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**Gibbs-Delaunay Tessellations**  
Simulation and estimation

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# Section 1

## Point processes

We're on  $(\mathbb{R}^d, \mathcal{B})$ .

Denote  $\mathcal{B}_0$  the set of bounded Borel sets.

## Definition. Poisson point process

Let  $\mu$  be a locally finite non-atomic measure on  $\mathbb{R}^d$ . A point process  $\Phi$  satisfying

- $\Phi(B) \sim \text{Pois}(\mu(B))$  for each  $B \in \mathcal{B}_0$ ,
- $\Phi(B_1), \dots, \Phi(B_n)$  for each  $n \in \mathbb{N}$  and  $B_1, \dots, B_n \in \mathcal{B}_0$  pairwise disjoint.

is called a **Poisson point process** with the **intensity measure**  $\mu$ .

If  $\mu = z\lambda^d$  we call the process **homogenous** and  $z$  the **intensity**.

For  $\Lambda \in \mathcal{B}_0$ , denote the distribution of  $\Theta \cap \Lambda$  as  $\pi_\Lambda^z$ .

For the case  $z = 1$ , use  $\pi_\Lambda$ .

$\Phi : (\Omega, \mathcal{A}, P) \rightarrow (\mathcal{F}_{lf}, \mathcal{F})$  where

- $\mathcal{F}_{lf} = \{\gamma \subset \mathbb{R}^d \mid \gamma \cap \Lambda \text{ is finite for all } \Lambda \in \mathcal{B}_0\}$  and
- $\mathcal{F}$  is generated by sets of the form  $\{\gamma \in \mathcal{F}_{lf} \mid N_\Lambda(\gamma) = n\}$ ,  $n \in \mathbb{N}$ ,  $\Lambda \in \mathcal{B}$ , where  $N_\Lambda(\gamma) = \text{Card}(\gamma \cap \Lambda)$ .

We can view  $\pi_\Lambda$  as a reference measure on  $(\mathcal{F}_{lf}, \mathcal{F}, \pi_\Lambda)$ .

Then we can define new point processes through defining their density w.r.t.  $\pi_\Lambda$ .

Poisson point process with intensity  $z$ :

$$\pi_\Lambda^z(d\gamma) \propto z^{N_\Lambda(\gamma)} \pi_\Lambda(d\gamma).$$

Add a new term to obtain the finite volume Gibbs point process:

$$z^{N_\Lambda(\gamma)} e^{-H(\gamma)} \pi_\Lambda(d\gamma).$$

Take  $\Lambda \in \mathcal{B}_0$ .

## Definition. Finite volume Gibbs point process

The **finite-volume Gibbs point process** (fGPP) is a point process defined by its density with respect to  $\pi_\Lambda$ :

$$f(\gamma) = \frac{1}{C_\Lambda^z} z^{N_\Lambda(\gamma)} e^{-H(\gamma)} \quad \gamma \in \mathcal{F}_\Lambda,$$

where

- $z > 0$ ,
- $H : \mathcal{F}_\Lambda \mapsto \mathbb{R} \cup \{+\infty\}$  is a measurable function called the **energy function**,
- $C_\Lambda^z = \int z^{N_\Lambda} e^{-H} d\pi_\Lambda$  is the normalizing constant.

- Physical motivation
- Other examples of energy functions
- Allows working explicitly with geometrical structures such as random tessellations

For  $\gamma \in \mathcal{F}_{lf}$  and  $x \in \mathbb{R}^d$ , define the **local energy** of  $x$  in  $\gamma$  by

$$h(x, \gamma) = H(\gamma \cup \{x\}) - H(\gamma).$$

**Proposition (Georgii, Nguyen, Zessin).** GZN equations

For any positive measurable function  $f : \mathbb{R}^d \times \mathcal{F}_{lf} \rightarrow \mathbb{R}$ ,

$$\int \sum_{x \in \gamma} f(x, \gamma \setminus \{x\}) P_{\Lambda}^z d(\gamma) = z \int \int_{\Lambda} f(x, \gamma) e^{-h(x, \gamma)} dx P_{\Lambda}^z(d\gamma).$$

