

# Ensembles convexes et mosaïques aléatoires

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1 Convex hulls of stochastic processes

2 Convex rearrangements

3 Random tessellations

4 Generalised peeling

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# Extremal points

Stochastic process  $X$

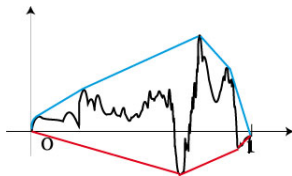
$\text{ext}(X)$  : Set of the extremal points of its graph.

Topics of interest:

- Topology of  $\text{ext}(X)$ ,
- 1-dimensional Hausdorff measure of  $\text{ext}(X)$ .

More precisely:

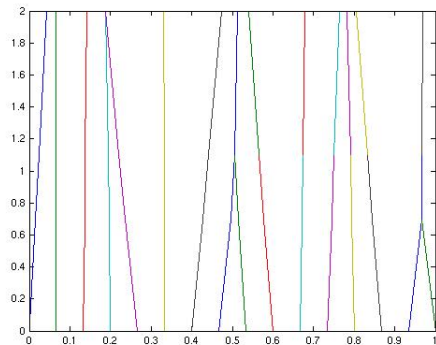
$\text{ext}^+(X)$ : Extremal points in the upper part of the convex hull.

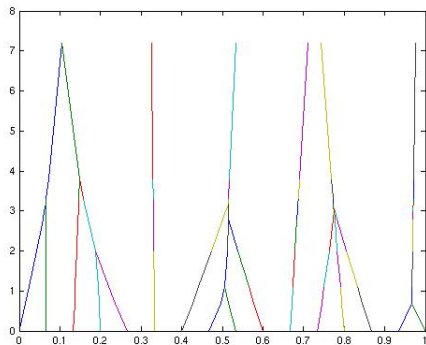


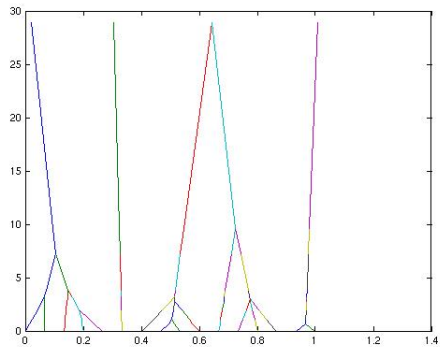
# Continuous model

- $I$ : Interval of  $\mathbb{R}$ .
- 1-dimensional fluid uniformly spread along  $I$ .
- $v(x, t)$ : Velocity field of the fluid at time  $t$ .
- $\psi(x, t)$ : Potential, defined by  $\partial_x \psi(x, t) = v(x, t)$ .

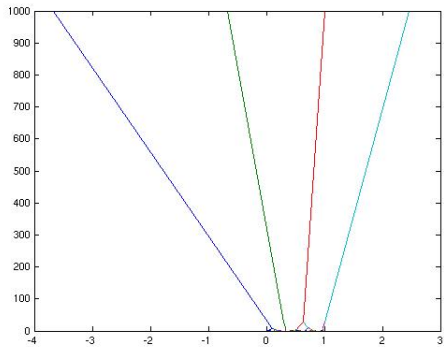
Evolution ruled by Burgers equation (inelastic shocks between particles).











- Initial data:  $X(x) = \psi(x, 0)$ , random process.
- $\mathcal{L}_t$ : Set of *Lagrangian regular points*, i.e. the points not involved in a shock up to time  $t$ .

