# Dynamic Systems Theory-State-space Control - Lecture 16 $^\ast$

November 1, 2012

#### 1 Summary of Controllability and Observability

$$\dot{x} = A(t)x(t) + B(t)u(t)$$
$$y = c(t)x(t)$$

 $x \sim \dim n, y \sim \dim m, u \sim \dim p$ 

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

These grammians tell the whole story. The system is <u>controllable</u>  $\Leftrightarrow W(t_0, t_1)$  is invertible. The system is <u>observable</u>  $\Leftrightarrow M(t_0, t_1)$  is invertible. For the constant coefficient case, the relevant objects are  $(B, AB, \ldots, A^{n-1}B)$  controllability: rank n $(C, CA, \ldots, CA^{n-1})^T$  observability: rank n

#### 2 Discrete time case

$$\begin{aligned} x(k+1) &= A(k)x(k) + B(k)u(k) \\ y(k) &= C(k)x(k) \end{aligned}$$

<sup>\*</sup>This work is being done by various members of the class of 2012

$$\begin{aligned} x(N+1) &= A(N)x(N) + B(N)u(N) \\ &= \underbrace{A(N)\dots A(1)}_{\Phi(N,1)} x(1) + A(N)\dots A(2)B(1)u(1) + \dots + B(u)u(N) \\ &= \Phi(N,1)x(1) + \sum_{j=1}^{N-1} \Phi(N,j+1)B(j)u(j) + B(N)u(N) \end{aligned}$$

#### 2.1 Discrete Time Theorem

<u>Discrete Time Theorem</u>: There exists a control sequence  $u(1), \ldots, u(N)$  that steers the state of

$$x(k+1) = A(k)x(k) + B(k)u(k)$$

from  $x(t) = x_0$  to  $x(N) = x_1$  if and only if

$$x_1 - \Phi(N-1,1)x_0$$

belongs to the range space of

$$W = \sum_{j=1}^{N-2} \Phi(N-1, j+1)B(j)B(j)^T \Phi(N-1, j+1)^T + B(N-1)B(N-1)^T$$

<u>Proof</u>: Suppose there is an  $\eta$  such that

$$x_1 - \Phi(N-1,1)x_0 = W\eta$$

Let

$$u(j) = \begin{cases} B(j)^T \Phi(N-1, j+1)^T \eta & j = 1, \dots N-2\\ B(N-1)^T \eta & \text{for } j = N-1 \end{cases}$$

Then

$$\begin{aligned} x(N) &= \Phi(N-1,1)x_0 + \sum_{j=1}^{N-2} \Phi(N-1,j+1)B(j)(B(j)^T \Phi(N-1,j+1)^T)\eta \\ &+ B(N-1)B(N-1)^T \eta \\ &= \Phi(N-1,1)x_0 + W\eta \\ &= x_1 \end{aligned}$$

Going the other way, suppose that  $x_1 - \Phi(N-1,1)x_0$  is not in the range space of W. Then there exists a  $y \in \mathbb{R}^n$  such that  $y \circ (x_1 - \Phi(N-1,1)x_0) \neq 0$ , but  $W^T y = 0$ . Following the same reasoning as in the continuous time case, this would imply

$$0 = y^T W y = \sum_{j=1}^{N} ||B(j)^T \Phi(N, j)^T y||^2$$
  

$$\Rightarrow B(j)^T \Phi(N, j)^T y = 0 \text{ for } j = 1, \dots N$$

This contradiction shows that if there is an input sequence achieving specified boundary conditions for all possible choices of boundary conditions, then W has full rank.

#### 2.2 Constant coefficient case

$$\begin{aligned} x(N+1) &= Ax(N) + Bu(N) \\ &= A^2 x(N-1) + ABu(N-1) + Bu(N) \\ & \dots \\ &= A^N x(1) + A^{N-1} Bu(1) + A^{N-2} Bu(2) + \dots + ABu(N-1) + Bu(N) \end{aligned}$$

The controllability rank condition [rank  $(B, AB, \ldots, A^{N-1}B) = n$ ] is equivalent to the condition of whether or not we can find a sequence  $u(1), \ldots, u(N)$  which steers between  $x_0$  and  $x_1$  for any choice of values  $x_0, x_1 \in \mathbb{R}^n$ .

We can always choose  $N \leq n$ . Hence the controllability condition is

$$B, AB, \ldots, A^{n-1}B$$

has rank n.

### 2.3 Observability in the constant coefficient discrete time case

$$y(k) = Cx(k) \qquad x(k+1) = Ax(k)$$

we observe

$$\begin{array}{rcl} y(k) & = & Cx(k) \\ y(k+1) & = & CAx(k) \\ & \vdots & \\ y(n-1+k) & = & CA^{n-1}x(k) \end{array} \right\} \text{observe eq.}$$

We can always solve the observe eq. for x(k) in terms the sequence  $y(k), y(k+1), \cdots, y(k+n-1) \Leftrightarrow \operatorname{rank} (C, CA, \cdots, CA^{n-1})^T = n$ 

## 3 Sampled systems hybrid continuous discrete systems

$$\dot{x} = A(t)x(t) + B(t)u(t)$$
  

$$y(t) = C(t)x(t)$$
(1)

