

# On estimating the structure factor of a point process, with applications to hyperuniformity

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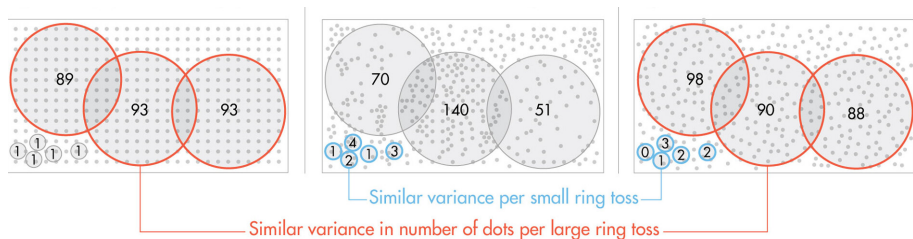
- 1 Hyperuniformity
- 2 Hyperuniformity test
- 3 Numerical experiment
- 4 Code availability 😊

# Hyperuniformity

Let  $\mathcal{X}$  be a stationary point process of  $\mathbb{R}^d$  of intensity  $\rho$ ,  $\mathcal{X}$  is hyperuniform iff

- Variance:

$$\lim_{R \rightarrow \infty} \frac{\text{Var}(\text{Card}(\mathcal{X} \cap B(0, R)))}{|B(0, R)|} = 0.$$



S. Torquato, *Hyperuniform States of Matter*, 2018.

S. Coste, *Order, Fluctuations, Rigidities*, 2021.

# Hyperuniformity using the structure factor

$$\mathcal{X} \text{ is hyperuniform} \iff \lim_{R \rightarrow \infty} \frac{\text{Var}(\text{Card}(\mathcal{X} \cap B(0, R)))}{|B(0, R)|} = 0$$

- Structure factor  $S$  of  $\mathcal{X}$

$$S(\mathbf{k}) = 1 + \rho \mathcal{F}(g - 1)(\mathbf{k})$$

- $\mathcal{X}$  is hyperuniform iff

$$S(\mathbf{0}) = 0$$

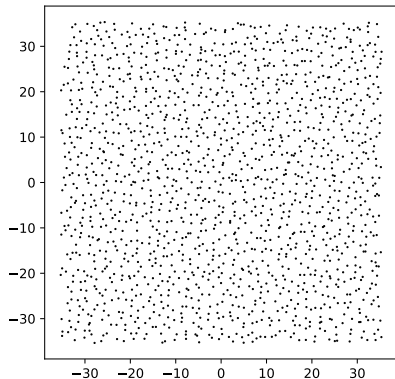
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# Example: Ginibre point process

- $\rho_{\text{Ginibre}} = 1/\pi$
- $S_{\text{Ginibre}}(k) = 1 - \exp(-k^2/4)$
- $S(0) = 0$



# Multiscale hyperuniformity test

- Given: Realizations  $\{\mathcal{X}_W\}$  of  $\mathcal{X}$  in the window  $W$  of lengthside  $L$  (e.g.,  $W = [-L/2, L/2]^d$ )
- Need: Check if  $S(\mathbf{0}) = 0$  using  $\{\mathcal{X}_W\}$
- Problem: We don't have an unbiased estimator of  $S(\mathbf{0})$



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- We have:  $S(\mathbf{k}) = \lim_{L \rightarrow \infty} \mathbb{E} \left[ \widehat{S}(\mathbf{k}) \right]$  for  $\mathbf{k} \in \mathbb{A}_W$ , with  $\|\mathbf{k}_{min}\|_2 \sim \frac{c}{L}$

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- How one can construct an unbiased estimator when only biased estimators are available?

- Need: estimate  $\mathbb{E}[Y] := \bar{Y}$
- Able to generate a sequence of r.v.  $(Y_m)_m$  s.t.  $\bar{Y} = \lim_{m \rightarrow \infty} \mathbb{E}[Y_m]$

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C. Rhee and P.W. Glynn. *Unbiased estimation with square root convergence for SDE models*, 2015.

