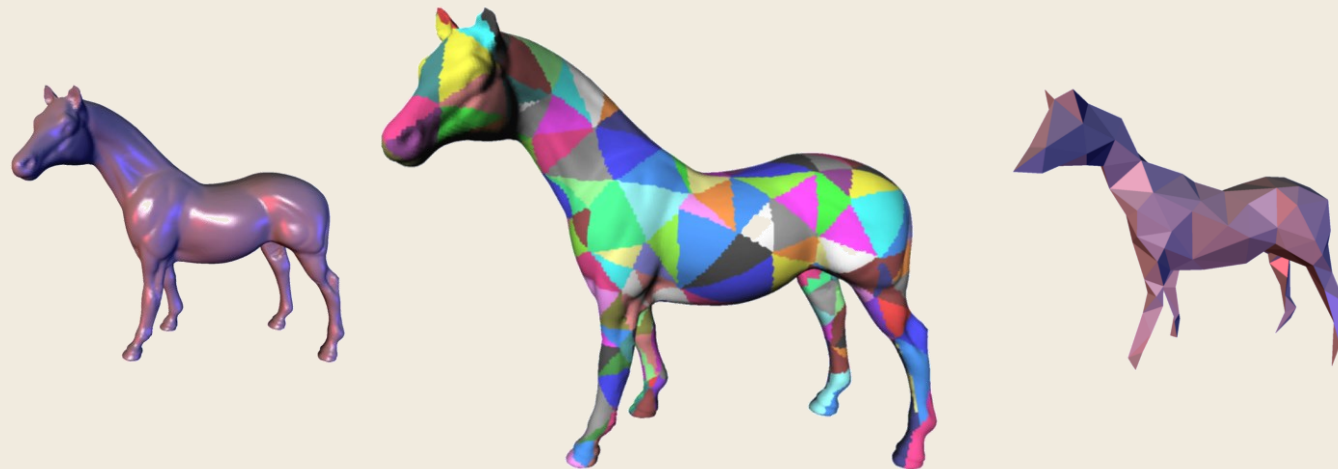


# Multiresolution Adaptive Parameterization of Surfaces

*Article by Aaron W. F. Lee, Princeton University - Wim Sweldens, Bell Laboratories - Peter Schröder, Caltech - Lawrence Cowsar, Bell Laboratories - David Dobkin, Princeton University*

Clémentine Grethen – Alexis Gosselin – Léo Meissner – Héloïse Lafargue



# I

## **Vertex Removal**

Removing a  
maximally independent  
set of vertices with low  
outdegree

# II

## **Priority queue**

Vertices with small and  
flat 1-ring  
neighborhoods

# III

## **Retriangulation**

Flattening :  
Conformal map &  
Delaunay triangulation

# IV

## **Decompression**

Decompression method  
implemented



# V

## Results

Images, videos,  
quantification

# VI

## Issues

Encountered, solved and  
remaining problems

# I - Vertex removal

## # Initialization

None of the vertices are marked and the set to be removed is empty  
`set_removed_vertices = []`

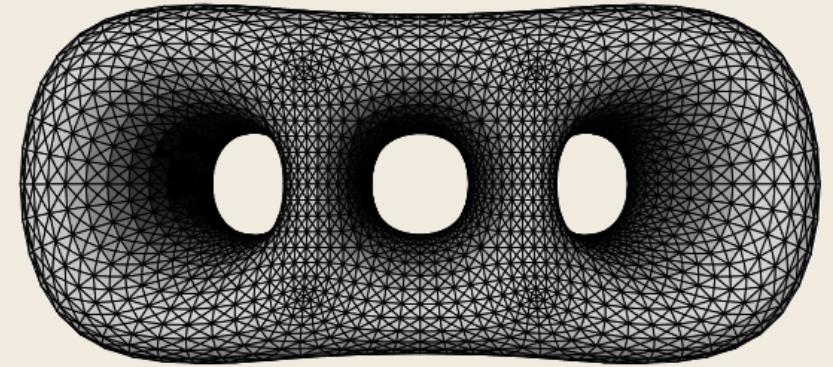
## # Hierarchical loop

While there is a non-marked vertex of degree  $< 12$   
    Selection of the 1st vertex in the priority queue  
        **Remove the vertex** and its star from  $K^l$   
        Marks its neighbors as unremovable  
        `set_removed_vertices.append(vertex)`

