

# INTRODUCTION TO MACHINE LEARNING

## Basic Concepts, Supervised Learning, Unsupervised Learning

Author: Eng. Carlos Andrés Sierra, M.Sc.  
cavirguezs@udistrital.edu.co

Full-time Adjunct Professor  
Computer Engineering Program  
School of Engineering  
Universidad Distrital Francisco José de Caldas

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# Outline

- 1 Fundamentals of Machine Learning
- 2 Supervised Machine Learning
- 3 Unsupervised Machine Learning



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# Key Concepts in Machine Learning

## Machine Learning

- **Machine learning** is a method of data analysis that **automates** analytical model building.
- It is a **branch** of **artificial intelligence** based on the idea that systems can **learn from data**, **identify patterns** and **make decisions** with minimal human intervention.

• Supervised Learning: The model is trained on labeled data.

• Unsupervised Learning: The model is trained on unlabeled data.



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- **Reinforcement Learning**: The model learns by **interacting** with an environment.



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# Typical Machine Learning Problems

- **Classification:** Predicting a **label**.
- **Regression:** Predicting a **continuous value**.
- **Clustering:** Grouping **similar data** points.
- **Dimensionality Reduction:** Reducing the number of features.
- **Anomaly Detection:** Identifying **unusual data** points.
- **Association Rule Learning:** Identifying **relationships** between variables.





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# The Machine Learning Workflow

- **Data Collection:** Gathering the **data**.
- Data Preprocessing: Cleaning and preparing the **data**.
- Feature Engineering: Creating **new features**.
- Model Selection: Choosing the **best model**.
- Model Training: Training the **model** on the data.
- Model Evaluation: Assessing the **model's performance**.
- Model Deployment: Putting the model into production.



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# Algorithmic Bias

- **Algorithmic bias** is a **systematic error** in a model that results in **unfair outcomes**.
- It can be caused by **biased training data**, biased algorithms, or biased decision-making.



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# Introduction to Supervised Machine Learning

## Definition

- **Supervised learning** is a type of **machine learning** where the model is trained on **labeled data**.
- It involves training a model to **map input data to output data** based on example **input-output pairs**.



# Overfitting and Underfitting

## Overfitting

**Overfitting** occurs when a model learns the training data too well and performs poorly on new data.

## Underfitting

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# Supervised Learning Datasets

- **Training Dataset:** The data used to **train the model**.
- **Validation Dataset:** The data used to **tune the model hyperparameters**.
- **Test Dataset:** The data used to **evaluate the model performance**.



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# Cross-Validation

- **Cross-validation** is a technique for **assessing the performance** of a model.
- It involves **splitting** the data into multiple subsets, training the model on some subsets, and evaluating it on others.
- Common cross-validation **techniques** include **k-fold cross-validation** and **leave-one-out cross-validation**.
- Cross-validation helps to **reduce overfitting** and provides a more **accurate estimate** of the model's **performance**.



# K-Nearest Neighbors

- **K-Nearest Neighbors (KNN)** is a simple algorithm that stores all available cases and classifies new cases based on a **similarity measure**.
- It can be used for both **classification** and **regression** tasks.
- For **classification**, the output is the **class label** of the majority of the k-nearest neighbors.



# Linear Regression with Least Squares

## Linear Regression

- **Linear regression** is a type of **regression analysis** used for predicting the value of a **continuous dependent variable**.
- It works by finding the **line that best fits the data**.

## Least Squares

**Least squares** is a method for finding the **best-fitting line** by **minimizing the sum** of the squared differences between the **predicted** and **actual** values.



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# Decision Trees

- **Decision trees** are a type of **machine learning model** that can be used for both **classification** and **regression** tasks.
- They work by recursively **partitioning** the data into **subsets** based on the values of the features.



# Random Forest

- **Random forest** is an **ensemble learning** method that combines **multiple decision trees** to create a strong predictive model.
- It works by building **multiple trees** and averaging their predictions to **reduce overfitting**.



# Neural Networks

- **Neural networks** are a type of **machine learning model** inspired by the **human brain**.
- They consist of **layers** of interconnected nodes that process **input data** and produce **output data**.



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# Introduction to Unsupervised Machine Learning

## Definition

- **Unsupervised learning** is a type of **machine learning** where the model is trained on **unlabeled data**.
- It involves finding **patterns** and **relationships** in the data without any predefined labels.



# Clustering

- **Clustering** is a type of **unsupervised learning** that involves **grouping** similar **data points** together.
- Common clustering algorithms include **k-means**, **hierarchical clustering**, and **DBSCAN**.



# K-means

- **K-means** is a popular **clustering algorithm** that partitions the data into  $K$  **distinct clusters** based on feature similarity.
- It works by iteratively assigning data points to the nearest **cluster centroid** and updating the centroids based on the assigned points.



# Anomaly Detection

- **Anomaly detection** is a type of **unsupervised learning** that involves identifying **unusual data points** in a dataset.
- Common anomaly detection algorithms include **Isolation Forest**, **One-Class SVM**, and **Autoencoders**.





# Autoencoders

- **Autoencoders** are a type of **neural network** used for unsupervised learning.
- They work by **encoding** the input data into a lower-dimensional representation and then **decoding** it back to the original data.



# Dimensionality Reduction

- **Dimensionality reduction** is a technique for **reducing** the number of **features** in a dataset while retaining as much information as possible.
- Common dimensionality reduction techniques include **Principal Component Analysis (PCA)** and **t-Distributed Stochastic Neighbor Embedding (t-SNE)**.



# Principal Component Analysis (PCA)

- **Principal Component Analysis (PCA)** is a **statistical technique** used for **dimensionality reduction**.
- It works by transforming the data into a **new coordinate system** where the greatest variance lies along the first principal **component**, the second greatest variance along the second principal **component**, and so on.



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# Thanks!

## Questions?



Repo: <https://github.com/EngAndres/ud-public/tree/main/courses/machine-learning>

