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# Discovery of Sensor Network Geometry

Jie Gao

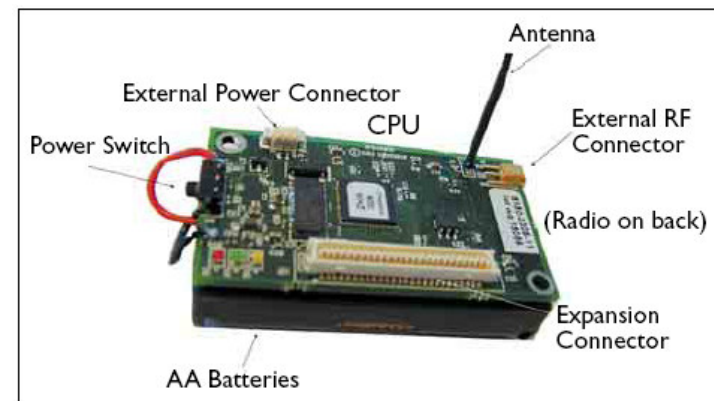
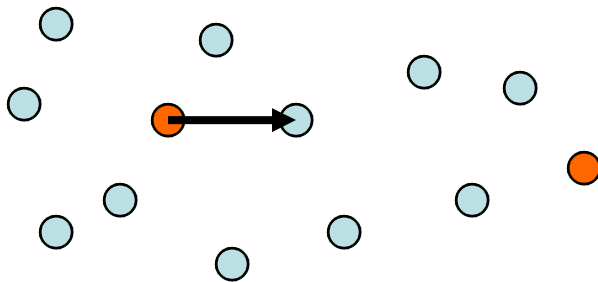
State University of New York at Stony Brook

Joint work with Sol Lederer and Yue Wang

# Localization of sensor nodes

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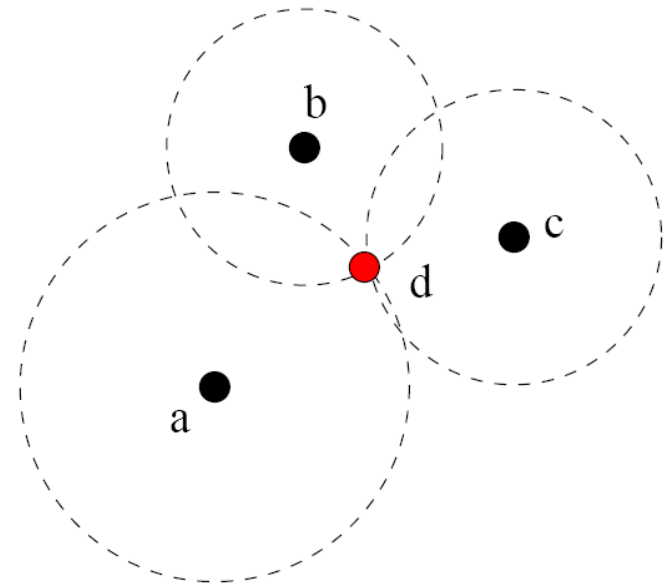
- Find out where the sensors are.
  - For the integrity of sensor data.
  - Useful for network organization and functioning (e.g., geographical routing).



# Existing localization techniques

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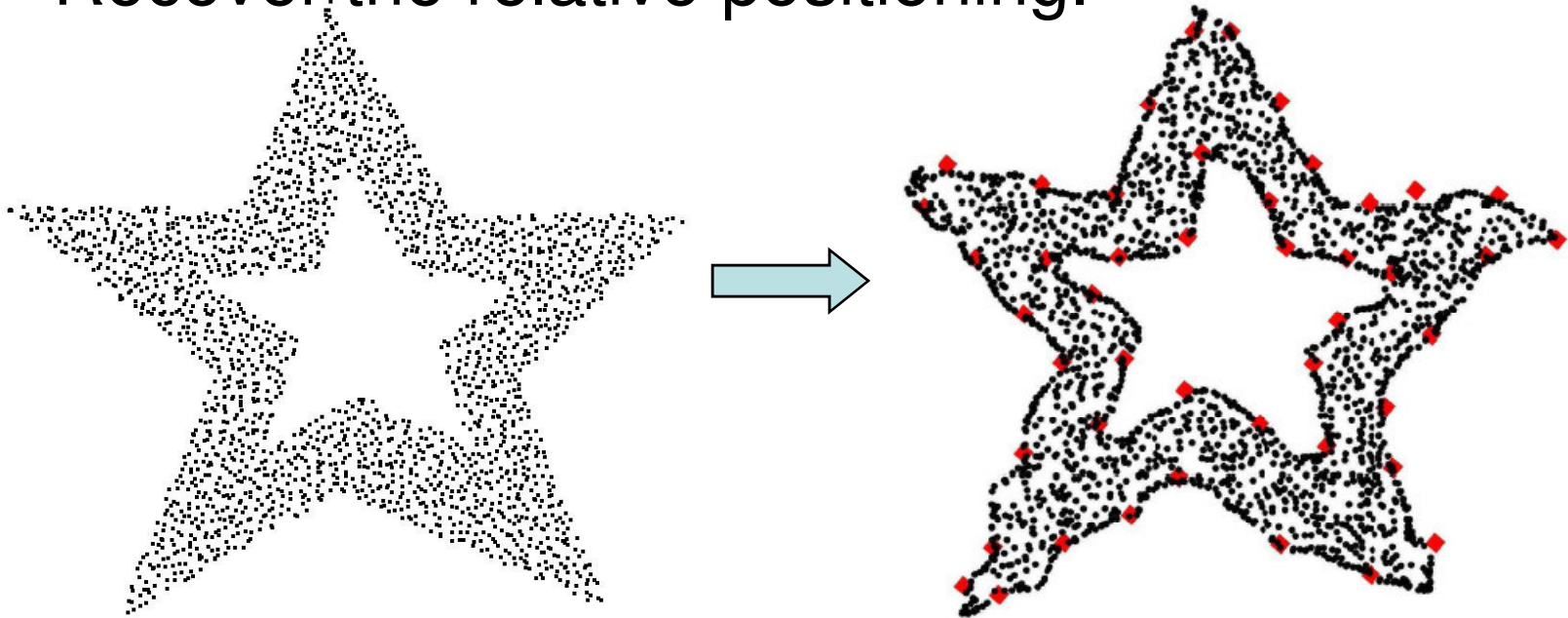
- GPS
  - expensive, large form factor, does not work in-door.
- Anchor-based localization.
  - Anchors have GPS.
  - Non-anchor nodes measure distances to anchors.
- This talk:
  - **Anchor-free** localization with connectivity information only.



# Localization of large-scale sensor network

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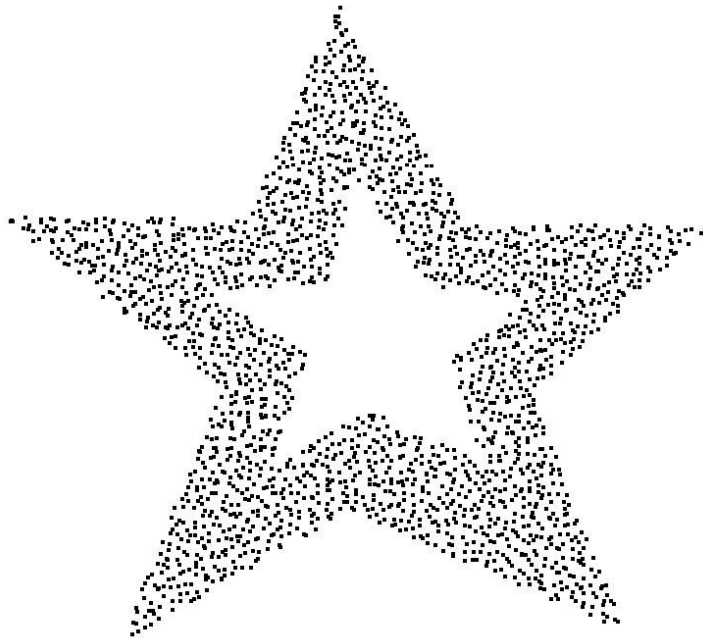
- Large-size sensor field with complex geometry.
- Nearby nodes are able to communicate.
- Use connectivity information only.
- Recover the relative positioning.