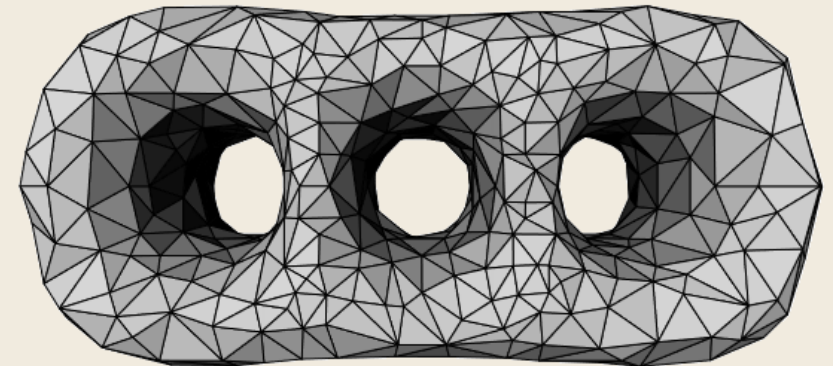
### **Retriangularization**

update level

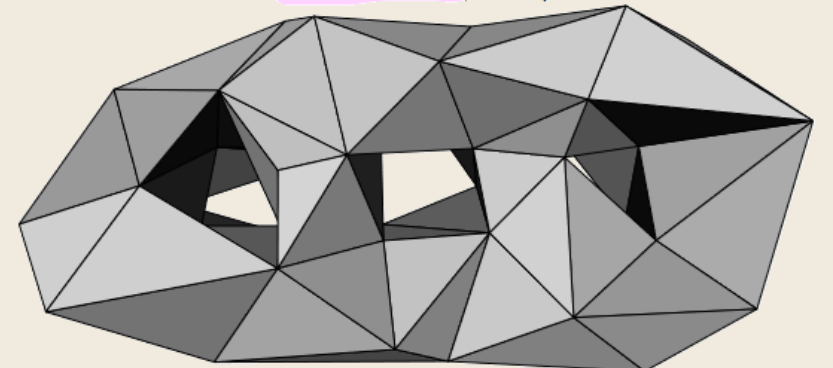
*Original mesh (level 14)*



*Intermediate mesh (level 6)*



*Coarsest mesh (level 0)*



## II – Priority queue

**The priority queue** : vertices with small and flat 1-ring neighborhoods will be chosen

At level  $l$ , for a vertex  $p_i \in P^l$ , we consider its 1-ring neighborhood  $\phi(|\text{star}(i)|)$  and compute its area  $a(i)$  and estimate its curvature  $\kappa(i)$ . These quantities are computed relative to  $K^l$ , the current level. We assign a priority to  $\{i\}$  inversely proportional to a convex combination of relative area and curvature.

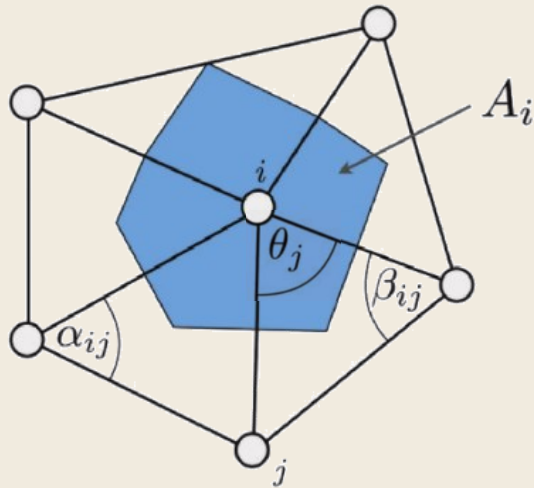
$$w(\lambda, i) = \lambda \frac{a(i)}{\max_{p_i \in P^l} a(i)} + (1 - \lambda) \frac{\kappa(i)}{\max_{p_i \in P^l} \kappa(i)}.$$

- area  $A_i$
  - connectivity
  - angles  $\theta(i)$
  - curvature  $\kappa(i)$
- **Sorted weights list  $w$**

# II – Priority queue

## Curvature method 1

Curvature of a triangle mesh, definition and computation - Rodolphe Vaillant



$$\kappa(i) = |\kappa_1| + |\kappa_2|$$

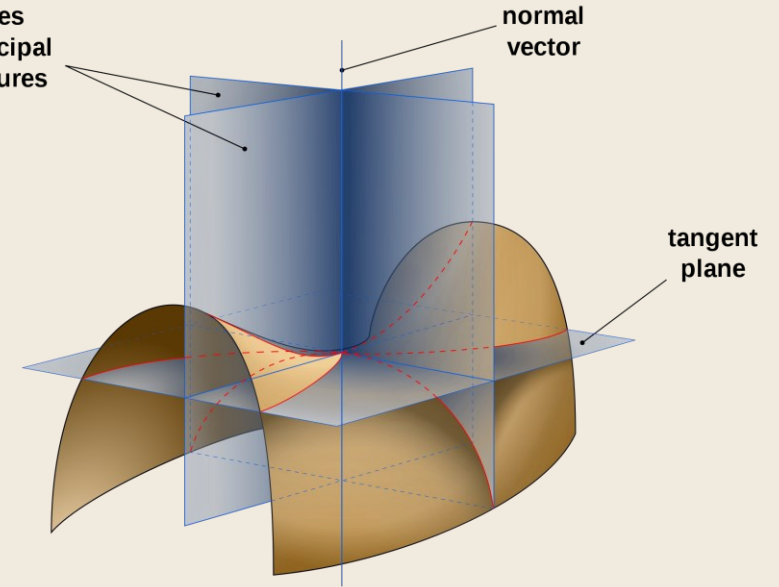
- Min and max

$$k_1 = H + \sqrt{H^2 - k_g}$$

$$k_2 = H - \sqrt{H^2 - k_g}$$

- Gaussian curvature

$$k_g = \frac{2\pi - \sum \theta_j}{A_i}$$



- Voronoï area

$$A_i = \frac{1}{3} \sum_{T_j \in \mathcal{N}(i)} \text{area}(T_j)$$

- Mean curvature

$$|H| = \frac{\|\Delta p_i\|}{2}$$

sign of the mean-curvature H

$$\text{dot}(\vec{n}_i, -\Delta p_i)$$

- Discrete Laplace operator

$$\Delta \vec{p}_i = \frac{1}{2A_i} \sum_{j \in \mathcal{N}(i)} (\cot \alpha_{ij} + \cot \beta_{ij})(p_j - p_i)$$

