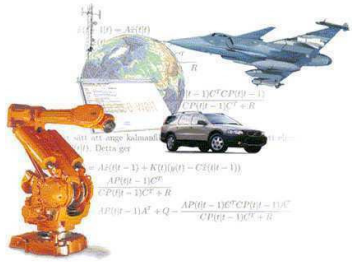


Dynamic Systems

Lecture 2 Observability



Torkel Glad
Reglerteknik, ISY, Linköpings Universitet

Observability

Continuous time model:

$$\dot{x}(t) = f(x(t), u(t)), \quad y(t) = h(x(t), u(t))$$

Discrete time model:

$$x(t+1) = f(x(t), u(t)), \quad y(t) = h(x(t))$$

Can you compute the state x from the output y and the input u ?

Observability, discrete time:

Look at several points in time to extract more information

$$y(t) = h(x(t))$$

$$y(t+1) = h(x(t+1)) = h(f(x(t), u(t))) = h^{(1)}(x(t), u(t))$$

$$y(t+2) = h^{(1)}(x(t+1), u(t+1)) = h^{(1)}(f(x(t), u(t)), u(t+1)) = h^{(2)}(x(t), u(t), u(t+1))$$

$$\vdots$$

$$y(t+N) = h^{(N)}(x(t), u(t), u(t+1), \dots, u(t+N-1))$$

This is a system of nonlinear equations in $x(t)$. Solvability implies observability.

Observability, continuous time:

Differentiate the output to extract more information

$$y = h(x)$$

$$\dot{y} = h_x(x)\dot{x} = h_x(x)f(x, u) = h^{(1)}(x, u)$$

$$\ddot{y} = h_x^{(1)}(x, u)\dot{x} + h_u^{(1)}(x, u)\dot{u} = h_x^{(1)}(x, u)f(x, u) + h_u^{(1)}(x, u)\dot{u} = h^{(2)}(x, u, \dot{u})$$

$$\vdots$$

$$y^{(N)} = h^{(N)}(x, u, \dot{u}, \dots, u^{(N-1)})$$

This is a system of nonlinear equations in x . Solvability implies observability.

No input. Alternative description

Define

$$L_f = \sum f_i(x) \frac{\partial}{\partial x_i}$$

Then

$$y = h(x)$$

$$\dot{y} = (L_f h)(x)$$

$$\ddot{y} = (L_f^2 h)(x)$$

\vdots

$$y^{(N)} = (L_f^N h)(x)$$

Solvability usually depends on u

Example:

$$\dot{x}_1 = x_2 u, \quad \dot{x}_2 = x_1 x_2, \quad y = x_1$$

The system of equations

$$y = x_1$$

$$\dot{y} = x_2 u$$

can not be solved for x_2 if $u = 0$.

Mathematics of equation solving

- Linear equations: solvability determined by rank test
- Polynomial equations: extensive (and difficult) mathematical theory (ideals, Gröbner bases, characteristic sets, elimination theory, cylindrical algebraic decomposition). Very high computational complexity.
- Local properties of general equations: Implicit function theorem.

Implicit function theorem

Consider the equation

$$f(x, y) = 0$$

where the dimensions of x and f are equal (same number of unknowns and equations).

Assume that

$$f(x_0, y_0) = 0, \quad f_x(x_0, y_0) \text{ nonsingular}$$

(f_x is the Jacobian of f with respect to x)

Then, for all y close to y_0 the equation has a solution x which is locally unique.

Local observability

Compute the Jacobian

$$J = \begin{bmatrix} h_x(x) \\ h_x^{(1)}(x, u, \dot{u}) \\ \vdots \\ h_x^{(N)}(x, u, \dot{u}, \dots, u^{(N-1)}) \end{bmatrix}$$

J full rank at $x_0 \Rightarrow x$ is uniquely determined by u, y and their derivatives in a neighborhood of x_0 (implicit function theorem)

Linear time invariant systems

$$\dot{x} = Ax + Bu, \quad y = Cx$$

The Jacobian J is constant and has the form

$$J = \begin{bmatrix} C \\ CA \\ \vdots \\ CA^{N-1} \end{bmatrix}$$

- J full rank $\Rightarrow x$ can be solved uniquely (globally)
- This is the classical observability test for linear systems
- Cayley-Hamilton \Rightarrow no need to take $N > n$.

Linear time-varying systems

$$\dot{x} = A(t)x + B(t)u, \quad y = C(t)x$$

The Jacobian J has the form

$$J = \begin{bmatrix} C \\ \frac{\partial}{\partial t}C + CA \\ \vdots \end{bmatrix}$$

See the exercises.

