

Delaunay triangulations of manifolds

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Delaunay triangulations of manifolds

- 1 Delaunay triangulations in Euclidean and Laguerre geometry
- 2 Good triangulations
- 3 Computational Topology
- 4 Shape reconstruction
- 5 Delaunay triangulation of manifolds

Bibliography

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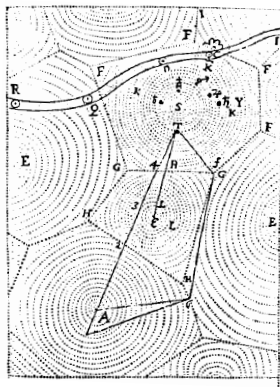
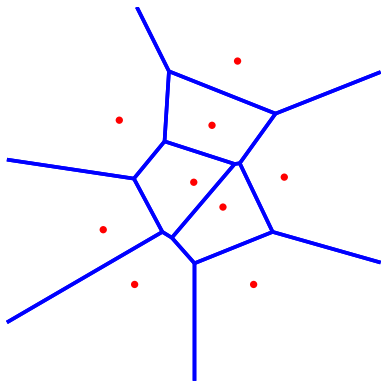
1 Voronoi diagrams and Delaunay triangulations

2 Molecules, Laguerre geometry and affine diagrams

3 Growth models and algebraic varieties

Voronoi diagrams

Fonction distance et croissance



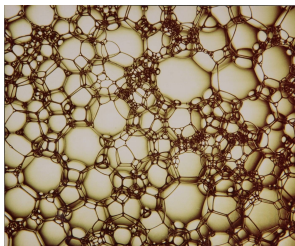
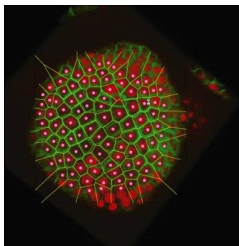
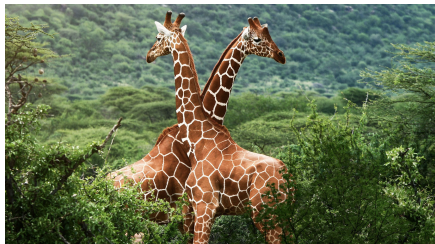
G. Voronoi
(1868-1908)



R. Descartes
(1596-1650)

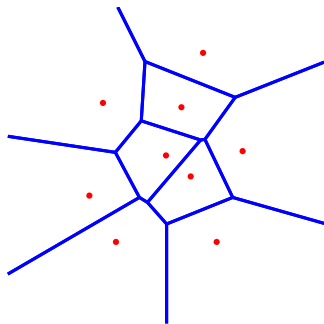
Voronoi diagrams

Voronoi diagrams in nature



Voronoi diagrams

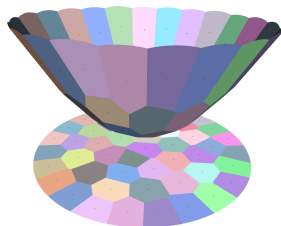
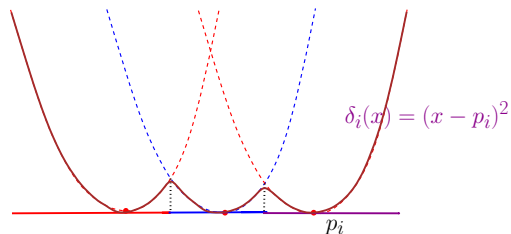
A set of points \mathcal{P} in \mathbb{R}^d



Voronoi cell $V(p_i) = \{x : \|x - p_i\| \leq \|x - p_j\|, \forall j\}$

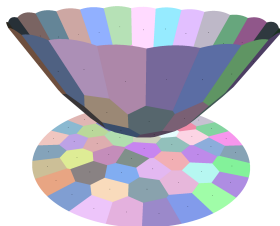
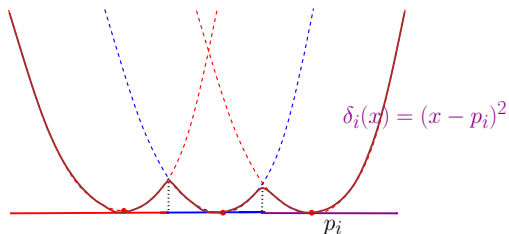
Voronoi diagram (\mathcal{P}) $= \{ \text{set of cells } V(p_i), p_i \in \mathcal{P} \}$

Lower envelopes of functions



- $\text{Vor}(p_i) = \{x : \delta_i(x) \leq \delta_j(x), \forall j\}$
- $\text{Vor}(\mathcal{P})$ is the projection of the lower envelope of the δ_i
= **minimization diagram** of the δ_i
- $\delta_i(x) \leq \delta_j(x) \Leftrightarrow h_{p_i} = p_i \cdot x - p_i^2 \geq h_{p_j} = p_j \cdot x - p_j^2$
- $\text{Vor}(\mathcal{P})$ is the projection of the lower envelope of the h_{p_i}
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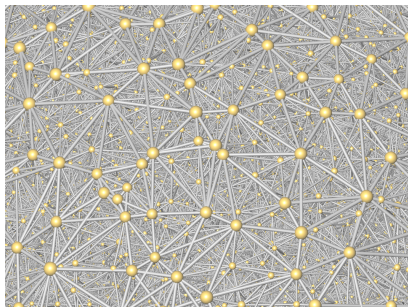
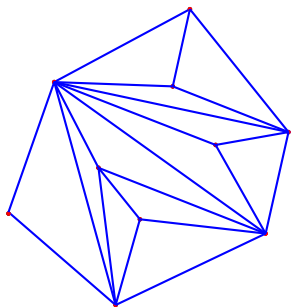
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Triangulations of finite point sets of \mathbb{R}^d

Gluing simplices together

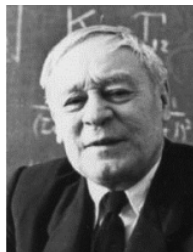
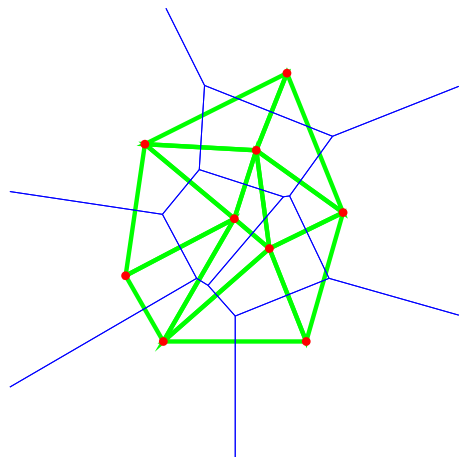


Triangulation of \mathcal{P} : a maximal set of d -simplices s.t.

