

# Graph models of wireless networks

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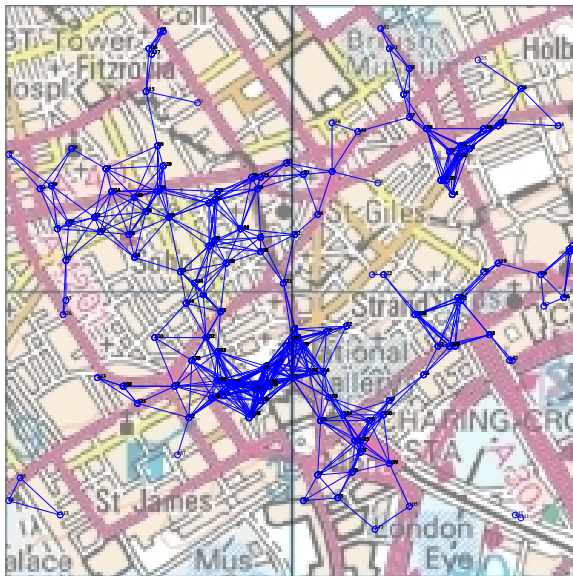
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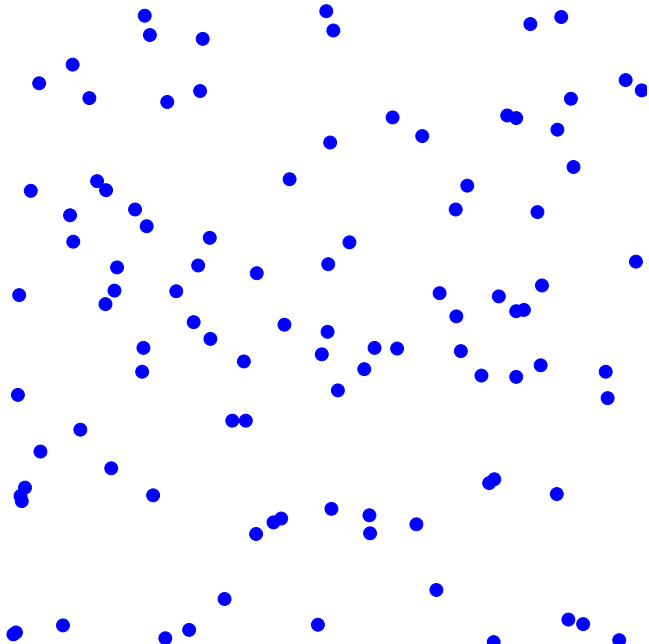
Corrected version 2009-02-06