Quantitative measure of observability

$$\dot{x}(t) = A(t)x(t), \quad y(t) = C(t)x(t)$$

The energy in the output is given by

$$\int_{t_0}^{t_1} y(t)^T y(t) dt = x(t_0)^T M(t_0, t_1) x(t_0)$$

where M is the *Observability Gramian*:

$$M(t_0, t_1) = \int_{t_0}^{t_1} \Phi^T(t, t_0) C^T(t) C(t) \Phi(t, t_0) dt$$

Observability and the Gramian

From the definition of the observability Gramian:

$$\int_{t_0}^{t_1} \Phi^T(t, t_0) C^T(t) y(t) dt = M(t_0, t_1) x(t_0)$$

Theorem:

observability on $[t_0, t_1] \Leftrightarrow M(t_0, t_1)$ nonsingular

Note: This is a smoothing estimate of $x(t_0)$ based on $y(t)$,
 $t_0 \leq t \leq t_1$.

Observability, discrete time

Iterating the state equation with $u = 0$ gives

$$\begin{bmatrix} y(t_0) \\ y(t_0 + 1) \\ \vdots \\ y(t_1 - 1) \end{bmatrix} = \begin{bmatrix} C(t_0) \\ C(t_0+1)A(t_0) \\ \vdots \\ \underbrace{C(t_1-1)A(t_1-2) \cdots A(t_0)}_{O(t_0, t_1)} \end{bmatrix} x(t_0)$$

$x(t_0)$ can be computed from $y(t_0), \dots, y(t_1 - 1)$ if $O(t_0, t_1)$ has full rank.

$O^T O$ is the discrete time observability Gramian.

For constant A and C one can define

$$\mathcal{O} = \begin{pmatrix} C \\ CA \\ \vdots \\ CA^{n-1} \end{pmatrix}$$

in both continuous and discrete time.

Theorem

The range and null spaces of $M(t_0, t_1)$
coincide with
the range and null spaces of $\mathcal{O}^T \mathcal{O}$
for all $t_1 > t_0$.

The corresponding discrete time result is trivial.

Change of state variables

$$x = T(z), \quad T \text{ one-to-one } T_z \text{ nonsingular}$$

gives

$$\dot{z} = T_z^{-1} f(T(z), u) = \tilde{f}(z, u), \quad y = h(T(z)) = \tilde{h}(z)$$

To test observability, we have to compute

$$\tilde{h}^{(1)}(z, u) = \tilde{h}_z(z) \tilde{f}(z, u) = h_x(T(z)) T_z T_z^{-1} f(T(z), u) = h^{(1)}(T(z), u)$$

$$\text{In general } \tilde{h}^{(j)}(z, u, \dot{u}, \dots, u^{(j-1)}) = h^{(j)}(T(z), u, \dot{u}, \dots, u^{(j-1)})$$

Change of state variables, cont'd

The Jacobian is

$$\tilde{J} = \begin{bmatrix} \tilde{h}_z(z) \\ \tilde{h}_z^{(1)}(z, u, \dot{u}) \\ \vdots \\ \tilde{h}_z^{(N)}(z, u, \dot{u}, \dots, u^{(N-1)}) \end{bmatrix} = \begin{bmatrix} h_x(T(z))T_z \\ h_x^{(1)}(T(z), u, \dot{u})T_z \\ \vdots \\ h_x^{(N)}(T(z), u, \dot{u}, \dots, u^{(N-1)})T_z \end{bmatrix} = JT_z$$

Since T_z is nonsingular, the rank of J is the same in both coordinate systems.

Change of state variables, linear systems

$$x = Tz, \quad T \text{ nonsingular matrix}$$

gives

$$\dot{z} = \underbrace{T^{-1}AT}_A z + \underbrace{T^{-1}B}_B u, \quad y = \underbrace{CT}_C z$$

$$\tilde{\mathcal{O}} = \begin{bmatrix} \tilde{C} \\ \tilde{C}\tilde{A} \\ \vdots \end{bmatrix} = \begin{bmatrix} CT \\ CTT^{-1}AT \\ \vdots \end{bmatrix} = \mathcal{O}T$$

Unobservability and observers

Consider the systems

$$\dot{x} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} x$$

$$y = [1 \quad 1] x$$

$$\dot{x} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} x$$

$$y = [1 \quad -1] x$$

- Both systems are unobservable
- One of them can be given a converging observer
- For the other one this is not possible
- Why?

