State estimation

Observabil

observer



Linear time-invariant processes: Estimation CHEM-E7190 (was E7140), 2023

Francesco Corona

Chemical and Metallurgical Engineering School of Chemical Engineering

State estimation

bservability

Asymptotic

State-estimation

LTI systems - Estimation

State estimation

State estimation

Observabili

Asymptoti

$$x(t) = e^{A(t-t_0)}x(t_0) + \int_{t_0}^t e^{A(t-\tau)}Bu(\tau)d\tau$$

From the forced response, we saw that controllability must depend only on A and B

We made some simplifying assumptions that helped us focusing on control

• We stated that the assumptions have no implications

We assumed that we can measure all state variables (and no feedthrough)

• That is, we have C = I and D = 0

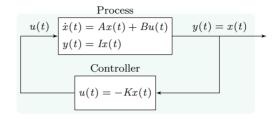
$$\underbrace{\begin{bmatrix} y_1(t) \\ y_2(t) \\ \vdots \\ y_{N_y}(t) \end{bmatrix}}_{y(t)} = \underbrace{\begin{bmatrix} 1 & 0 & \cdots & 0 \\ 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & 1 \end{bmatrix}}_{C=I} \underbrace{\begin{bmatrix} x_1(t) \\ x_2(t) \\ \vdots \\ x_{N_x}(t) \end{bmatrix}}_{x(t)} + \underbrace{\begin{bmatrix} 0 & 0 & \cdots & 0 \\ 0 & 0 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & 0 \end{bmatrix}}_{D=0} \underbrace{\begin{bmatrix} u_1(t) \\ u_2(t) \\ \vdots \\ u_{N_u}(t) \end{bmatrix}}_{u(t)}$$

State estimation (cont.)

State estimation

Observabili

observer



For linear time-invariant system, function h(x(t)) = -Kx(t) is optimal, in some sense

$$\underbrace{\begin{bmatrix} u_{1}(t) \\ u_{2}(t) \\ \vdots \\ u_{N_{u}}(t) \end{bmatrix}}_{u(t)} = - \underbrace{\begin{bmatrix} k_{11} & k_{12} & \cdots & k_{1,N_{x}} \\ k_{21} & k_{22} & \cdots & k_{2,N_{x}} \\ \vdots & \vdots & \ddots & \vdots \\ k_{N_{u},1} & k_{N_{u},2} & \cdots & k_{N_{u},N_{x}} \end{bmatrix}}_{K} \underbrace{\begin{bmatrix} x_{1}(t) \\ x_{2}(t) \\ \vdots \\ x_{N_{x}}(t) \end{bmatrix}}_{x(t)}$$

- For a system in state x(t), the optimal control action is u(t) = -Kx(t)
- (We will briefly also discuss the underlaying optimality criterion)

State estimation (cont.))

State estimation

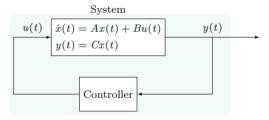
Observabili

observer

We assumed perfect measurement (observation) variables y(t) (from system's sensors)

• We assumed that y(t) returns all state variables x(t), or y(t) = Ix(t)

Not always the case, what if we cannot measure the state of the system, $x(t) \neq y(t)$?



How to design controllers that define an optimal control u(t), when x(t) is unknown?

State estimation

Observabilit

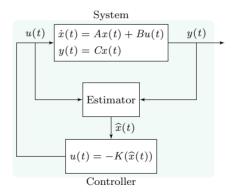
Asymptoti

State estimation (cont.)

We could design another device capable to provide an estimate the state of the system

- Based on the measurable quantities of the system (data)
- That is, measurements y(t) and inputs u(t)

A device that approximates the state of the system is a state observer, or estimator



Were the state estimate $\widehat{x}(t)$ accurate, we could use it with the optimal controller (-K)



State estimation (cont.)

