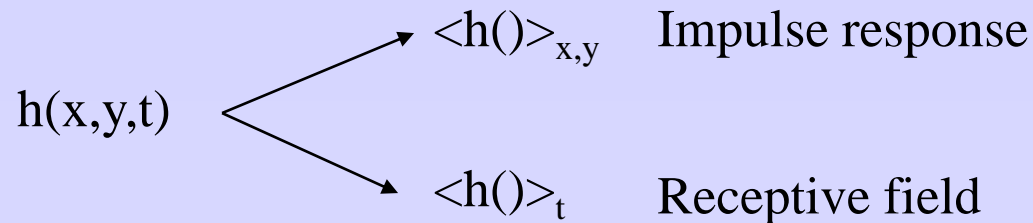


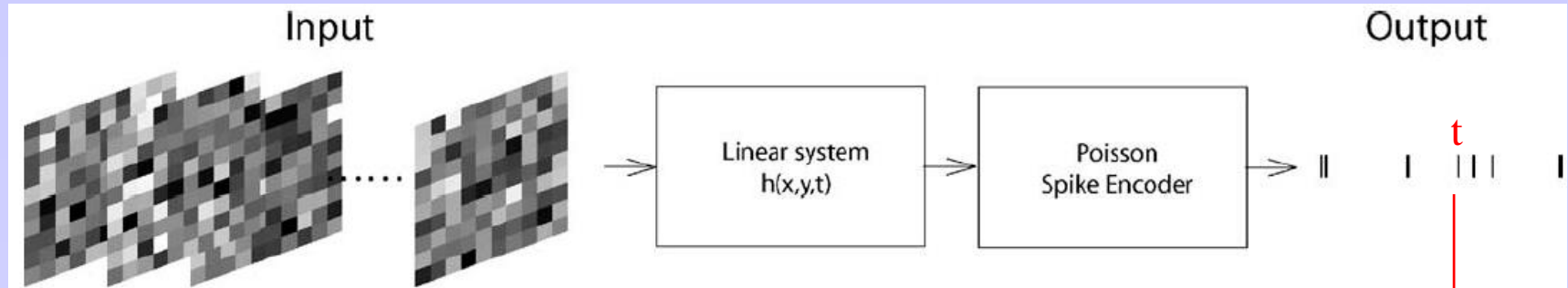
# Where are we?

- Estimate the neuron response, given a stimulus. The impulse response  $h(t)$ .
- Case of discrete response: use STA, case of continuous response use Wiener kernel/linear approximation. If white noise stimulus, use STA.
- Example of V1 (Ringach & Shapley, 2004).  $h(x,y,t)$  by subspace reverse correlation. Gabor kernel. Use  $h(t)$  to study the orientation selectivity of V1 cells, and its time course.
- Next: Example of V1 (Usrey, Sceniak and Chapman, 2003)



# V1 - Spatial receptive field

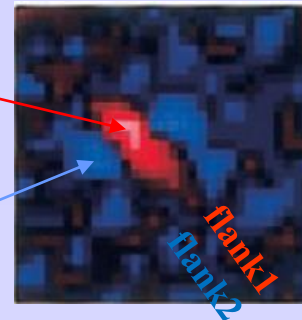
→ Receptive field = Average spatial response. No temporal information.



Temporal average as a function of  $(x,y)$  – Ferret V1, Layer 4



(Usrey et al, 2003)



There are center-surround V1 cells!

on

off

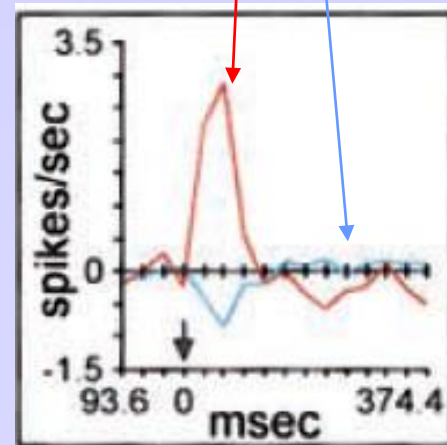
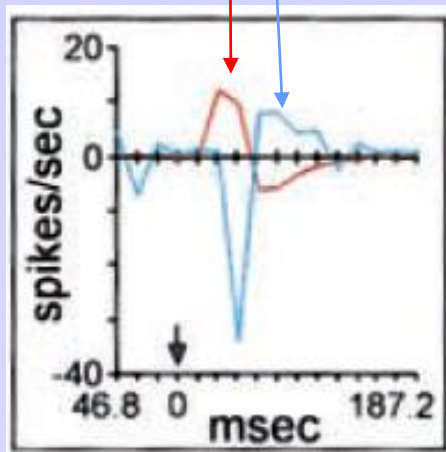
- Separating the excitatory and inhibitory spatial regions

# V1 - Impulse function

- Impulse response calculated for each spatial subregions. Separating the 'excitation' and 'inhibition' time courses.

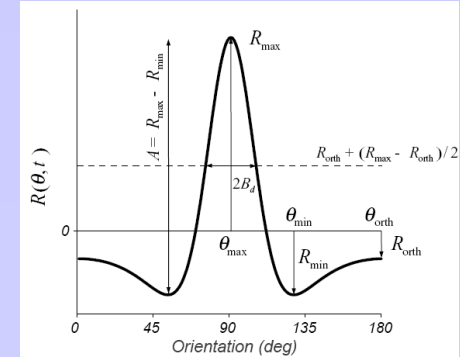


Spatial average as a function of  $\tau$

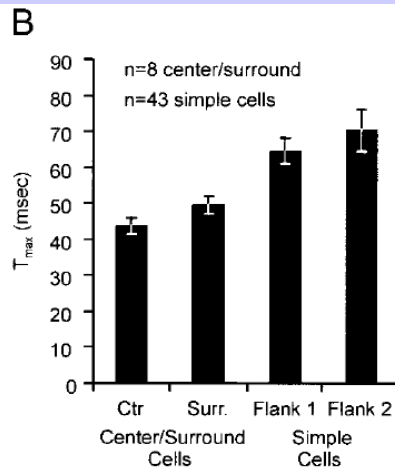
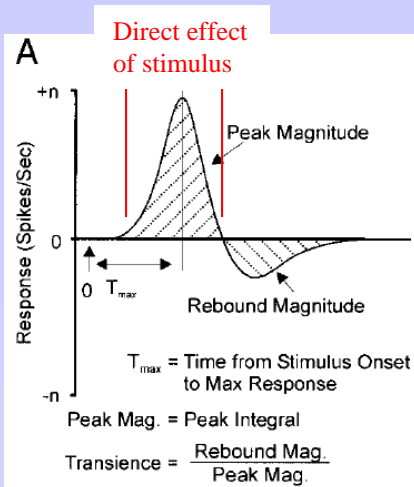


# Impulse function

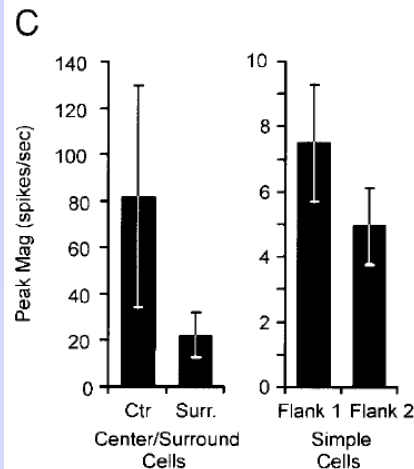
- How does one quantify an impulse function?



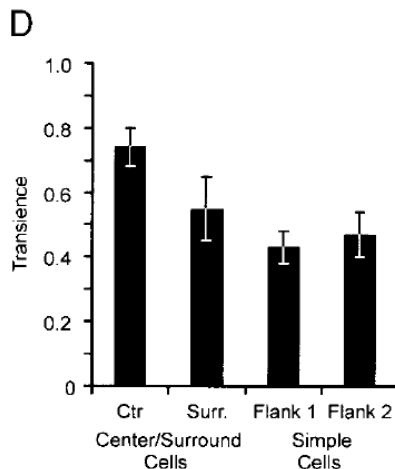
(Ringach shapley)



- B: Max response fastest for center/ surround cells.



- C: Magnitude of peak smallest for center/ surround cells.



- D: 'Transience' largest for center/ surround cells.

(Usrey et al, 2003)

