

Role of the cerebellum in an attractor model for saccadic control and learning

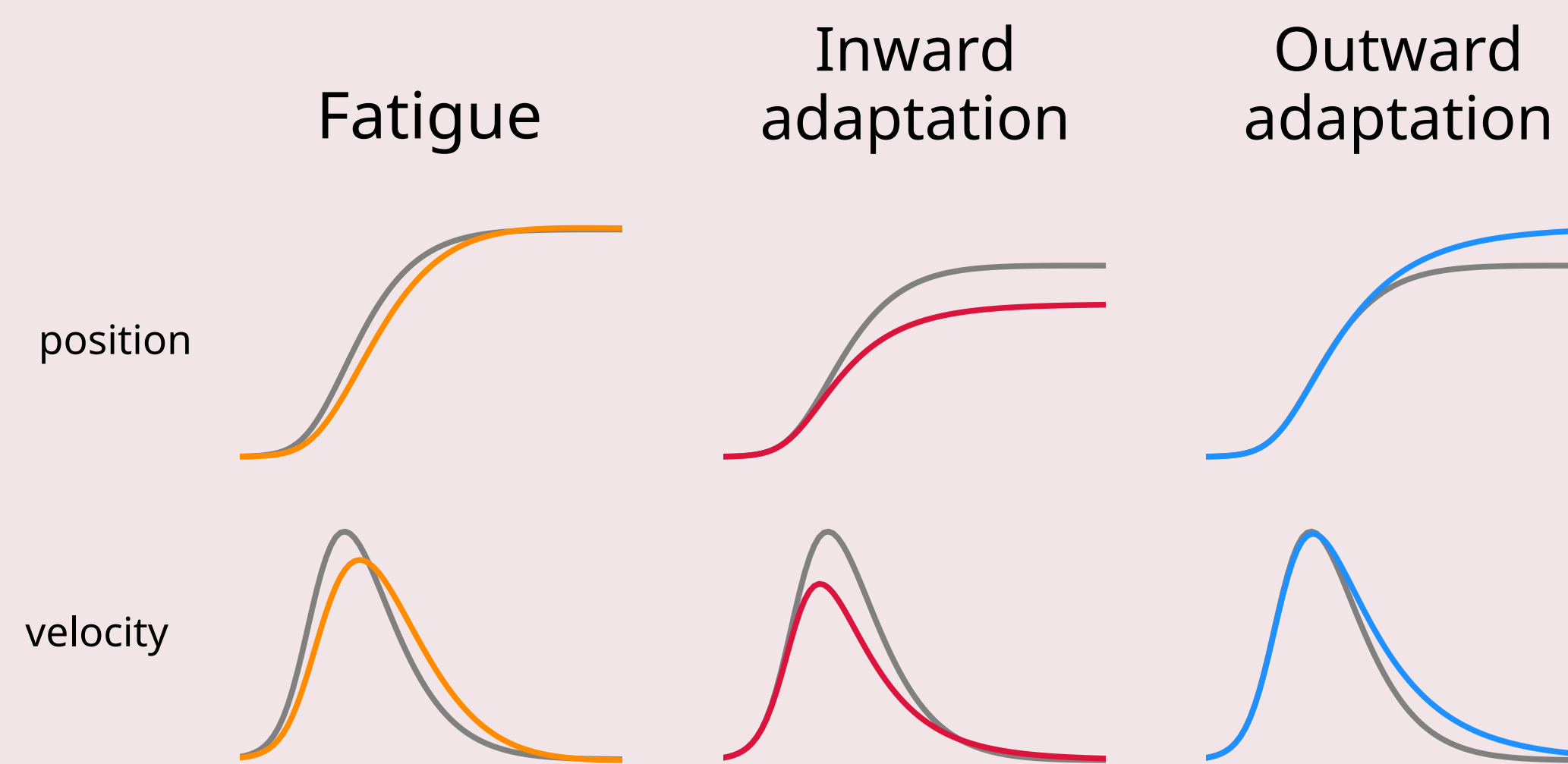
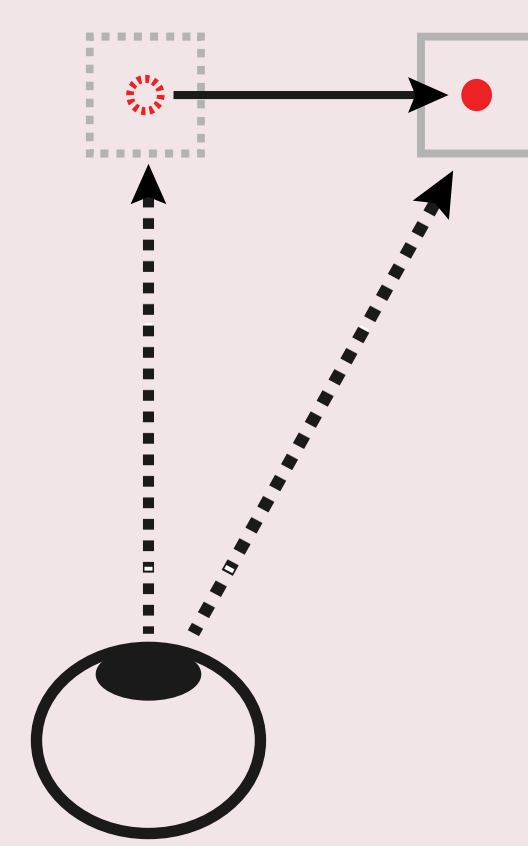
How does the cerebellum control saccades?

