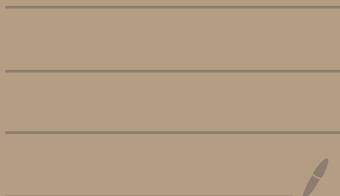


Intro to Robotics

Part(3)

(Joint Motion Control)



Joint Motion Control

Q: What is the goal of robot control ?

- find the time history torques to be sent to actuators in order to execute a task.
 - Regulation (constant reference)
 - Tracking (time-varying reference)

Challenges:

- External Disturbance
- Unmodeled dynamics & uncertain robot parameters
- Discretization

Open-loop Control

- The output of trajectory generator $\theta^d, \dot{\theta}^d, \ddot{\theta}^d$ is passed to the robot dynamic model:

$$\tau = \underbrace{M(\theta^d) \ddot{\theta}^d}_{\substack{\text{Inertia} \\ \text{Matrix}}} + \underbrace{C(\theta^d, \dot{\theta}^d)}_{\substack{\text{Coriolis} \\ \text{& Centrifugal} \\ \text{forces}}} + \underbrace{G(\theta^d)}_{\substack{\text{Gravitational} \\ \text{Forces}}}$$

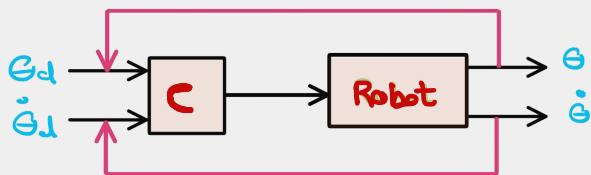
Problems

- Disturbances
- Model inaccuracies



Tracking Errors

Feedback Control



Inputs:

- Position/velocity of the joints : $\Theta(t)$, $\dot{\Theta}(t)$ [Measurements]
- Reference trajectory: $\Theta_d(t)$, $\dot{\Theta}_d(t)$

Control can be :

- Decentralized control (mostly used) [Independent Joint Control]
- Centralized control (accounts for dynamic interaction between joints)

Controller Specifications

Frequency Domain

- closed loop BW

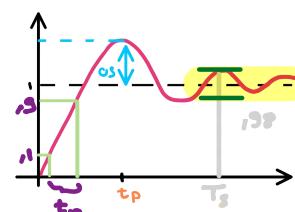
[Highest frequency sinusoid it can track]

Time Domain

- rise time (T_r)

- settling time (T_s)

- Overshoot (os)%



Design
spec.

Standard
Form

Poles

Gains (k)

MATLAB

T_r
 T_s
Os

ζ, ω_n

P_1, P_2

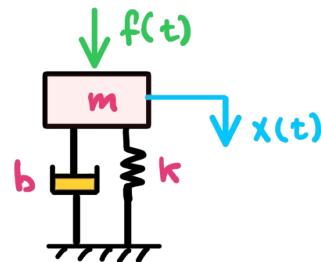
place
acker

Review on Spring-Mass-Damper

$$m \ddot{x} = f(t) - kx - bx'$$

$$m \ddot{x} + bx + kx = f(t)$$

(2nd order linear system)



I) Study of transient (natural dynamics) : $f(t) = 0$

$$x(t) = C_1 e^{\bar{\omega}_1 t} + C_2 e^{\bar{\omega}_2 t}$$

Where $\bar{\omega}_1, \bar{\omega}_2$ are the roots of the characteristics equation:

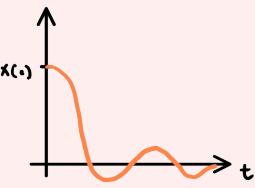
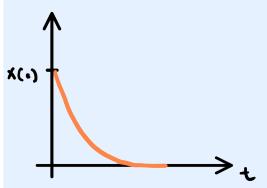
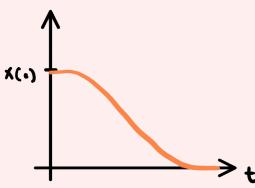
$$m \bar{\omega}^2 + b \bar{\omega} + k = 0$$

