

# Bayesian Phylogenetic Inference of Stochastic Block Model on Random Graphs

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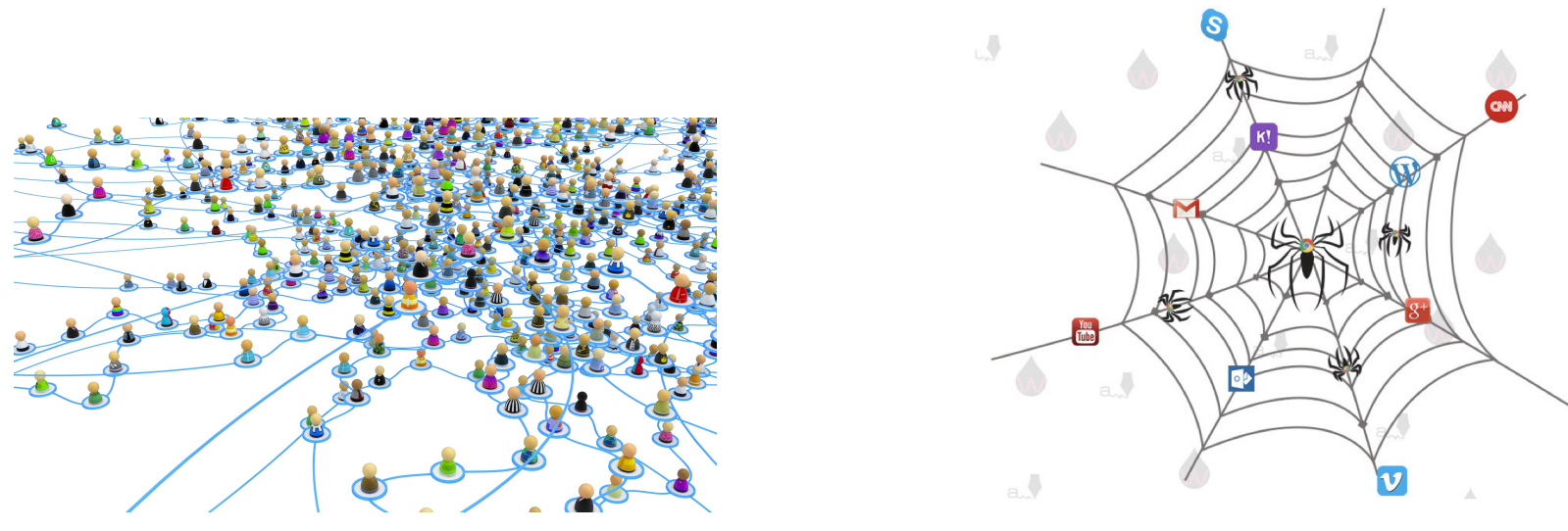
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## 1 Motivations

Networks are ubiquitous:

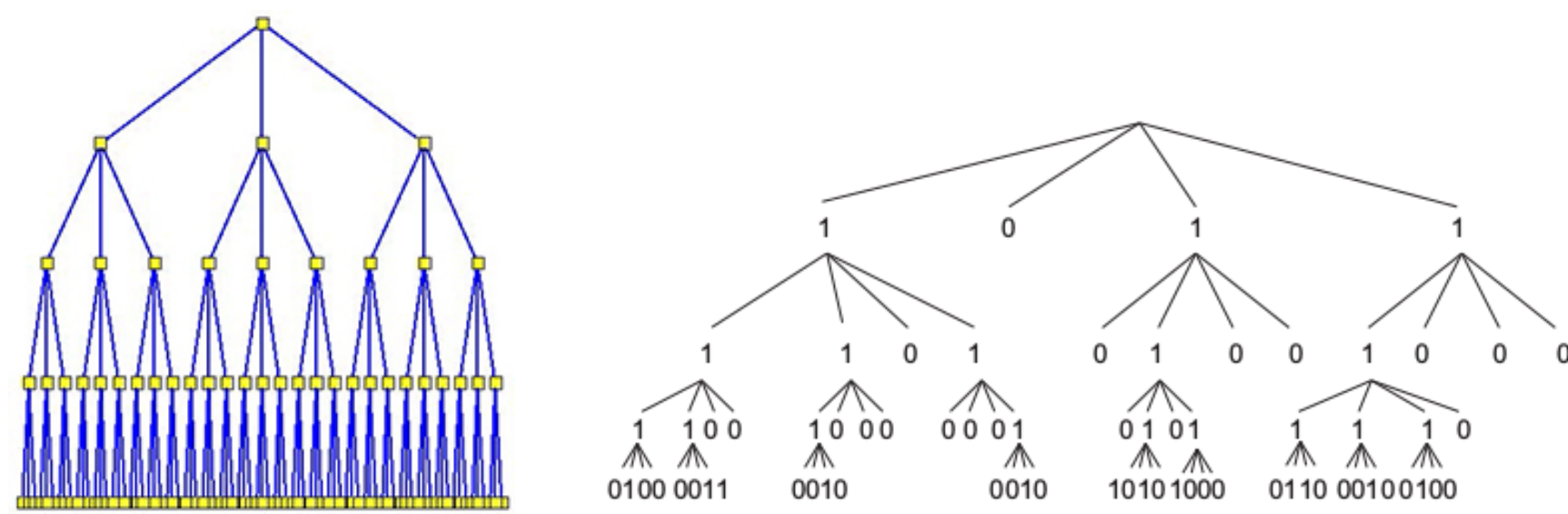
- Social networks; Biological networks
- Data science: Network sampling; Clustering in graphs



