

Randomized Algorithms

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Back to Incremental Convex Hull

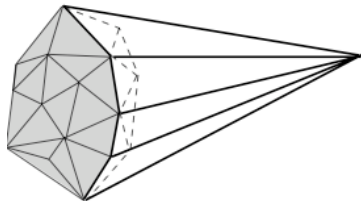
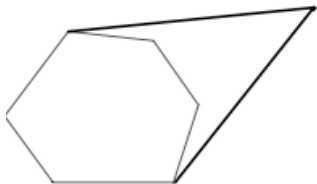
\mathcal{P} a set of n points in \mathbb{R}^d

- 1 Sort \mathcal{P} by lexicographic order

$$\mathcal{P} = \{p_1, \dots, p_n\}$$

- 2 For $i = 1, \dots, n$,

$$\text{insert } p_{i+1}: \text{conv}(\mathcal{P}_i) \longrightarrow \text{conv}(\mathcal{P}_i \cup p_{i+1})$$



Back to Incremental Convex Hull

$$\mathcal{P}_i = \{p_1, \dots, p_i\}$$

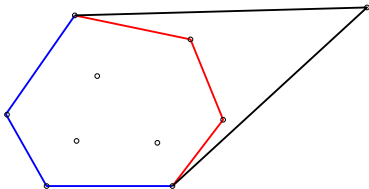
Updating $\text{conv}(\mathcal{P}_i)$ into $\text{conv}(\mathcal{P}_i \cup p_{i+1})$: a color story

Facet A facet f of $\text{conv}(\mathcal{P}_i)$ with supporting hyperplan h_f is:

- **red** iff $\text{conv}(\mathcal{P}_i)$ and p_{i+1} are on different side of h_f .
- **blue** iff $\text{conv}(\mathcal{P}_i)$ and p_{i+1} are on the same side of h_f .

k -Faces with $k < d - 1$ of $\text{conv}(\mathcal{P}_i)$ is:

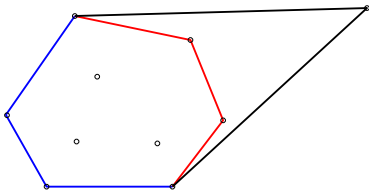
- **red** iff it is the intersection of **red** facets
- **blue** iff it is the intersection of **blue** facets
- **purple** iff it is included in **red** and **blue** facets



Back to Incremental Convex Hull

Updating $\text{conv}(\mathcal{P}_i)$ into $\text{conv}(\mathcal{P}_i \cup p_{i+1})$:

- 1 find a first **red** facets (there is one incident to p_i)
- 2 find the set of **red** facets (they form a connected set)
- 3 delete **red** faces
install new faces = $\text{conv}(e \cup p_{i+1})$
where e is a **purple** face of $\text{conv}(\mathcal{P}_i)$.

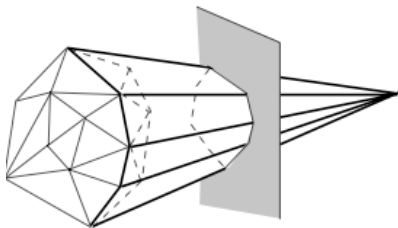


Back to Incremental Convex Hull

Complexity

Sorting + the total number of created facets

$$O\left(n \log n + \sum_{i=1}^n i \lfloor \frac{d-1}{2} \rfloor\right) = O\left(n \log n + n \lfloor \frac{d+1}{2} \rfloor\right)$$

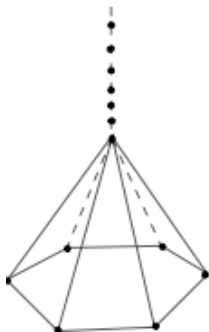


Randomized incremental convex hull: $O\left(n \log n + n \lfloor \frac{d}{2} \rfloor\right)$

Optimal convex hull: $O\left(n \log n + n \lfloor \frac{d}{2} \rfloor\right)$

Back to Incremental Convex Hull

A bad case in 3D



The bad case arises with a special set of points inserted in a special order.
Hence the idea of randomization.

