

# Connectivity of nodes

Keith Briggs

Keith.Briggs@bt.com

[research.btexact.com/teralab/keithbriggs.html](http://research.btexact.com/teralab/keithbriggs.html)

The logo for BTexact, featuring the letters 'BT' in a bold, blue, sans-serif font, followed by the word 'exact' in a purple, lowercase, sans-serif font.

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TYPESET IN PDF $\text{\LaTeX}$  ON A LINUX SYSTEM

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# Introduction

I consider connectivity of nodes with a radio range  $\rho$  placed uniformly and randomly in a bounded region under various models:

- *Poisson 1d model*: the nodes exist on all of  $\mathbb{R}$  with a exponential distribution of separation with parameter  $\lambda$ , and a window of unit length is placed over them. The number of nodes visible through the window is Poisson distributed.
- *fixed- $n$  1d model*: there are exactly  $n$  nodes independently and uniformly placed in  $[0, 1]$ .
- *Poisson 2d model*: the nodes exist on all of  $\mathbb{R}^2$  with a intensity  $\lambda$ , and a finite-area window is placed over them. The number of nodes visible through the window is Poisson distributed.
- *fixed- $n$  2d model*: there are exactly  $n$  nodes independently and uniformly placed in a bounded region  $R$ .

## Notation:

- ▷ *pdf=probability density function*
- ▷ *cdf=cumulative distribution function*
- ▷ *The notation is sloppy in not distinguishing a RV  $X$  and its values  $x$*
- ▷  *$[[x]]$  is the indicator function: 1 if  $x$  is true, else 0*

# Theory for the Poisson 1d model

- $\lambda$  is the intensity of nodes per unit length
- The pdf of the internode distance  $d$  is  $f(d) = \lambda e^{-\lambda d}$
- The cdf of the internode distance is  $F(d) = 1 - e^{-\lambda d}$
- The expectation of  $d$  is  $\mathbb{E}[d] = 1/\lambda$

## More theory for the Poisson 1d model

We now place a unit length window over  $\mathbb{R}$  and assume that  $n$  nodes are visible. The following results are conditional on  $n$

- There are  $n-1$  internode intervals, and the cdf of the maximum interval is  $F_{n-1}(d) = (1 - e^{-\lambda d})^{n-1}$
- the cdf of the minimum interval is  $F_1(d) = 1 - e^{-n\lambda d}$
- the pdf of the minimum interval is  $f_1(d) = n\lambda e^{-n\lambda d}$
- The expectation of the minimum interval is  $\mathbb{E}[d_{(1)}] = 1/(2\lambda)$ , so is half the expectation of the internode distance
- The intervals have correlation  $-1/n$
- The probability of full connectivity for the  $n$  nodes is thus approximately (i.e. ignoring correlation and edge effects)  $F_{n-1}(\rho) = (1 - e^{-\lambda\rho})^{n-1}$
- This result is only approximate. We should expect deviations small  $n$

The exact theory for the fixed- $n$  case is here

# Order statistics theory [dav70]

Let  $x_1, x_2, \dots, x_n$  be RVs uniformly distributed in  $[0, 1]$ .

Sort them in increasing order as  $x_{(1)} \leq x_{(2)} \leq \dots \leq x_{(n)}$ .

- The pdf of  $x_{(k)}$  is

$$\frac{1}{(k-1)!} \binom{n}{k} x^{k-1} (1-x)^{n-k}$$

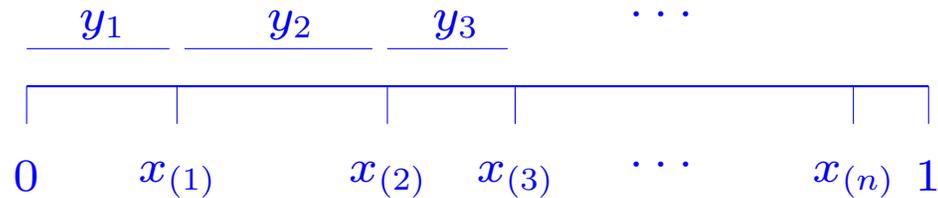
- The cdf of the range  $r = x_{(n)} - x_{(1)}$  is  $nr^{n-1} - (n-1)r^n$
- If  $w_{rs} = x_{(s)} - x_{(r)}$ , then the pdf of  $w_{rs}$  is

$$w_{rs}^{s-r-1} (1-w_{rs})^{n-s+r} / \text{B}(s-r, n-s+r+1)$$

- For the special case of adjacent nodes ( $s = r+1$ ), this becomes  $n(1-w_{r,r+1})^{n-1}$ , which gives a cdf of  $1 - (1-w_{r,r+1})^n$
- However, the  $w_{r,r+1}$  are **not independent** random variables, so the probability that the maximum of  $n-1$  samples of  $w_{r,r+1}$  is less than a constant  $\rho$ , is NOT  $[1 - (1-\rho)^n]^{n-1}$ 
  - ▷ *But this is approximately correct for large  $n$  and  $\rho$  near 1 and is plotted in blue on the graphs of simulation results*
  - ▷ *As  $\rho \rightarrow 1$ , this becomes  $1 - (n-1)(1-\rho)^n$ . cf. the exact equation*

# Exact theory for the fixed- $n$ 1d model

- Let  $y_k = x_{(k)} - x_{(k-1)}$  be the gaps ( $k = 2, \dots, n$ ), with  $y_1 = x_{(1)}$



- Their joint pdf is (for  $1 \leq m \leq n$  and  $\sum_{i=1}^m y_i \leq 1$ )

$$f(y_1, y_2, \dots, y_m) = \frac{n!}{(n-m)!} \left(1 - \sum_{i=1}^m y_i\right)^{n-m}$$

