

$$x_1 = \theta$$

$$M(\theta) \rightarrow M(x_1)$$

$$x_2 = \dot{\theta} = \dot{x}_1$$

Linearize about  $(\theta_d, 0)$

$$x_1 = \theta_d$$

$$x_2 = 0$$

$$x_1 = \theta - \theta_d$$

$$M(\theta) \rightarrow M(x_1 + \theta_d)$$

$$x_2 = \dot{\theta} = \dot{x}_1$$

Linearize about  $(0, 0)$

$$(x_1 = 0, x_2 = 0)$$

Linearized system about  $(\theta = \theta_d, \dot{\theta} = 0)$

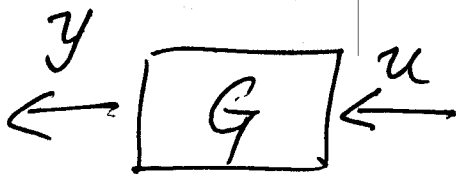
$$\dot{x} = Ax + Bu \rightarrow \text{input torques}$$

$$y = Cx$$

output angles

Create MATLAB LTI object:

$$G = \text{ss}(A, B, C, \underline{0})$$



$\text{zeros}(2, 2)$   
rows columns

$$G(s) = C (sI - A)^{-1} B$$

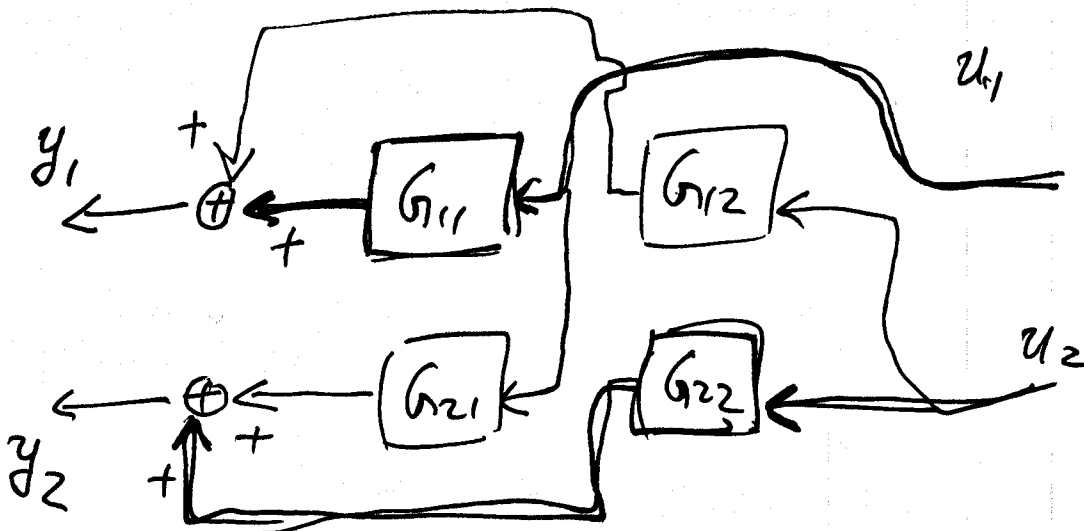
↳ 2x2 identity matrix  $\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$

$$sI = \begin{bmatrix} s & 0 \\ 0 & s \end{bmatrix}$$

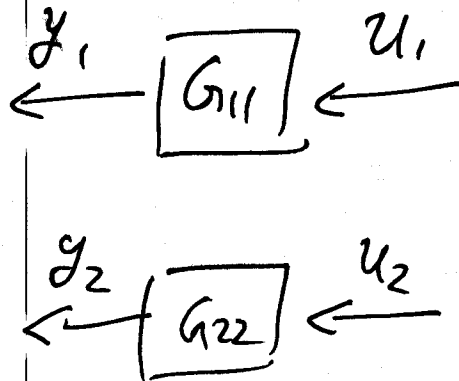
$$= \frac{C \text{Adj}(sI - A) B}{\det(sI - A)}$$

