

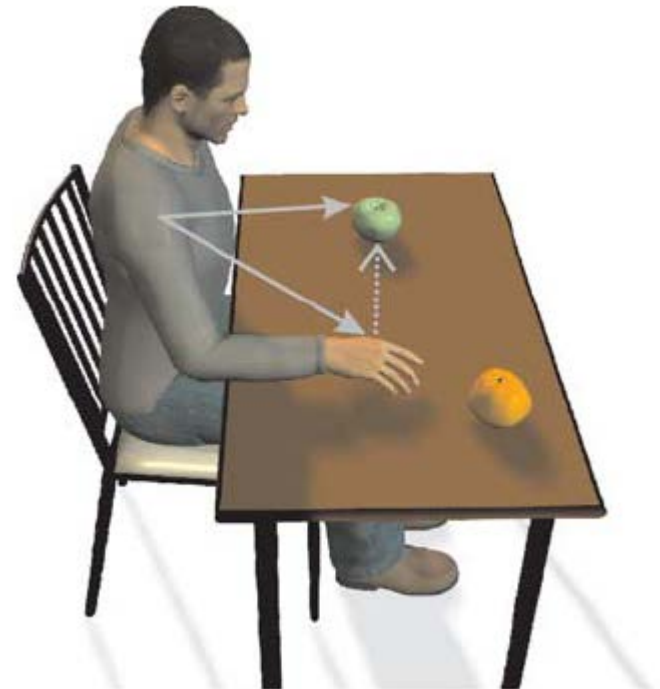
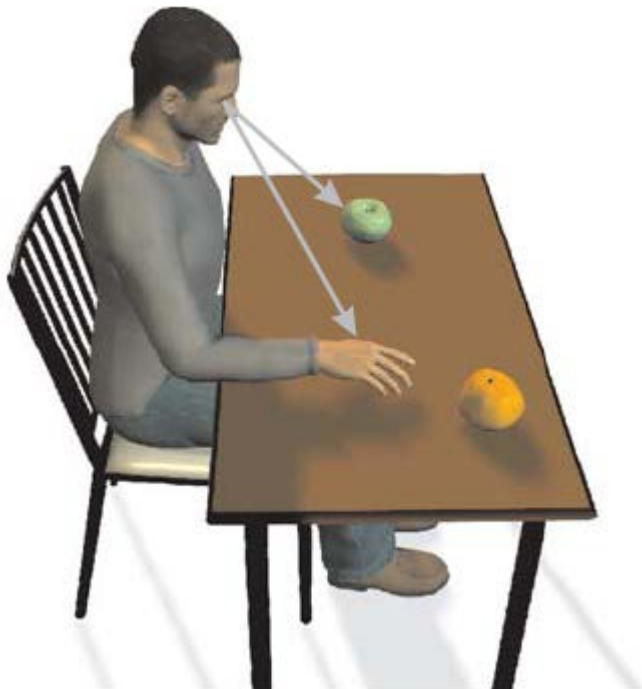
Sensory-motor control intro

Day 1

CoSMo 2017
Gunnar Blohm

Motor planning & execution

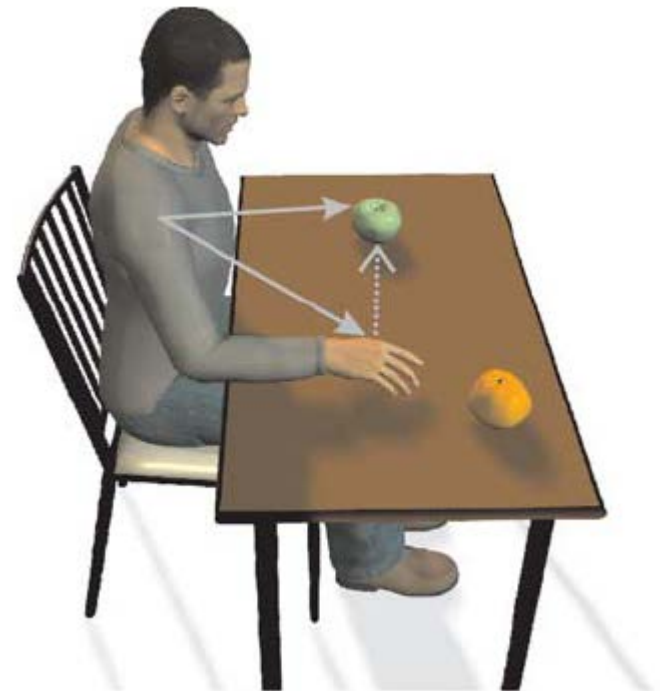
- ▶ Processes involved in the sensory guidance of action



Blohm et al. 2009

Motor planning & execution

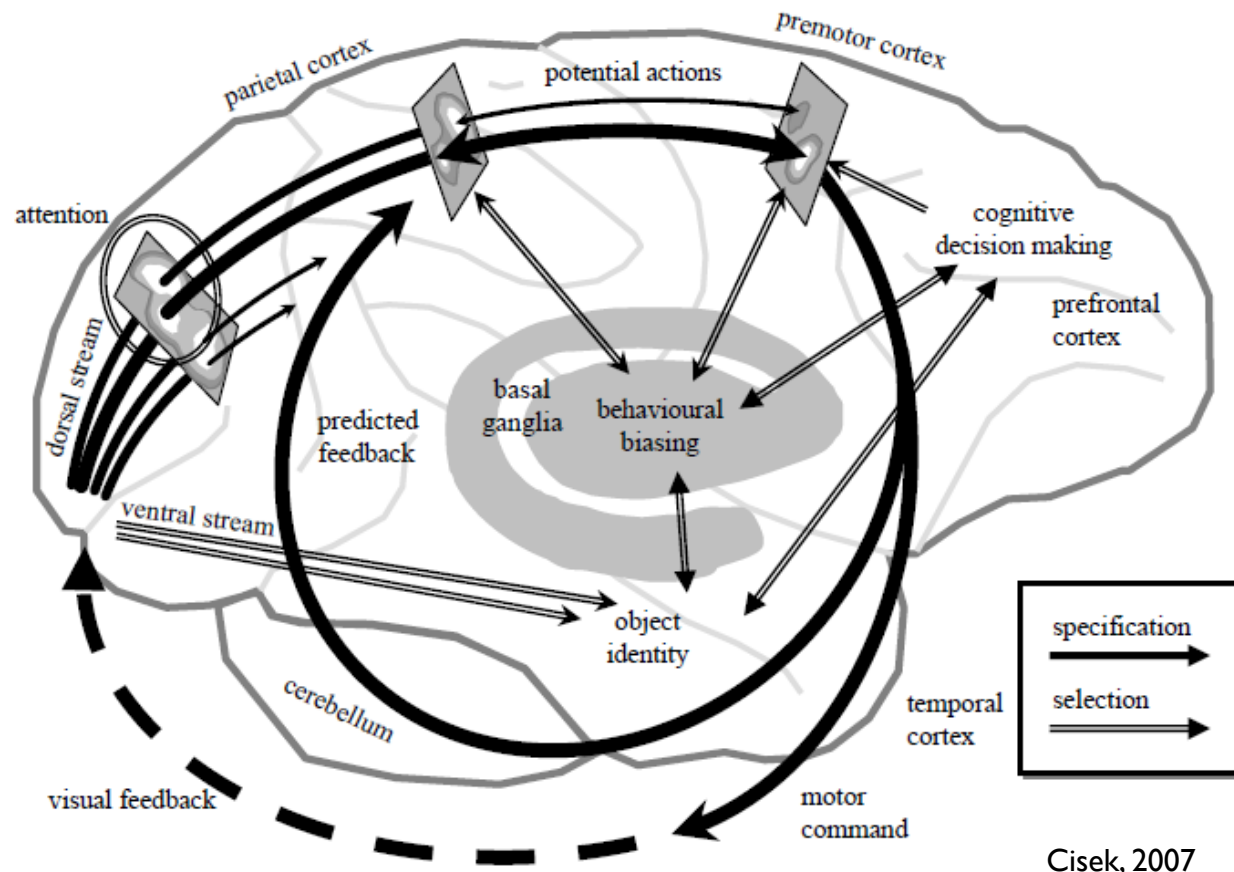
- ▶ Processes involved in the sensory guidance of action
 - ▶ Sensory processing
 - ▶ Multi-sensory integration
 - ▶ Reference frame transformations
 - ▶ Target selection
 - ▶ Decision making
 - ▶ Move or not
 - ▶ Which effector, which target
 - ▶ When to move (timing)
 - ▶ Motor planning
 - ▶ Motor control
 - ▶ Error corrections...



Blohm et al. 2009

Motor planning & execution

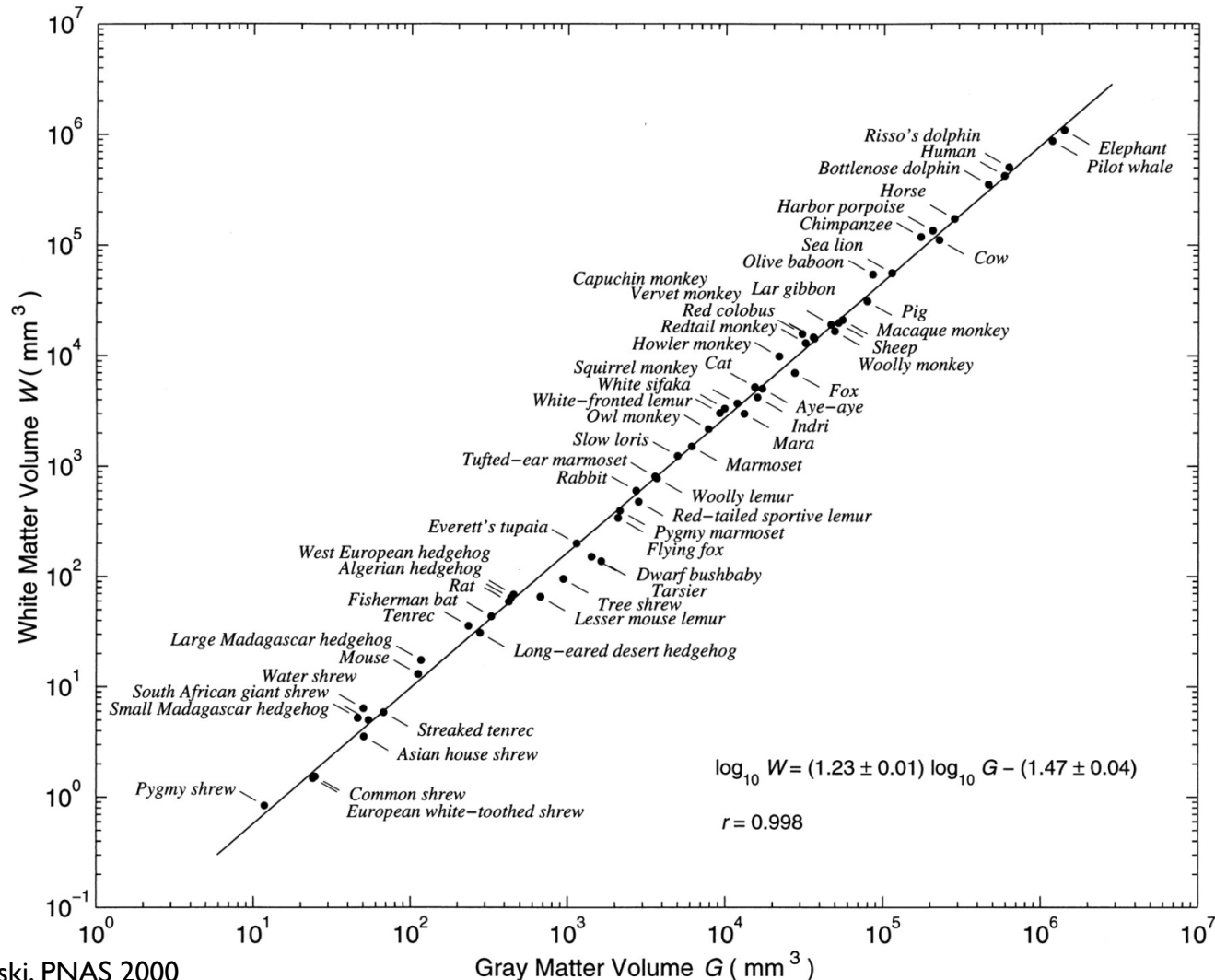
- Processes involved in the sensory guidance of action



