

# Entropy of Random Geometric Graphs

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*with thanks to N. Warsi, O. Georgiou and C. P. Dettmann*

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# Random Geometric Graphs and Spatially Embedded Networks

The model...

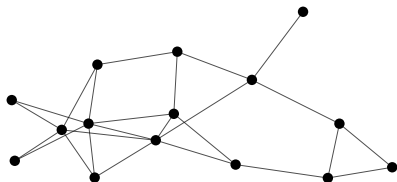
- ▶ Node positions are random
- ▶ Existence of an edge  $(i, j)$  depends on distance between nodes  $i$  and  $j$
- ▶ Notion of distance implies embedding in some space (e.g., Euclidean)

Applications...

- ▶ Human/animal interactions
- ▶ Nano processes (e.g., carbon nanotubes on a polymer substrate)
- ▶ Wireless communication networks

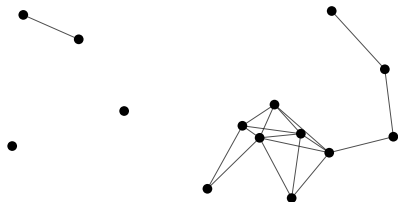
# Erdős-Rényi Graph vs Random Geometric Graph

ER Graphs



- ▶ Edge probability  $\varphi$
- ▶ Independent of embedding

RG Graphs



- ▶ Edge probability  $\varphi(s_{i,j})$
- ▶ Distribution of  $s_{i,j}$  important

# Graph Entropy

Consider a graph  $G = (\mathcal{V}, \mathcal{E})$  to be a set of vertices (nodes) and a set of edges (links).

- ▶  $\binom{n}{2} = n(n-1)/2$  possible edge configurations
- ▶  $2^{n(n-1)/2}$  possible graphs without considering node locations

**The entropy of  $G$  is a measure of disorder or the amount of information contained in the graph distribution.**

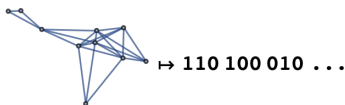
$$H(G) = \mathbb{E}[-\log(\mathbb{P}(G))]$$

\* *Logarithms are base  $e$ .*

# ER Graph Entropy

For an ER graph, all edges occur independently with probability  $\varphi$ .  
To calculate entropy...

- ▶ Map  $G$  to  $n(n-1)/2$  Bernoulli variables  $X_1, \dots, X_{n(n-1)/2}$



- ▶ Independence of  $\{X_i\}$  implies

$$H(G) = H(X_1, \dots, X_{n(n-1)/2}) = \sum_i H(X_i) = \binom{n}{2} H(\varphi)$$

- ▶ Entropy of single edge is

$$H(\varphi) = -\varphi \log \varphi - (1 - \varphi) \log(1 - \varphi)$$

# ER Graph Entropy

## Examples

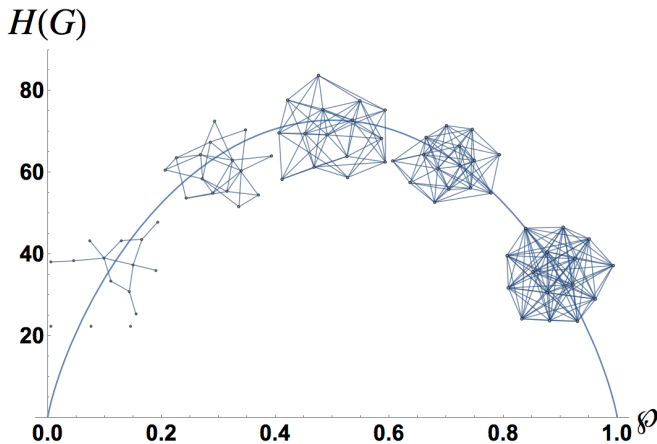


Figure : Entropy of ER graph with  $n = 15$  nodes.

# RG Graph Entropy

For an RG graph, edge  $(i, j)$  occurs with probability  $\varphi_{i,j}$ , which depends on the distance  $s_{i,j} = \|\mathbf{r}_i - \mathbf{r}_j\|$ . Consider *conditional entropy*...

$$\begin{aligned} H(G|S) &= \mathbb{E}[H(G|s_{1,2}, \dots, s_{n-1,n})] \\ &\leq \sum_{i < j} \mathbb{E}[H(X_{i,j}|s_{i,j})] && \text{(independence bound)} \\ &= \binom{n}{2} \mathbb{E}[H(X|s)] \\ &\leq \binom{n}{2} H(\bar{\varphi}) && \text{(Jensen's inequality)} \end{aligned}$$

where

$$\bar{\varphi} := \mathbb{E}[\varphi(s_{i,j})].$$

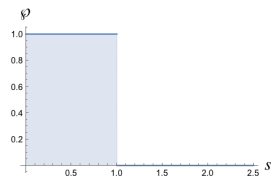
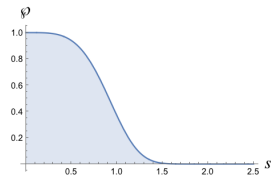
# Pair Connection Function

