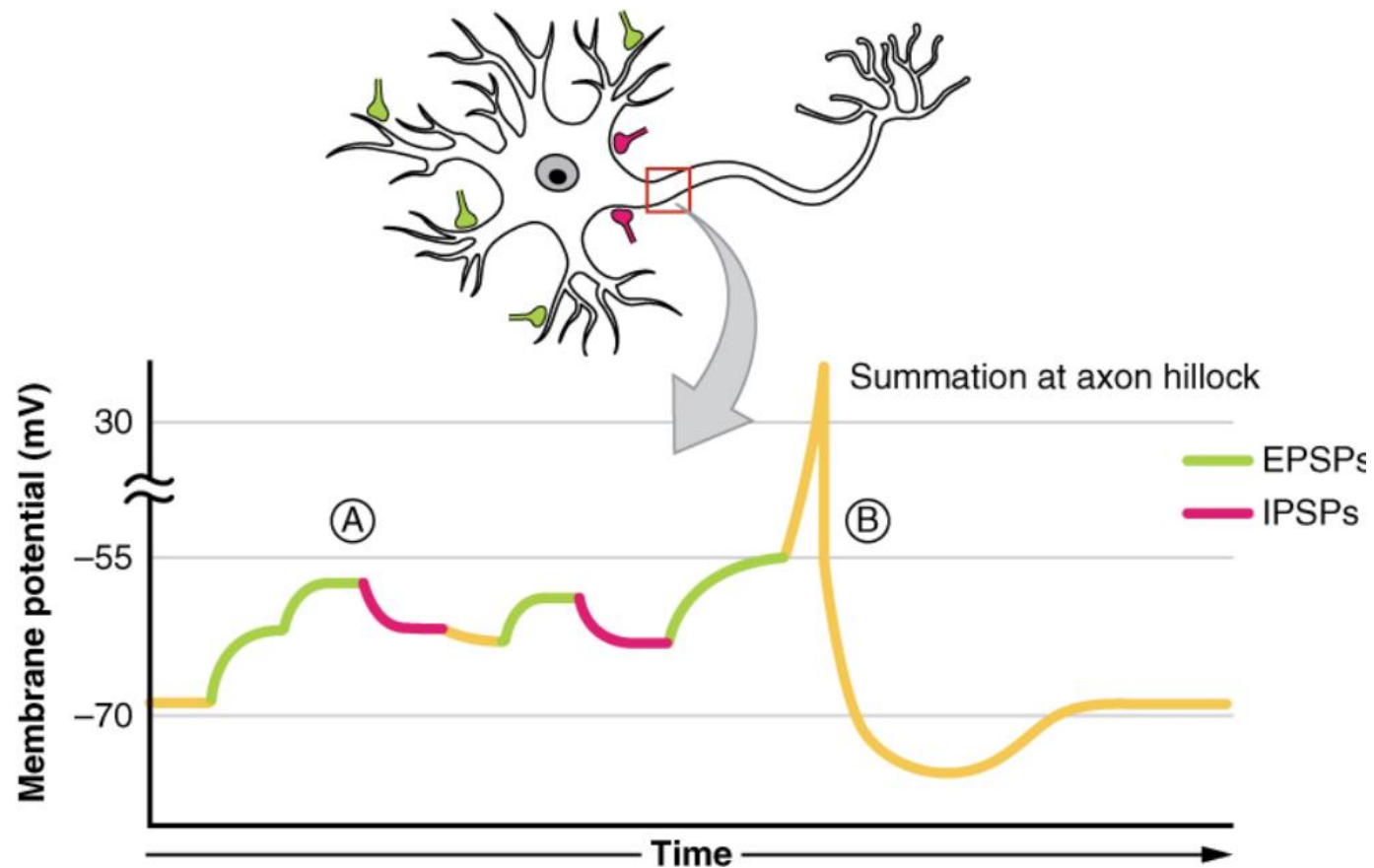


Stochastic optimal control

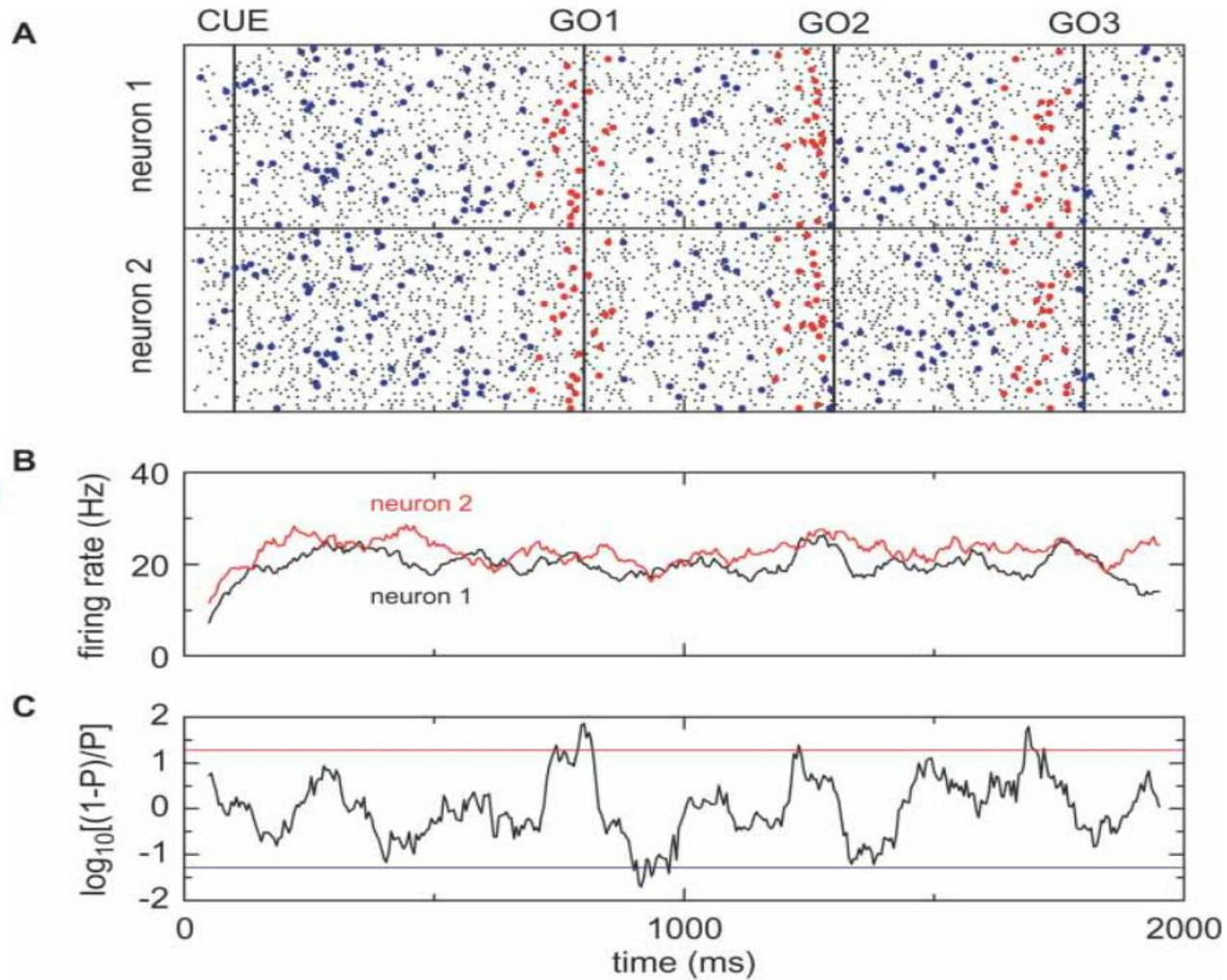
A. De Comité

LGBIO2072 – Practical Assignements

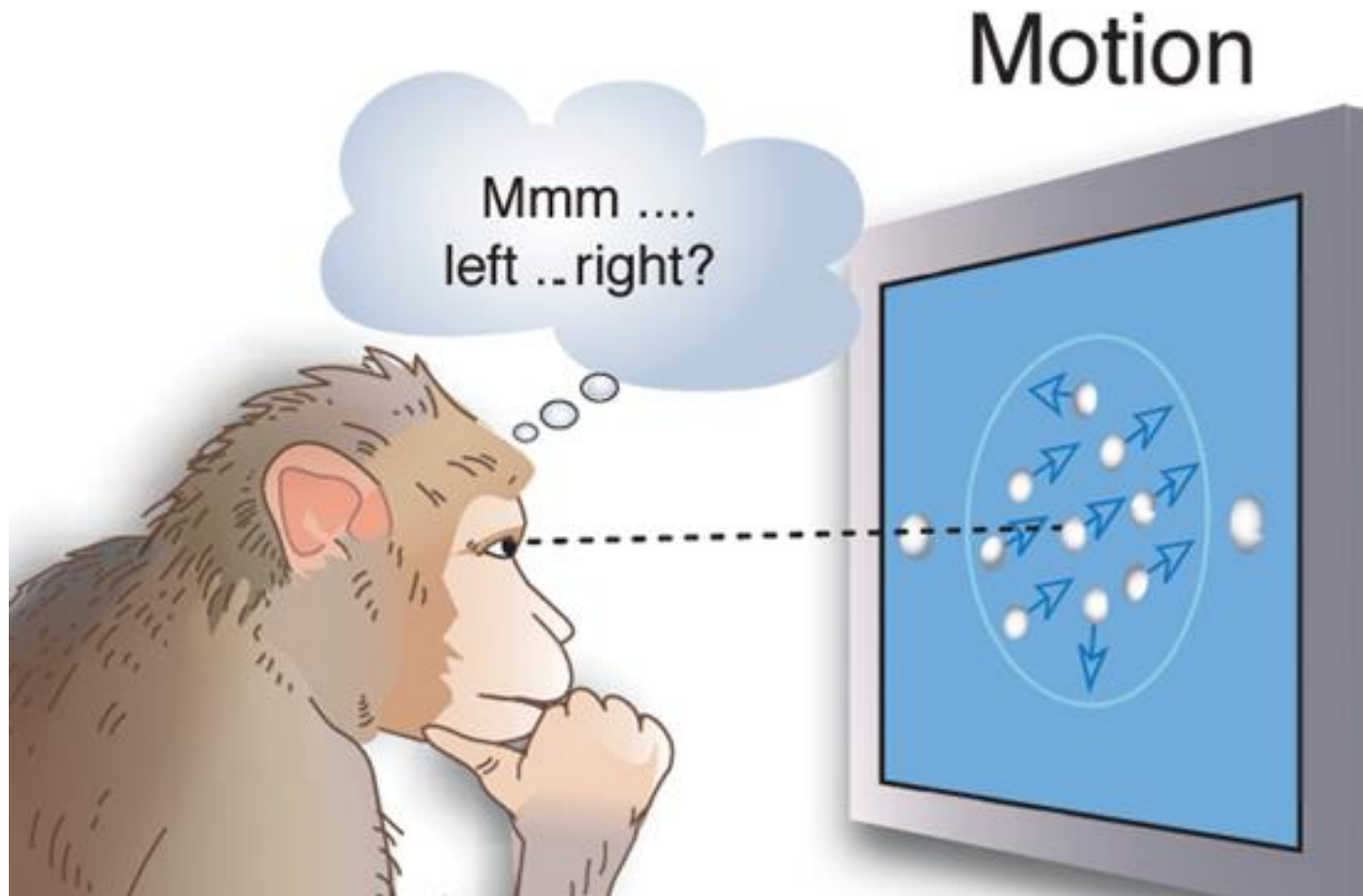
Introduction : Previous sessions



