

Lecture 3

- More on Adjoints
- Periodic Systems
- Internal Stability
- Discrete Time Systems

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Proof of (7)

$$\begin{aligned}
 Mx = 0 &\Rightarrow M^*Mx = 0 \\
 &\Rightarrow 0 = \langle x, M^*Mx \rangle = \langle Mx, Mx \rangle \\
 &\Rightarrow Mx = 0
 \end{aligned}$$

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Properties of the Adjoint

Let M be a bounded linear operator between two real Hilbert spaces.
Then

$$\begin{aligned}
 (1) \quad M^{**} &= M \\
 (2) \quad [\mathcal{R}(M)]^\perp &= \mathcal{N}(M^*) \\
 (3) \quad \overline{\mathcal{R}(M)} &= [\mathcal{N}(M^*)]^\perp \\
 (4) \quad [\mathcal{R}(M^*)]^\perp &= \mathcal{N}(M) \\
 (5) \quad \overline{\mathcal{R}(M^*)} &= [\mathcal{N}(M)]^\perp \\
 (6) \quad \mathcal{N}(M^*) &= \mathcal{N}(MM^*) \\
 (7) \quad \mathcal{N}(M) &= \mathcal{N}(M^*M)
 \end{aligned}$$

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Example: Shift Operator on l_2

$$\begin{aligned}
 l_2 &= \{x = (x_1, x_2, x_3, \dots) : \sum_{i=1}^{\infty} x_i^2 < \infty\} \\
 x &= (x_1, x_2, x_3, \dots) \\
 y &= (y_1, y_2, y_3, \dots) \\
 Sx &= (0, x_1, x_2, \dots) \\
 S^*y &= (y_2, y_3, \dots) \\
 \langle y, Sx \rangle &= \sum_{i=1}^{\infty} y_{i+1}x_i = \langle S^*y, x \rangle \\
 \mathcal{R}(S) &= \{(0, *, *, \dots)\} \\
 \mathcal{N}(S^*) &= \{(*, 0, 0, \dots)\}
 \end{aligned}$$

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Example: Observability Gramian

For $x^0 \in \mathbf{R}^n$, $y \in \mathbf{L}_2^m[t_0, t_1]$, introduce

$$(Mx^0)(t) = C(t)\Phi(t, t_0)x^0, \quad t \in [t_0, t_1]$$

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

Then the “unobservable” initial states can be computed as

$$\mathcal{N}(M) = \mathcal{N}(M^*M) = \mathcal{N}\left(\int_{t_0}^{t_1} \Phi(t, t_0)^T C(t)^T C(t)\Phi(t, t_0) dt\right)$$

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

will later be called the observability Gramian of the system.

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Proof

Define R and $P(t)$ by

$$e^{RT} = \Phi(T, 0)$$

$$P(t) = \Phi(t, 0)e^{-Rt}$$

Then

$$\Phi(t, \tau) = \Phi(t, 0)\Phi(\tau, 0)^{-1} = P(t)e^{R(t-\tau)}P(\tau)^{-1}$$

$$P(t+T) = \Phi(t+T, 0)e^{-R(t+T)}$$

$$= \Phi(t+T, T)\Phi(T, 0)e^{-RT}e^{-Rt}$$

$$= \Phi(t+T, T)e^{-Rt}$$

$$= \Phi(t, 0)e^{-Rt}$$

$$= P(t)$$

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Floquet Decomposition

Let $A(t)$ be continuous and T -periodic. Then for

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

the transition matrix can be written

$$\Phi(t, \tau) = P(t)e^{R(t-\tau)}P(\tau)^{-1}$$

where $R \in \mathbf{C}^{n \times n}$ is constant and $P(t)$ is continuous and T -periodic.

With the variable transformation $x(t) = P(t)z(t)$, this gives $\dot{z} = Rz$.

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Example: Sinusoidal Input

Consider the equation

$$\dot{x} = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix} x + \begin{bmatrix} 0 \\ \sin t \end{bmatrix} \quad x(0) = x^0$$

Laplace transform:

$$\mathbf{x}_2(s) = C(sI - A)^{-1}(Bu(s) + x^0) = \frac{s}{(1+s^2)^2} + \frac{1}{1+s^2} \begin{bmatrix} 1 & s \end{bmatrix} x^0$$

$$x_2(t) = \frac{t}{2} \sin t + \begin{bmatrix} \sin t & \cos t \end{bmatrix} x^0$$

For what systems does periodic input give periodic solution?

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Condition for Periodic Solutions

Let $A(t)$ be continuous and T -periodic and

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

The following statements are then equivalent:

- (i) No nontrivial T -periodic solution exists for $f \equiv 0$.
- (ii) A unique T -periodic solution exists for every T -periodic f .

