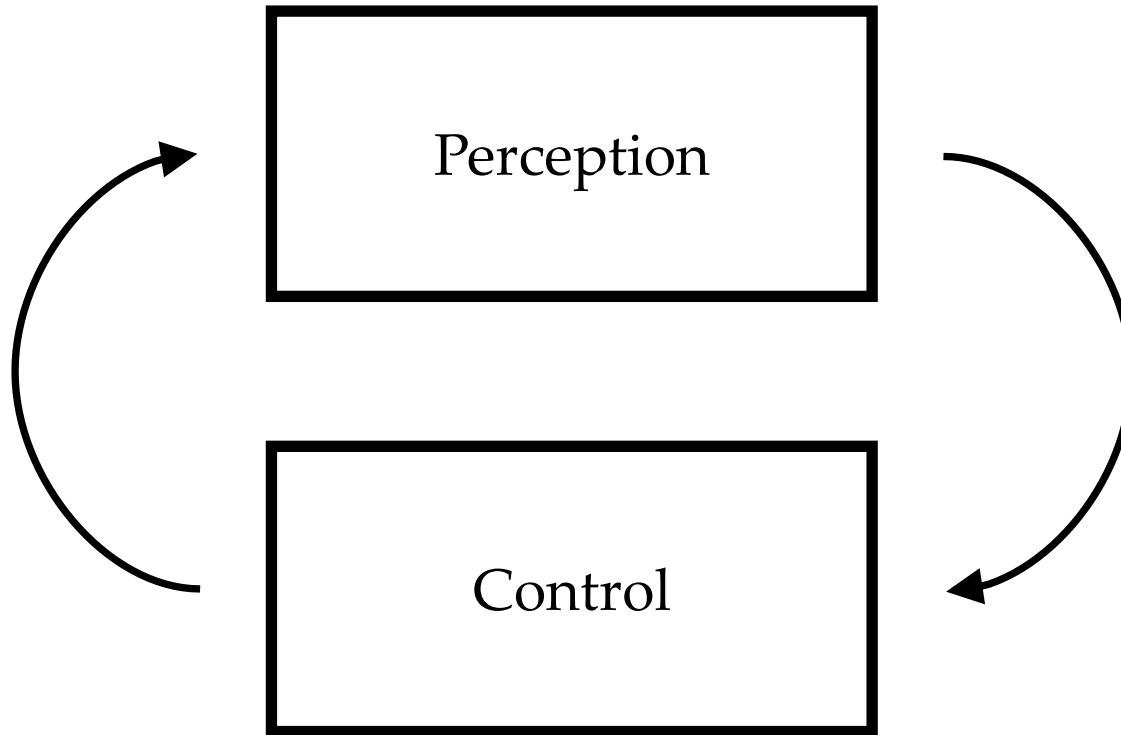
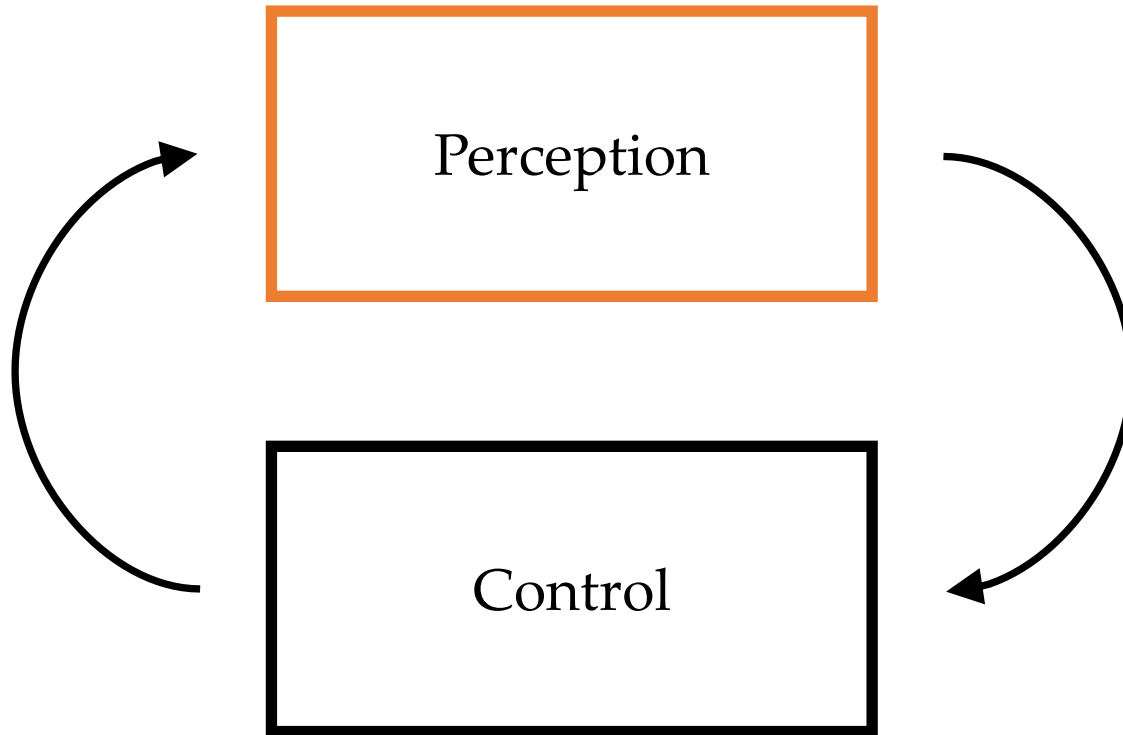


