

# Mouse Project

Final Results

Rose, Jessie, Shuting

# Agenda — what we did

- **Identify research subjects: Zero\_Maze experiment**

Mouse 409 w/ 25 neurons and all neurons

All mice w/ all neurons

- **Apply different models**

Logistic Model with PCA, Simple Neural Network, RNN, Bidirectional LSTM

- **Outputs**

Confusion Matrix for each model and Accuracy Comparison plot

# Conclusions

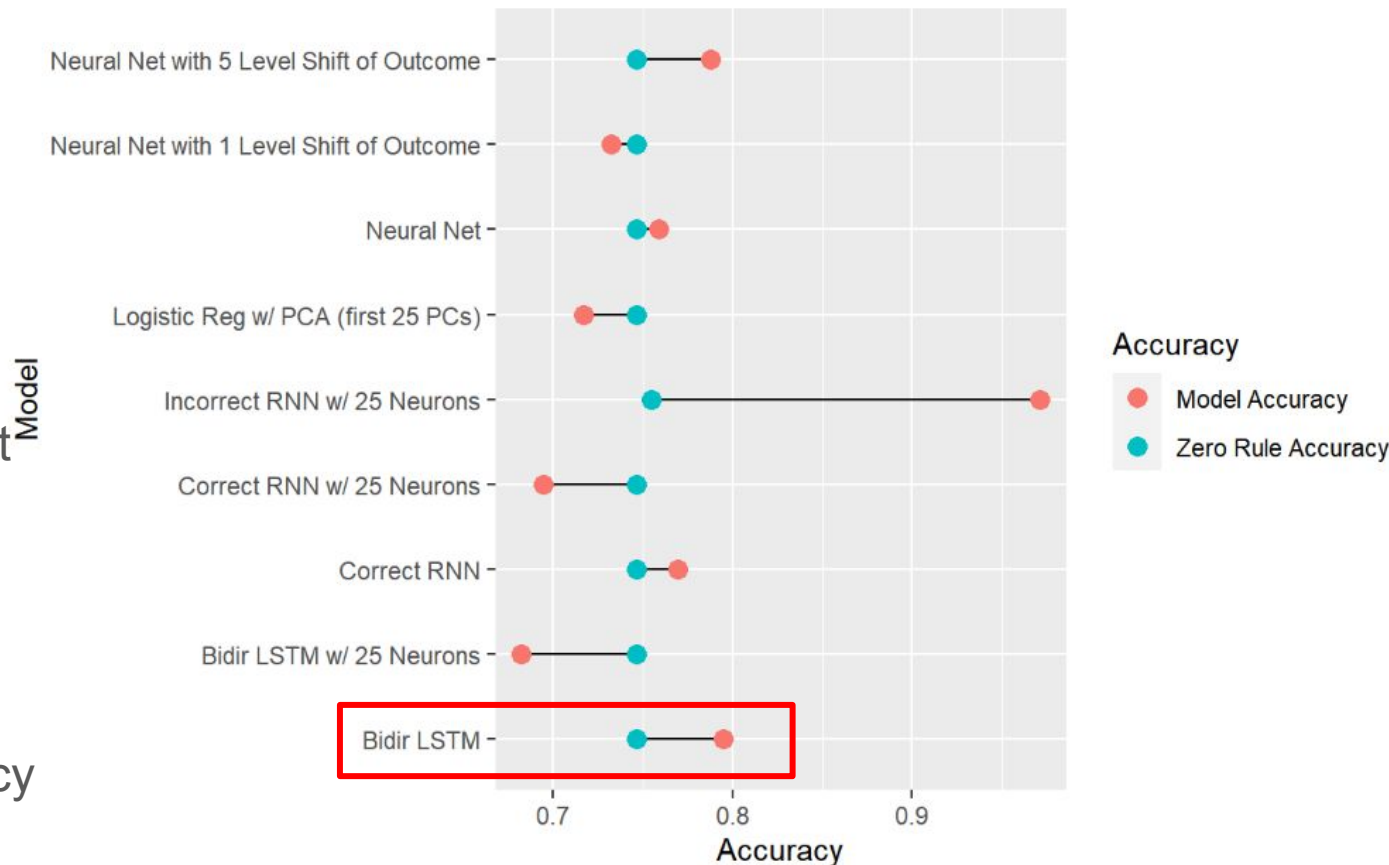
- **Split train / test in chunks** instead of randomly selected to ensure the sequence of time series
- Take **all neurons** into consideration to improve the predicted accuracy
- **Try different lags**, and it turns out that past events / neuron activity may not influence current outcome
- **Shift the behavior** to explore the causal relationship between neuron activation and behaviors

# Results — what we found

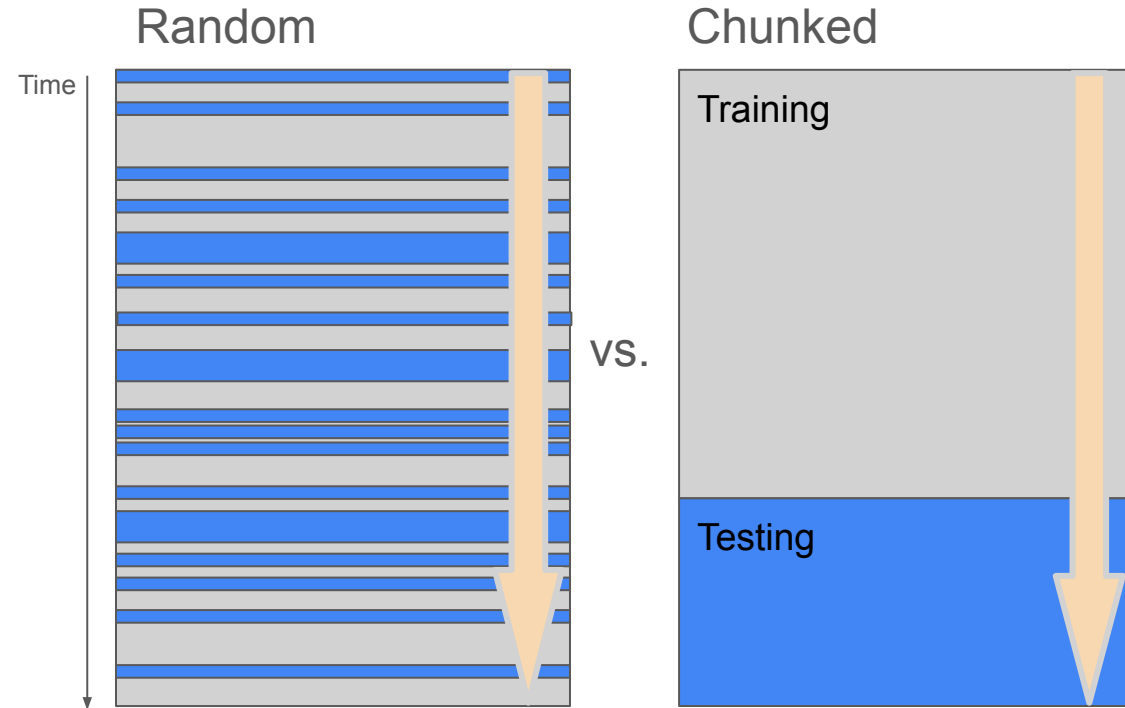
- **Model Comparisons with Mouse 409**

- **Zero Rule Accuracy =**  
Selecting the most common class

- **Goal =**  
Model Accuracy > Zero Rule Accuracy



# Train/Test Split of Data



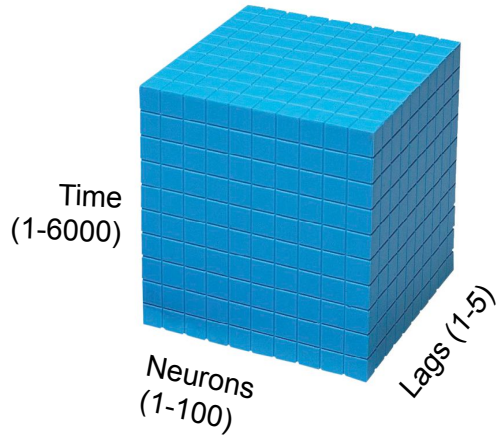
Random - learns a little bit of everything but loose sequence