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Corollary: LTI Systems

For $A \in \mathbf{R}^{n \times n}$ and

$$\dot{x}(t) = Ax(t) + f(t), \quad x(t_0) = x^0$$

the following statements are equivalent:

- (i) No eigenvalue of A has zero real part.
- (ii) A unique T -periodic solution exists for every T -periodic f .

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Proof

$$\begin{aligned}
 \Leftrightarrow \quad & \Phi(t_0 + T, t_0)x^0 = x^0 \quad \Rightarrow x^0 = 0 \\
 \Leftrightarrow \quad & \det(\Phi(t_0 + T, t_0) - I) \neq 0 \\
 \Leftrightarrow \quad & \forall f : \exists x^0 : \\
 & x^0 = \Phi(t_0 + T, t_0)x^0 + \int_{t_0}^{t_0+T} \Phi(t_0 + T, \tau)f(\tau)d\tau \\
 \Leftrightarrow \quad & \text{(ii)}
 \end{aligned}$$

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Definition of Uniform Stability

The system $\dot{x}(t) = A(t)x(t)$ is called

uniformly stable if $\exists \gamma > 0$ such that

$$|x(t)| < \gamma |x(t_0)| \quad \forall t \geq t_0 \geq 0$$

uniformly asymptotically stable if it is uniformly stable and

$$\forall \delta > 0 : \exists T > 0 :$$

$$|x(t)| < \delta |x(t_0)| \quad \forall t \geq t_0 + T$$

uniformly exponentially stable if $\exists \gamma, \lambda$ such that for $t \geq t_0 \geq 0$ one has

$$|x(t)| < \gamma |x(t_0)| e^{-\lambda(t-t_0)}$$

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Transition Matrix Conditions

The system $\dot{x}(t) = A(t)x(t)$ is

uniformly stable if $\exists \gamma > 0$ such that

$$\|\Phi(t, t_0)\| < \gamma \quad \forall t \geq t_0 \geq 0$$

uniformly asymptotically stable if it is uniformly stable and

$$\forall \delta > 0 : \exists T > 0 :$$

$$\|\Phi(t, t_0)\| < \delta \quad \forall t \geq t_0 + T$$

uniformly exponentially stable if $\exists \gamma, \lambda$ such that for $t \geq t_0 \geq 0$ one has

$$\|\Phi(t, t_0)\| < \gamma e^{-\lambda(t-t_0)}$$

Proof: Use $x(t) = \Phi(t, t_0)x(t_0)$ and the definition of matrix norm.

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Proof of (iii) \Rightarrow (ii)

Let $\alpha = \sup_t \|\Lambda(t)\|$. Then

$$\begin{aligned} \|\Phi(t, \tau)\| &= \left\| I - \int_{\tau}^t \frac{\partial}{\partial \sigma} \Phi(t, \sigma) d\sigma \right\| = \left\| I + \int_{\tau}^t \Phi(t, \sigma) \Lambda(\sigma) d\sigma \right\| \\ &\leq 1 + \alpha \int_{\tau}^t \|\Phi(t, \sigma)\| d\sigma \leq 1 + \alpha \beta \end{aligned}$$

$$\begin{aligned} \|\Phi(t, \tau)\| &= \frac{1}{t-\tau} \int_{\tau}^t \|\Phi(t, \tau)\| d\sigma \\ &\leq \frac{1}{t-\tau} \int_{\tau}^t \|\Phi(t, \sigma)\| \cdot \|\Phi(\sigma, \tau)\| d\sigma \leq \frac{\beta}{t-\tau} (1 + \alpha \beta) \end{aligned}$$

which proves asymptotic stability.

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Criterion for Exponential Stability

For the equation $\dot{x}(t) = A(t)x(t)$ with $\|A(t)\|$ bounded, the following three conditions are equivalent:

- (i) The equation is uniformly exponentially stable.
- (ii) The equation is uniformly asymptotically stable.
- (iii) There exists a $\beta > 0$ such that

$$\int_{\tau}^t \|\Phi(t, \sigma)\| d\sigma \leq \beta \quad \forall \tau \leq t$$

Proof: (i) \Rightarrow (iii) is obvious.

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Proof of (ii) \Rightarrow (i)

Assume asymptotic stability. To prove exponential stability, select $\gamma, T > 0$ such that

$$\|\Phi(t, t_0)\| \leq \gamma \quad \forall t \geq t_0; \quad \|\Phi(t_0 + T, t_0)\| \leq \frac{1}{2} \quad \forall t \geq t_0 + T$$

Then

$$\begin{aligned} \|\Phi(t_0 + kT, t_0)\| &\leq \|\Phi(t_0 + kT, t_0 + (k-1)T)\| \cdots \|\Phi(t_0 + T, t_0)\| \\ &\leq \frac{1}{2^k} \quad k = 1, 2, \dots \\ \|\Phi(t, t_0)\| &\leq \|\Phi(t, t_0 + kT)\| \cdot \|\Phi(t_0 + kT, t_0)\| \leq \frac{\gamma}{2^k} \quad t \geq t_0 + kT \end{aligned}$$

This proves exponential stability with $\lambda = \ln 2 / 2T$.

