



中国科学技术大学  
University of Science and Technology of China



GAMES 102在线课程

# 几何建模与处理基础

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GAMES 102在线课程：几何建模与处理基础

# 采样与剖分

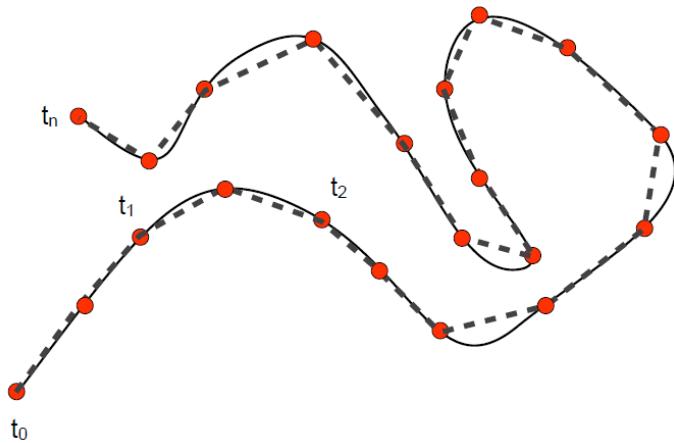
采样 (Sampling)

# 从连续到离散

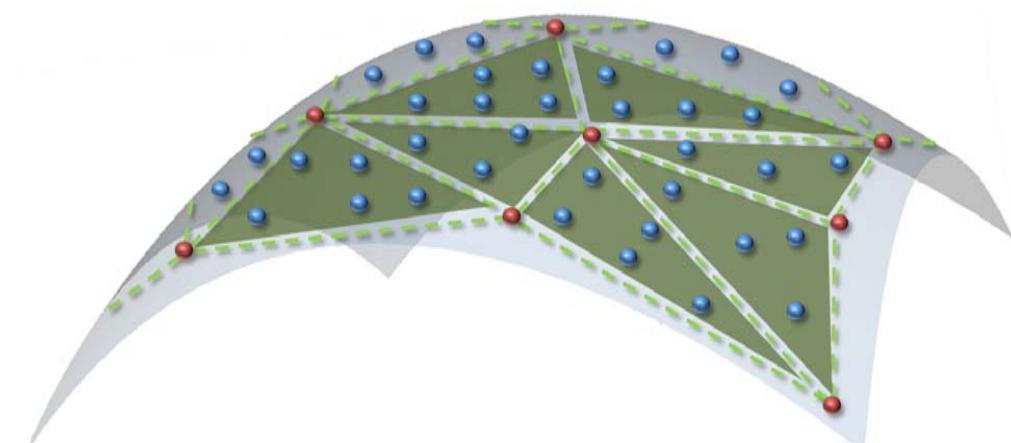
- 对象的表达
  - 在数学上，连续表达与计算
  - 在计算机中，离散表达与计算
- 数值方法：数值微分、数值积分、数值优化
  - 数值分析：离散计算对精确计算的近似程度
  - Fourier分析/变换：离散Fourier分析/变换
  - 卷积（滤波）
- 在计算机科学（计算机图形学）中，采样无处不在
  - 计算机只能表达离散的数值
  - 例子：int型的数据（量化）

# 曲线曲面的离散表达

- 曲线的绘制：
  - GDI/OpenGL 绘制基本单元：点、线段
  - 曲线须离散成多边形
- 曲面的绘制：
  - OpenGL 绘制基本单元：点、线、三角形
  - 曲面须离散成三角形网格



曲线的离散

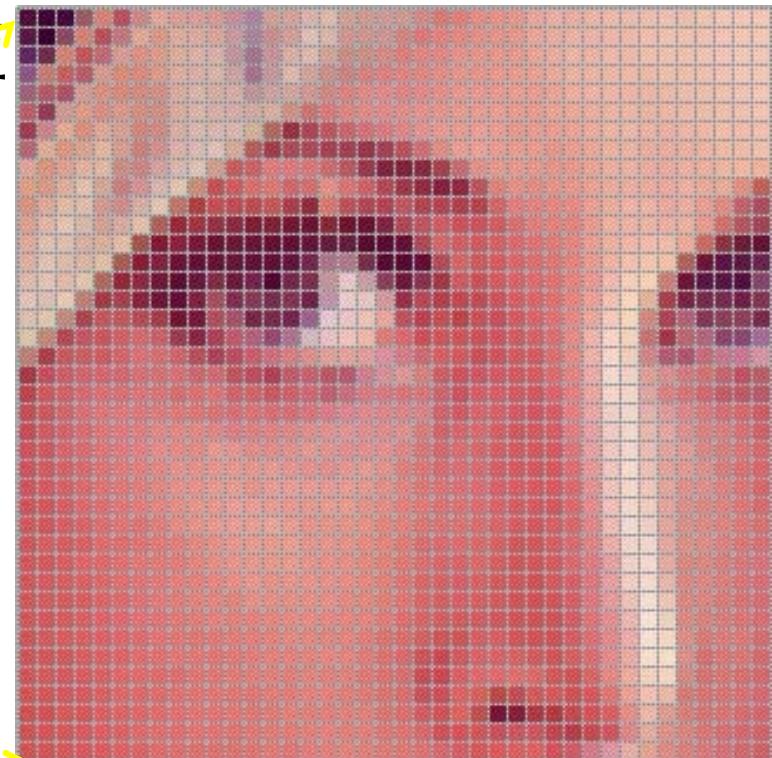
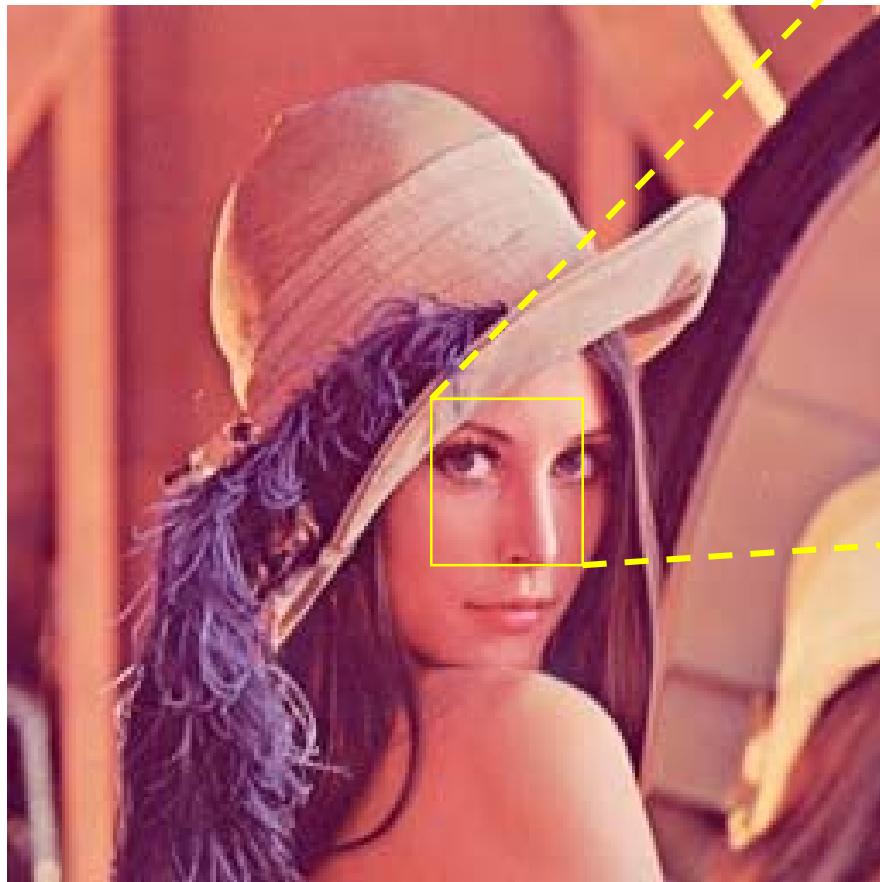


曲面的离散

# 离散的本质：采样 (Sampling)

- 曲线曲面的采样
  - 在参数域上采样
  - 直接在原始曲线曲面采样
- NURBS曲线曲面的采样误差估计
  - 可进行理论上的误差分析
- 逆向工程：
  - 采样点的获取
    - 通过扫描硬件设备得到采样点
    - 通过（多视点几何）重建算法计算得到采样点
  - 重建问题：如何通过采样点重构原始曲线/曲面
    - 连续重建：用连续函数来拟合表达
    - 离散重建：直接得到离散基元表达

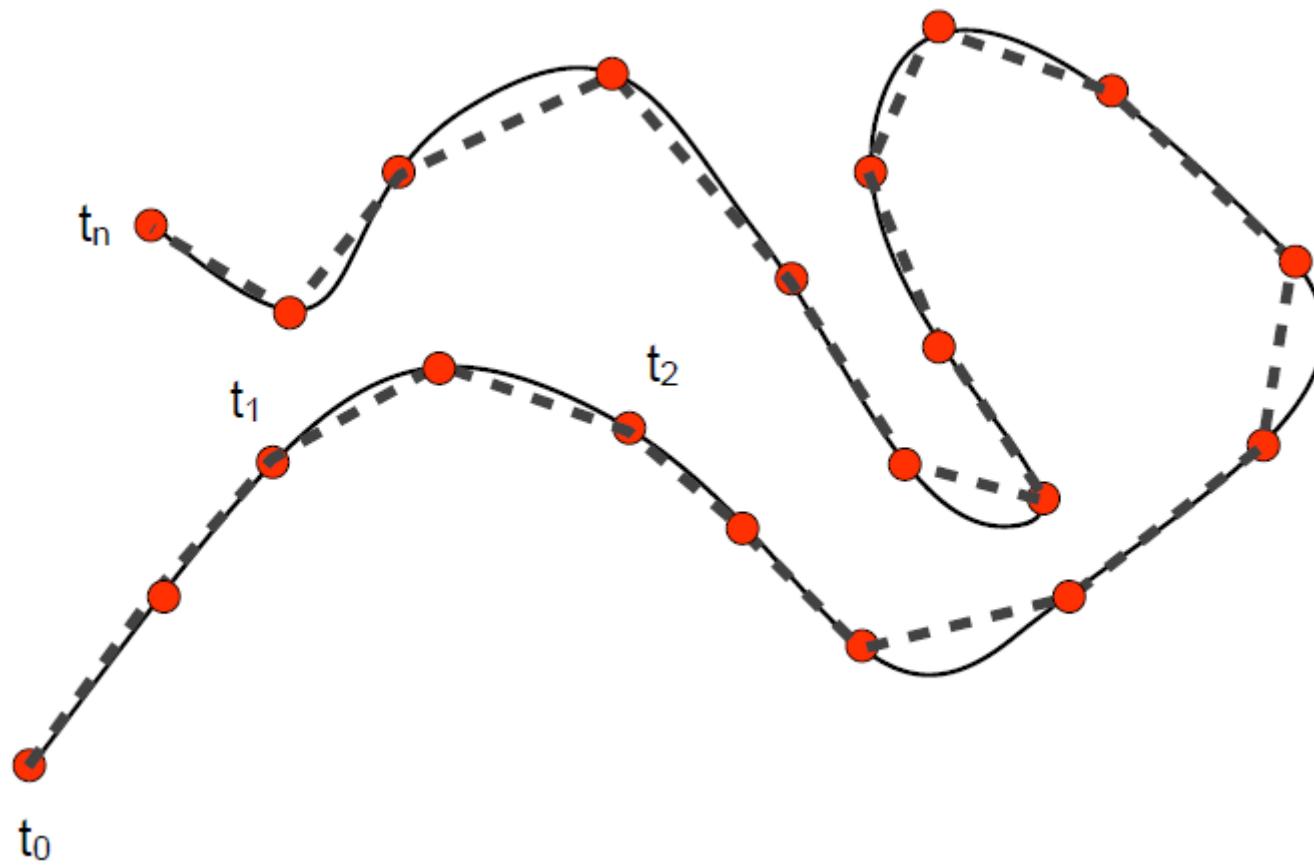
图像：区域的采样



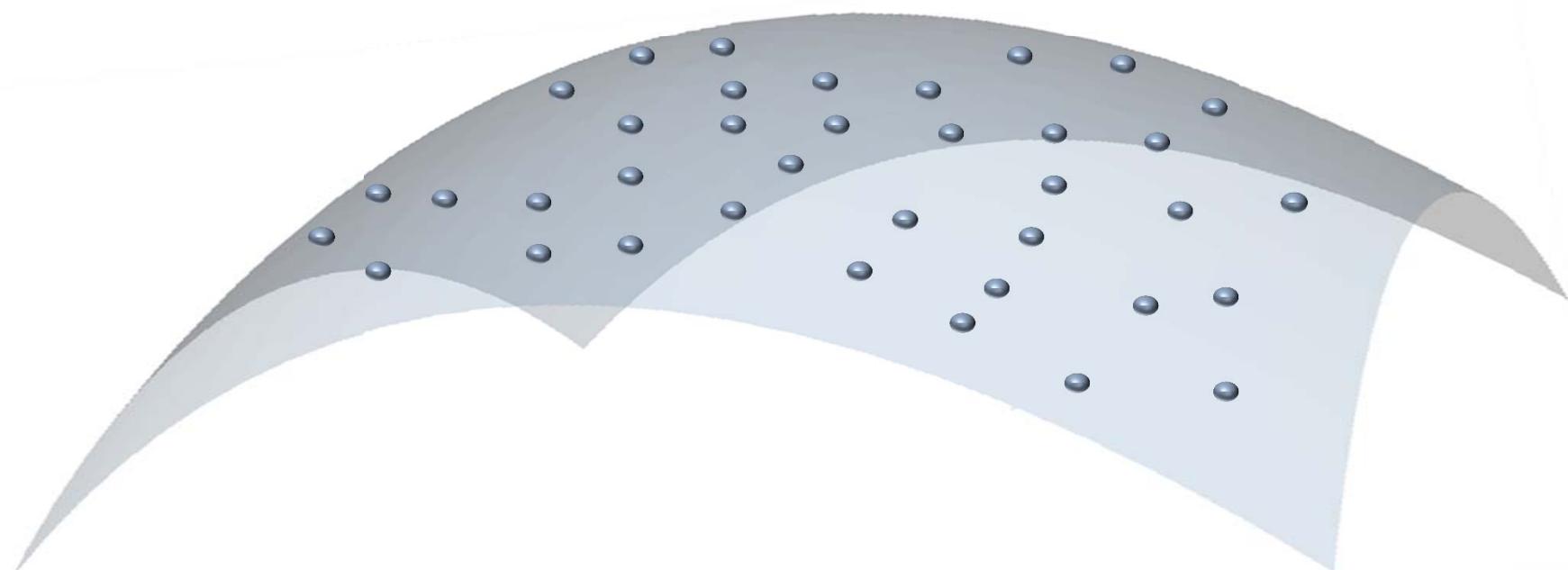
# 视频：时间的采样



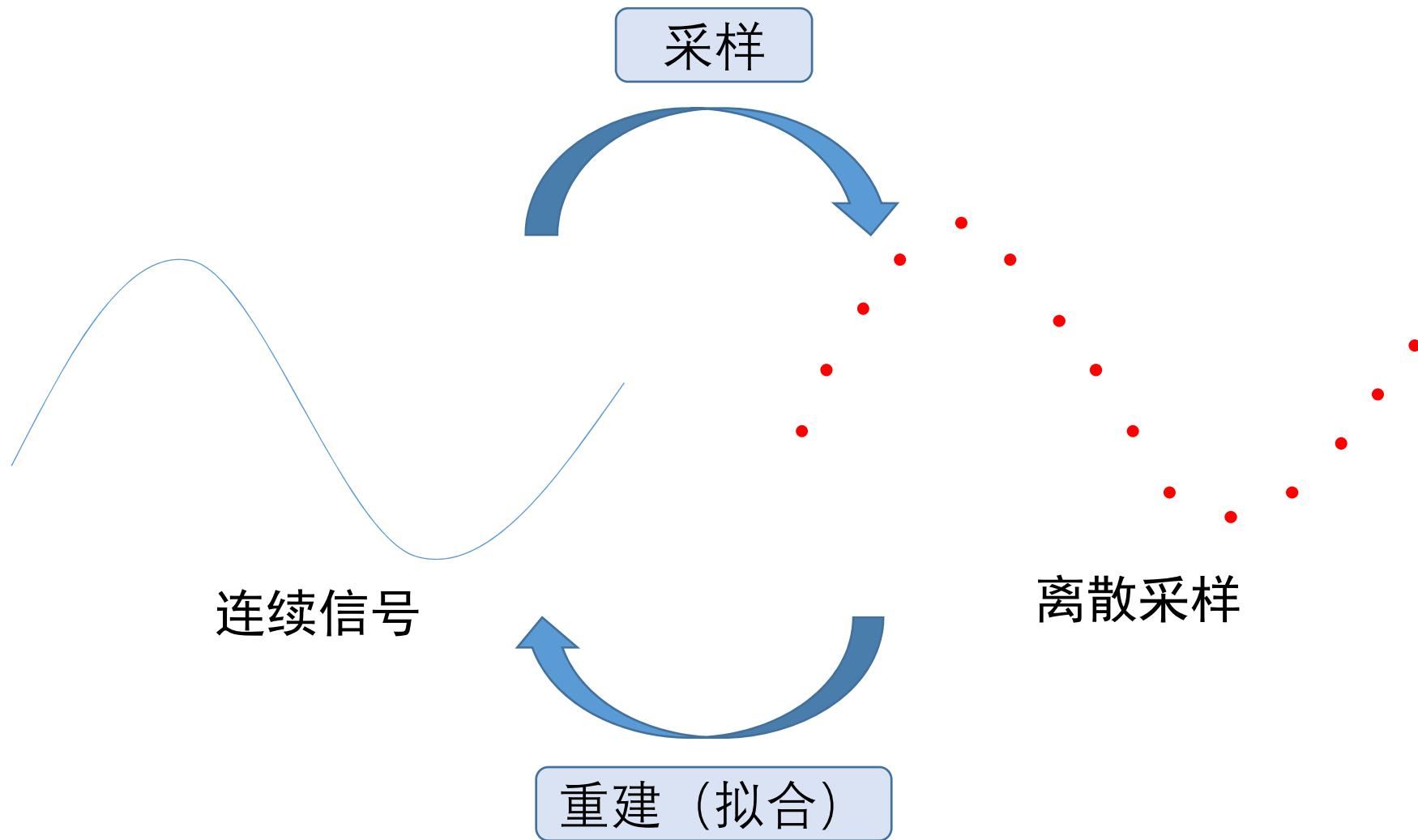
# 曲线的采样



# 曲面的采样



# 采样与重建



# Sampling Theorem

- Nyquist–Shannon sampling theorem

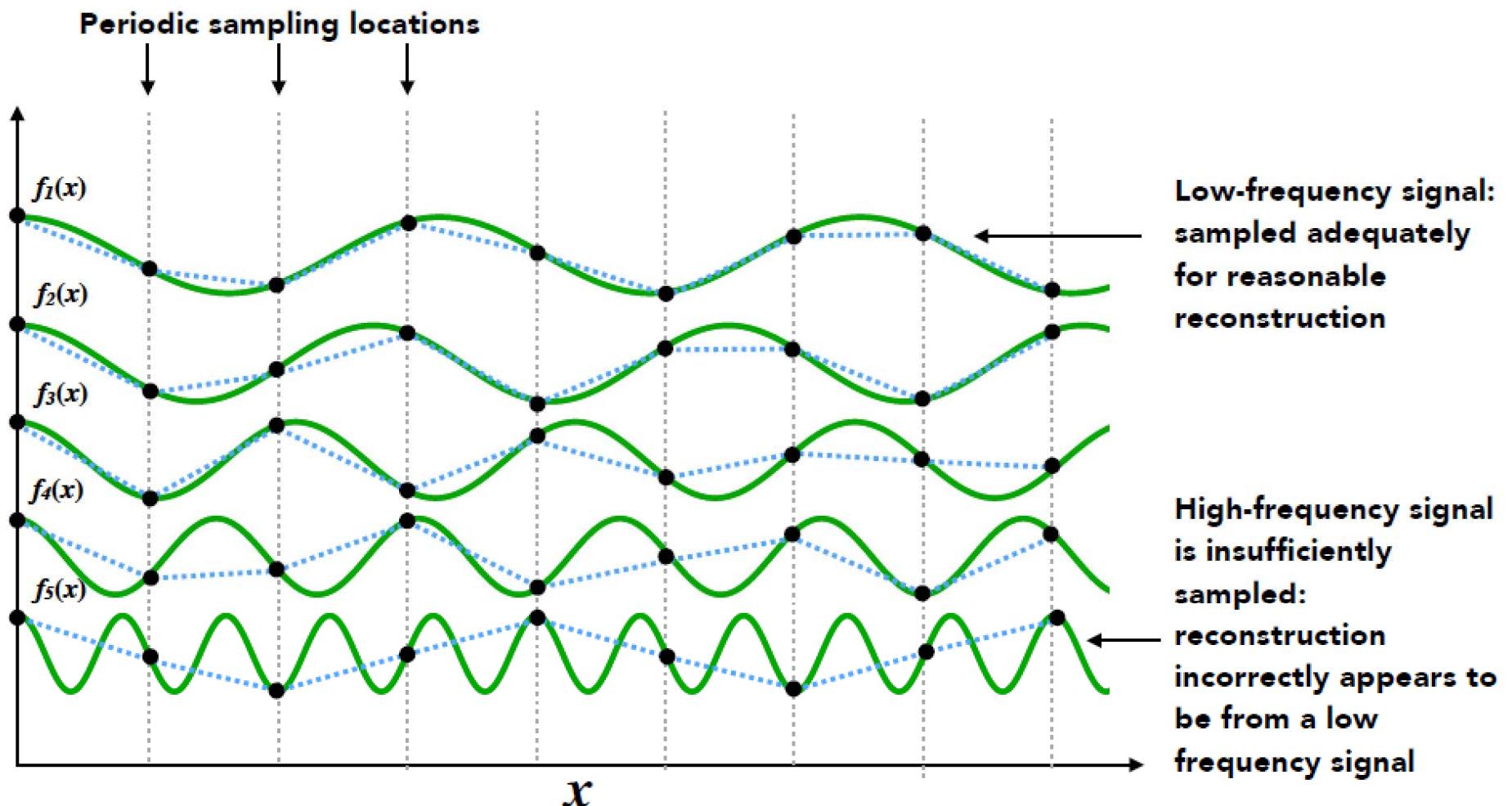
*If a function  $x(t)$  contains no frequencies higher than  $B$  hertz, it is completely determined by giving its ordinates at a series of points spaced  $1/(2B)$  seconds apart.*



