

Large deviations for random networks & applications

$G \sim G(n, p)$ a random graph
on n vertices
where every edge
occurs ind. w.p. p .

$$H = (V(H), E(H))$$

- Take H to be $K_3 \triangle$

X_H - no of copies of H in G .

$$E(X_H) \approx n^3 p^3$$

$$P\left(X_H \geq (1+\delta) E(X_H) \right) \quad \delta > 0.$$

A - Infamous upper
tail problem.

(Janson-Rucinski
'02).

- Geometric question

What does the random graph G look
like given the event A .

Fact: X_H is a polynomial of independent
Bernoulli variables.

Recall some classical concentration & large deviations results for linear functions

Azuma-Hoeffding inequality

X_1, X_2, \dots, X_n are independent mean zero random variables such that $a_i \leq X_i \leq b_i$ almost surely.

$$S_n = \sum_{i=1}^n X_i$$

- $P(|S_n| > t)$
- strategy is to compute exponential moments & then apply Markov.

$$E(e^{\theta X_i}) \leq e^{\theta^2 \underbrace{(b_i - a_i)^2}_{c_i} / 8}$$

$$E(e^{\theta S_n}) \leq e^{\theta^2 \sum_{i=1}^n c_i / 8}$$

$$P(S_n > t) \leq e^{\theta^2 \sum c_i / 8 - \theta t}.$$

#0.

- optimize over θ .

- X_i are coin tosses.

$$X_i \stackrel{iid}{\sim} \text{Ber}(p)$$

$$S_n = np + O(\sqrt{n})$$

$$P(S_n > nq)$$

$$\Lambda(\theta) = \log(pe^\theta + (1-p))$$

$$\leq e^{(n\Lambda(\theta) - n\theta q)}$$

$$\sup_{\theta} (\theta q - \Lambda(\theta))$$

$$= \mathbb{I}_p(q) = q \log \frac{q}{p} + (1-q) \log \frac{1-q}{1-p}$$

$$P(S_n > nq) \leq e^{-n \mathbb{I}_p(q)} \text{ (error free bound)}$$

- Lower bound - (Tilting)
- strategy is to do a change of measure which makes the atypical event typical.

- To get back to the original measure estimate the R-N derivative between the two measures

$$P(A) = \int_A e^{\log \frac{dP}{dQ}} dQ$$

Recall X_H is the number of copies of H in G .

$$t(H, G) = \frac{1}{n^{|V(H)|}} \sum_{i_1, i_2, \dots, i_k} \prod_{(x, y) \in E(H)} a_{i_x i_y}$$

G graph on n vertices with adjacency matrix $A = (a_{ij})$

- What does $G(n, p)$ look like
give $t(H, G)$ is large.

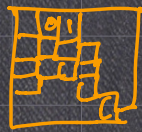
- $G(n, p)$ continues to look like
an Erdos-Renyi type graph
but with different densities.
(inhomogeneous random graph)

It will be convenient to define a
metric on graphs & embed them
in the same space.

Defn Let \mathcal{W} be the set of all
sym. mble function from $[0, 1]^2 \rightarrow [0, 1]$
- Graphons.



Note that any finite graph
naturally embeds in \mathcal{W} as
a $\{0, 1\}$ valued step function.



For $f, g \in \mathcal{W}$

$$d_{\Pi}(f, g) = \sup_{S, T \subset [0,1]} \left| \int_{S \times T} (f - g) \right|$$

cut distance.

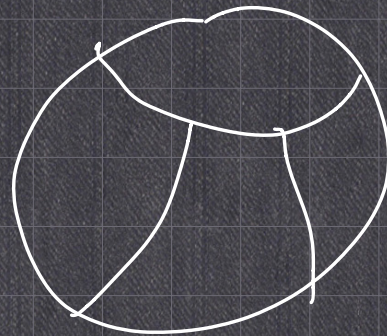
- Since we don't care about labels of the graphs, we should identify graphs/graphons which are the same up to a relabelling.

