# Connectivity of random 1-dimensional networks 

Vitaliy Kurlin, Durham
Mila Mihaylova, Lancaster
Simon Maskell, QinetiQ
http://maths.dur.ac.uk/~dma0vk

## Initial motivations

- dimension 1: monitoring roads, boundaries of restricted areas
- random: automatic deployment along riversides difficult of access


## Distributing along roads

Transmission radius $R>$ road width $W$.
Then a 2-dim network of $\left(x_{i}, z_{i}\right)$ is connected ff the 1-dim network of $x_{i}$ with radius $\sqrt{R^{2}-W^{2}}$ is connected.


## Filling a 2-dimensional area

Distributing sensors along a snake-like path fills an area if the distance between adjacent branches $D \leq R \sqrt{3} / 2$.


## What is random?

- common: all (positions of) sensors have a prescribed density function
- practical: deploy sensors one by one along a trajectory of a vehicle,
so the distance between successive sensors has a prescribed density


## Our assumptions

- $R$ is a transmission radius
- sensors are deployed in $[0, L]$, a sink node is fixed at $x_{0}=0$
- $f_{1}, \ldots, f_{n}$ are independent densities of distances between sensors: $P\left(0 \leq x_{i}-x_{i-1} \leq R\right)=\int_{0}^{R} f_{i}(s) d s$.


## Connectivity and coverage

 For a given probability and densities- find a minimal number of randomly deployed sensors in $[0, L]$ such that the resulting network is connected;
- find a minimal number of random sensors such that the network is connected and covers $[0, L]$.


## Key steps of our solution

- For arbitrary densities $f_{1}, \ldots, f_{n}$, compute the probability $P_{n}$ that the network of $n$ sensors is connected.
- Find estimates of $n$ such that $P_{n}$ is greater than the given probability.


## Conditional probabilities

Given densities $f_{1}, \ldots, f_{n}$ of distances, $y_{1}, \ldots, y_{n}$ are naturally defined on $[0, L]$, but the network should be proper, i.e. all sensors are in $[0, L]$ or $\sum_{i=1}^{n} y_{i} \leq L$.

We compute the probability that the network is connected if it is proper.

## 2-sensor networks

A network of 2 sensors with distances
$y_{1}=x_{1}-0, y_{2}=x_{2}-x_{1}$ is represented by $\left(y_{1}, y_{2}\right) \in\left\{y_{1}, y_{2} \geq 0 y_{1}+y_{2} \leq L\right\}$.


## Simplest non-trivial case

## The probability of connectivity is



## Connectivity Theorem

The probability of connectivity is

$$
P_{n}=v_{n}(R, L) / v_{n}(L, L), \text { where }
$$

$$
v_{0}(r, l)=1, r, I>0 ;
$$

$$
v_{n}(r, I)=0, r \leq 0 \text { or } I \leq 0 ;
$$

$$
v_{n}(r, I)=1, r \geq I>0, n>0
$$

$$
v_{n}(r, I)=\int_{0}^{r} f_{n}(s) v_{n-1}(r, I-s) d s, r<l
$$

# $P_{n}=v_{n}(R, L) / v_{n}(L, L)$ 

+ closed formula for finite networks
+ arbitrary different densities
- can be computationally difficult
+ explicit for important distributions
+ implies simple estimates for $n$


## The recursive function

$v_{n}(r, l)$ is the probability that random distances having densities $f_{1}, \ldots, f_{n}$
satisfy $\sum_{i=1}^{n} y_{i} \leq l$ and $0 \leq y_{i} \leq r$, e.g.
$v_{1}(r, I)=\int_{0}^{r} f_{1}(s) d s, r<l$,
$v_{2}(r, I)=\int_{0}^{r} f_{2}(s) v_{1}(r, I-s) d s$.
$v_{n}(L, L)$ : the network is proper,
$v_{n}(R, L)$ : the network is connected.

## Coverage Theorem

The probability of coverage is

$$
\left(v_{n}(R, L)-v_{n}(R, L-R)\right) / v_{n}(L, L) .
$$

$\frac{v_{n}(R, L)}{v_{n}(L, L)}$ : connected if proper on $[0, L]$,
$v_{n}(R, L-R) / v_{n}(L, L)$ : connected network if proper on $[0, L-R]$.