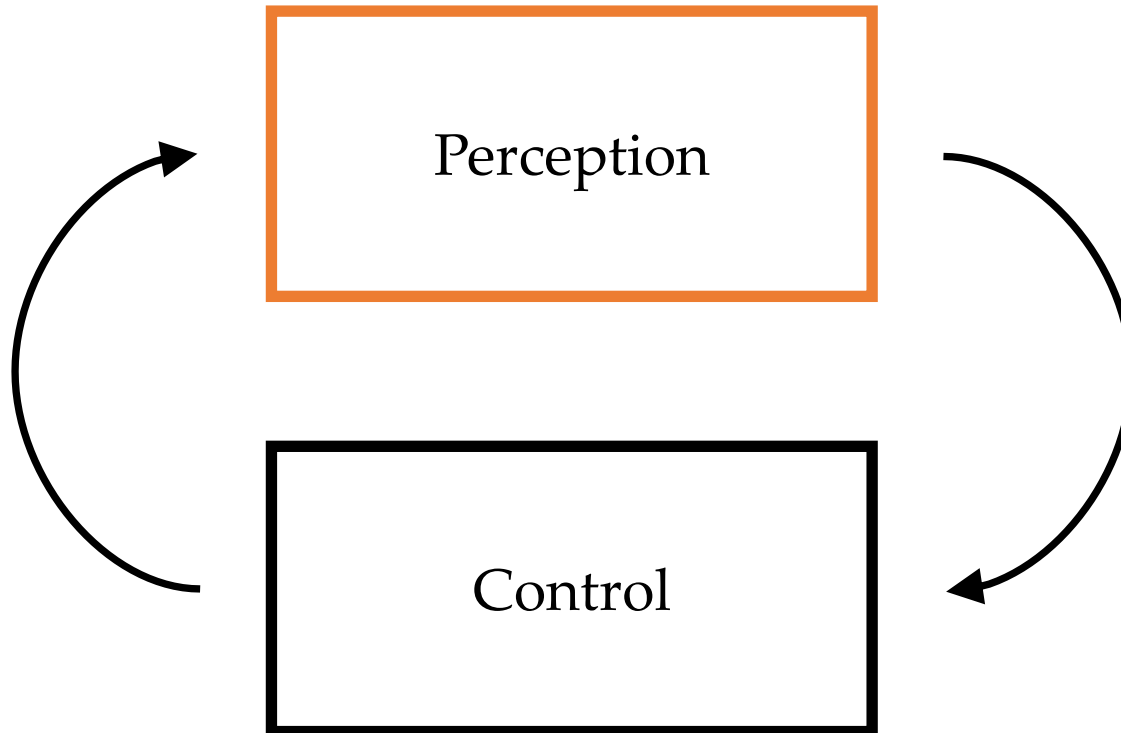
## **Class Overview**

CSCI 545 Introduction to Robotics  
Instructor: Stefanos Nikolaidis

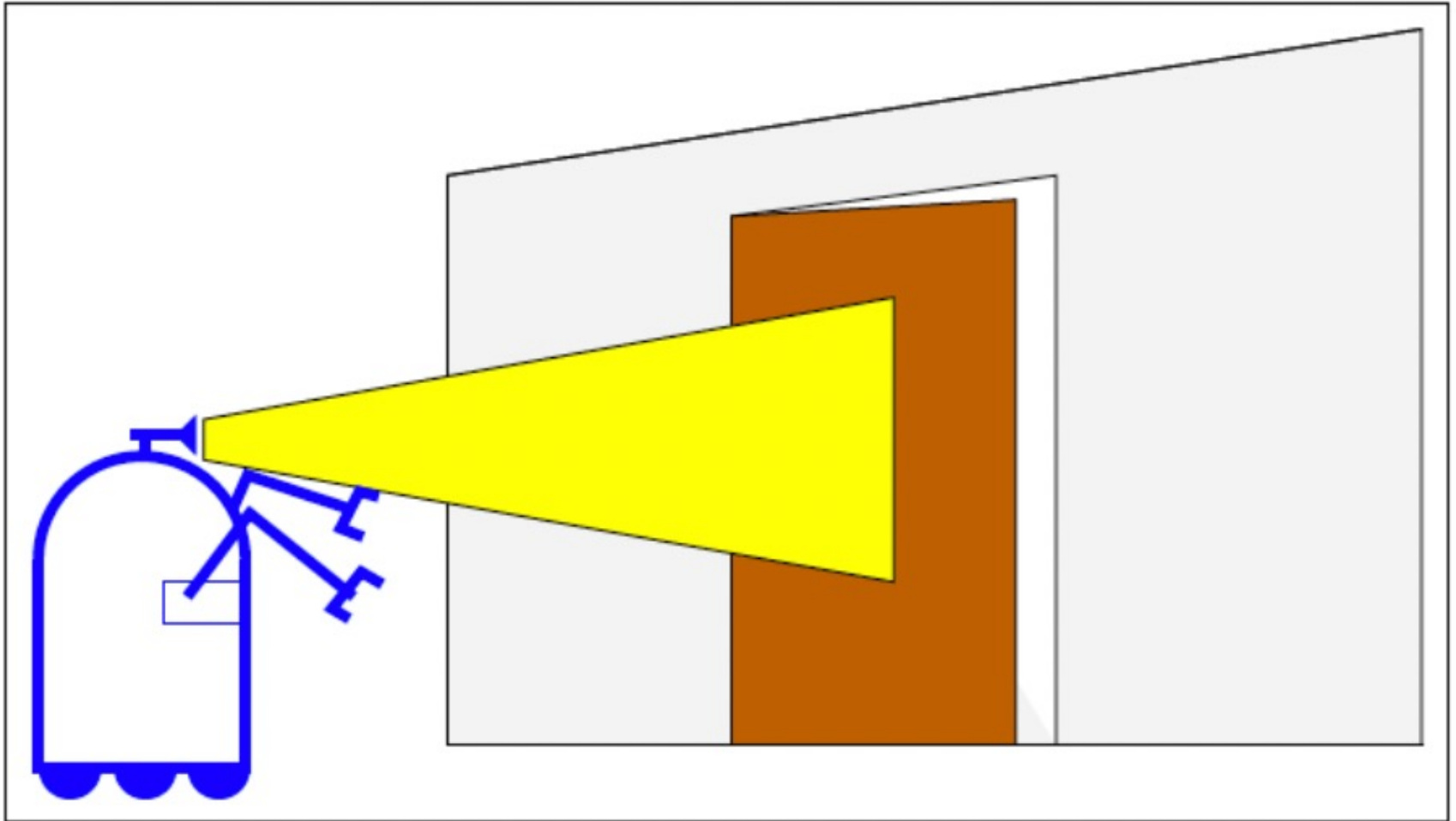




## Bayesian Filtering

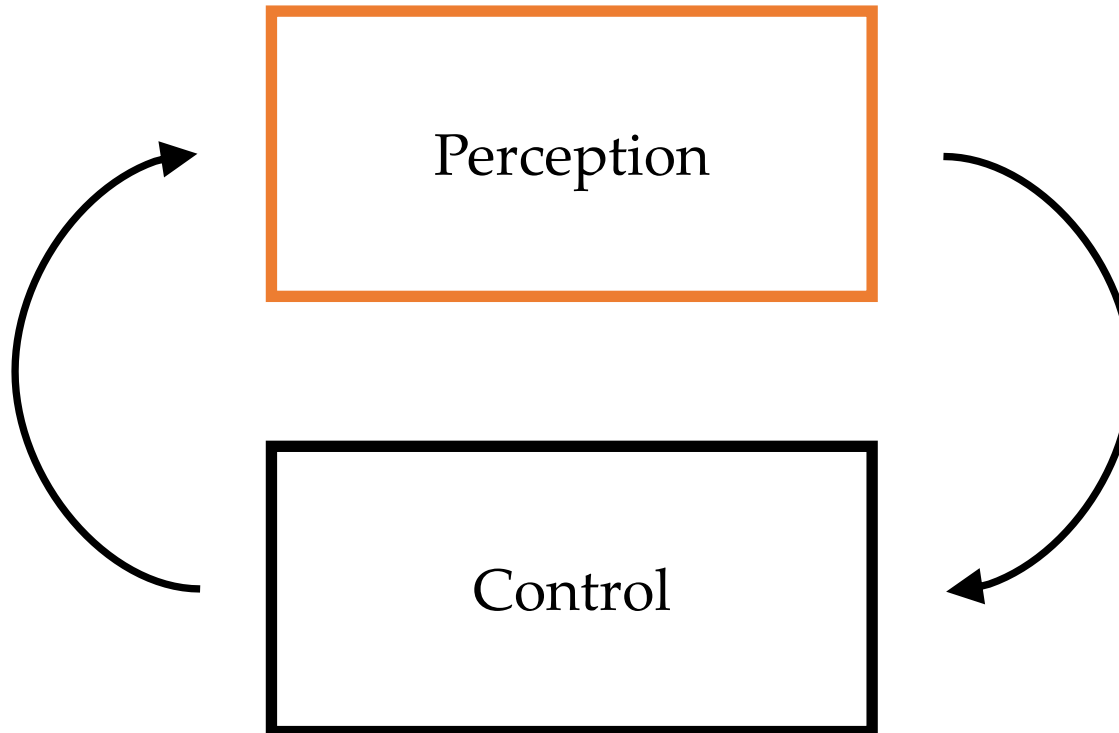


## Bayesian Filtering



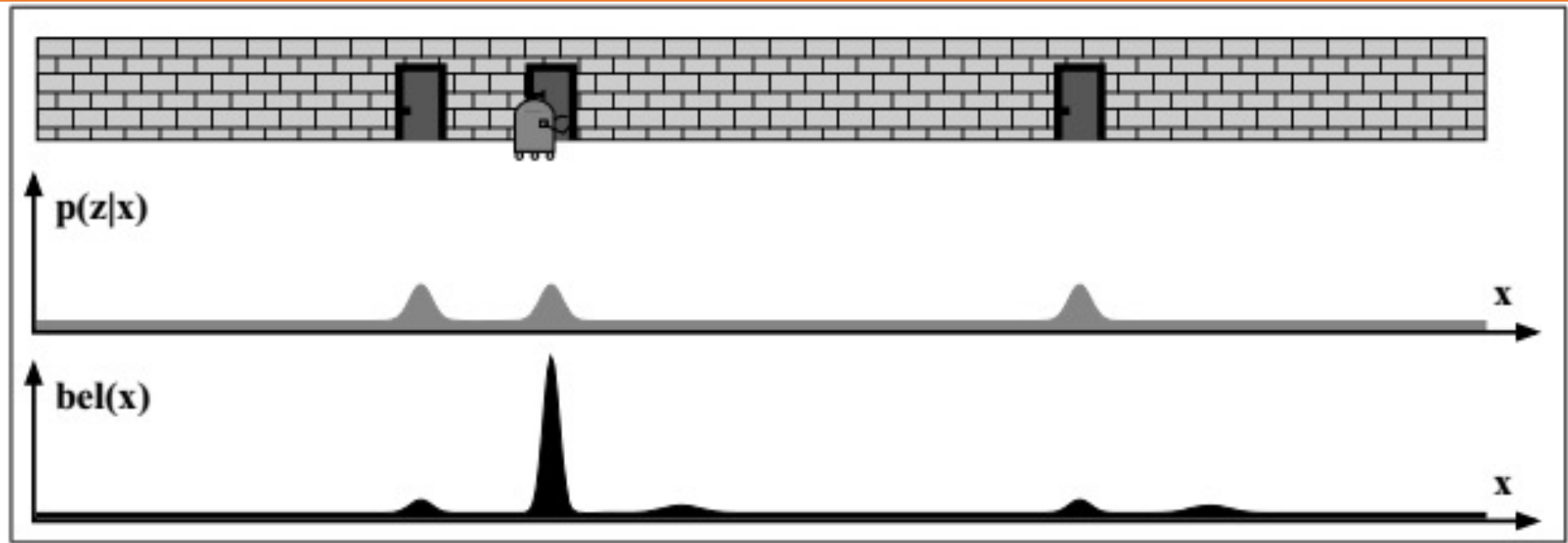
Bayesian Filtering

KF / EKF



## Bayesian Filtering

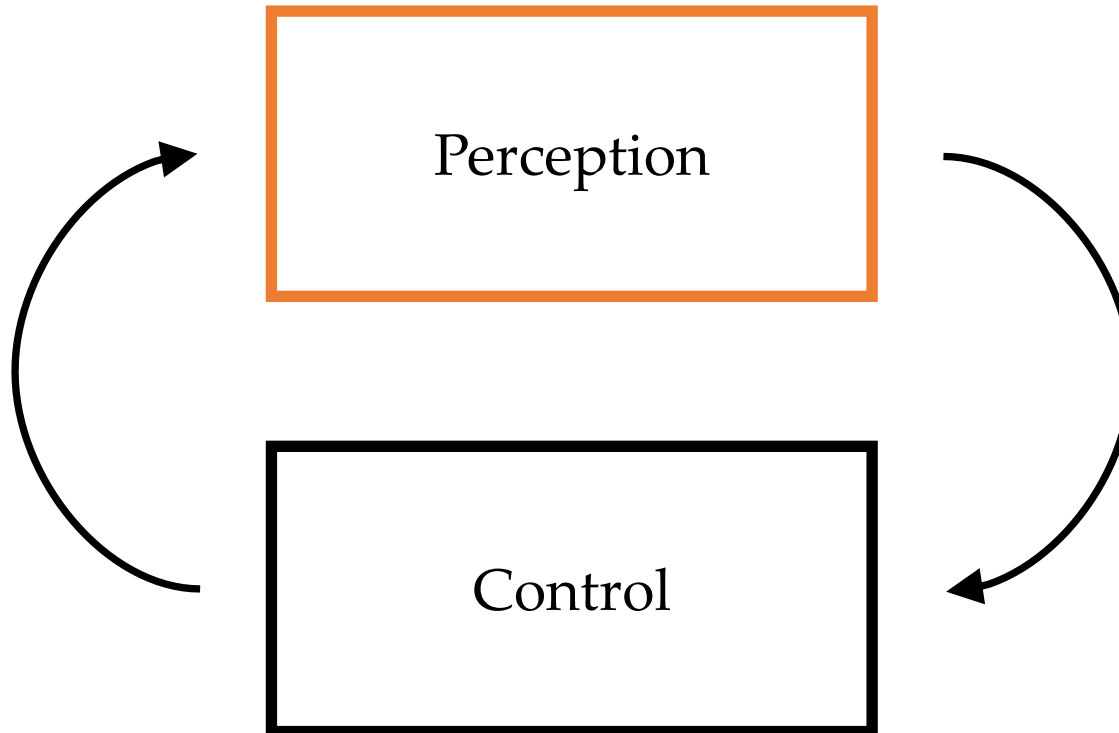
### KF / EKF



Bayesian Filtering

KF / EKF

Particle Filtering



