

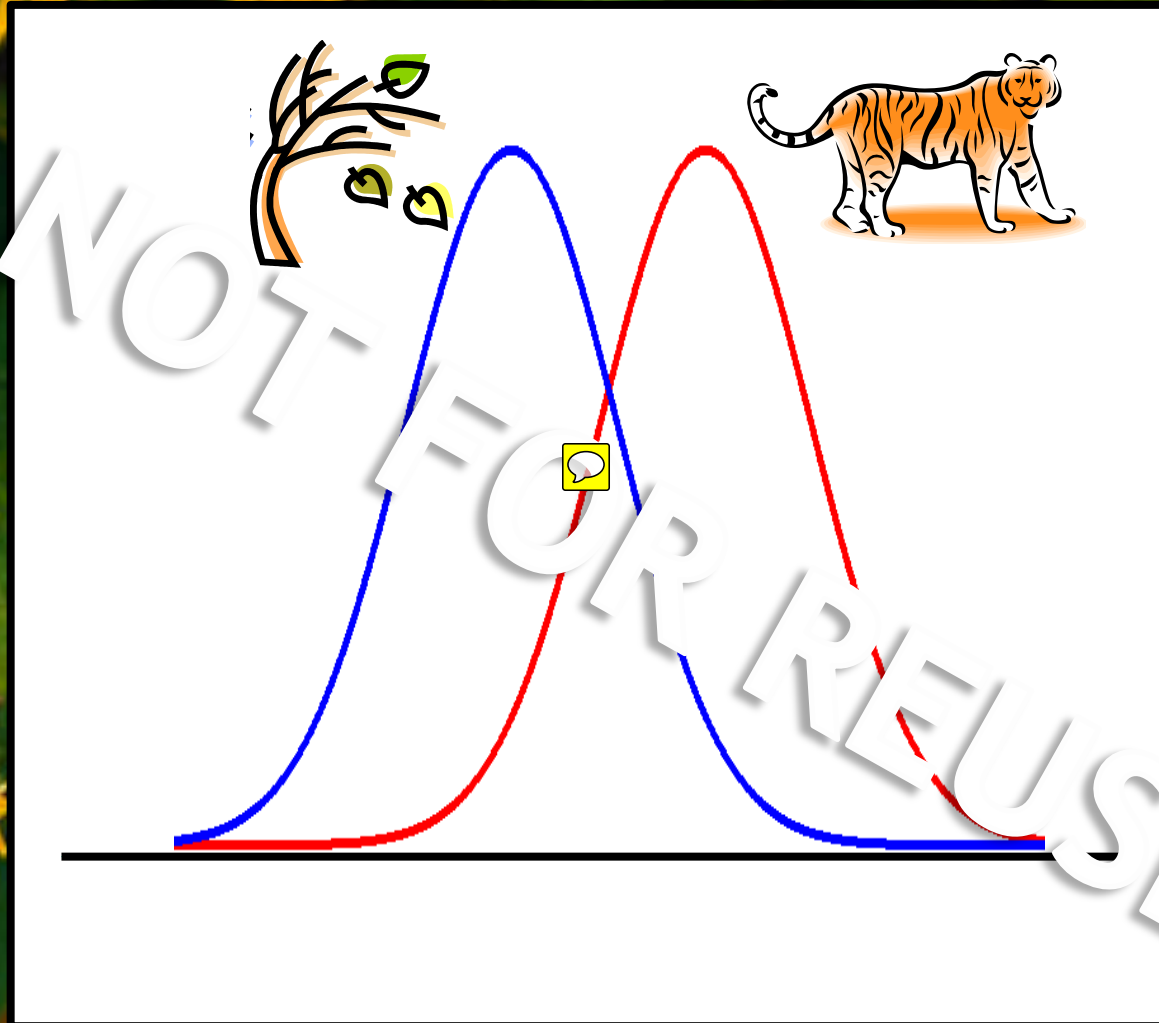
# Decoding

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How well can we learn what the stimulus is by looking at the neural responses?

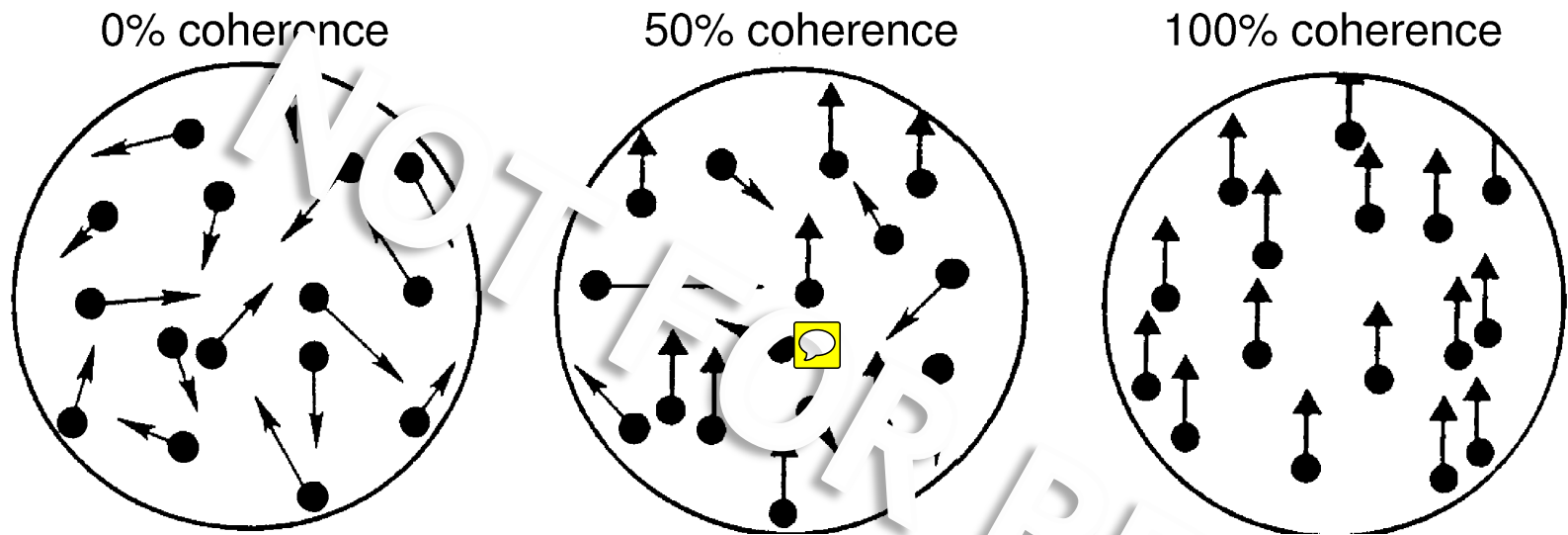
NOT FOR REUSE

Do I stay or do I go?

