### Linear System Theory

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Chapter 6: Controllability & Observability

Chapter 7: Minimal Realizations

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### Recap

- State space equation
- ► Linear Algebra
- Solutions of LTI and LTV system
- Stability

#### We will study

- ► Controllability & Observability
- ► Kalman Decomposition
- Minimal realizations

The first chapter dealing with the control input and output variable in the system

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

- x: state
- ▶ *u*: control
- ▶ *y*: output

Controllability (informal): we want to know whether the state of the system is controllable or not from the input

- ▶ Analyze the system structure from the input
- With the input, we want to move the state to the desired point in a finite time.

Observability (informal): we want to observe the initial state of the system from the output and input to quantify the behavior of the system

- ▶ State: position, velocity, acceleration, etc
- Sensors are required to measure the state. We are not able to use many sensors in real applications.

#### Controllability & Observability

- ▶ Important concepts in control, estimation, and filtering problems
- Optimal control (LQG, Kalman filtering, etc.)

#### Example

$$\dot{x} = \begin{pmatrix} -1 & 0 \\ 0 & 3 \end{pmatrix} x + \begin{pmatrix} b_1 \\ b_2 \end{pmatrix} u, \ x = (x_1 \ x_2)^T$$

- ▶  $(b_1, b_2)^T = (-1, 1)^T$ : can move both eigenvalues  $\Leftrightarrow$  can control the state  $x_1$  and  $x_2$
- ▶  $(b_1, b_2)^T = (1, 0)^T$ : cannot move the eigenvalue  $3 \Leftrightarrow$  cannot control state  $x_2$
- ▶  $(b_1, b_2)^T = (1, 0)^T$ : No matter input,  $x_2$  diverges  $\Leftrightarrow$  we cannot control  $x_2$

#### Example

$$\dot{x} = \begin{pmatrix} -1 & 0 \\ 0 & 3 \end{pmatrix} x, \ y = \begin{pmatrix} c_1 & c_2 \end{pmatrix} x$$

- $(c_1, c_2) = (1, 1)$ : can observe the state  $x_1$  and  $x_2$
- $(c_1, c_2) = (1, 0)$ : cannot observe the state  $x_2$
- $(c_1, c_2) = (1, 0)$ : Output is always stable, but the system is internally unstable

$$\dot{x}(t) = Ax(t) + Bu(t), \ x \in \mathbb{R}^n, \ u \in \mathbb{R}^m$$

Definition (Definition 6.1)

The state equation with the pair (A, B) is said to be controllable if for any initial state  $x(0) = x_0$ , any final state  $x_1$ , there exists an input that transfers  $x_0$  to  $x_1$  in a finite time.

#### Equivalent Definition:

A system is controllable at time  $t_0$  if there exists a finite time  $t_f$  such that for any initial condition  $x_0$ , and any final state  $x_f$ , there is a control input u defined on  $[t_0, t_f]$  such that  $x(t_f) = x_f$ .

- We need an input u to transfer the state from the initial to the final state
- ▶ Given initial and finial state conditions in  $\mathbb{R}^n$ , is it possible to steer x(t) to the final state by choosing an appropriate input u(t)?

## Controllability: A Preview

#### Discrete-time LTI system

$$x(k+1) = Ax(k) + Bu(k), \ x(0) = 0, \ x \in \mathbb{R}^{n}, \ u \in \mathbb{R}^{m}$$

$$x(1) = Bu(0)$$

$$x(2) = Ax(1) + Bu(1) = ABu(0) + Bu(1)$$

$$x(3) = A^{2}Bu(0) + ABu(1) + Bu(2)$$

$$\vdots$$

$$x(r) = A^{r-1}Bu(0) + A^{r-1}Bu(1) + \dots + Bu(r-1)$$

$$x(r) = (B \quad AB \quad \dots A^{r-1}B) \begin{pmatrix} u(r-1) \\ u(r-2) \\ \vdots \\ u(0) \end{pmatrix}$$

### Controllability: A Preview

$$x(r) = \begin{pmatrix} B & AB & \cdots & A^{r-1}B \end{pmatrix} \begin{pmatrix} u(r-1) \\ u(r-2) \\ \vdots \\ u(0) \end{pmatrix}$$

$$R((B AB \cdots A^{r-1}B)) = \{z \in \mathbb{R}^n, \ z = (B AB \cdots A^{r-1}B)p, \ p \in \mathbb{R}^{nm}\}$$
If  $x_f \in R((B AB \cdots A^{r-1}B))$ , then  $x_f$  is reachable

