

### Solution to Final Examination

1. (a) False. For example,

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

is an orthogonal matrix since  $\mathbf{Q}\mathbf{Q}^T = \mathbf{I}$ , but we have  $\det(\mathbf{Q}) = -1$ .

- (b) False. Let

$$\mathbf{A} = \frac{1}{6} \begin{bmatrix} 5 & 2 & -1 \\ 2 & 2 & 2 \\ -1 & 2 & -5 \end{bmatrix}.$$

We have

$$\mathbf{A}^2 = \frac{1}{36} \begin{bmatrix} 30 & 12 & 4 \\ 12 & 12 & -8 \\ 4 & -8 & 30 \end{bmatrix} \neq \mathbf{A}.$$

Therefore,  $\mathbf{A}$  is not a projection matrix.

- (c) True. Let

$$\mathbf{A} = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$

and

$$\mathbf{B} = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 3 \end{bmatrix}.$$

We can find the eigenvalues of  $\mathbf{A}$  with

$$\begin{aligned} \det(\mathbf{A} - \lambda\mathbf{I}) &= \begin{vmatrix} 1 - \lambda & 1 & 1 \\ 1 & 1 - \lambda & 1 \\ 1 & 1 & 1 - \lambda \end{vmatrix} \\ &= -\lambda^2(\lambda - 3) = 0. \end{aligned}$$

Therefore, the eigenvalues of  $\mathbf{A}$  are 0, 0, 3. Since  $\mathbf{A}$  is symmetric, it is diagonalizable and is similar to

$$\mathbf{B} = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 3 \end{bmatrix}.$$

- (d) True. Let  $r$  be the rank of  $\mathbf{A}$ , and then  $r \leq n < m$ . Since  $\mathbf{A}\mathbf{A}^T$  is  $m$  by  $m$  and  $\text{rank}(\mathbf{A}\mathbf{A}^T) = \text{rank}(\mathbf{A}) = r < m$ ,  $\mathbf{A}\mathbf{A}^T$  is singular. Therefore,  $\det(\mathbf{A}\mathbf{A}^T) = 0$ , which yields that  $\mathbf{A}\mathbf{A}^T$  cannot be positive definite.

2. (a) By Gaussian elimination, we can obtain

$$\begin{bmatrix} 1 & 0 \\ -1/2 & 1 \end{bmatrix} \mathbf{A} = \begin{bmatrix} 2 & 1 \\ 0 & 3/2 \end{bmatrix}.$$

Therefore,

$$\mathbf{A} = \begin{bmatrix} 1 & 0 \\ 1/2 & 1 \end{bmatrix} \begin{bmatrix} 2 & 1 \\ 0 & 3/2 \end{bmatrix} = \mathbf{LU}.$$

(b) Since the columns of  $\mathbf{A}$  are independent, let  $\mathbf{a}_1 = (2, 1)^T$  and  $\mathbf{a}_2 = (1, 2)^T$ . By the Gram-Schmidt process, we can have

$$\begin{aligned} \mathbf{A}_1 = \mathbf{a}_1 = \begin{bmatrix} 2 \\ 1 \end{bmatrix} &\implies \mathbf{q}_1 = \frac{\mathbf{A}_1}{\|\mathbf{A}_1\|} = \frac{1}{\sqrt{5}} \begin{bmatrix} 2 \\ 1 \end{bmatrix} \\ \mathbf{A}_2 = \mathbf{a}_2 - (\mathbf{a}_2^T \mathbf{q}_1) \mathbf{q}_1 = \begin{bmatrix} -3/5 \\ 6/5 \end{bmatrix} &\implies \mathbf{q}_2 = \frac{\mathbf{A}_2}{\|\mathbf{A}_2\|} = \frac{1}{\sqrt{5}} \begin{bmatrix} -1 \\ 2 \end{bmatrix}. \end{aligned}$$

Therefore,

$$\mathbf{A} = [\mathbf{a}_1 \quad \mathbf{a}_2] = [\mathbf{q}_1 \quad \mathbf{q}_2] \begin{bmatrix} \mathbf{q}_1^T \mathbf{a}_1 & \mathbf{q}_1^T \mathbf{a}_2 \\ 0 & \mathbf{q}_2^T \mathbf{a}_2 \end{bmatrix} = \mathbf{QR}$$

which gives

$$\begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} = \begin{bmatrix} 2/\sqrt{5} & -1/\sqrt{5} \\ 1/\sqrt{5} & 2/\sqrt{5} \end{bmatrix} \begin{bmatrix} 5/\sqrt{5} & 4/\sqrt{5} \\ 0 & 3/\sqrt{5} \end{bmatrix}.$$

