

# Practice Problems: Mathematical Foundations for Machine Learning

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## 1 Solution of Linear Systems and Vector Spaces

**Question 1** (Easy). Solve the following system of linear equations:

$$\begin{aligned}x + 2y - z &= 3 \\2x + y + z &= 6 \\x - y + 2z &= 3\end{aligned}$$

**Solution.** We write this system as an augmented matrix and use Gaussian elimination to reduce it to row echelon form.

$$\left[ \begin{array}{ccc|c} 1 & 2 & -1 & 3 \\ 2 & 1 & 1 & 6 \\ 1 & -1 & 2 & 3 \end{array} \right] \xrightarrow[R_3 \rightarrow R_3 - R_1]{R_2 \rightarrow R_2 - 2R_1} \left[ \begin{array}{ccc|c} 1 & 2 & -1 & 3 \\ 0 & -3 & 3 & 0 \\ 0 & -3 & 3 & 0 \end{array} \right] \xrightarrow{R_2 \rightarrow -\frac{1}{3}R_2} \left[ \begin{array}{ccc|c} 1 & 2 & -1 & 3 \\ 0 & 1 & -1 & 0 \\ 0 & -3 & 3 & 0 \end{array} \right] \xrightarrow{R_3 \rightarrow R_3 + 3R_2} \left[ \begin{array}{ccc|c} 1 & 2 & -1 & 3 \\ 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right]$$

The system is now in row echelon form. The matrix corresponds to the equations:

$$\begin{aligned}x + 2y - z &= 3 \\y - z &= 0\end{aligned}$$

From the second equation, we have  $y = z$ . Since the third row is all zeros, we have a free variable. Let  $z = t$ , where  $t$  is any real number. This implies  $y = t$ . Substitute  $y = t$  and  $z = t$  into the first equation:  $x + 2(t) - t = 3 \implies x + t = 3 \implies x = 3 - t$ . The general solution is a line in  $\mathbb{R}^3$  described by the vector  $(x, y, z) = (3 - t, t, t)$  for any  $t \in \mathbb{R}$ .

**Question 2** (Easy). Determine if the following set of vectors in  $\mathbb{R}^3$  is linearly independent:

$$S = \{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\} = \left\{ \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}, \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} \right\}$$

**Solution.** To check for linear independence, we form a matrix  $A$  whose columns are the vectors in  $S$  and check its determinant. The vectors are linearly independent if and only if the determinant is non-zero.

$$A = \begin{pmatrix} 1 & -1 & 0 \\ 0 & 1 & 1 \\ 1 & 0 & 1 \end{pmatrix}$$

$$\begin{aligned}\det(A) &= 1 \begin{vmatrix} 1 & 1 \\ 0 & 1 \end{vmatrix} - (-1) \begin{vmatrix} 0 & 1 \\ 1 & 1 \end{vmatrix} + 0 \begin{vmatrix} 0 & 1 \\ 1 & 0 \end{vmatrix} \\ &= 1(1 \cdot 1 - 1 \cdot 0) + 1(0 \cdot 1 - 1 \cdot 1) + 0 = 1(1) + 1(-1) = 1 - 1 = 0\end{aligned}$$

Since the determinant is 0, the set of vectors is **linearly dependent**.

Alternatively, we can try to find a non-trivial solution to  $c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + c_3\mathbf{v}_3 = \mathbf{0}$ . This gives the system:

$$\begin{aligned}c_1 - c_2 &= 0 \implies c_1 = c_2 \\ c_2 + c_3 &= 0 \implies c_3 = -c_2 \\ c_1 + c_3 &= 0\end{aligned}$$

Substituting the first two relations into the third gives  $(c_2) + (-c_2) = 0$ , which is  $0 = 0$ . This means there are infinitely many solutions. For a non-trivial solution, pick  $c_2 = 1$ . Then  $c_1 = 1$  and  $c_3 = -1$ . Indeed,  $1\mathbf{v}_1 + 1\mathbf{v}_2 - 1\mathbf{v}_3 = \mathbf{0}$ . Since a non-trivial solution exists, the vectors are linearly dependent.

**Question 3** (Moderate). Let  $\mathcal{B} = \{\mathbf{v}_1, \mathbf{v}_2\}$  be a basis for a subspace  $W \subset \mathbb{R}^3$ , where  $\mathbf{v}_1 = (1, 2, 0)$  and  $\mathbf{v}_2 = (3, 1, -1)$ . Determine if the vector  $\mathbf{u} = (-1, 3, 1)$  is in the subspace  $W$ . If it is, express  $\mathbf{u}$  as a linear combination of the basis vectors.

**Solution.** The vector  $\mathbf{u}$  is in the subspace  $W$  if and only if it can be written as a linear combination of the basis vectors  $\mathbf{v}_1$  and  $\mathbf{v}_2$ . We are looking for scalars  $c_1, c_2$  such that  $c_1\mathbf{v}_1 + c_2\mathbf{v}_2 = \mathbf{u}$ .

$$c_1 \begin{pmatrix} 1 \\ 2 \\ 0 \end{pmatrix} + c_2 \begin{pmatrix} 3 \\ 1 \\ -1 \end{pmatrix} = \begin{pmatrix} -1 \\ 3 \\ 1 \end{pmatrix}$$

This vector equation corresponds to a system of three linear equations:

$$\begin{aligned}c_1 + 3c_2 &= -1 \\ 2c_1 + c_2 &= 3 \\ -c_2 &= 1\end{aligned}$$

From the third equation, we immediately get  $c_2 = -1$ . Substitute  $c_2 = -1$  into the second equation:  $2c_1 + (-1) = 3 \implies 2c_1 = 4 \implies c_1 = 2$ . Now we have a potential solution  $(c_1, c_2) = (2, -1)$ . We must check if these values satisfy the remaining equation (the first one) to ensure the system is consistent. Check the first equation:  $c_1 + 3c_2 = (2) + 3(-1) = 2 - 3 = -1$ . The equation holds true. Since we found a consistent solution, the vector  $\mathbf{u}$  is in the subspace  $W$ . The linear combination is  $\mathbf{u} = 2\mathbf{v}_1 - \mathbf{v}_2$ .

**Question 4** (Moderate). Find a basis for the column space and the rank of the matrix  $A$ :

$$A = \begin{pmatrix} 1 & 2 & 0 & 1 \\ 2 & 4 & 1 & 3 \\ 0 & 0 & 1 & 1 \end{pmatrix}$$

**Solution.** To find a basis for the column space  $\text{Col}(A)$ , we reduce the matrix  $A$  to its row echelon form to identify the pivot columns.

$$\begin{pmatrix} 1 & 2 & 0 & 1 \\ 2 & 4 & 1 & 3 \\ 0 & 0 & 1 & 1 \end{pmatrix} \xrightarrow{R_2 \rightarrow R_2 - 2R_1} \begin{pmatrix} 1 & 2 & 0 & 1 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 1 & 1 \end{pmatrix} \xrightarrow{R_3 \rightarrow R_3 - R_2} \begin{pmatrix} 1 & 2 & 0 & 1 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

The leading 1's (pivots) are in the first and third columns of the row echelon form. Therefore, the corresponding columns of the **original matrix**  $\mathbf{A}$  form a basis for its column space. The pivot columns are column 1 and column 3. A basis for  $\text{Col}(\mathbf{A})$  is  $\left\{ \begin{pmatrix} 1 \\ 2 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} \right\}$ . The rank of a matrix is the dimension of its column space (which is equal to the number of pivot columns). There are 2 pivots, so the **rank of  $\mathbf{A}$  is 2**.

**Question 5** (Moderate). Let  $W$  be the set of all vectors in  $\mathbb{R}^3$  of the form  $(a, b, a + b)$ . Show that  $W$  is a subspace of  $\mathbb{R}^3$ .

**Solution.** To show that  $W$  is a subspace of  $\mathbb{R}^3$ , we must verify three properties:

1. **The zero vector is in  $W$ :** The zero vector in  $\mathbb{R}^3$  is  $\mathbf{0} = (0, 0, 0)$ . We must check if it can be written in the form  $(a, b, a + b)$ . If we choose  $a = 0$  and  $b = 0$ , the form becomes  $(0, 0, 0 + 0) = (0, 0, 0)$ . So,  $\mathbf{0} \in W$ . The set is non-empty.
2. **Closure under vector addition:** Let  $\mathbf{u}$  and  $\mathbf{v}$  be two arbitrary vectors in  $W$ . Then they can be written as:  $\mathbf{u} = (a_1, b_1, a_1 + b_1)$  for some scalars  $a_1, b_1$ .  $\mathbf{v} = (a_2, b_2, a_2 + b_2)$  for some scalars  $a_2, b_2$ . Their sum is:  $\mathbf{u} + \mathbf{v} = (a_1 + a_2, b_1 + b_2, (a_1 + b_1) + (a_2 + b_2)) = (a_1 + a_2, b_1 + b_2, (a_1 + a_2) + (b_1 + b_2))$ . Let  $a' = a_1 + a_2$  and  $b' = b_1 + b_2$ . Then the sum is of the form  $(a', b', a' + b')$ , which matches the definition of a vector in  $W$ . Thus,  $W$  is closed under addition.
3. **Closure under scalar multiplication:** Let  $\mathbf{u} = (a, b, a + b)$  be a vector in  $W$  and  $c$  be any scalar.  $c\mathbf{u} = c(a, b, a + b) = (ca, cb, c(a + b)) = (ca, cb, ca + cb)$ . Let  $a'' = ca$  and  $b'' = cb$ . Then the resulting vector is  $(a'', b'', a'' + b'')$ , which is in the form required for vectors in  $W$ . Thus,  $W$  is closed under scalar multiplication.

Since all three properties hold,  $W$  is a subspace of  $\mathbb{R}^3$ .

**Question 6** (Tough). Consider the system of equations:

$$\begin{aligned} x + y + kz &= 1 \\ x + ky + z &= 1 \\ kx + y + z &= -2 \end{aligned}$$

Find the values of  $k$  for which the system has: (a) a unique solution, (b) no solution, (c) infinitely many solutions.

**Solution.** The nature of the solution depends on the determinant of the coefficient matrix

$$A = \begin{pmatrix} 1 & 1 & k \\ 1 & k & 1 \\ k & 1 & 1 \end{pmatrix}.$$

$$\begin{aligned} \det(A) &= 1(k \cdot 1 - 1 \cdot 1) - 1(1 \cdot 1 - k \cdot 1) + k(1 \cdot 1 - k \cdot k) \\ &= (k - 1) - (1 - k) + k(1 - k^2) \\ &= k - 1 - 1 + k + k - k^3 = -k^3 + 3k - 2 \end{aligned}$$

To find the roots of this polynomial, we can test integer factors of -2. If  $k = 1$ ,  $\det(A) = -1 + 3 - 2 = 0$ . So  $(k - 1)$  is a factor. If  $k = -2$ ,  $\det(A) = -(-8) + 3(-2) - 2 = 8 - 6 - 2 = 0$ . So  $(k + 2)$  is a factor. By polynomial division or synthetic division, we find the complete factorization:  $\det(A) = -(k - 1)^2(k + 2)$ .

- (a) **Unique solution:** A unique solution exists if and only if  $\det(A) \neq 0$ . This occurs when  $-(k-1)^2(k+2) \neq 0$ , which means  $k \neq 1$  and  $k \neq -2$ .
- (b) **No solution or infinitely many solutions:** This occurs when  $\det(A) = 0$ , i.e., for  $k = 1$  or  $k = -2$ . We must analyze these cases using the augmented matrix.

**Case 1:  $k = 1$**

$$\left[ \begin{array}{ccc|c} 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & -2 \end{array} \right] \xrightarrow{R_3 \rightarrow R_3 - R_1} \left[ \begin{array}{ccc|c} 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 0 & 0 & 0 & -3 \end{array} \right]$$

The last row represents the equation  $0x + 0y + 0z = -3$ , which is  $0 = -3$ . This is a contradiction. Therefore, for  $k = 1$ , there is **no solution**.

