

# ENPM-667

## Problem set - 3

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① from spectral norm of a matrix we know that

a) 
$$\|A\| = \max_{\|x\|=1} \|Ax\|$$

Now let's consider that  $y = cx$  where  $c$  is an arbitrary constant and not equal to zero

So in terms of  $y$

$$\|A\| = \max_{\|y\|=1} \|Ay\|$$

which is also equal to

$$\|A\| = \max_{\|y\|=1} \frac{\|Ay\|}{\|y\|}$$

Now put  $y = cx$

$$\Rightarrow \max_{\|y\|=1} \frac{\|A(cx)\|}{\|cx\|}$$

from properties we know that  $\|cx\| = |c| \|x\|$

when  $\|y\| = 1$

$c$  is constant  $c \neq 0$

$\|cx\| = 1$

So  $x \neq 0$

So

$$\|A\| = \max_{\|x\| \neq 0} \frac{\cancel{|x|} \|Ax\|}{\cancel{|x|} \|x\|}$$

Therefore we get the result

$$\|A\| = \max_{\|x\| \neq 0} \frac{\|Ax\|}{\|x\|}$$

⑥ To prove that for any  $n \times 1$  vector  $x$ .

$$\|Ax\| \leq \|A\| \|x\|$$

lets assume that for any vector  $x$  of  $n \times 1$

$$\|Ax\| > \|A\| \|x\|$$

Now,

$$\|Ax\| \cdot \frac{1}{\|x\|} > \|A\|$$

$$\Rightarrow \frac{\|Ax\|}{\|x\|} > \|A\| \longrightarrow \textcircled{1}$$

By definition  $\|A\| = \max_{\|x\| \neq 0} \frac{\|Ax\|}{\|x\|}$

The max is  $\geq \frac{\|Ax\|}{\|x\|}$  so the result  $\textcircled{1}$  is wrong

So, the assumption

$$\|Ax\| > \|Ax\| \|x\| \text{ is not true}$$

Therefore in conclusion we get

$$\|Ax\| \leq \|Ax\| \|x\|$$

② Using the conclusion above prove that for compatible matrices  $A$  and  $B$

$$\|AB\| \leq \|A\| \|B\|$$

from the definition we know that

$$\|A\| = \max_{\|x\|=1} \|Ax\|$$

$$\text{Now consider } \|AB\| = \max_{\|x\| \neq 0} \frac{\|ABx\|}{\|x\|}$$

using the conclusion above

$$\text{we know } \|Ax\| \leq \|A\| \|x\|$$

So applying the same for  $AB$

$$\max_{\|x\|=1} \|AB\| \leq \max_{x \neq 0} \frac{\|A\| \|Bx\|}{\|x\|}$$

$$\|AB\| \leq \max_{x \neq 0} \frac{\|A\| \|B\| \cancel{\|x\|}}{\cancel{\|x\|}}$$

$\Rightarrow$  So we conclude that

$$\|AB\| \leq \|A\| \|B\|$$

② given definition  $\sigma(A) = \max\{\text{eigen values of } A\}$

Prove that  $\sigma(A) \leq \|A\|$  for all  $n \times n$  of  $A$   
from results of the above Problem

we know that  $\|Ax\| \leq \|A\| \|x\| \rightarrow \textcircled{1}$

from properties of matrix we know that

$$AX = \lambda X \rightarrow \text{eigen vector} \rightarrow \textcircled{2}$$

$\downarrow$   
eigen value

So applying  $\textcircled{1}$  on  $\textcircled{2}$

$$\text{we get } |\lambda| \|x\| \leq \|A\| \|x\|$$

Since  $\lambda$  is constant

$$|\lambda| \|x\| \leq \|A\| \|x\|$$

$$|\lambda| \leq \|A\|$$

And spectral radius  $\sigma(A)$  will be given  
as  $\sigma(A) = \max\{|\lambda|\}$

$$\therefore \sigma(A) \leq \|A\|$$

hence Proved.

