

Lecture 2

- Existence of Solutions
- Uniqueness by Grönwall's Inequality
- Transition Matrix properties
- Adjoint Equation

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Transition Matrix

$$\begin{aligned}\Phi(t, t_0) &= I + \int_{t_0}^t A(\sigma_1) d\sigma_1 \\ &+ \int_{t_0}^t A(\sigma_1) \int_{t_0}^{\sigma_1} A(\sigma_2) d\sigma_2 d\sigma_1 \\ &+ \int_{t_0}^t A(\sigma_1) \int_{t_0}^{\sigma_1} A(\sigma_2) \int_{t_0}^{\sigma_2} A(\sigma_3) d\sigma_3 d\sigma_2 d\sigma_1 \\ &\dots\end{aligned}$$

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General Solution

For continuous $A(t)$, the equation

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

has the unique solution

$$x(t) = \Phi(t, t_0)x^0$$

where $\Phi(t, t_0)$ is the *transition matrix*.

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Example: Time-invariant System

For

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

the transition matrix is

$$\begin{aligned}\Phi(t, t_0) &= I + \int_{t_0}^t A d\sigma_1 + \int_{t_0}^t A \int_{t_0}^{\sigma_1} A d\sigma_2 d\sigma_1 + \dots \\ &= I + A(t - t_0) + A^2 \frac{(t - t_0)^2}{2} + A^3 \frac{(t - t_0)^3}{6} + \dots \\ &= e^{A(t - t_0)}\end{aligned}$$

so the solution is

$$x(t) = e^{A(t - t_0)}x^0$$

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Example: Scalar Time-variation

Consider

$$\dot{x} = Aa(t)x(t)$$

The transition matrix is

$$\begin{aligned}\Phi(t, t_0) &= I + A \int_{t_0}^t a(\sigma_1) d\sigma_1 + A^2 \int_{t_0}^t a(\sigma_1) \int_{t_0}^{\sigma_1} a(\sigma_2) d\sigma_2 d\sigma_1 + \dots \\ &= \sum_{k=0}^{\infty} \frac{1}{k!} A^k \left[\int_{t_0}^t a(\sigma) d\sigma \right]^k = \exp \left(A \int_{t_0}^t a(\sigma) d\sigma \right)\end{aligned}$$

Second equality is nontrivial.

(Recall Two Tank Example)

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Existence Proof on $[t_0, t_0 + T]$

To solve:

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

Recursive approximations:

$$\begin{aligned}x_0(t) &= x^0 \\ x_1(t) &= x^0 + \int_0^t A(\sigma)x_0(\sigma)d\sigma \\ x_2(t) &= x^0 + \int_0^t A(\sigma)x_1(\sigma)d\sigma \\ &\vdots\end{aligned}$$

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General Criterion

If

$$A(t) \int_{t_0}^t A(\sigma) d\sigma = \int_{t_0}^t A(\sigma) d\sigma A(t)$$

then

$$\Phi(t, t_0) = \exp \left\{ \int_{t_0}^t A(\sigma) d\sigma \right\}$$

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Existence Proof Continued

Explicit form:

$$\begin{aligned}x_k(t) &= x^0 + \int_0^t A(\sigma_1)x^0 d\sigma_1 \\ &+ \dots + \\ &+ \int_0^t A(\sigma_1) \int_0^{\sigma_1} A(\sigma_2) \dots \int_0^{\sigma_{k-1}} A(\sigma_k)x^0 d\sigma_k \dots d\sigma_1\end{aligned}$$

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Existence Proof Continued

Let

$$\alpha = \max_{t_0 \leq t \leq t_0+T} \|A(t)\|$$

The maximum exists for continuous $A(t)$. Then

$$\begin{aligned} |x_1(t) - x_0(t)| &= \left| \int_{t_0}^t A(\sigma) x^0 d\sigma \right| \\ &\leq \int_{t_0}^t \alpha |x^0| d\sigma \\ &\leq (t - t_0) \alpha |x^0| \end{aligned}$$

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Existence Proof Continued

Hence for $t \in [t_0, T + t_0]$

$$\begin{aligned} \sum_{k=1}^{\infty} |x_k(t) - x_{k-1}(t)| &\leq \sum_{k=0}^{\infty} \frac{\alpha^k T^k}{k!} |x^0| \\ &= e^{\alpha T} |x^0| \end{aligned}$$

and by Cauchy's criterion, the sequence x_1, x_2, \dots converges uniformly on $[t_0, T + t_0]$.