(c) We can have

$$\begin{aligned} \det(\mathbf{A} - \lambda \mathbf{I}) &= \begin{vmatrix} 2 - \lambda & 1 \\ 1 & 2 - \lambda \end{vmatrix} = (3 - \lambda)(1 - \lambda) \\ \implies \begin{cases} \lambda_1 = 3 & \iff \mathbf{v}_1 = \frac{1}{\sqrt{2}}(1, 1)^T \\ \lambda_2 = 1 & \iff \mathbf{v}_2 = \frac{1}{\sqrt{2}}(1, -1)^T. \end{cases} \end{aligned}$$

Therefore,

$$\mathbf{A} = \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{bmatrix} \begin{bmatrix} 3 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{bmatrix}^T = \mathbf{Q}\mathbf{\Lambda}\mathbf{Q}^T.$$

(d) From (a), we can obtain

$$\begin{aligned} \mathbf{A} &= \begin{bmatrix} 1 & 0 \\ 1/2 & 1 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & 3/2 \end{bmatrix} \begin{bmatrix} 1 & 1/2 \\ 0 & 1 \end{bmatrix} \\ &= \begin{bmatrix} 1 & 0 \\ 1/2 & 1 \end{bmatrix} \begin{bmatrix} \sqrt{2} & 0 \\ 0 & \sqrt{6}/2 \end{bmatrix} \begin{bmatrix} \sqrt{2} & 0 \\ 0 & \sqrt{6}/2 \end{bmatrix}^T \begin{bmatrix} 1 & 0 \\ 1/2 & 1 \end{bmatrix}^T \\ &= \begin{bmatrix} \sqrt{2} & 0 \\ \sqrt{2}/2 & \sqrt{6}/2 \end{bmatrix} \begin{bmatrix} \sqrt{2} & 0 \\ \sqrt{2}/2 & \sqrt{6}/2 \end{bmatrix}^T = \mathbf{C}\mathbf{C}^T. \end{aligned}$$

3. (a) Let  $\mathbf{u}_k = \begin{bmatrix} x_k \\ y_k \end{bmatrix}$ . The relation between  $\mathbf{u}_{k+1} = \begin{bmatrix} x_{k+1} \\ y_{k+1} \end{bmatrix}$  and  $\mathbf{u}_k = \begin{bmatrix} x_k \\ y_k \end{bmatrix}$  is given by

$$\mathbf{u}_{k+1} = \begin{bmatrix} x_{k+1} \\ y_{k+1} \end{bmatrix} = \begin{bmatrix} 4/5 & 1/10 \\ 1/5 & 9/10 \end{bmatrix} \begin{bmatrix} x_k \\ y_k \end{bmatrix} = \mathbf{A}\mathbf{u}_k.$$

To find  $\mathbf{A}^k$ , we first find the eigenvalues and corresponding eigenvectors of  $\mathbf{A}$ .

$$\begin{aligned} \det(\mathbf{A} - \lambda\mathbf{I}_2) &= \begin{vmatrix} (4/5) - \lambda & 1/10 \\ 1/5 & (9/10) - \lambda \end{vmatrix} \\ &= \lambda^2 - \frac{17}{10}\lambda + \frac{7}{10} \\ &= (\lambda - 1)(\lambda - (7/10)) = 0 \\ \implies \begin{cases} \lambda_1 = 1 & \iff \mathbf{v}_1 = (1, 2)^T \\ \lambda_2 = 7/10 & \iff \mathbf{v}_2 = (1, -1)^T. \end{cases} \end{aligned}$$

Therefore, we have

$$\mathbf{A} = \mathbf{S}\mathbf{\Lambda}\mathbf{S}^{-1} = \begin{bmatrix} 1 & 1 \\ 2 & -1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 7/10 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 2 & -1 \end{bmatrix}^{-1}.$$

We can write  $\mathbf{u}_0$  as a linear combination of  $\mathbf{v}_1$  and  $\mathbf{v}_2$  as follows:

$$\begin{aligned} \mathbf{u}_0 = \begin{bmatrix} x_0 \\ y_0 \end{bmatrix} &= \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 2 & -1 \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \end{bmatrix} \\ \implies \begin{bmatrix} c_1 \\ c_2 \end{bmatrix} &= \begin{bmatrix} 1/3 \\ 2/3 \end{bmatrix} \\ \implies \mathbf{u}_0 &= \frac{1}{3}\mathbf{v}_1 + \frac{2}{3}\mathbf{v}_2. \end{aligned}$$