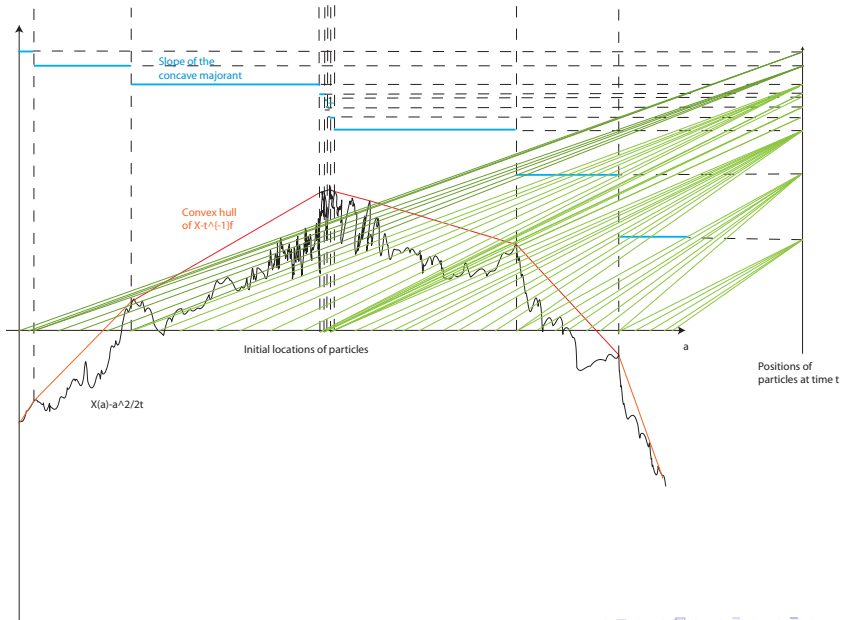
Hopf (1950), Cole (1951): Explicit solution in terms of  $\psi(x, t)$ .

Call  $f(x) = -x^2/2$ .  $\mathcal{L}_t$  is the set of points of  $\text{ext}^+(X + t^{-1}f)$  that are isolated neither on the left nor on the right.

### Theorem

$\text{ext}^+(X + t^{-1}f) \rightarrow \text{ext}^+(X)$  in the Hausdorff metric as  $t \rightarrow \infty$ .

Thus the set  $\mathcal{L}_t$  tends to “shrink” around the points of  $\text{ext}^+(X)$ .  
 Explicit description of  $\mathcal{L}_t$ : Bertoin, Winkel, Giraud,...



# Results of negligibility for $\text{ext}^+(X)$

## Theorem

Assume we are in one of the following situations:

- $X$  is a reversible “nice” Markov process, meaning for all  $s > 0$ ,

$$\{X(s+t) - X(s) : 0 \leq t < s\} \stackrel{(d)}{=} \{X(s) - X((s-t)^-) : 0 \leq t < s\}.$$

(It includes Lévy processes without drift)

- $X$  is an Ito process which Brownian component vanishes on no open interval.

Then  $\text{ext}(X)$  has zero 1-dimensional Hausdorff measure.

# Topological results

Variation of  $X$  on an interval  $I$ :

$$V(X, I) = \sup_{x_i, q} \sum_{i=1}^q |X(x_i) - X(x_{i+1})|.$$

for  $0 < x_1 < \dots < x_q \in I$ .

If  $X$  Lévy process,

- either  $V(X, I) = \infty$  for all open  $I$ ,
- or  $V(X, I) < \infty$  for all bounded  $I$ .

A Lévy process  $X$  with bounded variation only varies by jump,

$$X(a) - X(0) - ba = \sum_{0 \leq s \leq a} (X(s^+) - X(s^-)),$$

$b$ : Drift, irrelevant here. We assume  $b = 0$ .

$T_{\max}$ : Time where  $X$  approaches its maximal value if  $I$  is a compact interval (otherwise,  $T_{\max} = +\infty$ ).

### Theorem

- $T_{\max} \in \text{ext}^+(X)$ ,
- $\text{ext}^+(X)$  is discrete away from  $T_{\max}$ ,
- $T_{\max}$  is a left (resp. right) accumulation point of  $\text{ext}^+(X)$  iff  $X$  takes strictly positive (resp. negative) values arbitrary close from 0.

## Concave drift

The set  $\mathcal{L}_t$  is determined by  $\text{ext}^+(X + t^{-1}f)$ , where  $f(x) = -x^2/2$ .

