

# Random graphs, forest fires and self-organised criticality

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## Erdős-Rényi random graph

Random graph  $G(n, p)$  on  $n$  vertices: each edge is present independently with probability  $p$ .

Let  $C_k^n$  be the size of the  $k$ th largest component.

**Phase transition:** let  $p = c/n$  where  $c > 0$ .

$c < 1$  :  $C_1^n$  on order  $\log n$

$c = 1$  :  $C_1^n$  on order  $n^{2/3}$

$c > 1$  :  $C_1^n \sim \theta(c)n$  for some  $\theta(c) > 0$ .

## Erdős-Rényi random graph process

Random graph **process**: start with empty graph on  $n$  vertices at time  $t = 0$ . Each absent edge arrives at rate  $1/n$ .

State at time  $t$  has distribution  $G(n, p)$  where  $p = 1 - e^{-t/n} \approx t/n$ .

Let  $C_k^n(t)$  be the size of the  $k$ th largest component at time  $t$ .

$t < 1$  :  $C_1^n(t)$  on order  $\log n$

$t = 1$  :  $C_1^n(t)$  on order  $n^{2/3}$

$t > 1$  :  $C_1^n(t) \sim \theta(t)n$ .

“multiplicative coalescence”: blocks size  $a, b$  merge at rate  $\propto ab$ .

Define  $v_k^n(t) = \frac{\# \text{ vertices in components of size } k \text{ at time } t}{n}$ .

Then  $v_k^n(t) \xrightarrow{d} v_k(t)$  as  $n \rightarrow \infty$  for each  $k$  and each  $t$ , where

$$v_k(t) = \frac{k^{k-1}}{k!} e^{-kt} t^{k-1}.$$

$t < 1$  :  $v_k(t)$  decays exponentially,  $\sum_k v_k(t) = 1$ .

$t = 1$  :  $v_k(t)$  on order  $k^{-3/2}$  as  $k \rightarrow \infty$ .

$t > 1$  :  $v_k(t)$  decays exponentially,  $\sum_k v_k(t) = 1 - \theta(t) < 1$ .

(simple approximation by a **branching process** with Poisson( $t$ ) offspring distribution)

## Mean-field forest-fire model (Ráth and Tóth 2009):

Start with empty graph on  $n$  vertices. Each absent edge arrives at rate  $1/n$ . In addition, each vertex is **struck by lightning** at rate  $\lambda(n)$ . When lightning strikes a vertex, **remove all the edges** in its component.

## Mean-field “frozen percolation” (Ráth 2009):

Start with empty graph on  $n$  vertices. Each absent edge arrives at rate  $1/n$ . In addition, each vertex is **struck by lightning** at rate  $\lambda(n)$ . When lightning strikes a vertex, remove **all the edges AND all the vertices** in its component.

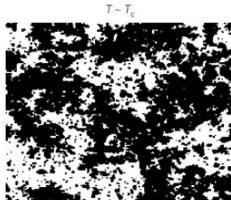
The frozen percolation model has a useful distributional property. At any time  $t$ , conditional on the graph having  $m$  remaining vertices, its distribution is that of  $G(m, 1 - e^{-t/n}) \approx G(m, t/n)$ .

Particularly interesting cases: when  $1/n \ll \lambda(n) \ll 1$  as  $n \rightarrow \infty$ . Then we see **“self-organised criticality”**.

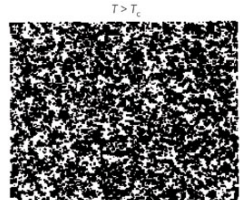
## Criticality: Ising model



Subcritical

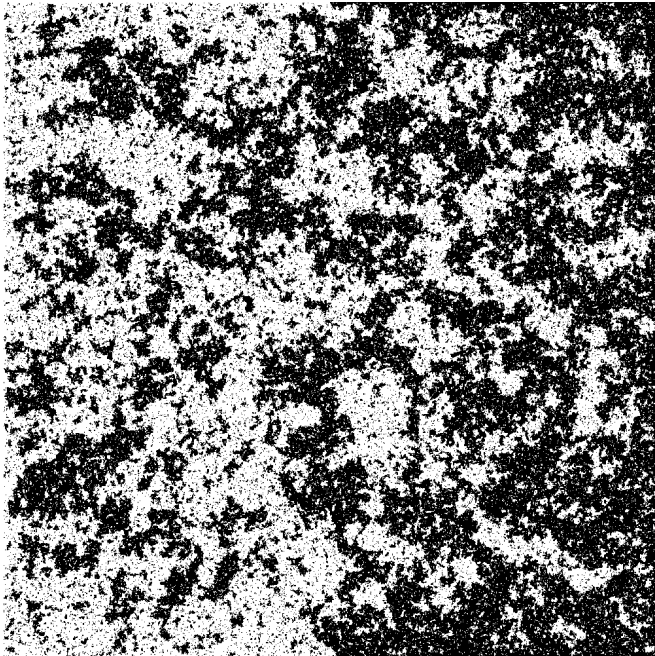


Critical



Supercritical

Critical Ising simulation ( $1000 \times 1000$ ):





$$1/n \ll \lambda(n) \ll 1.$$

$$v_k^n(t) = \frac{\# \text{ vertices in components of size } k \text{ at time } t}{n}.$$

