Supervised Learning

- Analogy: Think of it like learning with a teacher or a supervisor. You are given examples
 with correct answers (labels), and you learn to predict the answers for new, unseen
 examples.
- **Data:** Uses **labeled data**. This means for each data point in the training set, there's a known outcome or "correct answer" associated with it.
 - Example: Emails labeled as "spam" or "not spam"; houses with their known sale prices.

• Types of Problems:

- Classification: The output label is a category (discrete value).
 - Examples: Spam detection (spam/not spam), image recognition (cat/dog/car), sentiment analysis (positive/negative/neutral), predicting customer churn (yes/no).
- Regression: The output label is a continuous value.
 - Examples: Predicting house prices, forecasting stock prices, predicting temperature, estimating student scores.
- Common Algorithms: Linear Regression, Logistic Regression, Support Vector Machines (SVM), k-Nearest Neighbors (k-NN), Decision Trees, Random Forests, Naive¹ Bayes, Neural Networks (often used for complex supervised tasks).
- **Feedback Mechanism:** During training, the algorithm makes predictions, compares them to the actual labels, and adjusts its internal parameters to minimize the errors.



2. Unsupervised Learning

- Analogy: Think of it like learning without a teacher. You are given a lot of data and asked to find interesting patterns, structures, or relationships within it on your own, without any predefined answers.
- **Data:** Uses **unlabeled data**. The training data consists only of input data points (X) without any corresponding output labels (Y).
- **Goal:** To model the underlying structure or distribution in the data to learn more about it. It's about discovering hidden patterns, groupings, or anomalies.
- Types of Problems:
 - Clustering: Grouping similar data points together based on their features. The algorithm discovers these groups (clusters).

- Examples: Customer segmentation (grouping customers with similar purchasing behavior), grouping similar documents, image segmentation.
- Association Rule Learning: Discovering rules that describe large portions of your data, like "people who buy X also tend to buy Y".
 - Examples: Market basket analysis (what items are frequently bought together in a supermarket), recommendation systems.
- Dimensionality Reduction: Reducing the number of variables (features) while preserving important information.
 - Examples: Feature extraction, data compression, noise reduction.
- Anomaly Detection: Identifying rare items, events, or observations which differ significantly from the majority of the data.
 - Examples: Fraud detection, identifying defective products, network intrusion detection.
- Common Algorithms: K-Means Clustering, Hierarchical Clustering, DBSCAN, Principal Component Analysis (PCA), Singular Value Decomposition (SVD), Apriori Algorithm, Isolation Forest.
- Feedback Mechanism: There are no correct answers to compare against. Instead, algorithms often work by minimizing/maximizing an internal objective function (e.g., minimizing distance within clusters in K-Means) or identifying statistical patterns.



