

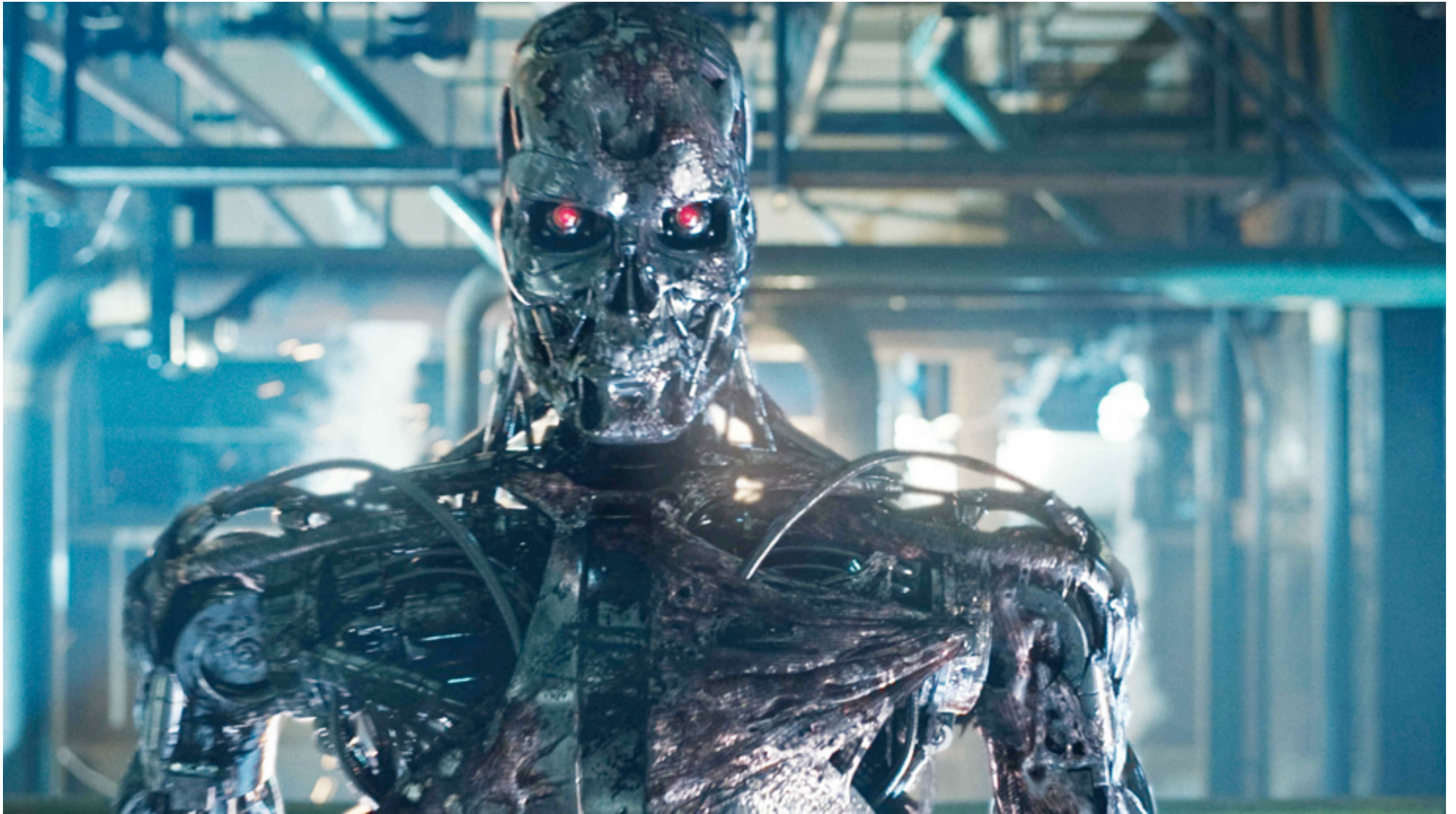
Computational Vision

Primary visual cortex

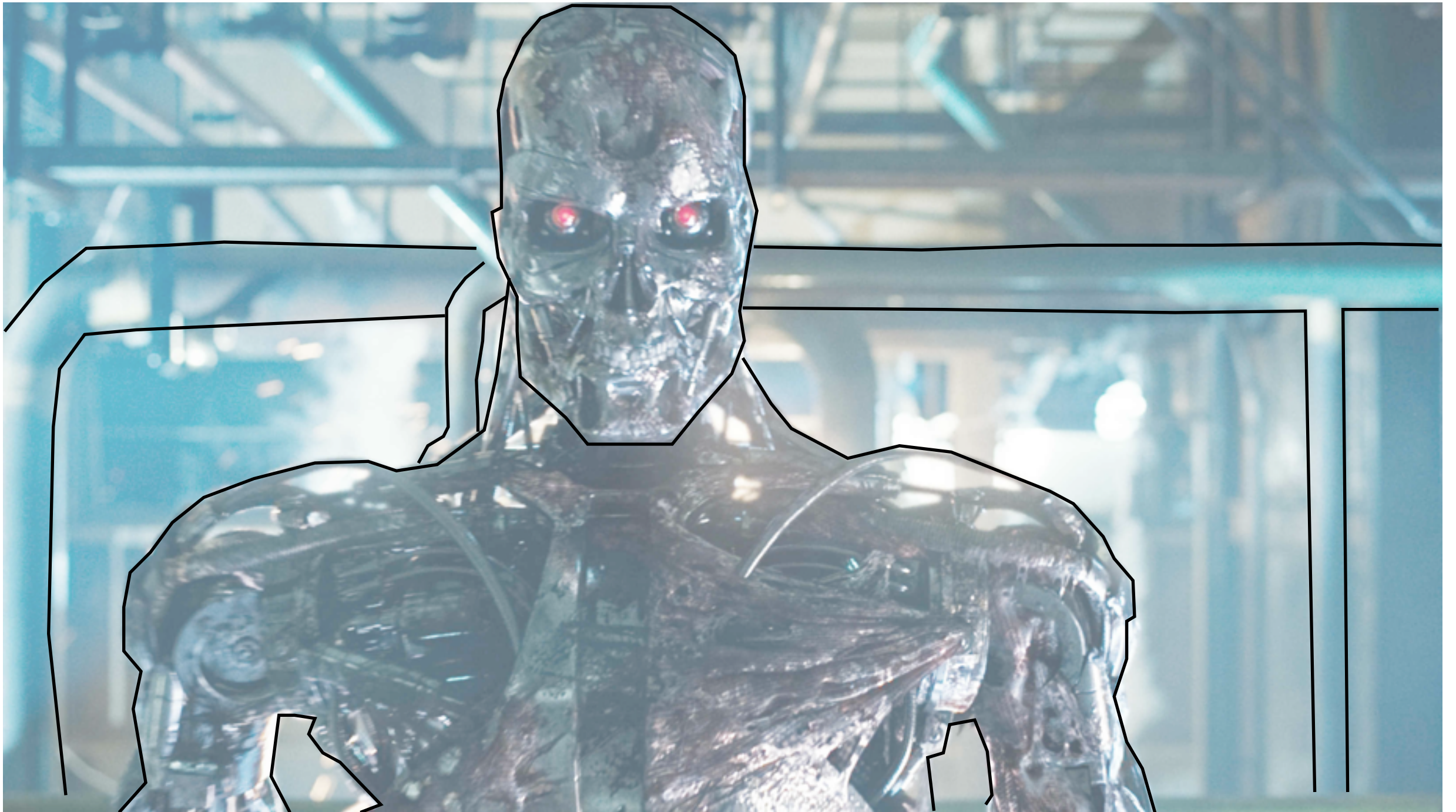
- Orientation selectivity
- Spatial frequency
- Color opponency
- Normalization



General announcement (from your TA)