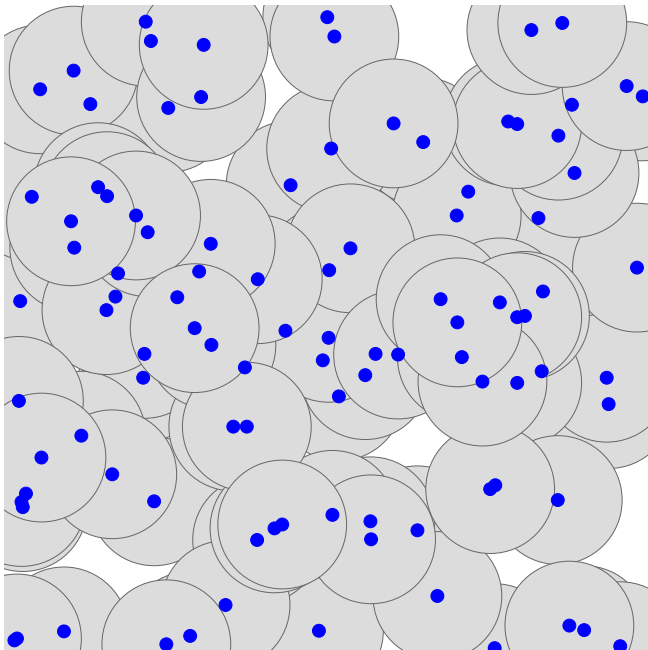
# BT Research - Adastral Park

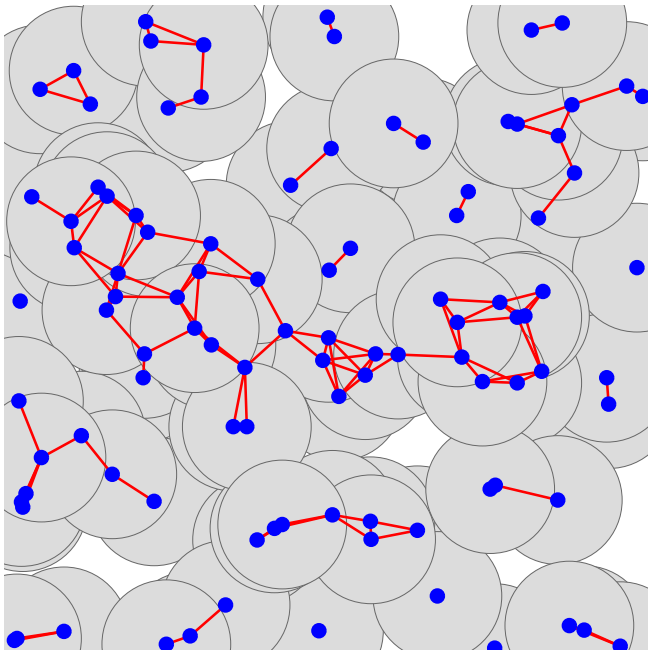


# Wireless networks









# PPPP( $\lambda$ ): definitions and statistical properties

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- ▶ also binomial PP
- ▶ also nonhomogeneous case

## PPPP( $\lambda$ ): radial generation

▶  $s=0$

▶ do

$$s \leftarrow s - \log(\text{Uniform}(0, 1))$$

$$\theta = 2\pi \text{Uniform}(0, 1)$$

$$r = \sqrt{s / (\pi\lambda)}$$

$$x = r \cos \theta$$

$$y = r \sin \theta$$

▶ while  $r <$  desired maximum radius

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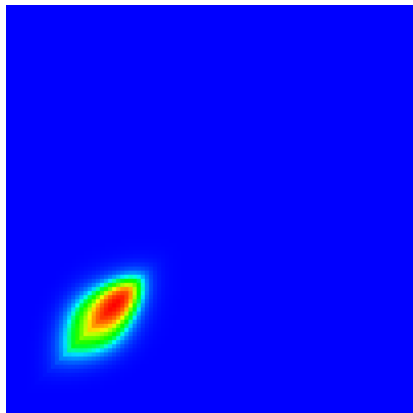
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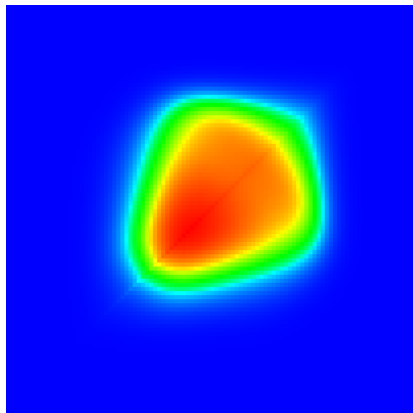
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- ▶ the cluster coefficient is  $1 - \frac{3\sqrt{3}}{4\pi}$
- ▶ surprisingly, the degree-degree correlation is the same, independently of  $\lambda$  and  $\rho$  !

# GRG(20, $\rho$ ) degree-degree distribution

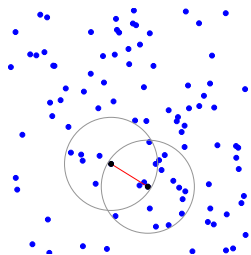


$\rho=0.3$



$\rho=0.5$

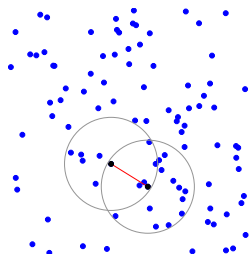
## GRG( $\lambda, \rho$ ) degree-degree correlation