Introduction : Previous sessions



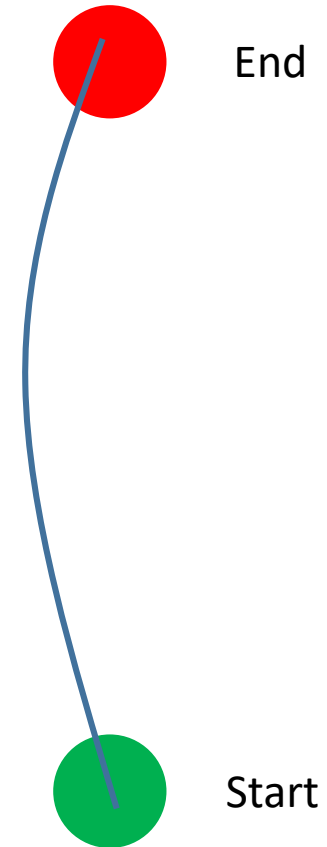
Introduction : Previous sessions



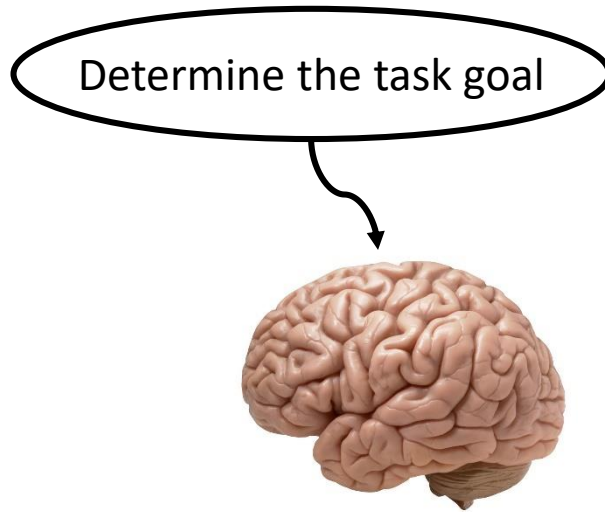
Introduction : Reaching movements



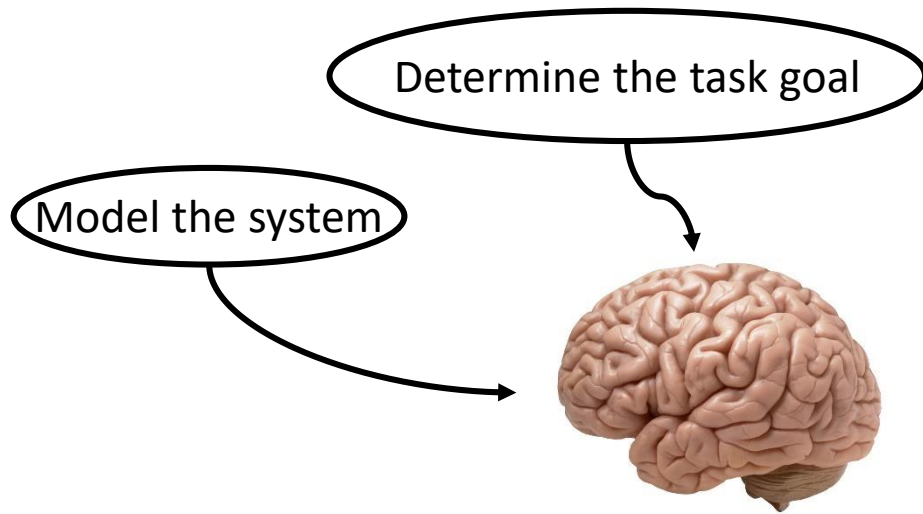
Introduction : Reaching movements



