

8/12 - L14 - MST COMPUTATIONS ON GW TREES

$$G_W(P) = P = (P(k))_{k \geq 0} = \text{pmf of}$$

the offspring random variable N .

Assume $P(0) = P(1) = 0$.

$$EN^2 = \sum_{k \geq 2} k^2 P(k) < \infty.$$

$$(T, \omega) \triangleq (G_W(P))$$

$$P(\hat{N}=k) = \frac{k P(k)}{\mathbb{E}[N]}, \quad k \geq 1.$$

$$\omega = (\omega(e))_{e \in E(T, \omega)} \text{ i.i.d. } \sim U[0, 1].$$

$$\tau_*(T) = \frac{1}{2} \mathbb{E} \left[\sum_{v \in V} \omega(v) \mathbf{1}_{\{(v, 0) \in \text{MSF}(T)\}} \right]$$

$$\text{MSF}(T) = \text{MSF}((T, \omega)).$$

[Assume $\exists G_n \ni (G_n, \omega) \xrightarrow{LW} (T, \omega, \omega)$
& $\frac{\tau(G_n)}{\sqrt{n}} \rightarrow \tau_*(T).$]

$$\begin{aligned} \mathbb{E} \tau_* &= \sum_{k \geq 1} k P(k) \mathbb{E} [\omega(0, 1) \mathbf{1}_{\{(0, 1) \in \text{MSF}(T)\}}] \\ &= \sum_{k \geq 1} k P(k) \int_0^1 t \mathbb{P}[(0, 1) \in \text{MSF}(T) | \omega(0, 1) = t, N=k] dt \end{aligned}$$

$$\begin{aligned} T'_{t'} &\stackrel{\text{def}}{=} \{e \in T : \omega(e) \leq t'\} \quad \uparrow \text{ & } T'_{t'} \text{ are} \\ T_t &\stackrel{\text{def}}{=} \{e \in T : \omega(e) \leq t\} \quad \text{independents} \end{aligned}$$

$$\begin{aligned} \textcircled{1} &= \sum_{k \geq 1} k P(k) \int_0^1 t (1 - \mathbb{P}(T'_{t'} = \infty)) \\ &= \sum_{k \geq 1} k P(k) \int_0^1 t (1 - \hat{\varphi}(t)) \mathbb{P}(T'_{t'} = \infty | N=k) dt \end{aligned}$$

$$\begin{aligned} \hat{\varphi}(t) &= 1 - \hat{\Phi}(t), \quad \hat{\Phi}(t) = \mathbb{P}_{\text{ext}}(\hat{T}_t) \\ \text{MTP} \Rightarrow & \sum_{k \geq 1} k P(k) \mathbb{P}(T'_{t'} = \infty | N=k) = \mathbb{E}N \cdot \hat{\varphi}(t) \end{aligned}$$

$$\begin{aligned} \varphi(t) &= \mathbb{E}N = \sum_{k \geq 1} k P(k); \quad \varphi(x) = \mathbb{E}[x^N] \\ \hat{\varphi}(x) &= \varphi(x) / \varphi'(x) = \hat{T} \end{aligned}$$

$$\begin{aligned} \textcircled{1} + \textcircled{2} \Rightarrow & \mathbb{E} \tau_* = \varphi(1) \int_0^1 t (1 - \hat{\varphi}(t)^2) dt \\ \hat{\varphi}(t), \text{ smallest root } r \Rightarrow & \hat{\varphi}(r) = r \\ \hat{\varphi}(t) &= 1 - \hat{\Phi}(1 - \hat{\varphi}(t)) \quad \hat{\Phi} = \mathbb{E}[e^{-\hat{T}^2}] \\ &= 1 - \hat{\Phi}(1 - t \hat{\varphi}(t)) \end{aligned}$$

$$\Rightarrow \tau_* = -\varphi(1) \int_0^1 (1-x) \ln(1 - \hat{\varphi}(x)) dx$$

[See Lemma 3.07 of Bordenave "Counting for full details" unimodular...]

$$1. (T, \omega) = T_d - d\text{-regular tree. } \varphi(1) = d, \quad \hat{\varphi}(x) = x^{d+1}$$

$$\Rightarrow \tau_*(T_d) = -d \int_0^1 (1-x) \ln(1 - x^{d+1}) dx.$$

$$2. (T, \omega) \triangleq G_W(\text{Poi}(\lambda)) \quad \hat{\varphi} = \varphi, \quad \varphi'(1) = \lambda$$

$$\tau_*(\lambda) = -\lambda \int_0^1 (1-x) \ln(1 - e^{-\lambda(1-x)}) dx$$

$$= -\int_0^\infty (\lambda x) \ln(1 - e^{-\lambda x}) dx$$

$$\textcircled{1} = \int_0^\infty t (1 - \hat{\varphi}(t)^2) dt$$

$$\begin{aligned} \hat{\varphi}(t) &= \mathbb{E}[e^{-\lambda t}] = 1 - \mathbb{E}[e^{-\lambda t}] \\ &= 1 - e^{-\lambda(1-\hat{\varphi}(t))} \end{aligned}$$