A saccade is a rapid movement **without sensory feedback**.
However, thanks to the cerebellum, it is still **fast** and **accurate**
and able to **adapt** to new environments

The cerebellum adjusts peak velocity (PV),
duration and amplitude resorting **internal model**
of the saccadic eye movement

Question:

How does the cerebellum fine-tune the parameters
of this internal model?



	Fatigue	Inward	Outward
Amp.		↓	↑
PV	↓	↓	
dur.	↑		↑

Theoretical approach

What kind of control is required to make a saccade fast and precise?

Estimation of motor command

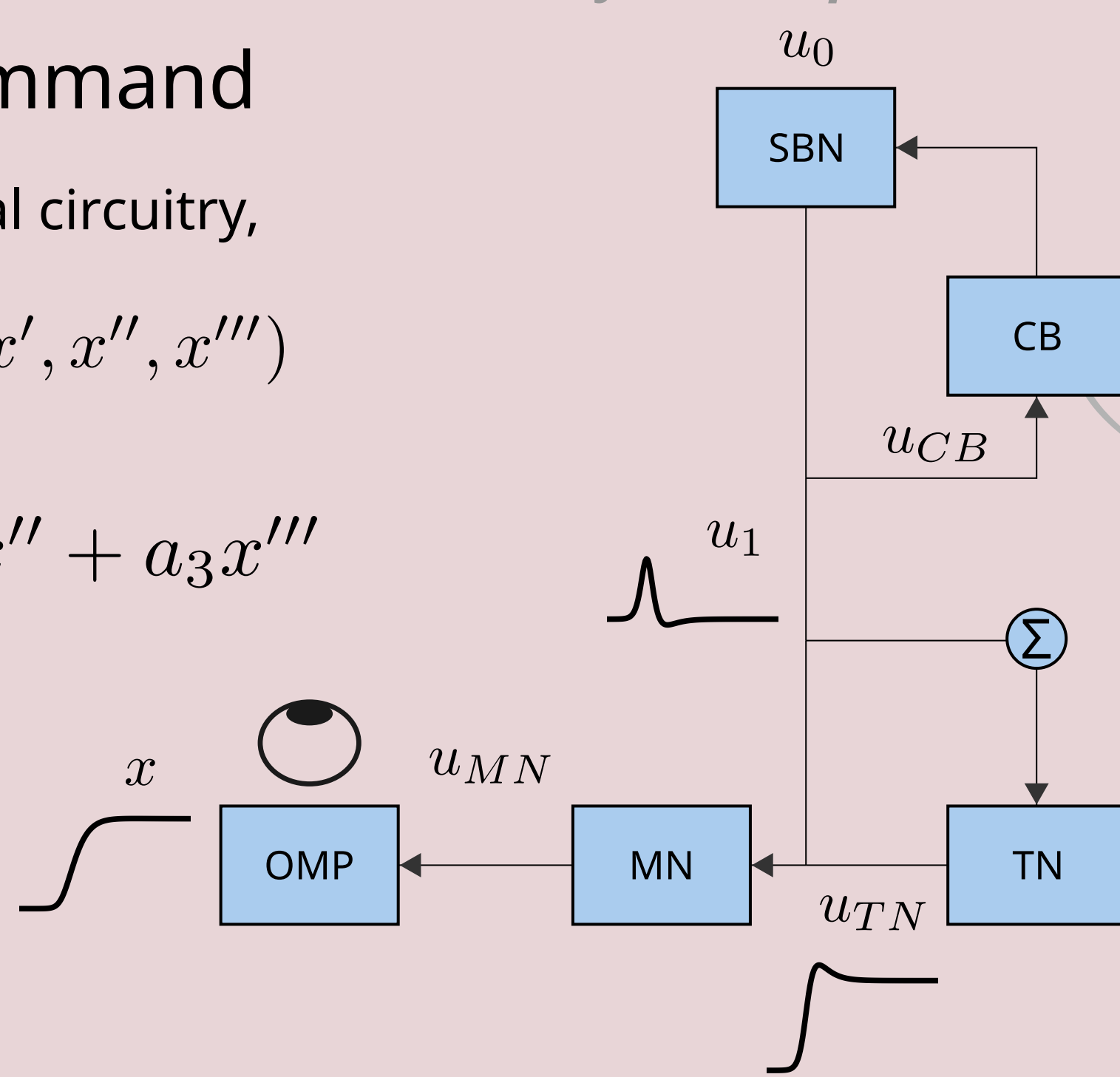
From biomechanics and neuronal circuitry,
estimate motor command u_1
from movement kinematics (x, x', x'', x''')

