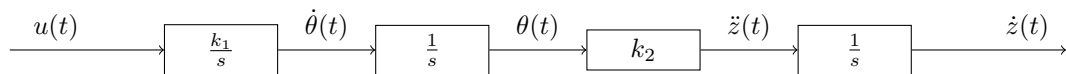
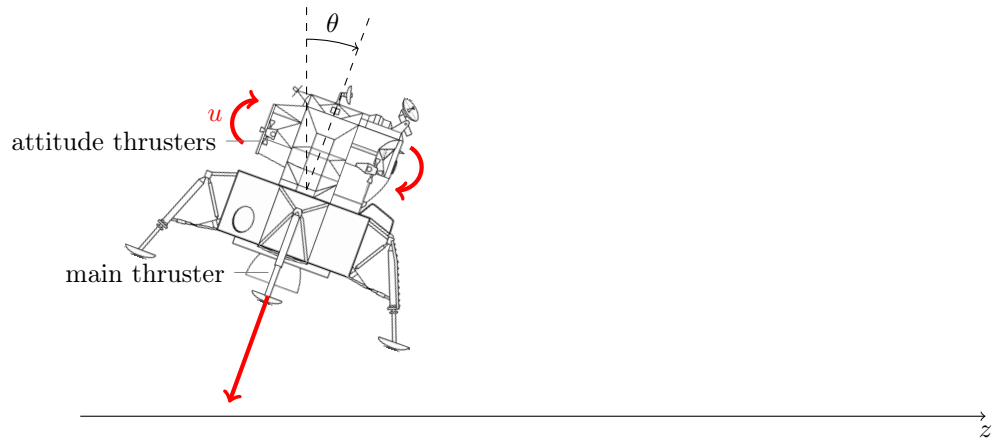


# Control computarizado - Modelos en espacio de estados

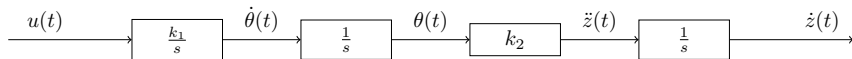
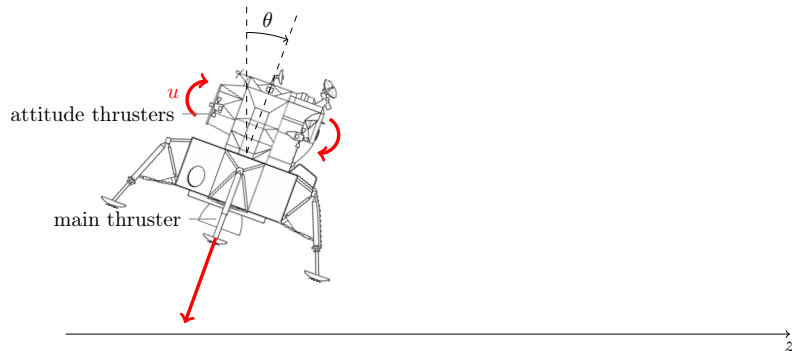
Kjartan Halvorsen

July 21, 2020

## Ejemplo - El modulo lunar de Apollo



## Ejemplo - El modulo lunar de Apollo



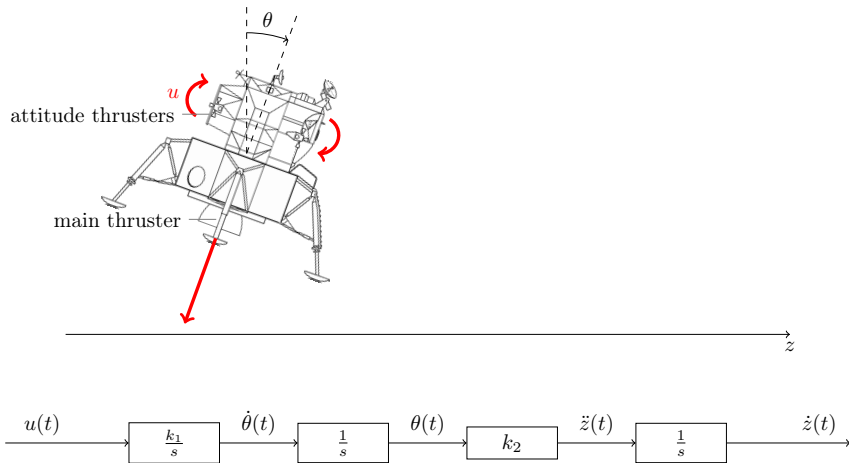
**Actividad** ¿Cuál es la función de transferencia del sistema?

