Lecture 17: Linear Discrete-Time Systems; Reachability and Controllability

# 5.3 Reachability and Controllability for Discrete-Time Linear Systems

### 5.3.1 Time-Varying Case

Consider the discrete-time linear time-varying state-space system:

$$x(j+1) = A(j)x(j) + B(j)u(j), \ x(r) = x_r$$
  

$$y(j) = C(j)x(j) + D(j)u(j)$$
(5.12)

where  $x(j) \in \mathbb{R}^n$ ,  $u(j) \in \mathbb{R}^m$ ,  $y(j) \in \mathbb{R}^p$ . We found the state response to be

$$x(j) = \Phi(j,r)x_r + \sum_{i=r}^{j-1} \Phi(j,i+1)B(i)u(i).$$
(5.13)

and the state transition matrix to be:

$$\Phi(k,r) = \prod_{i=r}^{k-1} A(i) .$$
(5.14)

The definitions of reachability and controllability for (5.12) are entirely similar to the continuous-time case. However, complete controllability, controllability at  $t_0$ , and reachability at  $t_j$  are not always equivalent as was the case for continuous-time state-space systems. This is because  $\Phi(j,r)$  is not always invertible, e.g., A(i) may be singular for some  $m \le i \le j-1$ .

Similar to the continuous-time case, the admissible input functions  $u(\cdot)$  belong to the space  $l_2^m[r,k], r \le j \le k$ :

$$\mathcal{U}_{ad} := l_2^{\ m}[r,k], \tag{5.15}$$

i.e., the space of (finite-dimensional) square-summable sequences, which we will view as finite-dimensional vectors in  $\mathbb{R}^{k-r+1}$ 

### Definition: Controllability Operator

The controllability operator for the system (5.12) is the transformation

$$\mathcal{L}: \mathcal{U}_{ad} \to \mathbb{R}^n, \quad \mathcal{L}u \coloneqq \sum_{i=r}^{k-1} \Phi(k, i+1)B(i)u(i)$$
(5.16)

which maps admissible inputs into state vectors.

### The Controllability Equation

The input which produces a terminal state  $z \in \mathbb{R}^n$  from the zero initial state  $\theta$  is any solution of the *controllability equation*:

$$\mathcal{L}u = z, \quad u \in \mathbb{R}^{m(k-r+1)}, \quad z \in \mathbb{R}^n$$
(5.17)

More generally, the input u produces a change in terminal state  $z = x_1 - \Phi(k, r)x_r$  from what it would be as a free trajectory starting at  $(x_r, r)$ .

The controllability equation can be expressed as:

$$\mathcal{L}u := \Phi(k, r+1)B(r)u(r) + \Phi(k, r+2)B(r+1)u(r+1) + \dots + \Phi(k, k)B(k-1)u(k-1)$$

$$= \underbrace{\left[\Phi(k, r+1)B(r) \quad \Phi(k, r+2)B(r+1) \quad \dots \quad \Phi(k, k)B(k-1)u(k-1)\right]}_{\mathcal{L}} \underbrace{\left[\begin{array}{c}u(r)\\u(r+1)\\\vdots\\u(k-1)\\\vdots\\u(k-1)\\\end{matrix}\right]}_{u}}_{u} = z$$
(5.18)

That is, the controllability operator is a matrix, with typically more columns than rows, and hence a solution to the controllability equation is composed of a component  $u_1 \in S_{sp} \{\mathcal{L}\}$  and any component in the nullspace  $u_N \in \mathcal{N} \{\mathcal{L}\}$ .



### Proposition:

The system (5.12) is CC  $\Leftrightarrow \mathcal{L}$  is onto.

- If the system is CC, then  $z \in \mathcal{R}{\mathcal{L}}$  and there are many solutions to the controllability equation.
- If the system is not CC, and  $z \notin \mathcal{R} \{ \mathcal{L} \}$  and there is no solution to the controllability equation.