Bayesian Filtering

Particle Filtering

KF / EKF

Localization

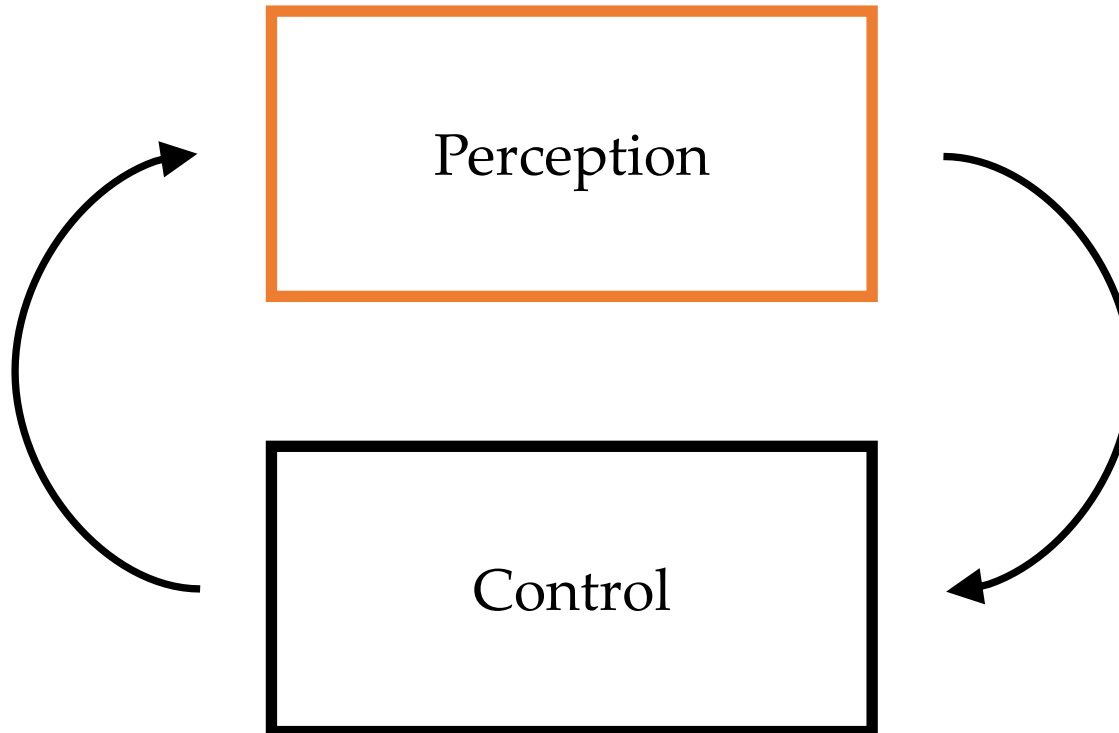


Bayesian Filtering

KF / EKF

Particle Filtering Mapping

Localization

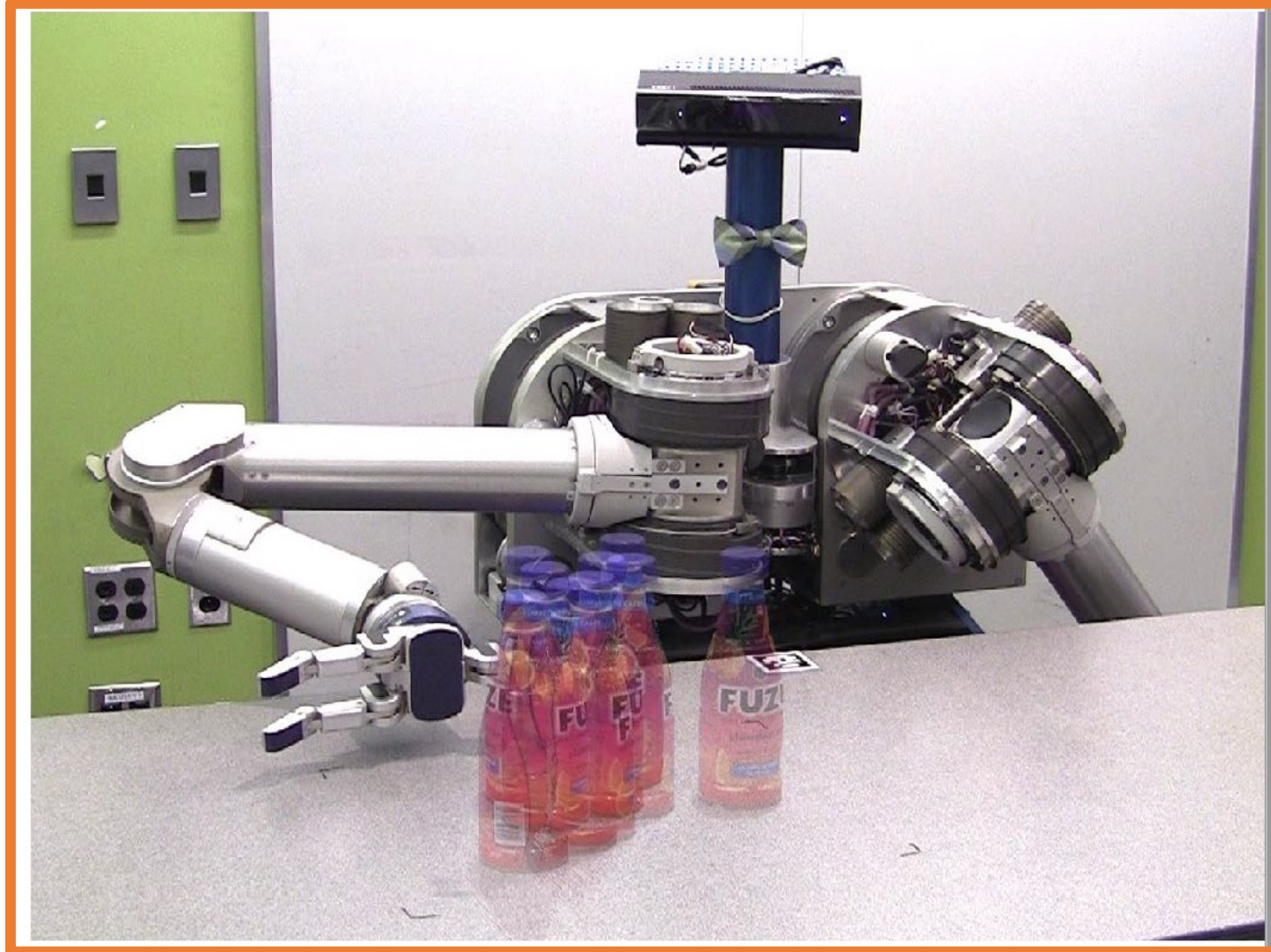


Bayesian Filtering

Particle Filtering Mapping

KF / EKF

Localization

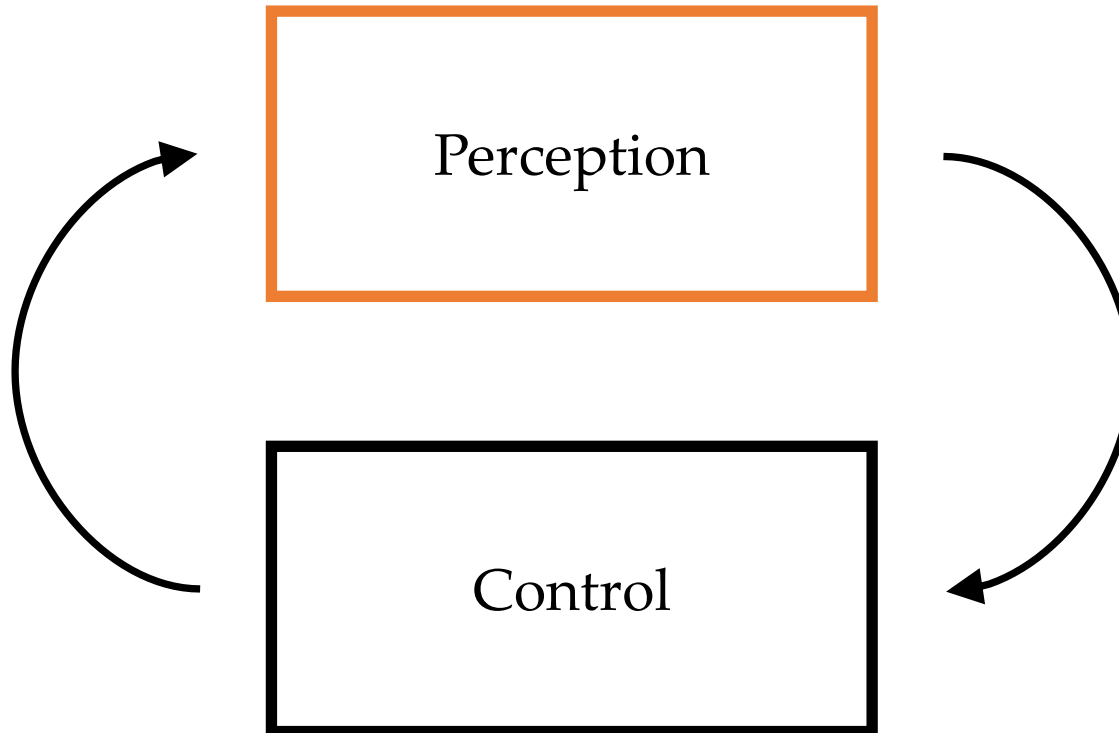


Bayesian Filtering

KF / EKF

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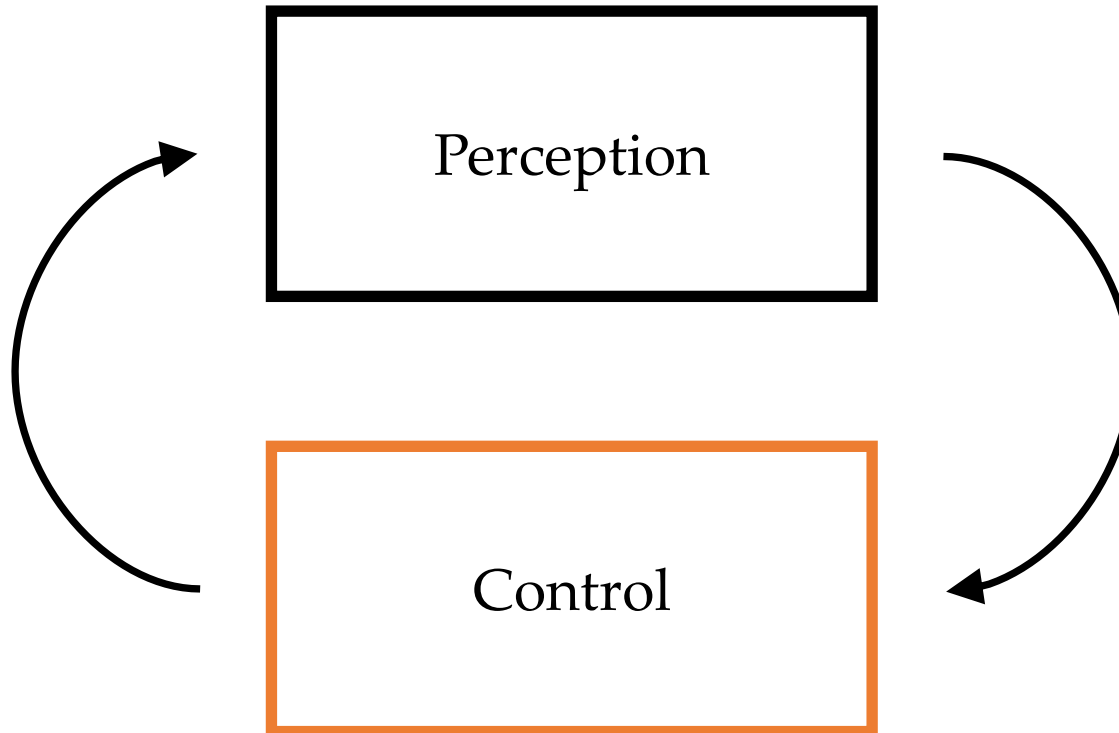


