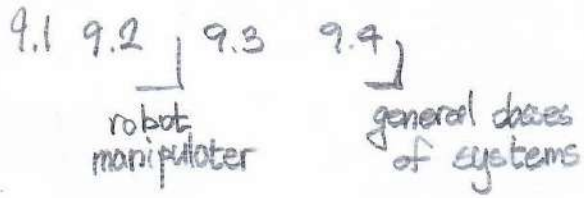


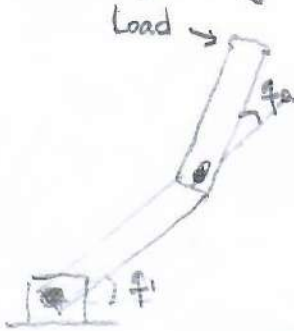
Ch 9. Control of Multi-Input Physical Systems.

- How to design a controller in a physical stance rather than mathematical.



9.1. Position & Trajectory Control.

9.2 Adaptive Trajectory Control.



$$H(q) \ddot{q} + C(q, \dot{q}) \dot{q} + g(q) = \tau, \quad q = [q_1, q_2]^T$$

where

- $H(q) \in \mathbb{R}^{2 \times 2}$: inertia matrix (sym & pos. def.)
- $C(q, \dot{q}) \dot{q}$: Centripetal & Coriolis torque
- $g(q)$: gravitational torque
- τ : control input to each joint (control law)

Q: How to construct τ to achieve $q \rightarrow q_d$.

- q_d : const. : position control
- $q_d = q_d(t)$: trajectory control.

• 2 Important Notes before we start.

① $E_k = \frac{1}{2} \dot{q}^T H(q) \dot{q}$ ($\approx \frac{1}{2} m v^2$)

② $H(q)$ is uniformly pos. definite i.e. $H(q) \geq \alpha I_2$ for some $\alpha > 0$

Pf) If not, $\exists \{q_n\} \subset \mathbb{R}^2$ s.t. $\lambda_{\min}(H(q_n)) \rightarrow 0$ as $n \rightarrow \infty$.

$\Rightarrow \exists \{q_n\} \subset \{q_n\}$ s.t. $q_n \rightarrow q_\infty$ (B-W thm)

$\Rightarrow H(q_\infty)$ has zero eigenvalue

$\Rightarrow \exists \dot{q} \neq 0$ s.t. $E_k = \frac{1}{2} \dot{q}^T H(q_\infty) \dot{q} = 0 \quad \nabla$

workspace is closed (compact)

9.1.1. Positional Control (q_d : const. vector)

We use a P.D controller for each joint independently.

i.e. $\tau_j = -k_{p_j} \tilde{q}_j - k_{d_j} \dot{\tilde{q}}_j$ where $\tilde{q}_j = q_j - q_{d_j}$ ($j = 1, 2, k_{p_j}, k_{d_j} > 0$)

$\Rightarrow \tau = -K_p \tilde{q} - K_D \dot{\tilde{q}}$

manip. on horizontal plane.

Transferring the problem to Hamiltonian form, assuming $g(q) = 0$,

$\frac{1}{2} \frac{d}{dt} (\dot{q}^T H \dot{q}) = \frac{1}{2} \ddot{q}^T H \dot{q} + \frac{1}{2} \dot{q}^T (\dot{H} \dot{q} + H \ddot{q})$

$\approx \frac{1}{2} \ddot{q}^T H \dot{q} + \frac{1}{2} \dot{q}^T H \ddot{q}$

$\dot{q}^T \dot{H} \dot{q} \approx 0$ often neglected in control context.

$= \dot{q}^T H \ddot{q}$

$= \dot{q}^T (\tau - C \dot{q} - g)$

$\dot{q}^T C \dot{q}$ neglected.

(related with \dot{H})

$= \dot{q}^T \tau$

$= \dot{q}^T (-K_p \tilde{q} - K_D \dot{\tilde{q}})$

pot. E ($\approx \frac{1}{2} k \tilde{q}^2$)

mech. Energy $V = \frac{1}{2} [\dot{q}^T H \dot{q} + \tilde{q}^T K_p \tilde{q}]$

$\Rightarrow \dot{V} = \dot{q}^T (-K_p \tilde{q} - K_D \dot{\tilde{q}}) + \frac{1}{2} \frac{d}{dt} (\tilde{q}^T K_p \tilde{q})$

$= \dot{q}^T (-K_p \tilde{q} - K_D \dot{\tilde{q}}) + \tilde{q}^T K_p \dot{\tilde{q}}$

$\dot{\tilde{q}} = \frac{d}{dt} (q - q_d) = \dot{q}$

$= -\dot{q}^T K_D \dot{\tilde{q}} \leq 0$

$\dot{q}^T K_p \tilde{q} = \tilde{q}^T K_p \dot{q}$

By Barbalat's lemma (\dot{V} is bounded $\Rightarrow V: UC$), $\dot{V} \rightarrow 0$ as $t \rightarrow \infty$.

We should also check system doesn't get stuck. (goal: $q \rightarrow q_d$)

i.e. $\dot{V} = 0$ only when $\tilde{q} = 0$

PF) $\dot{V} = 0 \Leftrightarrow \dot{q} = 0$

$\Rightarrow \ddot{q} = H^{-1} \tau = H^{-1} (-K_p \tilde{q} - K_D \dot{\tilde{q}})$

$= -H^{-1} K_p \tilde{q}$

$\Rightarrow \dot{q}$ is identically 0 only if $\tilde{q} = 0$.

