Adaptive control tutorial



Gaze stabilization in animals



Rapid reflexive responses

Vestibulo-ocular reflex

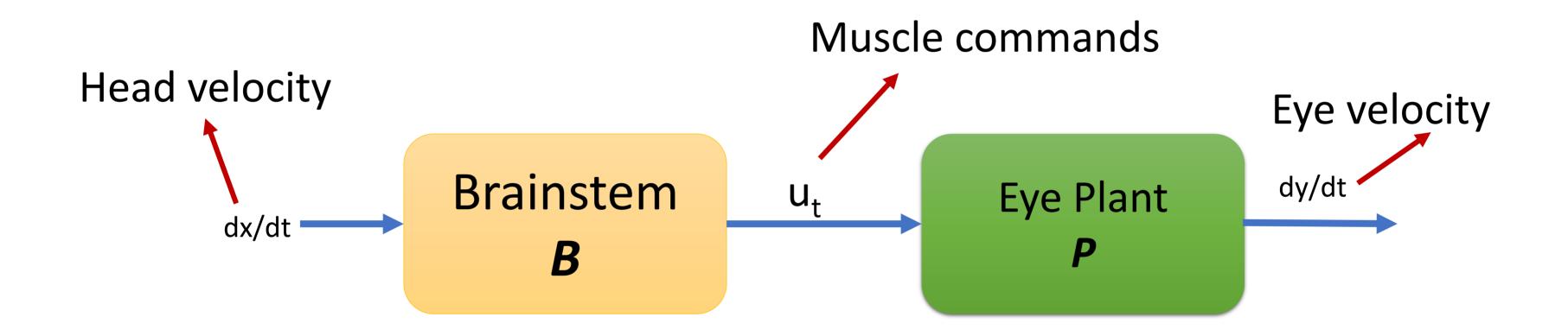
A COMPREHENSIVE GAZE STABILIZATION CONTROLLER BASED ON CEREBELLAR INTERNAL MODELS

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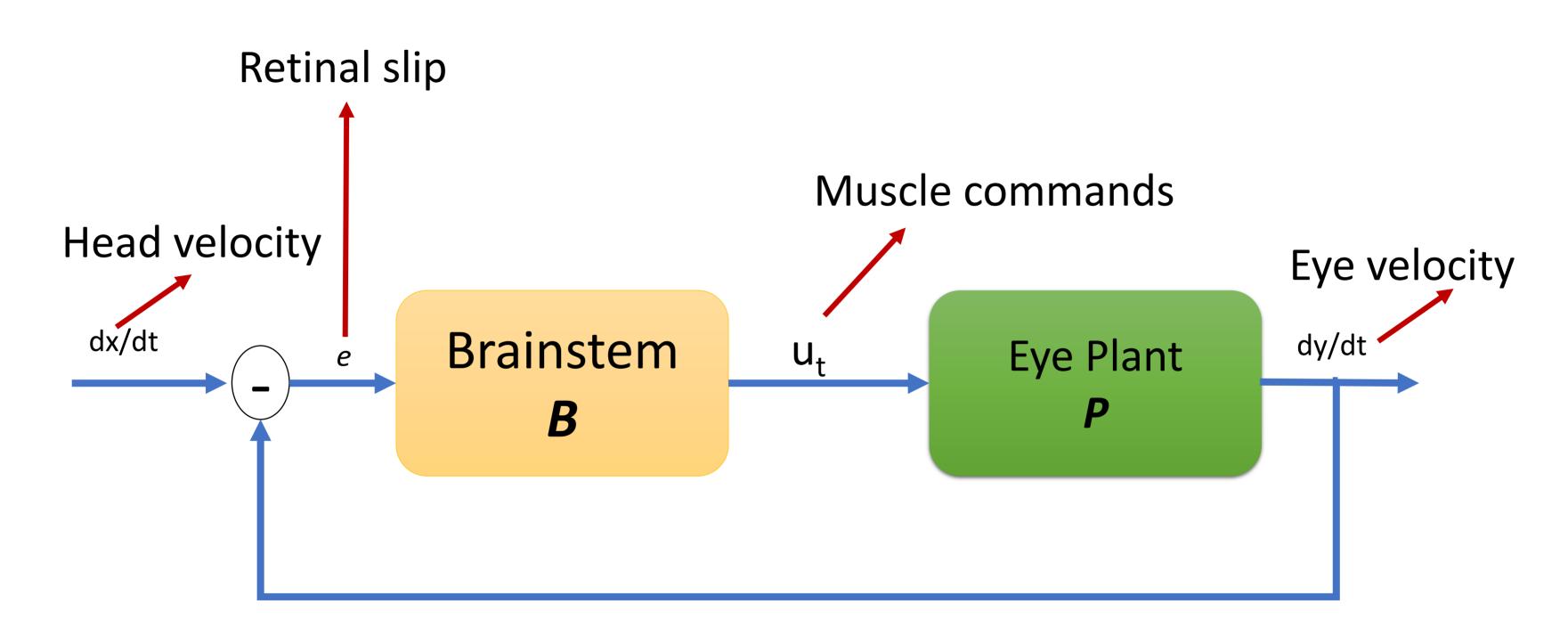
A direct mapping between head-eye velocity



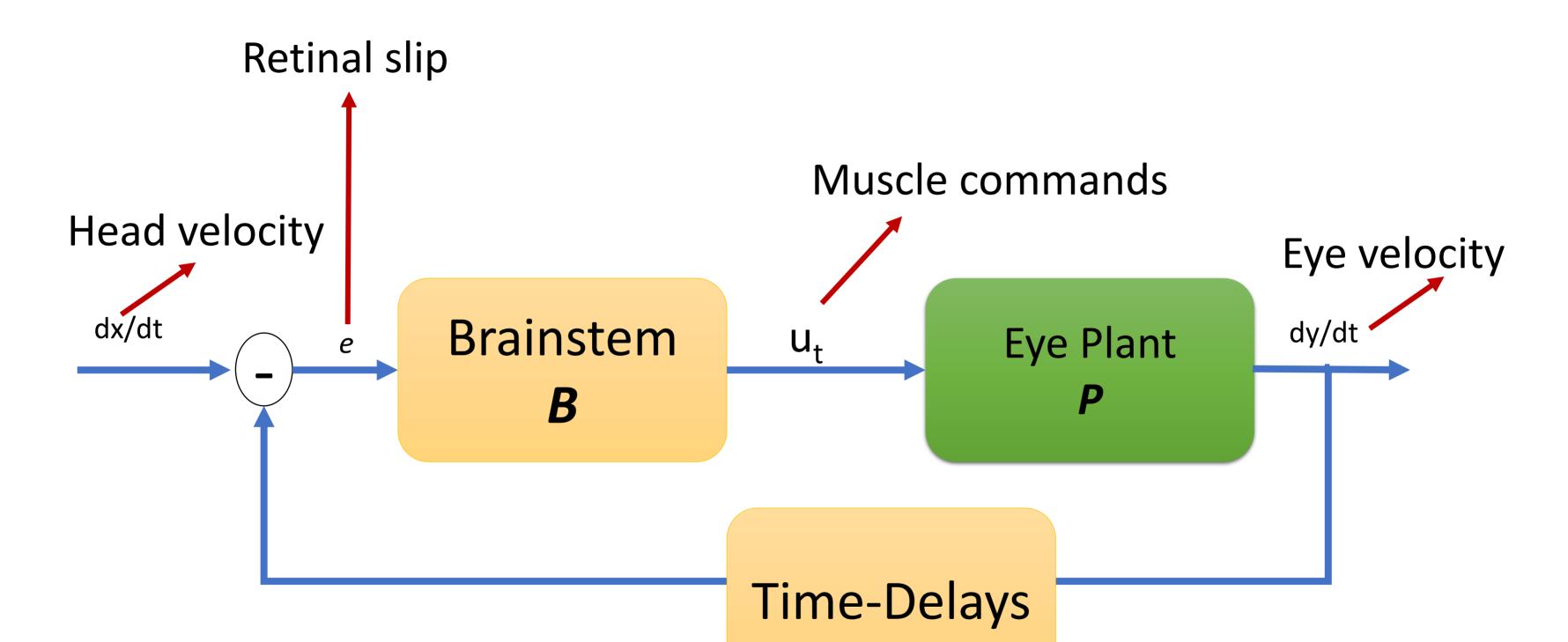
For a given head velocity, there exist a mapping that outputs motor commands To generate equal and opposite eye velocity to stabilize the visual image

