

Neurobiology HMS 130/230  
Harvard/GSAS 78454

Visual Object Recognition

## **Primary Visual Cortex**

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# Outline

**Visual system**

**Anatomy and  
physiology**

**Functional organization**

**Receptive field models**

**Neural populations**

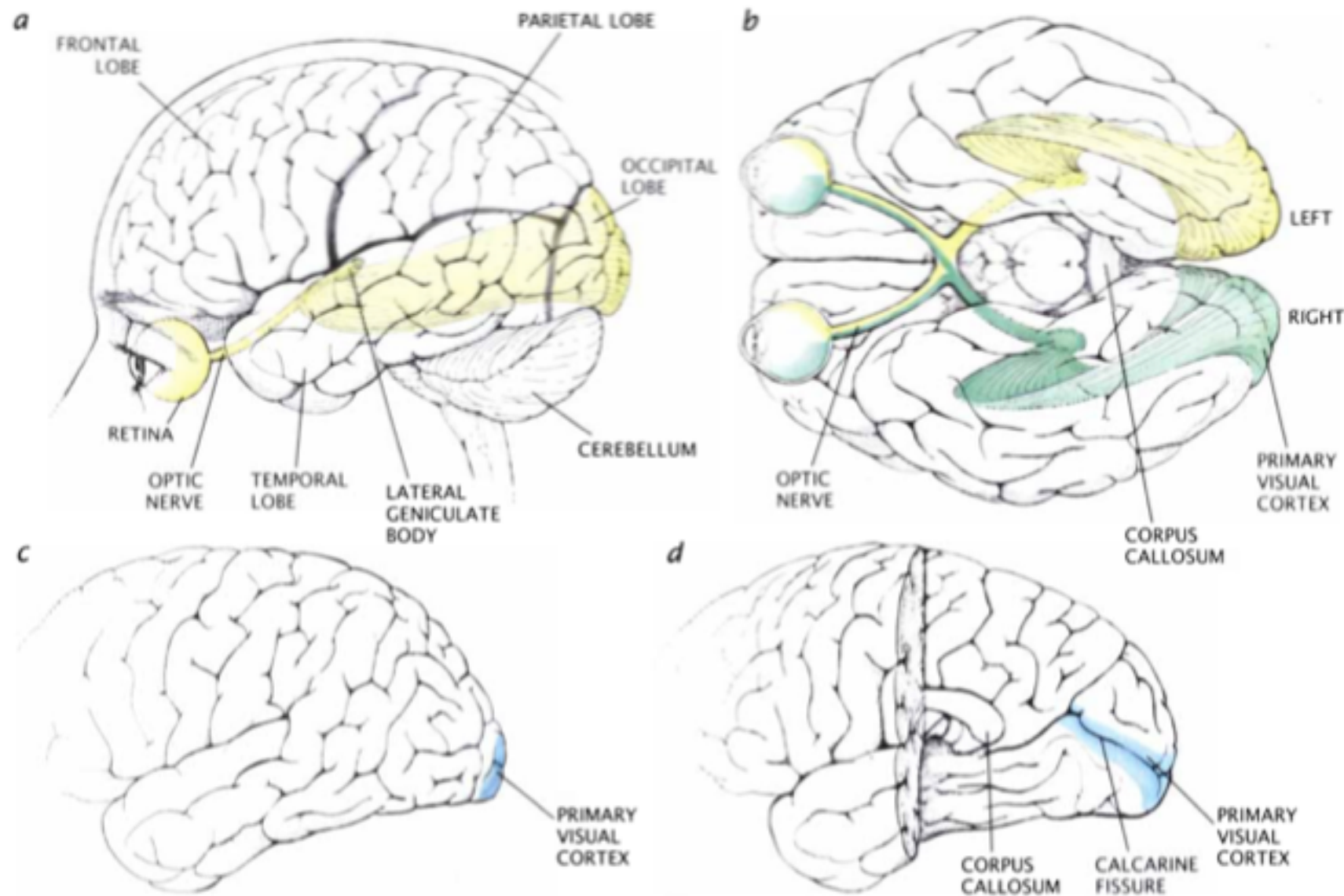
**Neural correlates of  
behavior**

**The Unknown**

**Computation:**

**How does the brain  
make us see?**

## From the retina to the cortex



# Studies of wounded soldiers revealed topographic visual deficits



Russo-Japanese War of 1904-5

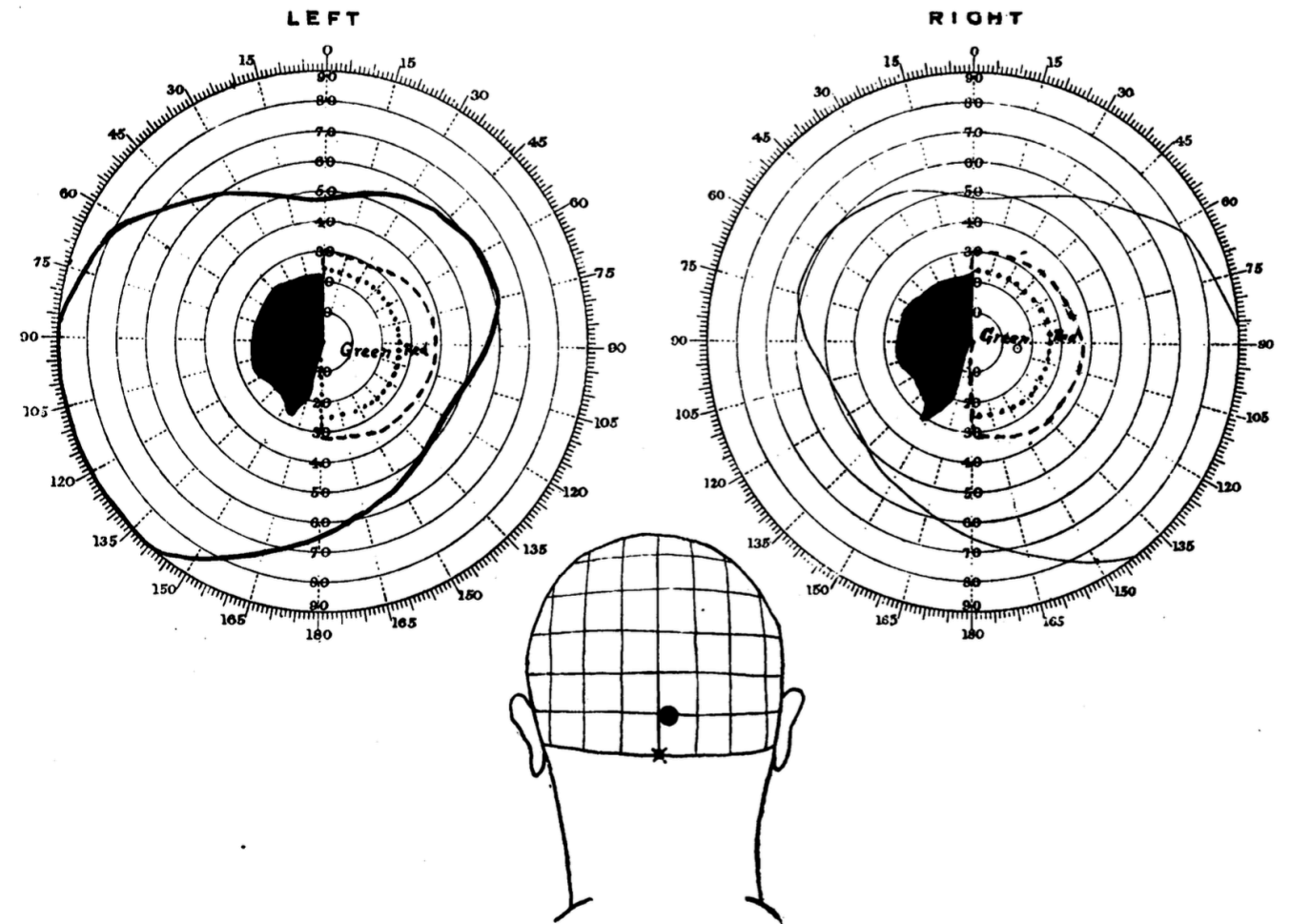


FIG. 1.



Acuity is much higher near the fovea

Fixate here



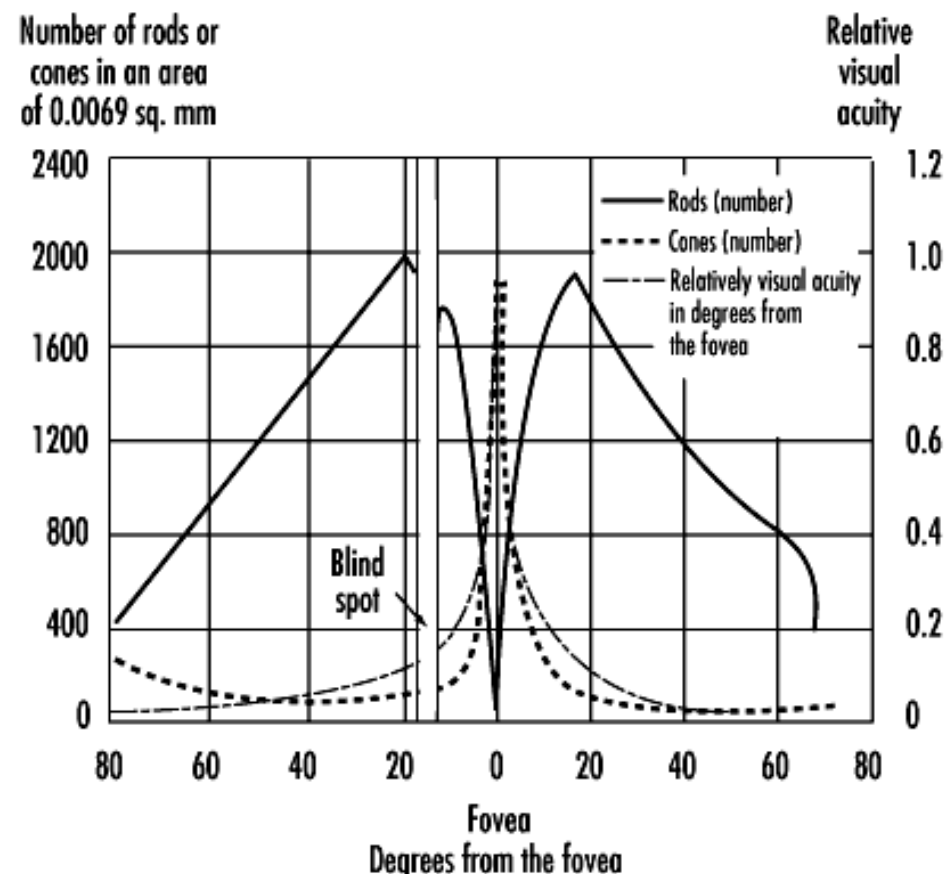
**TRY READING THIS [44]**  
**Retinal photoreceptor density [36]**

**Cortical magnification factor [28]**

**Why is it that we do not see things upside down? [20]**

Or the split between the two hemifields? [12]

And do not forget about the importance of crowding! [8]

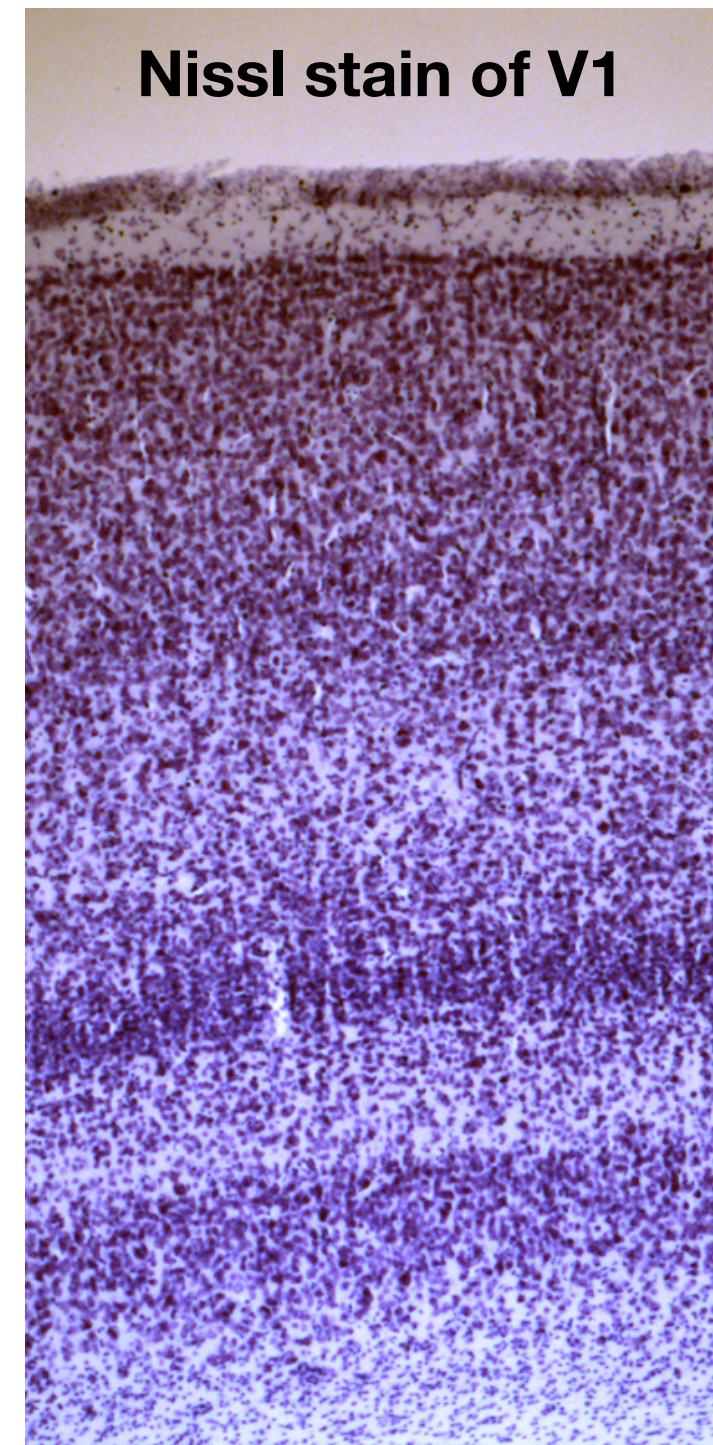


Vision is **deceptively unlike**  
camera photography

# The complex circuitry of the cortex



GOLGI-STAINED NERVE TISSUE from the visual cortex of a rat was sketched by Cajal in 1888. The numbers along the right-hand margin identify cellular layers; the capital letters label individual neurons. One of Cajal's most important contributions to neurobiology was to establish the neuron as a discrete, well-defined cell rather than as part of a continuous network.



