

Probabilistic Symmetry and Network Models

Lecture 2

Harry Crane

October 10, 2019

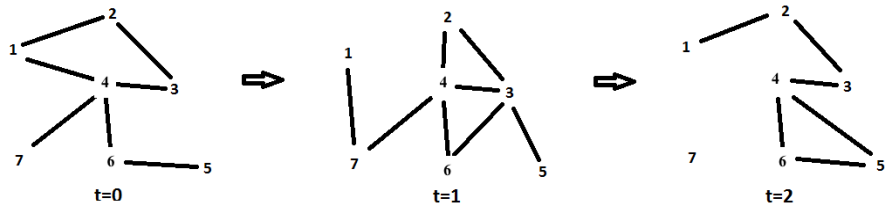
- Lecture 1: Basic symmetries and network sampling.
 - H. Crane and W. Dempsey. (2018). Edge exchangeable models for interaction networks. *Journal of the American Statistical Association*.
 - H. Crane and W. Dempsey. (2019). Relational exchangeability. *Journal of Applied Probability*.
 - H. Crane and H. Towsner. (2018). Relatively exchangeable structures. *Journal of Symbolic Logic*.
- **Lecture 2: Dynamic network models.**
 - H. Crane. (2018). *Probabilistic Foundations of Statistical Network Analysis*.
 - H. Crane. (2015). Time-varying network models. *Bernoulli*, **21**(3):1670–1696.
 - H. Crane. (2016). Dynamic random networks and their graph limits. *Annals of Applied Probability*.
 - H. Crane. (2017). Exchangeable graph-valued Feller processes. *Probability Theory and Related Fields*.
 - H. Crane. (2018). Combinatorial Lévy processes. *Annals of Applied Probability*.
 - H. Crane and H. Towsner. (2019+). The structure of combinatorial Markov processes.



Book website: <http://www.harrycrane.com/networks.html>

Time-varying network models

- $\{0, 1\}^{\mathbb{N} \times \mathbb{N}}$: graphs with vertex set $\mathbb{N} = \{1, 2, \dots\}$.
- $\{0, 1\}^{n \times n}$: graphs with vertex set $[n] := \{1, \dots, n\}$.



Basic assumptions:

$(\Gamma_t)_{t \geq 0}$ is a Markov process on $\{0, 1\}^{\mathbb{N} \times \mathbb{N}}$ satisfying

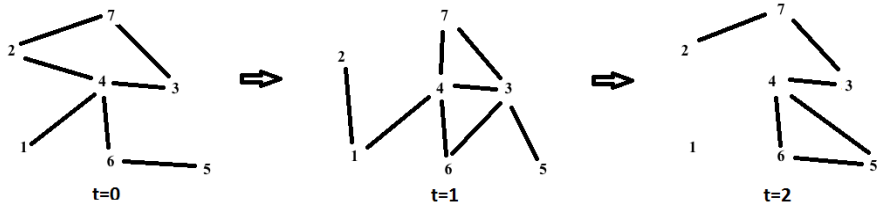
- exchangeability: for all $\sigma : \mathbb{N} \rightarrow \mathbb{N}$, the transition probabilities satisfy

$$\mathbb{P}\{\Gamma_{t+1} \in \cdot \mid \Gamma_t = G\} = \mathbb{P}\{\Gamma_{t+1}^\sigma \in \cdot \mid \Gamma_t^\sigma = G\}, \quad \text{for all } G \in \{0, 1\}^{\mathbb{N} \times \mathbb{N}}.$$

- projective Markov property: $(\Gamma_t|_{[n]})_{t \geq 0}$ is a Markov chain on $\{0, 1\}^{n \times n}$, for every $n = 1, 2, \dots$

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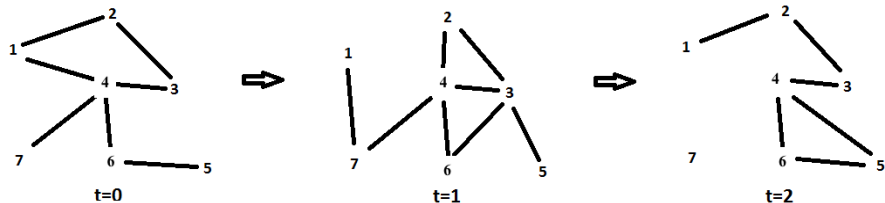
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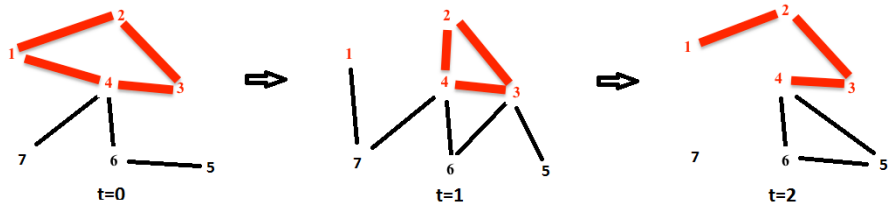
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Exchangeable, Feller \iff Exchangeable, projective