Generally, **a amount of samples** are required to **recover** complex signal

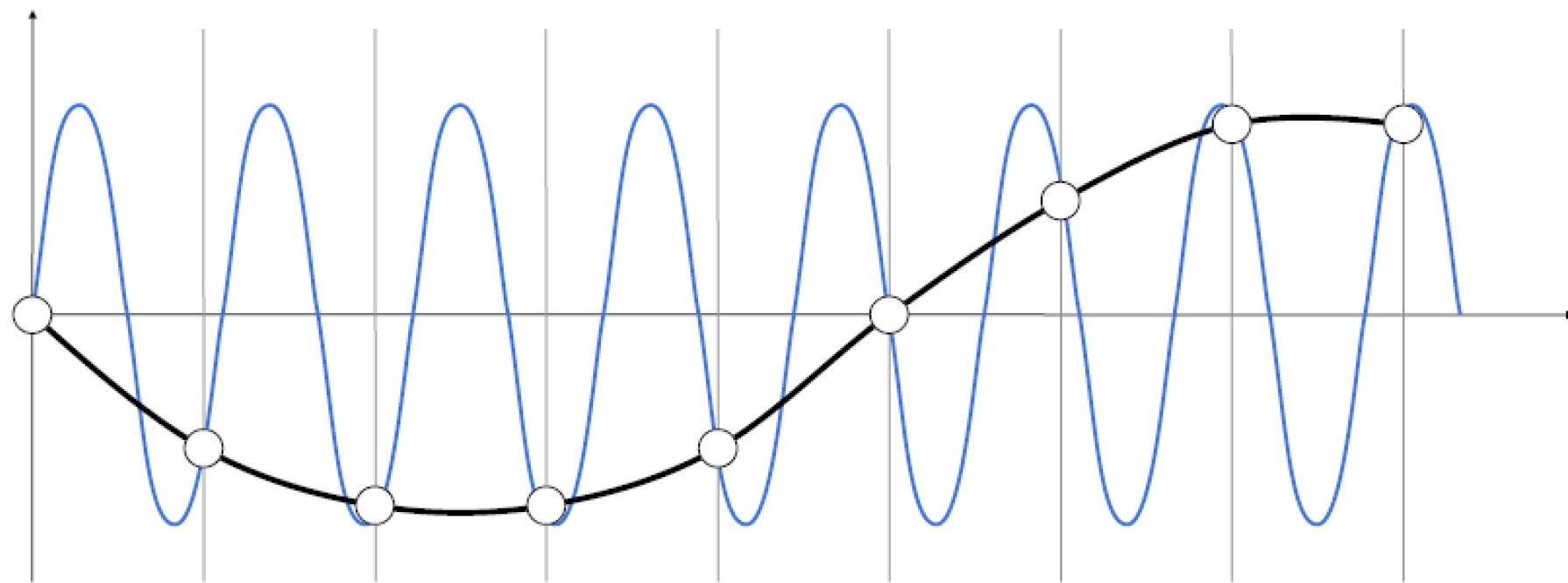
# 采样与信号频率

Fourier Analysis

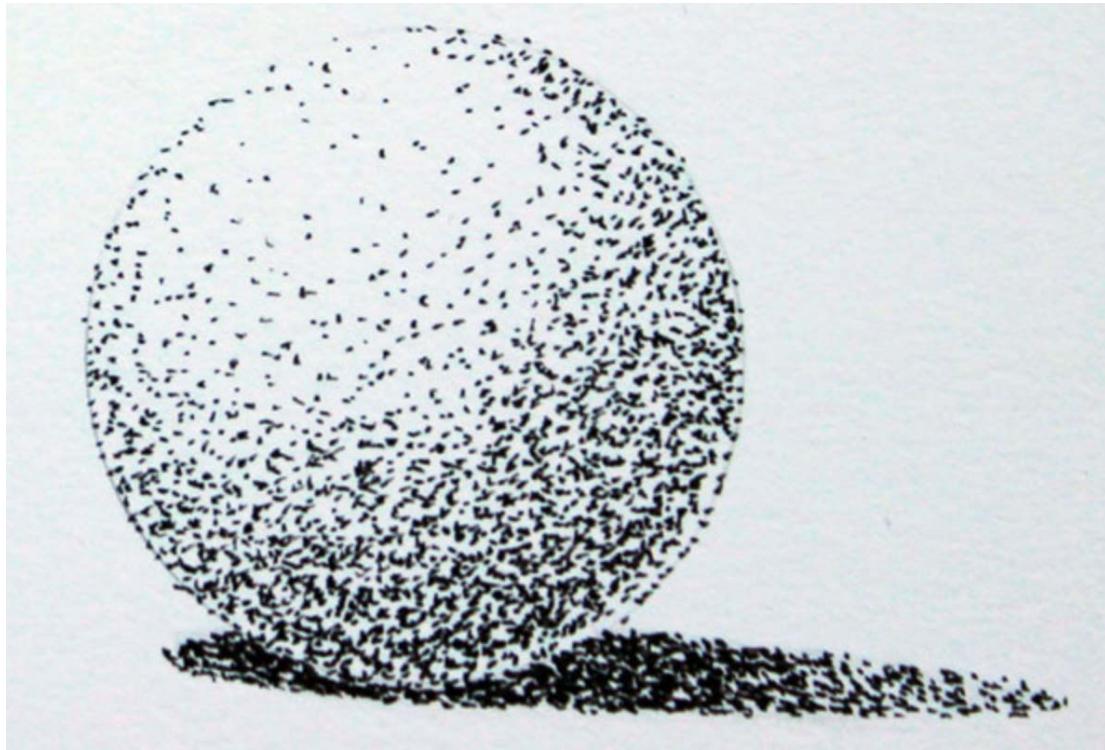


# 欠采样产生频率的走样

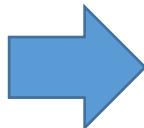
- 高频函数拟合低频信号：过拟合
- 低频函数拟合高频信号：欠拟合



# Stippling



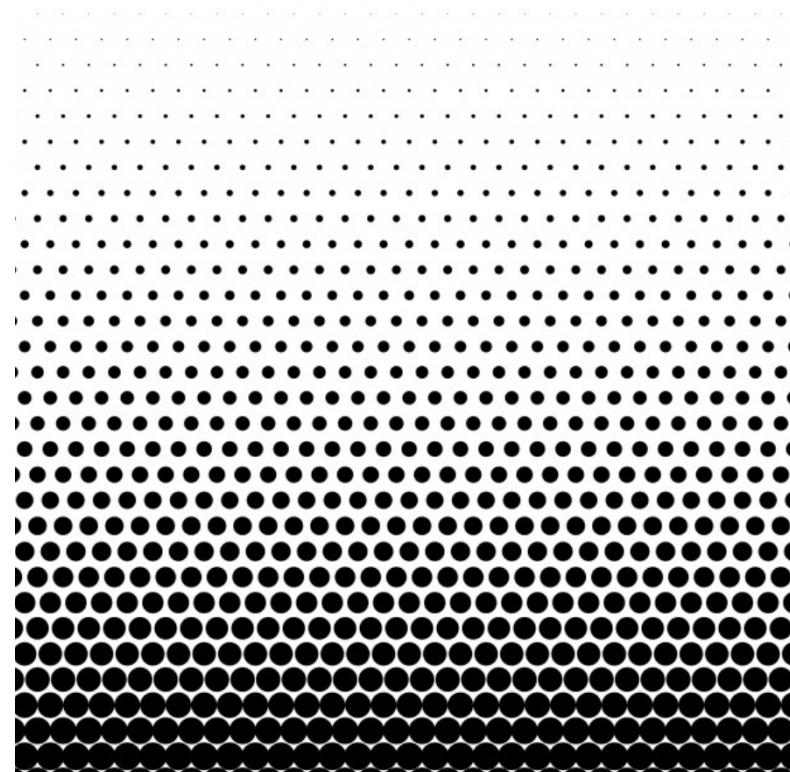
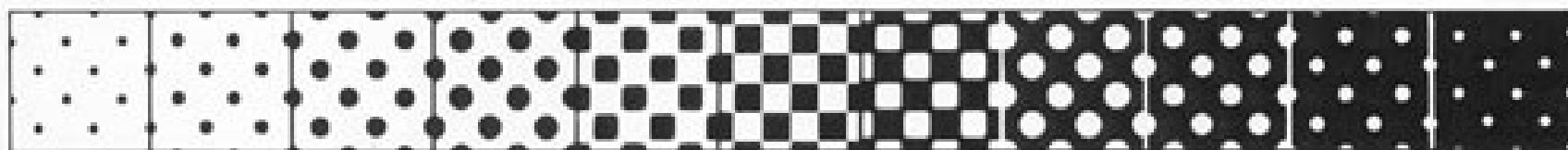
# Stippling



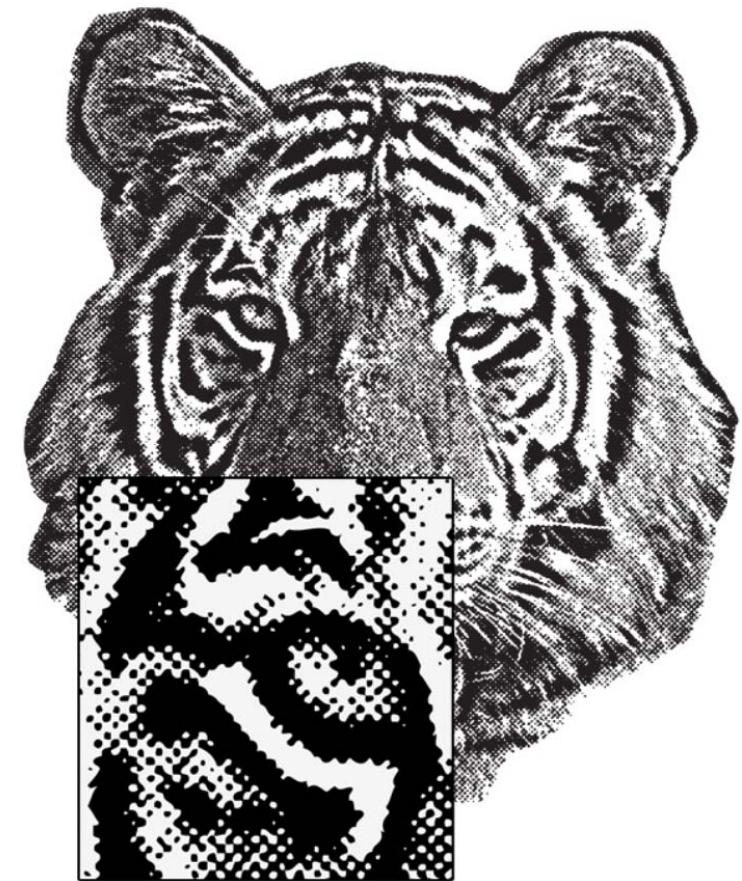
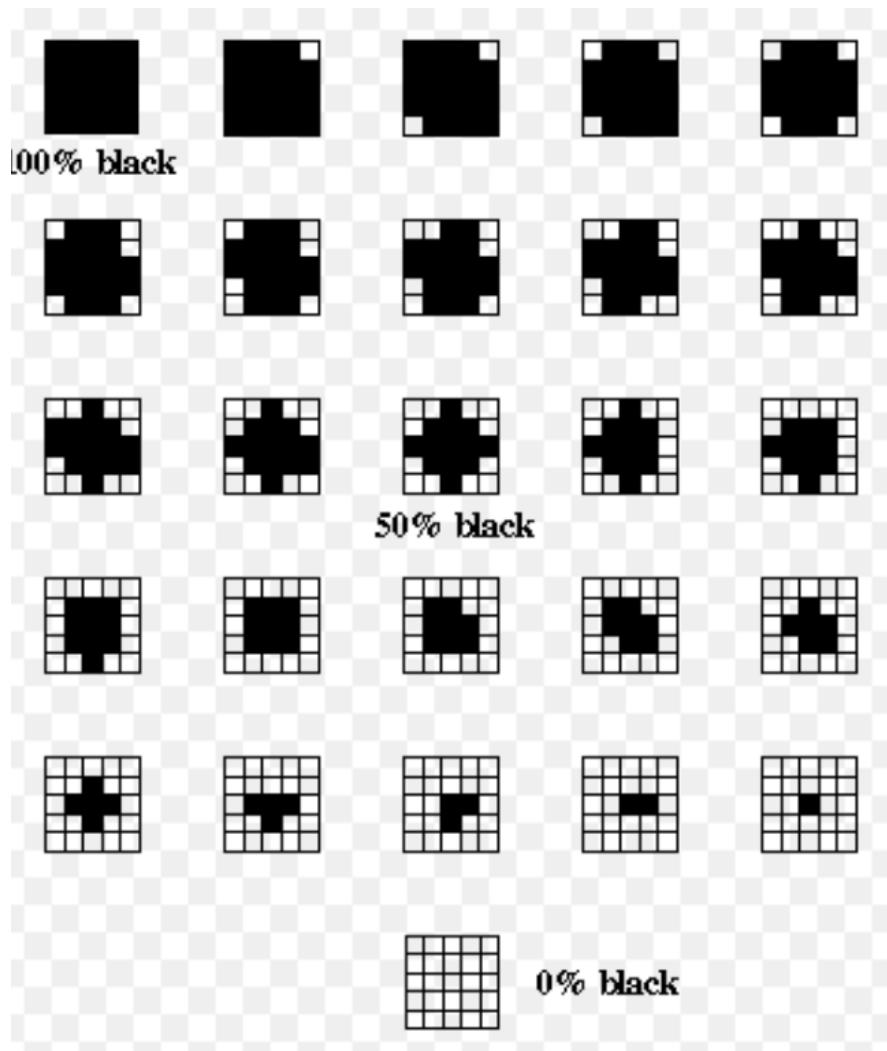
# Half-tone (dithering)

Enlarged view of grayscale percentages.

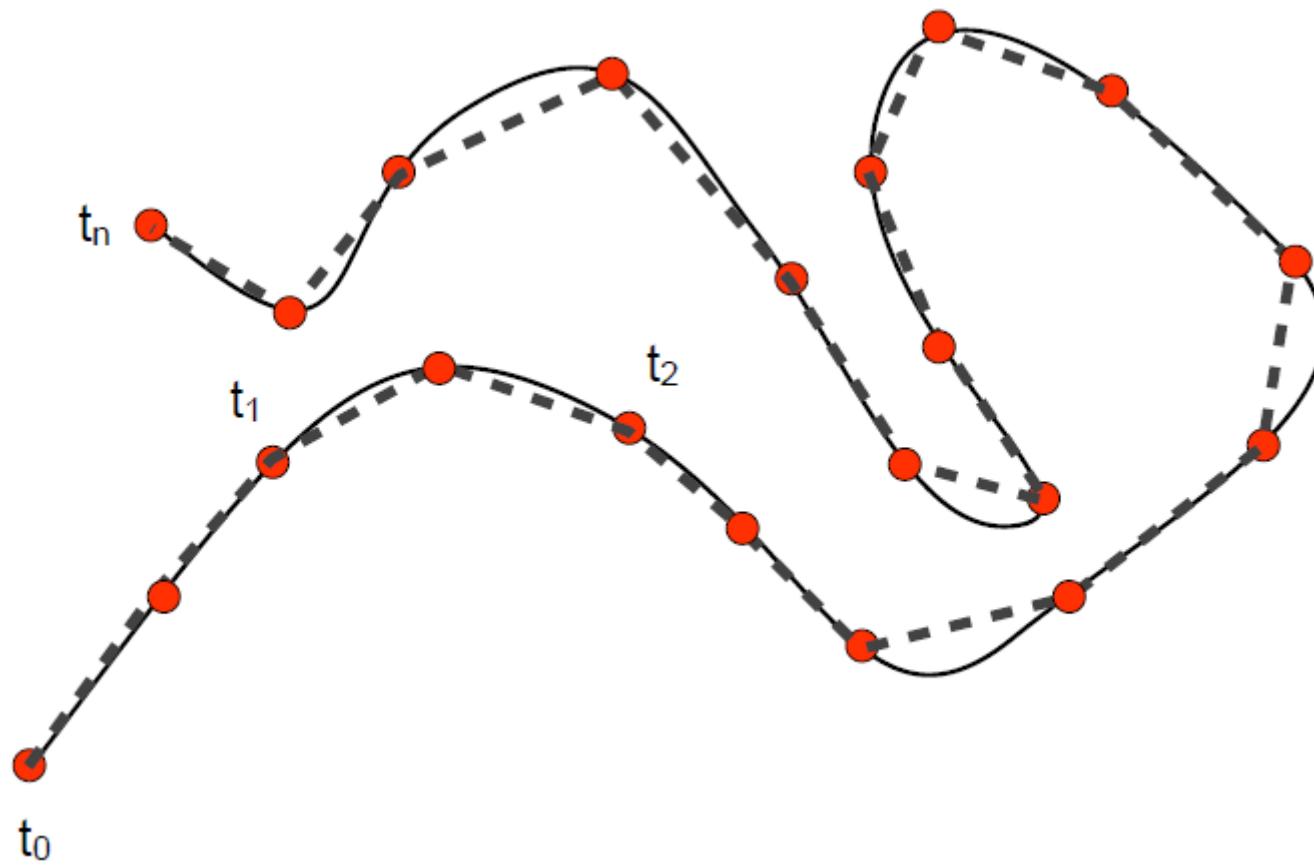
5%    10    20    30    40    50    60    70    80    90    95