The computational anatomy of sensory-motor control

Hierarchies

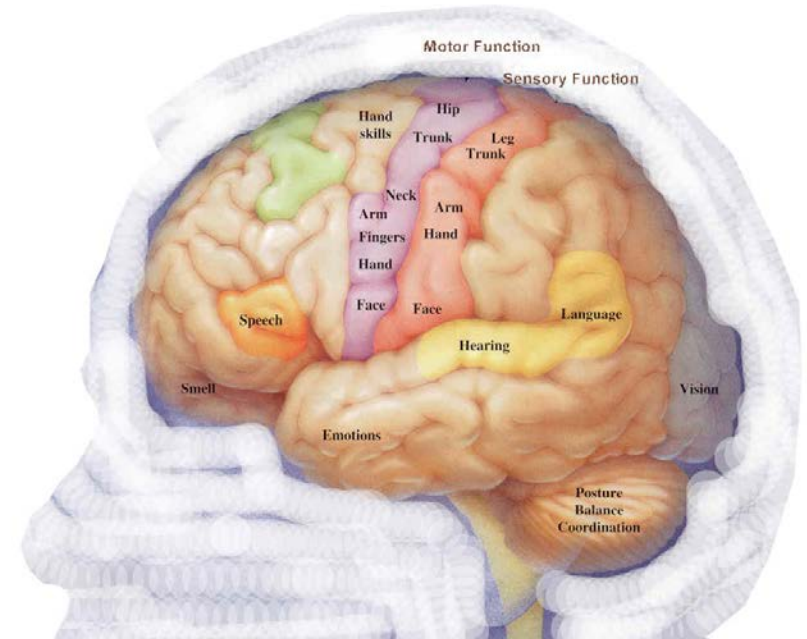
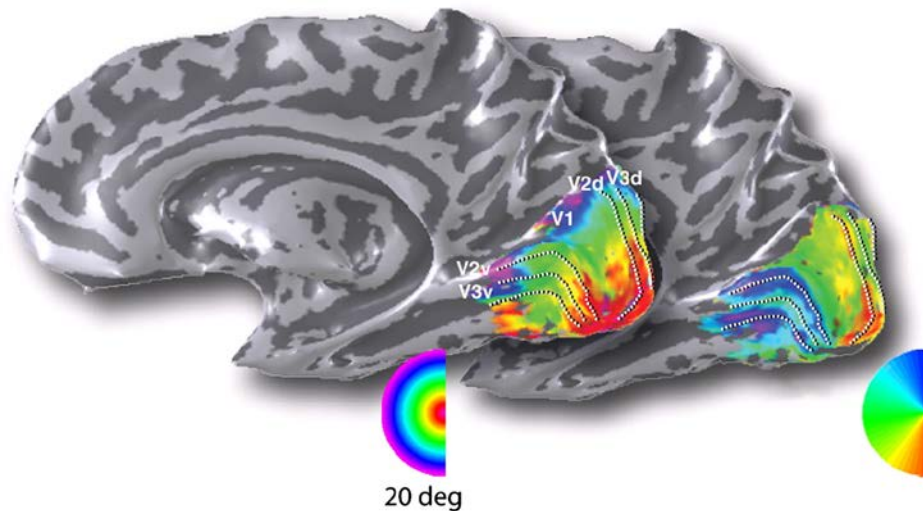
Computational anatomy of the brain



Zhang & Sejnowski, PNAS 2000

Computational hierarchy of the brain

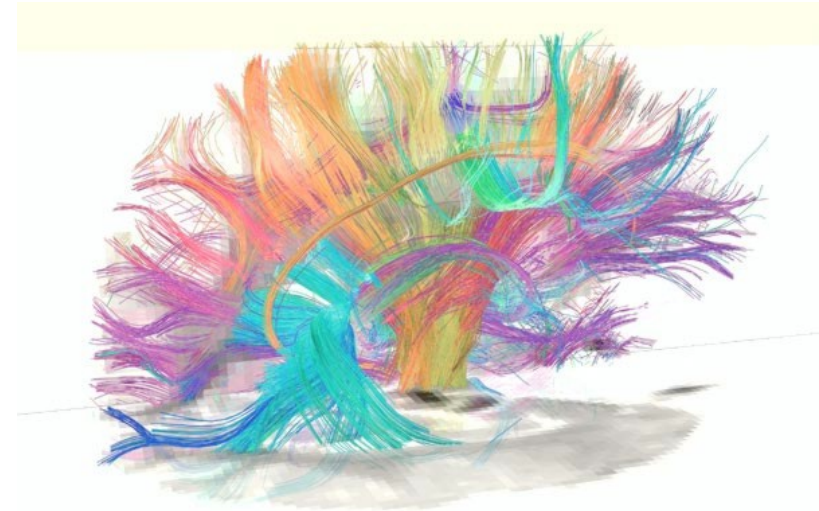
- ▶ Brodmann's areas
- ▶ Functional areas
- ▶ Maps



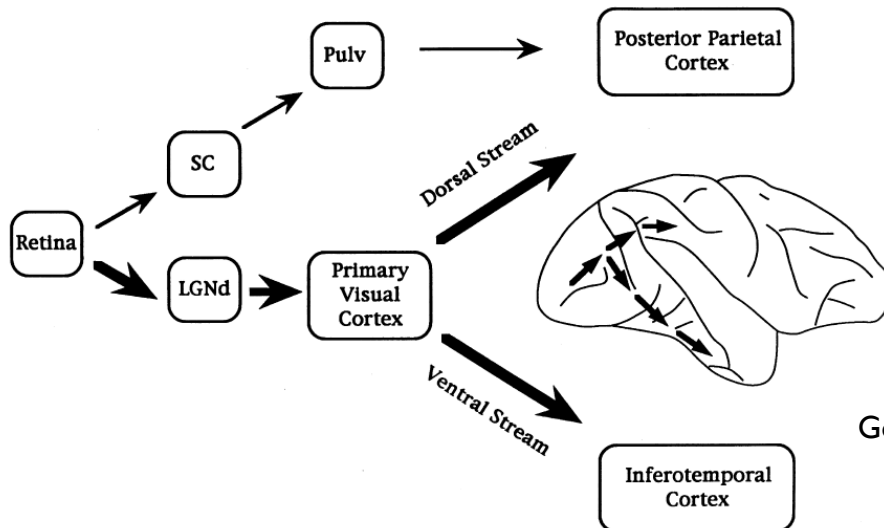
Wandell, et al., Neuron (2007)

Computational hierarchy of the brain

- ▶ Brodmann's areas
- ▶ Functional areas
- ▶ Maps
- ▶ Connectivity
- ▶ Functional pathways



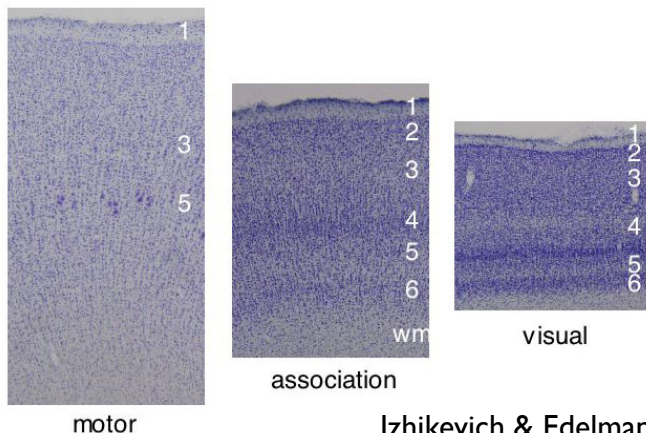
Courtesy of Kat Reinhart



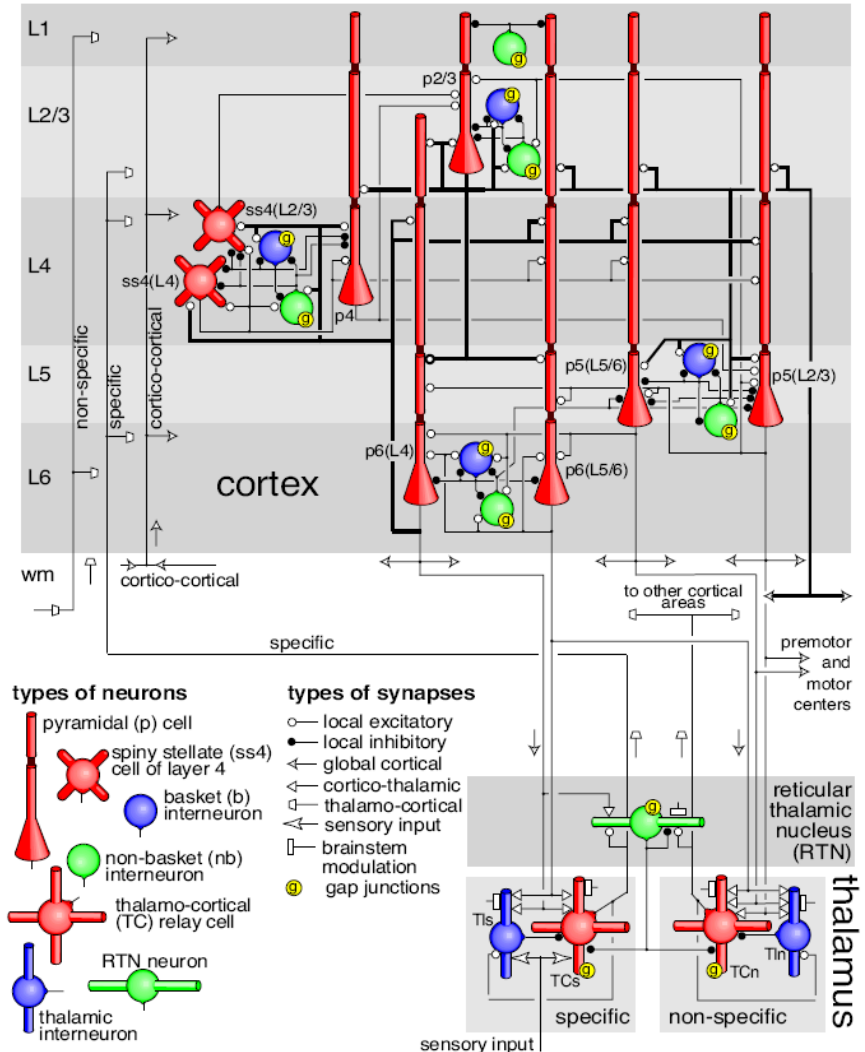
Goodale & Humphrey, 1998

Computational hierarchy of the brain

- ▶ Brodmann's areas
- ▶ Functional areas
- ▶ Maps
- ▶ Connectivity
- ▶ Functional pathways
- ▶ Detailed structure

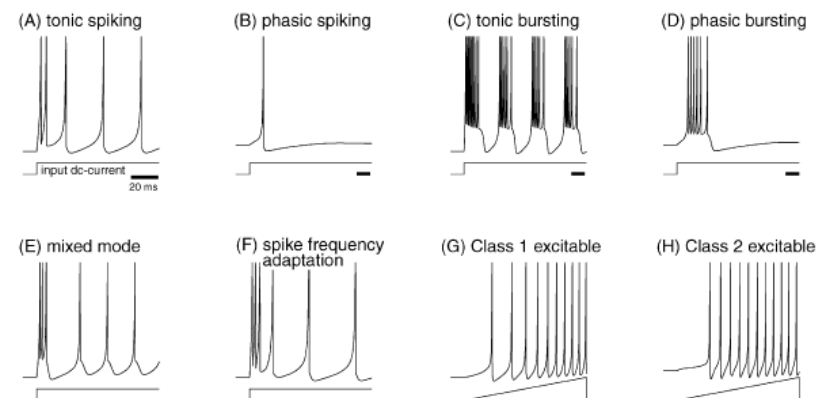
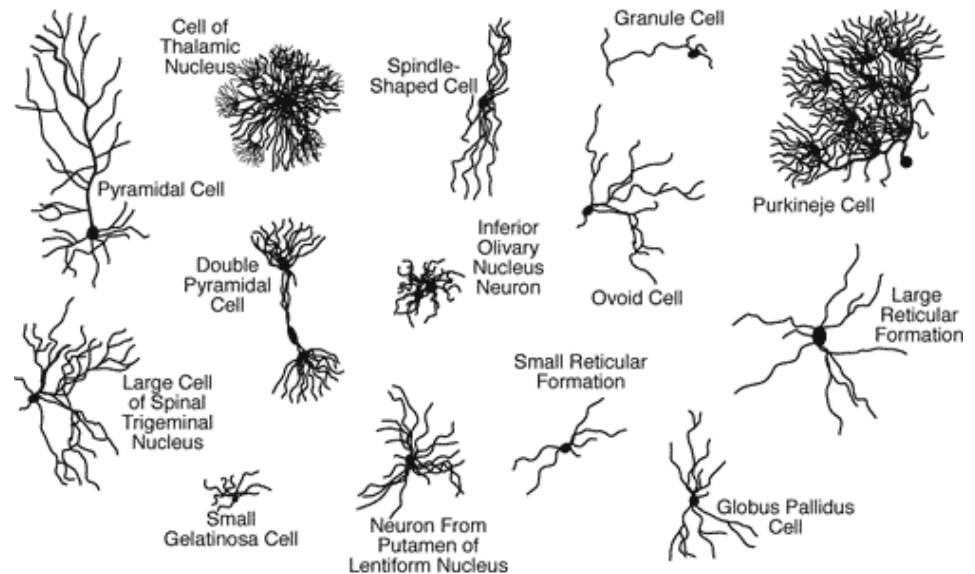


Izhikevich & Edelman PNAS (2008)



Computational hierarchy of the brain

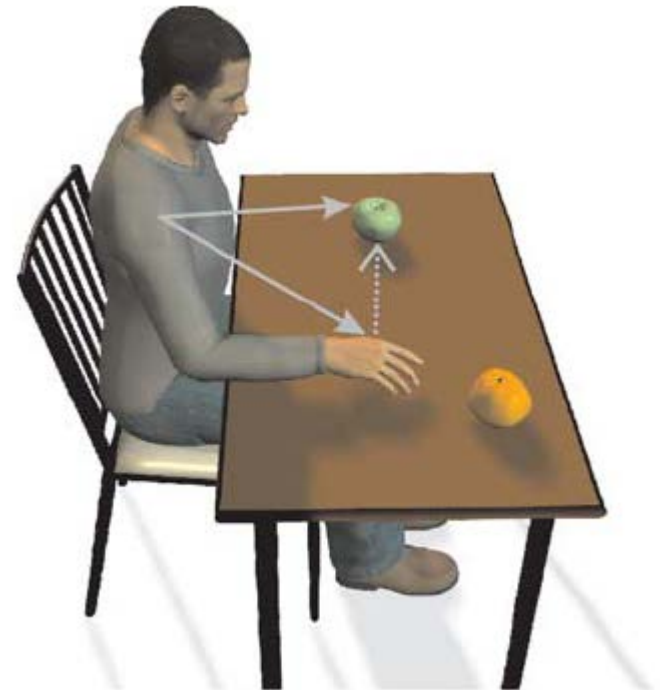
- ▶ Brodmann's areas
- ▶ Functional areas
- ▶ Maps
- ▶ Connectivity
- ▶ Functional pathways
- ▶ Detailed structure
- ▶ Varied anatomy
- ▶ Heterogeneous behavior
- ▶ Chemical & molecular complexity
- ▶ ...