③ Given that  $A$  is continuously differentiable  $n \times n$  matrix and it is invertible at each  $t$

We know that

$$A(t)A^{-1}(t) = I$$

differentiate both sides

$$\Rightarrow \frac{d}{dt} (A(t)A^{-1}(t)) = \frac{d}{dt} (I) \quad \left[ \because \frac{d}{dt} I = 0 \right]$$

$$\Rightarrow \frac{d}{dt} A(t) \cdot A^{-1}(t) + A(t) \frac{d}{dt} A^{-1}(t) = 0$$

$$\Rightarrow \dot{A}(t) \cdot A^{-1}(t) + A(t) \dot{A}^{-1}(t) = 0$$

$$\Rightarrow \dot{A}(t) \cdot A^{-1}(t) = -A(t) \dot{A}^{-1}(t)$$

multiply on both sides with  $A^{-1}(t)$

$$\Rightarrow -A^{-1}(t) \dot{A}(t) A^{-1}(t) = -\dot{A}^{-1}(t) \cancel{A(t)} \cancel{A^{-1}(t)}$$

$$\Rightarrow \dot{A}^{-1}(t) = -\dot{A}(t) A^{-1}(t)$$

$$\Rightarrow \boxed{\frac{d}{dt} A^{-1}(t) = -\dot{A}(t) A^{-1}(t)}$$

4) given  $\dot{x} = ax(t) + b(t)u(t)$

using laplace both sides

$$L\{\dot{x}\} = L\{ax(t)\} + L\{b(t)u(t)\}$$

$$L\{\dot{x}\} \rightarrow s \cdot x(s) - x(0)$$

$$L\{ax(t)\} \rightarrow a \cdot x(s)$$

$\Rightarrow$  putting these into equation

$$\Rightarrow s \cdot x(s) - x(0) = a \cdot x(s) + L\{b(t)u(t)\}$$

$$\Rightarrow s \cdot x(s) - a \cdot x(s) = x(0) + L\{b(t)u(t)\}$$

$$\Rightarrow (s-a)x(s) = x(0) + L\{b(t)u(t)\}$$

$$\Rightarrow x(s) = \frac{x(0)}{s-a} + \frac{L\{b(t)u(t)\}}{s-a}$$

consider  $L\{b(t)u(t)\} = B(s)U(s)$

$$x(s) = \frac{x(0)}{s-a} + \frac{B(s)U(s)}{s-a}$$

now lets apply inverse Laplace on both sides.

$$\mathcal{L}^{-1}[x(s)] = \mathcal{L}^{-1}\left[\frac{x(0)}{s-a}\right] + \mathcal{L}^{-1}\left[\frac{B(s)u(s)}{s-a}\right]$$

Now consider  $B(s)u(s) = D(s)$

and  $\frac{1}{(s-a)} = G(s)$

$$\mathcal{L}^{-1}[x(s)] = \mathcal{L}^{-1}\left[\frac{x(0)}{s-a}\right] + \mathcal{L}^{-1}[D(s)G(s)]$$

$$x(t) = x_0 e^{at} + \mathcal{L}^{-1}[D(s) \cdot G(s)]$$

$$\text{since } \mathcal{L}^{-1}[D(s) \cdot G(s)] = \int_0^t D(u) G(t-u) du$$

$$\Rightarrow \int_0^t B(\tau) u(\tau) e^{a(t-\tau)} d\tau$$

So,

$$x(t) = \int_0^t B(\tau) u(\tau) e^{a(t-\tau)} d\tau + x(0) e^{at}$$



$$\textcircled{5} \quad y^n(t) + a_{n-1} t^{-1} y^{(n-1)}(t) + a_{n-2} t^{-2} y^{(n-2)}(t) + \dots + a_1 t^{-n+1} y^{(1)}(t) + \dots + a_0 t^{-n} y(t) = 0$$

$$\text{where } y^n(t) = \frac{d^n y(t)}{dt^n}$$

$$\Rightarrow a_0 t^{-n} y(t) + a_1 t^{-n+1} y'(t) + \dots + a_{n-2} t^{-2} y^{(n-2)}(t) + a_{n-1} t^{-1} y^{(n-1)}(t) + y^n(t)$$

The possible state space can be

$$x(t) = \begin{bmatrix} t^{-n} y(t) & t^{-n+1} y'(t) & \dots & t^{-1} y^{(n-1)}(t) \end{bmatrix}^T$$

$$x(t) = \begin{bmatrix} x_1 & x_2 & \dots & x_n \end{bmatrix}^T$$

$$\text{where } x_1 = t^{-n} y(t) \quad x_2 = t^{-n+1} y'(t) \quad \dots \quad x_n = t^{-1} y^{(n-1)}(t)$$

