EE263 Autumn 2015 S. Boyd and S. Lall

Solution via Laplace transform and matrix exponential

- ► Laplace transform
- ightharpoonup solving $\dot{x}=Ax$ via Laplace transform
- state transition matrix
- matrix exponential
- qualitative behavior and stability

Laplace transform of matrix valued function

suppose $z: \mathbb{R}_+ \to \mathbb{R}^{p \times q}$

Laplace transform: $Z = \mathcal{L}(z)$, where $Z : D \subseteq \mathbb{C} \to \mathbb{C}^{p \times q}$ is defined by

$$Z(s) = \int_0^\infty e^{-st} z(t) \ dt$$

- ▶ integral of matrix is done term-by-term
- ▶ convention: upper case denotes Laplace transform
- ▶ D is the domain or region of convergence of Z
- ▶ D includes at least $\{s \mid \Re s > a\}$, where a satisfies $|z_{ij}(t)| \leq \alpha e^{at}$ for $t \geq 0$, $i = 1, \ldots, p, \ j = 1, \ldots, q$

Derivative property

$$\mathcal{L}(\dot{z}) = sZ(s) - z(0)$$

to derive, integrate by parts:

$$\mathcal{L}(\dot{z})(s) = \int_0^\infty e^{-st} \dot{z}(t) dt$$
$$= e^{-st} z(t) \Big|_{t=0}^{t \to \infty} + s \int_0^\infty e^{-st} z(t) dt$$
$$= sZ(s) - z(0)$$

Laplace transform solution of $\dot{x} = Ax$

consider continuous-time time-invariant (TI) LDS

$$\dot{x} = Ax$$

for $t \geq 0$, where $x(t) \in \mathbb{R}^n$

- ▶ take Laplace transform: sX(s) x(0) = AX(s)
- ightharpoonup rewrite as (sI-A)X(s)=x(0)
- ▶ hence $X(s) = (sI A)^{-1}x(0)$
- ▶ take inverse transform

$$x(t) = \mathcal{L}^{-1} ((sI - A)^{-1}) x(0)$$

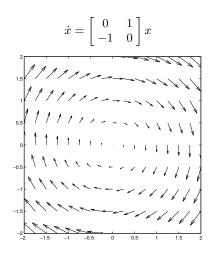
Resolvent and state transition matrix

- $(sI A)^{-1}$ is called the *resolvent* of A
- resolvent defined for $s\in\mathbb{C}$ except eigenvalues of $A,\ i.e.,\ s$ such that $\det(sI-A)=0$
- ▶ $\Phi(t) = \mathcal{L}^{-1}\left((sI A)^{-1}\right)$ is called the *state-transition matrix*; it maps the initial state to the state at time t:

$$x(t) = \Phi(t)x(0)$$

(in particular, state x(t) is a linear function of initial state x(0))

Example 1: Harmonic oscillator



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$$sI - A = \begin{bmatrix} s & -1 \\ 1 & s \end{bmatrix}$$
, so resolvent is

$$(sI - A)^{-1} = \begin{bmatrix} \frac{s}{s^2 + 1} & \frac{1}{s^2 + 1} \\ \frac{-1}{s^2 + 1} & \frac{s}{s^2 + 1} \end{bmatrix}$$

(eigenvalues are $\pm i$)

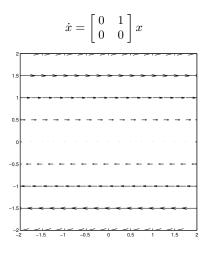
state transition matrix is

$$\Phi(t) = \mathcal{L}^{-1} \left(\begin{bmatrix} \frac{s}{s^2+1} & \frac{1}{s^2+1} \\ \frac{-1}{s^2+1} & \frac{s}{s^2+1} \end{bmatrix} \right) = \begin{bmatrix} \cos t & \sin t \\ -\sin t & \cos t \end{bmatrix}$$

a rotation matrix (-t radians)

so we have
$$x(t) = \begin{bmatrix} \cos t & \sin t \\ -\sin t & \cos t \end{bmatrix} x(0)$$

Example 2: Double integrator



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$$sI-A=\left[egin{array}{cc} s & -1 \\ 0 & s \end{array}
ight]$$
, so resolvent is

$$(sI - A)^{-1} = \begin{bmatrix} \frac{1}{s} & \frac{1}{s^2} \\ 0 & \frac{1}{s} \end{bmatrix}$$