Randomized Algorithms

What is a randomized algorithms

- A randomized algorithm computes the exact solution of a deterministic problem
- The algorithm performs internal random choices. that influence the algorithm's behaviour but not the output.
- Analysis of the algorithm: expectation over the random choices.

Randomized incremental algorithms

random choices: random insertion order of the data

Randomized analysis of on line algorithms

- An on line algorithm handles data according to the given input order.
- Expected complexity with respect to a random input order.

The Formalism of Regions and Conflicts

Data : a finite set \mathcal{P} of objects in \mathcal{O}

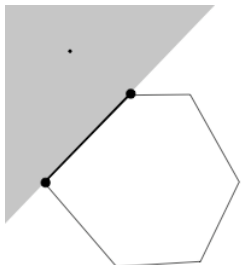
Region - a subset of $i < b$ objects that defines the region
- a subset of \mathcal{O} in conflict with the region

Problem of interest build the set of regions defined by \mathcal{P}
without conflict on \mathcal{P}

Regions and Conflicts for the Convex hull

- objects are points in \mathbb{R}^d
- regions are halfspaces of \mathbb{R}^d
- each region is defined by d -points
- p in conflict with h^+ iff $p \in h^+$

facets of $\text{conv}(\mathcal{P}) = \left\{ \begin{array}{l} \text{set of regions} \\ \text{defined by } \mathcal{P} \\ \text{with no conflict on } \mathcal{P} \end{array} \right.$



Randomized Incremental Construction

Notations

Let \mathcal{P} be a finite set of objects.

$\mathcal{F}(\mathcal{P})$ the set of regions defined by objects in \mathcal{P}

$\mathcal{F}_j(\mathcal{P})$ the set of regions defined by objects in \mathcal{P}
in conflict with j objects in \mathcal{P} .

Incremental construction of $\mathcal{F}_0(\mathcal{P})$.

At each step:

- a new point p of \mathcal{P} is added in a subset $\mathcal{R} \subset \mathcal{P}$
- the set $\mathcal{F}_0(\mathcal{R})$ is updated.

Randomized hypothesis

The order of insertion of objects in \mathcal{P} is random.

At each step r , \mathcal{R} is a r -sample of \mathcal{P} , i.e. a random subset with size r .

Randomized Incremental Construction

Expected number of constructed regions

\mathcal{R} random subset of \mathcal{P} with size r

$f_0(r)$ expected number of regions defined on \mathcal{R} and without conflict on \mathcal{R}

Theorem (First RIC theorem)

The expected total number of regions constructed by the RIC is:

$$O\left(\sum_{r=1}^n \frac{f_0(r)}{r}\right)$$

Proof.

probability for $f \in \mathcal{F}(\mathcal{P})$ to be in $\mathcal{F}_0(\mathcal{R})$: $p_f(r)$

Probability for $f \in \mathcal{F}(\mathcal{P})$ to appear in $\mathcal{F}_0(\mathcal{R})$ at step r : $\frac{b}{r}p_f(r)$

$$f_0(r) = \sum_{f \in \mathcal{F}(\mathcal{P})} p_f(r)$$

$$\begin{aligned} \text{Expected total number of regions} \\ \text{constructed by the RIC} \end{aligned} = \sum_r \sum_f \frac{b}{r} p_f(r) = O\left(\sum_{r=1}^n \frac{f_0(r)}{r}\right)$$

Randomized Incremental Convex Hull

Expected number of constructed regions

In the convex hull case $f_0(r) = O(r^{\lfloor \frac{d}{2} \rfloor})$

$$\begin{array}{l} \text{Expected total number of regions} \\ \text{constructed by the convex hull RIC} \end{array} = O\left(\sum_{r=1}^n \frac{f_0(r)}{r}\right) = O\left(n^{\lfloor \frac{d}{2} \rfloor}\right)$$

This not the end of the story for the RIC of convex hulls:
since points are inserted in random order
we can no more rely on the lexicographic order
to find the first *red* facet at each insertion.

