



Parallel Meshing by Enumerating the Vertices of the Voronoi cells

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ALICE Géométrie & Lumière
CENTRE INRIA Nancy Grand-Est

OVERVIEW

Part 1. Introduction - Motivations

Part 2. Blowing Bubbles: CVT

Part 3. Anisotropy

Part 4. Journey in the 6th dimension

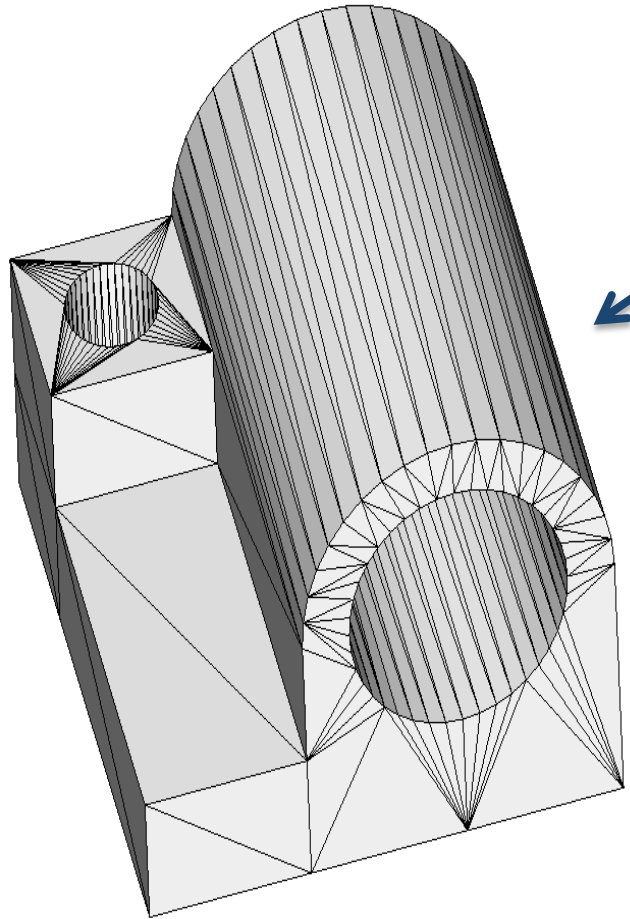
Part 5. The algorithm

Part 6. Results and conclusions

1

Introduction - Motivations

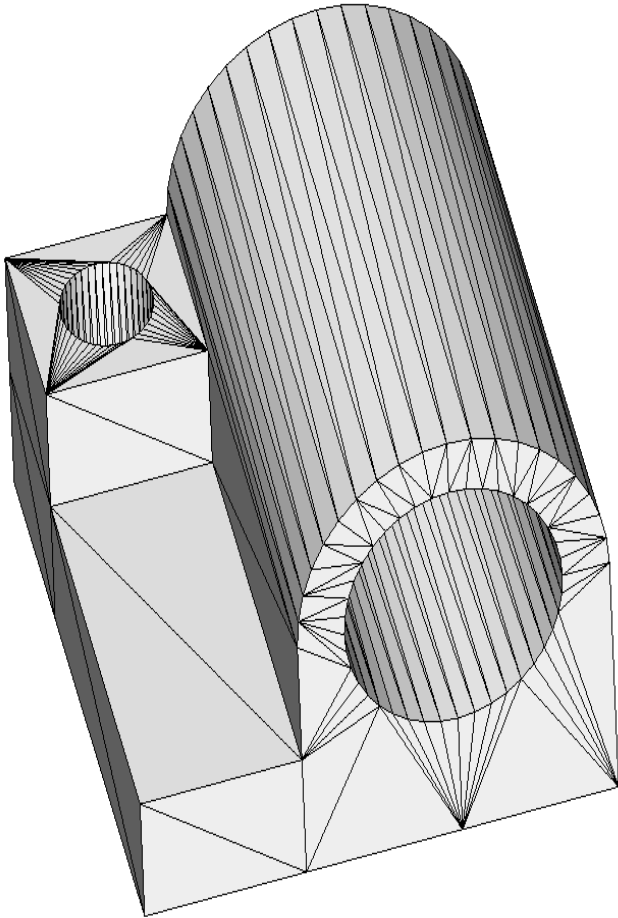
Part 1. Introduction: meshing (and re-meshing)



Input geometry, with “bad” triangles

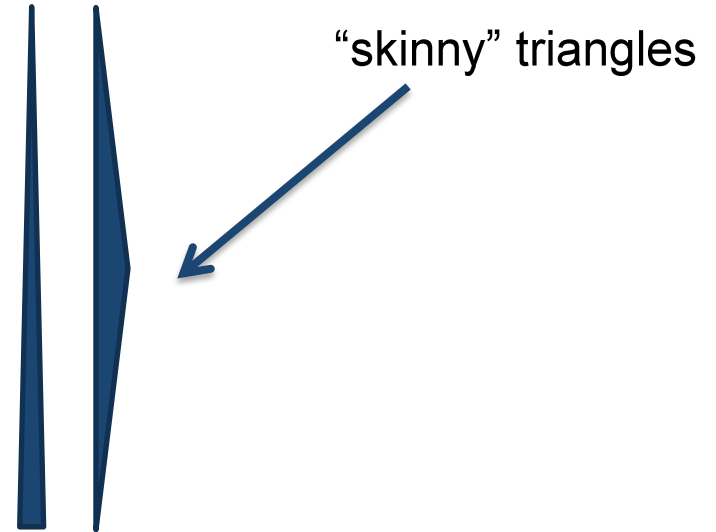
(Re) meshing [Du et.al], [Alliez et.al], [Yan, L et.al]

Part 1. Introduction: meshing (and re-meshing)



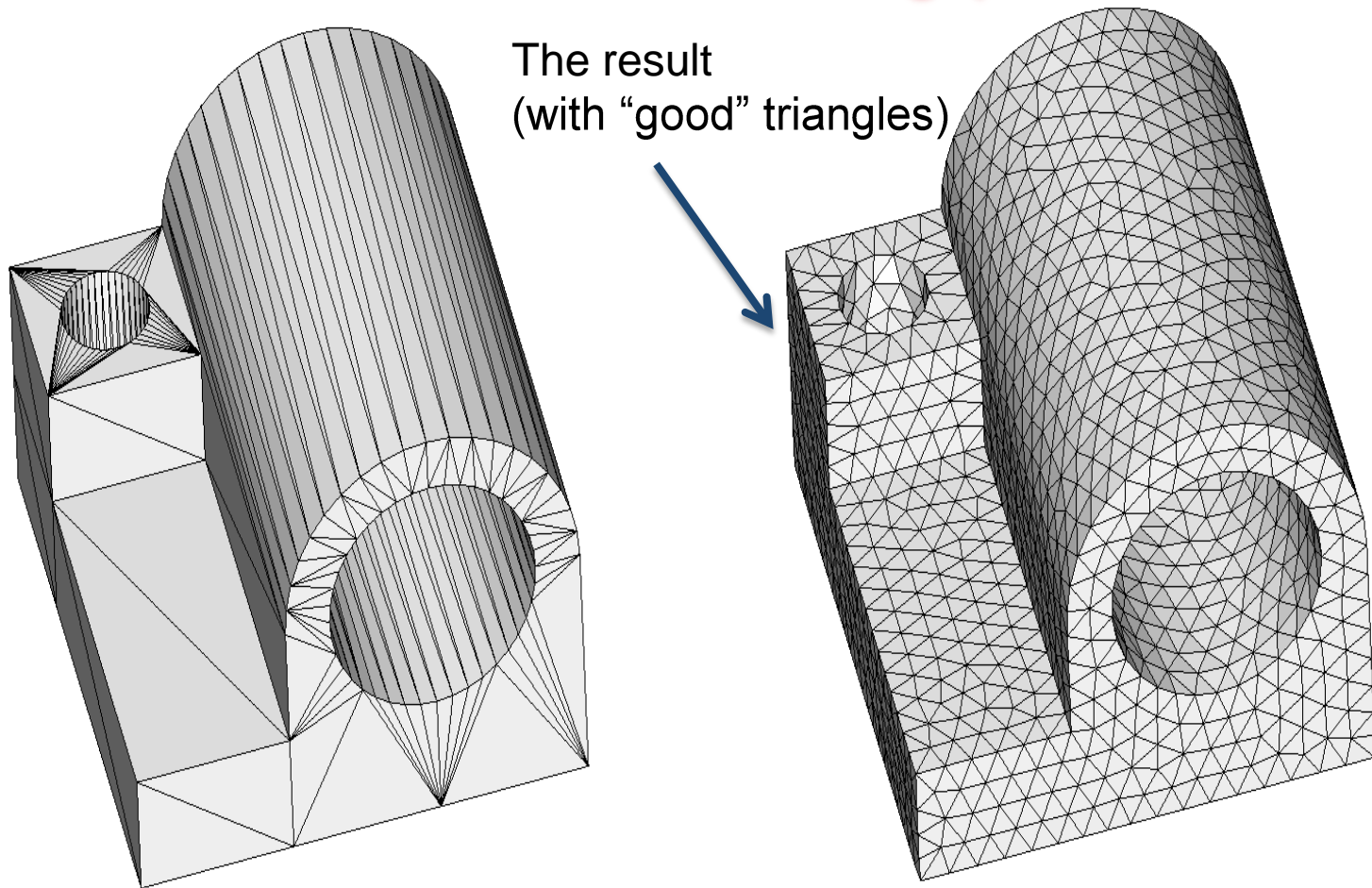
Why “bad” ?

Because extreme angles (near 0° or 180°) can cause numerical instabilities.



(Re)-meshing [Du et.al], [Alliez et.al], [Yan, L et.al]

Part 1. Introduction: meshing (and re-meshing)

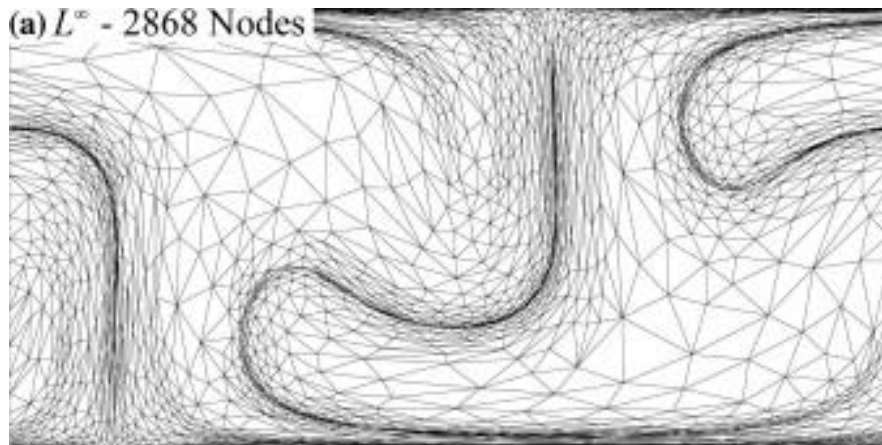


(Re)-meshing [Du et.al], [Alliez et.al], [Yan, L et.al]

Part 1. Introduction: meshing (and re-meshing)



It has skinny triangles where we want them and oriented as we like (following the variation of the physics)

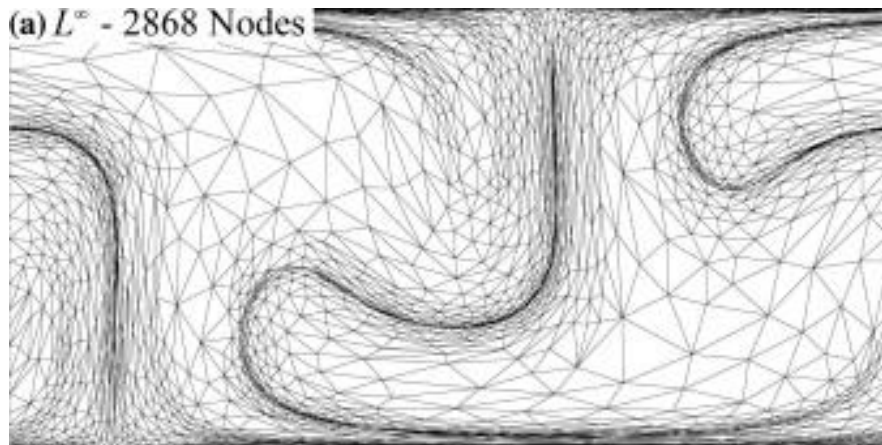


Solution-adapted mesh [Miron et.al, Journal of Computational Physics, 2010]

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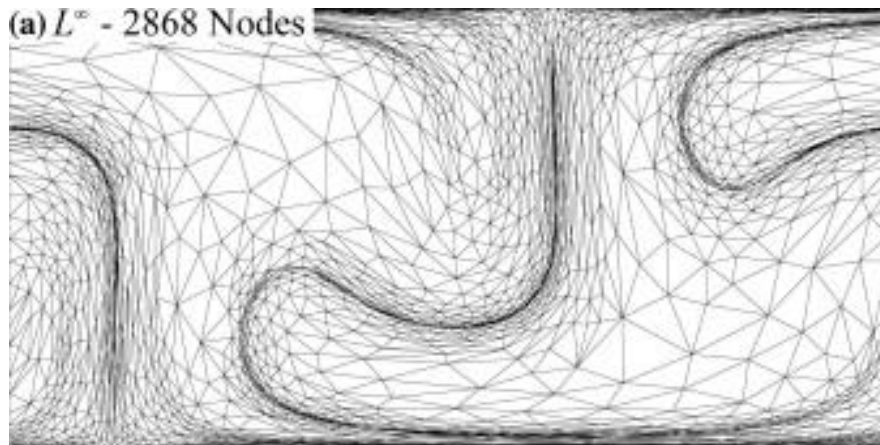
Benefit: higher accuracy with smaller number of elements.

Solution-adapted mesh [Miron et.al, Journal of Computational Physics, 2010]

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It has skinny triangles where we want them and oriented as we like (following the variation of the physics)

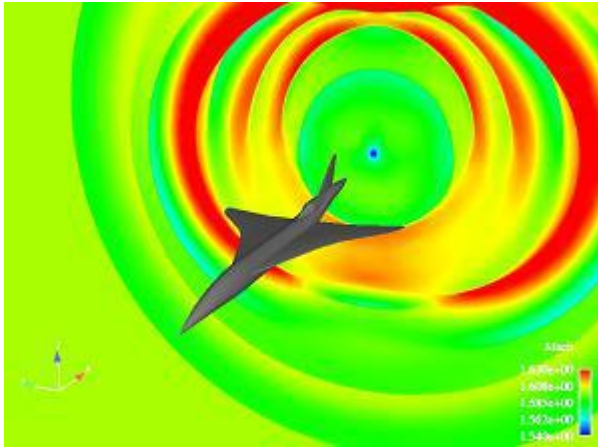


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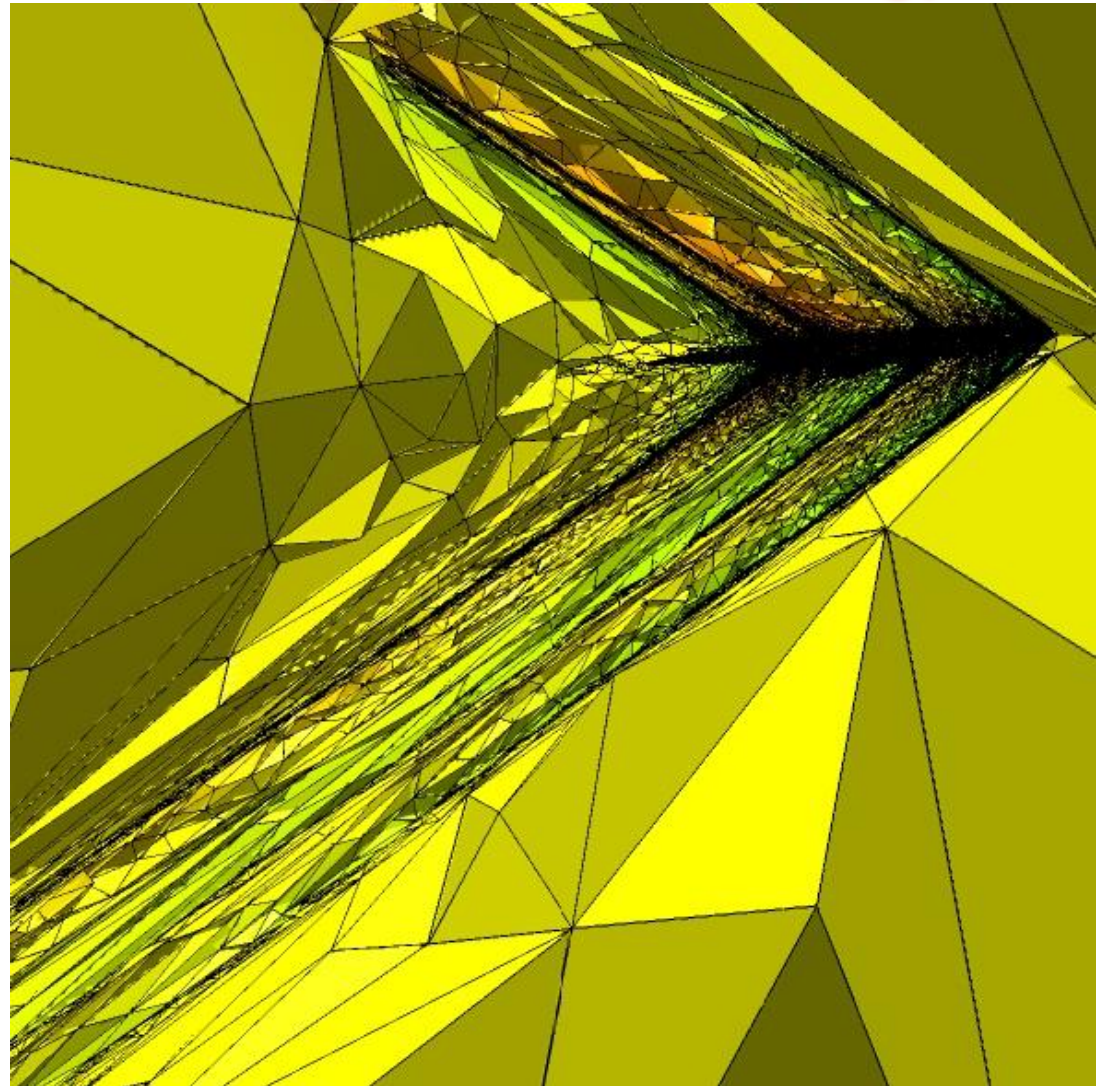
Skinny triangles: not always “bad”, even sometimes **desired** but with **controlled** shape, size and orientation.

Solution-adapted mesh [Miron et.al, Journal of Computational Physics, 2010]

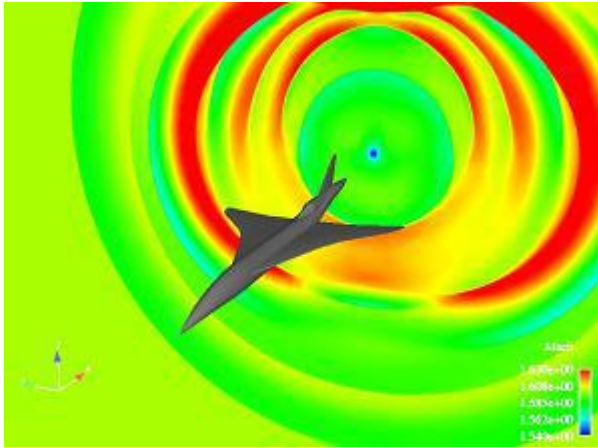
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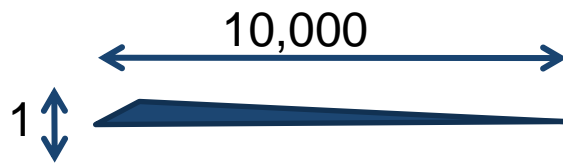
Supersonic flight
[Alauzet et.al]



Part 1. Introduction: meshing (and re-meshing)

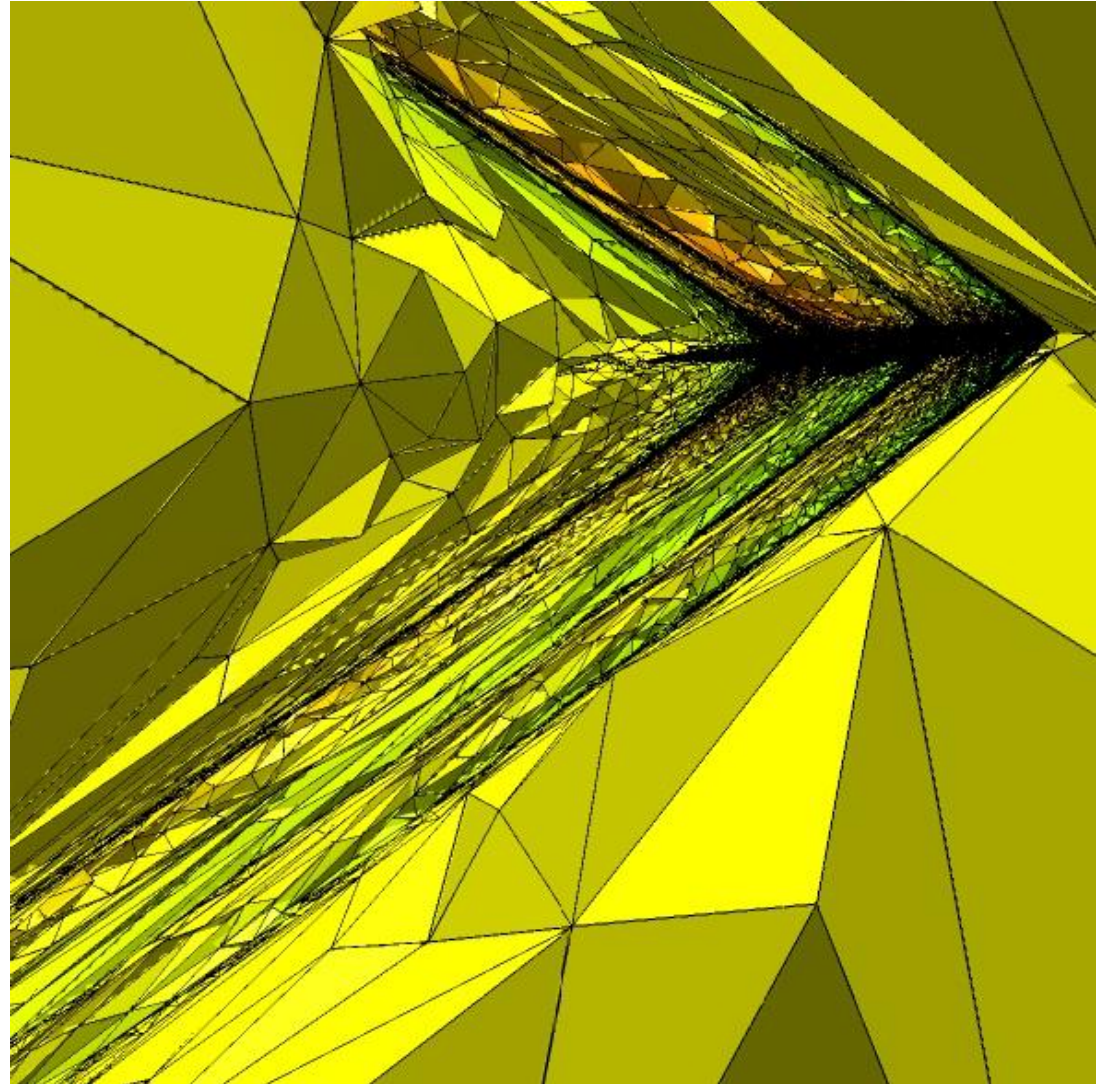


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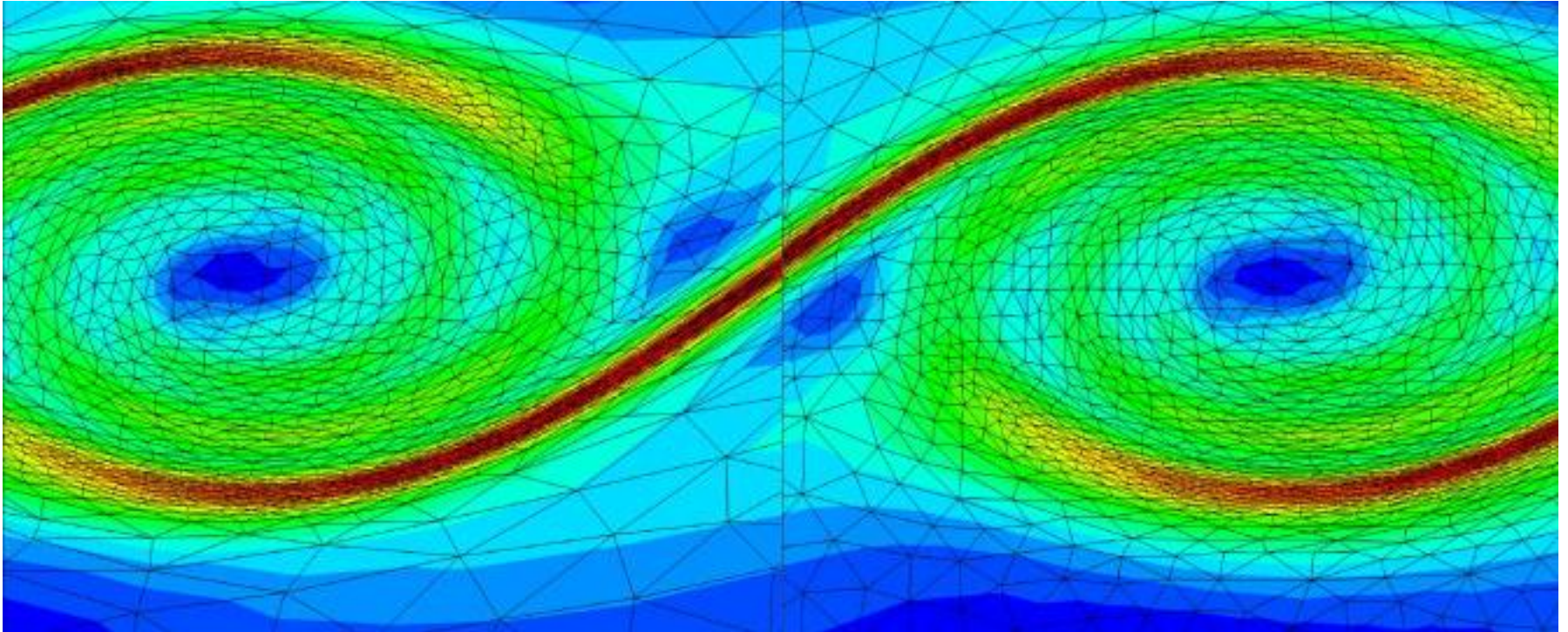


