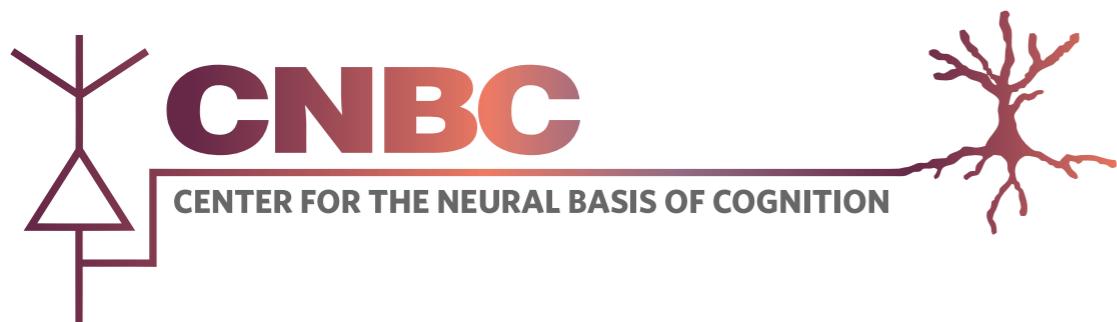


Pairwise Correlation

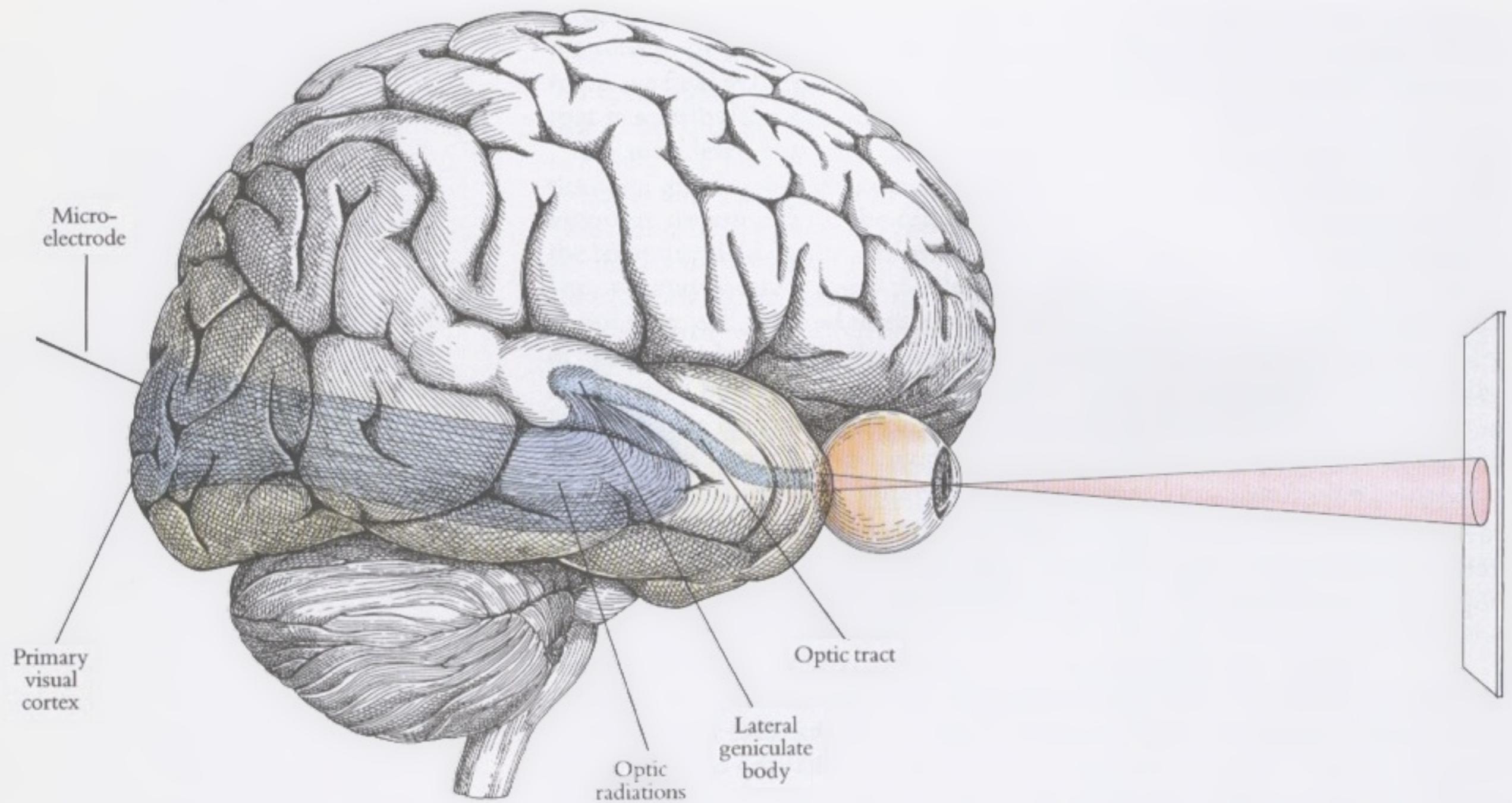
Cold Spring Harbor
Neural Data Science 2019

Matthew A Smith

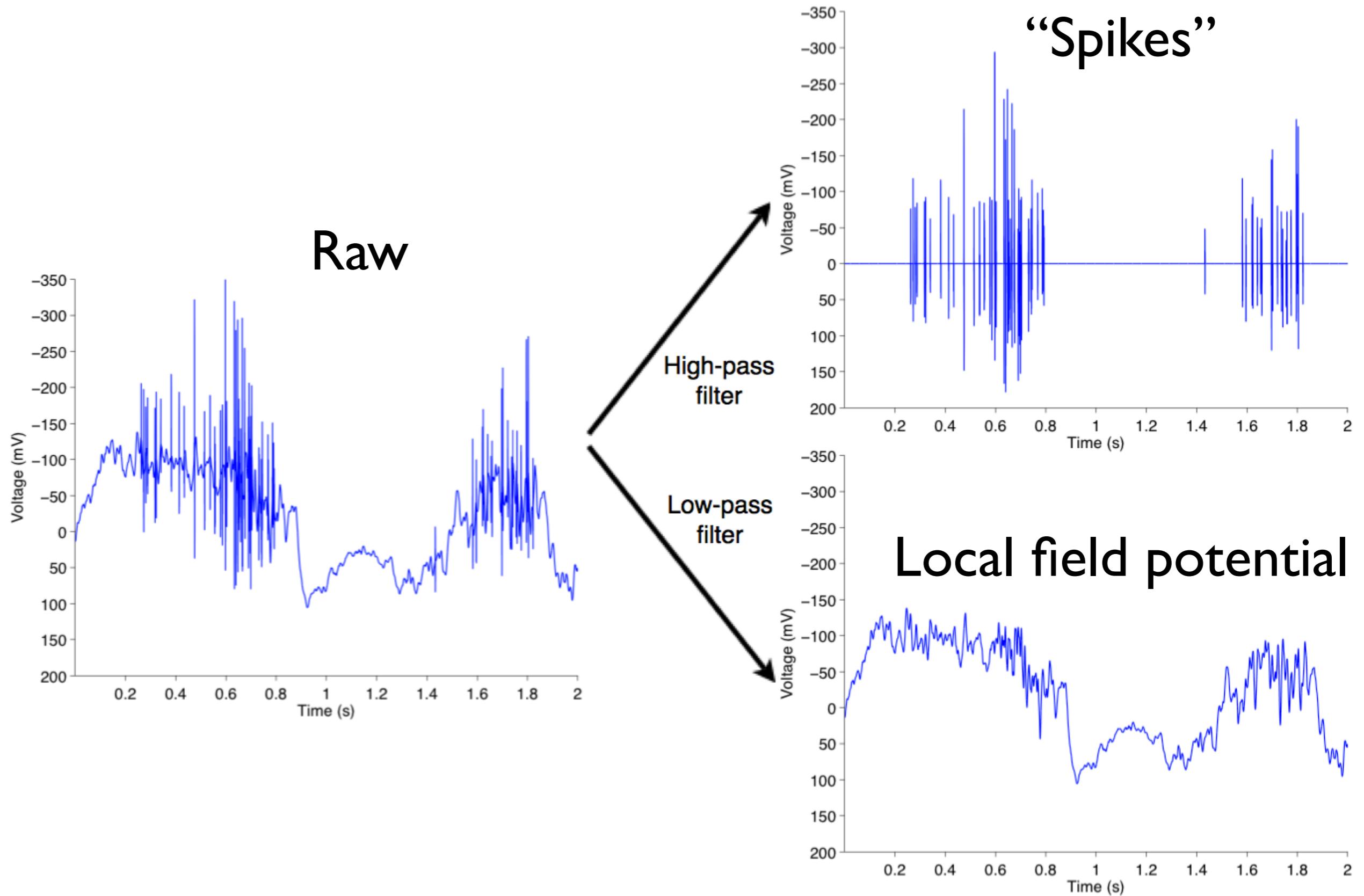
Departments of Biomedical Engineering
and Neuroscience Institute
(as of September 1)

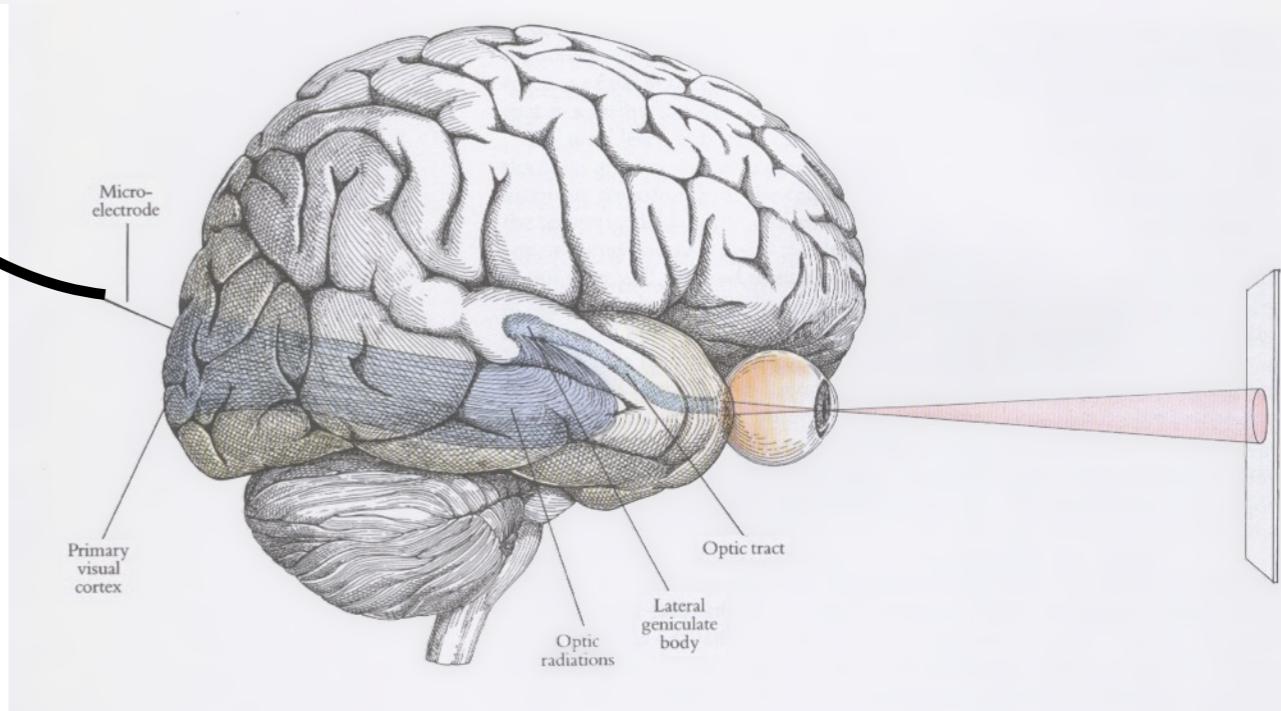
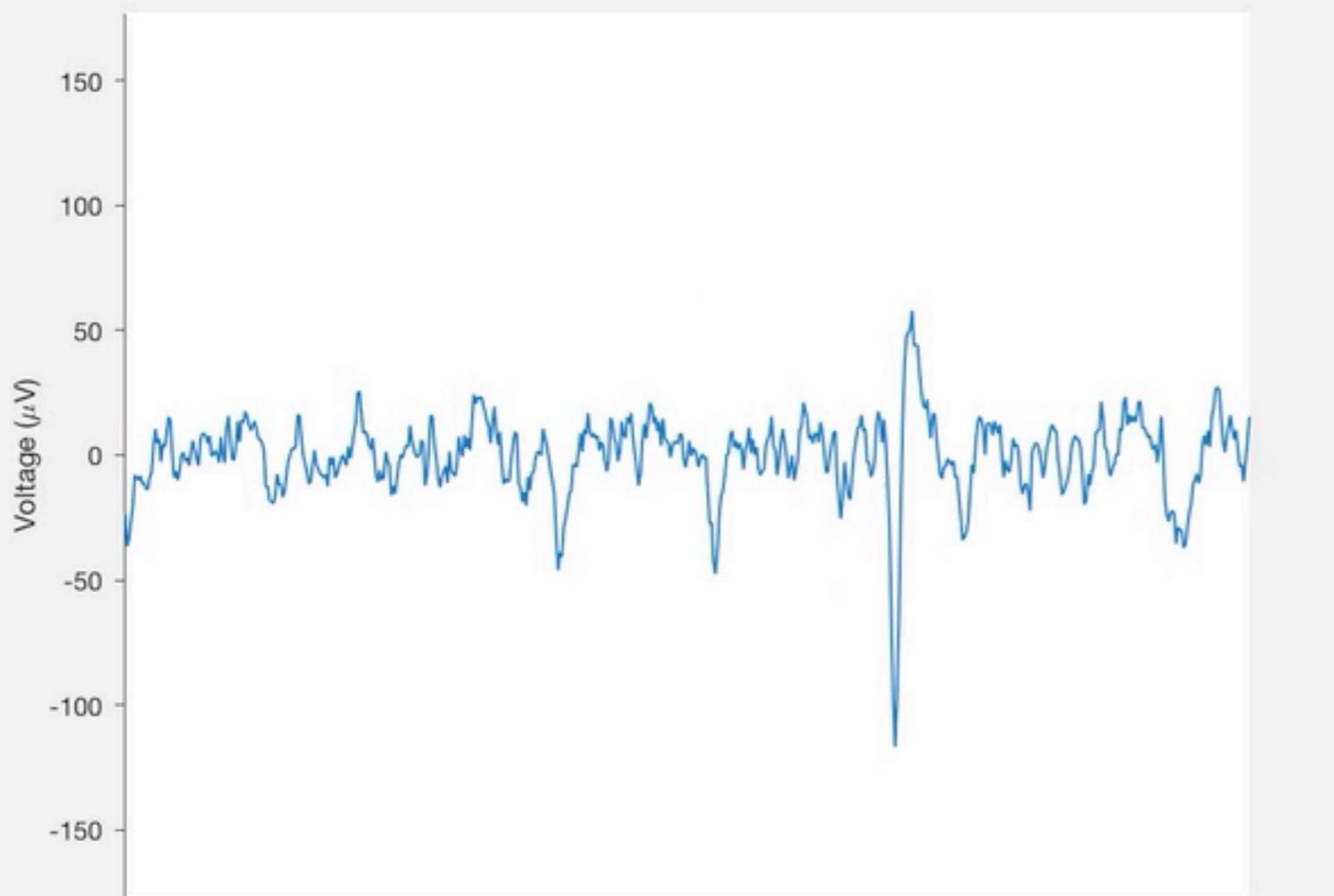


Carnegie
Mellon
University

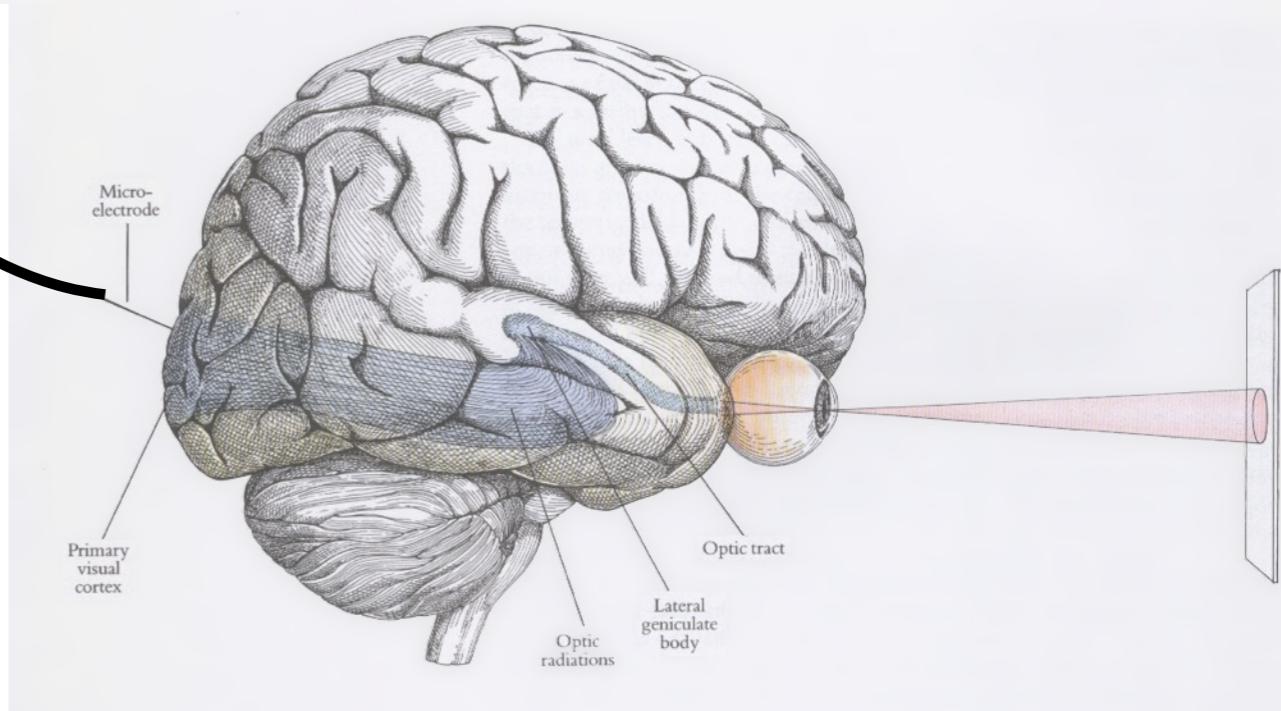
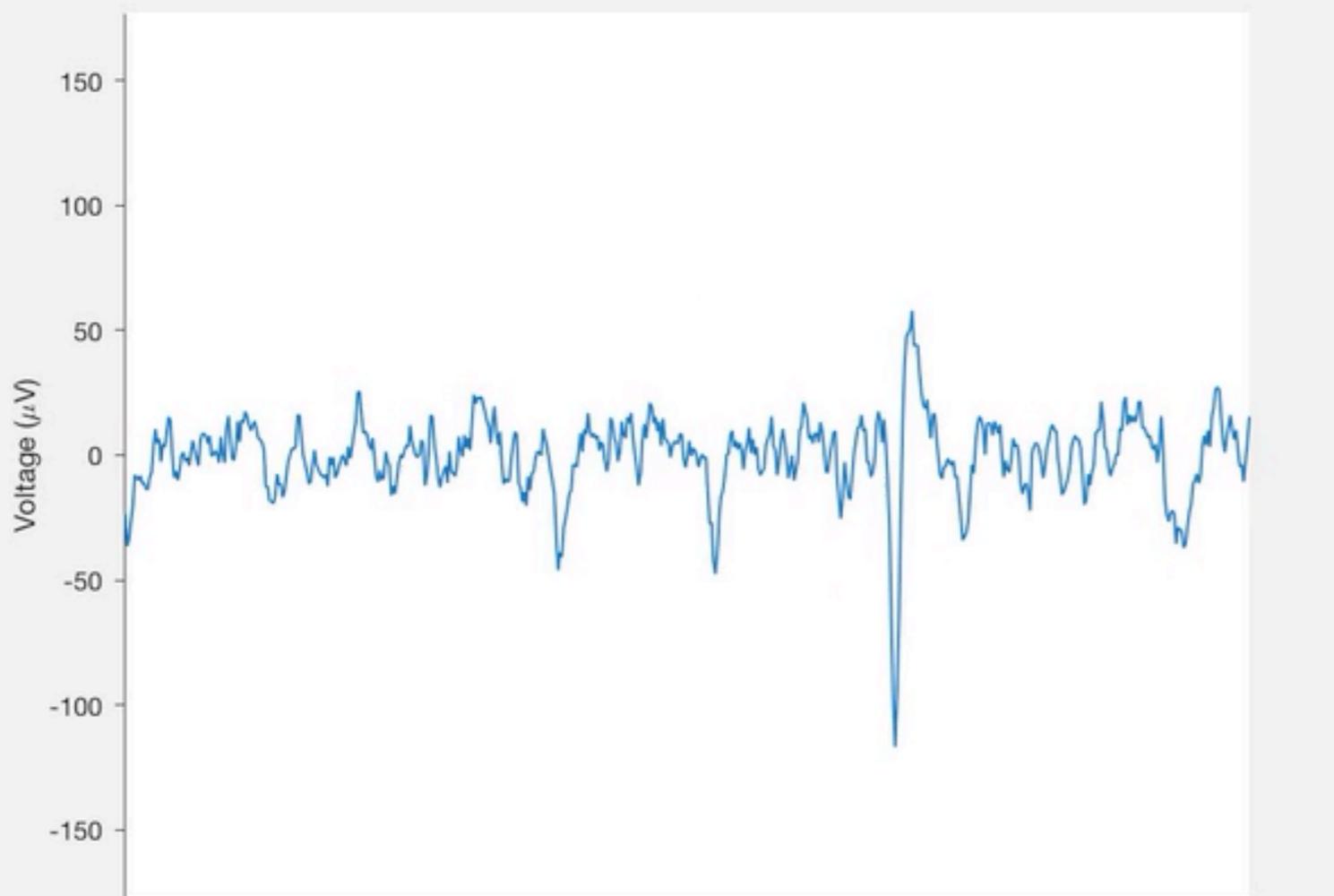


Signals from microelectrodes



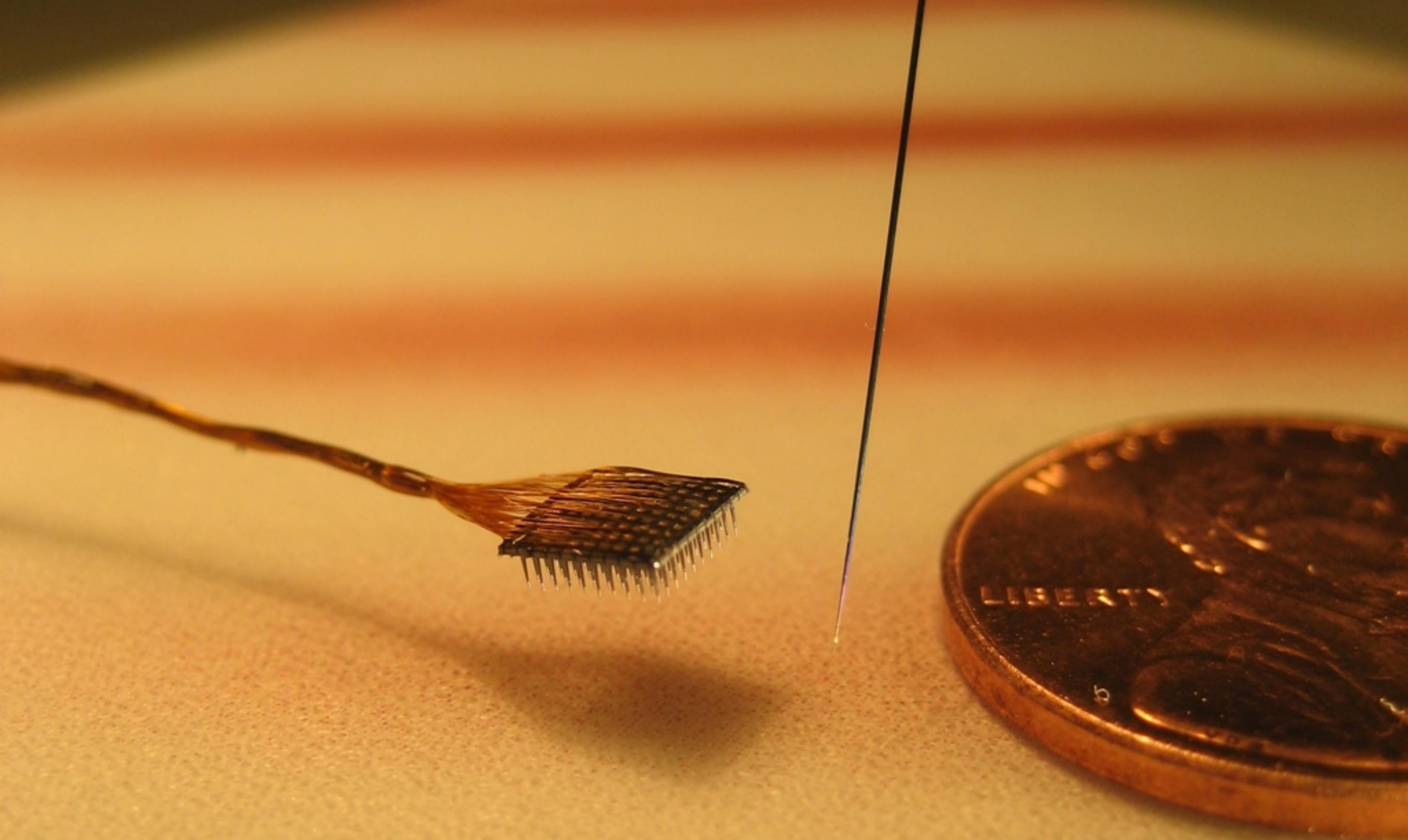


Hubel, 1988

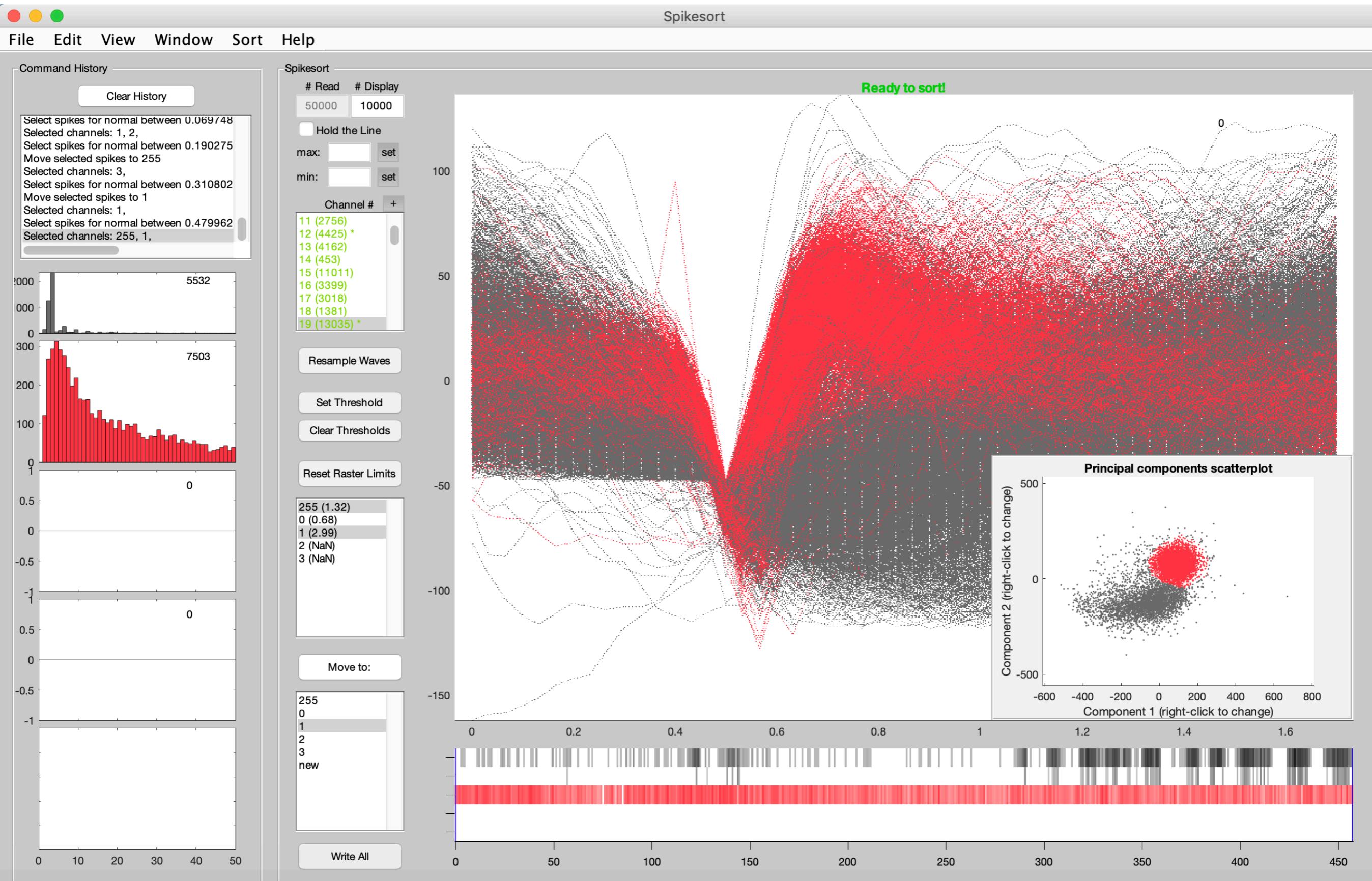


Hubel, 1988

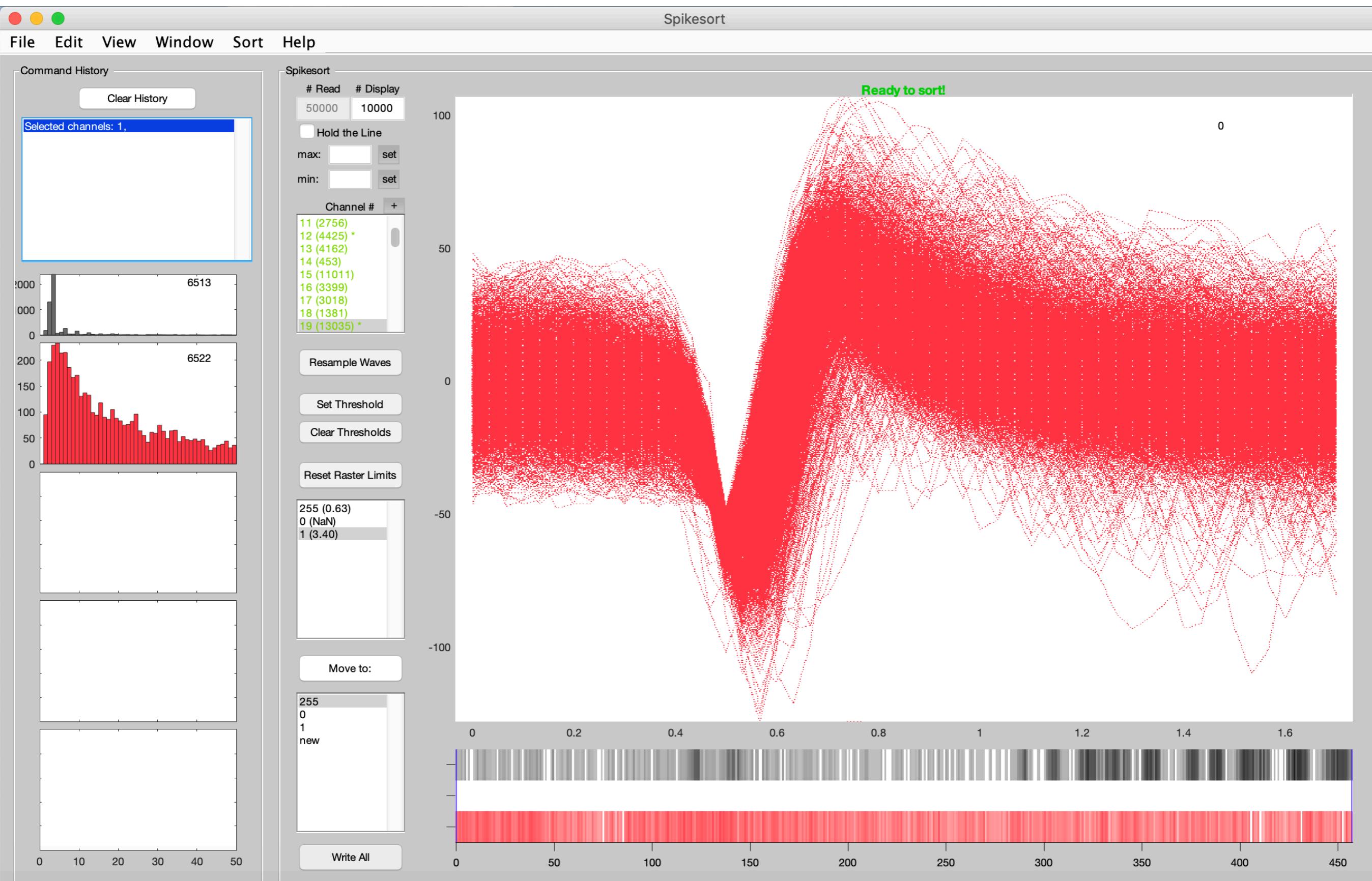




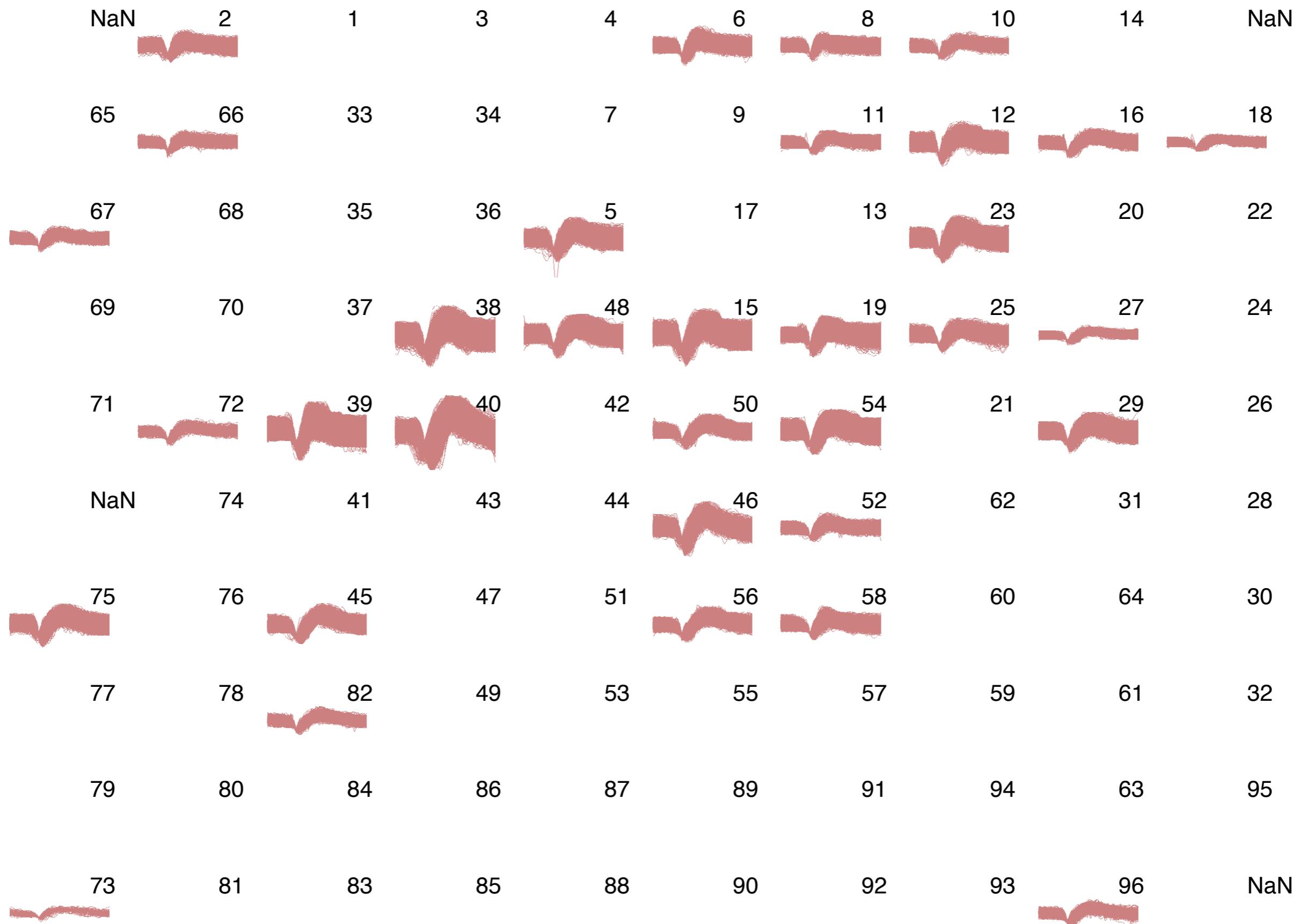
Kelly, Smith, Samonds, Kohn, Bonds & Movshon, 2007

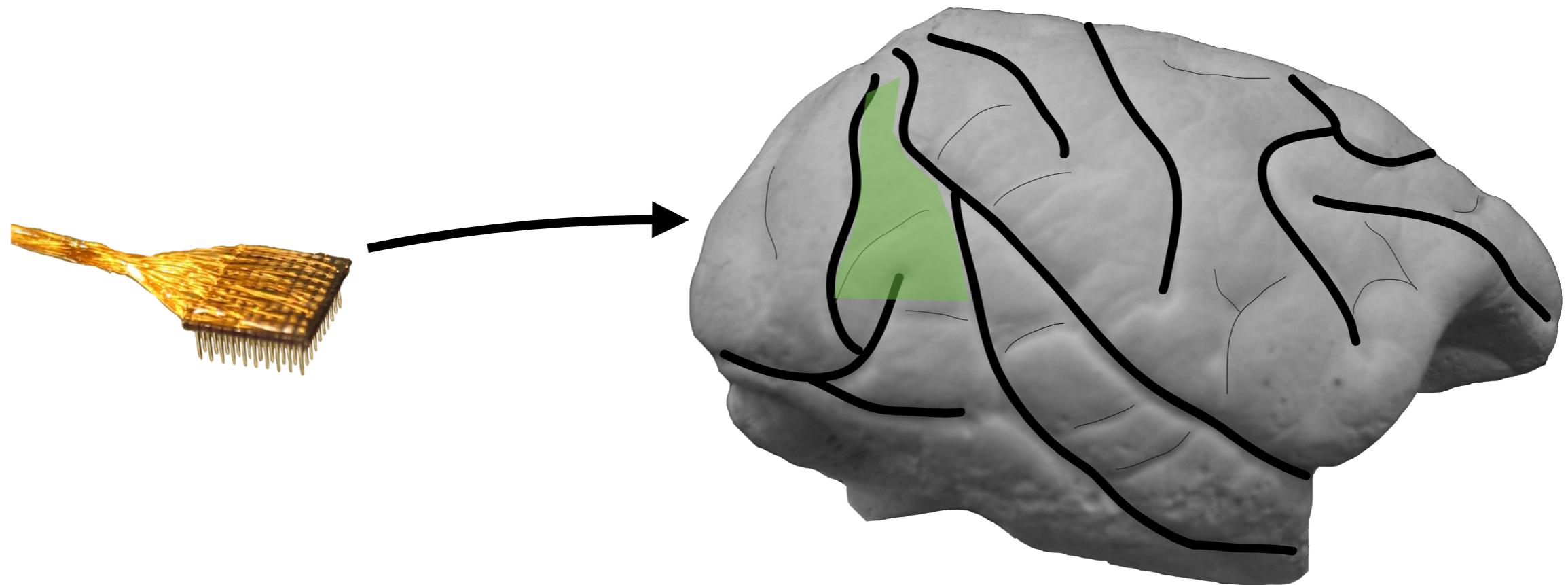
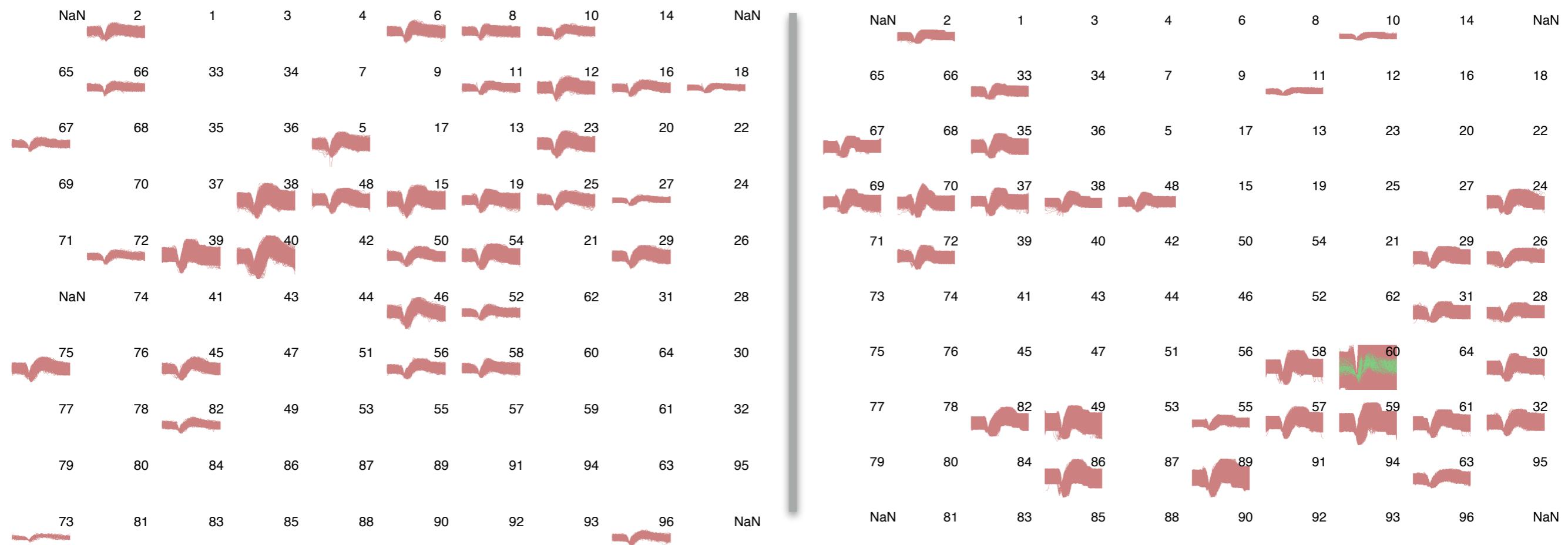