Randomized Incremental Convex hull

A first solution : the conflict graph

The conflict graph

A bipartite graph on $\mathcal{P} \times \mathcal{F}_0(\mathcal{R})$:

$\forall p \in \mathcal{P} \setminus \mathcal{R}$, an edge (p, f) where $f \in \mathcal{F}_0(\mathcal{R})$ conflicts with p .

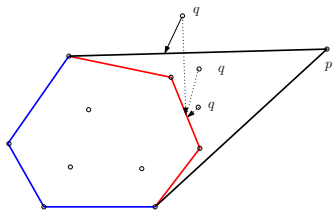
Using conflict graph

$\text{conv}(\mathcal{R}) \longrightarrow \text{conv}(\mathcal{R} \cup p)$

- the conflict graph edge of p provides the first **red** facet.
- Find all the **red** facets by walking on facets of $\text{conv}(\mathcal{R})$ in conflict with p
- Update the convex hull
- Update the conflict graph
 - for each $q \neq p$ in $\mathcal{P} \setminus \mathcal{R}$, find a new conflict graph edge by walking on facets of $\text{conv}(\mathcal{R})$ conflicting with p and q

Randomized Incremental Convex hull

Updating the conflict graph



Update the conflict graph

Insertion of $p : \text{conv}(\mathcal{R}) \longrightarrow \text{conv}(\mathcal{R} \cup p)$

Let $q \in \mathcal{P} \setminus \mathcal{R}$, $q \neq p$.

Walk on the facets f of $\text{conv}(\mathcal{R})$ in conflict with p and q

- if a neighbor g of f conflicts with q but not with p : $\text{edge}(q, g)$

- if a neighbor g of f conflicts neither with q nor f

f' neighbor of g in $\text{conv}(\mathcal{R} \cup p)$ s.t. $g \cap f = g \cap f'$

$\text{edge}(q, f')$ if q and f' conflict.

- if neither happens, discard q .

Randomized Incremental Constructions

Conflict graph complexity

Total complexity of convex hull

- Update of convex hulls : total number of constructed regions
- Finding conflicts + updating the conflict graphs :
total number of conflicts with constructed regions

Theorem (Second RIC theorem)

The expected total number of conflicts with constructed regions is:

$$O\left(\sum_{r=1}^n \frac{n-r}{r^2} f_0(r)\right)$$

Expected complexity of convex hull RIC

$$f_0(r) = O\left(r^{\lfloor \frac{d}{2} \rfloor}\right), \quad O\left(\sum_r \frac{n-r}{r^2} f_0(r)\right) = O\left(n \log n + n^{\lfloor \frac{d}{2} \rfloor}\right)$$

Proof of 2d RIC Theorem

Statistics on regions

$\mathcal{R} \subset \mathcal{P}$

$\mathcal{F}_j(\mathcal{P})$ the set of regions defined by \mathcal{P} with j conflicts in \mathcal{P} .

$\mathcal{F}_j(\mathcal{R})$ the set of regions defined by $\mathcal{R} \subset \mathcal{P}$ with j conflicts in \mathcal{R} .

$f_j(r)$ expected size of $\mathcal{F}_j(\mathcal{R})$.

