# Dynamic Field Theory: Selection Decisions

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### Recall from last lecture ...

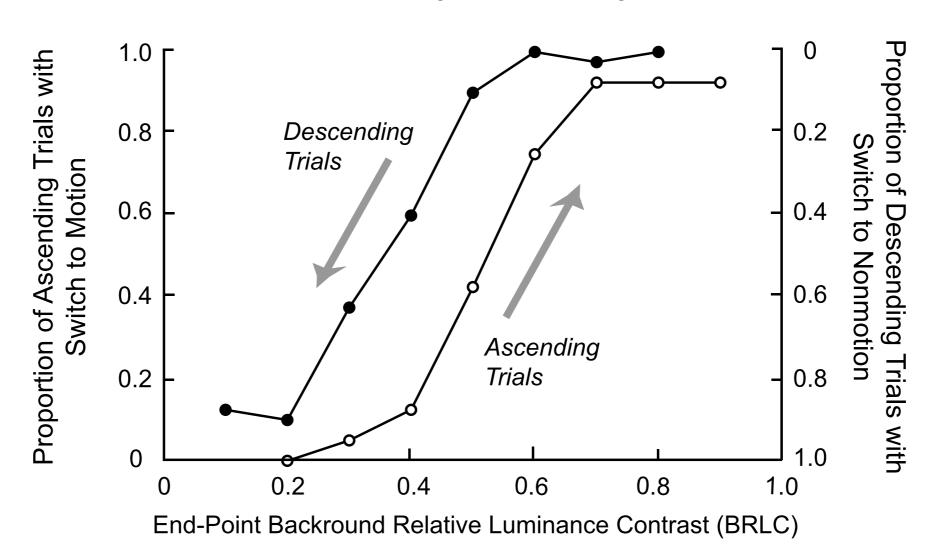
### Solutions and instabilities

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

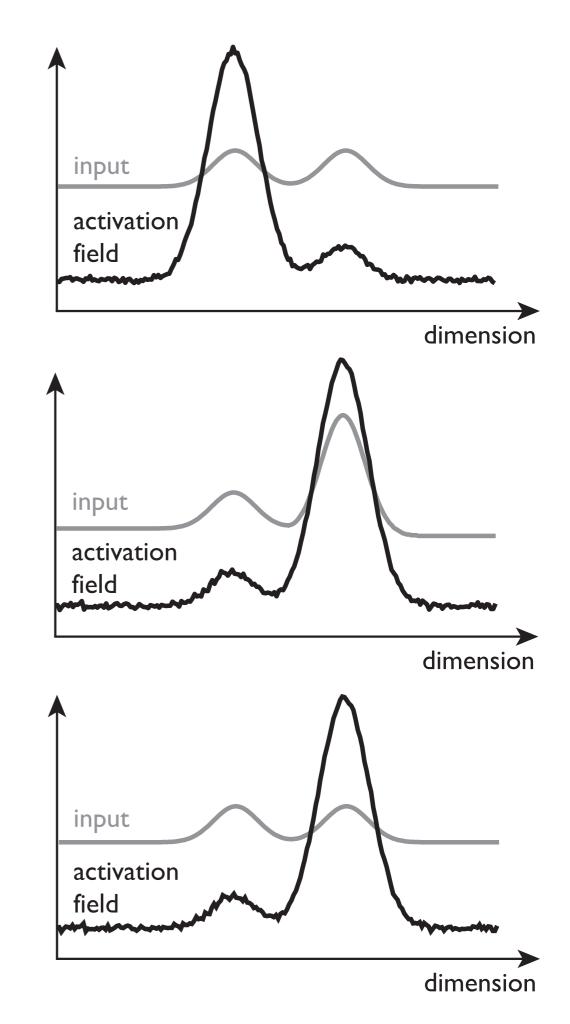
# Psychophysical evidence for the detection instability

perceptual hysteresis of motion detection

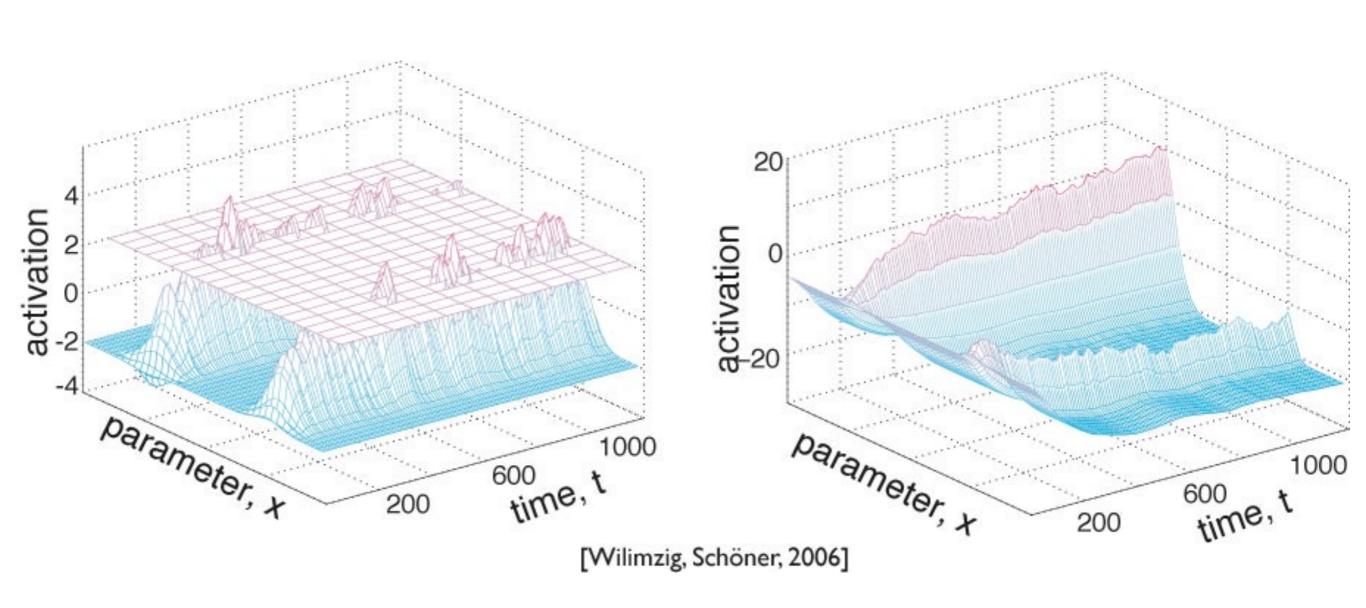
H. S. Hock, G. Schöner / Seeing and Perceiving 23 (2010) 173–195