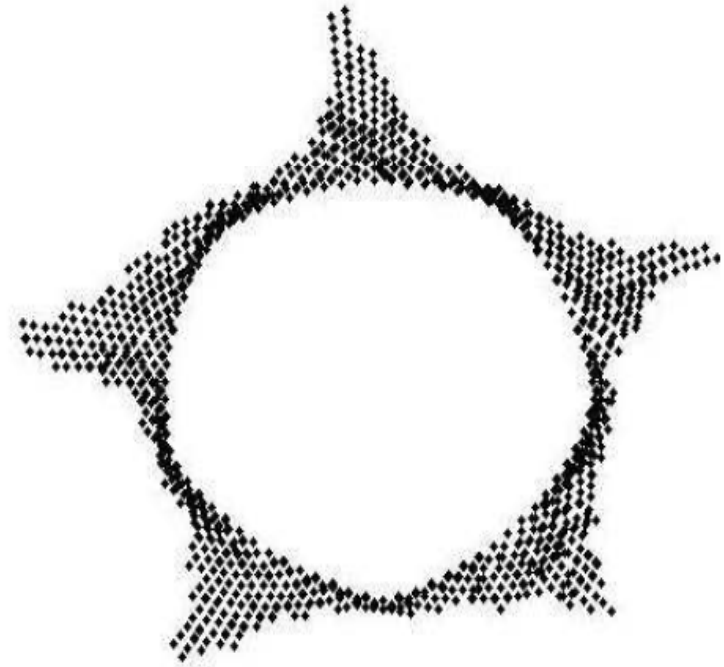
# Comparison: multi-dimensional scaling

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Ground Truth



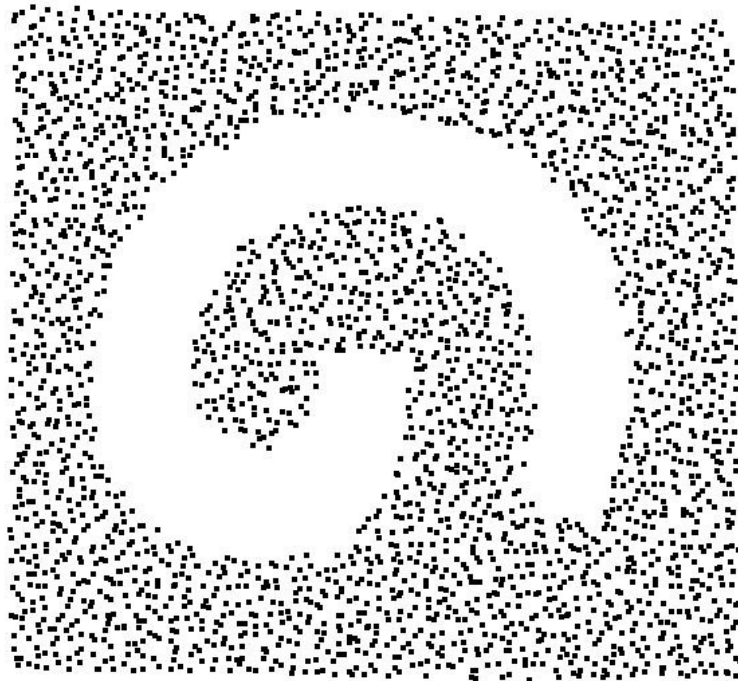
Multi-dimensional Scaling



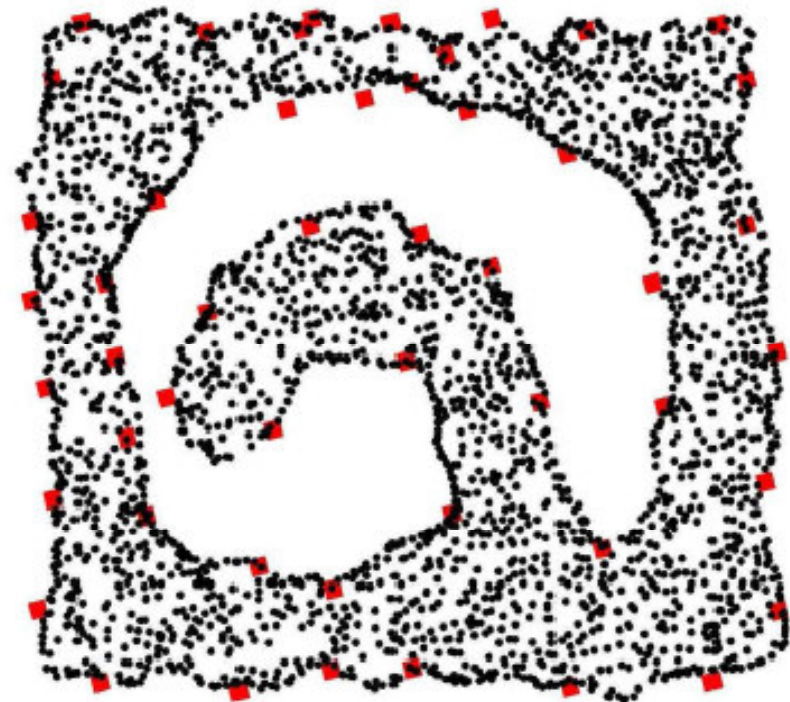
# Our results

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Ground Truth



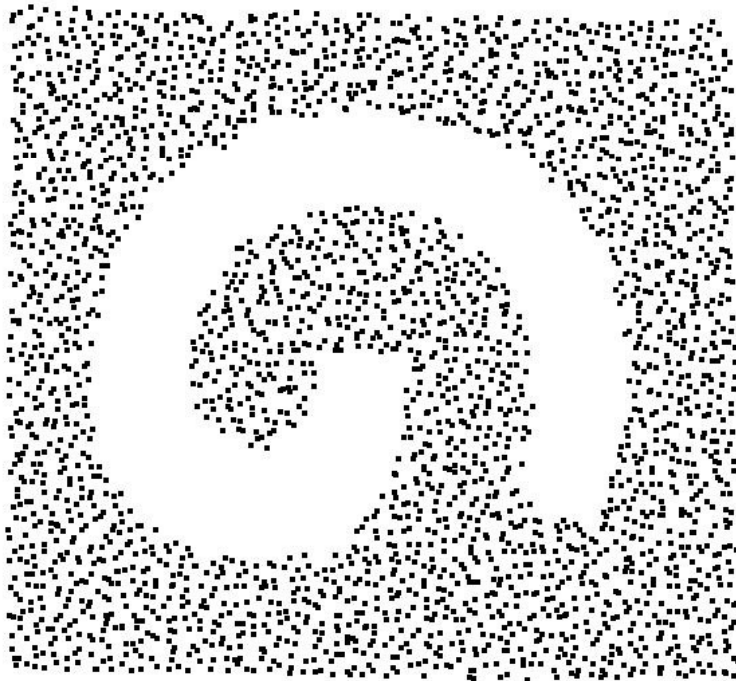
Our localization



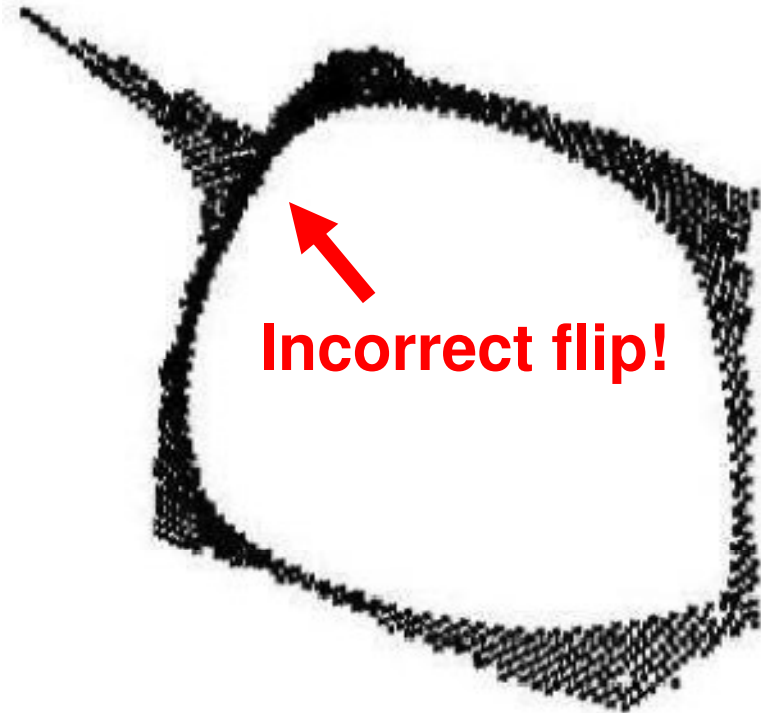
# Comparison: multi-dimensional scaling

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Ground Truth



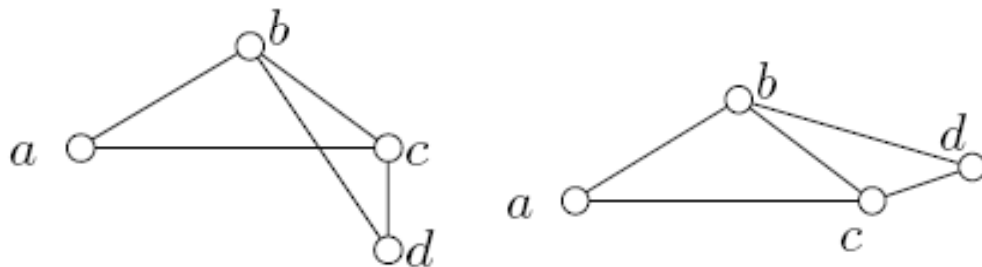
MDS



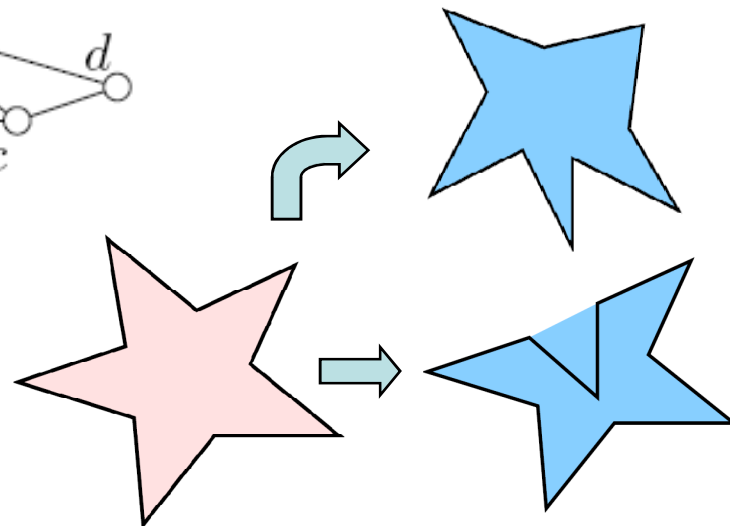
# The challenge: flip ambiguity

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- Given two triangles with **fixed** edge length, one can flip one triangle relative to the other.