$$1: G(s) = \frac{k_1 k_2}{s^2}$$

$$2: G(s) = \frac{k_1 k_2}{s(s^2 + 1)}$$

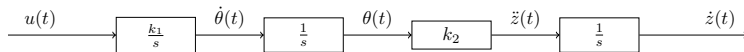
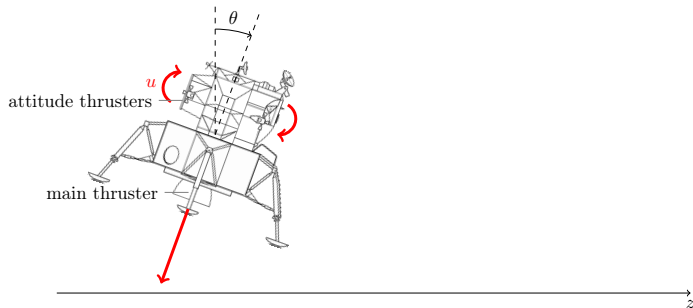
$$3: G(s) = \frac{k_1 k_2}{s^3}$$

## Ejemplo - El modulo lunar de Apollo



**Actividad** ¿Que sensores relevantes se puede usar para el control?

## Ejemplo - El modulo lunar de Apollo



Variables del estado:  $x = [x_1 \ x_2 \ x_3]^T = [\dot{\theta} \ \theta \ \dot{z}]^T$ . Con dinamica

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

## Ejemplo - El modulo lunar de Apollo

Variables del estado:  $x = [x_1 \ x_2 \ x_3]^T = [\dot{\theta} \ \theta \ \dot{z}]^T$ . Con dinamica

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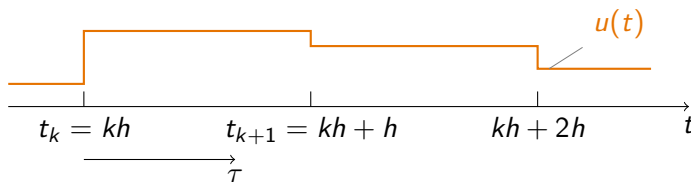
**Actividad** Llena los matrices y vectores en el modelo de espacio de estados

$$\dot{x} = \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \\ \dot{x}_3 \end{bmatrix} = \underbrace{\begin{bmatrix} \phantom{\dot{x}_1} \\ \phantom{\dot{x}_2} \\ \phantom{\dot{x}_3} \end{bmatrix}}_A \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} + \underbrace{\begin{bmatrix} \phantom{\dot{x}_1} \\ \phantom{\dot{x}_2} \\ \phantom{\dot{x}_3} \end{bmatrix}}_B u$$

# Discretización

Solución general de un sistema lineal en espacio de estados

$$x(t_k + \tau) = e^{A(\tau)}x(t_k) + \int_0^\tau e^{As}Bu((t_k + \tau) - s)ds$$



$$\begin{aligned} x(kh + h) &= e^{Ah}x(kh) + \int_0^h e^{As}Bu(kh + h - s)ds \\ &= \underbrace{e^{Ah}}_{\Phi(h)}x(kh) + \underbrace{\left(\int_0^h e^{As}Bds\right)}_{\Gamma(h)}u(kh) \end{aligned}$$

