



Vidyavardhini's College of Engineering and Technology

Department of Artificial Intelligence & Data Science

AY: 2025-26

Class:	TE	Semester:	V
Course Code:	CSC504	Course Name:	Data Warehousing & mining

Name of Student:	Pranita Kumbhar
Roll No. :	70
Assignment No.:	03
Title of Assignment:	Classification
Date of Submission:	
Date of Correction:	

Evaluation

Performance Indicator	Max. Marks	Marks Obtained
Completeness	5	5
Demonstrated Knowledge	3	3
Legibility	2	2
Total	10	10

Performance Indicator	Exceed Expectations (EE)	Meet Expectations (ME)	Below Expectations (BE)
Completeness	5	3-4	1-2
Demonstrated Knowledge	3	2	1
Legibility	2	1	0

Checked by

Name of Faculty : Ms. Neha Raut
Signature :
Date :

Assignment 3.

Q.1] Explain the steps of Min-max normalization and apply them to transform an attribute value from an old range to a new range.

Min-Max normalization is a technique used in data preprocessing to rescale the range of features (attributes) into a specific range, usually $[0, 1]$ or $[-1, 1]$. This ensures that no feature dominates others because of its scale.

• Steps of Min-Max Normalization:

1] Identify the old minimum (Old_min) and maximum (Old_max) values of the attribute.

2] Decide the new-range (New_min, New_max) where you want to scale the values.

3] Apply the formula:

$$V_{\text{new}} = \frac{(V_{\text{old}} - \text{Old}_{\text{min}})}{(\text{Old}_{\text{max}} - \text{Old}_{\text{min}})} \times (\text{New}_{\text{max}} - \text{New}_{\text{min}}) + \text{New}_{\text{min}}$$

where,

V_{old} = original value.

V_{new} = normalized value.

Example :

Suppose we have exam marks in the range $[20, 80]$, and we want to normalize them into $[0, 1]$.

- Old-min = 20,
- Old-max = 80,
- New-min = 0,
- New-max = 1
- Suppose $V_{old} = 50$

$$V_{new} = \frac{(50 - 20)}{(80 - 20)} \times (1 - 0) + 0$$
$$= \frac{30}{60}$$

$$V_{new} = \underline{\underline{0.5}}$$

The normalized value of 50 in $[0, 1]$ range is 0.5.

Q.2] Explain the term True Positive, False Positive, True Negative, and False Negative in a confusion matrix. Apply them in the case of spam email detection.

→

A confusion matrix is used to evaluate the performance of a classification model. It compares the predicted values with actual values.

- Confusion matrix for Spam detection:

	Predicted Spam	Predicted Not Spam
Actual Spam	TP	FN
Actual Not Spam	FP	TN

Q.3] For the given dataset. Apply Naïve - Bayes algorithm and predict the outcome for Car = (Red, Domestic, SUV).

Car	Type	Origin	Stolen
Red	Sports	Domestic	Y
Red	Sports	Domestic	N
Red	Sports	Domestic	Y
Yellow	Sports	Domestic	N
Yellow	Sports	Imported	Y
Yellow	SUV	Imported	N
Yellow	SUV	Imported	Y
Yellow	SUV	Domestic	N
Red	SUV	Imported	N
Red	Sports	Imported	Y

→

$$P(A/B) = \frac{P(B/A) \cdot P(A)}{P(B)}$$

$$X = [\text{Red, Domestic, SUV}]$$

- True Positive (TP) :
The model correctly predicts a positive class.
- In spam detection :
An email is actually spam and the model predicts it as spam.
- False Positive (FP) :
The model incorrectly predicts positive when it is negative.
- In spam detection :
An email is not spam (ham), but the model predicts it as spam. (Type I error).
- True Negative (TN) :
The model correctly predicts the negative class.
- In spam detection :
An email is not spam and the model predicts it as not spam.
- False Negative (FN) :
The model incorrectly predicts negative when it is positive.
- In spam detection :
An email is spam but the model predicts it as not spam. (Type II error).

$$i] P(\text{Red} / \text{Yes}) = \frac{P(\text{Yes} / \text{Red}) \cdot P(\text{Red})}{P(\text{Yes})}$$

=

$$\frac{3/5 \cdot 5/10}{5/10}$$

$$= \underline{\underline{3/5}}$$

$$ii] P(\text{Domestic} / \text{Yes}) = 2/5$$

$$iii] P(\text{SUV} / \text{Yes}) = 1/5$$

$$P(X / \text{Yes}) = \frac{3}{5} \times \frac{2}{5} \times \frac{1}{5} = \frac{6}{125}$$

$$= \underline{\underline{0.024}}$$

$$i] P(\text{Red} / \text{No}) = \frac{P(\text{No} / \text{Red}) \cdot P(\text{Red})}{P(\text{No})}$$

$$= \frac{2/5 \cdot 5/10}{5/10}$$

$$\frac{5}{10}$$

$$= \underline{\underline{2/5}}$$

$$ii] P(\text{Domestic} / \text{No}) = 3/5$$

$$iii] P(\text{SUV} / \text{No}) = 2/5$$

$$P(X / \text{No}) = \frac{2}{5} \times \frac{3}{5} \times \frac{2}{5} = \underline{\underline{0.072}} \quad \underline{\text{No.}}$$

The predicted outcome for car = [Red, Domestic, SUV]
for Yes = 0.024

and No is 0.072.