

# Background: Neural constraints

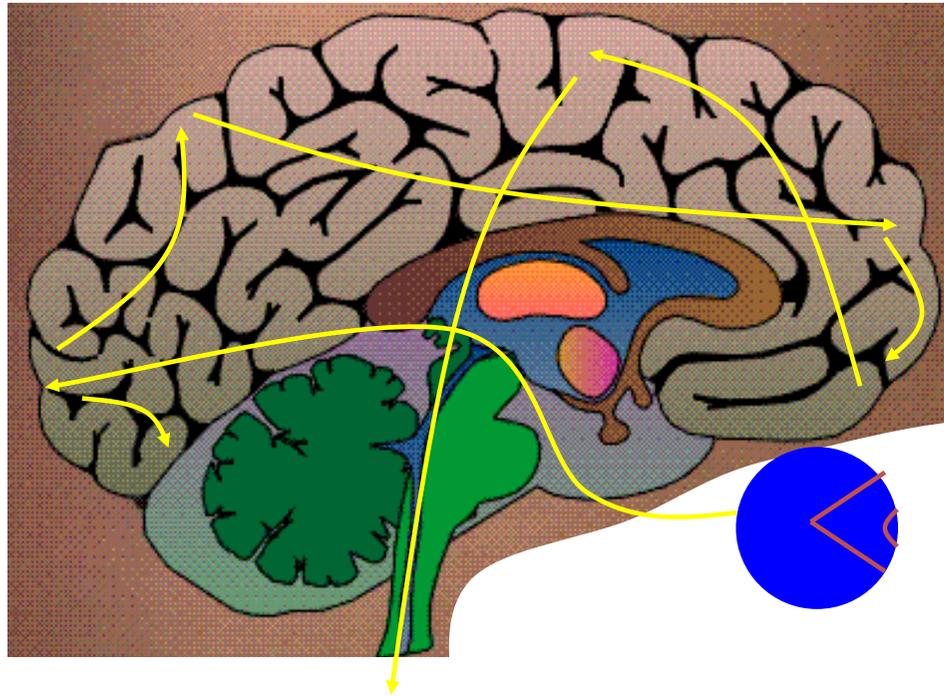
Gregor Schöner  
[dynamicfieldtheory.org](http://dynamicfieldtheory.org)

# the brain

motor  
cortex

frontal  
cortex

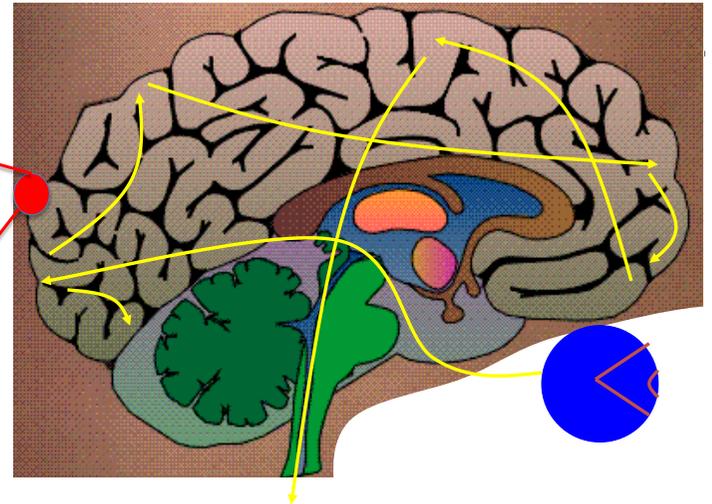
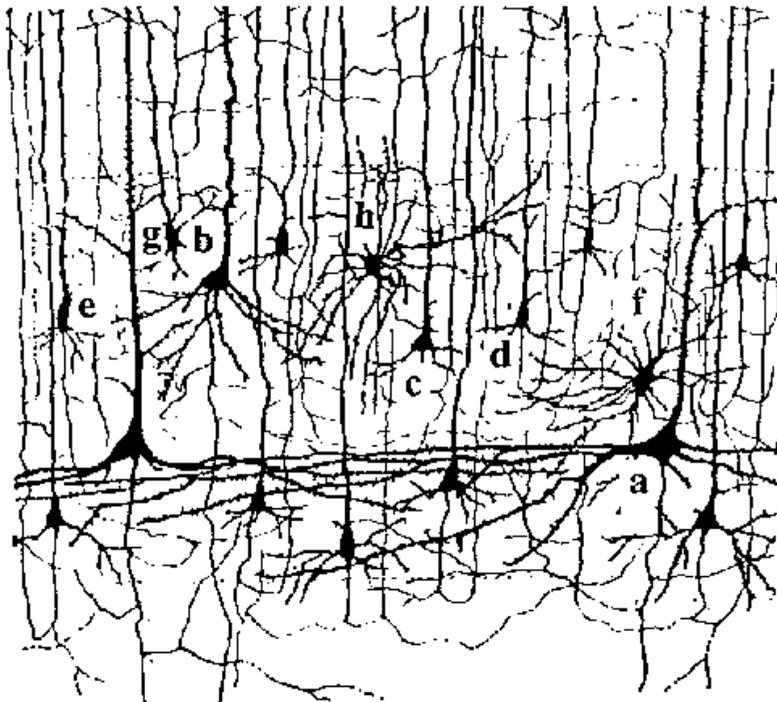
visual  
cortex



to motor  
output

[from: Tresilian: Sensorimotor  
Control and Learning: An  
introduction to the behavioral  
neuroscience of action 2012]

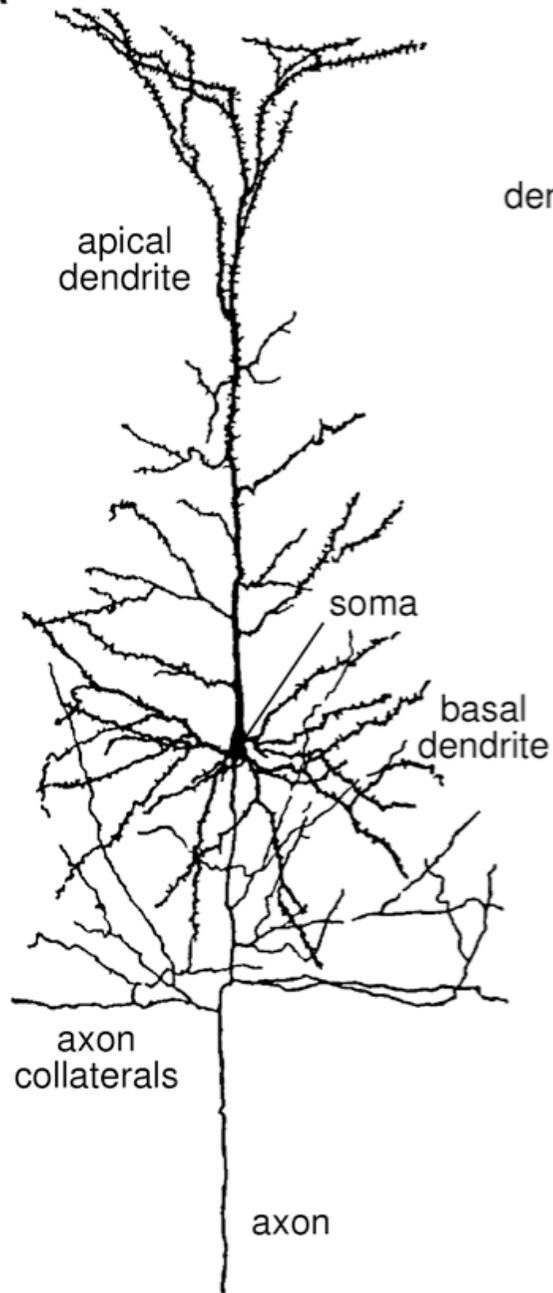
# neurons



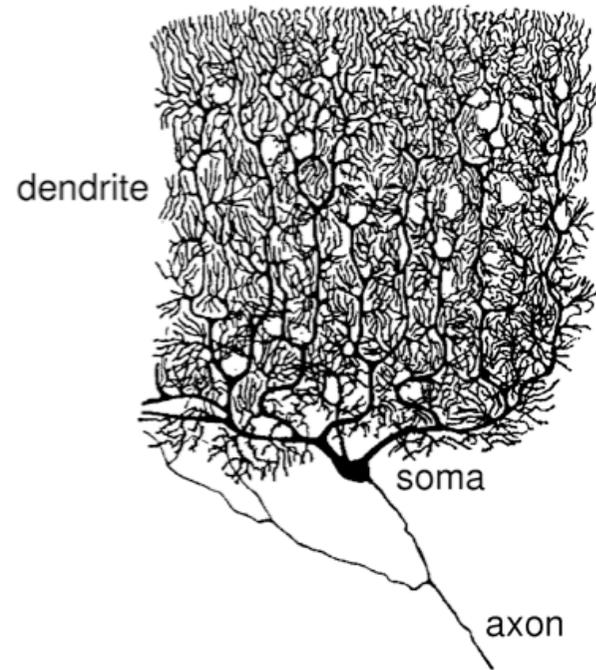
$\sim 10^{11}$  with 10000 synapses each

# neurons

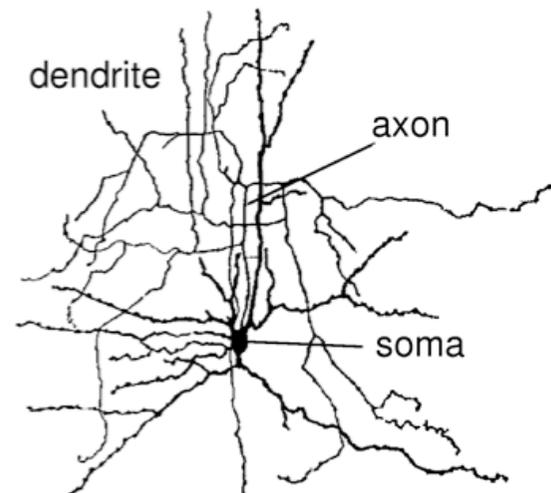
A



B

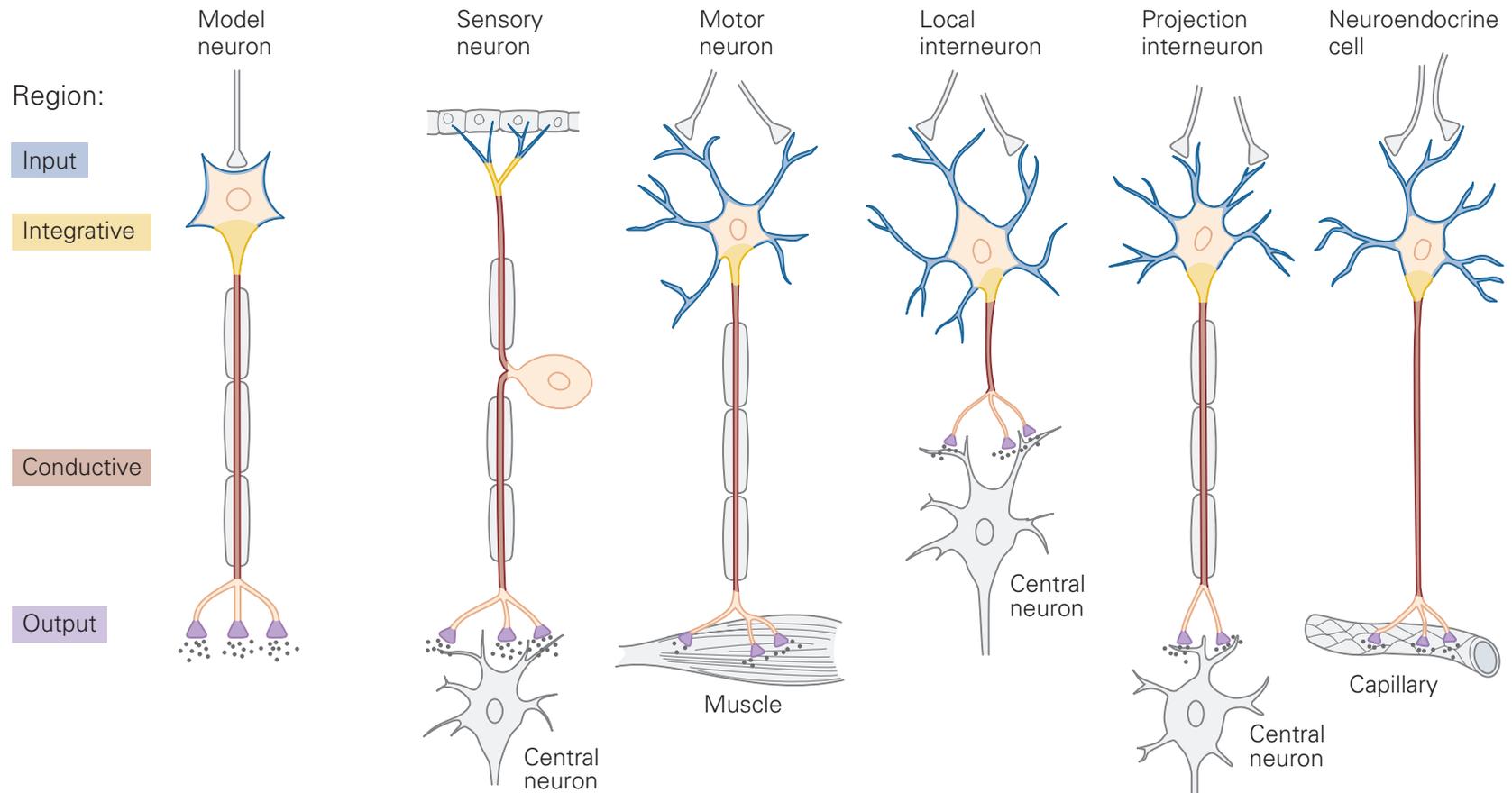


C



# neurons

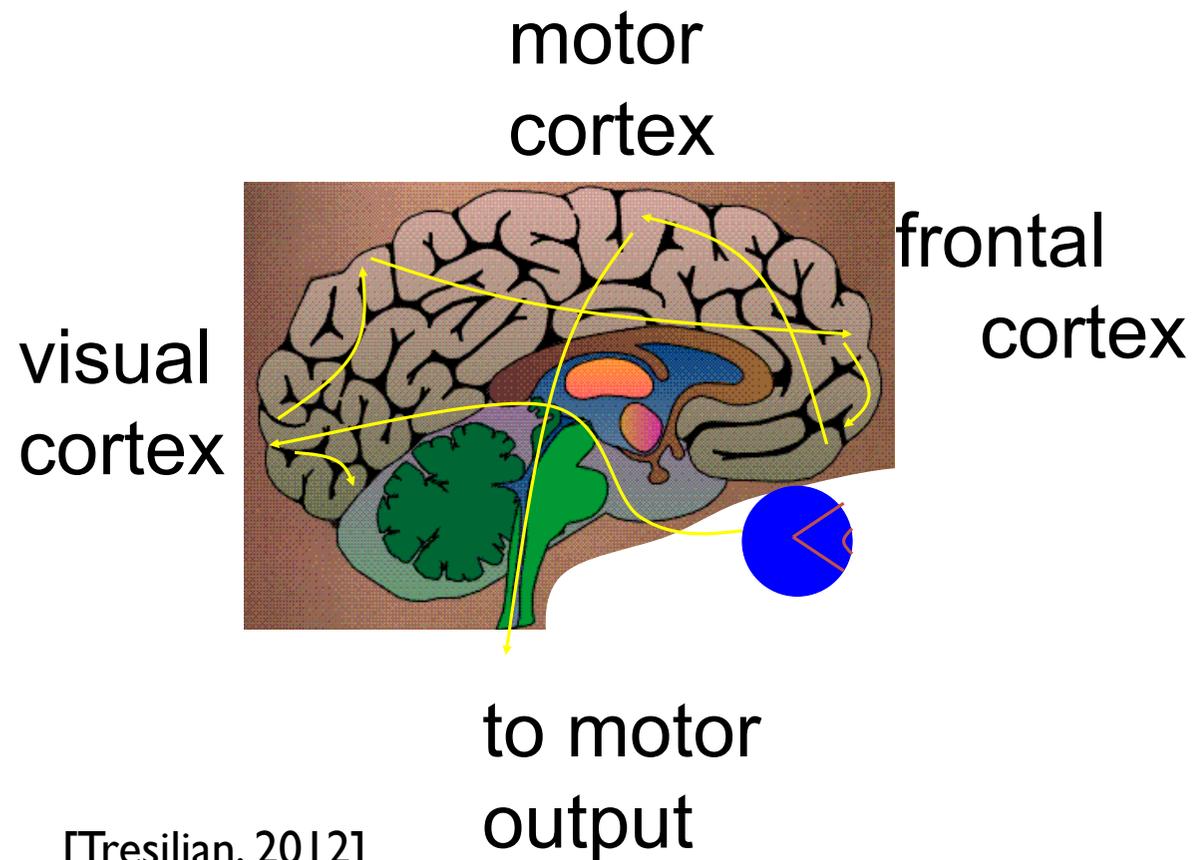
## ■ four components of neurons



# Functional analysis of the brain

■ vary conditions “outside” the brain: stimulus, motor task, cognitive task and relate to neural activity: **coding/decoding, cognitive neuroscience**

■ or the reverse: vary neural substrate (lesioning, optogenetics, etc) and observe what happens to behavior/competence: **neuropsychology**



- Neurophysics
- Sensors, actuators, rate code
- Receptive fields, tuning curves
- Maps
- Distributions of population representation
- Patterns of connectivity
- Synaptic dynamics