Interesting properties arise from the **distribution of the pair distance** and the **pair connection function**. Let...

$$\wp(s) = \exp(-(s/s_0)^\eta)$$

- ▶ Models Rayleigh fading in wireless networks
- ▶ Versatility offered by parameter  $\eta$ : probabilistic to deterministic

$$\Rightarrow \bar{\wp} = \int_0^D f(s) \exp(-(s/s_0)^\eta) ds$$





# Uniform Node Distributions in a Compact Domain

In  $d$  dimensions, the **set covariance** of a convex set  $\mathcal{K}$  is given by

$$c_{\mathcal{K}}(\mathbf{s}) = \int_{\mathbb{R}^d} \mathbf{1}_{\mathcal{K}}(\mathbf{x})\mathbf{1}_{\mathcal{K}}(\mathbf{x} - \mathbf{s}) d\mathbf{x}, \quad \mathbf{s} \in \mathbb{R}^d$$

where  $\mathbf{1}_{\mathcal{K}}(\mathbf{x})$  is the indicator function for  $\mathbf{x} \in \mathcal{K}$ .

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The **isotropised set covariance** is given by

$$\bar{c}_{\mathcal{K}}(s) = \int_{\mathcal{S}^{d-1}} c_{\mathcal{K}}(s\mathbf{u}) \, d\mathbf{u}, \quad s \geq 0$$

where  $\mathbf{u}$  is a vector denoting a point on the unit sphere  $\mathcal{S}^{d-1}$ .

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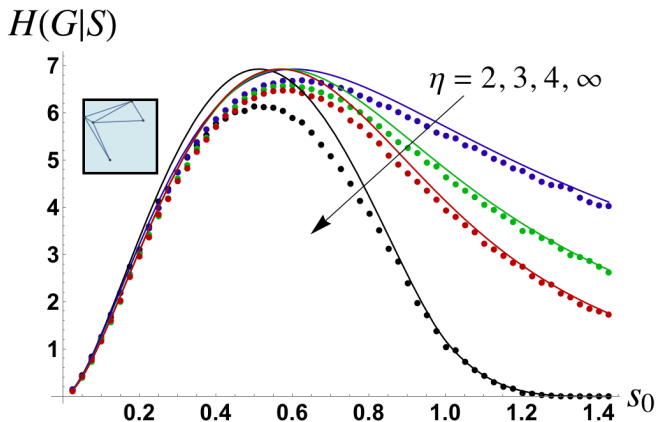
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The **pair distance** probability density function is given by

$$f(s) = \frac{2\pi^{d/2} s^{d-1} \bar{c}_{\mathcal{K}}(s)}{\Gamma(d/2) \text{vol}(\mathcal{K})^2}.$$

# Entropy vs Connection Range $s_0$

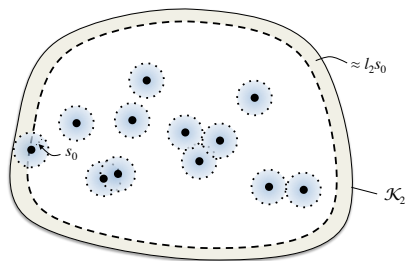


**Figure :** Entropy of RG graph with  $n = 5$  nodes and pair connection functions with typical range  $s_0$  for a unit square. Solid line: upper bound. Markers: numerical simulations.

# Uniform Node Distributions in a 2D Compact Domain

## Small Typical Connection Range

Consider a compact domain  $\mathcal{K}_2$  with area  $a_2$ , boundary length  $l_2$  and small typical connection range  $s_0 \ll \pi a_2 / l_2 \ll D$ .



- ▶ Pair distance distribution
- ▶ Soft pair connection probability
- ▶ Hard pair connection probability

$$f(s) = \frac{2\pi s}{a_2} \left(1 - \frac{l_2 s}{\pi a_2}\right)$$

$$\bar{\rho} = \frac{2\pi s_0^2 \Gamma(2/\eta)}{\eta a_2} \left(1 - \frac{l_2 s_0 \Gamma(3/\eta)}{\pi a_2 \Gamma(2/\eta)}\right)$$

$$\bar{\rho} = \frac{\pi s_0^2}{a_2} \left(1 - \frac{2l_2 s_0}{3\pi a_2}\right)$$

## Entropy for Small $s_0$

Let  $u_2 = 2\pi\Gamma(2/\eta)/(\eta a_2)$  denote the *fractional area of a unit soft disc*. For  $s_0 \rightarrow 0$ , we have

$$H(G|S) \leq \binom{n}{2} \left( 2u_2 \log \left( \frac{1}{s_0} \right) s_0^2 + u_2(1 - \log u_2)s_0^2 + O(s_0^3 \log s_0) \right)$$

- ▶ For fixed  $s_0$ , the bound increases with  $n$ .
- ▶ For fixed  $n$ , the bound decreases with  $s_0$  (toward a completely disconnected graph).