**Bayesian Filtering**

**KF / EKF**

**Particle Filtering    Mapping**

**Localization**

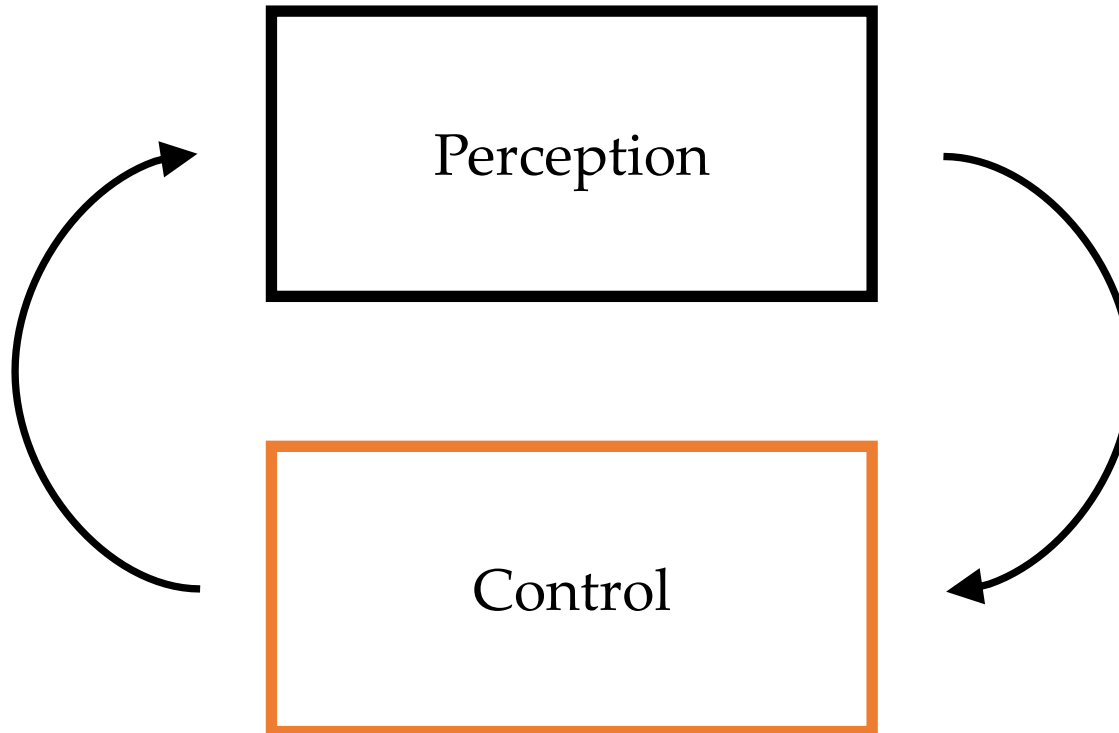


**Bayesian Filtering**

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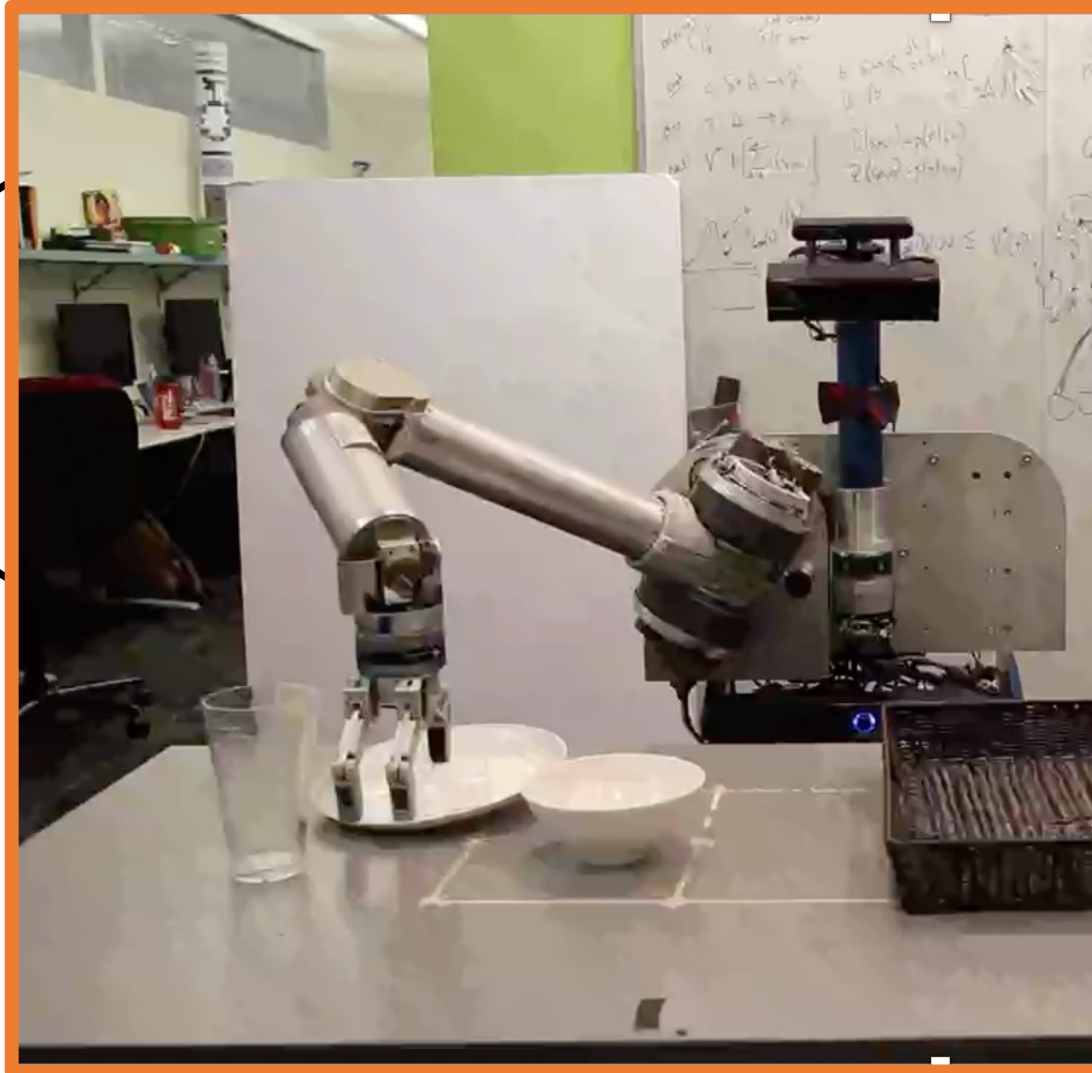


Bayesian Filtering

Particle Filtering Mapping

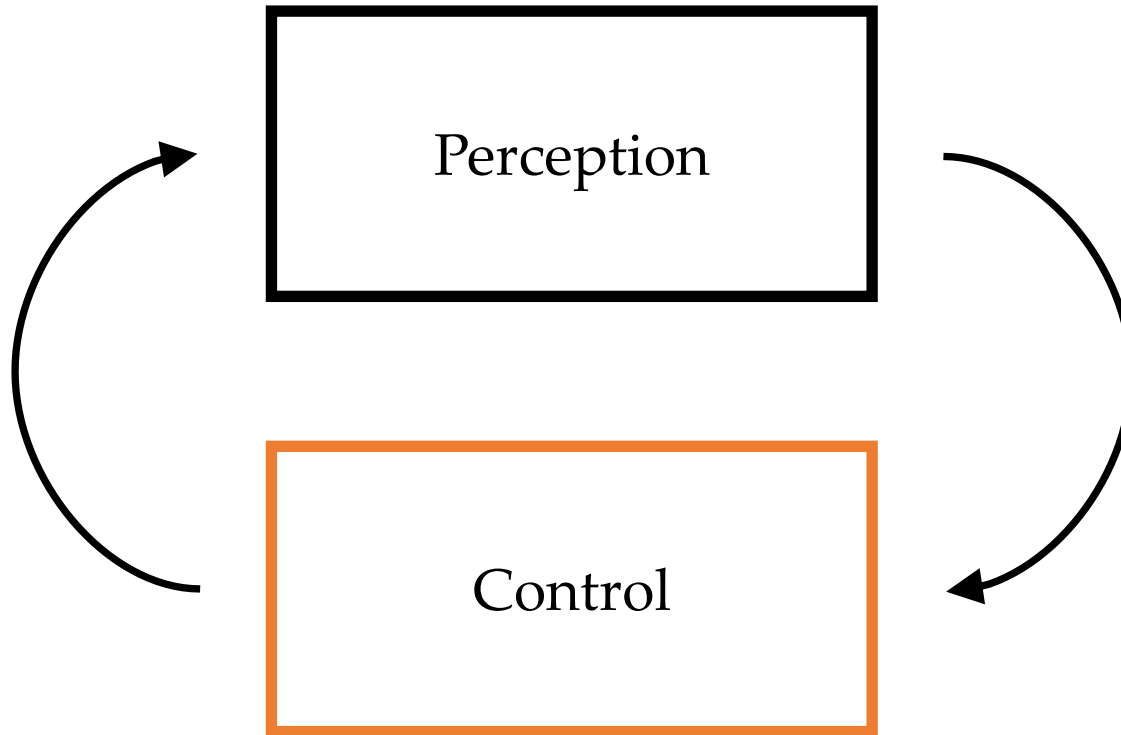
KF / EKF

Localization



**Bayesian Filtering**  
KF / EKF

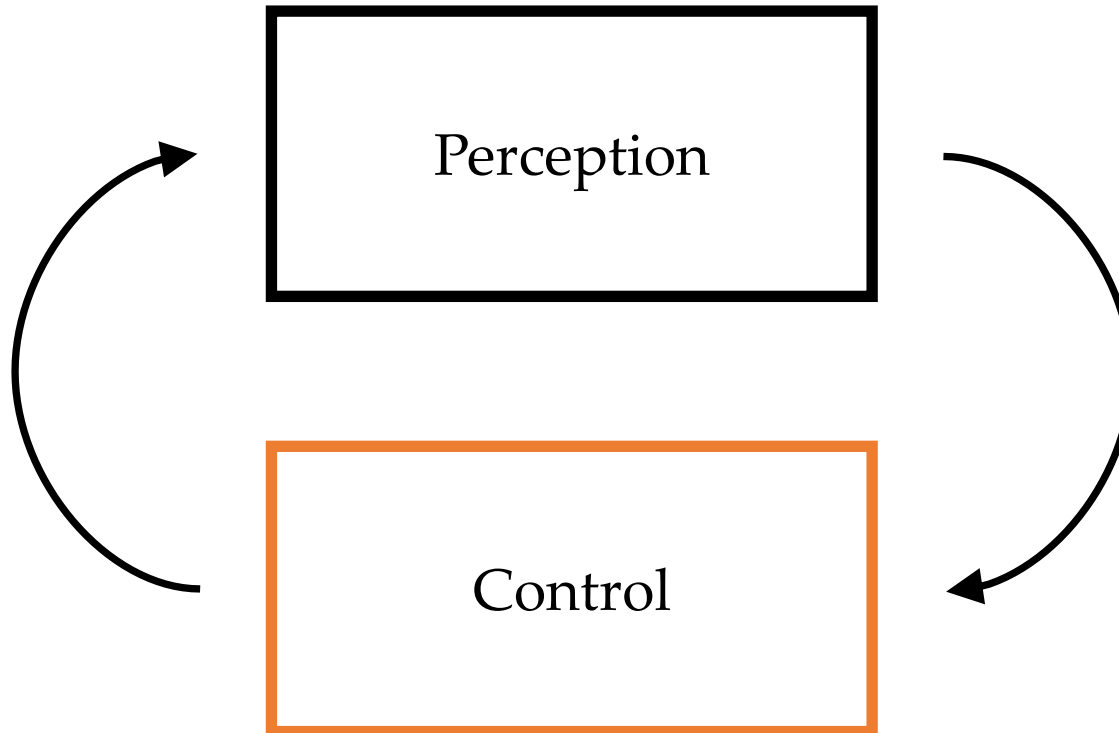
**Particle Filtering**   **Mapping**  
**Localization**