Aspect ratio:

1:10,000 (typically)



Part 1. Introduction: meshing (and re-meshing)



Solution-adapted **anisotropic** mesh

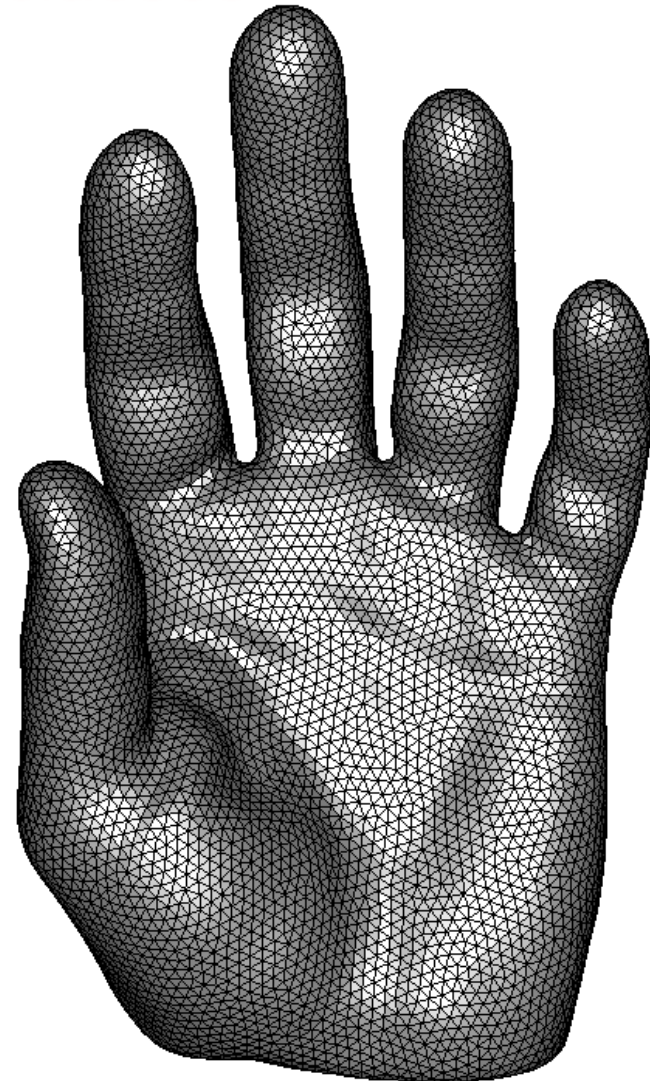
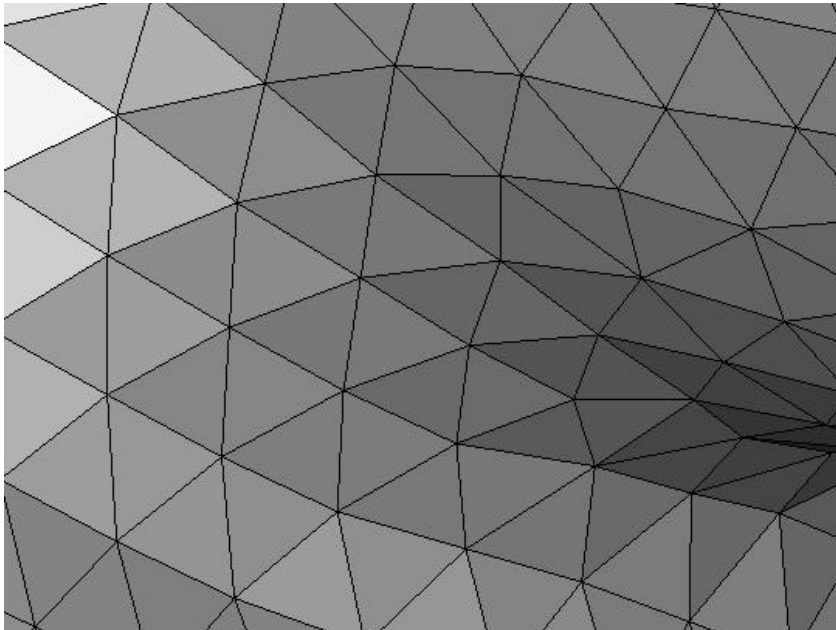
[Miron et.al, Journal of Computational Physics, 2010]

Part 1. Introduction : mesh classes

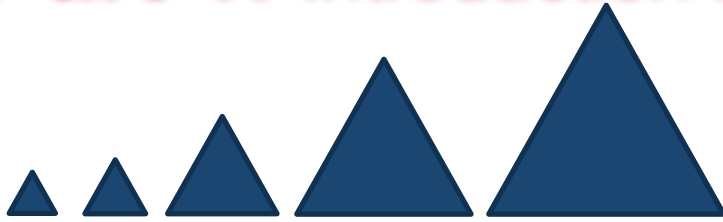


Isotropic mesh: All the triangles have

- * the same shape (equilateral)
- * the same size

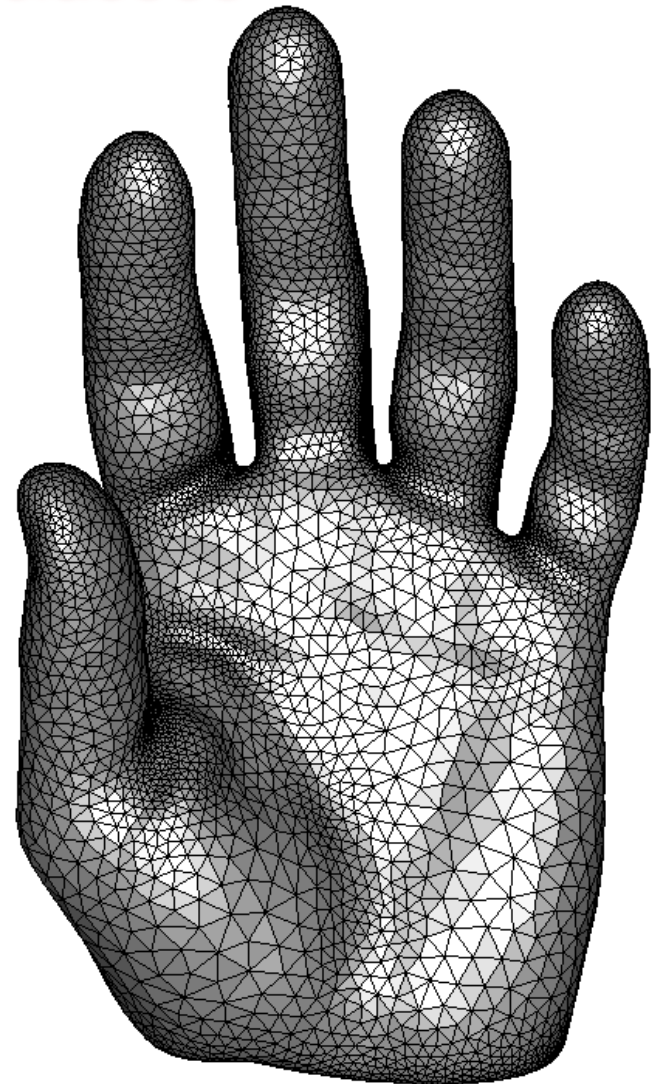
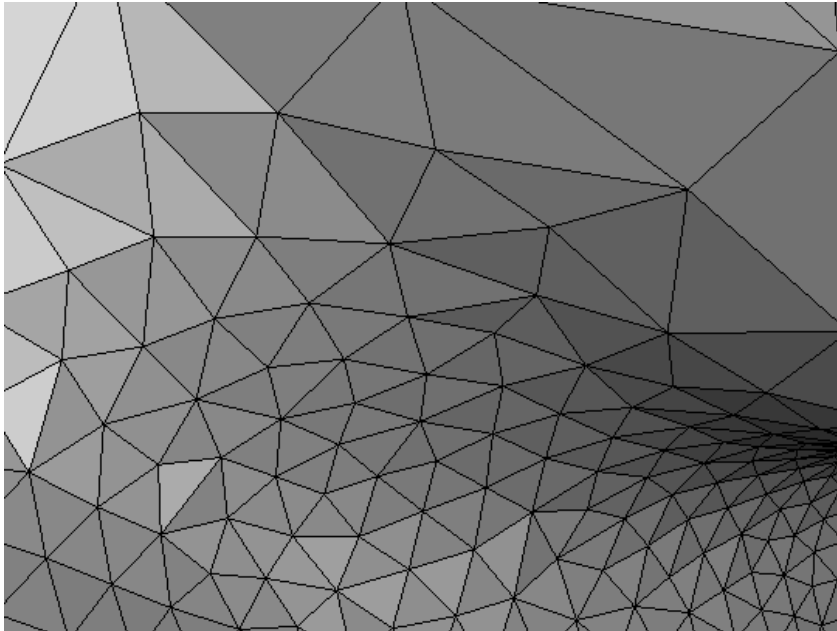


Part 1. Introduction : mesh classes



Isotropic graded mesh:

- * the same shape (equilateral)
- * size can vary

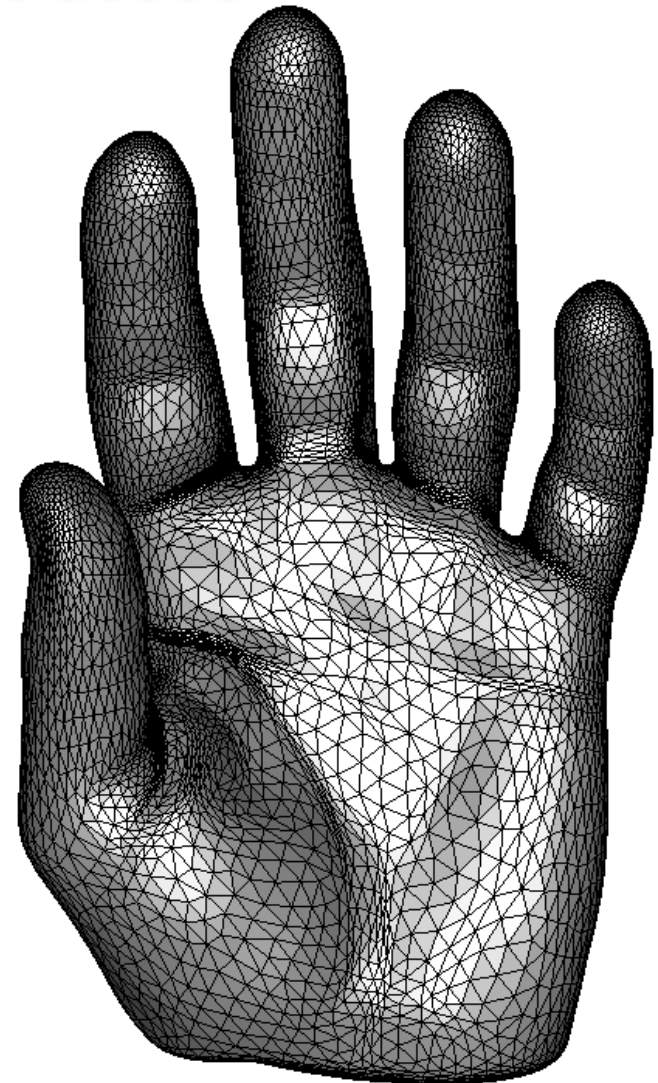
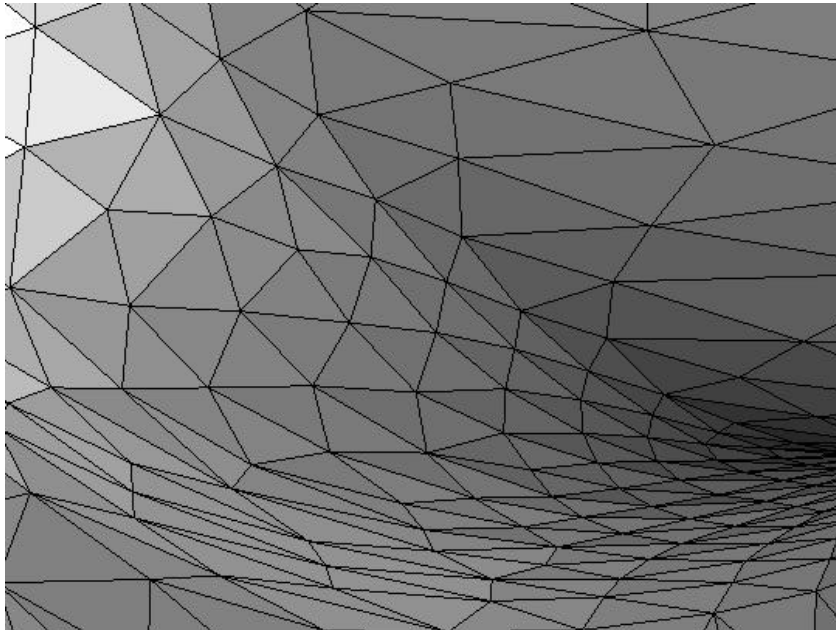


Part 1. Introduction : mesh classes



Anisotropic mesh:

- * shape can vary
- * size can vary



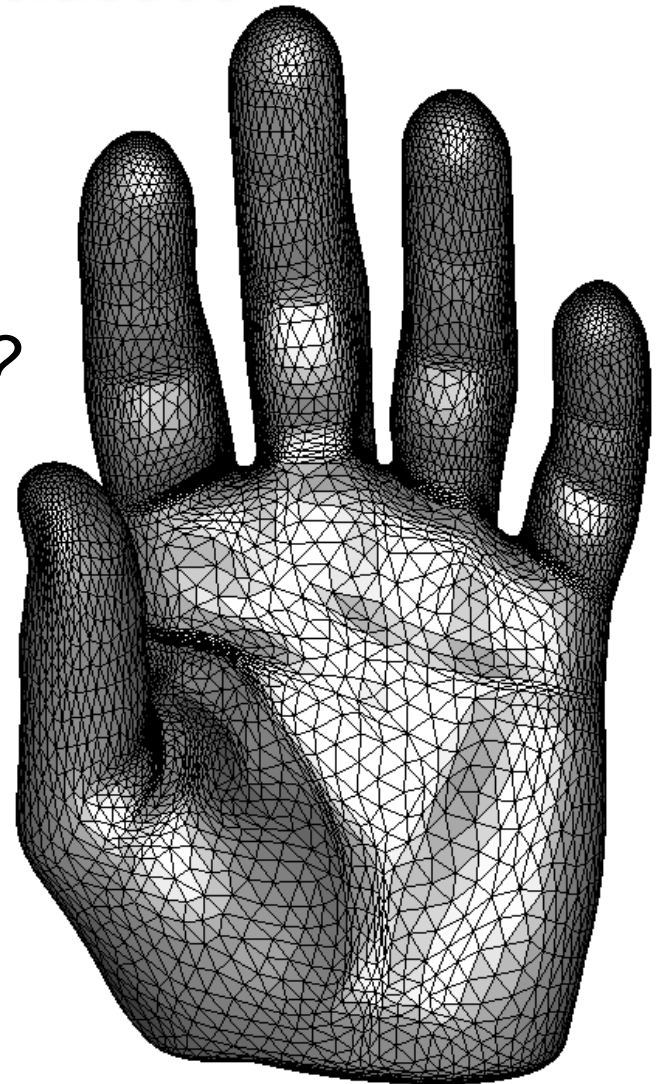
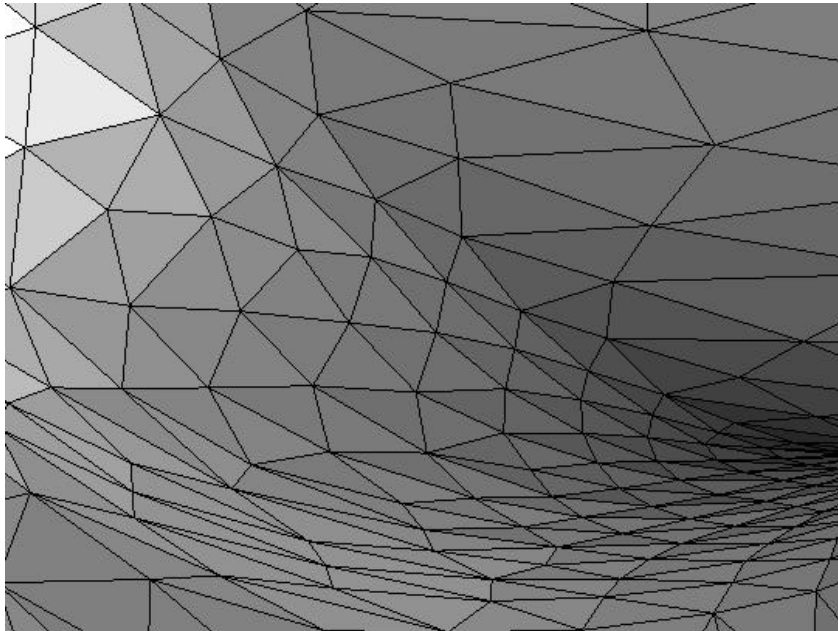
Part 1. Introduction : mesh classes



Anisotropic mesh:

- * shape can vary
- * size can vary

Q: How to generate an anisotropic surface mesh ?



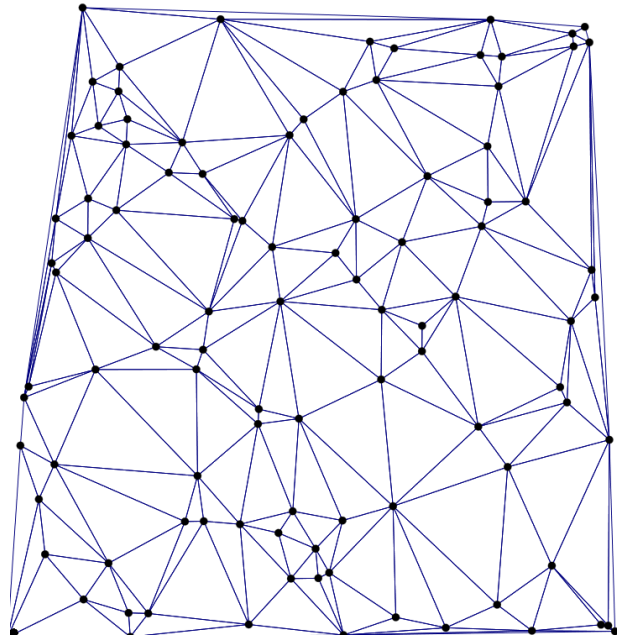
2

Blowing Bubbles

Centroidal Voronoi Tessellations

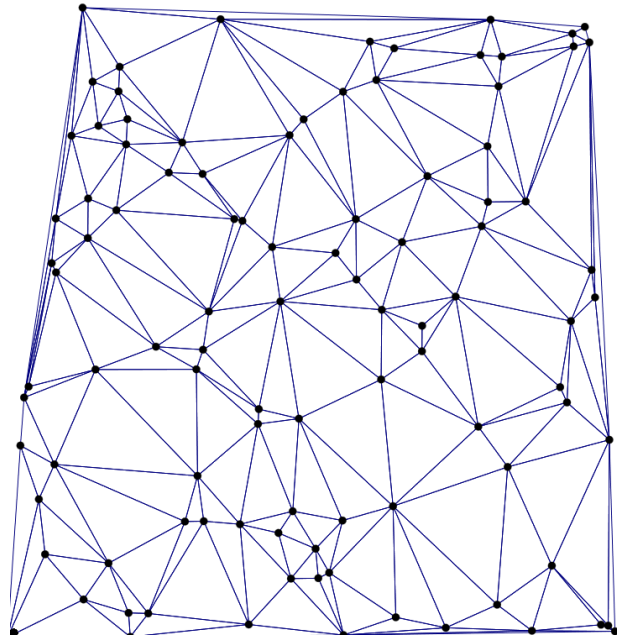
Part. 2. Centroidal Voronoi Tessellation

Optimize a Voronoi diagram from the point of view of sampling regularity
(quantization noise power)



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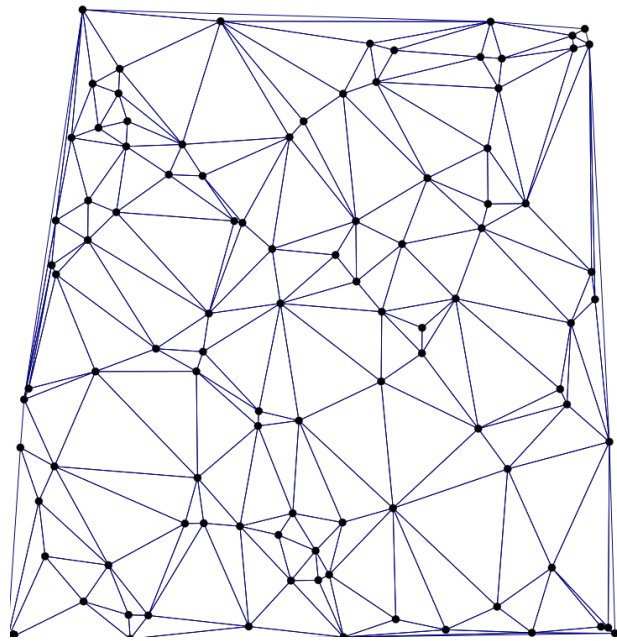