Chunked - preserves sequence

*Does a neuron convey the same information throughout the trial?*

# Bidirectional LSTM (Long Short Term Memory)

## Input Data



## LSTM Overview

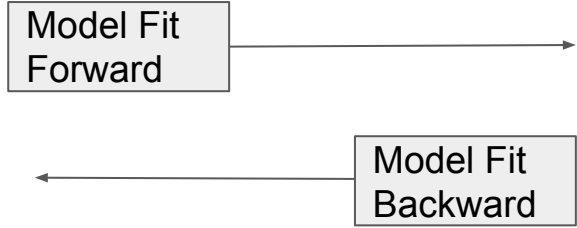
### Memory

Model that retains prior sequence information

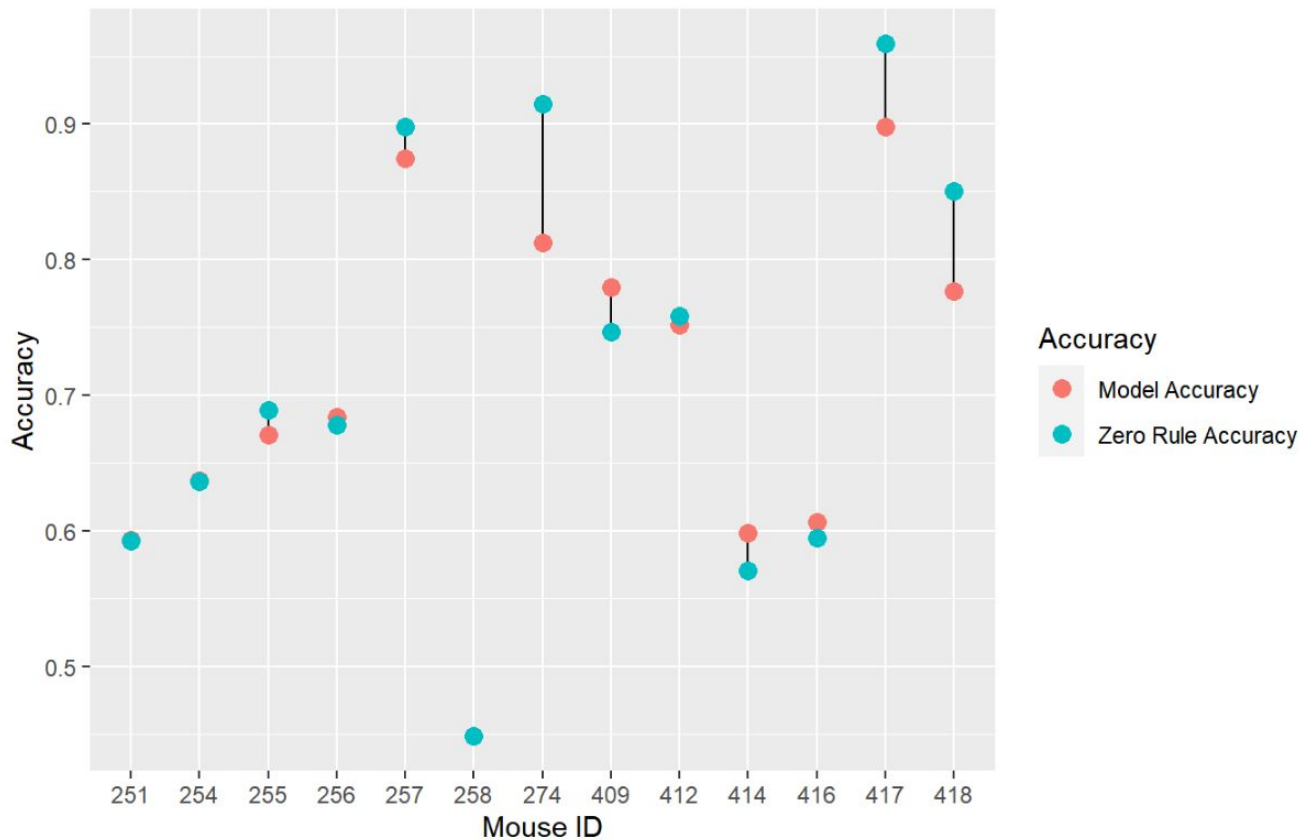
## Bidirectional

### Two directions

Information from before and after time point of interest

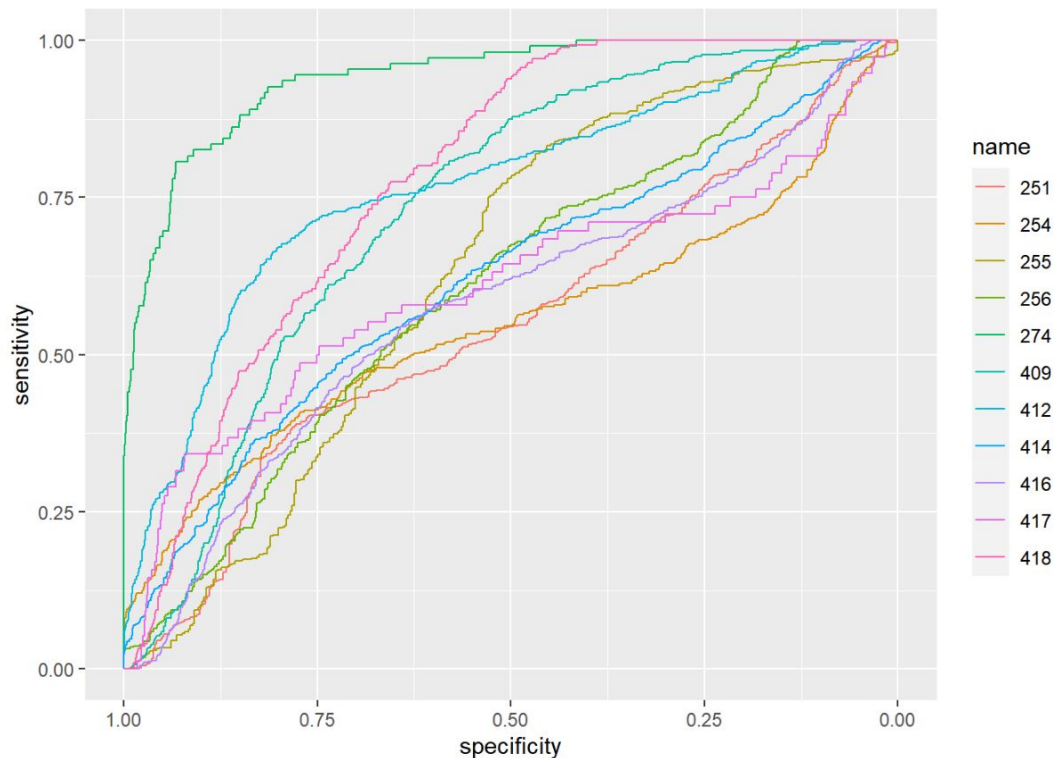
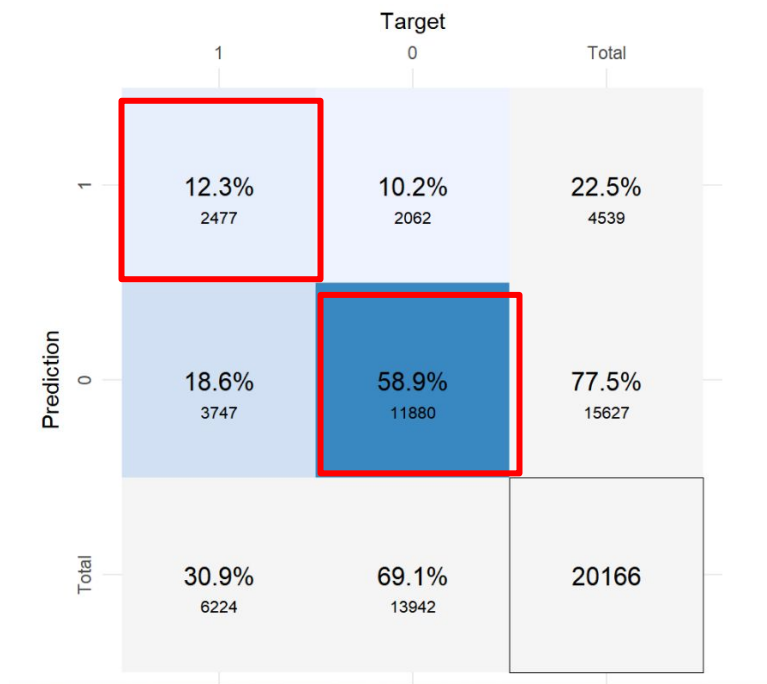