## Controllability: A Preview

This implies that we can reach arbitrary  $x_f \in \mathbb{R}^n$  at time  $t_f = r$  if and only if  $R((B \ AB \ \cdots \ A^{r-1}B)) = \mathbb{R}^n$  that is equivalent to  $rank((B \ AB \ \cdots \ A^{r-1}B)) = n$ 

Rank of  $(B AB \cdots A^{r-1}B)$ 

- ▶ By C-H theorem,  $A^k$  is a linear combination of  $\{I, A, ..., A^{n-1}\}$
- ▶ For  $r \ge n$ , the rank of  $(B \ AB \ \cdots \ A^{r-1}B))$  cannot increase

Hence, if  $rank((B AB \cdots A^{n-1}B)) = n$ , then we can find u for an arbitrary  $x_f \in \mathbb{R}^n$  for any finite time

$$\dot{x}(t) = Ax(t) + Bu(t), \ x \in \mathbb{R}^n$$
  $x(t) = e^{At}x(0) + \int_0^t e^{A(t-\tau)}Bu(\tau)d\tau$ 

The system is controllable (page 213 of the textbook)

- ▶  $\Leftrightarrow$  for any  $x_0$ , there exists u(t) on  $[t_0, t_f]$  that transfers  $x_0$  to the origin at  $t_f$  (controllability to the origin)
- ▶  $\Leftrightarrow$  there exists u(t) on  $[t_0, t_f]$  that transfers state from the origin to any final state  $x_f$  at  $t_f$  (reachability)

Proof: Exercise!! (note that  $e^{A(t-t_0)}$  is always invertible!)

Basically, we need the surjectivity of  $\int_0^t e^{A(t- au)} Bu( au) d au$ 

$$\dot{x}(t) = Ax(t) + Bu(t), \ x \in \mathbb{R}^n, \ u \in \mathbb{R}^m, \ x_0 = 0$$

▶ Set of reachable state for a fixed time *t*:

$$\mathcal{R}_t = \{ \xi \in \mathbb{R}^n, \text{ there exists } u \text{ such that } x(t) = \xi \}$$

Note that  $\mathcal{R}_t$  is a subspace of  $\mathbb{R}^n$ 

► Controllability matrix and controllability subspace

$$C_{AB} = \{ \xi \in \mathbb{R}^n : \ \xi = \begin{pmatrix} B & AB & \cdots & A^{n-1}B \end{pmatrix} z, \ z \in \mathbb{R}^{nm} \}$$

 $\mathcal{C}_{AB}$ : range space of  $\mathcal{C}$ , where  $\mathcal{C} = (B \ AB \ \cdots \ A^{n-1}B) \in \mathbb{R}^{n \times nm}$ 

► Controllability Gramian

$$W_t = \int_0^t e^{A(t-\tau)}BB^T e^{A^T(t-\tau)}d\tau = \int_0^t e^{A\tau}BB^T e^{A^T\tau}d\tau \ge 0$$

 $R(W_t)$ : the range space of  $W_t$ ,  $W_t$  is a symmetric positive semi-definite matrix

Theorem: Controllability (Theorem 6.1 of the textbook) For each time t > 0, the following set equality holds:

$$\mathcal{R}_t = \mathcal{C}_{AB} = R(W_t).$$

- $\mathcal{C} = (B \ AB \ \cdots \ A^{n-1}B)$ : controllability matrix
- ▶ Hence if dim  $C_{AB} = rank((B \ AB \ \cdots \ A^{n-1}B)) = n$ , the system is controllable
- ▶ Due to  $C_{AB}$ , the controllability is independent of the time
- ▶ If the system is controllable, then  $\mathcal{R}_t = \mathbb{R}^n$ , all the states are reachable by an appropriate choice of the control u

#### We will show that

 $ightharpoonup \mathcal{R}_t \subset \mathcal{C}_{AB}, \, \mathcal{C}_{AB} \subset R(W_t), \, R(W_t) \subset \mathcal{R}_t$ 

#### Required tools

▶ C-H theorem (Chapter 3),  $R(A^T) = (N(A))^{\perp}$ : Problem 1 in HW3



Theorem:  $\mathcal{R}_t \subset \mathcal{C}_{AB}$ 

Proof:

Fix t > 0, and choose any reachable state  $\xi \in \mathcal{R}_t$ . We need to show that  $\xi \in \mathcal{R}_t$  implies  $\xi \in \mathcal{C}_{AB}$ .

