

# Probability & Geometry on Configuration Spaces

Berlin,  
July '23

- 1) Basics in discrete percolation
- 2) Recent results on  $k$ -nearest-neighbor percolation
- 3) Basics in continuum percolation
- 4) Continuum percolation in random environment (dynamics)

## Section I

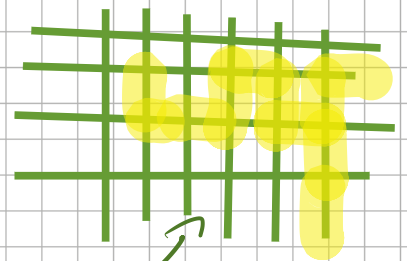
### References

- 1) Grimmett: Percolation, 99
- 2) Bollobás, Riordan: -4-, 06
- 3) Dominic-Copin: Lecture notes 18

and many more

Consider iid  $\text{Ber}(p)$ -distributed RV  $(e_i)_{i \in B}$   
 $B$  bonds of  $\mathbb{Z}^d$ ,  $p \in [0, 1]$

If  $e_i = 1$  we call bond open and if  
 $e_i = 0 \rightarrow$  closed.



real. of conp.  
of open bonds in  $\mathbb{Z}^d$

Percolation-probability

$$\Theta(p) := \mathbb{P}_p(0 \overset{\text{via open bonds}}{\longleftrightarrow} \infty)$$

Critical parameter

$$p_c := \inf \{ p > 0 : \Theta(p) > 0 \}$$

Thm 1 (Nontriviality of Bond-percolation)

For  $d \geq 2$  we have  $0 < p_c < 1$

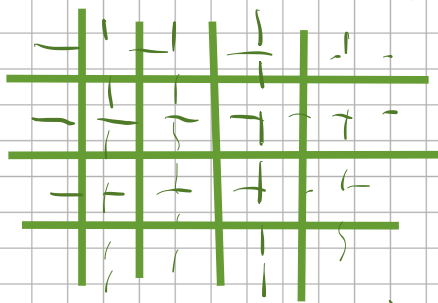
Remark: For  $p < p_c$  the system is subcritical.  
 For  $p > p_c$  is supercritical.

Proof: " $0 < p_c$ " via 1-st moment bound.

$$\Theta(p) \stackrel{V_n}{\leq} P_p(\exists \text{ selfavoiding open path of length } n \text{ starting at } o)$$

$$\stackrel{\text{Markov}}{\leq} \sum_{\text{Such path}} p^n \leq (2dp)^n \xrightarrow{\text{for } p < \frac{1}{2d}} 0 \Rightarrow p_c \geq \frac{1}{2d}$$

" $p_c < 1$ " via Peierls' argument for  $d=2$  (perc. in  $d=2 \Rightarrow$  perc. in  $d>2$ )  
 $\mathbb{Z}^d := \mathbb{Z}^d + (\frac{1}{2}, \frac{1}{2})$



$e$  closed  $\Leftrightarrow e'$  closed

$$1 - \Theta(p) \leq P_p(\exists \text{ closed interface in } \mathbb{Z}^{d+2} \text{ surrounding the origin})$$

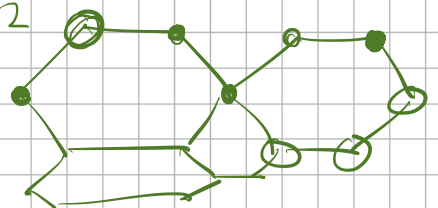
$$\leq \sum_{n \geq 0} P_p(\exists \text{ closed path in } \mathbb{Z}^{d+2} \text{ of length } 2n+4 \text{ passing through } (n+\frac{1}{2}, 0))$$

$$\leq \sum_{n \geq 0} \underbrace{(4(1-p))^{2n+4}}_{\leq 2d} < 1 \text{ for } p \text{ suff. close to 1.}$$

Rem: 1)  $E[\text{degree}(o)] = p 2d \Rightarrow$  for  $p < \frac{1}{2d}$   $E[\text{degree}(o)] < 1$

and people call  $\frac{1}{2d}$  the branching bound. The criterion holds for a large class of graphs. In  $\mathbb{Z}^d$   $p_c(d) \cdot 2d \xrightarrow{d \rightarrow \infty} 1$ .

2) Similar arguments lead to  $0 < p_c^{\text{site}} < 1$  for Bond-site-percolation. For example for triangular lattice  $p_c^{\text{site}} = \frac{1}{2}$  in  $d=2$



3) In  $d=1$   $p_c = 1$

Note that  $\Theta(p) > 0 \iff P_p(\underbrace{\exists \text{ infinite connected open component}}_{\text{cluster}}) > 0$

Indeed:  $\Theta(p) > 0 \Rightarrow P_p(\underbrace{\text{cluster}}_{\text{cluster}}) > 0$

and if  $\Theta(p) = 0$  then  $P_p(\text{cluster}) = P_p(\exists x \in \mathbb{Z}^d: x \leftrightarrow \infty)$   
 $\leq \sum_{x \in \mathbb{Z}^d} \underbrace{P_p(x \leftrightarrow \infty)}_{\Theta(p)} = 0$  (translation invariance)

Is it possible that  $P_p(\exists \text{ inf. cluster}) \in (0, 1)$ ?

Answer: No, by ergodicity. Consider

translation-invariant sets:  $A \subset \{0, 1\}^{\mathbb{Z}^d}$  with  $A+x = A \ \forall x \in \mathbb{Z}^d$   
 ex:  $\{\exists \text{ inf. cluster}\} = A$

translation-invariant measure:  $\mu(A+x) = \mu(A) \ \forall \text{ events } A, x \in \mathbb{Z}^d$   
 ergodic measure  $\mu(A) \in \{0, 1\} \ \forall \text{ transl. inv. sets } A$

Lemma 2 (Ergodicity)  $P_p$  is ergodic.

proof: It suffices to prove  $P_p(A) \leq P_p(A)^2 \ \forall \text{ transl.-inv. sets } A$ .

Note that  $\forall \varepsilon > 0 \exists \text{ event } B \text{ depending on only finitely many bonds and such that } P_p(A \Delta B) < \varepsilon$ . Since

$B$  dep. on only finitely many bonds  $\exists x \in \mathbb{Z}^d$  such that

$$B \cap (B+x) = \emptyset \Rightarrow P_p(B \cap (B+x)) = P_p(B)^2$$

by transl.-invariance.  $\Rightarrow P_p(A) = P_p(A \cap A) = P_p(A \cap (A+x))$

$$\leq P_p(B \cap (B+x)) + 2\varepsilon = P_p(B)^2 + 2\varepsilon$$

$$\leq P_p(A)^2 + 4\varepsilon \xrightarrow{\varepsilon \downarrow 0} P_p(A)^2 \quad \square$$

This implies that for  $p > p_c$   $P_p(\exists \text{ inf. cluster}) = 1$

What about uniqueness?

Thm 3 For  $p \in [0, 1]$  with  $\Theta(p) > 0$ :  $P_p(\exists \text{ a unique inf. cluster}) = 1$

## Section II (Recent results on $k$ -nearest-neighbor percolation)

Ref: 1) Percolation in lattice  $k$ -nearest-neighbor graphs

(J. Köppl, A. Tóbiás, B. Loebl, 87)

2) Oriented percolation in  $d \geq 4$

(Cox & Durrett, 83)

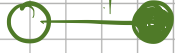
23)

3) Percolation (Grimmett)

We can consider a huge variety of lattice per. models. For example

1) Site percolation

2) Mixed percolation in  $d=2$ : open horizontal bonds with prob.  $p_h$   
open vertical — " —  $p_v$

3) AB percolation: iid field of occupied / unoccupied sites  
draw edge whenever 

4) Long-range percolation.

5) First-passage percolation.  $(\tau_i)_{i \in \mathbb{Z}}$  iid passage times on edges  
wet site at time  $t$

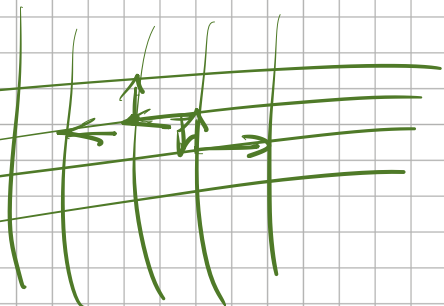
$$w_t = \{x \in \mathbb{Z}^d: \exists \text{ path } \gamma: 0 \rightarrow x \text{ with } \sum_{i \in \gamma} \tau_i \leq t\}$$

6) Directed percolation:

e.g.  $d=2$  iid vertex  $x$  has an edge facing north with prob.  $p$   
(and south otherwise) and an edge facing west — " —  
(and east otherwise)

$$P_p(0 \leftrightarrow \infty) = 0 \text{ for } p = \frac{1}{2}$$

$$P_p(0 \leftrightarrow \infty) > 0 \text{ for } p \neq \frac{1}{2}?$$



Consider  $k$ -neighbor percolation: each vertex chooses precisely  $k$  out of its  $2d$  nearest neighbors uniformly at random and we open an edge towards them.



This gives rise to (at least) three models

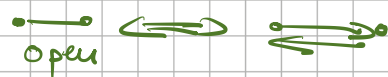
i) directed  $k$ -ng ( $k$ -Dng) as the resulting random graph of directed open edges



ii) undirected  $k$ -ng ( $k$ -Ung) forget arrows



iii) bidirectional  $k$ -ng ( $k$ -Bng)



We are interested in  $\Theta^*(k, d) = \mathbb{P}(0 \rightarrow \infty \text{ in } k\text{-ng in } \mathbb{Z}^d)$

$$\Theta^B(k, d) \leq \Theta^D(k, d) \leq \Theta^U(k, d) \quad k \in \{B, D, U\}$$

## Sec II.1 ( $k$ -Dng)

Prop. 4.  $\Theta^D(1, d) = 0$  and  $\Theta^D(k, d) = 1$  for  $k \geq d+1$

proof:  $\mathbb{P}(\exists \text{ path starting at } o \text{ of length } n) \leq (2d)^n \xrightarrow{n \rightarrow \infty} 0$ .  
Further, for  $k \geq d+1$

$$\max_{x \in G_n} \|x\|_1 \leq \max_{x \in G_{n+1}} \|x\|_1 \text{ a.s. where}$$

$$G_n := \{x \in \mathbb{Z}^d: \exists \text{ open path of length } n \text{ from } 0 \rightarrow x\}$$

Thm 5  $\Theta^D(k, d) > 0$  for  $d, k \geq 4$  or  $d \geq 5, k = 3$ .

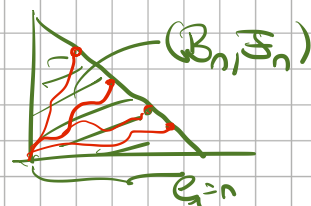
proof: We use Hacking argument. Consider 1st quadrant, and let

$\mathcal{F}_n$  be the  $\sigma$ -algebra up to  $\ell_1$ -distance  $n$ . Let

$$N_n := \# \text{ open paths in } R_n$$

$\uparrow$  set of directed paths

$$0 \leadsto \delta B_n$$



Note that 
$$\mathbb{E}[N_{n+1} | \mathcal{F}_n] = \frac{k}{2} N_n$$

$\uparrow$

$\Rightarrow W_n := \left(\frac{2}{k}\right)^n N_n$  is a  $\mathcal{F}_n$ -Martingale.

$\hookrightarrow$  Expected # of outwards open edges.

Since  $E[W_1] = 1$

$W_n \xrightarrow{n \uparrow \infty} W$  IP-a.s. and  $P(W > 0) > 0$  if

$$\lim_{n \uparrow \infty} E[W_n^2] < \infty$$

But  $E[N_n^2] = \sum_{s, t \in \mathcal{R}_n} P(s, t \text{ open}) \leq \sum_{s, t \in \mathcal{R}_n} p^{2n - K(s, t)}$

by negative correlation

$p = \frac{k}{2d}$   $\uparrow$  # joint edges in  $s, t$ .

$$= (dp)^{2n} E[p^{-K(n)}]$$

$\uparrow$  # joint edges of two independent random walks  $(S_m, S'_m)_{m \leq n}$

$$\Rightarrow E[W_n^2] \xrightarrow{n \uparrow \infty} E[p^{-K}]$$

$\uparrow$  total # of joint rw edges.

$$P(K=k) = g_d^k (1 - g_d) \text{ strong}$$

$\uparrow P(\exists m: S_m = S'_m, S_{m+1} = S'_{m+1})$

$$\Rightarrow E[p^{-K}] < \infty \iff \frac{k}{2d} > g_d$$

$\uparrow$  precise evaluation yields the result  $\square$

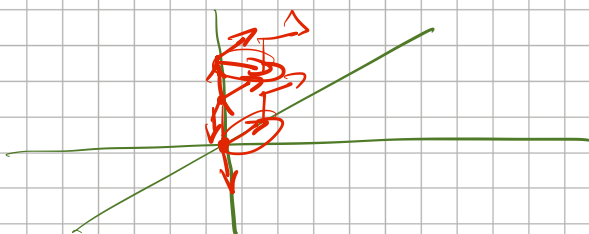
Open problem:  $\forall d \geq 1$   $\Theta^D(2, d) > 0$

overwhelming numerical evidence!!

Theorem 6  $\Theta^D(k+1, d+1) \geq \Theta^D(k, d)$

Proof idea: We use coupling arguments.  $k, d \geq 2$

algorithm: Go up/down until you find two non-vertex edges.



$$\begin{aligned}
 P_{2,2}(\hat{A}_n^{(2)}) &= \mathbb{E}_{3,3} \left[ \gamma(\hat{A}_n^{(2)} | \cdot) \right] \\
 &\quad \downarrow \{0 \rightarrow \delta B_n \text{ in } \mathbb{Z}^2\} \quad \downarrow \text{coupling measure} \\
 &= \mathbb{E}_{3,3} \left[ \gamma(\hat{A}_n^{(2)} | \cdot) 1_{A_n^{(3)}} \right] + \mathbb{E}_{3,3} \left[ \gamma(\hat{A}_n^{(2)} | \cdot) 1_{A_n^{(3)c}} \right] \\
 &\leq \mathbb{P}_{3,3}(\hat{A}_n^{(3)}) \leq 1
 \end{aligned}$$

□

## Sec II 2 (k-ung)

Thm 7  $\Theta^u(1,d) = 0 \quad \forall d \geq 1$  but  $\Theta^u(2,2), \Theta^u(3,3) > 0$

Proof: 1) Assume  $\mathbb{P}(A) > 0$  with  $A = \{0 \rightarrow \infty \text{ in } 1\text{-ung}\}$   
and  $0 = x_1, x_2, \dots \rightarrow \infty$  in  $A$

Let  $K := \inf \{k \geq 0 : (x_k, x_{k+1}) \text{ not in } 1\text{-Dng}\}$

$L = \inf \{k > K : (x_k, x_{k+1}) \text{ in } 1\text{-Dng}\}$



Since  $\Theta^D(1,d) = 0$   $\mathbb{P}(K = \infty) = 0$   $\mathbb{P}(L < \infty) = 1$

Spatially, let  $\kappa := \inf \{k > 0 : \bigcup_{\text{directed}} (0) \subset B_k\}$   
 $\mathbb{P}(\kappa = k_0 | A) \geq \varepsilon$  for some  $k_0$   
 since  $\mathbb{P}(\kappa = 2) = 1$   
 ↑ outgoing directed cluster of 0

$$\Rightarrow \varepsilon \mathbb{P}(A) \leq \mathbb{P}(\kappa = k_0, A)$$

$$\begin{aligned}
 &\stackrel{k_n}{\leq} \mathbb{P}(\exists \text{ selfavoiding path of length } n \text{ starting} \\
 &\quad \text{from } \delta B_n \text{ in } 1\text{-Dng}) \xrightarrow{n \uparrow \infty} 0 \text{ by Prop 4}
 \end{aligned}$$

## Section III (Continuum percolation) Ref. - Cont. Percol. Meester & Rogers

1) Goueree '08 '96

Consider point cloud in  $\mathbb{R}^d$  without interaction, the Poisson point process (PPP)

$X^\lambda = (X_i^\lambda)_{i \in I}$  with intensity  $\lambda dx$ , i.e.