Izhikevich, IEEE Transactions on Neural Networks (2004)

Motor planning & execution

- ▶ Processes involved in the sensory guidance of action
 - ▶ Sensory processing
 - ▶ Multi-sensory integration
 - ▶ Reference frame transformations
 - ▶ Target selection
 - ▶ Decision making
 - ▶ Move or not
 - ▶ Which effector, which target
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 - ▶ Error corrections...



Blohm et al. 2009

Sensory-motor transformations

Blohm

Sensory-motor transformations

- ▶ Justin: DLR robot – ball catching
 - ▶ Sensory ref frames \sim motor ref frame...
 - ▶ Sensory code \sim motor code...
 - ▶ Movie...

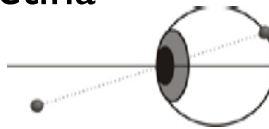


Sensory-motor transformations

► Reference frames

► Determined by sensory and motor apparatus

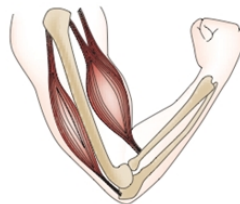
- Vision: attached to the retina



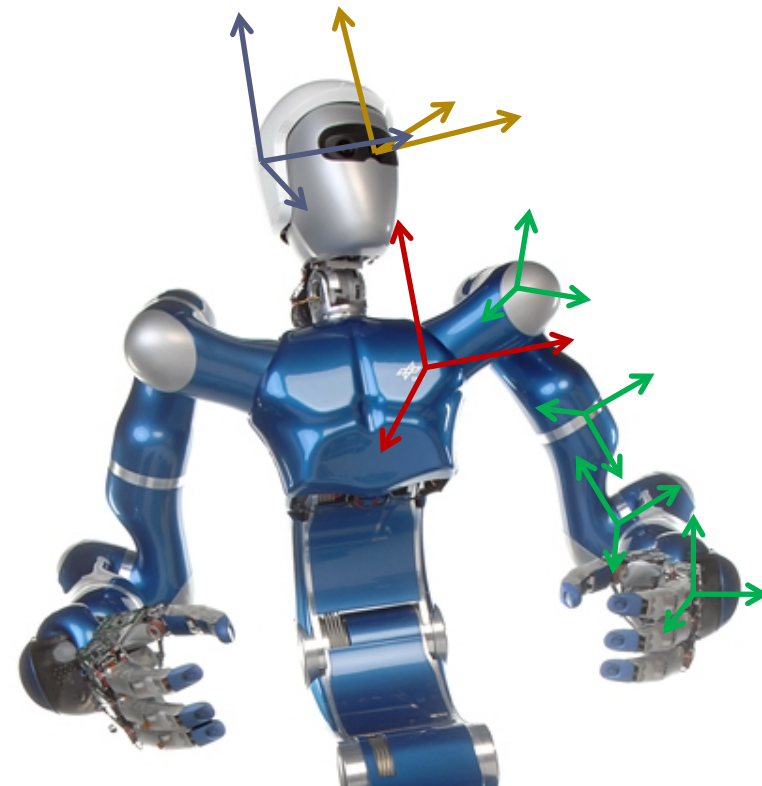
- Audition: attached to the head



- Proprioception: relative joint angles



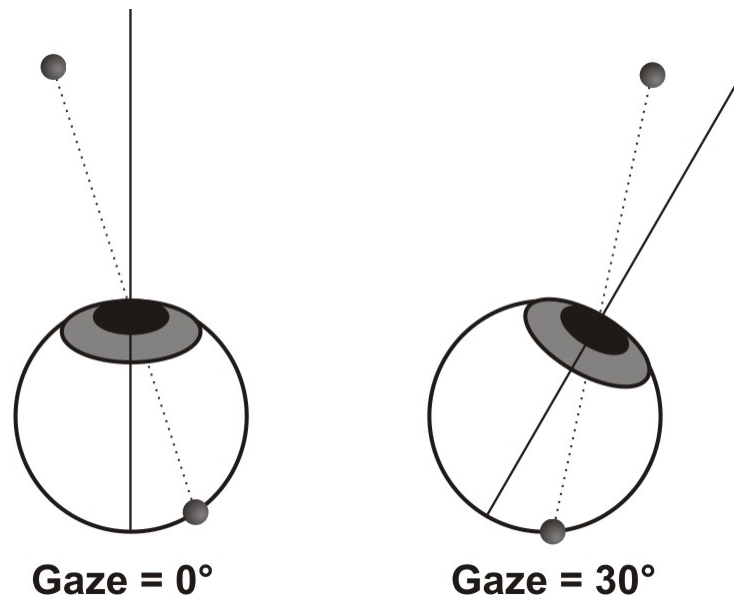
- Arm movement: relative to attachment at shoulder



Sensory-motor transformations

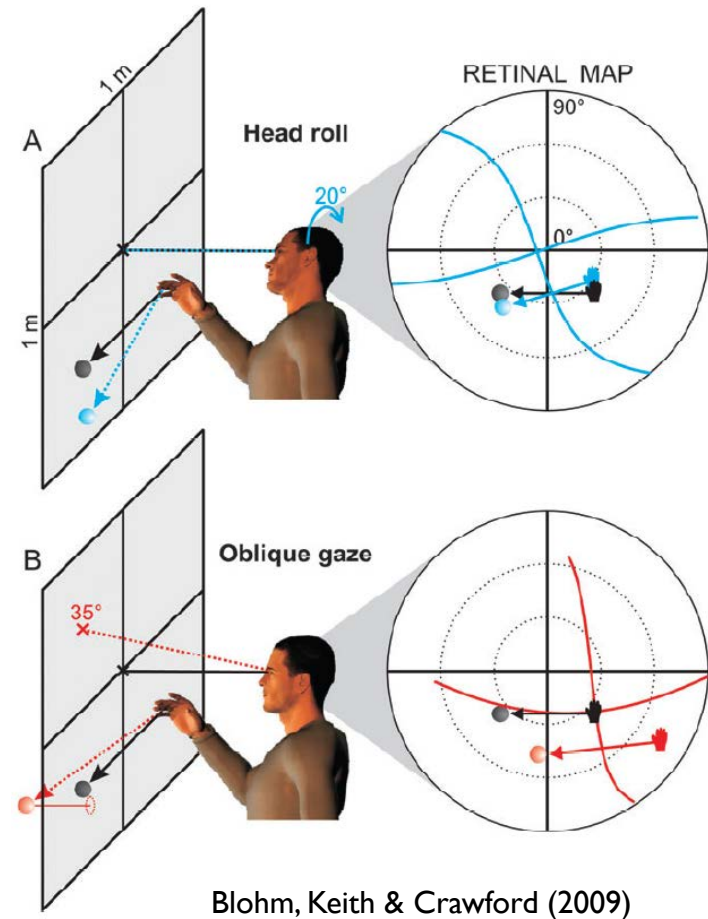
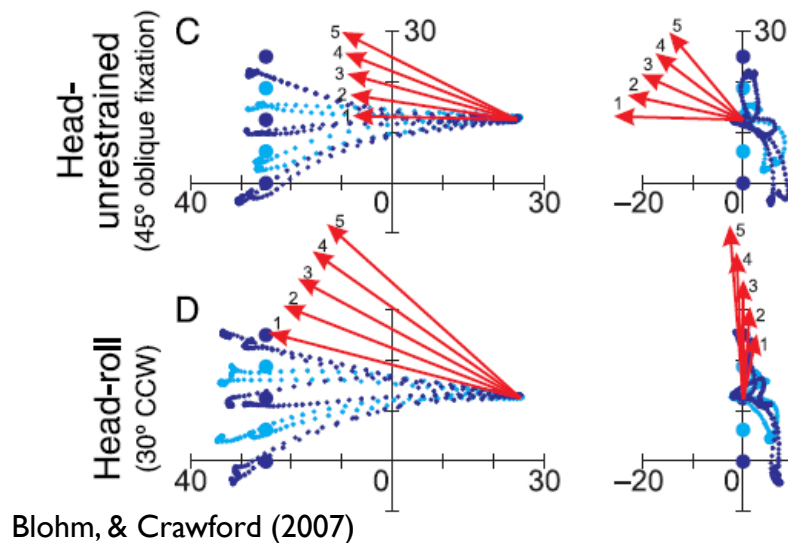
▶ Reference frames

- ▶ Knowledge about reference frames is required to localize sensory and motor events
 - ▶ Same retinal image – different spatial locations



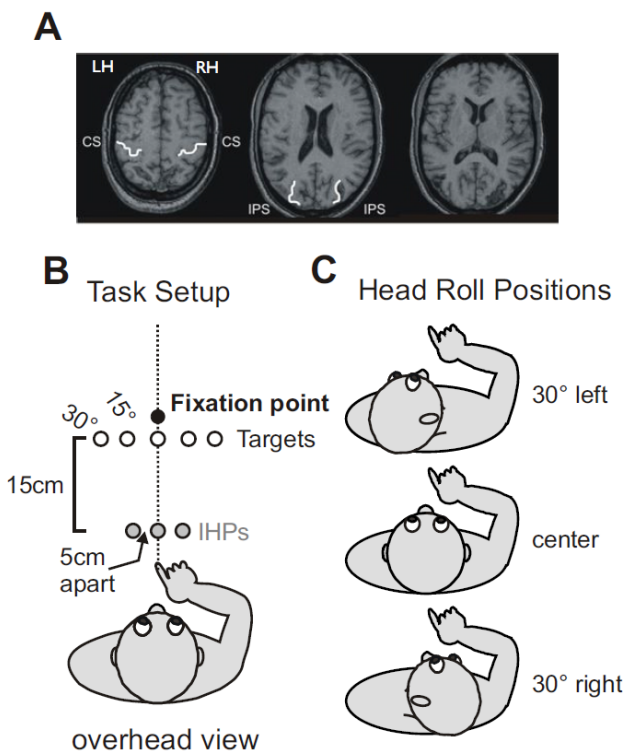
Examples: reference frame transformations