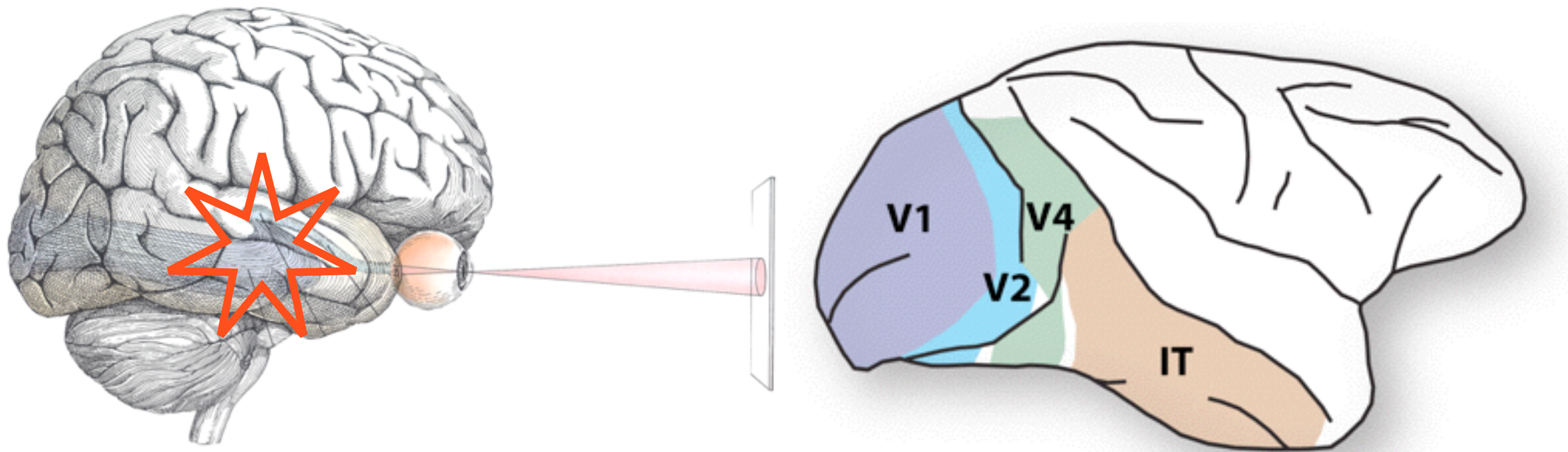
Boundary annotation



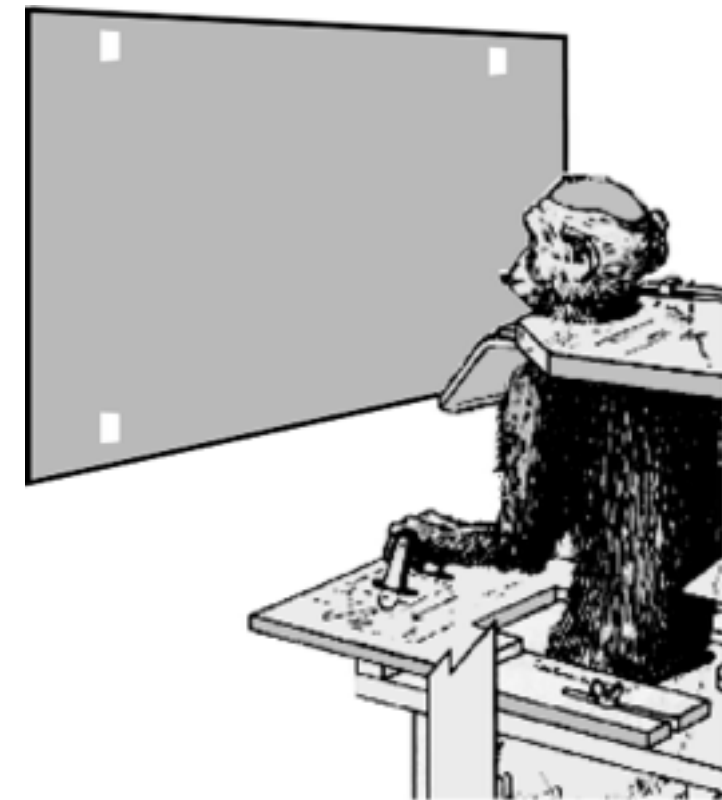
Here's the deal

- Annotate the boundaries between things in 100 natural scenes
- \$5 an hour (est. 10 hours of work)
- You have to complete within two weeks
- Speed bonus: \$50 if you complete within one week
- contact:
david_mely@brown.edu



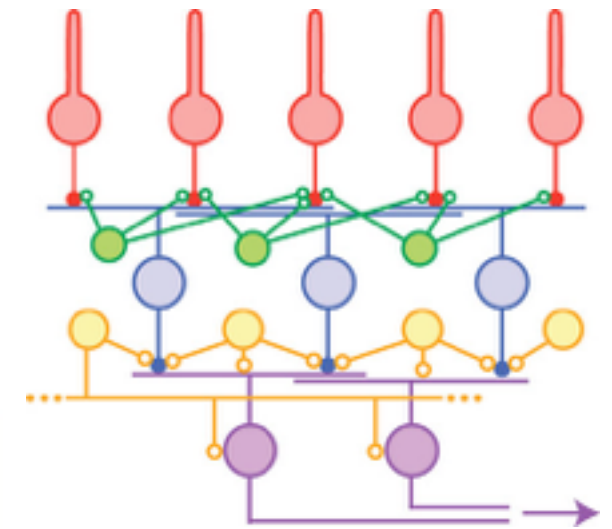
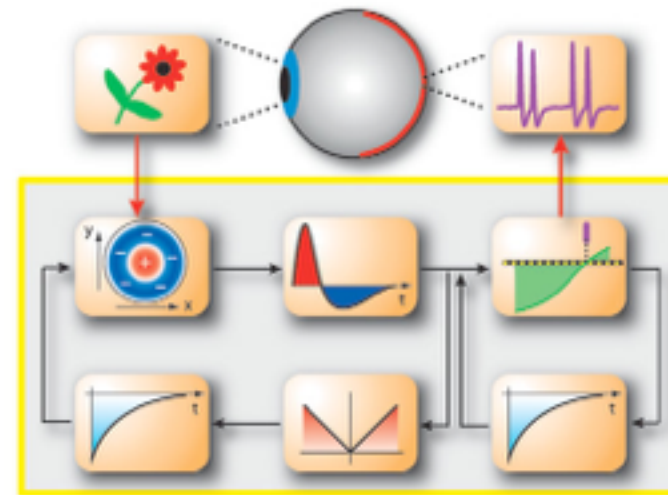


The visual system



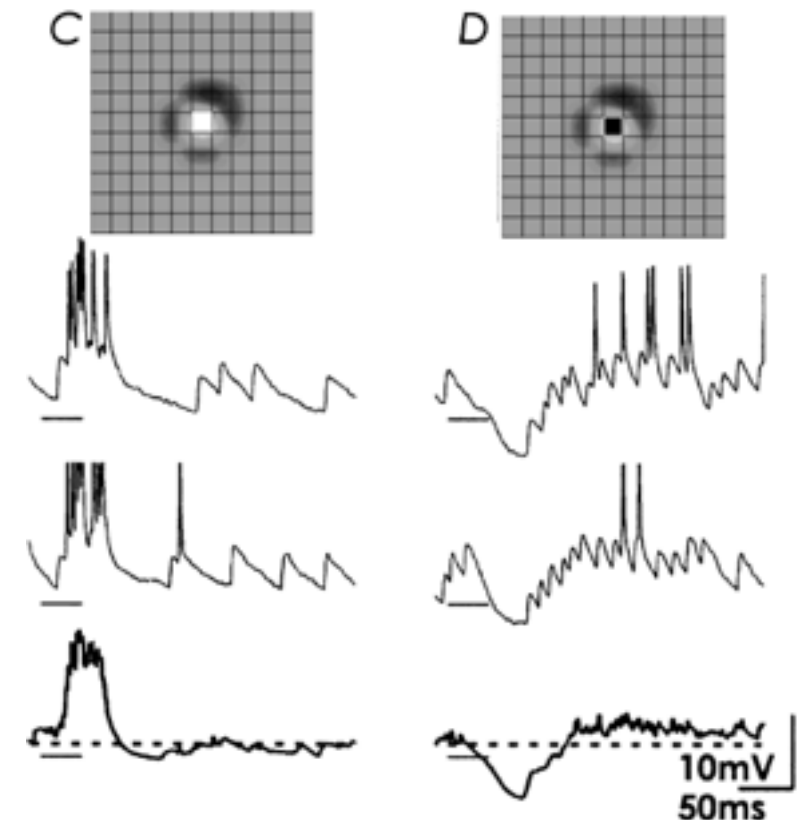
Computing with LGN

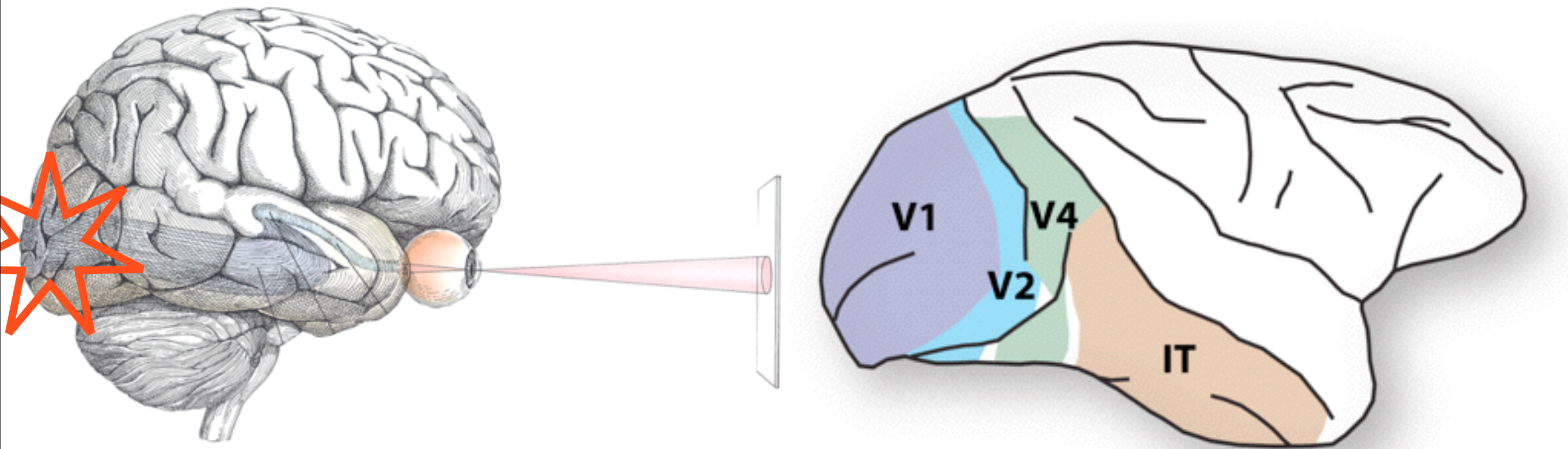
- Basic anatomy and physiology
- Center-surround processing
- Color opponency channels
- Convolution and filtering