1)  $\mathbb{P}(X(A) = n) = \text{Pois}_{\lambda|A|}(n)$

$\uparrow$  Lebesgue volume of  $A$

2) Total independence  $\mathbb{E}[f(X_A^\lambda)g(X_B^\lambda)]$

$= \mathbb{E}[f(X_A^\lambda)]\mathbb{E}[g(X_B^\lambda)]$

$\forall f, g \geq 0$  mb.  $A, B \in \mathcal{B}(\mathbb{R}^d)$   $A \cap B = \emptyset$ .

$X^\lambda$  is stationary & ergodic wrt shifts in  $\mathbb{R}^d$

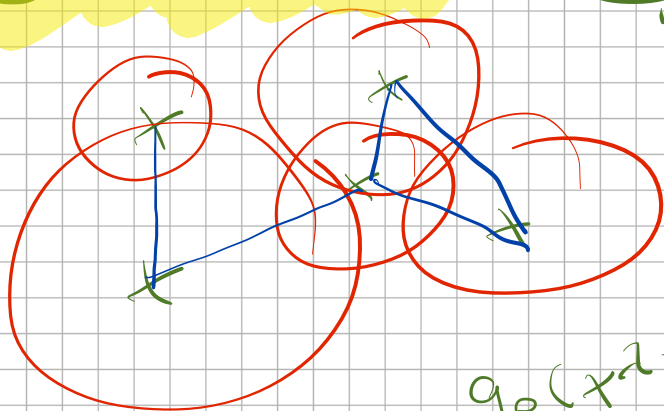
Attach iid marks  $(g_i)_{i \in I}$  with  $g \geq 0$

$\mathbb{P}(g > 0) > 0$  and consider

Boolean model

$C = \bigcup_{i \in I} B_{g_i}(x_i)$

$\uparrow$  balls with radius  $g_i$  centered at  $x_i$



Associated random graph

$g_g(X^\lambda)$  with vertices  $X^\lambda$  and edges  $(x_i, x_j) \iff B_{g_i}(x_i) \cap B_{g_j}(x_j) \neq \emptyset$ .

1)  $g \equiv r$  :  $g_r(X^\lambda)$  is called Gilbert graph.

Percolation:  $\lambda_c(r) := \inf \{ \lambda > 0 : \mathbb{P}(g_r(X^\lambda) \text{ contains an unbounded component}) > 0 \}$

Observe: i)  $\{g_r(X^\lambda) \text{ perc.}\}$  is shift inv., hence

$$P(g_r(x^d) \text{ per.}) > 0 \iff P(g_r(x^d) \text{ per.}) = 1$$

ii) Scale invariance

$$P(g_r(x^d) \text{ per.}) = P(g_1(x^{\frac{rd}{2}}) \text{ per.})$$

$$[\text{degree: } E[X^2(B_r(0))] = 2\lambda_d r^d]$$

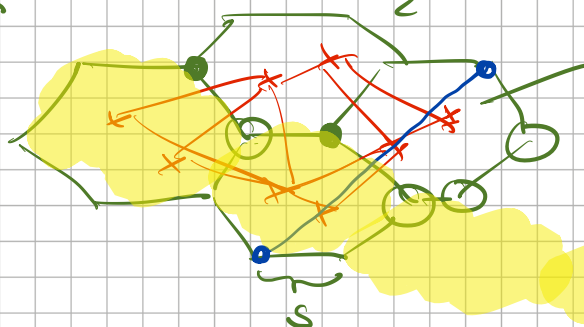
$$Is \quad 0 < \lambda_c(1) < \infty ?$$

$\uparrow$  subcr.       $\uparrow$  supercr.

Sec III.1 Supercritical phase

Thm 10 For  $d \geq 2$   $\lambda_c(1) < \infty$

proof: via discretization: for  $d=2$ :



Face percolation in hexagons  
= site percolation in  
triangular lattice  $p_c = \frac{1}{2}$

Call face  $\star$  open  $\iff X(\star) > 0$

Note: 1)  $|A| = c s^2$  (  $c = \frac{3\sqrt{3}}{2}$  )

$$2) P(X(\star) = 0) = e^{-2|A|}$$

3) If  $1 - e^{-2|A|} > \frac{1}{2}$  we see face percolation.

Now  $|x_i - x_j| < c's \quad \forall x_i, x_j \text{ in neighb. faces}$

$\Rightarrow$  If  $c's < 1$  points  $x_i, x_j$  in  $\sim$  are  
conn. in  $g_1(x^d)$

$\Rightarrow$  Choose  $s < \frac{1}{c}$  and  $\lambda > \frac{\log 2}{c s^2}$  we see

face proc. implying proc. is  $g_r(x^2)$   $\square$

This can be extended to random marks  $g_i$ .

What about  $d=1$ ?

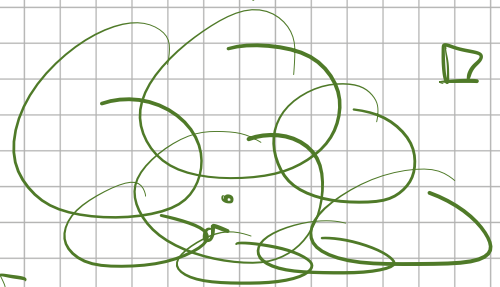
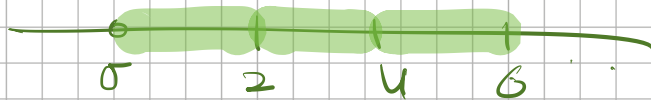
Note  $P(\exists \text{ inf. clust}) > 0 \Leftrightarrow P^0(0 \rightarrow \infty) > 0$

since  $\Leftarrow^u$  clus and  $\Rightarrow^u \exists n$  st.

$$\begin{aligned} 0 &< P(\exists \text{ inf. clust } C: C \cap X_{B_n} \neq \emptyset) \\ &\leq E \left[ \sum_{x \in B_n} 1\{x_i \in C\} \right] \underset{\text{Mecke } B_n}{=} \int dx P(x \in C) \\ &= |B_n| P^0(0 \rightarrow \infty) \end{aligned}$$

Thm 11 For  $d=1$   $\lambda_c(1) = \infty$

proof: Note that  $\Theta_n := P^0(0 \rightsquigarrow \delta B_n)$   
 $\leq (1 - e^{-\lambda_2})^{\frac{1}{2}} \xrightarrow{n \uparrow \infty} 0$



Sec III.2 Subcritical phase

Problem:  $E^0[\text{degree}(o)] = E^0[\#\{x_i: |x_i| < s_0 + s_1\}]$   
 $= E_{s_0, s_1}[\lambda |B_{s_0+s_1}|] = \lambda v_d E[(s_0+s_1)^d]$

Indeed if  $E[s^d] = \infty$   $\in (0, \infty]$

$\Rightarrow P(C = \mathbb{R}^d) = 1$  full coverage  
 true for any stationary point process



But even if  $E[g^d] < \infty$ , do we have a subcritical regime?

Thm 12 For any  $d \geq 1$  if  $E[g^d] < \infty$

$$\lambda_C := \inf \{ \lambda > 0 : \mathbb{P}^0(C_0 \text{ is unbounded}) > 0 \} > 0$$

$\uparrow$  con. com. of origin in  $C$

proof: via multiscale argument (Gouérou '08)

$$G(x, \alpha) := \{ C_\alpha(x) \cup B_{g_\alpha}(x) \}$$

connected component attached  
to  $B_{g_\alpha}(x)$  using points  
only in  $X|_{B_{10\alpha}(x)}$

Two error events

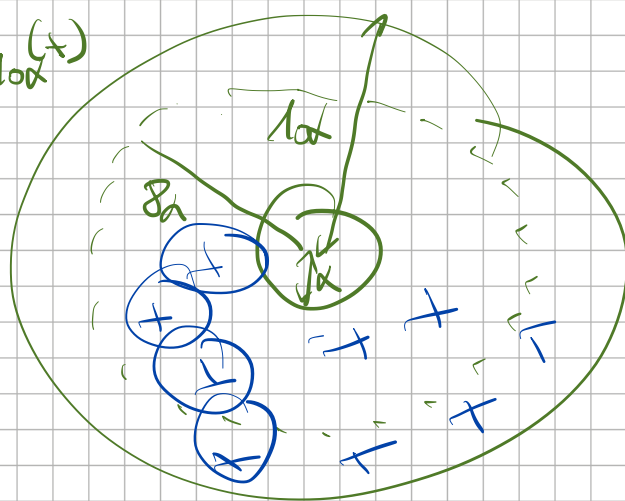
$$H(\alpha) := \{ \exists x_i \in B_{10\alpha}^c :$$

$$B_{g_\alpha}(x_i) \cap B_{g_\alpha} \neq \emptyset \}$$

(help from outside)

$$F(\alpha) := \{ \exists x_i \in B_{10\alpha} : g_i \geq \alpha \}$$

(large balls inside)



$M := \sup_{x \in C_0} \|x\|$  radius of  $C_0$  then

$$\{ M \geq g_\alpha \} \subset G(0, \alpha) \cup H(\alpha)$$

where

$$\begin{aligned}
 \mathbb{P}(H(\alpha)) &\leq \mathbb{E} \left[ \sum_{x_i \in B_{10\alpha}^c} \mu \{ B_g(x_i) \cap B_{g\alpha} \neq \emptyset \} \right] \\
 &= 2 \int dx \int \mathbb{P}(dr) \mu \{ x \in B_{g\alpha+r} \setminus B_{10\alpha} \} \\
 &\leq 2 \int_{\alpha}^{\infty} \mathbb{P}(dr) |B_{10r}| = 2 C' \int_{\alpha}^{\infty} r^d \mathbb{P}(dr) \\
 &\quad \xrightarrow{\alpha \uparrow \infty} 0
 \end{aligned}$$

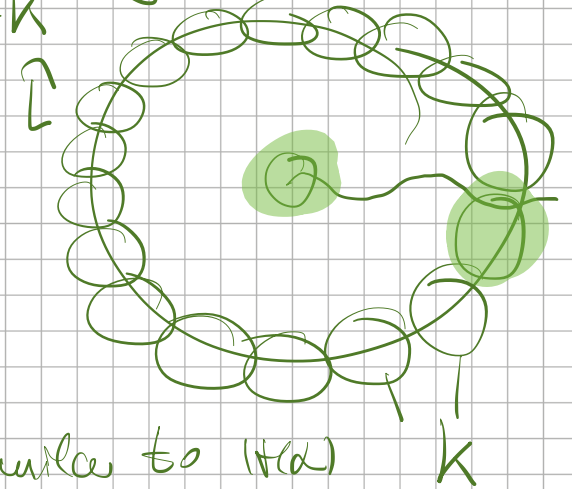
also

$$G(0, 10\alpha) \setminus F(\alpha)$$

$$\subset G(0, \alpha) \cap \bigcup_{k \in K} G(\alpha k, \alpha)$$

$$\begin{aligned}
 \Rightarrow \mathbb{P}(G(0, 10\alpha)) \\
 &\leq C'' \mathbb{P}(G(0, \alpha))^2 \\
 &\quad + \underbrace{\mathbb{P}(F(\alpha))}_{\rightarrow 0}
 \end{aligned}$$

$\xrightarrow{\alpha \uparrow \infty} 0$  similar to  $H(\alpha)$   $K$



Hence we have an inequality of the form

$$\begin{aligned}
 f(\alpha) &\leq c f\left(\frac{\alpha}{10}\right)^2 + g(\alpha) \quad (*) \\
 &\quad \uparrow \quad \quad \quad \rightarrow 0 \\
 &\quad \alpha \rightarrow 0 \text{ very fast}
 \end{aligned}$$

Input: Given for  $f \leq \frac{1}{2}$  on  $[1, 10]$   $g \leq \frac{1}{4}$  on  $[1, \infty)$  and  $(*)$ .

If  $g(\alpha) \xrightarrow{\alpha \uparrow \infty} 0$  then  $f(\alpha) \xrightarrow{\alpha \uparrow \infty} 0$ .

Sol  $g(\alpha) = 2C \int_{\alpha}^{\infty} r^d P(dr)$  small for  $\lambda$  small  
 $f(\alpha) \leq IP(\exists x_i \in B_{10\alpha})$  small for  $\lambda$  small  
in  $[1, 10]$

□

# Continuum Percolation in Random Environments



Benedikt Jahnel

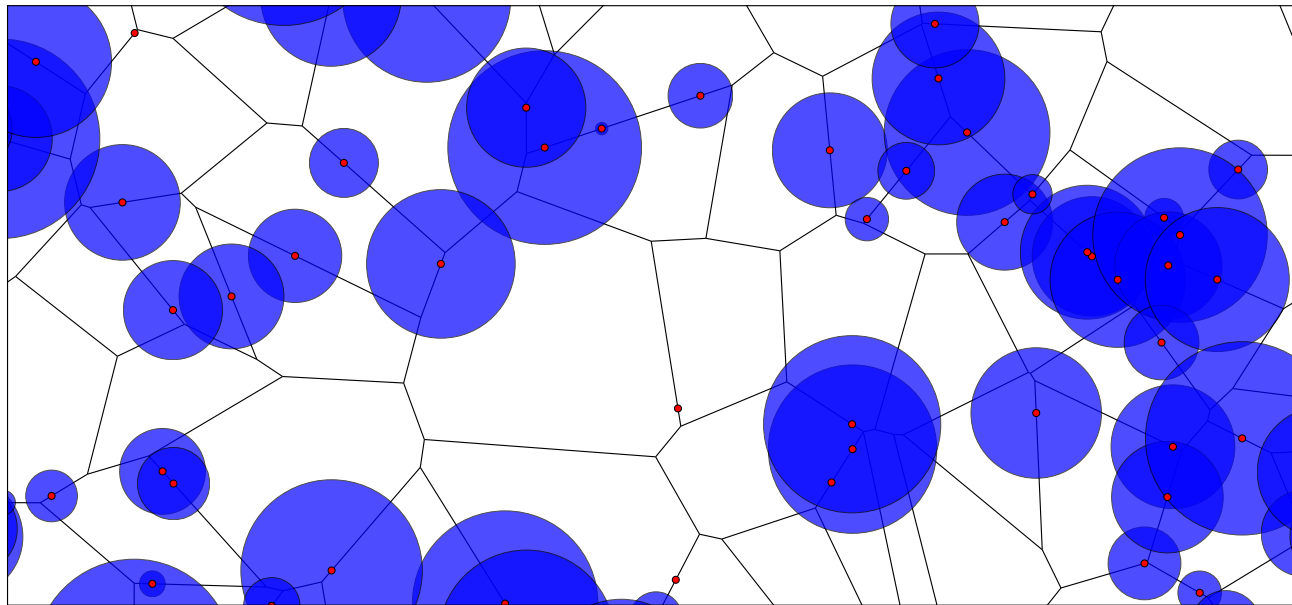
Joint with:

Christian Hirsch (RUG)  
András Tóbiás (TUB)  
Élie Calie (Orange)  
Alexander Hinsén (WIAS)  
Anh Duc Vu (WIAS)  
Sanjoy Jhawar (WAIS)

Stoch. Proc. Appl. (2019)  
Braz. J. Probab. Stat. (2022)  
Journ. Appl. Probab. (2023)  
IEEE Info. Theory (2023)

# Setting: Boolean models for Cox point processes

- $X = (X_i)_{i \in I}$  = stationary **Cox point process** in  $\mathbb{R}^d$ , i.e., Poisson point process (PPP) with stationary **random intensity measure**  $\Lambda$  where  $\mathbb{E}[\Lambda([0, 1]^d)] < \infty$
- $\varrho_i$  = **i.i.d. interaction radius** of  $X_i$  with  $\mathbb{P}(\varrho_i > 0) > 0$
- $\mathcal{C} = \bigcup_{i \in I} B_{\varrho_i}(X_i)$  = **Cox–Boolean model**, where  $B_r(x)$  = ball with radius  $r \geq 0$  centered at  $x \in \mathbb{R}^d$



**Figure:** Realization of a Cox–Boolean model (blue) based on a Cox point process (red) with directing measure given by a realization of a Poisson–Voronoi tessellation (black).

