

Lecture: 03

→ Supervised Learning

uses labeled data with paired inputs and outputs

• classification • Regression] Types of supervised learning.

→ unsupervised Learning

* no labeled data

* model identify patterns and group similar data points

→ Reinforcement Learning

* model (agent) interacts with an environment

* Receive feedback (reward or penalties) based on its actions

* goal is to maximize rewards

→ Applications of supervised Learning

Spam filtering - Audio to text conversion

Language translation - predicting ad clicks - Detecting

vehicles using radar - detecting product defects

→ Types of supervised learning:

1) classification (categorizing data into predefined labels)

2) Regression (Predicting continuous values)

→ Types of classification

1) Binary classification (two possible outcomes)

2) Multi-class classification (More than two outcomes)

→ common supervised Learning Algorithms

KNN

Logistic Regression

SVM

Log Linear Regression

Random forest

Decision tree

→ Applications of unsupervised Learning

clustering

Dimensionality Reduction

Anomaly Detection

→ common Algorithms of unsupervised Learning

K-Mean clustering

DBSCAN

Hierarchical clustering

one-class SVM

principal component Analysis (PCA)

→ common Algorithms of Reinforcement Learning

Q-Learning

SARSA

Deep Q-Network

policy Gradient Methods

Actor critic Method

Markov

Markov Decision process

→ Machine Learning workflow

- ① Data collection
- ② Data preprocessing
- ③ Model selection
- ④ Training the model
- ⑤ Evaluation
- ⑥ Deployment

→ challenges in ML

- * Data imbalance (some classes have much more data than others)
- * Feature selection
- * overfitting (work well on training data but poor on new ^{data})

Lecture: 04

• what is an Attribute?

An attribute is a distinct characteristic or property of a data object that can be measured, quantified or categorized.

Attribute → a column in database table

Machine Learning → (feature) an independent variable used for prediction or classification

variable → (statistical term) A measurable property that can take different values

→ Data object

sample (in statistics)

observation (in research)

data point (in visualization)

→ Types of attributes

Nominal → (categorical data) (no ordering)

↳ can represent using symbols or codes

↳ not supports mathematical operations

↳ mode ^{can be} ~~is~~ use calculated.

Binary → only two values

↳ represented using 0 or 1

↳ symmetric vs asymmetric

↓

Both values
are important

↓

one value is more
important

ordinal → inherent order or ranking exists

↳ (e.g) grading, ranks

↳ Discretization & ordinal attributes (converting numeric attributes into an ordinal attribute)

↳ operations (mode, median)

→ arithmetic operations are invalid

Numeric → Quantitative attributes

↳ Ratio scale

↳ Interval scale

→ Discrete attributes

finite or countable
(integer values)

→ continuous Attribute

floating point values

practical work on Google colab. →

Lecture: 05

Data preprocessing in ML

→ why do we need data preprocessing?

- * improve performance
- * Reduced complexity
- * Algorithm compatible
- * Ensure that data is in format : suitable for model interpretation

→ Data splitting (training, testing, validation)

Training data

- used for model learning
- should be accurate to ensure good learning
- it could be 80-20, 75-25, 70-30, etc

Testing data.

- used for checking the model learning
- if testing results incorrect, model has not learned well

→ Target variable selection

$X \rightarrow$ input data (independent variable)

$y \rightarrow$ target variable

→ Training data

→ Testing data

→ Data cleaning

- * Remove duplicate values / illogical values

- * remove incorrect data

- * handling missing values etc.

- * handling outliers

→ label encoding :- Assign a unique number to each category (e.g) Red $\rightarrow 0$, Blue $\rightarrow 1$, pink $\rightarrow 2$ etc

→ one-hot encoding :- ^{it} creates binary column for each category

→ ordinal encoding :- It encode categories by meaningful order

Low $\rightarrow 0$, medium $\rightarrow 1$, high $\rightarrow 2$.

practical work on google colab ----