## 2 Model Setup

Consider the broadcasting process as a discrete, irreducible, aperiodic, and reversible Markov chain.

- $\mathbb{T} = (\mathbb{V}, \mathbb{E}, \rho)$  is a tree with nodes  $\mathbb{V}$ , edges  $\mathbb{E}$  and root  $\rho \in \mathbb{V}$
- $d$ -ary tree is the infinite rooted tree where every vertex has exactly  $d$  offspring.
- Define a finite characters set  $\mathcal{C}$ , whose elements are configurations on  $\mathbb{T}$ , denoted by  $\sigma$ .
- The state of the root  $\rho$ , denoted by  $\sigma_\rho$ , is chosen according to an initial distribution  $\pi$  on  $\mathcal{C}$ .
- Denote by Probability transition matrix  $\mathbf{P} = (p_{ij})$  as the noisy communication channel on each edge.
- Let  $\sigma(n)$  denote the spins at distance  $n$  from the root and let  $\sigma^i(n)$  denote  $\sigma(n)$  conditioned on  $\sigma_\rho = i$ .



**RECONSTRUCTION:** Does this configuration contain a non-vanishing information transmitted by the root, as  $n$  goes to  $\infty$ ?

**Definition 1.** The reconstruction problem for the infinite tree  $\mathbb{T}$  is solvable if for some  $i, j \in \mathcal{C}$ ,

$$\limsup_{n \rightarrow \infty} d_{TV}(\sigma^i(n), \sigma^j(n)) > 0$$

where  $d_{TV}$  is the total variation distance, i.e.

$$d_{TV}(\sigma^i(n), \sigma^j(n)) = \sup_A |\mathbf{P}(\sigma(n) = A \mid \sigma_\rho = i) - \mathbf{P}(\sigma(n) = A \mid \sigma_\rho = j)|$$

When the  $\limsup$  is 0, the model has non-reconstruction on  $\mathbb{T}$ .

## 3 Background

- The second largest eigenvalue in absolute value of the transition matrix  $\mathbf{P}$ , say,  $\lambda$ , plays the crucial role in reconstruction problems.
- $d|\lambda|^2 > 1$  (Kesten-Stigum bound): The reconstruction problem is solvable. (Kesten and Stigum, 1966)
- For larger noise  $d|\lambda|^2 < 1$ : reconstruction depends on the channel.
- The binary symmetric channel: the reconstruction problem is solvable if and only if  $d\lambda^2 > 1$  (Bleher et al, 1995).
- The binary asymmetric model with sufficiently large asymmetry: Mossel (2004) showed that the Kesten-Stigum bound is **NOT** the bound for reconstruction.
- The first exact reconstruction threshold in roughly a decade was obtained by Borgs et al (2006) for the asymmetric Ising channel, i.e.

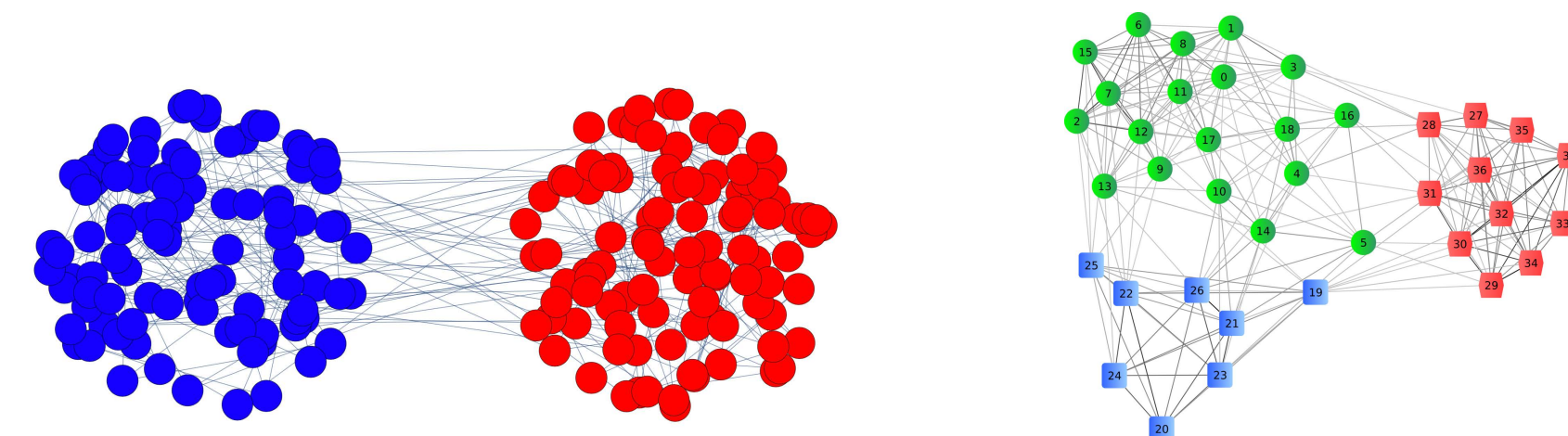
$$\mathbf{P} = \frac{1}{2} \left[ \begin{pmatrix} 1+\theta & 1-\theta \\ 1-\theta & 1+\theta \end{pmatrix} + \Delta \begin{pmatrix} -1 & 1 \\ 1 & -1 \end{pmatrix} \right]$$