By samlpling the state every h seconds, we get

$$y(kh) = C(kh)x(kh)$$

Letting

$$\overline{x}(k) = x(kh), \overline{A}(k) = \Phi((k+1)h, kh)$$
$$\overline{B}(k) = \int_{kh}^{(k+1)h} \Phi((k+1)h, s)B(s)ds$$
$$\overline{C}(k) = c(kh)$$

We find the state and output dynamics of

$$x(k+1) = \overline{A}(k)\overline{x}(k) + \overline{B}(k)\overline{u}(k)$$
(2)

is the same as a sampled system, sampled every **h** units of time

$$\dot{x} = A(t)x(t) + B(t)u_c(t)$$
  

$$y(t) = C(t)x(t)$$
(3)

where

$$u_c(t) = \overline{u}(k)$$
 for  $kh \le t < (k+1)h$ 

(2) is called the sampled version of (1).

<u>Note:</u> Constant coefficient continuous time systems give rise to constant coefficient sampled systems.

$$\overline{A} = e^{Ah}$$

$$\overline{B} = \int_{kh}^{(k+1)h} e^{A[(k+1)h-s]} B ds = \int_0^h e^{A(h-s)} B ds$$

The Laplace transform of

$$\dot{x} = A(t)x(t) + B(t)u_c(t)$$
$$y(t) = C(t)x(t)$$

is

$$\hat{y}(s) = C(Is - A)^{-1}B\hat{u}_c(s)$$

Intelligent Machines

A discrete sequence  $x(0), x(1), x(2), \cdots$  has an <u>associated</u> z-transform

$$\hat{x}(z) = \sum_{k=0}^{\infty} x(k) z^{-k}$$

Apply the z-transform to both sides of

$$\begin{aligned} x((k+1)h) &= Ax(kh) + Bu(kh) \\ y(kh) &= Cx(kh) \end{aligned}$$

$$\sum_{k=0}^{\infty} x((k+1)h)z^{-k}) = A \sum_{k=0}^{\infty} x(kh)z^{-k} + B \sum_{k=0}^{\infty} u(kh)z^{-k}$$

The left hand side can be written

$$z\left[\sum_{k=0}^{\infty} x(kh)z^{-k} - x(0)\right]$$

Assume that x(0) = 0Denote

$$Y(z) = \sum_{k=0}^{\infty} y(kh) z^{-k}$$
$$X(z) = \sum_{k=0}^{\infty} x(kh) z^{-k}$$
$$U(z) = \sum_{k=0}^{\infty} u(kh) z^{-k}$$
$$zX(z) = AX(z) + BU(z)$$
$$Y(z) = CX(z)$$
$$= C(Iz - A)^{-1}BU(z)$$

We wish to have a formal procedure for relating the z-transform of a sample system to the Laplace transform of the continuous time system from which it was obtainted.

Given a continuous signal f(t), we define the sampled representation to be

$$f^*(t) = \sum_{k=-\infty}^{\infty} f(kh)\delta(t-kh)$$

where  $\delta(t)$  is the Direc delta function defined formally by

$$\int_{-\infty}^{\infty} y(s)\delta(s)ds = g(0)$$

The Laplace transform of the sampled system  $f^*(t)$  is

$$\sum_{0}^{\infty} f^*(t)e^{-st}dt = \int_0^{\infty} \sum_{k=-\infty}^{\infty} f(kh)e^{-st}\delta(t-kh)dt$$
$$= f(0) + f(h)e^{-sh} + f(2h)e^{-s2h} + \cdots$$
$$= \sum_{k=0}^{\infty} f(kh)(e^{-sh})^k$$

This is the z-transform provided f(t) = 0 for t < 0 and provided we make the identification  $z = e^{sh}$ .

REMARKS ON THE ZERO-ORDER HOLD:

Given the sampled signal  $\{f(kh), k = 0, 1, 2 \cdots \}$ , one useful reconstruction is

$$f_c(t) = f(kh)$$
 for  $kh \le t < (k+1)h$ 

Applying this to our sampled control input  $u_c(t)$ , the Laplace transform is

$$\begin{split} \hat{u}_{c}(s) &= \int_{0}^{\infty} e^{-st} u_{c}(t) dt \\ &= \sum_{k=0}^{\infty} \int_{kh}^{(k+1)h} e^{-st} u(kh) dt \\ &= \sum_{k=0}^{\infty} u(kh) (-\frac{1}{s} e^{-st} \mid_{t=kh}^{t=(k+1)h}) \\ &= \sum_{k=0}^{\infty} u(kh) \frac{1}{s} (e^{-skh} - e^{-s(k+1)h}) \\ &= \sum_{k=0}^{\infty} u(kh) e^{-skh} \frac{1}{s} (1 - e^{-sh}) \\ &= \frac{1}{s} (1 - e^{-sh}) \mathfrak{L}(u^{*}(t)) \\ &= \frac{1}{s} (1 - e^{-sh}) u_{c}(z) \end{split}$$

Note that  $\frac{1}{s}(1-e^{-sh})$  is the Laplace transform of the zero-order hold; i.e. the Laplace transform of

$$x(t) = \begin{cases} 1 & 0 \le t < h \\ 0 & \text{elsewhere} \end{cases}$$