← numerator 2x2 polynomial matrix  
← characteristic polynomial

$$G_{tf} = tf(G); \quad \begin{pmatrix} \underline{G_{tf. num}} \\ \underline{G_{tf. den}} \end{pmatrix}$$

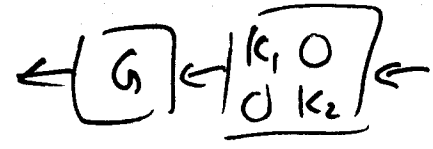


# Approximation for SISO design:



closed loop poles:  
pole (feedback  $(G \times K)$ )  
2x2

Use root locus + loop shaping to design  $K_1, K_2$  for  $G_{11}, G_{22}$ , respectively.

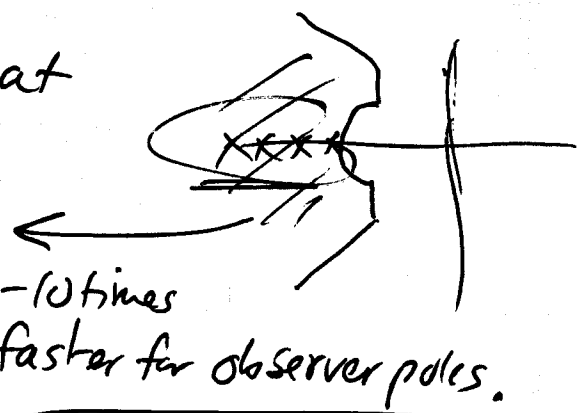


## Direct State Space Design (MIMO technique)

1. Full state feedback no need to use SISO approximation.

2. Observer design  
Find  $F$  such that  $\text{eig}(A - BF)$  are at desired locations.

Find  $L$  such that  $\text{eig}(A - LC)$  are at desired locations.



$$F = \text{place}(A, B, \text{desired poles})$$

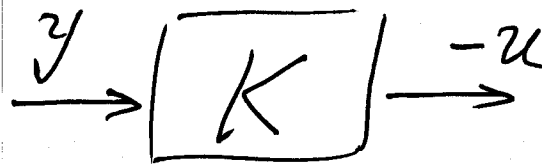
(as a vector)

$$L = \text{place}(A^T, C^T, \text{desired observer poles})^T$$

# Observer based controller:

$$\dot{\hat{x}} = A\hat{x} + Bu + L(y - C\hat{x})$$

estimated state  $\hat{x}$       output  $y$       estimated output  $C\hat{x}$   
 replica of plant      corrective gain  $L$   
 $u = -F\hat{x}$   
 replace state by estimated state.



Dynamic Controller (observer-based)  $\leftarrow$  LTI object

state space representation of  $K$ :

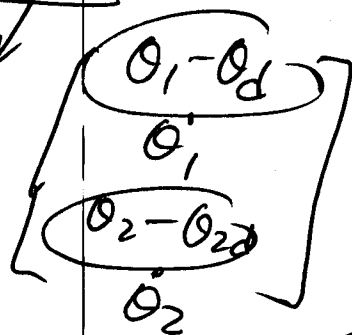
$$\begin{aligned} \dot{\hat{x}} &= \underline{A}\hat{x} - \underline{BF}\hat{x} + Ly - \underline{LC}\hat{x} \\ &= \underline{(A - BF - LC)}\hat{x} + \underline{L}y \\ -u &= \underline{F}\hat{x} \end{aligned}$$

$$K = ss(A - B * F - L * C, L, F, 0);$$

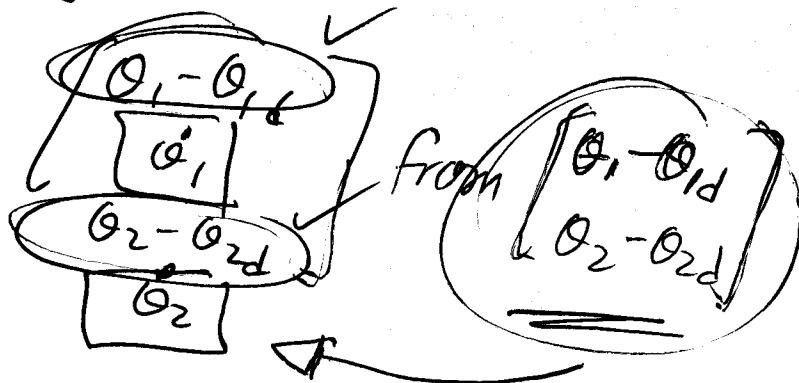
↑  
zeros(2, 2)

