

6/10/20: L6 - LWC FOR RANDOM GRAPHS

Thm: Let G_n be as above. $G_n \xrightarrow{W} (G, \rho)$

iff $\forall H_* \in \mathcal{G}_*$ we have that

$$\frac{1}{n} \sum_{u \in V_n} \mathbb{1}[B_r^{(G_n)}(u) \cong H_*] \rightarrow \mathbb{P}(B_r^{(G)}(o) \cong H_*)$$

Proof: $\Rightarrow$ as $\mathbb{1}[B_r^{(G)}(o) \cong H_*]$ is cts. (see A1, Thm 1)
 $\Leftarrow$ (non-trivial & more useful).

Let $h: \mathcal{G}_* \rightarrow \mathbb{R}$ be bdd cts & $\varepsilon > 0$.

By Portmanteau theorem, we can assume that h is uniformly cts as well.

$$\text{So } \exists \delta > 0 \ni d((G_i, o_i), (G_i', o_i')) < \delta$$

$$\Rightarrow |h(G_i, o_i) - h(G_i', o_i')| < \varepsilon/4$$

Choose $r = r(\delta) \ni \forall (G_i, o_i)$

$$d((G_i, o_i), B_r(o_i)) \leq \frac{1}{r+1} < \delta.$$

By Δ ineq,

$$\begin{aligned} & |\mathbb{E}_n[h(G_n, o_n)] - \mathbb{E}[h(G, o)]| \\ & \leq |\mathbb{E}_n[h(B_r^{(G_n)}(o_n))] - \mathbb{E}[h(B_r^{(G)}(o))]| + \varepsilon/2. \end{aligned}$$

if $\exists$ only finitely many graphs of depth r then
 easy to approximate $h(B_r^{(G_n)}(o_n)) \approx h(B_r^{(G)}(o))$
 by our assumption.

One obstruction to have finitely many graphs of
 depth r is that vertices can have arbitrarily
 large degrees. But this prob can be made small.

Define $E_{r,k}(G,o) = \{\exists v \in B_r(o) \ni d_v(G) > k\}$
 $E_{r,k}(G,o) \downarrow \{\exists v \in B_r(o) \ni d_v(G) = \infty\}$ as $k \uparrow \infty$.
 $\Rightarrow \forall \varepsilon > 0, \exists k \ni P(E_{r,k}(G,o)) \leq \varepsilon \quad (P(\vec{\cdot}) = 0)$

Consider $E_{r,k}(G,o)^c$ - $\exists$ fin. many rooted graphs

$H_1, \dots, H_m \ni E_{r,k}(G,o)^c = \bigcup_{i=1}^m B_r(o) \cong H_i$

||ly $E_{r,k}(G_n, o_n)^c = \bigcup_{i=1}^m \{B_r(o_n) \cong H_i\}$

By our assumption, we've that

$P_n(B_r(o_n) \cong H_i) \rightarrow P(B_r(o) \cong H_i) \quad 1 \leq i \leq m$

$\Rightarrow P_n(E_{r,k}(G_n, o_n)^c) \rightarrow P(E_{r,k}(G,o)^c) \quad (*)$

$\Rightarrow \exists n_0 \ni \forall n \geq n_0$

$P(E_{r,k}(G_n, o_n)) \leq 2\varepsilon$

$\Rightarrow |E_n[h(B_r(o_n))] - E[h(B_r(o))]|$

$\leq |E_n[h(B_r(o_n)) \mathbb{1}_{E_{r,k}(G_n, o_n)^c}] - E[h(B_r(o)) \mathbb{1}_{E_{r,k}(G,o)}]|$
 $+ 2\varepsilon \|h\|_\infty, \quad \forall n \geq n_0.$

Now since $E_{r,k}(G_n, 0_n)^c$ is determined by finitely many graphs, from our assumption ^(*) we obtain that $E[h(Br(n)) \mathbb{1}_{E_{r,k}(G_n, 0_n)^c}] \rightarrow E[h(Br(0)) \mathbb{1}_{E_{r,k}(G, 0)^c}]$.

$\Rightarrow \exists n_1 \geq n_0 \exists \forall n \geq n_1$

$$\left| E[h(Br(n)) \mathbb{1}_{E_{r,k}(G_n, 0_n)^c}] - E[h(Br(0)) \mathbb{1}_{E_{r,k}(G, 0)^c}] \right| \leq \varepsilon/4$$

Choosing ε small & plugging in the above bounds, the proof is complete. ■

What about random graphs G_n ?

$$E_n[h(G_n, 0_n)] = \frac{1}{n} \sum_{i \in [n]} h(G_n, i) \text{ — random variable.}$$

So how do we interpret convergence?