- Incorrect global flip.

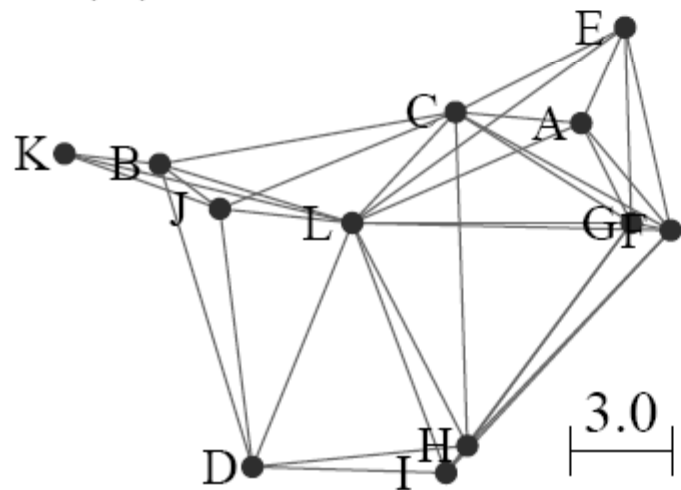


# In practice...

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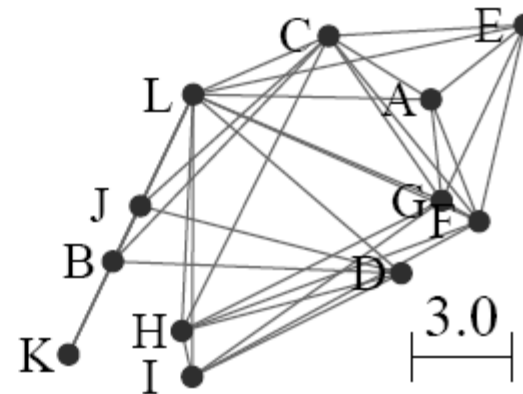
- Many optimization-based localization algorithms get stuck in local minimum.

(a) Ground truth



$$\sigma_{err} = 0.37$$

(b) Alternate realization



$$\sigma_{err} = 0.34$$

Reference: Anchor-Free Distributed Localization in Sensor Networks,  
2008/8/25 Nissanka B. Priyantha, Hari Balakrishnan, Erik Demaine, and Seth  
Teller, Technical Report MIT-LCS-TR-892, 2003.

# Outline

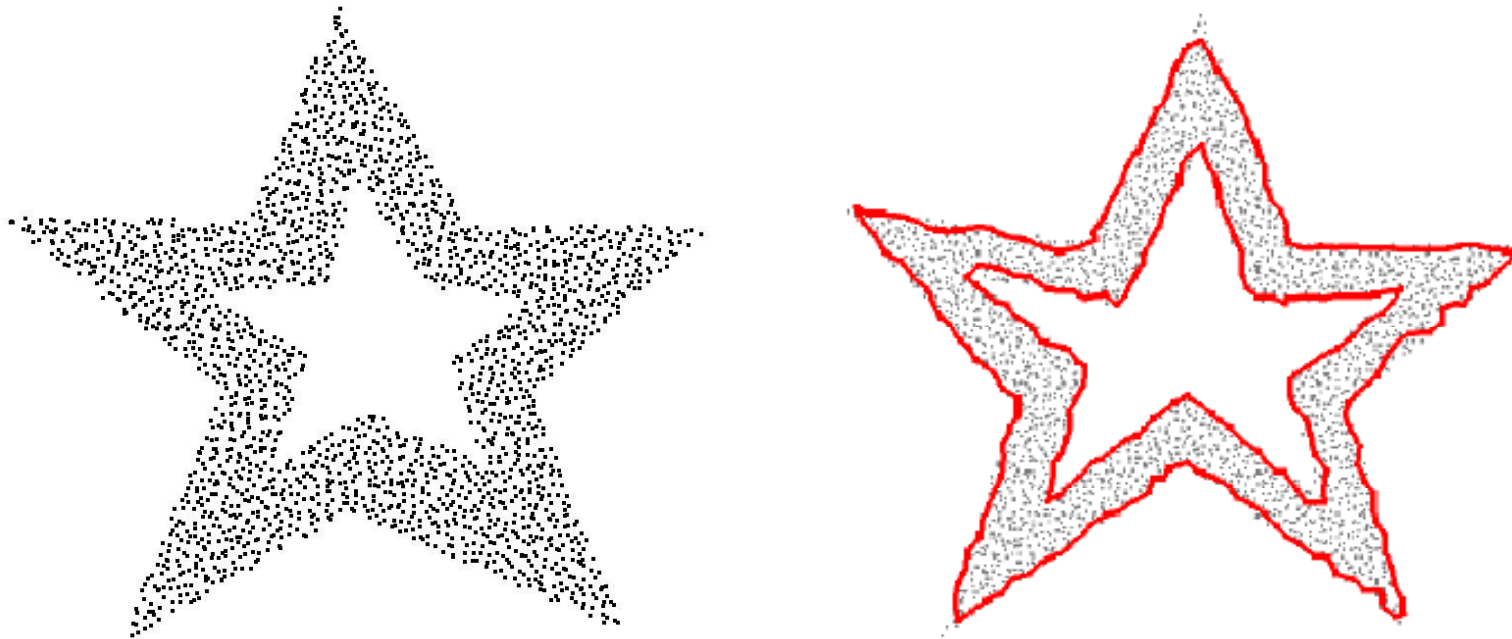
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- The localization algorithm.
- Why we do not have flip ambiguity.
- Theoretical foundations.

# Our algorithm

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## 1. Detect network boundary.



Existing algorithms by Sandor Fekete & Alexandar Kroller, Stefan Funke, our group.

# Our algorithm, cont.

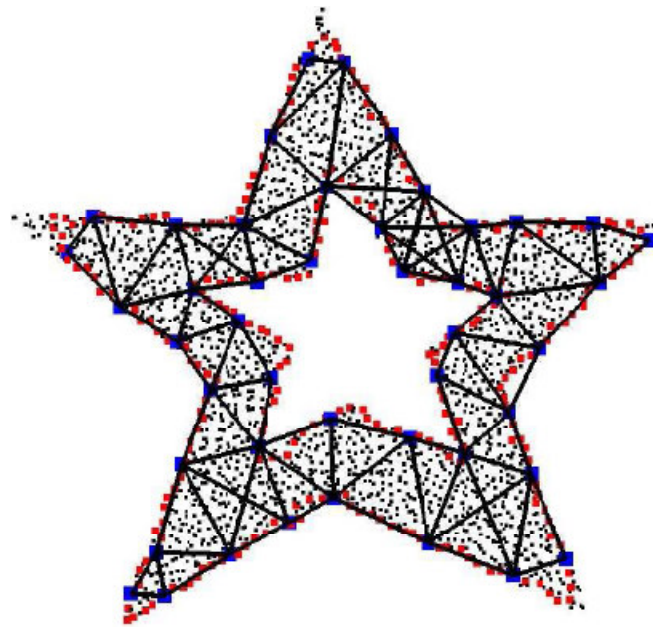
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## 2. Select landmarks on the boundary.