Full state feedback:

$$\underline{u = -F x} \rightarrow \dot{x} = (A - B F) x$$



Observer reconstructs



Alternatively:

Estimate  $\begin{bmatrix} \dot{\theta}_1 \\ \dot{\theta}_2 \end{bmatrix}$  by using washout filters

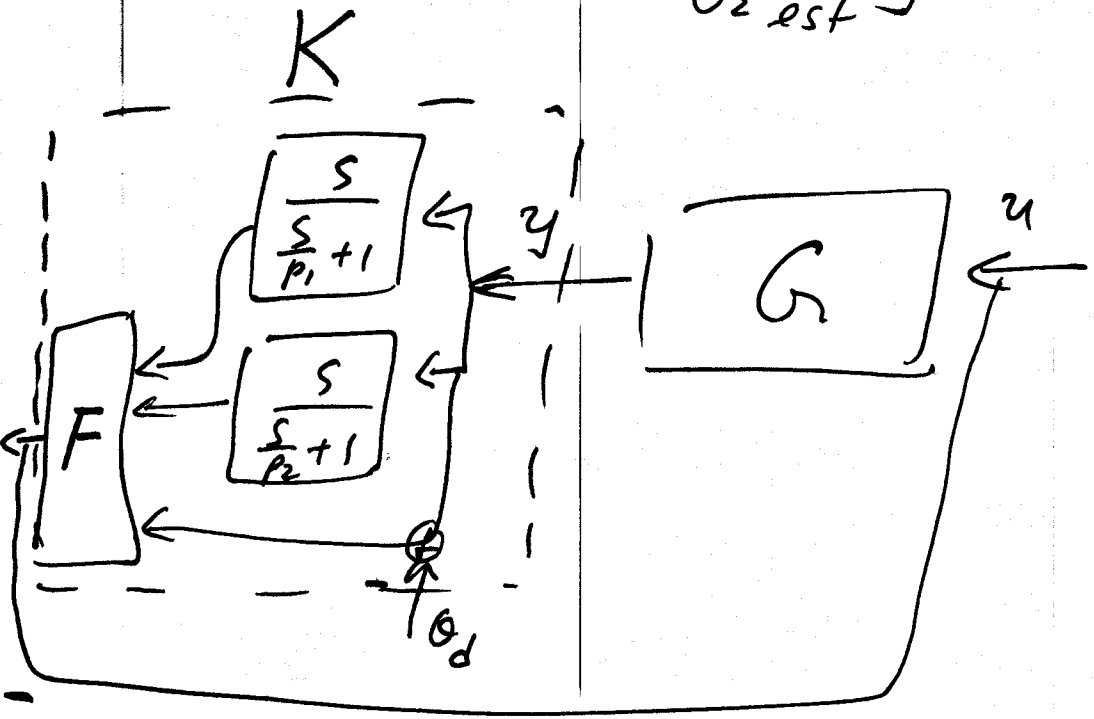
$$\dot{\theta}_{1\text{est}} = \frac{s}{\frac{s}{p_1} + 1} \theta_1$$