Then we can obtain

$$\begin{aligned} \mathbf{u}_k &= \mathbf{A}^k\mathbf{u}_0 \\ &= \frac{1}{3}\mathbf{A}^k\mathbf{v}_1 + \frac{2}{3}\mathbf{A}^k\mathbf{v}_2 \\ &= \frac{1}{3}\mathbf{v}_1 + \frac{2}{3}\left(\frac{7}{10}\right)^k\mathbf{v}_2 \\ &= \begin{bmatrix} (1/3) + (2/3)(7/10)^k \\ (2/3) - (2/3)(7/10)^k \end{bmatrix}. \end{aligned}$$

Therefore,

$$\lim_{k \rightarrow \infty} y_k = \lim_{k \rightarrow \infty} \left( \frac{2}{3} - \frac{2}{3} \left( \frac{7}{10} \right)^k \right) = \frac{2}{3}$$

i.e., after a long time, 2/3 of the NTHUEE students prefer linear algebra to calculus.

(b) Let  $\mathbf{u} = \begin{bmatrix} x(t) \\ y(t) \end{bmatrix}$ . We then have

$$\frac{d\mathbf{u}}{dt} = \begin{bmatrix} 3 & -4 \\ 2 & -3 \end{bmatrix} \mathbf{u} = \mathbf{A}\mathbf{u}.$$

To find the eigenvalues and corresponding eigenvectors of  $\mathbf{A}$ , we calculate

$$\begin{aligned} \det(\mathbf{A} - \lambda\mathbf{I}_2) &= \begin{vmatrix} 3 - \lambda & -4 \\ 2 & -3 - \lambda \end{vmatrix} \\ &= \lambda^2 - 1 \\ &= (\lambda - 1)(\lambda + 1) = 0 \\ \implies \begin{cases} \lambda_1 = 1 & \iff \mathbf{v}_1 = (2, 1)^T \\ \lambda_2 = -1 & \iff \mathbf{v}_2 = (1, 1)^T. \end{cases} \end{aligned}$$

Hence, we can obtain

$$\mathbf{u} = c_1 e^t \begin{bmatrix} 2 \\ 1 \end{bmatrix} + c_2 e^{-t} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 2c_1 e^t + c_2 e^{-t} \\ c_1 e^t + c_2 e^{-t} \end{bmatrix}.$$

At  $t = 0$ , we have

$$\begin{cases} x(0) = 1 = 2c_1 + c_2 \\ y(0) = 0 = c_1 + c_2 \end{cases} \implies \begin{cases} c_1 = 1 \\ c_2 = -1. \end{cases}$$

Therefore,

$$x(t) = 2e^t - e^{-t}$$

and

$$y(t) = e^t - e^{-t}.$$

4. (a) A matrix  $\mathbf{A}$  is diagonalizable if and only if each of its eigenvalues has the same algebraic multiplicity (AM) and geometric multiplicity (GM). Hence we should find the eigenvalues and corresponding eigenvectors of  $\mathbf{A}$ . Since

$$\det(\mathbf{A} - \lambda\mathbf{I}) = \begin{vmatrix} -\lambda & 0 & 0 \\ 1 & -\lambda & 0 \\ 0 & 1 & -\lambda \end{vmatrix} = -\lambda^3 = 0$$

we have  $\lambda = 0, 0, 0$ . For  $\lambda = 0$ , the AM of  $\lambda$  is 3. Besides,

$$\mathbf{A} - \lambda\mathbf{I} = \begin{bmatrix} 0 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$

and the corresponding eigenvector is

$$\begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}.$$

This gives the GM of  $\lambda$  is 1, which is smaller than the AM of  $\lambda$ . As a result,  $\mathbf{A}$  is not diagonalizable.

- (b) Since the AM and GM of the eigenvalue 0 are 3 and 1, respectively, the Jordan form  $\mathbf{J}$  for  $\mathbf{A}$  is

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

- (c) To obtain the singular value decomposition of  $\mathbf{A}$ , we first find the eigenvalues and corresponding orthonormal eigenvectors of  $\mathbf{A}^T \mathbf{A}$ . After some calculations, we can obtain

$$\mathbf{A}^T \mathbf{A} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} \implies \begin{cases} \lambda_1 = 1, \lambda_2 = 1 & \longleftrightarrow \mathbf{v}_1 = (1, 0, 0)^T, \mathbf{v}_2 = (0, 1, 0)^T \\ \lambda_3 = 0 & \longleftrightarrow \mathbf{v}_3 = (0, 0, 1)^T. \end{cases}$$

For  $\lambda_1 = \lambda_2 = 1$ , we have the singular values  $\sigma_1 = \sqrt{\lambda_1} = 1$ ,  $\sigma_2 = \sqrt{\lambda_2} = 1$ , and

$$\mathbf{u}_1 = \frac{\mathbf{A}\mathbf{v}_1}{\sigma_1} = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \quad \mathbf{u}_2 = \frac{\mathbf{A}\mathbf{v}_2}{\sigma_2} = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}.$$