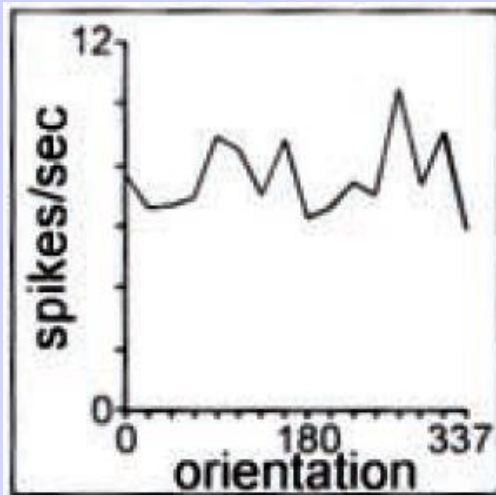
# Orientation selectivity

Stimuli:  
Moving gratings

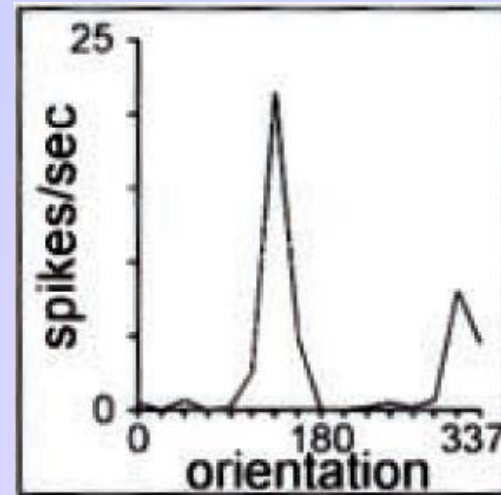


→ Orientation tuning curve: Average response (firing rate) per orientation

Not selective



Selective

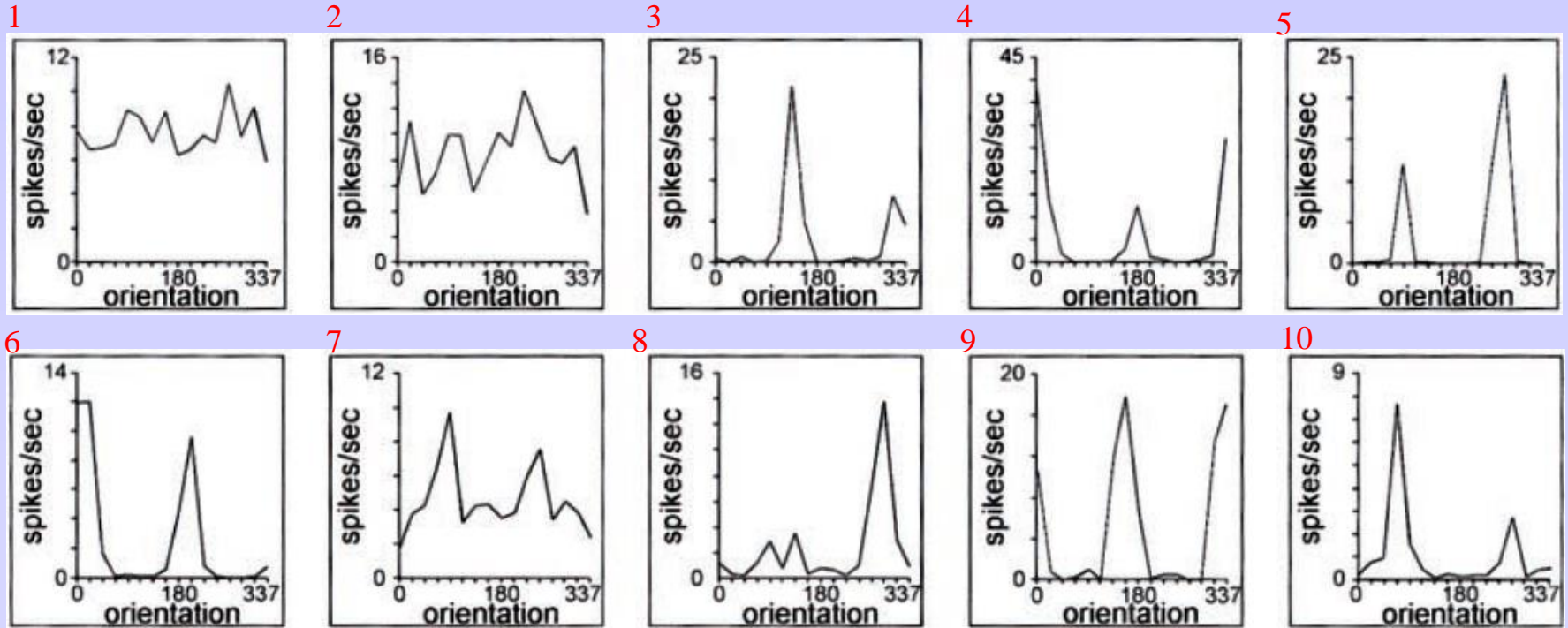


(Usrey et al, 2003)

→ Again....No temporal information...

# Orientation selectivity plots

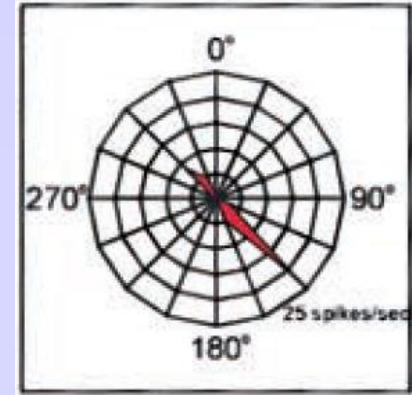
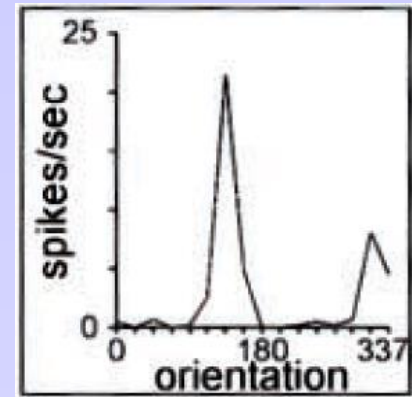
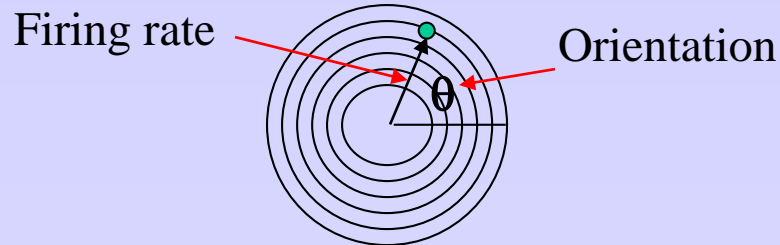
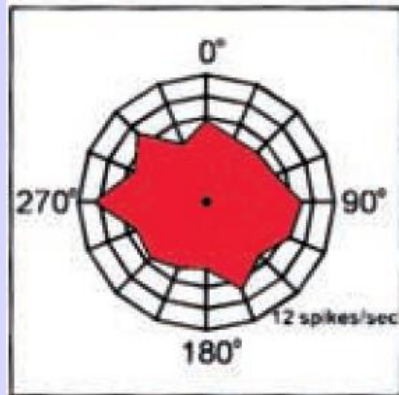
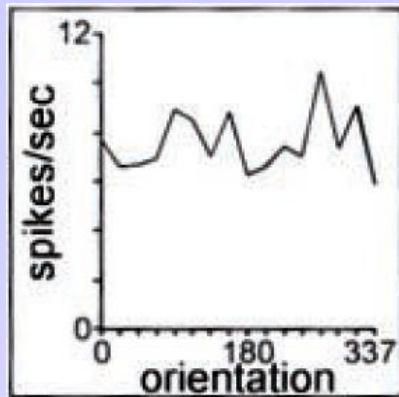
- Different types of selectivity: What do you see, qualitatively ?



(Usrey et al, 2003)

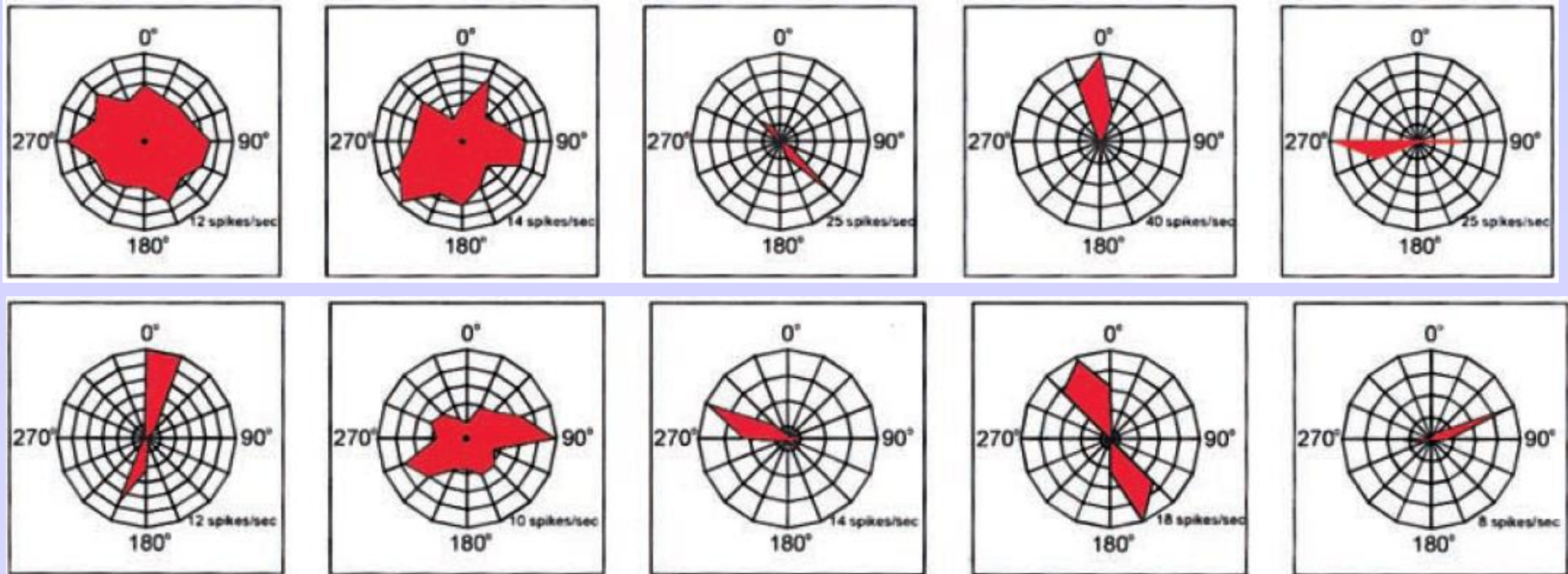
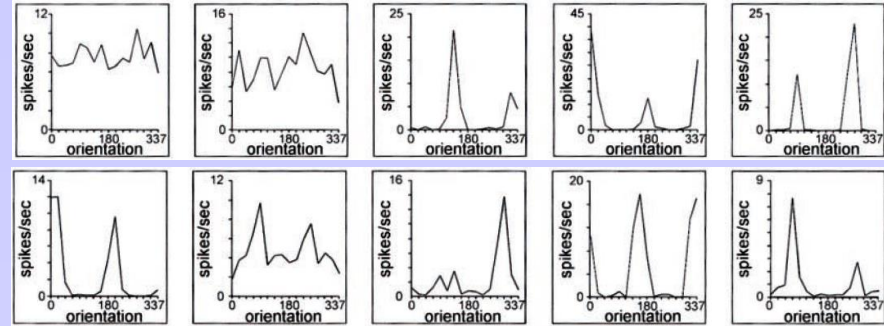
# Orientation selectivity plots

- Can we find *one graphic* per cell that characterizes its orientation selectivity?
- From Cartesian to Polar plots.
- Intuitive graphics... quick visualization ... no loss of information



# Orientation selectivity plots

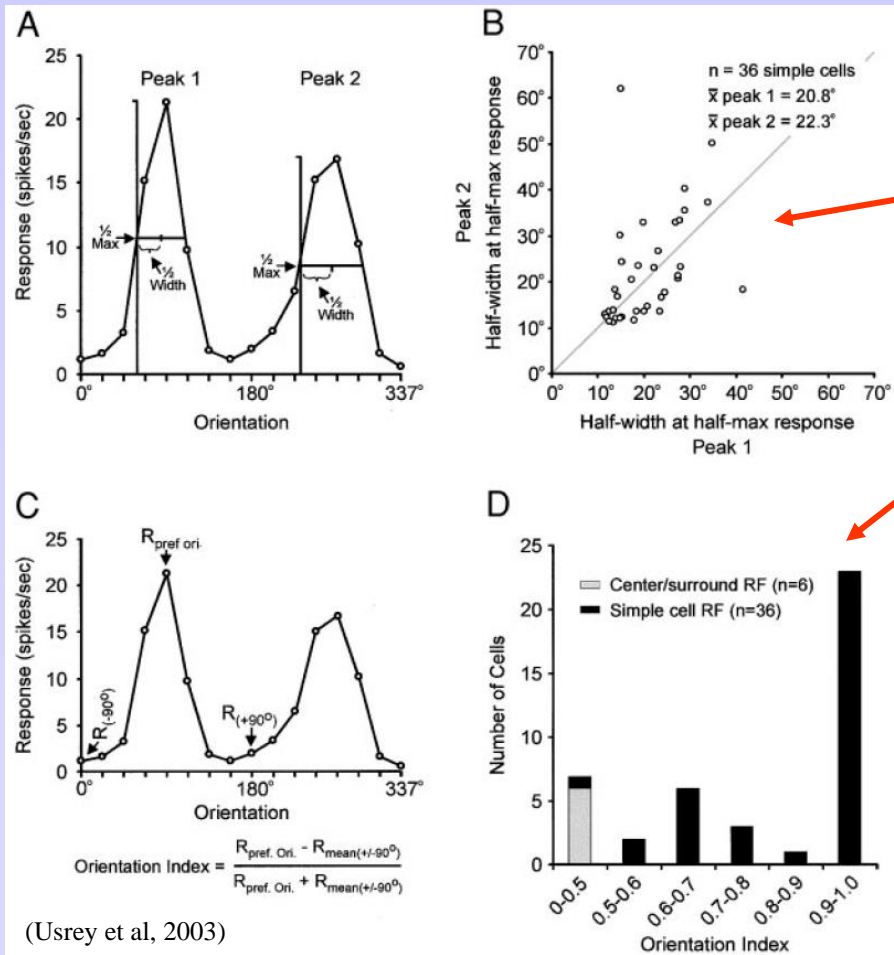
- Different types of selectivity: What do you see ?