### Theorem

$X$ : Levy process without drift and with bounded variation.

$f$ : Concave function.

$$\text{ext}^+(X - f) \subset \text{ext}^+(X) \subset \text{ext}^+(X + f).$$

Call  $\bar{g}$  the concave majorant of a function  $g$ .

A left (resp. right) accumulation point  $s$  in  $\text{ext}^+(X + f)$  satisfies

$$\begin{aligned}\overline{X + f}'(s^-) &= f'(s), \\ \text{resp. } \overline{X + f}'(s^+) &= f'(s).\end{aligned}$$

If  $f = 0$ , you retrieve the fact that every accumulation point  $s$  of  $\text{ext}^+(X)$  satisfies

$$\overline{X}'(s) = 0,$$

whence  $X(s)$  is an extremum of  $X$ .

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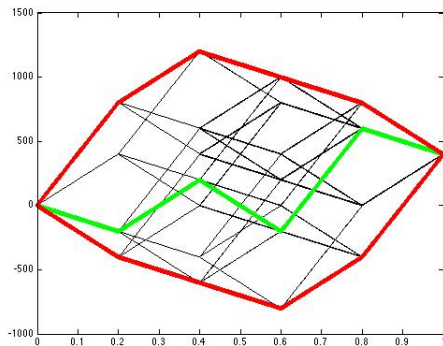


# Associated convex body of a 1-dimensional function

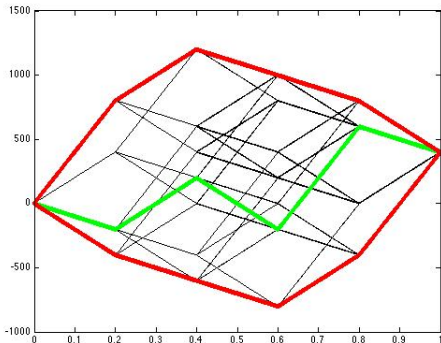
Resource distributed to a population of size  $N$ .

- Member labelled  $k$  receives  $r_k$ .
- Cumulative income function:  $f(n) = \sum_{k \leq n} r_k$ .

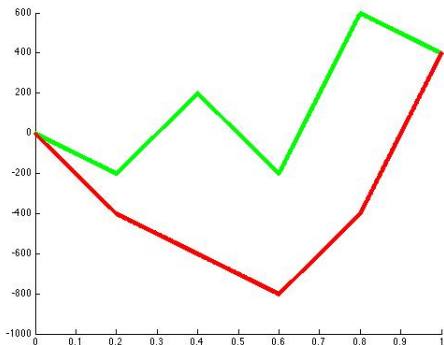
$f$  is extended to a piece-wise linear function on  $[0, N]$ .



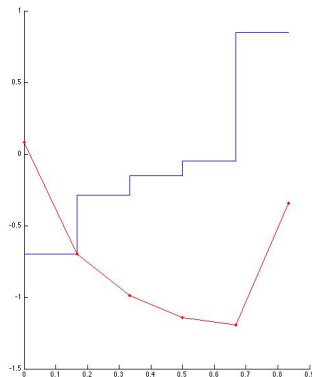
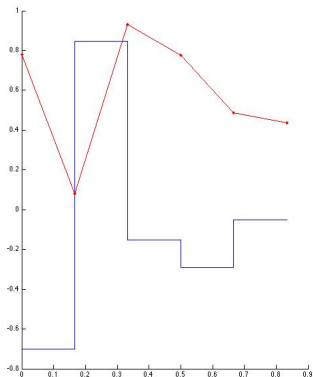
The area of the convex body can measure the inequalities over this particular resource (consider the equality case, where  $r_k$  is equal for all  $k$ )



Lower part (red): *convex rearrangement of  $f$* , denoted by  $\mathcal{C}f$ .