# Examples: Singular random environments

## ■ Singular random intensity measure, e.g.,

- $\Lambda(dx) = \nu_1(S \cap dx)$ , where  $\nu_1$  is one-dimensional Hausdorff measure and  $S \subset \mathbb{R}^d$  = stationary random segment process such as
  - Poisson–Voronoi tessellation (PVT),
  - Poisson–Delaunay tessellation (PDT),
  - Poisson line tessellation (PLT),
  - Manhattan grid (MG), etc.

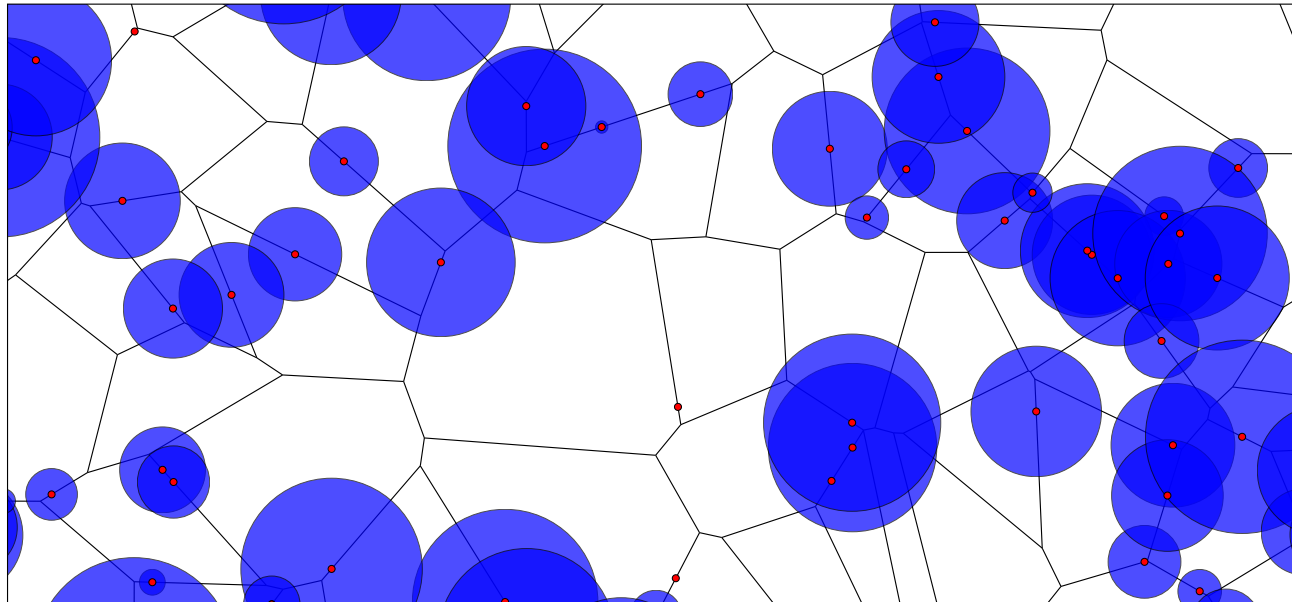
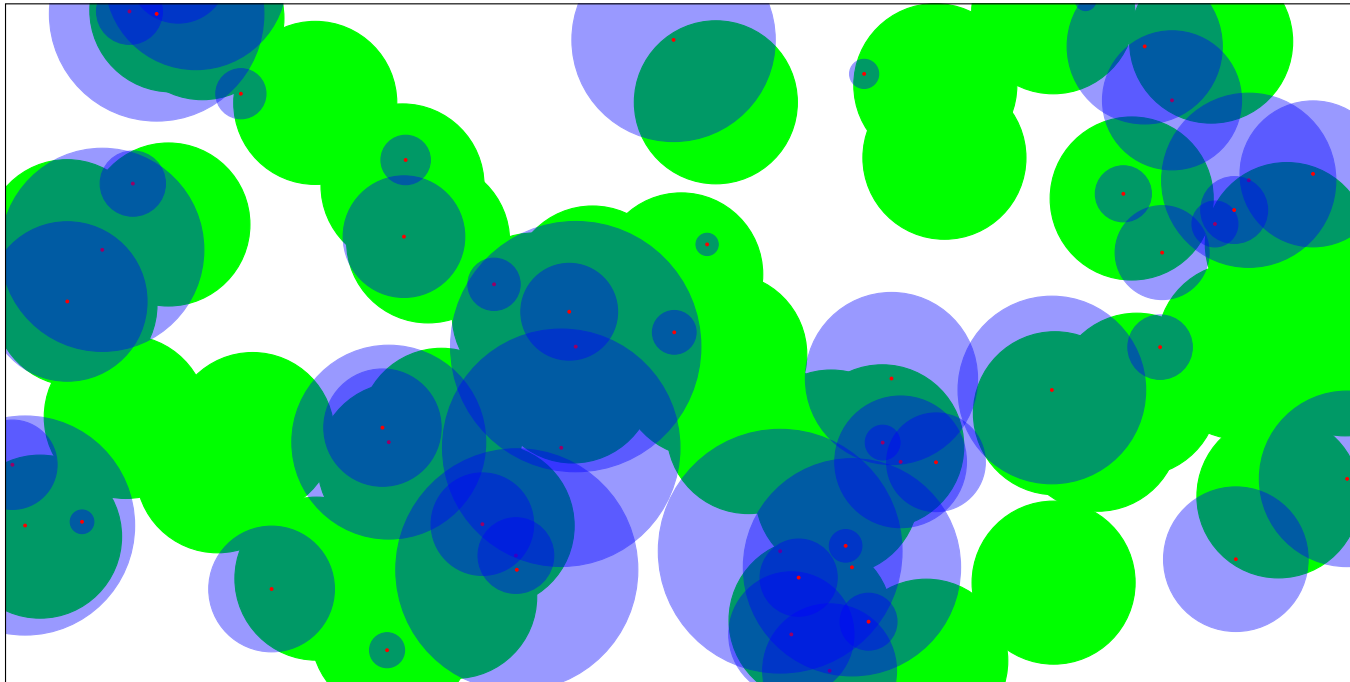


Figure: Realization of a Cox–Boolean model (blue) based on a Cox point process (red) with directing measure given by a realization of a Poisson–Voronoi tessellation (black).



# Examples: Absolutely continuous random environments

- $\Lambda(dx) = \ell_x dx =$  **absolutely continuous random intensity measure**, e.g.,
  - $\ell_x = \lambda \mathbb{1}\{x \in \Xi\} + \lambda' \mathbb{1}\{x \notin \Xi\}$ , where  $\Xi \subset \mathbb{R}^d$  **stationary random closed set** and  $\lambda, \lambda' \geq 0$
  - $\ell_x = \sum_{i \in I} \kappa(Y_i - x) =$  **shot-noise field**, where  $(Y_i)_{i \in I}$  stationary PPP and  $\kappa: \mathbb{R}^d \rightarrow [0, \infty)$  integrable
  - $\ell_x = \lambda \sum_{i \in I} \mathbb{1}\{|Y_i - x| < \rho_i\}$  where  $(Y_i)_{i \in I}$  stationary PPP with **i.i.d. marks**  $\rho_i$  and  $\lambda > 0$



**Figure:** Realization of a Cox–Boolean model (blue) based on a Cox point process (red) with directing measure given by an independent Poisson–Boolean model with fixed radii (green).

# Complete coverage and one-dimensional triviality

## Proposition (No complete coverage)

*For all  $d \geq 1$ ,  $\mathbb{E}[\varrho^d] < \infty$  implies  $\mathbb{P}(\mathcal{C} = \mathbb{R}^d) < 1$ . If additionally  $\Lambda$  is ergodic, then  $\mathbb{P}(\mathcal{C} = \mathbb{R}^d) = 0$ .*

## Proposition (One-dimensional triviality)

*Let  $d = 1$  and  $\Lambda$  ergodic, then  $\mathbb{E}[\varrho] < \infty$  implies  $\mathbb{P}(\mathcal{C} \text{ contains unbounded connected component}) = 0$ .*

- **Stationary point processes:**  $\mathbb{E}[\varrho^d] = \infty \Rightarrow \mathbb{P}(\mathcal{C} = \mathbb{R}^d) = 1$

Meester & Roy '96: *Continuum Percolation*, Cambridge University Press

- **Stationary PPP,  $d = 1$ :**  $\mathbb{E}[\varrho] < \infty \Rightarrow \mathbb{P}(\mathcal{C} \text{ contains unbounded component}) = 0$

Meester & Roy '96: *Continuum Percolation*, Cambridge University Press

- **Rectangular lilypond models:** Hirsch '16: *On the absence of percolation in a line-segment based lilypond model*, Ann. Henri Poincaré 52

- **2-degree random graphs:** Tóbiás & J '22: *Absence of percolation in graphs based on stationary point processes with degrees bounded by two*, Random Structures and Algorithms

# Essentially connected environments

## Definition (Essential $r$ -connectedness)

Let  $r > 0$  and  $Q_\alpha = [-\alpha, \alpha]^d$ . The stationary random measure  $\Lambda$  is **essentially  $r$ -connected**, if there exists a random field of **connectivity radii**  $R = \{R_x\}_{x \in \mathbb{R}^d}$ , defined on the same probability space as  $\Lambda$ , such that

1.  $(\Lambda, R)$  are jointly stationary,
2.  $\lim_{\alpha \uparrow \infty} \mathbb{P}(\sup_{y \in Q_\alpha \cap \mathbb{Q}^d} R_y \geq \alpha) = 0$ , and
3. for all  $\alpha \geq 1$ , whenever  $\sup_{y \in Q_{2\alpha} \cap \mathbb{Q}^d} R_y < \alpha/2$ , we have that for all  $x, y \in \text{supp}(\Lambda_{Q_\alpha})$  there exists a **finite sequence of points**  $(x_1, \dots, x_l) \subset \text{supp}(\Lambda_{Q_{2\alpha}})$  such that  $|x_i - x_{i+1}| < r$  for all  $i \in \{0, 1, \dots, l+1\}$  where  $x = x_0$  and  $y = x_{l+1}$ .

- $\ell_x = \lambda_1 \mathbb{1}\{x \in \Xi\} + \lambda_2 \mathbb{1}\{x \notin \Xi\}$  with  $\lambda_1, \lambda_2 > 0$  and  $\Lambda(dx) = \nu_1(S \cap dx)$  for  $S$  a PVT or PDT or MG, are **essentially  $r$ -connected** for any  $r > 0$
- Shot-noise fields and  $\ell_x = \lambda \sum_{i \in I} \mathbb{1}\{|Y_i - x| < \rho_i\}$  are in general **not essentially  $r$ -connected**

# Uniqueness of infinite cluster

## Theorem (Uniqueness)

Let  $d \geq 1$  and  $\Lambda$  ergodic. If  $r = \text{esssup}(\varrho) < \infty$  and  $\Lambda$  is *essentially  $r$ -connected*, then  $\mathbb{P}(\mathcal{C} \text{ contains at most one unbounded connected component}) = 1$ .

- Proof: adaptation of **Burton–Keane argument** and **FKG inequality**
- Stationary point processes:  $\text{esssup}(\varrho) = \infty$  or  $\Lambda \equiv 1 \Rightarrow$  uniqueness  
Meester & Roy '96: *Continuum Percolation*, Cambridge University Press
- Nearest neighbor graphs: Häggström & Meester '96: *Nearest neighbor and hard sphere models in continuum percolation*, Random Struct. Algorithms 167
- Insertion-tolerant point processes: Holroyd & Soo '13: *Insertion and deletion tolerance of point processes*, Electron. J. Probab. 18
- Levelset percolation: Broman & Meester '17: *Phase transition and uniqueness of levelset percolation*, J. Stat. Phys. 167

- $X = (X_i)_{i \in I} =$  **Cox point process** in  $\mathbb{R}^d$  with stationary **random intensity measure**  $\lambda\Lambda$ , where  $\mathbb{E}[\Lambda([0, 1]^d)] = 1$  and  $\lambda \geq 0$
- $\mathcal{C} \supset C_o =$  **connected component containing the origin**
- For  $A \subset \mathbb{R}^d$ ,  $|A| =$  Lebesgue volume;  $\text{diam}(A) =$  diameter;  $X(A) =$  number of Cox points
- **Critical values:**

$$\lambda_{vp} = \inf\{\lambda > 0: \mathbb{P}(|C_o| = \infty) > 0\}$$

$$\lambda_{dp} = \inf\{\lambda > 0: \mathbb{P}(\text{diam}(C_o) = \infty) > 0\}$$

$$\lambda_{np} = \inf\{\lambda > 0: \mathbb{P}(X(C_o) = \infty) > 0\}$$

$$\lambda_v(s) = \inf\{\lambda > 0: \mathbb{E}[|C_o|^s] = \infty\}, \text{ for } s > 0$$

$$\lambda_d(s) = \inf\{\lambda > 0: \mathbb{E}[\text{diam}(C_o)^s] = \infty\}, \text{ for } s > 0$$

$$\lambda_n(s) = \inf\{\lambda > 0: \mathbb{E}[X(C_o)^s] = \infty\}, \text{ for } s > 0$$

# Pathological percolation behavior for Cox point processes

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- **Insufficiently connected environments:**

for all  $s > 0$ ,  $\lambda_{vp} = \lambda_{dp} = \lambda_{np} = \lambda_v(s) = \lambda_d(s) = \lambda_n(s) = \infty$

- **Unbounded environments:**  $Z dx$  with  $\text{esssup}(Z) = \infty$

for all  $s > 0$ ,  $\lambda_{vp} = \lambda_{dp} = \lambda_{np} = \lambda_v(s) = \lambda_d(s) = \lambda_n(s) = 0$   
(non-ergodic, mixed PPP)

- **Attractive point processes:** Blaszczyzyn & Yogeshwaran '13: *Clustering and percolation of point processes*, Electron. J. Probab.
- **SINR percolation I:** Dousse, Baccelli & Thiran '05: *Impact of interferences on connectivity in ad hoc networks*, IEEE Trans. Netw.
- **SINR percolation II:** Dousse, Franceschetti, Macris, Meester & Thiran '06: *Percolation in the signal to interference ratio graph*, J. Appl. Probab.
- **Cox-SINR percolation I:** Tóbiás '20: *Signal to interference ratio percolation for Cox point processes*, Lat. Am. J. Probab. Math. Stat.
- **Cox-SINR percolation II:** Tóbiás & J '22: *SINR percolation for Cox point processes with random powers*, Adv. Appl. Probab.



# Stabilizing environments

## Definition (Stabilization)

The stationary random measure  $\Lambda$  is  **$\varphi$ -stabilizing** if there exists a random field of **stabilization radii**  $R = \{R_x\}_{x \in \mathbb{R}^d}$ , defined on the same probability space as  $\Lambda$ , with

1.  $(\Lambda, R)$  are jointly stationary,
2.  $\lim_{\alpha \uparrow \infty} \varphi(\alpha) = 0$ , where  $\varphi(\alpha) = \mathbb{P}(\sup_{y \in Q_\alpha \cap \mathbb{Q}^d} R_y \geq \alpha)$ , and
3. for all  $\alpha \geq 1$ , the random variables

$$\left( f(\Lambda_{Q_\alpha(x)}) \mathbb{1}_{\left\{ \sup_{y \in Q_\alpha(x) \cap \mathbb{Q}^d} R_y < \alpha \right\}} \right)_{x \in \psi}$$

are **independent** for all non-negative bounded measurable functions  $f$  and finite  $\psi \subset \mathbb{R}^d$ , as long as  $\text{dist}(x, \psi \setminus x) > 3\alpha$  for all  $x \in \psi$ .