The **semigroup** $(\mathbf{P}_t)_{t \in \mathcal{T}}$ of a Markov process Γ acts on bounded measurable functions $g : \{0, 1\}^{\mathbb{N} \times \mathbb{N}} \rightarrow \mathbb{R}$ by

$$\mathbf{P}_t g(G) := \mathbb{E}(g(\Gamma_t) \mid \Gamma_0 = G), \quad G \in \{0, 1\}^{\mathbb{N} \times \mathbb{N}}.$$

Definition (Feller property)

We say that Γ possesses the **Feller property** if for all bounded, continuous $g : \{0, 1\}^{\mathbb{N} \times \mathbb{N}} \rightarrow \mathbb{R}$,

- (i) $G \mapsto \mathbf{P}_t g(G)$ is continuous for every $t > 0$ and
- (ii) $\lim_{t \downarrow 0} \mathbf{P}_t g(G) = g(G)$ for all $G \in \{0, 1\}^{\mathbb{N} \times \mathbb{N}}$.

Theorem

The following are equivalent for any exchangeable Markov process Γ on $\{0, 1\}^{\mathbb{N} \times \mathbb{N}}$.

- (i) Γ has the projective Markov property.
- (ii) Γ has the Feller property.

Proof:

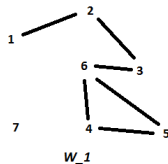
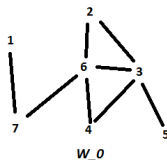
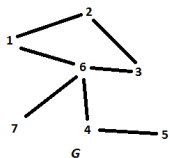
(i) \Rightarrow (ii): Compactness + Stone–Weierstrass theorem.

(ii) \Rightarrow (i): Use definition of Feller on test functions $\psi_F(G) := \mathbf{1}(G|_{[n]} = F)$.

Rewiring operator

A $\{0, 1\} \times \{0, 1\}$ -valued array $W = (W_{ij})_{i,j \geq 1}$ defines a map $\{0, 1\}^{\mathbb{N} \times \mathbb{N}} \rightarrow \{0, 1\}^{\mathbb{N} \times \mathbb{N}}$,
 $G \mapsto W(G) = (G'_{ij})_{i,j \geq 1}$ with

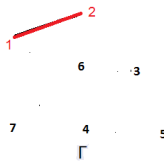
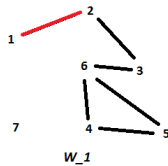
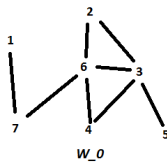
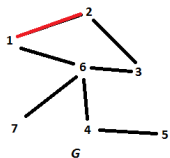
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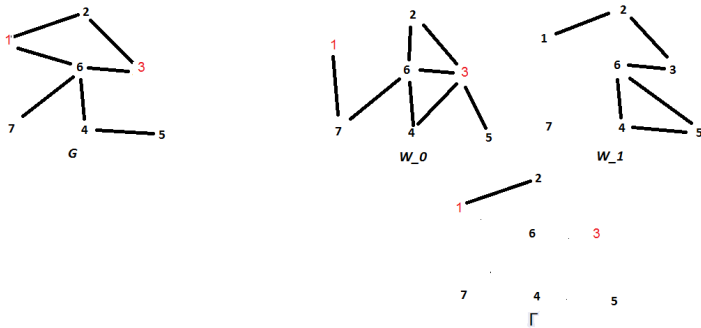
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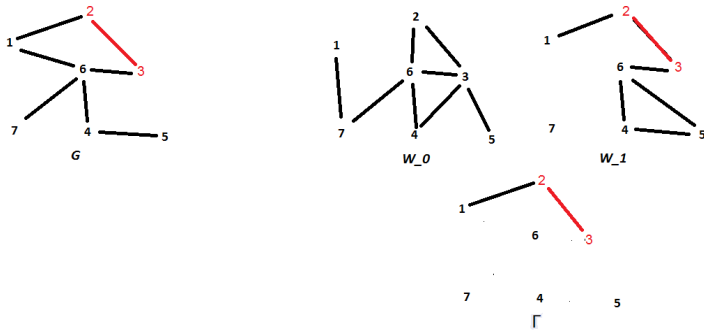
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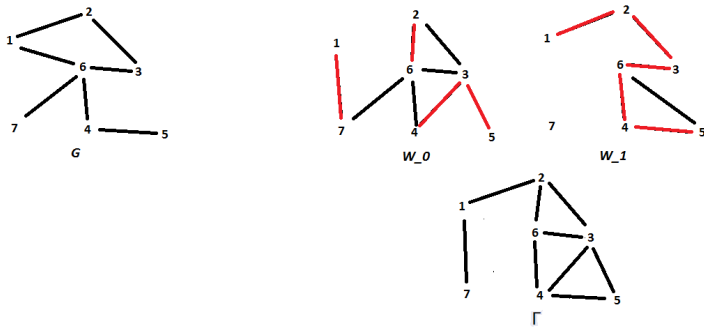
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- $G \mapsto W(G)$ is Lipschitz continuous.

