

Embodied Neural Dynamics

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The dynamics activation fields

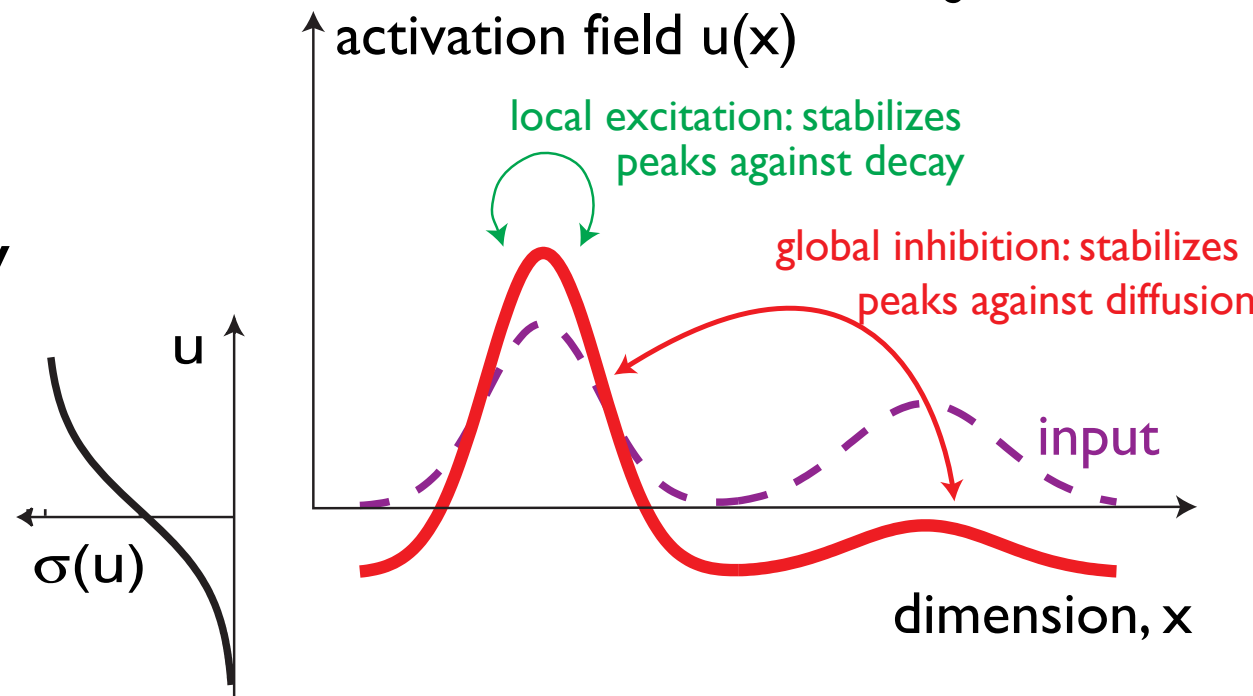
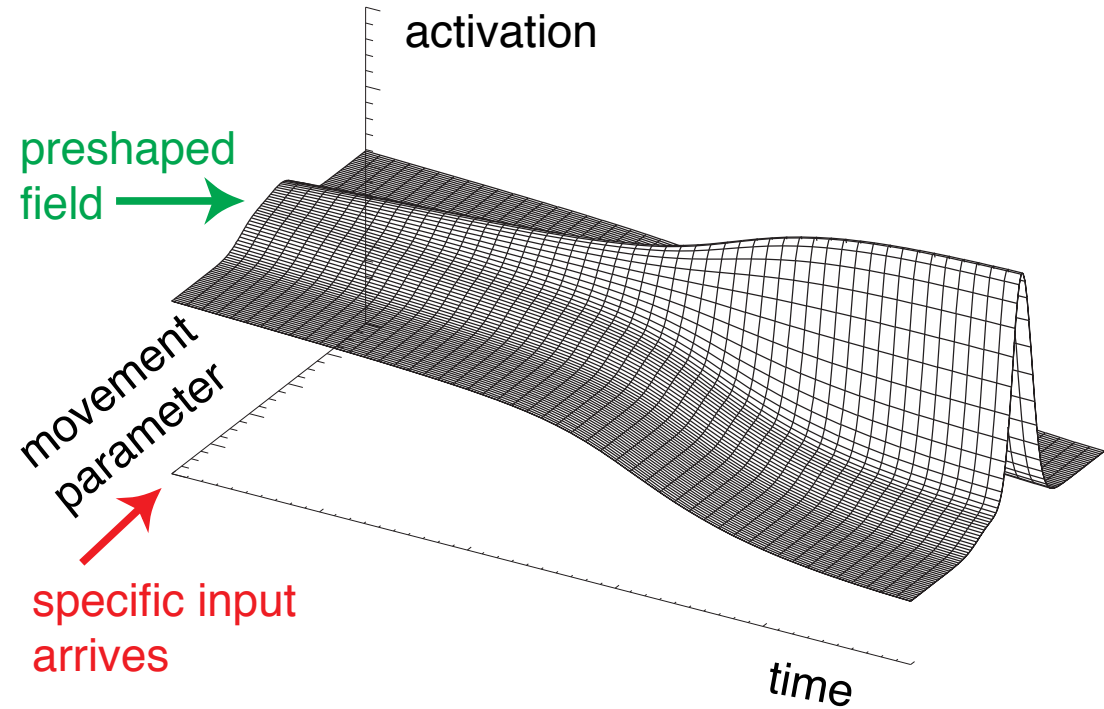
- field dynamics combines input

- with strong interaction:

 - local excitation

 - global inhibition

- \Rightarrow generates stability of peaks



=>

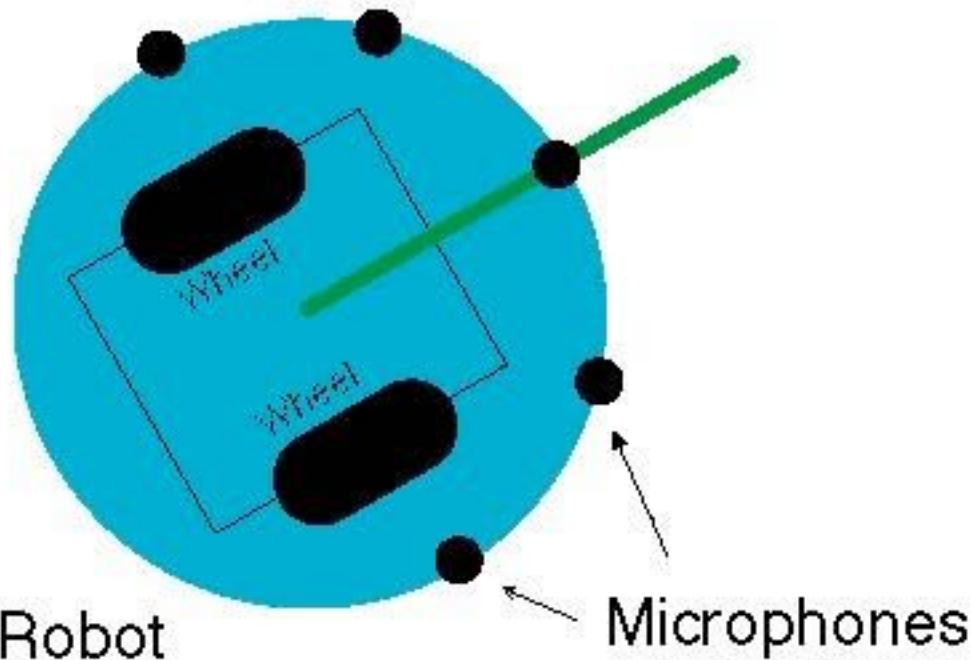
■ attractor states

- input driven solution (sub-threshold)
- self-stabilized solution (peak, supra-threshold)

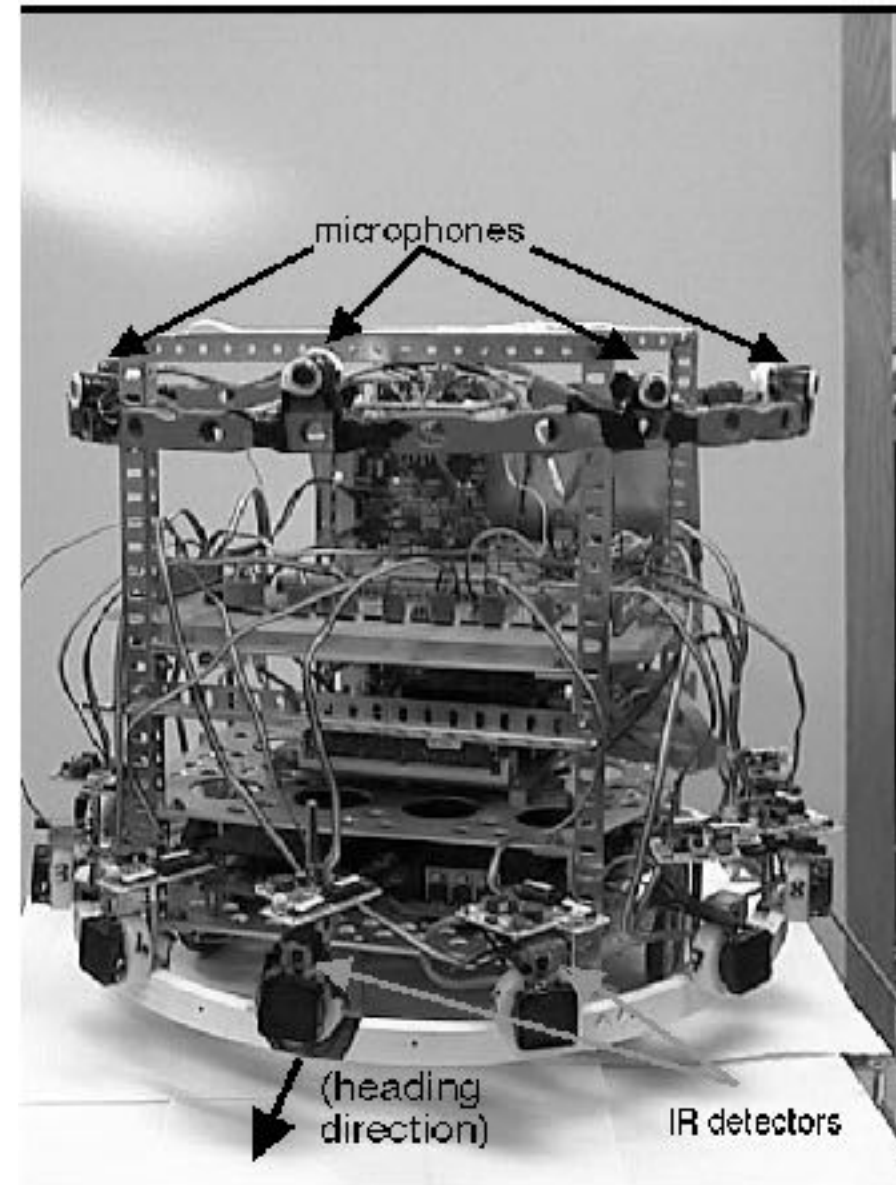
■ instabilities

- detection instability (from localize input or boost)
- reverse detection instability
- selection instability
- memory instability

Illustration: linking to sensors

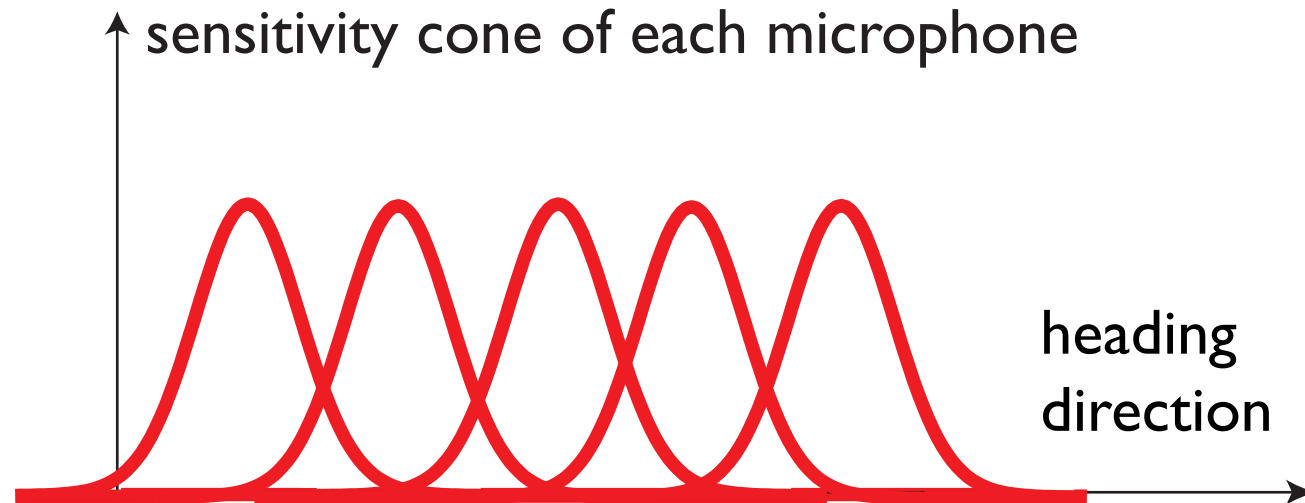


[from Bicho, Mallet, Schöner, Int J Rob Res, 2000]

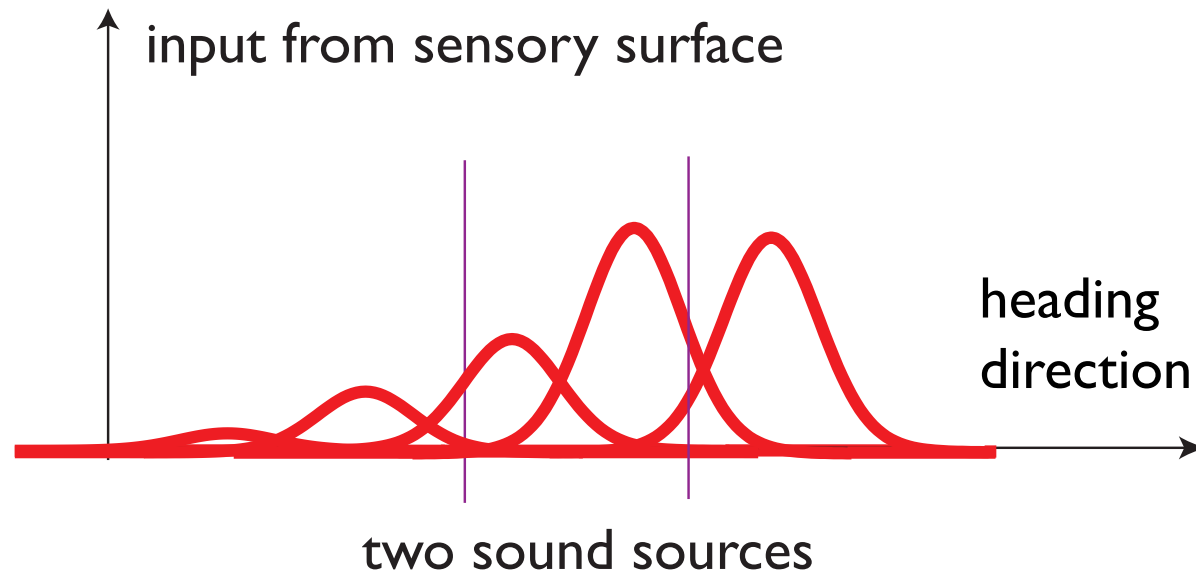
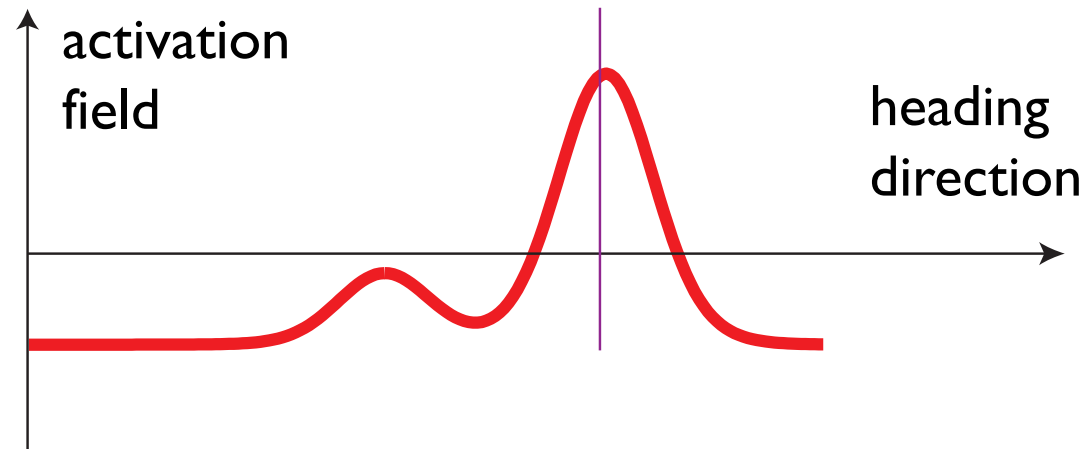


Sensory surface

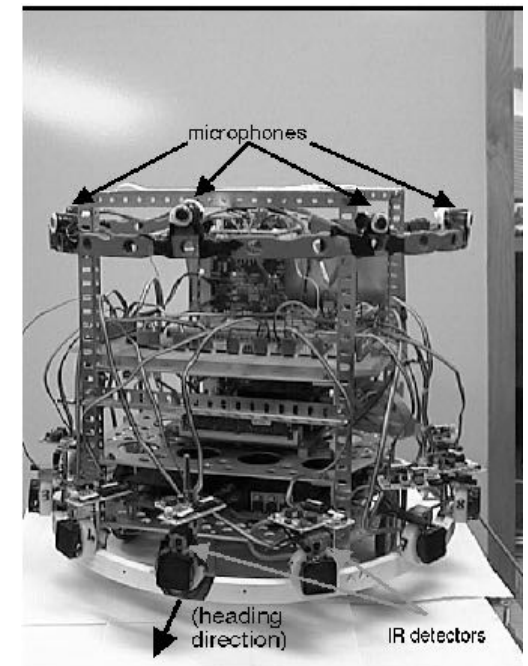
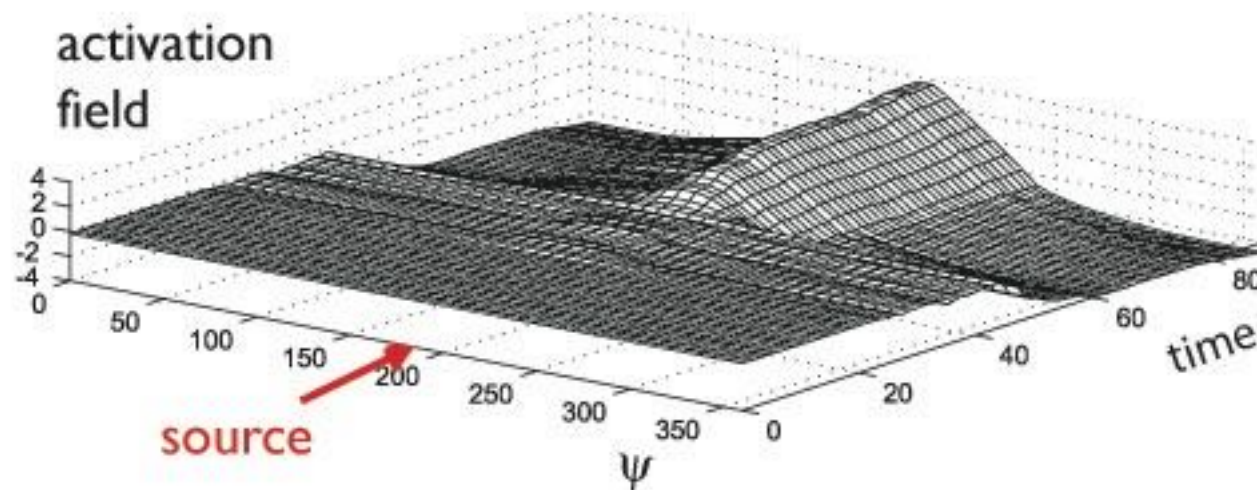
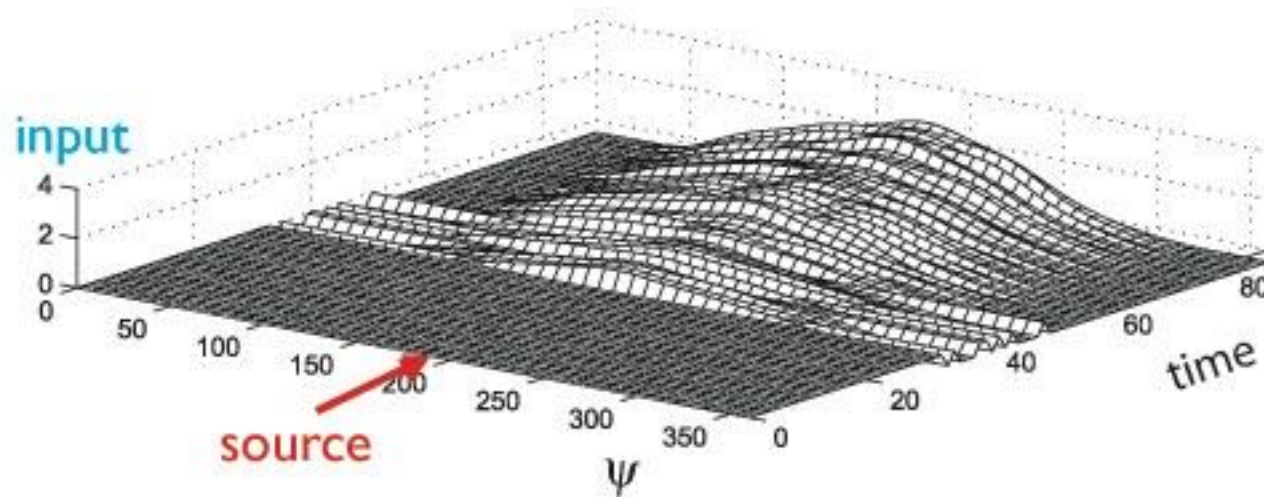
- each microphone samples heading direction