These direct mappings are problematic if there is any undesired effect such as Noise, changes in the plant characteristics etc.,

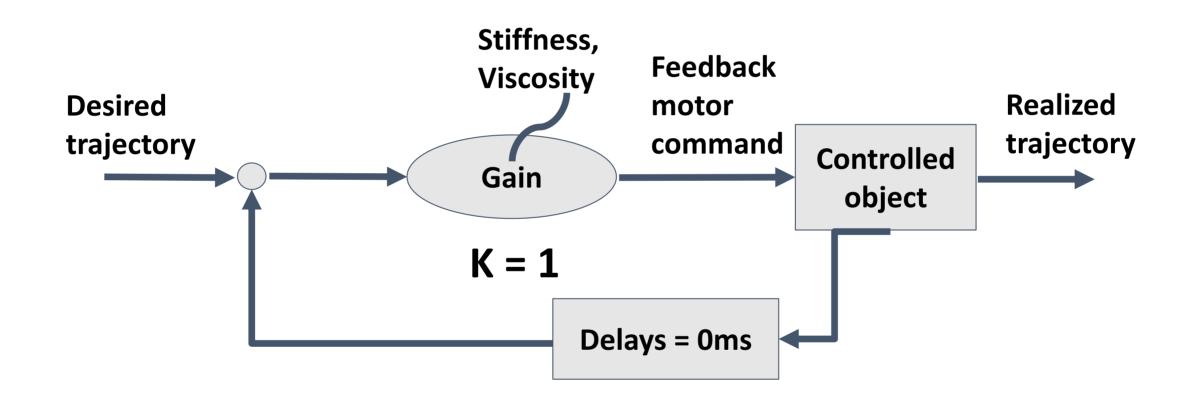
Simple Feedback Control



Problems with feedback delays

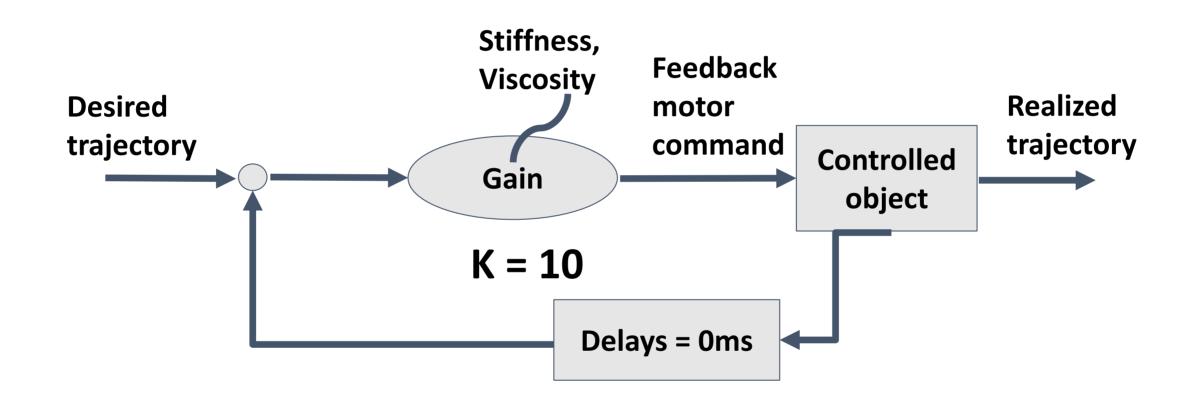


Consider a simple feedback control loop with proportional feedback gain



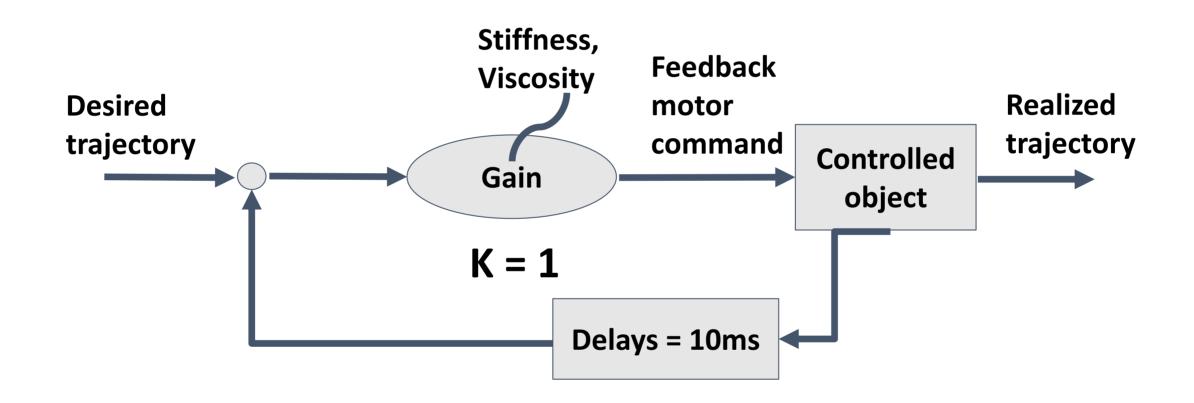
Low gain + No delay