**Inverse Kinematics**

**Bayesian Filtering**  
KF / EKF

**Particle Filtering**   **Mapping**  
**Localization**



**Inverse Kinematics**  
**Task Space Regions**

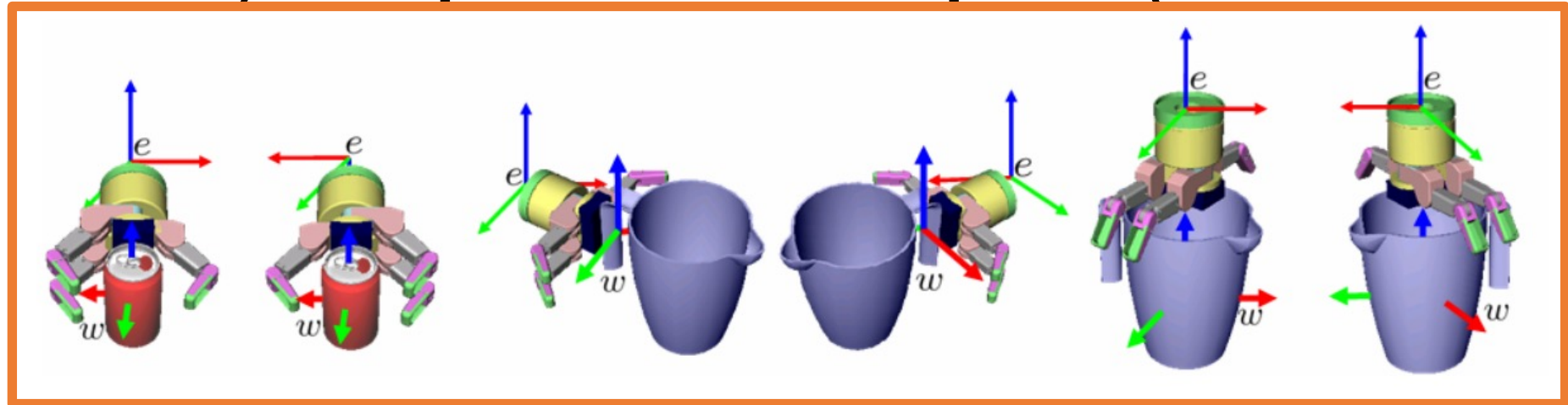
Bayesian Filtering

KF / EKF

Particle Filtering Mapping

Localization

Perception



Inverse Kinematics

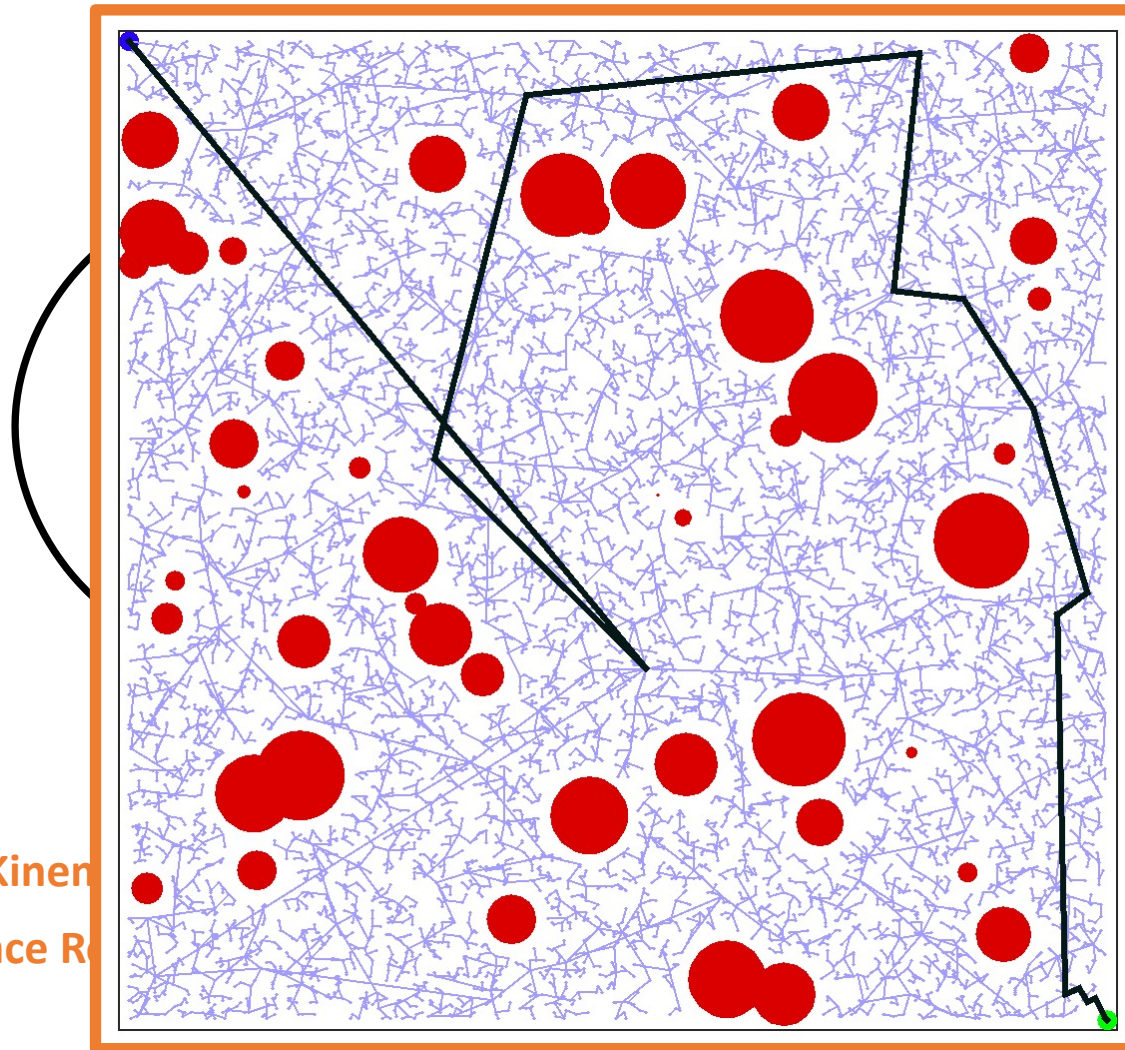
Task Space Regions

Bayesian Filtering

Particle Filtering

Mapping

KF / EKF



Inverse Kinematics

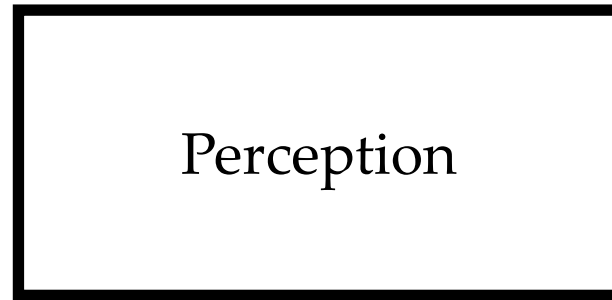
Task Space Replanning

**Bayesian Filtering**

**KF / EKF**

**Particle Filtering    Mapping**

**Localization**

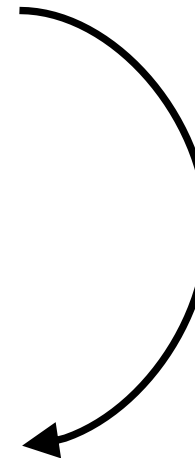
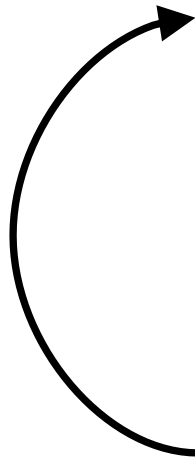


**Inverse Kinematics**

**Task Space Regions**

**Motion Planning**

**Dynamics**

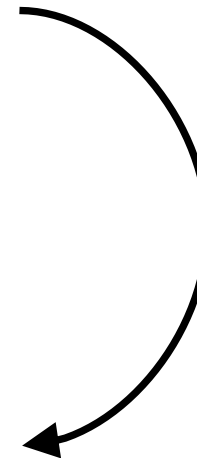
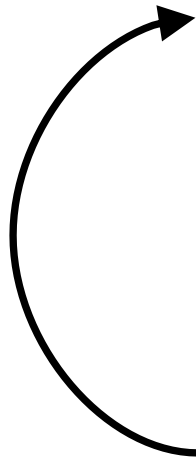
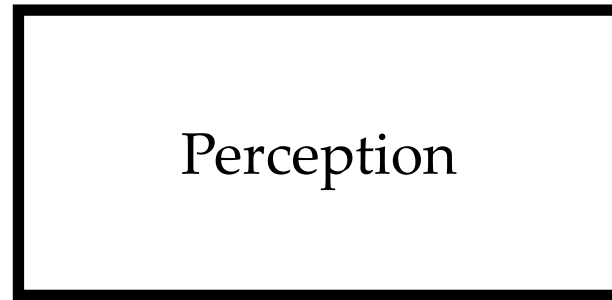


