How to determine if a random graph with a fixed degree sequence has a giant component Felix Joos, Guillem Perarnau, Dieter Rautenbach, Bruce Reed

A walkthrough by Angus Southwell

Monash University

Let $\boldsymbol{d} = (d_1, \ldots, d_n)$. Then let $G(\boldsymbol{d})$ be a uniformly chosen simple graph with labelled vertices $\{1, \ldots, n\}$ and degree sequence \boldsymbol{d} . The probability space of such graphs is $\mathcal{G}_{n,\boldsymbol{d}}$.

- Pro: these graphs are much more like most real-world graphs than G(n, p).
- Con: they are much more complicated to analyse.

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Switchings



A switching is an operation that takes G(d) to G'(d).

They are used to find the probability of certain events occurring that we previously got via configuration model, such as:

- probability of specific edges being present,
- probability that a given undiscovered vertex is found at the next step,
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Problem

Given a random graph model, what is the distribution of the size of the largest connected component?

Older results "Double jump" threshold for Erdős–Rényi random graphs at around $\frac{1}{2}n$ edges.

- Below the threshold, all components are order $O(\log n)$.
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Theorem (Molloy and Reed (1995)) Let D(n) be a "well behaved" degree sequence with max. degree at most $n^{\frac{1}{4}-\varepsilon}$. Then define

$$Q(D) := \frac{1}{n} \sum_{j \in [n]} d(j)(d(j) - 2).$$

- If Q(D) < 0, then all components have size $O(\log n)$.
- If Q(D) > 0, then there exists a component with at least αn vertices and βn cycles for $\alpha, \beta > 0$.

- Breadth first search on the graph.
- Keep track of X_t, the number of "half edges" in your component that can be explored still.



Show E_{t-1} [X_t − X_{t-1}] stays positive (or negative) for a sufficiently long time.

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Figure 1: "Superman using his laser vision on the four-vertex empty graph, soon to be the three-vertex empty graph" - Tim

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- Handling of large degree vertices is nonexistent.
- Criterion does not extend to general degree sequences:

Consider $n = k^2$ for large odd k, and d = (1, ..., 1, 2k). Then $Q(D) \approx 3$, so we would expect a giant component according to the MR criterion.



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Theorem (Joos et al. (2018)) For any function $\delta \to 0$ as $n \to \infty$, for every $\gamma > 0$, if $R_D \le \delta M_D$, the probability that G(D) has a component of order at least γn is o(1).

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- Now with added preprocessing!



- Suppression of degree 2 vertices.
- Switchings used to work in the graph model to get edge probabilities.

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Let H be the graph G with all degree 2 vertices contracted, let $\omega > 0$ be small such that $R_D \le \omega M_D$.

Let S be the smallest set of vertices of H such that $\sum_{i \in S} d_i \ge 5\omega^{1/4}M$ and no vertex outside of S is larger. Define initial exploration set S_0 to be $S \cup \{v\}$ for any vertex v. Define X'_t by

$$X'_{0} = \sum_{u \in S_{0}} d(u),$$
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$$egin{aligned} X_0' &= \sum_{u \in S_0} d(u), \ X_t' &= X_0' + \sum_{i=1}^t (d(w_i) - 2). \end{aligned}$$

We get the following results about the initial stages of the exploration:

Lemma

- $\sum_{w \in V \setminus S} d(w) (d(w) 2) \leq -4\omega^{1/4} M$,
- there is a vertex in S of degree at most $\omega^{-1/4}$.

So the vertices outside the preprocessing set are "small".

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- All degrees outside *S*_{*t*-1} being low helps the bounds on switchings.

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The game is now to bound

$$\mathbb{E}_{t-1}\left[d(w_t)-2\right],$$

the expected increase between X'_{t-1} and X'_t .