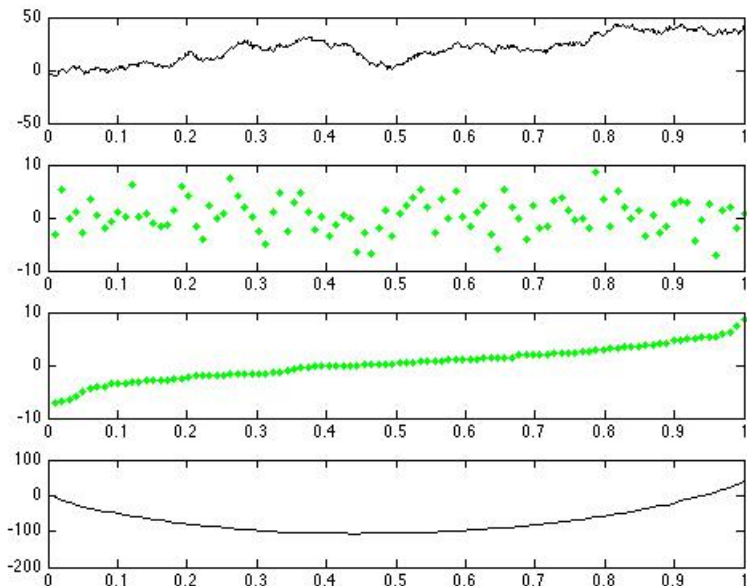
# Rearrangement of the derivative



It corresponds to rearranging the derivative in a monotone way. If  $f'$  is the derivative of  $f$ , and  $(\mathcal{C}f)'$  the derivative of  $\mathcal{C}f$ , we have

$$\lambda_1 f'^{-1} = \lambda_1 (\mathcal{C}f)'^{-1}.$$

# Asymptotic rearrangement of the Brownian motion



## Theorem (Davydov, Zitikis 2004)

$X$ : Brownian motion.

$X_n$ : Piece-wise linear interpolation of  $X$  on  $\{0, 1/n, \dots, 1\}$ .

Then

$$\sup_{x \in [0,1]} \left| \frac{1}{\sqrt{n}} \mathfrak{C}X_n(x) - L(x) \right| \rightarrow 0,$$

$L$ : Lorenz curve.

Other asymptotic convex rearrangements in Davydov & Vershik 1998.

# Multidimensional rearrangement

Let  $f : [0, 1]^d \rightarrow \mathbb{R}$ , differentiable a.e. such that

$$\int_{[0,1]^d} \|\nabla f(x)\| dx < +\infty.$$

A convex function  $C$  is a convex rearrangement of  $f$  if

$$\lambda_d \nabla f^{-1} = \lambda_d \nabla C^{-1}.$$

## Theorem (Brenier, 91)

*Every function  $f$  with finite gradient mass has a convex rearrangement.  
The convex rearrangement is unique up to a constant.*

## Asymptotic rearrangement

- $f$ : “irregular function”
- $f_n$ : Functions with finite gradient mass, the  $f_n$  converge to  $f$ . Is there a function  $C$ , and positive numbers  $\{b_n; n \geq 1\}$ , such that

$$b_n \mathfrak{C} f_n \rightarrow C?$$

If yes,  $C$  is a *asymptotic convex rearrangement*.

### Theorem

$\{f_n; n \geq 1\}$ : Functions with finite gradient mass,

$\{b_n; n \geq 1\}$ : Positive numbers.

The following assertions are equivalent

- (i) Weak convergence  $\lambda_d \nabla(b_n f_n)^{-1} \Rightarrow \mu$ .
- (ii)  $b_n \mathfrak{C} f_n(z) \rightarrow C(z)$ , for  $z \in \text{int}([0, 1]^d)$ ,
- (iii)  $\nabla(b_n \mathfrak{C} f_n)^{-1} \rightarrow \nabla C$  in the  $L^1$  sense on every sub-compact, whence  $C \in \mathfrak{C}f$ .

In this case:  $\mu = \lambda_d \nabla C^{-1}$ .