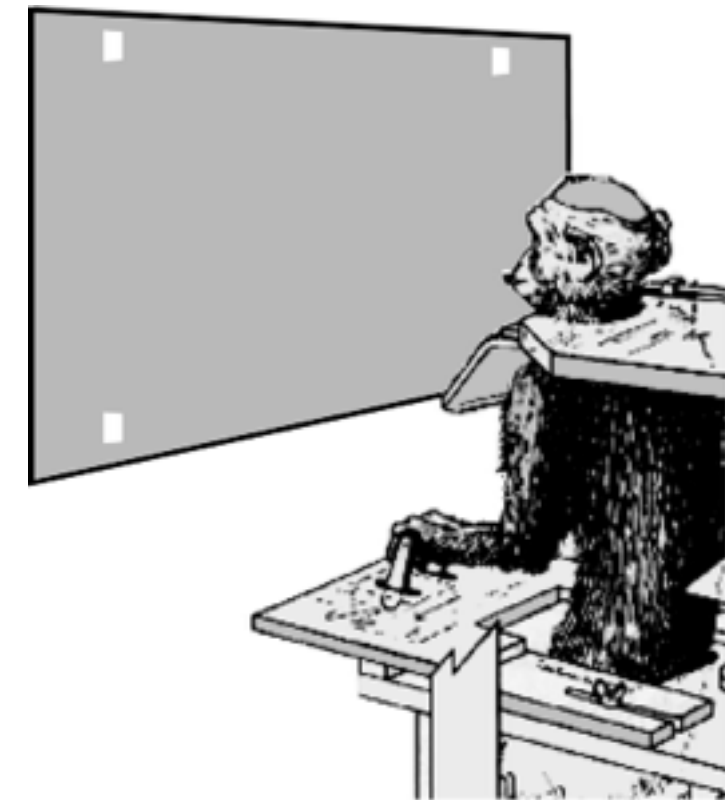
source: unknown

Thalamic Relay Cell





The visual system



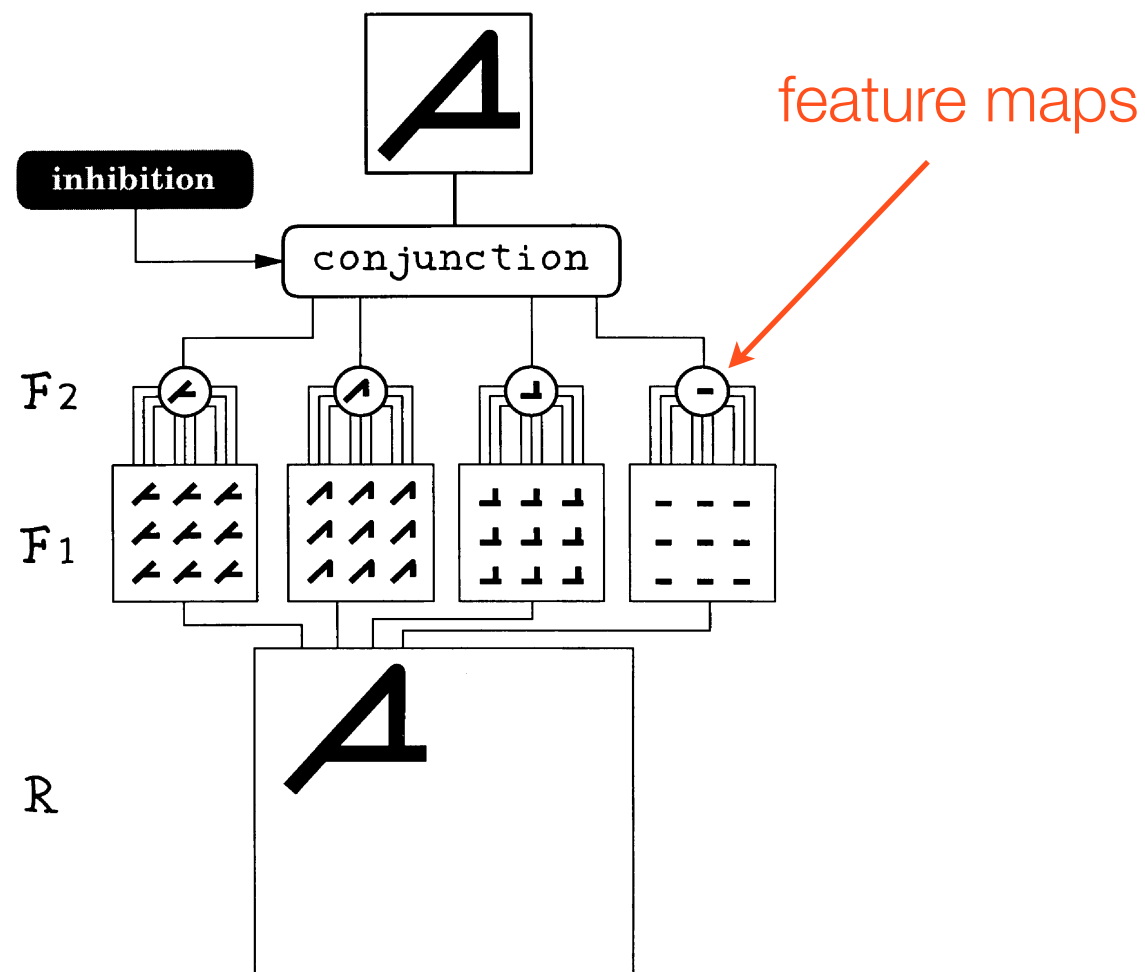
Seeing with brain maps

Orientation tuning



Visual analysis: Cortex vs. computers

Brains: Full-replication scheme



Computers: Convolution

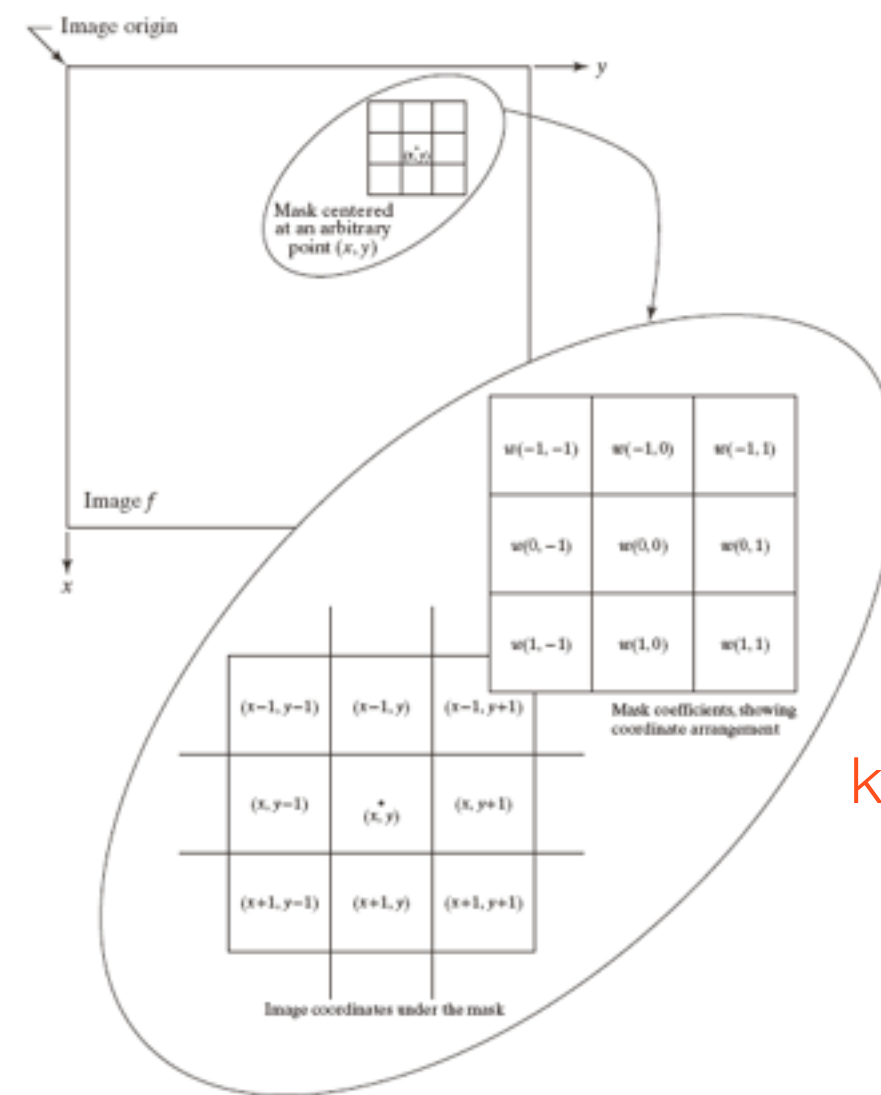
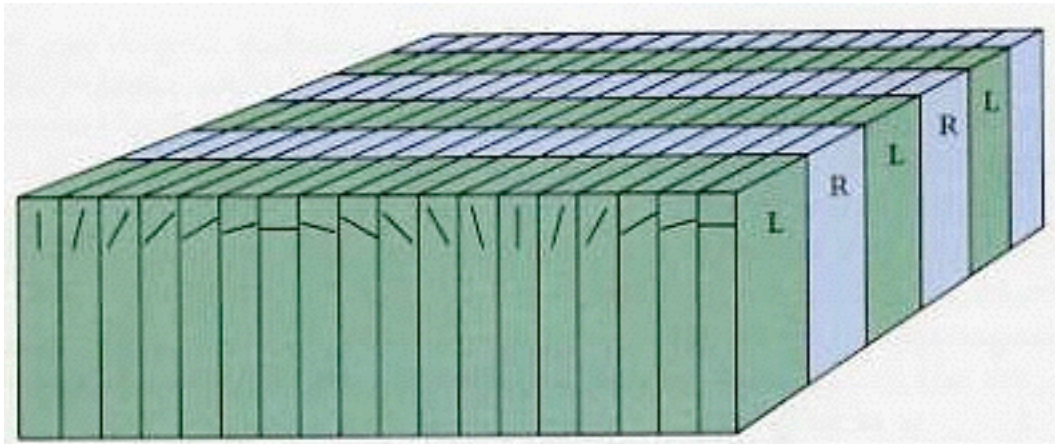