(eigenvalues are 0, 0)

state transition matrix is

$$\Phi(t) = \mathcal{L}^{-1} \left(\begin{bmatrix} \frac{1}{s} & \frac{1}{s^2} \\ 0 & \frac{1}{s} \end{bmatrix} \right) = \begin{bmatrix} 1 & t \\ 0 & 1 \end{bmatrix}$$

so we have
$$x(t) = \begin{bmatrix} 1 & t \\ 0 & 1 \end{bmatrix} x(0)$$

Characteristic polynomial

 $\mathcal{X}(s) = \det(sI - A)$ is called the *characteristic polynomial* of A

- \triangleright $\mathcal{X}(s)$ is a polynomial of degree n, with leading $(i.e., s^n)$ coefficient one
- ightharpoonup roots of $\mathcal X$ are the eigenvalues of A
- $m \mathcal X$ has real coefficients, so eigenvalues are either real or occur in conjugate pairs
- lacktriangle there are n eigenvalues (if we count multiplicity as roots of \mathcal{X})

Eigenvalues of A and poles of resolvent

i,j entry of resolvent can be expressed via Cramer's rule as

$$(-1)^{i+j} \frac{\det \Delta_{ij}}{\det(sI - A)}$$

where Δ_{ij} is sI - A with jth row and ith column deleted

- ▶ det Δ_{ij} is a polynomial of degree less than n, so i, j entry of resolvent has form $f_{ij}(s)/\mathcal{X}(s)$ where f_{ij} is polynomial with degree less than n
- lacktriangle poles of entries of resolvent must be eigenvalues of A
- but not all eigenvalues of A show up as poles of each entry (when there are cancellations between $\det \Delta_{ij}$ and $\mathcal{X}(s)$)

Matrix exponential

define matrix exponential as

$$e^M = I + M + \frac{M^2}{2!} + \cdots$$

- ightharpoonup converges for all $M \in \mathbb{R}^{n \times n}$
- ▶ looks like ordinary power series

$$e^{at} = 1 + ta + \frac{(ta)^2}{2!} + \cdots$$

with square matrices instead of scalars ...

Matrix exponential

$$(I - C)^{-1} = I + C + C^{2} + C^{3} + \cdots$$
 (if series converges)

series expansion of resolvent:

$$(sI - A)^{-1} = (1/s)(I - A/s)^{-1} = \frac{I}{s} + \frac{A}{s^2} + \frac{A^2}{s^3} + \cdots$$

(valid for |s| large enough) so

$$\Phi(t) = \mathcal{L}^{-1} \left((sI - A)^{-1} \right) = I + tA + \frac{(tA)^2}{2!} + \cdots$$

with this definition, state-transition matrix is

$$\Phi(t) = \mathcal{L}^{-1} ((sI - A)^{-1}) = e^{tA}$$

Matrix exponential solution of autonomous LDS

solution of $\dot{x} = Ax$, with $A \in \mathbb{R}^{n \times n}$ and constant, is

$$x(t) = e^{tA}x(0)$$

generalizes scalar case: solution of $\dot{x}=ax$, with $a\in\mathbb{R}$ and constant, is

$$x(t) = e^{ta}x(0)$$

Properties of matrix exponential

- ▶ matrix exponential is *meant* to look like scalar exponential
- some things you'd guess hold for the matrix exponential (by analogy with the scalar exponential) do in fact hold
- but many things you'd guess are wrong

example: you might guess that $e^{A+B}=e^Ae^B$, but it's false (in general)

$$A = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}, \qquad B = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$$

$$e^{A} = \begin{bmatrix} 0.54 & 0.84 \\ -0.84 & 0.54 \end{bmatrix}, \qquad e^{B} = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$$

$$e^{A+B} = \begin{bmatrix} 0.16 & 1.40 \\ -0.70 & 0.16 \end{bmatrix} \neq e^{A}e^{B} = \begin{bmatrix} 0.54 & 1.38 \\ -0.84 & -0.30 \end{bmatrix}$$

Properties of matrix exponential

$$e^{A+B} = e^A e^B$$
 if $AB = BA$

i.e., product rule holds when A and B commute

thus for
$$t,\ s\in\mathbb{R}$$
 , $e^{(tA+sA)}=e^{tA}e^{sA}$ with $s=-t$ we get
$$e^{tA}e^{-tA}=e^{tA-tA}=e^0=I$$

so e^{tA} is nonsingular, with inverse

$$\left(e^{tA}\right)^{-1} = e^{-tA}$$

Example: matrix exponential

let's find
$$e^A$$
, where $A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$

we already found

$$e^{tA} = \mathcal{L}^{-1}(sI - A)^{-1} = \begin{bmatrix} 1 & t \\ 0 & 1 \end{bmatrix}$$

so, plugging in t=1, we get $e^A=\begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$

let's check power series:

$$e^A = I + A + \frac{A^2}{2!} + \dots = I + A$$

since
$$A^2 = A^3 = \cdots = 0$$

Time transfer property

for $\dot{x} = Ax$ we know

$$x(t) = \Phi(t)x(0) = e^{tA}x(0)$$

interpretation: the matrix e^{tA} propagates initial condition into state at time t more generally we have, for any t and τ ,

$$x(\tau + t) = e^{tA}x(\tau)$$

(to see this, apply result above to $z(t) = x(t+\tau)$)

interpretation: the matrix e^{tA} propagates state t seconds forward in time (backward if t<0)