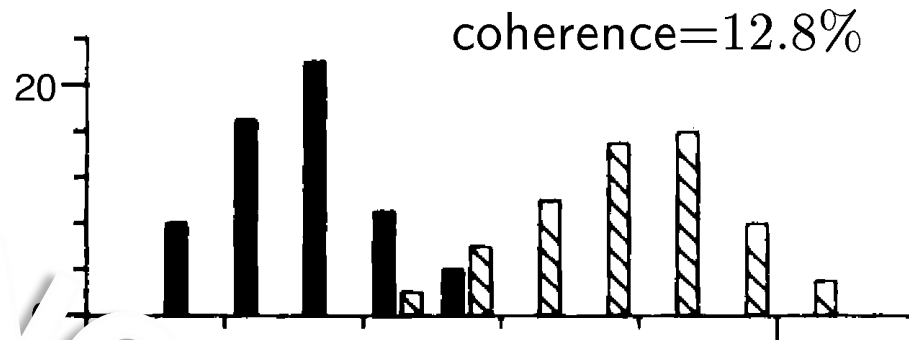


# Making a decision

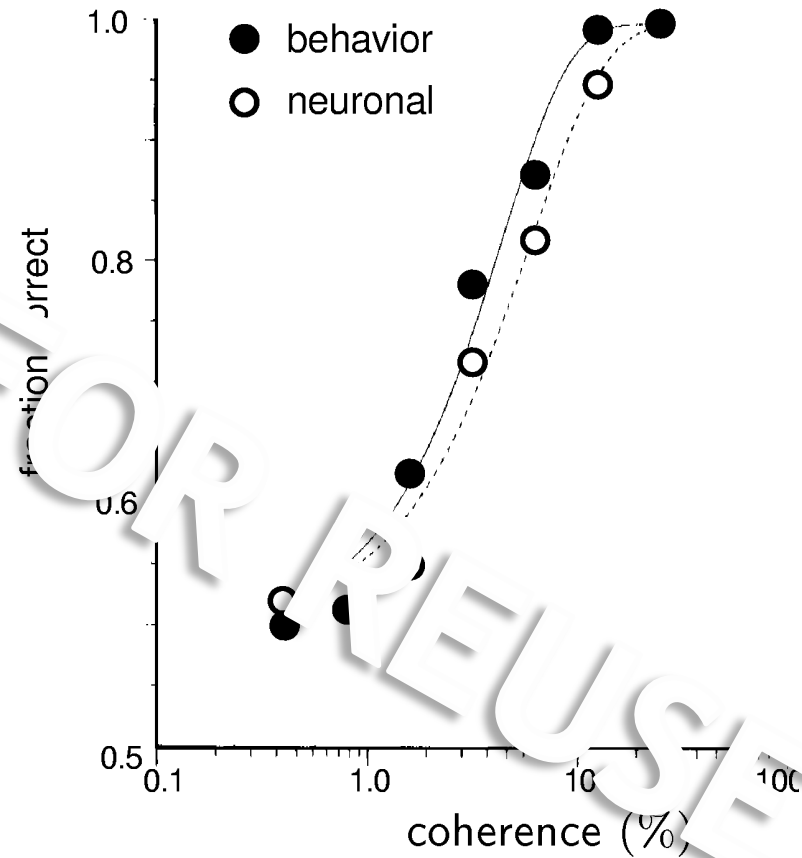
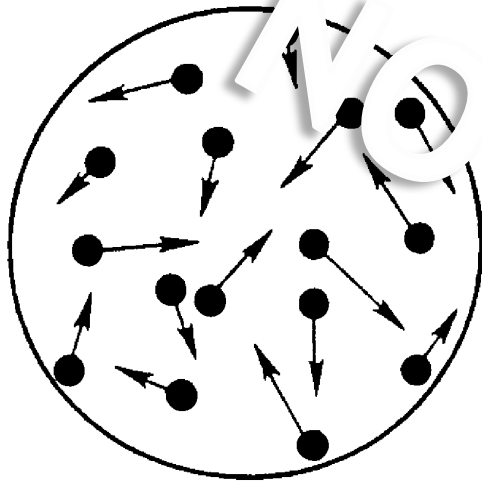
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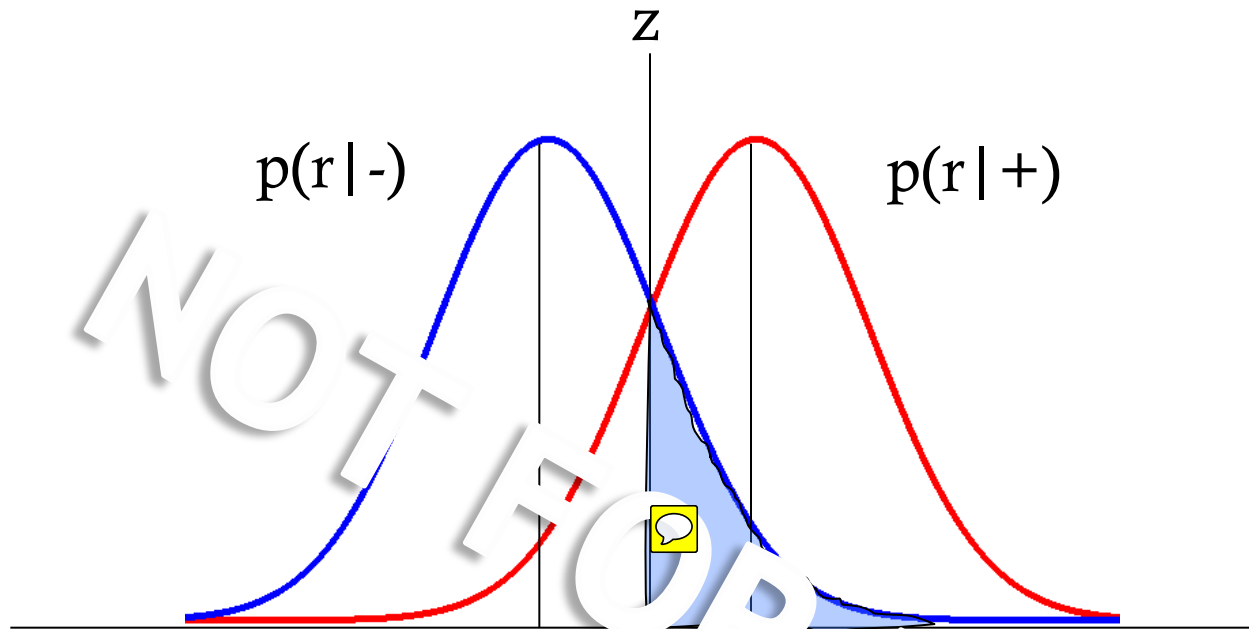
# Predictable from neural activity?