- the intersection of two simplices is either empty or a common face of the two simplices
- the union of the simplices = $\text{conv}(\mathcal{P})$

Delaunay Triangulations

Sur la sphère vide (On the empty sphere), Boris Delaunay (1934)



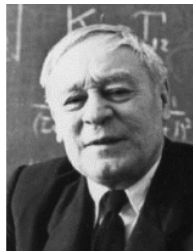
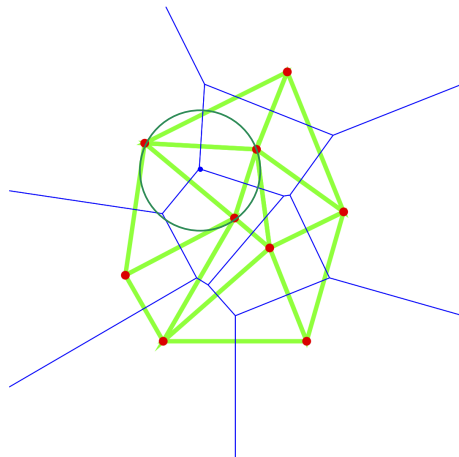
$\text{Del}(\mathcal{P})$ is the **nerve** of $\text{Vor}(\mathcal{P})$

Theorem

If no hypersphere contains $d + 2$ points of \mathcal{P} , alors
 $\text{Del}(\mathcal{P})$ is a triangulation of \mathcal{P}

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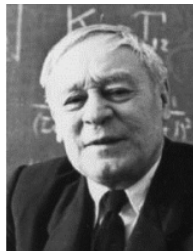
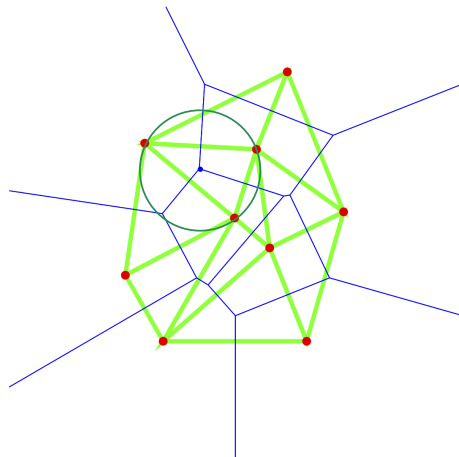
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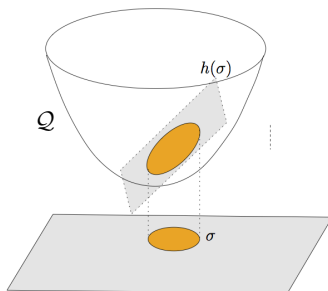
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Proof of Delaunay's theorem

Linearization



σ hypersphere of equation $\sigma(x) = 0$

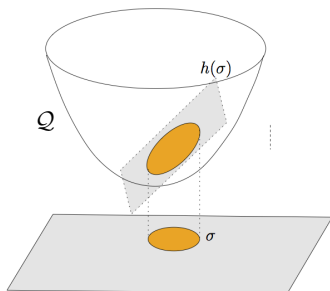
$$\sigma(x) = x^2 - 2c \cdot x + s, \quad s = c^2 - r^2$$

$$\sigma(x) < 0 \Leftrightarrow \begin{cases} z < 2c \cdot x - s & (h_{\sigma}^{-}) \\ z = x^2 & (Q) \end{cases}$$

$$\Leftrightarrow \hat{x} = (x, x^2) \in h_{\sigma}^{-}$$

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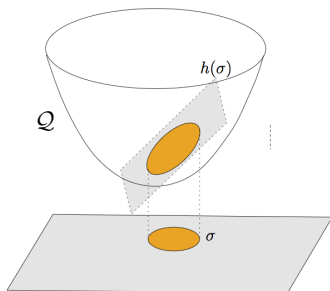
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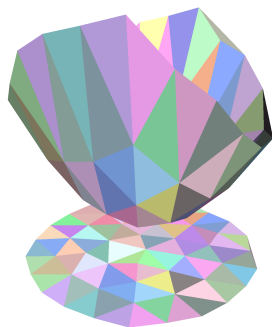
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$$\Leftrightarrow \hat{x} = (x, x^2) \in h_{\sigma}^-$$

Proof of Delaunay's theorem

Delaunay's triangulation and convex hull

\mathcal{P} in general position wrt spheres $\Leftrightarrow \hat{\mathcal{P}}$ in general position



σ a simplex, B_σ its circumscribing ball

$\sigma \in \text{Del}(\mathcal{P}) \Leftrightarrow \forall i, p_i \notin B_\sigma$

