

Solutions to HW 10

EML 4312

Problem 1:

$$\left. \begin{aligned} \dot{x} &= -\sigma x + \sigma y \\ \dot{y} &= -xz + r x - y \\ \dot{z} &= xy - bz \end{aligned} \right\} \textcircled{1}$$

It is not a linear system.

Informally, it is non-linear since xz and xy terms are non-linear functions of states x, y, z .

Formally, it is not linear because it cannot be expressed as $\dot{\mathbf{x}} = \mathbf{A}\mathbf{x}$ (there is no \mathbf{U} , so $\mathbf{B}\mathbf{U}$ is not needed)

If you try, you'll get something like

$$\begin{bmatrix} \dot{x} \\ \dot{y} \\ \dot{z} \end{bmatrix} = \begin{bmatrix} -\sigma & \sigma & 0 \\ -z+r & -1 & 0 \\ y & 0 & -b \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix}$$

\mathbf{A} : is a function of the state z, y ,

whereas for a linear system \mathbf{A} is required to be

~~either a matrix of constants or a matrix of~~

either a constant matrix or a matrix that is an

explicit function of time.

$$2. \quad \frac{Y}{U} = \frac{1}{s^2 + 5s + 26} = G(s)$$

$$\Rightarrow (s^2 + 5s + 26)Y(s) = U(s)$$

$$\Rightarrow \ddot{y} + 5\dot{y} + 26y = u \quad (\text{by taking inverse Laplace transform})$$

$$\text{Define } x_1 = y$$

$$x_2 = \dot{x}_1 (= \dot{y})$$

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

$$\dot{x}_2 = \ddot{y} = u - 5\dot{y} - 26y = u - 5x_2 - 26x_1$$

$$\Rightarrow \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \underbrace{\begin{bmatrix} 0 & 1 \\ -26 & -5 \end{bmatrix}}_A \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u$$

eigenvalues of A : roots of $\det(\lambda I - A) = 0$

$$\Leftrightarrow \det \begin{pmatrix} \lambda & -1 \\ 26 & \lambda + 5 \end{pmatrix} = 0$$

$$\Leftrightarrow \lambda^2 + \lambda 5 + 26 = 0 \quad \text{--- (2)}$$

$$\Leftrightarrow \lambda = \frac{-5 \pm \sqrt{25 - 104}}{2} = \frac{-5 \pm \sqrt{79}j}{2}$$

comparing (2) with the denominator of $G(s)$, we see that eigenvalues of A are precisely the poles of $G(s)$.

[It is no surprise that they are equal, since we have proved in class that they are.]

$$3/ \quad \dot{x} = \underbrace{\begin{bmatrix} 0 & 1 \\ -b & -k \end{bmatrix}}_A x + \underbrace{\begin{bmatrix} 0 \\ 1 \end{bmatrix}}_B u$$

To check (complete) controllability of (A, B) , evaluate

$$P_c = [B \quad AB] = \begin{bmatrix} 0 & 1 \\ 1 & -k \end{bmatrix}$$

To check if the rank of P_c is 2, evaluate

$$\det(P_c) = 0 - 1 = -1 \neq 0$$

$\Rightarrow \text{rank}(P_c) = 2 \Rightarrow (A, B)$ is completely controllable no matter what b and k are.

$$4/ \quad \dot{x} = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & -9.8 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 10.78 & 0 \end{bmatrix} x + \begin{bmatrix} 0 \\ 0.1 \\ 0 \\ -0.1 \end{bmatrix} u \quad \left(\text{this is the cart-n-pendulum system} \right)$$

$$n = 4.$$

$$P_c = [B \quad AB \quad A^2B \quad A^3B]$$

$$= \begin{bmatrix} 0 & 0.1 & 0 & 0.098 \\ 0.1 & 0 & 0.098 & 0 \\ 0 & -0.1 & 0 & -1.078 \\ -0.1 & 0 & -1.078 & 0 \end{bmatrix}$$

$$\text{rank}(P_c) = 4 \quad \text{since} \quad \det(P_c) = 0.0096 \quad (\neq 0)$$

\Rightarrow completely controllable.