- ▶ If  $X_0 \sim \text{Poi}(\lambda_0)$ ,  $X_1 \sim \text{Poi}(\lambda_1)$ ,  $X_2 \sim \text{Poi}(\lambda_1)$  are independent, and  $Y_1 = X_1 + X_0$ ,  $Y_2 = X_2 + X_0$ , then

$$\text{corr}(Y_1, Y_2) = \frac{\lambda_0}{\lambda_0 + \lambda_1}$$

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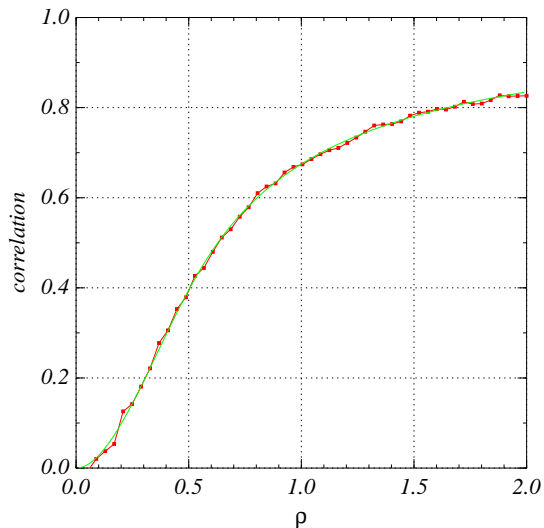
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- ▶ For PPP, degree-degree correlation is  $E[\text{corr}]$

$$\begin{aligned} &= \int_0^\rho \frac{2\rho^2 \arccos(x/(2\rho)) - (x/2)\sqrt{4\rho^2 - x^2}}{\pi\rho^2} \frac{2x}{\rho^2} dx \\ &= 1 - 3\sqrt{3}/(4\pi) \simeq 0.5865 \end{aligned}$$

# GRG( $\lambda, \rho$ ) degree-degree correlation - square



exact (doable but messy); simulation



## GRG( $\lambda, \rho$ , unit circle): degree distribution

- ▶ pdf of distance of a random point from the centre, given that it is within  $1 - \rho$  of the edge:

$$f_{\rho}(x) = \frac{(4 - 2\rho)x + 2\rho - 2}{\rho^3 - 2\rho^2 + 2\rho} \mathbb{I}[1 - \rho < x < 1]$$

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$$A(x) = \rho^2 \arccos\left(\frac{x^2 + \rho^2 - 1}{2x\rho}\right) + \arccos\left(\frac{x^2 - \rho^2 + 1}{2x}\right) - \frac{1}{2}[(1 - x + \rho)(x + \rho - 1)(x - \rho + 1)(x + \rho + 1)]^{1/2}$$

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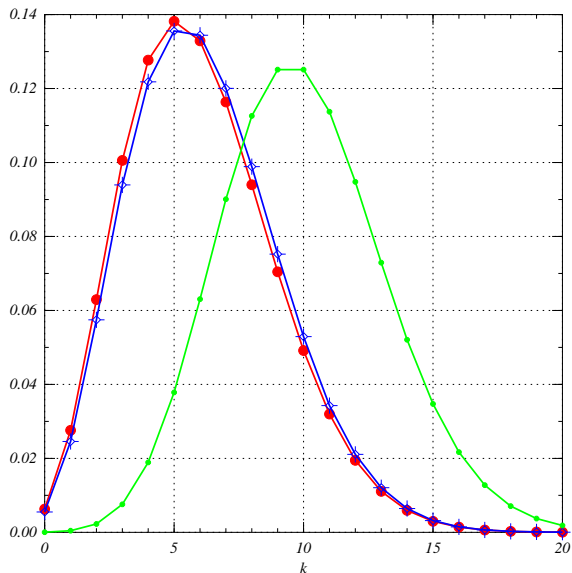
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- ▶  $\text{Prob}[d=k] = (1 - \rho)^2 \text{Poi}(A(x), k) + \rho(2 - \rho) \int_{1 - \rho}^1 \text{Poi}(A(x)\lambda) f_{\rho}(x) dx$
- ▶ where  $\text{Poi}(\mu, k) = e^{-\mu} \mu^k / k!$

# GRG( $\lambda, \rho$ , unit circle) degree distribution



exact; simulation;  
Poisson - ignores  
edge effect.