Consider a simple feedback control loop with proportional feedback gain



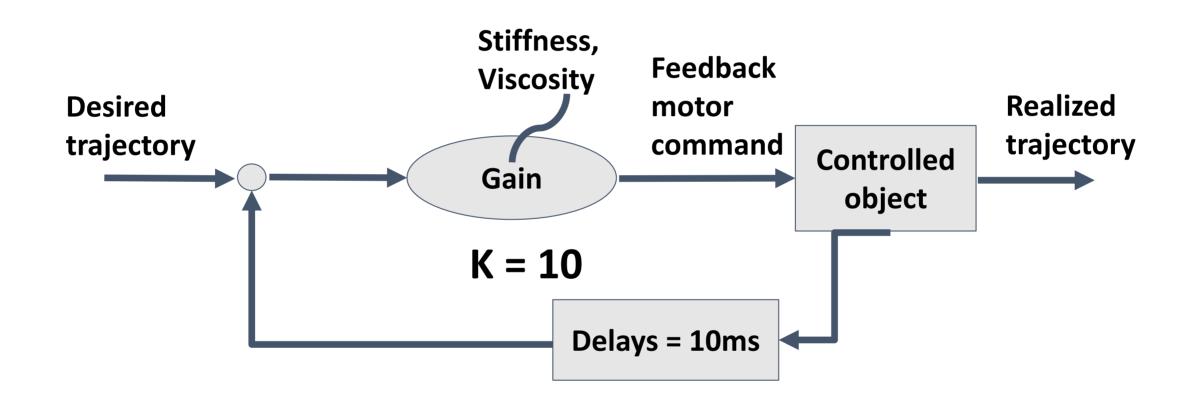
High gain + No delay

Consider a simple feedback control loop with proportional feedback gain



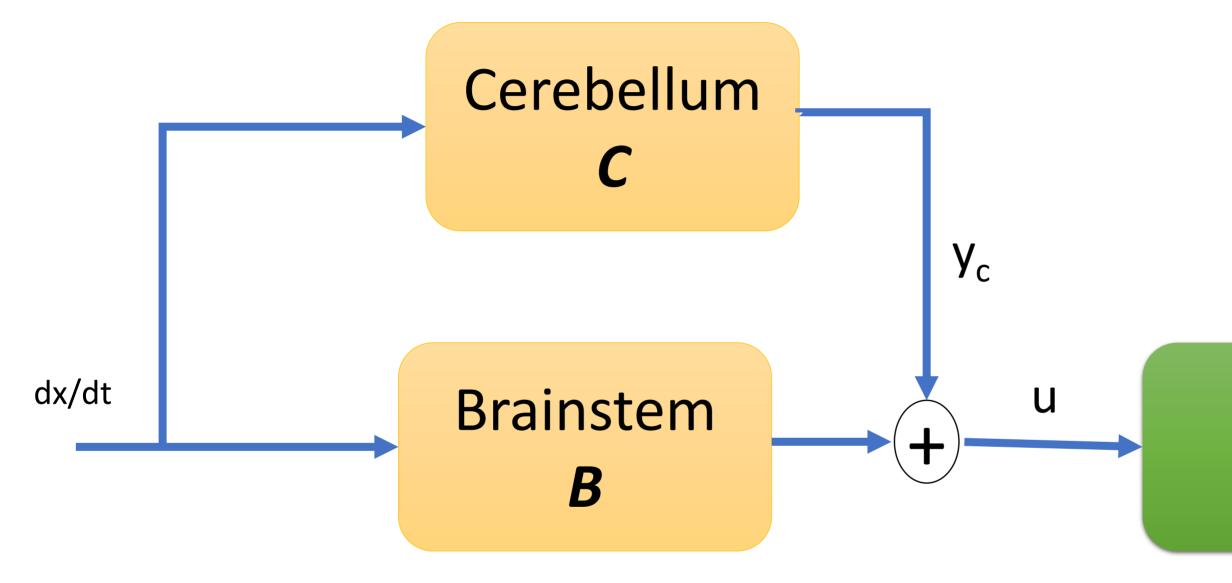
Low gain + delay

Consider a simple feedback control loop with proportional feedback gain



High gain + delay

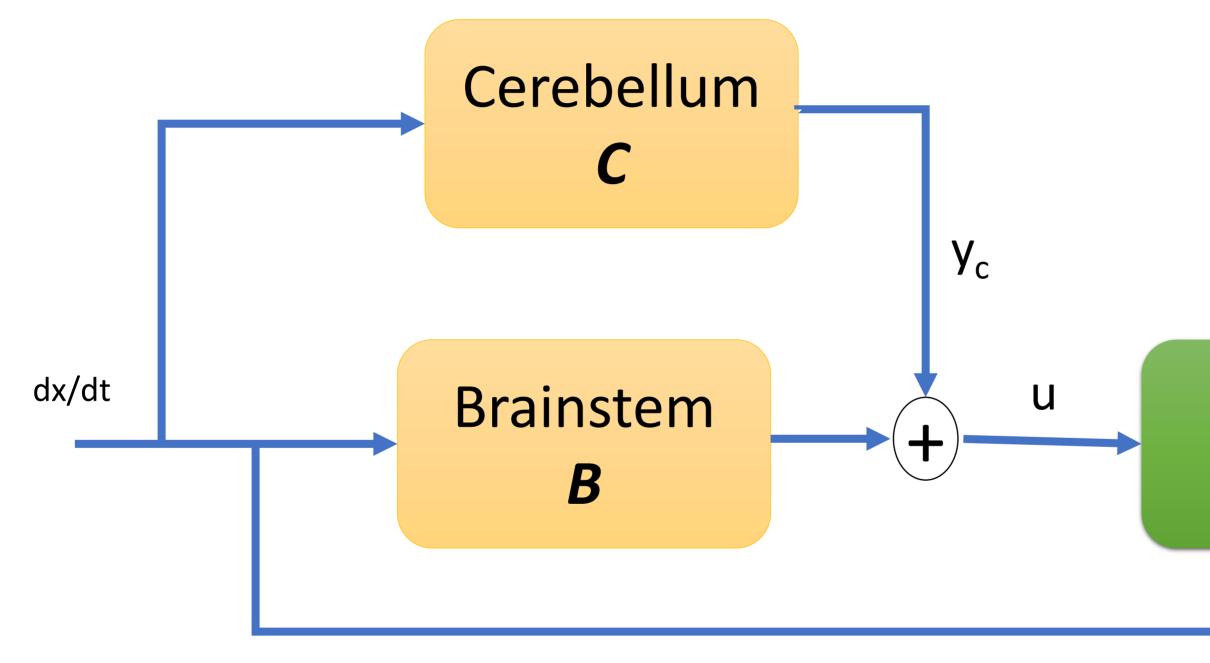
Cerebellum based adaptive control – inverse model