- each microphone provides input to the field

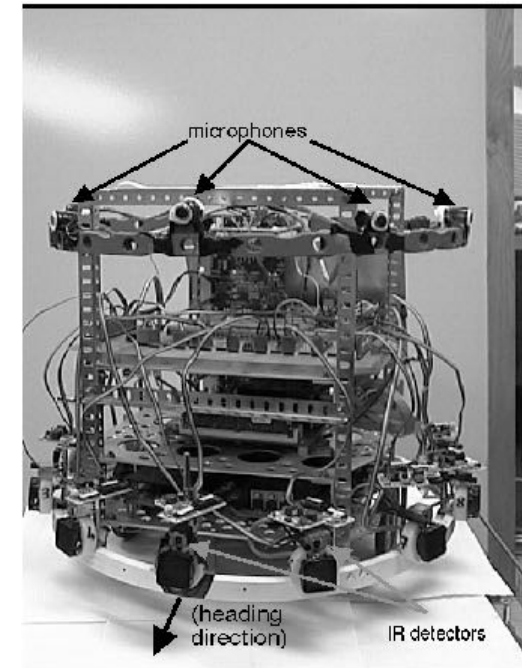
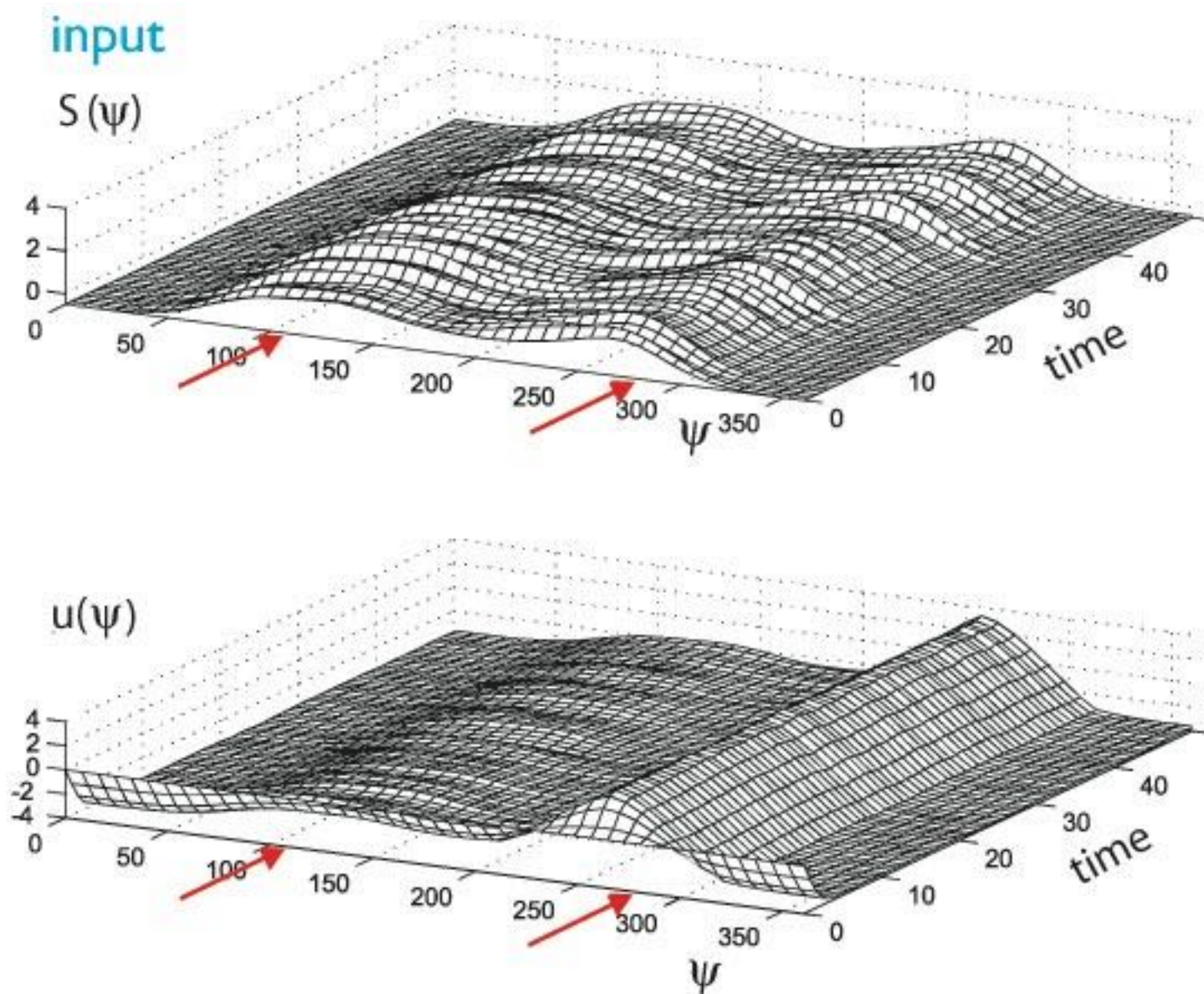


Detection instability induced by increasing intensity of sound source

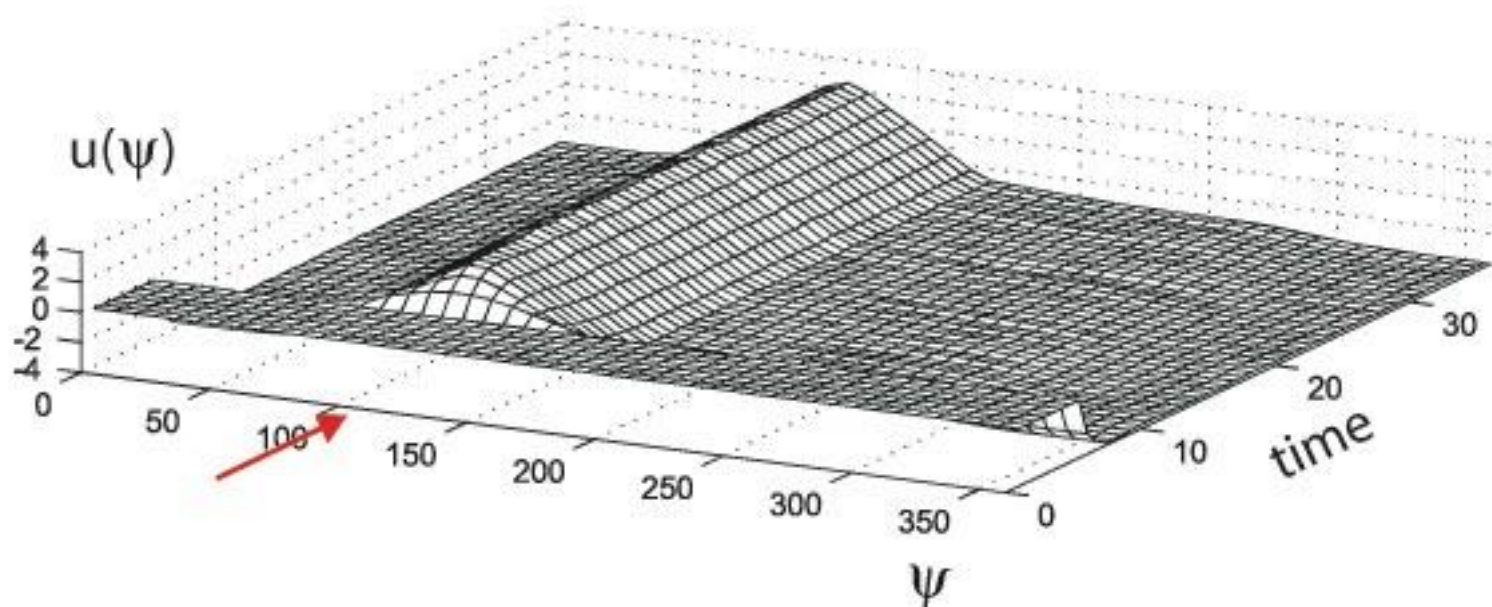
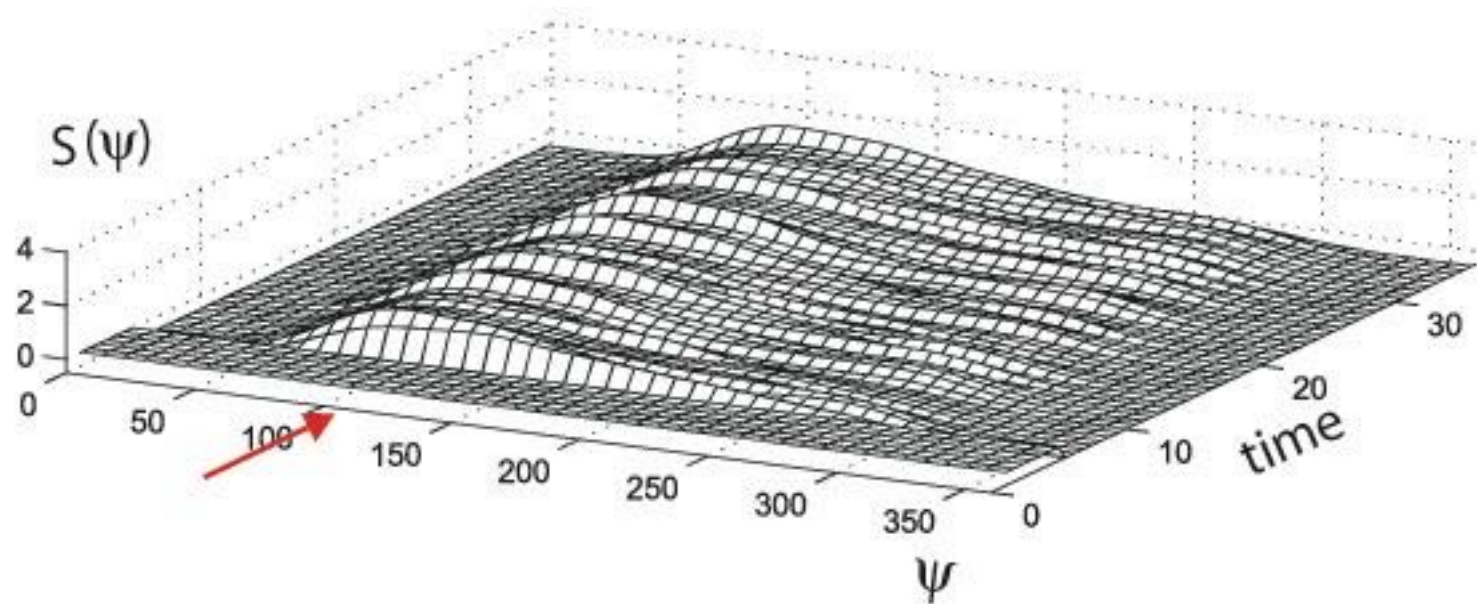


[from Bicho, Mallet, Schöner: Int. J. Rob. Res., 2000]

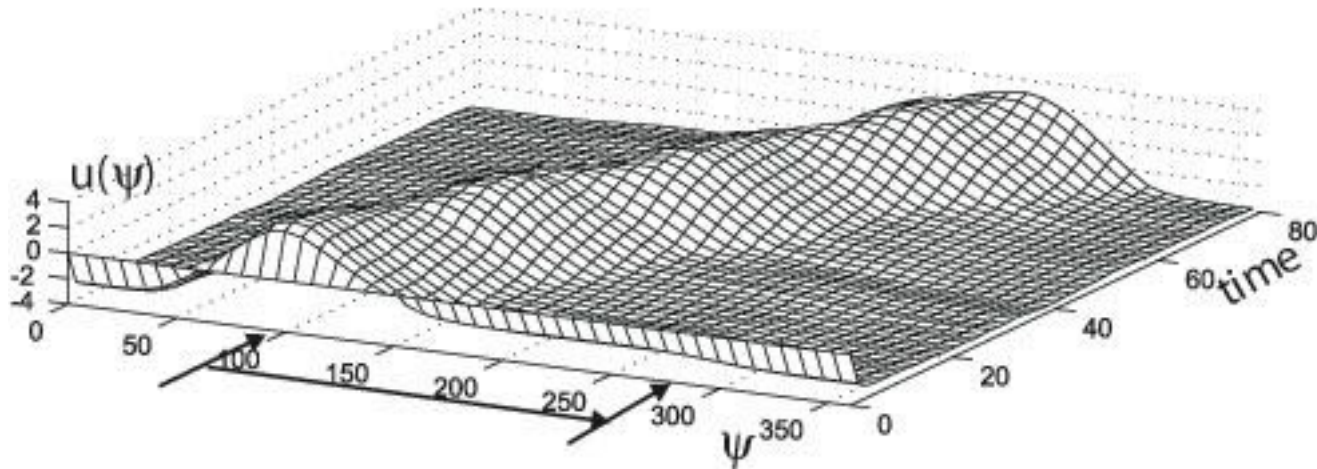
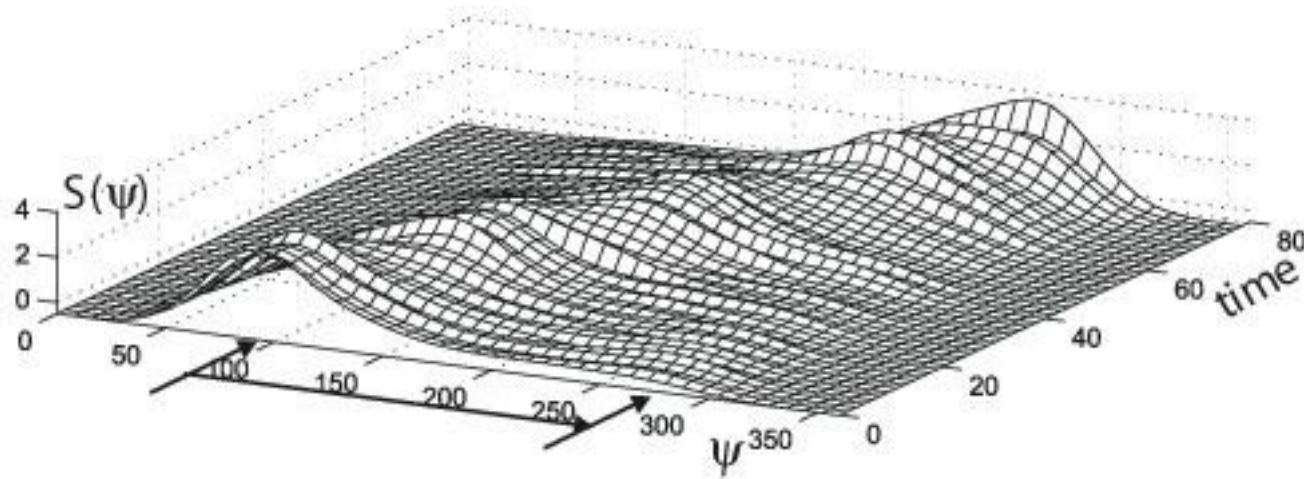
Target selection in the presence of two sources



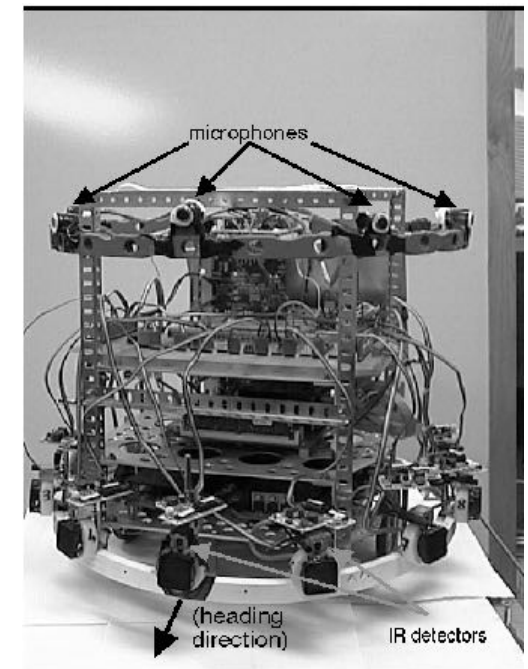
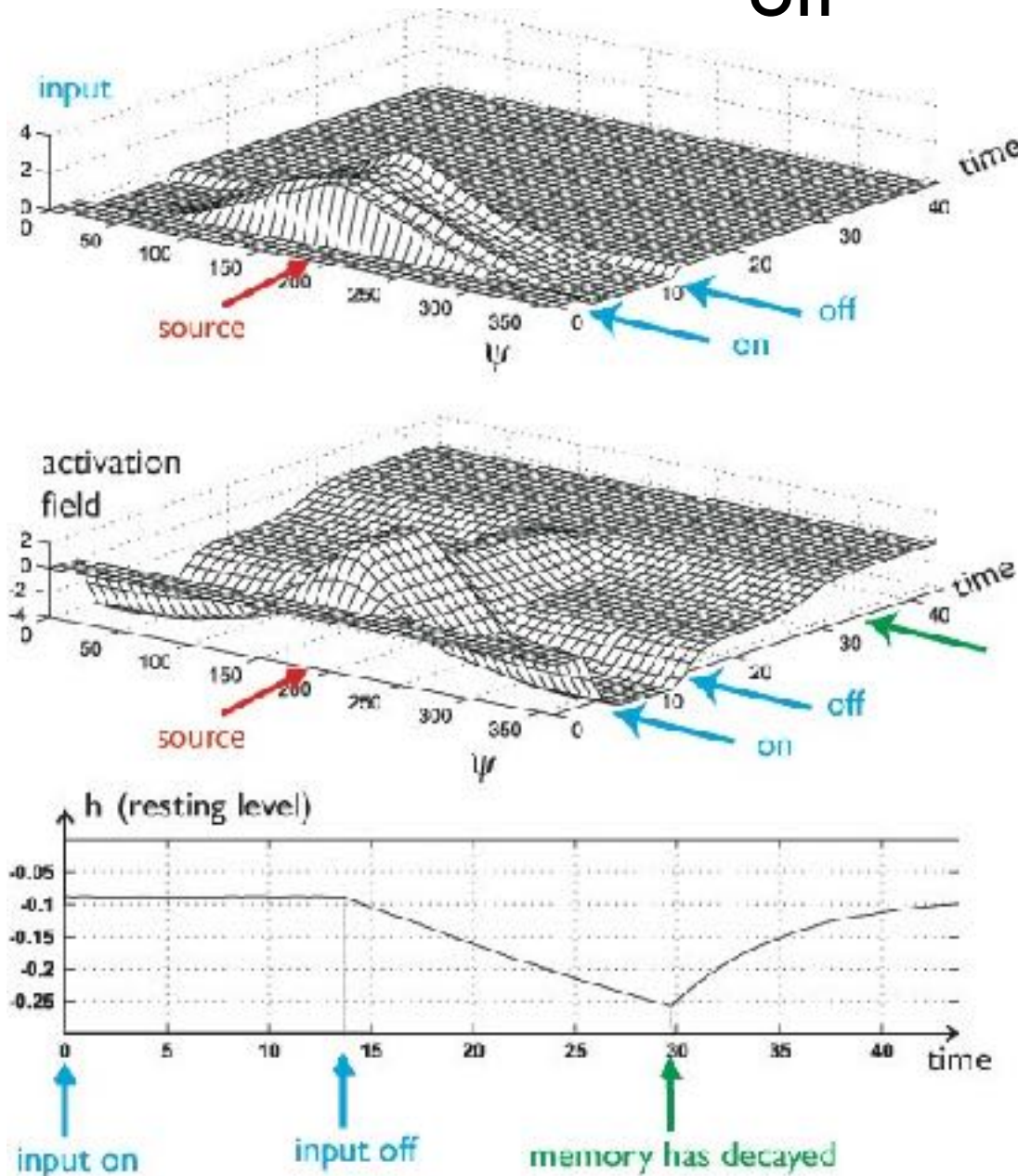
Robust estimation in the presence of outliers



Tracking when sound source moves



Memory (and forgetting) when sound source is turned off



[from Bicho, Mallet, Schöner: Int J Rob Res 19:424(2000)]

Illustration of instabilities

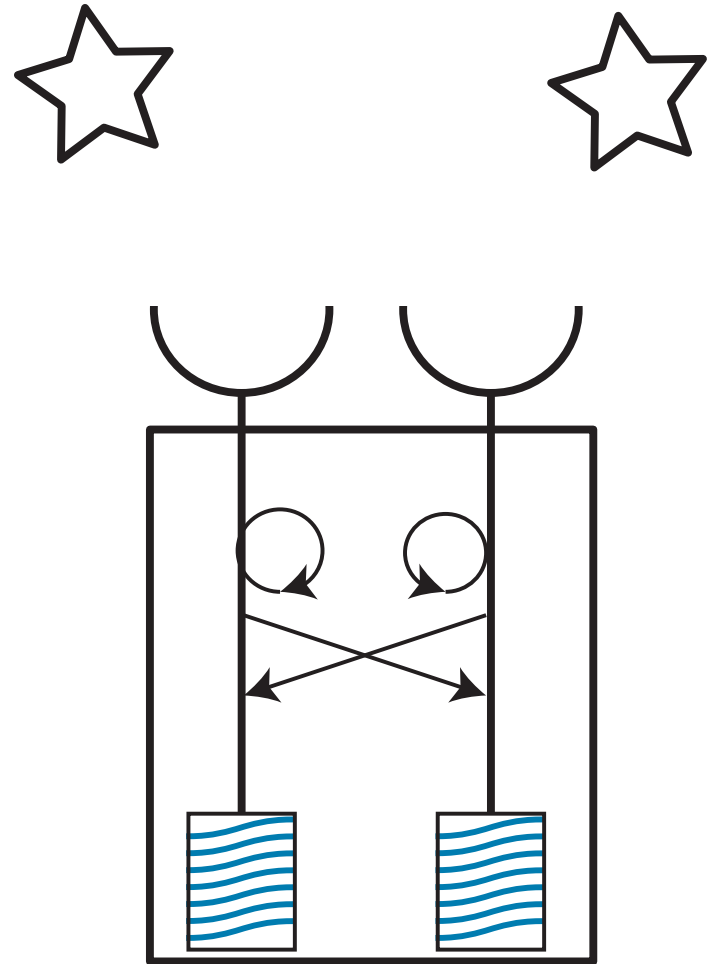


Motor behavior

- so far, the neural field was in open loop: received input from sensors, but didn't drive around and thus did not influence its own sensor input