State estimation

Observabilit

observer

The state estimation problem has a solution only for those systems that are observable

Definition

One, formal, definition of observability

A linear and time-invariant system with state equation $\dot{x}(t) = Ax(t)$ and measurement equation y(t) = Cx(t) is said to be observable if and only if it is possible to determine its state x(t) from the force-free response of its measurements over a finite time $(t_f < \infty)$, from any arbitrary initial state $x(t_0)$

state estimatio

Asymptotic

Observability

LTI systems - Estimation

Observability

State estimation

Observabili

observer

Observability refers to the possibility for estimate the system state from measurements

Observability for linear and time-invariant systems depends only on the pair (A, C)

$$\begin{cases} \dot{x}(t) = Ax(t) \\ y(t) = Cx(t) \end{cases}$$

We present a formal definition of observability for linear time-invariant systems

- Necessary and sufficient conditions
- (Invariance under similarity)

Observability

State estimation

Observabili

observer

We analyse the observabilty of a linear time-invariant system by using three criteria

- Observability gramian
- Observability matrix
- (Popov-Belevich-Hautus test)

As with controllability, all these criteria are complementary as for their usefulness

State estimation

Observabili

observer

Definition

Observability gramian

Consider the linear and time-invariant system (A, C), with $x(t) \in \mathcal{R}^{N_x}$ and $y(t) \in \mathcal{R}^{N_y}$

$$\begin{cases} \dot{x}(t) = Ax(t) \\ y(t) = Cx(t) \end{cases}$$

The system's observability gramian is a $(N_x \times N_x)$ matrix, real and symmetric

$$W_o(t) = \int_0^t e^{A^T \tau} C^T C e^{A\tau} d\tau$$

State estimation

Observabili

observer

Theorem

Observability test (I)

Consider the linear and time-invariant system (A, C), with $x(t) \in \mathcal{R}^{N_x}$ and $y(t) \in \mathcal{R}^{N_y}$

$$\begin{cases} \dot{x}(t) = Ax(t) \\ y(t) = Cx(t) \end{cases}$$

Let $W_o(t) = \int_0^t e^{A^T \tau} C^T C e^{A\tau} d\tau$ be the observability gramian of the system

• The system is observable iff $W_o(t)$ is non-singular, for all t>0

State estimation

Observability

observer

Observability (cont.)

Proof (Sufficient condition)

From the second Lagrangian equation, we have the force-free evolution of the output

$$y(\tau) = Ce^{A\tau}x(0)$$

We left-multiply the equation by $e^{A^T\tau}$, then we integrate between 0 and some t_f

$$\int_0^{t_f} e^{A^T \tau} y(\tau) d\tau = \int_0^{t_f} e^{A^T \tau} C e^{A \tau} x(0) d\tau$$
$$= W_o(t_f) x(0)$$

Thus, we have

$$x(0) = W_o^{-1}(tf) \int_0^{t_f} e^{A^T \tau} Cy(\tau) d\tau$$

The initial state is given as a function of the inverse of the observability gramian $W_o(tf)$ and the integral $\int_0^{t_f} e^{A^T \tau} C e^{A\tau} y(\tau) d\tau$ which can be computed from measurements $y(\tau)$

- The observability gramian need be non-singular
- Needed for the inverse to exist

State estimation

Observabilit

observer

Example

Consider the linear and time-invariant system (A, B), with $x(t) \in \mathbb{R}^2$ and $u(t) \in \mathbb{R}$

$$\begin{cases} \dot{x}(t) = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} x(t) \\ y(t) = \begin{bmatrix} 1 & 0 \end{bmatrix} x(t) \end{cases}$$

We are interested in verifying the observability of the system and initial state x(0)

- We have measured the output, $y(\tau) = 1 + 2\tau$
- Over a finite interval, $\tau \in [0, 1]$

We compute the observability gramian, then we determine whether its invertible

Observability (cont.)