# Results — what we found



# Logistic Model for all mice—— other model we tried (1)

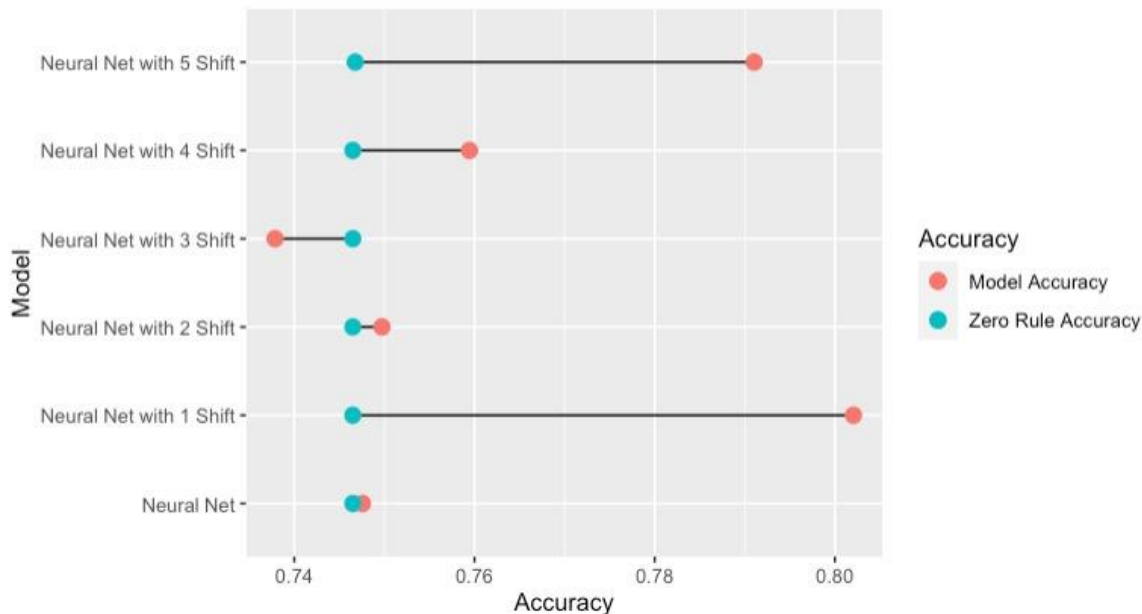
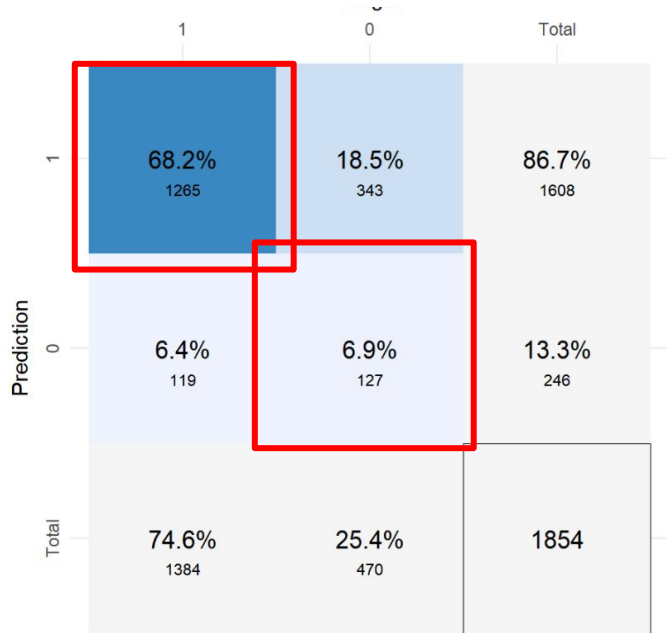
- Select neurons (predictors) with PCA=25, make confusion matrix and ROC lines.
- Accuracy is around 71%





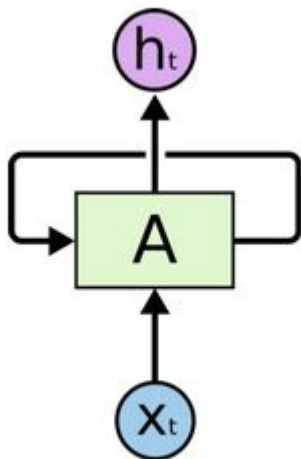
# Simple Neural Network — other model we tried (2)

- Features: not take into account the time series predicts the current location based on the current state of the neurons
- Accuracy = 75%
- **Shift the behavior (Y)** with 1 to 5 position forward



# RNN — other model we tried (3)

- Sequential RNN
- Accuracy: 77%

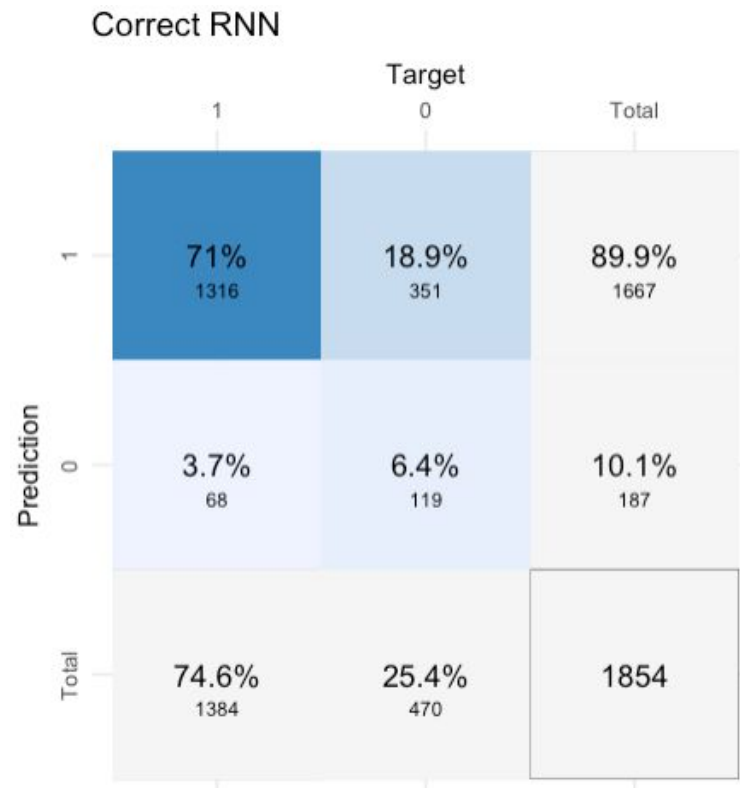
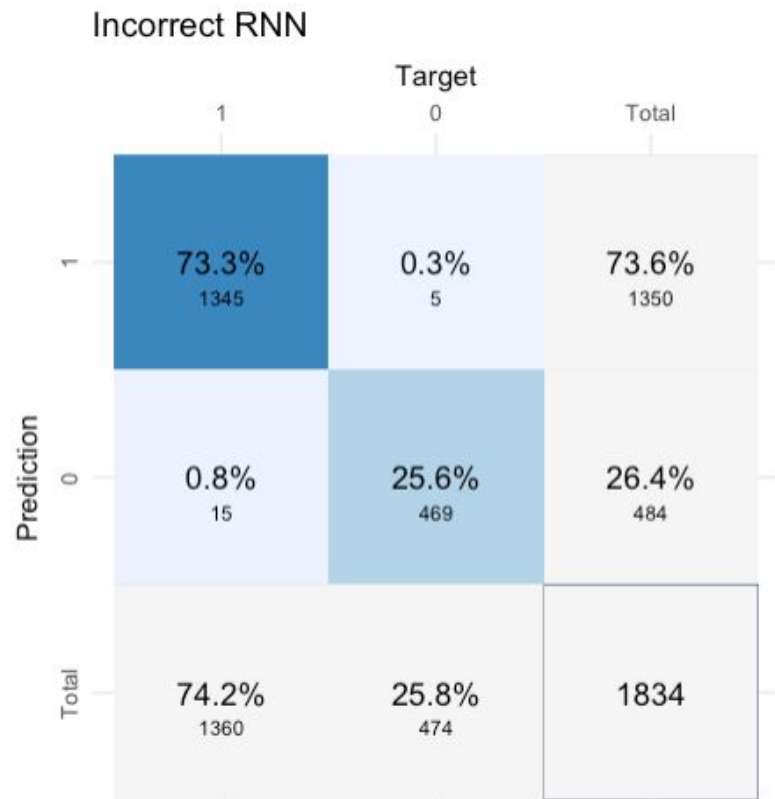


		Target		Total
		1	0	
Prediction	1	71% 1316	18.9% 351	89.9% 1667
	0	3.7% 68	6.4% 119	10.1% 187
Total		74.6% 1384	25.4% 470	1854

Thank you

# Appendix

- Randomly Selected vs. Sequential RNN



# Mouse Project

Final Results - Flipgrid Presentation

Rose, Jessie, Shuting

# Agenda — what we did

- **Identify research subjects**

Zero Maze with Mouse 409 (25 neurons and all neurons)

