

3. (a) Show that the characteristic polynomial of the $n \times n$ matrix

$$\begin{bmatrix} 0 & 1 & 0 & \cdots & 0 \\ 0 & 0 & 1 & \cdots & 0 \\ \cdots & \cdots & \cdots & \cdots & \cdots \\ 0 & 0 & 0 & \cdots & 1 \\ -a_0 & -a_1 & -a_2 & \cdots & -a_{n-1} \end{bmatrix}$$

is $f(\lambda) := a_0 + a_1\lambda + \cdots + a_{n-1}\lambda^{n-1} + \lambda^n$. (The matrix is called the *companion matrix* of $f(\lambda)$.)

- (b) Given complex numbers $\lambda_1, \lambda_2, \dots, \lambda_n$, not necessarily distinct, show that there is a matrix with characteristic roots $\lambda_1, \lambda_2, \dots, \lambda_n$. Prove this in two ways: (i) directly and (ii) using (a).
4. Show that the characteristic roots of the $n \times n$ permutation matrix \mathbf{P} with $p_{i,i+1} = 1$ for $i = 1, \dots, n-1$ and $p_{n1} = 1$, are the n -th roots of unity.
7. Find a rank-factorization of the matrix

$$\mathbf{C} = \begin{bmatrix} 2 & 4 & 2 & 4 & 4 \\ 1 & 2 & 1 & 2 & 2 \\ 3 & 0 & 3 & 3 & 0 \\ 0 & -4 & 0 & -2 & -4 \\ 5 & 2 & 5 & 6 & 2 \end{bmatrix}$$

and hence the characteristic roots of \mathbf{C} .

8. Express the characteristic polynomial of $\alpha\mathbf{I} + \beta\mathbf{A}$ in terms of that of \mathbf{A} . Hence find the characteristic roots of $\alpha\mathbf{I} + \beta\mathbf{A}$. What are the characteristic roots of $-\mathbf{A}$?
-
9. Show that if β is a characteristic root of \mathbf{A} and \mathbf{A} is non-singular, $1/\beta$ is a characteristic root of \mathbf{A}^{-1} .
10. Show that the characteristic roots of a matrix do not determine rank (except when zero occurs as a characteristic root at most once;
12. Let \mathbf{A} be a 2×2 matrix. Then show that $|\mathbf{I} + \mathbf{A}| = 1 + |\mathbf{A}|$ iff $\text{tr}(\mathbf{A}) = 0$.
3. Let α be an eigenvalue of \mathbf{A} . Then show that $\text{ES}(\mathbf{A}^k, \alpha^k) \supseteq \text{ES}(\mathbf{A}, \alpha)$ if $k \geq 1$. Extend the result to $k = -1$ if \mathbf{A} is non-singular. Show also that proper inclusion is possible.
4. If k, ℓ and n are integers such that $1 \leq k \leq \ell \leq n$, show that there exists an $n \times n$ matrix \mathbf{A} and an eigenvalue α of \mathbf{A} such that k and ℓ are the geometric and algebraic multiplicities of α with respect to \mathbf{A} .
5. If $\alpha_1, \dots, \alpha_k$ are the distinct eigenvalues of an $n \times n$ matrix \mathbf{A} with geometric multiplicities n_1, \dots, n_k respectively, then $n_1 + \cdots + n_k \leq n$.

6. (a) Let δ be an eigenvalue of \mathbf{A} with algebraic multiplicity a and let $\beta \neq 0$. Then show that $\alpha + \beta\delta$ is an eigenvalue of $\alpha\mathbf{I} + \beta\mathbf{A}$ with algebraic multiplicity a and $\text{ES}(\alpha\mathbf{I} + \beta\mathbf{A}, \alpha + \beta\delta) = \text{ES}(\mathbf{A}, \delta)$.
- (b) Prove or disprove: if δ is an eigenvalue of \mathbf{A} , the algebraic and geometric multiplicities of $f(\delta)$ with respect to $f(\mathbf{A})$ are the same as those of δ with respect to \mathbf{A} for any polynomial f .
8. If \mathbf{A} is an $n \times n$ singular matrix with k distinct eigenvalues, show that $k - 1 \leq \rho(\mathbf{A}) \leq n - 1$. Also show by construction that $\rho(\mathbf{A})$ can take any value ℓ between $k - 1$ and $n - 1$.
10. Let $\mathbf{A} = \mathbf{u}\mathbf{u}^*$ where \mathbf{u} is a non-null vector.
- (a) Show that the eigenvalues of \mathbf{A} are 0 and $\mathbf{u}^*\mathbf{u}$.
- (b) Show that $\mathbf{u}^*\mathbf{u}$ is a simple eigenvalue of \mathbf{A} .
- (c) Identify $\text{ES}(\mathbf{A}, 0)$ and $\text{ES}(\mathbf{A}, \mathbf{u}^*\mathbf{u})$ and deduce the result in (b).
- (d) Show that \mathbf{A} is similar to a diagonal matrix.
11. Find the eigenvalues and their algebraic and geometric multiplicities for each of the real $n \times n$ matrices (i) $(\alpha - \beta)\mathbf{I} + \beta\mathbf{1}\mathbf{1}^T$ and (ii) $\alpha\mathbf{I} + \mathbf{u}\mathbf{1}^T + \mathbf{1}\mathbf{u}^T$. Here $\mathbf{1}$ denotes a vector with all entries 1 and \mathbf{u} is an arbitrary vector.
12. If α is an eigenvalue of \mathbf{A} , then it is an eigenvalue of \mathbf{A}^T also. An eigenvector of \mathbf{A}^T corresponding to α , i.e., a vector $\mathbf{x} \neq \mathbf{0}$ such that $\mathbf{x}^T\mathbf{A} = \alpha\mathbf{x}^T$, is called a *left eigenvector of \mathbf{A} corresponding to α* . Viewed in the same spirit, eigenvectors of \mathbf{A} as defined in Definition 8.3.1 are called *right eigenvectors*. Let λ_1 and λ_2 be distinct eigenvalues of \mathbf{A} . If \mathbf{x} is a left eigenvector of \mathbf{A} corresponding to λ_1 and \mathbf{y} is a right eigenvector of \mathbf{A} corresponding to λ_2 , then show that $\mathbf{x}^T\mathbf{y} = 0$.
13. Let λ be an eigenvalue of \mathbf{A} . Let r be the geometric multiplicity of λ . Show that the dimension of the space of the left eigenvectors of \mathbf{A} corresponding to λ is also r .
- *15. Let \mathbf{A} be an $n \times n$ matrix and let