$\Leftrightarrow \forall i, \hat{p}_i \in h_\sigma^+ = \text{aff}(\hat{\sigma})$

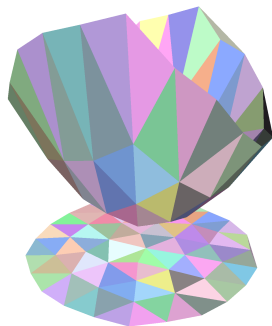
$\Leftrightarrow \hat{\sigma}$ is a face of $\text{conv}^-(\hat{\mathcal{P}})$

$$\text{Del}(\mathcal{P}) = \text{proj}(\text{conv}^-(\hat{\mathcal{P}}))$$

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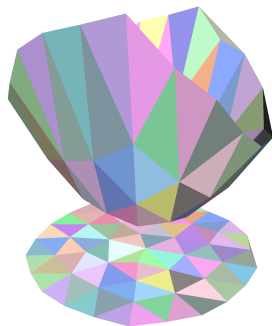
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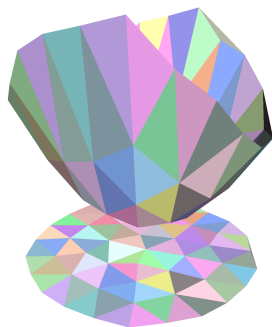
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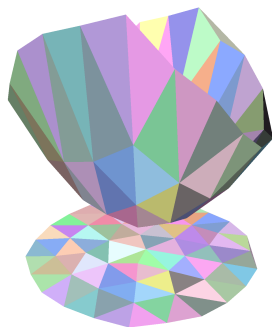
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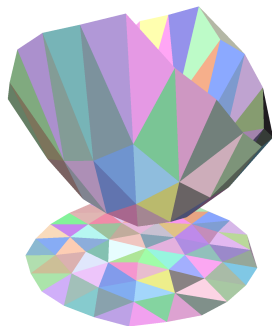
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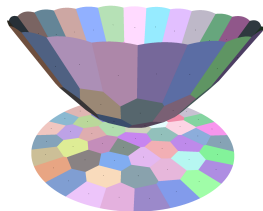
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Correspondence between structures

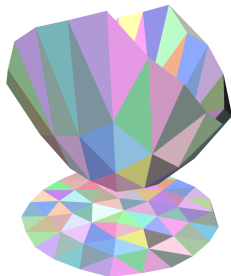
$$h_{p_i} : x_{d+1} = 2p_i \cdot x - p_i^2$$



$$\mathcal{V}(\mathcal{P}) = h_{p_1}^+ \cap \dots \cap h_{p_n}^+$$

Voronoi diagram of \mathcal{P}

$$\hat{p}_i = (p_i, p_i^2) = h_{p_i}^*$$



$$\mathcal{D}(\mathcal{P}) = \text{conv}^-(\{\hat{p}_1, \dots, \hat{p}_n\})$$

Delaunay triang. of \mathcal{P}

duality
 \longrightarrow

nerve
 \longrightarrow

The diagram commutes if \mathcal{P} is in general position wrt spheres

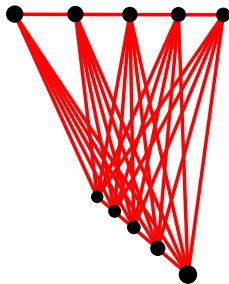
Happy consequences

Combinatorial complexity

The combinatorial complexity of the **Voronoi diagram** of n points of \mathbb{R}^d is the same as the combinatorial complexity of the intersection of n half-spaces of \mathbb{R}^{d+1}

The combinatorial complexity of the **Delaunay triangulation** of n points of \mathbb{R}^d is the same as the combinatorial complexity of the convex hull of n points of \mathbb{R}^{d+1}

The two complexities are the same by duality



$$\Theta(n^{\lceil \frac{d}{2} \rceil})$$

Quadratic in \mathbb{R}^3

Happy consequences

Algorithmic complexity

Construction of $\text{Del}(\mathcal{P})$, $\mathcal{P} = \{p_1, \dots, p_n\} \subset \mathbb{R}^d$

- 1 Lift the points of \mathcal{P} onto the paraboloid $x_{d+1} = x^2$ of \mathbb{R}^{d+1} :
 $p_i \rightarrow \hat{p}_i = (p_i, p_i^2)$
- 2 Compute $\text{conv}(\{\hat{p}_i\})$
- 3 Project the lower hull $\text{conv}^-(\{\hat{p}_i\})$ onto \mathbb{R}^d

Complexity : $\Theta(n \log n + n^{\lfloor \frac{d+1}{2} \rfloor})$ [Clarkson & Shor 1989] [Chazelle 1993]

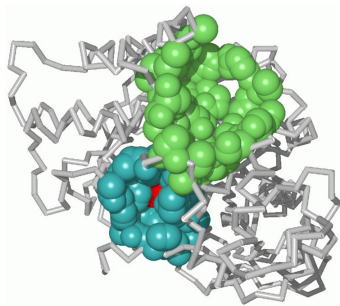
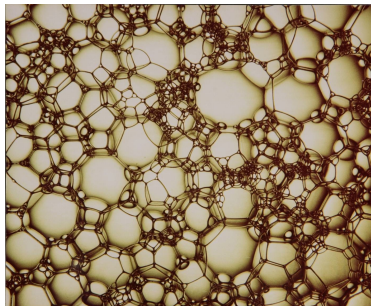
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Foam and molecules

Diagrams of spheres and unions of balls



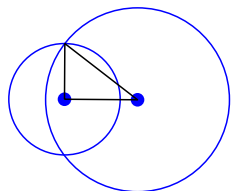
Orthogonal balls

Ball : $b(p, r) = \{x \in \mathbb{R}^d : \|p - x\| \leq r\}$

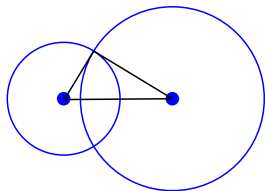
(Hyper)-sphere : $\partial b(p, r) = \{x \in \mathbb{R}^d : \|p - x\| = r\}$

