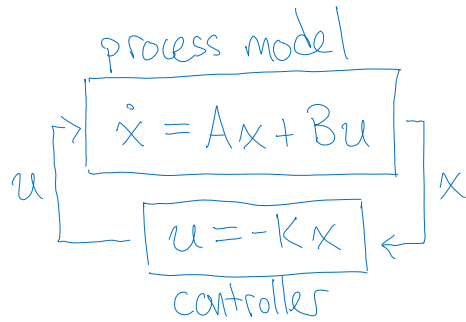


goal: synthesize stabilizing controllers for state-space systems

(a) state feedback

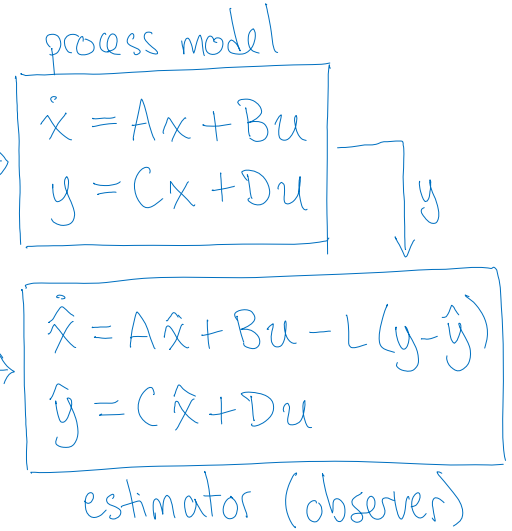


choose K s.t. $\dot{\hat{x}} = (A - BK)x$
stable, i.e. $\text{Re } \lambda(A - BK) < 0$

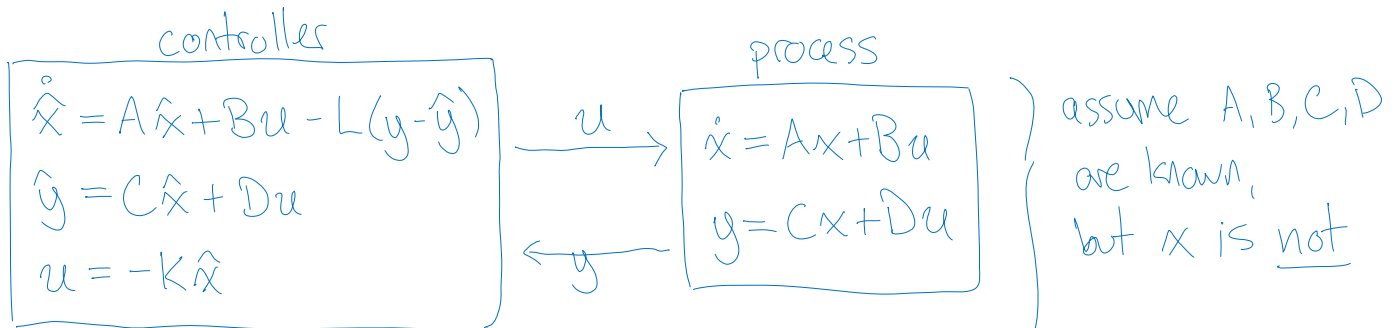
choose L s.t.
 $\text{Re } \lambda(A + LC) < 0$,
i.e. error dynamics

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

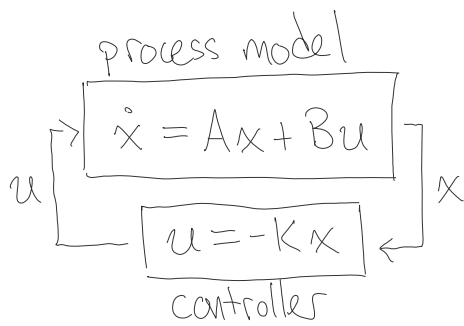
(b) state estimation



(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 / eigenvale 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
 $= s^n + a_1(K)s^{n-1} + a_2(K)s^{n-2} + \dots + a_n(K)$

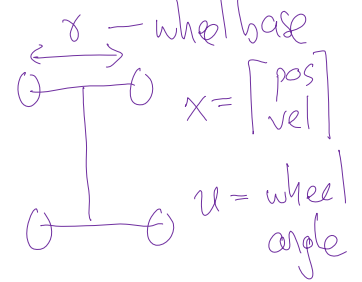
2° expand $\prod_{j=1}^n (s - \lambda_j)$ — symbolically or numerically
 $= s^n + a_1^* s^{n-1} + a_2^* + \dots + a_n^*$