Exchangeable rewiring chain

Random W defines a transition probability on $\{0, 1\}^{\mathbb{N} \times \mathbb{N}}$ by

$$P(G, \cdot) := \mathbb{P}(\{W : W(G) \in \cdot\}), \quad G \in \{0, 1\}^{\mathbb{N} \times \mathbb{N}}.$$

Let ω be an exchangeable probability measure on rewiring arrays.

For initial state $G \in \{0, 1\}^{\mathbb{N} \times \mathbb{N}}$, construct $\Gamma_G = (\Gamma_m)_{m \geq 0}$ by

- 1 taking W_1, W_2, \dots i.i.d. ω and, given W_1, W_2, \dots , putting
- 2 $\Gamma_0 = G$ and
- 3 $\Gamma_{m+1} = W_{m+1}(\Gamma_m)$ for $m \geq 0$.

Define the ω -rewiring chain with initial distribution ν by $\Gamma^{\omega, \nu} = \Gamma_G$, for G chosen from initial distribution ν .

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Proposition

For ν an exchangeable distribution on $\{0, 1\}^{\mathbb{N} \times \mathbb{N}}$, $\Gamma^{\omega, \nu}$ is exchangeable and has the Feller property.

Proof.

Exercise or see H. Crane. (2015). Time-varying network models. *Bernoulli*. □

Key claim: the converse holds.

Characterization of graph-valued Markov processes

Let $(\Gamma_m)_{m=0,1,\dots}$ be a discrete-time Markov chain on $\{0,1\}^{\mathbb{N} \times \mathbb{N}}$ (countable graphs) which is

- **exchangeable:** $(\Gamma_m)_{m \geq 0} =_{\mathcal{D}} (\Gamma_m^\sigma)_{m \geq 0}$ for all permutations $\sigma : \mathbb{N} \rightarrow \mathbb{N}$.
- **projective:** $(\Gamma_m|_{[n]})_{m \geq 0}$ is a Markov chain for all $n \geq 1$.

Theorem (Crane 2017)

Then there exists an exchangeable probability measure ω on $\{0,1\} \times \{0,1\}$ -valued arrays so that $(\Gamma_m) =_{\mathcal{D}} (\Gamma_m^*)$ with $\Gamma_0^* =_{\mathcal{D}} \Gamma_0$ and for each $m \geq 1$

$$\Gamma_m^* = W_m(\Gamma_{m-1}^*) = (W_m \circ \dots \circ W_1)(\Gamma_0^*), \quad m \geq 1,$$

for W_1, W_2, \dots i.i.d. ω and $G' = W(G)$ defined by

$$G'_{ij} = \begin{cases} W_{ij}^0, & G_{ij} = 0, \\ W_{ij}^1, & G_{ij} = 1. \end{cases}$$

Proof.

Exchangeable and Feller $\Rightarrow \Gamma_{t+1}$ is relatively exchangeable with respect to Γ_t for each t . Use representation for relatively exchangeable graphs to construct ω . □

$\Gamma = (\Gamma_{ij})_{i,j \geq 1}$: adjacency array of a random graph with vertex set \mathbb{N} .

Definition (Exchangeable random graph)

Γ is exchangeable if $\Gamma^\sigma = (\Gamma_{\sigma(i)\sigma(j)})_{i,j \geq 1} \stackrel{\mathcal{D}}{=} \Gamma$ for all permutations $\sigma : \mathbb{N} \rightarrow \mathbb{N}$.

Theorem (Aldous–Hoover)

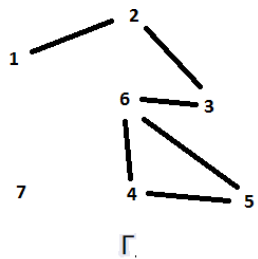
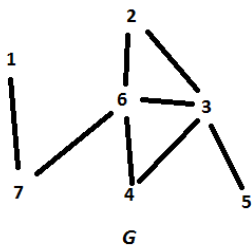
There exists a measurable function $\phi : [0, 1]^4 \rightarrow \{0, 1\}$ such that $\Gamma \stackrel{\mathcal{D}}{=} \Gamma^* = (\Gamma_{ij}^*)_{i,j \geq 1}$ with