# Behavioral performance



# Signal detection theory



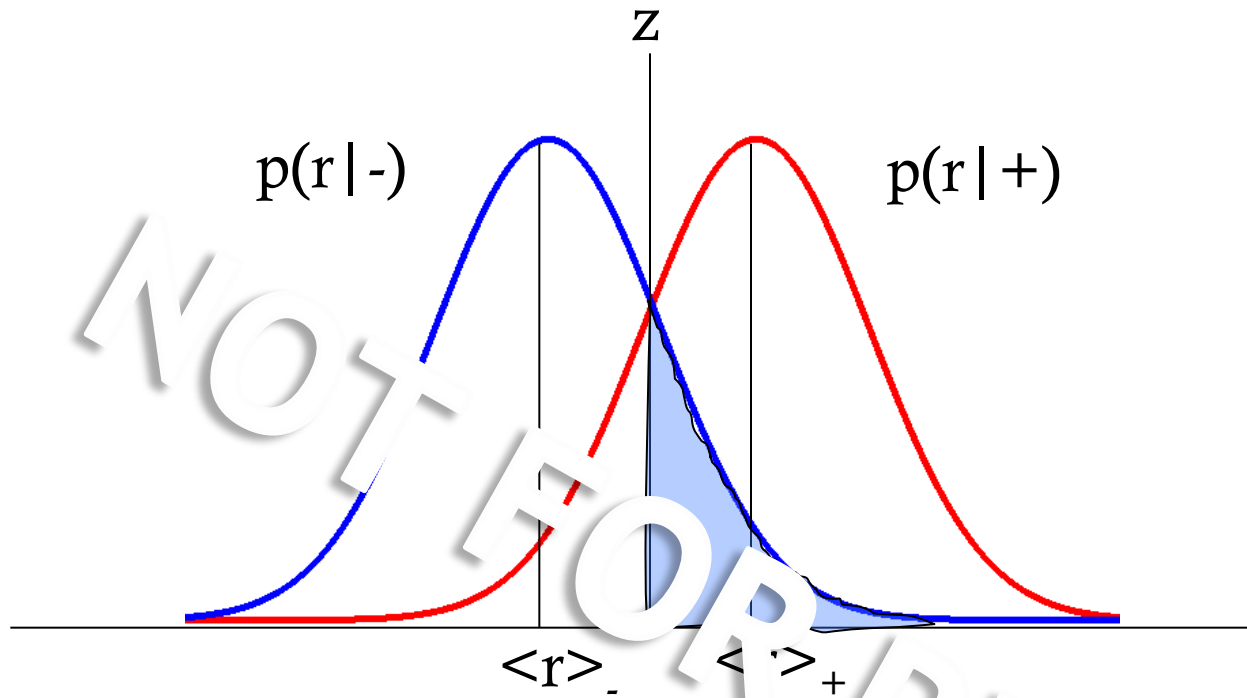
How many errors are you going to make

False alarms:  $P[r \geq z|-]$

Good calls =  $P[r \geq z|+]$

This choice of  $z$  maximizes  $P[\text{correct}]$

# Likelihood ratio



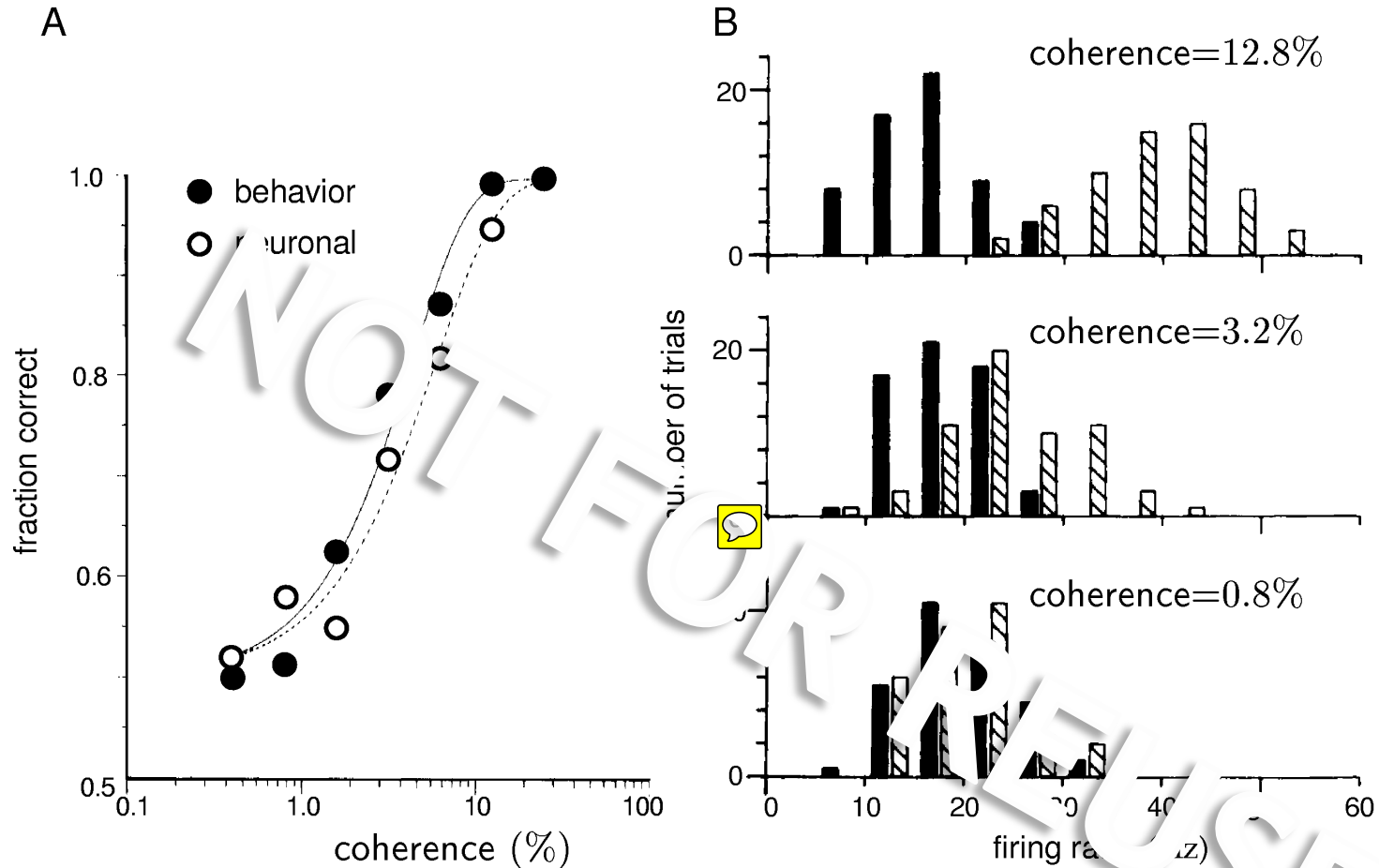
The likelihood ratio test is the most efficient statistic, in that it has the most power for a given size.