$$u_{MN} = a_0x + a_1x' + a_2x'' + a_3x'''$$

$$u_{MN} = \tau u_1 + u_{TN}$$

$$u_{TN} = \int u_1 dt$$

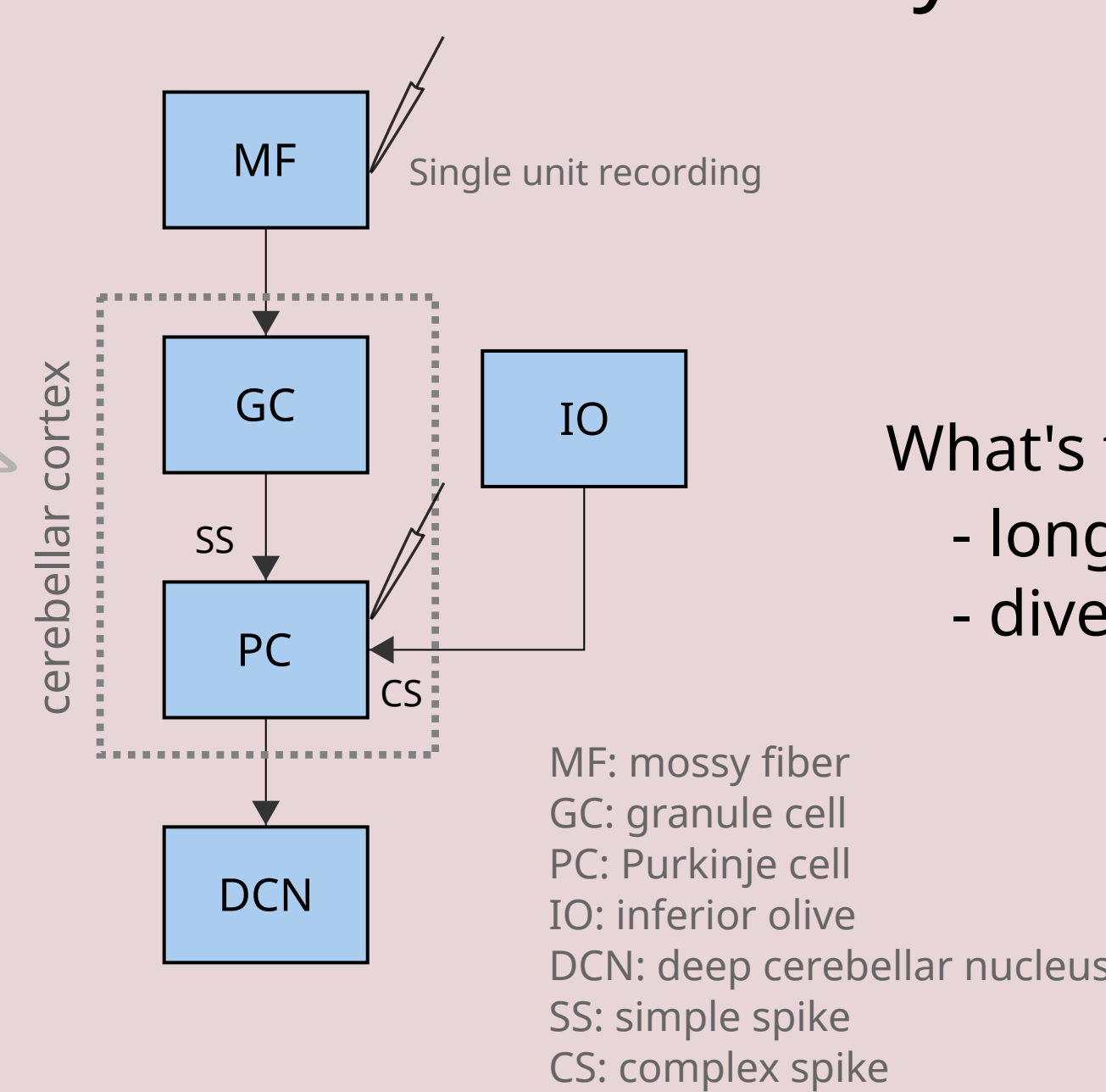
OMP: oculomotor plant
MN: motoneuron
TN: tonic neuron