State estimation

Observabili

observer

To compute the observability gramian, we already computed the state transition matrix

$$e^{A\tau} = \begin{bmatrix} 1 & \tau \\ 0 & 1 \end{bmatrix}$$

We can compute the observability gramian of the system, by applying the definition

$$W_o(t) = \int_0^t e^{A^T \tau} C^T C e^{A\tau} d\tau$$

$$= \int_0^t \begin{bmatrix} 1 & 0 \\ \tau & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} \begin{bmatrix} 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & \tau \\ 0 & 1 \end{bmatrix} d\tau$$

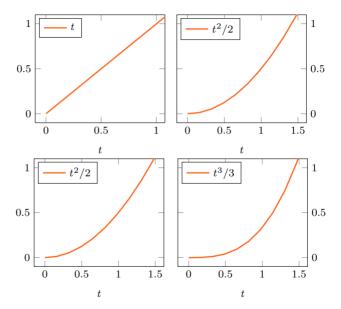
$$= \begin{bmatrix} t & t^2/2 \\ t^2/2 & t^3/3 \end{bmatrix}$$

State estimat

Observability

Asymptoti

Observability (cont.)



Observability (cont.)

State estimation

Observabili

observer

$$W_o(t) = \begin{bmatrix} t & t^2/2 \\ t^2/2 & t^3/3 \end{bmatrix}$$

To verify the singularity of the observability gramian $W_o(t)$, we check the determinant

- We need to check whether it is zero or it is non-zero
- Whatever the value of t (that is, at any time)

$$\det(W_o(t)) = t^4/12 > 0 \quad (\forall t > 0)$$

Since det $(W_o(t)) \neq 0$ for all t > 0, we can conclude that the system is observable

Observability (cont.)

State estimation

observer

$$x(0) = W_o^{-1}(tf) \int_0^{t_f} e^{A^T \tau} Cy(\tau) d\tau$$

For the observability gramian and its inverse at the final time $t_f = 1$, we have

$$W_o(1) = \begin{bmatrix} t & t^2/2 \\ t^2/2 & t^3/3 \end{bmatrix}$$
$$= \begin{bmatrix} 1 & 1/2 \\ 1/2 & 1/3 \end{bmatrix}$$

Then,

$$W_o^{-1}(1) = \begin{bmatrix} 4 & -6 \\ -6 & 12 \end{bmatrix}$$

We also have,

$$\int_0^1 e^{A^T \tau} C^T y(\tau) d\tau = \int_0^1 \begin{bmatrix} 1 & 0 \\ \tau & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} (1 + 2\tau) d\tau$$
$$= \begin{bmatrix} 2 \\ 7/6 \end{bmatrix}$$

State estimation

Observabili

observer

After substituting, we have the initial state

$$x(0) = W_o^{-1}(1) \int_0^1 e^{A^T \tau} C^T y(\tau) d\tau$$
$$= \begin{bmatrix} 4 & 6 \\ -6 & 12 \end{bmatrix} \begin{bmatrix} 2 \\ 7/6 \end{bmatrix}$$
$$= \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

The trajectory of the state variable from the initial state x(0),

$$x(t) = e^{At}x(0) + \int_0^t e^{A(t-\tau)}Bu(\tau)d\tau$$

State estimation

Asymptot: observer

Observability (cont.)

Theorem

Observability matrix and observability test (II)

Consider a linear and time-invariant system (A, C), with $x(t) \in \mathbb{R}^{N_x}$ and $y(t) \in \mathbb{R}^{N_y}$

$$\begin{cases} \dot{x}(t) = Ax(t) \\ y(t) = Cx(t) \end{cases}$$

We define the $(N_x \times (N_y \times N_x))$ observability matrix

$$\mathcal{O} = egin{bmatrix} C \ CA \ CA^2 \ dots \ CA^{N_x-1} \end{bmatrix}$$

Necessary and sufficient condition for observability,

$$rank(\mathcal{O}) = N_x$$

State estimation

Observabilit

observer

Example

Consider the linear and time-invariant system (A, C), with $x(t) \in \mathbb{R}^{N_x}$ and $y(t) \in \mathbb{R}^{N_y}$

$$\begin{cases} \dot{x}(t) = \begin{bmatrix} 2 & 4 & 0.5 \\ 0 & 4 & 0.5 \\ 0 & 0 & 2 \end{bmatrix} x(t) \\ y(t) = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 3 \end{bmatrix} x(t) \end{cases}$$

