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Section A

- ① A Classification problem involves predicting a categorical label for a given output based on the learned patterns from the Training data. The goal is to assign inputs to one of the predefined categories or classes.

Key Differences from Regression problem

- ① Output Type : Classification : The output is ~~continuous~~ categorical variable such as spam or not spam.

Regression: The output is continuous variable such as predicting house price or Temperature

- ② Evaluation Metrics :

Classification: Performance is often valued using metrics like accuracy, precision, recall, F1 score or the area under ROC curve

Regression: Performance is typically measured using metrics like mean squared error MSE, Mean Absolute Error MAE or R squared.

Three Algorithms

- 1) Decision Tree
- 2) Support Vector Machines (SVM)
- 3) K nearest Neighbours (K-NN)

b) In Logistic Regression, the odd ratio represent the ratio of the odds of an event occurring to the odds of it not occurring. It is used to measure the association between the predictor variable and the ~~outcome~~ outcome.

c) Factor Analysis

It is a statistical ~~and~~ method used to describe variability among observed variables in terms of fewer unobserved variables called factors. The goal is to identify the underlying relationships between the observed variables.

Applications

- 1) Data Compression: Simplifying complex data by uncovering hidden patterns and guiding strategic decision making. ~~across~~

1) Feature Extraction: We can identify the most significant features in data set. Helps in identifying distinct customer segments by understanding factors in customers' behaviour & preferences [customer segmentation].

Section 3 Part A

2) Dilemmas

1) Nature of the data: Time series problems: involves sequential data points ordered in time. The objective is often to predict future values based on past observations.

Regression Problem: involves predicting a continuous target variable based on a set of independent variables. The order of data points is not inherently important.

2) Dilemmas: Time Series Problem: Assumes that observations are dependent on the previous time points, or previous time points, in displaying temporal dependence.

Regression Problem: Assumes that observations are independent of each other.

Test-Train Split process

Time Series Problem: The data is typically split ~~chronologically~~ chronologically to preserve the time order. The Training set consists of an earlier time period & the test set consists of later time periods to simulate future predictions.

Regression Problem: The data can be split randomly since the order of observation does not affect the outcome. This ensures that both training & test set represent the same distribution.

b) Stationary. A time series is stationary if its statistical properties such as mean, variance & autocorrelation are constant over time.

Importance: It is crucial because many time series modelling techniques like ARIMA assume the time series is stationary. Non-stationary data can lead to misleading results & poor model performance.

Checking

- Visual Inspection : Plotting the time series to check for constant mean & variance
- Statistical Test : Using test like ADF test to formally check for stationarity

Common Test : Augmented Dickey Fuller (ADF)

C) Formatting Data object

In Time Series modelling the ^{date} object is usually formatted as 'datetime' to facilitate time based operations & analysis.

Eg import pandas as pd

Example Data string

data_string = "25-07-2024"

Convert to datetime object

data_object = pd.to_datetime(data_string, format='%d-%m-%Y')

Common Evaluation Metrics

- ▷ Mean Absolute Error
- ▷ Mean Squared Error