Minimize

$$F = \sum_i \int_{\text{Vor}(i)} \left\| \mathbf{x}_i - \mathbf{x} \right\|^2 d\mathbf{x}$$

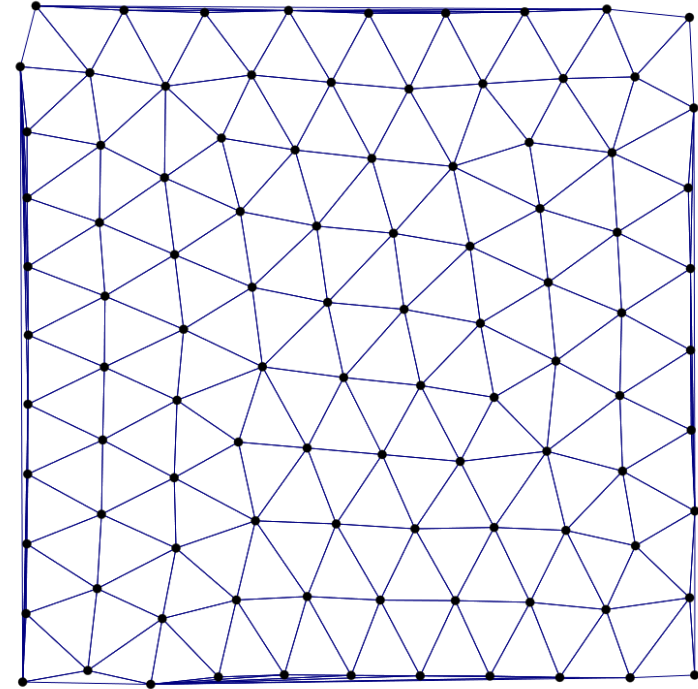
Part. 2. Centroidal Voronoi Tessellation

Optimize a Voronoi diagram from the point of view of sampling regularity
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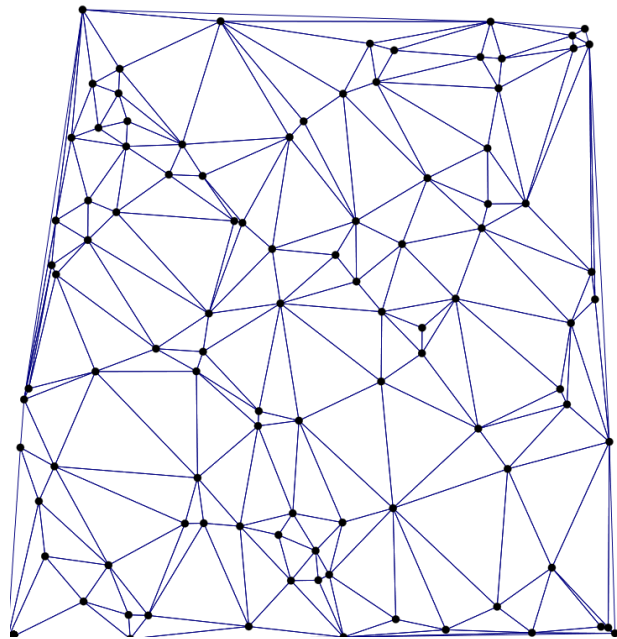
Minimize

$$F = \sum_i \int_{\text{Vor}(i)} \| \mathbf{x}_i - \mathbf{x} \|^2 dx$$



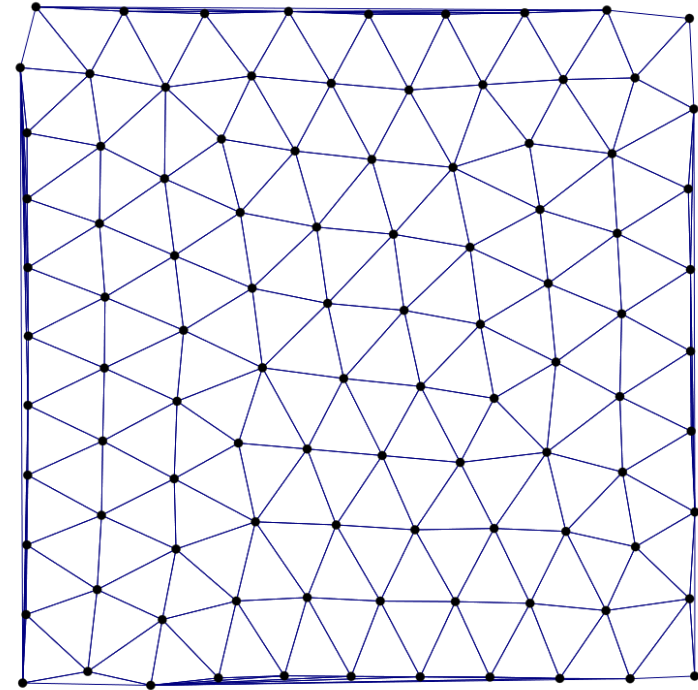
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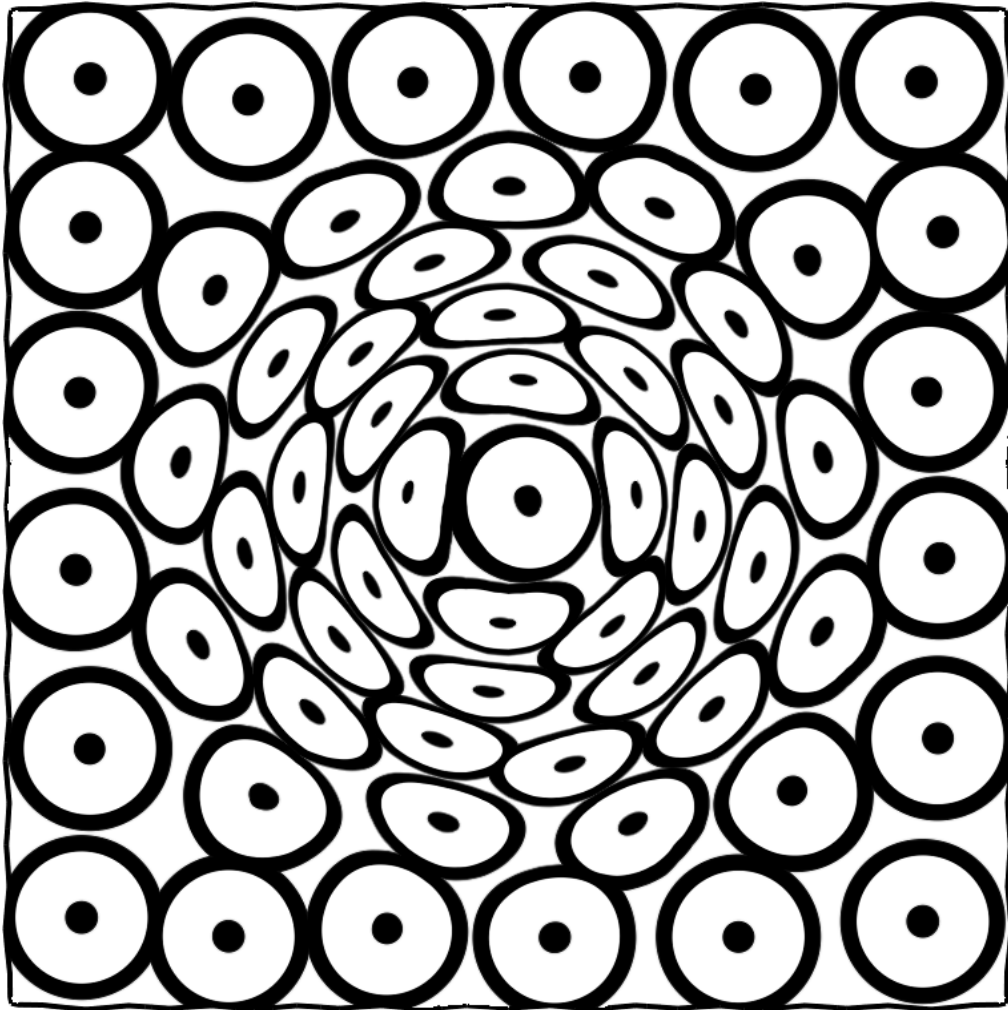
Theorem: F is of class C^2 [Liu, Wang, L, Yan, Lu, ACM TOG 2008]

Part. 2. Centroidal Voronoi Tessellation

3

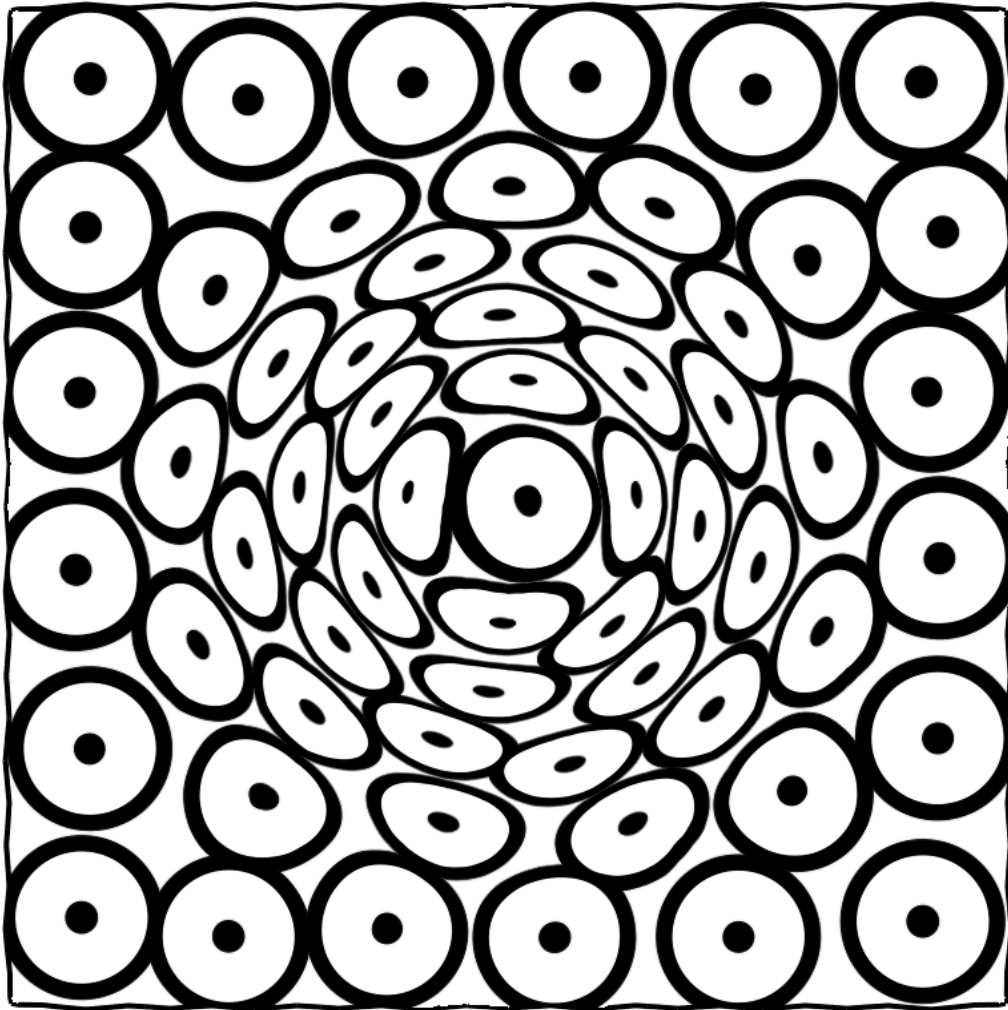
Anisotropy

Part. 3. Anisotropy



The input: anisotropy field
Specifies shape and orientation

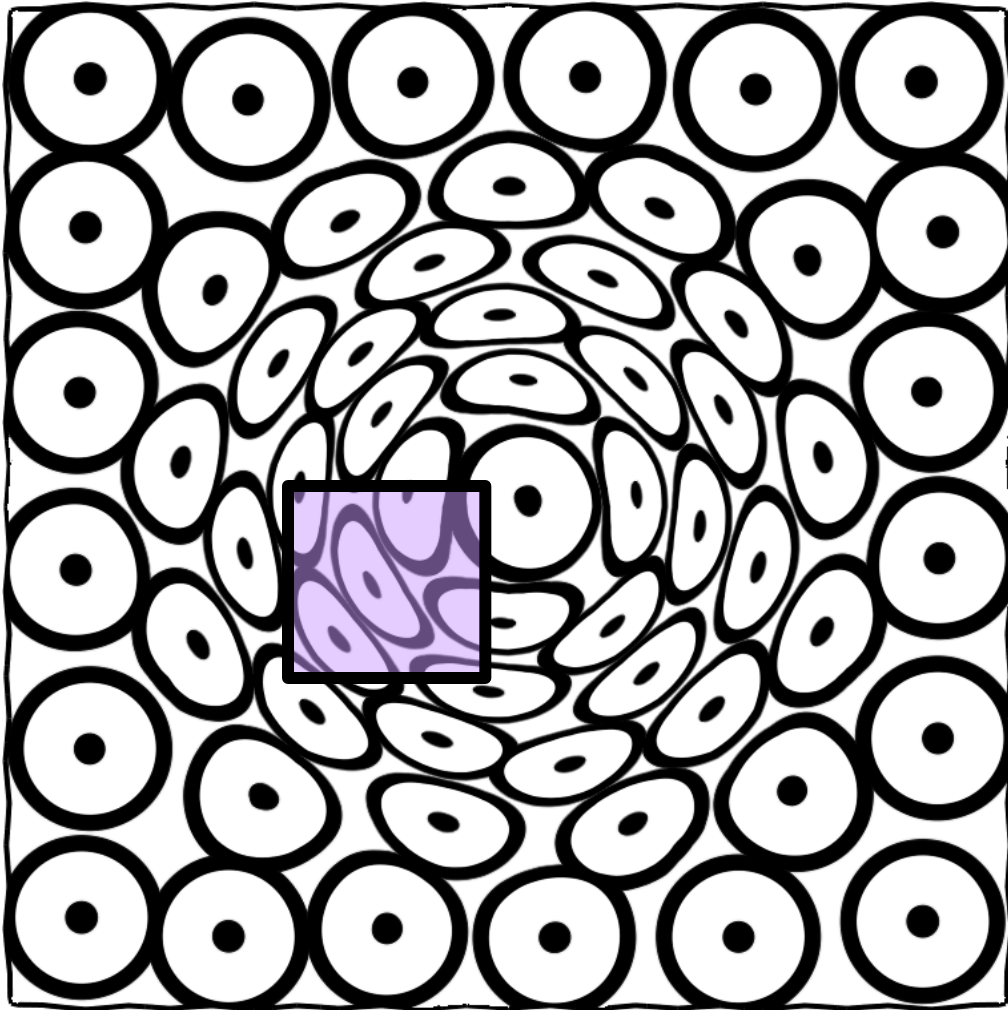
Part. 3. Anisotropy



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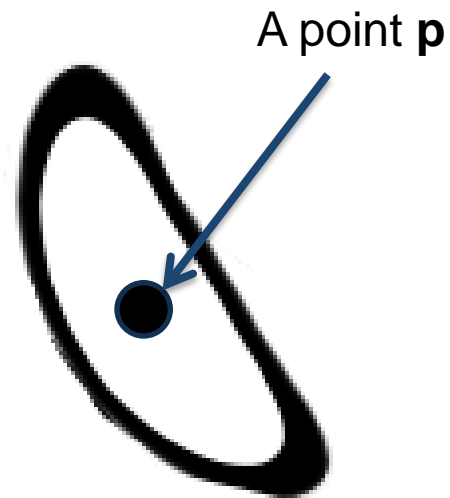
Anisotropy: An “alteration” of
of distances and angles.

Part. 3. Anisotropy

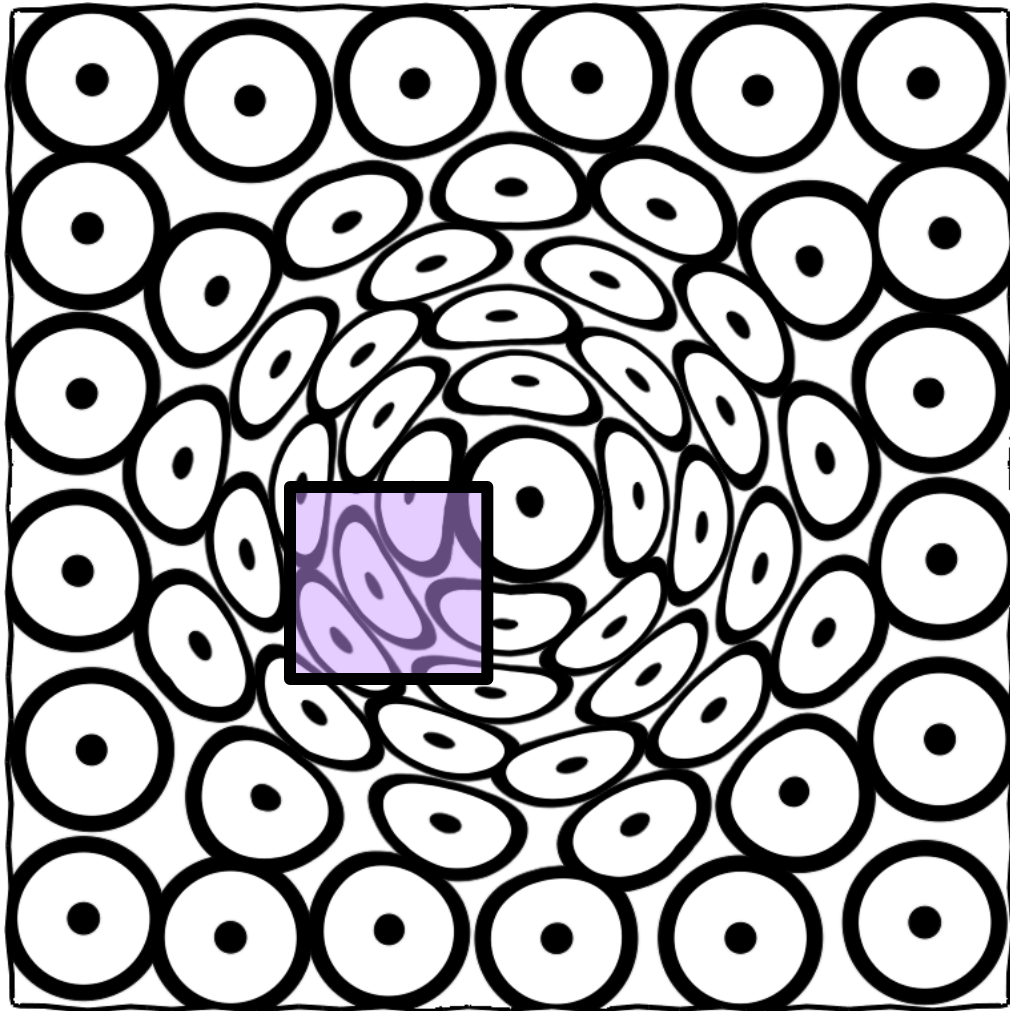


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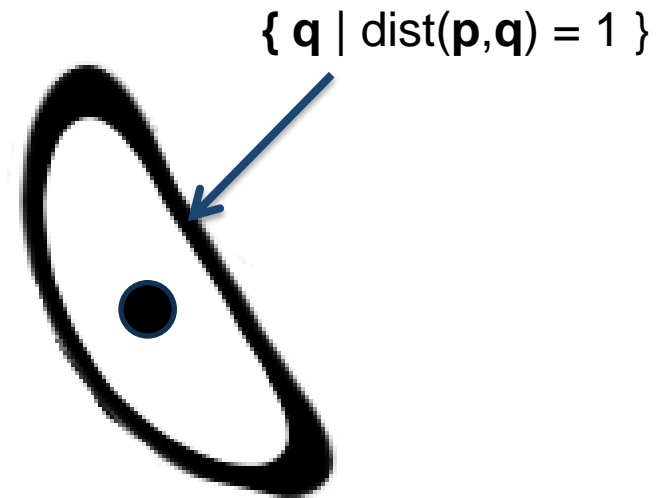


Part. 3. Anisotropy

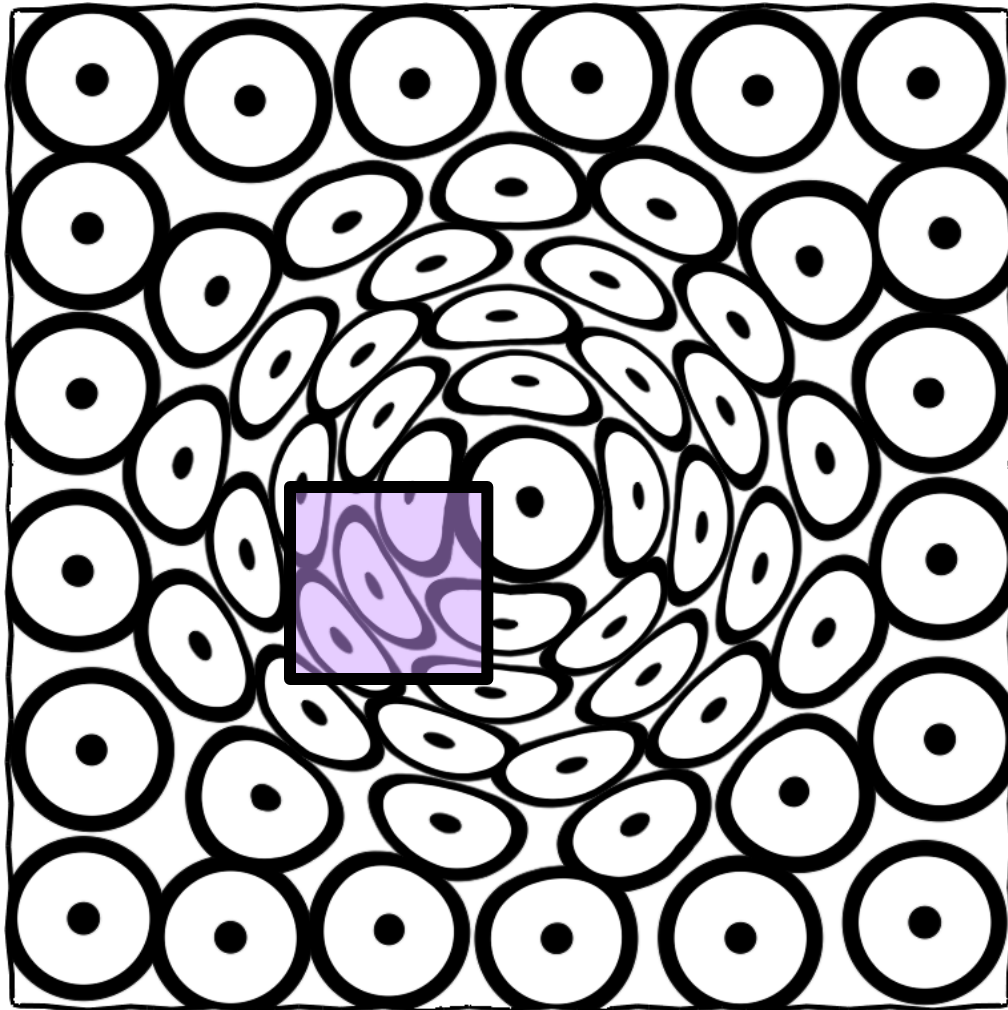


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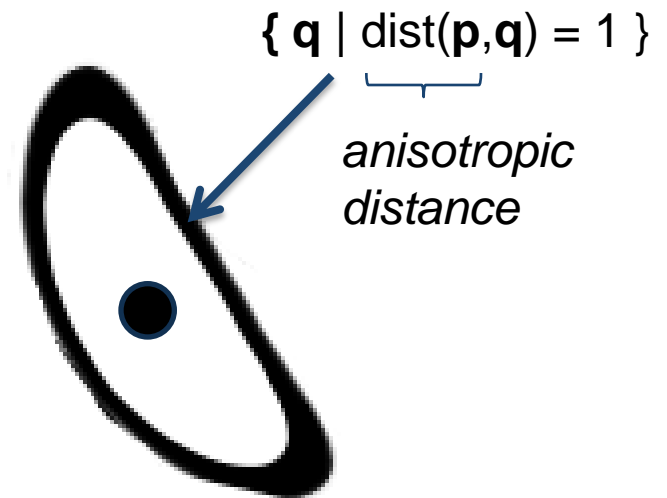


