

Justify all responses with clear explanations and in complete sentences unless otherwise stated. Write up your solutions cleanly and neatly and submit via Gradescope, making sure to select page submissions for each problem. Use of generative AI in any manner is not allowed on this or any other course assignments.

Part I: No justifications are required for these problems. Answers will be graded on correctness.

1. Identify each of the following statements as true or false:

- (a) The function $Q(x, y) = xy$ on \mathbb{R}^2 is a quadratic form.
- (b) The function $Q(x, y, z) = x^2 - 4xy + xyz + z^2$ on \mathbb{R}^3 is a quadratic form.
- (c) The function $Q(f) = \int_0^1 x f(x)^2 dx$ on $\mathbb{R}[x]$ is a quadratic form.
- (d) Every quadratic form over \mathbb{R} is a bilinear form.
- (e) Every quadratic form over an arbitrary field is a bilinear form.
- (f) The second derivatives test classifies any critical point as a local minimum, local maximum, or saddle.
- (g) If both eigenvalues of the 2×2 real symmetric matrix S are positive, then the graph of $(x, y) \cdot S \cdot (x, y)^T = 1$ in \mathbb{R}^2 will be an ellipse.
- (h) If one eigenvalue of the 2×2 real symmetric matrix S is zero and the other is nonzero, then the graph of $(x, y) \cdot S \cdot (x, y)^T = 1$ in \mathbb{R}^2 will be a hyperbola.
- (i) The singular values of $T : V \rightarrow V$ are the absolute values of the eigenvalues of T .
- (j) If T is Hermitian, the singular values of $T : V \rightarrow V$ are absolute values of the eigenvalues of T .
- (k) The singular value decomposition of a matrix is unique.
 - (l) If $T : V \rightarrow W$ is linear, the pseudoinverse T^\dagger satisfies $TT^\dagger(\mathbf{w}) = \mathbf{w}$ for all $\mathbf{w} \in \text{im}(T)$.
- (m) If $T : V \rightarrow W$ is linear, the pseudoinverse T^\dagger satisfies $TT^\dagger(\mathbf{w}) = \mathbf{w}$ for all $\mathbf{w} \in \text{im}(T)^\perp$.
- (n) If $T : V \rightarrow V$ is an isomorphism, then $T^\dagger = T^{-1}$.

2. Consider the bilinear form $\Phi[(a, b), (c, d)] = 4ac - 2ad - 2bc + 7bd$ on \mathbb{R}^2 with associated quadratic form Q .

- (a) Write down Q explicitly and also find $[\Phi]_\beta$ for $\beta = \{(1, 0), (0, 1)\}$.
- (b) Find an orthonormal basis γ for \mathbb{R}^2 such that $[\Phi]_\gamma$ is diagonal, and compute the diagonalization $[\Phi]_\gamma$.
- (c) Describe the shape of the quadratic variety $Q(x, y) = 1$ in \mathbb{R}^2 as one of the 3 standard conic sections.
- (d) Classify the critical point of $Q(x, y)$ at $(0, 0)$ as a local minimum, local maximum, or saddle point.
- (e) Calculate the signature and index of Q , and determine the definiteness of Q .

3. Consider the quadratic form $Q(x, y, z) = 11x^2 + 40xy - 16xz - 16y^2 - 16yz + 5z^2$ on \mathbb{R}^3 .

- (a) Find the symmetric matrix S associated to the underlying bilinear form for Q with respect to the standard basis $\beta = \{(1, 0, 0), (0, 1, 0), (0, 0, 1)\}$.
- (b) Give an explicit orthonormal change of basis that diagonalizes Q , and find the resulting diagonalization.
- (c) Describe the shape of the quadratic variety $Q(x, y, z) = 1$ in \mathbb{R}^3 as one of the 9 standard quadric surfaces.
- (d) Classify the critical point of $Q(x, y, z)$ at $(0, 0, 0)$ as a local minimum, local maximum, or saddle point.
- (e) Calculate the signature and index of Q , and determine the definiteness of Q .

