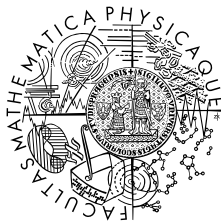


# Stochastic reconstruction for inhomogeneous point processes



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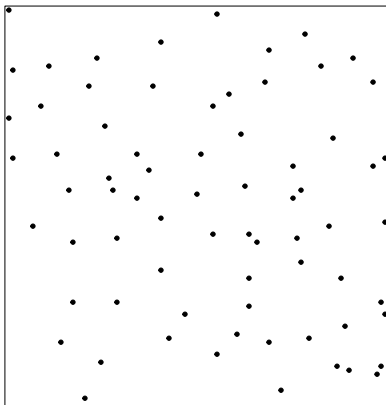
**25. 1. 2018**

# Overview

- Planar point processes
  - ▶ Definition
  - ▶ Intensity measure
  - ▶ Summary characteristics
- Importance of simulations
- Stochastic reconstruction
  - ▶ Algorithm
  - ▶ Applications
- Inhomogeneous case

# Planar point processes

- A random countable subset of the two-dimensional Euclidean space.

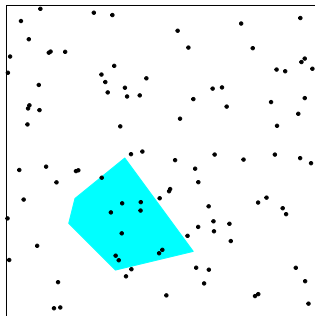


# Planar point processes

- Planar point process  $X \rightarrow$  random variable  $X(B)$ ,  $B \in \mathfrak{B}^2$ .
- Intensity measure:

$$\Lambda(B) = \mathbb{E}[X(B)] = \int_B \lambda(u) \, du, \quad B \in \mathfrak{B}^2$$

- Constant  $\lambda$  is called **intensity**.



# Planar point processes

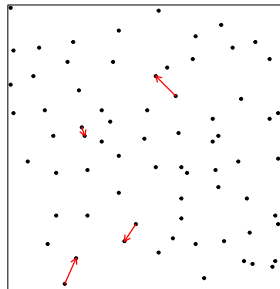
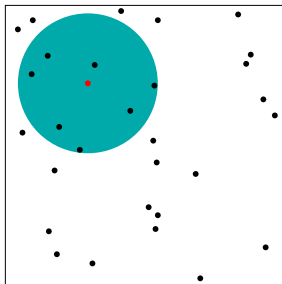
Summary characteristics:

- **numerical**

- ▶ intensity,

- **functional**

- ▶ K-function,
- ▶ nearest neighbour distance distribution function.



# Importance of simulations

- Simulations may be used to:
  - ▶ test statistical hypotheses,
  - ▶ assess estimation variance,
  - ▶ ...
- Simulation methods:
  - ▶ generate samples from explicit point process model,
  - ▶ parameter estimation.

# Stochastic reconstruction (Tscheschel and Stoyan 2006)

- Algorithmic procedure.
- Useful tool for generating point patterns with prescribed summary characteristic.
- No need to specify any explicit model assumptions.

Notation:

- $X, Y \dots$  point patterns
- $n_i \dots$   $i$ -th numerical summary characteristic
- $f_j \dots$   $j$ -th functional summary characteristic
- $R_j \dots$  constant, depends on the observation window
- $\hat{n}_i \dots$  empirical estimator of the  $i$ -th numerical s. c.
- $\hat{f}_j \dots$  empirical estimator of the  $j$ -th functional s. c.

# Stochastic reconstruction

$$E_{n_i}(X, Y) = [\hat{n}_i(X) - \hat{n}_i(Y)]^2$$

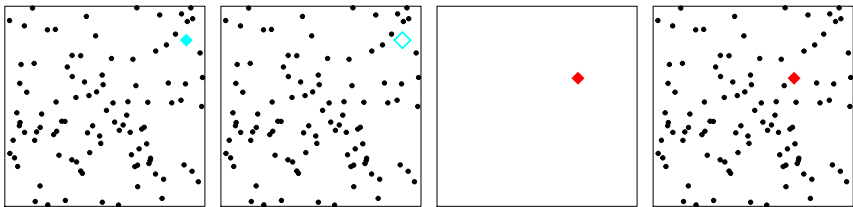
$$E_{f_j}(X, Y) = \int_0^{R_j} [\hat{f}_j(r, X) - \hat{f}_j(r, Y)]^2 dr$$

- Energy functional:

$$E(X, Y) = \sum_{i=1}^I E_{n_i}(X, Y) + \sum_{j=1}^J E_{f_j}(X, Y)$$



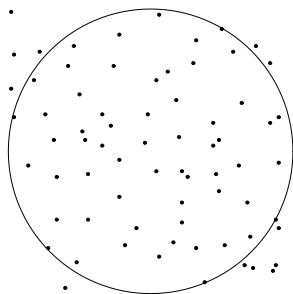
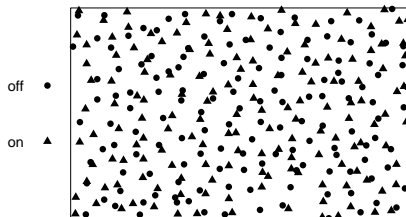
# Stochastic reconstruction



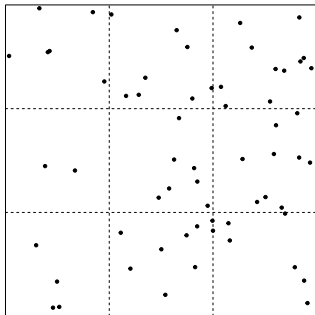
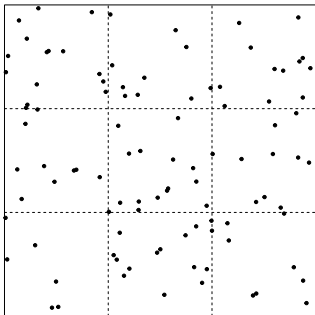
# Stochastic reconstruction

## Applications:

- test of independence of two different patterns
  - ▶ Getzin, Wiegand and Hubbell (2014)
- conditional reconstruction
  - ▶ Pommerening (2006)
  - ▶ Tscheschel and Chiu (2008)







# Inhomogeneous case



## Modifications of the algorithm:

- initial configuration + generating new points,
- quality of the reconstruction,
- energy functional.

# References

-  A. Pommerening, “Evaluating structural indices by reversing forest structural analysis”, *Forest Ecology and Management*, vol. 224, pp. 266–277, 2006.
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-  S. Getzin, T. Wiegand and S. P. Hubbell, “Stochastically driven adult-recruit associations of tree species on Barro Colorado Island”, *Proceedings of the Royal Society, Series B*, vol. 281, 2014.