

LARGE DEVIATION PRINCIPLES FOR EMPIRICAL MEASURES OF MULTITYPE RANDOM NETWORKS

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OUTLINE

- Multitype Random Network(MRN) Model;
Fienberg, Meyer and Wasserman(1985).
- Empirical Measures on Multitype Random Networks;
Doku-Amponsah (2006).
- Statement and Discussion of Results.
- Further Work
Hellman and Staudigl(2014)

MULTITYPE RANDOM NETWORK MODEL

Functions: $\kappa_n, \ell_n : \Omega \times \Omega \rightarrow [0, \infty]$ and $\kappa, \ell : \Omega \times \Omega \rightarrow [0, \infty]$

Type Law: $\eta : \Omega \rightarrow (0, 1)$

Define **Multitype Random Network** X with n sites as follows:

- Partition the set of sites, $[n] = \{1, 2, 3, \dots, n\}$, into finitely many types $\Omega = \{a_1, a_2, a_3, \dots, a_m\}$ independently according to the type law η .
- For a couple of sites (i, j) take the intensities of link creation and link destruction to be $w_{ij}^{(n)}(a, b) = (1 - A_{ij})\kappa_n(a, b)$,
 $\nu_{ij}^{(n)}(a, b) = A_{ij}\ell_n(a, b)$,
- $W^{(n)}$ is called the *attachment mechanism* while $V^{(n)}$ is called the *volatility mechanism*.

TYPES OF MULTITYPE RANDOM NETWORK

- $X = X(\kappa_n, \ell_n) := \{(X(i), \in \Omega), E\}$ under the joint law of the type and network,
- $X(i)$: the type of site i and X as MRN.
- X is called *symmetric MRN* if both functions κ_n and ℓ_n are *symmetric*
- Otherwise we call it *asymmetric MRN*.
- MRN = coloured random graph with edge probability is

$$p_n(a, b) = \frac{\kappa_n(a, b)}{\kappa_n(a, b) + \ell_n(a, b)}.$$

See Hellmann and Staudigl(2014).

EMPIRICAL TYPE AND CO-OPERATIVE MEASURES OF THE MRN

- **Empirical type measure:**

$$L^1(a) := \frac{1}{n} \sum_{i \in [n]} \delta_{X(i)}(a), \quad a \in \Omega.$$

- **Empirical cooperative measure on $\Omega \times \Omega$**

Symmetric X :

$$L^2(a, b) = \frac{1}{n} \sum_{(i,j) \in E} [\delta_{(X(i), X(j))} + \delta_{((X(j), X(i)))}](a, b)$$

Asymmetric X :

$$L^2(a, b) = \frac{2}{n} \sum_{(i,j) \in E} \delta_{(X(i), X(j))}(a, b).$$

EMPIRICAL LOCALITY MEASURES OF THE MRN

- $M^1(a, l) := \frac{1}{n} \sum_{i \in [n]} \delta_{(X(i), L(i))}(a, l),$
- $L(i) = (l^i(b), b \in \Omega), l^i(b) = \sum_{j \sim i} 1_{\{X(j)=b\}}.$
- $Deg_Z(k) = \sum_{(a, \ell) \in \Omega \times N(\Omega)} \delta_k(\|\ell\|) M^1(a, \ell)$
- $L^2(a, b) = \sum_{\ell \in N(\Omega)} \ell(b) M^1(a, \ell)$
- $W_0 = \sum_{a \in \Omega} M^1(a, 0).$ O'Connell(1998), Doku-Amponsah (2016)

NOTATION

- $N(\Omega)$ is the space of counting measures on Ω , i.e. those measures taking values in $N \cup \{0\}$, endowed with the discrete topology.
- $M(\Omega)$ is the space of probability measures on Ω endowed with the weak topology.
- $\tilde{M}_*(\Omega \times \Omega)$ is the subspace of symmetric measures in $\tilde{M}(\Omega \times \Omega)$ (the space of finite measures on $\Omega \times \Omega$) also equipped with the weak topology.
- $M(\Omega \times N(\Omega))$ is the space of probability measures on Ω endowed with the weak topology.

RELATIVE ENTROPY FOR EMPIRICAL PAIR MEASURES

- For a measure $\pi \in \tilde{M}(\Omega \times \Omega)$ and a measure $\rho \in M(\Omega)$, define (a relative entropy) $\mathfrak{H}_{\kappa/\ell}$ by

$$\mathfrak{H}_{\kappa/\ell}(\pi \parallel \rho) := H(\pi \parallel \frac{\kappa}{\ell} \rho \otimes \rho) + \|\frac{\kappa}{\ell} \rho \otimes \rho\| - \|\pi\|,$$

- $\frac{\kappa}{\ell} \rho \otimes \rho(a, b) = \frac{\kappa(a,b)}{\ell(a,b)} \rho(a) \rho(b)$, for $a, b \in \Omega$
- Note $\mathfrak{H}_{\kappa/\ell}(\pi \parallel \rho) \geq 0$ with equality iff $\pi = \frac{\kappa}{\ell} \rho \otimes \rho$.