\mathcal{R} being a random sample
probability for $f \in \mathcal{F}_j(\mathcal{P})$
to be in $\mathcal{F}_k(\mathcal{R})$

$$p_{j,k}(r) = \frac{\binom{j}{k} \binom{n-b-j}{r-b-k}}{\binom{n}{r}}$$

$$f_k(r) = \sum_j |\mathcal{F}_j(\mathcal{P})| p_{j,k}(r)$$

$$f_0(r) = \sum_j |\mathcal{F}_j(\mathcal{P})| p_j(r)$$

Proof of 2d RIC Theorem

Moments

$\mathcal{R} \subset \mathcal{P}$

$\mathcal{P}(f)$ set of objects in \mathcal{P} in conflict with region f

Moment of order k of \mathcal{R} with respect to \mathcal{P}

$$m_k(\mathcal{R}, \mathcal{P}) = \sum_{f \in \mathcal{F}_0(\mathcal{R})} \binom{|\mathcal{P}(f)|}{k}.$$

Expectation for a r -sample

$$m_k(r) = \sum_f \binom{|\mathcal{P}(f)|}{k} \text{proba}(f \in \mathcal{F}_0(\mathcal{R})) = \sum_j |\mathcal{F}_j(\mathcal{P})| \binom{j}{k} p_j(r).$$

RIC and first order moment

Expected total number of conflicts with constructed regions

$$\sum_r \sum_j |\mathcal{F}_j(\mathcal{P})| \binom{j}{1} \frac{b}{r} p_j(r) = \sum_r b \frac{m_1(r)}{r}$$

Proof of 2d RIC Theorem

Moments Theorem

Theorem (Moments theorem)

$$m_k(\mathcal{R}) \leq f_k(\mathcal{R}) \frac{(n-r+k)!}{(n-r)!} \frac{(r-b-k)!}{(r-b)!}$$

Proof
 $m_k(r) =$

$$\sum_j |\mathcal{F}_j(\mathcal{P})| \binom{j}{k} \frac{\binom{n-b-j}{r-b}}{\binom{n}{r}}$$

$f_k(r) =$

$$\sum_j |\mathcal{F}_j(\mathcal{P})| \binom{j}{k} \frac{\binom{n-b-j}{r-b-k}}{\binom{n}{r}}$$

First order moment : $m_1(r) = f_1(r) \frac{n-r+1}{r-b} = O\left(\frac{n-r}{r} f_1(r)\right)$

Proof of 2d RIC Theorem

Backward Analysis

Theorem

$$f_1(r) = O(f_0(r))$$

\mathcal{R} a subset of \mathcal{P} of size r

Consider a random sample \mathcal{R}' of \mathcal{R} of size $r - 1$

$$\text{Exp}(|\mathcal{F}_0(\mathcal{R}')|) = \frac{1}{r}|\mathcal{F}_1(\mathcal{R})| + \frac{r-b}{r}|\mathcal{F}_0(\mathcal{R})|$$

$$f_0(r-1) = \frac{1}{r}f_1(r) + \frac{r-b}{r}f_0(r)$$

$$\frac{b}{r}f_0(r) \geq \frac{1}{r}f_1(r)$$

(assuming that $f_0(r)$ is a growing function of r).



Proof of 2d RIC Theorem

Summary

RIC and first order moment

Expected total number of conflicts with constructed regions

$$\sum_r \sum_j |\mathcal{F}_j(\mathcal{P})| \binom{j}{1} \frac{b}{r} p_j(r) = \sum_r b \frac{m_1(r)}{r}$$

First order moment theorem

$$m_1(r) = f_1(r) \frac{n-r+1}{r-b} = O\left(\frac{n-r}{r} f_1(r)\right)$$

Backward Analysis

$$f_1(r) = O(f_0(r))$$

Conclusion

The expected total number of conflicts with constructed regions is:

$$O\left(\sum_{r=1}^n \frac{n-r}{r^2} f_0(r)\right)$$

On Line Algorithms

The influence graph

- A directed acyclic connected graph,
- with one node for each constructed region.
- The region of a node is included in the union of the regions of its parents

Localisation

To find the conflicts when inserting a new object p visit all the nodes in the influence graph in conflict with p

Randomised complexity: $O\left(\sum_{r=1}^n \frac{n-r}{r^2} f_0(r)\right)$
provided that the outdegree of each node is bounded.

