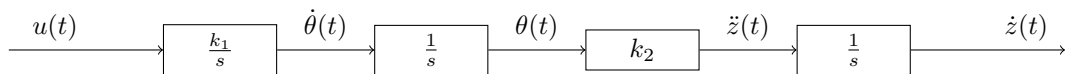
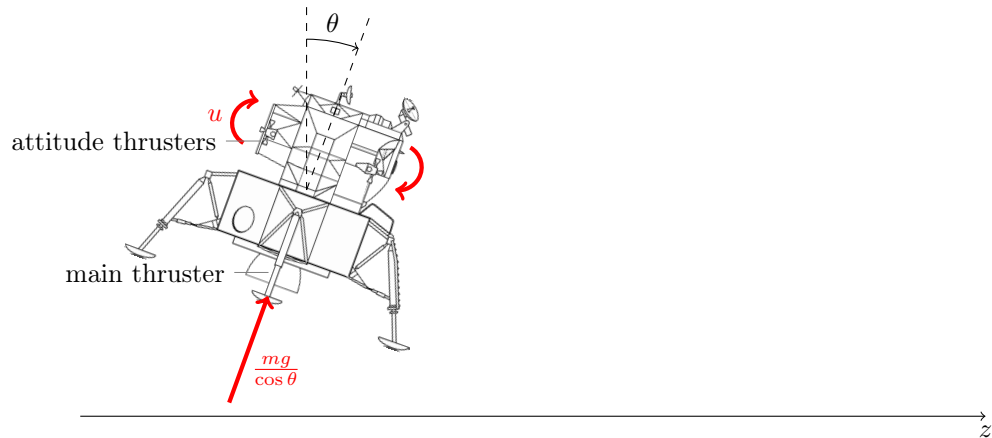


Output feedback (observer)

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Example - The Apollo lunar module



Example - The Apollo lunar module

State variables: $x = [x_1 \ x_2 \ x_3]^T = [\dot{\theta} \ \theta \ \dot{z}]^T$. With dynamics

$$\begin{cases} \dot{x}_1 = \ddot{\theta} = k_1 u \\ \dot{x}_2 = \dot{\theta} = x_1 \\ \dot{x}_3 = \ddot{z} = k_2 \theta = k_2 x_2 \end{cases}$$

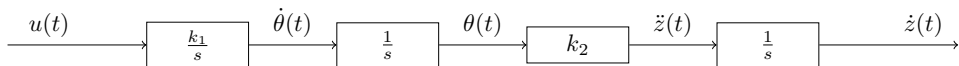
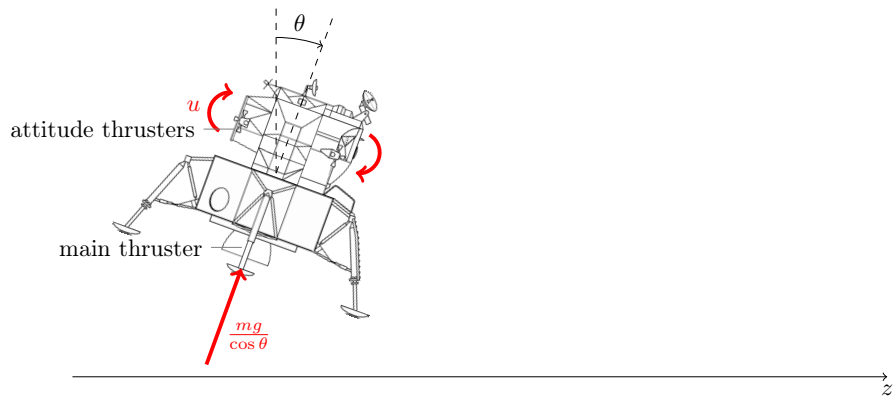
$$\dot{x} = \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \\ \dot{x}_3 \end{bmatrix} = \underbrace{\begin{bmatrix} 0 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & k_2 & 0 \end{bmatrix}}_A \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} + \underbrace{\begin{bmatrix} k_1 \\ 0 \\ 0 \end{bmatrix}}_B u$$

