

## Lecture 4.2: Linear Systems Analysis

### Today

1. Matrix exponential
2. Jordan form
3. Stability of linear ODEs
4. Impulse response and convolution
5. Linearization

### Matrix exponential

Claim  $x(t) = e^{At} x_0$  is the solution of  $\dot{x} = Ax$ ,  $x(0) = x_0$

Pf By defn,  $e^{At} = I + At + \frac{1}{2!}(At)^2 + \frac{1}{3!}(At)^3 + \dots$

Differentiating  $x(t)$ , we have

$$\begin{aligned}
 \dot{x} &= \frac{d}{dt} (e^{At} x_0) \\
 &= \frac{d}{dt} \left( I + At + \frac{1}{2!}(At)^2 + \frac{1}{3!}(At)^3 + \dots \right) x_0 \\
 &= \left( 0 + A + A^2 t + \frac{1}{2!} A^3 t^2 + \dots \right) x_0 \\
 &= A \left( I + At + \frac{1}{2!}(At)^2 + \frac{1}{3!}(At)^3 + \dots \right) x_0 \\
 &= A (e^{At}) x_0 = Ax //
 \end{aligned}$$

To explore the behavior of  $\dot{x} = Ax$ , we must understand the structure of  $e^{At}$ . Specifically;

- when does  $e^{At} \rightarrow 0$  as  $t \rightarrow \infty$  (stability)
- what directions does  $e^{At}$  grow/shrink fastest (stability)

## Jordan form

Suppose we change coordinates and write  $z = Tx$  where  $T \in \mathbb{R}^{n \times n}$  is an invertible matrix

$$\dot{z} = T\dot{x} = TAx = TAT^{-1}z$$

If  $x=0$  is stable in  $x$  coordinates then  $z=0$  stable in  $z$  coords  $\Rightarrow$  these systems are equivalent. The transformation  $A \mapsto TAT^{-1}$  is called a similarity transformation.

Q: Can we find  $T \in \mathbb{C}^{n \times n}$  that makes  $TAT^{-1}$  particularly simple/  
useful

A: Jordan form:  $J = TAT^{-1}$

$$J = \begin{bmatrix} J_1 & 0 & \dots & 0 \\ & J_2 & & \vdots \\ & & \ddots & 0 \\ 0 & & & J_q \end{bmatrix} \quad J_i = \begin{bmatrix} \lambda_i & 1 & 0 \\ & \ddots & \vdots \\ 0 & & \lambda_i \end{bmatrix} \in \mathbb{R}^{m_i \times m_i}$$

$\lambda_i = i$ th eigenvalue of  $A$

$m_i =$  size of the  $i$ th Jordan block

### Remarks

1. Jordan form is unique (up to permutation in blocks)
2. Can have multiple blocks w/ same eigenvalue
3. Use this to compute matrix exponential

$$e^{At} = e^{(T^{-1}JT)t} = T^{-1}e^{Jt}T = T^{-1} \begin{bmatrix} e^{J_1} & 0 \\ & \ddots \\ 0 & e^{J_q} \end{bmatrix} T$$

The exponential of a Jordan block determines behavior of a portion of a linear system:

$$e^{J_i t} = e^{\lambda_i t} \begin{bmatrix} 1 & t & \frac{1}{2}t^2 & \dots & \frac{1}{m_i!} t^{m_i-1} \\ & 1 & t & \dots & \frac{1}{(m_i-1)!} t^{m_i-2} \\ & & & \ddots & \vdots \\ 0 & & & & 1 \end{bmatrix}$$

If  $\operatorname{Re}(\lambda_i) < 0 \Rightarrow e^{J_i t} z_0 \rightarrow 0$

If  $\operatorname{Re}(\lambda_i) > 0 \Rightarrow e^{J_i t} z_0 \rightarrow \infty$

If  $\operatorname{Re}(\lambda_i) = 0 \Rightarrow e^{J_i t} z_0 = \begin{bmatrix} 1 & t & \dots & \frac{1}{m_i!} t^{m_i-1} \\ \vdots & \vdots & \ddots & \vdots \\ 1 & & & \end{bmatrix} z_0 \rightarrow \infty$

So non-trivial Jordan blocks give unstable behavior when  $\operatorname{Re}(\lambda_i) = 0$ .

