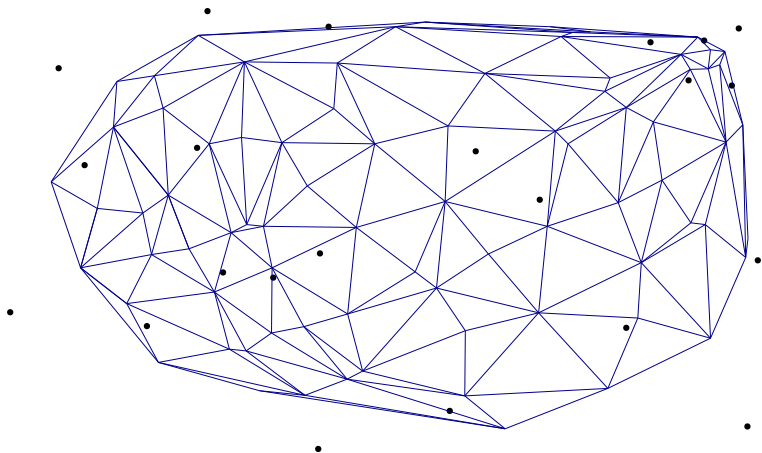


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- We have a large number of balls but we can restrict to the cells of the  $k$ -order Voronoi diagram.
- The result is a weighted Delaunay triangulation.

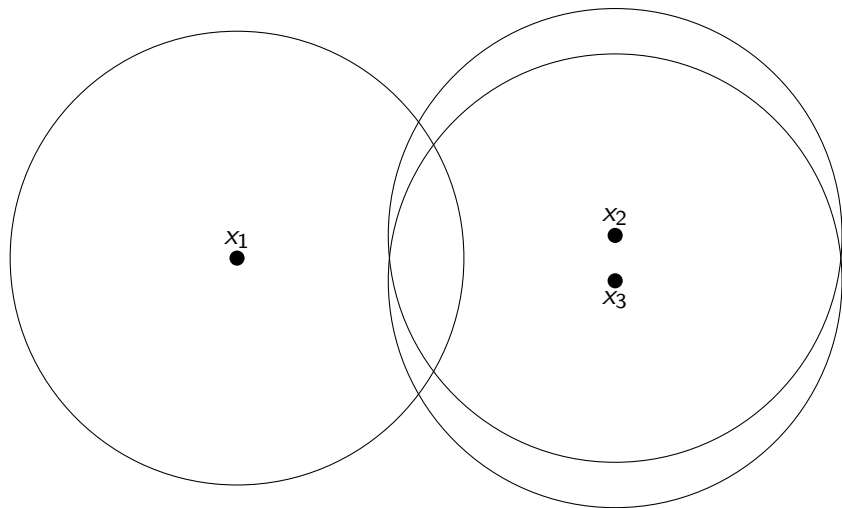
# Resulting triangulation



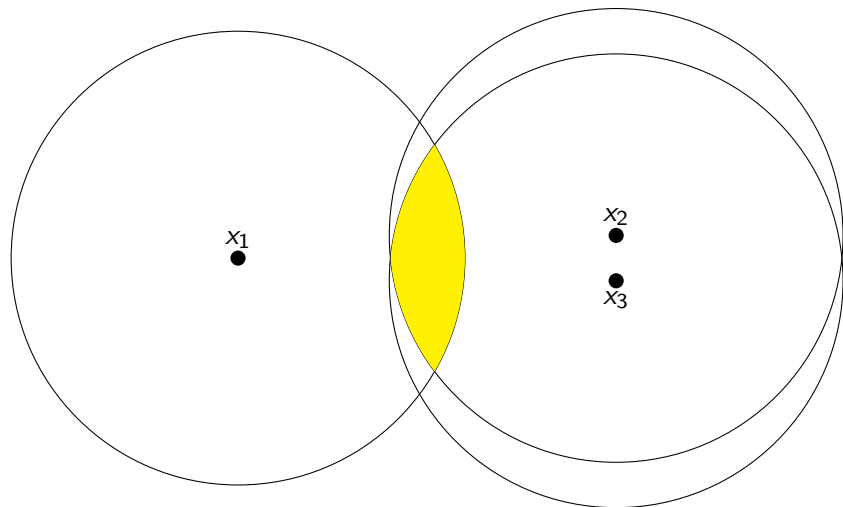
## Theorem

$$\forall k \geq 2, \forall \alpha > 0, \mathcal{K}_\alpha^{(k)} \subset \mathcal{A}_\alpha \subset \mathcal{K}_{\sqrt{k}\alpha}^{(k)}$$

