Linear ODEs

Linear ODEs

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Linear ODEs

Definition (Linear ODE)

A linear ODE is a differential equation taking the form

$$\frac{d}{dt}x = A(t)x + B(t), \qquad (LNH)$$

where $A(t) \in \mathcal{M}_n(\mathbb{R})$ with continuous entries, $B(t) \in \mathbb{R}^n$ with real valued, continuous coefficients, and $x \in \mathbb{R}^n$. The associated IVP takes the form

$$\frac{d}{dt}x = A(t)x + B(t)$$

$$x(t_0) = x_0.$$
(1)

Types of systems

- x' = A(t)x + B(t) is linear nonautonomous (A(t) depends on t) nonhomogeneous (also called *affine* system).
- ► x' = A(t)x is linear nonautonomous homogeneous.
- x' = Ax + B, that is, A(t) ≡ A and B(t) ≡ B, is linear autonomous nonhomogeneous (or affine autonomous).
- ► x' = Ax is linear autonomous homogeneous.
- If A(t + T) = A(t) for some T > 0 and all t, then linear periodic.

Existence of solutions to linear IVPs

Existence and uniqueness of solutions

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Theorem (Existence and Uniqueness)

Solutions to (1) exist and are unique on the whole interval over which A and B are continuous. In particular, if A, B are constant, then solutions exist on \mathbb{R} .

Existence of solutions to linear IVPs

Definition

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A set of n linearly independent solutions of (LH) on J, $\{\phi_1,\ldots,\phi_n\},$ is called a fundamental set of solutions of (LH) and the matrix

$$\Phi = [\phi_1 \ \phi_2 \ \dots \ \phi_n]$$

is called a fundamental matrix of (LH).

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Fundamental matrix

Theorem

Consider the homogeneous system

$$\frac{d}{dt}x = A(t)x,$$
 (LH)

with A(t) defined and continuous on an interval J. The set of solutions of (LH) forms an n-dimensional vector space.

The vector space of solutions

Fundamental matrix solution

Let $X \in \mathcal{M}_n(\mathbb{R})$ with entries $[x_{ij}]$. Define the derivative of X, X' (or $\frac{d}{dt}X$) as

$$\frac{d}{dt}X(t) = [\frac{d}{dt}x_{ij}(t)].$$

The system of n^2 equations

$$\frac{d}{dt}X = A(t)X$$

is called a matrix differential equation.

Theorem

A fundamental matrix Φ of (LH) satisfies the matrix equation X'=A(t)X on the interval J.-

Abel's formula

Theorem

If Φ is a solution of the matrix equation X' = A(t)X on an interval J and $\tau \in J$, then

$$\det\Phi(t) = \det\Phi(\tau) \exp\left(\int_{\tau}^{t} \operatorname{tr} A(s) ds\right)$$

for all
$$t \in J$$
.

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The resolvent matrix

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Definition (Resolvent matrix)

Let $t_0 \in J$ and $\Phi(t)$ be a fundamental matrix solution of (LH) on J. Since the columns of Φ are linearly independent, it follows that $\Phi(t_0)$ is invertible. The resolvent (or state transition matrix, or principal fundamental matrix) of (LH) is then defined as

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

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Proposition

The resolvent matrix satisfies the Chapman-Kolmogorov identities

1. $\mathcal{R}(t, t) = I$, 2. $\mathcal{R}(t, s)\mathcal{R}(s, u) = \mathcal{R}(t, u)$,

as well as the identities

3. $\mathcal{R}(t,s)^{-1} = \mathcal{R}(s,t),$ 4. $\frac{\partial}{\partial \mathcal{R}}\mathcal{R}(t,s) = -\mathcal{R}(t,s)\mathcal{A}(s),$

5.
$$\frac{\partial}{\partial t}\mathcal{R}(t,s) = A(t)\mathcal{R}(t,s).$$

Proposition

 $\mathcal{R}(t, t_0)$ is the only solution in $\mathcal{M}_n(\mathbb{K})$ of the initial value problem

$$\frac{d}{dt}M(t) = A(t)M(t)$$
$$M(t_0) = \mathbb{I},$$

with $M(t) \in M_n(\mathbb{K})$.