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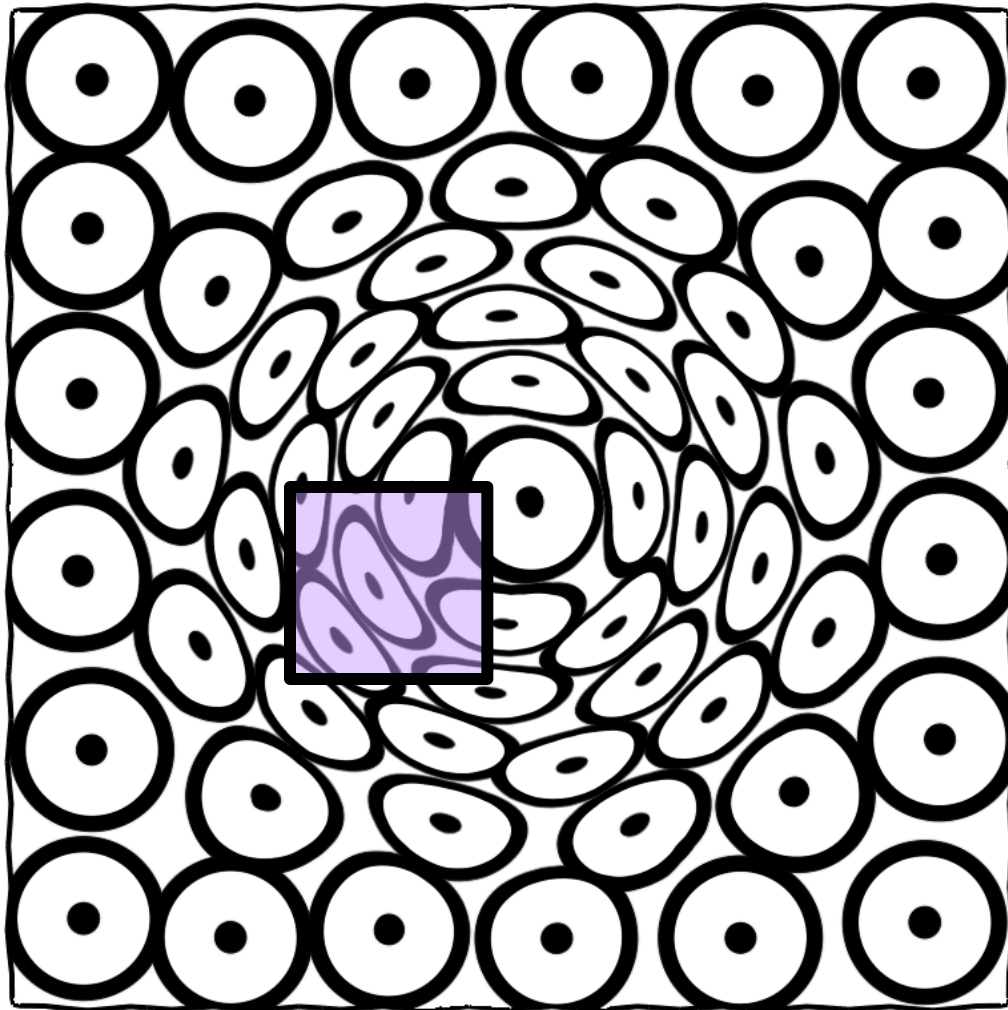


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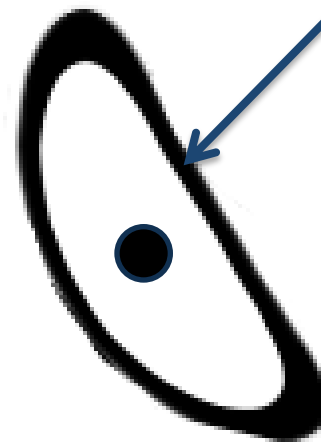
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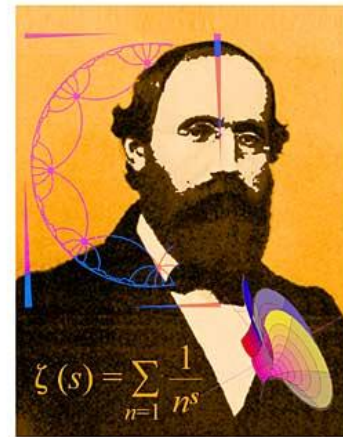
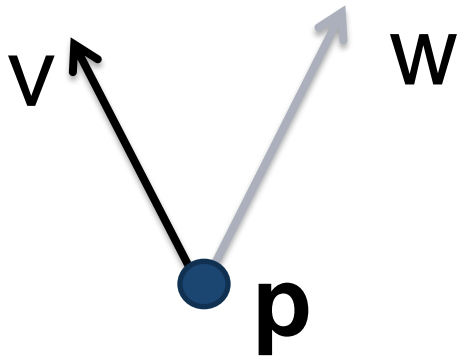
This is a circle !
 $\{ q \mid \text{dist}(p, q) = 1 \}$



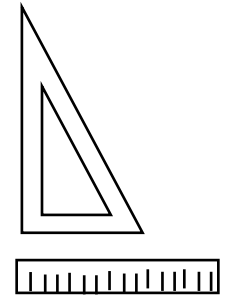
*anisotropic
distance*

Part. 3. Anisotropy

The dot product: a geometric tool



Götting Friedrich Bernhard Riemann. 1826 - 1866

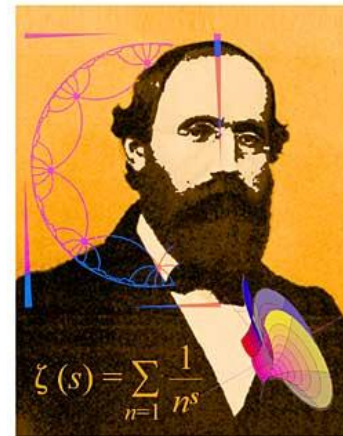
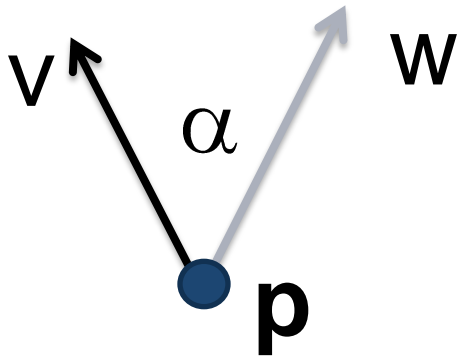


$$v \cdot w = \langle v, w \rangle = v^t w$$

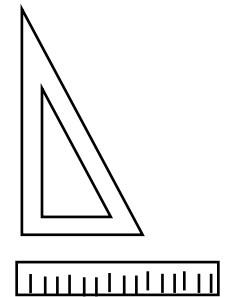
Part. 3. Anisotropy

The dot product: a geometric tool

Measuring angles



Göttinger Friedrich Bernhard Riemann. 1826 - 1866

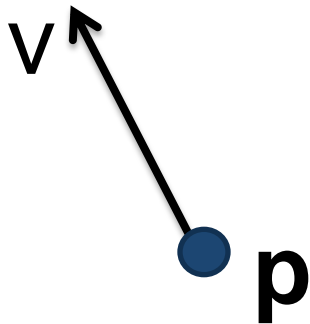


$$\cos(\alpha) = \frac{\langle v.w \rangle}{\sqrt{\langle v.v \rangle \langle w.w \rangle}}$$

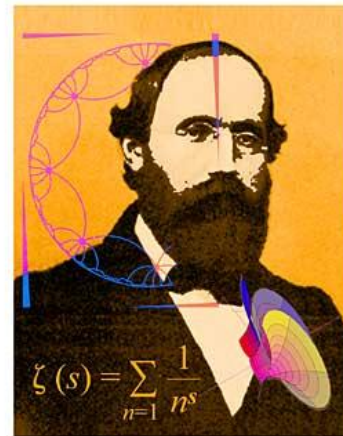
Part. 3. Anisotropy

The dot product: a geometric tool

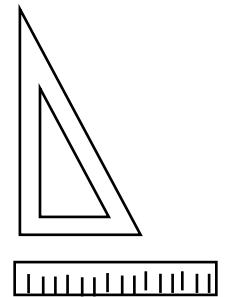
Measuring length



$$\|v\| = \sqrt{\langle v, v \rangle}$$



Götting Friedrich Bernhard Riemann. 1826 - 1866



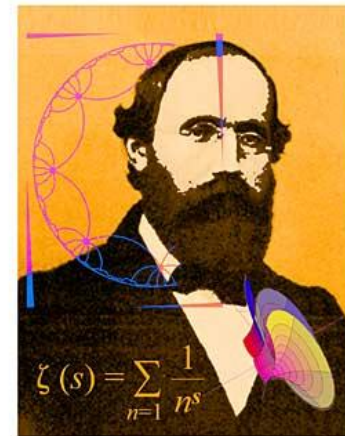
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The dot product: a geometric tool

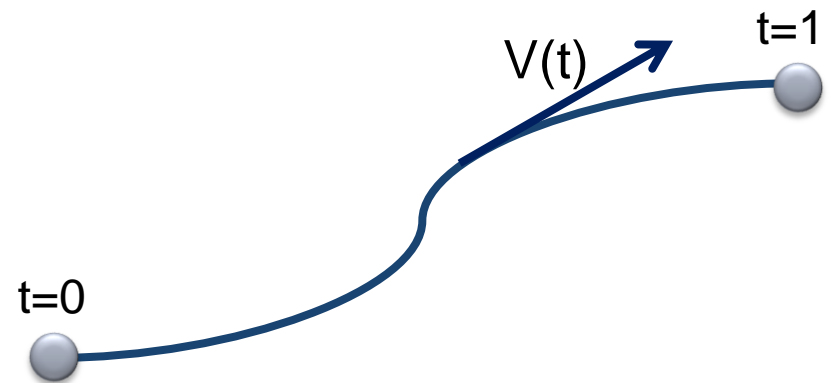
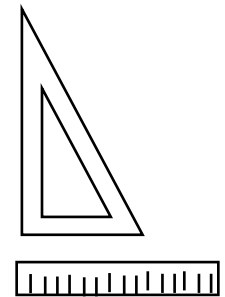
Measuring the length of a curve

$$l(C) = \int_{t=0}^1 \|v(t)\| dt$$

$$= \int_{t=0}^1 \sqrt{\langle v(t), v(t) \rangle} dt$$



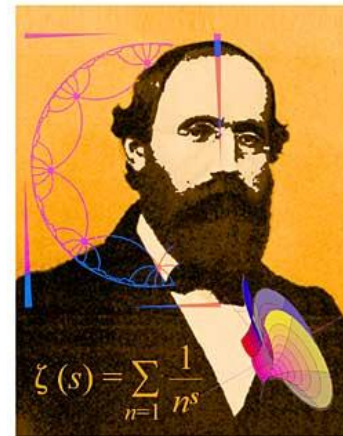
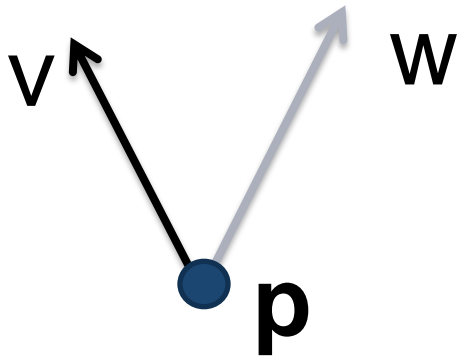
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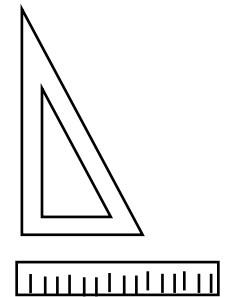
Part. 3. Anisotropy

The dot product: a geometric tool

Changing the dot product



Götting Friedrich Bernhard Riemann. 1826 - 1866

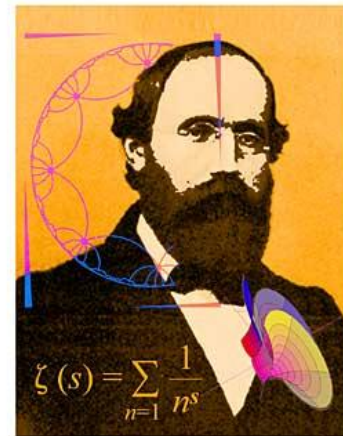
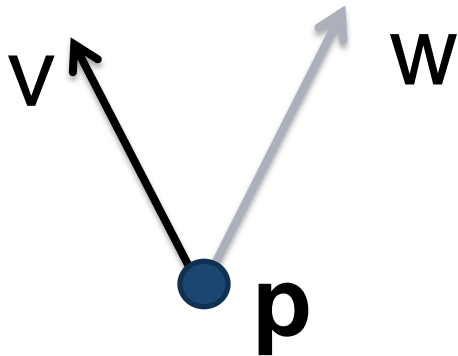


$$v \cdot w = \langle v, w \rangle = v^t \text{Id } w$$

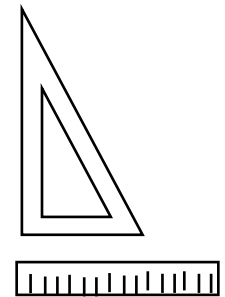
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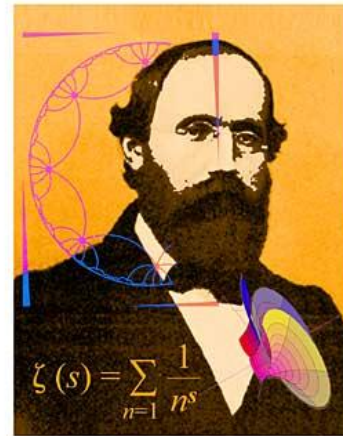
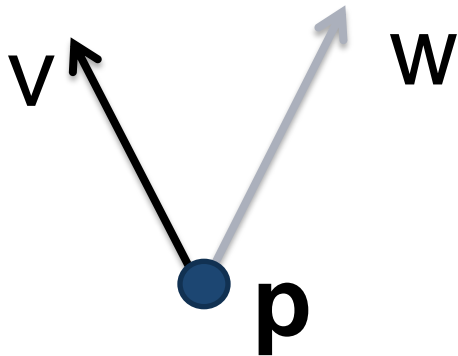
$$v \cdot w = \langle v, w \rangle = v^t \text{Id } w$$

$$\langle v, w \rangle_G = v^t G(p) w$$

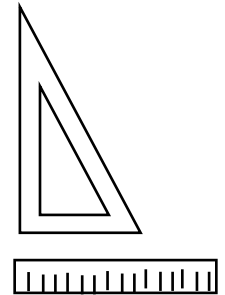
Part. 3. Anisotropy

The dot product: a geometric tool

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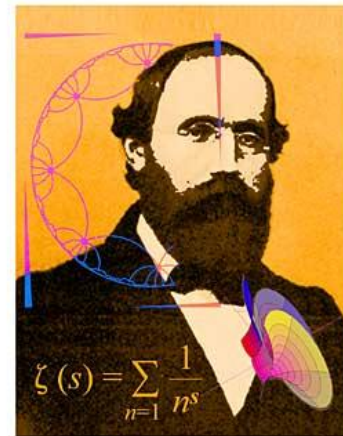
$$\langle v, w \rangle_G = v^t \mathbf{G}(p) w$$

A 2x2 symmetric matrix that depends on p

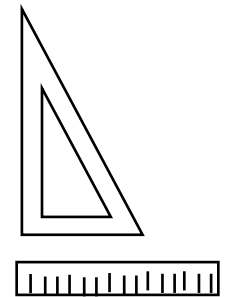
Part. 3. Anisotropy

The dot product: a geometric tool

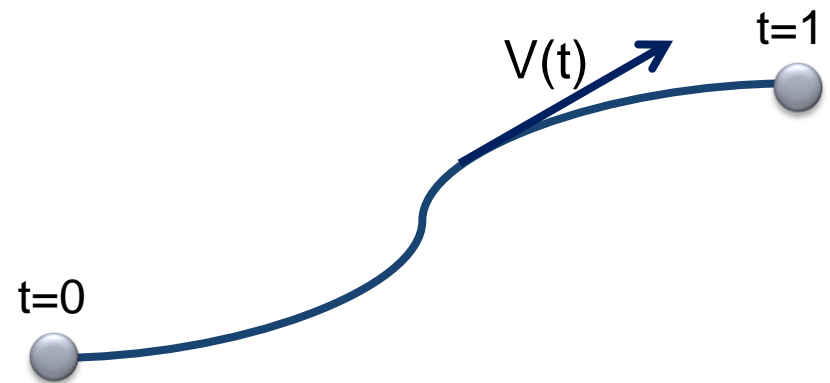
Measuring the anisotropic length of a curve



Götting Friedrich Bernhard Riemann. 1826 - 1866



$$l_G(C) = \int_{t=0}^1 \sqrt{v(t)^t G(t) v(t)} dt$$

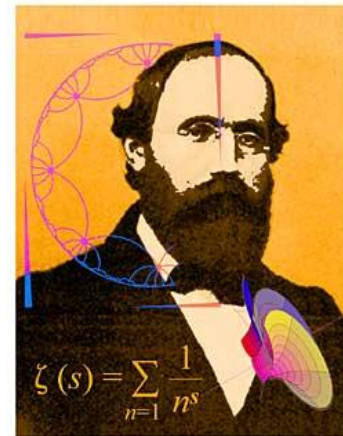


Part. 3. Anisotropy

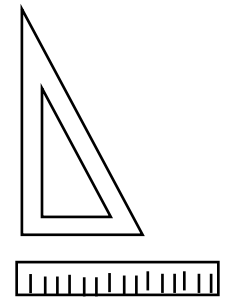
The dot product: a geometric tool

Anisotropic distance between \mathbf{p} and \mathbf{q} w.r.t. G

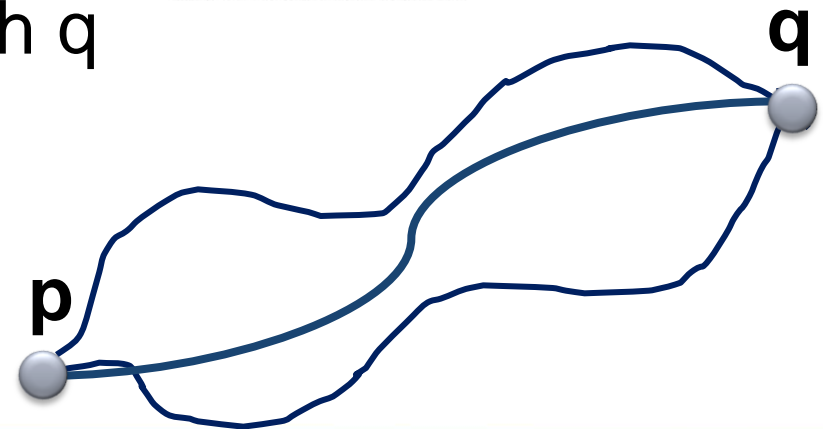
$d_G(\mathbf{p}, \mathbf{q}) =$ (anisotropic) length of shortest curve that connects \mathbf{p} with \mathbf{q}



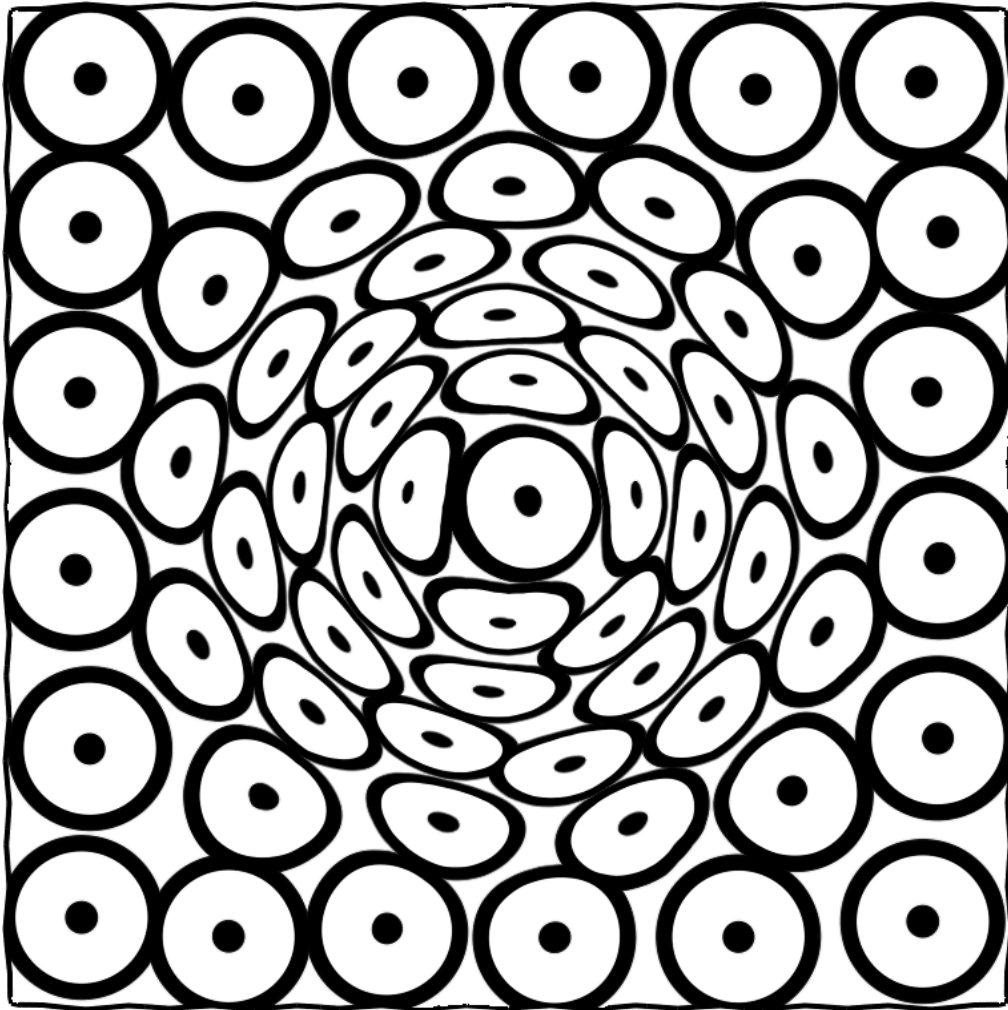
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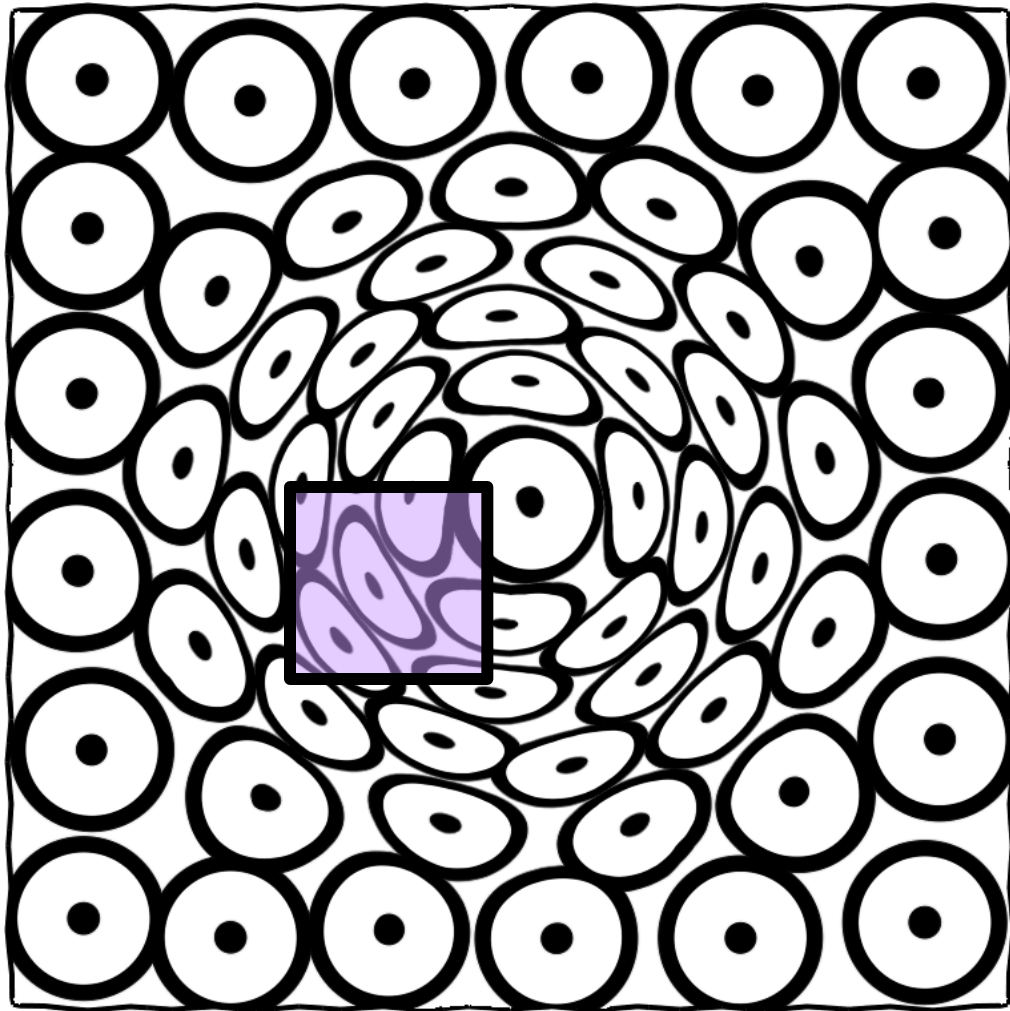
Part. 3. Anisotropy