$$\rho_i = \sum_{\substack{j=1 \\ j \neq i}}^n |a_{ij}| \quad (i = 1, \dots, n)$$

- (a) If α is an eigenvalue of \mathbf{A} , show that $|\alpha - a_{ii}| \leq \rho_i$ for at least one i .
16. Let α be an eigenvalue of \mathbf{A} . Then show that $|\alpha| \leq \|\mathbf{A}\|$ where $\|\cdot\|$ is the matrix norm induced by any vector norm.
17. (a) Let \mathbf{A} be an $n \times n$ idempotent matrix. Then show that $\mathcal{E}(\mathbf{A}) = \text{ES}(\mathbf{A}, 1)$ and $\mathcal{E}(\mathbf{I} - \mathbf{A}) = \text{ES}(\mathbf{A}, 0)$ and that \mathbf{A} has n linearly independent eigenvectors.
- (b) If each eigenvalue of \mathbf{A} is 0 or 1, does it follow that \mathbf{A} is idempotent?

- *18. Let f be a linear operator on a complex vector space V .
- (a) Show that there exists a subspace S of V with $d(S) = 1$ such that $f(S) \subseteq S$. An S satisfying the latter condition is said to be *invariant under f* .
19. Let \mathbf{A} be an $n \times n$ matrix and let \mathbf{D} be the $n \times n$ matrix with (i, j) -th element $\text{tr}(\mathbf{A}^{i+j-2})$. Show that the characteristic roots of \mathbf{A} are distinct iff \mathbf{D} is non-singular.
6. Prove Cayley-Hamilton theorem thus: let $\mathbf{H} := (\lambda \mathbf{I} - \mathbf{A})^{\otimes} = \mathbf{H}_0 + \lambda \mathbf{H}_1 + \cdots + \lambda^{n-1} \mathbf{H}_{n-1}$. Then $(\lambda \mathbf{I} - \mathbf{A})\mathbf{H} = \chi_{\mathbf{A}}(\lambda) \mathbf{I}$. Multiply by \mathbf{A}^i the equation obtained by comparing the coefficients of λ^i on the two sides and sum up to get $\chi_{\mathbf{A}}(\mathbf{A}) = \mathbf{0}$.
8. \mathbf{A} is said to be nilpotent if $\mathbf{A}^k = \mathbf{0}$ for some positive integer k . Show that \mathbf{A} is nilpotent iff all the characteristic roots of \mathbf{A} are 0.
9. Let \mathbf{A} be nilpotent.
- (a) If $\mathbf{A} \neq \mathbf{0}$, show that \mathbf{A} cannot be similar to a diagonal matrix.
- (b) What can you say about the minimal polynomial of \mathbf{A} ?
10. Show that the constant term in the minimal polynomial of \mathbf{A} is non-zero iff \mathbf{A} is non-singular.
13. Let \mathbf{A} and \mathbf{B} be square matrices of the same order and let $\mathbf{C} = \mathbf{AB} - \mathbf{BA}$. Show that $\mathbf{I} - \mathbf{C}$ is not nilpotent.
14. What is the minimal polynomial of $\alpha \mathbf{A}$?
15. Find the minimal polynomial of the $n \times n$ matrix $\mathbf{J} = \mathbf{11}^T$.
16. Prove that the minimal polynomial of $\text{diag}(\mathbf{A}, \mathbf{B})$ is the L.C.M. of the minimal polynomials of \mathbf{A} and \mathbf{B} .
1. If \mathbf{A} is a 2×2 matrix such that $\mathbf{A}^2 = \mathbf{0}$, show that either $\mathbf{A} = \mathbf{0}$ or \mathbf{A} is similar to $\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$.
2. Show that $\mathbf{A} = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$ is semi-simple iff either \mathbf{A} is a scalar matrix or $(a - d)^2 + 4bc \neq 0$.
3. If $\mathbf{A}^k = \mathbf{I}$ for some positive integer k , show that \mathbf{A} is semi-simple.
4. Show that \mathbf{A} is idempotent iff each eigenvalue of \mathbf{A} is 0 or 1 and \mathbf{A} is semi-simple.
5. If \mathbf{A} is a semi-simple matrix such that $\mathbf{A}^2 = \mathbf{A}^3$, show that \mathbf{A} is idempotent. Show also that the condition that \mathbf{A} is semi-simple cannot be dropped.
7. If \mathbf{A} is semi-simple, show that any polynomial in \mathbf{A} is also semi-simple.

