

# "GENERAL" MODEL

$$\bar{\pi}_n(\mathcal{g}) = \frac{\exp \left\{ \frac{\alpha}{n^2} \sum_i \sum_j g_{ij} + \frac{\beta}{2n^2} \sum_i \sum_j g_{ij} g_{ji} + \frac{\gamma}{n^3} \sum_i \sum_j \sum_{k \neq i,j} g_{ij} g_{jk} \right\}}{c(\alpha, \beta, \gamma, \mathcal{g}_n)}$$

$$\frac{1}{n^2} \sum_i \sum_j g_{ij} = \text{direct link density}$$

$$\frac{1}{n^2} \sum_i \sum_j g_{ij} g_{ji} = \text{reciprocity density}$$

$$\frac{1}{n^2} \sum_i \sum_j \sum_{k \neq i,j} g_{ij} g_{jk} = \text{density of directed two-paths}$$

$$\leftarrow c(\alpha, \beta, \gamma, \mathcal{g}_n)$$

- Connection with the exponential family in statistics
- Smooth behaviour if  $c(\cdot)$  is analytic in  $\mathcal{D} = [\alpha, \beta, \gamma]$
- unstable on phase transition if  $c(\cdot)$  is not analytic in  $\mathcal{D}$ .

# CONVERGENT GRAPH SEQUENCES

$H, G$  simple graphs (~~undirected~~, unweighted, no loops, no multiple edges)

**Homomorphism**: arc-preserving map from  $V(H)$  to  $V(G)$

$$t(H, G) = \frac{|\text{hom}(H, G)|}{|V(G)|^{|V(H)|}} \quad \begin{array}{l} \text{homomorphism density} \\ (= \text{prob. that a random mapping is a homomorphism}) \end{array}$$

Let  $G_n \rightarrow G$  as  $n \rightarrow \infty$ , then for any simple  $H$

$$t(H, G_n) \rightarrow t(H, G)$$

Need a graphon!

## GRAPHS

Let  $h \in W$ ,  $W =$  set of all measurable (~~symmetric~~) functions from  $[0,1]^2$  to  $[0,1]$

### Interpretation

- $h$  represents a "continuous" graph, i.e. a graph s.t.  $|V(G)|$  is non denumerable
- $(x, y) \sim \text{Uniform}([0,1]^2)$
  - $h(x, y) =$  (conditional) probability that  $x$  and  $y$  are connected

Let  $H$  be s.t.  $V(H) = \{1, 2, \dots, k\}$

$$t(H, h) = \int_{[0,1]^k} \prod_{(i,j) \in E(H)} h(x_i, x_j) dx_1, dx_2, \dots, dx_k = \text{probability of homomorphism from } H \text{ to } G=h$$

Let  $V(H) = \{1, 2\}$ , then

$$t(H, h) = \int_{[0,1]^2} h(x_1, x_2) dx_1 dx_2$$

Any finite simple graph can be represented as a graphon.

Let  $V(G_n) = \{1, 2, \dots, n\}$

$$f^G(x, y) = \begin{cases} 1 & \text{if } (\Gamma_{nx}, \Gamma_{ny}) \in E(G) \\ 0 & \text{otherwise} \end{cases}$$

## EXCHANGEABILITY

$X$  = adjacency matrix

$$\text{Exchangeability} \Leftrightarrow (X_{ij}) \stackrel{d}{=} (X_{\sigma(i)\sigma(j)})$$

Notice that if node permutations are equally likely, then any permutation is a measure preserving transformation

## CUT DISTANCE

Let  $f, g \in W$

$$d_{\square}(f, g) = \sup_{S, T \subseteq [0, 1]} \left| \int_{S \times T} f(x, y) - g(x, y) dx dy \right|$$

Let  $\Sigma$  be the space of all the measure preserving bijections

$$\sigma: [0, 1] \rightarrow [0, 1] \quad \text{s.t.} \quad \mu(A) = \mu(\sigma^{-1}A)$$