# Orientation selectivity: building an 'index'

- Can we find *one number* per cell that characterizes its orientation selectivity?
- From Qualitative to *Quantitative*: How does one quantify an orientation selectivity curve?



- 'tightness' or 'selectivity' of orientation.

- 'strength' of orientation selectivity (orientation index)

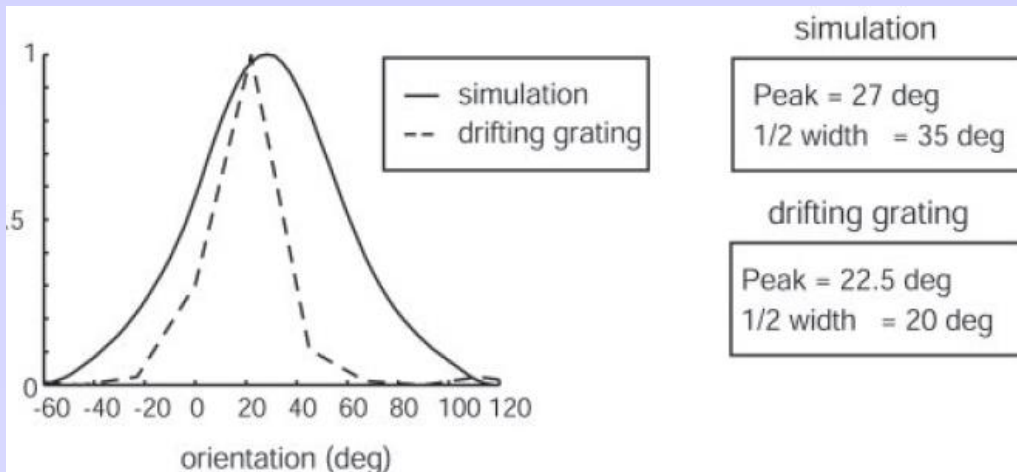
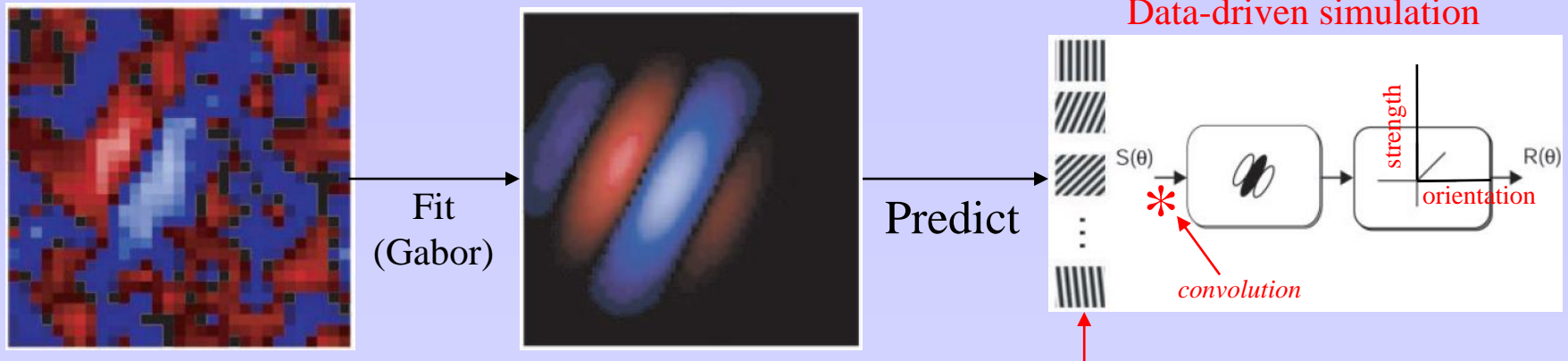
$$OI = \frac{R_{pref\ Orient} - R_{mean(\pm 90^\circ)}}{R_{pref\ Orient} + R_{mean(\pm 90^\circ)}} \left. \begin{array}{l} 0 \rightarrow ? \\ 1 \rightarrow ? \end{array} \right\}$$

(Usrey et al, 2003)

# Orientation Selectivity: data-driven simulation

- Now that we have a number ...

→ Can orientation selectivity be (linearly) predicted from the receptive field structure?



(Usrey et al, 2003)

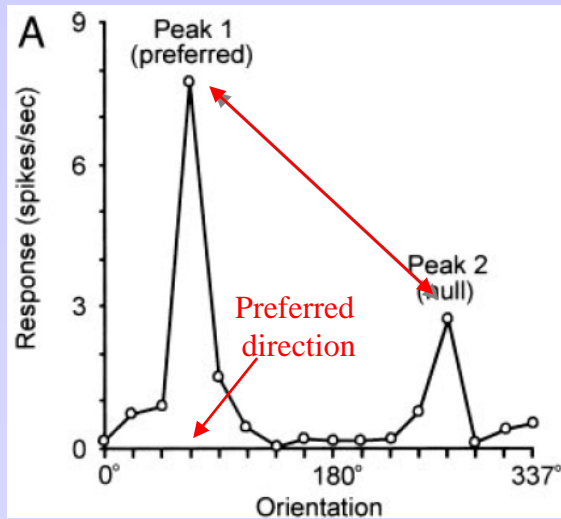
- Good for peaks  
i.e. orientation preference
- Bad for half-widths  
i.e. sharpness/selectivity

→ Selectivity = non linear consequence of receptive field?

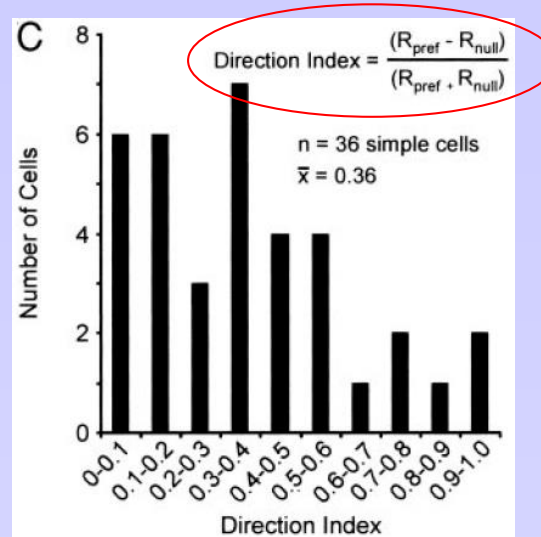
# Direction selectivity

- Now that we have a number ...

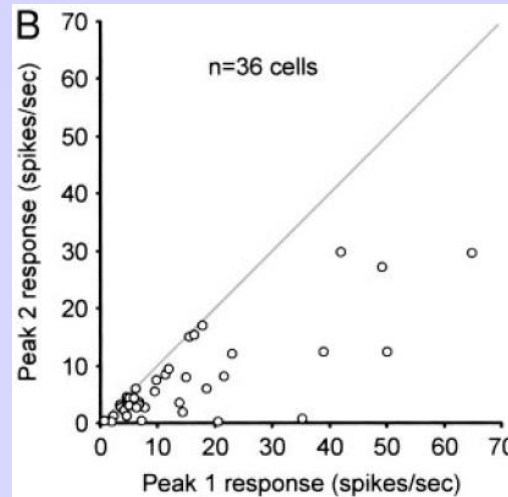
→ The same analyses can be done for other features/indexes. Example: direction selectivity



(Usrey et al, 2003)



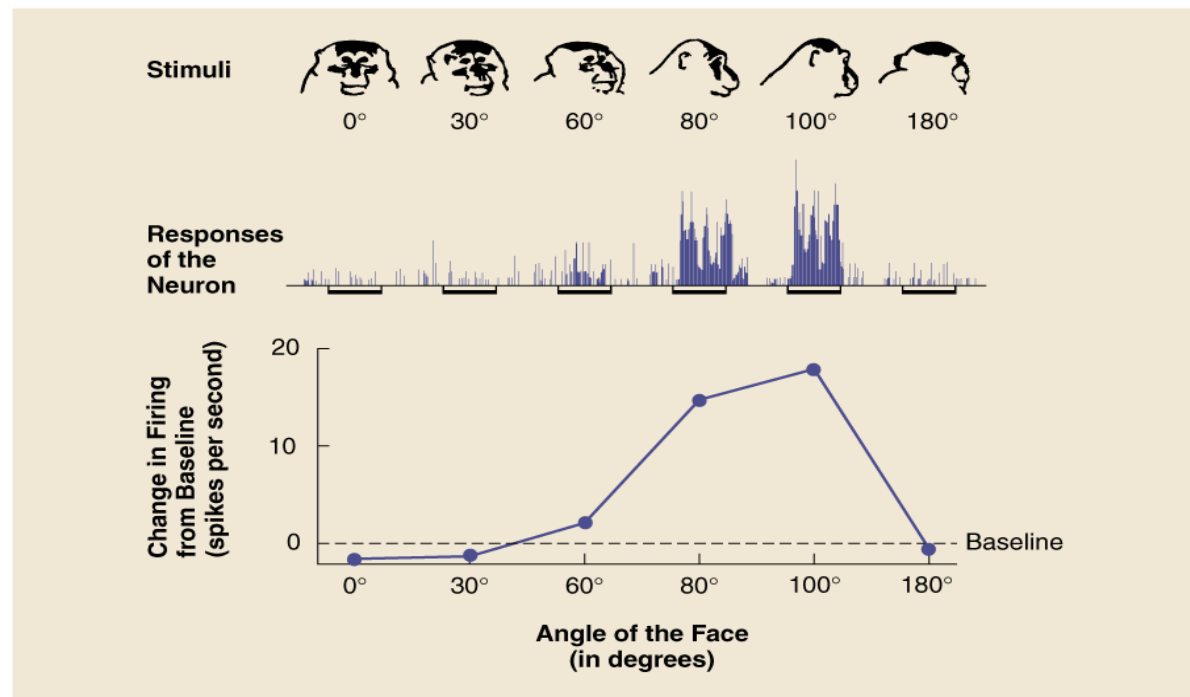
1 → strong direction preference



# Tuning curves

- Relationship: response  $\leftrightarrow$  stimulus is not necessarily linear
- Tuning curves: Firing rate Vs. stimulus feature