Problem 5: A in both cases is the same: $\begin{bmatrix} 0 & 1 \\ -b & -k \end{bmatrix}$

case 1 $c = [1 \ 0]$

$\Rightarrow P_0 = \begin{bmatrix} c \\ cA \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$: rank 2 \Rightarrow completely reconstructible

case 2 $c = [0 \ 1]$

$\Rightarrow P_0 = \begin{bmatrix} c \\ cA \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -b & -k \end{bmatrix}$ $\det(P_0) = b \begin{cases} \neq 0 & \text{if } b \neq 0 \\ = 0 & \text{if } b = 0 \end{cases}$

$\therefore \text{rank}(P_0) = 2$ if and only if $b \neq 0$

$\Rightarrow (A, c)$ is reconstructible if $b \neq 0$, otherwise not.

Problem 6

One option is to compute the transfer function by directly applying the formula

$$G(s) = c(sI - A)^{-1}B$$

But that requires inversion of a 3×3 matrix.

The simpler method is to take Laplace transform of each of the state equations and eliminate all the states except $x_1(s)$, (which is $y(s)$)

We have the state equations as

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

$$\dot{x}_2 = -2x_1 - x_2 + u \quad (\text{and } y = x_1)$$

$$\dot{x}_3 = ax_3$$

Taking Laplace transform, we get

$$sX_1(s) = X_2(s)$$

$$sX_2(s) = -2X_1(s) - X_2(s) + U(s)$$

$$sX_3(s) = aX_3(s)$$

$$\Rightarrow U(s) = (s^2 + s + 2)X_1(s)$$

$$\Rightarrow \frac{X_1(s)}{U(s)} = \frac{1}{s^2 + s + 2}$$

Since $Y(s) = X_1(s)$, we get

$$\frac{Y(s)}{U(s)} = \frac{X_1(s)}{U(s)} = \frac{1}{s^2 + s + 2}$$

Note: x_3 plays no role; since neither the input u affects x_3 (even indirectly through another state) nor x_3 affect the output y (even indirectly).

That is why the transfer function from $U(s)$ to $Y(s)$ is second order (only the dynamics of x_1 & x_2 have an impact in the T.F.) even though the state dimension is three.

Problem 7

states are position and velocity.

call $p(t) =$ position

$v(t) =$ velocity

$$\dot{p} = v \quad \leftarrow \text{(by definition of } p \text{ \& } v)$$

$$\dot{v} = -av + b\theta \quad \leftarrow \text{given}$$

$$\Rightarrow \begin{bmatrix} \dot{p} \\ \dot{v} \end{bmatrix} = \underbrace{\begin{bmatrix} 0 & 1 \\ 0 & -a \end{bmatrix}}_A \begin{bmatrix} p \\ v \end{bmatrix} + \begin{bmatrix} 0 \\ b \end{bmatrix} \theta$$

$$X = \text{state vector} = \begin{bmatrix} p \\ v \end{bmatrix}, \quad U = \text{input vector} = [\theta]$$

$$Y = \text{measured output} = v \quad (\text{given})$$

$$= \underbrace{\begin{bmatrix} 0 & 1 \end{bmatrix}}_C \begin{bmatrix} p \\ v \end{bmatrix}$$

reconstructibility :

$$P_0 = \begin{bmatrix} C \\ CA \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 0 & -a \end{bmatrix} \Rightarrow \det(P_0) = 0 \Rightarrow \text{rank}(P_0) < 2$$

$\Rightarrow (A, C)$ is not completely reconstructible.

Note 1 : The elements of the state vector X are usually written as x_1, x_2, \dots, x_n , but there is nothing sacred about that notation. In this example, p and v are the states.

Note 2: It makes sense that the state vector $x(t)$ cannot be uniquely determined since ~~the~~ the velocity is measured, since the position cannot be determined from measurement of velocity. If the position were measured, the system would have been found to be reconstructible. Think about it.

Problem 8

$$P_o = \begin{bmatrix} C \\ CA \\ CA^2 \end{bmatrix} \quad C = [1 \ 0 \ 0]$$

$$A = \begin{bmatrix} 0 & 1 & 0 \\ -2 & -1 & 0 \\ 0 & 0 & a \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ -2 & -1 & 0 \end{bmatrix}$$

$\det(P_o) = 0$ (since the last column is $\underline{0}$, you can see the det is 0 without computing it)

$$\Rightarrow \text{rank}(P_o) < 3$$

$\Rightarrow (A, C)$ is not completely reconstructible.

Note: Again, the answer makes intuitive sense since the state x_3 does not even indirectly affect the output y . There is no hope of determining x_3 from y and u .

