

## CS 556-B: Mathematical Foundations of Machine Learning

### Homework 1: Linear Algebra (100 points)

Note: Calculators allowed for trigonometric operations & arithmetic operations (i.e., addition, subtraction, multiplication or division of *scalars*). All solutions methods must be full explained.

## Vectors

1. (5 points) Find the magnitude of the vector  $\mathbf{x} = \begin{bmatrix} 2 \\ -3 \\ 2 \\ 4 \\ -4 \end{bmatrix}$

**Solution:** The magnitude of a vector

$$\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_m \end{bmatrix}$$

is calculated using the formula:

$$\|\mathbf{x}\| = \sqrt{x_1^2 + x_2^2 + \dots + x_m^2}$$

Hence,

$$\|\mathbf{x}\| = \sqrt{2^2 + (-3)^2 + 2^2 + 4^2 + (-4)^2} = \sqrt{49} = 7$$

2. (5 points) Consider two vectors  $\begin{bmatrix} 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \end{bmatrix}$  (in vector space  $\mathbb{R}^2$ ), what is their span? Briefly explain your reasoning leveraging the definition of the span of a set of vectors.

**Solution:** The span of a set of vectors in set  $S$  is defined as all possible vectors that are *reachable* using only the vectors in  $S$ . Specifically, with linear combinations of vectors in  $S$ .

In the given set, we have two vectors  $S = \left\{ \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \end{bmatrix} \right\}$  and the space of interest is  $\mathbb{R}^2$ . Since the vectors in the given set are linearly independent (as they are orthogonal - implied by a dot product of zero between the two vectors), it implies that all vectors in  $\mathbb{R}^2$  are reachable using linear combinations of vectors in  $S$  (i.e.,  $S$  forms a basis in  $\mathbb{R}^2$ ). Hence the span is the entire vector space  $\mathbb{R}^2$ .

## Dot Product

3. (10 points) If two vectors  $\mathbf{a}, \mathbf{b}$  have magnitudes 3 and 5 respectively and the angle between them is  $\frac{\pi}{2}$  radians, what is their dot product?

**Solution:** We know that the dot product of two vectors is given by  $\mathbf{u} \cdot \mathbf{v} = \|\mathbf{u}\| \|\mathbf{v}\| \cos(\theta)$

We also know that the cosine of the angle between two perpendicular vectors (i.e.,  $\cos(\frac{\pi}{2})$ ) is zero. Hence it must be the case that the dot product of vectors  $\mathbf{u}$  and  $\mathbf{v}$  is zero. i.e.,  $\mathbf{u} \cdot \mathbf{v} = 0$ .

4. (10 points) Let vector  $\mathbf{u} = \begin{bmatrix} 1 \\ 3 \end{bmatrix}$   $\mathbf{v} = \begin{bmatrix} -2 \\ 7 \end{bmatrix}$ , calculate the dot product of  $\mathbf{u}$  and  $\mathbf{v}$  also calculate the angle between (i.e., not the cosine of the angle but the actual angle in radians or degrees)  $\mathbf{u}$  and  $\mathbf{v}$ .

**Solution:** We know that the dot product of two vectors is given by  $\mathbf{u} \cdot \mathbf{v} = \|\mathbf{u}\| \|\mathbf{v}\| \cos(\theta)$

$$\mathbf{u} \cdot \mathbf{v} = 1 \times (-2) + 3 \times 7 = 19$$

$$\|\mathbf{u}\| = \sqrt{1^2 + 3^2} = \sqrt{10}$$

$$\|\mathbf{v}\| = \sqrt{(-2)^2 + 7^2} = \sqrt{53}$$

$$\theta = \frac{19}{\sqrt{10} \times \sqrt{53}} = 0.6005 \text{ radians}$$

or

$$\text{Answer in Degrees: } \theta = 34.3804^\circ$$

## Linear Independence

5. (15 points) Check if the vectors  $\mathbf{x} = \begin{bmatrix} -1 \\ 0 \\ 2 \end{bmatrix}$ ,  $\mathbf{y} = \begin{bmatrix} 3 \\ -2 \\ 2 \end{bmatrix}$ ,  $\mathbf{z} = \begin{bmatrix} 5 \\ 2 \\ -6 \end{bmatrix}$

are linearly independent. Note: The condition for linear independence is that given a set  $S$  of vectors  $\mathbf{x}, \mathbf{y}, \mathbf{z}$  and coefficients  $a, b, c$ ,  $a\mathbf{x} + b\mathbf{y} + c\mathbf{z} = \mathbf{0}$  if and only if  $a = b = c = 0$ .