**Case 2:  $k = -2$**

$$\left[ \begin{array}{ccc|c} 1 & 1 & -2 & 1 \\ 1 & -2 & 1 & 1 \\ -2 & 1 & 1 & -2 \end{array} \right] \xrightarrow[R_3 \rightarrow R_3 + 2R_1]{R_2 \rightarrow R_2 - R_1} \left[ \begin{array}{ccc|c} 1 & 1 & -2 & 1 \\ 0 & -3 & 3 & 0 \\ 0 & 3 & -3 & 0 \end{array} \right] \xrightarrow{R_3 \rightarrow R_3 + R_2} \left[ \begin{array}{ccc|c} 1 & 1 & -2 & 1 \\ 0 & -3 & 3 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right]$$

The last row is  $0 = 0$ , which is consistent. Since there is a row of zeros, we have at least one free variable, and the system is consistent. Therefore, for  $k = -2$ , there are **infinitely many solutions**.

**Question 7 (Tough).** Let the matrix  $A$  be given by:

$$A = \begin{pmatrix} 1 & -2 & 0 & 3 \\ 2 & -4 & 1 & 8 \\ -1 & 2 & -2 & -9 \end{pmatrix}$$

- (a) Find the Reduced Row Echelon Form (RREF) of  $A$ .
- (b) Find a basis for the null space,  $\text{Null}(A)$ .
- (c) Find a basis for the column space,  $\text{Col}(A)$ .
- (d) State the dimensions of the null space and column space and verify the Rank-Nullity Theorem.

**Solution.** (a) **RREF of  $A$ :** We perform Gaussian elimination.

$$\left[ \begin{array}{cccc} 1 & -2 & 0 & 3 \\ 2 & -4 & 1 & 8 \\ -1 & 2 & -2 & -9 \end{array} \right] \xrightarrow[R_3 \rightarrow R_3 + R_1]{R_2 \rightarrow R_2 - 2R_1} \left[ \begin{array}{cccc} 1 & -2 & 0 & 3 \\ 0 & 0 & 1 & 2 \\ 0 & 0 & -2 & -6 \end{array} \right] \xrightarrow{R_3 \rightarrow R_3 + 2R_2} \left[ \begin{array}{cccc} 1 & -2 & 0 & 3 \\ 0 & 0 & 1 & 2 \\ 0 & 0 & 0 & -2 \end{array} \right]$$

To get RREF, continue reducing:

$$\xrightarrow{R_3 \rightarrow -1/2 R_3} \left[ \begin{array}{cccc} 1 & -2 & 0 & 3 \\ 0 & 0 & 1 & 2 \\ 0 & 0 & 0 & 1 \end{array} \right] \xrightarrow[R_1 \rightarrow R_1 - 3R_3]{R_2 \rightarrow R_2 - 2R_3} \left[ \begin{array}{cccc} 1 & -2 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{array} \right]$$

This is the **RREF of  $A$** .

- (b) **Basis for  $\text{Null}(\mathbf{A})$ :** We solve  $\mathbf{A}\mathbf{x} = \mathbf{0}$  using the RREF. Let  $\mathbf{x} = (x_1, x_2, x_3, x_4)^T$ . The RREF gives the system:  $x_1 - 2x_2 = 0 \implies x_1 = 2x_2$   $x_3 = 0$   $x_4 = 0$ . The variable  $x_2$  corresponds to a non-pivot column, so it is a free variable. Let  $x_2 = t$ . The solution vector is  $\mathbf{x} = (2t, t, 0, 0)^T = t(2, 1, 0, 0)^T$ . A basis for  $\text{Null}(\mathbf{A})$  is the vector that spans this space:  $\left\{ \begin{pmatrix} 2 \\ 1 \\ 0 \\ 0 \end{pmatrix} \right\}$ .
- (c) **Basis for  $\text{Col}(\mathbf{A})$ :** The pivot columns in the RREF are columns 1, 3, and 4. We take the corresponding columns from the **original matrix  $\mathbf{A}$** . A basis for  $\text{Col}(\mathbf{A})$  is  $\left\{ \begin{pmatrix} 1 \\ 2 \\ -1 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ -2 \end{pmatrix}, \begin{pmatrix} 3 \\ 8 \\ -9 \end{pmatrix} \right\}$ .
- (d) **Rank-Nullity Theorem:** The dimension of the null space,  $\text{nullity}(\mathbf{A})$ , is the number of free variables, which is 1. The dimension of the column space,  $\text{rank}(\mathbf{A})$ , is the number of pivot columns, which is 3. The Rank-Nullity Theorem states that  $\text{rank}(\mathbf{A}) + \text{nullity}(\mathbf{A}) = n$ , where  $n$  is the number of columns of  $\mathbf{A}$ . Here,  $n = 4$ . We check:  $3 + 1 = 4$ . The theorem is verified.

## 2 Analytic Geometry

**Question 8** (Easy). Given the vectors  $\mathbf{u} = (2, -1, 3)$  and  $\mathbf{v} = (1, 1, -2)$  in  $\mathbb{R}^3$ .

(a) Find the Euclidean norm (length) of  $\mathbf{u}$ .

(b) Find the distance between  $\mathbf{u}$  and  $\mathbf{v}$ .

**Solution.** (a) The Euclidean norm  $\|\mathbf{u}\|$  is given by the square root of the sum of the squares of its components.

$$\|\mathbf{u}\| = \sqrt{2^2 + (-1)^2 + 3^2} = \sqrt{4 + 1 + 9} = \sqrt{14}$$

(b) The distance between  $\mathbf{u}$  and  $\mathbf{v}$  is the norm of their difference,  $\|\mathbf{u} - \mathbf{v}\|$ . First, find the difference vector:

$$\mathbf{u} - \mathbf{v} = (2 - 1, -1 - 1, 3 - (-2)) = (1, -2, 5)$$

Now, find the norm of this vector:

$$\|\mathbf{u} - \mathbf{v}\| = \sqrt{1^2 + (-2)^2 + 5^2} = \sqrt{1 + 4 + 25} = \sqrt{30}$$

**Question 9** (Easy). For the vectors  $\mathbf{a} = (1, 2, -3)$  and  $\mathbf{b} = (3, 1, 2)$ :

(a) Calculate the inner product (dot product)  $\mathbf{a} \cdot \mathbf{b}$ .

(b) Determine if the vectors are orthogonal.

**Solution.** (a) The inner product is calculated as the sum of the products of corresponding components:

$$\mathbf{a} \cdot \mathbf{b} = (1)(3) + (2)(1) + (-3)(2) = 3 + 2 - 6 = -1$$

(b) Two vectors are orthogonal if their inner product is zero. Since  $\mathbf{a} \cdot \mathbf{b} = -1 \neq 0$ , the vectors are **not orthogonal**.

**Question 10** (Moderate). Find a non-zero vector  $\mathbf{w}$  that is orthogonal to both  $\mathbf{u} = (1, 1, -2)$  and  $\mathbf{v} = (2, -1, 1)$ .

**Solution.** A vector that is orthogonal to two other vectors in  $\mathbb{R}^3$  can be found by computing their cross product. The cross product  $\mathbf{u} \times \mathbf{v}$  produces a vector that is orthogonal to both  $\mathbf{u}$  and  $\mathbf{v}$ .

$$\mathbf{w} = \mathbf{u} \times \mathbf{v} = \begin{vmatrix} \mathbf{i} & \mathbf{j} & \mathbf{k} \\ 1 & 1 & -2 \\ 2 & -1 & 1 \end{vmatrix}$$

where  $\mathbf{i}, \mathbf{j}, \mathbf{k}$  are the standard basis vectors for  $\mathbb{R}^3$ .

$$\mathbf{w} = \mathbf{i}(1 \cdot 1 - (-2) \cdot (-1)) - \mathbf{j}(1 \cdot 1 - (-2) \cdot 2) + \mathbf{k}(1 \cdot (-1) - 1 \cdot 2)$$

$$\mathbf{w} = \mathbf{i}(1 - 2) - \mathbf{j}(1 + 4) + \mathbf{k}(-1 - 2)$$

$$\mathbf{w} = -1\mathbf{i} - 5\mathbf{j} - 3\mathbf{k} = (-1, -5, -3)$$

To verify, we check the dot products:

$$\mathbf{w} \cdot \mathbf{u} = (-1)(1) + (-5)(1) + (-3)(-2) = -1 - 5 + 6 = 0$$

$$\mathbf{w} \cdot \mathbf{v} = (-1)(2) + (-5)(-1) + (-3)(1) = -2 + 5 - 3 = 0$$

Since both dot products are zero, the vector  $\mathbf{w} = (-1, -5, -3)$  is orthogonal to both  $\mathbf{u}$  and  $\mathbf{v}$ . Any non-zero scalar multiple of  $\mathbf{w}$ , such as  $(1, 5, 3)$ , is also a valid answer.

**Question 11** (Moderate). An inner product on  $\mathbb{R}^2$  can be defined by a symmetric matrix. Consider the inner product defined by the matrix  $A = \begin{pmatrix} 2 & -1 \\ -1 & 3 \end{pmatrix}$ , where  $\langle \mathbf{x}, \mathbf{y} \rangle = \mathbf{x}^T A \mathbf{y}$ .

Let  $\mathbf{x} = \begin{pmatrix} 1 \\ 2 \end{pmatrix}$  and  $\mathbf{y} = \begin{pmatrix} 3 \\ -1 \end{pmatrix}$ .

(a) Calculate the inner product  $\langle \mathbf{x}, \mathbf{y} \rangle$ .

(b) Calculate the norm of  $\mathbf{x}$  induced by this inner product,  $\|\mathbf{x}\|_A = \sqrt{\langle \mathbf{x}, \mathbf{x} \rangle}$ .

**Solution.** (a) **Calculate the inner product:**

$$\langle \mathbf{x}, \mathbf{y} \rangle = \mathbf{x}^T A \mathbf{y} = \begin{pmatrix} 1 & 2 \end{pmatrix} \begin{pmatrix} 2 & -1 \\ -1 & 3 \end{pmatrix} \begin{pmatrix} 3 \\ -1 \end{pmatrix}$$

First, compute  $A\mathbf{y}$ :

$$A\mathbf{y} = \begin{pmatrix} 2(3) + (-1)(-1) \\ -1(3) + 3(-1) \end{pmatrix} = \begin{pmatrix} 7 \\ -6 \end{pmatrix}$$

Now compute the full product:

$$\langle \mathbf{x}, \mathbf{y} \rangle = \begin{pmatrix} 1 & 2 \end{pmatrix} \begin{pmatrix} 7 \\ -6 \end{pmatrix} = 1(7) + 2(-6) = 7 - 12 = -5$$

(b) **Calculate the norm of  $\mathbf{x}$ :** The squared norm is  $\|\mathbf{x}\|_A^2 = \langle \mathbf{x}, \mathbf{x} \rangle = \mathbf{x}^T A \mathbf{x}$ .

$$\|\mathbf{x}\|_A^2 = \begin{pmatrix} 1 & 2 \end{pmatrix} \begin{pmatrix} 2 & -1 \\ -1 & 3 \end{pmatrix} \begin{pmatrix} 1 \\ 2 \end{pmatrix}$$

First, compute  $A\mathbf{x}$ :

$$A\mathbf{x} = \begin{pmatrix} 2(1) + (-1)(2) \\ -1(1) + 3(2) \end{pmatrix} = \begin{pmatrix} 0 \\ 5 \end{pmatrix}$$

Now compute the full product:

$$\|\mathbf{x}\|_A^2 = \begin{pmatrix} 1 & 2 \end{pmatrix} \begin{pmatrix} 0 \\ 5 \end{pmatrix} = 1(0) + 2(5) = 10$$

The norm is therefore  $\|\mathbf{x}\|_A = \sqrt{10}$ .

**Question 12** (Moderate). Given the vectors  $\mathbf{v}_1 = (1, 1, 0)$ ,  $\mathbf{v}_2 = (1, 0, 1)$ , and  $\mathbf{v}_3 = (0, 1, 1)$ , which form a basis for  $\mathbb{R}^3$ . Use the Gram-Schmidt process to find an orthonormal basis  $\{\mathbf{q}_1, \mathbf{q}_2, \mathbf{q}_3\}$ .

## Solution

The Gram-Schmidt process transforms a set of linearly independent vectors into an orthogonal basis, which can then be normalized to create an orthonormal basis. Let the given vectors be  $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$ . We will first construct an orthogonal basis  $\{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3\}$  and then normalize each vector to get the orthonormal basis  $\{\mathbf{q}_1, \mathbf{q}_2, \mathbf{q}_3\}$ .