<https://github.com/smithlabvision/spikesort>



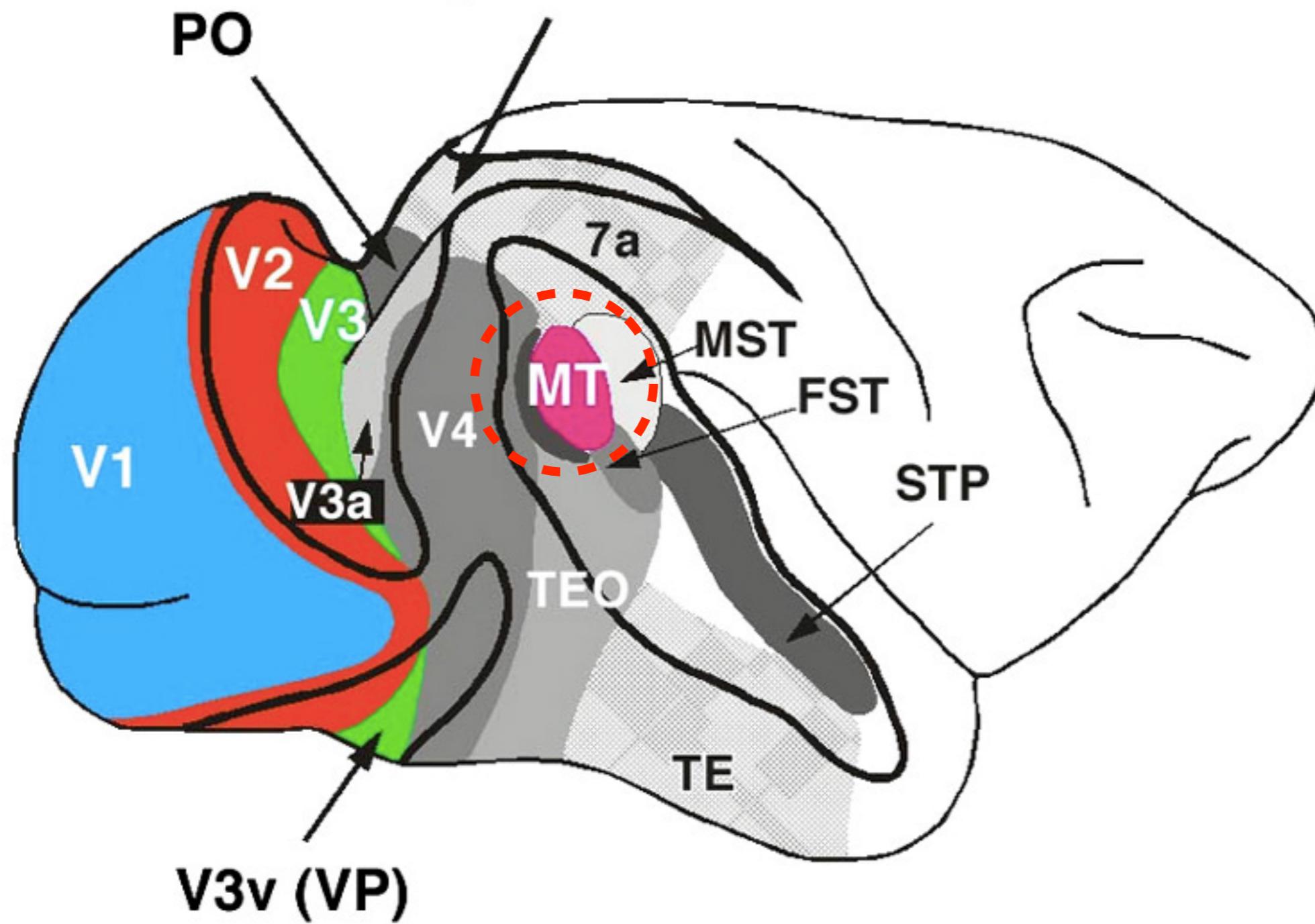
<https://github.com/smithlabvision/spikesort>



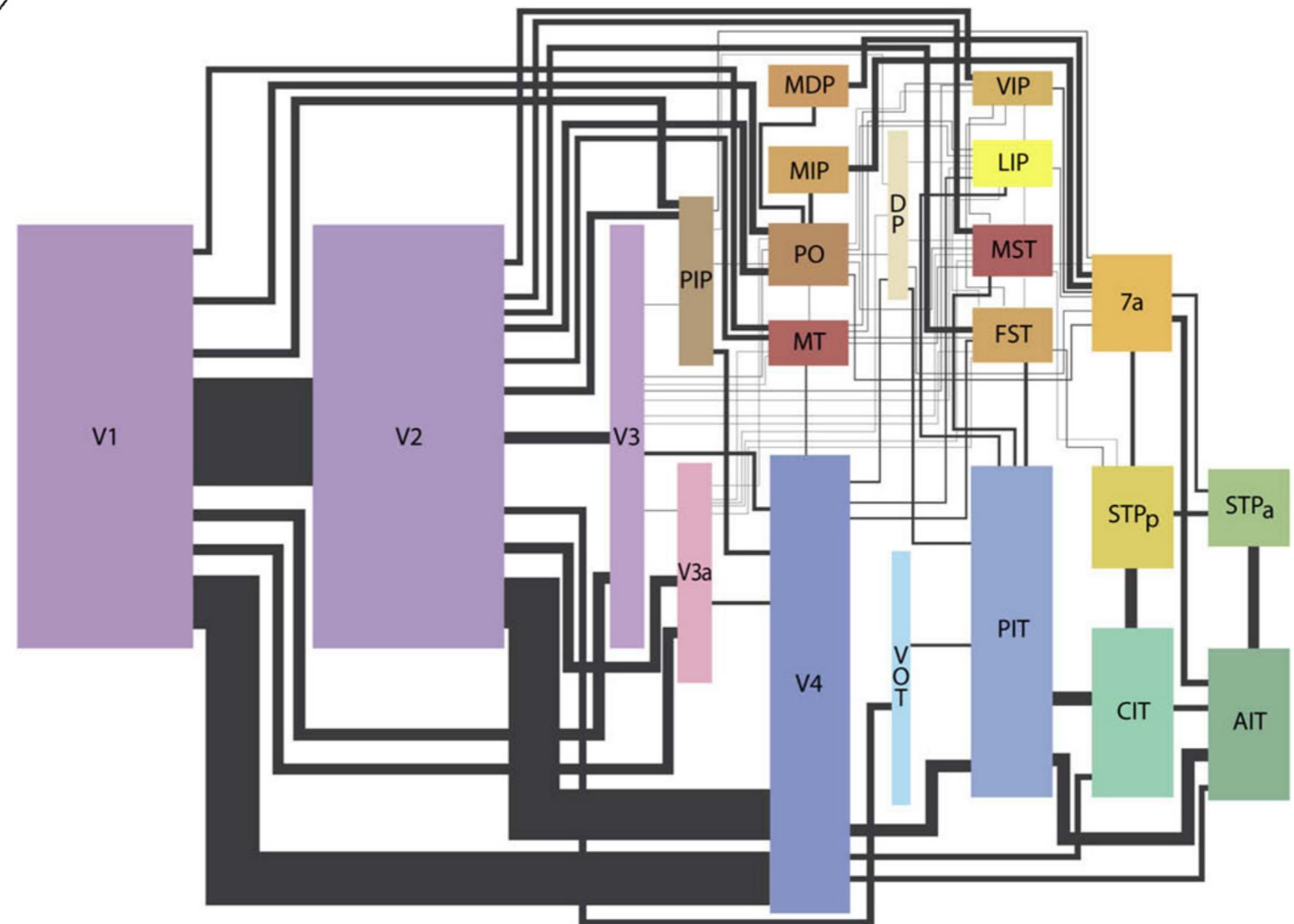
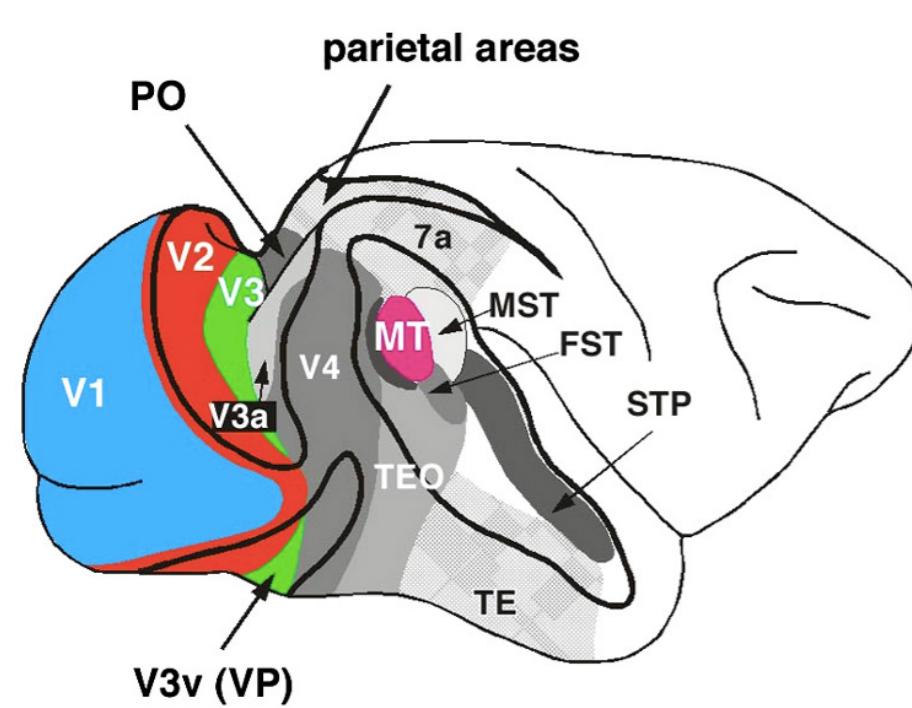


Macaque Cortex

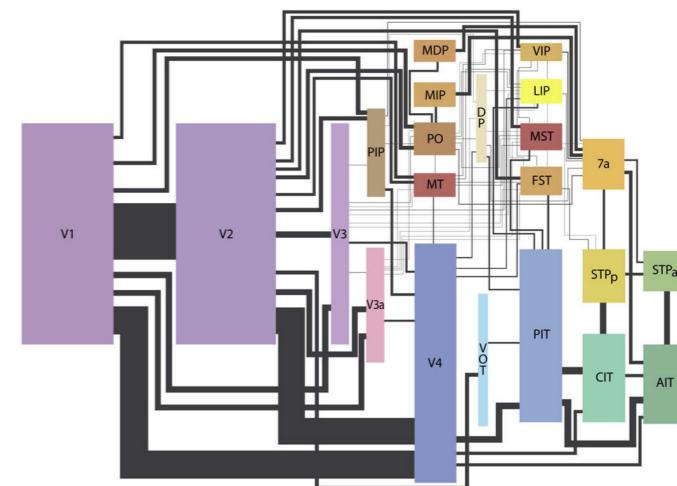
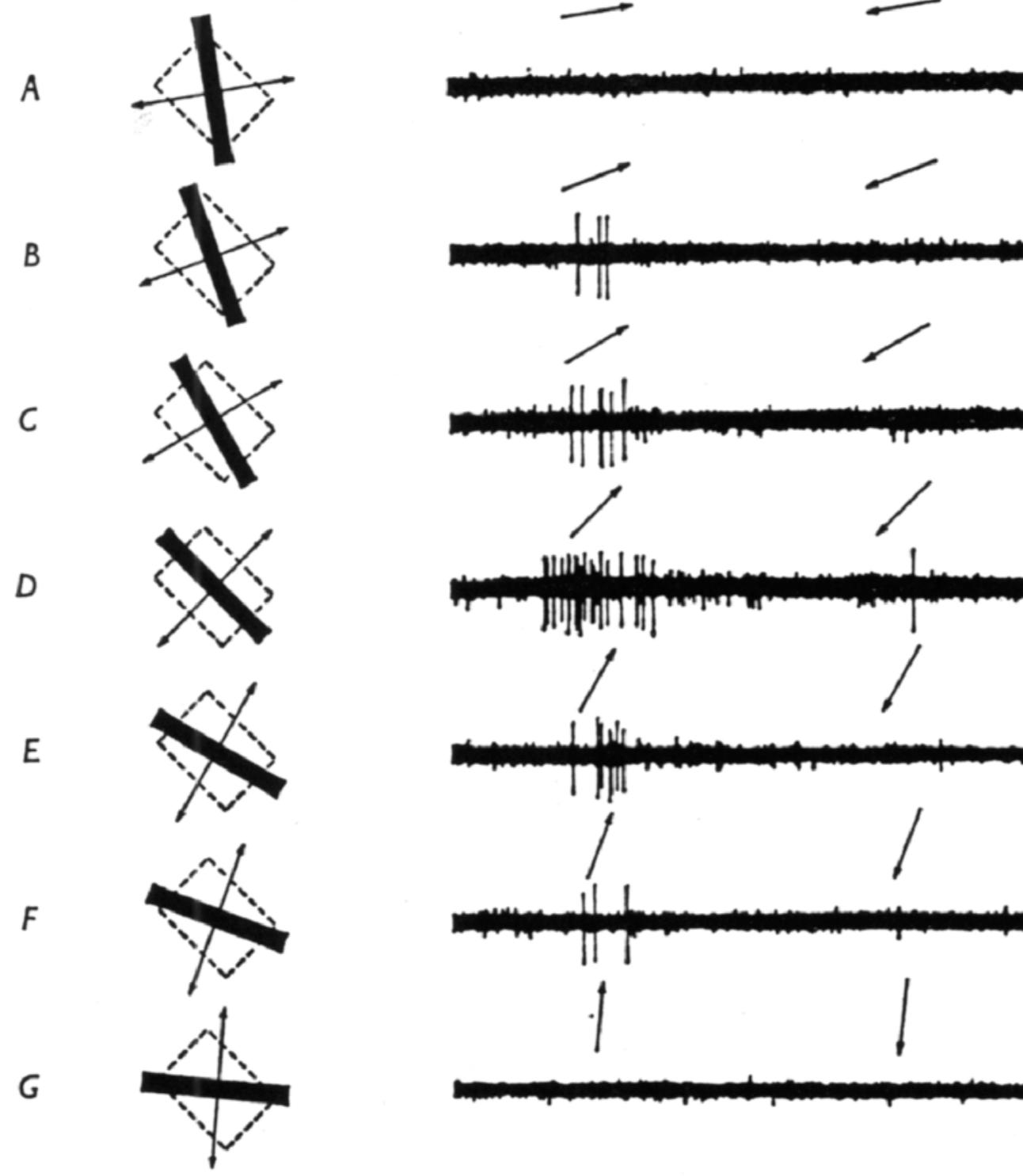
parietal areas



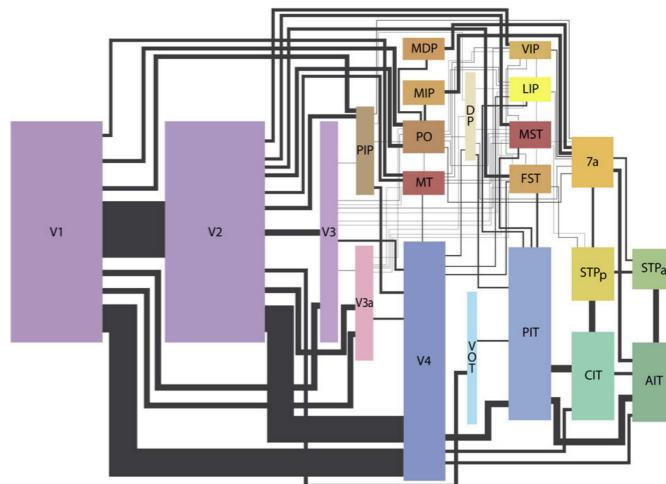
Macaque Cortex



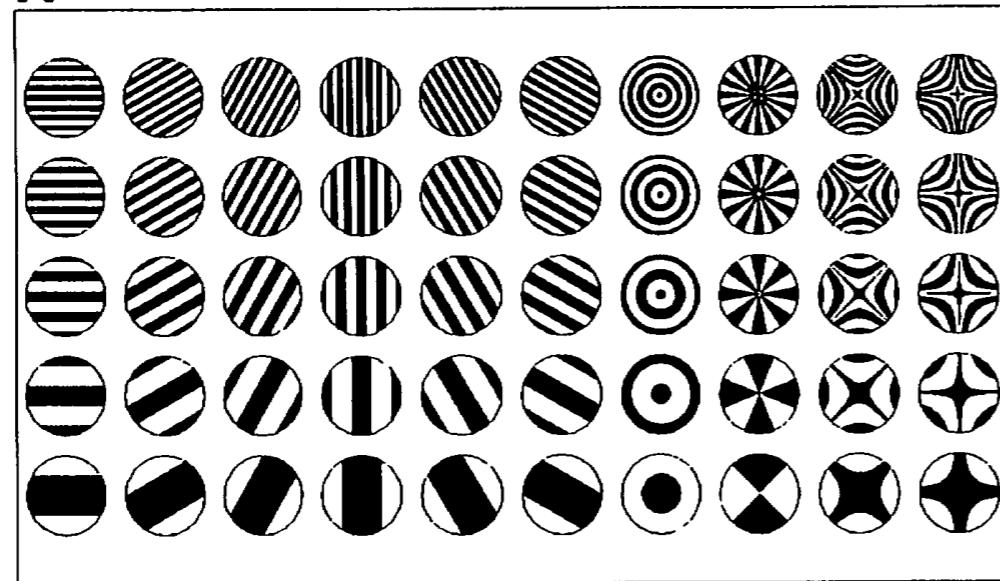
Visual Cortex (V1)



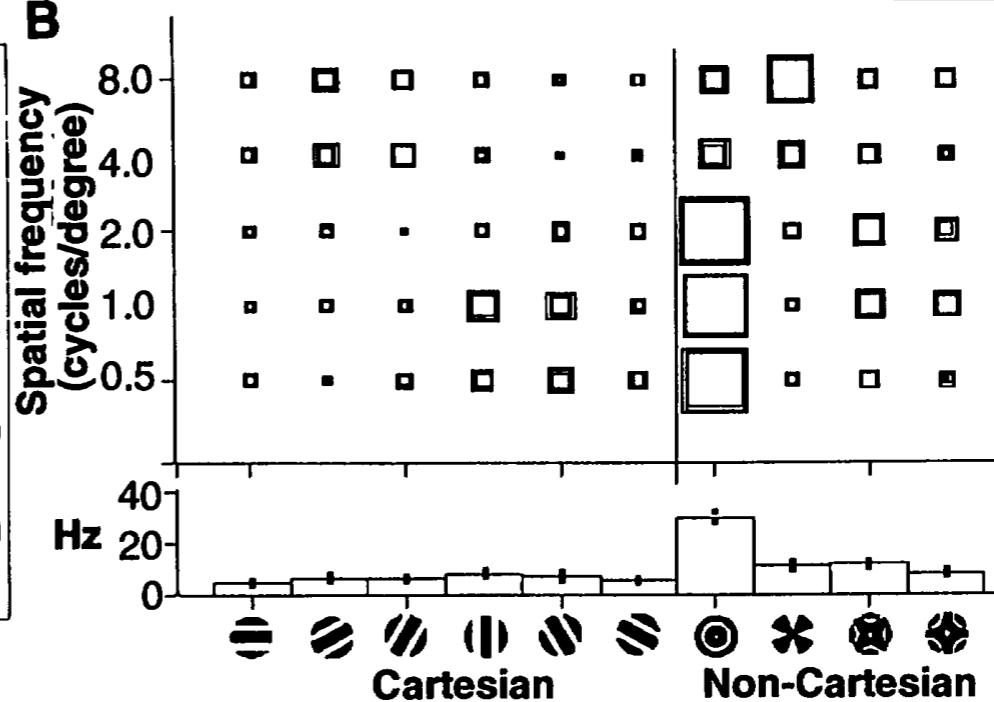
Visual Cortex (V4)



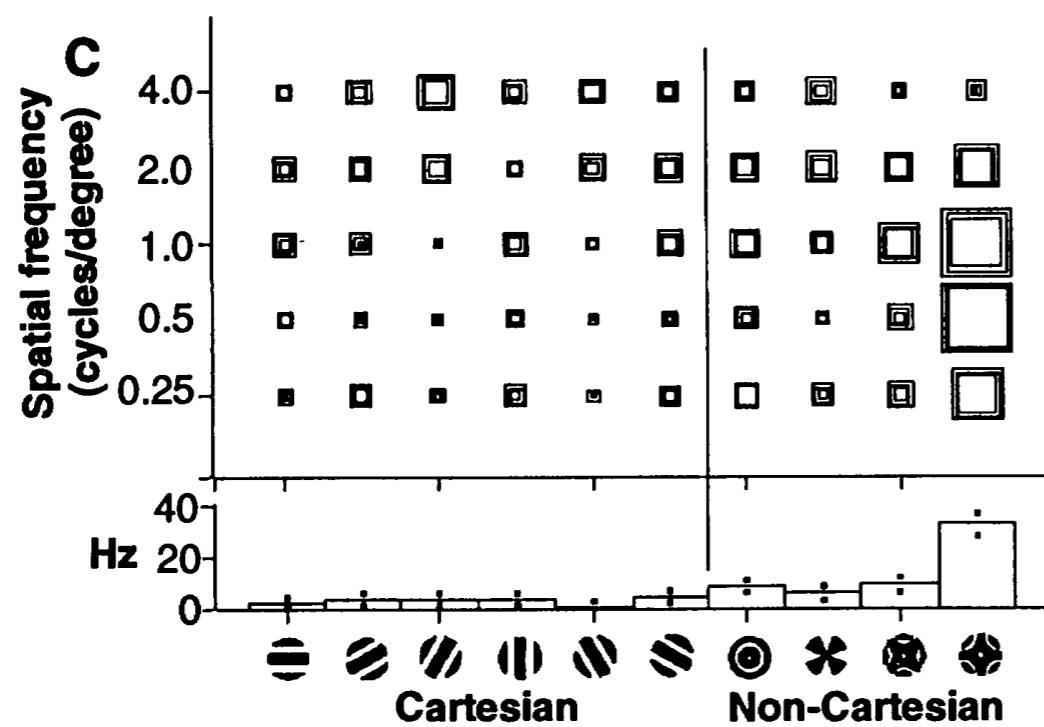
A



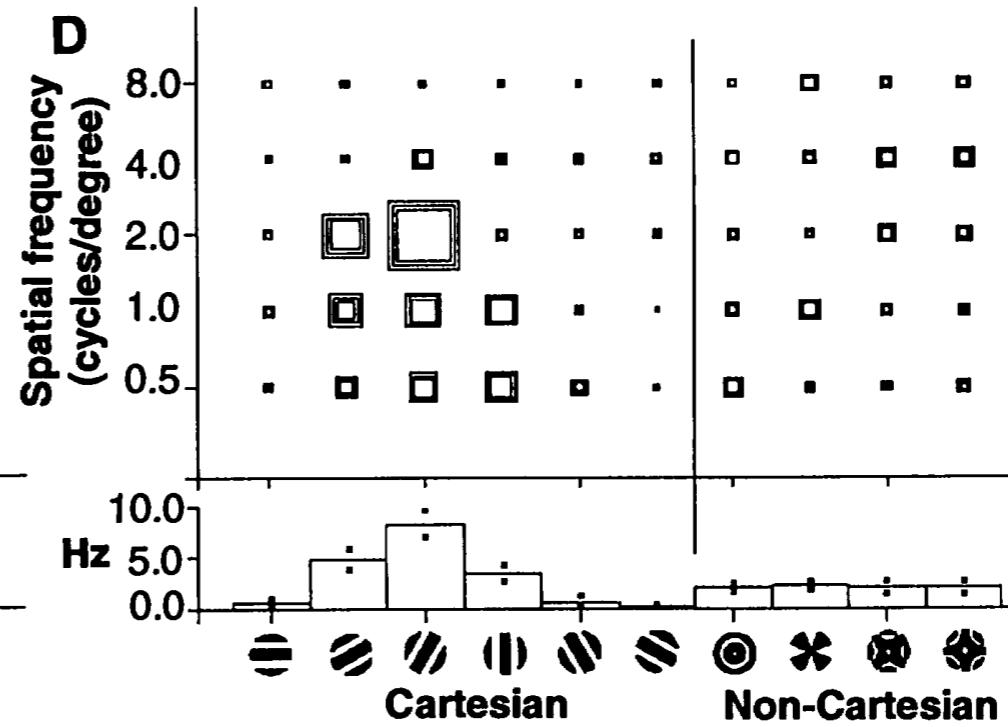
B



C



D



Experimental Task



look at fixation dot

Experimental Task



Experimental Task

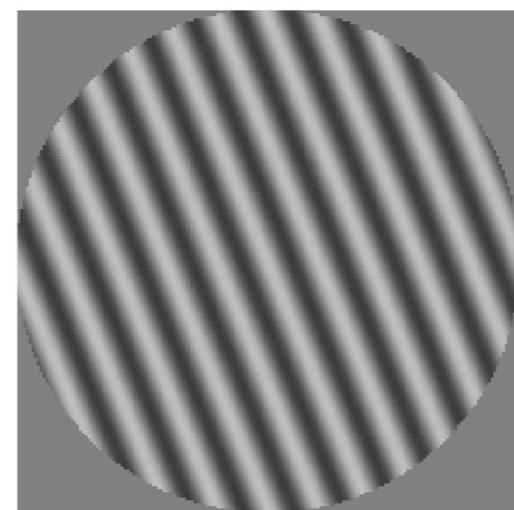


maintain fixation (0.5 or 1.0 s)

Experimental Task



Experimental Task



maintain fixation (1.0 s stimulus duration)

Experimental Task

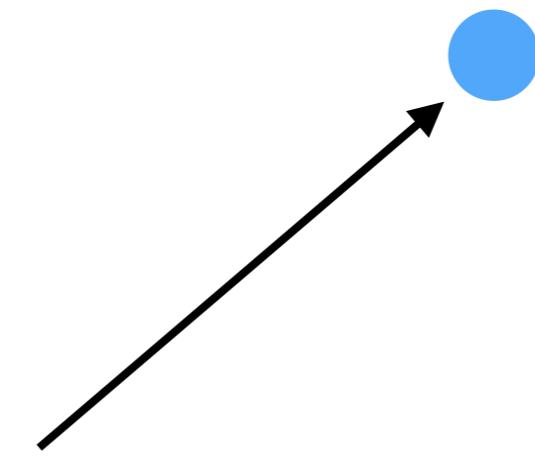


Experimental Task



fixation dot jumps to random location

Experimental Task



saccade to dot to receive reward

Experimental Task

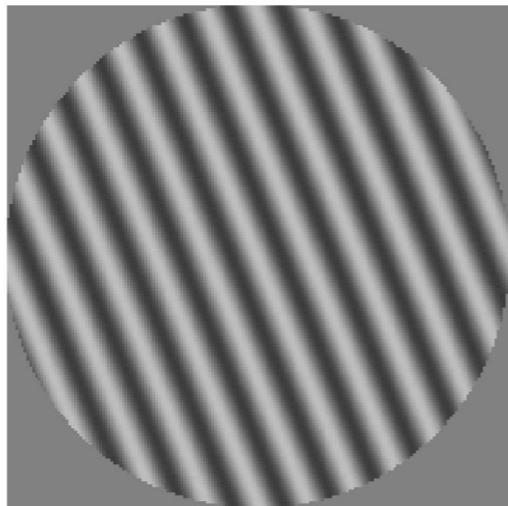


saccade to dot to receive reward

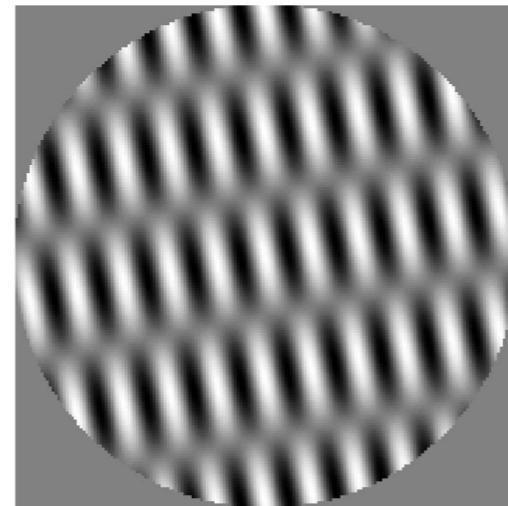
Visual stimuli

Grating

8 orientations
spaced by 22.5°
(or a blank)



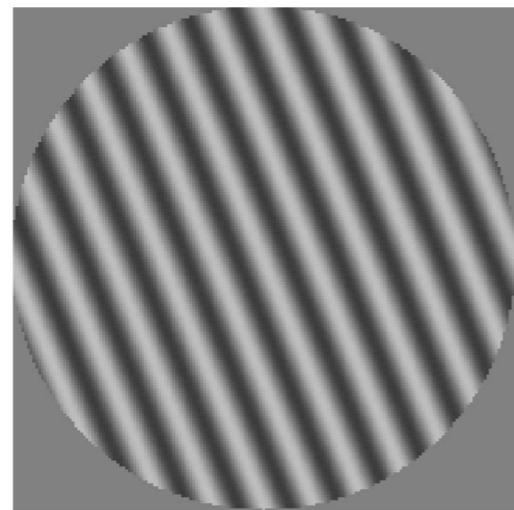
Plaid



81 possible
stimuli

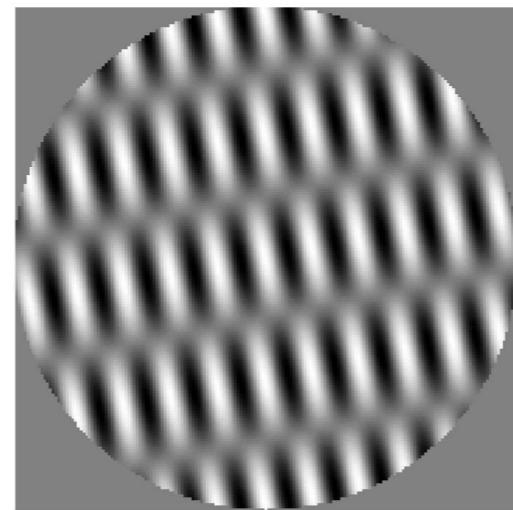
Visual stimuli

Grating



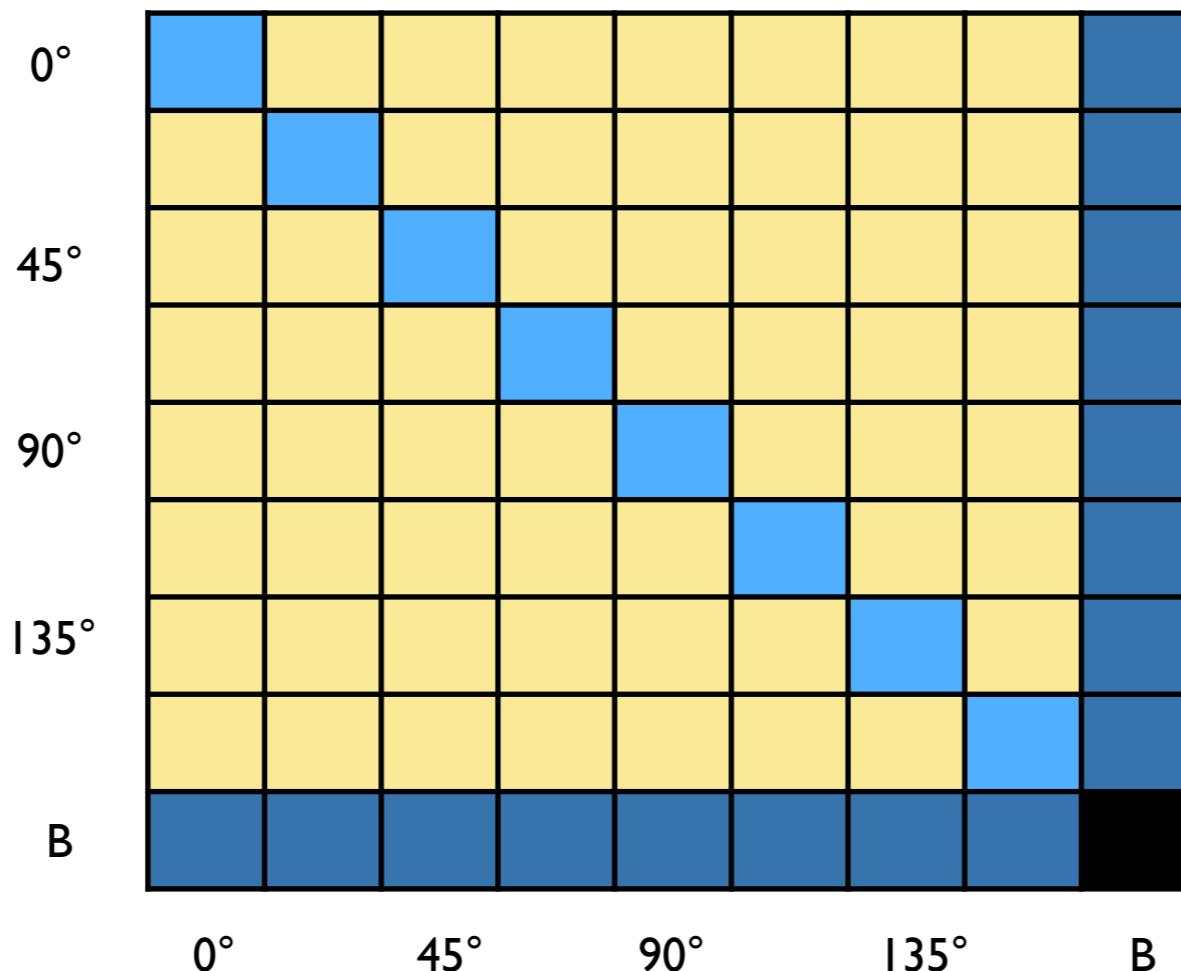
8 orientations
spaced by 22.5°
(or a blank)

Plaid



81 possible
stimuli

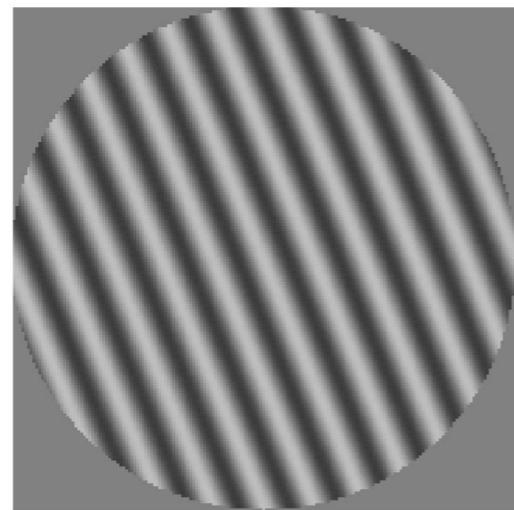
Grating 1



Grating 2

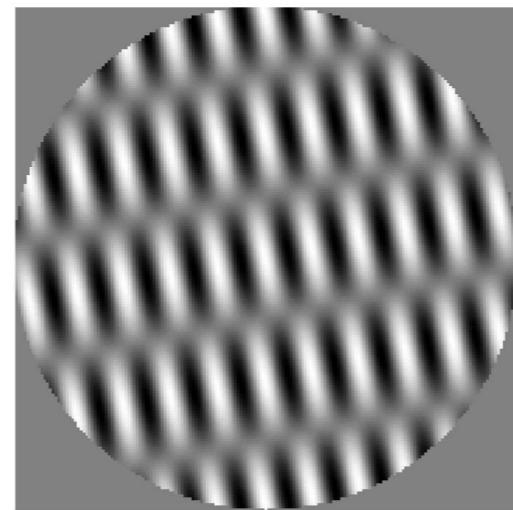
Visual stimuli

Grating



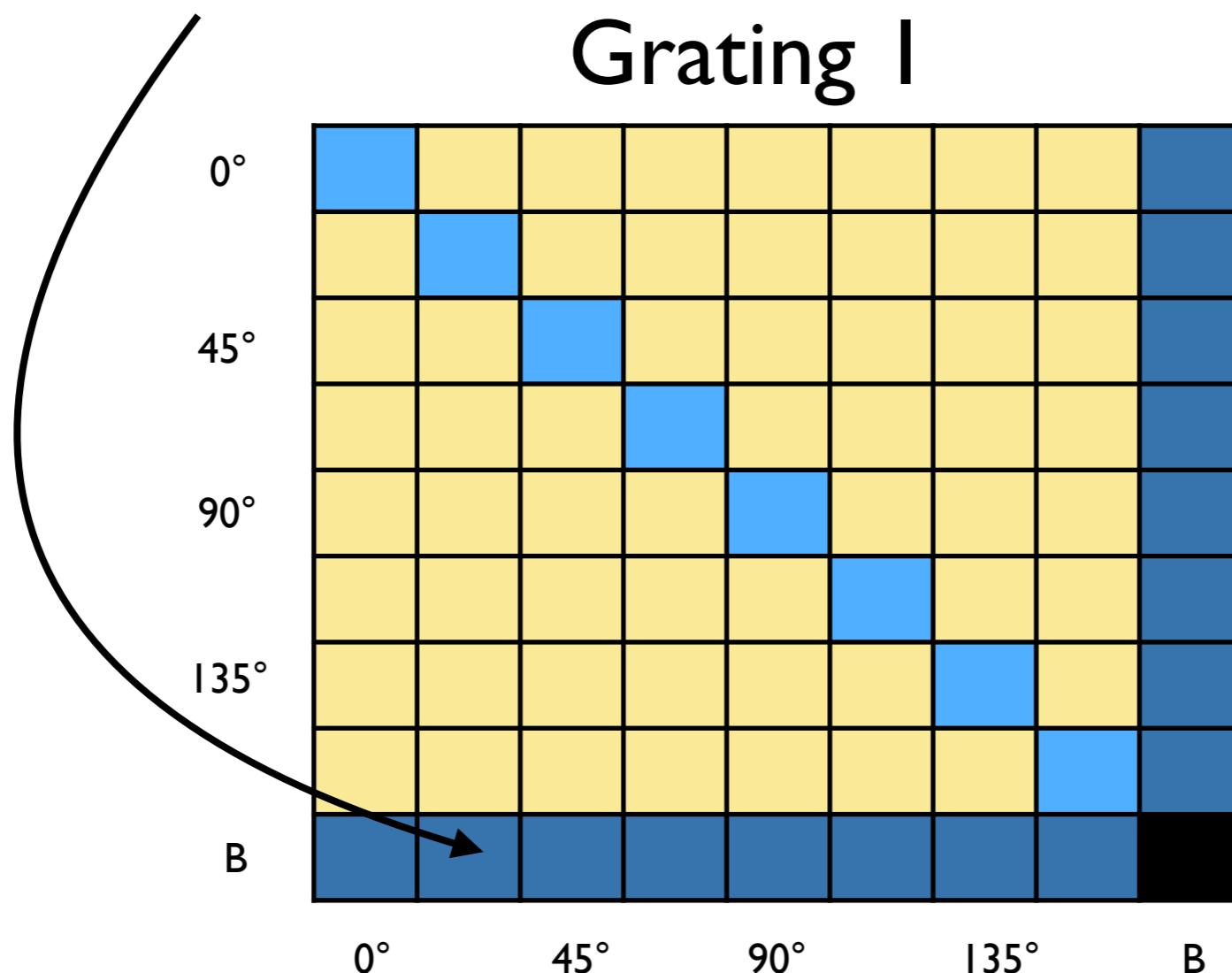
8 orientations
spaced by 22.5°
(or a blank)