## Uniform Corollary

If all $f_{i}=1 / L$ then the probability is

$$
P_{n}^{u}=\sum_{i=0}^{i<L / R}(-1)^{i}\binom{n}{i}(1-i R / L)^{n} .
$$

$P_{1}^{u}=R / L$ : connected with the sink.

$$
P_{2}^{u}= \begin{cases}2(R / L)^{2} & \text { if } R \leq L / 2 \\ 4(R / L)-2(R / L)^{2}-1 & \text { if } R \geq L / 2\end{cases}
$$

## Uniform case: simulations

$L=1 \mathrm{~km}, R=50 \mathrm{~m}, n \leq 200$ sensors.


## Uniform case: estimate

Set $Q=(L / R)-1$. The network is connected with a probability $p>2 / 3$ if
$n \geq \frac{3}{2}(1-Q)+\sqrt{\frac{(3 Q-1)^{2}}{4}+6 Q^{2}\left(\frac{Q}{1-p}-1\right)}$.

| Transmission Radius, m. | 200 | 100 | 50 | 25 |
| :--- | :---: | :---: | :---: | :---: |
| Min Number of Sensors | 29 | 69 | 157 | 349 |
| Estimate of Min Number | 83 | 283 | 905 | 2610 |

# Uniform case: conclusions 

- less effective than non-random - rough estimate, not optimal + quadratic estimate is used later + can be improved using Taylor approximations of degrees 4,5 + non-trivial inequalities $0 \leq P_{n}^{u} \leq 1$


## A constant density: graph

 Let $f=1 /(b-a)$ over $[a, b] \subset[0, L]$. $\uparrow f(l)$$$
n=1: P\left(0 \leq y_{1} \leq R\right)=(R-a) /(b-a)
$$

## Constant Corollary

If all $f_{i}=1 /(b-a)$ then the probability is

$$
\begin{aligned}
& P_{n}^{c}=\frac{\sum_{k=0}^{n}(-1)^{k}\binom{n}{k}(L-a(n-k)-R k)^{n}}{\sum_{k=0}^{n}(-1)^{k}\binom{n}{k}(L-a(n-k)-b k)^{n}} . \\
& P_{1}^{c}=\frac{(L-a)-(L-R)}{(L-a)-(L-b)}=\frac{R-a}{b-a} .
\end{aligned}
$$

## Constant case: simulations

$$
L=1 \mathrm{~km}, R=50 \mathrm{~m}, a=10 \mathrm{~m}, b=80 \mathrm{~m} .
$$



## Constant case: estimate

The network is connected with
a probability $p$ if $\frac{a+b}{2} \leq R \leq b$ and
$n \geq \max \left\{\frac{3}{2}+\sqrt{\frac{1+5 p}{1-p}}, 1+\frac{L-b}{a}\right\}$.
For all $p$ not too close to 1 , the 2 nd estimate holds: $L+a-b \leq a n<L$.

## Constant case: conclusions

Constant density over [0.2R, 1.6R].

| Transmission Radius, m. | 200 | 150 | 100 | 50 | 25 |
| :--- | :---: | :---: | :---: | :---: | :---: |
| Min Number of Sensors | 14 | 19 | 30 | 63 | 132 |
| Estimate of Min Number | 18 | 27 | 43 | 93 | 193 |
| Max Number of Sensors | 25 | 34 | 50 | 100 | 200 |

+ minimal practical assumptions + very simple effective estimate + non-trivial inequalities $0 \leq P_{n}^{c} \leq 1$


## Exponential Corollary

If the distances between successive sensors have the density $f(s)=c e^{-\lambda s}$ on $[0, L]$, then the probability of
connectivity is $P_{n}^{e}=\frac{v_{n}(R, L)}{v_{n}(L, L)}, v_{n}(r, I)=$
$\sum_{i=0}^{i<1 / r}(-1)^{i}\binom{n}{i} \frac{e^{-i \lambda r}}{\lambda^{n}}\left(1-e^{-\lambda(I-i r)} \sum_{j=0}^{n-1} \frac{\lambda^{j}(I-i r)^{j}}{j!}\right)$

## Exponential conclusions

Estimate: as in the uniform case.
The denominator tends to 0 fast:
$v_{n}(L, L)=1-e^{-\lambda L} \sum_{j=0}^{n-1}(\lambda L)^{j} / j!$

- unpractical: throw on the alert
- sensors are too close to each other


## Normal distribution

If $f(s)=\frac{C}{\sigma \sqrt{2 \pi}} e^{-(s-\mu)^{2} / 2 \sigma^{2}}$ on $[0, L]$
then the distances between successive sensors are close to the mean $\mu$, e.g. very likely to be in $[\mu-3 \sigma, \mu+3 \sigma$ ]

Reasonable to assume: $\mu<R, n \mu<L$.