Resolvent matrix

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A variation of constants formula

Theorem (Variation of constants formula) Consider the IVP

$$x' = A(t)x + g(t, x)$$
(2a)

$$x(t_0) = x_0$$
, (2b)

where $g : \mathbb{R} \times \mathbb{R}^n \to \mathbb{R}^n$ a smooth function, and let $\mathcal{R}(t, t_0)$ be the resolvent associated to the homogeneous system x' = A(t)x, with \mathcal{R} defined on some interval $J \ni t_0$. Then the solution ϕ of (2) is given by

$$\phi(t) = \mathcal{R}(t, t_0) x_0 + \int_{t_0}^t \mathcal{R}(t, s) g(\phi(s), s) ds, \qquad (3)$$

on some subinterval of J.

Theorem

The solution to the IVP consisting of the linear homogeneous nonautonomous system (LH) with initial condition $x(t_0) = x_0$ is given by

$$\phi(t) = \mathcal{R}(t, t_0)x_0$$

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Consider the autonomous affine system

 $\frac{d}{dt}x = Ax + B,\tag{A}$

and the associated homogeneous autonomous system

 $\frac{d}{dt}x = Ax.$ (L)

Autonomous linear systems

Exponential of a matrix

Definition (Matrix exponential)

Let $A \in \mathcal{M}_n(\mathbb{K})$ with $\mathbb{K} = \mathbb{R}$ or \mathbb{C} . The *exponential* of A, denoted e^{At} , is a matrix in $\mathcal{M}_n(\mathbb{K})$, defined by

$$e^{At} = \mathbb{I} + \sum_{k=1}^{\infty} \frac{t^k}{k!} A^k,$$

where I is the identity matrix in $\mathcal{M}_n(\mathbb{K})$.

Properties of the matrix exponential

- Φ(t) = e^{At} is a fundamental matrix for (L) for t ∈ ℝ.
- ► The resolvent for (L) is given for t ∈ J by

 $\mathcal{R}(t, t_0) = e^{A(t-t_0)} = \Phi(t-t_0).$

- $e^{At_1}e^{At_2} = e^{A(t_1+t_2)}$ for all $t_1, t_2 \in \mathbb{R}$. 1
- $Ae^{At} = e^{At}A$ for all $t \in \mathbb{R}$.
- $(e^{At})^{-1} = e^{-At}$ for all $t \in \mathbb{R}$.
- ▶ The unique solution ϕ of (L) with $\phi(t_0) = x_0$ is given by

$$\phi(t) = e^{A(t-t_0)}x_0$$

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Computing the matrix exponential

Let P be a nonsingular matrix in $\mathcal{M}_n(\mathbb{R})$. We transform the IVP

$$\frac{d}{dt} x = Ax$$
 (L_IVP)
$$x(t_0) = x_0$$

using the transformation x = Py or $y = P^{-1}x$.

The dynamics of y is

$$y' = (P^{-1}x)'$$
$$= P^{-1}x'$$
$$= P^{-1}Ax$$
$$= P^{-1}APy$$

The initial condition is $y_0 = P^{-1}x_0$.

Autonomous linear systems

Diagonalizable case

We have thus transformed IVP (L_IVP) into

$$\frac{d}{dt}y = P^{-1}APy$$

$$y(t_0) = P^{-1}x_0$$
(L_IVP_y)

From the earlier result, we then know that the solution of (L_IVP_y) is given by

$$\psi(t) = e^{P^{-1}AP(t-t_0)}P^{-1}x_0$$

and since x = Py, the solution to (L_IVP) is given by

$$\phi(t) = P e^{P^{-1}AP(t-t_0)} P^{-1} x_0.$$

So everything depends on $P^{-1}AP$.

