

Attitude Control Crash Course

- We're trying to point our satellite with only torque coils (no reaction wheels)
- This is hard because the system is underactuated

* Underactuated Systems:

- Manipulator Equation:

$$\underbrace{M(q) \ddot{v}}_{\text{mass matrix}} + \underbrace{C(q, v)v}_{\text{Coriolis term}} + \underbrace{G(q)}_{\text{potential term}} = \underbrace{B(q)u}_{\substack{\text{Input} \\ \text{Jacobain}}}$$

q = "configuration" or "generalized coordinates"

v = Velocity

u = control input

- A system is fully actuated if $\text{rank}(B(q)) = \dim(v)$.
- If this is true, we can cancel out the natural dynamics and do whatever we want:

$$u = -B^{-1}(M\ddot{v} + Cv + G)$$

B must be invertable!

* Satellite Dynamics :

$$\underbrace{\mathbf{J} \dot{\omega} + \omega \times \mathbf{J} \omega = \boldsymbol{\gamma},}_{\text{Euler's Equation}} \quad \boldsymbol{\gamma} = \frac{\mathbf{m}}{T} \times \mathbf{B} \xrightarrow{\substack{\text{magnetic} \\ \text{moment}}} \xrightarrow{\substack{\text{Earth's} \\ \text{magnetic field}}}$$

- In manipulator form:

$$\mathbf{M}(q) = \mathbf{J} \quad , \quad (\mathbf{C}_{q,v}) = \hat{\omega}^T \quad , \quad \mathbf{B}(q) = -\hat{\mathbf{b}}$$

- Hat matrix :

$$\hat{\omega} = \begin{bmatrix} 0 & -\omega_3 & \omega_2 \\ \omega_3 & 0 & -\omega_1 \\ -\omega_2 & \omega_1 & 0 \end{bmatrix} \Rightarrow \hat{\omega} \mathbf{x} = \omega \times \mathbf{x}$$

- Note $\text{rank}(\hat{\mathbf{b}}) = 2 \Rightarrow \text{underactuated!}$

Trajectory Optimization

- We're going to turn the control problem into the following optimization problem!

$$\underset{\substack{\mathbf{x}_{1:N}, \mathbf{u}_{1:N-1} \\ \text{states} \quad \text{controls}}}{\text{minimize}} \quad \mathbf{J} = \sum_{k=1}^{N-1} \underbrace{\mathcal{L}(\mathbf{x}_k, \mathbf{u}_k)}_{\substack{\text{cost} \\ \text{function}}} + \underbrace{\mathcal{L}_F(\mathbf{x}_N)}_{\substack{\text{stage} \\ \text{cost}}} + \underbrace{\mathcal{L}_T(\mathbf{x}_N)}_{\substack{\text{Terminal} \\ \text{cost}}}$$

$$\text{s.t.} \quad \mathbf{x}_{n+1} = \mathbf{f}(\mathbf{x}_n, \mathbf{u}_n) \leftarrow \text{dynamics constraints}$$

$$u_{\min} \leq u_n \leq u_{\max} \leftarrow \text{input constraints}$$

- We're going to assume our cost functions are quadratic

$$L(x, u) = \frac{1}{2} x^T Q x + q^T x + \frac{1}{2} u^T R u + r^T u$$

$$L_F(x) = \frac{1}{2} x^T Q_F x + q_F^T x$$

- We will also need to linearize our dynamics :

$$x_{n+1} + \delta x_{n+1} = f(x_n + \delta x_n, u_n + \delta u_n) \approx f(x_n, u_n) + \underbrace{A_n}_{\frac{\partial f}{\partial x}} \delta x_n + \underbrace{B_n}_{\frac{\partial f}{\partial u}} \delta u_n$$

$$\Rightarrow \delta x_{n+1} = A_n \delta x_n + B_n \delta u_n$$

- If the dynamics were actually linear, this would be an LQR problem. We're going to iteratively linearize + solve LQR problems until convergence.

* LQR (Linear-Quadratic Regulator)

- We're going to calculate the solution backward from the goal.
- Define the "cost-to-go" function

$$V_n(x) = \frac{1}{2} x^T S_n x + s_n^T x$$

- Starting from the end:

$$V_N(x) = L_F(x) = \frac{1}{2} x^T Q_F x + q_F^T x$$

$$\Rightarrow S_N = Q_F, \quad s_N = q_F$$

- Backing up one step:

$$V_{N-1}(x) = \min_{u_{N-1}} L(x_{N-1}, u_{N-1}) + V_N(A_{N-1}x_{N-1} + B_{N-1}u_{N-1})$$

$$\Rightarrow \frac{\partial L}{\partial u_{N-1}} + \underbrace{\frac{\partial V_N}{\partial x_{N-1}} \frac{\partial x_{N-1}}{\partial u_{N-1}}}_{B_{N-1}^T} = 0$$

$$\Rightarrow R u_{N-1} + r + B_{N-1}^T S_N (A x_{N-1} + B u_{N-1}) + B_{N-1}^T s_N = 0$$

$$\Rightarrow \boxed{u_{N-1} = -(R + B_{N-1}^T S_N B_{N-1})^{-1} (r + B_{N-1}^T s_N + B_{N-1}^T S_N A x_{N-1})}$$

$$= -l_{N-1} - K_{N-1} x_{N-1}$$

- Plug u_m back into cost-to-go:

$$V_{N-1}(x) = L(x_{m+1}, -l_{m+1} - K_{m+1}x_{m+1}) + V_{N-1}([A_{m+1} - B_{m+1}K_{m+1}]x_m - B_{m+1}l_{m+1}) \\ = \frac{1}{2} x^T S_{m+1} x + s_{m+1}^T x$$

$$\Rightarrow \boxed{S_{N-1} = Q + K_{m+1}^T R K_{m+1} + (A_{m+1} - B_{m+1}K_{m+1})^T S_N (A_{m+1} - B_{m+1}K_{m+1})} \\ \boxed{s_{N-1} = q - K_{m+1}^T r + K_{m+1}^T R l_{m+1} + (A_{m+1} - B_{m+1}K_{m+1})^T (s_N - S_N B_{m+1} l_{m+1})}$$

- Now we're back to where we started and we can continue backwards until we get to $k=1$