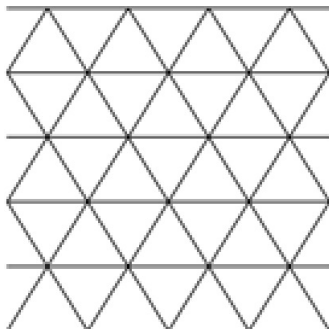
## Gaussian fields

$X_n$  : Approximations of a Gaussian field  $X$  on  $[0, 1]^d$ .

$X_n$  is obtained by interpolation of  $X$  on a triangulation  $\mathcal{T}_n$ .

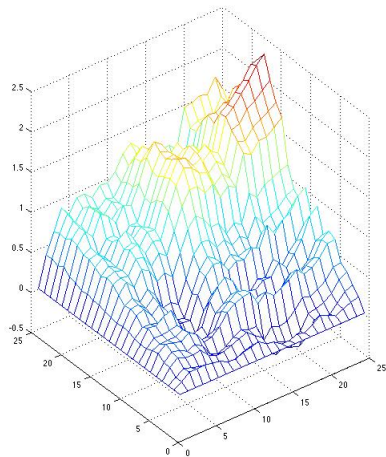
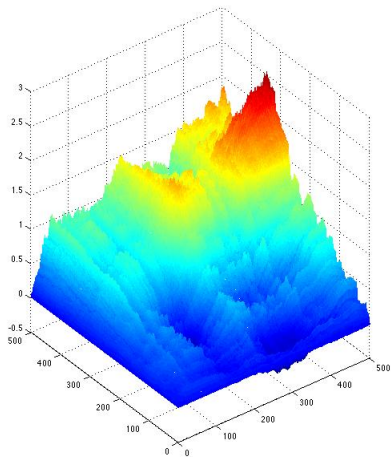
There are regular simplices  $T_1, \dots, T_k$ , and a discrete group  $\Gamma$  of  $\mathbb{R}^d$  such that

$$\mathcal{T}_n = \left\{ \frac{1}{n}(\gamma + T_j) : \gamma \in \Gamma, 1 \leq j \leq k \right\}.$$



Call  $\mathcal{V}_n$  the set of vertices of  $\mathcal{T}_n$ .

# Brownian sheet approximation



# Results

## Theorem

Assume the following:

Convergence of the mean: There is a deterministic measure  $\mu$  such that, for all Borel set  $B$ ,

$$\mathbb{E} \int_{[0,1]^d} \mathbf{1}_{\{b_n \nabla X_n(z) \in B\}} dz = \mathbb{E}(\lambda_d(b_n \nabla X_n)^{-1}(B)) \rightarrow \mu(B).$$

$\Theta$  : Set of points  $(x, y)$  where the covariance of  $X$  is not of class  $C^1$

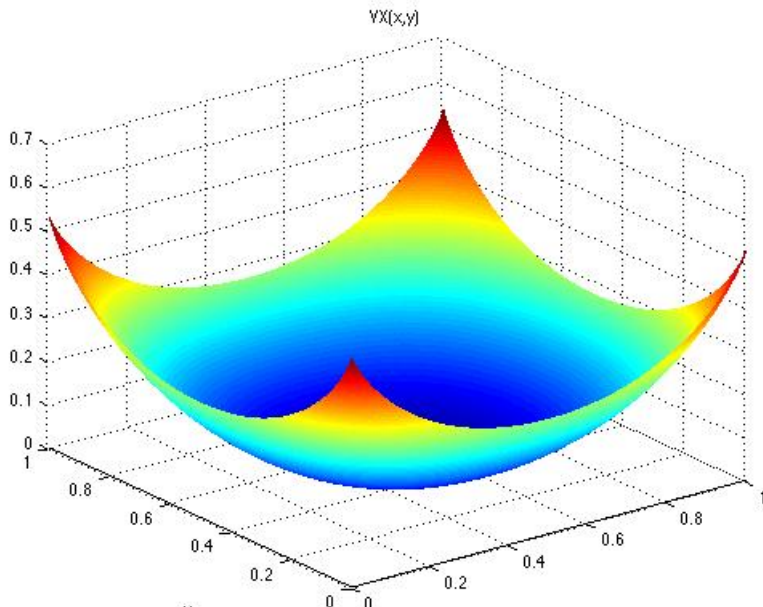
- $\sum_n b_n^{-4} < +\infty$ ,
- $\sum_n n^{-2d} \text{card}(\Theta \cap \mathcal{V}_n^2) < +\infty$ ,
- $\sup_{n \in \mathbb{N}, z \in [0,1]^d} \text{cov}(\nabla b_n X_n(z)) < +\infty$ .