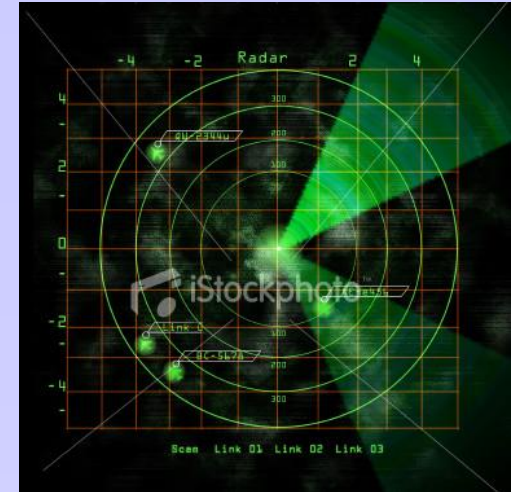
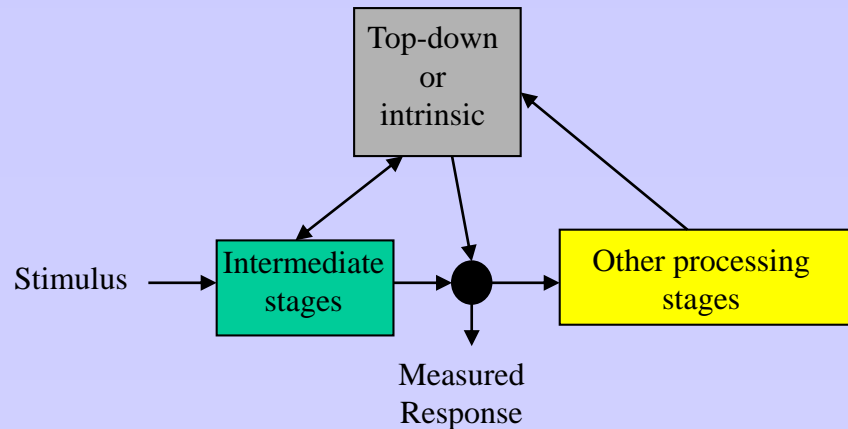
► Firing Rate of a Monkey Inferotemporal Cortex Neuron



Source: Adapted from Gross et al., *Pattern Recognition Mechanisms*, p. 179-201. Berlin: Springer-Verlag 1985.

- Other tuning curves: Odors? Sounds?....  $\longrightarrow$  Need 'metric'

# Signal Detection Theory



Measured response = Spike train = ‘Signal(s)’ + noise

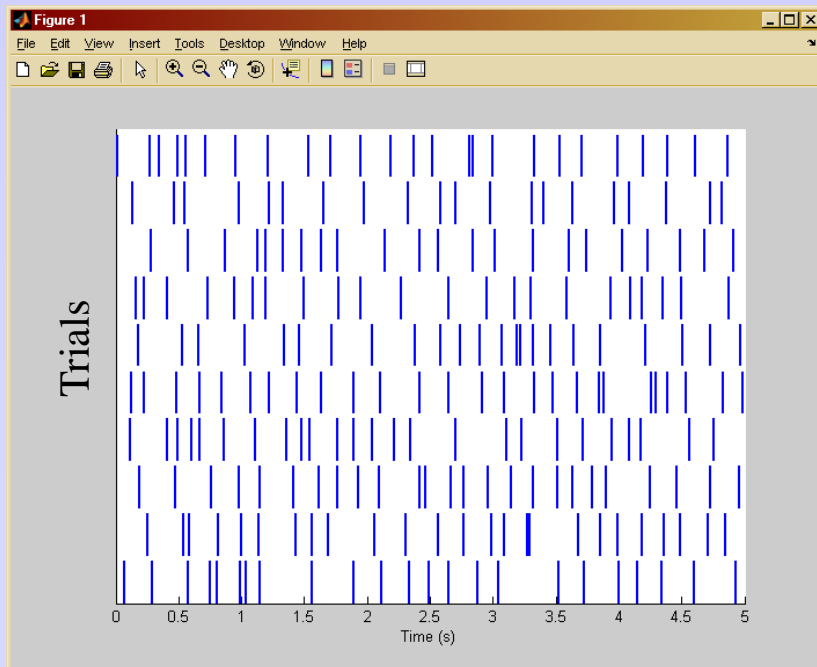
- **Detection** of a signal: signal? or Noise?
  - > Is the neuron responding at all? (PSTH...)
- **Discrimination** between 2 responses: same signals? different signals?
  - > Are two responses of the same cell ‘equal’? (tuning curves...)
- **Classification**: How many signals? Does the response contain one of several signals?
  - > Are there different ways to respond to the same stimulus? (Information theory)

# Discriminability

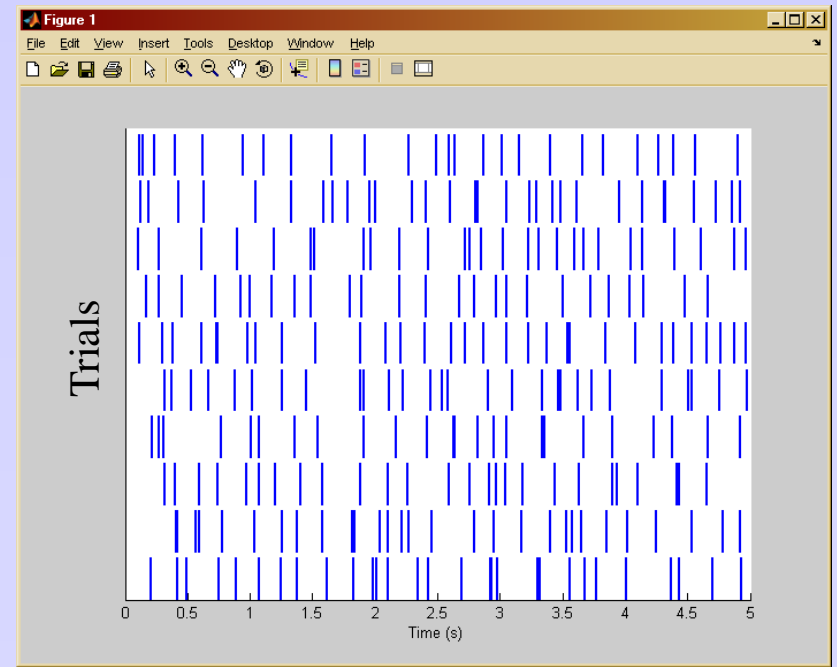
- Noise = spontaneous activity = ‘background signal’

General problem: Detection Vs. Discrimination

- Are these two experimental conditions different?



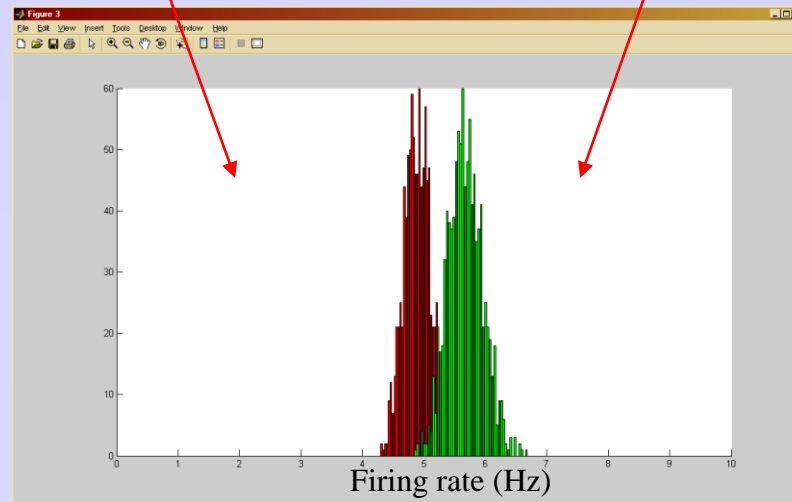
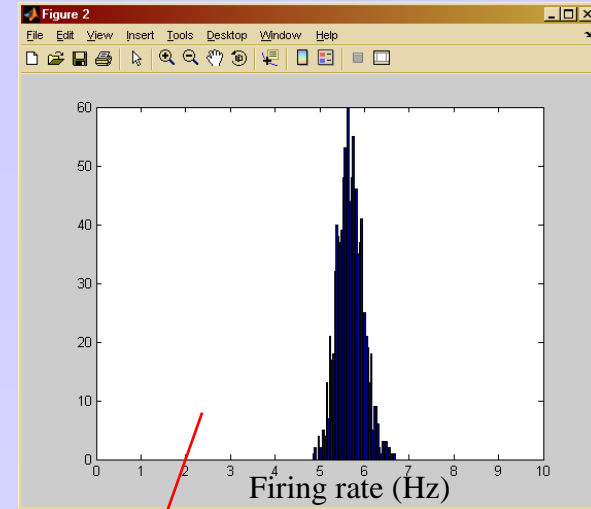
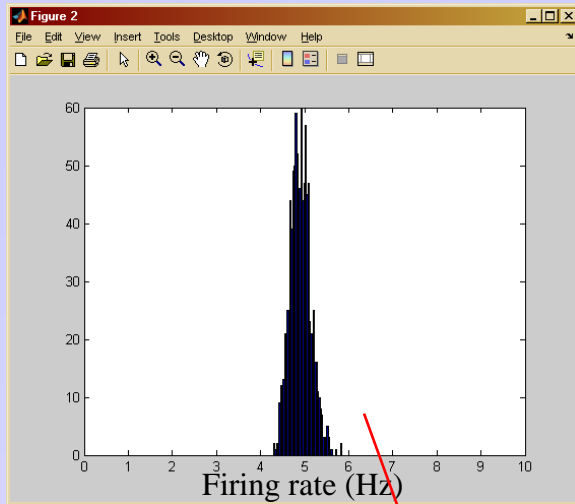
Stimulus