Landmark Voronoi diagram:  
Nodes identify closest landmark



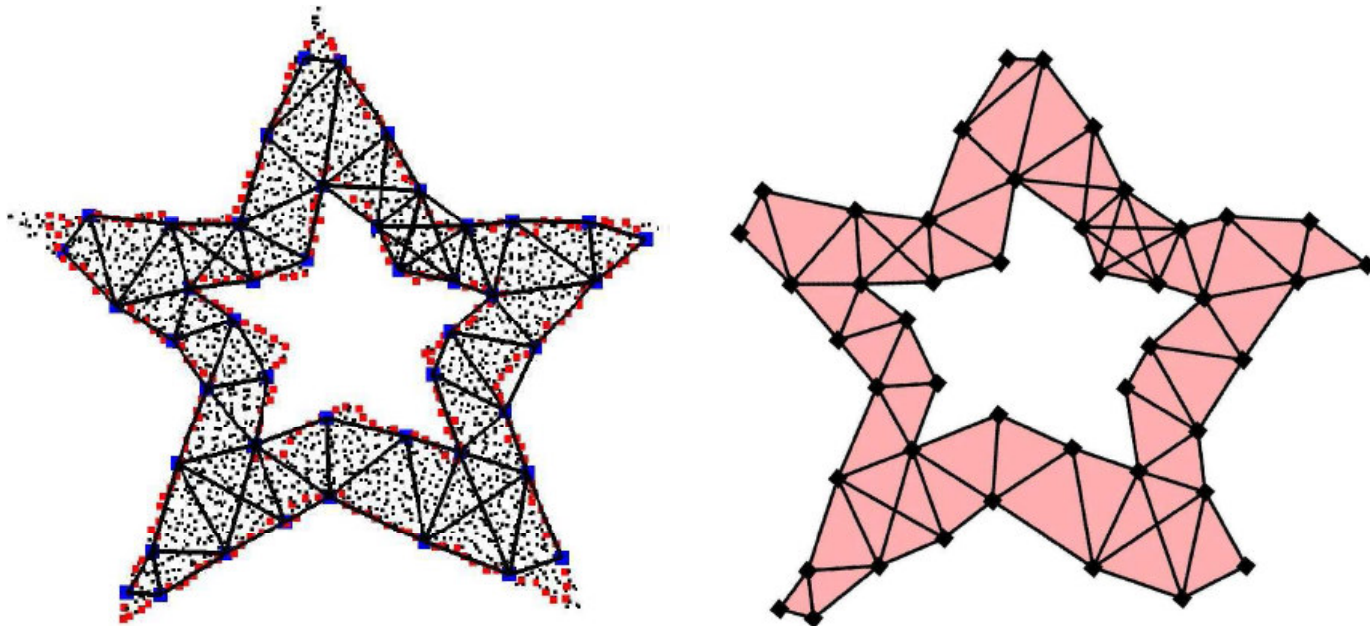
Combinatorial Delaunay complex  
Neighboring landmarks  
connected by an edge



# Our algorithm, cont

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3. Embed the combinatorial Delaunay complex (purely combinatorial, no geometry), find a realization (simplicial complex, with geometry)

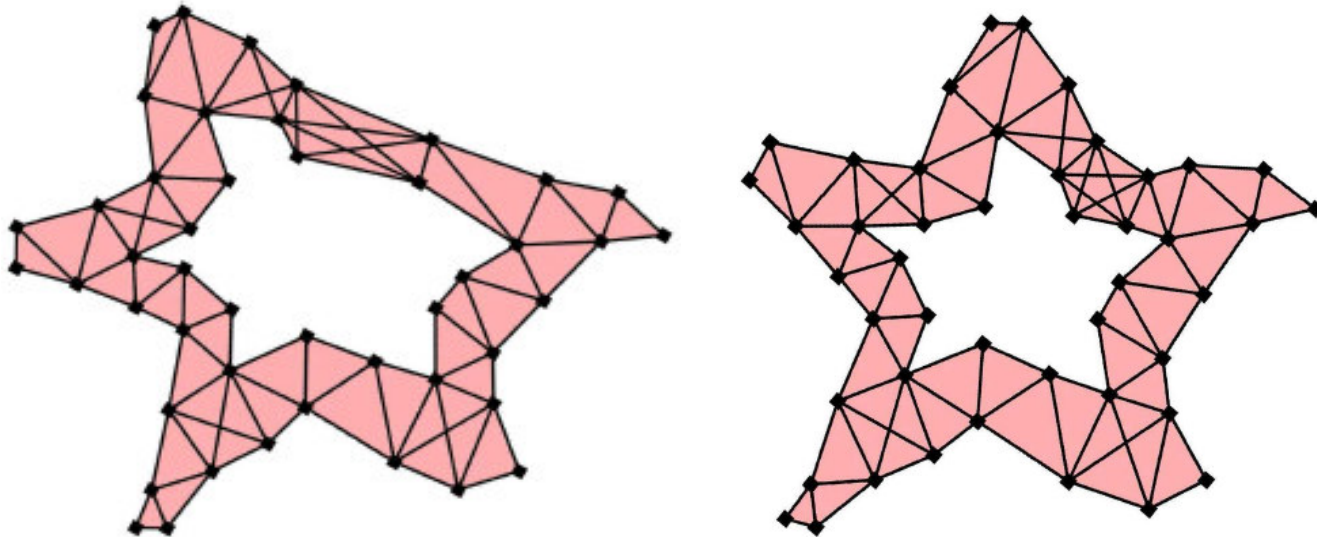


# Our algorithm, cont

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## 3. Embed the landmarks.

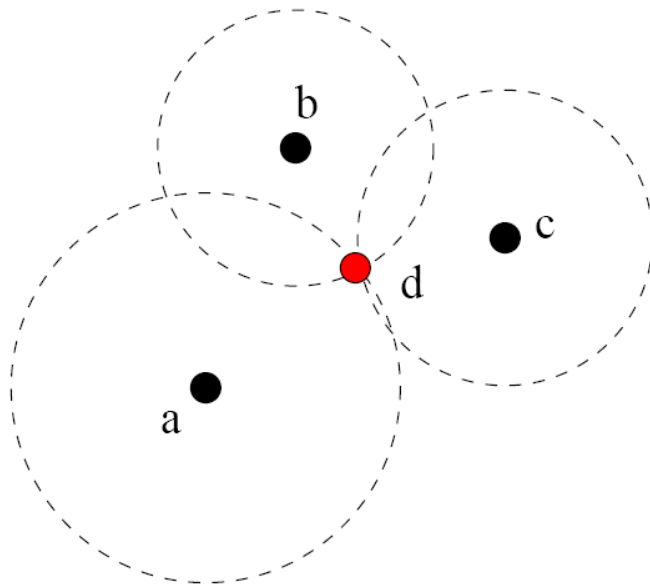
1. Each edge has a length = min hop count.
2. Glue the triangles incrementally.
3. Do a mass-spring relaxation to smooth out errors.



# Our algorithm, cont

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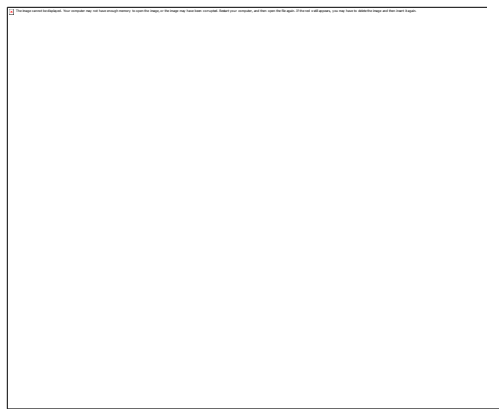
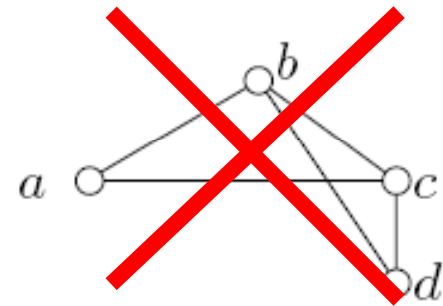
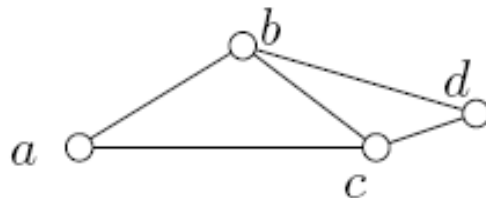
4. Embed the rest of the nodes.
  - Each node embeds itself with hop-count distances to nearby 3 landmarks.



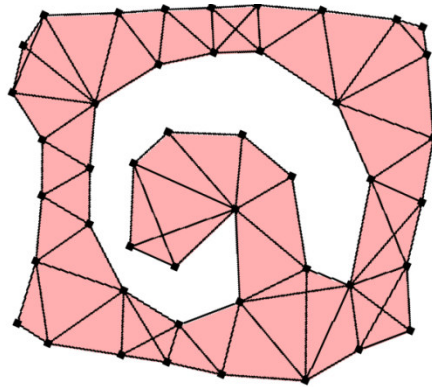
# Now: no flip ambiguity of Delaunay triangles

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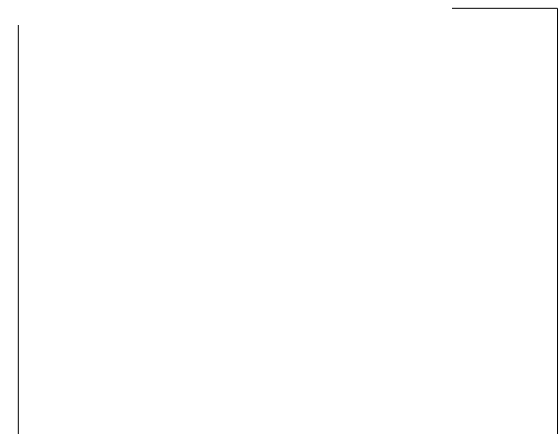
- Two adjacent Delaunay triangles must be embedded as disjoint.
- Key insight: they are “solid”.