Canonical form for observability

$$\dot{z} = \underbrace{T^{-1}AT}_A z + \underbrace{T^{-1}B}_B u, \quad y = \underbrace{CT}_C z$$

Theorem Let the rank of \mathcal{O} be r . Then T can be chosen so that

$$\tilde{A} = \begin{pmatrix} \tilde{A}_{11} & 0 \\ \tilde{A}_{21} & \tilde{A}_{22} \end{pmatrix}, \quad \tilde{C} = (\tilde{C}_{11} \quad 0)$$

where \tilde{A}_{11} is $r \times r$, \tilde{C}_{11} is $p \times r$ and $\tilde{A}_{11}, \tilde{C}_{11}$ observable. *Non-unique!*

Note: The partition of eigenvalues between \tilde{A}_{11} and \tilde{A}_{22} is unique: it makes sense to speak of “observable” and “unobservable” eigenvalues.

Similar theorem for time-varying systems, see Rugh.

PBH tests

Popov 1966, Belevitch 1968, Hautus 1969:

Theorem: A, C observable if and only if

$$Ap = \lambda p, Cp = 0 \Rightarrow p = 0$$

Theorem: A, C observable if and only if

$$\text{rank} \begin{pmatrix} C \\ sI - A \end{pmatrix} = n$$

for all complex s .

More on nonlinear observability

The definition of nonlinear observability is tricky. Is it for instance reasonable to say that the following system is observable?

$$\dot{x} = 1$$
$$y = \begin{cases} 0 & x < 10^{100} \\ (x - 10^{100})^2 & x \geq 10^{100} \end{cases}$$

Detectability

Detectable system: All eigenvalues of the unobservable part lie strictly in left half plane.

The following are equivalent:

- The system is detectable.
- It is possible to choose the observer gain K so that $A - KC$ has all its eigenvalues strictly in the left half plane.

Proof: PBH and canonical form.

Some terminology

$$\dot{x} = f(x, u), \quad y = h(x), \quad x(0) = x_0$$

has solution $x(t) = \pi(t, x_0, u)$

x_1, x_2 **indistinguishable:**

$h(\pi(t; x_1, u)) = h(\pi(t; x_2, u)), \text{ all } t \geq 0, \text{ all } u$

$I(x) =$ all points indistinguishable from x

- system **observable at** $x_0: I(x_0) = \{x_0\}$
- system **observable:** $I(x_0) = \{x_0\}$ all x_0

More terminology

Often sufficient to distinguish points that are close:

- system **weakly observable** at x_0 : exists neighborhood (nbh) U such that $I(x_0) \cap U = \{x_0\}$

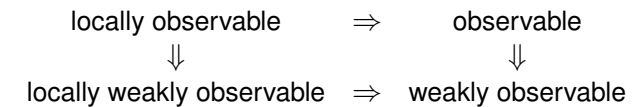
Often necessary to distinguish points without moving to far:

- x_1, x_2 **U-indistinguishable** if they are distinguishable as long as both trajectories lie entirely in U .
- $I_U(x) =$ all points U-indistinguishable from x

Local weak observability

- System **locally observable** at x_0 : $I_U(x_0) = \{x_0\}$ every open nbh U of x_0 .
- System **locally weakly observable** at x_0 : Exists open nbh U of x_0 such that $I_V(x_0) = \{x_0\}$ for every open nbh V of x_0 with $V \subset U$
- System **locally weakly observable**: locally weakly observable at every x_0 ;
“ x can instantaneously be distinguished from its neighbors”

Relationships:



For a *linear* system they are all equivalent.

Lie derivative evaluation (piecewise constant u)

Define $f_1(x) = f(x, u_1)$

$$(L_{f_1}h)(x) = h_x(x)f_1(x) = \dot{y}$$

The *Lie-derivative* in the direction f_1 .

More generally, if $f = f_1$ for t_1 units of time, $f = f_2$ for t_2 units of time,...

$$\left(\frac{\partial^k}{\partial t_1 \dots \partial t_k} y(t_1 + t_2 + \dots + t_k) \right) \Big|_{t_1 = \dots = t_k = 0} = (L_{f_1} L_{f_2} \dots L_{f_k} h)(x_0)$$

Test for local weak observability

Consider all elements of the form

$$h_x, (L_{f_1}h)_x, (L_{f_2}h)_x, \dots, (L_{f_1}L_{f_2}h)_x, \dots, (L_{f_1}L_{f_2} \dots L_{f_k}h)_x, \dots$$

for all possible choices of u .

- Observability rank condition at x_0 : n linearly independent rows among these elements
- Hermann and Krener 1977:
 - Observability rank condition at $x_0 \Rightarrow$ local weak observability at x_0
 - Local weak observability for all $x \Rightarrow$ observability rank condition generic (satisfied on open dense subset)