*Power* = probability of a false negative

*Size* = probability of a false positive

Neyman-Pearson lemma

# Neurons vs organisms



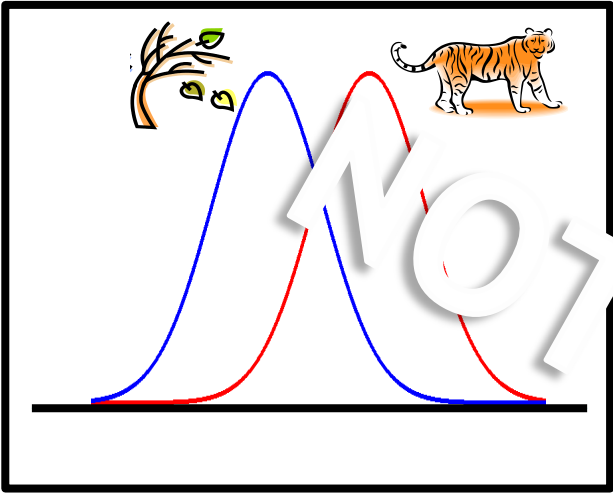
Close correspondence between neuron decoding and behavior..

So why so many neurons?

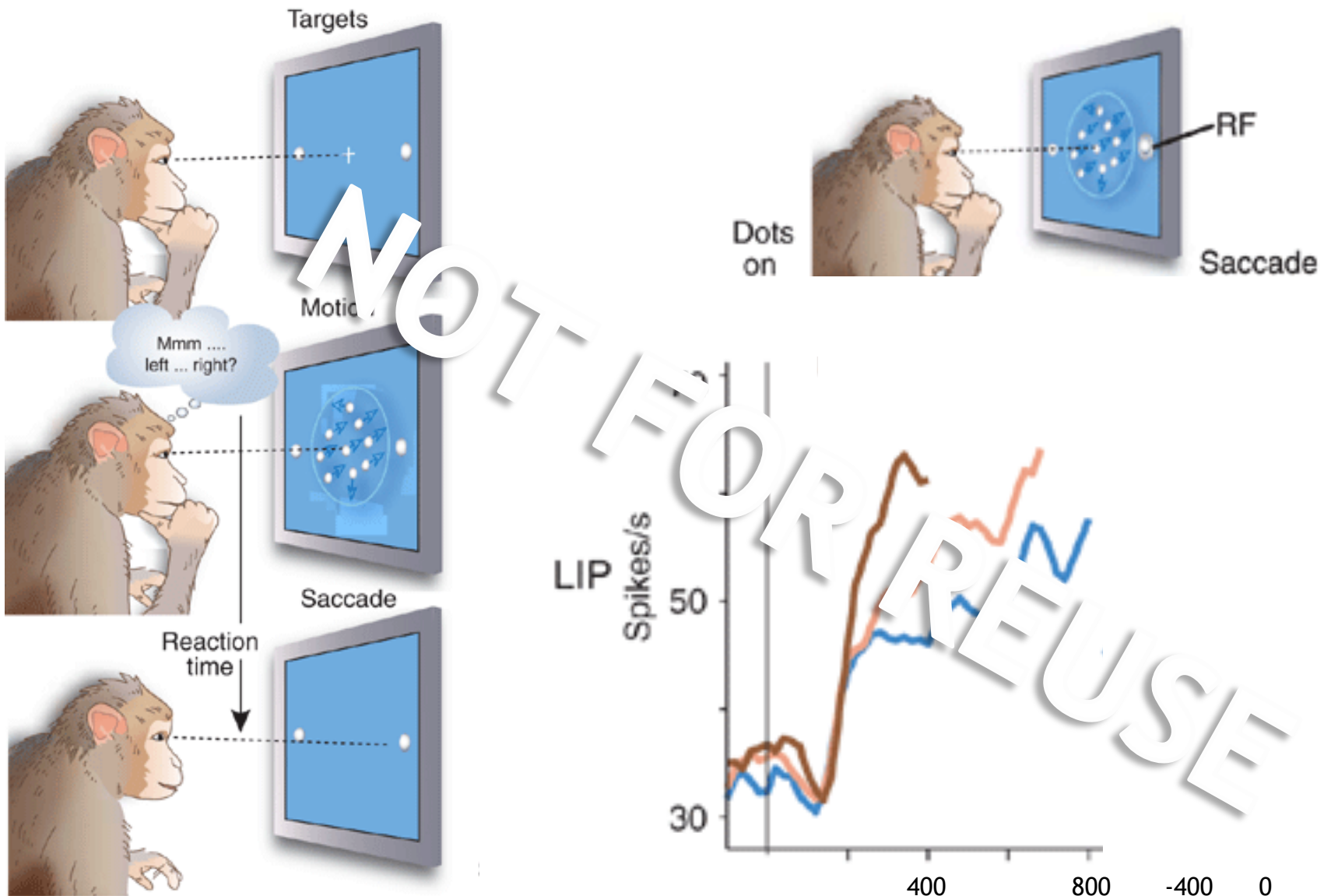


# Let's just consider for a moment

Now let's say we don't have to decide immediately...

