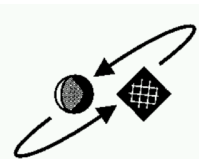


Anticipation in cybernetic systems: A case against mindless antirepresentationalism

Lambert Schomaker



Kunstmatige Intelligentie / RuG



Overview

- From data to explanation: competing theories
- Neural representations
- Anticipation and attention: phenomena requiring representation
- Conclusions

War of worlds/words

- behaviorism & associationism

$$Stim \rightarrow Resp$$

- traditional symbolic cognitive science

$$Act = Cogn(Perc)$$

- ecological approaches

$$Act \Leftrightarrow Perc$$

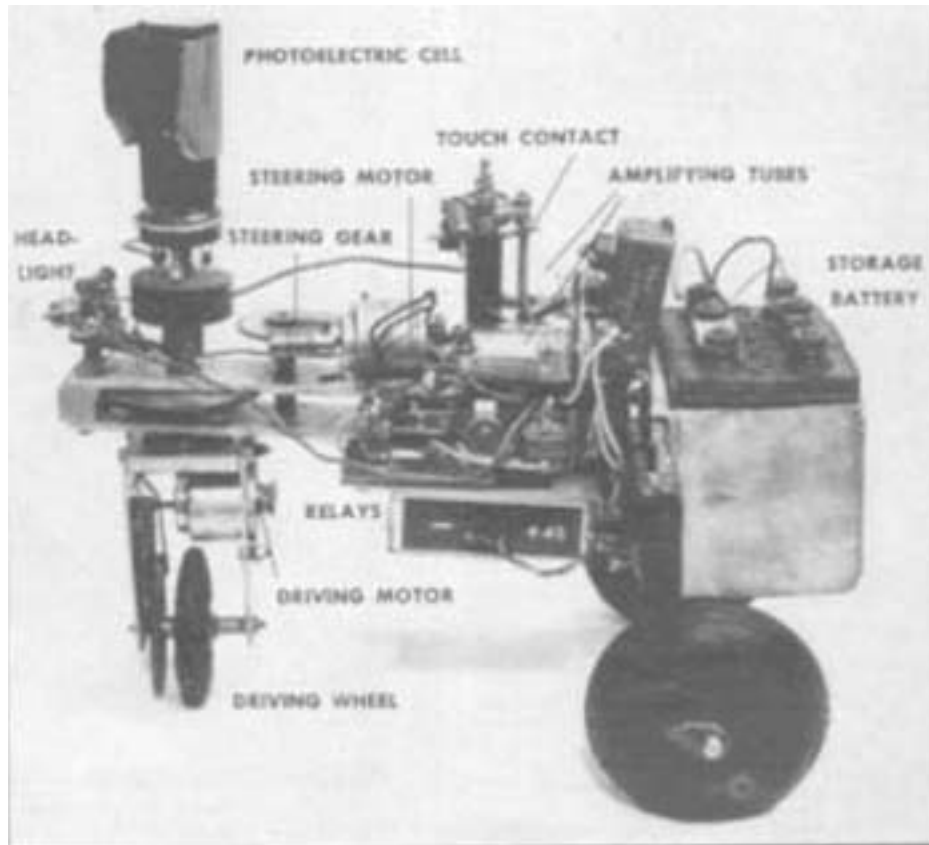
- the brain-imaging revolution

$$Act = Brain(Perc)$$

Cognitive theories vs(?) Non-linear dynamic systems theories

- **Grey Walter (1948)**
Emergent behavior in Turtle bots
- **JJ Gibson (1960-1970)**
Ecological perception & action
- **Scott Kelso (198x)**
Action-Perception as a pattern formation process
- **Rodney Brooks (1991)**
Intelligence without representation

Grey Walter (194x): behavioral complexity through simple perception/action mechanisms



“Elsie the artificial tortoise”

- light sensor
- thermionic valve
- simple steering

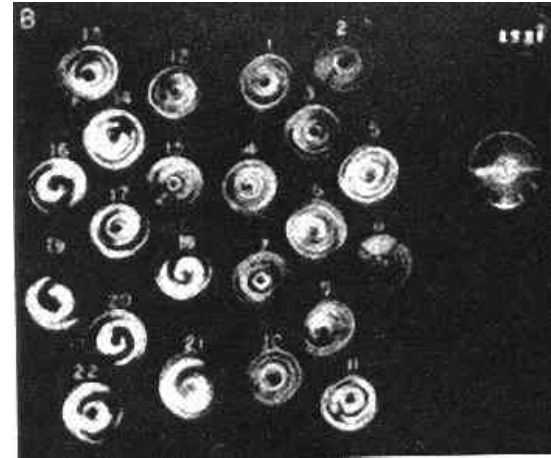
- Nonlinearity, e.g.:
go towards faint light,
avoid bright light

Grey Walter (194x): turtle dance

two electromechanical turtles,
each with a non-linear light
sensor and a light source over
its shell, produce a strange
movement,
**“like the mating behavior of
animals”**



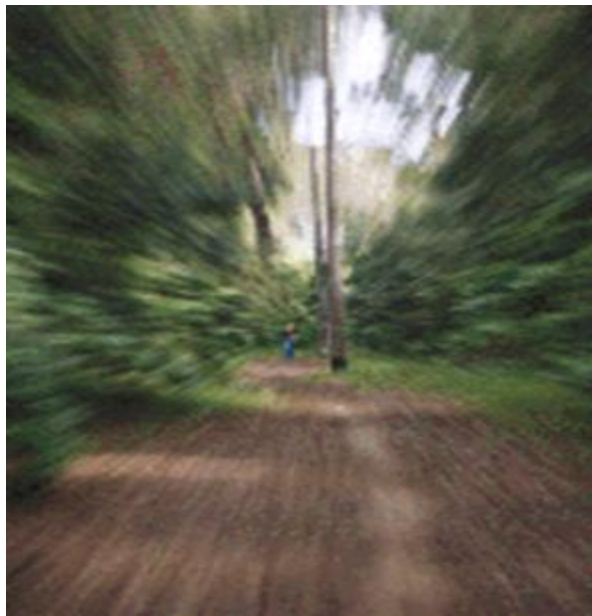
Grey Walter, Wiener et al. 40's/50's...



even in the early days there is a strong sense of friction between
“behavioral complexity through a few simple rules”
and
“brain complexity through many simple neurons”