« Distance » between balls : $D(b_1, b_2) = (p_1 - p_2)^2 - r_1^2 - r_2^2$

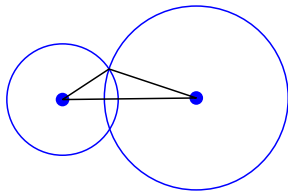
Orthogonal balls : $D(b_1, b_2) = 0$



$$D(b_1, b_2) < 0$$



$$D(b_1, b_2) = 0$$

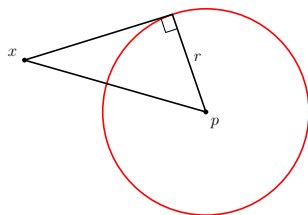


$$D(b_1, b_2) > 0$$

Power of a point wrt to a ball

Power of x wrt b : $D(x, b) = (x - p)^2 - r^2$

Z : D is **not** a distance



$$x \in \text{int}b \iff D(x, b) < 0$$

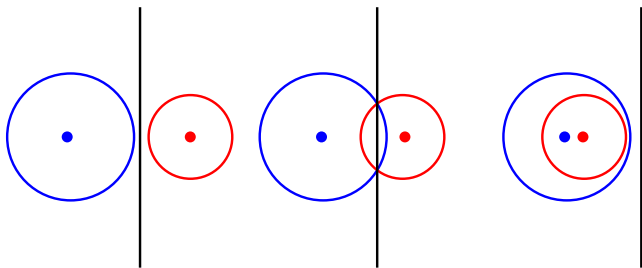
$$x \in \partial b \iff D(x, b) = 0$$

$$x \notin b \iff D(x, b) > 0$$

Radical hyperplane

- The set of points that have a same power wrt two balls $b_1(p_1, r_1)$ and $b_2(p_2, r_2)$ is a **hyperplane**

$$\begin{aligned}D(x, b_1) = D(x, b_2) &\iff (x - p_1)^2 - r_1^2 = (x - p_2)^2 - r_2^2 \stackrel{\text{def}}{=} r_x^2 \\ &\iff -2p_1x + p_1^2 - r_1^2 = -2p_2x + p_2^2 - r_2^2 \\ &\iff 2(p_2 - p_1)x + (p_1^2 - r_1^2) - (p_2^2 - r_2^2) = 0\end{aligned}$$

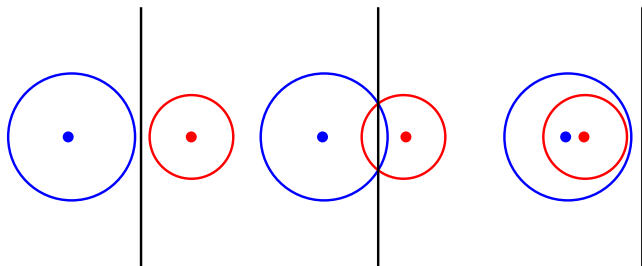


- The radical hyperplane is the set of centres x of the balls $B(x, r_x)$ that are **orthogonal** to b_1 and b_2

Radical hyperplane

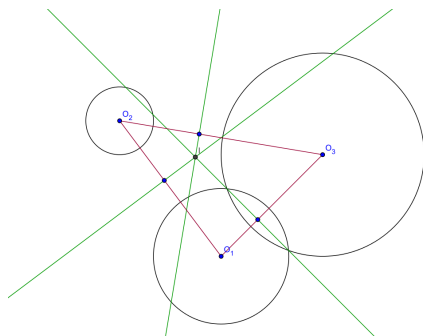
- The set of points that have a same power wrt two balls $b_1(p_1, r_1)$ and $b_2(p_2, r_2)$ is a **hyperplane**

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- The radical hyperplane is the set of centres x of the balls $B(x, r_x)$ that are **orthogonal** to b_1 and b_2

Radical centre



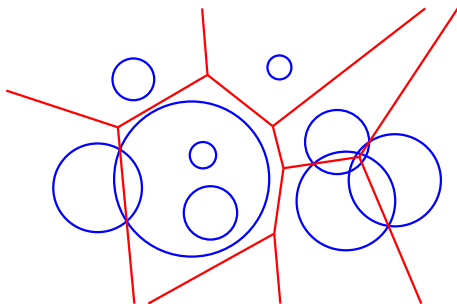
There exists a **unique** point with a same power wrt $d + 1$ balls b_0, \dots, b_d of \mathbb{R}^d

this point is the centre of the unique ball that is orthogonal to b_0, \dots, b_d

Set of balls \mathcal{B} in general position : no ball is orthogonal to $d + 2$ balls of \mathcal{B}

Laguerre (power) diagrams

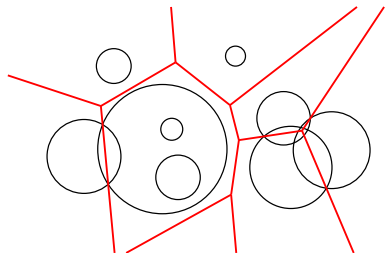
$$\mathcal{B} = \{b_1, \dots, b_n\}$$



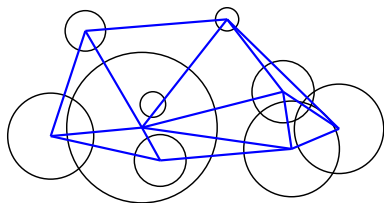
Voronoi cell : $V(b_i) = \{x : D(x, b_i) \leq D(x, b_j) \forall j\}$

Voronoi diagram of \mathcal{B} : $= \{ \text{set of cells } V(b_i), b_i \in \mathcal{B} \}$

