

Lecture 3: Random Graphs

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The Erdős-Rényi class $\mathcal{G}_{n,p}$

Definition

Today we discuss about random graphs. The *Erdős-Rényi class* $\mathcal{G}_{n,p}$ of random graphs is defined as follows.

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Today we discuss about random graphs. The *Erdős-Rényi class* $\mathcal{G}_{n,p}$ of random graphs is defined as follows.

Let \mathcal{V} denote the set of n vertices, $\mathcal{V} = \{1, 2, \dots, n\}$, and let \mathcal{G} denote the

set of all graphs with vertices \mathcal{V} . There are exactly $2^{\binom{n}{2}}$ such graphs.

The probability mass function on \mathcal{G} , $P : \mathcal{G} \rightarrow [0, 1]$, is obtained by assuming that, as random variables, edges are independent from one another, and each edge occurs with probability $p \in [0, 1]$. Thus a graph $G \in \mathcal{G}$ with m edges will have probability $P(G)$ given by

$$P(G) = p^m (1 - p)^{\binom{n}{2} - m}.$$

(explain why)

The Erdős-Rényi class $\mathcal{G}_{n,p}$

Probability space

Formally, $\mathcal{G}_{n,p}$ stands for the the probability space (\mathcal{G}, P) composed of the set \mathcal{G} of all graphs with n vertices, and the probability mass function P defined above.

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A reformulation of P : Let $G = (\mathcal{V}, \mathcal{E})$ be a graph with n vertices and m edges and let A be its adjacency matrix. Then:

$$\begin{aligned} P(G) &= \prod_{(i,j) \in \mathcal{E}} \text{Prob}((i,j) \text{ is an edge}) \prod_{(i,j) \notin \mathcal{E}} \text{Prob}((i,j) \text{ is not an edge}) = \\ &= \prod_{1 \leq i < j \leq n} p^{A_{i,j}} (1-p)^{1-A_{i,j}} \end{aligned}$$

where the product is over all ordered pairs (i,j) with $1 \leq i < j \leq n$. Note:

$$|\{(i,j), 1 \leq i < j \leq n\}| = \binom{n}{2} \quad \& \quad |\{(i,j) \in \mathcal{E}\}| = |\mathcal{E}| = m = \sum_{1 \leq i < j \leq n} A_{i,j}.$$

The Erdős-Rényi class $\mathcal{G}_{n,p}$

Computations in $\mathcal{G}_{n,p}$

How to compute the expected number of edges of a graph in $\mathcal{G}_{n,p}$?

The Erdős-Rényi class $\mathcal{G}_{n,p}$

Computations in $\mathcal{G}_{n,p}$

How to compute the expected number of edges of a graph in $\mathcal{G}_{n,p}$?

Let $X_2 : \mathcal{G}_{n,p} \rightarrow \{0, 1, \dots, \binom{n}{2}\}$ be the random variable of *number of edges of a graph G* .

$$X_2 = \sum_{1 \leq i < j \leq n} 1_{(i,j)} \quad , \quad 1_{(i,j)}(G) = \begin{cases} 1 & \text{if } (i,j) \text{ is edge in } G \\ 0 & \text{if otherwise} \end{cases}$$

Use linearity and the fact that $\mathbb{E}[1_{(i,j)}] = \text{Prob}((i,j) \in \mathcal{E}) = p$ to obtain:

$$\mathbb{E}[\text{Number of Edges}] = \binom{n}{2} p = \frac{n(n-1)}{2} p$$

The Erdős-Rényi class $\mathcal{G}_{n,p}$

MLE of p

Given a realization G of a graph with n vertices and m edges, how to estimate the most likely p that explains the graph.

Concept: The Maximum Likelihood Estimator (MLE).

In statistics: The MLE of a parameter θ given an observation x of a random variable $X \sim p_X(x; \theta)$ is the value θ that maximizes the probability $P_X(x; \theta)$:

$$\theta_{MLE} = \operatorname{argmax}_{\theta} P_X(x; \theta).$$

In our case: our observation G has m edges. We know

$$P(G; p) = p^m (1 - p)^{\binom{n}{2} - m}.$$

The Erdős-Rényi class $\mathcal{G}_{n,p}$

MLE of p

Lemma

Given a random graph with n vertices and m edges, the MLE estimator of p is

$$p_{MLE} = \frac{m}{\binom{n}{2}} = \frac{2m}{n(n-1)}.$$

The Erdős-Rényi class $\mathcal{G}_{n,p}$ MLE of p

Lemma

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Why

Note $\log(P(G; p)) = m \log(p) + \left(\binom{n}{2} - m \right) \log(1-p)$ and solve for p :

$$\frac{d \log(P)}{dp} = \frac{m}{p} - \frac{\binom{n}{2} - m}{1-p} = 0.$$

The Erdős-Rényi class $\mathcal{G}_{n,p}$

Method of Moments Estimator for p

An alternative parameter estimation method is the moment matching method. Given a likelihood function for observed data $p(x; \theta)$ and a sequence of observations (x_1, x_2, \dots, x_N) , the moment matching method computes the parameters $\theta \in \mathbb{R}^d$ by solving the system of equations:

$$\mathbb{E}[X] = \frac{1}{N} \sum_{t=1}^N x_t \quad \dots \quad \mathbb{E}[X^d] = \frac{1}{N} \sum_{t=1}^N x_t^d$$

(or unbiased estimates of the moments). In particular, for the Erdős-Rényi class, we match the first moment with the observation: $\frac{n(n-1)}{2} p = m$.

