

# Lipschitz-Killing curvatures of excursion sets for 2D random fields

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# Collaboration

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# Motivation : Shot noise random fields

A **(Poisson) shot noise random field** is a random function  $X : \mathbb{R}^d \rightarrow \mathbb{R}$  given by

$$\forall x \in \mathbb{R}^d, \quad X(x) = \sum_{i \in I} g_{m_i}(x - x_i), \text{ where}$$

- $\{x_i\}_{i \in I}$  is a Poisson point process of intensity  $\lambda > 0$  in  $\mathbb{R}^d$ ,
- $\{m_i\}_{i \in I}$  are independent « marks » with distribution  $F(dm)$  on  $\mathbb{R}^k$ , and independent of  $\{x_i\}_{i \in I}$ .
- The functions  $g_m$  are real-valued deterministic functions, called **spot functions**, such that

$$\int_{\mathbb{R}^k} \int_{\mathbb{R}^d} |g_m(y)| dy F(dm) < +\infty.$$

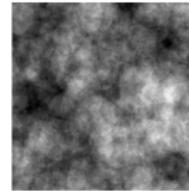
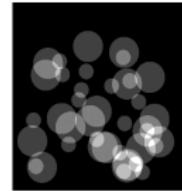
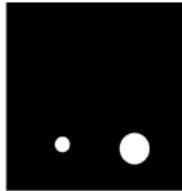
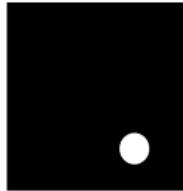
Here we consider  $d = 2$  and, for sake of simplicity,  $k \leq 2$  with a single  $L^1(\mathbb{R}^2)$  function  $g$  randomly weighted and dilated :  $(W, R) \sim F$  is a probability measure on  $\mathbb{R} \times (0, +\infty)$  and for  $m = (w, r)$

$$g_m(x) = w g(x/r).$$

## Example 1 : disk with random radius

Let  $d = 2$ ,  $T$  a bounded closed rectangle of  $\mathbb{R}^2$  and  $g = \mathbf{1}_D$ . Consider random disk of radius  $r = r_1$  or  $r = r_2$  with  $0 < r_1 < r_2$  (each with probability  $1/2$ ), same weights  $W = 1$  a.s. and intensity  $\lambda > 0$

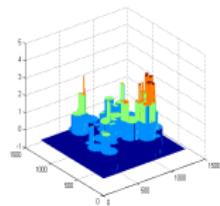
- The number  $n$  of centers in  $T$  is a Poisson random variable of parameter  $\lambda|T|$
- The centers  $x_1, \dots, x_n$  are thrown uniformly, independently on  $T$
- The radius  $R_1, \dots, R_n$  are attached to each center by flipping a coin to choose between  $r_1$  or  $r_2$ .



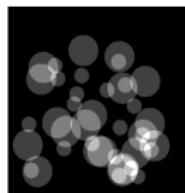
# Excursion set

We consider the excursion set or the level set of level  $u \in \mathbb{R}$  of  $X$  in  $T$  defined by

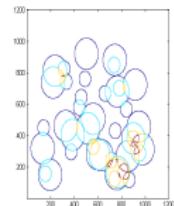
$$E_X(u) \cap T := \{x \in T; X(x) \geq u\} \text{ with } E_X(u) = \{X \geq u\}.$$



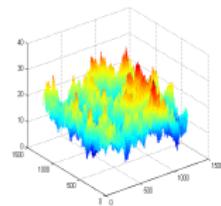
view 3D



view 2D



some level lines



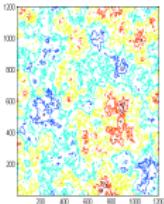
$u = 0.5$



$u = 1.5$

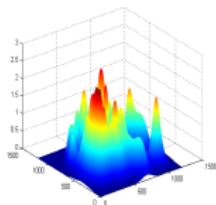


$u = 2.5$

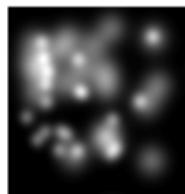


## Example 2 : Gaussian kernel

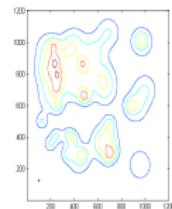
Let us choose  $g(x) = e^{-\frac{\|x\|^2}{2}}$  instead of  $\mathbf{1}_D$ .



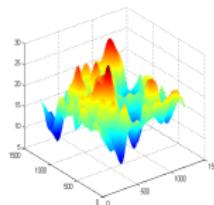
view 3D



view 2D



some level lines



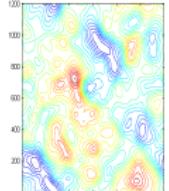
$u = 0.5$



$u = 1$



$u = 1.5$



# Main questions

What can be said about "mean" geometry of excursion sets ? Area ?  
Perimeter ? Euler Characteristic = # connected components – # holes ?

Known results for

- Boolean model : Mecke (2001), Mecke, Wagner (1991)
- Smooth Gaussian and related random fields : Adler (2000), Adler, Taylor (2007), Azaïs, Wschebor (2009), ...
- High levels for some infinitely divisible fields : Adler, Samorodnitsky, Taylor (2010,2013),...