We are interested in verifying its observability using the observability matrix

The observability matrix has size $((N_y = 2 \times N_x = 3) \times N_x = 3) = (6 \times 3)$

$$\mathcal{O} = \begin{bmatrix} C \\ CA \\ CA^2 \end{bmatrix}$$

Observability (cont.)

$$\mathcal{O} = \begin{bmatrix} C \\ CA \\ CA^2 \end{bmatrix}$$

We know C, we need to compute CA and CA^2 ,

$$CA = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 3 \end{bmatrix} \begin{bmatrix} 2 & 4 & 0.5 \\ 0 & 4 & 0.5 \\ 0 & 0 & 2 \end{bmatrix}$$
$$= \begin{bmatrix} 2 & 4 & 0.5 \\ 0 & 0 & 6 \end{bmatrix}$$

$$CA^{2} = (CA)A$$

$$= \begin{bmatrix} 2 & 4 & 0.5 \\ 0 & 0 & 6 \end{bmatrix} \begin{bmatrix} 2 & 4 & 0.5 \\ 0 & 4 & 0.5 \\ 0 & 0 & 2 \end{bmatrix}$$

$$= \begin{bmatrix} 4 & 24 & 4 \\ 0 & 0 & 12 \end{bmatrix}$$

State estimation

Observabilit

Asymptot

Thus, we have the observability matrix

$$C = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 3 \\ 2 & 4 & 0.5 \\ 0 & 0 & 6 \\ 4 & 24 & 4 \\ 0 & 0 & 12 \end{bmatrix}$$

We check observability from its rank,

$$rank(\mathcal{C}) = 3$$
$$= N_x$$

Observability (cont.)

State estimation

Observabili

observer

```
>> help ctrb
                                            % OBSV computes the observability matrix
                                            % of pair (A,C)
                                            % Read about it and how to use it
3
4
  >> A = [?]:
                                            % Define state matrix A
  >> C = [?]:
                                            % Define control matrix C
  >> [Ny,Nx] = size(C);
                                           % Ny and Nx
  >> Omat = obsv(A,C)
                                            % Observability matrix
  >> rnkOmat = rank(Omat)
                                           % Rank of the observability matrix
14 >> rnkOmat == Nx
                                           % Return 0/1 for observability
```

State estimation

Observabilit

Asymptoti

Example

Consider the linear and time-invariant system (A, B, C), $x \in \mathbb{R}^{N_x}$, $u \in \mathbb{R}^{N_u}$, $y \in \mathbb{R}^{N_y}$

$$\begin{cases} \dot{x}(t) = \begin{bmatrix} 1 & 2 \\ 1 & 0 \end{bmatrix} x(t) + \begin{bmatrix} -2 \\ 3 \end{bmatrix} u(t) \\ y(t) = \begin{bmatrix} 1 & 0 \end{bmatrix} x(t) \end{cases}$$

Determine the stability, controllability, and observability using matrix A, C, and O

What changes when we have $C = \begin{bmatrix} 1 & 1 \end{bmatrix}$?

State estimation

Observabili

Asymptoti

Example

Consider the linear and time-invariant system (A, B, C), $x \in \mathbb{R}^{N_x}$, $u \in \mathbb{R}^{N_u}$, $y \in \mathbb{R}^{N_y}$

$$\begin{cases} \dot{x}(t) = \begin{bmatrix} 2 & 1 & 3 \\ 3 & 1 & 4 \\ 1 & 0 & 1 \end{bmatrix} x(t) + \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix} u(t) \\ y(t) = \begin{bmatrix} 1 & 0 & 1 \end{bmatrix} x(t) \end{cases}$$

Determine the stability, controllability, and observability using matrix A, C, and O