Hence

$$p_{MM} = \frac{2m}{n(n-1)},$$

same as the MLE estimator.

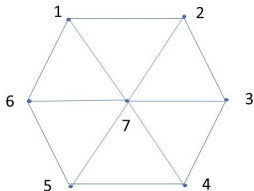
Cliques

q -cliques

Definition

Given a graph $G = (\mathcal{V}, \mathcal{E})$, a subset of q vertices $S \subset \mathcal{V}$ is called a q -clique if the subgraph $(S, \mathcal{E}|_S)$ is complete.

In other words, S is a q -clique if for every $i \neq j \in S$, $(i, j) \in \mathcal{E}$ (or $(j, i) \in \mathcal{E}$), that is, (i, j) is an edge in G .



- Each edge is a 2-clique.

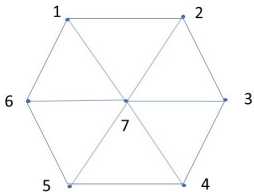
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- Each edge is a 2-clique.
- $\{1, 2, 7\}$ is a 3-clique. And so are $\{2, 3, 7\}$, $\{3, 4, 7\}$, $\{4, 5, 7\}$, $\{5, 6, 7\}$, $\{1, 6, 7\}$

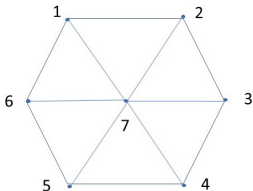
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- There is no k -clique, with $k \geq 4$.

The Erdős-Rényi class $\mathcal{G}_{n,p}$

Computations in $\mathcal{G}_{n,p}$: q -cliques

How to compute the expected number of q -cliques?

The Erdős-Rényi class $\mathcal{G}_{n,p}$

Computations in $\mathcal{G}_{n,p}$: q -cliques

How to compute the expected number of q -cliques?

For $k = 2$ we computed earlier the number of edges, which is also the number of 2-cliques.

We shall compute now the number of 3-cliques: triangles, or 3-cycles.

Let $X_3 : \mathcal{G}_{n,p} \rightarrow \mathbb{N}$ be the random variable of number of 3-cliques. Note

the maximum number of 3-cliques is $\binom{n}{3}$.

Let S_3 denote the set of all distinct 3-cliques of the complete graph with n vertices, $S_3 = \{(i, j, k) , 1 \leq i < j < k \leq n\}$.

Let

$$1_{(i,j,k)}(G) = \begin{cases} 1 & \text{if } (i, j, k) \text{ is a 3-clique in } G \\ 0 & \text{if otherwise} \end{cases}$$

The Erdős-Rényi class $\mathcal{G}_{n,p}$

Expectation of the number of 3-cliques

Note: $X_3 = \sum_{(i,j,k) \in \mathcal{S}_3} 1_{(i,j,k)}$. Thus

$$\mathbb{E}[X_3] = \sum_{(i,j,k) \in \mathcal{S}_3} \mathbb{E}[1_{(i,j,k)}] = \sum_{(i,j,k) \in \mathcal{S}_3} \text{Prob}((i,j,k) \text{ is a clique}).$$

Since $\text{Prob}((i,j,k) \text{ is a clique}) = p^3$ we obtain:

$$\mathbb{E}[\text{Number of 3-cliques}] = \binom{n}{3} p^3 = \frac{n(n-1)(n-2)}{6} p^3.$$

The Erdős-Rényi class $\mathcal{G}_{n,p}$

Number of 3 cliques

Assume we observe a graph G with n vertices and m edges. What would be the expected number N_3 of 3-cliques?

$$\mathbb{E}[X_3 | X_2 = m] = \frac{1}{L} \sum_{k=1}^L X_3(G_k)$$

where L denotes the number of graphs with m edges and n vertices, and G_1, \dots, G_L is an enumeration of these graphs.

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We approximate:

$$\mathbb{E}[X_3 | X_2 = m] \approx \mathbb{E}[X_3; p = p_{MLE}(m)]$$

and obtain:

$$E[X_3 | X_2 = m] \approx \frac{4(n-2)}{3n^2(n-1)^2} m^3.$$

The Erdős-Rényi class $\mathcal{G}_{n,p}$

Expectation of the number of q -cliques

Let $X_q : \mathcal{G}_{n,p} \rightarrow \mathbb{N}$ be the random variable of number of q -cliques. Note the maximum number of q -cliques is $\binom{n}{q}$.