Layer  
1  
2  
3  
4  
5  
6

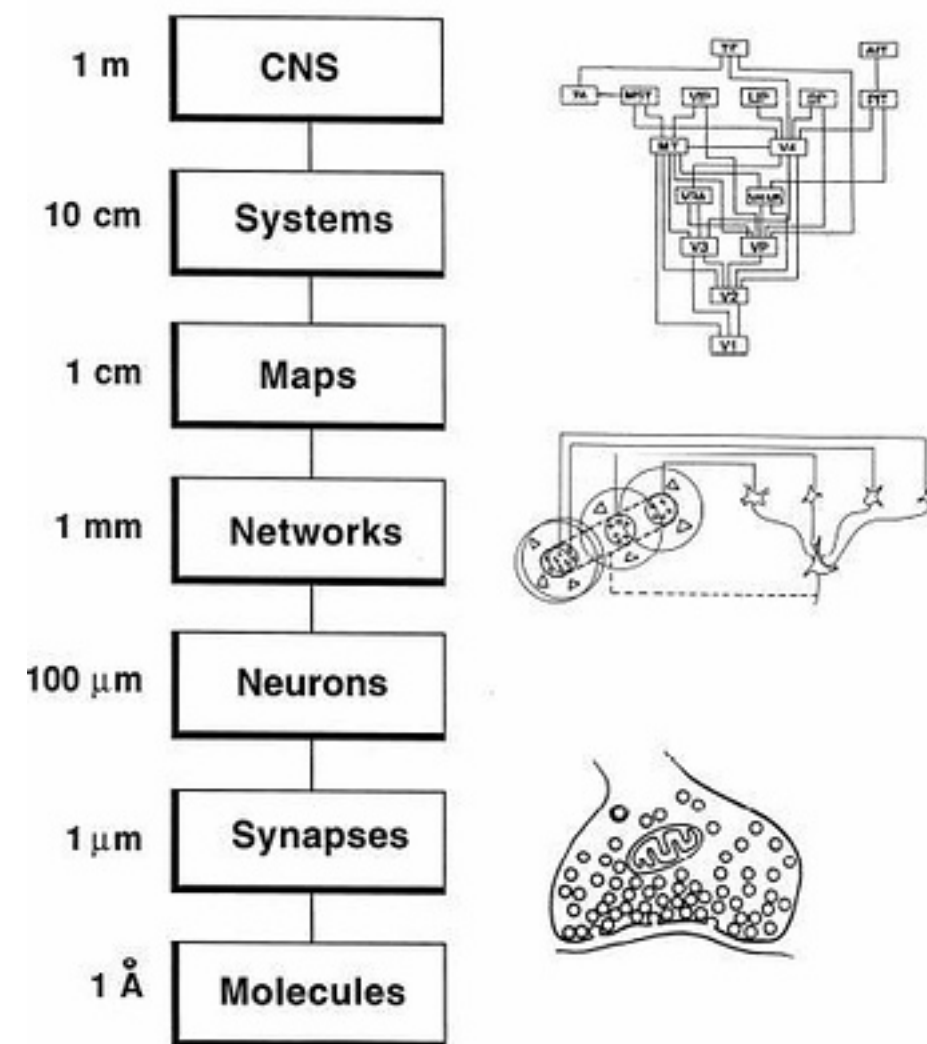


0.5 mm

Ramón y Cajal (1852–1934)



# Spatial scales of the nervous system

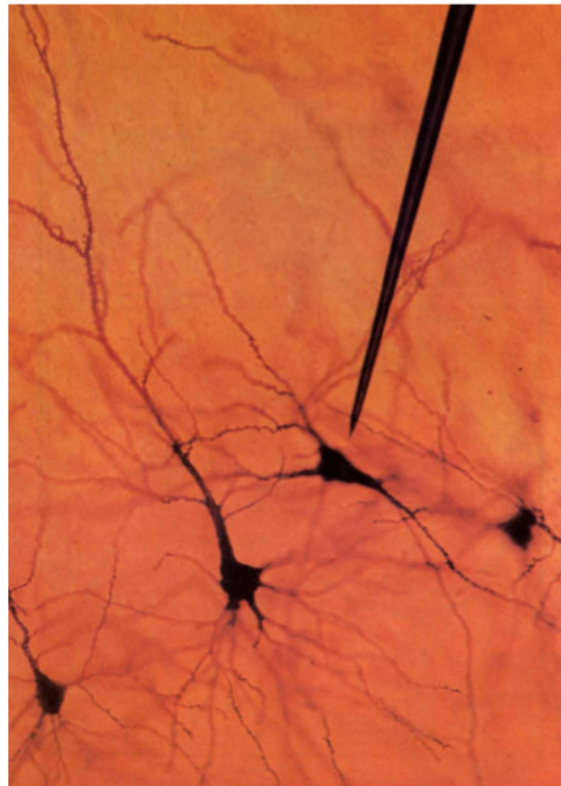


Churchland and Sejnowski, 1992

# Physiological access using the microelectrode

Edgar Adrian 1926

Neuronal resolution  
Sub-millisecond temporal resolution  
Direct examination of action potentials

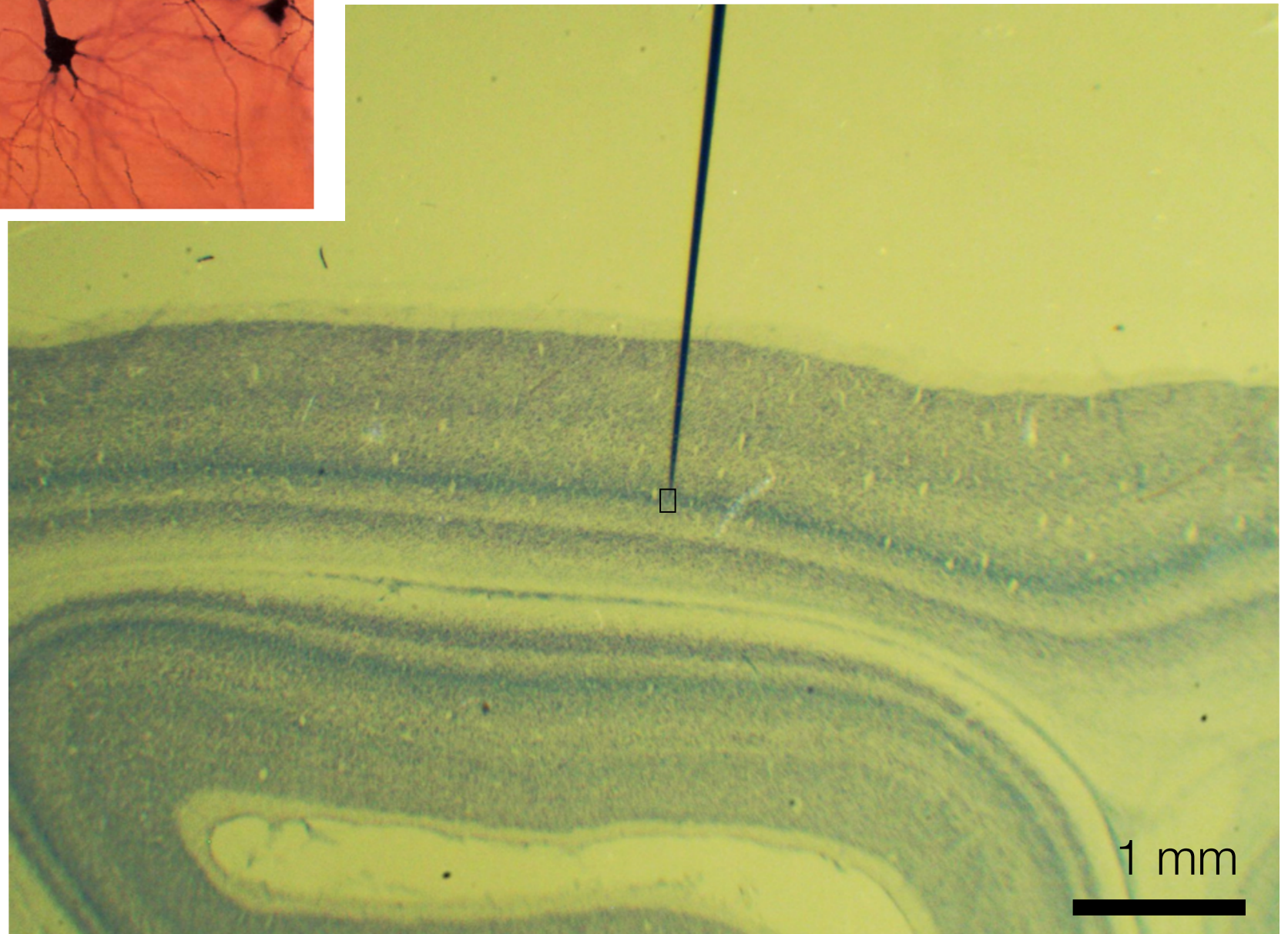


Hubel, D. (1979). The Visual Brain.  
SCIENTIFIC AMERICAN 241, 45-53.



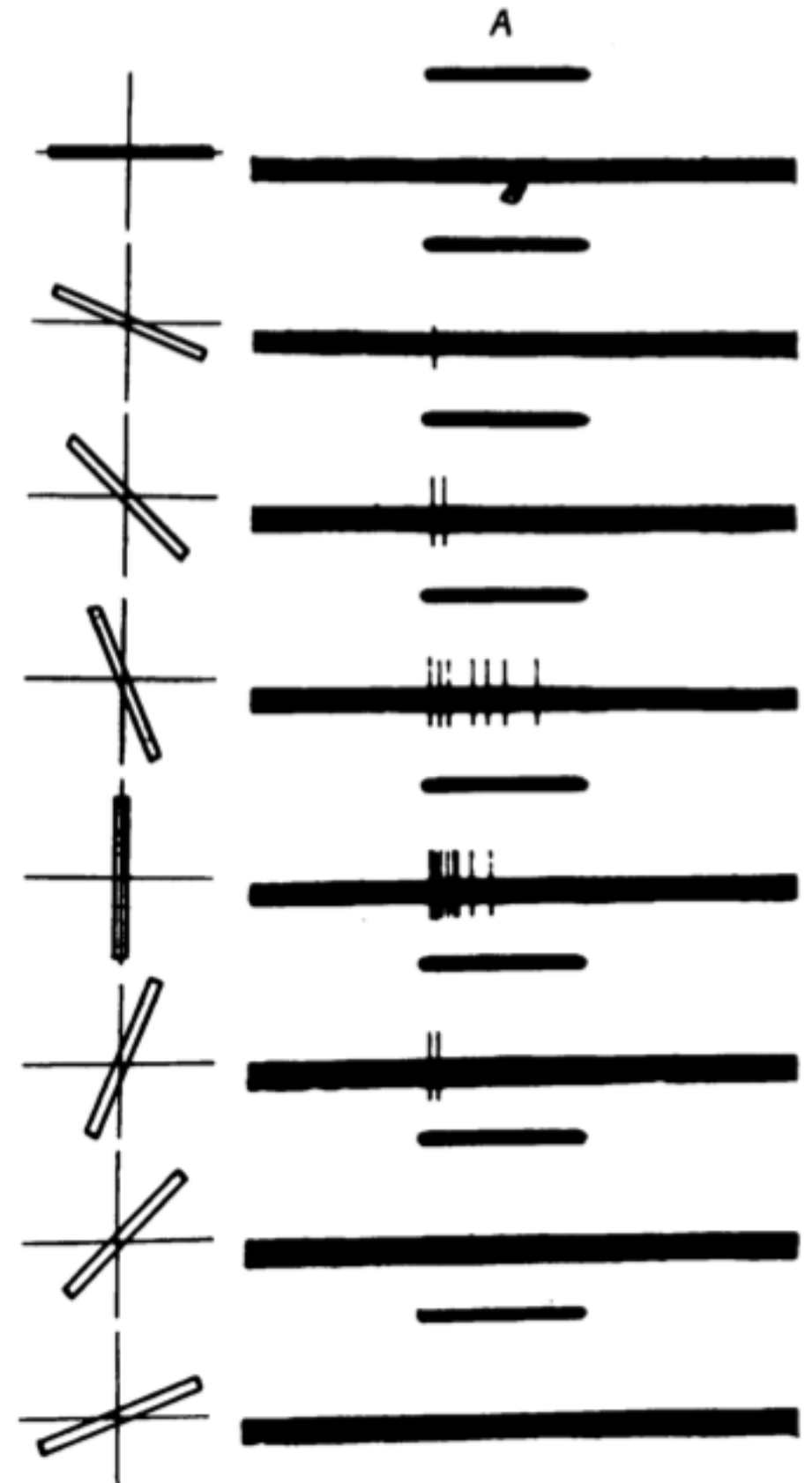
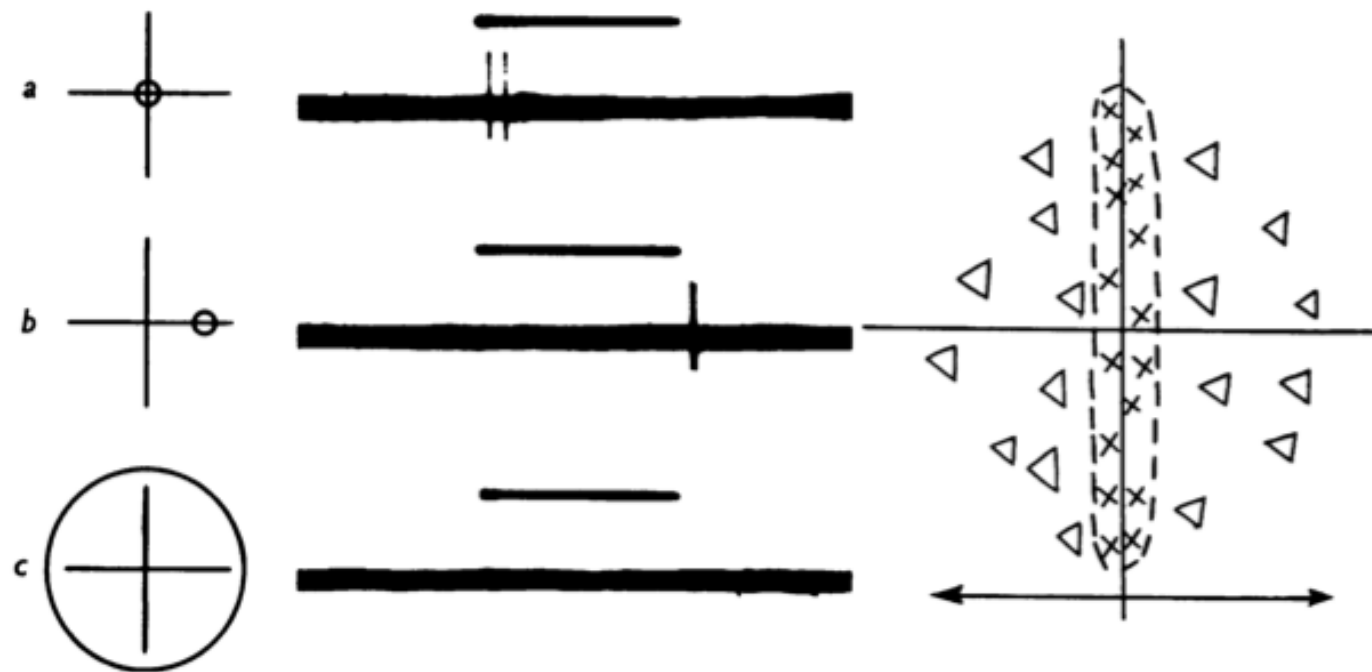
Hubel