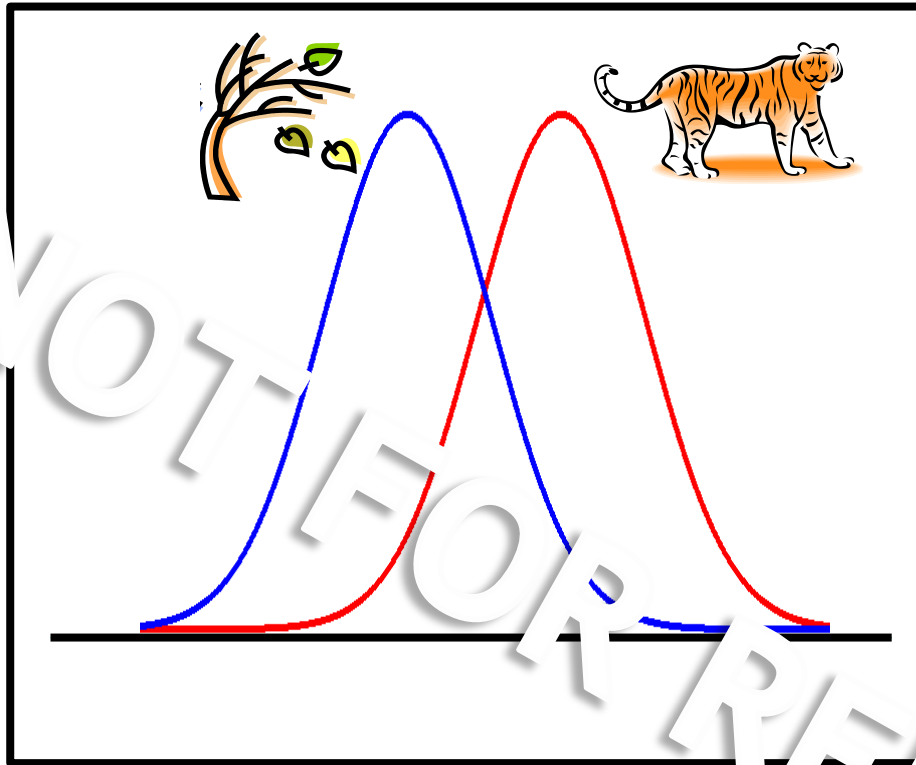


# Accumulated evidence for accumulated evidence



Kiani, Hanks & Shadlen, Nature Neuroscience (2006)

# Back to one trial: building in what we already know



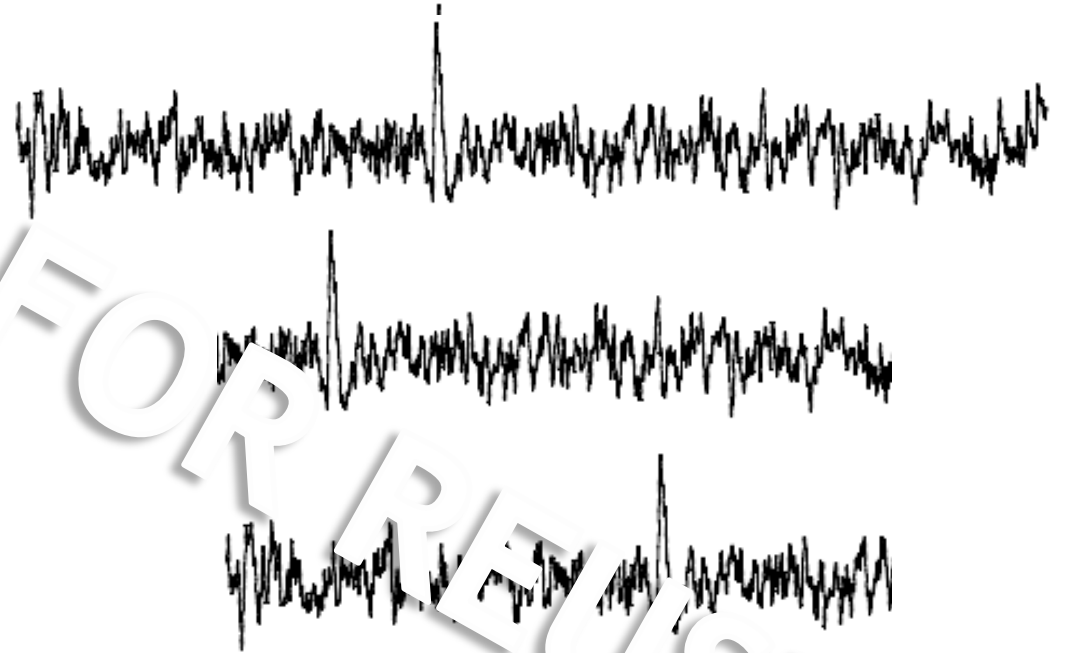
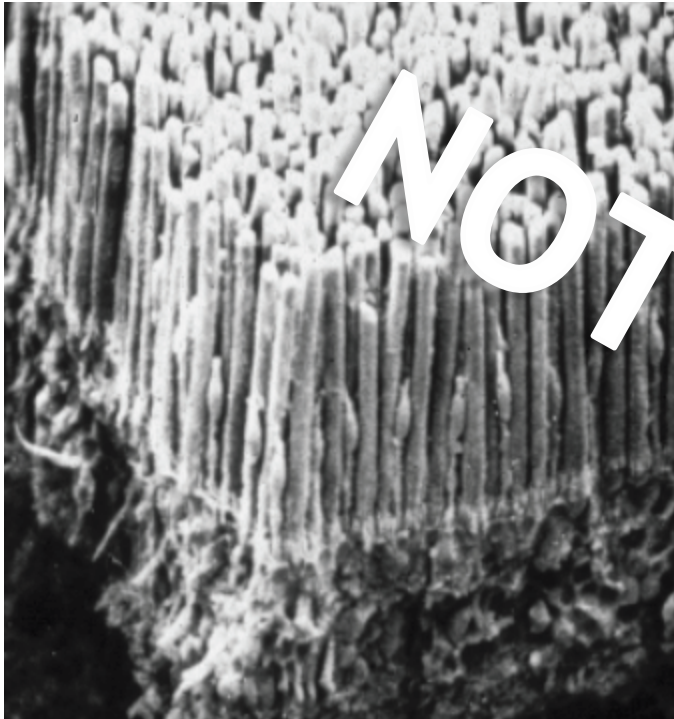
Role of *priors*:

Find  $z$  by maximizing  $P[\text{correct}] = p[+] b(z) + p[-](1 - a(z))$

# The wind or a tiger?

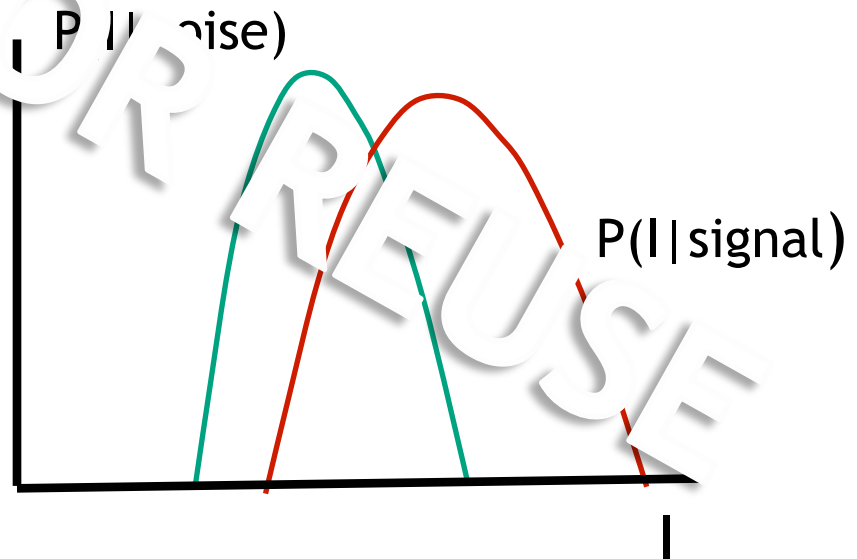
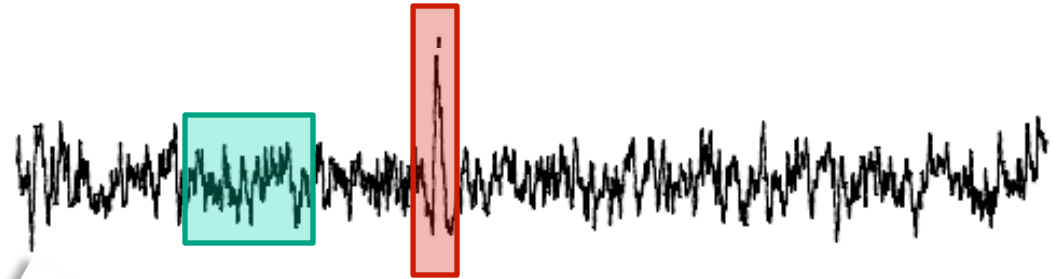
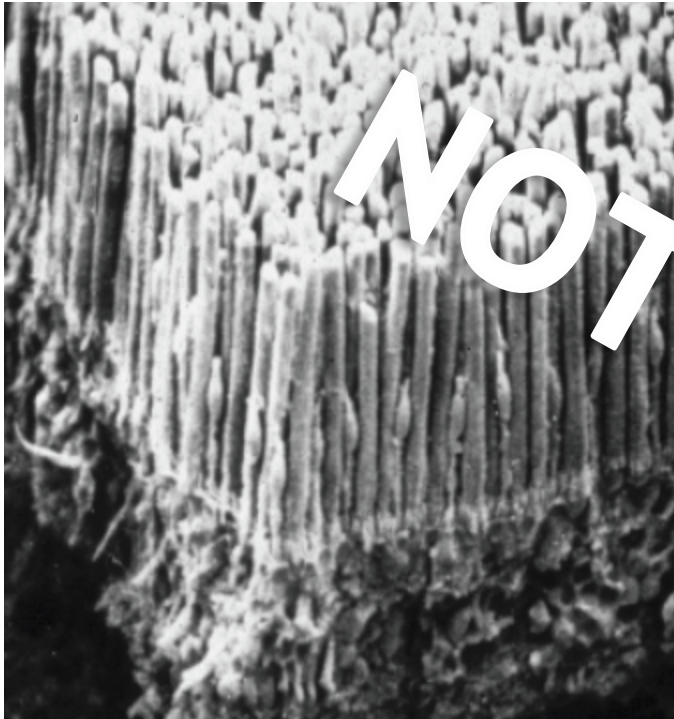
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Classification of noisy data: single photon responses



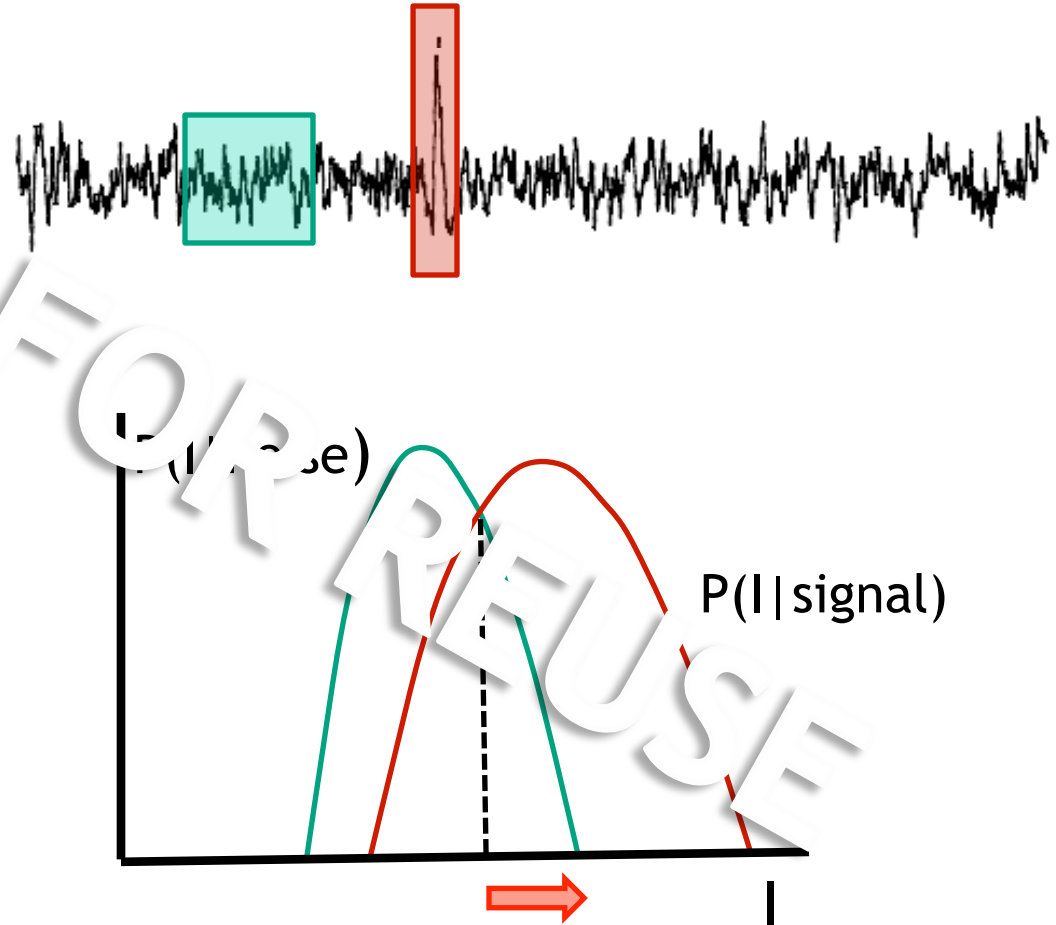
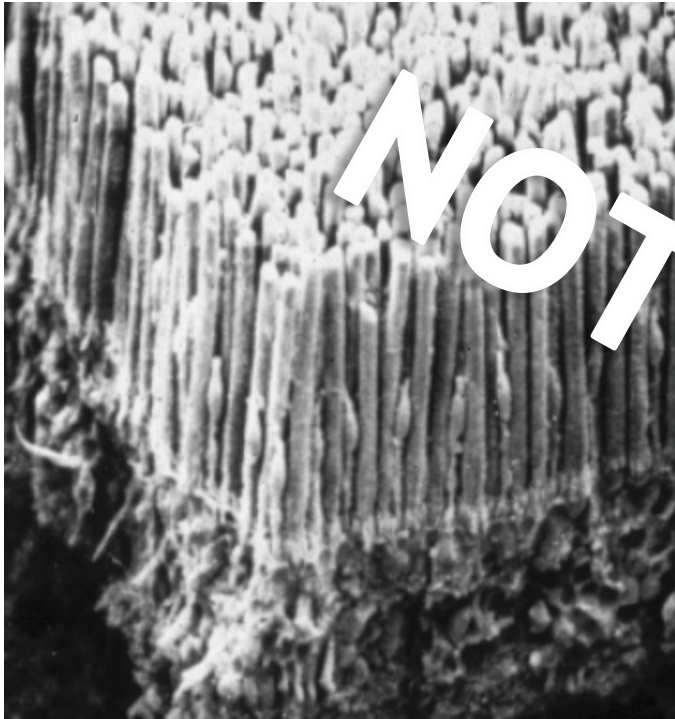
# Nonlinear separation of signal and noise

