Experiment. No.08

Title: classify the email using the binary classification method. Email spam detection has two states a) Normal state - Not spam

b) Abnormal state - spam

Use k-negrest neighbours and support vactor machine

Use k-negrest neighbours and support vector machine for classification, Analyze their performance

objective: To classify email using the binary classification and implemented email spam detection technique by using k-Nearest Neighbours and support vector machine algorithm:

Pataset Description:

the csv file contains s172 rows. each row for each email. These are 3002 columns. The first column indicates Email name the name has been set with numbers and not recipient's name to protect privacy. The last column has the labels for prediction: 1 for spam, o for not spam.

Theory:

Data Preprocessing:

A real-world data generally contains noises, missing values & maybe a in an unsable format which cannot be directly used for machine learning models.

pata preprocessing is required tasks for cleaning the data & making it suitable for a ML model which also increases the accuracy & efficiency of a ML model.

steps 1. Get dataset 2. Import libraries a import dataset 4. Finding missing data 5. Encoding categorical data 6. splitting dataset into training & test set 7. Feature scaling Binary classification In binary classification, the goal is 10 classify the input into one of two classes 08 categories. Example - on the basis of the given health conditions of a person, we have to determine whet the person has a certain diseases or not Binary classification X 2 k-nearest neighbours-KNN algorithm at the training phase stores the dataset and when it gets new data then it classifies that Sata into a category much similar to the new data. Example - suppose, we have an image creature that looks similar to cat &

want to know either it is a	cat or dog . our KNN
model will find similar feature	s of the new dataset
to the cats & dog images &	based on the most similar
features.	1 alacrithm
It is a non-parametric as	well as lazy algorithm
Before KNN X	ofter KNN
category B	cat B
3,010,010,010	Assigned Assigned
New data point	Assigned cat B
category A	· · · cat A
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The state of the s	
applications because megation data and non linear relation main objective of sur	Variety of tasks such as fication, spam detection, ficient in a variety of a manage high-dimensional
space. X2	
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It then	separate data into two different da
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	t and split Procedure:
The	train_test_split() method is used
Spiir our a	lata into train and test sets.
Train sat	
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Final mode Final mode First, we and labels (Dataframe & Y-test X-train & model if Conclusion:	tilized to give an accurate evaluation and subset of the training data into features need to divide our data into features gets divided into x-train, x-test, y-train