Problem 9.1

For a system $\dot{x} = Ax + Bu$, $y = Cx$,
the poles of the T.F. from u to y are
the eigenvalues of A .

$$\text{eig}(A) = \text{# roots of } \det(\lambda I - A) = 0$$

$$\det(\lambda I - A) = \det\left(\begin{bmatrix} \lambda & 0 \\ 0 & \lambda \end{bmatrix} - \begin{bmatrix} 0 & 1 \\ -0.02 & -0.2 \end{bmatrix}\right)$$

$$= \det\begin{pmatrix} \lambda & -1 \\ 0.02 & \lambda + 0.2 \end{pmatrix}$$

$$= \lambda(\lambda + 0.2) + 0.02$$

$$= \lambda^2 + 0.2\lambda + 0.02$$

$$\therefore \text{eigenvalues: } \lambda^2 + 0.2\lambda + 0.02 = 0$$

$$\Rightarrow \lambda = \frac{-0.2 \pm \sqrt{(0.2)^2 - 4 \times 0.02}}{2}$$

$$= -0.1 \pm 0.1j$$



9.2 Closed loop eigenvalues can be placed anywhere
we want if ~~the~~ (A, B) is completely controllable.

check:

$$P_c = [B \quad AB] = \begin{bmatrix} 0 & 1 \\ 1 & -0.2 \end{bmatrix} \quad : \text{rank } 2. \\ \Rightarrow \text{controllable.}$$

So, yes, the closed loop eigenvalues can be placed at
 $-0.1 \pm 0.1j$ (or anywhere we want) by state feedback.

To design the state feedback controller ($u = -Kx$), that is, to design the gain K , so that the eigenvalues of $A - BK$ are at $-\frac{1}{1} \pm 0.1j$, we use Ackerman's formula.

step 1: define $q(s) = (s - \lambda_1^{(d)})(s - \lambda_2^{(d)})$

$$= \left[s - \left(\frac{-1}{1} + 0.1j \right) \right] \left[s - \left(\frac{-1}{1} - 0.1j \right) \right]$$

$$= s^2 + 2s + 1.01$$

$$\Rightarrow q(A) = A^2 + 2A + 1.01 \cdot I_{2 \times 2}$$

$$= \begin{bmatrix} -0.02 & -0.2 \\ 0.004 & 0.02 \end{bmatrix} + 2 \begin{bmatrix} 0 & 1 \\ -0.02 & -0.2 \end{bmatrix} + 1.01 \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$= \begin{bmatrix} 0.99 & 1.8 \\ -0.036 & 0.63 \end{bmatrix}$$

step 2: determine $P_e^{-1} = \begin{bmatrix} 0 & 1 \\ 1 & -0.2 \end{bmatrix}^{-1}$

$$= \frac{1}{-1} \begin{bmatrix} -0.2 & -1 \\ -1 & 0 \end{bmatrix}$$

$$= \begin{bmatrix} 0.2 & 1 \\ 1 & 0 \end{bmatrix}$$

step 3: Apply Ackerman's formula: $K = [0 \ 1] P_e^{-1} q(A)$

$$\Rightarrow K = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 0.2 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 0.99 & 1.8 \\ -0.036 & 0.63 \end{bmatrix}$$

$$\Rightarrow K = \begin{bmatrix} 1 & 0 \end{bmatrix} \begin{bmatrix} 0.99 & 1.8 \\ -0.036 & 0.63 \end{bmatrix}$$

$$= \begin{bmatrix} 0.99 & 1.8 \end{bmatrix}$$

check: Are the eigenvalues of $A-BK$ really at $-1 \pm 0.1j$?

$$A-BK = \begin{bmatrix} 0 & 1 \\ -0.02 & -0.2 \end{bmatrix} - \begin{bmatrix} 0 \\ 1 \end{bmatrix} \begin{bmatrix} 0.99 & 1.8 \end{bmatrix}$$

$$= \begin{bmatrix} 0 & 1 \\ -0.02 & -0.2 \end{bmatrix} - \begin{bmatrix} 0 & 0 \\ 0.99 & 1.8 \end{bmatrix}$$

$$= \begin{bmatrix} 0 & 1 \\ -1.01 & -2 \end{bmatrix}$$

eigenvalues of $A-BK$ can be computed now to show that they are indeed $-1 \pm 0.1j$