Let S_q denote the set of all distinct q -cliques of the complete graph with n vertices, $S_q = \{(i_1, i_2, \dots, i_q), 1 \leq i_1 < i_2 < \dots < i_q \leq n\}$. Note

$$|S_q| = \binom{n}{q}.$$

Let

$$1_{(i_1, i_2, \dots, i_q)}(G) = \begin{cases} 1 & \text{if } (i_1, i_2, \dots, i_q) \text{ is a } q\text{-clique in } G \\ 0 & \text{if otherwise} \end{cases}$$

The Erdős-Rényi class $\mathcal{G}_{n,p}$

Expectation of the number of q -cliques

Since $X_q = \sum_{(i_1, \dots, i_q) \in \mathcal{S}_q} \mathbf{1}_{i_1, \dots, i_q}$ and

$Prob((i_1, \dots, i_q) \text{ is a clique}) = p^{\binom{q}{2}}$ we obtain:

$$\mathbb{E}[\text{Number of } q\text{-cliques}] = \binom{n}{q} p^{q(q-1)/2}.$$

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Using a similar argument as before, if G has m edges, then

$$\mathbb{E}[X_q | X_2 = m] \approx \binom{n}{q} \left(\frac{2m}{n(n-1)} \right)^{q(q-1)/2}.$$

Computation of Number of Cliques

An Iterative Algorithm

We discuss two algorithms to compute X_q : iterative, and adjacency matrix based algorithm.

Framework: we are given a sequence $(G_t)_{t \geq 0}$ of graphs on n vertices, where G_{t+1} is obtained from G_t by adding one additional edge:

$G_t = (\mathcal{V}, \mathcal{E}_t)$, $\emptyset = \mathcal{E}_0 \subset \mathcal{E}_1 \subset \dots$ and $|\mathcal{E}_t| = t$.

Iterative Algorithm: Assume we know $X_q(G_t)$, the number of q -cliques of graph G_t . Then $X_q(G_{t+1}) = X_q(G_t) + D_q(e; G_t)$ where $D_q(e; G_t)$ denotes the number of q -cliques in G_{t+1} formed by the additional edge $e \in \mathcal{E}_{t+1} \setminus \mathcal{E}_t$.

Computation of Number of Cliques

An Analytic Formula

Laplace Matrix $\Delta = D - A$ contains all connectivity information.

Idea: Note the (i, j) element of A^2 is

$$(A^2)_{i,j} = \sum_{k=1}^n A_{i,k}A_{k,j} = |\{k : i \sim k \sim j\}|.$$

This means $(A^2)_{i,j}$ is the number of paths of length 2 that connect i to j .

Remark: The diagonal elements of $A(A^2 - D)$ represent twice the number of 3-cycles (= 3-cliques) that contain that particular vertex.

Conclusion:

$$X_3 = \frac{1}{6} \text{trace}\{A(A^2 - D)\} = \frac{1}{6} \text{trace}(A^3).$$

Computation of Number of Cliques

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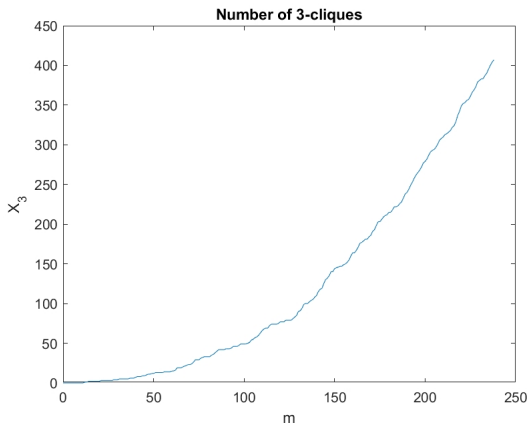
$$X_3 = \frac{1}{6} \text{trace}\{A(A^2 - D)\} = \frac{1}{6} \text{trace}(A^3).$$

Exercise: Generalize this formula for X_4 .

Numerical results

Graph of X_3 for the BKOFF dataset

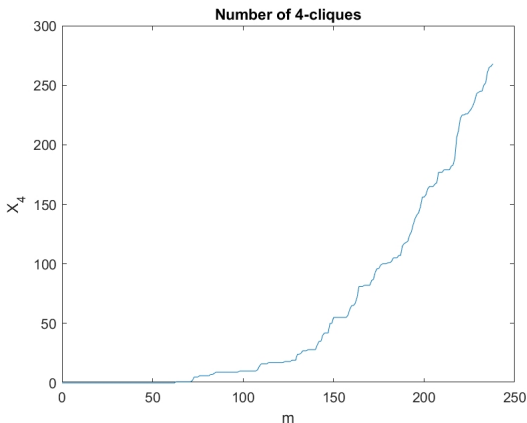
Recall the dataset Bernard & Killworth Office. Weighted graph: Ordered $m = 238$ edges for $n = 40$ nodes. The plot of X_3 the number of 3-cliques:










Numerical results

Plot of X_4 for the BKOFF dataset

Weighted graph: Ordered $m = 238$ edges for $n = 40$ nodes. The plot of X_4 the number of 4-cliques:



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