# selection instability



## stabilizing selection decisions

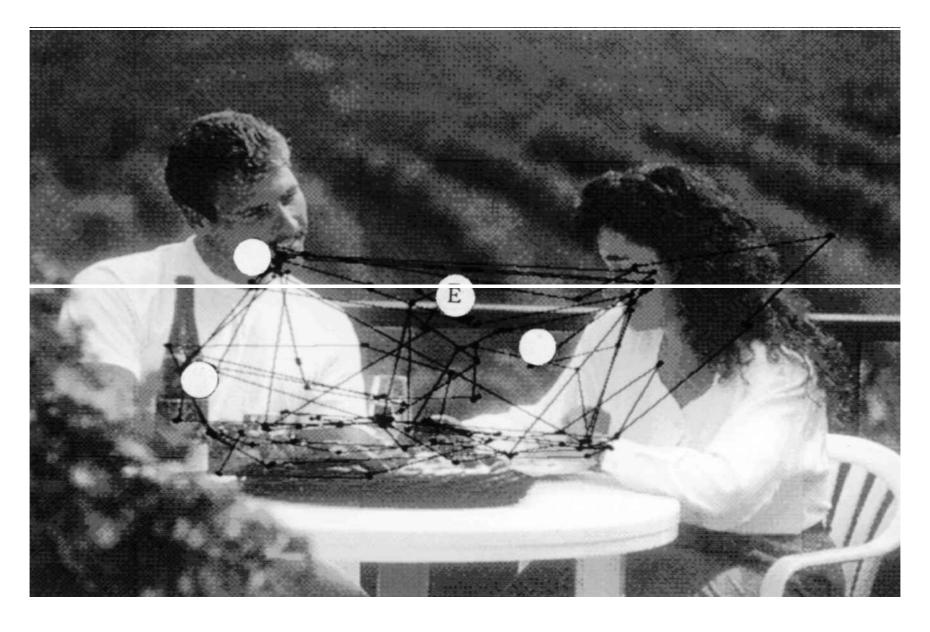


# behavioral signatures of selection decisions

- In most experimental situations, the correct selection decision is cued by an "imperative signal" leaving no actual freedom of "choice" to the participant (only the freedom of "error")
- reasons are experimental
- when performance approaches chance level, then close to "free choice"
- because task set plays a major role in such tasks, I will discuss these only a little later

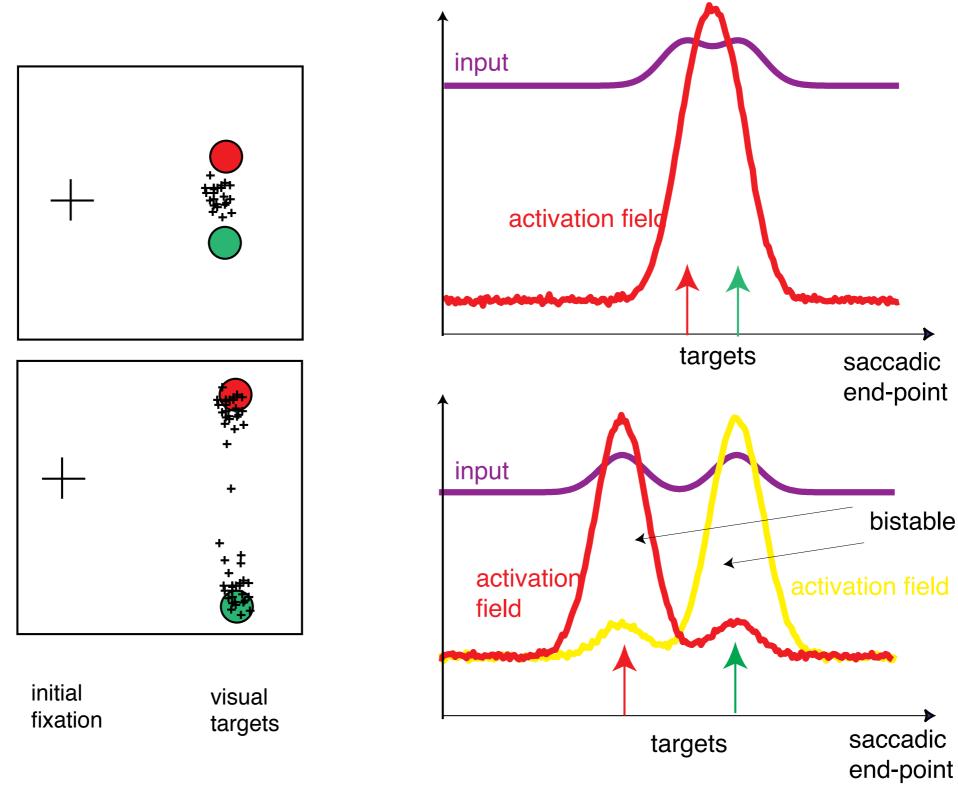
# one system of "free choice"

selecting a new saccadic location



[O'Reagan et al., 2000]

## saccade generation

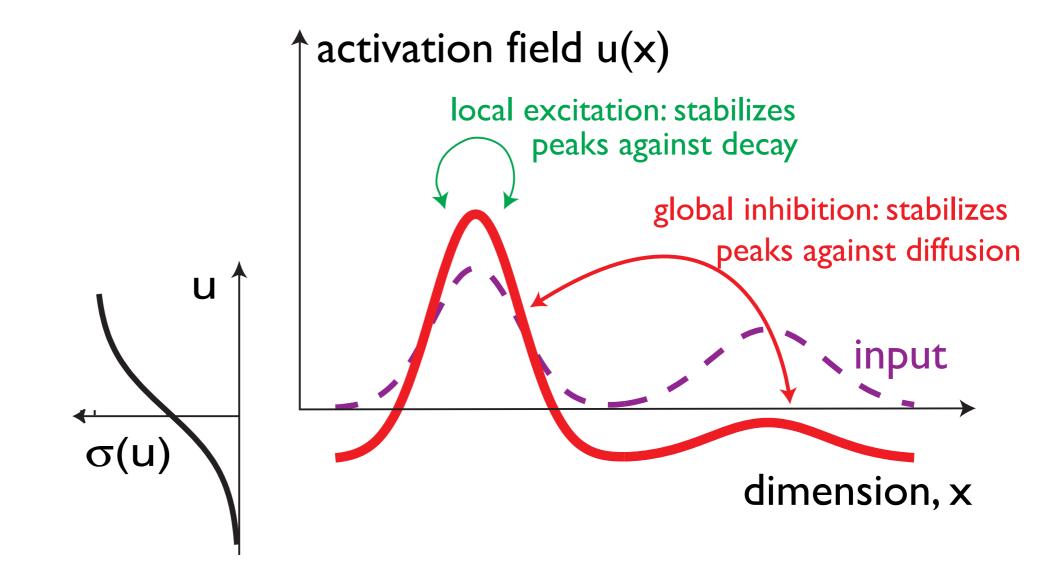


[after: Ottes et al., Vis. Res. 25:825 (85)]

[after Kopecz, Schöner: Biol Cybern 73:49 (95)]

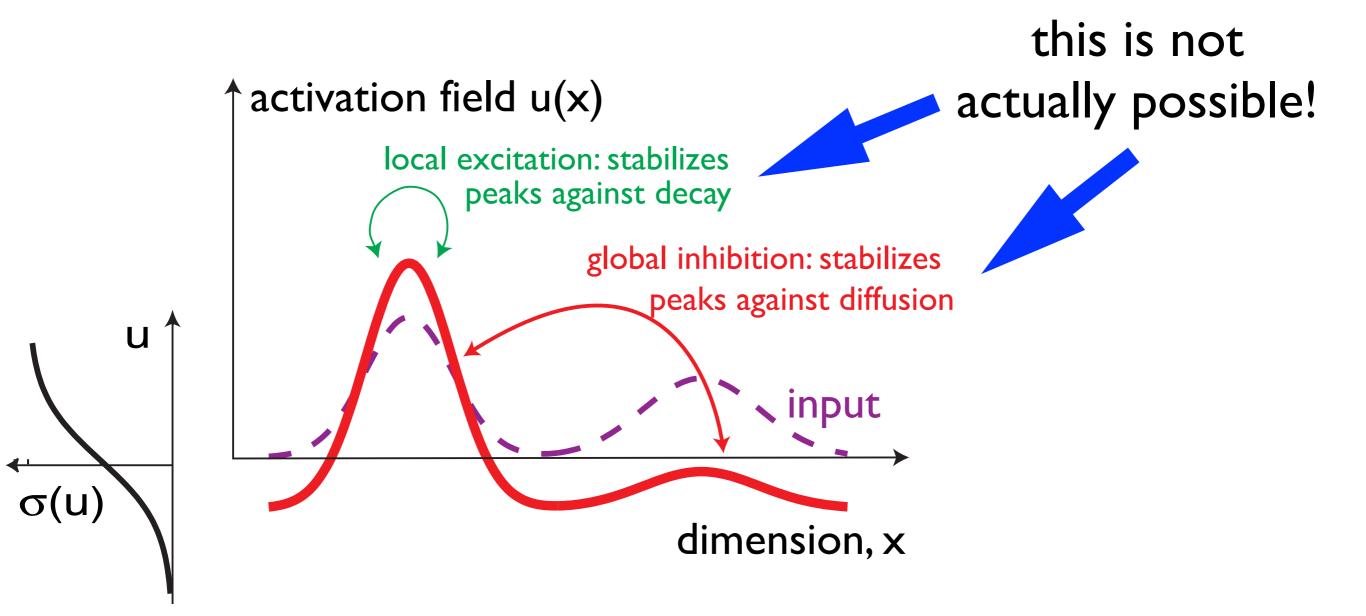
### ... so far we assumed

that a single population of activation variable mediates both the excitatory and the inhibitory coupling required to make peaks attractors