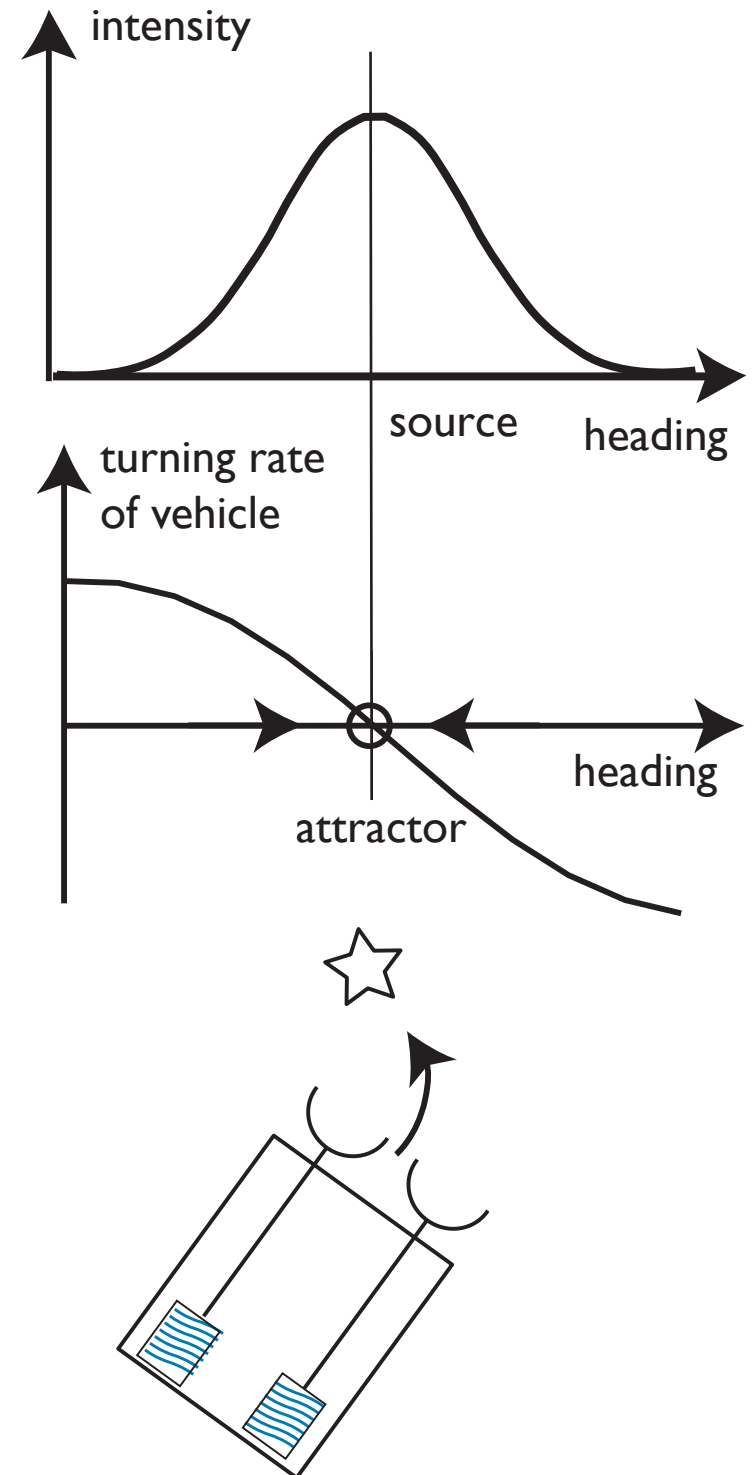
Braitenberg

■ in terms of the Braitenberg vehicle, we only looked at the “inner” neural dynamics



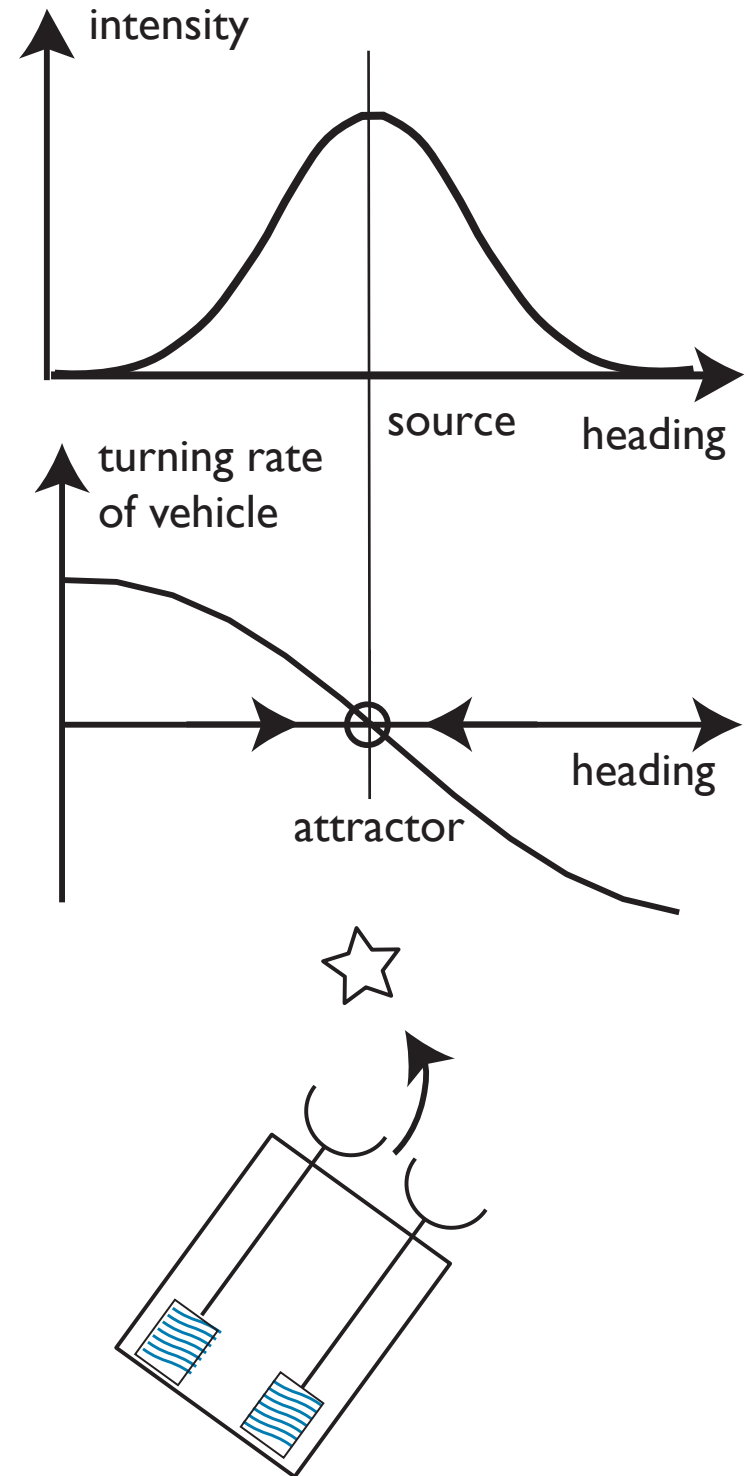
Braitenberg

- we did not yet look at the emergence of (motor) behavior given a representation of sensory information



Braitenberg

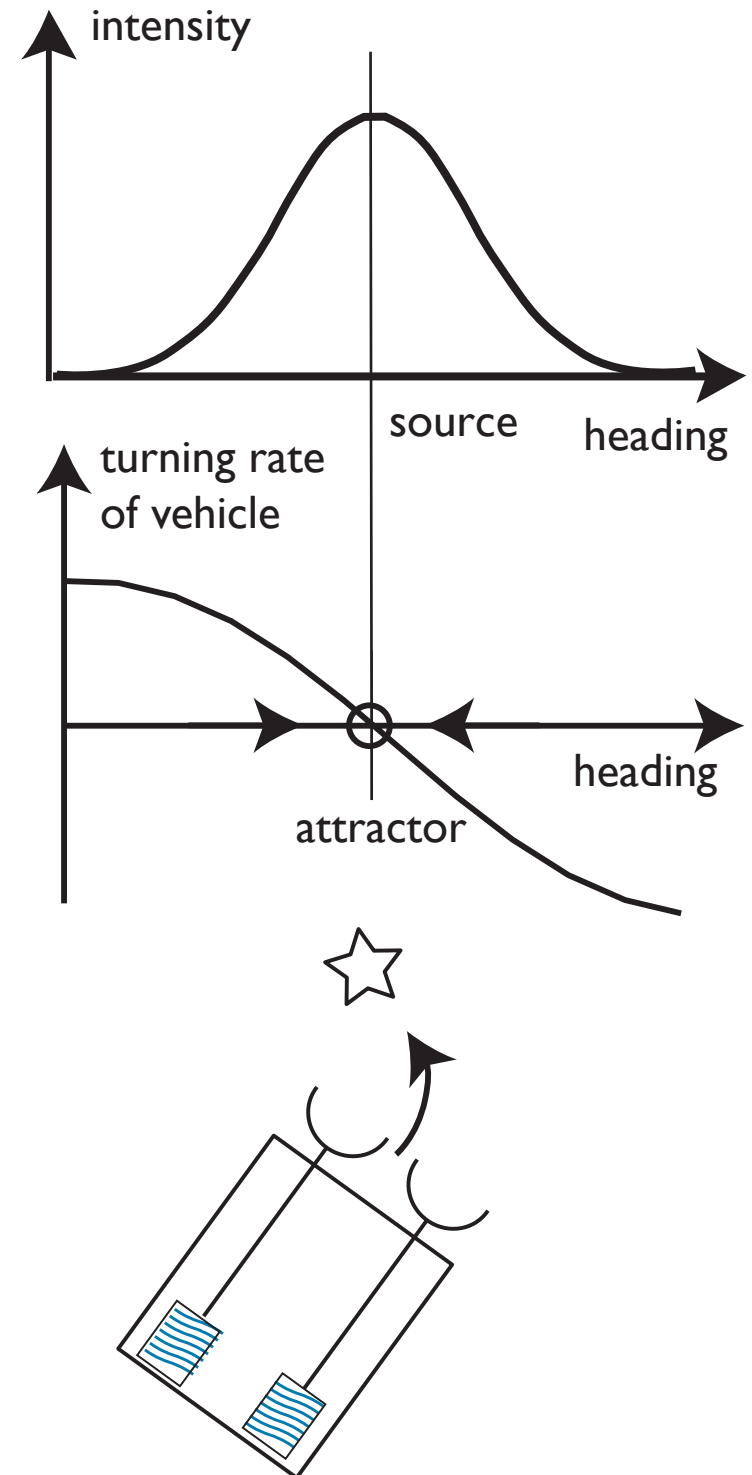
- overt movement behavior is generated by a behavioral dynamics
- how may the neural representations of DFT couple into behavioral dynamics “standing in for” sensory inputs?



Braitenberg

■ two problems

- how do we go from a field to an attractor dynamics? => space to rate code issue
- how does the field emulate “closed loop” behavior? => coordinate transforms

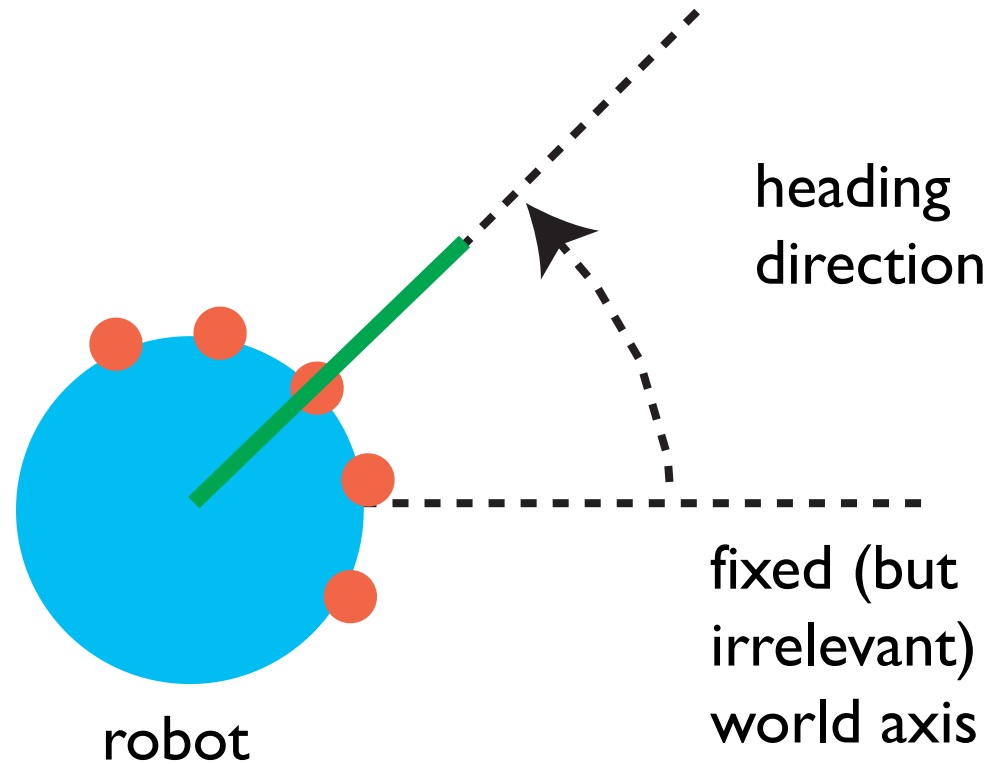


Basic ideas: behavioral dynamics

- behavioral variables
- time courses from dynamical system:
attractors
- tracking attractors
- bifurcations for flexibility

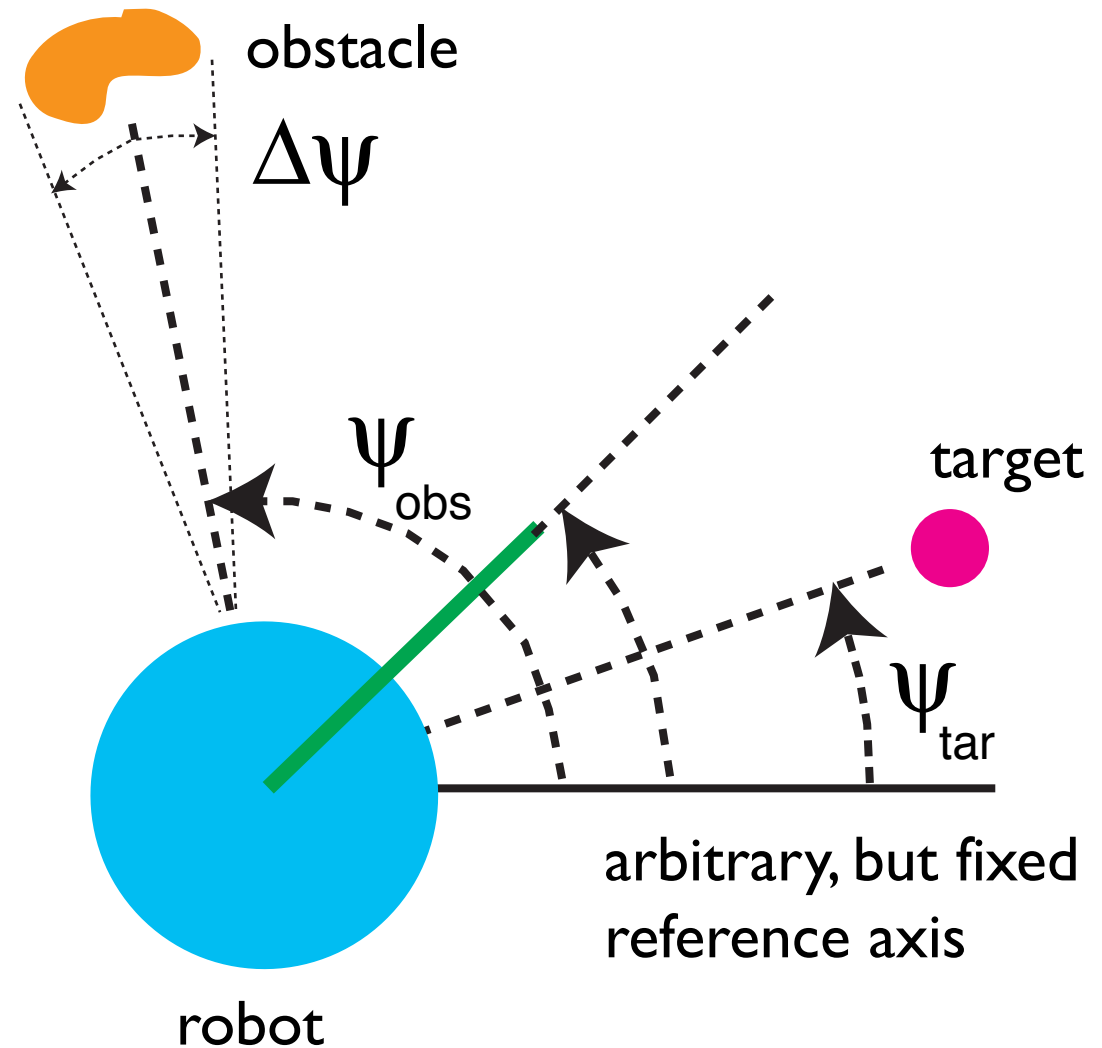
Behavioral variables: example

■ vehicle moving in
2D: heading
direction



Behavioral variables: example

- constraints:
obstacle avoidance
and target
acquisition



Behavioral variables

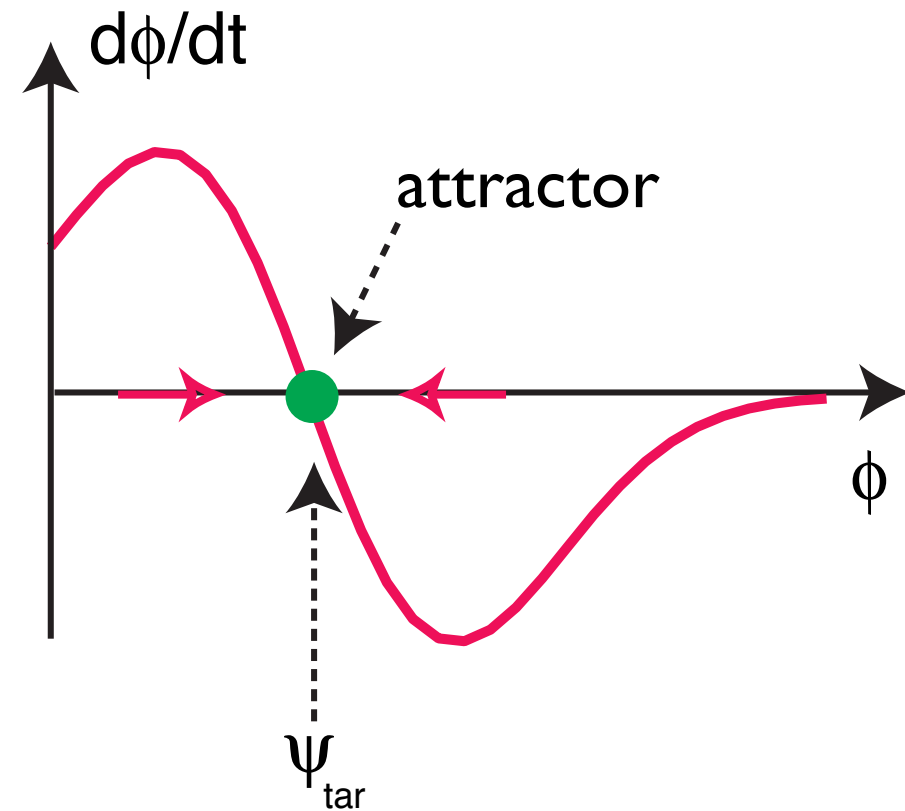
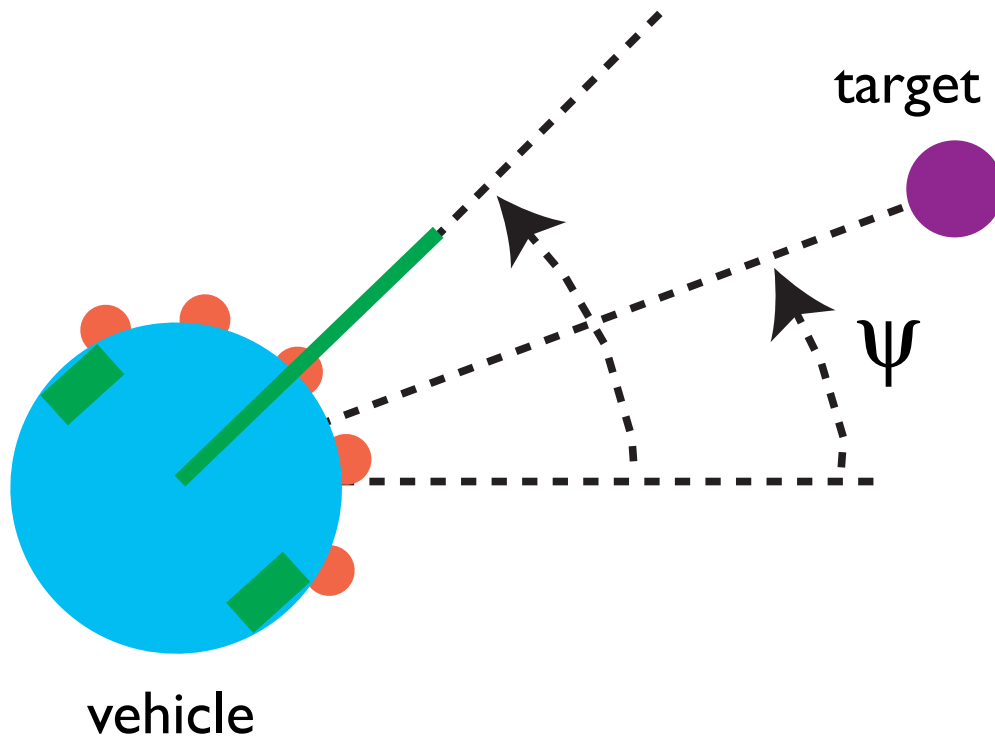
- describe desired motor behavior
- “enactable”
- express constraints as values/value ranges
- appropriate level of invariance