On Line Convex Hull

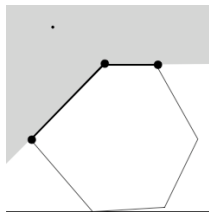
Regions are now defined by $d + 1$ points: $\{p_0, p_1, \dots, p_d\}$

h_0^+ halfspace bounded by hyperplan through $\{p_1, \dots, p_d\}$
and not including p_0

h_d^+ halfspace bounded by hyperplan through $\{p_0, p_1, \dots, p_{d-1}\}$
and not including p_d

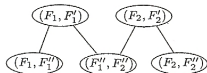
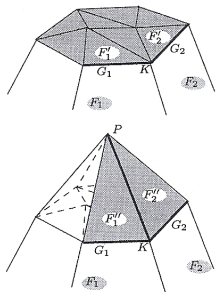
region: $h_0^+ \cup h_d^+$

Regions defined by \mathcal{P}
and without conflict in \mathcal{P}
are in bijection with the
 $(d - 2)$ -faces of $\text{conv}(\mathcal{P})$



On Line Convex Hull

Updating the influence graph



- Each node is attached to one or two parents
- Each node received $1 + (d - 1) = d$ children

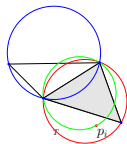
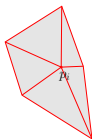
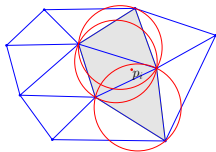
Incremental Delaunay Triangulation

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

$\text{Del}(\mathcal{P})$ can be obtained from a convex hull in \mathbb{R}^{d+1}

Inserting a new point p_i :

1. Location : Find all current cells whose circumball includes p_i
2. Update the Delaunay triangulation : star the hole from p_i



Incremental Delaunay Triangulation

conflict graph

Objects : point in \mathbb{R}^d

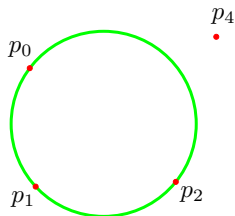
Regions : d -simplex in \mathbb{R}^d

p conflicts with τ iff

$p \in \text{circumball}(\tau)$

Conflict test

in 2D $\text{incircle}(p_i, p_j, p_k, p_l)$



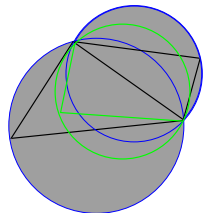
$$\text{sign} \left(\begin{array}{cccc} 1 & 1 & 1 & 1 \\ x_i & x_j & x_k & x_l \\ y_i & y_j & y_k & y_l \\ x_i^2 + y_i^2 & x_j^2 + y_j^2 & x_k^2 + y_k^2 & x_l^2 + y_l^2 \end{array} \middle| * \begin{array}{ccc} 1 & 1 & 1 \\ x_i & x_j & x_k \\ y_i & y_j & y_k \end{array} \right)$$

Incremental Delaunay Triangulation

Influence graph: the Delaunay tree

Objects : points in \mathbb{R}^d

Regions : union of two balls circumscribed
to adjacent d -simplex,
defined by $d + 2$ points



Updating the Delaunay tree

Each constructed region is attached to one or two parents in the Delaunay tree

Incremental Delaunay Triangulation

The Delaunay hierarchy

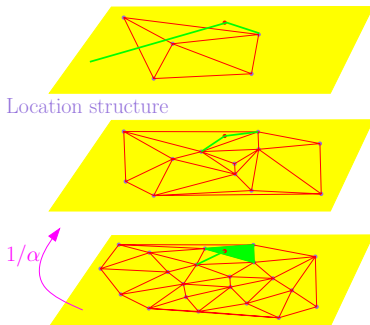
The Delaunay hierarchy

A location data structure widely used.