### Step 1: Construct $\mathbf{q}_1$ from $\mathbf{v}_1$

First, we set the initial orthogonal vector  $\mathbf{u}_1$  to be equal to  $\mathbf{v}_1$ .

$$\mathbf{u}_1 = \mathbf{v}_1 = (1, 1, 0)$$

Next, we find its magnitude (norm):

$$\|\mathbf{u}_1\| = \sqrt{1^2 + 1^2 + 0^2} = \sqrt{2}$$

The first orthonormal vector  $\mathbf{q}_1$  is found by normalizing  $\mathbf{u}_1$ :

$$\mathbf{q}_1 = \frac{\mathbf{u}_1}{\|\mathbf{u}_1\|} = \frac{1}{\sqrt{2}}(1, 1, 0)$$

### Step 2: Construct $\mathbf{q}_2$ from $\mathbf{v}_2$

The second orthogonal vector,  $\mathbf{u}_2$ , is found by subtracting the projection of  $\mathbf{v}_2$  onto  $\mathbf{u}_1$  from  $\mathbf{v}_2$ .

$$\mathbf{u}_2 = \mathbf{v}_2 - \text{proj}_{\mathbf{u}_1}(\mathbf{v}_2) = \mathbf{v}_2 - \frac{\mathbf{v}_2 \mathbf{u}_1}{\|\mathbf{u}_1\|^2} \mathbf{u}_1$$

First, we calculate the dot product  $\mathbf{v}_2 \mathbf{u}_1$ :

$$\mathbf{v}_2 \mathbf{u}_1 = (1, 0, 1)(1, 1, 0) = 1(1) + 0(1) + 1(0) = 1$$

Now, substitute this into the equation for  $\mathbf{u}_2$ :

$$\mathbf{u}_2 = (1, 0, 1) - \frac{1}{2}(1, 1, 0) = \left(1 - \frac{1}{2}, 0 - \frac{1}{2}, 1 - 0\right) = \left(\frac{1}{2}, -\frac{1}{2}, 1\right)$$

To simplify the arithmetic in the next step, we can use a scaled version of  $\mathbf{u}_2$ , as its direction is what matters. Let's use  $\mathbf{u}'_2 = 2\mathbf{u}_2$ :

$$\mathbf{u}'_2 = 2\left(\frac{1}{2}, -\frac{1}{2}, 1\right) = (1, -1, 2)$$

Now, we normalize this cleaner vector to find  $\mathbf{q}_2$ :

$$\|\mathbf{u}'_2\| = \sqrt{1^2 + (-1)^2 + 2^2} = \sqrt{1 + 1 + 4} = \sqrt{6}$$

$$\mathbf{q}_2 = \frac{\mathbf{u}'_2}{\|\mathbf{u}'_2\|} = \frac{1}{\sqrt{6}}(1, -1, 2)$$

### Step 3: Construct $\mathbf{q}_3$ from $\mathbf{v}_3$

The third orthogonal vector,  $\mathbf{u}_3$ , is found by subtracting the projections of  $\mathbf{v}_3$  onto the previously found orthogonal vectors,  $\mathbf{u}_1$  and  $\mathbf{u}'_2$ .

$$\mathbf{u}_3 = \mathbf{v}_3 - \text{proj}_{\mathbf{u}_1}(\mathbf{v}_3) - \text{proj}_{\mathbf{u}'_2}(\mathbf{v}_3) = \mathbf{v}_3 - \frac{\mathbf{v}_3 \mathbf{u}_1}{\|\mathbf{u}_1\|^2} \mathbf{u}_1 - \frac{\mathbf{v}_3 \mathbf{u}'_2}{\|\mathbf{u}'_2\|^2} \mathbf{u}'_2$$

We calculate the required dot products first:

$$\mathbf{v}_3 \mathbf{u}_1 = (0, 1, 1)(1, 1, 0) = 0(1) + 1(1) + 1(0) = 1$$

$$\mathbf{v}_3 \mathbf{u}'_2 = (0, 1, 1)(1, -1, 2) = 0(1) + 1(-1) + 1(2) = 1$$



Now, we substitute these values into the equation for  $\mathbf{u}_3$ :

$$\begin{aligned}\mathbf{u}_3 &= (0, 1, 1) - \frac{1}{2}(1, 1, 0) - \frac{1}{6}(1, -1, 2) \\ &= (0, 1, 1) - \left(\frac{1}{2}, \frac{1}{2}, 0\right) - \left(\frac{1}{6}, -\frac{1}{6}, \frac{1}{3}\right) \\ &= \left(0 - \frac{1}{2} - \frac{1}{6}, 1 - \frac{1}{2} + \frac{1}{6}, 1 - 0 - \frac{1}{3}\right) \\ &= \left(-\frac{3}{6} - \frac{1}{6}, \frac{6}{6} - \frac{3}{6} + \frac{1}{6}, \frac{3}{3} - \frac{1}{3}\right) \\ &= \left(-\frac{4}{6}, \frac{4}{6}, \frac{2}{3}\right) = \left(-\frac{2}{3}, \frac{2}{3}, \frac{2}{3}\right)\end{aligned}$$

Again, we can simplify by scaling. Let's use  $\mathbf{u}'_3 = \frac{3}{2}\mathbf{u}_3$  to eliminate the fractions:

$$\mathbf{u}'_3 = \frac{3}{2} \left(-\frac{2}{3}, \frac{2}{3}, \frac{2}{3}\right) = (-1, 1, 1)$$

Finally, we normalize  $\mathbf{u}'_3$  to get  $\mathbf{q}_3$ :

$$\begin{aligned}\|\mathbf{u}'_3\| &= \sqrt{(-1)^2 + 1^2 + 1^2} = \sqrt{3} \\ \mathbf{q}_3 &= \frac{\mathbf{u}'_3}{\|\mathbf{u}'_3\|} = \frac{1}{\sqrt{3}}(-1, 1, 1)\end{aligned}$$

## Conclusion: The Orthonormal Basis

The resulting orthonormal basis for  $\mathbb{R}^3$  is  $\{\mathbf{q}_1, \mathbf{q}_2, \mathbf{q}_3\}$ , where:

$$\begin{aligned}\mathbf{q}_1 &= \frac{1}{\sqrt{2}}(1, 1, 0) \\ \mathbf{q}_2 &= \frac{1}{\sqrt{6}}(1, -1, 2) \\ \mathbf{q}_3 &= \frac{1}{\sqrt{3}}(-1, 1, 1)\end{aligned}$$

**Question 13** (Tough). Let  $W$  be the subspace of  $\mathbb{R}^4$  spanned by the orthogonal vectors  $\mathbf{u}_1 = (1, 1, 0, 1)$  and  $\mathbf{u}_2 = (1, -1, 1, 0)$ . Let  $\mathbf{y} = (2, 3, 2, 1)$ . Find the orthogonal projection of  $\mathbf{y}$  onto the subspace  $W$ , and find the shortest distance from  $\mathbf{y}$  to  $W$ .

**Solution.** The **orthogonal projection** of  $\mathbf{y}$  onto the subspace  $W$ , denoted  $\hat{\mathbf{y}}$ , is the point in  $W$  closest to  $\mathbf{y}$ . Since  $W$  is spanned by an orthogonal set of vectors  $\{\mathbf{u}_1, \mathbf{u}_2\}$ , the formula is:

$$\hat{\mathbf{y}} = \text{proj}_W(\mathbf{y}) = \frac{\mathbf{y} \cdot \mathbf{u}_1}{\mathbf{u}_1 \cdot \mathbf{u}_1} \mathbf{u}_1 + \frac{\mathbf{y} \cdot \mathbf{u}_2}{\mathbf{u}_2 \cdot \mathbf{u}_2} \mathbf{u}_2$$

First, we calculate the required dot products:

$$\mathbf{y} \cdot \mathbf{u}_1 = (2)(1) + (3)(1) + (2)(0) + (1)(1) = 2 + 3 + 0 + 1 = 6$$

$$\mathbf{u}_1 \cdot \mathbf{u}_1 = \|\mathbf{u}_1\|^2 = 1^2 + 1^2 + 0^2 + 1^2 = 3$$

$$\mathbf{y} \cdot \mathbf{u}_2 = (2)(1) + (3)(-1) + (2)(1) + (1)(0) = 2 - 3 + 2 + 0 = 1$$

$$\mathbf{u}_2 \cdot \mathbf{u}_2 = \|\mathbf{u}_2\|^2 = 1^2 + (-1)^2 + 1^2 + 0^2 = 3$$

Now, substitute these values into the projection formula:

$$\hat{\mathbf{y}} = \frac{6}{3}\mathbf{u}_1 + \frac{1}{3}\mathbf{u}_2 = 2(1, 1, 0, 1) + \frac{1}{3}(1, -1, 1, 0)$$

$$\hat{\mathbf{y}} = (2, 2, 0, 2) + \left(\frac{1}{3}, -\frac{1}{3}, \frac{1}{3}, 0\right) = \left(\frac{7}{3}, \frac{5}{3}, \frac{1}{3}, 2\right)$$

This vector  $\hat{\mathbf{y}}$  is the orthogonal projection of  $\mathbf{y}$  onto  $W$ .

The **shortest distance** from  $\mathbf{y}$  to  $W$  is the norm of the component of  $\mathbf{y}$  orthogonal to  $W$ , which is the vector  $\mathbf{z} = \mathbf{y} - \hat{\mathbf{y}}$ .

$$\mathbf{z} = (2, 3, 2, 1) - \left(\frac{7}{3}, \frac{5}{3}, \frac{1}{3}, 2\right) = \left(2 - \frac{7}{3}, 3 - \frac{5}{3}, 2 - \frac{1}{3}, 1 - 2\right) = \left(-\frac{1}{3}, \frac{4}{3}, \frac{5}{3}, -1\right)$$

The distance is the norm of  $\mathbf{z}$ :

$$\|\mathbf{z}\| = \sqrt{\left(-\frac{1}{3}\right)^2 + \left(\frac{4}{3}\right)^2 + \left(\frac{5}{3}\right)^2 + (-1)^2} = \sqrt{\frac{1}{9} + \frac{16}{9} + \frac{25}{9} + \frac{9}{9}}$$

$$\|\mathbf{z}\| = \sqrt{\frac{1 + 16 + 25 + 9}{9}} = \sqrt{\frac{51}{9}} = \frac{\sqrt{51}}{3}$$

The shortest distance is  $\frac{\sqrt{51}}{3}$ .

**Question 14** (Tough). Find the orthogonal projection of the point  $P(1, 0, 0)$  onto the plane defined by the equation  $x + 2y + z = 6$ .

**Solution.** Let the plane be  $\Pi$  and the point be  $P$ . Let the projection of  $P$  onto the plane be  $Q$ . **Step 1: Identify the Normal Vector** The vector  $\vec{PQ}$  must be orthogonal to the plane. The normal vector to the plane  $x + 2y + z = 6$  is given by the coefficients of  $x, y, z$ . So,  $\mathbf{n} = \langle 1, 2, 1 \rangle$ .

**Step 2: Define the line from  $P$  to  $Q$**  The vector  $\vec{PQ}$  must be parallel to the normal vector  $\mathbf{n}$ . This means we can define a line from  $P$  to  $Q$  as  $Q(t) = P + t\mathbf{n}$  for some scalar  $t$ . Let the coordinates of the projection point  $Q$  be  $(q_1, q_2, q_3)$ .  $Q(t) = (1, 0, 0) + t(1, 2, 1) = (1 + t, 2t, t)$ . So,  $q_1 = 1 + t$ ,  $q_2 = 2t$ ,  $q_3 = t$ .

**Step 3: Solve for  $t$**  Since the point  $Q$  lies on the plane  $\Pi$ , its coordinates must satisfy the plane's equation. Substitute the expressions for  $q_1, q_2, q_3$  into the plane equation:

$$(1 + t) + 2(2t) + (t) = 6$$

$$1 + t + 4t + t = 6$$

$$6t + 1 = 6 \implies 6t = 5 \implies t = \frac{5}{6}$$

**Step 4: Find the coordinates of  $Q$**  Now we find the coordinates of the projection point  $Q$  by substituting  $t = 5/6$  back into our expressions from Step 2:  $q_1 = 1 + \frac{5}{6} = \frac{11}{6}$ ,  $q_2 = 2\left(\frac{5}{6}\right) = \frac{10}{6} = \frac{5}{3}$ ,  $q_3 = \frac{5}{6}$ . The orthogonal projection of the point  $P(1, 0, 0)$  onto the plane is the point  $Q\left(\frac{11}{6}, \frac{5}{3}, \frac{5}{6}\right)$ .