- $\varphi$ -stabilization implies ergodicity
- $\Lambda$   **$b$ -dependent** if  $\varphi(\alpha) = 0$  for all  $\alpha > b$
- $\Lambda$  **exponentially stabilizing** if  $\varphi(\alpha) \leq \exp(-c\alpha)$

# Examples: stabilizing environments

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- **Shot-noise fields** are  $b$ -dependent for compactly supported kernels; kernels with unbounded support lead to **absence of  $\varphi$ -stabilization**
- $\Lambda(\mathrm{d}x) = \nu_1(S \cap \mathrm{d}x)$  for  $S$  a PVT or PDT, are **exponentially stabilizing**,  
 $R_x = \min\{|x - X_i| : X_i \in \mathcal{X}\}$   
Hirsch, Cali & J '19: *Continuum percolation for Cox point processes*, Stoch. Proc. Appl.
- $\Lambda(\mathrm{d}x) = \nu_1(S \cap \mathrm{d}x)$  for  $S$  a PLT or MG are **not  $\varphi$ -stabilizing**
- $\ell_x = \lambda \sum_{i \in I} \mathbb{1}\{|Y_i - x| < \rho_i\}$  with  $b = \text{esssup}(\rho) < \infty$  are  **$b$ -dependent**
- $\ell_x = \lambda \sum_{i \in I} \mathbb{1}\{|Y_i - x| < \rho_i\}$  with  $\text{esssup}(\rho) = \infty$  and  $\mathbb{E}[\rho^{d+s}] < \infty$  for  $s > 0$ , are  **$\varphi$ -stabilizing** with  $\int_0^\infty \alpha^{s-1} \varphi(\alpha) \mathrm{d}\alpha < \infty$

# Phase transitions for Cox–Boolean models

## Theorem (Phase transitions)

Let  $d \geq 2$ ,  $s > 0$  and  $\Lambda$  be stationary.

1.  $\Lambda$   *$\varphi$ -stabilizing* with sufficiently large  $\text{esssup}(\varrho)$  implies  $\lambda_{\text{vp}}, \lambda_{\text{dp}}, \lambda_{\text{np}} < \infty$ .
2.  $\mathbb{E}[\varrho^d] < \infty$  and  $\Lambda$   *$\varphi$ -stabilizing* implies  $\lambda_{\text{vp}}, \lambda_{\text{dp}}, \lambda_{\text{np}} > 0$ .
3.  $\mathbb{E}[\varrho^{d+s}] < \infty$  and  $\Lambda$   *$\varphi$ -stabilizing* with  $\int_0^\infty \alpha^{s-1} \varphi(\alpha) d\alpha < \infty$  implies  $\lambda_v(s/d), \lambda_d(s) > 0$ .  $\lambda_n(s/d) > 0$  if additionally

$$\int_0^\infty \alpha^{s-1} \mathbb{P}(\Lambda(B_\alpha) \geq c\alpha^d) d\alpha < \infty \quad \text{for some } c > 0. \quad (1)$$

4.  $\mathbb{E}[\varrho^{d+s}] = \infty$  and  $\Lambda$  *ergodic* implies  $\lambda_v(s/d) = \lambda_d(s) = \lambda_n(s/d) = 0$ .

- Proof via *multi-scale and branching-process arguments* (see below)
- Overshoot-Condition (1) satisfied by canonical examples (see below)

- $0 < \lambda_{vp} < \infty$  for fixed radii

Hirsch, Cali & J '19: *Continuum percolation for Cox point processes*, Stoch. Proc. Appl.

Tóbiás '20: *Signal to interference ratio percolation for Cox point processes*, Lat. Am. J. Probab. Math. Stat.

- $\lambda_{vp} > 0, \lambda_{dp} > 0, \lambda_v(s/d) > 0$  and  $\lambda_d(s) > 0$  alternatively via

Gouéré, '09: *Subcritical regimes in some models of continuum percolation*, Ann. Appl. Probab.

- Version for **Poisson point processes**

Meester & Roy '96: *Continuum Percolation*, Cambridge University Press

Gouéré '08: *Subcritical regimes in the Poisson Boolean model of continuum percolation*, Ann. Probab.

Gouéré & Thérét '18: *Equivalence of some subcritical properties in continuum percolation*, Bernoulli

- Percolation for **Gibbs point processes** with fixed radii

Mürrmann '75: *Equilibrium distributions of physical clusters*, Comm. Math. Phys.

Stucki '13: *Continuum percolation for Gibbs point processes*, Electron. Commun. Probab.

Jansen '16: *Continuum percolation for Gibbsian point processes with attractive interactions*, Electron. J. Probab.

Magazinov '18: *On percolation of two-dimensional hard disks*, Comm. Math. Phys.

# Related work

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- Percolation for repelling point processes in  $\mathbb{R}^2$  include Ginibre ensembles and Gaussian zero processes with fixed radii

Ghosh, Krishnapur & Peres '16: *Continuum percolation for Gaussian zeroes and Ginibre eigenvalues*, Ann. Probab.

- Percolation for negatively associated point processes, including determinantal point processes and some perturbed lattices

Georgii & Yoo '05: *Conditional intensity and Gibbsianness of determinantal point processes*, J. Stat. Phys.

Błaszczyszyn & Yogeshwaran '13: *Clustering and percolation of point processes*, Electron. J. Probab.

# Criteria for Condition (1)

## Lemma

*Condition (1) holds if any of the following two conditions holds,*

$$\begin{aligned} \limsup_{\alpha \uparrow \infty} \alpha^{-d} \log \mathbb{E}[\exp(\beta \Lambda(B_\alpha))] &< \infty && \text{for some } \beta > 0, \\ \limsup_{\alpha \uparrow \infty} \alpha^{s-d\beta+\varepsilon} \mathbb{E}[|\Lambda(B_\alpha) - |B_\alpha||^\beta] &< \infty && \text{for some } \beta \geq 1 \text{ and } \varepsilon > 0. \end{aligned}$$

## Applicable:

- Shot-noise field with compactly supported kernel  $\kappa$
- $\Lambda(dx) = \ell_x dx$  with  $\ell_x = \lambda \sum_{i \in I} \mathbb{1}\{|Y_i - x| < \rho_i\}$  and  $\mathbb{E}[\rho^{d+2s+\varepsilon}] < \infty$
- $\Lambda(dx) = \nu_1(S \cap dx)$  with  $S$  edge set of PDT in  $\mathbb{R}^2$



# Ideas of proof

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- **Part 1:** Existence of supercritical percolation regimes for sufficiently-large radii via coupling with percolating Cox–Boolean model with large fixed radii
- **Part 2 & 3:** Existence of subcritical percolation regimes via multi-scale argument; main idea, let  $f(\alpha) = \mathbb{P}(o \rightsquigarrow B_\alpha^c(o))$  and  $g(\alpha) = \lambda \mathbb{E}[\varrho^d \mathbb{1}\{\varrho \geq \alpha\}] + 2\varphi(10\alpha)$ , then there exists  $c = c(d) < \infty$  such that

$$f(10\alpha) \leq c(f(\alpha)^2 + g(\alpha))$$

[Gou   , '08]:  $\lim_{\alpha \uparrow \infty} g(\alpha) = 0$  implies  $\lim_{\alpha \uparrow \infty} f(\alpha) = 0$  and  $\int_1^\infty \alpha^s g(\alpha) d\alpha < \infty$  implies  $\int_1^\infty \alpha^s f(\alpha) d\alpha < \infty$

- **Part 4:** Absence of subcritical percolation regimes via branching-process arguments

# Percolation probabilities and Palm calculus

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- Fix  $r > 0$  and consider **Cox–Gilbert graph**  $g_r(X)$
- Assume normalization  $\mathbb{E}[\Lambda(Q_1)] = 1$  with  $Q_r = [-r/2, r/2]^d$
- Define **Palm version**  $X^*$  of  $X$  via

$$\mathbb{E}[f(X^*)] = \frac{1}{\lambda} \mathbb{E} \left[ \sum_{X_i \in Q_1} f(X - X_i) \right]$$

- Define **percolation probability** by

$$\theta(\lambda, r) = \mathbb{P}(o \longleftrightarrow \infty \text{ in } g_r(X^*))$$

- Define **Palm version**  $\Lambda^*$  of  $\Lambda$  via

$$\mathbb{E}[f(\Lambda^*)] = \mathbb{E} \left[ \int_{Q_1} \Lambda(dx) f(\Lambda - x) \right]$$

# Large-radius limit of the percolation probability

- For **Poisson–Gilbert graph**:  $\lim_{r \uparrow \infty} r^{-d} \log(1 - \theta(\lambda, r)) = -|B_1(o)|\lambda$   
Penrose '91: *On a continuum percolation model*, Adv. Appl. Probab.
- Large-deviations rate function given by **isolation probability**
- Remains true for  $b$ -dependent Cox processes with  $\Lambda$ -dependent rate:

## Theorem

*If  $\lambda > 0$ , then*

$$\liminf_{r \uparrow \infty} r^{-d} \log(1 - \theta(\lambda, r)) \geq \liminf_{r \uparrow \infty} r^{-d} \log \mathbb{E}[\exp(-\lambda \Lambda^*(B_r(o)))].$$

*If, additionally,  $\Lambda$  is  $b$ -dependent and  $\Lambda([0, 1]^d)$  has all exponential moments, then the limit*

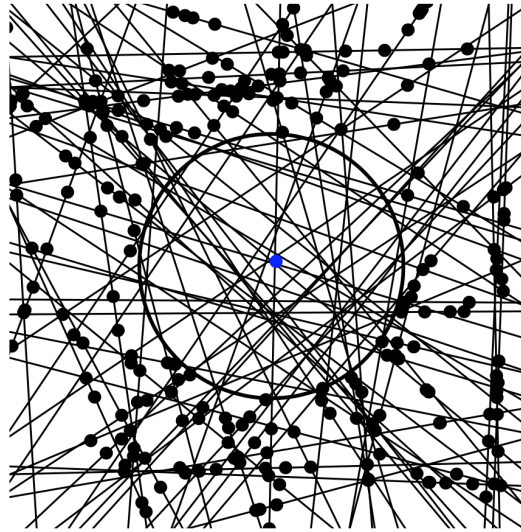
$$I^* = - \lim_{r \uparrow \infty} r^{-d} \log \mathbb{E}[\exp(-\lambda \Lambda^*(Q_r))]$$

*exists and*

$$\lim_{r \uparrow \infty} r^{-d} \log(1 - \theta(\lambda, r)) = -|B_1(o)|I^*.$$

# Large-radius limit of the percolation probability

---



- Global lower bound given by isolation probability
- Upper bound applicable for example for environment based on  $\Xi$  or shot-noise fields
- Limiting Laplace transform computable for example for [shot-noise fields](#),

$$\lim_{r \uparrow \infty} r^{-d} \log \mathbb{E}[\exp(-\lambda \Lambda(Q_r))] = \lambda_s (e^{-\lambda K} - 1)$$

where  $K = \int dx k(x)$

# Large-intensity limit of the percolation probability

- For Poisson–Gilbert graph, due to **scale invariance**, large-intensity statements can be obtained from large-radius statements
- In general **not true** for Cox–Gilbert graphs
- Cox case: connectivity structure of the support of  $\Lambda^*$  becomes prominent
- Define  **$r$ -boundary** of set  $A \subset \mathbb{R}^d$  by  $\partial_r A = \{x \in \mathbb{R}^d : \text{dist}(x, A) < r\} \setminus A$
- Define  $\mathcal{R}_r$  = set of all compact sets that contain the origin and are  **$r$ -connected**

## Theorem

*Let  $r > 0$ . Then,*

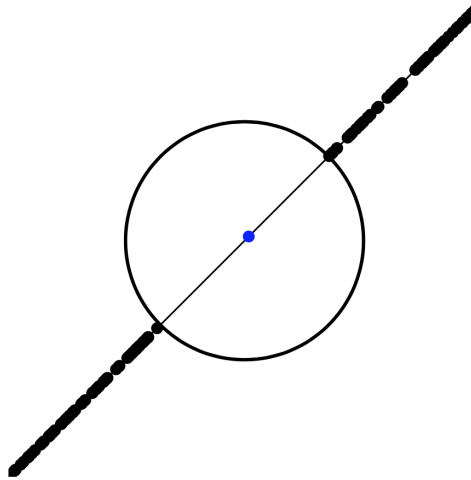
$$\liminf_{\lambda \uparrow \infty} \lambda^{-1} \log(1 - \theta(\lambda, r)) \geq - \inf_{A \in \mathcal{R}_r} \text{essinf}(\Lambda^*(\partial_r A)).$$

*If, additionally,  $\Lambda$  is  $b$ -dependent and  $\text{essinf}(\Lambda(Q_\delta)) > 0$  for every  $\delta > 0$ , then*

$$\limsup_{\lambda \uparrow \infty} \lambda^{-1} \log(1 - \theta(\lambda, r)) \leq - \lim_{\varepsilon \downarrow 0} \inf_{A \in \mathcal{R}_{r+\varepsilon}} \text{essinf}(\Lambda^*(\partial_{r-\varepsilon} A)).$$

# Large-intensity limit of the percolation probability

---



- Global lower bound **not necessarily** given by isolation probability
- For example for environment based on  $\Xi$  with  $\lambda_1, \lambda_2 > 0$ , r.h.s. optimal for  $A = \{o\}$
- For PVT r.h.s. of the lower bound can be computed to be  $-2r$
- Condition  $\text{essinf}(\Lambda(Q_\delta)) > 0$  **not satisfied** for shot-noise field and singular examples

# Sharp thresholds for finite-range Cox–Boolean models

- Environments with uniformly **bounded dependencies** and **bounded intensity**, e.g., PVT superposed with sparse grids and thickened edges or capped intensities
- **Finite-volume percolation probability**

$$\theta_n(\lambda) = \mathbb{P}(o \longleftrightarrow [-n/2, n/2]^c \text{ in } g_1(X^\lambda))$$

## Theorem

1.  $\limsup_{n \rightarrow \infty} n^{-1} \log \theta_n(\lambda) < 0$  holds for every  $\lambda < \lambda_c$ , and
2.  $\liminf_{\lambda \downarrow \lambda_c} \theta(\lambda)/(\lambda - \lambda_c) > 0$ .

- Proof based on OSSS inequality and coarse-graining construction

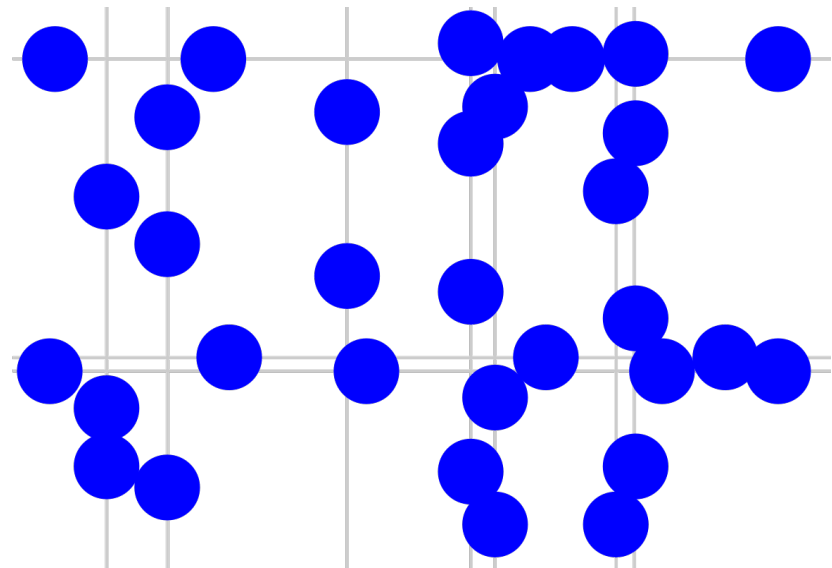
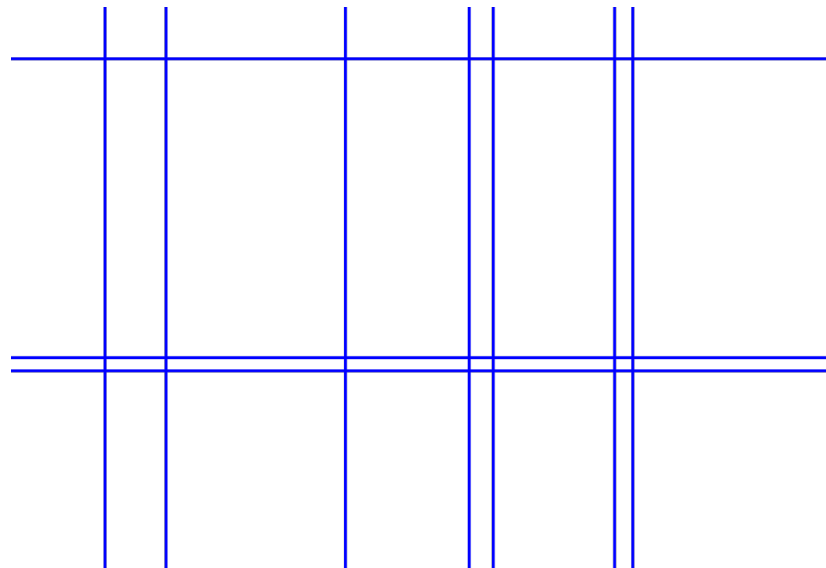
Hirsch, Muirhead & J '22: *Sharp phase transition for Cox percolation*, Elec. Comm. Probab.

