

# ME 599/699 Robot Modeling & Control

## Feedback Control

Hasan Poonawala

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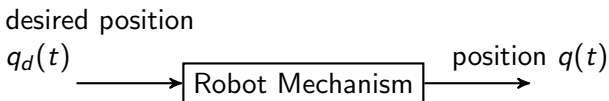
The task of **motion control** is to achieve that change

# Ideal Behavior

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Ideally,  $q_d(t) = q(t)$  at all times



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If we can apply large enough forces, we may assume that we can **practically** instantaneously set velocities. Eg: low inertia wheels.

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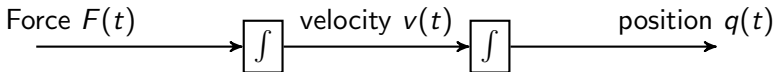
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In other words,  $q(t) = \int_0^t \int_0^t F(t) dv dq$ , where  $v(t) = \dot{q}(t)$ .



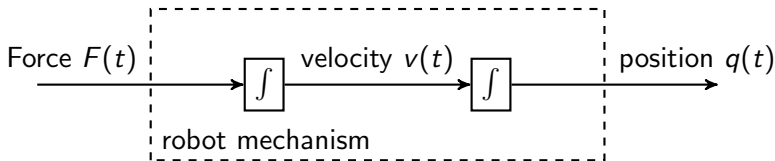
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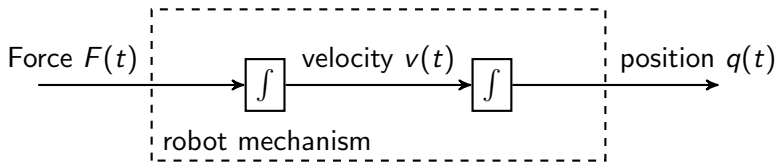
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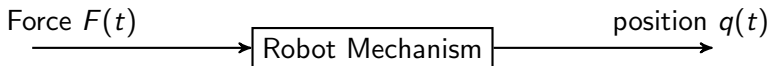
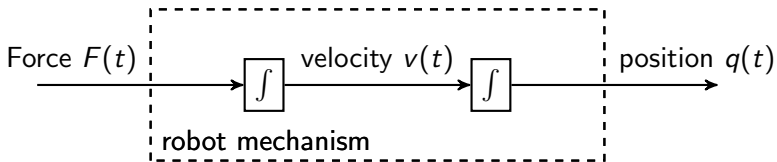
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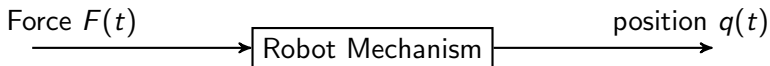
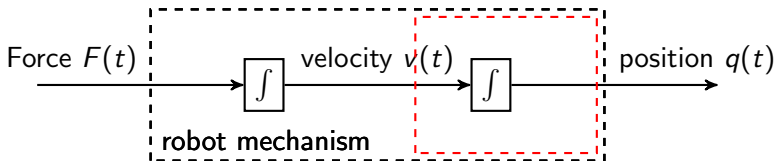
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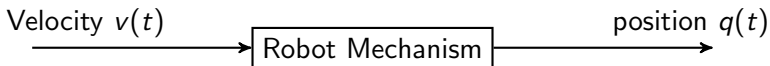
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Low inertia



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One approach: Ensure that  $q(0) = q_d(0)$ ,  $\dot{q}(0) = \dot{q}_d(0)$  and make  $F(t) = \ddot{q}_d(t)$  (model inversion).