SBN: saccadic burst neuron
CB: cerebellum



Experimental approach

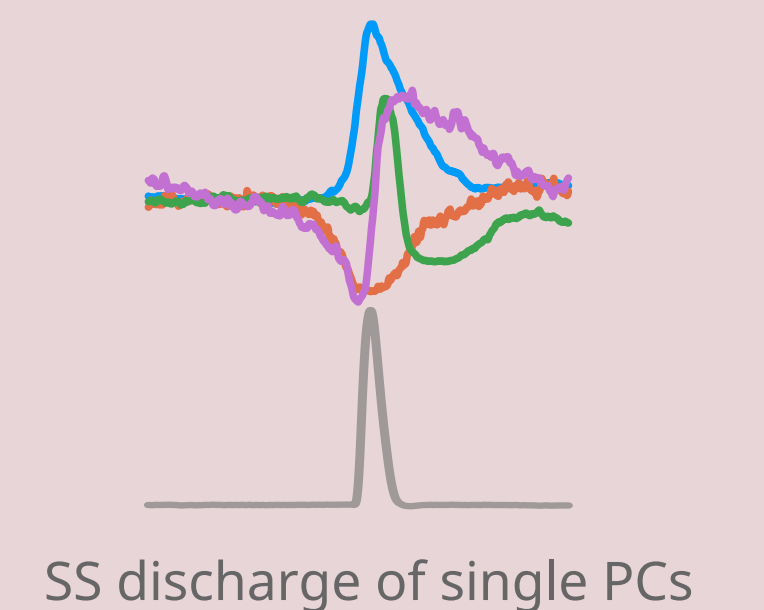
What is the relationship of the model elements to neurons in the cerebellar cortex?

