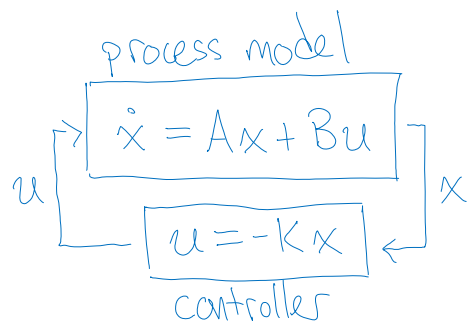


goal: synthesize stabilizing controllers for state-space systems

(a) state feedback

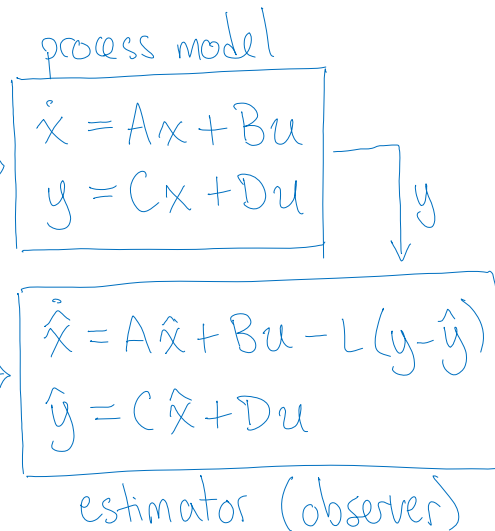


choose  $K$  s.t.  $\dot{\hat{x}} = (A - BK)x$   
stable, i.e.  $\text{Re } \lambda(A - BK) < 0$

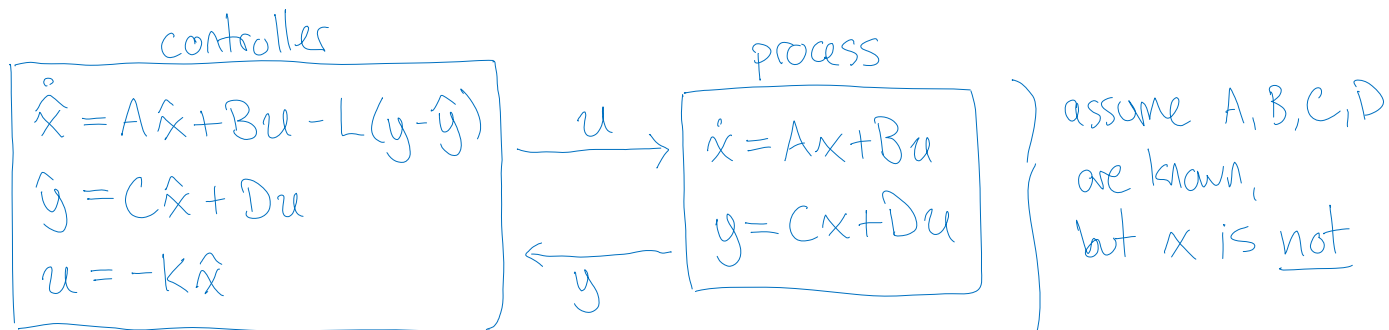
choose  $L$  s.t.  
 $\text{Re } \lambda(A + LC) < 0$ ,  
i.e. error dynamics

$(\dot{x} - \hat{x}) = (A + LC)(x - \hat{x})$   
are stable,  
so  $\hat{x} \rightarrow x$

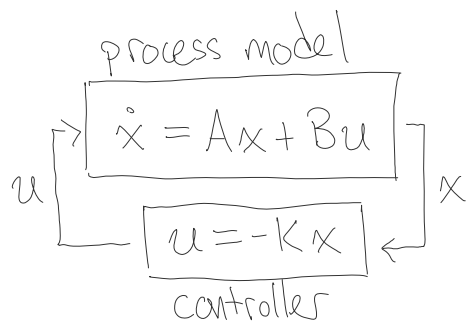
(b) state estimation



(c) stabilizing controller = (b) state estimation + (a) state feedback



(a) state feedback [AMv2 ch 7] [Nv7 ch 12.2]



• assume given: process model  $\dot{x} = Ax + Bu$   
→ for now, we'll assume we (i.e. our controller) gets to see the entire state vector  $x \in \mathbb{R}^n$

\* measuring all voltages/currents in a circuit  
positions/velocities in mechanical sys

goal: determine  $u$  given  $x$  so that  $x \rightarrow 0$  (i.e. closed-loop system stable)

\* if we choose  $u$  as a linear function of  $x$ ,  $u = -Kx$ ,  $K \in \mathbb{R}^{p \times n}$   
then the closed-loop system is linear:  $\dot{x} = Ax + Bu$   
 $= Ax - BKx = (A - BK)x$

