# Dynamic Field Theory: Neural Grounding

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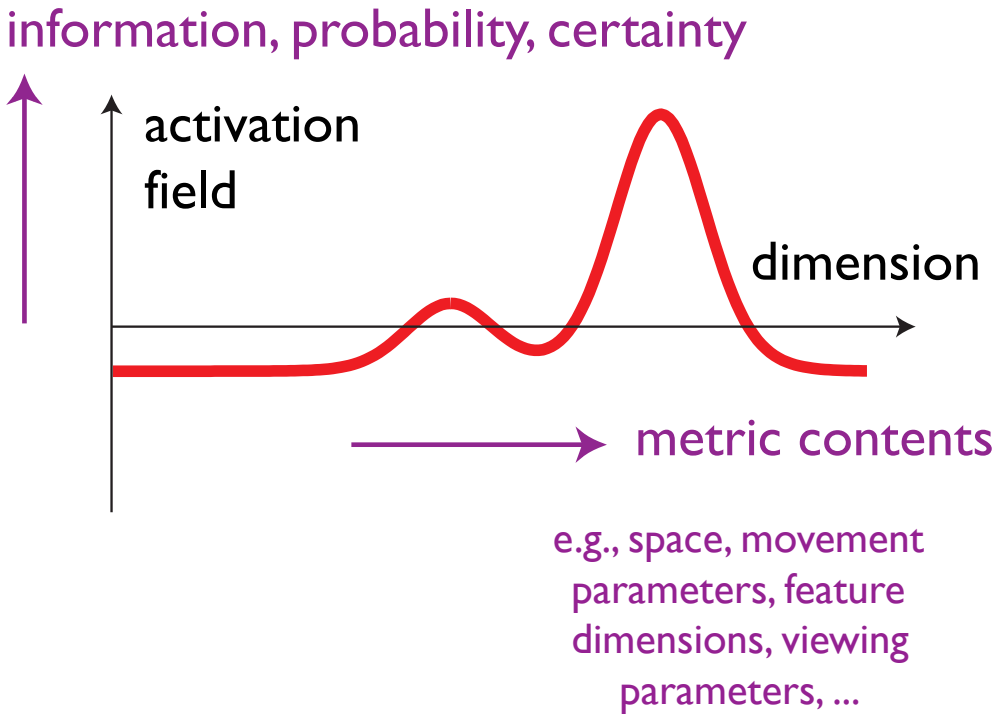
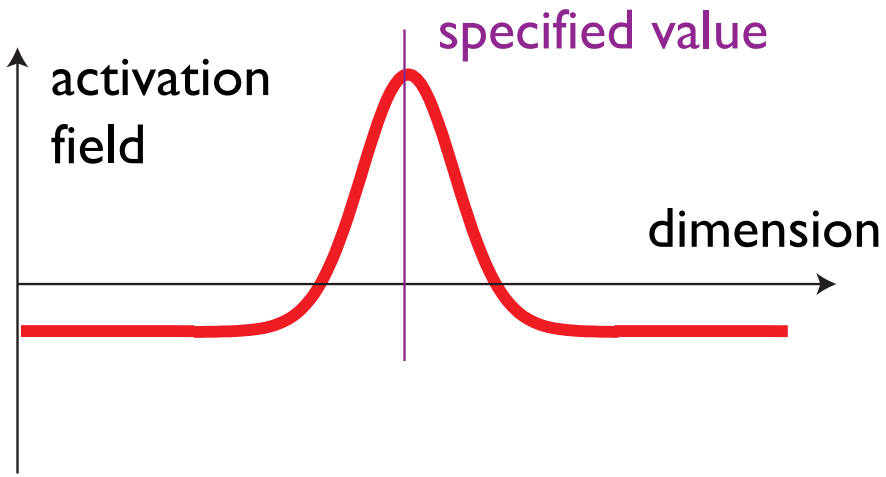
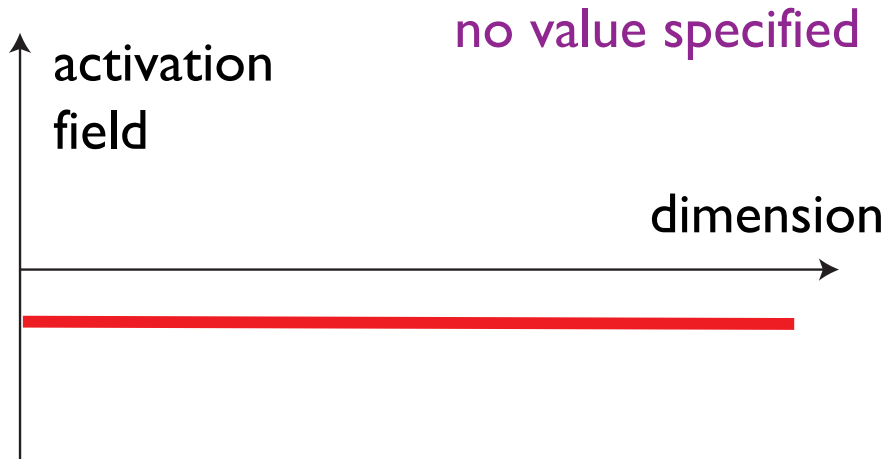
# … last week: activation fields

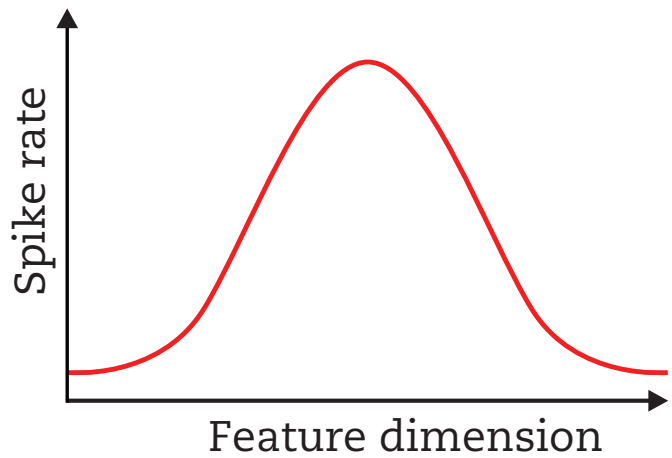


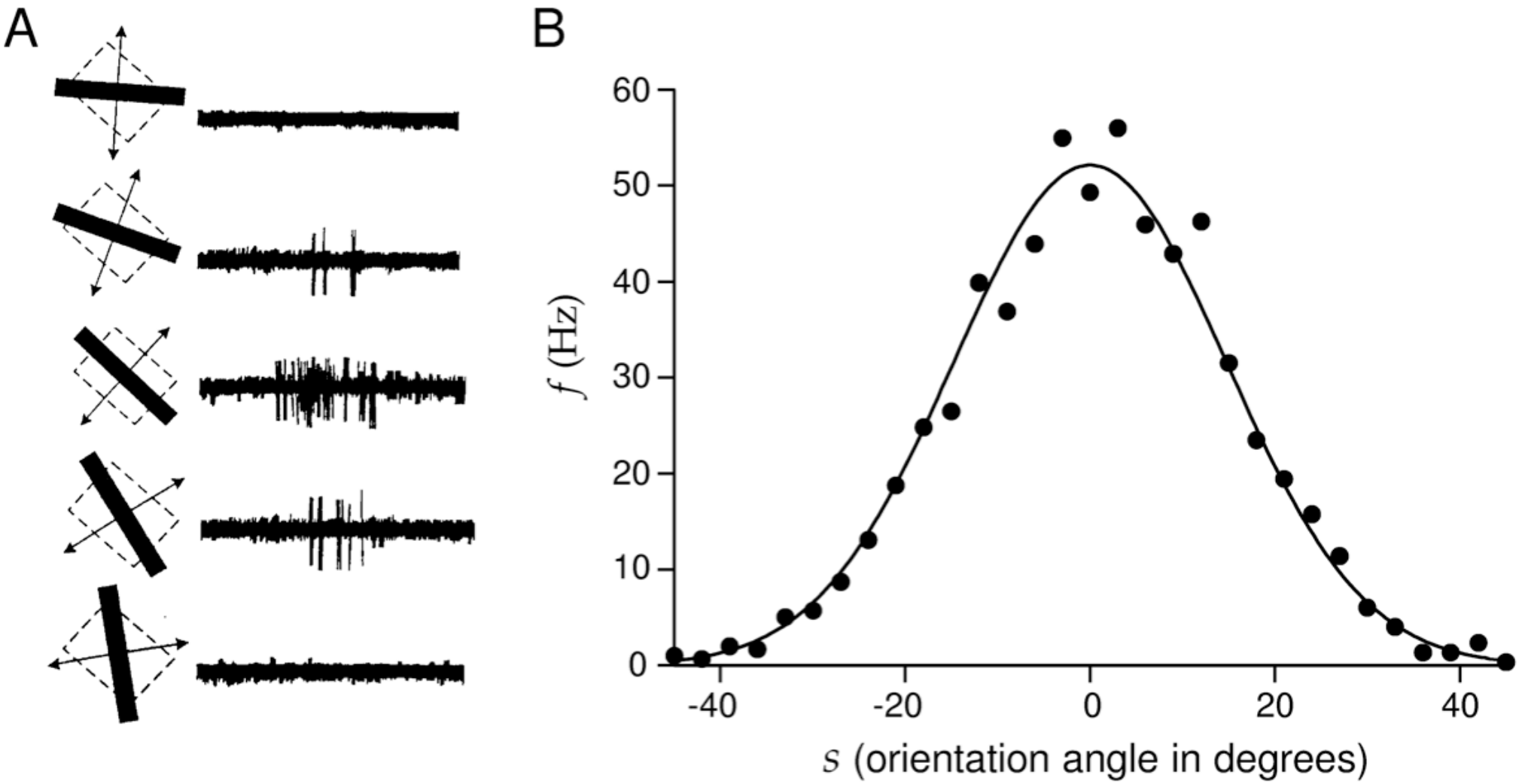



# On the link between DFT and neurophysiology

- What do neurons represent?
  - notion of a tuning curve that links something outside the nervous system to the state of a neuron (e.g. through firing rate)
  - based on the forward picture in which
    - the connectivity from the sensory surface
    - or the connectivity from the neuron to the motor surface
  - determine the activity of the neuron