Cerebellar anatomy



SS:
- result of cerebellar computation
CS:
- convey error information
- induce learning

What's the true population activity of PC-SS?
- longer modulation than actual movement
- diverse & complex firing



Control strategy

- The difference between x and u_{TN} drives the eye
- u_{TN} transiently overshoots the target
- But finally u_{TN} returns to the target such as to ensure that the eyes reach the target in one stroke
- estimate of kinematics \hat{x} and \hat{x}' are necessary

Attractor model

$$u_1 = u_0 w_{burst} \left(\left(1 - \frac{\hat{x}}{w_{amp}} \right) - w_{vel} \hat{x}' \right)$$

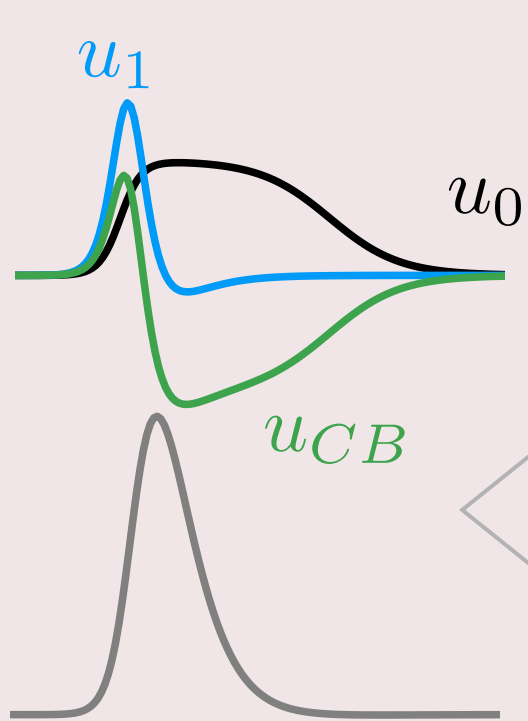
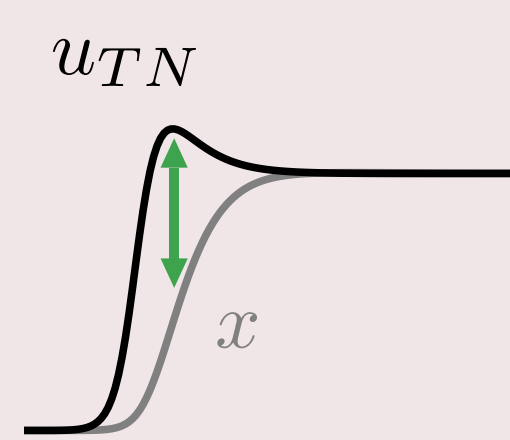
$$u_{CB} = u_1 - u_0$$

Approximation of current kinematics
(**internal model**)

$$u_1(t) \approx a \hat{x}'(t + \Delta t) + b \hat{x}''(t + \Delta t)$$

Model parameters

$$w_{burst}, w_{amp}, w_{vel}, a, b$$



PC clustering by error tuning

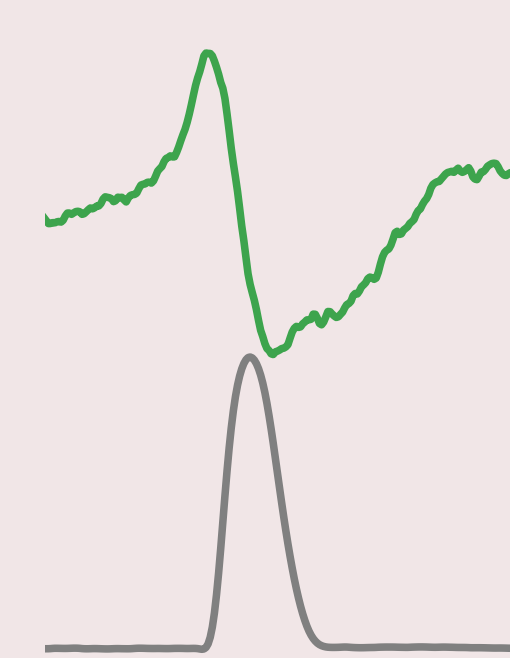
CSs convey error information and induce learning