Stimulus

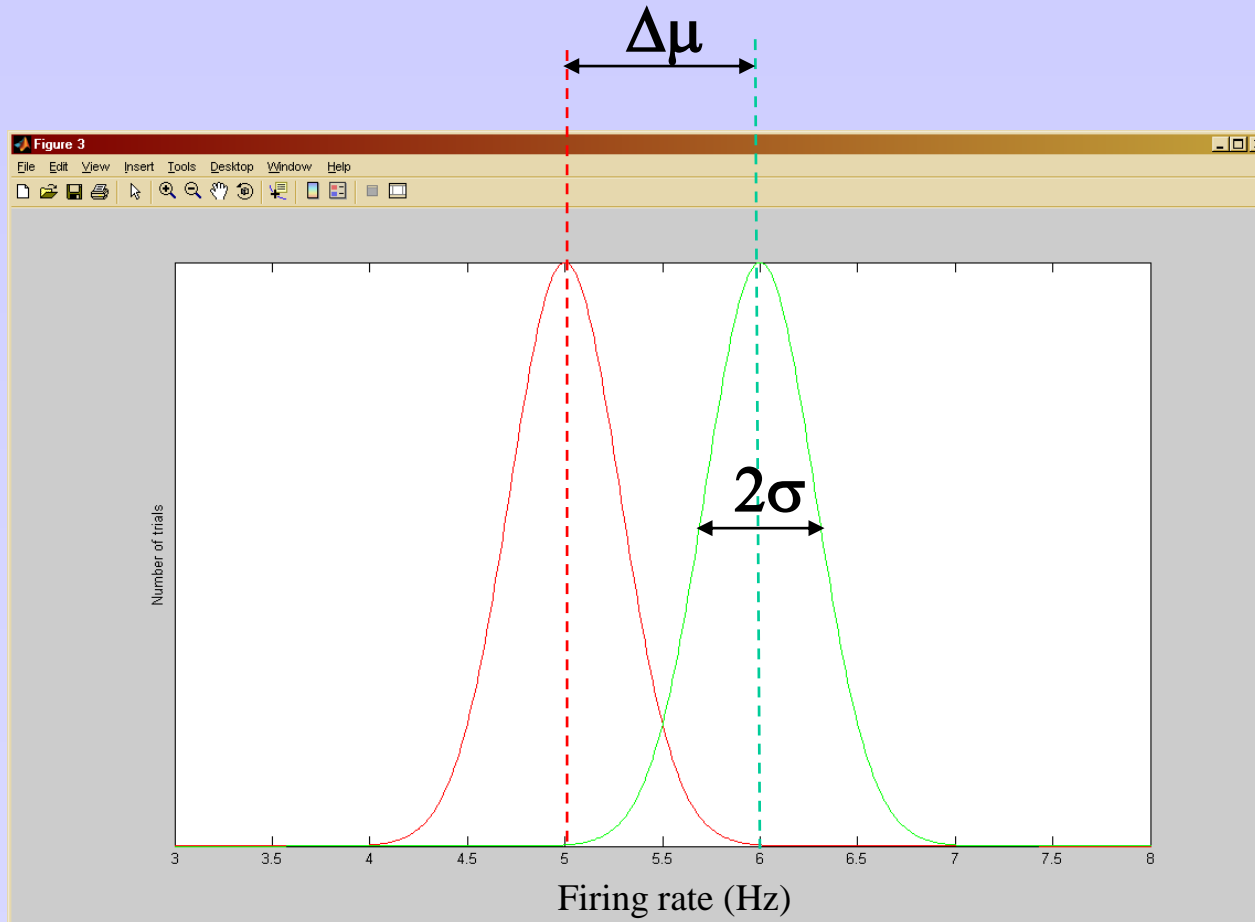
# Discriminability

- Are the two experimental conditions different?
  - Are the firing rate histograms different?



# Discriminability: $d'$

- Assuming the firing rate distributions are *displaced* Gaussian:



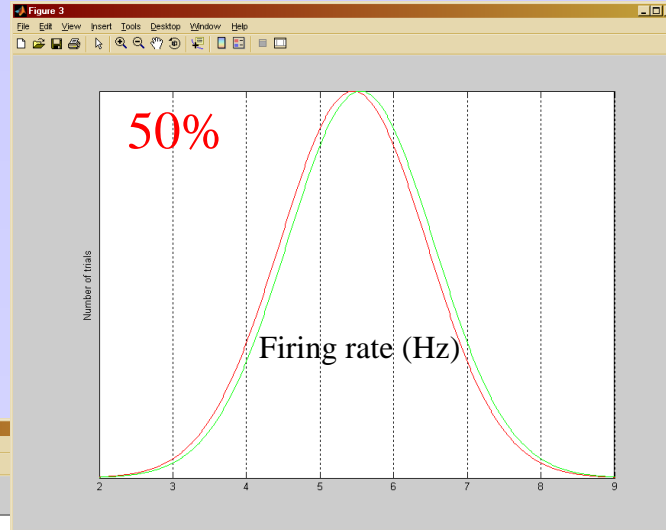
$$d' = \frac{\Delta\mu}{\sigma}$$



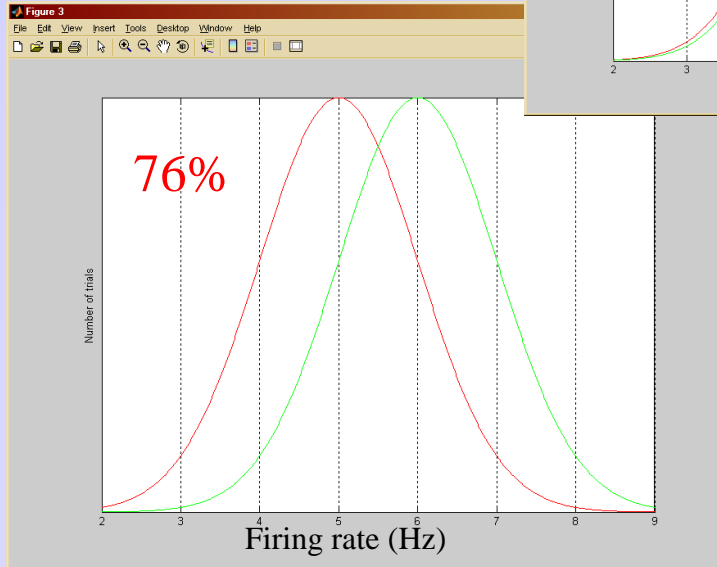
# Discriminability: $d'$

- Interpreting  $d'$ : If you were given a number within the red distribution... how likely are you of being right calling it 'red' ?

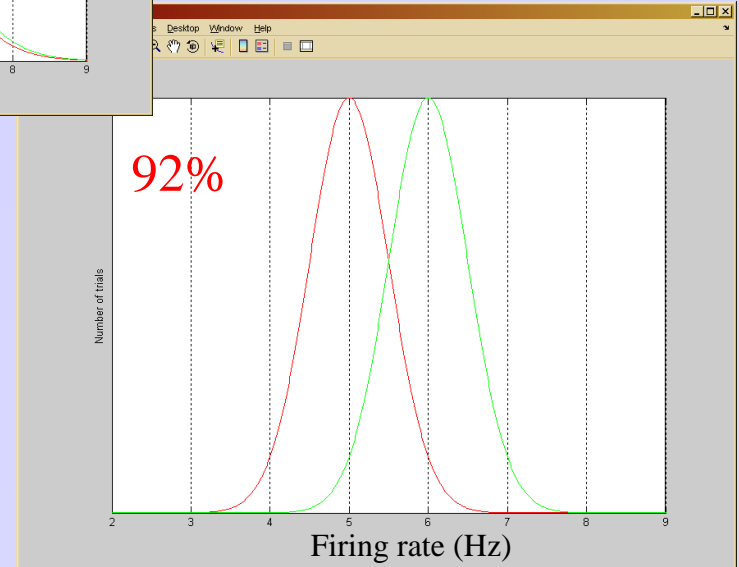
$d'=0$



$d'=1$



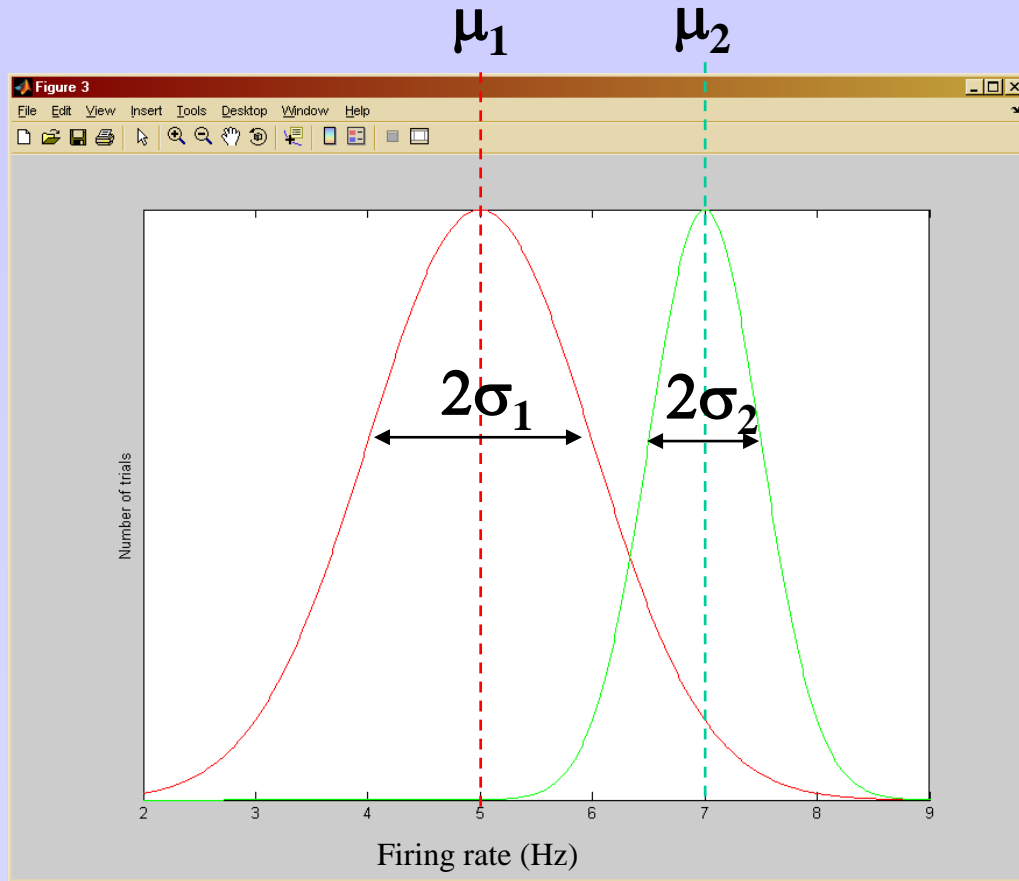
$d'=2$



2-alternative-forced-choice

# Discriminability: $d'$

- What if the two distributions have different standard deviations?