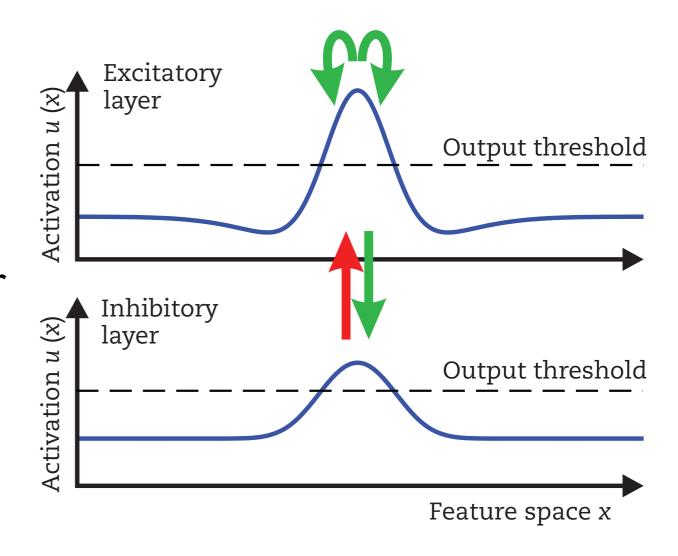
### But: Dale's law

- says: every neuron forms with its axon only one type of synapse on the neurons it projects onto
- and that is either excitatory or inhibitory



# 2 layer neural fields

- inhibitory coupling is mediated by inhibitory interneurons that
  - are excited by the excitatory layer
  - and in turn inhibit the inhibitory layer



[chapter 3 of the book]

# 2 layer Amari fields

$$\tau_{u}\dot{u}(x,t) = -u(x,t) + h_{u} + s(x,t) + \int k_{uu}(x-x')g(u(x',t))dx' - \int k_{uv}(x-x')g(v(x',t))dx'$$

$$\tau_{v}\dot{v}(x,t) = -v(x,t) + h_{v} + \int k_{vu}(x-x')g(u(x',t))dx'$$

$$t(x,t) = -v(x,t) + h_{v} + \int k_{vu}(x-x')g(u(x',t))dx'$$

#### with projection kernels

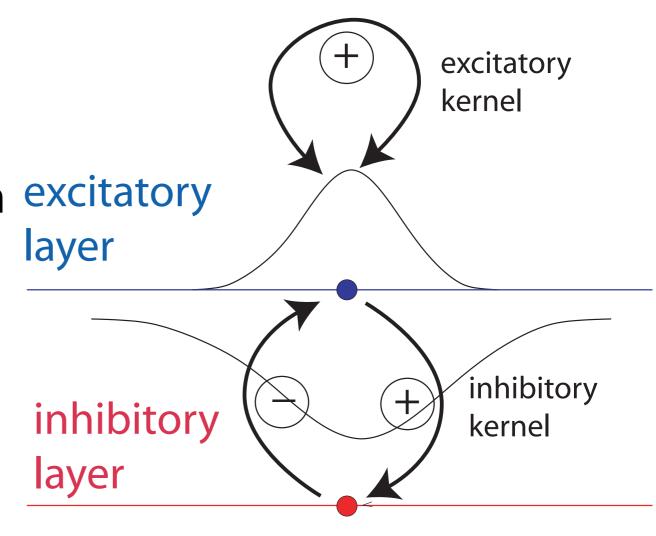
$$k_{uu}(x-x') = c_{uu} \cdot \exp\left(-\frac{(x-x')^2}{2\sigma_{uu}^2}\right)$$

σ

## simulation

# **Implications**

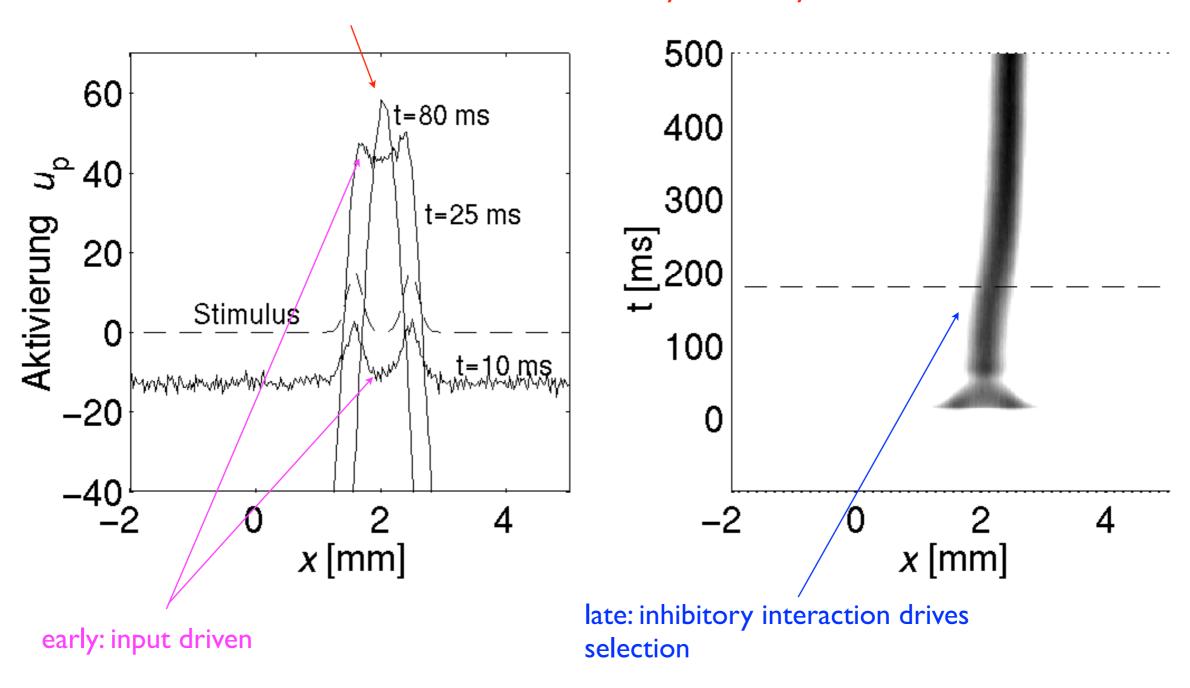
- the fact that inhibition arises only after excitation has been induced has observable consequences in excitatory the time course of decision layer making:
  - initially input-dominated
  - early excitatory interaction
  - late inhibitory interaction



[figure: Wilimzig, Schneider, Schöner, Neural Networks, 2006]