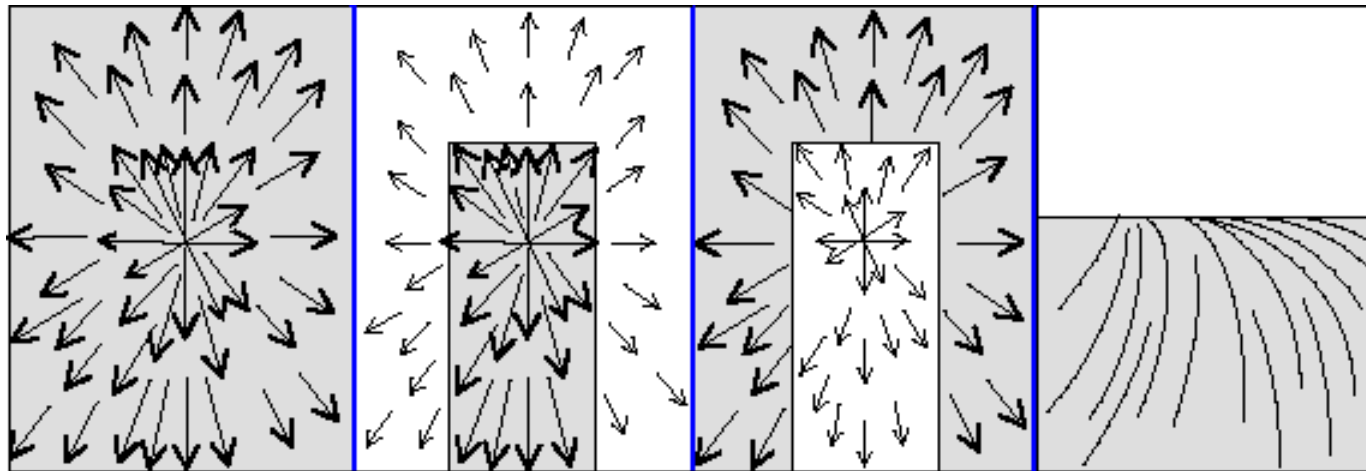
JJ Gibson 70's, Scott Kelso, 80's

- Perception/Action: seamless integration into the world. Example: ego motion and optic flow



JJ Gibson 70's, Scott Kelso, 80's

- Perception/Action: seamless integration into the world. Example: ego motion and optic flow



Approach

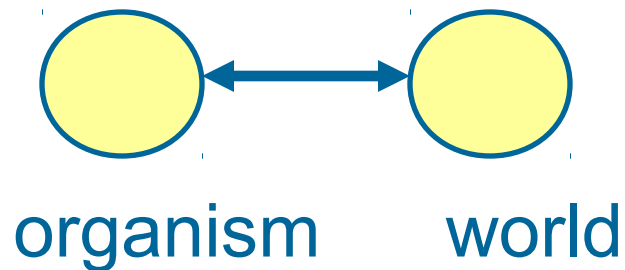
Approach
obstacle

Approach
hole

Curvilinear
heading

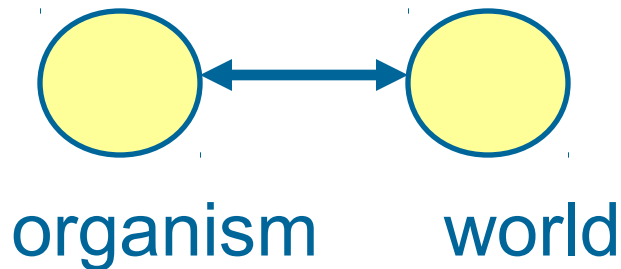
JJ Gibson 70's, Scott Kelso, 80's

- Perception/Action: seamless integration into the world

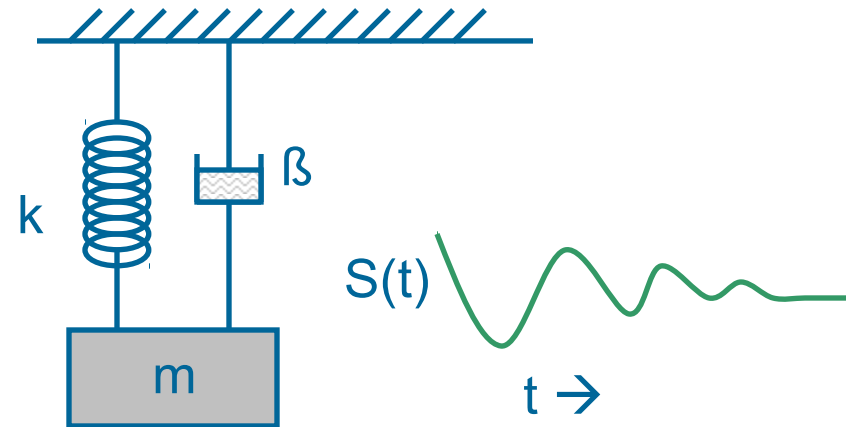


JJ Gibson 70's, Scott Kelso, 80's

- Perception/Action: seamless integration into the world



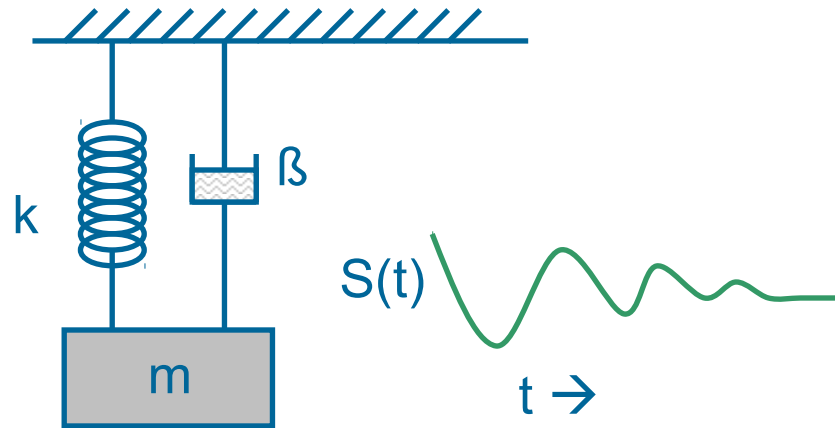
Like in:



mass, spring & friction:
what causes the motion?

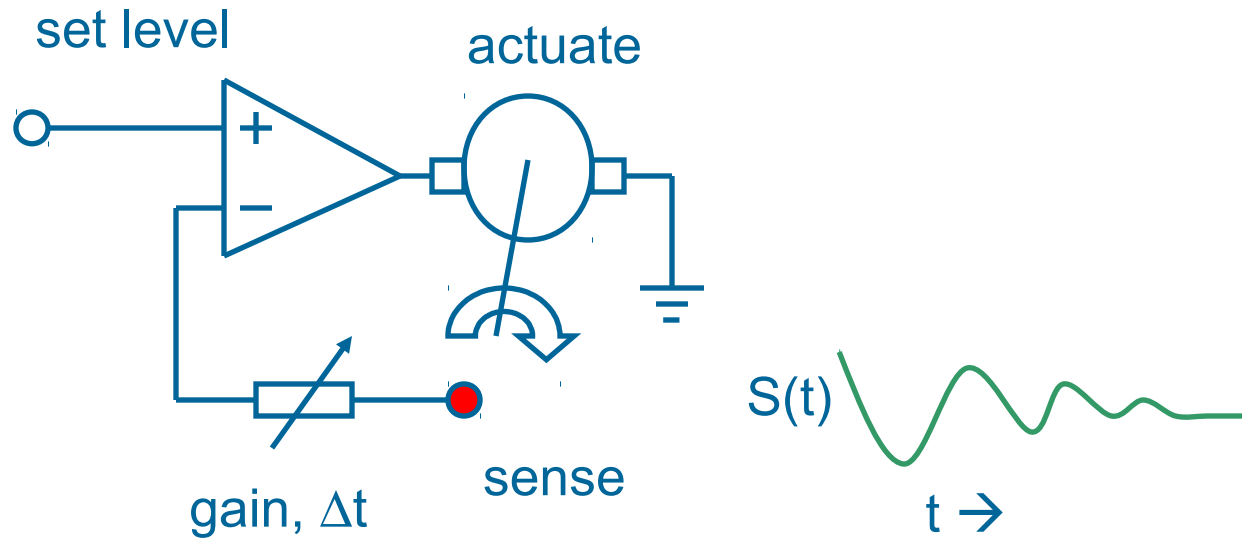
$$m\ddot{x}_t + \beta\dot{x}_t + kx_t = c$$