**Bayesian Filtering**

**KF / EKF**

**Particle Filtering   Mapping**

**Localization**



**Inverse Kinematics**

**Task Space Regions**

**Motion Planning**

**Dynamics**

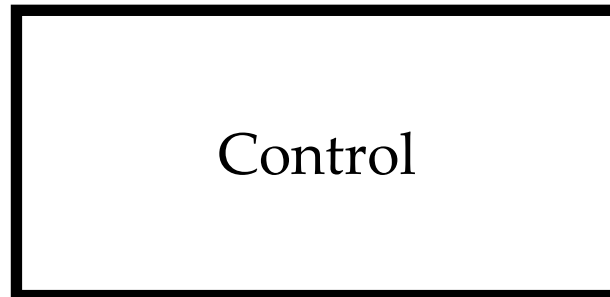
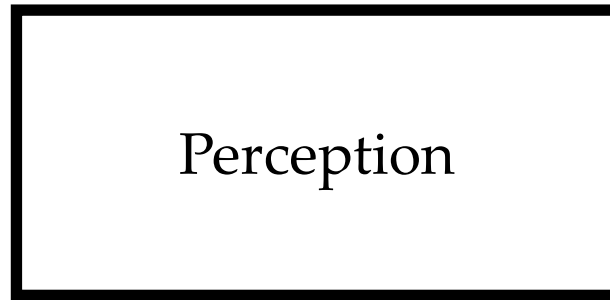
**Linear / Non-Linear Control**

**Bayesian Filtering**

**KF / EKF**

**Particle Filtering    Mapping**

**Localization**



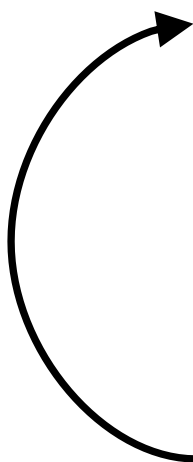
**Inverse Kinematics**

**Task Space Regions**

**Motion Planning**

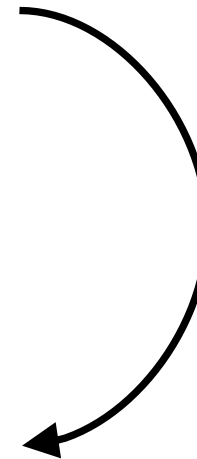
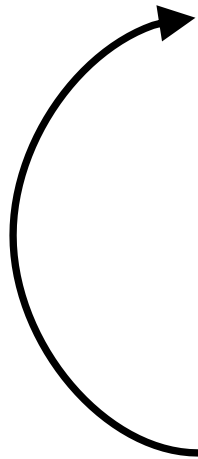
**Dynamics**

**Linear / Non-Linear Control**



**Bayesian Filtering**  
KF / EKF

**Particle Filtering**   **Mapping**  
**Localization**



**Decision Theory**  
**Planning Under**  
**Uncertainty**

**Inverse Kinematics**  
**Task Space Regions**

**Motion Planning**   **Linear / Non-Linear Control**  
**Dynamics**

Bayesian Filtering

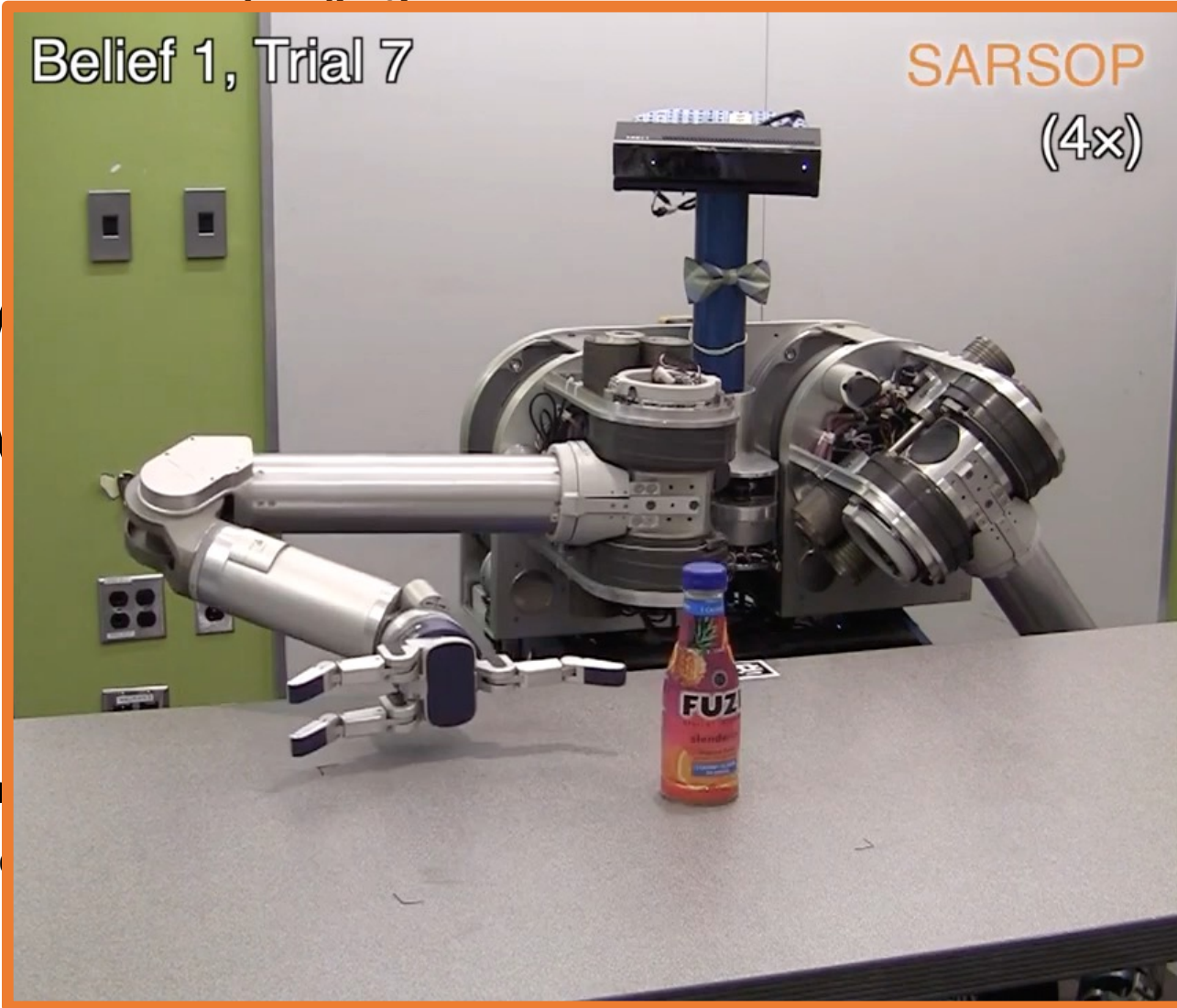
Particle Filtering

Mapping

KF / EKF

Belief 1, Trial 7

SARSOP  
(4x)



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Inverse Kin

Task Space

# Non-linear Control

CSCI 545 Introduction to Robotics  
Instructor: Stefanos Nikolaidis

Resources: CSCI 545 Lecture Notes by Prof. Stefan Schaal

# Dynamics Equations

$$B(q)\ddot{q} + C(q, \dot{q})\dot{q} + G(q) = \tau$$

These equations are non-linear.

# Non-linear Control

Joint-Space Control: Create motor commands in joint space

Operational Space Control: Create motor commands in end-effector space

# Non-linear Control

**Joint-Space Control: Create motor commands in joint space**

Operational Space Control: Create motor commands in end-effector space

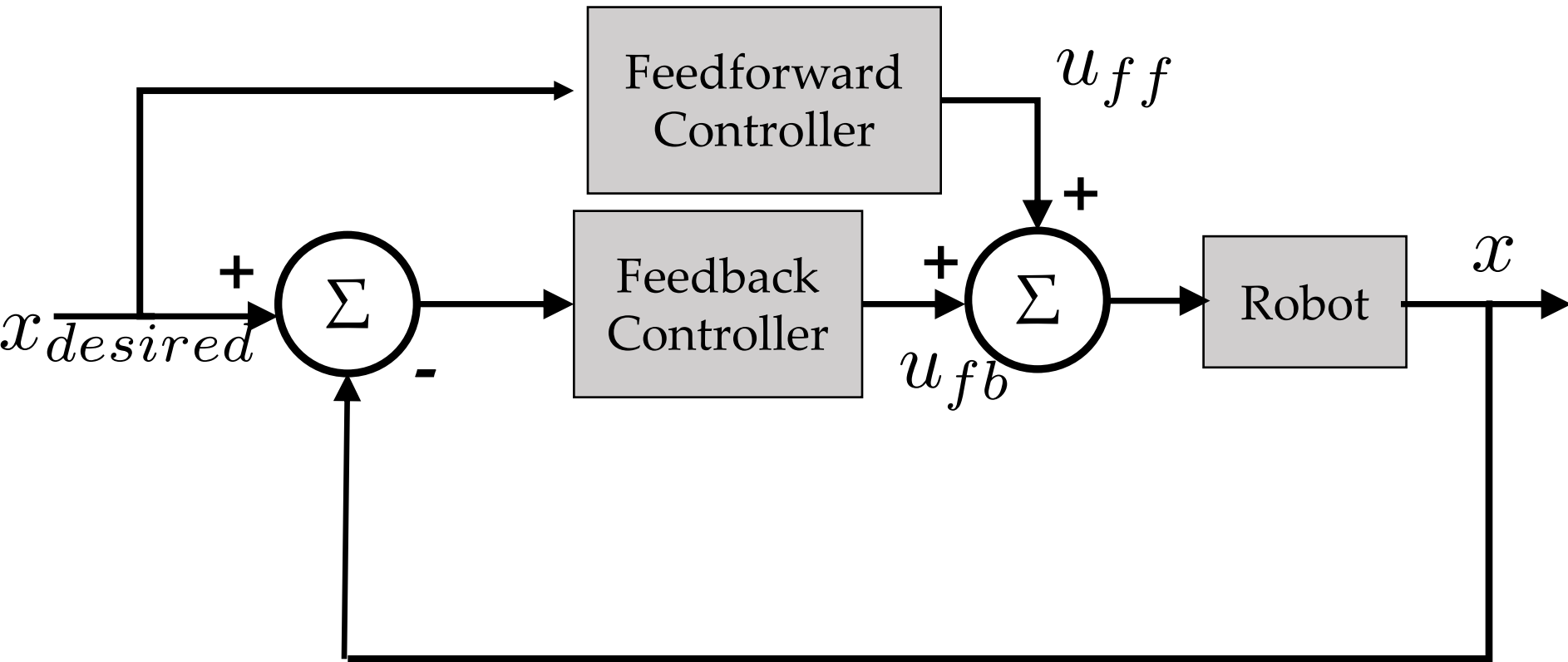
# Joint-Space Control

- Independent (Decentralized) Joint-Space Control
  - Appropriate when coupling terms are negligible, e.g., robot is decoupled (two 1DOF robots)
  - I focus on the diagonal elements of  $B, C$  and I treat the rest as disturbances
  - I create a PD / PID controller for each joint-space independently