## Section 2

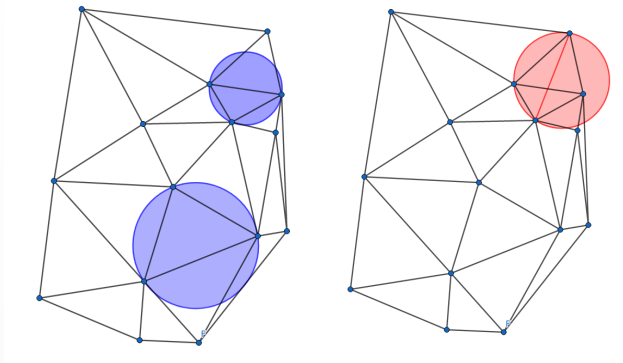
# Triangulations



# Delaunay triangulation

Through empty sphere property

A  $d + 1$ -tuple  $\{x_1, \dots, x_{d+1}\} \subset \gamma$  has the **empty sphere property** if the open circumscribed ball  $B(T)$  does not contain any points from  $\gamma$ .



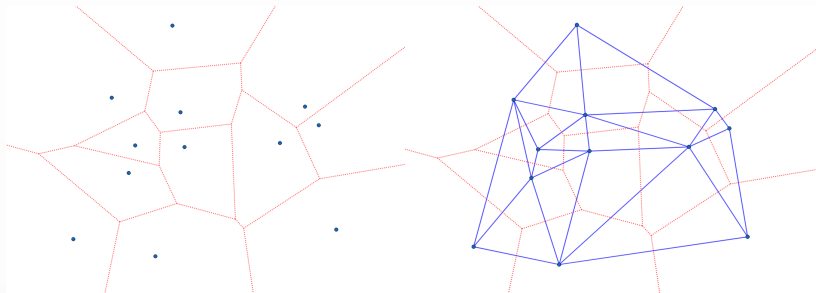
Additional assumption on  $\gamma$  (**No cospherical points**): no  $d + 2$  points  $x_1, \dots, x_{d+2}$  are cospherical, i.e. there is no point  $x \in \mathbb{R}^d$  such that  $d(x, x_1) = \dots = d(x, x_2)$ .

# Delaunay triangulation

Through Voronoi tessellation

For  $x \in \gamma$ , the **Voronoi cell** of  $x$  in  $\gamma$  is

$$C(x, \gamma) = \{z \in \mathbb{R}^d : \|x - z\| \leq \|y - z\| \ \forall y \in \gamma\}.$$

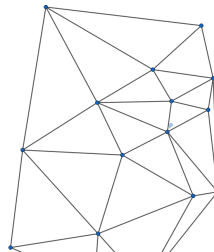
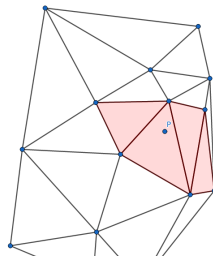
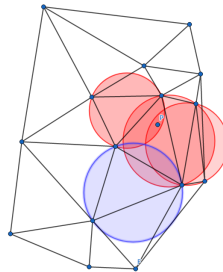
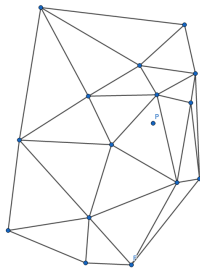


Then the Delaunay tessellation can be defined as

$$Del(\gamma) = \{\{x, y\} \subset \gamma : C(x, \gamma) \cap C(y, \gamma) \neq \emptyset\}.$$

# Delaunay triangulation

Building a Delaunay triangulation



# Delaunay triangulation

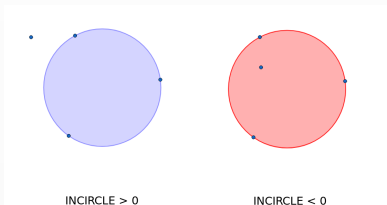
## Geometric predicates, 2D

In 2D

$$\text{INCIRCLE}(P_1, P_2, P_3, P_4) = \begin{vmatrix} x_1 & y_1 & w_1 & 1 \\ x_2 & y_2 & w_2 & 1 \\ x_3 & y_3 & w_3 & 1 \\ x_4 & y_4 & w_4 & 1 \end{vmatrix}$$

where  $w_i = x_i^2 + y_i^2, i = 1, \dots, 4$  and

$$\text{ORIENTATION}(P_1, P_2, P_3) = \begin{vmatrix} x_1 & y_1 & 1 \\ x_2 & y_2 & 1 \\ x_3 & y_3 & 1 \end{vmatrix} > 0$$