By the Gram-Schmidt process, we can obtain

$$\mathbf{u}_3 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}.$$

As a result, we have the singular value decomposition of  $\mathbf{A}$  as

$$\begin{aligned} \mathbf{A} &= \mathbf{U}\mathbf{\Sigma}\mathbf{V}^T \\ &= [\mathbf{u}_1 \quad \mathbf{u}_2 \quad \mathbf{u}_3] \begin{bmatrix} \sigma_1 & 0 & 0 \\ 0 & \sigma_2 & 0 \\ 0 & 0 & 0 \end{bmatrix} [\mathbf{v}_1 \quad \mathbf{v}_2 \quad \mathbf{v}_3]^T \\ &= \begin{bmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}. \end{aligned}$$

- (d) According to what was taught in class, we have  $\mathbf{v}_3 = (0, 0, 1)^T$  and  $\mathbf{u}_3 = (1, 0, 0)^T$  form an orthonormal basis for  $\mathcal{N}(\mathbf{A})$  and  $\mathcal{N}(\mathbf{A}^T)$ , respectively.

5. (a) Let  $\beta = \{1, x, x^2\}$ . Since

$$\begin{aligned} L(1) &= 0 &= 0 \cdot 1 + 0 \cdot x + 0 \cdot x^2 \\ L(x) &= x &= 0 \cdot 1 + 1 \cdot x + 0 \cdot x^2 \\ L(x^2) &= 2x^2 + 2 &= 2 \cdot 1 + 0 \cdot x + 2 \cdot x^2 \end{aligned}$$

we have

$$\mathbf{A} = [L]_{\beta} = \begin{bmatrix} 0 & 0 & 2 \\ 0 & 1 & 0 \\ 0 & 0 & 2 \end{bmatrix}.$$

(b) Let  $\gamma = \{1, x, 1 + x^2\}$ . Since

$$\begin{aligned} L(1) &= 0 &= 0 \cdot 1 + 0 \cdot x + 0 \cdot (1 + x^2) \\ L(x) &= x &= 0 \cdot 1 + 1 \cdot x + 0 \cdot (1 + x^2) \\ L(1 + x^2) &= 2x^2 + 2 &= 0 \cdot 1 + 0 \cdot x + 2 \cdot (1 + x^2) \end{aligned}$$

we have

$$\mathbf{B} = [L]_{\gamma} = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 2 \end{bmatrix}.$$

(c) Let  $I$  be the identity transformation, i.e.,  $I(p(x)) = p(x)$ . According to what was taught in class, we have  $\mathbf{M} = [I]_{\gamma}^{\beta}$ . Since

$$\begin{aligned} I(1) &= 1 &= 1 \cdot 1 + 0 \cdot x + 0 \cdot x^2 \\ I(x) &= x &= 0 \cdot 1 + 1 \cdot x + 0 \cdot x^2 \\ I(1 + x^2) &= 1 + x^2 &= 1 \cdot 1 + 0 \cdot x + 1 \cdot x^2 \end{aligned}$$

we can obtain

$$\mathbf{M} = [I]_{\gamma}^{\beta} = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}.$$

(d) Using  $\gamma$  as the basis for  $P_2$ , we can have

$$p(x) = b_0 \cdot 1 + b_1 \cdot x + b_2 \cdot (1 + x^2) \iff \begin{bmatrix} b_0 \\ b_1 \\ b_2 \end{bmatrix}$$

and

$$L(p(x)) \iff [L]_{\gamma} \begin{bmatrix} b_0 \\ b_1 \\ b_2 \end{bmatrix}.$$

To find  $L^n(p(x))$ , we should diagonalize  $[L]_{\gamma}$ . Since  $[L]_{\gamma}$  is already a diagonal matrix, we can obtain

$$([L]_{\gamma})^n = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 2 \end{bmatrix}^n = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 2^n \end{bmatrix}.$$

Therefore,

$$([L]_\gamma)^n \begin{bmatrix} b_0 \\ b_1 \\ b_2 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 2^n \end{bmatrix} \begin{bmatrix} b_0 \\ b_1 \\ b_2 \end{bmatrix} = \begin{bmatrix} 0 \\ b_1 \\ 2^n b_2 \end{bmatrix}$$

$$\begin{aligned} \implies L^n(p(x)) &= 0 \cdot 1 + b_1 \cdot x + 2^n b_2 \cdot (1 + x^2) \\ &= 2^n b_2 + b_1 x + 2^n b_2 x^2. \end{aligned}$$