Behavioral dynamics

- generate behavior by generating time courses of behavioral variables
- generate time course of behavioral variables from attractor solutions of a (designed) dynamical system
- that dynamical system is constructed from contributions expressing behavioral constraints

Behavioral dynamics: example

■ behavioral constraint: target acquisition



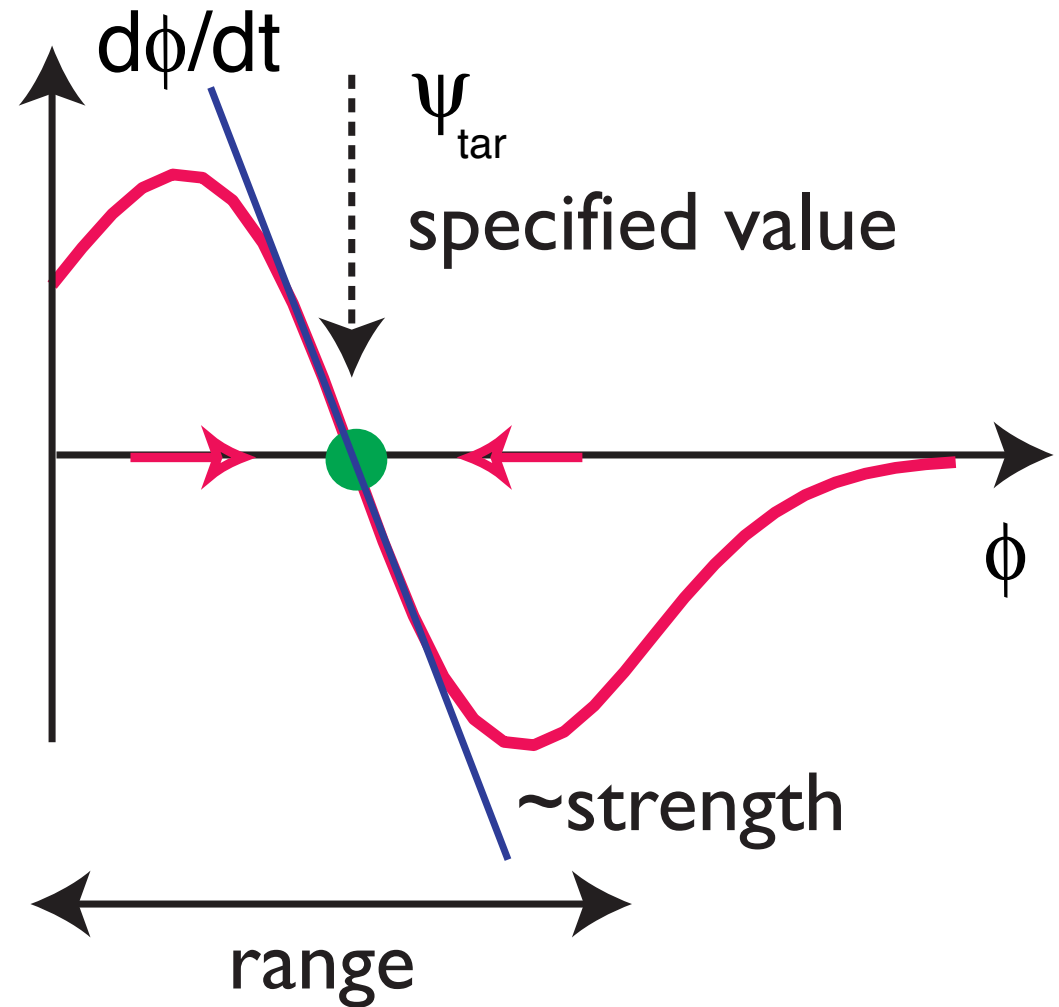
Behavioral dynamics

■ each contribution is a “force-let” with

■ specified value

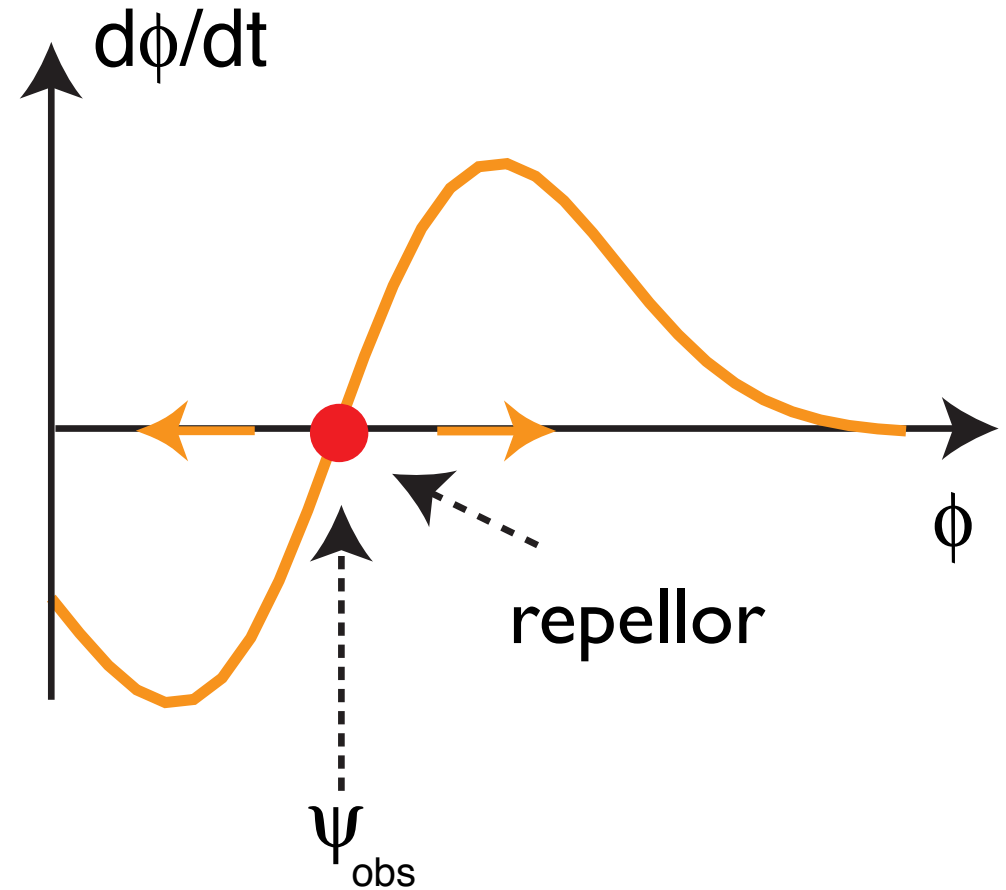
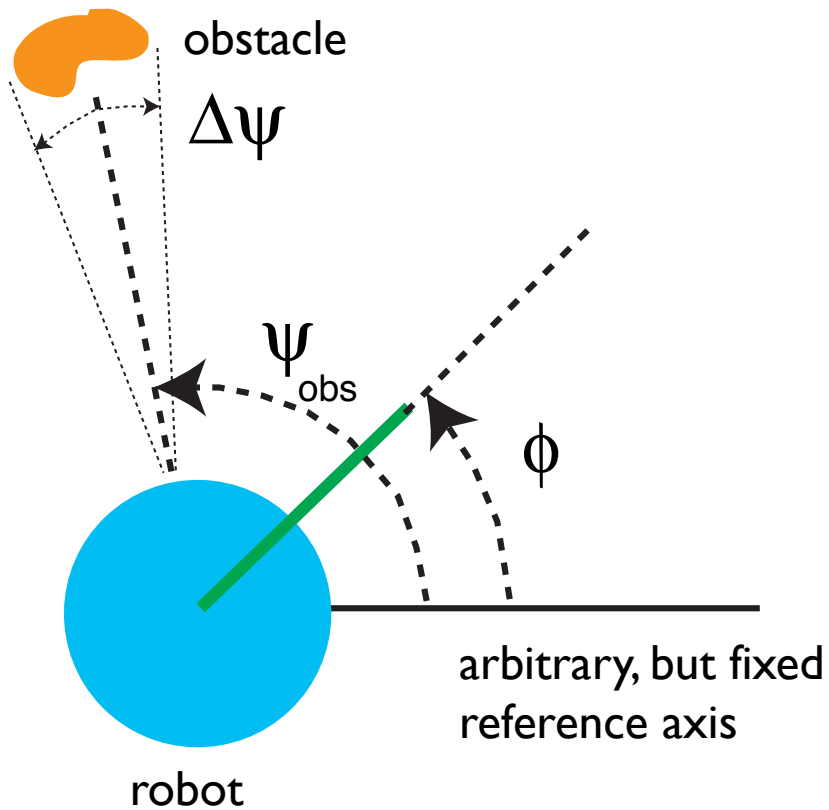
■ strength

■ range



Behavioral dynamics: example

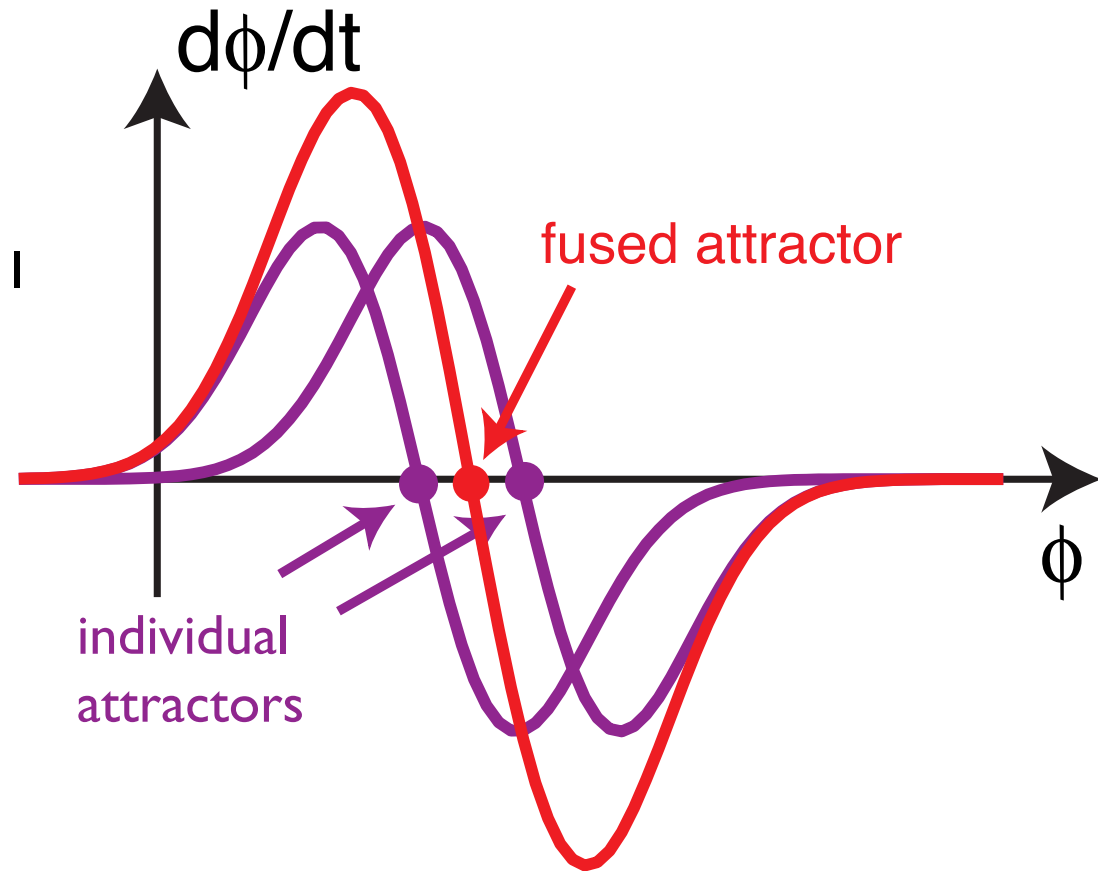
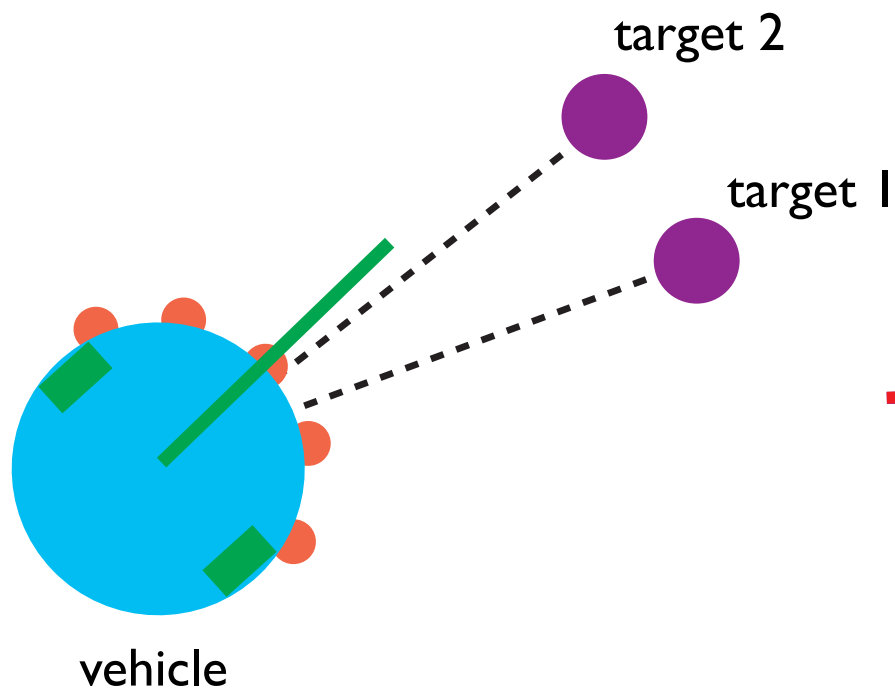
■ behavioral constraint: obstacle avoidance



Behavioral dynamics

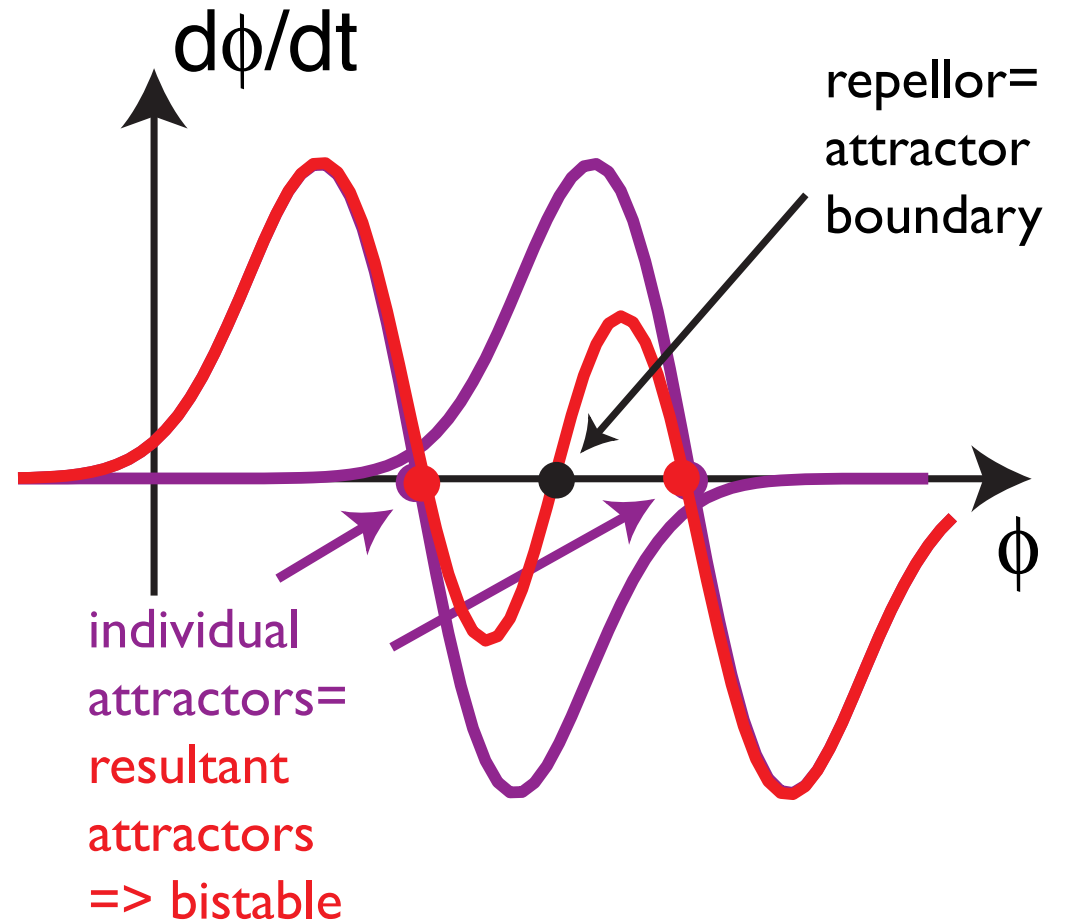
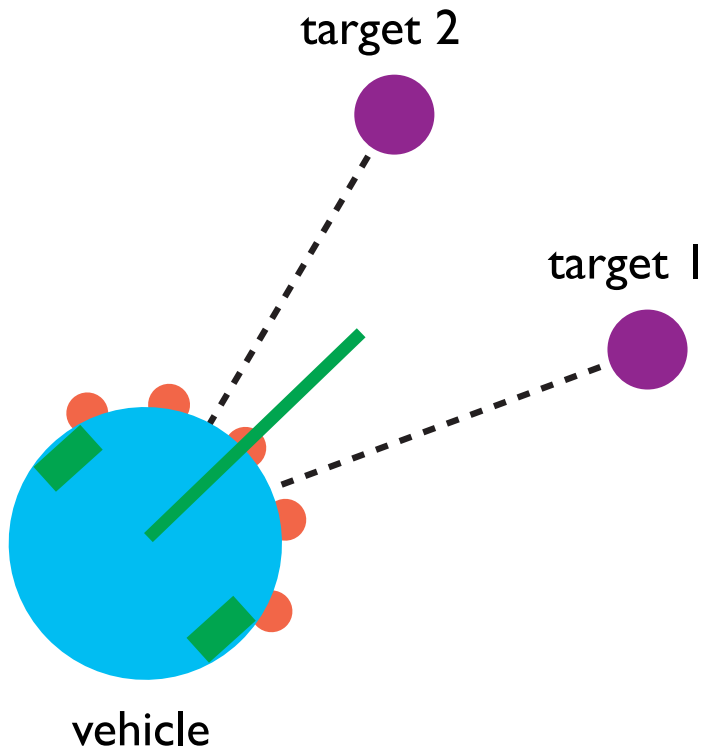
■ multiple constraints: superpose “force-lets”

■ fusion



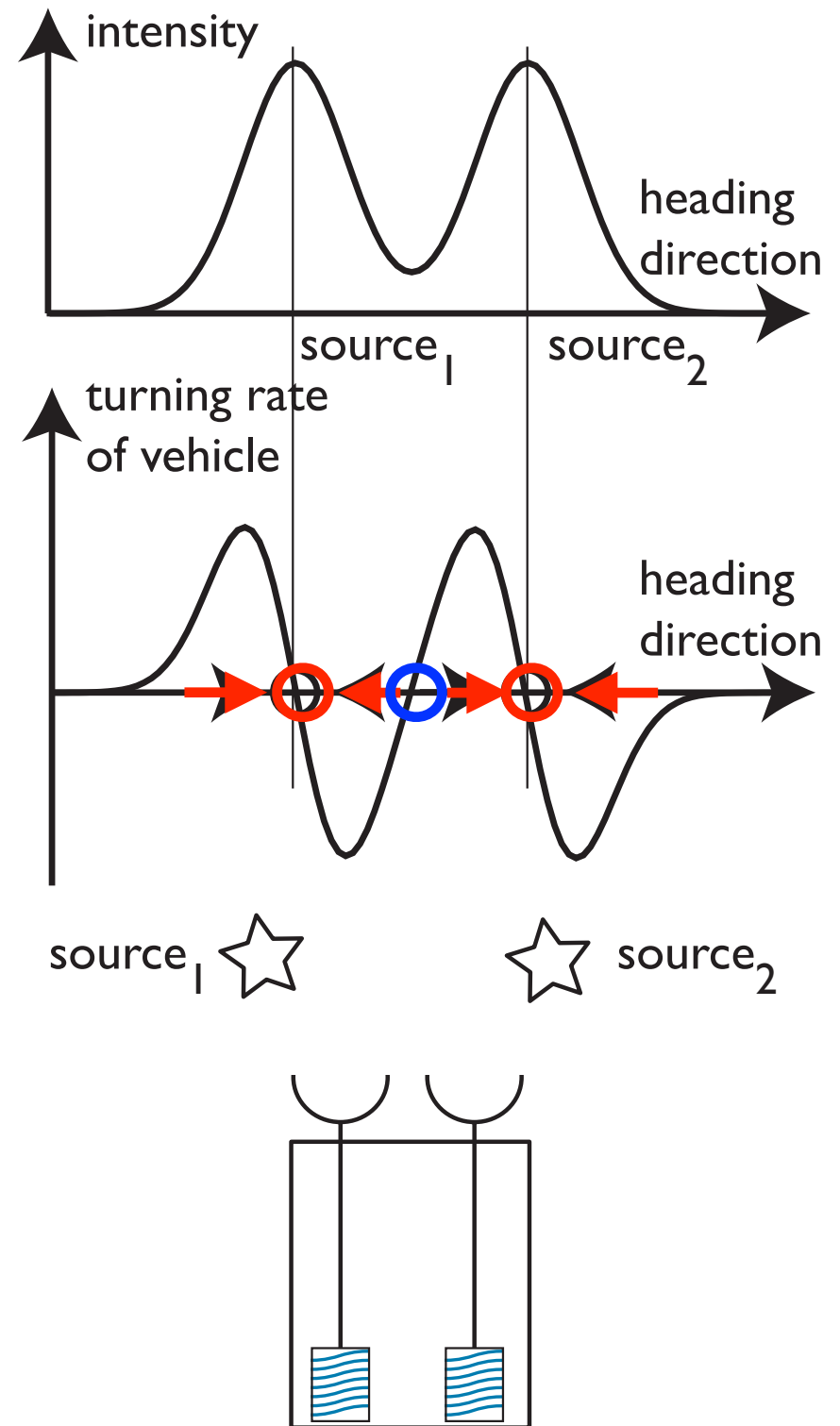
Behavioral dynamics

■ decision making



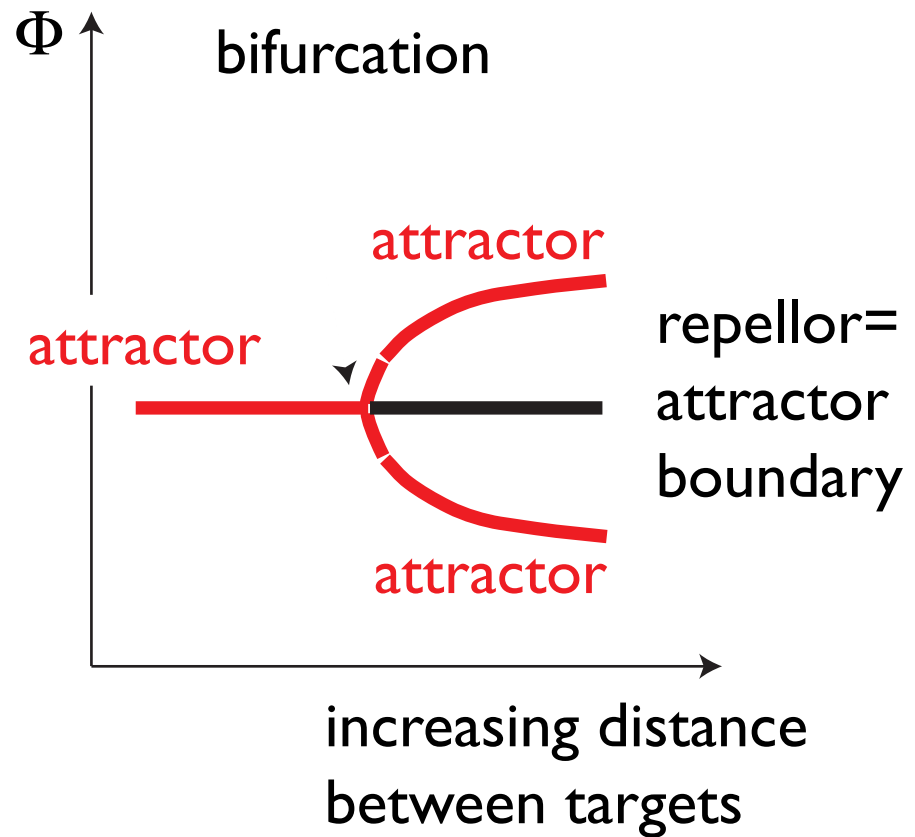
Braitenberg

- bistable dynamics for bimodal intensity distribution
- \Rightarrow nonlinear dynamics makes selection decision