Wiesel



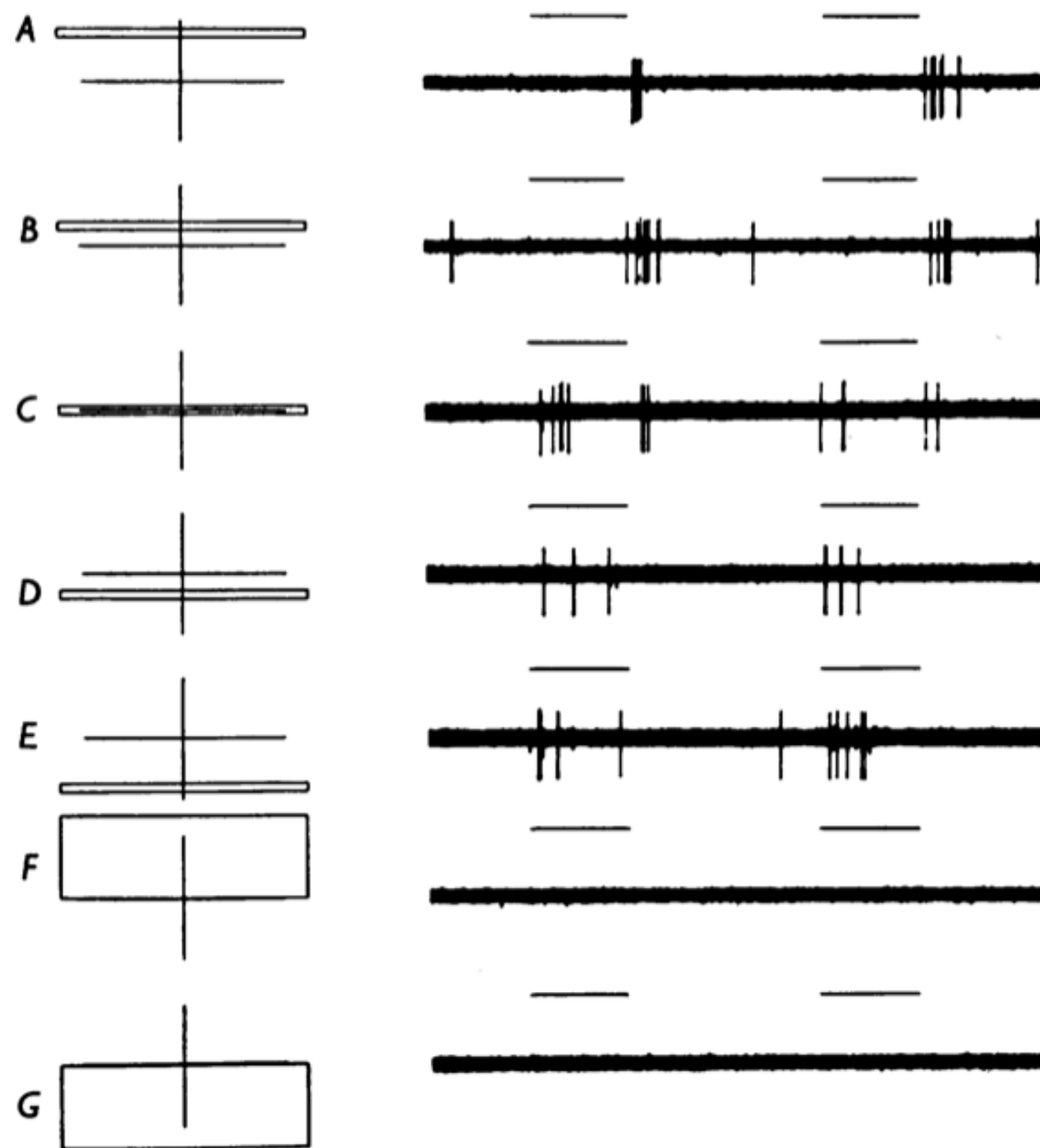
# Electrophysiological recordings from V1

## Orientation selectivity of simple fields





# Selectivity and tolerance of complex fields

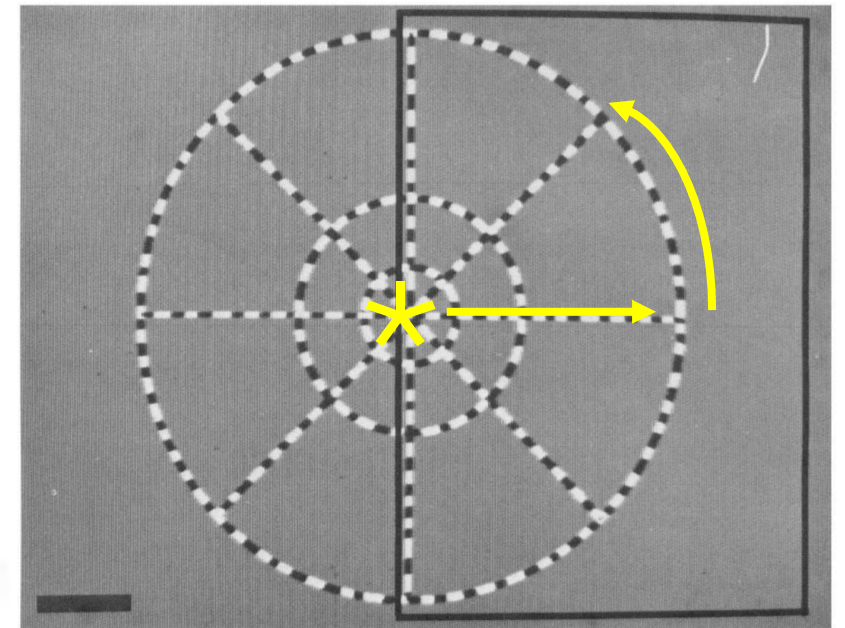


# Hubel and Wiesel mapping V1 neurons

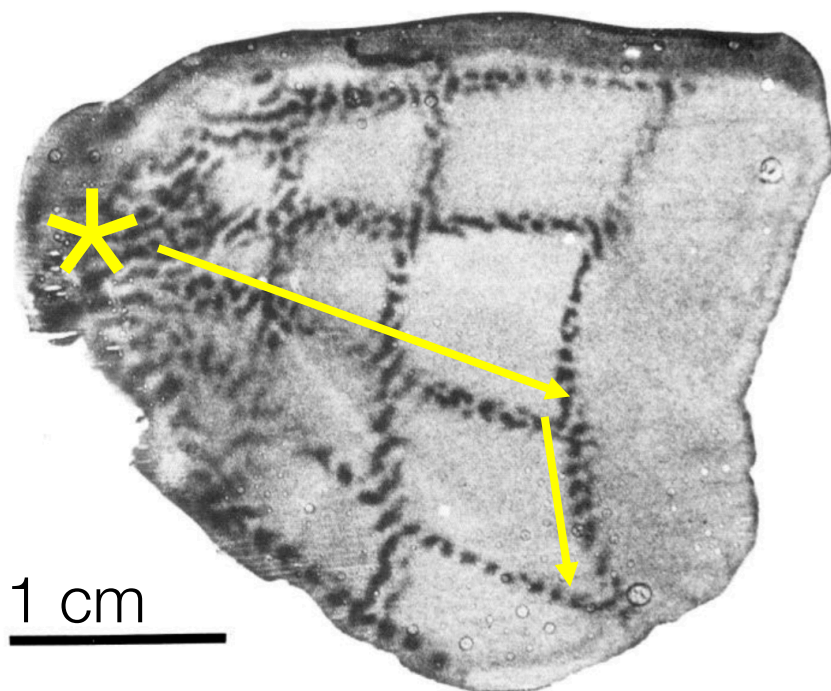
[www.youtube.com/watch?v=8VdFf3egwfg](http://www.youtube.com/watch?v=8VdFf3egwfg)

# Retinotopical map in the cortex

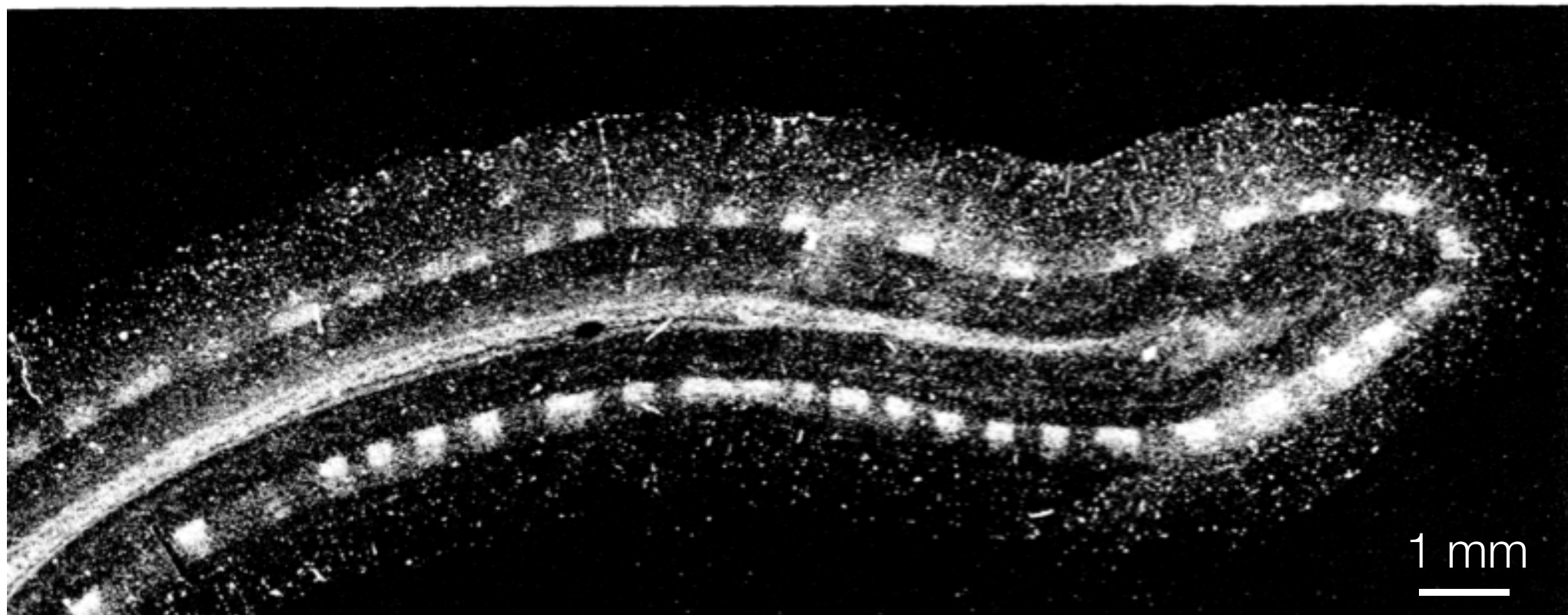
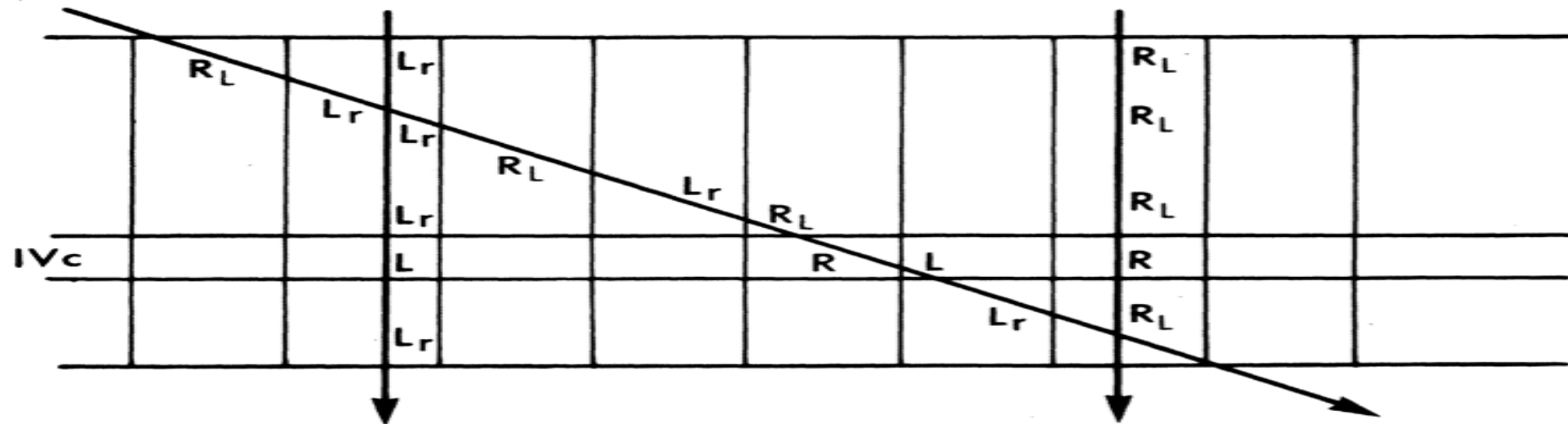
## Functional organization



Left hemisphere V1



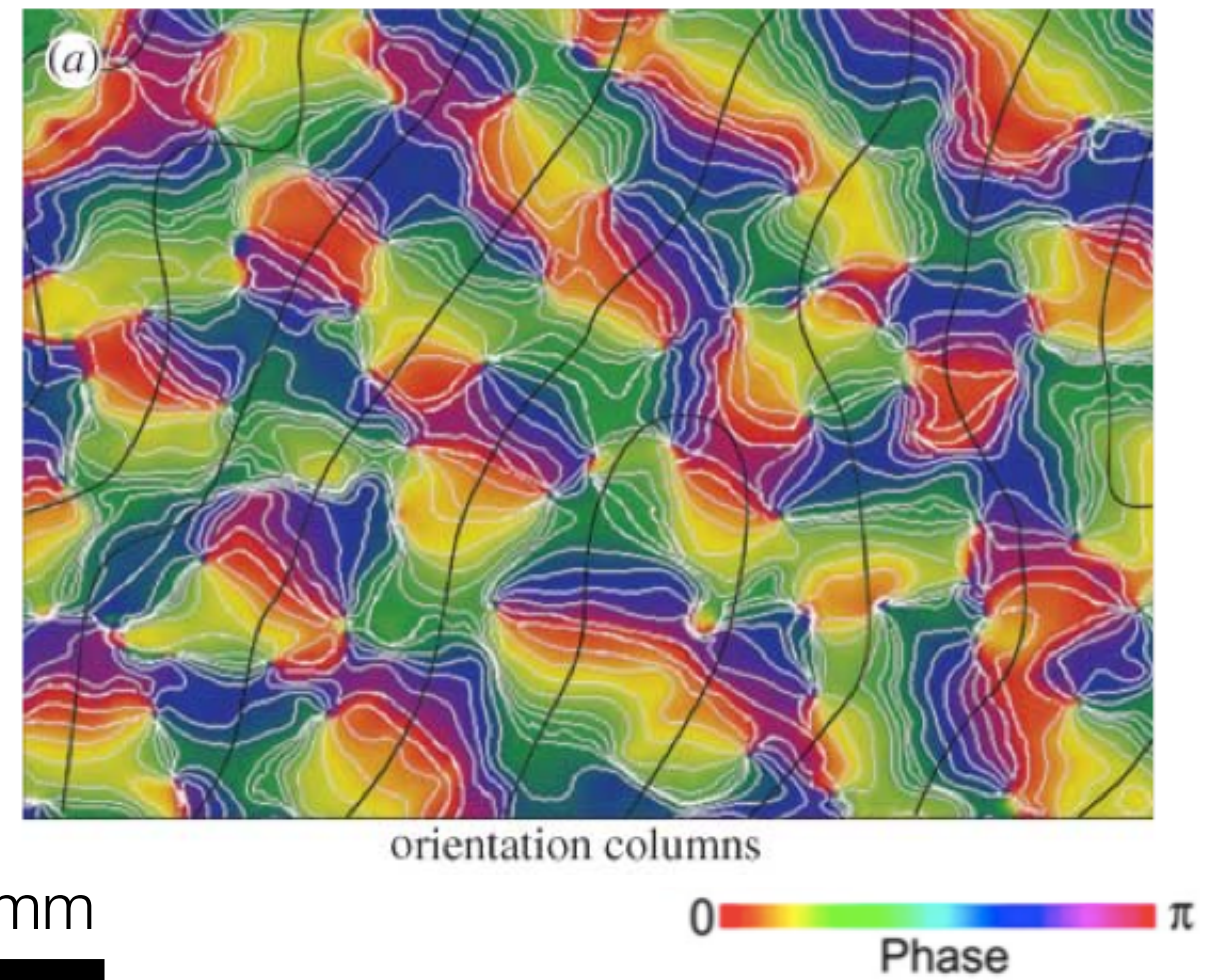
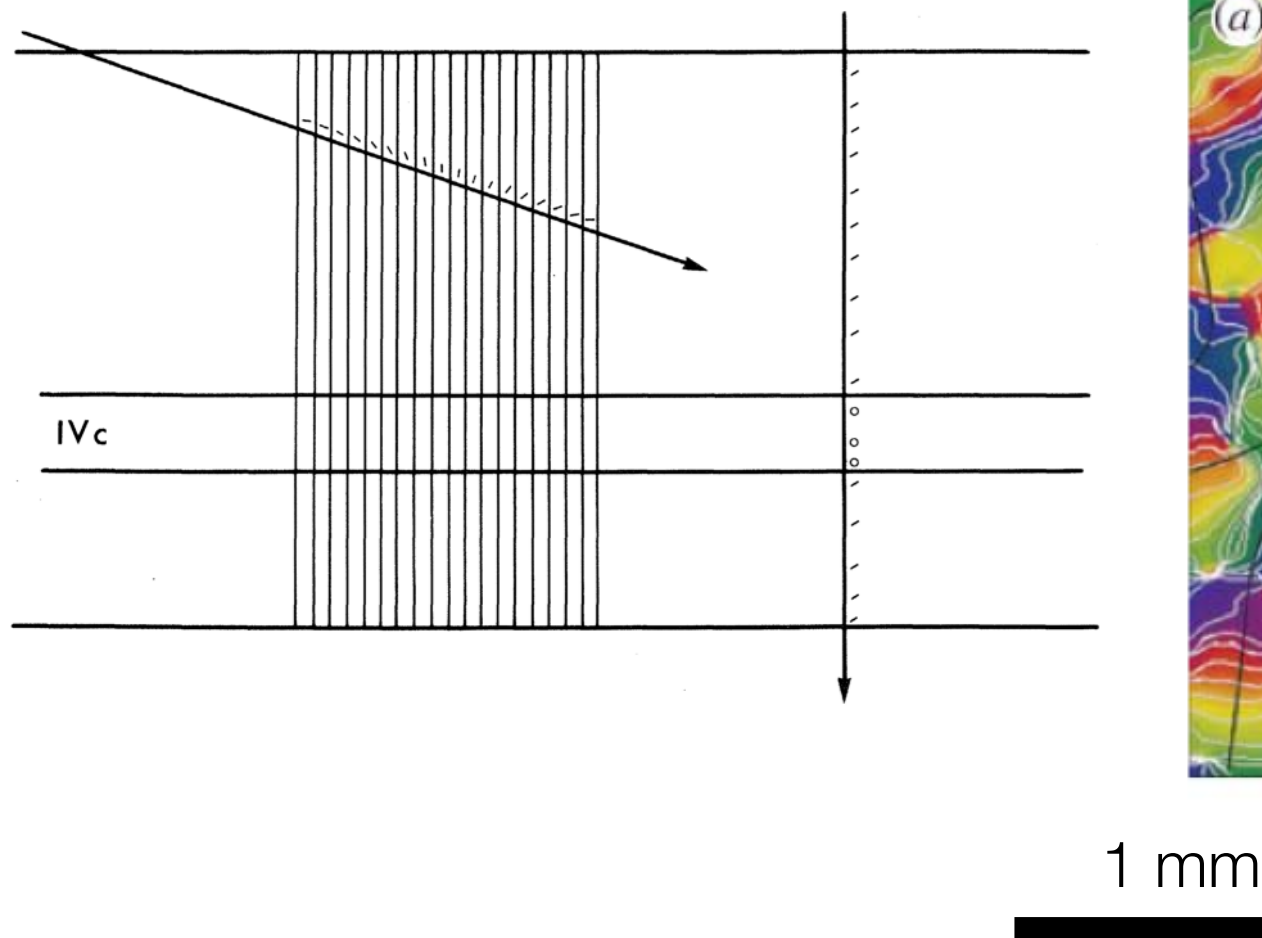
# Ocular dominance columns



