

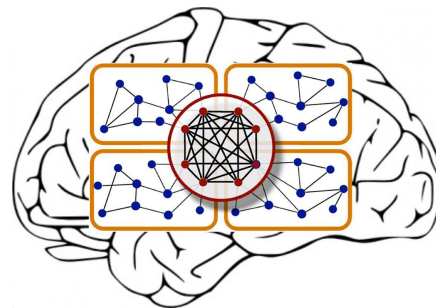
# 3D Distribution of Synapses in Cortex

Kelly Chang, Andrew Cheng, San-He Wu

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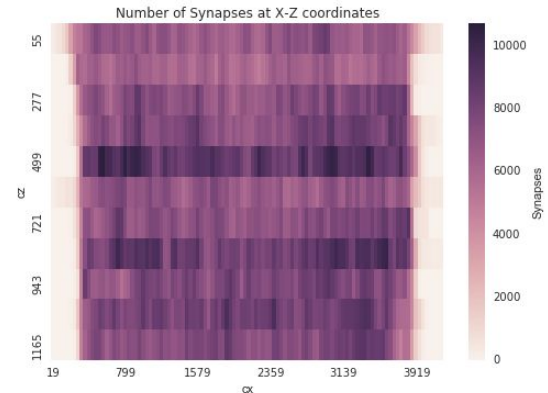
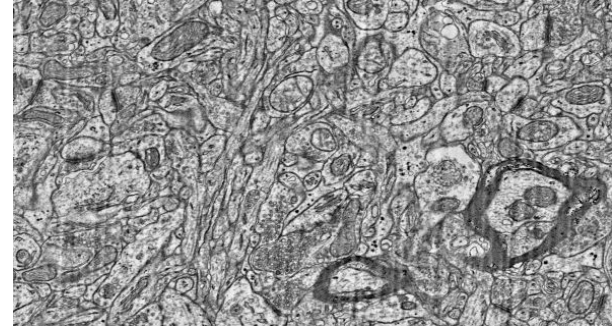
# Significance

- The ultimate goal of systems neuroscience is to fully map the human connectome
- Traditional neuroscience techniques involve studying individual neurons in isolation. This cannot capture:
  - ***Spatial Distribution***
  - Connectivity
  - Information processing



# Need/Gap

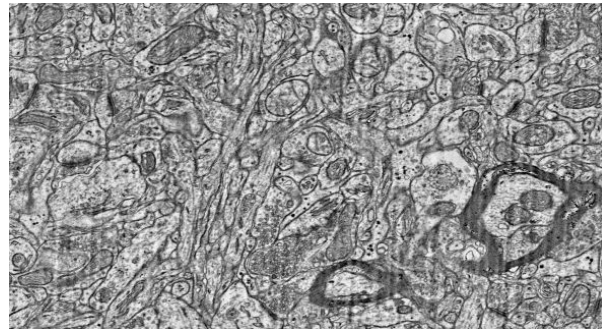
- Electron microscopy now affords the ability to map entire sections of cortex in detail
- It is now possible to assess the distribution of synapses in cortex
  - Also possible to assess Connectivity
- This project will focus on synaptic density across “layers” of cortex



# Challenges

- Annotation of neural markers

- Labeling of synapses
- Labeling of neurons / cell types
- Separating cortical layers



- Stratifying synapses

- This project is focused on synaptic density across “layers”
- Resolving different cortical layers from the EM data is non-trivial.
- We considered different Z-layers instead of cortical