- If  $c_i$  are constants such that  $\sum_{i=1}^m c_i \leq 1$ , then by integrating the pdf we obtain

$$\Pr [y_1 > c_1, y_2 > c_2, \dots] = \left(1 - \sum_{i=1}^m c_i\right)^{n-1}$$

- Boole's law for the probability of at least one event  $A_i$  of  $n$  events  $A_1, A_2, \dots, A_n$  occurring is

$$\Pr \left[ \bigcup_{i=1}^n A_i \right] = \sum_i \Pr [A_i] - \sum_{i < j} \Pr [A_i A_j] + \dots + (-1)^{n-1} \Pr [A_1 A_2 \dots A_n]$$

## Exact theory for the fixed- $n$ 1d model (cotd.)

- We don't care about  $y_1$ , so we put  $c_1 = 0$
- Using Boole's law, the probability that the largest  $y_k$  exceeds some constant  $\rho$  is

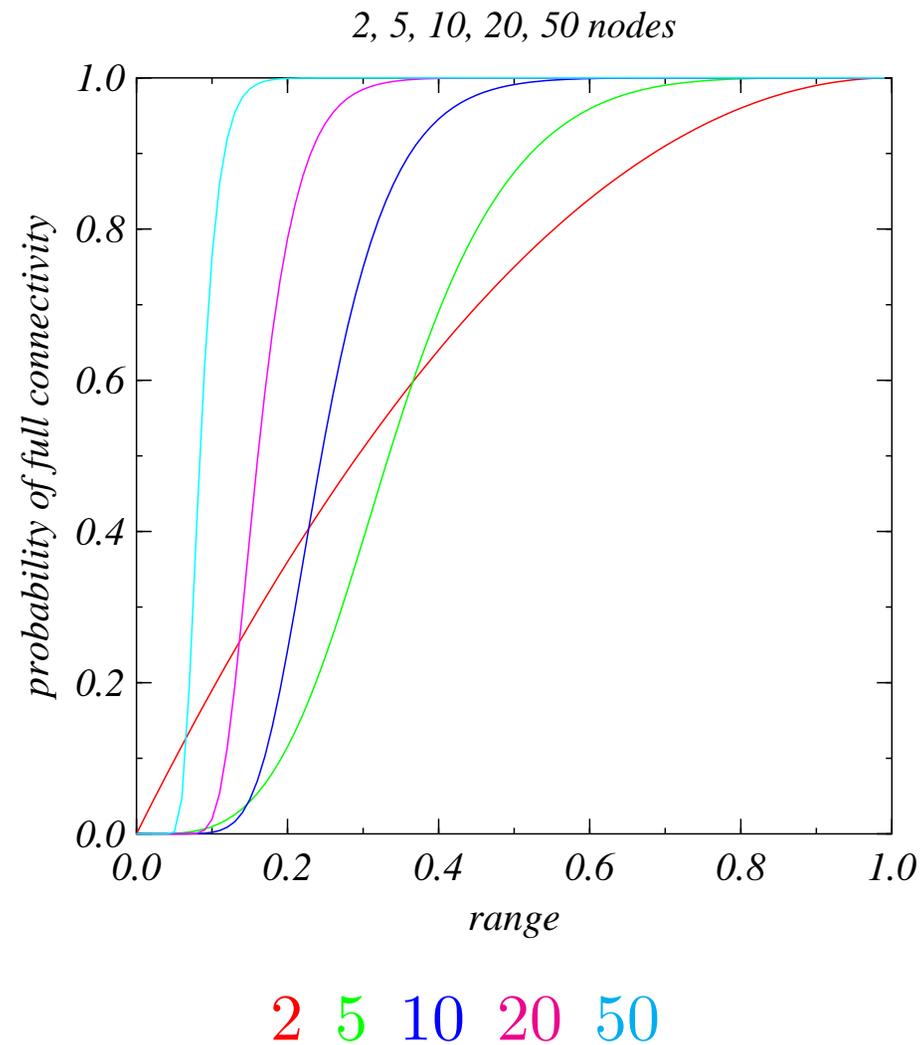
$$\Pr [y_{(n)} > \rho] = (n-1) \Pr [y_1 > \rho] - \binom{n-1}{2} \Pr [y_1 > c_1, y_2 > c_2] + \dots$$

- Thus

$$\Pr [\text{fully connected}] = 1 - \sum_{i=1}^{\lfloor 1/\rho \rfloor} (-1)^{i+1} \binom{n-1}{i} (1-i\rho)^n$$

- This is plotted as a red line on the following pages
- Note that for  $\rho > 1/2$ , this is exactly  $1 - (n-1)(1-\rho)^n$ .
  - ▷ *cf. an approximation*