Defn: $G_n = ([n], E(G_n))$ is a seq. of RBGs (possibly disconn.)

(1) $G_n \xrightarrow{\text{LW-d}} (G, 0)$ if $E[h(G_n, 0_n)] \rightarrow E[h(G, 0)]$
 ∇ odd cls $h: \mathcal{G}_* \rightarrow \mathbb{R}$.

$$\mathbb{E}[h(G_n, o_n)] = \mathbb{E}[\mathbb{E}_n[h(G_n, o_n)]]$$

$\downarrow$
 Exp over G_n & o_n

$$= \frac{1}{n} \mathbb{E}\left[\sum_{i \in [n]} h(G_n, i)\right]$$

$$(2) \quad G_n \xrightarrow{\text{LW-P}} (G, o) \text{ if } \mathbb{E}_n[h(G_n, o_n)] \xrightarrow{P} \underbrace{\mathbb{E}[h(G, o)]}_{\mu}$$

$\forall$ bdd cts h .

Rem: (1) One can define a.s. convergence as well but not so much of interest to us now. In general, not common.

(2) In LW-P, we have assumed that the limit (G, o) has prob. distribn μ on $\mathcal{G}_*$.

But we can also assume μ is a random prob. measure on $\mathcal{G}_*$ i.e., μ can vary with realizations of G_n .
 (Ex(A): Construct examples).

(3) Our focus will purely be with deterministic limits.

THM: let $(G_n)_{n \geq 1}$ be a sequence of graphs (random or deterministic)

(a) $G_n \xrightarrow{\text{LW-d}} (G, 0)$ iff $\forall H_* \in \mathcal{G}_*$,

$$\mathbb{E}[p^{(n)}(H_*)] = \frac{1}{n} \sum_{u \in [n]} \mathbb{P}(B_r^{(G_n)}(u) \cong H_*) \\ \rightarrow \mathbb{P}(B_r^{(G,0)} \cong H_*) \quad (1)$$

$$[p^{(n)}(H_*)] = \frac{1}{n} \sum_{u \in [n]} \mathbb{1}[B_r^{(G_n)}(u) \cong H_*]$$

(b) $G_n \xrightarrow{\text{LW-p}} (G, 0)$ iff $\forall H_* \in \mathcal{G}_*$

$$p^{(n)}(H_*) \xrightarrow{p} \mathbb{P}(B_r^{(G,0)} \cong H_*) \quad (2)$$

Prms: (1) if G_n 's are deterministic,
then $p^{(n)}(H_*) = \mathbb{E}[p^{(n)}(H_*)]$

$$\Rightarrow G_n \xrightarrow{\text{LW-d}} (G, 0) \Leftrightarrow G_n \xrightarrow{\text{LW-p}} (G, 0).$$

(2) criteria holds for a.s. convergence also.

Proof: Follows from deterministic Criteria Thm.
(Ex- complete the details). $\square$

THM: Let $\mathcal{I}_* \subseteq \mathcal{G}_*$ be a subset of rooted graphs. Let $\mathcal{I}_*(r)$ be rooted graphs in $\mathcal{I}_*$ with depth/height at most r .

Assume $(\mathcal{G}, \rho)$ random graph $\Rightarrow P((\mathcal{G}, \rho) \in \mathcal{I}_*) \equiv 1$

Let $(G_n)_{n \geq 1}$ be a sequence of graphs (random/det)

Then $G_n \xrightarrow{LW-d} (\mathcal{G}, \rho)$ if (C1) holds $\forall H_* \in \mathcal{I}_*(r)$
& $r \geq 1$.

Also $G_n \xrightarrow{LWP} (\mathcal{G}, \rho)$ if (C2) holds $\forall r \geq 1$
& $\forall H_* \in \mathcal{I}_*(r)$.