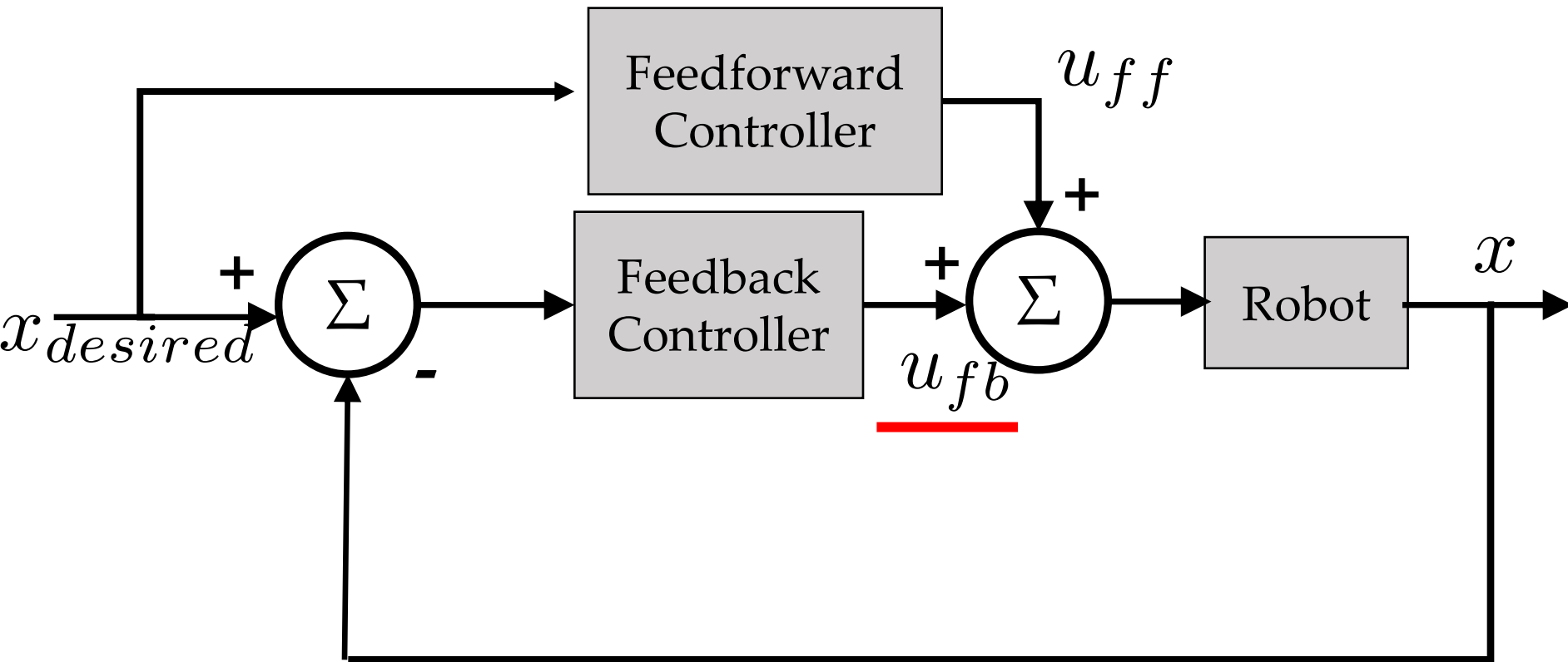
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  - I create a PD / PID controller for each joint-space independently
- Dependent Joint-Space Control
  - I cannot ignore the coupling terms

# Negative Feedback & Feedforward Control



# Negative Feedback & Feedforward Control



# Negative Feedback Control

- Based on linear control:

- Proportional Control (“Position Error”)

$$u_P = \pi(x - x_{des}, \alpha, t) = K_P(x_{des}(t) - x(t))$$

- Derivative Control (“Damping”)

$$u_D = \pi(x - x_{des}, \alpha, t) = K_D(\dot{x}_{des}(t) - \dot{x}(t))$$

- Integral Control (“Steady State Error”)

$$u_I(t) = K_I \int_{\tau=0}^{\tau=t} (x_{des}(t) - x(t)) dt$$

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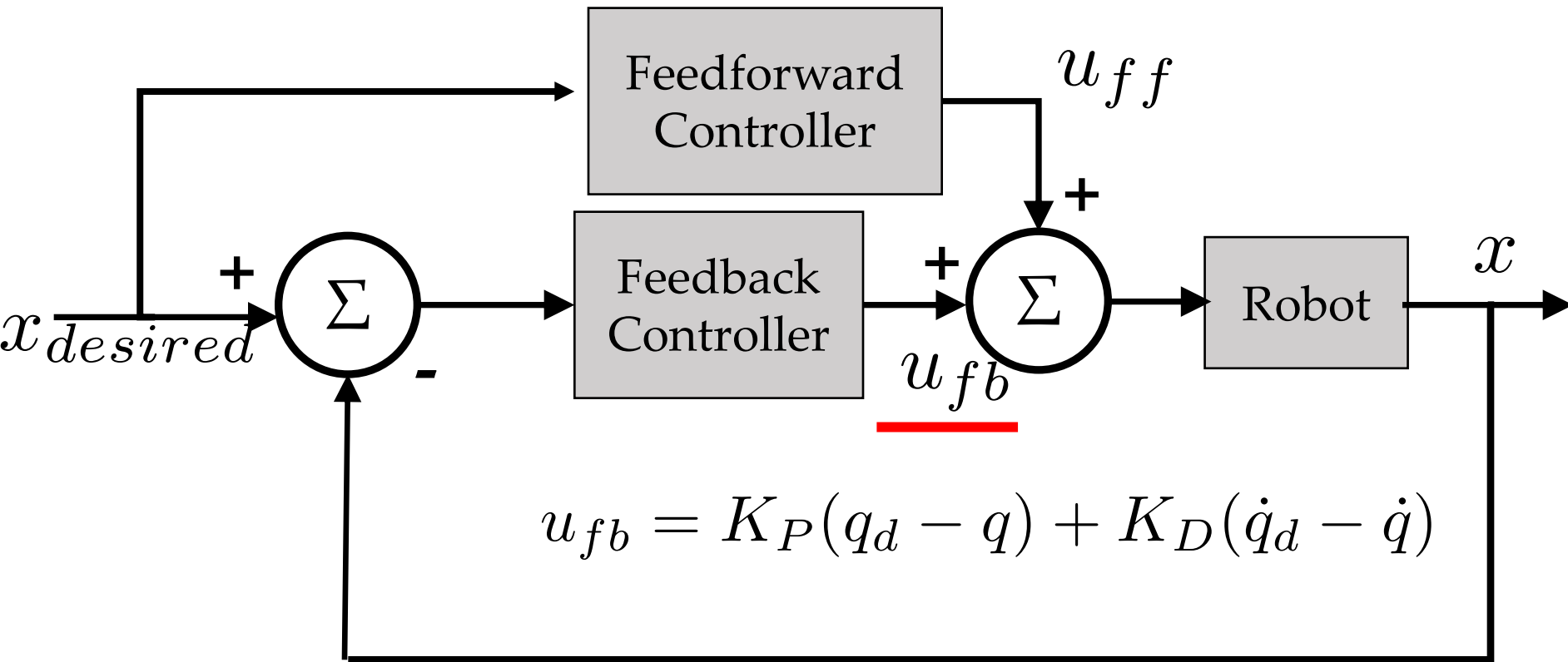
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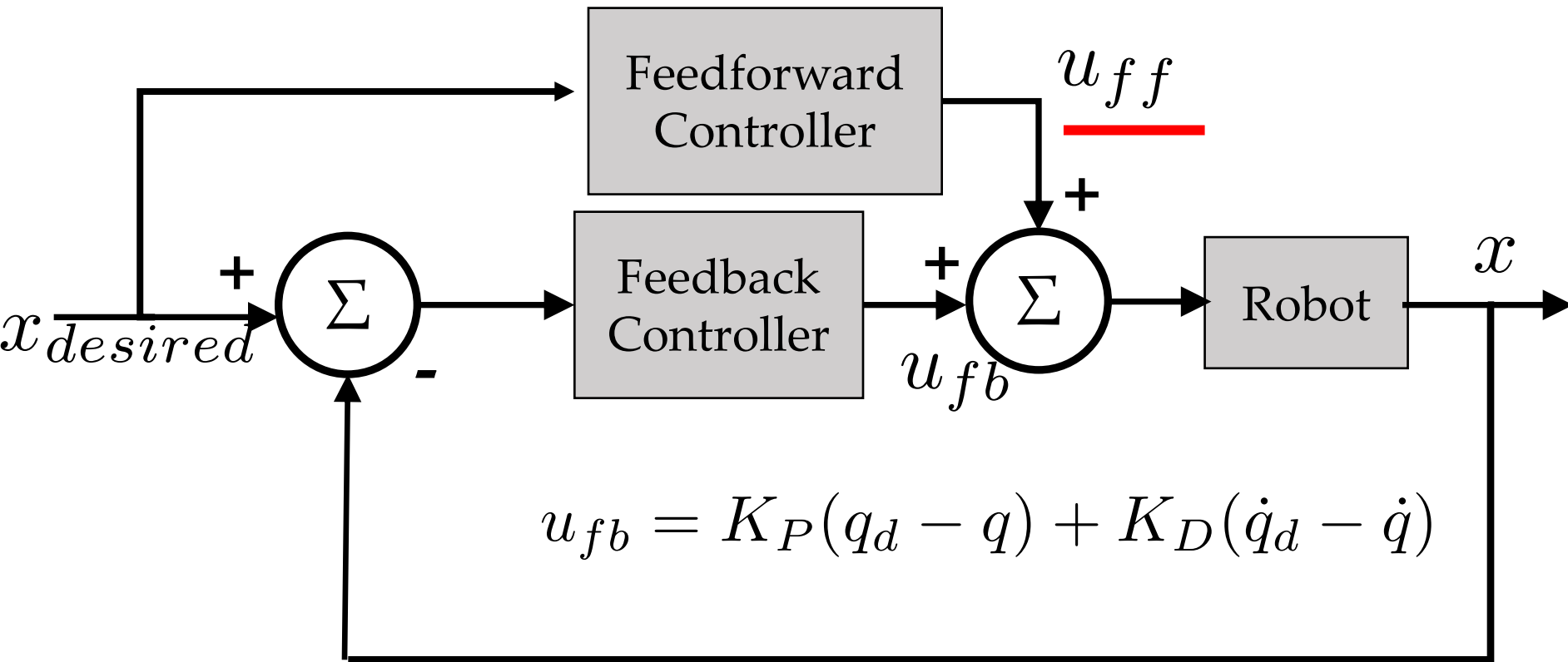
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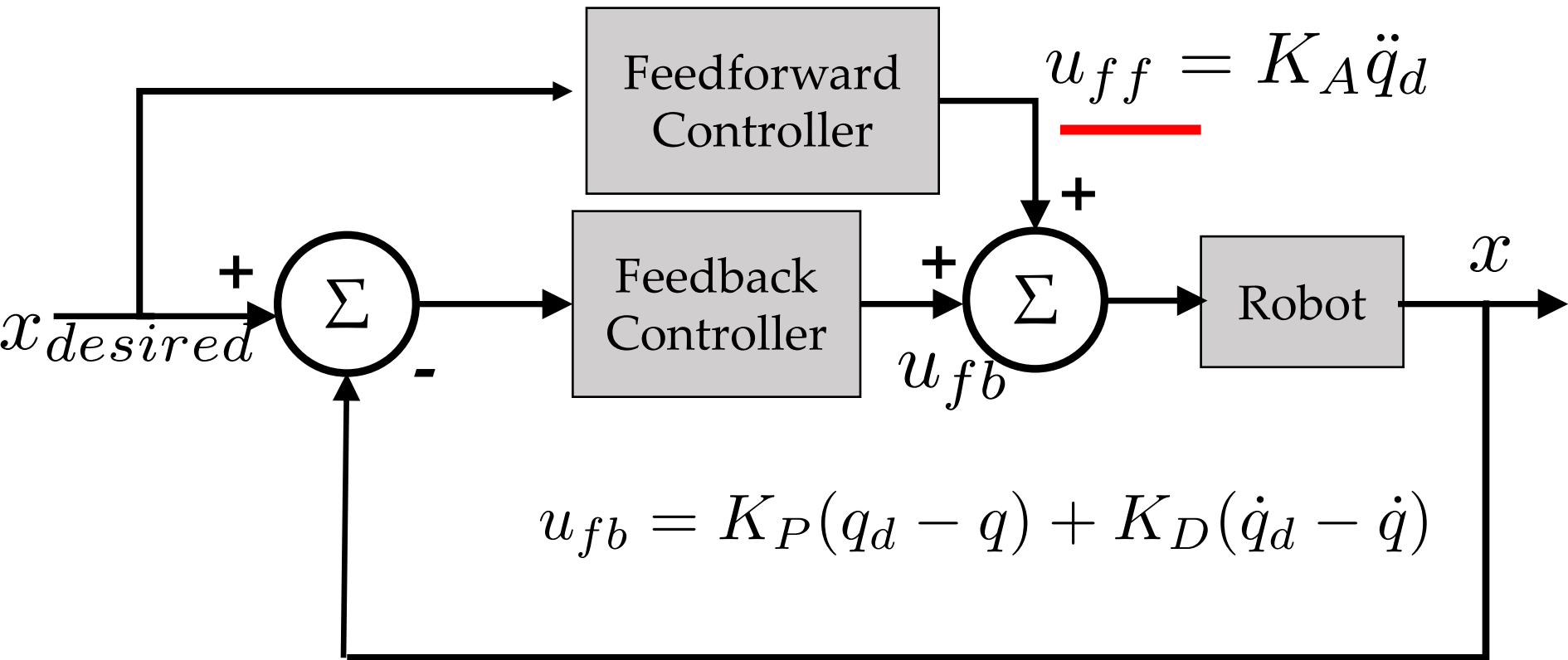
# Negative Feedback & Feedforward Control