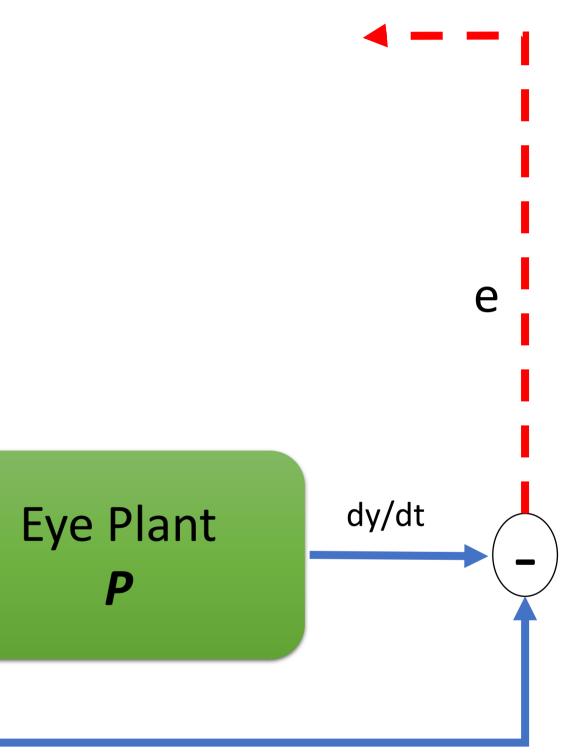


Eye Plant P

dy/dt

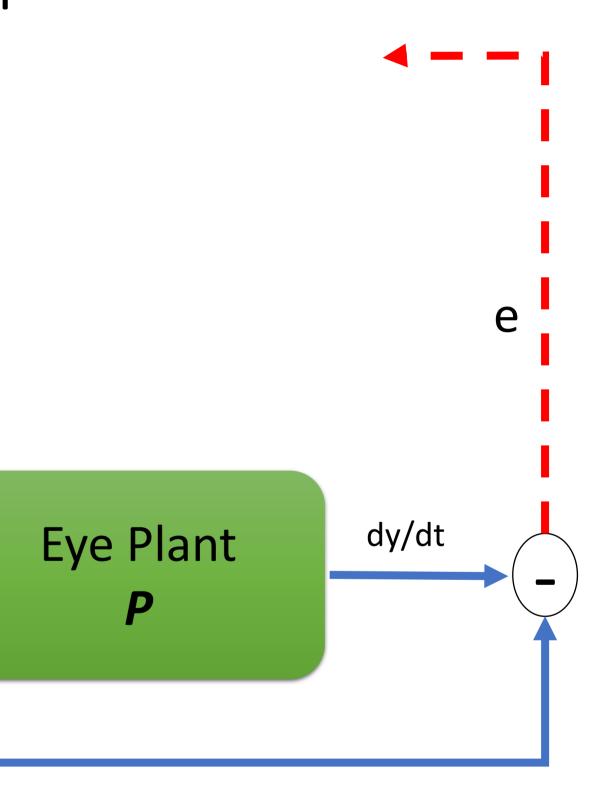
Cerebellum based adaptive control – inverse model



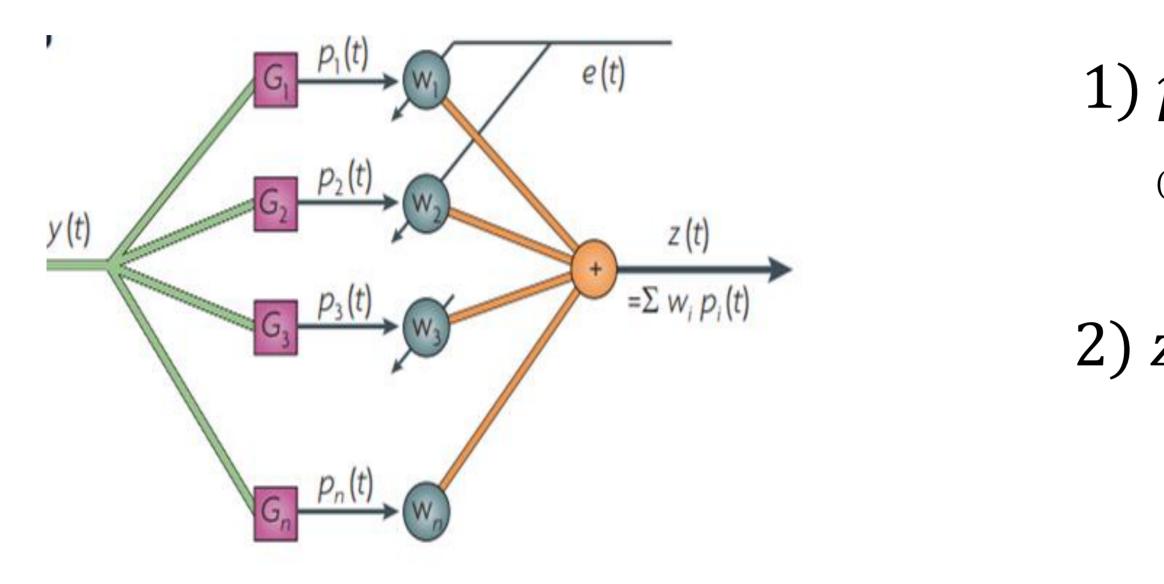


Cerebellum based adaptive control – inverse model Cerebellum $\mathbf{Y}_{\mathbf{C}}$ dx/dt U **Brainstem**

B



The computational circuit of cerebellum - adaptive filter



Note: 'W' represents the vector of weights {w1, w2, w3....,wn}. 'G' represents the input signal weighting

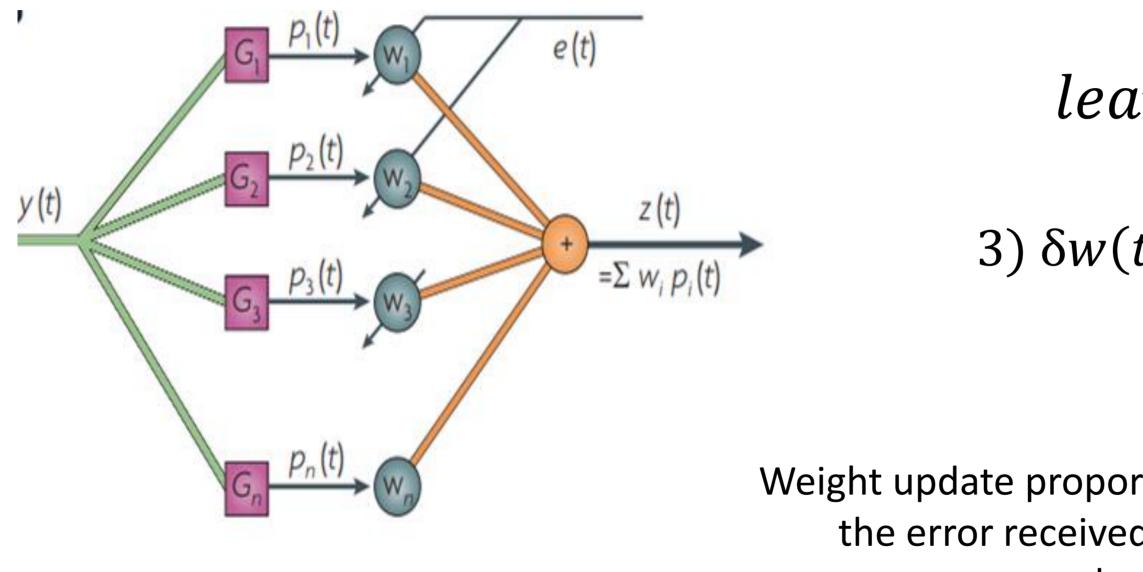
1) p(t) = G * y(t)

(n X 1) = (n X m) * (m X 1)s.t., n >= m

2) z(t) = W * p(t)

