

# Journées MAS 2016

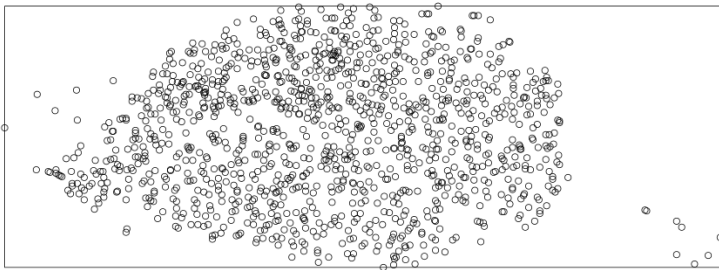
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(joint works with J.S. Gomez, A. Vergne, P. Martins, L. Decreusefond, W. Chen)

Processus Ponctuel de  $\beta$ -Ginibre et Déploiement d'un  
Réseau Cellulaire

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# Introduction



**Figure:** Positions of all base stations in Paris.

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## Framework

- $\mathbb{Y}$  a locally compact metric space
- $\mu$  a diffuse and locally finite measure of reference on  $\mathbb{Y}$
- $N_{\mathbb{Y}}$  the space of configurations on  $\mathbb{Y}$
- $\widehat{N}_{\mathbb{Y}}$  the space of finite configurations on  $\mathbb{Y}$

# Determinantal point process

## Definition

- **Correlation function**  $\rho$  of a point process  $\Phi$ :

$$\mathbb{E}\left[\sum_{\substack{\alpha \in \hat{N}_{\mathbb{Y}} \\ \alpha \subset \Phi}} f(\alpha)\right] = \sum_{k=0}^{+\infty} \frac{1}{k!} \int_{\mathbb{Y}^k} f \cdot \rho(\{x_1, \dots, x_k\}) \mu(dx_1) \dots \mu(dx_k)$$

$\rho(\alpha) \approx$  probability of finding a point in at least each point of  $\alpha$

## Definition

Determinantal point process  $\text{DPP}(C, \mu)$ :

$$\rho(\{x_1, \dots, x_k\}) = \det(C(x_i, x_j), 1 \leq i, j \leq k)$$

## $\beta$ -Ginibre point process (1/2)

### Definition

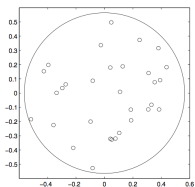
- Ginibre point process with intensity  $\lambda = \frac{\gamma}{\pi}$ :

$$C_{\gamma}(x, y) = \frac{\gamma}{\pi} e^{-\frac{\gamma}{2}(|x|^2 + |y|^2)} e^{\gamma x \bar{y}}$$

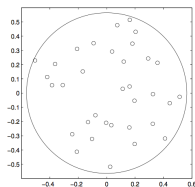
- $\beta$ -Ginibre point process with intensity  $\lambda = \frac{\gamma}{\pi}$ :

$$C_{\gamma, \beta}(x, y) = \frac{\gamma}{\pi} e^{-\frac{\gamma}{2\beta}(|x|^2 + |y|^2)} e^{\frac{\gamma}{\beta} x \bar{y}}$$

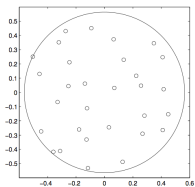
# $\beta$ -Ginibre point process (2/2)



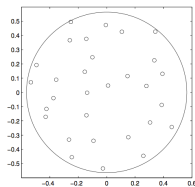
(a) PPP



(b)  $1/4$ -GPP



(c)  $3/4$ -GPP



(d) 1-GPP

# J-function (1/2)

## Definition

- $\Phi$  a stationary point process
- Empty space function:

$$F(r) = \mathbb{P}(\|u - \Phi\| \leq r)$$

- Nearest-neighbor distance distribution function:

$$G(r) = \mathbb{P}(\|u - \Phi \setminus \{u\}\| \leq r)$$

- J-function:

$$J(r) = \frac{1 - G(r)}{1 - F(r)}$$



## J-function (2/2)

### Interpretation

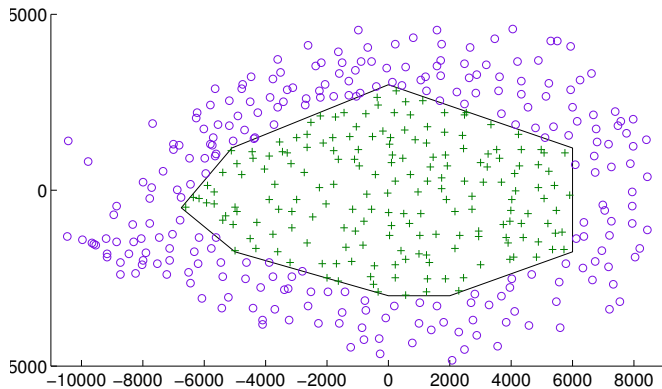
- $J < 1 \implies \Phi$  is attractive
- $J > 1 \implies \Phi$  is repulsive
- $J = 1 \implies \Phi$  is a Poisson point process

### Proposition (Haenggi et al.)

- J-function of a  $\beta$ -Ginibre point process:

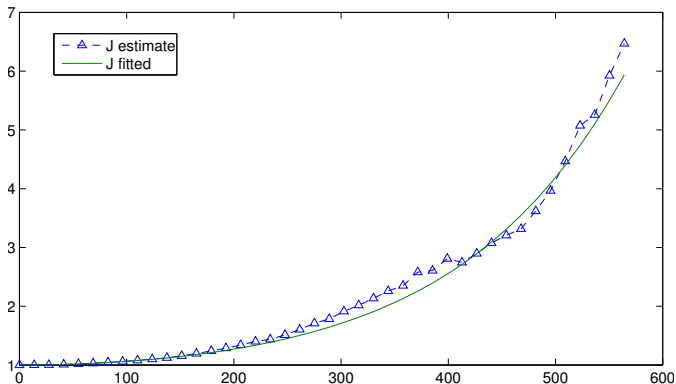
$$J(r) = \frac{1}{1 - \beta + \beta e^{-\frac{\gamma}{\beta} r^2}}$$

## Fitting method



**Figure:** Example of data sample for one GSM operator. The  $J$ -function is fitted on the points within the rectangular window.

# Example of J-function fitting



**Figure:** Example of  $J$ -function fitting for SFR on the 3G 900 MHz band.  $\lambda = 1.92$  base stations per kilometre square  $\beta = 0.97$ . Residuals: 0.74.

# Results per technology and operator

**Table:** Numerical values of  $\beta$  per technology and operator.

	Orange	SFR	Bouygues	Free
GSM 900	0.81	0.76	0.65	NA
GSM 1800	0.84	0.85	0.71	NA
UMTS 900	NA	0.97	0.53	0.89
UMTS 2100	1.04	0.65	0.82	0.89
LTE 800	1.02	0.93	0.67	NA
LTE 1800	NA	NA	0.75	NA
LTE 2600	0.93	0.67	0.63	0.89

# Results per operator and for the superposition

**Table:** Numerical values of  $\beta$  and  $\lambda$  per operator and for the superposition of all the sites.

	Orange	SFR	Bouygues	Free	<b>Superposition</b>
$\beta$	0,94	0,70	0,81	0,89	<b>0,17</b>
$\lambda$	3,48	3,70	4,23	1,05	<b>10,28</b>
Number of sites	185	197	225	56	<b>547</b>

# Papangelou intensity

## Definition

- **Papangelou intensity  $c$**  of a point process  $\Phi$ :

$$\mathbb{E}\left[\sum_{x \in \Phi} f(x, \Phi \setminus \{x\})\right] = \int_{\mathbb{Y}} \mathbb{E}[c(x, \Phi) f(x, \Phi)] \mu(dx)$$

$c(x, \xi) \approx$  conditionnal probability of finding a point in  $x$  given  $\xi$

## Proposition (Georgii et al.)

Papangelou intensity of  $\text{DPP}(C, \mu)$ :

$$c(x_0, \{x_1, \dots, x_k\}) = \frac{\det(H(x_i, x_j), 0 \leq i, j \leq k)}{\det(H(x_i, x_j), 1 \leq i, j \leq k)}$$

where  $H = (I - C)^{-1}C$ .