- Liu and Ning (2018) improved the result: when  $\Delta^2 > (1-\theta)^2/3$ , for every  $d$  the Kesten-Stigum bound is not tight. In other words, the reconstruction problem is solvable for some  $\theta$  even if  $d\theta^2 < 1$ ; when  $\Delta^2 < (1-\theta)^2/3$ , there exists a  $D = D(\pi) > 0$  such that for  $d > D$  the Kesten-Stigum bound is sharp.

## 4 Reconstruction of Stochastic Block Models

### 4.1 Stochastic Block Models

- The number  $n$  of vertices;
- a partition of the vertex set  $\{1, \dots, n\}$  into disjoint subsets  $C_1, \dots, C_r$ , called communities; a symmetric  $r \times r$  matrix  $\mathbf{P}$  of edge probabilities.
- The edge set is then sampled at random as follows: any two vertices  $u \in C_i$  and  $v \in C_j$  are connected by an edge with probability  $p_{ij}$ .



### Different In-block and Out-block Mutations

- Characters set:  $\mathcal{C} = \mathcal{C}_1 \cup \mathcal{C}_2$ , consisting of two categories  $\mathcal{C}_1 = \{1, \dots, q\}$  and  $\mathcal{C}_2 = \{q+1, \dots, 2q\}$ .
- The state of the root variable  $\rho$  is chosen according to the uniform distribution on  $\mathcal{C}$ .
- The information flow in the tree according to a general symmetric  $2q \times 2q$  probability transition matrix  $\mathbf{P}$  with different in-community and out-community transition probabilities, defined as follows:

$$\mathbf{P} = \begin{pmatrix} p_0 & p_1 & \cdots & p_1 & p_2 & \cdots & \cdots & p_2 \\ p_1 & p_0 & \cdots & p_1 & \vdots & \cdots & & \vdots \\ \vdots & \vdots & \ddots & \vdots & \vdots & \ddots & & \vdots \\ p_1 & p_1 & \cdots & p_0 & p_2 & \cdots & \cdots & p_2 \\ p_2 & \cdots & \cdots & p_2 & \bar{p}_0 & \bar{p}_1 & \cdots & \bar{p}_1 \\ \vdots & \ddots & & \vdots & \bar{p}_1 & \bar{p}_0 & \cdots & \bar{p}_1 \\ \vdots & \cdots & \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ p_2 & \cdots & \cdots & p_2 & \bar{p}_1 & \bar{p}_1 & \cdots & \bar{p}_0 \end{pmatrix}_{2q \times 2q}$$

The eigenvalues of  $\mathbf{P}$  are 1 and

$$\lambda_1 = p_0 - p_1, \quad \lambda_2 = p_0 + (q-1)p_1 - qp_2, \quad \lambda_3 = \bar{p}_0 - \bar{p}_1.$$

### Kimura 1980 Model

Specially if set  $p_0 = \bar{p}_0$ , the preceding model becomes K80 DNA sequence evolution model, which has two mutation classes with  $q$  states in each class and distinguishes between transitions and transversions.