- **Apply different models**

Logistic Model with PCA, Simple Neural Network, RNN, Bidirectional LSTM

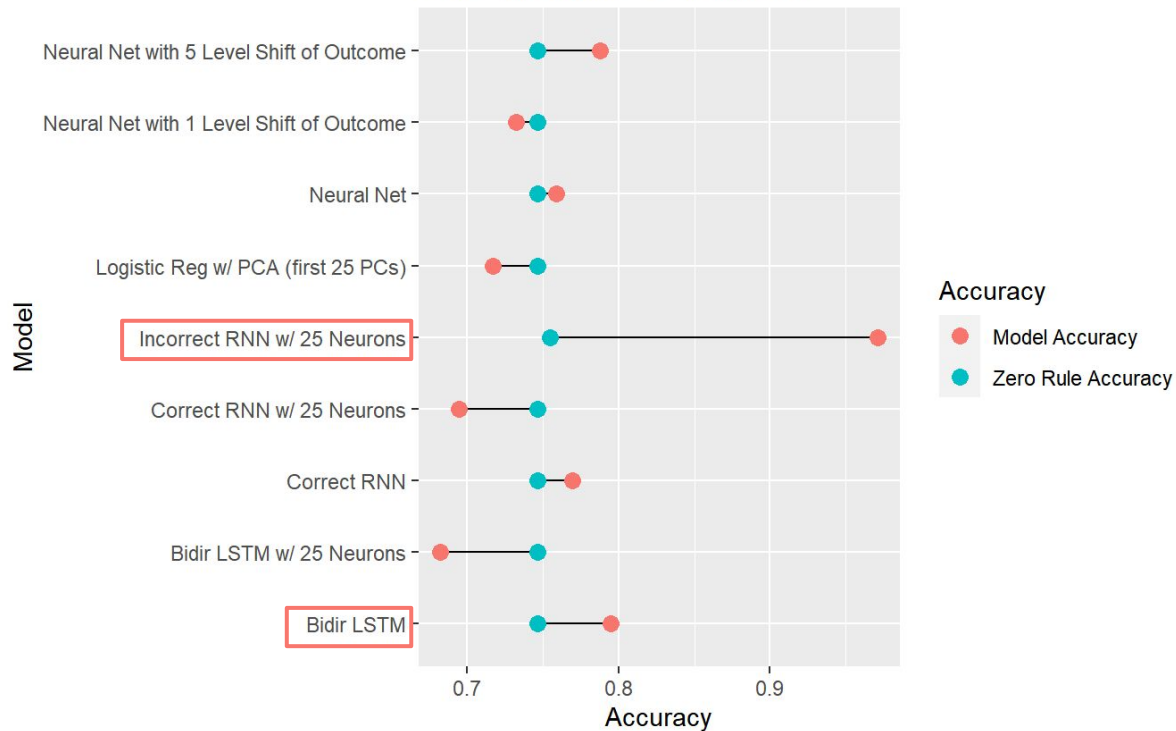
- **Detailed Updated**

Train/test splitting matters

# Model Comparisons with Mouse 409

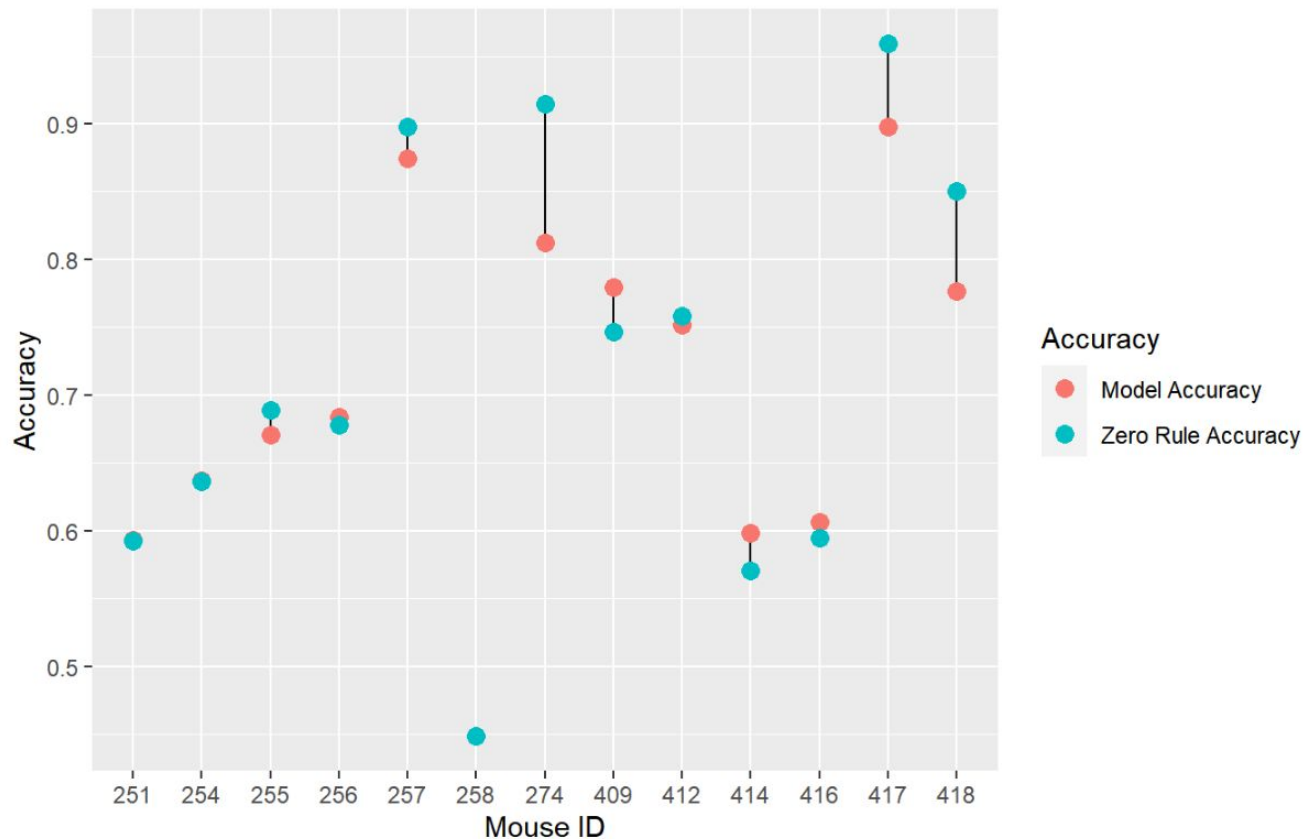
**Zero Rule Accuracy =**  
Selecting the most  
common class

**Goal =**  
Model Accuracy > Zero  
Rule Accuracy



# Results — what we found

- Accuracy of Test Data with Bidirectional LSTM
- **All Mice w/ all neurons**





# Conclusions

- Split train / test in **chunks**
- Take **all neurons** into consideration
- Try different **lags**

- **Key Questions:**

Does a neuron carry the same information throughout the 10 minute trial?

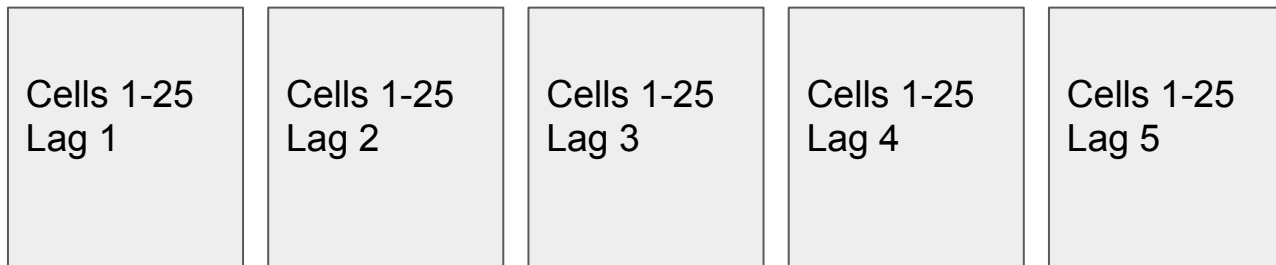
Does behavior influence neural activity? OR Does neural activity influence behavior?

