

(Random) Finite graphs from Graphons:-

$K: [0,1]^2 \rightarrow [0,1]$ - Symmetric and measurable

n -vertices $\{1, \dots, n\} \equiv V$

Edge $i \sim j$ with probability $K(u_i, u_j) \equiv E$

where $\{u_i\}_{i=1}^n$ are independent $U[0,1]$ random variables.

$G(n, K) \equiv (V, E).$

$i \sim j$ with probability

$K(u_i, u_j)$

$V = \{1, \dots, n\}$

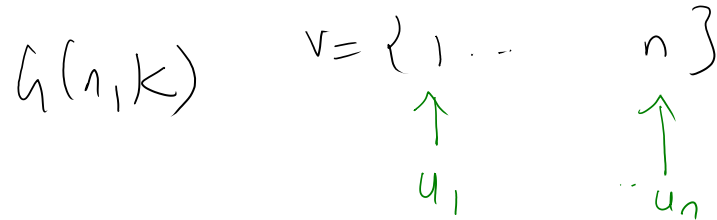
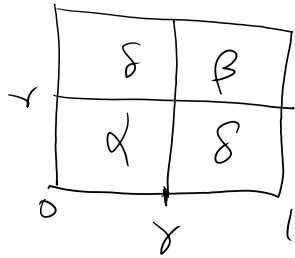
labels	u_1	u_n
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 $\equiv U[0,1]$

Examples:- • $k(\cdot, \cdot) = p$ $0 < p < 1$

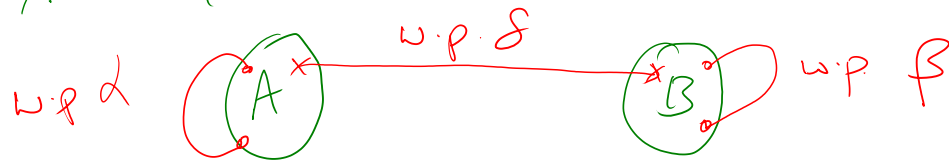
Then $G(n, p) \equiv$ Erdős-Rényi

- Each edge $i \sim j$ w.p. p
 $V = \{1, \dots, n\}$



$i \sim j$ with probability $k(u_i, u_j)$

$A = \{i \in V \mid u_i < \sigma\}$ $B = \{i \in V \mid u_i > \sigma\}$



Show :- $G(n, K) \xrightarrow{\quad ? \quad} K$ in d_{sub} as $n \rightarrow \infty$

Theorem : Let $\{U_i\}_{i \geq 1}$ be a sequence of i.i.d uniform $[0,1]$ random variables. Let $K: [0,1]^2 \rightarrow [0,1]$ be a graphon and be continuous a.e on $([0,1] \times [0,1], dx \times dy)$

Then $d_{\text{sub}}(G(n, K), K) \xrightarrow{\text{as } n \rightarrow \infty} 0$ with

probability 1.

Remark : - $U^{(n)} = (U_1^{(n)}, \dots, U_i^{(n)})$ where $\{U_i^{(j)}\}_{i \geq 1, j \geq 1}$ are i.i.d $U([0,1])$ - Theorem V
(Resample edges at each n)

• ~~Proof that we will present breaks down if~~ (? Result)
Continuity assumption does not hold.

• [LS 06] + Theorem $\Rightarrow$ There is always an "intertwined" random graph sequence that converges to the graphon for any Cauchy sequence $\{G_n\}_{n \geq 1}$ of dense graphs.

Proof:-

To show

$$d_{\text{sub}}(g(n, k), k) \rightarrow 0 \text{ as } n \rightarrow \infty \text{ w.p. 1}$$

where:

$$d_{\text{sub}}(g(n, k), k) = \sum_{i=1}^{\infty} \frac{1}{2^i} |S(F_i, g(n, k)) - S(F_i, k)|$$

of vertices F_i
 $\equiv m$

$$S(F_i, g(n, k)) = \frac{X_{F_i}(g(n, k))}{X_{F_i}(k_n)} \in$$

$$S(F_i, k) = \int_{[0,1]^m} \prod_{j=1}^m dx_j \prod_{(x, u) \in e(F)} k(x_x, x_u)$$

Its enough to show:

$$S(F_i, g(n, k)) \rightarrow S(F_i, k) \text{ as } n \rightarrow \infty$$

with probability one.