Delaunay triangulations of balls



$\text{Vor}(\mathcal{B})$

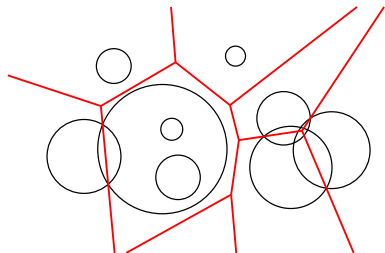


$\text{Del}(\mathcal{B})$ is the **nerve** of $\text{Vor}(\mathcal{B})$

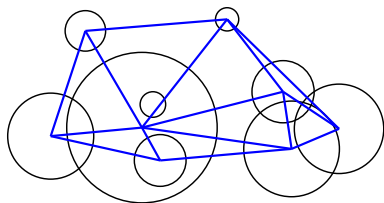
Theorem

If the balls are in general position, then $\text{Del}(\mathcal{B})$ is a triangulation of a subset $\mathcal{P}' \subseteq \mathcal{P}$ of the points

Delaunay triangulations of balls



$\text{Vor}(\mathcal{B})$



$\text{Del}(\mathcal{B})$ is the **nerve** of $\text{Vor}(\mathcal{B})$

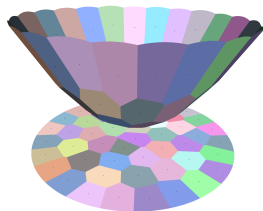
Theorem

If the balls are in general position, then $\text{Del}(\mathcal{B})$ is a triangulation of a subset $\mathcal{P}' \subseteq \mathcal{P}$ of the points

Correspondence between structures

$$h_{b_i} : x_{d+1} = 2p_i \cdot x - p_i^2 + r_i^2$$

$$\hat{b}_i = (p_i, p_i^2 - r_i^2) = h_{b_i}^*$$

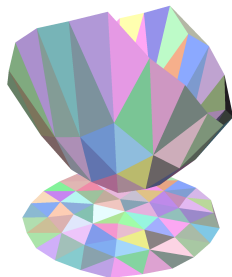


$$\mathcal{V}(\mathcal{B}) = h_{b_1}^+ \cap \dots \cap h_{b_n}^+$$

Voronoi diagram of \mathcal{B}

duality
 \longrightarrow

nerve
 \longrightarrow



$$\mathcal{D}(\mathcal{B}) = \text{conv}^-(\{\hat{b}_1, \dots, \hat{b}_n\})$$

Delaunay triang. of \mathcal{B}

The diagram commutes if \mathcal{B} is in general position

Affine diagrams

Sites + distance functions s.t. the bisectors are hyperplanes

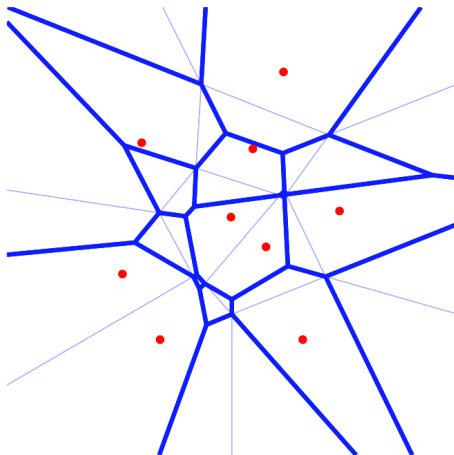
Theorem [Aurenhammer]

Any affine diagram of \mathbb{R}^d is the Laguerre diagram of a set of balls of \mathbb{R}^d

Example : The intersection of a Voronoi diagram with an affine space is a Laguerre diagram

Voronoi diagram of order k

An example of an affine diagram



Each cell is the set of points that have the same k nearest sites

Voronoi diagrams of order k are Laguerre diagrams

S_1, S_2, \dots the subsets of k points of \mathcal{P}

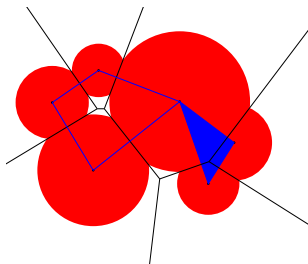
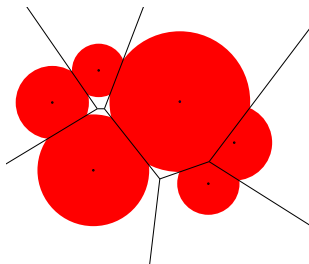
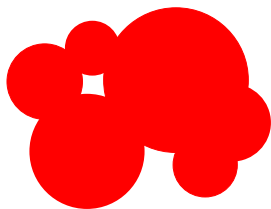
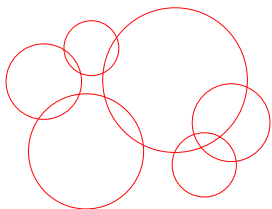
$$\begin{aligned}\delta(x, S_i) &= \frac{1}{k} \sum_{p \in S_i} (x - p)^2 \\ &= x^2 - \frac{2}{k} \sum_{p \in S_i} p \cdot x + \frac{1}{k} \sum_{p \in S_i} p^2 \\ &= D(b_i, x)\end{aligned}$$

where b_i is the ball centered at $c_i = \frac{1}{k} \sum_{p \in S_i} p$

of radius $r_i^2 = c_i^2 - \frac{1}{k} \sum_{p \in S_i} p^2$

$$x \in \text{Vor}_k(S_i) \Leftrightarrow \delta(x, S_i) \leq \delta(x, S_j) \quad \forall j$$