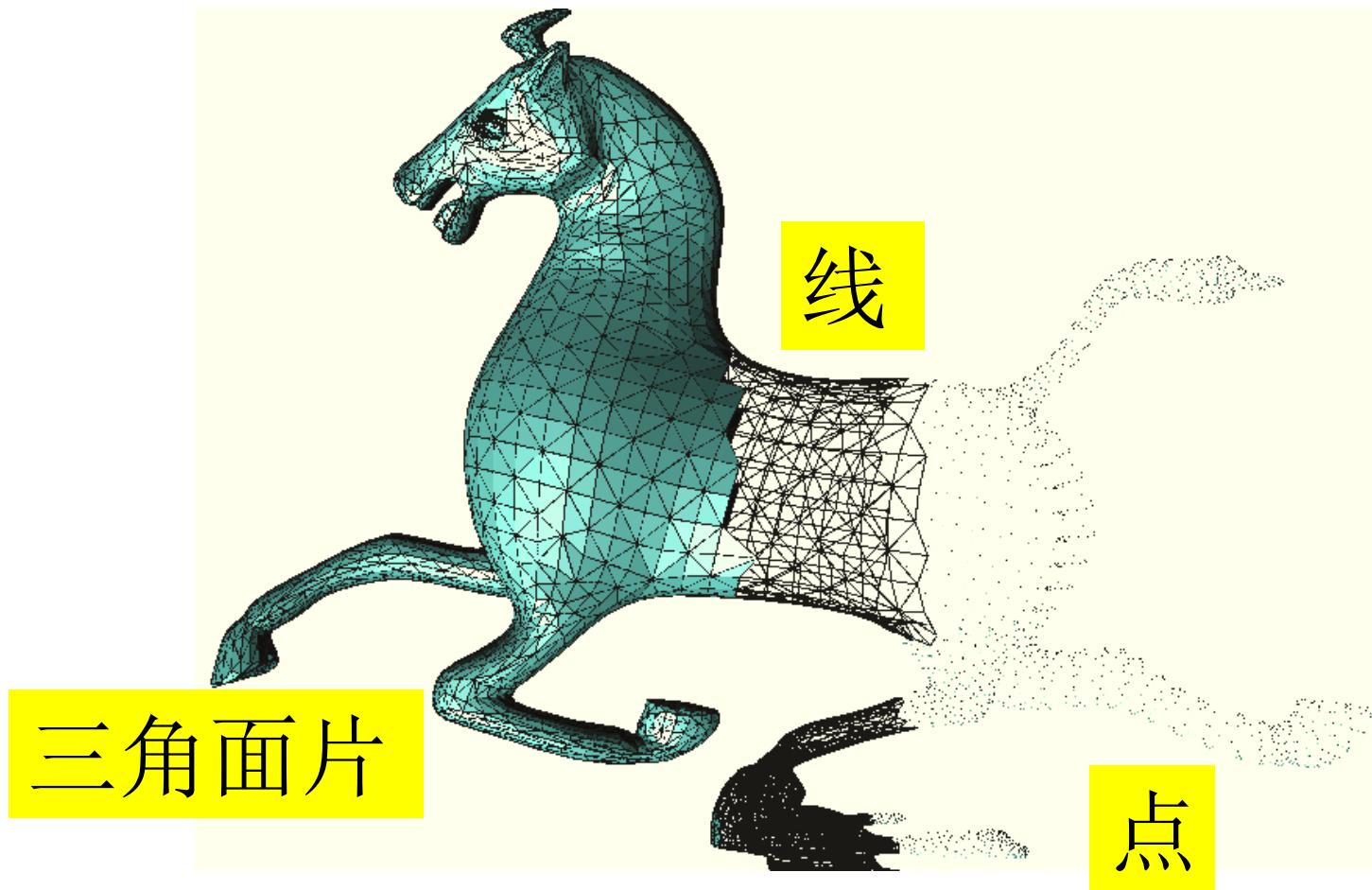
# Halftoning Printing



# 1D曲线的采样：分段线性逼近表达

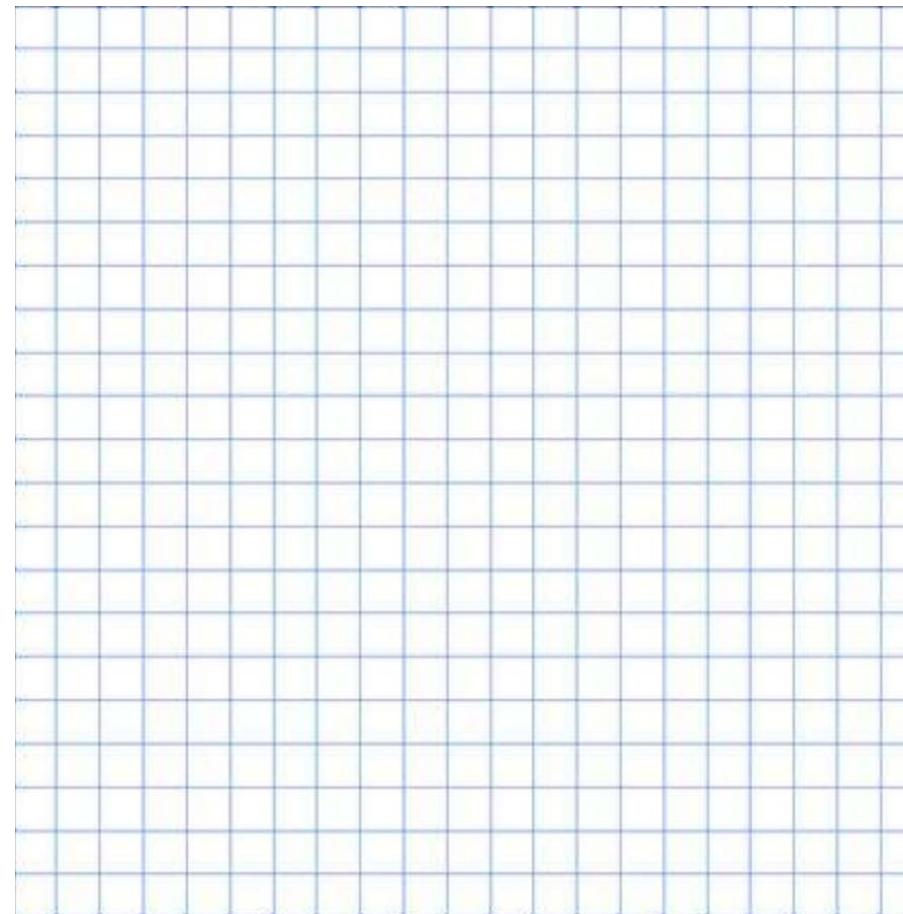


# 2D曲面的采样：分片线性逼近表达

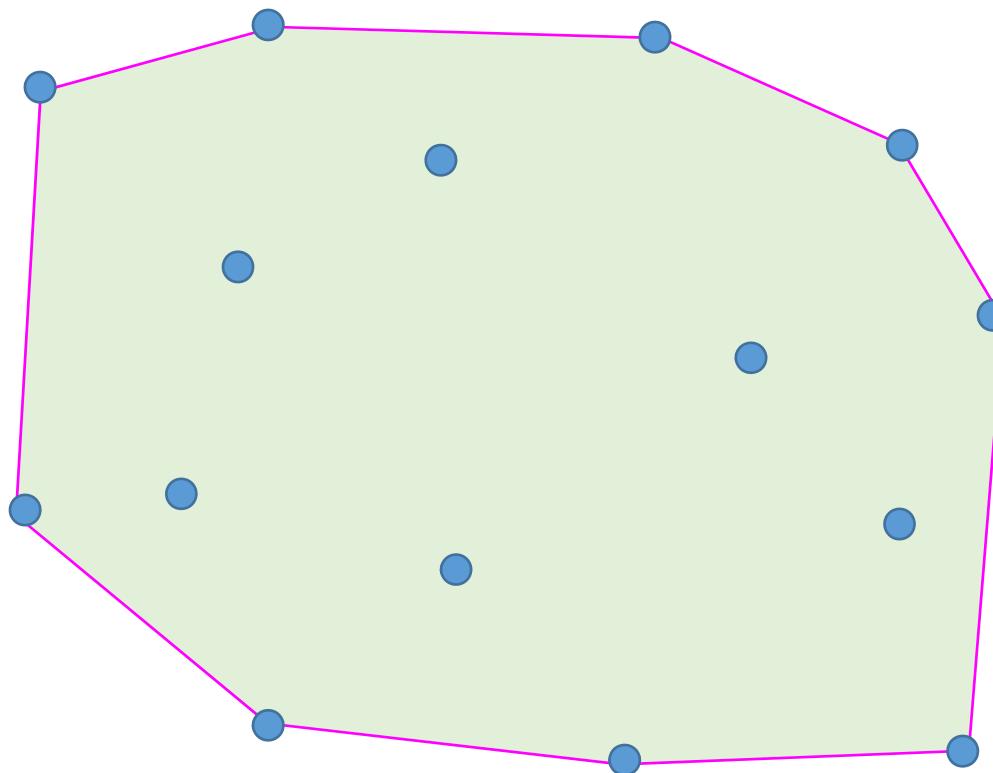


# 平面规则区域的采样

- 将一个区域分解为若干个小区域

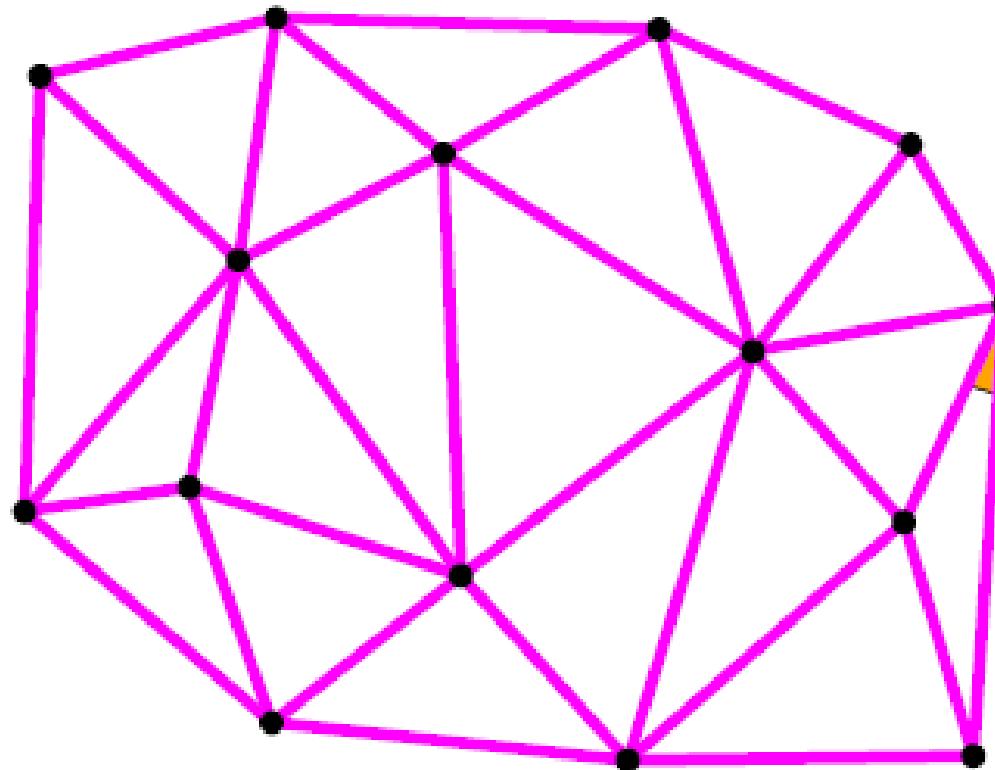


# 平面区域的不规则采样



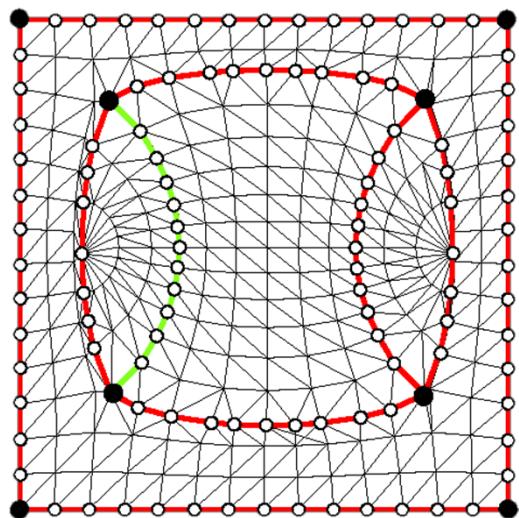
# 平面区域的不规则采样

- 将一个区域分解为若干个小区域

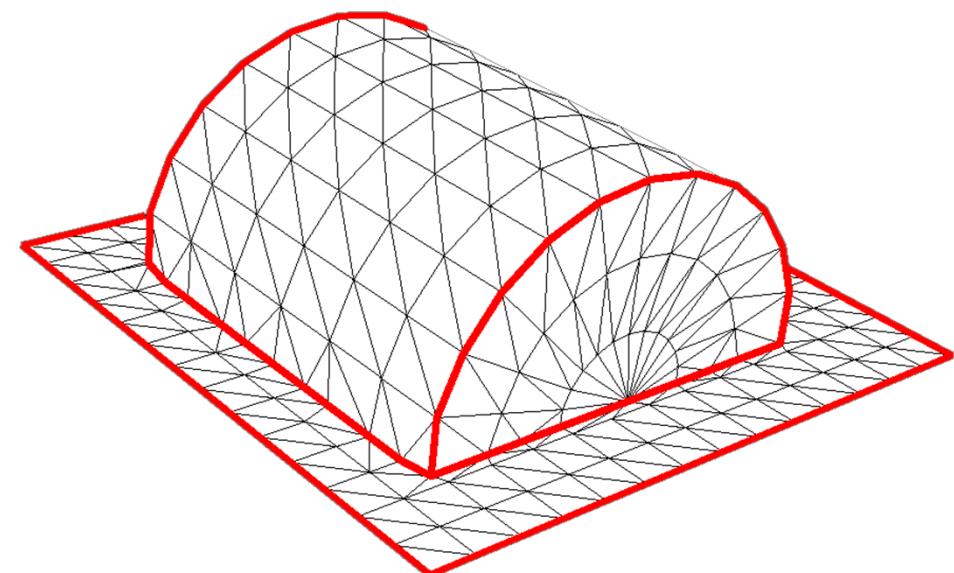


# Triangulation

- 复杂函数的分片线性逼近 (piece-wise linear approximation)



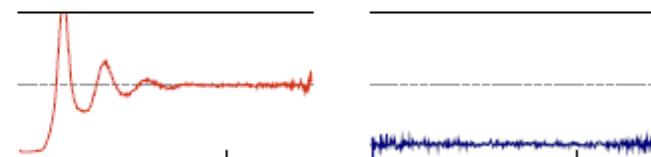
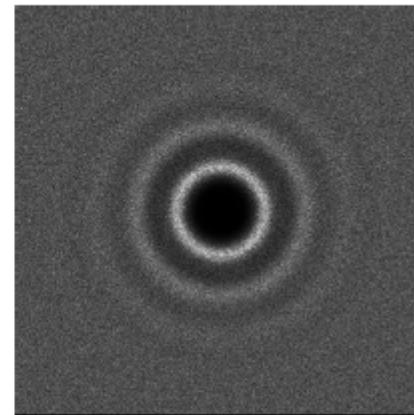
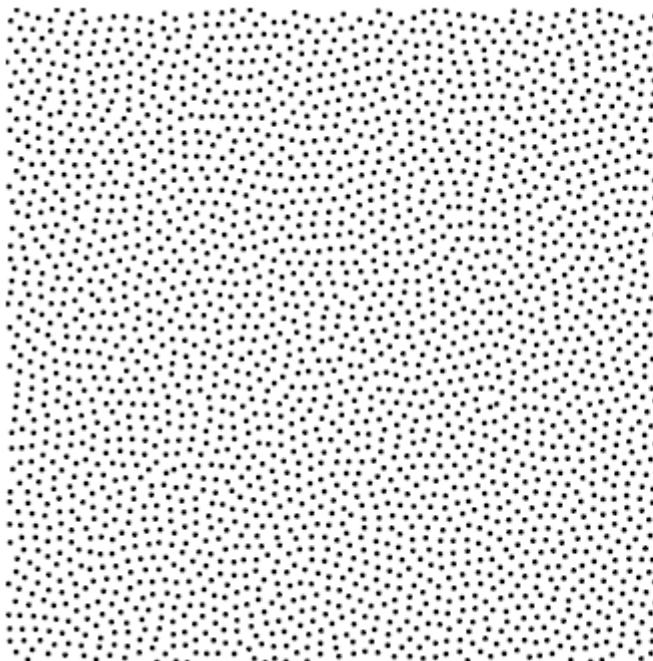
2D



3D

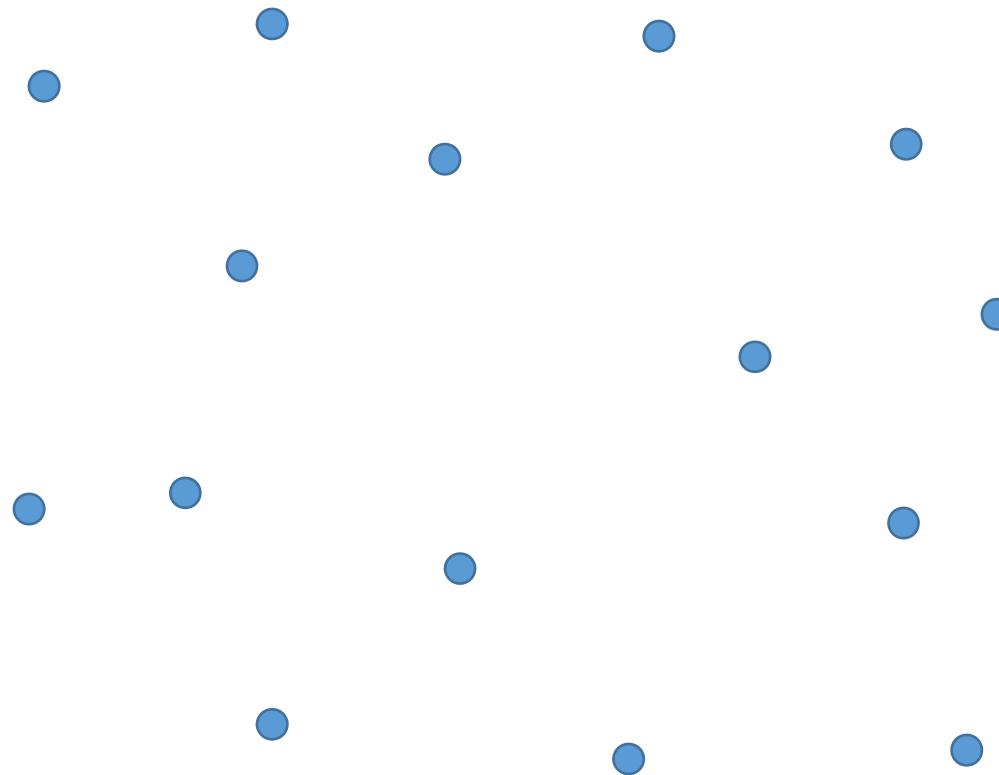
# Blue Noise Sampling

- (Left) A uniformly distributed yet randomly located point set
- (Right) The typical power spectrum, radially averaged power spectrum and anisotropy of blue noise distributions.

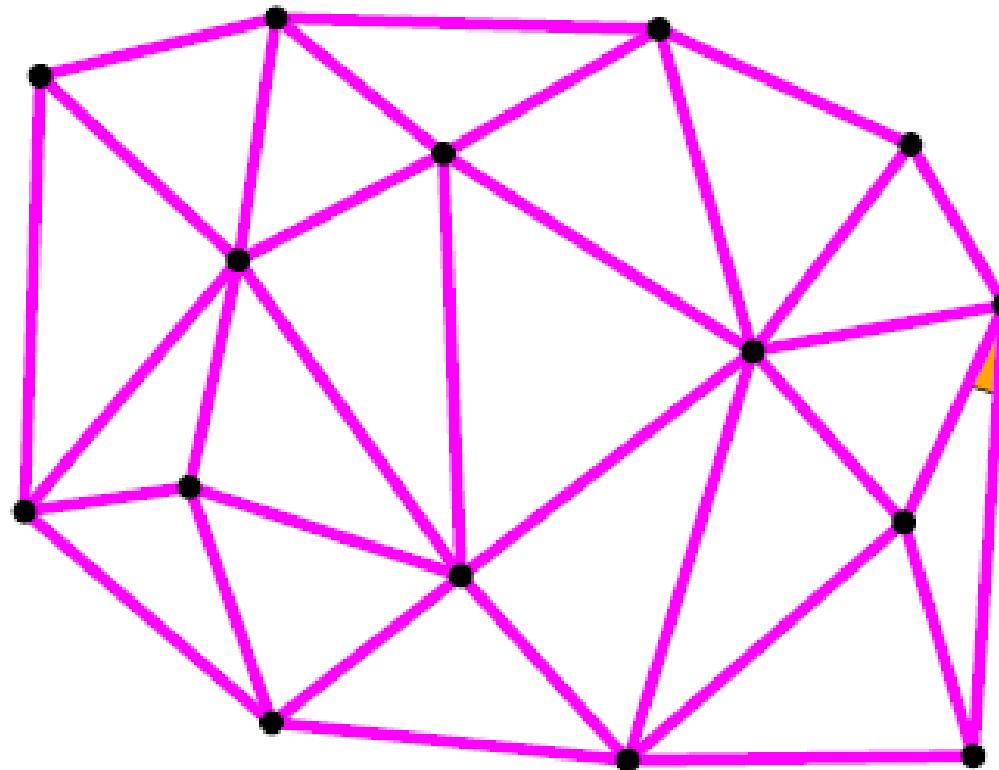


平面三角网格

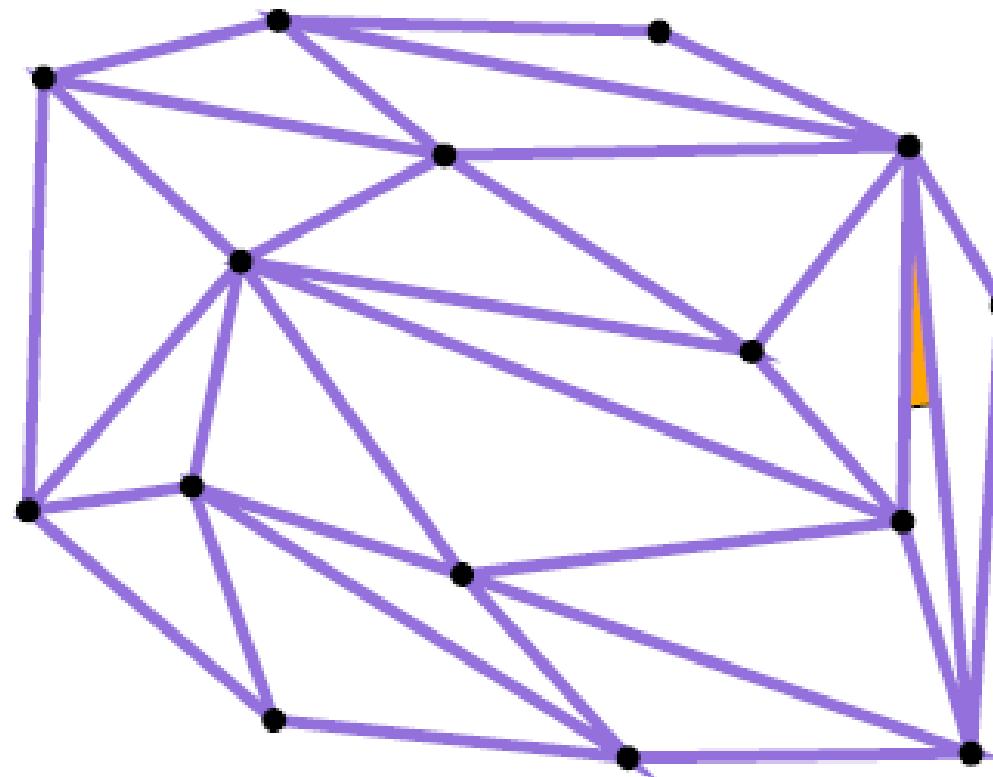
给定平面上一些点，  
如何生成三角剖分？



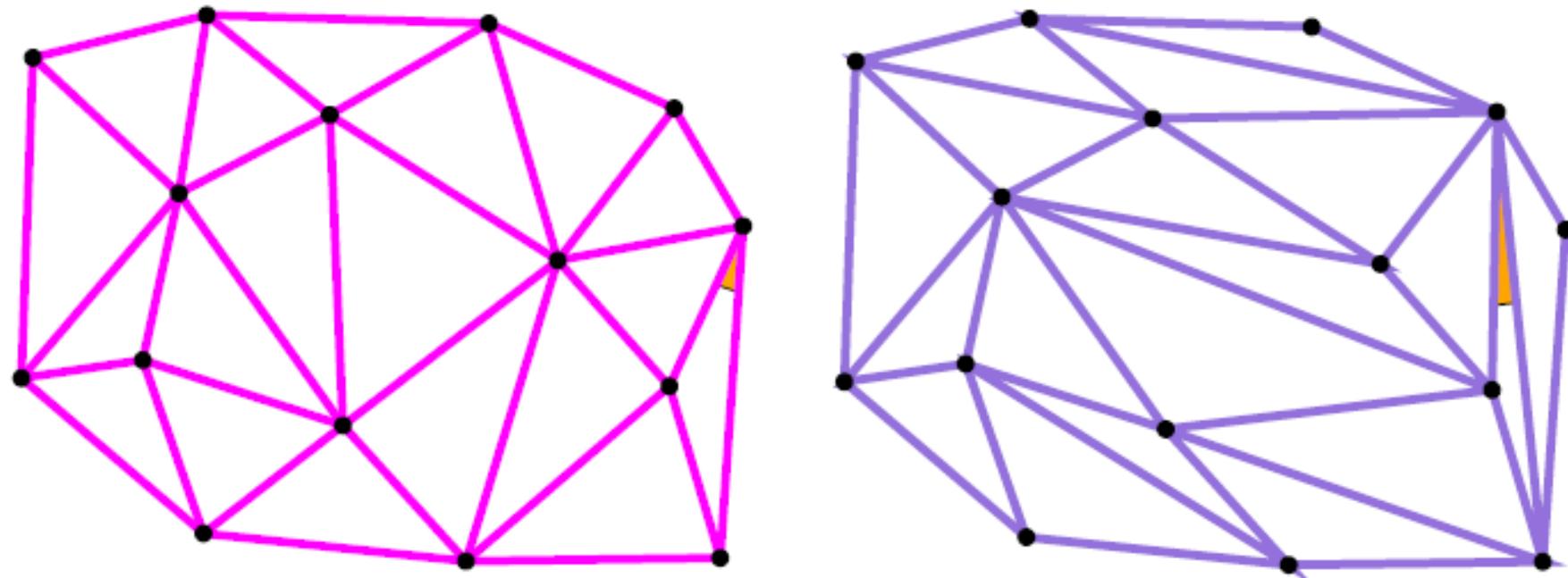
给定平面上一些点，  
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给定平面上一些点，  
如何生成三角剖分？



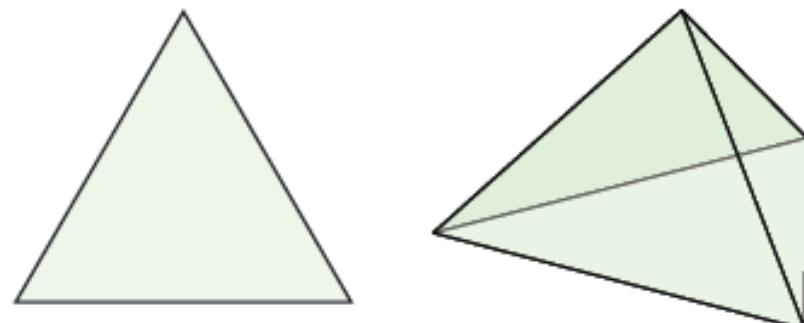
给定平面上一些点，如何生成  
**比较好的**三角剖分？



有无“最好”的三角剖分？

# Mesh Quality

- What do we mean a “good” mesh/simplex (triangle)?
  - Minimal angle
  - Mean ratio
  - Aspect/radius ratio
  - **Singular values**
  - ...
- It is not easy to define a universal mesh quality acceptable by everyone. But everyone agrees on the "best" simplex: equilateral triangle and tetrahedra.



