

Day 25

Optimal observers

AE 353

Spring 2022

Bretl

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

$$y = Cx$$

$\leftarrow n_x$ number of states
 n_u number of inputs
 n_y number of outputs

OPTIMAL CONTROLLER

$$u = -Kx \quad \text{where} \quad K = \text{lqr}(A, B, Q_c, R_c)$$

diagonal w/ positive numbers $\left\{ \begin{array}{l} Q_c \text{ is } n_x \times n_x \quad \leftarrow \text{error} \\ R_c \text{ is } n_u \times n_u \quad \leftarrow \text{effort} \end{array} \right.$

OPTIMAL OBSERVER

$$\dot{\hat{x}} = A\hat{x} + Bu - L(C\hat{x} - y) \quad \text{where} \quad L = \text{lqr}(A^T, C^T, R_o^{-1}, Q_o^{-1})^T$$

diagonal w/ positive numbers $\left\{ \begin{array}{l} Q_o \text{ is } n_y \times n_y \quad \leftarrow \text{sensors} \\ R_o \text{ is } n_x \times n_x \quad \leftarrow \text{dynamics} \end{array} \right.$

height of quadrator

$$y = x$$

$$y = 1$$

$$\hat{x} = 1$$

$$y_1 = x + n_1$$

$$y_2 = x + n_2$$

$$y_1 = 1$$

$$y_2 = 2$$

$$\hat{x} = \cancel{1.5}$$

$$\cancel{2.0}$$



1.8?

1.9?

\hat{x}	n_1	n_2
1.5	-0.5	0.5
2.0	-1.0	0.0
1.8	-0.8	0.2

minimize
 x

$$g_1(y_1 - x)^2 + g_2(y_2 - x)^2$$

minimize $u[t_0, \infty)$

$$\int_{t_0}^{\infty} \left(\underbrace{x(t)^T Q_c x(t)}_{\text{error}} + \underbrace{u(t)^T R_c u(t)}_{\text{effort}} \right) dt$$

$$u = -Kx$$

subject to

$$\begin{aligned} \dot{x}(t) &= Ax(t) + Bu(t) \quad \text{for } t \in [t_0, \infty) \\ x(t_0) &= x_0 \end{aligned}$$

OPTIMAL
CONTROLLER

minimize

$x(t_1), n(-\infty, t_1], d(-\infty, t_1]$

$$\int_{-\infty}^{t_1} \left(\underbrace{n(t)^T Q_o n(t)}_{\text{sensor noise}} + \underbrace{d(t)^T R_o d(t)}_{\text{process disturbance}} \right) dt$$

OPTIMAL
OBSERVER

subject to

$$\left. \begin{aligned} \dot{x}(t) &= Ax(t) + Bu(t) + d(t) \\ y(t) &= Cx(t) + n(t) \end{aligned} \right\} \text{for } t \in (-\infty, t_1]$$

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

$$d(t) = \dot{x}(t) - (Ax(t) + Bu(t))$$

$$-n(t) = Cx(t) - y(t)$$

$\left. \begin{aligned} \text{if } t_1 = t_\alpha \text{ then } x(t_1) = \alpha \\ \text{if } t_1 = t_\beta \text{ then } x(t_1) = \beta \end{aligned} \right\}$

integrate this starting at

$$\hat{x}(t_\alpha) = \alpha$$

then the solution is $\hat{x}(t_\beta) = \beta$