So two problems of interest here would be:

- 1. Find  $\mathcal{R} \{ \mathcal{L} \}$ , the space of attainable states, and
- 2. Find the smallest-norm input  $u \in \mathcal{U}_{ad}$  that produces the state z .

Problem 1 can be solved by reduction of L, the matrix representation of  $\mathcal{L}$ , into the echelon form.

To address Problem 2, we will use the adjoint  $\mathcal{L}^*$  in the next section. But first, let us look at the LTI case.

# 5.3.2 Time-Invariant Case

Consider the discrete-time linear time-invariant state-space system:

$$x(j+1) = Ax(j) + Bu(j), \ x(r) = x_r$$
  

$$y(j) = Cx(j) + Du(j)$$
(5.19)

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The controllability equation is given by:

$$\mathcal{L}u = \underbrace{\left[\Phi(k, r+1)B(r) \quad \Phi(k, r+2)B(r+1) \quad \cdots \quad \Phi(k, k)B(k-1)\right]}_{L} \underbrace{\left[\begin{array}{c}u(r)\\u(r+1)\\\vdots\\u(k-1)\end{array}\right]}_{u}$$

$$= \underbrace{\left[A^{k-r-1}B \quad A^{k-r-2}B \quad \cdots \quad B\right]}_{L} \underbrace{\left[\begin{array}{c}u(r)\\u(r+1)\\\vdots\\u(k-1)\end{array}\right]}_{u} = z$$
(5.20)

which we can rearrange by reversing the order of the input samples:

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$$\mathcal{L}u = \begin{bmatrix} B & AB & \cdots & A^{k-r-1}B \end{bmatrix} \begin{bmatrix} u(k-1) \\ u(k-2) \\ \vdots \\ u(r) \end{bmatrix} = z$$
(5.21)

Starting from the initial time r = 0:

### **Definition: Controllability Matrix**

The controllability matrix of system (5.19) is defined as:

$$L_{n} := \begin{bmatrix} B & AB & \cdots & A^{n-1}B \end{bmatrix} \in \mathbb{R}^{n \times mn}$$
(5.23)

#### Proposition:

The time-invariant system is reachable (or controllable-from-the-origin) iff the controllability matrix  $L_n$  has full row rank, i.e., rank  $\{L_n\} = n$ .

#### Proof:

The reason for using  $L_n$  rather than  $L_k$ , k > n to define the controllability matrix is because if a transfer from the origin cannot be accomplished in n time steps of the control sequence, it cannot be accomplished by taking more than n time steps. That is, it can be shown that:

$$\mathcal{R}\left\{L_n\right\} = \mathcal{R}\left\{L_k\right\}, \ k \ge n$$

but note that  $\mathcal{R}\{L_n\} \supset \mathcal{R}\{L_k\}, \ k < n$ .

It follows that it is possible to transfer the state from some vector  $x(0) = x_0$  to some other vector  $x(n) = x_1$  in *n* time steps iff there exists an *n*-step input sequence  $\{u(0), u(1), \dots, u(n-1)\}$  that satisfies:

$$x_1 - A^n x_0 = L_n u^n \,. \tag{5.24}$$

which has a solution iff  $x_1 - A^n x_0 \in \mathcal{R} \{L_n\}$ . Clearly, the LTI system is reachable from the origin, implying that any  $x_1$  can be reached from  $x(0) = x_0$  in finite time iff  $\operatorname{rank} \{L_n\} = n$  so that  $\mathcal{R} \{L_n\} = \mathbb{R}^n$ .

Note that  $\mathcal{R}\{L_n\}$  is called the *reachable subspace* of the system.