Delaunay triangulation restricted to a molecule

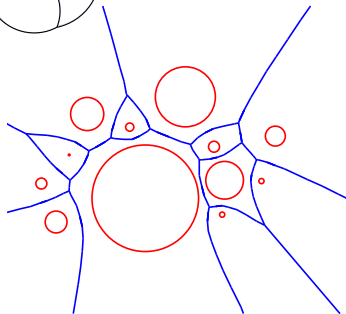
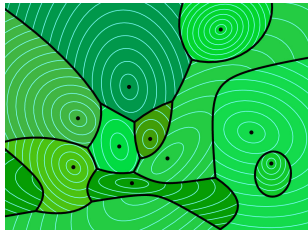
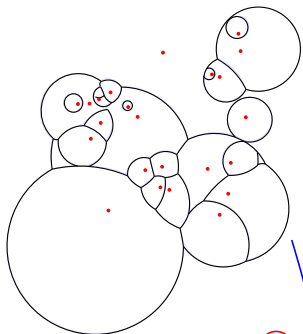


$$C(b) = b \cap V(b)$$
$$U = \bigcup_{b \in \mathcal{B}} C(b)$$

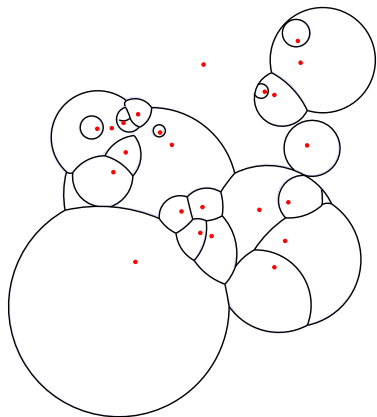
$$\text{Vor}_{|U}(\mathcal{B}) = \{f \in \text{Vor}(\mathcal{B}), f \cap U \neq \emptyset\}$$
$$\text{Del}_{|U}(\mathcal{B})$$

- 1 Voronoi diagrams and Delaunay triangulations
- 2 Molecules, Laguerre geometry and affine diagrams
- 3 Growth models and algebraic varieties**

Distance functions and growth models



Möbius diagrams



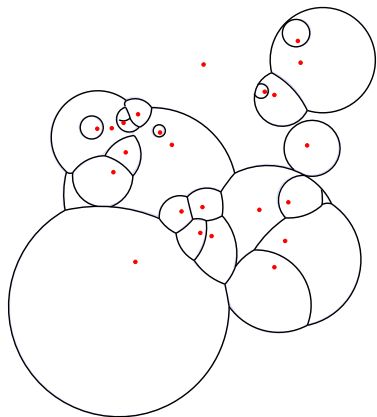
$$W_i = (p_i, \lambda_i, \mu_i)$$

$$\delta_M(x, W_i) = \lambda_i \|x - p_i\|^2 - \mu_i$$

$$\text{Mob}(W_i) = \{x, \delta(x, \sigma_i) \leq \delta(x, \sigma_j)\}$$

Bisectors are **hyperspheres** (hyperplanes or \emptyset)

Möbius diagrams



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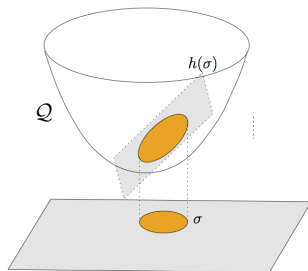
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Bisectors are **hyperspheres** (hyperplanes or \emptyset)

Linearization

A Möbius diagram of \mathbb{R}^d is the restriction of an affine diagram of \mathbb{R}^{d+1}



Lift the spherical bisectors onto \mathcal{Q} and take the polar hyperplanes

The hyperplanes define an affine diagram $\text{Vor}(\mathcal{B})$ in \mathbb{R}^{d+1}

The faces of the Möbius diagram are the projections of the faces of $\text{Vor}(\mathcal{B}) \cap \mathcal{Q}$

Corollaries

- 1 *Any spherical diagram (i.e. whose bisectors are hyperspheres) is a Möbius diagram*
- 2 *The set of Möbius diagrams is stable under Möbius transformation*
- 3 *The intersection of a spherical diagram with an affine subspace is a spherical diagram*
- 4 \boxed{Z} : *the nerve of a Möbius diagram has a realization in \mathbb{R}^{d+1} but is not (in general) a triangulation of \mathbb{R}^d*

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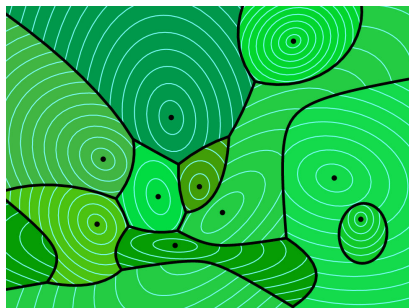
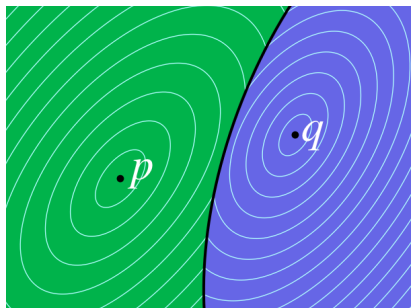
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Anisotropic Voronoi diagrams

Metric at p : M_p : $d \times d$ matrix that is symmetric, positive definite

$$d_p(x, y) = \sqrt{(x - y)^t M_p (x - y)}$$



[Labelle & Shewchuk 2003]

$$V(p) = \{x : d_p(x, p) \leq d_q(x, q) \text{ for all } p, q \in P\}$$

Linearization

To any point $x = (x_1, \dots, x_d) \in \mathbb{R}^d$, associate the two points

$$\tilde{x} = (x_i x_j, 1 \leq i \leq j \leq d) \in \mathbb{R}^{\frac{d(d+1)}{2}}$$

$$\hat{x} = (x, \tilde{x}) \in \mathbb{R}^D, \quad D = \frac{d(d+3)}{2}$$

- Observe that $\mathcal{Q} = \{\hat{x}, x \in \mathbb{R}^d\}$ is a « surface » of dimension d in \mathbb{R}^D

- By elementary calculations : $d_p(x, p)^2 = -2\hat{p}^t \hat{x} + p^t M_p p$

which implies

$$d_p(x, p) < d_q(x, q) \iff (\hat{x} - \hat{p})^2 - (\hat{p}^2 - p^t M_p p) < (\hat{x} - \hat{q})^2 - (\hat{q}^2 - q^t M_q q)$$

Theorem

The anisotropic Voronoi diagram of \mathcal{P} is the projection of the restriction of the Laguerre diagram of a set of n balls restricted to \mathcal{Q}

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Information geometry

Statistical spaces

A point represents a probability density function (pdf), for example the isotropic gaussian defined in \mathbb{R}^d

$$f(x, \mu, \sigma) = \frac{1}{\sqrt{2\pi}\sigma} \exp\left(\frac{-\|x - \mu\|}{2\sigma^2}\right)$$

can be represented by the point (μ, σ) in the space

$$H = \{(\mu, \sigma) \in \mathbb{R}^{d+1}, \sigma > 0\}$$

- What distance in those spaces ?
- Can we define and construct Voronoi diagrams in statistical spaces ?

