Navigation System for Low-Cost Lunar Lander with LOS Measurements

ECE/MAE 7560: Optimal Estimation

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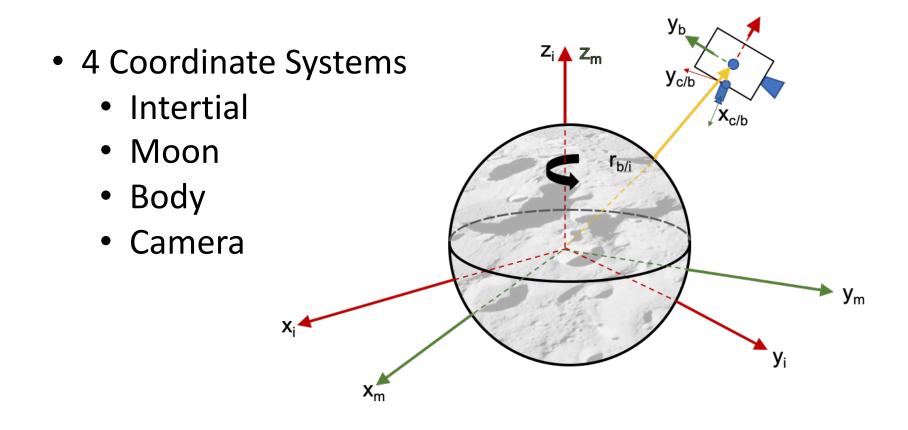
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Introduction

- Returning to the moon is high-priority
- Automation is desirable for establishing a long-term presence
- Any automated lander needs to be able to make precision landings, navigate unmapped areas, and should be as low-cost as possible
- Many autonomous landing and navigation schemes already exist, but the good ones tend to be very expensive (6+ sensors = \$\$\$)
- Trade-off: efficacy vs. cost
- Proposed solution: Minimal sensor suite (monocular camera and accelerometer)

Coordinate Systems



Truth vs. Design States

• Truth State

$$oldsymbol{x_t} = \left[egin{array}{c} oldsymbol{r_{b/i}^i} \ oldsymbol{v_{b/i}^i} \ q_i^m \ q_b^c \ b_r \ oldsymbol{\epsilon_g^i} \ h_t \ oldsymbol{b}^b \end{array}
ight] = \left[egin{array}{c} oldsymbol{x_v} \ oldsymbol{p} \end{array}
ight] & oldsymbol{x_v} = \left[egin{array}{c} oldsymbol{r_{b/i}^i} \ oldsymbol{v_{b/i}^i} \ oldsymbol{q_i^m} \end{array}
ight] & oldsymbol{p} = \left[egin{array}{c} oldsymbol{r_{b/i}^i} \ oldsymbol{v_{b/i}^i} \ oldsymbol{q_i^c} \ oldsymbol{h_t} \ oldsymbol{b_n^c} \ oldsymbol{h_t} \ oldsymbol{b_n^c} \end{array}
ight] & oldsymbol{p} = \left[egin{array}{c} oldsymbol{r_{b/i}^i} \ oldsymbol{v_{b/i}^i} \ oldsymbol{q_i^c} \ oldsymbol{h_t} \ oldsymbol{b_n^c} \ oldsymbol{h_t} \ oldsymbol{b_n^c} \end{array}
ight] & oldsymbol{p} = \left[egin{array}{c} oldsymbol{r_{b/i}^i} \ oldsymbol{v_{b/i}^i} \ oldsymbol{h_t} \ oldsymbol{b_n^c} \ oldsymbol{h_t} \ oldsymbol{b_n^c} \end{array}
ight] & oldsymbol{q_i^c} \end{array}$$

Design State

Truth vs. Design States

• Truth State

Design State

$$egin{aligned} \dot{oldsymbol{x}} &= \left[egin{aligned} \dot{oldsymbol{r}}_{b/i}^i \ \dot{oldsymbol{b}}_{b/i}^i \ \dot{oldsymbol{b}}_{b/i}^i \ \dot{oldsymbol{b}}_{b/i}^i \ \dot{oldsymbol{b}}_{b/i}^i \ \dot{oldsymbol{c}}_{b/i}^i \ \dot{oldsymbol{c}}_{b/$$

Linearization of Model

- Perturbation Model
- General form

$$\delta \dot{\boldsymbol{x}} = F(\hat{\boldsymbol{x}})\delta \boldsymbol{x} + B\omega$$