Problem 10

$$m\ddot{h} = -F(i, h) + mg + w$$

$$F(i, h) = \frac{\mu_0 N^2 A}{4} \cdot \frac{i^2}{h^2}$$

10.1 to determine i_0 so that ~~$\ddot{h} = 0$~~ ~~$\dot{h} = 0$~~ ~~$h = 0$~~

$$mg = F(i_0, h_0), \text{ set}$$

$$\frac{\mu_0 N^2 A}{4} \frac{i_0^2}{h_0^2} = mg$$

$$\Rightarrow \boxed{i_0 = h_0 \sqrt{mg \cdot \frac{\mu_0 N^2 A}{4}}} \quad \left(\begin{array}{l} \text{ignore the} \\ \text{negative root} \end{array} \right)$$

(= 26.061 Amp)

When $w(t) \equiv 0$ ($\neq 0$ for all t), $i(t) = i_0 \quad t \geq \bar{t}$,

and $h(\bar{t}) = 0$, then

$$m \ddot{h}(\bar{t}) = -F(i(\bar{t}), h(\bar{t})) + mg + 0$$

$$= -F(i_0, h_0) + mg$$

$$= 0. \quad (\text{since } i_0 \text{ by definition}$$

satisfies $mg = F(i_0, h_0)$)

Since $\ddot{h}(\bar{t}) = 0$, both ~~\dot{h}~~ $h(t)$ and $\dot{h}(t)$ will not change

for $t \geq \bar{t} \Rightarrow h(t) = h_0$ for all $t \geq \bar{t}$.

10.2 $\tilde{i} = i - i_0$, $x_1 = h$, $x_2 = \dot{h}$: states
 \tilde{i} , w : external inputs.

to find state space description, find

$$\dot{x}_1 = f_1(x_1, x_2, \tilde{i}, w) ?$$

$$\dot{x}_2 = f_2(x_1, x_2, \tilde{i}, w) ?$$

$$\dot{x}_1 = \dot{h} = x_2 \text{ by definition. } \leftarrow \leftarrow f_1$$

$$\dot{x}_2 = \ddot{h} = \frac{1}{m} \left(- \frac{\mu_0 N^2 A}{4} \cdot \frac{i^2}{h^2} + mg + w \right)$$

$$= \frac{1}{m} \left(- \frac{\mu_0 N^2 A}{4} \frac{(i_0^2 + \tilde{i}^2)}{x_1^2} + mg + w \right). \leftarrow \leftarrow f_2$$

this is not a linear model, since it cannot be expressed as $\dot{x} = Ax + Bu$, where

$$x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \text{ and } u = \begin{bmatrix} \tilde{i} \\ w \end{bmatrix}. \text{ The reason is that}$$

f_2 is a non-linear function of x_1, \tilde{i} .

10.3 To find equilibria of $\dot{x} = f(x, u)$,

$$\text{set } f(x^*, 0) = 0 \text{ and solve for } x^* = \begin{bmatrix} x_1^* \\ x_2^* \end{bmatrix}$$

$$\Rightarrow \left. \begin{array}{l} f_1(x_1^*, x_2^*, 0, 0) = 0 \Rightarrow \boxed{x_2^* = 0} \\ f_2(x_1^*, x_2^*, 0, 0) = 0 \Rightarrow - \frac{\mu_0 N^2 A}{4} \frac{i_0^2}{x_1^{*2}} + mg = 0 \end{array} \right\}$$

$\Rightarrow x_1^* = h_0$ (use the formula for i_0 to figure this out. I am also ignoring $x_1^* = -h_0$, which is another solution).

\Rightarrow An equilibrium point of the system is

$$X^* = \begin{bmatrix} h_0 \\ 0 \end{bmatrix}.$$

10.4: $\tilde{x}_1 = x_1 - x_1^* = x_1 - h_0$

$$\tilde{x}_2 = x_2 - x_2^* = x_2$$

To linearize, we ~~express~~ find equations for the ~~derivations~~ $\dot{\tilde{x}}_1$ and $\dot{\tilde{x}}_2$ and simplify them by assuming that the deviations \tilde{x}_1 and \tilde{x}_2 are small.