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Assume P nonsingular in $\mathcal{M}_n(\mathbb{R})$ such that

$$P^{-1}AP = \begin{pmatrix} \lambda_1 & 0 \\ & \ddots \\ 0 & & \lambda_n \end{pmatrix}$$

with all eigenvalues $\lambda_1, \ldots, \lambda_n$ different.

We have

$$e^{P^{-1}AP} = \mathbb{I} + \sum_{k=1}^{\infty} \frac{t^k}{k!} \begin{pmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{pmatrix}^k$$

For a (block) diagonal matrix M of the form

$$M = \begin{pmatrix} m_{11} & 0 \\ & \ddots & \\ 0 & & m_{nn} \end{pmatrix}$$

there holds

Autonomous linear systems

$$M^k = \begin{pmatrix} m_{11}^k & 0 \\ & \ddots \\ 0 & & m_{nn}^k \end{pmatrix}$$

$$\begin{split} e^{P^{-1}AP} &= \mathbb{I} + \sum_{k=1}^{\infty} \frac{t^k}{k!} \begin{pmatrix} \lambda_1^k & 0\\ 0 & \ddots \\ 0 & \lambda_n^k \end{pmatrix} \\ &= \begin{pmatrix} \sum_{k=0}^{\infty} \frac{t^k}{k!} \lambda_1^k & 0\\ 0 & \sum_{k=0}^{\infty} \frac{t^k}{k!} \lambda_n^k \end{pmatrix} \\ &= \begin{pmatrix} e^{\lambda_1 t} & 0\\ 0 & e^{\lambda_n t} \end{pmatrix} \end{split}$$

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Nondiagonalizable case

The Jordan canonical form is

$$P^{-1}AP = \begin{pmatrix} J_0 & & 0 \\ & \ddots & \\ 0 & & J_s \end{pmatrix}$$

so we use the same property as before (but with block matrices now), and

$$e^{P^{-1}APt} = \begin{pmatrix} e^{J_0t} & 0 \\ & \ddots & \\ 0 & e^{J_st} \end{pmatrix}$$

And so the solution to (L_IVP) is given by

$$\phi(t) = P \begin{pmatrix} e^{\lambda_1 t} & 0 \\ & \ddots & \\ 0 & e^{\lambda_n t} \end{pmatrix} P^{-1} x_0.$$

The first block in the Jordan canonical form takes the form

$$J_0 = \begin{pmatrix} \lambda_0 & & 0 \\ & \ddots & \\ 0 & & \lambda_k \end{pmatrix}$$

and thus, as before,

$$e^{J_0 t} = \begin{pmatrix} e^{\lambda_0 t} & 0 \\ & \ddots & \\ 0 & & e^{\lambda_k t} \end{pmatrix}$$

Other blocks J_i are written as

$$J_i = \lambda_{k+i} \mathbb{I} + N_i$$

with I the $n_i \times n_i$ identity and N_i the $n_i \times n_i$ nilpotent matrix

$$N_i = \begin{pmatrix} 0 & 1 & 0 & 0 \\ & \ddots & & \\ 0 & & & 1 \\ 0 & & & 0 \end{pmatrix}$$

 $\lambda_{k+i}\mathbb{I}$ and N_i commute, and thus

$$e^{J_i t} = e^{\lambda_{k+i} t} e^{N_i t}$$

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Since N_i is nilpotent, $N_i^k = 0$ for all $k \ge n_i$, and the series $e^{N_i t}$ terminates, and

$$e^{J_{i}t} = e^{\lambda_{k+i}t} \begin{pmatrix} 1 & t & \cdots & \frac{t^{\eta_{i}-1}}{\eta_{i}-1!} \\ 0 & 1 & \cdots & \frac{t^{\eta_{i}-2}}{(\eta_{i}-2)!} \\ 0 & 1 \end{pmatrix}$$

Theorem

For all $(t_0, x_0) \in \mathbb{R} \times \mathbb{R}^n$, there is a unique solution x(t) to (LJVP) defined for all $t \in \mathbb{R}$. Each coordinate function of x(t) is a linear combination of functions of the form

$$t^k e^{\alpha t} \cos(\beta t)$$
 and $t^k e^{\alpha t} \sin(\beta t)$

where $\alpha + i\beta$ is an eigenvalue of A and k is less than the algebraic multiplicity of the eigenvalue.