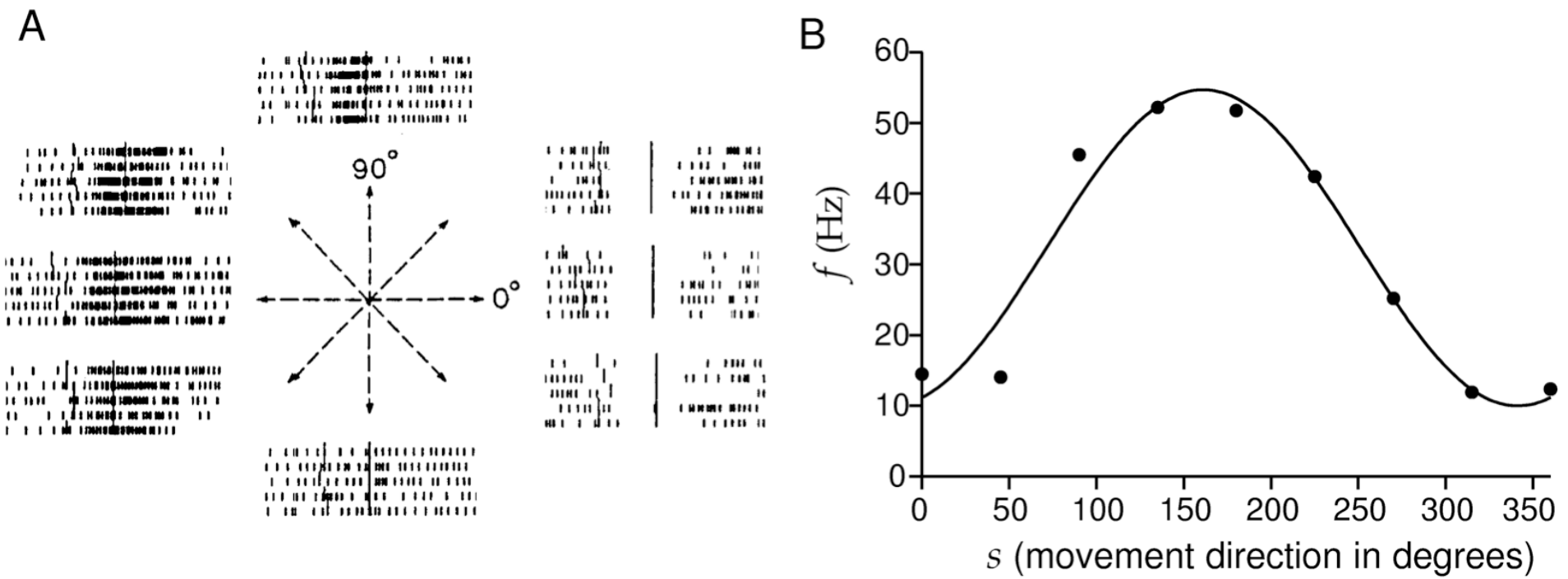


# Example tuning curve in primary visual cortex (monkey)




[Hubel, Wiesel, 1962]

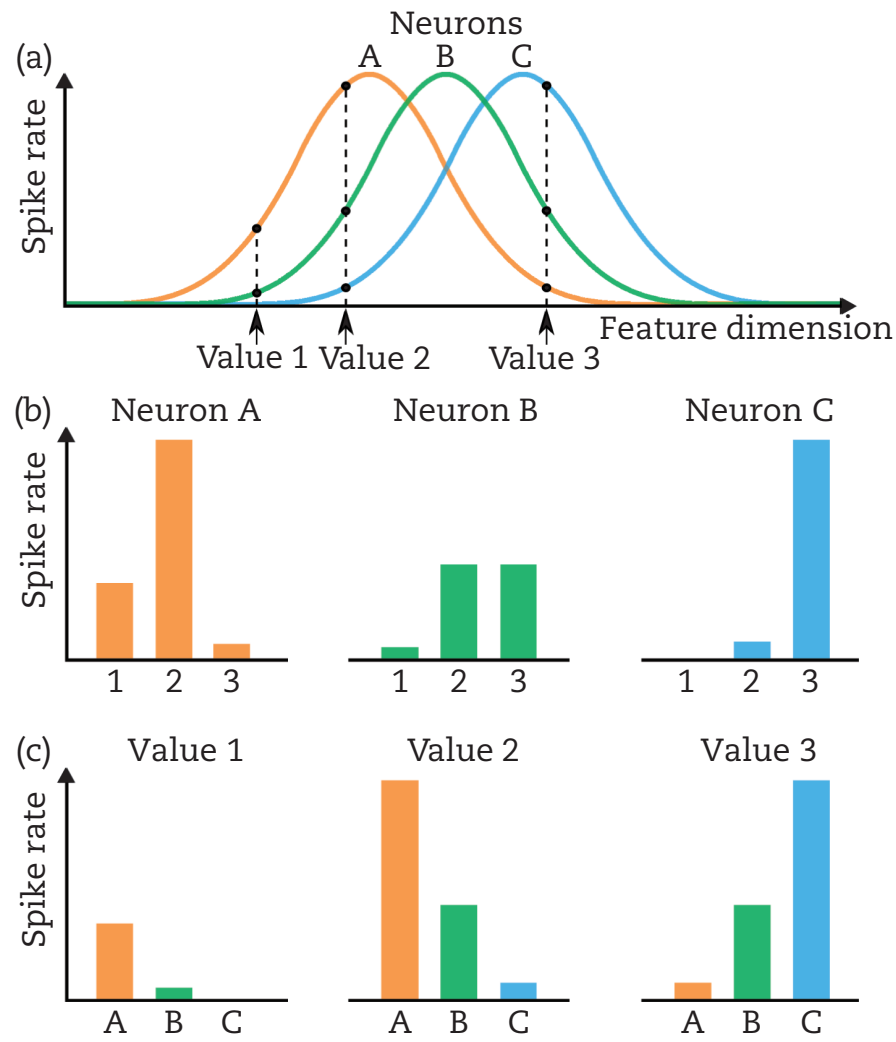
# Example: tuning curve in primary motor cortex (monkey)




[Georgopoulos, Schwartz, Kalaska, 1986]

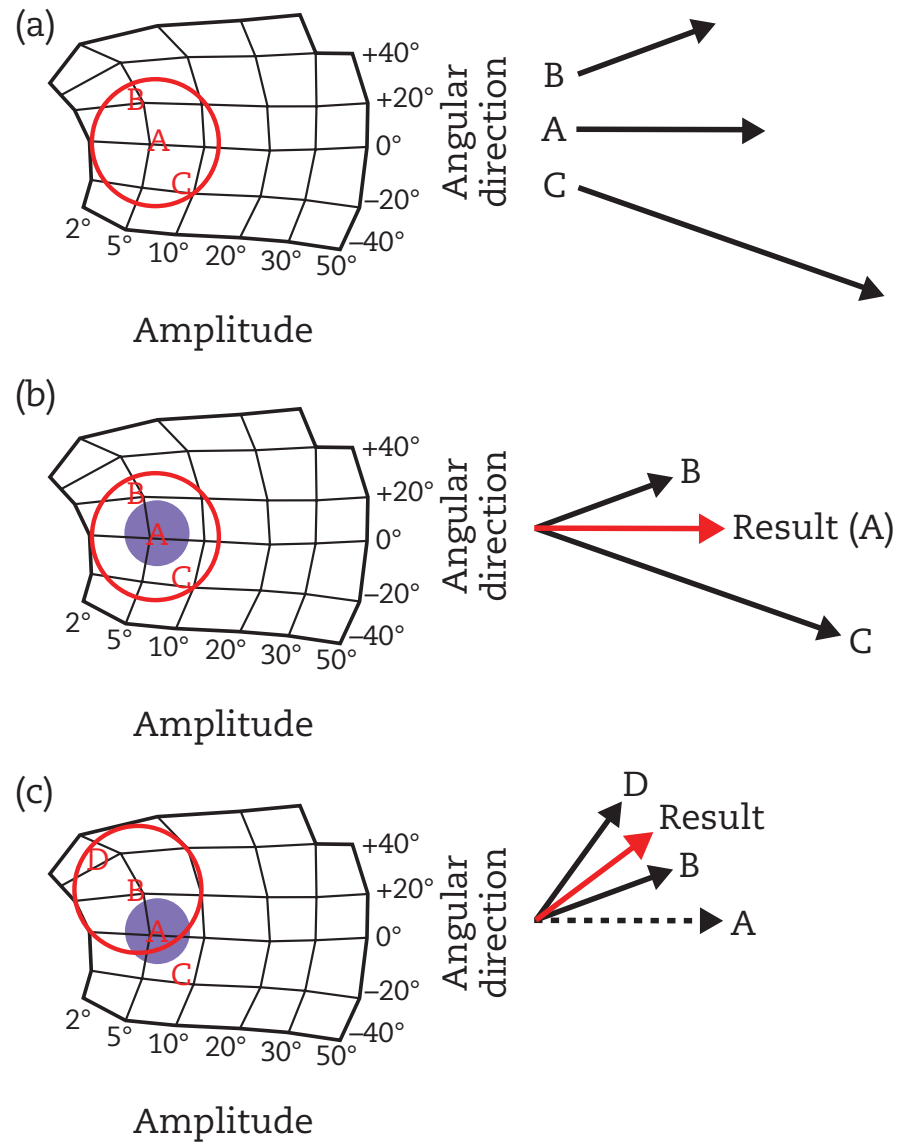
# What do ensembles of neurons represent?