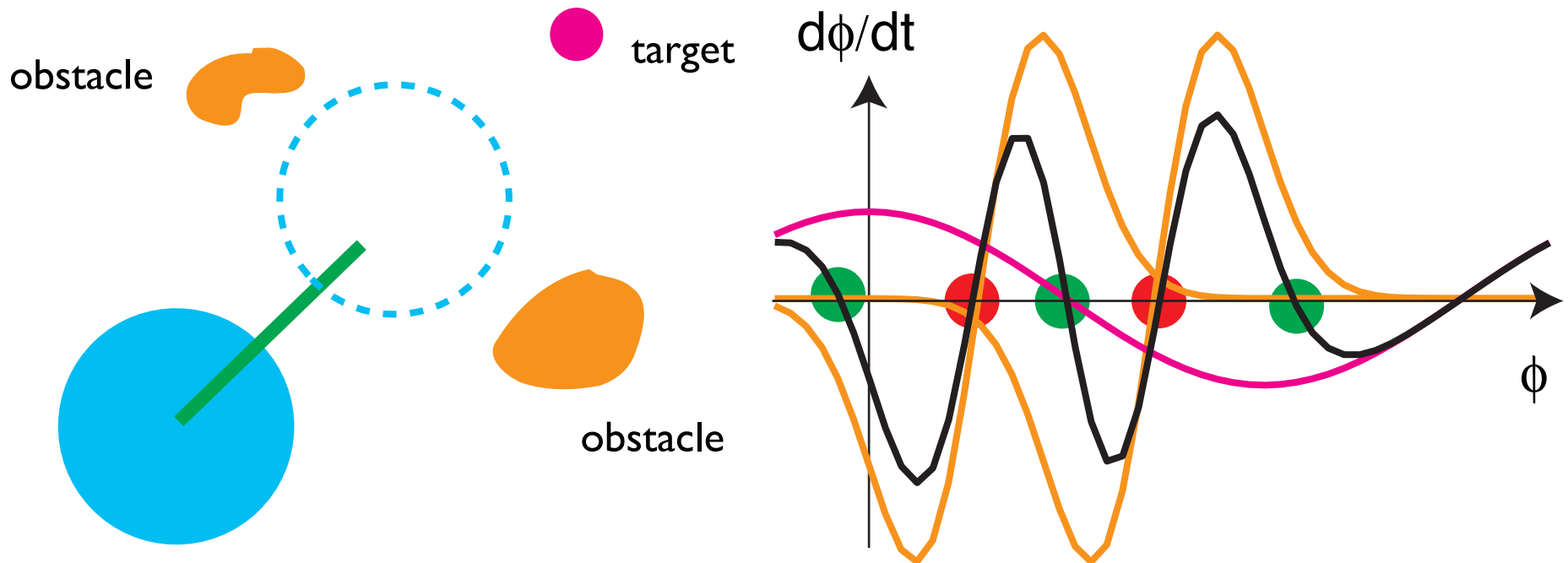
Behavioral dynamics

- Bifurcations switch between fusion and decision making



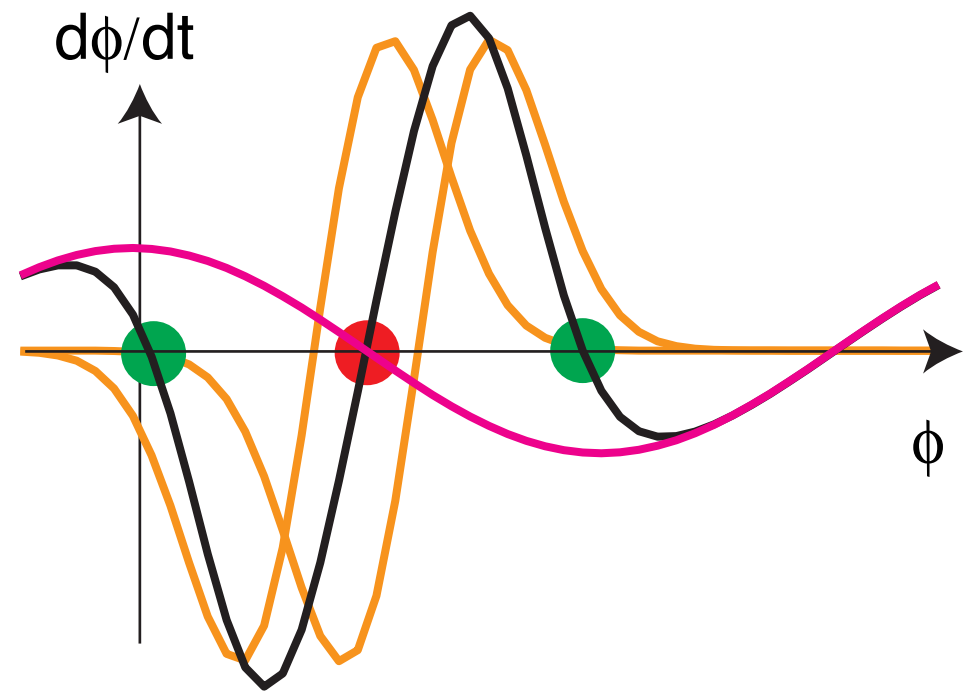
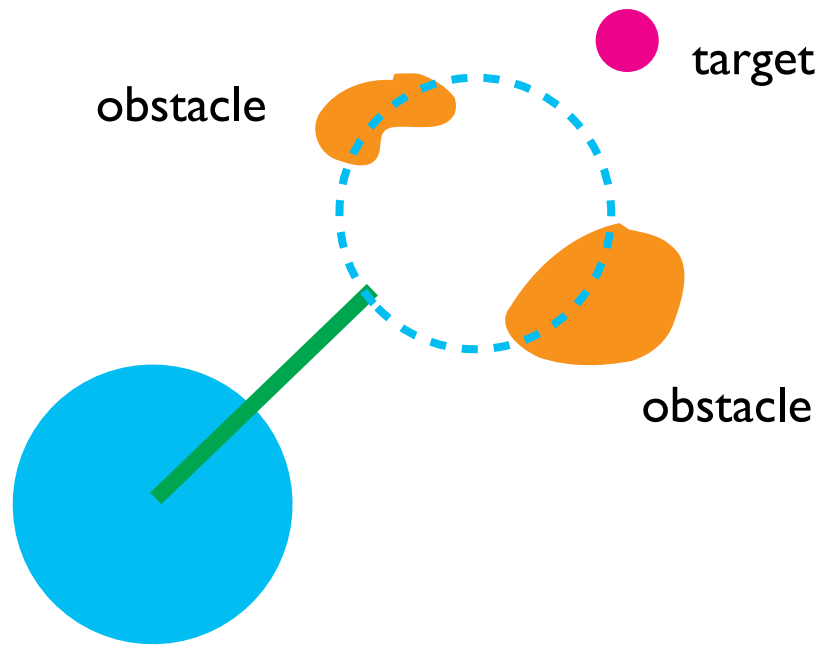
Behavioral dynamics

- an example closer to “real life”: bifurcations in obstacle avoidance and target acquisition
- constraints not in conflict



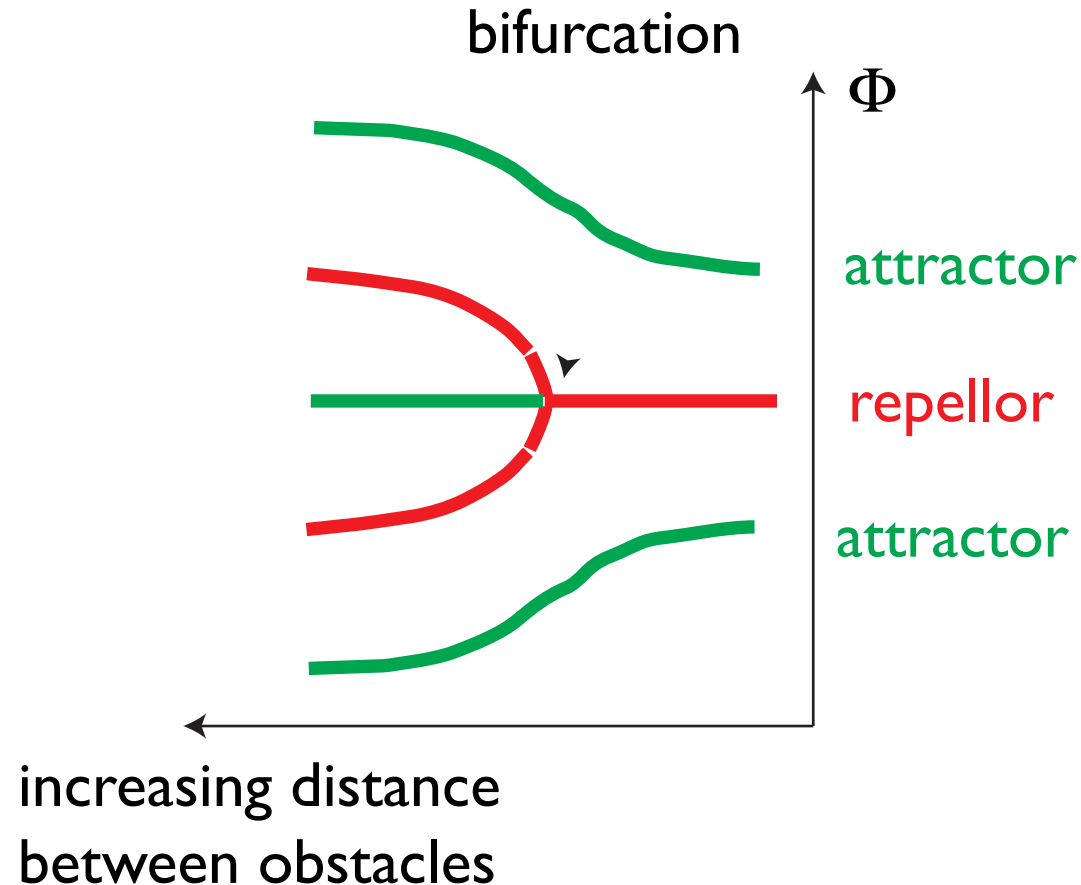
Behavioral dynamics

■ constraints in conflict

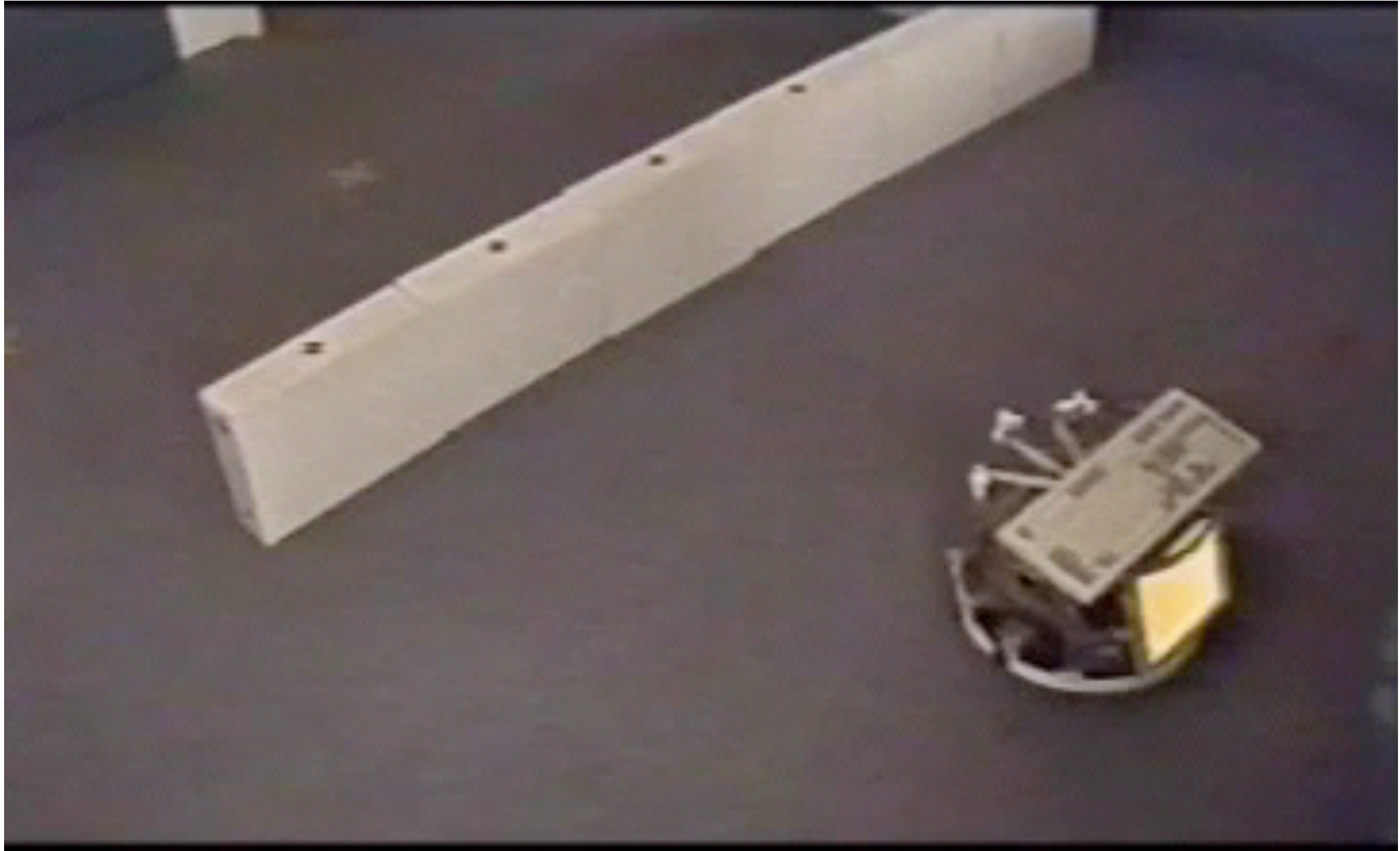


Behavioral dynamics

- transition from “constraints not in conflict” to “constraints in conflict” is a bifurcation

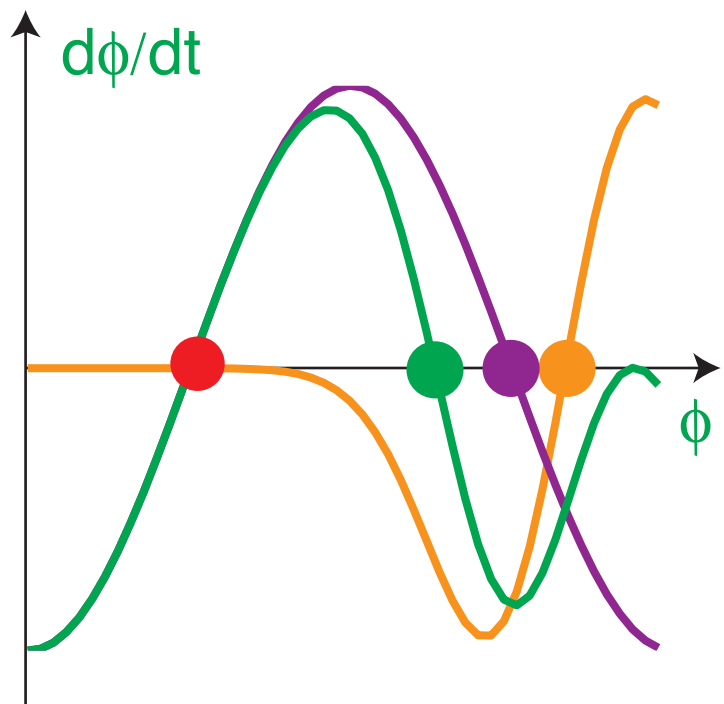
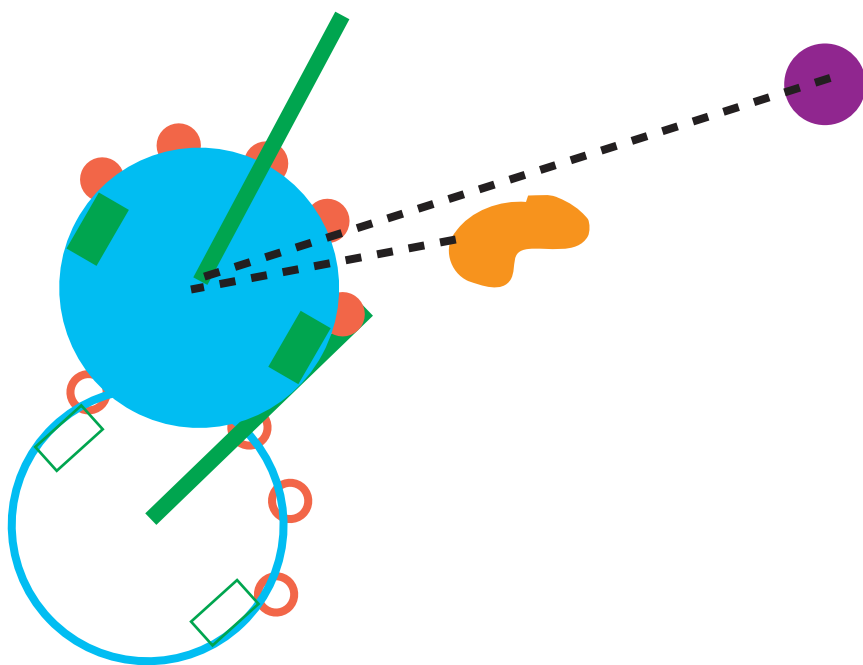
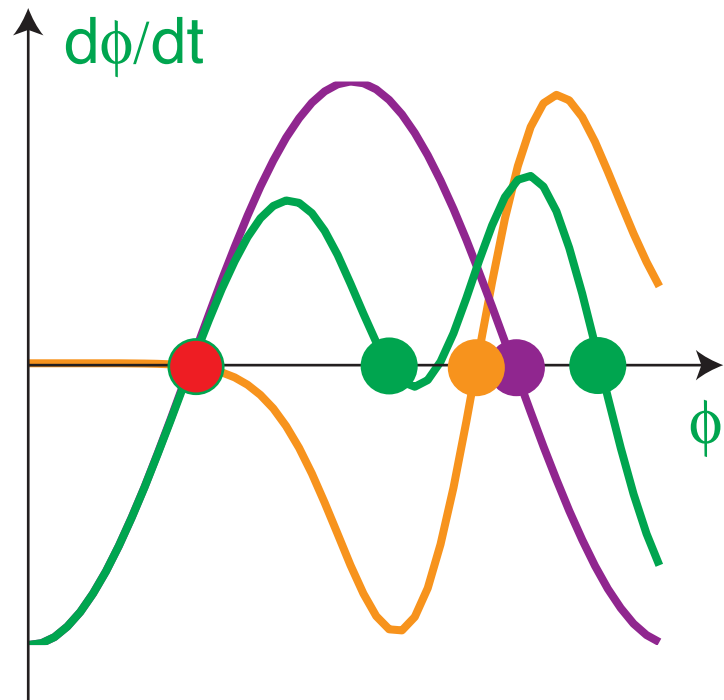
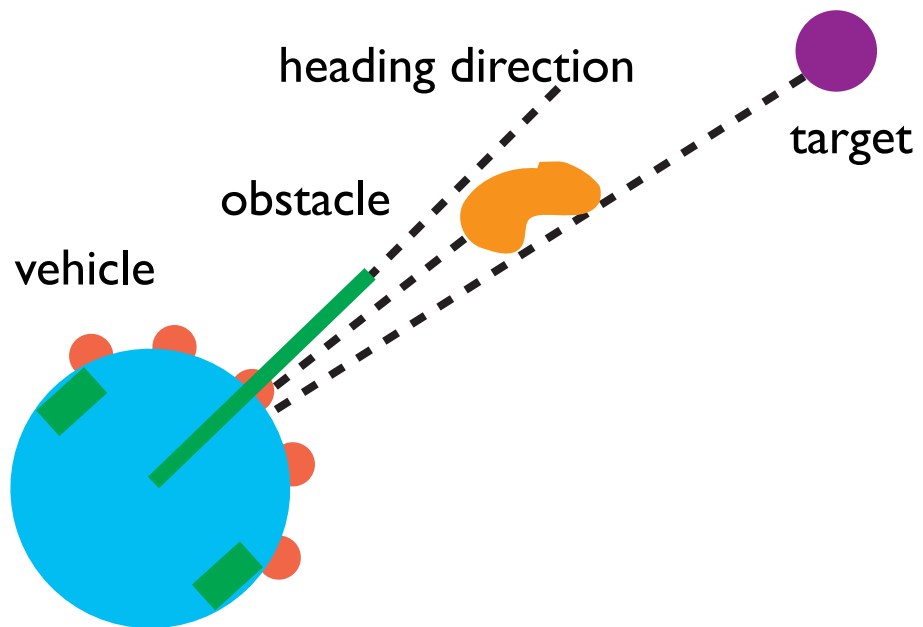