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Lemma

If $t \leq \omega^{1/9} M$ and $X'_{t-1} \leq \omega^{1/5} M$, and $X_{t'} > 0$ for all t' < t, then:

• If
$$w \in V ackslash S_{t-1}$$
 and $d(w) = 1$, then

$$\mathbb{P}(w_t = w) \ge (1 - 9\omega^{1/5}) \frac{1}{M_{t-1}}$$

• If
$$w \in V \setminus S_{t-1}$$
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$$\mathbb{P}(w_t = w) \leq (1 + 9\omega^{1/5}) \frac{d(w)}{M_{t-1}}.$$

Switchings II



Want to find the number of forward and backward switchings.

Number of forward switchings is at most M_{t-1} . How many of these are 'bad'?

- x or $y \in S_{t-1}$
- $v_t \sim x$
- $w \sim y$
- Vertices overlap

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Lemma Define $Y_t = d(w_t) - \mathbb{E}_{t-1}(d(w_t))$. The probability that there exists a *t* such that $\sum_{t' \le t} Y_{t'} > M^{2/3}$ is less than $e^{-M^{1/4}}$.

Lemma For $t \leq \lfloor rac{\omega^{1/9}M}{2}
floor$, we have that

$$\mathbb{E}_{t-1}\left(d(w_t)-2\right) \leq -\frac{t}{M} + 19\omega^{1/5}$$

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With probability greater than $1 - e^{-M^{1/4}}$, there exists a time $t \leq \lfloor \frac{\omega^{1/9}M}{3} \rfloor$ such that $X_t = 0$.

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Same exploration process, but different preprocessing.

Preprocessing – supercritical case Expose all components in *H* containing a vertex of degree larger than $\frac{\sqrt{M}}{\log M}$. Call the set of exposed vertices *U*.

Analysis splits into two cases:

- $\sum_{u \in U} d(u) \geq \frac{R}{100}$ then U contains a giant component,
- $\sum_{u \in U} d(u) < \frac{R}{100}$ same exploration as in the subcritical case.

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Let U be a set of vertices containing all vertices with degree greater than $\frac{\sqrt{M}}{\log M}$ and let $\frac{1}{4} < c < 1$ be such that $\sum_{u \in U} d(u) \leq cR$. Then

$$\sum_{u\in V\setminus U} d(u) \left(d(u)-2\right) \geq \frac{(1-c)}{2}R.$$

Supercritical switching analysis



Same switching, more complicated bounds: need a lower bound on $\# \mathsf{backward}.$

Bad backward switchings are:

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Let $\beta = 10^{-6}\varepsilon^2$ be a fixed constant. If $M_{t-1} \ge \frac{3M}{4}$ and $X_{t-1} \le \beta M$, then for every $w \in S_{t-1}$,

$$(1-10\sqrt{eta})rac{d(w)}{M_{t-1}} \leq \mathbb{P}\left(w=w_t
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Lemma

For $t \leq \tau$, $\mathbb{E}[d(w_t) - 2] \geq \frac{\varepsilon}{4}$, $\mathbb{E}[d'_t(w_t)] \leq \frac{\mathbb{E}[d(w_t) - 2]}{3}$, and thus $\mathbb{E}[X_t - X_{t-1}] \geq \frac{\varepsilon}{12}$.

Here τ is the smallest t for which either $X_t \ge \beta M$ or $M_t \le \left(1 - \frac{R}{4M}\right) M_0$.

 $X_t \ge X_{t-1} + (d(w_t) - 2) - 2d'_t(w_t)$, so we can use this recursively to get...

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$$X_{\tau} \geq \mathbb{E}\left[X_{\tau}\right] + \sum_{t \leq \tau} A_t + \sum_{t \leq \tau} B_t,$$

where $A_t = d(w_t) - \mathbb{E}[d(w_t)]$ and $B_t = d'_s(w_t) - \mathbb{E}[d'_t(w_t)]$.

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Lemma

With probability 1 - o(1), $X_{\tau} \ge \beta M$.

- We found bounds on the number of edges in each component, not vertices!
- What about degree 2 vertices?
 - Degree 2 vertices in components of H(D)
 - Degree 2 vertices in cyclic components
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Thank you!

References

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