# Mouse Project

Simple RNN(Based on Mouse 409 in Zero-Maze)

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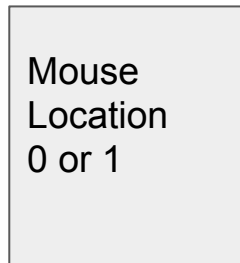
# Input Data



Subsample of 25 neurons with 5 time lags.  
Total Rows = 6179 (70% training; 30% testing)

# Output Data

Total Rows = 6179



# Model Details

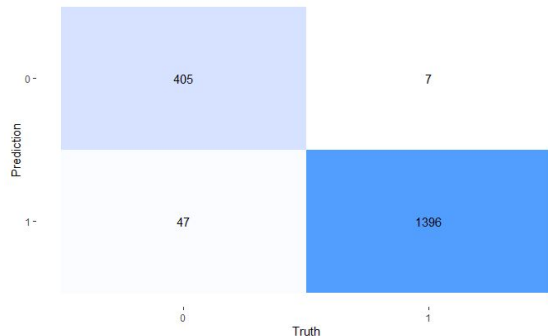
```
model <- keras_model_sequential () %>%  
  layer_simple_rnn ( units = 100 ,  
                    input_shape = list (n_lags , n_neurons) ,  
                    dropout = 0.1 , recurrent_dropout = 0.1) %>%  
  layer_dense ( units = 50, activation = "relu") %>%  
  layer_dense ( units = 1, activation = "sigmoid")  
model %>% compile ( optimizer = optimizer_rmsprop() ,  
                  loss = "binary_crossentropy",  
                  metrics = c("accuracy") )
```

- Three layers
- Activation Function: ReLu, Sigmoid
- Loss Function: Cross-entropy

## Result

Epochs = 50

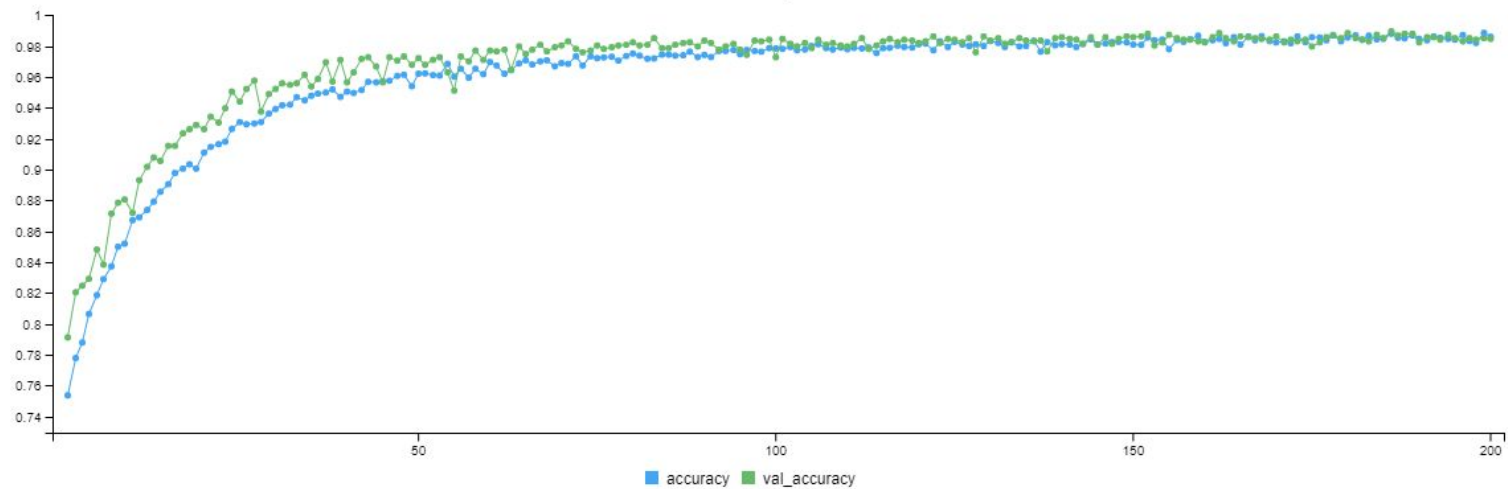
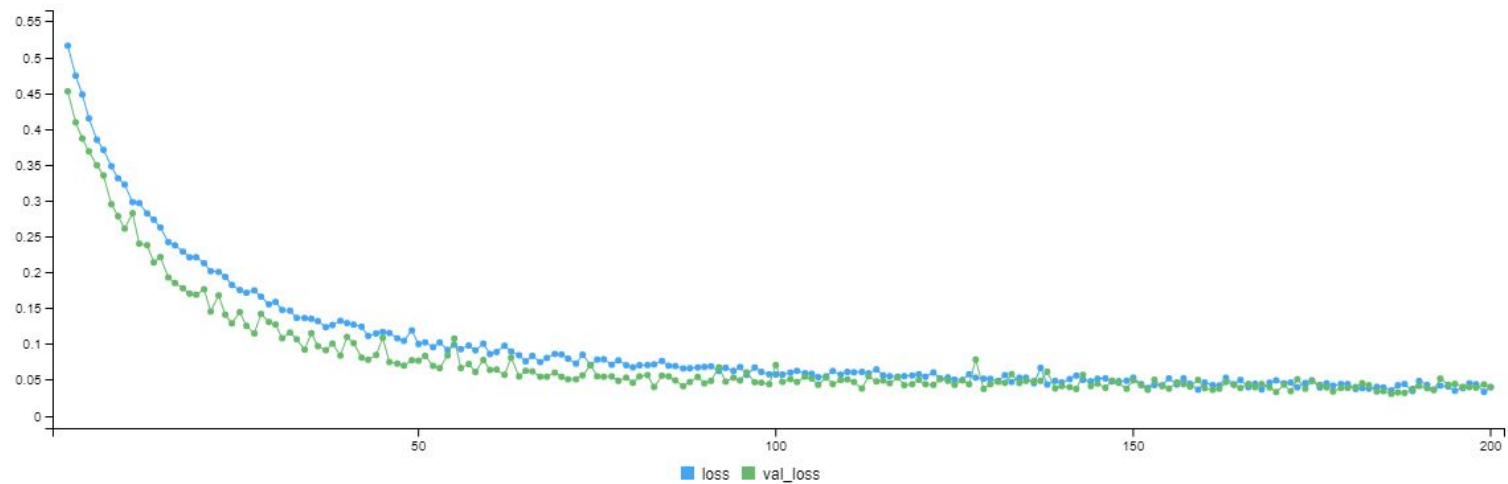
Accuracy: 0.962



Epochs = 200

Accuracy: 0.985





## Next Steps:

- Shuffle the neurons for the RNN model to get a more general result.
  - Try neural network with PCA.
  - Try add some noise (more neurons).
- Try different “windows”: 10 seconds, 25 seconds, etc.
- Make ROC lines for every mouse and try the different number of components for PCA.
- Explore the correlated neurons according to the consistency of correlated and PCA plots. Are they distributed in a cluster or randomly (space or distant)?

# Mouse Project

Baseline Model (Based on Mouse 409 in Zero-Maze)

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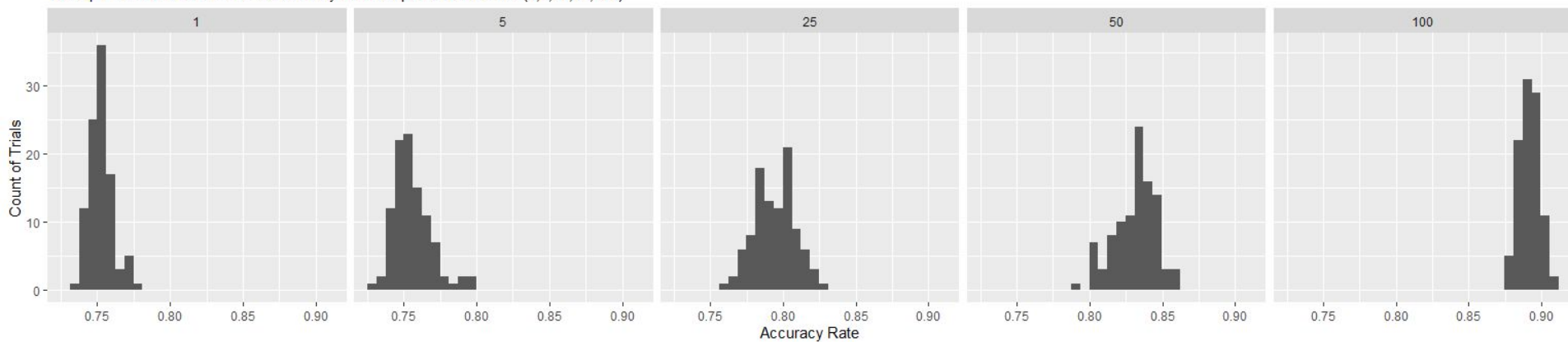
# Logistic Regression

“The **logistic regression** was trained and tested using an increasing interval of randomly selected cells with 1000 repetitions for each number of cells”