Then

$$b_n \nabla X_n(z) \rightarrow C(z), \quad z \in \text{int}([0, 1]^d)$$

where  $C$  is convex and the gradient distribution of  $C$  is  $\mu$ .

# Asymptotic rearrangement of the Brownian sheet



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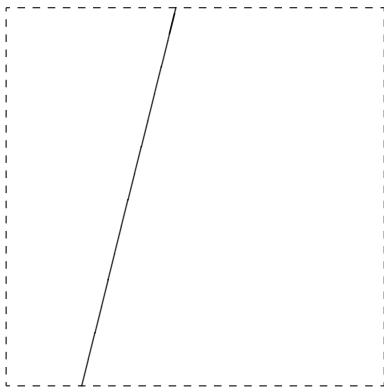
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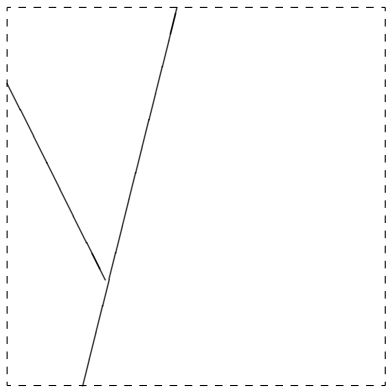
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Construction on a bounded convex window

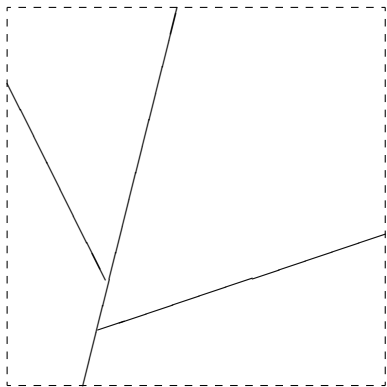
Wait an exponential time with parameter  $\text{Per}(W)$ . Then draw a random line.



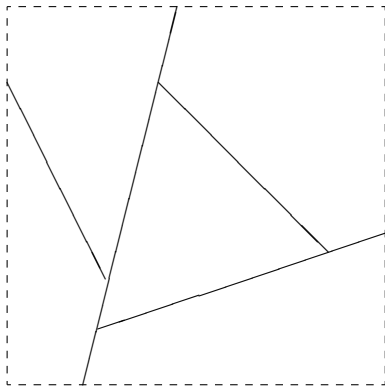
Construction on a bounded convex window  $W$ . Each cell created behaves independently. On each cell, wait an exponential time with parameter the perimeter, and then draw a random line inside the cell.

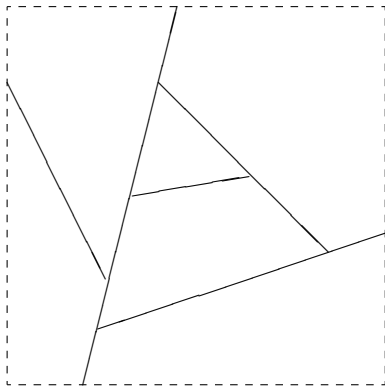


Construction on a bounded convex window  $W$ . Each cell created behaves independently. On each cell, wait an exponential time with parameter the perimeter, and then draw a random line inside the cell.

