

6.2) Compute A^t and e^{At} for the following matrices

$$A_1 = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}, \quad A_2 = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 1 \end{bmatrix}, \quad A_3 = \begin{bmatrix} 2 & 0 & 0 & 0 \\ 2 & 2 & 0 & 0 \\ 0 & 0 & 3 & 3 \\ 0 & 0 & 0 & 3 \end{bmatrix}$$

$$A) \quad A_1 A_1 = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 1+1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$A^2 = \begin{bmatrix} 1 & 2 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 1+2 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$A^t = \begin{bmatrix} 1 & t & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} \lambda^t & t\lambda^{t-1} & 0 \\ 0 & \lambda^t & 0 \\ 0 & 0 & \lambda^t \end{bmatrix} \quad \checkmark$$

$$e^{At} = \mathcal{L}^{-1}((sI - A)^{-1})$$

$$\Rightarrow \begin{bmatrix} s-1 & -1 & 0 \\ 0 & s-1 & 0 \\ 0 & 0 & s-1 \end{bmatrix} \quad \text{cofactor} \begin{bmatrix} (s-1)^2 & 0 & 0 \\ (s-1) & (s-1)^2 & 0 \\ 0 & 0 & (s-1)^2 \end{bmatrix} \quad \text{adjugate} = \begin{bmatrix} (s-1)^2 & (s-1) & 0 \\ 0 & (s-1)^2 & 0 \\ 0 & 0 & (s-1)^2 \end{bmatrix}$$

$$\det(sI - A) = -(s-1)^3$$

$$(sI - A)^{-1} = \begin{bmatrix} \frac{1}{s-1} & \frac{1}{s-1} & 0 \\ 0 & \frac{1}{s-1} & 0 \\ 0 & 0 & \frac{1}{s-1} \end{bmatrix} \quad \mathcal{L}^{-1}(sI - A)^{-1} = e^t \begin{bmatrix} 1 & t & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} U(t) \quad \checkmark$$

$$A_2) \quad A = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 1 \end{bmatrix} \quad A^2 = \begin{bmatrix} 1 & 1 & 1+1 \\ 0 & 0 & 1 \\ 0 & 0 & 1 \end{bmatrix} \quad \text{thm } A^t = \begin{bmatrix} 1 & 1 & t+1 \\ 0 & 0 & 1 \\ 0 & 0 & 1 \end{bmatrix} \quad \checkmark$$

$$e^{At} = \mathcal{L}^{-1}((sI - A)^{-1})$$

$$\Rightarrow \begin{bmatrix} s-1 & -1 & 0 \\ 0 & s & -1 \\ 0 & 0 & s-1 \end{bmatrix} \quad \text{cofactor} \begin{bmatrix} s(s-1) & 0 & 0 \\ s-1 & (s-1)^2 & 0 \\ 1 & s-1 & s(s-1) \end{bmatrix} \quad \text{adjugate} = \begin{bmatrix} s(s-1) & s-1 & 1 \\ 0 & (s-1)^2 & s-1 \\ 0 & 0 & s(s-1) \end{bmatrix}$$

$$\det(sI - A) = s(s-1)^2$$

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

$$\frac{1}{s(s-1)^2} = \frac{1}{s} + \frac{1}{s-1} + \frac{1}{(s-1)^2}$$

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

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

$$A_3 = \begin{bmatrix} 2 & 0 & 0 & 0 \\ 2 & 2 & 0 & 0 \\ 0 & 0 & 3 & 3 \\ 0 & 0 & 0 & 3 \end{bmatrix} = \begin{bmatrix} B_1 & 0 \\ 0 & B_2 \end{bmatrix} \text{ where } B_1 = \begin{bmatrix} 2 & 0 \\ 2 & 2 \end{bmatrix} \text{ \& } B_2 = \begin{bmatrix} 3 & 3 \\ 0 & 3 \end{bmatrix}$$

$$A_3^k = \begin{bmatrix} B_1^k & 0 \\ 0 & B_2^k \end{bmatrix} \text{ \& } e^{A_3 t} = \begin{bmatrix} e^{B_1 t} & 0 \\ 0 & e^{B_2 t} \end{bmatrix}$$