# Probability of connectivity for the fixed- $n$ 1d model



# Theory for the Poisson 2d model [cre91]

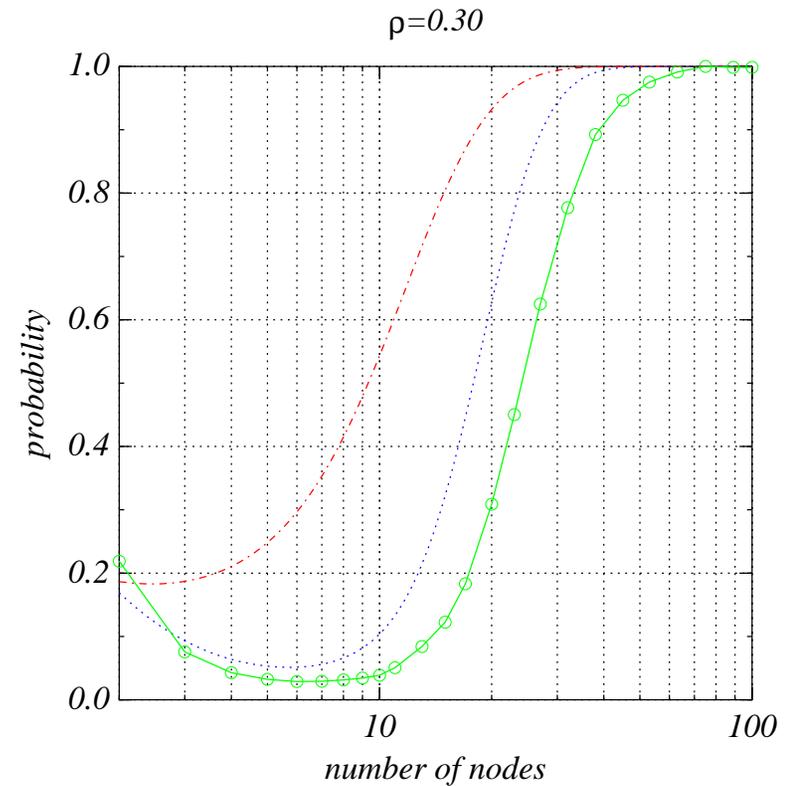
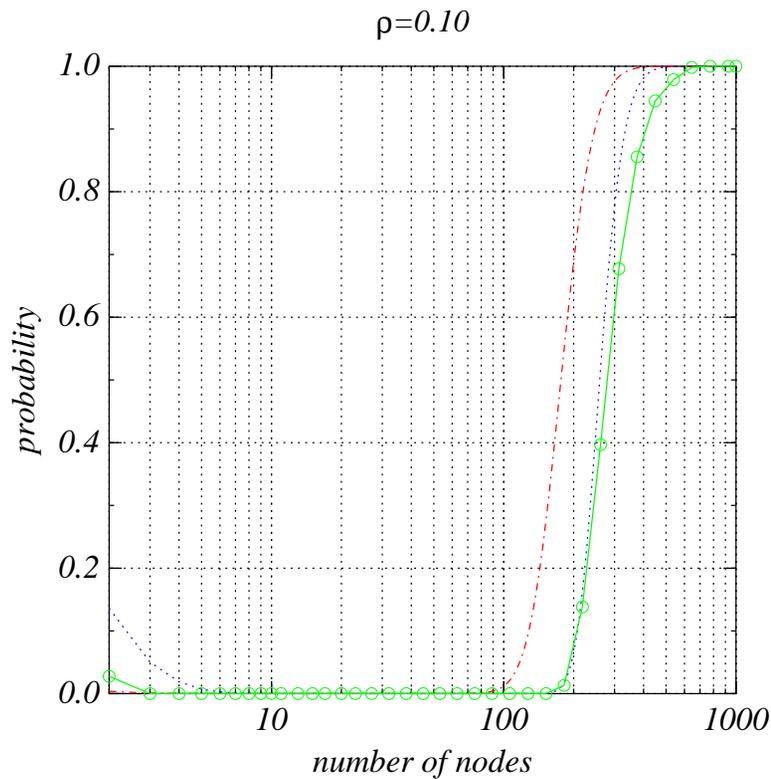
- $\lambda$  is the intensity of nodes per unit area
- The pdf of the nearest neighbour distance  $d$  is  $f(d) = 2\pi\lambda d e^{-\lambda\pi d^2}$
- The cdf of  $d$  is  $F(d) = 1 - e^{-\pi\lambda d^2}$
- The expectation of  $d$  is  $\mathbb{E}[d] = 1/(2\lambda^{1/2})$
- The variance of  $d$  is  $(4 - \pi)/(4\pi\lambda)$
- The probability of a node being isolated (i.e. having no neighbour within range  $\rho$ ) is  $e^{-\pi\lambda\rho^2}$

# Theory for the Poisson 2d model

We now place a window  $R$  of area  $A$  over  $\mathbb{R}^2$

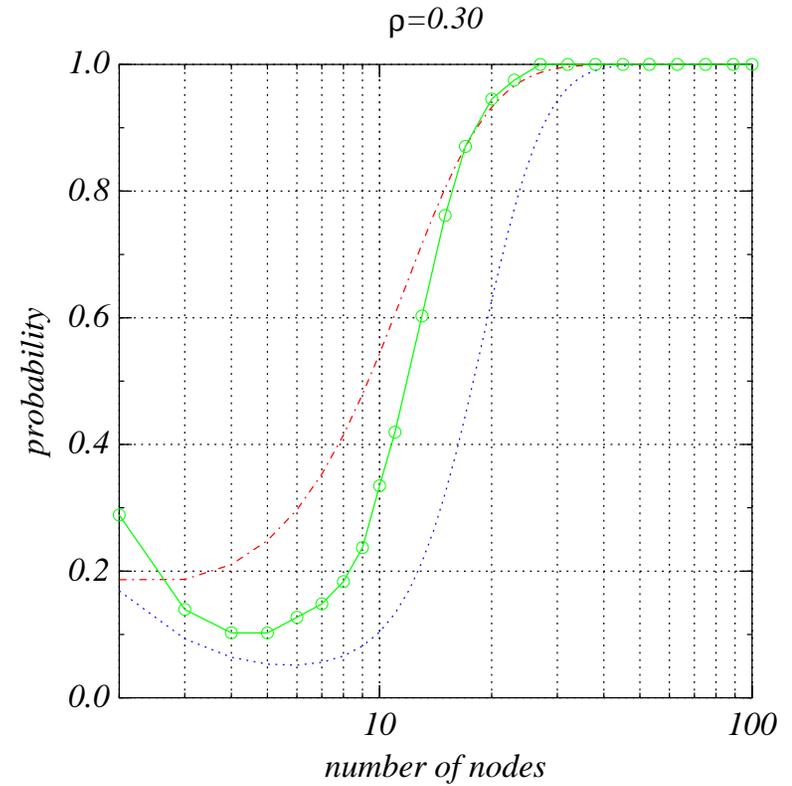
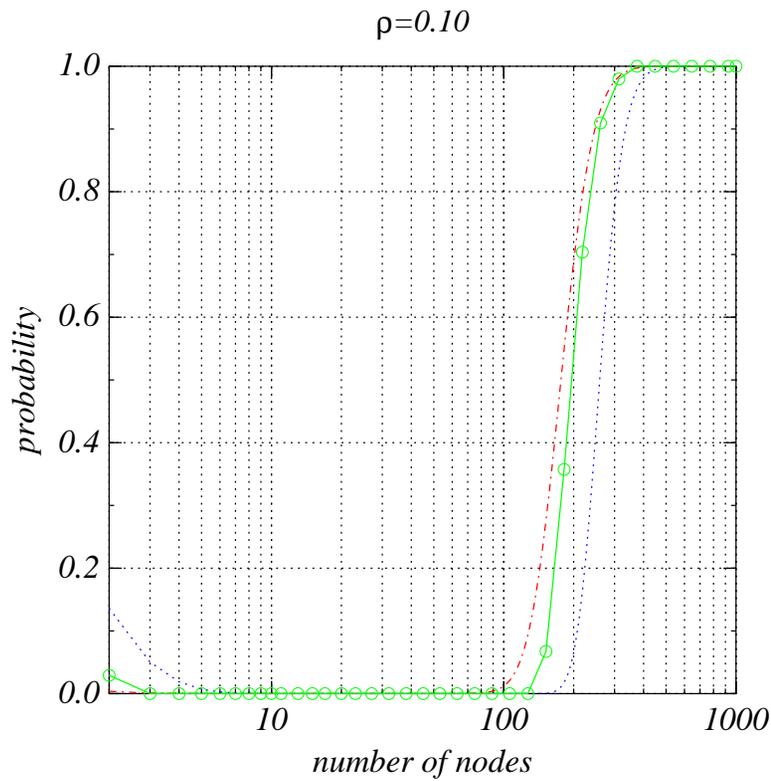
- The number of nodes visible will be Poisson distributed with mean  $\lambda A$
- Conditional on  $n$  nodes being visible, and if the nearest neighbour distances were independent (which is *not* the case) the probability of no node being isolated would be  $\left(1 - e^{-\pi\lambda\rho^2}\right)^n$
- There is no simple way to compute the probability of full connectivity. However, since a necessary condition is that no node is isolated, the last expression is an approximate upper bound for the fixed- $n$  model and is plotted in **red** on the following graphs
- The **blue** curve is the asymptotic probability of the whole region  $R$  being covered, using this theory