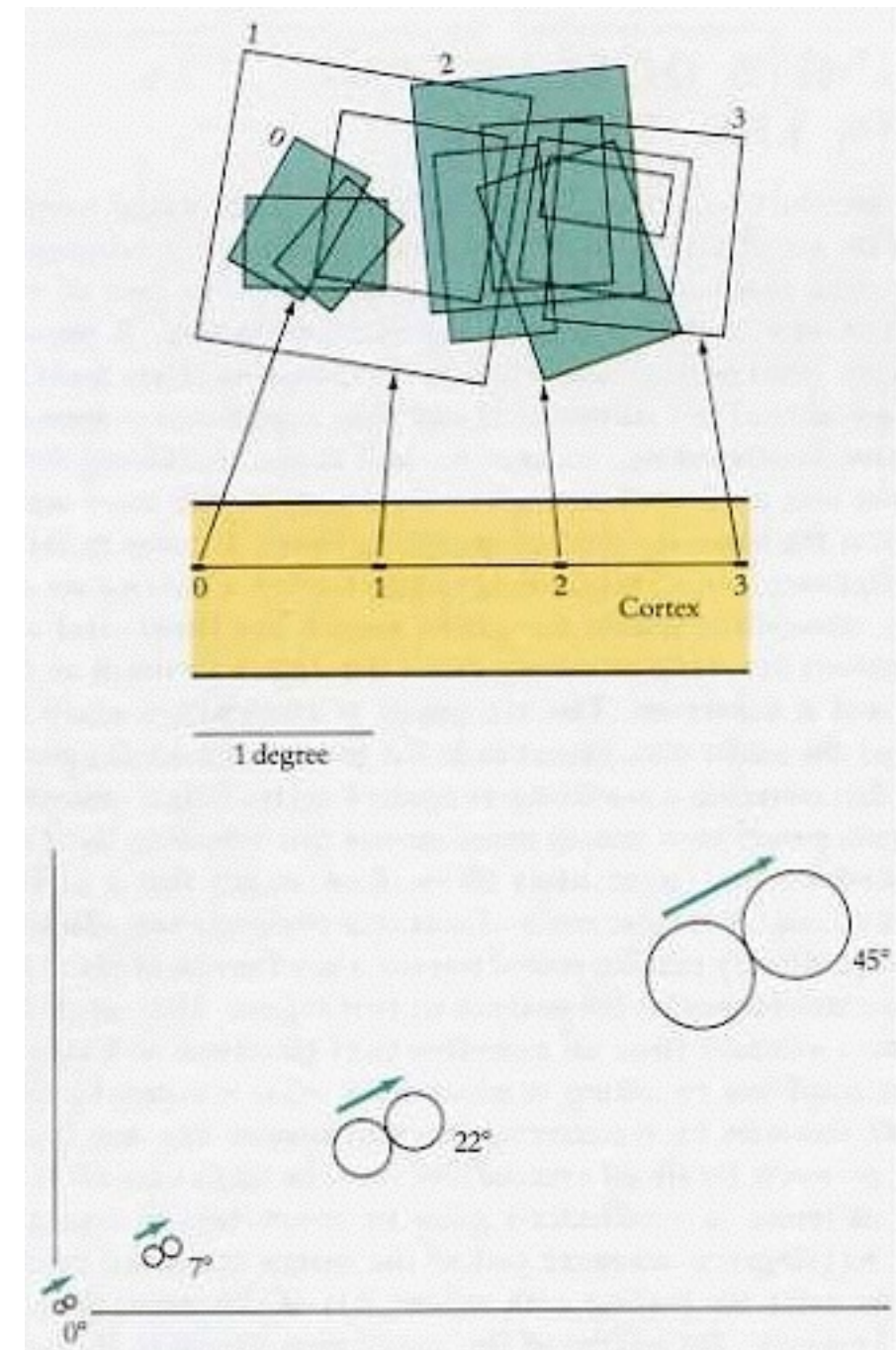
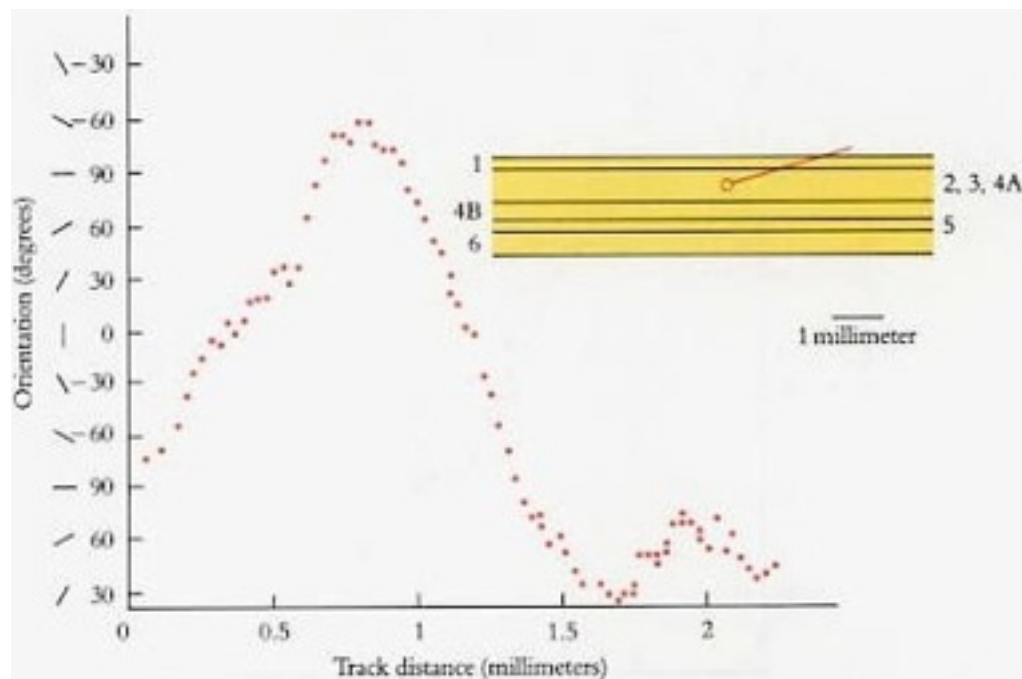


FIGURE 3.13
The mechanics of linear spatial filtering. The magnified drawing shows a 3x3 filter mask and the corresponding image neighborhood directly under it. The image neighborhood is shown displaced out from under the mask for ease of readability.