**Theorem 4.1** (Liu et al, 2018). When  $q \geq 4$ , for every  $d$  the Kesten-Stigum bound is not tight.

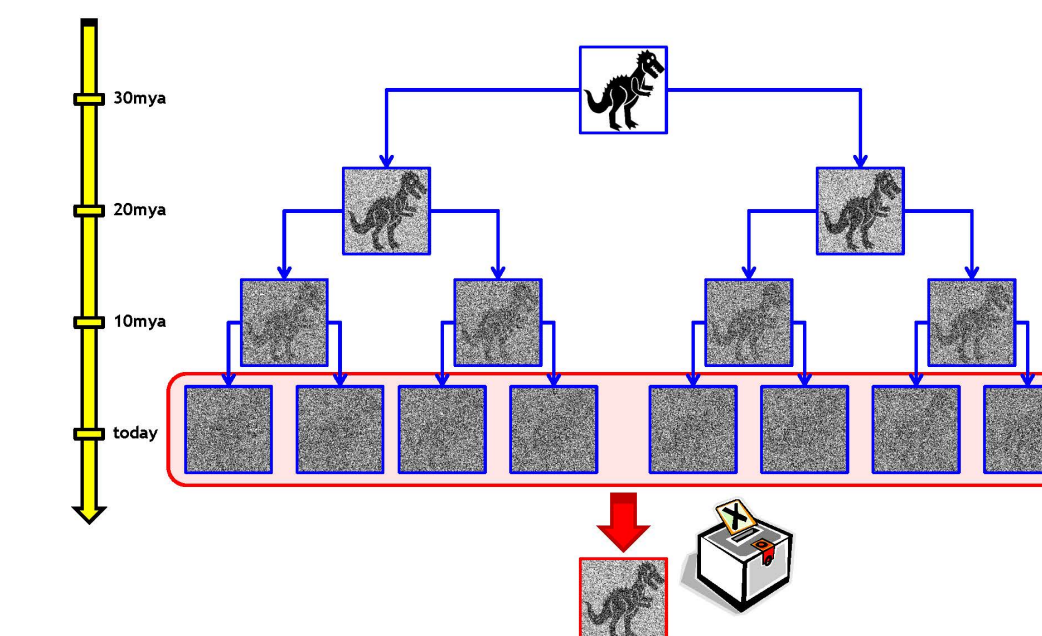
### 4.2 Research Questions

The corresponding information reconstruction problem in molecular phylogenetics will be explored, by means of the refined analysis of moment recursion on a weighted version of the magnetization, concentration investigation and in-depth investigation on the resulting nonlinear second order dynamical system. Our purpose is to figure out

- Under what conditions of  $p_0, p_1, \bar{p}_0, \bar{p}_1, p_2$  is the Kesten-Stigum reconstruction bound tight, i.e. the reconstruction is unsolvable when  $d\lambda_2 < 1$ ?
- If Kesten-Stigum bound is not sharp, then we are interested in figuring out the new reconstruction threshold.

### 4.3 Application

**Unsupervised Learning.** Classification problem in unsupervised learning setting using deep generative hierarchical network.



**Clustering Problem.** Clustering problem in unsupervised learning setting using the stochastic block model.

**Phylogenetic Reconstruction.** Construct the ancestry tree of a collection of species, given the information of present species.

## 5 Method and Materials

Let  $u_1, \dots, u_d$  be the children of  $\rho$  and  $\mathbb{T}_v$  be the subtree of descendants of  $v \in \mathbb{T}$ . Furthermore, if we set  $d(\cdot, \cdot)$  as the graph-metric distance on  $\mathbb{T}$ , denote the  $n$ th level of the tree by  $L_n = \{v \in \mathbb{V} : d(\rho, v) = n\}$  and then let  $\sigma_j(n)$  denote the spins on  $L_n \cap \mathbb{T}_{u_j}$ . For a configuration  $A$  on  $L_n$  define the posterior function

$$f_n(i, A) = \mathbf{P}(\sigma_\rho = i \mid \sigma(n) = A) = \mathbf{P}(\sigma_{u_j} = i \mid \sigma_j(n+1) = A).$$

We will research the asymptotic behavior of the objective quantities:

$$x_n = \mathbf{E} \left( f_n(1, \sigma^1(n)) - \frac{1}{2q} \right); \quad \bar{x}_n = \mathbf{E} \left( f_n(q+1, \sigma^{q+1}(n)) - \frac{1}{2q} \right).$$

If the reconstruction problem is solvable, then  $\sigma(n)$  contains significant information on the root variable.

**Theorem 5.1.** The non-reconstruction is equivalent to

$$\lim_{n \rightarrow \infty} x_n = \lim_{n \rightarrow \infty} \bar{x}_n = 0.$$

### Distributional Recursion

- Key Idea: analyze the recursive relation between  $x_n, \bar{x}_n$  and  $x_{n+1}, \bar{x}_{n+1}$  by Markov random field property.
- Establish the distributional recursion and moment recursion.
- Display the interactions between spins become very weak if they are sufficiently far away from each other.
- The traditional method is to analyze the the stability of fixed points of the preceding dynamical system of  $x_n$  and  $\bar{x}_n$ .

## 6 Extensions and Further Discussion

Our technology would be generalized to handle the probabilistic examination of the underlying microscopic systems with large populations on general random graphs.