Physics



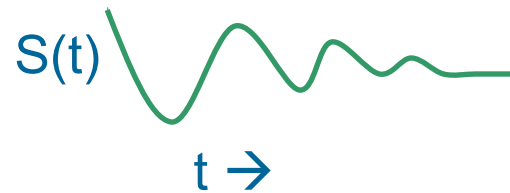
$$m x''_t + \beta x'_t + k x_t = c$$

Cybernetics

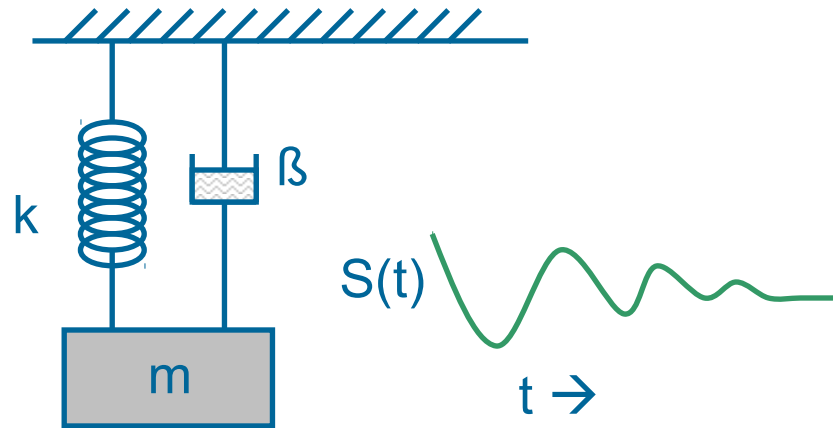


Informatics

```
while (true) {  
    S := sense(state);  
    if ( S < set_level ) {  
        actuate(s + gain * ( set_level - S));  
    }  
    sleep(dt);  
}
```



Physics... but in a wholistic sense



$$m\ddot{x}_t + \beta\dot{x}_t + kx_t = c$$

cf. Example by van Gelder, Watt's governor:
no representation, still behavior

meanwhile, in AI

- Cognitive Science & AI:
Perception → **Cognition** → Action
- ... does not seem to **work** that well in robotics
- Brooks: GOFAI needs representations & logic, but that does not help me in creating robots with believable intelligent behaviors
(*Elephants don't play chess*, Brooks, 1990)

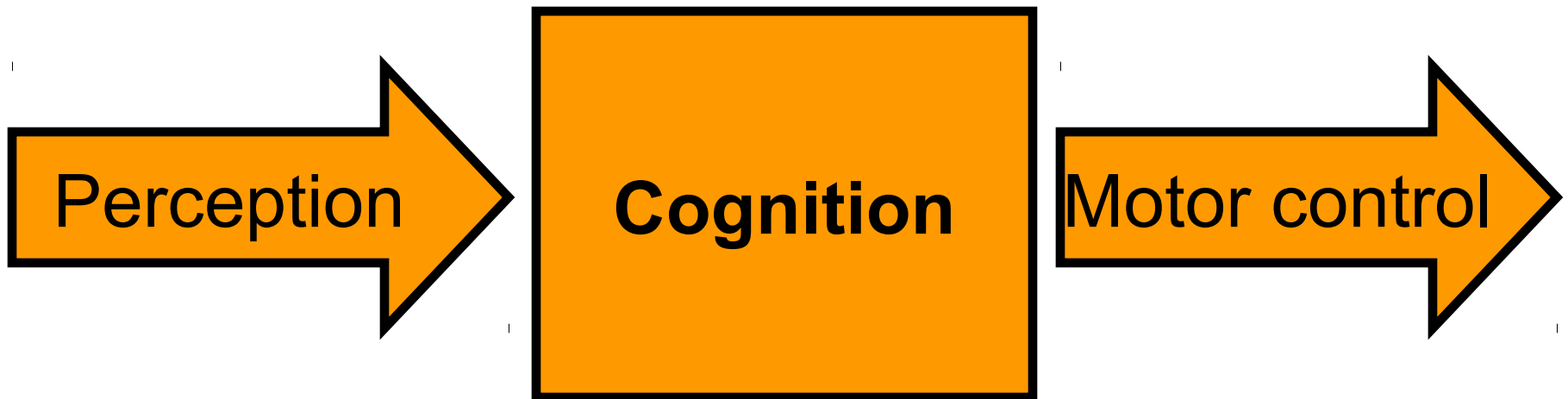
late 1990's

- → behavior-based robotics
- → Artificial Life

- → representation avoiders

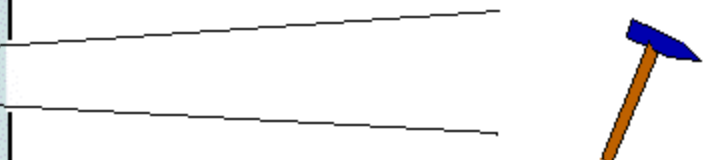
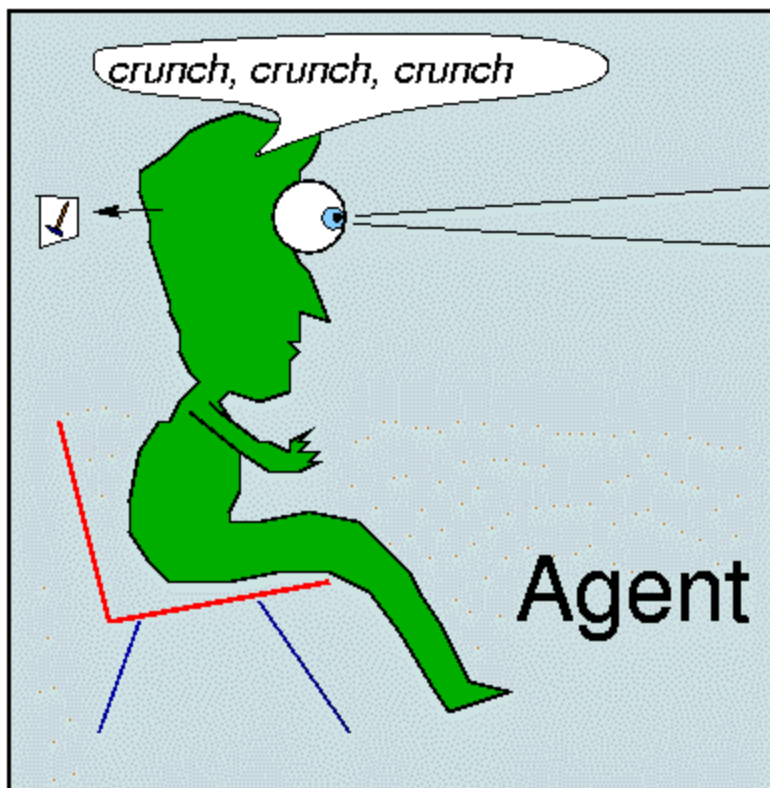