8. Let $\mathbf{A} = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}$. Obtain a spectral decomposition of \mathbf{A} . Hence find \mathbf{A}^{10} and write down a spectral decomposition of \mathbf{A}^{-1} .
9. A matrix \mathbf{A} is said to be *stochastic* if $a_{ij} \geq 0$ for all i and j and $\sum_j a_{ij} = 1$ for all i . Let \mathbf{A} be a 2×2 stochastic matrix $\neq \mathbf{I}$.
- Find a spectral decomposition of \mathbf{A} .
 - Obtain an expression for \mathbf{A}^k where k is an arbitrary positive integer.
 - Show that there exists a 3×3 stochastic (upper triangular) matrix which is not semi-simple.
13. Let \mathbf{A} be semi-simple and let $\mathbf{A} = \sum_{i=1}^k \alpha_i \mathbf{E}_i$ be the spectral form of \mathbf{A} . Then prove that \mathbf{B} commutes with \mathbf{A} iff \mathbf{B} commutes with \mathbf{E}_i for $i = 1, \dots, k$.
14. Let \mathbf{A} be a square matrix with real eigenvalues such that $\rho(\mathbf{A}) = \rho(\mathbf{A}^2)$ and $\text{tr}(\mathbf{A}^2) \neq 0$. Then show that

$$\rho(\mathbf{A}) \geq \frac{(\text{tr}(\mathbf{A}))^2}{\text{tr}(\mathbf{A}^2)}$$

16. Let \mathbf{A} and \mathbf{B} be $n \times n$ semi-simple matrices. Show that the following statements are equivalent:
- $\mathbf{AB} = \mathbf{BA}$,
 - \mathbf{A} and \mathbf{B} are simultaneously diagonalizable (i.e., there exists a non-singular matrix \mathbf{P} such that $\mathbf{P}^{-1}\mathbf{AP}$ and $\mathbf{P}^{-1}\mathbf{BP}$ are diagonal),
 - \mathbf{A} and \mathbf{B} are polynomials in a common semi-simple matrix.
4. Let \mathbf{A} be a real skew-symmetric matrix of order n .
- If n is odd, show that $|\mathbf{A}| = 0$.
 - If n is even, show that $|\mathbf{A}| \geq 0$.
 - For any n , show that $|\mathbf{I} + \mathbf{A}| \geq 1$.
- (b) Find the spectral form of the $(k+1) \times (k+1)$ matrix $\mathbf{B} = \begin{bmatrix} \mathbf{0} & \mathbf{1} \\ \mathbf{1}^T & \mathbf{0} \end{bmatrix}$.
(Hint: use rank-factorization.)
8. Find a normal matrix which is none of: hermitian, skew-hermitian, unitary and diagonal.
9. Show that a normal matrix is unitary iff every eigenvalue has unit modulus.
11. Prove or disprove: every complex symmetric matrix is normal.
12. Show that the $n \times n$ matrix $\mathbf{1}\mathbf{1}^T$ is similar to the $n \times n$ matrix $\begin{bmatrix} n & \mathbf{0} \\ \mathbf{0} & \mathbf{0} \end{bmatrix}$.

13. Show that \mathbf{A} is an orthogonal projector iff \mathbf{A} is hermitian and each eigenvalue of \mathbf{A} belongs to $\{0, 1\}$.
14. Prove that \mathbf{A} is normal iff $\|\mathbf{A}^* \mathbf{x}\| = \|\mathbf{A} \mathbf{x}\|$ for all \mathbf{x} .
15. Let \mathbf{A} be a real symmetric matrix.
 - (a) If $\mathbf{A}^k = \mathbf{I}$ for some positive integer k , show that $\mathbf{A}^2 = \mathbf{I}$.
 - (b) If the eigenvalues of \mathbf{A} are all positive and if $\mathbf{A}^k = \mathbf{I}$ for some positive integer k then show that $\mathbf{A} = \mathbf{I}$.
 - (c) If $\mathbf{A}^k = \mathbf{0}$ for some positive integer k , then show that $\mathbf{A} = \mathbf{0}$.