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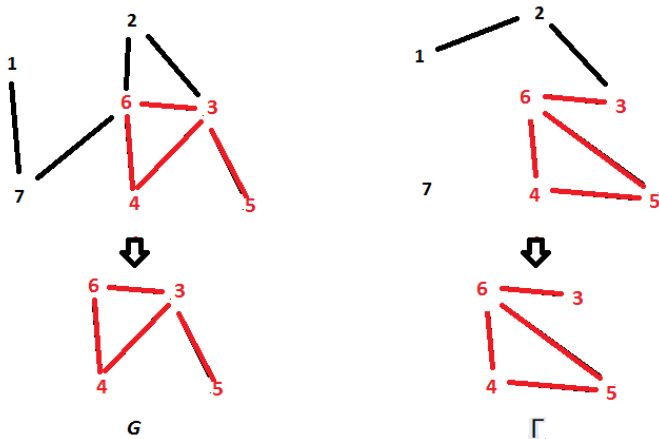
Definition (Relatively exchangeable random graph)

Γ is relatively exchangeable with respect to G if, for all $S \subseteq \mathbb{N}$, $\Gamma|_S^\sigma \stackrel{D}{=} \Gamma|_S$ for all automorphisms σ of $G|_S$.



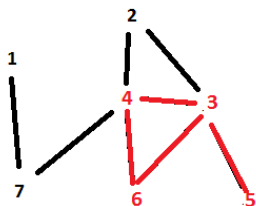
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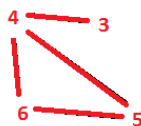
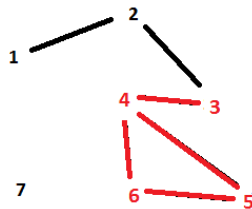


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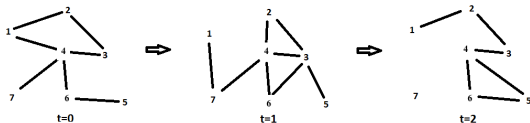
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G



Γ



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- exchangeability: for all $\sigma : \mathbb{N} \rightarrow \mathbb{N}$, $(\Gamma_t^\sigma)_{t \geq 0}$ has same finite-dimensional distributions as Γ .
- projective Markov property: $(\Gamma_t|_{[n]})_{t \geq 0}$ is a Markov chain on $\{0, 1\}^{n \times n}$, for every $n = 1, 2, \dots$

Transition from G to Γ :

Observation

Exchangeability and projectivity \implies for all $S \subseteq \mathbb{N}$, the conditional distribution of $\Gamma|_S$ given G is invariant under automorphisms of $G|_S$.

Theorem (Crane (2017))

$G = (\mathbb{N}; E)$ an undirected graph* and Γ a random graph relatively exchangeable with respect to G . Then there exists $\phi : \{0, 1\} \times [0, 1]^4 \rightarrow \{0, 1\}$ such that

$\Gamma \stackrel{\mathcal{D}}{=} \Gamma^* = (\Gamma_{ij}^*)_{i,j \geq 1}$ with

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Application:

- Generate a pair of graphs $W = (W_0, W_1)$ (jointly exchangeable) by

$$W_0(i, j) = \phi(0, \xi_{\emptyset}, \xi_{\{i\}}, \xi_{\{j\}}, \xi_{\{i,j\}}) \quad \text{and}$$

$$W_1(i, j) = \phi(1, \xi_{\emptyset}, \xi_{\{i\}}, \xi_{\{j\}}, \xi_{\{i,j\}}).$$

- $W = (W_0, W_1)$ defines a random operator $\{0, 1\}^{\mathbb{N} \times \mathbb{N}} \rightarrow \{0, 1\}^{\mathbb{N} \times \mathbb{N}}$ by $G \mapsto G' = W(G)$ with

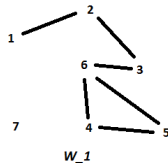
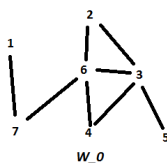
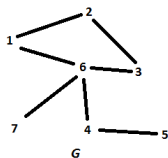
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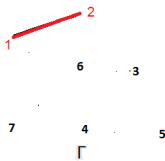
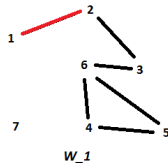
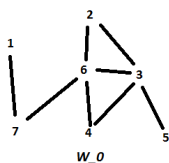
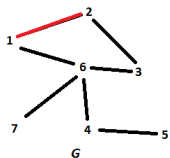
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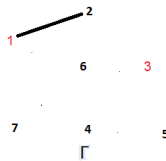
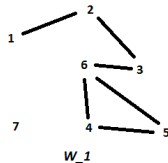
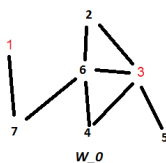
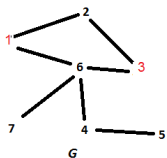
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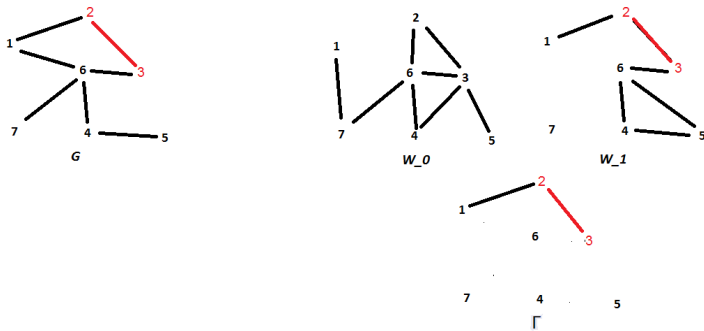