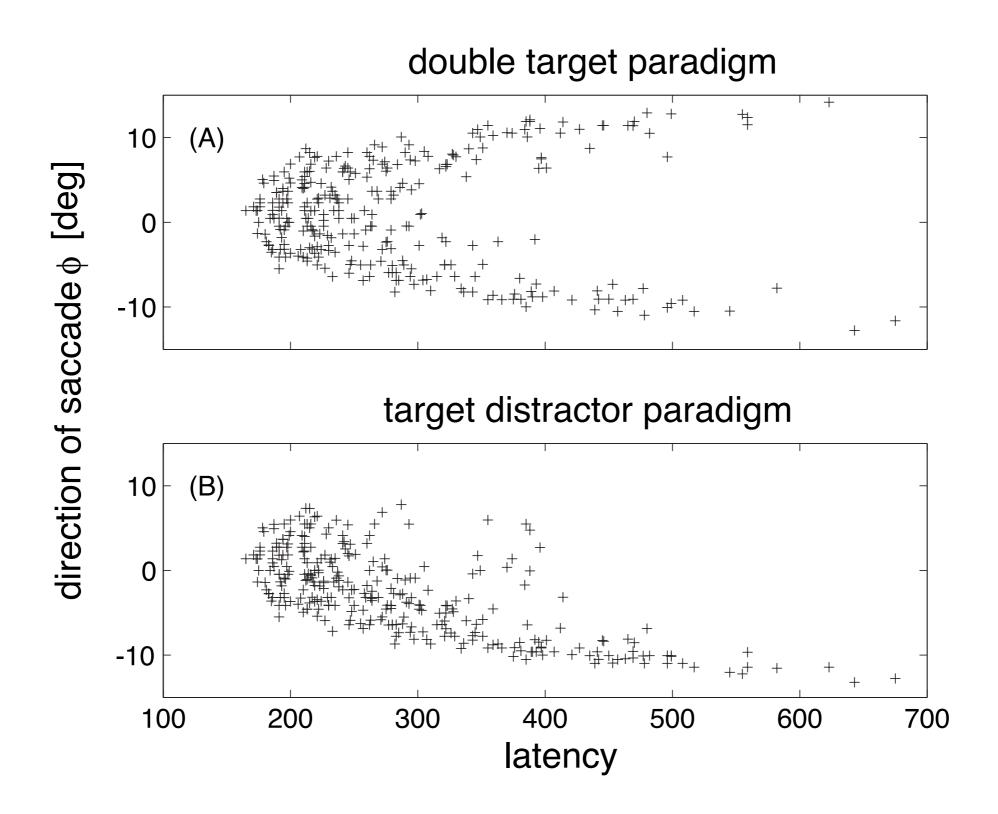
### time course of selection

intermediate: dominated by excitatory interaction



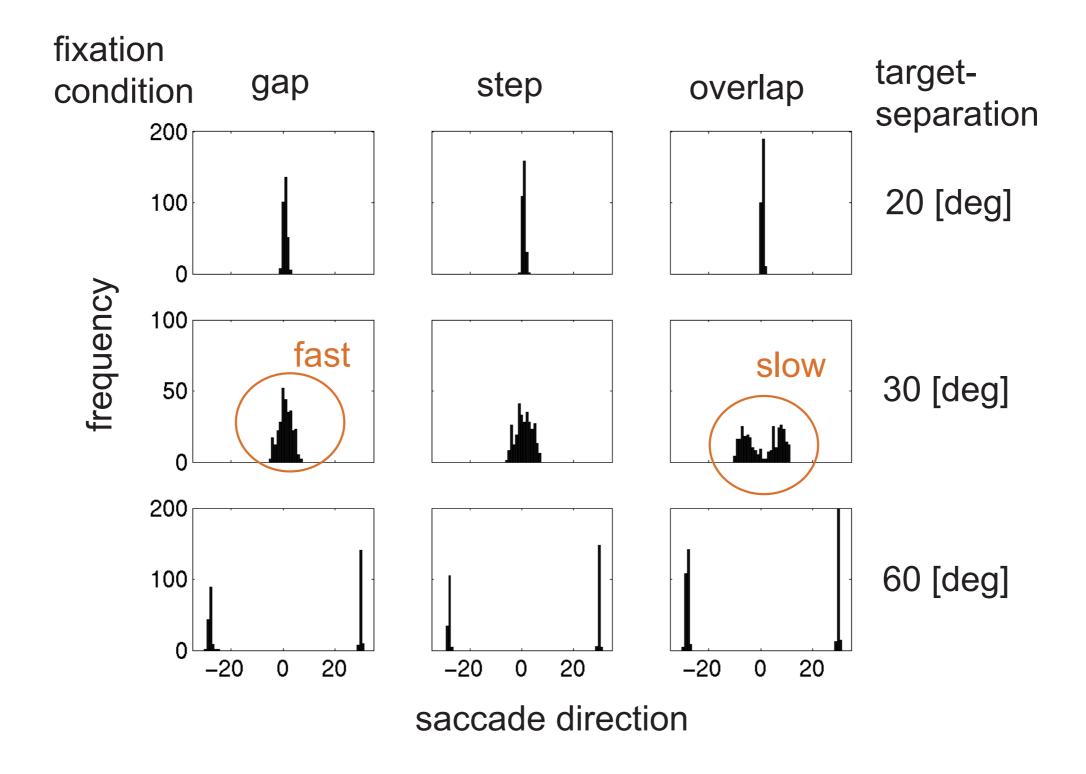
[figure: Wilimzig, Schneider, Schöner, Neural Networks, 2006]

# => early fusion, late selection



[figure: Wilimzig, Schneider, Schöner, Neural Networks, 2006]

### fixation and selection



[figure: Wilimzig, Schneider, Schöner, Neural Networks, 2006]

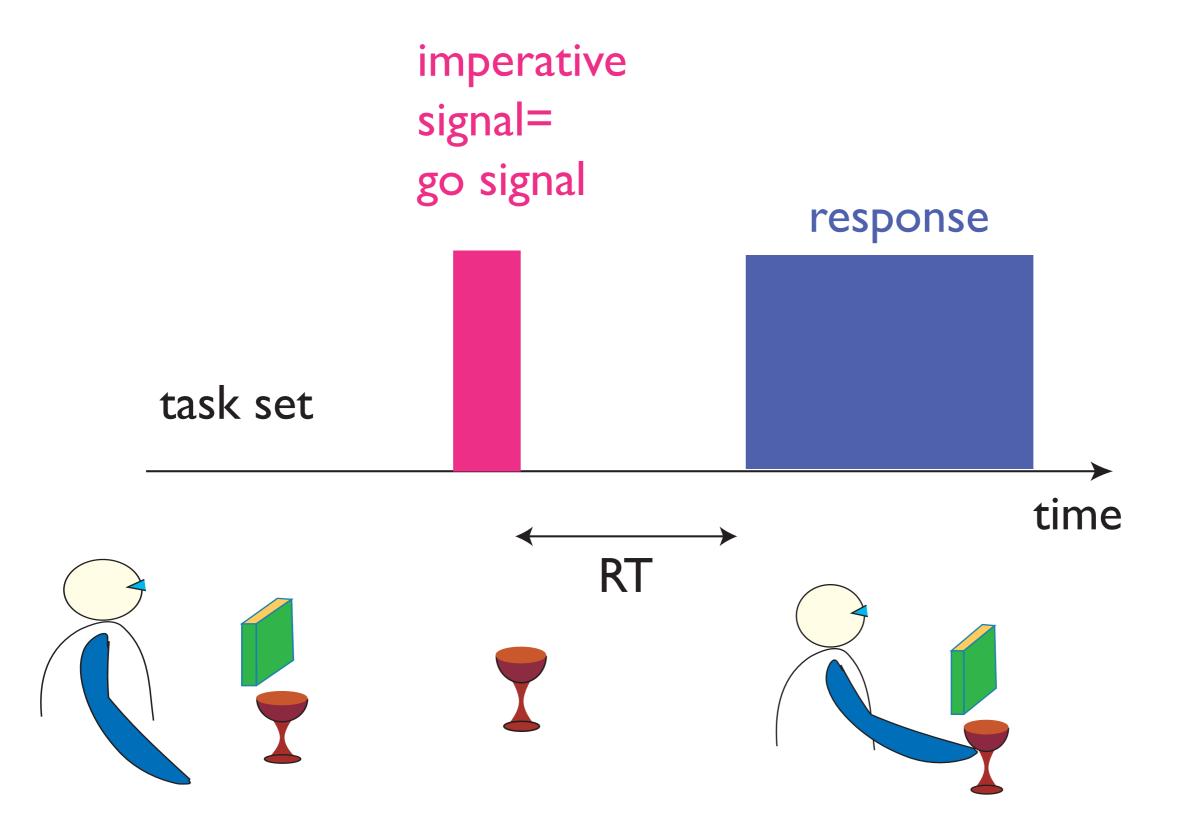
# 2 layer fields afford oscillations

- => simulation
- (oscillatory states for enhanced coupling among fields)
- (generic nature of oscillations)