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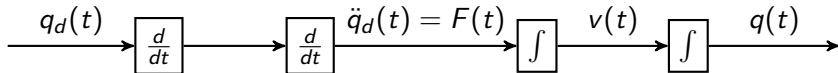
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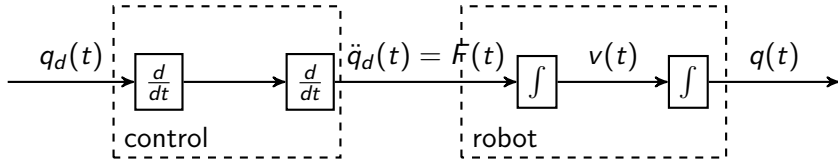
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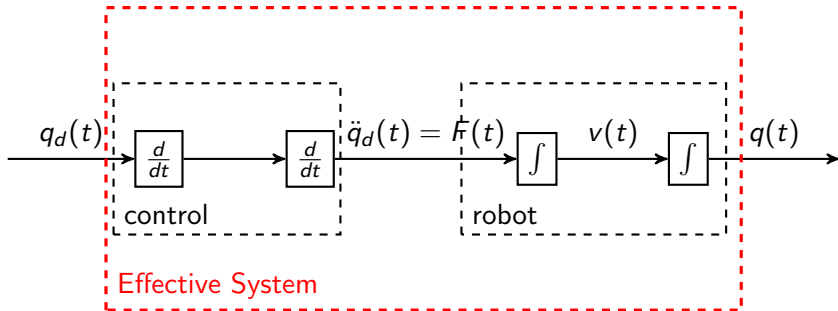
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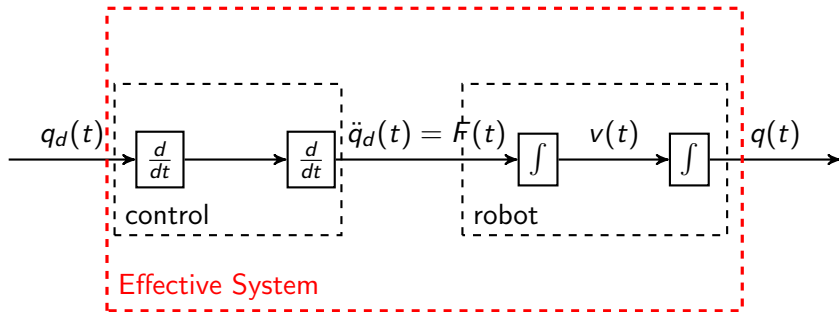
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# Model Inversion: Open-Loop

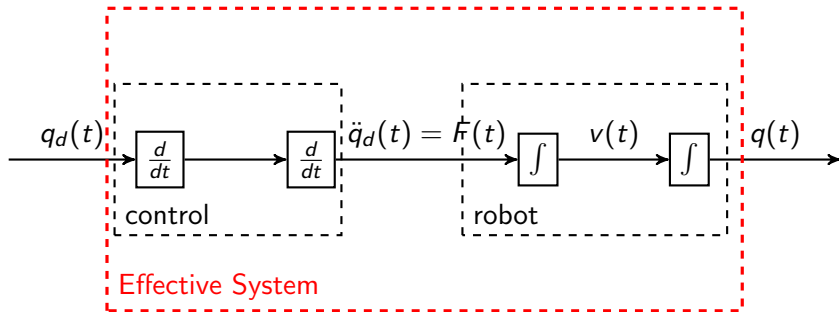


We've managed to make our robot react to desired position instantly, by

- ▶ Setting the initial condition
- ▶ Knowing the 'input'  $q_d(t)$  perfectly



# Model Inversion: Open-Loop



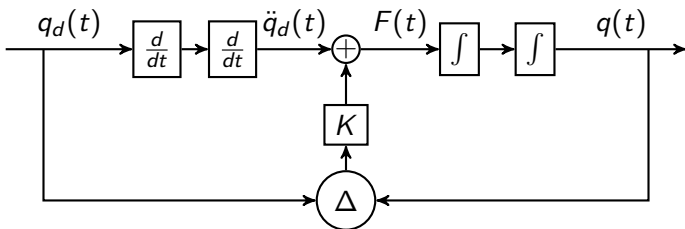
Issues:

- ▶ What if desired position is not known ahead of time?
- ▶ What if a disturbance force  $f_d(t)$  acts, so that input is  $F(t) + f_d(t) \neq \ddot{q}_d(t)$ ?

# Closed-Loop Control

One way to account for these issues is to also react to errors in position:

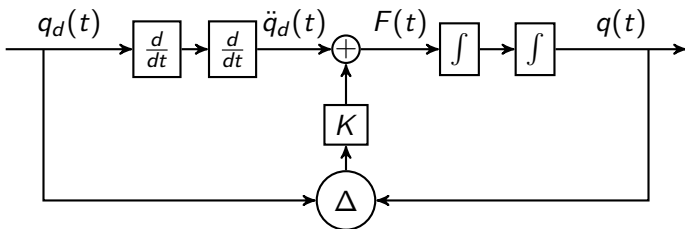
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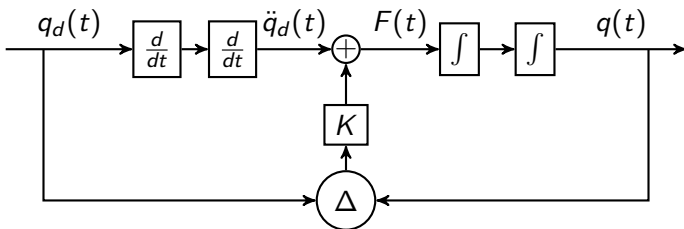