Theorem (Crane (2017))

$G = (\mathbb{N}; E)$ an undirected graph* and Γ a random graph relatively exchangeable with respect to G . Then there exists $\phi : \{0, 1\} \times [0, 1]^4 \rightarrow \{0, 1\}$ such that $\Gamma \stackrel{\mathcal{D}}{=} \Gamma^* = (\Gamma_{ij}^*)_{i,j \geq 1}$ with $\Gamma_{ij}^* = \phi(G_{ij}, U_\emptyset, U_{\{i\}}, U_{\{j\}}, U_{\{i,j\}})$ for $i, j \geq 1$.

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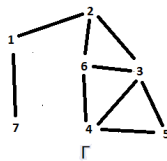
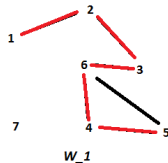
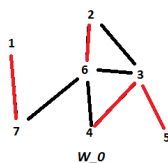
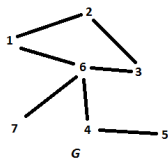
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Application:

- Generate a pair of graphs (W_0, W_1) (jointly exchangeable).



- Let $\Gamma = (\Gamma(t))_{t=0,1,\dots}$ be exchangeable and have the projective Markov property.
- Draw G from Erdős–Rényi(1/2) and consider $\mathbb{P}\{\Gamma(t+1) \in \cdot \mid \Gamma(t) = G\}$.
- Given $\Gamma(t) = G$, $\Gamma(t+1)$ is relatively exchangeable with respect to G .
 $\implies \phi : \{0, 1\} \times [0, 1]^4 \rightarrow \{0, 1\}$ such that

$$\Gamma_{ij}(t+1) =_{\mathcal{D}} (\phi(G_{ij}, \xi_{\emptyset}, \xi_{\{i\}}, \xi_{\{j\}}, \xi_{\{i,j\}}))_{i,j \geq 1}.$$

- Now construct $W = (W_{ij})_{i,j \geq 1}$ with $W_{ij} = (W_{ij}(0), W_{ij}(1))$ by

$$W_{ij}(0) = \phi(0, \xi_{\emptyset}, \xi_{\{i\}}, \xi_{\{j\}}, \xi_{\{i,j\}})$$

$$W_{ij}(1) = \phi(1, \xi_{\emptyset}, \xi_{\{i\}}, \xi_{\{j\}}, \xi_{\{i,j\}}).$$

Let ω be the induced probability law of W .

- For any $G' \in \{0, 1\}^{\mathbb{N} \times \mathbb{N}}$, there exists (w.p. 1) $\psi_{G',G} : \mathbb{N} \rightarrow \mathbb{N}$ such that $G^{\psi_{G',G}} = G'$.
- Given $W \sim \omega$ and any $G' \in \{0, 1\}^{\mathbb{N} \times \mathbb{N}}$, we have

$$\begin{aligned} \mathbb{P}\{W(G') \in \cdot\} &= \mathbb{P}\{\Gamma^{\psi_{G',G}}(t+1) \in \cdot \mid \Gamma(t) = G\} \\ &= \mathbb{P}\{\Gamma^{\psi_{G',G}}(t+1) \in \cdot \mid \Gamma^{\psi_{G',G}}(t) = G'\} = P(G', \cdot) \end{aligned}$$

by the exchangeability & Feller property.

- Let ω be an exchangeable σ -finite measure on $\mathcal{W}_{\mathbb{N}}$, i.e.,

$$\omega(\{\mathbf{Id}_{\mathbb{N}}\}) = \mathbf{0} \quad \text{and} \quad \omega(\{\mathbf{W} \in \mathcal{W}_{\mathbb{N}} : \mathbf{W}|_{[2]} \neq \mathbf{Id}_{[2]}\}) < \infty.$$

- Let $\mathbf{W} := \{(t, W_t)\} \subseteq (0, \infty) \times (\{0, 1\} \times \{0, 1\})^{\mathbb{N}}$ be a Poisson point process with intensity $dt \otimes \omega$.
- Given \mathbf{W} , construct $\Gamma_{\omega}^* = (\Gamma^*(t))_{t \geq 0}$ with initial state $G \in \{0, 1\}^{\mathbb{N} \times \mathbb{N}}$ by putting $\Gamma^*(0) = G$ and for each $n \geq 1$ and $t > 0$,
 - $\Gamma_t^{*[n]} = W_t^{[n]}(\Gamma_{t-}^{*[n]})$, if $t > 0$ is an atom of \mathbf{W} for which $W_t^{[n]} \neq \mathbf{Id}_{[n]}$, or
 - $\Gamma_t^{*[n]} = \Gamma_{t-}^{*[n]}$, otherwise.