Proof: T.S.T. $\forall H_* \notin \mathcal{I}_*(r), r \geq 1$

$$P(B_r^{(G_n)} \equiv H_*) \rightarrow 0 = P(B_r^{(\mathcal{G})}(\rho) \equiv H_*)$$

$\mathcal{G}_x(r)$ is countable

$\mathcal{G}_x(r)$ is countable $\forall \varepsilon > 0 \exists m$

$$\& \mathcal{G}_x(r, m) \subseteq \mathcal{G}_x(r) \Rightarrow |\mathcal{G}_x(r, m)| \leq m$$

$$\Rightarrow P(B_r^{(\mathcal{G})}(0) \in \mathcal{G}_x(r, m)) \geq 1 - \varepsilon$$

$$\lim_{n \rightarrow \infty} P(B_r^{(\mathcal{G}_n)}(0_n) \notin \mathcal{G}_x(r))$$

$$= (1 - \lim_{n \rightarrow \infty} P(B_r^{(\mathcal{G}_n)}(0_n) \in \mathcal{G}_x(r)))$$

$$\leq 1 - \lim_{n \rightarrow \infty} P(B_r^{(\mathcal{G}_n)}(0_n) \in \mathcal{G}_x(r, m))$$

$$= 1 - P(B_r^{(\mathcal{G})}(0) \in \mathcal{G}_x(r, m))$$

$$\left[\begin{array}{l} \text{(Since } \mathcal{G}_x(r, m) \text{ has only finitely many graphs)} \\ \& P(B_r(0_n) \cong H_x) \rightarrow P(B_r(0) \cong H_x) \\ \forall H_x \in \mathcal{G}_x(r, m) \end{array} \right]$$

$\leq \varepsilon$

$$\Rightarrow \lim_{n \rightarrow \infty} P(B_r^{(\mathcal{G}_n)}(0_n) \notin \mathcal{G}_x(r)) = 0$$

Ex: Extend the proof for LW-p.

THM: Let $G_n = G(n, p)$, $p = \frac{\lambda}{n}$, $\lambda < \infty$ be the ER random graph. Let $\mathcal{T}_\lambda$ be $\text{BBW}(\text{Bisect})$ offspring distribution. Then

$$G_n \xrightarrow{\text{LW-p}} \mathcal{T}_\lambda.$$

Proof: $E[p^{(n)}(t_*)] = \frac{1}{n} \sum_{u \in [n]} P(B_r^{(G_n)}(u) \cong t_*)$

$(B_r^{(G_n)}(u))$ are identically distributed in G_n

$$= P(B_r^{(G_n)}(u) \cong t_*)$$

(same in
prev.
classes)

$$\rightarrow P(B_r^{(\mathcal{T}_\lambda)}(\phi) \cong t_*) \text{ if}$$

t_* is a tree.

Choosing $\mathcal{T}_\lambda$ as rooted trees, we can

apply the previous theorem to conclude
that $G_n \xrightarrow{\text{LW-d}} \mathcal{T}_\lambda$.

let $T \in \mathcal{T}_\lambda(r)$, $r \geq 1$.

$$N_n(T) = n p^{(n)}(T) = \sum_{i \in [n]} \mathbb{1}[B_r^{(G_n)}(i) \cong T]$$

$$\frac{\mathbb{E}[N_n(T)]}{n} \rightarrow \mathbb{P}(B_r^{(\mathcal{T}_\lambda)}(\phi) \cong T) \quad (*)$$

we want to show

$$\frac{N_n(T)}{n} \xrightarrow{P} \downarrow \quad - (1)$$

CLT

$$\frac{\text{Var}(N_n(T))}{\mathbb{E}[N_n(T)]^2} \rightarrow 0. \quad - (2)$$

$$E(N_n(t)^2) = n P(B_r^{(G_n)}(i) \cong T)$$

$$+ n(n-1) P(B_r^{(G_n)}(i) \cong T, B_r^{(G_n)}(j) \cong T)$$

[by expanding $N_n(t)^2$ & using that $(B_r^{(G_n)}(i), B_r^{(G_n)}(j)) \stackrel{d}{=} (B_r^{(G_n)}(i), B_r^{(G_n)}(i))$ $i \neq j$.

