

Lecture 4

- Controllability
- Observability
- Controller and Observer Forms
- Balanced Realizations

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Operator Interpretation

Define $M : \mathbf{L}_2^n[t_0, t_f] \rightarrow \mathbf{R}^n$ by

$$Mu = \int_{t_0}^{t_f} \Phi(t_0, \tau) B(\tau) u(\tau) d\tau$$

Then

$$\begin{aligned} x(t_f) &= \Phi(t_f, t_0)[x(t_0) + Mu] \\ (M^*x)(t) &= B(t)^T \Phi(t_0, t)^T x \\ MM^* &= \int_{t_0}^{t_f} \Phi(t_0, \tau) B(\tau) B(\tau)^T \Phi(t_0, \tau)^T d\tau \\ &= W(t_0, t_f) \end{aligned}$$

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Controllability

The equation

$$\dot{x}(t) = A(t)x(t) + B(t)u(t), \quad x(t_0) = x^0$$

is called *controllable on* (t_0, t_f) , if for any x^0 , there exists $u(t)$ such that $x(t_f) = 0$. The matrix function

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

is called *controllability Gramian*.

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Degree of Controllability

The minimal input, in terms of $|u|$, to go from $x(t_0) = x_0$ to $x(t_f) = 0$ can be used to evaluate degree of controllability.

From Lecture 2: Minimize $|u|$ under the constraint $x_0 + Mu = 0$.

$$\begin{aligned} \hat{u} &= -M^*(MM^*)^{-1}x_0 \quad (\text{if } MM^* \text{ invertible}) \\ |\hat{u}|^2 &= x_0^T (MM^*)^{-1} x_0 \\ &= x_0^T W(t_0, t_f)^{-1} x_0 \end{aligned}$$

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Theorem 1: Controllability Criterion

The system $\dot{x}(t) = A(t)x(t) + B(t)u(t)$ is controllable on (t_0, t_f) if and only if $W(t_0, t_f) > 0$. The minimal cost $\int_{t_0}^{t_f} |u|^2 dt$ to reach 0 from x_0 is $x_0^T W(t_0, t_f)^{-1} x_0$.

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Theorem 2: Time-Invariant Controllability

The following four conditions are equivalent:

- (i) The system $\dot{x}(t) = Ax(t) + Bu(t)$ is controllable.
 - (ii) $\text{rank}[B \ AB \ A^2B \ \dots \ A^{n-1}B] = n$.
 - (iii) $\lambda \in \mathbb{C}, p^T A = \lambda p^T, p^T B = 0 \Rightarrow p = 0$ (PBH-test)
 - (iv) $\text{rank}[\lambda I - A \ B] = n \quad \forall \lambda \in \mathbb{C}$. (PBH-test)
- Popov-Belevitch-Hautus (PBH), see p221.
Notice the Rugh Example 9.6 on p147.

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Proof of Theorem 1

Controllability on (t_0, t_f)

$$\begin{aligned}
 &\Leftrightarrow \forall x_f : \exists u : x(t_f) = 0 \\
 &\Leftrightarrow \forall x_f : \exists u : x_0 + Mu = 0 \\
 &\Leftrightarrow \mathcal{R}(M) = \mathbb{R}^n \\
 &\Leftrightarrow \mathcal{N}(M^*) = \{0\} \\
 &\Leftrightarrow \mathcal{N}(MM^*) = \{0\} \\
 &\Leftrightarrow \mathcal{N}[W(t_0, t_f)] = \{0\} \\
 &\Leftrightarrow W(t_0, t_f) > 0
 \end{aligned}$$

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Theorem 3: Uncontrollable State Equation

Suppose that $0 < q < n$ and

$$\text{rank} \begin{bmatrix} B & AB & A^2B & \dots & A^{n-1}B \end{bmatrix} = q < n$$

Then there exists an invertible $P \in \mathbb{R}^{n \times n}$ such that

$$P^{-1}AP = \begin{bmatrix} \hat{A}_{11} & \hat{A}_{12} \\ 0 & \hat{A}_{22} \end{bmatrix} \quad P^{-1}B = \begin{bmatrix} \hat{B}_{11} \\ 0 \end{bmatrix}$$

where \hat{A}_{11} is $q \times q$, \hat{B}_{11} is $q \times m$, and

$$\text{rank}[\hat{B}_{11} \quad \hat{A}_{11}\hat{B}_{11} \dots \hat{A}_{11}^{q-1}\hat{B}_{11}] = q$$

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Proof of Theorem 3

Let p_1, \dots, p_q be linearly independent columns from

$$[B \quad AB \dots A^{n-1}B]$$

Let $p_{q+1} \dots p_n$ be additional columns that make

$$P = [p_1 \dots p_q p_{q+1} \dots p_n]$$

invertible.