Bifurcations

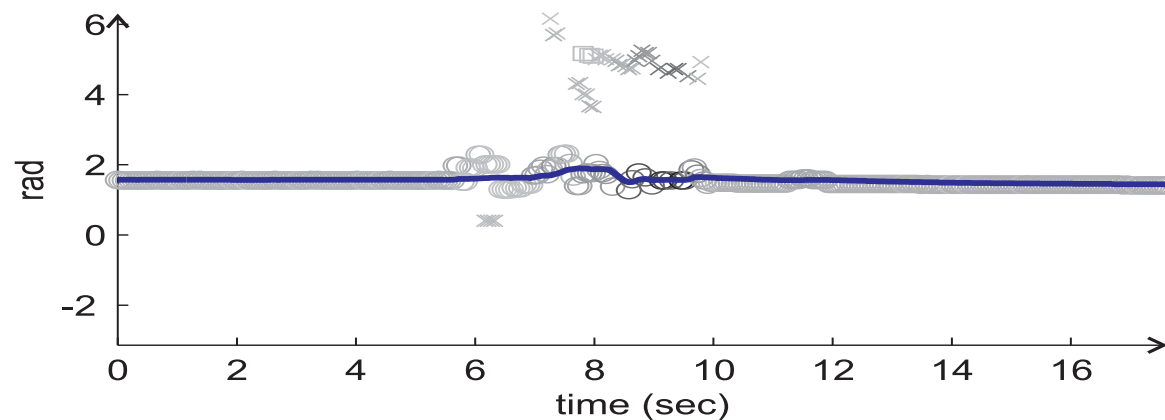
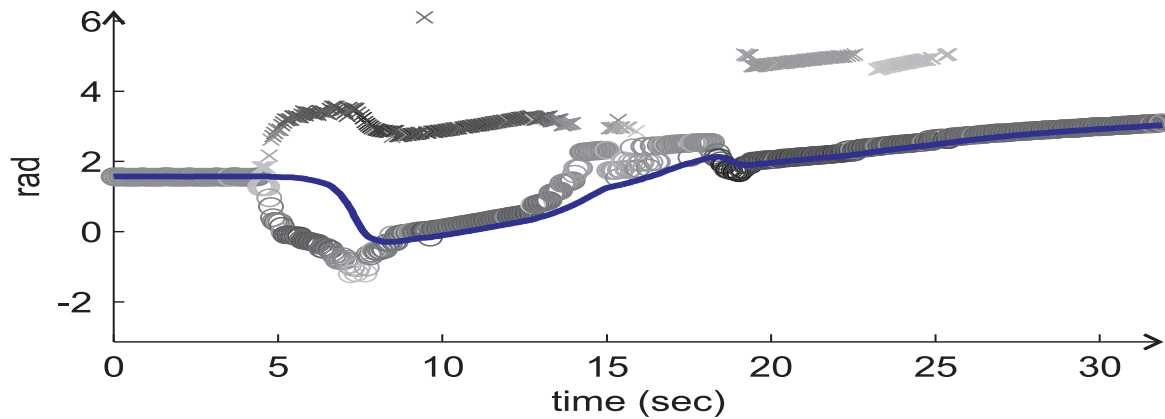
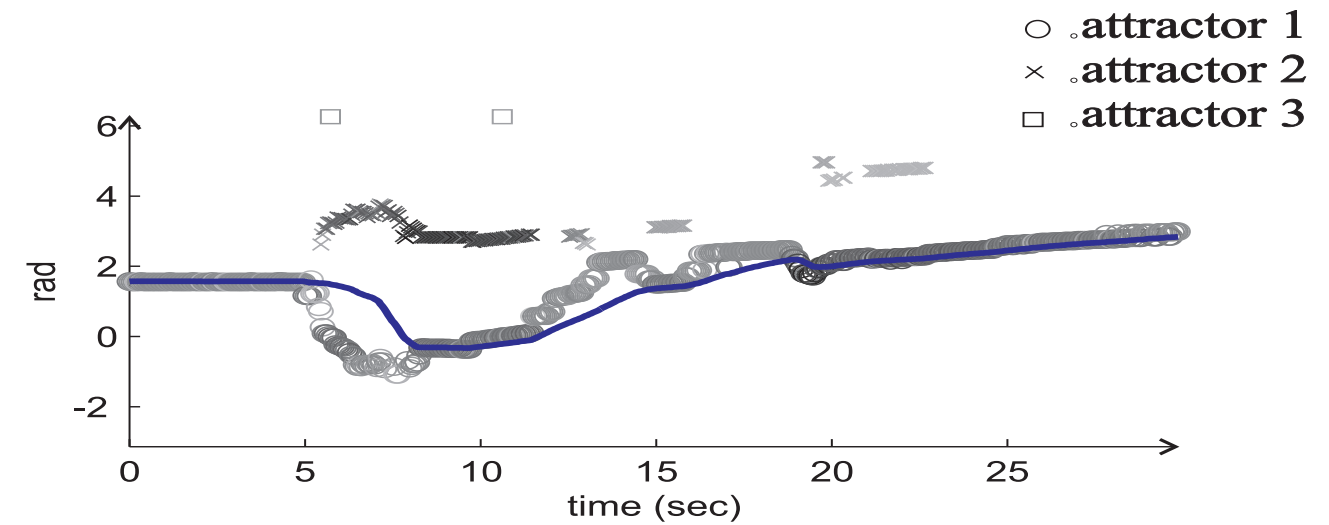


Behavioral dynamics

- heading direction is always in or close to attractor... which is why bifurcations matter...
- But how may complex behavior be generated while “sitting” in an attractor?



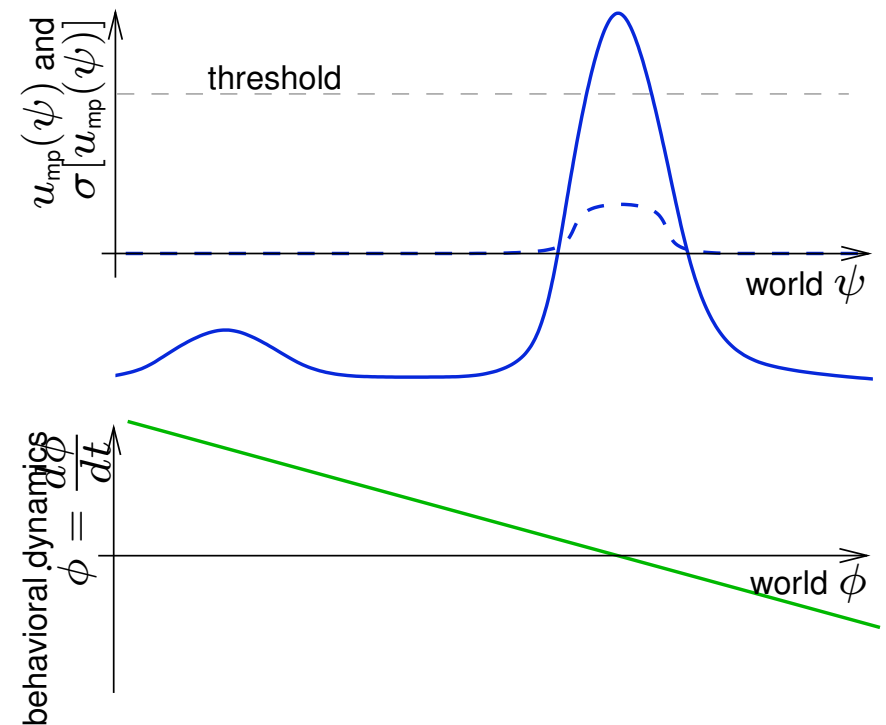
Tracking attractors



[Bicho, 2000]

Steering the behavioral dynamics

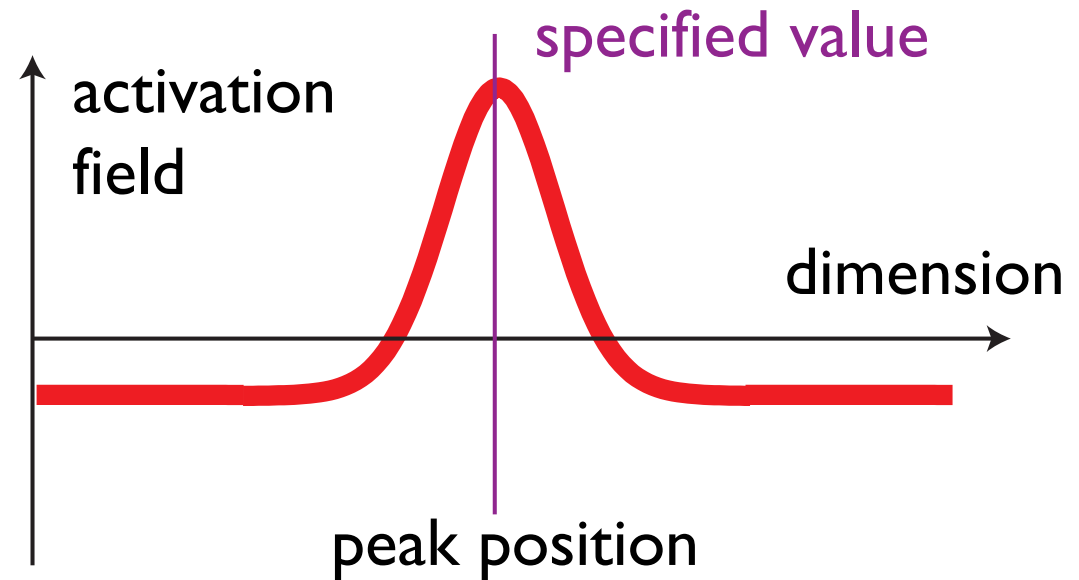
- so far, we took for granted that there is perceptual information about the constraints: targets, obstacles
- these constraints emerge from a neural dynamics: couple a peak in the neural field of target bearing into the dynamics of heading direction as an attractor



Problem number 1:

“Reading out” from the neural field?

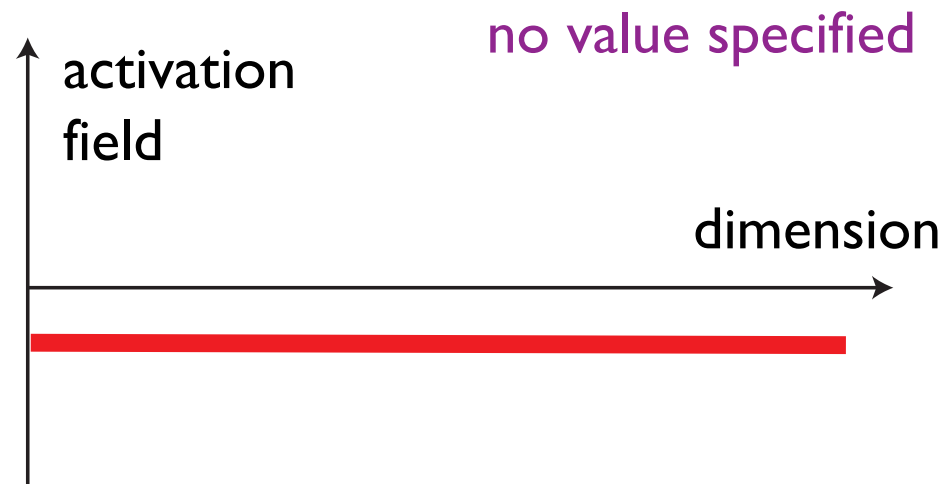
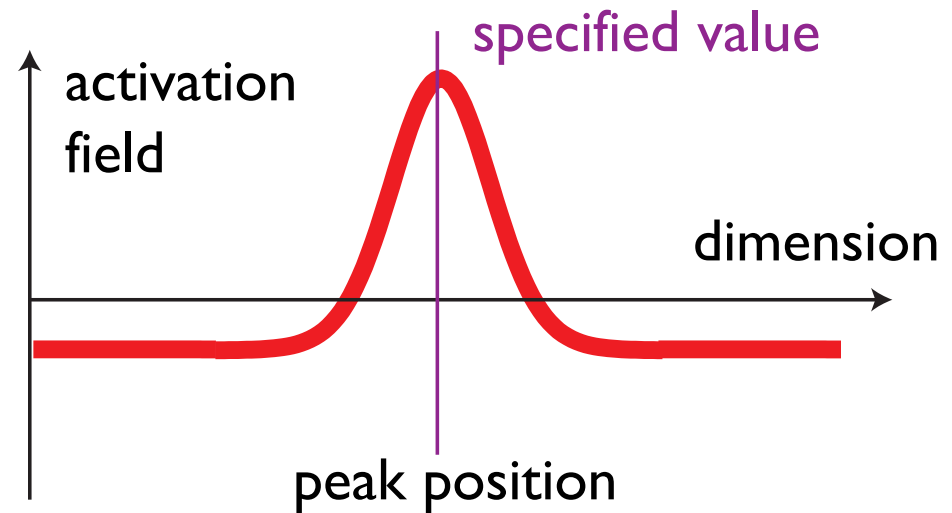
- peak specifies value of the field dimension over which it is located...
- but how to “read out” that value?



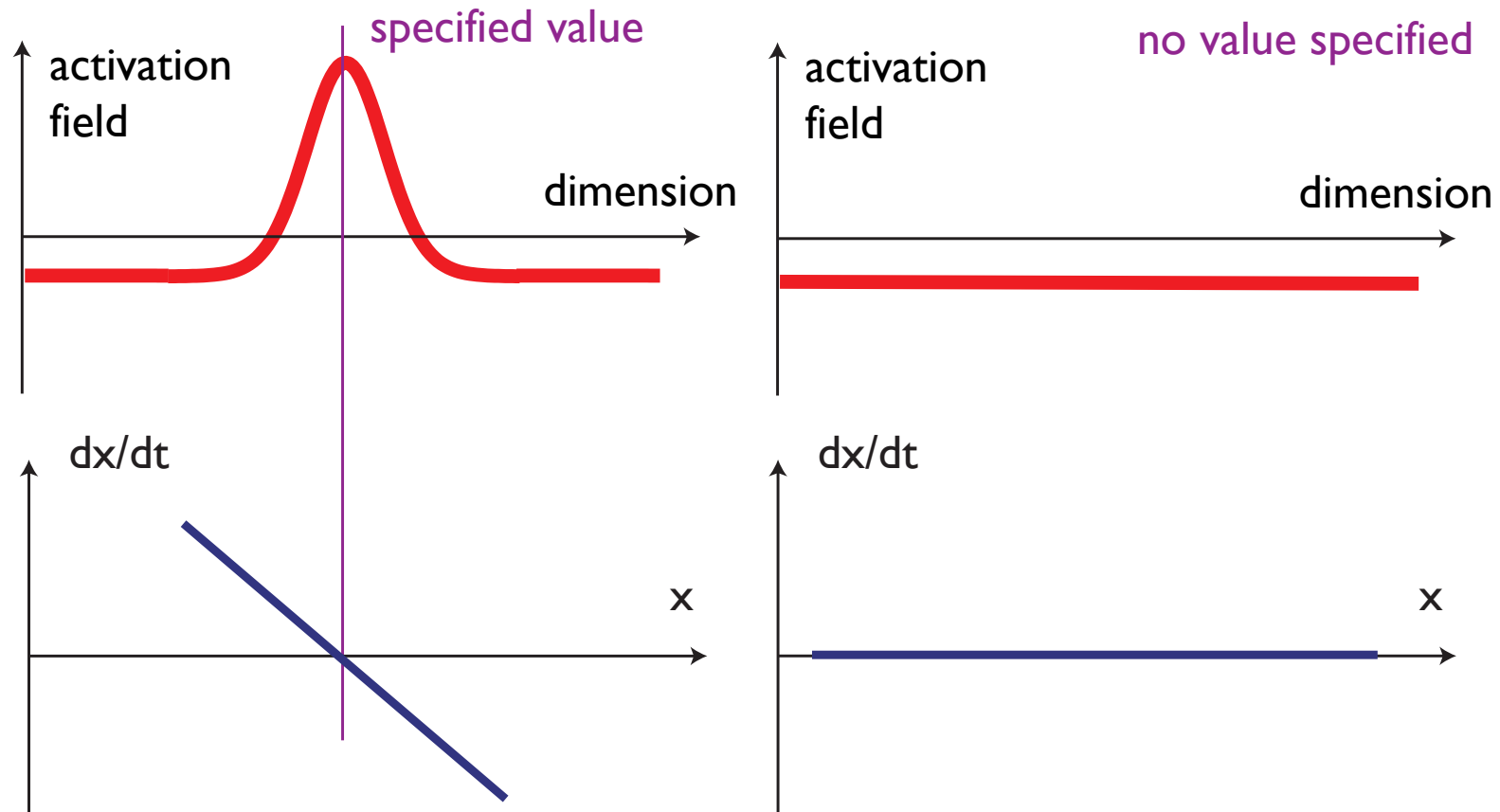
“reading out” from the neural field?

- standard idea: treat supra-threshold field as a probability density
- but: need to normalize the activation pattern
- => problem when there is no peak: divide by zero!

$$x_{\text{peak}} = \frac{\int dx \, x \, \sigma(u(x, t))}{\int dx \, \sigma(u(x, t))}$$



“reading out” from the neural field?



from DFT to DST

■ solution: peak sets attractor

■ location of attractor: peak location

■ strength of attractor: summed supra-threshold activation

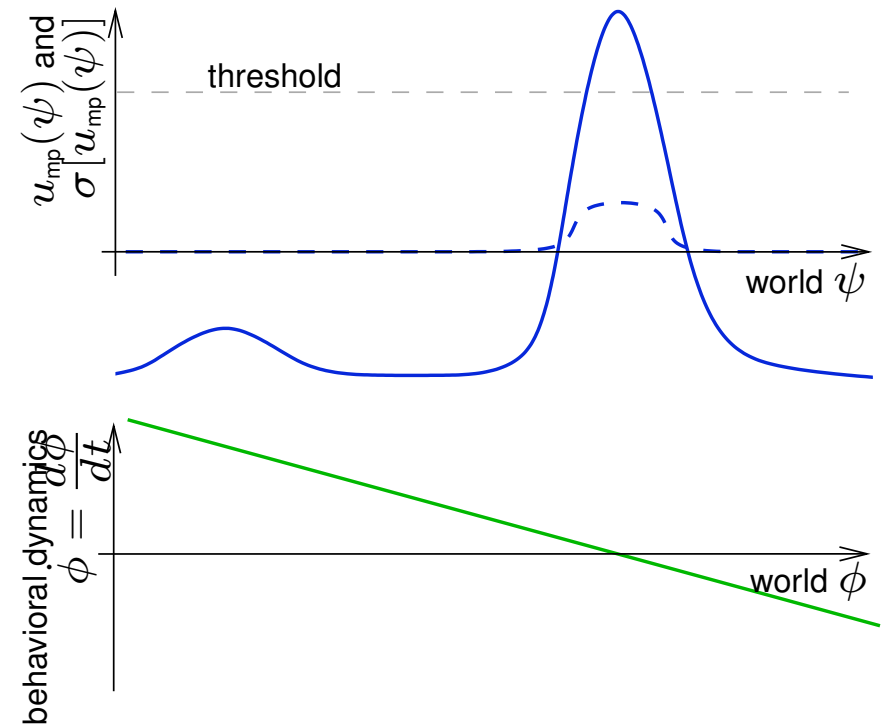
$$x_{\text{peak}} = \frac{\int dx \, x \, \sigma(u(x, t))}{\int dx \, \sigma(u(x, t))}$$

$$\dot{x} = - \left[\int dx \, \sigma(u(x, t)) \right] (x - x_{\text{peak}})$$

$$\Rightarrow \dot{x} = - \left[\int dx \, \sigma(u(x, t)) \right] x + \left[\int dx \, x \, \sigma(u(x, t)) \right]$$

Problem number 2: closed loop

- the target representation is invariant in space, defined over heading direction
- and so is the motor dynamics...
- how does the “heading direction” then capture the physical state of the body in the world ~ behavioral dynamics?

