

Statistical Connectomics HW 3

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1 Introduction

In the Bock paper, the sample space was defined as $\mathcal{G}_n = (\mathcal{X}, \mathcal{Y}, \mathcal{Z})$ where $\mathcal{X} = (0, 1)^{n \times n}$ is the set of nodes (i.e. neurons) in the graph, $\mathcal{Y} = (0, 1)^n$ determines whether the neuron is excitatory or inhibitory and \mathcal{Z} describes the preferred orientation of the neuron. The tuning property of the neuron can range from 0 to 2π , which would give us $\mathcal{Z} = (0, 2\pi)^n$. However, since the neurons must be categorized into discrete blocks, the Bock paper defined \mathcal{Z} by partitioning it into 8 blocks, which gives us $\mathcal{Z} = [8]^n$. In this assignment, we will try to redefine this definition of \mathcal{Z} to (hopefully!) come up with a better model and also define the structure of the block model parameters, ρ and β , used in this model

2 Redefining \mathcal{Z}

Rather than defining \mathcal{Z} into 8 partitions, we decided to use 18 partitions instead for two reasons. 1) The range of orientation sensitivity for a neuron is around 10 degrees, so this may be a more biologically accurate representation of neurons in the brain and 2) More blocks allow a more detailed categorization of neurons and could prevent oversimplifying our model.

3 Structure of Block Model Parameters ρ and β

For ρ and β we will use the following definitions we learned from class: $\rho : \mathcal{Z} \in \Delta_k$ and $\beta = (0, 1)^{k \times k}$. Now we just have to define k . We propose to define $k=18$, since there will be 18 blocks in our model due to how we partitioned \mathcal{Z} in the previous section above. Therefore, $\rho : \mathcal{Z} \in \Delta_{18}$.