Delaunay三角化

# Boris N. Delaunay



- Russian mathematician
- March 15, 1890 - July 17, 1980
- Introduce Delaunay triangulation in 1934

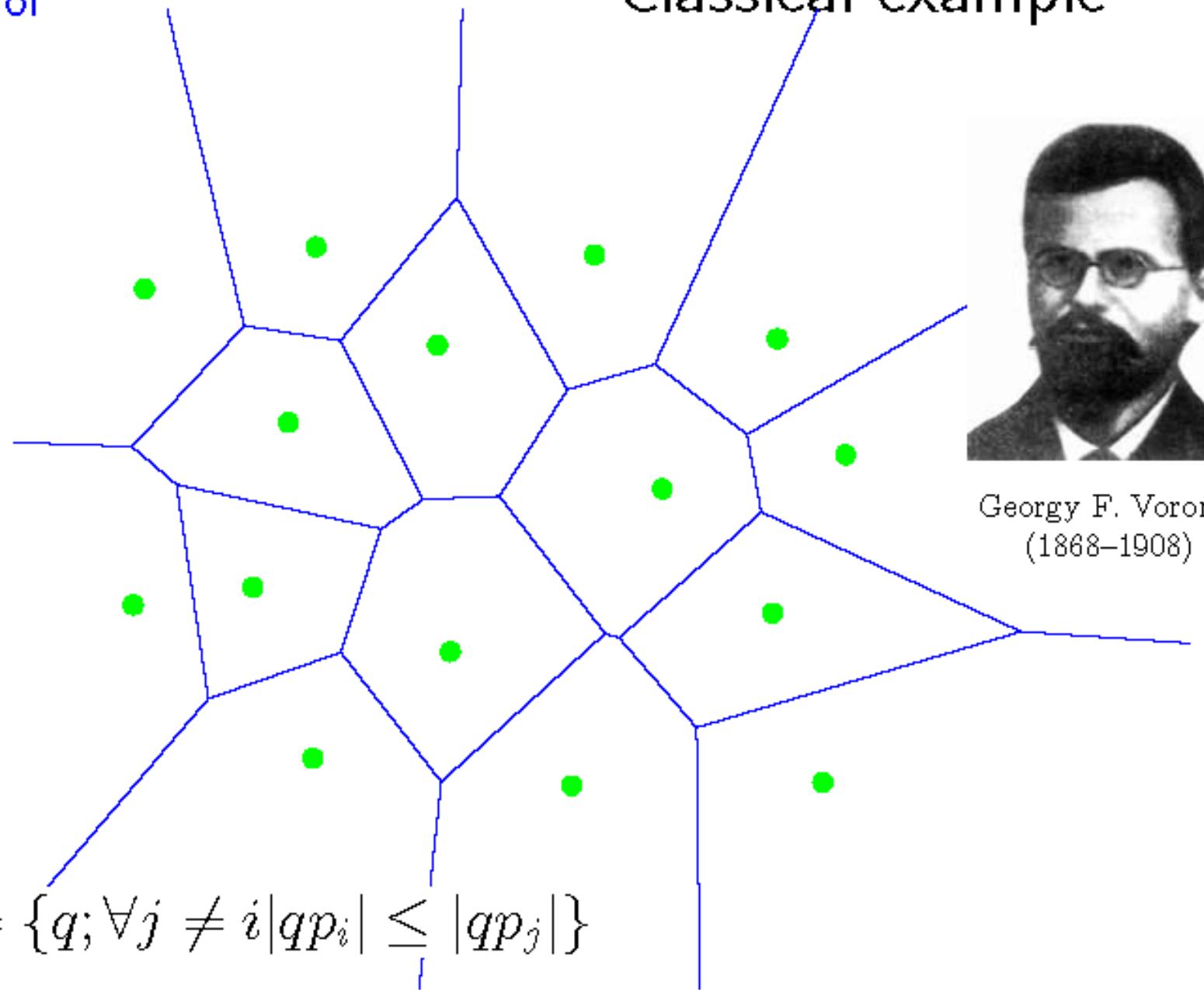
# Georgy F. Voronoi



- Russian mathematician
- April 28, 1868 - November 20, 1908

Voronoi

Classical example

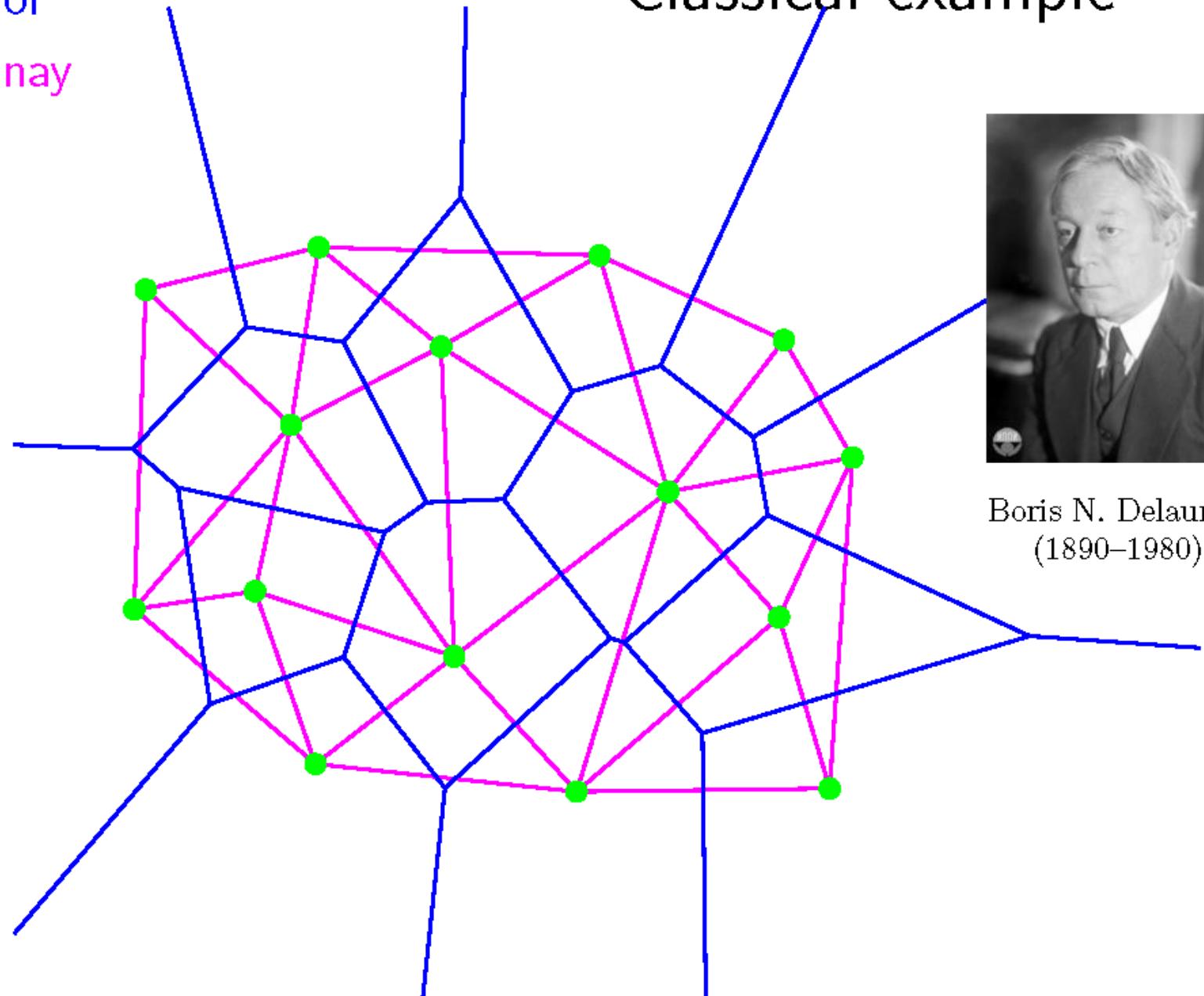


Voronoi  
Delaunay

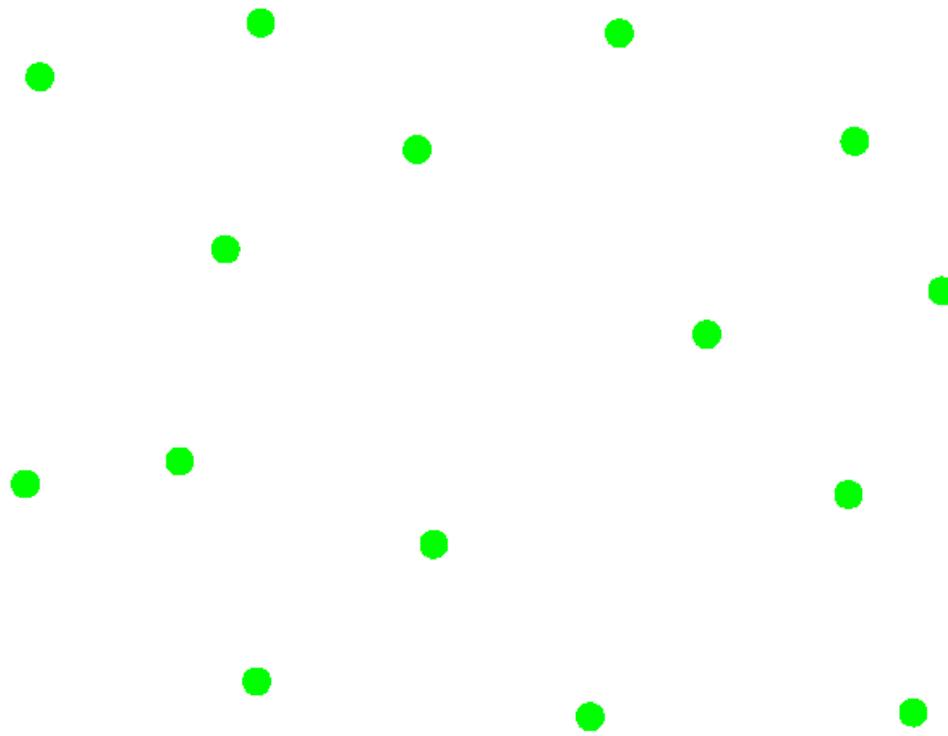
# Classical example



Boris N. Delaunay  
(1890–1980)

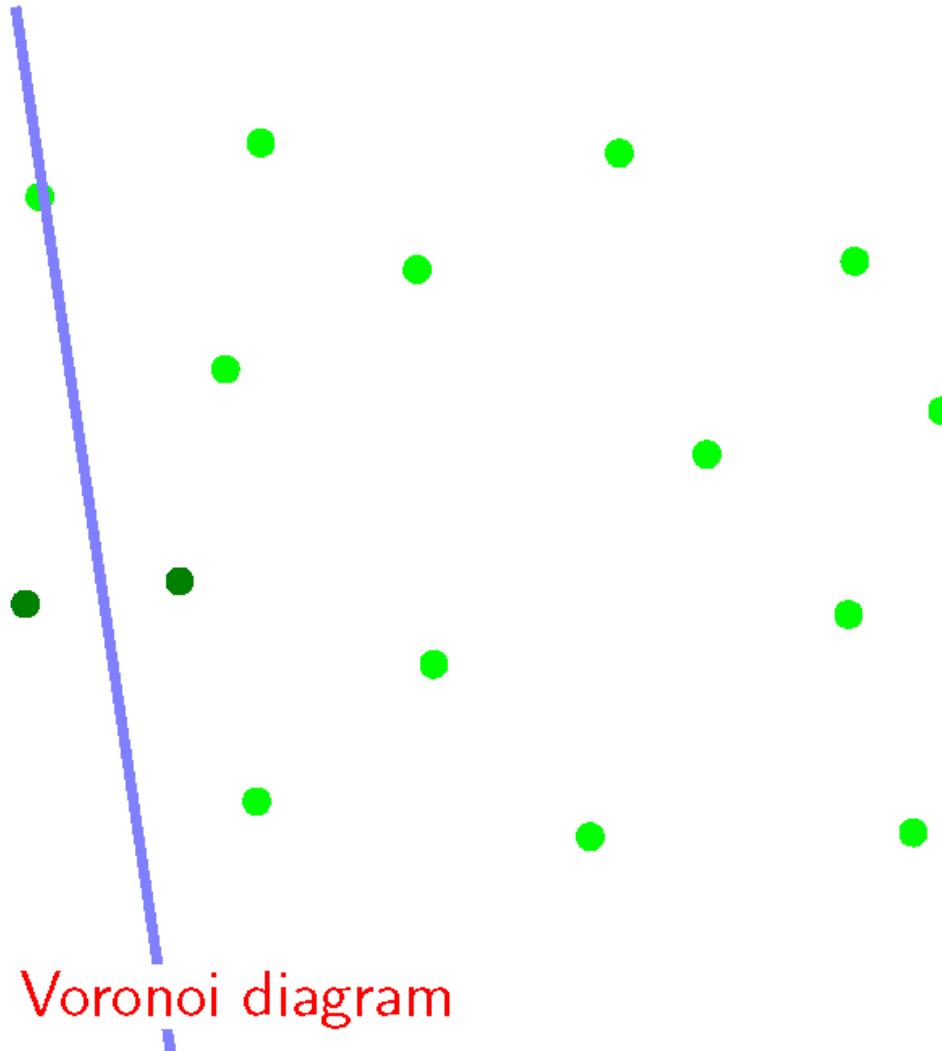


Voronoi



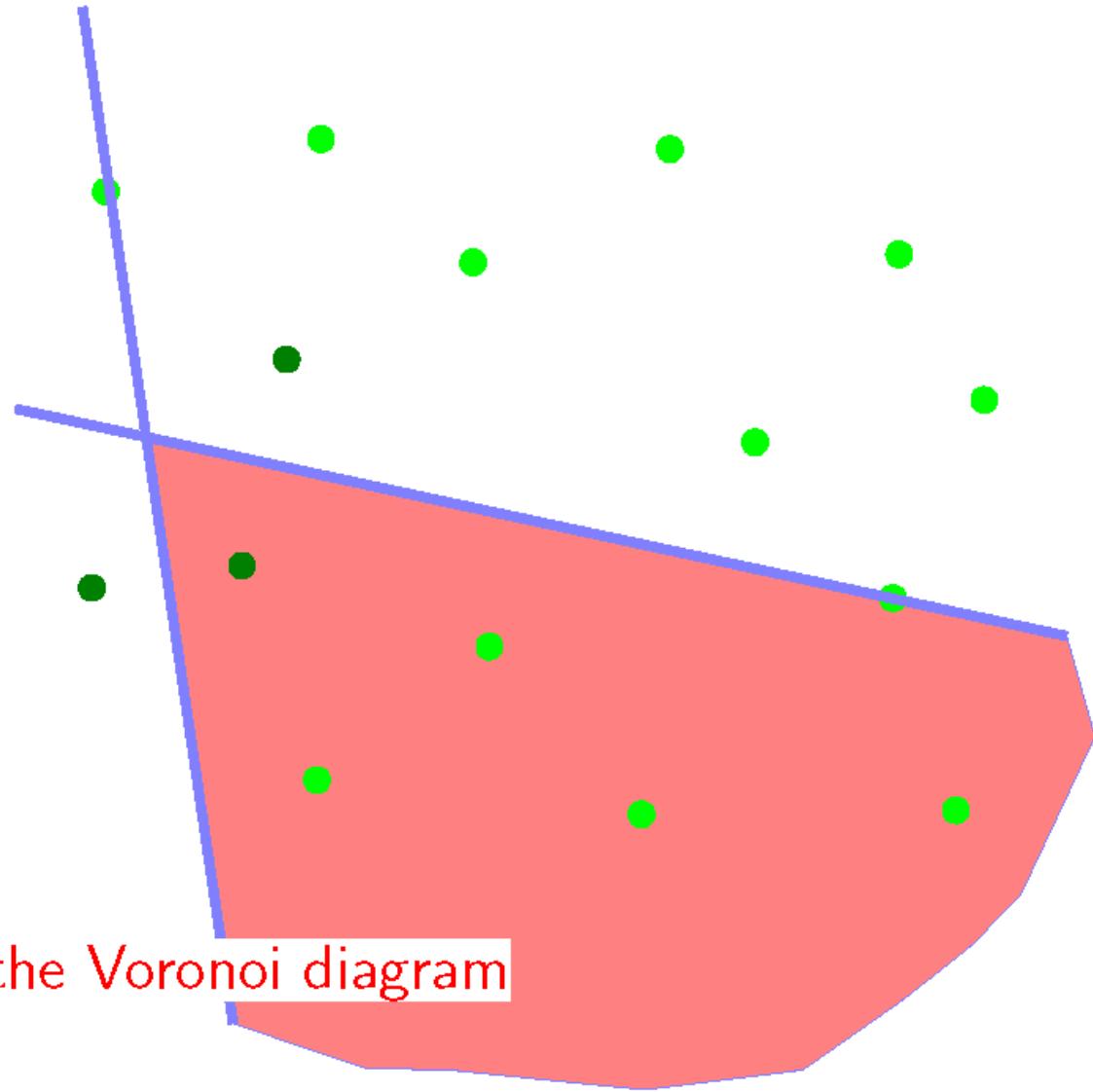
faces of the Voronoi diagram

Voronoi

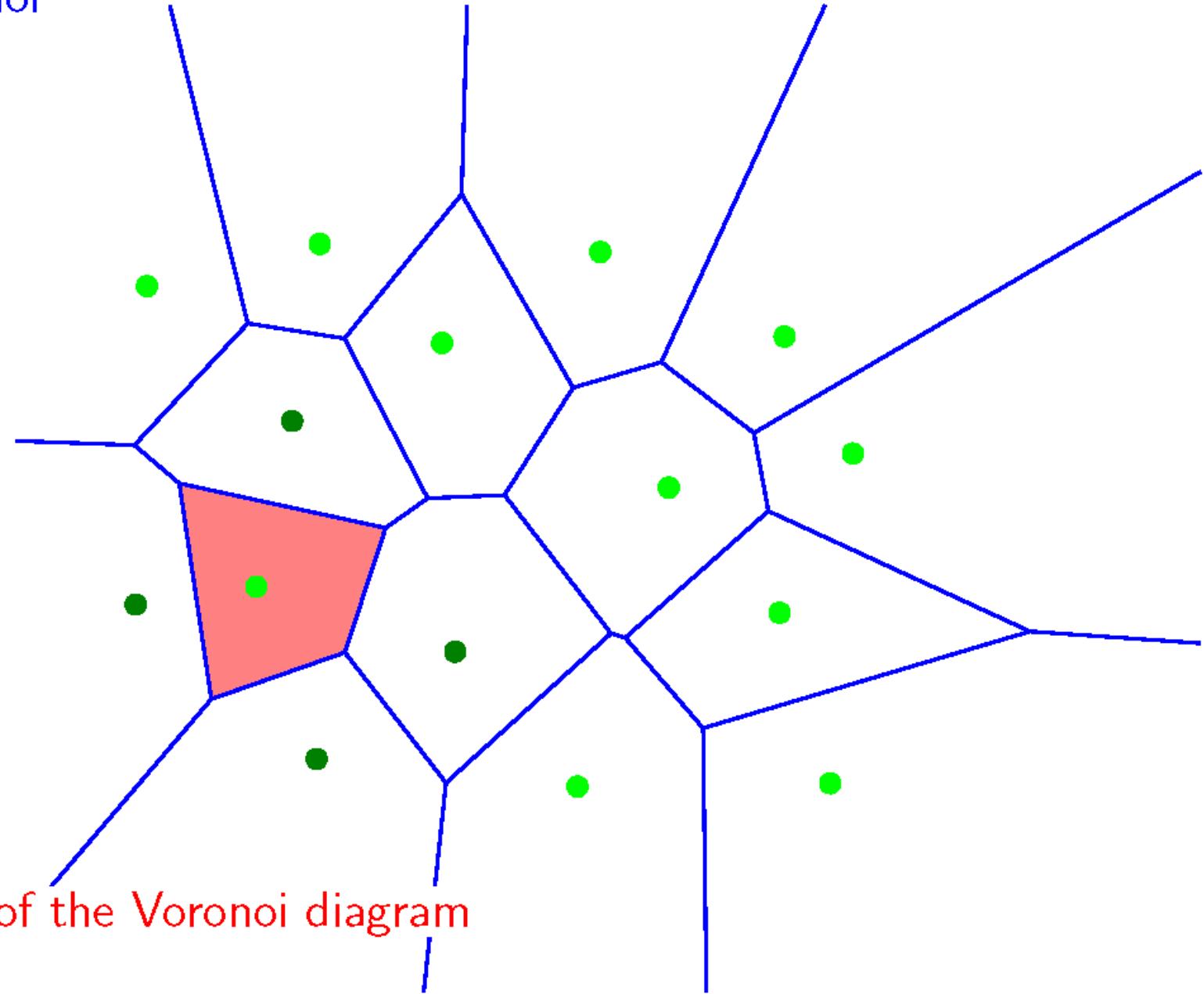


faces of the Voronoi diagram

Voronoi



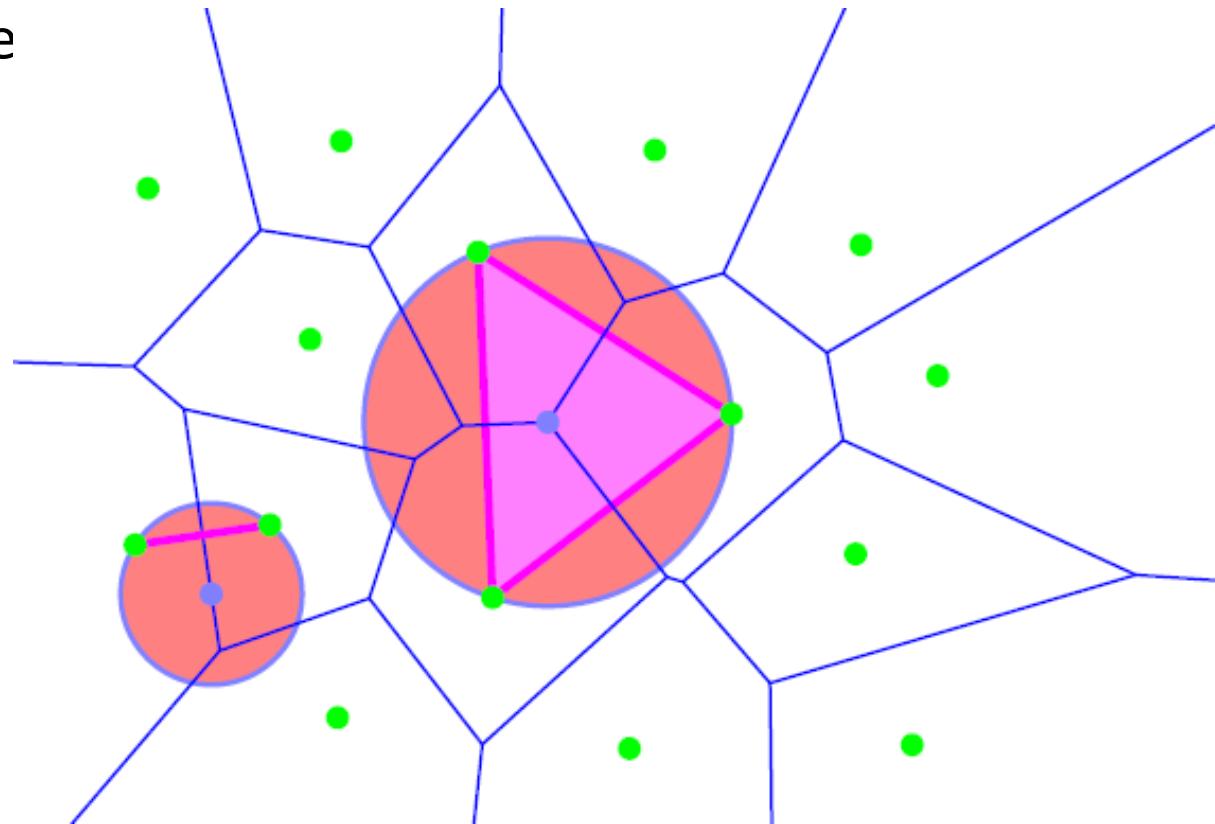
Voronoi



faces of the Voronoi diagram

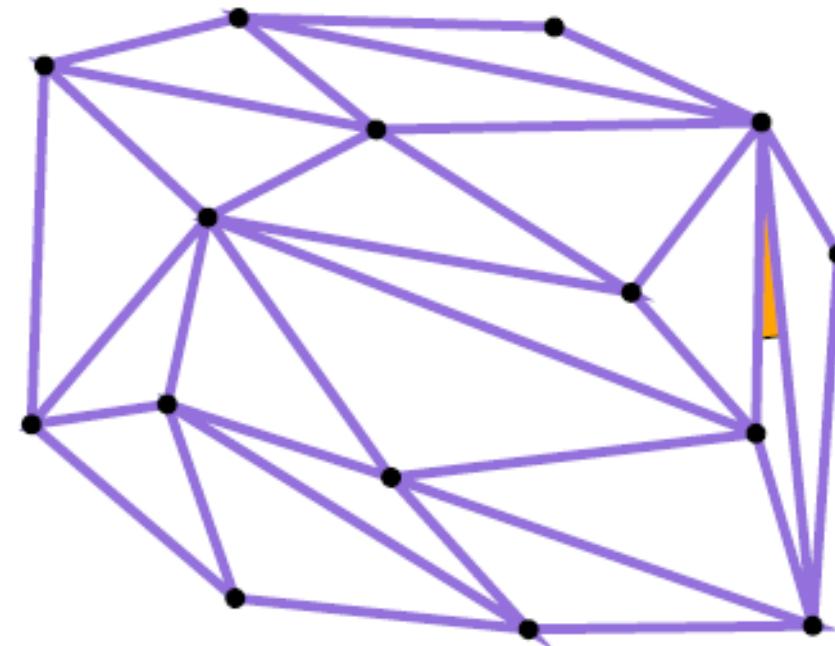
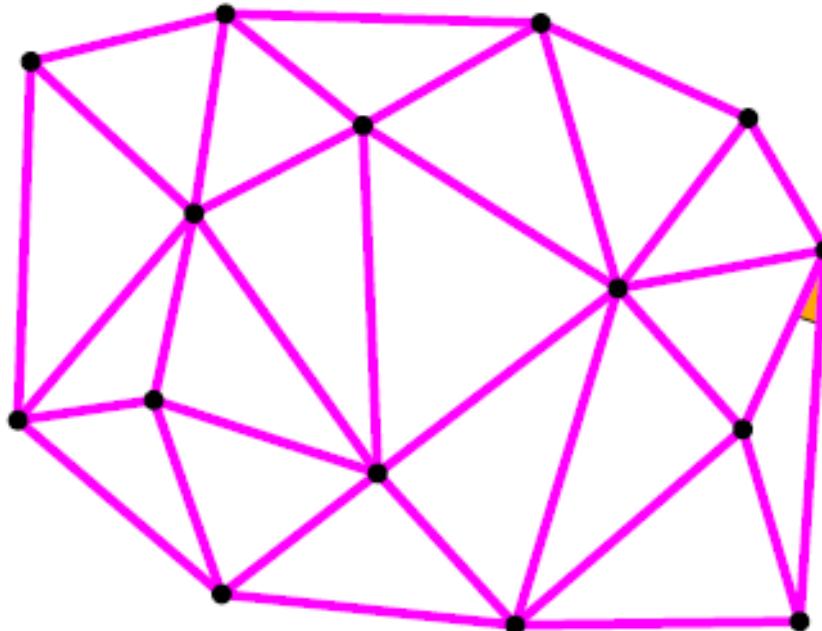
# Properties of DT (1)

