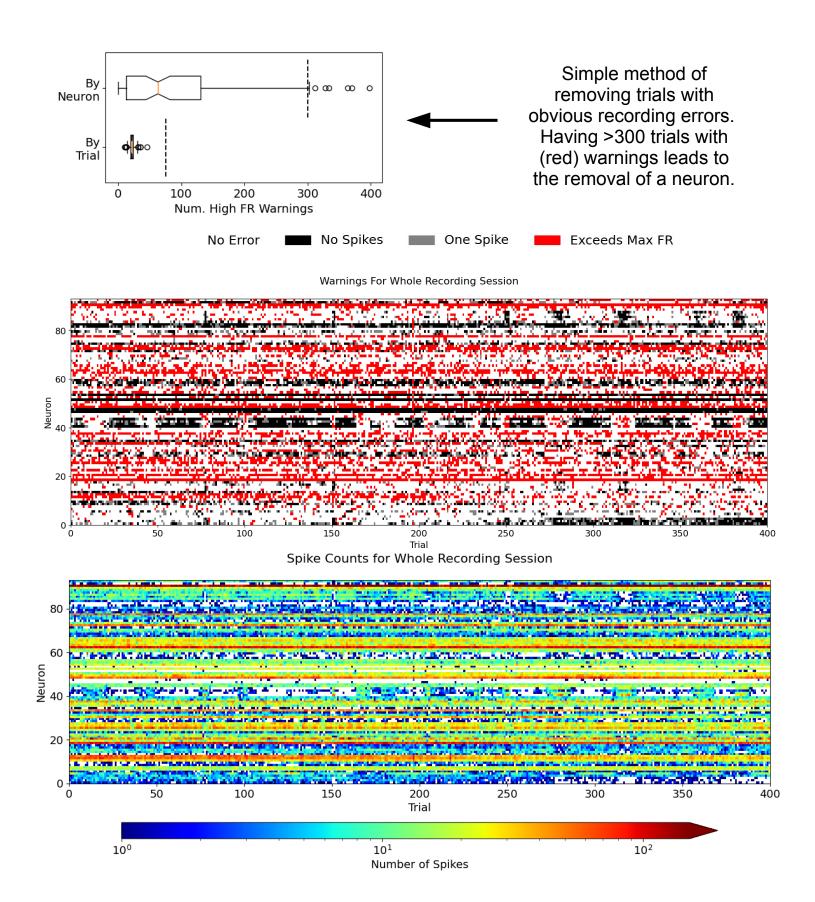
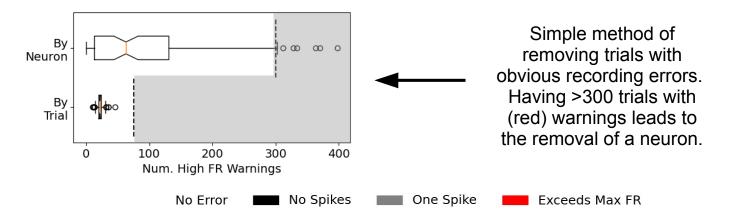
Source: https://github.com/timtyree/tbins.git



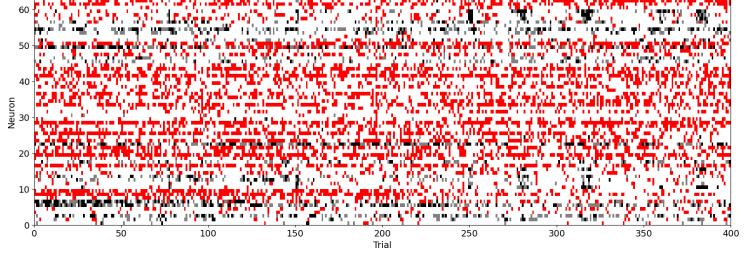
Source: https://github.com/timtyree/tbins.git

Remove neurons with unreasonably fast firing rates for most of the session.