► Reaching / pointing

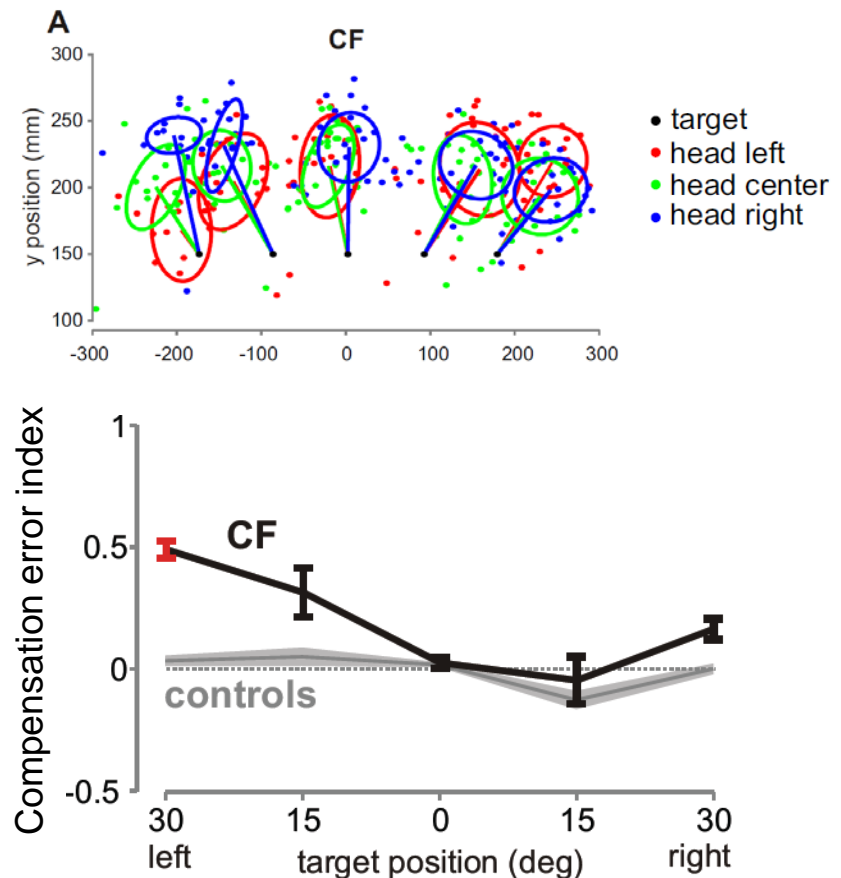


Examples: reference frame transformations

► Reference frame transformation deficits in optic ataxia



Khan, Pisella, Blohm (2013)



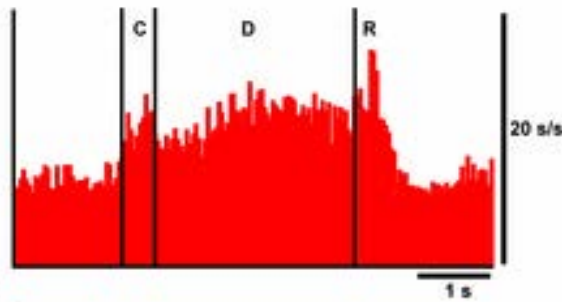
Working memory

- ▶ Retain important information for immediate action
 - ▶ Relies on persistent mnemonic activity through a balance of local excitation and global inhibition

Macaque



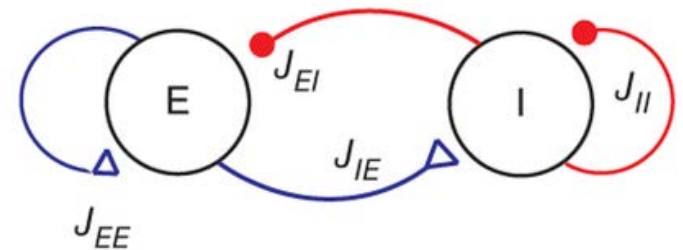
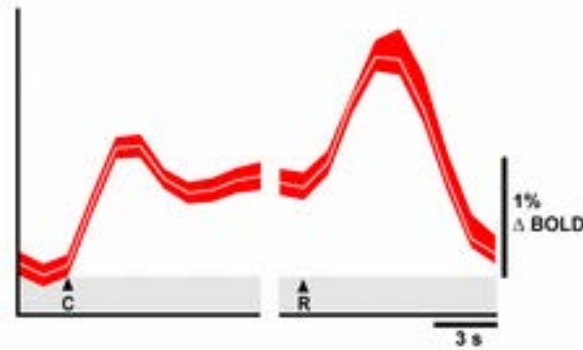
Funahashi, Bruce, Goldman-Rakic, 1989



Human



Strikal & Curtis, 2008

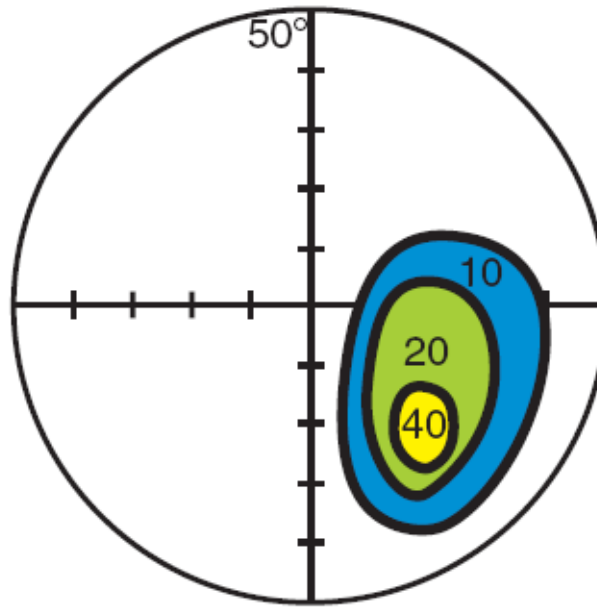


Lim & Goldman (2013)

Coding information in the brain

► Receptive fields

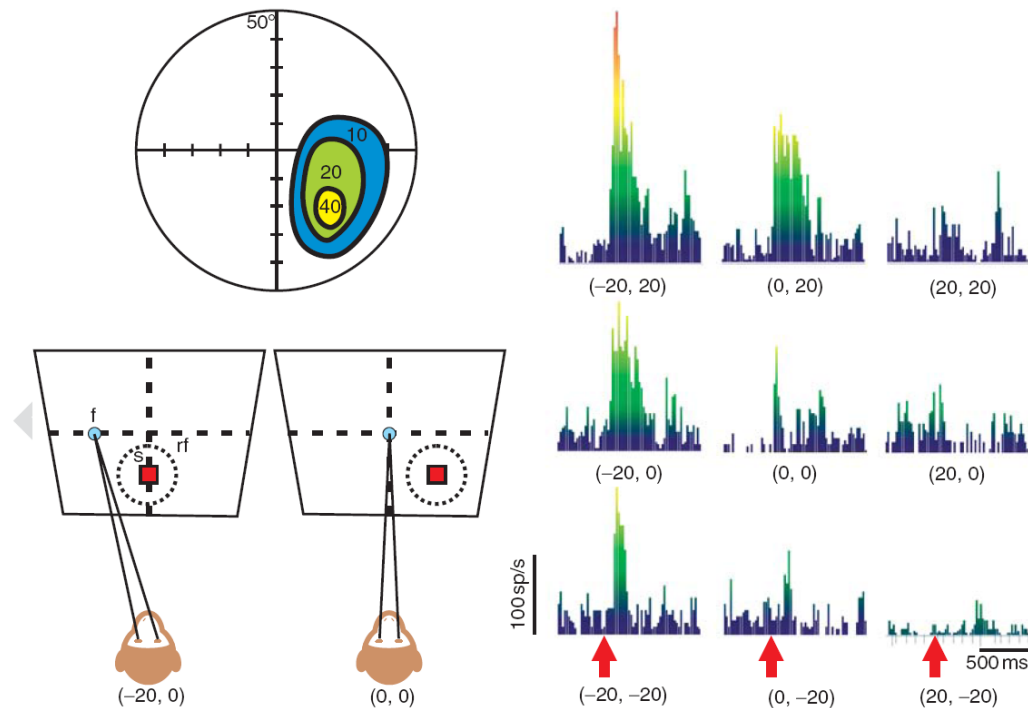
- = activation pattern of a neuron for targets across space
 - We assume that the brain explicitly “codes” certain information
 - AND that we can decode it!



Blohm, Khan, Crawford, 2009 (adapted from Andersen, et al., 1985)

Gain modulation

- ▶ = change of receptive field strength with secondary input
 - ▶ E.g. eye position gain modulation of visual receptive fields in posterior parietal cortex

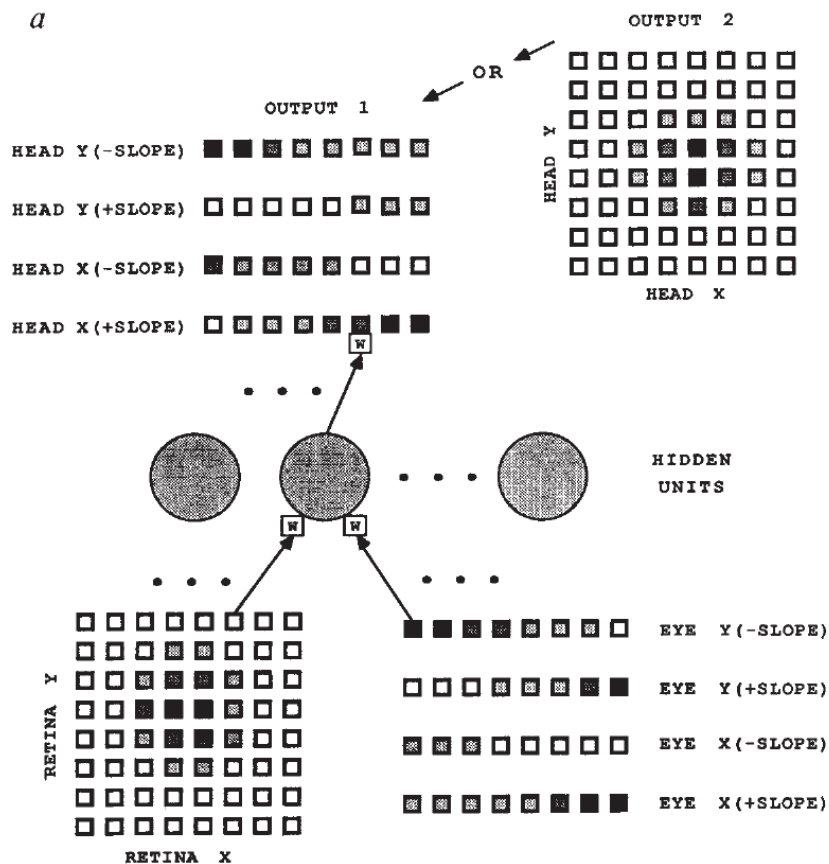


Blohm, Khan, Crawford, 2009 (adapted from Andersen, et al., 1985)

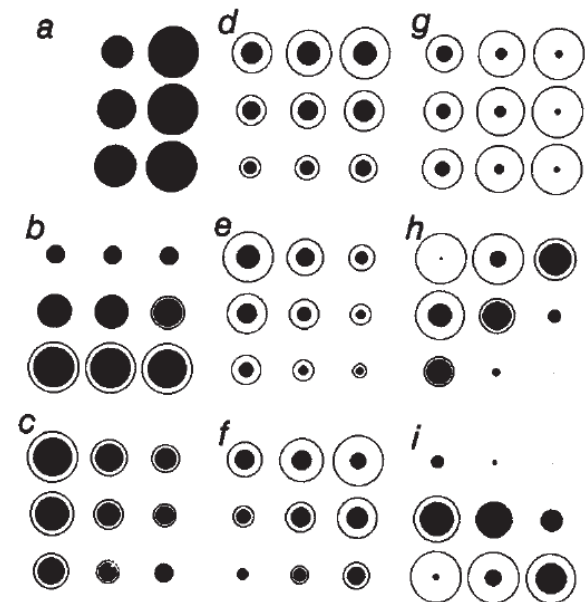
Gain modulation

► Reference frame transformations

► Zipser & Andersen, Nature 1988

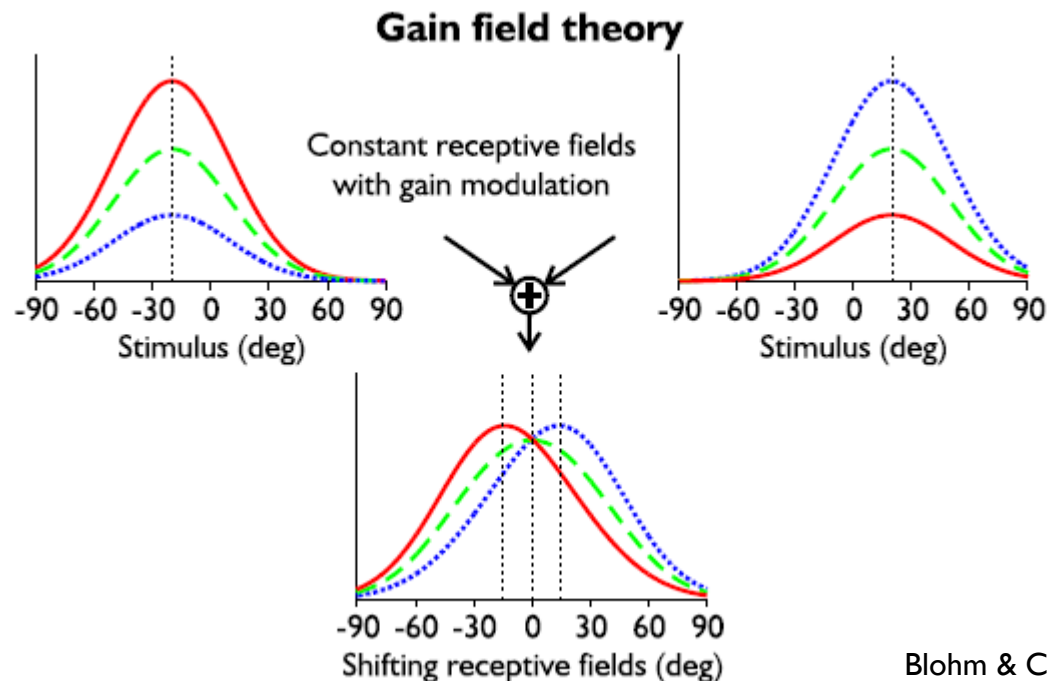


Eye position gain modulation of hidden layer units



Gain modulation

- ▶ Powerful computational means for
 - ▶ Cue combination
 - ▶ Reference frame transformations
 - ▶ Multi-sensory integration...