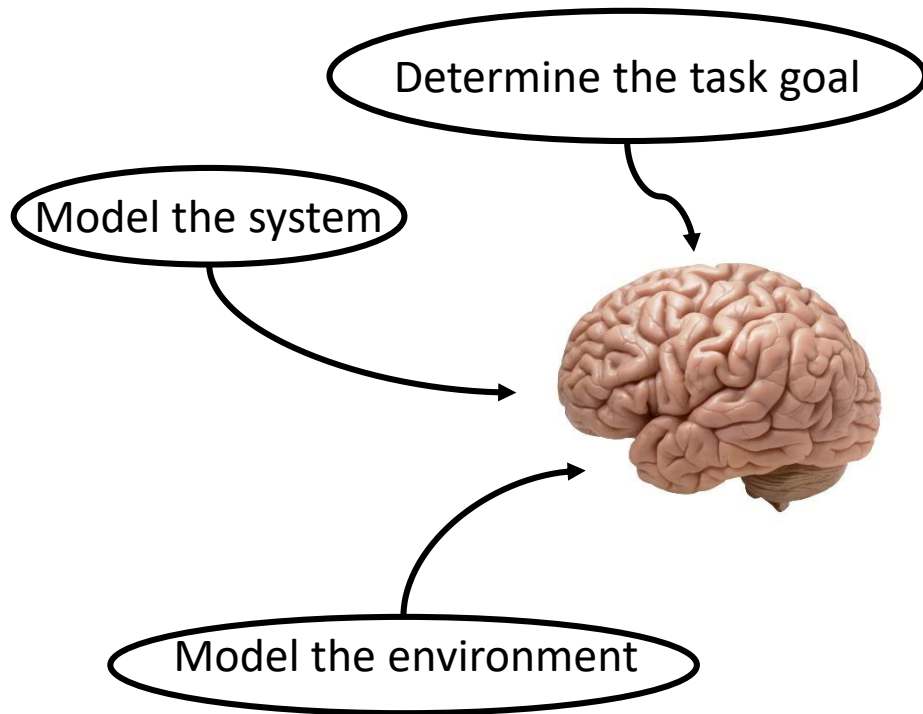
How does the brain implement this?



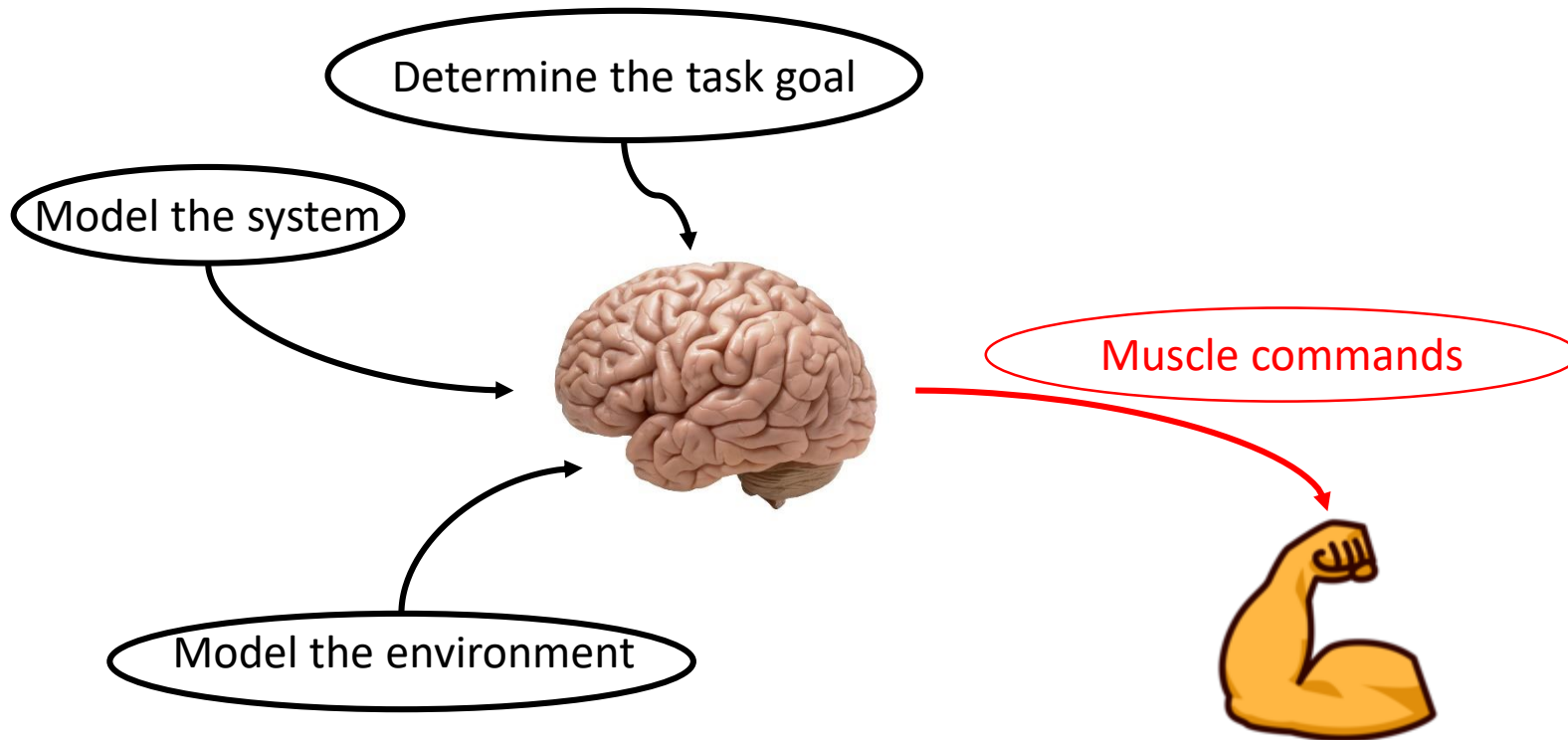
How does the brain implement this?



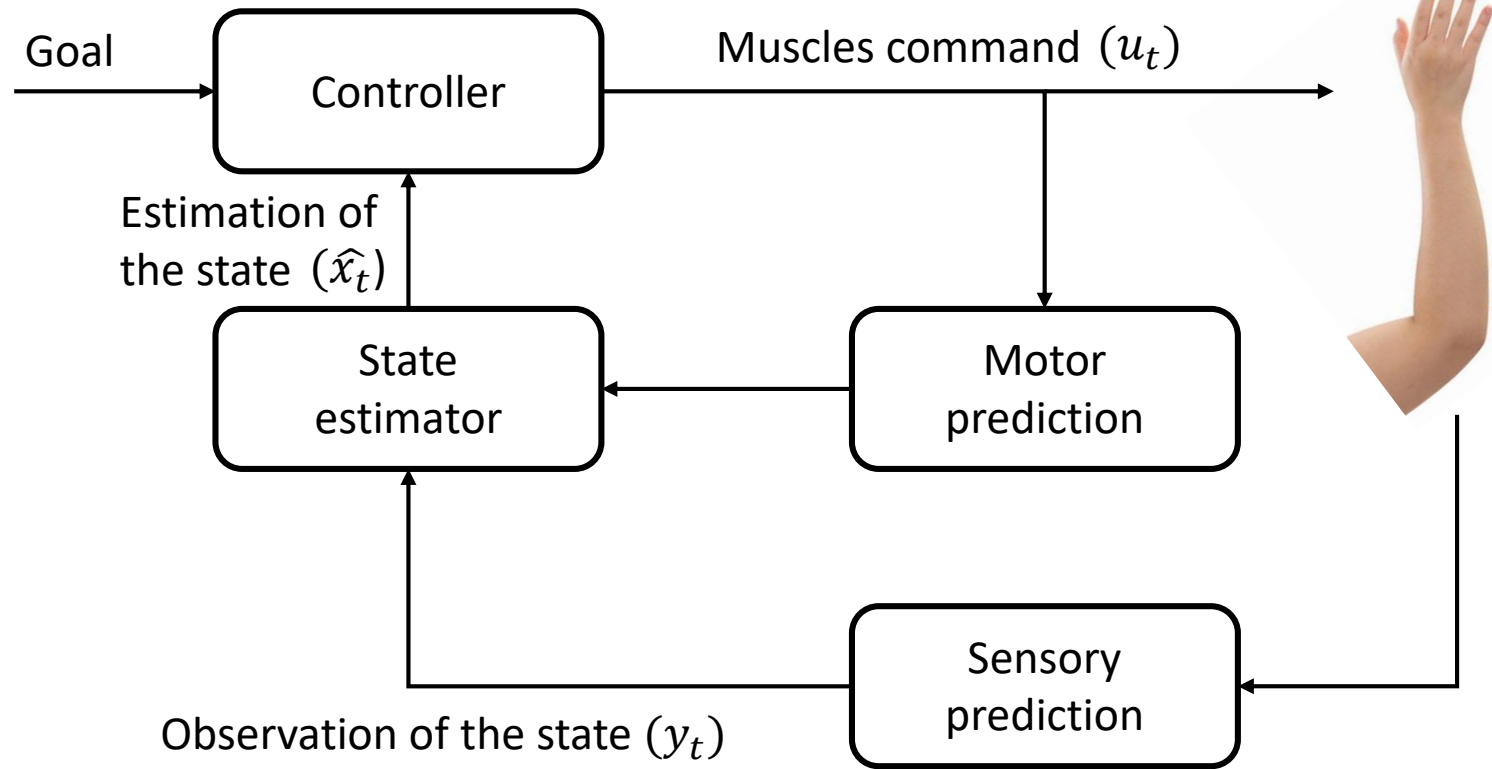
How does the brain implement this?