### 3 Determinant, Eigenvalues, SVD

**Question 15** (Easy). For the matrix  $A = \begin{pmatrix} 1 & -2 & 3 \\ 0 & 4 & -1 \\ 2 & 1 & 5 \end{pmatrix}$ :

(a) Calculate the Trace of  $A$ .

(b) Calculate the Determinant of  $A$ .

**Solution.** (a) **Trace of  $A$ :** The trace of a square matrix is the sum of the elements on its main diagonal.

$$\text{Tr}(A) = 1 + 4 + 5 = 10$$

(b) **Determinant of  $A$ :** The determinant can be calculated using cofactor expansion. Let's expand along the first row.

$$\begin{aligned} \det(A) &= 1 \begin{vmatrix} 4 & -1 \\ 1 & 5 \end{vmatrix} - (-2) \begin{vmatrix} 0 & -1 \\ 2 & 5 \end{vmatrix} + 3 \begin{vmatrix} 0 & 4 \\ 2 & 1 \end{vmatrix} \\ &= 1(4(5) - (-1)(1)) + 2(0(5) - (-1)(2)) + 3(0(1) - 4(2)) \\ &= 1(20 + 1) + 2(0 + 2) + 3(0 - 8) \\ &= 1(21) + 2(2) + 3(-8) \\ &= 21 + 4 - 24 = 1 \end{aligned}$$

**Question 16** (Easy). Find the eigenvalues of the matrix  $B = \begin{pmatrix} 4 & -1 \\ 2 & 1 \end{pmatrix}$ .

**Solution.** To find the eigenvalues, we solve the characteristic equation  $\det(B - \lambda I) = 0$ , where  $I$  is the identity matrix and  $\lambda$  represents the eigenvalues.

$$B - \lambda I = \begin{pmatrix} 4 & -1 \\ 2 & 1 \end{pmatrix} - \lambda \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 4 - \lambda & -1 \\ 2 & 1 - \lambda \end{pmatrix}$$

Now, we compute the determinant of this matrix:

$$\begin{aligned} \det(B - \lambda I) &= (4 - \lambda)(1 - \lambda) - (-1)(2) \\ &= 4 - 4\lambda - \lambda + \lambda^2 + 2 \\ &= \lambda^2 - 5\lambda + 6 \end{aligned}$$

Set the characteristic polynomial to zero to find the eigenvalues:

$$\lambda^2 - 5\lambda + 6 = 0$$

Factoring the quadratic equation:

$$(\lambda - 2)(\lambda - 3) = 0$$

The eigenvalues are  $\lambda_1 = 2$  and  $\lambda_2 = 3$ .

**Question 17** (Moderate). Consider the symmetric matrix  $A$ :

$$A = \begin{pmatrix} 4 & 2 & 2 \\ 2 & 4 & 2 \\ 2 & 2 & 4 \end{pmatrix}$$

- (a) Given that  $\lambda_1 = 2$  is an eigenvalue of  $A$ , use the properties of trace and determinant to find the other two eigenvalues.
- (b) Find a basis for the eigenspace corresponding to the eigenvalue  $\lambda = 2$ .

**Solution.** (a) **Find other eigenvalues using Trace and Determinant:** Let the eigenvalues be  $\lambda_1, \lambda_2, \lambda_3$ . We are given  $\lambda_1 = 2$ . The trace of a matrix is the sum of its eigenvalues:  $\text{Tr}(A) = \lambda_1 + \lambda_2 + \lambda_3$ . The trace is also the sum of the diagonal elements:  $\text{Tr}(A) = 4 + 4 + 4 = 12$ . So,  $2 + \lambda_2 + \lambda_3 = 12 \implies \lambda_2 + \lambda_3 = 10$ . (Eq. 1)

The determinant of a matrix is the product of its eigenvalues:  $\det(A) = \lambda_1 \lambda_2 \lambda_3$ .  $\det(A) = 4(16 - 4) - 2(8 - 4) + 2(4 - 8) = 4(12) - 2(4) + 2(-4) = 48 - 8 - 8 = 32$ . So,  $2\lambda_2\lambda_3 = 32 \implies \lambda_2\lambda_3 = 16$ . (Eq. 2)

We solve the system from Eq. 1 and Eq. 2. From Eq. 1,  $\lambda_3 = 10 - \lambda_2$ . Substitute this into Eq. 2:  $\lambda_2(10 - \lambda_2) = 16 \implies 10\lambda_2 - \lambda_2^2 = 16 \implies \lambda_2^2 - 10\lambda_2 + 16 = 0$ . Factoring this quadratic:  $(\lambda_2 - 2)(\lambda_2 - 8) = 0$ . This gives  $\lambda_2 = 2$  and  $\lambda_3 = 8$ . The three eigenvalues are  $\{2, 2, 8\}$ .

- (b) **Find basis for the eigenspace of  $\lambda = 2$ :** We need to find the basis for the null space of  $(A - 2I)$ .

$$A - 2I = \begin{pmatrix} 4-2 & 2 & 2 \\ 2 & 4-2 & 2 \\ 2 & 2 & 4-2 \end{pmatrix} = \begin{pmatrix} 2 & 2 & 2 \\ 2 & 2 & 2 \\ 2 & 2 & 2 \end{pmatrix}$$

We solve  $(A - 2I)\mathbf{x} = \mathbf{0}$  by row reducing the matrix:

$$\begin{pmatrix} 2 & 2 & 2 \\ 2 & 2 & 2 \\ 2 & 2 & 2 \end{pmatrix} \xrightarrow{R_1 \rightarrow 1/2 R_1} \begin{pmatrix} 1 & 1 & 1 \\ 2 & 2 & 2 \\ 2 & 2 & 2 \end{pmatrix} \xrightarrow{\substack{R_2 \rightarrow R_2 - 2R_1 \\ R_3 \rightarrow R_3 - 2R_1}} \begin{pmatrix} 1 & 1 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

This gives the single equation  $x_1 + x_2 + x_3 = 0$ . There is one pivot and two free variables. Let  $x_2 = s$  and  $x_3 = t$ . Then  $x_1 = -s - t$ . The general solution (the eigenvectors) is  $\mathbf{x} = \begin{pmatrix} -s-t \\ s \\ t \end{pmatrix} = s \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix} + t \begin{pmatrix} -1 \\ 0 \\ 1 \end{pmatrix}$ . A basis for the eigenspace is the set of vectors spanning this solution space:

$$\mathcal{B}_{\lambda=2} = \left\{ \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} -1 \\ 0 \\ 1 \end{pmatrix} \right\}$$

**Question 18** (Moderate). Find the eigenvalues and corresponding eigenvectors of the matrix:

$$A = \begin{pmatrix} 2 & 1 & 0 \\ 1 & 2 & 1 \\ 0 & 1 & 2 \end{pmatrix}$$

**Solution.** First, find the eigenvalues from  $\det(A - \lambda I) = 0$ .

$$\begin{aligned}\det(A - \lambda I) &= \begin{vmatrix} 2 - \lambda & 1 & 0 \\ 1 & 2 - \lambda & 1 \\ 0 & 1 & 2 - \lambda \end{vmatrix} \\ &= (2 - \lambda) \begin{vmatrix} 2 - \lambda & 1 \\ 1 & 2 - \lambda \end{vmatrix} - 1 \begin{vmatrix} 1 & 1 \\ 0 & 2 - \lambda \end{vmatrix} \\ &= (2 - \lambda)((2 - \lambda)^2 - 1) - 1(2 - \lambda) \\ &= (2 - \lambda)[(2 - \lambda)^2 - 1 - 1] = (2 - \lambda)[\lambda^2 - 4\lambda + 4 - 2] \\ &= (2 - \lambda)(\lambda^2 - 4\lambda + 2) = 0\end{aligned}$$

One eigenvalue is  $\lambda_1 = 2$ . For the others, we use the quadratic formula on  $\lambda^2 - 4\lambda + 2 = 0$ :  
 $\lambda = \frac{-(-4) \pm \sqrt{(-4)^2 - 4(1)(2)}}{2(1)} = \frac{4 \pm \sqrt{16 - 8}}{2} = \frac{4 \pm \sqrt{8}}{2} = \frac{4 \pm 2\sqrt{2}}{2} = 2 \pm \sqrt{2}$ . The other eigenvalues are  $\lambda_2 = 2 + \sqrt{2}$  and  $\lambda_3 = 2 - \sqrt{2}$ .

Now, find the eigenvector for each eigenvalue by solving  $(A - \lambda I)\mathbf{v} = \mathbf{0}$ .

- For  $\lambda_1 = 2$ :  $(A - 2I)\mathbf{v} = \begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \mathbf{0}$ . This gives the system  $x_2 = 0$  and  $x_1 + x_3 = 0 \implies x_1 = -x_3$ . Let  $x_3 = t$ . Then  $x_1 = -t, x_2 = 0$ . An eigenvector is  $\mathbf{v}_1 = \begin{pmatrix} -1 \\ 0 \\ 1 \end{pmatrix}$ .

- For  $\lambda_2 = 2 + \sqrt{2}$ :  $(A - (2 + \sqrt{2})I)\mathbf{v} = \begin{pmatrix} -\sqrt{2} & 1 & 0 \\ 1 & -\sqrt{2} & 1 \\ 0 & 1 & -\sqrt{2} \end{pmatrix} \mathbf{v} = \mathbf{0}$ . From the first row:  $-\sqrt{2}x_1 + x_2 = 0 \implies x_2 = \sqrt{2}x_1$ . From the third row:  $x_2 - \sqrt{2}x_3 = 0 \implies x_2 = \sqrt{2}x_3$ . These together imply  $x_1 = x_3$ . Let  $x_1 = t$ . Then  $x_3 = t$  and  $x_2 = \sqrt{2}t$ . An eigenvector is  $\mathbf{v}_2 = \begin{pmatrix} 1 \\ \sqrt{2} \\ 1 \end{pmatrix}$ .

- For  $\lambda_3 = 2 - \sqrt{2}$ :  $(A - (2 - \sqrt{2})I)\mathbf{v} = \begin{pmatrix} \sqrt{2} & 1 & 0 \\ 1 & \sqrt{2} & 1 \\ 0 & 1 & \sqrt{2} \end{pmatrix} \mathbf{v} = \mathbf{0}$ . From the first row:  $\sqrt{2}x_1 + x_2 = 0 \implies x_2 = -\sqrt{2}x_1$ . From the third row:  $x_2 + \sqrt{2}x_3 = 0 \implies x_2 = -\sqrt{2}x_3$ . These imply  $x_1 = x_3$ . Let  $x_1 = t$ . Then  $x_3 = t$  and  $x_2 = -\sqrt{2}t$ . An eigenvector is  $\mathbf{v}_3 = \begin{pmatrix} 1 \\ -\sqrt{2} \\ 1 \end{pmatrix}$ .

**Question 19** (Moderate). Given the matrix  $A = \begin{pmatrix} 1 & 4 \\ 1 & 1 \end{pmatrix}$ . Find matrices  $P$  and  $D$  such that  $A = PDP^{-1}$ , where  $D$  is a diagonal matrix.

**Solution.** This is an eigendecomposition problem. We must find the eigenvalues and corresponding eigenvectors of  $A$ . **Step 1: Find Eigenvalues** Characteristic equation:

$$\det(A - \lambda I) = 0.$$

$$\det \begin{pmatrix} 1 - \lambda & 4 \\ 1 & 1 - \lambda \end{pmatrix} = (1 - \lambda)^2 - 4 = 0$$

$$(1 - \lambda)^2 = 4 \implies 1 - \lambda = \pm 2$$

This gives two possibilities:  $1 - \lambda = 2 \implies \lambda_1 = -1$ .  $1 - \lambda = -2 \implies \lambda_2 = 3$ . **Step 2: Find Eigenvectors**

- For  $\lambda_1 = -1$ : Solve  $(A - (-1)I)\mathbf{v} = \mathbf{0}$ .  $\begin{pmatrix} 2 & 4 \\ 1 & 2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \mathbf{0}$ . This gives the equation  $x_1 + 2x_2 = 0 \implies x_1 = -2x_2$ . Let  $x_2 = 1$ , then  $x_1 = -2$ . An eigenvector is  $\mathbf{v}_1 = \begin{pmatrix} -2 \\ 1 \end{pmatrix}$ .
- For  $\lambda_2 = 3$ : Solve  $(A - 3I)\mathbf{v} = \mathbf{0}$ .  $\begin{pmatrix} -2 & 4 \\ 1 & -2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \mathbf{0}$ . This gives the equation  $x_1 - 2x_2 = 0 \implies x_1 = 2x_2$ . Let  $x_2 = 1$ , then  $x_1 = 2$ . An eigenvector is  $\mathbf{v}_2 = \begin{pmatrix} 2 \\ 1 \end{pmatrix}$ .

