ME 599/699 Robot Modeling & Control Fall 2021

Independent Joint Control

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Introduction

- Derived an EL model for a torque-driven robot
- Torques from voltage-controlled PMDC motor ⇒ individual motor model
- Combine motor and link models for link angle dynamics
- Extend model by including flexibility

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- ▶ Proportional Control: $u(t) = -k_p(\theta_m(t) \theta_d(t))$
- Closed-loop model:

$$(Js^2 + Bs)\Theta_m(s) = -k_p\Theta_m(s) + k_p\Theta_d(s) - D(s), \quad (2)$$

or

$$\Theta_m(s) = \frac{k_p}{Js^2 + Bs + k_p} \Theta_d(s) - \frac{1}{Js^2 + Bs + k_p} D(s).$$
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 - In practice, it's a bad idea due to sensor noise and actuator nonlinearities

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$$\Theta_m(s) = \frac{k_p}{Js^2 + (B+k_d)s + k_p}\Theta_d(s) - \frac{1}{Js^2 + (B+k_d)s + k_p}D(s).$$

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- As long as k_p > 0, k_d > 0, and disturbances are bounded, the closed loop system is stable.
- Similar behavior as P control for set-point regulation, with some more control on transient behavior
- In practice, full control on transients limited by actuator saturation

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Proportional-Integral-Derivative Control

Add an integral term:

$$u(t) = -k_p(\theta_m(t) - \theta_d(t)) - K_d \dot{\theta}_m(t) - k_I \int_0^t (\theta_m(\eta) - \theta_d(\eta)) d\eta,$$

Closed loop:

$$\Theta_m(s) = rac{k_p s + k_l}{J s^3 + (B + k_d) s^2 + k_p s + k_l} \Theta_d(s) \ - rac{s}{J s^3 + (B + k_d) s^2 + k_p s + k_l} D(s).$$

- ▶ If system is stable, $\theta_m(t) \rightarrow \theta_d(t)$, even when non-zero constant disturbance present
- Stability requires $k_I < \frac{k_p(B+k_d)}{J}$
 - We can use lower gains to track set-point θ_d
 - However, may need to know J, B to avoid instability
 - Also, integrators interact poorly with actuator saturation (wind-up problem)

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 Eg:

Model:
$$(Js^2 + B)\Theta_m(s) = U(s) - D(s)$$

 $\implies b(s) = (Js^2 + B)$
 $\implies u_{FF}(t) = J\ddot{\theta}_d(t) + B\dot{\theta}_d(t)$