
Question1 (5 points)

- (a) (1 point) How many $n \times n$ matrices are both diagonal and orthogonal?

Solution:

0M 2^n

$$\begin{bmatrix} \pm 1 & 0 & \cdots & 0 \\ 0 & \pm 1 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \pm 1 \end{bmatrix}$$

- (b) (1 point) How many $n \times n$ matrices are diagonal and unitary?

Solution:

0M There are infinitely many because each diagonal entry can be any point on the unit circle in the complex plane.

- (c) (1 point) Find a unitary matrix \mathbf{U} that diagonalizes

$$\mathbf{A} = \begin{bmatrix} 5 & 0 & 0 \\ 0 & -1 & -1+i \\ 0 & -1-i & 0 \end{bmatrix}$$

Solution:

0M Find the eigenvalues

$$\mathbf{D} = \begin{bmatrix} -2 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 5 \end{bmatrix}$$

0M Find and normalize the eigenvectors corresponding to those eigenvalues,

$$\mathbf{U} = \begin{bmatrix} 0 & 0 & 1 \\ \frac{1-i}{\sqrt{3}} & \frac{-1+i}{\sqrt{6}} & 0 \\ \frac{1}{\sqrt{3}} & \frac{2}{\sqrt{6}} & 0 \end{bmatrix}$$

- (d) (1 point) Prove that if \mathcal{V} is a complex inner product space, then

$$\langle \mathbf{u}, \mathbf{v} \rangle = \frac{\|\mathbf{u} + \mathbf{v}\|^2 - \|\mathbf{u} - \mathbf{v}\|^2 + \|\mathbf{u} + i\mathbf{v}\|^2 - \|\mathbf{u} - i\mathbf{v}\|^2}{4}i, \quad \text{for all } \mathbf{u}, \mathbf{v} \in \mathcal{V}.$$

Solution:

0M Expanding each of those top terms

$$\begin{aligned}\|\mathbf{u} + \mathbf{v}\|^2 &= \|\mathbf{u}\|^2 + \langle \mathbf{u}, \mathbf{v} \rangle + \overline{\langle \mathbf{u}, \mathbf{v} \rangle} + \|\mathbf{v}\|^2 \\ \|\mathbf{u} - \mathbf{v}\|^2 &= \|\mathbf{u}\|^2 - \langle \mathbf{u}, \mathbf{v} \rangle - \overline{\langle \mathbf{u}, \mathbf{v} \rangle} + \|\mathbf{v}\|^2 \\ \|\mathbf{u} + i\mathbf{v}\|^2 &= \|\mathbf{u}\|^2 - i\langle \mathbf{u}, \mathbf{v} \rangle + i\overline{\langle \mathbf{u}, \mathbf{v} \rangle} + \|\mathbf{v}\|^2 \\ \|\mathbf{u} - i\mathbf{v}\|^2 &= \|\mathbf{u}\|^2 + i\langle \mathbf{u}, \mathbf{v} \rangle - i\overline{\langle \mathbf{u}, \mathbf{v} \rangle} + \|\mathbf{v}\|^2\end{aligned}$$

0M Putting those terms together,

$$\begin{aligned}\|\mathbf{u} + \mathbf{v}\|^2 - \|\mathbf{u} - \mathbf{v}\|^2 &= 2 \left(\langle \mathbf{u}, \mathbf{v} \rangle + \overline{\langle \mathbf{u}, \mathbf{v} \rangle} \right), \\ i \left(\|\mathbf{u} + i\mathbf{v}\|^2 - \|\mathbf{u} - i\mathbf{v}\|^2 \right) &= 2 \left(\langle \mathbf{u}, \mathbf{v} \rangle - \overline{\langle \mathbf{u}, \mathbf{v} \rangle} \right)\end{aligned}$$

0M Thus together these show that

$$\langle \mathbf{u}, \mathbf{v} \rangle = \frac{\|\mathbf{u} + \mathbf{v}\|^2 - \|\mathbf{u} - \mathbf{v}\|^2 + \|\mathbf{u} + i\mathbf{v}\|^2 i - \|\mathbf{u} - i\mathbf{v}\|^2 i}{4}, \quad \text{for all } \mathbf{u}, \mathbf{v} \in \mathcal{V}.$$

(e) (1 point) Show the eigenvalues of a skew-Hermitian matrix are 0 or pure imaginary.