**Step 3: Construct P and D** The diagonal matrix  $D$  contains the eigenvalues on its diagonal, and the matrix  $P$  contains the corresponding eigenvectors as its columns.

$$D = \begin{pmatrix} -1 & 0 \\ 0 & 3 \end{pmatrix}, \quad P = \begin{pmatrix} -2 & 2 \\ 1 & 1 \end{pmatrix}$$

To complete the decomposition  $A = PDP^{-1}$ , we need  $P^{-1}$ . For a  $2 \times 2$  matrix  $\begin{pmatrix} a & b \\ c & d \end{pmatrix}$ , the inverse is  $\frac{1}{ad-bc} \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}$ .

$$\det(P) = (-2)(1) - (2)(1) = -4$$

$$P^{-1} = -\frac{1}{4} \begin{pmatrix} 1 & -2 \\ -1 & -2 \end{pmatrix} = \begin{pmatrix} -1/4 & 1/2 \\ 1/4 & 1/2 \end{pmatrix}$$

$$\text{The full diagonalization is } A = \begin{pmatrix} -2 & 2 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} -1 & 0 \\ 0 & 3 \end{pmatrix} \begin{pmatrix} -1/4 & 1/2 \\ 1/4 & 1/2 \end{pmatrix}.$$

**Question 20 (Tough).** Two  $n \times n$  matrices  $A$  and  $B$  are simultaneously diagonalizable if they share a common basis of eigenvectors ( $A = PD_A P^{-1}$  and  $B = PD_B P^{-1}$ ). This is possible if and only if  $A$  and  $B$  commute ( $AB = BA$ ). You are given:

$$A = \begin{pmatrix} 1 & -1 & 0 \\ -1 & 2 & -1 \\ 0 & -1 & 1 \end{pmatrix}, \quad D_B = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & -1 \end{pmatrix}$$

Assuming  $A$  and  $B$  are simultaneously diagonalizable, find the matrix  $B$ .

**Solution. Step 1: Find the eigenvector matrix P from A:** First, we find the eigenvalues of  $A$ . The characteristic equation  $\det(A - \lambda I) = 0$  is:  $\det(A - \lambda I) = (1 - \lambda)((2 - \lambda)(1 - \lambda) - 1) + 1(-(1 - \lambda)) = (1 - \lambda)[\lambda^2 - 3\lambda + 2 - 1 - 1] = (1 - \lambda)(\lambda^2 - 3\lambda) = \lambda(1 - \lambda)(\lambda - 3) = 0$ . The eigenvalues of  $A$  are  $\lambda_1 = 0, \lambda_2 = 1, \lambda_3 = 3$ .

Since  $A$  is symmetric, its eigenvectors will be orthogonal. We find normalized eigenvectors to form an orthogonal matrix  $P$ .

- For  $\lambda_1 = 0$ :  $\begin{pmatrix} 1 & -1 & 0 \\ -1 & 2 & -1 \\ 0 & -1 & 1 \end{pmatrix} \mathbf{x} = \mathbf{0} \implies \mathbf{v}_1 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$ . Normalized:  $\mathbf{p}_1 = \frac{1}{\sqrt{3}} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$ .
- For  $\lambda_2 = 1$ :  $\begin{pmatrix} 0 & -1 & 0 \\ -1 & 1 & -1 \\ 0 & -1 & 0 \end{pmatrix} \mathbf{x} = \mathbf{0} \implies \mathbf{v}_2 = \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}$ . Normalized:  $\mathbf{p}_2 = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}$ .
- For  $\lambda_3 = 3$ :  $\begin{pmatrix} -2 & -1 & 0 \\ -1 & -1 & -1 \\ 0 & -1 & -2 \end{pmatrix} \mathbf{x} = \mathbf{0} \implies \mathbf{v}_3 = \begin{pmatrix} 1 \\ -2 \\ 1 \end{pmatrix}$ . Normalized:  $\mathbf{p}_3 = \frac{1}{\sqrt{6}} \begin{pmatrix} 1 \\ -2 \\ 1 \end{pmatrix}$ .

The matrix  $P$  has these eigenvectors as columns, ordered according to the eigenvalues of  $A$  (0, 1, 3).

$$P = \begin{pmatrix} 1/\sqrt{3} & 1/\sqrt{2} & 1/\sqrt{6} \\ 1/\sqrt{3} & 0 & -2/\sqrt{6} \\ 1/\sqrt{3} & -1/\sqrt{2} & 1/\sqrt{6} \end{pmatrix}$$

**Step 2: Compute  $B$ :** We use the diagonalization formula  $B = PD_B P^{-1}$ . Since  $P$  is an orthogonal matrix (its columns are orthonormal), we have  $P^{-1} = P^T$ .

$$B = PD_B P^T = \begin{pmatrix} 1/\sqrt{3} & 1/\sqrt{2} & 1/\sqrt{6} \\ 1/\sqrt{3} & 0 & -2/\sqrt{6} \\ 1/\sqrt{3} & -1/\sqrt{2} & 1/\sqrt{6} \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & -1 \end{pmatrix} \begin{pmatrix} 1/\sqrt{3} & 1/\sqrt{3} & 1/\sqrt{3} \\ 1/\sqrt{2} & 0 & -1/\sqrt{2} \\ 1/\sqrt{6} & -2/\sqrt{6} & 1/\sqrt{6} \end{pmatrix}$$

First, compute  $PD_B$ :

$$PD_B = \begin{pmatrix} 1(1/\sqrt{3}) & 2(1/\sqrt{2}) & -1(1/\sqrt{6}) \\ 1(1/\sqrt{3}) & 2(0) & -1(-2/\sqrt{6}) \\ 1(1/\sqrt{3}) & 2(-1/\sqrt{2}) & -1(1/\sqrt{6}) \end{pmatrix} = \begin{pmatrix} 1/\sqrt{3} & 2/\sqrt{2} & -1/\sqrt{6} \\ 1/\sqrt{3} & 0 & 2/\sqrt{6} \\ 1/\sqrt{3} & -2/\sqrt{2} & -1/\sqrt{6} \end{pmatrix}$$

Now, multiply by  $P^T$ :  $B = (PD_B)P^T$ .

$$B_{11} = \frac{1}{\sqrt{3}} \frac{1}{\sqrt{3}} + \frac{2}{\sqrt{2}} \frac{1}{\sqrt{2}} - \frac{1}{\sqrt{6}} \frac{1}{\sqrt{6}} = \frac{1}{3} + \frac{2}{2} - \frac{1}{6} = \frac{2+6-1}{6} = \frac{7}{6}$$

$$B_{12} = \frac{1}{\sqrt{3}} \frac{1}{\sqrt{3}} + \frac{2}{\sqrt{2}}(0) - \frac{1}{\sqrt{6}} \frac{-2}{\sqrt{6}} = \frac{1}{3} + \frac{2}{6} = \frac{4}{6}$$

$$B_{13} = \frac{1}{\sqrt{3}} \frac{1}{\sqrt{3}} + \frac{2}{\sqrt{2}} \frac{-1}{\sqrt{2}} - \frac{1}{\sqrt{6}} \frac{1}{\sqrt{6}} = \frac{1}{3} - \frac{2}{2} - \frac{1}{6} = \frac{2-6-1}{6} = -\frac{5}{6}$$

By symmetry ( $B$  will be symmetric as  $D_B$  is diagonal and  $P$  is orthogonal),  $B_{21} = B_{12}$ ,  $B_{31} = B_{13}$ , etc.

$$B_{22} = \frac{1}{3} + 0 + \frac{2}{\sqrt{6}} \frac{-2}{\sqrt{6}} = \frac{1}{3} - \frac{4}{6} = \frac{2-4}{6} = -\frac{2}{6}$$

$$B_{23} = \frac{1}{3} + 0 + \frac{2}{\sqrt{6}} \frac{1}{\sqrt{6}} = \frac{1}{3} + \frac{2}{6} = \frac{4}{6}$$

$$B_{33} = \frac{1}{3} + \frac{-2}{\sqrt{2}} \frac{-1}{\sqrt{2}} - \frac{1}{\sqrt{6}} \frac{1}{\sqrt{6}} = \frac{1}{3} + \frac{2}{2} - \frac{1}{6} = \frac{2+6-1}{6} = \frac{7}{6}$$

$$B = \frac{1}{6} \begin{pmatrix} 7 & 4 & -5 \\ 4 & -2 & 4 \\ -5 & 4 & 7 \end{pmatrix}$$

**Question 21** (Tough). Find the Singular Value Decomposition (SVD) of the matrix  $A = \begin{pmatrix} 1 & 1 \\ 0 & 1 \\ 1 & 0 \end{pmatrix}$ . Then, find the best rank-1 approximation of  $A$ .

**Solution.** The SVD of  $A$  is  $A = U\Sigma V^T$ .

- Find  $V$  and  $\Sigma$  from  $A^T A$ :**  $A^T A = \begin{pmatrix} 1 & 0 & 1 \\ 1 & 1 & 0 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 0 & 1 \\ 1 & 0 \end{pmatrix} = \begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix}$ . Find eigenvalues of  $A^T A$ :  $\det(A^T A - \lambda I) = (2-\lambda)^2 - 1 = 0 \implies 2-\lambda = \pm 1$ . The eigenvalues are  $\lambda_1 = 3, \lambda_2 = 1$ . The singular values are their square roots:  $\sigma_1 = \sqrt{3}, \sigma_2 = 1$ . The matrix  $\Sigma$  is an  $m \times n$  matrix (same size as  $A$ ) with singular values on the diagonal:  $\Sigma = \begin{pmatrix} \sqrt{3} & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix}$ .

Find normalized eigenvectors of  $A^T A$  for  $V$ : For  $\lambda_1 = 3$ :  $(A^T A - 3I)\mathbf{v} = \begin{pmatrix} -1 & 1 \\ 1 & -1 \end{pmatrix} \mathbf{v} = \mathbf{0} \implies -x_1 + x_2 = 0$ . Eigenvector is  $\begin{pmatrix} 1 \\ 1 \end{pmatrix}$ . Normalized:  $\mathbf{v}_1 = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ . For  $\lambda_2 = 1$ :  $(A^T A - I)\mathbf{v} = \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix} \mathbf{v} = \mathbf{0} \implies x_1 + x_2 = 0$ . Eigenvector is  $\begin{pmatrix} 1 \\ -1 \end{pmatrix}$ . Normalized:  $\mathbf{v}_2 = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ -1 \end{pmatrix}$ . So,  $V = \begin{pmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{pmatrix}$ .

- Find  $U$ :** The columns of  $U$  are found by  $\mathbf{u}_i = \frac{1}{\sigma_i} A \mathbf{v}_i$ .  $\mathbf{u}_1 = \frac{1}{\sqrt{3}} \begin{pmatrix} 1 & 1 \\ 0 & 1 \\ 1 & 0 \end{pmatrix} \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \frac{1}{\sqrt{6}} \begin{pmatrix} 2 \\ 1 \\ 1 \end{pmatrix}$ .  $\mathbf{u}_2 = \frac{1}{1} \begin{pmatrix} 1 & 1 \\ 0 & 1 \\ 1 & 0 \end{pmatrix} \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ -1 \end{pmatrix} = \frac{1}{\sqrt{2}} \begin{pmatrix} 0 \\ -1 \\ 1 \end{pmatrix}$ . Since  $U$  must be a  $3 \times 3$  orthogonal matrix, we need a third vector  $\mathbf{u}_3$  orthogonal to  $\mathbf{u}_1, \mathbf{u}_2$ . We can find it with the cross product of the non-normalized vectors  $(2, 1, 1)$  and  $(0, -1, 1)$ .  $(2, 1, 1) \times (0, -1, 1) = \det \begin{pmatrix} \mathbf{i} & \mathbf{j} & \mathbf{k} \\ 2 & 1 & 1 \\ 0 & -1 & 1 \end{pmatrix} = \mathbf{i}(1 - (-1)) - \mathbf{j}(2 - 0) + \mathbf{k}(-2 - 0) = (2, -2, -2)$ .