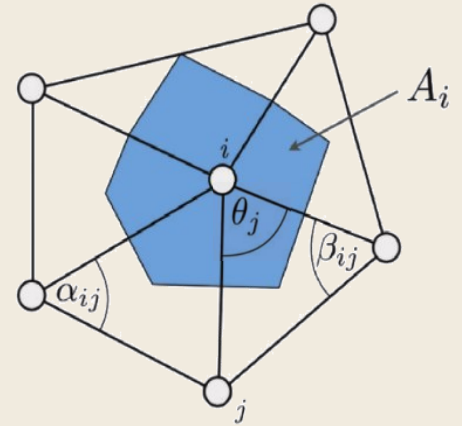
$$\Delta \vec{p}_i = \frac{1}{|\mathcal{N}(i)|} \sum_{j \in \mathcal{N}(i)} 1 \cdot (p_j - p_i)$$

## II – Priority queue

### Curvature method 2

*PC-MSDM: A quality metric for 3D point clouds - Gabriel Meynet, Julie Digne, Guillaume Lavoué*

- Local approximation of surface near 3D points
- Computation of mean curvature
- Steps:
  - Principal Component Analysis for approximate an orthonormal frame (done with SVD)
  - Local least squares fitting of a quadric surface  $Q(x, y) = ax^2 + by^2 + cxy + dx + ey + f$
  - That minimizes:  $\sum_i \|z_i - Q(x_i, y_i)\|^2$
  - Derivatives of the quadric surface's coefficients used to estimate the curvature



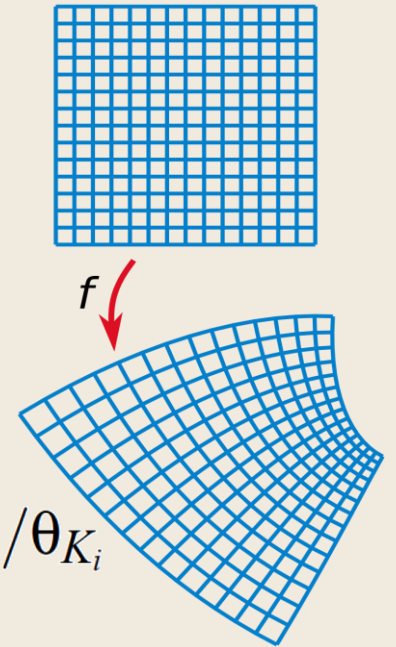
$$Curv(p) = \frac{(1 + d^2)a + (1 + e^2)b - 4abc}{(1 + e^2 + d^2)^{\frac{3}{2}}}$$

# III - Retriangulation

- We need to retriangulate the holes left by removing the independent set.
- Use the **conformal map**  $z^a$  which minimizes metric distortion to **map the neighborhood of a removed vertex into the plane**.

Let  $\{i\}$  be a vertex to be removed. Enumerate cyclically the  $K_i$  vertices in the 1-ring  $N(i) = \{j_k \mid 1 \leq k \leq K_i\}$  such that  $\{j_{k-1}, i, j_k\} \in K^l$  with  $j_0 = j_{K_i}$ . A piecewise linear approximation of  $z^a$ , which we denote by  $\mu_i$ , is defined by its values for the center point and 1-ring neighbors; namely,  $\mu_i(p_i) = 0$  and  $\mu_i(p_{j_k}) = r_k^a \exp(i\theta_k a)$ , where :

$$r_k = \|p_i - p_{j_k}\| \quad \theta_k = \sum_{l=1}^k \angle(p_{j_{l-1}}, p_i, p_{j_l}) \quad a = 2\pi/\theta_{K_i}$$

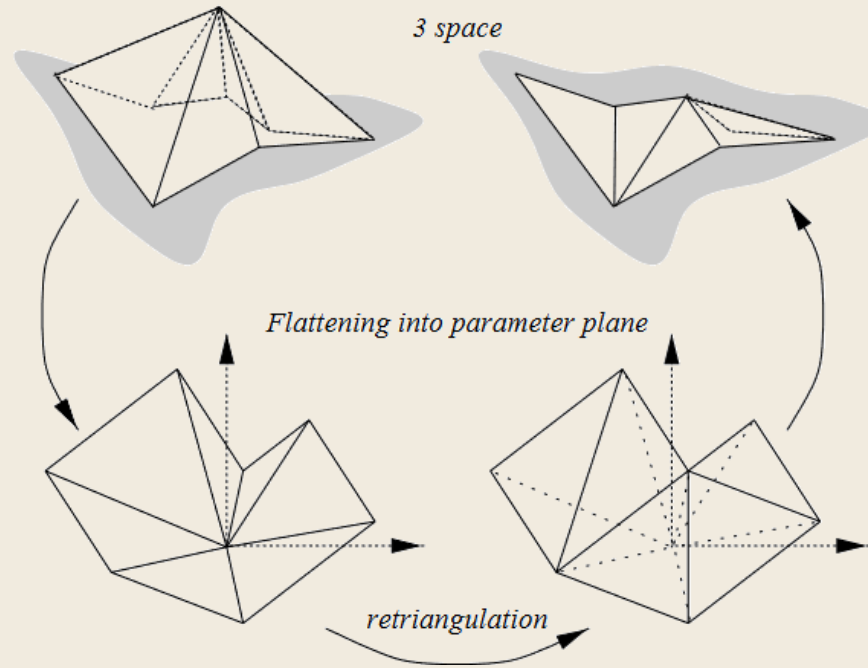


- When the vertex to be removed is a **boundary vertex**, we map to a half disk by setting  $a = \pi/\theta_{K_i}$



# III - Retriangulation

- **Advantages** of the conformal map:  
always exists + easy to compute + minimizes metric distortion + bijection
- Once the 1-ring is flattened, we can retriangulate the hole using, with a **constrained Delaunay triangulation**. This tells us how to build  $K^{l-1}$ .