$$\dot{\tilde{x}}_1 = \frac{d}{dt}(x_1 - x_1^*) = \dot{x}_1 = x_2 = \tilde{x}_2 \Rightarrow \boxed{\dot{\tilde{x}}_1 = \tilde{x}_2} \quad \text{--- (a)}$$

$$\dot{\tilde{x}}_2 = \frac{1}{m} \left(-\frac{\mu_0 N^2 A}{4} \frac{(i_0 + \tilde{i})^2}{(h_0 + \tilde{x}_1)^2} + mg + w \right)$$

$$= \frac{1}{m} \left[-\frac{\mu_0 N^2 A}{4} \frac{i_0^2 (1 + \tilde{i}/i_0)^2}{h_0^2 (1 + \tilde{x}_1/h_0)^2} + mg + w \right]$$

$$= \frac{1}{m} \left[-\frac{\mu_0 N^2 A i_0^2}{4 h_0^2} \left(1 + \frac{\tilde{i}}{i_0}\right)^2 \left(1 + \frac{\tilde{x}_1}{h_0}\right)^{-2} + mg + w \right]$$

$$= \frac{1}{m} \left[-mg \left(1 + 2\frac{\tilde{i}}{i_0} + \frac{\tilde{i}^2}{i_0^2}\right) \left(1 - 2\frac{\tilde{x}_1}{h_0} + O\left(\frac{\tilde{x}_1^2}{h_0^2}\right)\right) + mg + w \right]$$

\uparrow since $F_0 = mg$

$$\Rightarrow \ddot{\tilde{x}}_2 = \frac{1}{m} \left[-mg \left(1 - 2 \frac{\tilde{x}_1}{h_0} + 2 \frac{\tilde{i}}{i_0} + \text{terms involving } \left(\frac{\tilde{i}}{i_0} \right)^2, \left(\frac{\tilde{x}_1}{h_0} \right)^2, \left(\frac{\tilde{i}}{i_0}, \frac{\tilde{x}_1}{h_0} \right), \text{ and even higher order} \right) + mg + w \right]$$

$$= \frac{1}{m} \left[-\cancel{mg} + 2mg \left(\frac{\tilde{x}_1}{h_0} + \frac{\tilde{i}}{i_0} \right) + O\left(\left(\frac{\tilde{x}_1}{h_0} \right)^2, \left(\frac{\tilde{i}}{i_0} \right)^2, \dots \right) + \cancel{mg} + w \right]$$

$$\approx \frac{1}{m} \left[2mg \left(\frac{\tilde{x}_1}{h_0} - \frac{\tilde{i}}{i_0} \right) \right] + \frac{w}{m} \left(\begin{array}{l} \text{assuming } \left| \frac{\tilde{x}_1}{h_0} \right| \ll 1, \\ \left| \frac{\tilde{i}}{i_0} \right| \ll 1 \end{array} \right)$$

∴ linear approximation of $\ddot{\tilde{x}}_2$ is :

$$\boxed{\ddot{\tilde{x}}_2 = \frac{2g}{h_0} \tilde{x}_1 - \frac{2g}{i_0} \tilde{i}} \quad \text{--- (b)}$$

(a) and (b) can now be written in the form:

$$\begin{bmatrix} \ddot{\tilde{x}}_1 \\ \ddot{\tilde{x}}_2 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ \frac{2g}{h_0} & 0 \end{bmatrix} \begin{bmatrix} \tilde{x}_1 \\ \tilde{x}_2 \end{bmatrix} + \begin{bmatrix} 0 \\ -\frac{2g}{i_0} \end{bmatrix} \tilde{i} + \begin{bmatrix} 0 \\ \frac{1}{m} \end{bmatrix} w$$

$$= \begin{bmatrix} 0 & 1 \\ \frac{2g}{h_0} & 0 \end{bmatrix} \begin{bmatrix} \tilde{x}_1 \\ \tilde{x}_2 \end{bmatrix} + \begin{bmatrix} 0 & 0 \\ -\frac{2g}{i_0} & \frac{1}{m} \end{bmatrix} \begin{bmatrix} \tilde{i} \\ w \end{bmatrix}$$

Since we did not decide what is being measured, there is no y . (So, no $y = cx$)

10.5

In matlab, a LTI system $\dot{x} = Ax + Bu$, $y = Cx + Du$ can be defined using `ss(A, B, C, D)`.

In our case, since output y is not specified, we arbitrarily choose $y = x$ (full state as measured output)

so in matlab you use `C = eye(2)`, `D = zeros(2)`.

since you are asked to simulate with $i(t) = i_0$ for all t , it means $\tilde{i}(t) = 0$ for all time, (and also $w(t) = 0$). So $W(t) = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$ for all t .