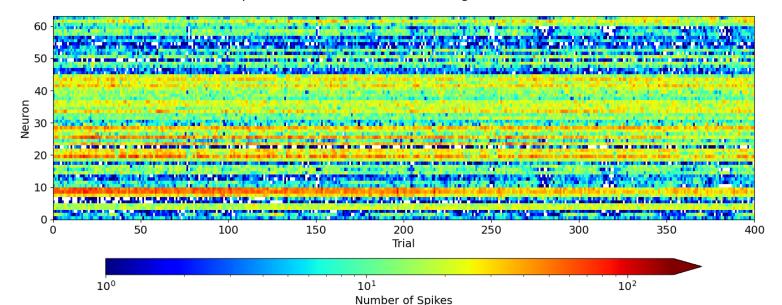
Similarly, remove neurons that aren't firing for unreasonably large amounts of time.



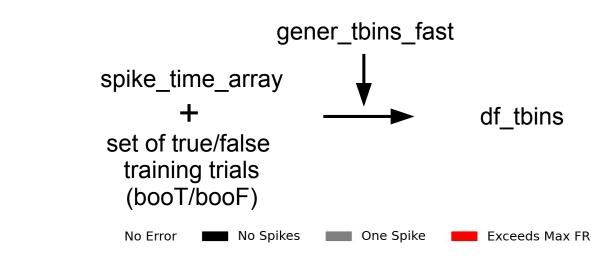
Warnings with Obvious Recording Errors Removed



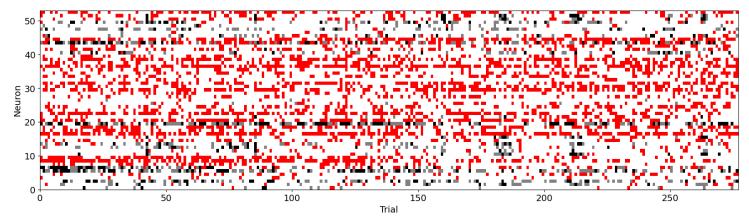
Spike Counts with Obvious Recording Errors Removed



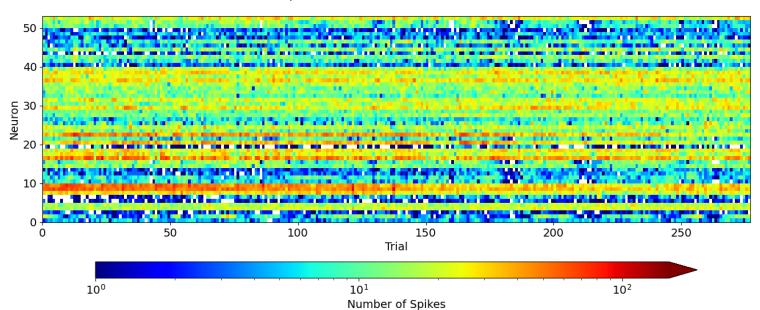
# **Apparent Predictive Neurons (#46)**



#### Warnings For Predictive Neurons

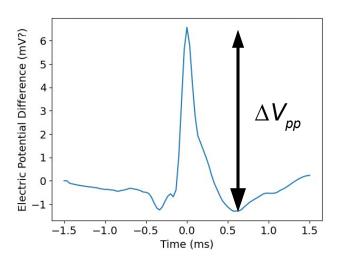


#### Spike Counts For Predictive Neurons



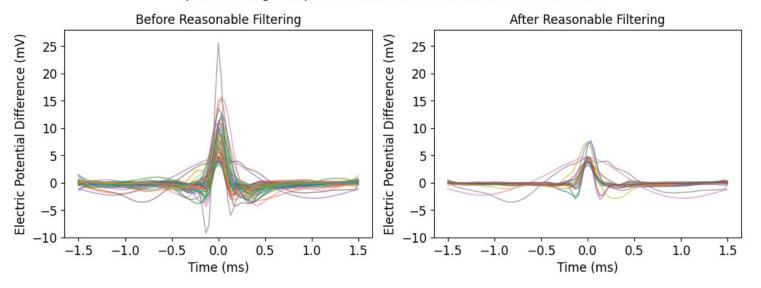
## **Apparent Predictive Neurons (#46)**

Larger spike amplitudes tend to offer worse prediction from an apparent predictive time bins.



