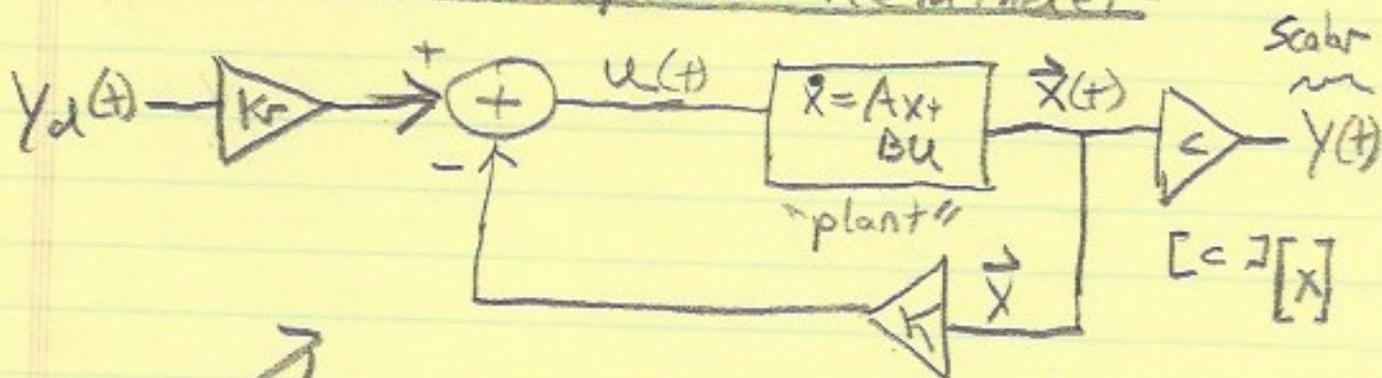


6.310 11/20/23

①

Today Observers in C.T. (D.T. after Holiday)

C.T. State Space Reminder



$$Kx \leftarrow [K][x]$$

scalar

Diagram

Equation

$$\dot{\hat{x}} = A\hat{x} + Bu \quad u = K_r y_d - K\hat{x}$$

$$\dot{\hat{x}} = (A - BK)\hat{x} + BK_r y$$

state
output

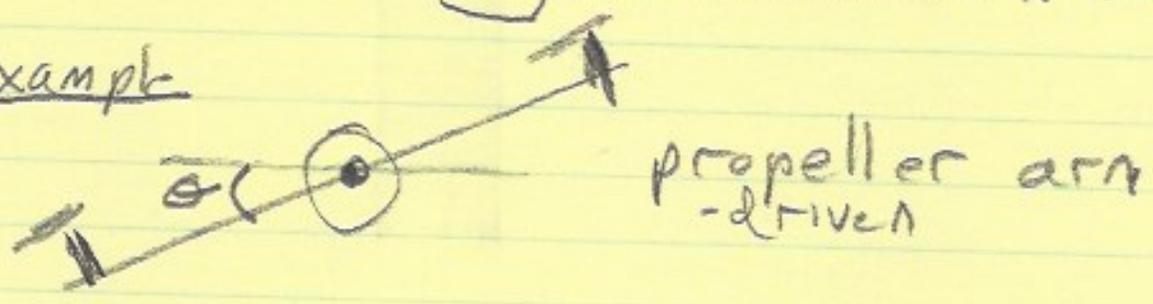
AND

$$y(t) = C\hat{x}(t)$$

Ex: Arm-angle, umbrella position

What if x is not measurable (or some of x isn't) ②

Example



We can measure θ

We can approximate $\omega = \frac{\theta(t) - \theta(t-\Delta T)}{\Delta T}$

We can not measure:

Differential Propeller speed

Left speed - Right Prop Speed (proportional to Torque)

Set up an Estimator

original system:

$$\dot{x} = Ax + Bu$$

Estimation system
(computed using simulation)

$$\dot{\hat{x}} = A\hat{x} + Bu$$

↑
estimate state

We generate u so we know it. same for both systems

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Use estimated (e.g. simulated) state for control

$$u(t) = \underbrace{K_r y_d(t)}_{\text{KNOWN}} - K \hat{x}$$

↑
estimated state

"Obvious Problems"

1) is $x(t)|_{t=0} = \hat{x}(t)|_{t=0}$?
(Do initial conds match)

2) Must simulate ^{Estimation} a system as part of control

Examine initial condition issue

Define $e_{err}(t) = x(t) - \hat{x}(t)$

- (1) $\frac{d}{dt}x(t) = Ax(t) + Bu(t)$
- (2) $\frac{d}{dt}\hat{x}(t) = A\hat{x}(t) + Bu(t)$

Subtracting (2) from (1) (Assuming A, B model match exactly)

$$\frac{d}{dt}(\underbrace{x(t) - \hat{x}(t)}_{e_{err}(t)}) = A(\underbrace{x(t) - \hat{x}(t)}_{e_{err}(t)}) + \cancel{Bu(t)} - \cancel{Bu(t)}$$

$$\frac{d}{dt}e_{err}(t) = Ae_{err}(t)$$

Correct $\hat{x}(t)$ using $y(t) = C\hat{x}(t)$

(5)

(1) Physical system $\dot{x} = Ax + Bu$

(2) Estimation system $\dot{\hat{x}} = A\hat{x} + \underbrace{Bu}_{\substack{\text{We} \\ \text{know}}} + [L](y - \hat{y})$