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Proof of Theorem 2

(i) \Rightarrow (ii) If (ii) fails, then after a coordinate change

$$\begin{bmatrix} \hat{x}_1 \\ \hat{x}_2 \end{bmatrix} = Px$$

as in Theorem 3, \hat{x}_2 is unaffected by the input, so (i) fails.

(ii) \Rightarrow (i) If $p^T W(t_0, t_f)p = 0$ for some $p \neq 0$, then

$$p^T e^{A(t_0-t)} B = 0 \quad \forall t \in [t_0, t_f]$$

Differentiation with respect to t at $t = t_0$, gives

$$p^T [B \quad AB \dots A^{n-1}B] = 0,$$

so (ii) fails.

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Proof continued

$$\text{With } [\hat{A}_1 \quad \hat{A}_2] = P^{-1}AP, \hat{B} = P^{-1}B$$

$$\mathcal{R}(P\hat{B}) = \mathcal{R}(B) \subset \mathcal{R}([p_1 \dots p_q])$$

$$\Rightarrow \hat{B} = \begin{bmatrix} \hat{B}_1 \\ 0 \end{bmatrix}$$

$$\mathcal{R}(P\hat{A}_1) = \mathcal{R}(A[p_1 \dots p_q]) \subset \mathcal{R}([p_1 \dots p_q])$$

$$\Rightarrow \hat{A}_1 = \begin{bmatrix} \hat{A}_{11} \\ 0 \end{bmatrix}$$

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Proof continued

(ii) \Rightarrow (iii) If $p^T A = \lambda p^T$ and $p^T B = 0$ then $p^T [B \quad AB \dots A^{n-1}B] = 0$, so (ii) fails.

(iii) \Rightarrow (ii) If $\text{rank}[B \dots A^{n-1}B] = q < n$ then let P be defined as in Theorem 3 and let $p_2^T \hat{A}_{22} = \lambda p_2^T$ and $p^T = [0 \quad p_2^T] P^{-1}$.

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Proof continued

Then

$$p^T B = \begin{bmatrix} 0 & p_2^T \end{bmatrix} \begin{bmatrix} \hat{B}_{11} \\ 0 \end{bmatrix} = 0$$

$$p^T A = \begin{bmatrix} 0 & p_2^T \end{bmatrix} \begin{bmatrix} \hat{A}_{11} & \hat{A}_{12} \\ 0 & \hat{A}_{22} \end{bmatrix} P^{-1} = \lambda \begin{bmatrix} 0 & p_2^T \end{bmatrix} P^{-1} = \lambda p^T$$

so (iii) fails.

$$(iv) \Leftrightarrow \{p^T \lambda - A \quad B\} = 0 \Rightarrow p = 0 \Leftrightarrow (iii)$$

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Observability

The equation

$$\begin{aligned} \dot{x}(t) &= A(t)x(t), & x(t_0) &= x^0 \\ y(t) &= C(t)x(t) \end{aligned}$$

is called *observable on* $[t_0, t_f]$ if any initial state x^0 is uniquely determined by the output $y(t)$ for $t \in [t_0, t_f]$.

It is called *reconstructable on* $[t_0, t_f]$ if the state $x(t_f)$ is uniquely determined by the output $y(t)$ for $t \in [t_0, t_f]$.

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Reachability

The equation

$$\dot{x}(t) = A(t)x(t) + B(t)u(t), \quad x(t_0) = 0$$

is called *reachable on* (t_0, t_f) , if for any x_f , there exists $u(t)$ such that $x(t_f) = x_f$. The matrix function

$$\begin{aligned} W_r(t_0, t_f) &= \int_{t_0}^{t_f} \Phi(t_f, t) B(t) B(t)^T \Phi(t_f, t)^T dt \\ &= \Phi(t_f, t_0) W(t_0, t_f) \Phi(t_f, t_0)^T \end{aligned}$$

is called *reachability Gramian*. If $A(t)$ is continuous, then controllability and reachability are equivalent.