$$\bar{\omega}_{1,2} = \frac{-b \pm \sqrt{b^2 - 4ac}}{2m}$$

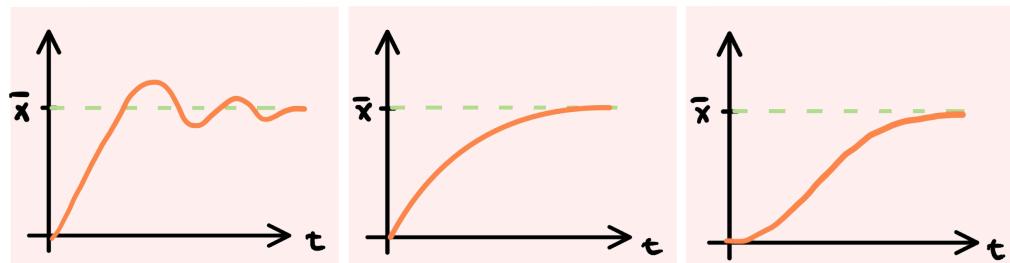
System is always **stable** if the poles have **negative real part**

- Let's take a look at different cases for system response.

I) Study of transient (natural dynamics) : $f(t) = 0$

Case	When?	Solution	Response
Under Damped	$D - 4mk < 0$ $D < 2\sqrt{mk}$	Complex conjugate pair of roots	
Critically Damped	$D - 4mk = 0$ $D = 2\sqrt{mk}$	2 real roots $\lambda_1 = \lambda_2$	
Over Damped	$D - 4mk > 0$ $D > 2\sqrt{mk}$	2 real roots $\lambda_1 \neq \lambda_2$	

II) case when : $f(t) \neq 0 = cost$, $x(0) = 0$



Under Damped

Critically Damped

Over Damped

Notes on Under-damped 2nd Order Systems

$$\ddot{x} + \frac{b}{m} \dot{x} + \frac{k}{m} x = 0 \quad \rightarrow \quad \ddot{x} + 2\{\omega_n \dot{x} + \omega_n^2 x = 0$$

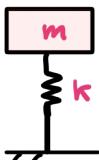
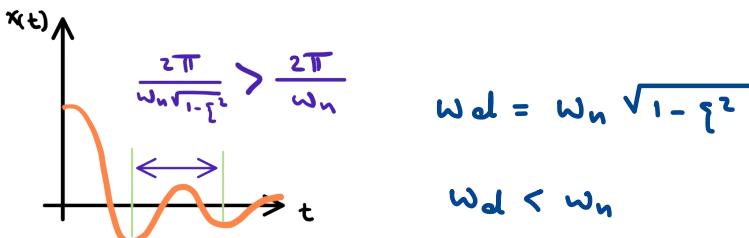
$$\omega_n = \sqrt{\frac{k}{m}}$$

[Natural Frequency]

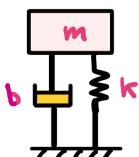
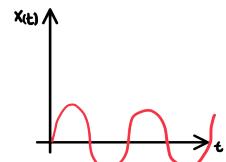
$$\frac{b}{m} = 2\{\omega_n \quad \rightarrow \quad \{\ = \frac{b}{2\cdot\omega_n\cdot m} \quad [\text{Natural Damping Frequency}]$$

When $\{\ = 1$, $b = 2\sqrt{k\cdot m}$ [Critically Damped]

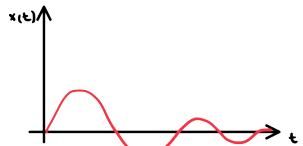
Time Response for $\{\ < 1$ [Under-Damped]



The system oscillates bouncing energy between mass(kinetic) and spring(potential)



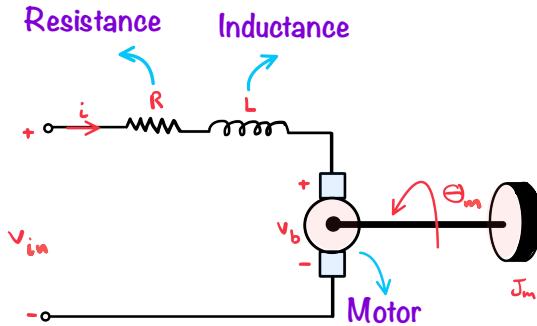
Unless there is damping element



Notes:

- From dynamics we know that **robot** is a **2nd order nonlinear system**.
- we can apply **linear control** to the robot if:
 - Gear ratio (N) > 100
 - Robot operate at low speeds
- Under these assumptions we can
 - decouple the dynamics of each joint
 - Consider coupling as disturbance
- To see why this works, we will take a look at **motor dynamics**

Motor Modelling



Electrical Domain :

Lenz-Law: $v_b = k_b \dot{\theta}_m$ "back emf" (as motor spins it creates) some voltage

Loop-Equation: $v_{in} = i \cdot R + \frac{di}{dt} \cdot L + v_b$

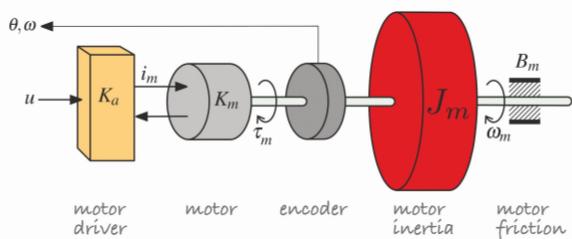
Lorenz-Law: $\tau_m = k_t \cdot i$

Power in Elec.domain: $P_m = i \cdot v_b = i (k_b \dot{\theta}_m)$

Mechanical Domain :

$$\tau_m = J_m \ddot{\theta}_m + b_m \dot{\theta}_m$$

$$P_m = \tau_m \dot{\theta}_m$$
 "mechanical Power"



Motor Modelling

- from mechanical power and electrical power we notice that:

$$P_m = \underbrace{\tau_m \cdot \ddot{\theta}_m}_{\text{Mechanical}} = k_t \cdot i \cdot \ddot{\theta}_m = k_b \cdot \dot{\theta}_m \cdot i = \underbrace{k_b \cdot \dot{\theta}_m \cdot i}_{\text{Electrical}}$$

$\therefore k_t = k_b$

Coupling between Mechanical and Electrical Domains :

Motor torque: $\tau_m = k_t \cdot i = J_m \ddot{\theta}_m + b \dot{\theta}$

Input Voltage: $v_{in} = L_a \frac{di}{dt} + R \cdot i + k_b \cdot \dot{\theta}_m$

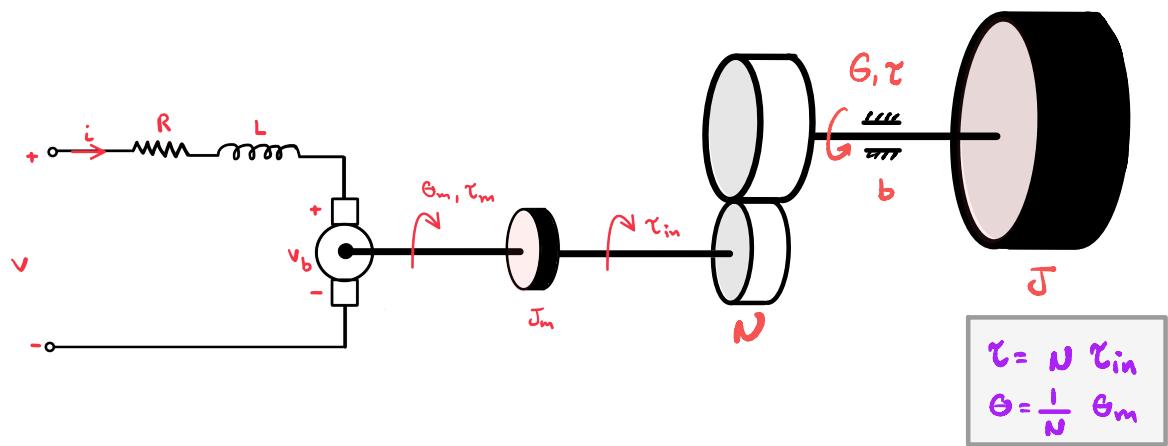
≈ 0
as L_a is so small

$$v_{in} = \frac{R}{k_t} J_m \ddot{\theta}_m + \left(k_b + \frac{R \cdot b_m}{k_t} \right) \dot{\theta}_m$$