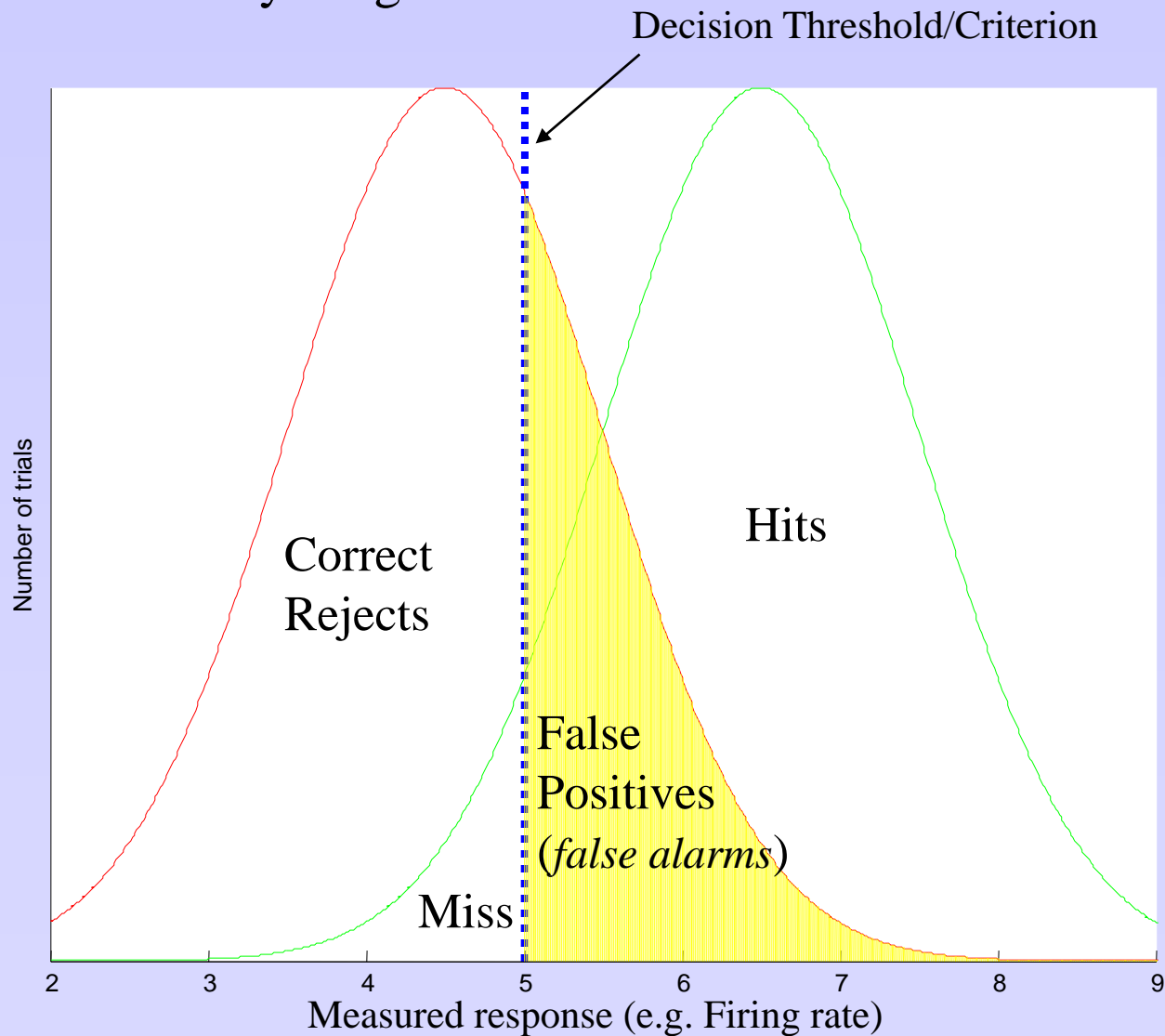


$$d' = \sqrt{2} \frac{|\mu_2 - \mu_1|}{\sqrt{\sigma_1^2 + \sigma_2^2}}$$

- Other issues: different amplitudes, bimodal, non Gaussian....

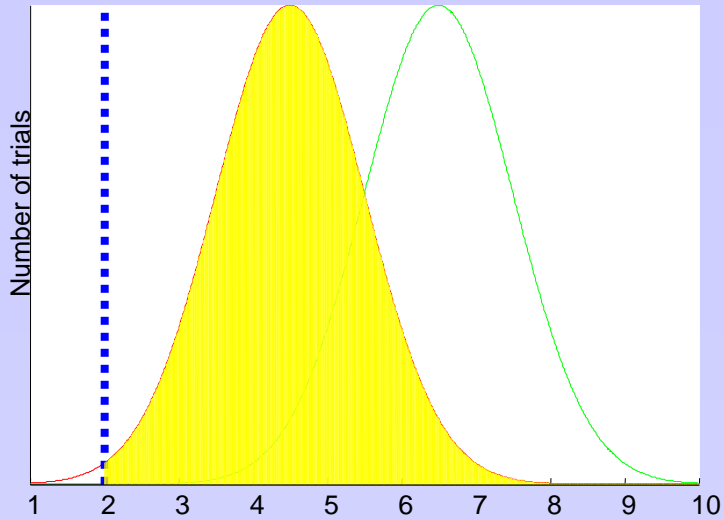
# Signal Detection: ROC curves

- More general approach...
- Which spike trains carry a signal?