Duminil-Copin, Raoufi & Tassion'19: *Sharp phase transition for the random-cluster and Potts models via decision trees*, Ann. of Math.

# Beyond stabilization: Manhattan–Boolean models

- **Absence of stabilization** features long (infinite)-range dependencies
- **Manhattan grid** based on two independent homogeneous Poisson point processes  $\Phi^x, \Phi^y \subset \mathbb{R}$  with intensities  $\mu_x$  and  $\mu_y$

$$\Lambda(A) := \int_{\mathbb{R}} \int_{\mathbb{R}} \mathbb{1}\{(x, y) \in A\} dy \Phi^x(dx) + \int_{\mathbb{R}} \int_{\mathbb{R}} \mathbb{1}\{(x, y) \in A\} dx \Phi^y(dy)$$





# No sharp thresholds in Manhattan–Boolean models

- $g_r(X^\lambda)$  Gilbert graph based on Poisson point process with intensity  $\lambda\Lambda$
- Accumulation of streets is probabilistically cheap, leading to subexponential decay

Vu, Jhawar & J '22: *Continuum percolation in a nonstabilizing environment*, arXiv:2205.15366

## Proposition

*For all  $r > \sqrt{2}$  and  $\mu_x, \mu_y, \lambda > 0$  we have that*

$$\liminf_{n \rightarrow \infty} \lfloor \lambda^{-1} \log n \rfloor! n^{\lambda^{-1} \log(\mu_x \wedge \mu_y)} \theta_n(r, \mu_x, \mu_y, \lambda) > 0,$$

*in particular, for every  $\varepsilon > 0$*

$$\liminf_{n \rightarrow \infty} n^{(1+\varepsilon)\lambda^{-1} \log(\log n)} \theta_n(r, \mu_x, \mu_y, \lambda) = \infty.$$

# Supercritical regimes in Manhattan–Boolean models

## Theorem

*Let  $r > 0$  be arbitrary.*

- 1. For every  $\lambda > 0$ , there exists  $\mu_c(r, \lambda) > 0$ , such that for all  $\mu_x, \mu_y \geq \mu_c(r, \lambda)$ :  $\mathcal{C}(r, \mu_x, \mu_y, \lambda)$  percolates almost surely.*
- 2. For every  $\mu_x, \mu_y > 0$ , there exists  $\lambda_c(r, \mu_x, \mu_y) > 0$ , such that for all  $\lambda \geq \lambda_c(r, \mu_x, \mu_y)$ :  $\mathcal{C}(r, \mu_x, \mu_y, \lambda)$  percolates almost surely.*
- 3. For every  $\lambda, \mu_x > 0$ , there exists  $\mu_{y,c}(r, \mu_x, \lambda) > 0$ , such that for all  $\mu_y \geq \mu_{y,c}(r, \mu_x, \lambda)$ :  $\mathcal{C}(r, \mu_x, \mu_y, \lambda)$  percolates almost surely.*

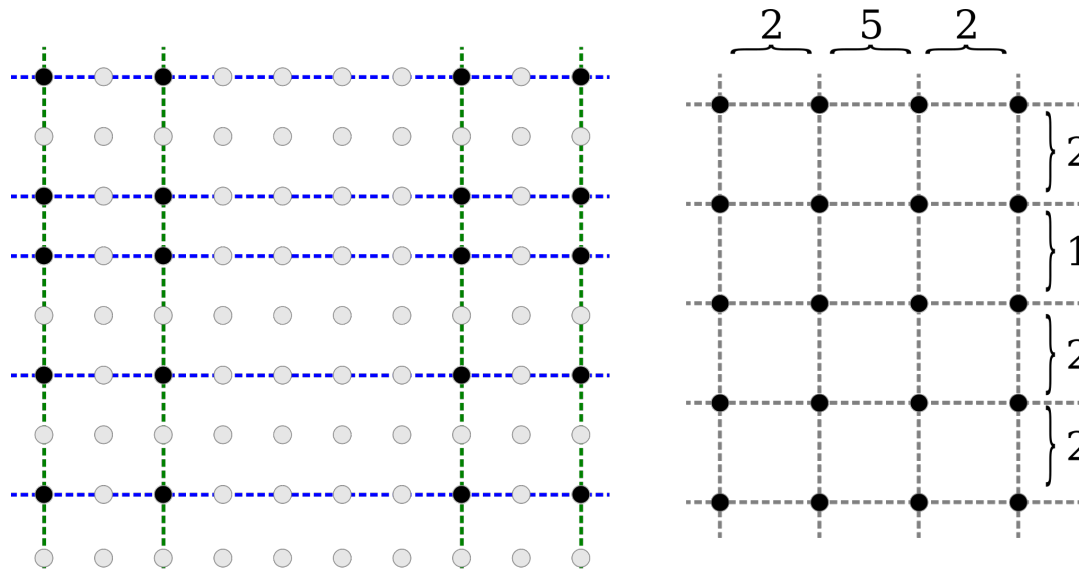
- Proof by comparison to supercritical bond percolation in **randomly stretched lattice**

Hoffman '05: *Phase transition in dependent percolation*, Comm. Math. Phys.

# Randomly stretched lattice (RSL)

- $N^{(x)} := (N_i^{(x)})_{i \in \mathbb{Z}}$  and  $N^{(y)} := (N_j^{(y)})_{j \in \mathbb{Z}}$  = families of mutually independent positive random variables and  $p \in (0, 1)$

$$\mathbb{P}((i, j) \leftrightarrow (i + 1, j) \text{ is open} \mid N^{(x)}, N^{(y)}) := p^{N_i^{(x)}}$$



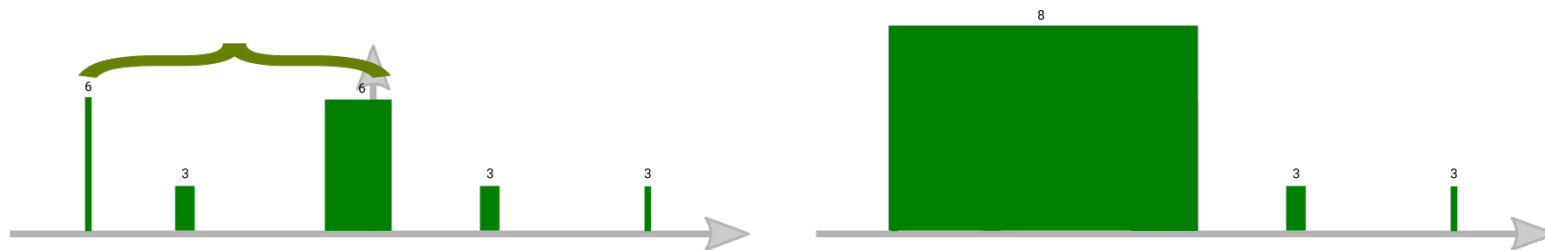
- Hoffman: If  $\mathbb{P}(N_i^{(x)} \geq l), P(N_j^{(y)} \geq l) \leq 2^{-1000l}$ , then there exists  $p_c \in (0, 1)$  such that RSL percolates almost surely for all  $p \geq p_c$ .

# Subcritical regimes in Manhattan–Boolean models

## Theorem

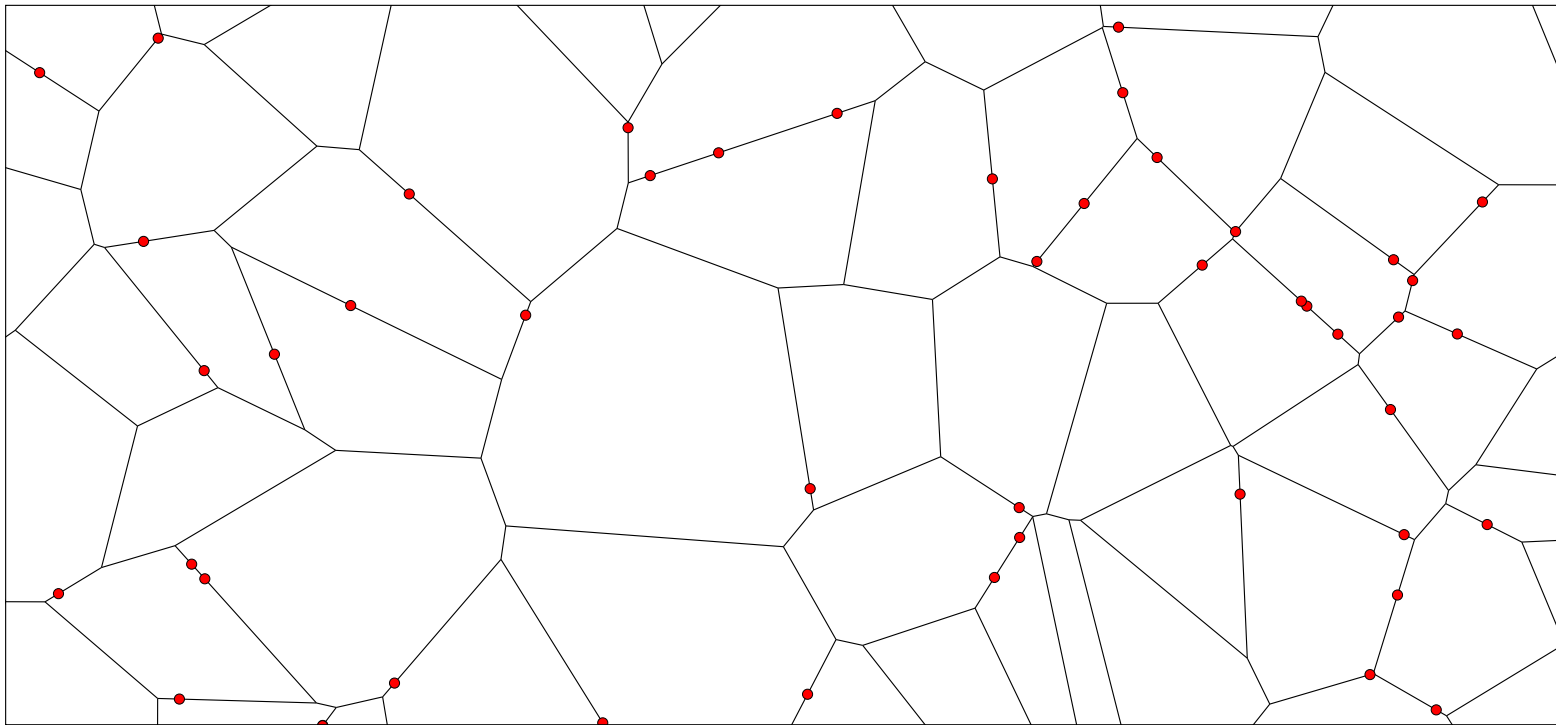
*For every  $r > 0$  and every  $\mu_x, \mu_y \geq 0$ , there exists a  $\lambda_c(r, \mu_x, \mu_y) > 0$  such that for all  $\lambda \leq \lambda_c(r, \mu_x, \mu_y)$ :  $\mathcal{C}(r, \mu_x, \mu_y, \lambda)$  does not percolate almost surely.*

- Proof significantly more tedious via Peierls' argument and existence of closed circuits in dual of subcritical RSL
- Discretization of streets into merging bands and labels mimicking and extending the strategy of Hoffman



# Setting: Device locations

- **Devices at time zero:** Stationary Cox point process  $X^\lambda = (X_i)_{i \in I}$  on random segment processes  $S$  with intensity measure  $\Lambda_S(dx) = \lambda |S \cap dx|$ , where  $\lambda > 0$  and  $\mathbb{E}[\Lambda_S([0, 1]^d)] = 1$



**Figure:** Realization of a Cox–Boolean model (blue) based on a Cox point process (red) with directing measure given by a realization of a Poisson–Voronoi tessellation (black).

# Stabilizing environments

## Definition (Stabilization)

A stationary random segment process  $S$  is called **stabilizing** if there exists a random field of **stabilization radii**  $R = \{R_x\}_{x \in \mathbb{R}^2}$ , defined on the same probability space as  $S$ , with

1.  $(S, R)$  are jointly stationary,
2.  $\lim_{n \uparrow \infty} \mathbb{P}(\sup_{y \in Q_n \cap \mathbb{Q}^2} R_y < n) = 1$  and
3. for all  $n \geq 1$ , the random variables

$$(f(S_{Q_n(x)}) \mathbb{1}_{\{\sup_{y \in Q_n(x) \cap \mathbb{Q}^2} R_y < n\}})_{x \in \psi}$$

are **independent** for all non-negative bounded measurable functions  $f$  and finite  $\psi \subset \mathbb{R}^2$ , as long as  $\text{dist}(x, \psi \setminus x) > 3n$  for all  $x \in \psi$ .

- stabilization implies ergodicity
- $S$   **$b$ -dependent** if  $\mathbb{P}(\sup_{y \in Q_n \cap \mathbb{Q}^2} R_y \geq n) = 0$  for all  $n > b$
- $S$  **exponentially stabilizing** if  $\mathbb{P}(\sup_{y \in Q_n \cap \mathbb{Q}^2} R_y \geq n) \leq C \exp(-cn)$

# Asymptotic essential connected environments

---

- $\Lambda_S$  for  $S$  a PVT or PDT are **exponentially stabilizing**,  $R_x = \min\{|x - X_i| : X_i\}$   
Hirsch, Cali & J '19: *Continuum percolation for Cox point processes*, Stochastic Processes and Applications
- $\Lambda_S$  for  $S$  a PLT or MG are **not stabilizing**

# Asymptotic essential connected environments

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- $\Lambda_S$  for  $S$  a PLT or MG are **not stabilizing**

## Definition (Asymptotic essential connectedness)

A stabilizing random segment process  $S$  with stabilization radii  $R$  is **asymptotically essentially connected** if, for all sufficiently large  $n \geq 1$ , whenever  $\sup_{y \in Q_{2n} \cap \mathbb{Q}^2} R_y < n/2$ , we have that

1.  $|S \cap Q_n| > 0$  and
2.  $S \cap Q_n$  is contained in one of the connected components of  $S \cap Q_{2n}$ .