(k X 1) = (k X n) * (n X 1)s.t., k <= n

The computational circuit of cerebellum - adaptive filter

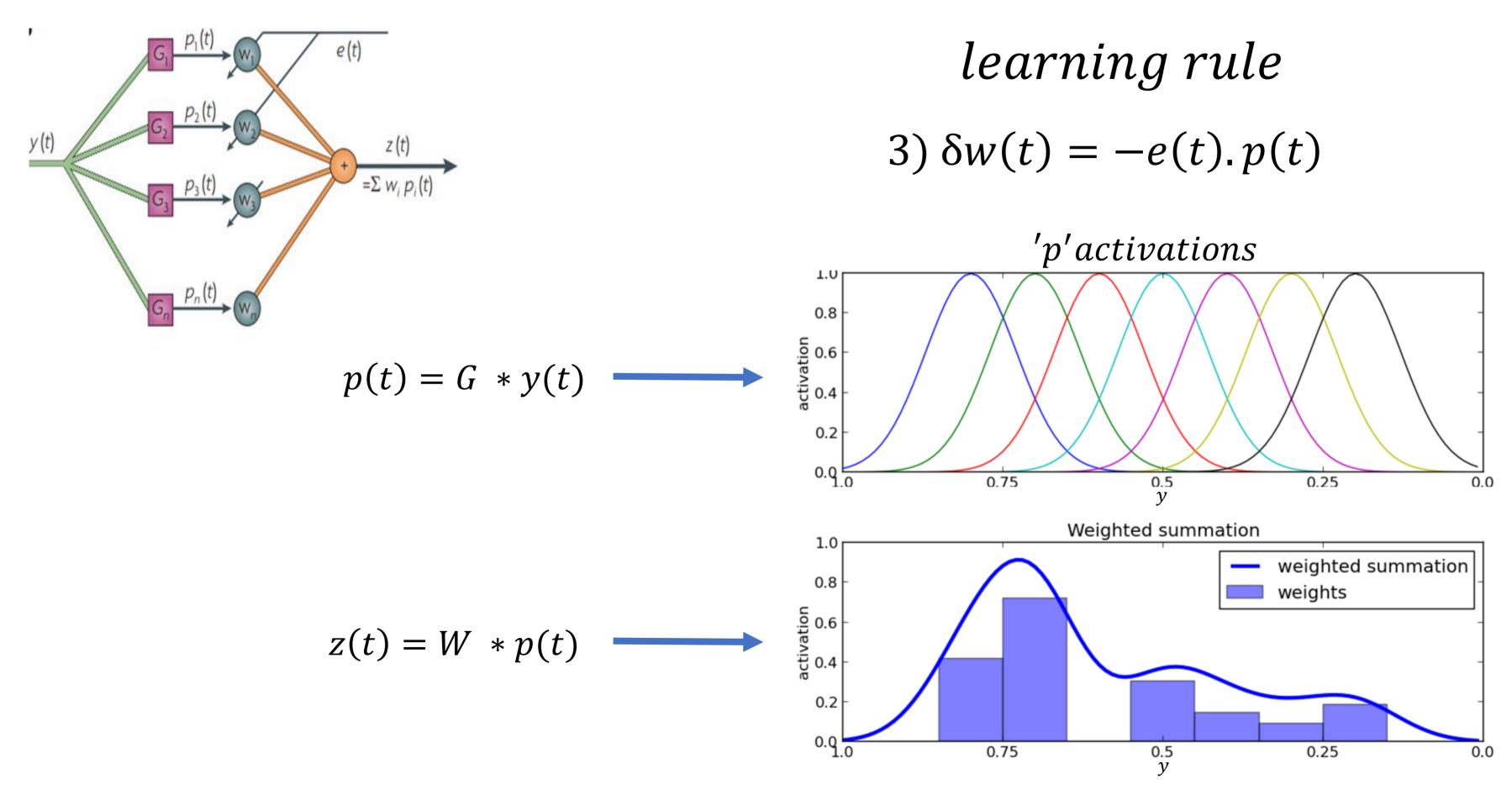


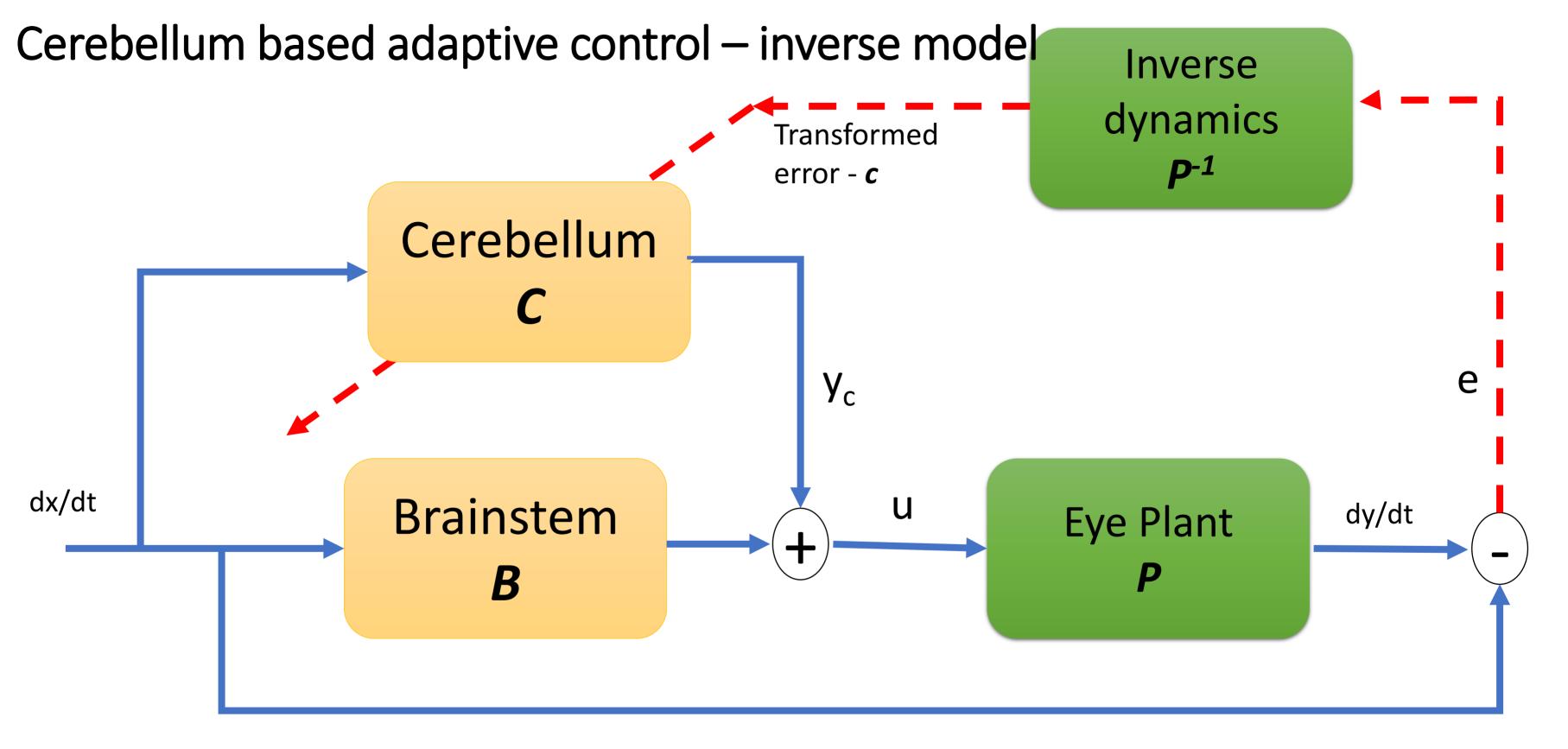
learning rule