The input: anisotropy field

$$G(x,y) = \begin{bmatrix} a(x,y) & b(x,y) \\ b(x,y) & c(x,y) \end{bmatrix}$$

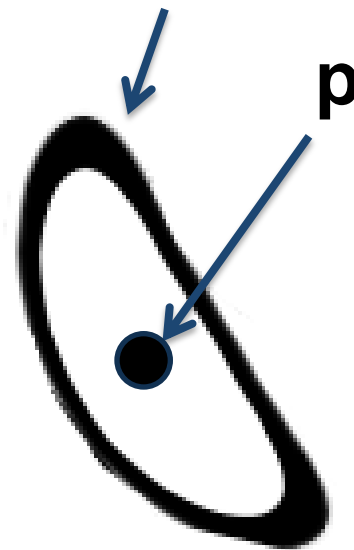
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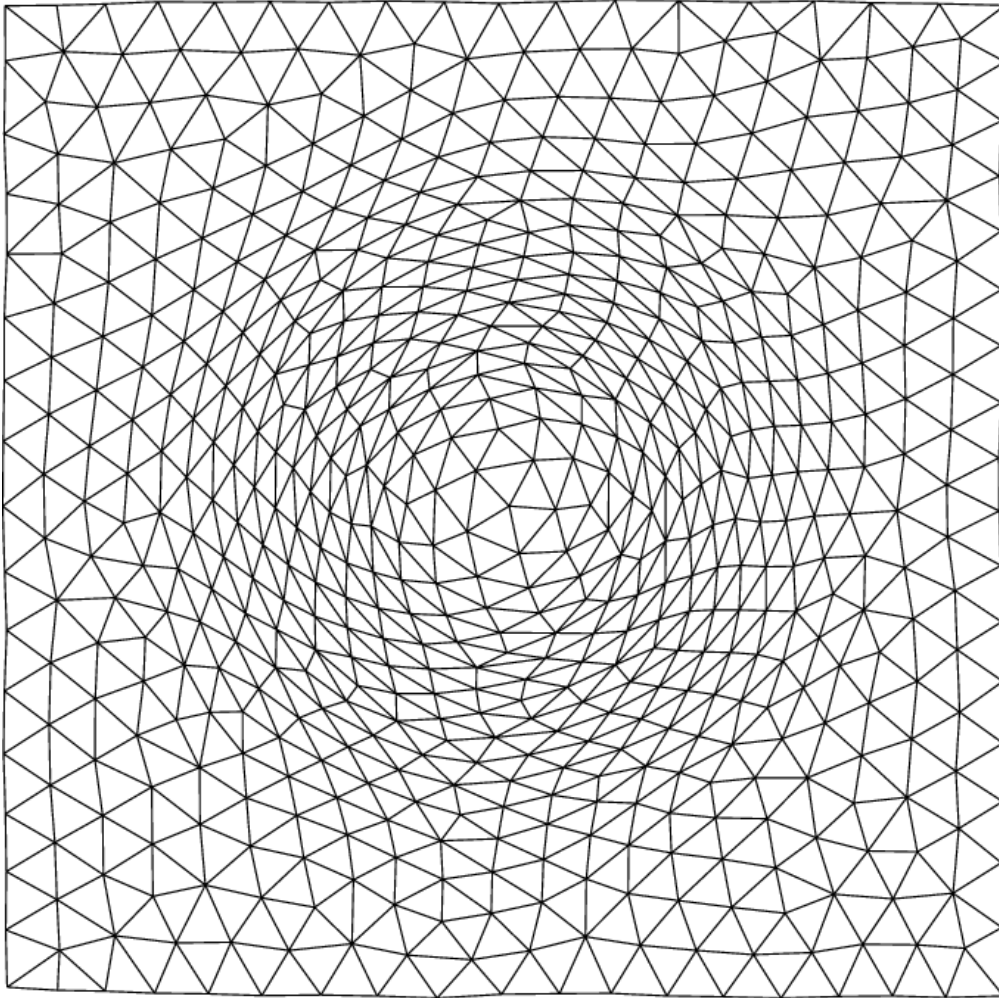
The input: anisotropy field

$$G(x,y) = \begin{bmatrix} a(x,y) & b(x,y) \\ b(x,y) & c(x,y) \end{bmatrix}$$

$$\{ \mathbf{q} \mid d_G(\mathbf{p}, \mathbf{q}) = 1 \}$$

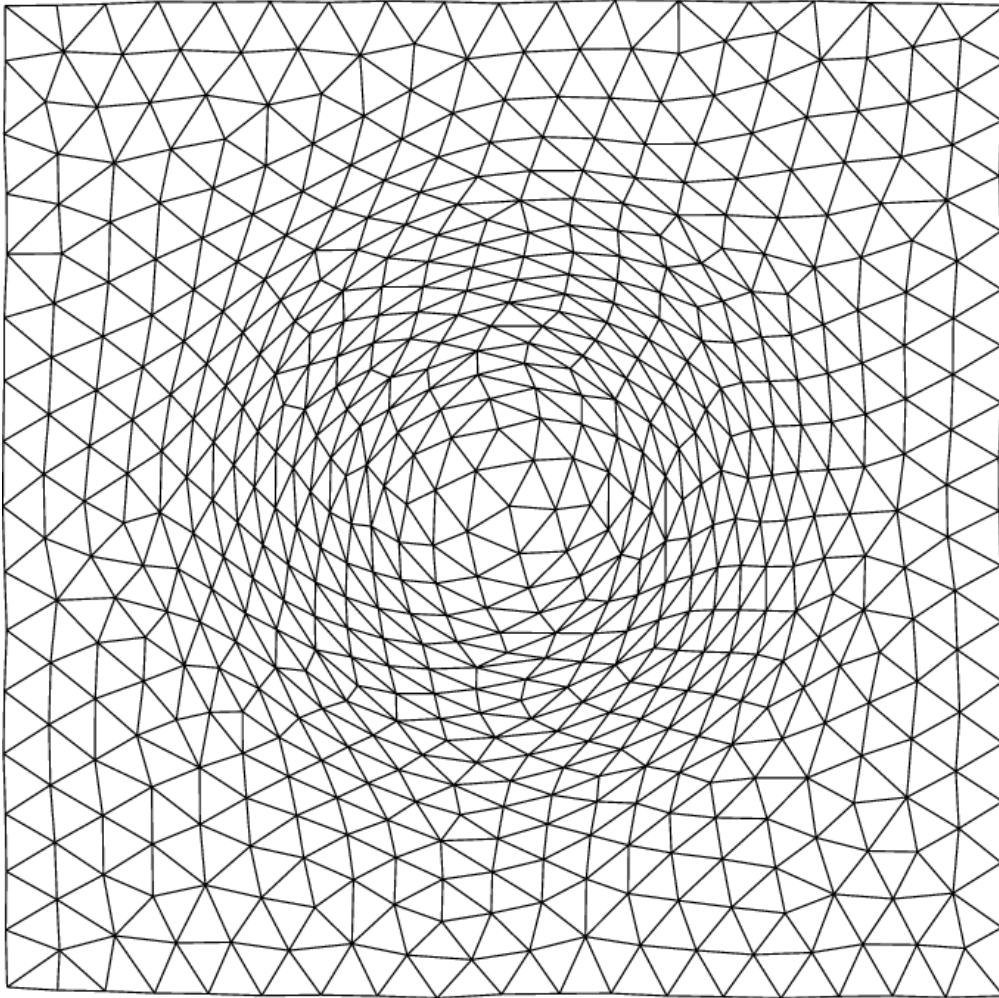


Part. 3. Anisotropy



The result: triangles are “deformed” by the anisotropy.

Part. 3. Anisotropy



The result: triangles are “deformed” by the anisotropy.

*Q: How to compute an **Anisotropic Centroidal Voronoi Tessellation** ?*

4

Journey in the 6th dimension

... and beyond

Part. 4 Journey in the 6th dimension

The key idea

This example:

Anisotropic mesh in 2d  Isotropic mesh in 3d

Part. 4 Journey in the 6th dimension

The key idea

This example:

Anisotropic mesh in 2d  Isotropic mesh in 3d

Replace **anisotropy** with **additional dimensions**

Part. 4 Journey in the 6th dimension

The key idea

Replace **anisotropy** with **additional dimensions**

Note: more dimensions may be needed

Part. 4 Journey in the 6th dimension

The key idea

Replace **anisotropy** with **additional dimensions**

Note: more dimensions may be needed

How many ?

Part. 4 Journey in the 6th dimension

The key idea

Replace **anisotropy** with **additional dimensions**

Note: more dimensions may be needed

How many ?

John Nash's isometric embedding theorem:

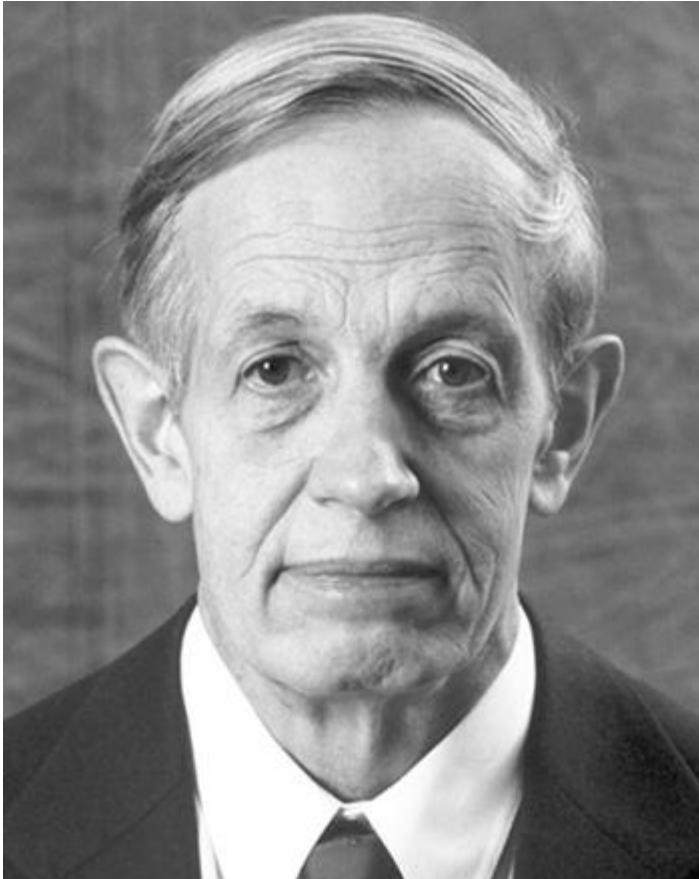
Maximum: depending on desired smoothness

$C^1 : 2n$ [Nash-Kuiper]

$C^k : \text{bounded by } n(3n+11)/2$ [Nash, Nash-Moser]

Part. 4 Journey in the 6th dimension

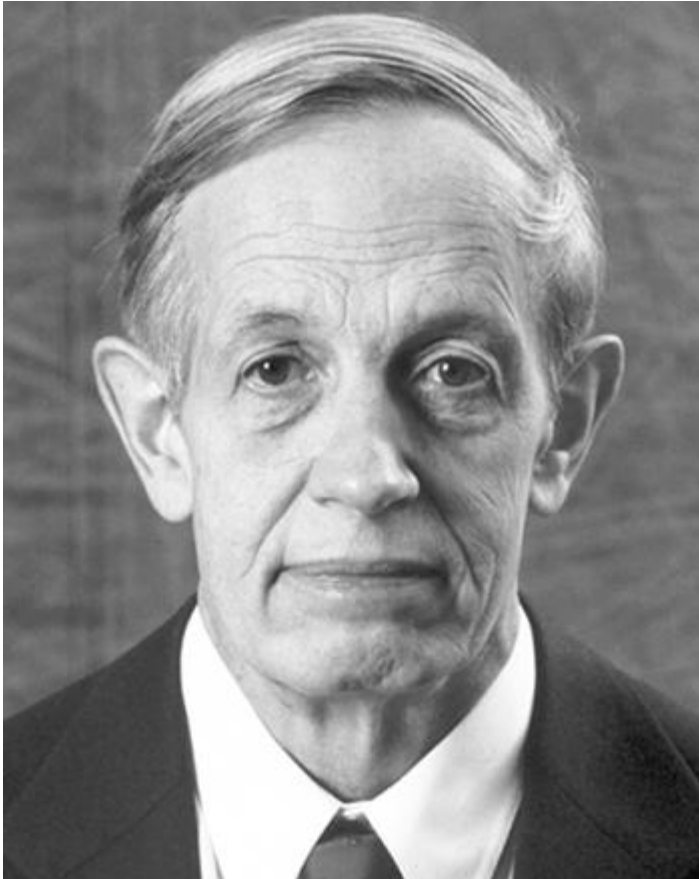
Two words about John Nash



- Isometric embedding theorem
- Nash Equilibrium → Nobel prize of economy

Part. 4 Journey in the 6th dimension

Two words about John Nash



- Isometric embedding theorem
- Nash Equilibrium → Nobel prize of economy

The **existence** is proved, but it does not tell me **how to compute** the embedding given a specified surface and anisotropy field.

Part. 4 Journey in the 6th dimension

Convex integration – Flat Torus



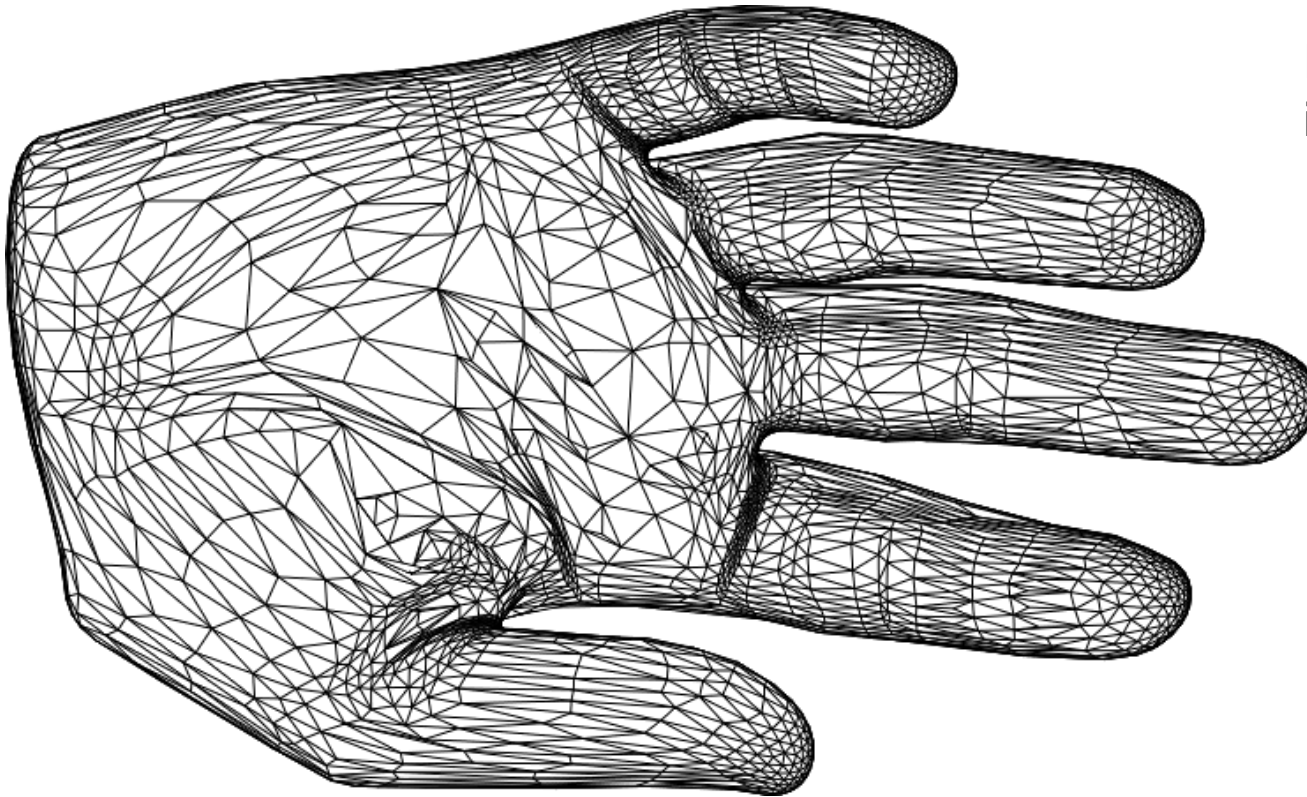
[Borelli, Jabrane, Lazarus, Thibert]

Part. 4 Journey in the 6th dimension

A 6d embedding for curvature-adapted meshing

What we want:

Equilateral triangles
in spherical zones

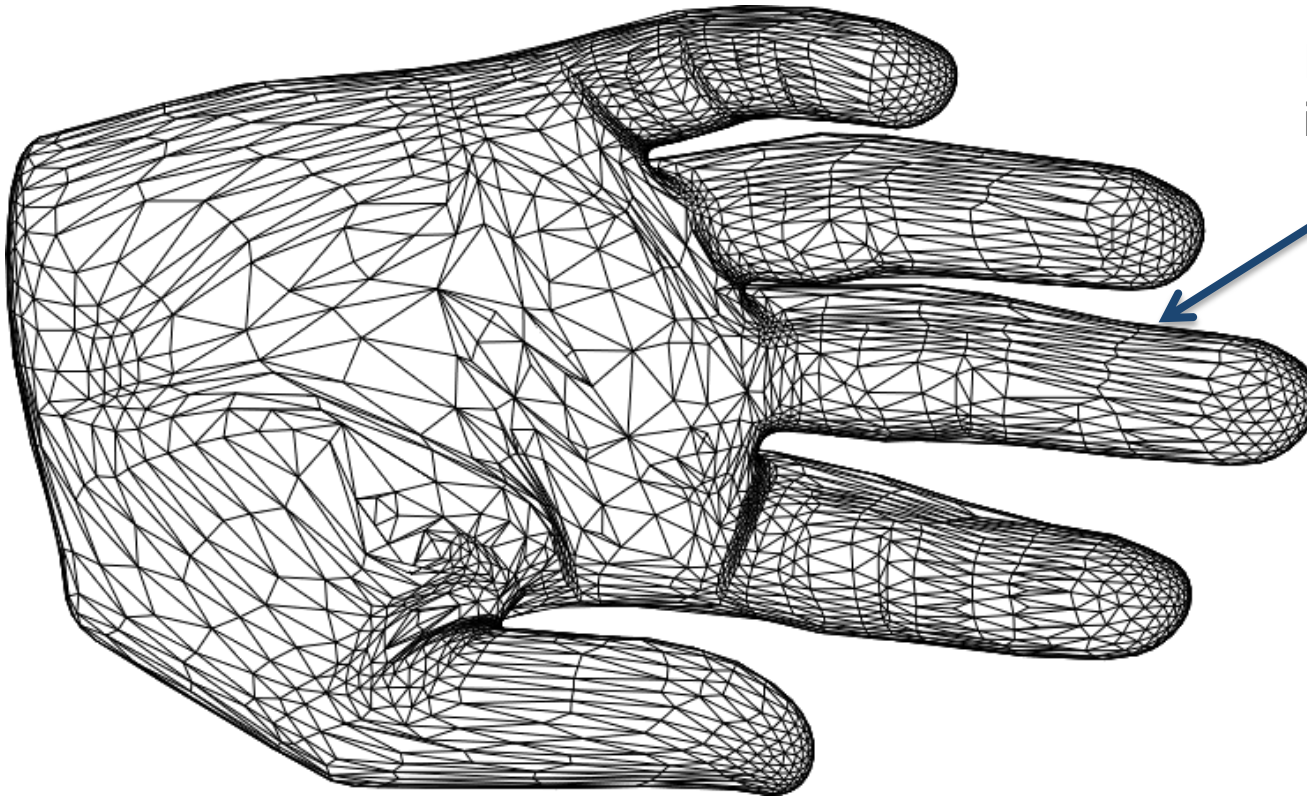


Part. 4 Journey in the 6th dimension

A 6d embedding for curvature-adapted meshing

What we want:

Elongated triangles
in cylindrical zones



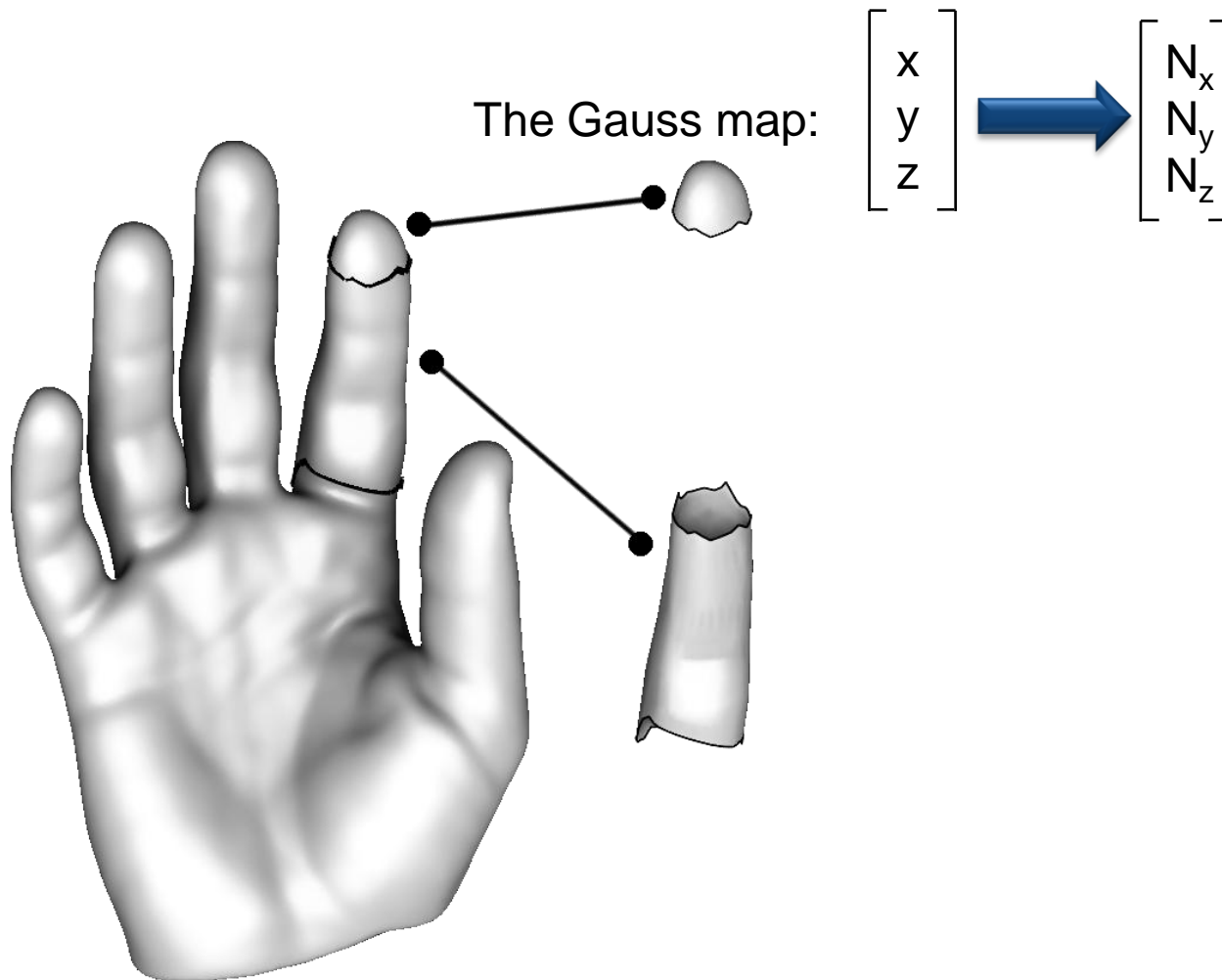
Part. 4 Journey in the 6th dimension

A 6d embedding for curvature-adapted meshing

The Gauss map: $\begin{bmatrix} x \\ y \\ z \end{bmatrix} \rightarrow \begin{bmatrix} N_x \\ N_y \\ N_z \end{bmatrix}$

Part. 4 Journey in the 6th dimension

A 6d embedding for curvature-adapted meshing

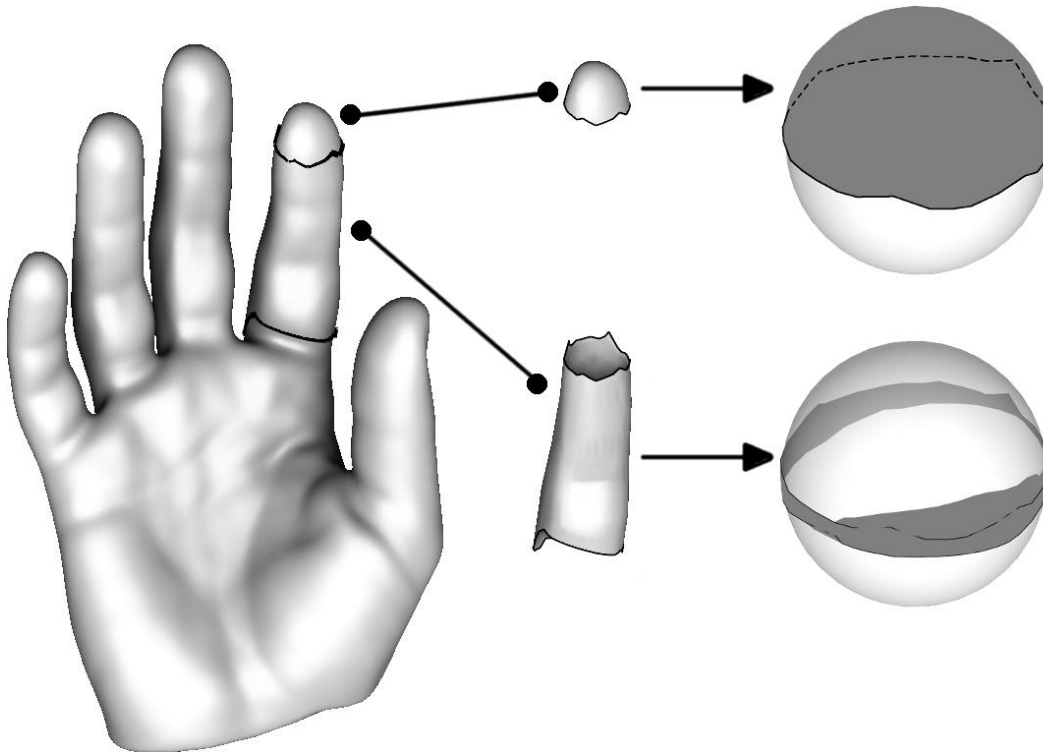


Part. 4 Journey in the 6th dimension

A 6d embedding for curvature-adapted meshing

The Gauss map:

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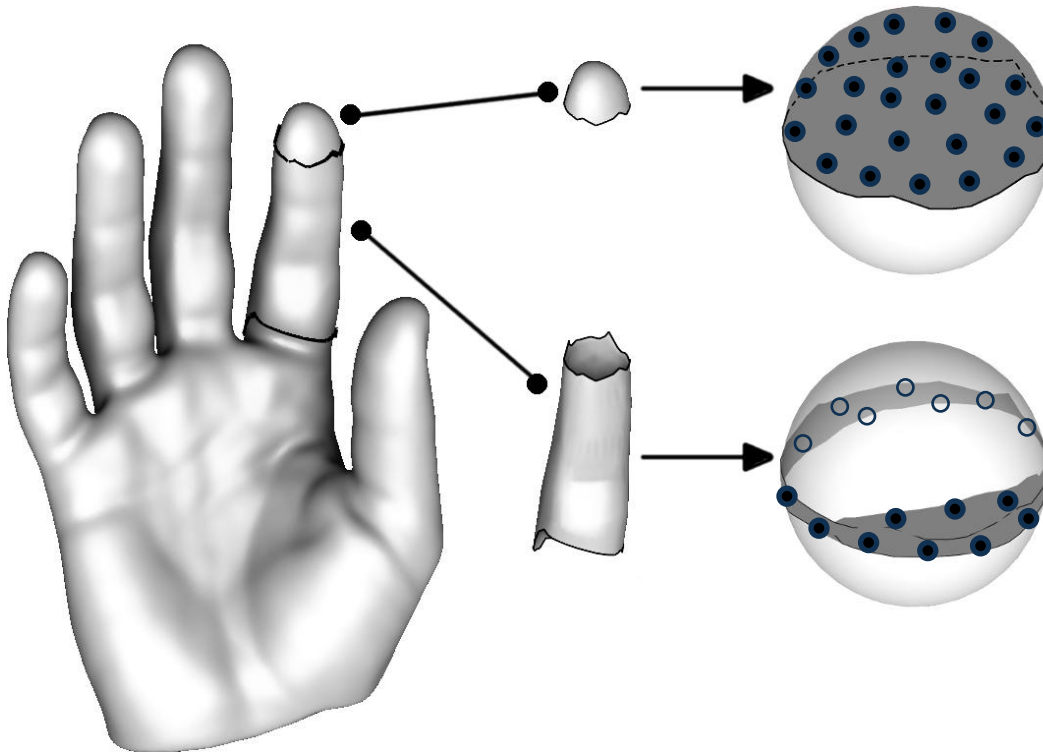


Part. 4 Journey in the 6th dimension

A 6d embedding for curvature-adapted meshing

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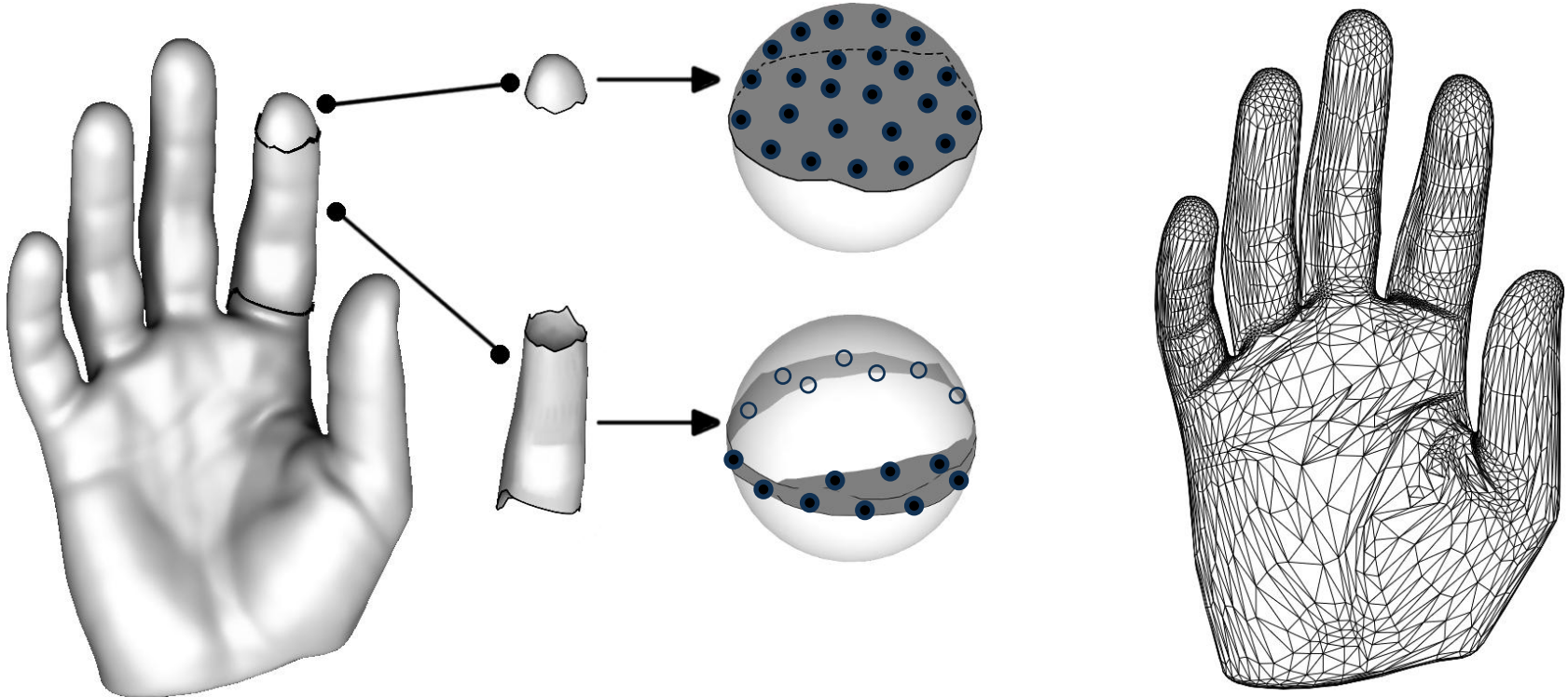


Part. 4 Journey in the 6th dimension

A 6d embedding for curvature-adapted meshing

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Part. 4 Journey in the 6th dimension

A 6d embedding for curvature-adapted meshing

The Gauss map: $\begin{bmatrix} x \\ y \\ z \end{bmatrix} \rightarrow \begin{bmatrix} N_x \\ N_y \\ N_z \end{bmatrix}$

The Gauss-map is **non-bijective** in general (*bijective only if convex object*)

Part. 4 Journey in the 6th dimension

A 6d embedding for curvature-adapted meshing

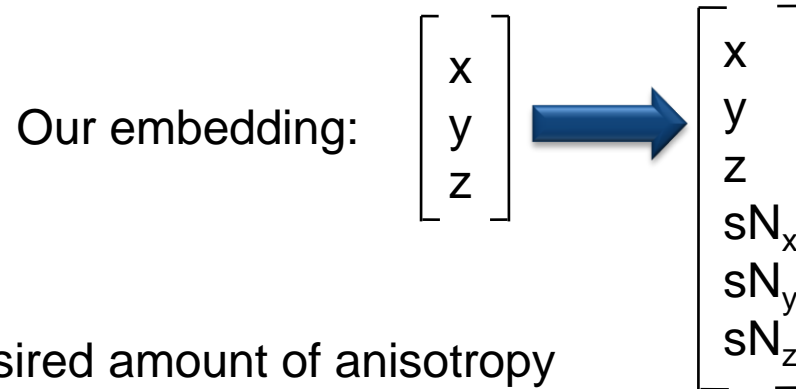
The Gauss map:
$$\begin{bmatrix} x \\ y \\ z \end{bmatrix} \longrightarrow \begin{bmatrix} N_x \\ N_y \\ N_z \end{bmatrix}$$

The Gauss-map is **non-bijective** in general (*bijective only if convex object*)

Our embedding:
$$\begin{bmatrix} x \\ y \\ z \end{bmatrix} \longrightarrow \begin{bmatrix} x \\ y \\ z \\ sN_x \\ sN_y \\ sN_z \end{bmatrix}$$

Part. 4 Journey in the 6th dimension

A 6d embedding for curvature-adapted meshing



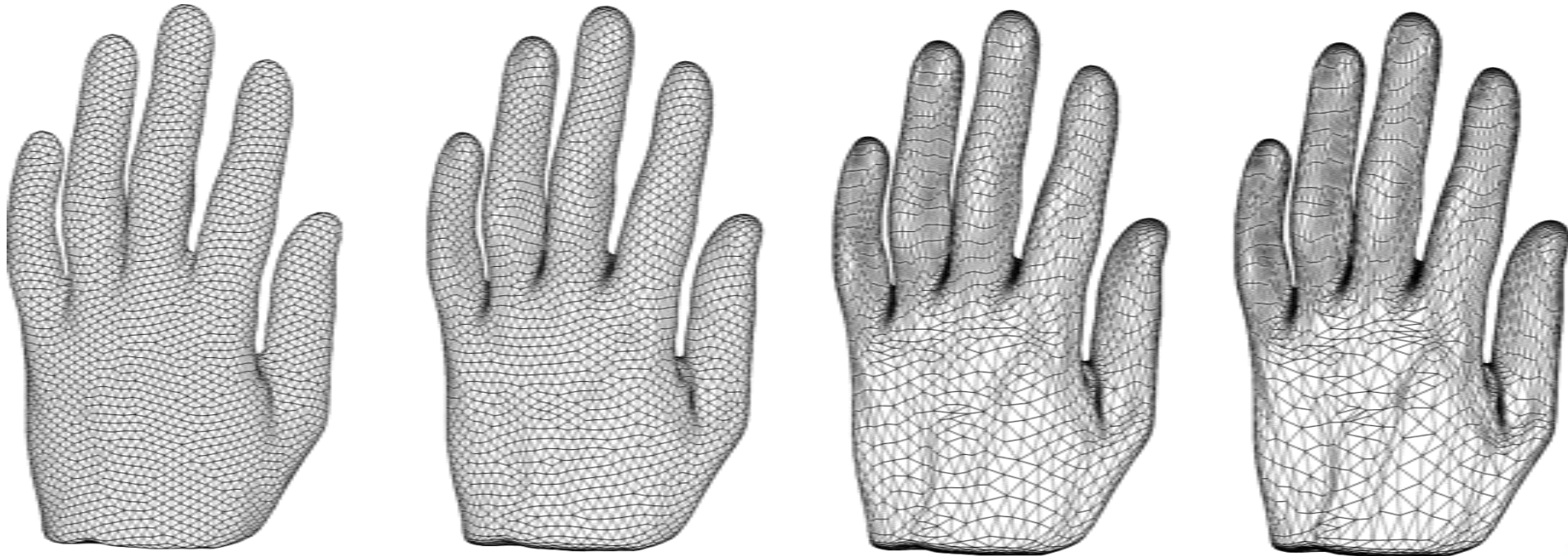
Part. 4 Journey in the 6th dimension

A 6d embedding for curvature-adapted meshing

s : desired amount of anisotropy

Small s

Large s



5

The algorithm

Part. 5 The algorithm

Lloyd relaxation in \mathbb{R}^6 (Naïve version)

(1) Embed the surface \mathbf{S} into \mathbb{R}^6

Part. 5 The algorithm

Lloyd relaxation in \mathbb{R}^6 (Naïve version)

- (1) Embed the surface \mathbf{S} into \mathbb{R}^6
- (2) Compute initial point distrib. \mathbf{X}

Part. 5 The algorithm

Lloyd relaxation in \mathbb{R}^6 (Naïve version)

- (1) Embed the surface **S** into \mathbb{R}^6
 - (2) Compute initial point distrib. **X**
- While convergence is not reached

Part. 5 The algorithm

Lloyd relaxation in \mathbb{R}^6 (Naïve version)

- (1) Embed the surface \mathbf{S} into \mathbb{R}^6
 - (2) Compute initial point distrib. \mathbf{X}
- While convergence is not reached
- (3) Compute Vor(\mathbf{X})

Part. 5 The algorithm

Lloyd relaxation in \mathbb{R}^6 (Naïve version)

- (1) Embed the surface \mathbf{S} into \mathbb{R}^6
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- (3) Compute $\text{Vor}(\mathbf{X})$
 - (4) Compute $\text{Vor}(\mathbf{X}) \cap \mathbf{S}$

Part. 5 The algorithm

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 - (4) Compute $\text{Vor}(\mathbf{X}) \cap \mathbf{S}$
 - (5) Move each \mathbf{x}_i to the centroid of $\text{Vor}(\mathbf{x}_i) \cap \mathbf{S}$

Part. 5 The algorithm

Lloyd relaxation in \mathbb{R}^6 (Naïve version)

(1) Embed the surface \mathbf{S} into \mathbb{R}^6

(2) Compute initial point distrib. \mathbf{X}

While convergence is not reached

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Costs $d!$ for dimension d

(4) Compute $\text{Vor}(\mathbf{X}) \cap \mathbf{S}$

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Part. 5 The algorithm

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While convergence is not reached

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Costs $d!$ for dimension d
 $d = 6$; $d! = 720$

(4) Compute $\text{Vor}(\mathbf{X}) \cap \mathbf{S}$

(5) Move each \mathbf{x}_i to the centroid of $\text{Vor}(\mathbf{x}_i) \cap \mathbf{S}$

Part. 5 The algorithm

Lloyd relaxation in \mathbb{R}^6 (Naïve version)

(1) Embed the surface \mathbf{S} into \mathbb{R}^6

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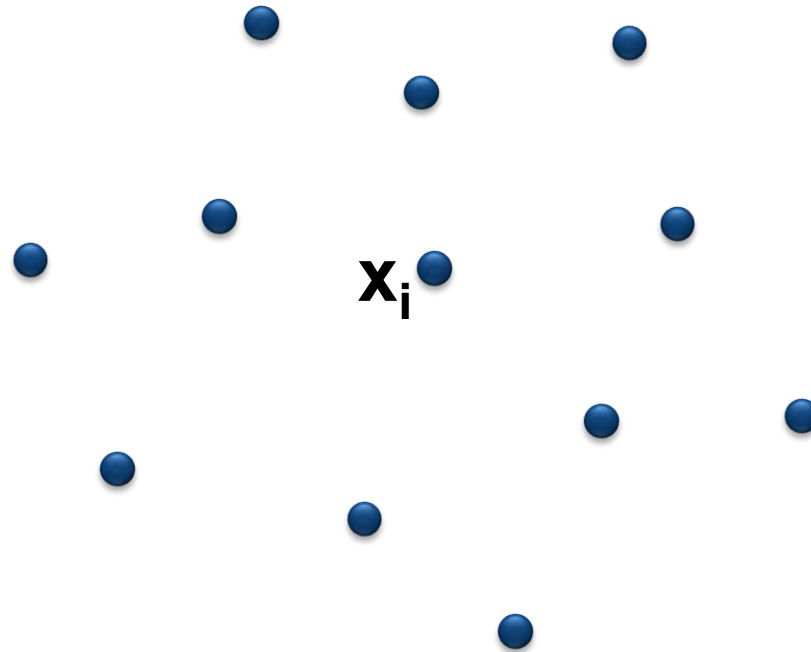
Curse of dimensionality

Some theoretical results

existence of bounds – Tangent Delaunay Complex [Boissonnat et.al.]