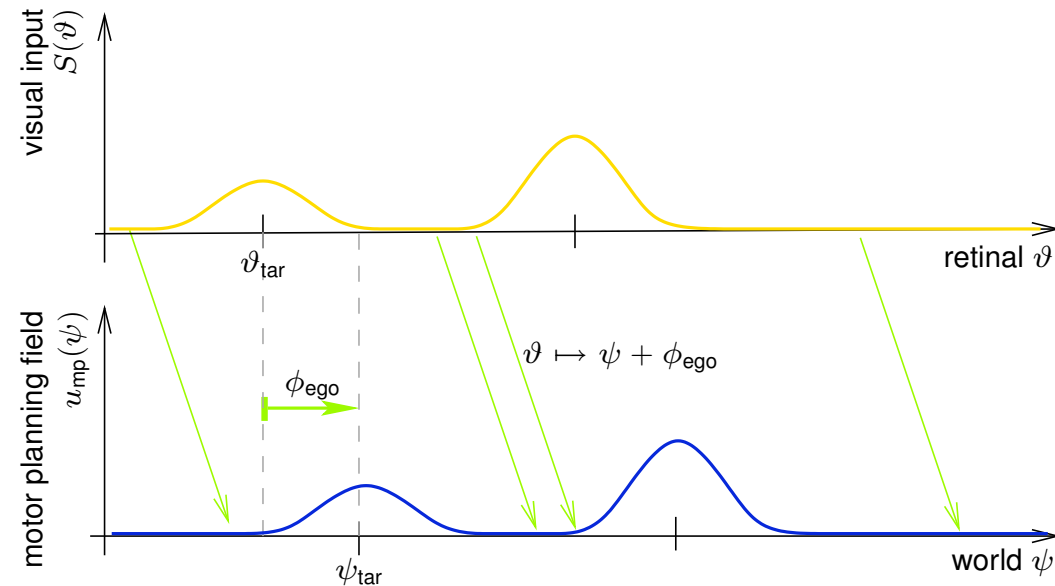


Answer

- the target representation must be invariant under a change in heading because it is in that frame that working memory about the target and neural state about target selection is meaningful... this is a property of the world
- and the same argument applies to the motor dynamics: only when the dynamics is invariant under change of heading is it a meaningful dynamics

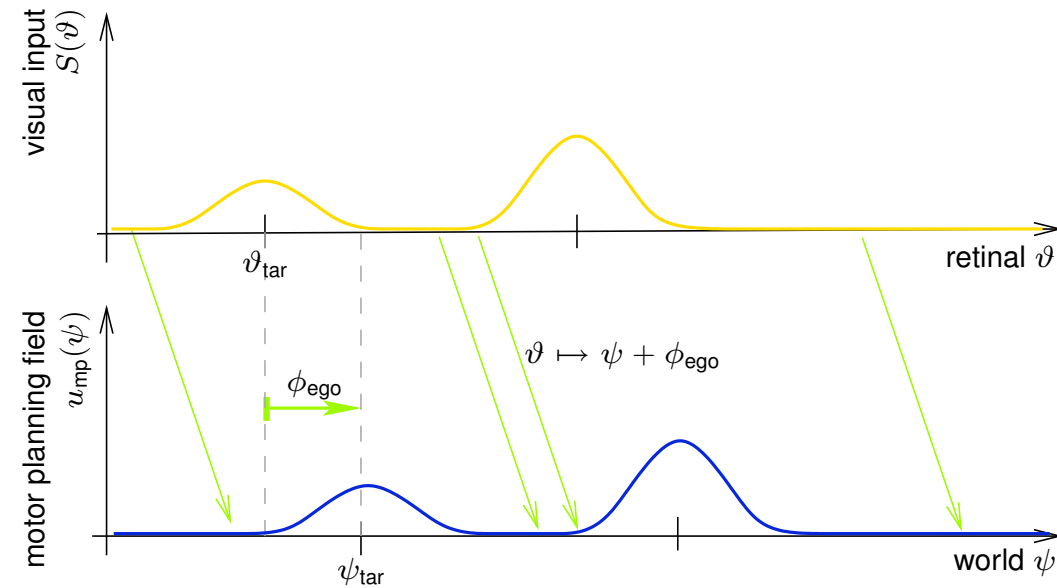
Answer

- to make this consistent with coupling to sensory information, we must perform a coordinate transform from the sensory surface (“retina”) to the invariant world frame!
- and that requires knowing the heading direction in the world...



Answer

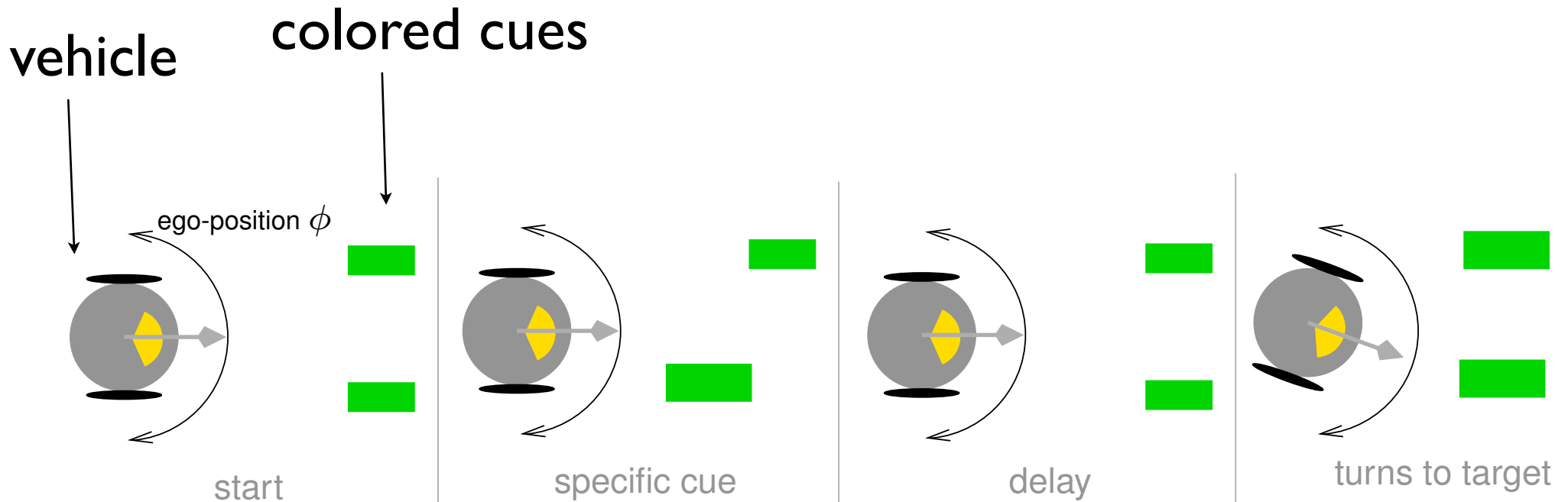
- this is a steerable neural map... and we'll cover that in the next lecture





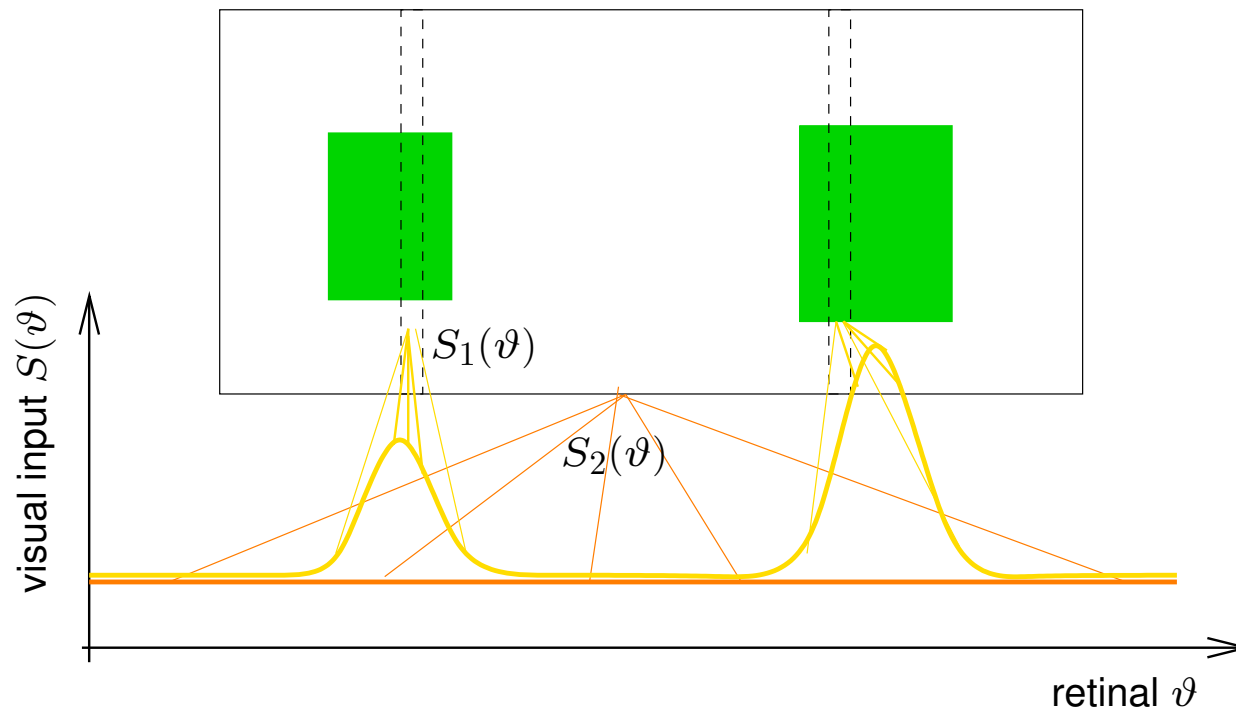
Embodied A not B

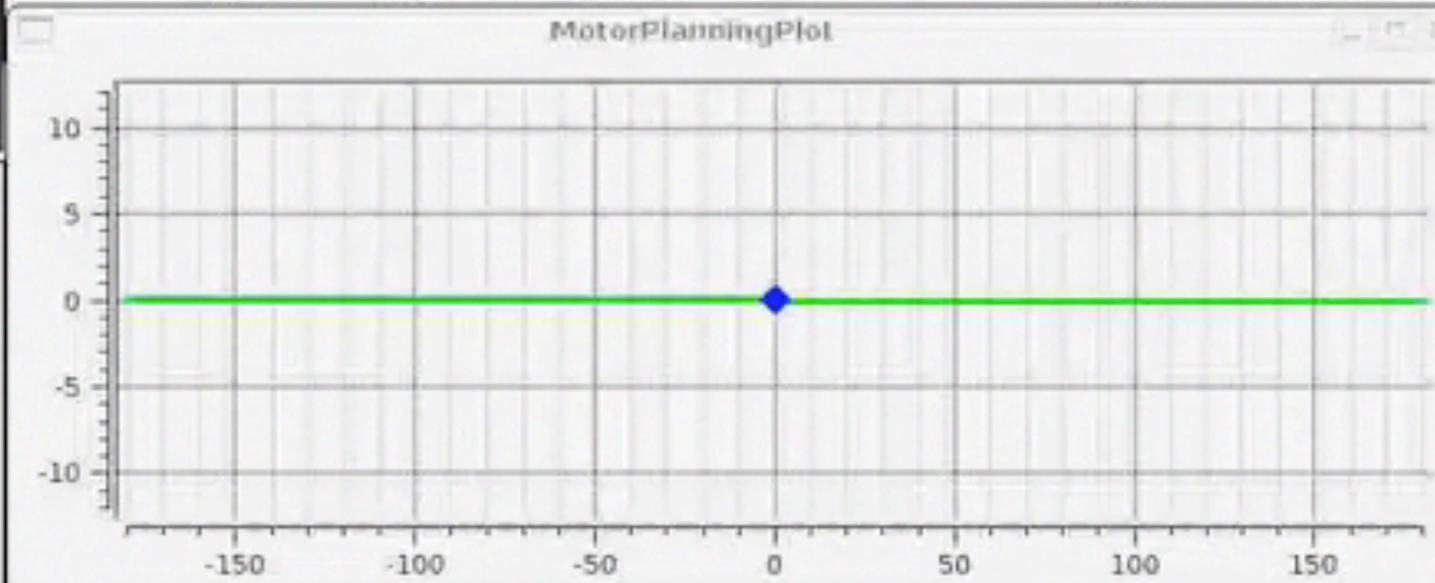
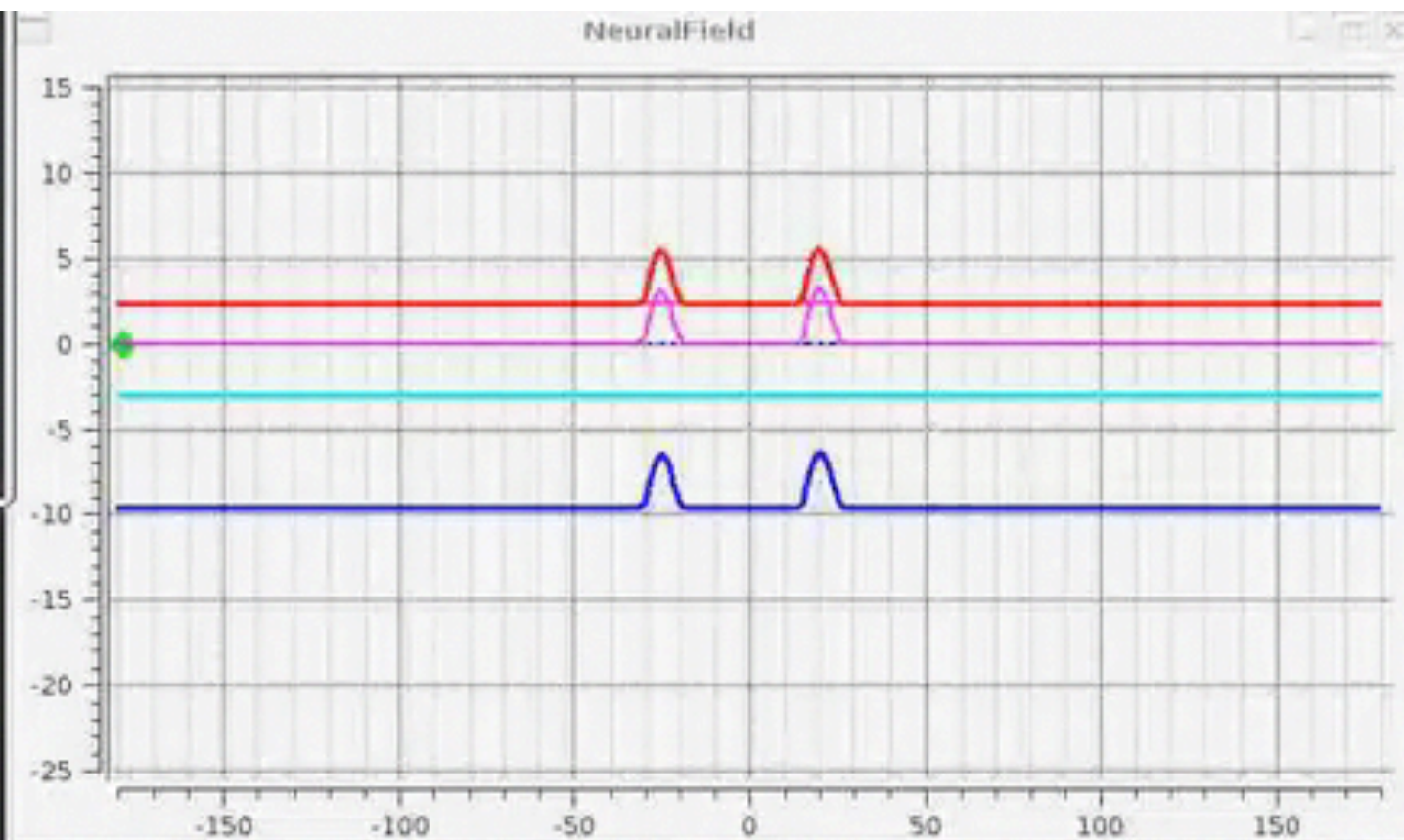
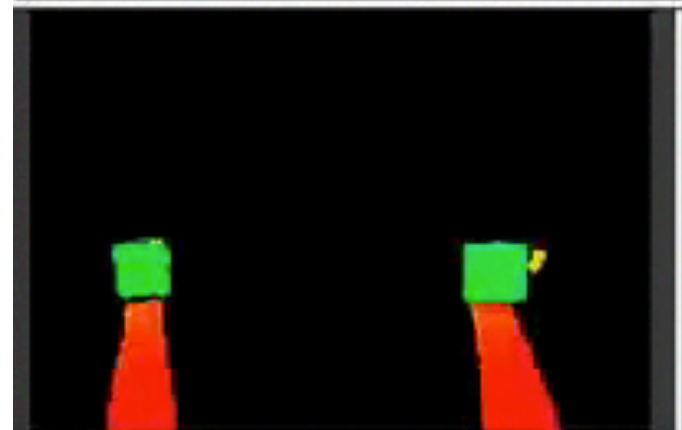
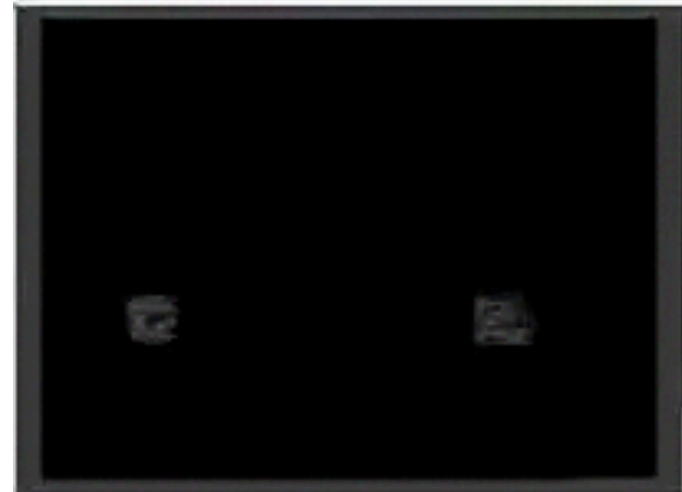
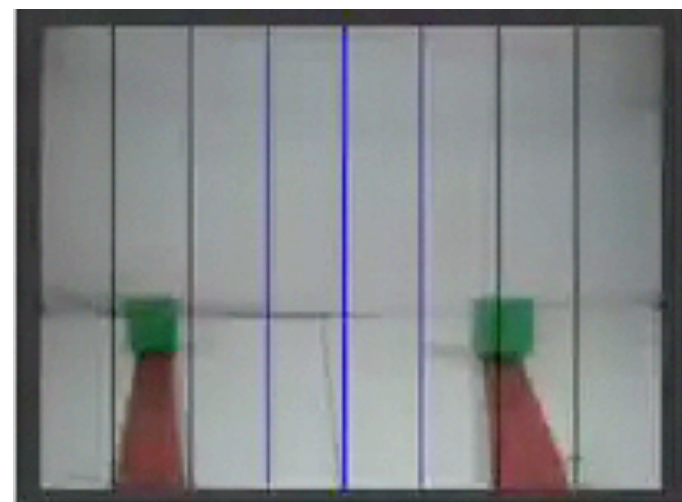
- implementing the A not B model on a autonomous robot with continuous link to sensory and motor surfaces...



Visual input

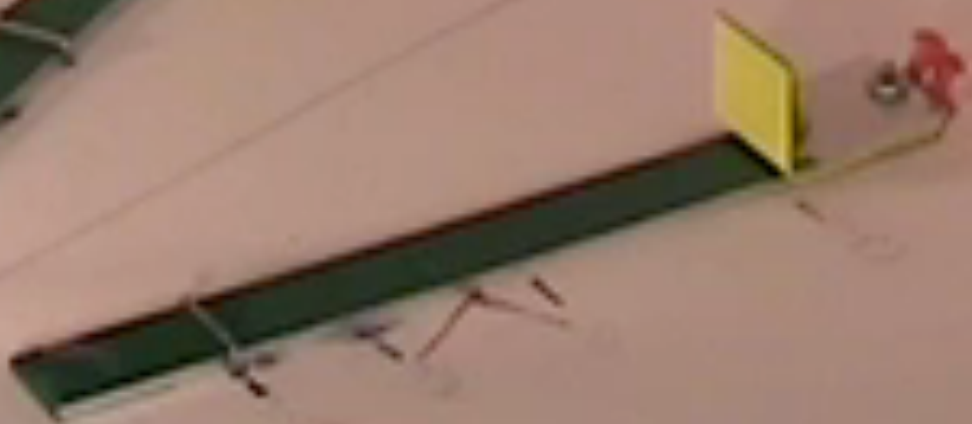
- color-based segmentation
- summing color pixels within color slot along the vertical
- spatially filter at two resolutions