In order to remove a vertex  $p_i$ , its star (i) is mapped from 3-space to a plane using the map  $z^a$ . In the plane the central vertex is removed and the resulting hole retriangulated (bottom right).

# III - Decompression

While the number of vertices > desired number of vertices:

- Calculate 'w' for each vertex

- While 'w' is not empty:

  - Retrieve the vertex with the highest weight

  - Lock neighboring vertices

  - Retriangulate the faces associated with the vertex

  - Populate the operations list

- Retrieve faces that have not been retriangulated in this iteration to integrate into the model

- Reindex the vertices and faces of the new model

Add the last vertices and faces to the operations list

Reverse operations

Create the model

```
class MappingFaceIndex:
    def __init__(self, faces):
        self.faces = faces
        self.face_indices = {}

    def assign_indices(self):
        return {self.face_to_tuple(face): index for index, face in enumerate(self.faces)}

    def init_indices(self):
        self.face_indices = self.assign_indices()

    def face_to_tuple(self, face):
        return (face.a, face.b, face.c)

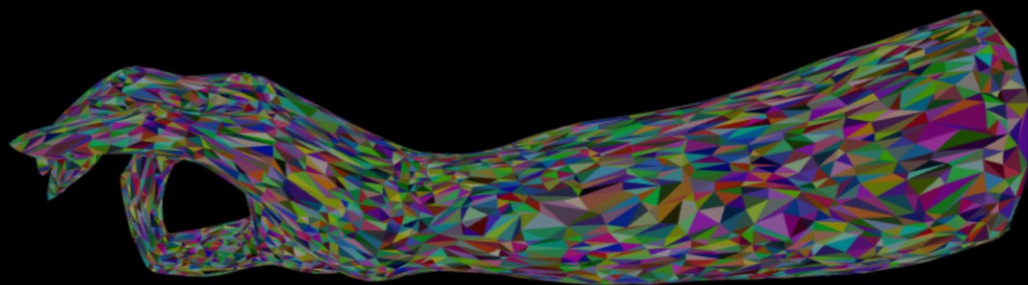
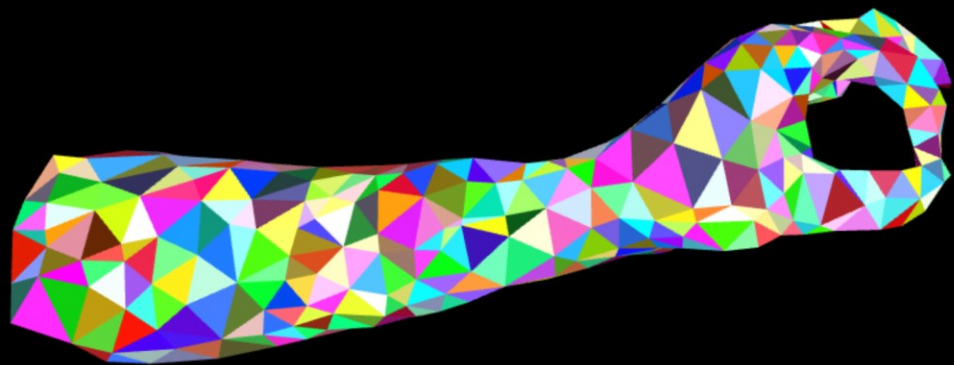
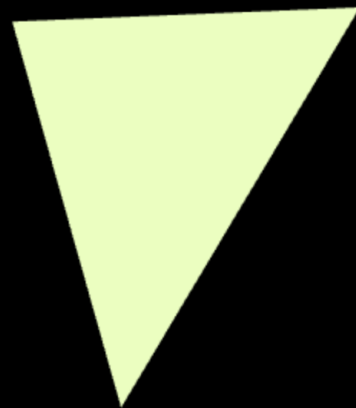
    def modify_face(self, old_face, new_face):