- $\Lambda_S$  for  $S$  a PVT or PDT are **asymptotically essentially connected**

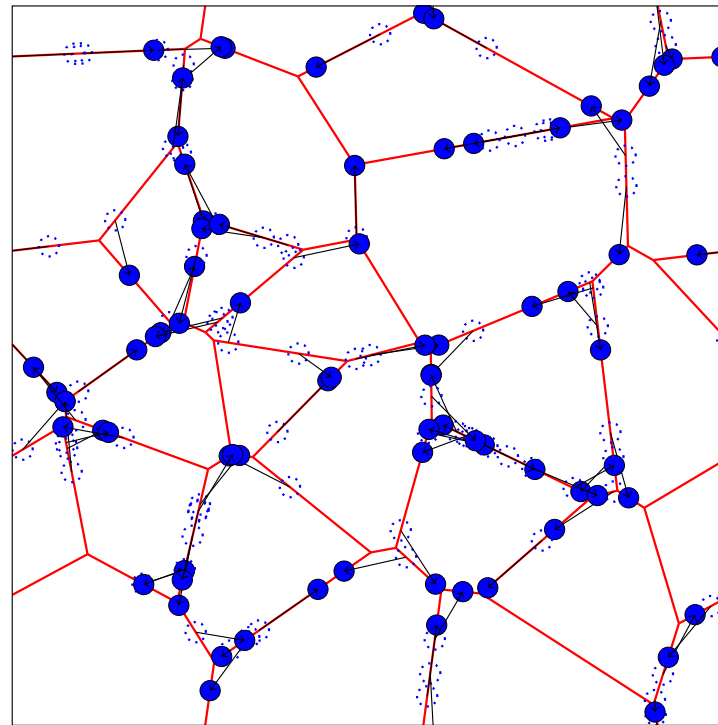


# Mobility in urban device-to-device networks

- **Waypoint kernel:** device  $X_i$  picks target location  $Y_i$  independently via stationary kernel

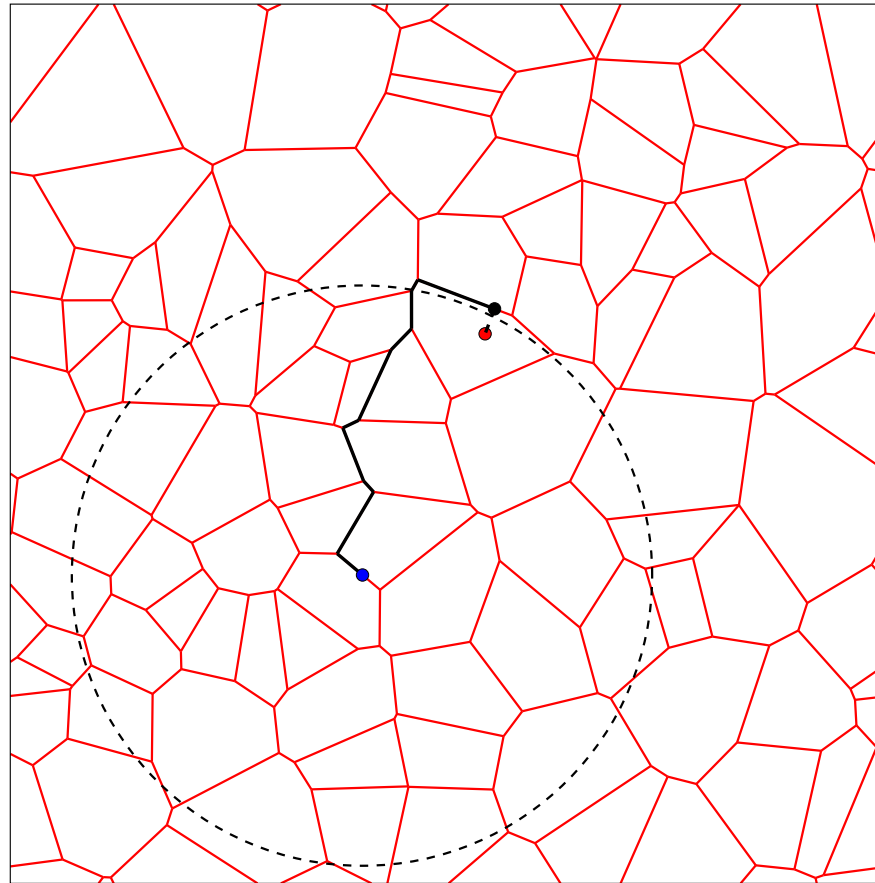
$$\kappa^S(X_i, dy)$$

- Example:  $\kappa^S(x, dy) = |S \cap B_R(x)|^{-1} |dy \cap S \cap B_R(x)|$



# Mobility

- **Movement:** device  $X_i$  moves to  $Y_i$  and back with iid **velocity**  $V_i$  (governed by  $\mu_v$ , with  $\text{supp}(\mu_v) = [v_{\min}, v_{\max}] \subset (0, \infty)$ ) to waypoint via **shortest path** on  $\mathcal{S}$



# Part 1: Connectivity in mobile device-to-device networks

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- **Contact times:** device  $X_i$  and  $X_j$  with trajectories  $(T_{i,t})_{t \geq 0}$  and  $(T_{j,t})_{t \geq 0}$  have contact at times

$$Z(X_i, X_j) = \{0 \leq t \leq T : |T_{i,t} - T_{j,t}| < r \text{ and } T_{i,t}, T_{j,t} \text{ are on same street}\}$$

- **Connection:** device  $X_i$  and  $X_j$  are connected if for some  $t \geq 0$ ,

$$[t, t + \rho] \subset Z(X_i, X_j)$$

where  $\rho = \rho_o + \rho_1$  with **initialization time**  $\rho_o \geq 0$  and **transmission time**  $\rho_1 > 0$

Analysis of **clustering behavior** of graph

$$g_{T, \mu_v, \rho, r}(X^\lambda)$$

with respect to system parameters.

# Illustration of connectivity graph $g_{T,\mu_v,\rho,r}(X^\lambda)$

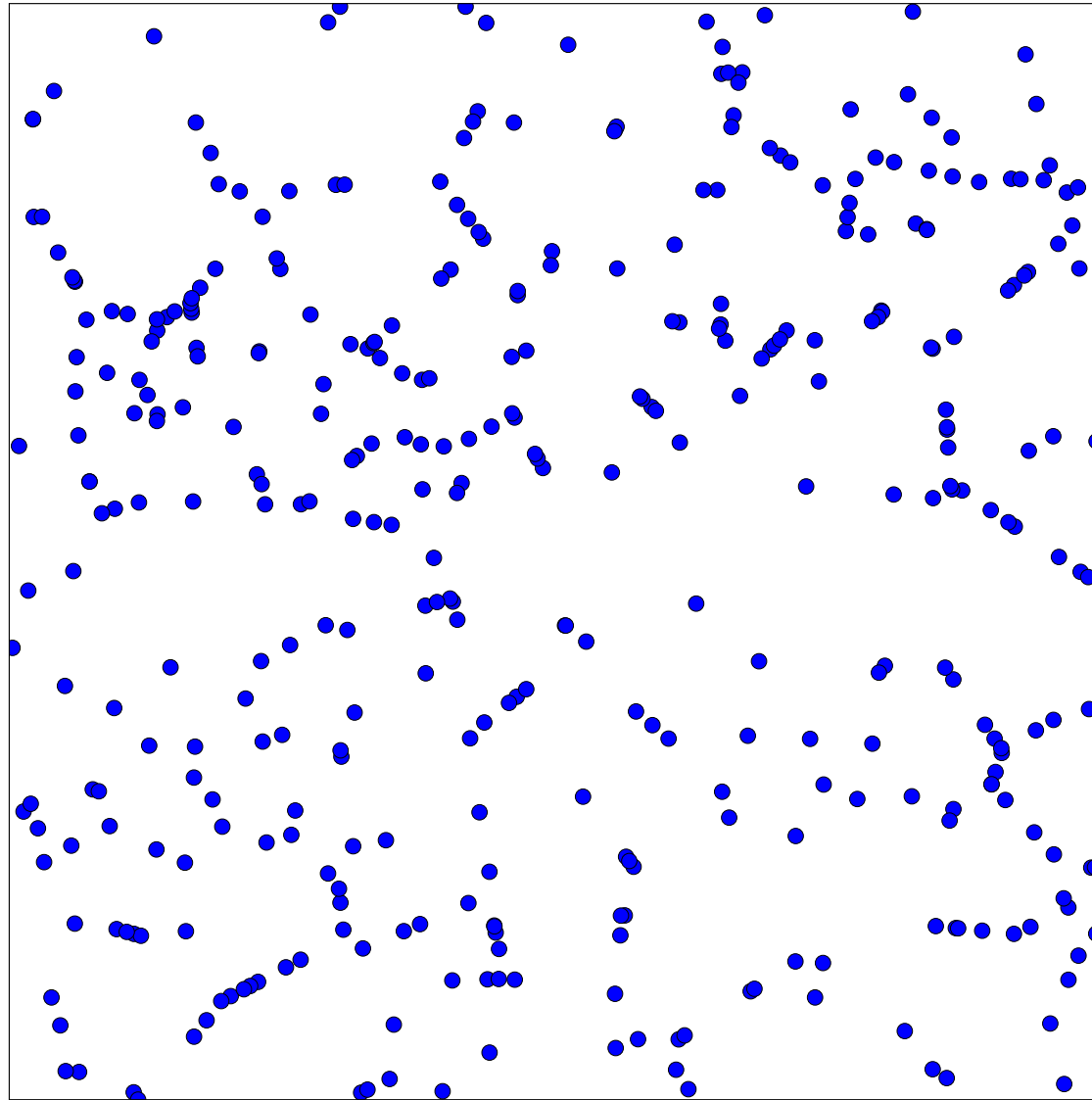


Figure: Realization of  $g_{T,\mu_v,\rho,r}(X^\lambda)$  with  $T = 0$ . The street system is suppressed.

# Illustration of connectivity graph $g_{T,\mu_v,\rho,r}(X^\lambda)$

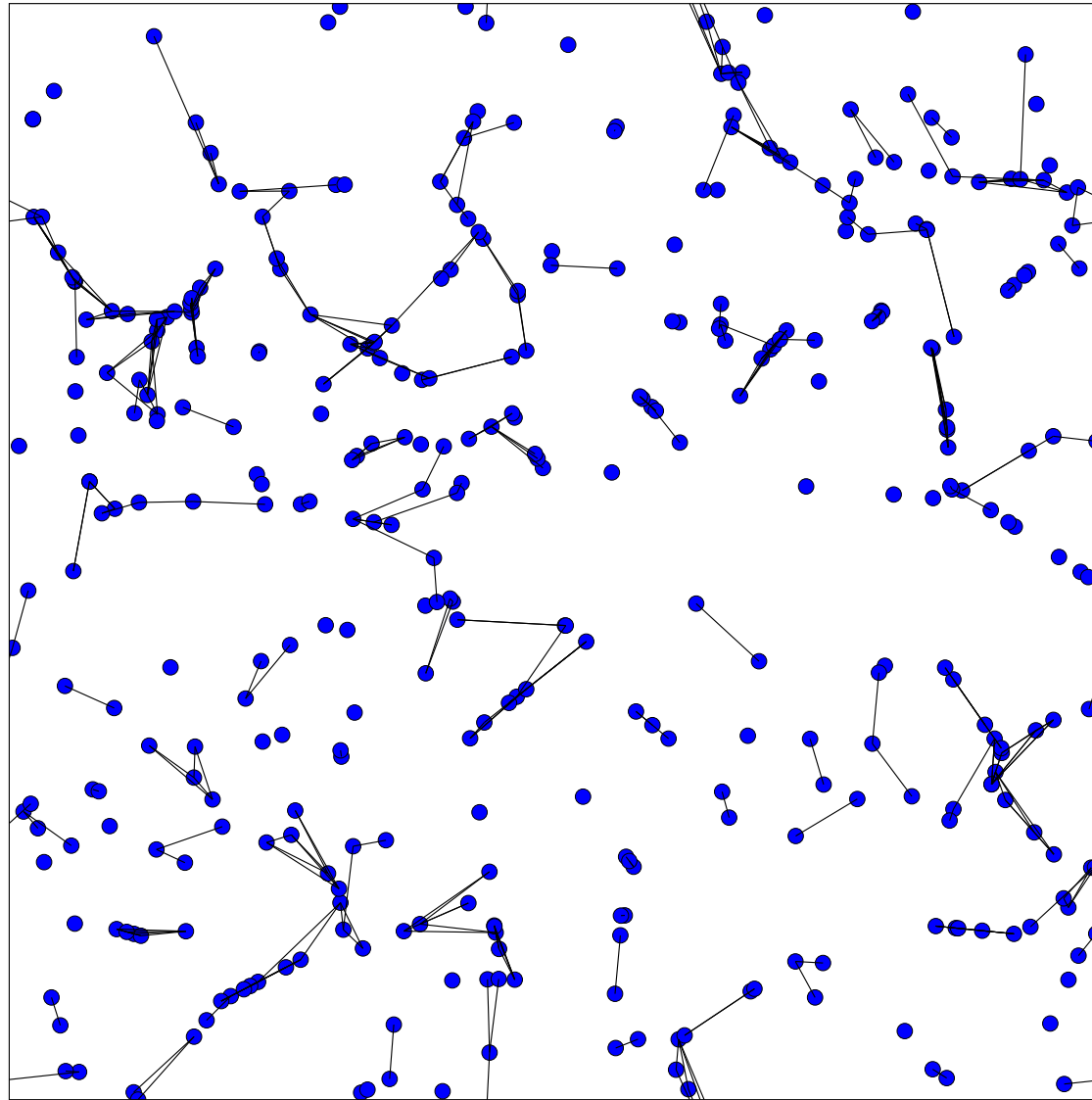


Figure: Realization of  $g_{T,\mu_v,\rho,r}(X^\lambda)$  with  $T = 1$ . The street system is suppressed.

# Illustration of connectivity graph $g_{T,\mu_v,\rho,r}(X^\lambda)$

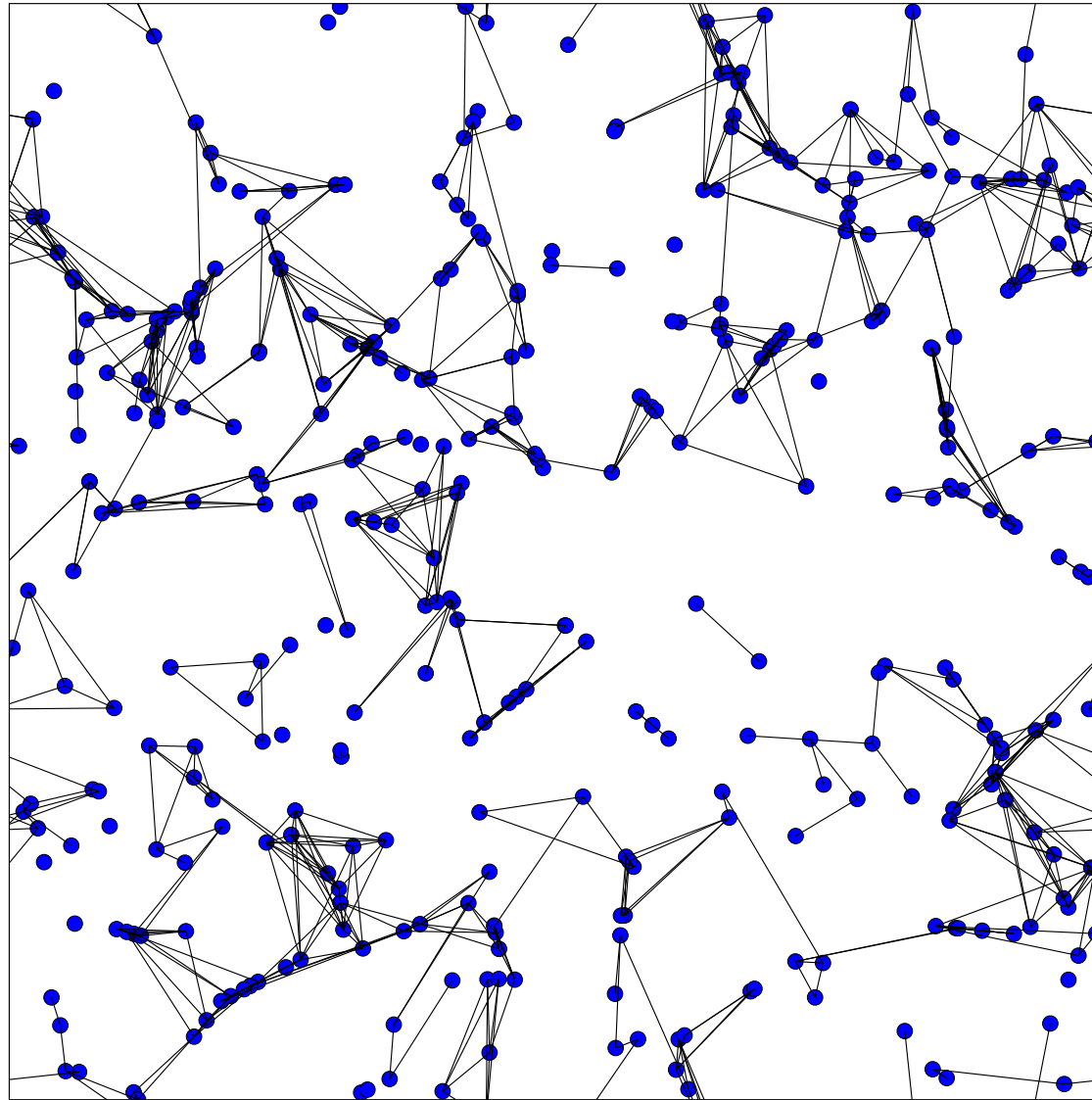


Figure: Realization of  $g_{T,\mu_v,\rho,r}(X^\lambda)$  with  $T = 2$ . The street system is suppressed.