# Repulsive point process

## Definition

- Point process repulsive if

$$\phi \subset \xi \implies c(x, \xi) \leq c(x, \phi)$$

- Point process weakly repulsive if

$$c(x, \xi) \leq c(x, \emptyset)$$

## Proposition (Georgii et al.)

- $\text{DPP}(C, \mu)$  is repulsive.

# Kantorovich-Rubinstein distance

- Total variation distance:

$$d_{TV}(\nu_1, \nu_2) := \sup_{\substack{A \in \mathcal{F}_Y \\ \nu_1(A), \nu_2(A) < \infty}} |\nu_1(A) - \nu_2(A)|$$

- $F : N_Y \rightarrow \mathbb{R}$  is 1-Lipschitz ( $F \in \text{Lip}_1$ ) if

$$|F(\phi_1) - F(\phi_2)| \leq d_{TV}(\phi_1, \phi_2) \text{ for all } \phi_1, \phi_2 \in N_Y$$

- Kantorovich-Rubinstein distance:

$$d_{KR}(\mathbb{P}_1, \mathbb{P}_2) = \sup_{F \in \text{Lip}_1} \left| \int_{N_Y} F(\phi) \mathbb{P}_1(d\phi) - \int_{N_Y} F(\phi) \mathbb{P}_2(d\phi) \right|$$

- Convergence in K.-R. distance  $\xRightarrow{\text{strictly}}$  Convergence in law



## Upper bound theorem

### Theorem (LD, AV)

- $\Phi$  a finite point process on  $\mathbb{Y}$
- $\zeta_M$  a PPP with finite control measure  $M(dy) = m(y)\mu(dy)$ .

Then, we have:

$$d_{KR}(\mathbb{P}_\Phi, \mathbb{P}_{\zeta_M}) \leq \int_{\mathbb{Y}} \mathbb{E}[|m(y) - c(y, \Phi)|] \mu(dy).$$

# Superposition of weakly repulsive point processes

- $\Phi_{n,1}, \dots, \Phi_{n,n}$ :  $n$  independent, finite and **weakly repulsive** point processes on  $\mathbb{Y}$
- $\Phi_n := \sum_{i=1}^n \Phi_{n,i}$
- $R_n := \int_{\mathbb{Y}} \left| \sum_{i=1}^n \rho_{n,i}(x) - m(x) \right| \mu(dx)$
- $\zeta_M$  a PPP with control measure  $M(dx) = m(x)\mu(dx)$

# Superposition of weakly repulsive point processes

## Proposition (LD, AV)

- $\Phi_n = \sum_{i=1}^n \Phi_{n,i}$
- $\zeta_M$  a PPP with control measure  $M(dx) = m(x)\mu(dx)$

$$d_{\text{KR}}(\mathbb{P}_{\Phi_n}, \mathbb{P}_{\zeta_M}) \leq R_n + \max_{1 \leq i \leq n} \int_{\mathbb{Y}} \rho_{n,i}(x) \mu(dx)$$

## $\beta$ -Ginibre point processes

### Proposition (LD, AV)

- $\Phi_n$  the  $\beta_n$ -Ginibre process reduced to a compact set  $\Lambda$
- $\zeta$  the PPP with intensity  $1/\pi$  on  $\Lambda$

$$d_{\text{KR}}(\mathbb{P}_{\Phi_n}, \mathbb{P}_{\zeta}) \leq \frac{2}{\pi} \frac{\beta_n}{1 - \beta_n} |\Lambda|$$

## References

- J.S. Gomez, A. Vasseur, A. Vergne, P. Martins, L. Decreusefond, and W. Chen, A Case Study on Regularity in Cellular Network Deployment, IEEE Wireless Communications Letters, 2015.
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- N. Deng, W. Zhou, and M. Haenggi, The Ginibre Point Process as a Model for Wireless Networks with Repulsion, 2014.
- H.O. Georgii, and H.J. Yoo, Conditional intensity and gibbsianness of determinantal point processes, J. Statist. Phys. (118), 2004.

Thank you ...

... for your attention. Questions?