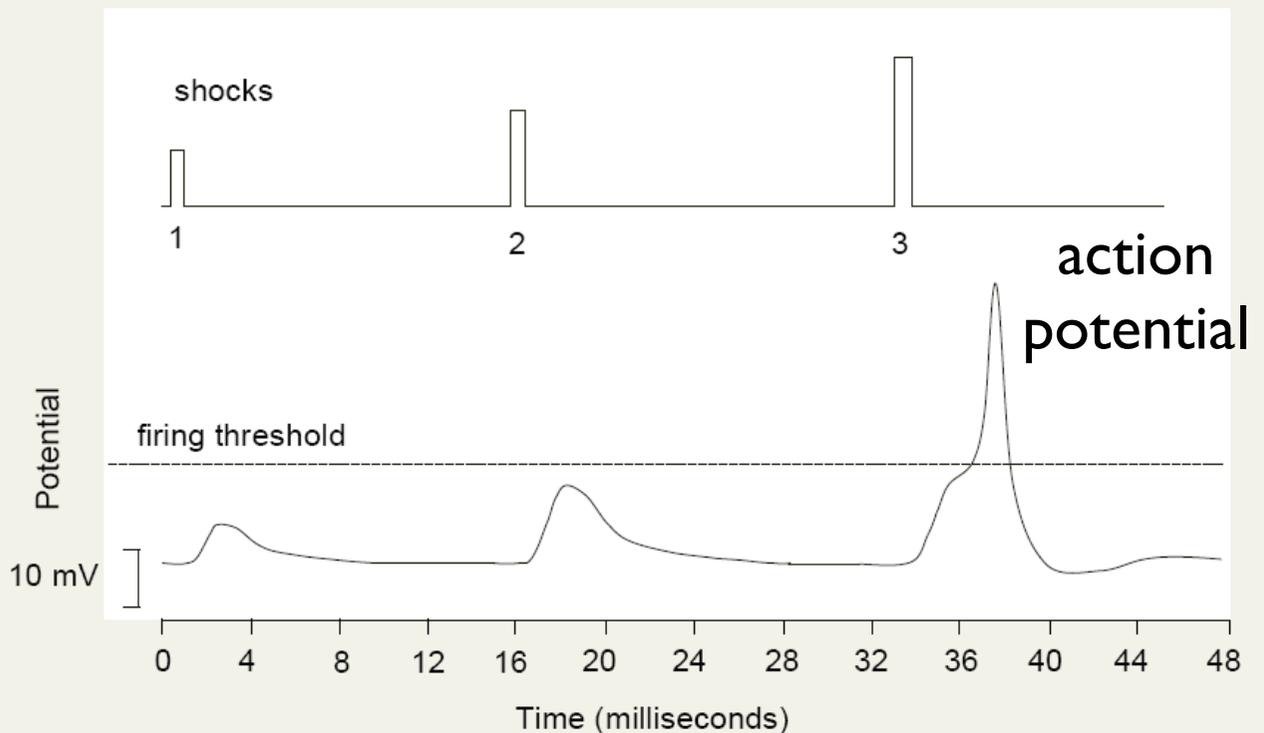
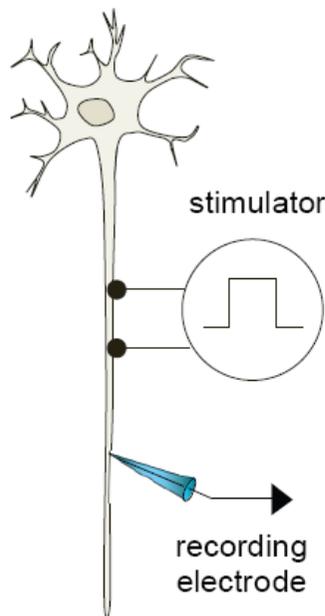
# Neuro-physics

- membrane potential,  $u(t)$ , evolves as a dynamical system

$$\tau \dot{u}(t) = -u(t) + h + \text{input}(t)$$

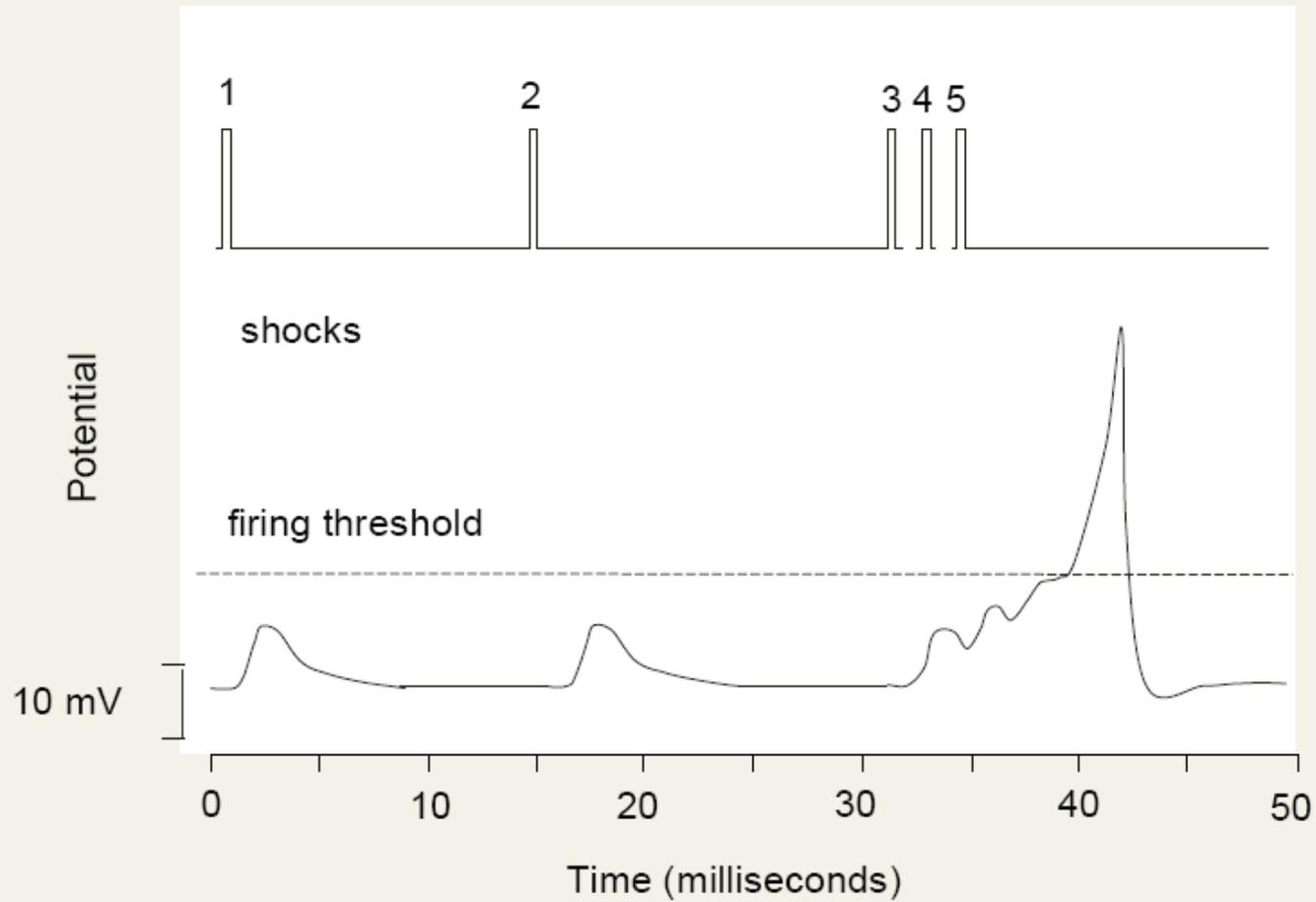
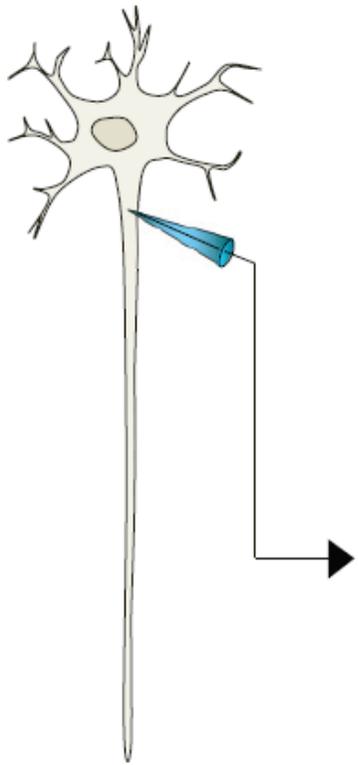
$\tau \approx 10$  ms time scale

- only when membrane potential exceeds a threshold is activation transmitted to downstream neurons



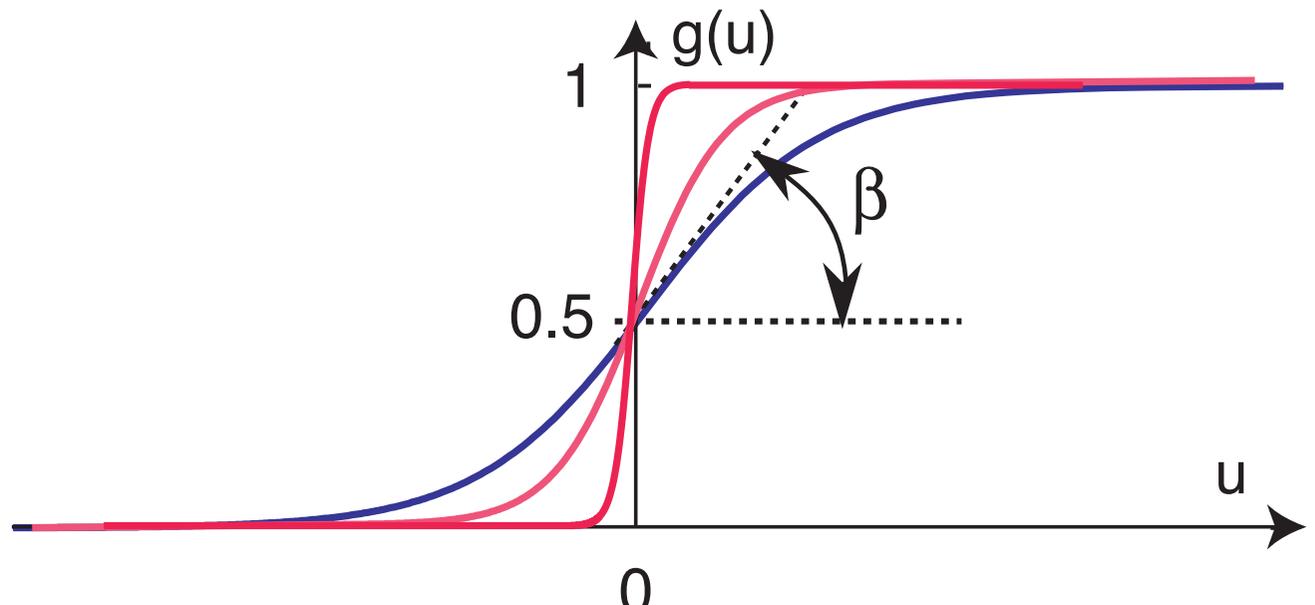
[from: Tresilian, 2012]

# temporal summation



# Neural dynamics

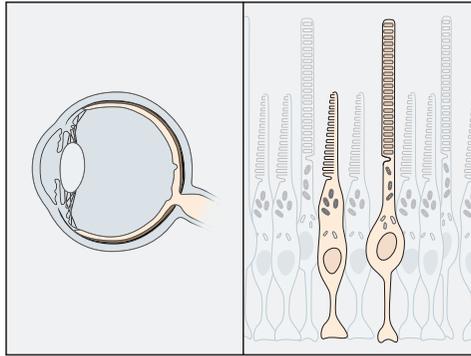
- replace spiking mechanism by sigmoid:
  - low levels of activation: not transmitted to downstream systems
  - high levels of activation: transmitted to downstream systems
- abstracting from biophysical details ~ **population level membrane potential**



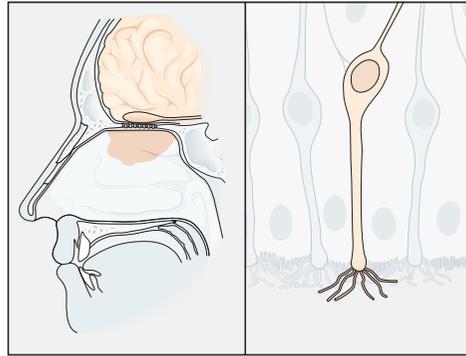
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# Sensor cells