# Illustration of connectivity graph $g_{T,\mu_v,\rho,r}(X^\lambda)$

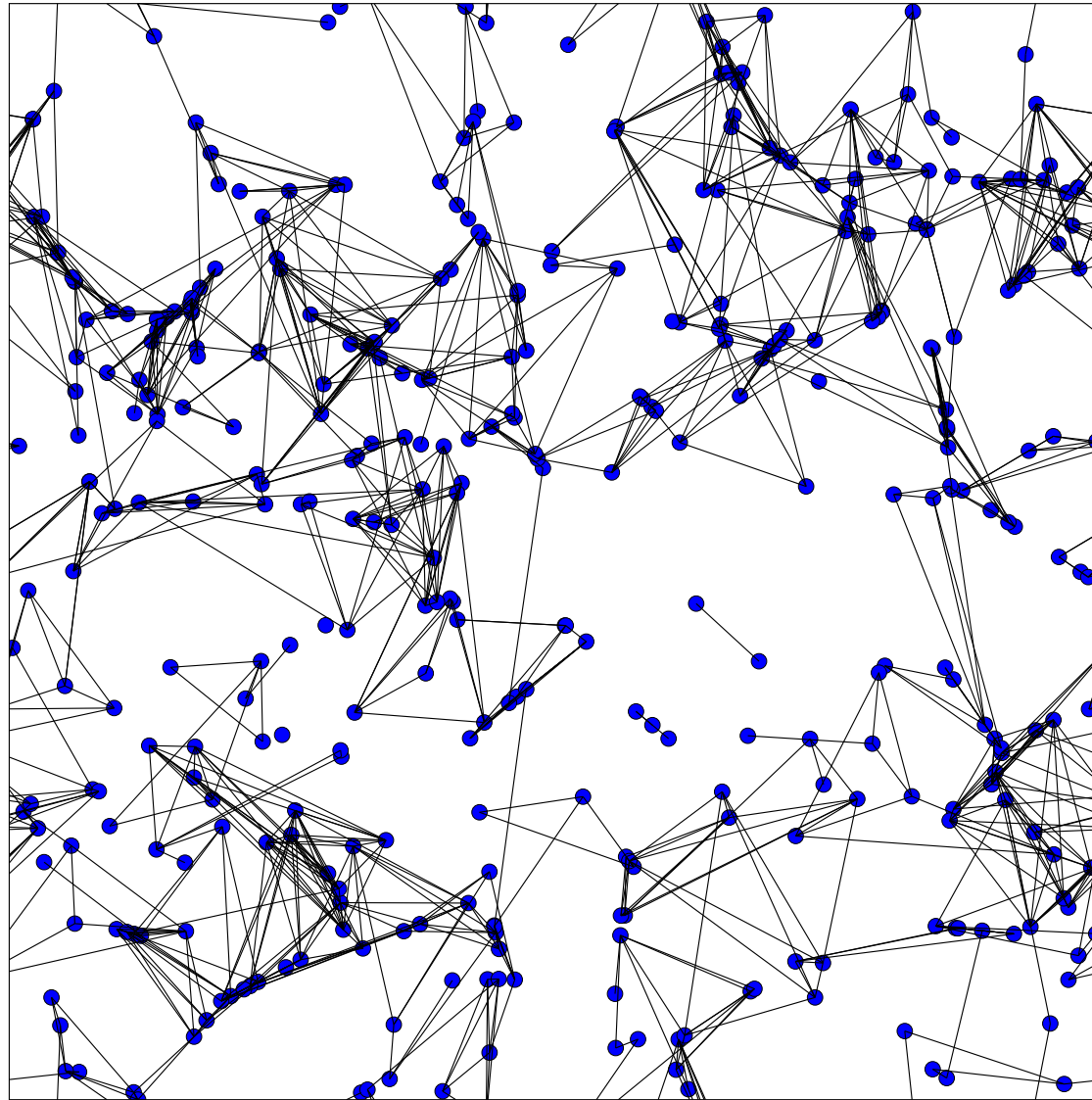


Figure: Realization of  $g_{T,\mu_v,\rho,r}(X^\lambda)$  with  $T = 3$ . The street system is suppressed.

# Illustration of connectivity graph $g_{T,\mu_v,\rho,r}(X^\lambda)$

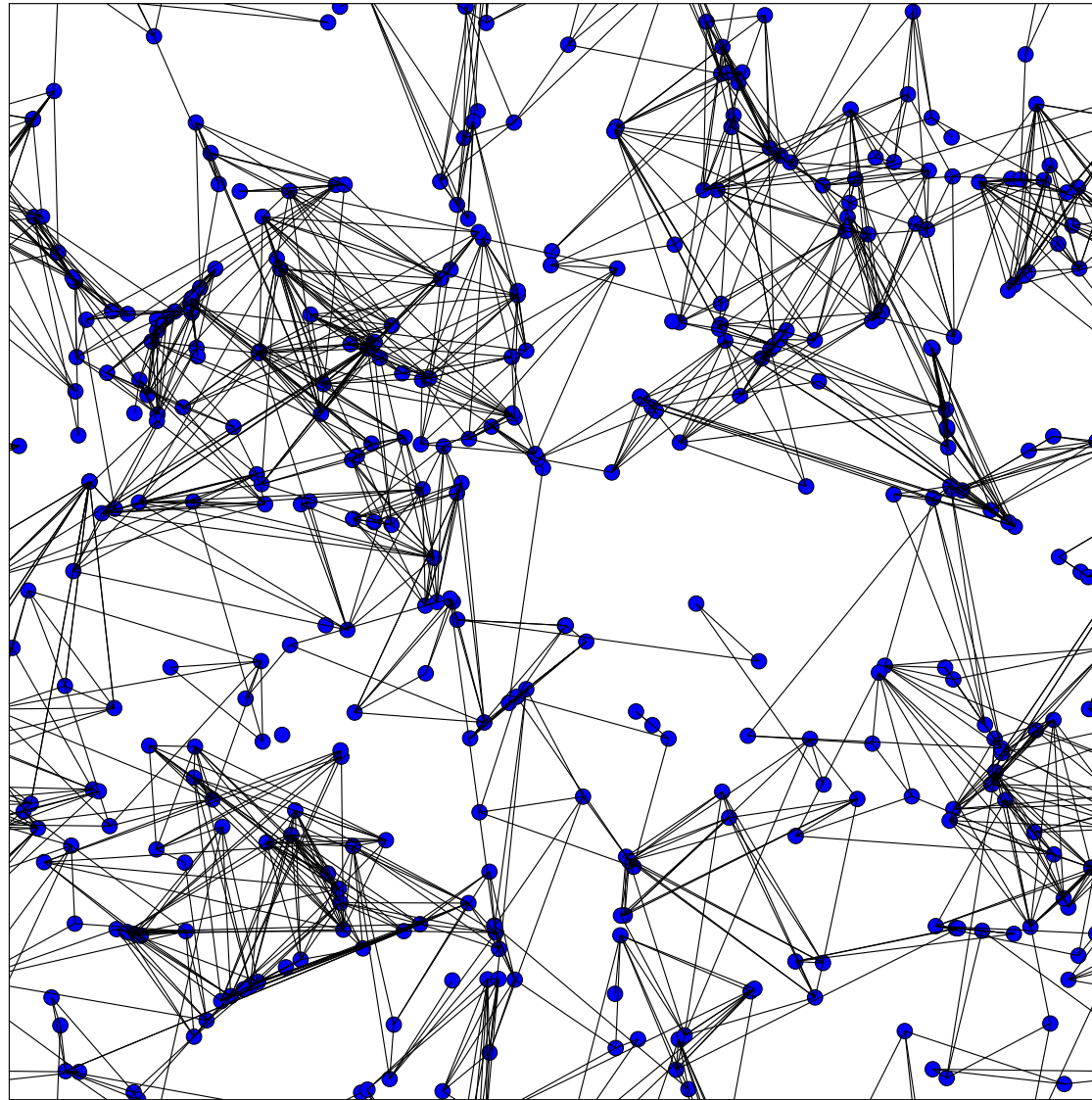


Figure: Realization of  $g_{T,\mu_v,\rho,r}(X^\lambda)$  with  $T = 4$ . The street system is suppressed.



# Illustration of connectivity graph $g_{T,\mu_v,\rho,r}(X^\lambda)$

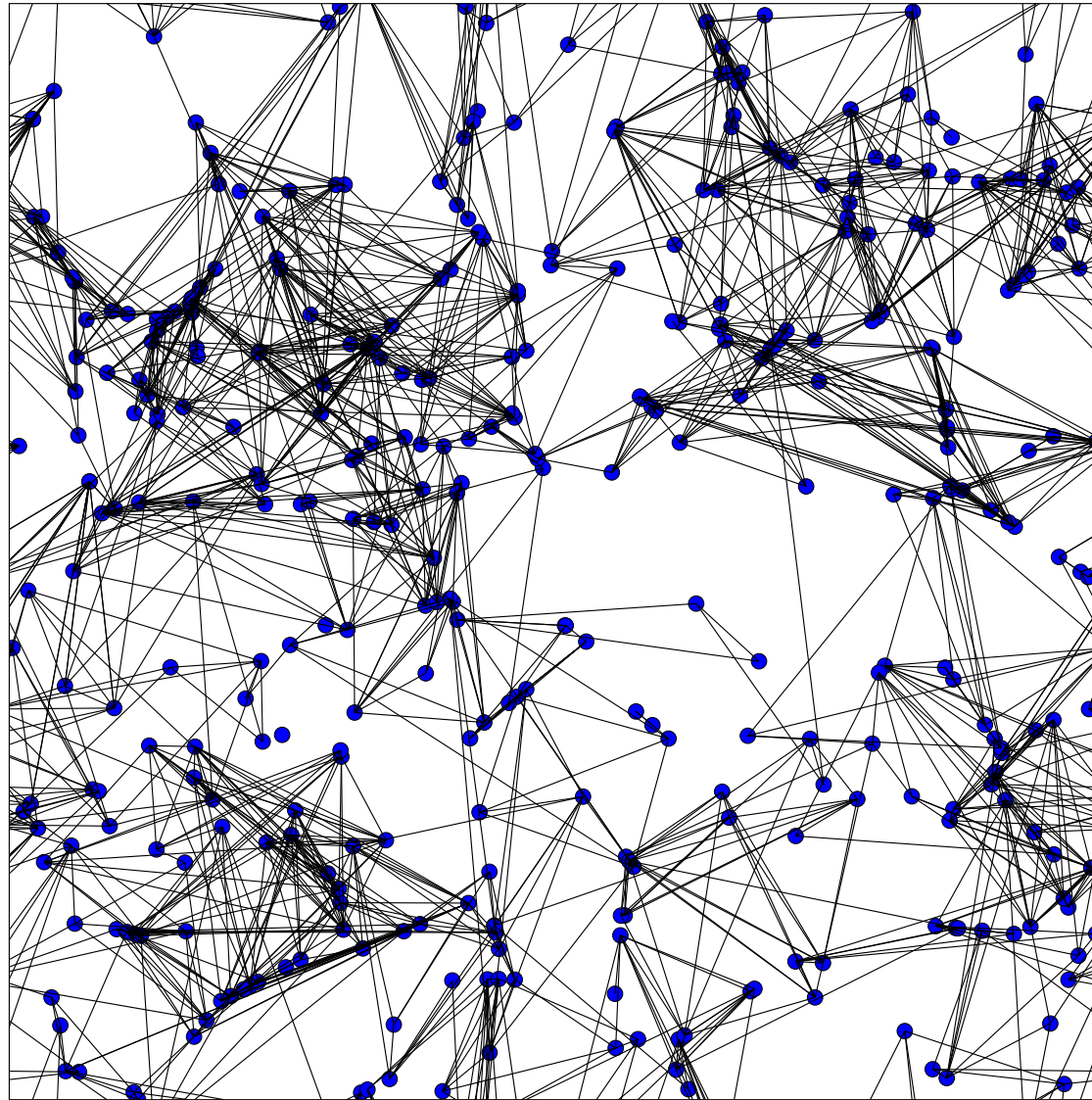


Figure: Realization of  $g_{T,\mu_v,\rho,r}(X^\lambda)$  with  $T = 5$ . The street system is suppressed

# Critical parameters for percolation

---

- Define **critical parameters for percolation**:

$$T_c = T_c(\lambda, \mu_v, \rho, r) := \inf\{T > 0: \mathbb{P}(g_{T, \mu_v, \rho, r}(X^\lambda) \text{ percolates}) > 0\}$$

$$\lambda_c = \lambda_c(T, \mu_v, \rho, r) := \inf\{\lambda > 0: \mathbb{P}(g_{T, \mu_v, \rho, r}(X^\lambda) \text{ percolates}) > 0\}$$

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$$r_c = r_c(T, \lambda, \mu_v, \rho) := \inf\{r > 0: \mathbb{P}(g_{T, \mu_v, \rho, r}(X^\lambda) \text{ percolates}) > 0\}$$

$$v_c = v_c(T, \lambda, \rho, r) := \inf\{av_{\max} > 0: \mathbb{P}(g_{T, \mu_v^a, \rho, r}(X^\lambda) \text{ percolates}) > 0\}$$

$$v^c = v^c(T, \lambda, \rho, r) := \sup\{av_{\min} > 0: \mathbb{P}(g_{T, \mu_v^a, \rho, r}(X^\lambda) \text{ percolates}) > 0\}$$

with  $\mu_v^a(dv) := \mu_v(dv/a)$

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with  $\mu_v^a(dv) := \mu_v(dv/a)$

- **Partial monotonicities:**

- $T \mapsto \mathbb{P}(g_{T, \mu_v, \rho, r}(X^\lambda) \text{ percolates})$  increasing
- $\lambda \mapsto \mathbb{P}(g_{T, \mu_v, \rho, r}(X^\lambda) \text{ percolates})$  increasing
- $r \mapsto \mathbb{P}(g_{T, \mu_v, \rho, r}(X^\lambda) \text{ percolates})$  increasing
- $\rho \mapsto \mathbb{P}(g_{T, \mu_v, \rho, r}(X^\lambda) \text{ percolates})$  decreasing
- $a \mapsto \mathbb{P}(g_{T, \mu_v^a, \rho, r}(X^\lambda) \text{ percolates})$  **non monotone**

# Absence of percolation I

---

Theorem ( $T_c, \lambda_c, v_c > 0$ )

*For any  $\kappa$  and all  $r, \rho \geq 0$  the following holds:*

- (1) *For all  $\lambda \geq 0$  and  $\mu_v$  we have that  $T_c(\lambda, \mu_v, \rho, r) > 2\rho$ ,*
- (2) *for all  $T \geq 0$  and  $\mu_v$  we have that  $\lambda_c(T, \mu_v, \rho, r) > 0$  and*
- (3) *for all  $T, \lambda \geq 0$  we have that  $v_c(T, \lambda, \rho, r) > 0$ .*

- Ideas for the proof: Comparison to Cox–Boolean models and discretization

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- Ideas for the proof: Comparison to Cox–Boolean models and discretization