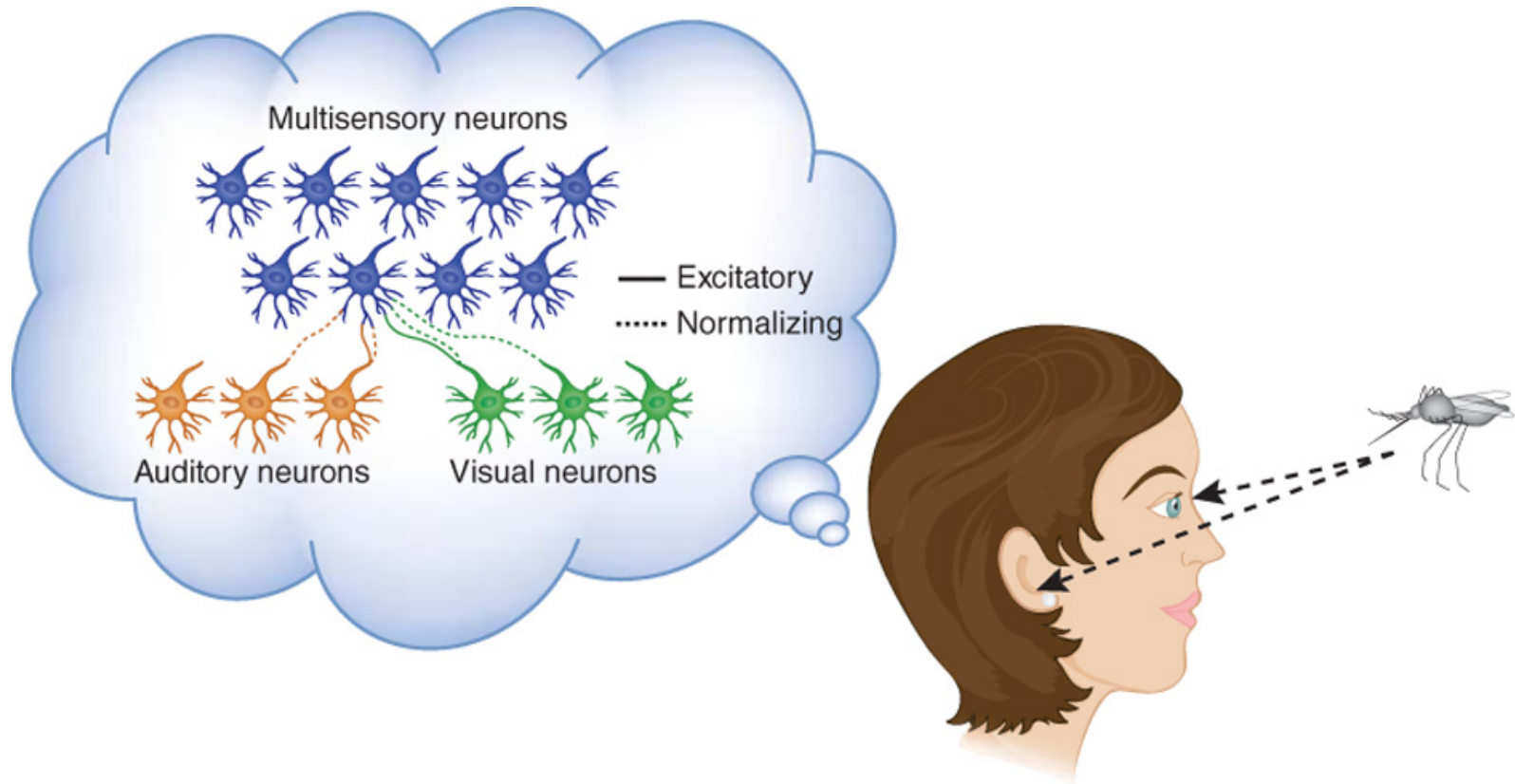
Blohm & Crawford, 2009

A mosaic artwork is displayed within a brick archway. The central figure is a nude person, possibly a woman, seen from behind, standing in a room. The room is composed of various colored blocks, including blue, orange, yellow, and green, creating a pixelated or cubist effect. The figure is surrounded by these blocks, which seem to form the walls and floor of the room. The overall style is reminiscent of a modern mosaic or a digital art piece.

Multi-sensory integration

Kording & Schrater
Drugowitsch

Multi-sensory integration



Churchland (2011)

Mathematical framework for Bayesian integration

► Cue combination

- Optimal Bayesian observer $p(X | A, V) = \frac{p(A, V | X) \cdot p(X)}{p(A, V)}$
- Independent observations A, V

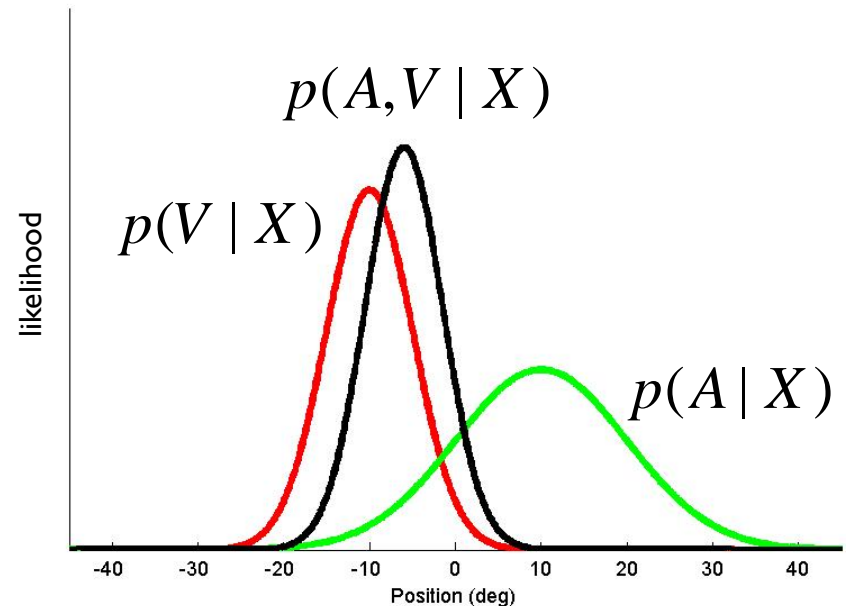
$$p(A, V | X) = p(V | X) \cdot p(A | X)$$

- If uniform priors, then

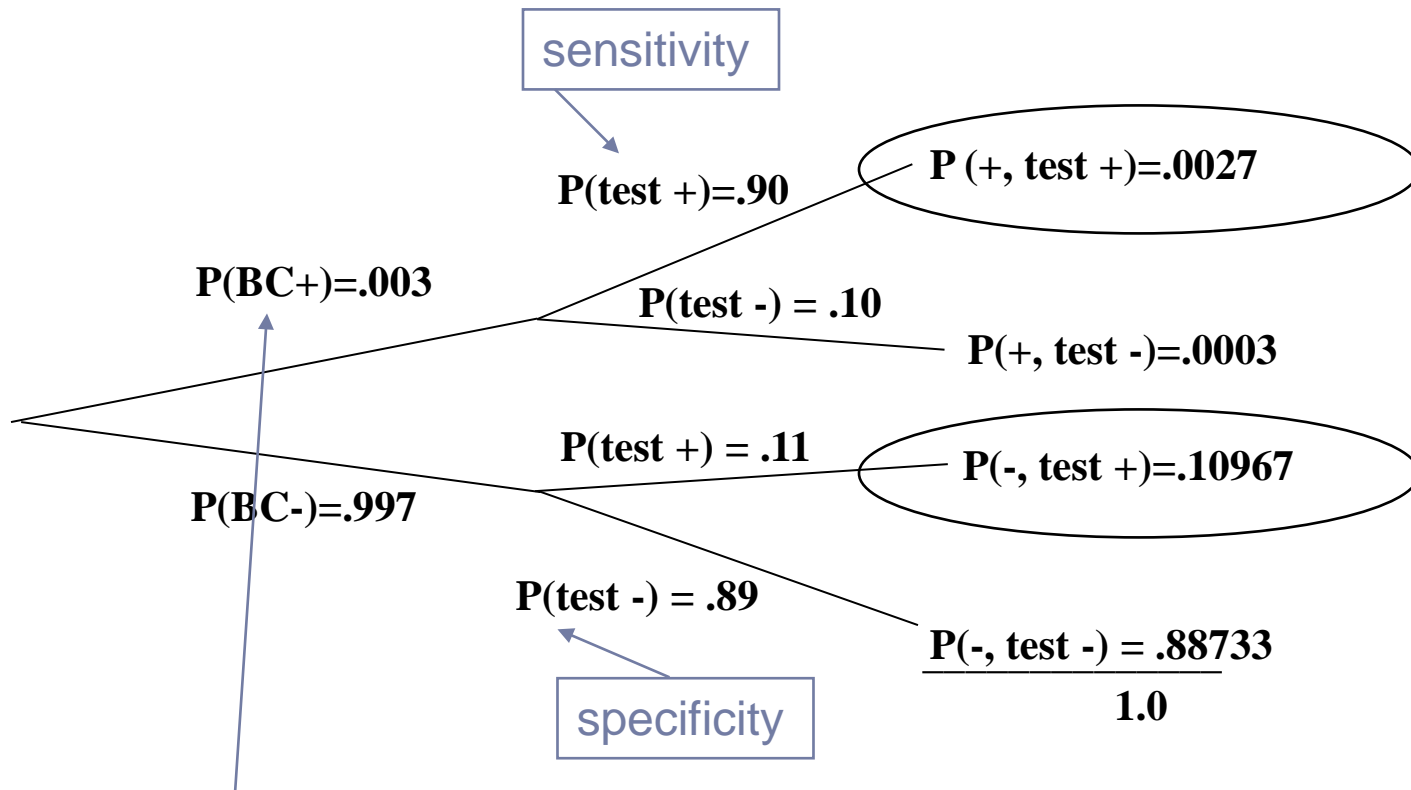
$$p(X | A, V) \propto$$

$$p(V | X) \cdot p(A | X)$$

- The brain always uses all available useful information.
- Information from different sources is combined in a statistically optimal fashion



Example: breast cancer

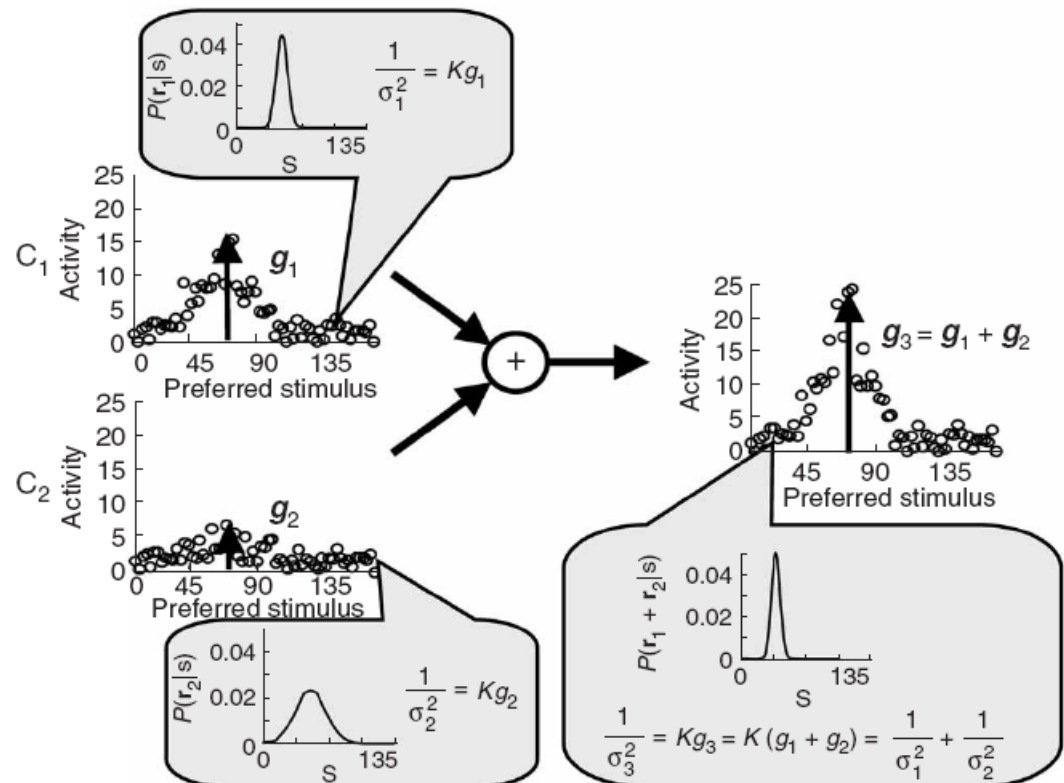


Marginal probabilities of breast cancer....(prevalence among all 54-year olds)

$$P(\text{BC}/\text{test}+) = .0027 / (.0027 + .10967) = 2.4\%$$

Bayesian computations in population codes

- ▶ Representing uncertainty with population codes
 - ▶ Probabilistic population codes
 - ▶ Poisson-like neural noise
 - ▶ Variance inversely related to gains of population code