Two different frameworks

- 1 Piecewise constant fields (elementary)
- 2 Smooth fields : at least  $C^2$

# Curvature measures

Let  $E \subset \mathbb{R}^2$  be a "nice set". Its curvature measures  $\Phi_j(E, \cdot)$ , for  $j = 0, 1, 2$ , are defined for any Borel set  $U \subset \mathbb{R}^2$  by

- $\Phi_2(E, U) = |E \cap U|,$
- $\Phi_1(E, U) = \frac{1}{2} \mathcal{H}^1(\partial E \cap U) = \frac{1}{2} \text{Per}(E, U)$
- $\Phi_0(E, U) = \frac{1}{2\pi} \text{TC}(\partial E, U),$

where  $\mathcal{H}^1(\partial E \cap U)$  is the lenght and  $\text{TC}(\partial E, U)$  the total curvature of the positively oriented curve  $\partial E$  in  $U$ .

Ref : Schneider, Weil, Stochastic and Integral Geometry

# Piecewise regular curve

A Jordan curve  $\Gamma \subset \mathbb{R}^2$  is piecewise regular if  $\Gamma = \mathcal{R}_\Gamma \cup \mathcal{C}_\Gamma$  with  $\#\mathcal{C}_\Gamma < +\infty$

- for  $x \in \mathcal{R}_\Gamma$  one has  $x = \gamma(s)$  for some  $s \in (0, \varepsilon)$  with  $\gamma : (0, \varepsilon) \rightarrow \Gamma$   $C^2$ , arc length parametrized. Then,  $|\gamma(0, \varepsilon)|_1 = \varepsilon$ .  
The **signed curvature**  $\kappa_\Gamma(x)$  of  $\Gamma$  at  $x$  is

$$\kappa_\Gamma(x) = \langle \gamma''(s), \gamma'(s)^\perp \rangle.$$

- for  $x \in \mathcal{C}_\Gamma$  one has  $x = \gamma(0)$  with  $\gamma : (-\varepsilon, \varepsilon) \rightarrow \Gamma$  continuous and  $C^2$  on  $(-\varepsilon, \varepsilon) \setminus \{0\}$  s.t.  $\gamma'$  admits limits  $\gamma'(0^-) \in S^1$  and  $\gamma'(0^+) \in S^1$  at 0. Then,  $|\gamma(-\varepsilon, \varepsilon)|_1 = 2\varepsilon$ .  
The **turning angle** at a corner point  $x = \gamma(0) \in \mathcal{C}_\Gamma$  is the angle  $\alpha_\Gamma(x) \in (-\pi, \pi)$  between the tangent “before” and the one “after”  $x$

$$\alpha_\Gamma(x) = \operatorname{Arg} \gamma'(0^+) - \operatorname{Arg} \gamma'(0^-) \quad \in (-\pi, \pi).$$

# Total curvature and Euler characteristic

The **total curvature** of  $\Gamma$  in  $U$  is defined as

$$\text{TC}(\Gamma, U) := \int_{\mathcal{R}_\Gamma \cap U} \kappa_\Gamma(x) \mathcal{H}^1(dx) + \sum_{x \in \mathcal{C}_\Gamma \cap U} \alpha_\Gamma(x).$$

**Gauss-Bonnet Theorem :** Let  $E \subset U$  be a regular region ie  $E = \overline{E}$  such that  $\partial E$  is formed by  $n$  piecewise regular positively oriented disjoint Jordan curves  $\Gamma_1, \dots, \Gamma_n$  then

$$\text{TC}(\partial E, U) := \sum_{i=1}^n \text{TC}(\Gamma_i, U) = 2\pi\chi(E),$$

where  $\chi(E)$  is the **Euler characteristic** of  $E$ ,

$$\chi(E) = \#\text{connected components} - \#\text{holes}.$$

It follows that  $\Phi_0(E, U) = \chi(E)$ .

# Geometry of excursion sets

Let  $X = (X(x))_{x \in \mathbb{R}^2}$  be a stationary "nice" random field and  $T$  a bounded closed rectangle with  $\overset{\circ}{T} \neq \emptyset$ . For  $u \in \mathbb{R}$ , we consider the excursion set of level  $u$  in  $T$

$$E_X(u) \cap T := \{x \in T; X(x) \geq u\}.$$

the **LK curvatures** of the excursion set  $E_X(u)$  within  $T$  are

$$C_j(X, u, T) := \Phi_j(E_X(u) \cap T, T), \text{ for } j = 0, 1, 2.$$

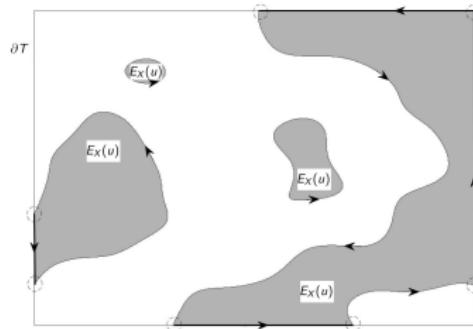
and, assuming the limits exist, the associated **LK densities** are

$$C_j^*(X, u) := \lim_{T \nearrow \mathbb{R}^2} \frac{\mathbb{E}[C_j(X, u, T)]}{|T|}, \text{ for } j = 0, 1, 2,$$

where  $\lim_{T \nearrow \mathbb{R}^2}$  stands for the limit along any sequence of bounded rectangles that grows to  $\mathbb{R}^2$ . Note that

$$C_j^*(\sigma X + m, u) = C_j^*(X, (u - m)/\sigma).$$

# LK densities



Note that

$$C_j(X, T, u) = \Phi_j(E_X(u) \cap T, T) = \Phi_j(E_X(u), \overset{o}{T}) + \Phi_j(T \cap E_X(u), \partial T).$$