How does the brain implement this?



Introduction : What is optimal control?



The muscle commands are optimal (they minimise the cost function)

Outline

- How to model the system and the environment
- How to implement the task goal?
- How to compute the optimal serie of commands?
- Presentation of the project

How to model the system?

System = Hand and forearm

└ Dynamics defined by :

$$\dot{\mathbf{x}} = A\mathbf{x} + B\mathbf{u}$$


State vector

Command vector

How to model the system?

System = Hand and forearm

└ Dynamics defined by :

$$\dot{\mathbf{x}} = \mathbf{A}\mathbf{x} + \mathbf{B}\mathbf{u}$$


State vector

- State of the system
- State of the environment
- State of the target

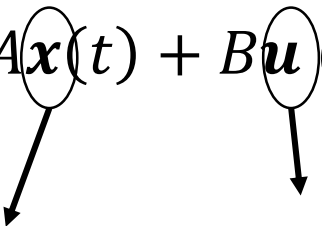
Command vector

- Contains the muscle command sent by the brain to the arms

How to model the system?

System = Hand and forearm

└ Dynamics defined by :

$$\dot{\mathbf{x}}(t) = A\mathbf{x}(t) + B\mathbf{u}(t)$$


State vector

- State of the system
- State of the environment
- State of the target

Command vector

- Contains the muscle command sent by the brain to the arms

How to obtain this equation?

Equations of movement

Using Newton's law... (mainly the second)

$$\sum \vec{F} = m \vec{a} \xrightarrow{\text{In both directions!!! (if working in 2D)}}$$

$$\sum \vec{T} = I \vec{a}_R \quad \Leftrightarrow \quad I \ddot{\theta} = -G \dot{\theta} + T$$

$$\text{Filter on the muscle : } \tau \dot{T} = u - T$$

Equations of movement

Using Newton's law... (mainly the second)

$$\sum \vec{F} = m \vec{a} \xrightarrow{\text{In both directions!!! (if working in 2D)}}$$

$$\sum \vec{T} = I \vec{a}_R \quad \Leftrightarrow \quad I \ddot{\theta} = -G \dot{\theta} + T$$

Filter on the muscle : $\tau \dot{T} = u - T$

To obtain this equation : $\dot{\mathbf{x}}(t) = \mathbf{A}\mathbf{x}(t) + \mathbf{B}\mathbf{u}(t)$

1. Write all the movement equations
2. Put all the variables describing the system in the state vector
3. Fill the A and B matrices

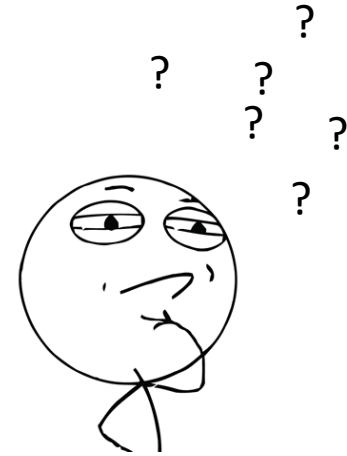
You should obtain a continuous time matricial differential equation

Equations of movement

$$\dot{\mathbf{x}}(t) = \mathbf{A}\mathbf{x}(t) + \mathbf{B}\mathbf{u}(t) \quad \text{Continuous}$$

Finite difference schemes

Discrete



We need discrete-time representation because the controller is designed to work on finite horizon.

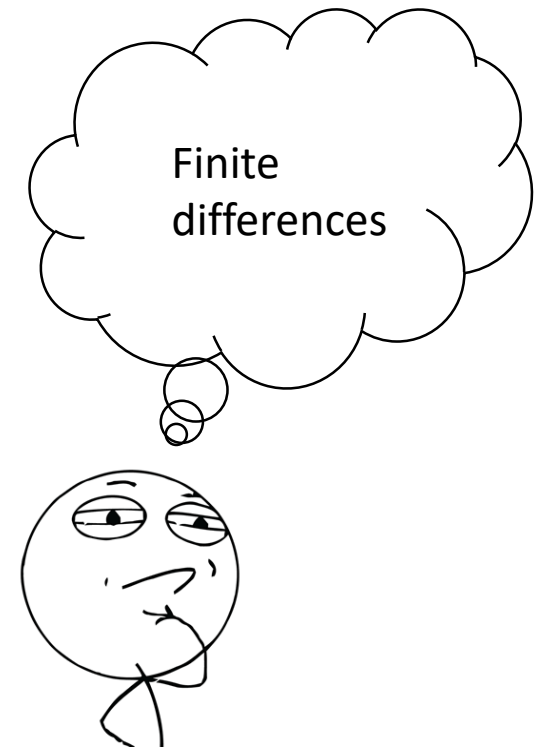
Equations of movement

$$\dot{\mathbf{x}}(t) = \mathbf{A}\mathbf{x}(t) + \mathbf{B}\mathbf{u}(t) \quad \text{Continuous}$$