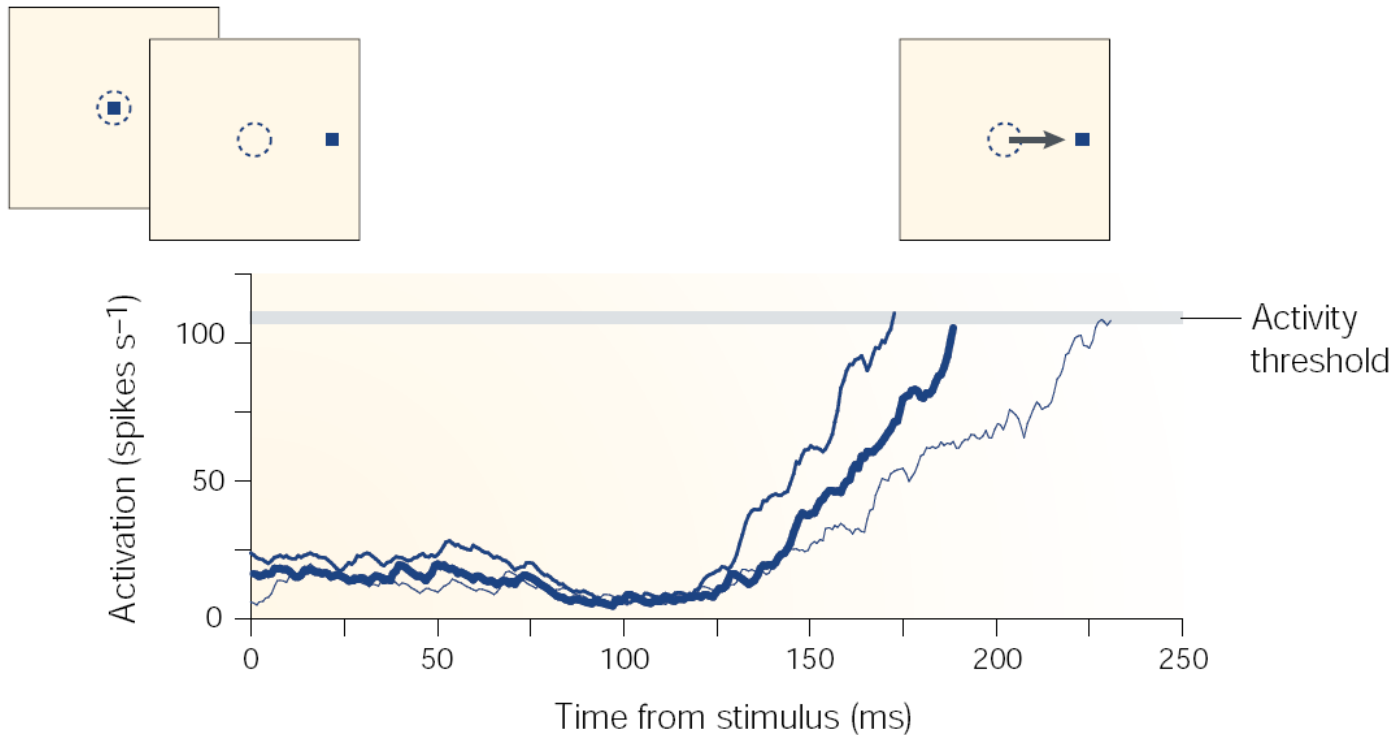
Ma et al. (2006)

Target selection & decision making

Drugowitsch, Schrater

Choice & decisions

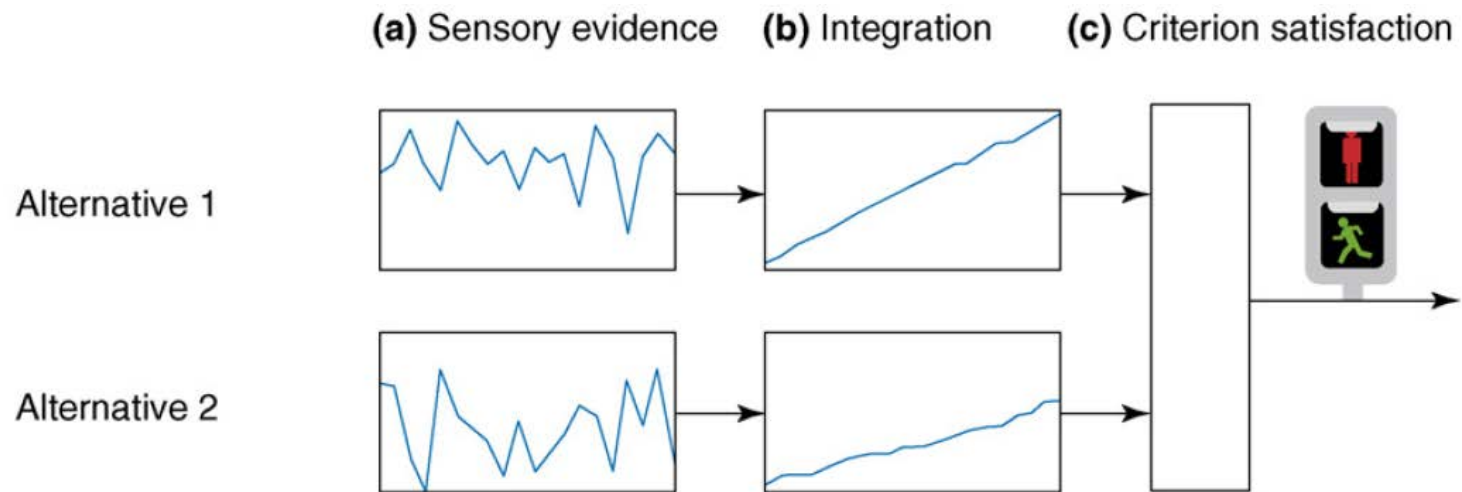
- ▶ Competition of alternatives
- ▶ Example: areas LIP / FEF



Schall, 2001

Current theories

- ▶ Target selection
 - ▶ Through the interaction of bottom-up and top-down attention
- ▶ Decision making
 - ▶ Different versions of rise-to-threshold models



Bogacz, 2007

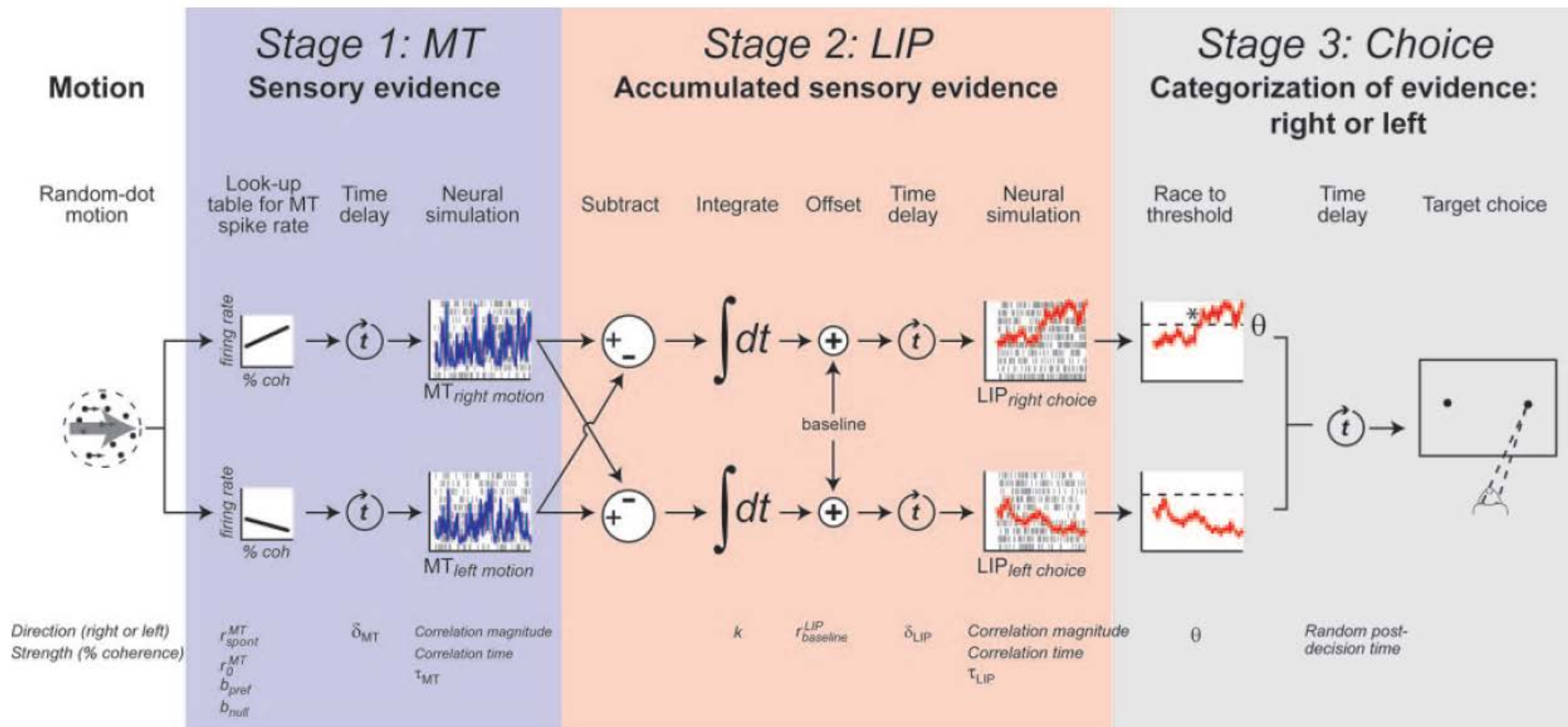
TRENDS in Cognitive Sciences

Current theories

- ▶ **Target selection**
 - ▶ Through the interaction of bottom-up and top-down attention
- ▶ **Decision making**
 - ▶ Different versions of rise-to-threshold models
- ▶ **Computational principles at work?**
 - ▶ Competitive processing
 - ▶ Divisive normalization
 - ▶ Gain modulation

Diffusion models

- ▶ Example: left-right decisions
 - ▶ Integrated decision model (Mazurek, et al. 2003)