# Simulation results - square, $\rho = 0.1, 0.3$



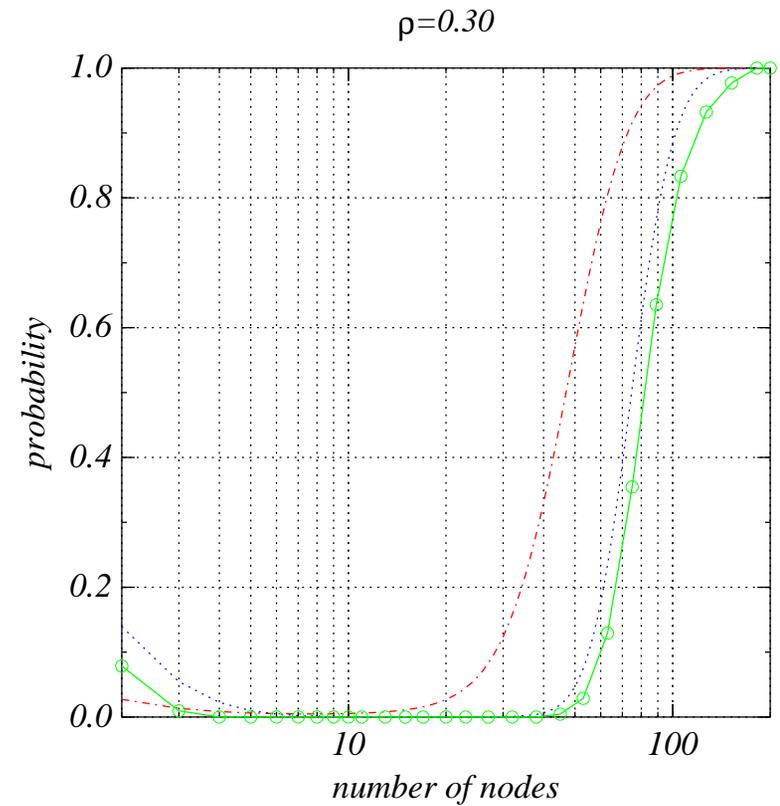
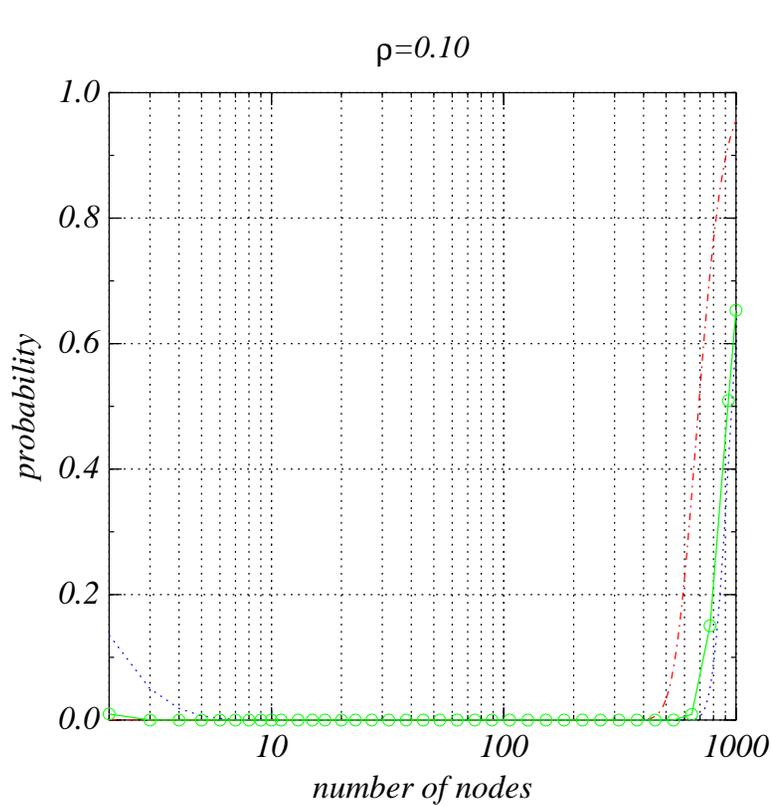
$$\left[1 - e^{-\pi \lambda \rho^2}\right]^n \quad \text{simulation asymptotic}$$