- Empty sphere property: no points inside the circum-sphere of any simplex
  - Delaunay edge



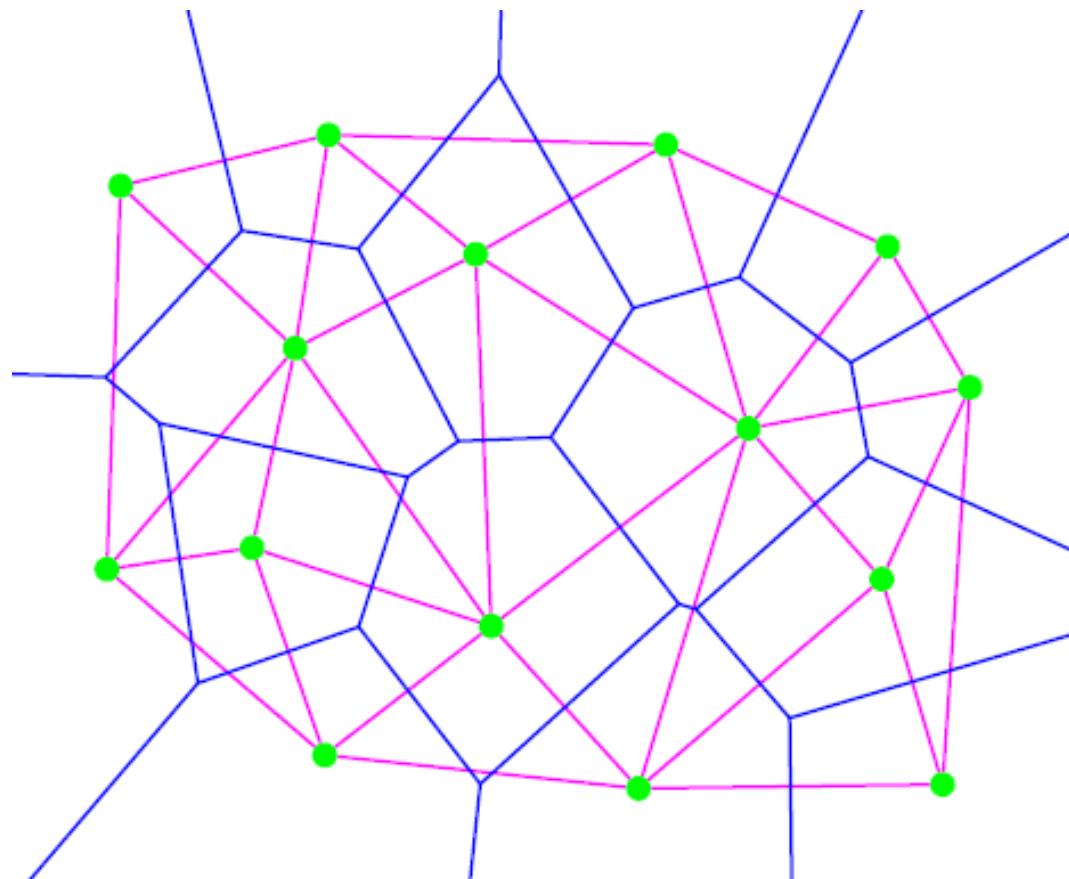
# Properties of DT (2)

- DT maximizes the smallest angle
  - [Lawson 1977] and [Sibson 1978]



# Properties of DT (3)

- Convex hull: union of all triangles

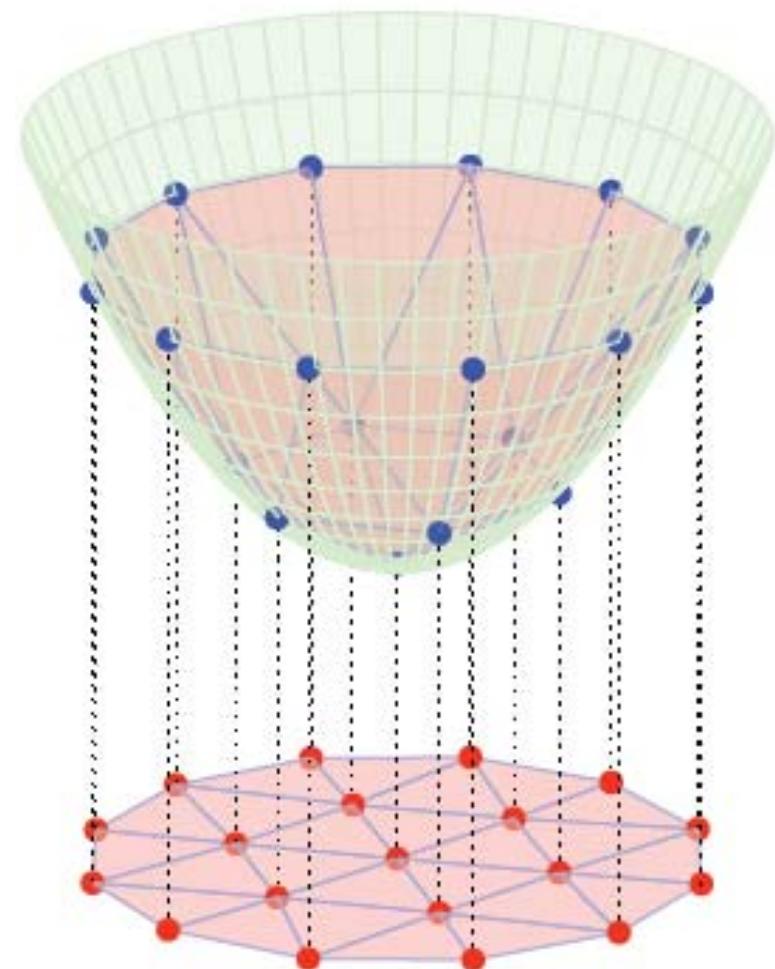


# Properties of DT (4)

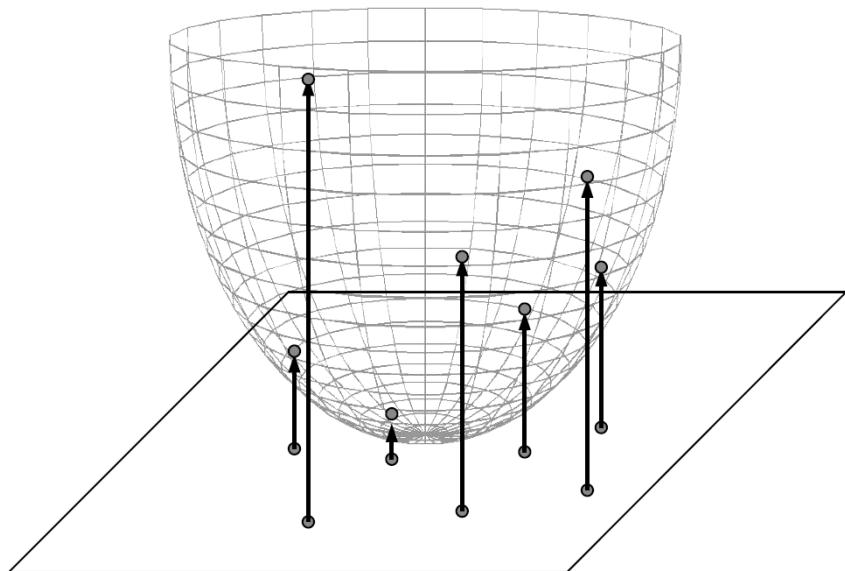
- DT maximizes the arithmetic mean of the radius of inscribed circles of the triangles.
  - [Lambert 1994]
- DT minimizes roughness (the Dirichlet energy of any piecewise-linear scalar function)
  - [Rippa 1990]
- DT minimizes the maximum containing radius (the radius of the smallest sphere containing the simplex)
  - [Azevedo and Simpson 1989], [Rajan 1991]

# Properties of DT (5)

- The DT in  $d$ -dimensional spaces is the projection of the points of convex hull onto a  $(d+1)$ -dimensional paraboloid.
  - [Brown 1979]

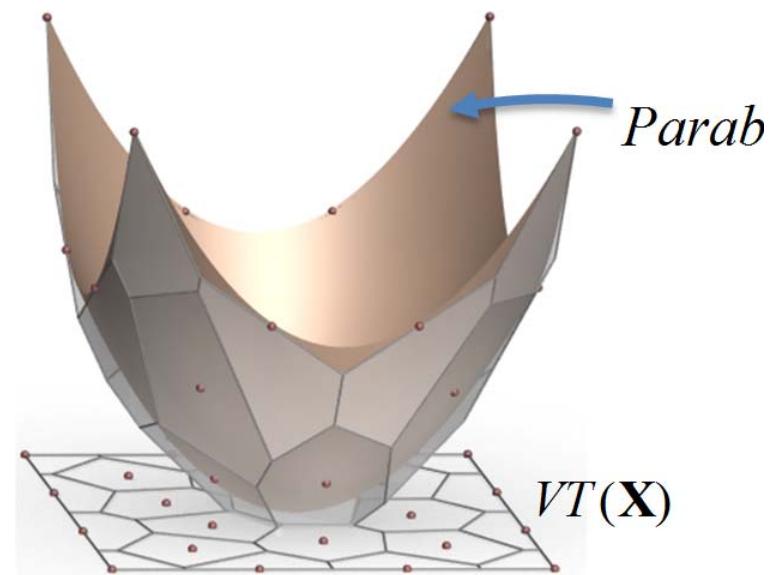


# Properties of DT (5)



Paraboloid:  $f(\mathbf{x}) = \|\mathbf{x}\|^2$

Lifting Operator:  $\mathbf{x}^* = (\mathbf{x}, \|\mathbf{x}\|^2)$



Intersections of tangent planes

# Properties of DT (6)

- DT minimizes the spectrum of the geometric Laplacian (spectral characterization)
  - [Chen et al. 2010]

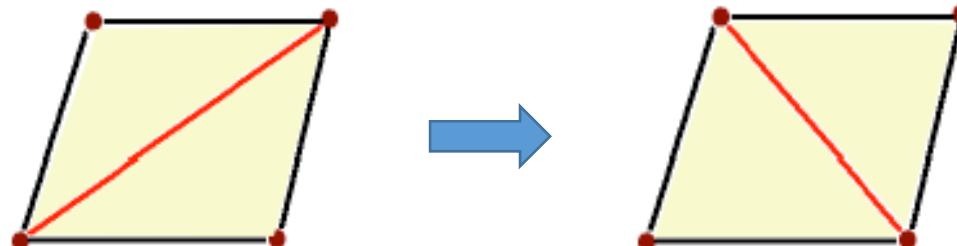
## Delaunay Spectral Theorem

The spectrum of the geometric Laplacian obtains its minimum on a Delaunay triangulation. Namely if  $\{\lambda_1 = 0, \lambda_2, \dots, \lambda_n\}$  and  $\{\mu_1 = 0, \mu_2, \dots, \mu_n\}$  are the sequences of non-decreasing eigenvalues of the geometric Laplacian of a Delaunay triangulation and of any other triangulation of the same set of points, respectively, then  $\lambda_i \leq \mu_i$  for  $i = 1, \dots, n$ .

# Simple Method: Edge Swapping/Flipping

[Sibson 1978]

- Start with any triangulation
  - 1. find any two adjacent triangles that form a convex quadrilateral that does not satisfy empty sphere condition
  - 2. swap the diagonal of the quadrilateral to be a Delaunay triangulation of that four points
  - 3. repeat step 1,2 until stuck.



Convergence? Is it possible to end with an infinite loop?

# Algorithms for Voronoi Diagrams

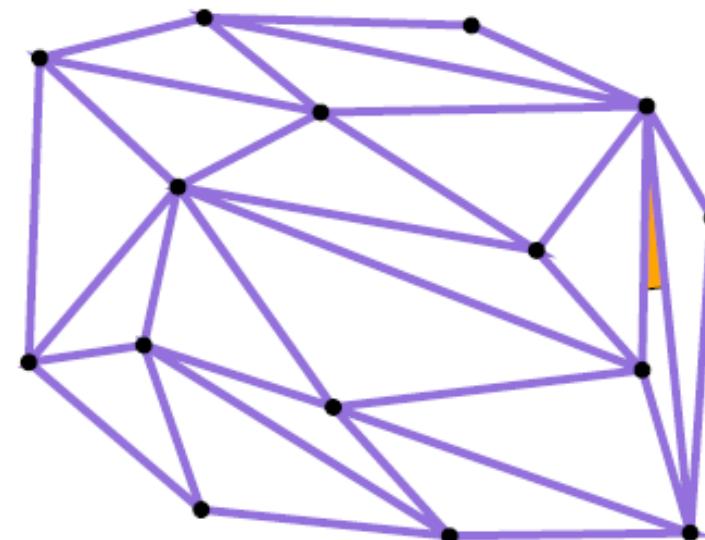
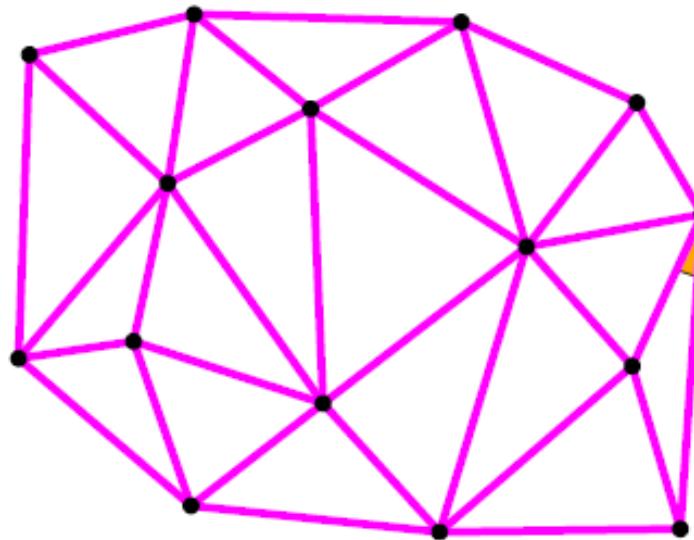
- Compute the intersection of  $n-1$  half-planes for each site, and “merge” the cells into the diagram
- Divide-and-conquer (1975, Shamos & Hoey)
- Plane sweep (1987, Fortune)
- Randomized incremental construction (1992, Guibas, Knuth& Sharir)

Available in open source library CGAL:  
<http://www.cgal.org>

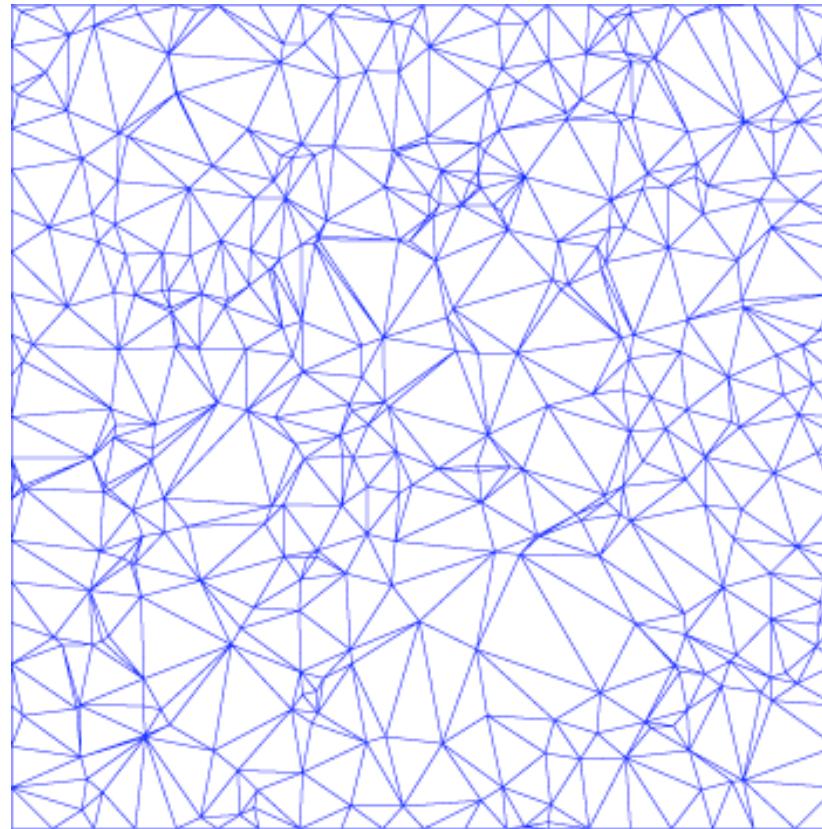
# Mesh Generation

# Mesh Generation

- Given a fixed point set, Delaunay triangulation will try to make the triangulation more shape regular and thus is considered as a “good” unstructured mesh.



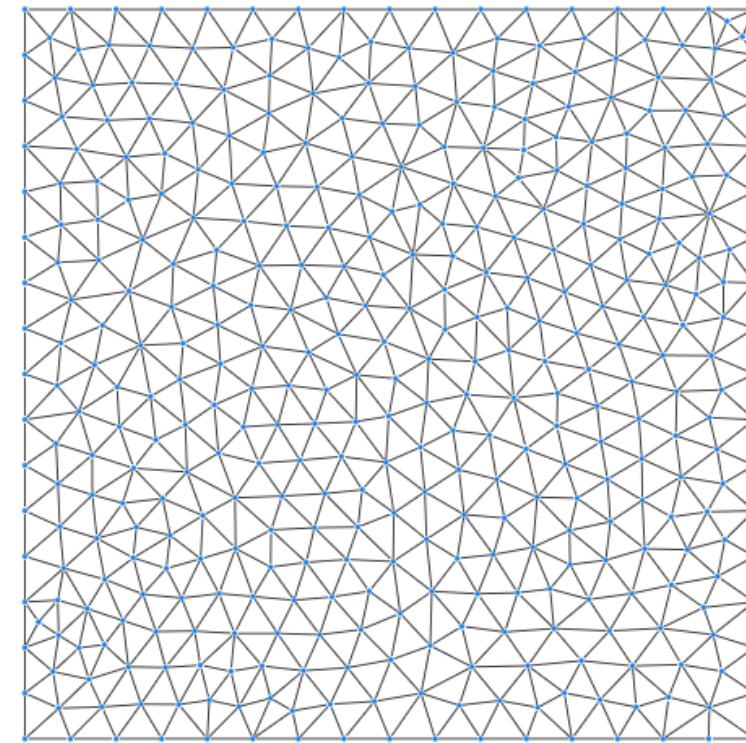
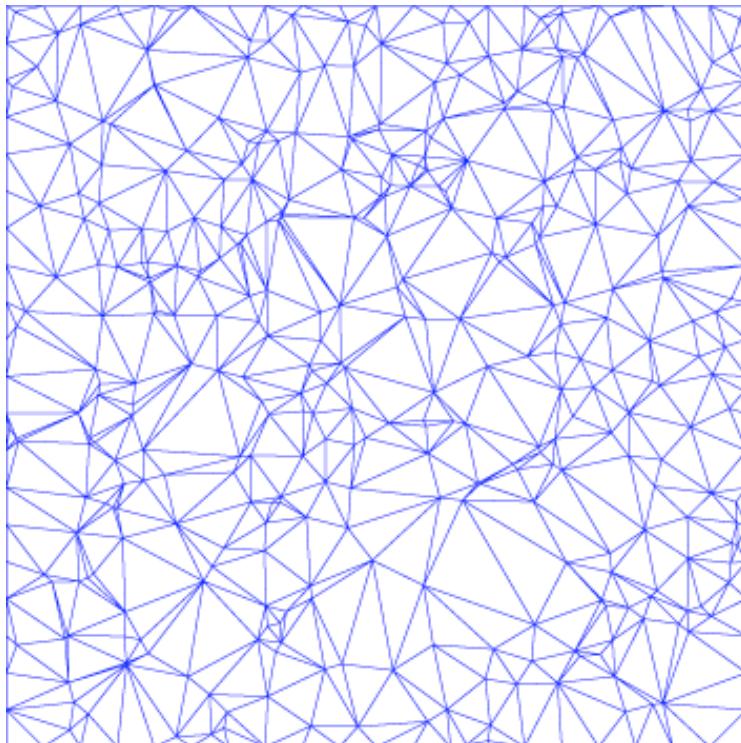
# DT is not necessary a good mesh



DT only optimize the **connectivity** when points are fixed. The **distribution of points** is more important for a good mesh.