We'll show

$$P(B_r^{(G_n)}(i) \cong T, B_r^{(G_n)}(j) \cong T)$$

$$\rightarrow P(B_r^{(G_n)}(i) \cong T)^2 \quad \text{--- ③}$$

$$\text{From ③} \quad \frac{Var(N_n(t))}{n^2} = \frac{P(B_r^{(G_n)}(i) \cong T)}{n}$$

$$+ P(B_r^{(G_n)}(i) \cong T, B_r^{(G_n)}(j) \cong T)$$

$$\rightarrow 0 \quad \Rightarrow \quad \text{②}$$

③ An $r=1$ case in $\boxed{A1}$.

$$\mathbb{P}(B_r^n(1) \cong T, B_r^n(2) \cong T)$$

①
1

$$= \mathbb{P}(B_r^n(1) \cong T, B_r^n(2) \cong T, d_{G_n}(1,2) > 2r) \\ + \mathbb{P}(B_r^n(1) \cong T, B_r^n(2) \cong T, d_{G_n}(1,2) \leq 2r)$$

②
+2

$$\mathbb{P}(\dots, d_{G_n}(1,2) \leq 2r)$$

$$\leq \mathbb{P}(B_r^n(1) \cong T, d_{G_n}(1,2) \leq 2r)$$

$$= \mathbb{E}[\mathbb{1}_{(B_r^n(1) \cong T)} \mathbb{1}_{[2 \in B_r^n(1)]}]$$

$$= \frac{1}{n+1} \mathbb{E}[\mathbb{1}_{(B_r^n(1) \cong T)} \sum_{j=2}^n \mathbb{1}_{[j \in B_r^n(1)]}]$$

(since $\mathbb{1}_{[j \in B_r^n(1)]}$ is id. distributed)

$$\leq \frac{1}{n} \mathbb{E}[|B_{2r}^n(1)|]$$

(From A2) $E[|B_{2r}^{(n)}(v)| | B_{2r}^{(n)}(u)|] \leq \lambda^k.$

$$\frac{1}{n} E[|B_{2r}^{(n)}(v)|] \leq \frac{1}{n} \sum_{k=0}^{2r} \lambda^k \rightarrow 0.$$

$$\Rightarrow \textcircled{f2} \rightarrow 0. \quad \textcircled{4}$$

$$\textcircled{f1} = P(B_r^n(u) \cong T, d_{G_n}(1,2) > 2r)$$

$$P(B_r^n(u) \cong T \mid B_r^n(u) \cong T, d_{G_n}(1,2) > 2r)$$

By $\textcircled{4}$ & $\textcircled{*}$

$$P(B_r^n(u) \cong T, d_{G_n}(1,2) > 2r)$$

$$\rightarrow P(B_r^{(n)}(u) \cong T)$$

Let $t = |V(T)|.$

$$B_r^{(n)}(u) \mid \{ B_r^{(n)}(u) \cong T, d_{G_n}(1,2) > 2r \}$$

$$\stackrel{d}{=} B_{r-t}^{G_n-t}(u) \quad \text{where } G_n-t \text{ is an}$$

$\textcircled{5}$

ER random graph on n vertices
with $p = \frac{\lambda}{n}$.

$$\Rightarrow P(B_r^{(G_n)}(2) \cong T \mid \dots \dots) \\ = P(B_r^{(n-t)}(2) \cong T) \rightarrow P(B_r^{(G_n)}(\emptyset) \cong T)$$

Shows that (T1) $\rightarrow P(B_r^{(G_n)}(\emptyset) \cong T)^2$

& so we have (3)

Explanation for (5):

$$P(B_r^{(G_n)}(2) \cong T \mid B_r^{(G_n)}(1) \cong T, d_{G_n}(1,2) > 2r) \\ = \sum_{\substack{1 \in F \subseteq [n] \\ |F|=k}} P(B_r^{(G_n)}(2) \cong T \mid B_r^{(G_n)}(1) \cong T, V_n = F, d_{G_n}(1,2) > 2r) \\ \times P(V_n = F \mid B_r^{(G_n)}(1) \cong T, d_{G_n}(1,2) > 2r) \\ [V_n = V(B_r^{(G_n)}(1))] \quad - (6)$$

$G_n^F = G_n \setminus F$ - ER graph on $V_n \setminus F$.

$$= \sum_{\substack{1 \in F \subseteq [n] \\ |F|=k}} P(B_r^{(G_n^F)}(2) \cong T \mid \dots \dots) P(\dots \mid \dots)$$

$$= \sum_{\substack{1 \leq i \leq n \\ |A|=k}} P(B_r^{(G_n^F)}(i) \cong T) P(\dots | \dots)$$

$$= \sum_{\dots} P(B_r^{(G_{n-t})}(1) \cong T) P(\dots | \dots) \\ (G_n^F \trianglelefteq G_{n-t})$$

$$= P(B_r^{(G_{n-t})}(1) \cong T)$$

$$\longrightarrow P(B_r^{(G)}(\phi) \cong T) \quad \blacksquare$$