Level 0 is $\text{Del}(\mathcal{P})$

Each data point p in level l is introduced in level $l + 1$

with probability $\beta = \frac{1}{\alpha}$



Incremental Delaunay Triangulation

The Delaunay hierarchy

Location of point q :

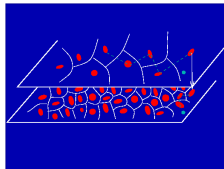
find the nearest neighbor of q in \mathcal{P}

$n_l(q)$: nearest neighbor of q in \mathcal{P}_l

Locate q in highest level

From $n_{l+1}(q)$ to $n_l(q)$:

- use the pointer of $n_{l+1}(q)$ to level l
- walk in level l from $n_{l+1}(q)$ to $n_l(q)$



The number of steps performed at level l : m_l

at most k if $n_{l+1}(p)$ is the k th neighbor of q in \mathcal{P}_l

$$\begin{aligned} \text{Exp}(m_l) &\leq \sum_{k=1}^{n_l} k(1-\beta)^{k-1}\beta \\ &\leq \beta \left[-\frac{\partial}{\partial \beta} \sum_k (1-\beta)^k \right] = \frac{1}{\beta} \end{aligned}$$

Expected total number of steps: $O(\log n)$.

Randomization

A tool for combinatorial results

Theorem (The sampling theorem)

\mathcal{P} a set of n objects

$\mathcal{F}_{\leq k}(\mathcal{P})$ regions defined by \mathcal{P} with at most k conflicts on \mathcal{P}

b the number of objects to define a region

$f_0(r)$ the expected number of regions defined and
without conflict on a random r -sample.

For $2 \leq k \leq \frac{n}{b+1}$,

$$|\mathcal{F}_{\leq k}(\mathcal{P})| \leq 4(b+1)^b k^b f_0\left(\left\lfloor \frac{n}{k} \right\rfloor\right).$$

Randomization

Proof of the sampling theorem 1.

$$f_0(r) = \sum_j |\mathcal{F}_j(\mathcal{P})| \frac{\binom{n-b-j}{r-b}}{\binom{n}{r}} \geq |\mathcal{F}_{\leq k}(\mathcal{P})| \frac{\binom{n-b-k}{r-b}}{\binom{n}{r}}$$

then, we prove that $\frac{\binom{n-b-k}{r-b}}{\binom{n}{r}} \geq \frac{1}{4(b-1)^b k^b}$

$$\frac{\binom{n-b-k}{r-b}}{\binom{n}{r}} = \underbrace{\frac{r!}{(r-b)!} \frac{(n-b)!}{n!}}_{\geq \frac{1}{4}} \underbrace{\frac{(n-r)!}{(n-r-k)!} \frac{(n-b-k)!}{(n-b)!}}_{\geq \frac{1}{(b-1)^b k^b}}$$

Randomization

Proof of the sampling theorem 2.

$$\begin{aligned}\frac{(n-r)!}{(n-r-k)!} \frac{(n-b-k)!}{(n-b)!} &\geq \left(\frac{n-r-k+1}{n-b-k+1}\right)^k \\ &\geq \left(\frac{n-n/k-k+1}{n-k}\right)^k \\ &\geq (1-1/k)^k \geq 1/4 \text{ pour } (2 \leq k),\end{aligned}$$

$$\begin{aligned}\frac{r!}{(r-b)!} \frac{(n-b)!}{n!} &= \prod_{l=0}^{b-1} \frac{r-l}{n-l} \geq \prod_{l=1}^b \frac{r+1-b}{n} \\ &\geq \prod_{l=1}^b \frac{n/k-b}{n} \\ &\geq 1/k^b (1 - \frac{bk}{n})^b \geq \frac{1}{k^b (b+1)^b} \text{ pour } (k \leq \frac{n}{b+1}).\end{aligned}$$

Bound on the number of k -sets

using randomization

k -sets

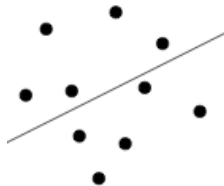
\mathcal{P} a set of n points in \mathbb{R}^d .

A k -set of \mathcal{P} is a subset \mathcal{P}' of \mathcal{P} with size k that can be separated from $\mathcal{P} \setminus \mathcal{P}'$ by a hyperplan.