Plaid



81 possible
stimuli

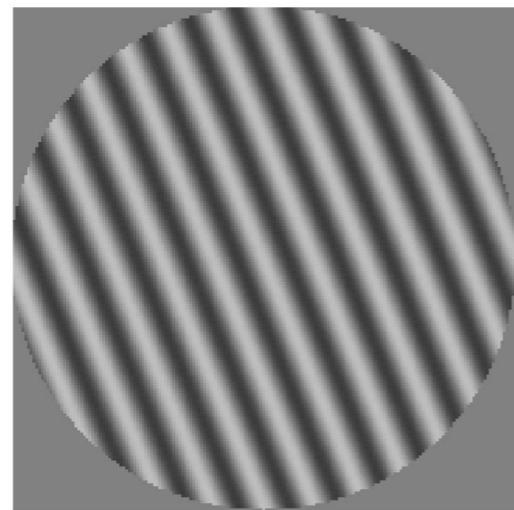
Grating 1



Grating 2

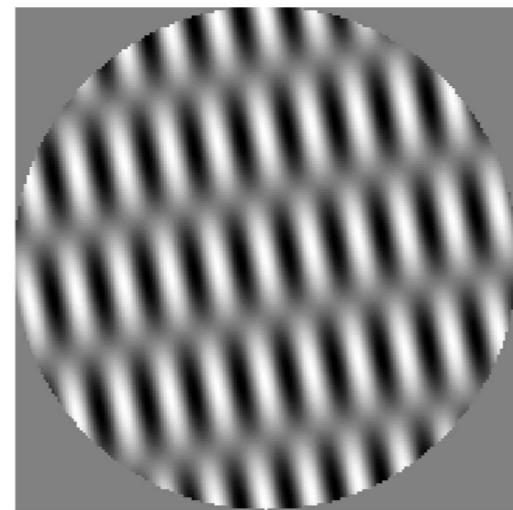
Visual stimuli

Grating



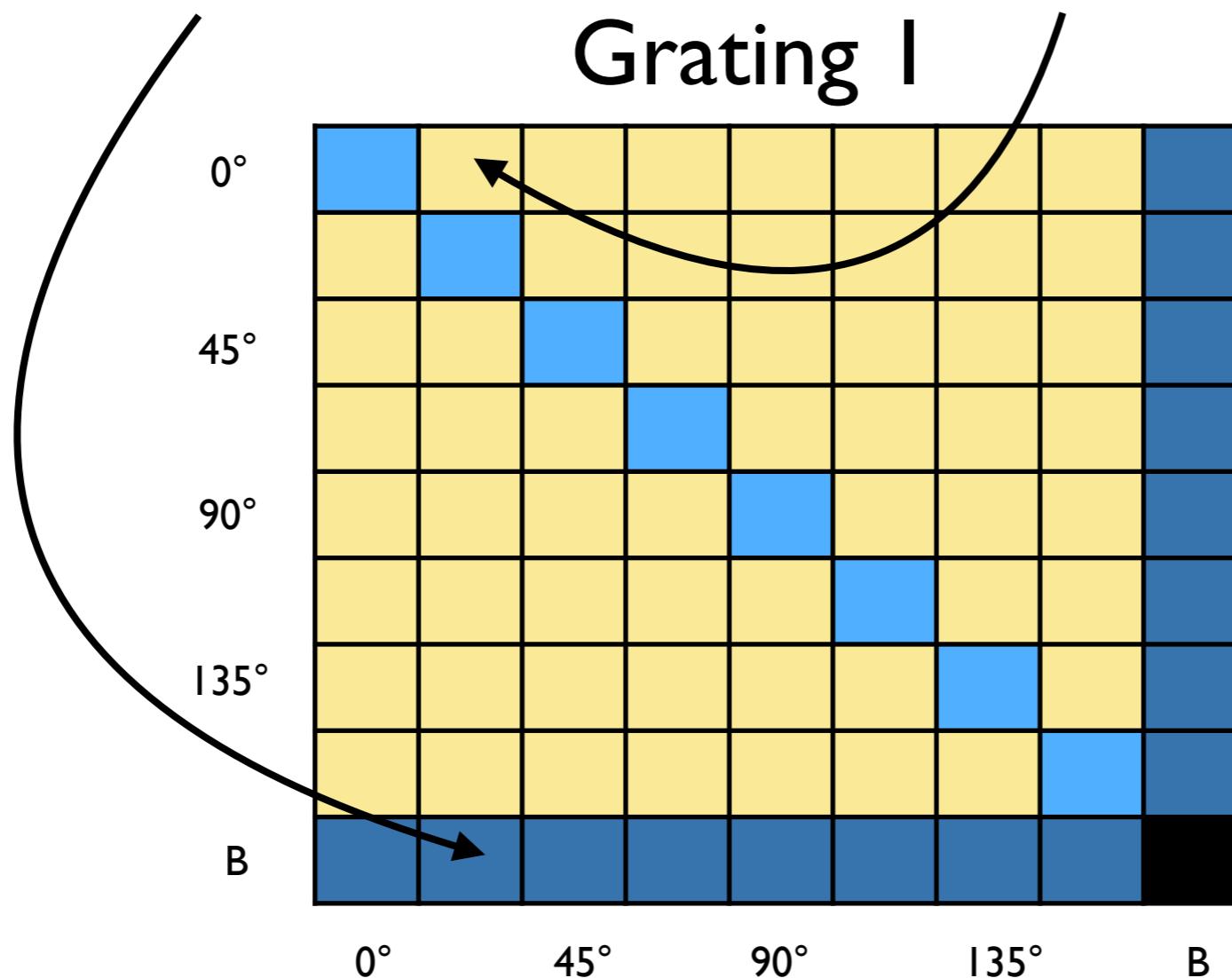
8 orientations
spaced by 22.5°
(or a blank)

Plaid



81 possible
stimuli

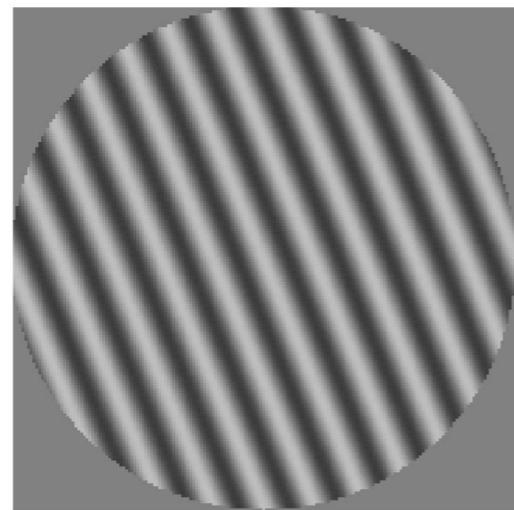
Grating 1



Grating 2

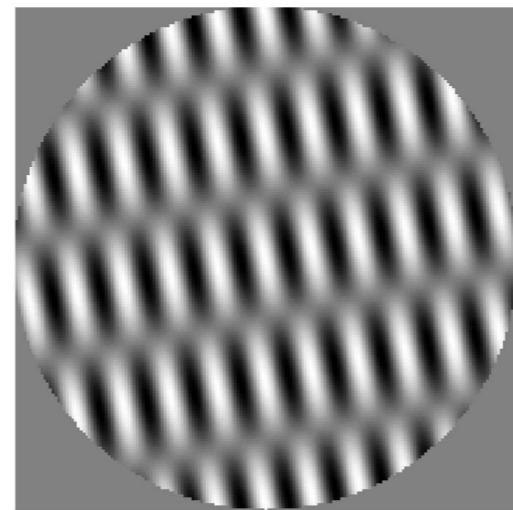
Visual stimuli

Grating



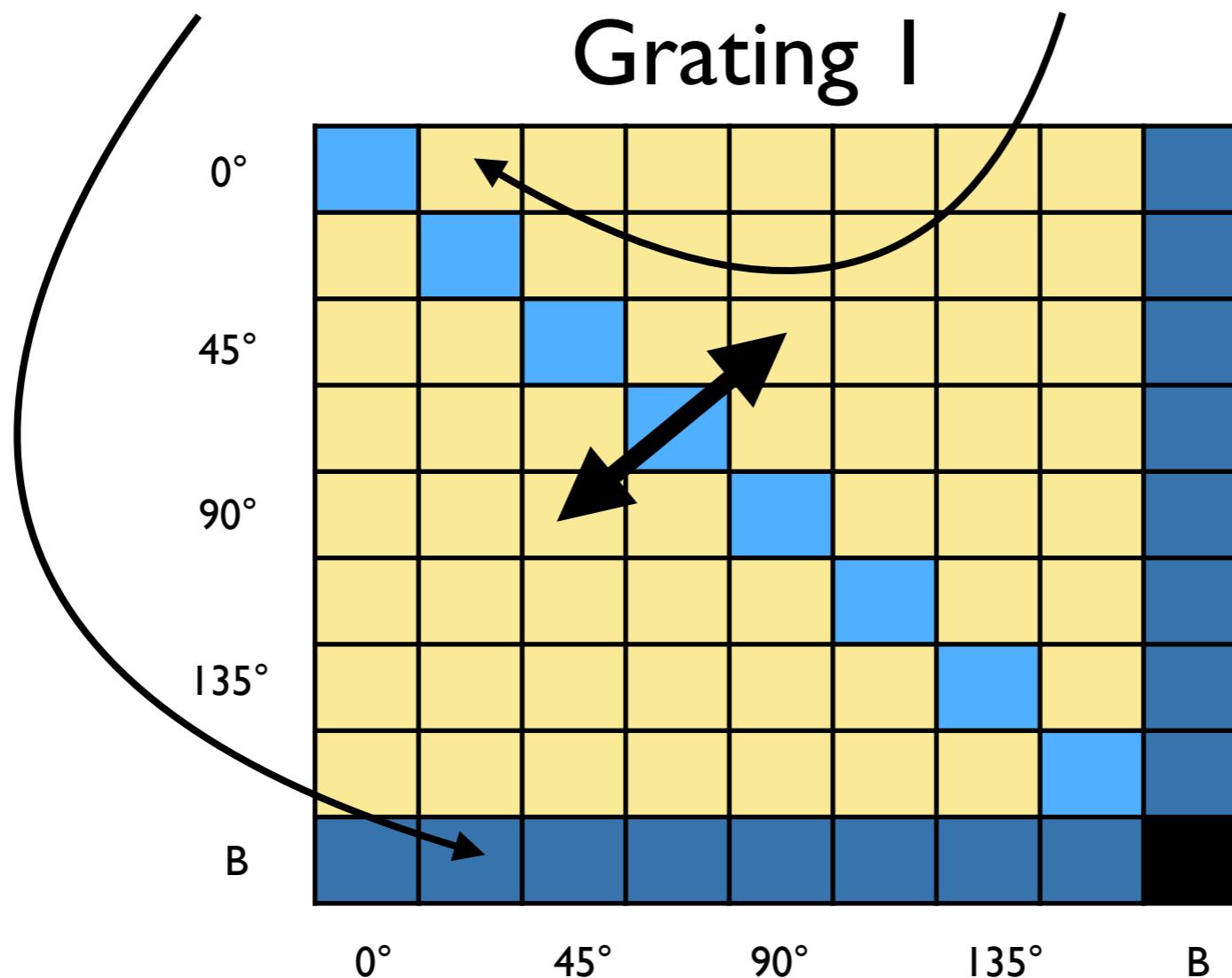
8 orientations
spaced by 22.5°
(or a blank)

Plaid



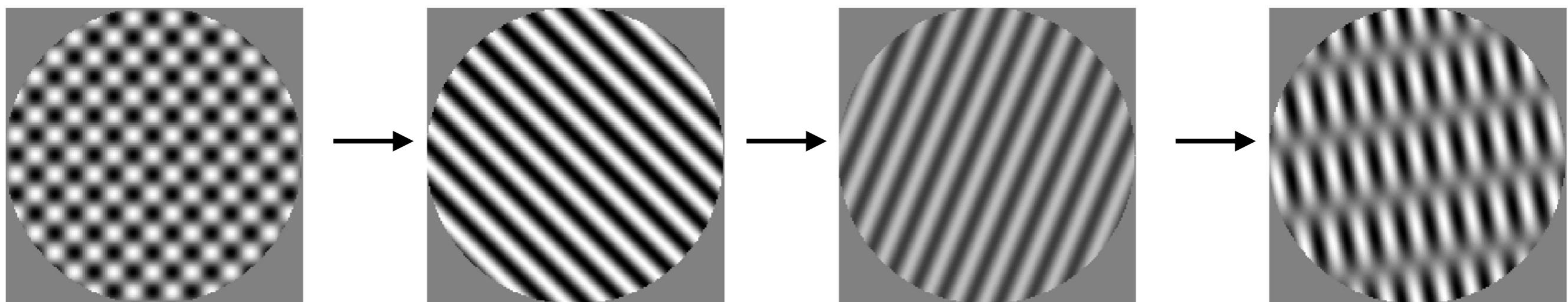
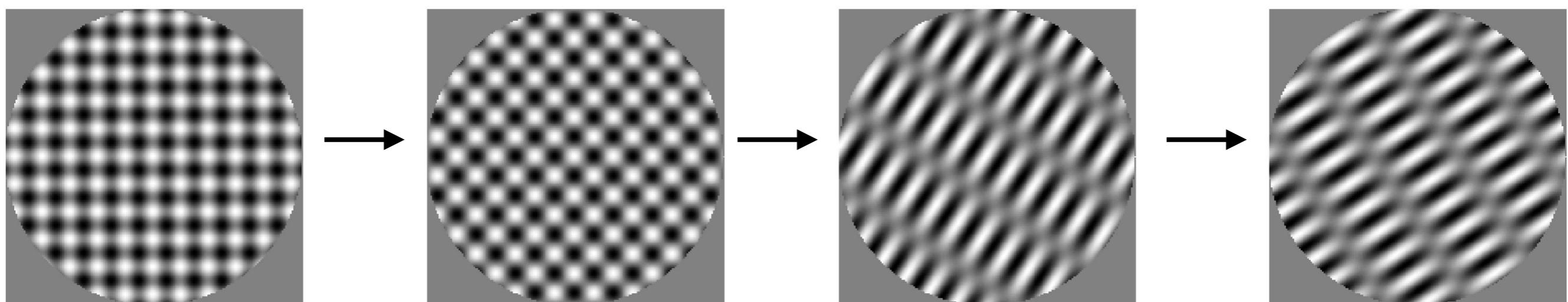
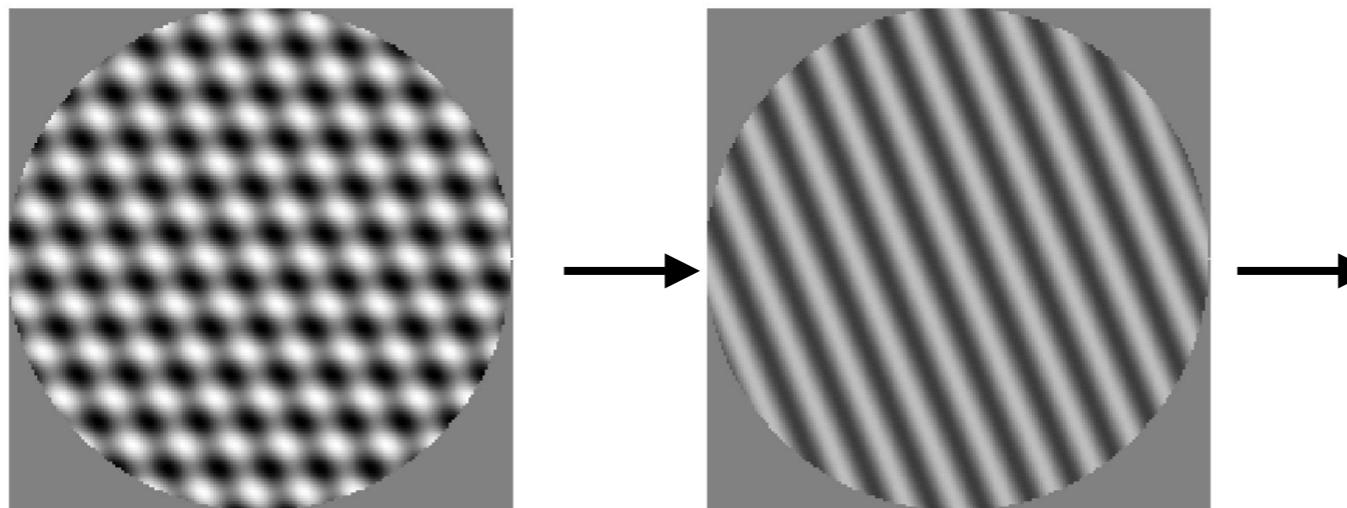
81 possible
stimuli

Grating 1

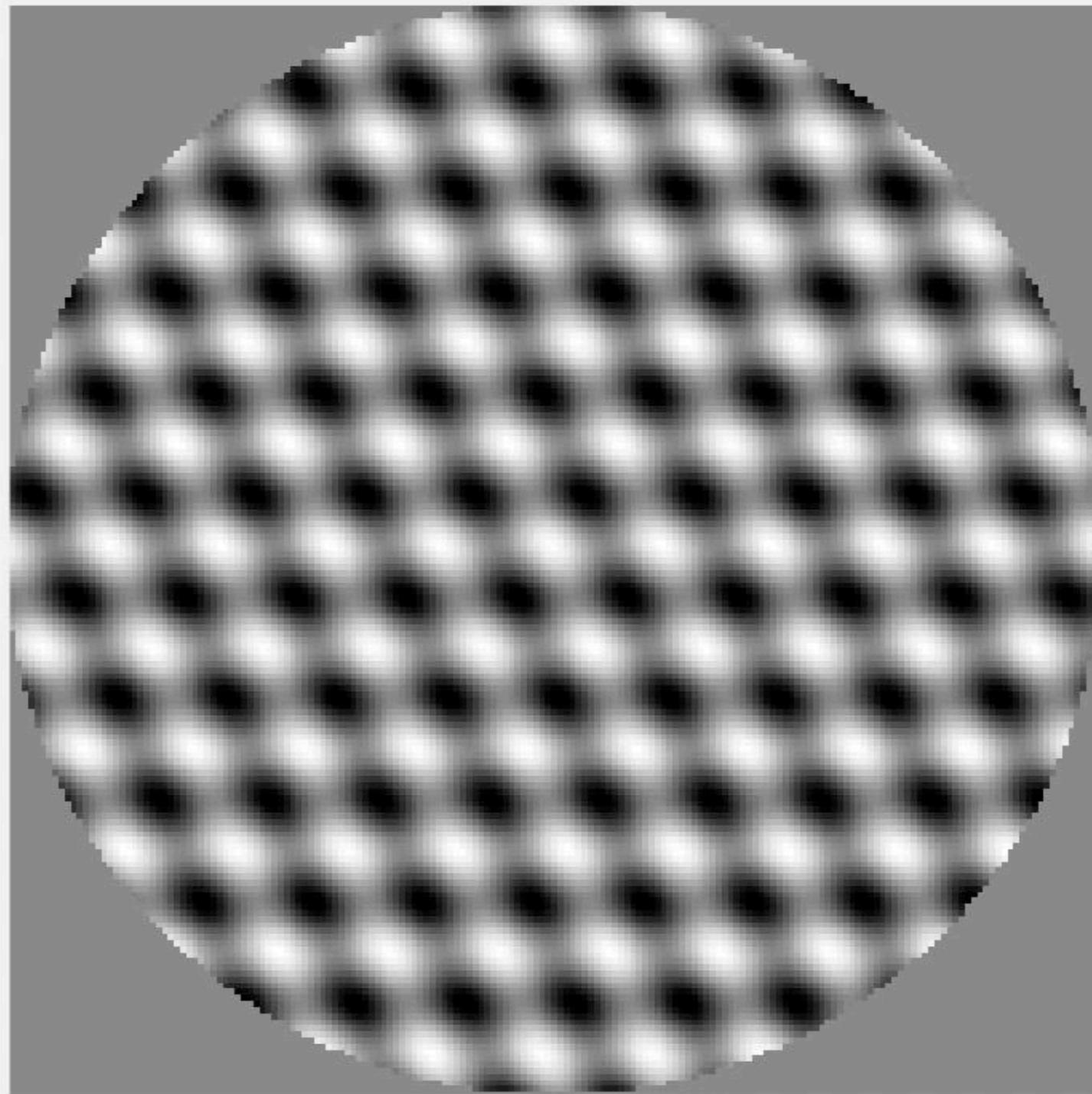


Grating 2

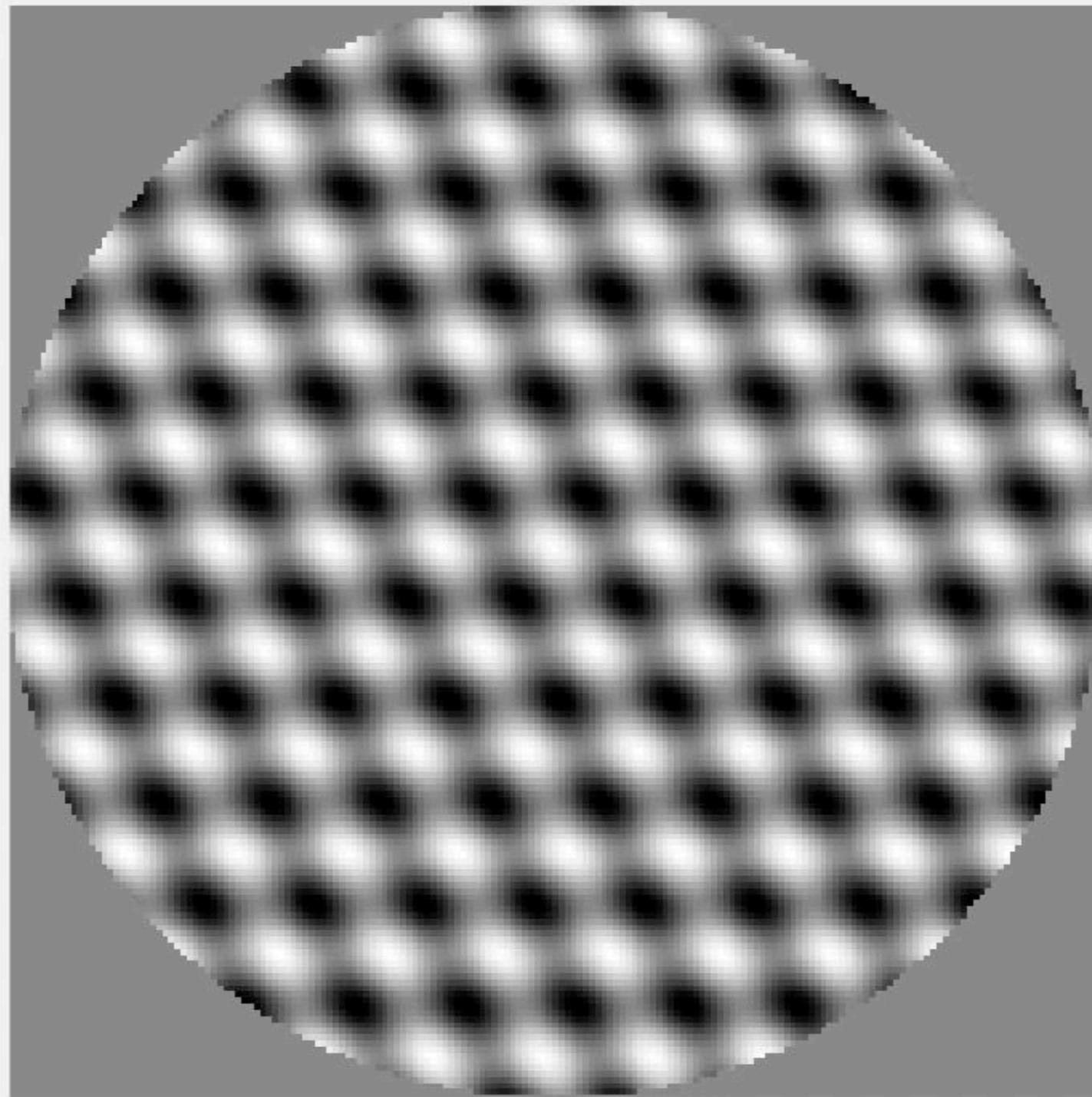
Visual stimuli (10 stimuli per fixation, 100 ms each)