**Solution:** We can re-write the condition for linear independence using vectors  $\mathbf{x}, \mathbf{y}, \mathbf{z}$  as follows:

$$a\mathbf{x} + b\mathbf{y} + c\mathbf{z} = \mathbf{0}$$

An alternate method to represent the relationship would be as a multiplication of a matrix by a vector to yield the  $\mathbf{0}$  vector.

$$\begin{bmatrix} -1 & 3 & 5 \\ 0 & -2 & 2 \\ 2 & 2 & -6 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

Essentially, we need to demonstrate that the above matrix vector produce has  $a = b = c = 0$  as its solution. To do this, we can employ Gaussian Elimination.

We first re-write the above matrix vector produce in augmented matrix form.

$$\left[ \begin{array}{ccc|c} -1 & 3 & 5 & 0 \\ 0 & -2 & 2 & 0 \\ 2 & 2 & -6 & 0 \end{array} \right]$$

Our goal in applying Gaussian elimination is to reduce the above matrix vector product to row-echelon form so that a solution can be obtained for  $a, b, c$ .

$$\begin{aligned} \left[ \begin{array}{ccc|c} -1 & 3 & 5 & 0 \\ 0 & -2 & 2 & 0 \\ 2 & 2 & -6 & 0 \end{array} \right] &\xrightarrow{R_3+2R_1} \left[ \begin{array}{ccc|c} -1 & 3 & 5 & 0 \\ 0 & -2 & 2 & 0 \\ 0 & 8 & 4 & 0 \end{array} \right] \xrightarrow{\frac{1}{2}R_2} \left[ \begin{array}{ccc|c} -1 & 3 & 5 & 0 \\ 0 & -1 & 1 & 0 \\ 0 & 8 & 4 & 0 \end{array} \right] \\ &\xrightarrow{R_1+3R_2} \left[ \begin{array}{ccc|c} -1 & 0 & 8 & 0 \\ 0 & -1 & 1 & 0 \\ 0 & 8 & 4 & 0 \end{array} \right] \xrightarrow{R_3+8R_2} \boxed{\left[ \begin{array}{ccc|c} -1 & 0 & 8 & 0 \\ 0 & -1 & 1 & 0 \\ 0 & 0 & 12 & 0 \end{array} \right]} \end{aligned}$$

The final state (matrix highlighted with a black frame) is in *row echelon* form and we can derive that values of  $c = 0$  (from row  $R_3$ ),  $b = 0$  (from row  $R_2$ ) and  $a = 0$  (from row  $R_1$ ).

Hence we can conclude that the vectors  $\mathbf{x}, \mathbf{y}, \mathbf{z}$  are independent.

6. (15 points) Given a subset of vectors  $S = \{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_k\}$  for  $k \in \mathbb{N}$  of a vector space  $V$ , prove that  $S$  is linearly independent iff a linear combination of elements of  $S$  with non-zero coefficients does not yield  $\mathbf{0}$ . **Hint:** To prove *iff* statements, i.e.,  $A \text{ iff } B$  ( $A \iff B$ ), first prove  $A \rightarrow B$ , then prove  $B \leftarrow A$ .

**Solution:** The given condition implies an if and only if condition requiring us to prove both directions of the statement.

**Part 1:** Set  $S$  of vectors is linearly independent  $\rightarrow$  a linear combination of elements of  $S$  with non-zero coefficients does not yield  $\mathbf{0}$ .

We shall prove part 1 by contradiction.

**Condition 1:** Set  $S$  of vectors is linearly independent  $\rightarrow$  a linear combination of elements of  $S$  with non-zero coefficients yields  $\mathbf{0}$

Mathematically, we can write the right hand side of *condition 1* as follows:

$$\lambda_1 \mathbf{x}_1 + \lambda_2 \mathbf{x}_2 + \dots + \lambda_k \mathbf{x}_k = \mathbf{0} \quad (1)$$

Further, according to *condition 1*, let us assume some scalar  $\lambda_i \neq 0$ . Hence Eq. 1 can be re-written as follows:

$$\lambda_1 \mathbf{x}_1 + \lambda_2 \mathbf{x}_2 + \dots + \lambda_{i-1} \mathbf{x}_{i-1} + \lambda_{i+1} \mathbf{x}_{i+1} + \dots + \lambda_k \mathbf{x}_k = -\lambda_i \mathbf{x}_i \quad (2)$$

Dividing both sides of Eq. 2 by  $-\lambda_i$  we have:

$$\frac{\lambda_1}{-\lambda_i} \mathbf{x}_1 + \frac{\lambda_2}{-\lambda_i} \mathbf{x}_2 + \dots + \frac{\lambda_{i-1}}{-\lambda_i} \mathbf{x}_{i-1} + \frac{\lambda_{i+1}}{-\lambda_i} \mathbf{x}_{i+1} + \dots + \frac{\lambda_k}{-\lambda_i} \mathbf{x}_k = \mathbf{x}_i \quad (3)$$

However Eq. 3 implies that  $\mathbf{x}_i$  is a linear combinations of the other vectors present in  $S$ , which is a contradiction of *condition 1*.