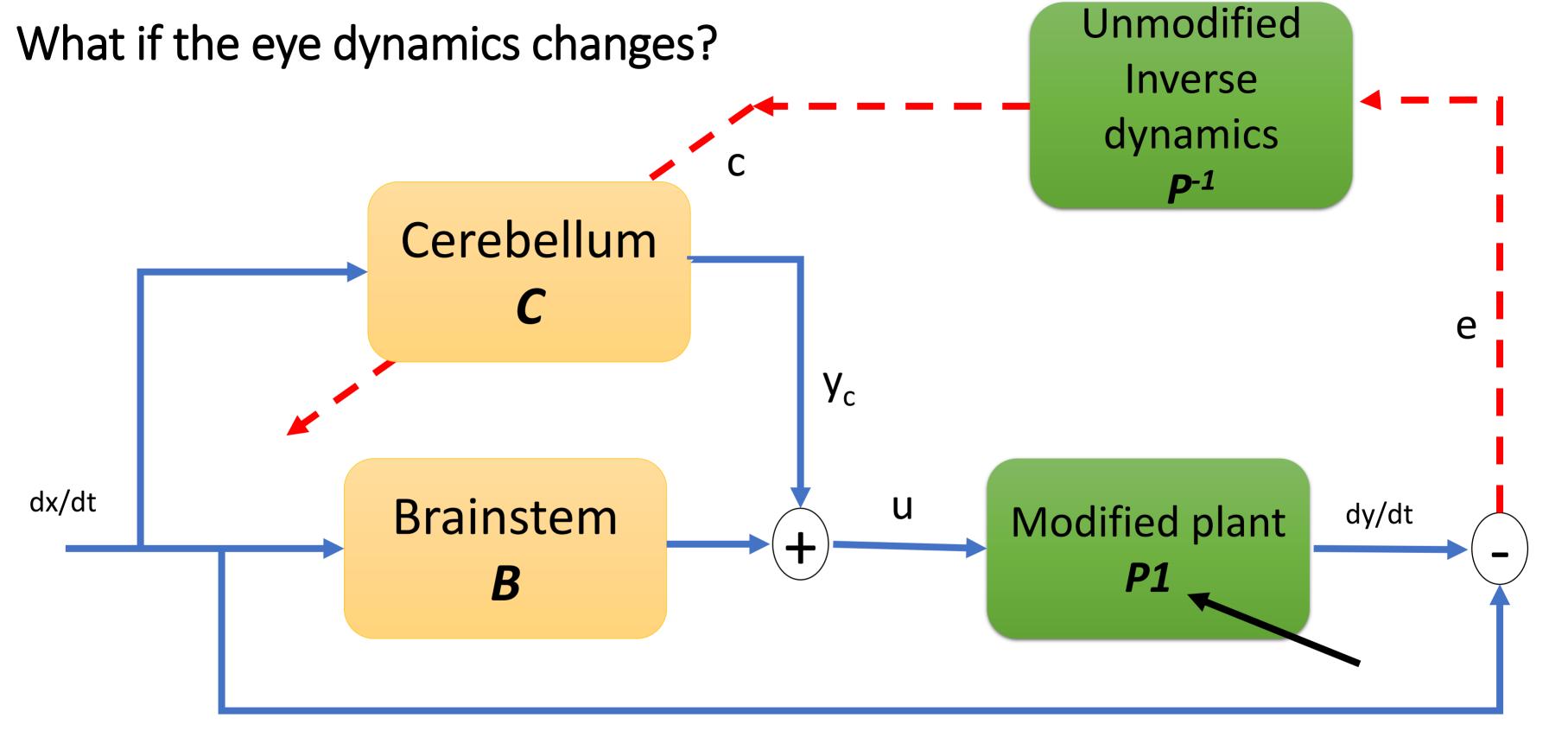
3) $\delta w(t) \alpha - e(t) p(t)$

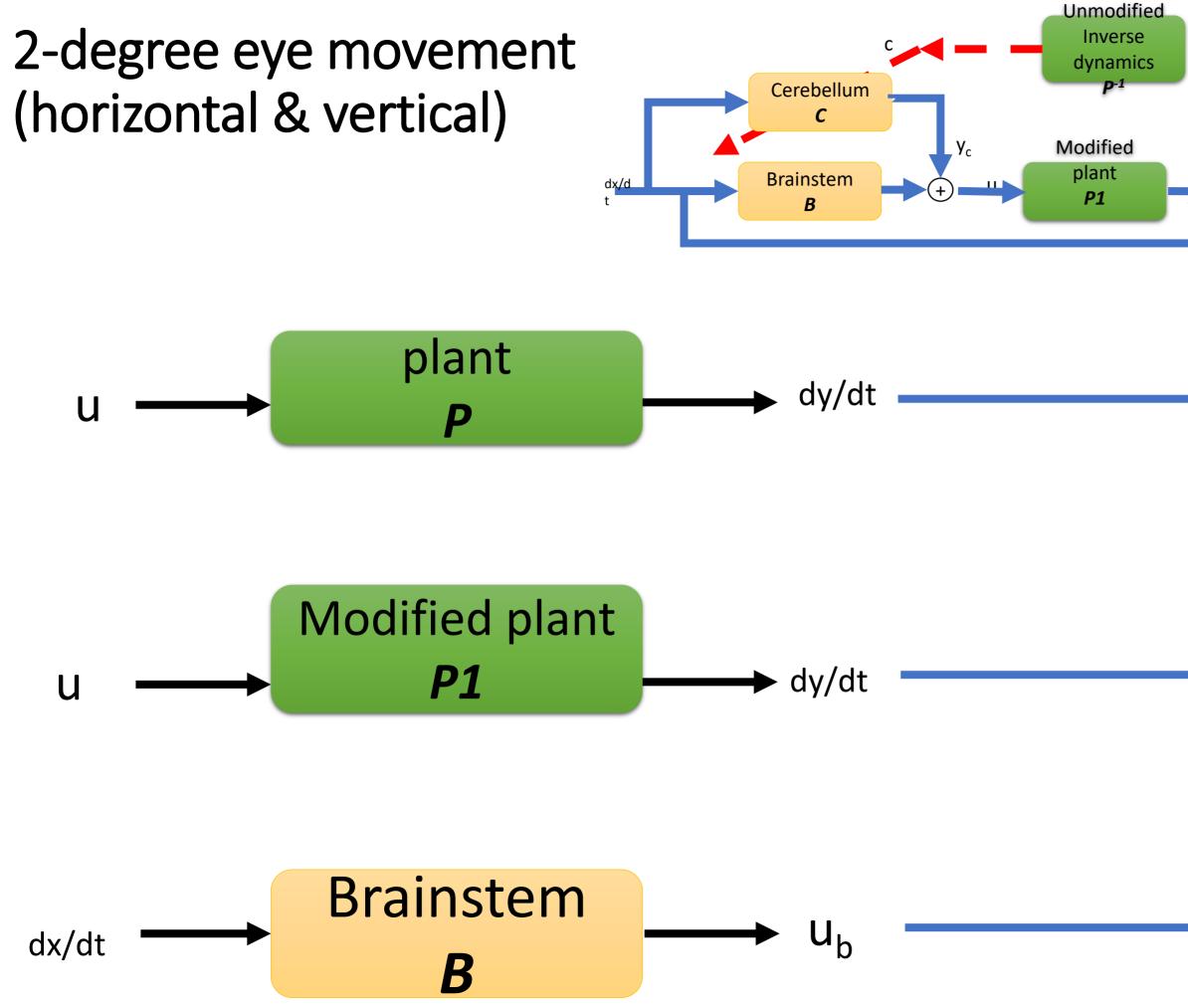
Weight update proportional to the negative product of the error received by the cerebellum and the basis activity 'p'