# studying selection decisions in the laboratory

using an imperative signal...

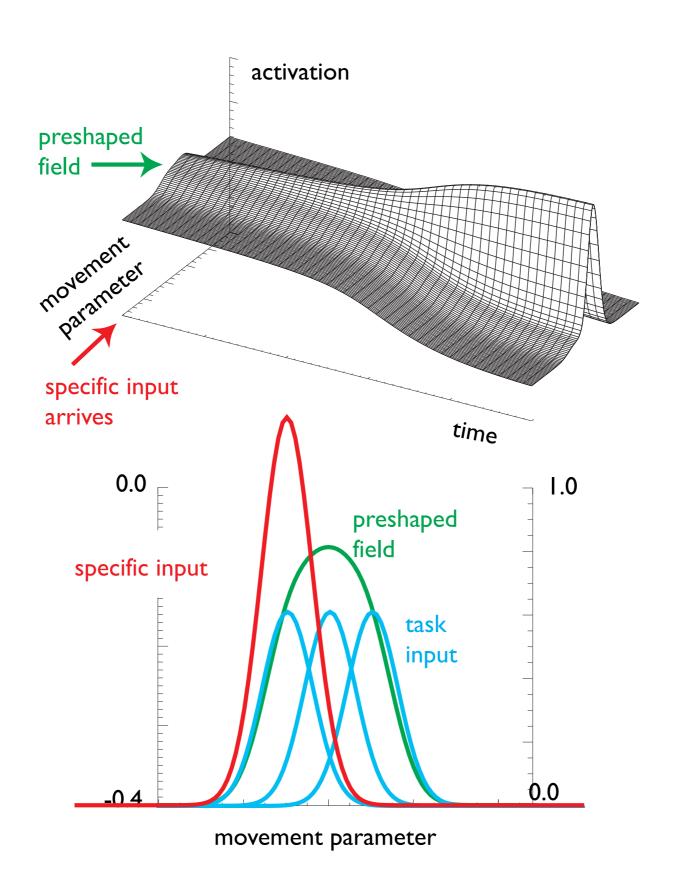
# reaction time (RT) paradigm



#### the task set

- is the critical factor in such studies of selection: which perceptual/action alternative/choices are available...
  - e.g., how many choices
  - e.g., how likely is each choice
  - e.g., how "easy" are the choices to recognize/perform
- because the task set is known to the participant prior to the presentation of the imperative signal, one may think of the task set as a "preshaping" of the underlying representation (pre=before the decision)

# notion of preshape



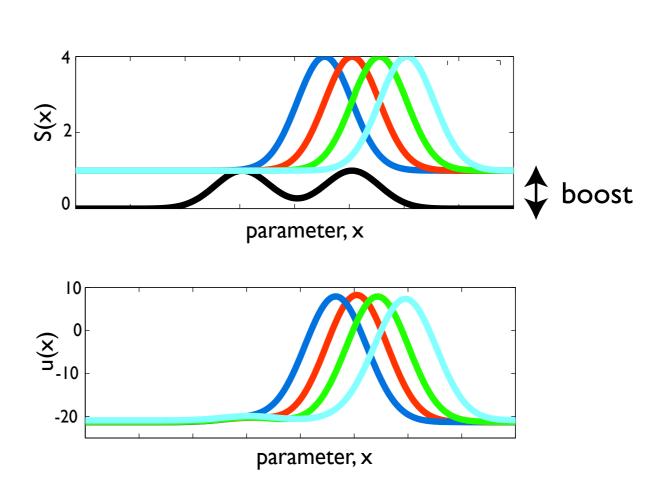
# weak preshape in selection

specific input + boost in different conditions

(X) n uoite in different conditions

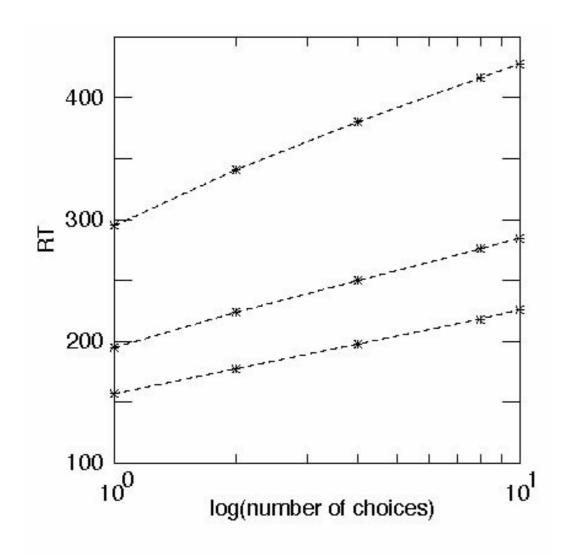
Parameter, x 0 preshape

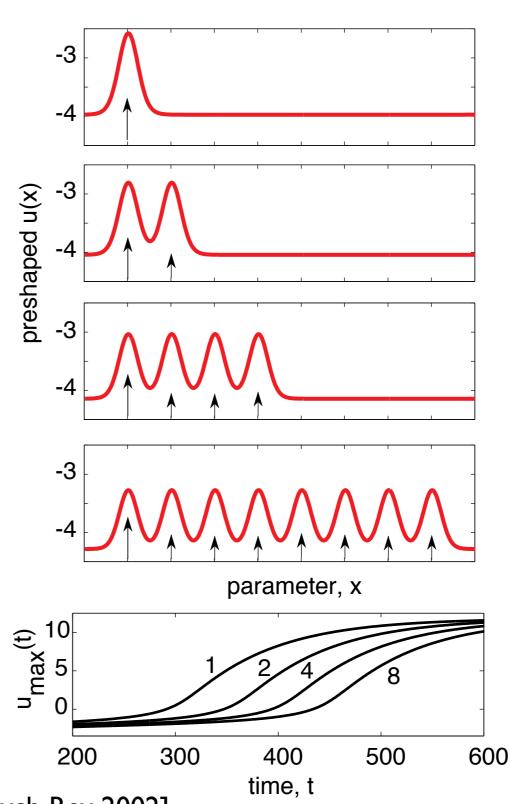
specific (imperative) input dominates and drives detection instability



