## Assignment: Clustering Vs Classification and Regression Vs Classification

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## **Differences Between Clustering and Classification**

## Clustering

- 1. **Definition**: Clustering is an unsupervised learning technique where the objective is to group a set of objects in such a way that objects in the same group (called a cluster) are more similar to each other than to those in other groups.
  - **Example**: Grouping customers based on purchasing behavior without predefined labels.
- 2. **Labels**: Clustering does not use predefined labels; it finds patterns and relationships in data to form clusters.
  - **Example**: Segmenting news articles into topics without knowing the topics in advance.
- 3. **Algorithm Types**: Common algorithms include K-Means, Hierarchical Clustering, and DBSCAN.
  - Example: Using K-Means to group images based on color similarity.
- 4. **Use Case**: Clustering is often used in exploratory data analysis to discover natural groupings in data.
  - **Example**: Identifying different species of plants in a botanical garden based on physical characteristics.
- 5. **Outcome**: The output of clustering is a set of clusters, where each cluster represents a group of similar data points.
  - Example: Grouping social media users based on their interaction patterns.

#### Classification

- 1. **Definition**: Classification is a supervised learning technique where the objective is to assign predefined labels to new observations based on a training dataset.
  - o **Example**: Classifying emails as spam or not spam.

- 2. **Labels**: Classification uses predefined labels and trains the model to learn the mapping from input features to these labels.
  - Example: Diagnosing diseases from medical images where the labels are known diagnoses.
- 3. **Algorithm Types**: Common algorithms include Decision Trees, Random Forest, Support Vector Machines, and Neural Networks.
  - **Example**: Using a Random Forest to classify whether a tumor is benign or malignant.
- 4. **Use Case**: Classification is used when the categories are known and the goal is to categorize new data points into these predefined categories.
  - Example: Categorizing customer support tickets into different issue types.
- 5. **Outcome**: The output of classification is a label for each input data point, predicting the category it belongs to.
  - **Example**: Predicting the species of a flower based on its petal and sepal measurements.

# **Differences Between Regression and Classification**

# Regression

- 1. **Definition**: Regression is a supervised learning technique where the goal is to predict a continuous output based on input features.
  - Example: Predicting the price of a house based on its size, location, and other features.
- 2. **Output**: The output of regression is a continuous value, which can be any real number.
  - o **Example**: Forecasting the temperature for the next week.
- 3. **Algorithm Types**: Common algorithms include Linear Regression, Polynomial Regression, and Support Vector Regression.
  - Example: Using Linear Regression to predict a student's final exam score based on study hours.
- 4. **Use Case**: Regression is used when the target variable is continuous and the goal is to predict a specific value.
  - o **Example**: Estimating the sales revenue for the next quarter.
- 5. **Evaluation Metrics**: Common evaluation metrics for regression include Mean Absolute Error (MAE), Mean Squared Error (MSE), and R-squared.
  - **Example**: Evaluating the performance of a regression model predicting stock prices using MSE.

### Classification

- 1. **Definition**: Classification is a supervised learning technique where the objective is to assign predefined labels to new observations based on a training dataset.
  - o **Example**: Classifying images as either of cats or dogs.
- 2. **Output**: The output of classification is a discrete label from a set of predefined categories.
  - o **Example**: Identifying whether a given email is spam or not spam.
- 3. **Algorithm Types**: Common algorithms include Logistic Regression, Decision Trees, Random Forest, and Naive Bayes.
  - **Example**: Using Logistic Regression to classify whether a patient has a certain disease based on medical test results.
- 4. **Use Case**: Classification is used when the target variable is categorical and the goal is to predict the category of new data points.
  - Example: Determining the genre of a song based on its audio features.
- 5. **Evaluation Metrics**: Common evaluation metrics for classification include Accuracy, Precision, Recall, F1 Score, and ROC-AUC.
  - Example: Measuring the performance of a classification model predicting loan defaults using Precision and Recall.