# Entropy for Small $s_0$

What about  $s_0 = g(n)$ ?

## Fact

*For typical connection distances that decay according to the relation*

$$s_0^2 \log \left( \frac{1}{s_0} \right) = o \left( \frac{1}{n^2} \right)$$

*the entropy  $H(G|S) \rightarrow 0$  as  $n \rightarrow \infty$ .*

# An Entropy Limit for Small $s_0$

## Fact

*The upper bound on the entropy of a graph in  $\mathcal{K}_2$  will tend to a limit  $\nu > 0$  as  $n \rightarrow \infty$  if*

$$\begin{aligned} s_0(n) &= \exp\left(\frac{1}{2}W_m\left(-\frac{2\nu}{u_2 n^2}\right)\right) \\ &= \frac{\sqrt{\nu}}{n\sqrt{u_2 \log n}}\left(1 + O\left(\frac{1}{\log n}\right)\right) \end{aligned}$$

*where  $W_m(x)$  is the lower branch ( $-1/e \leq x < 0$  and  $W_m \leq -1$ ) of the solution to  $x = W \exp W$ .*



# Entropy vs Number of Nodes $n$

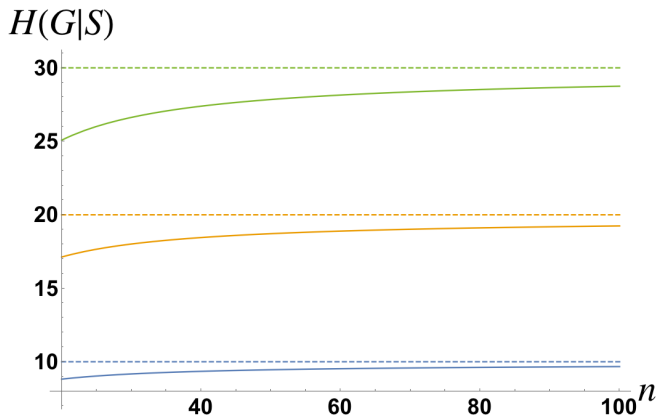


Figure : Upper bound on the entropy of an RG graph in a unit square with  $s_0 = \exp\left(\frac{1}{2} W_m \left(-\frac{2\nu}{u_2 n^2}\right)\right)$ . Solid line: theory. Dashed line: limit.

## Connectivity for Small $s_0$

### Fact

*The probability that a graph in  $\mathcal{K}_2$  is completely disconnected is well approximated by*

$$P = (1 - \bar{\rho})^{n(n-1)/2}.$$

*Thus, for  $s_0 \rightarrow 0$  and  $n \rightarrow \infty$ , a graph is almost surely completely disconnected if*

$$s_0 = o\left(\frac{1}{n}\right).$$

*The probability of a completely disconnected graph will tend to a limit  $\varphi \in (0, 1)$  if*

$$s_0(n) = \frac{1}{n} \sqrt{\frac{2}{u_2} \log\left(\frac{1}{\varphi}\right)}$$

# Connectivity vs Entropy for Small $s_0$

## Interesting Fact

*Letting  $s_0$  tend to zero such that the upper bound on  $H(G|S)$  tends to  $\nu > 0$  yields*

$$P \sim 1 - \frac{\nu}{2 \log n}.$$

*I.e., typical connection distances that yield a positive limit on the entropy bound result in almost surely completely disconnected graphs (albeit with a slow convergence).*

## Arbitrary Node Configurations for $s_0 \gg D$

For the **hard disc model**, the graph  $G$  is complete, so  $H(G|S) = H(G) = 0$ .

For the **soft model** (i.e.,  $\eta < \infty$ )

$$\bar{\rho} = 1 - \frac{\mathbb{E}[s^\eta]}{s_0^\eta} + O\left(\frac{1}{s_0^{2\eta}}\right)$$

provided the  $\eta$ th,  $2\eta$ th,  $\dots$  moments exist. The conditional entropy is thus bounded by

$$H(G|S) \leq \binom{n}{2} \frac{\eta \mathbb{E}[s^\eta] \log s_0}{s_0^\eta} + O\left(\frac{1}{s_0^\eta}\right), \quad \eta < \infty$$

## Room for Improvement...

- ▶ Consider  $d > 2$ ; all small  $s_0$  results generalise
- ▶ More to consider for  $s_0 \gg D$
- ▶ Exact calculations (difficult)
- ▶ Strengthen the bound
- ▶ Consider other pair connection functions
- ▶ Explicit calculations for specific bounding geometries
- ▶ Maximum entropy points
- ▶ Structural entropy
- ▶ Applications (e.g., routing/forwarding tables, protocol design)