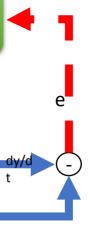
The computational circuit of cerebellum - adaptive filter







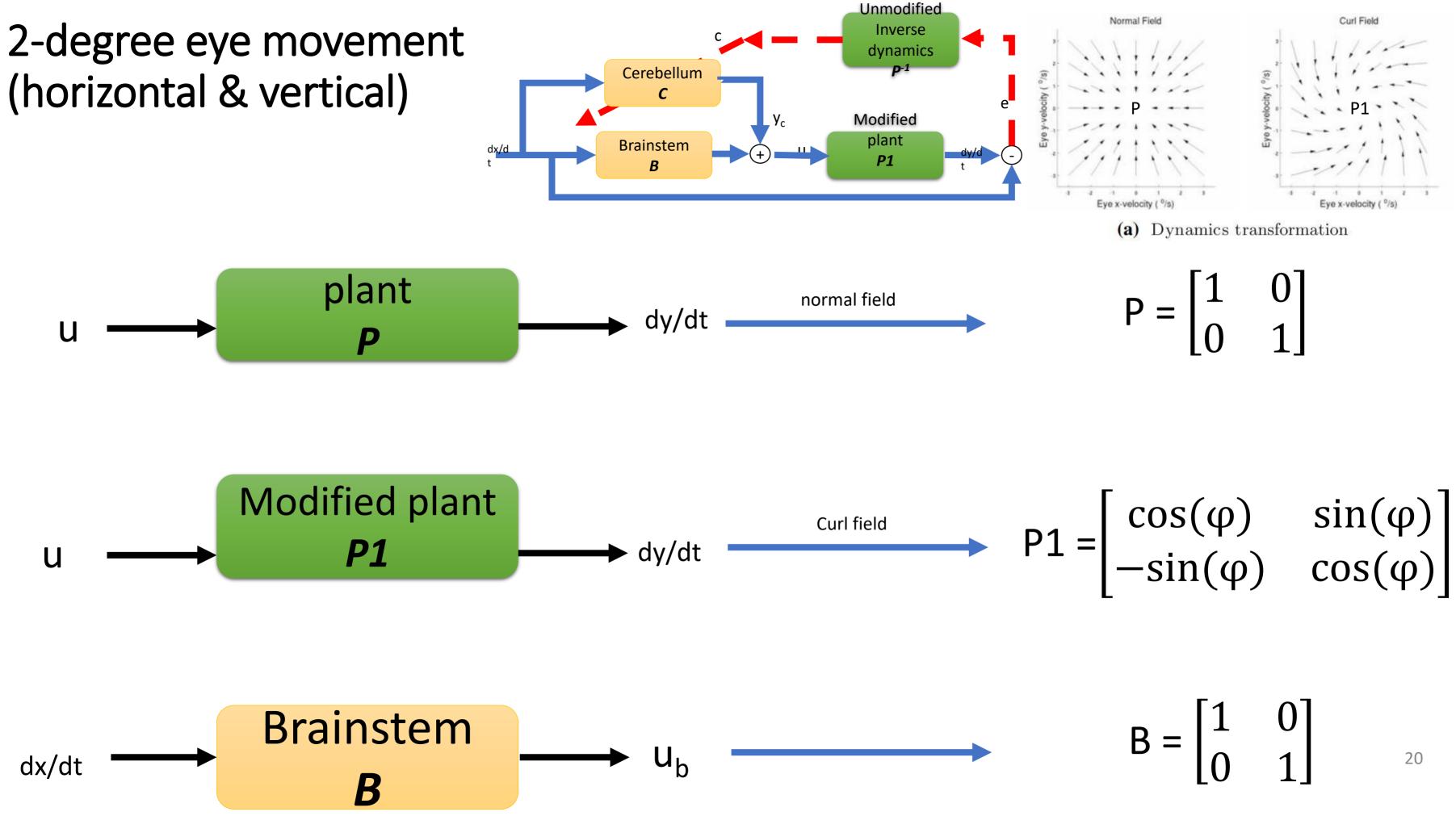


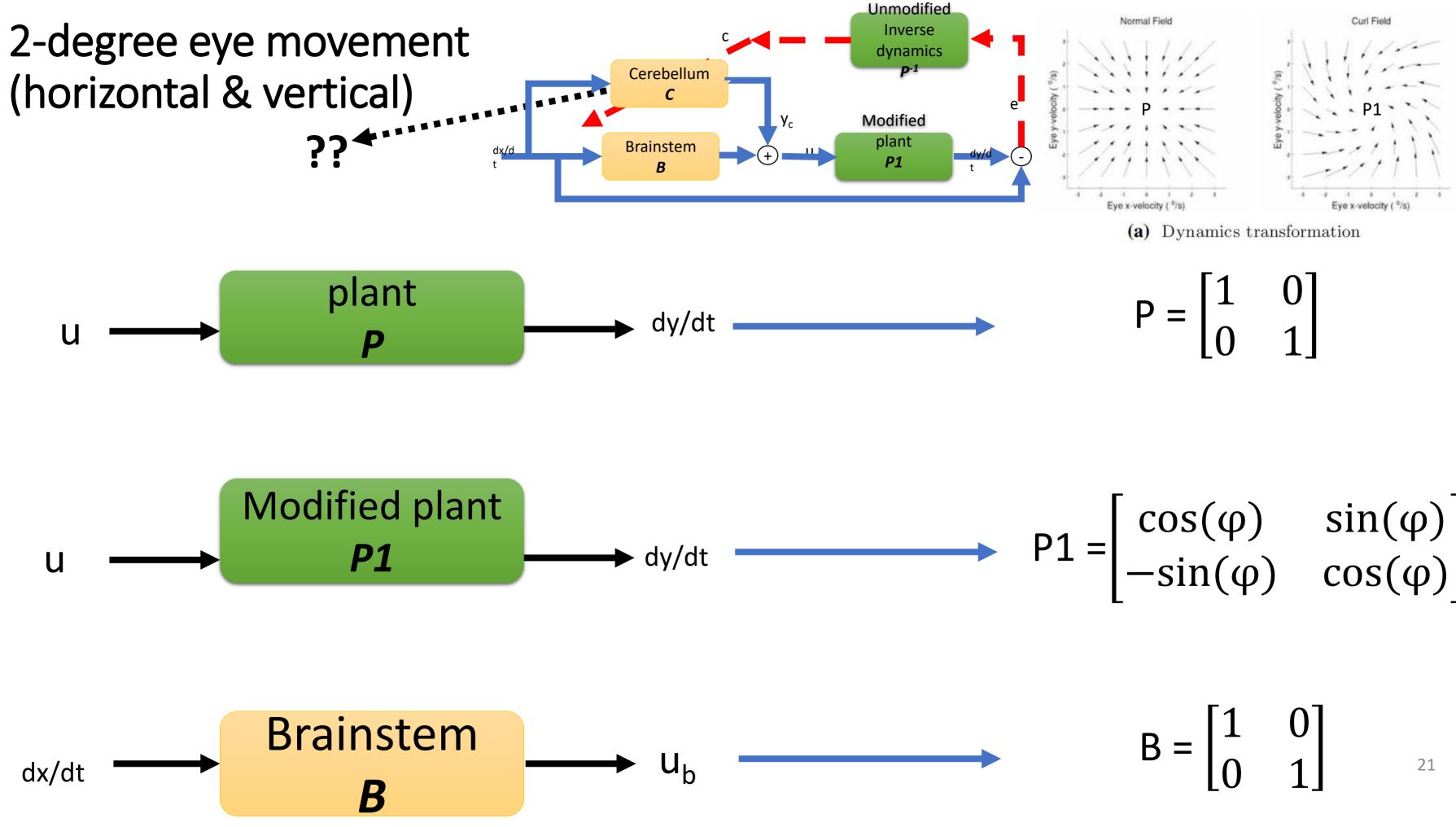


$P = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ Curl/rotation $P1 = \begin{bmatrix} \cos(\phi) & \sin(\phi) \\ -\sin(\phi) & \cos(\phi) \end{bmatrix}$

$\mathsf{B} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$

19





$P1 = \begin{vmatrix} \cos(\varphi) & \sin(\varphi) \\ -\sin(\varphi) & \cos(\varphi) \end{vmatrix}$

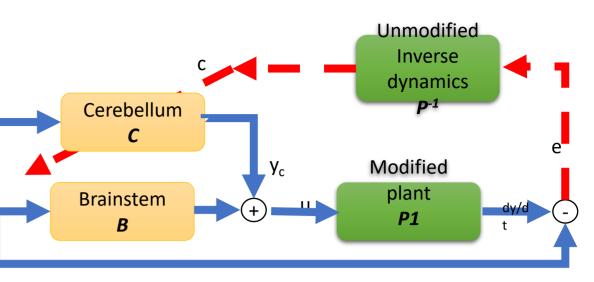
Demonstration

- start a sinusoidal head movement •
- Use the adaptive control method to cause compensatory eye movement •

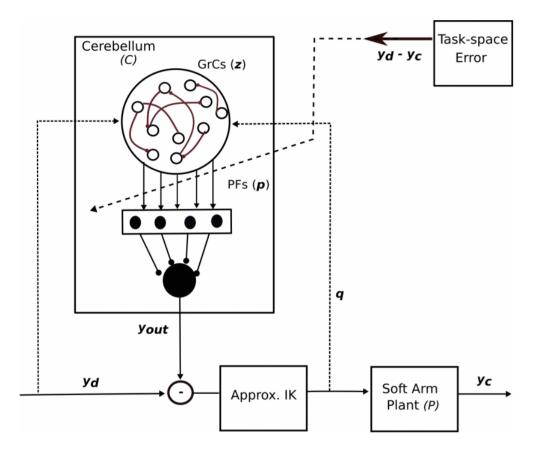
algorithm

- Initialize cerebellum as an adaptive filter
- Initialize learning rate = Ir
- for t = 1:T
 - Sense head velocity: x dot 0
 - Compute brain-stem output : u b 0