### Determining the size of Jordan blocks

The size of the Jordan blocks,  $m_i$ , are determined by the characteristic polynomial

$$\lambda(s) = \det(sI - A) = (s - \lambda_1)^{m_1} (s - \lambda_2)^{m_2} \dots (s - \lambda_q)^{m_q}$$

$m_i$  = algebraic multiplicity

$\dim \mathcal{N}(\lambda I - A) = \#$  Jordan blocks w/ eigenvalue  $\lambda$

$$\dim \mathcal{N}(\lambda I - A)^k = \sum_{\lambda_i = \lambda} \min\{k, m_i\}$$

Using these relationships, we can figure out the length of the Jordan blocks  $m_1, \dots, m_q$

Example

$$J = \begin{bmatrix} \lambda_1 & & & \\ & \lambda_1 & & \\ & & \lambda_1 & 1 \\ & & & \lambda_1 \end{bmatrix}$$

$$(\lambda_1 I - J) = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix} \Rightarrow \mathcal{N}(\lambda_1 I - J) = 3$$

$$(\lambda_1 I - J)^2 = \begin{bmatrix} 0 & & & \\ & 0 & & \\ & & 0 & \\ & & & 0 \end{bmatrix} \Rightarrow \mathcal{N}(\lambda_1 I - J)^2 = 4 = \sum \min\{2, m_i\}$$

$$\Rightarrow m_1 + m_2 + m_3 = 4$$

$$m_1 = 1, m_2 = 1, m_3 = 2$$

(or some permutation)

See ACM 95/100 for more general treatment.

Computing T

To compute T, we write

$$T = [T_1 | T_2 | \dots | T_q] \quad T_i = [V_{i1} \ V_{i2} \ \dots \ V_{in_i}] \in \mathbb{R}^{n \times n_i}$$

$$T^{-1}AT = J \Rightarrow AT_i = T_i J_i$$

From the form of  $J_i$ , we have

$$AV_{i1} = \lambda_i V_{i1} \quad j=1 \quad \text{eigenvector}$$

$$AV_{ij} = V_{i,j-1} + \lambda_i V_{ij} \quad j=2, \dots, n_i \quad \text{generalized eigenvector}$$

Stability of linear ODEs

Thm A linear system  $\dot{x} = Ax$  is asymptotically stable iff  $\operatorname{Re}(\lambda_i) < 0$  for all eigenvalues  $\lambda_1, \dots, \lambda_n$  of  $A$

PF Write  $z = Tx$  where  $T^{-1}AT = J$  is in Jordan form. Solution is composed of terms of the form  $t^k e^{\lambda_i t}$ . If  $\operatorname{Re}(\lambda_i) < 0$  then  $t^k e^{\lambda_i t} \rightarrow 0$  as  $t \rightarrow \infty$

Remarks

1. If  $\operatorname{Re}(\lambda_i) = 0$  then stability depends on Jordan structure for the eigenvalue

$$\dot{x}_1 = x_2$$

$$\dot{x}_2 = -x_1$$

$$\lambda = \pm j$$

$$J = \begin{bmatrix} j & 0 \\ 0 & -j \end{bmatrix}$$

$$x_1(t) = a \cos \omega t + b \sin \omega t$$

$$x_2(t) = -b \sin \omega t + a \cos \omega t$$

Stable (but not asymptotically stable)

$$\dot{x}_1 = x_2$$

$$\dot{x}_2 = 0$$

$$\lambda = \{0, 0\}$$

$$J = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$$

$$x_1(t) = x_{10} + t x_{20}$$

$$x_2(t) = x_{20}$$

Unstable

2. Can also show  $x=0$  is unstable if any  $\lambda_i$  has positive real part.

# Solutions of linear ODEs

RMM 20 Oct 04  
~~22 Oct 02~~  
 (1) (2)

Claim  $x(t) = e^{At} y_0 + \int_0^t e^{A(t-\tau)} B u(\tau) d\tau$

PA Recall that  $\frac{d}{dt} \left( \int_0^t f(t, \tau) d\tau \right) = f(t, t) + \int_0^t \frac{\partial f}{\partial t}(t, \tau) d\tau$