        # Trouver l'indice de l'ancienne face
        old_face_tuple = self.face_to_tuple(old_face)
        if old_face_tuple in self.face_indices:
            index = self.face_indices[old_face_tuple]

            # Remplacer l'ancienne face par la nouvelle dans la liste
            self.faces[index] = new_face

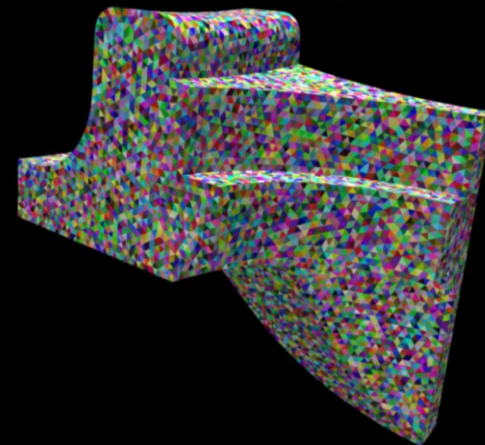
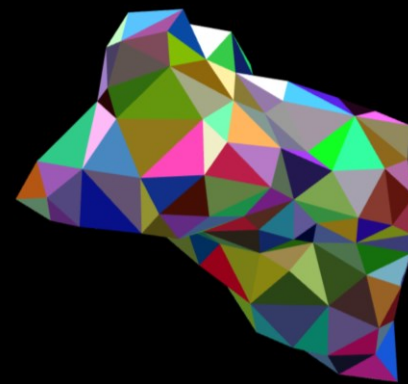
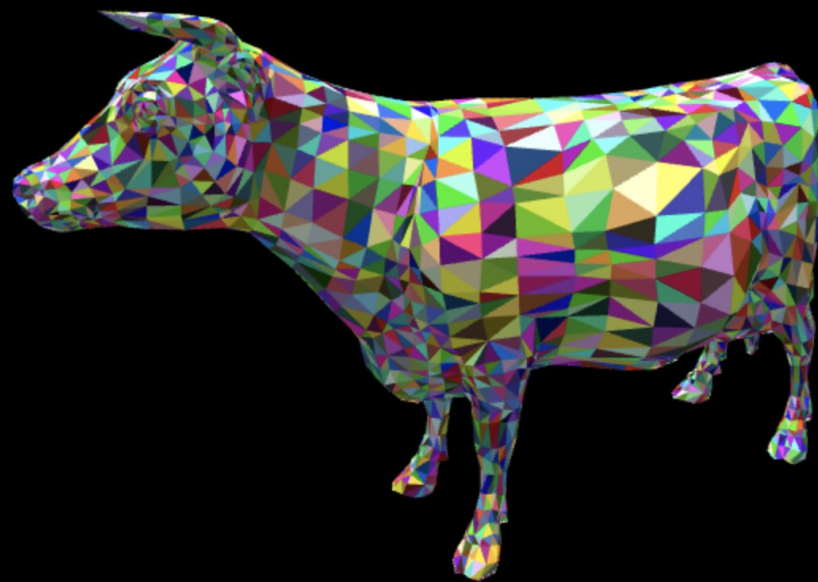
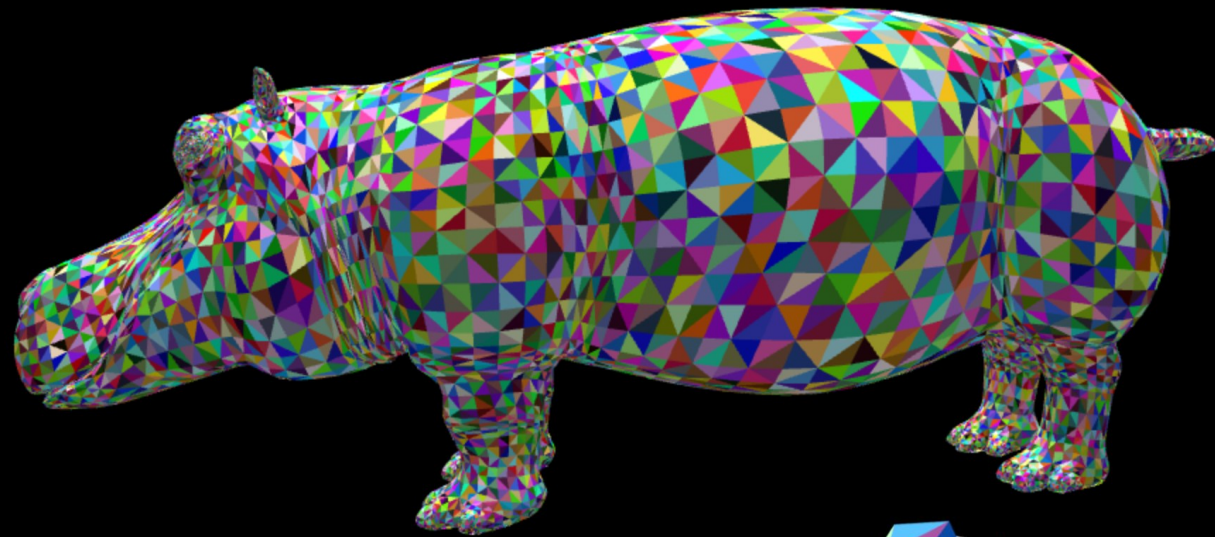
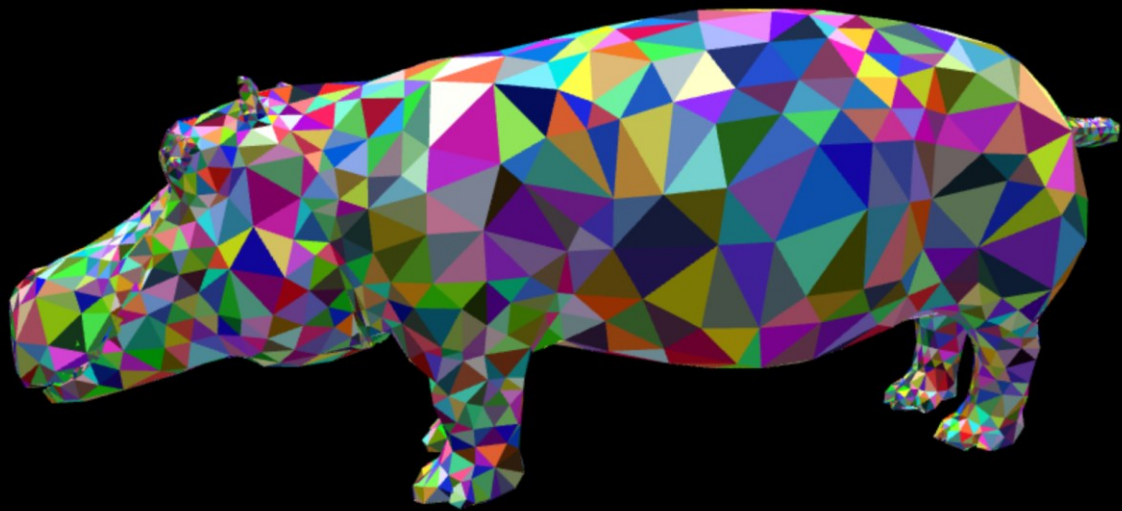
            # Mettre à jour le dictionnaire de mapping
            del self.face_indices[old_face_tuple]
            new_face_tuple = self.face_to_tuple(new_face)
            self.face_indices[new_face_tuple] = index
        else:
            raise ValueError("L'ancienne face n'existe pas dans le modèle.")

    def add_face(self, face, index_face):
        # Ajouter la face à la liste
        #self.faces.append(face)
        compteur = 0
        if index_face < len(self.faces):
            while compteur < len(self.faces):
                if self.faces[compteur] == face:
                    break
                compteur += 1
            if compteur == len(self.faces):
                self.faces.append(face)
        else:
            self.faces[index_face] = face
```