Johnson et al. (2022)

## Accuracy of Logistic Regression Models

100 repetitions of each number of randomly selected predictor neurons (1,5,25,50,100)





# Logistic Regression Considerations

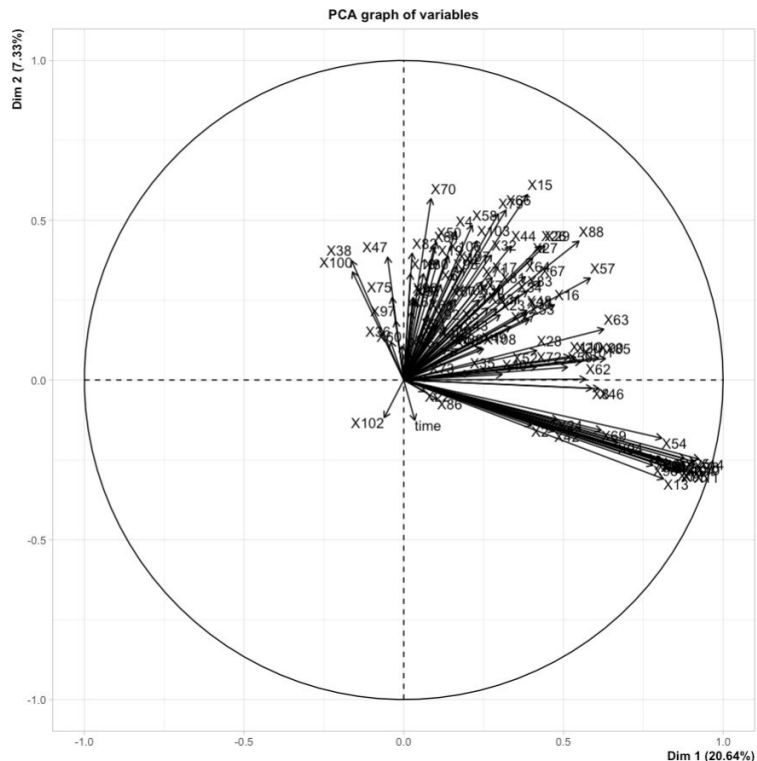
- Does not account for time and autocorrelation
- Correlation between neurons
- Unbalanced data (In total: 1 = 4647 rows; 0 = 153 )

Prediction	Reference	
	1	0
1	2137	207
0	144	606

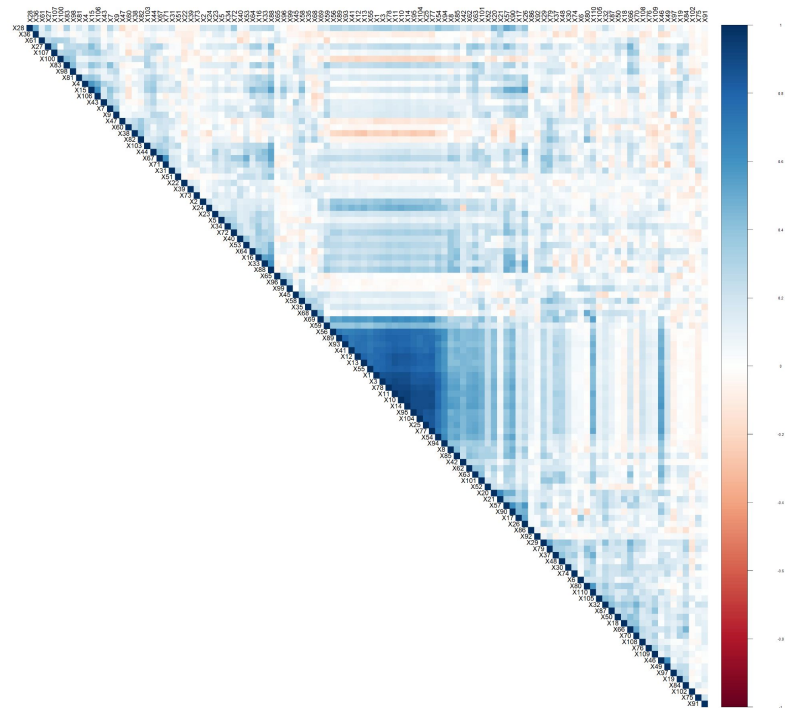
Accuracy : 0.8866

# Deal with Correlation between Neurons (PCA)

## PCA Plot



## Correlation Plot



# Logistic Regression and PCA

## Eigenvalue & Contribution Format

	eigenvalue	variance.percent	cumulative.variance.percent
Dim.1	22.90938441	20.63908506	20.63909
Dim.2	8.13994788	7.33328637	27.97237
Dim.3	5.94709912	5.35774695	33.33012
Dim.4	4.75210760	4.28117802	37.61130
Dim.5	3.56853359	3.21489513	40.82619
Dim.6	2.89839073	2.61116282	43.43735
Dim.7	2.63542867	2.37426007	45.81161
Dim.8	2.45687416	2.21340015	48.02501
Dim.9	2.03942374	1.83731868	49.86233
Dim.10	1.87910008	1.69288296	51.55522
Dim.11	1.84050439	1.65811207	53.21333
Dim.12	1.71331261	1.54352488	54.75685
Dim.13	1.59762885	1.43930527	56.19616
Dim.14	1.50689683	1.35756472	57.55372

Result when we choose first 40 principal components:

```
Call:
glm(formula = open ~ ., family = binomial("logit"), data = training)
```

Deviance Residuals:

Min	1Q	Median	3Q	Max
-3.07276	-0.06894	0.24742	0.52941	2.66530

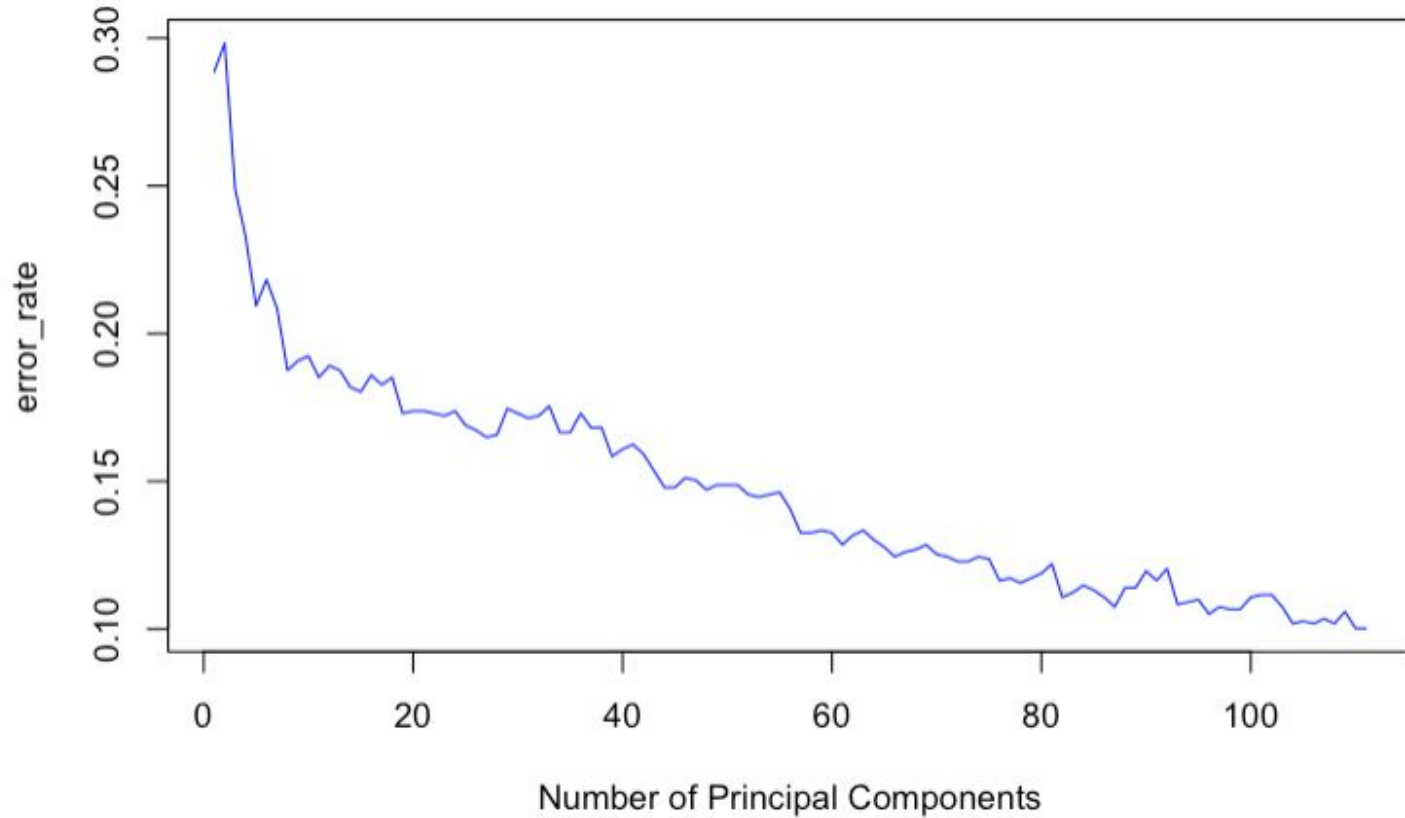
Coefficients:

	Estimate	Std. Error	z value	Pr(> z )
(Intercept)	1.951235	0.062801	31.070	< 2e-16 ***
Dim.1	-0.031187	0.010739	-2.904	0.003683 **
Dim.2	0.046791	0.018421	2.540	0.011083 *
Dim.3	-0.611384	0.027581	-22.167	< 2e-16 ***
Dim.4	0.324319	0.024087	13.465	< 2e-16 ***
Dim.5	0.408173	0.029219	13.969	< 2e-16 ***
Dim.6	-0.133940	0.033148	-4.041	5.33e-05 ***
Dim.7	0.152302	0.033419	4.557	5.18e-06 ***
Dim.8	-0.423454	0.035565	-11.907	< 2e-16 ***
Dim.9	-0.006772	0.033604	-0.202	0.840291
Dim.10	0.291025	0.036563	7.959	1.73e-15 ***

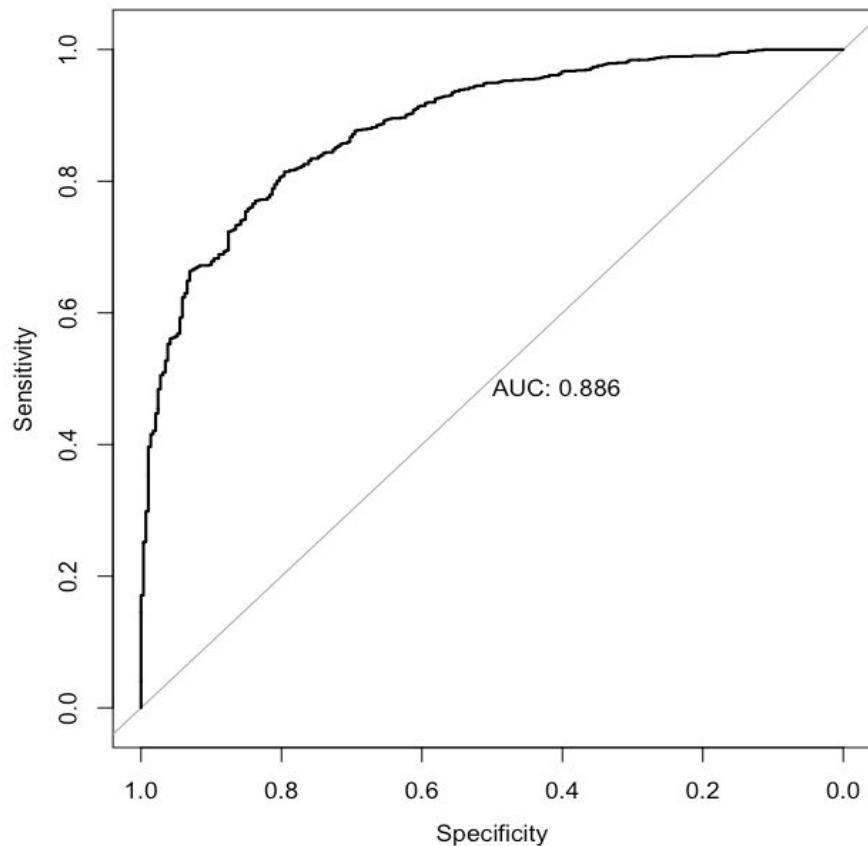
error\_rate

[1] 0.1657235

# Error Rate VS. Number of PC



# Deal with Unbalanced Data (ROC Curve)



- The AUC of our logistic model (number of components = 40) is 0.886, which means this classifier performs good even in unbalanced data

# Mouse Project

EDA

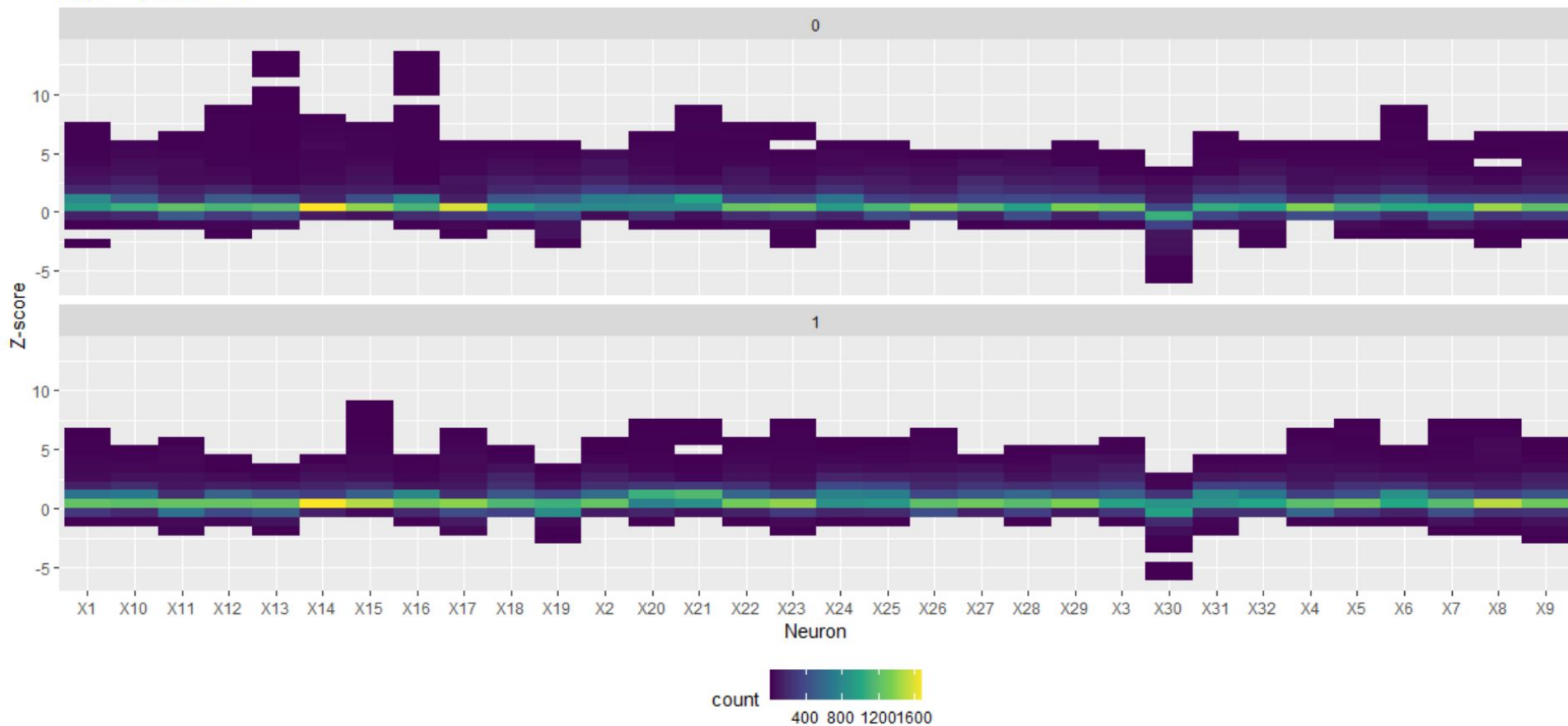
Rose, Jessie, Shuting

# Background

- Objective: decoding neural performance to analyse mice's social behavior
- Method: use machine learning skills to predict mice's behavior based on their neural recordings
- This Week: use mouse 251 in Zero\_Maze experiment as example, to do exploratory data analysis
- Next Steps: Continue EDA. Understand "Machine Learning for Neural Decoding" paper. Consider potential models.