Laplace Transform:

$$T.F = \frac{\Theta_m(s)}{V_{in}(s)} = \frac{k_t / R}{s[J_m \cdot s + (b_m + \frac{k_b \cdot k_t}{R})]}$$

Dc Motor Model with Gearbox:



- Torque at output of gearbox:

$$\tau = N \cdot \tau_{in} = N [\tau_m - b_m \dot{\theta}_m - J_m \ddot{\theta}_m] = b \dot{\theta} + J \ddot{\theta}$$

$$N \tau_m = (b + N^2 b_m) + \underbrace{(J + N^2 J_m)}_{\text{inertia at o/p}} \ddot{\theta}$$

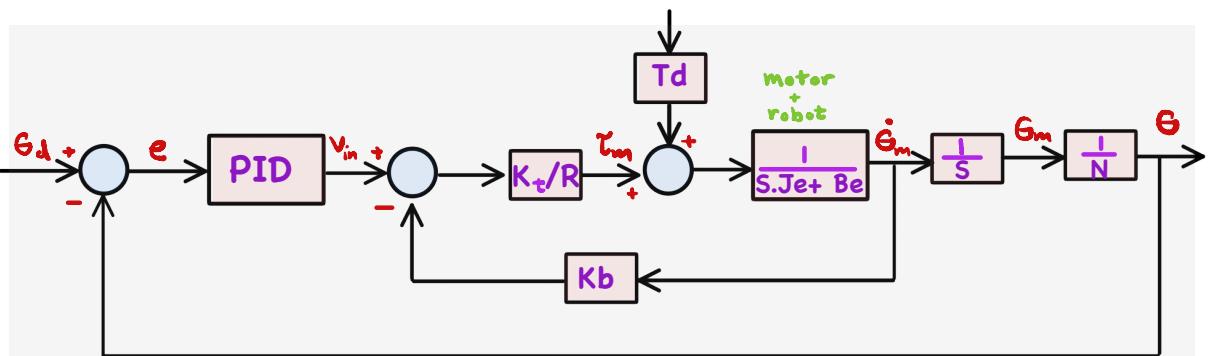
- In terms of motor angles:

$$\tau_m = \underbrace{(b_m + \frac{b}{N^2})}_{J_{eff}} \dot{\theta}_m + \underbrace{(J_m + \frac{J}{N^2})}_{B_{eff}} \ddot{\theta}_m$$

Notes:

- In multi-joint robot J is **not fixed**.
- When N is **large**, motor dynamics dominates. "That's why Linear control works"
- Industrial robots $N > 100$.
- Design **critically damped** for max J .

Block Diagram



T_d : disturbance torque due to

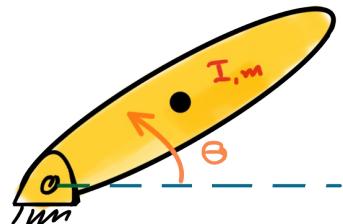
- Inertial coupling
- Centrifugal & Coriolis effect
- Gravitational effects

Regulation Problem

Approach(1): PD Control (linear)

Consider:

- 1-link moving in a plane (\perp to gravity)



Dynamics: $\tau = I \ddot{\theta}$ [No Gravity]

PD Idea: Create virtual spring/damper to derive the link towards a desired position θ_d

Let: $\tau = k_p(\theta_d - \theta) + k_d(\dot{\theta}_d - \dot{\theta})$ (Torque Input)

Then: closed loop dynamics will be

$$I \ddot{\theta} + k_d \dot{\theta} + k_p \theta = k_p \theta_d + k_d \dot{\theta}_d$$

Q: What happens at steady-state?