Actually,  $C_j^*(X, u) = \frac{\mathbb{E}[\Phi_j(E_X(u), \overset{o}{T})]}{|T|}$ . Moreover, by stationarity

$$C_2^*(X, u) = \mathbb{P}(X(0) \geq u).$$

# What for smooth deterministic functions?

Assume that  $f : W \rightarrow \mathbb{R}$  is  $C^2$  with  $W$  open s.t.  $T \subset W$  and note that  $\partial E_f(u) \cap \dot{T} = \{x \in \dot{T}; f(x) = u\}$ .

- 1 By Morse-Sard theorem, the image by  $f$  of the set of critical values of  $f$  has measure 0 in  $\mathbb{R}$ .
- 2 Let  $u$  be such a non-critical value. For a curve  $\gamma$  given by an implicit form  $f(\gamma(s)) = u$ , we have  $\gamma'(s)^\perp = \nabla f(\gamma(s))/\|\nabla f(\gamma(s))\|$  and therefore the curvature at  $x = \gamma(s)$  is given by

$$\kappa_f(x) = -\frac{D^2f(x).(\nabla f^\perp(x), \nabla f^\perp(x))}{\|\nabla f(x)\|^3}.$$

- 3 The **coarea formula** for Lipschitz mappings states that, for any  $\mathcal{L}$ -integrable function  $g$ ,

$$\int_{\mathbb{R}} \int_{\partial E_f(u) \cap \dot{T}} g(x) \mathcal{H}^1(dx) du = \int_{\dot{T}} g(x) \|\nabla f(x)\| dx.$$

## Weak formula for $\Phi_1$ and $\Phi_0$

Let us choose  $h : \mathbb{R} \rightarrow \mathbb{R}$  a bounded continuous function (test function).

**Coarea formula** with  $g(x) = h(f(x))$  :

$$\int_{\mathbb{R}} h(u) \Phi_1(E_f(u), \mathring{T}) du = \frac{1}{2} \int_{\mathring{T}} h(f(x)) \|\nabla f(x)\| dx.$$

**Coarea formula** with  $g(x) = h(f(x)) \kappa_f(x)$  for

$$\kappa_f(x) = -\frac{D^2 f(x) \cdot (\nabla f(x)^\perp, \nabla f(x)^\perp)}{\|\nabla f(x)\|^3} \mathbf{1}_{\|\nabla f(x)\| > 0},$$

$$\begin{aligned} & \int_{\mathbb{R}} h(u) \Phi_0(E_f(u), \mathring{T}) du \\ &= -\frac{1}{2\pi} \int_{\mathring{T}} h(f(x)) \frac{D^2 f(x) \cdot (\nabla f(x)^\perp, \nabla f(x)^\perp)}{\|\nabla f(x)\|^2} \mathbf{1}_{\|\nabla f(x)\| > 0} dx. \end{aligned}$$

# Expectation under stationarity

When  $X$  is a stationary field a.s.  $C^2$  with  $X(0)$ ,  $\nabla X(0)$  and  $D^2X(0)$   $L^1$

$$\int_{\mathbb{R}} h(u) C_1^*(X, u) du = \frac{1}{2} \mathbb{E}(h(X(0)) \|\nabla X(0)\|)$$

$$\int_{\mathbb{R}} h(u) C_0^*(X, u) du = \frac{-1}{2\pi} \mathbb{E} \left( h(X(0)) \frac{D^2X(0) \cdot (\nabla X(0)^\perp, \nabla X(0)^\perp)}{\|\nabla X(0)\|^2} \mathbf{1}_{\|\nabla X(0)\| > 0} \right)$$

↳ Allows information for a.e  $u \in \mathbb{R}$

We also obtain for  $h = 1$

- $TV^*(X) = 2 \int_{\mathbb{R}} C_1^*(X, u) du ;$
- $LTC^*(X) = 2\pi \int_{\mathbb{R}} C_0^*(X, u) du.$

# Expectation under stationarity and isotropy

When  $X$  is also **isotropic**, we get

$$\mathbb{E}(h(X(0))\|\nabla X(0)\|) = \frac{\pi}{2}\mathbb{E}(h(X(0))|X_1(0)|).$$

Moreover,

$$\mathbb{E}\left(h(X(0))\frac{D^2X(0).(\nabla X(0)^\perp, \nabla X(0)^\perp)}{\|\nabla X(0)\|^2} \mathbf{1}_{\|\nabla X(0)\|>0}\right) = \alpha_0(h) + 2\alpha_2(h),$$

with  $\alpha_0(h) = \mathbb{E}(h(X(0))X_{11}(0))$ , and

$$\alpha_2(h) = -2\mathbb{E}\left(h(X(0))X_{12}(0)\frac{X_1(0)X_2(0)}{\|\nabla X(0)\|^2} \mathbf{1}_{\|\nabla X(0)\|>0}\right).$$