2008/9/25  
Ground truth



Our result



MDS

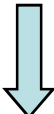
# Abstract simplicial complex & simplicial complex

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- Abstract simplicial complex (defined in terms of sets).
  - An edge: a set of two vertices.
  - A triangle: a set of three vertices, and so on.
  - An abstract simplicial complex: a collection  $A$  of sets s.t. if  $\alpha \in A$  and  $\beta \subset \alpha$ , then  $\beta \in A$ .

# Abstract simplicial complex & simplicial complex

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- Abstract simplicial complex (defined in terms of sets). 
- Simplicial complex (with geometry): any two simplices are either disjoint or intersect at a common face (subsimplex).

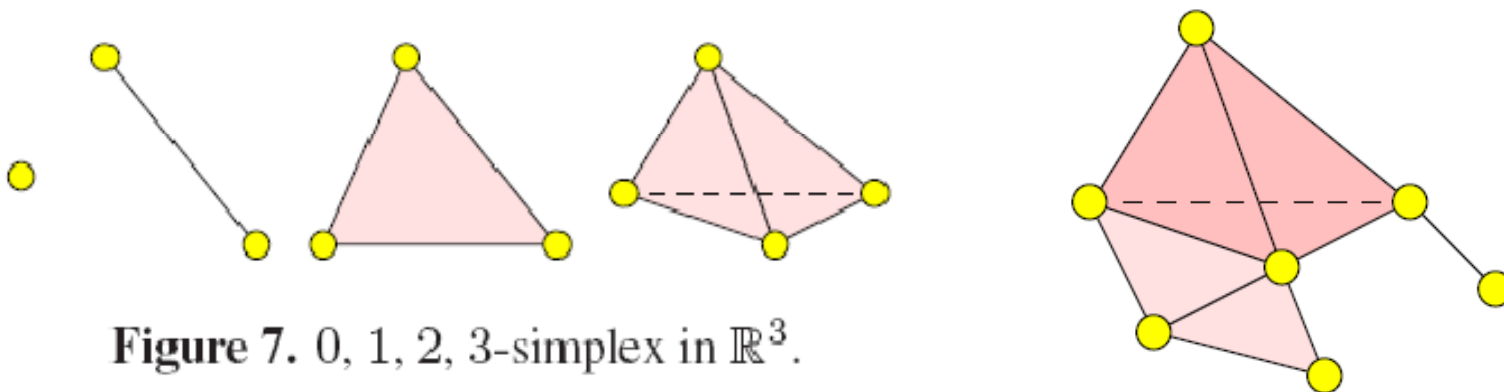
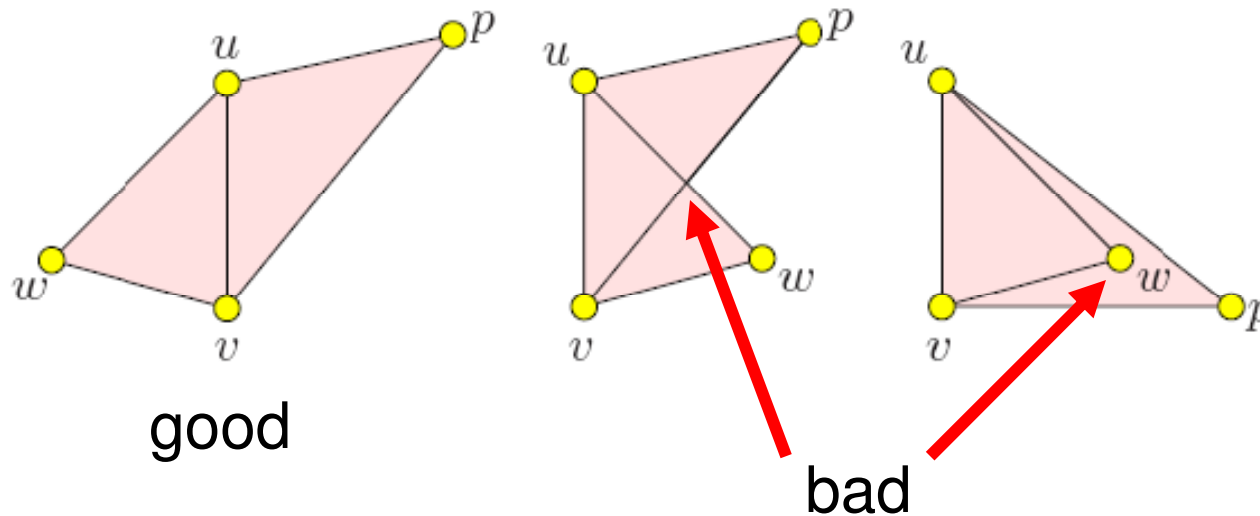


Figure 7. 0, 1, 2, 3-simplex in  $\mathbb{R}^3$ .

# No flip ambiguity for Delaunay triangles

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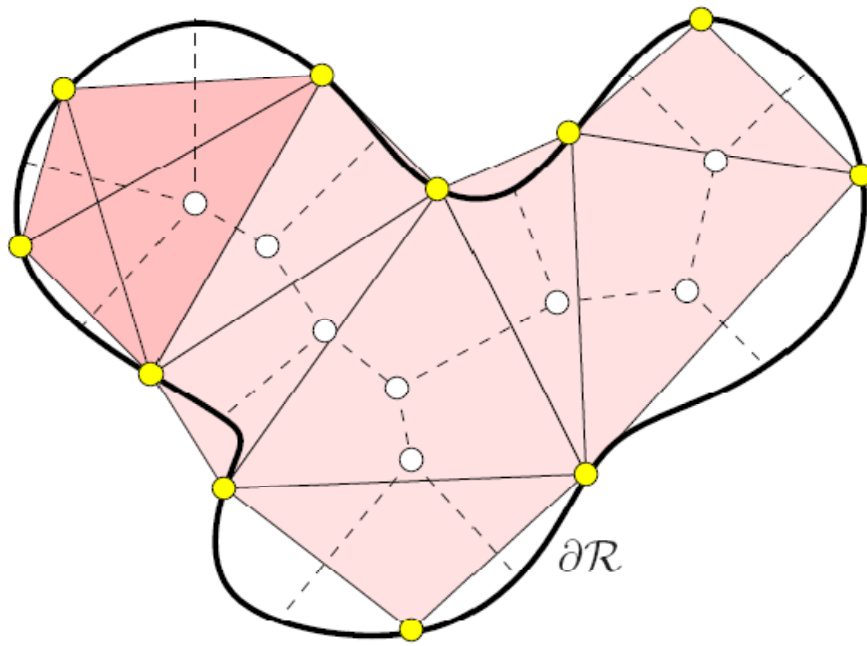
- Recall: we want to embed as a simplicial complex.
  - Intersection of two simplices is empty or a common face.



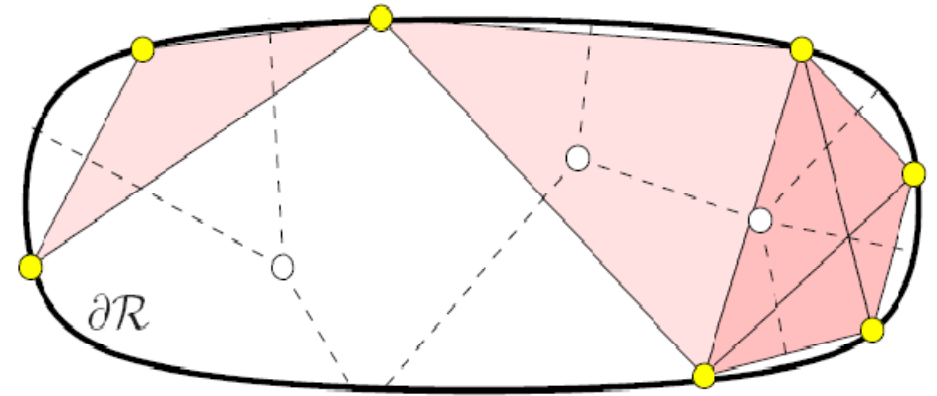
# What we need: make sure the Delaunay complex is rigid.

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- A graph is rigid if one can not deform the shape without changing the edge lengths.