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- Consider an  $\mathbb{N}$ -r.v.  $M$  s.t.,  $\mathbb{P}(M \geq j) > 0$  for all  $j$ , and let  $Y_0 = 0$

$$Z_m = \sum_{j=1}^{m \wedge M} \frac{Y_j - Y_{j-1}}{\mathbb{P}(M \geq j)}, \quad m \geq 1$$

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# Coupled sum estimator

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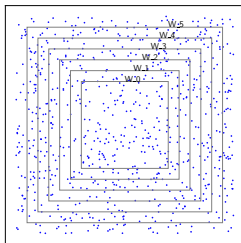
- $\mathbb{E}[Z_m] = \mathbb{E}[Y_m]$  and  $Z_m \xrightarrow[m \rightarrow \infty]{\text{a.s.}} Z := \sum_{j=1}^M \frac{Y_j - Y_{j-1}}{\mathbb{P}(M \geq j)}$ .
- If  $Y_m \xrightarrow[m \rightarrow \infty]{L^2} Y$  + some hypotheses, then  $\mathbb{E}[Z] = \bar{Y}$

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# Multiscale hyperuniformity test

- Consider an increasing sequence of sets  $(\mathcal{X} \cap W_m)_{m \geq 1}$ , with  $\{W_m\}_m \uparrow$  and  $W_\infty = \mathbb{R}^d$
- $\mathbf{k}_m^{\min}$  minimum wavevector of  $\mathbb{A}_{W_m}$ ,  $\mathbf{k}_m^{\min} \xrightarrow{m \rightarrow \infty} \mathbf{0}$



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- Take  $Y_m = 1 \wedge \widehat{S}_m(\mathbf{k}_m^{\min})$
- $Z = \sum_{j=1}^M \frac{Y_j - Y_{j-1}}{\mathbb{P}(M \geq j)}$  with  $M$  is an  $\mathbb{N}$ -r.v. such that  $\mathbb{P}(M \geq j) > 0$  for all  $j$ , and  $Y_0 = 0$

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## Proposition

Assume that  $M \in L^p$  for some  $p \geq 1$ . Then  $Z \in L^p$  and  $Z_m \rightarrow Z$  in  $L^p$ .  
Moreover,

- 1 If  $\mathcal{X}$  is hyperuniform, then  $\mathbb{E}[Z] = 0$ .
- 2 If  $\mathcal{X}$  is not hyperuniform and  $\sup_m \mathbb{E}[\widehat{S}_m^2(\mathbf{k}_m^{\min})] < \infty$ , then  $\mathbb{E}[Z] \neq 0$ .

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Preprint: D. Hawat, G. Gautier, R. Bardenet, R. Lachièze-Rey. *On estimating the structure factor of a point process, with applications to hyperuniformity, 2022.*



Need: Check  $\mathbb{E}[Z] = 0$ , with  $Z = \sum_{j=1}^M \frac{Y_j - Y_{j-1}}{\mathbb{P}(M \geq j)}$

## Test:

- $M$  a Poisson r.v. of parameter  $\lambda$
- i.i.d. pairs  $(\mathcal{X}_a, M_a)_{a=1}^A$  of realizations of  $(\mathcal{X}, M)$
- Asymptotic confidence interval  $CI[\mathbb{E}[Z]]$  of level  $\zeta$

$$CI[\mathbb{E}[Z]] = \left[ \bar{Z}_A - z\bar{\sigma}_A A^{-1/2}, \bar{Z}_A + z\bar{\sigma}_A A^{-1/2} \right]$$

with  $\mathbb{P}(-z < \mathcal{N}(0, 1) < z) = \zeta$

- Assessing whether 0 lies in  $CI[\mathbb{E}[Z]]$

- $\mathcal{X}$  a stationary point process
- $\mathcal{X}_p$  an independent  $p$ -thinning with  $p \in (0, 1)$
- Structure factor:  $S_p(\mathbf{k}) = pS(\mathbf{k}) + 1 - p$
- $\mathcal{X}$  is hyperuniform  $\implies S_p(\mathbf{0}) = 1 - p$

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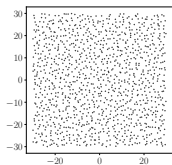
J. Kim and S. Torquato. *Effect of imperfections on the hyperuniformity of many-body systems*, 2018.