PCs are tuned to specific error directions of CSs
(e.g. inward error \ominus and outward error \oplus)

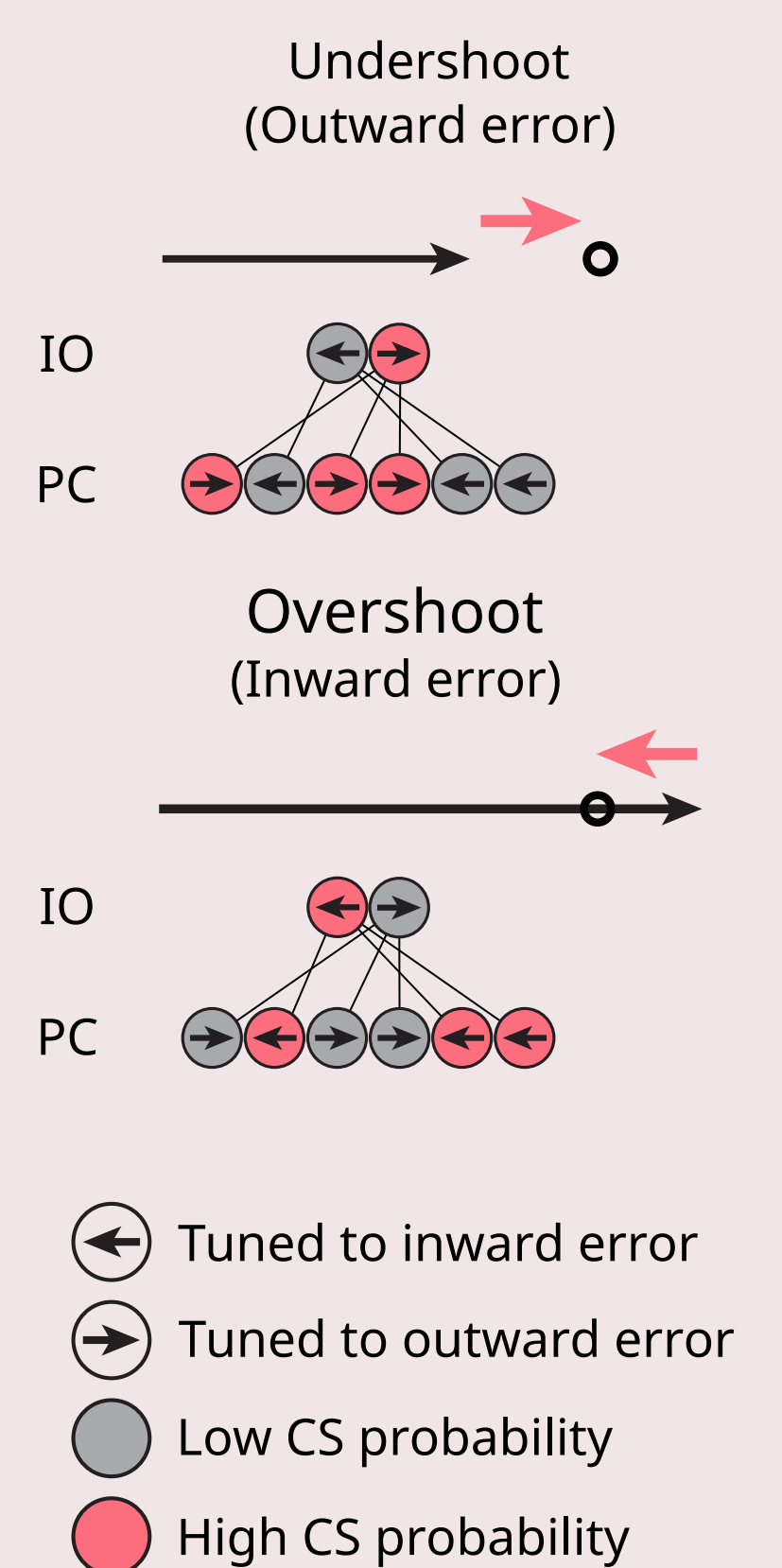
They have functionally opposite roles

Net modulation is the difference
between the two groups \ominus and \oplus

net population PC-SS



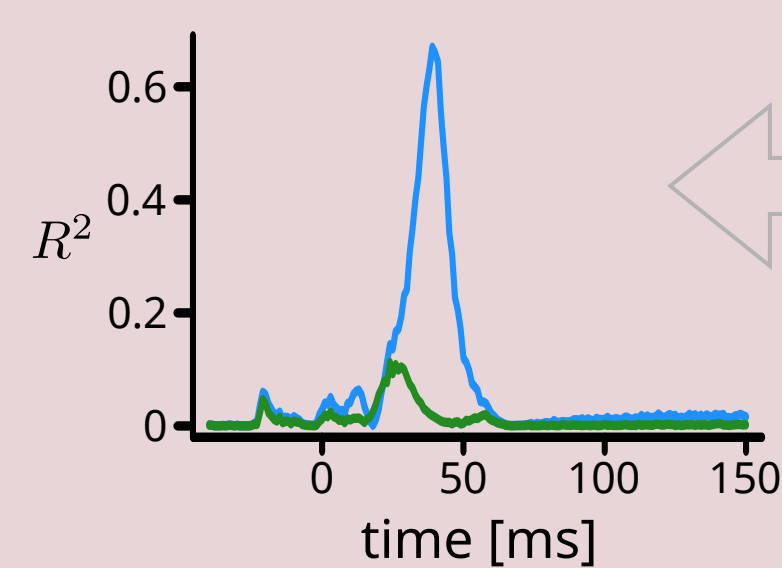
- Early peak (acceleration)