# using preshape to account for classical RT data

Hick's law: RT increases with the number of choices

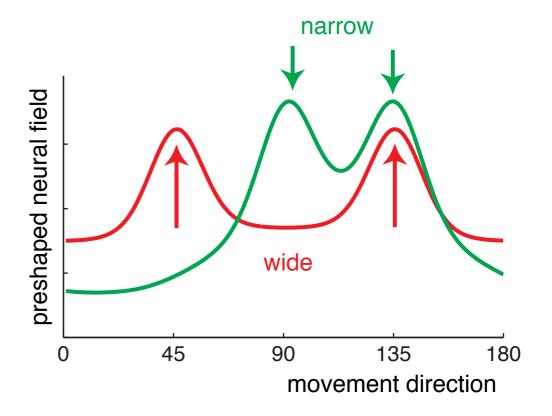


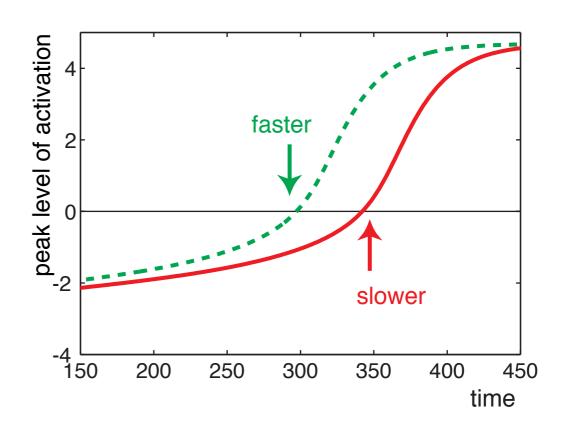


[Erlhagen, Schöner, Psych Rev 2002]

#### metric effect

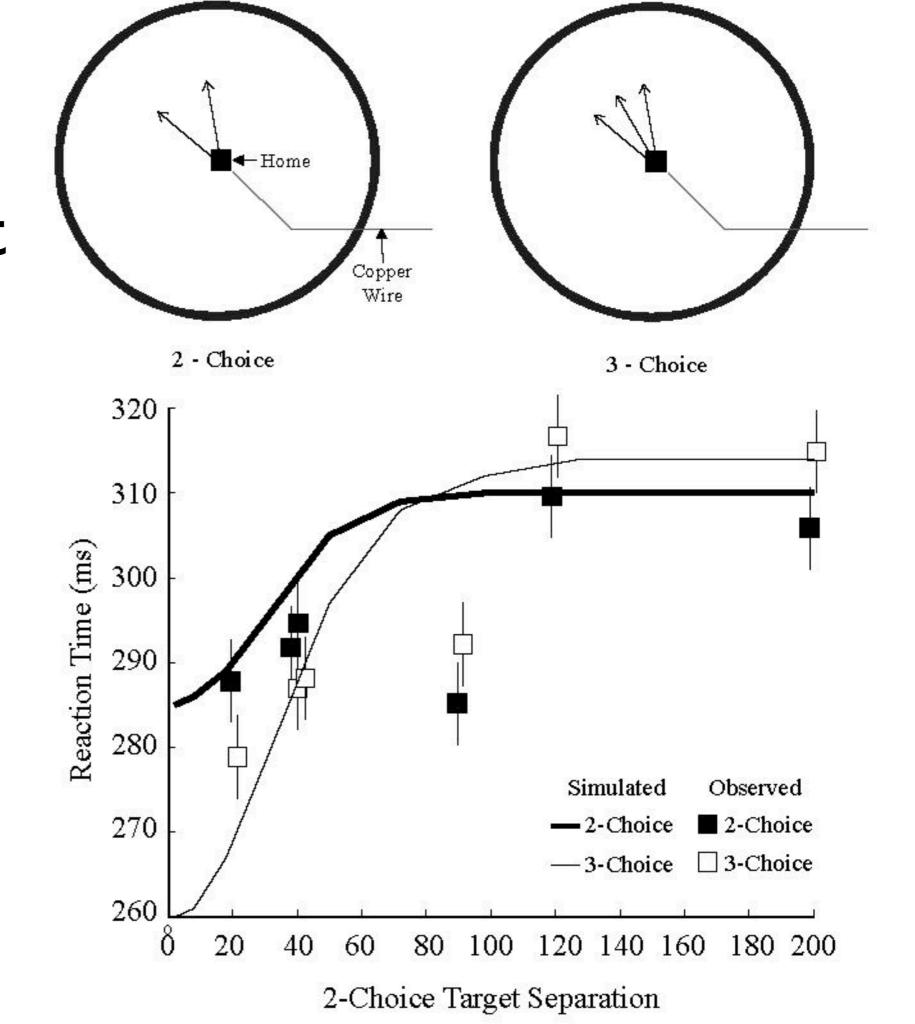
predict faster response times for metrically close than for metrically far choices

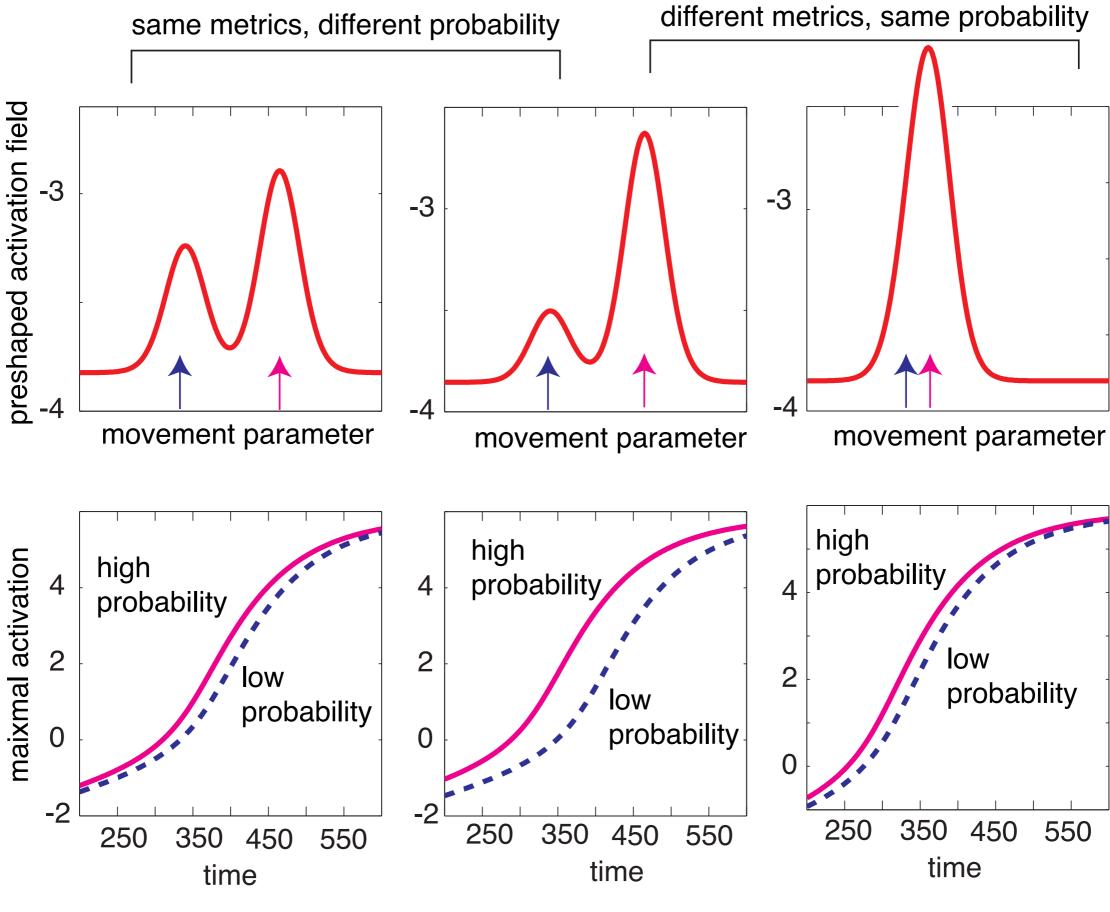




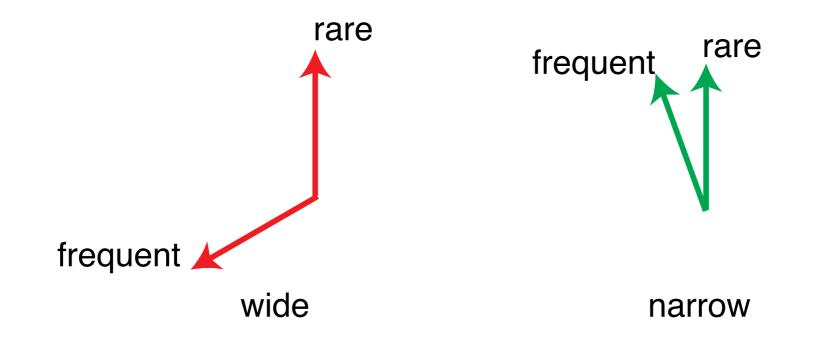
[from Schöner, Kopecz, Erlhagen, 1997]