Example - The Apollo lunar module

$$\begin{aligned}x(kh + h) &= e^{Ah}x(kh) + \int_0^h e^{As}Bu(kh + h - s)ds \\&= \underbrace{e^{Ah}}_{\Phi(h)}x(kh) + \underbrace{\left(\int_0^h e^{As}Bds\right)}_{\Gamma(h)}u(kh) \\&= \begin{bmatrix} 1 & 0 & 0 \\ h & 1 & 0 \\ \frac{h^2 k_2}{2} & hk_2 & 1 \end{bmatrix} x(kh) + k_1 \begin{bmatrix} h \\ \frac{h^2}{2} \\ \frac{k_2 h^3}{6} \end{bmatrix} u(kh)\end{aligned}$$

State feedback with reconstructed states

State feedback with reconstructed states



State feedback

Given

$$\begin{aligned}x(k+1) &= \Phi x(k) + \Gamma u(k) \\ y(k) &= Cx(k)\end{aligned}\tag{1}$$

and measurements (or estimates) of the state vector $x(k)$.

Linear state feedback is the control law

$$\begin{aligned}u(k) &= f((x(k), u_c(k))) = -l_1x_1(k) - l_2x_2(k) - \cdots - l_nx_n(k) + l_0u_c(k) \\ &= -Lx(k) + l_0u_c(k),\end{aligned}$$

where

$$L = [l_1 \quad l_2 \quad \cdots \quad l_n] .$$

Substituting the control law in the state space model (8) gives

$$\begin{aligned}x(k+1) &= (\Phi - \Gamma L) x(k) + l_0\Gamma u_c(k) \\ y(k) &= Cx(k)\end{aligned}\tag{2}$$

Observer design

Given model

$$\begin{aligned}x(k+1) &= \Phi x(k) + \Gamma u(k) \\ y(k) &= Cx(k)\end{aligned}$$

and measurements of the output signal $y(k)$.

The observer is given by

$$\hat{x}(k+1) = \underbrace{\Phi \hat{x}(k) + \Gamma u(k)}_{\text{simulation}} + \underbrace{K(y(k) - C\hat{x}(k))}_{\text{correction}} = (\Phi - KC) \hat{x}(k) + \Gamma u(k) + Ky(k)$$

with poles given by the eigenvalues of the matrix $\Phi_o = \Phi - KC$

Rule-of-thumb Choose the poles of the observer (eigenvalues of $\Phi - KC$) at least twice as fast as the poles (eigenvalues) of $\Phi - \Gamma L$.

Observer design

Rule-of-thumb Choose the poles of the observer (eigenvalues of $\Phi - KC$) at least twice as fast as the poles (eigenvalues) of $\Phi - \Gamma L$.

In continuous time (the s-plane), choosing a pole to be twice as fast, means moving the pole to twice the distance from the origin. Given a discrete pole p_1 , the discrete pole in

$$p_2 = \exp\left(2\frac{\ln p_1}{h}h\right) = \exp(2 \ln p_1) = p_1^2$$

corresponds to a response that is twice as fast.

Control by feedback from reconstructed states

The design problem can be separated into two problems

1. Determine the gain vector L and the gain l_0 of the control law

$$u(k) = -L\hat{x}(k) + l_0 u_c(k)$$

so that the closed-loop system has good reference tracking.

2. Determine the gain vector K of the observer

$$\hat{x}(k+1) = \Phi\hat{x}(k) + \Gamma u(k) + K(y(k) - C\hat{x}(k))$$

to get a good balance between disturbance rejection and noise attenuation.

Computing the observer gain

A matrix M and its transpose M^T have the same eigenvalues. Hence, the problem of determining the gain K to obtain desired eigenvalues of

$$\Phi - KC$$

is equivalent to determining the gain K in

$$(\Phi - KC)^T = \Phi^T - C^T K^T.$$

The last problem has the exact same form as the problem of determining L to obtain desired eigenvalues of

$$\Phi - \Gamma L$$

So, the same matlab function can be used for both problems.

Computing the observer gain

1. Ackerman's method

```
K = acker(Phi', C', po)'
```

2. More numerically stable method

```
K = place(Phi', C', pd)'
```