**Frozen percolation:**  $v_k^n(t) \xrightarrow{d} v_k(t)$  as  $n \rightarrow \infty$  for each  $k$  and each  $t$ , where

$$v_k(t) = \begin{cases} \frac{k^{k-1}}{k!} e^{-kt} t^{k-1}, & t \leq 1 \\ \frac{1}{t} v_k(1), & t > 1. \end{cases}$$

(Rath 2009)

$$1/n \ll \lambda(n) \ll 1.$$

$$v_k^n(t) = \frac{\# \text{ vertices in components of size } k \text{ at time } t}{n}.$$

**Forest fire:**  $v_k^n(t) \xrightarrow{d} v_k(t)$  as  $n \rightarrow \infty$  for each  $k$  and each  $t$ , where

$$t \leq 1 : v_k(t) = \frac{k^{k-1}}{k!} e^{-kt} t^{k-1}$$

$$t > 1 : \sum_{l=k}^{\infty} v_l(t) \sim c(t) k^{-1/2}.$$

We expect that as  $t \rightarrow \infty$ ,

$$v_k(t) \rightarrow v_k(\infty) = 2 \binom{2k-2}{k-1} \frac{4^{-k}}{k}.$$

$v_k(\infty)$  is of order  $k^{-3/2}$  as  $k \rightarrow \infty$ , and corresponds to the distribution of the number of leaves of a critical binary branching process. (Rath and Toth 2009)

## Erdős-Rényi process: scaling window

Let  $C_k^n(t)$  be the size of the  $k$ th largest component at time  $t$ .

$t < 1$  :  $C_1^n(t)$  on order  $\log n$

$t = 1$  :  $C_1^n(t), \dots, C_k^n(t)$  all on order  $n^{2/3}$

$t > 1$  :  $C_1^n(t) \sim \theta(t)n$ ,  $C_2^n(t)$  on order  $\log t$

**Scaling window** of width order  $n^{-1/3}$ .

If  $t = 1 + u_n n^{-1/3}$ , then

if  $u_n \rightarrow -\infty$  :  $\frac{C_1^n(t)}{n^{2/3}} \xrightarrow{d} 0$

if  $u_n \rightarrow u \in (-\infty, \infty)$  :  $\frac{C_1^n(t)}{n^{2/3}}$  converges in distribution to a non-trivial limit.

if  $u_n \rightarrow \infty$  :  $\frac{C_1^n(t)}{n^{2/3}} \xrightarrow{d} \infty$ .

## Scaling window: convergence of the process to the multiplicative coalescent

Under appropriate rescaling, the evolution of large components of the random graph in the “scaling window” around  $t = 1$  converges as  $n \rightarrow \infty$ :

$$\left( n^{-2/3} C_1^n \left( 1 + un^{-1/3} \right), n^{-2/3} C_2^n \left( 1 + un^{-1/3} \right), \dots \right)_{u \in \mathbb{R}} \\ \implies (\mathcal{X}_1(u), \mathcal{X}_2(u), \dots)_{u \in \mathbb{R}}.$$

$\mathcal{X}(u)$  is a version of the *multiplicative coalescent*, by which we mean a random process with state space

$$\ell_2^\downarrow = \{x_1, x_2, \dots : x_1 \geq x_2 \geq \dots \geq 0, \sum x_i^2 < \infty\}$$

in which any pair of blocks of size  $a$  and  $b$  merge at rate  $ab$  to form a block of size  $a + b$ .

This version is called the *standard multiplicative coalescent*.

## Aldous's description of a state of the standard multiplicative coalescent

Let  $B(s)$ ,  $s \geq 0$  be a standard Brownian motion. Define

$$W^u(s) = B(s) + us - \frac{s^2}{2},$$

a Brownian motion with drift  $u - s$  at time  $s$ .

### Theorem

*Let  $u \in \mathbb{R}$ . The excursions of  $W^u(s)$ ,  $s \geq 0$  above its past minimum have the same distribution as the sizes of the blocks of  $\mathcal{X}(u)$ , put into size-biased order.*

**Exploration process** to find the components of a graph.

Choose  $v_1$  uniformly at random in the vertex set  $V$ .

Let  $v_2, \dots, v_k$  be the neighbours of  $v_1$ .

Let  $v_{k+1}, \dots, v_{k+m}$  be the neighbours of  $v_2$  in  $V \setminus \{v_1, \dots, v_k\}$ .

And so on, continuing until we “run out of vertices” – we have finished exploring one whole component, of size  $r$  say. Then choose the next vertex  $v_{r+1}$  uniformly at random among those remaining, and continue as before.

Let  $X_i = \#$  neighbours of  $v_i$  which are not neighbours of any vertex  $v_1, \dots, v_{i-1}$ .