We have  $\xi \in \mathcal{R}_t$ , which implies  $\xi = \int_0^t e^{A(t-\tau)} Bu(\tau) d\tau$ . Then by C-H theorem,  $e^{At} = \beta_0(t)I + \cdots + \beta_{n-1}(t)A^{n-1}$  ( $\beta_i(t)$ : scalar function).

Hence

$$\xi = B \int_0^t \beta_0(t-\tau)u(\tau)d\tau + \dots + A^{n-1}B \int_0^t \beta_{n-1}(t-\tau)u(\tau)d\tau$$

$$= (B \quad AB \quad \dots \quad A^{n-1}B) \underbrace{\begin{pmatrix} \int_0^t \beta_0(t-\tau)u(\tau)d\tau \\ \vdots \\ \int_0^t \beta_{n-1}(t-\tau)u(\tau)d\tau \end{pmatrix}}_{\in \mathbb{D}^{nm}}$$

Hence,  $\xi \in \mathcal{C}_{AB}$ 

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Theorem: \mathcal{C}_{AB} \subset R(W_t)

Proof:

Since \mathcal{C}_{AB} \subset R(W_t) is equivalent to \mathcal{C}_{AB}^{\perp} \supset R(W_t)^{\perp} (proof: exercise!!), we will show that \mathcal{C}_{AB}^{\perp} \supset R(W_t)^{\perp}.

From Problem 1 in HW3, R(W_t) = (N(W_t))^{\perp}, which is equivalent to (R(W_t))^{\perp} = N(W_t), and similarly, \mathcal{C}_{AB}^{\perp} = N((B AB \cdots A^{n-1}B)).

Hence we need to show that if \xi \in N(W_t), then \xi \in N((B AB \cdots A^{n-1}B)).
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Let  $\xi \in N(W_t)$ , then  $W_t \xi = 0 \in \mathbb{R}^n$ , which also implies  $\xi^T W_t \xi = 0 \in \mathbb{R}$ . Then

$$0 = \xi^T \int_0^t e^{A\tau} B B^T e^{A^T \tau} d\tau \xi = \int_0^t \|B^T e^{A^T \tau} \xi\|^2 d\tau \Leftrightarrow B^T e^{A^T \tau} \xi = 0, \ \forall \tau \in [0, t]$$

Since  $y(\tau) = \xi^T e^{A\tau} B = 0$ ,  $\forall \tau \in [0, t]$ , we have

$$\xi^{T} \left( \frac{d^{k}}{d\tau^{k}} e^{A\tau} \right) \Big|_{\tau=0} B = \xi^{T} A^{k} B = 0, \ \forall k \ge 0$$
  
$$\Rightarrow \xi^{T} \left( B \quad AB \quad \cdots \quad A^{n-1} B \right) = 0 \Rightarrow \xi \in \mathcal{N}((B \ AB \quad \cdots \quad A^{n-1} B)) = \mathcal{C}_{AB}^{\perp}$$

Theorem:  $R(W_t) \subset \mathcal{R}_t$ 

Proof:

Let  $\xi \in R(W_t)$ . Then there exists  $v \in \mathbb{R}^n$  such that

$$\xi = W_t v = \int_0^t e^{A\tau} B B^T e^{A^T \tau} v d\tau$$

Define  $u(\tau) = B^T e^{A^T(t-\tau)} v, \ \tau \in [0, t]$ 

Then, since  $\dot{x} = Ax + Bu$  with x(0) = 0, we have

$$x(t) = \int_0^t e^{A(t-\tau)} Bu(\tau) d\tau$$
$$= \int_0^t e^{A(t-\tau)} BB^T e^{A^T(t-\tau)} v d\tau = W_t v = \xi$$

This means that  $\xi \in \mathcal{R}_t$ , since we have found the control u that steers the state to  $\xi$  from the origin.