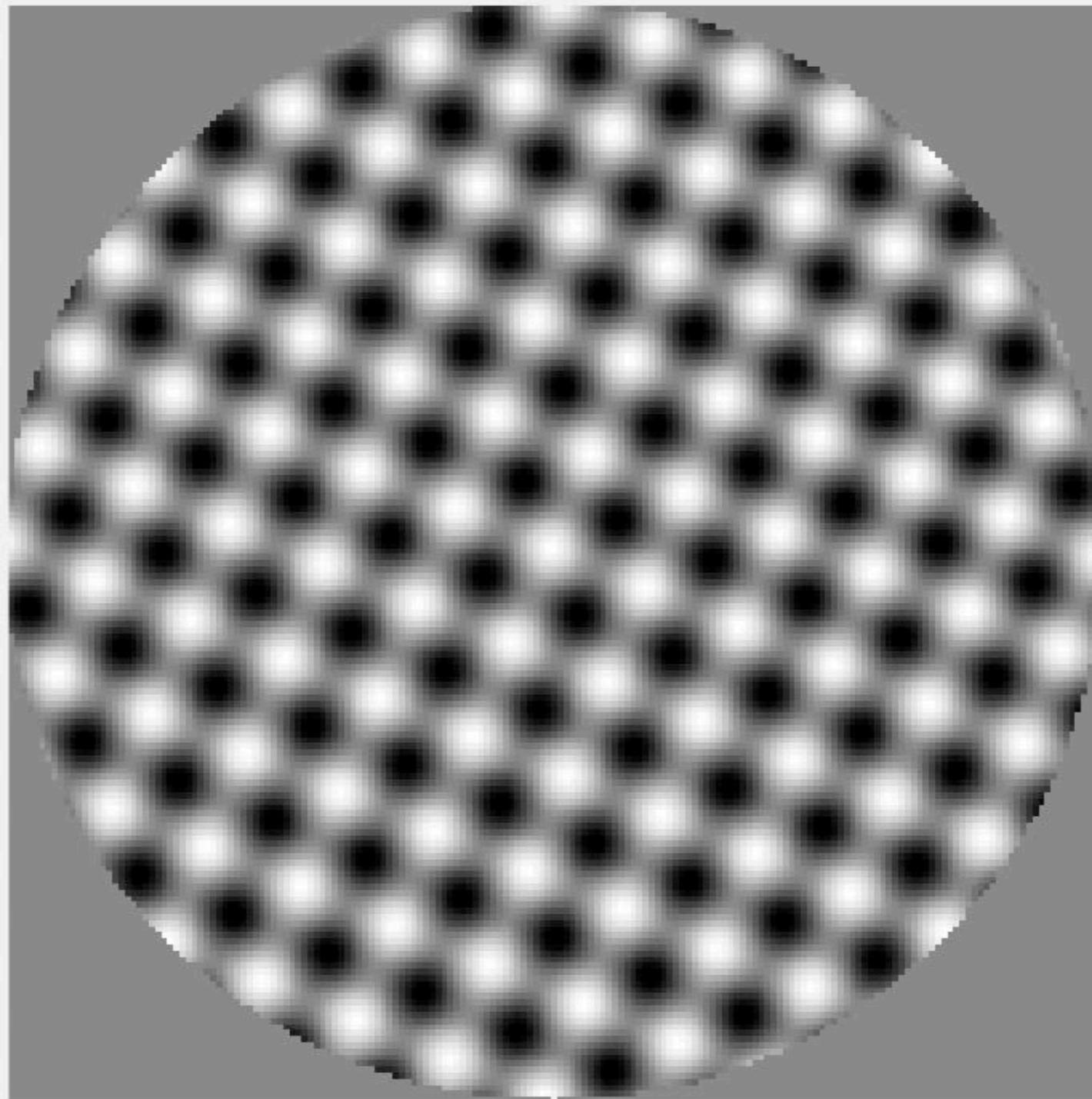
Visual Stimulus



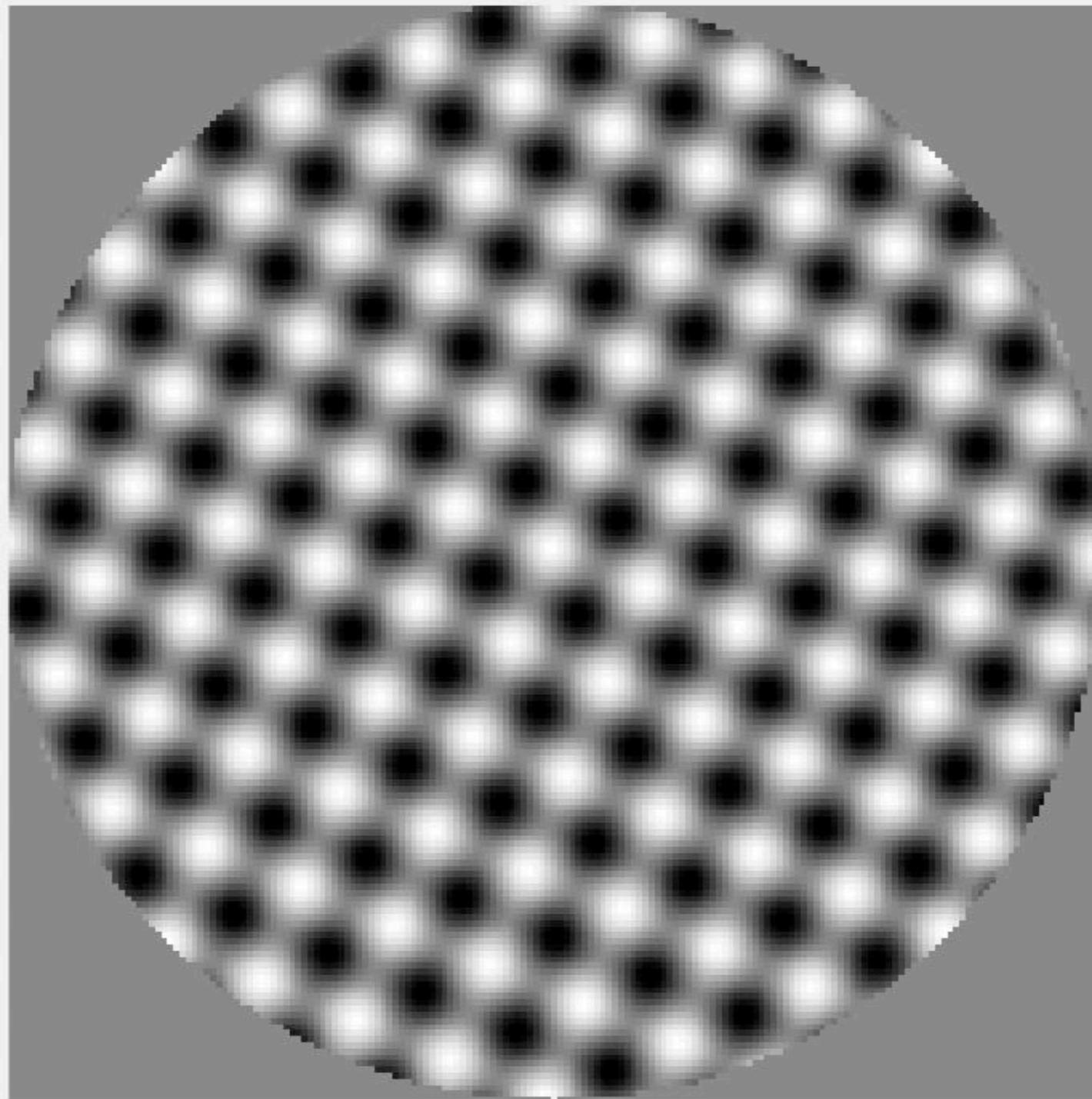
Visual Stimulus



Visual Stimulus



Visual Stimulus

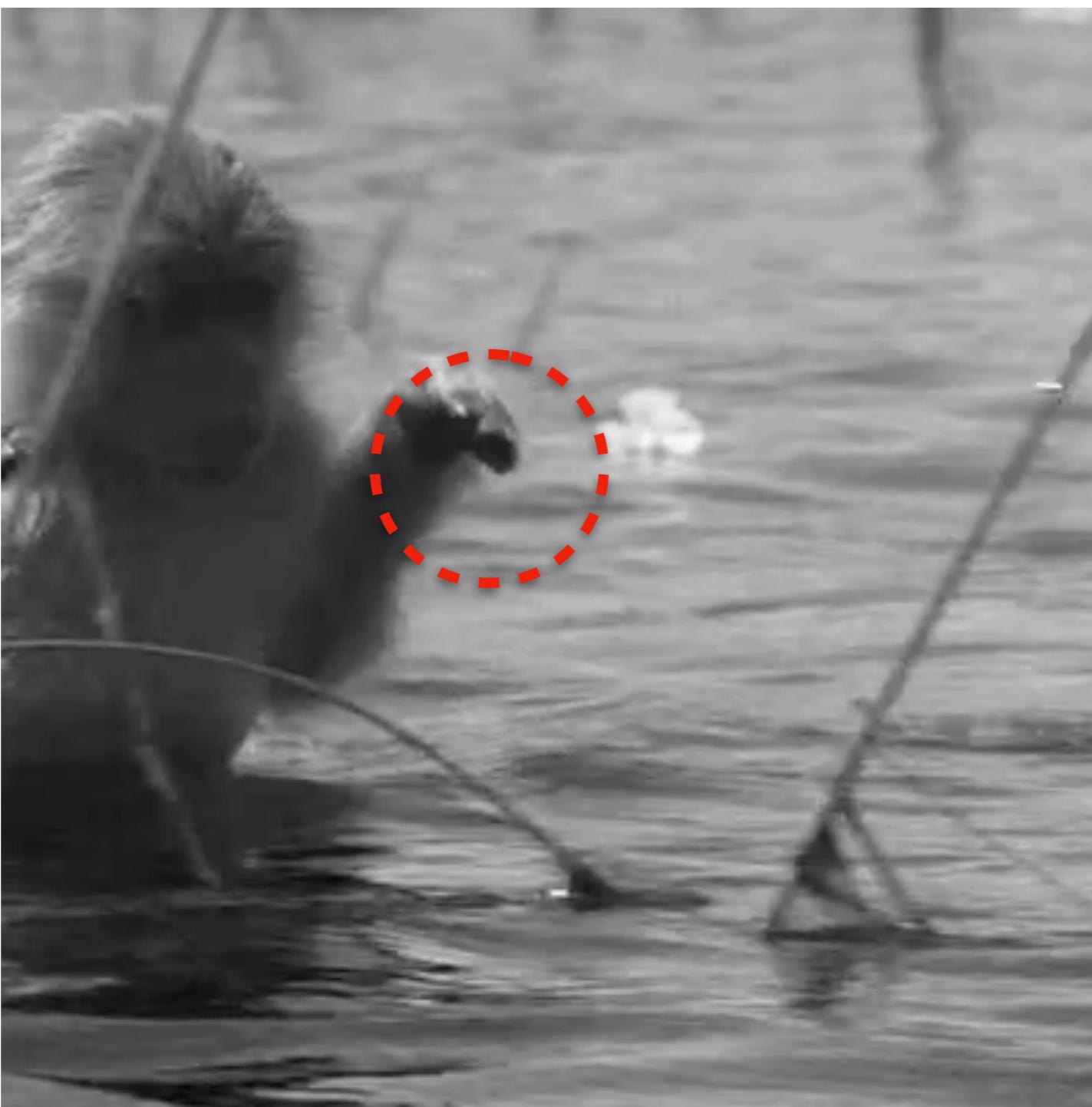




Neurons are noisy



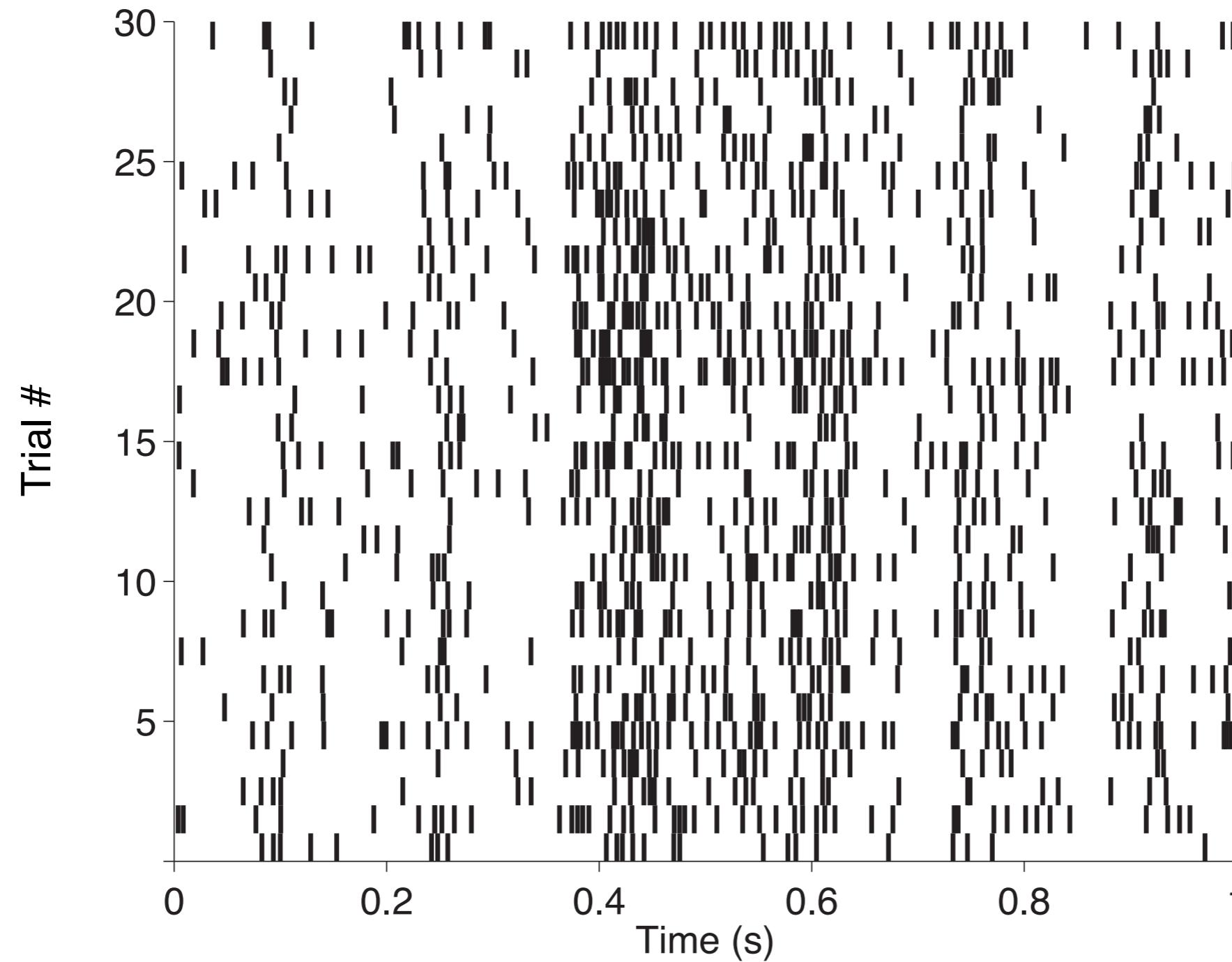
Neurons are noisy



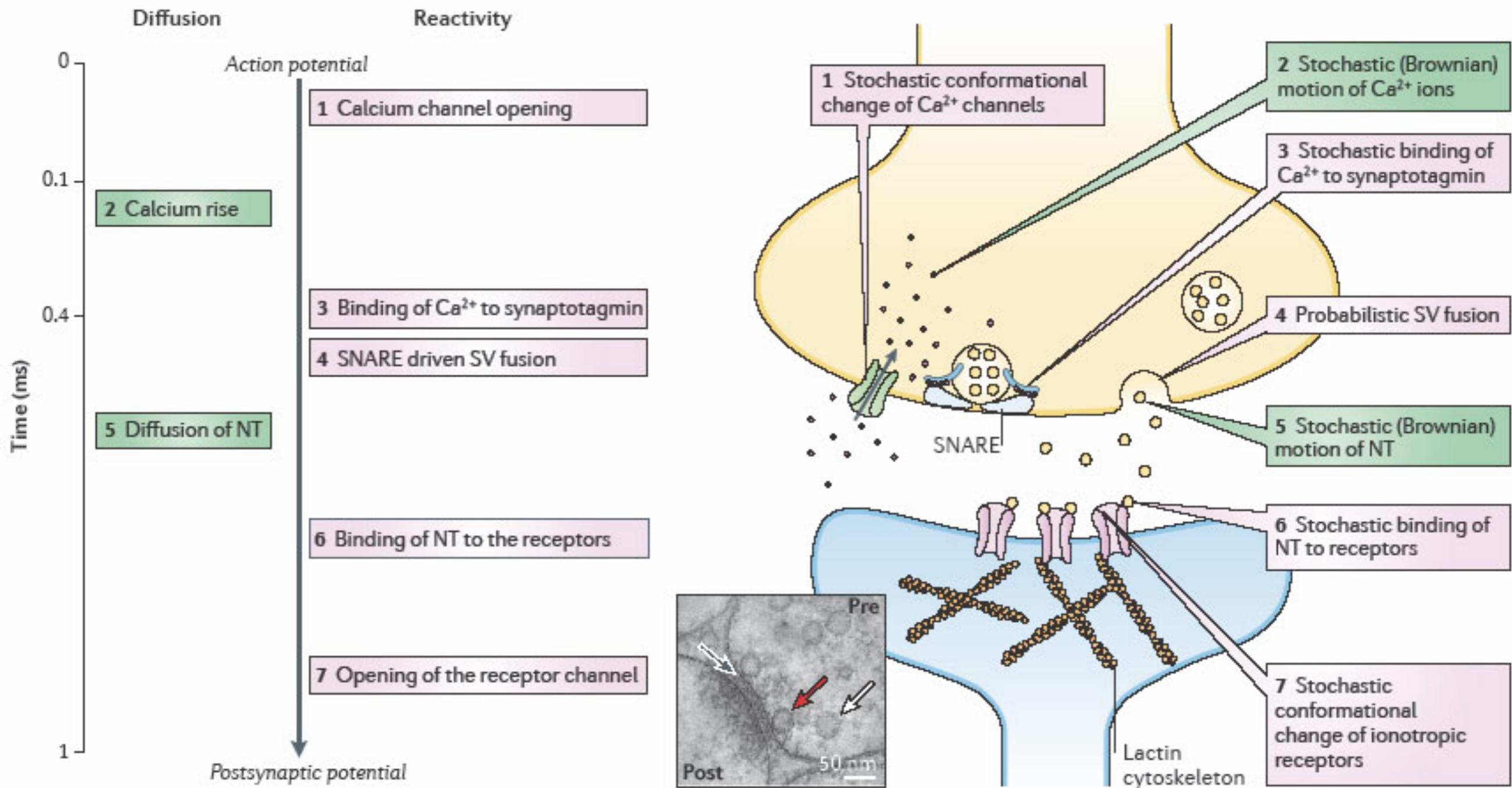
Neurons are noisy



Neurons are noisy



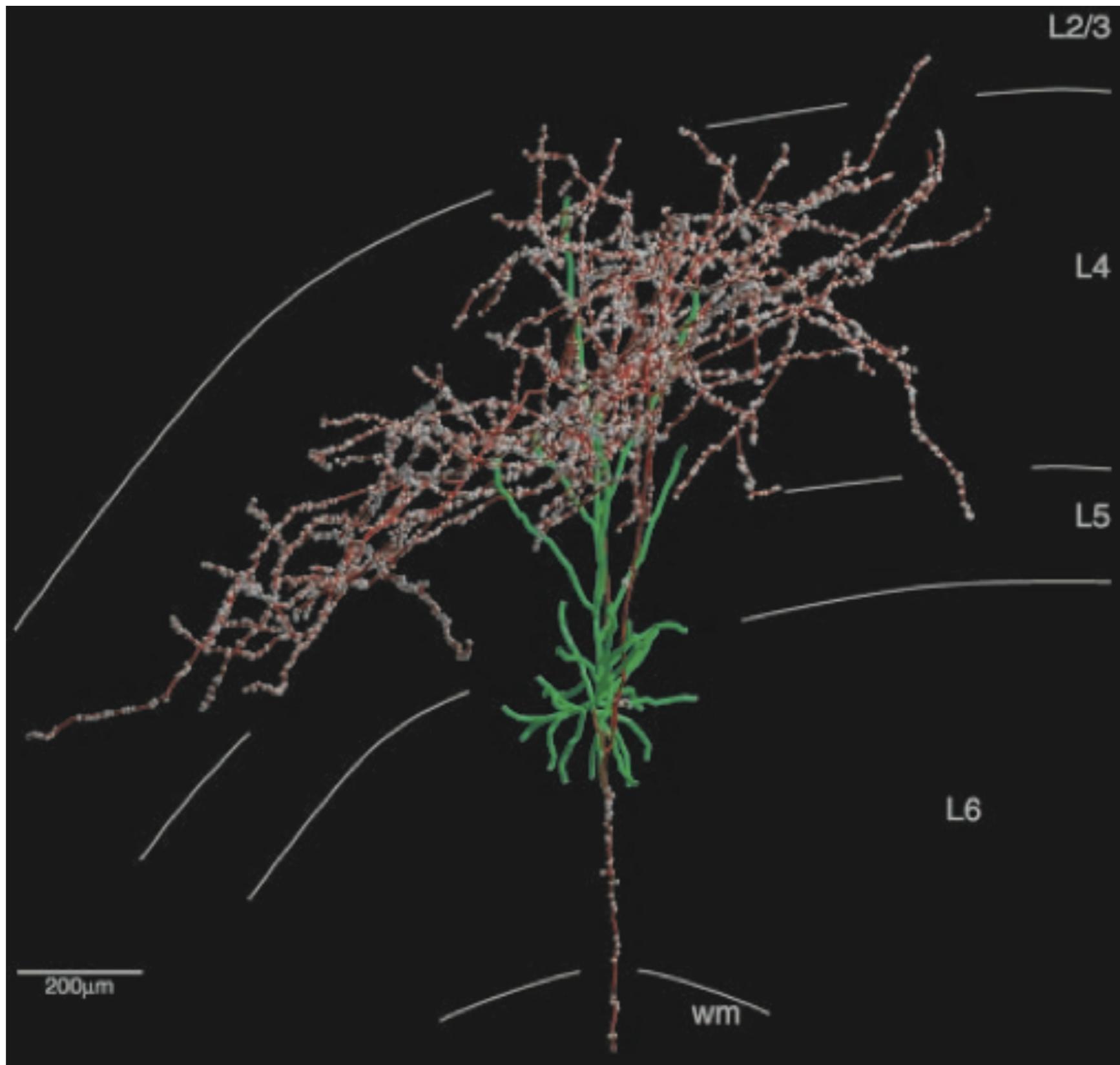
Neurons are noisy



Nature Reviews | Neuroscience

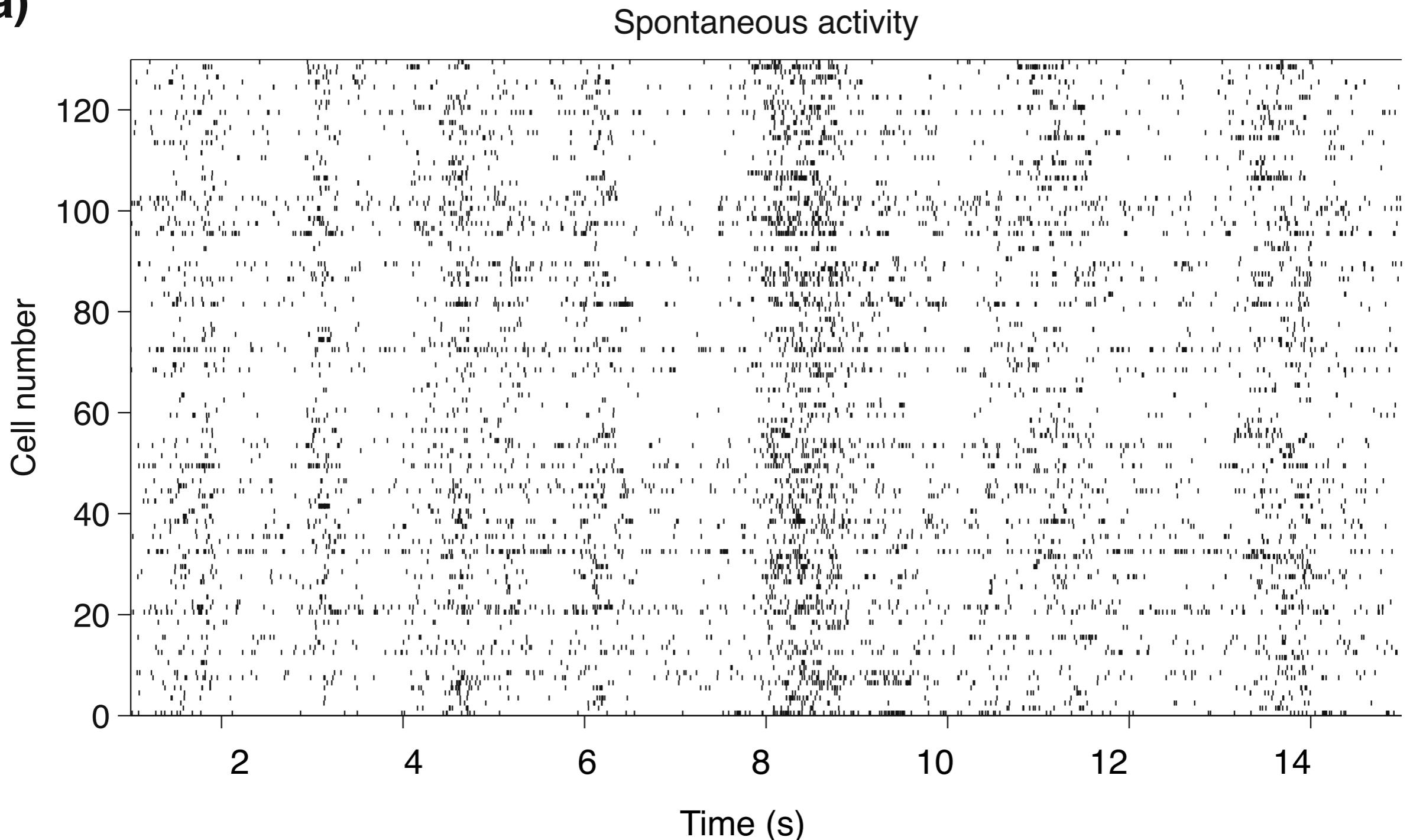
Ribraut, Sekimoto & Triller, 2011

Neurons are noisy



Neuronal populations have correlated noise

(a)



Correlation: a fundamental way to study interactions

Covariation in X and Y

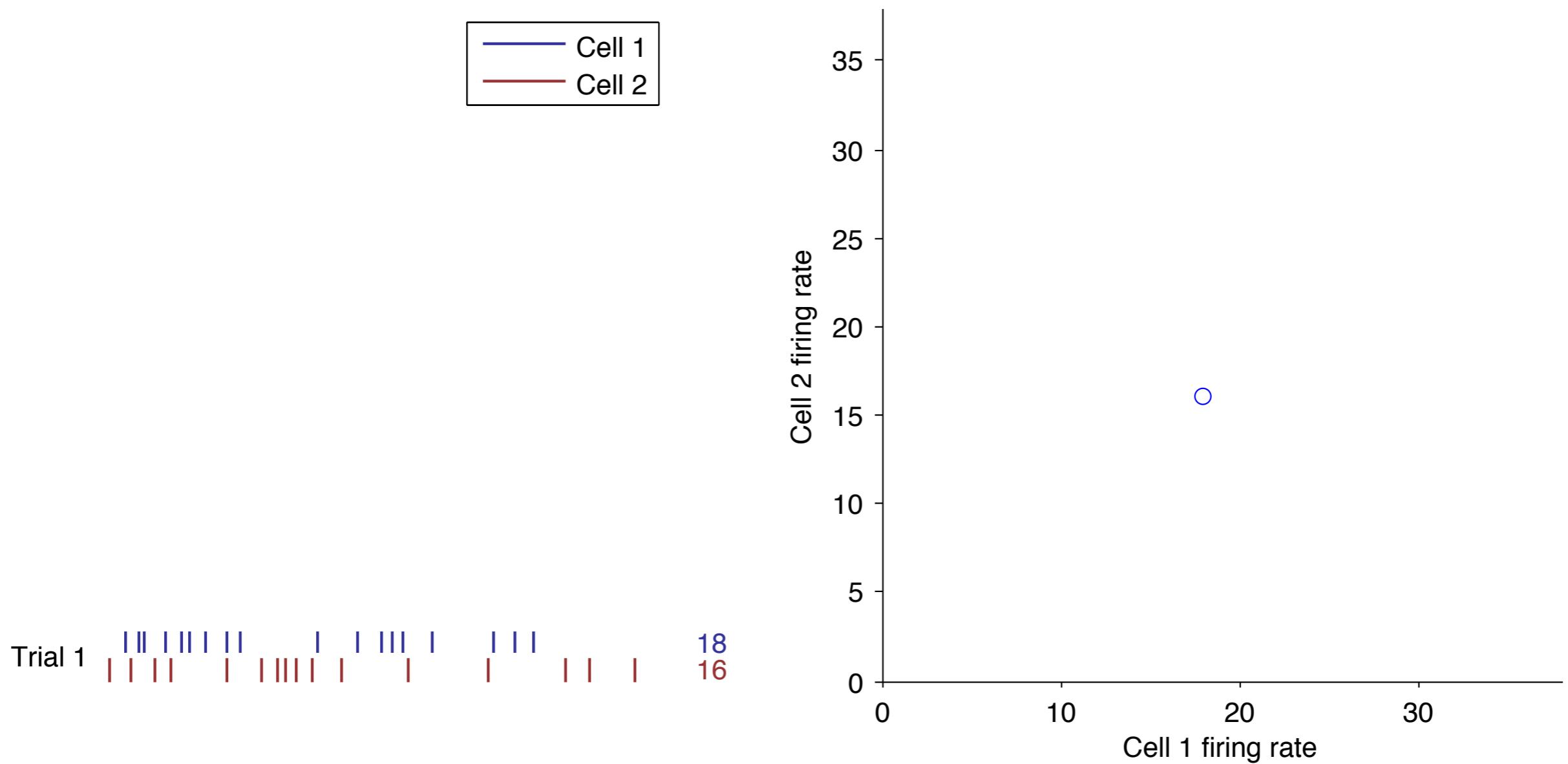
$$r = \frac{\sum_{i=1}^n ((x_i - \bar{x})(y_i - \bar{y}))}{\sqrt{\sum_{i=1}^n (x_i - \bar{x})^2 \sum_{i=1}^n (y_i - \bar{y})^2}}$$

↓

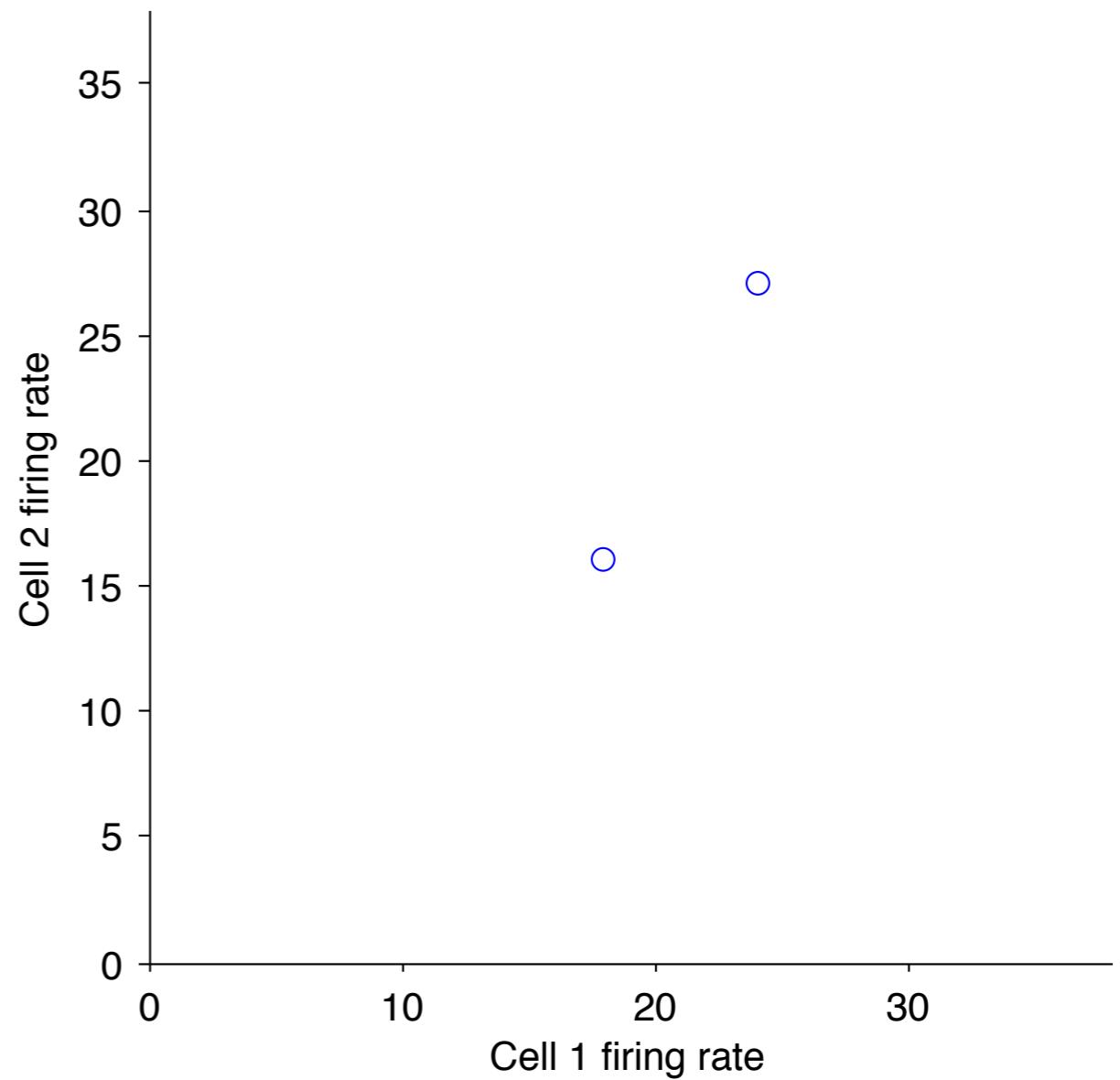
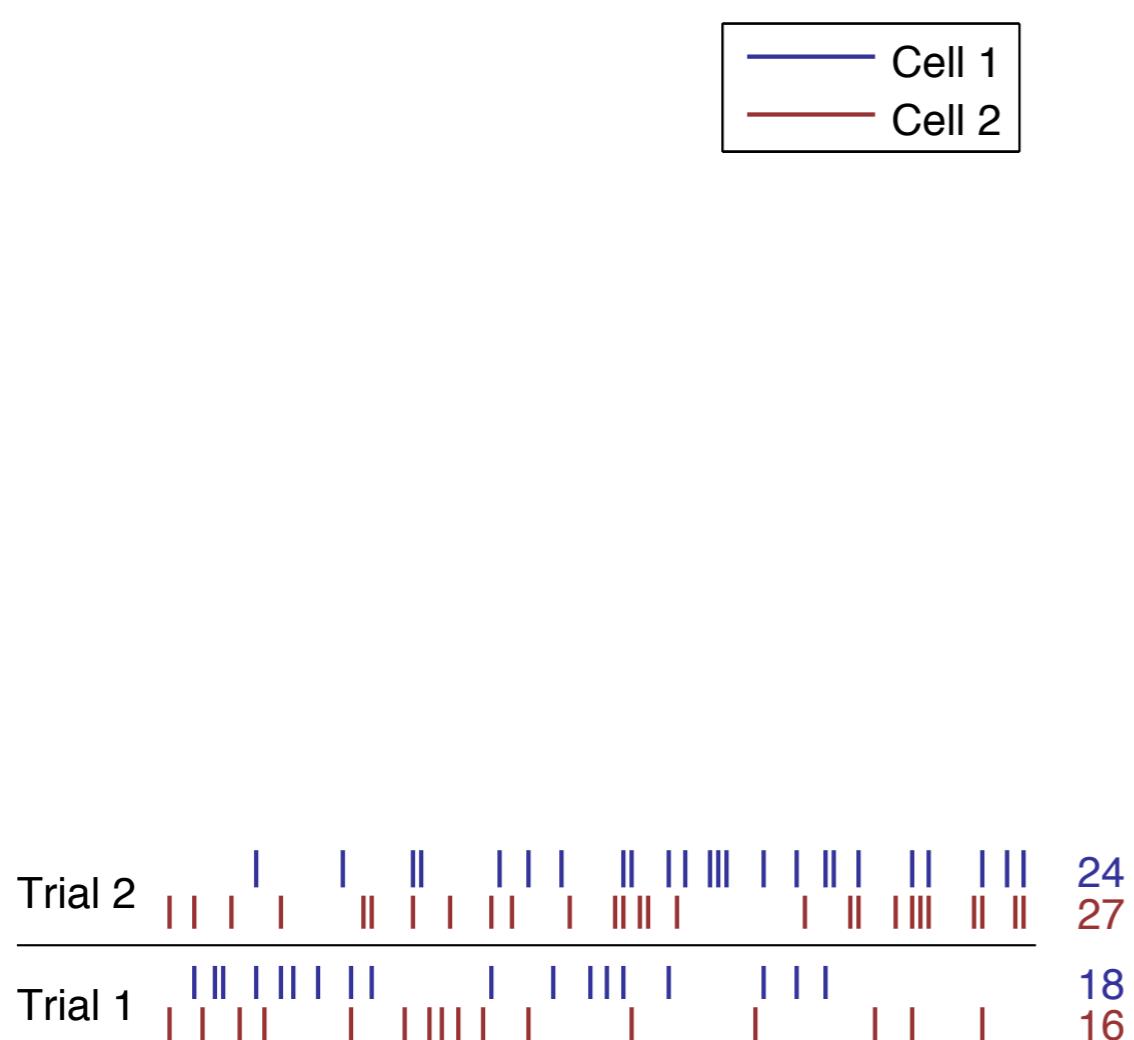
Variation in X Variation in Y

The diagram illustrates the formula for correlation coefficient r . At the top, the text "Covariation in X and Y" is centered above a downward-pointing arrow. Below the arrow is the formula for r , which consists of two main parts: the numerator and the denominator. The numerator is the sum of the products of the deviations of x_i and y_i from their respective means \bar{x} and \bar{y} , for all i from 1 to n . The denominator is the square root of the product of two sums: the first sum is the sum of the squared deviations of x_i from \bar{x} for all i from 1 to n ; the second sum is the sum of the squared deviations of y_i from \bar{y} for all i from 1 to n . Two arrows point from the text "Variation in X" and "Variation in Y" at the bottom to the terms $(x_i - \bar{x})^2$ and $(y_i - \bar{y})^2$ respectively in the denominator.

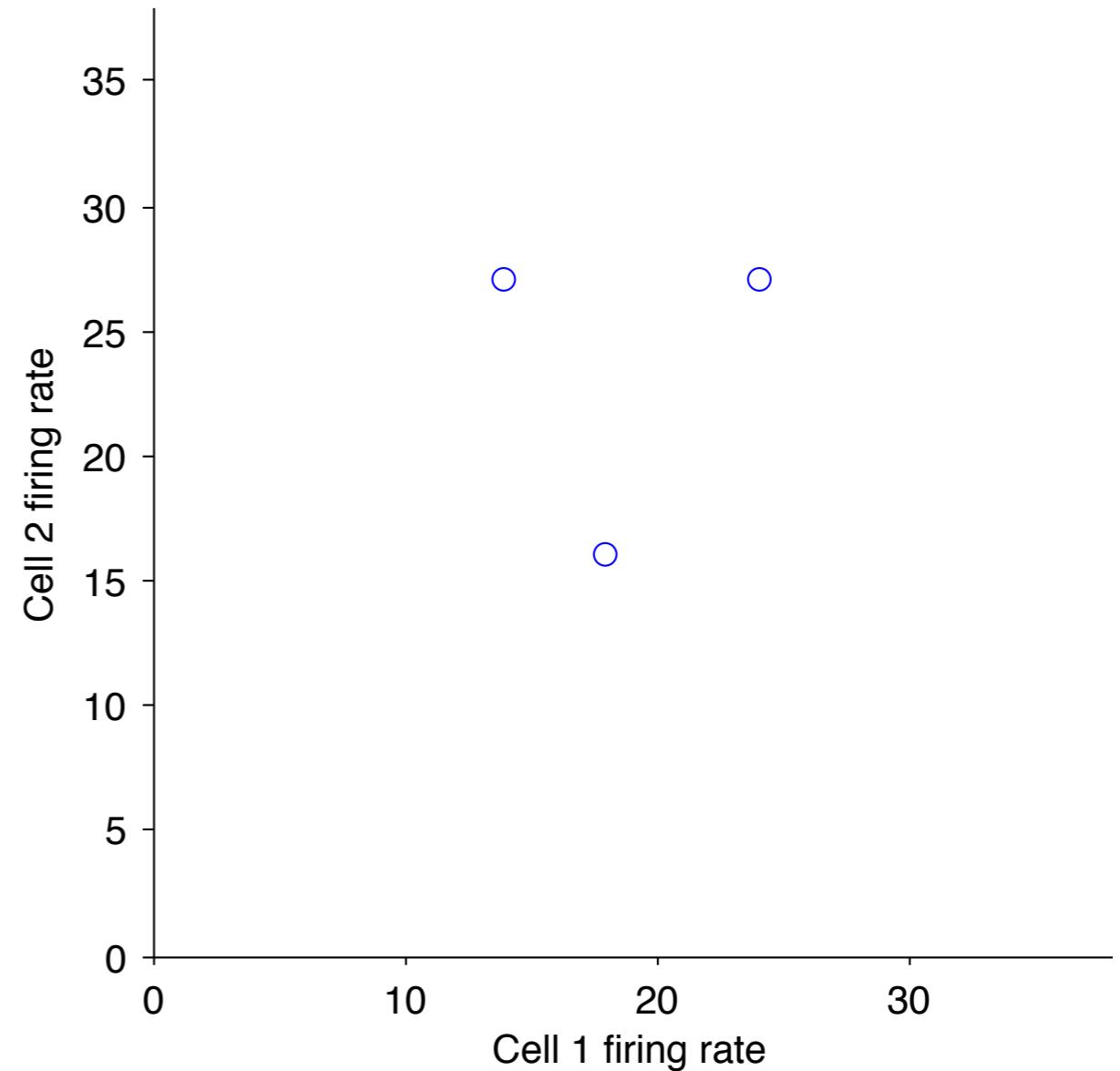
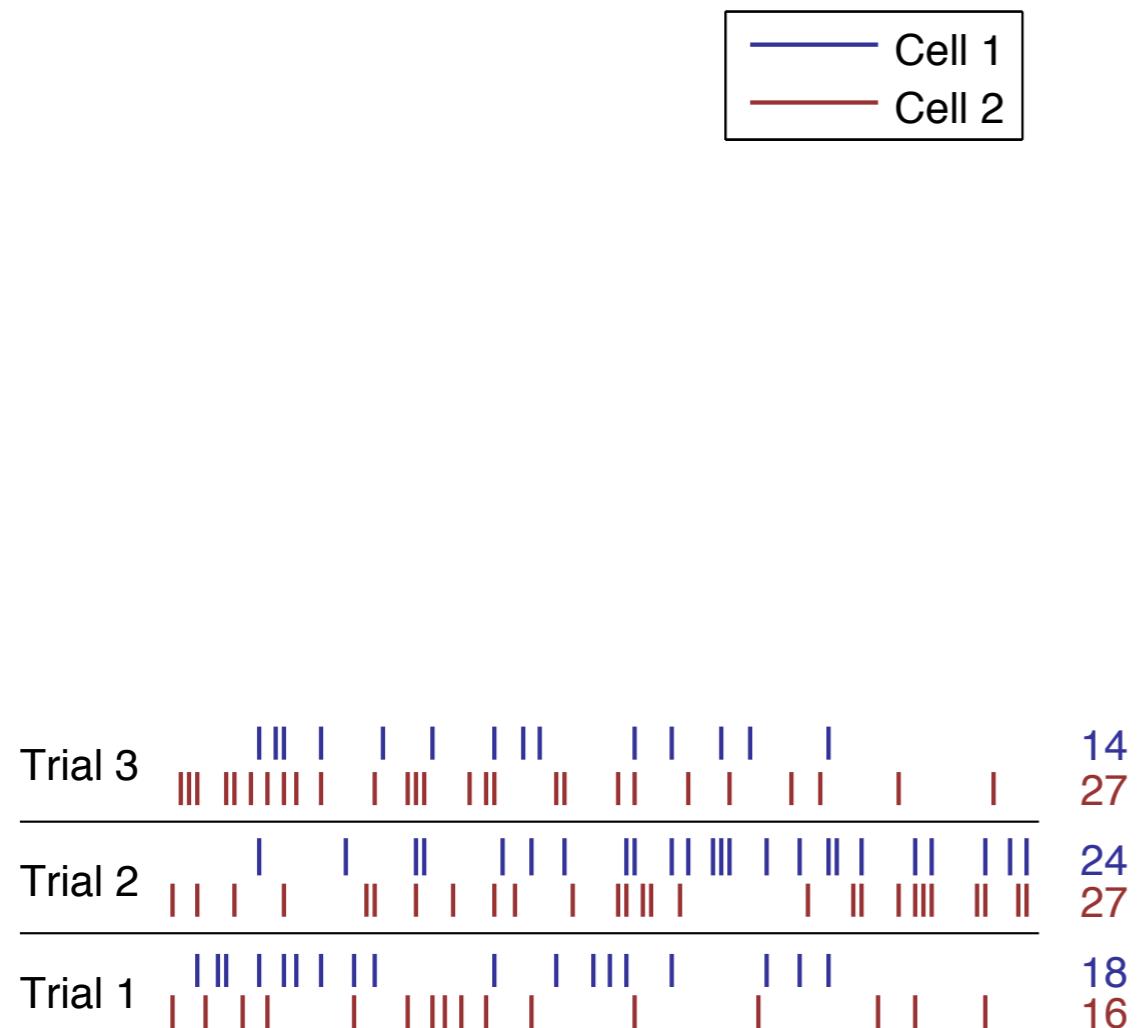
Spike count correlation (r_{sc})



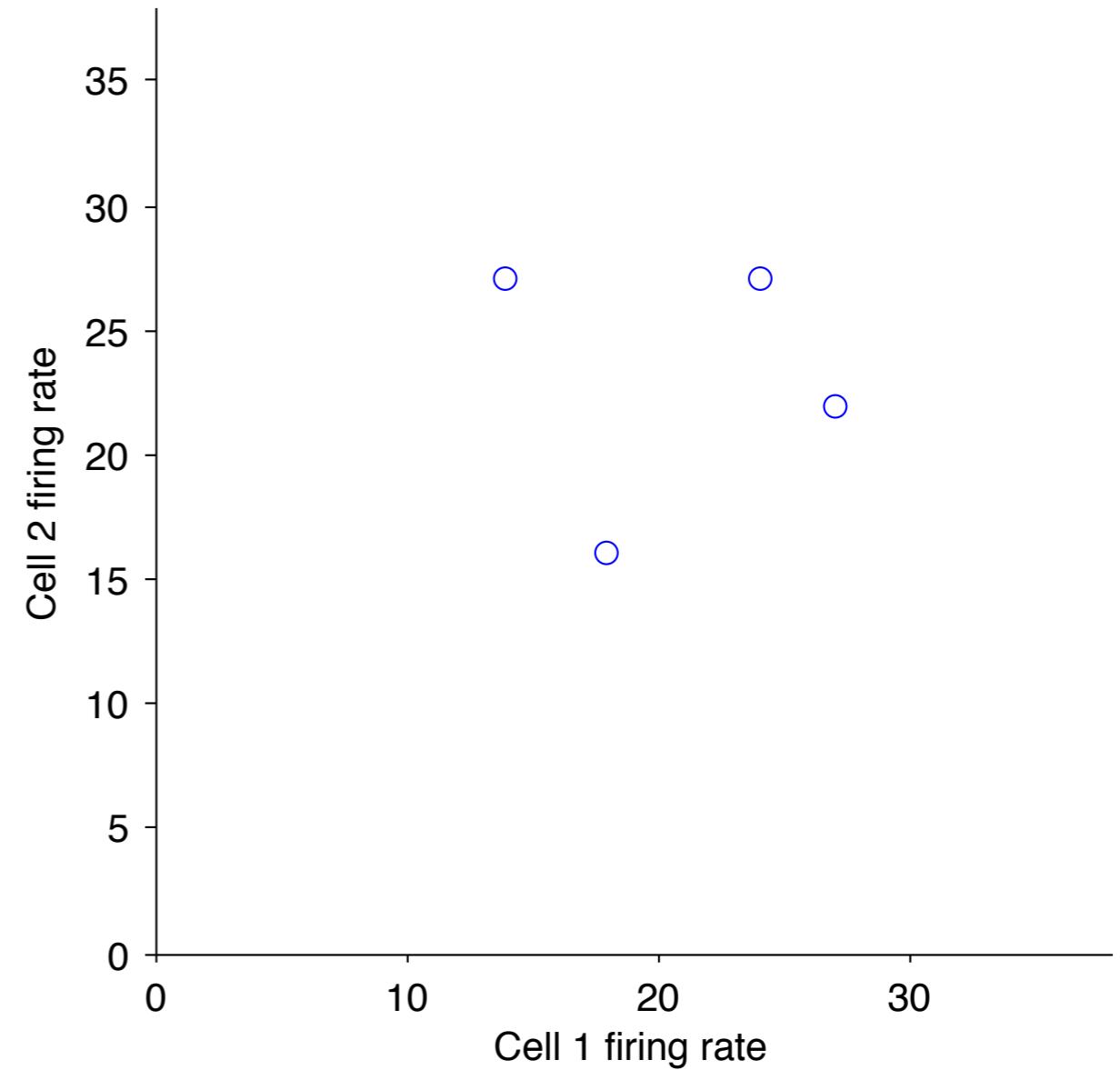
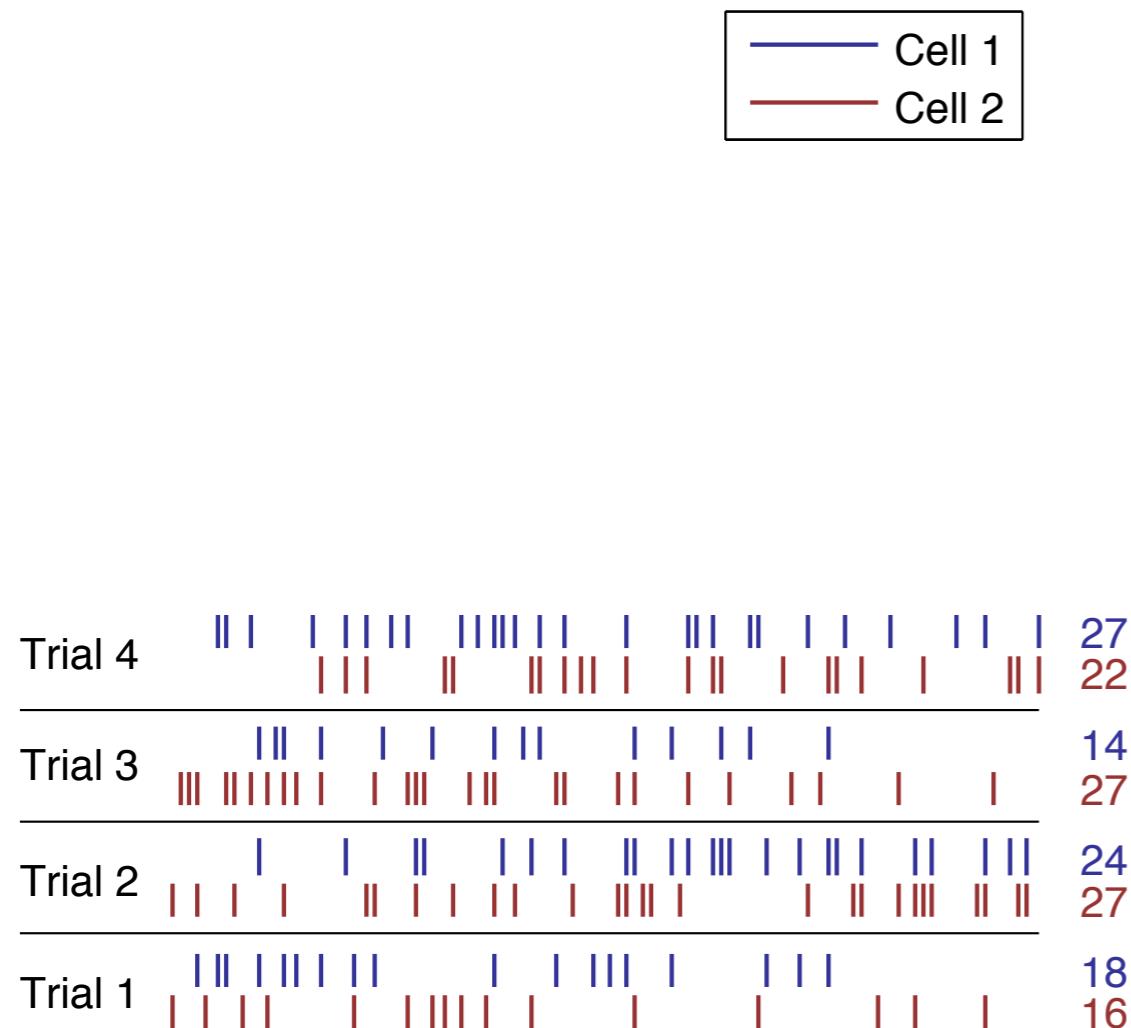
Spike count correlation (r_{sc})