4. For each matrix M , find (i) the singular values of M , (ii) a singular value decomposition $M = U\Sigma V^*$ where U and V are unitary and Σ is a rectangular diagonal matrix, and (iii) the pseudoinverse M^\dagger of M :

- (a) $\begin{bmatrix} 4 & 2 \\ 2 & 7 \end{bmatrix}$.
- (b) $\begin{bmatrix} 1 & 2 \\ 4 & 8 \end{bmatrix}$.
- (c) $\begin{bmatrix} -2 & 2 & 3 \\ 2 & 1 & 6 \end{bmatrix}$.
- (d) $\begin{bmatrix} -2 & 2 \\ 2 & 1 \\ 3 & 6 \end{bmatrix}$.
- (e) $\begin{bmatrix} 1 & i & -1 & -i \\ 2 & 2 & 2 & 2 \end{bmatrix}$.

5. Let $A = \begin{bmatrix} 2 & -8 & 2 \\ 6 & 6 & -9 \end{bmatrix}$.

- Find singular value decompositions for A and for A^T .
- Find the pseudoinverses A^\dagger and $(A^T)^\dagger$.
- Find the solution \mathbf{x} to the system $A\mathbf{x} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$ of minimal norm.
- Find the least-squares solution to the inconsistent system $A^T\mathbf{x} = \begin{bmatrix} 1 \\ -1 \\ 4 \end{bmatrix}$.

Part II: Solve the following problems. Justify all answers with rigorous, clear explanations.

6. In multivariable calculus, the following more explicit version of the second derivative test is often taught¹:

- Theorem** (Second Derivatives Test in \mathbb{R}^2): Suppose P is a critical point of $f(x, y)$, and let D be the value of the discriminant $f_{xx}f_{yy} - f_{xy}^2$ at P . If $D > 0$ and $f_{xx} > 0$, then the critical point is a minimum. If $D > 0$ and $f_{xx} < 0$, then the critical point is a maximum. If $D < 0$, then the critical point is a saddle point. If $D = 0$, then the test is inconclusive.

Using our general version of the second derivatives test, prove this variation. [Hint: Note $D = \det(H) = \lambda_1\lambda_2$.]

7. By the singular value decomposition theorem, if $T : V \rightarrow W$ is a linear transformation of rank r , then there exist orthonormal bases $\beta = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$ of V and $\gamma = \{\mathbf{w}_1, \dots, \mathbf{w}_m\}$ of W along with scalars $\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_r > 0$ with $T(\mathbf{v}_i) = \sigma_i\mathbf{w}_i$ for $1 \leq i \leq r$ and $T(\mathbf{v}_i) = \mathbf{0}$ for $i > r$.

- Show that $T^*(\mathbf{w}_i) = \sigma_i\mathbf{v}_i$ for $1 \leq i \leq r$ and $T^*(\mathbf{w}_i) = \mathbf{0}$ for $i > r$. [Hint: Consider $[T]_\beta^\gamma$ and $[T^*]_\gamma^\beta$.]
- Show that $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$ is a set of eigenvectors for T^*T with corresponding eigenvalues $\sigma_1^2, \dots, \sigma_r^2, 0, \dots, 0$, and that $\{\mathbf{w}_1, \dots, \mathbf{w}_m\}$ is a set of eigenvectors for TT^* with corresponding eigenvalues $\sigma_1^2, \dots, \sigma_r^2, 0, \dots, 0$.
- Deduce that the nonzero eigenvalues of T^*T and TT^* are the same, and hence that the nonzero singular values of T and T^* are the same.
- Show that if $A \in M_{m \times n}(\mathbb{C})$, then the singular values of A and A^* are the same, and that if A^* has a singular value decomposition $A^* = U\Sigma V^*$ then A has a singular value decomposition $A = V\Sigma^T U^*$.