Theorem (Theorem 6.1 of the textbook)

If (A, B) is controllable, and A is stable (eigenvalues of A have negative real parts), then there exists a unique solution of

$$AP + PA^T = -BB^T$$
,

where 
$$P = \int_0^\infty e^{A au} B B^T e^{A^T au} d au > 0$$

- ▶ Note that  $BB^T \ge 0$
- ▶ In Chapter 5,  $AP + PA^T = -Q$  where Q > 0

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Theorem (Theorem 6.1 of the textbook) (A, B) is controllable if and only if rank((A - \lambda I B)) = n for all eigenvalues, \lambda, of A.
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► Hautus-Rosenbrock test

Theorem (Theorem 6.1 of the textbook) (A,B) is controllable if and only if  $W_t>0$ , that is, the controllability Gramian is non-singular

Theorem (Theorem 6.2 of the textbook) Let  $\bar{A} = PAP^{-1}$  and  $\bar{B} = PB$ . Then (A, B) is controllable if and only if  $(\bar{A}, \bar{B})$  is controllable

Controllability is invariant under the similarity transformation

Fact: The state space equation with the controllable canonical form is always controllable.

$$\dot{x} = Ax + Bu, \ G(s) = \frac{X(s)}{U(s)} = (sI - A)^{-1}B$$

Kalman Decomposition Theorem (Theorem 6.6 of the textbook): Suppose that  $\mathcal{C}_{AB} = r < n$ . Let

$$P = (v_1, \ldots, v_r, v_{r+1}, \ldots, v_n)$$

where  $v_i$ , i = 1, 2, ..., r is eigenvectors of C, and  $v_{r+1}, ..., v_n$  are arbitrary vectors that guarantees P being nonsingular. Let z = Px.

Then

$$\begin{split} \dot{z} &= PAP^{-1}z + PBu \\ \bar{A} &= PAP^{-1} = \begin{pmatrix} \bar{A}_{11} & \bar{A}_{12} \\ 0 & \bar{A}_{22} \end{pmatrix}, \ \bar{B} = PB = \begin{pmatrix} \bar{B}_1 \\ 0 \end{pmatrix} \\ \bar{A}_{11} &\in \mathbb{R}^{r \times r}, \ \bar{B}_1 \in \mathbb{R}^{r \times m} \end{split}$$

Also,  $(\bar{A}_{11}, \bar{B}_1)$  is controllable, and  $G(s) = (sI - \bar{A}_{11})^{-1}\bar{B}_1$ .

$$\dot{x} = Ax + Bu, \ y = Cx, \ x \in \mathbb{R}^n, \ y \in \mathbb{R}^p$$

Definition (Definition 6.01)

The state-space equation is said to be observable if for any unknown initial condition, there exists a finite  $t_1$  such that the knowledge of the input and the output over  $[0, t_1]$  is suffices to determine uniquely the initial condition x(0).

W.L.G., 
$$u = 0$$
, (since  $u$  is completely known)

Note that

$$y(t) = Ce^{At}x(0)$$

Hence, if  $N(Ce^{At}) = \emptyset$ , i.e.,  $\dim(N(Ce^{At})) = nullity(Ce^{At}) = 0$ , then the system is observable.

 $\triangleright$   $N(Ce^{At})$ : unobservable subspace

Let

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

 $\mathcal{O}$ : Observability matrix,  $\mathcal{O} \in \mathbb{R}^{pn \times n}$ 

Theorem:  $N(Ce^{At}) = N(\mathcal{O})$ 

Proof: We will show that  $N(Ce^{At}) \subset N(\mathcal{O})$  and  $N(Ce^{At}) \supset N(\mathcal{O})$ .

If  $x_0 \in N(Ce^{At})$ , then

$$0 = Ce^{At}x_0 \Rightarrow 0 = C\left(\frac{d}{dt}e^{At}\right)\Big|_{t=0}x_0 \Rightarrow 0 = CA^kx_0, \ \forall k \ge 0$$

Hence,  $x_0 \in N(\mathcal{O})$ .

If  $x_0 \in N(\mathcal{O})$ . then  $x_0 \in N(Ce^{At})$ , since by C-H Theorem, we have

$$Ce^{At} = C\beta_0(t)I + \cdots + CA^{n-1}\beta_{n-1}(t)$$

If  $N(Ce^{At}) = \emptyset$ , i.e.,  $\dim(N(Ce^{At})) = nullity(Ce^{At}) = 0$ , then the system is observable.