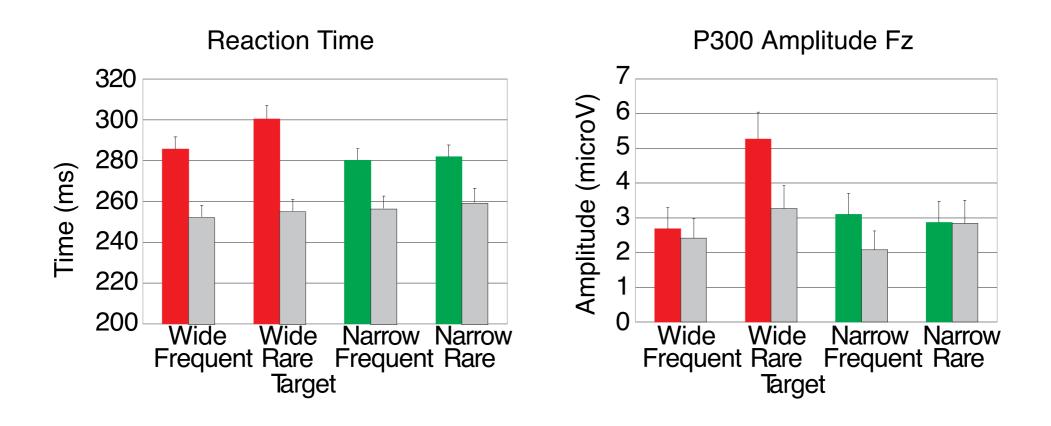
# experiment: metric effect





[from Erlhagen, Schöner: Psych. Rev. 2002]

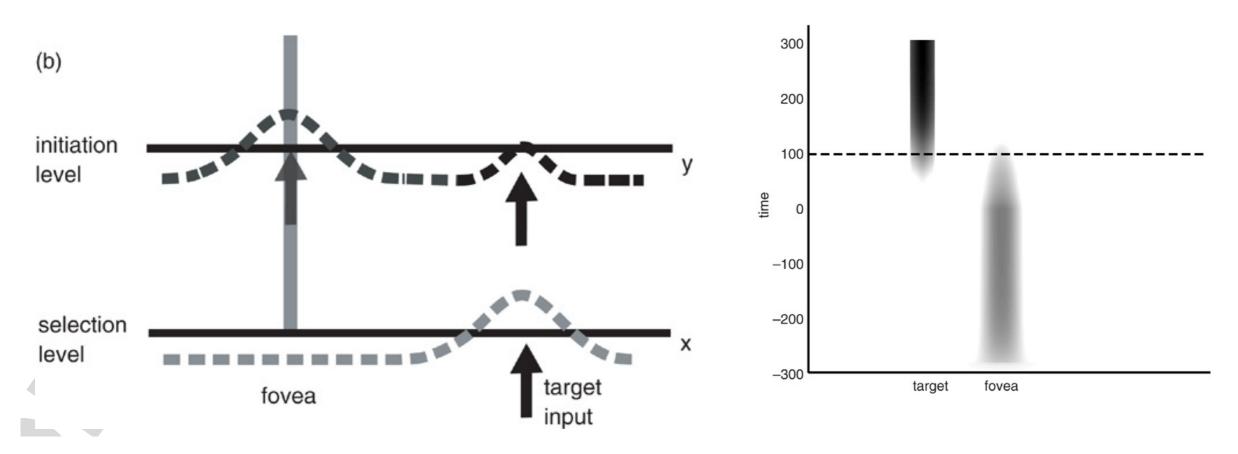




[from McDowell, Jeka, Schöner, Hatfield, 2002]

### detection-selection: overcoming fixation

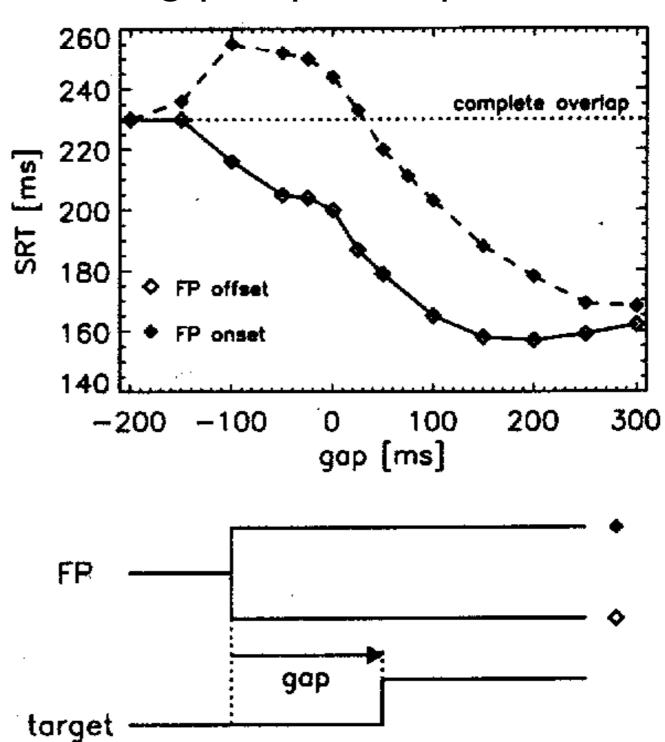
- detection can be like selection: initiating an action means terminating the non-action=fixation or posture
- example: saccade initiation



[Wilimzig, Schneider, Schöner, 2006]