$$\Rightarrow \dot{x}_1 \Rightarrow -n t^{-n-1} y(t) + t^{-n} y'(t)$$

$$\Rightarrow -n t^{-1} x_1 + t^{-1} x_2$$

$$\text{Similarly } \Rightarrow \dot{x}_2 = -(n-1) t^{-1} x_2 + t^{-1} x_3$$

$$x_n = -t^{-1} x_n + t^{-1} y^n(t)$$

$$y_n \Rightarrow -a_{n-1} t^{-1} y^{(n-1)}(t) \dots - a_0 t^{-n} y(t)$$

$$\Rightarrow -a_{n-1} x_n - a_{n-2} x_{n+1} \dots - a_0 x_1$$

$$\dot{x}(t) = \begin{bmatrix} t^{-1}(-n) & t^{-1} & 0 & 0 & \dots & 0 \\ 0 & -(n-1)t^{-1} & t^{-1} & & & \\ 0 & 0 & & & & \\ \vdots & \vdots & & & & \\ -a_0 & -a_1 & \dots & \dots & \dots & -a_{n+1} \end{bmatrix} x(t)$$

$$\tilde{x}(t) = t^{-1} \begin{bmatrix} -n & 1 & 0 & 0 & 0 & \dots & 0 & 0 \\ 0 & -(n-1) & 1 & & & & & \\ 0 & 0 & & & & & & \\ \vdots & \vdots & & & & & & \\ -a_0 & -a_1 & & & & & & -a_{n+1} \end{bmatrix} x(t)$$


  
 $A$

So, we get  $\dot{\tilde{x}}(t) = t^{-1} A \tilde{x}(t)$

$$6) \frac{\partial}{\partial \tau} \Phi(t, \tau) = -\Phi(t, \tau) A(\tau)$$

using the property

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

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

using the above property we can write that

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

now differentiating on both sides

$$\Rightarrow \frac{\partial}{\partial \tau} [\Phi(t, \tau) \cdot \Phi(\tau, t)] = \frac{\partial}{\partial \tau} (I)$$

$$\Rightarrow \frac{\partial}{\partial \tau} \Phi(t, \tau) \Phi(\tau, t) + \Phi(t, \tau) \frac{\partial}{\partial \tau} \Phi(\tau, t) = 0$$

$\Rightarrow$  we know that for Transition matrix

$$\frac{\partial}{\partial t} \Phi(t, t_0) = A(t) \Phi(t, t_0)$$

$$\Rightarrow \frac{\partial}{\partial \tau} \Phi(t, \tau) \Phi(\tau, t) = -\Phi(t, \tau) \cdot A(\tau) \Phi(\tau, t)$$

$$\Rightarrow \frac{\partial}{\partial \tau} \Phi(t, \tau) = -\Phi(t, \tau) \cdot A(\tau) \underbrace{\Phi(\tau, t) \Phi^{-1}(\tau, t)}_{I}$$

So we get

$$\frac{\partial}{\partial \tau} \Phi(t, \tau) = -\Phi(t, \tau) A(\tau)$$

⑦ Given that  $A = \begin{bmatrix} 1 & 0 \\ 1 & \eta(t) \end{bmatrix}$

where  $\eta$  is a bounded and continuous function of  $t$

$$\dot{\vec{x}} = A(t)\vec{x}(t)$$

$$\Rightarrow \begin{bmatrix} \dot{x}_1(t) \\ \dot{x}_2(t) \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 1 & \eta(t) \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix}$$

after multiplication

we get

$$\dot{x}_1 = x_1(t)$$

$$\Rightarrow \frac{dx_1}{dt} = x_1(t)$$

$$\Rightarrow dx_1 = x_1(t)dt \Rightarrow \frac{dx_1}{x_1(t)} = dt$$

on integrating we get

$$\Rightarrow \ln x_1(t) - \ln x_1(t_0) = t - t_0$$

$$\Rightarrow \ln \left( \frac{x_1(t)}{x_1(t_0)} \right) = t - t_0$$

$$\Rightarrow x_1(t) = e^{(t-t_0)} x_1(t_0) \rightarrow \textcircled{1}$$

Similarly  $\dot{x}_2(t) = x_1(t) + x_2(t) \cdot \eta(t)$

$$\Rightarrow \frac{d}{dt} x_2(t) - x_2(t) \eta(t) = x_1(t)$$

we know that for the form  $\frac{dy}{dx} + y \cdot p(x) = Q$

solution is given by

$$x_2(t) \cdot IF = \int (IF) Q(x) dt$$

$$\text{so } IF = e^{\int -\eta(t) dt}$$

$$\Rightarrow x_2(t) e^{\int -\eta(t) dt} = \int x_1(t) e^{-\int \eta(t) dt} dt + C$$

$$\Rightarrow x_2(t) e^{\int -\eta(t) dt} = \int x_1(0) e^{t-t_0} \cdot e^{-\int \eta(t) dt} dt + C$$

$$\Rightarrow x_2(t) e^{-\int \eta(t) dt} = x_1(0) \int e^{t-t_0} e^{-\int \eta(t) dt} dt + C$$