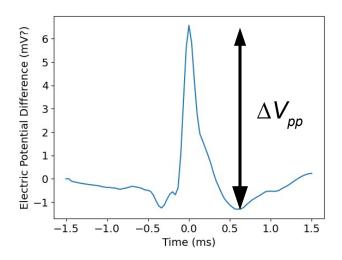
# (note the change in max $\Delta V_{pp}$ .)

Spike Sorting Templates of Predictive Neurons (Session #46)

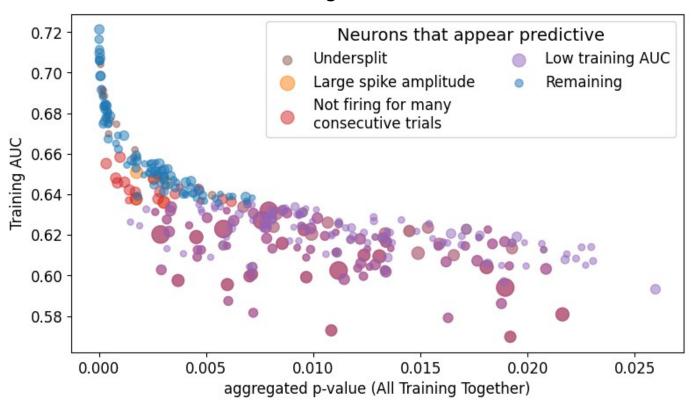


#### **Apparent Predictive Neurons (#46)**

Larger spike amplitudes tend to offer worse prediction from an apparent predictive time bins.



Large spike amplitudes together with other reasonable arguments lead to the selective removal of thins that tend to fail to generalize.

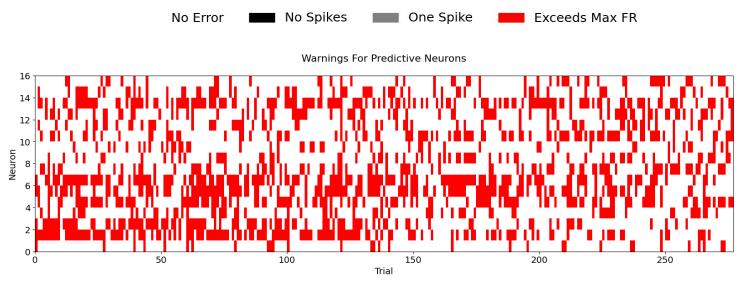


Sparsely firing neurons might be oversplit during spike sorting.

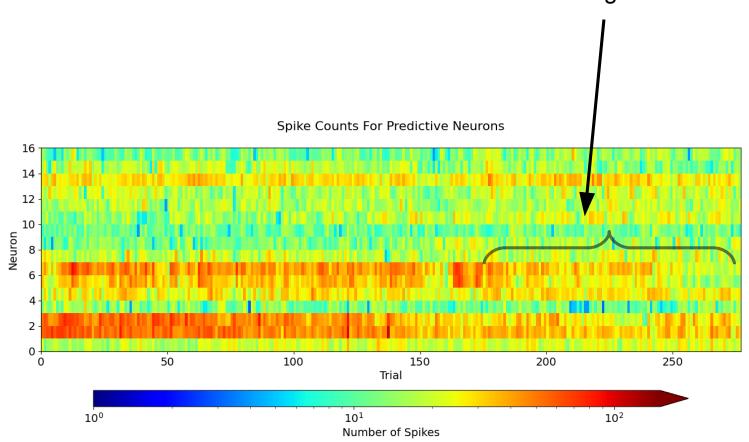
It's exceptionally difficult to tell oversplit units from properly categorized pyramidal cells.