- Late trough (deceleration)



Kinematic encoding

What is the relationship between
the motor command & the resulting kinematics?
→ signal dependent noise added to u_0

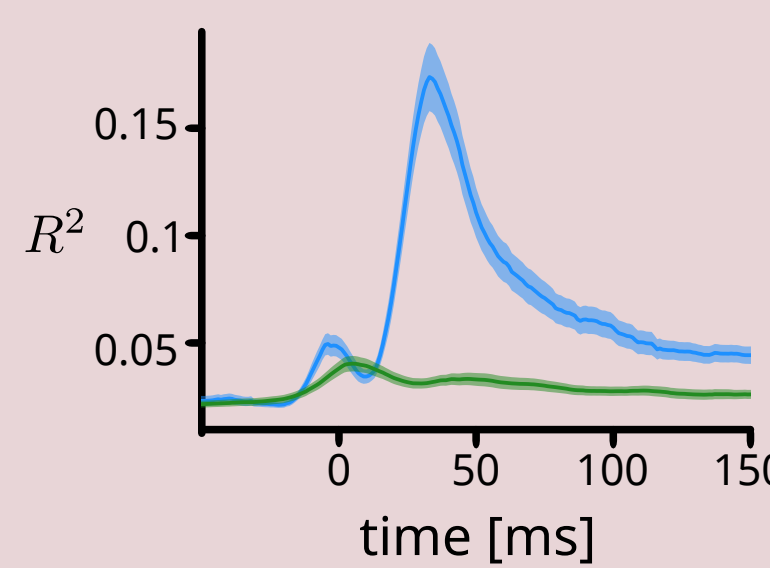
u_1 is highly correlated to kinematics (PV & duration)
 u_{CB} corrects the motor command from noisy u_0



Kinematic encoding

What do MFs and PCs encode?