## IV - Results







# IV - Results

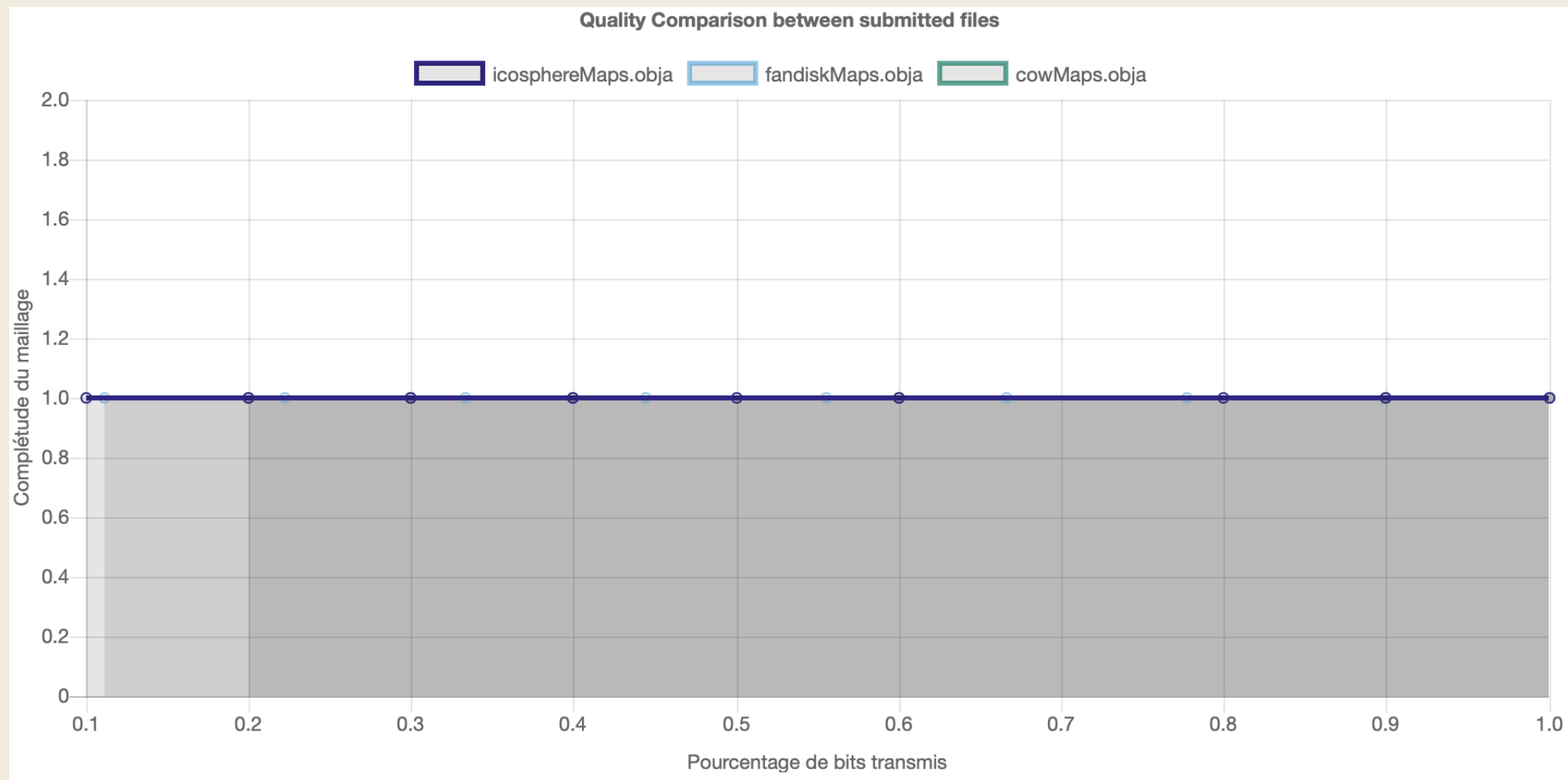
Table showing the effect of compression on the number of vertices

