

# CDS 101/110: Lecture 5.2

## Observability & State Estimation

October 26, 2016

### Goals:

- Introduce notion of Observability
- Observability Test and Observable Canonical Form.
- Observer Design
- State feedback of estimated state

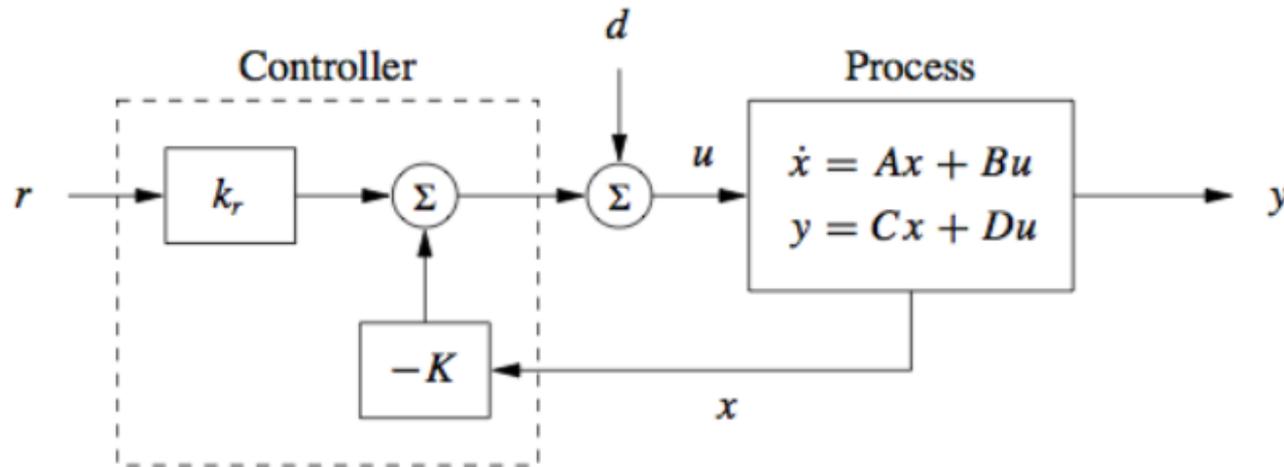
### Reading:

- Åström and Murray, Feedback Systems-2e, Section 8.1-8.2, and first half of 8.3

# Types of Feedback

**State Feedback:**  $u(t) = -Kx(t)$

- Can place poles arbitrarily if system is reachable
- Can relate poles to performance criteria, such as overshoot.
- Can add “dynamic compensator”, such as integral feedback, which overcomes modeling errors or uncertainty.
- *But*, states cannot always be measured, as needed for feedback.



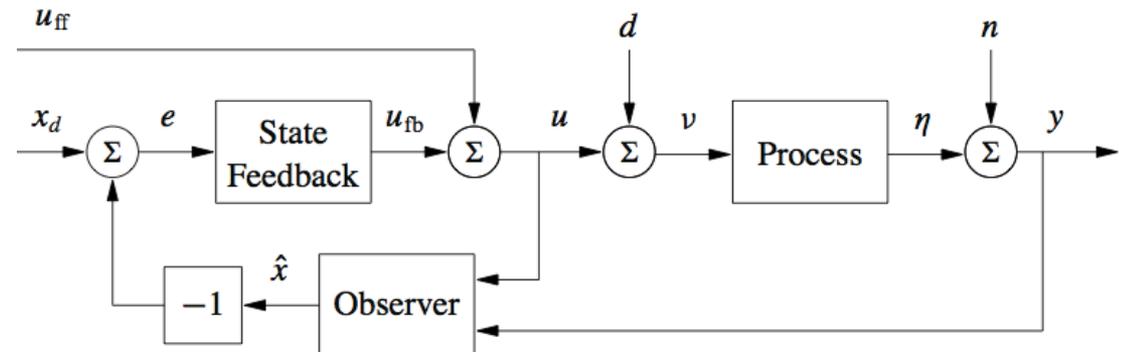
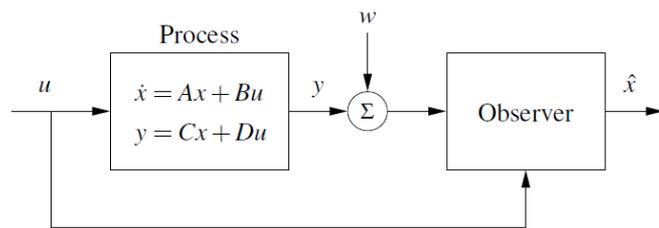
# Types of Feedback