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Observability Gramian

The matrix function

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

is called the *observability Gramian* of the system

$$\begin{aligned} \dot{x}(t) &= A(t)x(t) \\ y(t) &= C(t)x(t) \end{aligned}$$

Operator interpretation:

$$M(t_0, t_f) = L^* L$$

where $L : \mathbf{R}^n \rightarrow L_2^n(t_0, t_f)$ with

$$(Lx^0)(t) = C(t)\Phi(t, t_0)x^0, \quad x^0 \in \mathbf{R}^n$$

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Theorem 4: Observability Criterion

The following two conditions are equivalent

- (i) The system defined by $\{A(t), C(t)\}$ is observable on $[t_0, t_f]$.
- (ii) $M(t_0, t_f) > 0$

Degree of Observability Consider $y = Lx^0 + e$, where e is white noise with unit variance.

$|y - Lx^0|$ is minimized for $\hat{x}^0 = (L^*L)^{-1}L^*y$ and the variance of the estimate is

$$(L^*L)^{-1} = M(t_0, t_f)^{-1}.$$

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Theorem 6: Unobservable State Equation

Suppose that $0 < l < n$ and rank

$$\begin{bmatrix} C \\ CA \\ \vdots \\ CA^{n-1} \end{bmatrix} = l < n.$$

Then there exists an invertible $Q \in \mathbf{R}^{n \times n}$ such that

$$Q^{-1}AQ = \begin{bmatrix} \hat{A}_{11} & 0 \\ \hat{A}_{21} & \hat{A}_{22} \end{bmatrix} \quad CQ = \begin{bmatrix} \hat{C}_{11} & 0 \end{bmatrix}$$

where \hat{A}_{11} is $l \times l$, \hat{C}_{11} is $p \times l$, and the system $\dot{x}(t) = \hat{A}_{11}x(t)$, $y(t) = \hat{C}_{11}x(t)$ is observable.

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Theorem 5: Time-Invariant Observability

The following four conditions are equivalent:

- (i) The system $\dot{x}(t) = Ax(t)$, $y(t) = Cx(t)$ is observable.

$$(ii) \text{ rank } \begin{bmatrix} C \\ \vdots \\ CA^{n-1} \end{bmatrix} = n.$$

- (iii) $\exists p \in \mathbf{C}^n, \lambda \in \mathbf{C} : Ap = \lambda p, Cp = 0$ (PBH-test)

$$(iv) \text{ rank } \begin{bmatrix} \lambda I - A \\ C \end{bmatrix} = n \quad \forall \lambda \in \mathbf{C}. \quad (\text{PBH-test})$$

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Definition: Controllability Index

Let $B = [B_1 \dots B_m]$. For $j = 1, \dots, m$, the *controllability index* ρ_j is the smallest integer such that $A^{\rho_j} B_j$ is linearly independent on the column vectors occurring to the left of it in the controllability matrix

$$\begin{bmatrix} B & AB & \dots & A^{n-1}B \end{bmatrix}$$

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Notation for Controller Form

Given a pair $\{A, B\}$, with controllability indices ρ_1, \dots, ρ_m , define

$$M = \begin{bmatrix} M_1 \\ \vdots \\ M_n \end{bmatrix} := \begin{bmatrix} B_1 & AB_1 & \dots & A^{\rho_1-1}B_1 & \dots & B_m & \dots & A^{\rho_m-1}B_m \end{bmatrix}^{-1}$$

$$P^c = \begin{bmatrix} P_1 \\ \vdots \\ P_m \end{bmatrix}, \quad P_i = \begin{bmatrix} M_{\rho_1+\dots+\rho_i} \\ M_{\rho_1+\dots+\rho_i}A \\ \vdots \\ M_{\rho_1+\dots+\rho_i}A^{\rho_i-1} \end{bmatrix}$$

Notice that it is rather easy to write Matlab code for this.