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Issue: How do we choose  $K$ ? Why will a good choice be possible?  
Another issue: How do you know what  $q(t)$  is?

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Configuration is  $q(t)$ , input is a force  $F(t)$

We then related force to change in  $q(t)$  using an ordinary differential equation:

$$\frac{d^2}{dt^2}q(t) = F(t).$$

This ODE is a forward dynamics model: how the **state** and input affect the change in state.

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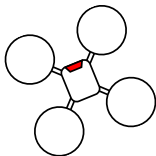
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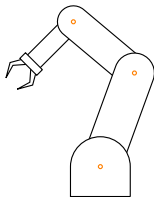
$$\dot{x} = \begin{bmatrix} 0 \\ F(t) \end{bmatrix} = f(x(t), F(t)).$$

# Dynamics

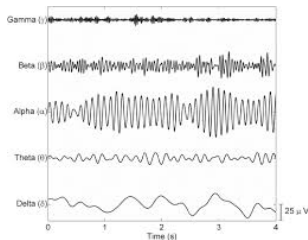
- The system has a state  $\mathbf{x}$
- The dynamics are captured by how  $\mathbf{x}$  changes with time



**Quadrotor**

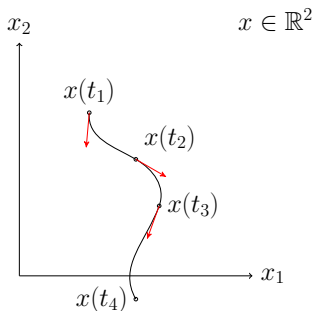


**Robot Arm**

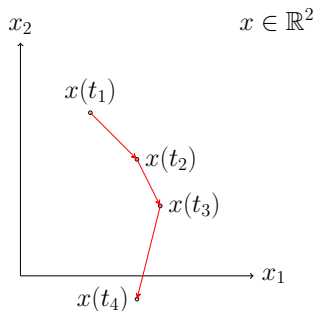


**EEG**

# Dynamics



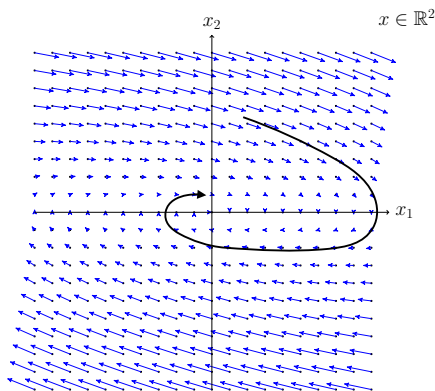
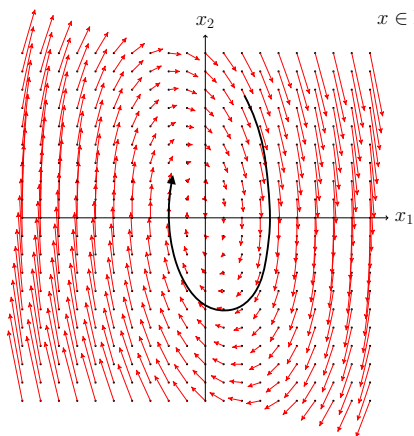
**Continuous-time**



**Discrete-time**

**Think of vectors assigned to each point  $x$   
(A vector field)**

# Dynamics



Two continuous vector fields

# Forward Dynamics

Continuous Time:

$$\dot{x}(t) = f(x(t), u(t), t),$$

where the independent variable is time  $t \in \mathbb{R}$

Discrete Time:

$$x_{t+1} = f_t(x_t, u_t),$$

where the independent variable is  $t \in \mathbb{N}$

# Inverse Dynamics

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For example, for the point mass, we said that we want  $\ddot{q}(t)$  to equal  $\ddot{q}_d(t)$ , so we choose

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