Finite difference schemes

Discrete

$$\mathbf{x}[k + 1] = \mathbf{A}\mathbf{x}[k] + \mathbf{B}\mathbf{u}[k]$$

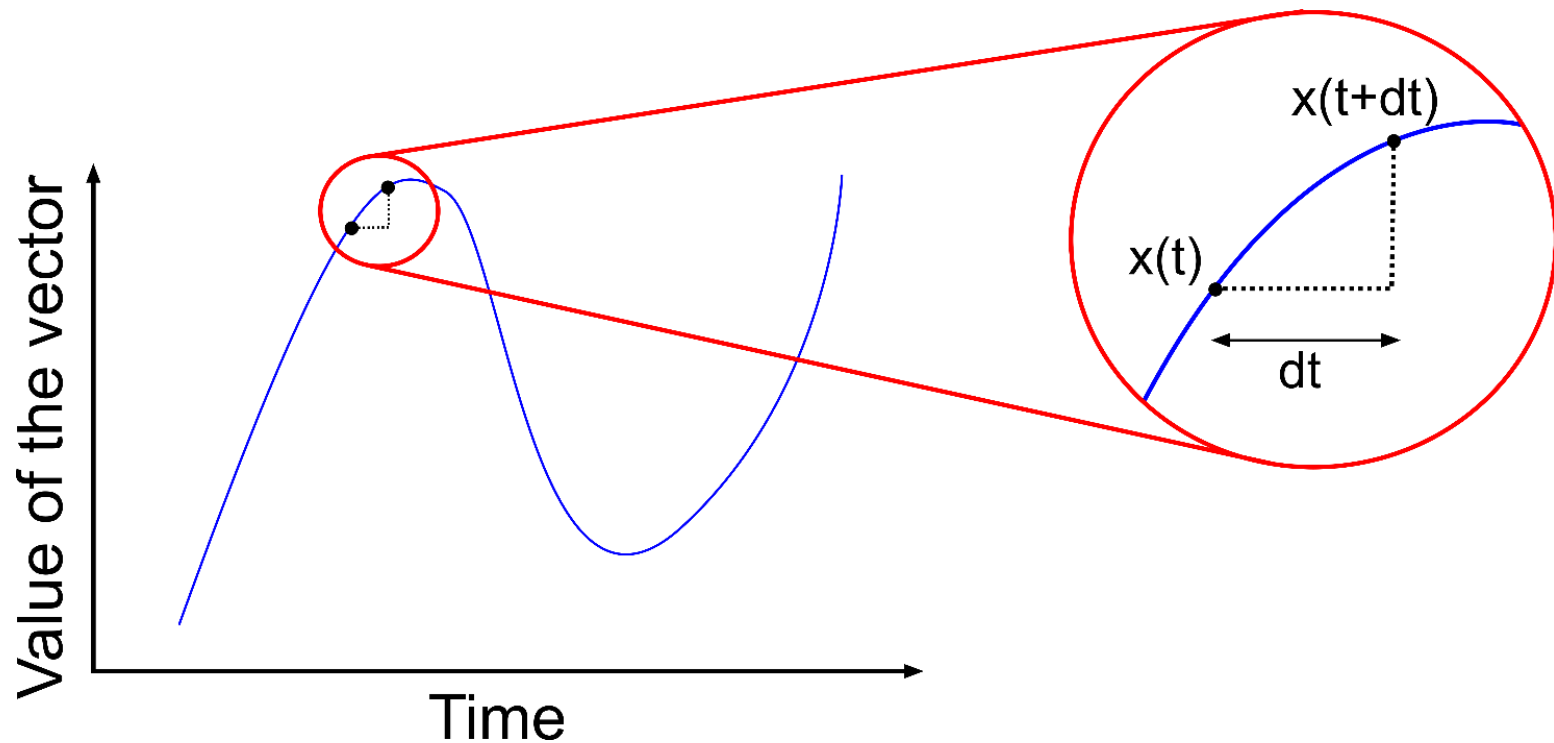


We need discrete-time representation because the controller is designed to work on finite horizon.

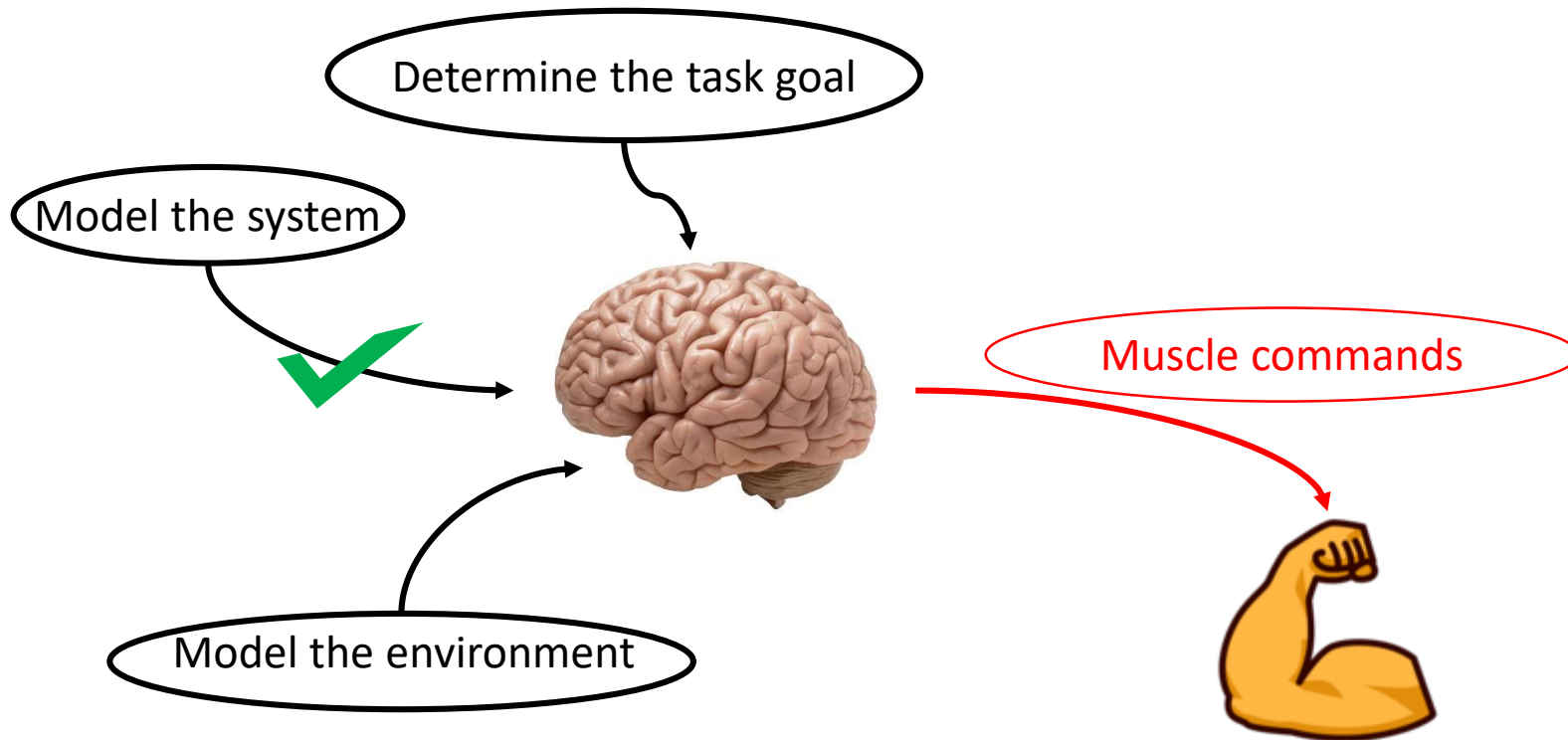
Reminder : Euler implicit (one integration method for more see LFSAB1104)

Allows to approximate a derivative by a finite difference

$$\dot{x}(k) \cong \frac{x(k+1) - x(k)}{\delta t}$$



How does the brain implement this?



How to determine and model the task goal ?

Task goal is associated by cost function :

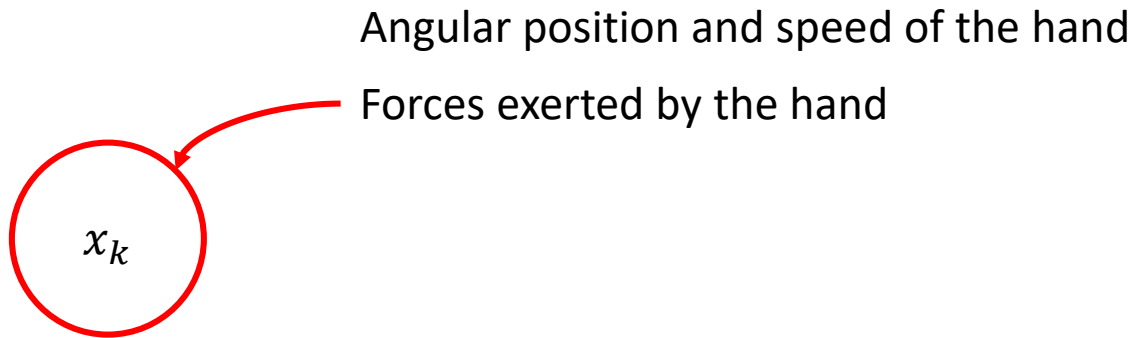
Cost function = end-point error + energetic cost

$$J(x, u) = x_N^T Q_N x_N + \sum_{k=1}^{N-1} (x_k^T Q_k x_k + u_k^T R u_k)$$

How does this take the position of the target into account?