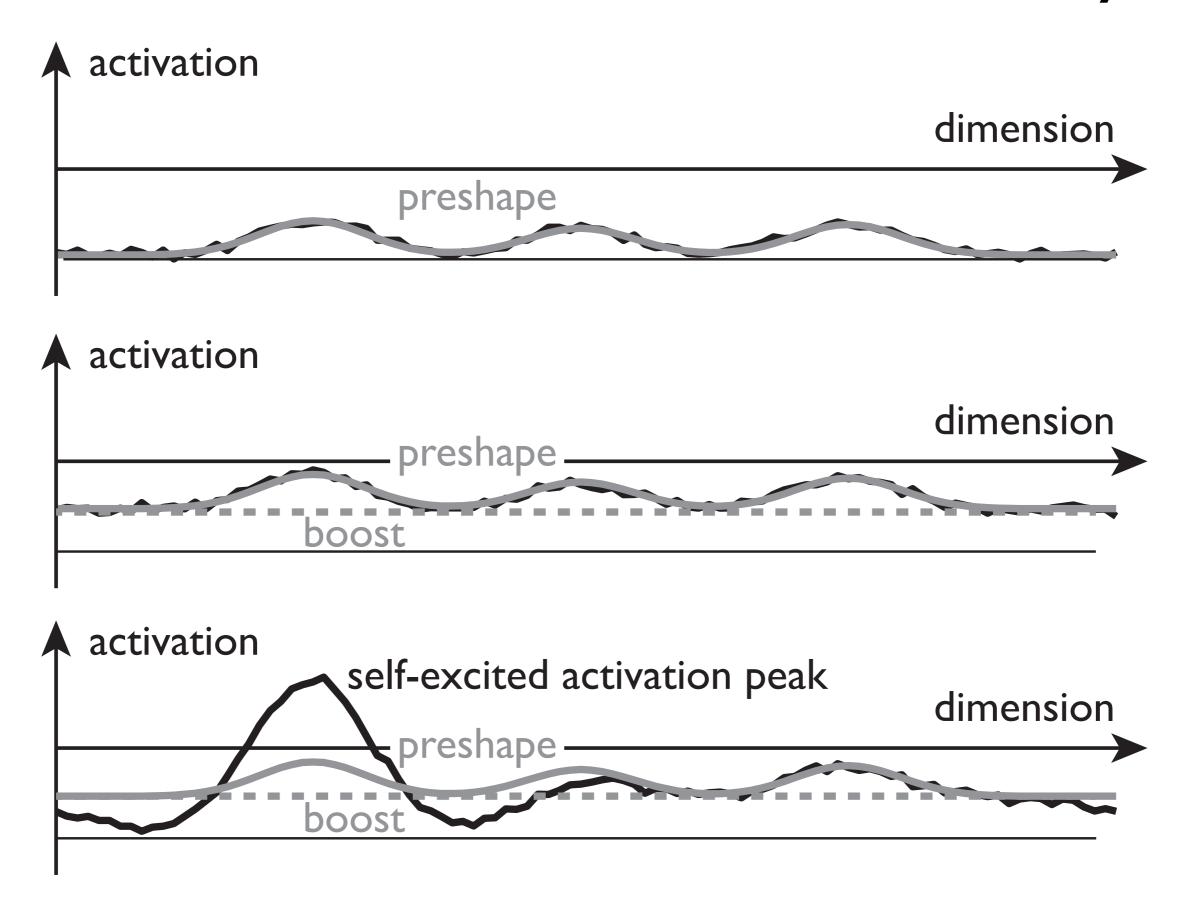
#### initiation vs. fixation

such models account for the gap-step-overlap effect



[Kopecz, 95]

# boost-induced detection instability



## boost-driven detection instability

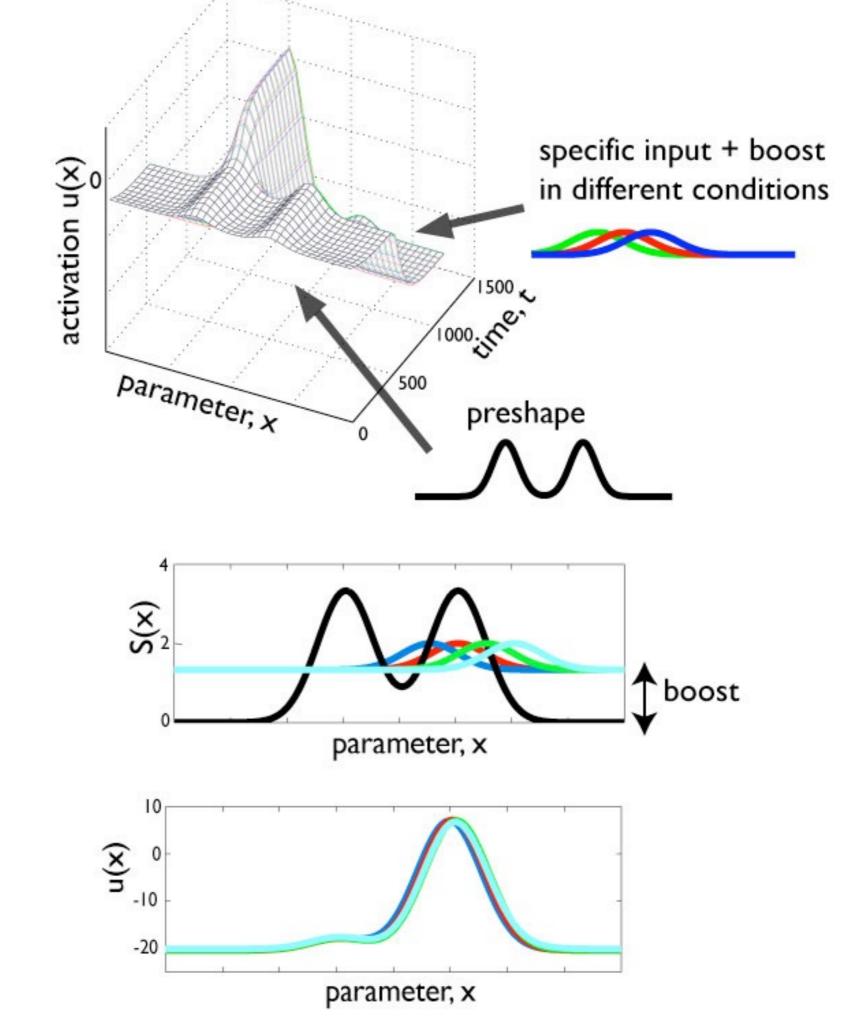
- ■inhomogeneities in the field existing prior to a signal/stimulus that leads to a macroscopic response="preshape"
- the boost-driven detection instability amplifies preshape into macroscopic selection decisions

## ... emergence of categories?

if we understand, how such inhomogeneities come about, we understand the emergence of categories...

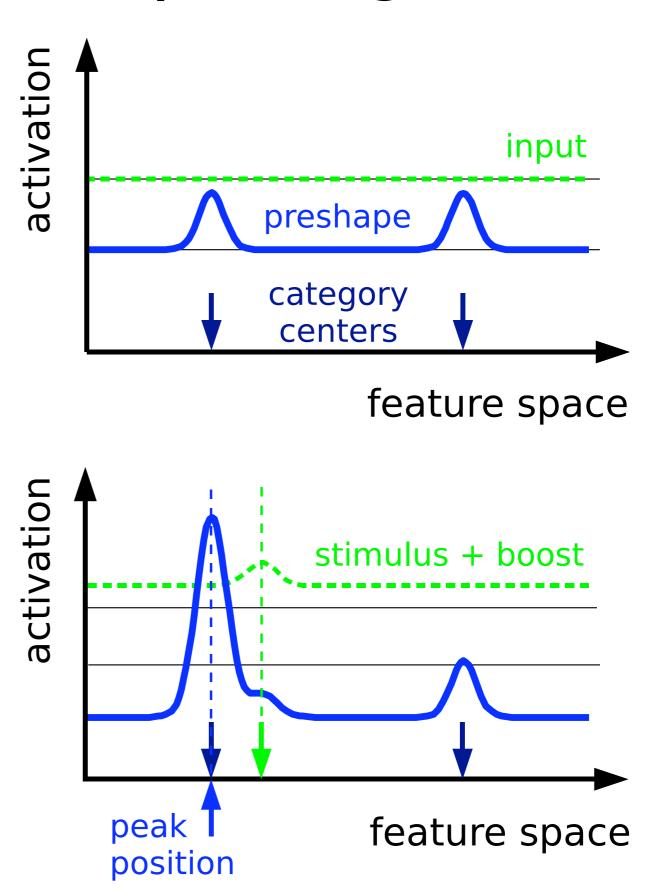
# this supports categorical behavior

when preshape dominates



### categorical responding

 based on categorical memory trace and boost-driven detection instability



#### distance effect

common in categorical tasks... e.g., decide which of two sticks is longer => RT is larger when sticks are more similar in length (1930s')

### interaction metrics-probability

- opposite to that predicted for input-driven detection instabilities:
- metrically close choices show larger effect of probability