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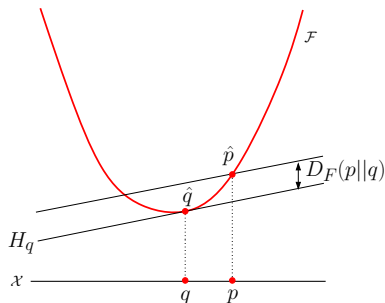
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Bregman divergences

F a strictly convex and differentiable function defined on a convex set \mathcal{X}

$$D_F(\mathbf{p}, \mathbf{q}) = F(\mathbf{p}) - F(\mathbf{q}) - \langle \mathbf{p} - \mathbf{q}, \nabla F(\mathbf{q}) \rangle$$



D_F is **not** a distance but $D_F(\mathbf{p}, \mathbf{q}) \geq 0$ and $D_F(\mathbf{p}, \mathbf{q}) = 0$ iff $\mathbf{p} = \mathbf{q}$

Examples of Bregman divergences

- $F(x) = x^2$: Squared euclidean distance

$$D_F(\mathbf{p}, \mathbf{q}) = F(\mathbf{p}) - F(\mathbf{q}) - \langle \mathbf{p} - \mathbf{q}, \nabla F(\mathbf{q}) \rangle = \mathbf{p}^2 - \mathbf{q}^2 - \langle \mathbf{p} - \mathbf{q}, 2\mathbf{q} \rangle = \|\mathbf{p} - \mathbf{q}\|^2$$

- $F(p) = \sum p(x) \log_2 p(x)$ (Shannon entropy)
 $D_F(p, q) = \sum_x p(x) \log_2 \frac{p(x)}{q(x)}$ (K-L divergence)
- $F(p) = -\sum_x \log p(x)$ (Burg entropy)
 $D_F(p, q) = \sum_x \left(\frac{p(x)}{q(x)} \log \frac{p(x)}{q(x)} - 1 \right)$ (Itakura-Saito)

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Bregman diagrams

[Boissonnat, Nielsen, Nock 2010]

$$D_F(\mathbf{p}, \mathbf{q}) = F(\mathbf{p}) - F(\mathbf{q}) - \langle \mathbf{p} - \mathbf{q}, \nabla F(\mathbf{q}) \rangle$$

Two types of bisectors

$$H_{pq} : D_F(\mathbf{x}, \mathbf{p}) = D_F(\mathbf{x}, \mathbf{q}) \quad (\text{hyperplane})$$

$$H_{pq}^* : D_F(\mathbf{p}, \mathbf{x}) = D_F(\mathbf{q}, \mathbf{x}) \quad (\text{hypersurface})$$

Bregman diagrams

- Two types of Bregman diagrams
- By the Legendre duality : $D_F(\mathbf{x}, \mathbf{y}) = D_{F^*}(\mathbf{y}', \mathbf{x}')$ ($\mathbf{x}' = \nabla F(\mathbf{x})$)

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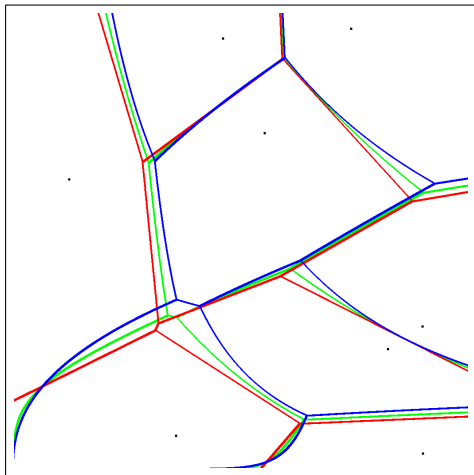
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Bregman diagrams and Laguerre diagrams

The Bregman diagram of the **1st type** of a set of n sites of \mathcal{P} is identical to the **Laguerre** diagram of n euclidian balls centered at the points \mathbf{p}'_i

$$\begin{aligned} & D_F(\mathbf{x}, \mathbf{p}_i) \leq D_F(\mathbf{x}, \mathbf{p}_j) \\ \iff & -F(\mathbf{p}_i) - \langle \mathbf{x} - \mathbf{p}_i, \mathbf{p}'_i \rangle \leq -F(\mathbf{p}_j) - \langle \mathbf{x} - \mathbf{p}_j, \mathbf{p}'_j \rangle \\ \iff & \langle \mathbf{x}, \mathbf{x} \rangle - 2\langle \mathbf{x}, \mathbf{p}'_i \rangle - 2F(\mathbf{p}_i) + 2\langle \mathbf{p}_i, \mathbf{p}'_i \rangle \leq \langle \mathbf{x}, \mathbf{x} \rangle - 2\langle \mathbf{x}, \mathbf{p}'_j \rangle - 2F(\mathbf{p}_j) + 2\langle \mathbf{p}_j, \mathbf{p}'_j \rangle \\ \iff & \langle \mathbf{x} - \mathbf{p}'_i, \mathbf{x} - \mathbf{p}'_i \rangle - r_i^2 \leq \langle \mathbf{x} - \mathbf{p}'_j, \mathbf{x} - \mathbf{p}'_j \rangle - r_j^2 \end{aligned}$$

où $r_l^2 = \langle \mathbf{p}'_l, \mathbf{p}'_l \rangle + 2(F(\mathbf{p}_l) - \langle \mathbf{p}_l, \mathbf{p}'_l \rangle)$

