

Last Time:

- LQG
- Kalman

Today:

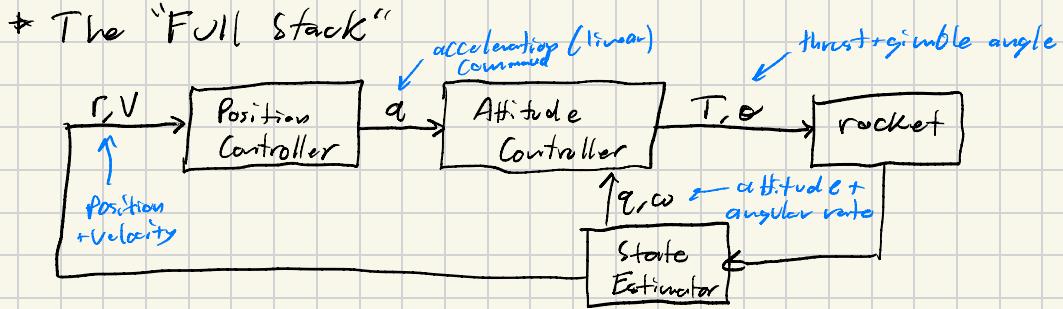
- Rocket Soft Landing

- * The Rocket Soft-Landing Problem:

- Go from some initial state x_0 to some final position r_f with $Z_f = 0$, $V_f \approx 0$
- Minimize some combination of fuel consumption and/or landing position error $\|r_f - r_g\|$
- Respect thrust limits + safety constraints

- * Examples:

- NASA Curiosity "Sky Crane" (2012)
- SpaceX Falcon, Starship
- NASA Perseverance w/TRN (2021)



- State Estimation

SpaceX: GPS + IMU with good filtering $\sim 1\text{ m}$ position accuracy, $< 1\text{ cm/s}$ velocity, $\sim 1^\circ$ attitude.

Mars: No GPS. IMU + Radar Altimeter + Vision $\sim 30\text{ meter}$ Accuracy. Avoid Boulders.

- Decoupled Control Loop:

High-Level Position Controller: Uses a point-mass model. Reasons about safety, thrust limits, and fuel. Generating acceleration commands. Runs at $\sim 6\text{ Hz}$

Low-Level Attitude Controller: Reasons about attitude, flexible modes, fluid slosh, generates thrust + gimble commands to track desired acceleration. Runs at $\sim 10\text{ Hz}$

Rocket Dynamics



Rigid Body

$$\ddot{\mathbf{r}} = -\mathbf{g} + \frac{\mathbf{T}}{m} \leftarrow \text{point mass}$$

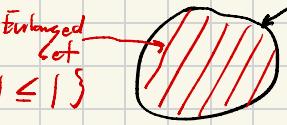
$$\dot{m} = -\alpha T \quad \leftarrow \text{fuel burn}$$

$$\begin{aligned} \text{Attitude} & \left\{ \begin{array}{l} \mathbf{T}_{\text{ci}} + \omega \times \mathbf{T}_{\text{c}\omega} = \underline{l} \times \underline{T} \\ \text{Controller} \\ (\text{fast}) \end{array} \right. & \left. \begin{array}{l} \text{Inertia} \\ \text{torque} \end{array} \right\} & \leftarrow \text{attitude} \end{aligned}$$

- Fuel can be 80% + of initial vehicle mass. Have to account for this.
- Fluid slosh: Highly nonlinear, time-varying, hard to model. Standard model: pendulum.
- Flexible Modes: Rockets are built to be ~~light~~ \Rightarrow not stiff \Rightarrow low-frequency bending modes. First modes $\approx 1\text{Hz}$. Dealt with by adding notch filters to the attitude controller at bending frequencies.
- Aerodynamic Forces: Mostly ignore.
- Lots of model error \Rightarrow linear robust control ideas are used in the attitude controller.

* Background: Convex Relaxations

- Sometimes we have a nonconvex constraint that can be expressed as the boundary of a convex set!


Original set S_2
 $S_2 = \{x \mid \|x\| \leq 1\}$
 $S_1 = \{x \mid \|x\| = 1\}$

$$S_1 = \partial S_2$$

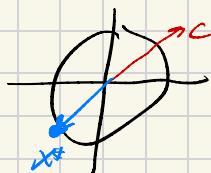
- Replacing the original constraint with larger convex one is called "convex relaxation"
- Sometimes if the cost is "nice" we can still get the answer to the original problem by solving the relaxed version:

$$\min c^T x$$

$$\text{s.t. } \|x\| = 1$$



$$\|x\| \leq 1$$

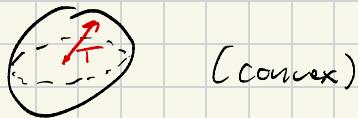


- When this happens we call it a "tight" relaxation.

* Convex Relaxation of Thrust Constraint

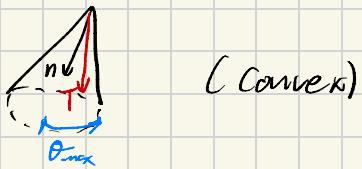
- Maximum thrust constraint

$$T \in \mathbb{R}^3, \|T\| \leq T_{\max}$$



- Thrust angle constraint

$$\frac{n^T T}{\|T\|} \leq \cos(\theta_{\max})$$



- Rocket engines also have a minimum thrust constraint:

$$T_{\min} \leq \|T\| \leq T_{\max}$$



- Let's add a new "slack variable" $\Gamma \in \mathbb{R}$ that equals the thrust magnitude:

$$\begin{aligned} 1) \quad \|T\| = \Gamma & \quad \left. \right\} \text{Boundary of a convex set (sphere)} \\ 2) \quad T_{\min} \leq \Gamma \leq T_{\max} \\ 3) \quad n^T T \leq \Gamma \cos(\theta_{\max}) \end{aligned} \quad \left. \right\} \text{Convex}$$

- Now we can convexify the constraint by relaxing 1):

$$\begin{aligned} 1') \quad \|T\| \leq \Gamma \\ 2) \quad T_{\min} \leq \Gamma \leq T_{\max} \\ 3) \quad n^T T \leq \Gamma \cos(\theta_{\max}) \end{aligned}$$

- The paper proves that this relaxation is tight using Pontryagin's minimum principle.