- the pattern of neural activity across multiple neurons represents a feature value much more precisely than individual neurons do



# Do all activated neurons contribute?

- superior colliculus: topographic map of saccadic endpoint
- deactivate portions of the population: observe predicted deviations of saccadic endpoint




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

# Population code

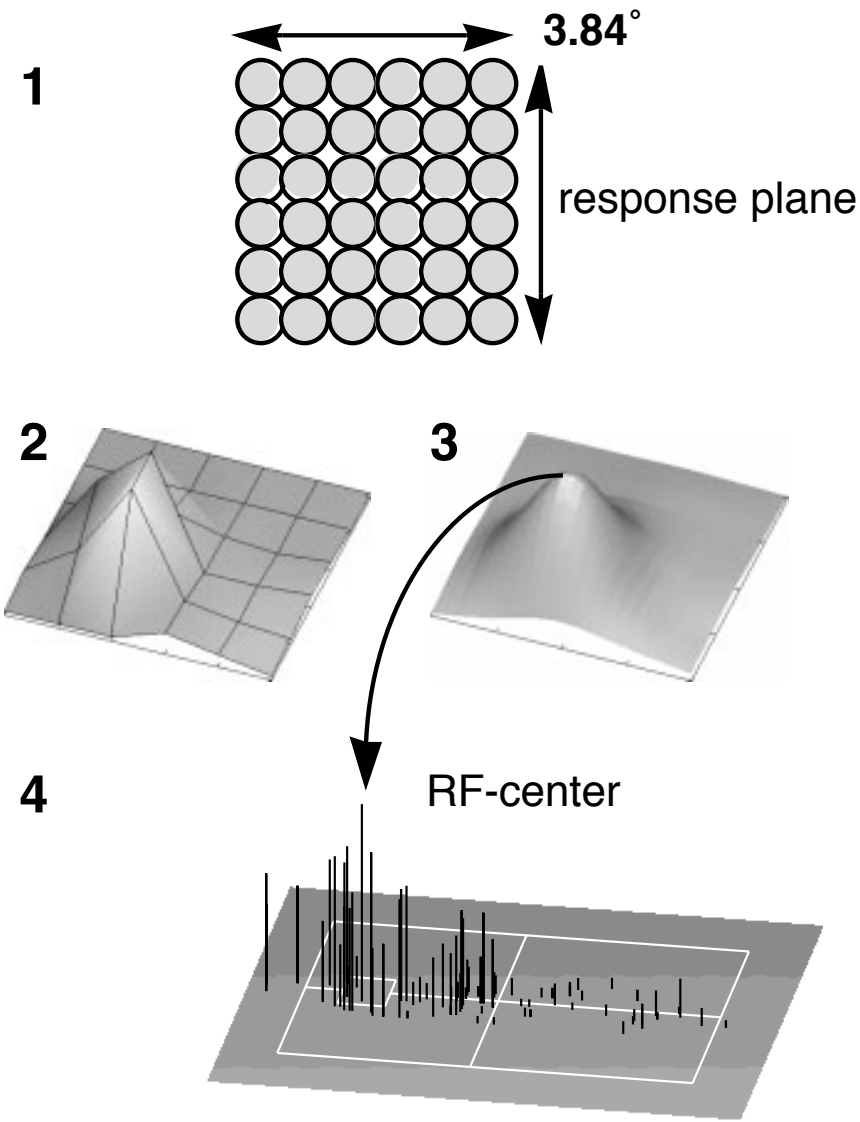
- similar work in MT
  - Purushothaman, G., & Bradley, Da. C. (2005). Neural population code for fine perceptual decisions in area MT. Nature Neuroscience, 8(1), 99–106.
- consensus, that localized populations of neurons best correlated with behavior
  - there are subtle issues of noise and correlation in populations
  - e.g., Cohen, Newsome J Neurosci 2009: about 1000 neurons needed to match behavioral performance
  - review: Shamir, M. (2014). Emerging principles of population coding: In search for the neural code. *Current Opinion in Neurobiology*, *25*, 140–148.

# Neurophysiological grounding of DFT

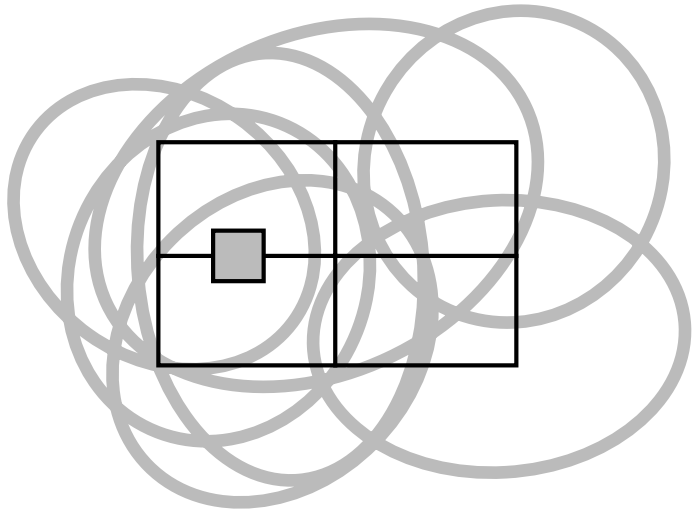
- Example 1: primary visual cortex A17 in the cat, population representation of retinal location

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

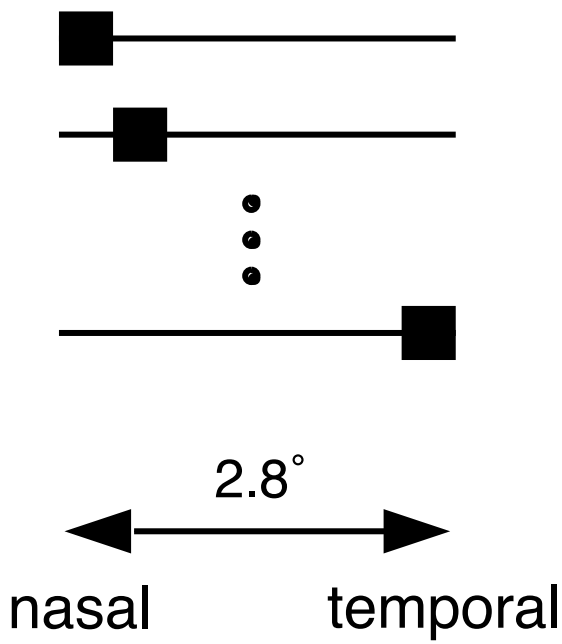
- determine RF profile for each cell
- it's center determines what that neuron codes for
- compute a distribution of population activation by superposing RF profiles weighted with current neural firing rate



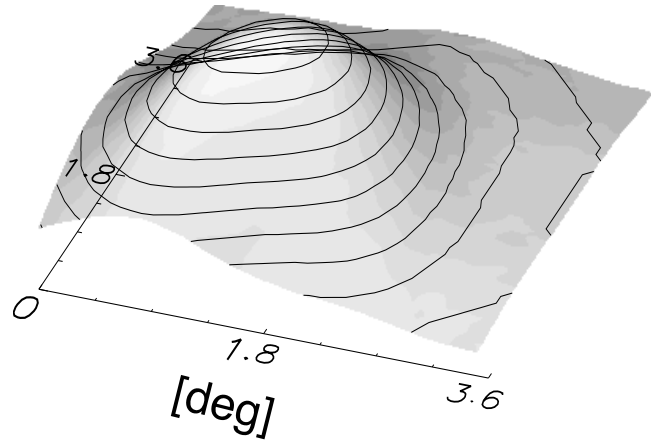
- The current response refers to a stimulus experienced by all neurons
- Reference condition: localized points of light

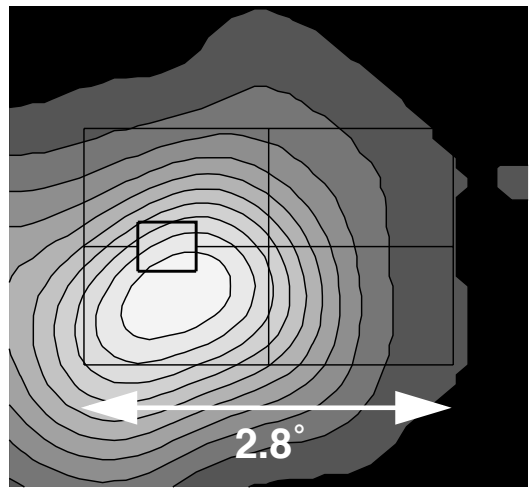


elementary stimuli

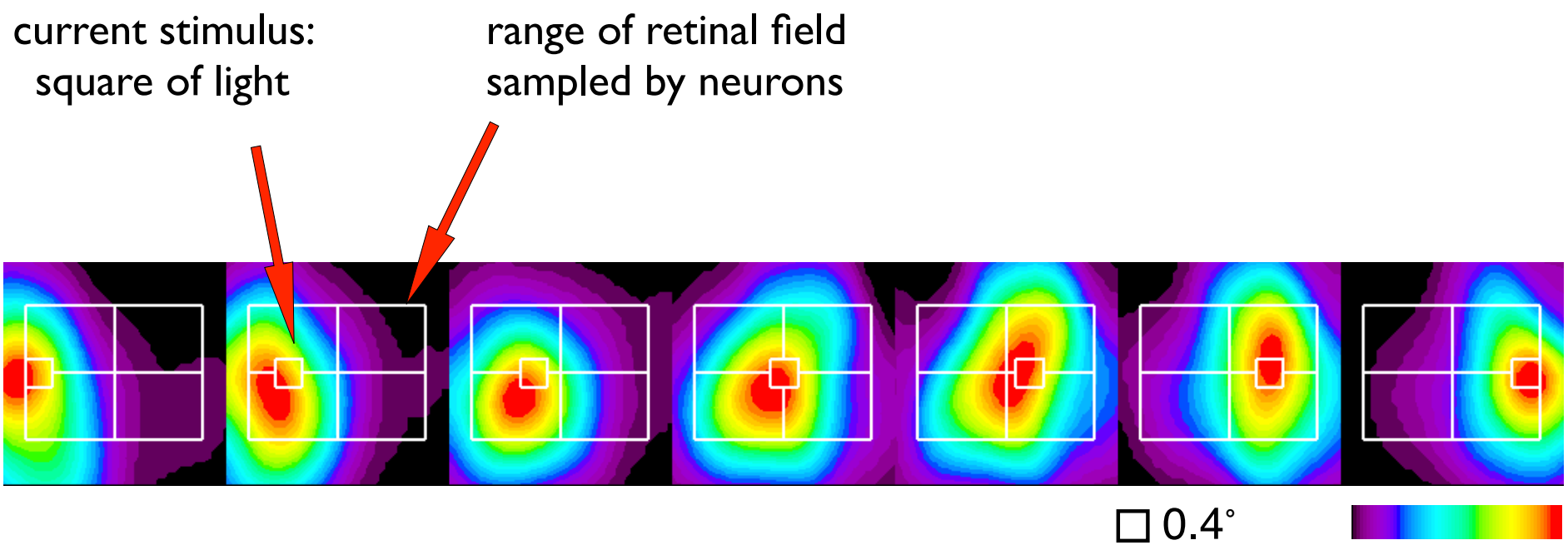


- result: population distribution of activation defined over retinal space = representation of visual location