## Predictive Neurons (#46)

# After reasonable filtering, we see fewer warnings

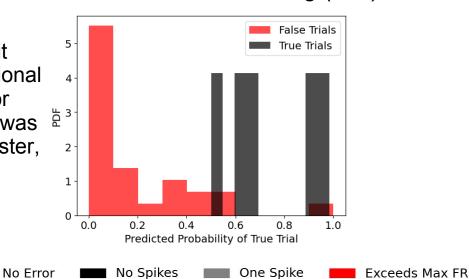


It looks like some neurons are firing less as the session goes on...

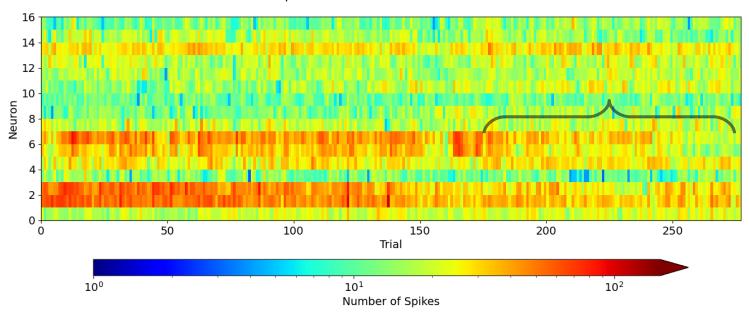


# **Predictive Neurons** *after reasonable filtering (#46)*

A random train-test split revealed exceptional separation for whether Hades was observing her sister, Hermes.



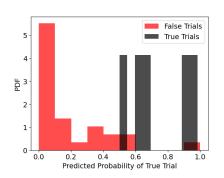
#### Spike Counts For Predictive Neurons

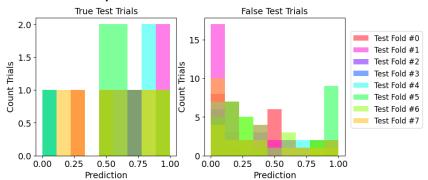


#### Identity-Specific Decoding of an Individual's Identity

Hades observing the face or voice of her sister, Hermes (#46).

# Histograms of the predicted probability that the face or voice of Hermes is present

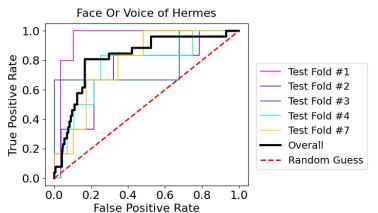




#### All Testing Folds

#### Face Or Voice of Hermes Test Fold #0 1.0 Test Fold #1 Test Fold #2 True Positive Rate 0.8 Test Fold #3 Test Fold #4 0.6 Test Fold #5 Test Fold #6 0.4 Test Fold #7 Overall 0.2 --- Random Guess 0.0 0.0 0.2 0.4 0.6 0.8 1.0 False Positive Rate

#### All Testing Folds with AUC>0.65



(left)
i\_on\_i: AUC: 0.7395, num\_trials: 277
 threshold as 0.5000 ==> tpr: 0.8372, fpr: 0.2991

```
*** test results aggregated over testing folds ***
```

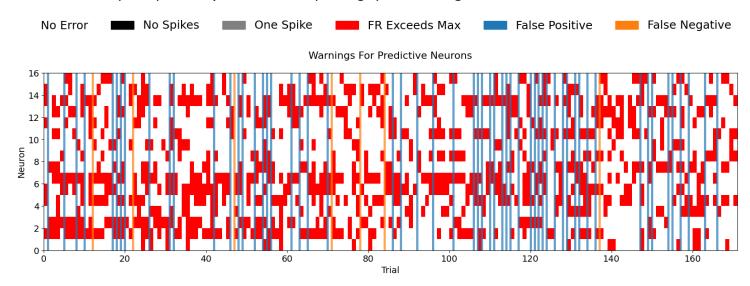
	200 <b>,</b>	506,	156,	Count	
testing auc:	[0.62823276	0.75	0.80288283	8.	]
testing accuracy:	[0.6752451	0.73529412	0.76638655	8.	]
testing sensitivity:	[0.16346154	0.28181818	0.4	8.	]
testing precision:	[0.44642857	0.66666667	0.66666667	8.	]
testing negative_predictive_value:	[0.71162403	0.76785714	0.80018473	8.	]