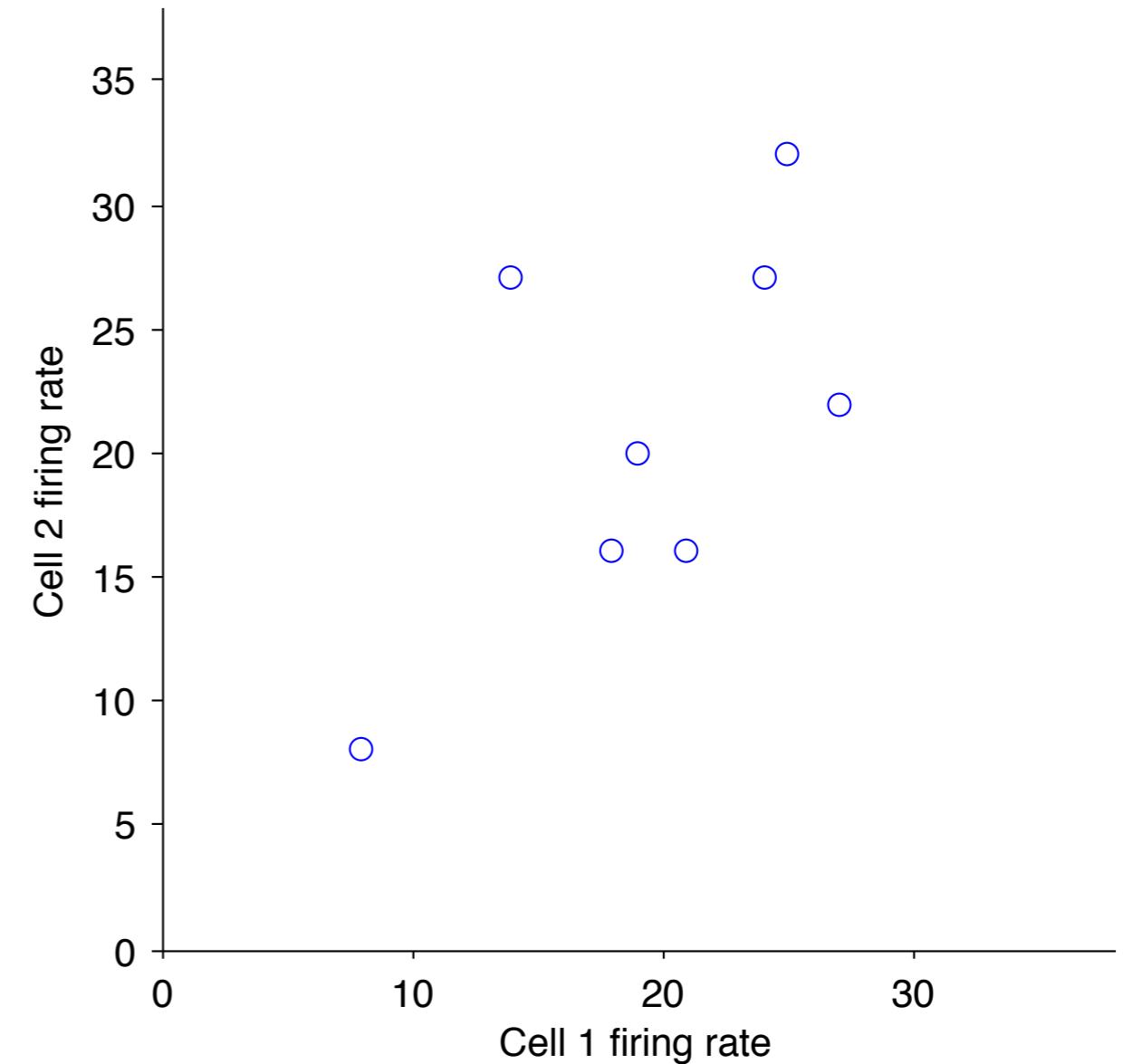
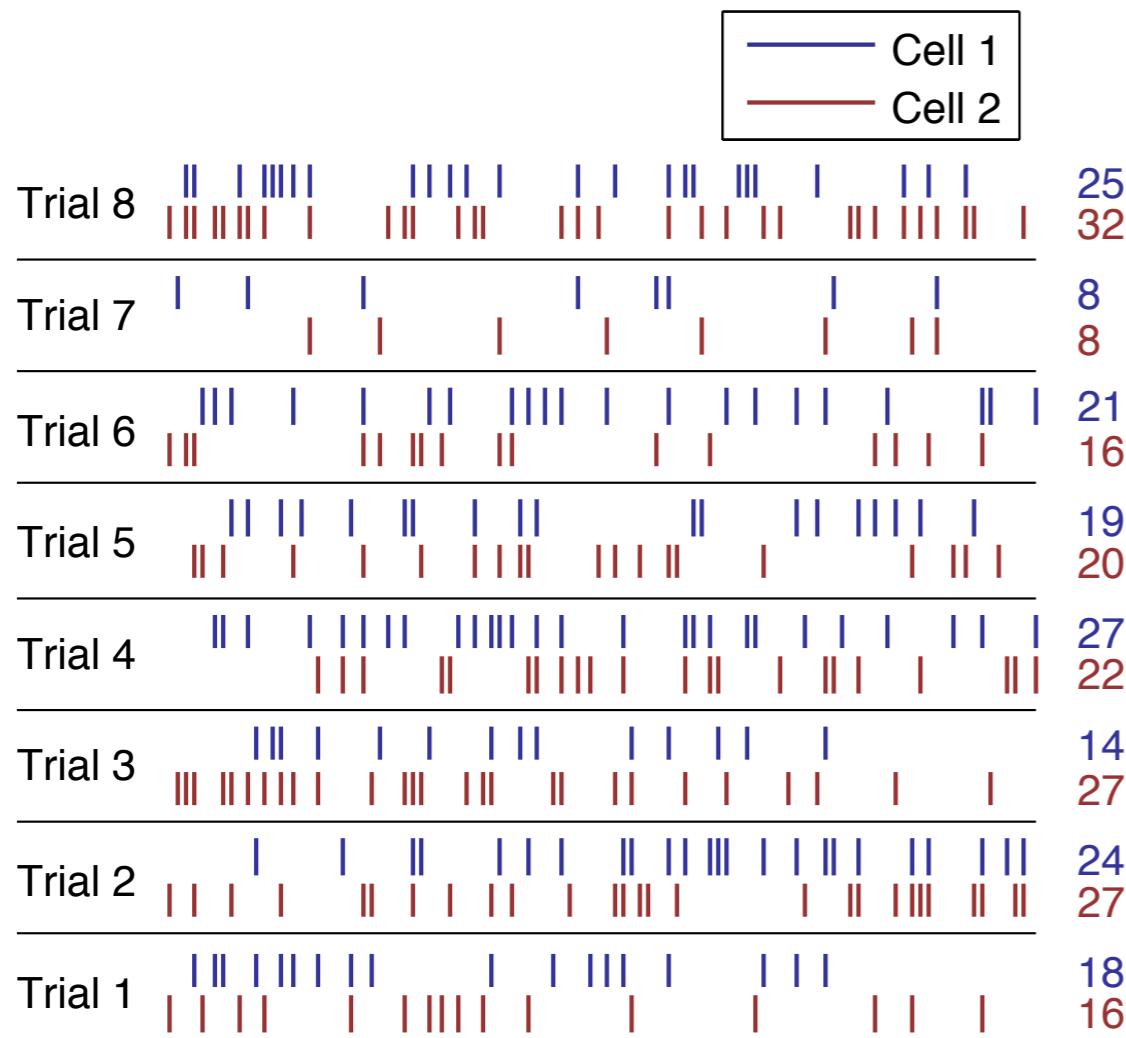
Spike count correlation (r_{sc})



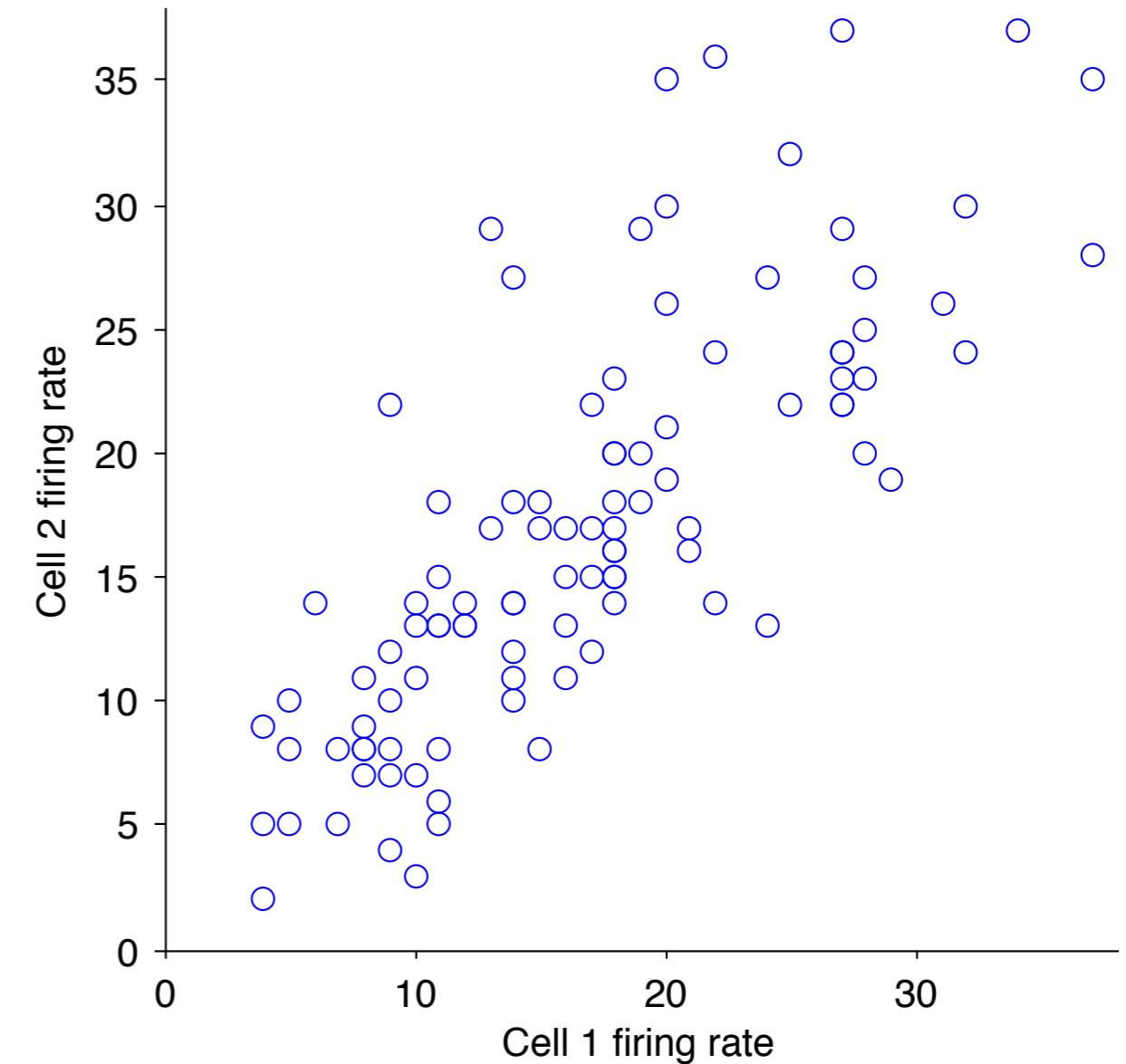
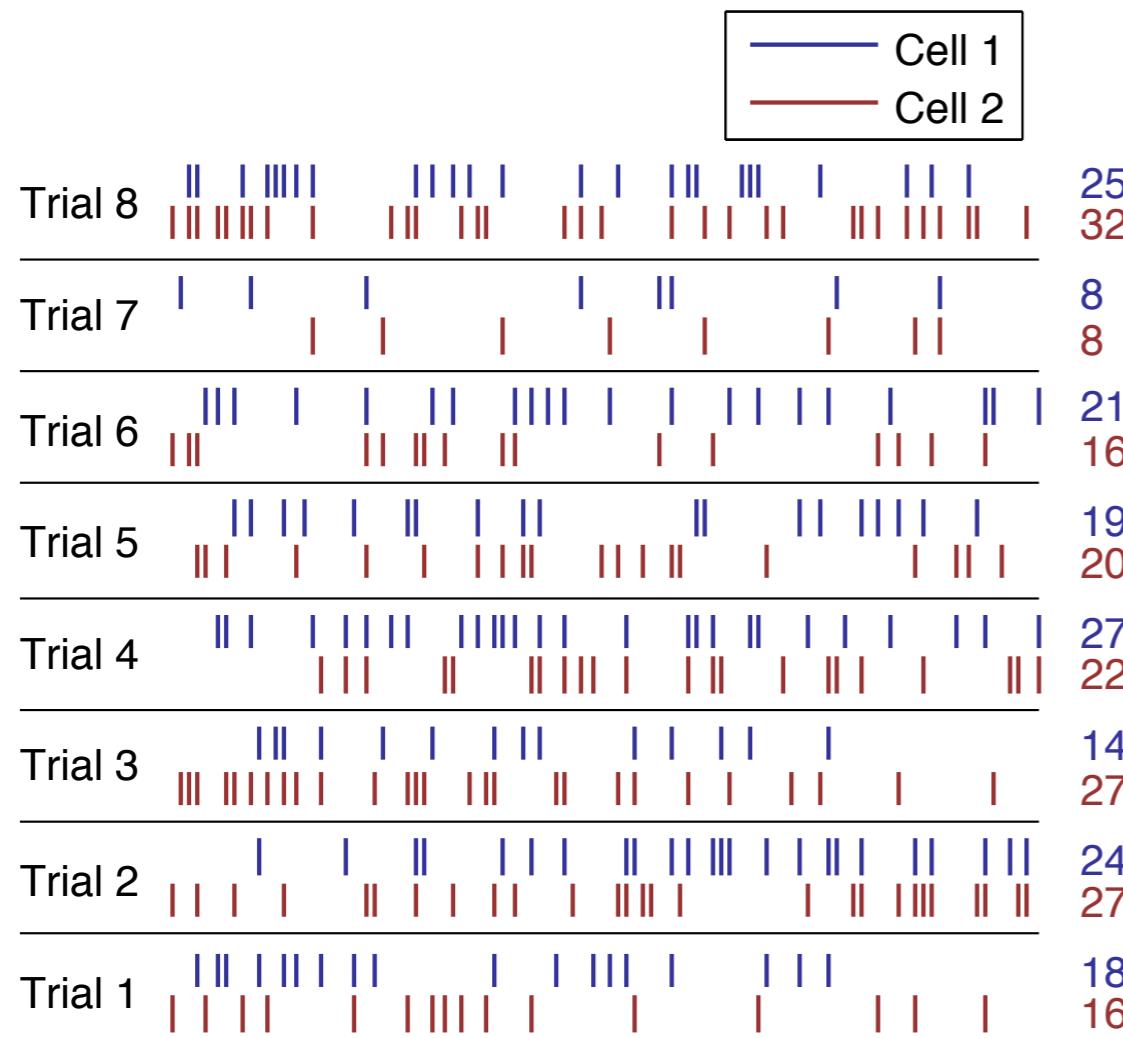
Spike count correlation (r_{sc})



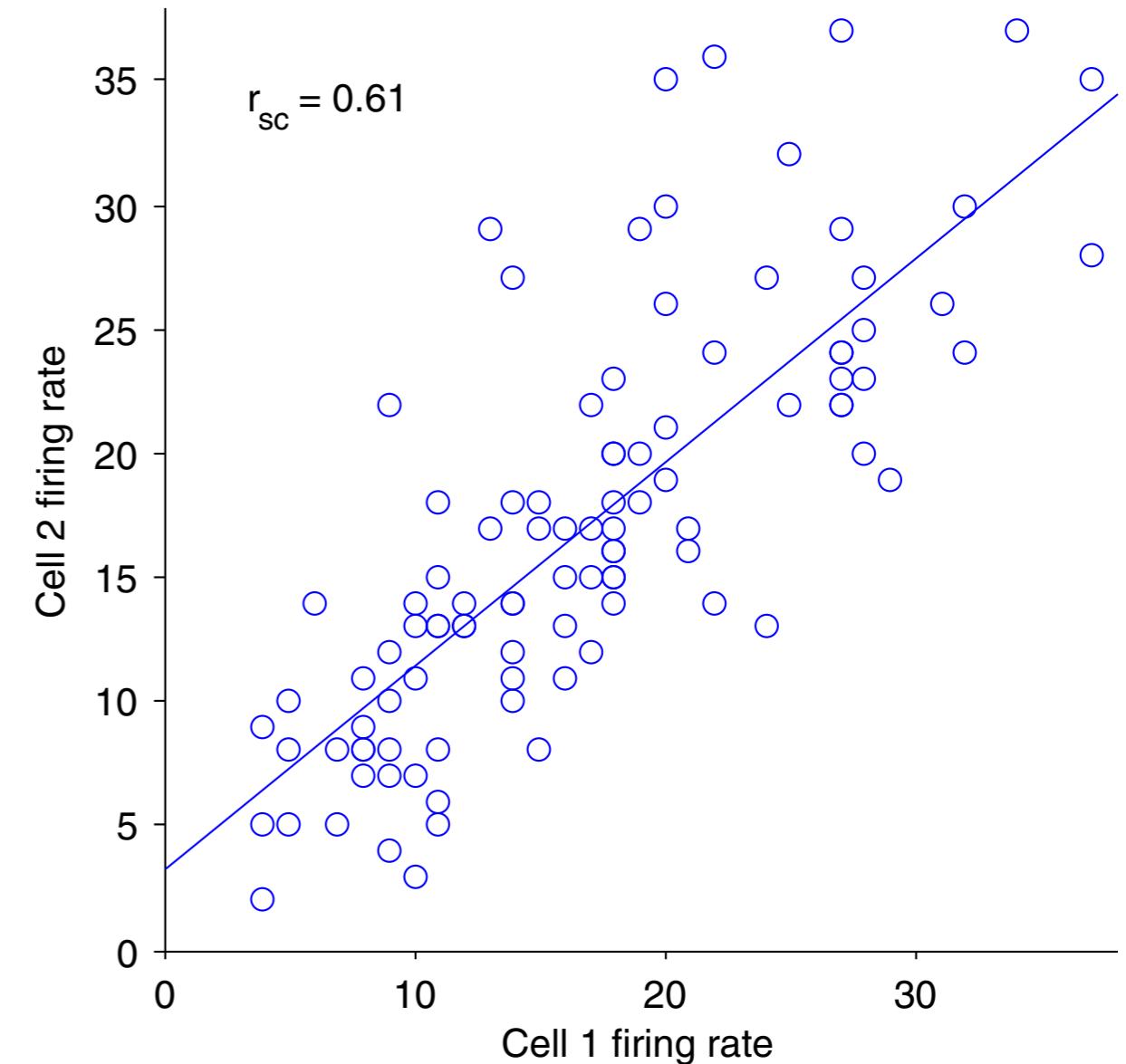
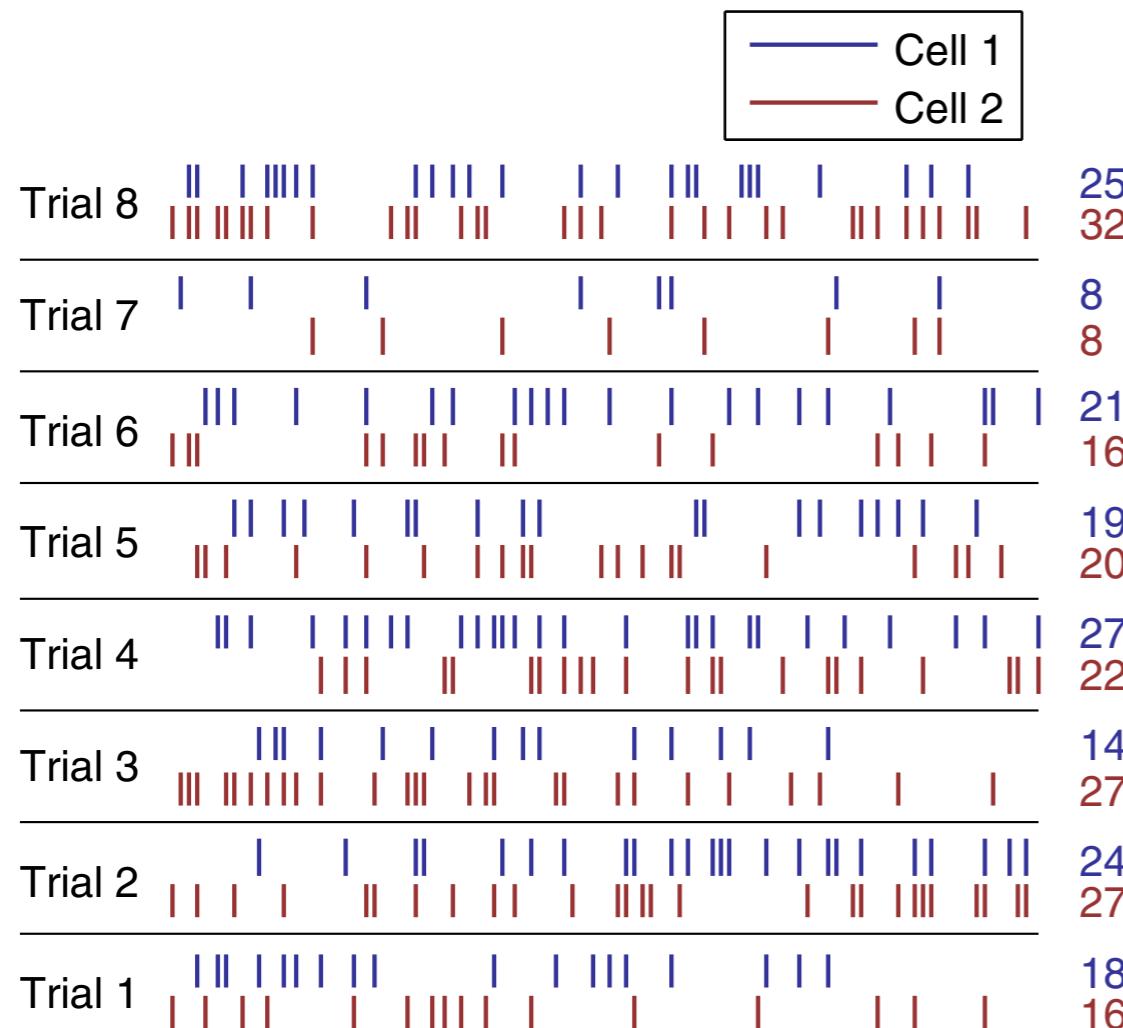
Spike count correlation (r_{sc})



Spike count correlation (r_{sc})

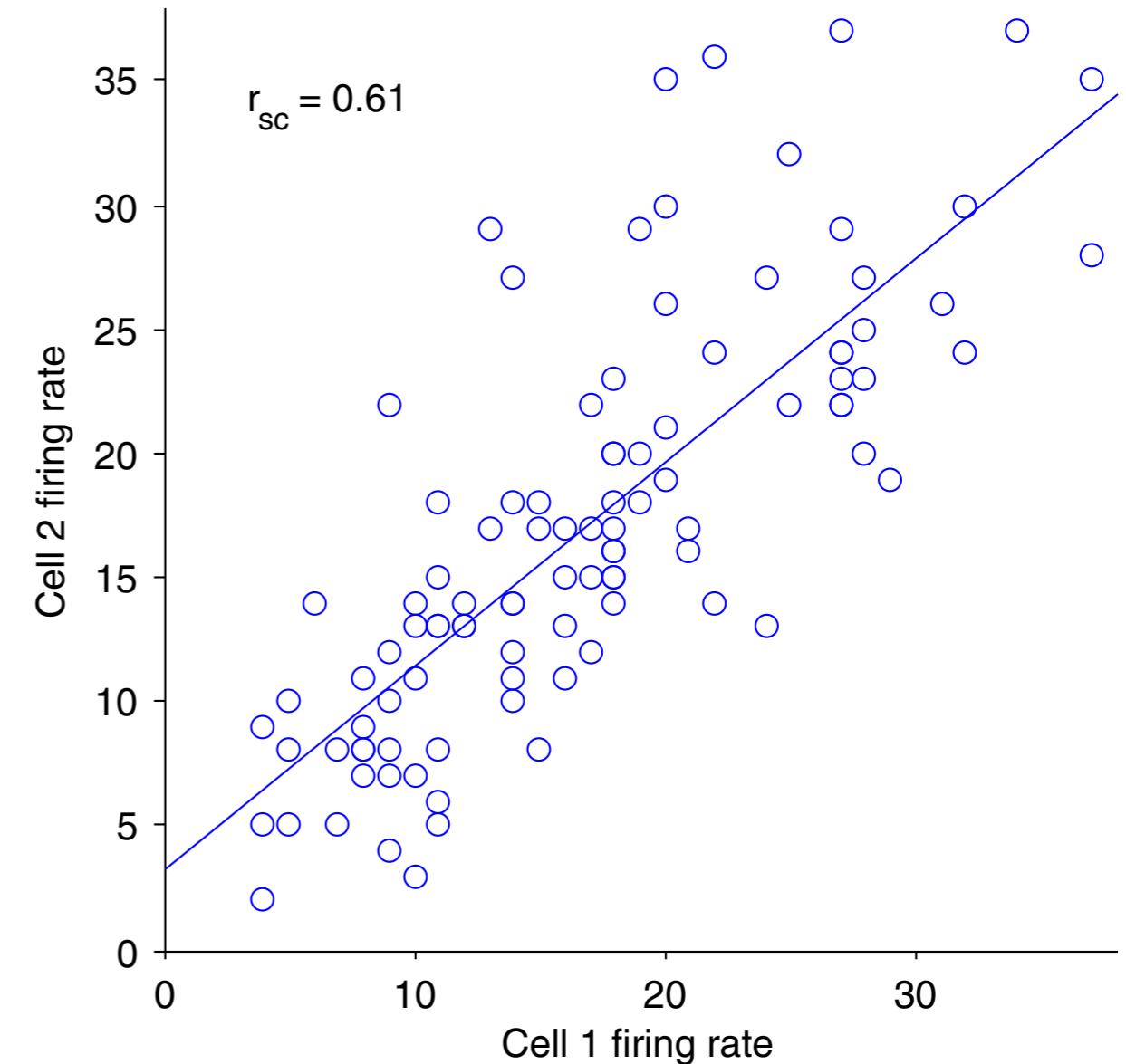
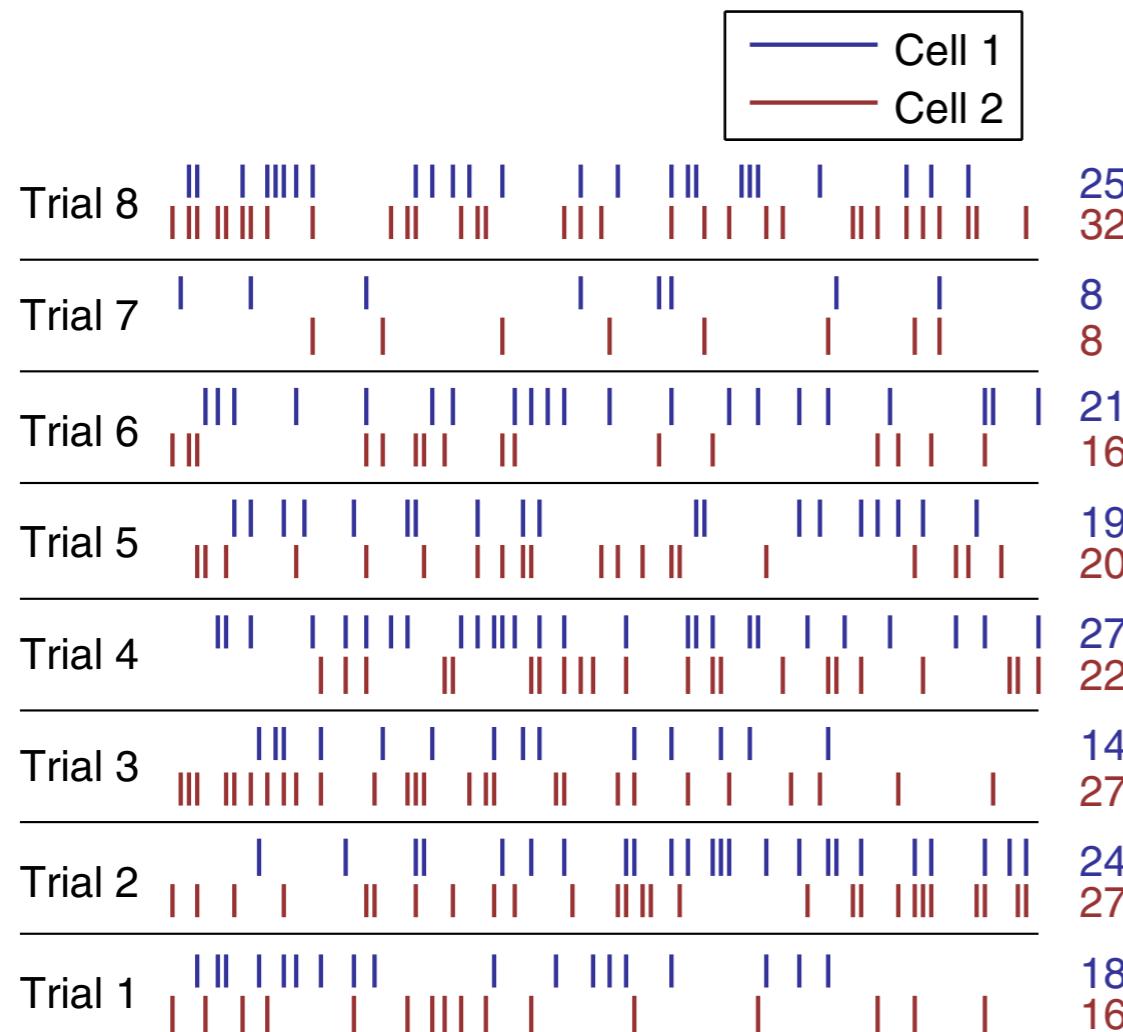


Spike count correlation (r_{sc})

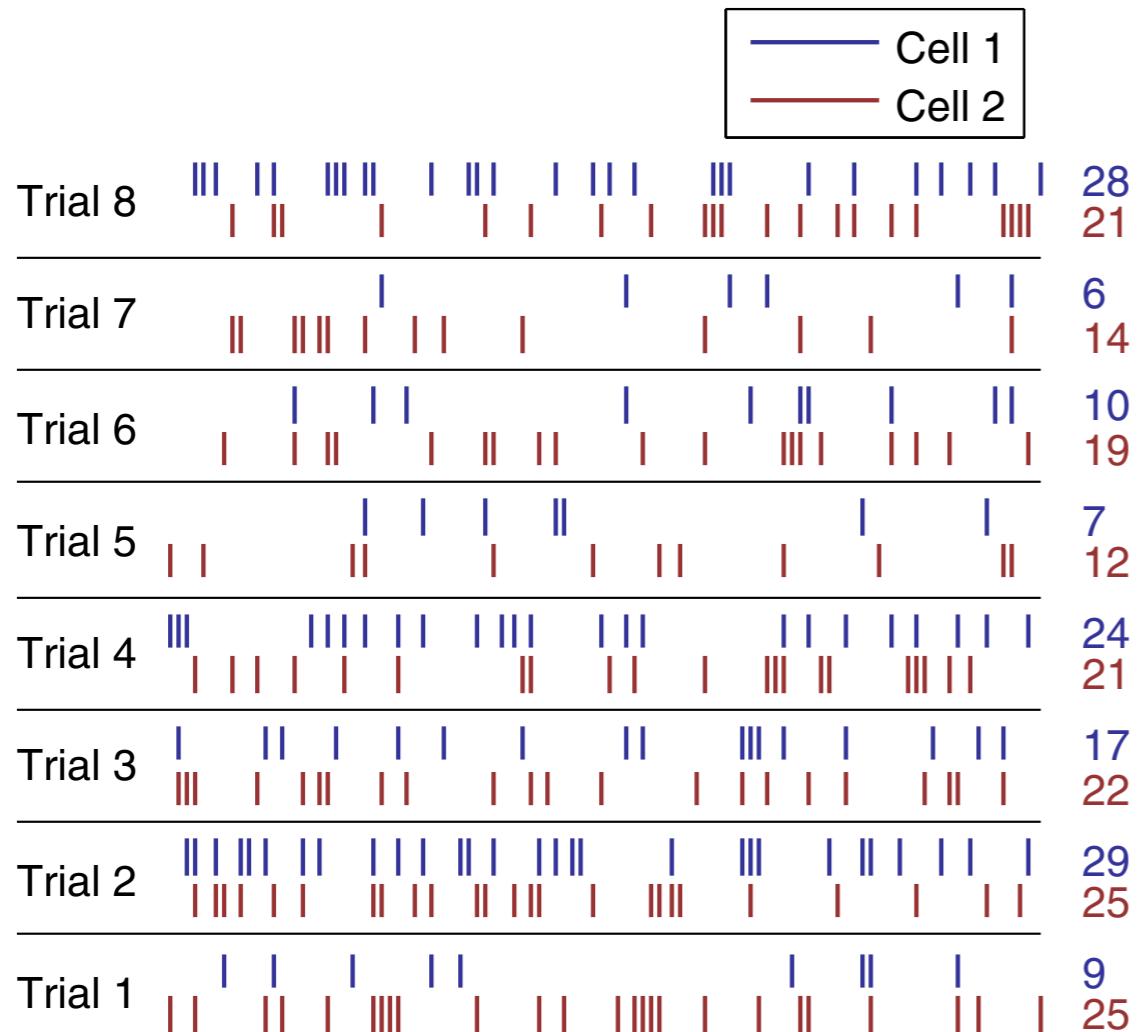


Spike count correlation (r_{sc})

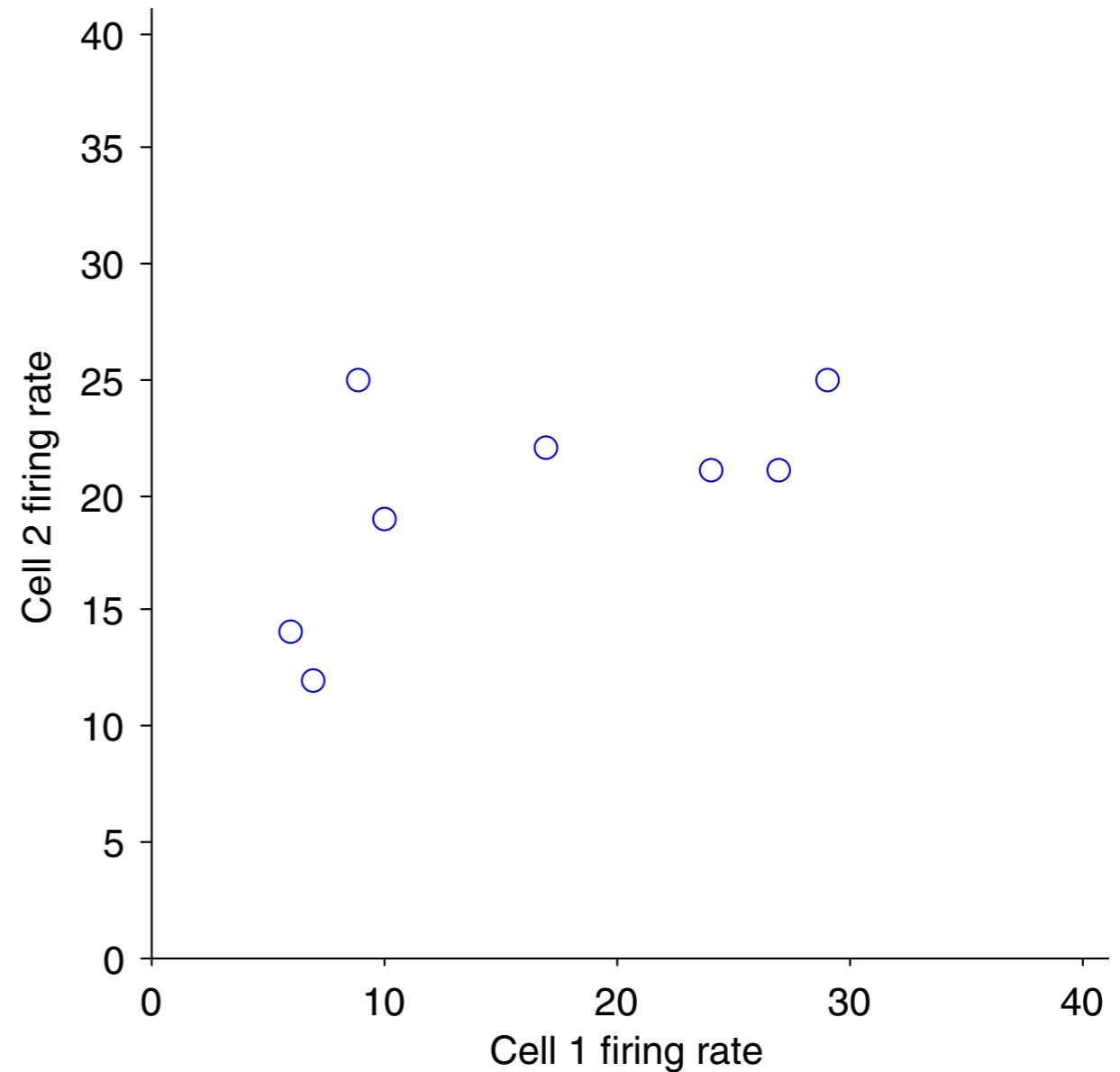
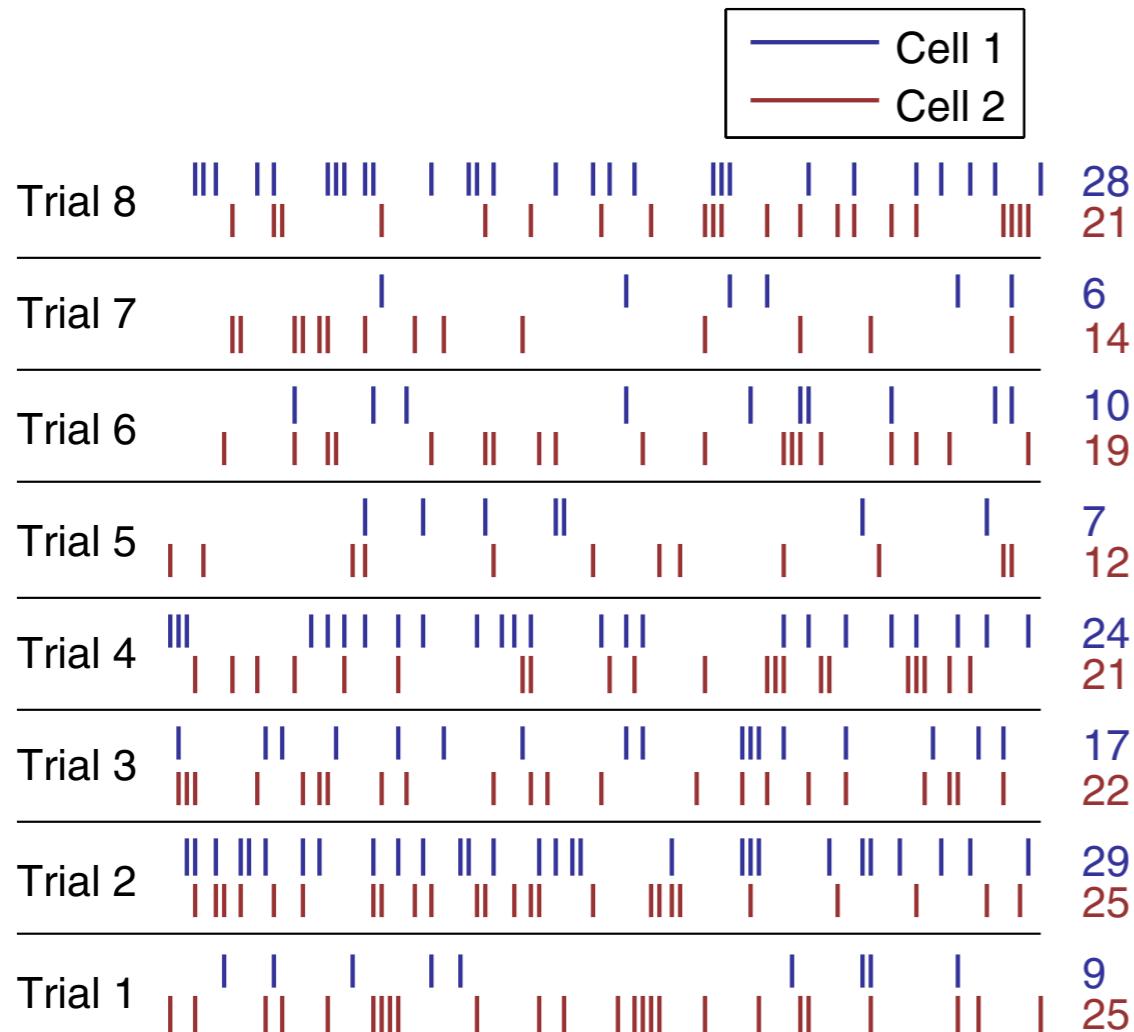
Slow time scale correlation