Consider the random walk with step sizes

$$X_i - 1 \in \{-1, 0, 1, 2 \dots\}.$$

- ▶ The size of the component of  $v_1$  is the time until the random walk hits  $-1$ .
- ▶ More generally, the component sizes of the graph are the times between successive minima in the random walk.

The components appear in **size-biased order** in the exploration process.

For  $G(n, p)$ ,  $n$  large,

$$X_1, X_2, X_3, \dots \approx \text{i.i.d. Binomial}(n, p).$$

Note: there is some freedom in the order in which we “expand” the vertices in this exploration process (depth first? breadth first? something else?)

## Aldous's description of a state of the standard multiplicative coalescent

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Aldous: how to describe the process  $\mathcal{X}(u), u \in \mathbb{R}$ ?

### Theorem (Armendariz)

*Aldous's construction, taken simultaneously for each  $u$ , based on the same underlying Brownian motion  $B$ , corresponds to the whole process  $\mathcal{X}(u), u \in \mathbb{R}$ .*

## 1D representation of random graph process

Let  $v_1$  = vertex chosen uniformly at random

$v_2$  = first vertex to be joined by an edge to  $v_1$

$v_3$  = first vertex outside  $\{v_1, v_2\}$  to be joined by an edge to  $\{v_1, v_2\}$

$\vdots$

$v_{k+1}$  = first vertex outside  $\{v_1, \dots, v_k\}$  to be  
joined by an edge to  $\{v_1, \dots, v_k\}$ .

Then at every time  $t$  in the random graph process, each component consists of an interval of vertices of the form  $\{v_a, v_{a+1}, \dots, v_{a+m}\}$ ; all coalescences involve neighbouring blocks.

The components appear in size-biased order.

This **coupling** of the order of expansion of the vertices for the exploration processes at different times leads to the claimed limit for the process in the scaling window.

## 1D representation of the frozen percolation model

- ▶ only neighbouring blocks can coalesce
- ▶ only the leftmost block can be hit by lightning

To achieve this, we define

$v_1$  = first vertex to be hit by lightning

$v_2$  = first vertex in  $V \setminus \{v_1\}$  to be

hit by lightning or joined to  $v_1$  by an edge

⋮

$v_{k+1}$  = first vertex in  $V \setminus \{v_1, v_2, \dots, v_k\}$  to be

hit by lightning or joined to  $\{v_1, \dots, v_k\}$  by an edge.

Frozen percolation with  $\lambda(n) = \lambda n^{-1/3}$  for some fixed  $\lambda > 0$ .

### Theorem

Let  $b_n = 1 + \frac{1}{3}\lambda \log(n)n^{-1/3}$ .

$$\left( n^{-2/3} C_1^n \left( b_n + sn^{-1/3} \right), n^{-2/3} C_2^n \left( b_n + sn^{-1/3} \right), \dots \right)_{s \in \mathbb{R}} \\ \implies \left( \mathcal{X}_1^\lambda(s), \mathcal{X}_2^\lambda(s), \dots \right)_{s \in \mathbb{R}}.$$

Here  $\mathcal{X}^\lambda$  is a

**“multiplicative coalescent with linear deletion”**

which is an  $\ell_2^\downarrow$ -valued process such that

- ▶ any pair of blocks with size  $a$  and  $b$  merge at rate  $ab$
- ▶ any block of size  $a$  is **deleted** at rate  $\lambda a$ .

The multiplicative coalescent with linear deletion process  $\mathcal{X}^\lambda(s)$  can be described through its *window process*  $U(s)$ .

$U(s)$  is a Markov process which drifts up at rate 1 (representing coalescence) and jumps down (representing deletion).

$U(s)$  can be read off as a deterministic function of the state  $\mathcal{X}^\lambda(s)$ . Conditional on  $U(s) = u$ ,  $\mathcal{X}^\lambda(s)$  has the distribution of  $\mathcal{X}(u)$  (where  $\mathcal{X}(\cdot)$  is the standard multiplicate coalescent, without deletion).

“Window process” because  $U$  describes the corresponding position in the scaling window of the original random graph process.

$U(s)$  converges to a stationary distribution as  $s \rightarrow \infty$ , and correspondingly so does the whole process  $\mathcal{X}^\lambda(s)$  itself.

## Convergence results / conjectures

Fix  $\lambda > 0$  and let  $\lambda(n) = \lambda n^{-1/3}$ .

- (1) Already stated above: for frozen percolation

$$n^{-2/3} \mathbf{C}^n \left( 1 + \frac{\lambda \log n}{3} n^{-1/3} + sn^{-1/3} \right)_{s \in \mathbb{R}} \implies \mathbf{x}^\lambda(s)_{s \in \mathbb{R}}$$

- (2) Same for forest fire.

- (3) For frozen percolation, for any  $t > 1$ ,

$$n^{-2/3} \mathbf{C}^n \left( t + sn^{-1/3} \right)_{s \in \mathbb{R}}$$

converges to a **stationary** version of the MCLD.

- (4) Conjecture: same for forest fire.

- (5) Conjecture: same for forest fire **in stationarity**.