State estimation

Observabili

Asymptoti

Example

Consider the linear and time-invariant system (A, B, C), $x \in \mathbb{R}^{N_x}$, $u \in \mathbb{R}^{N_u}$, $y \in \mathbb{R}^{N_y}$

$$\begin{cases} \dot{x}(t) = \begin{bmatrix} 1.8 & 0.6 & -0.2 \\ 0.8 & 1.6 & -0.2 \\ -0.4 & -0.8 & 2.6 \end{bmatrix} x(t) + \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} u(t) \\ y(t) = \begin{bmatrix} 0 & 1 & 2 \end{bmatrix} x(t) \end{cases}$$

Determine the stability, controllability, and observability using matrix A, C, and O

State estimation

Observabili

Asymptoti observer

Example

Consider the linear and time-invariant system (A, B, C), $x \in \mathbb{R}^{N_x}$, $u \in \mathbb{R}^{N_u}$, $y \in \mathbb{R}^{N_y}$

$$\begin{cases} \dot{x}(t) = \begin{bmatrix} -2 & -2 \\ 1 & 0 \end{bmatrix} x(t) + \begin{bmatrix} 1 \\ 1 \end{bmatrix} u(t) \\ y(t) = \begin{bmatrix} 1 & 1 \end{bmatrix} x(t) \end{cases}$$

Determine the stability, controllability, and observability using matrix $A,\,\mathcal{C},$ and \mathcal{O}

State estimation

Observability

Asymptotic observer

Asymptotic state observers

LTI systems - Estimation

Asymptotic state observers

State estimation

 $\begin{array}{c} {\bf Asymptotic} \\ {\bf observer} \end{array}$

We show that for linear and time-invariant systems which are observable it is possible to exactly estimate the state vector at some infinite $t \to \infty$ (that is, asymptotically)

$$\lim_{t \to \infty} ||x(t) - \widehat{x}(t)|| = 0$$

The device that estimates the state of s system is itself a dynamic system

Asymptotic state observers

State estimation

Asymptotic observer

Definition

Luenberger observer

Consider a linear and time-invariant dynamical system

$$\begin{cases} \dot{x}(t) = Ax(t) + Bu(t) \\ y(t) = Cx(t) \end{cases},$$

with $x(t) \in \mathcal{R}^{N_x}$, $u(t) \in \mathcal{R}^{N_u}$, and $y(t) \in \mathcal{R}^{N_y}$

The linear and time-invariant dynamical system

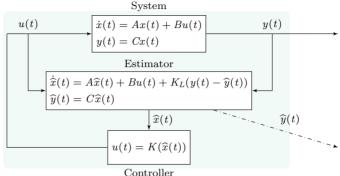
$$\begin{cases} \dot{\widehat{x}}(t) = A\widehat{x}(t) + Bu(t) + K_L (y(t) - \widehat{y}(t)) \\ \widehat{y}(t) = C\widehat{x}(t) \end{cases},$$

with $\hat{x} \in \mathcal{R}^{N_x}$, $\hat{y}(t) \in \mathcal{R}^{N_y}$ is a Luenberger observer of the system iff $K_L \in \mathcal{R}^{N_x \times N_y}$ is any matrix such that the eigenvalues of matrix $A - K_L C$ all have a negative real part

State estimation

Asymptotic observer

Asymptotic state observers (cont.)



Controller

$$\begin{split} \dot{\widehat{x}}(t) &= A\widehat{x}(t) + Bu(t) + K_L y(t) - K_L \widehat{y}(t) \\ &= A\widehat{x}(t) + Bu(t) + K_L y(t) - K_L C \widehat{x}(t) \\ &= (A - K_L C)\widehat{x}(t) + Bu(t) + K_L y(t) \\ &= \underbrace{(A - K_L C)}_{A_L} \widehat{x}(t) + \underbrace{\begin{bmatrix} B & K_L \end{bmatrix}}_{B_L} \begin{bmatrix} u(t) \\ y(t) \end{bmatrix} \end{split}$$

Asymptotic state observers (cont.)