From feature detectors to population codes



Columnar organization



Computing neurons

$$y(x) = \mathbf{w}^T \mathbf{x}$$

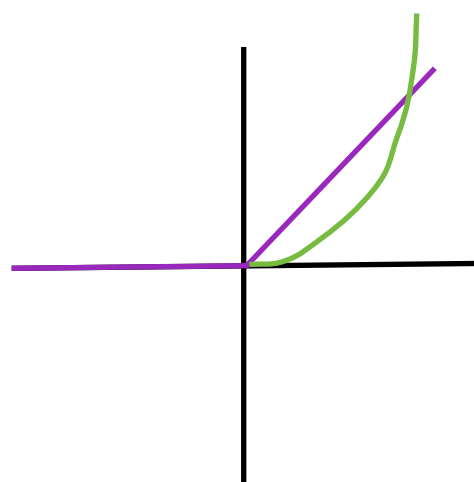
Synaptic efficacies

$$y(x) = \sum w_i x_i$$

Summation at the soma

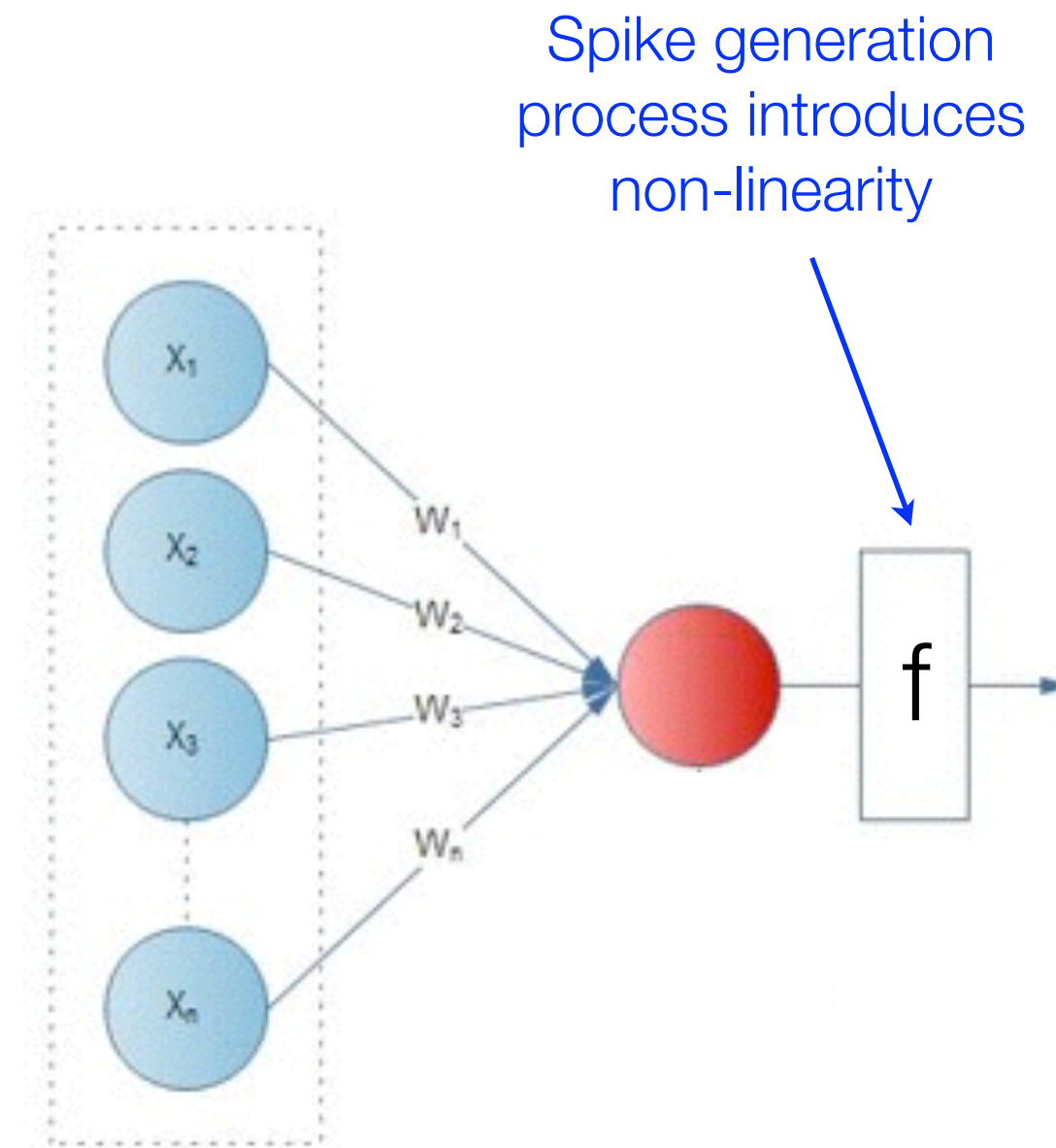
Spike trains / firing rates
coming from each
afferent neuron

rectification



$f(u) = u$ if $u > 0$
and 0 otherwise

$f(u) = u^2$ if $u > 0$
and 0 otherwise



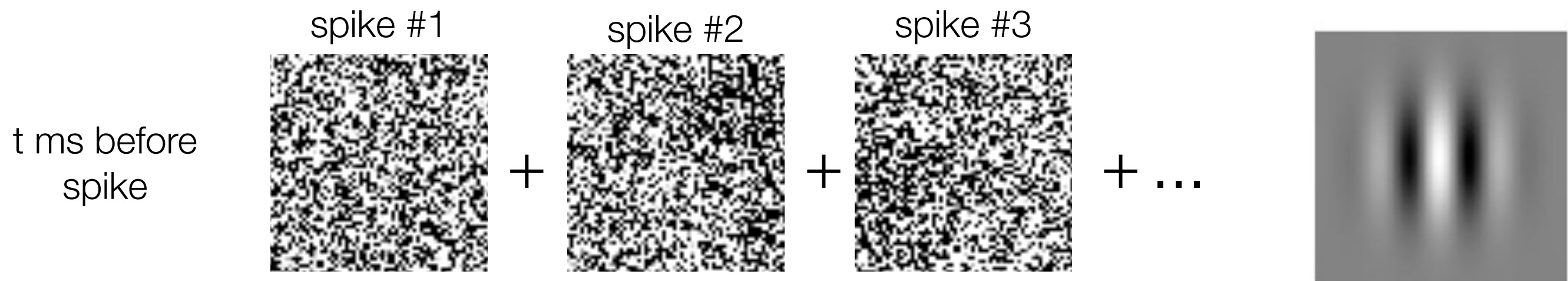
**several layers
of processing
in the retina**

**ganglion cells =
retinal output /
LGN**

Reverse-correlation / spike-triggered average

$$\begin{aligned} y(x) &= \sum w_i x_i \\ &= \mathbf{w} \cdot \mathbf{x} \\ &= ||\mathbf{w}|| ||\mathbf{x}|| \cos(\theta) \end{aligned}$$

finding out the neuron selectivity = template

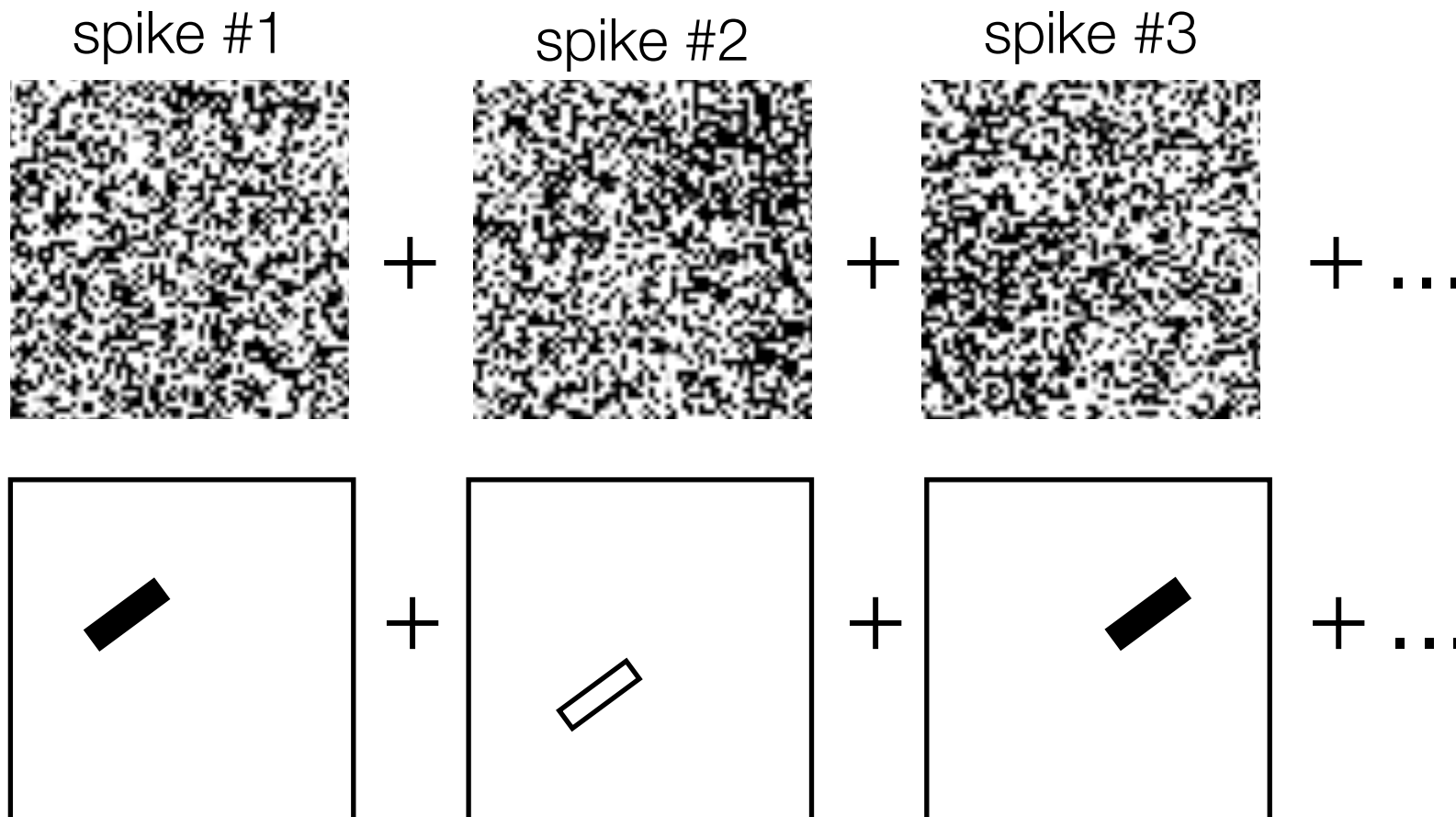


Reverse-correlation / spike-triggered average

finding out the neuron selectivity = template

$$\begin{aligned}y(x) &= \sum w_i x_i \\&= \mathbf{w} \cdot \mathbf{x} \\&= ||\mathbf{w}|| ||\mathbf{x}|| \cos(\theta)\end{aligned}$$