$$\frac{1}{2} \int_0^\infty t (1 - \hat{\varphi}(t)^2) dt = \zeta(3)$$

$$\begin{aligned} \text{Riemann-Zeta fn.} & \quad \text{by thinning property of Poisson nov.} \\ \text{Riemann-Zeta fn.} & \quad \text{Riemann-Zeta fn.} \end{aligned}$$

$$\begin{aligned} \text{Eg 5: } K_n - \text{complete graph. } \omega(e) - \text{i.i.d. } U[0, 1] \\ \mathbb{E}[\tau(K_n)]. \end{aligned}$$

$$\mathbb{E}[\tau(K_n)] = \mathbb{E}[\tau(G(n, t))] \quad \stackrel{\text{def}}{=} \mathbb{E}[\tau(G(n, t))]$$

$$K_n(t) = (K_n, \{e : \omega(e) < t\}) = G(n, t), \quad t \in [0, 1].$$

$$(u, v) \in \text{NST}(K_n) \Leftrightarrow u \leftrightarrow v \text{ in } G(n, \omega)$$

$$\mathbb{E}[\tau(G(n, t))] = \sum_{e \in K_n} \int_0^1 t \mathbb{P}(u \leftrightarrow v \text{ in } G(n, t)) dt$$

$$(t = \frac{s}{n}) \approx \frac{1}{n^2} \sum_{e \in K_n} \int_0^1 s \mathbb{P}(u \leftrightarrow v \text{ in } G(n, s/n)) ds$$

$$[\text{IP}(G(n, t) \text{ is connected}) \rightarrow 1 \text{ if } t = \frac{\log n + b_n}{n}, b_n \rightarrow 0]$$

$$\approx \frac{1}{n^2} \sum_{e \in K_n} \int_0^\infty s \mathbb{P}(u \leftrightarrow v \text{ in } G(n, s)) ds$$

$$G(n, s/n) = \frac{1}{2} \int_0^\infty t (1 - \hat{\varphi}(t)^2) dt$$

$$G(n, s/n) \rightarrow (T_s, \omega) \quad T_s = G_W(\text{Poi}(s))$$

$$\mathbb{P}(v \in B_r(u)) \rightarrow 0 \quad \forall r \geq 1$$

$$\Rightarrow T_s \text{ & } T'_s \text{ are indep. & disjoint } G_W(\text{Poi}(s)) \text{ trees}$$

$$\Rightarrow \{ (u, v) \notin \text{MSF}(G(n, s/n)) \}$$

$$\approx \{ (u, v) \notin \text{MSF}(G(n, s)) \}$$

$$= \{ |T_s| = |T'_{s'}| = \infty \}$$

$$\Rightarrow \mathbb{P}(u, v \in \text{MSF}(G(n, s/n))) \rightarrow \mathbb{P}(|T_s| = \infty)$$

$$= q(s)^2$$

$$\text{Freeze } \mathbb{E}[\tau(K_n)] \rightarrow \frac{1}{2} \int_0^\infty s (1 - q(s)^2) ds = \zeta(3).$$

$$[\text{LWC Proofs } (K_n, \omega) \quad \omega(e) \text{ i.i.d. EXP}(1).]$$

$$w_h(e) = n \omega(e) \triangleq \text{EXP}(n).$$

$$K_n^1 = (K_n, w_h) \quad \frac{1}{n} \tau(K_n^1) = \tau(K_n).$$

$$K_n^1 \xrightarrow{LW} \text{PWT}(1)$$

$$\text{Poisson weighted } \infty \text{ tree.}$$

$$S \text{ is a Poisson process on } \mathbb{R}_+ \text{ with mean } t$$

$$\text{i.e., } S_{t+h} - S_t \text{ are i.i.d. EXP}(1) \text{ r.v.'s.}$$

$$\mathbb{E}[\tau_*(\text{PWT})] = \frac{1}{2} \mathbb{E} \left[\sum_{v \in V} \omega(0, v) \mathbf{1}_{\{(0, v) \in \text{MSF}(\text{PWT})\}} \right]$$

$$[\text{See Pg 125 \& 126 }] \quad = \frac{1}{2} \int_0^\infty s \mathbb{P}((0, v) \in \text{MSF}(\text{PWT}) / w(0, v)) ds$$

$$= \frac{1}{2} \int_0^\infty s (1 - q(s)^2) ds = \zeta(3).$$

$$[\text{Big messages Tag model for combinatorial optimization problems. Put i.i.d. EXP}(n) weights on } K_n.]$$

$$\frac{\alpha_n}{n} = \min_{1 \leq h \leq K_n} \alpha(h) \rightarrow \alpha(\text{PWT})$$

$$\rightarrow G(n, b) - |M| - \text{size of maximal matching}$$

$$\frac{1}{n} \tau(G(n, b)) \rightarrow ? \quad \text{LWC theory in Ch. 4 of Bordenave's notes.}$$

$$\frac{1}{n} \tau(G(n, b)) \rightarrow ?? \quad \text{IM - Induced matching}$$

$$\text{Matchings in Hypergraphs.} \quad \rightarrow \boxed{\text{??}}$$

$$\text{I.P.d. weights. } \mathbb{E}[\min \text{ cost matching}] \rightarrow \zeta(2)$$

$$= \mathbb{E}[\dots] \rightarrow \boxed{\text{??}}$$

$$= \mathbb{E}[\dots] \rightarrow \boxed{\text{??}}$$

SUMMARY.