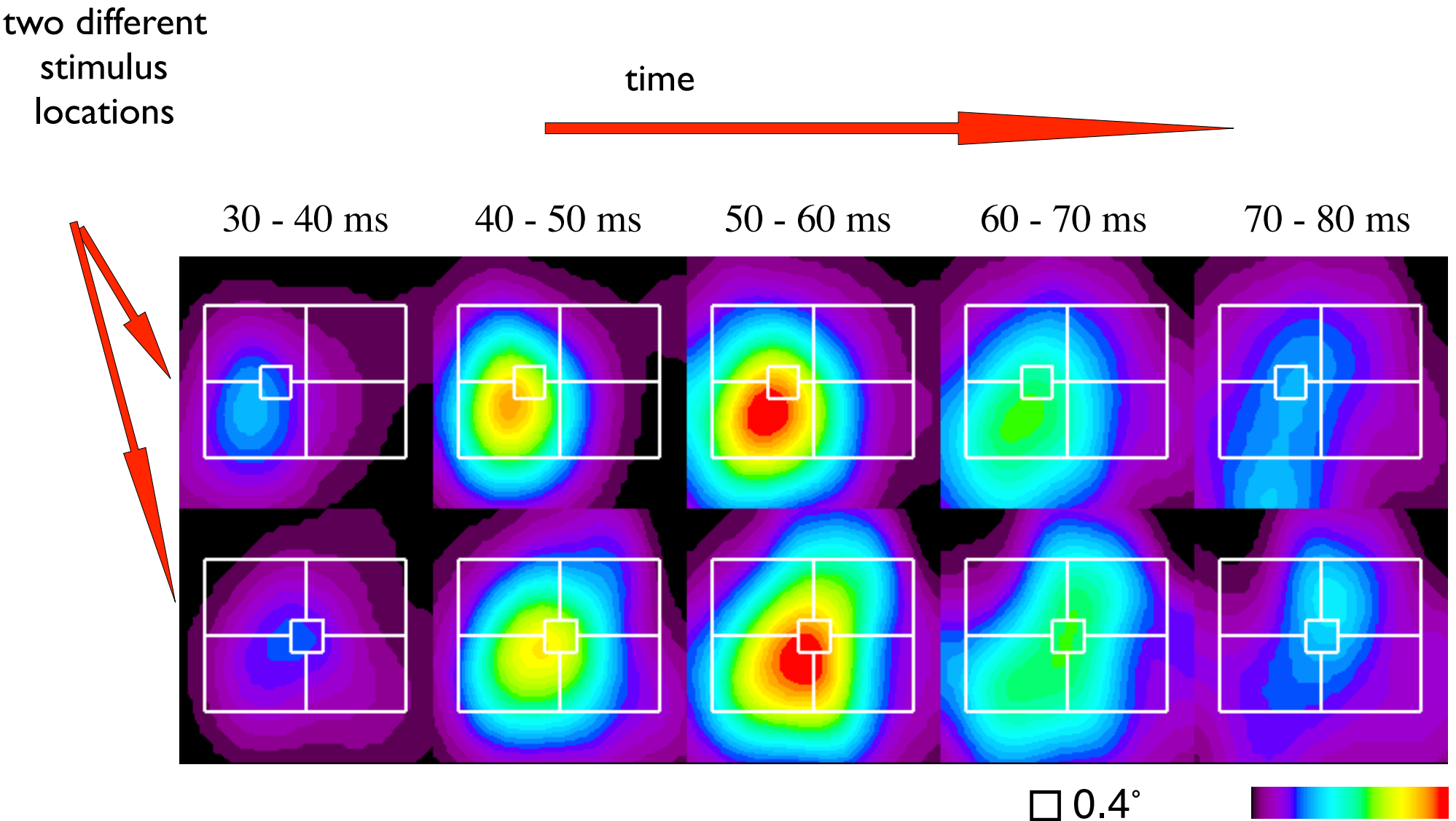


=> does a decent job estimating retinal position



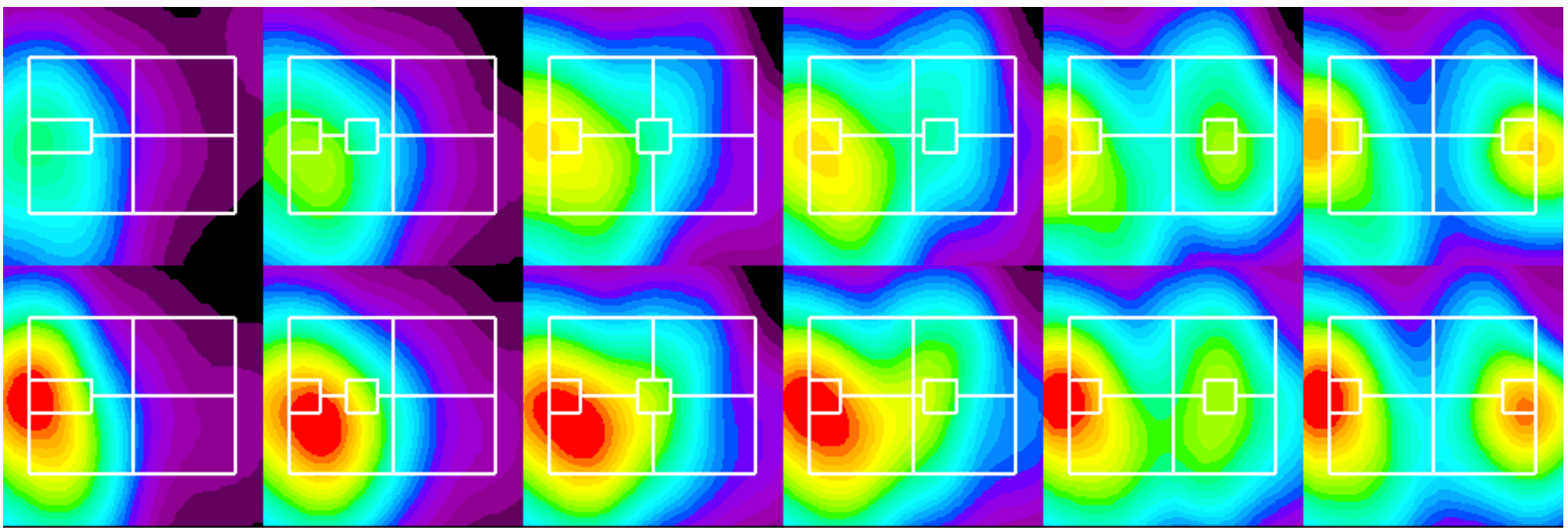
- Extrapolate measurement device to new conditions
- e.g., time resolved



- or when complex stimuli are presented (here: two spots of light)

response to composite stimuli

increasing distance between the two squares of light

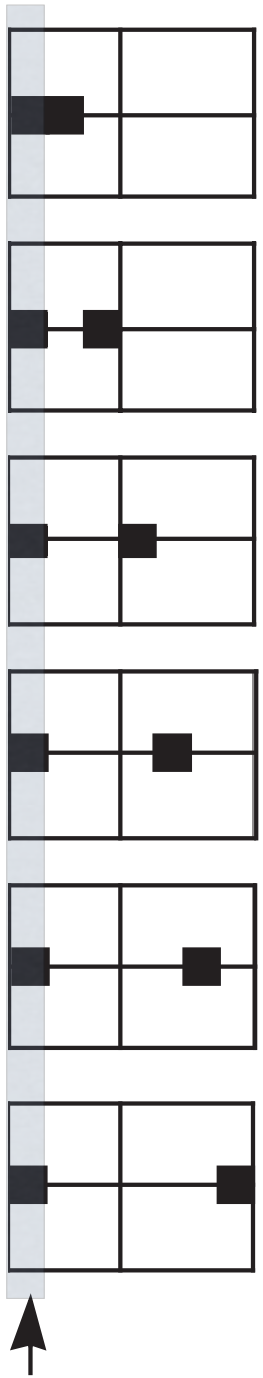


superposition of responses to each elemental stimulus

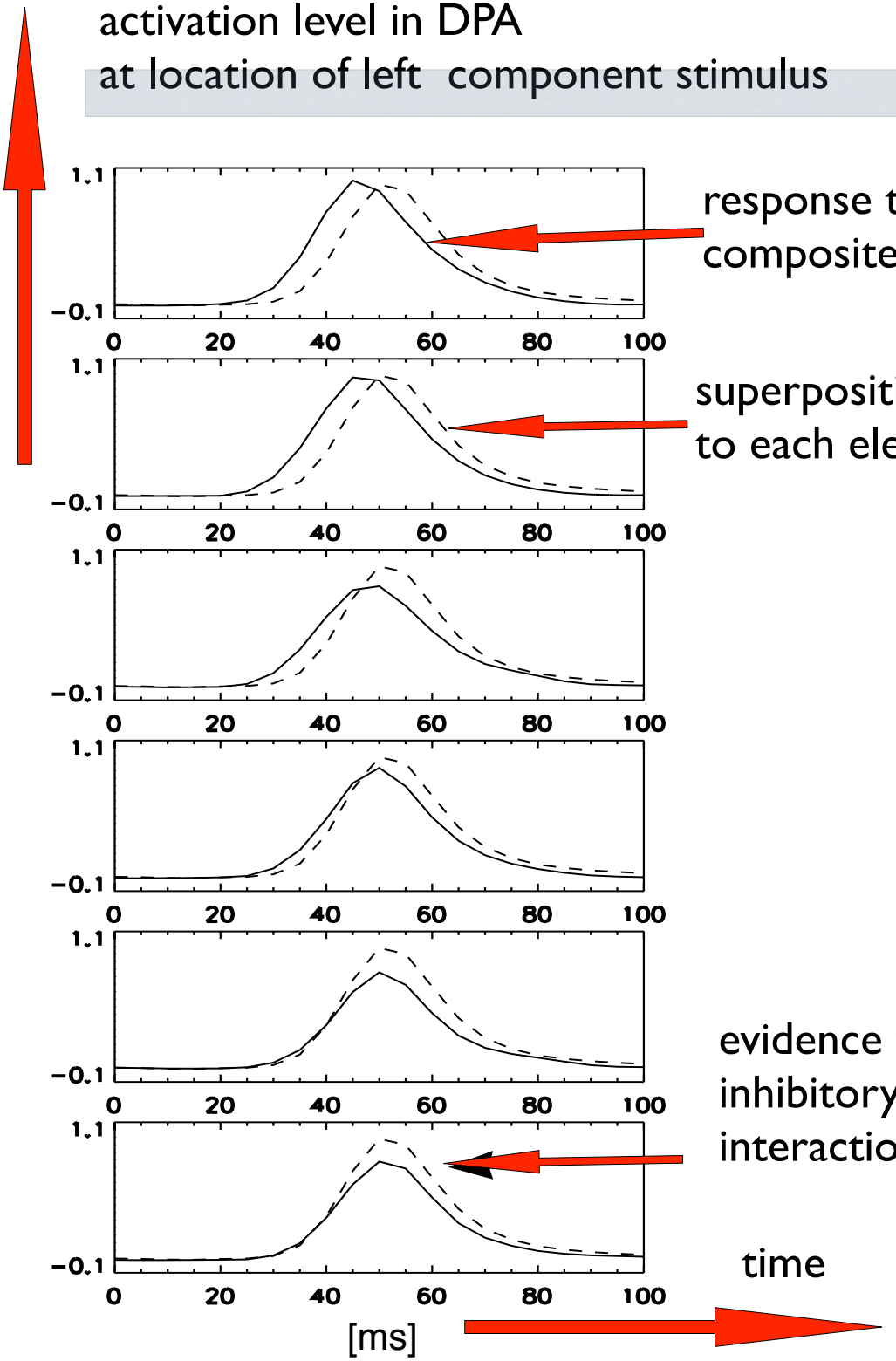
0.4°

- by comparing DPA of composite stimuli to superposition of DPAs of the two elementary stimuil obtain evidence for interaction
  - early excitation
  - late inhibition