Therefore

$$TV^*(X) = \frac{\pi}{2}\mathbb{E}(|X_1(0)|) \text{ and } LTC^*(X) = 4\mathbb{E}\left(X_{12}(0)\frac{X_1(0)X_2(0)}{\|\nabla X(0)\|^2} \mathbf{1}_{\|\nabla X(0)\|>0}\right).$$

## Gaussian case

For  $X$  a stationary standard isotropic  $C^2$  Gaussian random field we note  $\rho(x) = \text{Cov}(X(x), X(0))$ , and the second spectral moment

$$\lambda_2 = -\partial_k^2 \rho(0) = -\text{Cov}(X(0), X_{kk}(0)) = \text{Var}(X_k(0)).$$

By stationarity  $\text{Cov}(X(0), X_k(0)) = \text{Cov}(X_k(0), X_{12}(0)) = 0$  and

$$\mathbb{E}(h(X(0))|X_1(0)|) = \mathbb{E}(h(X(0)))\mathbb{E}(|X_1(0)|) = \sqrt{\frac{2\lambda_2}{\pi}}\mathbb{E}(h(X(0)));$$

$$\begin{aligned}\alpha_0(h) &= \mathbb{E}(h(X(0))X_{11}(0)) = \mathbb{E}(h(X(0))\mathbb{E}(X_{11}(0)|X(0))) \\ &= -\lambda_2\mathbb{E}(h(X(0))X(0));\end{aligned}$$

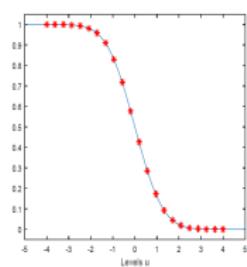
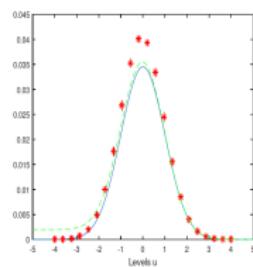
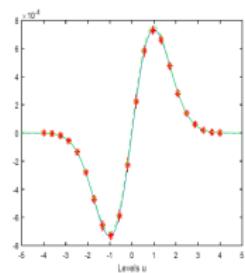
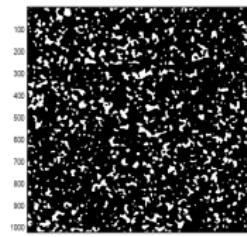
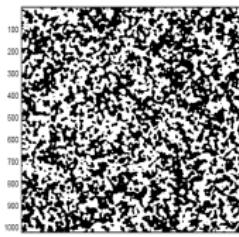
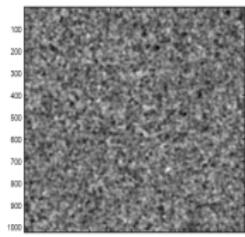
$$\begin{aligned}\alpha_2(h) &= -2\mathbb{E}\left(h(X(0))X_{12}(0)\frac{X_1(0)X_2(0)}{\|\nabla X(0)\|^2}\mathbf{1}_{\|\nabla X\|>0}\right) \\ &= -2\mathbb{E}(h(X(0))X_{12}(0))\mathbb{E}\left(\frac{X_1(0)X_2(0)}{\|\nabla X(0)\|^2}\mathbf{1}_{\|\nabla X\|>0}\right) = 0.\end{aligned}$$

Hence  $TV^*(X) = \sqrt{\frac{\pi\lambda_2}{2}}$  and  $LTC^*(X) = 0$ .

# Gaussian case

For  $X(0) \sim \mathcal{N}(0, 1)$ , this yields to for a.e.  $u \in \mathbb{R}$

$$C_0^*(X, u) = \frac{1}{(2\pi)^{3/2}} \lambda_2 u e^{-\frac{u^2}{2}} \text{ and } C_1^*(X, u) = \frac{1}{4} \lambda_2^{1/2} e^{-\frac{u^2}{2}}.$$



$\rho(x) = e^{-\kappa^2 \|x\|^2}$ , for  $\kappa = 100/2^{10}$  in a domain of size  $2^{10} \times 2^{10}$  pixels.

# Comments

- If one knows that  $u \mapsto C_1^*(X, u)$  or  $u \mapsto C_0^*(X, u)$  are continuous then a.e. is enough ! In Berzin, Latour, Leon (2017) general assumptions to ensure that  $u \mapsto C_1^*(X, u)$  is continuous ;
- For isotropic stationary  $C^3$  Gaussian field the formulas hold for all level (weakest assumptions cf Adler, Taylor (2007)) ; Moreover by kinematic formula

$$\mathbb{E}[C_0(X, T, u)] = C_0^*(X, u)|T| + \frac{1}{\pi} C_1^*(X, u)\mathcal{H}^1(\partial T) + C_2^*(X, u),$$
$$\mathbb{E}[C_1(X, T, u)] = C_1^*(X, u)|T| + \frac{1}{2} C_2^*(X, u)\mathcal{H}^1(\partial T).$$

- Due to Adler and Taylor, using Gaussian kinematic and Tube formulas, computations for fields of Gaussian type :  $X = F(\mathbf{G})$  where  $F : \mathbb{R}^k \mapsto \mathbb{R}^{C^2}$  and  $\mathbf{G} = (G_1, \dots, G_k)$  with  $G_1, \dots, G_k$  iid  $C^3$  homogeneous Gaussian rf.

