

Observers

Consider

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

$$x \in \mathbb{R}^n, u \in \mathbb{R}^m$$

$$y = Cx$$

$$y \in \mathbb{R}^p$$

If $x(t)$ is available, we can design $u = Kx$ to solve ex. pole placement.

However, if only $y(t) = Cx(t)$ is available, what can we do?

1. use $u = Ly$ output feedback
 $= LCx$

2. estimate first $x(t)$ (based on $y(t)$) (denotes the estimation), then use $u = K\hat{x}$.

a. how to estimate $x(t)$? Assume (C, A) is observable.

- We estimate $x(0) \Rightarrow \hat{x}(t)$ X
- non robust

b. We estimate $x(t)$ online \Rightarrow
the best we can have is $\hat{x}(t) \rightarrow x(t)$
as $t \rightarrow \infty$.

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

- observer we can just assume $\hat{x}(0) = 0$

our goal is to design $L_{n \times p}$ such that

$$\tilde{x}(t) = x(t) - \hat{x}(t) \rightarrow 0 \text{ as } t \rightarrow \infty,$$

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

\Rightarrow we need to find L st. $A - LC$ is a stable matrix!

since $(A - LC)^T = A^T - C^T L^T$ has the same eigenvalues of $A - LC$,

$\Rightarrow \exists L$ st. $A^T - C^T L^T$ is a stable matrix?

If we let $\bar{A} = A^T$, $\bar{B} = C^T$, $\bar{K} = -L^T \Rightarrow$
 $A^T - C^T L^T = \bar{A} + \bar{B} \bar{K}!$

solvable if (A^T, C^T) is reachable, i.e.

$[C^T \ A^T C^T \ \dots \ (A^T)^{n-1} C^T]$ has full rank

i.e. $\Omega = \begin{bmatrix} C \\ CA \\ \vdots \\ CA^{n-1} \end{bmatrix}$ has full rank

$\Leftrightarrow (C, A)$ is observable!

then we can use $u = K\hat{x}$ as control.

$$\Rightarrow \begin{cases} \dot{x} = Ax + BK\hat{x} & (\dot{\hat{x}} = A\hat{x} + BKx) \\ \dot{\hat{x}} = A\hat{x} + BK\hat{x} + L(Cx - C\hat{x}) \end{cases}$$

We use x , $\tilde{x} = x - \hat{x}$ as the new coordinates

$$\begin{aligned} \dot{\hat{x}} &= Ax + BK(x - \tilde{x}) \\ &= (A + BK)x - BK\tilde{x} \end{aligned}$$

$$\dot{\tilde{x}} = (A - LC)\tilde{x}$$

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

eigenvalues are eigenvalues of $A + BK$ and
eigenvalues of $A - LC$

— Principle of separation: design
state feedback and observer independent.

Linear-Quadratic optimal control

Consider

$$\begin{aligned}\dot{x} &= Ax + Bu & x \in \mathbb{R}^n \\ x(t_0) &= x_0 & u \in \mathbb{R}^m\end{aligned}$$

Find the control on the time interval $[t_0, t_1]$ such that the following functional (Cost) is minimized

$$J(u) = \int_{t_0}^{t_1} [x(t)^T Q x(t) + u(t)^T R u(t)] dt + x(t_1)^T S x(t_1)$$

$$Q \geq 0, \quad R > 0, \quad S \geq 0.$$

$$\text{Let } V(x, t) = x^T P(t) x$$

Ansatz and we conjecture V is the optimal cost for $t_0 = t$, $t \in [t_0, t_1]$, and $x(t_0) = x$,

$$\begin{aligned}\frac{d}{dt} V(x(t), t) &= \dot{x}(t)^T P(t) x + x^T(t) \dot{P} x + x^T(t) P(t) \dot{x}(t) \\ &= (Ax + Bu)^T P x + x^T \dot{P} x + x^T P (Ax + Bu) \\ &= x^T (\dot{A} P + PA + \dot{P}) x + u^T B^T P x + x^T P B u\end{aligned}$$

\Rightarrow

$$V(x(t_1), t_1) - V(x_0, t_0)$$

$$= \int_{t_0}^{t_1} [x^T(t) (\dot{A}^T P + PA + \dot{P}) x(t) + u^T(t) B^T P x(t) + x^T(t) P B u(t)] dt$$

$$J(u) - V(x_0, t_0)$$

$$= \int_{t_0}^{t_1} [x^T(t) (\dot{A}^T P + PA + Q + \dot{P}) x + u^T R u + u^T B^T P x + x^T P B u] dt$$

$$x^T(t_1) S x(t_1) - x^T(t_0) P(t_0) x(t_0) \geq 0$$

$$P(t_1) = S$$

$$\left(\begin{aligned} u^T u + u^T B^T P x + x^T P B u &= r (u^2 + u^T B^T P x + x^T P B u) \\ &= r (u + B^T P x)^2 - x^T P B R^{-1} B^T P x \end{aligned} \right)$$

$$= \int_{t_0}^{t_1} [x^T(t) (\dot{A}^T P + PA + Q - P B R^{-1} B^T P) x(t) + (u + R^{-1} B^T P x)^T R (u + R^{-1} B^T P x)] dt$$

$$\Rightarrow \begin{cases} \dot{P} = -\dot{A}^T P - PA + P B R^{-1} B^T P - Q \\ P(t_1) = S \end{cases} \quad \text{- Dynamical Riccati equations}$$

$$\Rightarrow J(u) - V(x_0, t_0) = \int_{t_0}^{t_1} (u + R^{-1} B^T P x)^T R (u + R^{-1} B^T P x) dt$$

$$\geq 0$$

Furthermore $J(u) = V(x_0, t_0) = x_0^T P(t_0) x_0$

if $u = -R^{-1} B^T P x$

$\Rightarrow u = -R^{-1} B^T P(t) x(t)$ is the optimal control

and $x_0^T P(t_0) x_0$ is the optimal cost!