t ms before
spike



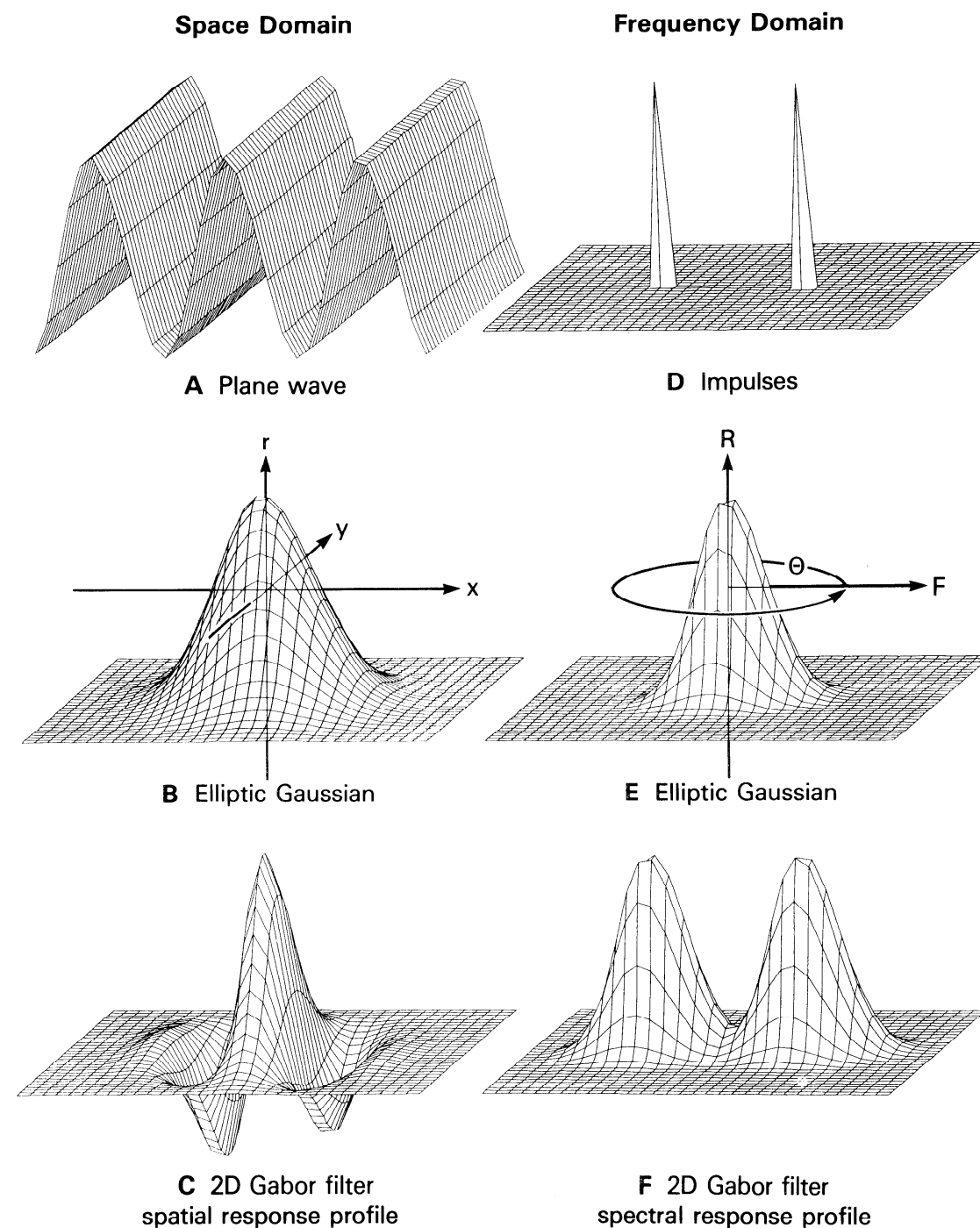
Reverse-correlation / spike-triggered average

finding out the neuron selectivity = template

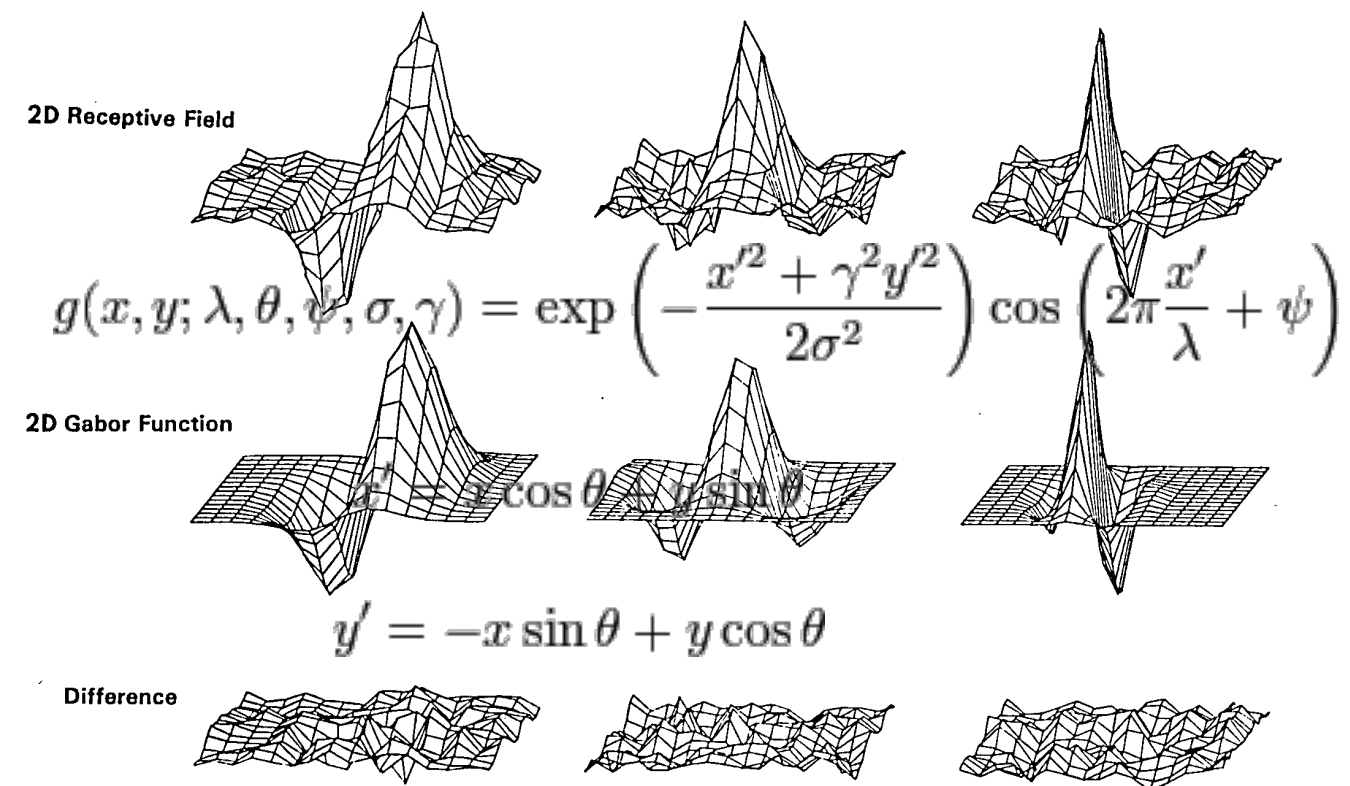
$$\begin{aligned}y(x) &= \sum w_i x_i \\&= \mathbf{w} \cdot \mathbf{x} \\&= ||\mathbf{w}|| ||\mathbf{x}|| \cos(\theta)\end{aligned}$$

Receptive Field Organization
[Reverse Correlation Analysis]

Gabor functions as models of simple cells

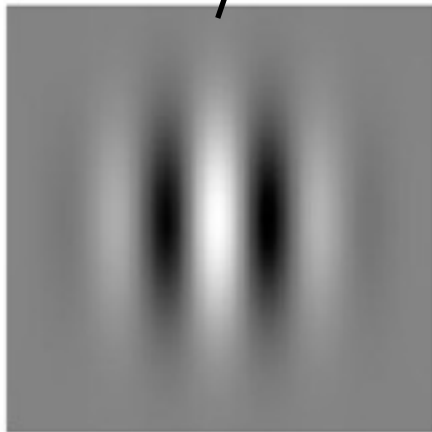


Gabor filters optimize the general uncertainty relations for joint spatial-spectral information resolution



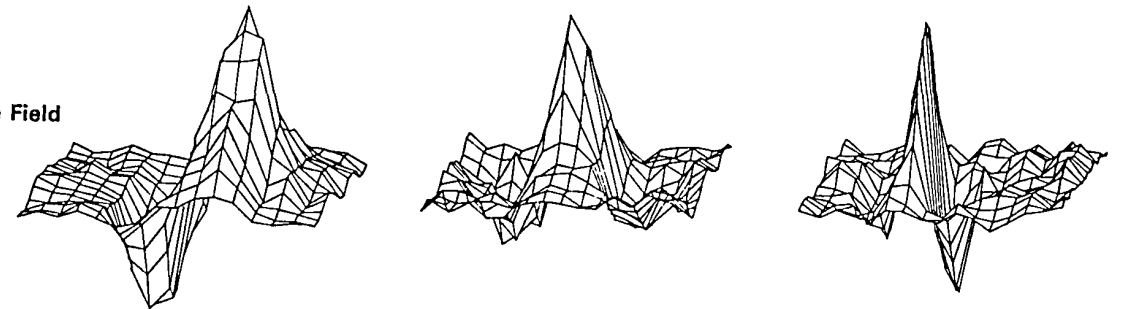
Gabor functions as models of simple cells

$$y(x) = \mathbf{w}^T \mathbf{x}$$

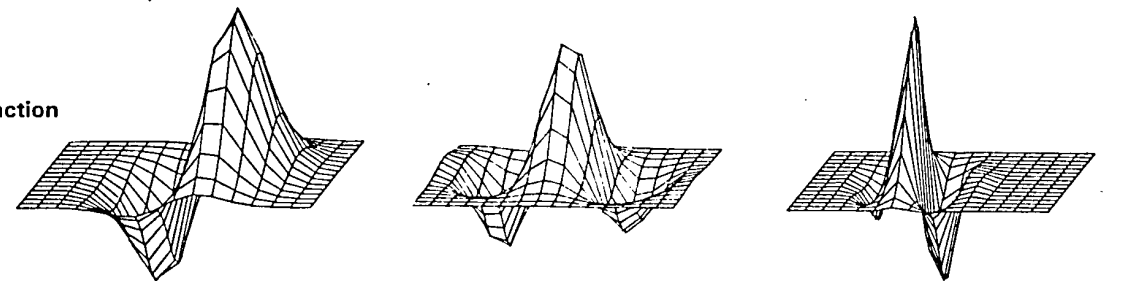


Gabor filters optimize the general uncertainty relations for joint spatial-spectral information resolution