# Simulation results - torus, $\rho = 0.1, 0.3$



$$\left[1 - e^{-\pi \lambda \rho^2}\right]^n \text{ simulation asymptotic}$$

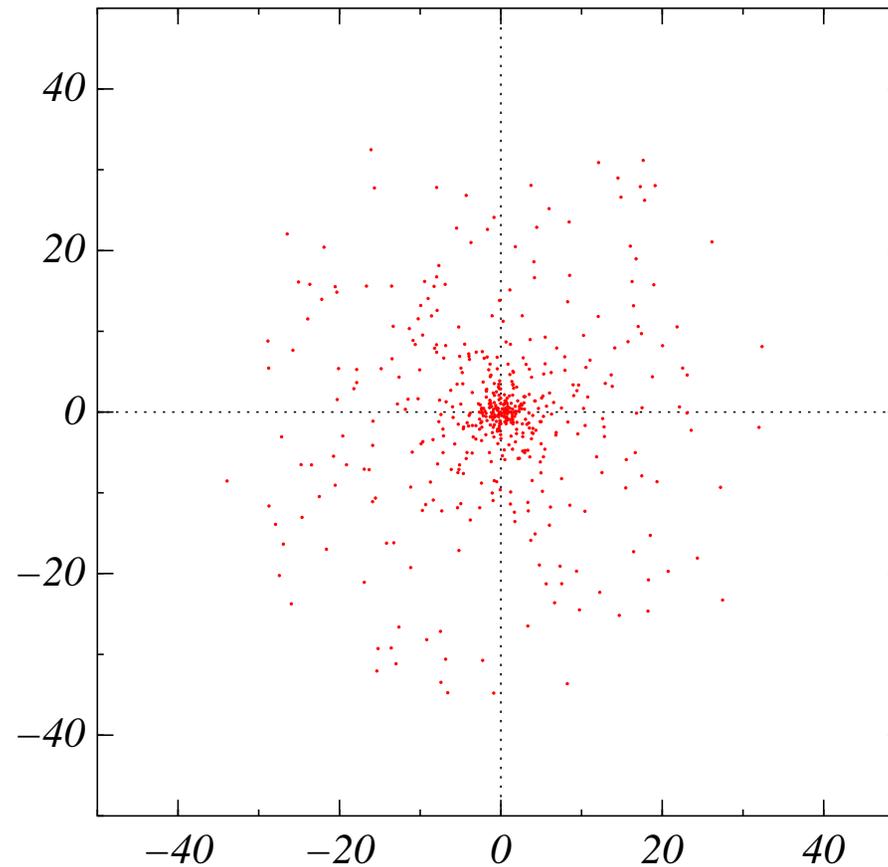
# Simulation results - unit-radius disk, $\rho = 0.1, 0.3$



$$\left[1 - e^{-\pi \lambda \rho^2}\right]^n \quad \text{simulation asymptotic}$$