# Delaunay triangulation

Geometric predicates, 3D

In 3D

$$INCIRCLE(P_1, P_2, P_3, P_4, P_5) = \begin{vmatrix} x_1 & y_1 & z_1 & w_1 & 1 \\ x_2 & y_2 & z_2 & w_2 & 1 \\ x_3 & y_3 & z_3 & w_3 & 1 \\ x_4 & y_4 & z_4 & w_4 & 1 \\ x_5 & y_5 & z_5 & w_5 & 1 \end{vmatrix}$$

where  $w_i = x_i^2 + y_i^2 + z_i^2, i = 1, \dots, 5$   
if the following condition is satisfied

$$ORIENTATION(P_1, P_2, P_3, P_4) = \begin{vmatrix} x_1 & y_1 & z_1 & 1 \\ x_2 & y_2 & z_2 & 1 \\ x_3 & y_3 & z_3 & 1 \\ x_4 & y_4 & z_4 & 1 \end{vmatrix} > 0$$

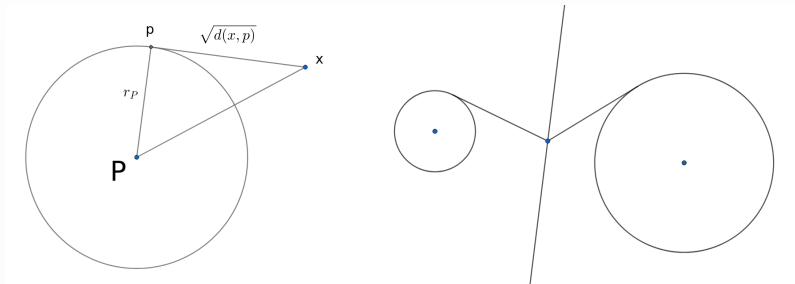
# Laguerre-Delaunay triangulation

Power metric

Generators are not points, but **spheres**.

Metric is not Euclidean, but **power distance**.

$$d(x, P) = d(x, p)^2 - r_P^2$$



# Laguerre-Delaunay triangulation

Inscribed sphere and empty sphere property

# Laguerre-Delaunay triangulation

Geometric predicates, 3D

$$INCIRCLE(P_1, P_2, P_3, P_4, P_5) = \begin{vmatrix} x_1 & y_1 & z_1 & w_1 & 1 \\ x_2 & y_2 & z_2 & w_2 & 1 \\ x_3 & y_3 & z_3 & w_3 & 1 \\ x_4 & y_4 & z_4 & w_4 & 1 \\ x_5 & y_5 & z_5 & w_5 & 1 \end{vmatrix}$$

where  $w_i = x_i^2 + y_i^2 + z_i^2 - r_i^2, i = 1, \dots, 5$   
if the following condition is satisfied

$$ORIENTATION(P_1, P_2, P_3, P_4) = \begin{vmatrix} x_1 & y_1 & z_1 & 1 \\ x_2 & y_2 & z_2 & 1 \\ x_3 & y_3 & z_3 & 1 \\ x_4 & y_4 & z_4 & 1 \end{vmatrix} > 0$$

Why? Because both are **regular triangulations** - convex hulls of lifted sets of points.



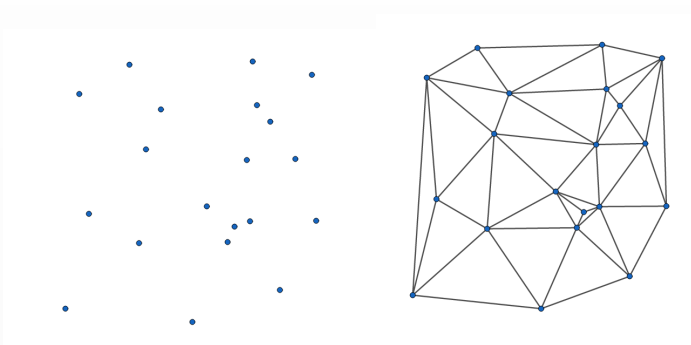


## Section 3

# Random triangulations

# Poisson-Delaunay triangulation

For Poisson point process  $\Phi$ , define **Poisson-Delaunay triangulation** as  $Del(\Phi)$ .



