

Last Time:

- LQG
- Kalman Filter

Today:

- Rocket Soft Landing
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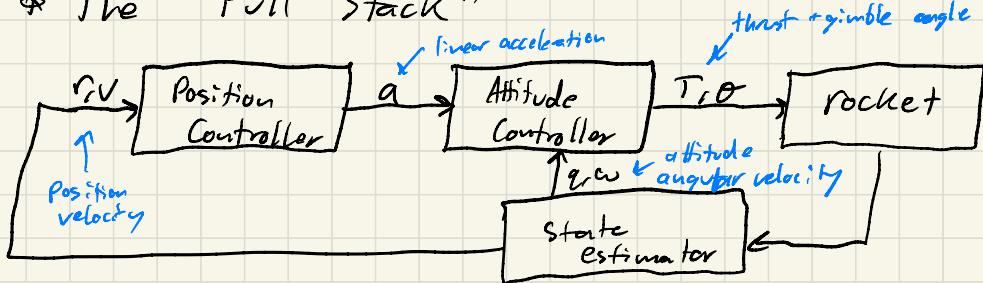
* The Rocket Soft-Landing Problem

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

* Examples:

- NASA Curiosity "Sky Crane" (2012)
- SpaceX Falcon 9, Starship
- NASA Perseverance w/TRN (2021)

* The "Full Stack"



- State Estimation:

SpaceX: GPS + IMU with good filtering, ~1m position accuracy, < 1 cm/sec velocity, ~1° attitude. Enables precision landing.

Mars: No GPS. IMU + Radar Altimeter + Vision ~ 30 meter accuracy. Avoid Boulders.

- Decoupled Control Loop:

High-level Position Controller: Uses a point-mass model. Reasons about safety, thrust, and fuel. Generates acceleration commands. Runs at ~6Hz

Low-level Attitude Controller: Reasons about attitude, flexible modes, fluid slosh, generates thrust+gimble commands to track desired acceleration. Runs at ~10 Hz

* Rocket Dynamics



- Rigid Body

$$\text{Position} \quad \dot{\vec{r}} = -\vec{g} + \frac{\vec{T}}{m} \leftarrow \text{point mass}$$

$$\text{Control} \quad (\text{slow}) \quad \dot{\vec{m}} = -\vec{x} T \leftarrow \text{fuel burn}$$

$$\text{attitude} \quad \begin{cases} \text{controller} & J \dot{\vec{\omega}} + \vec{\omega} \times J \vec{\omega} = \underline{\vec{T}_{inertia}} \\ \text{(fast)} & \end{cases} \leftarrow \text{attitude}$$

$\vec{T}_{inertia}$

- Fuel can be 80% + of initial vehicle mass. Have to account for this

- Fluid slosh: Highly nonlinear, hard to model



\rightarrow pendulum with ℓ_p and m_p fit to data

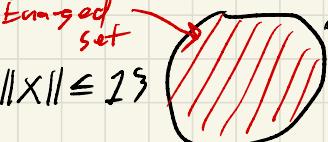
- Flexible modes: Rockets are built to be very light
 \Rightarrow not stiff \Rightarrow low-frequency bending modes. First mode $\approx 1\text{Hz}$. Dealt with by adding notch filters to the attitude controller at the bending frequencies to avoid exciting it.

- Aerodynamic Forces: Mostly ignore. Velocity constraints in position controller to make sure these stay small.

- Lots of model uncertainty. Linear robust control ideas are used in the attitude controller.

* Background: Convex Relaxation

- Sometimes we have a non-convex constraint that can be expressed as the boundary of a larger convex set:

 original set S_2

$$S_2 = \{x \mid \|x\| \leq 1\}$$

$$S_1 = \{x \mid \|x\| = 1\}$$

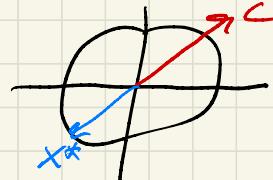
$$S_1 = \partial S_2$$

- 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:

$$\min C^T x$$

$$\text{st. } \|x\| = 1$$

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



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

* Convex Relaxation of Thrust Constraint

- Maximum thrust constraint

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



(convex)

- Thrust angle constraint

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



(convex)

- Rocket engine also has minimum thrust constraint:

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



Not convex!

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

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

- Now we can convexify the constraint by relaxing 1)

$$\begin{aligned} 1) \|T\| \leq \Gamma \\ 2) T_{\min} \leq \Gamma \leq T_{\max} \\ 3) n^T T \leq \Gamma \cos(\theta_{\max}) \end{aligned}$$

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