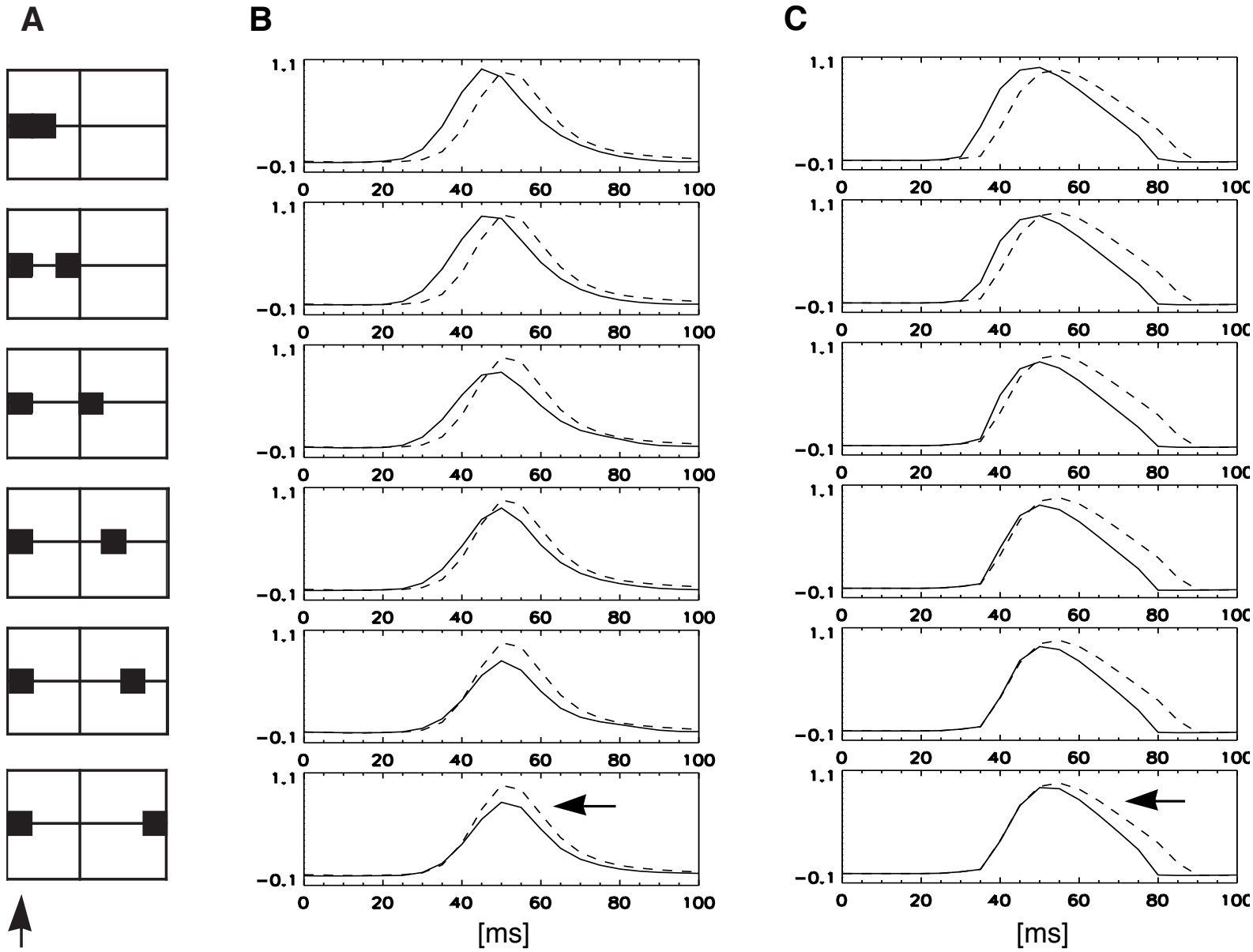
# interaction



activation level in DPA
at location of left component stimulus



# model by dynamic field:




stimulus experiment DFT model

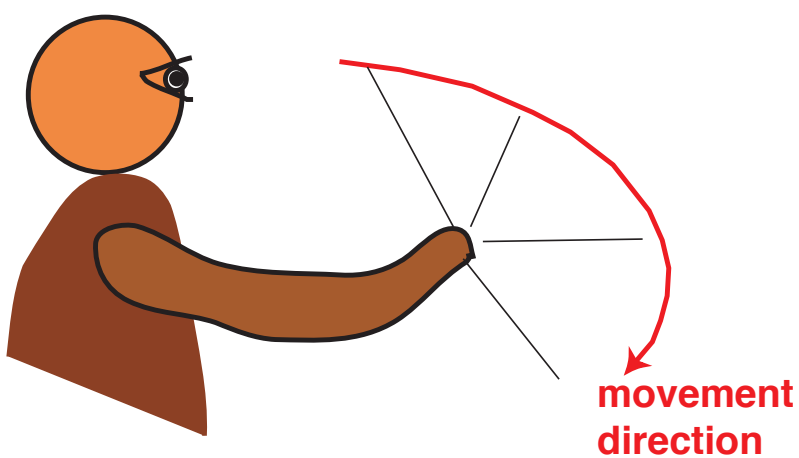
# Neurophysiological grounding of DFT

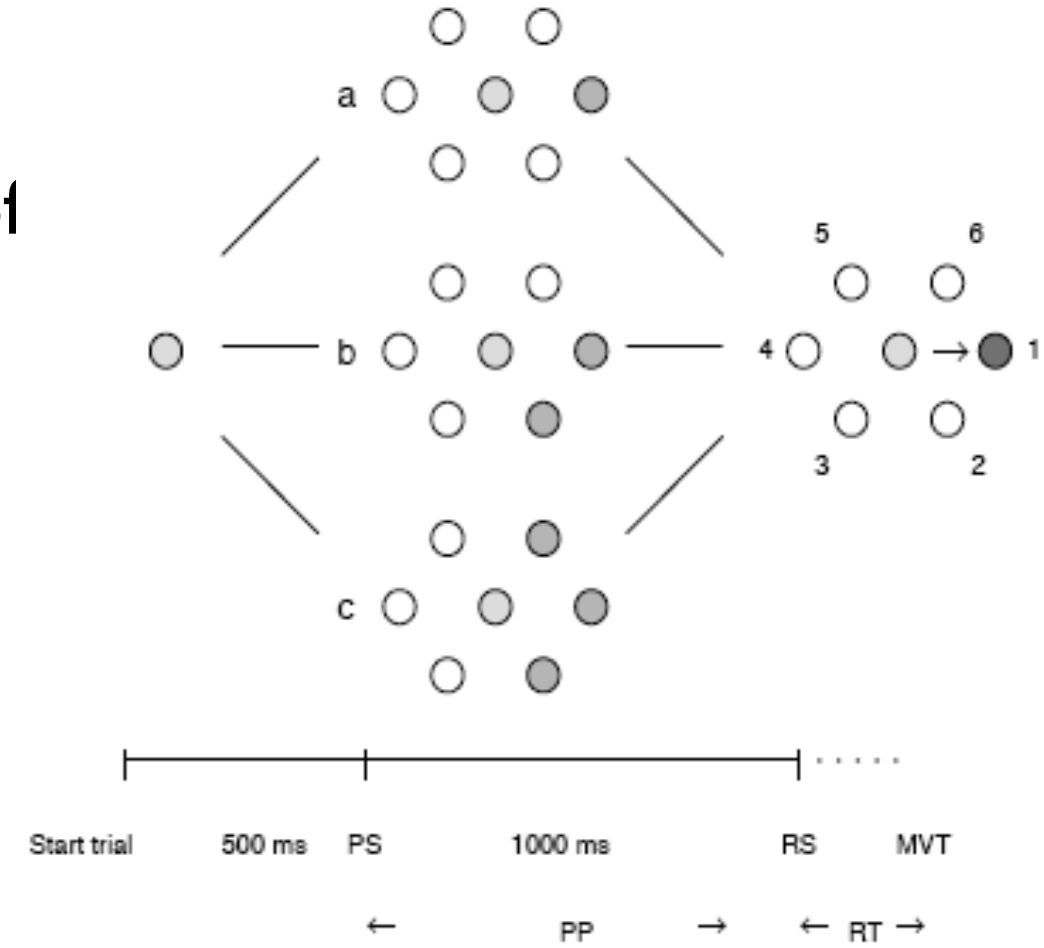
- Example 2: primary motor cortex (M1), population representation of movement direction of the hand

Bastian, Riehle, Schöner, 2003

# Task

- center-out movement task for macaque
- with varying amounts of prior information