$$B_1^2 = \begin{bmatrix} 2 & 0 \\ 2 & 2 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 2 & 2 \end{bmatrix} = \begin{bmatrix} 2^2 & 0 \\ 2 \cdot 2 & 2^2 \end{bmatrix} \rightarrow \begin{bmatrix} \lambda^k & 0 \\ k \lambda^{k-1} & \lambda^k \end{bmatrix} = \begin{bmatrix} 2^k & 0 \\ k 2^{k-1} & 2^k \end{bmatrix}$$

$$B_2^k = \begin{bmatrix} \lambda^k & k \lambda^{k-1} \\ 0 & \lambda^k \end{bmatrix} \text{ From first part } = \begin{bmatrix} 3^k & k 3^{k-1} \\ 0 & 3^k \end{bmatrix}$$

$$A^k = \begin{bmatrix} 2^k & 0 & 0 & 0 \\ k 2^{k-1} & 2^k & 0 & 0 \\ 0 & 0 & 3^k & k 3^{k-1} \\ 0 & 0 & 0 & 3^k \end{bmatrix} \quad \checkmark$$

$$e^{A_3 t} = \begin{bmatrix} e^{B_1 t} & 0 \\ 0 & e^{B_2 t} \end{bmatrix}$$

$$B_1 = \begin{bmatrix} 2 & 0 \\ 2 & 2 \end{bmatrix} \quad \mathcal{L}^{-1}(sI - B_1) = \mathcal{L}^{-1} \begin{bmatrix} s-2 & 0 \\ 2 & s-2 \end{bmatrix} \frac{1}{(s-2)^2} = \begin{bmatrix} e^{2t} & 0 \\ 2te^{2t} & e^{2t} \end{bmatrix}$$

$$B_2 = \begin{bmatrix} 3 & 3 \\ 0 & 3 \end{bmatrix} \quad \mathcal{L}^{-1}(sI - B_2) = \begin{bmatrix} e^{3t} & 3e^{3t} \\ 0 & e^{3t} \end{bmatrix}$$

$$e^{A_3 t} = \begin{bmatrix} e^{2t} & 0 & 0 & 0 \\ 2te^{2t} & e^{2t} & 0 & 0 \\ 0 & 0 & e^{3t} & 3te^{3t} \\ 0 & 0 & 0 & e^{3t} \end{bmatrix}$$

7.2 2nd Edition)

Consider the matrix $A = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 2 & 0 \\ 0 & 1 & 2 \end{bmatrix}$

a) compute the characteristic & minimum polynomial of A .

$$\det(\lambda I - A) = \begin{vmatrix} \lambda-1 & -1 & 0 \\ 0 & \lambda-2 & 0 \\ 0 & 0 & \lambda-2 \end{vmatrix} = (\lambda-1)(\lambda-2)^2 \Rightarrow (\lambda-1)(\lambda^2-4\lambda+4) \\ \Rightarrow \lambda^3-4\lambda^2+4\lambda-\lambda^2+4\lambda-4 = \lambda^3-5\lambda^2+8\lambda-4 \quad \checkmark$$

$$\lambda^3 - 5\lambda^2 + 8\lambda - 4 = 0 \quad \checkmark$$

b) Is this matrix diagonalizable? If so diagonalize it, otherwise compute its Jordan normal form.

$$\lambda_1 (\lambda I - A) = \begin{bmatrix} 0 & -1 & 0 \\ 0 & -1 & 0 \\ 0 & -1 & -1 \end{bmatrix} \begin{matrix} x_2=0 \\ x_2=0 \\ -x_2-x_3=0 \end{matrix} \quad \begin{matrix} v_1 \\ 1 \\ 0 \\ 0 \end{matrix}$$

$$\lambda_2 (A - \lambda I) = \begin{bmatrix} -1 & 1 & 0 \\ 0 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \begin{matrix} -x_1+x_2=0 \\ x_3=\text{anything} \\ x_2=0 \end{matrix} \quad \begin{matrix} v_2 \\ 0 \\ 0 \\ 1 \end{matrix}$$