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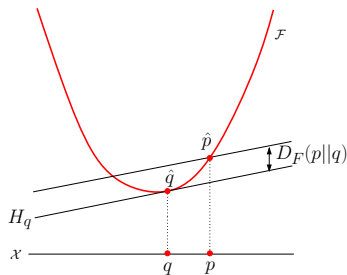
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Bregman spheres

Definition : $\sigma(\mathbf{c}, r) = \{\mathbf{x} \in \mathcal{X} \mid D_F(\mathbf{x}, \mathbf{c}) = r\}$

Lemma The image $\hat{\sigma}$ of a Bregman sphere σ by the lifting map onto \mathcal{F} is contained in a hyperplane H_σ

Conversely, the intersection of any hyperplane H with \mathcal{F} projects vertically onto a Bregman sphere



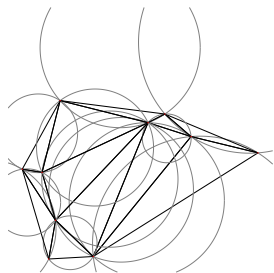
Lemma A d -simplex has a unique circumscribing Bregman hypersphere

Bregman triangulations

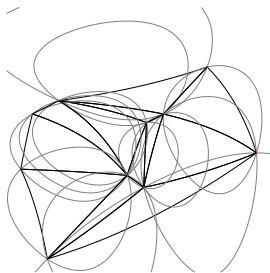
Definition : the Bregman triangulation of \mathcal{P} , $BT(\mathcal{P})$ is the nerve of the Bregman diagram of \mathcal{P} of the 1st type and therefore also of a Laguerre diagram of \mathcal{P}'

Characteristic property : The Bregman sphere circumscribing any simplex of $BT(\mathcal{P})$ is empty

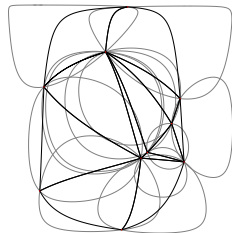
Examples



(a) Ordinary Delaunay

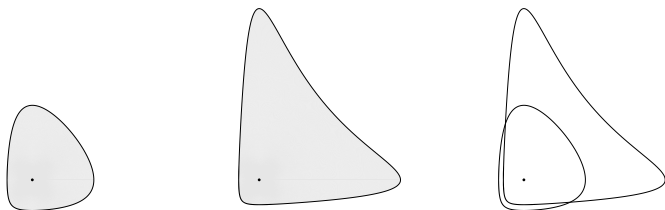


(b) Exponential loss



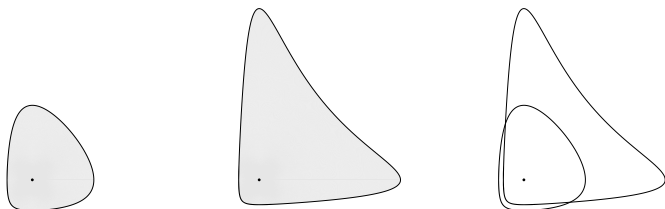
(c) Hellinger-like divergence

Unions of Bregman balls of types 1 and 2



- The **combinatorial and algorithmic complexity** of a union of Bregman balls is the same as for euclidean balls
- The same is true for unions of balls of the 2nd type (via Legendre transform which is a homeomorphism (if F is a Legendre function))

Unions of Bregman balls of types 1 and 2



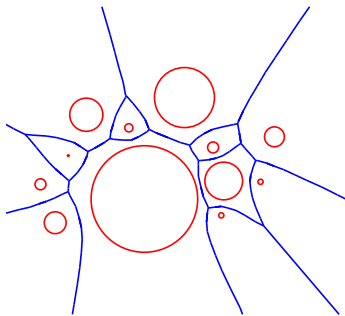
- The **combinatorial and algorithmic complexity** of a union of Bregman balls is the same as for euclidean balls
- The same is true for unions of balls of the 2nd type (via Legendre transform which is a homeomorphism (if F is a Legendre function))

Next lectures

- Practical combinatorial and algorithmic complexity
- More general metrics
- Triangulation of surfaces and other curved spaces

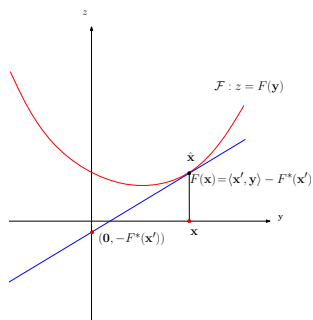
Open question

- Combinatorial complexity of the additively weighted Voronoi diagram



Legendre duality

Convex conjugate : $F^*(x') = x \cdot x' - F(x)$



Gradient space : $\Omega' = \{\nabla F(x), x \in \Omega\}$

F is a function of Legendre type if Ω' is convex

Legendre duality

Properties of functions of Legendre type

- $(F^*)^* = F$
- F^* is strictly convex and differentiable
- Writing $\mathbf{y}' = \nabla_F(\mathbf{y})$:
$$F^*(\mathbf{y}') = -F(\mathbf{y}) + \langle \mathbf{y}, \mathbf{y}' \rangle$$
$$\nabla_{F^*} = \nabla_F^{-1}$$

$$\begin{aligned} D_F(\mathbf{x}, \mathbf{y}) &= F(\mathbf{x}) - F(\mathbf{y}) - \langle \mathbf{x} - \mathbf{y}, \mathbf{y}' \rangle \\ &= -F^*(\mathbf{x}') + \langle \mathbf{x}, \mathbf{x}' \rangle + F^*(\mathbf{y}') - \langle \mathbf{x}, \mathbf{y}' \rangle \\ &= D_{F^*}(\mathbf{y}', \mathbf{x}') \end{aligned}$$