## Discretización - La exponencial de una matriz

Matriz  $A$  cuadrada. Variable  $t$  escalar.

$$e^{At} = 1 + At + \frac{t^2}{2!}A^2 + \frac{t^3}{3!}A^3 + \dots$$

Transformada de Laplace:

$$\mathcal{L}\{e^{At}\} = (sI - A)^{-1}$$



## Discretización - ejemplo

$$\begin{aligned}x(kh + h) &= e^{Ah}x(kh) + \int_0^h e^{As}Bu(kh + h - s)ds \\&= \underbrace{e^{Ah}}_{\Phi(h)}x(kh) + \underbrace{\left(\int_0^h e^{As}Bds\right)}_{\Gamma(h)}u(kh)\end{aligned}$$

$$A = \begin{bmatrix} 0 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & k_2 & 0 \end{bmatrix}, \quad A^2 = \begin{bmatrix} 0 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & k_2 & 0 \end{bmatrix} \begin{bmatrix} 0 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & k_2 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ k_2 & 0 & 0 \end{bmatrix}, \quad A^3 = 0$$

Entonces,

$$\Phi(h) = e^{Ah} = 1 + Ah + A^2h^2/2 + \dots$$

$$= \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} + \begin{bmatrix} 0 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & k_2 & 0 \end{bmatrix} h + \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ k_2 & 0 & 0 \end{bmatrix} \frac{h^2}{2} = \begin{bmatrix} 1 & 0 & 0 \\ h & 1 & 0 \\ \frac{h^2k_2}{2} & hk_2 & 1 \end{bmatrix}$$

## Discretización - ejemplo

$$\begin{aligned}x(kh + h) &= e^{Ah}x(kh) + \int_0^h e^{As}Bu(kh + h - s)ds \\&= \underbrace{e^{Ah}}_{\Phi(h)}x(kh) + \underbrace{\left(\int_0^h e^{As}Bds\right)}_{\Gamma(h)}u(kh)\end{aligned}$$

$$e^{As}B = \begin{bmatrix} 1 & 0 & 0 \\ h & 1 & 0 \\ \frac{s^2 k_2}{2} & sk_2 & 1 \end{bmatrix} \begin{bmatrix} k_1 \\ 0 \\ 0 \end{bmatrix} = k_1 \begin{bmatrix} 1 \\ s \\ \frac{k_2 s^2}{2} \end{bmatrix}$$

$$\Gamma(h) = \int_0^h e^{As}Bds = k_1 \int_0^h \begin{bmatrix} 1 \\ s \\ \frac{k_2 s^2}{2} \end{bmatrix} ds = k_1 \begin{bmatrix} h \\ \frac{h^2}{2} \\ \frac{k_2 h^3}{6} \end{bmatrix}$$

## Discretización - ejemplo

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

# Goal of today's lecture

- ▶ Understand state feedback design

# Stability

A system

$$x(k+1) = \Phi x(k), \quad x(0) = x_0$$

is *asymptotically stable* if  $\lim_{t \rightarrow \infty} x(kh) = 0$  for all  $x_0 \in \mathbb{R}^n$ .

A system is asymptotically stable if and only if **all eigenvalues of  $\Phi$  are inside the unit circle.**

## Reachability (controllability)

Reachability is the answer to the question "Can we by choosing a suitable input sequence  $u(k)$ ,  $k = 0, 1, 2, \dots, n - 1$  reach any point in the state space?"