Hubel & Wiesel, *Proc. R. Soc. Lond. B*, 1977



# Visual orientation columns

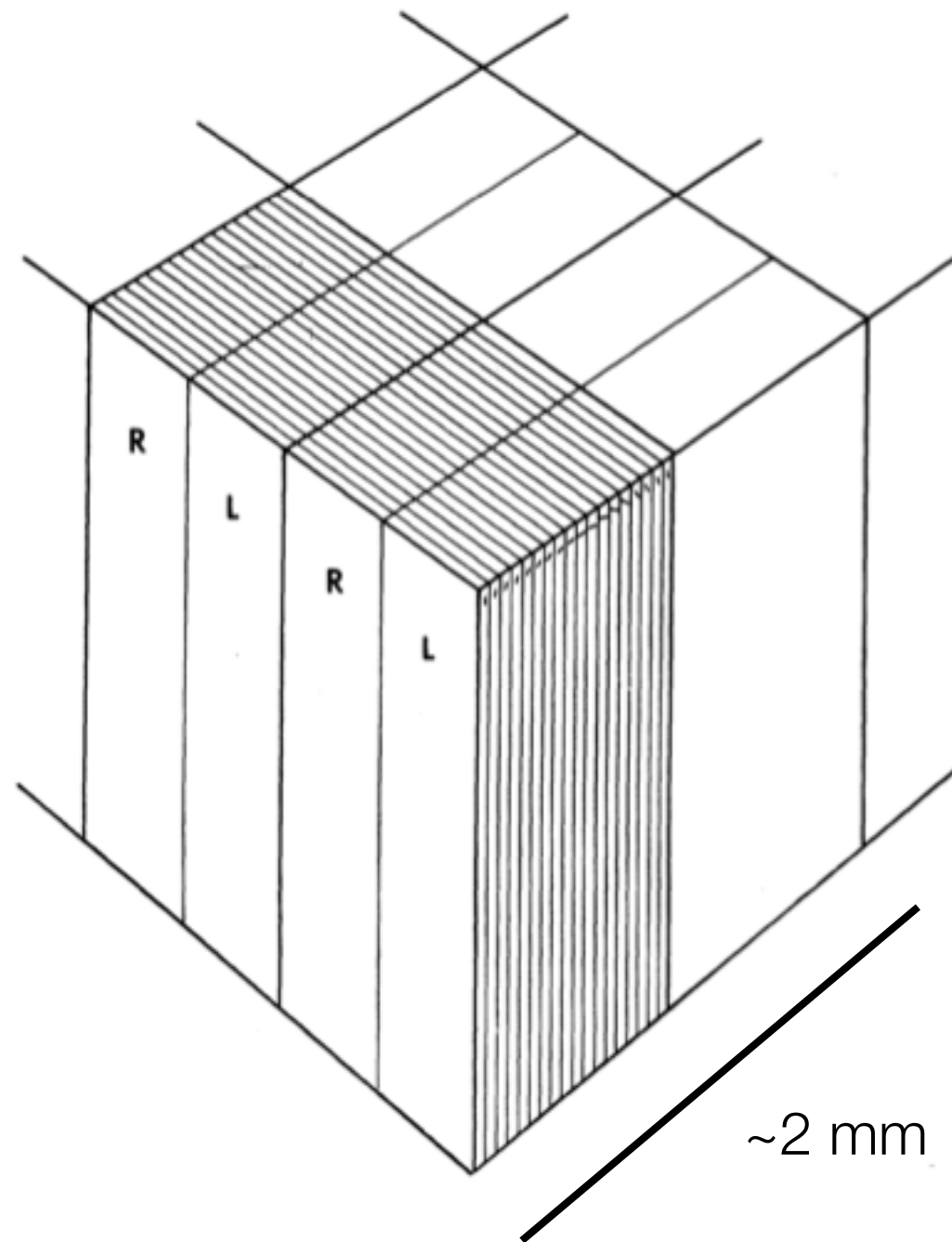


Hubel & Wiesel, *Proc. R. Soc. Lond. B*, 1977

Horton & Adams, *Phil. Trans. R. Soc. B*, 2005



# Putting it all\* together: the “hypercolumn”



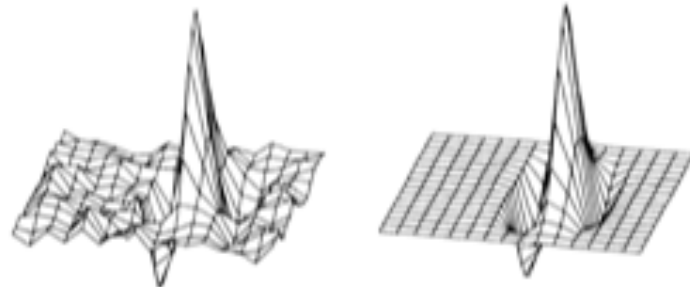
\*all is more than ocularity and orientation. Many V1 neurons are also selective for:

- Direction & speed
- Depth
- Color

# Mathematical description of a receptive field

## Receptive field models

Spatial receptive field      Gabor fit



Cell 1

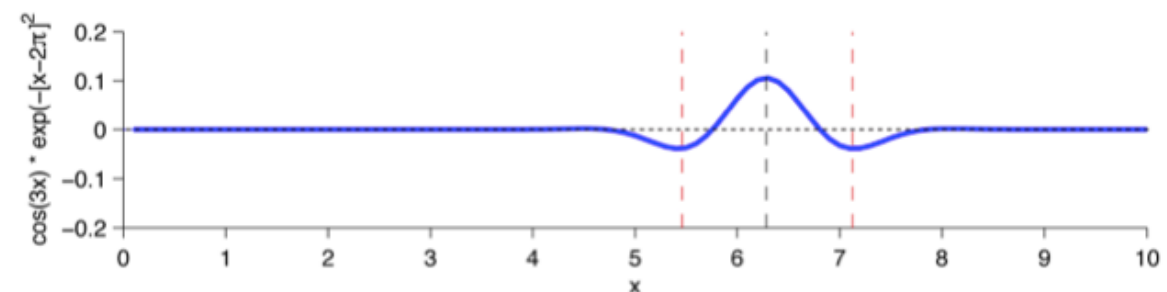
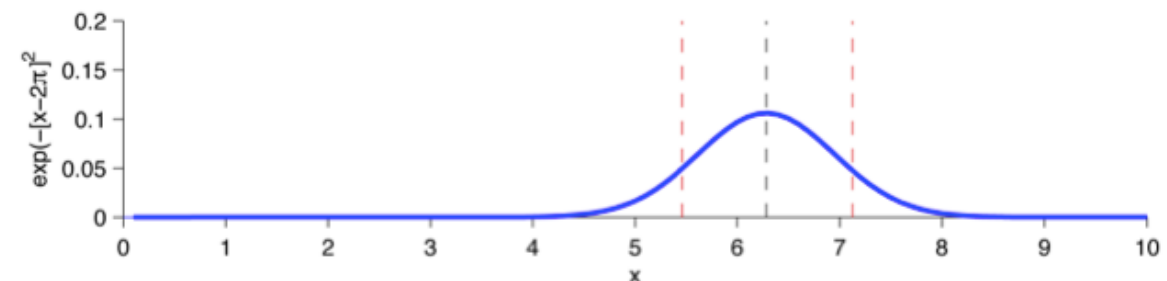
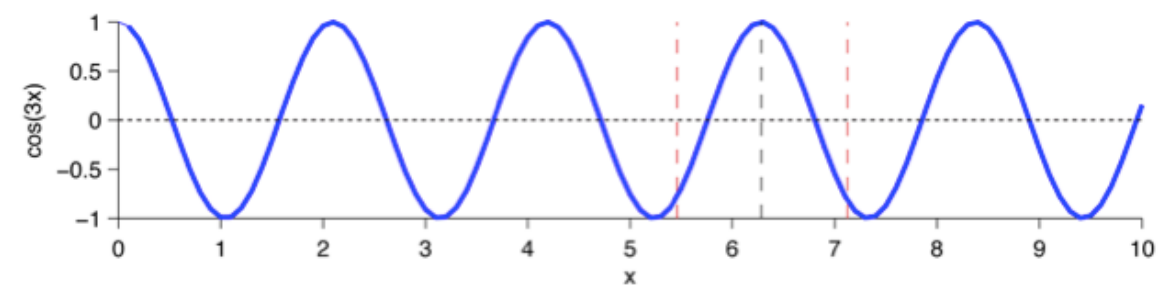


Cell 2

Spatial receptive field  
Cat primary visual cortex (area 17)  
Jones and Palmer 1987

Gabor function

$$D(x,y) = \frac{1}{2\pi\sigma_x\sigma_y} \exp\left[-\frac{x^2}{2\sigma_x^2} - \frac{y^2}{2\sigma_y^2}\right] \cos(kx - \phi)$$



# Interlude : MATLAB

An easy way to write computer code

- <http://www.mathworks.com/index.html>
- “High-level” computer programming language
- Quite powerful!

```
theta_rad=(2*pi/360)*theta;  
x=(-2*sigma_x):bin:(2*sigma_x);nx=length(x);  
y=(-2*sigma_y):bin:(2*sigma_y);ny=length(y);
```

```
% theta angle in radians  
% define x axis  
% define y axis
```

```
factor1=1/(2*pi*sigma_x*sigma_y);
```

```
for i=1:nx
```

```
    for j=1:ny
```

```
        curr_x=x(i)*cos(theta_rad)+y(j)*sin(theta_rad);
```

```
        curr_y=y(j)*cos(theta_rad)-x(i)*sin(theta_rad);
```

```
        factor2=exp(-curr_x^2/(2*sigma_x^2)-curr_y^2/(2*sigma_y^2));
```

```
        factor3=cos(k*curr_x-phi);
```

```
        Ds(i,j)=factor1*factor2*factor3;
```

```
    end
```

```
end
```

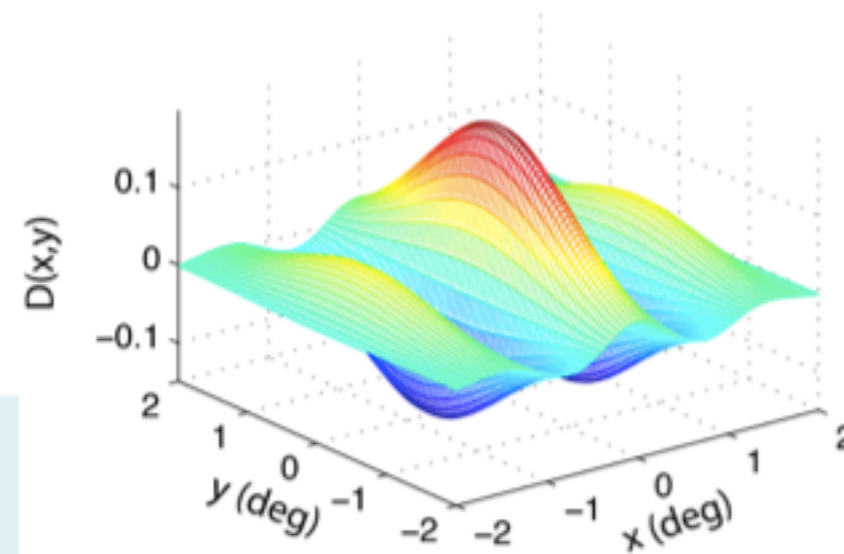
$$D(x,y) = \frac{1}{2\pi\sigma_x\sigma_y} \exp\left[-\frac{x^2}{2\sigma_x^2} - \frac{y^2}{2\sigma_y^2}\right] \cos(kx - \phi)$$

[ Time permitting ]