- We need  $N(Ce^{At}) = N(\mathcal{O}) = \emptyset$
- ▶ Hence, by the rank-nullity theorem, the system is observable if  $rank(\mathcal{O}) = n$
- ▶ We say that the system is observable if and only if the pair (C, A) is observable
- Observability also does not depend on the time (by C-H Theorem)

Duality Theorem (Theorem 6.5 of the textbook) The following are equivalent:

- $\triangleright$  (C, A) is observable
- $\triangleright$   $(A^T, C^T)$  is controllable

Proof: 
$$(A^T, C^T)$$
 is controllable if and only if 
$$\mathcal{O}^T = (C^T \ A^T C^T \ \cdots \ (A^T)^{n-1} C^T)$$
 
$$rank(\mathcal{O}^T) = n = rank(\mathcal{O})$$
 
$$\mathcal{O} = \begin{pmatrix} C \\ CA \\ \vdots \\ CA^{n-1} \end{pmatrix}$$

Theorem (Theorem 6.01) If (A, C) is observable, and A is stable, then there exists a unique solution of

$$A^T P + PA = -C^T C$$

where 
$$P = \int_0^\infty e^{A^T \tau} C^T C e^{A \tau} d\tau > 0$$
.

Theorem (Theorem 6.O1) (C, A) is observable if and only if

$$rank \begin{pmatrix} C \\ A - \lambda I \end{pmatrix} = n$$

Theorem (Theorem 6.01) (C, A) is observable if and only if the observability Gramian

$$Q_t = \int_0^t e^{A^T \tau} C^T C e^{A \tau} d\tau > 0$$

Theorem (Theorem 6.O3) Let  $\bar{A} = PAP^{-1}$  and  $\bar{C} = CP^{-1}$ . Then (C, A) is observable if and only if  $(\bar{C}, \bar{A})$  is observable.

$$\dot{x} = Ax + Bu, \ y = Cx, \ G(s) = \frac{Y(s)}{U(s)} = C(sI - A)^{-1}B$$

Kalman Decomposition Theorem (Theorem 6.06 of the textbook): Suppose that  $rank(\mathcal{O}) = q < n$ . Let

$$P = egin{pmatrix} v_1 \ dots \ v_q \ v_{q+1} \ dots \ v_n \end{pmatrix}, \ v_1, \dots, v_q dots \ ext{eigenvectors}.$$

Let 
$$z=Px$$
. Then  $\dot{z}=PAP^{-1}z+PBu,\ y(t)=CP^{-1}z,$  and 
$$\bar{A}=PAP^{-1}=\begin{pmatrix}\bar{A}_{11}&0\\\bar{A}_{21}&\bar{A}_{22}\end{pmatrix},\ \bar{B}=PB=\begin{pmatrix}\bar{B}_{1}\\\bar{B}_{2}\end{pmatrix},\ \bar{C}=\begin{pmatrix}\bar{C}_{1}&0\end{pmatrix}$$
  $\bar{A}_{11}\in\mathbb{R}^{q\times q},\ \bar{C}_{1}\in\mathbb{R}^{p\times q}$ 

Also,  $(\bar{C}_1, \bar{A}_{11})$  is observable, and  $G(s) = \bar{C}_1(sI - A_{11})^{-1}\bar{B}_1$ .

### Kalman Decomposition Theorem

Theorem (Theorem 6.7 of the textbook)
We can extract the state that is controllable and observable.

Fact: The state space equation with the observable canonical form is always observable

## Discrete-Time LTI System

#### Discrete-time LTI system

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

- ▶ (A, B) is controllable if and only if rank(C) = n
- ▶ (C, A) is observable if and only if  $rank(\mathcal{O}) = n$

# Minimum Energy Control (page 189)

$$\dot{x}=Ax+Bu, \ x(0)=x_0, \ x(t_1)=x_f, \ (A,B)$$
 controllable  $W_{t_1}=\int_0^{t_1}e^{A\tau}BB^Te^{A^T\tau}d au>0, \ ext{invertible}$ 

Let

$$u^{*}(t) = -B^{T} e^{A^{T}(t_{1}-t)} W_{t_{1}}^{-1}(e^{At_{1}}x_{0} - x_{f})$$

$$x(t_{1}) = e^{At_{1}}x_{0} - \underbrace{\left(\int_{0}^{t_{1}} e^{A(t_{1}-\tau)}BB^{T} e^{A^{T}(t_{1}-\tau)}d\tau\right)}_{W_{t_{1}}} W_{t_{1}}^{-1}(e^{At_{1}}x_{0} - x_{f}) = x_{f}$$