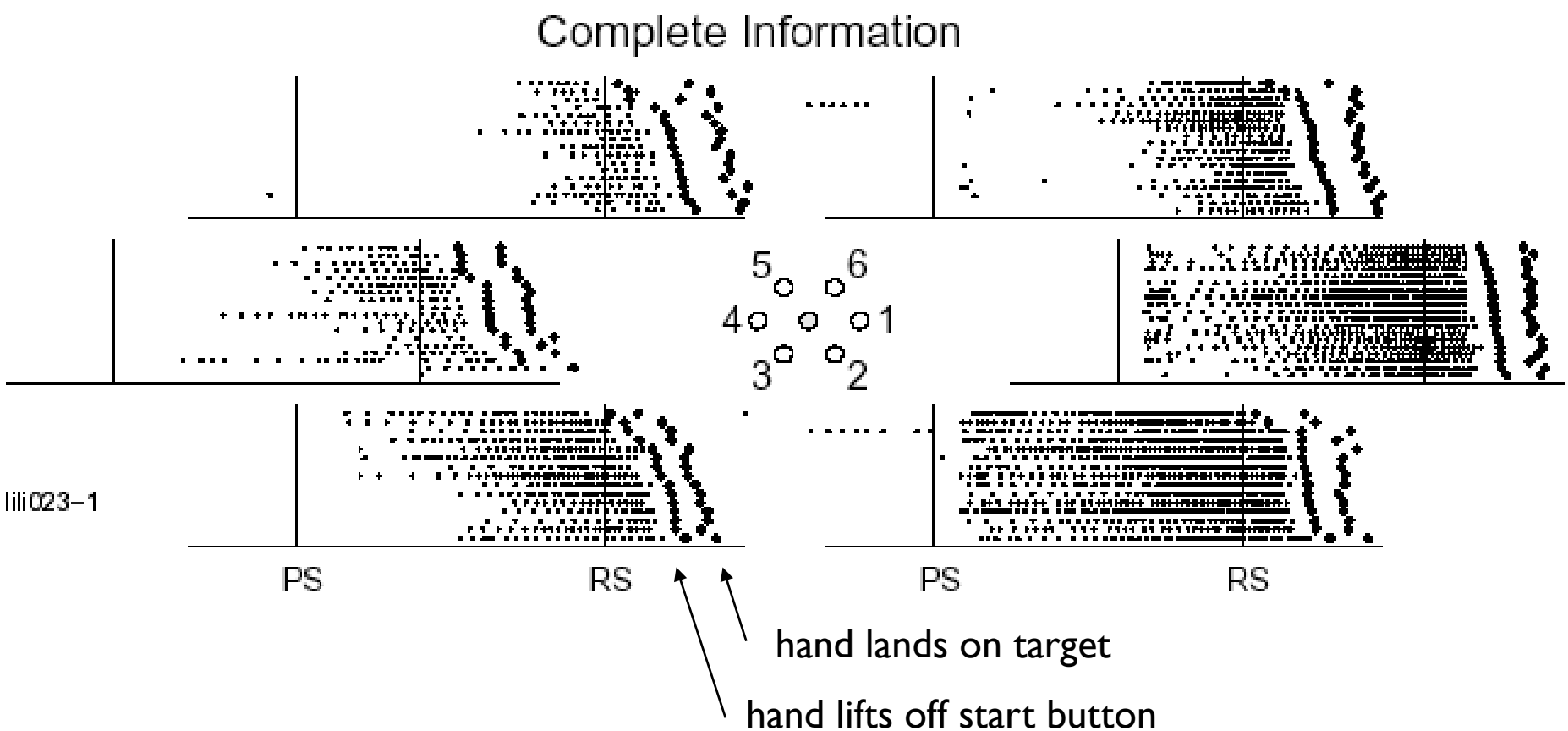



Bastian, Riehle, Schöner, 2003

# Tuning of neurons in MI to movement direction

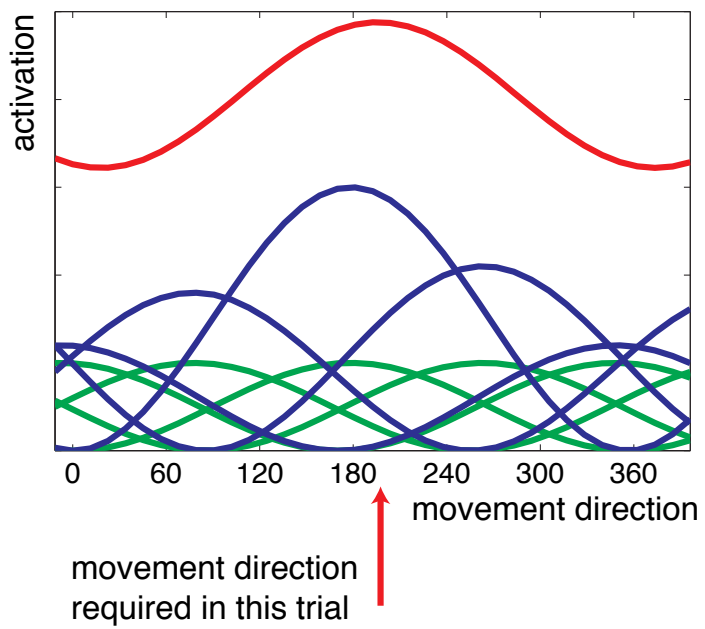
- trials aligned by go signals, ordered by reaction time

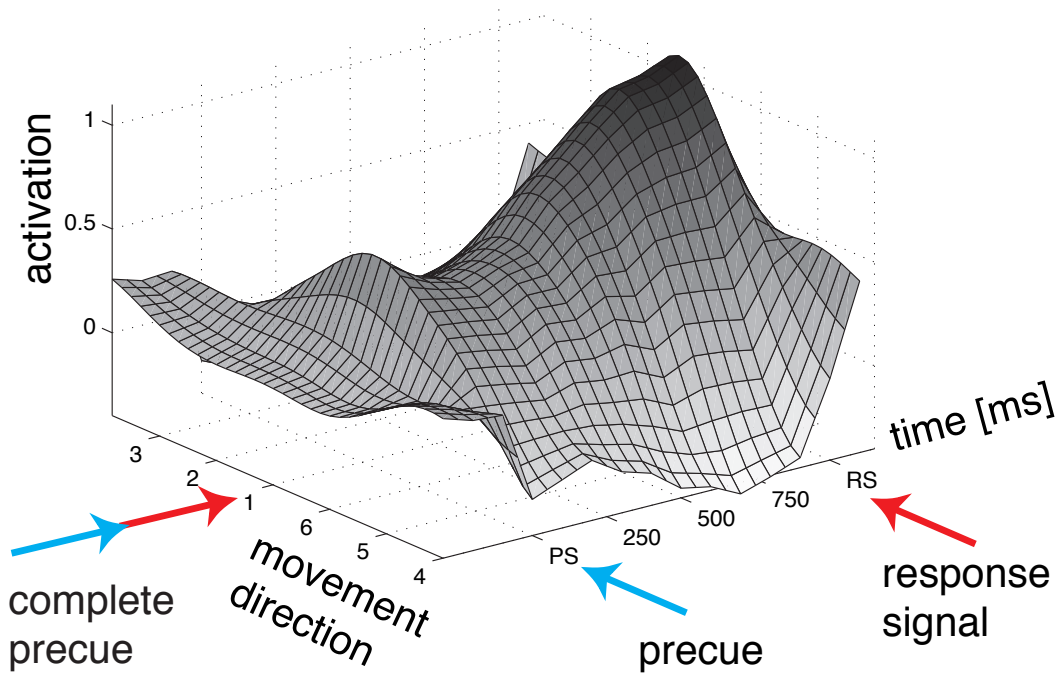


# Distribution of Population Activation (DPA)

Distribution of population activation =

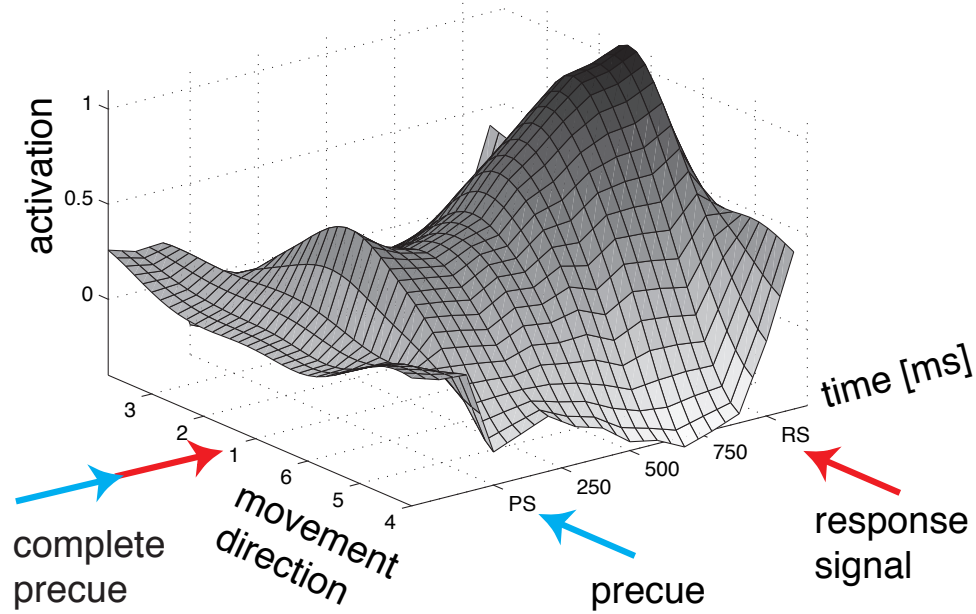
$\Sigma_{\text{neurons}}$ tuning curve * current firing rate

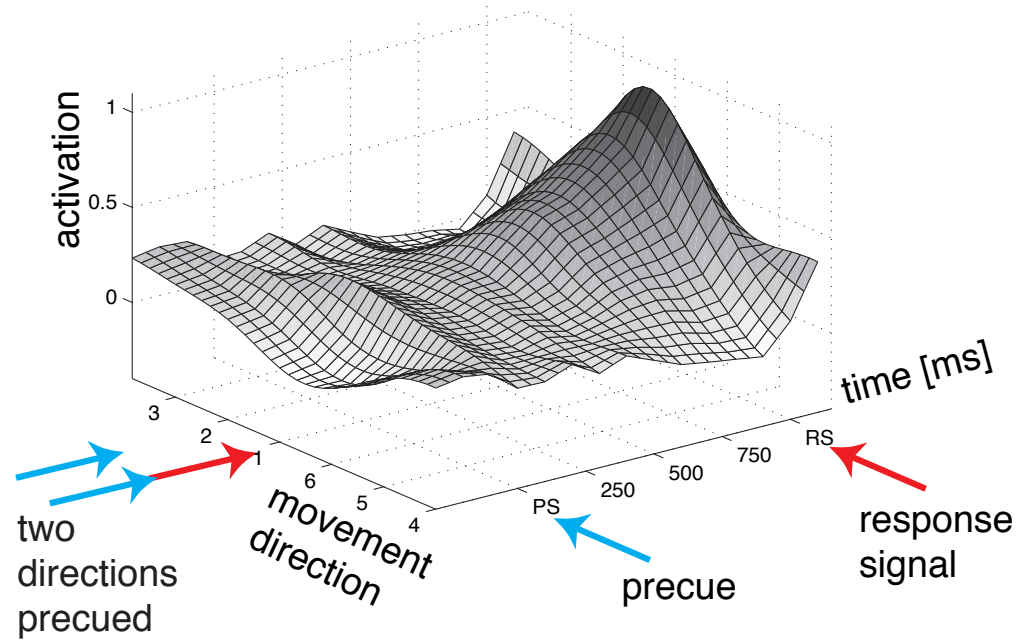


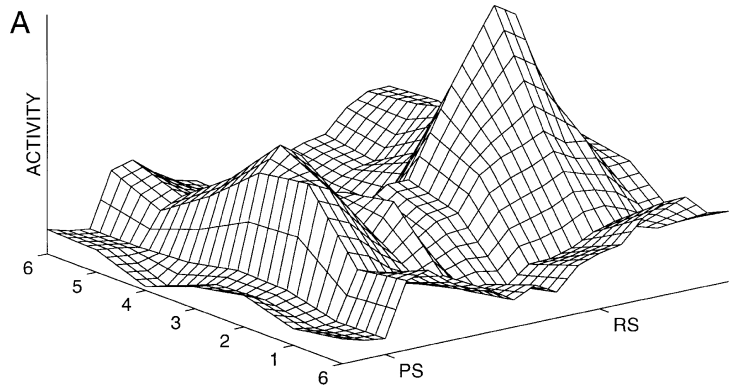

[Bastian, Riehle, Schöner, 2003]