→ we know how to assess stability:

closed-loop system  $\dot{x} = (A - BK)x$  is stable

$$x \rightarrow 0 \iff \operatorname{Re} \lambda(A - BK) < 0$$

all eigenvalues of  $A - BK$  have negative real part

\* our general approach: pole placement / eigenvalue assignment

– if we want the eigenvalues to be  $\lambda(A - BK) = \{\lambda_j\}_{j=1}^n \subset \mathbb{C}$ ,  
we just need to ensure characteristic polynomial

$$\det(sI - (A - BK)) = (s - \lambda_1) \cdot (s - \lambda_2) \cdots (s - \lambda_n) = \prod_{j=1}^n (s - \lambda_j)$$

1°. determine  $\det(sI - (A - BK))$  – symbolically or numerically

1°. determine  $\det(sI - (A - BK))$  — symbolically or numerically  
 $= s^n + a_1(K)s^{n-1} + a_2(K)s^{n-2} + \dots + a_n(K)$

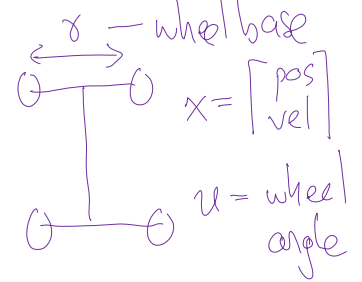
2°. expand  $\prod_{j=1}^n (s - \lambda_j)$  — symbolically or numerically  
 $= s^n + a_1^* s^{n-1} + a_2^* + \dots + a_n^*$

3°. solve  $a_1(K) = a_1^*, a_2(K) = a_2^*, \dots, a_n(K) = a_n^*$  for  $K \in \mathbb{R}^{p \times n}$   
 — symbolically or numerically

ex: (symbolically)  $A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, B = \begin{bmatrix} \gamma \\ 1 \end{bmatrix}$  [AMv2 Ex 7.4]

1°. determine  $\det(sI - (A - BK))$ ,  $K = [k_1, k_2]$

(recall:  $\det \begin{bmatrix} a & b \\ c & d \end{bmatrix} = ad - bc$ )



→  $\det(sI - (A - BK))$  — note:  $BK = \begin{bmatrix} \gamma \\ 1 \end{bmatrix} [k_1, k_2] = \begin{bmatrix} \gamma k_1 & \gamma k_2 \\ k_1 & k_2 \end{bmatrix}$   
 $= \det \begin{bmatrix} s + \gamma k_1 & -1 + \gamma k_1 \\ k_1 & s + k_2 \end{bmatrix} = (s + \gamma k_1) \cdot (s + k_2) - (\gamma k_1 - 1) \cdot k_1$   
 $= s^2 + (\gamma k_1 + k_2)s + k_1$

2°. expand  $\prod_{j=1}^n (s - \lambda_j)$ ,  $\lambda_{\pm} = -\sigma \pm j\omega$ ,  $\sigma > 0$   
 $= (s - (-\sigma - j\omega)) \cdot (s - (-\sigma + j\omega)) = s^2 + 2\sigma s + \sigma^2 + \omega^2$

3°. → solve  $a(K) = a^*$  to determine  $k_1, k_2$

$$* s^2 + (\gamma k_1 + k_2)s + k_1 = s^2 + 2\sigma s + \sigma^2 + \omega^2$$

$$\Leftrightarrow k_1^* = \sigma^2 + \omega^2 \quad \gamma k_1^* + k_2^* = 2\sigma, \text{ i.e. } k_2^* = 2\sigma - \gamma k_1^* = 2\sigma - \gamma(\sigma^2 + \omega^2)$$

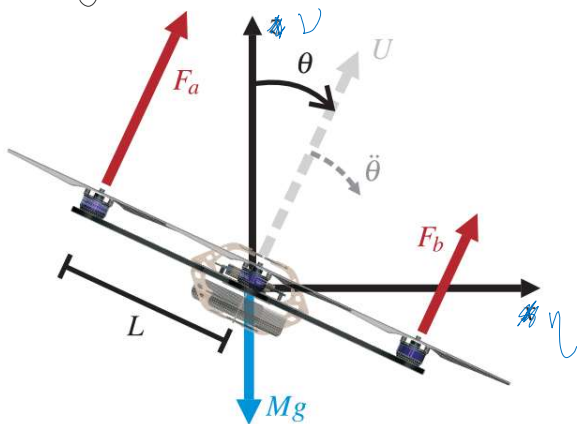
$$* u = -Kx, \quad K = [k_1^* \quad k_2^*] \Rightarrow \dot{x} = (A - BK)x \text{ is stable:}$$