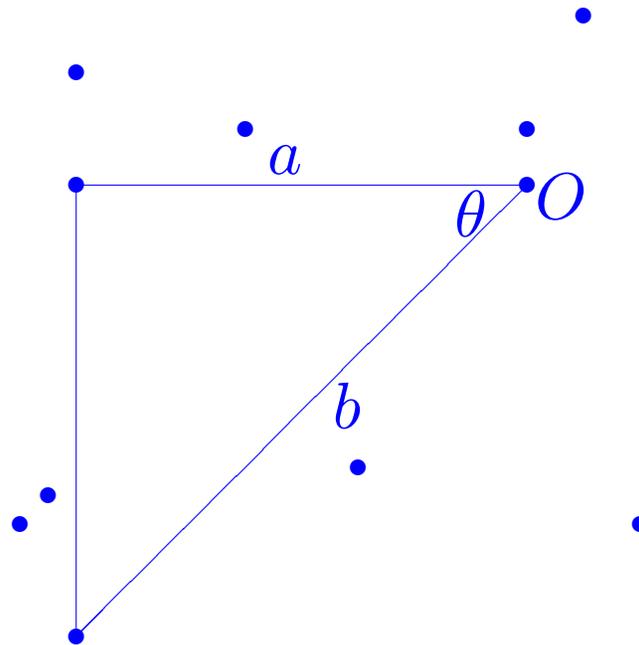
# Non-homogeneous Poisson process

Example: intensity falls off exponentially from an access point



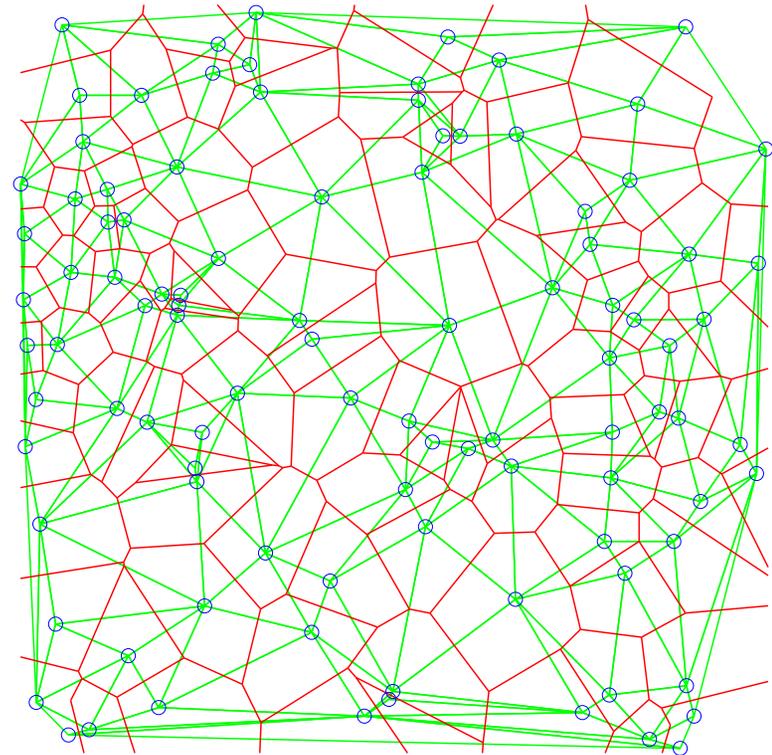
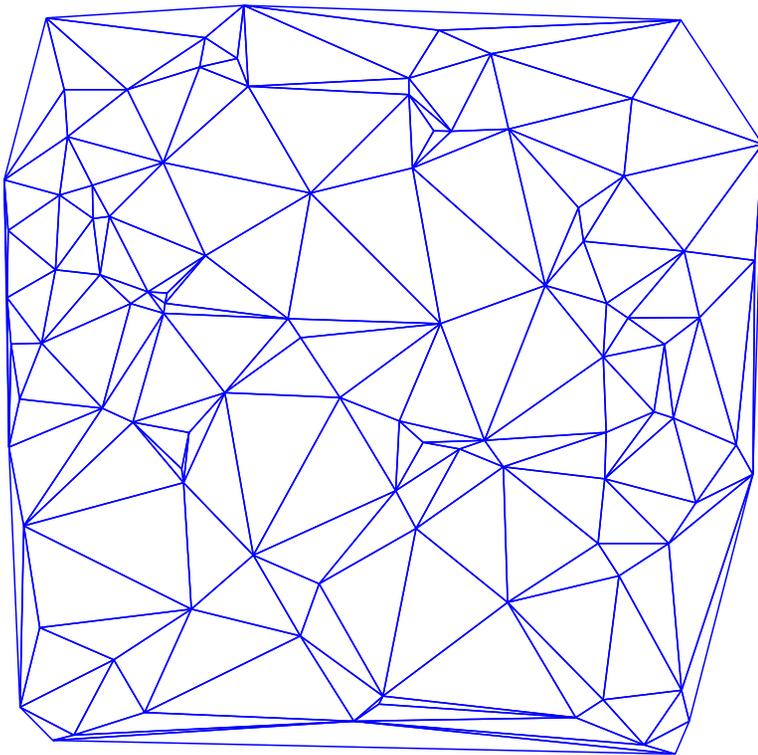
## Delaunay triangles 1 [kla90]

- Pick one node  $O$  from a planar Poisson process of intensity  $\lambda$
- Consider triangles formed by two other nodes
- Call it *empty* if no other nodes are in the triangle
- Call it *very empty* if no other nodes are in the circumcircle of the triangle
- An empty triangle:



## Delaunay triangles 2

The Delaunay triangulation consists of very empty triangles only. The second figure shows the Voronoi tessellation superimposed.



## Delaunay triangles 3

- Let  $a$  and  $b$  be the lengths of the two edges adjacent to  $O$  and  $\theta$  the angle
- For very empty triangles, the joint pdf is

$$2\pi ab\lambda^4 \exp \left[ -\pi\lambda^2 \frac{a^2 + b^2 - 2ab \cos \theta}{4 \sin^2 \theta} \right]$$

- For very empty triangles, the pdf of the area  $A$  is  $\lambda^2 A \exp(-\lambda A)$
- For very empty triangles, the mean of  $a$  is  $\frac{32}{9\pi\lambda}$
- For empty triangles, the pdf is  $2\pi ab\lambda^4 \exp[-\lambda^2 ab \sin(\theta)/2]$
- In both cases, the mean number of triangles at  $O$  is 6

## Delaunay triangles 4

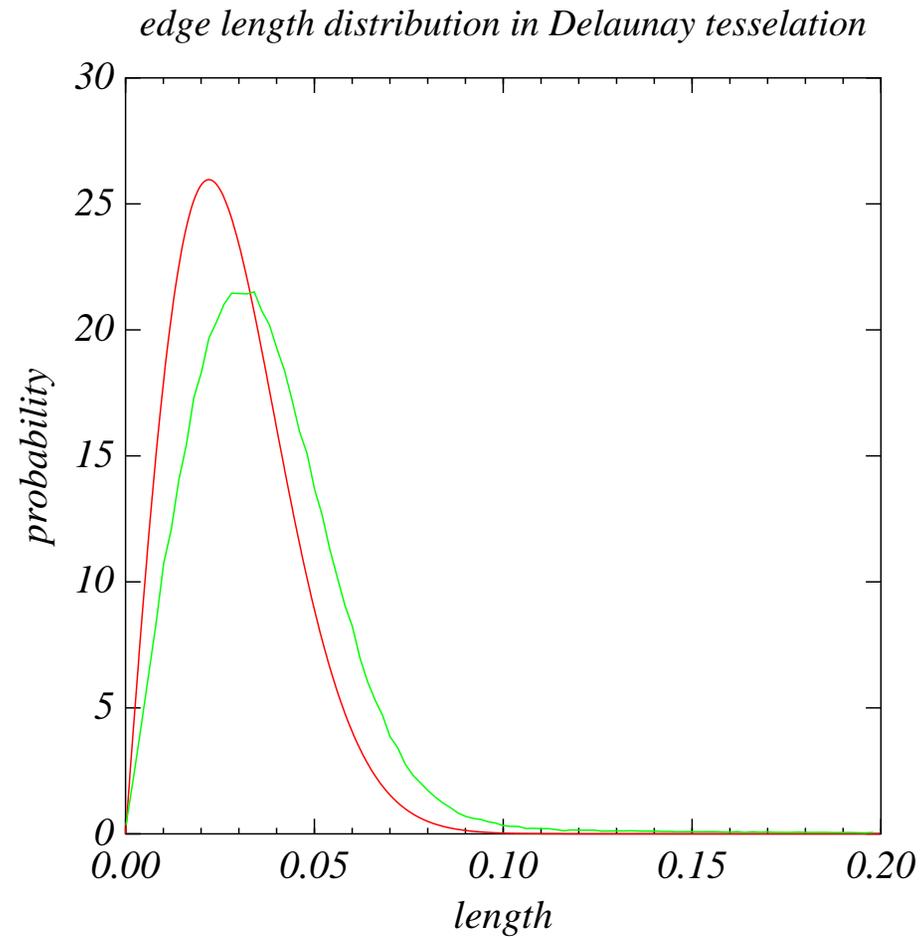
Integrating out over side  $b$  and angle  $\theta$ , we get for the pdf of side  $a$ :