 - Compute cerebellum output : y_c = w*x_dot 0
 - Total $u = u_b + y_c$ 0
 - Compute eye velocity : y dot = P1 * u 0
 - Compute retinal slip/error: e = x_dot y_dot 0
 - Convert retinal slip to cerebellum error : $c = P^{-1} * e$ in general (or) simply c = e for small curl fields 0
 - Update cerebellum parameters : dw = lr * c * x_dot 0

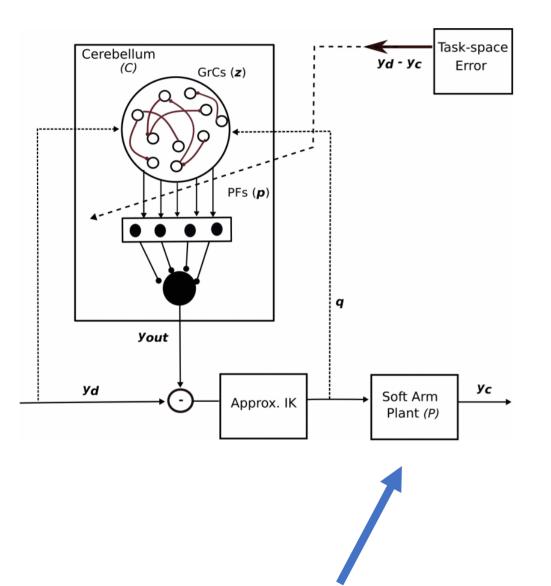




Applications - Soft robot simulation video

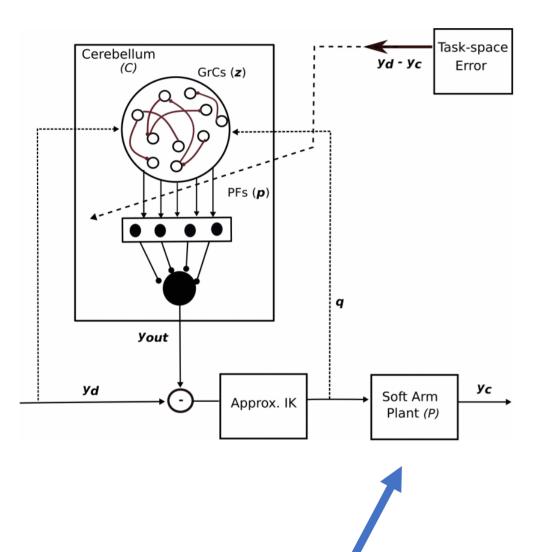


Applications - Soft robot simulation video



Can the cerebellum forward model compensate for changes in the soft arm dynamics?

Applications - Soft robot simulation video



Can the cerebellum forward model compensate for changes in the soft arm dynamics?

Cerebellum-inspired approach for adaptive kinematic control of soft robots

- (IEEE RoboSoft 2019)

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