MF firing is better correlated with saccade kinematics
PCs correct the movement



Changing model parameters

Behavioral changes

$$w_{amp} \sim a$$

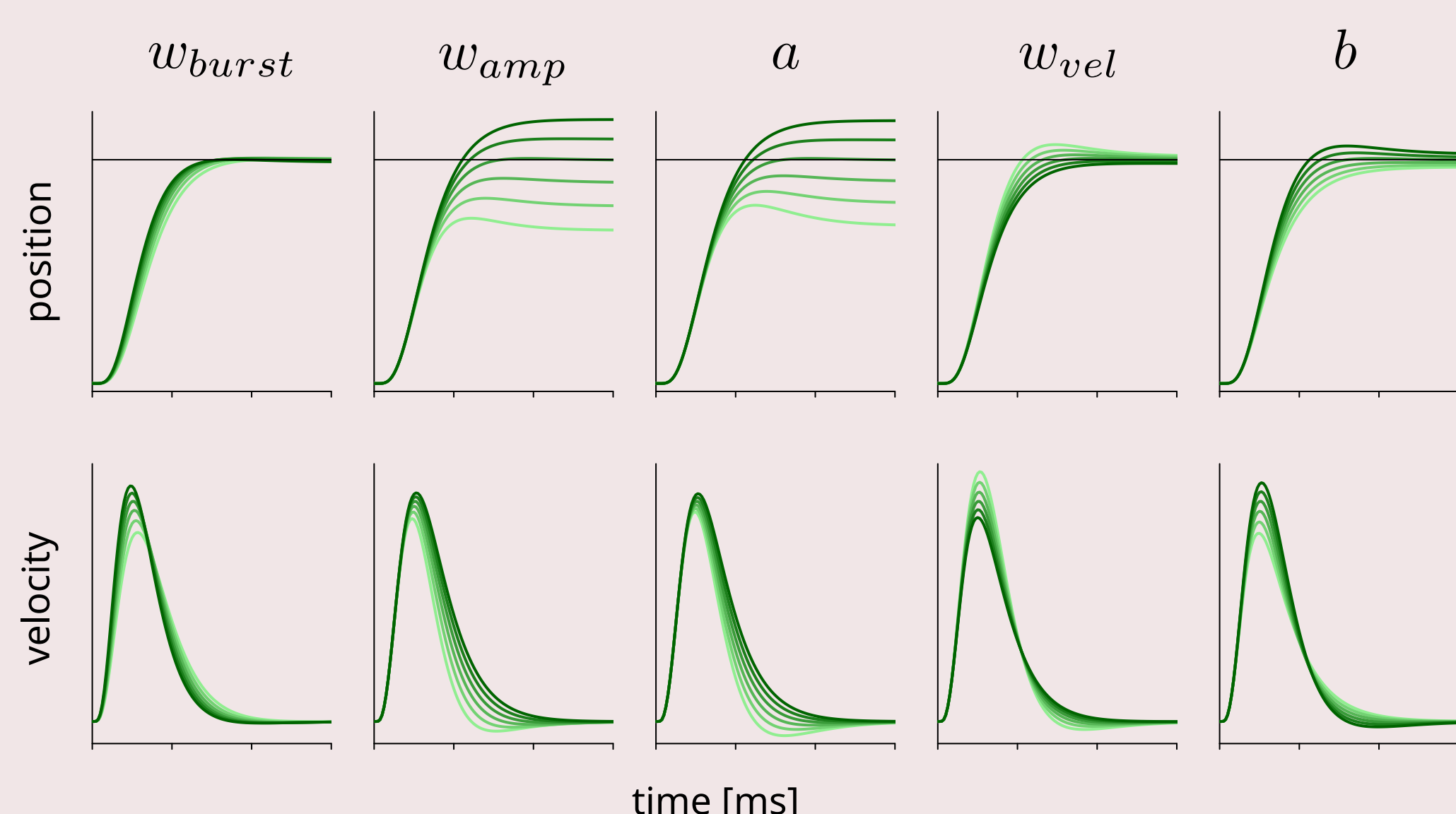
$$w_{vel} \sim b$$

Short-term adaptation

update: (a, b)
internal model parameters

Long-term adaptation

update: (w_{amp}, w_{vel})
downstream network weights



Summary & Conclusion

The attractor model can explain effectively the behavior of the PC-SS discharge
- early acceleration & late deceleration phases of the saccades
- little encoding of saccade kinematics
- PC-SS works as an online correction of the noisy signal

This could be particularly important for the implementation of
learning-based changes of saccade amplitudes
- multidimensional parameter updates for
short- and long-term adaptation