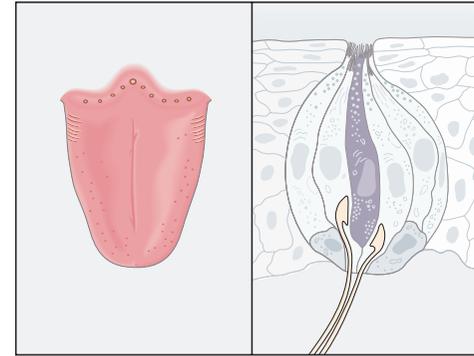
Vision



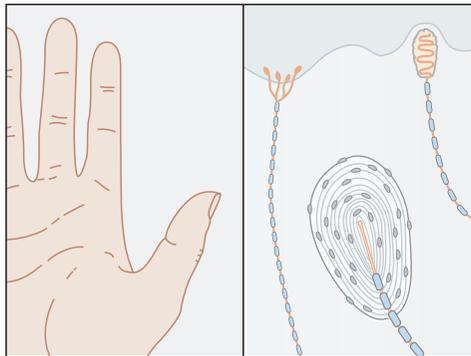
Smell



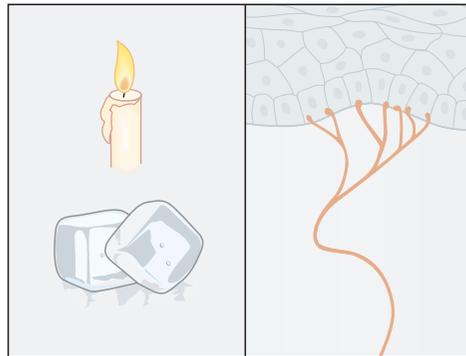
Taste



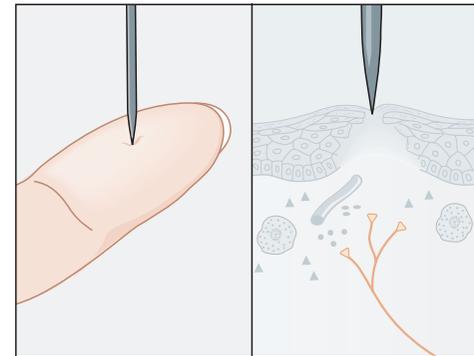
Touch



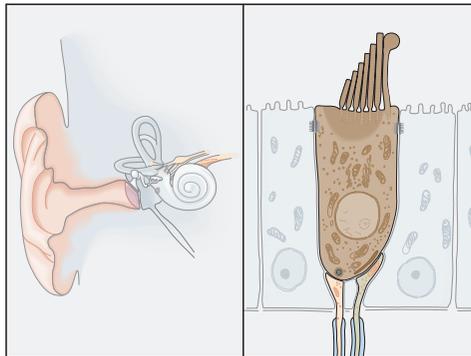
Thermal senses



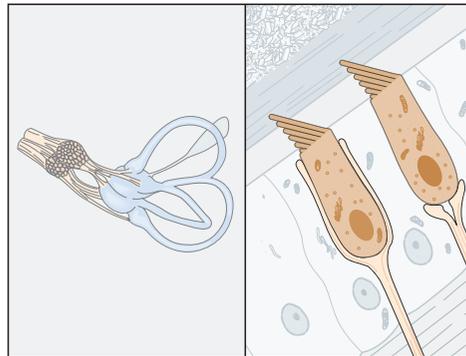
Pain



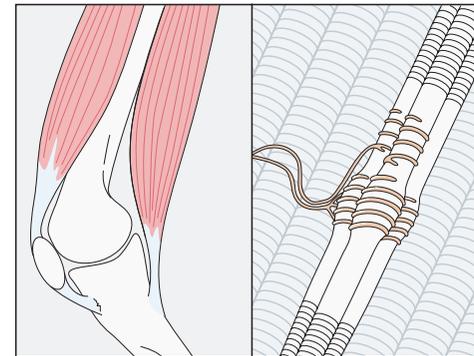
Hearing



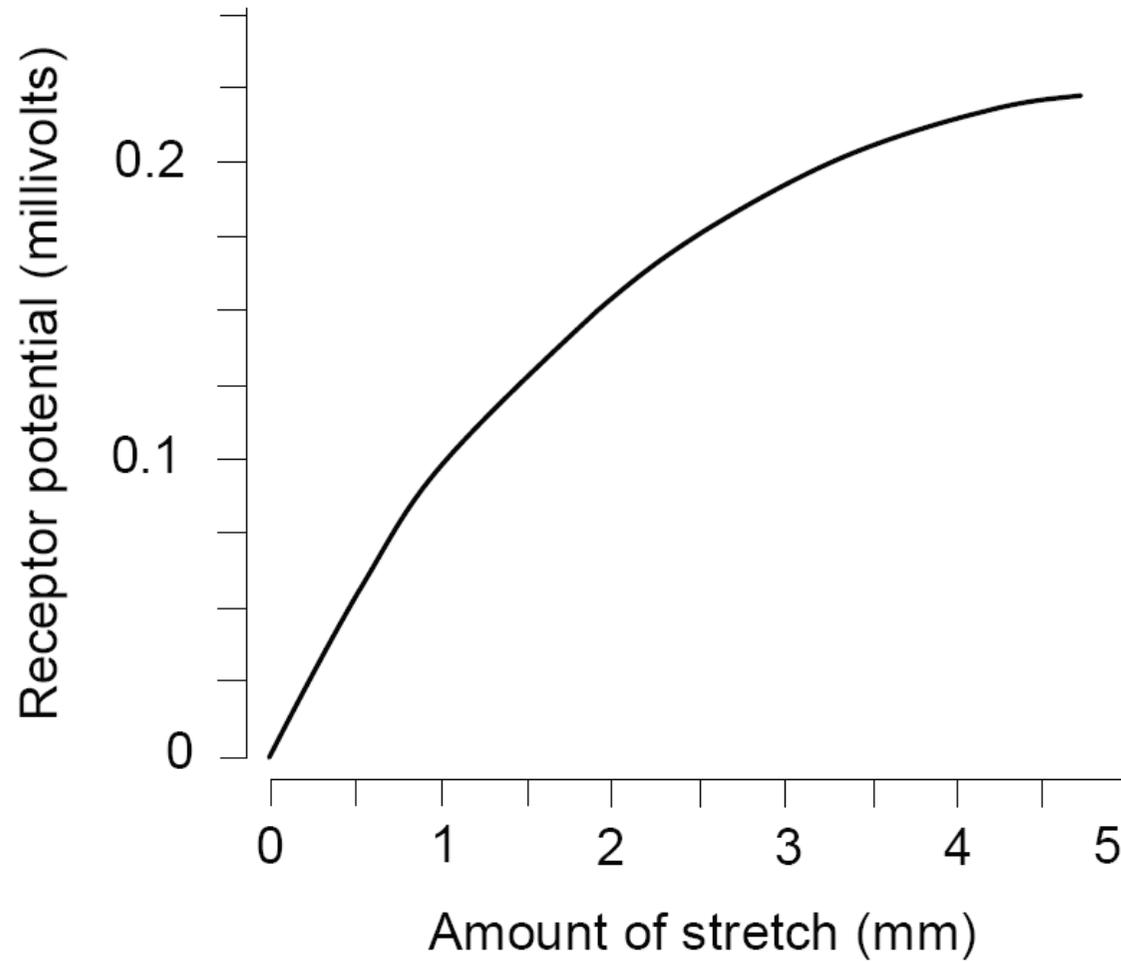
Balance



Proprioception



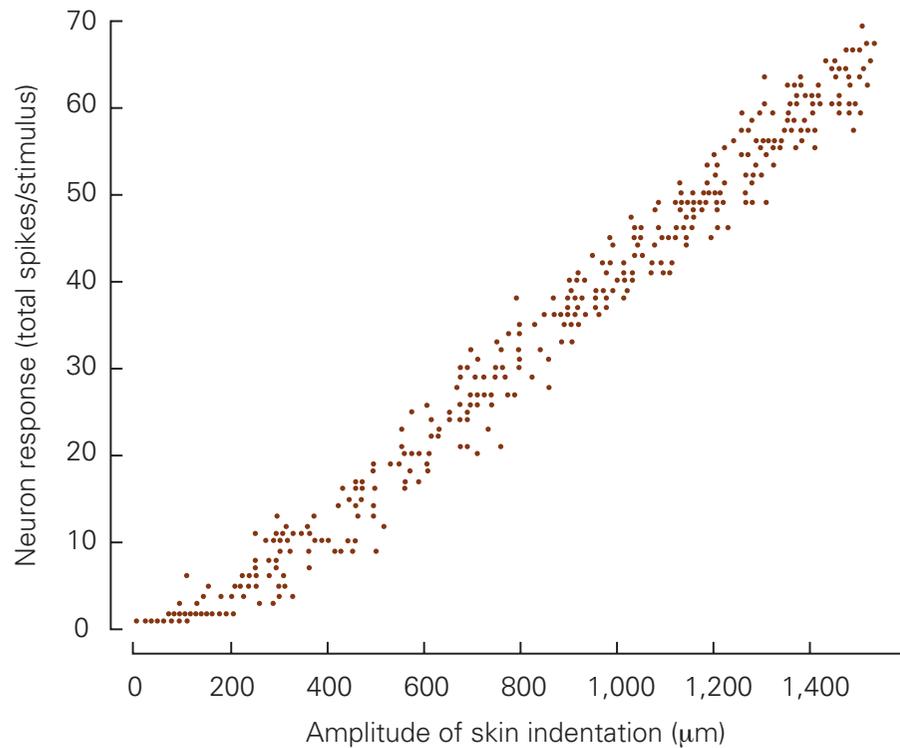
# Sensor characteristic



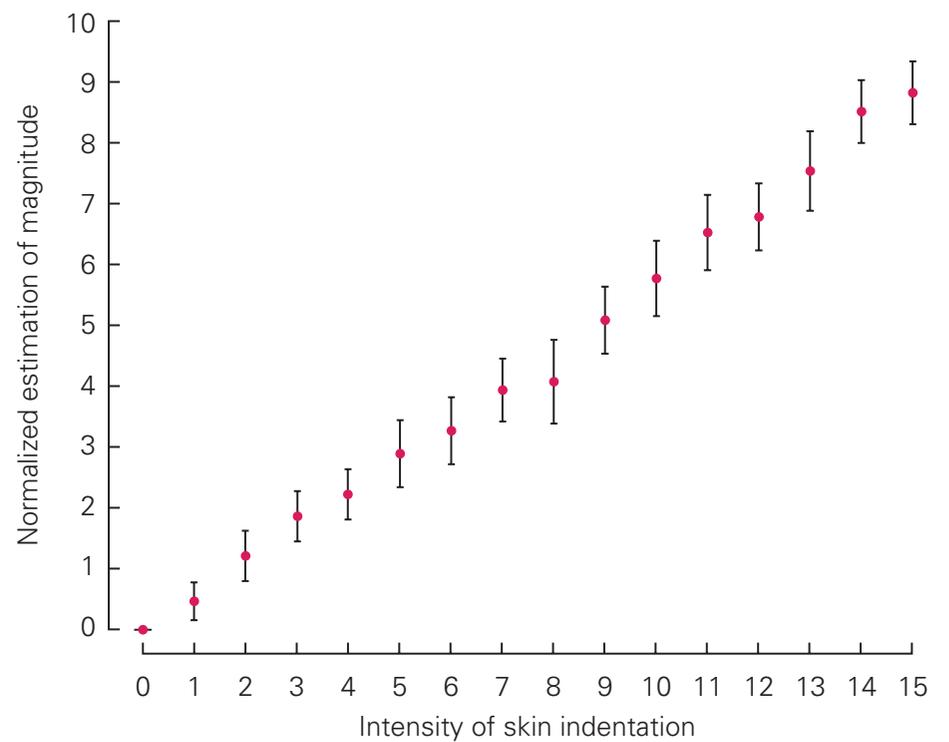
[from: Tresilian, 2012]

# Sensor characteristic

A Neural code of stimulus magnitude



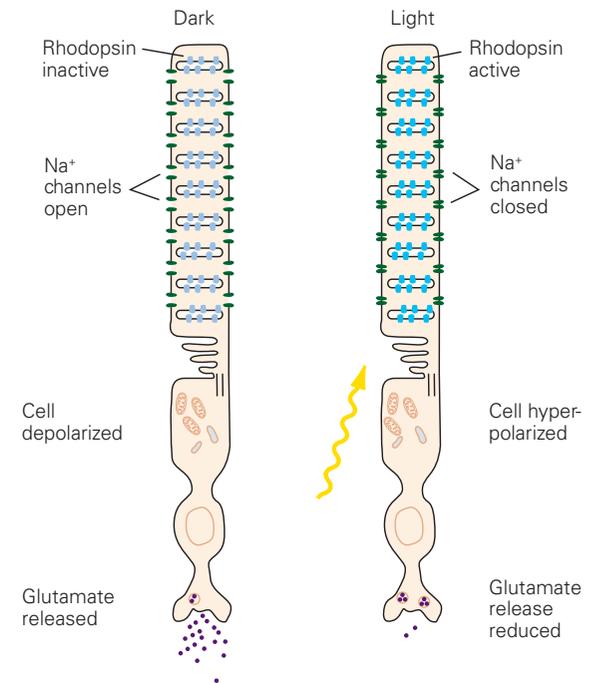
B Perceived sensation intensity



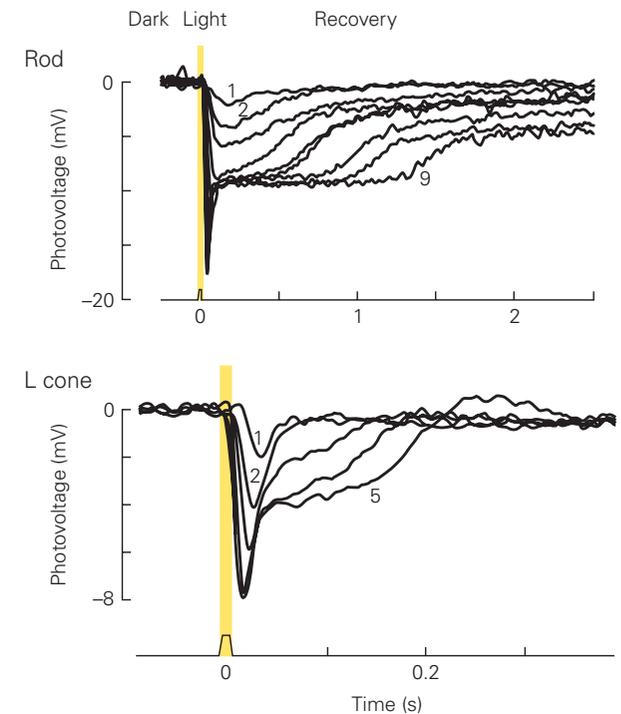
# Sensor characteristic

## ■ photoreceptors

A Phototransduction and neural signaling

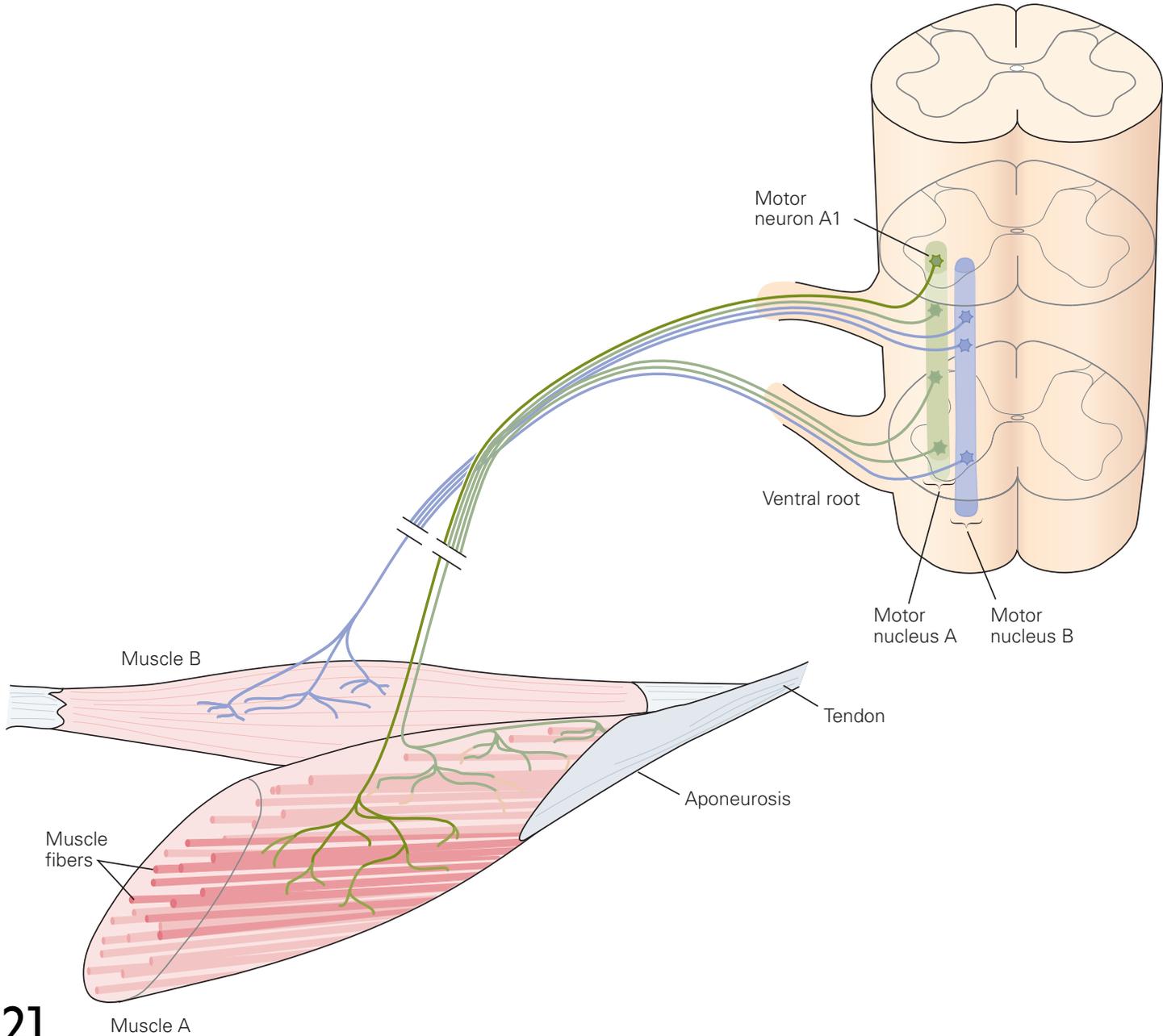


C Voltage response to light



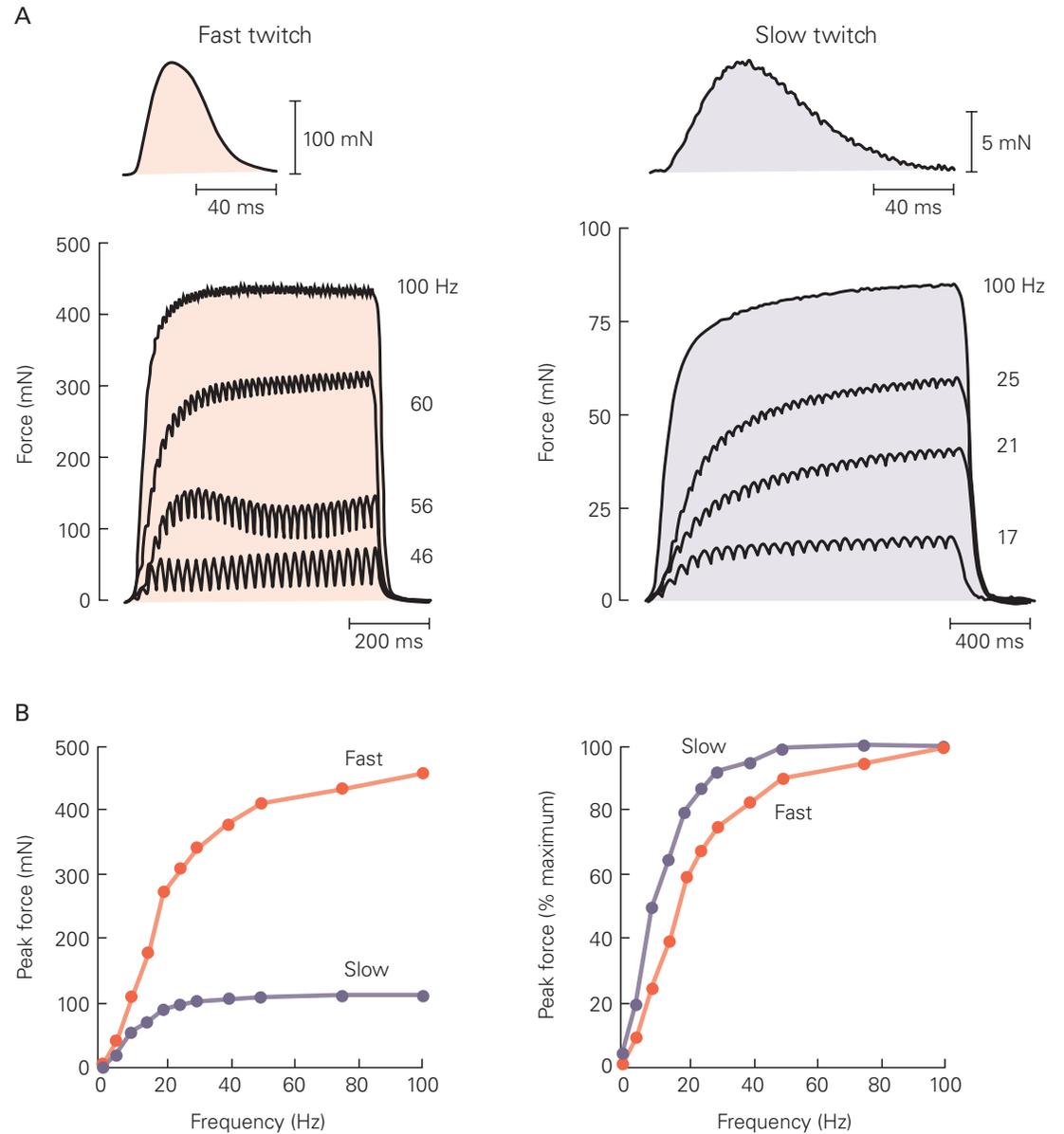
[from: Tresilian, 2012]

# Motor neurons



[from: Tresilian, 2012]

# Motor neurons

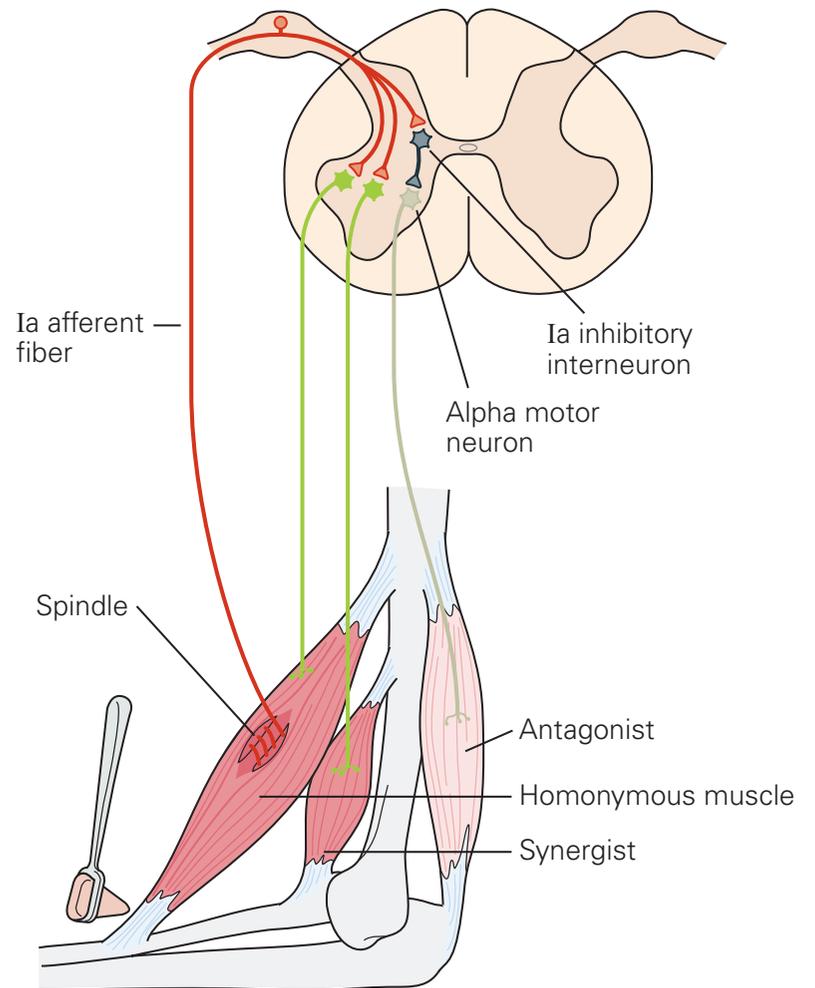


[from: Tresilian, 2012]