Solution:

0M Let \mathbf{A} be skew-Hermitian, and $\alpha = \mathbf{x}^H \mathbf{A} \mathbf{x}$, then

$$\bar{\alpha} = \alpha^H = (\mathbf{x}^H \mathbf{A} \mathbf{x})^H = \mathbf{x}^H \mathbf{A}^H \mathbf{x} = -\mathbf{x}^H \mathbf{A} \mathbf{x} = -\alpha$$

therefore $\alpha = \mathbf{x}^H \mathbf{A} \mathbf{x}$ is zero or pure imaginary.

0M And, since norm is always real,

$$\alpha = \mathbf{x}^H \mathbf{A} \mathbf{x} = \mathbf{x}^H \lambda \mathbf{x} = \lambda \|\mathbf{x}\|^2 \implies \lambda \text{ is zero or pure imaginary as well.}$$

Question2 (5 points)

(a) (1 point) Find a symmetric matrix \mathbf{B} such that

$$\mathbf{B}^2 = \begin{bmatrix} 17 & 16 & -16 \\ 16 & 41 & -32 \\ -16 & -32 & 41 \end{bmatrix}$$

Solution:

0M Note \mathbf{B}^2 is symmetric, so we can find the orthogonal diagonalization for \mathbf{B}^2 .

$$\mathbf{B}^2 = \mathbf{Q} \mathbf{D} \mathbf{Q}^T = \begin{bmatrix} \frac{2}{\sqrt{5}} & \frac{-2}{3\sqrt{5}} & \frac{-1}{3} \\ 0 & \frac{5}{3\sqrt{5}} & \frac{-2}{3} \\ \frac{1}{\sqrt{5}} & \frac{4}{3\sqrt{5}} & \frac{2}{3} \end{bmatrix} \begin{bmatrix} 9 & 0 & 0 \\ 0 & 9 & 0 \\ 0 & 0 & 81 \end{bmatrix} \begin{bmatrix} \frac{2}{\sqrt{5}} & 0 & \frac{1}{\sqrt{5}} \\ \frac{-2}{3\sqrt{5}} & \frac{5}{3\sqrt{5}} & \frac{4}{3\sqrt{5}} \\ \frac{-1}{3} & \frac{-2}{3} & \frac{2}{3} \end{bmatrix}$$

1M Therefore

$$\mathbf{B} = \begin{bmatrix} \frac{2}{\sqrt{5}} & \frac{-2}{3\sqrt{5}} & \frac{-1}{3} \\ 0 & \frac{5}{3\sqrt{5}} & \frac{-2}{3} \\ \frac{1}{\sqrt{5}} & \frac{4}{3\sqrt{5}} & \frac{2}{3} \end{bmatrix} \begin{bmatrix} 3 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 9 \end{bmatrix} \begin{bmatrix} \frac{2}{\sqrt{5}} & 0 & \frac{1}{\sqrt{5}} \\ \frac{-2}{3\sqrt{5}} & \frac{5}{3\sqrt{5}} & \frac{4}{3\sqrt{5}} \\ \frac{-1}{3} & \frac{-2}{3} & \frac{2}{3} \end{bmatrix}$$

- (b) (1 point) Use the singular value decomposition to find the least square solution of the system $\mathbf{Ax} = \mathbf{b}$ that has the smallest 2-norm, where

$$\mathbf{A} = \begin{bmatrix} 5 & 2 & 4 \\ 5 & 2 & 4 \\ 3 & 6 & 0 \\ 3 & 6 & 0 \end{bmatrix} \quad \text{and} \quad \mathbf{b} = \begin{bmatrix} 5 \\ 1 \\ -1 \\ 9 \end{bmatrix}$$

Solution:

0M Suppose \mathbf{A} has the following SVD,

$$\mathbf{A} = \mathbf{U}\mathbf{\Sigma}\mathbf{V}^T$$

then the first k columns of \mathbf{U} form an orthonormal basis for $\text{col}(\mathbf{A})$