The system is unstable, (eigenvalues of A are ± 49.4975) so the states will blow up.

see the code provided in the website. ~~###~~

Note: it might appear that states stay close to 0 until the very end, then at the end they shoot up to very large numbers. This impression is created because of the scale of the y -axis in the plots.

If you plot in log scale you'll see that the states are continuously increasing, at all times.

(see the code)

Problem 11

$$\frac{Y}{U} = \frac{1}{s^2 + 2s}$$

$$\Rightarrow \ddot{y} + 2\dot{y} = u$$

states $x_1 = y$

$$x_2 = \dot{y}$$

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

$$\dot{x}_2 = \ddot{y} = u - 2\dot{y} = u - 2x_2$$

$$\Rightarrow \begin{pmatrix} \dot{x}_1 \\ \dot{x}_2 \end{pmatrix} = \underbrace{\begin{pmatrix} 0 & 1 \\ 0 & -2 \end{pmatrix}}_A \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} + \begin{pmatrix} 0 \\ 1 \end{pmatrix} u$$

$$y = x_1 = \underbrace{\begin{pmatrix} 1 & 0 \end{pmatrix}}_C \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

observer: $\hat{x}(t)$: estimate of the state $x(t)$

$$\dot{\hat{x}} = A \hat{x} + Bu + L(y - C\hat{x})$$

$$\begin{pmatrix} \dot{\hat{x}}_1 \\ \dot{\hat{x}}_2 \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ 0 & -2 \end{pmatrix} \begin{pmatrix} \hat{x}_1 \\ \hat{x}_2 \end{pmatrix} + \begin{pmatrix} 0 \\ 1 \end{pmatrix} u + \begin{pmatrix} l_1 \\ l_2 \end{pmatrix} (y - \hat{x}_1), \quad \begin{pmatrix} \hat{x}(0) \\ \text{arbitrary} \end{pmatrix}$$

L , (that is, l_1 and l_2) need to be designed so that the eigenvalues of $A - LC$ are at $-3 \pm 4j$

Use Ackerman's formula for observer design:

$$L = q(A) \cdot P_0^{-1} \cdot \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

steps : find $q(\lambda)$, P_0^{-1} & then compute L

$$q(s) = (s - \lambda_1^d)(s - \lambda_2^d)$$

↑ ↑
desired eigenvalues of $A - LC$

$$\Rightarrow \lambda_1^d = -3 + 4j \quad (\text{specified})$$
$$\lambda_2^d = -3 - 4j$$

$$= (s + 3 + 4j)(s + 3 - 4j) = s^2 + 6s + 25$$

$$\Rightarrow q(A) = A^2 + 6A + 25I = \begin{bmatrix} 0 & 1 \\ 0 & -2 \end{bmatrix}^2 + 6 \begin{bmatrix} 0 & 1 \\ 0 & -2 \end{bmatrix} + \begin{bmatrix} 25 & 0 \\ 0 & 25 \end{bmatrix}$$

$$= \begin{bmatrix} 0 & -2 \\ 0 & 4 \end{bmatrix} + \begin{bmatrix} 0 & 6 \\ 0 & -12 \end{bmatrix} + \begin{bmatrix} 25 & 0 \\ 0 & 25 \end{bmatrix}$$

$$= \begin{bmatrix} 25 & 4 \\ 0 & 17 \end{bmatrix}$$

$$P_0 = \begin{bmatrix} C \\ CA \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \Rightarrow P_0^{-1} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$\Rightarrow L = \begin{bmatrix} 25 & 4 \\ 0 & 17 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 25 & 4 \\ 0 & 17 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

$$= \begin{bmatrix} 4 \\ 17 \end{bmatrix}$$

Now check if the eigenvalues of $A - LC$ are indeed $-3 \pm 4j$

$$A - LC = \begin{bmatrix} 0 & 1 \\ 0 & -2 \end{bmatrix} - \begin{bmatrix} 4 \\ 17 \end{bmatrix} \begin{bmatrix} 1 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 0 & -2 \end{bmatrix} - \begin{bmatrix} 4 & 0 \\ 17 & 0 \end{bmatrix} = \begin{bmatrix} -4 & 1 \\ -17 & -2 \end{bmatrix}$$

eigenvalues of $\begin{bmatrix} -4 & 1 \\ -17 & -2 \end{bmatrix}$ are indeed $-3 \pm 4j$. done!