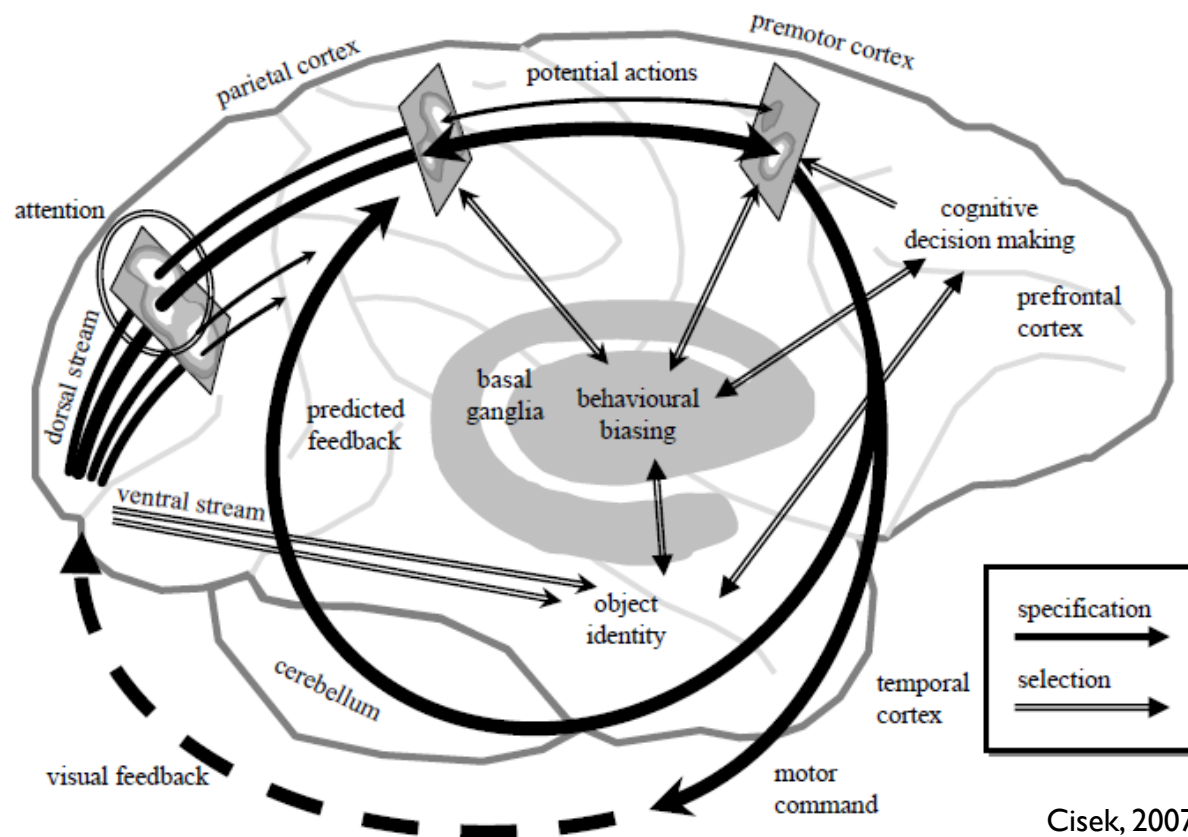


Motor planning, motor control & error corrections

Kording, Schrater, Blohm
Shadmehr & Ahmed
Sternad

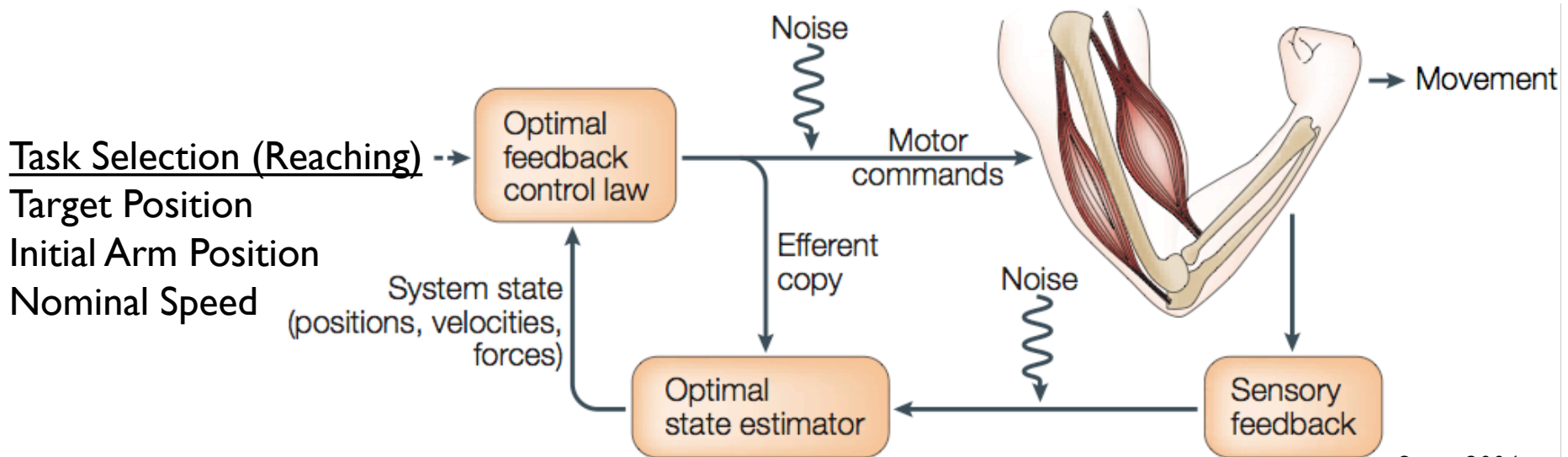
Motor planning & execution

- Planning, execution and error correction are all part of the same system



Motor planning & control

- ▶ Motor planning is the result of all previous steps...
 - ▶ Sensory processing
 - ▶ Transformations & multi-sensory integration
 - ▶ Target selection & decision making
- ▶ Motor control
 - ▶ Execution of the motor plan...



Scott, 2004

Eye/head plants

- ▶ Eye and neck muscles properties
- ▶ Damped spring-mass system equivalent

Equation of motion of eye ball:

$$J \cdot \frac{d^2\theta}{dt^2} + F_p = F_m$$

Muscle force applied:

$$F_m = F_0 - R_m \cdot \frac{d\theta}{dt} - \frac{R_m}{K_{se}} \cdot \frac{dF_m}{dt}$$

Robinson (1964), Scudder (2009)

Passive muscle/tissue force:

$$F_p = \frac{R_1 R_2 \cdot \frac{d^2\theta}{dt^2} + (R_1 K_2 + R_2 K_1) \cdot \frac{d\theta}{dt} + K_1 K_2 \cdot \theta - (R_1 + R_2) \cdot \frac{dF_p}{dt}}{(K_1 + K_2)}$$

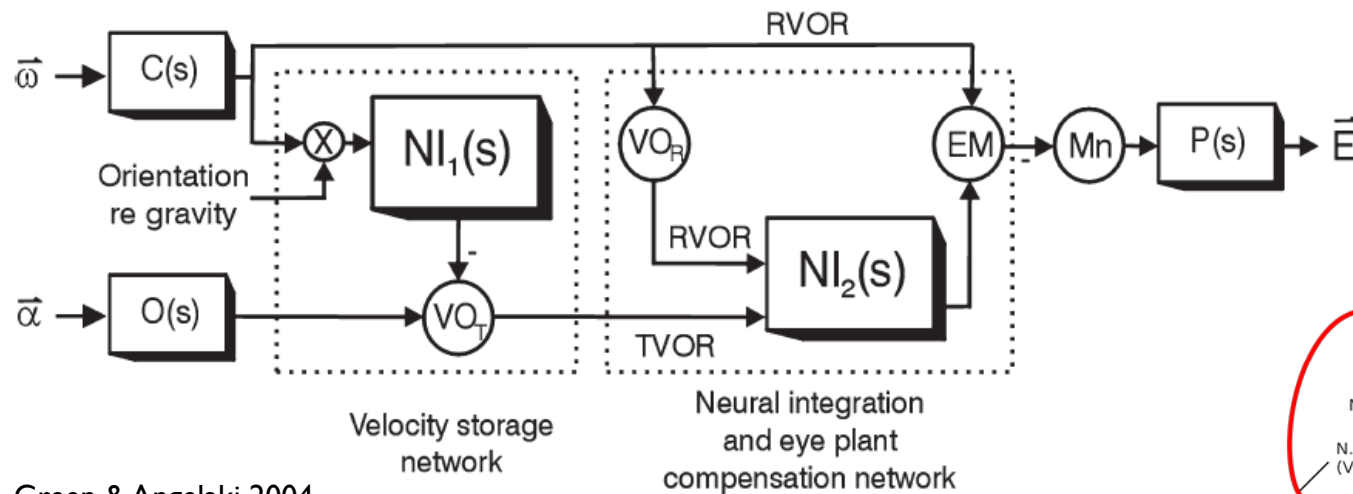
J: moment of inertia

F_p : passive force (muscle tissue)

F_m : active muscle force

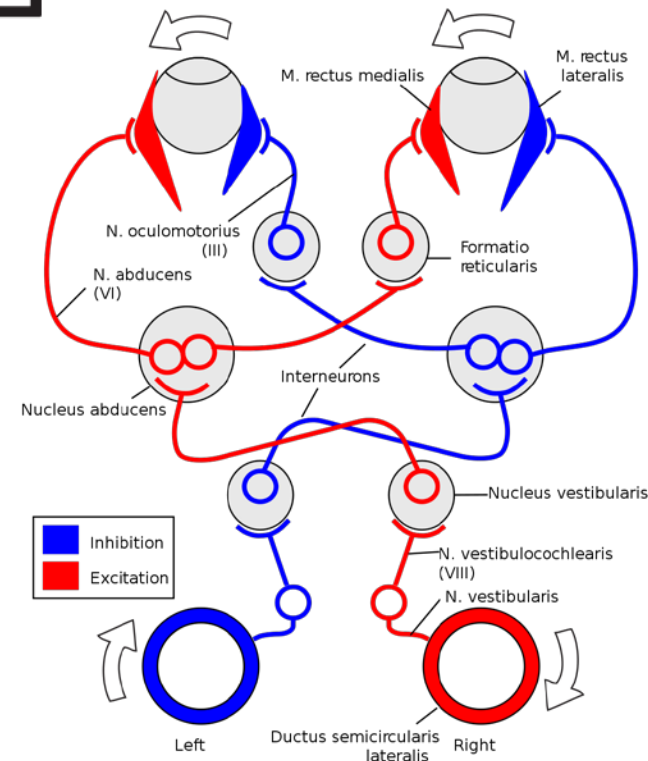
Classical control

► Vestibulo-ocular reflex



Green & Angelaki 2004

- ❑ Angular velocity from semi-circular canals: ω
- ❑ Linear acceleration senses by otolith organs: α
- ❑ VO: vestibular-only neurons
- ❑ EM: eye movement sensitive neurons



Optimal feedback control

Sensory state of our body and the world we interact with

$$\mathbf{x}^{(k+1)} = A\mathbf{x}^{(k)} + \overset{\text{motor command}}{C\mathbf{u}^{(k)}} + \underset{\text{motor noise}}{\boldsymbol{\varepsilon}_u^{(k)}}$$

What we can observe about the state

$$\mathbf{y}^{(k)} = B\mathbf{x}^{(k)} + \boldsymbol{\varepsilon}_y^{(k)}$$

sensory noise

Cost to minimize

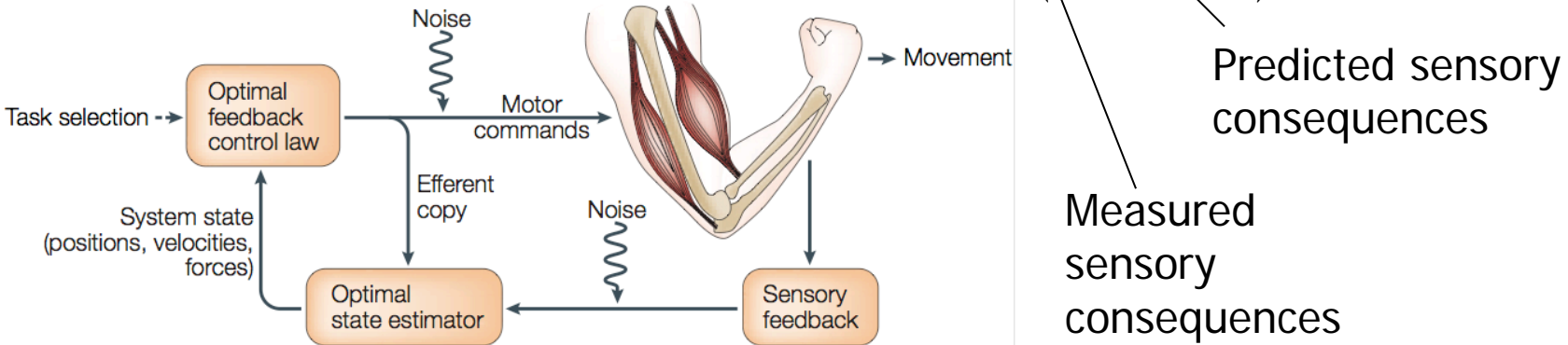
$$J = \sum_{k=0}^{p-1} \mathbf{u}^{(k)T} L^{(k)} \mathbf{u}^{(k)} + \mathbf{y}^{(k+1)T} T^{(k+1)} \mathbf{y}^{(k+1)}$$

Feedback control policy

$$\mathbf{u}^{(k)} = G^{(k)} \hat{\mathbf{x}}^{(k)}$$

Belief about state

$$\hat{\mathbf{x}}^{(k+1)} = \hat{A}\hat{\mathbf{x}}^{(k)} + \hat{A}K^{(k)} \left(\mathbf{y}^{(k)} - \hat{\mathbf{y}}^{(k)} \right) + \hat{C}\mathbf{u}^{(k)}$$