σ -measure pres. bijection on $[0,1]$
 $f \sim g$ are equivalent if
 $\exists \sigma$ st $d_{\Pi}(f, g \circ \sigma) = 0$

- We would work with the quotient space $\tilde{\mathcal{W}}$.

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A_1	P	P	P	P
A_2	P	P	P	P
A_3	P	P	P	P
A_4	P	P	P	P



$$G \sim G(n, p)$$

$d(G, P)$ is typically small.

- A related question is what is the prob the graph looks like

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A_1	P_1	P_2	P_3	P_4
A_2	P_2	P	P	P
A_3	P_3	P	P	P
A_4	P_4	P	P	P

- Notice that this is exactly the coin tossing problem.

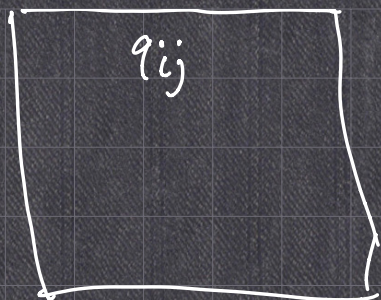
The prob of this, using the same reasoning as the coin tossing is

$$e^{-\left[\binom{A_1}{2} I_p(P_1) + \binom{A_1}{A_2} I_p(P_2) \dots \right]}$$

- We want to find the best possible block graph which makes the atypical event typical.

$$\phi(H, n, P, \delta) = \min \left(\sum_{1 \leq i < j \leq n} I_p(q_{ij}) \right) :$$

$$t(H, \delta) \geq (1+\delta) E_p(t(H, \delta))$$

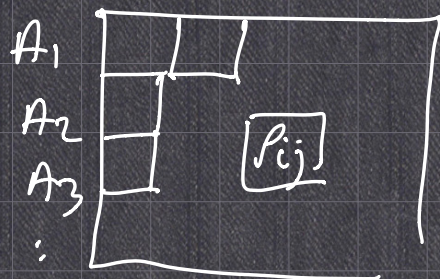


$$Q = (q_{ij})$$

is the new weighted graph.

- $\phi(H, n, p, \delta)$ is the best prob ϵ one can get by the strategy of considering inhomogeneous random graphs.
- How to prove that this is indeed the optimal strategy.
- Szemerédi's regularity lemma
 - roughly this says "any" graph can be approximated by a block random graph where the no of blocks is only a function of the error and not the size of the graph.
 - (Weak reg. lemma - Frieze & Kannan)
 - Given any graph $G = (V, E)$ there exists a partition of V into

K classes $A_1, A_2, \dots, A_K \subset \mathcal{P}$
 such that



$$p_{ij} = \frac{E_G(A_i, A_j)}{|A_i| |A_j|}$$

$$d_G(G, G_p) \leq o\left(\frac{1}{\sqrt{\log k}}\right)$$

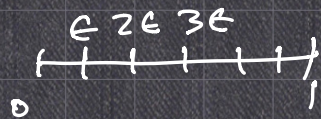
One crucial prop of the cut dist
 - (Counting Lemma)

Fix H , and graphs f, g

$$|t(H, f) - t(H, g)| \leq O_H(d_G(f, g))$$

- Using the above two facts
 one can compute the probability
 that G looks like a given
 block graph and then union bound
 over all possible choices of block
 graphs (consider all possible
 partition of V into K blocks
 and all possible edge densities

β_{ij} up to an ϵ error)



- If the union bd is over a not too big set, the upper bd one gets is $e^{-\phi(H, n, p, \delta)}$ smaller order.
- This fails if p is going to zero with n faster than a poly log.
- Full LDP on graphons for a fixed p was proven by Chatterjee & Varadhan (2011)
- The argument above which is more combinatorial. - Leberky-Zhao (2015).