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Existence Proof Continued

$$\begin{aligned} |x_2(t) - x_1(t)| &= \left| \int_{t_0}^t A(\sigma) (x_1(\sigma) - x_0(\sigma)) d\sigma \right| \\ &\leq \int_{t_0}^t \alpha |x_1(\sigma) - x_0(\sigma)| d\sigma \\ &\leq \alpha \int_{t_0}^t (\sigma - t_0) \alpha |x^0| d\sigma \\ &= \alpha^2 (t - t_0)^2 |x^0| / 2 \\ &\vdots \\ |x_k(t) - x_{k-1}(t)| &\leq \frac{\alpha^k (t - t_0)^k}{k!} |x^0| \end{aligned}$$

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Existence Proof Continued

Taking limits in

$$x_k(t) = x^0 + \int_{t_0}^t A(\sigma) x_{k-1}(\sigma) d\sigma$$

gives

$$\begin{aligned} x &= x^0 + \int_{t_0}^t A(\sigma) x(\sigma) d\sigma \\ \dot{x}(t) &= A(t) x(t), \quad x(t_0) = x^0 \end{aligned}$$

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Example: Non-uniqueness

Let

$$A(t) = \begin{cases} 0 & \text{for } t \leq t_0 \\ 1/(t - t_0) & \text{for } t > t_0 \end{cases}$$

Then the equation

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

has two solutions:

$$x \equiv 0 \text{ and } x(t) = \begin{cases} 0 & \text{for } t \leq t_0 \\ t - t_0 & \text{for } t > t_0 \end{cases}$$

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The Grönwall Inequality

Let $\phi(t)$ and $\nu(t) \geq 0$ be continuous and

$$\phi(t) \leq \psi + \int_{t_0}^t \nu(\sigma)\phi(\sigma)d\sigma, \quad t \geq t_0$$

for some constant ψ . Then

$$\phi(t) \leq \psi e^{\int_{t_0}^t \nu(\sigma)d\sigma}, \quad t \geq t_0$$

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Proof of Grönwall

Let

$$r(t) = \psi + \int_{t_0}^t \nu(\sigma)\phi(\sigma)d\sigma$$

Then

$$\dot{r}(t) = \nu(t)\phi(t) \leq \nu(t)r(t)$$

$$[\dot{r}(t) - \nu(t)r(t)]e^{-\int_{t_0}^t \nu(\sigma)d\sigma} \leq 0$$

$$\frac{d}{dt} \left[r(t)e^{-\int_{t_0}^t \nu(\sigma)d\sigma} \right] \leq 0$$

$$r(t)e^{-\int_{t_0}^t \nu(\sigma)d\sigma} \leq r(t_0) = \psi$$

$$\phi(t) \leq r(t) \leq \psi e^{\int_{t_0}^t \nu(\sigma)d\sigma}$$

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Uniqueness of Solutions

Let

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

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

$$z(t) = x_a(t) - x_b(t)$$

Then

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

$$|z(t)| = \left| \int_0^t A(\sigma)z(\sigma)d\sigma \right| \leq \int_0^t \|A(\sigma)\| \cdot |z(\sigma)|d\sigma$$

and by Grönwall's inequality

$$|z(t)| \equiv 0$$

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Characterization of $\Phi(t, t_0)$

The unique solution of the equation

$$\frac{d}{dt}X(t) = A(t)X(t); \quad X(t_0) = I$$

is $X(t) = \Phi(t, t_0)$.

Proof. Let $x(t) = X(t)x^0$. Then

$$\dot{x}(t) = \frac{d}{dt}X(t)x^0 = A(t)X(t)x^0 = A(t)x(t)$$

so

$$x'(t) = \Phi(t, t_0)x^0$$

Hence $\Phi(t, t_0)x^0 = X(t)x^0$ for every x^0 , so

$$\Phi(t, t_0) = X(t)$$

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Input-driven System

The equation

$$\begin{aligned} \dot{x}(t) &= A(t)x(t) + B(t)u(t) \\ x(t_0) &= x^0 \end{aligned}$$

has the unique solution

$$x(t) = \Phi(t, t_0)x^0 + \int_{t_0}^t \Phi(t, \sigma)B(\sigma)u(\sigma)d\sigma$$

Proof: Differentiate!

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Example

$$\dot{x}(t) = \begin{bmatrix} 1 & \cos t \\ 0 & 0 \end{bmatrix} x(t)$$

$$x_2(t) \equiv x_2^0$$

$$\dot{x}_1(t) = x_1(t) + x_2^0 \cos t$$

$$x_1(t) = e^t x_1^0 + \int_0^t e^{t-\sigma} \cos \sigma x_2^0 d\sigma = e^t x_1^0 + \frac{1}{2}(e^t + \sin t - \cos t)x_2^0$$

$$\Phi(t, 0) = \begin{bmatrix} e^t & \frac{1}{2}(e^t + \sin t - \cos t) \\ 0 & 1 \end{bmatrix}$$

$$\Phi(t, t_0) = \begin{bmatrix} e^{t-t_0} & \frac{1}{2}(e^{t-t_0} + \sin(t-t_0) - \cos(t-t_0)) \\ 0 & 1 \end{bmatrix}$$