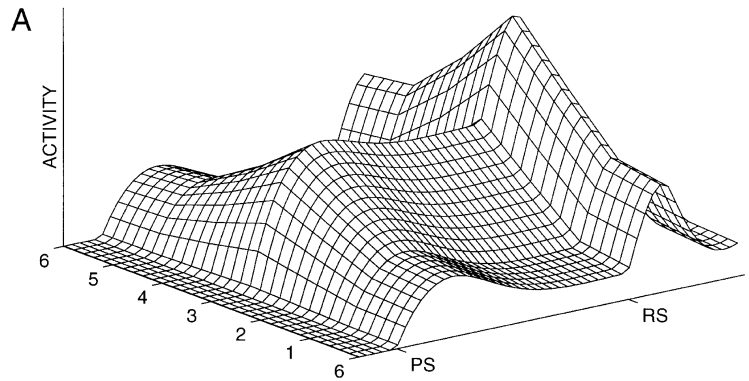
- look at temporal evolution of DPA
- or DPAs in new conditions, here: DPA reflects prior information

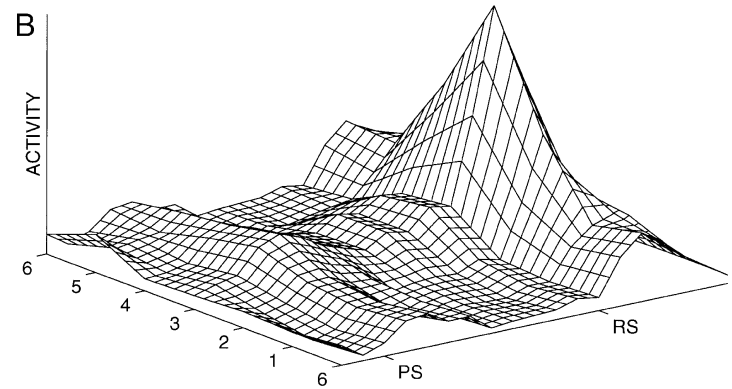


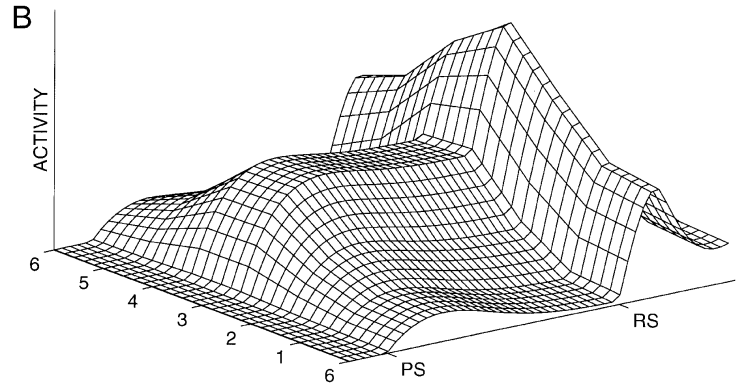



# Theory-Experiment



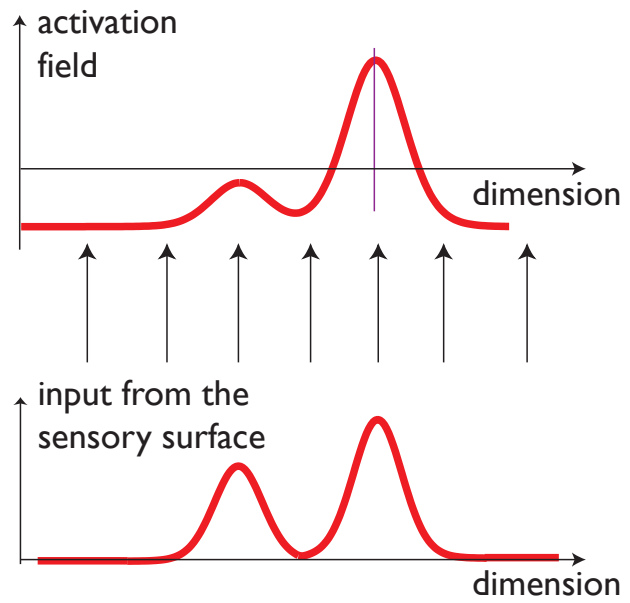

[Bastian, Riehle, Erlhagen, Schöner, 98]

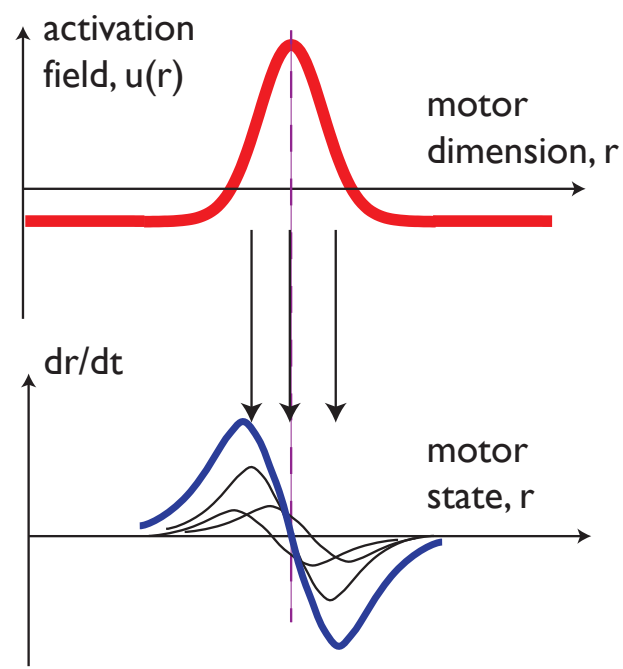
# Distributions of Population Activation are abstract

- 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

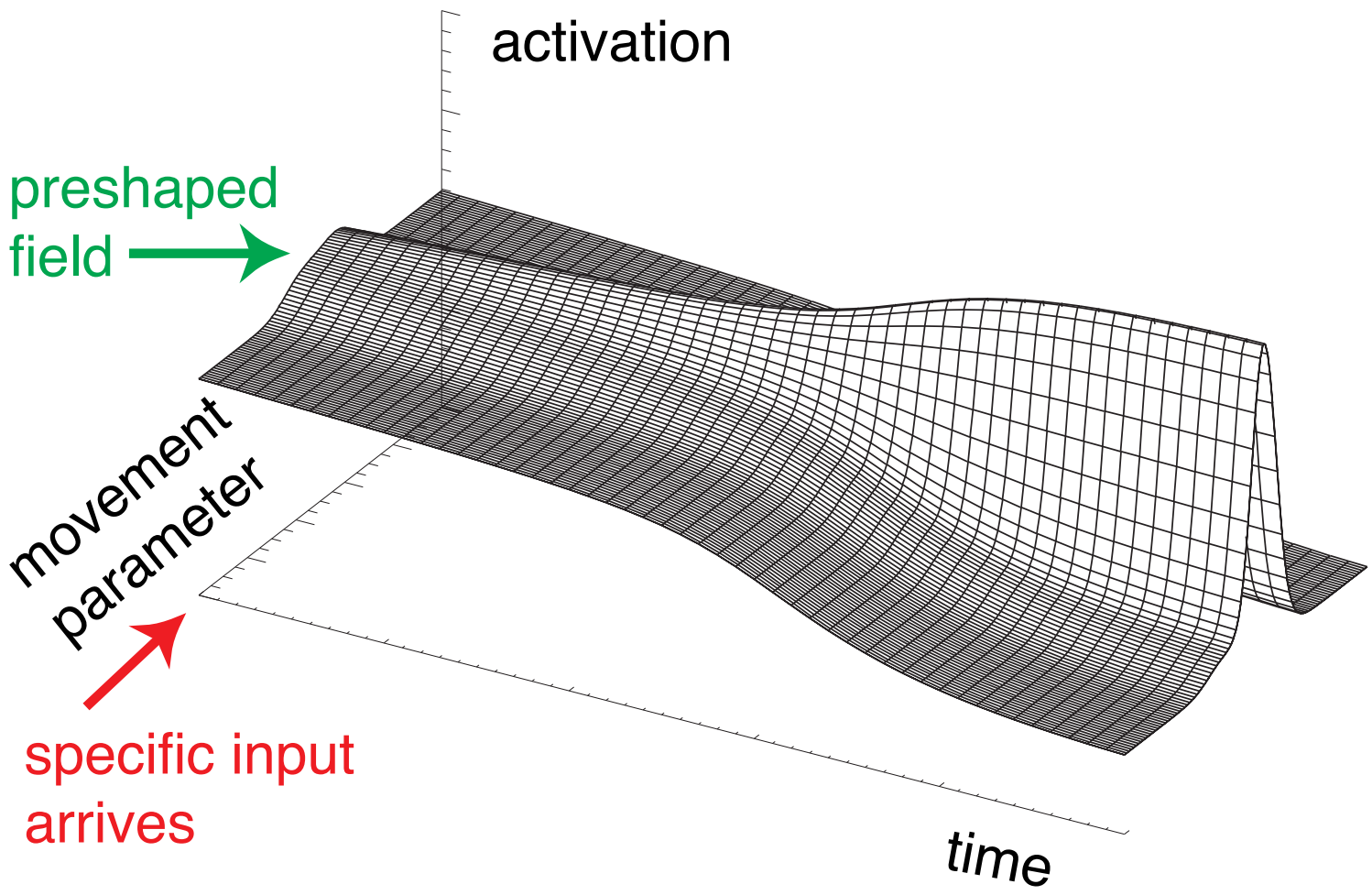
# … back to the activation fields

- that are "defined" over the appropriate dimension just as population code is…
- in building DFT models, we must ensure that this is actually true by setting up the appropriate input/output connectivity