2D Receptive Field



2D Gabor Function

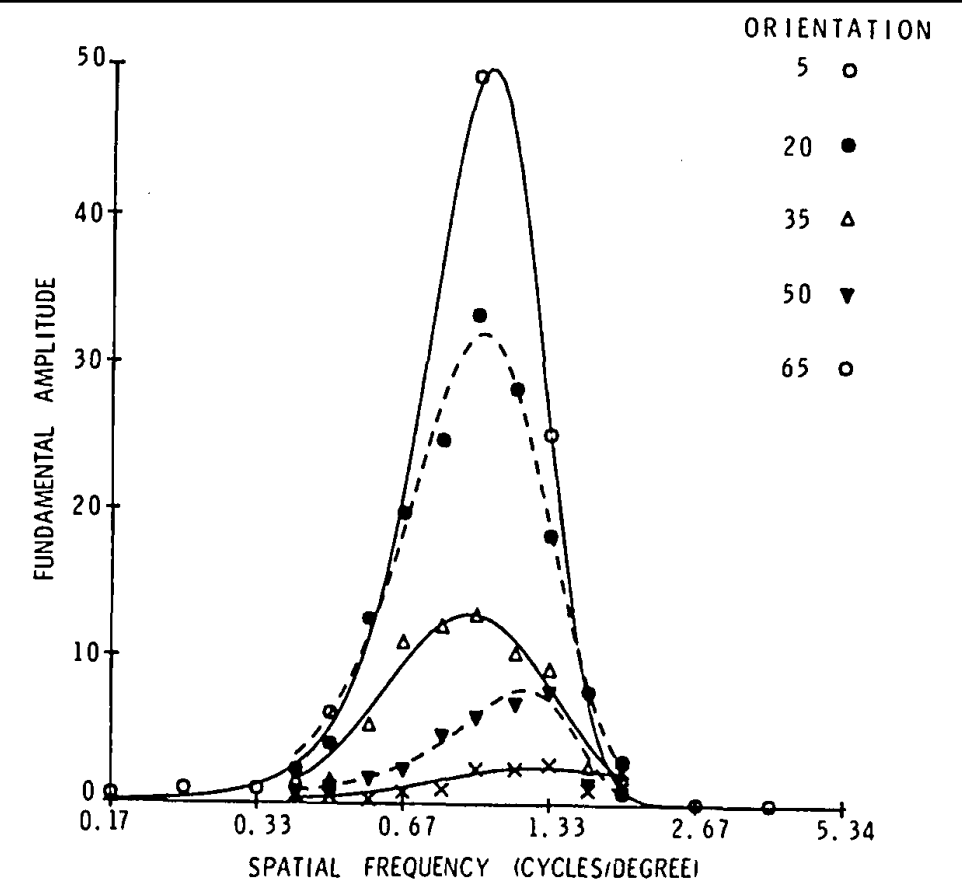


Difference

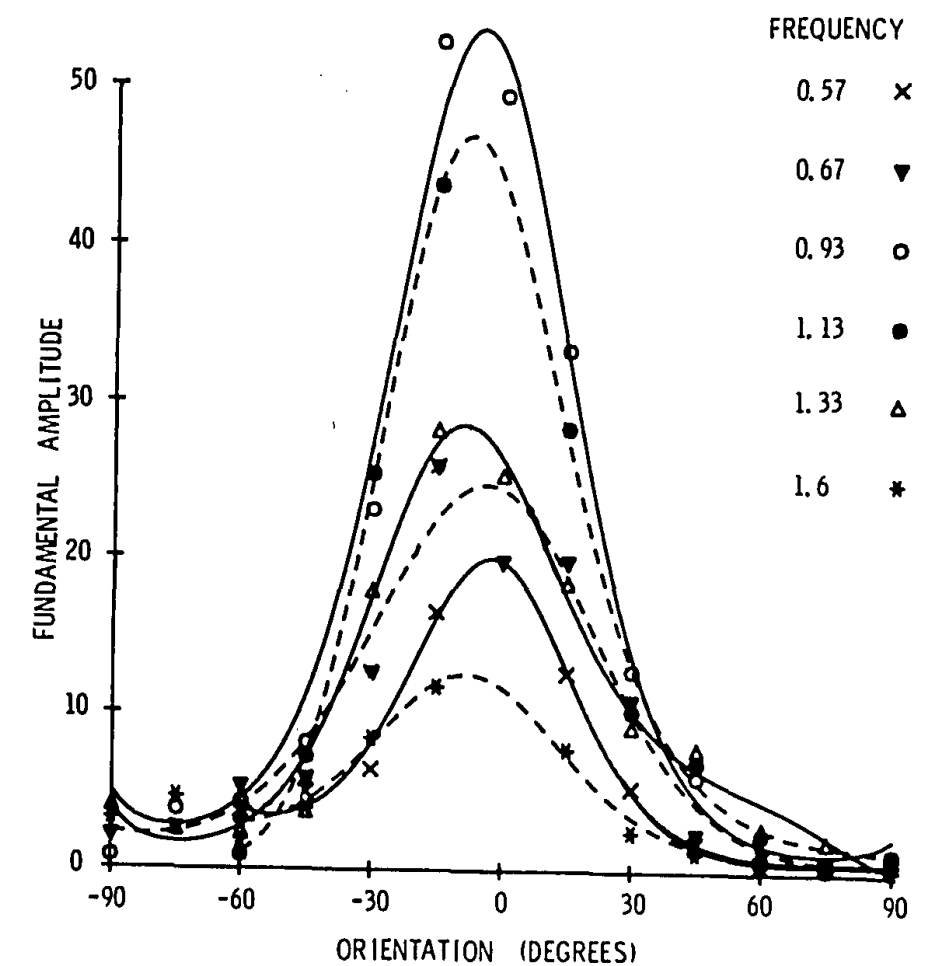


Tuning in the visual cortex

Next assignment!

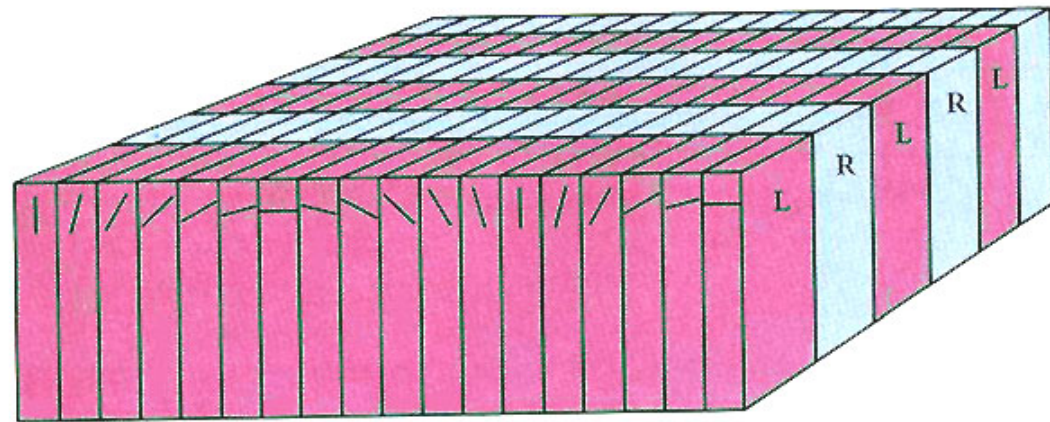


A
CAT SIMPLE CELL (#3)