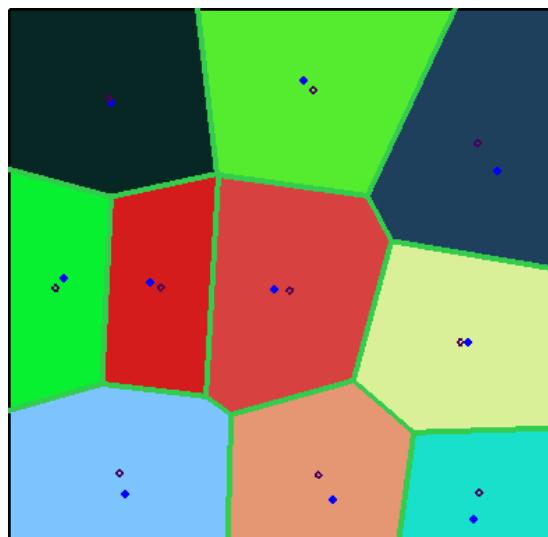
# Mesh Generation

- How to sample points to generate high-quality meshes?

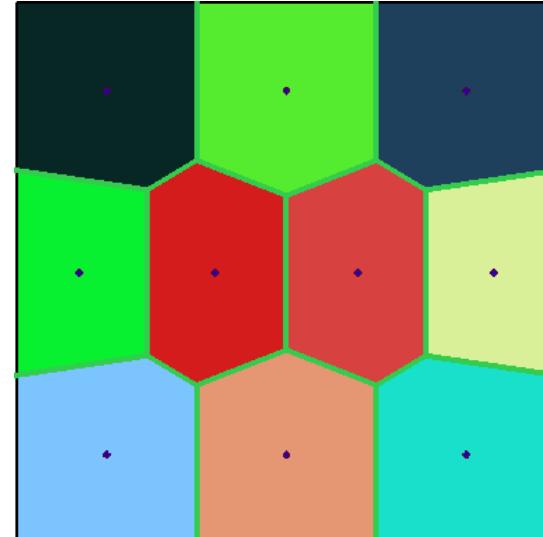


# Centroidal Voronoi Tessellation

- *Definition:* The VT is a centroidal Voronoi tessellation (CVT), if each seed coincides with the **centroid** of its Voronoi cell

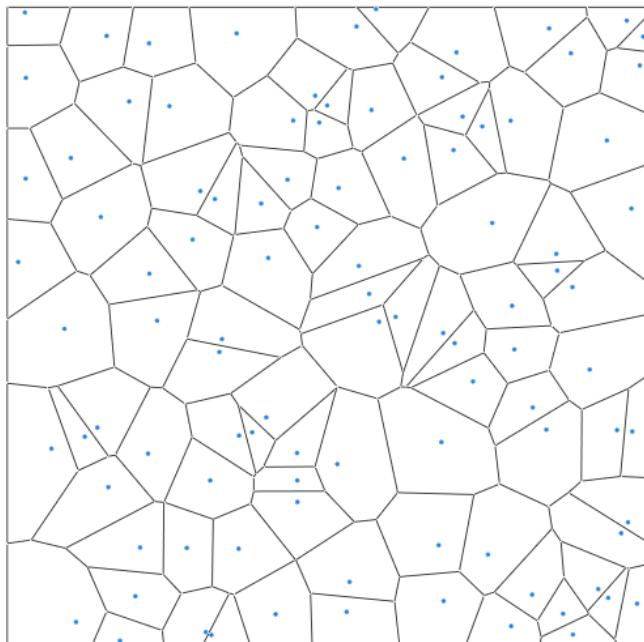


Generic VT

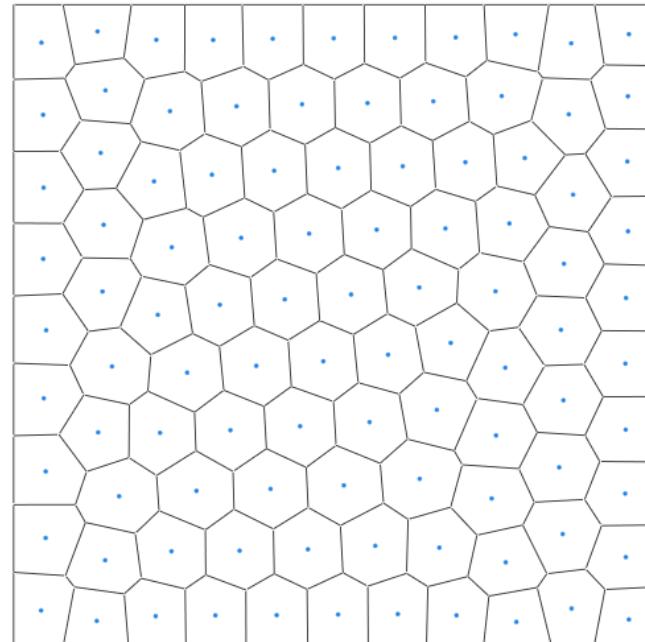


CVT

# Centroidal Voronoi Tessellation



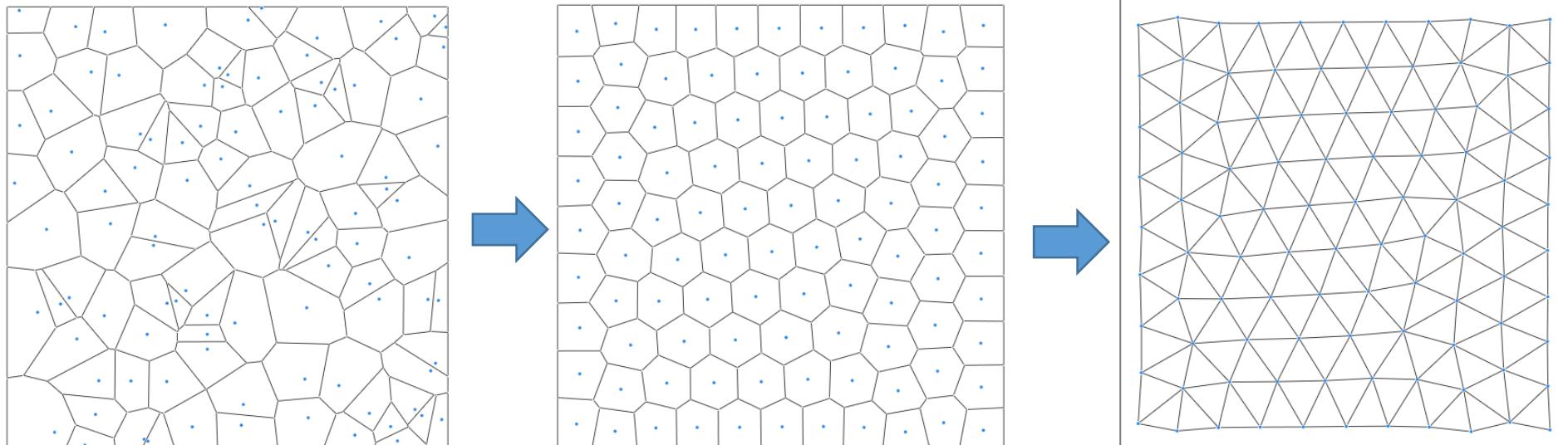
Generic VT



CVT

# Lloyd Algorithm

- Construct the VT associated with the points
- Compute the centroids of the Voronoi regions
- Move the points to the centroids
- Iterate until convergent



# Centroidal Voronoi Tessellation

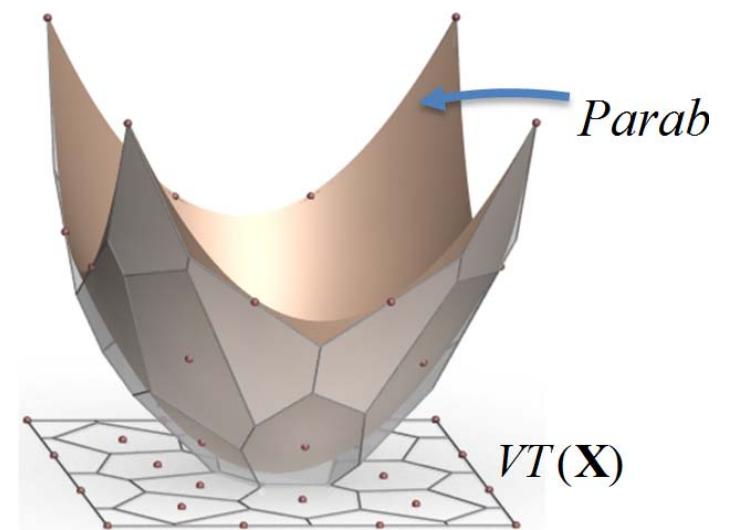
- *Definition* (Variational point of view)
  - CVT energy function:

$$F(X) = \sum_{i=1}^N \int_{V_i} \rho(\mathbf{x}) \|\mathbf{x} - \mathbf{x}_i\|^2 d\mathbf{x}$$

- CVT is a critical point of  $F(X)$ , an *optimal* CVT is a global minimizer of  $F(X)$

# CVT Energy Function

- Geometric interpretation
  - The CVT energy with  $\rho(\mathbf{x})$  identical to 1, is the volume difference between the circumscribed polytope and the paraboloid.



# The Gradient of CVT Energy

- The gradient of  $F(\mathbf{X})$  is [Iri et al. 1984; Asami 1991; Du et al. 1999]:

$$\frac{\partial F}{\partial \mathbf{x}_i} = 2m_i(\mathbf{x}_i - \mathbf{c}_i),$$

where

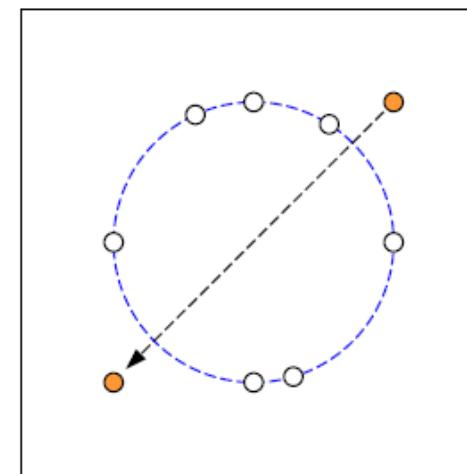
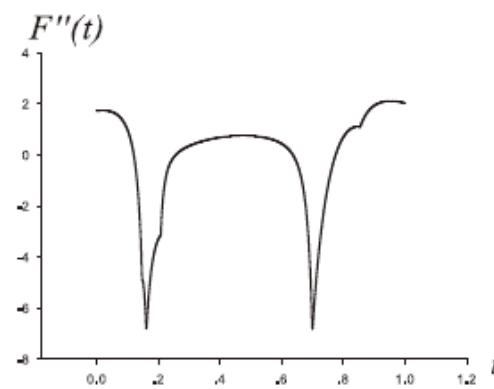
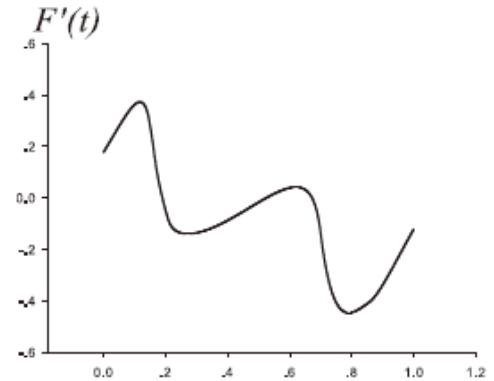
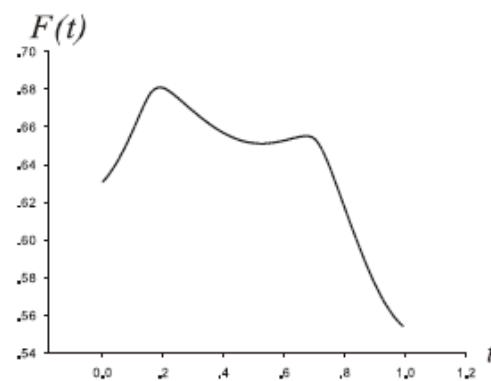
$$m_i = \int_{\mathbf{x} \in \Omega_i} \rho(\mathbf{x}) d\sigma$$

- Lloyd's method is a gradient descent method, thus has linear convergence

# Smoothness of $F(\mathbf{X})$

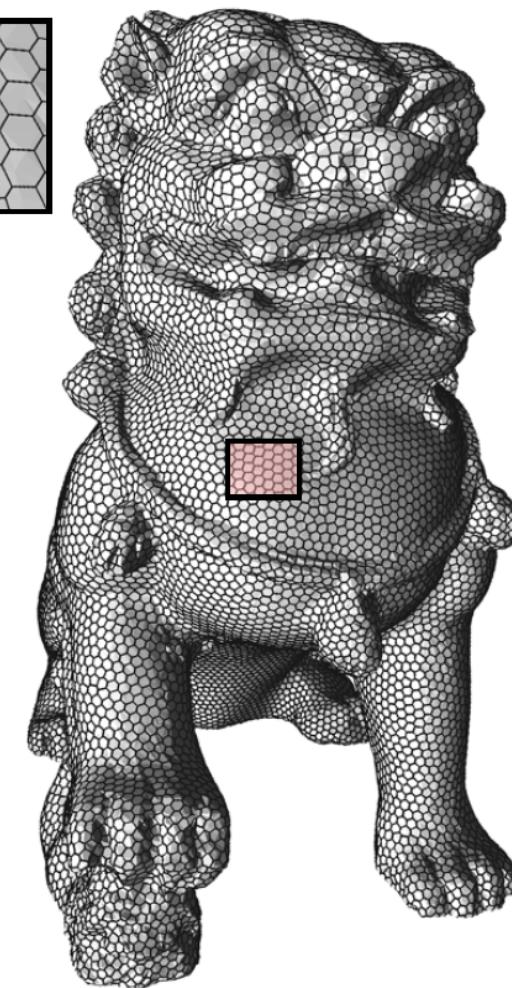
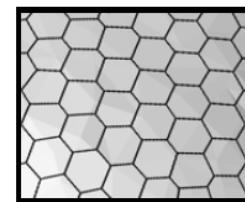
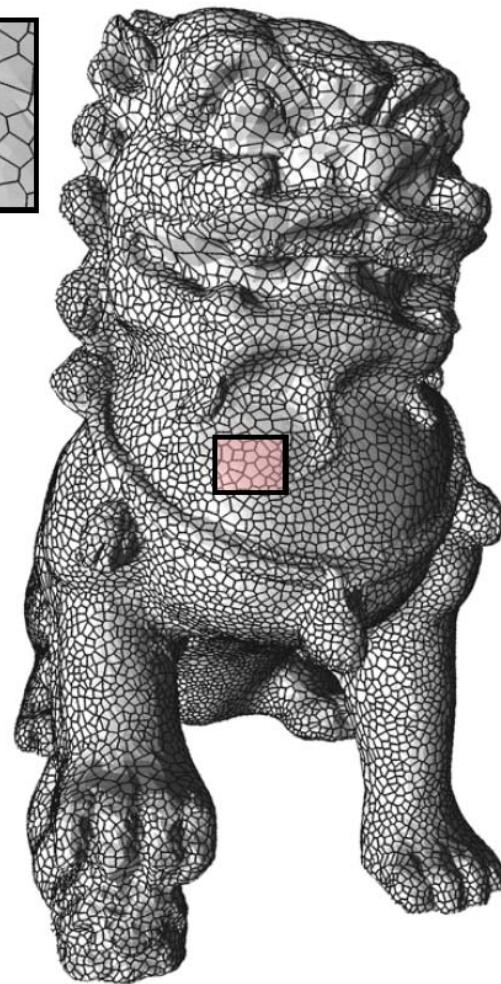
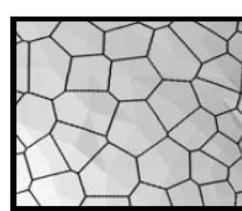
- Can BFGS method be applied to computing CVT? Or does CVT energy  $F(\mathbf{X})$  have required  $C^2$  smoothness?
- Results:
  - It has been noted that  $F(\mathbf{X})$  is non-smooth [Iri et al. 1984], but without proof
  - It has been proved that  $F(\mathbf{X})$  is  $C^1$  [Cortes et al. 2005]
  - $F(\mathbf{X})$  is  $C^2$  in a **convex** domain in 2D and 3D [Liu et al. 2009]

# $C^2$ Continuity of $F(X)$ – Illustration



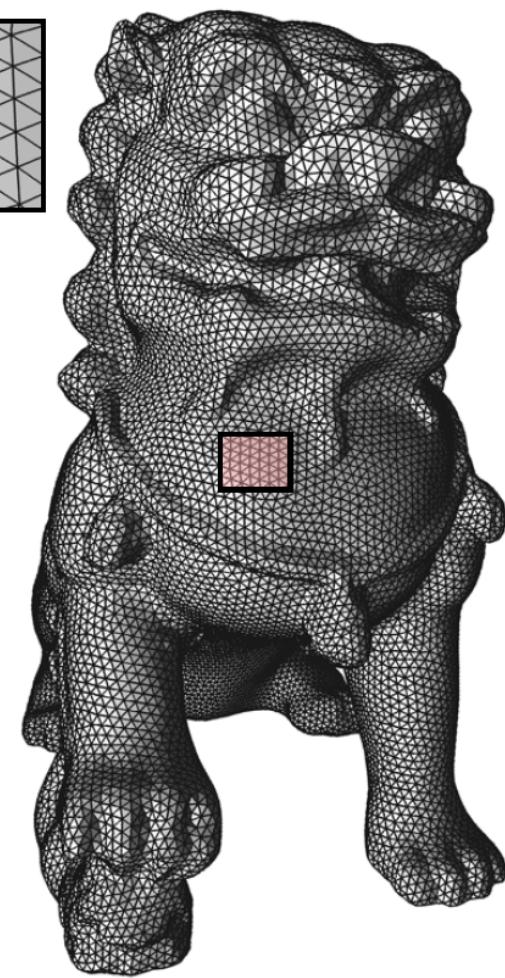
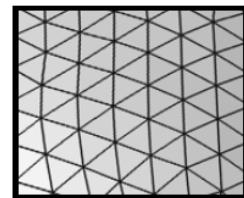
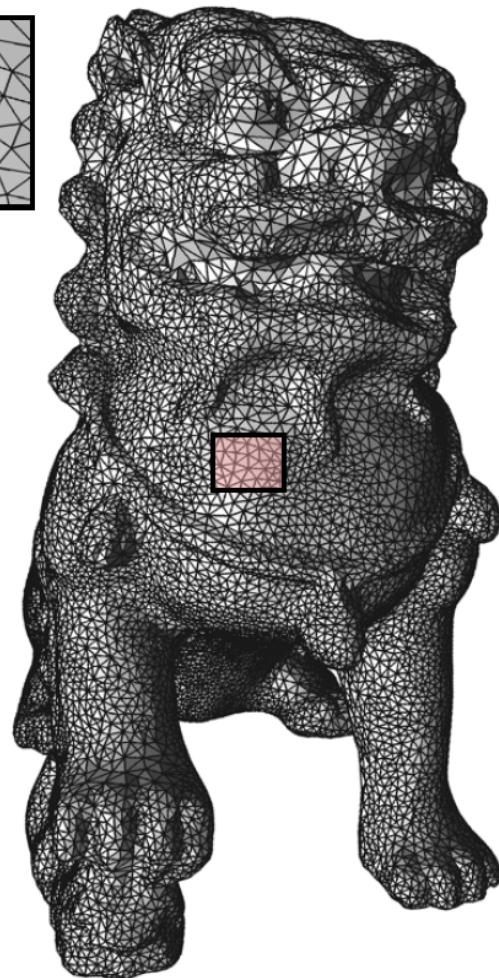
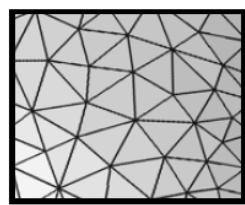
**Figure:** Illustration of  $C^2$  smoothness of CVT energy in 2D

# CVT on Surface



Left: initial Voronoi; Right: CVT

# CVT for Remeshing



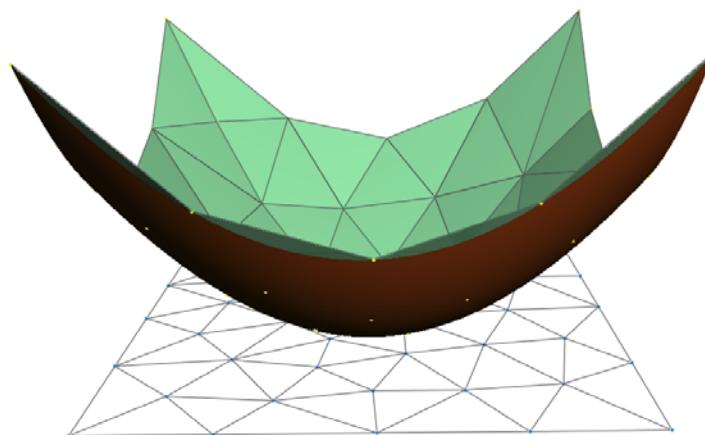
Left: initial mesh; Right: dual mesh of CVT

# Optimal Delaunay Triangulation

- ODT energy function:

$$\begin{aligned} E(X) &= \|f - f_{I,\mathcal{T}}\|_{L^1(\Omega)} \\ &= \sum_{\tau \in \mathcal{T}} \int_{\tau} f_I(\mathbf{x}) d\mathbf{x} - \int_{\Omega} f(\mathbf{x}) d\mathbf{x} \end{aligned}$$