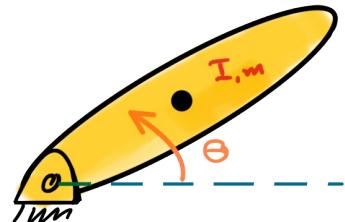
$$\cancel{I \ddot{\theta} + k_d \dot{\theta} + k_p \theta = k_p \theta_d + k_d \dot{\theta}_d}$$

At steady-state: $\dot{\theta} = \ddot{\theta} = 0 \rightarrow \theta = \theta_d$

Regulation Problem

Approach(2): PD+Feedforward Control (linear)

Dynamics: $\tau = I \ddot{\theta}$



PD Idea: use our knowledge of the model dynamics to generate the required torque as **feedforward control** + the previous **PD feedback control**

Let: $\tau = I \ddot{\theta}_d + k_p(\theta_d - \theta) + k_d(\dot{\theta}_d - \dot{\theta})$

$I \ddot{\theta}_d$ $k_p(\theta_d - \theta)$ + $k_d(\dot{\theta}_d - \dot{\theta})$
 From Dynamics PD Control

Then: closed loop dynamics will be

$$I(\ddot{\theta}_d - \ddot{\theta}) + k_d(\dot{\theta}_d - \dot{\theta}) + k_p(\theta_d - \theta) = 0$$

$\ddot{\theta}$ $\dot{\theta}$ θ



$$I \ddot{e} + k_d \dot{e} + k_p e = 0 \quad \text{"Error Dynamics"}$$

Notes:

- Error dynamics converges to zero
- Transient response depends on (I, k_d, k_p)

Q: How to tune PD gains? 🤔

- Knowing **link inertia (I)** from CAD, we can achieve **tracking performance** by proper selection for k_p, k_d

Example:

Let: $I = 2,5$

Performance Specifications:

- Settling Time $t_s = 2 \text{ s}$ for reaching (98%)

$$\text{2nd order systems } 1 - e^{-\zeta \omega_n t} = 0,98$$

$$e^{-\zeta \omega_n t_s} = 0,02 \rightarrow \zeta \omega_n t_s = \log \frac{1}{0,02} = 4$$

$$t_s = \frac{4}{\zeta \omega_n}$$

- Critically damped system: $\zeta = 1$, $\omega_n = \frac{4}{t_s} = \frac{4}{2} = 2 \text{ rad/s}$

$$\omega_n = \sqrt{\frac{k_p}{I}} = 2 \rightarrow k_p = 4 \cdot I = 10$$

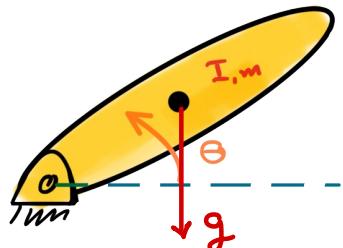
$$k_d = 2 \sqrt{k_p I} = 2 \sqrt{10 \cdot 2,5} = 10$$

$$k_p = k_d = 10$$

Regulation Problem

Approach(3): PID Control (linear)

Dynamics: $I \ddot{\theta} + f_{\text{dist}} + g = \tau$



PD Idea:

- Adding **integral action** can eliminate steady-state error against unknown constant disturbances.

Let: $\tau = k_p \cdot e + k_i \int e \, d\tau + k_d \cdot \dot{e}$

where, $e = \theta_d - \theta$, $\dot{e} = \dot{\theta}_d - \dot{\theta}$

Then: closed loop dynamics will be

$$I \ddot{\theta} + f_{\text{dist}} + g = k_p (\theta_d - \theta) + k_d (\dot{\theta}_d - \dot{\theta}) + k_i \int (\theta_d - \theta) \, d\tau$$

At steady-state: $\dot{\theta} = \ddot{\theta} = 0$

$$g + f_{\text{dist}} = k_p (\theta_d - \theta) + k_i \int (\theta_d - \theta) \, d\tau$$



$$\frac{d}{dt}$$

What happens when time evolves?

$$\theta = k_p \dot{e} + k_i e$$



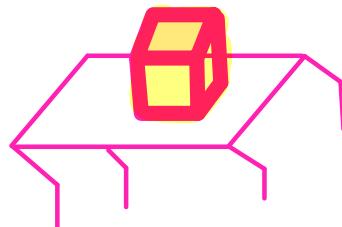
Error Converges to zero

Notes:

- The **I-Term** grows magnifying the **error** till the disturbance is compensated.
- Not good with **coulomb friction** -> **oscillations**

Applications:

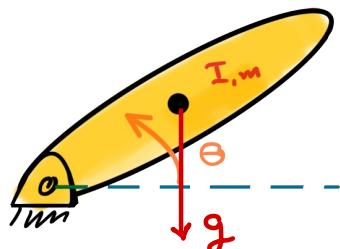
- Unknown payload
- Constant torque offset



Regulation Problem

Approach(4): PD+Gravity Compensation (non-linear)

Dynamics: $I \ddot{\theta} + g(\theta) = \tau$



Q: Why we didn't use PD control only?