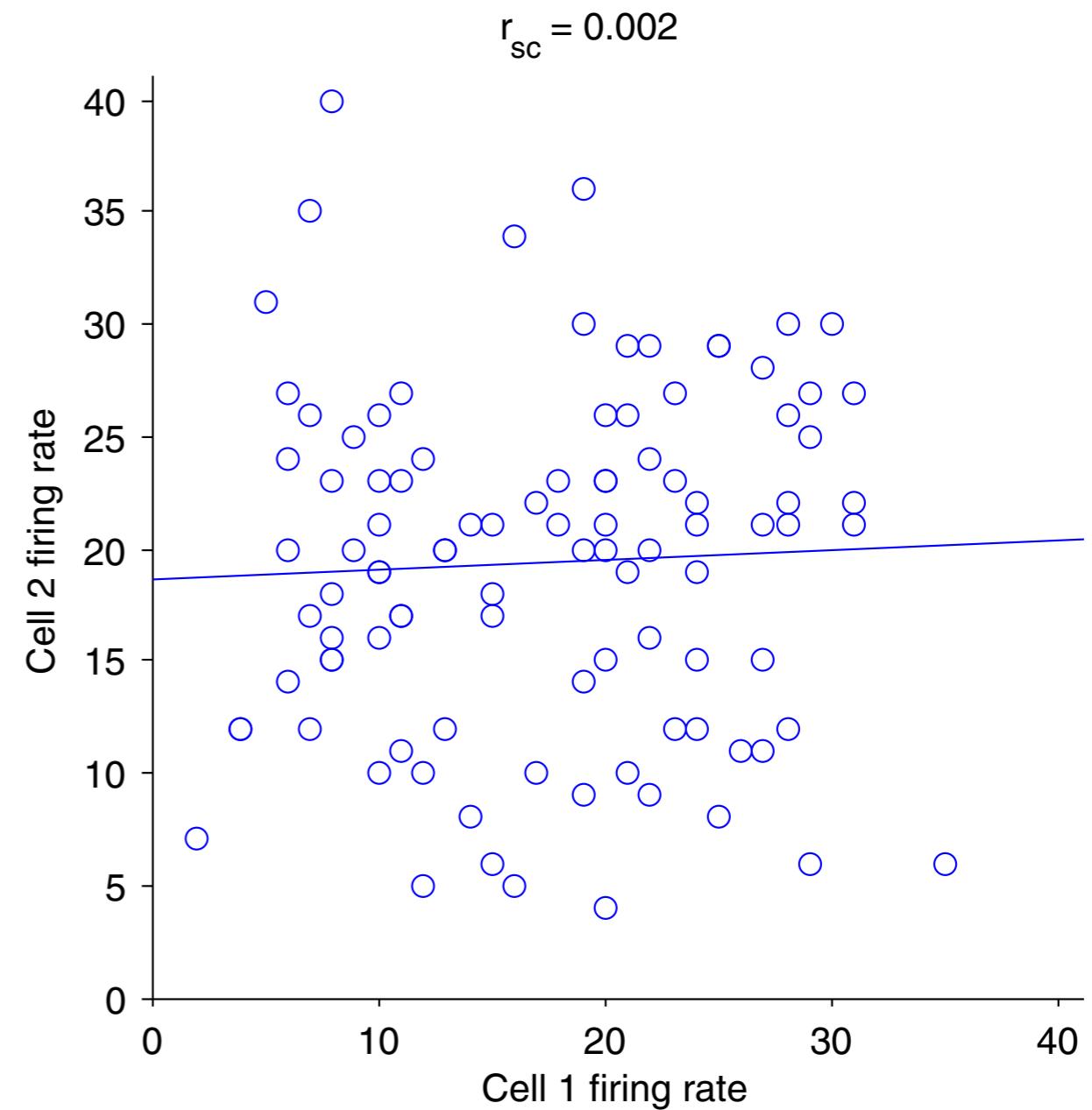
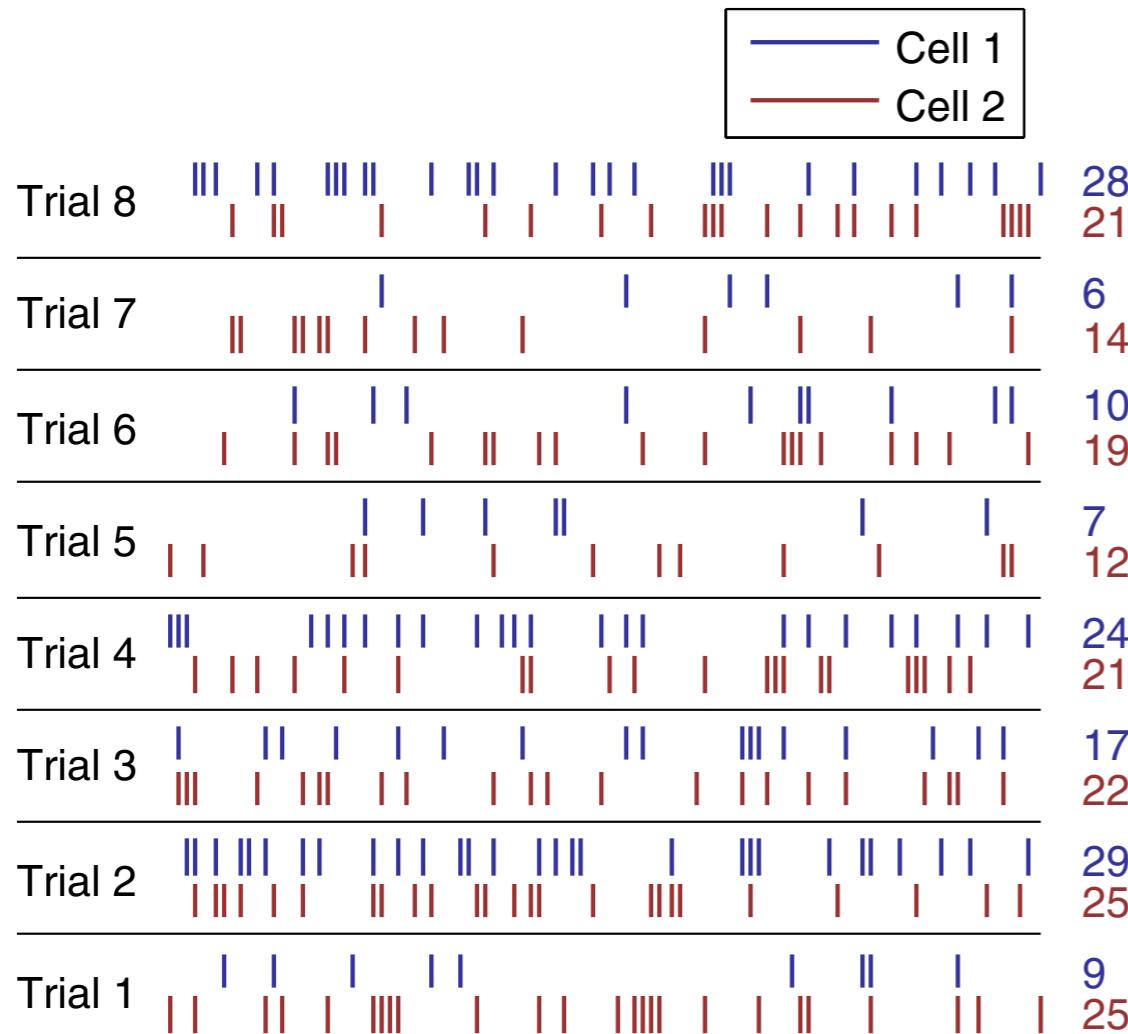
Spike count correlation (r_{sc})



Spike count correlation (r_{sc})

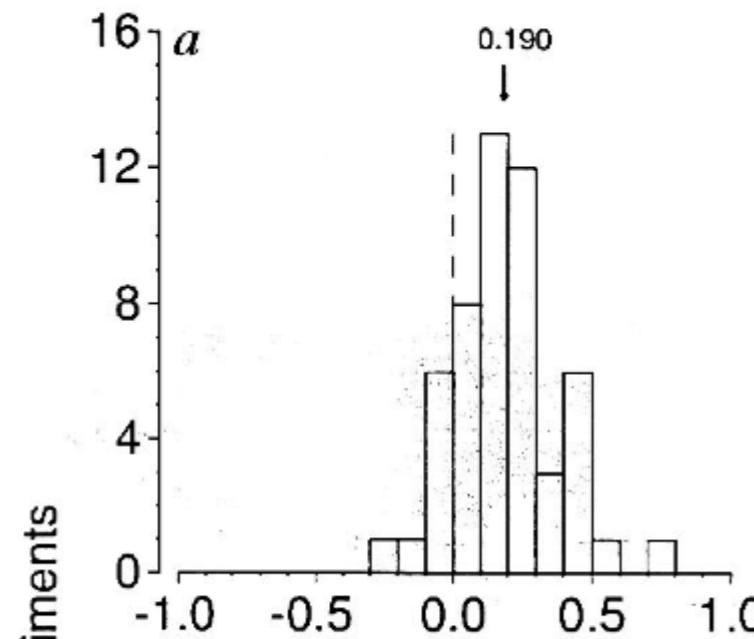


Spike count correlation (r_{sc})

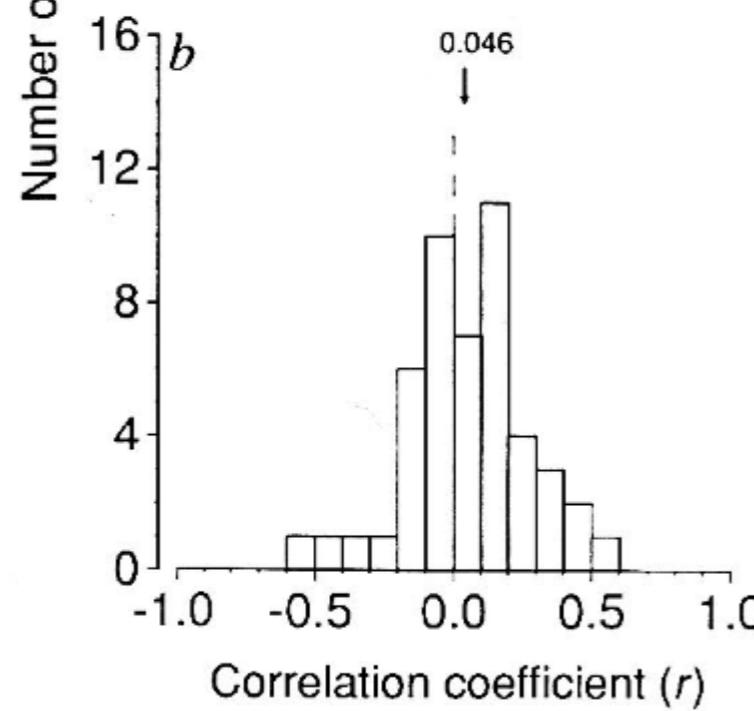


Correlation: a fundamental way to study interactions

similar direction preference

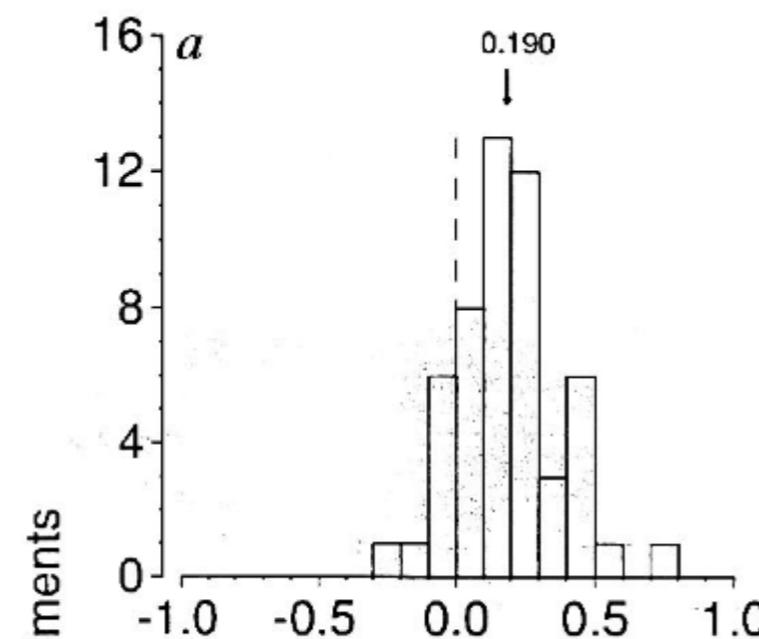


different direction preference

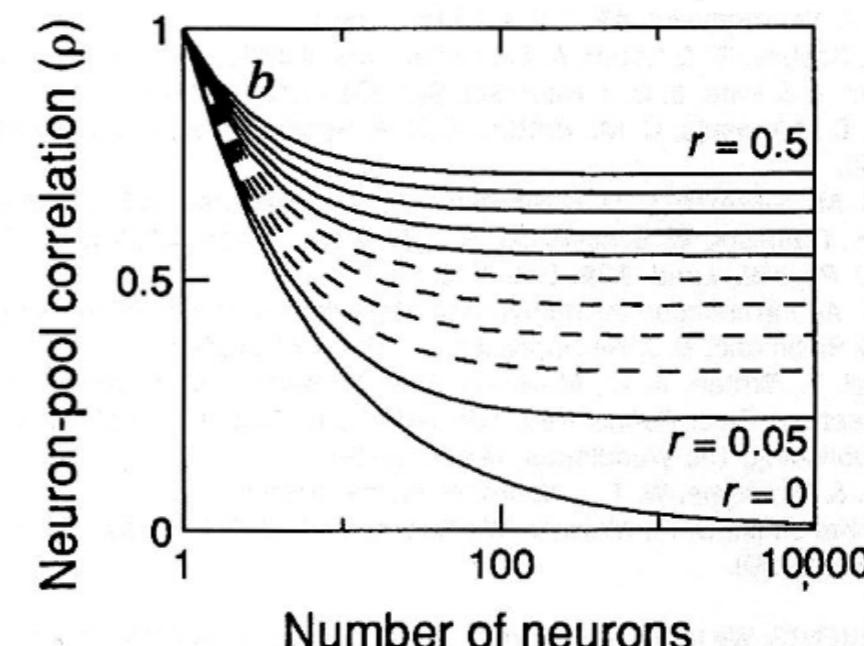
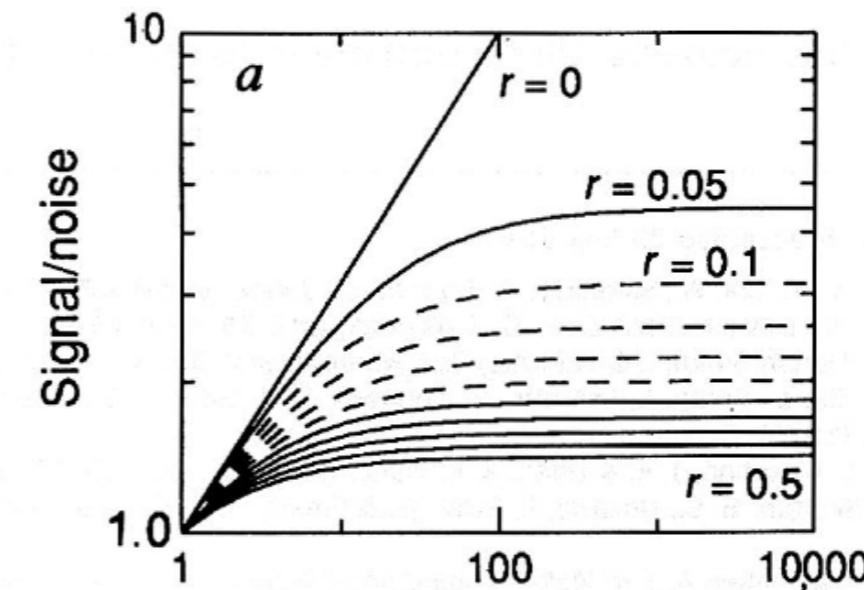
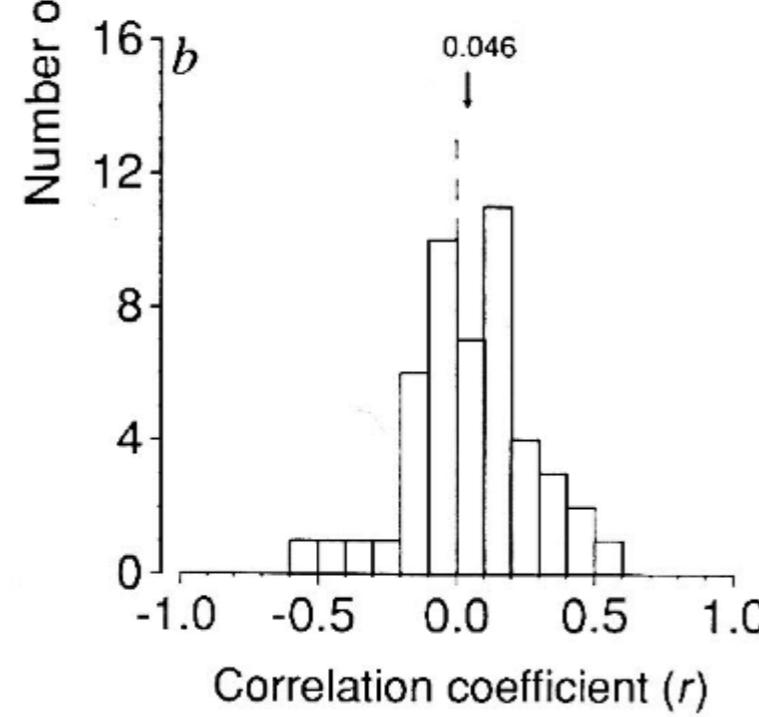


Correlation: a fundamental way to study interactions

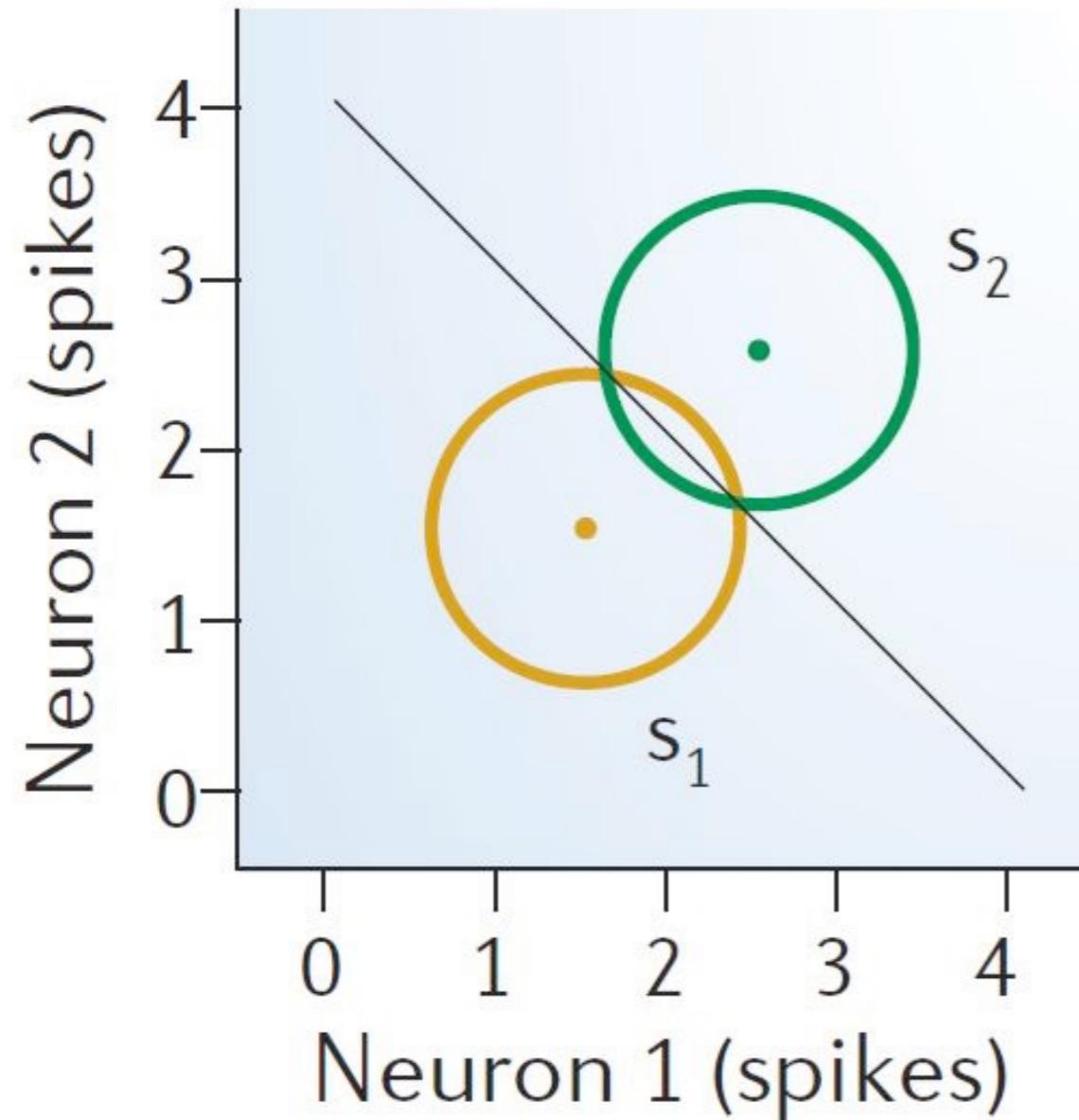
similar direction preference



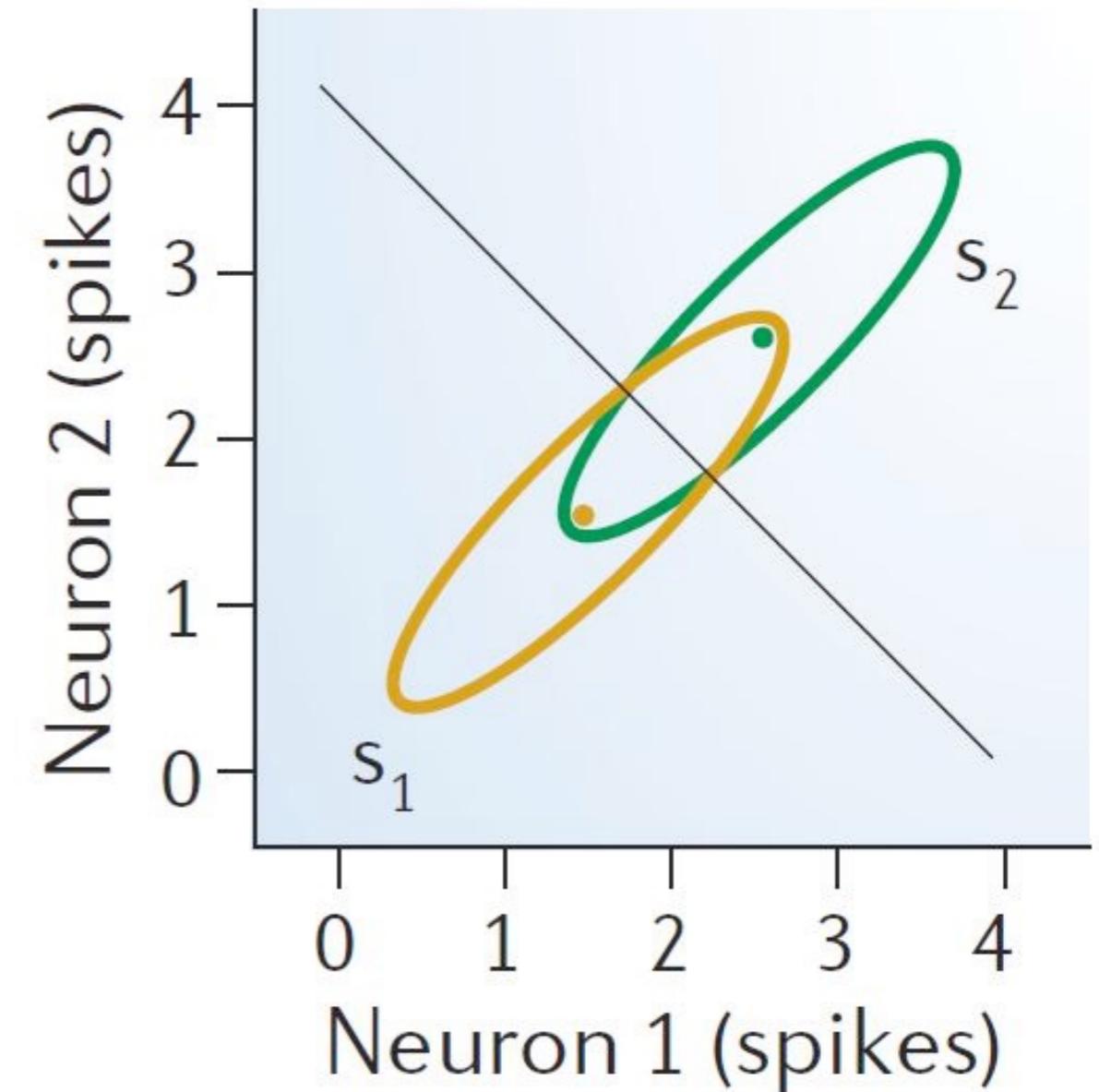
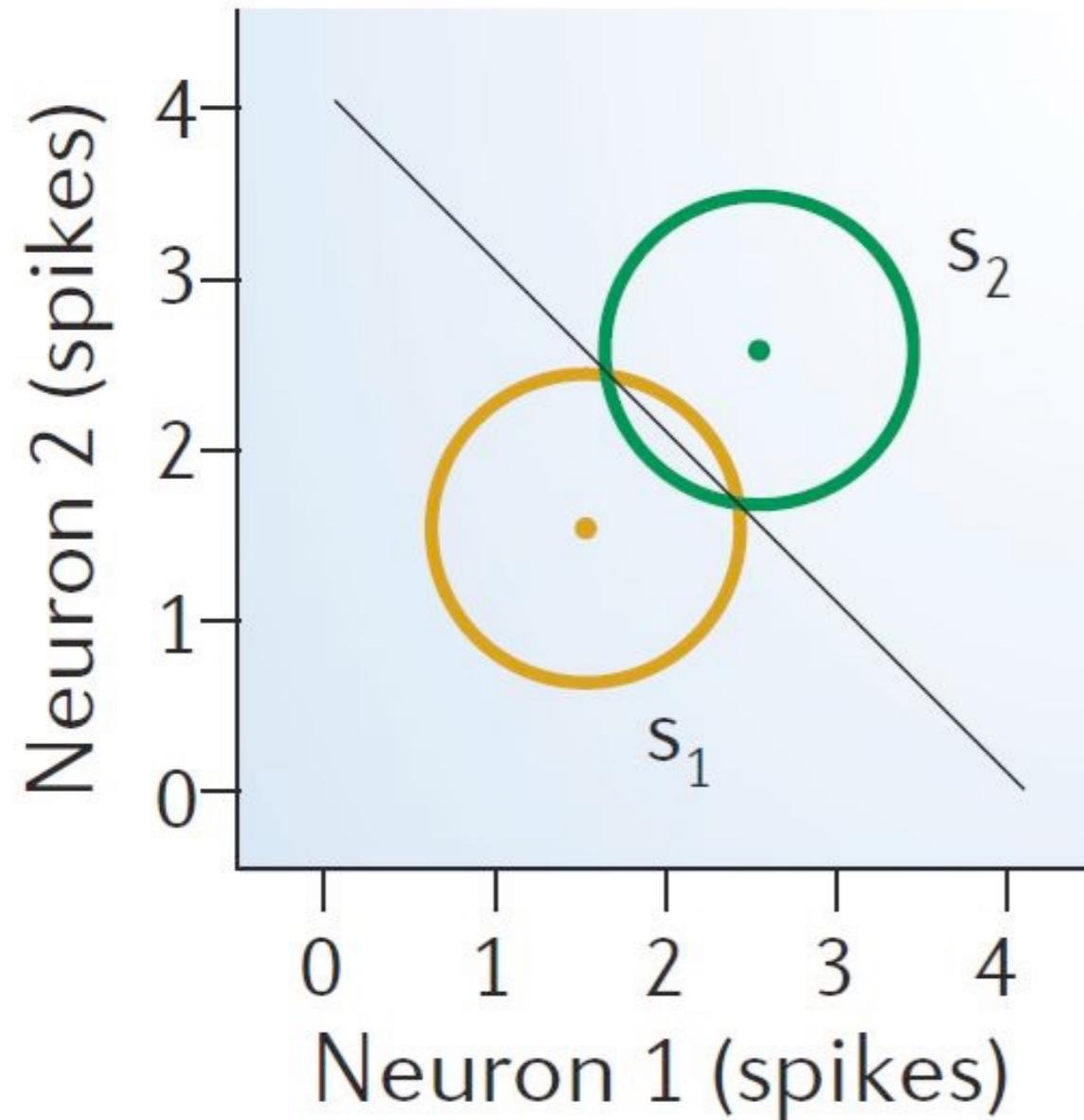
different direction preference



Correlated noise affects information capacity

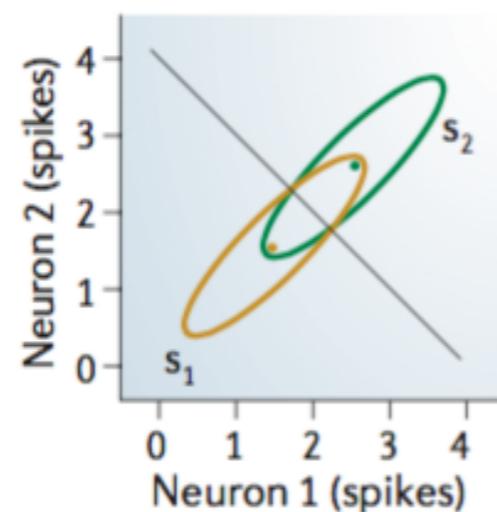


Correlated noise affects information capacity



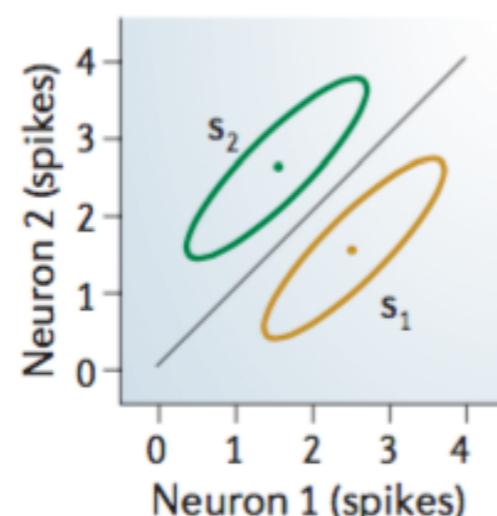
**Information (I) in
unshuffled responses**

a $\Delta I_{\text{shuffled}} < 0$

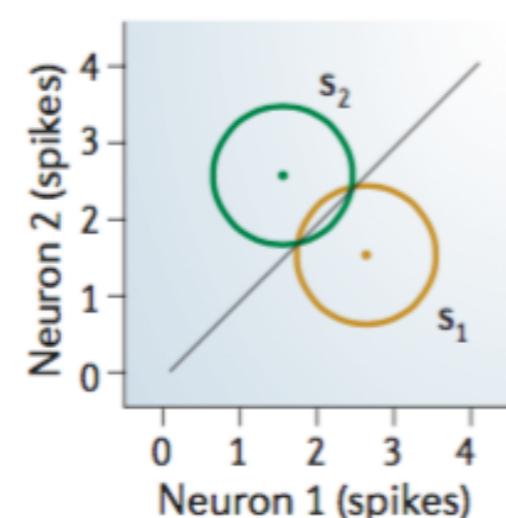
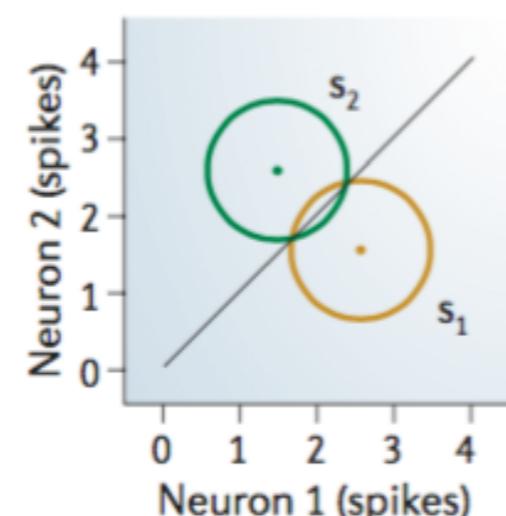
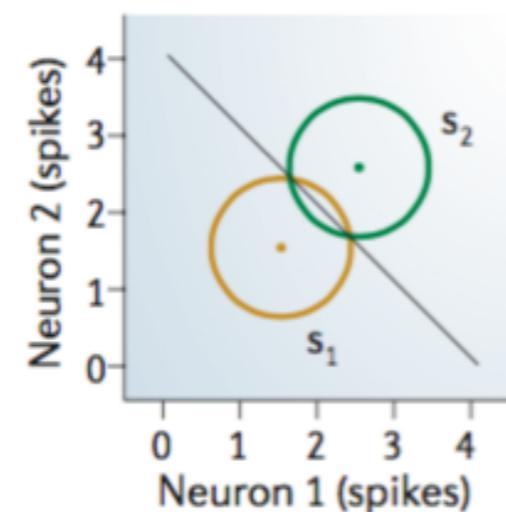
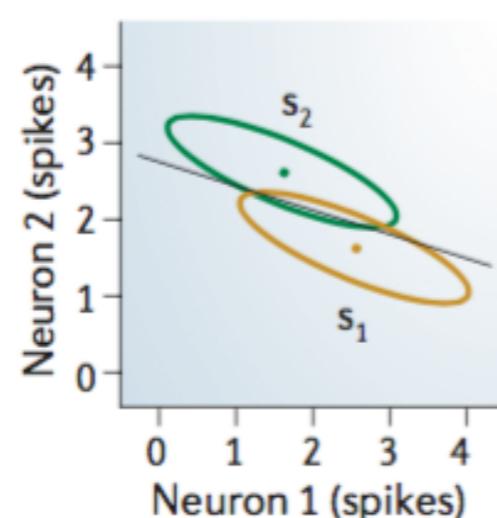


**Information (I_{shuffled})
in shuffled responses**

b $\Delta I_{\text{shuffled}} > 0$



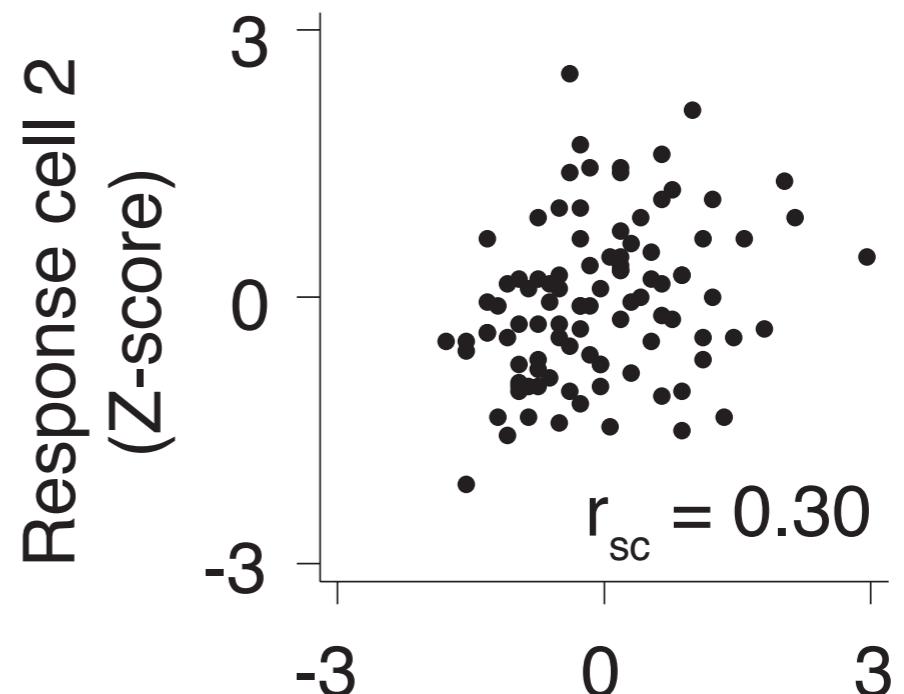
c $\Delta I_{\text{shuffled}} = 0$



Other uses of correlation measures

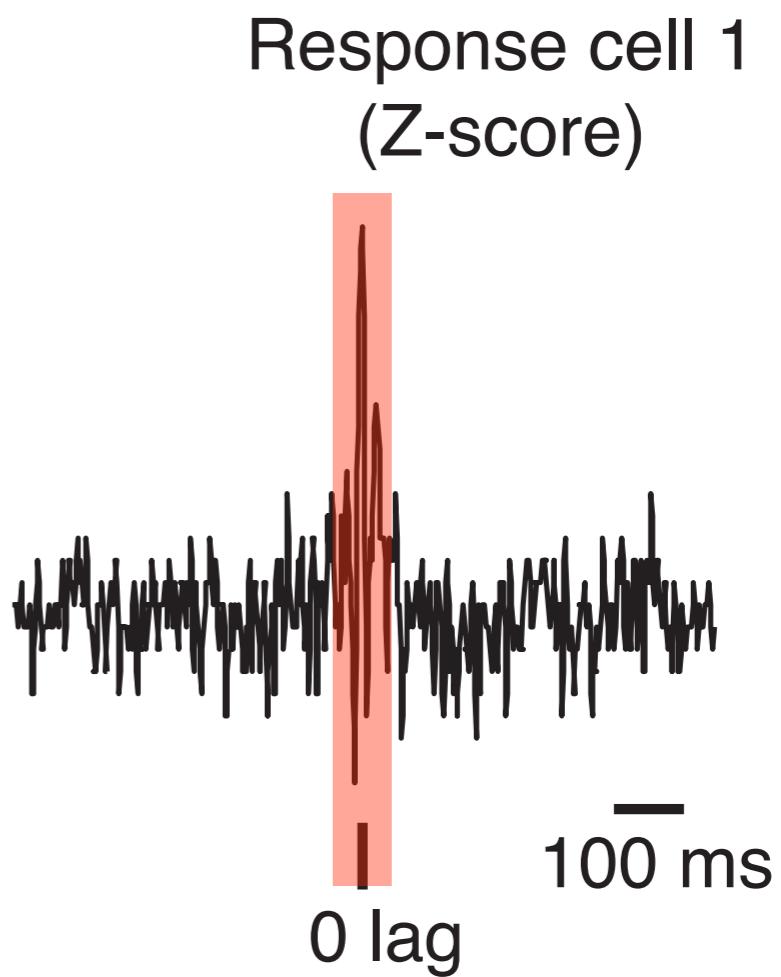
Correlation of trial-to-trial response variability (a.k.a. “noise correlation”, spike count correlation, r_{sc})

Slow time scale



Synchronous spiking
(CCG or cross-correlogram)

Fast time scale

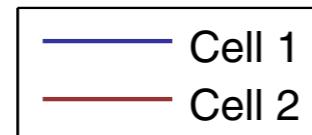


Spike timing correlation (synchrony)

Fast time scale correlation

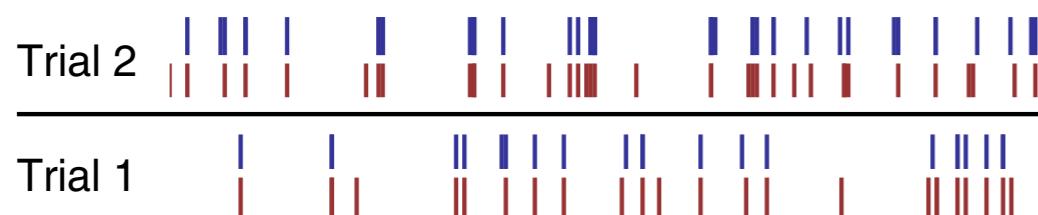
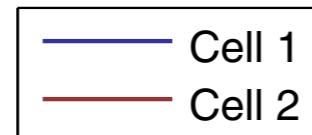
Spike timing correlation (synchrony)

Fast time scale correlation



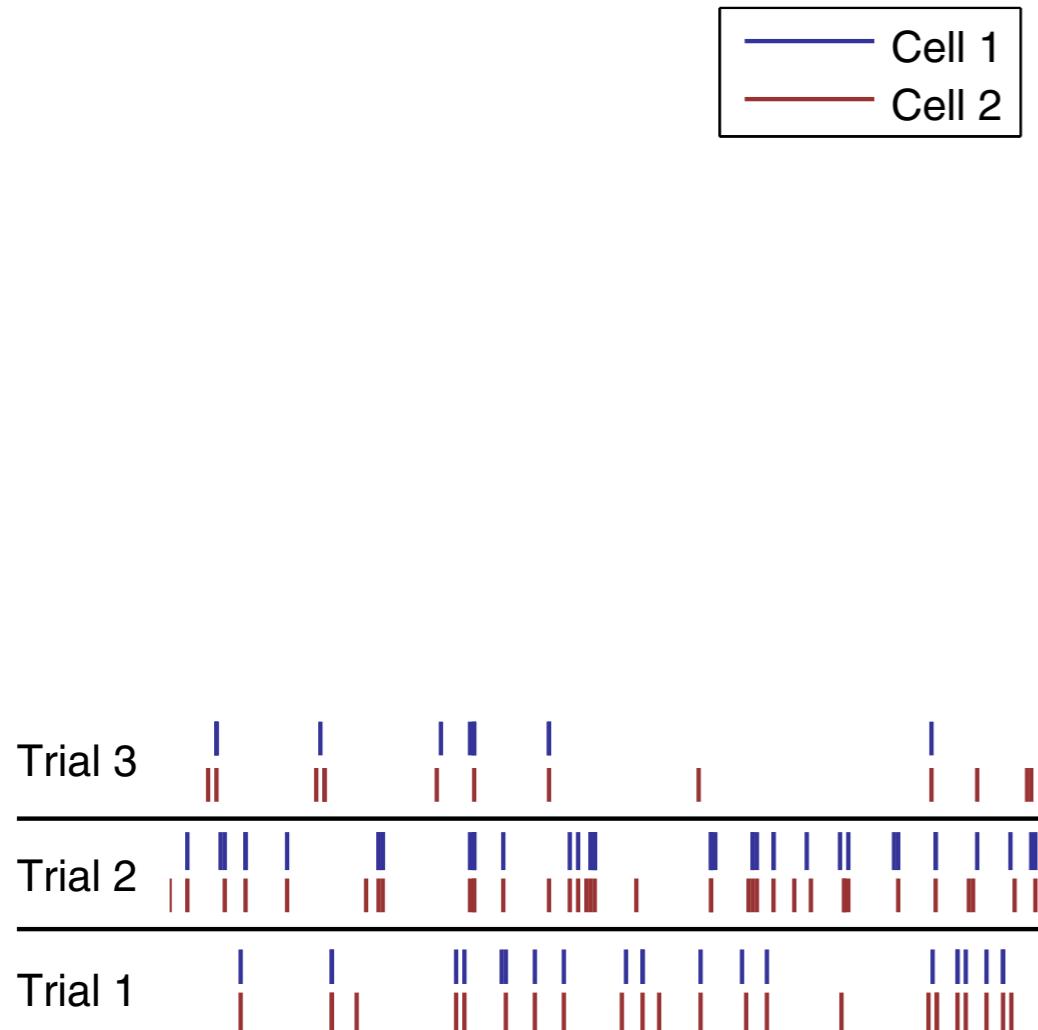
Spike timing correlation (synchrony)