E - expectation over the edge probabilities

E - expectation over the labels.

Condition on $u_1, u_2, \dots, u_n = (x_1, \dots, x_n) \in [0, 1]^n$

let $G(n, k) \big|_{(u_1, \dots, u_n) = (x_1, \dots, x_n)} \equiv G(n, x, k)$

let F be a finite graph.

Q: How large is $|S(F, G(n, x, k)) - \mathbb{E}(S(F, G(n, x, k)))|$ as $n \rightarrow \infty$?

Concentration - inequalities } to understand the above
can be used

McDiarmid's Concentration Inequality

$X_1, X_2, \dots, X_n$ be independent, $X_i \in \mathcal{X}_i$
 $f: \prod_{i=1}^n \mathcal{X}_i \rightarrow \mathbb{R}$ that is "Lipschitz":

i.e. $x, x' \in \prod_{i=1}^n \mathcal{X}_i$ that differ only in the
 k^{th} coordinate then

$$|f(x) - f(x')| \leq \sigma_k.$$

let $Y = f(X_1, \dots, X_n)$. Then $\forall \alpha > 0$

$$\mathbb{P}(|Y - \mathbb{E}[Y]| > \alpha) \leq 2 \exp\left(-\frac{2\alpha^2}{\sum_{i=1}^n \sigma_i^2}\right)$$

$S(F_i, G(n, x, k)) =$ function on all the edges.

- random variables in the Theorem are Edge random variables

Q - How much will $S(F_i, G(n, x, k))$ change if
we remove/add a single edge?
- Bound for change

ANS:-

F - finite graph - on k vertices

G - graph on n vertices

G' - graph on n vertices

G' differs from G by a
Single edge

Ex:-

$$|S(F, G) - S(F, G')| \leq \frac{k(k-1)}{n(n-1)} \equiv \begin{matrix} 0 \\ \text{Same} \\ \text{+ edge} \\ \text{change} \end{matrix}$$

Fix $\varepsilon > 0$:- $\mathbb{P}(|S(F, G(n, x, k)) - \mathbb{E}[S(F, G(n, x, k))]| > \varepsilon)$

F - graph
on k - vertices

$$\leq 2 \exp \left[- \frac{2 \varepsilon^2}{n G_2 \left(\frac{k(k-1)}{n(n-1)} \right)^2} \right]$$

$$\equiv 2 \exp(-a_n) \quad a_n = O(n^2)$$

$\Rightarrow$

Ex:- $|S(F, G(n, x, k)) - \mathbb{E}[S(F, G(n, x, k))]|$
 $\longrightarrow 0$ with probability one

To show the result we need to show:

$$S(F, h(n, k)) \equiv S(F, h(n, (u_1, \dots, u_n), k))$$

We showed :-

$$|S(F, h(n, u_1, \dots, u_n, k)) - E(S(F, h(n, u_1, \dots, u_n, k)))| \longrightarrow 0 \quad \text{with probability one}$$

We need to show:

Generalised U-statistic
LLN

$$E(S(F, h(n, u_1, \dots, u_n, k))) \xrightarrow[\text{with probability 1}]{\text{Generalised U-statistic LLN}} S(F, k)$$

$$\rightarrow S(F, k) = E\left(\prod_{(i,j) \in e(F)} K(v_i, v_j)\right) \quad \begin{array}{l} \text{independent} \\ v_i = U(o_i) \\ \text{d.v.'s} \end{array}$$

"U-statistic"

Please:

Zoom Contacts



add

Sivaathreya @ yahoo.co.in
as contact