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Example: Stability by Coordinate Change

Note that the scalar system

$$\dot{x} = x$$

is not stable, but the change of coordinates $z(t) = e^{-2t}x(t)$ gives the stable equation

$$\dot{z} = -z$$

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Stability Preserved

Both uniform stability and uniform exponential stability are preserved under a coordinate transformation $x(t) = P(t)z(t)$ defined by a Lyapunov transformation.

Proof. This follows immediately from the relations

$$\begin{aligned} \|\Phi_x(t, t_0)\| &= \|P(t)\Phi_z(t, t_0)P(t_0)^{-1}\| \\ &\leq \rho^2 \|\Phi_z(t, t_0)\| \\ \|\Phi_z(t, t_0)\| &= \|P(t)^{-1}\Phi_x(t, t_0)P(t_0)\| \\ &\leq \rho^2 \|\Phi_x(t, t_0)\| \end{aligned}$$

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Lyapunov Transformation

An $n \times n$ continuously differentiable matrix function is called a *Lyapunov transformation* if there exist $\rho > 0$ such that

$$\|P(t)\| \leq \rho, \quad \|P(t)^{-1}\| \leq \rho \quad \forall t$$

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Discrete Time Systems

Given a matrix sequence $A(0), A(1), \dots$ the equation

$$x(k+1) = A(k)x(k), \quad x(k_0) = x^0$$

has the unique solution

$$x(k) = \Phi(k, k_0)x^0$$

defined by the *transition matrix*

$$\Phi(k, k_0) = \begin{cases} A(k-1) \cdots A(k_0), & k > k_0 \\ I, & k = k_0 \end{cases}$$

Proof by inspection.

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Input-driven System

The equation

$$\begin{aligned} x(k+1) &= A(k)x(k) + B(k)u(k) \\ x(k_0) &= x^0 \end{aligned}$$

has the unique solution

$$x(k) = \Phi(k, k_0)x^0 + \sum_{k_0}^{k-1} \Phi(k, \sigma)B(\sigma)u(\sigma)$$

Proof by inspection.

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Properties of $\Phi(k, k_0)$

$$\begin{aligned} \Phi(k+1, j) &= A(k)\Phi(k, j), \quad k \geq j \\ \Phi(k, j-1) &= \Phi(k, j)A(j-1), \quad k \geq j \\ \Phi(k, i) &= \Phi(k, j)\Phi(j, i), \quad i \leq j \leq k \end{aligned}$$

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Recursive definition of $\Phi(k, k_0)$

Define $X(k)$ recursively as

$$\begin{aligned} X(k+1) &= A(k)X(k), \quad k \geq k_0 \\ X(k_0) &= I \end{aligned}$$

Then $\Phi(k, k_0) = X(k)$.

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Inversion

If the $n \times n$ matrix $A(k)$ is invertible for each k , then $\Phi(k, j)$ is invertible for each $k \geq j$ and $\Phi(j, k)$ can be defined as

$$\Phi(j, k) = \Phi(k, j)^{-1}$$

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Change of Variables

The equation

$$x(k+1) = A(k)x(k), \quad x(k_0) = x^0$$

with new variables

$$x(k) = P(k)z(k)$$

writes

$$z(k+1) = [P(k+1)^{-1}A(k)P(k)]z(k)$$

and has transition matrix

$$\Phi_z(k, j) = P(k)^{-1}\Phi_x(k, j)P(j)$$

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2-periodic Example

$$A(k) = \begin{bmatrix} (-1)^k & 0 \\ 0 & 1 \end{bmatrix}$$

$$R^2 = \Phi(2, 0) = \begin{bmatrix} -1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$R = \begin{bmatrix} i & 0 \\ 0 & 1 \end{bmatrix}$$

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Floquet Decomposition

Let $A(k)$ be K -periodic. Then for

$$x(k+1) = A(k)x(k), \quad x(k_0) = x^0$$

the transition matrix can be written

$$\Phi(k, j) = P(k)R^{(k-j)}P(j)^{-1}$$

where $R \in \mathbb{C}^{n \times n}$ and $P(k)$ is K -periodic.

With $x(k) = P(k)z(k)$, this gives

$$z(k+1) = Rz(k)$$

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Condition for Periodic Solutions

Let $A(k)$ be K -periodic and

$$x(k+1) = A(k)x(k) + f(k)$$

The following statements are then equivalent:

- (i) No nontrivial K -periodic solution exists for $f \equiv 0$.
- (ii) A unique K -periodic solution exists for every K -periodic f .

Proof. Analogous to continuous time.

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Next Lecture

- Controllability and Observability
- Controller and Observer forms
- Gramians
- Balanced Realizations