Volume between the  
lift-up of DT and the  
paraboloid



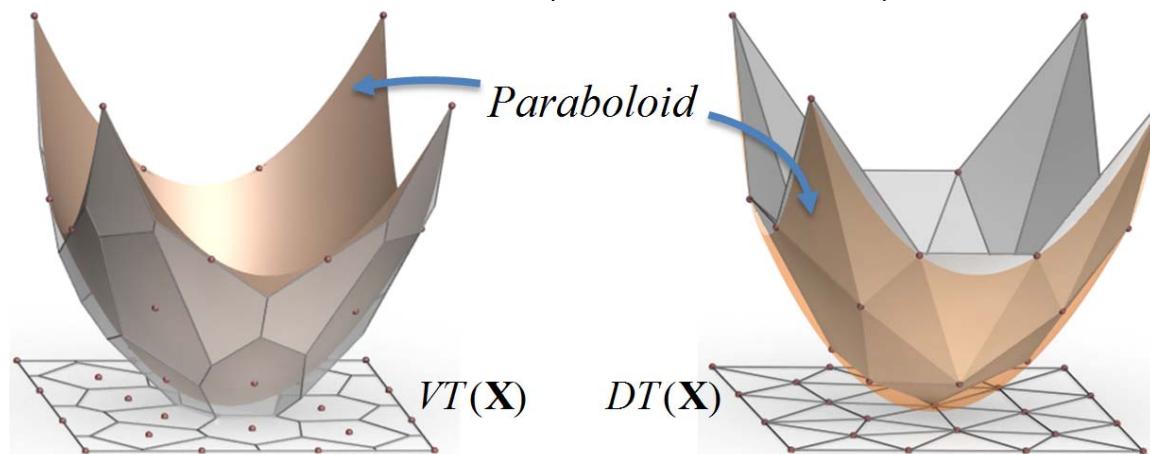
# CVT & ODT Energy

- CVT energy

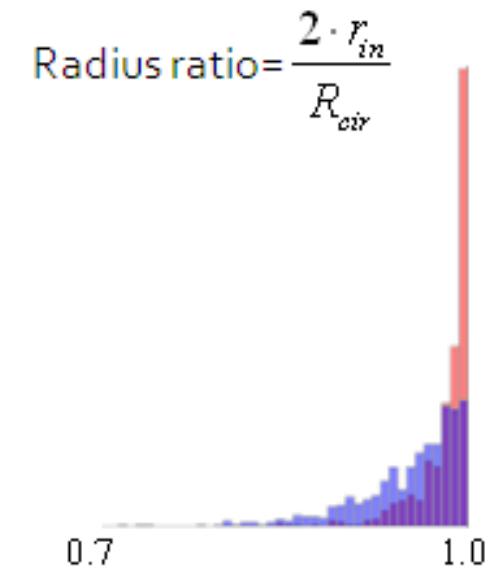
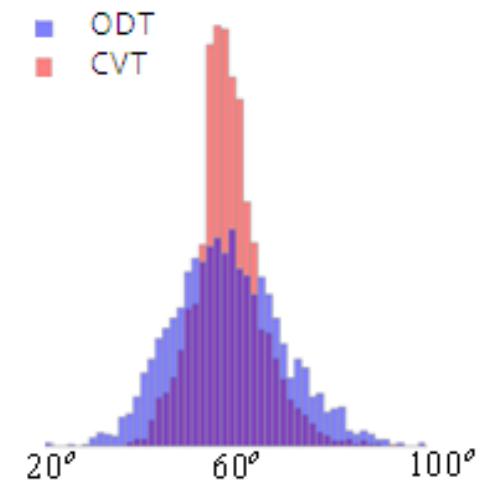
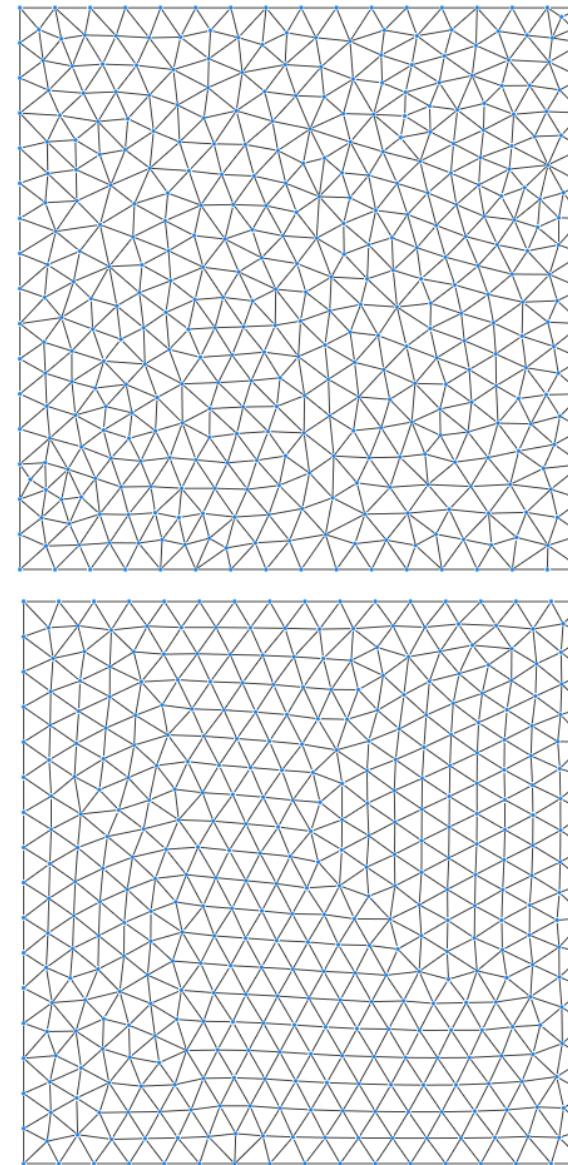
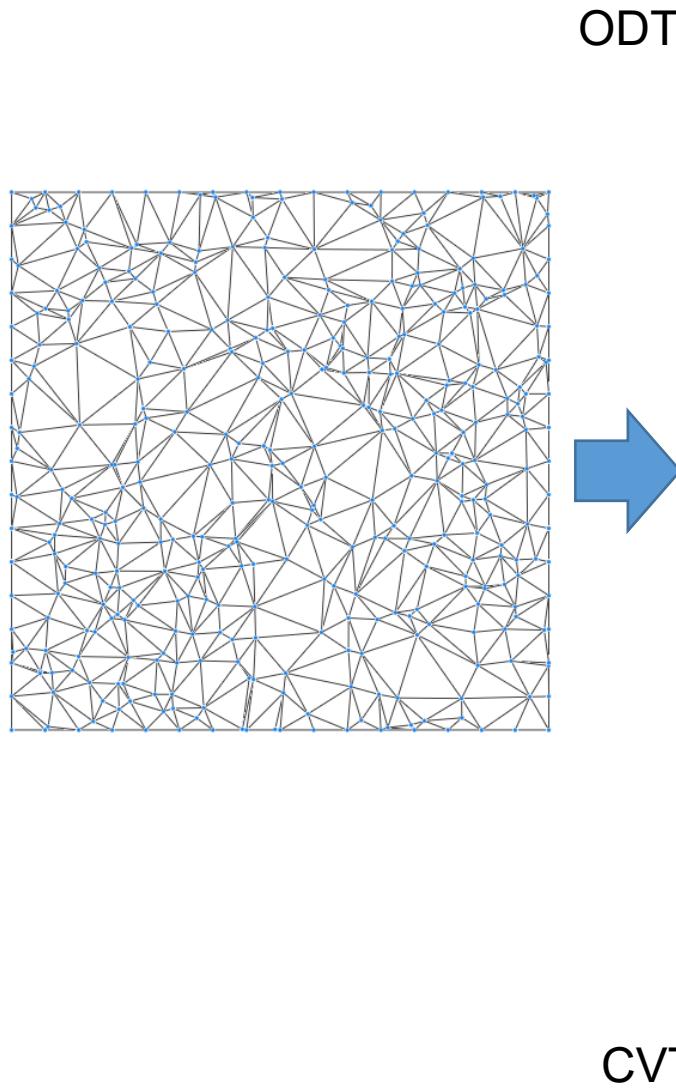
$$F(X) = \sum_{i=1}^N \int_{V_i} \|\mathbf{x} - \mathbf{x}_i\|^2 d\mathbf{x}$$

- ODT energy

$$E(X) = \sum_{\tau \in \mathcal{T}} \int_{\tau} f_I(\mathbf{x}) d\mathbf{x} - \int_{\Omega} f(\mathbf{x}) d\mathbf{x}$$



# Compare ODT and CVT



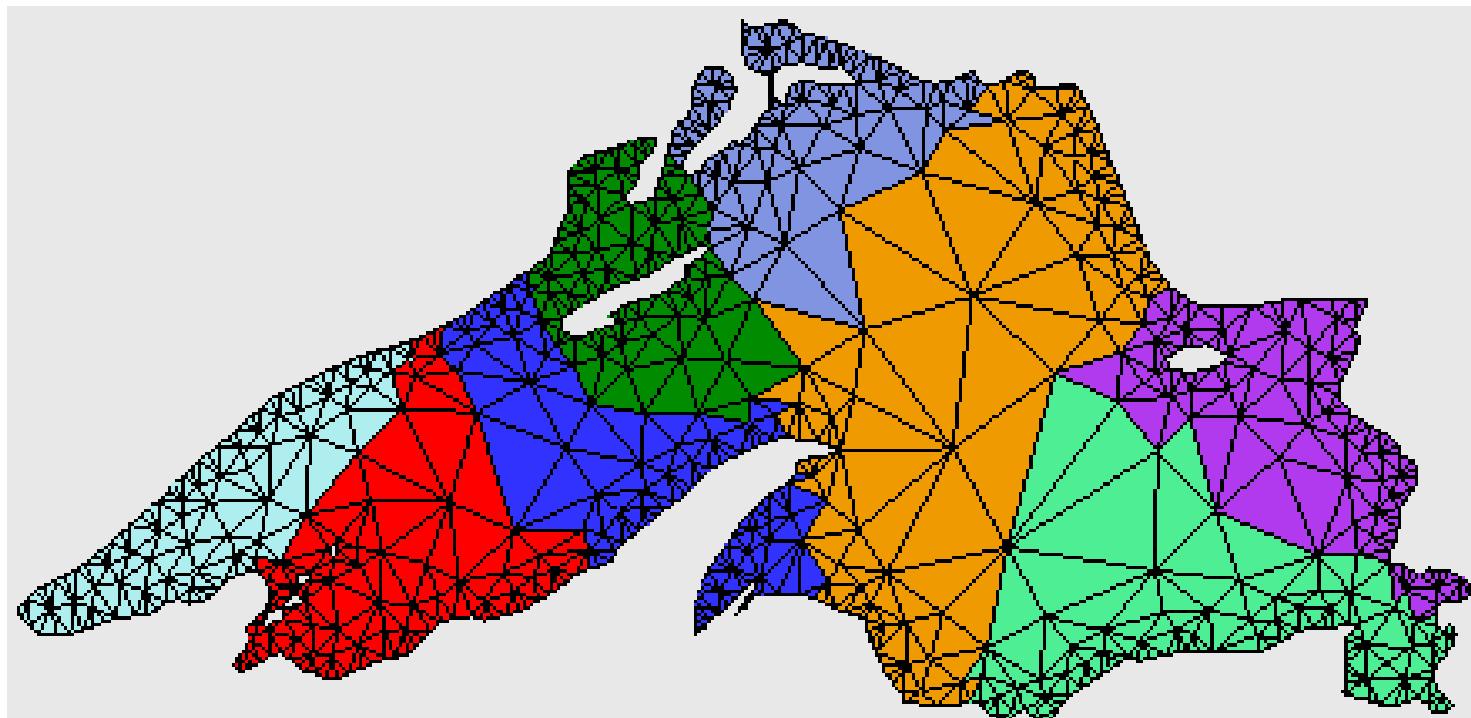
# Triangle Library

A Two-Dimensional Quality Mesh Generator and  
Delaunay Triangulator

# Use the library “Triangle”!

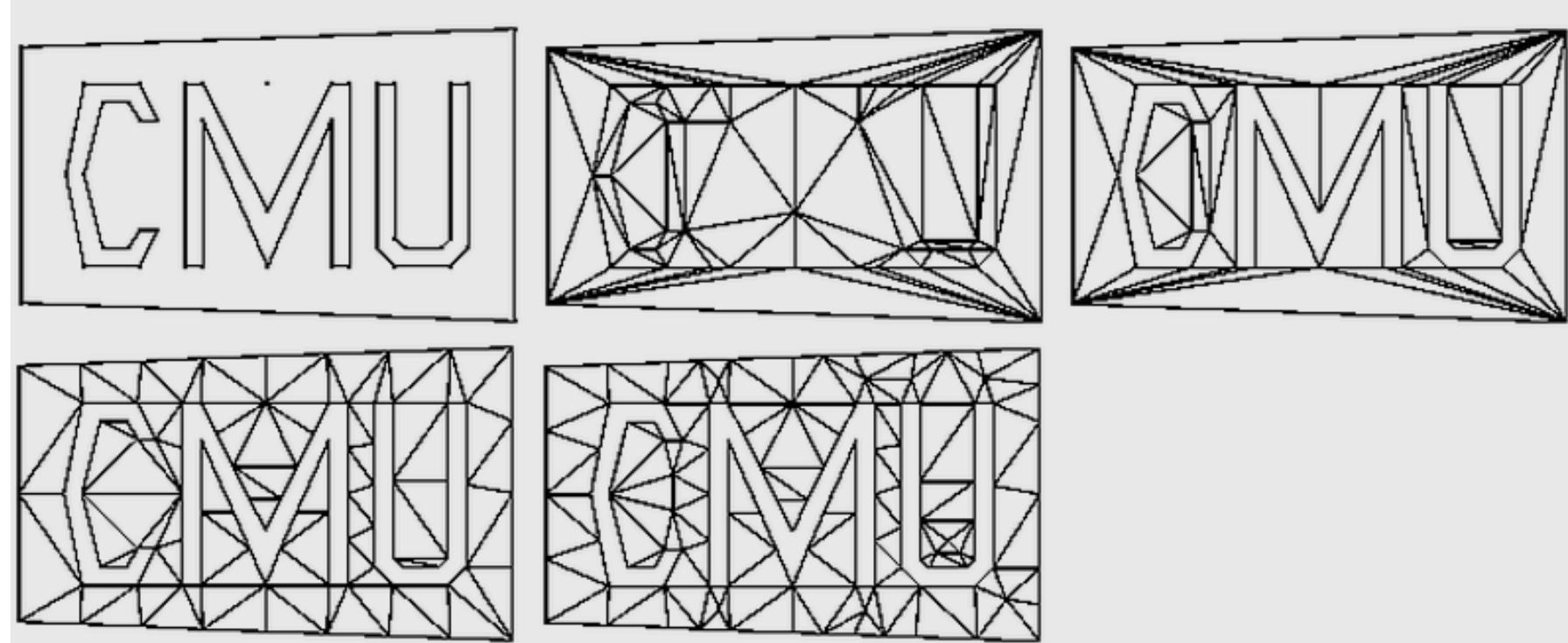


- <http://www.cs.cmu.edu/~quake/triangle.html>



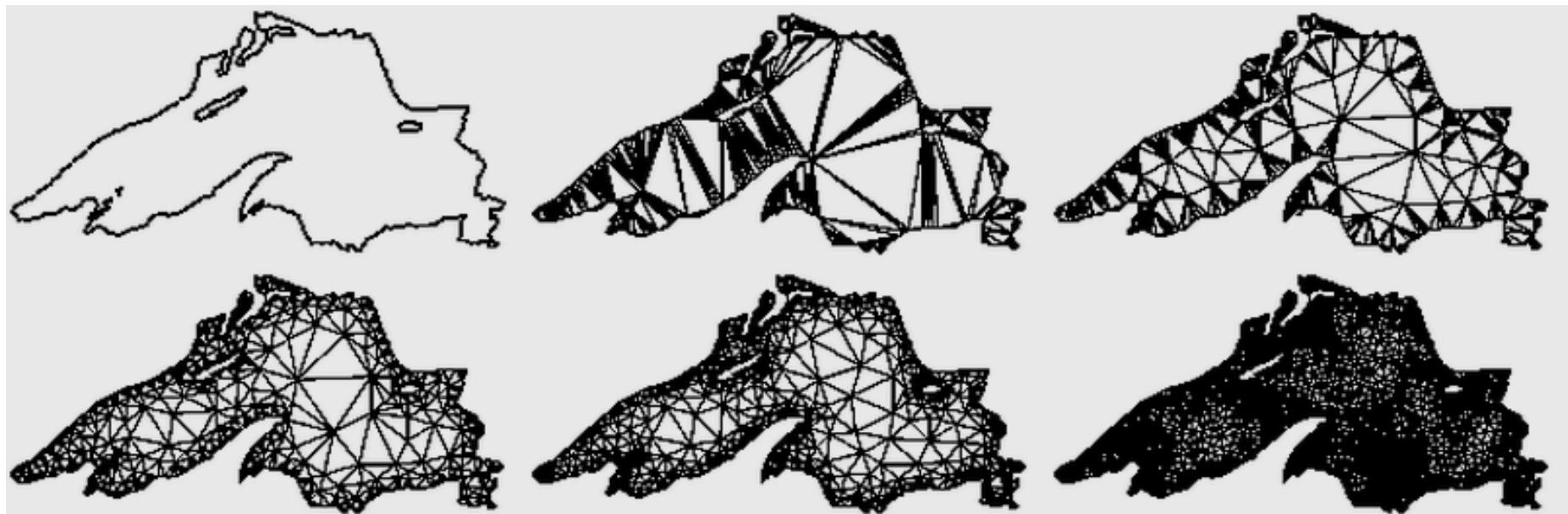
# Examples of “Triangle”

- Planar Straight Line Graph (PSLG)
- a Delaunay triangulation of its vertices
- a constrained Delaunay triangulation of the PSLG
- a conforming Delaunay triangulation of the PSLG
- a quality conforming DT of the PSLG with no angle smaller than 25 degree



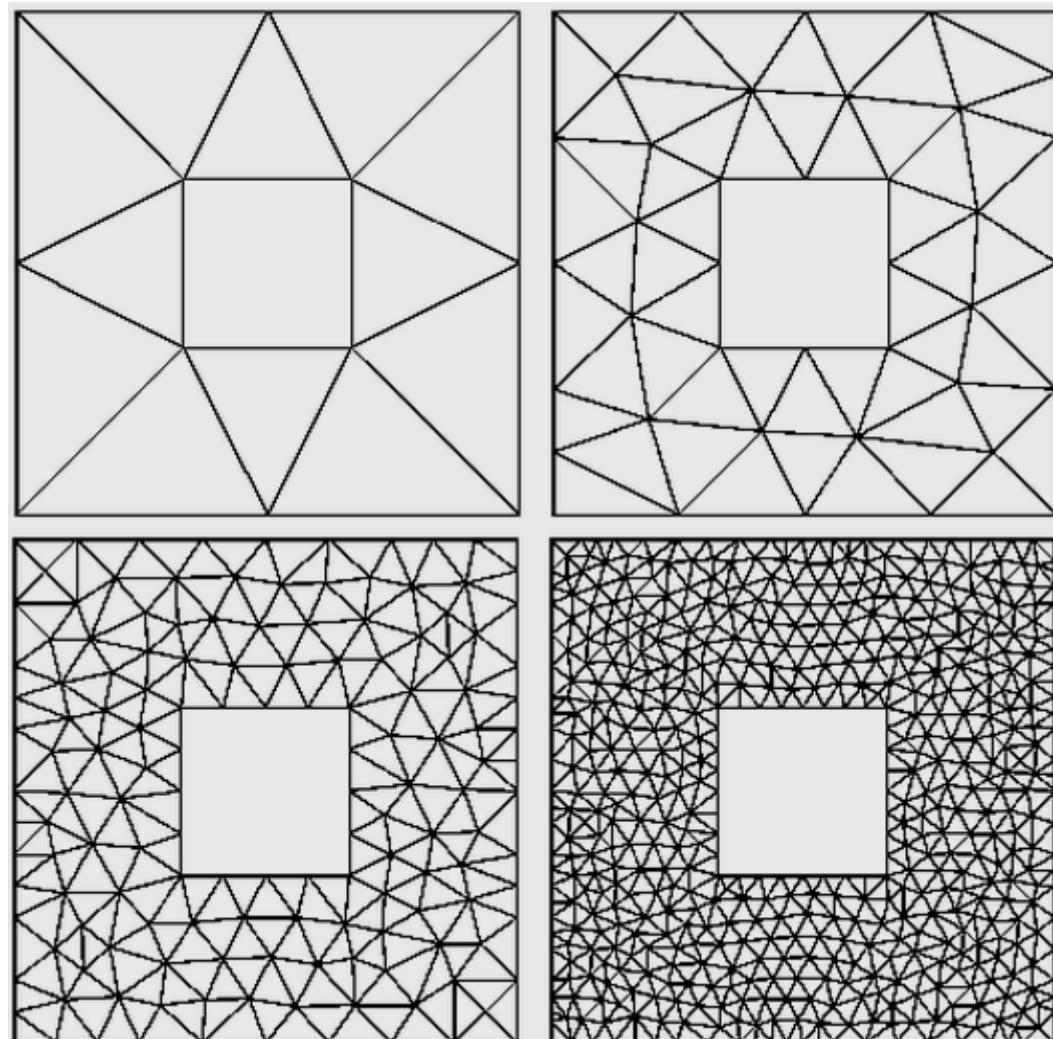
# Examples of “Triangle”

- a PSLG of Lake Superior
- triangulations having minimum angles of 0, 5, 15, 25, and 33.8 degrees.



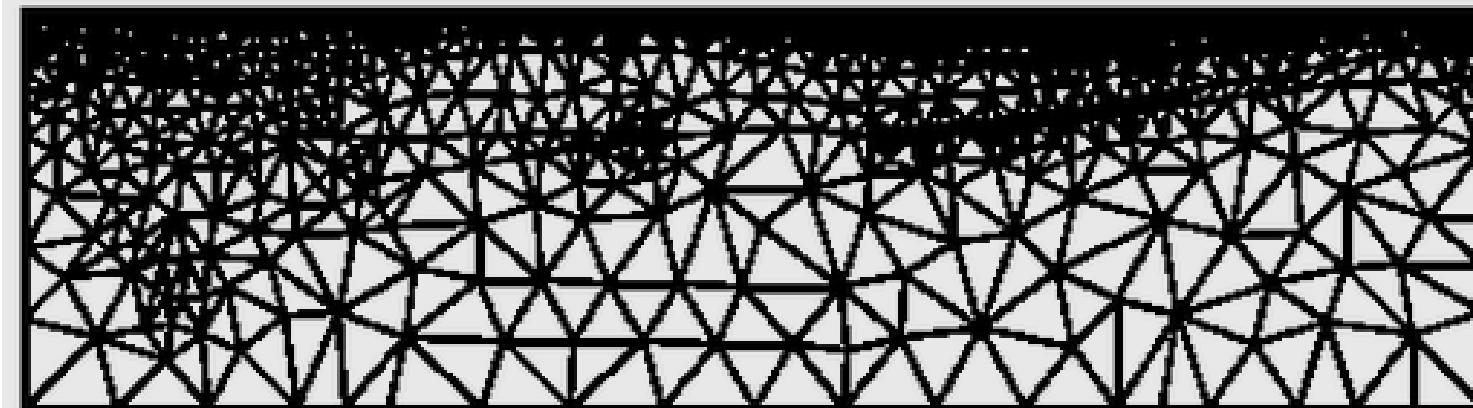
# Examples of “Triangle”

- Maximum area constraints

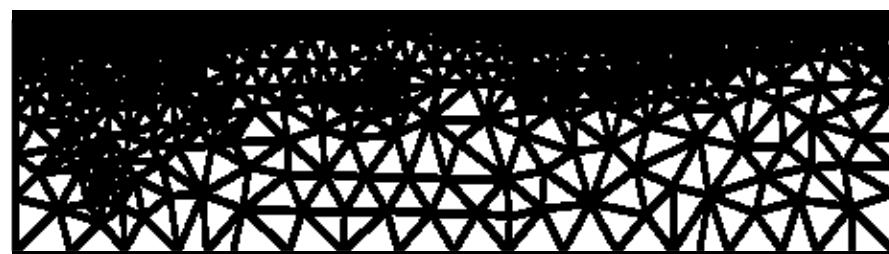
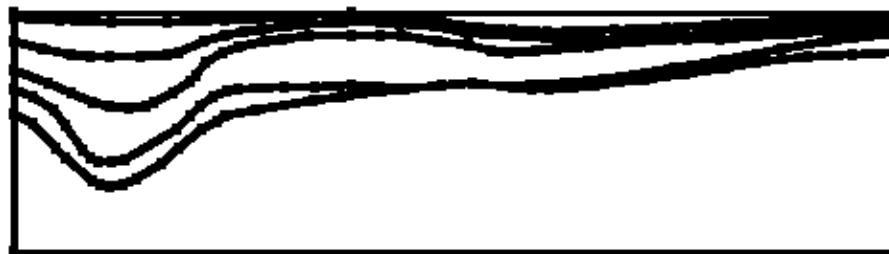
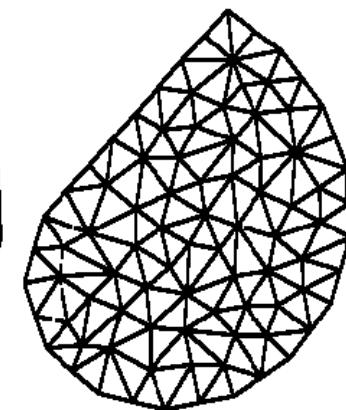
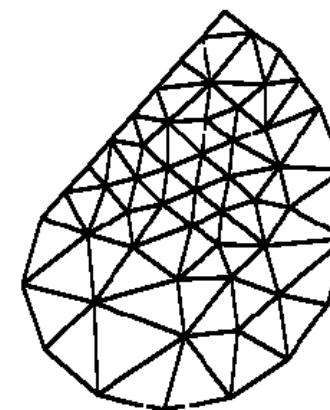
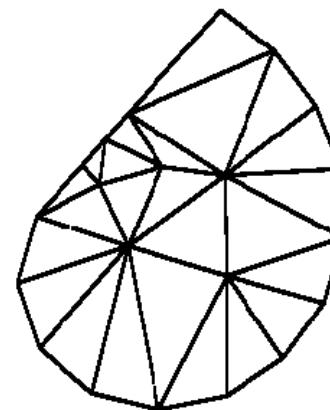
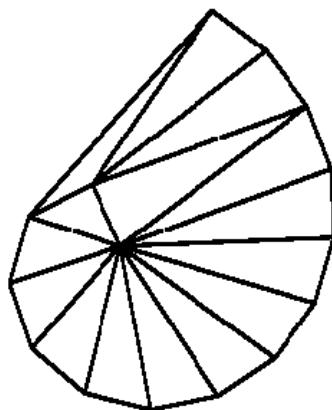