Geometric aspects of the triangulation can be used to define  $H$ .  
In general, the energy can have the form

$$H(\gamma) = \sum_{T \in Del(\gamma)} V_1(T) + \sum_{\{T, T'\} \subset Del(\gamma)} V_2(T, T')$$

to take interaction into account.  $V_1$  and  $V_2$  can be any function from  $d$ -dimension simplices to  $\mathbb{R} \cup \{+\infty\}$ .

Add example(s)?

In the model we used, the energy function is of the form

$$H(\gamma) = \sum_{T \in Del_{\lambda}(\gamma)} V_1(T),$$

with  $V_1$  defined as

$$V_1(T) = \begin{cases} \infty & \text{if } a(T) \leq \epsilon, \\ \infty & \text{if } R(T) \geq \alpha, \\ \theta Sur(T) & \text{otherwise,} \end{cases} \quad (1)$$

where

- $a(T)$  is the area of the smallest face of the tetrahedron  $T$ .
- $R(T)$  is the circumradius of  $T$ .
- $Sur(T)$  is the surface area of the tetrahedron.

# Section 4

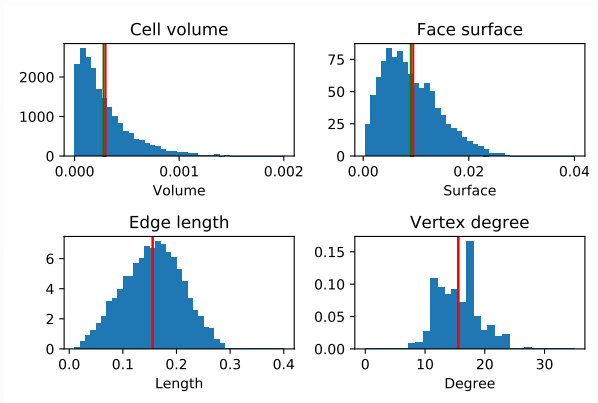
## Simulation



$$\pi_{\Lambda}^z \propto z^{N_{\Lambda}} \pi_{\Lambda}$$

$$P_{\Lambda}^z \propto z^{N_{\Lambda}} e^{-\theta H} \pi_{\Lambda}$$

$\theta = 0 \Rightarrow$  GPP becomes PPP with intensity  $z$ .





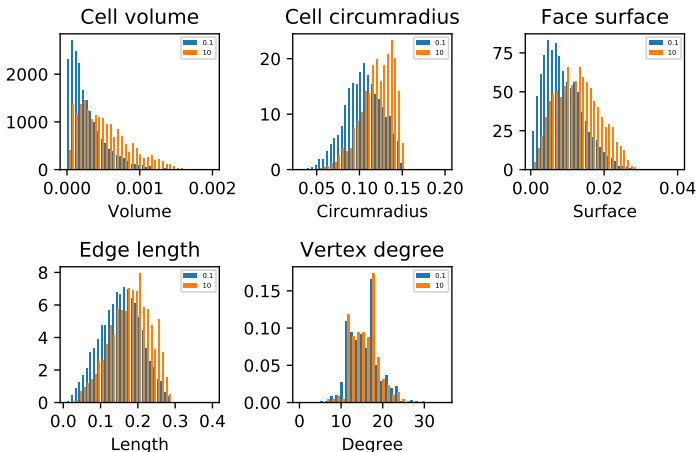


# Role of the parameter $\theta$

$\theta$  positive

The model prefers configurations with lower energy.

$\theta$  multiplies the total surface area of all cells, thus with higher  $\theta$ , the cells are forced to become large.

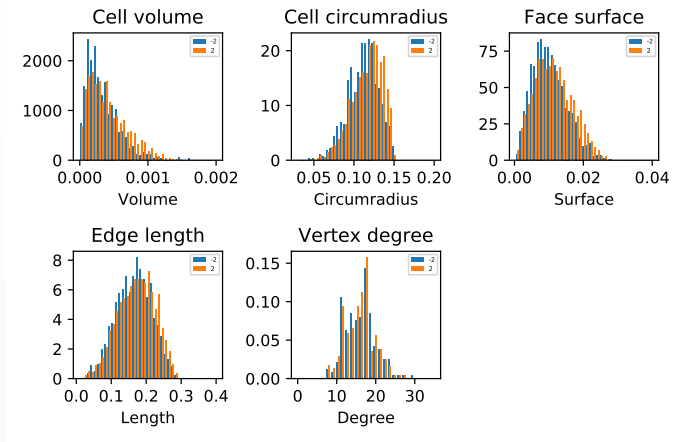


# Role of the parameter $\theta$

$\theta$  negative

The model prefers configurations with lower energy.