# Minimum Energy Control (page 189)

We can show that the controller  $u^*$  is the minimum energy controller in the sense that for any controller u that transfers the state from  $x_0$  to  $x_f$ , we have

$$\int_0^{t_1} \|u(t)\|^2 dt \ge \int_0^{t_1} \|u^*(t)\|^2 dt, \ \forall u$$

## Stabilizability & Detectability

Weaker notions of controllability and observability

A system is stabilizable if and only if  $\bar{A}_{22}$  is stable and  $(\bar{A}_{11}, \bar{B}_1)$  is controllable

A system is detectable if and only if  $\bar{A}_{22}$  is stable and  $(\bar{C}_1, \bar{A}_{11})$  is observable

How about the example on pages 4-5. Is it stabilizable? Is it detectable?

## Controllability & Observability: LTV system

$$\begin{split} \dot{x}(t) &= A(t)x(t) + B(t)u(t), \ y(t) = C(t)x(t) \\ W_t &= \int_0^t \Phi(t,\tau)B(\tau)B^T(\tau)\Phi^T(t,\tau)d\tau \ge 0, \ \forall t \ge 0 \\ Q_t &= \int_0^t \Phi^T(t,\tau)C^T(\tau)C(\tau)\Phi(t,\tau)d\tau \ge 0, \ \forall t \ge 0 \end{split}$$

#### The LTV system is

- lacktriangle is controllable if and only if there exists  $t_f>0$  such that  $W_{t_f}>0$
- lacktriangle is observable if and only if there exists  $t_f>0$  such that  $Q_{t_f}>0$
- $\triangleright$   $W_t$ : controllability Graminan
- ▶ *Q<sub>t</sub>*: observability Graminan

We have seen that the realization of the state-space equation is not unique.

$$\begin{split} \dot{x} &= Ax + Bu, \ y = Cx + Du, \ x(0) = 0 \\ \dot{x} &= A_1x + B_1u, \ y = C_1x + D_1u, \ x(0) = 0 \\ y(t) &= C \int_0^t e^{A(t-\tau)}Bu(\tau)d\tau = C_1 \int_0^t e^{A_1(t-\tau)}B_1u(\tau)d\tau \end{split}$$

#### Lemma (not in the textbook)

▶ Two system realizations (A, B, C, D) and  $(A_1, B_1, C_1, D_1)$  are equivalent if and only if  $D = D_1$  and

$$Ce^{At}B = C_1e^{A_1t}B_1, \ \forall t \geq 0$$

▶ Two system realizations (A, B, C, D) and  $(A_1, B_1, C_1, D_1)$  are equivalent if and only if  $D = D_1$  and

$$CA^kB = C_1A_1^kB, \ \forall k \geq 0$$

In view of the Kalman decomposition, we have the following result:  $\Rightarrow$  Suppose (A, B, C, D) is a system realization. If either (C, A) is not observable or (A, B) is not controllable, then there exists a lower-order realization  $(A_1, B_1, C_1, D_1)$  for the system

Definition (page 233 of the textbook)

Realizations with the smallest possible dimension are called minimal realizations

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Theorem (Theorem 7.M2 (page 254)) (A, B, C, D) is a minimial realization of the transfer function G(s) if and only if (A, B) is controllable and (C, A) is observable
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If the system is not controllable or not observable (or not controllable and observable), then there are pole-zero cancellations in a transfer function.

### MATLAB Commands

- controllability matrix: ctrb(A, B)
- ▶ observability matrix: ctrb(A<sup>T</sup>, C<sup>T</sup>)
- ▶ minimal realization: minreal(A, B, C, D) ⇒ reduce the system order that has only controllable and observable state
- Mostly, we use the balanced realization (Chapter 7.4) ⇒ related to controllability and observability Gramians (robust control, advanced control topic)

### **Conclusions**

#### In this chapter

- Controllability
- Observability
- Duality
- ► Kalman decomposition
- Minimal realization

Next chapter: control system design

- ► Pole-placement
- observer design
- ▶ Optimum system design (LQR + Kalman filter)