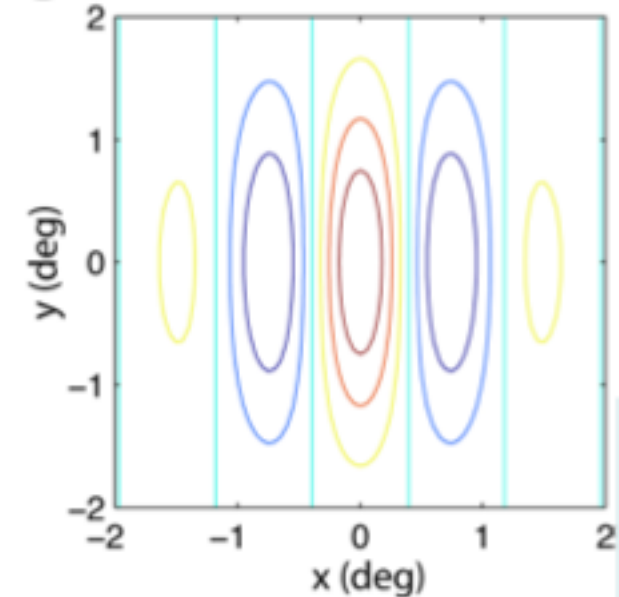
# Interlude : MATLAB

An easy way to make plots

**A**



**B**



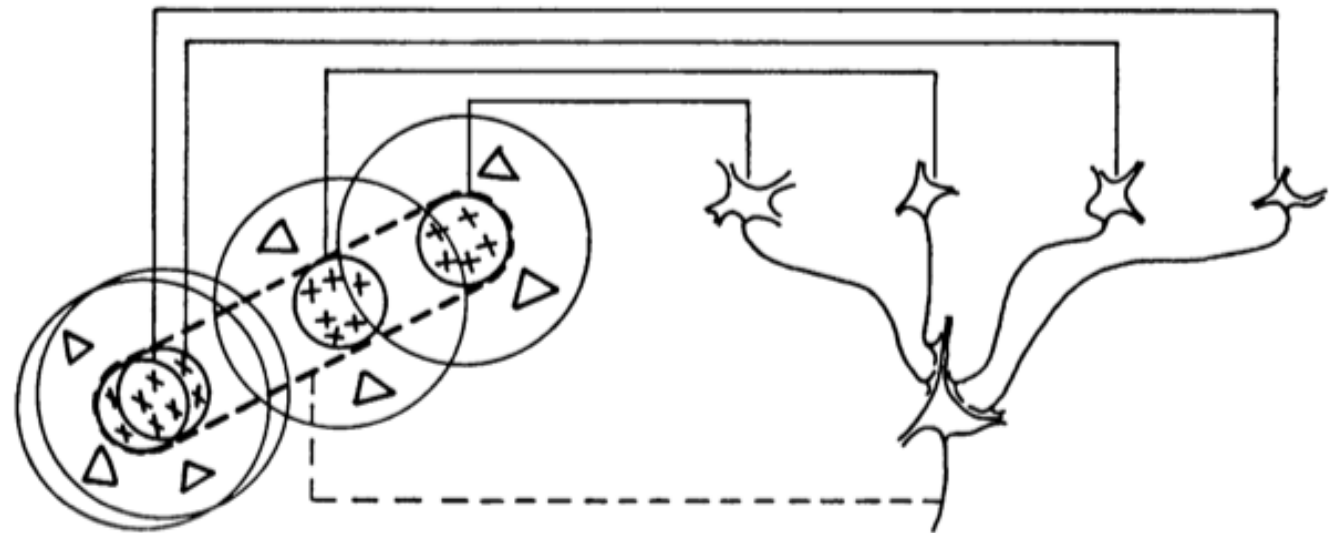
```
sigma_x=1;
sigma_y=1;
bin=0.05;
k=1/0.25;
theta=0;
i=0;
phi=0;
[Ds,x,y]=mygabor1(sigma_x,sigma_y,k,phi,theta,bin);
subplot(2,2,1);
mesh(x,y,Ds');
axis([min(x) max(x) min(y) max(y) min(Ds(:)) max(Ds(:))]);
subplot(2,2,2);
contour(x,y,Ds');
axis square;
```

$$D(x,y) = \frac{1}{2\pi\sigma_x\sigma_y} \exp\left[-\frac{x^2}{2\sigma_x^2} - \frac{y^2}{2\sigma_y^2}\right] \cos(kx - \phi)$$

# Stimulus “selectivity” and “tolerance”

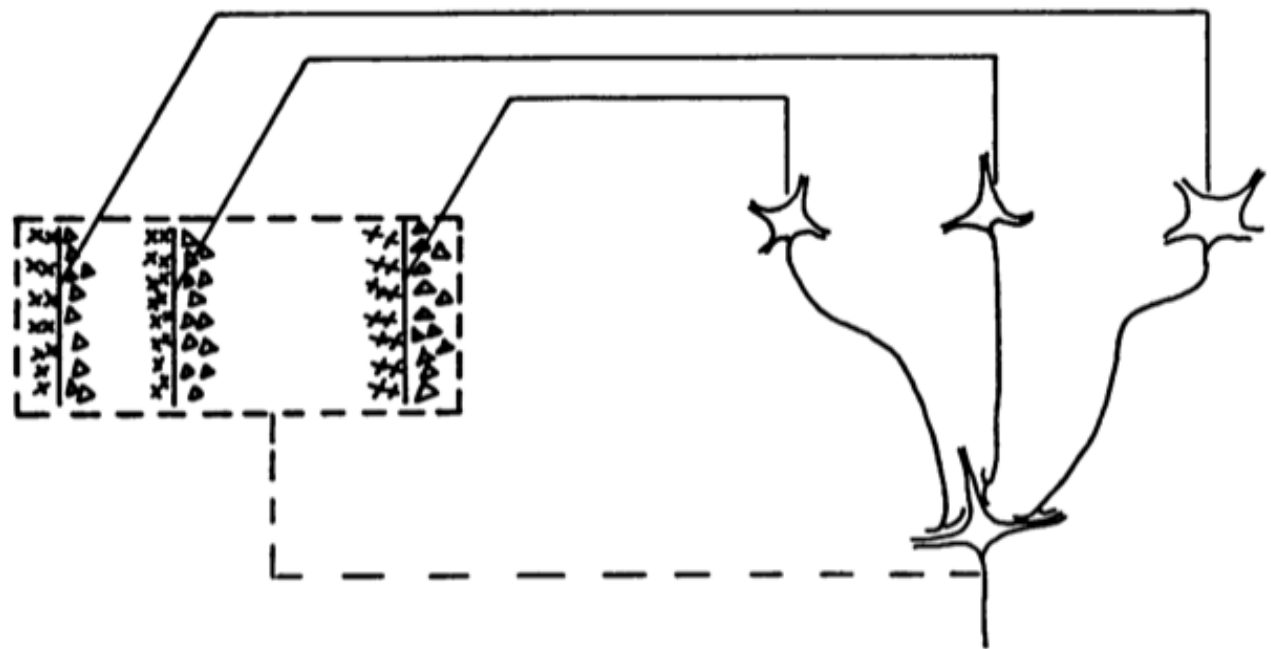
## Orientation selectivity of a simple cell:

boolean ‘AND’ operation  
over circular ON fields  
with *different positions*



## Position tolerance of a complex cell:

boolean ‘OR’ operation  
over simple fields with  
*same orientation preference*

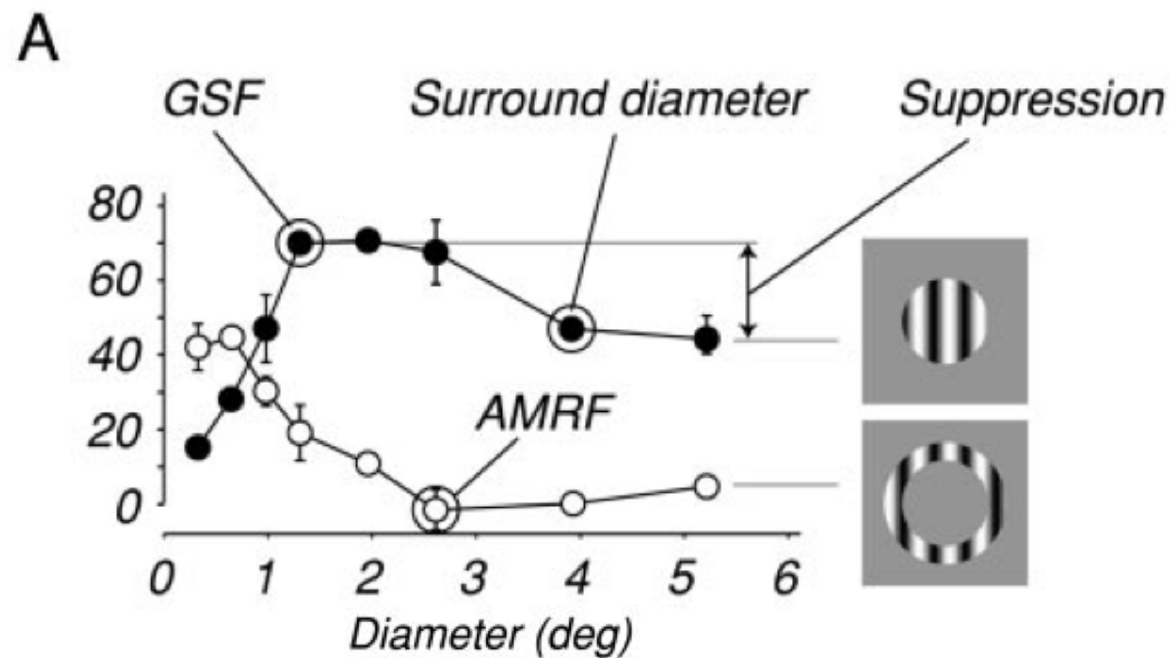


Hubel & Wiesel, *J. Physiol.*, 1962

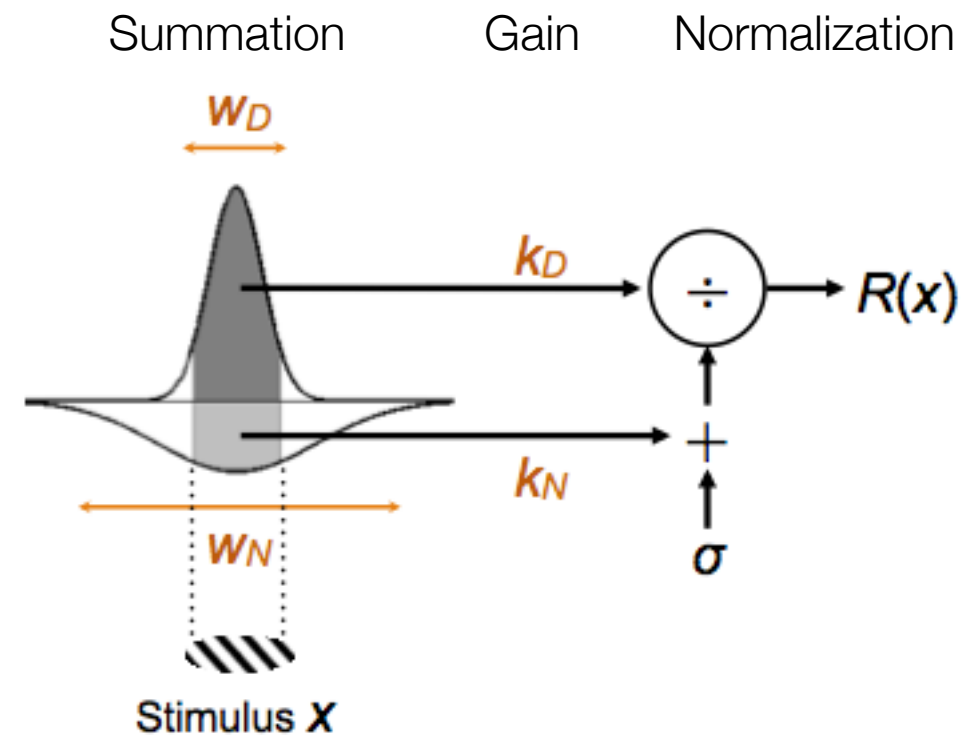
**Question:** The circuits are essentially identical, so why call one ‘AND’ and the other ‘OR’?



# More is not always better: the surround can suppress the responses of neurons in V1



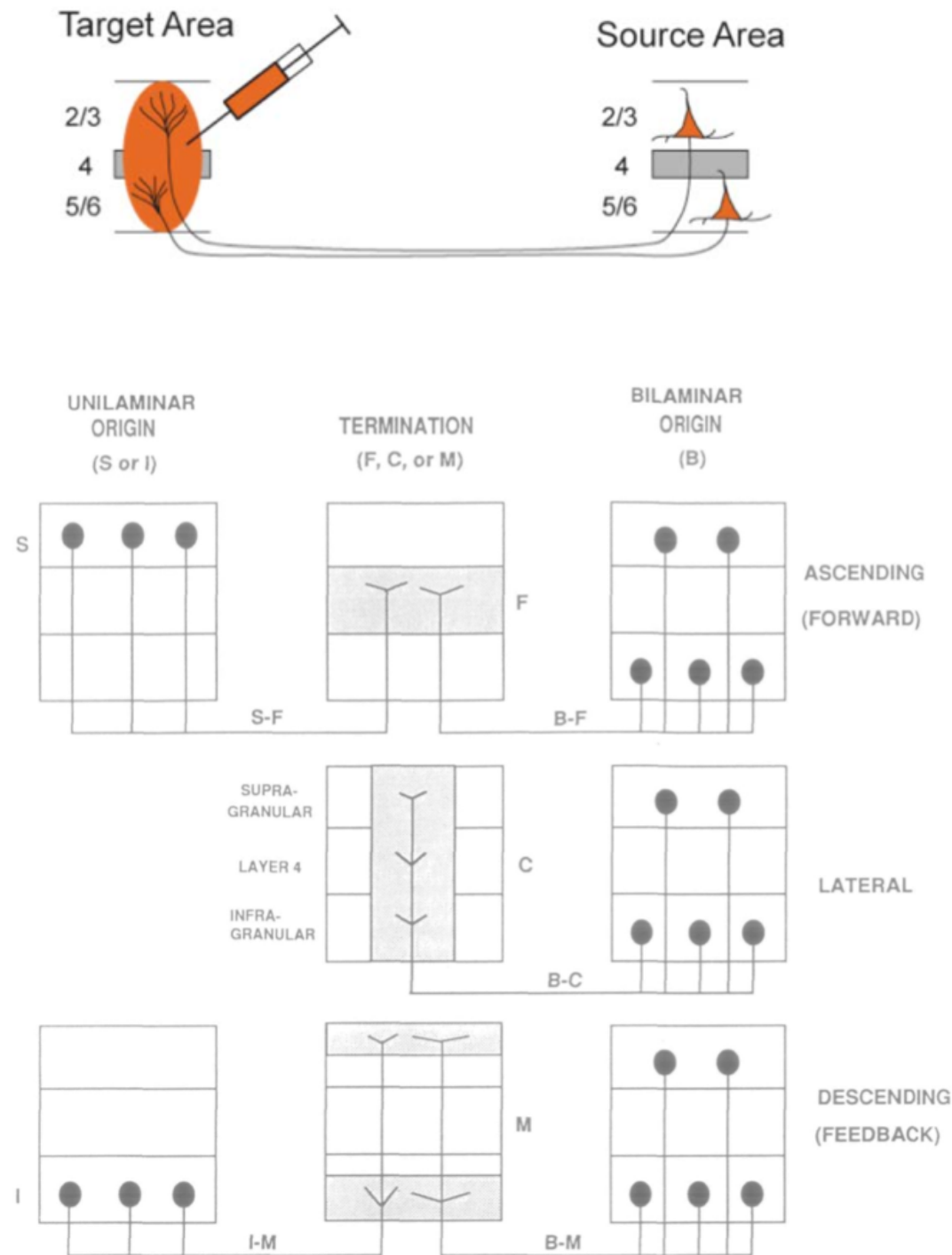
Cavanaugh et al., *J. Neurophysiol.*, 2002



$$R_{\text{ROG}}(x) = R_0 + \frac{k_D [w_D \text{erf}(x/2w_D)]^2}{\sigma + k_N [w_N \text{erf}(x/2w_N)]^2}$$