Traditional paradigm



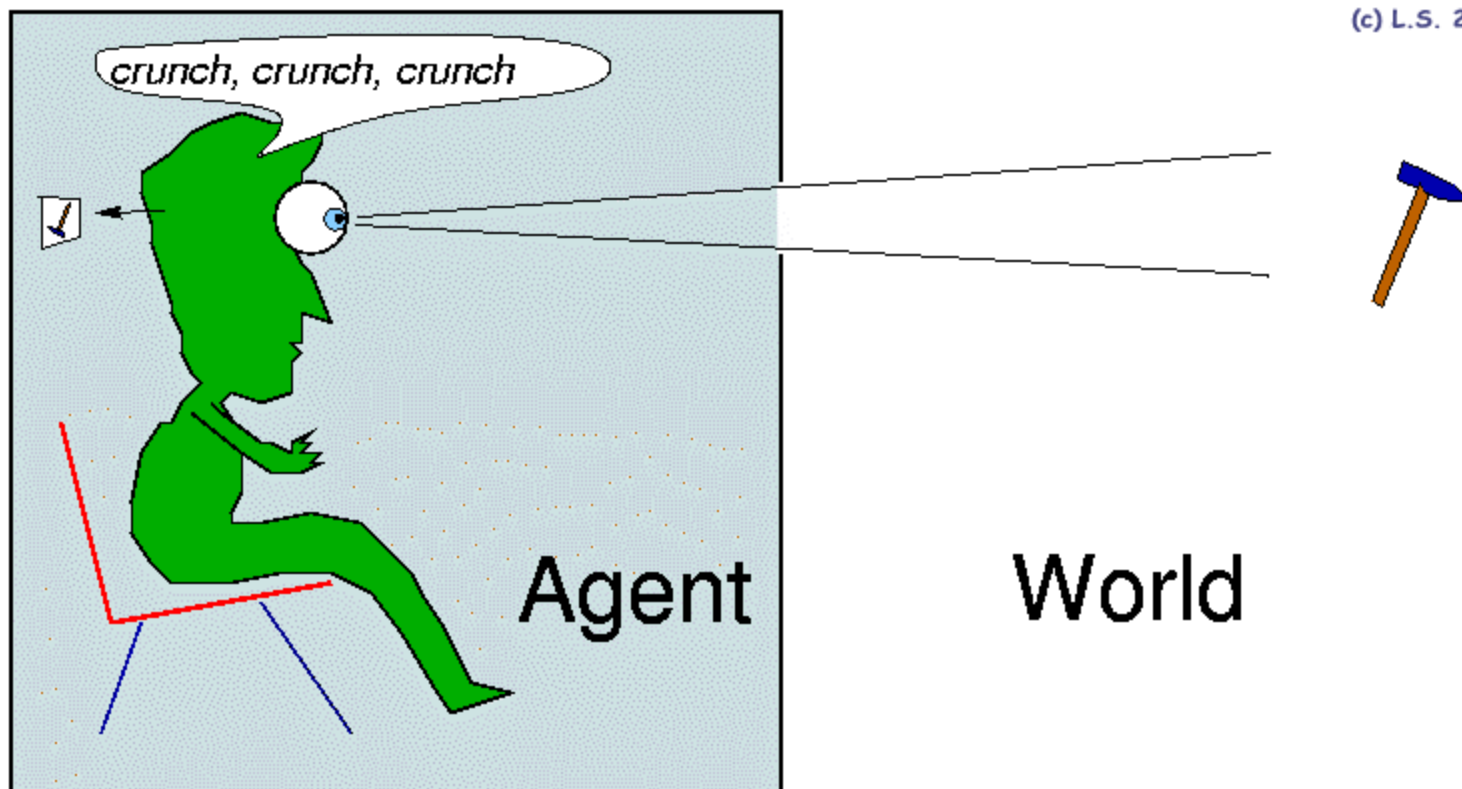
Epistemological Overspecialisation





World

How Visual Perception is viewed

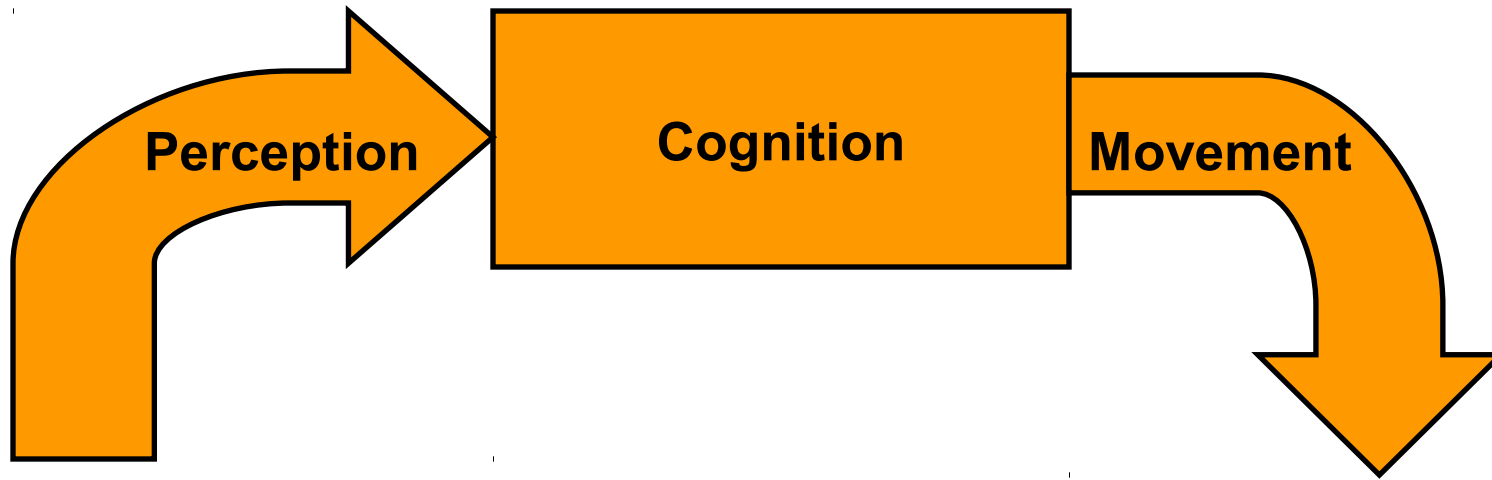


Couch-potato Peep-hole Perception

- a common paradigm in experimental psychology AND in computer vision!