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

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

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

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

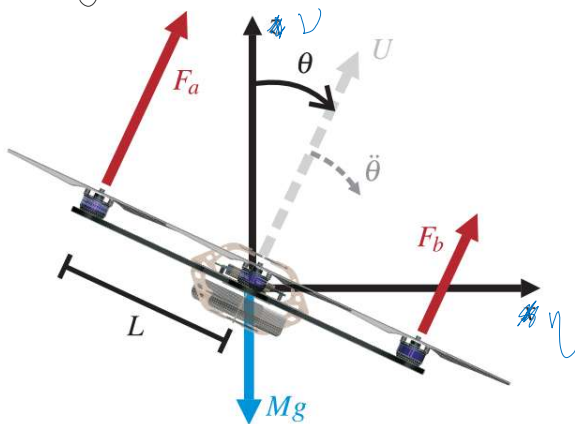
$$* u = -Kx, \quad K = [k_1^* \quad k_2^*] \Rightarrow \dot{x} = (A - BK)x \text{ is stable:}$$

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

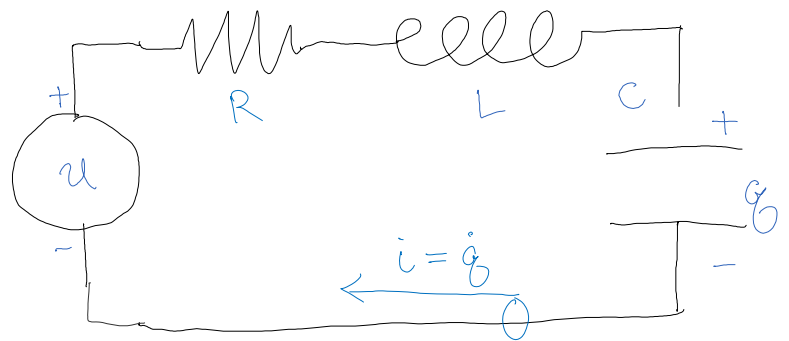
ex: (numerical) \rightarrow see lecture examples notebook § 6

(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?)

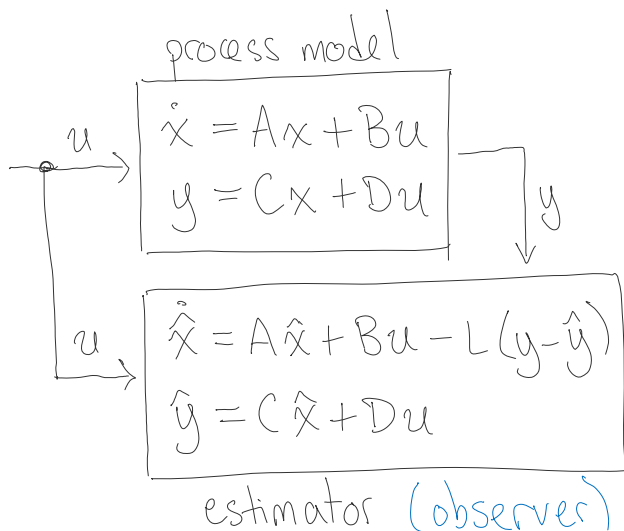
	quadrotor		circuit	
state:	positions	velocities	voltages	currents
sensor:	GPS	gyrometer	voltmeter	ammeter

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



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

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

ex: positions (not velocities)

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

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

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

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

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

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

→ 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

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

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

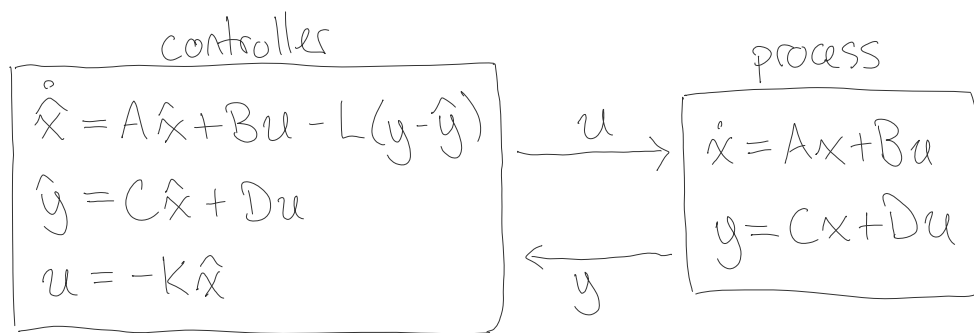
• we want to choose L (i.e. l_1, l_2) s.t. $\operatorname{Re} \lambda(A+Lc) < 0$

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

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

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

(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})

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

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

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

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

$\Rightarrow \underline{x \text{ \& } e} \rightarrow 0 \quad \checkmark$ i.e. the combined state estimator
& state feedback controller
stabilizes both systems