Bound on the number of k -sets

For a set of n points in \mathbb{R}^d ,
the number of l -sets with $l \leq k$ is:

$$O\left(k^{\lceil \frac{d}{2} \rceil} n^{\lfloor \frac{d}{2} \rfloor}\right)$$



Bound on the number of k -sets

using randomization

$c_k(\mathcal{P})$ number of k -sets of \mathcal{P}

$c'_k(\mathcal{P})$ number of k -sets separated
by a hyperplan through points of \mathcal{P}

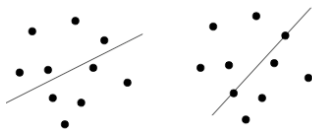
$$c'_{\leq k}(\mathcal{P}) = \sum_{l \leq k} c'_l(\mathcal{P})$$

$$c'_{\leq k}(n) = \sup_{|\mathcal{P}|=n} c'_{\leq k}(\mathcal{P})$$

Objects : points of \mathbb{R}^d

Regions : halfspaces in \mathbb{R}^d , $b = d$

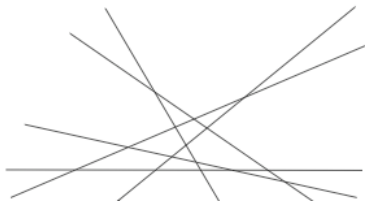
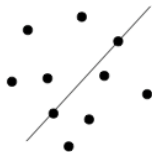
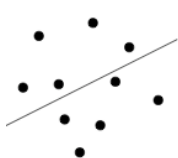
Conflict between p and h^+ : $p \in h^+$



$$\left. \begin{array}{l} \text{Sampling th: } c'_{\leq k}(\mathcal{P}) \leq 4(b+1)^b k^b f_0\left(\lfloor \frac{n}{k} \rfloor\right) \\ \text{Upper bound th: } f_0\left(\lfloor \frac{n}{k} \rfloor\right) = O\left(\frac{n \lfloor \frac{d}{2} \rfloor}{k \lfloor \frac{d}{2} \rfloor}\right) \end{array} \right\} \Rightarrow c'_{\leq k}(n) = O\left(k^{\lceil \frac{d}{2} \rceil} n^{\lfloor \frac{d}{2} \rfloor}\right)$$

Hyperplan Arrangements

k-sets and *k*-levels



Duality:

$$p \in \mathbb{R}^d \longrightarrow p^* : x_d + p_d - 2 \sum_{i=1}^{d-1} p_i \cdot x_i = 0$$

$$\mathcal{P} \in \mathbb{R}^d \longrightarrow \mathcal{P}^* = \{p^* : p \in \mathcal{P}\}$$

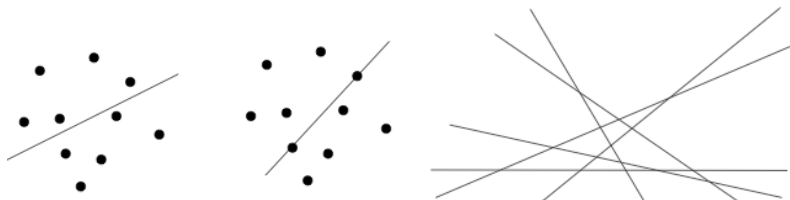
Arrangement : $\mathcal{A}(\mathcal{P}^*)$

$$c_k(\mathcal{P}) \leftrightarrow \text{cells with level } k \text{ or } n - k$$

$$c'_k(\mathcal{P}) \leftrightarrow \text{vertices of } \mathcal{A}(\mathcal{P}^*) \text{ with level } k \text{ or } n - k$$

Bound on the number of k -sets

End



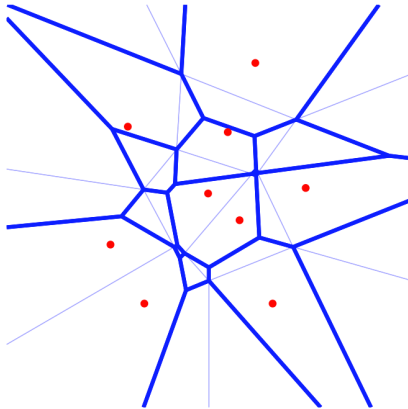
Each cell in $\mathcal{A}(\mathcal{P}^*)$ has at most one leftmost and one rightmost vertex.
Each vertex v in $\mathcal{A}(\mathcal{P}^*)$ is the rightmost or leftmost vertex of a single cell f .