Nassi et al., *Front. Syst. Neurosci.*, 2014

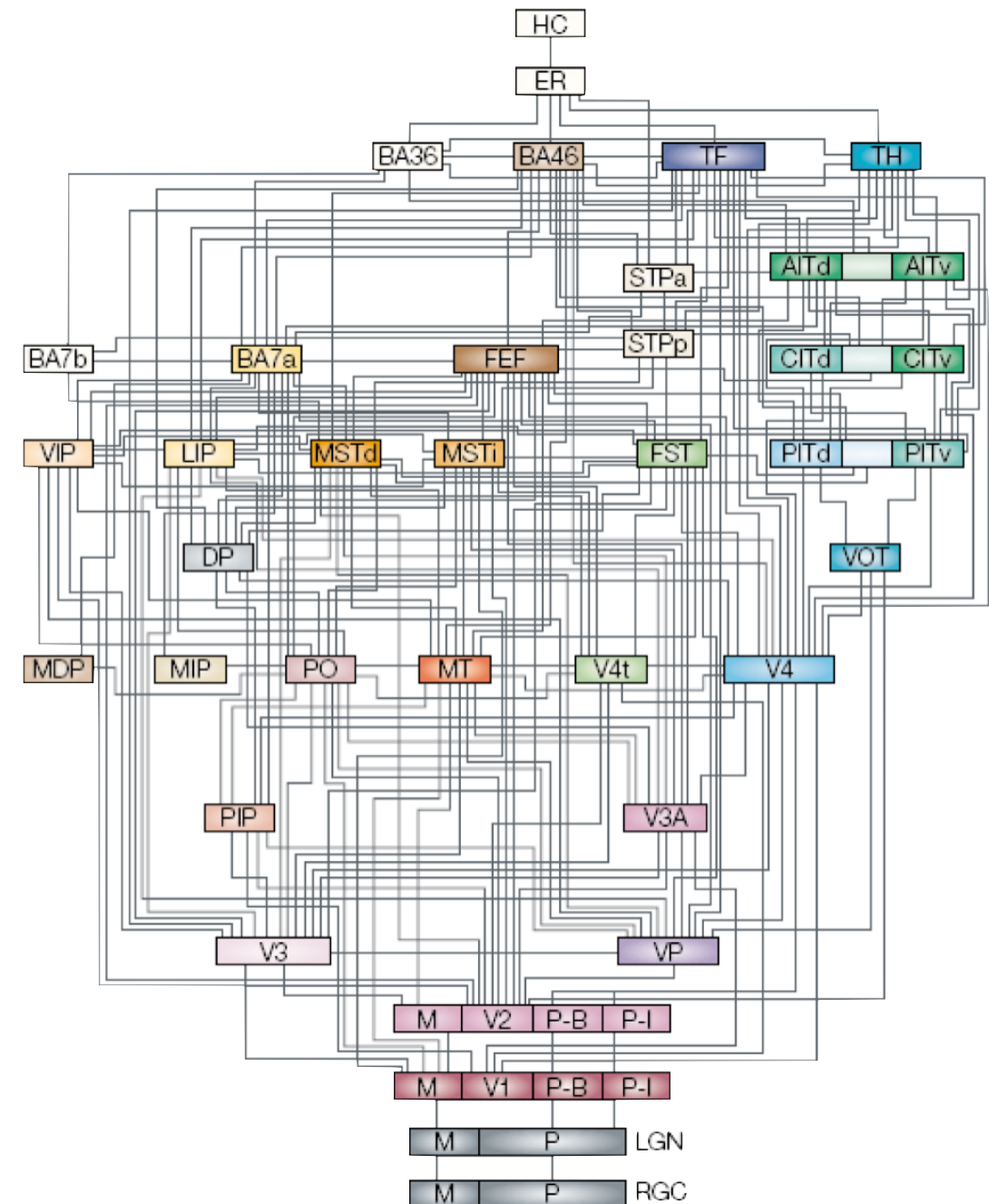
# Neurons “work” together!



Maunsell & Van Essen, *J. Neurosci.*, 1983

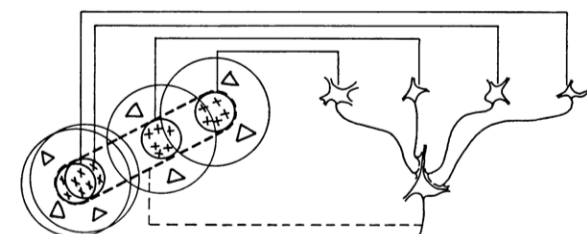
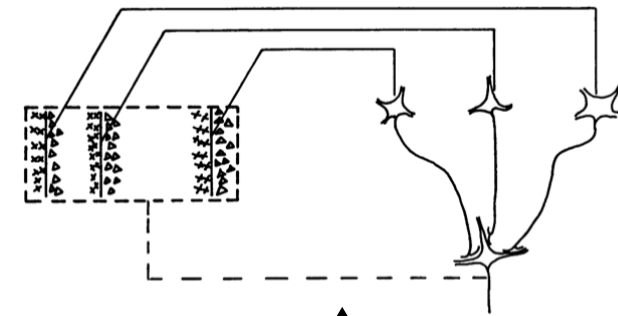
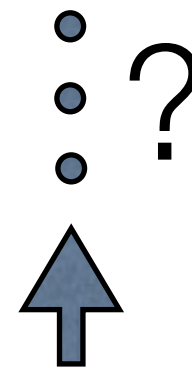
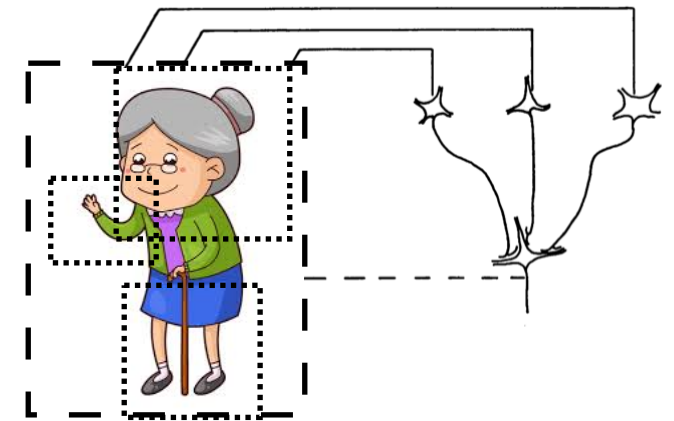
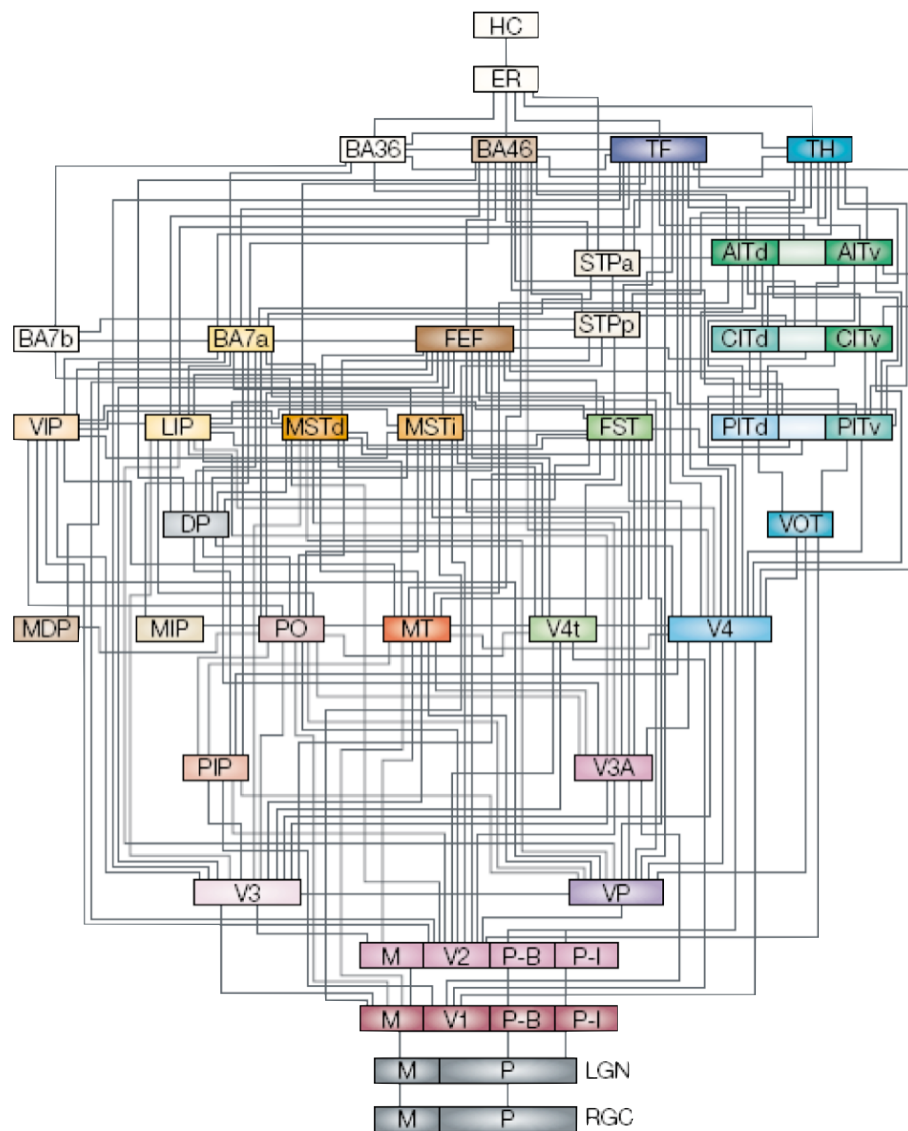
## Neural populations

Cortical areas are hierarchically organized



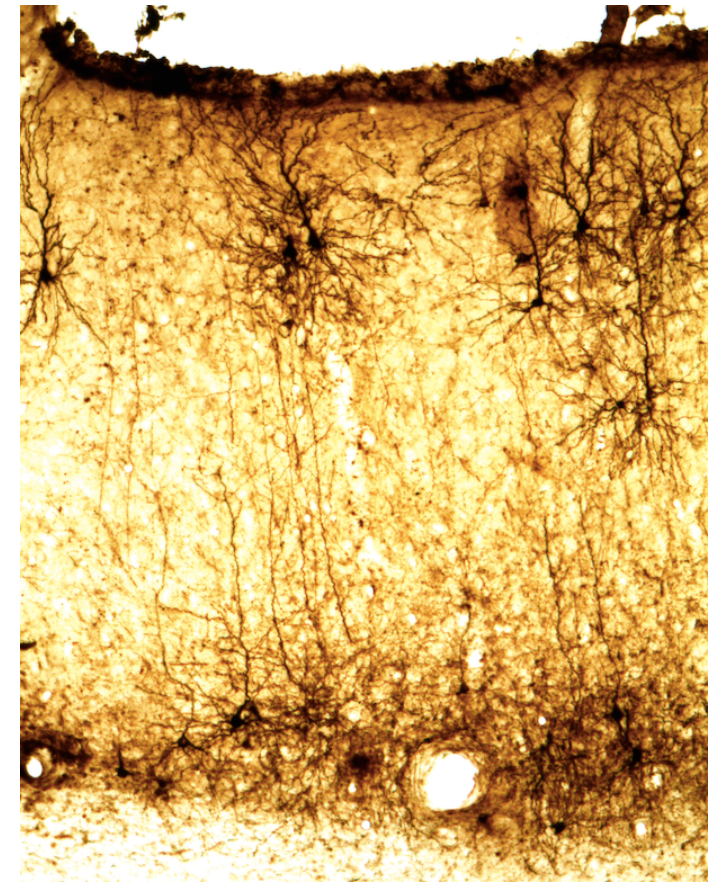
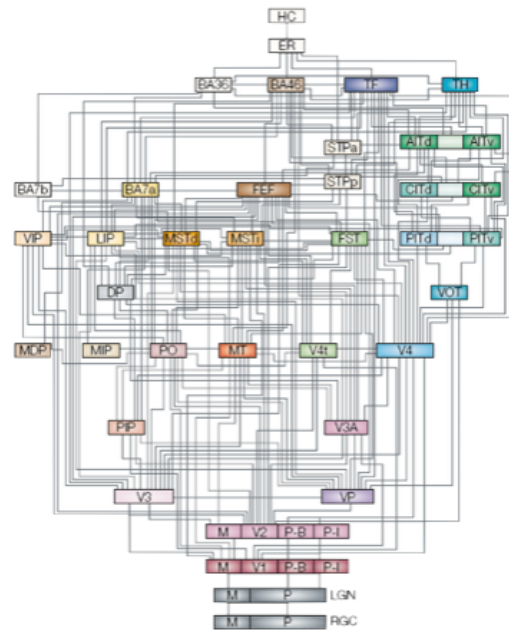
Felleman & Van Essen, *Cereb. Cortex*, 1991