# Poisson maxima 1

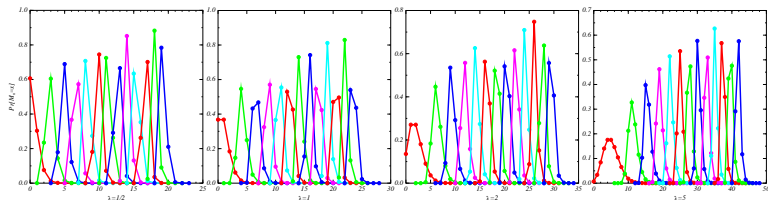
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The distribution of the maximum of Poisson $_{\lambda}$  variables for  $\lambda=1/2, 1, 2, 5$  (left to right) and  $n=10^0, 10^2, 10^4, \dots, 10^{24}$

## Poisson maxima 2

- ▶ Anderson:  $\exists I_n \in \mathbb{Z}$  s.t.  $\Pr[M_n \in \{I_n, I_n + 1\}] \rightarrow 1$

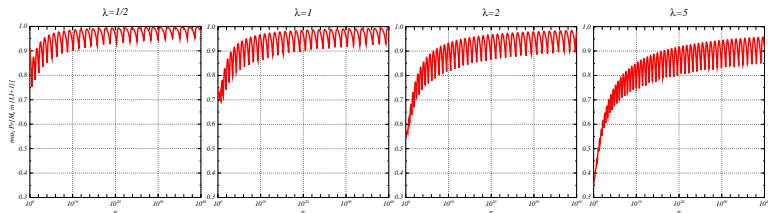


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The maximal probability (with respect to  $I_n$ ) that  $M_n \in \{I_n, I_n + 1\}$  for  $\lambda=1/2, 1, 2, 5$  (left to right) and  $10^0 \leq n \leq 10^{40}$ . The curves show the probability that  $M_n$  takes either of its two most frequently occurring values.

## Poisson maxima 3

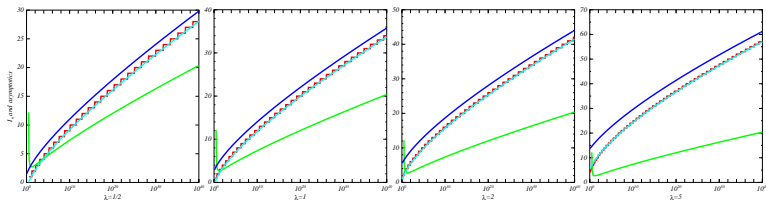
►  $M_n \sim x_0 \equiv \log n / W\left(\frac{\log n}{e\lambda}\right)$

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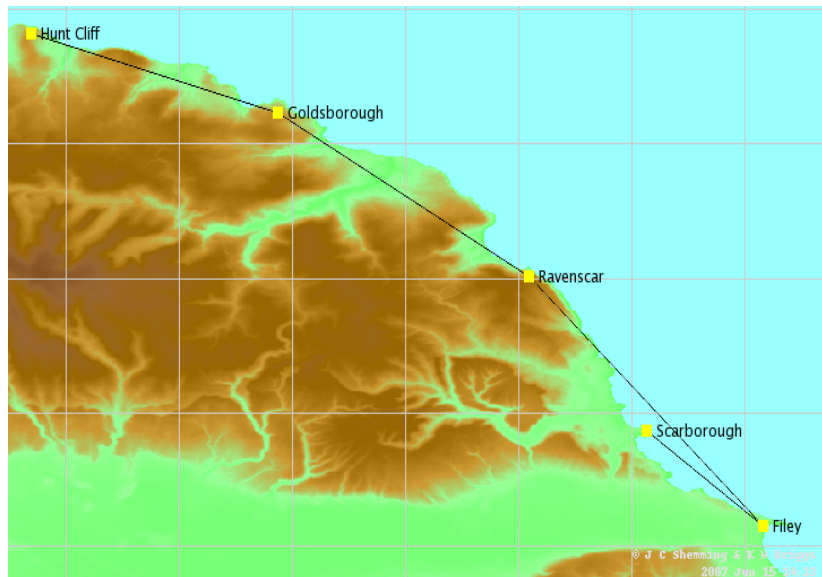
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# Roman networks



# Anglo-Saxon networks