## Chi2 case

For  $k \geq 1$ ,  $Z_k = G_1^2 + \dots + G_k^2$  and normalized field

$$\tilde{Z}_k(t) := \frac{1}{\sqrt{2k}}(Z_k(t) - k), \quad t \in \mathbb{R}^2.$$

Then, for all  $u \in \mathbb{R}$ ,

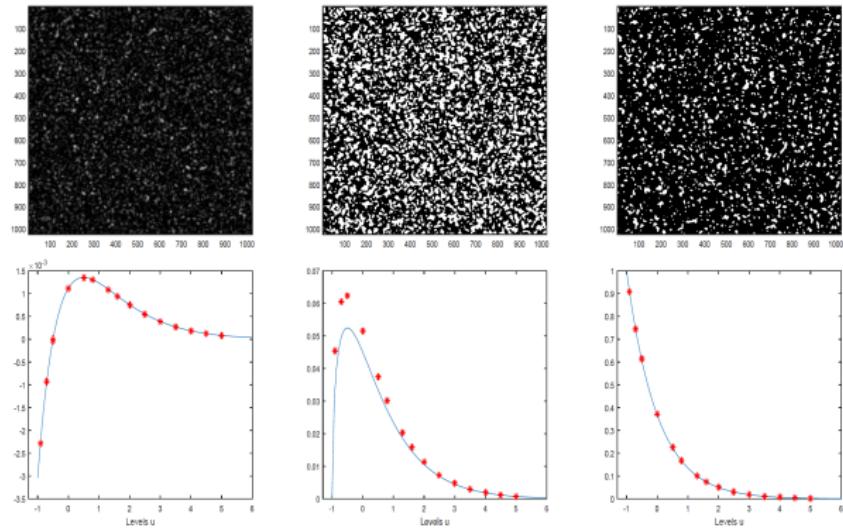
$$C_0^*(\tilde{Z}_k, u) = \frac{\lambda_2}{\pi 2^{k/2} \Gamma(k/2)} \left( k + u\sqrt{2k} \right)^{(k-2)/2} \left( u\sqrt{2k} + 1 \right) \exp \left( -\frac{k + u\sqrt{2k}}{2} \right)$$

$$C_1^*(\tilde{Z}_k, u) = \frac{\sqrt{\pi \lambda_2}}{2^{(k+1)/2} \Gamma(k/2)} \left( k + u\sqrt{2k} \right)^{(k-1)/2} \exp \left( -\frac{k + u\sqrt{2k}}{2} \right),$$

$$C_2^*(\tilde{Z}_k, u) = \mathbb{P} \left( \chi_k^2 \geq k + u\sqrt{2k} \right).$$

$$\text{Hence } TV^*(\tilde{Z}_k) = \sqrt{\frac{2\pi\lambda_2}{k}} \frac{\Gamma((k+1)/2)}{\Gamma(k/2)} \text{ and } LTC^*(\tilde{Z}_k) = \sqrt{\frac{2}{k}} \lambda_2.$$

# Chi2 case



$\tilde{Z}_k$  for  $k = 2$  and for iid standard  $G_1, \dots, G_k$  with covariance function  $\rho(x) = e^{-\kappa^2 \|x\|^2}$ , for  $\kappa = 100/2^{10}$  in a domain of size  $2^{10} \times 2^{10}$  pixels.

## Student case

For  $k \geq 3$ ,  $T_k = G_{k+1}/\sqrt{Z_k/k}$  and normalized field

$$\tilde{T}_k(t) := \sqrt{(k-2)/k} T_k(t), \quad t \in \mathbb{R}^2.$$

Then,

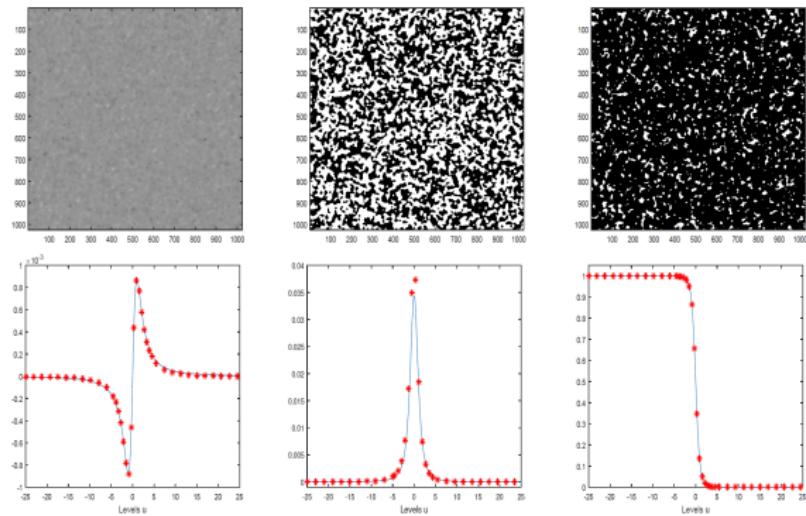
$$C_0^*(\tilde{T}_k, u) = \frac{\lambda_2(k-1)}{4\pi^{\frac{3}{2}}} \frac{u}{\sqrt{k-2}} \frac{\Gamma\left(\frac{k-1}{2}\right)}{\Gamma\left(\frac{k}{2}\right)} \left(1 + \frac{u^2}{k-2}\right)^{\frac{1-k}{2}},$$

$$C_1^*(\tilde{T}_k, u) = \frac{\sqrt{\lambda_2}}{4} \left(1 + \frac{u^2}{k-2}\right)^{\frac{1-k}{2}},$$

$$C_2^*(\tilde{T}_k, u) = \mathbb{P}(Student(k) \geq u\sqrt{k/(k-2)}).$$

Hence  $TV^*(\tilde{T}_k) = \sqrt{(k-2)\pi\lambda_2} \frac{\Gamma((k-2)/2)}{2\Gamma((k-1)/2)}$  and  $LTC^*(\tilde{T}_k) = 0$ .