# Formal Statement of Problem

$N = \# \{ \text{Synapses} \}$

$X_i = \text{X position} \quad Y_i = \text{Y position}$

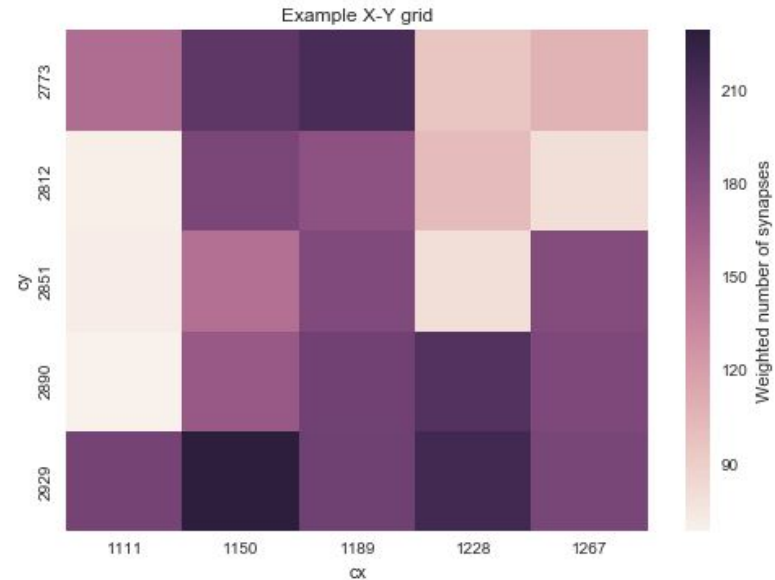
$X_i, Y_i, N \sim F := \{F_{X,Y,N} (.,\theta): \theta \in \Theta\}$

$W_i = \{ \text{High Density}, \text{Low Density} \}$

$$L = \sum_{-\infty}^{\infty} I(\hat{W} \neq W)$$

$$E[L] = \sum_{-\infty}^{\infty} I(\hat{W} \neq W) / N$$

- Z layers labeled (8 low, 3 high density)
- Classify grids by mean synapses per bin
- Get the expectation of loss function



# Model Assumptions

- Grid means are independent of X, Y dimensions

$$X_i, Y_i \parallel U_i$$

- Grid means are i.i.d

$$(u_1, u_2, \dots, u_i) \sim F = \prod F_i, F_i = F_j, \forall i \neq j$$

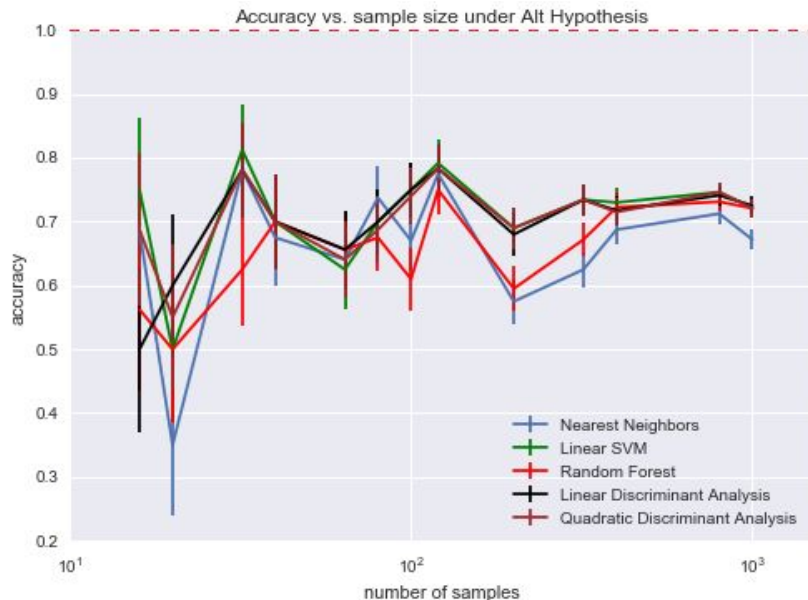
- Class conditional difference exist

$$f(x) = y + \varepsilon, \varepsilon \sim \text{Normal}(\mu, \sigma^2)$$

# Formal Statement of Algorithm

- Linear Discriminant Analysis (LDA)
- Quadratic Discriminant Analysis (QDA)
- K-Nearest Neighbors (kNN)
- Support Vector Machines (SVM)
- Random Forest (RF)