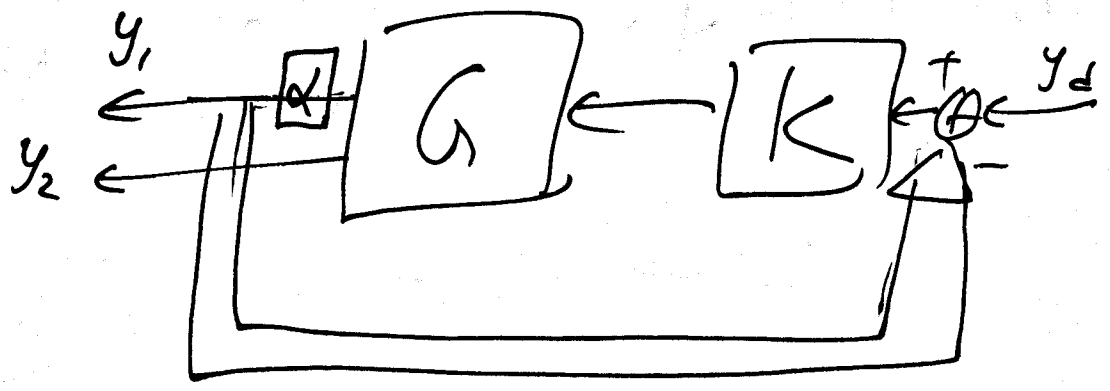
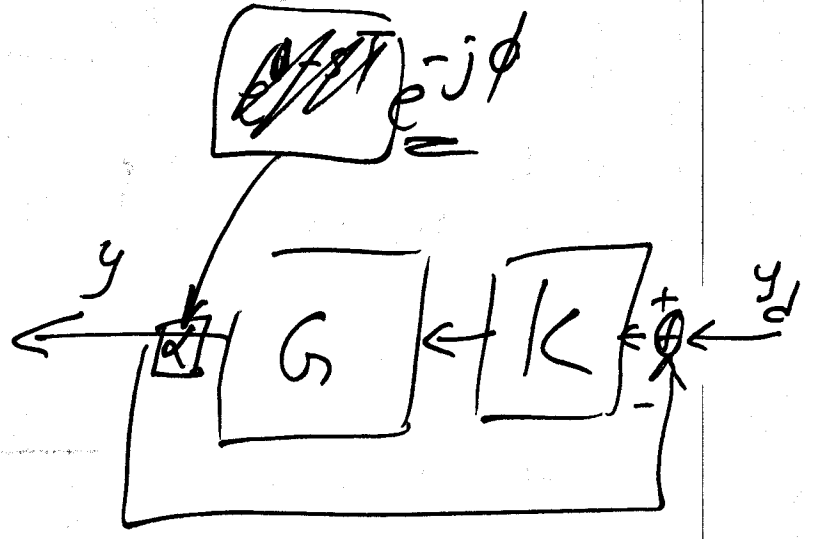
washout filter

$$\dot{\theta}_{2\text{est}} = \frac{s}{\frac{s}{p_2} + 1} \theta_2$$

Now apply the full state feedback:

$$u = -F \begin{bmatrix} \theta_1 - \theta_{1d} \\ \dot{\theta}_1 \text{ est} \\ \theta_2 - \theta_{2d} \\ \dot{\theta}_2 \text{ est} \end{bmatrix}$$





✓ loop gain

$$L_g = \underline{G} * \underline{K};$$

margin ( $L_g(1,1)$ )

margin ( $L_g(1,2)$ )

margin ( $L_g(2,2)$ )

margin ( $L_g(2,1)$ )

# Discretization of Observer

$$\hat{\dot{x}} = (A - L(C + DF) - BF)\hat{x} + Ly$$

$$u = -F\hat{x}$$

~~[Ad, Bd] c2d (A-L\*(C-DF)-B\*F, I);~~ 0.001

'zoh'  
'tustin'

$$\hat{x}_{k+1} = A_d \hat{x}_k + B_d y_k$$

$$u_k = -F \hat{x}_k$$

obsv = c2d(ss(A-L\*(C-DF)-B\*F, B, -F, 0));

~~[Ad, Bd, Cd, Dd] = ssdata(obsv);~~

$$K = ss(A - L(C + DF), L, F, 0);$$

$$\hat{x}_{k+1} = A_d \hat{x}_k + B_d y_k$$

$$u_k = C_d \hat{x}_k + D_d y_k$$

$K_d = c2d(K, 0.001);$