Example:

$$x(j+1) = Ax(j) + Bu(j)$$

$$A = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix}, \quad B = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$
(5.25)

- Controllability matrix:  $L_2 = \begin{bmatrix} B & AB \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix}$  has rank  $\{L_2\} = 2$ . Hence the system is reachable.
- Any state, say  $x_1 = \begin{bmatrix} a \\ b \end{bmatrix}$ , can be reached from the origin in n = 2 time steps:

$$\begin{bmatrix} a \\ b \end{bmatrix} = L_2 u^2 = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} u(1) \\ u(0) \end{bmatrix}$$
$$\Rightarrow \begin{bmatrix} u(1) \\ u(0) \end{bmatrix} = \begin{bmatrix} b - a \\ a \end{bmatrix}$$

Check:

$$x(j+1) = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} x(j) + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u(j)$$
$$x(1) = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 0 \\ 0 \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u(0) = \begin{bmatrix} 0 \\ a \end{bmatrix}$$
$$x(2) = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 0 \\ a \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u(1) = \begin{bmatrix} a \\ a \end{bmatrix} + \begin{bmatrix} 0 \\ b-a \end{bmatrix} = \begin{bmatrix} a \\ b \end{bmatrix}$$

• Since the system is reachable, it is possible to transfer the state from some vector  $x(0) = x_0$  to some other vector  $x(n) = x_1$  in n = 2 time steps. Let  $x_0 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$  and  $x_1 = \begin{bmatrix} a \\ b \end{bmatrix}$ . Then:

$$\begin{aligned} x_{1} - A^{2}x_{0} &= L_{2}u^{2} = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} u(1) \\ u(0) \end{bmatrix} \\ \begin{bmatrix} a \\ b \end{bmatrix} - \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} u(1) \\ u(0) \end{bmatrix} \\ \begin{bmatrix} a \\ b \end{bmatrix} - \begin{bmatrix} 2 \\ 3 \end{bmatrix} = \begin{bmatrix} u(0) \\ u(0) + u(1) \end{bmatrix} \\ \Rightarrow \begin{bmatrix} u(1) \\ u(0) \end{bmatrix} = \begin{bmatrix} b - a - 1 \\ a - 2 \end{bmatrix} \end{aligned}$$

Notes:

- The system is controllable to the origin when  $A^n x_0 \in \mathcal{R} \{L_n\}, \ \forall x_0 \in \mathbb{R}^n$ .
- If rank  $\{A\} = n$ , the system is controllable when rank  $\{L_n\} = n$  (i.e., when the reachability condition is satisfied) because then  $-A^n x_0 = L_n u^n$  can be solved for any  $x_0$ . In this case,

$$A^{-n}L_n := \begin{bmatrix} A^{-n} & B & A^{-n+1} & \cdots & A^{-1} & B \end{bmatrix} \in \mathbb{R}^{n \times mn}$$

is of interest, and the system is controllable iff  $\operatorname{rank} \{A^{-n}L_n\} = \operatorname{rank} \{L_n\} = n$ . If, however,  $\operatorname{rank} \{A\} < n$ , then controllability does not imply reachability.

• The above example is controllable-to-the-origin. Let  $x_1 = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$  and  $x_0 = \begin{bmatrix} a \\ b \end{bmatrix}$ . Then:

$$-A^{2}x_{0} = L_{2}u^{2} = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} u(1) \\ u(0) \end{bmatrix}$$
$$-\begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} u(1) \\ u(0) \end{bmatrix}$$
$$-\begin{bmatrix} a+b \\ a+2b \end{bmatrix} = \begin{bmatrix} u(0) \\ u(0)+u(1) \end{bmatrix}$$
$$\Rightarrow \begin{bmatrix} u(1) \\ u(0) \end{bmatrix} = \begin{bmatrix} -b \\ -a-b \end{bmatrix}$$

# 5.4 Solutions to the Discrete-Time Controllability Problem

# 5.4.1 Linear Time-Varying Case

Since the controllability operator  ${\mathcal L}$  is a map between finite-dimensional spaces, the pseudoinverse solution

$$u_{opt} = \mathcal{L}^* \left( \mathcal{L} \mathcal{L}^* \right)^{-1} z \tag{5.26}$$

is a solution of the controllability equation  $\mathcal{L}u = z$ .