Time transfer property

▶ recall first order (forward Euler) *approximate* state update, for small *t*:

$$x(\tau + t) \approx x(\tau) + t\dot{x}(\tau) = (I + tA)x(\tau)$$

exact solution is

$$x(\tau + t) = e^{tA}x(\tau) = (I + tA + (tA)^2/2! + \cdots)x(\tau)$$

forward Euler is just first two terms in series

Sampling a continuous-time system

suppose
$$\dot{x}=Ax$$
 sample x at times $t_1\leq t_2\leq \cdots$: define $z(k)=x(t_k)$ then $z(k+1)=e^{(t_{k+1}-t_k)A}z(k)$ for uniform sampling $t_{k+1}-t_k=h$, so
$$z(k+1)=e^{hA}z(k),$$

a discrete-time LDS (called discretized version of continuous-time system)

Piecewise constant system

consider *time-varying* LDS $\dot{x} = A(t)x$, with

$$A(t) = \begin{cases} A_0 & 0 \le t < t_1 \\ A_1 & t_1 \le t < t_2 \\ \vdots & \end{cases}$$

where $0 < t_1 < t_2 < \cdots$ (sometimes called jump linear system)

for $t \in [t_i, t_{i+1}]$ we have

$$x(t) = e^{(t-t_i)A_i} \cdots e^{(t_3-t_2)A_2} e^{(t_2-t_1)A_1} e^{t_1A_0} x(0)$$

(matrix on righthand side is called state transition matrix for system, and denoted $\Phi(t)$)

Qualitative behavior of x(t)

suppose $\dot{x} = Ax$, $x(t) \in \mathbb{R}^n$

then
$$x(t) = e^{tA}x(0)$$
; $X(s) = (sI - A)^{-1}x(0)$

ith component $X_i(s)$ has form

$$X_i(s) = \frac{a_i(s)}{\mathcal{X}(s)}$$

where a_i is a polynomial of degree < n

thus the poles of X_i are all eigenvalues of A (but not necessarily the other way around)

Qualitative behavior of x(t)

first assume eigenvalues λ_i are distinct, so $X_i(s)$ cannot have repeated poles then $x_i(t)$ has form

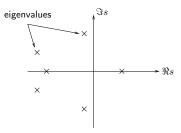
$$x_i(t) = \sum_{j=1}^n \beta_{ij} e^{\lambda_j t}$$

where β_{ij} depend on x(0) (linearly) eigenvalues determine (possible) qualitative behavior of x:

- eigenvalues give exponents that can occur in exponentials
- \blacktriangleright real eigenvalue λ corresponds to an exponentially decaying or growing term $e^{\lambda t}$ in solution
- \blacktriangleright complex eigenvalue $\lambda=\sigma+i\omega$ corresponds to decaying or growing sinusoidal term $e^{\sigma t}\cos(\omega t+\phi)$ in solution

Qualitative behavior of x(t)

- ▶ $\Re \lambda_j$ gives exponential growth rate (if > 0), or exponential decay rate (if < 0) of term
- $\Im \lambda_j$ gives frequency of oscillatory term (if $\neq 0$)



Repeated eigenvalues

now suppose A has repeated eigenvalues, so X_i can have repeated poles express eigenvalues as $\lambda_1,\dots,\lambda_r$ (distinct) with multiplicities n_1,\dots,n_r , respectively $(n_1+\dots+n_r=n)$

then $x_i(t)$ has form

$$x_i(t) = \sum_{j=1}^r p_{ij}(t)e^{\lambda_j t}$$

where $p_{ij}(t)$ is a polynomial of degree $< n_j$ (that depends linearly on x(0))

Stability

we say system $\dot{x}=Ax$ is *stable* if $e^{tA}\to 0$ as $t\to \infty$ meaning:

- \blacktriangleright state x(t) converges to 0, as $t \to \infty$, no matter what x(0) is
- ▶ all trajectories of $\dot{x} = Ax$ converge to 0 as $t \to \infty$

fact: $\dot{x} = Ax$ is stable if and only if all eigenvalues of A have negative real part:

$$\Re \lambda_i < 0, \quad i = 1, \dots, n$$

Stability

the 'if' part is clear since

$$\lim_{t\to\infty}p(t)e^{\lambda t}=0$$

for any polynomial, if $\Re \lambda < 0$

we'll see the 'only if' part next lecture

more generally, $\max_i \Re \lambda_i$ determines the maximum asymptotic logarithmic growth rate of x(t) (or decay, if < 0)