Now lets assume  $\Rightarrow \int e^{t-t_0} e^{-\int \eta(t) dt} = a(t)$

and  $e^{-\int \eta(t) dt} = u(t)$

so simplified equation will be

$$x_2(t) u(t) = x_1(0) a(t) + C$$

$$\text{at } t=t_0 \quad x_2(t_0) = x_2(0)$$

$$\text{So, } c = x_2(0)u(0) - x_1(0)t(0)$$

$$x_2(t)u(t) = x_1(0)a(t) + x_2(0)u(0) - x_1(0)t(0)$$

$$x_2 = \frac{x_1(0)a(t)}{u(t)} + \left( \frac{x_2(0)u(0) - x_1(0)a(0)}{u(t)} \right)$$

$$\Rightarrow \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix} = \begin{bmatrix} e^{t-t_0} & 0 \\ \frac{a(t)}{u(t)} - \frac{a(0)}{u(t)} & \frac{x_2(0)u(0)}{u(t)} \end{bmatrix} \begin{bmatrix} x_1(0) \\ x_2(0) \end{bmatrix}$$

2) Exponential matrix of  $A \Rightarrow \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 6 & 5 & -2 \end{bmatrix}$

first lets diagonalize the matrix  
finding eigen values.

$$|A - \lambda I| = 0$$

$$\Rightarrow \det \begin{bmatrix} -\lambda & 1 & 0 \\ 0 & -\lambda & 1 \\ 6 & 5 & -2-\lambda \end{bmatrix} = 0$$

$$-\lambda(+2\lambda + \lambda^2 - 5) - 1(-6) = 0$$

$$\Rightarrow -2\lambda^2 - \lambda^3 + 5\lambda + 6 = 0$$

$$\Rightarrow \lambda^3 + 2\lambda^2 - 5\lambda - 6 = 0$$

$$\Rightarrow -(\lambda - 2)(\lambda + 1)(\lambda + 3) = 0$$

so  $\lambda = 2, -1, -3 \rightarrow$  these are the eigen values

put  $\lambda = 2$

$$\Rightarrow \begin{bmatrix} -2 & 1 & 0 \\ 0 & -2 & 1 \\ 6 & 5 & 4 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\Delta \begin{cases} -2x_1 + x_2 + 0x_3 = 0 \\ 0x_1 - 2x_2 + x_3 = 0 \\ 6x_1 + 5x_2 - 4x_3 = 0 \end{cases} \rightarrow \text{Apply Cramer's rule}$$

$$\frac{x_1}{\begin{vmatrix} 1 & 0 \\ -2 & 1 \end{vmatrix}} = \frac{-x_2}{\begin{vmatrix} -2 & 0 \\ 0 & 1 \end{vmatrix}} = \frac{x_3}{\begin{vmatrix} -2 & 1 \\ 0 & -2 \end{vmatrix}} = k$$

$$\Rightarrow \frac{x_1}{1} = \frac{-x_2}{-2} = \frac{x_3}{+4} = k$$

$$\Rightarrow \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} +1 \\ +2 \\ +4 \end{bmatrix} \Rightarrow \begin{bmatrix} 1/4 \\ 1/2 \\ 1 \end{bmatrix}$$

for  $\lambda = -1$

$$\Rightarrow \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 6 & 5 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\Rightarrow \begin{cases} x_1 + x_2 + 0x_3 = 0 \\ 0x_1 + x_2 + x_3 = 0 \\ 6x_1 + 5x_2 - x_3 = 0 \end{cases} \rightarrow \text{Cramer's rule}$$

$$\frac{x_1}{\begin{vmatrix} 1 & 0 \\ 0 & 1 \end{vmatrix}} = \frac{-x_2}{\begin{vmatrix} 1 & 0 \\ 0 & 1 \end{vmatrix}} = \frac{x_3}{\begin{vmatrix} 1 & 1 \\ 0 & 1 \end{vmatrix}} = k$$



$$\frac{x_1}{1} = \frac{x_2}{-1} = \frac{x_3}{1}$$

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix} \rightarrow \text{eigen vector}$$