$f(x, y) \sim g(x, y)$  if  $f(x, y) = g_\sigma(x, y) := g(\sigma x, \sigma y)$  for some  $\sigma \in \Sigma$

$\tilde{g}$  represents the closure of  $\{g_\sigma\}$  in  $(W, d_\square)$

$\tilde{W}$  denotes the quotient space and  $\tilde{\tau}: g \rightarrow \tilde{g}$

We can define a natural distance on  $\tilde{W}$ :

$$d_\square(\tilde{f}, \tilde{g}) := \inf_{\sigma} d_\square(f, g_\sigma) = \inf_{\sigma} d_\square(f_\sigma, g) = \inf_{\sigma_1, \sigma_2} d_\square(f_{\sigma_1}, g_{\sigma_2})$$

$\tilde{W}$  is a compact space

$t(H, \cdot)$  is continuous for any finite  $H$

# LARGE DEVIATIONS

$$\text{let } I_p(u) := \frac{1}{2} u \log \frac{u}{p} + \frac{1}{2} (1-u) \log \frac{1-u}{1-p}$$

and

$$I_p(h) = \int \int I_p(h(x,y)) dx dy, \quad h \in W$$

$$I_p(\tilde{h}) := I_p(h), \quad \tilde{h} \in \tilde{W} \quad \text{and} \quad h \sim \tilde{h}$$

Erdős Renyi:  $G(n,p)$  induces  $\mathbb{P}_{n,p}$  measure on  $W$  through

the map  $G \rightarrow f^G$ , hence

$$\tilde{\mathbb{P}}_{n,p} \text{ on } \tilde{W} \text{ through } G \rightarrow f^G \rightarrow \tilde{f}^G = \tilde{G}$$



LARGE DEVIATION PRINCIPLE FOR  $\tilde{P}_{n,p}$  ON  $(\tilde{W}, d_{\square})$

For each fixed  $p \in (0, 1)$ , the sequence  $\tilde{P}_{n,p}$  obeys a large deviation principle in the space  $\tilde{W}$  (equipped with the ent metric) with rate function  $I_p$ . This means that, for any closed

set  $\tilde{F} \subseteq \tilde{W}$

$$\limsup_{n \rightarrow \infty} \frac{1}{n^2} \log \tilde{P}_{n,p}(\tilde{F}) \leq - \inf_{h \in \tilde{F}} I_p(h)$$

and for any open set  $\tilde{U} \subseteq \tilde{W}$

$$\liminf_{n \rightarrow \infty} \frac{1}{n^2} \log \tilde{P}_{n,p}(\tilde{U}) \geq - \inf_{h \in \tilde{U}} I_p(h)$$

## MAIN RESULT

Let  $T: \tilde{W} \rightarrow \mathbb{R}$  a bounded continuous function on  $(\tilde{W}, \tau_{\square})$

$\mathcal{G}_n$  = set of simple graphs with  $n$  vertices

$$p_n(G) := e^{n^2 (T(\tilde{G}) - \psi_n)} \quad \tilde{G} \in \tilde{W}$$

$$\psi_n = \frac{1}{n^2} \log \sum_{G \in \mathcal{G}_n} e^{n^2 T(\tilde{G})}$$

Let  $\mathbb{I}(u) = \frac{1}{2} u \log(u) + \frac{1}{2} (1-u) \log(1-u)$

and

$$\mathbb{I}(\tilde{h}) = \int_0^1 \int_0^1 \mathbb{I}(h(x,y)) dx dy \quad h \sim \tilde{h}$$

Then

$$\psi := \lim_{n \rightarrow \infty} \psi_n = \sup_{\tilde{h} \in \tilde{W}} \left( T(\tilde{h}) - I(\tilde{h}) \right)$$

The maximiser might be

- a scalar  $\Rightarrow \psi$  analytic function of the parameters

asymptotically  $E-R$  graph

- scalar, but not unique

still  $E-R$  asymptotically but with phase transitions

small variations in the parameters determine radically different behaviours (from nearly complete to sparse graphs)

- a function

no more  $E-R$  asymptotically