Rigid

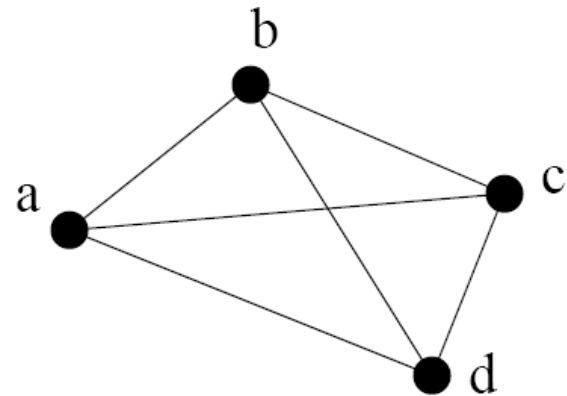
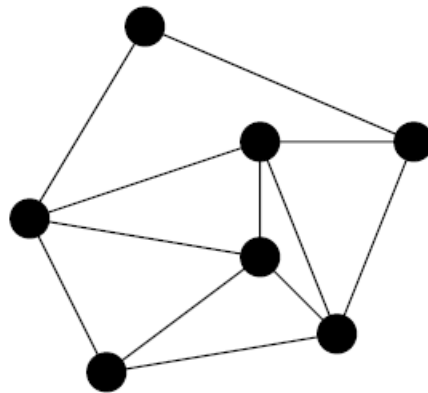
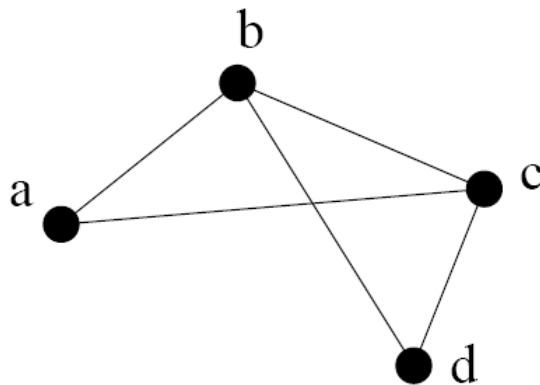


Non-rigid

# 2D rigidity: Laman condition

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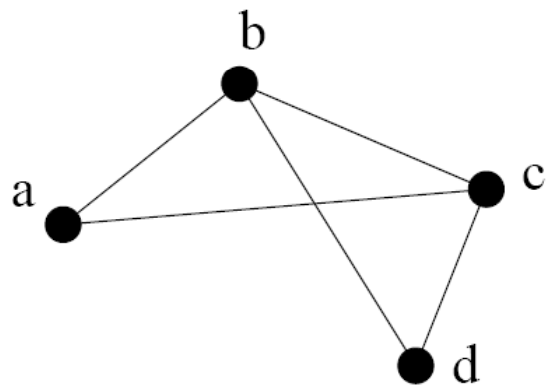
- A graph is rigid in 2D if and only if it contains a Laman graph on its vertices.
- A Laman graph has  $n$  vertices,  $2n-3$  edges and any subset of  $k$  vertices spans at most  $2k-3$  edges.



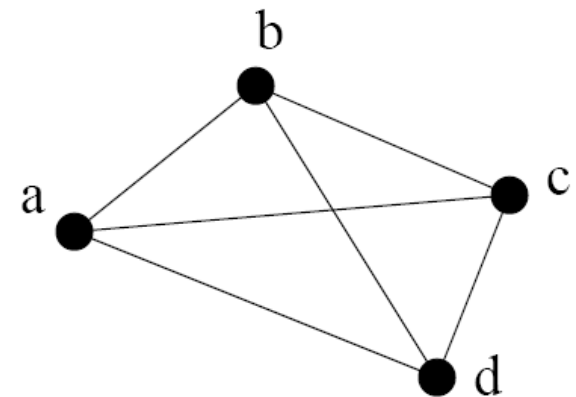
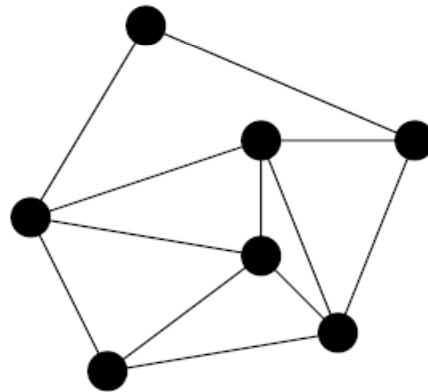
# Rigidity and global rigidity

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- A graph is globally rigid if it has a unique embedding in the plane.



Rigid, not globally rigid

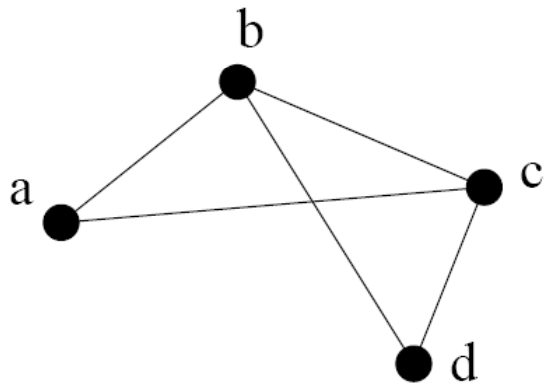


Globally rigid

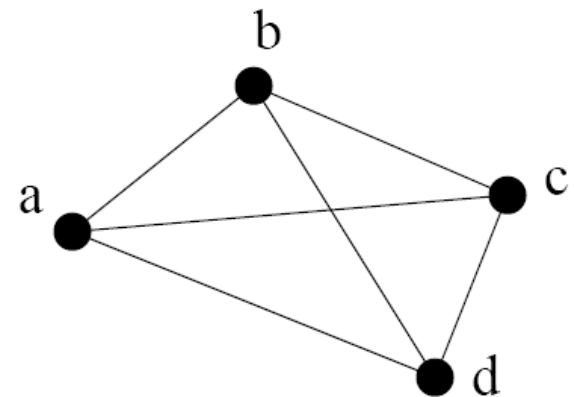
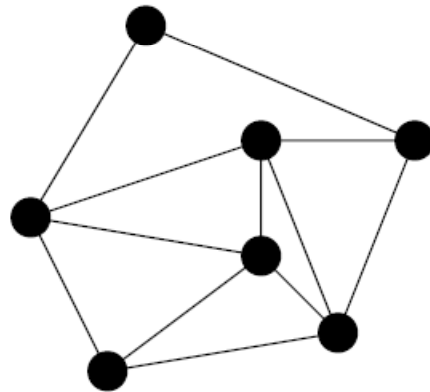
# Rigidity and global rigidity

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- A graph is globally rigid in 2D iff it is **3-connected** and **redundantly rigid** (rigid upon the removal of an edge).



Rigid, not globally rigid



Globally rigid

# Remark on global rigidity

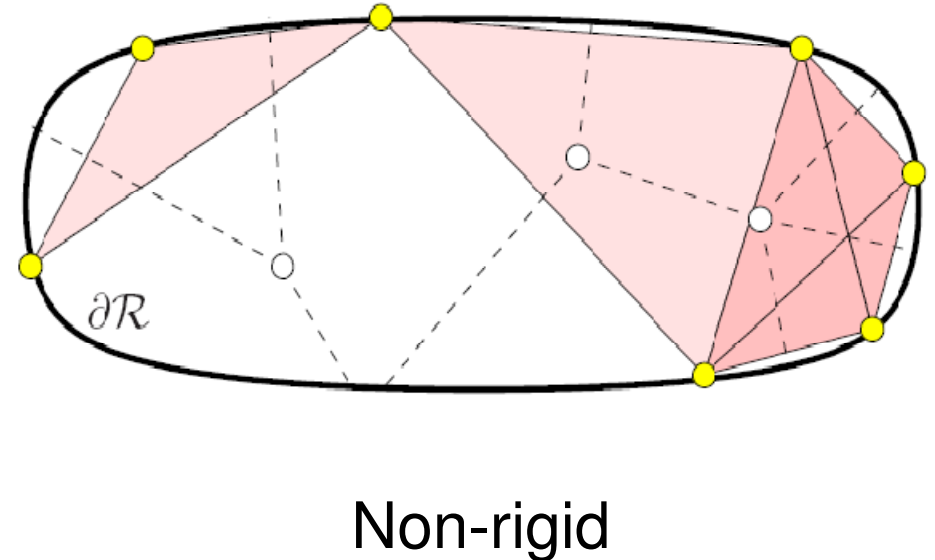
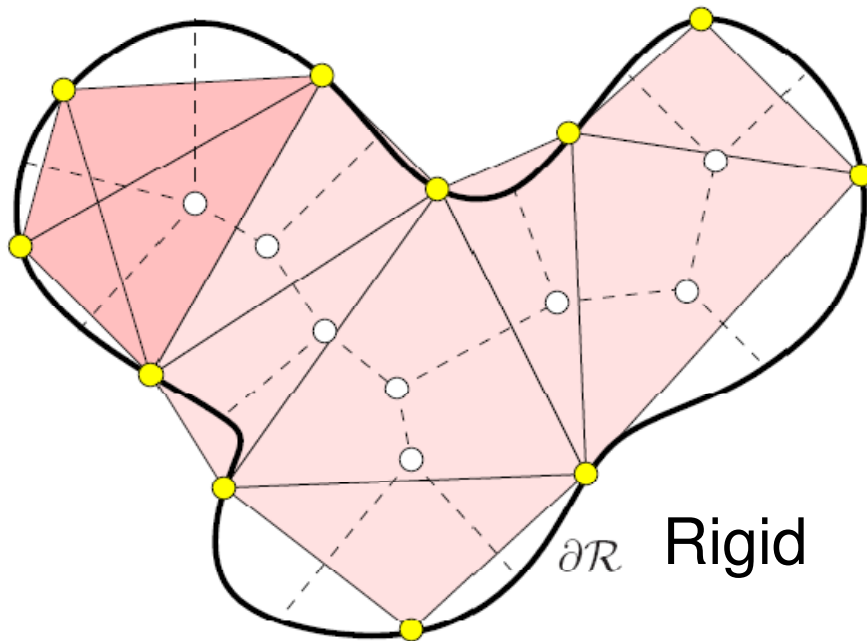
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- To embed **graphs**, **global rigidity** is desired.
- To embed **Delaunay complex**, **rigidity** is enough.
  