## Normal case: estimate

The network with normal distances is connected with a given probability $p$ if

$$
\begin{aligned}
& n \leq \min \left\{\frac{p(1-p)}{\varepsilon}, \frac{\left(\sqrt{4 \mu L+\sigma^{2} \phi^{-2}(p)}-\sigma \Phi^{-1}(p)\right)^{2}}{4 \mu^{2}}\right\} \\
& \Phi(x)=\frac{1}{\sqrt{2 \pi}} \int_{-\infty}^{x} e^{-s^{2} / 2} d s, \varepsilon=\Phi\left(-\frac{\mu}{\sigma}\right)+1-\Phi\left(\frac{R-\mu}{\sigma}\right) .
\end{aligned}
$$

## Normal case: example

 $\mu=0.6 R, \sigma=0.1 R, p=0.9975$.Then $\Phi^{-1}(p) \approx 2.8, \varepsilon \approx 0.000063$
$R=25 \mathrm{~m}: n \leq p(1-p) / \varepsilon \approx 40$.
$R \geq 50 \mathrm{~m}$ : the 2nd estimate is close to
$L / \mu \approx \frac{\left(\sqrt{4 \mu L+\sigma^{2} \Phi^{-2}(p)}-\sigma \Phi^{-1}(p)\right)^{2}}{4 \mu^{2}}$.

## Normal case: table

$$
\text { Let } L=1 \mathrm{~km}, \mu=0.6 R, \sigma=0.1 R \text {. }
$$

| Transmission Radius, m. | 200 | 150 | 100 | 50 | 25 |
| :--- | :---: | :---: | :---: | :---: | :---: |
| Estimate of Max Number | 7 | 11 | 16 | 33 | 40 |

6 non-random sensors are enough for the radius $R=150 \mathrm{~m}: 6 / 11 \approx \mu / R$.

## All cases: conclusions

- exponential: too dense networks
- normal: ideal density $\Rightarrow$ ideal results
- uniform: a useful theoretical exercise
+ constant over [a, b]: very reasonable
+ more complicated: piecewise constant?


## Ideas of proofs

- induction on the number of sensors: adding 1 sensor keeps connectivity if it is close to the previous one
- the key probability $v_{n}(r, I)$ is
an iterated convolution of densities computed by the Laplace transform


## More explicit formulae

- heterogeneous networks: distances have different constant densities
- building densities from blocks: any piecewise constant density
- more can be produced easily


## A 3-step density: graph

$\uparrow f(l)$

$C, R$ are chosen so that $\int_{0}^{L} f(s) d s=1$.

## A 3-step density: formula

The probability of connectivity is $P_{n}=$

$$
\frac{\sum_{m=0}^{n} \sum_{k_{1}=0}^{m} \sum_{k_{2}=0}^{n-m} \frac{(-1)^{k_{1}+k_{2}}\left(L-\left(2 k_{1}+k_{2}+n-m\right) R / 2\right)^{n}}{d_{m} k_{1}!\left(m-k_{1}\right)!k_{2}!\left(n-m-k_{2}\right)!}}{\sum_{m=0}^{n} \sum_{k_{1}=0}^{m} \sum_{k_{2}=0}^{n-m} \frac{(-1)^{k_{1}+k_{2}}\left(L-\left(2 k_{1}+2 k_{2}+n-m\right) R / 2\right)^{n}}{d_{m} k_{1}!\left(m-k_{1}\right)!k_{2}!\left(n-m-k_{2}\right)!}}
$$

$$
d_{m}=C^{-m}(1 / R-C)^{m-n}, \text { the sums are }
$$

over all $m, k_{1}, k_{2}$ if the terms $>0$.

## A 3-step density: simulations

 Let $L=1 \mathrm{~km}, R=50 \mathrm{~m}, C=0.9 / R$.

## A 3-step density: table

| Transmission Radius, m. | 250 | 200 | 150 | 100 | 50 |
| :--- | :---: | :---: | :---: | :---: | :---: |
| Min Number of Sensors | 12 | 17 | 25 | 44 | 105 |

+ flexible practical assumptions
+ reasonable estimates of min number
+ non-trivial inequalities $0 \leq P_{n} \leq 1$


## Open problem 1

Compute the exact probability of connectivity if the distances between successive sensors have a truncated normal density on $[0, L]$.

## Open problem 2

For a given segment $[0, L]$ and number $n$ of sensors, find an optimal density of distances between successive sensors to maximise the probabilities of connectivity and coverage.

## Open problem 3

Compute the probabilities of connectivity and coverage if sensors are randomly deployed along a non-straight trajectory filling a 2 -dimensional area.