9.1.2 Trajectory Control ($q_d = q_d(t)$)

• Feedback Linearization

$$\tau = H\ddot{v} + C\dot{q} + g \quad \text{where } v = \ddot{q}_d - 2\lambda\dot{\tilde{q}} - \lambda^2\tilde{q} \quad (\lambda > 0)$$

$\Rightarrow H\ddot{q} + C\dot{q} + g = \tau$ turns into

$$\ddot{q} = v = \ddot{q}_d - 2\lambda\dot{\tilde{q}} - \lambda^2\tilde{q}$$

$\Rightarrow \ddot{\tilde{q}} + 2\lambda\dot{\tilde{q}} + \lambda^2\tilde{q} = 0$. i.e. \tilde{q} exponentially converges to 0.

• Robust Control.

Aim to handle uncertainty & disturbances in the system.

We derive a robust traj. controller using sliding control. (chap 7)

sliding condition for multi-input case is

$$\frac{1}{2} \frac{d}{dt} (s^T s) \leq -\eta (s^T s)^{1/2} \quad (\eta > 0) \quad \text{where}$$

$$s = \overset{\text{velocity error}}{\dot{\tilde{q}}} + \overset{\text{sym. pos. def.}}{\lambda \tilde{q}} \\ = \dot{q} - \dot{q}_r \quad (\dot{q}_r = \dot{q}_d - \lambda \tilde{q})$$

Note

① Assuming bdd initial cond. s : bdd $\Leftrightarrow \tilde{q}$ & $\dot{\tilde{q}}$ bdd.

② $s \rightarrow 0 \Rightarrow \tilde{q} \text{ \& \ } \dot{\tilde{q}} \rightarrow 0$

③ Look at Ham. Form again with $g \neq 0$

$$\dot{q}^T (\tau - g) = \frac{1}{2} \frac{d}{dt} [\dot{q}^T H \dot{q}]$$

$$\Rightarrow \dot{q}^T (H\ddot{q} + C\dot{q} + g - g) = \dot{q}^T H\ddot{q} + \frac{1}{2} \dot{q}^T \dot{H}\dot{q}$$

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

$$\Rightarrow H - 2C \text{ is skew-symmetric}$$

$$\Rightarrow (H - 2C)^T = -(H - 2C)$$

$$\Rightarrow \dot{H} = C + C^T$$

↓ not quite obvious \Rightarrow Accept.
but proof is too long (p. 400).

9.2 Adaptive Robot Trajectory Control.

We deal with the situation when the manipulator's mass properties are unknown. But want $q \rightarrow q_d$.

a : unknown param. of mass properties (const. vector).

$$\tilde{a} = \hat{a} - a$$

$$V(t) = \frac{1}{2} [s^T H s + \tilde{a}^T \Gamma^{-1} \tilde{a}] \quad \text{where } \Gamma \text{ is a sym. pos. def. matrix}$$

$$\Rightarrow \dot{V}(t) = s^T ([-H\ddot{q}_r - C\dot{q}_r - g] + \hat{a}^T \Gamma^{-1} \dot{\tilde{a}})$$

Since H, C, g all linearly depends on a , we can construct Y s.t.

$$H\ddot{q}_r + C\dot{q}_r + g = Y(q, \dot{q}, \ddot{q}, \dot{q}_r, \ddot{q}_r) a$$

Now, take $L = Y\hat{a} - K_D s$ using P.D controller,

$$\Rightarrow \dot{V} = s^T (Y\hat{a} - K_D s - Y a) + \hat{a}^T \Gamma^{-1} \dot{\tilde{a}}$$

$$= s^T Y \tilde{a} - s^T K_D s + \hat{a}^T \Gamma^{-1} \dot{\tilde{a}}$$

update \hat{a} by $\dot{\hat{a}} = -\Gamma Y^T s$

$$\Rightarrow \dot{V} = -s^T K_D s \leq 0 \quad : \text{ global stability \& convergence of } \hat{q} \text{ to } 0$$

Dealing with prob.

Define $V(t) = \frac{1}{2} s^T H s$

$$\begin{aligned} \Rightarrow \dot{V}(t) &= s^T H \dot{s} + \frac{1}{2} s^T \dot{H} s \\ &= s^T H (\ddot{q} - \ddot{q}_r) + \frac{1}{2} s^T \dot{H} s \quad \downarrow \quad s = \dot{q} - \dot{q}_r \\ &= s^T (\underbrace{L - C\dot{q} - g}_{H\ddot{q} = L - C\dot{q} - g} - H\ddot{q}_r) + \frac{1}{2} s^T \dot{H} s \\ &= s^T \{ L - C(s + \dot{q}_r) - g - H\ddot{q}_r \} + \frac{1}{2} s^T \dot{H} s \\ &= s^T (L - H\ddot{q}_r - C\dot{q}_r - g) + \frac{1}{2} s^T (\dot{H} - 2C) s \end{aligned}$$

Choose $L = \hat{L} - K \cdot \text{sgn}(s)$ const. vector we can chose element wise Where $\hat{L} = \hat{H}\ddot{q}_r + \hat{C}\dot{q}_r + \hat{g}$ (makes $\dot{v} = 0$ given all dynamics known)

$$\Rightarrow \dot{V} = s^T [\tilde{H}\ddot{q}_r + \tilde{C}\dot{q}_r + \tilde{g}] - \sum_i k_i |s_i| \quad (\tilde{H} = \hat{H} - H)$$

Given bounds of modelling errors $\tilde{H}, \tilde{C}, \tilde{g}$, if we choose k_i s.t.

$$k_i \geq | [\tilde{H}\ddot{q}_r + \tilde{C}\dot{q}_r + \tilde{g}]_i | + \eta_i, \text{ then } \dot{V} \leq - \sum_i \eta_i |s_i|$$

we get the desired sliding cond.

\Rightarrow sliding surface $s = 0$ is reached in finite time.

$\hookrightarrow q \rightarrow q_d$ exponentially.