# Examples of “Triangle”

- Different maximum triangle area constraints

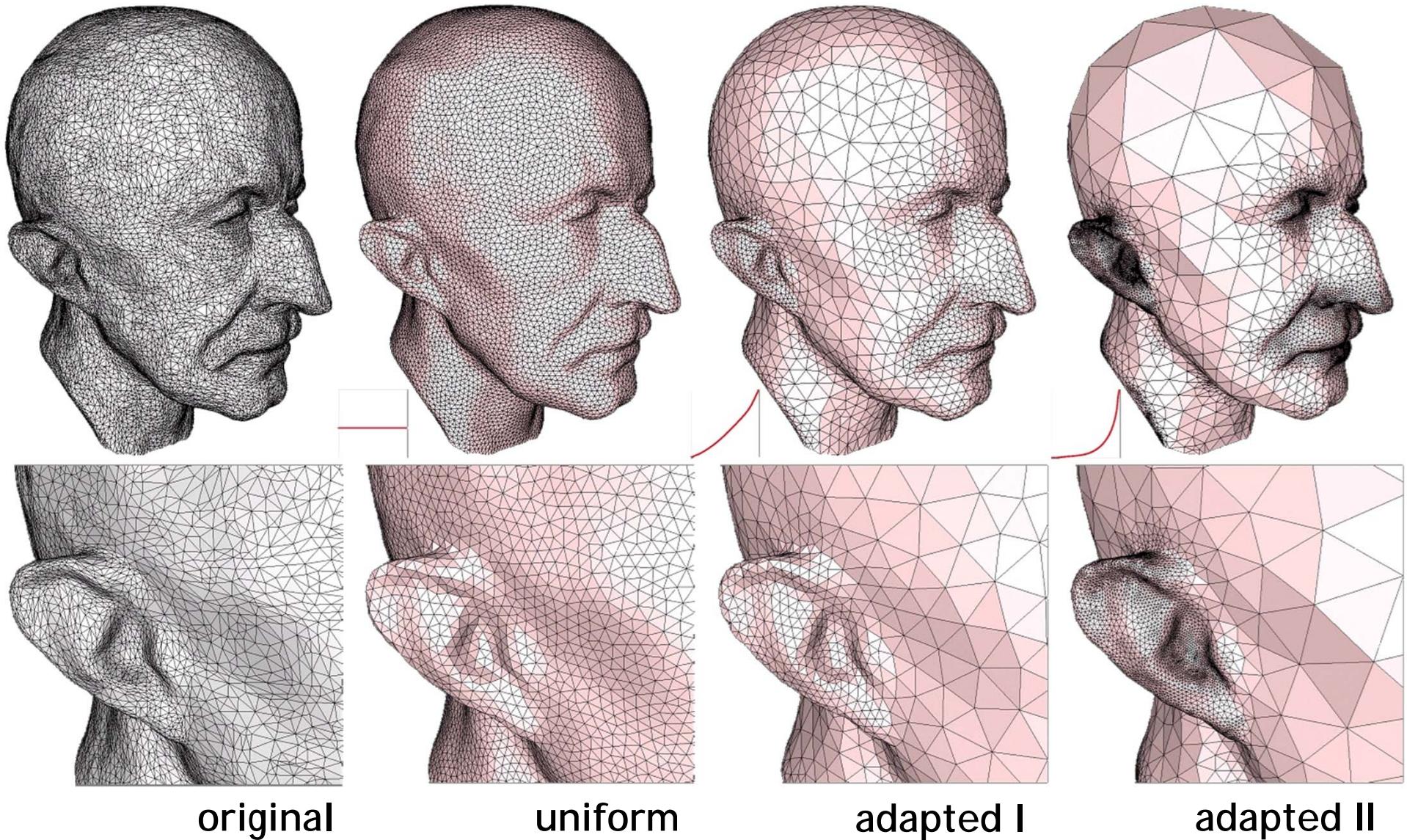


# Examples of “Triangle”

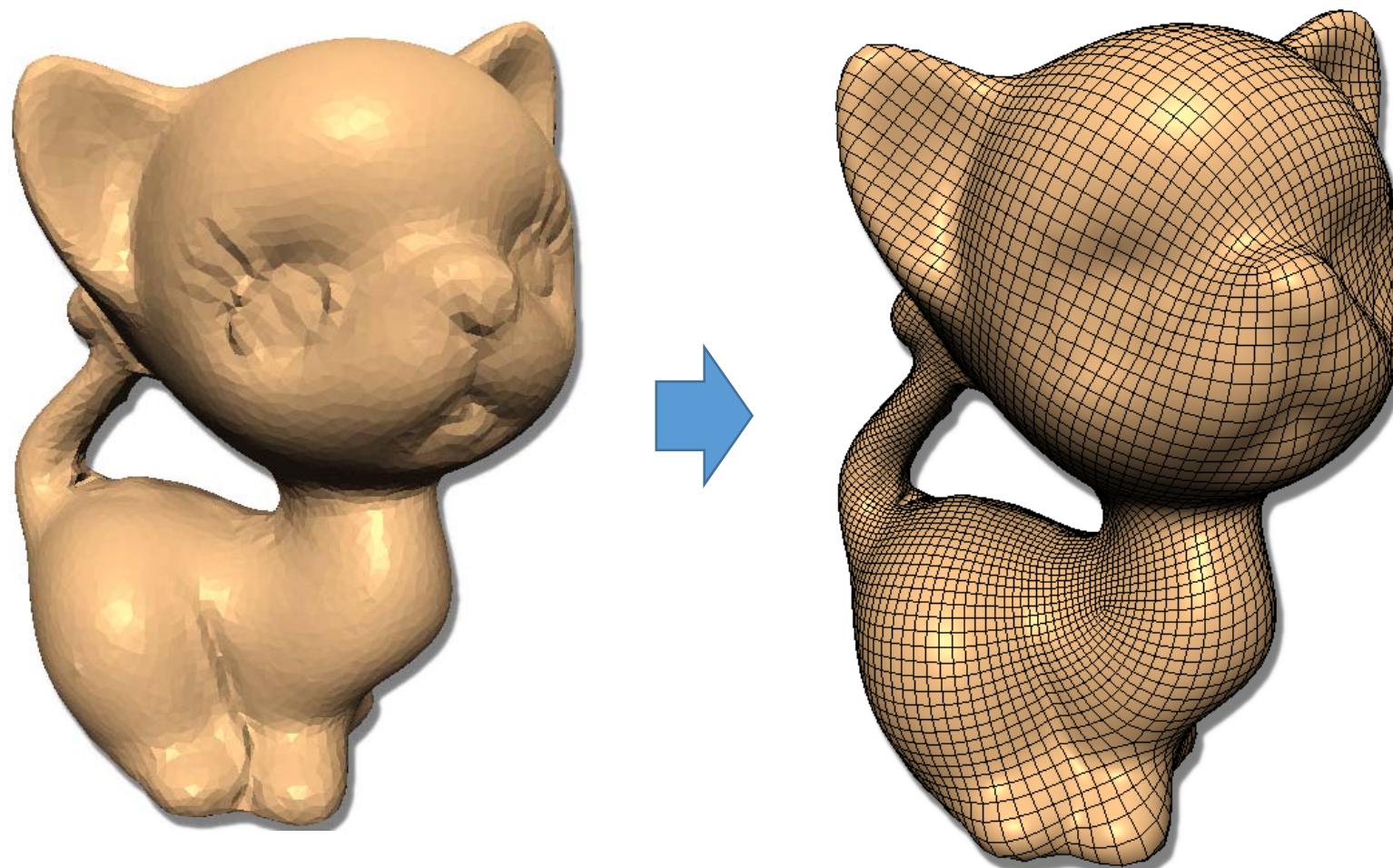


# 高维几何对象的 采样与剖分

# 二维流形曲面的采样与网格化

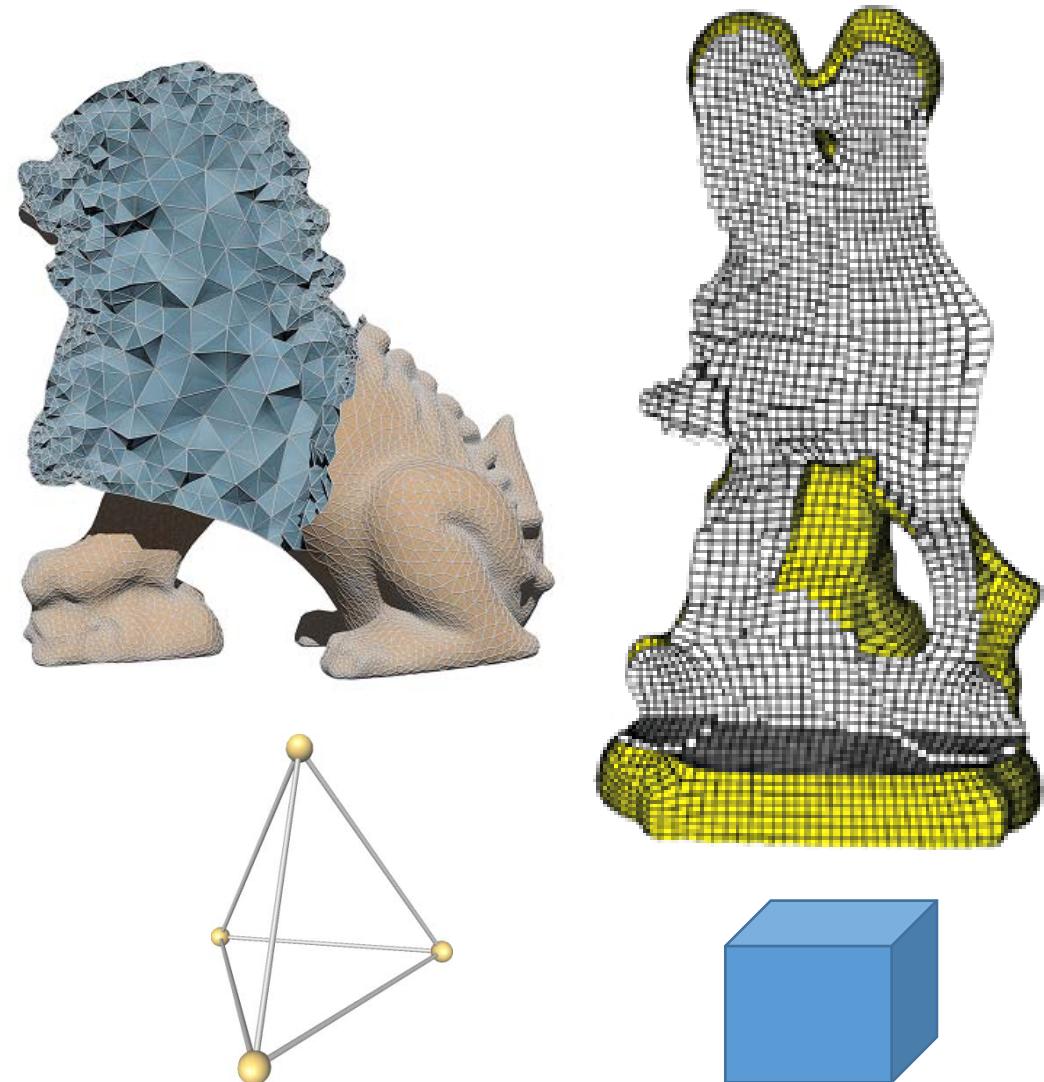


# 二维流形曲面的四边形网格化

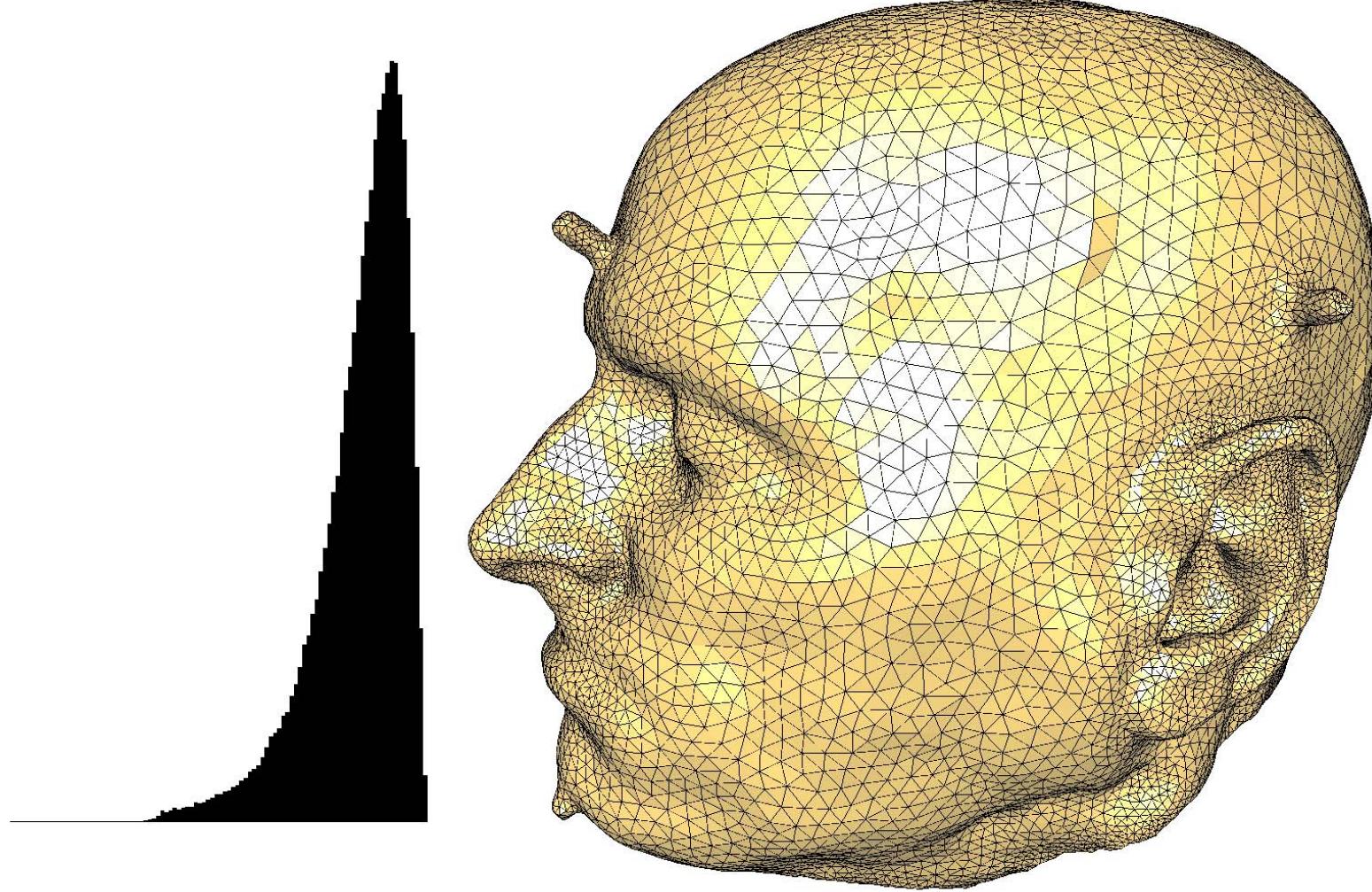


# 空间体的采样与剖分

- Interior of 3D shapes
  - FEM
  - Simulation
  - ...
- Two typical types
  - Tetrahedral meshes
  - Hexahedral meshes



# 空间体的四面体网格



# TETGEN (**Tetrahedron Generation**)

- A Quality Tetrahedral Mesh Generator and 3D Delaunay Triangulator
- <https://people.sc.fsu.edu/~jburkardt/examples/tetgen/tetgen.html>
- Author: Hang Si

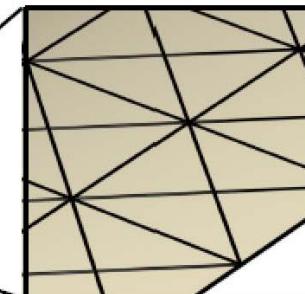
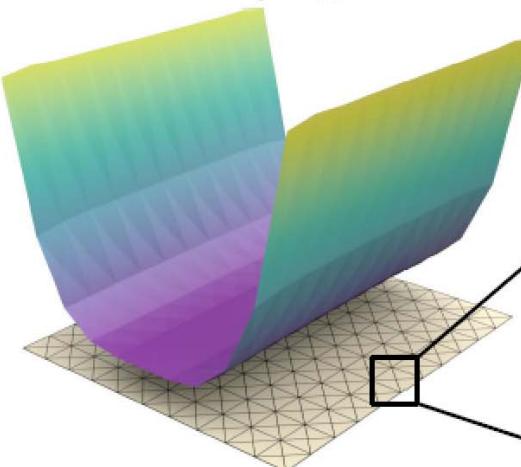
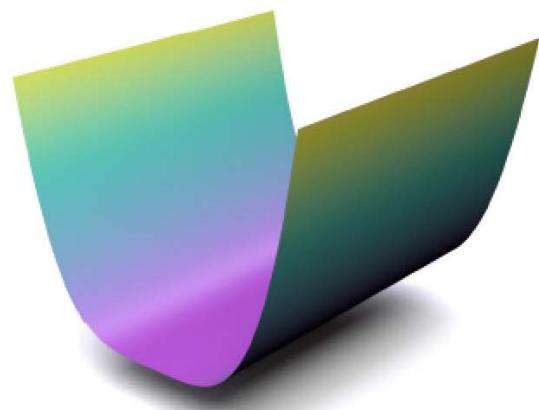
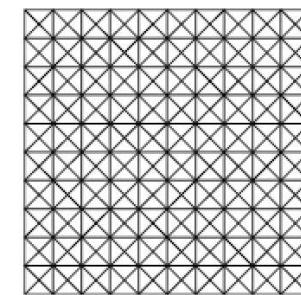
Meshting is still hard...

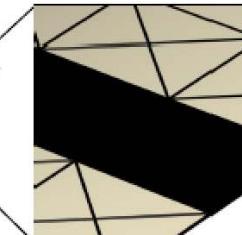
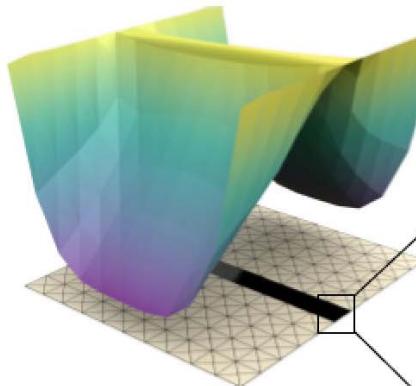
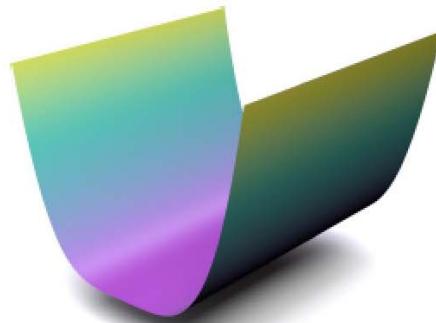
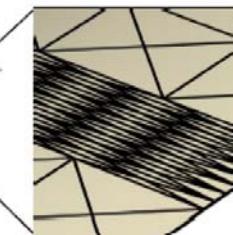
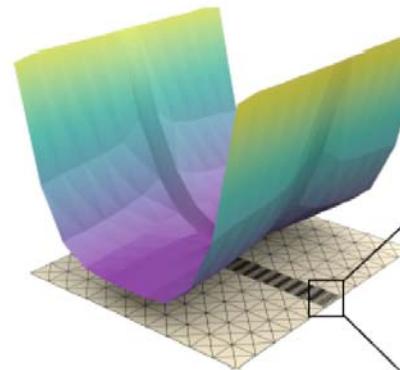
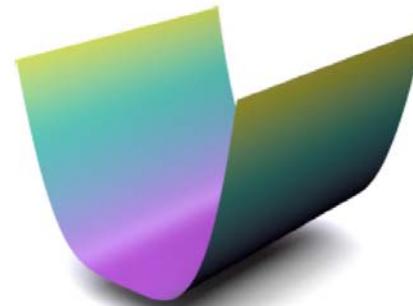
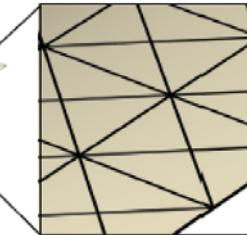
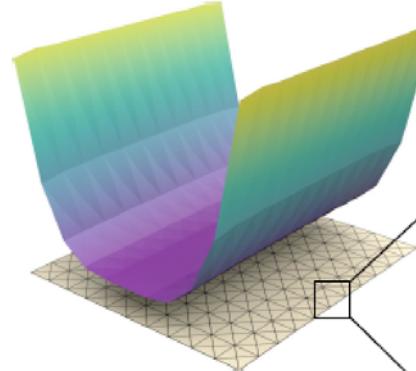
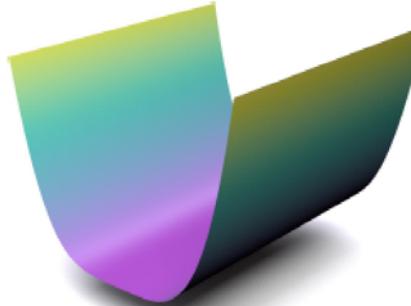
# Solving PDE (FEM)

- Resolution, basis order, element quality...

$$\Delta u = f$$

$$u = x^4 \quad \approx \quad U = \sum_{i=1}^n u_i \phi_i$$





# 作业8

- 任务：
  - 实现平面点集CVT的Lloyd算法
- 目的
  - 学习Voronoi算法、使用相关数学库（如Triangle、CGAL等）
- 步骤
  - 在给定的正方形区域内随机生成若干采样点
  - 生成这些点的Voronoi剖分
  - 计算每个剖分的重心，将采样点的位置更新到该重心
  - 迭代步骤2和3
- 思考（非必要、可选）
  - 如何在曲面上采样并生成CVT（如何将算法推广到流形曲面）？
- Deadline: **2020年12月19日晚**



中国科学技术大学  
University of Science and Technology of China

谢谢！