# Feedforward connectivity can enable highly selective (and tolerant) neurons

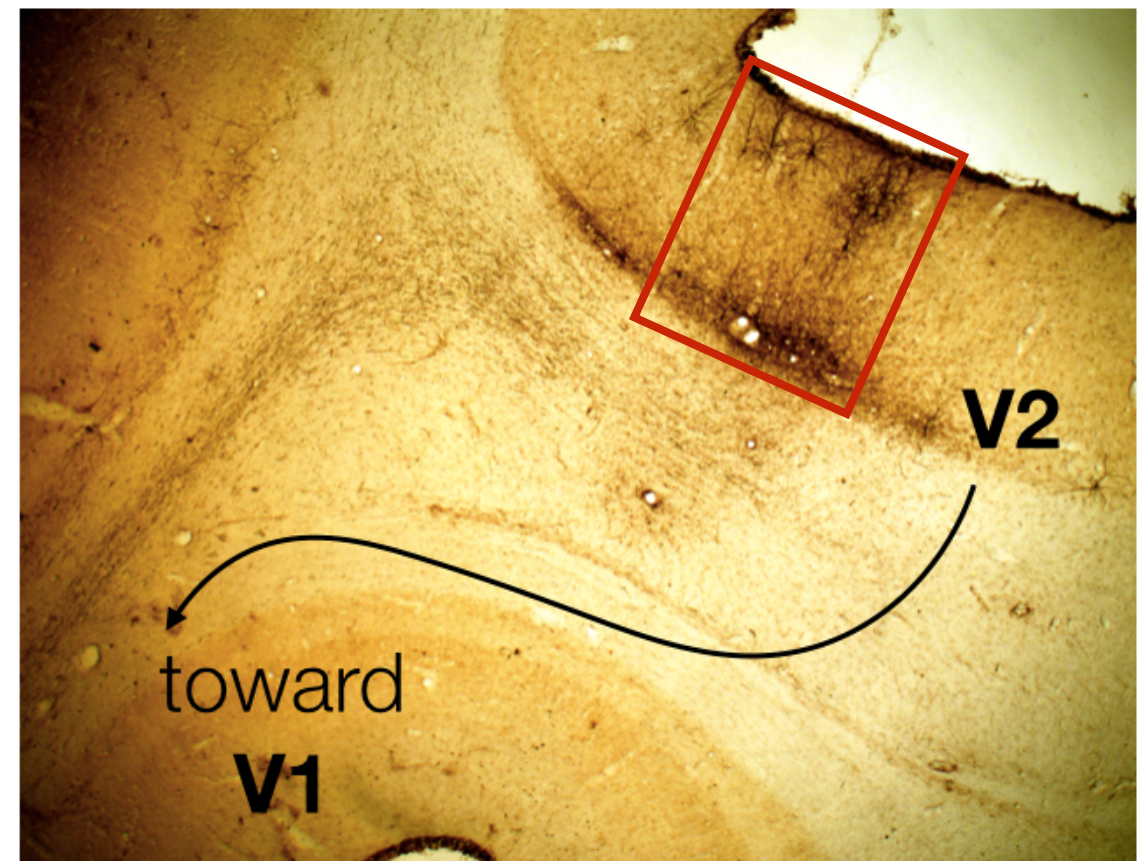
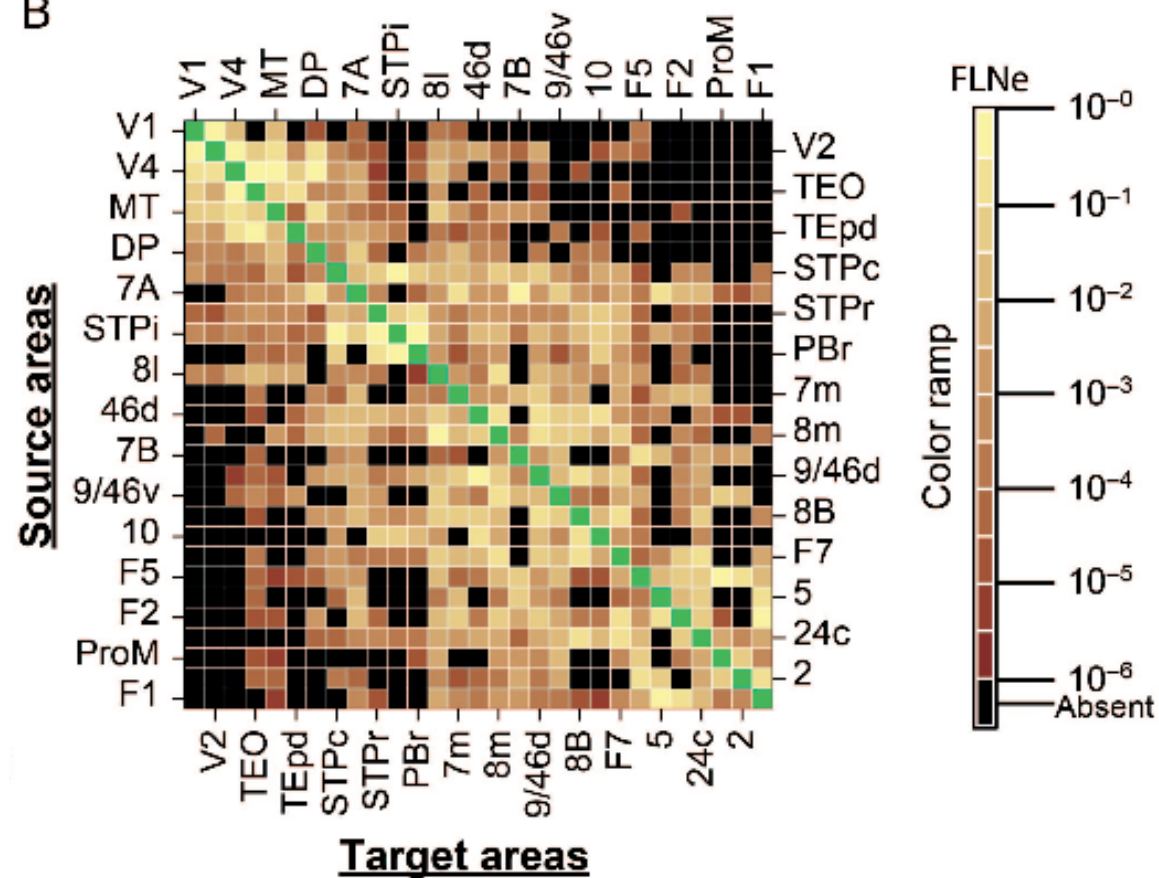




# So, what does cortical feedback do?

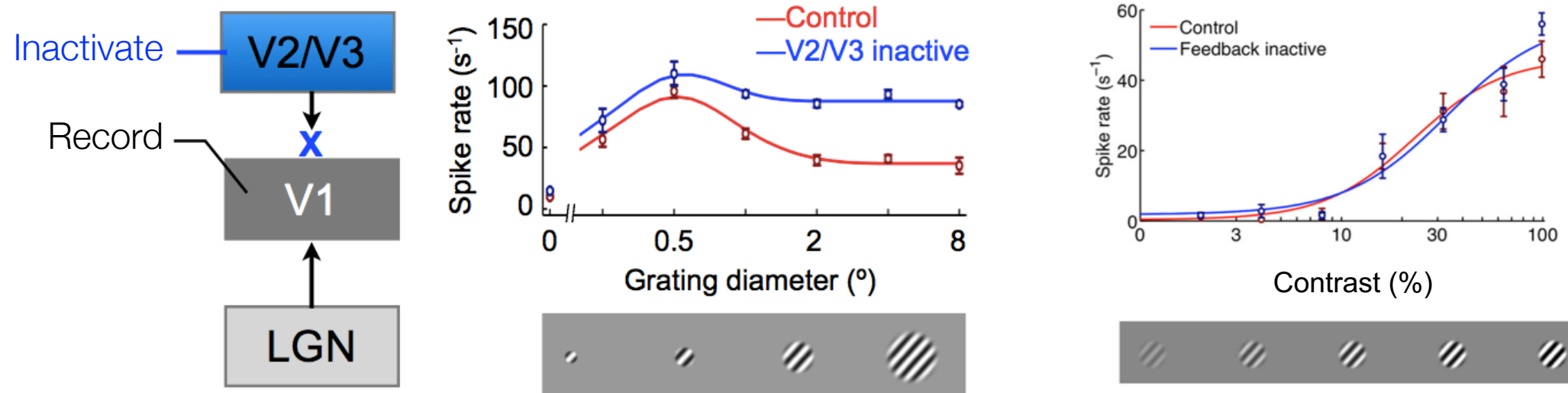


B



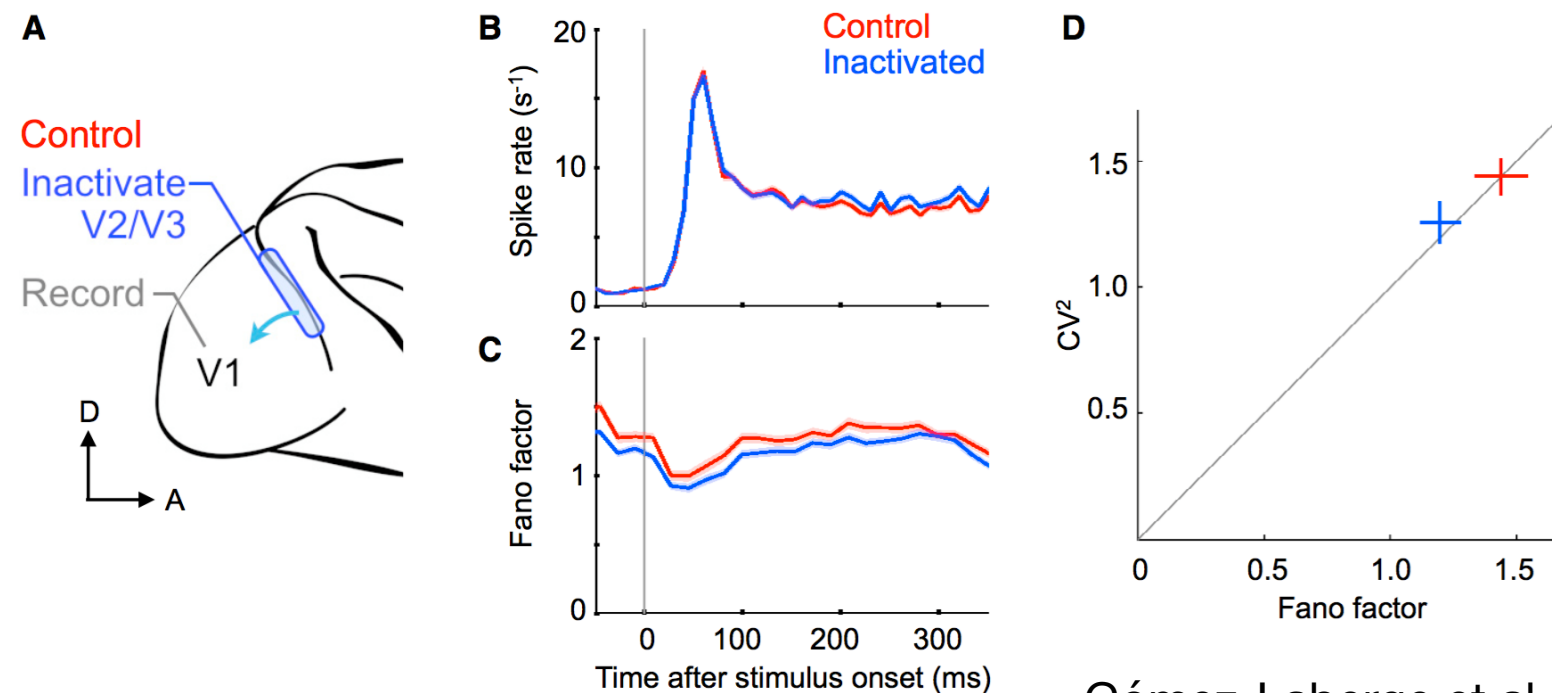
# Two basic things we've learned about feedback:

## Cortical feedback increases surround suppression to V1 neurons



Nassi et al., *Front. Syst. Neurosci.*, 2014

## Cortical feedback increases the **trial-to-trial variability** of V1 neurons



Gómez-Laberge et al., *Neuron*, 2016



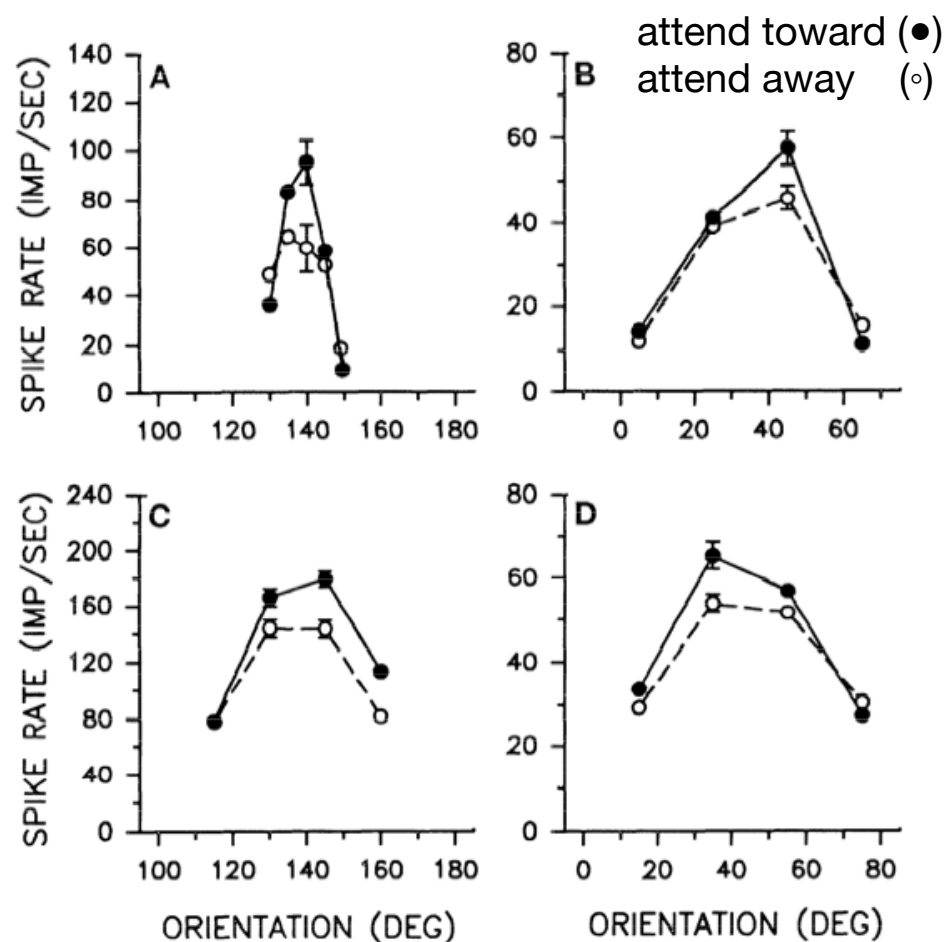
# Four orders of magnitude: from neuron to organism

## Neural correlates of behavior

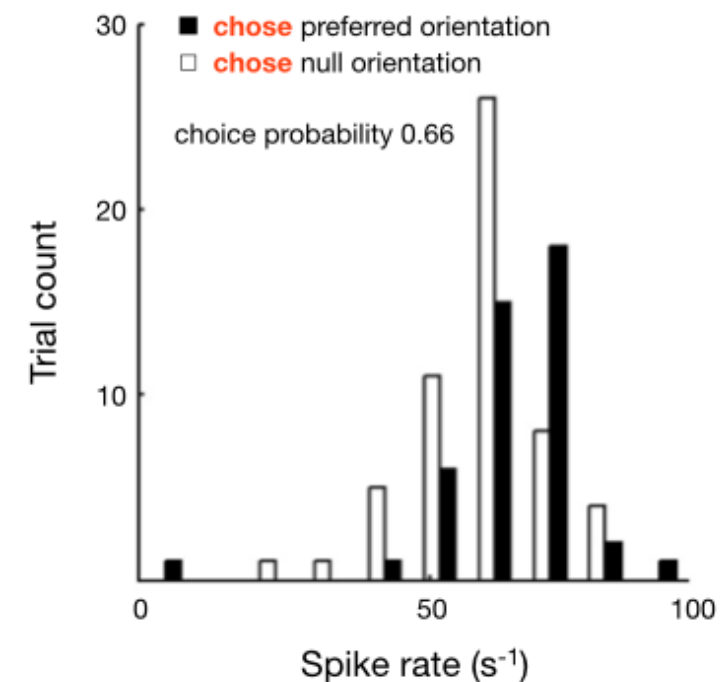
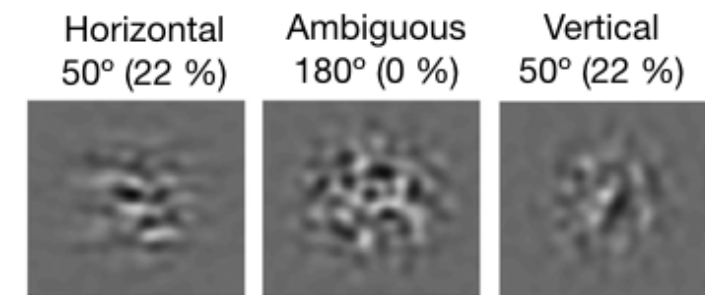
### The Emergence of a Neurobiology of Perception

The ready ability to record neuronal activity in alert nonhuman primates has allowed one of the most notable achievements of modern neuroscience: the establishment of direct links between the behavior of single neurons and that of the whole organism (e.g., Mountcastle et al. 1972, Newsome et al. 1989; see also Barlow 1972).