Put  $\lambda = -3$

$$\begin{bmatrix} 3 & 1 & 0 \\ 0 & 3 & 1 \\ 6 & 5 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\begin{aligned} 3x_1 + x_2 + 0x_3 &= 0 \\ 0x_1 + 3x_2 + x_3 &= 0 \\ 6x_1 + 5x_2 + x_3 &= 0 \end{aligned} \rightarrow \text{Cramer's rule}$$

$$\frac{x_1}{\begin{vmatrix} 1 & 0 \\ 3 & 1 \end{vmatrix}} = \frac{-x_2}{\begin{vmatrix} 3 & 0 \\ 0 & 1 \end{vmatrix}} = \frac{x_3}{\begin{vmatrix} 3 & 1 \\ 0 & 3 \end{vmatrix}}$$

$$\frac{x_1}{1} = \frac{-x_2}{3} = \frac{x_3}{9}$$

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 1/9 \\ -1/3 \\ 1 \end{bmatrix} \rightarrow \text{eigen vector}$$

So  $P = \begin{bmatrix} 1 & 1 & 1 \\ 2 & -1 & -3 \\ 4 & 1 & 9 \end{bmatrix} \Rightarrow \text{on computing we get}$

$$D = \begin{bmatrix} 2 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & -3 \end{bmatrix} \quad P^{-1} = \begin{bmatrix} \frac{1}{5} & \frac{4}{15} & \frac{1}{15} \\ 1 & -\frac{1}{6} & -\frac{1}{6} \\ -\frac{1}{5} & -\frac{1}{10} & \frac{1}{10} \end{bmatrix}$$

So  $A = P D P^{-1}$

Now, from exponential series

$$e^{xt} = 1 + xt + \frac{x^2 t^2}{2!} + \dots + \frac{(xt)^n}{n!}$$

$$e^{At} = I + (At) + \frac{(At)^2}{2!} + \dots + \frac{(At)^n}{n!}$$

$$e^{ht} = P D^n P^{-1}$$

$$e^A = P e^{nt} P^{-1}$$

The exponential of a diagonal matrix is

$$e^{Dt} = \begin{bmatrix} e^{2t} & 0 & 0 \\ 0 & e^{-t} & 0 \\ 0 & 0 & e^{-3t} \end{bmatrix}$$

So  $e^{At} = \begin{bmatrix} 1 & 1 & 1 \\ 2 & -1 & -3 \\ 4 & 1 & 9 \end{bmatrix} \begin{bmatrix} e^{2t} & 0 & 0 \\ 0 & e^{-t} & 0 \\ 0 & 0 & e^{-3t} \end{bmatrix} \begin{bmatrix} \frac{1}{5} & \frac{4}{15} & \frac{1}{15} \\ 1 & -\frac{1}{6} & -\frac{1}{6} \\ -\frac{1}{5} & -\frac{1}{10} & \frac{1}{10} \end{bmatrix}$

$$\Rightarrow e^{At} = \begin{bmatrix} e^{2t} & e^{-t} & e^{-3t} \\ 2e^{2t} & -e^{-t} & -3e^{-3t} \\ 4e^{2t} & e^{-t} & 9e^{-3t} \end{bmatrix} \begin{bmatrix} 1/5 & 4/15 & 1/15 \\ 1 & -1/6 & -1/6 \\ -1/5 & -1/10 & 1/10 \end{bmatrix}$$

$$\Rightarrow \frac{1}{30} \begin{bmatrix} 6e^{-3t}(e^{5t} + 5e^{2t} - 1) & e^{-3t}(8e^{5t} - 5e^{2t} - 3) & e^{-3t}(2e^{5t} - 5e^{2t} + 3) \\ 6e^{-3t}(2e^{5t} - 5e^{2t} + 3) & e^{-3t}(16e^{5t} + 5e^{2t} + 9) & e^{-3t}(4e^{5t} + 5e^{2t} - 9) \\ 6e^{-3t}(4e^{5t} + 5e^{2t} - 9) & e^{-3t}(32e^{5t} - 5e^{2t} - 27) & e^{-3t}(8e^{5t} - 5e^{2t} + 27) \end{bmatrix}$$

$$\text{So, } e^{At} = \frac{e^{-3t}}{30} \begin{bmatrix} 6(e^{5t} + 5e^{2t} - 1) & 8e^{5t} - 5e^{2t} - 3 & 2e^{5t} - 5e^{2t} + 3 \\ 6(2e^{5t} - 5e^{2t} + 3) & 16e^{5t} + 5e^{2t} + 9 & 4e^{5t} + 5e^{2t} - 9 \\ 6(4e^{5t} + 5e^{2t} - 9) & 32e^{5t} - 5e^{2t} - 27 & 8e^{5t} - 5e^{2t} + 27 \end{bmatrix}$$

↓  
C

$$\Rightarrow e^{At} = \frac{e^{-3t}}{30} [C]$$