Situated & Embodied systems: Close the Loop!

AGENT

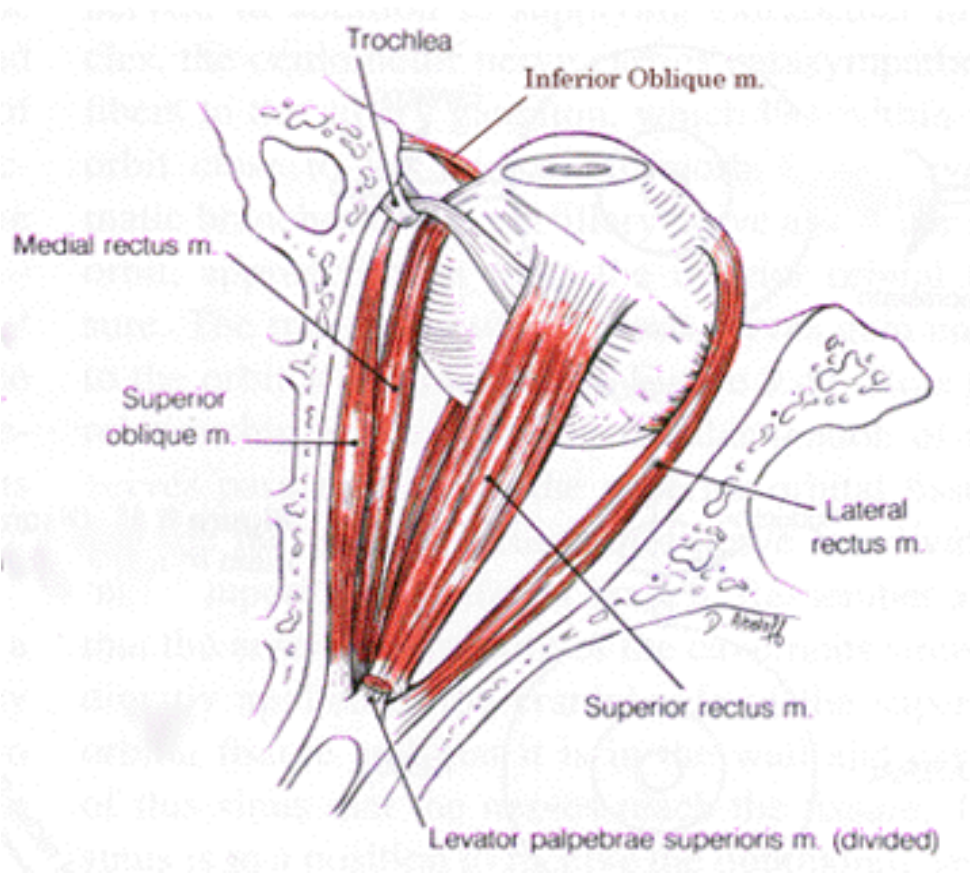


sensors



effectors

Input/Output are codependent



Input/Output are codependent



late 1990's

- → behavior-based robotics
- → Artificial Life
- → representation avoiders

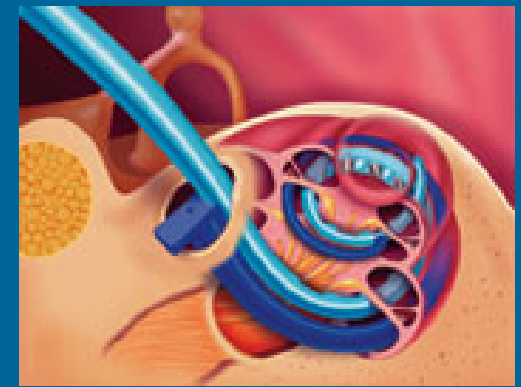
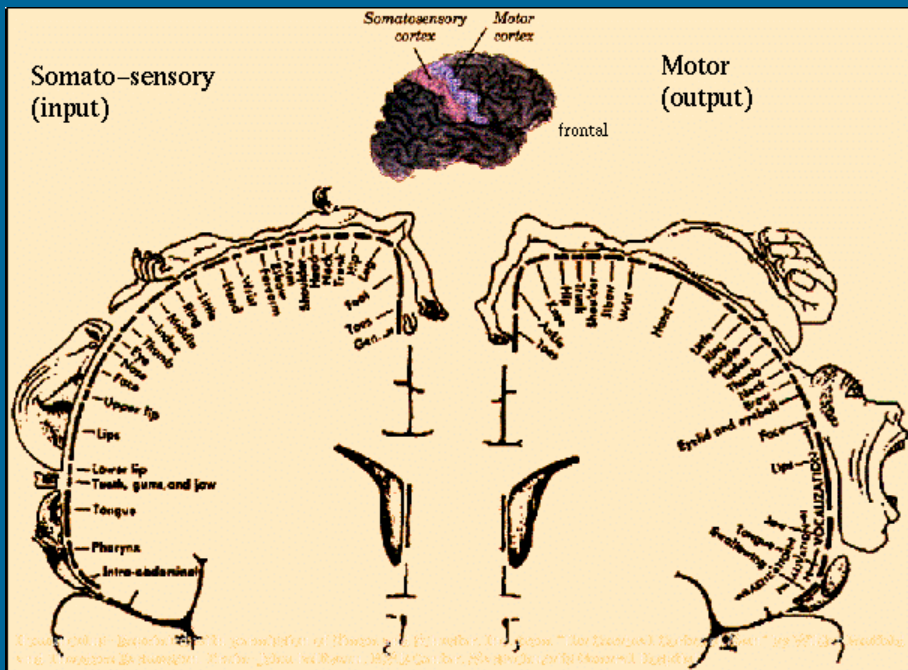
beware!