Classification of noisy data: single photon responses



# Nonlinear separation of signal and noise

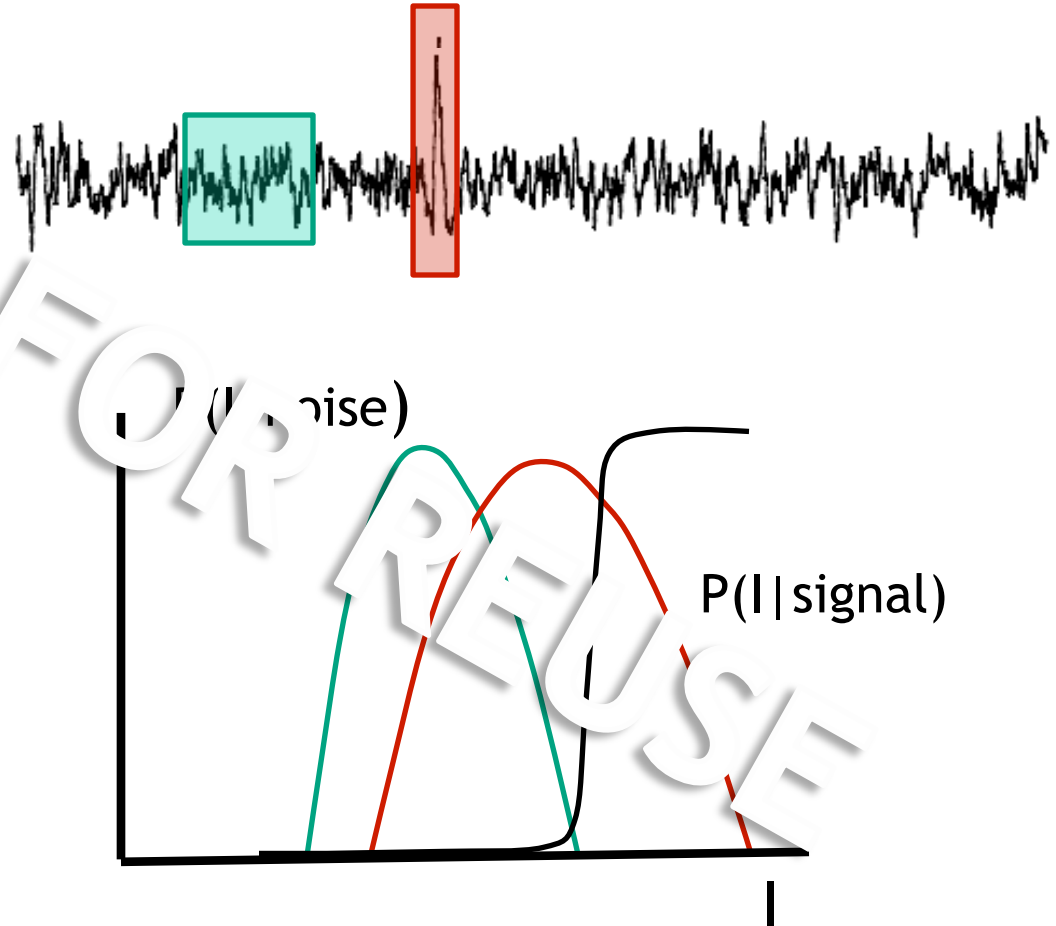
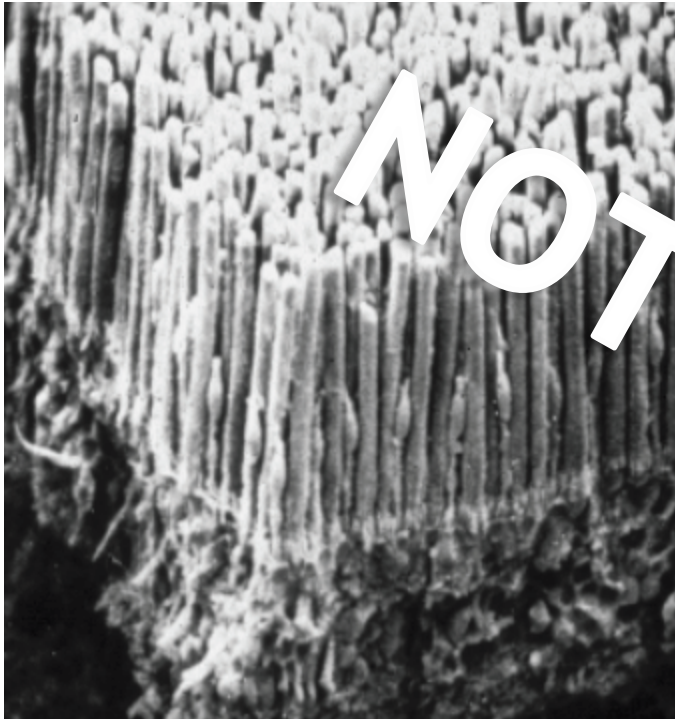
Classification of noisy data: single photon responses