We can use the simpler parallel vector  $(1, -1, -1)$ . Normalized,  $\mathbf{u}_3 = \frac{1}{\sqrt{3}} \begin{pmatrix} 1 \\ -1 \\ -1 \end{pmatrix}$ .



$$\text{So, } U = \begin{pmatrix} 2/\sqrt{6} & 0 & 1/\sqrt{3} \\ 1/\sqrt{6} & -1/\sqrt{2} & -1/\sqrt{3} \\ 1/\sqrt{6} & 1/\sqrt{2} & -1/\sqrt{3} \end{pmatrix}.$$

3. **Rank-1 Approximation:** The best rank- $k$  approximation is  $A_k = \sum_{i=1}^k \sigma_i \mathbf{u}_i \mathbf{v}_i^T$ .  
For rank-1, this is  $A_1 = \sigma_1 \mathbf{u}_1 \mathbf{v}_1^T$ .

$$A_1 = \sqrt{3} \left( \frac{1}{\sqrt{6}} \begin{pmatrix} 2 \\ 1 \\ 1 \end{pmatrix} \right) \left( \frac{1}{\sqrt{2}} (1 \ 1) \right)$$

$$A_1 = \frac{\sqrt{3}}{\sqrt{12}} \begin{pmatrix} 2 \\ 1 \\ 1 \end{pmatrix} (1 \ 1) = \frac{\sqrt{3}}{2\sqrt{3}} \begin{pmatrix} 2 & 2 \\ 1 & 1 \\ 1 & 1 \end{pmatrix} = \frac{1}{2} \begin{pmatrix} 2 & 2 \\ 1 & 1 \\ 1 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 1 \\ 1/2 & 1/2 \\ 1/2 & 1/2 \end{pmatrix}$$

## 4 Differentiation and Gradients

**Question 22** (Easy). Find the derivative of the univariate function  $f(x) = x^3 \sin(x)$ .

**Solution.** We use the product rule for differentiation, which states that  $(u(x)v(x))' = u'(x)v(x) + u(x)v'(x)$ . Let  $u(x) = x^3$  and  $v(x) = \sin(x)$ . Then their derivatives are  $u'(x) = 3x^2$  and  $v'(x) = \cos(x)$ . Applying the product rule:

$$f'(x) = (3x^2)(\sin(x)) + (x^3)(\cos(x)) = 3x^2 \sin(x) + x^3 \cos(x)$$

**Question 23** (Easy). Given the function  $f(x, y) = 3x^2y^3 + e^{2x} - \ln(y)$ , find the partial derivatives  $\frac{\partial f}{\partial x}$  and  $\frac{\partial f}{\partial y}$ .

**Solution.** To find  $\frac{\partial f}{\partial x}$ , we treat the variable  $y$  as a constant.

$$\begin{aligned}\frac{\partial f}{\partial x} &= \frac{\partial}{\partial x}(3x^2y^3) + \frac{\partial}{\partial x}(e^{2x}) - \frac{\partial}{\partial x}(\ln(y)) \\ &= (3y^3)(2x) + e^{2x}(2) - 0 = 6xy^3 + 2e^{2x}\end{aligned}$$

To find  $\frac{\partial f}{\partial y}$ , we treat the variable  $x$  as a constant.

$$\begin{aligned}\frac{\partial f}{\partial y} &= \frac{\partial}{\partial y}(3x^2y^3) + \frac{\partial}{\partial y}(e^{2x}) - \frac{\partial}{\partial y}(\ln(y)) \\ &= (3x^2)(3y^2) + 0 - \frac{1}{y} = 9x^2y^2 - \frac{1}{y}\end{aligned}$$

**Question 24** (Moderate). Find the gradient of the function  $f(x, y, z) = x^2y \cos(z)$  at the point  $P(2, 1, \pi)$ .

**Solution.** The gradient of  $f$  is the vector of its partial derivatives:  $\nabla f = \left\langle \frac{\partial f}{\partial x}, \frac{\partial f}{\partial y}, \frac{\partial f}{\partial z} \right\rangle$ .

$$\frac{\partial f}{\partial x} = \frac{\partial}{\partial x}(x^2y \cos(z)) = 2xy \cos(z)$$

$$\frac{\partial f}{\partial y} = \frac{\partial}{\partial y}(x^2y \cos(z)) = x^2 \cos(z)$$

$$\frac{\partial f}{\partial z} = \frac{\partial}{\partial z}(x^2y \cos(z)) = -x^2y \sin(z)$$

The gradient vector is  $\nabla f(x, y, z) = \langle 2xy \cos(z), x^2 \cos(z), -x^2y \sin(z) \rangle$ . Now, evaluate the gradient at the point  $P(2, 1, \pi)$ . We use  $\cos(\pi) = -1$  and  $\sin(\pi) = 0$ .

$$\left. \frac{\partial f}{\partial x} \right|_{(2,1,\pi)} = 2(2)(1) \cos(\pi) = 4(-1) = -4$$

$$\left. \frac{\partial f}{\partial y} \right|_{(2,1,\pi)} = (2)^2 \cos(\pi) = 4(-1) = -4$$

$$\left. \frac{\partial f}{\partial z} \right|_{(2,1,\pi)} = -(2)^2(1) \sin(\pi) = -4(0) = 0$$

The gradient at  $P$  is  $\nabla f(2, 1, \pi) = \langle -4, -4, 0 \rangle$ .

**Question 25** (Moderate). Let  $\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$  and let  $\mathbf{f} : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  be a vector-valued function defined by:

$$\mathbf{f}(\mathbf{x}) = \begin{pmatrix} f_1(x_1, x_2) \\ f_2(x_1, x_2) \end{pmatrix} = \begin{pmatrix} x_1^2 + x_2^2 \\ x_1 x_2 \end{pmatrix}$$

Find the Jacobian matrix of  $\mathbf{f}$ .

**Solution.** The Jacobian matrix  $J$  of  $\mathbf{f}$  is an  $m \times n$  matrix (here  $2 \times 2$ ) where the entry  $J_{ij}$  is  $\frac{\partial f_i}{\partial x_j}$ .

$$J = \begin{pmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} \end{pmatrix}$$

We calculate the four required partial derivatives:

$$\frac{\partial f_1}{\partial x_1} = \frac{\partial}{\partial x_1}(x_1^2 + x_2^2) = 2x_1$$

$$\frac{\partial f_1}{\partial x_2} = \frac{\partial}{\partial x_2}(x_1^2 + x_2^2) = 2x_2$$

$$\frac{\partial f_2}{\partial x_1} = \frac{\partial}{\partial x_1}(x_1 x_2) = x_2$$

$$\frac{\partial f_2}{\partial x_2} = \frac{\partial}{\partial x_2}(x_1 x_2) = x_1$$

Assembling these into the Jacobian matrix gives:

$$J = \begin{pmatrix} 2x_1 & 2x_2 \\ x_2 & x_1 \end{pmatrix}$$

**Question 26** (Moderate). In machine learning, L2 regularization is often added to a loss function to prevent overfitting. The loss function for Ridge Regression is given by  $f(\mathbf{w}) = \|\mathbf{y} - X\mathbf{w}\|_2^2 + \lambda\|\mathbf{w}\|_2^2$ , where  $X$  is the  $m \times n$  design matrix,  $\mathbf{y}$  is the  $m \times 1$  target vector,  $\mathbf{w}$  is the  $n \times 1$  vector of weights, and  $\lambda$  is the regularization parameter. Find the gradient of this function with respect to  $\mathbf{w}$ .

**Solution.** Let  $f(\mathbf{w}) = \|\mathbf{y} - X\mathbf{w}\|_2^2 + \lambda\|\mathbf{w}\|_2^2$ . We can rewrite the squared Euclidean norms using transposes:  $f(\mathbf{w}) = (\mathbf{y} - X\mathbf{w})^T(\mathbf{y} - X\mathbf{w}) + \lambda\mathbf{w}^T\mathbf{w}$ .

**Step 1: Expand the expression.** Using properties of transposes, the first term expands to:

$$(\mathbf{y}^T - \mathbf{w}^T X^T)(\mathbf{y} - X\mathbf{w}) = \mathbf{y}^T \mathbf{y} - \mathbf{y}^T X\mathbf{w} - \mathbf{w}^T X^T \mathbf{y} + \mathbf{w}^T X^T X \mathbf{w}$$

Since  $\mathbf{y}^T X\mathbf{w}$  is a scalar, it equals its transpose:  $(\mathbf{y}^T X\mathbf{w})^T = \mathbf{w}^T X^T \mathbf{y}$ . We can combine the middle terms.

$$f(\mathbf{w}) = \mathbf{y}^T \mathbf{y} - 2\mathbf{w}^T X^T \mathbf{y} + \mathbf{w}^T (X^T X) \mathbf{w} + \lambda \mathbf{w}^T \mathbf{w}$$

**Step 2: Differentiate with respect to  $\mathbf{w}$ .** We use two standard results from matrix calculus:  $\nabla_{\mathbf{w}}(\mathbf{a}^T \mathbf{w}) = \mathbf{a}$  and  $\nabla_{\mathbf{w}}(\mathbf{w}^T A \mathbf{w}) = (A + A^T)\mathbf{w}$ .

- $\nabla_{\mathbf{w}}(\mathbf{y}^T \mathbf{y}) = \mathbf{0}$  (no dependence on  $\mathbf{w}$ ).

- $\nabla_{\mathbf{w}}(-2\mathbf{w}^T X^T \mathbf{y}) = -2(X^T \mathbf{y})$ .
- $\nabla_{\mathbf{w}}(\mathbf{w}^T (X^T X) \mathbf{w})$ . The matrix  $A = X^T X$  is symmetric, so the derivative is  $2(X^T X) \mathbf{w}$ .
- $\nabla_{\mathbf{w}}(\lambda \mathbf{w}^T \mathbf{w}) = \nabla_{\mathbf{w}}(\lambda \mathbf{w}^T I \mathbf{w})$ . The matrix is  $\lambda I$ , which is symmetric. So the derivative is  $2(\lambda I) \mathbf{w} = 2\lambda \mathbf{w}$ .

**Step 3: Combine the results.**

$$\nabla_{\mathbf{w}} f(\mathbf{w}) = \mathbf{0} - 2X^T \mathbf{y} + 2X^T X \mathbf{w} + 2\lambda \mathbf{w}$$

$$\nabla_{\mathbf{w}} f(\mathbf{w}) = 2X^T X \mathbf{w} - 2X^T \mathbf{y} + 2\lambda \mathbf{w}$$

Factoring out common terms gives the final gradient:

$$\nabla_{\mathbf{w}} f(\mathbf{w}) = 2(X^T (X \mathbf{w} - \mathbf{y}) + \lambda \mathbf{w})$$

**Question 27 (Tough).** Let  $f(x, y, z) = x^2 e^y + y \ln(z)$ . Find the directional derivative of  $f$  at the point  $P(1, 0, 2)$  in the direction of the vector  $\mathbf{v} = \langle 2, -1, 2 \rangle$ . What does this value represent?

**Solution.** The directional derivative of  $f$  in the direction of a unit vector  $\mathbf{u}$ , denoted  $D_{\mathbf{u}} f$ , is calculated as the dot product of the gradient of  $f$  and the unit vector  $\mathbf{u}$ .

$$D_{\mathbf{u}} f = \nabla f \cdot \mathbf{u}$$

**Step 1: Find the gradient of  $f$ .** The gradient is the vector of partial derivatives:

$$\nabla f(x, y, z) = \left\langle \frac{\partial f}{\partial x}, \frac{\partial f}{\partial y}, \frac{\partial f}{\partial z} \right\rangle$$

$$\frac{\partial f}{\partial x} = 2xe^y$$

$$\frac{\partial f}{\partial y} = x^2 e^y + \ln(z)$$

$$\frac{\partial f}{\partial z} = \frac{y}{z}$$

So,  $\nabla f = \langle 2xe^y, x^2 e^y + \ln(z), \frac{y}{z} \rangle$ .