Representation in neural systems

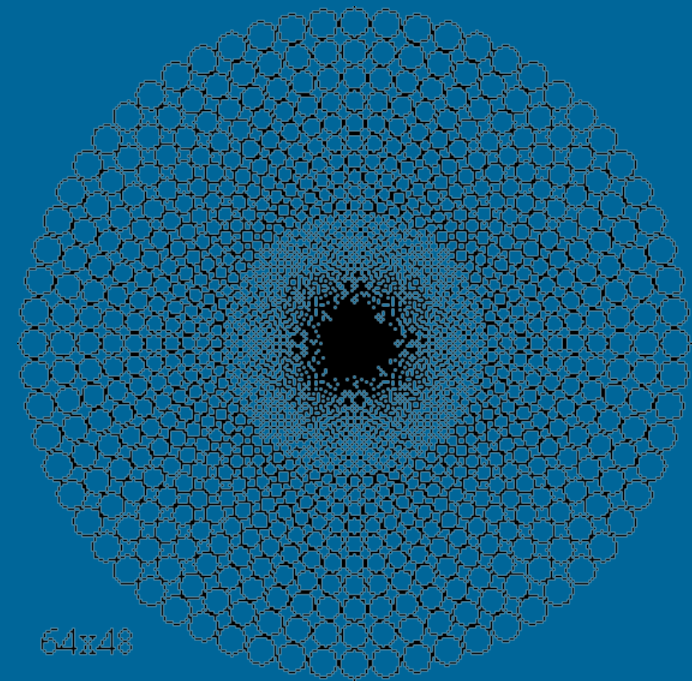
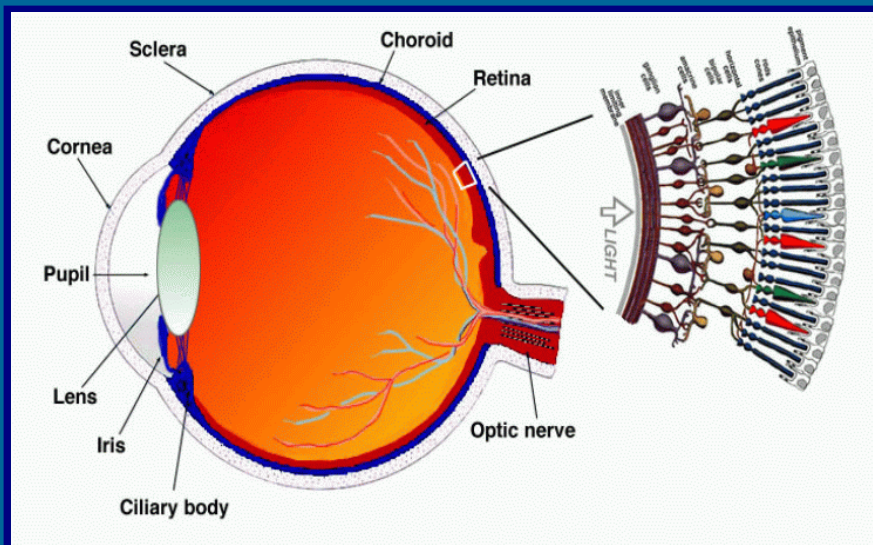
- Antirepresentationalists may throw away the baby with the bath water
- Representations are abundant in neural systems
- In order to apply simple rules, one may need complex representations!

Neural representations

- Topological: vision, hearing, tactile sensing
- Quantity coding: firing rate and recruitment
- Distributed representations
- Timing, vetoing, synchronisation, coherence