$$(A - \lambda I)v_3 = v_2 \quad \begin{bmatrix} -1 & 1 & 0 \\ 0 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \quad \begin{matrix} -x_1+x_2=0 & x_1=-1 \\ x_3=\text{anything} \\ x_2=1 \end{matrix} \quad v_3 = \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}$$

$$V = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \quad J = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 1 \\ 0 & 0 & 2 \end{bmatrix} \quad \checkmark$$

c) Compute e^{At}

$$e^{At} = V e^{Jt} V^{-1} = V \begin{bmatrix} e^t & 0 & 0 \\ 0 & e^{2t} & te^{2t} \\ 0 & 0 & e^{2t} \end{bmatrix} V^{-1}$$

8.2) For a given matrix A , construct vectors for which (8.2) holds for each of the three norms $\|\cdot\|_1$, $\|\cdot\|_2$, and $\|\cdot\|_\infty$.

(8.2) For every matrix $A \in \mathbb{R}^{m \times n}$ there is a vector $x^* \in \mathbb{R}^n$ for which $\|A\|_p = \frac{\|Ax^*\|_p}{\|x^*\|_p}$

$\|A\|_2 = \sigma_{\max}[A]$ where $\sigma_{\max}[A]$ denotes the largest singular value of A

let v_i correspond to the left eigen of σ_{\max}

$$\|A\|_2 = \|Ax^*\|_2 / \|x^*\|_2 = \|U \in v_i\|_2 / \|v_i\|_2 = \|U \in e_i\|_2 = \|U e_i\|_2 = \|U e_i\|_2 \sigma_{\max} = \|\sigma_{\max}\|$$

$$\|A\|_1 = \max_{1 \leq j \leq n} \sum_{i=1}^m |a_{ij}|$$

let x^* be $e_i \in \mathbb{R}^n$ where i indicates the non-zero element = to one

$$\|A\|_1 = \|Ax^*\|_1 / \|x^*\|_1 = \|A e_i\|_1 / \|e_i\|_1 = \max_{1 \leq j \leq n} \sum_{i=1}^m |a_{ij}|$$

$$\|A\|_\infty = \max_{1 \leq i \leq m} \sum_{j=1}^n |a_{ij}| \quad \text{or max row sum}$$

let x^* be $v \in \mathbb{R}^n$ where v is filled elements of the same value. whose sign is the same as the elements in the max abs row.

$$\|A\|_\infty = \|Ax^*\|_\infty / \|x^*\|_\infty = \|A v\|_\infty / \|v\|_\infty$$

$$= \left\| \begin{bmatrix} \alpha a_{11} + \alpha a_{12} + \dots + \alpha a_{1n} \\ \alpha a_{21} + \alpha a_{22} + \dots + \alpha a_{2n} \\ \alpha a_{m1} + \alpha a_{m2} + \dots + \alpha a_{mn} \end{bmatrix} \right\|_\infty / |\alpha| = \max_{1 \leq i \leq m} \sum_{j=1}^n |a_{ij}|$$

8.4) Consider a linear system with a state-transition matrix $\Phi(t, \tau)$ for which

$$\Phi(t, 0) = \begin{bmatrix} e^t \cos 2t & -e^t \sin 2t \\ -e^t \sin 2t & e^t \cos 2t \end{bmatrix}$$

See Matlab

c) compute the eigenvalues of $A(t)$

$$\det(A - sI) = \det \begin{bmatrix} 3\cos(\theta)/2 - 1/2 - s & 2 - 3\sin(\theta)/2 \\ -3\sin(\theta)/2 - 2 & -1/2 - 3\cos(\theta)/2 - s \end{bmatrix} \quad \text{where } \theta = 4t - 4t_0$$

$$= (3\cos(\theta)/2 - 1/2 - s)(-1/2 - 3\cos(\theta)/2 - s) - (2 - 3\sin(\theta)/2)(-3\sin(\theta)/2 - 2)$$

$$= -9\cos^2\theta/4 - 3\cos\theta/4 - 5\cos\theta/2 + 1/4 + 3\cos\theta/2 + 1/2s + 1/2s + 5\cos\theta/2 + s^2$$