This solution solves the problem with the minimum-norm input signal  $u_{opt}$ . Also of interest is the corresponding state trajectory  $u_{opt}$ .



We know that the system in (5.12) is CC  $\Leftrightarrow \mathcal{L}$  is onto  $\Leftrightarrow (\mathcal{LL}^*)^{-1}$  exists. Thus, complete controllability is equivalent to the existence of the inverse operator  $(\mathcal{LL}^*)^{-1}$ . In this case,  $u_{opt}$  as given in (5.26) exists, where

$$\mathcal{L}u = \sum_{i=r}^{k-1} \Phi(k, i+1)B(i)u(i).$$
(5.27)

The adjoint  $\mathcal{L}^*: \mathbb{R}^n \to l_2^{-m}[r, k-1]$  is obtained as follows:

$$\left\langle \mathcal{L}u, z \right\rangle_{\mathbb{R}^{n}} = \left( \sum_{i=r}^{k-1} \Phi(k, i+1) B(i) u(i) \right)^{*} z = \sum_{i=r}^{k-1} u(i)^{*} B(i)^{*} \Phi(k, i+1)^{*} z$$
  
=  $\left\langle u(\cdot), B(\cdot)^{*} \Phi(k, \cdot+1)^{*} z \right\rangle_{l_{2}[r,k-1]} = \left\langle u(\cdot), \mathcal{L}^{*} z \right\rangle_{l_{2}[r,k-1]} \quad \forall u \in l_{2}[r,k-1], \forall z \in \mathbb{R}^{n}$  (5.28)  
 $\Rightarrow \mathcal{L}^{*} = B(\cdot)^{*} \Phi(k, \cdot+1)^{*}$ 

Thus,

$$\mathcal{LL}^{*}z = \underbrace{\sum_{i=r}^{k-1} \Phi(k, i+1)B(i)B(i)^{*} \Phi(k, i+1)^{*}}_{M(r,k)} z$$

$$= M(r,k)z$$
(5.29)

#### **Definition: Controllability Grammian**

The matrix  $M(r,k) \in \mathbb{R}^{n \times n}$  is the above equation is called the *controllability grammian* of the system. It depends only on the time interval [r, k], but it is constant once this interval is fixed.

Finally, the optimal, minimum-norm input is given by:

$$u_{opt}(j) = \left(\mathcal{L}^{*}\left(\mathcal{L}\mathcal{L}^{*}\right)^{-1}z\right)(j) = B(j)^{*}\Phi(k, j+1)^{*}M^{-1}(r, k)z$$
(5.30)

Recall that the norm is  $\left\|u\right\|_2 = \sqrt{\sum_{i=r}^{k-1} \left\|u(i)\right\|^2}$ .

### 5.4.2 Linear Time-Invariant Case

Consider the system

$$x(j+1) = Ax(j) + Bu(j), \ x(r) = x_r.$$
(5.31)

Then,

$$x(k) = A^{k-r}x_r + \sum_{i=r}^{k-1} A^{k-i-1}Bu(i)$$
(5.32)

Assume the system is CC. Then, the controllability grammian is invertible:

$$M(r,k) = \sum_{i=r}^{k-1} A^{k-i-1} B B^* A^{*^{k-i-1}}$$
(5.33)

Fact:

The controllability grammian can be written in terms of the controllability matrix:

$$M(r,k) = L_k L_k^{*}$$
(5.34)

We have the adjoint:

$$\mathcal{L}^* = B^* A^{*^{k-(\cdot)+1}}$$
(5.35)

The optimal control input is:

$$u_{opt}(j) = B^* A^{*^{k-j+1}} M(r,k)(x_k - A^{k-r} x_r)$$
(5.36)

and finally,

$$\mathcal{R}\left\{L_{n}\right\} = \mathcal{R}\left\{M(r,k)\right\}.$$
(5.37)

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