	∠ეశ <b>,</b>	১∪ <b>৪,</b>	158,	count	
<pre>num_true_testing:</pre>	[4.75	6.	6.	8.	]
<pre>num_false_testing:</pre>	[28.	29.	29.5	8.	]
num_features:	[11.75	13.	14.	8.	]

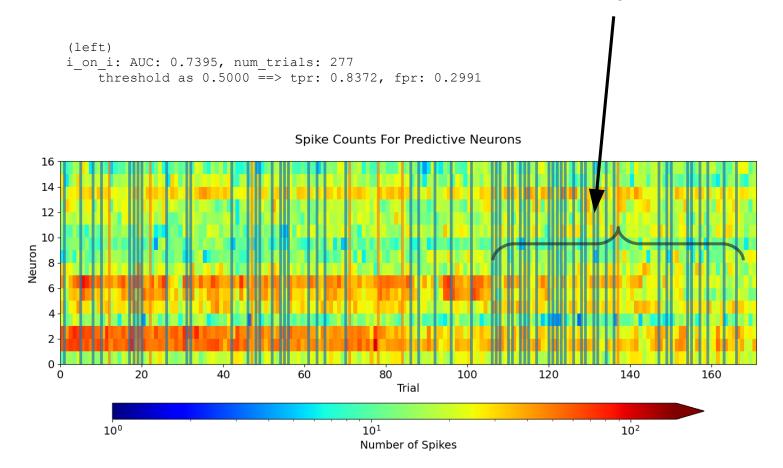
### Why did some folds perform poorly?

# Horizontal bars indicate all misclassifications predicted by a population-level neural decoder

Shown are (blue) false positives or (orange) false negatives.



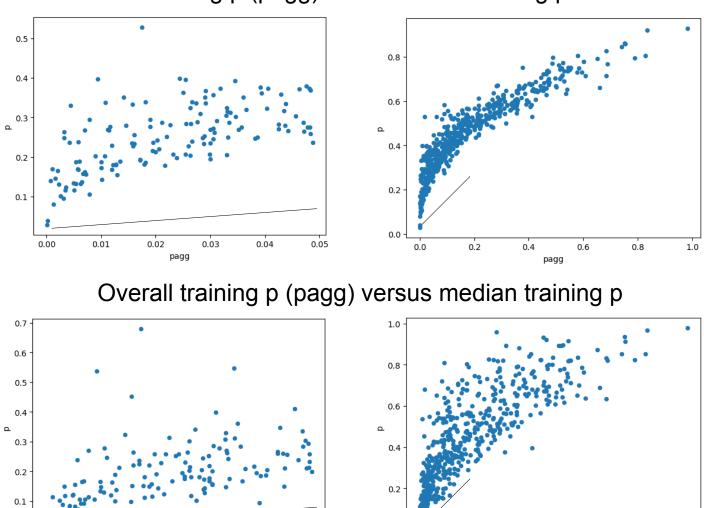
It looks like some neurons are firing less as the session goes on...



#### (Supplementary)

# Demonstration that the (y-axis) mean p-statistic is sufficient relative to (x-axis) Wilcoxon-Mann-Whitney test conducted over all of the training trials (#8)

#### Training p (pagg) versus mean training p



Note the apparent difference in spread.

The mean appears tighter and more stable in its statistical sufficiency. This supports our use of mean p versus median p.

0.8

pagg

0.05

0.04

0.03

0.0

0.00

0.01

0.02

pagg

All of these p-values were computed from the Wilcoxon-Mann-Whitney test for significant difference in median value, which makes no assumption of normality.