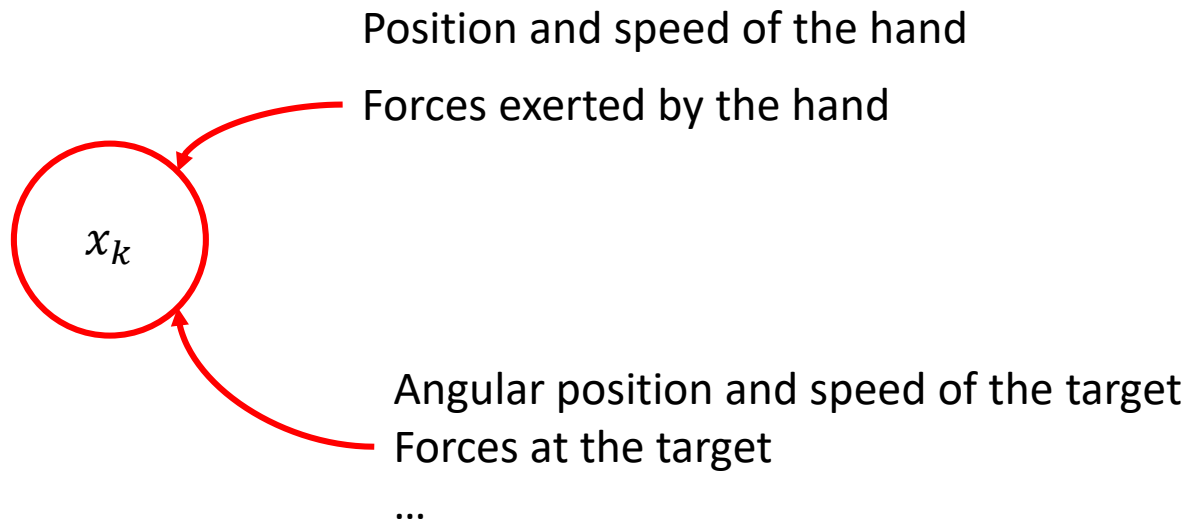
What's inside the state?

Let's take a look at the state of the system...



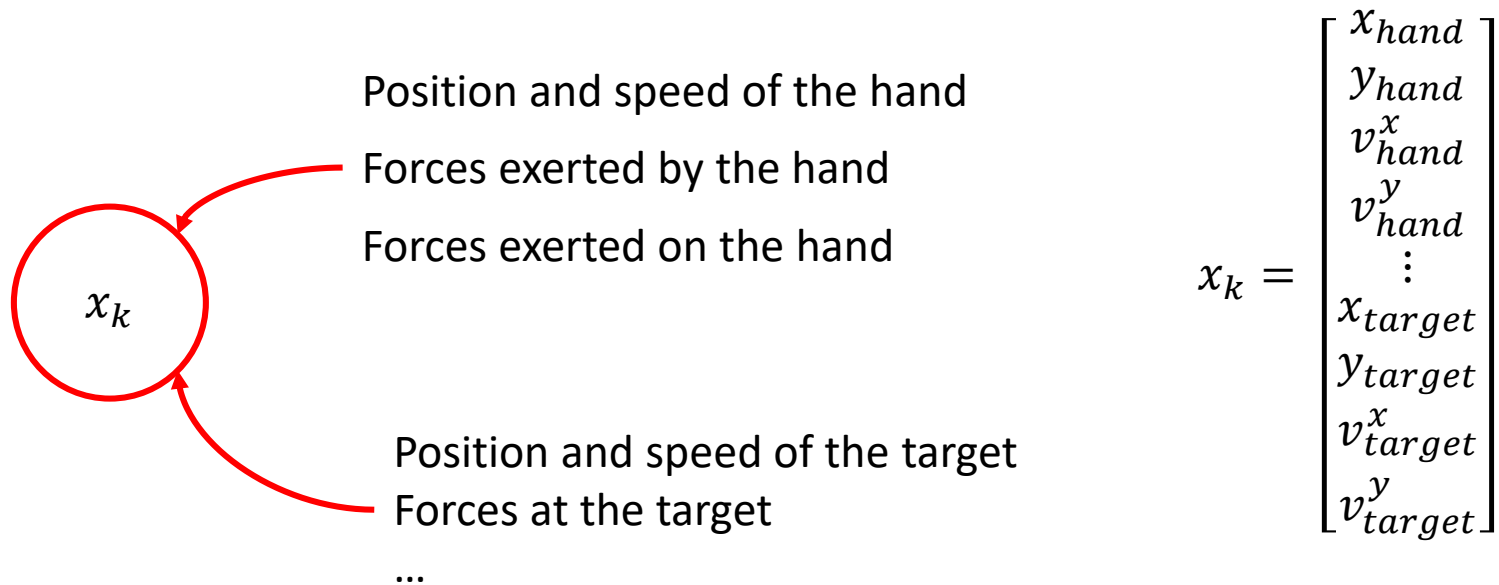
What's inside the state?

Let's take a look at the state of the system...



What's inside the state?

Let's take a look at the state of the system...



Extension of the state!!! ➔ Extension of the A and B matrices to match the dimensions!!!

Objective function?

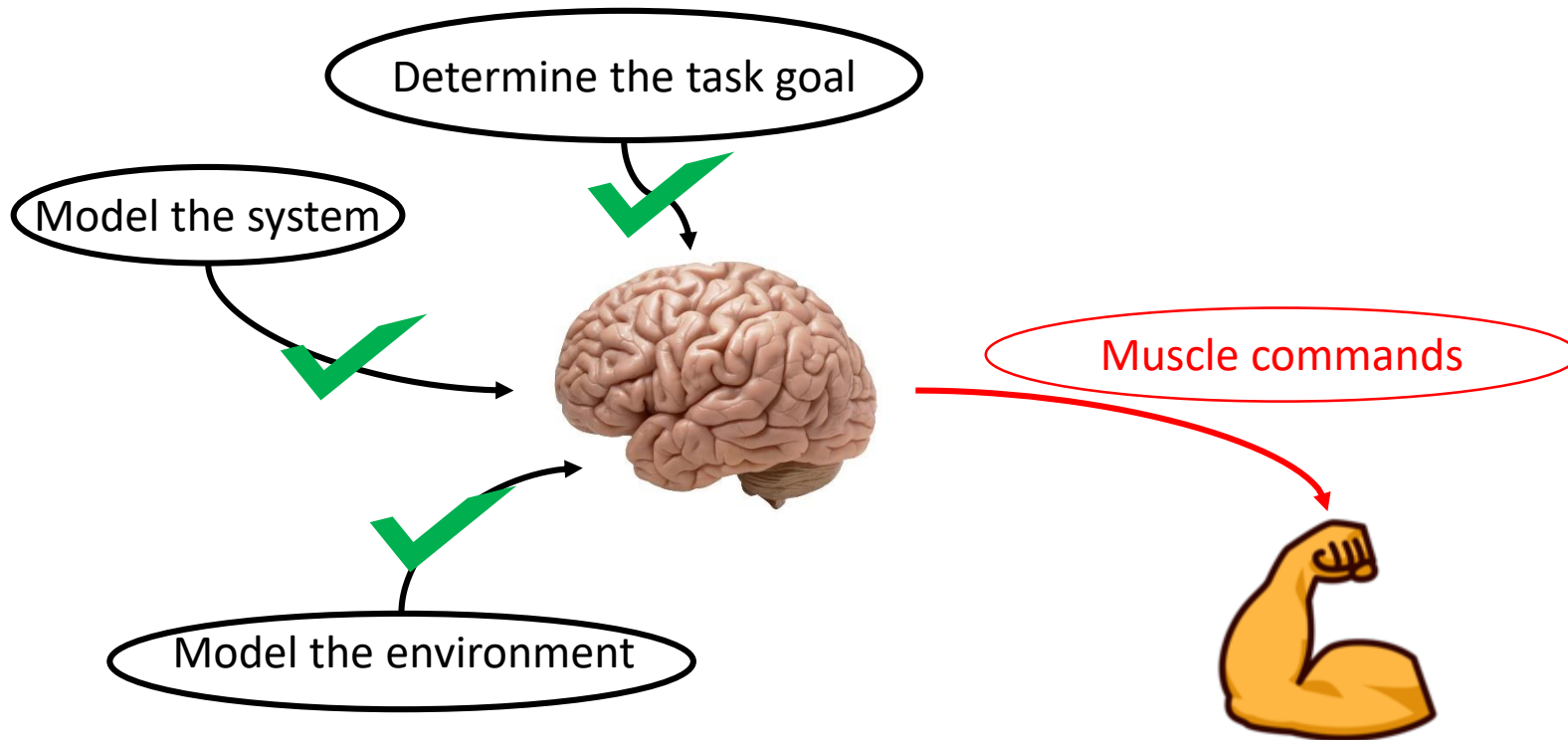
$$J(\mathbf{x}) = w_1(\theta_k - \theta_k^*)^2 + w_2(\dot{\theta}_k - \dot{\theta}_k^*)^2 + w_3(T_k - T_k^*)^2$$