0M Using the singular value decomposition, we know $\text{rank } \mathbf{A} = 2$

$$\mathbf{A} = \mathbf{U}\mathbf{\Sigma}\mathbf{V}^T$$

$$= \begin{bmatrix} -1/2 & -1/2 & 1/2 & 1/2 \\ -1/2 & -1/2 & -1/2 & 1/2 \\ -1/2 & 1/2 & 1/2 & -1/2 \\ -1/2 & 1/2 & 1/2 & 1/2 \end{bmatrix} \begin{bmatrix} 12 & 0 & 0 \\ 0 & 6 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} -2/3 & -1/3 & -2/3 \\ -2/3 & 2/3 & 1/3 \\ -1/3 & -2/3 & 2/3 \end{bmatrix}^T$$

0M It can be shown by the definition of SVD that

$$(\mathbf{A}^T \mathbf{A}) \hat{\mathbf{x}} = \mathbf{A}^T \mathbf{b}$$

$$\mathbf{\Sigma}^T \mathbf{\Sigma} \mathbf{V}^T \hat{\mathbf{x}} = \mathbf{\Sigma}^T \mathbf{U}^T \mathbf{b}$$

$$\begin{bmatrix} 144 & 0 & 0 \\ 0 & 36 & 0 \\ 0 & 0 & 0 \end{bmatrix} \mathbf{V}^T \hat{\mathbf{x}} = \begin{bmatrix} 12 & 0 & 0 & 0 \\ 0 & 6 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \mathbf{U}^T \mathbf{b}$$

$$\begin{bmatrix} 1/144 & 0 & 0 \\ 0 & 1/36 & 0 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 144 & 0 & 0 \\ 0 & 36 & 0 \\ 0 & 0 & 0 \end{bmatrix} \mathbf{V}^T \hat{\mathbf{x}} = \begin{bmatrix} 1/144 & 0 & 0 \\ 0 & 1/36 & 0 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 12 & 0 & 0 & 0 \\ 0 & 6 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \mathbf{U}^T \mathbf{b}$$

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} \mathbf{V}^T \hat{\mathbf{x}} = \begin{bmatrix} 1/12 & 0 & 0 & 0 \\ 0 & 1/6 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \mathbf{U}^T \mathbf{b}$$

0M Note the last row of the first matrix on both sides is redundant

$$\hat{\mathbf{x}} = \mathbf{V} \mathbf{\Sigma}^* \mathbf{U}^T \mathbf{b} = \begin{bmatrix} 1/3 \\ 1/2 \\ 1/12 \end{bmatrix}$$

where

$$\Sigma^* = \begin{bmatrix} 1/12 & 0 & 0 & 0 \\ 0 & 1/6 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

(c) (1 point) A **polar decomposition** of an $n \times n$ matrix \mathbf{A} is a factorization

$$\mathbf{A} = \mathbf{P}\mathbf{Q}$$

in which \mathbf{P} is a positive semidefinite $n \times n$ matrix with the same rank as \mathbf{A} , and \mathbf{Q} is an orthogonal $n \times n$ matrix. Show if

$$\mathbf{A} = \mathbf{U}\Sigma\mathbf{V}^T$$

is the SVD of \mathbf{A} , then

$$\mathbf{A} = (\mathbf{U}\Sigma\mathbf{U}^T)(\mathbf{U}\mathbf{V}^T)$$

is a polar decomposition of \mathbf{A} .

Solution:

0M Since $\mathbf{U}^T\mathbf{U}$ is the identity matrix \mathbf{I} ,

$$\mathbf{A} = \mathbf{U}\Sigma\mathbf{V}^T = \mathbf{U}\Sigma(\mathbf{U}^T\mathbf{U})\mathbf{V}^T = (\mathbf{U}\Sigma\mathbf{U}^T)(\mathbf{U}\mathbf{V}^T)$$

0M Since Σ can in general have zero diagonal elements, $\mathbf{P} = \mathbf{U}\Sigma\mathbf{U}^T$, which shares the same eigenvalues with Σ for they are similar, is positive semidefinite. And they share the same rank for the same reason.