- ER graph has many regimes
- It "looks like" a tree "locally" in the sparse regime
- Formalized via UWC - Extends to weighted graphs

[Other Gg : Configuration Model, Hypercube percolation]

(See any of references) Assignment

- Limits are Unimodular Rho.
 - UWC \Rightarrow bounded fns on bounded degree graphs are estimable!
- \Rightarrow Size of largest conn. component under some assumptions
- \rightarrow Aldous-Steele Cty thm for MST.
When limiting tree is GW, we can compute the limit.

See Bondarenko - "Counting" -- "for more eg. of non-trivial estimable functions."

Is Spec(A_G) estimable?

Adj. matrix

LWC Link:

$$(K_n, w) \xrightarrow{\text{LWC}} ?$$

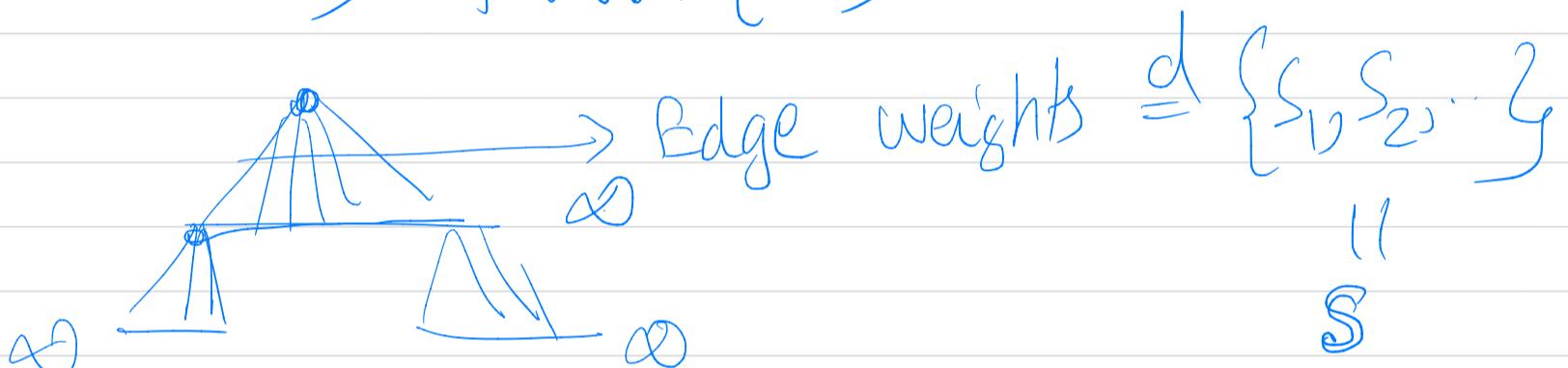
let $w(e)$ i.i.d. EXP(1).

$$P(w(e) \geq x) = e^{-x}, x \geq 0$$

$$P(w_h(e) \geq n^{-1}x) = e^{-x}, w_h(e) = nw(e).$$

$\Rightarrow w_h(e)$ i.i.d. EXP(n)

$$(K_n, w_n) \xrightarrow{\text{LWC}} \text{PWIT}(x)$$



S = Poisson process on \mathbb{R}_+ .

i.e., $S_0=0$, $S_{i+1}-S_i$ are i.i.d. EXP(1).

$$N_t = |S \cap [0, t]| \stackrel{\text{i.i.d.}}{\equiv} \text{Poi}(t) \sim N_0$$

$\{U_1, \dots, U_N\}$, U_i i.i.d. $U[0, t]$

$$\mathbb{E}[T_{\text{PWIT}}] = \frac{1}{2} \sum_{v \in V} \sum_{v' \in V} w(v, v') \mathbb{1}_{[(0, v), (0, v')] \in \text{NSF(PWIT)}}$$

$$= \frac{1}{2} \int_0^\infty s P(0, v) \text{NSF(PWIT)} |w(0, v) = s| ds$$

$$w(0,\theta) = \delta$$

$$(0,\theta) \in \text{NST}(\text{PWT}) \Leftrightarrow |(\text{PWT}(s), 0)| < \delta \text{ or } |(\text{PWT}(s), \theta)| < \delta$$

$$\Rightarrow P((0,\theta) \in \text{NSF}(\text{PWT})) = (1 - P(|\text{PWT}(s), \theta| = \delta))$$

$$(\text{PWT}(s), 0) \stackrel{d}{=} G_W(\text{Poi}(\delta))$$

$$\text{Since } |S_n(0, \delta)| \stackrel{d}{=} \text{Poi}(\delta)$$

$$I((k, \omega)) = \frac{1}{n} I((k_n, \omega_n))$$

$$E[I((k, \omega))] = \frac{1}{n} E[I((k_n, \omega_n))] \rightarrow E[I_{\text{PWT}}]$$

$$= q(3).$$