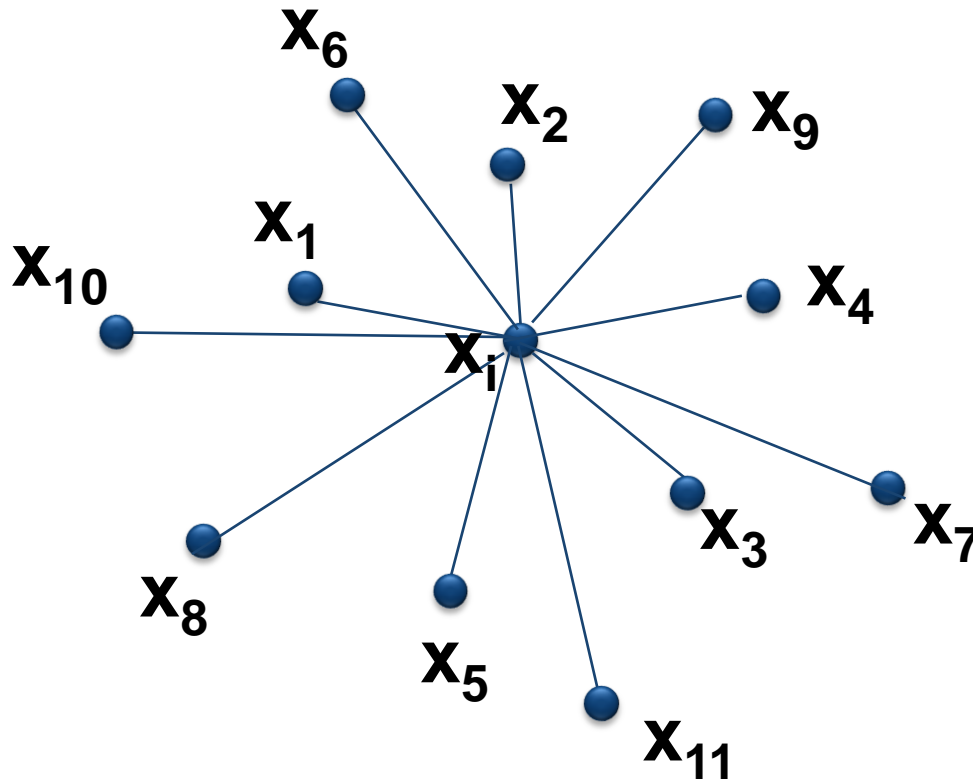
Part. 5 The algorithm

Voronoi cells as iterative convex clipping



Part. 5 The algorithm

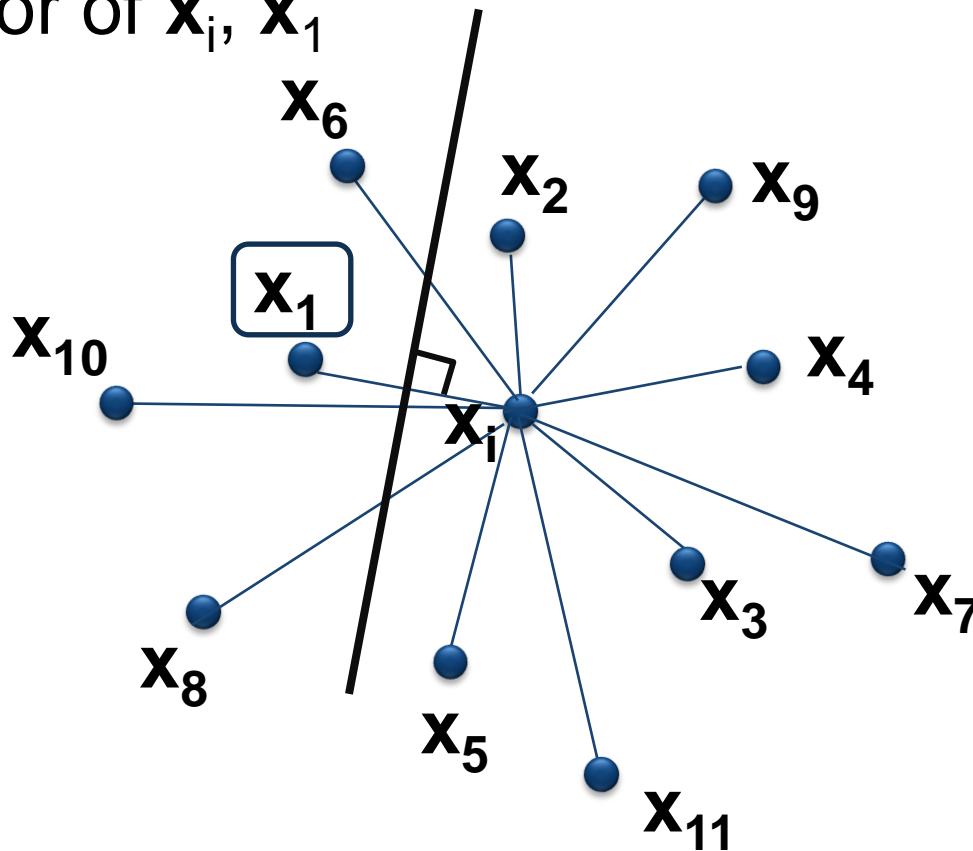
Voronoi cells as iterative convex clipping
Neighbors in increasing (6d) distance from x_i



Part. 5 The algorithm

Voronoi cells as iterative convex clipping

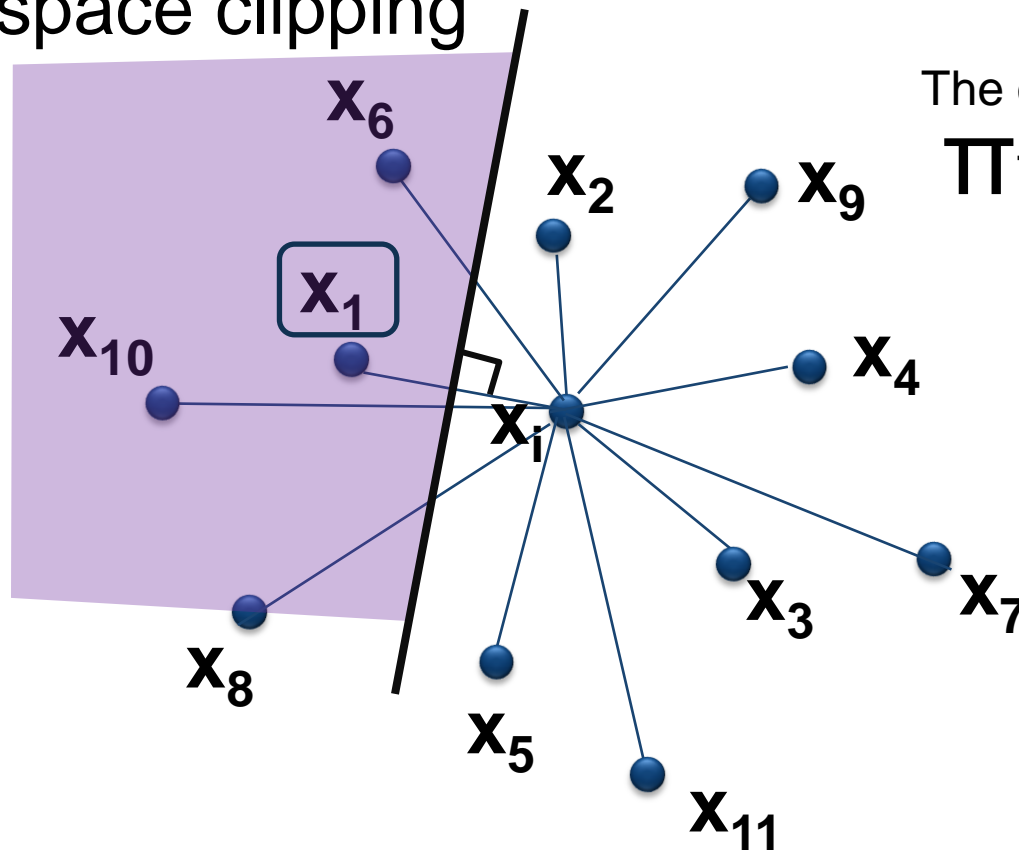
Bisector of x_i, x_1



Part. 5 The algorithm

Voronoi cells as iterative convex clipping
Half-space clipping

This side:
 $\Pi^-(i,1)$

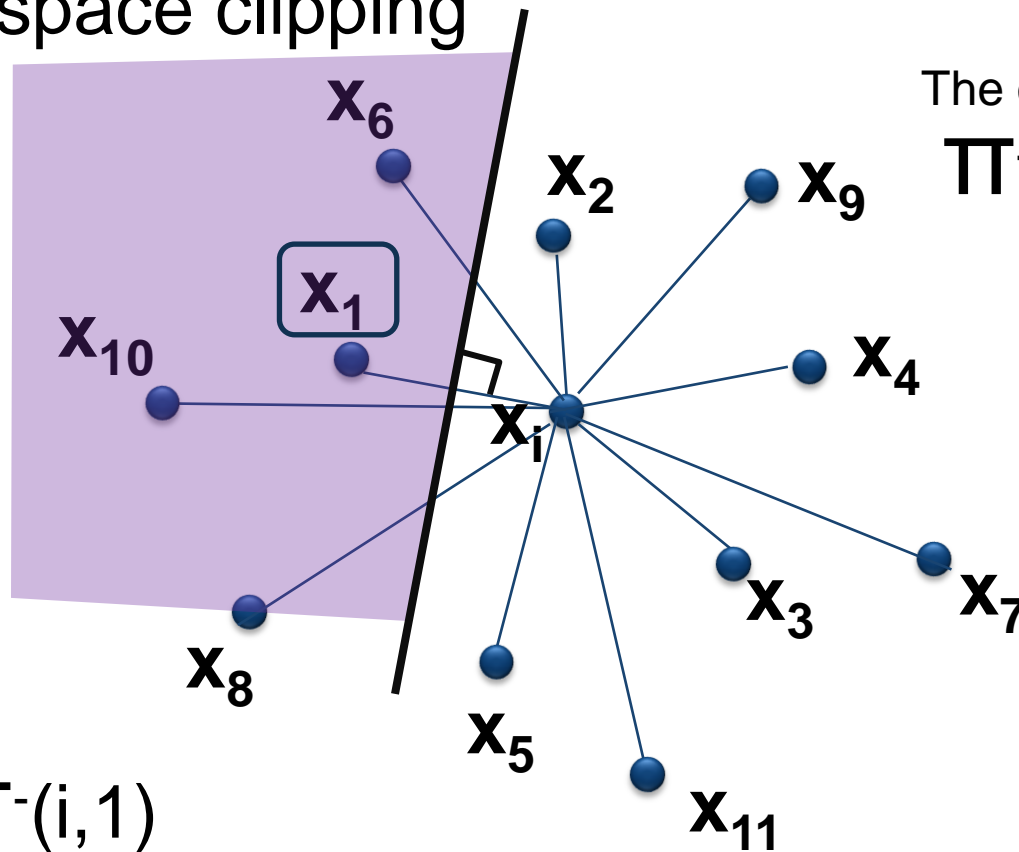


The other side:
 $\Pi^+(i,1)$

Part. 5 The algorithm

Voronoi cells as iterative convex clipping
Half-space clipping

This side:
 $\Pi^-(i,1)$

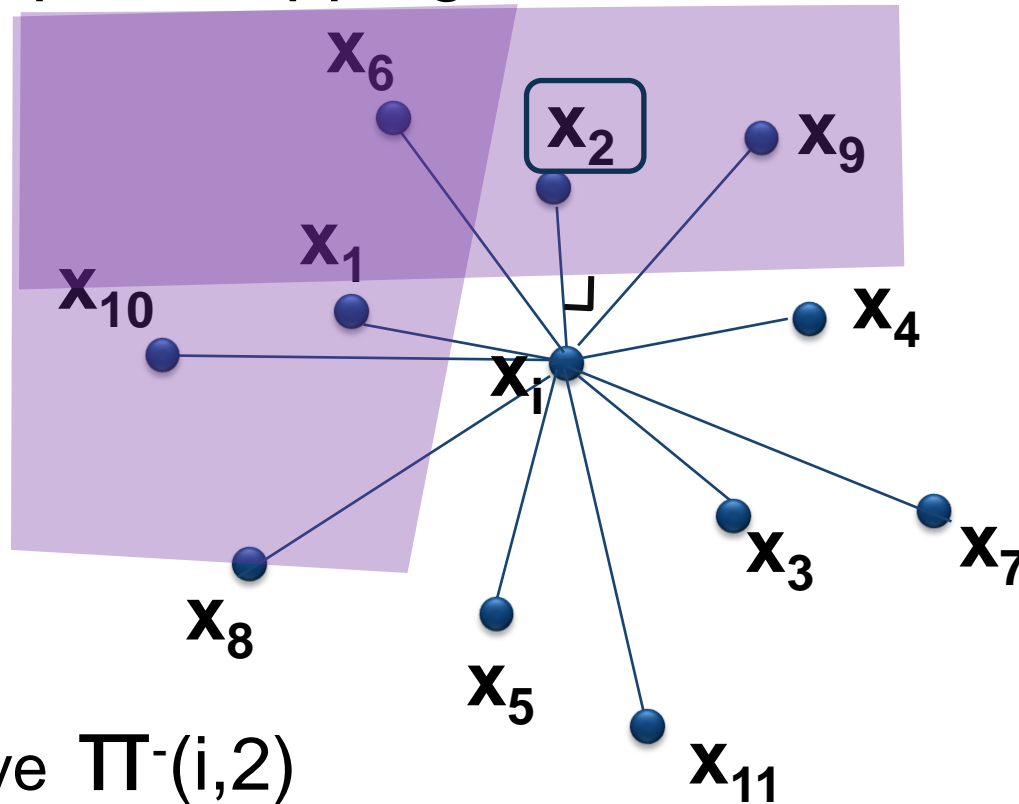


The other side:
 $\Pi^+(i,1)$

Remove $\Pi^-(i,1)$

Part. 5 The algorithm

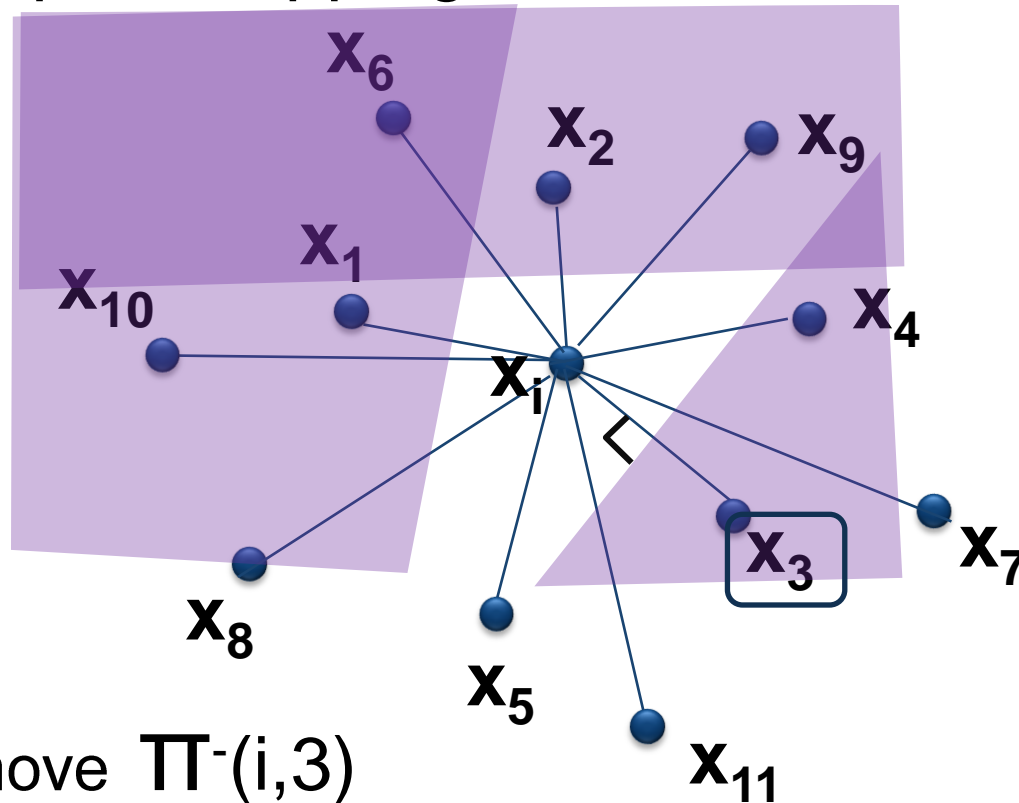
Voronoi cells as iterative convex clipping
Half-space clipping



Then remove $\Pi^-(i,2)$

Part. 5 The algorithm

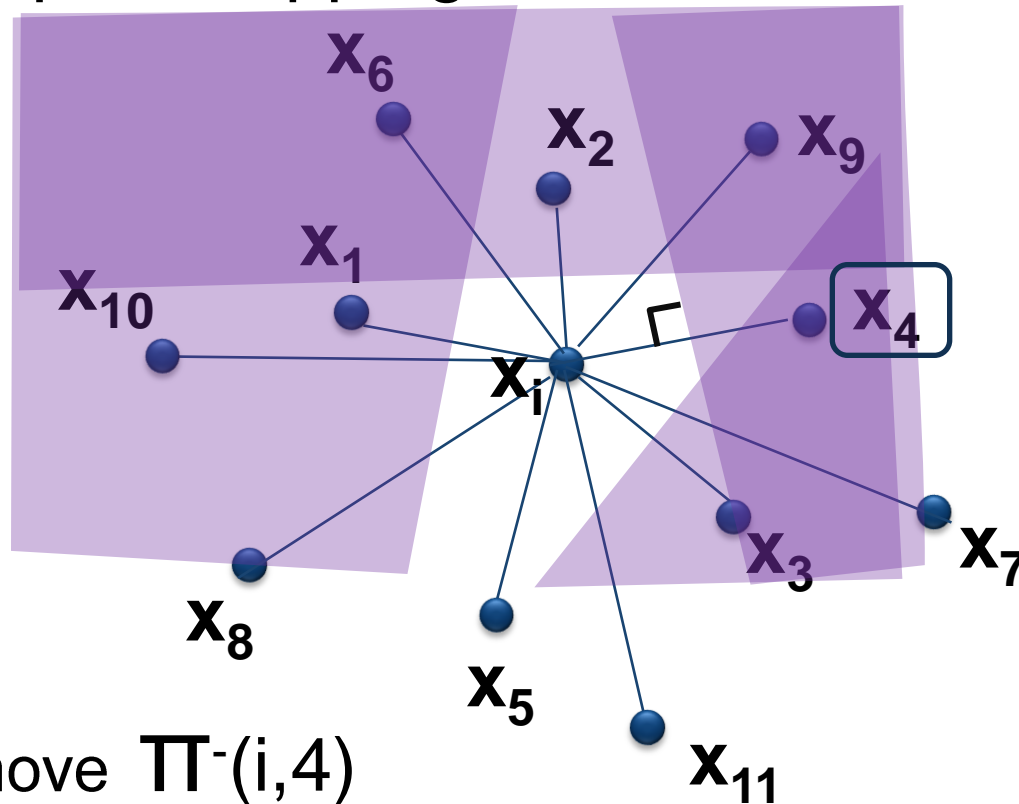
Voronoi cells as iterative convex clipping
Half-space clipping



... then remove $\Pi^-(i,3)$

Part. 5 The algorithm

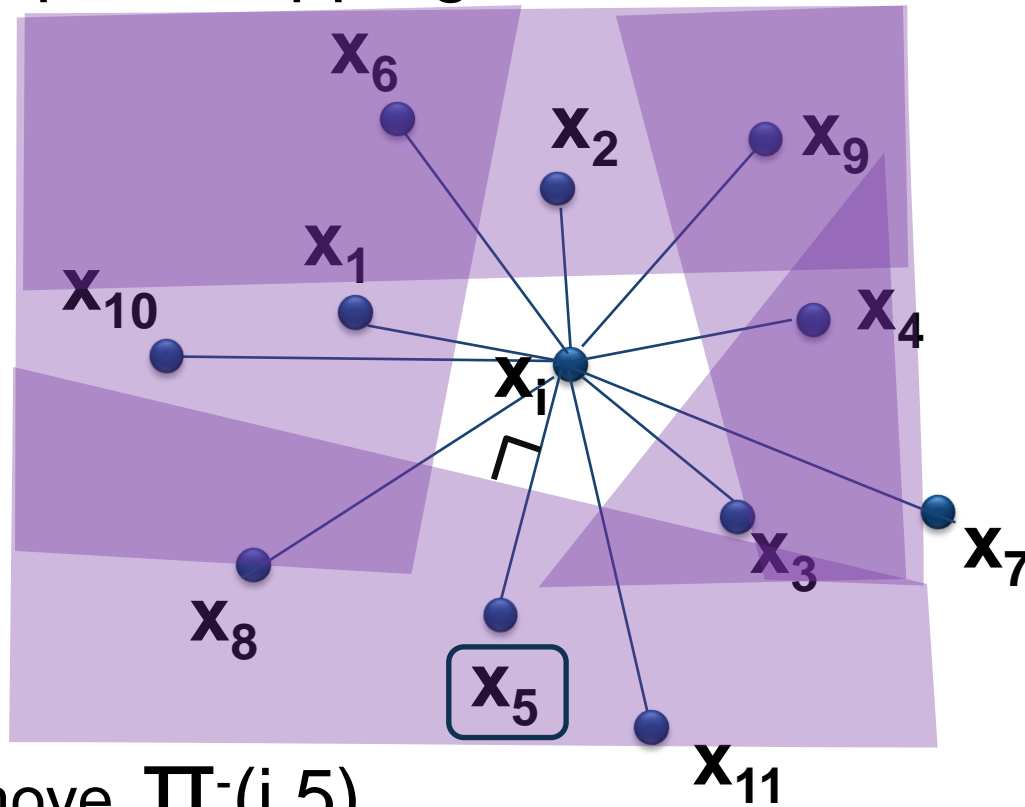
Voronoi cells as iterative convex clipping
Half-space clipping



... then remove $\Pi^-(i,4)$

Part. 5 The algorithm

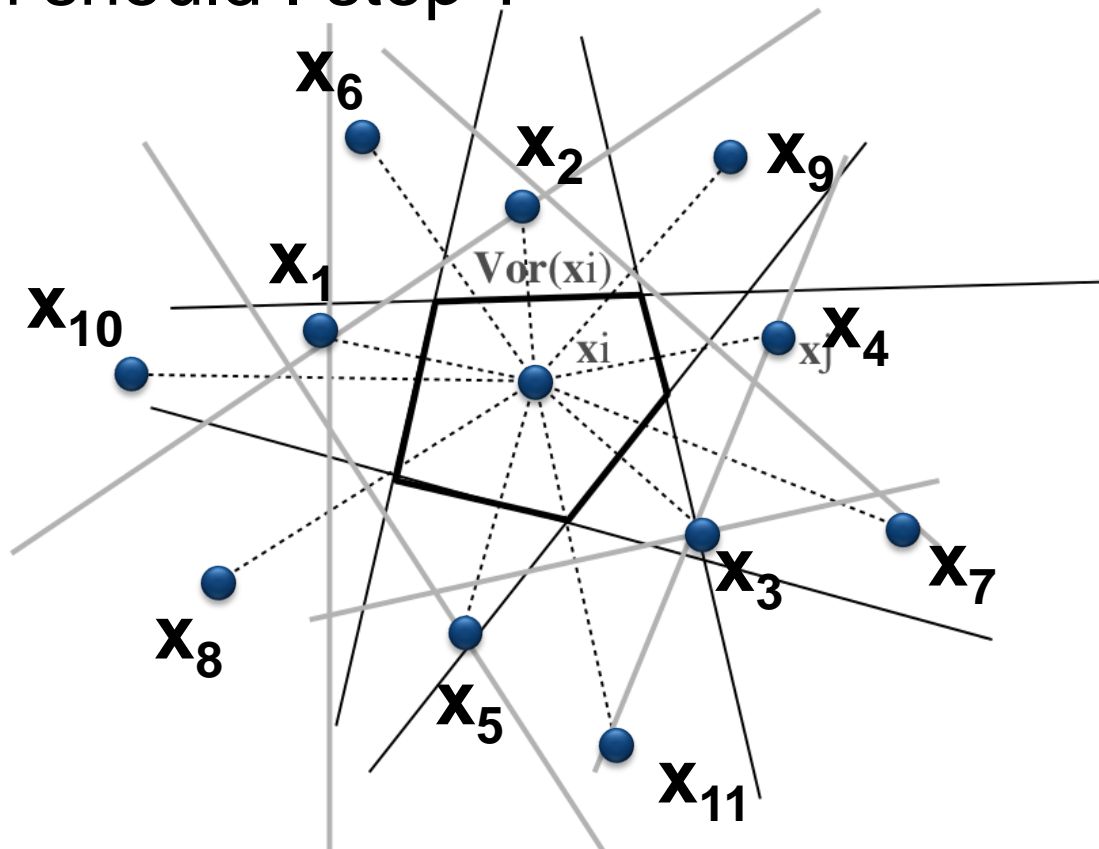
Voronoi cells as iterative convex clipping
Half-space clipping



... then remove $\Pi^-(i,5)$

Part. 5 The algorithm

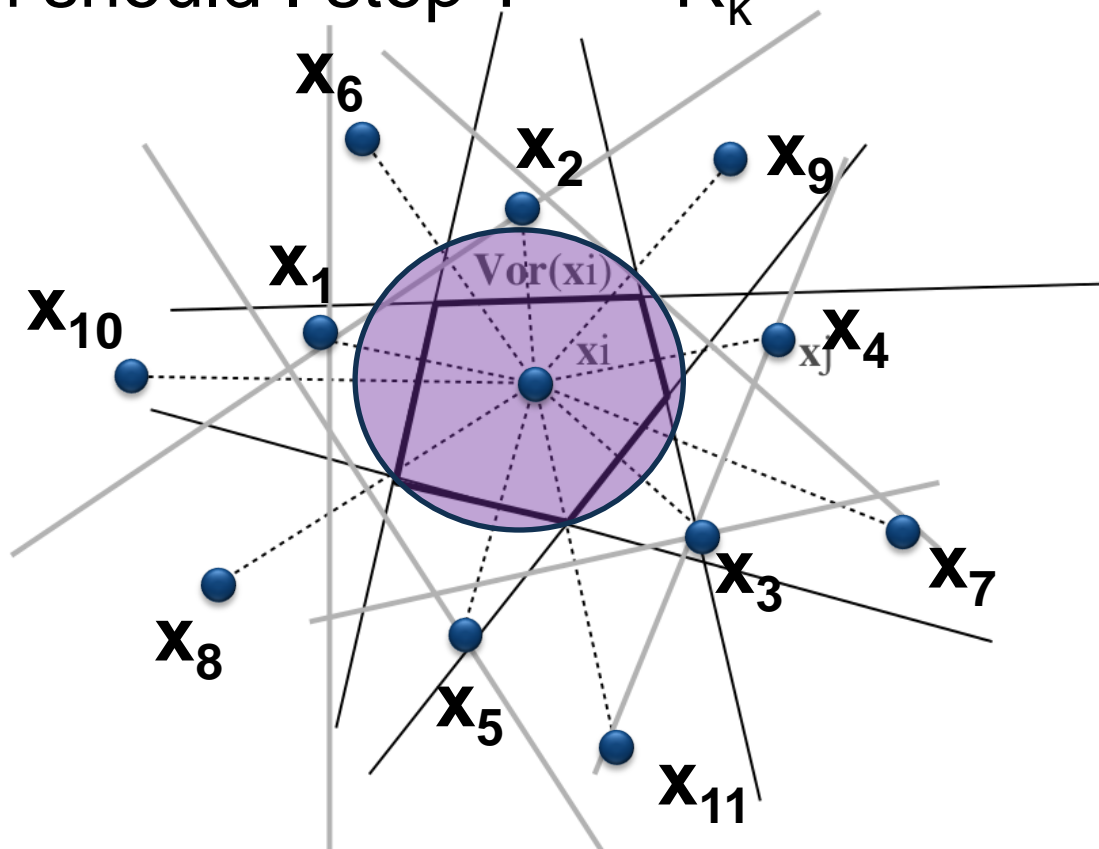
Voronoi cells as iterative convex clipping
When should I stop ?