$$J(\mathbf{x}) = \mathbf{x}[k]^T Q_N \mathbf{x}[k]$$

Goal : Find the entries of the Q_N matrix

How does the brain implement this?



How to compute the optimal serie of command?

$$\mathbf{x}[k + 1] = A\mathbf{x}[k] + B\mathbf{u}[k]$$

Dynamics of the system
(= how it can evolve)

$$J(x, u) = \mathbf{x}_N^T \mathbf{Q}_N \mathbf{x}_N + \sum_{k=1}^{N-1} (\mathbf{x}_k^T \mathbf{Q}_k \mathbf{x}_k + \mathbf{u}_k^T \mathbf{R} \mathbf{u}_k)$$

Where and how we want to perform
the task?

How to compute the optimal serie of command?

$$\mathbf{x}[k + 1] = A\mathbf{x}[k] + B\mathbf{u}[k]$$

Dynamics of the system
(= how it can evolve)

$$J(x, u) = \mathbf{x}_N^T \mathbf{Q}_N \mathbf{x}_N + \sum_{k=1}^{N-1} (\mathbf{x}_k^T \mathbf{Q}_k \mathbf{x}_k + \mathbf{u}_k^T \mathbf{R} \mathbf{u}_k)$$

Where and how we want to perform
the task?

$$\min_{u_1, u_2, \dots, u_N} J(x, u)$$

How to compute the optimal serie of command?

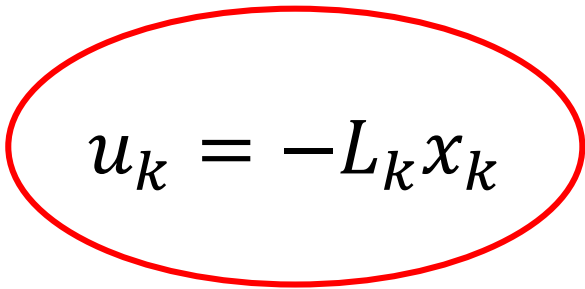
The solution is optimal in the sense that it **minimises the objective function**

In the fully observable case, we have the following recurrence

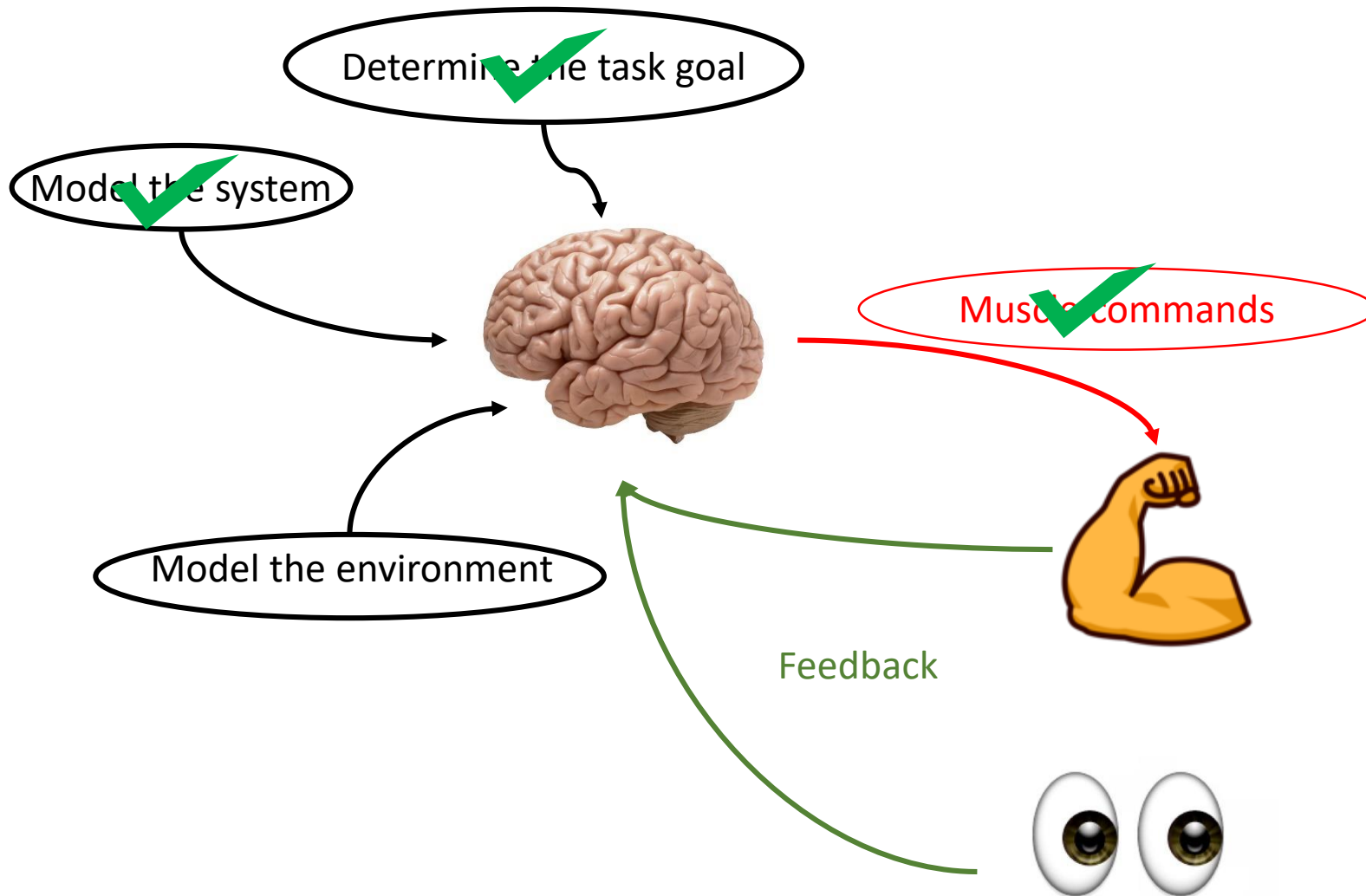
$$\begin{aligned}L_k &= (R + B^T S_{k+1} B)^{-1} B^T S_{k+1} A \\S_k &= Q_k + A^T S_{k+1} (A - B L_k) \\S_N &= Q_N, \quad S_N = 0 \\s_k &= s_{k+1} + \text{tr}(S_{k+1} \Omega_\xi)\end{aligned}$$

Simulation of the system

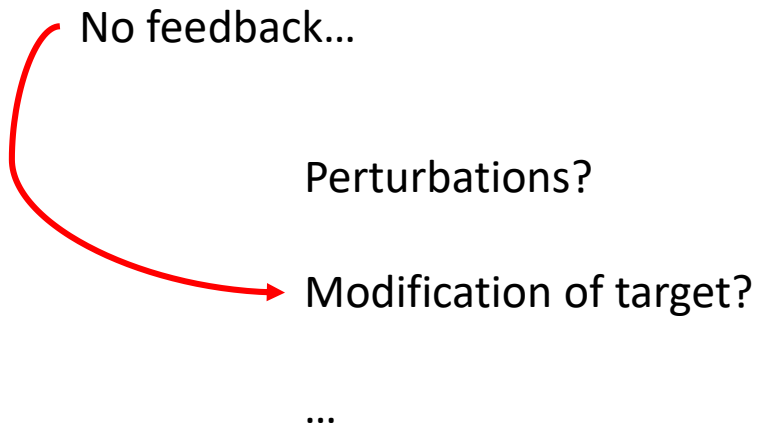
$$x_{k+1} = (A - B L_k) x_k + \xi_k$$


$$u_k = -L_k x_k$$

How does the brain implement this?