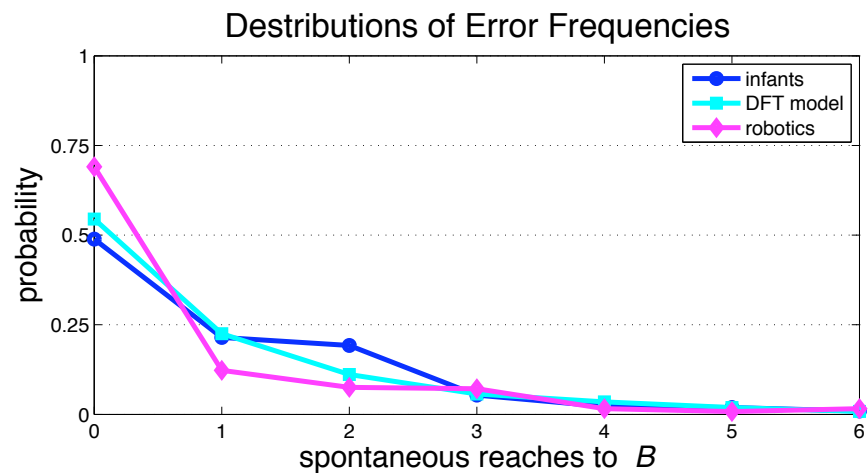
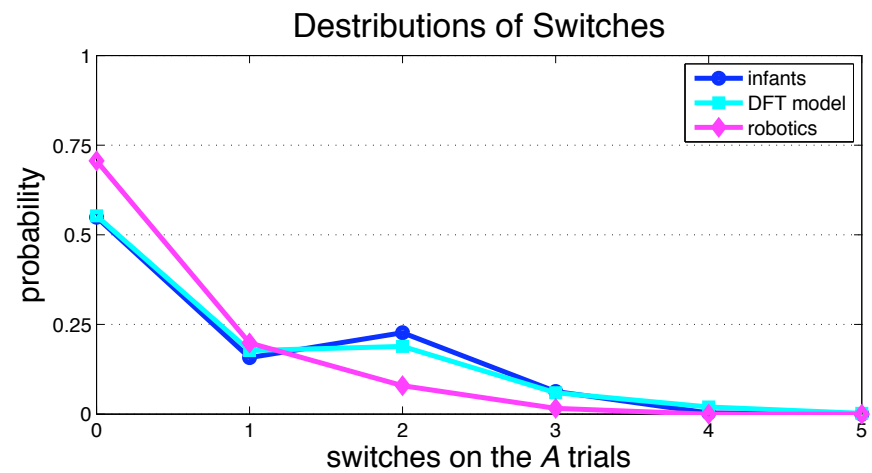
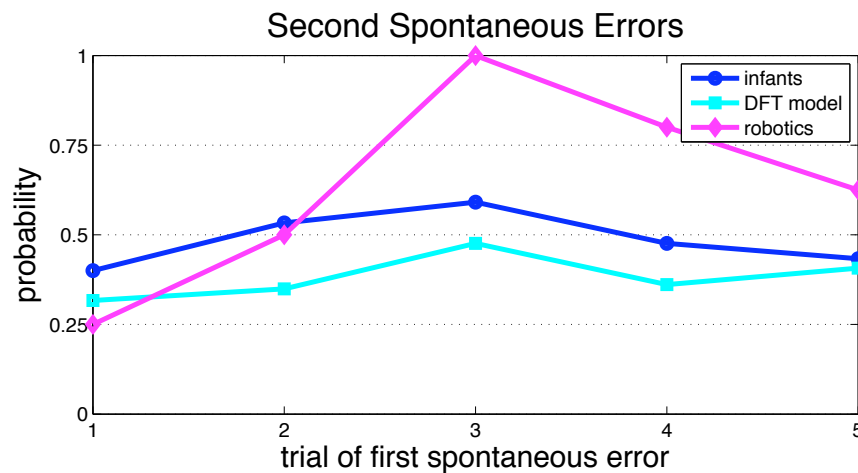
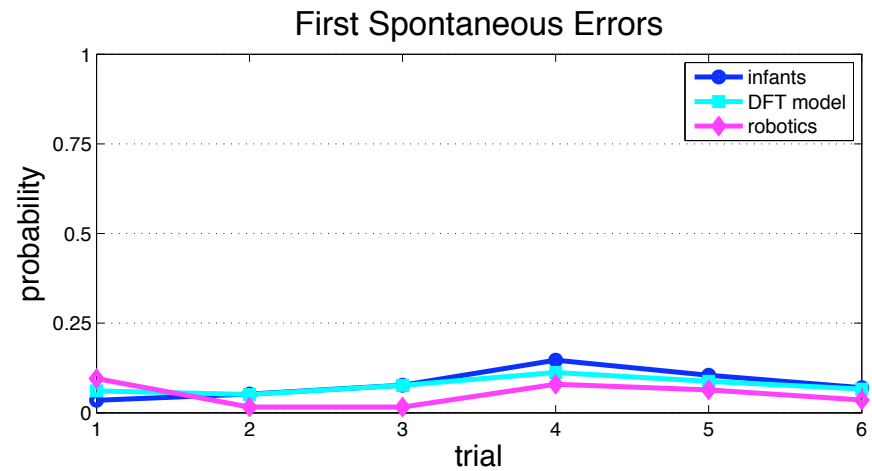
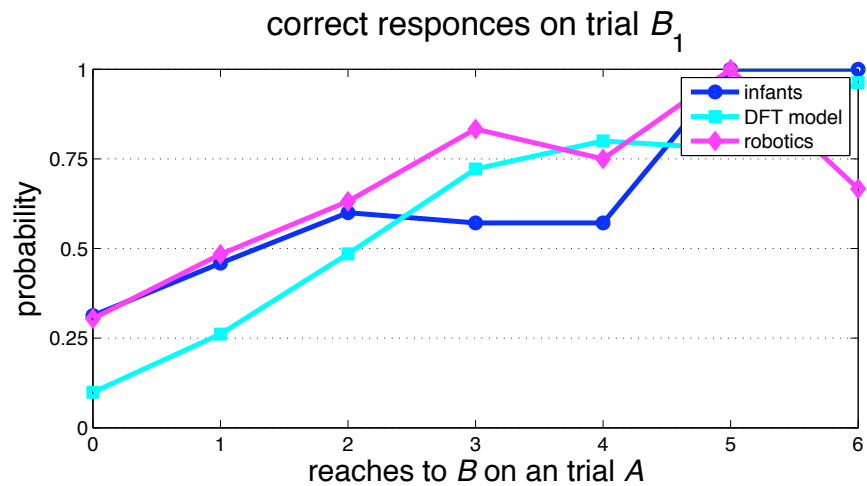




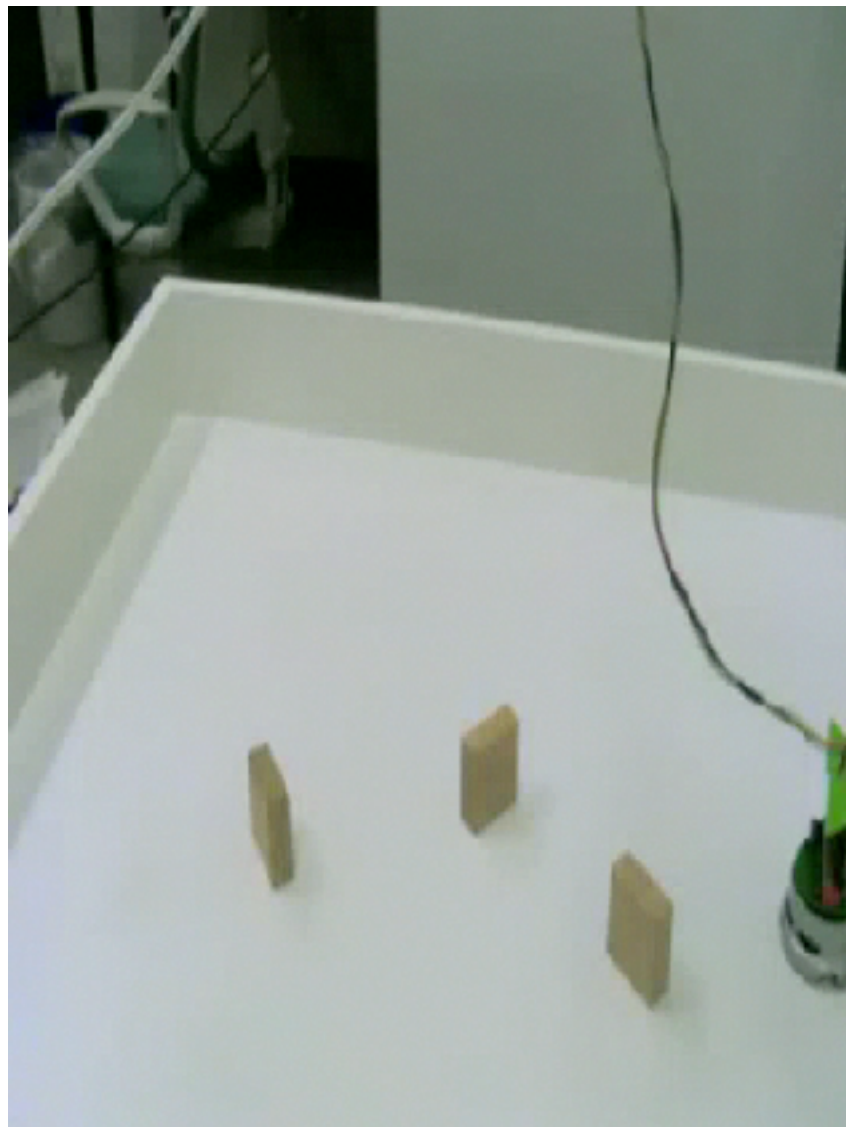
Exp #3

R

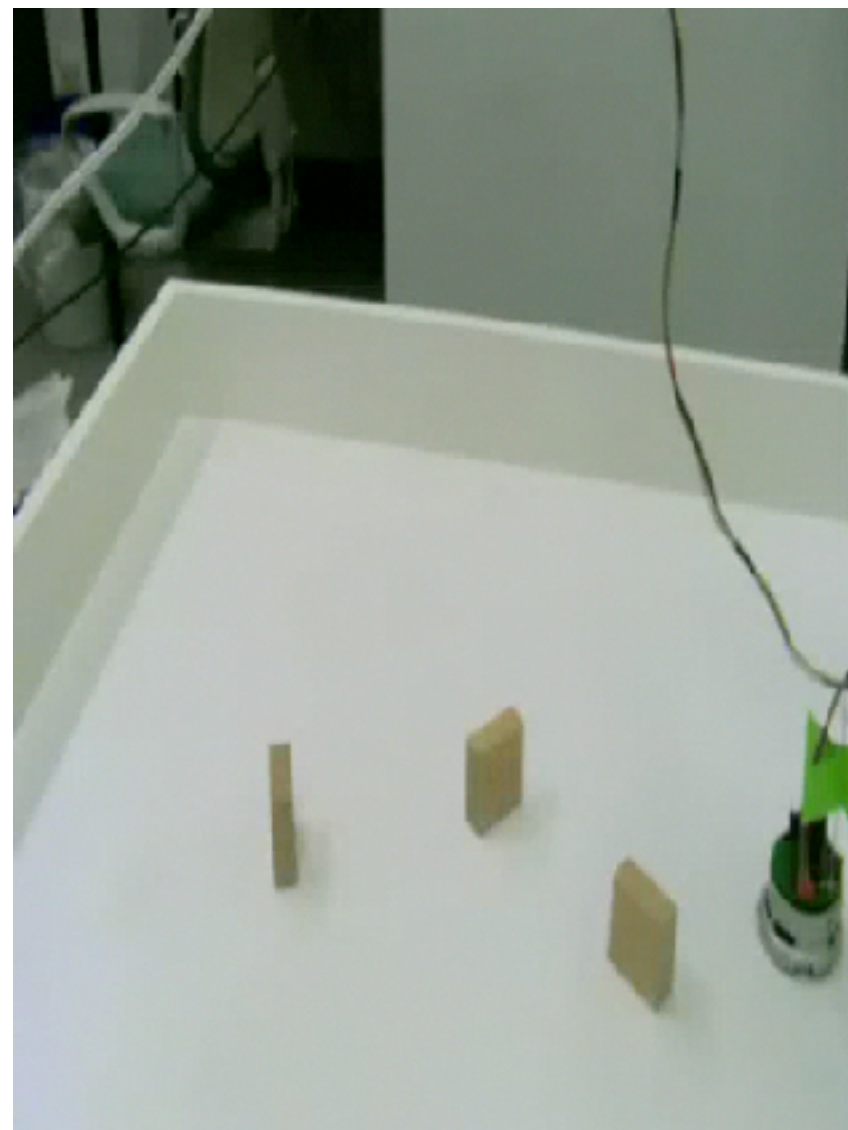




“young” robot

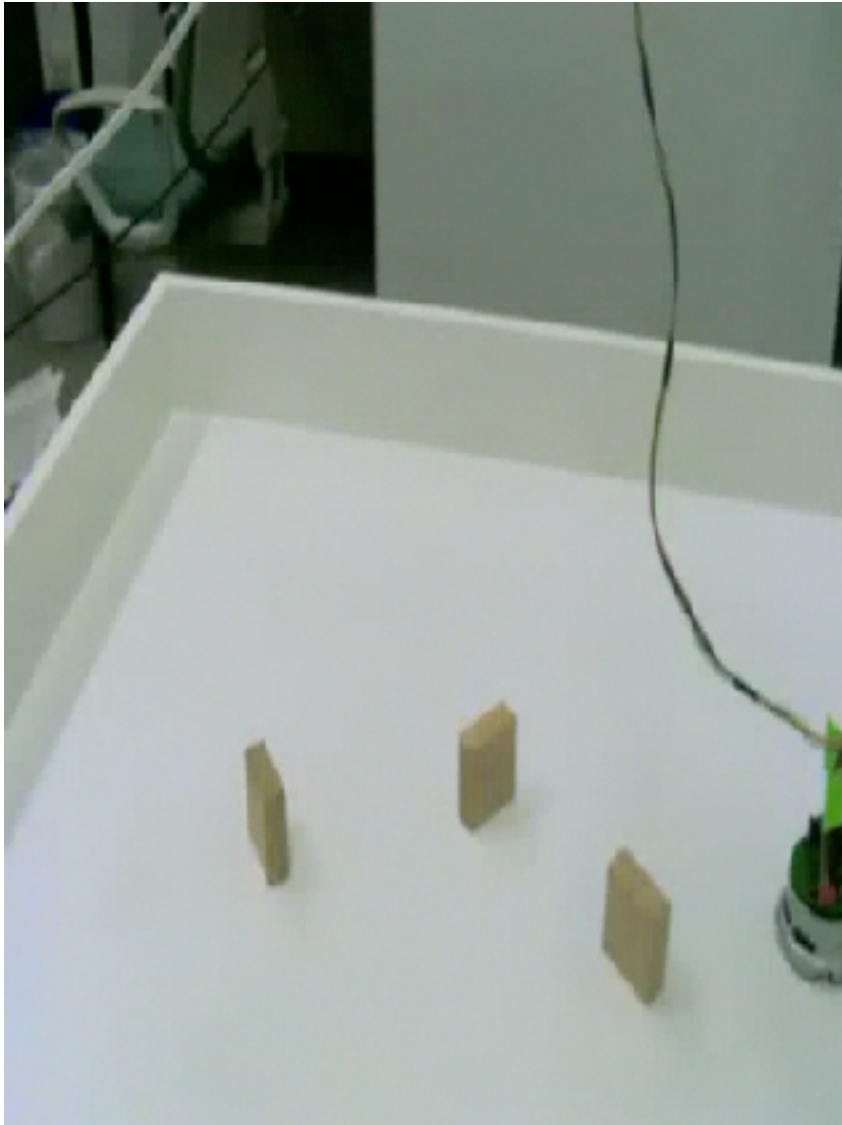
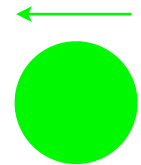


“old” robot



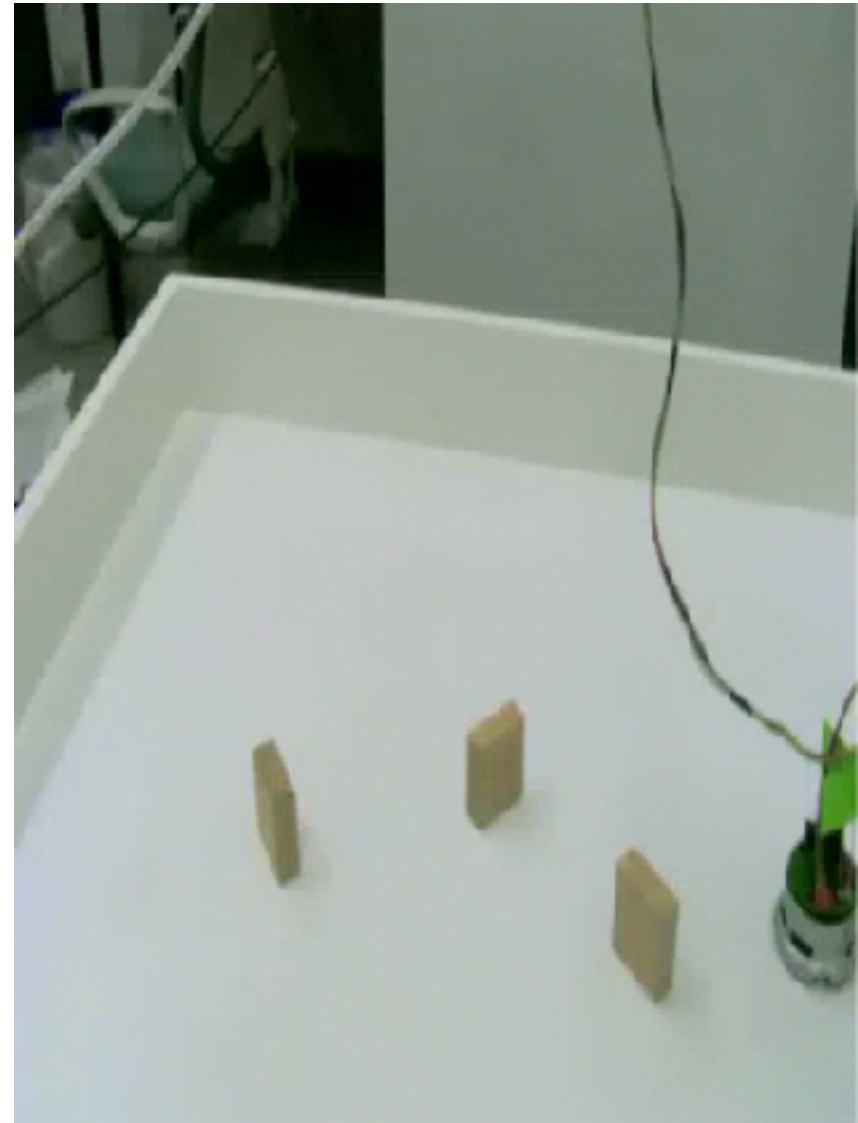
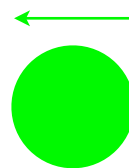
“young” robot

target

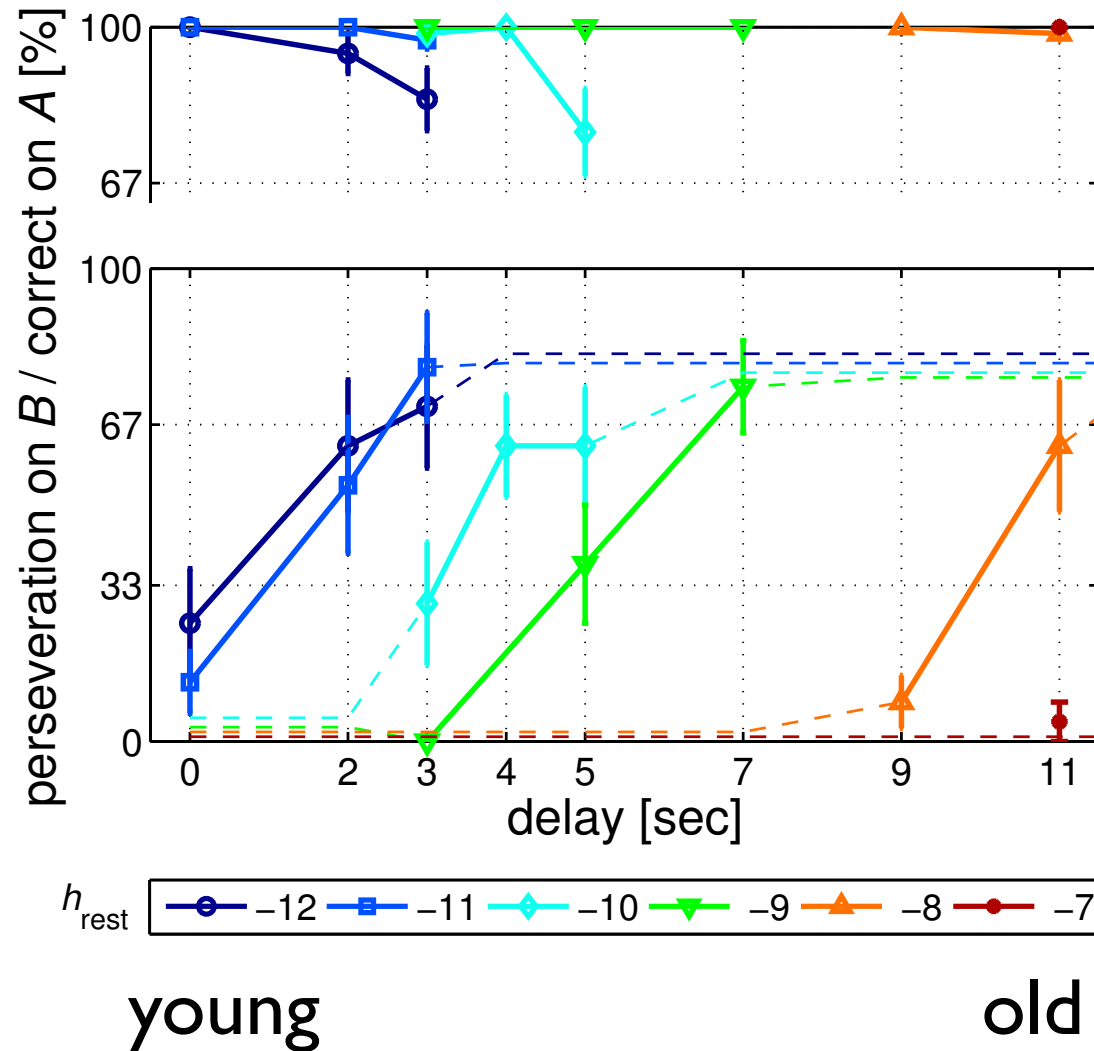


“young” robot with
memory trace

target



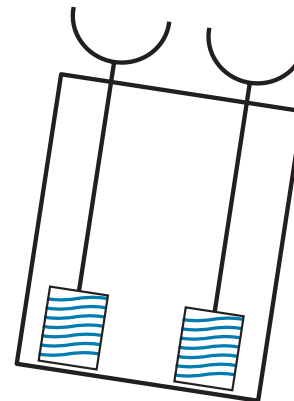
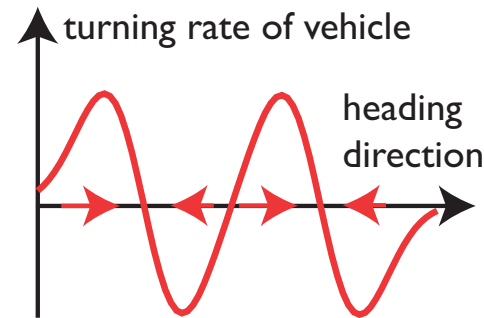
result: reproduce fundamental
age-delay trade-off in A not B



Conclusion

- neural dynamics directly driven by sensory input
- attractor dynamics all the way down to behavioral variables
- fields couple into behavioral dynamics by setting attractors => no more “read-out” of neural dynamics

behavioral dynamics



neural dynamics