Fixed points (equilibria)

Definition

A fixed point (or equilibrium point, or critical point) of an autonomous differential equation

x' = f(x)

is a point p such that f(p) = 0. For a nonautonomous differential equation

$$x' = f(t, x),$$

a fixed point satisfies f(t, p) = 0 for all t.

A fixed point is a solution.

Autonomous linear systems

Definition (Liapunov stable orbit)

The orbit of a point *p* is *Liapunov stable* for a flow ϕ_t if, given $\varepsilon > 0$, there exists $\delta > 0$ such that $d(x, p) < \delta$ implies that $d(\phi_t(x), \phi_t(p)) < \varepsilon$ for all $t \ge 0$. If *p* is a fixed point, then this is written $d(\phi_t(x), p) < \varepsilon$.

Definition (Asymptotically stable orbit)

The orbit of a point p is asymptotically stable (or attracting) for a flow ϕ_t if it is Liapunov stable, and there exists $\delta_1 > 0$ such that $d(x, p) < \delta_1$ implies that $\lim_{t \to \infty} d(\phi_t(x), \phi_t(p)) = 0$. If p is a fixed point, then it is asymptotically stable if it is Liapunov stable and there exists $\delta_1 > 0$ such that $d(x, p) < \delta_1$ implies that $\omega(x) = \{p\}$.

Orbits, limit sets

Orbits and limit sets are defined as for maps.

For the equation x' = f(x), the subset $\{x(t), t \in I\}$, where I is the maximal interval of existence of the solution, is an *orbit*.

If the maximal solution $x(t, x_0)$ of x' = f(x) is defined for all $t \ge 0$, where f is Lipschitz on an open subset V of \mathbb{R}^n , then the omega limit set of x_0 is the subset of V defined by

$$\omega(x_0) = \bigcap_{\tau=0}^{\infty} \left(\overline{\{x(t, x_0) : t \ge \tau\}} \cap V \} \right).$$

Proposition

A point q is in $\omega(x_0)$ iff there exists a sequence $\{t_k\}$ such that $\lim_{k\to\infty} t_k = \infty$ and $\lim_{k\to\infty} x(t_k, x_0) = q \in V$.

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Contracting linear equation

Theorem

Let $A \in \mathcal{M}_n(\mathbb{R})$, and consider the equation (L). Then the following conditions are equivalent.

 There is a norm || ||_A on ℝⁿ and a constant a > 0 such that for any x₀ ∈ ℝⁿ and all t ≥ 0,

$$||e^{At}x_0||_A \le e^{-at}||x_0||_A$$

 There is a norm || ||_B on ℝⁿ and constants a > 0 and C ≥ 1 such that for any x₀ ∈ ℝⁿ and all t ≥ 0,

$$\|e^{At}x_0\|_B \le Ce^{-at}\|x_0\|_B.$$

3. All eigenvalues of A have negative real parts.

In that case, the origin is a *sink* or *attracting*, the flow is a *contraction* (antonyms *source*, *repelling* and *expansion*).

Hyperbolic linear equation

Definition

The linear differential equation (L) is *hyperbolic* if A has no eigenvalue with zero real part.

Autonomous linear systems

We can write

$$\mathbb{R}^n = E^s \oplus E^u \oplus + E^c,$$

and in the case that $E^c =$, then $\mathbb{R}^n = E^s \oplus E^u$ is called a *hyperbolic splitting*.

The symbol \oplus stands for *direct sum*.