**Step 2: Evaluate the gradient at the point  $P(1, 0, 2)$ .**

$$\nabla f(1, 0, 2) = \langle 2(1)e^0, (1)^2 e^0 + \ln(2), \frac{0}{2} \rangle = \langle 2, 1 + \ln(2), 0 \rangle$$

**Step 3: Find the unit vector  $\mathbf{u}$  in the direction of  $\mathbf{v}$ .** First find the magnitude of  $\mathbf{v}$ :

$$\|\mathbf{v}\| = \sqrt{2^2 + (-1)^2 + 2^2} = \sqrt{4 + 1 + 4} = \sqrt{9} = 3$$

The unit vector  $\mathbf{u}$  is  $\mathbf{v}$  divided by its magnitude:

$$\mathbf{u} = \frac{\mathbf{v}}{\|\mathbf{v}\|} = \frac{1}{3} \langle 2, -1, 2 \rangle = \left\langle \frac{2}{3}, -\frac{1}{3}, \frac{2}{3} \right\rangle$$

**Step 4:** Calculate the dot product  $\nabla f(P) \cdot \mathbf{u}$ .

$$\begin{aligned} D_{\mathbf{u}}f(1, 0, 2) &= \langle 2, 1 + \ln(2), 0 \rangle \cdot \left\langle \frac{2}{3}, -\frac{1}{3}, \frac{2}{3} \right\rangle \\ &= (2) \left( \frac{2}{3} \right) + (1 + \ln(2)) \left( -\frac{1}{3} \right) + (0) \left( \frac{2}{3} \right) \\ &= \frac{4}{3} - \frac{1}{3} - \frac{\ln(2)}{3} = \frac{3 - \ln(2)}{3} = 1 - \frac{\ln(2)}{3} \end{aligned}$$

**Interpretation:** The value  $1 - \frac{\ln(2)}{3}$  represents the instantaneous rate of change of the function  $f$  at the point  $P(1, 0, 2)$  as we move from  $P$  in the direction of the vector  $\mathbf{v}$ . Since  $\ln(2) \approx 0.693$ , the value is positive, which means the function's value is increasing at that point along that specific direction.

## 5 Higher-Order Derivatives and Taylor's Series

**Question 28** (Easy). Find the first and second derivatives of the function  $f(x) = e^{x^2}$ .

**Solution. First derivative:** We use the chain rule,  $\frac{d}{dx}e^{u(x)} = e^{u(x)}u'(x)$ . Let  $u(x) = x^2$ , so  $u'(x) = 2x$ .

$$f'(x) = \frac{d}{dx}(e^{x^2}) = e^{x^2} \cdot \frac{d}{dx}(x^2) = 2xe^{x^2}$$

**Second derivative:** We differentiate  $f'(x)$  using the product rule,  $(uv)' = u'v + uv'$ . Let  $u(x) = 2x$  and  $v(x) = e^{x^2}$ . Then  $u'(x) = 2$  and  $v'(x) = 2xe^{x^2}$  (from our first derivative calculation).

$$\begin{aligned} f''(x) &= \frac{d}{dx}(2xe^{x^2}) = (2)(e^{x^2}) + (2x)(2xe^{x^2}) \\ f''(x) &= 2e^{x^2} + 4x^2e^{x^2} = 2e^{x^2}(1 + 2x^2) \end{aligned}$$

**Question 29** (Easy). For the function  $f(x, y) = x^3 - 3xy^2 + y^4$ , compute all second-order partial derivatives:  $f_{xx}, f_{yy}, f_{xy}, f_{yx}$ .

**Solution.** First, find the first-order partial derivatives:

$$\begin{aligned} f_x &= \frac{\partial f}{\partial x} = \frac{\partial}{\partial x}(x^3 - 3xy^2 + y^4) = 3x^2 - 3y^2 \\ f_y &= \frac{\partial f}{\partial y} = \frac{\partial}{\partial y}(x^3 - 3xy^2 + y^4) = -6xy + 4y^3 \end{aligned}$$

Now, find the second-order partial derivatives by differentiating the first-order derivatives:

$$\begin{aligned} f_{xx} &= \frac{\partial}{\partial x}(f_x) = \frac{\partial}{\partial x}(3x^2 - 3y^2) = 6x \\ f_{yy} &= \frac{\partial}{\partial y}(f_y) = \frac{\partial}{\partial y}(-6xy + 4y^3) = -6x + 12y^2 \\ f_{xy} &= \frac{\partial}{\partial y}(f_x) = \frac{\partial}{\partial y}(3x^2 - 3y^2) = -6y \\ f_{yx} &= \frac{\partial}{\partial x}(f_y) = \frac{\partial}{\partial x}(-6xy + 4y^3) = -6y \end{aligned}$$

As expected from Clairaut's theorem on the equality of mixed partials (since these derivatives are continuous), we find that  $f_{xy} = f_{yx}$ .

**Question 30** (Moderate). Find the Hessian matrix of the function  $f(x, y) = x \sin(y) + y \cos(x)$  at the point  $(\frac{\pi}{2}, 0)$ .

**Solution.** The Hessian matrix  $H$  is the matrix of second-order partial derivatives:

$$H = \begin{pmatrix} f_{xx} & f_{xy} \\ f_{yx} & f_{yy} \end{pmatrix}$$

First, find the first-order partial derivatives:

$$f_x = \frac{\partial}{\partial x}(x \sin(y) + y \cos(x)) = \sin(y) - y \sin(x)$$

$$f_y = \frac{\partial}{\partial y}(x \sin(y) + y \cos(x)) = x \cos(y) + \cos(x)$$

Now, find the second-order partial derivatives:

$$f_{xx} = \frac{\partial}{\partial x}(\sin(y) - y \sin(x)) = -y \cos(x)$$

$$f_{yy} = \frac{\partial}{\partial y}(x \cos(y) + \cos(x)) = -x \sin(y)$$

$$f_{xy} = \frac{\partial}{\partial y}(\sin(y) - y \sin(x)) = \cos(y) - \sin(x)$$

The Hessian matrix as a function of  $(x, y)$  is:

$$H(x, y) = \begin{pmatrix} -y \cos(x) & \cos(y) - \sin(x) \\ \cos(y) - \sin(x) & -x \sin(y) \end{pmatrix}$$

Now evaluate at the point  $(\frac{\pi}{2}, 0)$ . We use the values:  $\cos(\frac{\pi}{2}) = 0$ ,  $\sin(\frac{\pi}{2}) = 1$ ,  $\cos(0) = 1$ ,  $\sin(0) = 0$ .

$$H\left(\frac{\pi}{2}, 0\right) = \begin{pmatrix} -(0) \cos(\frac{\pi}{2}) & \cos(0) - \sin(\frac{\pi}{2}) \\ \cos(0) - \sin(\frac{\pi}{2}) & -\frac{\pi}{2} \sin(0) \end{pmatrix}$$

$$H\left(\frac{\pi}{2}, 0\right) = \begin{pmatrix} 0 & 1 - 1 \\ 1 - 1 & -\frac{\pi}{2} \cdot 0 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$$

**Question 31** (Moderate). Find the linearization (first-order Taylor expansion) of the function  $f(x, y) = \sqrt{x^2 + y^2}$  around the point  $(3, 4)$ . Use it to approximate  $f(3.01, 3.98)$ .

**Solution.** The linearization  $L(x, y)$  of a function  $f(x, y)$  at a point  $(a, b)$  is the tangent plane to the function at that point. The formula is:

$$L(x, y) = f(a, b) + f_x(a, b)(x - a) + f_y(a, b)(y - b)$$

Here, our function is  $f(x, y) = (x^2 + y^2)^{1/2}$  and the point is  $(a, b) = (3, 4)$ . **Step 1: Evaluate the function at the point.**

$$f(3, 4) = \sqrt{3^2 + 4^2} = \sqrt{9 + 16} = \sqrt{25} = 5$$

**Step 2: Find the partial derivatives.**

$$f_x = \frac{\partial}{\partial x}(x^2 + y^2)^{1/2} = \frac{1}{2}(x^2 + y^2)^{-1/2}(2x) = \frac{x}{\sqrt{x^2 + y^2}}$$

$$f_y = \frac{\partial}{\partial y}(x^2 + y^2)^{1/2} = \frac{1}{2}(x^2 + y^2)^{-1/2}(2y) = \frac{y}{\sqrt{x^2 + y^2}}$$

**Step 3: Evaluate the partial derivatives at  $(3, 4)$ .**

$$f_x(3, 4) = \frac{3}{\sqrt{3^2 + 4^2}} = \frac{3}{5}$$

$$f_y(3, 4) = \frac{4}{\sqrt{3^2 + 4^2}} = \frac{4}{5}$$

**Step 4: Construct the linearization formula.**

$$L(x, y) = 5 + \frac{3}{5}(x - 3) + \frac{4}{5}(y - 4)$$

This is the first-order Taylor expansion. To approximate  $f(3.01, 3.98)$ , we evaluate  $L(3.01, 3.98)$ . Here,  $x - a = 3.01 - 3 = 0.01$  and  $y - b = 3.98 - 4 = -0.02$ .

$$\begin{aligned} f(3.01, 3.98) &\approx L(3.01, 3.98) = 5 + \frac{3}{5}(0.01) + \frac{4}{5}(-0.02) \\ &= 5 + \frac{0.03}{5} - \frac{0.08}{5} = 5 - \frac{0.05}{5} = 5 - 0.01 = 4.99 \end{aligned}$$

**Question 32** (Tough). Find the second-order Taylor series expansion of the function  $f(x, y) = e^{x-y}$  around the point  $(a, b) = (1, 1)$ .

**Solution.** The second-order Taylor expansion of  $f(x, y)$  around  $(a, b)$  is given by:

$$\begin{aligned} T_2(x, y) &= f(a, b) + f_x(a, b)(x - a) + f_y(a, b)(y - b) \\ &\quad + \frac{1}{2!} (f_{xx}(a, b)(x - a)^2 + 2f_{xy}(a, b)(x - a)(y - b) + f_{yy}(a, b)(y - b)^2) \end{aligned}$$

Here,  $(a, b) = (1, 1)$ , so  $x - a = x - 1$  and  $y - b = y - 1$ . **Step 1: Evaluate the function and all required derivatives at  $(1, 1)$ .**

- $f(x, y) = e^{x-y} \implies f(1, 1) = e^{1-1} = e^0 = 1.$
- $f_x = e^{x-y} \cdot (1) = e^{x-y} \implies f_x(1, 1) = e^0 = 1.$
- $f_y = e^{x-y} \cdot (-1) = -e^{x-y} \implies f_y(1, 1) = -e^0 = -1.$
- $f_{xx} = \frac{\partial}{\partial x}(e^{x-y}) = e^{x-y} \implies f_{xx}(1, 1) = 1.$
- $f_{yy} = \frac{\partial}{\partial y}(-e^{x-y}) = -e^{x-y} \cdot (-1) = e^{x-y} \implies f_{yy}(1, 1) = 1.$
- $f_{xy} = \frac{\partial}{\partial y}(e^{x-y}) = -e^{x-y} \implies f_{xy}(1, 1) = -1.$

**Step 2: Substitute these values into the Taylor formula.**

$$\begin{aligned} T_2(x, y) &= 1 + (1)(x - 1) + (-1)(y - 1) \\ &\quad + \frac{1}{2} ((1)(x - 1)^2 + 2(-1)(x - 1)(y - 1) + (1)(y - 1)^2) \end{aligned}$$

$$T_2(x, y) = 1 + (x - 1) - (y - 1) + \frac{1}{2} ((x - 1)^2 - 2(x - 1)(y - 1) + (y - 1)^2)$$

This expression is the final answer. It can be simplified by noticing the quadratic part is a perfect square:

$$T_2(x, y) = 1 + (x - y) + \frac{1}{2} ((x - 1) - (y - 1))^2 = 1 + (x - y) + \frac{1}{2}(x - y)^2$$

This matches the standard Maclaurin series for  $e^u \approx 1 + u + u^2/2$  where  $u = x - y$ .

**Question 33** (Tough). Consider the function  $f(x, y) = \cos(x) - y^2 - xy$ .



- (a) Find the second-order Taylor expansion of  $f(x, y)$  around the point  $(0, 0)$ .
- (b) Use the Hessian matrix from the Taylor expansion to classify the critical point at  $(0, 0)$ .

**Solution.** (a) **Second-order Taylor expansion around  $(0, 0)$ :** The formula is:  
 $T_2(x, y) = f(0, 0) + f_x(0, 0)x + f_y(0, 0)y + \frac{1}{2}(f_{xx}(0, 0)x^2 + 2f_{xy}(0, 0)xy + f_{yy}(0, 0)y^2)$ .