If the level of f is k , the level of $v \in [k - d + 1, k - 1]$

hence, $c_k(n) \leq 2 \sum_{l=k-d+1}^{k-1} c'_l(n)$

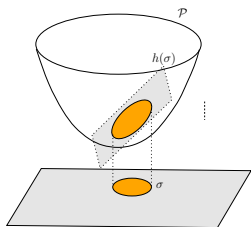
$$\left. \begin{array}{l} c_k(n) \leq 2 \sum_{l=k-d+1}^{k-1} c'_l(n) \\ c'_{\leq k}(n) = O\left(k^{\lceil \frac{d}{2} \rceil} n^{\lfloor \frac{d}{2} \rfloor}\right) \end{array} \right\} \Rightarrow c_{\leq k}(n) = O\left(k^{\lceil \frac{d}{2} \rceil} n^{\lfloor \frac{d}{2} \rfloor}\right)$$

k -order Voronoi Diagrams



Each cell is the locus of points with the same set of k -nearest neighbors.

Back to Space of Spheres



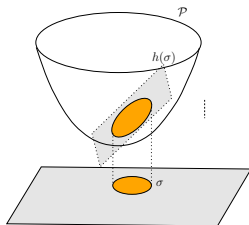
$$\begin{array}{ll} p \text{ point in } \mathbb{R}^d & \rightarrow \phi(p) = (p, p^2) \in \mathbb{R}^{d+1} \\ \sigma \text{ sphere de } \mathbb{R}^d & \rightarrow \phi(\sigma) \in \mathbb{R}^{d+1} \\ x^2 - 2c \cdot x + s = 0 & \phi(\sigma) = (c, s), \\ s = c^2 - r^2 & \text{hyperplane } \phi(\sigma)^* \\ & x_{d+1} - 2c \cdot x + s = 0 \end{array}$$

1. Si $\sigma = \{p\}$, $\phi(\sigma) \in \mathcal{P}$, $\phi(\sigma)^*$ hyperplan tangent à \mathcal{P}
2. L'intersection $\phi(\sigma)^* \cap \mathcal{P}$ se projette sur $x_{d+1} = 0$ selon σ

$$\begin{array}{lll} \sigma(x) = 0 & \iff & x^2 - 2c \cdot x + s = 0 \iff \phi(x) \in \phi(\sigma)^* \\ \sigma(x) < 0 & \iff & x^2 - 2c \cdot x + s < 0 \iff \phi(x) \in \phi(\sigma)^{-} \\ \sigma(x) > 0 & \iff & x^2 - 2c \cdot x + s > 0 \iff \phi(x) \in \phi(\sigma)^{+} \end{array}$$

k -Order Voronoi Diagrams

Back to space of spheres



Complexity of k -order Voronoi diagram

Let \mathcal{P} be a set of points in \mathbb{R}^d . The total number of faces in Voronoi diagrams of \mathcal{P} of order up to k is

$$O\left(k \binom{d+1}{2} n \lfloor \frac{d+1}{2} \rfloor\right)$$

Proof. Each cell in the k -order Voronoi diagram of \mathcal{P} corresponds to :

- a k -set of $\phi(\mathcal{P})$ in \mathbb{R}^{d+1} .
- a cell of level k in the arrangement $\mathcal{A}(\phi^*(\mathcal{P}))$ in \mathbb{R}^{d+1} .