## Theorem ( $v_c < \infty$ )

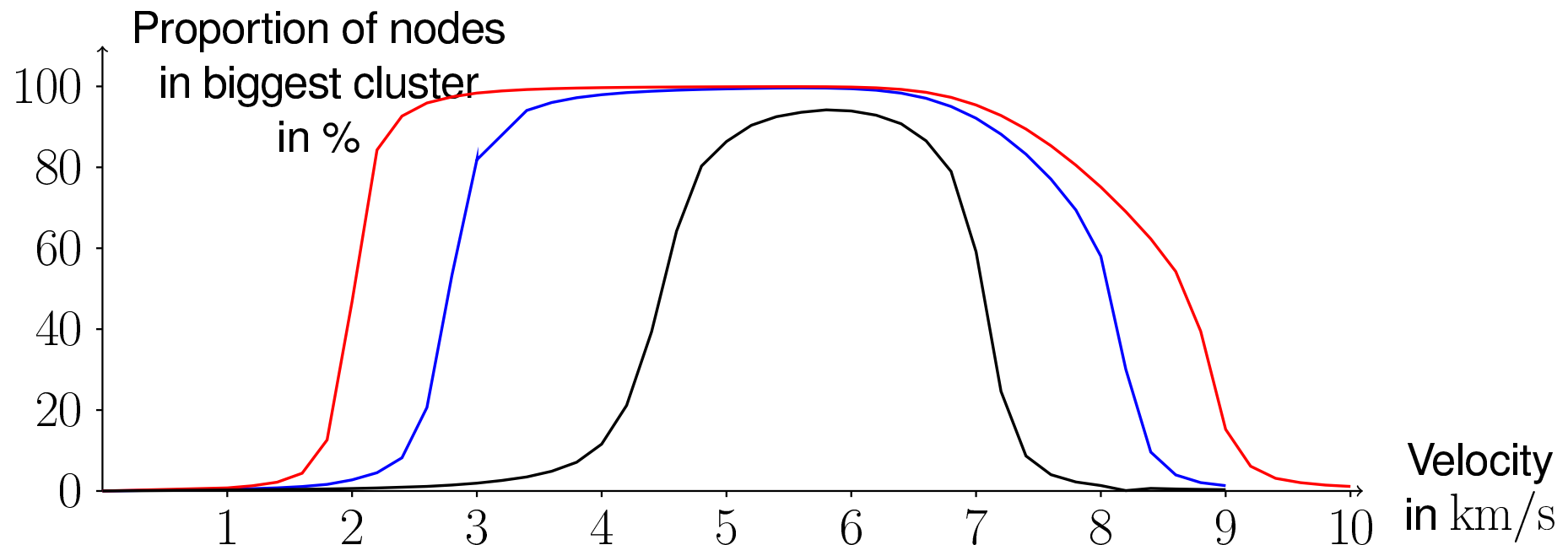
*Let  $S$  be a Poisson–Voronoi or Poisson–Delaunay tessellation. For any  $\kappa$  with bounded support and all  $T, \lambda, r \geq 0$  and all  $\rho > 0$  we have that  $v^c(T, \lambda, \rho, r) < \infty$ .*

- Ideas for the proof: Multiscale argument for geostatistical markings

# Absence of percolation I

## Take-home message

*In-and-out of percolation for  $v \mapsto g_{T,\delta_v,\rho,r}(X^\lambda)$ .*



**Figure:** In-and-out of percolation: Simulation of the largest connected cluster at different times  $T = 3\text{min}$  (black),  $T = 4.5\text{min}$  (blue), and  $T = 6\text{min}$  (red). Furthermore, we used  $\rho = 10\text{sec}$ ,  $r = 20\text{m}$ ,  $\lambda = 20\text{devices/km}$  and a street intensity of  $20\text{km/km}^2$ . For the velocities we chose a normal distribution, conditioned to be positive  $\mu_v = \mathcal{N}^+(v, v/5)$ .

# Absence of percolation II

Theorem ( $r = 0: \rho_c = 0$ )

For any  $\kappa, \mu_v$  and all  $T, \gamma, \lambda \geq 0$  we have that  $\rho_c(T, \lambda, \mu_v, 0) = 0$ .

Theorem ( $r > 0: \rho_c < \rho'_c(T, \mu_v)$ )

For any  $\kappa, \mu_v$  and  $T, \lambda, r \geq 0: \rho_c(T, \lambda, \mu_v, r) \leq \rho'_c(T, \mu_v) \wedge T/2$ .

Here  $\rho'_c(T, \delta_v) := \sup\{\rho > 0: \mathbb{P}(S^{v\rho, v(T-2\rho)} \text{ percolates}) > 0\}$ , with  $S^{a,b}$  given by edges  $\ell \subset S$  with  $|\ell| \geq a$  and edges between endpoints of  $\ell$  at distance  $\leq b$ .

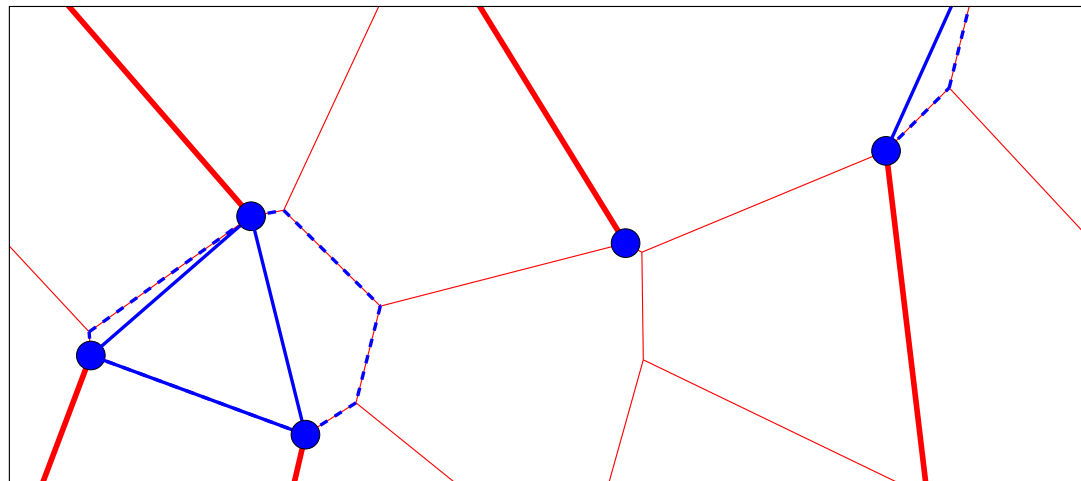


Figure: Realization of  $S^{v\rho, v(T-2\rho)}$  based on  $S$ .

# Thinned street systems

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- Consider  $a_c := \lim_{b \downarrow 0} a_c(b)$  where

$$a_c(b) := \sup\{a > 0: \mathbb{P}(S^{a,b} \text{ percolates}) > 0\}$$

## Theorem

*We have that  $a_c < \infty$ . Further, for any  $\kappa$  and  $T \geq 0$  as well as  $v > 4a_c/T$  it holds that  $\rho'_c(T, \delta_v) < T/2$ .*

- Proof via stabilization arguments



# Existence of percolation

- $\kappa$  is *c-well behaved* if for almost-all  $S$  and all  $x \in S$

$$B_c(x) \cap S \subset \text{supp}(\kappa^S(x, dy))$$

$\kappa$  is *well behaved* if  $\kappa$  is *c-well behaved* for some  $c > 0$

**Theorem ( $\rho = 0: \lambda_c < \infty$ )**

*For  $\kappa$  well behaved and  $T > 0$  then  $\lambda_c(T, \mu_v, 0, r) < \infty$ .*

**Theorem ( $\rho > 0: \lambda_c < \infty$ )**

*We have that  $0 < a_c(0)$ . Let  $\kappa$  be *c-well behaved* for some  $c > 0$ ,  $r, \rho > 0$  and  $T > 2\rho$ . Then, for all sufficiently small  $v_{\min}$  we have  $\lambda_c(T, \mu_v, \rho, r) < \infty$ .*

- Proof via coupling to system of good and bad streets and stabilization arguments

## Part 2: Chase escape in mobile device-to-device networks

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- **Contact times:** device  $X_i$  and  $X_j$  with trajectories  $(T_{i,t})_{t \geq 0}$  and  $(T_{j,t})_{t \geq 0}$  have contact at times

$$Z(X_i, X_j) = \{0 \leq t \leq T : |T_{i,t} - T_{j,t}| < r \text{ and } T_{i,t}, T_{j,t} \text{ are on same street}\}$$

- **Generalized connections:** device  $X_i$  and  $X_j$  are connected if for some  $t \geq 0$ ,

$$[t - \rho(X_i, X_j), t] \subset Z(X_i, X_j)$$

with  $\rho(X_i, X_j)$  iid **infection time** with distribution  $\varrho$

- **Shortest-path lengths:**  $\ell_S(x) = \sup\{|\ell_S(x, y)| : y \in \text{supp}(\kappa^S(x, dy))\} + r/2$   
where  $\ell_S(x, y)$  shortest path between  $x$  and  $y$  on  $S$
- **Degree:**  $\deg(X_i) = \#\{X_j \in X^\lambda \setminus X_i : |X_i - X_j| \leq \ell_S(X_i) + \ell_S(X_j)\}$
- **Local connectedness assumption:**  $\mathbb{P}(\exists X_i \in X^\lambda \text{ such that } \deg(X_i) = \infty) = 0$

## Proposition (Local connectedness)

*If  $\kappa$  has bounded support,  $S$  is exponentially stabilizing and  $|S \cap Q_1|$  has exponential moments, then the network is locally connected.*

- Proof via first-moment method for typical device
- Satisfied for standard examples PVT and PDT

# Devices and white knights

---

- **Initial infected device:**  $X_o$ , typical device at origin under Palm measure
- **Initial susceptible devices:** Cox process  $X^\lambda$  on  $S$
- **Initial white knights:** Cox process  $Y^{\lambda_W}$  on  $S$
- **Infection time distribution:**  $\varrho_I$  and **patching time distribution:**  $\varrho_W$
- **Minimal times:**  $\varrho_{I,\min} = \inf\{x \geq 0 : x \in \text{supp}(\varrho_I)\}$ , analogously for  $\varrho_{W,\min}$
- **Transmission mechanism:**
  - **Infection:** Infected  $X_i$  transmits to susceptible  $X_j$  after completion of  $\rho_I(X_i, X_j)$
  - **Patching:** White knight  $X_i$  transmits to infected  $X_j$  after completion of  $\rho_W(X_i, X_j)$
- **Process of susceptible and infected devices and white knights:**  $(S_t, I_t, W_t)_{t \geq 0}$

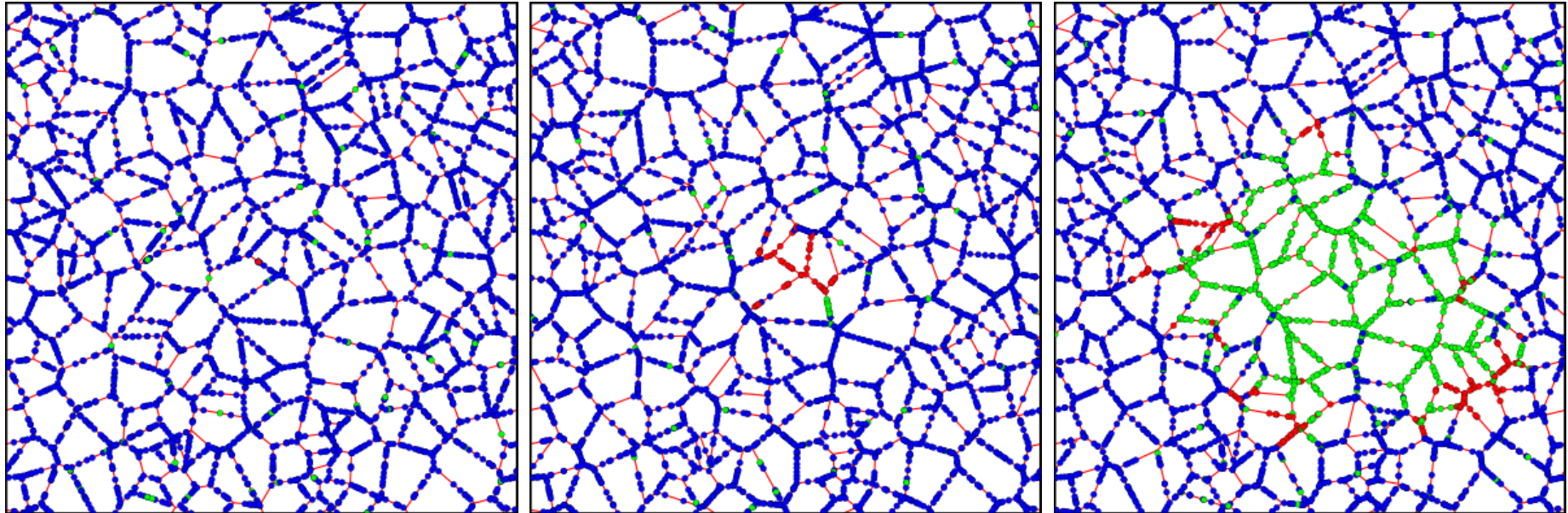


Figure: Propagation of infected devices (red) on a street system given by a Poisson–Voronoi tessellation. In the initial state (left) there is exactly one infected device present in the center and the remaining devices are either susceptible (blue) or white knights (green). At some small positive time (middle) further devices in the vicinity of the initially infected device have become infected and have started to make contact to white knights. At some later time (right) infected devices are only present along the boundary of the set of white knights in the center region.

- **Waypoint continuity:**  $\kappa$  is  $c$ -continuous if  $\inf_{y \in B_c(x)} \kappa^S(x, y) > 0$  where  $\kappa^S(x, y)$  Lebesgue density defined for almost-all  $S$  and  $x \in S$
- **Street connectedness:**  $S^a = \{s \in S : |s| \geq a\}$  and  $R_x^a$  distance to furthest point from  $x$  reachable without crossing  $S^a$ . Call graph  $R^a$ -connected if

$$\lim_{n \uparrow \infty} \mathbb{P}\left(\sup_{x \in Q_n \cap \mathbb{Q}^2} R_x^a < n\right) = 1$$

Let  $a_c := \sup\{a > 0 : S^a \text{ is } R^a\text{-connected}\}$

- **Global survival:**  $|\bigcup_{t \geq 0} I_t| = \infty$

## Theorem (Global survival)

*Let  $\kappa$  be  $c$ -continuous for some  $c > 0$  and  $\varrho_{W,\min} > \varrho_{I,\min} > 0$ . Then, for all sufficiently small  $v_{\min}$  such that  $0 < v_{\min} \varrho_{I,\min} < \min(a_c/2, r, c/2)$ , there exists  $\lambda_c > 0$  such that for all  $\lambda > \lambda_c$  and all  $\lambda_W \geq 0$ ,*

$$\mathbb{P}(\text{infection survives globally}) > 0.$$

# Extinction

---

- **Punchline:** If  $\varrho_{W,\min} > \varrho_{I,\min} > 0$ , then survival guaranteed for all  $\lambda_W \geq 0$
- **Proof idea:** Existence of infinite cluster of good streets on which infected wins against patching

# Extinction

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## Theorem (Extinction)

*Let  $S$  be a Poisson–Voronoi or Poisson–Delaunay tessellation. Let  $\kappa$  be well behaved with bounded support and assume that  $r > v_{\max} \varrho_{W,\min}$  and  $\varrho_{I,\min} > \varrho_{W,\min} > 0$ . Then, for all  $\lambda \geq 0$ , there exists  $\lambda_{W,c} > 0$  such that for all  $\lambda_W > \lambda_{W,c}$  we have that*

$$\mathbb{P}(\text{infection survives globally}) = 0.$$

- **Proof idea:** Multiscale argument



## Theorem (Speed-dependent survival and extinction)

*Let  $S$  be a Poisson–Voronoi or Poisson–Delaunay tessellation. Let  $\kappa$  have bounded support and be  $c$ -well behaved for some  $c > 0$ . Assume further that  $\varrho_W = \delta_{\rho_W}$  and  $\varrho_I = \delta_{\rho_I}$  with  $\rho_W > \rho_I$ , and let  $0 < v_o < \min(a_c/2, r, c/2)/\rho_I$ . Then, there exists  $\lambda_c > 0$  and  $\lambda_{W,c} > 0$  (independent of  $\lambda_c$  and  $v_o$ ), such that for all  $\lambda > \lambda_c$  and all  $\lambda_W > \lambda_{W,c}$  we have that*

- (1) *there exists  $v_o > v_c(\lambda, \lambda_W) > 0$  such that for all  $\mu_v = \delta_v$  with  $v < v_c(\lambda, \lambda_W)$  we have*

$$\mathbb{P}(\text{infection survives globally}) = 0,$$

- (2) *for  $\mu_v = \delta_{v_o}$  we have  $\mathbb{P}(\text{infection survives globally}) > 0$ , and*

- (3) *there exists  $\infty > v^c > v_o$  (independent of  $\lambda$  and  $\lambda_W$ ), such that for all  $\mu_v = \delta_v$  with  $v > v^c$  we have*

$$\mathbb{P}(\text{infection survives globally}) = 0.$$

# Thank you for your attention

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