

## Solution via matrix exponential

- ▶ matrix exponential
- ▶ solving  $\dot{x} = Ax$  via matrix exponential
- ▶ state transition matrix

## Matrix exponential

define **matrix exponential** as

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

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

$$e^a = 1 + a + \frac{a^2}{2!} + \dots$$

with square matrices instead of scalars ...

## Properties of the Matrix exponential

▶  $e^0 = I$  where  $0 \in \mathbb{R}^{n \times n}$

▶  $e^{(M^T)} = (e^M)^T$

▶  $\frac{d}{dt}e^{tA} = Ae^{tA} = e^{tA}A$

these follow from the definition

## Matrix exponential solution of autonomous LDS

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

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

the matrix  $e^{tA}$  is called the *state transition matrix*, usually written  $\Phi(t)$

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  $e^{A+B} = e^A e^B$ , but it's false (in general)

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

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

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

## Matrix exponential

- ▶ if  $A$  and  $B$  commute, then  $e^{A+B} = e^A e^B$
- ▶ because if  $AB = BA$ , then  $e^{At} B = B e^{At}$ , so

$$\begin{aligned}\frac{d}{dt} e^{At} e^{Bt} &= A e^{At} e^{Bt} + e^{At} B e^{Bt} \\ &= (A + B) e^{At} e^{Bt}\end{aligned}$$

- ▶ hence  $e^{At} e^{Bt} = e^{(A+B)t}$

## Properties of matrix exponential

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

*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

$$(e^{tA})^{-1} = e^{-tA}$$

## Example: matrix exponential

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

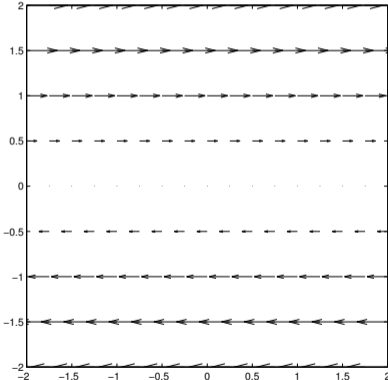
the power series gives

$$\begin{aligned} e^{tA} &= I + tA + \frac{t^2 A^2}{2} + \frac{t^3 A^3}{3!} + \dots \\ &= I + tA \quad \text{since } A^2 = 0 \\ &= \begin{bmatrix} 1 & t \\ 0 & 1 \end{bmatrix} \end{aligned}$$

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

# Example: Double integrator

$$\dot{x} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} x$$



## Example: Harmonic oscillator

$A = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}$ , so state transition matrix is

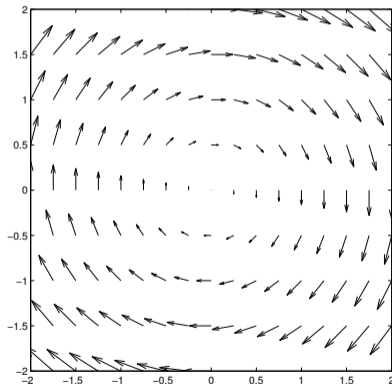
$$\begin{aligned} e^{tA} &= I + tA + \frac{t^2 A^2}{2} + \frac{t^3 A^3}{3!} + \frac{t^4 A^4}{4!} \dots \\ &= \left(1 - \frac{t^2}{2} + \frac{t^4}{4!} - \dots\right) I + \left(t - \frac{t^3}{3!} + \frac{t^5}{5!} - \dots\right) A \\ &= (\cos t)I + (\sin t)A \\ &= \begin{bmatrix} \cos t & \sin t \\ -\sin t & \cos t \end{bmatrix} \end{aligned}$$

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: Harmonic oscillator

$$\dot{x} = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix} x$$



## 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  $\tau$ ,

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

(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  $t < 0$ )

## Numerical integration of continuous system

compute approximate solution of  $\dot{x} = Ax$ ,  $x(0) = x_0$

suppose  $h$  is small time step ( $x$  doesn't change much in  $h$  seconds)

simple ('forward Euler') approximation:

$$x(t+h) \approx x(t) + hx(t) = (I + hA)x(t)$$

by carrying out this recursion (discrete-time LDS), starting at  $x(0) = x_0$ , we get approximation

$$x(kh) \approx (I + hA)^k x(0)$$

(forward Euler is never used in practice)

## Time transfer property

- ▶ 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

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

- ▶ forward Euler is just first two terms in series

## Sampling a continuous-time system

suppose  $\dot{x} = Ax$

sample  $x$  at times  $t_1 \leq t_2 \leq \dots$ : 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

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

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

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

$$x(t) = e^{(t-t_i)A_i} \dots e^{(t_3-t_2)A_2} e^{(t_2-t_1)A_1} e^{t_1 A_0} x(0)$$

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

## Laplace transform of matrix valued function

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

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

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

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

## Derivative property

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

to derive, integrate by parts:

$$\begin{aligned}\mathcal{L}(\dot{z})(s) &= \int_0^{\infty} e^{-st} \dot{z}(t) dt \\ &= e^{-st} z(t) \Big|_{t=0}^{t \rightarrow \infty} + s \int_0^{\infty} e^{-st} z(t) dt \\ &= s\hat{z}(s) - z(0)\end{aligned}$$

## 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:  $s\hat{x}(s) - x(0) = A\hat{x}(s)$
- ▶ rewrite as  $(sI - A)\hat{x}(s) = x(0)$
- ▶ hence  $\hat{x}(s) = (sI - A)^{-1}x(0)$
- ▶ take inverse transform

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

## 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  $\mathbf{det}(sI - A) = 0$
- ▶  $\mathcal{X}(s) = \mathbf{det}(sI - A)$  is called the *characteristic polynomial* of  $A$
- ▶  $\mathcal{X}(s)$  is a polynomial of degree  $n$ , with leading (*i.e.*,  $s^n$ ) coefficient one
- ▶ roots of  $\mathcal{X}$  are the eigenvalues of  $A$

## Matrix exponential

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

► series expansion of resolvent:

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

(valid for  $|s|$  large enough) so

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

► with this definition, state-transition matrix is

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

## Discrete time systems

- ▶ the discrete-time LDS is  $x(k+1) = Ax(k)$
- ▶ with solution  $x(k) = A^k x(0)$
- ▶ The  $\mathcal{Z}$ -transform is  $\hat{x} = \mathcal{Z}(x)$  where

$$\hat{x}(z) = \sum_{k=0}^{\infty} x(k) z^{-k}$$

for all  $z \in \mathbb{C}$  with  $|z|$  sufficiently large

- ▶ if  $y(k) = x(k+1)$  then  $\hat{y}(z) = z(\hat{x}(z) - x(0))$
- ▶ then  $x(k+1) = Ax(k)$  implies  $\hat{x}(z) = z(zI - A)^{-1}x(0)$  and

$$\mathcal{Z}(A^k) = z(zI - A)^{-1}$$