$$+ 3\cos\theta + 4 - 9\sin^2\theta/4 - 3\sin(\theta)$$

$$= -9/4 (\cos^2\theta + \sin^2\theta) + 1/4 + s + s^2 + 4 - 3s_0 + 3c_0$$

$$= s^2 + s + 2 - 3s_0 + 3c_0$$

$$\lambda = \frac{-1 \pm \sqrt{1 - 4(2 - 3s_0 + 3c_0)^2}}{2} = \frac{-1}{2} \pm \frac{\sqrt{1 - 4(4 - 6s_0 + 6c_0 - 6s_0^2 + 9c_0^2 - 9s_0c_0 + 6c_0 - 9s_0^2)}}{2}$$

$$\lambda = \frac{-1}{2} \pm \frac{\sqrt{1 - 4(13 - 12s_0 + 12c_0 - 18c_0s_0)}}{2}$$

using Matlab for the res

$$\lambda = -\frac{1}{2} - \frac{\sqrt{7}}{2}j \quad \text{and} \quad -\frac{1}{2} + \frac{\sqrt{7}}{2}j$$

d) classify this system in terms of Lyapunov stability

Looking at the state transition matrix it looks unstable.

3.5) verify that $(e^{At})' = e^{A't}$

$$e^{At} = \sum_{k=0}^{\infty} \frac{t^k}{k!} A^k \Rightarrow (e^{At})' = \left(\sum_{k=0}^{\infty} \frac{t^k}{k!} A^k \right)' = \sum_{k=0}^{\infty} \frac{t^k}{k!} (A^k)' = \sum_{k=0}^{\infty} \frac{t^k}{k!} (A')^k = e^{A't}$$

See Matlab

B.6) Consider the continuous-time LTI system

$$\dot{x} = Ax \quad x \in \mathbb{R}^n$$

and suppose that there exists a positive constant α and positive-definite matrices $P, Q \in \mathbb{R}^n$ for the Lyapunov equation

$$A^T P + P A + 2\alpha P = -Q$$

Show that all eigenvalues of A have real parts less than $-\alpha$.

For clarification Let $B = A + \alpha I$ then

$$B^T P_b + P_b B = -Q_b \Rightarrow (A^T + \alpha I) P_b + P_b (A + \alpha I) = -Q_b \Rightarrow A^T P_b + P_b A + 2\alpha P_b = -Q_b$$

Thus according to Theorem 8.2 all of the eigen values of B have strictly negative real parts. $\text{eig}(B) < 0$

$B = A + \alpha I = V^{-1} \Lambda V + \alpha I$ where $V \neq \Lambda$ are the eigen vectors and eigen values of A , then

$$B = V^{-1} \Lambda V + \alpha V V^{-1} V = V^{-1} (\Lambda + \alpha I) V \quad \text{where } \Lambda + \alpha I \text{ are the eigen values of } B \text{ thus } \Lambda + \alpha I < 0 \text{ and } \Lambda < -\alpha I$$

Thus all the eigen values of A have real parts less than $-\alpha$

8.7) Investigate whether or not the solutions to the following nonlinear systems converge to the given equilibrium point when they start close enough to it

a) The state space system

$$\begin{aligned} \dot{x}_1 &= -x_1 + x_1(x_1^2 + x_2^2) = f_1(x_1, x_2) \\ \dot{x}_2 &= -x_2 + x_2(x_1^2 + x_2^2) = f_2(x_1, x_2) \end{aligned}$$

with equilibrium points $x_1 = x_2 = 0$

$$\begin{aligned} \frac{\partial f_1}{\partial x_1} \bigg|_{e_0} &= -1 + x_1^2 + x_2^2 \bigg|_{e_0} = -1 \\ \frac{\partial f_1}{\partial x_2} \bigg|_{e_0} &= 0 + 2x_1 x_2 \bigg|_{e_0} = 0 \\ \frac{\partial f_2}{\partial x_1} \bigg|_{e_0} &= 0 + 2x_1 x_2 \bigg|_{e_0} = 0 \\ \frac{\partial f_2}{\partial x_2} \bigg|_{e_0} &= -1 + x_1^2 + x_2^2 \bigg|_{e_0} = -1 \end{aligned}$$

$$\dot{x} = \underbrace{\begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix}}_A x \quad \text{thus} \quad x(t) = e^A x(0) \quad \text{the eigen values of } A \text{ are negative thus the system will converge to the equilibrium point when near it.}$$

b) The second order system

$$\ddot{w} + g(w) \dot{w} + w = 0 \quad \Rightarrow \quad \ddot{w} = -g(w) \dot{w} - w = f(w, \dot{w})$$