**Part 2:** If a linear combinations of vectors in  $S$  with non-zero coefficients does not yield  $\mathbf{0} \rightarrow$  vectors in  $S$  are linearly independent.

We shall once again prove part 2 by contradiction.

**Condition 2:** If a linear combinations of vectors in  $S$  with non-zero coefficients does not yield  $\mathbf{0} \rightarrow$  vectors in  $S$  are linearly dependent.

Let us assume some vector  $\mathbf{x}_i \in S$  is linearly dependent on the rest of the vectors in  $S$ . This condition can be mathematically represented as follows:

$$\lambda_1 \mathbf{x}_1 + \lambda_2 \mathbf{x}_2 + \dots + \lambda_{i-1} \mathbf{x}_{i-1} + \lambda_{i+1} \mathbf{x}_{i+1} + \dots + \lambda_k \mathbf{x}_k = \mathbf{x}_i \quad (4)$$

Adding  $-\mathbf{x}_i$  to both sides of Eq. 4 we get:

$$\lambda_1 \mathbf{x}_1 + \lambda_2 \mathbf{x}_2 + \dots + \lambda_{i-1} \mathbf{x}_{i-1} + \lambda_{i+1} \mathbf{x}_{i+1} + \dots + \lambda_k \mathbf{x}_k - \mathbf{x}_i = \mathbf{0} \quad (5)$$

Eq. 5 has the vector  $\mathbf{x}_i$  with a non-zero coefficient i.e., -1, yet Eq. 5 yields  $\mathbf{0}$  which is a contradiction of condition 2. Hence by part 1, 2 we have proved the original statement of linear independence.

## Matrices

7. (10 points) Demonstrate the distributive property of matrix multiplication over addition.

Given  $A = \begin{bmatrix} 3 & 4 \\ 1 & 2 \end{bmatrix}$ ,  $B = \begin{bmatrix} 6 & 2 \\ 3 & 2 \end{bmatrix}$ ,  $C = \begin{bmatrix} 1 & -2 \\ 4 & -1 \end{bmatrix}$ , demonstrate:  $A(B+C) = AB + AC$

**Solution:** Given the matrices A, B, C, we first calculate AB, AC and B + C as follows.

$$AB = \begin{bmatrix} 3 \times 6 + 4 \times 3 & 3 \times 2 + 4 \times 2 \\ 1 \times 6 + 2 \times 3 & 1 \times 2 + 2 \times 2 \end{bmatrix} = \begin{bmatrix} 30 & 14 \\ 12 & 6 \end{bmatrix}$$

$$AC = \begin{bmatrix} 3 \times 1 + 4 \times 4 & 3 \times -2 + 4 \times -1 \\ 1 \times 1 + 2 \times 4 & 1 \times -2 + 2 \times -1 \end{bmatrix} = \begin{bmatrix} 19 & -10 \\ 9 & -4 \end{bmatrix}$$

$$AB + AC = \begin{bmatrix} 49 & 4 \\ 21 & 2 \end{bmatrix}$$

$$B + C = \begin{bmatrix} 6 + 1 & 2 + (-2) \\ 3 + 4 & 2 + (-1) \end{bmatrix} = \begin{bmatrix} 7 & 0 \\ 7 & 1 \end{bmatrix}$$

$$A(B + C) = \begin{bmatrix} 3 \times 7 + 4 \times 7 & 3 \times 0 + 4 \times 1 \\ 1 \times 7 + 2 \times 7 & 1 \times 0 + 2 \times 1 \end{bmatrix} = \begin{bmatrix} 49 & 4 \\ 21 & 2 \end{bmatrix}$$

Thus we have demonstrated that the distributive property holds for distribution of matrix multiplication over addition.

8. (15 points) Calculate the inverse of matrix  $A = \begin{bmatrix} 3 & 1 & 2 \\ -2 & -4 & 1 \\ 5 & -3 & 2 \end{bmatrix}$ . Note: It is acceptable to leave the final solution with fractional entities in the matrix (i.e., no requirement to convert fractions to decimal numbers).

**Solution:** The procedure we follow for calculating the inverse of a matrix is to first calculate the *minors* and *cofactors* followed by the *adjugate* matrix.  $A^{-1}$  is then obtained by multiplying the inverse of the determinant of matrix A with the adjugate matrix.