- Let's take a look at **steady-state error** when we use PD only

$$I \ddot{\theta} + g(\theta) = k_p (\underbrace{\theta_d - \theta}_e) + k_d (\underbrace{\dot{\theta}_d - \dot{\theta}}_{\dot{e}})$$

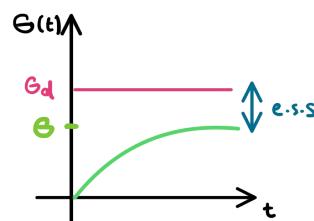
τ / Input

- At **steady-state**: $\dot{\theta} = \ddot{\theta} = 0$

$$k_p (\theta_d - \theta) = g(\theta)$$

$$\theta = \theta_d - \frac{g(\theta)}{k_p}$$

Steady State error



At equilibrium: $\theta_d \neq \theta$

Q: What if we compensate for gravity ? 🤔

Let: $\tau = k_p (\theta_d - \theta) + k_d (\dot{\theta}_d - \dot{\theta}) + g(\theta)$

old PD Control Compensate for Gravity

Then: closed loop dynamics will be

$$I \ddot{\theta} + g(\theta) = k_p (\theta_d - \theta) + k_d (\dot{\theta}_d - \dot{\theta}) + g(\theta)$$

Notes:

- **stability** ensured for all ($k_p, k_d > 0$).
- $g(\theta)$ is **configuration dependant** and we need to recompute it for each step.

Tracking Problem

Approach(1): Inverse Dynamics Control (non-linear)

Dynamics:

$$D(q) \ddot{q} + \underbrace{c(q, \dot{q}) \dot{q} + G(q)}_{=n(q, \dot{q})} + \text{friction} = \tau$$

Idea:

- Use our knowledge of robot dynamics to compute the required torque to follow trajectory of (q, \dot{q})

Let:

$$\tau = D(q_d) \ddot{q}_d + n(q_d, \dot{q}_d)$$

- Assume $q(0) = q_d$, $\dot{q}(0) = \dot{q}_d$ "Exact Initialization"

Then: closed loop dynamics will be

$$\ddot{\tilde{q}}_d = \ddot{q}$$

Problems:

- Initial state not matched to the desired trajectory q_d .
- External disturbance.
- Inaccurate model parameters.
- Unknown payload.
- Unmodeled dynamics.

Tracking Problem

Approach(2): Feedback Linearization + (PD+Feedforward)

Dynamics: $\ddot{\tau} = D(\dot{q}) \ddot{q} + n(q, \dot{q})$

Idea:

1. Use feedback for nonlinearity cancellation
2. Use linear control for stabilizing error to zero

Let:

$$\ddot{\tau} = D(\dot{q}) \left[\ddot{q}_d + k_p (q_d - q) + k_D (\dot{q}_d - \dot{q}) \right] + n(q, \dot{q})$$

Feedback
Linearization

Feedforward PD Linear Control

Then: closed loop dynamics will be

1. Feedback Linearization

$$D(\dot{q}) \ddot{q} + n(q, \dot{q}) = u = D(\dot{q}) a + n(q, \dot{q})$$

$$\ddot{q} = q$$

Nonlinear
Control

$$k_p = \begin{bmatrix} k_{p_1} & 0 \\ 0 & \dots \\ 0 & k_{p_n} \end{bmatrix}$$

$$k_D = \begin{bmatrix} k_{D_1} & 0 \\ 0 & \dots \\ 0 & k_{D_n} \end{bmatrix}$$

k_p, k_D : Positive Definite

2. Linear control

$$a = \ddot{q} + k_D (\dot{q}_d - \dot{q}) + k_p (q_d - q) \rightarrow \ddot{e} + k_D \dot{e} + k_p e = 0$$

Block Diagram