Problem?



Closed loop control = Blind control

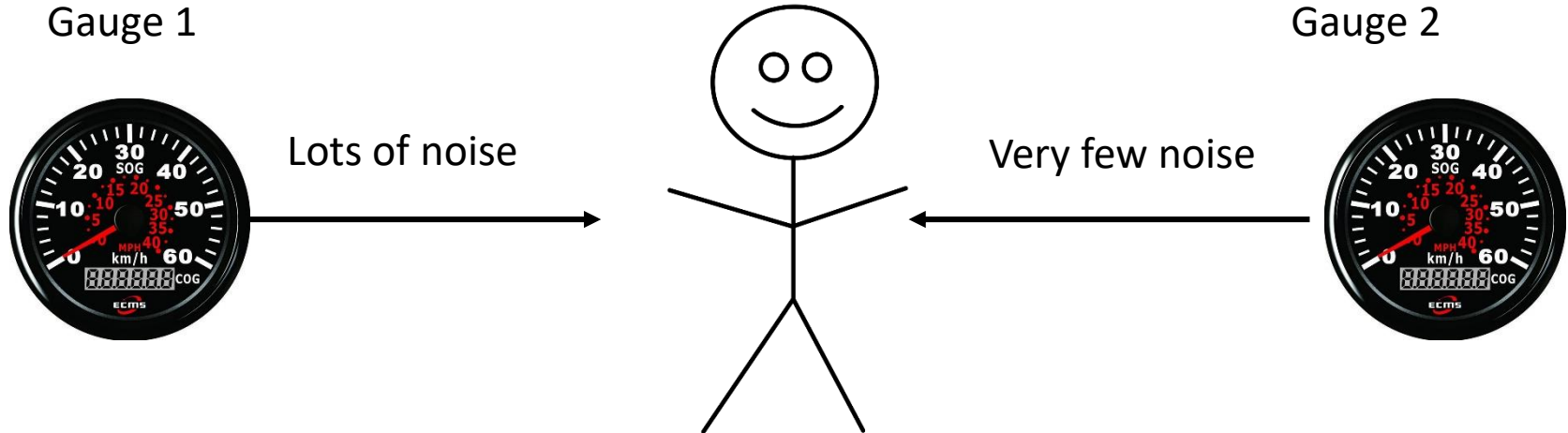
Human body has feedbacks...

Feedbacks from the human body

Eyes and proprioception

$$y_k = Hx_k + \omega_k$$

Noisy observations of the state

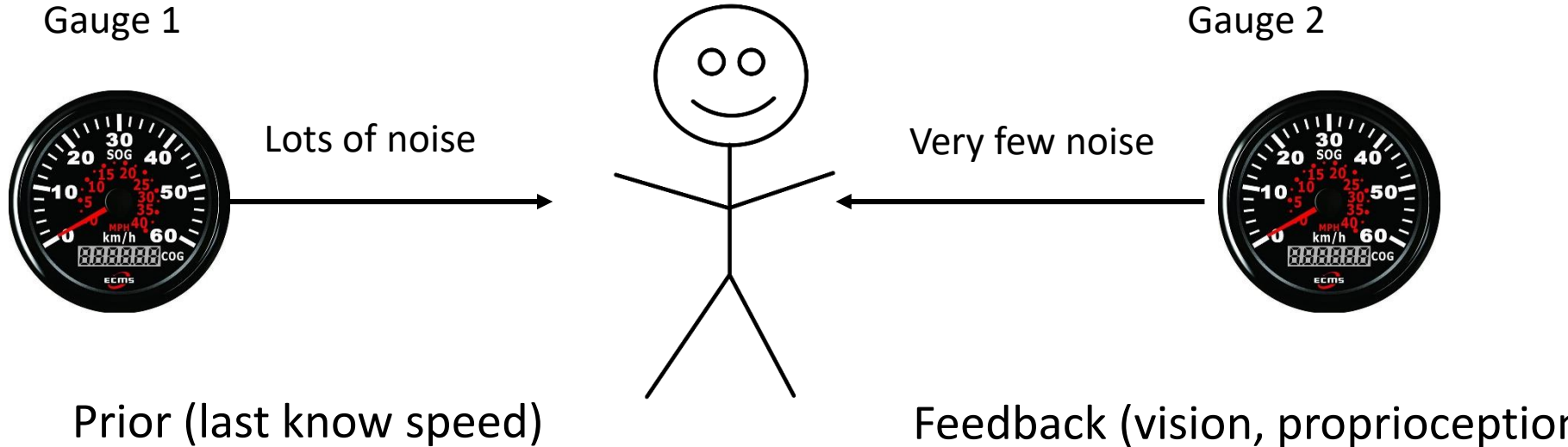


Feedbacks from the human body

Eyes and proprioception

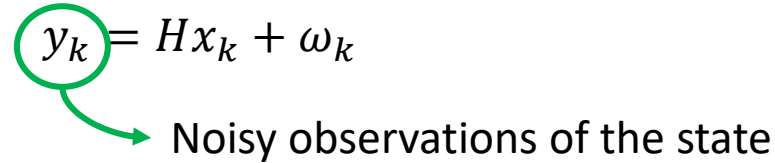
$$y_k = Hx_k + \omega_k$$

Noisy observations of the state



Feedbacks from the human body

Eyes and proprioception

$$y_k = Hx_k + \omega_k$$


Noisy observations of the state

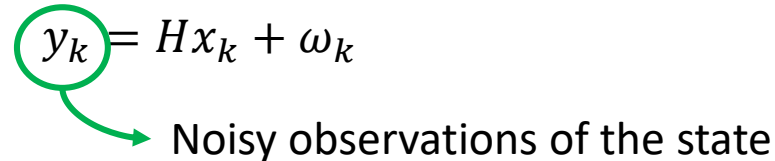
Combination of priors and feedback :

$$\hat{x}_{k+1} = (1 - K) * \text{prior} + K * \text{feedback}$$
$$\hat{x}_{k+1} = A \hat{x}_k + B u_k + K(y_k - H \hat{x}_k)$$

How to find the coefficient K?

Feedbacks from the human body

Eyes and proprioception

$$y_k = Hx_k + \omega_k$$


Noisy observations of the state

Combination of priors and feedback :

$$\hat{x}_{k+1} = (1 - K) * \text{prior} + K * \text{feedback}$$
$$\hat{x}_{k+1} = A \hat{x}_k + B u_k + K(y_k - H \hat{x}_k)$$

How to find the coefficient K?

Optimal estimation of the state



Ponderate source by their 'accuracy'

Computation of the different gains

$$\hat{x}_{k+1} = A \hat{x}_k + B u_k + K_k (y_k - H \hat{x}_k)$$

$$K_k = A \Sigma_k H^T (H \Sigma_k H^T + \Omega_\omega)^{-1}$$

$$\Sigma_{k+1} = \Omega_\xi + (A - K_k H) \Sigma_k A^T$$

The command becomes :

$$u_k = -L_k \hat{x}_k$$

Project

- See instructions on Moodle
- Groups of 2 students
- Write a report of 6 pages
- Due date : **17th of November 2020**