$$\lambda(A - BK) = -\sigma \pm j\omega, \quad \sigma > 0$$

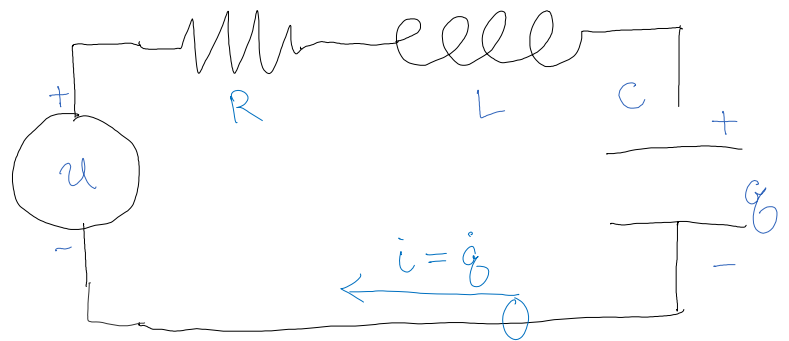
ex: (numerical)  $\rightarrow$  see lecture examples notebook § 6

(b) state estimation [AMv2 Ch 8] [Nv7 Ch 12.5]

ex: quadrotor



ex: RLC circuit



$\rightarrow$  what is the state  $x \in \mathbb{R}^n$ ?  $\rightarrow$  how would you measure each  $x_i$ ?  
(what is the sensor? how noisy is it? how expensive?)

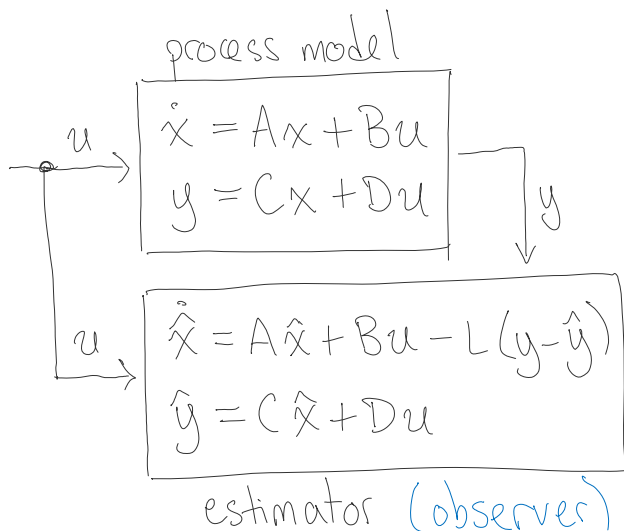
	quadrotor			circuit	
state:	positions	velocities		voltages	currents
sensor:	GPS	gyrometer		voltmeter	ammeter

sensor:	GPS	gyrometer	voltmeter	ammeter
	camera(s)	accelerometer	---	$E \frac{1}{2} M$
	LIDAR/RADAR			

\* different states are harder / more expensive to measure

\* it would be great if we could only measure a subset of states (e.g. positions, voltages) and estimate the rest (velocities, currents)

→ we will develop control system techniques for state estimation



•  $x \in \mathbb{R}^n$ ,  $u \in \mathbb{R}^p$ ,  $y \in \mathbb{R}^q$

\*  $q < n$ , i.e. we're measuring some (not all) of the states

ex: positions (not velocities)

$$x = \begin{bmatrix} p \\ \dot{p} \end{bmatrix}, \quad y = p = \begin{bmatrix} I & 0 \end{bmatrix} x + 0 \cdot u = Cx + \cancel{Du}^0$$

• given process model  $\dot{x} = Ax + Bu$  (i.e. given  $A, B, C, D$ )  
 $y = Cx + Du$

and assuming we know  $u: [0, \infty) \rightarrow \mathbb{R}^p$ ,  $y: [0, \infty) \rightarrow \mathbb{R}^q$   
 (b/c we choose  $u$ ) (b/c we measure  $y$ )

we construct another LTI system called an estimator (or observer):

$$\dot{\hat{x}} = A\hat{x} + Bu - L(y - \hat{y}) \quad \text{where } L \in \mathbb{R}^{n \times q} \text{ is an}$$

$$\begin{aligned}\dot{\hat{x}} &= A\hat{x} + Bu - L(y - \hat{y}) & \text{where } L \in \mathbb{R}^{n \times o} \text{ is an} \\ \hat{y} &= C\hat{x} + Du & \text{output error feedback matrix}\end{aligned}$$

→ to see why this works, determine the dynamics of  $e = x - \hat{x}$   
(your answer should be of the form  $\dot{e} = M \cdot e$ )

$$\begin{aligned}- e = x - \hat{x} &\Rightarrow \dot{e} = \dot{x} - \dot{\hat{x}} \\ &= (Ax + Bu) - (A\hat{x} + Bu - L(y - \hat{y})) \\ &= Ax - A\hat{x} + L(Cx - C\hat{x}) \\ &= A(\underbrace{x - \hat{x}}_{=e}) + LC(\underbrace{x - \hat{x}}_{=e}) = (A + LC)e\end{aligned}$$

