

Different priors

1 Class Model (Stochastic Block Model, SBM)

Likelihood

$$\begin{aligned}
 y_{i,i'} \mid \xi_i, \xi_{i'}, \{\eta_{k,\ell}\} &\stackrel{\text{ind}}{\sim} \text{Ber}(\text{expit } \eta_{\phi(\xi_i, \xi_{i'})}) \\
 p(\mathbf{Y} \mid \{\xi_i\}, \{\eta_{k,\ell}\}) &= \prod_{i=1}^{I-1} \prod_{i'=i+1}^I (\text{expit } \eta_{\phi(\xi_i, \xi_{i'})})^{y_{i,i'}} (1 - \text{expit } \eta_{\phi(\xi_i, \xi_{i'})})^{1-y_{i,i'}} \\
 &= \prod_{k=1}^K \prod_{\ell=k}^K (\text{expit } \eta_{k,\ell})^{s_{k,\ell}} (1 - \text{expit } \eta_{k,\ell})^{n_{k,\ell} - s_{k,\ell}}
 \end{aligned}$$

where $s_{k,\ell} = \sum_{\mathcal{S}_{k,\ell}} y_{i,i'}$, $n_{k,\ell} = \sum_{\mathcal{S}_{k,\ell}} 1$, with $\mathcal{S}_{k,\ell} = \{(i, i') : i < i' \text{ and } \phi(\xi_i, \xi_{i'}) = (k, \ell)\}$, and $\phi(x, y) = (\min\{x, y\}, \max\{x, y\})$ is a function to take into account that $\mathbf{Y} = [y_{i,i'}]$ is a symmetric matrix.

Prior

$$\begin{aligned}
 \eta_{k,\ell} \mid \zeta, \tau^2 &\stackrel{\text{iid}}{\sim} \text{N}(\zeta, \tau^2) \\
 \zeta &\sim \text{N}(\mu_\zeta, \sigma_\zeta^2) \\
 \tau^2 &\sim \text{IGam}(a_\tau, b_\tau) \\
 p(K, n) &\sim DP(\alpha) \\
 \alpha &\sim \text{might add this later}
 \end{aligned}$$

Parameters

$$\Upsilon = (\eta_{1,1}, \eta_{1,2}, \dots, \eta_{K,K}, K, n_1, \dots, n_K, \alpha),$$

where $\xi_i \in \{1, \dots, K\}$, $i = 1, \dots, I$, are the cluster assignments ($\xi_i = k$ means that actor i belongs to cluster k).

Hyper-parameters

$$(\mu_\zeta, \sigma_\zeta^2, a_\tau, b_\tau).$$

Posterior

$$p(\Upsilon \mid \mathbf{Y}) = p(\mathbf{Y} \mid \{\xi_i\}, \{\eta_{k,\ell}\}) p(\{\eta_{k,\ell}\} \mid \zeta, \tau^2) p(\zeta) p(\tau^2) p(K, n \mid \alpha) p(\alpha)$$

$$\propto \prod_{i=1}^{I-1} \prod_{i'=i+1}^I \theta_{i,i'}^{y_{i,i'}} (1 - \theta_{i,i'})^{1-y_{i,i'}} \times \exp\left\{-\frac{1}{2\sigma_\zeta^2}(\zeta - \mu_\zeta)^2\right\} \times (\tau^2)^{-(a_\tau-1)} \exp\left\{-\frac{b_\tau}{\tau^2}\right\} \\ \times \prod_{k=1}^K \prod_{\ell=k}^K (\tau^2)^{-1/2} \exp\left\{-\frac{1}{2\tau^2}(\eta_{k,\ell} - \zeta)^2\right\} \times \begin{cases} n_h, & \text{for } h = 1, \dots, H \\ \alpha, & \text{for } h = H + 1 \end{cases}$$

where $\theta_{i,i'} = \text{expit}(\eta_{\phi(\xi_i, \xi_{i'})})$ and $[\cdot]$ is the Iverson bracket.

MCMC

The algorithm proceeds by generating a new state $\Upsilon^{(b+1)}$ from a current state $\Upsilon^{(b)}$, $b = 1, \dots, B$, as follows:

1. Sample $\eta_{k,\ell}^{(b+1)}$, $\ell = k, \dots, K$ and $k = 1, \dots, K$, according to a Metropolis–Hastings Algorithm, considering the fd:

$$\log p(\eta_{k,\ell} \mid \text{rest}) \propto s_{k,\ell} \log(\text{expit } \eta_{k,\ell}) + (n_{k,\ell} - s_{k,\ell}) \log(1 - \text{expit } \eta_{k,\ell}) - \frac{1}{2\tau^2}(\eta_{k,\ell} - \zeta)^2 \\ = s_{k,\ell} \eta_{k,\ell} - n_{k,\ell} \log(1 + \exp \eta_{k,\ell}) - \frac{1}{2\tau^2}(\eta_{k,\ell} - \zeta)^2,$$

where $s_{k,\ell} = \sum_{\mathcal{S}_{k,\ell}} y_{i,i'}$ and $n_{k,\ell} = \sum_{\mathcal{S}_{k,\ell}} 1$, with $\mathcal{S}_{k,\ell} = \{(i, i') : i < i' \text{ and } \phi(\xi_i, \xi_{i'}) = (k, \ell)\}$.

2. Sample $\xi_i^{(b+1)}$, $i = 1, \dots, I$, from a categorical distribution on $\{1, \dots, K\}$, such that:

$$\mathbb{P}r[\xi_i = k \mid \text{rest}] \propto \prod_{i'=i+1}^I \eta_{\phi(k, \xi_{i'})}^{y_{i,i'}} (1 - \eta_{\phi(k, \xi_{i'})})^{1-y_{i,i'}} \times \prod_{i'=1}^{i-1} \eta_{\phi(\xi_{i'}, k)}^{y_{i',i}} (1 - \eta_{\phi(\xi_{i'}, k)})^{1-y_{i',i}} \times \begin{cases} n_h, & h = 1, \dots, H \\ \alpha, & h = H + 1 \end{cases}$$

3. Sample $\zeta^{(b+1)}$ from $\mathcal{N}(m, v^2)$, where

$$v^2 = \left(\frac{1}{\sigma_\zeta^2} + \frac{K(K+1)/2}{\tau^2} \right)^{-1} \quad \text{and} \quad m = v^2 \left(\frac{\mu_\zeta}{\sigma_\zeta^2} + \frac{1}{\tau^2} \sum_{k=1}^K \sum_{\ell=k}^K \eta_{k,\ell} \right).$$

4. Sample $(\sigma^2)^{(b+1)}$ from $p(\sigma^2 \mid \text{rest}) = \text{IGam}\left(\sigma^2 \mid a_\tau + \frac{K(K+1)}{4}, b_\tau + \frac{1}{2} \sum_{k=1}^K \sum_{\ell=k}^K (\eta_{k,\ell} - \zeta)^2\right)$.