Fast time scale correlation



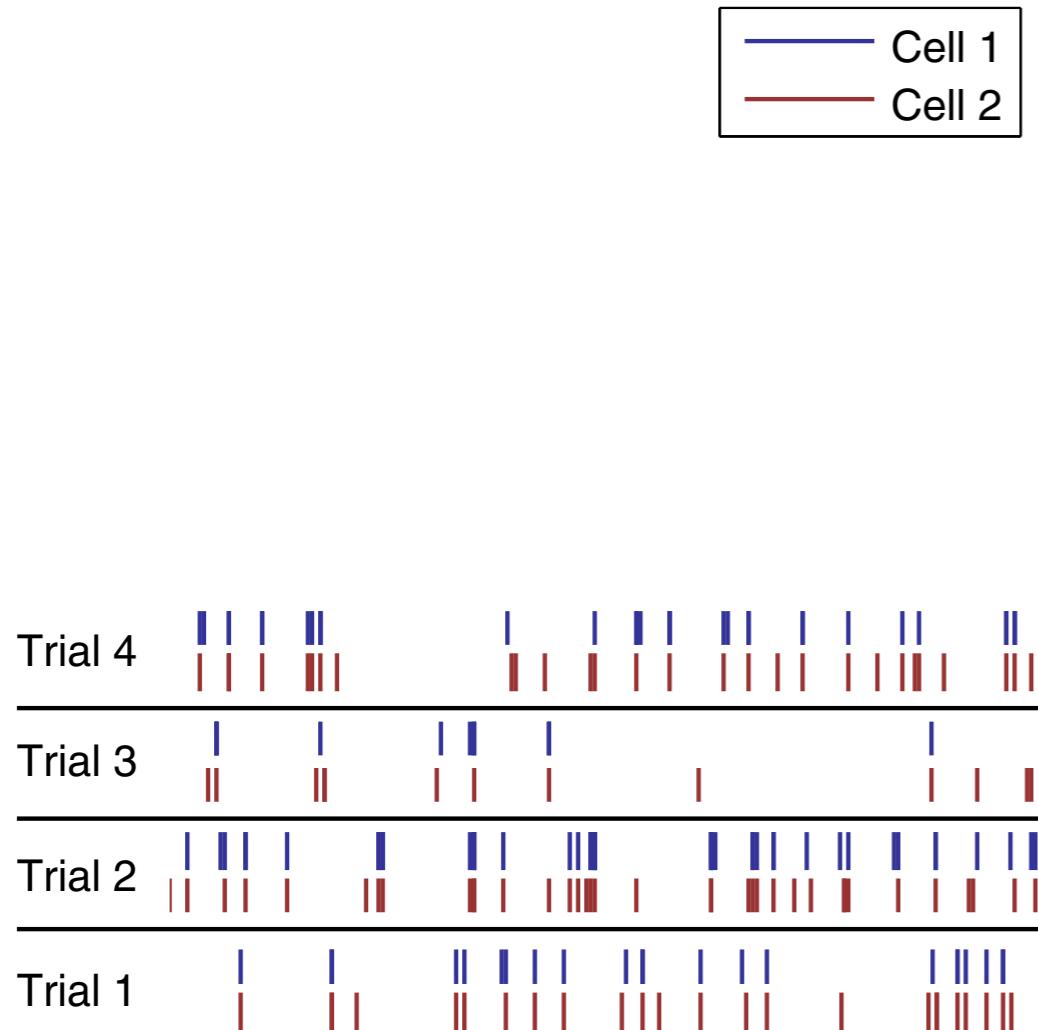
Spike timing correlation (synchrony)

Fast time scale correlation



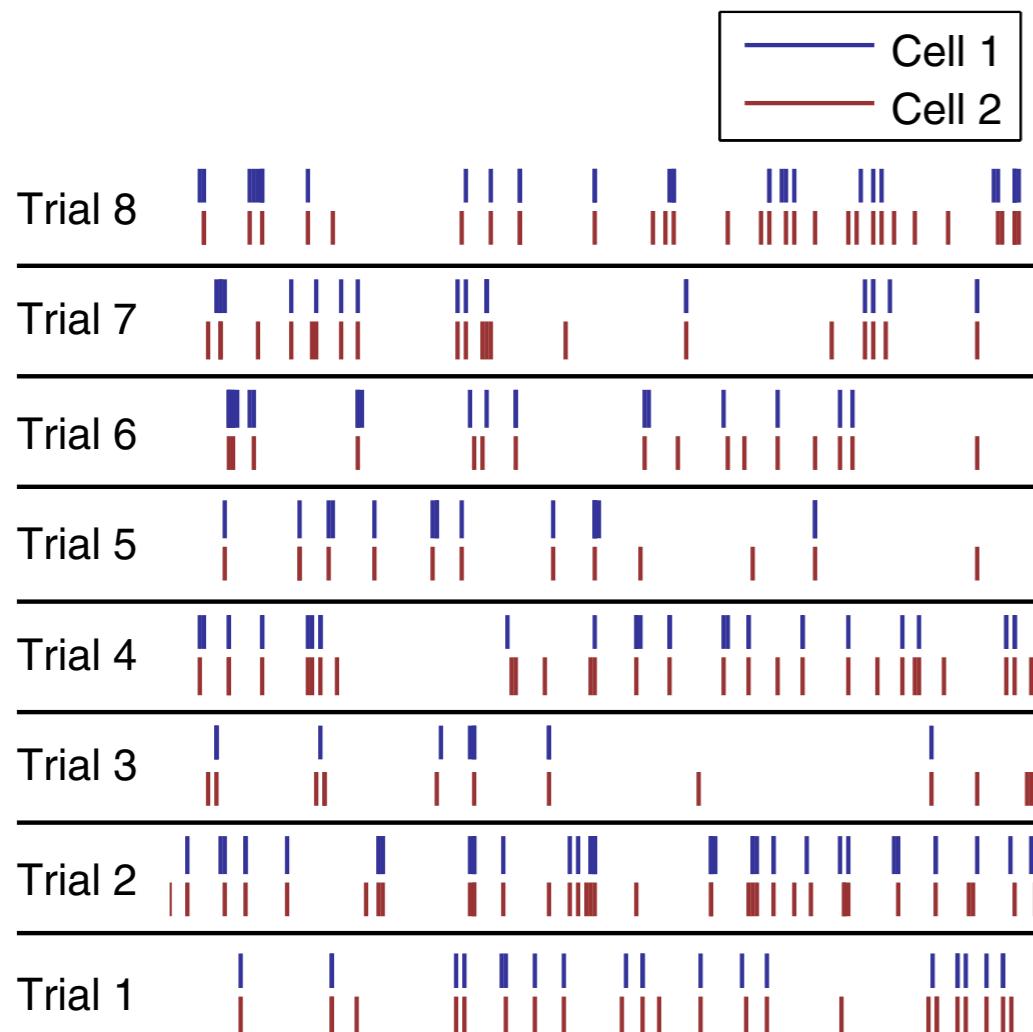
Spike timing correlation (synchrony)

Fast time scale correlation



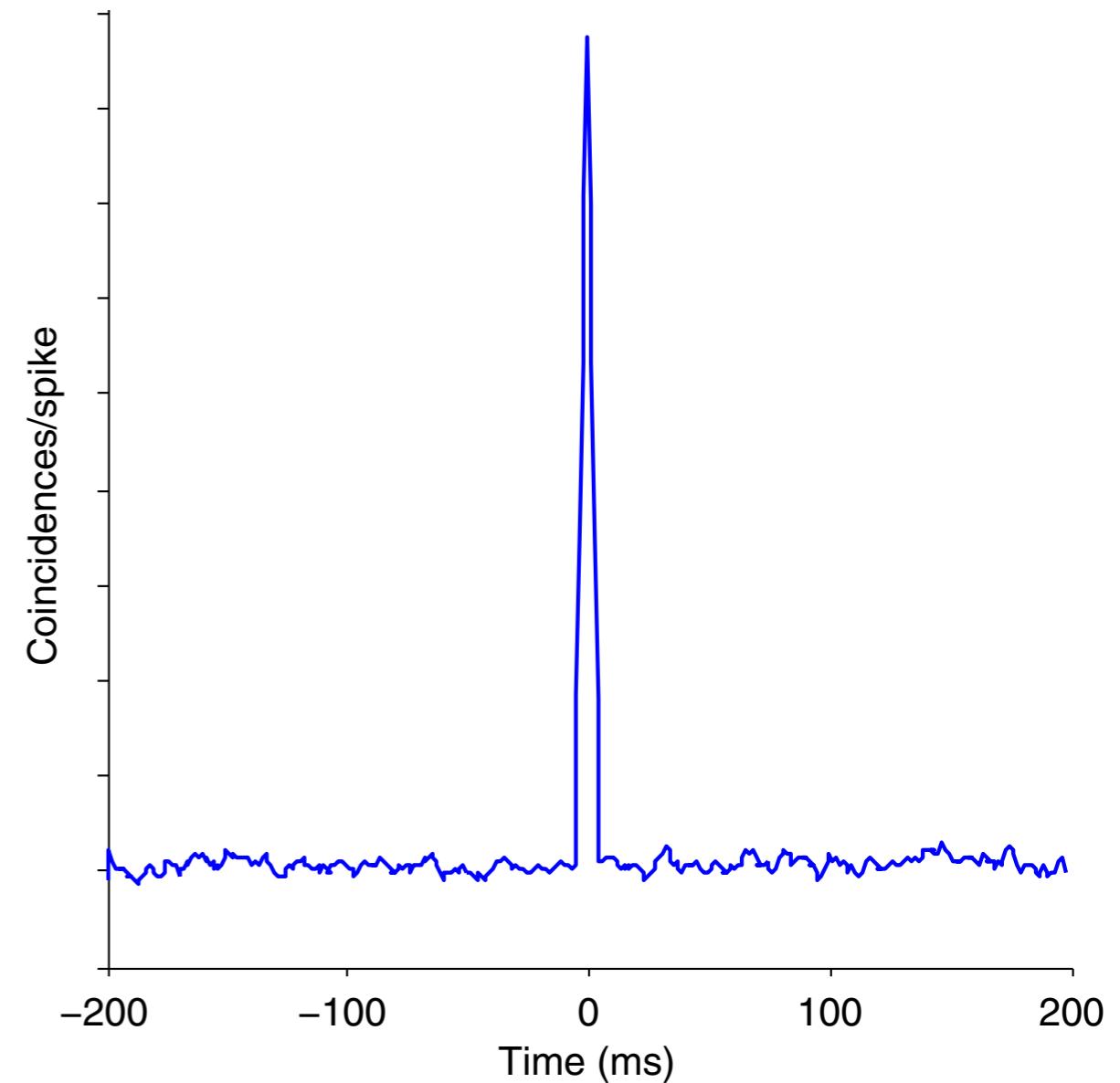
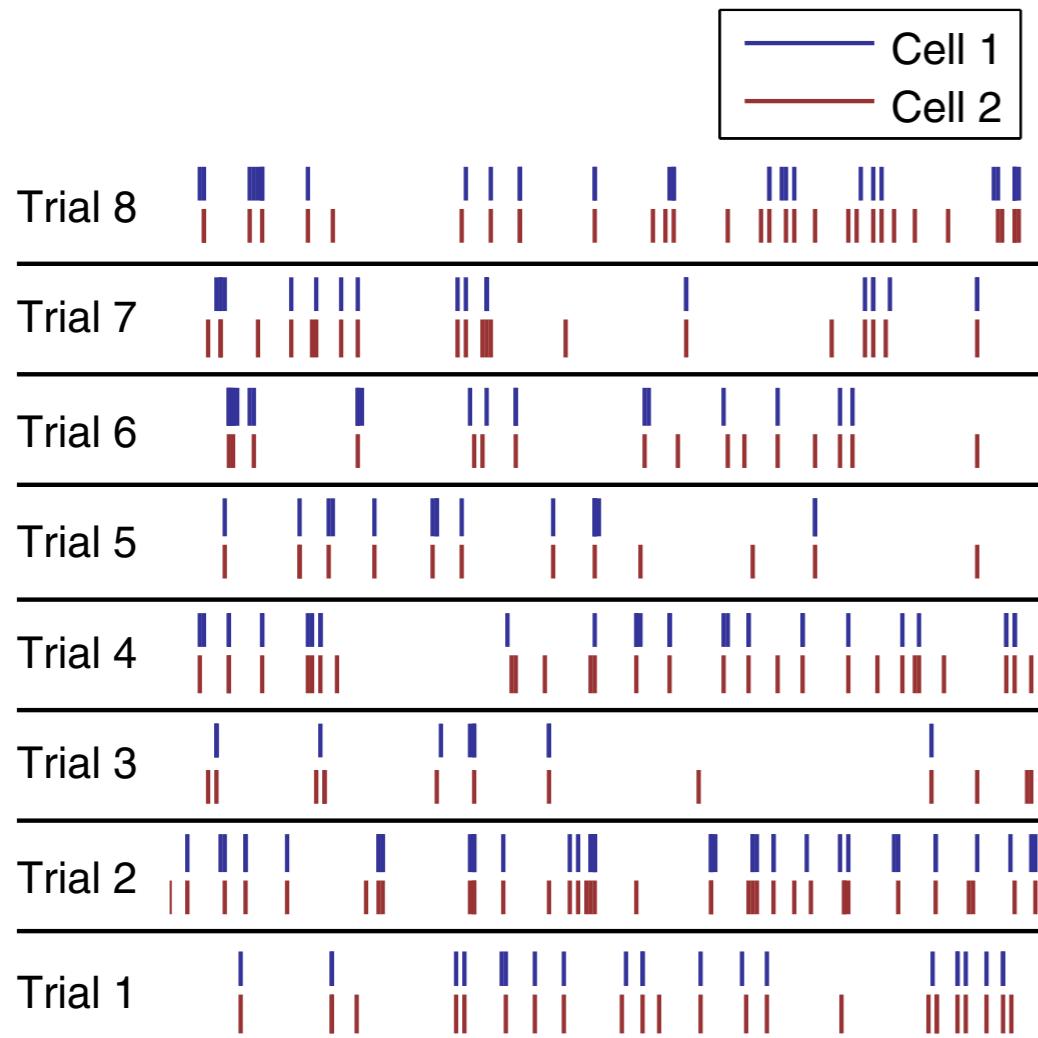
Spike timing correlation (synchrony)

Fast time scale correlation



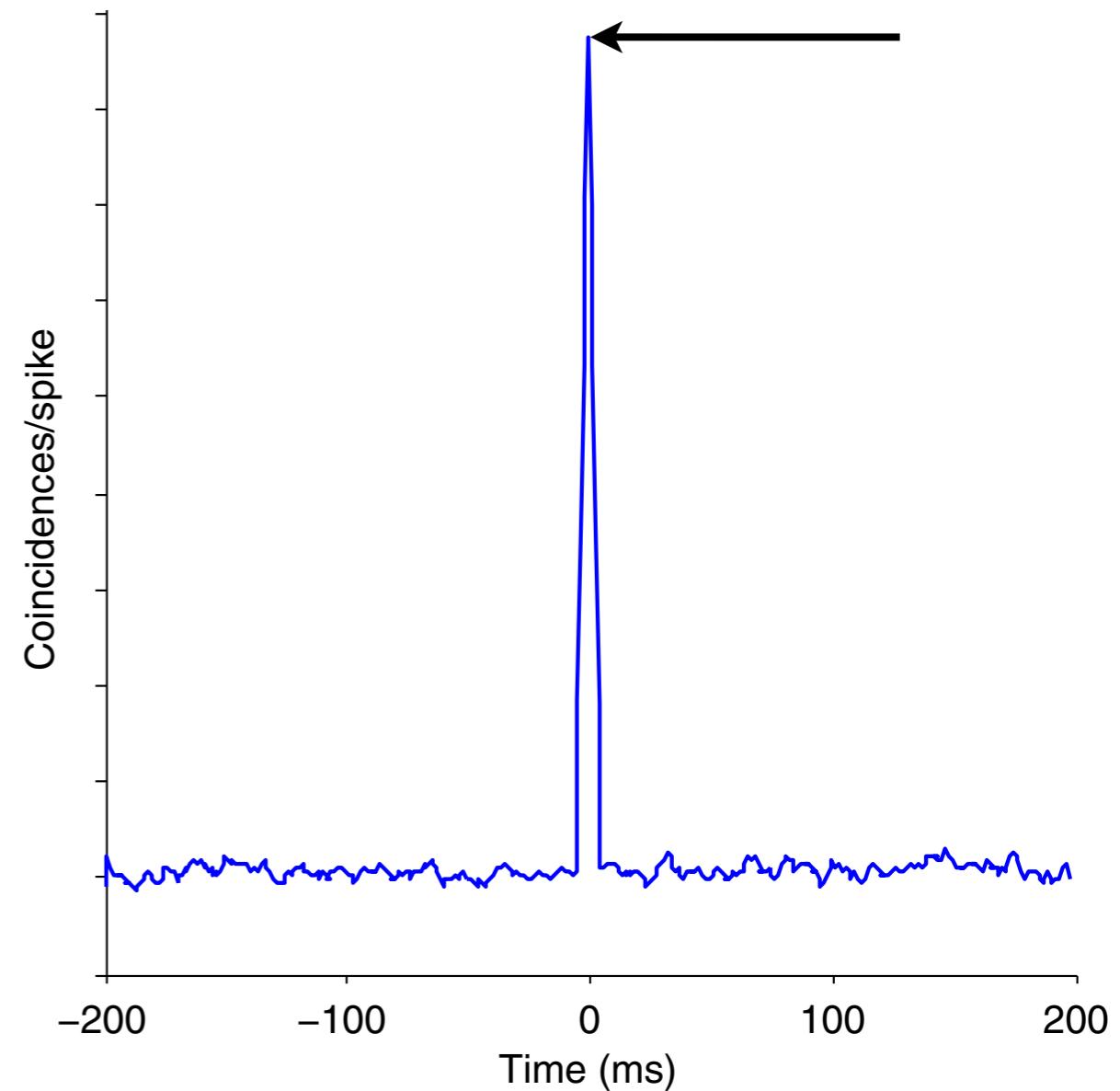
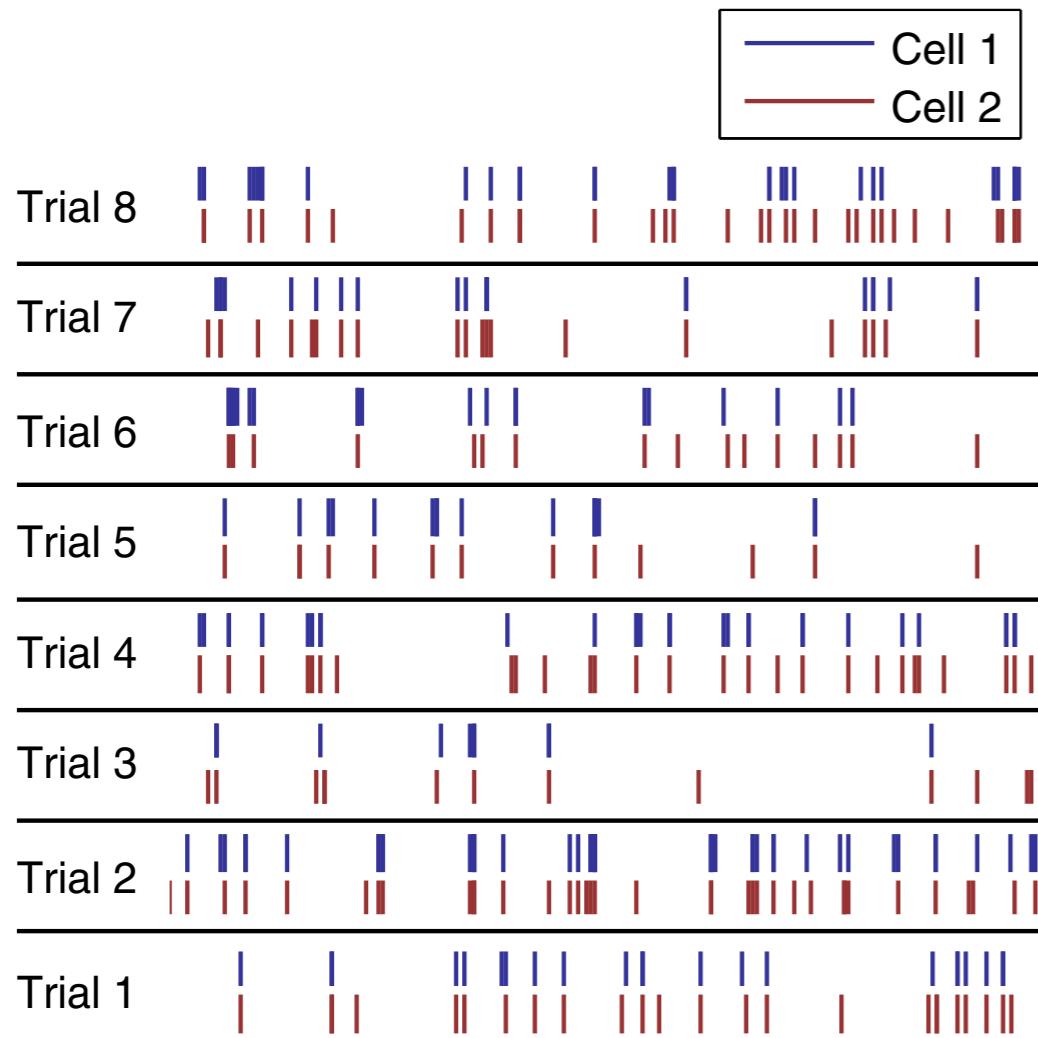
Spike timing correlation (synchrony)

Fast time scale correlation



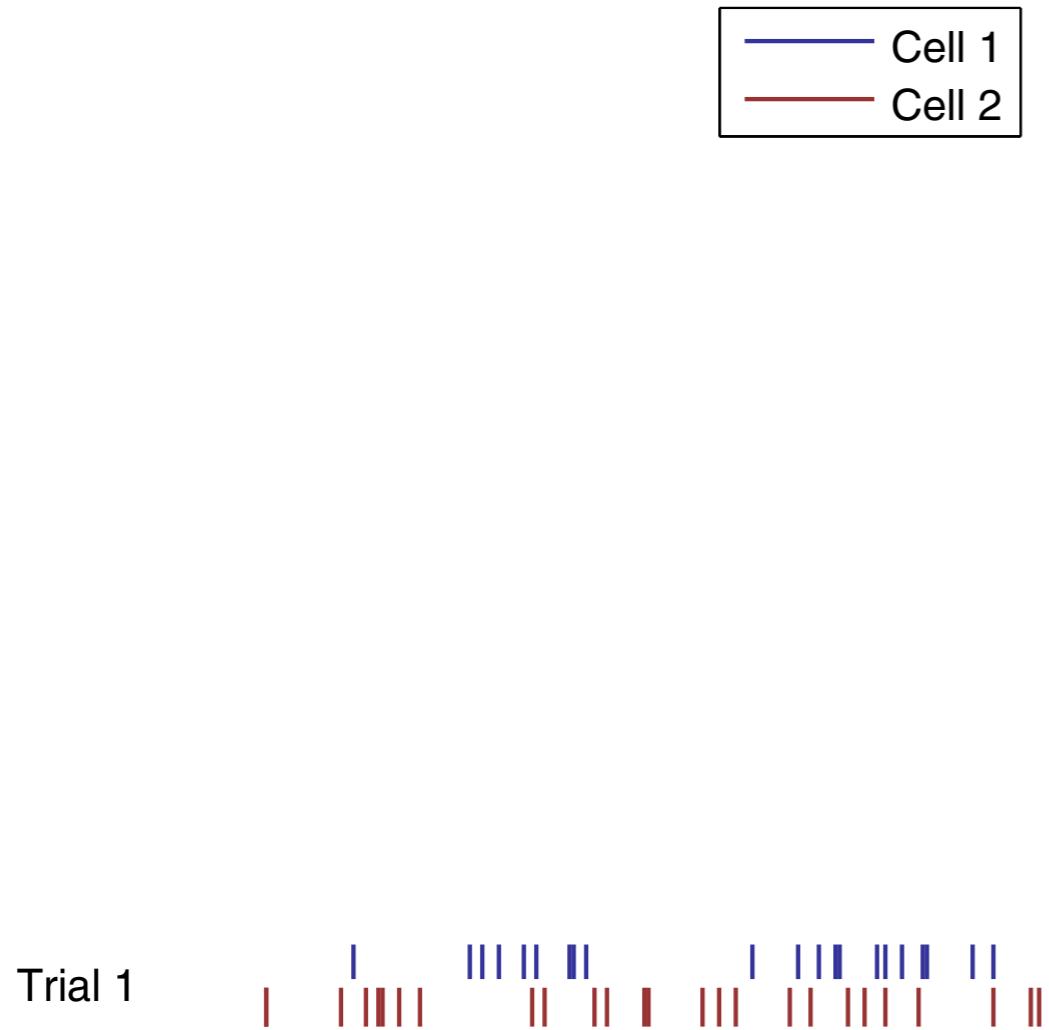
Spike timing correlation (synchrony)

Fast time scale correlation

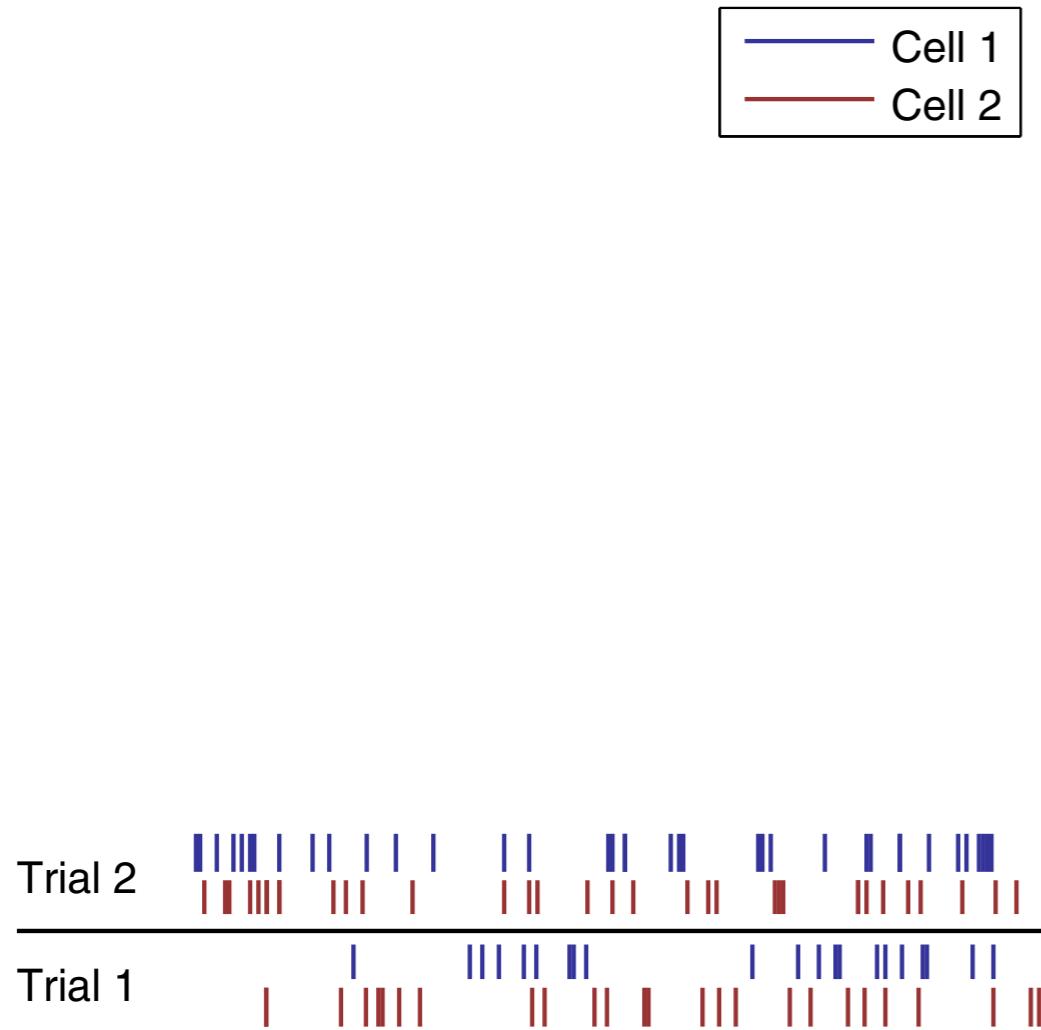


Spike timing correlation (synchrony)

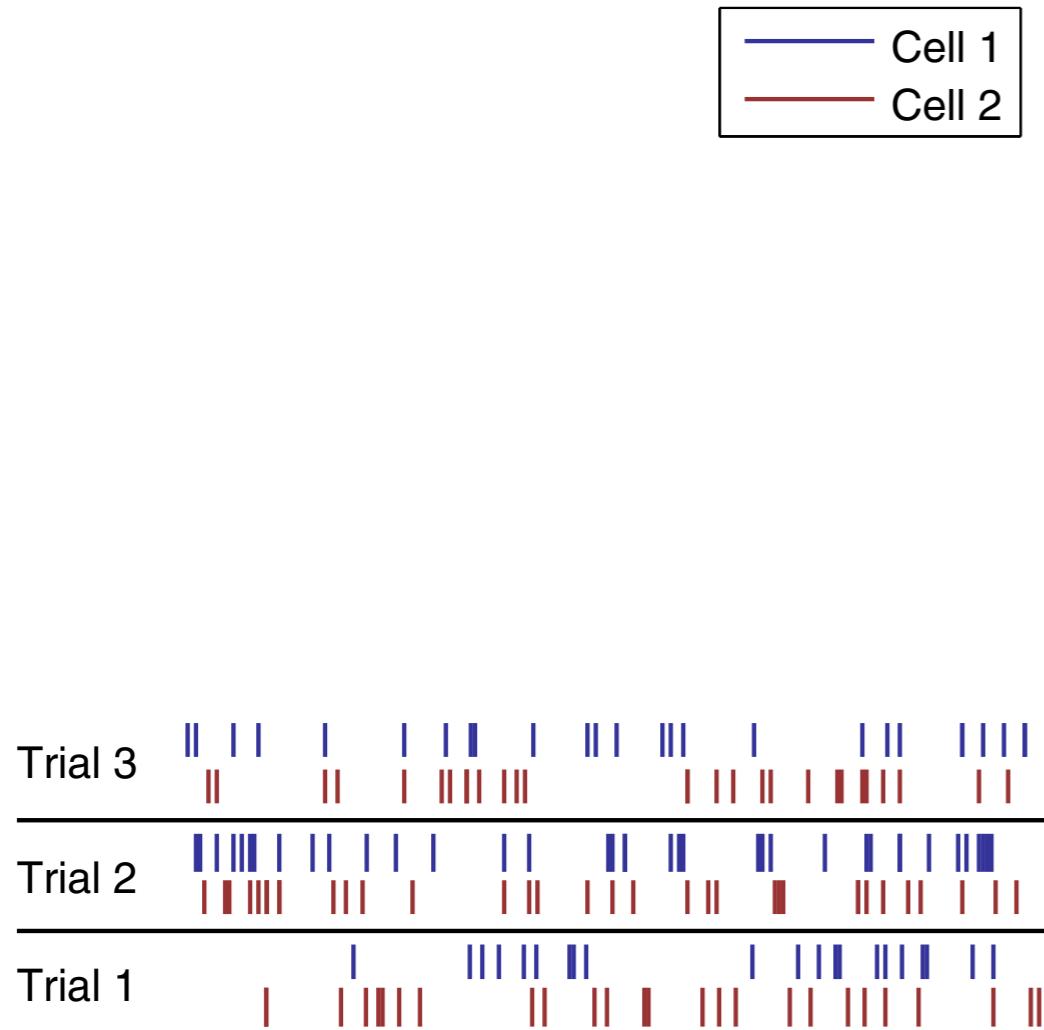
Spike timing correlation (synchrony)



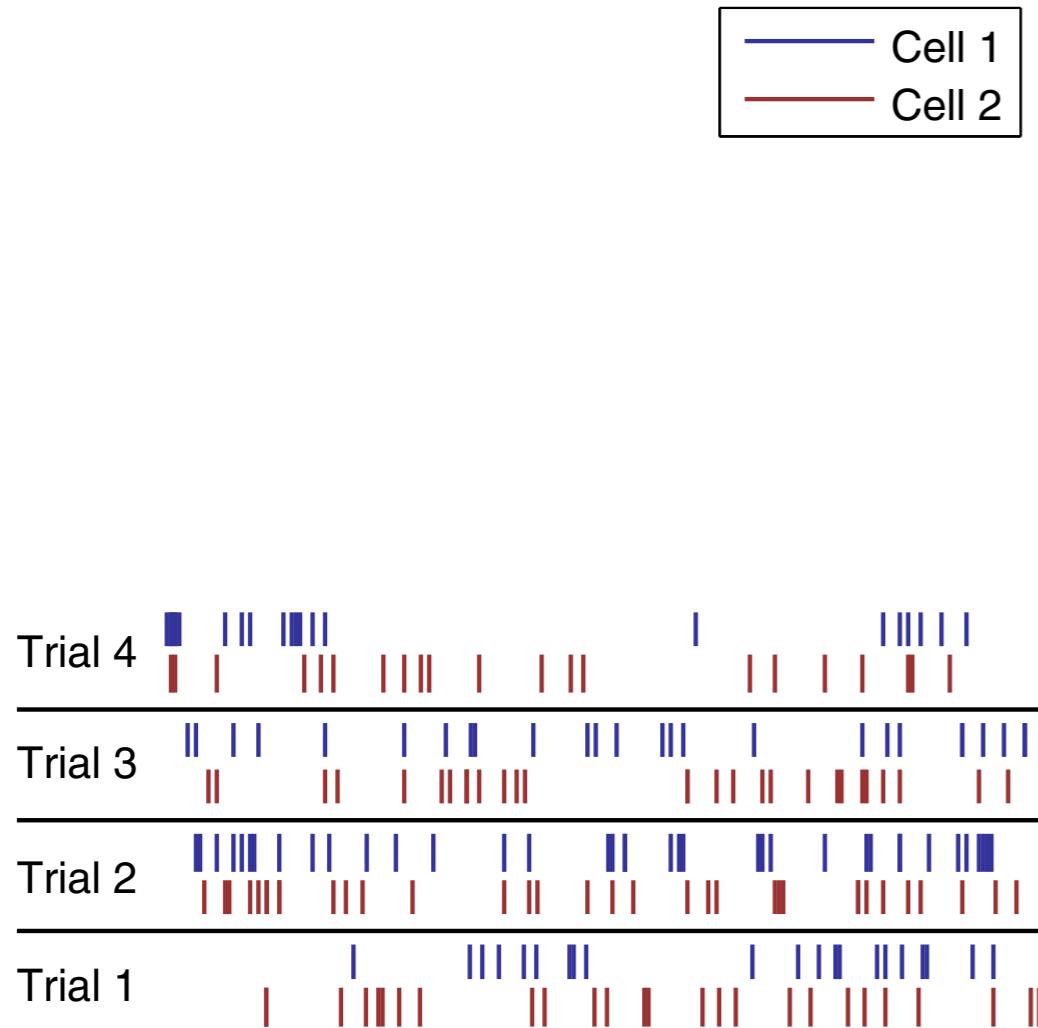
Spike timing correlation (synchrony)



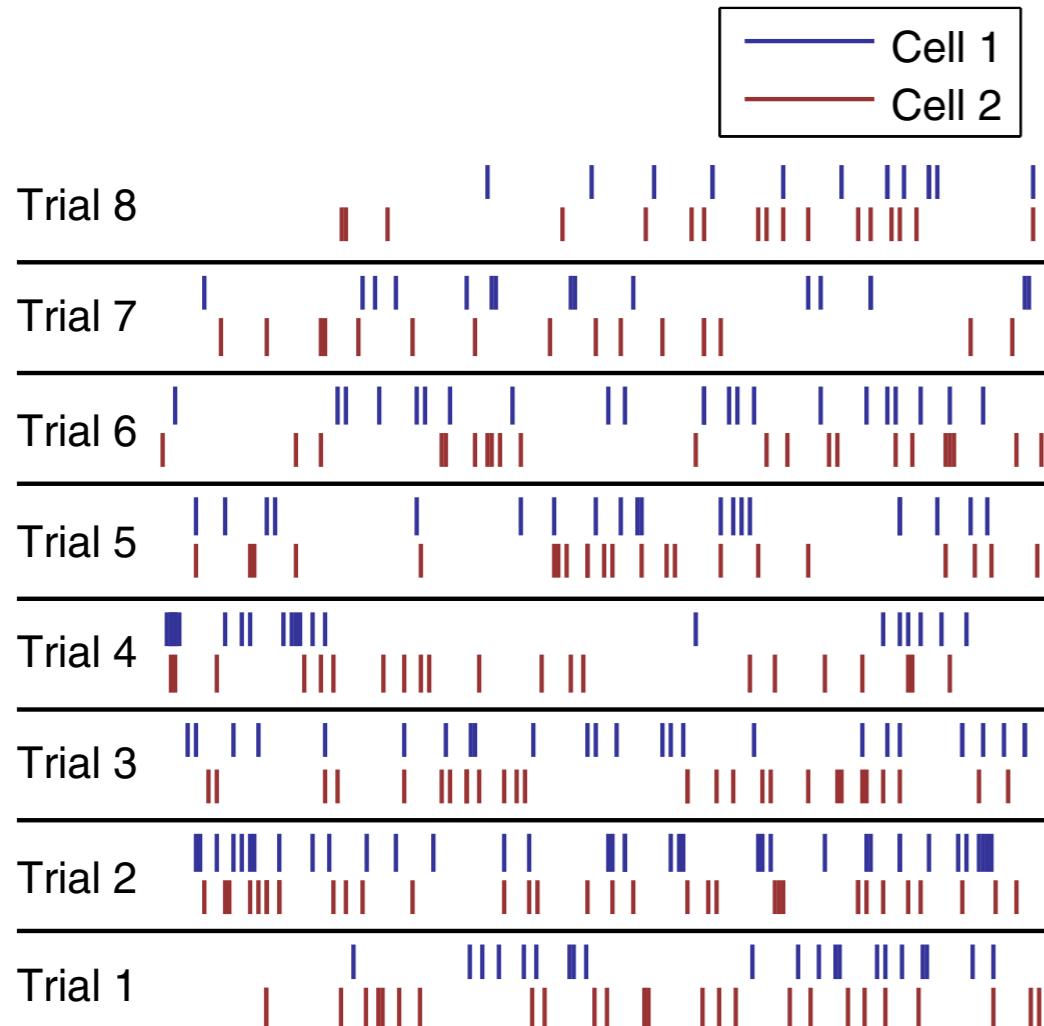
Spike timing correlation (synchrony)