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Comments

Reverse statevariable ordering in each block to get the SISO controller form of AK.
Reveals Structure, Pole Placement,
Minimal Realizations

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Theorem 7, Controller form

Suppose $\dot{x} = Ax + Bu$ is controllable, with controllability indices ρ_1, \dots, ρ_m . Then the variable transformation $z = P^c x$ gives $\dot{z} = A^c z + B^c u$ with

$$A^c = \quad \quad \quad B^c =$$

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Definition: Observability Index

Let $C^T = [C_1^T \dots C_p^T]^T$. For $j = 1, \dots, p$, the *observability index* η_j is the smallest integer such that $C_j A^{\eta_j}$ is linearly dependent on the row vectors occurring above it in the observability matrix

$$\begin{bmatrix} C \\ CA \\ \vdots \\ CA^{n-1} \end{bmatrix}$$

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Theorem 8: Observer form

Suppose $\dot{x} = Ax, y = Cx$ is observable, with observability indices η_1, \dots, η_p . Then the variable transformation $z = P^o x$ gives $\dot{z} = A^o z + B^o u$ with

$$A^o =$$

$$C^o =$$

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Proof of Theorem 9

Let $P = W_r(-\infty, 0) = \int_0^\infty e^{A\sigma} B B^T e^{A^T \sigma} d\sigma$. Then

$$\begin{aligned} P A^T + A P &= \int_0^\infty \frac{\partial}{\partial \sigma} \left(e^{A\sigma} B B^T e^{A^T \sigma} \right) d\sigma \\ &= \left[e^{A\sigma} B B^T e^{A^T \sigma} \right]_0^\infty = -B B^T \end{aligned}$$

The linear operator

$$L(P) = AP + PA^T$$

has $\mathcal{R}(L) = \mathbf{R}^{n \times n}$ so $\mathcal{N}(L) = \{0\}$ and the solution P is unique. (Lyapunov 1893)

The equation for the observability Gramian is obtained by replacing A, B with A^T, C^T .

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Theorem 9: Time-Invariant Gramian

Let A be exponentially stable. Then, the reachability Gramian $W_c(-\infty, 0)$ equals the unique solution P to the matrix equation

$$P A^T + A P = -B B^T$$

Similarly, the observability Gramian $M(0, \infty)$ equals the solution Q of

$$Q A + A^T Q = -C^T C$$

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Balanced Realization

For the stable system (A, B, C) , with Gramians P and Q , the variable transformation $\hat{x} = T x$ gives

$$\hat{P} = T P T^*, \quad \hat{Q} = T^{-*} Q T^{-1}$$

The choice of quadratic R, T , unitary U and diagonal Σ such that

$$\begin{aligned} Q &= R^* R \quad (\text{Choleski Factorisation}) \\ R P R^* &= U \Sigma^2 U^* \quad (\text{Singular Value Decomposition}) \\ T &= \Sigma^{-1/2} U^* R \end{aligned}$$

gives

$$\hat{P} = \hat{Q} = \Sigma$$

$(\hat{A}, \hat{B}, \hat{C})$ is called a *balanced realization* of the system (A, B, C) .

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Truncated Balanced Realization

Let the states be sorted such that Σ is decreasing. The diagonal elements of Σ measures "how controllable and observable" the corresponding states are. With

$$\hat{A} = \begin{bmatrix} \hat{A}_{11} & \hat{A}_{12} \\ \hat{A}_{21} & \hat{A}_{22} \end{bmatrix}, \quad \hat{B} = \begin{bmatrix} \hat{B}_1 \\ \hat{B}_2 \end{bmatrix}$$

$$\hat{C} = \begin{bmatrix} \hat{C}_1 & \hat{C}_2 \end{bmatrix}, \quad \Sigma = \begin{bmatrix} \Sigma_1 & 0 \\ 0 & \Sigma_2 \end{bmatrix}$$

$(\hat{A}_{11}, \hat{B}_1, \hat{C}_1)$ is called a *truncated balanced realization* of (A, B, C) .

See also R. Johansson: System Modeling & Identification, p236

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Next week

- Realization from Weighing Pattern
- Realization from Impulse Response
- Realization from Markov Parameters
- Minimal Realizations

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Example

$$C(sI - A)^{-1}B = \frac{1 - s}{s^6 + 3s^5 + 5s^4 + 7s^3 + 5s^2 + 3s + 1}$$

$$\Sigma = \text{diag}\{1.98, 1.92, 0.75, 0.33, 0.15, 0.0045\}$$

$$\hat{C}(sI - \hat{A})^{-1}\hat{B} = \frac{0.20s^2 - 0.44s + 0.23}{s^3 + 0.44s^2 + 0.66s + 0.17}$$

(Done with balreal in MATLAB!)

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