# Peripheral neural circuits

■ stretch reflex

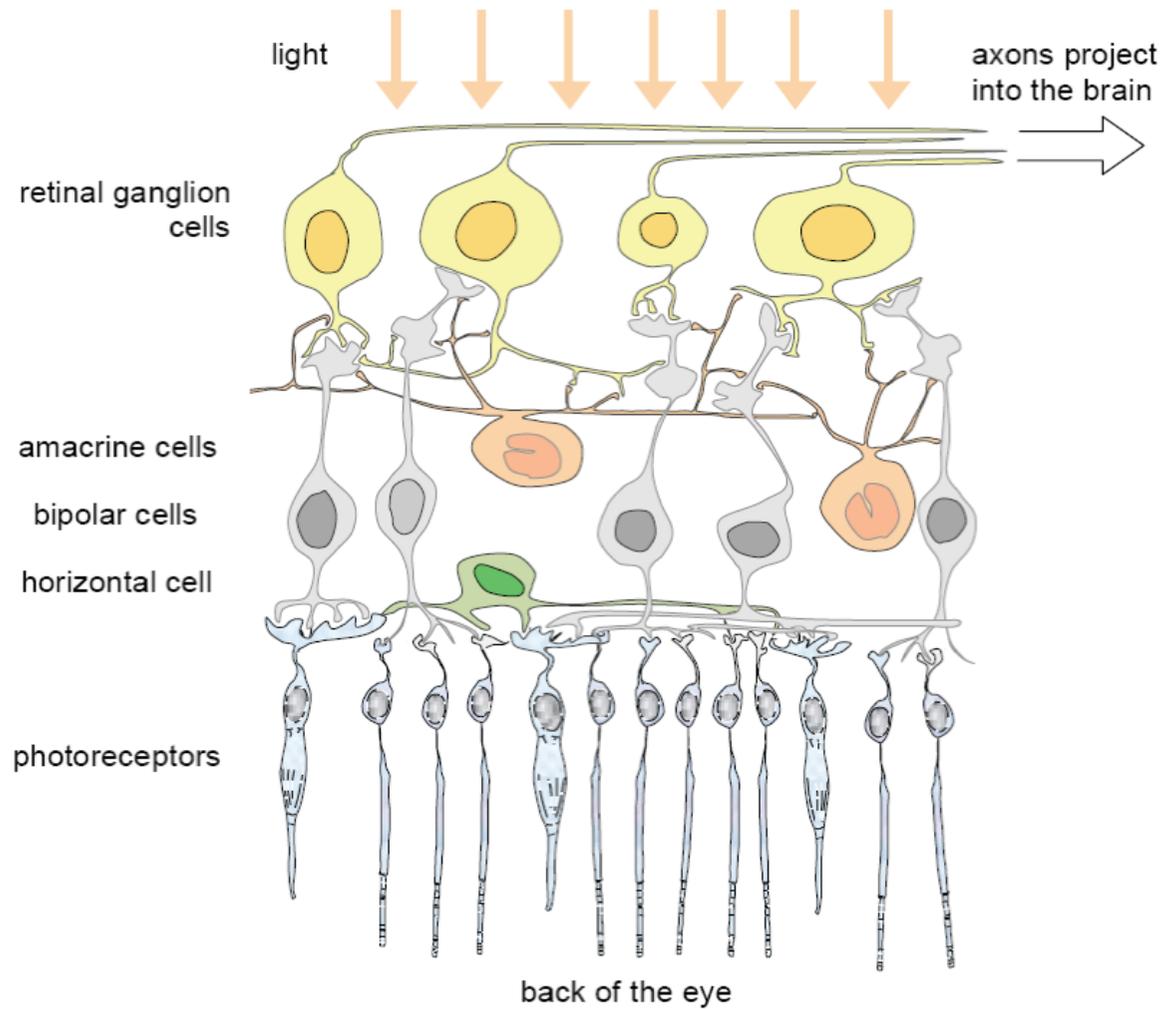
B Monosynaptic pathways (stretch reflex)



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# Sensory networks

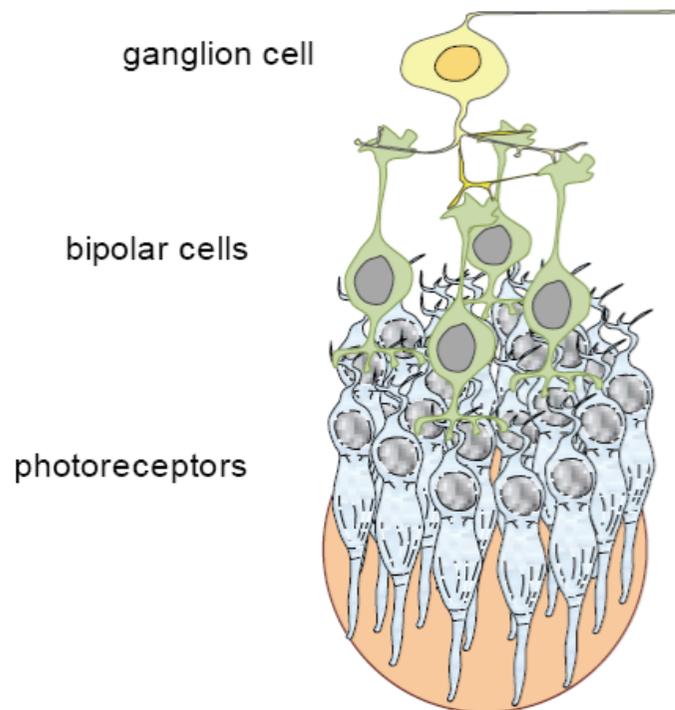
## retinal network



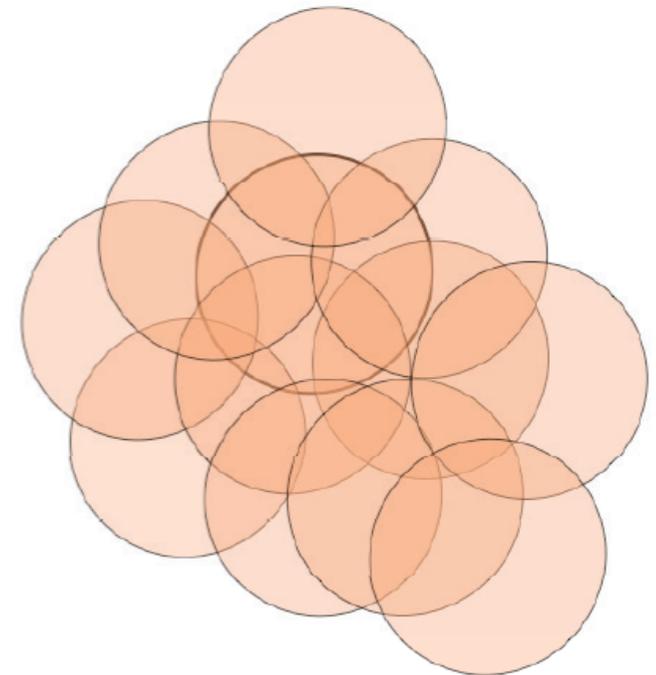
# Sensory networks

## ■ receptive fields

(A) A ganglion cell is connected to the photoreceptors occupying a roughly circular region



(B) Receptive fields of neighboring ganglion cells overlap

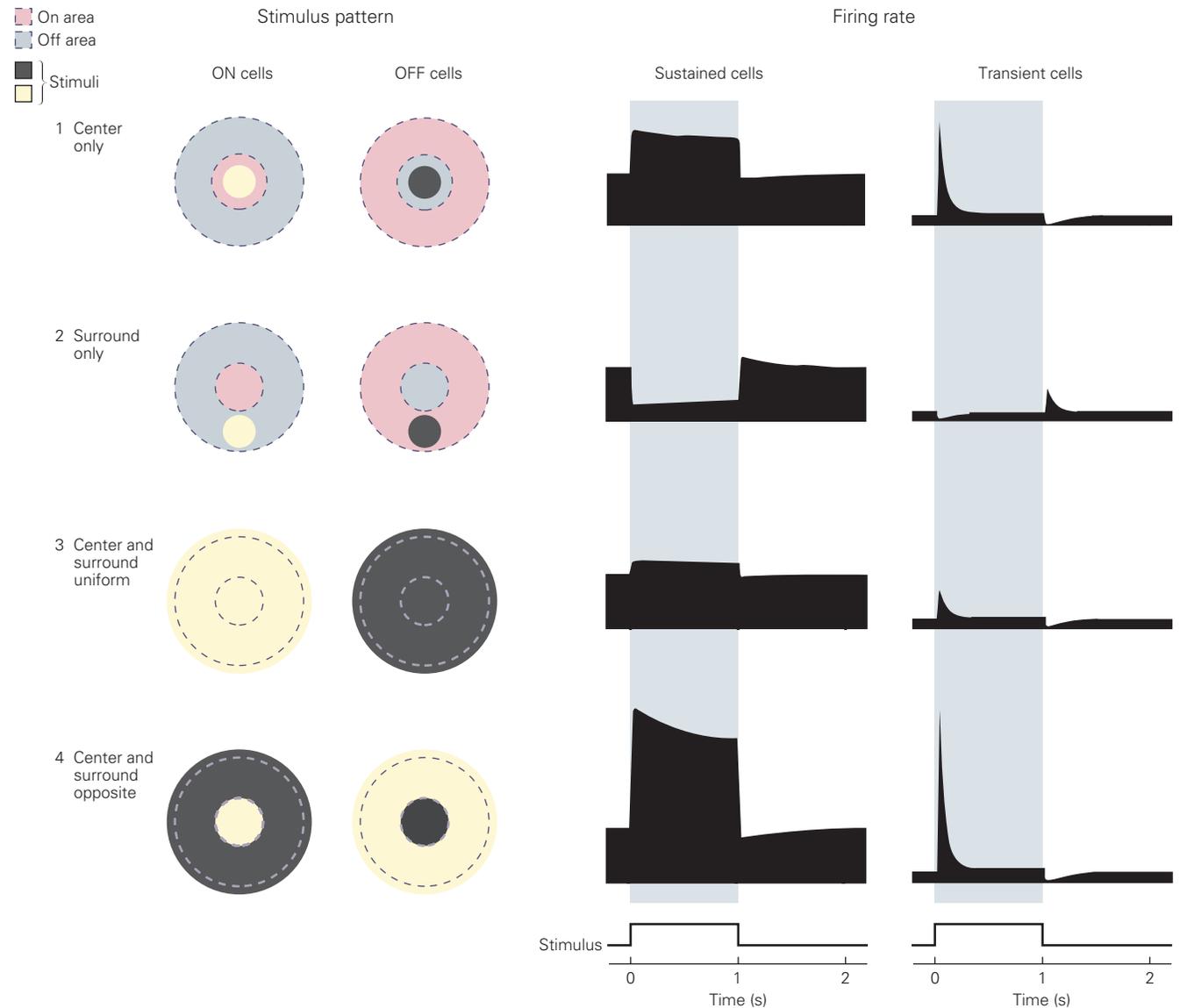


[from:  
Tresilian,  
2012]

# Sensory networks

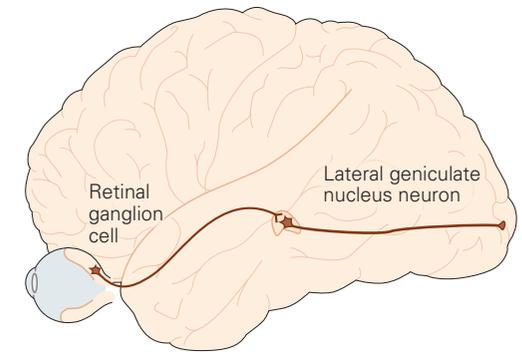
■ space-time structure of receptive fields

■ retinal ganglion cells

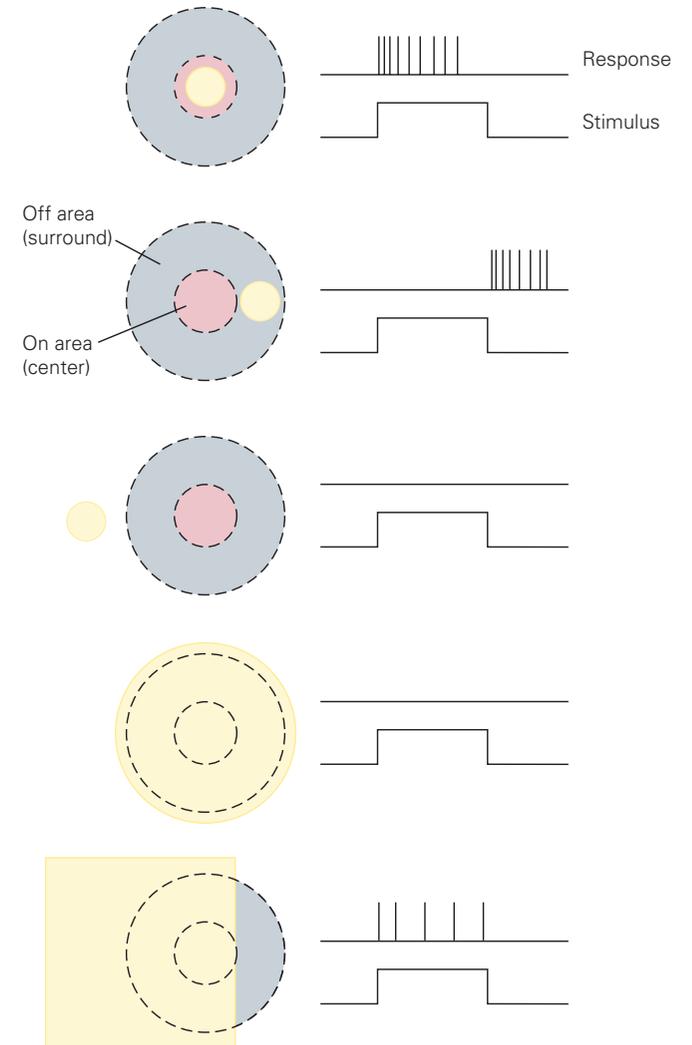


[from: Tresilian, 2012]

# Sensory networks



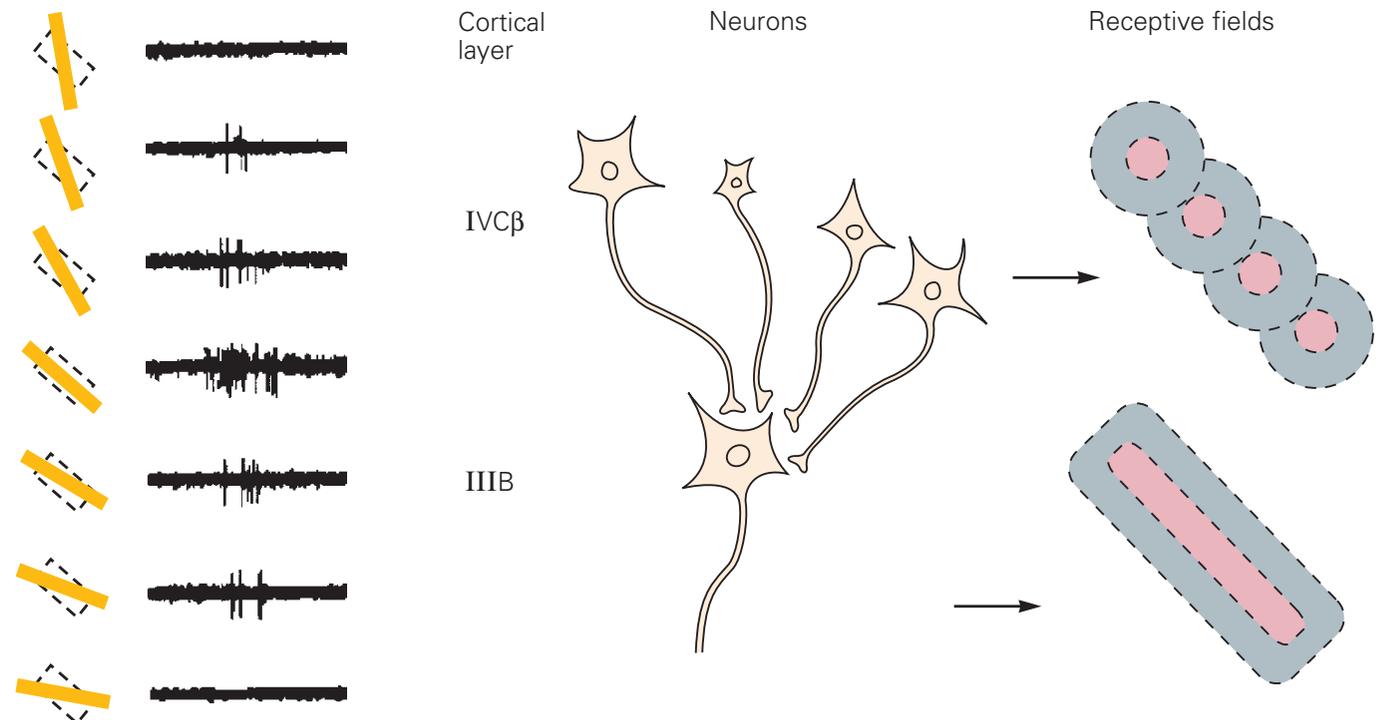
 LGN



[from: Tresilian, 2012]