# Negative Feedback & Feedforward Control



# Negative Feedback & Feedforward Control



# Independent Joint-Space Control

$$B(q)\ddot{q} + C(q, \dot{q})\dot{q} + G(q) = K_A\ddot{q}_d + K_P(q_d - q) + K_D(\dot{q}_d - \dot{q})$$

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$$B(q)\ddot{q} + C(q, \dot{q})\dot{q} + G(q) = K_A\ddot{q}_d + K_P(q_d - q) + K_D(\dot{q}_d - \dot{q})$$

We compensate for the coupling terms by treating them as noise

# Hybrid Joint-Space Control

- We can take into account the dynamics of the system

$$B(q)\ddot{q} + C(q, \dot{q}) + G(q) = \tau$$
$$B(q)\ddot{q} + C(q, \dot{q})\dot{q} + G(q) = B(q_d)\ddot{q}_d + C(q_d, \dot{q}_d)\dot{q}_d + G(q_d) + K_P(q_d - q) + K_D(\dot{q}_d - \dot{q})$$

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$$B(q) \approx B(q_d), C(q, \dot{q}) \approx C(q_d, \dot{q}_d)$$

$$B(q)\ddot{\tilde{q}} + (C(q, \dot{q}) + K_D)\dot{\tilde{q}} + K_P\tilde{q} = 0$$

# Hybrid Joint-Space Control

- Not accurate if the system deviates too much from the desired trajectory
- They can be computed offline

# Proper Inverse Dynamics Control

$$B(q)\ddot{q} + C(q, \dot{q}) + G(q) = \tau$$

$$B(q)\ddot{q} + C(q, \dot{q})\dot{q} + G(q) = B(q)[\ddot{q}_d + K_P(q_d - q) + K_D(\dot{q}_d - \dot{q})] + C(q, \dot{q})\dot{q} + G(q)$$

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$$\ddot{\tilde{q}} + K_D\dot{\tilde{q}} + K_P\tilde{q} = 0$$

- Requires online computation of the dynamics
- Requires perfect knowledge of the system

# A simpler example: PD with Gravity Compensation

$$B(q)\ddot{q} + C(q, \dot{q})\dot{q} + G(q) = \tau$$

$$B(q)\ddot{q} + C(q, \dot{q})\dot{q} + G(q) = G(q) + K_P(q_d - q) + K_D(\dot{q}_d - \dot{q})$$

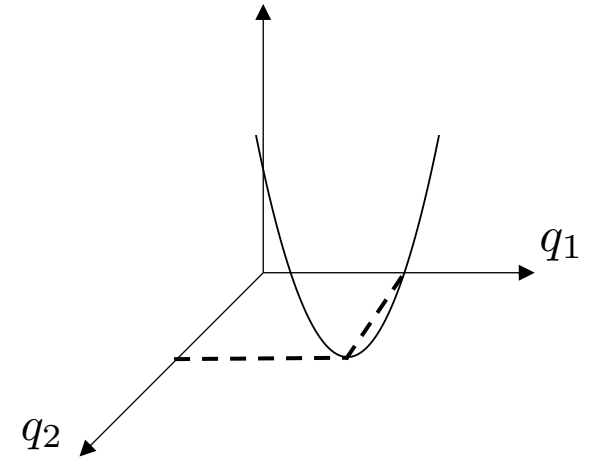
Assume  $\dot{q}_d = 0$

# Controller Stability

- We can use Lyapunov's method for proving the stability of the system
- It is a function of energy of the system.
- To prove stability, we need:
  - $V(x) = 0$  iff  $x = 0$
  - $V(x) > 0$  iff  $x \neq 0$
  - $\dot{V}(x) = \frac{d}{dt}V(x) = \sum_{i=1}^n \frac{\partial V}{\partial x_i} f_i(x) = \nabla V \cdot f(x) \leq 0, \forall x \neq 0$

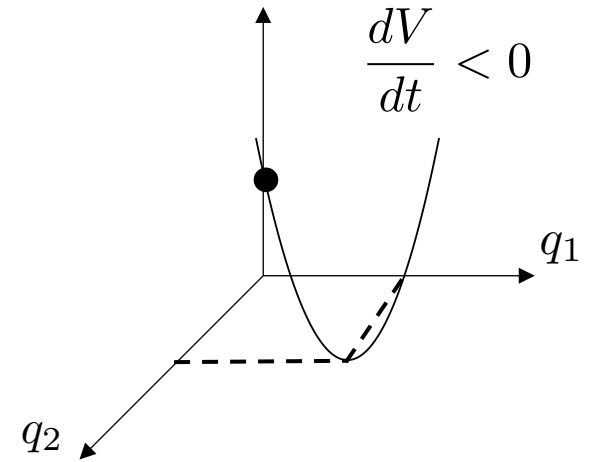
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# Stability of PD+Gravity Compensation Controller

- We consider the following Lyapunov function:

$$V(\dot{q}, \tilde{q}) = \frac{1}{2} \dot{q}^T B(q) \dot{q} + \frac{1}{2} \tilde{q}^T K_P \tilde{q} > 0, \forall \dot{q}, \tilde{q} \neq 0, \tilde{q} = q_d - q$$

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$$\dot{V} = \dot{q}^T B(q) \ddot{q} + \frac{1}{2} \dot{q}^T \dot{B}(q) \dot{q} - \dot{q}^T K_p \tilde{q}$$

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$(\dot{\tilde{q}} = -\dot{q})$

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$$\dot{q}^T (\dot{B} - 2C)\dot{q} = 0$$

(see SS, “7.2 Notable Properties of Dynamic Model”)

# Stability of PD+Gravity Compensation Controller

$$\dot{V} = \dot{q}^T \tau - \dot{q}^T C(q, \dot{q})\dot{q} - \dot{q}^T G(q) + \frac{1}{2} \dot{q}^T \dot{B}(q)\dot{q} - \dot{q}^T K_p \tilde{q}$$

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$\dot{B} - 2C = 0$  (see SS, “7.2 Notable Properties of Dynamic Model”)

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$$\dot{V} = \dot{q}^T \tau - \dot{q}^T C(q, \dot{q}) \dot{q} - \dot{q}^T G(q) + \frac{1}{2} \dot{q}^T \dot{B}(q) \dot{q} - \dot{q}^T K_p \tilde{q}$$

$$\dot{B} - 2C = 0 \quad (\text{see SS, "7.2 Notable Properties of Dynamic Model"})$$

$$\dot{V} = \dot{q}^T \tau - \dot{q}^T G(q) - \dot{q}^T K_p \tilde{q}$$

When is  $\dot{V} < 0$ ?

# Stability of PD+Gravity Compensation Controller

$$\dot{V} = \dot{q}^T \tau - \dot{q}^T C(q, \dot{q}) \dot{q} - \dot{q}^T G(q) + \frac{1}{2} \dot{q}^T \dot{B}(q) \dot{q} - \dot{q}^T K_p \tilde{q}$$

$$\dot{B} - 2C = 0 \quad (\text{see SS, "7.2 Notable Properties of Dynamic Model"})$$

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We can let  $\tau = G(q) + K_p \tilde{q} - K_d \dot{q}$

# Stability of PD+Gravity Compensation Controller

$$\dot{V} = \dot{q}^T \tau - \dot{q}^T C(q, \dot{q}) \dot{q} - \dot{q}^T G(q) + \frac{1}{2} \dot{q}^T \dot{B}(q) \dot{q} - \dot{q}^T K_p \tilde{q}$$

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When is  $\dot{V} < 0$ ?

We can let  $\tau = \underline{G(q) + K_p \tilde{q} - K_d \dot{q}}$

Feedback Linearization

# A simpler example: PD with Gravity Compensation

$$B(q)\ddot{q} + C(q, \dot{q})\dot{q} + G(q) = \tau$$

$$B(q)\ddot{q} + C(q, \dot{q})\dot{q} + G(q) = G(q) + K_P(q_d - q) + K_D(\dot{q}_d - \dot{q})$$

Assume  $\dot{q}_d = 0$