# Student case



$\tilde{T}_k$  for  $k = 4$  and and for iid standard  $G_1, \dots, G_{k+1}$  with covariance function  $\rho(x) = e^{-\kappa^2 \|x\|^2}$ , for  $\kappa = 100/2^{10}$  in a domain of size  $2^{10} \times 2^{10}$  pixels.

# Shot noise fields

The characteristic function is given by

$$\mathbb{E} \left( e^{itX(0)} \right) = \exp \left( \lambda \int_{\mathbb{R}^k \times \mathbb{R}^2} [e^{i[tg_m(x)]} - 1] F(dm) dx \right).$$

When  $g_m$  is smooth, we have also access to joint law of  $(X(0), \nabla X(0), D^2 X(0))$  via characteristic function and similar integral expression. In particular the joint characteristic function of  $X$  and  $\partial_1 X$  is

$$\begin{aligned}\varphi(t, s) &= \mathbb{E} \left( e^{itX(0)+is\partial_1 X(0)} \right) \\ &= \exp \left( \lambda \iint [e^{itg_m(x)+is\partial_1 g_m(x)} - 1] F(dm) dx \right)\end{aligned}$$

# Isotropic smooth Shot noise fields

The main idea is therefore to take  $h_t(u) = e^{itu}$  to compute

$\widehat{C}_j^*(t) = \int_{\mathbb{R}} e^{itu} C_j^*(X, u) du$ . We obtain integral formulas :

$$\widehat{C}_1^*(t) = \frac{1}{2} \int_0^{+\infty} \frac{\varphi(t, s)}{s} S_0(t, s) ds.$$

$$\widehat{C}_0^*(t) = S_1(t) \varphi(t, 0) + \int_0^{+\infty} \frac{\varphi(t, s)}{s} S_2(t, s) ds,$$

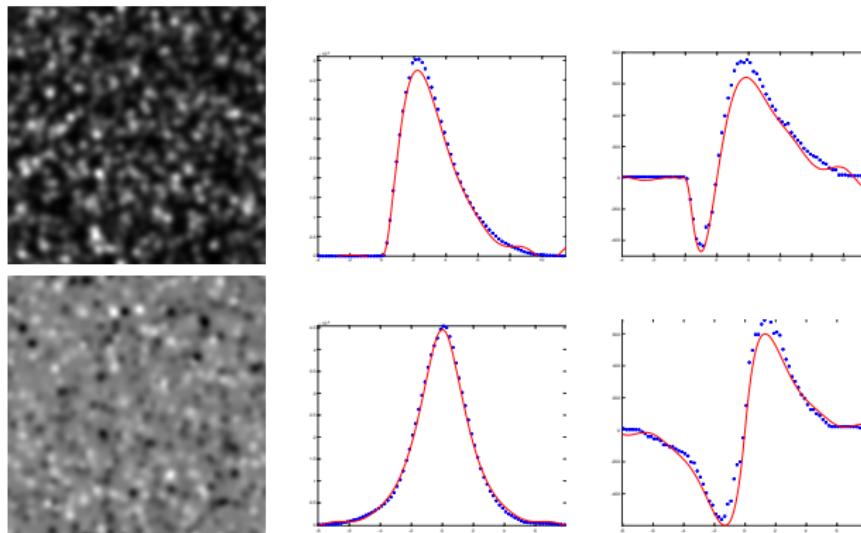
with

$$S_0(t) = -i\lambda \int_{\mathbb{R}} \int_{\mathbb{R}^2} \partial_1 g_m(x) e^{i[tg_m(x) + s\partial_1 g_m(x)]} dx F(dm)$$

$$S_1(t) = -\frac{\lambda}{2\pi} \int_{\mathbb{R}} \int_{\mathbb{R}^2} \partial_1^2 g_m(x) e^{itg_m(x)} dx F(dm)$$

$$S_2(t, s) = \frac{\lambda}{2\pi} \int_{\mathbb{R}} \int_{\mathbb{R}^2} [\partial_2^2 g_m(x) - \partial_1^2 g_m(x)] e^{i[tg_m(x) + s\partial_1 g_m(x)]} dx F(dm)$$

# Shot noise Gaussian examples



We chose  $g_m(x) = we^{-\frac{\|x/r\|^2}{2}}$  with  $R = 1/a > 0$  a.s.

- Top :  $W \sim \mathcal{E}(\mu)$ , we find  $\varphi(t) = \left(\frac{\mu}{\mu-it}\right)^{2\pi\lambda/a}$  and  $X(0) \sim \gamma(\mu, 2\pi\lambda/a)$ ;
- Bottom :  $W \sim \mathcal{L}(\mu)$ ,  $\varphi(t) = \left(\frac{\mu^2}{\mu^2+t^2}\right)^{\pi\lambda/a}$  and  $X(0) \sim GS\mathcal{L}(\mu, \pi\lambda/a)$ .