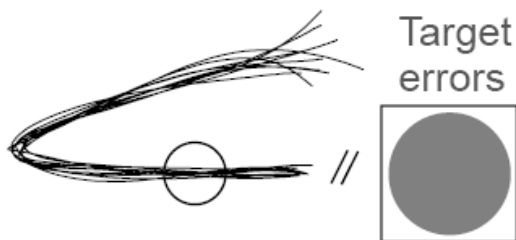


Optimal feedback control

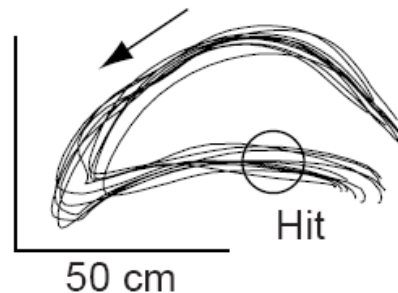
► Example: tennis



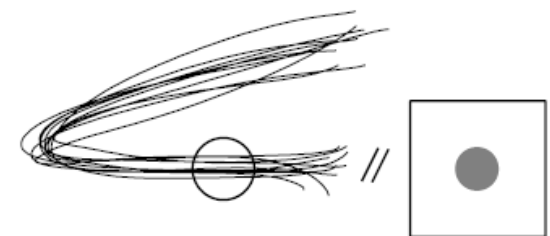
Desired trajectory



Experimental data



Optimal control

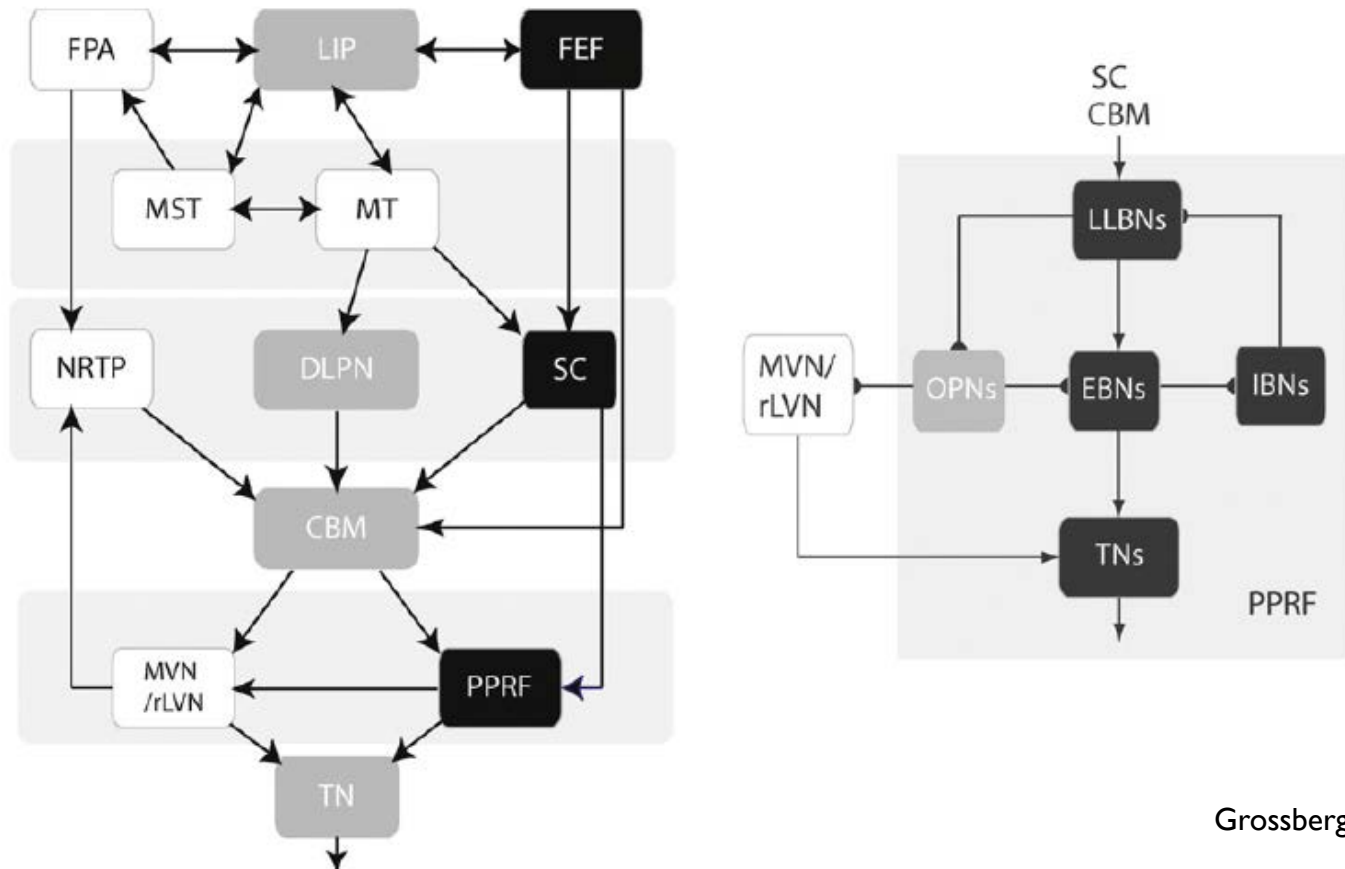


Torodov & Jordan, 2002

► Optimal control reproduces backward swing

Neural Networks

► Smooth pursuit – saccades coordination

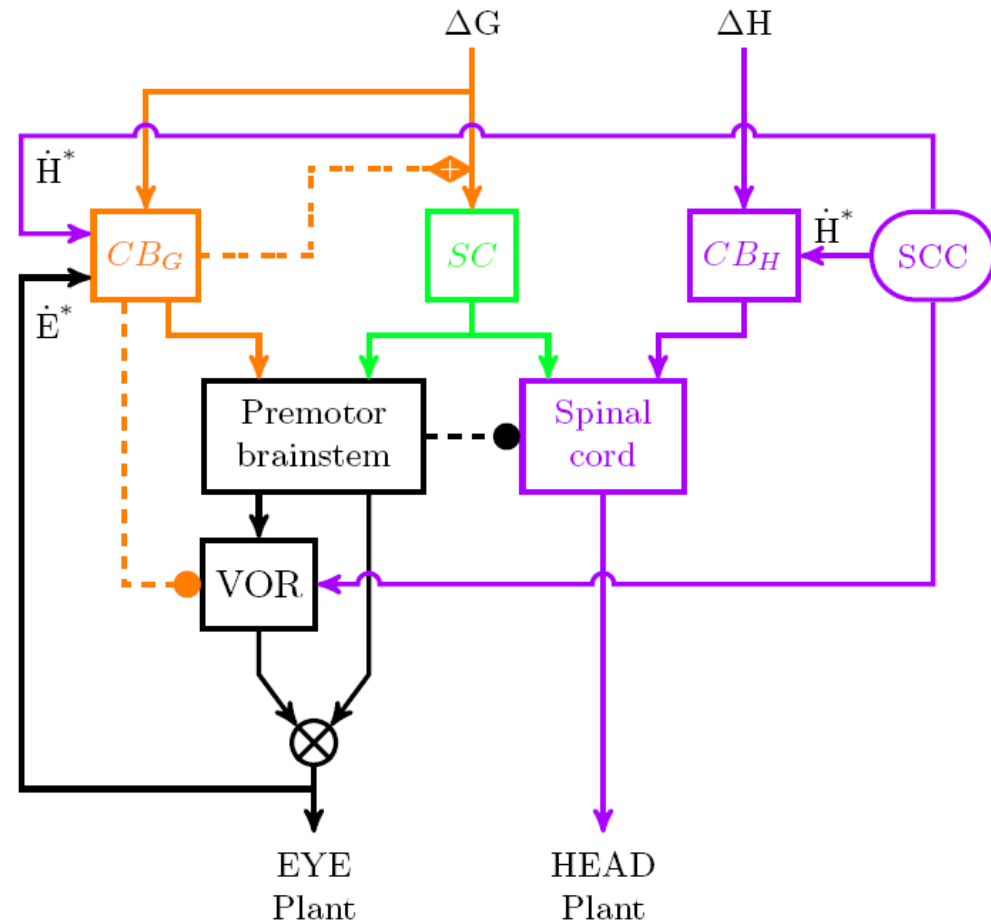


Grossberg, et al. (2012)

Hierarchical control

► Eye-head Saccades

- Endpoint control
 - Vs. trajectory control
- Head motion = perturbation to gaze goal



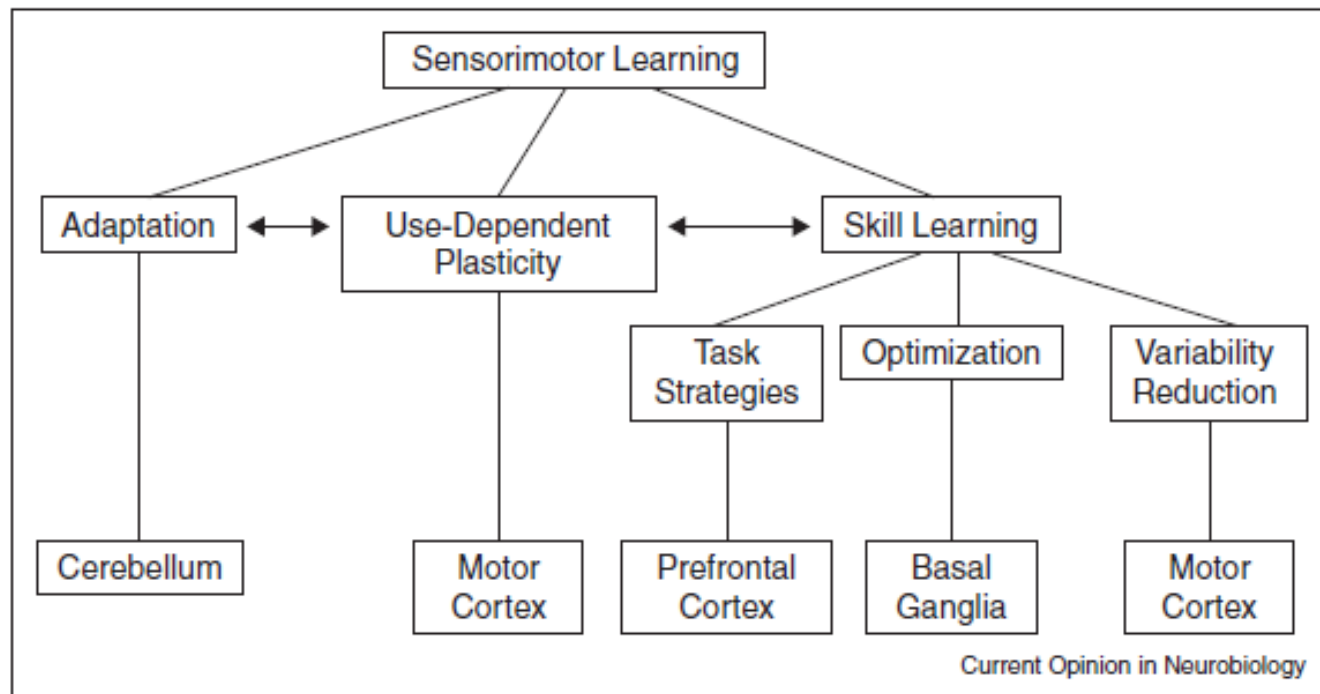
Daye, Optican, Blohm, Lefèvre (2014)

Adaptation & learning

Shadmehr & Ahmed

Learning theories

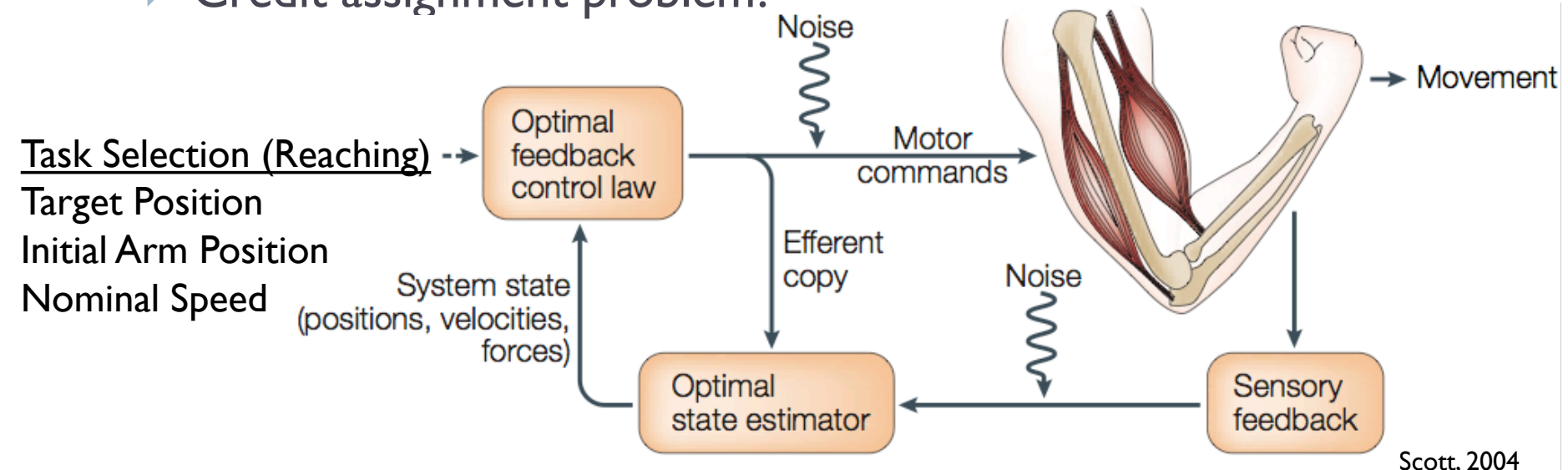
- ▶ Learning results in behavioral changes
 - ▶ Definition: “Any relatively permanent change in behavior that occurs as a result of experience.”



Krakauer & Mazzoni 2011

Learning theories

- ▶ Learning results in behavioral changes
 - ▶ Error-based learning: what went wrong?
 - ▶ Initial state?
 - ▶ OFC – inverse model?
 - ▶ State estimation – forward model? ...
 - ▶ Credit assignment problem!





That's all Folks!