• Taylor Series resulting in Jacobian

$$egin{bmatrix} \delta oldsymbol{r}_{b/i}^i \ \delta oldsymbol{v}_{b/i}^i \ \delta oldsymbol{b}_r^i \ \delta oldsymbol{e}_g^i \ \delta oldsymbol{h}_t \ \delta oldsymbol{b}_a^b \end{bmatrix} = egin{bmatrix} oldsymbol{r}_{b/i}^i \ oldsymbol{v}_{b/i}^i \ oldsymbol{b}_r^i \ oldsymbol{\epsilon}_g^i \ oldsymbol{h}_t \ oldsymbol{b}_a^b \end{bmatrix} - egin{bmatrix} oldsymbol{r}_{b/i}^i \ oldsymbol{\hat{v}}_{b/i}^i \ oldsy$$

$$F = \frac{df(x,u)}{dx} \Big|_{x=\hat{x}}$$

$$= \begin{bmatrix} 0_3 & I_3 & 0_{3\times 1} & 0_3 & 0_{3\times 1} & 0_3 \\ \frac{-\mu(I_3 - 3i_r i_r^T)}{\|\hat{r}\|^3} & 0_3 & 0_{3\times 1} & I_3 & 0_{3\times 1} & 0_3 \\ 0_3 & 0_3 & \frac{-1}{\tau_r} & 0_3 & 0_{3\times 1} & 0_3 \\ 0_3 & 0_3 & 0_{3\times 1} & \frac{\|\hat{V}_\perp\|}{dg} I_3 & 0_{3\times 1} & 0_3 \\ 0_3 & 0_3 & 0_{3\times 1} & 0_3 & \frac{\|\hat{V}_\perp\|}{dh} & 0_3 \\ 0_3 & 0_3 & 0_{3\times 1} & 0_3 & 0_3 & \frac{-1}{\tau_{accl}} \end{bmatrix}$$

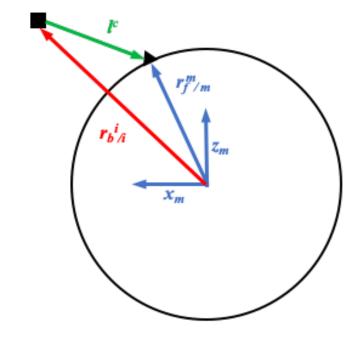
Measurement Model

• Camera measurements

$$oldsymbol{\ell}^c = T_b^c T_i^b (T_m^i oldsymbol{r}_{f/m}^m - oldsymbol{r}_{b/i}^i) = \left[egin{array}{c} \ell_x \ \ell_y \ \ell_z \end{array}
ight]$$

Non-linear Model

$$ilde{oldsymbol{z}} = \left[egin{array}{c} rac{\ell_x}{\ell_z} \ rac{\ell_y}{\ell_z} \end{array}
ight] + oldsymbol{v}_c$$



Measurement Model Linearization

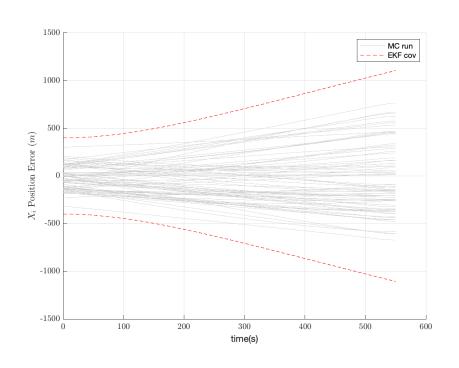
• Linearized model $\delta \tilde{z} = H(\hat{x})\delta x + G\nu$

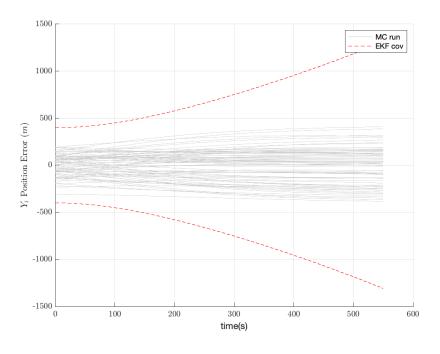
Linearization done in two steps

$$H = \frac{dh}{dl} \frac{dl}{dx} \Big|_{x=\hat{x}} \longrightarrow \frac{dh}{dl} = \begin{bmatrix} 1/\tilde{l}_z & 0 & -\tilde{l}_x/\tilde{l}_z^2 \\ 0 & 1/\tilde{l}_z & -\tilde{l}_y/\tilde{l}_z^2 \end{bmatrix}$$
$$\frac{dl}{dx} = \begin{bmatrix} -T_b^i T_i^b I_3 & 0_{3\times 11} \end{bmatrix}$$

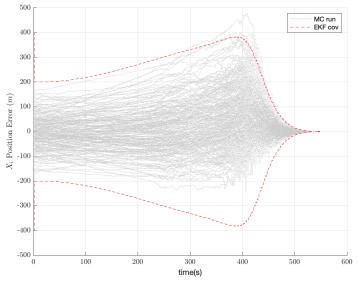
• Result $H = \begin{bmatrix} 1/\tilde{l}_z & 0 & -\tilde{l}_x/\tilde{l}_z^2 \\ 0 & 1/\tilde{l}_z & -\tilde{l}_y/\tilde{l}_z^2 \end{bmatrix} \begin{bmatrix} -T_b^i T_i^b I_3 & 0_{3\times 11} \end{bmatrix}$