With the equilibrium point $w = \dot{w} = 0$. Determine for which values of $g(0)$ we can guarantee convergence to the origin based on the local linearization

$$\frac{df}{dw}|_{q_0} = -1$$

$$\frac{df}{d\dot{w}}|_{q_0} = -g(w)|_{q_0} = -g(0)$$

$$\dot{\delta w} = -g(0) \delta \dot{w} - \delta w \quad \text{where } \delta = x(t) - x_{eq}$$

$$\text{Let } r(w, \dot{w}) = f(w, \dot{w}) - (f(w_{eq}, \dot{w}_{eq}) - g(0) \delta \dot{w} - \delta w) = O(\|\delta w\|^2, \|\delta \dot{w}\|^2)$$

basically r is the error in the Taylor Series approximation.

which means that there is a constant C and a ball \bar{B} around equilibrium for which

$$\|r\| \leq C[\|\delta w\|^2 + \|\delta \dot{w}\|^2] \quad \forall w, \dot{w} \in \bar{B}$$

Create the Lyapunov function

$$\begin{aligned} V(t) &= \frac{1}{2} \dot{w}(t)^2 + \frac{1}{2} w(t)^2 \\ \dot{V}(t) &= \delta \dot{w}(t) \delta \ddot{w}(t) + \dot{w}(t) f(x) \\ &= \delta \dot{w}(t) \delta \ddot{w}(t) + \delta \dot{w}(t) [-g(0) \delta \dot{w} - \delta w + r] \\ &= \delta \dot{w}(t) \delta \ddot{w}(t) - \delta \dot{w}(t) \delta w - \delta \dot{w}(t)^2 g(0) + \delta \dot{w}(t) r \\ &= -\delta \dot{w}^2(t) g(0) + \delta \dot{w}(t) r \\ &\leq -\delta \dot{w}^2(t) g(0) + C \delta \dot{w}(t) [\|\delta w\|^2 + \|\delta \dot{w}\|^2] \\ &\leq -\|\delta \dot{w}(t)\|^2 g(0) + C \|\delta \dot{w}(t)\| [\|\delta w\|^2 + \|\delta \dot{w}\|^2] \\ &\leq -\|\delta \dot{w}\|^2 [g(0) - C[\|\delta w\|^2 + \|\delta \dot{w}\|^2] / \|\delta \dot{w}\|] \end{aligned}$$

$$\text{we need } g(0) > C[\|\delta w\|^2 + \|\delta \dot{w}\|^2] / \|\delta \dot{w}\|$$

or in state space form

$$\begin{aligned} \dot{x} &= \begin{bmatrix} 0 & 1 \\ -1 & -g(0) \end{bmatrix} x \quad \text{eigen values are } \det(A - \lambda I) = \det \begin{bmatrix} -\lambda & 1 \\ -1 & -g(0) - \lambda \end{bmatrix} = -\lambda(-g(0) - \lambda) + 1 \\ &= \lambda^2 + g(0)\lambda + 1 \\ &= \frac{-g(0) \pm \sqrt{g(0)^2 - 4}}{2} \end{aligned}$$

$$g(0) \leq -g(0) \pm \sqrt{g(0)^2 - 4} < 0$$

$g(0) > 0$ the $\text{Re}\{\lambda_i\}$ are strictly negative ✓

$g(0) < 0$ the $\text{Re}\{\lambda_i\}$ are strictly positive ✓

$g(0) = 0$ the $\text{Re}\{\lambda_i\}$ are 0 and eigen values are complex ✓