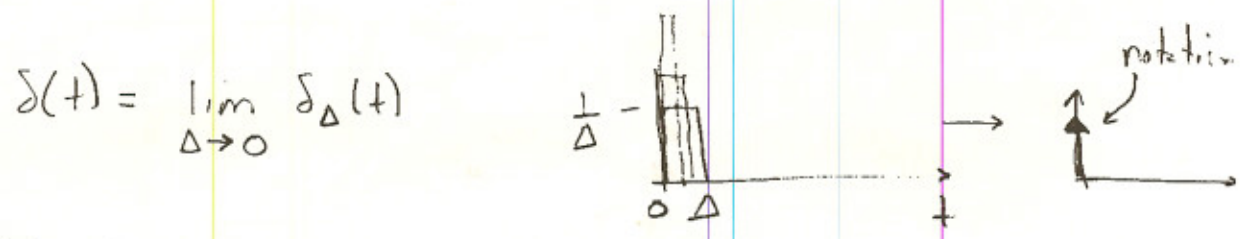
Differentiating the particular sol'n we have

$$\begin{aligned} \dot{x} &= \left[ e^{A(t-\tau)} B u(\tau) \right]_t + \int_0^t \frac{\partial}{\partial t} \left( e^{A(t-\tau)} B u(\tau) \right) d\tau \\ &= B u(t) + \int_0^t A e^{A(t-\tau)} B u(\tau) d\tau \\ &= A \int_0^t e^{A(t-\tau)} B u(\tau) d\tau + B u = Ax + Bu \end{aligned}$$

## Remarks

1. If  $y = Cx + Du$ ,  $y(t) = C e^{At} x_0 + \int_0^t C e^{A(t-\tau)} B u(\tau) d\tau + D u$ .
2.  $h(t) = C e^{A(t-\tau)} B$  is called the impulse response:

Defn Let  $\delta_\Delta(t)$  be a pulse from  $t=0$  to  $t=\Delta$  of height  $\frac{1}{\Delta}$ . The unit impulse  $\delta(t)$  is defined as



3. Can interpret  $\int_0^t \dots$  as the convolution of  $h(t)$  &  $u(t)$

$$(h * u)(t) = \int_0^t h(t-\tau) u(\tau) d\tau$$

Intuition: decompose  $u$  into "sum" of impulse & superimpose response

Linearization

$$\dot{x} = f(x, u)$$

$$y = h(x, u)$$

$$f(x_e, u_e) = 0$$

$$y_e = h(x_e, u_e)$$

Write Taylor series expansion around  $x_e, u_e, y_e$

$$\dot{x} = \underbrace{f(x_e, u_e)}_{\substack{\parallel \\ 0}} + \underbrace{\frac{\partial f}{\partial x} \Big|_{(x_e, u_e)}}_A \underbrace{(x - x_e)}_z + \underbrace{\frac{\partial f}{\partial u} \Big|_{(x_e, u_e)}}_B \underbrace{(u - u_e)}_v + \text{h.o.t.} \quad \text{ignore}$$

$$\dot{z} = Az + Bv + \text{h.o.t.} \leftarrow \text{can ignore for } z \ll 1$$

$$y = h(x_e, u_e) + \underbrace{\frac{\partial h}{\partial x} \Big|_{(x_e, u_e)}}_C \underbrace{(x - x_e)}_z + \underbrace{\frac{\partial h}{\partial u} \Big|_{(x_e, u_e)}}_D \underbrace{(u - u_e)}_v$$

Remarks

↳ Same as small angle approximation

Thm The eq pt  $x_e = 0$  of (14) is asymptotically stable ( $\lim_{k \rightarrow \infty} x[k] = 0$ ) if  $|\lambda_i| < 1$  for all eigenvalues  $\lambda_i \in \lambda(A)$ .

PF (scalar case)

$$x[k] = \lambda x[k]$$

$$x[k] = \lambda^k x[0] \Rightarrow \text{stable if } |\lambda| < 1$$

More general case (diagonalizable): HW #4, problem 3  
Most general case: use Jordan decomposition

HW: derive initial condition response and response to input  $u[k]$ .

Linearization in discrete time

Exactly same as continuous time

$$\begin{aligned} x^+ &= f(x, u) & \rightarrow & \quad x^+ = Ax + Bu \\ y &= h(x, u) & & \quad y = Cx + Du \end{aligned}$$

$$A = \left. \frac{\partial f}{\partial x} \right|_{x_e, u_e} \quad B = \left. \frac{\partial f}{\partial u} \right|_{x_e, u_e}$$

$$C = \left. \frac{\partial h}{\partial x} \right|_{x_e, u_e} \quad D = \left. \frac{\partial h}{\partial u} \right|_{x_e, u_e}$$

To be more careful, redefine

$$\begin{aligned} z &= x - x_e \\ v &= u - u_e \quad \dots \\ w &= y - y_e \end{aligned}$$