Estimation plots before filter

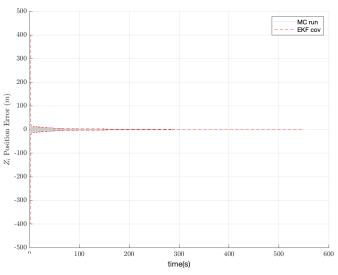


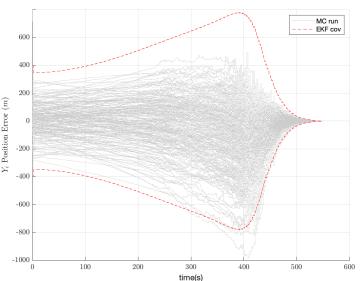


Estimation plots

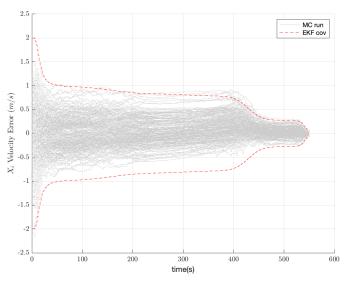


- X position looks okay, though biased
- Y position looks good, though biased
- Z position looks great

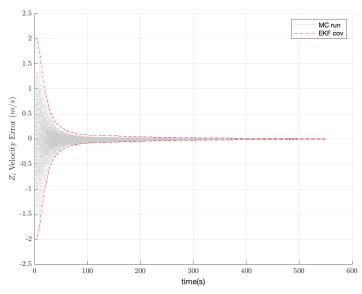


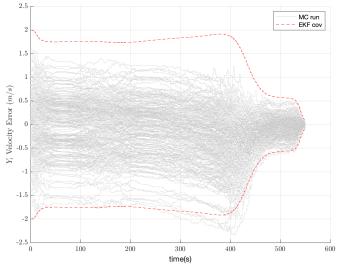


Estimation plots



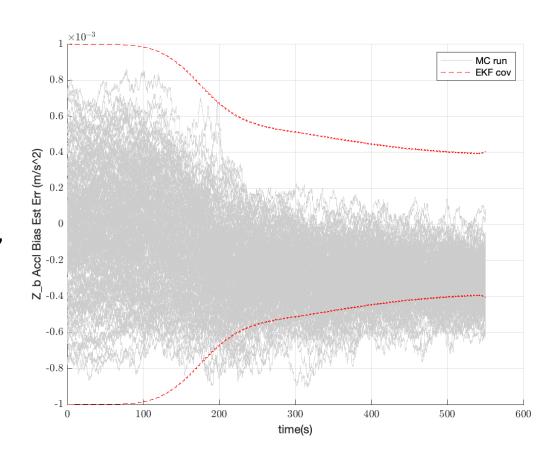
- X velocity looks okay, though biased
- Y velocity looks good, though biased
- Z velocity looks great



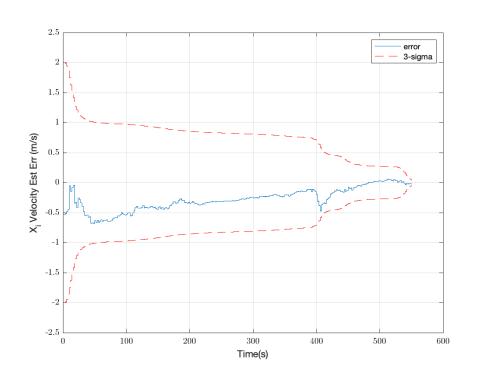


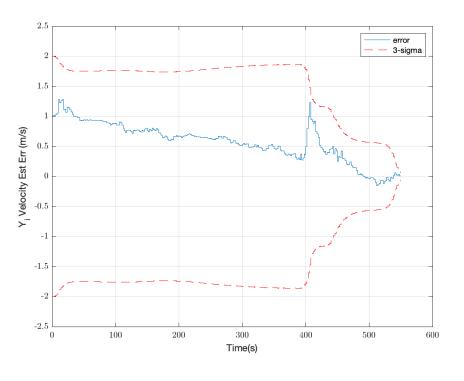
Estimation plots

- Not yet sure what is going on with this
- Probably what is causing the position and velocity problems
- Covariance looks good, probably a bug in the state propagation



Residual Plots from single Monte Carlo





Results

- The lander is very accurate in some states
 - Position within a couple hundred meters
 - Velocity within 0.5 m/s
- Executes a fairly soft landing, although it could be softer
- Using one or two more sensors could greatly improve the system – but it would be more expensive to add those
 - Radio altimeter
 - Gyro
 - Star Tracker

Conclusions

- Lunar lander with a minimal sensor suite is feasible
- Linearizations can occasionally be very difficult
- Communication between different parts of the simulation is key
- EKFs are very hard to debug
 - Isolate as much as possible
 - Debug in small chunks
 - It just takes a long time
 - Validate where possible
 - Coordinate systems are very important
- I could take the class again and learn just as much