**Step 1: Evaluate all derivatives at  $(0, 0)$ :**

- $f(x, y) = \cos(x) - y^2 - xy \implies f(0, 0) = \cos(0) - 0 - 0 = 1$ .
- $f_x = -\sin(x) - y \implies f_x(0, 0) = -\sin(0) - 0 = 0$ .
- $f_y = -2y - x \implies f_y(0, 0) = 0 - 0 = 0$ .
- $f_{xx} = -\cos(x) \implies f_{xx}(0, 0) = -\cos(0) = -1$ .
- $f_{yy} = -2 \implies f_{yy}(0, 0) = -2$ .
- $f_{xy} = -1 \implies f_{xy}(0, 0) = -1$ .

Since  $f_x(0, 0) = 0$  and  $f_y(0, 0) = 0$ , the point  $(0, 0)$  is indeed a critical point.

**Step 2: Assemble the Taylor expansion:**

$$T_2(x, y) = 1 + (0)x + (0)y + \frac{1}{2}((-1)x^2 + 2(-1)xy + (-2)y^2)$$

$$T_2(x, y) = 1 - \frac{1}{2}x^2 - xy - y^2$$

- (b) **Classify the critical point using the Hessian:** The Hessian matrix at a point  $(x, y)$  is  $H(x, y) = \begin{pmatrix} f_{xx} & f_{xy} \\ f_{yx} & f_{yy} \end{pmatrix}$ . At the point  $(0, 0)$ , we have:

$$H(0, 0) = \begin{pmatrix} f_{xx}(0, 0) & f_{xy}(0, 0) \\ f_{yx}(0, 0) & f_{yy}(0, 0) \end{pmatrix} = \begin{pmatrix} -1 & -1 \\ -1 & -2 \end{pmatrix}$$

To classify the critical point, we use the Second Derivative Test. We compute the determinant of the Hessian,  $D$ .

$$D = \det(H) = (-1)(-2) - (-1)(-1) = 2 - 1 = 1$$

The test is as follows:

- If  $D > 0$  and  $f_{xx} > 0$ , it is a local minimum.
- If  $D > 0$  and  $f_{xx} < 0$ , it is a local maximum.
- If  $D < 0$ , it is a saddle point.
- If  $D = 0$ , the test is inconclusive.

In our case,  $D = 1 > 0$  and  $f_{xx}(0, 0) = -1 < 0$ . Therefore, the critical point at  $(0, 0)$  is a **local maximum**.

## 6 Unconstrained Optimization

**Question 34** (Easy). Find the critical points of the function  $f(x) = x^3 - 6x^2 + 9x + 2$ .

**Solution.** To find the critical points of a single-variable function, we find the points where the first derivative is zero or undefined. First, find the derivative of  $f(x)$ :

$$f'(x) = \frac{d}{dx}(x^3 - 6x^2 + 9x + 2) = 3x^2 - 12x + 9$$

The derivative is a polynomial, so it is defined for all real numbers  $x$ . We find the critical points by setting the derivative to zero:

$$3x^2 - 12x + 9 = 0$$

Divide the entire equation by 3 to simplify:

$$x^2 - 4x + 3 = 0$$

Factor the quadratic equation:

$$(x - 1)(x - 3) = 0$$

This gives two solutions. The critical points are  $x = 1$  and  $x = 3$ .

**Question 35** (Easy). Find the critical points of the function  $f(x, y) = x^2 + y^2 - 2x - 6y + 14$ .

**Solution.** To find critical points of a multivariable function, we set its gradient to the zero vector,  $\nabla f = \mathbf{0}$ . The gradient is the vector of partial derivatives. First, find the partial derivatives:

$$\frac{\partial f}{\partial x} = \frac{\partial}{\partial x}(x^2 + y^2 - 2x - 6y + 14) = 2x - 2$$

$$\frac{\partial f}{\partial y} = \frac{\partial}{\partial y}(x^2 + y^2 - 2x - 6y + 14) = 2y - 6$$

Now, set both partial derivatives to zero to find the coordinates of the critical point(s):

$$2x - 2 = 0 \implies 2x = 2 \implies x = 1$$

$$2y - 6 = 0 \implies 2y = 6 \implies y = 3$$

The system of equations has only one solution. The only critical point is  $(1, 3)$ . (This function describes a paraboloid, which has a single minimum).

**Question 36** (Moderate). Find all critical points of the function  $f(x, y) = x^3 + y^3 - 3xy$  and classify each as a local maximum, local minimum, or saddle point.

**Solution. Step 1: Find critical points.** Set the gradient  $\nabla f = \langle f_x, f_y \rangle$  to  $\mathbf{0}$ .

$$f_x = 3x^2 - 3y = 0 \implies y = x^2 \quad (1)$$

$$f_y = 3y^2 - 3x = 0 \implies x = y^2 \quad (2)$$

Substitute the first equation into the second:  $x = (x^2)^2 \implies x = x^4 \implies x^4 - x = 0 \implies x(x^3 - 1) = 0$ . This gives two possible values for  $x$ :

- $x = 0$ . Substitute back into eq(1):  $y = 0^2 = 0$ . This gives the critical point  $(0, 0)$ .
- $x^3 - 1 = 0 \implies x = 1$ . Substitute back into eq(1):  $y = 1^2 = 1$ . This gives the critical point  $(1, 1)$ .

**Step 2: Classify points using the Second Derivative Test.** We need the Hessian matrix,  $H(x, y) = \begin{pmatrix} f_{xx} & f_{xy} \\ f_{yx} & f_{yy} \end{pmatrix}$ .

$$f_{xx} = 6x, \quad f_{yy} = 6y, \quad f_{xy} = -3$$

The determinant of the Hessian is  $D(x, y) = f_{xx}f_{yy} - (f_{xy})^2 = (6x)(6y) - (-3)^2 = 36xy - 9$ .

**Case 1: Point  $(0, 0)$**   $D(0, 0) = 36(0)(0) - 9 = -9$ . Since  $D < 0$ , the point  $(0, 0)$  is a **saddle point**.

**Case 2: Point  $(1, 1)$**   $D(1, 1) = 36(1)(1) - 9 = 27$ . Since  $D > 0$ , we check the sign of  $f_{xx}(1, 1)$ .  $f_{xx}(1, 1) = 6(1) = 6$ . Since  $D > 0$  and  $f_{xx} > 0$ , the point  $(1, 1)$  is a **local minimum**.

**Question 37 (Moderate).** A company wants to build a rectangular storage container with an open top. It must have a volume of 10 cubic meters. The material for the base costs \$10 per square meter, and the material for the sides costs \$6 per square meter. Find the dimensions (length  $l$ , width  $w$ , height  $h$ ) that will minimize the cost of the container.

**Solution. Step 1: Set up the objective and constraint equations.** Let the dimensions be length  $l$ , width  $w$ , and height  $h$ . The volume is the constraint equation:  $V = lwh = 10$ . The cost is the objective function we want to minimize. Cost of base = (Area of base)  $\times$  (Cost per area) =  $(lw) \cdot 10$ . Cost of sides = (Area of sides)  $\times$  (Cost per area) =  $(2lh + 2wh) \cdot 6$ . Total cost:  $C(l, w, h) = 10lw + 12lh + 12wh$ .

**Step 2: Reduce to a two-variable function.** From the volume constraint, we can express one variable in terms of the others:  $h = \frac{10}{lw}$ . Substitute this into the cost function:

$$C(l, w) = 10lw + 12l \left( \frac{10}{lw} \right) + 12w \left( \frac{10}{lw} \right) = 10lw + \frac{120}{w} + \frac{120}{l}$$

**Step 3: Find critical points.** Set the partial derivatives of  $C(l, w)$  with respect to  $l$  and  $w$  to zero.

$$C_l = \frac{\partial C}{\partial l} = 10w - \frac{120}{l^2} = 0 \implies 10wl^2 = 120 \implies w = \frac{12}{l^2} \quad (1)$$

$$C_w = \frac{\partial C}{\partial w} = 10l - \frac{120}{w^2} = 0 \implies 10lw^2 = 120 \implies l = \frac{12}{w^2} \quad (2)$$

Substitute the expression for  $w$  from (1) into (2):

$$l = \frac{12}{(12/l^2)^2} = \frac{12}{144/l^4} = \frac{12l^4}{144} = \frac{l^4}{12}$$

Since we are building a container,  $l > 0$ , so we can divide by  $l$ :  $1 = \frac{l^3}{12} \implies l^3 = 12 \implies l = \sqrt[3]{12}$ . Now find  $w$  using equation (1):

$$w = \frac{12}{l^2} = \frac{12}{(\sqrt[3]{12})^2} = \frac{12}{12^{2/3}} = 12^{1-2/3} = 12^{1/3} = \sqrt[3]{12}$$

So the optimal base is a square,  $l = w$ . Finally, find the height  $h$ :

$$h = \frac{10}{lw} = \frac{10}{\sqrt[3]{12} \cdot \sqrt[3]{12}} = \frac{10}{12^{2/3}}$$

The dimensions that minimize cost are  $l = w = \sqrt[3]{12} \approx 2.29$  m and  $h = \frac{10}{12^{2/3}} \approx 1.91$  m. (A check with the second derivative test would confirm this is a minimum).

**Question 38 (Tough).** Consider the function  $f(x, y) = (x^2 + y^2)^2 - k(x^2 - y^2)$ , where  $k$  is a constant parameter.

- Show that the origin  $(0, 0)$  is a critical point for any value of  $k$ .
- Find the Hessian matrix and use the second derivative test to determine how the classification of the critical point at  $(0, 0)$  (local min, max, or saddle) depends on the value of  $k$ .

**Solution.** (a) **Show  $(0, 0)$  is a critical point:** We find the first partial derivatives of  $f(x, y) = x^4 + 2x^2y^2 + y^4 - kx^2 + ky^2$ .  $f_x = 4x^3 + 4xy^2 - 2kx$ .  $f_y = 4x^2y + 4y^3 + 2ky$ . Now evaluate them at the origin  $(0, 0)$ :  $f_x(0, 0) = 4(0)^3 + 4(0)(0)^2 - 2k(0) = 0$ .  $f_y(0, 0) = 4(0)^2(0) + 4(0)^3 + 2k(0) = 0$ . Since  $\nabla f(0, 0) = \langle 0, 0 \rangle$  for any value of  $k$ , the origin is always a critical point.

- Classify the critical point based on  $k$ :** We find the second partial derivatives to construct the Hessian matrix.  $f_{xx} = 12x^2 + 4y^2 - 2k$ .  $f_{yy} = 4x^2 + 12y^2 + 2k$ .  $f_{xy} = 8xy$ . Now evaluate the Hessian at the origin  $(0, 0)$ :  $H(0, 0) = \begin{pmatrix} f_{xx}(0, 0) & f_{xy}(0, 0) \\ f_{yx}(0, 0) & f_{yy}(0, 0) \end{pmatrix} = \begin{pmatrix} -2k & 0 \\ 0 & 2k \end{pmatrix}$ . The determinant of the Hessian at the origin is  $D = (-2k)(2k) - (0)^2 = -4k^2$ .

We analyze the classification based on the values of  $D$  and  $f_{xx}$ :

- Case 1:**  $k \neq 0$ . In this case,  $k^2 > 0$ , so the determinant  $D = -4k^2$  will always be negative. If  $D < 0$ , the critical point is a **saddle point**. This is true for any non-zero value of  $k$ .
- Case 2:**  $k = 0$ . If  $k = 0$ , the determinant  $D = -4(0)^2 = 0$ . When  $D = 0$ , the second derivative test is inconclusive. We must investigate the function itself. If  $k = 0$ , the function becomes  $f(x, y) = (x^2 + y^2)^2$ . At the critical point  $(0, 0)$ ,  $f(0, 0) = 0$ . For any other point  $(x, y) \neq (0, 0)$ , the term  $x^2 + y^2$  is a positive number, and so is  $(x^2 + y^2)^2$ . Thus,  $f(x, y) > f(0, 0)$  for all points in the neighborhood of the origin. This means that for  $k = 0$ , the origin is a **local minimum** (in fact, it is a global minimum).

**Summary:**

- If  $k \neq 0$ , the origin is a **saddle point**.
- If  $k = 0$ , the origin is a **local minimum**.