Consider

$$x(k + 1) = \Phi x(k) + \Gamma u(k).$$

With initial state  $x(0)$  given. The solution at time  $n$  where  $n$  is the order of the system is

$$\begin{aligned} x(n) &= \Phi^n x(0) + \Phi^{n-1} \Gamma u(0) + \dots + \Gamma u(n-1) \\ &= \Phi^n x(0) + W_c U, \end{aligned} \tag{1}$$

where

$$\begin{aligned} W_c &= [\Gamma \quad \Phi \Gamma \quad \dots \quad \Phi^{n-1} \Gamma] \\ U &= [u(n-1) \quad u(n-2) \quad \dots \quad u(0)]^T \end{aligned}$$

## Reachability (controllability), contd

To find the input sequence that takes the state to  $x(n) = x_d$  we solve the equation

$$x_d = \Phi^n x(0) + W_c U$$

for  $U$ .

$$U = W_c^{-1} (x_d - \Phi^n x(0))$$

This requires the matrix  $W_x$  to be **invertible**. This gives Theorem 3.7 in Å&W:

**THEOREM 3.7 REACHABILITY** The state space system above is reachable if and only if the matrix  $W_c$  has rank  $n$ .

This is equivalent to

$$\det W_c \neq 0.$$

# State feedback

Have state space model

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

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

**Linear state feedback** is the control law

$$\begin{aligned}u(k) &= f((x(k), u_c(k))) = -l_1 x_1(k) - l_2 x_2(k) - \cdots - l_n x_n(k) + m u_c(k) \\ &= -Lx(k) + m u_c(k),\end{aligned}$$

where

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

Insert the control law into the state space model (3) to get



## State feedback

Have state space model

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$$u(k) = -l_1 x_1(k) - l_2 x_2(k) - \cdots - l_n x_n(k) + m u_c(k) = -Lx(k) + m u_c(k),$$

where

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

Insert the control law into the state space model (3) to get

$$\begin{aligned}x(k+1) &= (\Phi - \Gamma L) x(k) + m \Gamma u_c(k) \\ y(k) &= Cx(k)\end{aligned}\tag{4}$$

## Pole placement by state feedback

Assume the desired performance of the control system is given as a set of desired closed loop poles  $p_1, p_2, \dots, p_n$ , corresponding to the desired characteristic polynomial

$$a_c(z) = (z - p_1)(z - p_2) \cdots (z - p_n) = z^n + \alpha_1 z^{n-1} + \cdots + \alpha_n. \quad (5)$$

With state feedback we get the the closed-loop system

$$\begin{aligned} x(k+1) &= (\Phi - \Gamma L)x(k) + m\Gamma u_c(k) \\ y(k) &= Cx(k) \end{aligned} \quad (6)$$

with characteristic equation

$$\det(zI - (\Phi - \Gamma L)) = z^n + \beta_1(l_1, \dots, l_n)z^{n-1} + \cdots + \beta_n(l_1, \dots, l_n). \quad (7)$$

Equate the coefficients in (5) and (7) to get the system of equations

$$\beta_1(l_1, \dots, l_n) = \alpha_1$$

$$\beta_2(l_1, \dots, l_n) = \alpha_2$$

$$\vdots$$

$$\beta_n(l_1, \dots, l_n) = \alpha_n$$

## Pole placement by state feedback, contd.

The system of equations

$$\beta_1(l_1, \dots, l_n) = \alpha_1$$

$$\beta_2(l_1, \dots, l_n) = \alpha_2$$

$$\vdots$$

$$\beta_n(l_1, \dots, l_n) = \alpha_n$$

is always linear in the unknown controller parameters, so it can be written

$$AL^T = \alpha,$$

Where  $\alpha^T = [\alpha_1 \quad \alpha_2 \quad \cdots \quad \alpha_n]$ .

## Pole placement and reachability

It can be shown that the controllability matrix  $W_c$  is a factor of the matrix  $A$

$$A = \bar{A}W_c.$$

Hence, in general the system of equations

$$\bar{A}W_cL^T = \alpha \tag{8}$$

has a solution only if  $W_c$  is invertible, i.e. the system is *reachable*.

Note that equation (8) can still have a solution for unreachable systems if  $\alpha$  is in the *column space of  $A$* , i.e.  $\alpha$  can be written

$$\alpha = b_1A_{:,1} + b_2A_{:,2} + \cdots + b_mA_{:,m}, \quad m < n$$