Dataset	Input size (vertex)	Output size (vertex)
Bunny	2503	84
Cow	3 906	1 126
Hand	10 000	473
FanDisc	9 864	4
Icosphere	642	3
Hippo	32 144	6 393

Videos compression + decompression

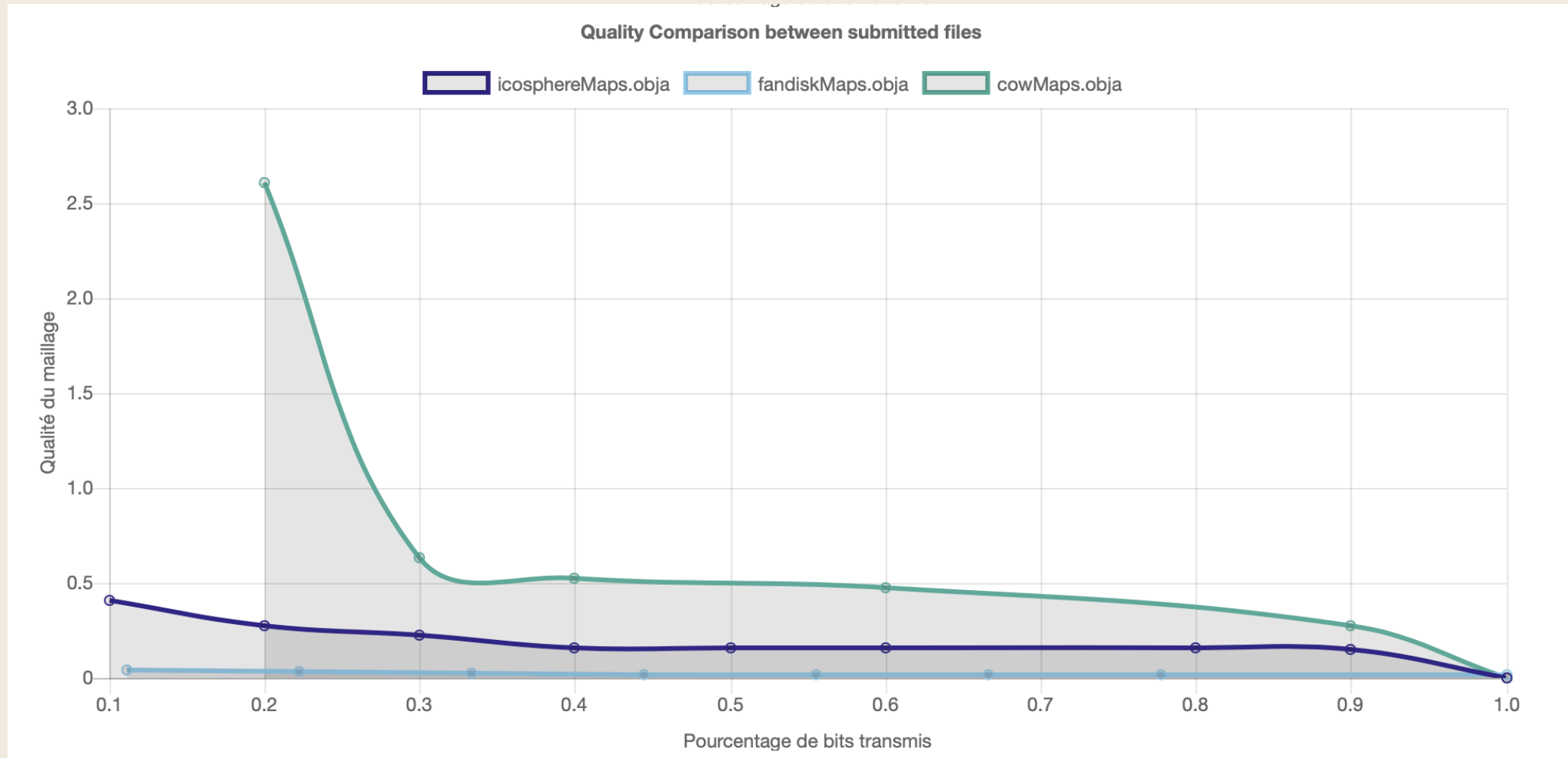
# IV - Results

## Quantification - Completude



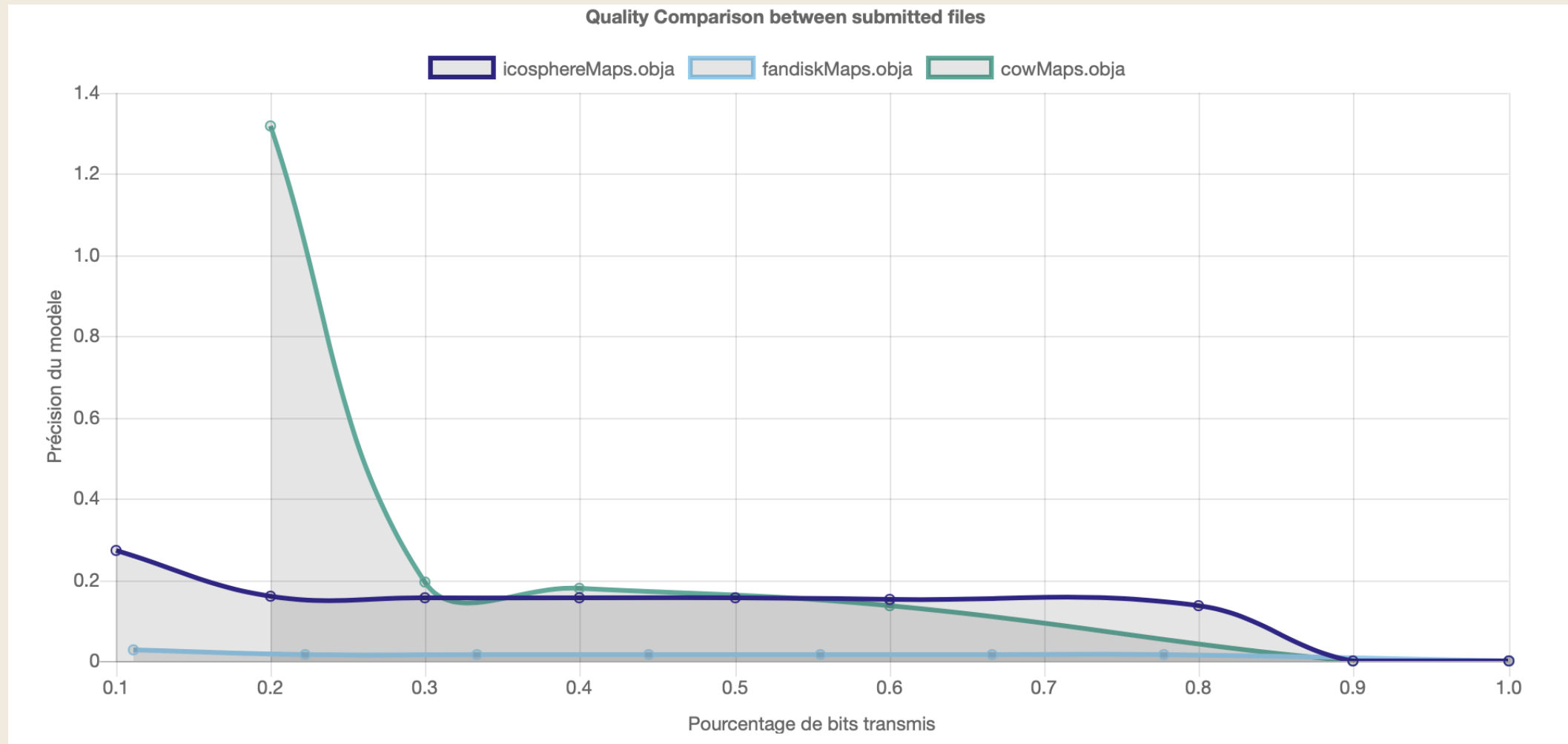
# IV - Results

## Quantification – Hausdorff distance



# IV - Results

## Quantification – Middlebury





# V - Issues

The issues we encountered and solved:

- Delaunay to impose constraints (segment indexing issue)
- Indexing problem (replace 'for' loops with mappings)
- Curvature calculation (switching methods based on another scientific article)

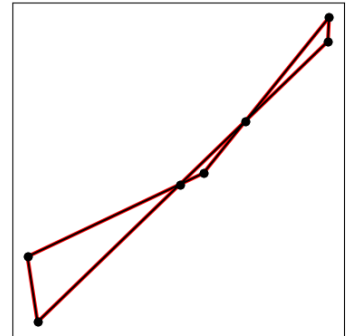