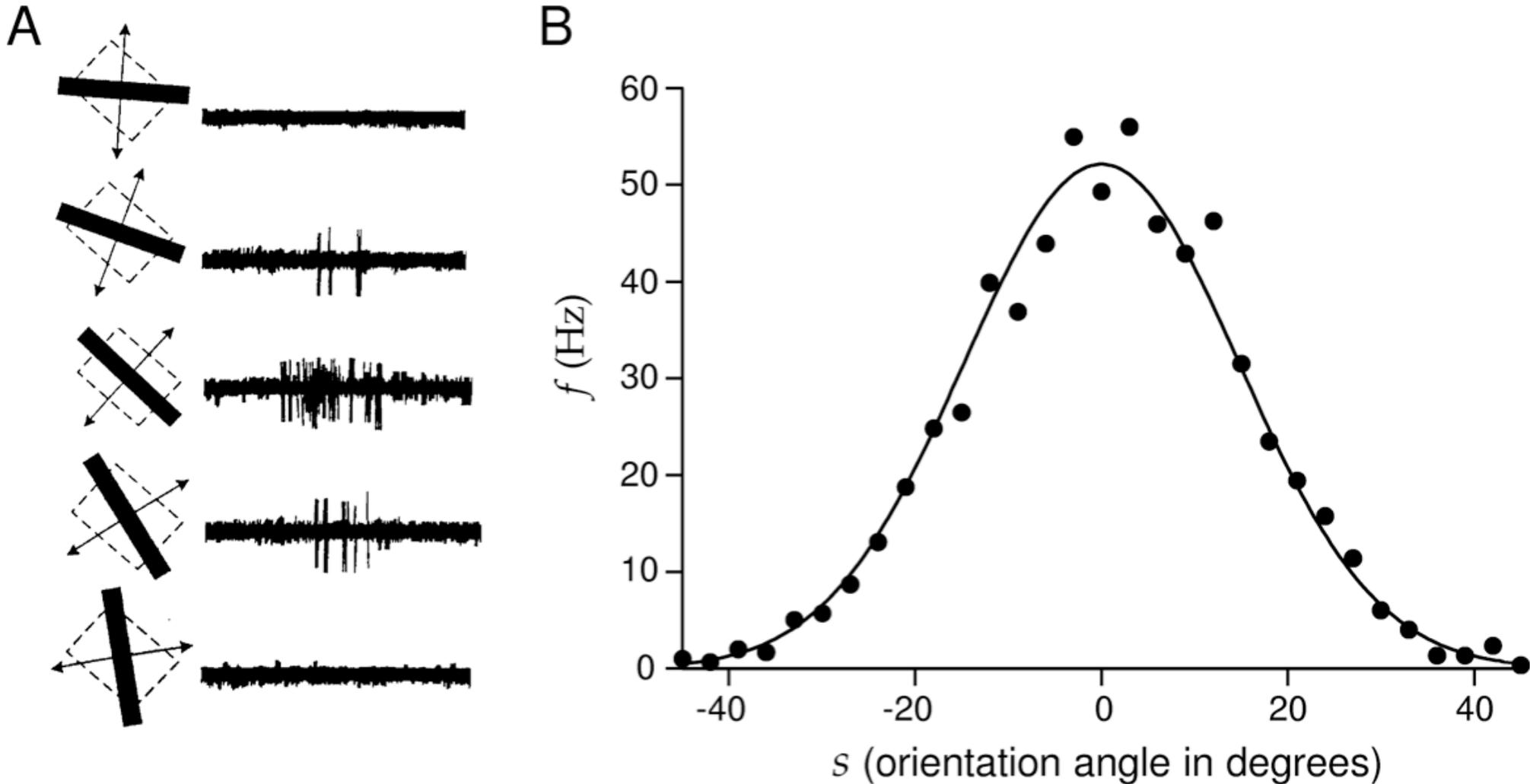
# Sensory networks

## ■ tuning curves in primary visual cortex

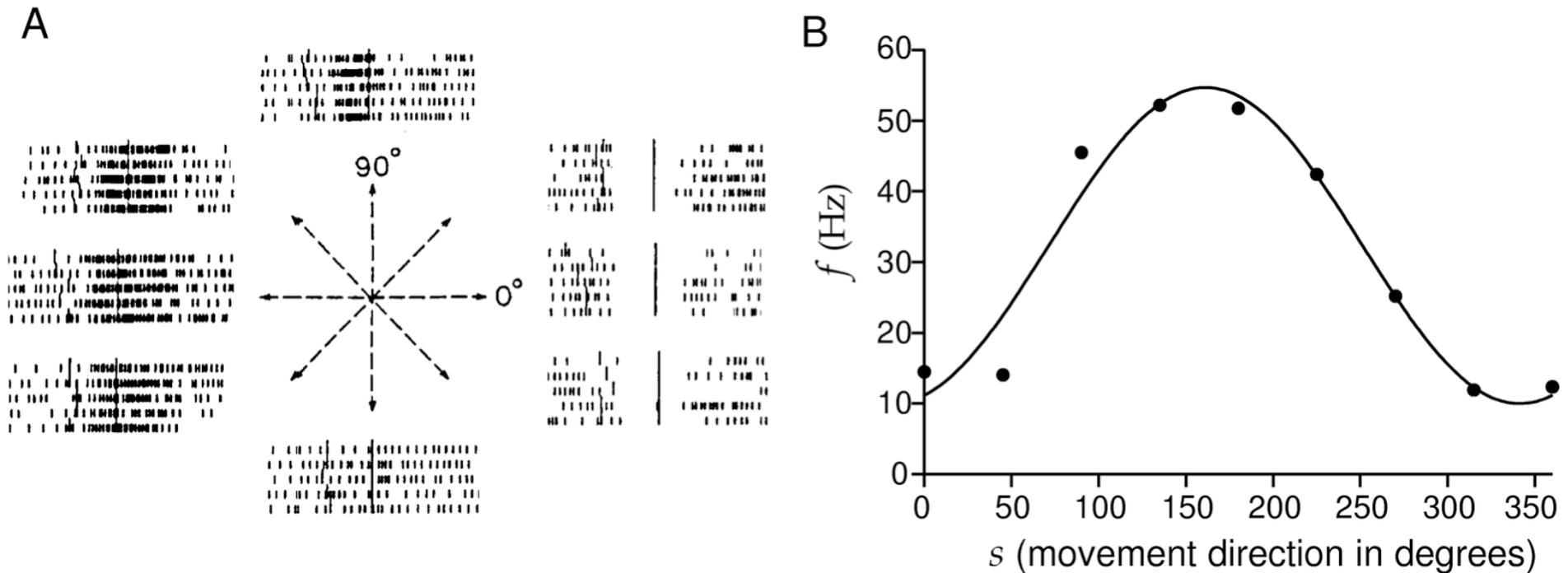


[from: Tresilian, 2012]

# Hubel, Wiesel: Tuning curve in macaque primary visual cortex

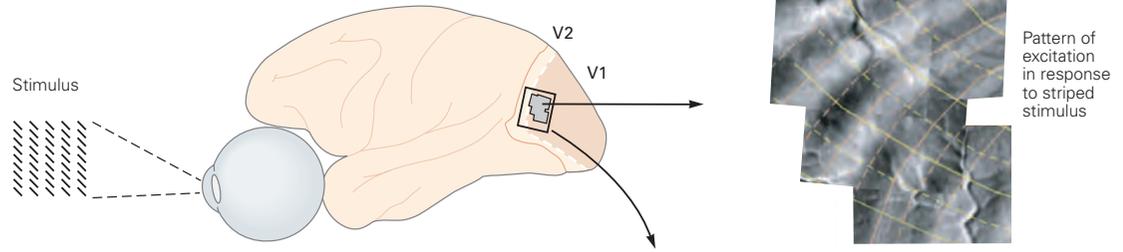


# Tuning curve in macaque primary motor cortex



- Neurophysics
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# Neural maps

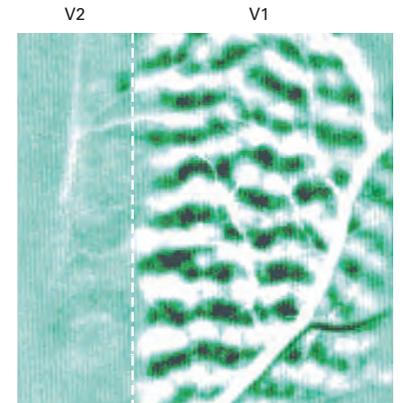
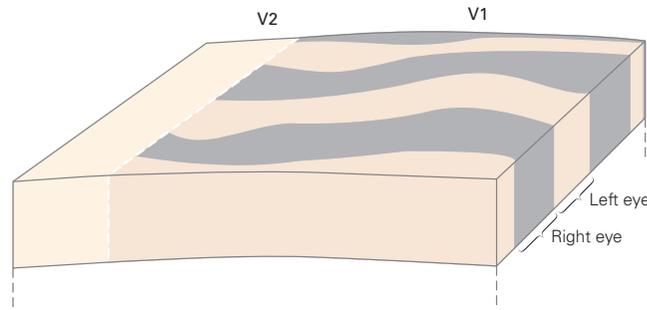


■ tuning curves studied systematically across the cortical surface

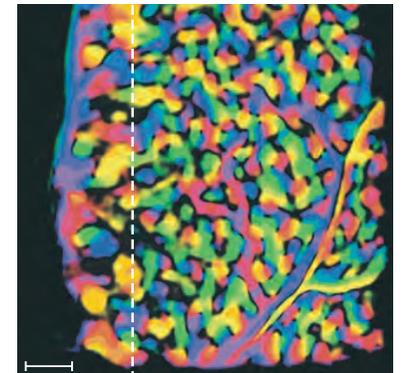
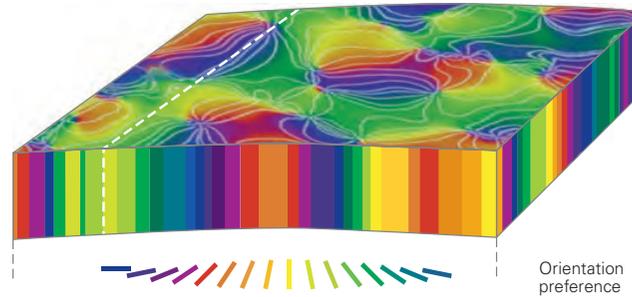
■ => feature maps

■ topography

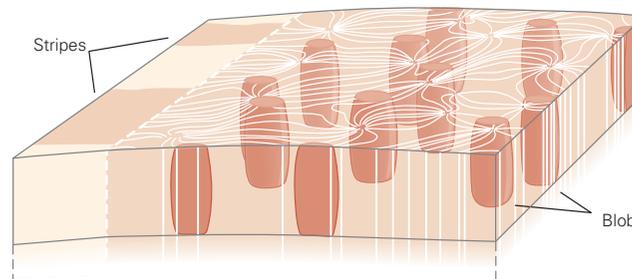
B Ocular dominance columns



C Orientation columns

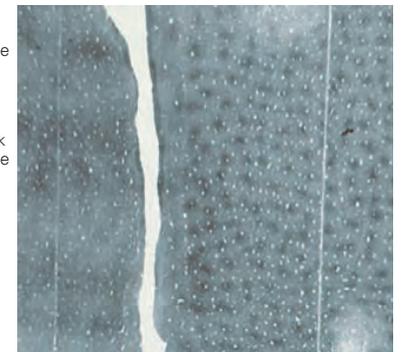


D Blobs, interblobs (V1), and stripes (V2)



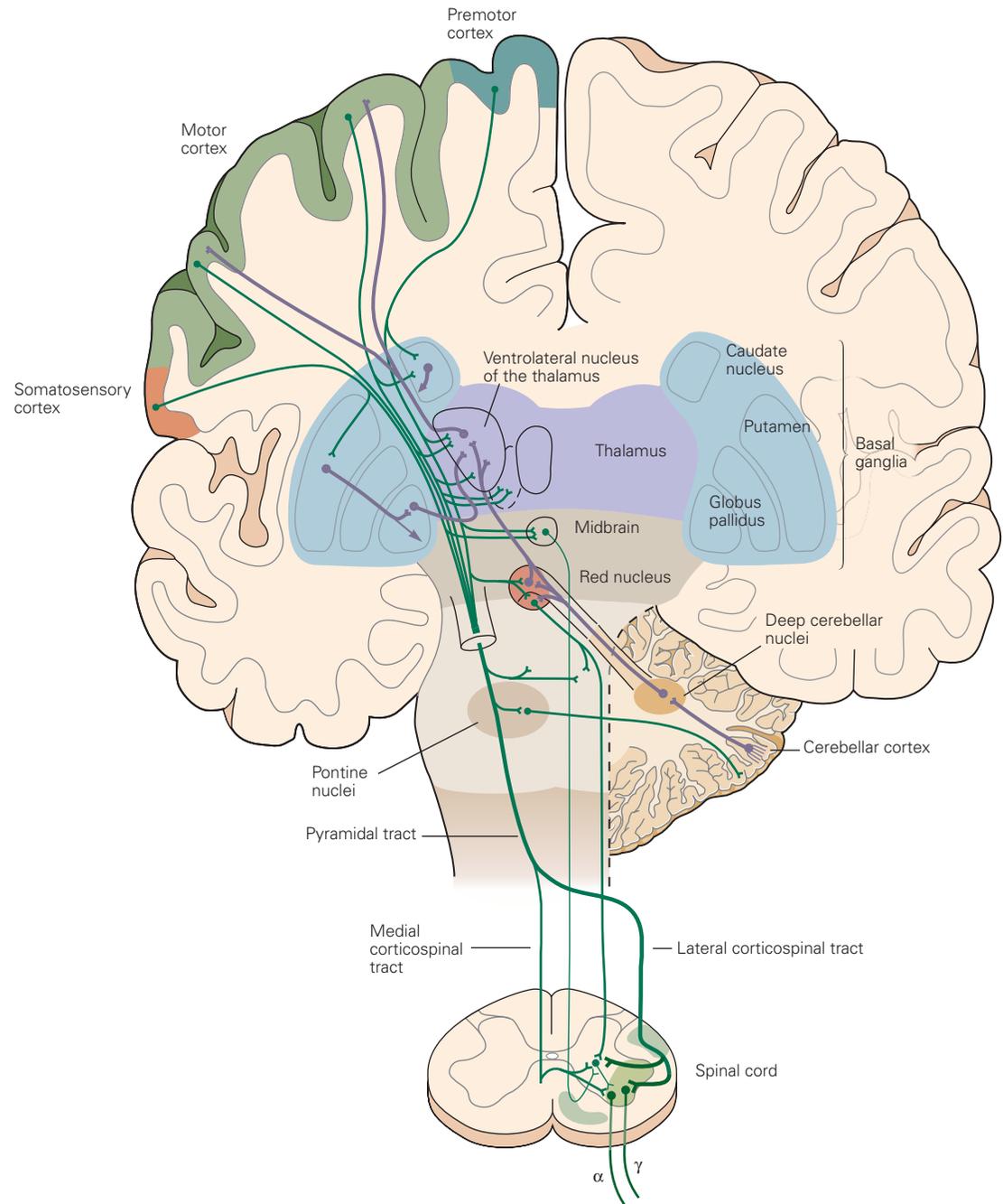
Thin stripe

Thick stripe



[from: Charles D. Gilbert, Aniruddha Das, Chapter 21 of Kandel et al 2021]

# Motor networks

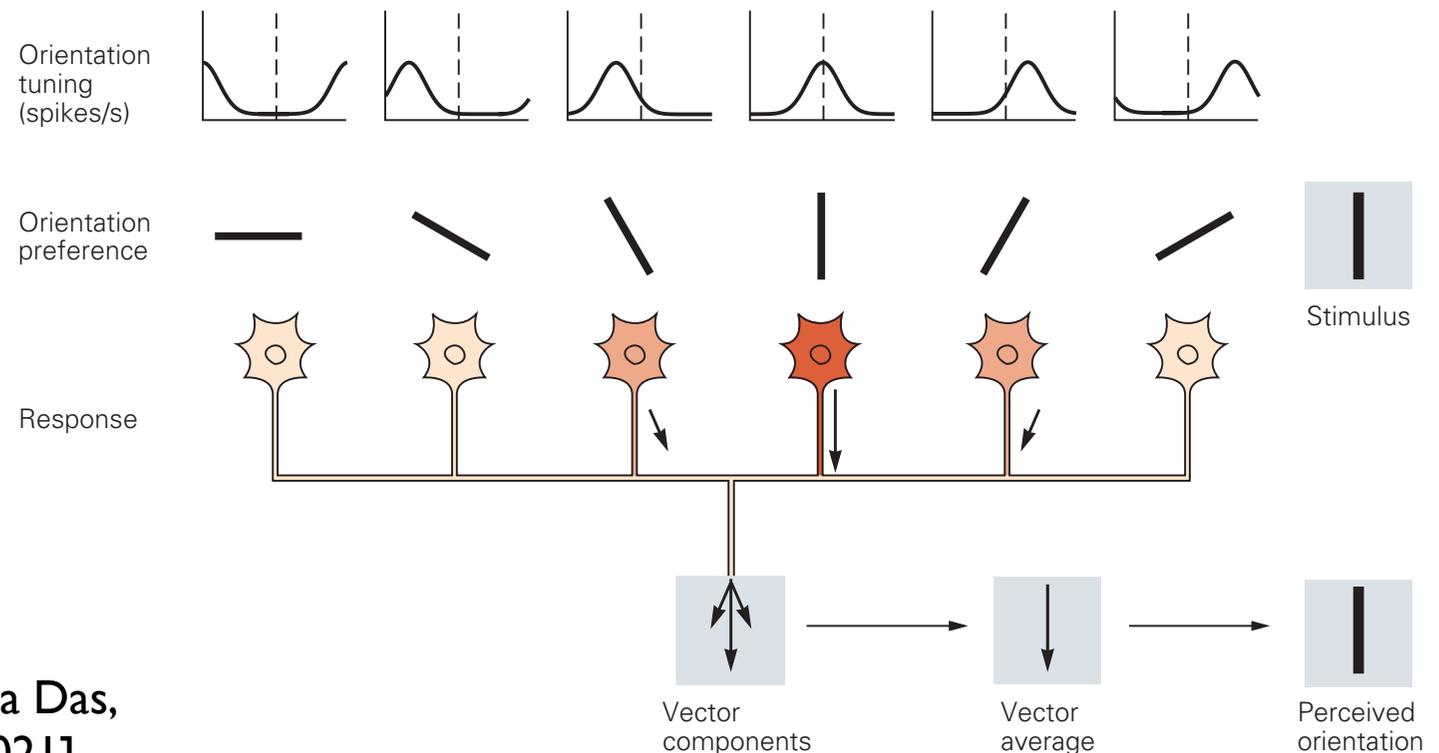


[from: Tresilian, 2012]