Theorem

Γ_{ω}^* constructed above is exchangeable and Feller.

Key claim: Converse holds.

Theorem (Crane 2017)

$\Gamma = (\Gamma_t)_{t \geq 0}$ an exchangeable, projective Markov process on $\{0, 1\}^{\mathbb{N} \times \mathbb{N}}$. Then there exists an exchangeable measure ω satisfying

$$\omega(\{\mathbf{Id}_{\mathbb{N}}\}) = 0 \quad \text{and} \quad \omega(\{W \in \mathcal{W}_{\mathbb{N}} : W|_{[2]} \neq \mathbf{Id}_{[2]}\}) < \infty$$

such that $\Gamma =_{\mathcal{D}} \Gamma_{\omega}^*$ as constructed above.

Conditions on ω :

- (identifiability) $\omega(\{\mathbf{Id}_{\mathbb{N}}\}) = 0$: $\mathbf{Id}_{\mathbb{N}}(G) = G$ for all G .
- (σ -finite) $\omega(\{W \in \mathcal{W}_{\mathbb{N}} : W|_{[2]} \neq \mathbf{Id}_{[2]}\}) < \infty$: ensures

$$\begin{aligned} \omega(\{W|_{[n]} \neq \mathbf{Id}_{[n]}\}) &= \omega\left(\bigcup_{1 \leq i, j \leq n} \{W|_{\{i, j\}} \neq \mathbf{Id}_{\{i, j\}}\}\right) \\ &\leq \sum_{1 \leq i, j \leq n} \omega(\{W|_{\{i, j\}} \neq \mathbf{Id}_{\{i, j\}}\}) \\ &= n^2 \omega(W|_{[2]} \neq \mathbf{Id}_{[2]}) < \infty. \end{aligned}$$

Theorem (C, 2017)

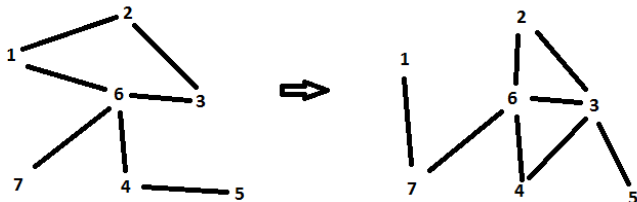
The measure ω above can be characterized by unique constants $\mathbf{e}_0, \mathbf{e}_1, \mathbf{v} \geq 0$, a unique probability measure Σ on 2×2 stochastic matrices, and a unique measure Υ on the space of graph limits satisfying

$$\Upsilon(\{\mathbf{I}\}) = 0 \quad \text{and} \quad \int (1 - v_*^{(2)}) \Upsilon(d\nu) < \infty,$$

such that

$$\omega = \underbrace{\Omega \Upsilon}_{\mu_I} + \mathbf{v} \Omega \Sigma + \mathbf{e}_0 \epsilon_0 + \mathbf{e}_1 \epsilon_1.$$

(I) **global jump**: a positive fraction of all edges change



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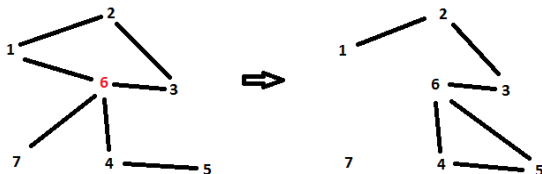
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- (II) **single-vertex jump**: a positive fraction of edges incident to a single vertex change, everything else stays the same



Theorem (C, 2017)

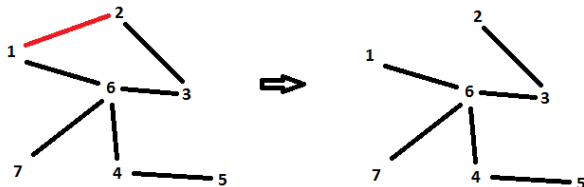
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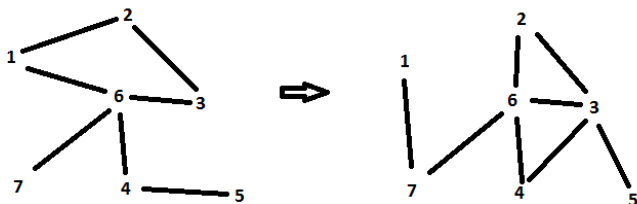
(III) **single-edge flip**: a single edge changes, everything else stays the same.