# Location of barycentres

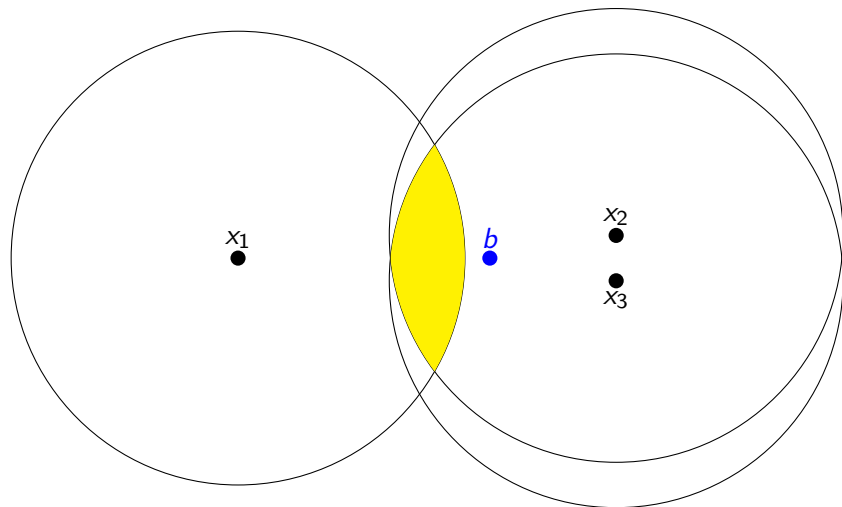


## Location of barycentres

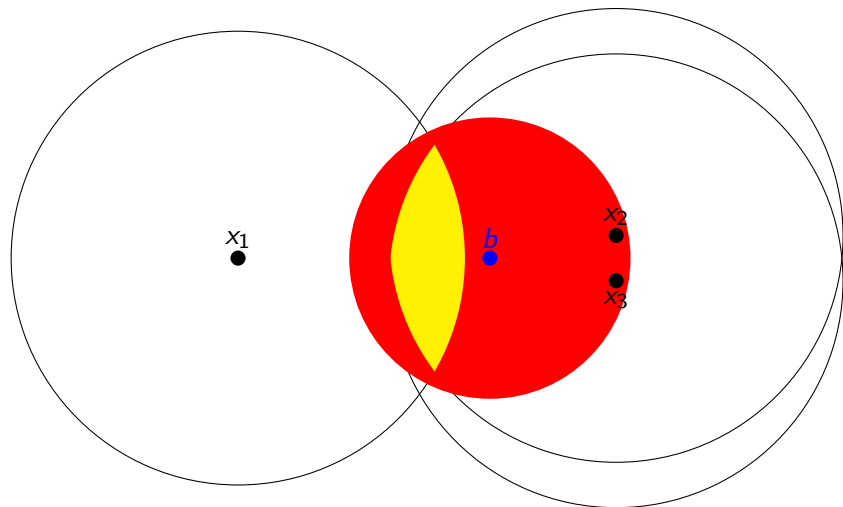




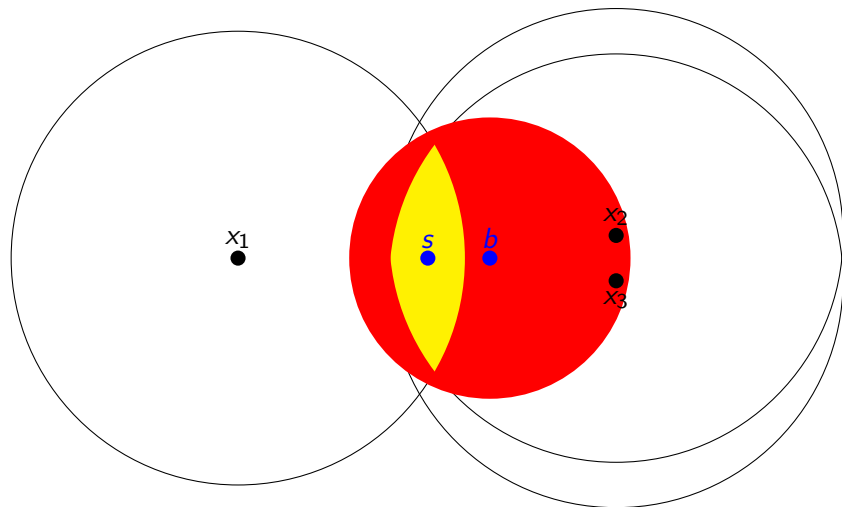
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# Minimum enclosing balls

- Replace barycenters by the center of the minimum enclosing ball.