- $\theta > 0$ . The sum needs to be minimized  $\Rightarrow$  fewer larger tetrahedra.
- $\theta < 0$ . The sum needs to be maximized  $\Rightarrow$  many smaller tetrahedra.



# Section 5

## Estimation

# Two-step procedure

Thanks to the fact that the hardcore parameter  $\epsilon$  satisfies

$$\text{if } \epsilon > \epsilon' \text{ then } \forall \Lambda, E_{\Lambda}^{\epsilon, \alpha, \theta}(\gamma_{\Lambda}, \gamma_{\Lambda^c}) < \infty \Rightarrow E_{\Lambda}^{\epsilon', \alpha, \theta}(\gamma_{\Lambda}, \gamma_{\Lambda^c}) < \infty,$$

and the hardcore parameter  $\alpha$  satisfies

$$\text{if } \alpha < \alpha' \text{ then } \forall \Lambda, E_{\Lambda}^{\epsilon, \alpha, \theta}(\gamma_{\Lambda}, \gamma_{\Lambda^c}) < \infty \Rightarrow E_{\Lambda}^{\epsilon, \alpha', \theta}(\gamma_{\Lambda}, \gamma_{\Lambda^c}) < \infty,$$

their consistent estimators are:

$$\hat{\epsilon} = \inf\{\epsilon > 0, E_{\Lambda}(\gamma_{\Lambda}, \gamma_{\Lambda}^c) < \infty\},$$

$$\hat{\alpha} = \sup\{\alpha > 0, E_{\Lambda}(\gamma_{\Lambda}, \gamma_{\Lambda}^c) < \infty\}.$$

In practice, the parameters are estimated as

$$\hat{\epsilon} = \min\{a(T), T \in Del_{\Lambda}(\gamma)\},$$

$$\hat{\alpha} = \max\{r(T), T \in Del_{\Lambda}(\gamma)\}.$$

The estimate  $\hat{\beta} = (\hat{\epsilon}, \hat{\alpha})$  is then used in the pseudo-likelihood function in the second estimation step.

$\mathbb{R}^\beta(\gamma)$  is the set of **removable points**.

$$PLL_{\Lambda_n}(\gamma, z, \beta, \theta) = \int_{\Lambda'_n} z \exp(-h^{\beta, \theta}(x, \gamma)) dx + \sum_{x \in \mathcal{R}^\beta(\gamma) \cap \Lambda_n} (h^{\beta, \theta}(x, \gamma \setminus \{x\}) - \ln z)$$

where  $\Lambda'_n$  is the set of all addable points in  $\Lambda_n$  and  $h^{\beta, \theta}(x, \gamma \setminus \{x\})$  is local energy of  $x$  in  $\gamma$  defined for every  $x \in \mathcal{R}^\beta(\gamma)$  by:

$$h^{\beta, \theta}(x, \gamma \setminus \{x\}) = H_{\Lambda}^{\beta, \theta}(\gamma_{\Lambda}, \gamma_{\Lambda^c}) - H_{\Lambda}^{\beta, \theta}(\gamma_{\Lambda} \setminus \{x\}, \gamma_{\Lambda^c}).$$

The estimates  $\hat{\theta}$  and  $\hat{z}$  are obtained through minimizing the  $PLL_{\Lambda_n}$  function.

$$(\hat{z}, \hat{\theta}) = \operatorname{argmin}_{z, \theta} PLL_{\Lambda_n}(\gamma, z, \hat{\beta}, \theta).$$

We obtain the estimate of  $\theta$  by substituting the expression for  $\hat{z}$  into the equation for  $\theta$ . This eventually leads to the equation

$$\int_{\Lambda'_n} \exp(-\theta h(x))(h(x) - c)dx = 0.$$

After  $\hat{\theta}$  is estimated, we then obtain the estimate  $\hat{z}$  with  $\hat{\theta}$  instead of  $\theta$ . All integrals are estimated by MC-integration.

are not great so far.



- Variational estimate
- Energy with explicit interaction
- Periodic outside configuration
- ...