Theorem (Crane 2016)

Let $\Gamma = (\Gamma_t)_{t \geq 0}$ be an exchangeable càdlàg Markov process on $\{0, 1\}^{\mathbb{N} \times \mathbb{N}}$. Then there are three types of discontinuity:

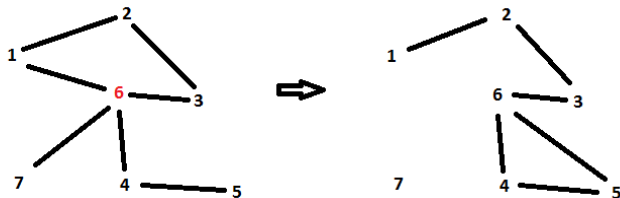
- (I) **global jump**: a positive fraction of all edges changes;
- (II) **single-vertex jump**: a positive fraction of edges incident to a single vertex change, everything else stays the same;
- (III) **single-edge flip**: a single edge changes, everything else stays the same.



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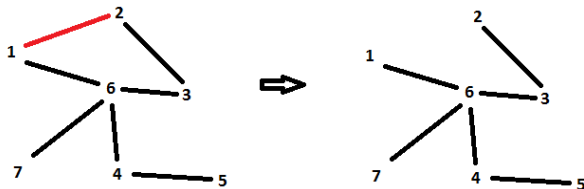
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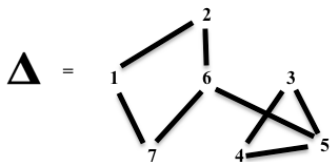
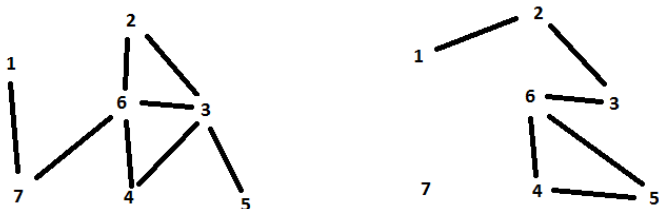


Combinatorial Lévy Processes

Combinatorial Lévy processes (CLPs)

Define the **increment between graphs** G and G' in $\{0, 1\}^{n \times n}$ ($n = 1, 2, \dots, \infty$) by $\Delta(G, G') \equiv (\Delta_{G, G'}(i, j))_{1 \leq i, j \leq n}$, where

$$\Delta_{G, G'}(i, j) = |G_{ij} - G'_{ij}| = \begin{cases} 0, & G_{ij} = G'_{ij}, \\ 1, & \text{otherwise.} \end{cases}$$



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$$\Delta_{G, G'}(i, j) = |G_{ij} - G'_{ij}| = \begin{cases} 0, & G_{ij} = G'_{ij}, \\ 1, & \text{otherwise.} \end{cases}$$

Definition (CLP)

A **graph-valued Lévy process** $\Gamma = (\Gamma_t)_{t \geq 0}$ on $\{0, 1\}^{n \times n}$ ($n = 1, 2, \dots, \infty$) satisfies

- $\Gamma_0 = \emptyset$ (the empty graph),
- stationary increments: for all $s, t \geq 0$, the increment $\Delta(\Gamma_s, \Gamma_{s+t}) \stackrel{D}{=} \Gamma_t$.
- independent increments: for all $t_0 < t_1 < \dots < t_n < \infty$, the increments $\Delta(\Gamma_{t_1}, \Gamma_{t_0}), \dots, \Delta(\Gamma_{t_n}, \Gamma_{t_{n-1}})$ are independent.
- càdlàg paths: Γ has càdlàg paths in the product-discrete topology on $\{0, 1\}^{\mathbb{N} \times \mathbb{N}}$.

Graph-valued random walk

Let μ be a probability distribution on $\{0, 1\}^{\mathbb{N} \times \mathbb{N}}$ and define $\Gamma = (\Gamma_t)_{t=0,1,\dots}$ with $\Gamma_0 = G$ by

$$\Gamma_{t+1} = \Gamma_t \triangle D_t = \Gamma_{t-1} \triangle D_{t-1} \triangle D_t = G \triangle D_0 \triangle \dots \triangle D_t,$$

for D_0, D_1, \dots i.i.d. μ .

Theorem

Any discrete time graph-valued Lévy process is uniquely determined by its initial distribution and the increment measure μ given above.

Open Problem

Study the convergence rate of CLPs and/or rewiring chains to their stationary distribution. Prove the cutoff phenomenon.

See related work:

H. Crane and S.P. Lalley. (2013). Convergence rates of Markov chains on spaces of partitions. Electronic Journal of Probability.