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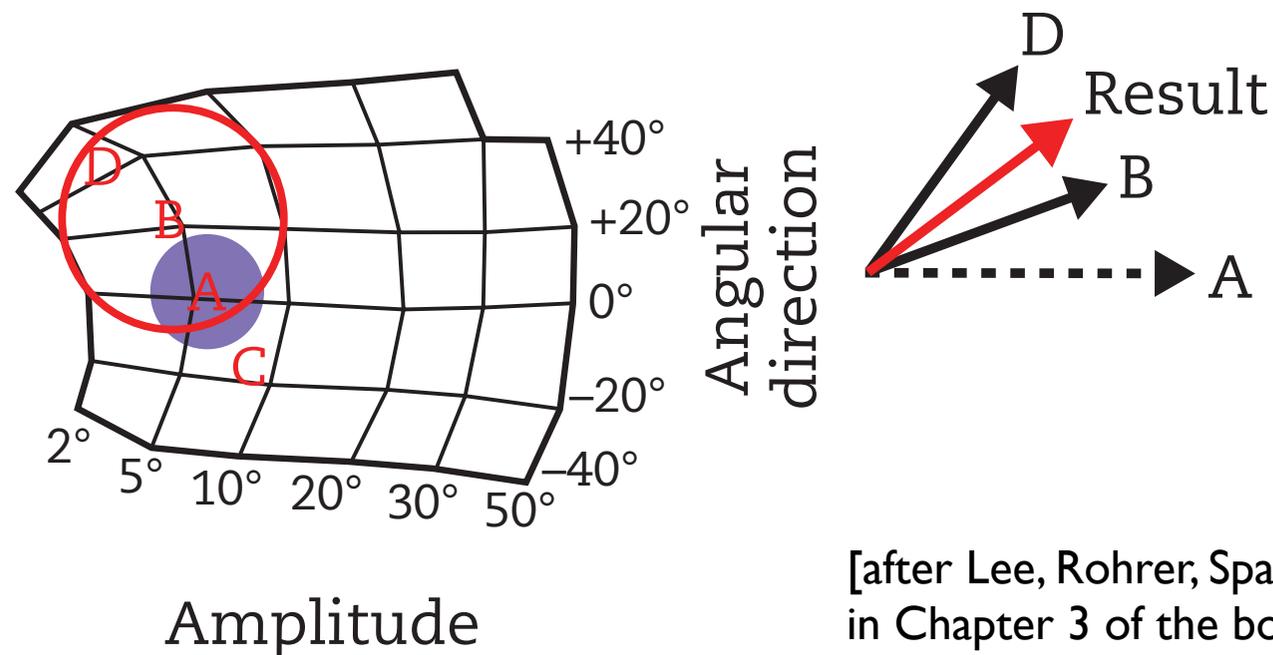
# Population code

- notion that all activated neurons contribute to feature representation according to their tuning curves



# Experimental evidence for population representations

- Lee, Rohrer, Sparks: use the topographic map of saccadic endpoint in superior colliculus
- to reversibly deactivate portions of the population: observe predicted deviations of saccadic endpoints



[after Lee, Rohrer, Sparks: Nature (1988)  
in Chapter 3 of the book]

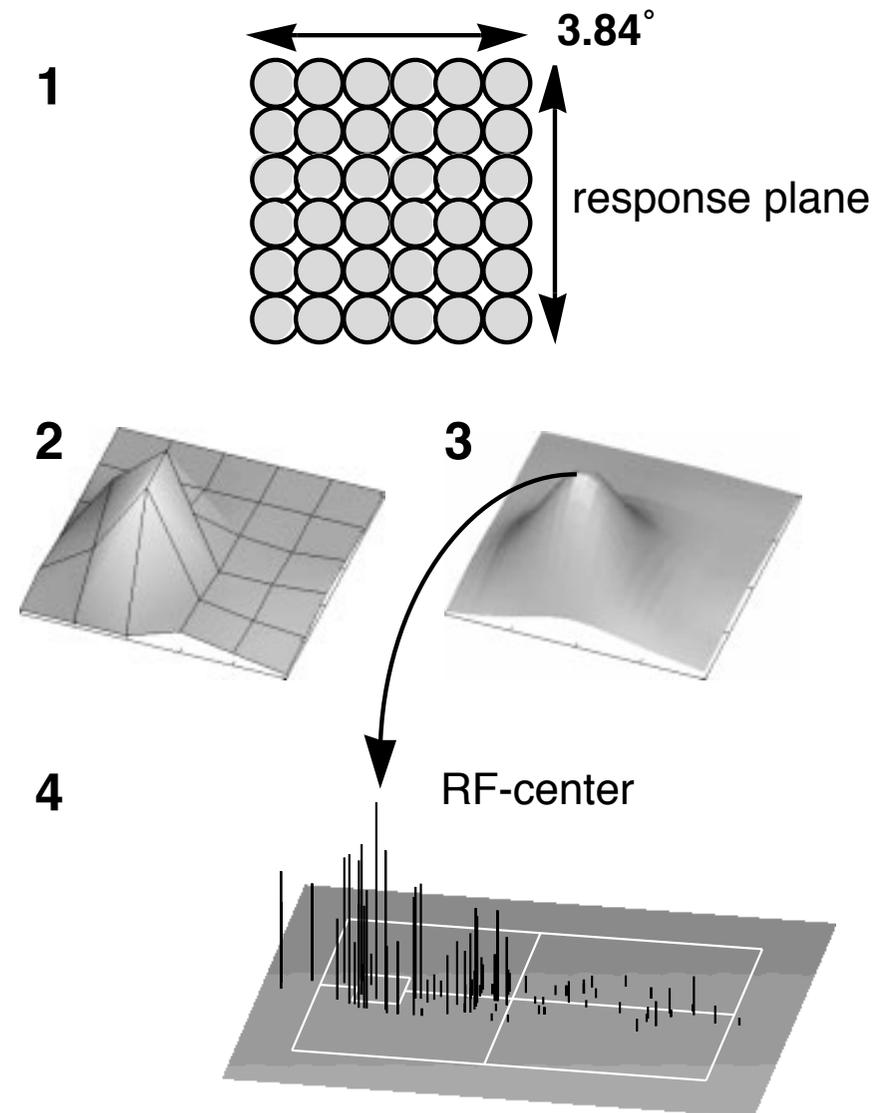
# Population representation in the visual system

- Example 1: Jancke et al: A17 in the cat, population representation of retinal location

Jancke, Erlhagen, Dinse, Akhavan, Giese, Steinhage,  
Schöner JNeurosci 19:9016 (99)

# Distribution of Population Activation (DPA)

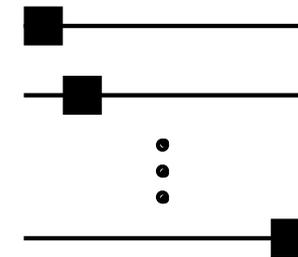
- determine tuning to retinal location for each cell
- superpose tuning curves weighted by current firing rate: **distribution of population activation DPA** representing retinal location



# Neural grounding of DFT: sensory

■ DPA of stimuli presented to all neurons

elementary stimuli



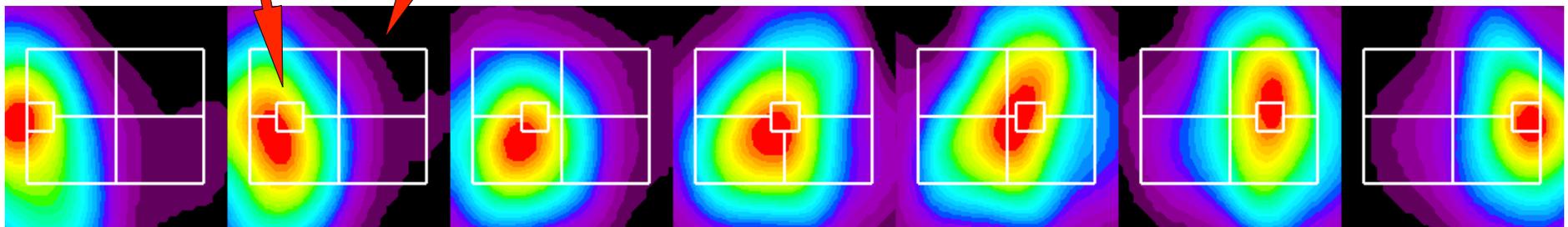
2.8°

nasal

temporal

current stimulus:  
square of light

range of retinal field  
sampled by neurons

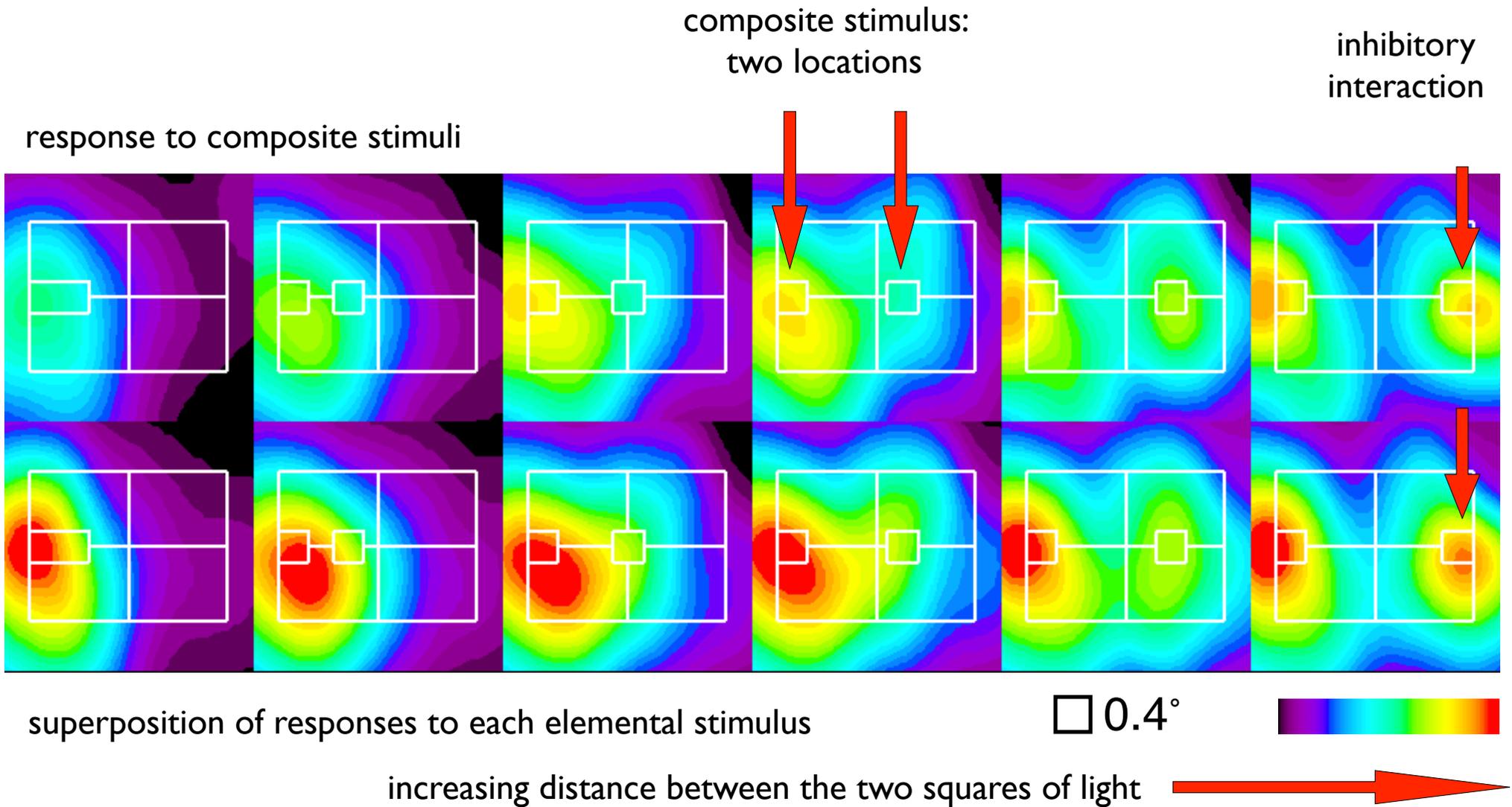


□ 0.4°

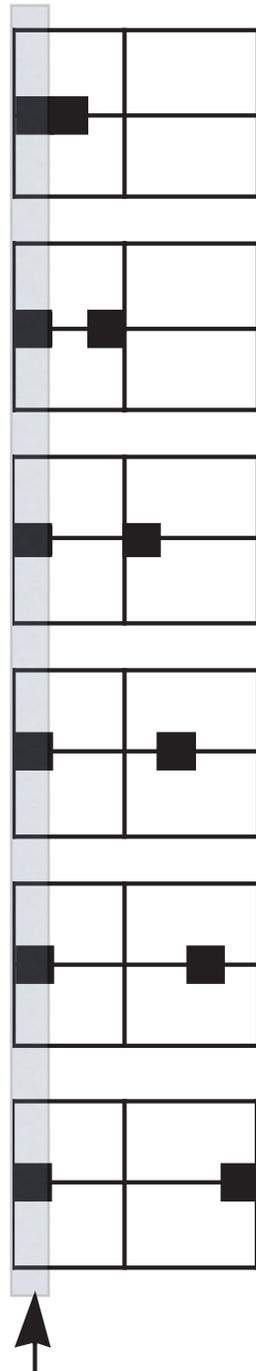


# Neural grounding of DFT: sensory

■ => observe interaction in DPA

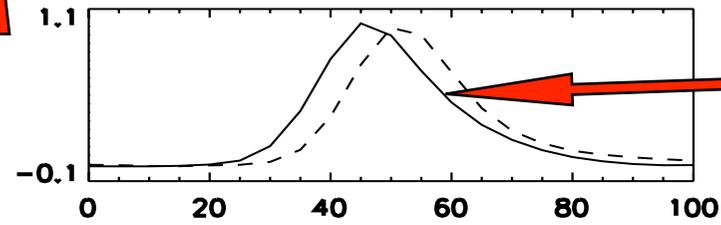


# interaction

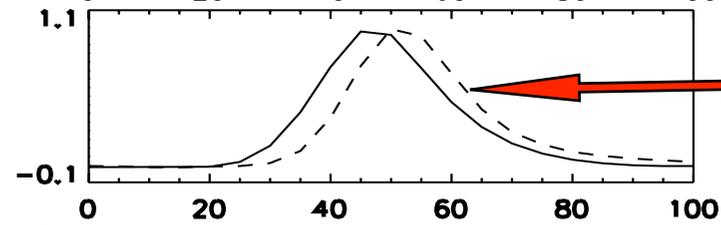


activation level in DPA

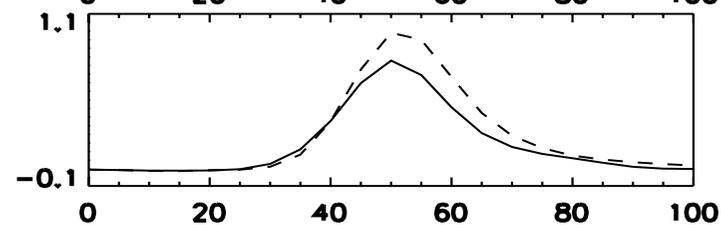
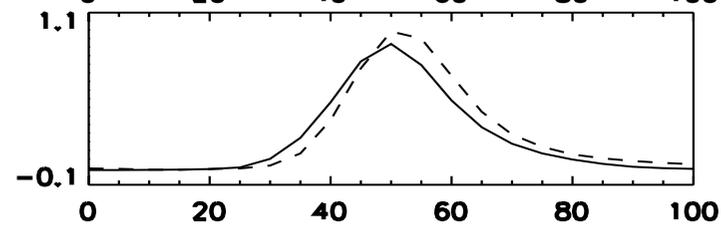
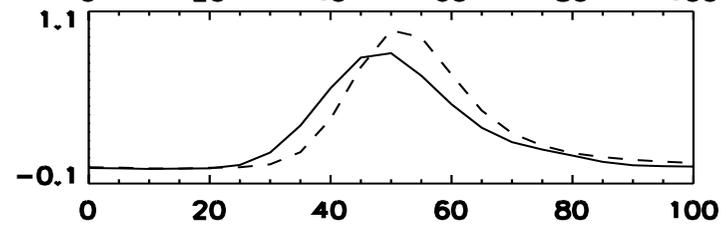
at location of left component stimulus