Remark: The results of this problem are useful in computing the SVD of a non-square matrix, since one may just find the nonzero eigenvalues and eigenvectors of the smaller of A^*A and AA^* to construct $\{\mathbf{v}_1, \dots, \mathbf{v}_r\}$ and $\{\mathbf{w}_1, \dots, \mathbf{w}_r\}$, and then compute $\ker(A)$ to get $\{\mathbf{v}_{r+1}, \dots, \mathbf{v}_n\}$ and $\ker(A^*)$ to get $\{\mathbf{w}_{r+1}, \dots, \mathbf{w}_m\}$.

8. [Challenge] A Hermitian matrix A is said to be positive-definite if $\mathbf{v}^*A\mathbf{v} > 0$ for every $\mathbf{v} \neq \mathbf{0}$. The goal of this problem is to prove Sylvester's criterion for positive-definiteness: if A is an $n \times n$ Hermitian matrix, then A is positive definite if and only if $\det A^{(k)} > 0$ for all $1 \leq k \leq n$, where $A^{(k)}$ is the upper $k \times k$ submatrix of A . So suppose $A \in M_{n \times n}(\mathbb{C})$ is Hermitian.

- If A is positive definite, show that $A^{(k)}$ is positive definite for each $1 \leq k \leq n$ and deduce that $\det A^{(k)} > 0$ for all $1 \leq k \leq n$.
- Suppose that A has two orthonormal eigenvectors $\mathbf{v}_1, \mathbf{v}_2$ whose eigenvalues λ_1, λ_2 are negative. Show that $A^{(k-1)}$ is not positive definite. [Hint: Show that there exists a linear combination $\mathbf{w} = a_1\mathbf{v}_1 + a_2\mathbf{v}_2$ whose last coordinate is zero, and then that $\mathbf{w}^*A\mathbf{w} < 0$.]
- Deduce that if $A^{(k-1)}$ is positive definite and $\det(A) > 0$, then all eigenvalues of A must be positive and hence A is positive definite.
- Suppose that $\det A^{(k)} > 0$ for all $1 \leq k \leq n$. Show that A is positive definite.

¹The statement of this theorem is copied directly from my multivariable calculus course notes, in fact!

9. [Extra Bonus Challenge] The goal of this problem is to discuss matrix square roots and the matrix analogue of the polar form $z = e^{i\theta}r$ of a complex number. Let $A \in M_{n \times n}(\mathbb{C})$.

- (a) Show that there exists a unitary matrix W and a positive semidefinite Hermitian matrix P such that $A = WP$; this is called a (right) polar decomposition of A with W being the analogue of $e^{i\theta}$ and P being the analogue of r . [Hint: Take $W = UV^*$ and $P = V\Sigma V^*$.]
- (b) Show that if B is a positive-semidefinite Hermitian matrix such that $B^2 = \mu I_n$ for some nonnegative scalar μ , then $B = \sqrt{\mu}I_n$.
- (c) Show that if A is a positive-semidefinite Hermitian matrix, then there exists a unique positive-semidefinite Hermitian matrix B satisfying $B^2 = A$ (i.e., a “square root” of A). [Hint: Reduce to the case where A is diagonal, and then use part (b) along with 5(a) from homework 8 on each eigenspace of A .]
- (d) Suppose P and Q are positive-semidefinite Hermitian matrices and $P^2 = Q^2$. Show that $P = Q$.
- (e) Show that the polar decomposition of an invertible matrix A is unique. [Hint: Show first that P is invertible and then that $WP = ZQ$ implies $P^2 = Q^2$.]

Remark: The usual procedure for finding the polar form of a complex number $z = e^{i\theta}r$ is to note that $r = \sqrt{|z|^2} = \sqrt{\bar{z}z}$ and then $e^{i\theta} = z/r$. For the polar decomposition $A = WP$ we have an analogous formula: $P = \sqrt{A^*A}$, where the square root here denotes the positive-semidefinite matrix square root of (c), and when P is positive-definite we obtain the unitary part W via $W = AP^{-1}$.