## Shot noise disk examples

Considering  $g_m(x) = \mathbf{1}_{rD}(x)$  we note

$$\bar{a} = \pi \mathbb{E}(R^2) \text{ and } \bar{p} = 2\pi \mathbb{E}(R)$$

and get

$$X(0) \sim \mathcal{P}(\lambda \bar{a}).$$

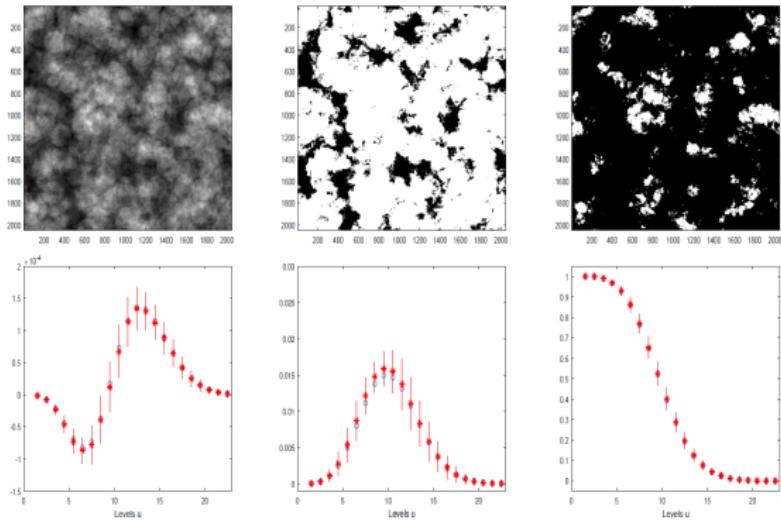
Moreover, for  $u \in \mathbb{R}^+ \setminus \mathbb{Z}^+$ , it holds that

$$C_0^*(u) = e^{-\lambda \bar{a}} \frac{(\lambda \bar{a})^{\lfloor u \rfloor}}{\lfloor u \rfloor!} \lambda \left( 1 - \lambda \frac{\bar{p}^2}{4\pi} + \lfloor u \rfloor \frac{\bar{p}^2}{4\pi \bar{a}} \right),$$

$$C_1^*(u) = \frac{1}{2} e^{-\lambda \bar{a}} \frac{(\lambda \bar{a})^{\lfloor u \rfloor}}{\lfloor u \rfloor!} \lambda \bar{p}$$

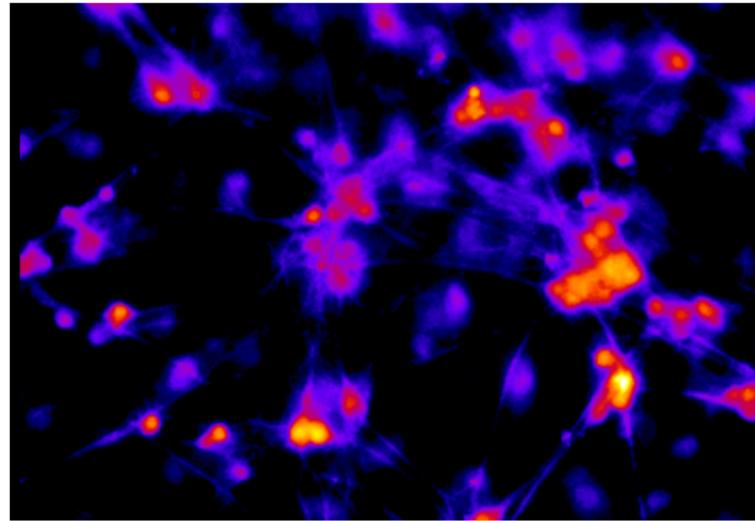
$$C_2^*(u) = e^{-\lambda \bar{a}} \sum_{k>u} \frac{(\lambda \bar{a})^k}{k!}.$$

# Shot noise disk case



$R = 50$  or  $R = 100$  each with probability  $1/2$  and  $\lambda = 5 \times 10^{-4}$  in a domain of size  $2^{10} \times 2^{10}$  pixels.

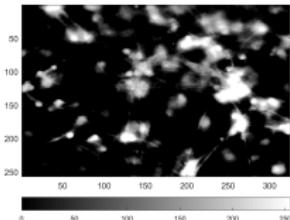
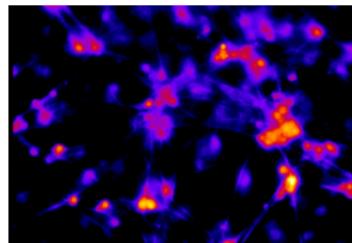
# Biological movies



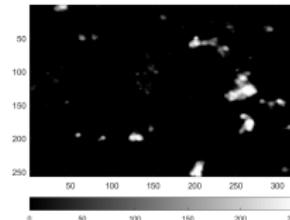
- Calcium imaging recordings for mouse nasal explants
- Context : study of pulsatility for Gonadotropin-Releasing Hormone-1 neurons and calcium event synchronization
- Collaboration : A. Duittoz (INRA Nouzilly) and C. Georgelin (IDP Tours)

# The data

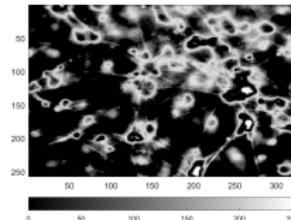
$D = 700$  images of size  $257 \times 325$  pixels in RGB coded in 8 bit :