Albright & Stoner, *Annu. Rev. Neurosci.*, 2002



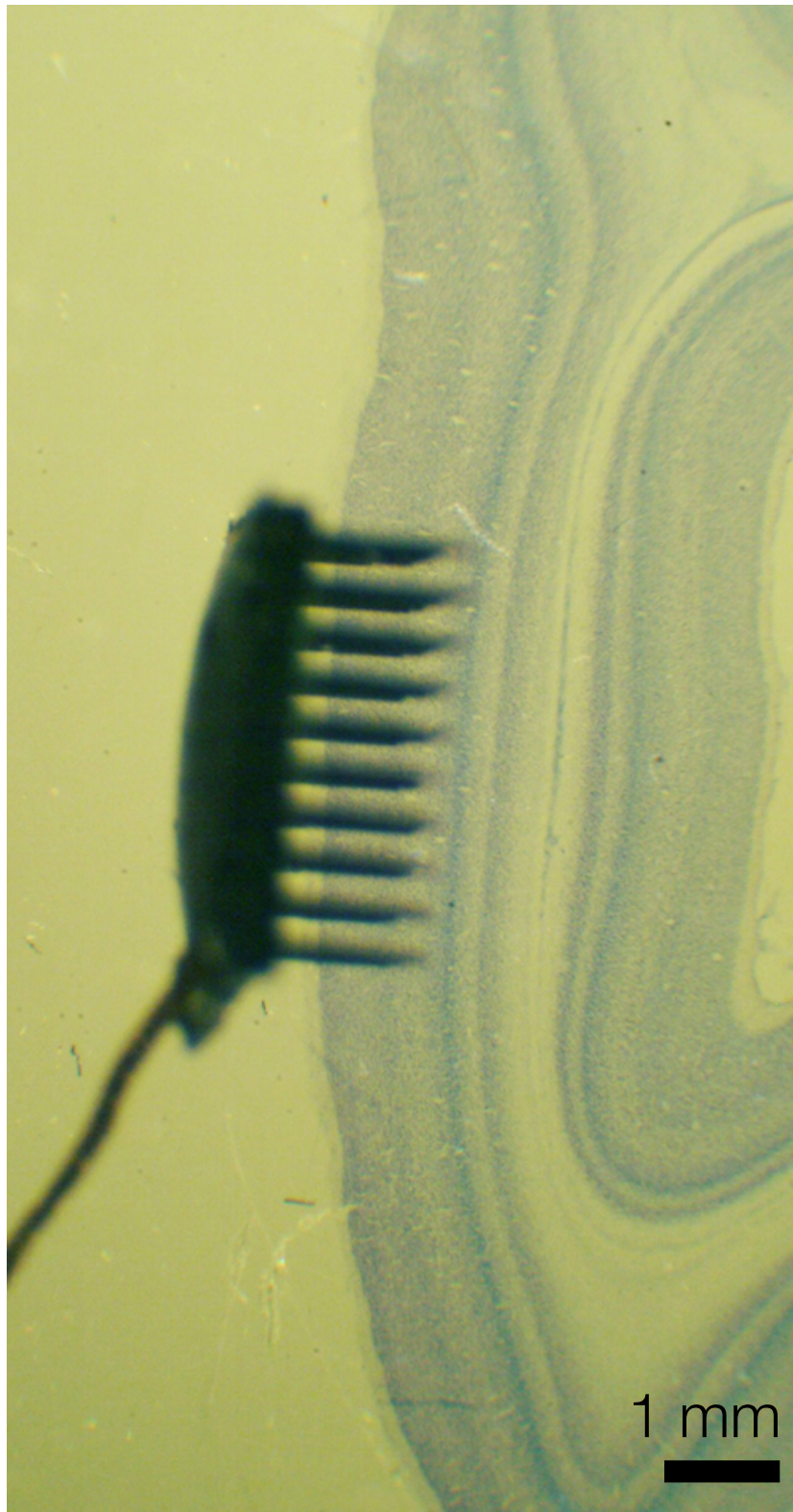
Motter, *J. Neurophysiol.*, 1993



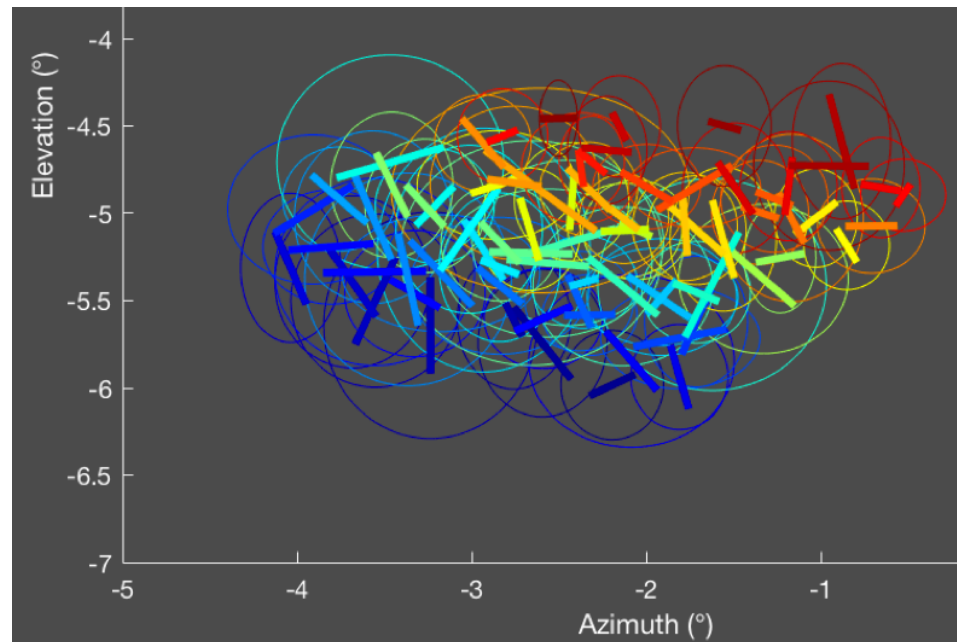
Nienborg & Cumming, *J. Neurosci.*, 2014

# Behavioral context is also related to neural co-variability...

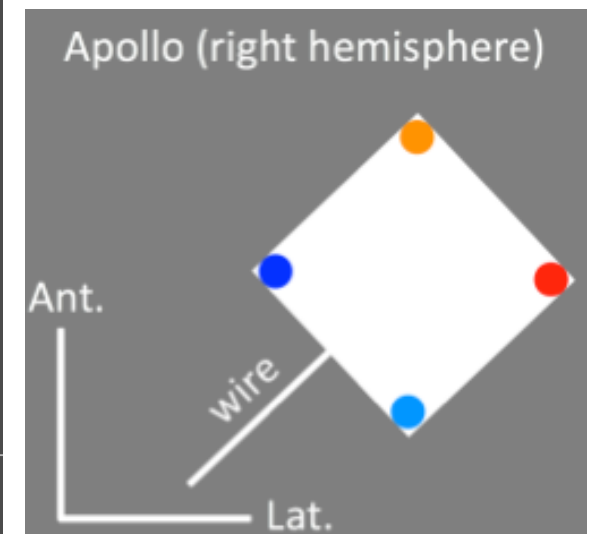
10 x 10 multi-electrode array



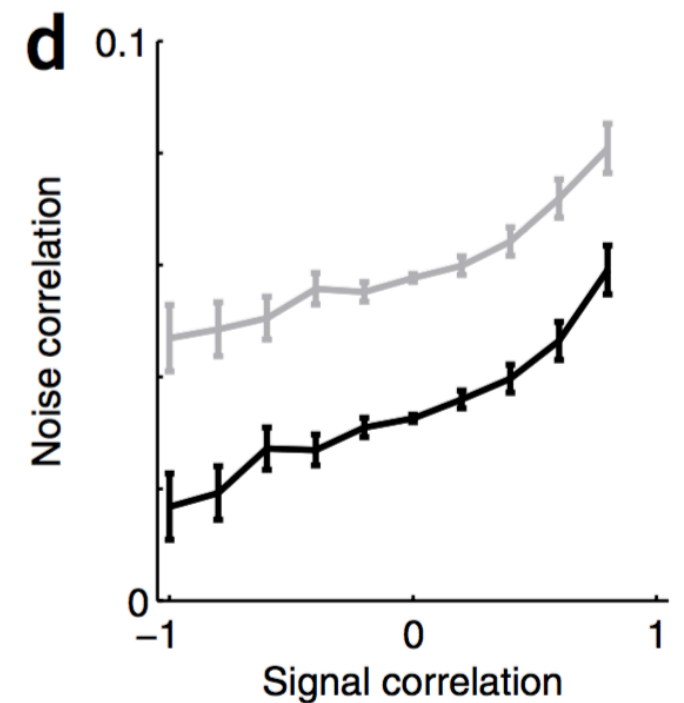
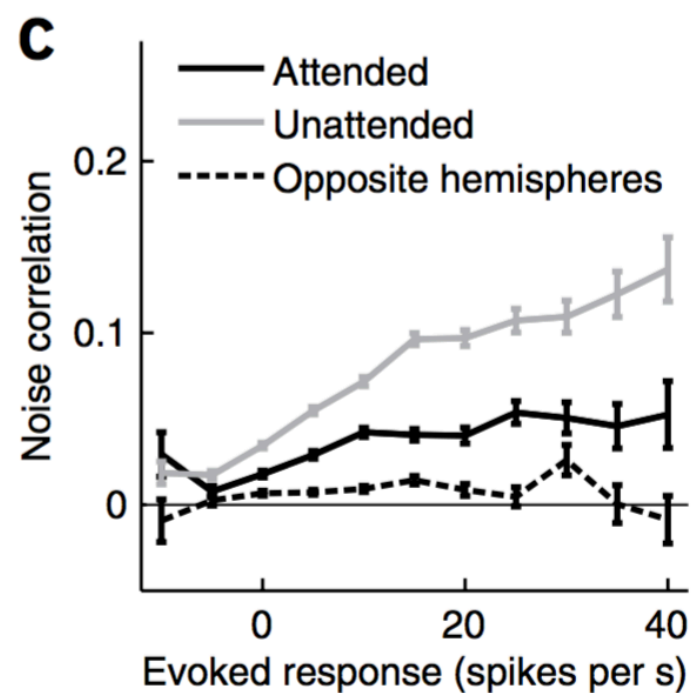
receptive field and preferred orientation



array placement in brain



correlated activity between neurons is prevalent in cortex



... which leads us to a unifying hypothesis (to be tested):  
feedback provides behavioral context to visual cortex

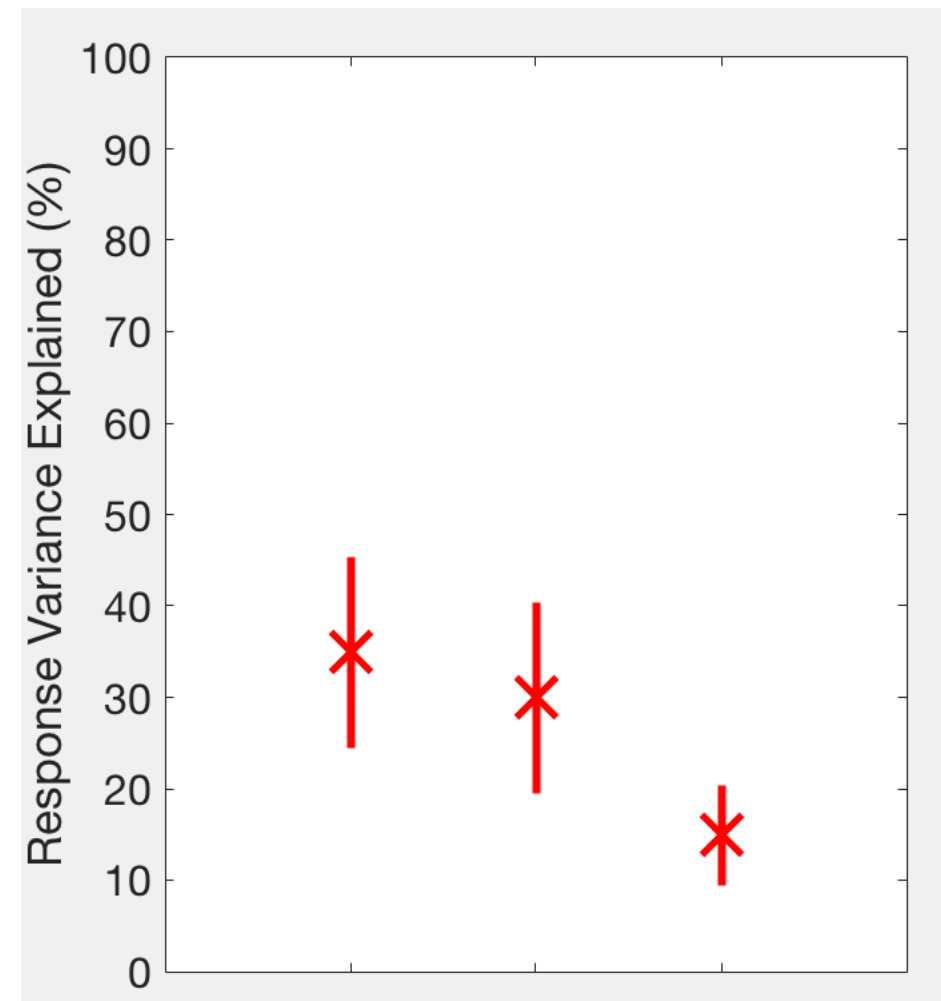
[ Some unpublished work will appear on this slide during the lecture ]

A grain (perhaps a block) of salt:  
But do we even *really* know what V1 does?

## The Unknown

What we currently understand is subject to important limitations:

- Biased sampling of neurons
- Biased visual stimuli
- Biased theories
- Contextual effects
- Internal connections and feedback



Carandini et al.,  
J. Neurosci., 2005

David & Gallant,  
Network, 2005

Olshausen & Field,  
Neural Comput., 2005



# Further reading

- **Wandell B. Foundations of Vision. Sinauer Books 1995.**
- **Dayan and Abbott. Theoretical Neuroscience. MIT Press 2001.**

## Papers cited in these slides (not exhaustive list):

1. Hubel DH, Wiesel TN (1959) Receptive fields of single neurones in the cat's striate cortex. *J Physiol (Lond)* 148:574–591.
2. Hubel DH, Wiesel TN (1962) Receptive fields, binocular interaction and functional architecture in the cat's visual cortex. *J Physiol (Lond)* 160:106–154.
3. Hubel DH, Wiesel TN (1977) Functional architecture of macaque monkey visual cortex. *Proc R Soc Lond B* 198:1–59.
4. Horton JC, Adams DL (2005) The cortical column: a structure without a function. *Philos Trans R Soc Lond, B, Biol Sci* 360:837–862.
5. Cavanaugh JR, Bair W, Movshon JA (2002) Nature and Interaction of Signals From the Receptive Field Center and Surround in Macaque V1 Neurons. *J Neurophysiol* 88:2530–2546.
6. Nassi JJ, Gómez-Laberge C, Kreiman G, Born RT (2014) Corticocortical feedback increases the spatial extent of normalization. *Front Syst Neurosci* 8:105.
7. Maunsell JHR, van Essen DC (1983) The connections of the middle temporal visual area (MT) and their relationship to a cortical hierarchy in the macaque monkey. *J Neurosci* 3:2563–2586.
8. Felleman DJ, van Essen DC (1991) Distributed hierarchical processing in the primate cerebral cortex. *Cereb Cortex* 1:1–47.
9. Markov NT et al. (2014) A weighted and directed interareal connectivity matrix for macaque cerebral cortex. *Cereb Cortex* 24:17–36.
10. Gómez-Laberge C, Smolyanskaya A, Nassi JJ, Kreiman G, Born RT (2016) Bottom-up and top-down input augment the variability of cortical neurons. *Neuron* 91:540–547.
11. Smith MA, Kohn A (2008) Spatial and temporal scales of neuronal correlation in primary visual cortex. *J Neurosci* 28:12591–12603.
12. Motter BC (1993) Focal attention produces spatially selective processing in visual cortical areas V1, V2, and V4 in the presence of competing stimuli. *J Neurophysiol* 70:909–919.
13. Albright TD, Stoner GR (2002) Contextual influences on visual processing. *Annu Rev Neurosci* 25:339–379.
14. Nienborg H, Cumming BG (2014) Decision-related activity in sensory neurons may depend on the columnar architecture of cerebral cortex. *J Neurosci* 34:3579–3585.
15. Cohen MR, Maunsell JHR (2009) Attention improves performance primarily by reducing interneuronal correlations. *Nat Neurosci* 12:1594–1600.
16. Lange RD, Haefner RM (2016) Inferring the brain's internal model from sensory responses in a probabilistic inference framework. *bioRxiv*. doi: 10.1101/081661