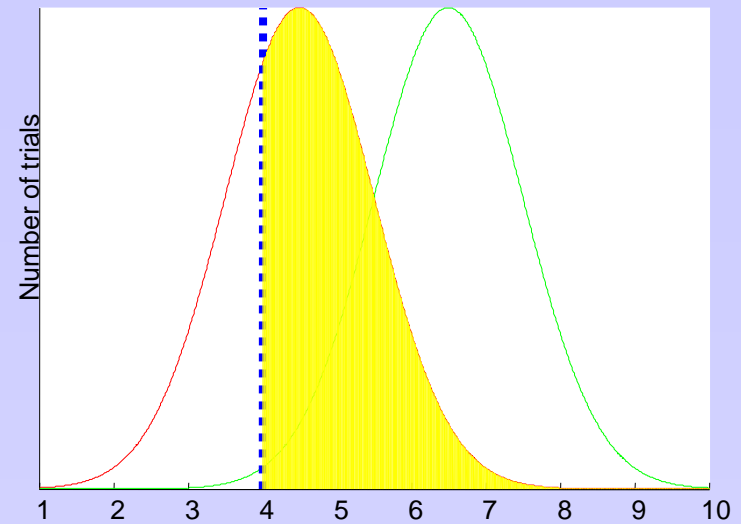


# ROC curves: Changing the detection threshold

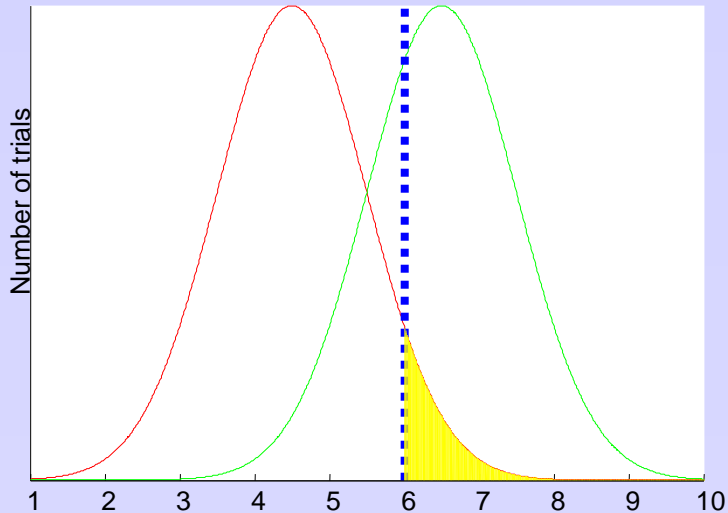
Hits=100%, False Alarms=99%



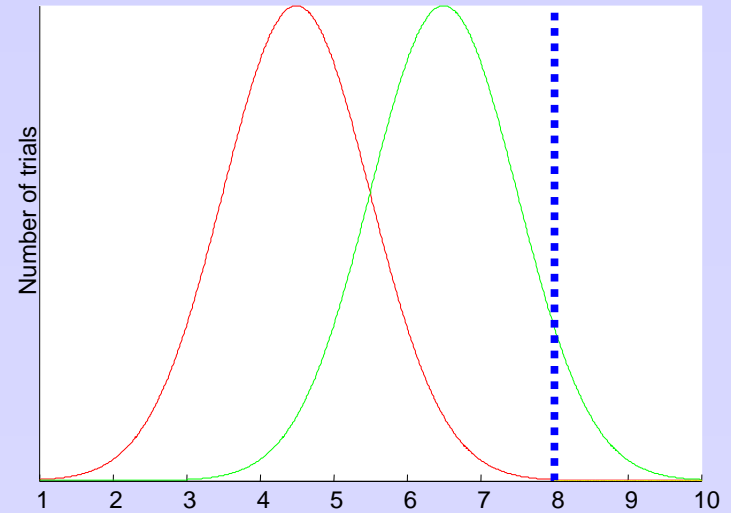
Hits=99%, False Alarms=69%



Hits=69%, False Alarms=6.6%

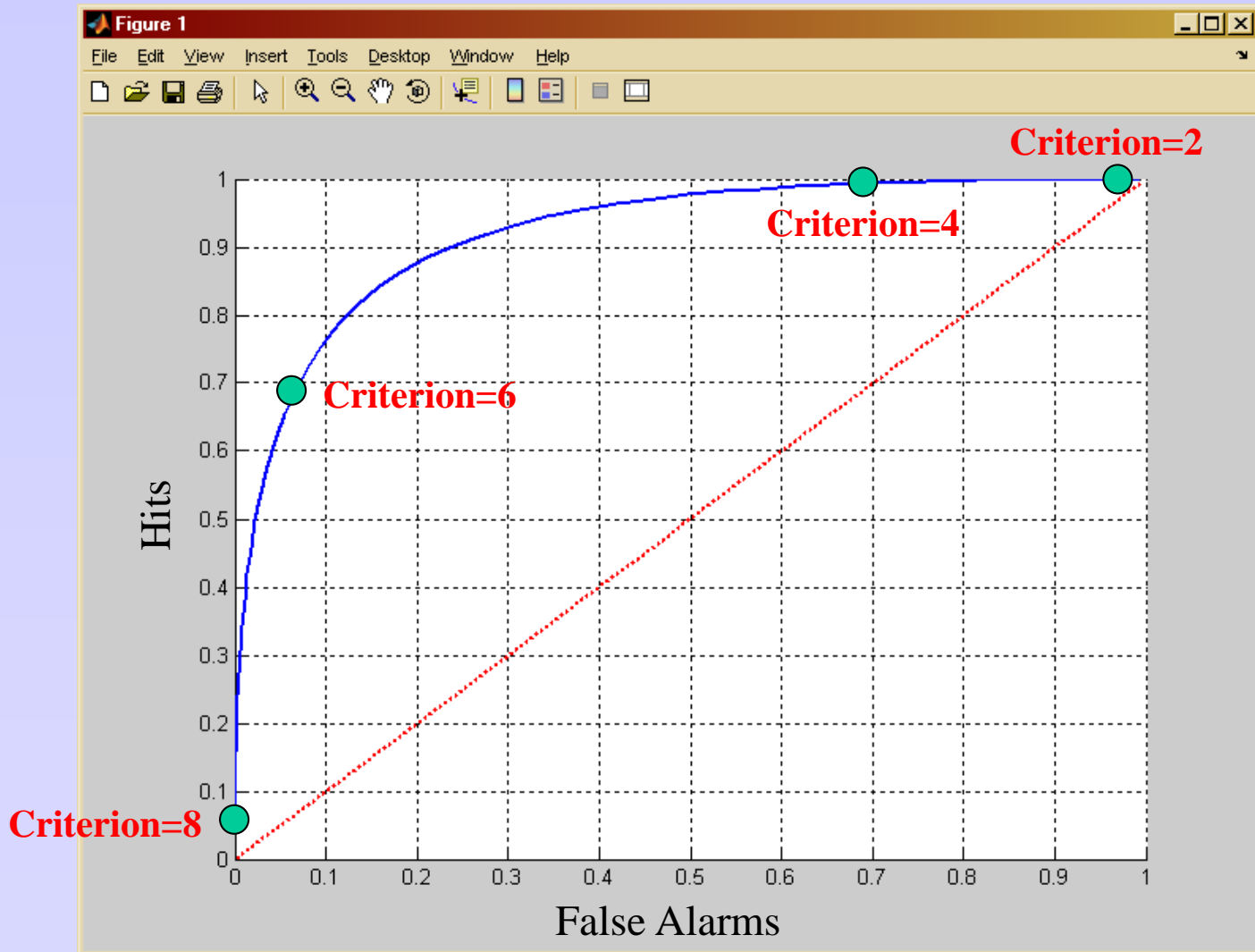


Hits=6.6%, False Alarms=0%



# ROC curves

- The **R**eceiver **O**perating **C**haracteristic curve
- In general, can be built empirically (no need to know the distributions)

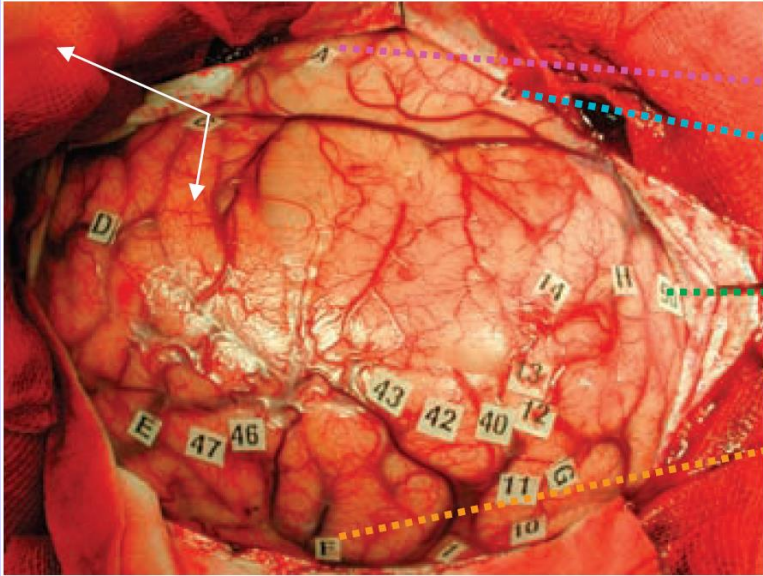


# ROC curves: Fun facts...

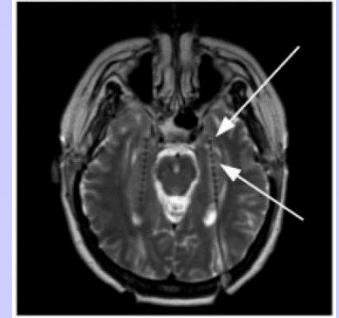
- Tradeoff between Hits and False alarms  $\Leftrightarrow$  Tradeoff between Misses and Correct Rejects.
- The closer the curve follows the left-hand border and then the top border of the ROC space, the more accurate the 'test'.
- The closer the curve comes to the 45-degree diagonal of the ROC space, the less accurate the 'test'. Diagonal=chance.
- The slope of the tangent line at any point gives the *likelihood ratio* (LR) for that value of the test.

*'how much better will my hit rate be, if I tolerate a few more false positive'*
- The area under the curve is a distribution-free measure of performance.

# Face and object cells in humans



(Thiebaut de Schotten et al 2005)



(Trautner et al, 2004)



