

Mathematical & Statistical Foundations for ML

- Linear Algebra Essentials

Vectors

- A vector is an ordered collection of numbers.
- Can be represented as a row or column.
- Used to represent features in ML.
- Example: $v = [2, 5, 7]$

Matrices

- A matrix is a 2D array of numbers arranged in rows and columns.
- Used to store datasets, transformations, etc.
- Example:
- $[1 \ 2 \ 3]$
- $[4 \ 5 \ 6]$
- $[7 \ 8 \ 9]$

Matrix Operations

- Addition: Element-wise.
- Multiplication: Dot product or matrix multiplication.
- Transpose: Flips rows to columns.
- Inverse: Exists if matrix is square and non-singular.

Eigen Decomposition

- Decomposes matrix into eigenvalues and eigenvectors.
- Used for understanding variance, dimensionality reduction.
- If $A^*v = \lambda^*v$
- Then:
- λ = eigenvalue
- v = eigenvector

Singular Value Decomposition (SVD)

- Decomposes matrix A into U, Σ , and V^T :
- $A = U \Sigma V^T$
- U: Left singular vectors
- Σ : Diagonal matrix of singular values
- V^T : Right singular vectors
- Applications: Image compression, recommendation systems

Use in Machine Learning

- Linear algebra is foundational to:
- - Principal Component Analysis (PCA)
- - Support Vector Machines (SVM)
- - Neural networks (weight matrices)
- - Optimization algorithms