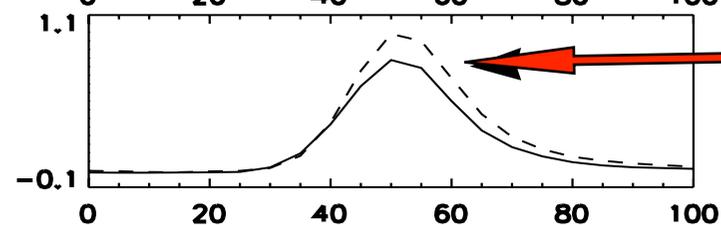
response to composite stimuli



superposition of responses to each elemental stimulus



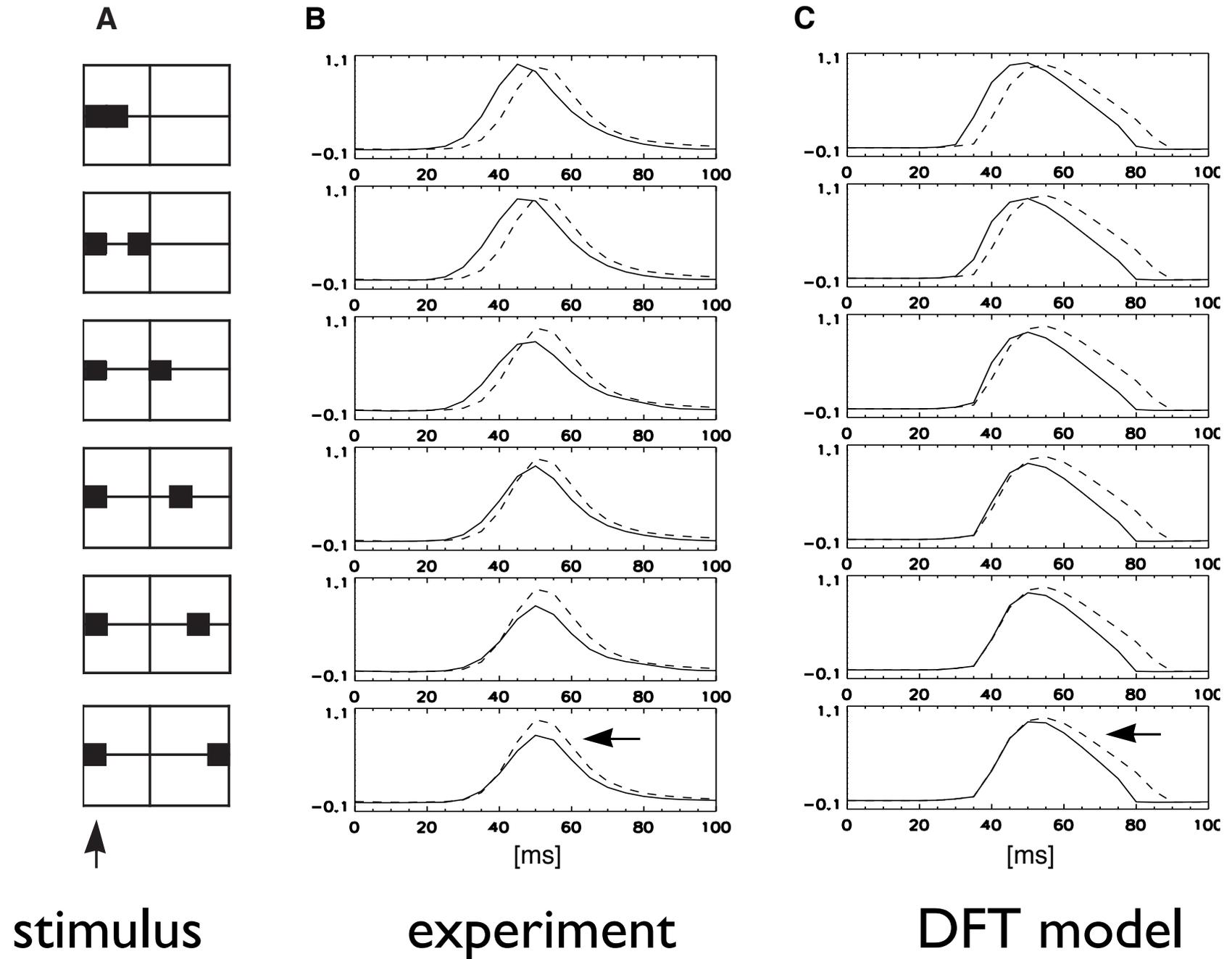
evidence for inhibitory interaction



time

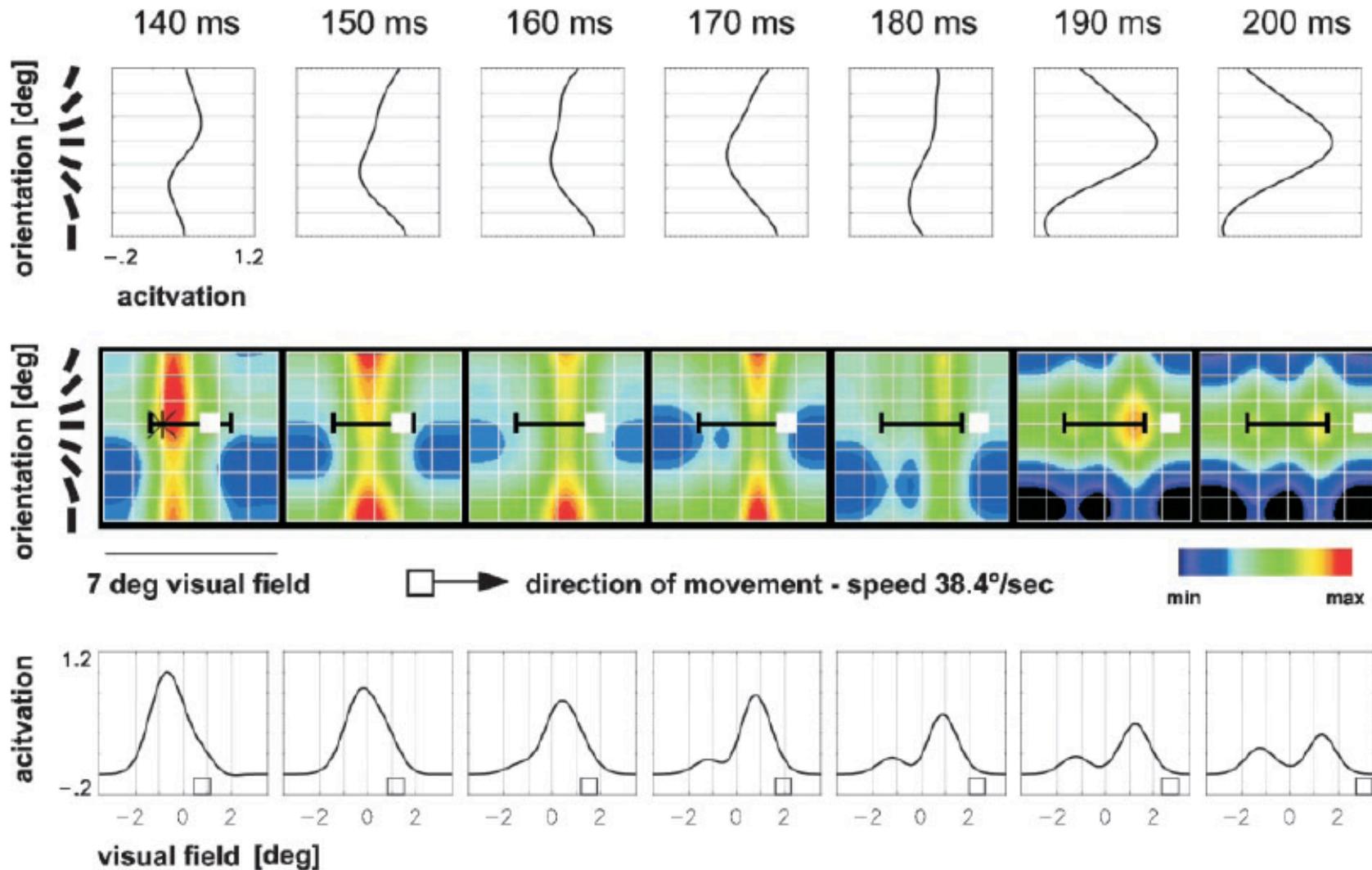
[ms]

# model by dynamic field:



# Neural grounding of DFT: sensory

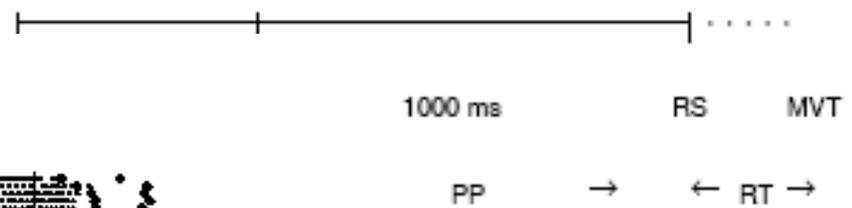
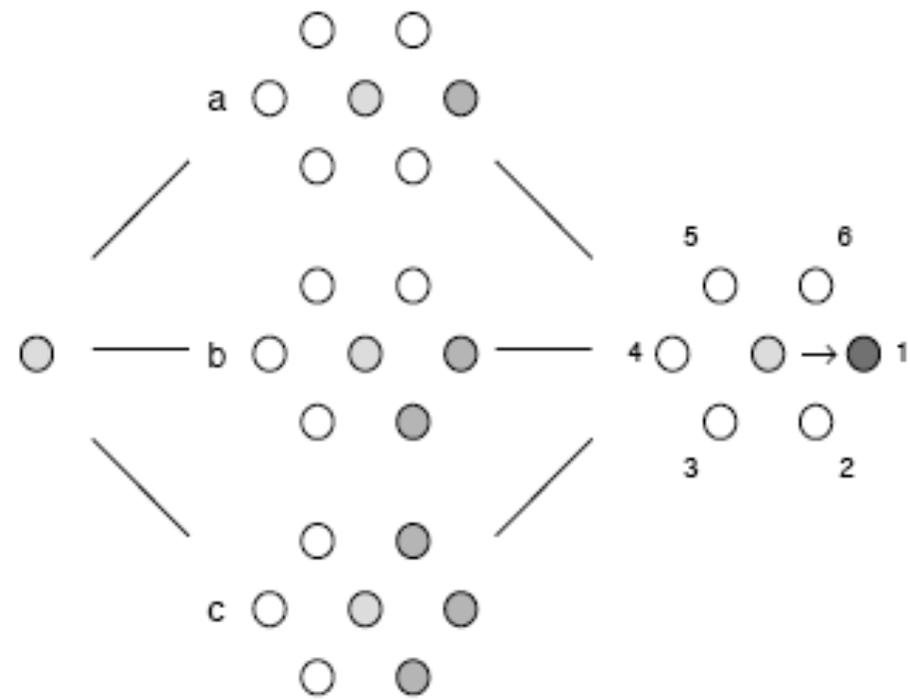
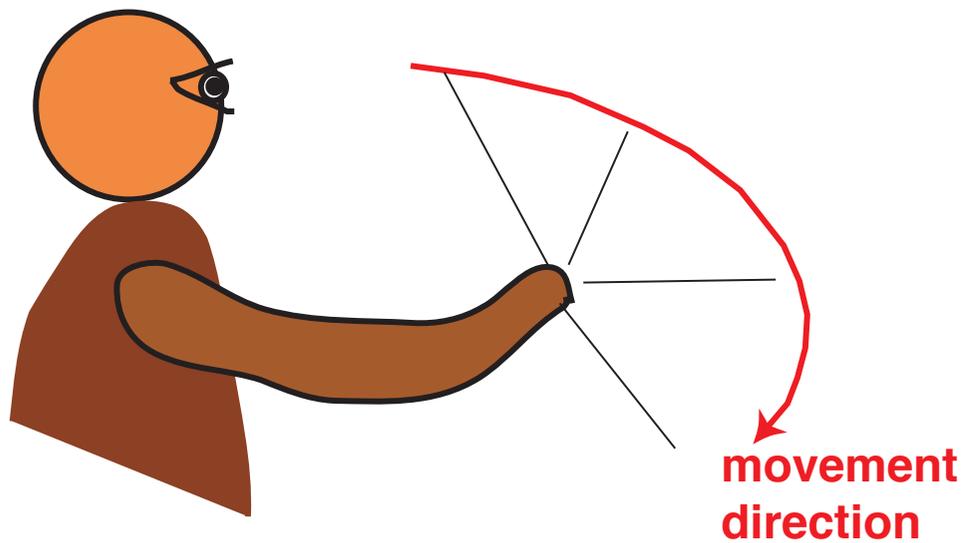
- DPA of orientation and (ID) retinal location



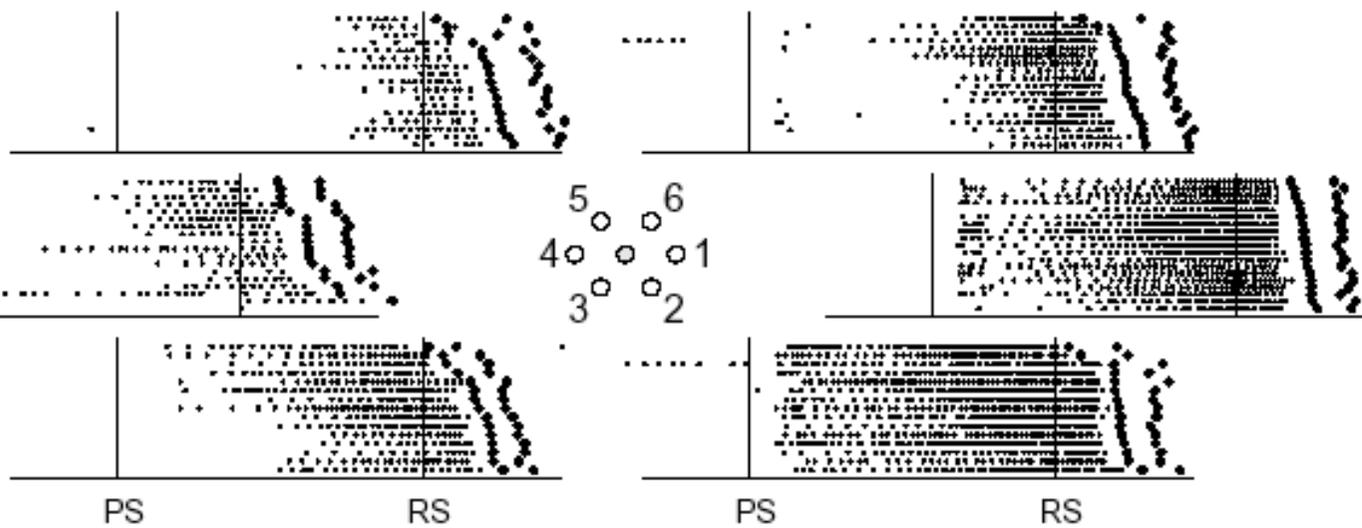
[Jancke, JNeurosci (2000)]

# Population representation motor system

- motor and pre-motor cortex (macaque)
- in behaving animal

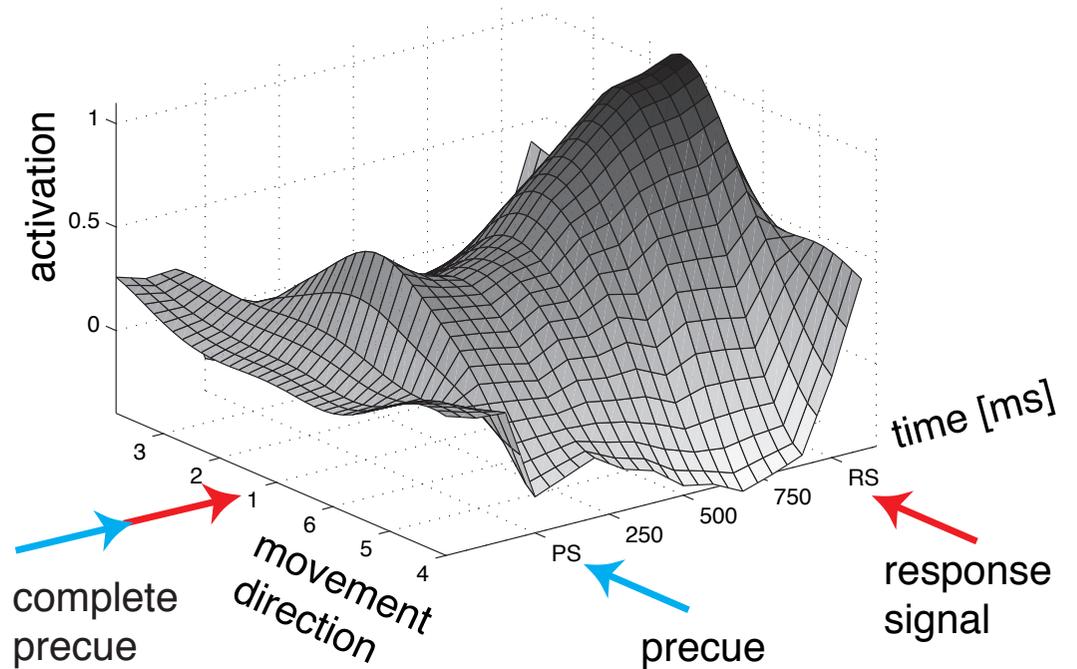
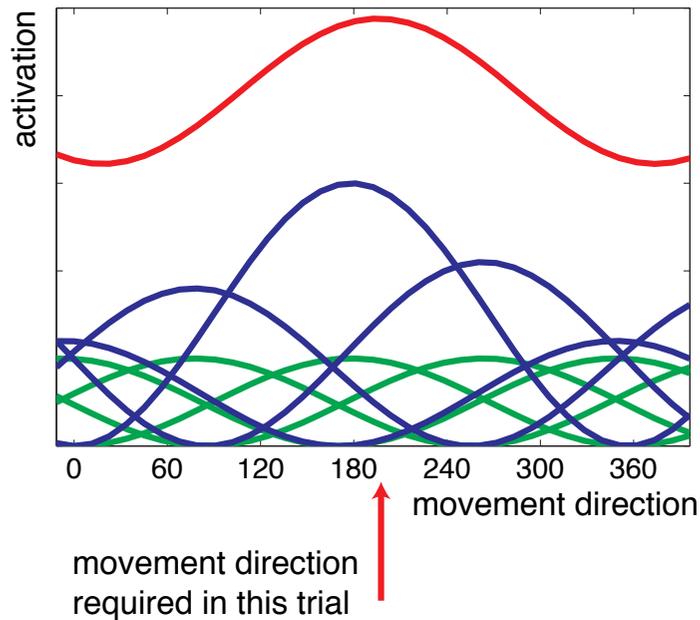


