# **Unsupervised Learning**

A Beginner's Introduction

Report Error 1

#### Outline

# Introduction

#### What is Unsupervised Learning?

- · Learning patterns from unlabeled data
- · No predefined output or target variable
- Goal: discover hidden structures, groupings, or representations
- · Common in exploratory data analysis

#### Why Unsupervised Learning?

- · Labeling data is expensive or impossible
- Understand data distribution and relationships
- · Useful for:
  - · Clustering customers, documents, images
  - · Reducing dimensionality for visualization or speed
  - Detecting anomalies or outliers

## Main Tasks

#### Clustering

- Group similar data points into clusters
- Examples:
  - k-Means: partition data into k groups by minimizing within-cluster variance
  - · Hierarchical Clustering: build a tree of clusters
  - · DBSCAN: density-based clustering, detects noise

### **Dimensionality Reduction**

- · Reduce number of features while preserving structure
- · Common methods:
  - PCA (Principal Component Analysis): finds directions of maximum variance
  - t-SNE, UMAP: nonlinear embeddings for visualization
  - Autoencoders: neural networks that learn compressed representations

#### Density Estimation and Anomaly Detection

- · Estimate probability distribution of data
- Detect outliers as points in low-density regions
- Examples:
  - Gaussian Mixture Models (GMM)
  - · Kernel Density Estimation (KDE)
  - One-class SVM

Popular Algorithms

#### k-Means Clustering

- · Initialize k centroids randomly
- · Assign points to nearest centroid
- Update centroids as mean of assigned points
- Repeat until convergence

#### Principal Component Analysis (PCA)

- · Linear projection to lower dimension
- · Finds orthogonal directions maximizing variance
- Useful for:
  - · Noise reduction
  - Visualization
  - Feature extraction

Challenges and Considerations

### Challenges in Unsupervised Learning

- No ground truth for evaluation
- Choosing number of clusters or components
- Sensitivity to initialization and parameters
- Scalability to large datasets
- Interpretability of results

# **Applications and Tools**

#### **Applications**

- Customer segmentation
- · Document and image organization
- · Anomaly detection in fraud, network security
- · Data compression and visualization

#### Popular Libraries

- · Scikit-learn: clustering, PCA, GMM, DBSCAN
- · TensorFlow/PyTorch: autoencoders and deep clustering
- HDBSCAN, UMAP packages for advanced clustering and visualization

# Summary

#### **Key Takeaways**

- · Unsupervised learning finds patterns without labels
- Key tasks: clustering, dimensionality reduction, density estimation
- · Many algorithms exist; choice depends on data and goal
- Evaluation is often subjective or uses proxy metrics