$y = Cx$ $\hat{y} = C\hat{x}$

Recall estimation error

$$e_{err}(t) \equiv x(t) - \hat{x}(t)$$

(1) - (2)

$$\begin{aligned} \dot{e}_{err}(t) &= A e_{err}(t) - L(y(t) - \hat{y}(t)) \\ &= A e_{err}(t) - L(Cx(t) - C\hat{x}(t)) \\ &= A e_{err}(t) - LC(e_{err}(t)) \\ &= (A - LC) e_{err}(t) \end{aligned}$$

Pick L so that $e_{err}(t) \rightarrow 0$ fast

Note

$$\begin{array}{ccc} \begin{array}{c} \xrightarrow{n} \\ [K] \\ \xleftarrow{n} \\ [B] \\ \xleftarrow{n} \\ (A - BK) \end{array} & \xleftrightarrow{\text{Transpose}} & \begin{array}{c} \xleftarrow{n} \\ [L] \\ \xleftarrow{n} \\ [C] \\ \xleftarrow{n} \\ (A - LC) \end{array} \end{array}$$

$K = \text{place}(A, B, \text{poles})$
 $K = \text{lqr}(A, B, \text{weights})$

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

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LQR $L = \text{lqr}(A^T, C^T, \text{~~Q~~, R)^T$

what's being minimized?

Pole Placement

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

$\text{eig}(A-LC) = \text{eig}((A-LC)^T)$

$= \text{eig}(A^T - C^T L^T)$

So place works the usual way

Estimator given L & K

$u = K_r y_d - K \hat{x}$

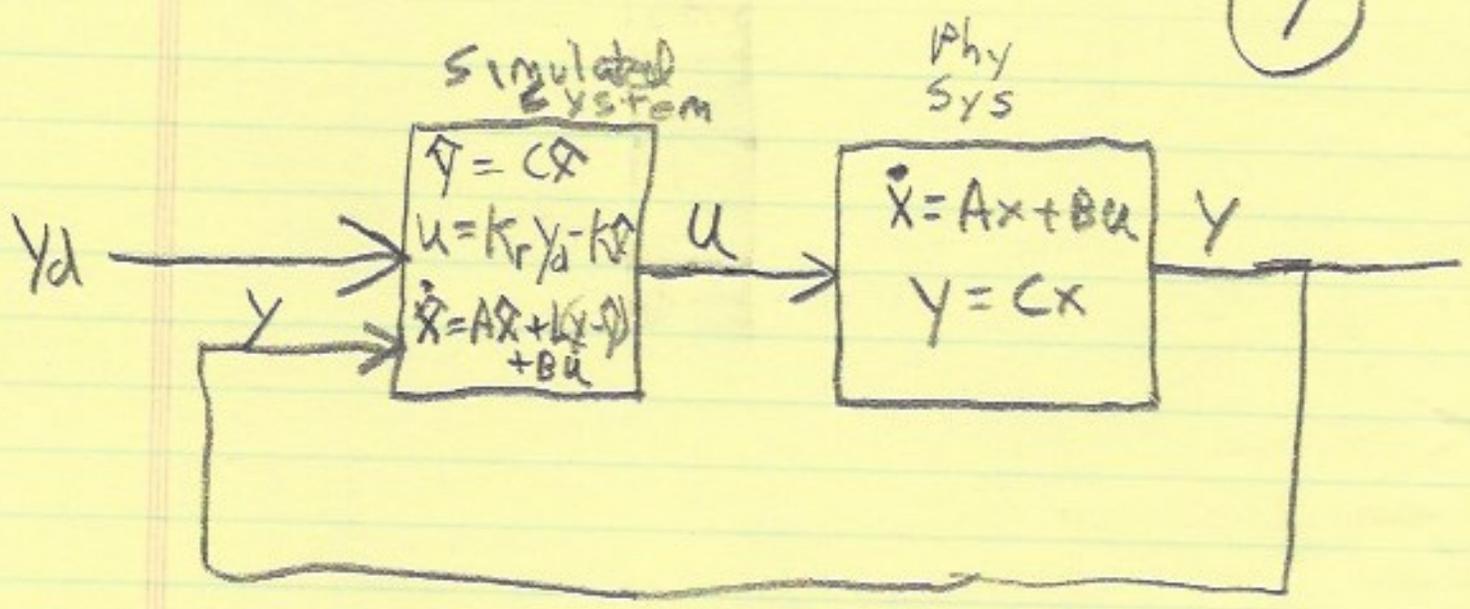
Est

$\hat{x} = (A - BK)\hat{x} + K_r y_d + L(y - \hat{y})$

Phys

$\hat{x} = (A - BK - LC)\hat{x} + K_r y_d + Ly$

Hmmm...



Analysis

(1) Phys $\dot{x} = Ax + Bu = Ax + B(K_r y_d - K\hat{x})$

(2) Est $\dot{\hat{x}} = (A\hat{x} + B\hat{u} + L(y - \hat{y})) + BK_r y_d$

error $e(t) \equiv x(t) - \hat{x}(t)$ $\hat{x}(t) = x(t) - e(t)$

(1)-(2) $\dot{e}(t) = (A - LC)e(t)$

$$\dot{x} = Ax + BK_r y_d - BK(x(t) - e(t))$$

$$= (A - BK)x + BK e(t) + BK_r y_d$$

$$\begin{bmatrix} \dot{x} \\ \dot{err} \end{bmatrix} = \begin{bmatrix} A - BK & BK \\ 0 & A - LC \end{bmatrix} \begin{bmatrix} x \\ err \end{bmatrix} + \begin{bmatrix} BK_r \\ 0 \end{bmatrix} y_d$$

decouples control from estimator!