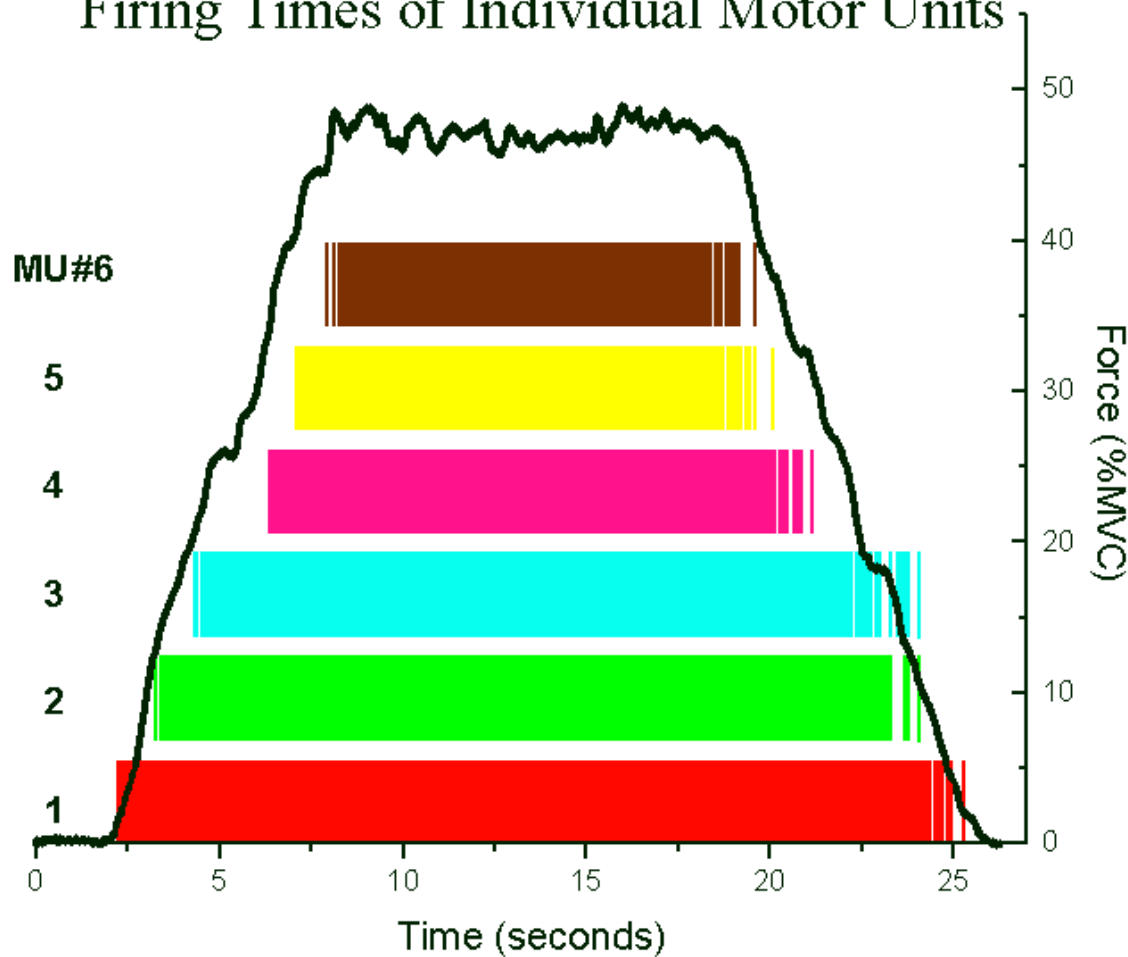


cochlea $\sim G(f)$



$x, y \rightarrow \log(r), \phi$

Firing Times of Individual Motor Units



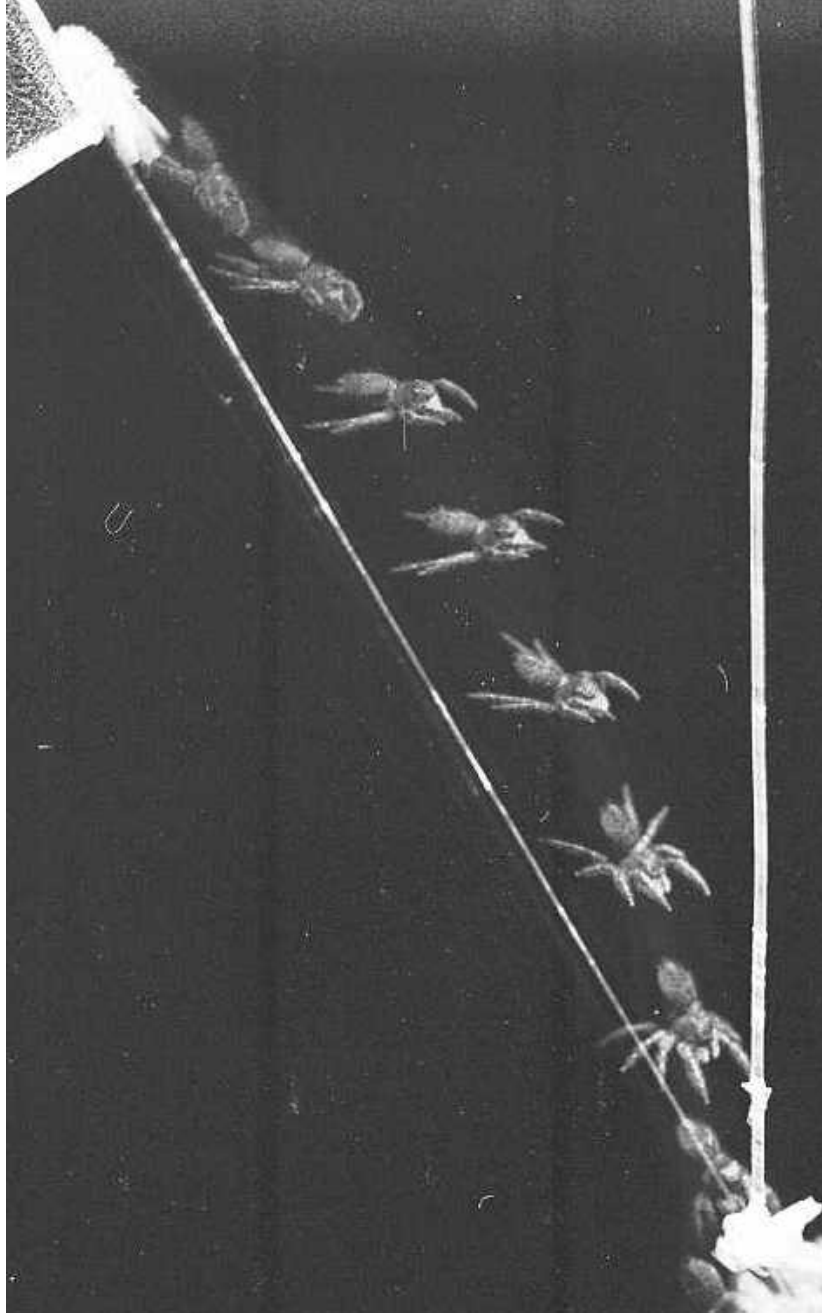
(Fig: neuromuscular research center)

“Quantity” = #units active (coarse control)
& their firing rate (fine control)

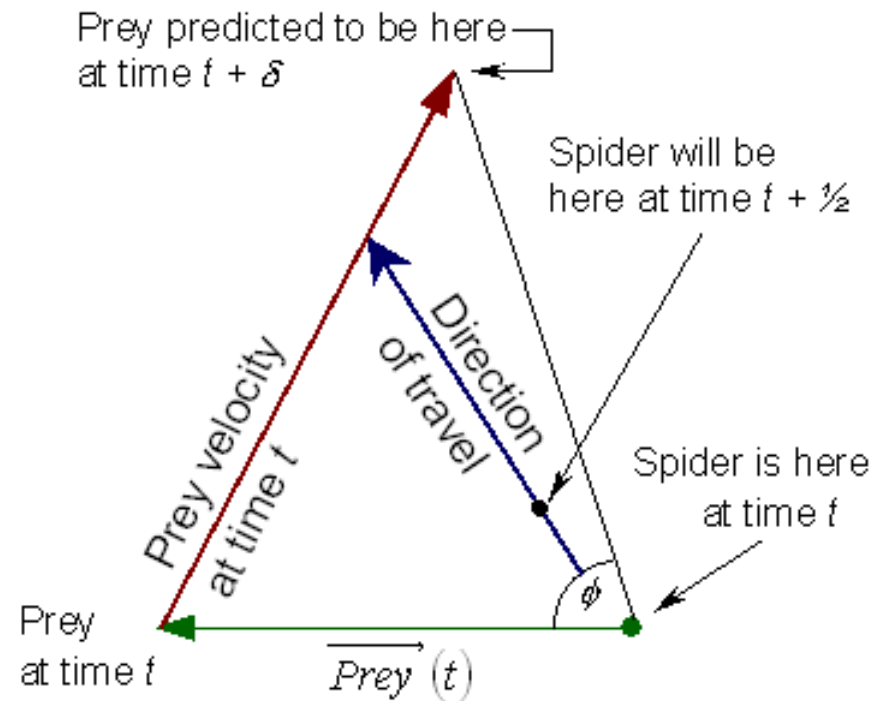
phidippus princeps



(Hill, 2001)



(Hill, 2001)