# Results

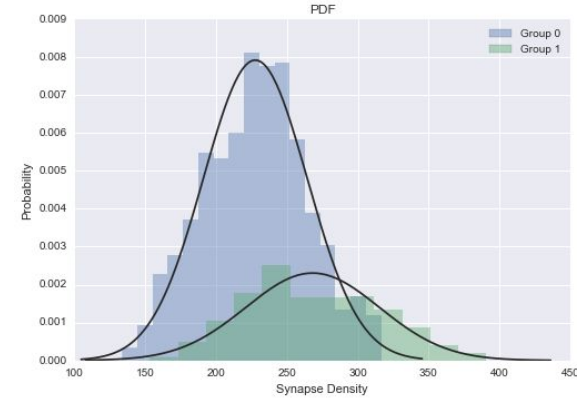
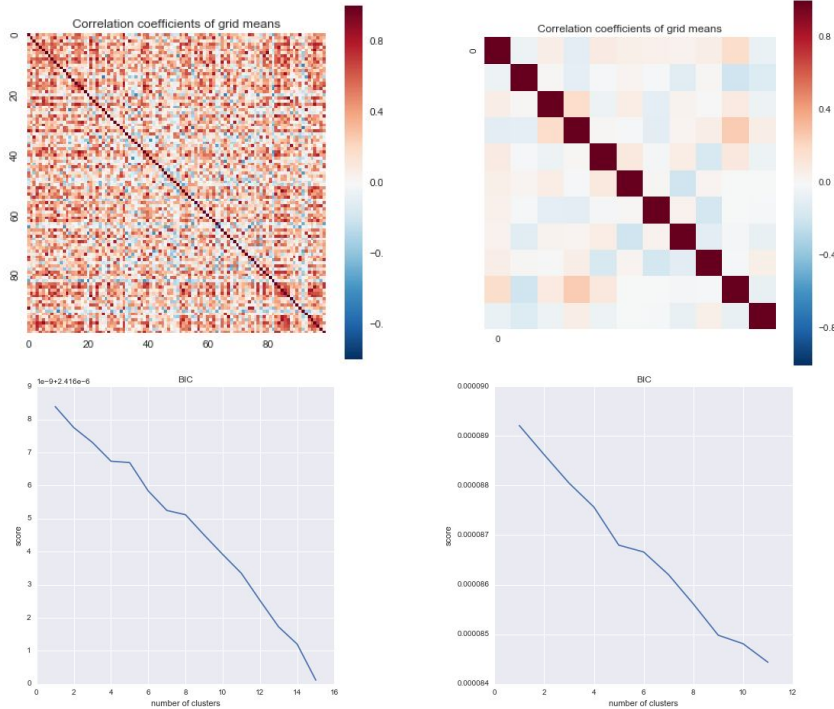


<u>Algorithm</u>	<u>Accuracy</u>
Nearest Neighbors	$0.74 \pm 0.01$
Linear SVM	$0.78 \pm 0.01$
Random Forest	$0.77 \pm 0.01$
LDA	$0.78 \pm 0.01$
QDA	$0.78 \pm 0.01$

- All classifiers performed similarly, with accuracy slightly above 73% (prior probability of low-density layer).
- Grid means don't seem to provide enough additional information.



# Model Checking



- Grid means not independent across (X,Y)
- Grid means independent across Z
- Grid means identically distributed in both (X,Y) and Z
- Grid means aren't sufficient to distinguish between layers

# Discussion / Future Work

- Framework for testing classifiers and identifying strengths/weaknesses
- Preliminary results indicate that more information is needed for successful classification of Z layers
  - Larger grids
  - Alternative metric (instead of means)
  - Quantify spatial organization
    - Clustering
- Functional information
  - Cortical layers vs. Z layers