Complete Information



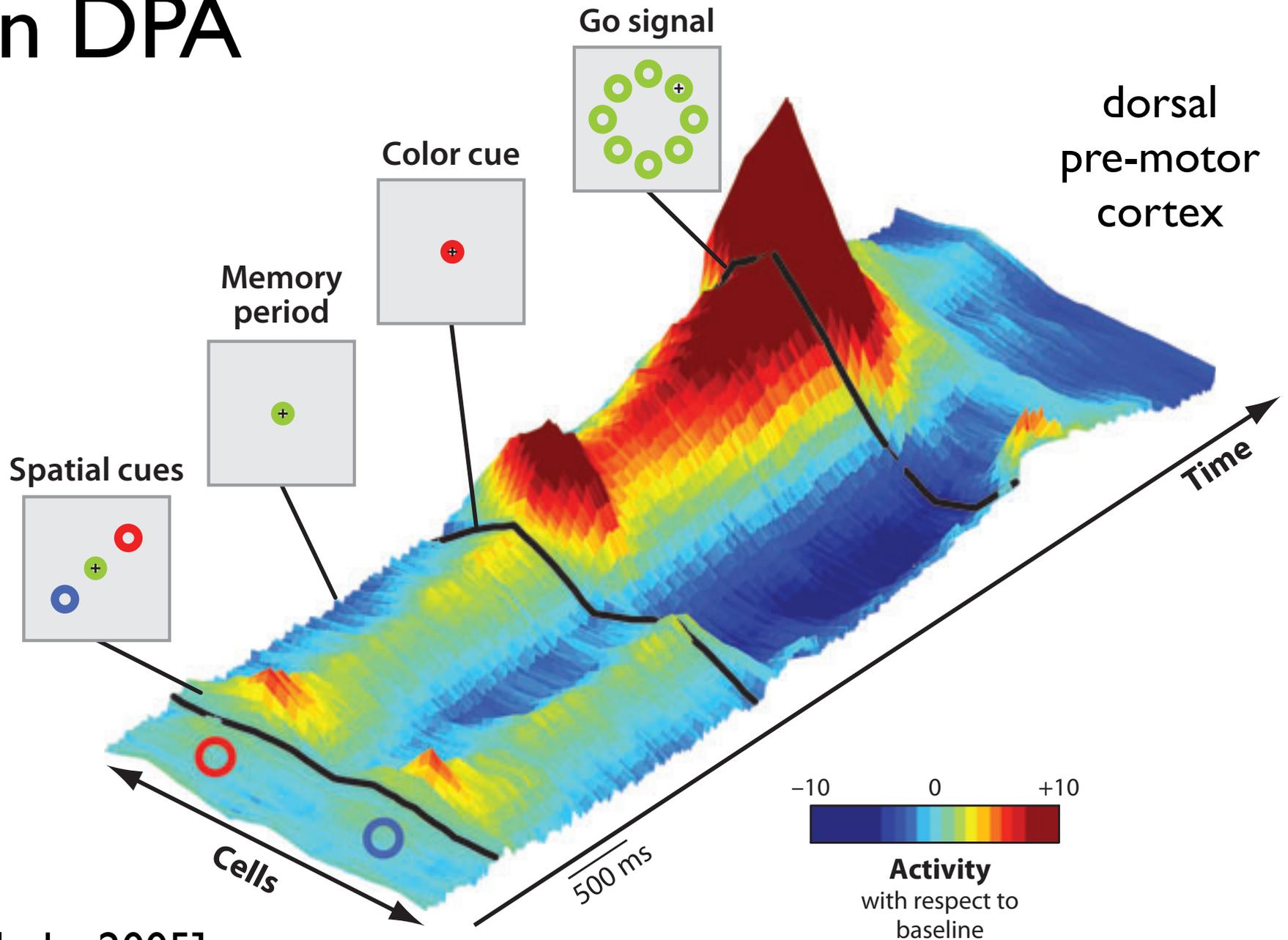
# Distribution of Population Activation (DPA)

Distribution of population activation =  $\sum_{\text{neurons}} \text{tuning curve} * \text{current firing rate}$



[Bastian, Riehle, Schöner, 2003]

# Decision making in DPA



[Cisek, Kalaska 2005]

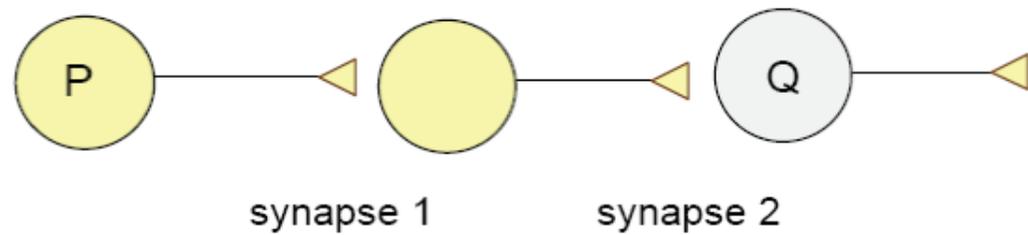
# Distributions of Population Activation (DPA)

- neurons are not localized within DPA!
- cortical neurons really are sensitive to many dimensions
  - motor: arm configuration, force direction
  - visual: many feature dimensions such as spatial frequency, orientation, direction...
- => DPA is a projection from that high-dimensional space onto a single dimension

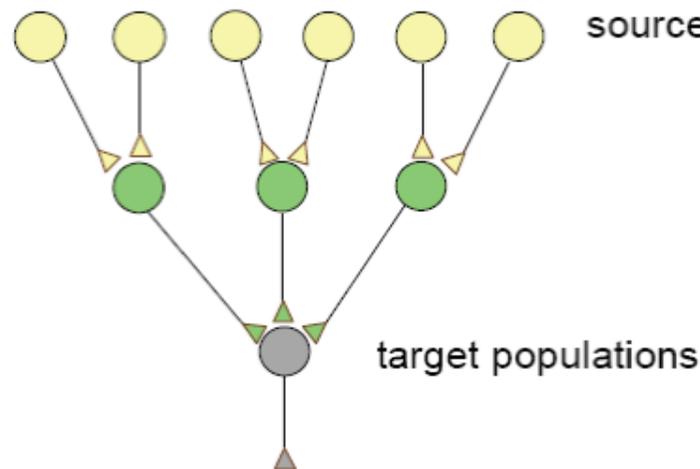
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# Forward connectivity

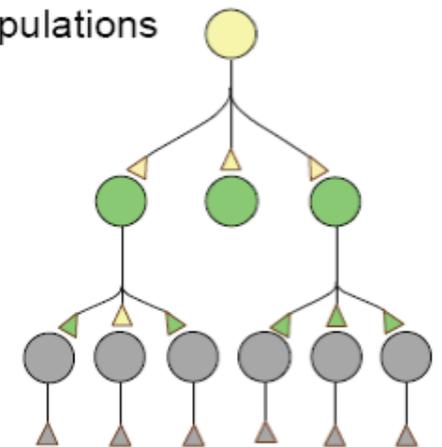
(A) Disynaptic connectivity between P and Q



(B) Convergent pathway

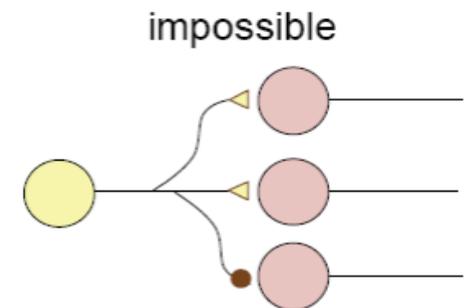
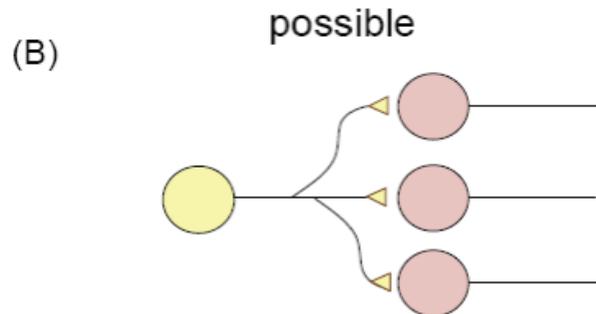


(C) Divergent pathway



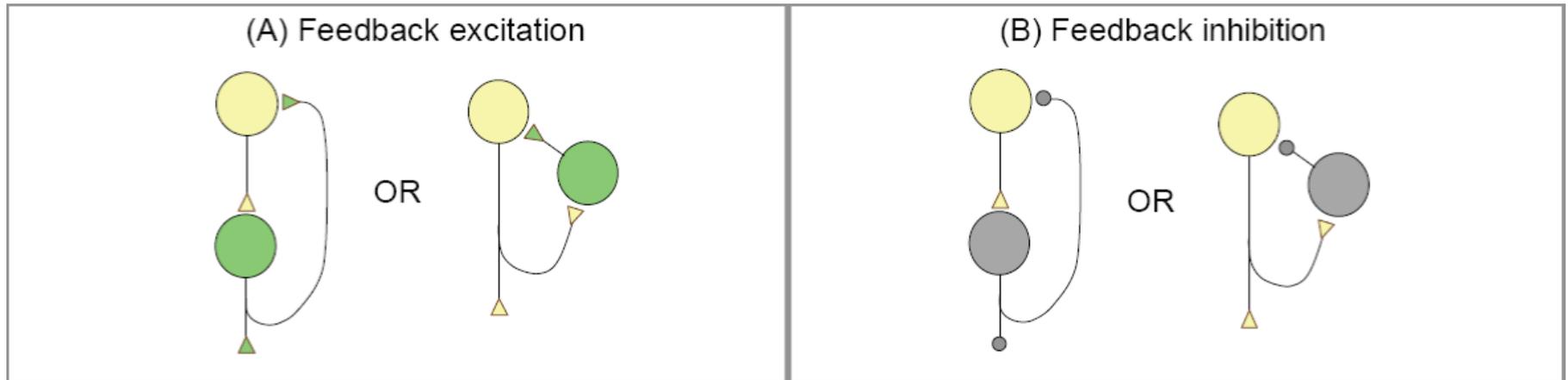
# Dale's law

- all synaptic connections coming from a given neuron are of the same type



# Patterns of connectivity

## ■ recurrent connectivity

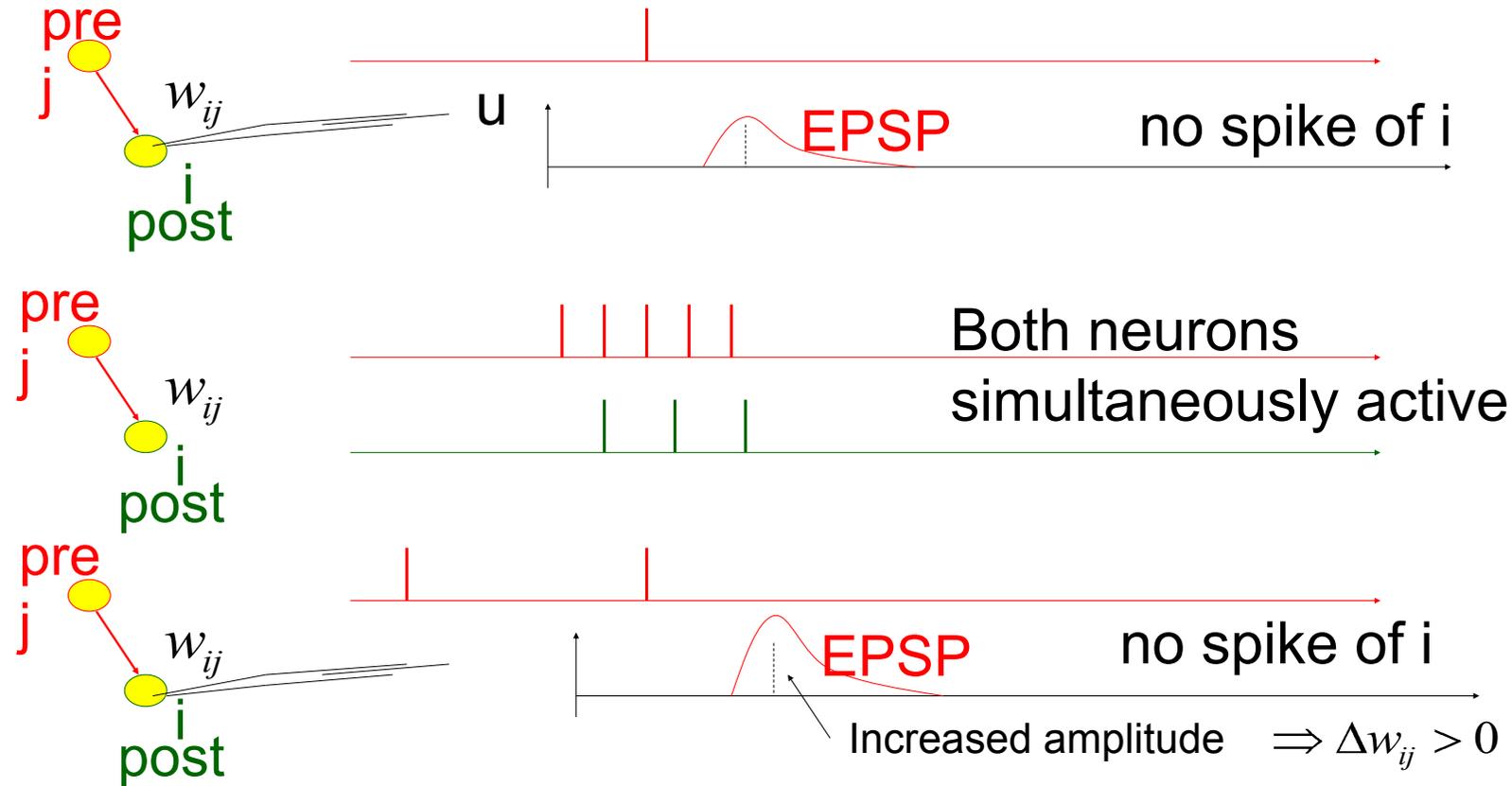


- Neurophysics
- Sensors, actuators, rate code
- Receptive fields, tuning curves
- Maps
- Distributions of population representation
- Patterns of connectivity
- Synaptic dynamics

# Learning by synaptic plasticity

- synaptic strengths change as a function of pre/post synaptic neural state

# Hebbian Learning in experiments (schematic)



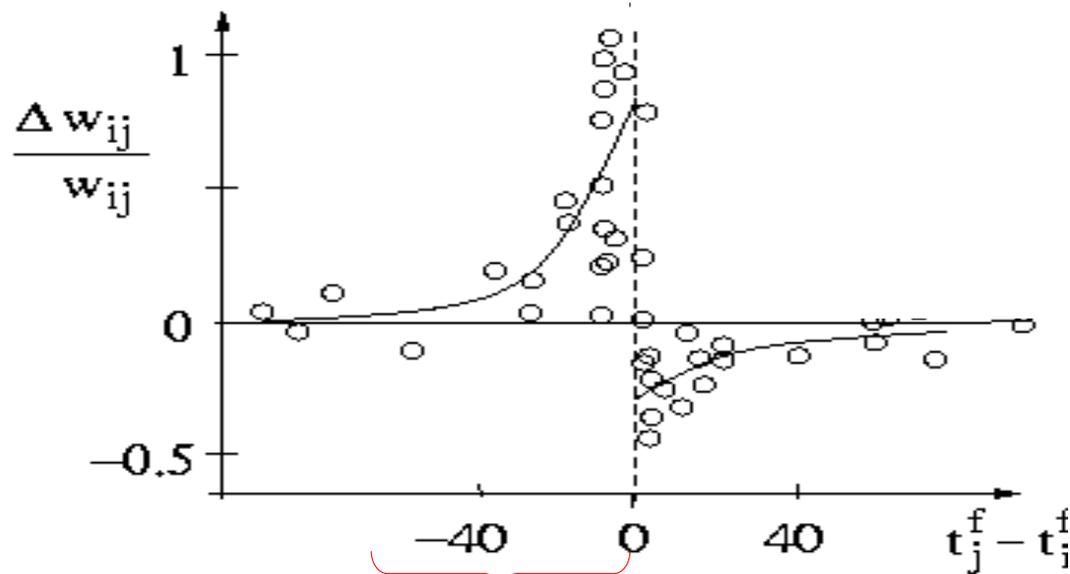
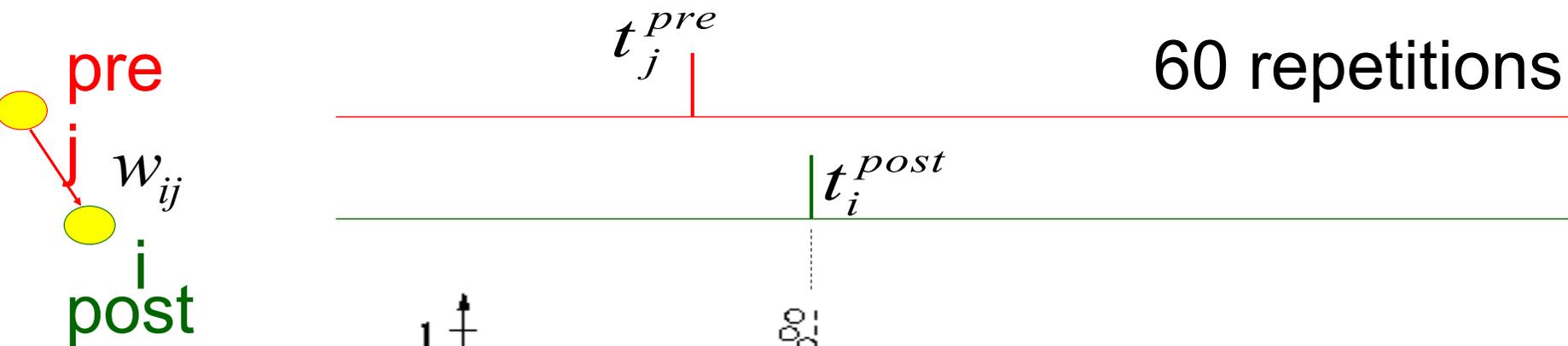
# Learning by plasticity

- spike-time dependent plasticity

- strengthening of synapses in which pre-synaptic spike precedes post-synaptic spike

- weakening synapses when the temporal order is the reverse...

# Spike-time dependent plasticity



Pre  
before post

[Gerstner et al, 2014]

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