## Output Feedback:

- Can we stabilize the system using only the output measurements,  $y(t)$ ?
- **Approach #1:** develop a theory which directly addresses this problem
- **Approach #2:** connect to state feedback theory by trying to determine if states is can be “reconstructed”, “inferred”, or “estimated” from  $y(t)$ .

## Key Questions:

- Are internal states,  $x(t)$ , “observable” from output  $y(t)$ ?
- How do we actually estimate  $x(t)$  from  $y(t)$ ?
- Does the use of output feedback limit performance?



# Observability

**System:**  $\dot{x} = Ax + Bu; \quad y = Cx + Du \quad (*)$

- **Definition:** The linear system (\*) is said to be **Observable** if for every  $T > 0$  it is possible to determine the system state  $x(T)$  through measurements  $y(t)$  and knowledge of  $u(t)$  on the interval  $[0, T]$ .
  - Note: some texts/papers are slightly different: Observable if  $x(t = 0)$  can be determined from measurements and inputs.
  - If (\*) is observable, then there are no “hidden” internal states. This is a practical issue in system design—do you have the right sensors?

## Testing for Observability:

- Simplify the problem by ignoring controls:  $\dot{x} = Ax; \quad y = Cx \quad (**)$ 
  - We can do this because of linearity!
  - If  $C$  is square and full rank, then solution is easy:  $x(T) = C^{-1}y(T)$ .
  - But that's generally not the case

# Observability (continued)

## Construct Estimate as follows:

- Take the derivative of  $y(t)$ :  $\dot{y} = C\dot{x} = CAx$ 
  - Note: some texts/papers are slightly different: Observable if  $x(t = 0)$  can be determined from measurements and inputs.
  - If (\*) is observable, then there are no “hidden” internal states. This is a practical issue in system design—do you have the right sensors?
  - Continue taking derivatives of output:  $\ddot{y}(t) = \frac{d}{dt}(CAx) = CA^2x(t), \dots$
  - Arrange all of the derivatives in a column

$$\begin{bmatrix} y(t) \\ \dot{y}(t) \\ \ddot{y}(t) \\ \vdots \\ W_O \end{bmatrix} = \begin{bmatrix} C \\ CA \\ CA^2 \\ \vdots \\ CA^{n-1} \end{bmatrix} x(t) \equiv W_O x(t)$$

Observability Matrix

- Why can we stop at  $A^{n-1}$ ?

# Observability (continued)

## Aside: Cayley-Hamilton Theorem

- Let  $A$  be an  $n \times n$  matrix.
- Let  $\lambda_A(s) = \det(sI - A) = s^n + a_1 s^{n-1} + \dots + a_{n-1}s + a_n$  be characteristic polynomial of  $A$ .
- $A$  satisfies its own characteristic polynomial:  $A^n + a_1 A^{n-1} + \dots + a_{n-1}A + a_n I = 0$ 
  - Hence,  $A^k$  for  $k \geq n$  are linear combinations of  $I, A, \dots, A^{n-1}$

## Theorem:

- A linear system (\*\*) is **observable** if and only if  $W_O$  (the observability matrix) is full rank. For a single output system, this is equivalent to  $W_O$

begin full rank. In this case:  $x(T) = W_O^{-1} \begin{bmatrix} y(T) \\ \vdots \\ \frac{d^{n-1}}{dt^{n-1}} y(T) \end{bmatrix}$