$$4a\lambda \int_0^\pi \sin^2 \theta \exp \left[ \frac{-\pi a^2 \lambda}{4 \sin^2 \theta} \right] \left[ 1 + e^{\alpha^2 \nu^2} \alpha |\nu| \pi^{1/2} (\operatorname{erf}(\alpha |\nu|) + \operatorname{sign}(\nu)) \right] d\theta$$

where

$$\alpha = \frac{\sin \theta}{(\pi \lambda)^{1/2}}$$
$$\nu = \frac{a \lambda \cos \theta}{2 \sin^2 \theta}$$

# Delaunay triangles 5



1000 nodes: **exact** **simulation**

## Distance distribution for some regions

Two points independently uniformly distributed in a region  $R$ ; the pdf of the distance  $d$  is  $f(d)$ , mean distance is  $\mu$ :

- $R =$  unit interval,  $f(d) = 2(1-d)[[0 \leq d \leq 1]]$ ,  $\mu = 1/3$
- $R =$  1-torus,  $f(d) = 2[[0 \leq d \leq 1/2]]$ ,  $\mu = 1/4$
- $R =$  2-torus

$$f(d) = \begin{cases} 2\pi d & \text{if } 0 \leq d < 1/2 \\ 2d [\pi - 4 \sec^{-1}(2d)] & \text{if } 1/2 \leq d \leq \sqrt{2} \end{cases}$$

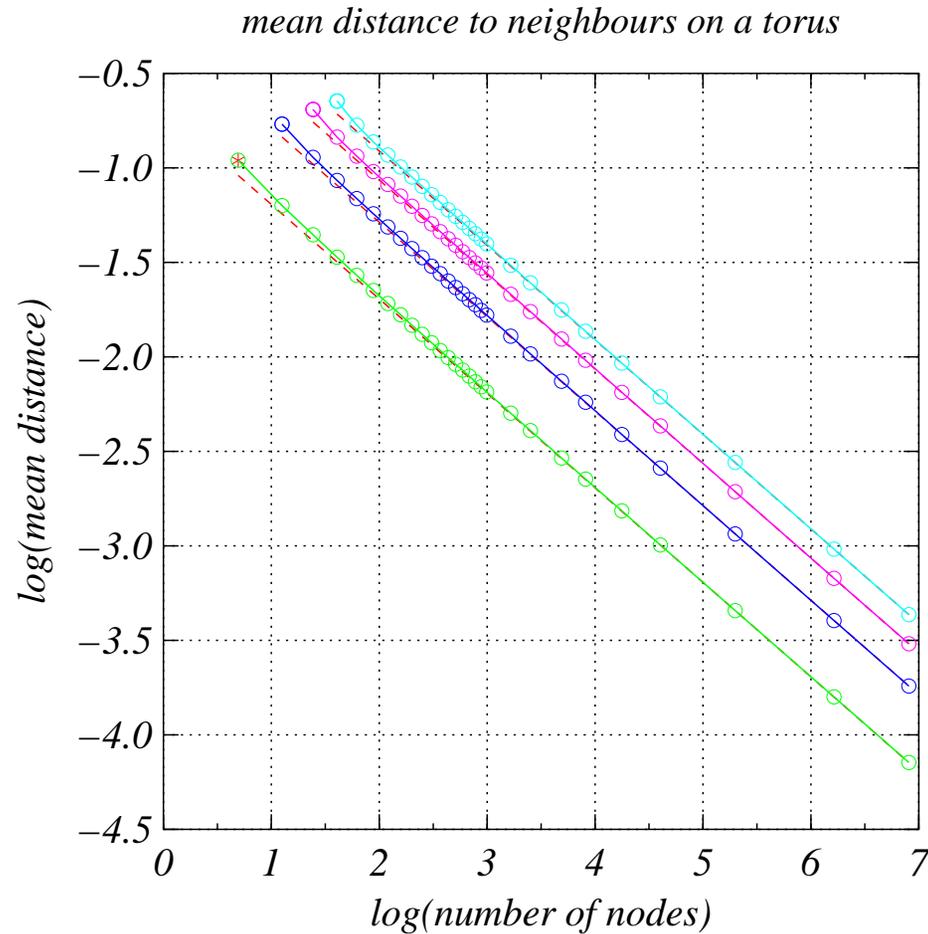
$$\mu = \left[ \sqrt{2} + \log(1 + \sqrt{2}) \right] / 6$$

- $R =$  unit radius disk,  $f(d) = d/\pi [4 \arctan(\sqrt{4-d^2}/d) - d\sqrt{4-d^2}] [[0 \leq d \leq 2]]$ ,  $\mu = 128/(45\pi)$
- $R =$  unit sphere,  $\mu = 36/35$
- $R =$  unit square, see next slide

# Asymptotics for near neighbours

- Put  $n$  points in a unit torus in  $\mathbb{R}^2$
- Let  $d_k =$  be the distance to  $k$ th nearest neighbour
- Then it is known that ([eva02]):  $\mathbb{E}[d_k] = \pi^{-1/2} \frac{\Gamma(k+1/2)}{\Gamma(k)} n^{-1/2} + \mathcal{O}(n^{-3/2})$
- So  $\mathbb{E}[d_1] = 1/2 n^{-1/2} + \mathcal{O}(n^{-3/2})$