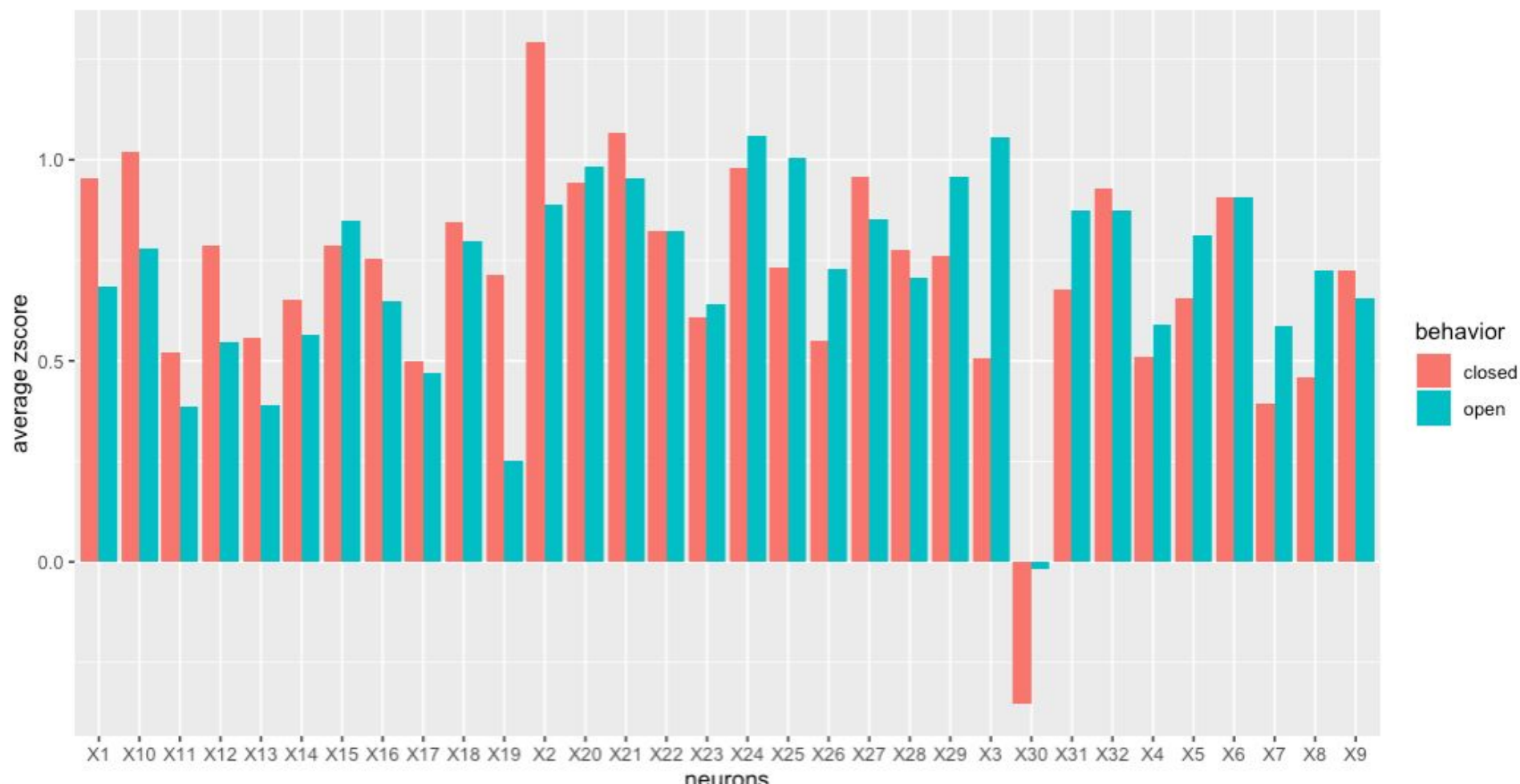
## Heatmap of Activation of One Mouse

Open = 1, Closed = 0

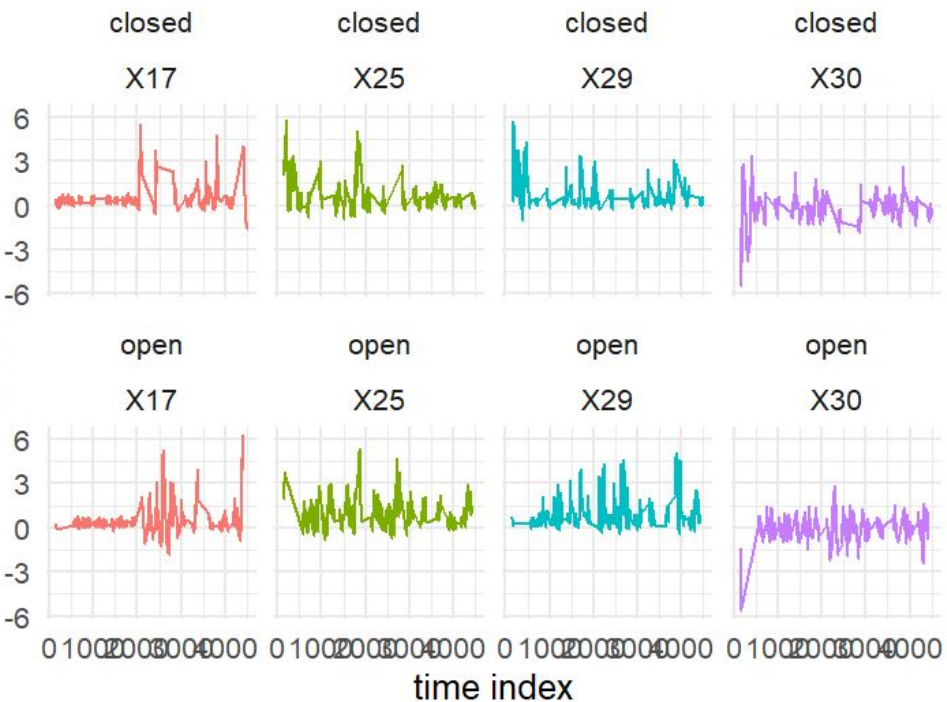




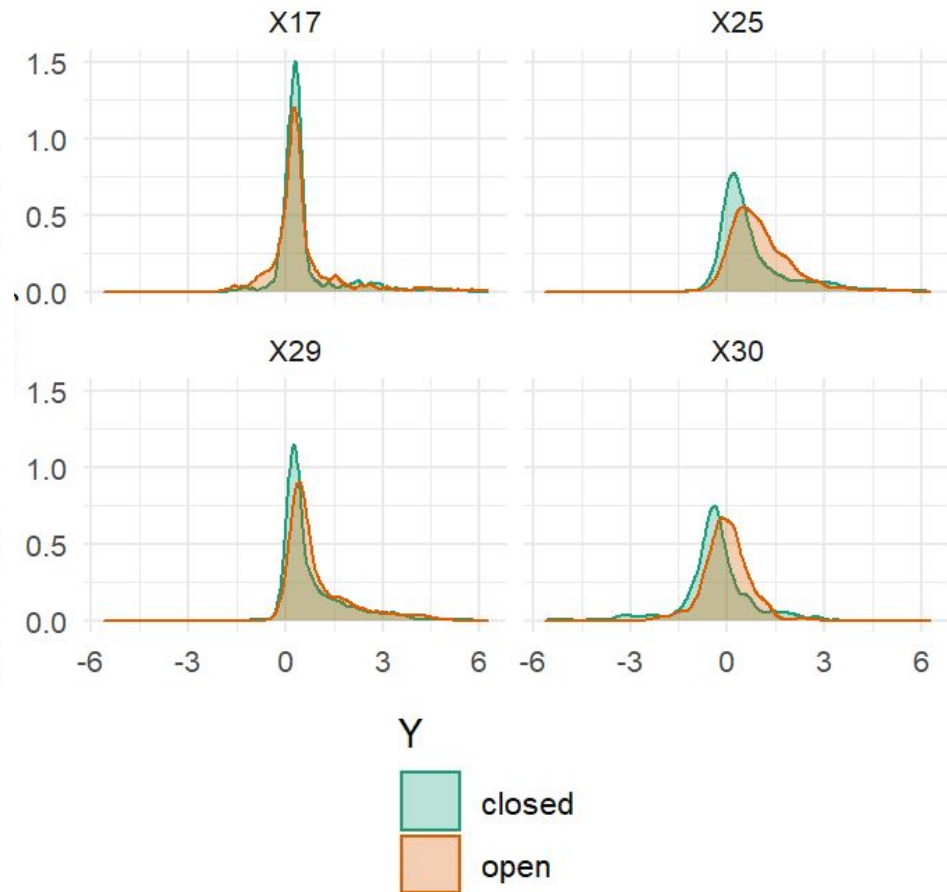
mouse 251 in Zero\_Maze experiment



# Time series plot of z-score



# Density plot of z-score



# Neuron Activity for Mouse 251

Subsample of Neurons