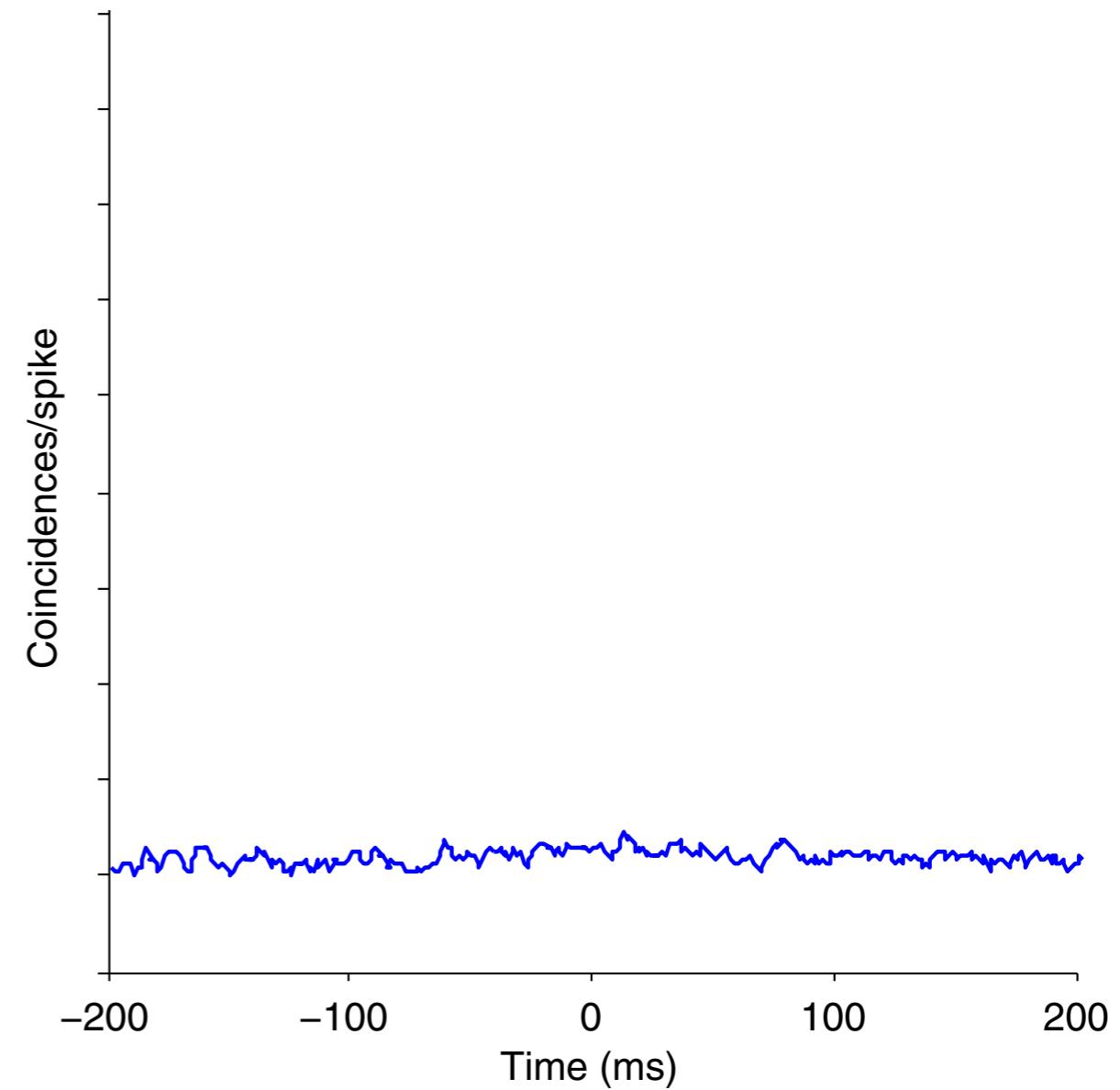
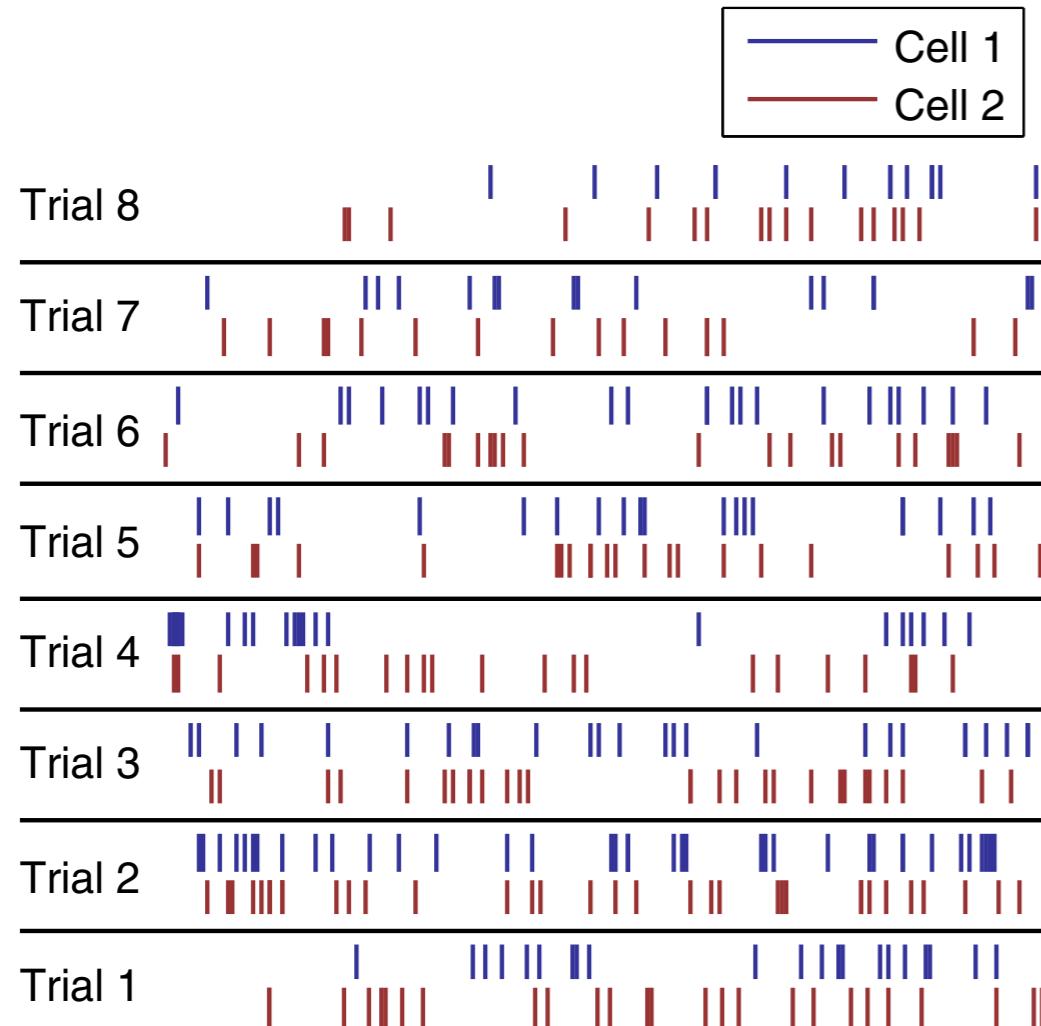
Spike timing correlation (synchrony)



Spike timing correlation (synchrony)



Spike timing correlation (synchrony)

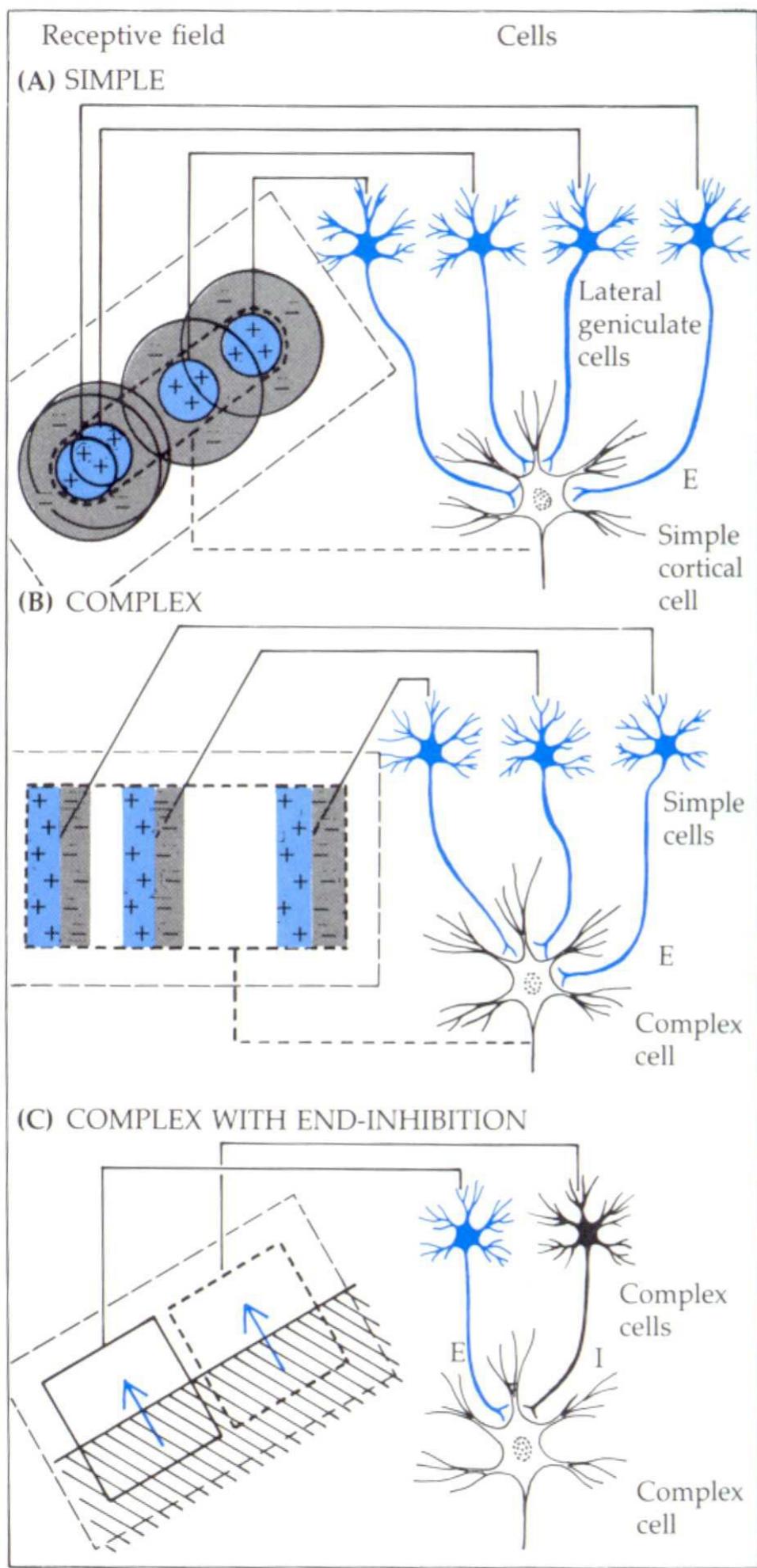


Cross-correlogram (CCG)

$$CCG(\tau) = \frac{\frac{1}{M} \sum_{i=1}^M \sum_{t=1}^N x_1^i(t)x_2^i(t + \tau)}{\theta(\tau) \sqrt{\lambda_1 \lambda_2}},$$

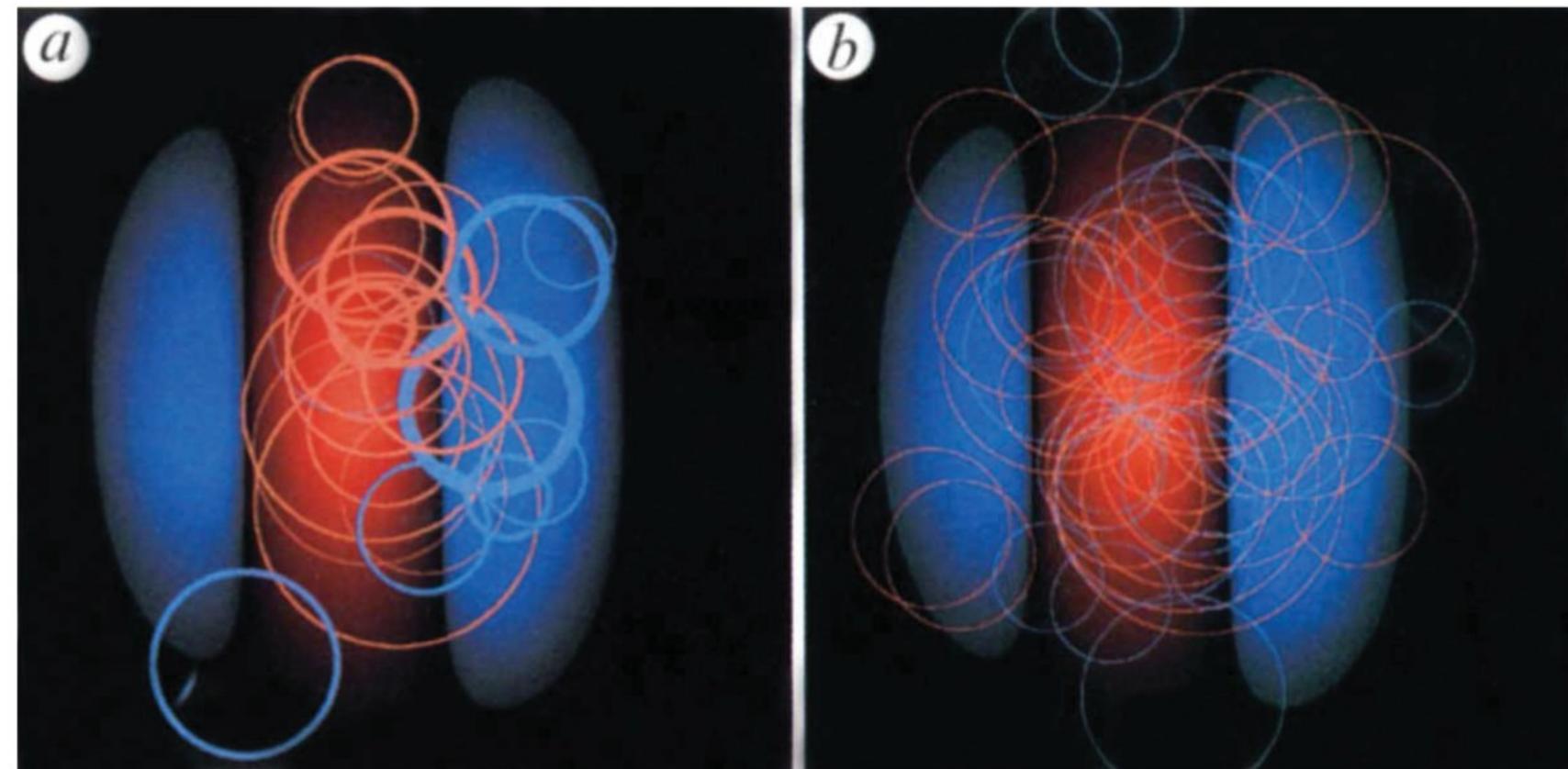
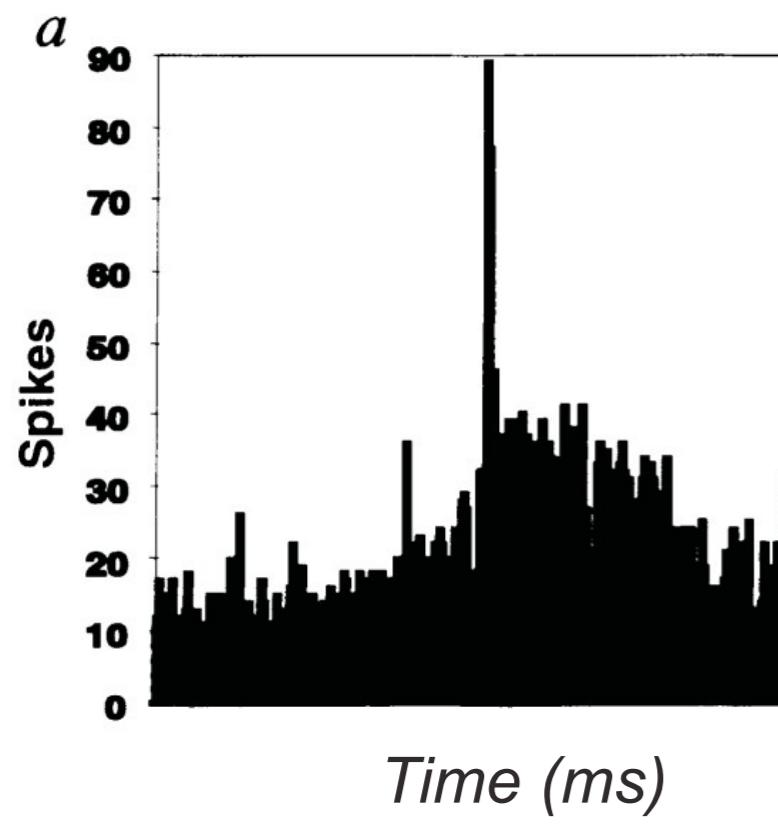
Diagram illustrating the components of the Cross-correlogram (CCG) formula:

- trials**: Points to the term $\frac{1}{M} \sum_{i=1}^M$.
- bins**: Points to the term $\sum_{t=1}^N$.
- spike trains**: Points to the term $x_1^i(t)x_2^i(t + \tau)$.
- triangular function**: Points to the term $\theta(\tau)$.
- mean f.r.**: Points to the term $\sqrt{\lambda_1 \lambda_2}$.
- time lag**: Points to the term $t + \tau$ in the spike train product.



Using the cross-correlogram to infer circuits

Using the cross-correlogram to infer circuits



Reid and Alonso (1995) Nature 378: 281

Correcting the cross-correlogram

raw CCG - surrogate data
=
corrected CCG

Correcting the cross-correlogram

raw CCG - surrogate data

=

corrected CCG

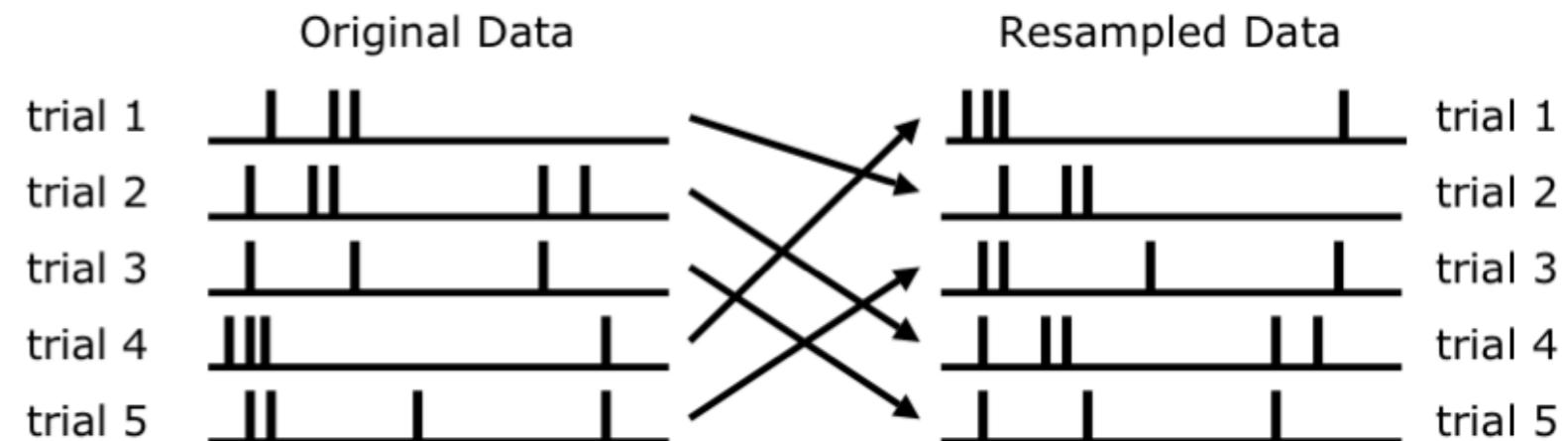
surrogates=

shuffle, shift, jitter, ?

How do we isolate fast correlation?

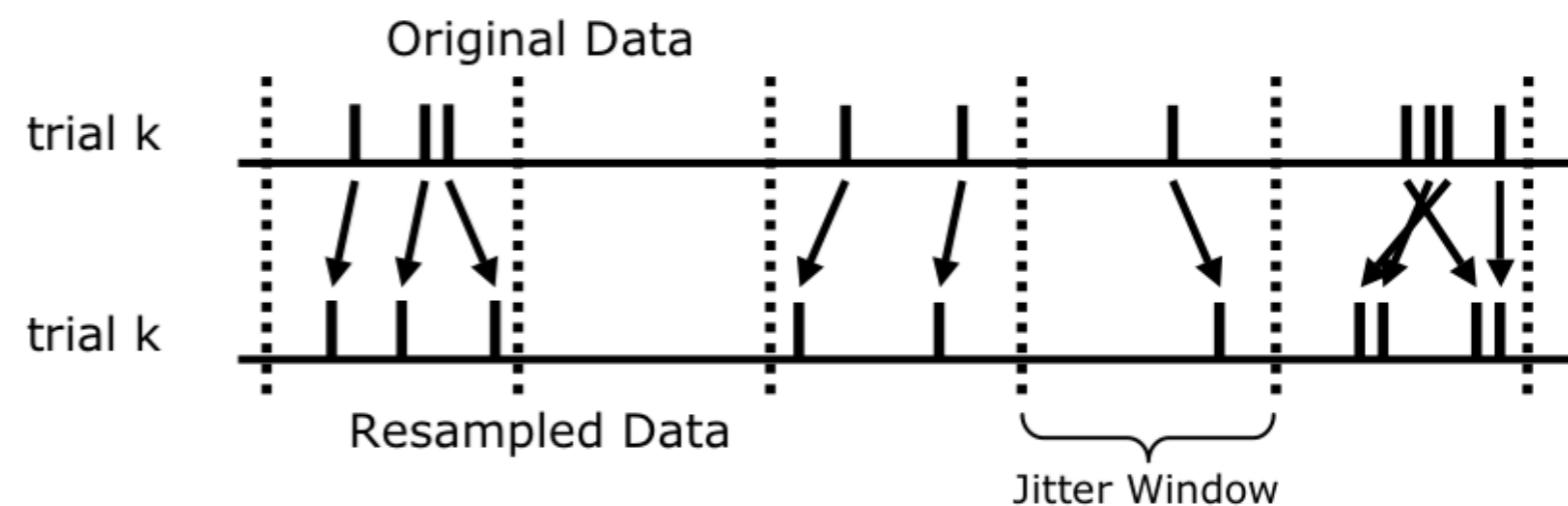
Removes event-locked correlation

Trial Shuffling

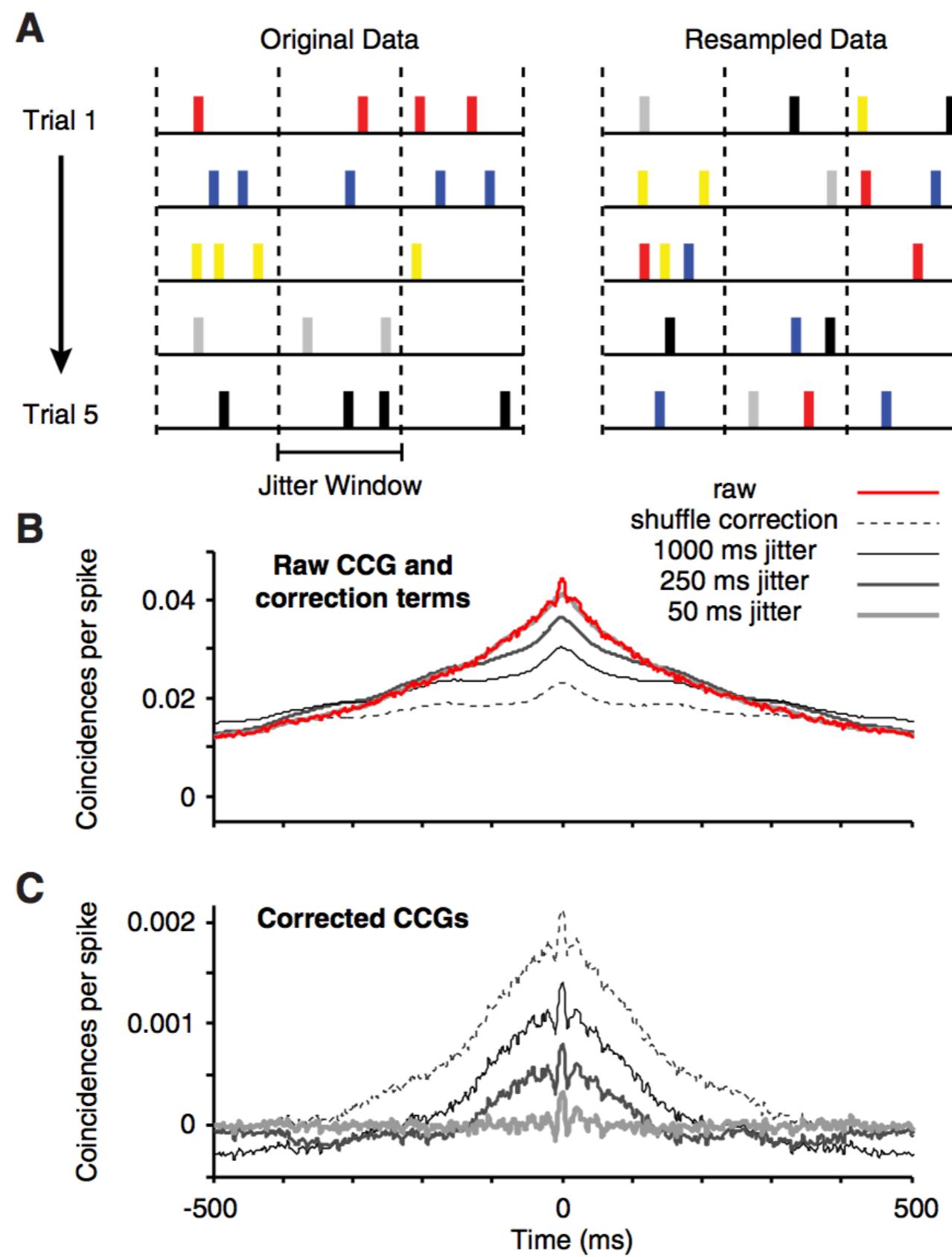


Removes all correlation on time scale < jitter window

Spike Jitter



Correcting the cross-correlogram



CCGs can allow us to infer anatomy

Autocorrelogram (ACG)

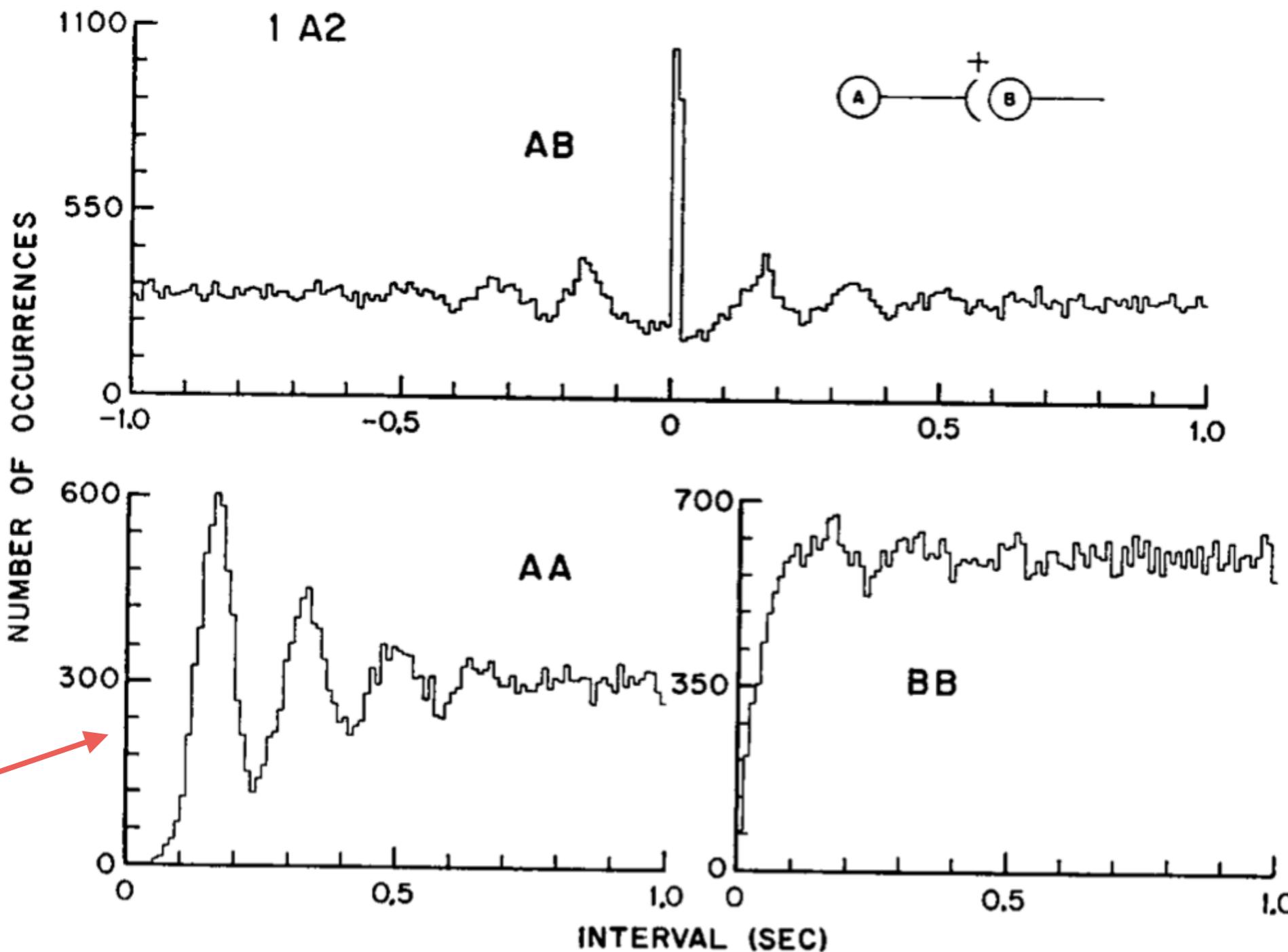


FIGURE 1 Monosynaptic excitation. Periodic presynaptic cell. In this and Figs. 2–6 the *A* cell is always the presynaptic cell, *B* the postsynaptic cell. Top: *AB* cross-correlation. Bottom, left: cell *A* autocorrelation. Right: *B* autocorrelation (run 1A2).

CCGs can allow us to infer anatomy

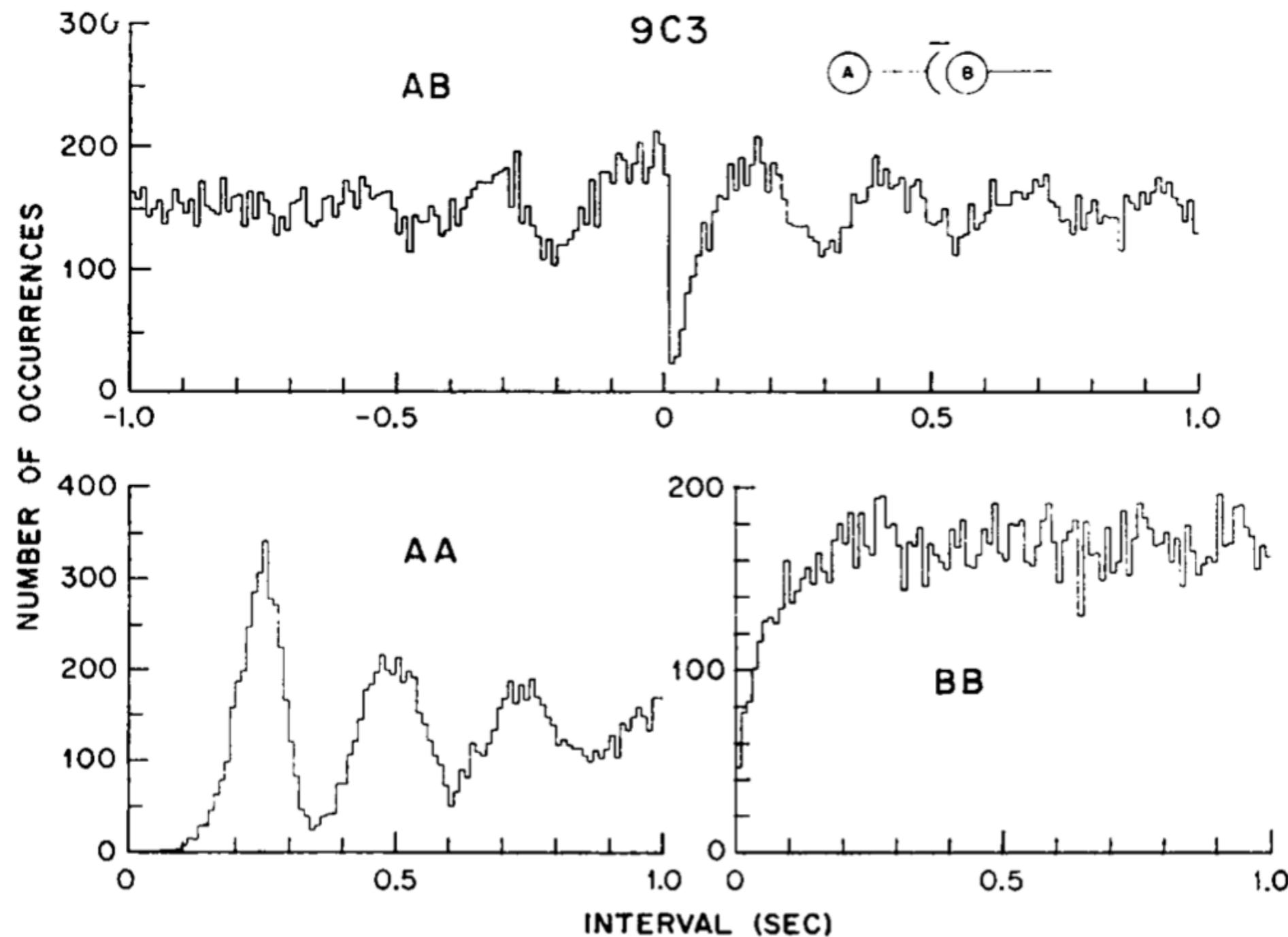


FIGURE 5 Monosynaptic inhibition. Periodically firing presynaptic cell. Top: *AB* cross-correlation. Initial trough is primary sign of monosynaptic inhibition. Bottom: *A* and *B* cell autocorrelations. Note appearance of input cell's periodicity in cross-correlation for positive and negative τ (run 9C3).

... but we're limited in that inference

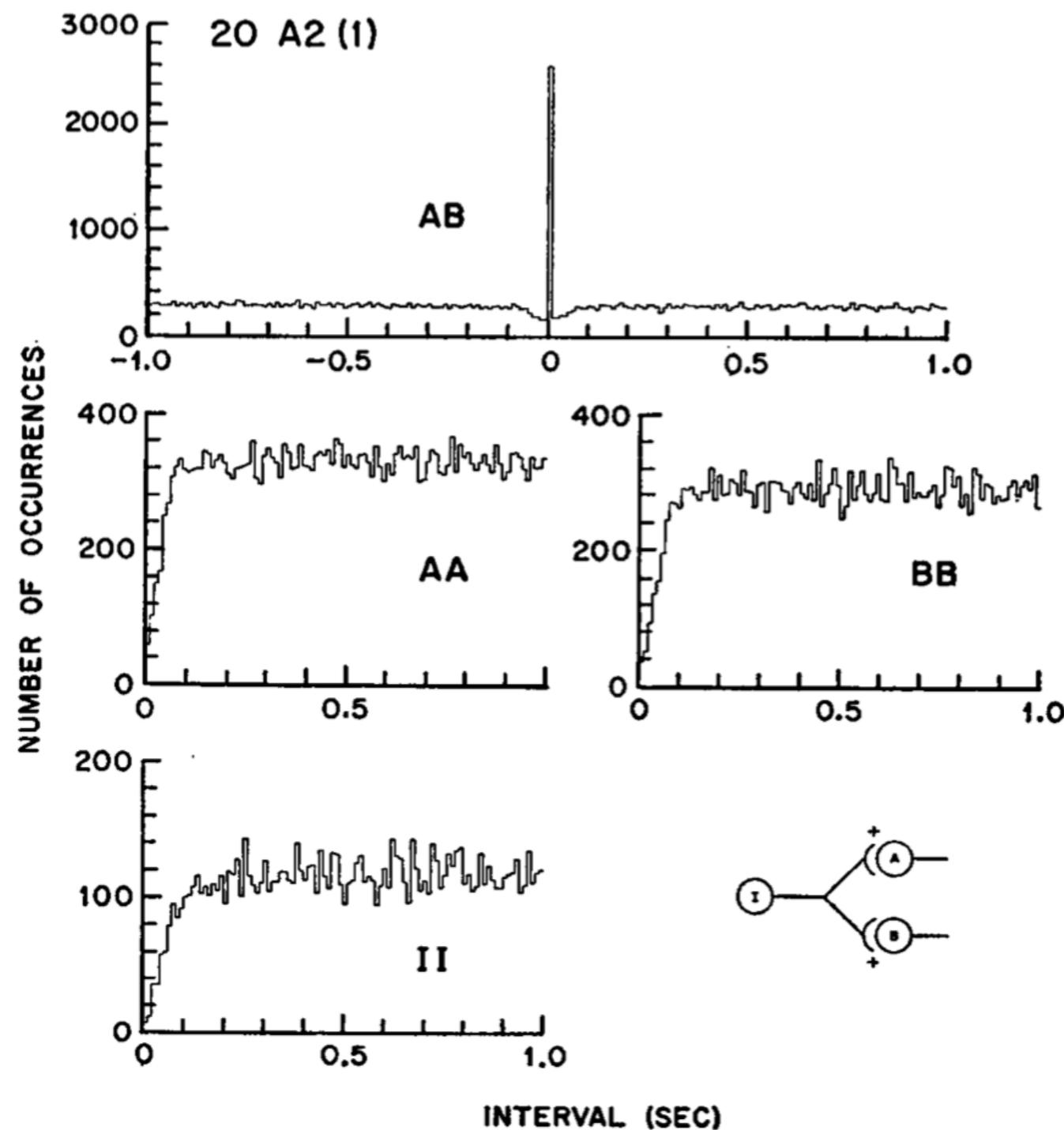


FIGURE 7 Shared excitatory synaptic input. Basic case in which common excitatory source and both postsynaptic cells have flat autocorrelations.

... but we're limited in that inference

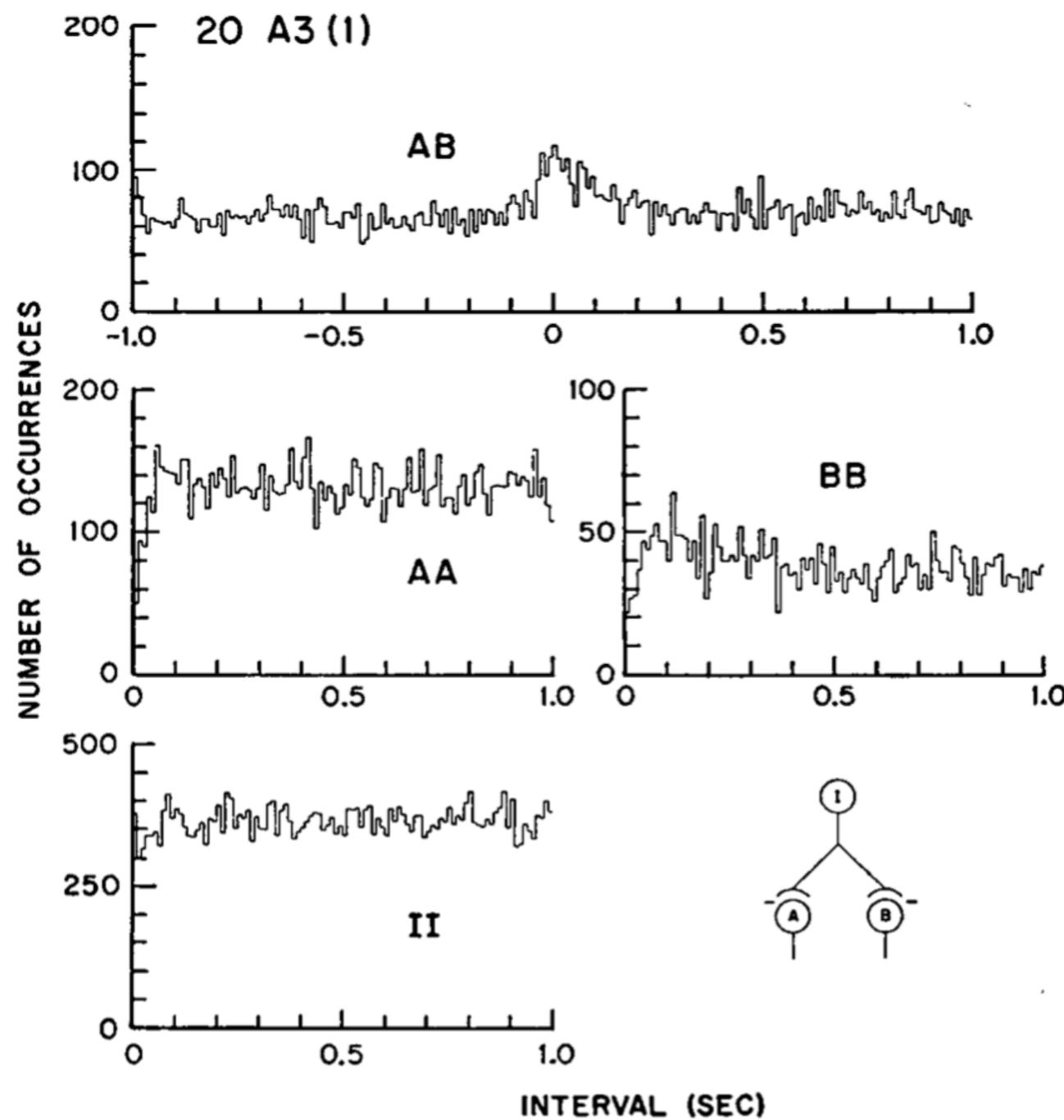


FIGURE 9 Shared inhibitory synaptic input. Basic case of shared inhibition in which all three cells have flat autocorrelation functions. Central broad peak is primary characteristic of the shared inhibition.

The CCG and r_{sc} are linked