**Minors Calculation** The minors matrix is first calculated by considering a determinant of each  $2 \times 2$  matrix obtained by disregarding the row and column of the matrix element under consideration.

$$\begin{bmatrix} \begin{vmatrix} -4 & 1 \\ -3 & 2 \end{vmatrix} & \begin{vmatrix} -2 & 1 \\ 5 & 2 \end{vmatrix} & \begin{vmatrix} -2 & -4 \\ 5 & -3 \end{vmatrix} \\ \begin{vmatrix} 1 & 2 \\ -3 & 2 \end{vmatrix} & \begin{vmatrix} 3 & 2 \\ 5 & 2 \end{vmatrix} & \begin{vmatrix} 3 & 1 \\ 5 & -3 \end{vmatrix} \\ \begin{vmatrix} 1 & 2 \\ -4 & 1 \end{vmatrix} & \begin{vmatrix} 3 & 2 \\ -2 & 1 \end{vmatrix} & \begin{vmatrix} 3 & 1 \\ -2 & -4 \end{vmatrix} \end{bmatrix} \xrightarrow{\text{Minors}} \begin{bmatrix} -5 & -9 & 26 \\ 8 & -4 & -14 \\ 9 & 7 & -10 \end{bmatrix}$$

**Cofactors:** The cofactors matrix is obtained by multiplying the minors matrix (elementwise) with the following matrix.

$$\begin{bmatrix} +1 & -1 & +1 \\ -1 & +1 & -1 \\ +1 & -1 & +1 \end{bmatrix}$$

The co-factors matrix is as follows:

$$\begin{bmatrix} -5 & 9 & 26 \\ -8 & -4 & 14 \\ 9 & -7 & -10 \end{bmatrix}$$

**Adjugate Matrix:** The adjugate matrix is obtained as a transpose of the cofactors matrix.

$$\begin{bmatrix} -5 & -8 & 9 \\ 9 & -4 & -7 \\ 26 & 14 & -10 \end{bmatrix}$$

**Determinant of A:** The determinant of matrix  $A$  i.e.,  $|A|$  is obtained by applying the *Laplace expansion* to the top row of  $A$ , which can be achieved by multiplying element-wise the top row of the cofactors matrix with the top row of  $A$ .  $|A| = 3 \times (-5) + 1 \times (9) + 2 \times 26 = -15 + 9 + 52 = \mathbf{46}$

$\mathbf{A}^{-1}$  : The inverse of  $A$  is given by  $\frac{1}{|A|} \text{adj}(A)$  where  $\text{adj}(A)$  represents the adjugate matrix of  $A$ .

$$A^{-1} = \frac{1}{46} \begin{bmatrix} -5 & -8 & 9 \\ 9 & -4 & -7 \\ 26 & 14 & -10 \end{bmatrix} \rightarrow \begin{bmatrix} \frac{-5}{46} & \frac{-8}{23} & \frac{9}{46} \\ \frac{9}{46} & \frac{-2}{23} & \frac{-7}{46} \\ \frac{13}{23} & \frac{7}{23} & \frac{-5}{23} \end{bmatrix}$$

## Change of Bases

9. (15 points) Consider the three columns in matrix  $A$  (problem 8) to be our new basis of interest in  $R^3$ . If a vector  $\mathbf{x} = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$  defined on the natural basis in  $R^3$  (i.e.,  $\mathbf{e}_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$ ,  $\mathbf{e}_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$ ,  $\mathbf{e}_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$ ), how would vector  $\mathbf{x}$  be represented in the basis defined by the matrix  $A$  in problem 8.

**Solution:** Given the vector  $\mathbf{x}$  defined on the standard basis and given the new basis defined by matrix  $A$  (from problem 8), we know the relationship that

$A\mathbf{x}_A = \mathbf{x}$  holds. Our goal is to find the value of vector  $\mathbf{x}_A$  which is the representation of  $\mathbf{x}$  in the basis defined by  $A$ .

$$\begin{aligned} A\mathbf{x}_A &= \mathbf{x} \\ \mathbf{x}_A &= A^{-1}\mathbf{x} \end{aligned}$$

substituting in values of  $\mathbf{x}$  and  $A^{-1}$  we have the following relationship.

$$\mathbf{x}_A = \begin{bmatrix} \frac{-5}{46} & \frac{-4}{23} & \frac{9}{46} \\ \frac{9}{46} & \frac{-2}{23} & \frac{-7}{46} \\ \frac{13}{23} & \frac{7}{23} & \frac{-5}{23} \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} = \begin{bmatrix} \frac{-5-16+27}{46} \\ \frac{9-8-21}{46} \\ \frac{26+28-30}{46} \end{bmatrix} = \begin{bmatrix} \frac{6}{46} \\ \frac{-20}{46} \\ \frac{24}{46} \end{bmatrix}$$

Hence we have (the vector highlighted in the black box) as the value of the vector  $\mathbf{x}$  in the new basis defined by matrix  $A$ . This new vector is represented as vector  $\mathbf{x}_A$ .