Continuous-time CLPs (Poissonian construction)

Let μ be a measure on $\{0, 1\}^{\mathbb{N} \times \mathbb{N}}$ such that

$$\mu(\{\mathbf{0}_{\mathbb{N}}\}) = 0 \quad \text{and} \quad \mu(\{G \in \{0, 1\}^{\mathbb{N} \times \mathbb{N}} : G|_{[n]} \neq \mathbf{0}_{[n]}\}) < \infty \quad \text{all } n \geq 1.$$

Let $\mathbf{D} := \{(t, D_t)\}$ be a Poisson point process on $(0, \infty) \times \{0, 1\}^{\mathbb{N} \times \mathbb{N}}$ with intensity $dt \otimes \mu$.

Construct $\Gamma^* = (\Gamma_t^*)_{t \geq 0}$ on $\{0, 1\}^{\mathbb{N} \times \mathbb{N}}$ by putting $\Gamma_t^{*[n]} = \mathbf{0}_{[n]}$ for each $n \geq 1$ and

- $\Gamma_t^{*[n]} = \Gamma_{t-}^{*[n]} \triangle D_t$ for all atoms (t, D_t) in \mathbf{D} for which $D_t|_{[n]} \neq \mathbf{0}_{[n]}$.
- $\Gamma_t^{*[n]} = \Gamma_{t-}^{*[n]}$ otherwise.

Theorem (Crane, 2018)

If μ is exchangeable, then it decomposes as

$$\mu = \mu_{(2)} + \mu_{(1,1)} + \mu_{(1)} + \mu_{\emptyset},$$

where

- $\mu_{(2)}$ governs jumps of a single loop.
- $\mu_{(1,1)}$ governs jumps of a single edge (off-diagonal).
- $\mu_{(1)}$ governs jumps of a single vertex.
- μ_{\emptyset} governs global jumps.

Theorem (C. 2016; C. 2017)

Let Γ be an exchangeable Markov process on $\{0, 1\}^{\mathbb{N} \times \mathbb{N}}$ (with the Feller property). Then the projection $\|\Gamma\| := (\|\Gamma_t\|)_{t \geq 0}$ into the space of graph limits \mathcal{D}^* exists a.s. (and has the Feller property) with respect to the topology induced by

$$d(D, D') := 2^{-n} \sum_{n \geq 1} \sum_{F, F' \in \{0, 1\}^{n \times n}} |D(F) - D(F')|, \quad D, D' \in \mathcal{D}^*.$$

Open Problem

Construct and study an analog to “Brownian motion” in the space of graph limits.

- S. Athreya, F. den Hollander and A. Röllin. (2019). Graphon-valued stochastic processes from population genetics. *arXiv:1908.06241*.

Open Problem

Study projection of graph-valued Lévy processes into the space of graph limits.

- Is there a suitable notion of Lévy process in this space?
- How does it relate to graph-valued Lévy processes?
- Address the problem for general combinatorial Lévy processes.

$(\Gamma_t)_{t \geq 0}$ is a Markov process on $\{0, 1\}^{\mathbb{N} \times \mathbb{N}}$ satisfying

- exchangeability: for all $\sigma : \mathbb{N} \rightarrow \mathbb{N}$, the transition probabilities satisfy

$$\mathbb{P}\{\Gamma_{t+1} \in \cdot \mid \Gamma_t = G\} = \mathbb{P}\{\Gamma_{t+1}^\sigma \in \cdot \mid \Gamma_t^\sigma = G\}, \quad \text{for all } G \in \{0, 1\}^{\mathbb{N} \times \mathbb{N}}.$$

- projective Markov property: $(\Gamma_t|_{[n]})_{t \geq 0}$ is a Markov chain on $\{0, 1\}^{n \times n}$, for every $n = 1, 2, \dots$

Theorem (Discrete-time processes, C. 2017)

Then there exists an exchangeable probability measure ω on $\{0, 1\} \times \{0, 1\}$ -valued arrays so that $(\Gamma_m) =_{\mathcal{D}} (\Gamma_m^*)$ with $\Gamma_0^* =_{\mathcal{D}} \Gamma_0$ and for each $m \geq 1$

$$\Gamma_m^* = W_m(\Gamma_{m-1}^*) = (W_m \circ \dots \circ W_1)(\Gamma_0^*), \quad m \geq 1,$$

for W_1, W_2, \dots i.i.d. ω and $G' = W(G)$ defined by

$$G'_{ij} = \begin{cases} W_{ij}^0, & G_{ij} = 0, \\ W_{ij}^1, & G_{ij} = 1. \end{cases}$$

$(\Gamma_t)_{t \geq 0}$ is a Markov process on $\{0, 1\}^{\mathbb{N} \times \mathbb{N}}$ satisfying

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Theorem (Lévy–Itô–Khintchine representation, C. 2017)

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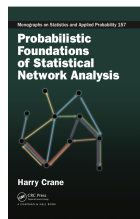
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* More information available at www.harrycrane.com/networks.html