# Minimum enclosing balls

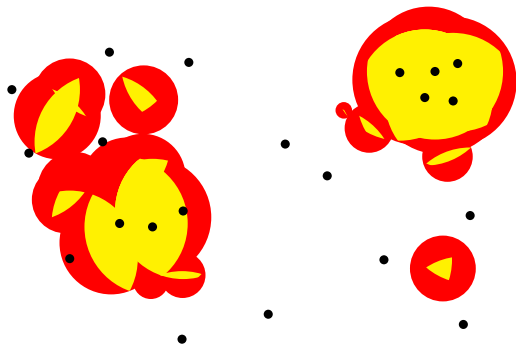
- Replace barycenters by the center of the minimum enclosing ball.
- Replace the weight by the radius  $R$  of the minimum enclosing ball.

# Minimum enclosing balls

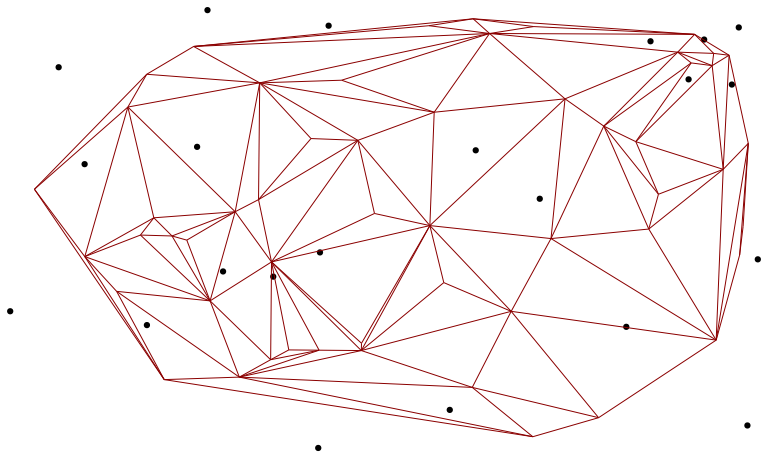
- Replace barycenters by the center of the minimum enclosing ball.
- Replace the weight by the radius  $R$  of the minimum enclosing ball.

$$f(x) = \min_{T=x_1, \dots, x_k} \sqrt{\|x - s_T\|^2 + R_T^2}$$

# Minimum enclosing balls covering



# MEB triangulation





## Theorem

$$\forall k \geq 2, \forall \alpha > 0, \mathcal{K}_\alpha^{(k)} \subset \mathcal{E}_\alpha \subset \mathcal{K}_{\sqrt{2}\alpha}^{(k)}$$

# The conjecture

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*Every non-empty cell in the MEB-diagram is also a non-empty cell in the  $k$ -order Voronoi diagram.*

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*Every simplex of the MEB-triangulation is also a simplex of the  $k$ -order Delaunay triangulation.*

## Summary (1)

- It is possible to approximate the  $k$ -fold filtration within a constant approximation factor using a weighted Delaunay triangulation.

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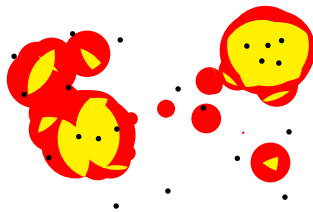
- It is possible to approximate the  $k$ -fold filtration within a constant approximation factor using a weighted Delaunay triangulation.
- We need to use a triangulation based on minimum enclosing balls rather than barycentres.

## Summary (1)

- It is possible to approximate the  $k$ -fold filtration within a constant approximation factor using a weighted Delaunay triangulation.
- We need to use a triangulation based on minimum enclosing balls rather than barycentres.
- This triangulation seems smaller than the  $k$ -order Delaunay triangulation.

# Summary (2)

Barycentres



MEBs