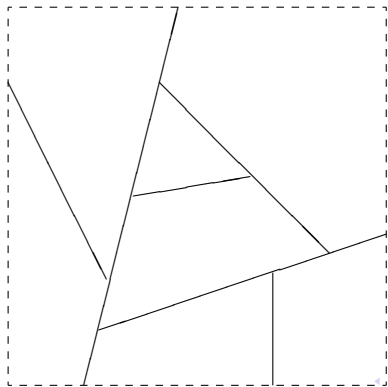








Stop the construction at some predetermined time  $a > 0$



Properties of the construction:

- The behaviour of two disjoint cells is independent.
- The process has no memory.

The parameters of the construction are

- The time parameter  $a > 0$ ,
- The law of the random lines.

The same construction can take place in  $\mathbb{R}^d$ , but lines are replaced by hyperplanes.

It can model the cracking of a drying soil, or the subdivision procedure of cells.

## Theorem

*There exists a tessellation  $T$  on all  $\mathbb{R}^d$  such that the law of  $T \cap W$ , for any compact  $W$ , corresponds to the above tessellation.*

Such a tessellation is called STIT tessellation.

## Mixing property

For two compact sets  $A$  and  $B$ , and a STIT tessellation  $T$ , one has

$$\mathbb{P}(A \cap T = \emptyset, (B + h) \cap T = \emptyset) \rightarrow \mathbb{P}(A \cap T = \emptyset)\mathbb{P}(B \cap T = \emptyset),$$

when  $\|h\| \rightarrow \infty$ .

It implies that  $T$  is ergodic. Moreover, one has the decay rate

$$|\mathbb{P}(A \cap T = \emptyset, (B + h) \cap T = \emptyset) - \mathbb{P}(A \cap T = \emptyset)\mathbb{P}(B \cap T = \emptyset)| = o(\|h\|^{-1}).$$

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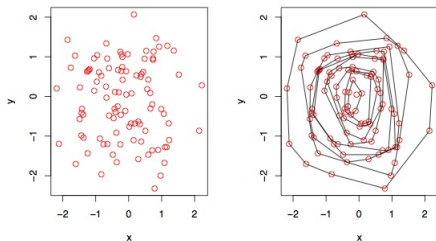
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# Convex peeling

Convex peeling of a set of points.



For a random set of points  $X$ , call  $X_{(k)}$  the points that are on the border of the  $k$ -th envelope. We have

$$X = \cup_k X_{(k)},$$

and the union is disjoint.

We set

$$\text{env}^k(X) = \text{conv}(X_{(k)}).$$



# Weak convergence for $\alpha$ -stable point processes

IID random variables:  $\xi_i, i \geq 1$ , which common law belongs to the attraction domain of an  $\alpha$ -stable law.

$$X_n = \{b_n \xi_1, \dots, b_n \xi_n\}$$

with proper  $b_n$ ,

$$X_n \rightarrow X,$$

$X$ :  $\alpha$ -stable Poisson point process.

Weak convergence:

$$\text{env}^k(X_n) \Rightarrow \text{env}^k(X),$$

The topology is that generated by the Hausdorff metric.

Davydov & Nagaev 2004 deduced the convergence of functionals based on  $X_{(k)}$  (extreme summands).

# The $\mathcal{F}$ -convexity

Closed convex hull

$$\text{conv}(A) = \bigcap_{A \subseteq H} H,$$

for  $H$  closed half-space.

Closed half-spaces  $\Rightarrow$  arbitrary family of level sets, of the form

$$H = \{x : f(x) \leq \alpha\},$$

for  $\alpha \in \mathbb{R}$ ,  $f$  continuous function,  $f \in \mathcal{F}$ .

Convex hull:  $\mathcal{F}$  is the class of linear forms.

# Convergence result

Define the  $\mathcal{F}$ -peeling similarly. We obtained sufficient conditions for having the same result of convergence

There is the weak convergence

$$\text{env}_{\mathcal{F}}^k(X_n) \Rightarrow C_{\mathcal{F}}^k,$$

$C_{\mathcal{F}}^k = \text{env}_{\mathcal{F}}^k(X)$ , for some  $\alpha$ -stable Poisson point process  $X$ .