State estimation

 $\begin{array}{c} {\bf Asymptotic} \\ {\bf observer} \end{array}$

Theorem

Consider a linear and time-invariant dynamical system

$$\begin{cases} \dot{x}(t) = Ax(t) + Bu(t) \\ y(t) = Cx(t) \end{cases},$$

with $x(t) \in \mathcal{R}^{N_x}$, $u(t) \in \mathcal{R}^{N_u}$, and $y(t) \in \mathcal{R}^{N_y}$

The Luenberger observer is an asymptotic state observer

$$\begin{cases} \dot{\widehat{x}}(t) = A\widehat{x}(t) + Bu(t) + K_L(y(t) - \widehat{y}(t)) \\ \widehat{y}(t) = C\widehat{x}(t) \end{cases}$$

with $\widehat{x} \in \mathcal{R}^{N_x}$, $\widehat{y}(t) \in \mathcal{R}^{N_y}$

Asymptotic state observers (cont.)

State estimation

Asymptotic observer

Proof

Let us define the estimation error e(t), the difference between true state and estimate

$$e(t) = x(t) - \widehat{x}(t)$$

We want to show that the error has linear and time-invariant homogenous dynamics

- \rightarrow The state matrix of this system is given by $A K_L C$
- \rightarrow Because $A K_L C$ is stable, the error vanishes
- → (Asymptotically)

State estimation

Observabilit

Asymptotic observer

Asymptotic state observers (cont.)

We differentiate with respect to time the definition of estimation error and rearrange

$$\begin{split} \dot{e}(t) &= \dot{x}(t) - \dot{\widehat{x}}(t) \\ &= \underbrace{\left(Ax(t) + Bu(t)\right)}_{\dot{x}(t)} - \underbrace{\left(A\widehat{x}(t) + Bu(t) + K_L\left(y(t) - \widehat{y}(t)\right)\right)}_{\dot{x}(t)} \\ &= \left(Ax(t) + Bu(t)\right) - \left(A\widehat{x}(t) + Bu(t) + K_L\left(y(t) - \widehat{y}(t)\right)\right) \\ &= A\left(\underbrace{x(t) - \widehat{x}(t)}_{e(t)}\right) - K_Ly(t) + K_L\widehat{y}(t) \\ &= A\left(\underbrace{x(t) - \widehat{x}(t)}_{e(t)}\right) - K_L\underbrace{\left(Cx(t)\right)}_{y(t)} - K_L\underbrace{\left(C\widehat{x}(t)\right)}_{\widehat{y}(t)} \\ &= A\left(\underbrace{x(t) - \widehat{x}(t)}_{e(t)}\right) - K_LC\left(\underbrace{x(t) - \widehat{x}(t)}_{\widehat{y}(t)}\right) \\ &= \left(A - K_LC\right)(x(t) - \widehat{x}(t)) \\ &= \left(A - K_LC\right)e(t) \end{split}$$

State estimation

Observabilit

Asymptotic observer

The dynamics of the state estimation error $e(0) = x(0) - \hat{x}(0)$,

$$\dot{e}(t) = \underbrace{(A - K_L C)}_{A_L} e(t),$$

We have $\lim_{t\to\infty} \|x(t) - \widehat{x}(t)\| = 0$, for all inputs u(t) and initial states x(0) and $\widehat{x}(0)$



Asymptotic state observers (cont.)

State estimation

Asymptotic observer

$$\begin{cases} \hat{x}(t) = A\hat{x}(t) + Bu(t) + K_L(y(t) - \hat{y}(t)) \\ \hat{y}(t) = \hat{x}(t) \end{cases}$$

A Luenberger observer is a linear and time-invariant dynamical system

- The observer has the same order N_x of the original system
- The observer state variable is the state estimate $\widehat{x}(t)$
- The observer input variables are u(t) and y(t)

The system and observer share the same measurement equation

While the estimation error tends to zero as time tends to infinity, the rate at which the error becomes practically negligible depends on the (estimated) initial state $\hat{x}(0)$

• The closer to the actual state x(0), the better