

Linear Algebra

Definition 1.1: A

ector space (or linear space) V over a field \mathbb{F} consists of a set on which two operations (called addition and multiplication respectively here) are defined so that;

- (A) (V is Closed Under Addition) For all $\mathbf{x}, \mathbf{y} \in V$, there exists a unique element $\mathbf{x} + \mathbf{y} \in V$.
- (M) (V is Closed Under Scalar Multiplication) For all elements $a \in \mathbb{F}$ and elements $\mathbf{x} \in V$, there exists a unique element $a\mathbf{x} \in V$.

Such that the following properties hold:

- (VS 1) (Commutativity of Addition) For all $\mathbf{x}, \mathbf{y} \in V$, we have $\mathbf{x} + \mathbf{y} = \mathbf{y} + \mathbf{x}$.
- (VS 2) (Associativity of Addition) For all $\mathbf{x}, \mathbf{y}, \mathbf{z} \in V$, we have $(\mathbf{x} + \mathbf{y}) + \mathbf{z} = \mathbf{x} + (\mathbf{y} + \mathbf{z})$.
- (VS 3) (Existence of The Zero/Null Vector) There exists an element in V denoted by $\mathbf{0}$, such that $\mathbf{x} + \mathbf{0} = \mathbf{x}$ for all $\mathbf{x} \in V$.
- (VS 4) (Existence of Additive Inverses) For all elements $\mathbf{x} \in V$, there exists an element $\mathbf{y} \in V$ such that $\mathbf{x} + \mathbf{y} = \mathbf{0}$.
- (VS 5) (Multiplicative Identity) For all elements $x \in V$, we have $1\mathbf{x} = \mathbf{x}$, where 1 denotes the multiplicative identity in \mathbb{F} .
- (VS 6) (Compatibility of Scalar Multiplication with Field Multiplication) For all elements $a, b \in \mathbb{F}$ and elements $\mathbf{x} \in V$, we have $(ab)\mathbf{x} = a(b\mathbf{x})$.
- (VS 7) (Distributivity of Scalar Multiplication over Vector Addition) For all elements $a \in \mathbb{F}$ and elements $\mathbf{x}, \mathbf{y} \in V$, we have $a(\mathbf{x} + \mathbf{y}) = a\mathbf{x} + a\mathbf{y}$.
- (VS 8) (Distributivity of Scalar Multiplication over Field Addition) For all elements $a, b \in \mathbb{F}$, and elements $\mathbf{x} \in V$, we have $(a + b)\mathbf{x} = a\mathbf{x} + b\mathbf{x}$.

General Information

1. Let V be a vector space and W a subset of V . Then W is a subspace of V iff the following 3 conditions hold for the operations defined in V .
 - (a) $\mathbf{0} \in W$
 - (b) $\mathbf{x} + \mathbf{y} \in W$ whenever $\mathbf{x} \in W$ and $\mathbf{y} \in W$.
 - (c) $c\mathbf{x} \in W$ whenever $c \in \mathbb{F}$ and $\mathbf{x} \in W$.
2. For any matrix, its row space, column space, and dimension are identical.
3. A system $\mathbf{Ax} = \mathbf{b}$ is *homogeneous* iff $\mathbf{b} = \mathbf{0}$; otherwise it is *nonhomogeneous*.
4. A system $\mathbf{Ax} = \mathbf{b}$ of m linear equations in n unknowns has a solution space of dimension $n - \text{rank}(\mathbf{A})$.
5. A system $\mathbf{Ax} = \mathbf{b}$ of linear equations is *consistent* iff its solution set is nonempty; otherwise it is *inconsistent*.
6. A system $\mathbf{Ax} = \mathbf{b}$ is consistent iff $\text{rank}(\mathbf{A}) = \text{rank}(\mathbf{A}|\mathbf{b})$.
7. A matrix is said to be in *reduced row echelon form* iff

- (a) Any row containing a nonzero entry precedes any row in which all the entries are zero (if any).
- (b) The first nonzero entry in each row is the only nonzero entry in its column.
- (c) The first nonzero entry in each row is 1 and it occurs in a column to the right of the first nonzero entry in the preceding row.

8. Gaussian elimination.

- (a) In the forward pass, the augmented matrix is transformed into an upper triangular matrix in which the first nonzero entry of each row is 1 and it occurs in a column to the right of the first nonzero entry of each preceding row.
- (b) In the backward pass, the upper triangular matrix is transformed into reduced row echelon form by making the first nonzero entry of each row the only nonzero entry of its column.

9. Let \mathbf{A} be an $m \times n$ matrix, and \mathbf{a}_j its j th column. For any $\mathbf{x} = (x_1 \ x_2 \ \cdots \ x_n)^\top$,

$$\mathbf{A}\mathbf{x} = \sum_{j=1}^n x_j \mathbf{a}_j.$$

10. Let \mathbf{A} and \mathbf{B} be matrices having n rows. For any matrix \mathbf{M} with n columns, we have

$$\mathbf{M}(\mathbf{A}|\mathbf{B}) = (\mathbf{M}\mathbf{A}|\mathbf{M}\mathbf{B}).$$

11. The determinant of a square matrix can be evaluated by cofactor expansion along any row. That is, if $\mathbf{A} \in M_{n \times n}(\mathbb{F})$, then for any integer $1 \leq i \leq n$,

$$\det(\mathbf{A}) = \sum_{j=1}^n (-1)^{i+j} \mathbf{A}_{ij} \cdot \det(\tilde{\mathbf{A}}_{ij}).$$

Here, $\tilde{\mathbf{A}}_{ij}$ is the $(n-1) \times (n-1)$ matrix obtained from \mathbf{A} by deleting its i th row and j th column.

12. The determinant of a square matrix can also be evaluated by cofactor expansion along any column, since

$$\det(\mathbf{A}) = \det(\mathbf{A}^\top).$$

13. A matrix \mathbf{A} is invertible iff its determinant is nonzero.

14. Let \mathbf{A} be an invertible $n \times n$ matrix. Then, for some elementary row matrices \mathbf{E}_1 to \mathbf{E}_p ,

$$\mathbf{E}_p \mathbf{E}_{p-1} \cdots \mathbf{E}_1 (\mathbf{A} | \mathbf{I}_n) = \mathbf{A}^{-1} (\mathbf{A} | \mathbf{I}_n) = (\mathbf{I}_n | \mathbf{A}^{-1}).$$

In other words, we can perform Gaussian elimination, so that $(\mathbf{A} | \mathbf{I}_n) \rightarrow (\mathbf{I}_n | \mathbf{A}^{-1})$.

15. Alternatively, letting \mathbf{C} be the cofactor matrix of \mathbf{A} , i.e. $c_{ij} = (-1)^{i+j} \det(\tilde{\mathbf{A}}_{ij})$, we have

$$\mathbf{A}^{-1} = \frac{1}{\det(\mathbf{A})} \mathbf{C}^\top.$$