R



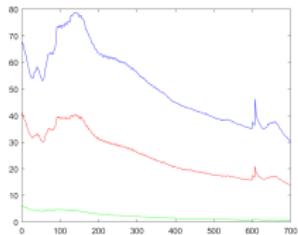
G



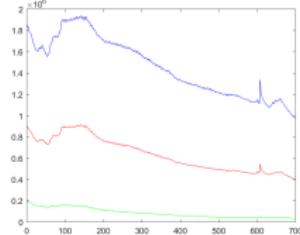
B

# Statistics of the data

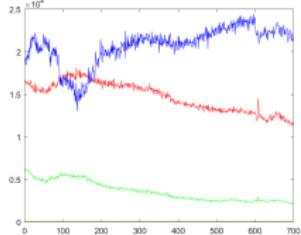
*Evolution according to time*



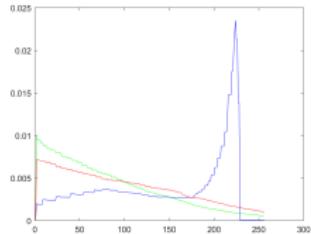
mean



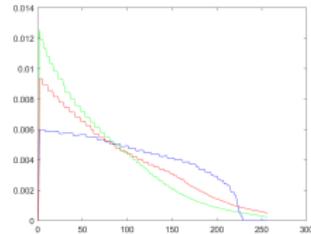
$TV^* \times |T|$



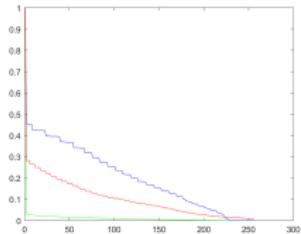
$LTC^* \times |T|$



$C_0^* / \int |C_0^*|$



$C_1^* / TV^*$

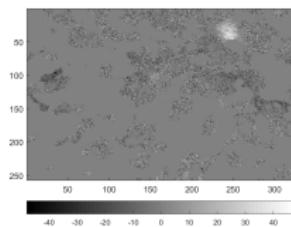
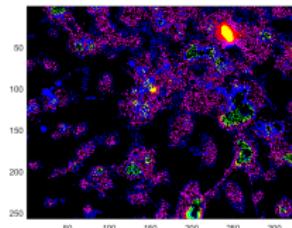


$C_2^*$

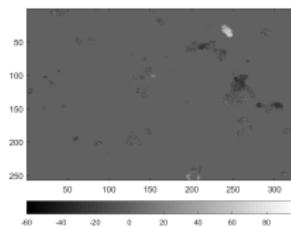
*Mean over time / according to values*

# The time differences

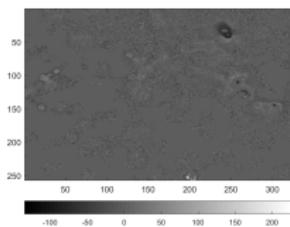
Consider the difference vs time  $\Delta X_t = X_{t+1} - X_t$



R



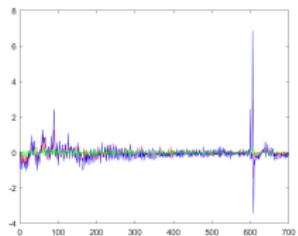
G



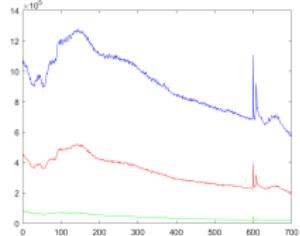
B

# Statistics of time differences

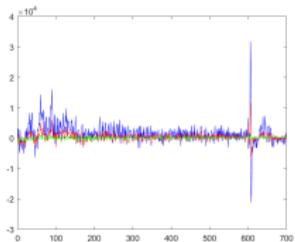
*Evolution according time*



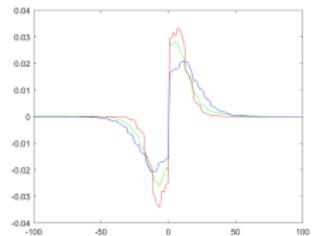
mean



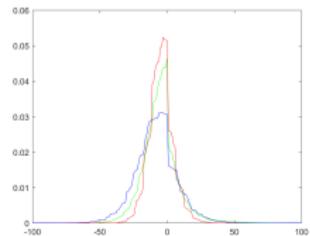
$TV^* \times |T|$



$LTC^* \times |T|$

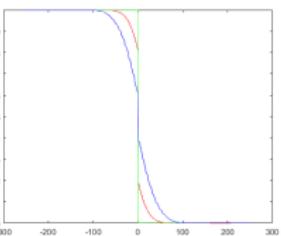


$C_0^* / \int |C_0^*|$



$C_1^* / TV^*$

Mean over time / according to values



$C_2^*$

# Perspectives

- Effect of discretization of random fields
- Use for statistical inference
- Detect main topological changes in a sequence of images
- Beyond expectation : higher order moments (variance at least, and then CLT)
- Higher dimension

# References

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-  H. Biermé, E. Di Bernardino, C. Duval, A. Estrade : Lipschitz Killing curvatures of excursion sets for 2D random fields *to appear in EJS*