Definition (Direct sum)

Let U, V be two subspaces of a vector space X. Then the span of U and V is defined by u + v for $u \in U$ and $v \in V$. If U and V are disjoint except for 0, then the span of U and V is called the *direct* sum of U and V, and is denoted $U \oplus V$.

Definition (Stable eigenspace) The stable eigenspace of $A \in M_n(\mathbb{R})$ is

$$\begin{split} E^s = \text{span} \{ v : v \text{ generalized eigenvector for eigenvalue } \lambda, \\ & \text{with } \Re(\lambda) < 0 \} \end{split}$$

Definition (Center eigenspace)

The center eigenspace of $A \in \mathcal{M}_n(\mathbb{R})$ is

$$\begin{split} E^{c} &= \operatorname{span}\{v: v \text{ generalized eigenvector for eigenvalue } \lambda, \\ & \text{with } \Re(\lambda) = 0 \end{split}$$

Definition (Unstable eigenspace) The unstable eigenspace of $A \in \mathcal{M}_n(\mathbb{R})$ is

$$\begin{split} E^u = \mathsf{span} \{ v : v \text{ generalized eigenvector for eigenvalue } \lambda, \\ & \text{with } \Re(\lambda) > 0 \} \end{split}$$

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Trichotomy

Define

$$\begin{split} V^s &= \{v: \text{there exists } a > 0 \text{ and } C \geq 1 \text{ such that} \\ &\|e^{At}v\| \leq Ce^{-at}\|v\| \text{ for } t \geq 0\}. \\ V^u &= \{v: \text{there exists } a > 0 \text{ and } C \geq 1 \text{ such that} \\ &\|e^{At}v\| \leq Ce^{-a|t|}\|v\| \text{ for } t \leq 0\}. \\ V^c &= \{v: \text{ for all } a > 0, \|e^{At}v\| e^{-a|t|} \to 0 \text{ as } t \to \pm\infty\}. \end{split}$$

Theorem

The following are true.

- 1. The subspaces $E^{s},\,E^{u}$ and E^{c} are invariant under the flow e^{At} .
- There holds that E^s = V^s, E^u = V^u and E^c = V^c, and thus e^{At}|_{E^u} is an exponential expansion, e^{At}|_{E^s} is an exponential contraction, and e^{At}|_{E^c} grows subexponentially as t → ±∞.

Topologically conjugate linear ODEs

Definition (Topologically conjugate flows)

Let ϕ_t and ψ_t be two flows on a space M. ϕ_t and ψ_t are topologically conjugate if there exists an homeomorphism $h:M\to M$ such that

$$h \circ \phi_t(x) = \psi_t \circ h(x)$$

for all $x \in M$ and all $t \in \mathbb{R}$.

Definition (Topologically equivalent flows)

Let ϕ_t and ψ_t be two flows on a space M. ϕ_t and ψ_t are topologically equivalent if there exists an homeomorphism $h: M \to M$ and a function $\alpha: \mathbb{R} \times M \to \mathbb{R}$ such that

 $h \circ \phi_{\alpha(t+s,x)}(x) = \psi_t \circ h(x),$

for all $x \in M$ and all $t \in \mathbb{R}$, and where $\alpha(t, x)$ is monotonically increasing in t for each x and onto all of \mathbb{R} .

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Theorem

Let $A, B \in \mathcal{M}_n(\mathbb{R})$.

- If all eigenvalues of A and B have negative real parts, then the linear flows e^{At} and e^{Bt} are topologically conjugate.
- Assume that the system is hyperbolic, and that the dimension of the stable eigenspace of A is equal to the dimension of the eigenspace of B. Then the linear flows e^{At} and e^{Bt} are topologically conjugate.

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Theorem

Let $A, B \in \mathcal{M}_n(\mathbb{R})$. Assume that e^{At} and e^{Bt} are linearly conjugate, i.e., there exists M with $e^{Bt} = Me^{At}M^{-1}$. Then A and B have the same eigenvalues.