# Asymptotics for nearest neighbours - simulations



nearest, second nearest, . . .

asymptotic, \* is exact value for  $n = 2, k = 1$ , namely  $[2^{1/2} + \log(1 + 2^{1/2})]/6$

# Exact theory for mean distances on a torus 1

Recall that our pdf and cdf are defined piecewise: I will use  $\langle$  and  $\rangle$  to indicate the pieces on  $[0, 1/2]$  and  $[1/2, 1/\sqrt{2}]$  respectively:

$$f^{\langle}(x) = 2\pi x$$

$$f^{\rangle}(x) = 2x \left[ \pi - 4 \sec^{-1}(2x) \right]$$

$$F^{\langle}(x) = \pi x^2$$

$$F^{\rangle}(x) = \sqrt{4x^2 - 1} + x^2 \left[ \pi - 4 \sec^{-1}(2x) \right]$$

## Exact theory for mean distances on a torus 2

I will use the subscript  $k:n$  to denote the  $k$ th order statistic in a sample of size  $n$

Thus

$$f_{k:n}^{<}(x) = \frac{\Gamma(n+1)}{\Gamma(k)\Gamma(n-k+1)} f^{<}(x) [F^{<}(x)]^{k-1} [1 - F^{<}(x)]^{n-k}$$

and similarly for  $f_{k:n}^{>}(x)$ .

So we have

$$f_{k:n}(x) = f_{k:n}^{<}(x) \llbracket 0 \leq x \leq 1/2 \rrbracket + f_{k:n}^{>}(x) \llbracket 1/2 < x \leq 1/\sqrt{2} \rrbracket$$

and

$$\mu_{k:n} = \mu_{k:n}^{<} + \mu_{k:n}^{>}, \quad 1 \leq k \leq n$$

## Exact theory for mean distances on a torus 3

To get the mean, we can do the lower integral exactly:

$$\begin{aligned}\mu_{k:n}^{\leq} &= \int_0^{1/2} t f_{k:n}^{\leq}(t) dt \\ &= \frac{(\pi/4)^k}{(2k+1)} \frac{\Gamma(n+1)}{\Gamma(k)\Gamma(n-k+1)} \mathbf{F}\left(\begin{matrix} k+1/2 & k-n \\ k+3/2 \end{matrix} \middle| \pi/4\right)\end{aligned}$$

but the upper integral

$$\mu_{k:n}^{\geq} = \int_{1/2}^{1/\sqrt{2}} t f_{k:n}^{\geq}(t) dt$$

will have to be approximated. Luckily, it is typically a very small correction term to  $\mu_{k:n}^{\leq}$ , and goes to zero geometrically with  $n$ .

## Exact theory for mean distances on a torus 4

Because  $k - n < 0$ , the hypergeometric function above is a terminating series:

$$F\left(\begin{matrix} k+1/2 & k-n \\ k+3/2 \end{matrix} \middle| \pi/4\right) = \sum_{i=0}^{n-k} \frac{(k+1/2)^{\bar{i}} (k-n)^{\bar{i}} (\pi/4)^i}{(k+3/2)^{\bar{i}} i!}$$

It is quite feasible to evaluate  $\mu_{k:n}$  exactly from this, but if desired we can use an integral representation of this function and Watson's lemma to find the large  $n$  asymptotics. I omit all the details of this. The results are on the next page.

## Exact theory for mean distances on a torus 5

We now do asymptotics ( $n \rightarrow \infty$ ) for the mean distance to the  $k$ th neighbour

$$\begin{aligned} \mu_{1:n}^{\leq} &\sim n \left[ \frac{1}{2} n^{-3/2} - \frac{3}{16} n^{-5/2} + \frac{25}{256} n^{-7/2} - \frac{105}{2048} n^{-9/2} + \dots \right] \\ \mu_{2:n}^{\leq} &\sim n^2 \left[ \frac{3}{4} (n-1)^{-5/2} - \frac{45}{32} (n-1)^{-7/2} + \frac{1155}{512} (n-1)^{-9/2} - \dots \right] \\ \mu_{3:n}^{\leq} &\sim n^3 \left[ \frac{15}{16} (n-2)^{-7/2} - \frac{525}{128} (n-2)^{-9/2} + \dots \right] \\ \mu_{4:n}^{\leq} &\sim n^4 \left[ \frac{35}{32} (n-3)^{-9/2} - \dots \right] \\ &\dots \\ \mu_{k:n}^{\leq} &\sim \Gamma(k+1/2)/\Gamma(k) n^{-1/2} \end{aligned}$$

Note: for the 2d Poisson process, we have  $\frac{1}{2} n^{-1/2}$  exactly for the nearest neighbour ( $k = 1$ )

## Exact theory for mean distances on a torus 6

To compute the contribution to the mean from the upper integral, we need to do:

$$\mu_{k:n}^> = \int_{1/2}^{1/\sqrt{2}} t f_{k:n}^>(t) dt$$

I do not know a way of approximating this for all  $n$  and  $k$ , but by making a series expansion of  $f_{1:n}^>$  around  $1/\sqrt{2}$  and just keeping the first term, for the nearest neighbour we get:

$$\mu_{1:n}^> \approx (3 - 2\sqrt{2})^n$$

Thus a good approximation for the mean distance to the nearest neighbour is

$$\mu_{1:n} = \mu_{1:n}^< + \mu_{1:n}^> \sim 1/2 n^{-1/2} - 3/16 n^{-3/2} + (3 - 2\sqrt{2})^n$$

## Almost sure connectivity results [mil70]

- Planar process of intensity  $\lambda$  in region  $R$
- Let  $p(r) = \Pr$  [every point of  $R$  covered by a disk radius  $r$ ]
- Then, as  $|R| \rightarrow \infty$

$$p(r) \sim \exp \left[ -\lambda |R| e^{-\pi \lambda r^2} (1 + \pi \lambda r^2) \right]$$

- This is plotted in **blue** on these graphs of simulation results

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