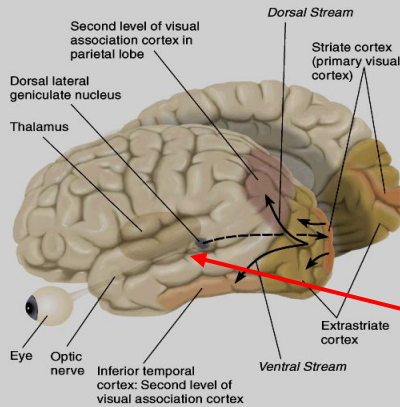
Chronic implants



Psychophysical testing  
Faces, landmarks, animals, objects

Medial Temporal  
Lobe

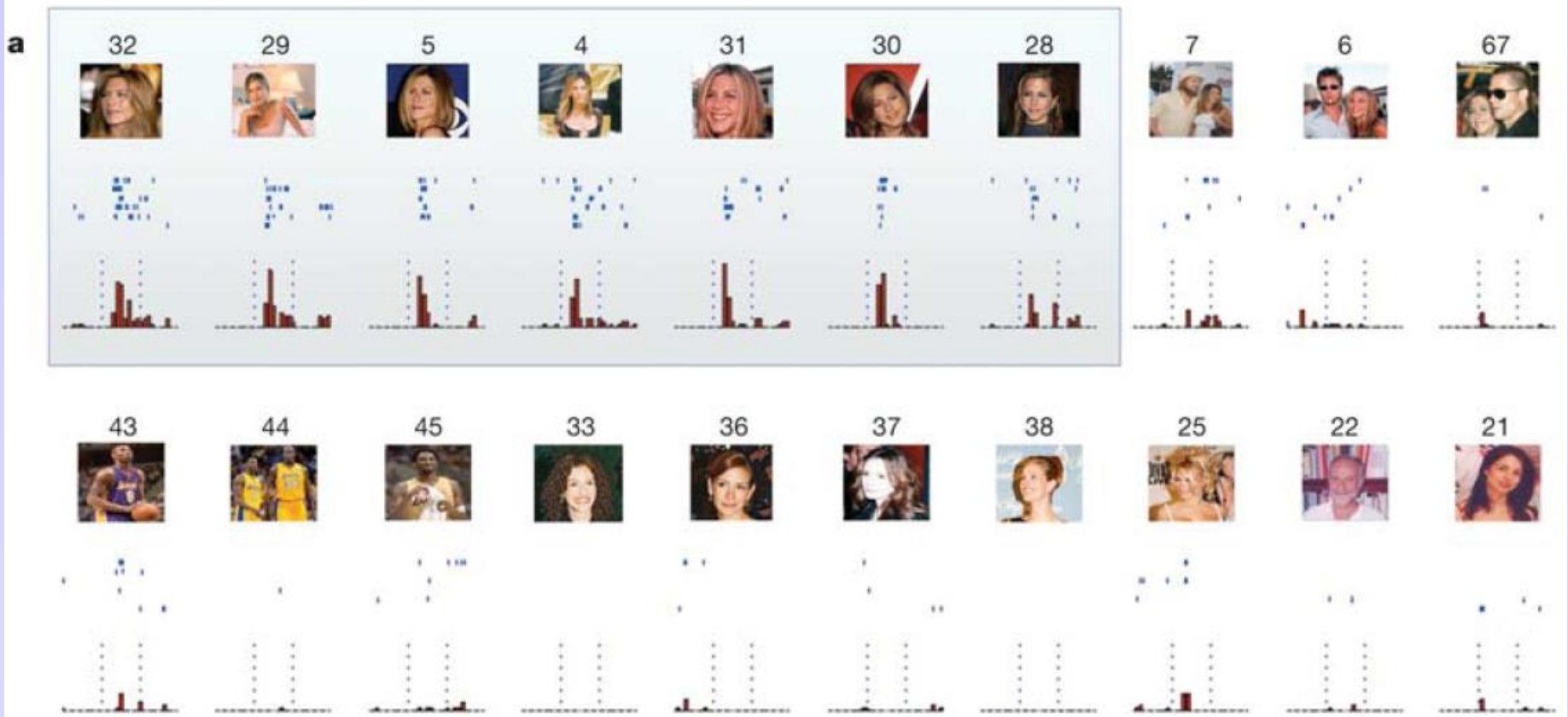
## ► The Human Visual System



# 'person' selective cell

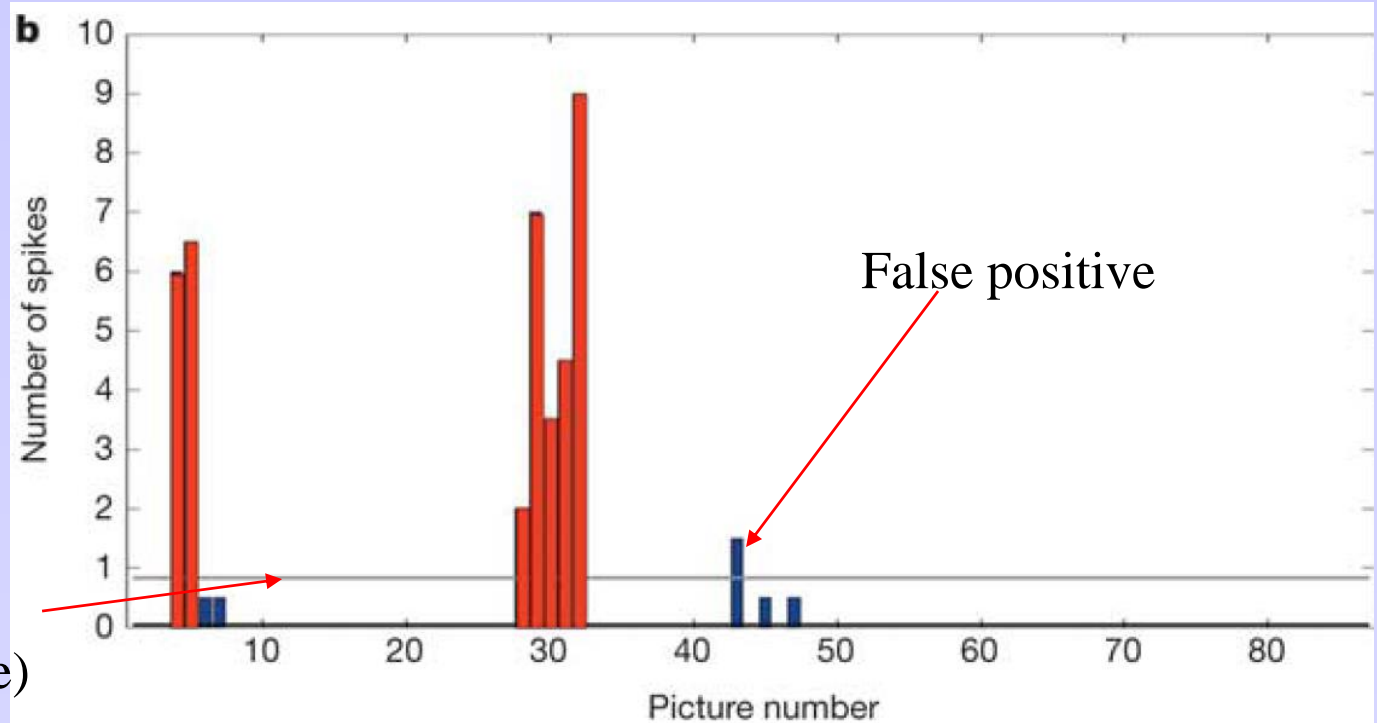


Left posterior hippocampus. 30/87 images





# 'person' selective cell



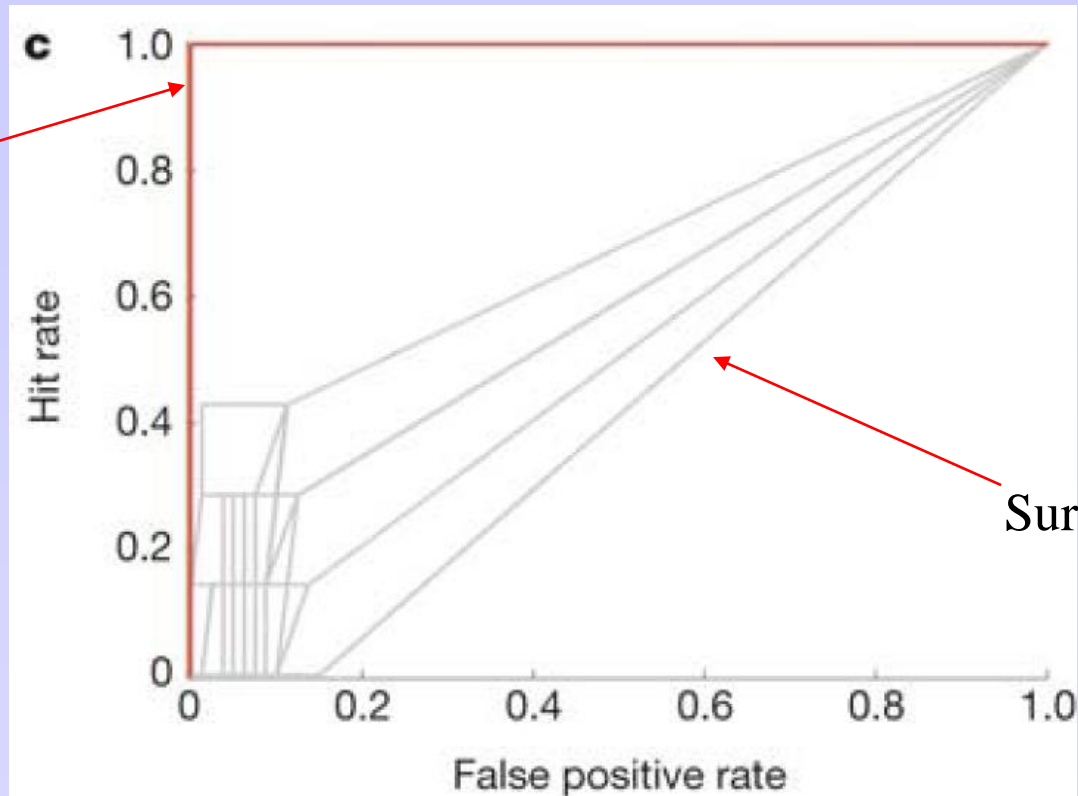
Detection criterion:  
Noise=baseline

$D_c = \text{Noise} + 5 \cdot \text{std}(\text{noise})$

(Quiroga et al, 2005)

# 'person' selective cell

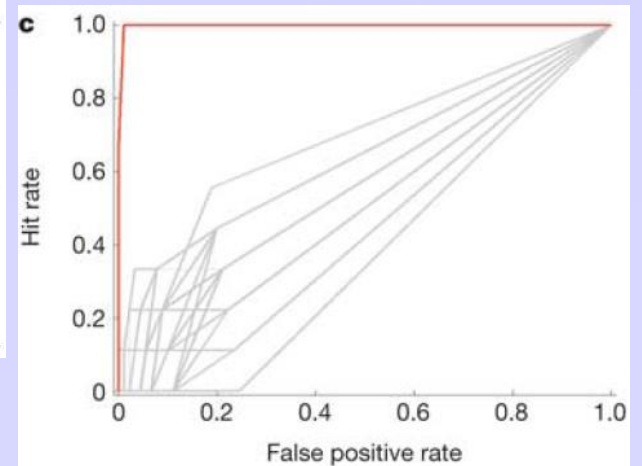
J. Aniston  
alone



Surrogate sets

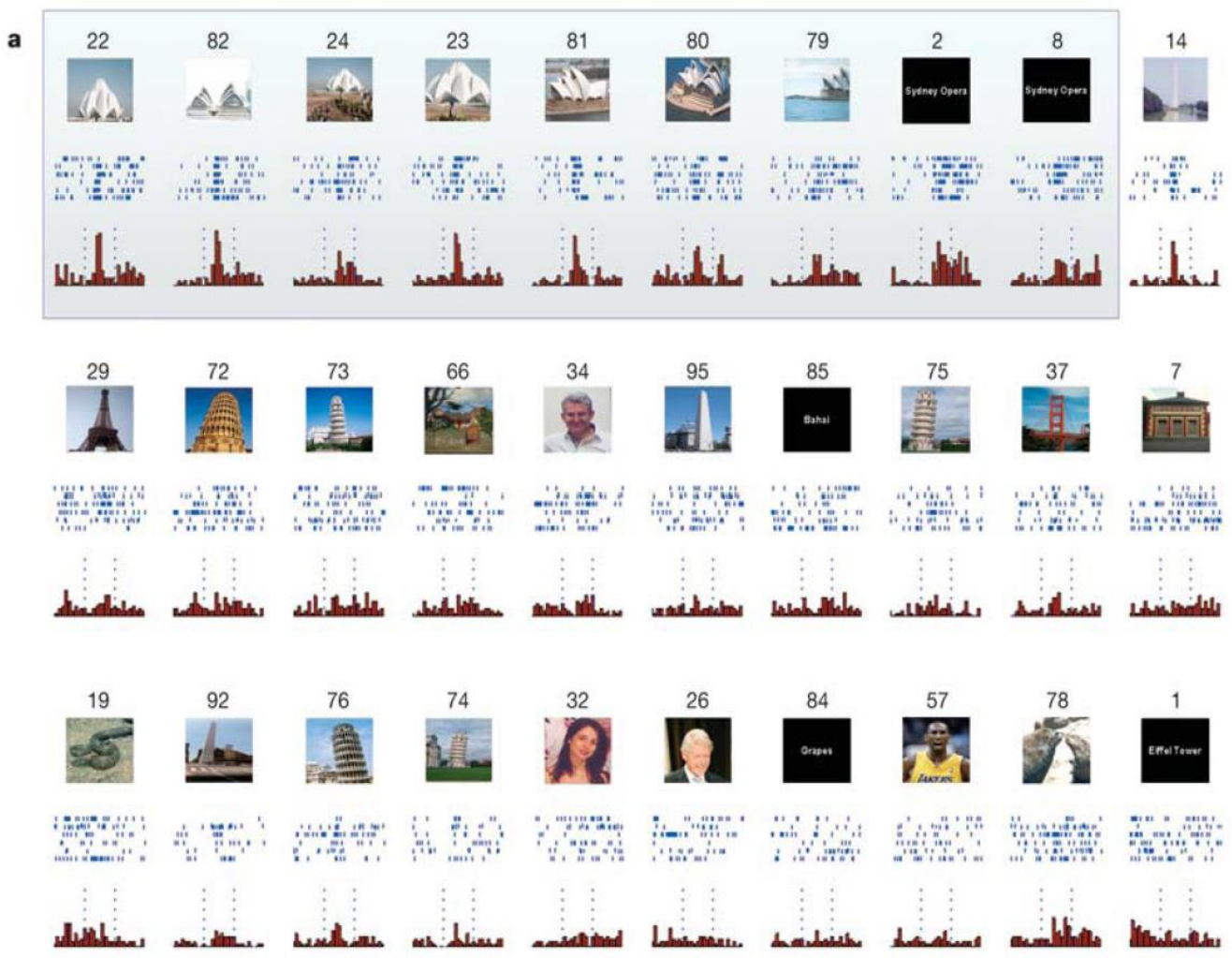
(Quiroga et al, 2005)

# 'person' selective cell

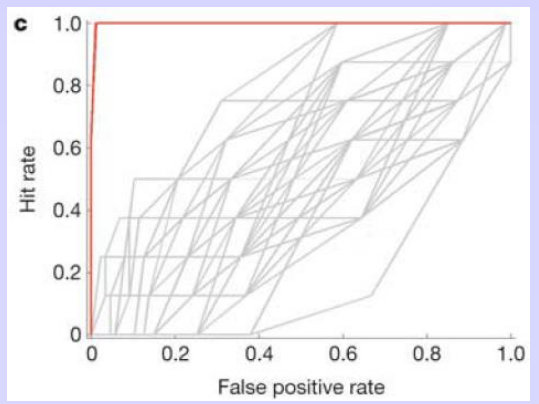


(Quiroga et al, 2005)

# 'Landmark' selective cell



(Quiroga et al, 2005)



Note: task = 'face?' or 'no face?'

# Homework 4: ROC

- Create a function that generate a Gaussian distribution of 1000 random values between 1 and 10, and a STD less than 4. Plot two such distributions with about 30% overlap. Give the value of  $d'$ .
- Write a function that takes a decision threshold (between 0 and 10) and computes the fraction of false alarms and hits.
- Plot the ROC curve, and quantify its departure from the diagonal.