Computing with V1

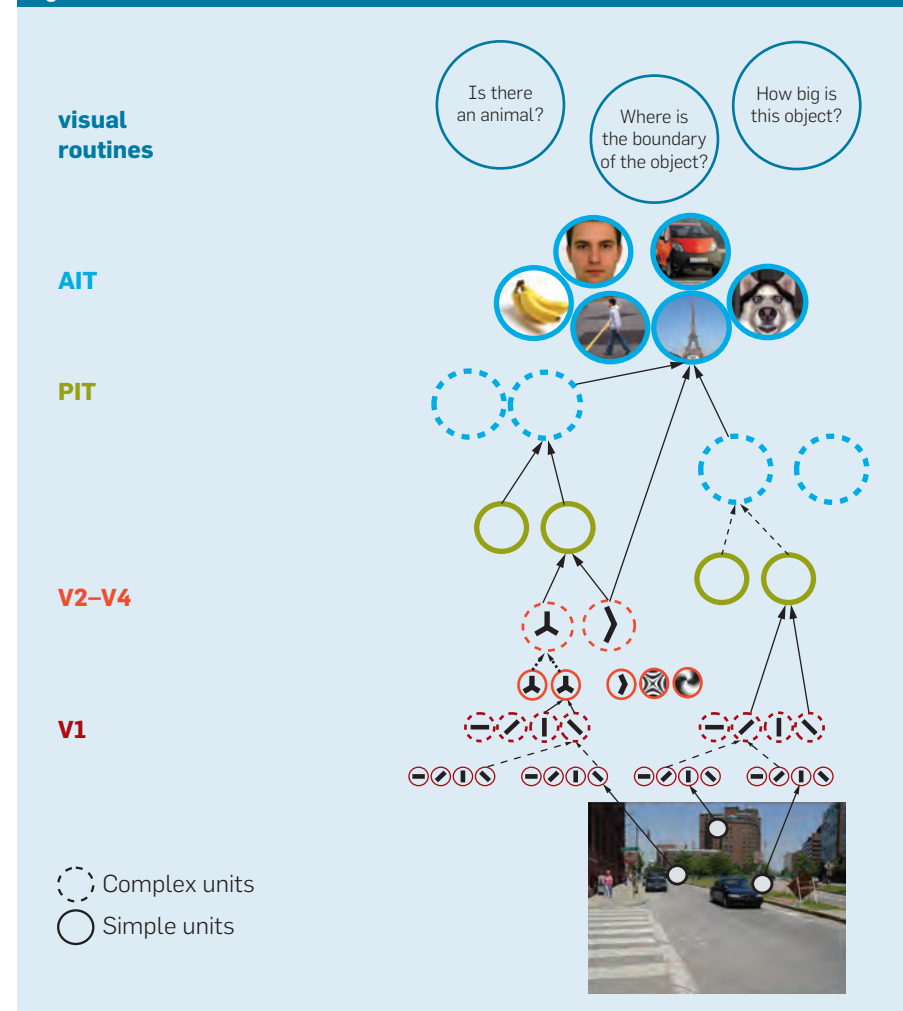
Gabor filters



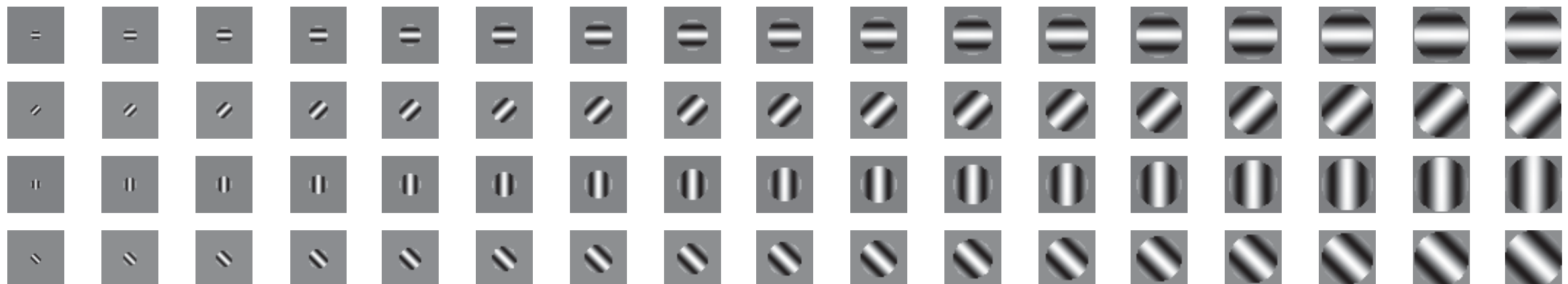
Orientation and ocular dominance columns

Figure 23. The ice-cube model of the cortex. It illustrates how the cortex is divided, at the same time, into two kinds of slabs, one set of ocular dominance (left and right) and one set for orientation. The model should not be taken literally: Neither set is as regular as this, and the orientation slabs especially are far from parallel or straight.

Figure 2. Hierarchical feedforward models of the visual cortex.

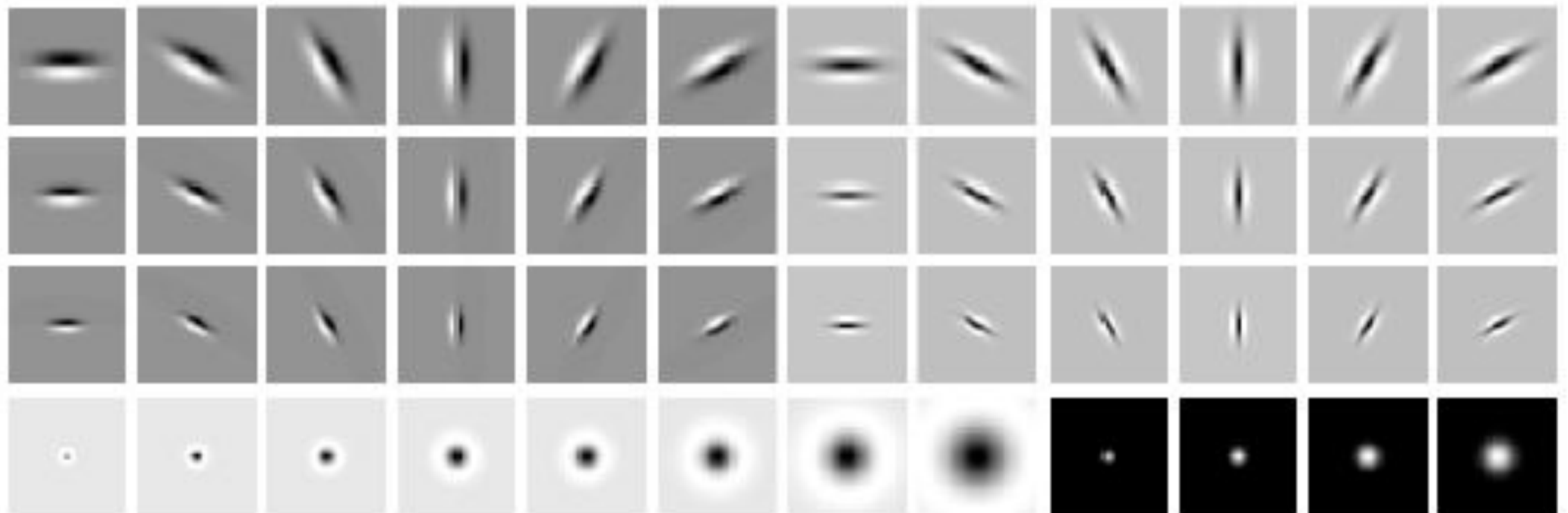
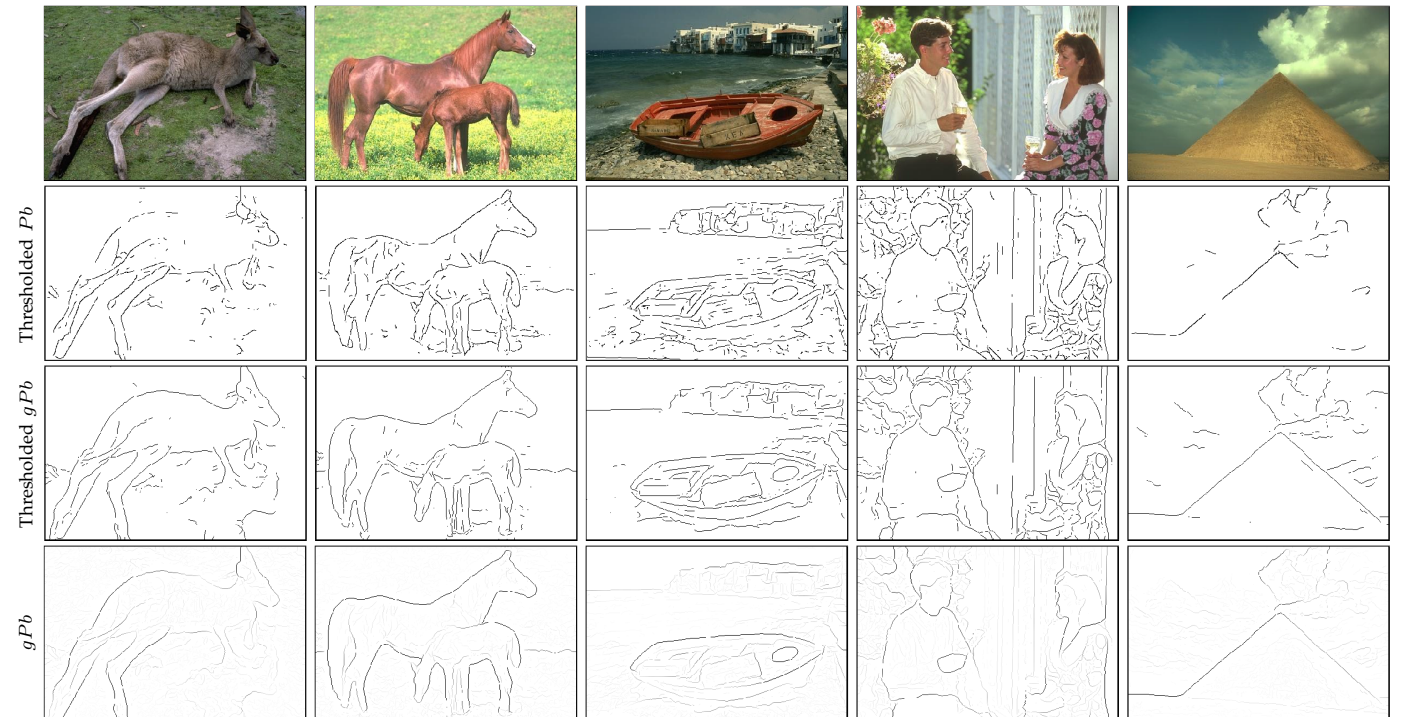


Gabor filters at multiple phases (one phase shown), orientations and spatial frequencies/scales (parameters derived from available experimental data)



Computing with V1

Gaussian derivatives



Computing with V1

Gist descriptor

