

Homework 3

1) Sample Space

We define our sample space as: $G_n = (X, Y, Z)$

- Adjacency Matrix: $X = (0,1)^{n \times n}$ -> this adjacency matrix details the connection of neurons
- $Y = \{0,1\}^n$ -> whether or not the neuron is inhibitory or excitatory
 - Inhibitory = 0
 - Excitatory = 1
- Z represents the orientation of the neuron into four divided regions equally
 - The tuning property
 - $Z = (0, 2\pi)^n$

2) The model being implemented is a stochastic block model

- $SBM_n^2(p, B) = \{P_\theta: \theta \rightarrow \vartheta\}, \vartheta \Delta_k \times (0,1)^{k \times k}$

The block model parameters of $p(z)$ and $B(z)$ must be defined in our stochastic block model. Specifically, $p(z) \in \Delta_k$ and $B(z) \in (0,1)^{k \times k}$ are the parameters. K is the number of blocks in our model while B , utilizing the Bernoulli distribution, is the model of the edges. In regards to the distribution of choice, the Bernoulli distribution is the simplest for B . The model has no dependence on the orientation of the neurons.

The interval for Z (between 0 and 2π) creates a large variance for the SBM. Interestingly, we must bin the orientation in order to reduce the variance. In the Bock paper, 8 bins were used. However, we can improve upon 8bins by increasing to 18bins taking into account the 0-180degree range. We must notice that decreasing our variance will increase bias due to the bias-variance tradeoff.

Interpreting and analyzing the graph studied would not be optimal. This is due to the initial assumption that the orientation of the neurons changes connectivity.