0M It is trivial to verify that $\mathbf{U}\mathbf{V}^T$ is orthogonal

$$(\mathbf{U}\mathbf{V}^T)^T(\mathbf{U}\mathbf{V}^T) = \mathbf{V}\mathbf{U}^T\mathbf{U}\mathbf{V}^T = \mathbf{I}$$

(d) (1 point) Find a symmetric matrix \mathbf{A} so that

$$f(\mathbf{x}) = \mathbf{x}^T\mathbf{A}\mathbf{x}$$

where

$$f(\mathbf{x}) = \frac{1}{9}(-2x_1^2 + 7x_2^2 + 4x_3^2 + 4x_1x_2 + 16x_1x_3 + 20x_2x_3),$$

Is it a positive definite form?

Solution:

0M When $\mathbf{x}^T\mathbf{A}\mathbf{x}$ is expanded, the coefficient of x_ix_j is given by $(a_{ij} + a_{ji})/2$.

0m Therefore,

$$\mathbf{A} = \frac{1}{9} \begin{bmatrix} -2 & 2 & 8 \\ 2 & 7 & 10 \\ 8 & 10 & 4 \end{bmatrix}$$

0M No, the form is indefinite.

(e) (1 point) Suppose $\mathbf{A} \in \mathbb{C}^{n \times n}$, and

$$p(\lambda) = \det(\mathbf{A} - \lambda \mathbf{I})$$

is the characteristic polynomial of \mathbf{A} . Prove

$$p(\mathbf{A}) = 0$$

Solution:

0M If \mathbf{A} is diagonalizable, then we can find an invertible matrix \mathbf{P} such that

$$\mathbf{A} = \mathbf{P}\mathbf{D}\mathbf{P}^{-1} \quad \text{where } \mathbf{D} \text{ is diagonal.}$$

since the power of a diagonal matrix is

$$\mathbf{D}^k = \begin{bmatrix} \lambda_1^k & 0 & \cdots & 0 \\ 0 & \ddots & & \vdots \\ \vdots & & \ddots & 0 \\ 0 & \cdots & 0 & \lambda_n^k \end{bmatrix}$$

then a polynomial of diagonal matrix is

$$p(\mathbf{D}) = \begin{bmatrix} p(\lambda_1) & 0 & \cdots & 0 \\ 0 & \ddots & & \vdots \\ \vdots & & \ddots & 0 \\ 0 & \cdots & 0 & p(\lambda_n) \end{bmatrix}$$

0M Since $p(\lambda_i) = 0$ for all i , thus

$$p(\mathbf{D}) = \mathbf{0}$$

and $\mathbf{A}^k = \mathbf{P}\mathbf{D}^k\mathbf{P}^{-1}$ for all k . we have

$$p(\mathbf{A}) = \mathbf{P}p(\mathbf{D})\mathbf{P}^{-1} \implies p(\mathbf{A}) = \mathbf{0}$$

This completes the proof of the Cayley-Hamilton theorem in this special case.

0M To show the Cayley-Hamilton theorem in general, we use the fact that any matrix $\mathbf{A} \in \mathbb{C}^{n \times n}$ can be approximated by diagonalizable matrices. More precisely, given any matrix $\mathbf{A} \in \mathbb{C}^{n \times n}$, we can find a sequence of matrices $\{\mathbf{A}_k, k \in \mathbb{N}\}$ such that $\mathbf{A}_k \rightarrow \mathbf{A}$ as $k \rightarrow \infty$ and each matrix \mathbf{A}_k has n distinct eigenvalues.

0M This can be proved by the Schur's theorem,

$$\mathbf{A} = \mathbf{U}\mathbf{R}\mathbf{U}^H \implies \mathbf{A} \approx \mathbf{U}\mathbf{R}^*\mathbf{U}^H$$

by changing some of the diagonal entries of \mathbf{R} by less than ϵ so that all of those diagonal elements become distinct, thus diagonalizable.

0M Since $p(\mathbf{A})$ can be written as a polynomial of \mathbf{A} . With

$$\lim_{k \rightarrow \infty} \mathbf{A}_k = \mathbf{A}$$

we conclude that

$$\lim_{k \rightarrow \infty} p_k(\mathbf{A}_k) = p(\mathbf{A})$$

Since $p_k(\mathbf{A}_k) = \mathbf{0}$ for every $k \in \mathbb{N}$, we must have $p(\mathbf{A}) = \mathbf{0}$.