Part. 5 The algorithm

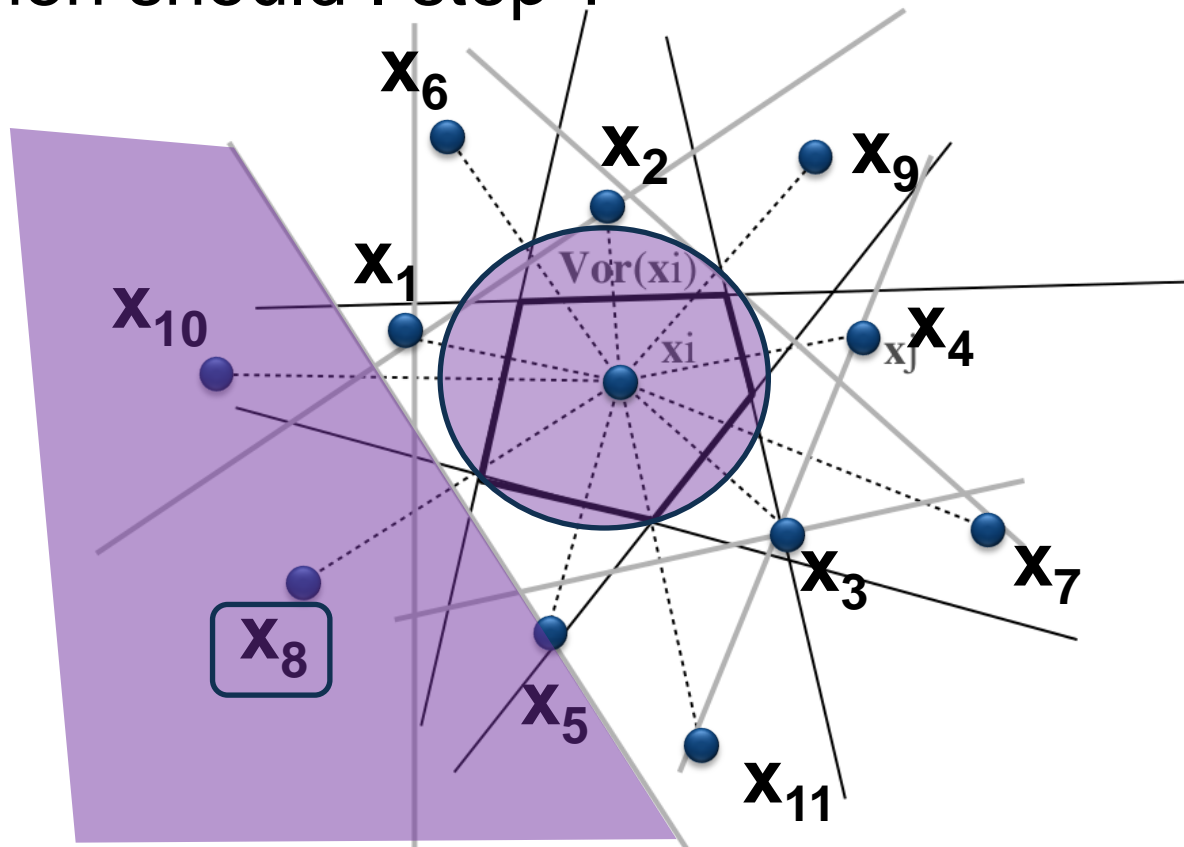
Voronoi cells as iterative convex clipping

When should I stop ? R_k



Part. 5 The algorithm

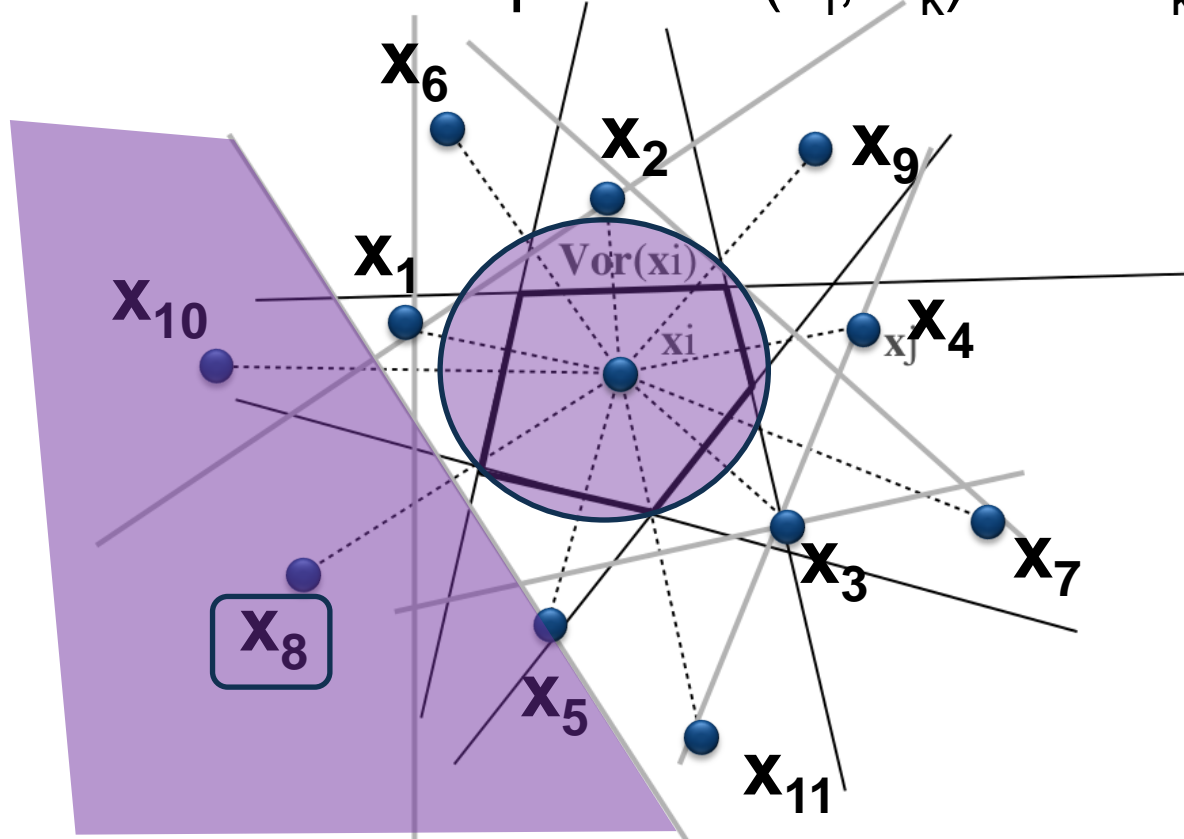
Voronoi cells as iterative convex clipping
When should I stop ?



Part. 5 The algorithm

Voronoi cells as iterative convex clipping

When should I stop ? $d(\mathbf{x}_i, \mathbf{x}_k) > 2 R_k$



Part. 5 The algorithm

Voronoi cells as iterative convex clipping

Theorem: $d(\mathbf{x}_i, \mathbf{x}_{k+1}) > 2R_k \rightarrow \bigcap \Pi^+(i,k) = \text{Vor}(\mathbf{x}_i)$

[L and Bonneel, IMR 2012]

Part. 5 The algorithm

Voronoi cells as iterative convex clipping

When should I stop ? $d(\mathbf{x}_i, \mathbf{x}_k) > 2 R_k$

“Radius of security” is reached

Part. 5 The algorithm

Voronoi cells as iterative convex clipping

When should I stop ? $d(\mathbf{x}_i, \mathbf{x}_k) > 2 R_k$

“Radius of security” is reached

Note: R_k decreases and $d(\mathbf{x}_i, \mathbf{x}_k)$ increases

Part. 5 The algorithm

Voronoi cells as iterative convex clipping

When should I stop ? $d(\mathbf{x}_i, \mathbf{x}_k) > 2 R_k$

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Note: R_k decreases and $d(\mathbf{x}_i, \mathbf{x}_k)$ increases

Advantages:

Part. 5 The algorithm

Voronoi cells as iterative convex clipping

When should I stop ? $d(\mathbf{x}_i, \mathbf{x}_k) > 2 R_k$

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Note: R_k decreases and $d(\mathbf{x}_i, \mathbf{x}_k)$ increases

Advantages:

(1) Compute $\text{Vor}(\mathbf{X}) \cap S$ directly (start with f and clip)

Part. 5 The algorithm

Voronoi cells as iterative convex clipping

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Note: R_k decreases and $d(\mathbf{x}_i, \mathbf{x}_k)$ increases

Advantages:

- (1) Compute $\text{Vor}(\mathbf{X}) \cap S$ directly (start with f and clip)
- (2) Replace Delaunay with ANN ! (no $d!$ factor)

Part. 5 The algorithm

Voronoi cells as iterative convex clipping

When should I stop ? $d(\mathbf{x}_i, \mathbf{x}_k) > 2 R_k$

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Note: R_k decreases and $d(\mathbf{x}_i, \mathbf{x}_k)$ increases

Advantages:

- (1) Compute $\text{Vor}(\mathbf{X}) \cap S$ directly (start with f and clip)
- (2) Replace Delaunay with ANN ! (no $d!$ factor)
- (3) Parallelization is trivial (partition S and // in parts)

6

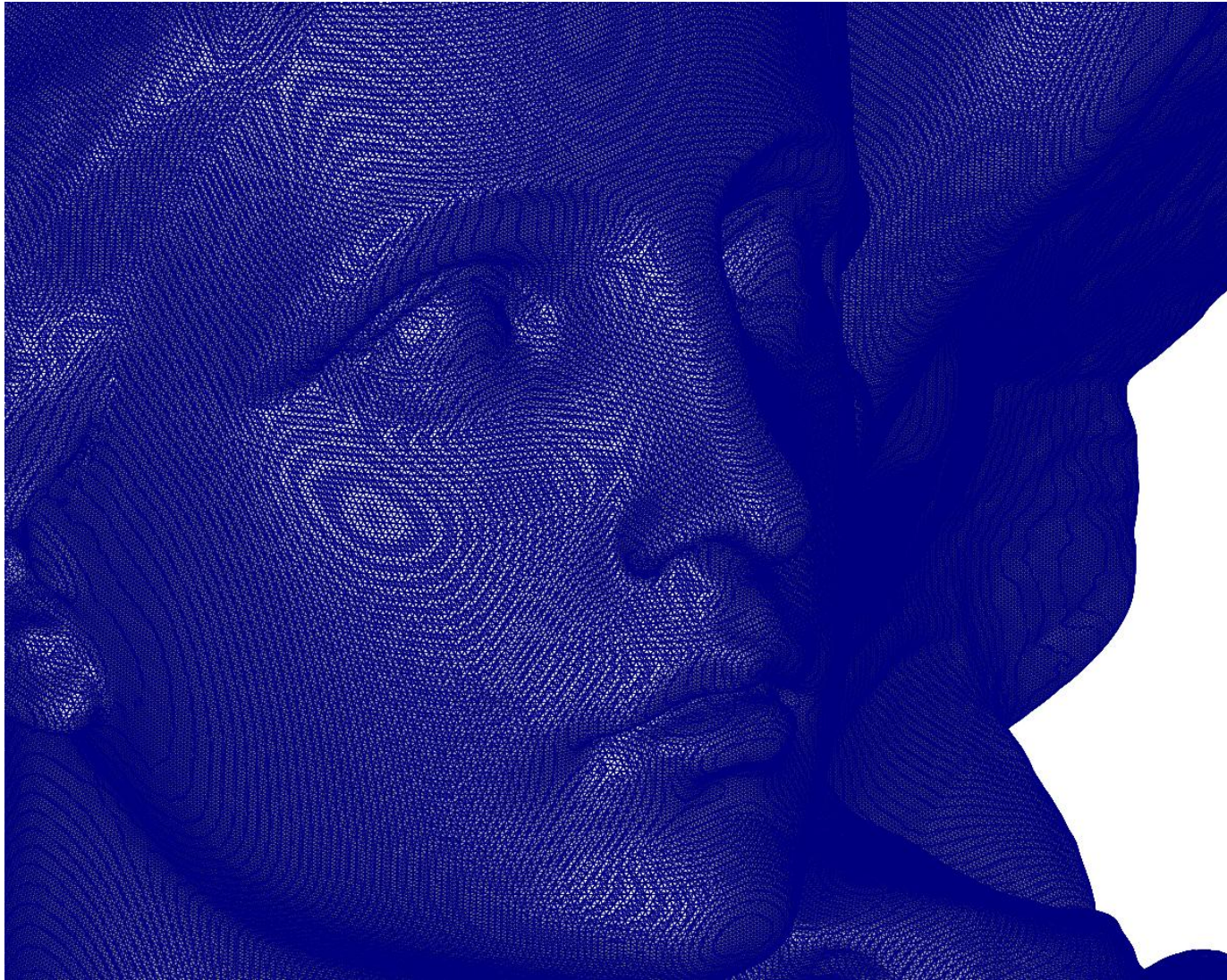
Some results

Part. 6 Some results – Filigree (Aim@Shape)

Part. 6 Some results – AntEater (Konstanz U.)

Part. 6 Some results

Lucy (Stanford): 28 million triangles



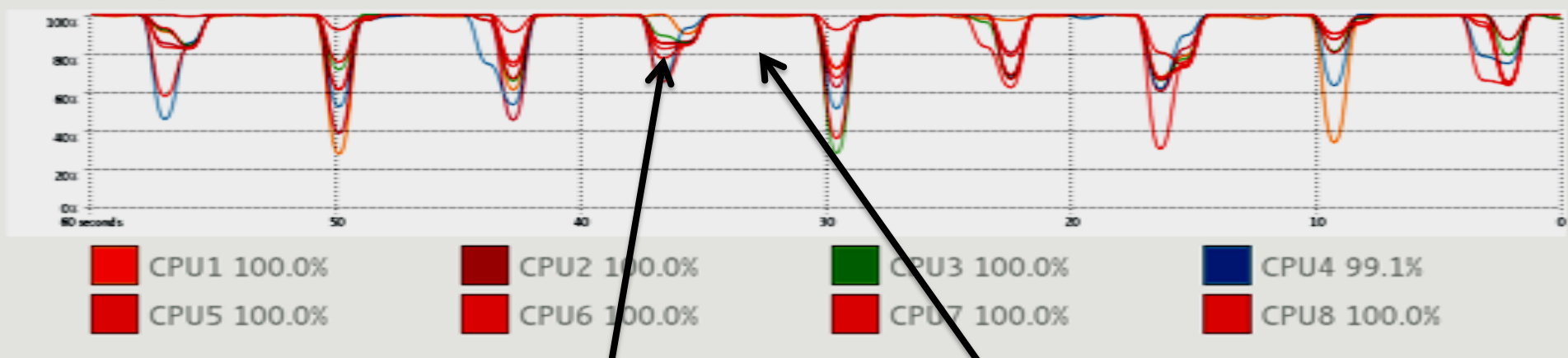
Parallel implementation, 8 threads, core i7 : 5min

Part. 6 Some results

Lucy (Stanford): 28 million triangles

One iteration

CPU History



ANN construction
(sequential)

Voronoi Parallel Linear Enumeration

Parallel implementation, 8 threads, core i7 : 5min



Part. 6 Some results

Vorpaline mesh, 100K vertices

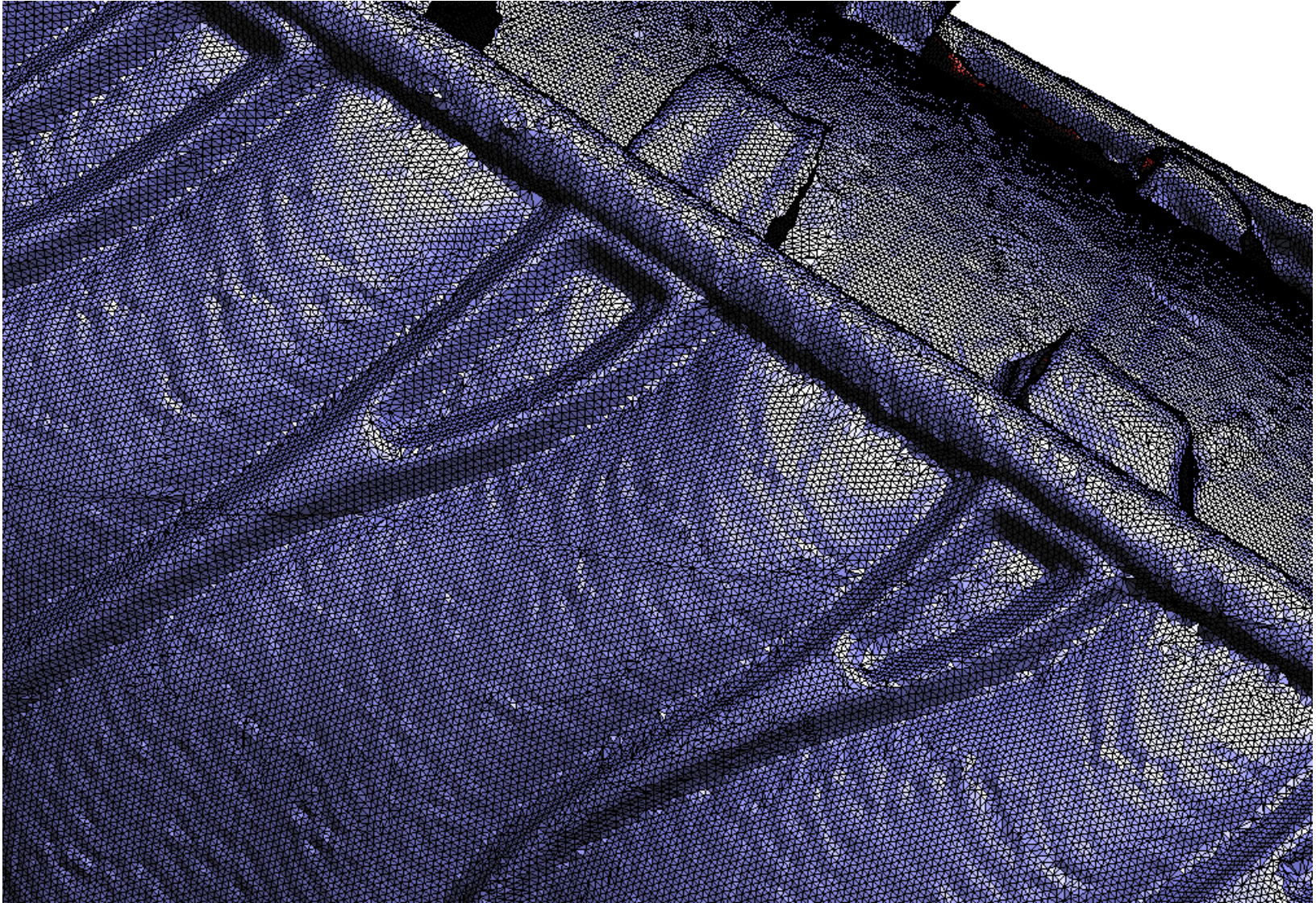
Part. 6 Some results – Porsche (Distene)

Part. 6 Some results – Plane (Distene)

Part. 6 Some results



Part. 6 Some results

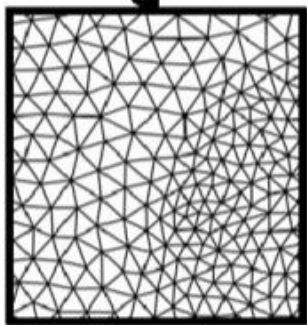
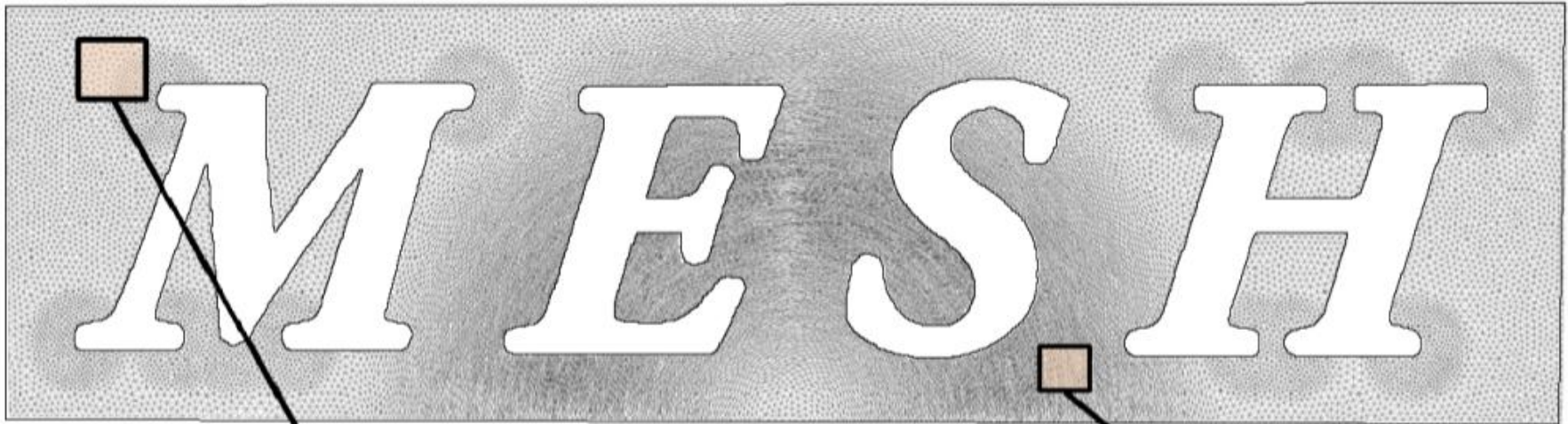


Part. 6 Some results

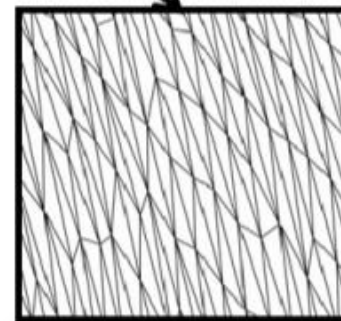
Part. 6 What about arbitrary anisotropy ?

Particle-based Anisotropic Surface Meshing

[Zhong, Guo, Wang, L, Sun, Liu and Mao, ACM SIGGRAPH 2013]



Non-Uniform Isotropic



Anisotropic

Part. 6 Reconstruction – Co3Ne (Concurrent Co Cone)

Thank you for your attention

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Wenping Wang (Hong-Kong U.)

Nicolas Bonneel (Harvard U.)

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ANR MORPHO, ANR BECASIM