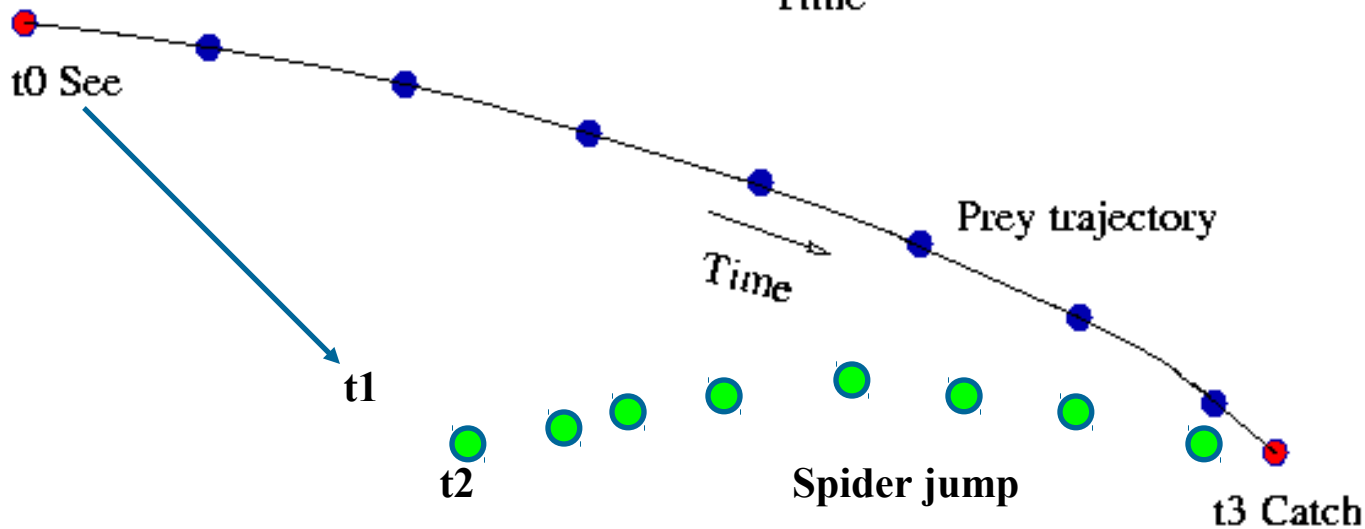
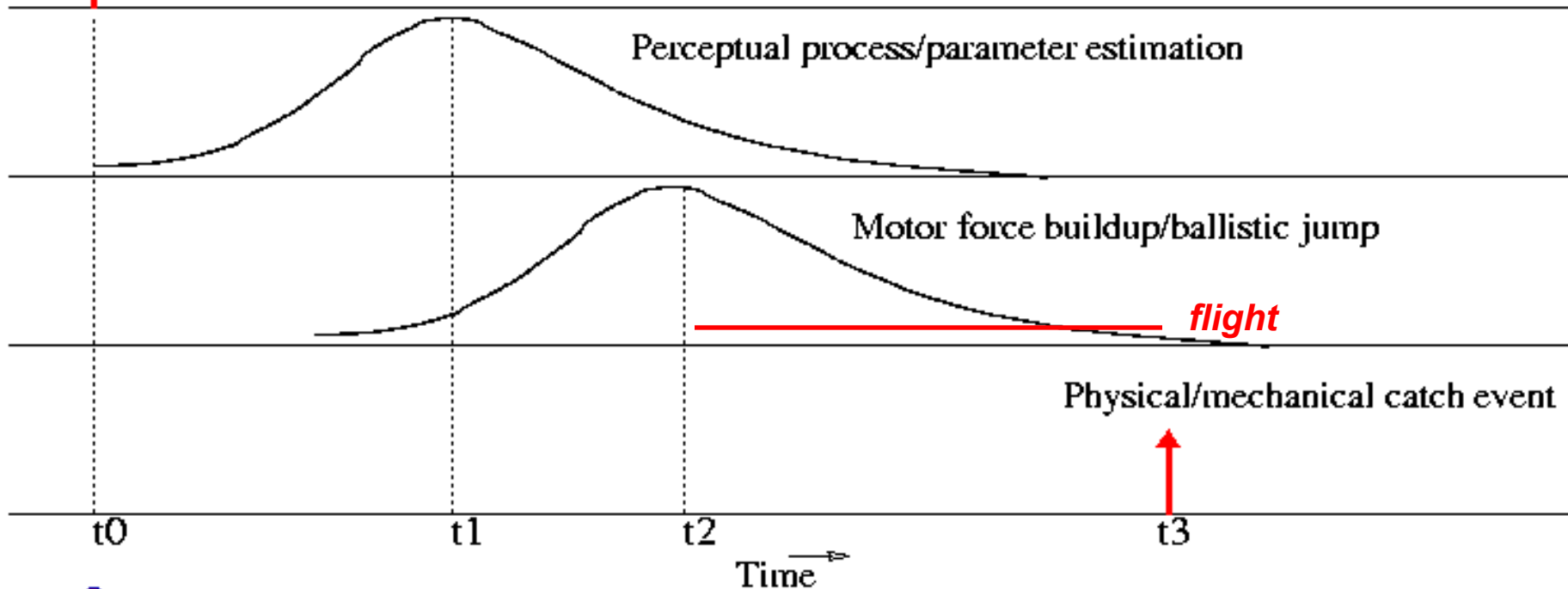


(Forster & Forster, 1999)

Properties of the spider jump

- Determination of prey velocity on the basis of optic flow
- **Preparation** of the muscle contraction amplitude, direction and timing,
in advance
- Jump
- Flight (almost no trajectory corrections possible!)
- Catch or miss

Physical/optical detection event



The spider jump ...

- is **not** purely reactive (i.e. non Brooksian)
- the jump is planned in a **pro-active** manner
- towards a position where there is **no visual percept** of the prey
- estimating a future time of arrival
- → there must be a represented estimate of a predicted state in the future

System models: stateless, reactive

- $A = F(P)$

Reactive, with perceptual memory

- $A = F(P_{[t_0, t]})$

reactive with perceptual and action memory

- $A = F(P_{[t_0, t]}, A_{[t_0, t-\Delta t]})$

proactive, with perceptual and action memory
and prediction window for perception and action

- $$A = F(P_{[t_0, t]}, A_{[t_0, t-\Delta t]}, P_{[t, t+dt]}, A_{[t, t+dt]})$$

proactive, with perceptual and action memory
and prediction window for perception and action

$$\blacksquare A = F(P_{[t_0, t]}, A_{[t_0, t-\Delta t]}, P_{[t, t+dt]}, A_{[t, t+dt]})$$

Prediction of the future perceptual and motor state is essential when there is any form of time delay within or outside the agent.

System models

- $A = F(P)$
- $A = F(P_{[t_0, t]})$
- $A = F(P_{[t_0, t]}, A_{[t_0, t-\Delta t]})$
- $A = F(P_{[t_0, t]}, A_{[t_0, t-\Delta t]}, \underline{P_{[t, \dots]}, A_{[t, \dots]}})$

cf: frontal and
prefrontal
cortex in
primates

Example: The non-linear IIR

IIR = infinite impulse response

$$y(t+\Delta t) = \mathbf{F} \left(\sum_{\tau} w_{\tau} x(t-\tau), \sum_{\tau} v_{\tau} y(t-\tau) \right)$$

Example: The multipurpose non-linear IIR

$$\mathbf{y}(t+\Delta t) = \mathbf{F} \left(\sum_{\tau} w_{\tau} \mathbf{x}(t-\tau), \sum_{\tau} v_{\tau} \mathbf{y}(t-\tau) \right)$$

“the next action is a non-linear function of (1) the weighted sum of *things x seen until now* and (2) the weighted sum of *things y done until now*”

Example: The multipurpose non-linear IIR

$$\mathbf{y}(t+\Delta t) = \mathbf{F} \left(\sum_{\tau} \alpha_{\tau} \mathbf{x}(t-\tau), \sum_{\tau} \beta_{\tau} \mathbf{y}(t-\tau) \right)$$

“the next action is a non-linear function of (1) the weighted sum of *things x seen until now* and (2) the weighted sum of *things y done until now*”

(it can be used for modeling a plethora of processes in physics, engineering and biology)

Conclusion (1)

- Behavior may be determined by simple rules
- *but the complexity of the brain is apparent (?)*

- Some may want to do away with representation
- *but neural representation is the essence of cognitive neuroscience*

Conclusion (2)

- Even “simple” animals may need to estimate the state of the world in the future

this can only be realized if a persistent representation of the relevant facets of that world is available for prediction