CONSISTENCY OF EMPIRICAL MEASURES

- We call a pair of measures $(\pi, \mu) \in \tilde{M}(\Omega \times \Omega) \times M(\Omega \times N(\Omega))$ **sub-consistent** if

$$\pi(a, b) \leq \sum_{\ell \in N(\Omega)} \ell(b) \mu(a, \ell), \quad \forall a, b \in \Omega.$$

- We call a pair of measures $(\pi, \mu) \in \tilde{M}(\Omega \times \Omega) \times M(\Omega \times N(\Omega))$ **consistent** if

$$\pi(a, b) = \sum_{\ell \in N(\Omega)} \ell(b) \mu(a, \ell), \quad \forall a, b \in \Omega.$$

- Write

$$\Gamma := \left\{ (\pi, \mu) : (\pi, \mu) \text{ is consistent} \right\}$$

LDP AND ASYMPTOTIC DISTRIBUTIONS

■ Large deviation principle

$U: G(\Omega) \rightarrow M$ is said to satisfy a *large deviation principle* with rate function I if, for all Borel sets $B \subset M$,

$$\begin{aligned} - \inf_{m \in B^\circ} I(m) &\leq \liminf_{n \rightarrow \infty} \frac{1}{n} \log P_n \{U(X) \in B\} \\ &\leq \limsup_{n \rightarrow \infty} \frac{1}{n} \log P_n \{U(X) \in B\} \leq - \inf_{m \in B^c} I(m). \end{aligned}$$

■ Limiting multivariate poisson distribution

For every $(\pi, \nu) \in \Omega$, we define a probability measure $Q = Q[\pi, \mu]$ on $\Omega \times N(\Omega)$ by

$$Q(a, l) = \mu_1(a) \prod_{b \in \Omega} \frac{e^{-[\pi(a,b)/\mu_1(a)]} [\pi(a,b)/\mu_1(a)]^{l(b)}}{l(b)!}.$$

ASUMPTIONS

As $n \rightarrow \infty$ we assume that κ_n, ℓ_n satisfy

- $n \frac{\kappa_n(a,b)}{\ell_n(a,b)} \rightarrow \frac{\kappa(a,b)}{\ell(a,b)}$
- $\frac{\kappa_n(a,b)}{\ell_n(a,b)} \rightarrow 0$

Theorem A

(L^1, L^2) obeys an LDP in $M(\Omega) \times M(\Omega \times \Omega)$ with good rate function,

$$I(\rho, \pi) = H(\rho \parallel \eta) + \frac{1}{2} \mathfrak{H}_{\kappa/\ell}(\pi \parallel \rho).$$

Proof of Theorem A

- Exponential change of measure; See Doku-Amponsah (2006).
- Local Large deviation principle; See Doku-Amponsah (2017).
- Large deviation for mixtures. See Biggins (2004).

Theorem B

(L^2, M^1) obeys an LDP in $M(\Omega \times \Omega) \times M(\Omega \times N(\Omega))$ with good rate function,

$$J_1(\pi, \mu) = \left\{ \begin{array}{ll} H(\mu \| Q) + H(\mu_1 \| \eta) + \frac{1}{2} \mathfrak{H}_{\kappa/\ell}(\pi \| \mu_1) & \text{if } (\pi, \mu) \in \Gamma \\ \infty & \text{otherwise.} \end{array} \right\}.$$

Proof of Theorem B

- Random Allocation of Coloured Balls into Coloured Bins; See Doku-Amponsah and Moerters (2010).
- Theorem A
- Large deviations for mixtures; See Biggins (2004)

DISCUSSION OF MAIN RESULTS

- Asymptotically, the empirical cooperative measure behave as

$$\frac{\kappa}{\ell} \eta \otimes \eta.$$

- Asymptotically, the empirical locality measure behave as

$$Q(a, l) = \eta(a) \prod_{b \in \Omega} \frac{e^{-\left[\frac{\kappa(a,b)}{\ell(a,b)} \eta(b)\right]} \left[\frac{\kappa(a,b)}{\ell(a,b)} \eta(b)\right]^{l(b)}}{l(b)!}.$$

EVOLUTION AND CO-EVOLUTION PROCESSES ON NETWORKS

- Evolution processes defines an irreducible Markov Chains on the set of networks, say $G[n]$, with a unique invariant distributions which is the law of an inhomogeneous random network. See, Jackson and Watts (2002a).
- Co-Evolution Processes has all the properties of the evolution processes plus additional parameter which models the action profile of a player. ie. its has Action Adjustment rule, Link creation rule and Link destruction rule. See, Jackson and Watts (2002a).
- Indeed, co-evolution processes may seen as extensions of random networks. See, Hellman and Staudigl(2014).

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THANK YOU