\* so if  $L \in \mathbb{R}^{n \times o}$  is chosen such that  $\operatorname{Re} \lambda(A + LC) < 0$   
then  $\dot{e} = (A + LC)e$  is stable ! i.e.  $e = x - \hat{x} \rightarrow 0$  !

ex: (vehicle steering)  $\begin{aligned}\dot{x} &= Ax + Bu \\ y &= Cx\end{aligned}$   $A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, B = \begin{bmatrix} \delta \\ 1 \end{bmatrix}$   
 $C = [1 \ 0], D = 0$

→ determine state estimate error dynamics matrix  $A + LC$   
and the characteristic polynomial

$$- L \in \mathbb{R}^{2 \times 1}, L = \begin{bmatrix} l_1 \\ l_2 \end{bmatrix}, A + LC = \begin{bmatrix} l_1 & 1 \\ l_2 & 0 \end{bmatrix}$$

$$- \det(sI - (A+Lc)) = s^2 - l_1 s - l_2$$

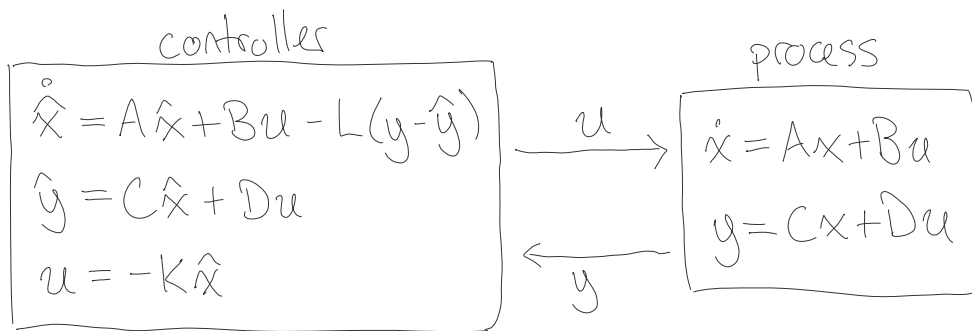
• we want to choose  $L$  (i.e.  $l_1, l_2$ ) s.t.  $\operatorname{Re} \lambda(A+Lc) < 0$

1°. characteristic polynomial  $\det(sI - (A+Lc)) = s^2 - l_1 s - l_2$

2°. want:  $(s + \zeta)^2 = s^2 + 2\zeta s + \zeta^2$ ,  $\zeta > 0$

3°. matching coefficients:  $l_2 = -\zeta^2$ ,  $l_1 = -2\zeta$

(c) stabilizing controller [AMv2 ch 8] [Nv7 ch 12.5]



\* assume  $A, B, C, D$  given and  $K, L$  are chosen such that:  
 $\operatorname{Re} \lambda(A - BK) < 0$  and  $\operatorname{Re} \lambda(A + LC) < 0$

→ already saw that  $e = x - \hat{x} \Rightarrow \dot{e} = (A + LC)e$   
 regardless of the input signal  $u$

→ determine dynamics of  $x$  when  $u = -K\hat{x}$

→ determine dynamics of  $x$  when  $u = -K\hat{x}$   
 (substitute to write  $\dot{x}$  in terms of  $x$  &  $e$ , not  $\hat{x}$ )

$$\begin{aligned} \dot{x} &= Ax + Bu = Ax - BK\hat{x} \\ &= Ax - BK(x - e) \quad \left. \begin{array}{l} e = x - \hat{x} \\ \text{so } \hat{x} = x - e \end{array} \right\} \\ &= (A - BK)x - BKe \end{aligned}$$

• with  $\bar{x} = \begin{bmatrix} x \\ e \end{bmatrix} \Rightarrow \dot{\bar{x}} = \begin{bmatrix} \dot{x} \\ \dot{e} \end{bmatrix} = \begin{bmatrix} A - BK & -BK \\ 0 & A + LC \end{bmatrix} \begin{bmatrix} x \\ e \end{bmatrix} = \bar{A} \bar{x}$

\* note:  $\det(sI - \bar{A}) = \det(sI - (A - BK)) \cdot \det(sI - (A + LC))$

→ so  $\operatorname{Re} \lambda(A - BK) < 0$  &  $\operatorname{Re} \lambda(A + LC) < 0 \Rightarrow \operatorname{Re} \lambda(\bar{A}) < 0$

$\Rightarrow x$  &  $e \rightarrow 0$  ✓ i.e. the combined state estimator  
 & state feedback controller  
 stabilizes both systems