

What does a Point Process Outside a Domain tell us about What's Inside ?

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joint work with Subhro Ghosh, Fedor Nazarov, Mikhail Sodin

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Background

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Key Characteristics:

- **Independence** between spatially **disjoint** domains
- In any given domain D , the number of points N follows **Poisson** distribution, with parameter $\propto \text{area}(D)$
- The N points are distributed **uniformly** in D and independently of each other

Limitation: Does not take into account **Spatial Correlation**

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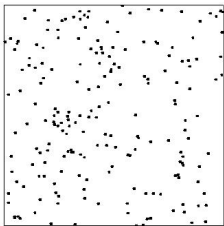
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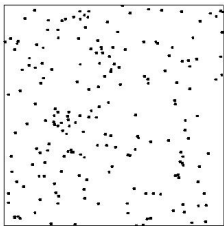
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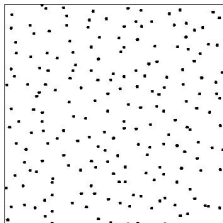


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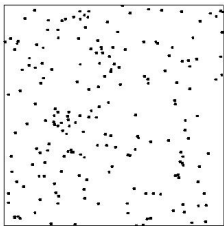


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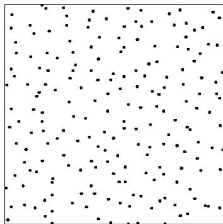


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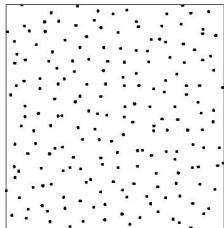
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- Determines expected number of points in a domain

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Convergence of Point Processes:

- $\mu_n \xrightarrow{d} \mu$ iff $\int \varphi d\mu_n \xrightarrow{d} \int \varphi d\mu \quad \forall \varphi \in C_c(\mathbb{C})$

- Finite n : $\mu_n =$ Eigenvalues of $G_n = ((\xi_{ij}))_{1 \leq i, j \leq n}$, ξ_{ij} i.i.d $N_{\mathbb{C}}(0, 1)$ (NO normalization by \sqrt{n})

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Theorem (Sodin rigidity)

$f(z)$ is the unique (up to a deterministic multiplier) Gaussian entire function with a translation invariant zero process of intensity 1.

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Motivation

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- In Poisson point process, the points inside and outside \mathbb{D} are independent of each other

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Theorem (Ghosh, Nazarov, P., Sodin, '12)

In the Ginibre ensemble,

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$M(\omega)$ and $m(\omega)$ positive constants

$\Delta(\zeta) = \prod_{i < j} (\zeta_i - \zeta_j)$ (Vandermonde)

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Limit as $n \rightarrow \infty$ is Poisson : Not Rigid !

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$\Sigma_{S(\omega)}$: constant sum hypersurface $\sum_{i=1}^{N(\omega)} \zeta_i = S(\omega)$ inside $\mathbb{D}^{N(\omega)}$

$M(\omega)$, $m(\omega)$ and $\Delta(\zeta)$ are as before.

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- Application to continuum percolation

Proof Sketch: Rigidity of Number of Points of GAF

- Given: outside zeroes of GAF
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- But $\int \varphi_L d\nu = n(\mathbb{D}) + \int_{\mathbb{D}_L \setminus \mathbb{D}} \varphi_L d\nu$
- Know outside zeroes \Rightarrow Know $\int_{\mathbb{D}_L \setminus \mathbb{D}} \varphi_L d\nu \Rightarrow$ Compute $n(\mathbb{D})$ approximately, now let $L \rightarrow \infty$

An Intriguing Consequence of our Estimates

Proposition (Reconstruction of Gaussian Analytic Function)

The zeroes of the GAF determine the function a.s. (up to a multiplicative factor of modulus 1). In other words, if ν denotes the zeroes of the GAF f , then \exists an analytic function

$$g(z) = \sum_{k=0}^{\infty} a_k(\nu) z^k \text{ such that } f(z) = \gamma \cdot g(z)$$

Here γ follows $\text{Unif}(S^1)$ and is independent of ν .