

Solutions for Statistical analysis of network data – Sheet 7

Recall that the Hamming distance between two adjacency matrices A_1 and A_2 is

$$d_1(A_1, A_2) = \frac{\|A_1 - A_2\|_1}{n^2},$$

and the Jaccard distance between A_1 and A_2 is

$$d_J(A_1, A_2) = \frac{\|A_1 - A_2\|_1}{\|A_1 + A_2\|_*},$$

where $\|\cdot\|_*$ is the nuclear norm.

1. Substituting the two matrices into the formulae above, we get $d_1(A_1, A_2) = 4/36$ and $d_J(A_1, A_2) = 0.22$. The estimated edge density is $\rho_1 = 9/15$ for A_1 and $\rho_2 = 11/15$ for A_2 .
2. Substituting the two matrices into the formulae, we get $d_1(A_1, A_2) = 4/36$ (similarly as in Problem 1) and $d_J(A_1, A_2) = 0.36$. The estimated edge density is $\rho_1 = 5/15$ for A_1 and $\rho_2 = 7/15$ for A_2 ; the relative change in edge density is higher in this case than in problem 1, and this is reflected by the increased Jaccard distance value.
3. The exponential random graph formula:

$$\Pr\{\mathbf{A} = \mathbf{a}\} = \left(\frac{1}{\kappa}\right) \exp \left\{ \sum_H \theta_H g_H(\mathbf{a}) \right\}.$$

- (a) Requiring independence is equivalent to require that the above probability be of a product form in the terms of $\mathbf{a}$. Any function g_H defined on any subgraph H other than a single edge is therefore excluded.
- (b) Let K_2 denote the complete graph on 2 nodes (that is, the edge graph). Then

$$\begin{aligned} \Pr\{\mathbf{A} = \mathbf{a}\} &= \left(\frac{1}{\kappa}\right) \exp \left\{ \sum_H \theta_H g_H(\mathbf{a}) \right\} \\ &= \left(\frac{1}{\kappa}\right) \exp \left\{ \sum_{K_2} \theta_{K_2} g_{K_2}(\mathbf{a}) \right\} \\ &= \left(\frac{1}{\kappa}\right) \exp \left\{ \sum_{i>j} \theta_{K_2} g_{K_2}(a_{ij}) \right\}. \end{aligned}$$

- (c) The Erdős-Rényi model can be recovered by choosing

$$\begin{aligned} g_{K_2}(a_{ij}) &= a_{ij}, \\ \theta_{K_2} &= \log \frac{p}{1-p}, \\ \kappa &= (1-p)^{-\binom{n}{2}} \end{aligned}$$