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Composition Property

For any t, τ, σ , the transition matrix satisfies

$$\Phi(t, \tau) = \Phi(t, \sigma)\Phi(\sigma, \tau)$$

Proof. Let $R(t) = \Phi(t, \sigma)\Phi(\sigma, \tau)$. Then

$$\begin{aligned} \frac{d}{dt}R(t) &= A(t)R(t) \\ R(\sigma) &= \Phi(\sigma, \tau) \end{aligned}$$

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Theorem by Abel-Jacobi-Liouville

Let $A(t)$ be continuous. Then

$$\det \Phi(t, t_0) = \exp \left(\int_{t_0}^t \text{Tr}[A(\sigma)] d\sigma \right)$$

Interpretation: Volume contraction

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Inversion

The transition matrix $\Phi(t, t_0)$ is invertible for any t, t_0 and

$$\Phi(t, t_0)^{-1} = \Phi(t_0, t)$$

Proof. Invertibility by the Abel-Jacobi-Liouville Theorem.

Formula by the composition rule

$$I = \Phi(t, t) = \Phi(t, t_0)\Phi(t_0, t)$$

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Proof of Abel-Jacobi-Liouville

$$\begin{aligned} \frac{d}{dt} \det \Phi(t, t_0) &= \sum_{i,j} \left(\frac{\partial}{\partial \phi_{ij}} \det \Phi(t, t_0) \right) \dot{\phi}_{ij}(t, t_0) \\ &= \sum_{i,j} c_{ij}(t, t_0) \dot{\phi}_{ij}(t, t_0) \\ &= \text{tr} \left(C(t, t_0)^T \dot{\Phi}(t, t_0) \right) \\ &= \text{tr} \left(\Phi(t, t_0) C(t, t_0)^T A(t) \right) \\ &= \text{tr} \left((\det \Phi(t, t_0) I) A(t) \right) \\ &= (\text{tr} A(t)) \det \Phi(t, t_0) \end{aligned}$$

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Change of Variables

The equation

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

with new variables $x(t) = P(t)z(t)$ writes

$$\dot{z}(t) = \left[P(t)^{-1} A(t) P(t) - P(t)^{-1} \dot{P}(t) \right] z(t)$$

Proof.

$$APz = Ax = \dot{x} = \dot{P}z + P\dot{z}$$

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Adjoint Operator

Let $M : X \rightarrow Y$ be a bounded linear operator. The *adjoint operator* $M^* : Y \rightarrow X$ is defined by the identity

$$\langle y, Mx \rangle = \langle M^*y, x \rangle$$

for $x \in X, y \in Y$.

Example: Matrix Adjoint

From the equalities

$$\langle y, Mx \rangle = y^T Mx = (M^T y)^T x = \langle M^T y, x \rangle$$

we see that the adjoint of a matrix is given by its transpose.

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Proof

The definition of \hat{u} gives

$$\langle Mx, M\hat{u} - v \rangle = \langle x, M^*M\hat{u} - M^*v \rangle = 0$$

With $x = u - \hat{u}$, it follows that

$$\begin{aligned} |Mu - v|^2 &= \langle M\hat{u} - v + Mx, M\hat{u} - v + Mx \rangle \\ &= \langle M\hat{u} - v, M\hat{u} - v \rangle + \langle Mx, Mx \rangle \\ &= |M\hat{u} - v|^2 + |Mx|^2 \\ &\geq |M\hat{u} - v|^2 \end{aligned}$$

This completes the proof

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Least Squares Problem I

Minimize $|Mu - v|$ with respect to u .

Solution:

$$\hat{u} = (M^*M)^{-1}M^*v \quad (\text{if } M^*M \text{ invertible})$$

Property:

$$0 = \langle Mx, M\hat{u} - v \rangle \quad \text{for all } x$$

Application: Fewer control signals than objectives

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Least Squares Problem II

Minimize $|u|$ under the constraint $Mu = v$.

Solution:

$$\hat{u} = M^*(MM^*)^{-1}v \quad (\text{if } MM^* \text{ invertible})$$

Property:

$$0 = \langle \hat{u}, \hat{u} - u \rangle \quad \text{for all } u \text{ with } Mu = v$$

Application: Reach certain state with minimal cost

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If $M : \mathbf{R}^n \rightarrow \mathbf{L}_2^m[0, \infty)$ is defined by

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

then the adjoint $M^* : \mathbf{L}_2^m[0, \infty) \rightarrow \mathbf{R}^n$ is given by

$$M^*y = \int_0^\infty \Phi(t, 0)^T y(t) dt$$

Proof:

$$\begin{aligned} \langle y, Mx \rangle &= \int_0^\infty y(t)^T \Phi(t, 0)x dt \\ &= \left(\int_0^\infty \Phi(t, 0)^T y(t) dt \right)^T x = \langle M^*y, x \rangle \end{aligned}$$

- Transition Matrices for Periodic Systems
- Transition Matrices for Discrete Time Systems
- Internal Stability