- Next: show the Delaunay graph is rigid given sufficiently dense landmarks.

# Connect to the rigidity result

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- If the Voronoi diagram (Voronoi edges and vertices) is connected in  $R$ , then the Delaunay graph is rigid.



# Proof by picture

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- Show that the Delaunay graph contains a Laman graph.

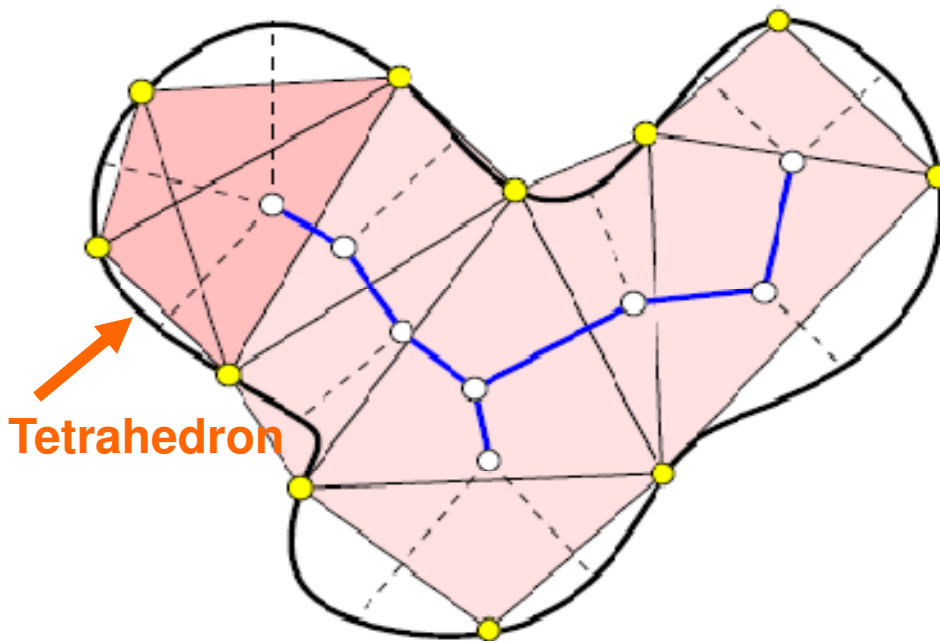
Take a tree connecting Voronoi vertices.

Count:

1. start with one Voronoi vertex (3 landmarks, 3 edges).
2. Add an adjacent Voronoi vertex (1 landmark, 2 edges).

Total:  $2n-3$  edges.

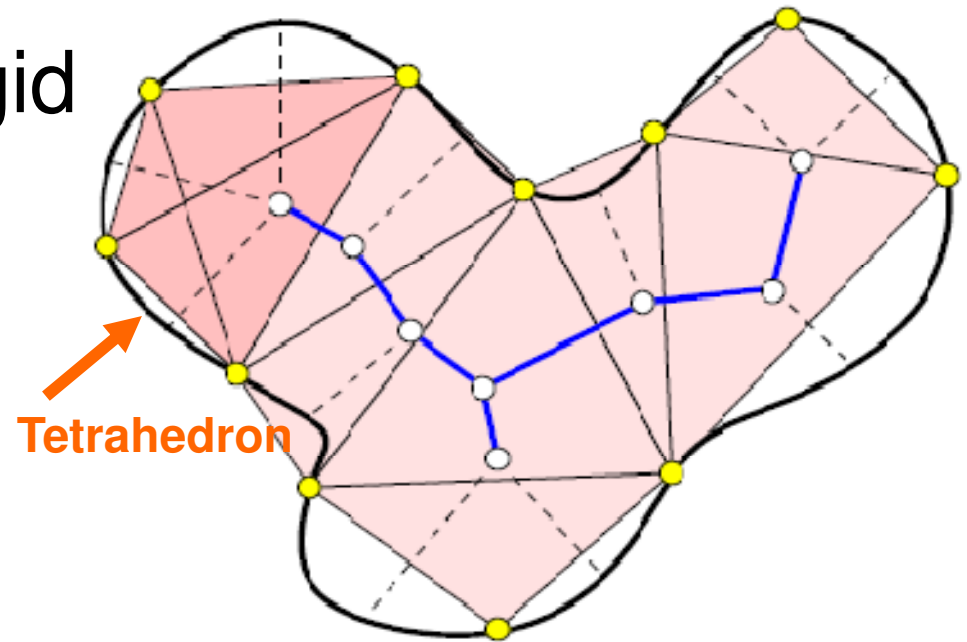
Also works for any subset of  $k$  vertices.



# A remark on degeneracy

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- For high-order simplices ( $k$ -simplex,  $k \geq 3$ ), we get a complete graph on 4 or more vertices.
- Thus it is globally rigid



# What is left

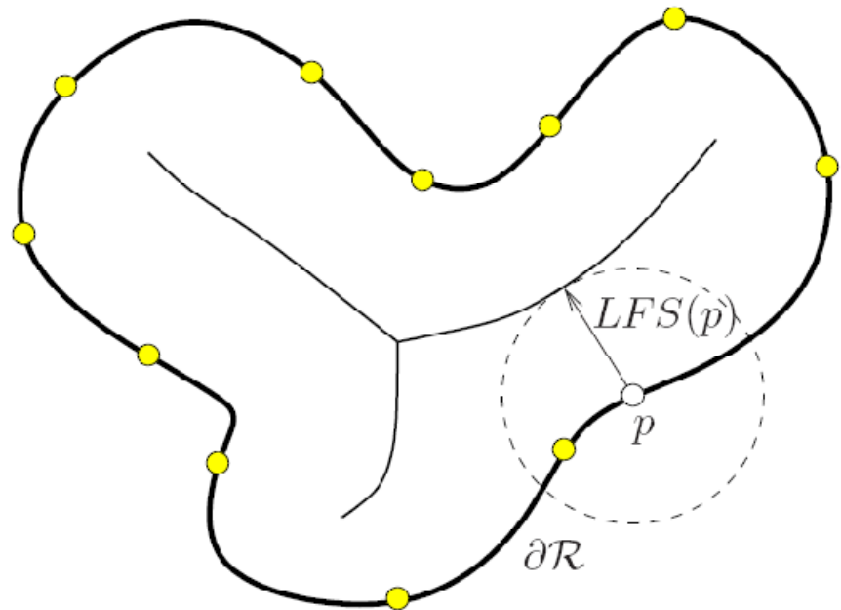
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- When the landmarks are **sufficiently dense** (subject to local geometric feature), the Voronoi graph is connected.

# Medial axis, local feature size

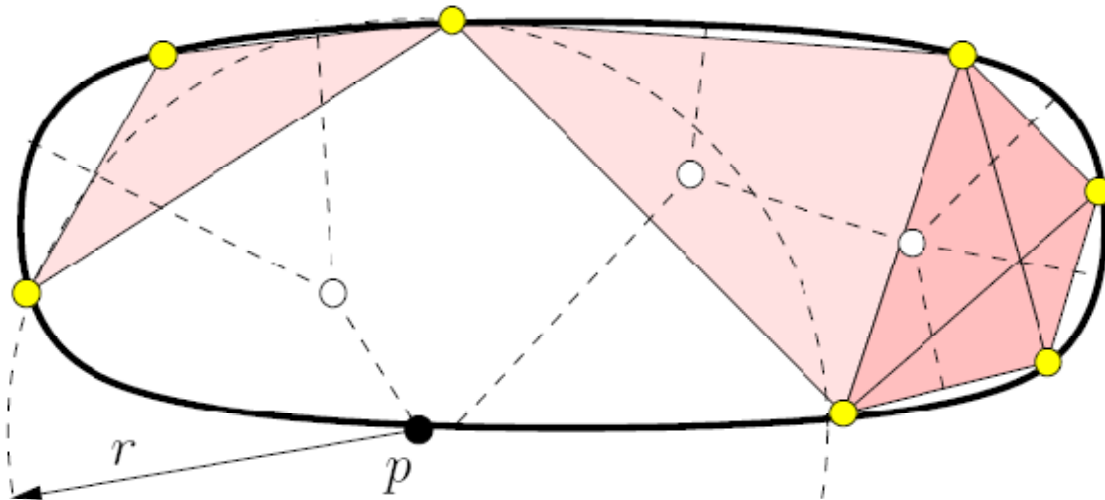
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- **Medial axis**: collection of points with two or more closest points on the boundary.
- **Local feature size**: the smallest distance to the medial axis.
- **Sampling criterion**: Each boundary point  $p$  has a landmark within  $LFS(p)$ .



# Proof by picture

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- Argue that the ball at  $p$  has a medial axis point inside.
- Thus this violates the sampling condition because  $r > \text{LFS}(p)$ .