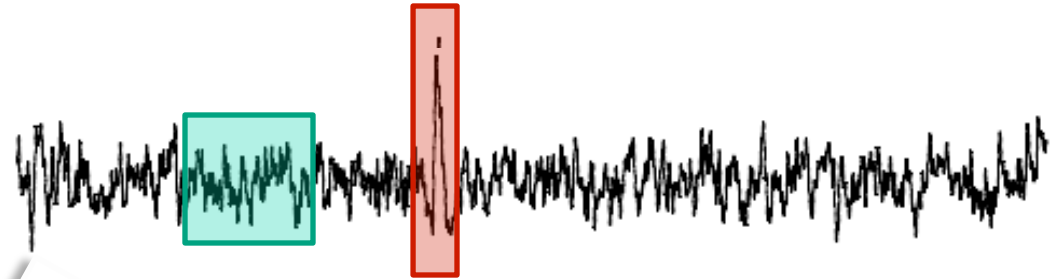
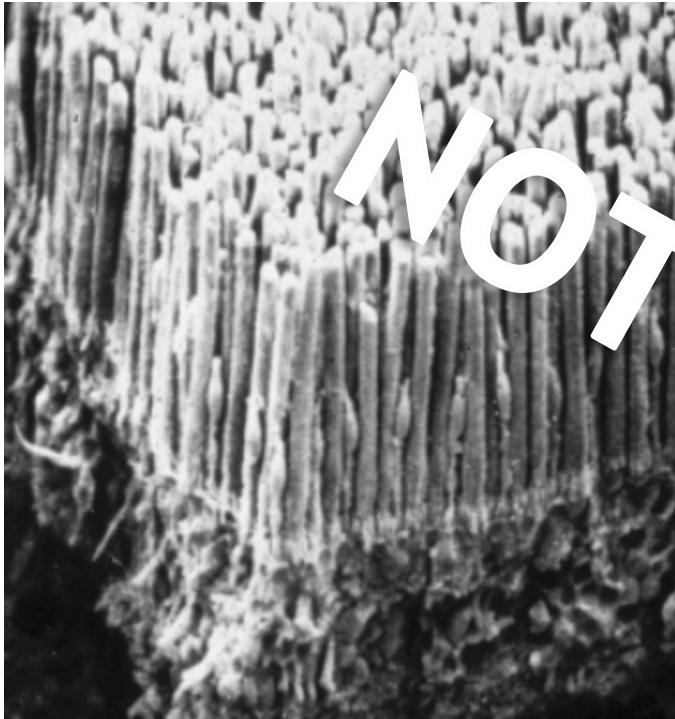
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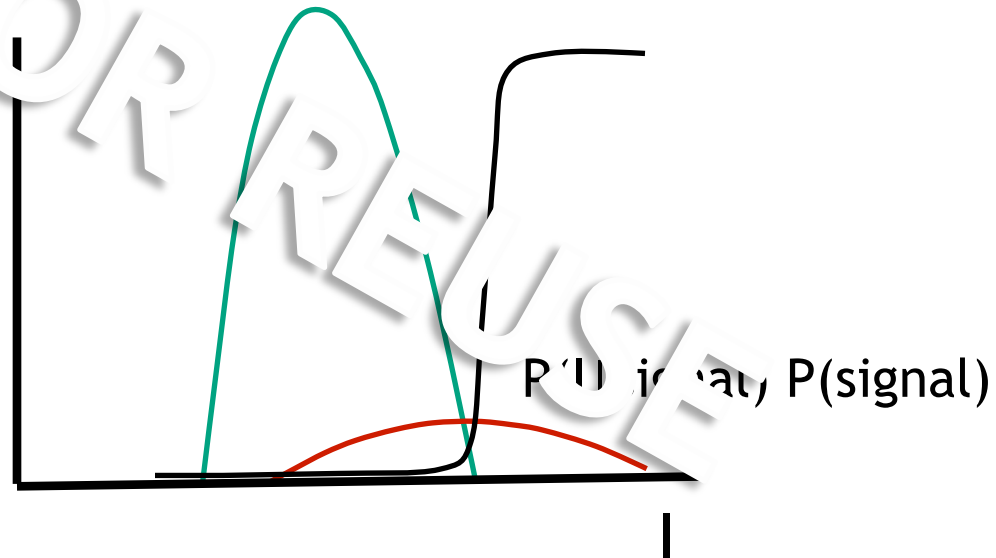


# Nonlinear separation of signal and noise

Classification of noisy data: single photon responses



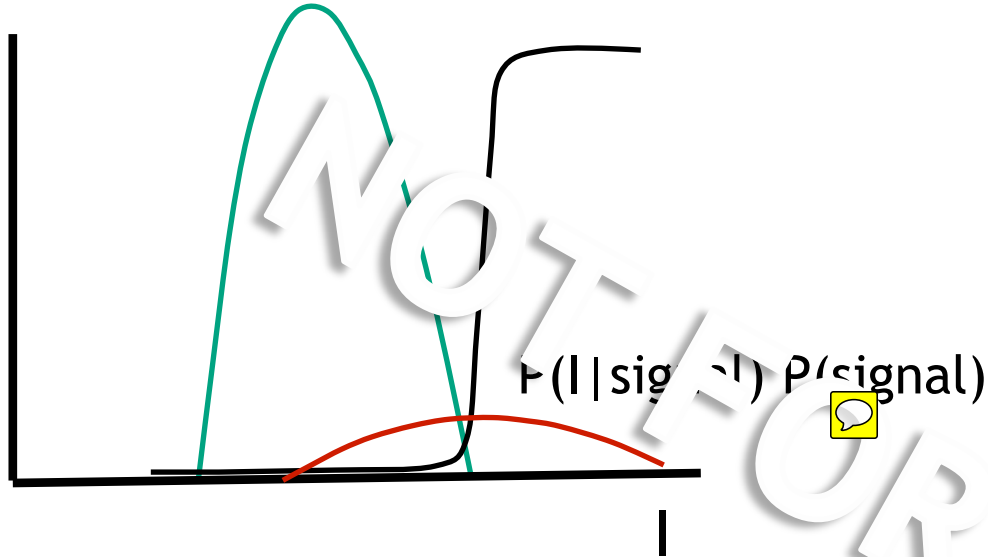
$P(\text{noise})$   $P(\text{noise})$





# That's prior knowledge: how about costs?

$P(I|\text{noise})$   $P(\text{noise})$



*the signal and the  
and the noise and  
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noise and the no  
why so many are  
predictions fail—  
but some don't t  
the noise and  
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noise and the no*

# Building in cost



Cut your losses: answer + when Loss<sub>+</sub> < Loss<sub>-</sub>

i.e.  $L_+P[- | r] < L_-P[+ | r]$ .

$$\rightarrow p[r | +] / p[r | -] > L_+P[-] / L_-P[+]$$