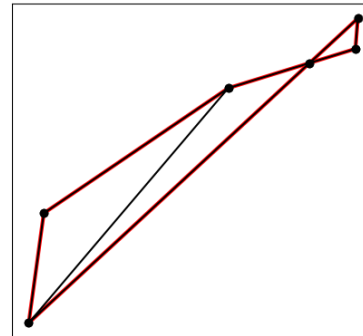
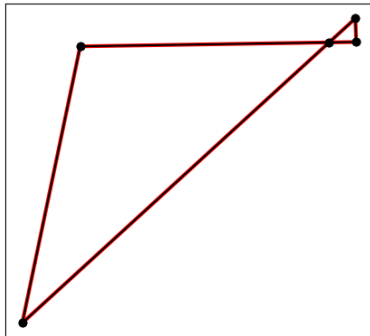
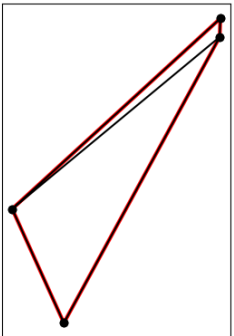
The remaining issues to address:

- Retriangulation issue for non-compact problems
- Edge handling (manual retriangulation)

# V - Issues

## Edge handling

- To find points on the borders:
  - an edge of the border is an edge that belongs to only one triangle.
  - a vertex on the border is a vertex that belongs to an edge on the border.
- Conformal map: for the borders, instead of mapping the neighbors of vertices to be removed by a disc, they are mapped by a half-disc.
- Issue during Delaunay retriangulation on the borders: handled manually.



**Thank you for your attention**