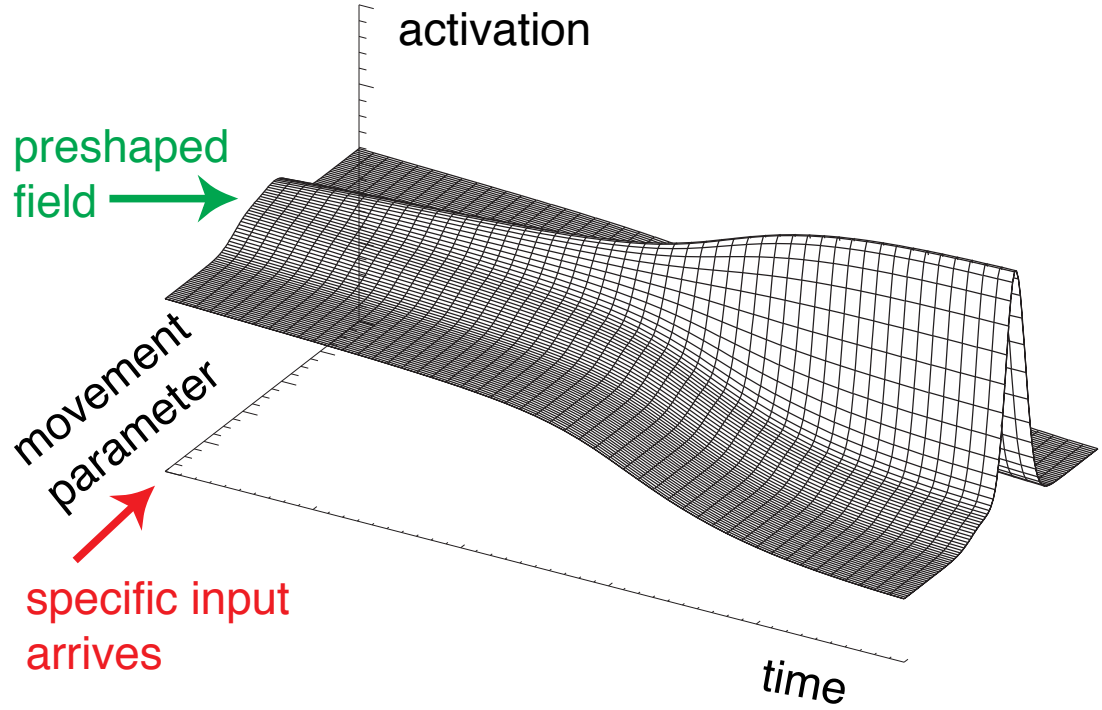


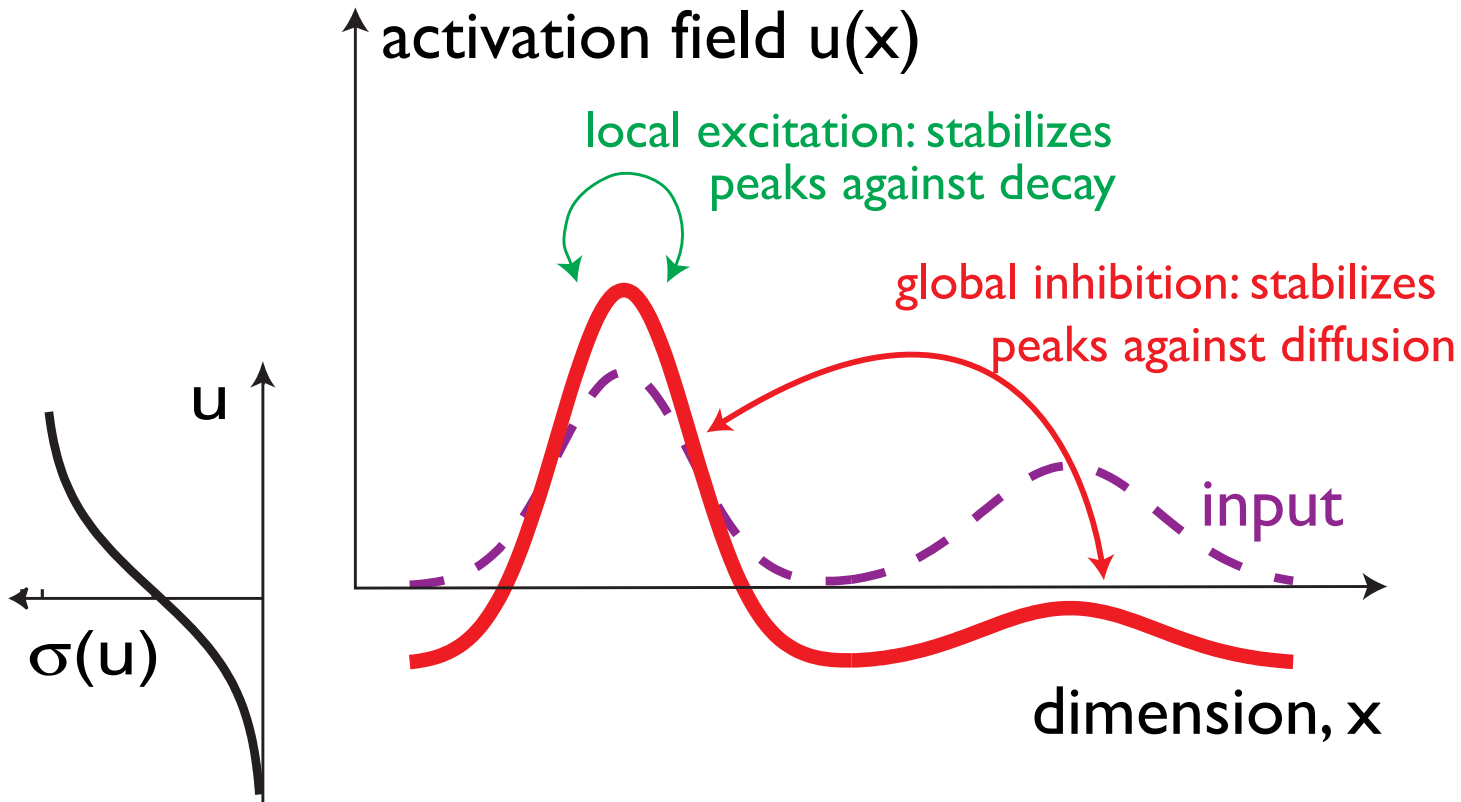



# Neural dynamics of activation fields



# The neural dynamics a activation fields is structured so that localized peaks are attractors



# mathematical formalization

Amari equation

$$\tau \dot{u}(x,t) = -u(x,t) + h + S(x,t) + \int w(x-x')\sigma(u(x',t))\ dx'$$

where

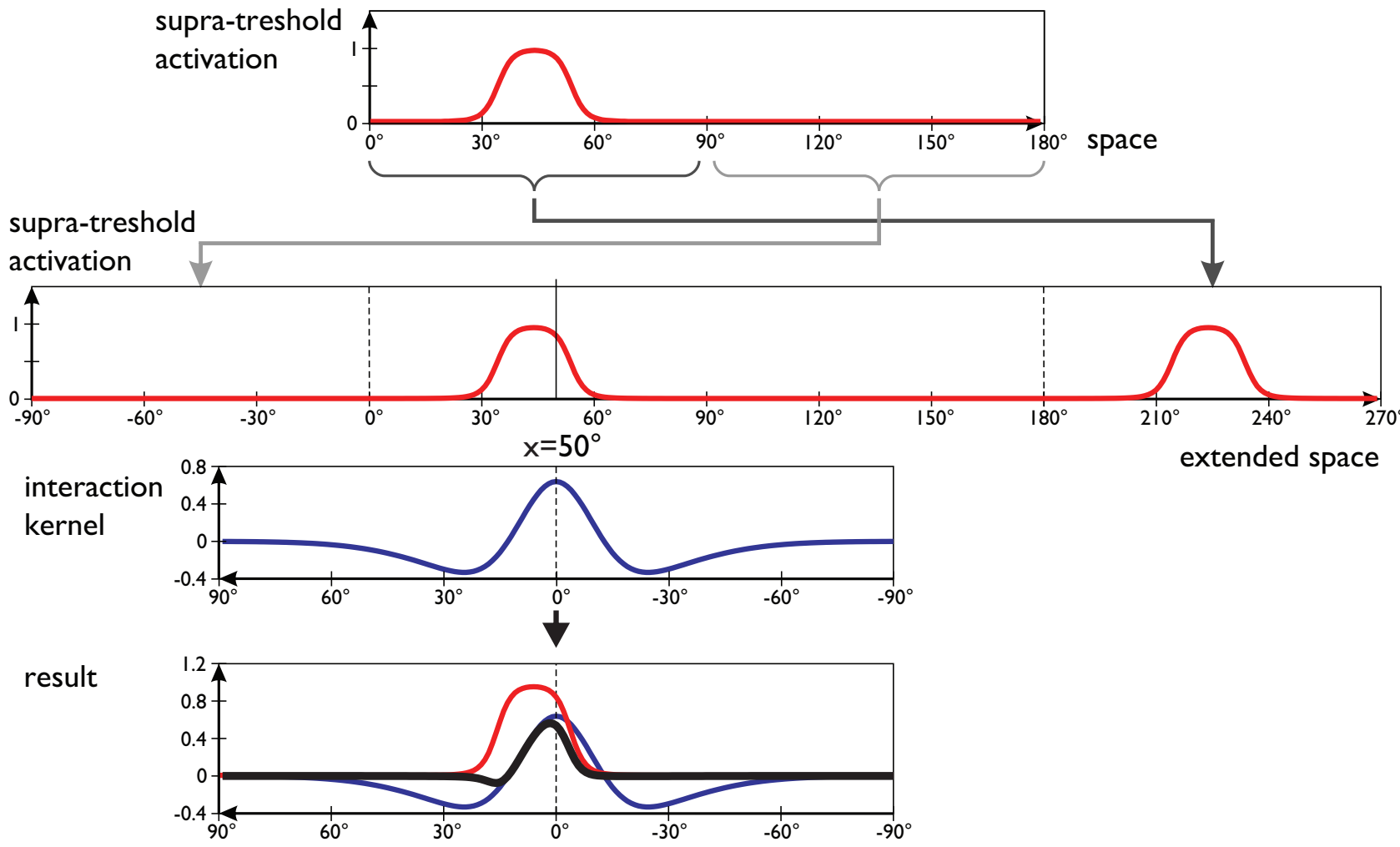
- time scale is $\tau$
- resting level is $h < 0$
- input is $S(x,t)$
- interaction kernel is

$$w(x-x') = w_i + w_e \exp\left[-\frac{(x-x')^2}{2\sigma_i^2}\right]$$

- sigmoidal nonlinearity is

$$\sigma(u) = \frac{1}{1+\exp[-\beta(u-u_0)]}$$

# Interaction: convolution



=> simulations

# Solutions and instabilities

- input driven solution (sub-threshold) vs. self-stabilized solution (peak, supra-threshold)
- detection instability
- reverse detection instability
- selection
- selection instability
- memory instability
- detection instability from boost

# Relationship to the dynamics of discrete activation variables