# More results

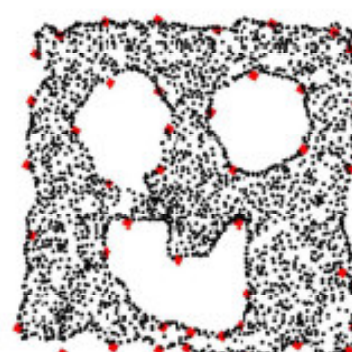
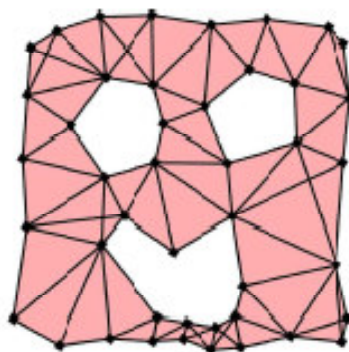
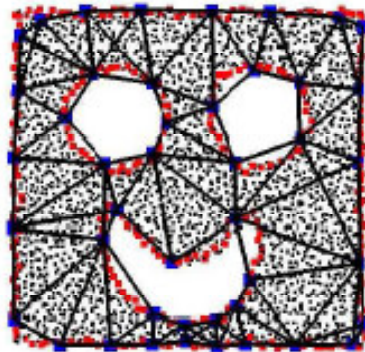
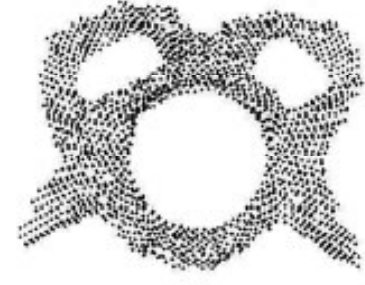
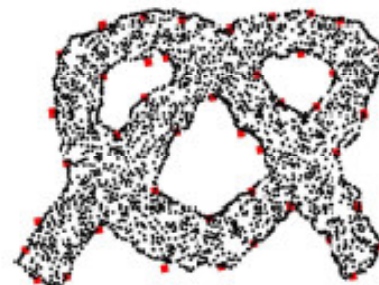
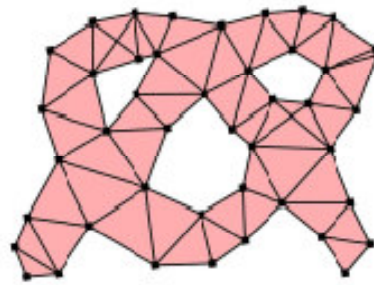
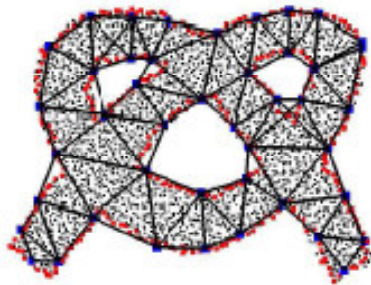
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Ground truth

Landmarks

All nodes

MDS



Unit disk graph model, Node average degree 6~10.

# More results

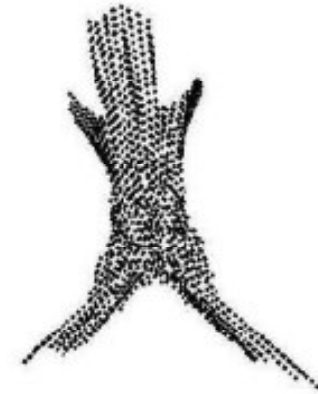
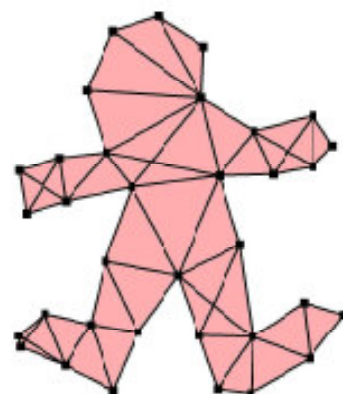
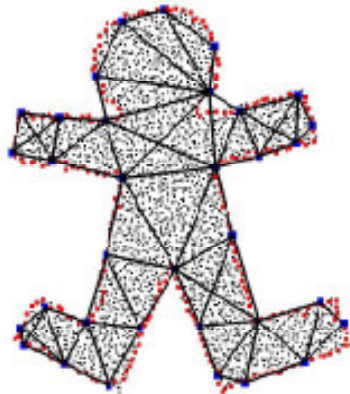
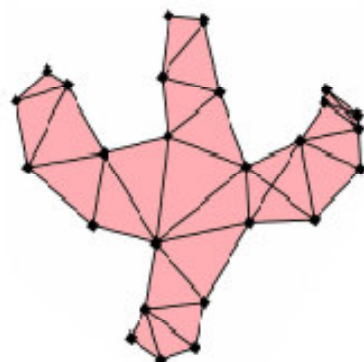
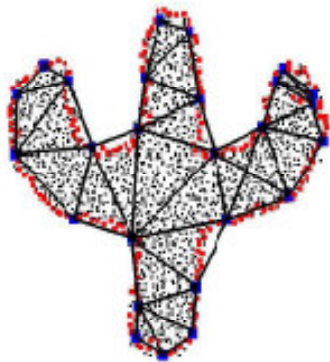
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Ground truth

Landmarks

All nodes

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Unit disk graph model, Node average degree 6~10.

# More results

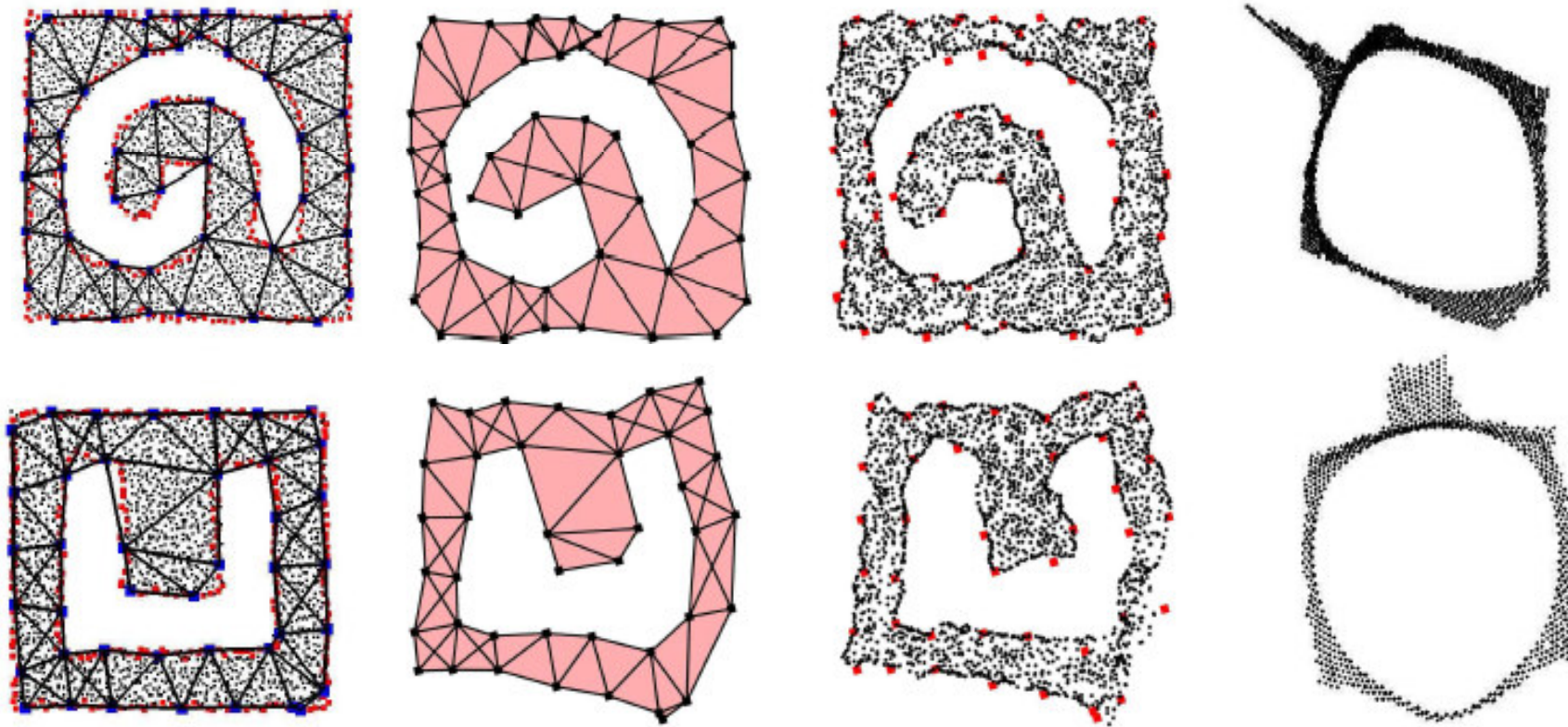
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Ground truth

Landmarks

All nodes

MDS



Unit disk graph model, Node average degree 6~10.

# Conclusion

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- Exploit the geometric and topological features of sensor networks.
- Simple algorithm, Surprising results.