M. A. Klatt, G. Last, and N. Henze. *A genuine test for hyperuniformity*, 2022.

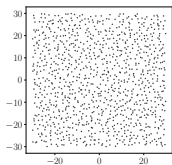
# Multiscale hyperuniformity test

Numerical experiment

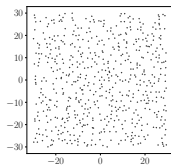
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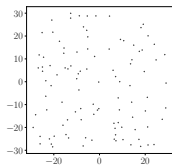
Ginibre,  $S(0) = 0$



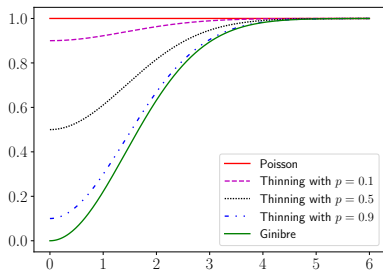
$p = 0.9$ ,  $S(0) = 0.1$



$p = 0.5$ ,  $S(0) = 0.5$

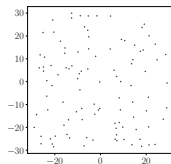
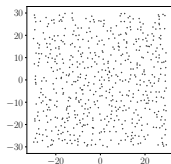
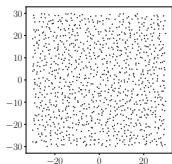
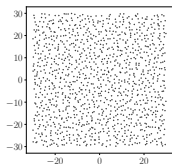


$p = 0.1$ ,  $S(0) = 0.9$



Structure factor

# Multiscale hyperuniformity test



Ginibre,  $S(0) = 0$


$\rho = 0.9$ ,  $S(0) = 0.1$

$\rho = 0.5$ ,  $S(0) = 0.5$

$\rho = 0.1$ ,  $S(0) = 0.9$

Table: Multiscale hyperuniformity test obtained using  $\widehat{S}_{BI}$  on the thinned Ginibre process.

	$\bar{Z}_A$	$C/\mathbb{E}[Z]$
Ginibre	0.0057	$[-0.0042, 0.0156]$
Thinning $\rho = 0.9$ , $S(\mathbf{0}) = 0.1$	0.0865	$[0.0411, 0.1318]$
Thinning $\rho = 0.5$ , $S(\mathbf{0}) = 0.5$	0.5722	$[0.4227, 0.7217]$
Thinning $\rho = 0.1$ , $S(\mathbf{0}) = 0.9$	0.611	$[0.2082, 1.0137]$

 Code availability

- 1 Open-source 🐍 Python toolbox called `structure_factor`<sup>1</sup>
- 2 Available on 🐙 GitHub and PyPI<sup>2</sup>
- 3 Detailed documentation<sup>3</sup>
- 4 Jupyter notebook tutorial<sup>4</sup>

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<sup>1</sup><https://github.com/For-a-few-DPPs-more/structure-factor>

<sup>2</sup><https://pypi.org/project/structure-factor/>

<sup>3</sup><https://for-a-few-dpps-more.github.io/structure-factor/>

<sup>4</sup><https://github.com/For-a-few-DPPs-more/structure-factor/tree/main/notebooks>

# Conclusion

- Statistical test of hyperuniformity using biased estimators of the structure factor
- Python toolbox `structure-factor`

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# THANK YOU

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Github



Documentation



Preprint

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Github: <https://github.com/For-a-few-DPPs-more/structure-factor>

Documentation: <https://for-a-few-dpps-more.github.io/structure-factor/>

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