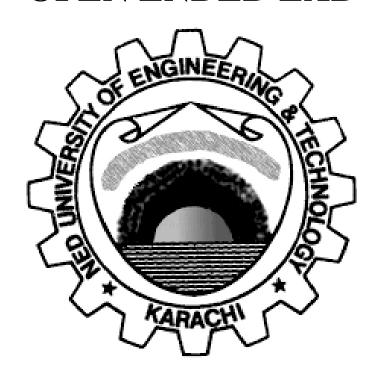
MACHINE LEARNING CS-324 OPEN ENDED LAB



FRAUD DETECTION FOR TWITTER ACCOUNTS

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INTRODUCTION

The primary objective of this project is to detect fraudulent Twitter accounts using machine learning algorithms. The dataset was sourced from Kaggle and includes various features that can help identify whether an account is fraudulent or not. The steps followed in this project include data collection, preprocessing, exploratory data analysis (EDA), feature engineering, model building, model evaluation, and the creation of a user-friendly interface for model interaction.

DATA COLLECTION

The dataset (twitter_human_bots_dataset) used in this project was obtained from Kaggle. It contains a comprehensive set of features about Twitter accounts, including user profile details, followers count, friends count, status count, average tweets per day and account age in days, etc. The dataset has 37438 records and 20 features, with a target variable indicating whether an account is fake or not.

DATA PREPROCESSING AND EXPLORATORY DATA ANALYSIS (EDA)

Data preprocessing involved cleaning the dataset by handling missing values and encoding categorical variables. Missing values were removed, and categorical variables were transformed into numerical values. Visualizations such as count plots and pair plots were used to understand the distributions and patterns in the data. EDA was conducted to gain deeper insights into the data. Heat maps were used to examine the distribution of variables, while a correlation matrix helped identify relationships between features.

MODEL BUILDING

Three machine learning algorithms were used to build predictive models: Logistic Regression, K-Nearest Neighbors (KNN), and Decision Trees. Each model was implemented using Python packages such as scikit-learn and also manually to understand the underlying mechanics.

- 1. Logistic Regression:
 - Implemented using scikit-learn's LogisticRegression class.
 - Manually implemented by computing the sigmoid function and gradient descent.
- 2. K-Nearest Neighbors (KNNs):
 - Implemented using scikit-learn's KNeighborsClassifier.
 - Manually implemented by calculating the Euclidean distance and majority voting.
- 3. Decision Trees:
 - Implemented using scikit-learn's DecisionTreeClassifier.
 - Manually implemented by constructing a tree based on information gain.

MODEL EVALUATION

The models were evaluated using various metrics such as accuracy, precision, recall, F1 score, and ROC-AUC. The performance of each model was compared to identify the best-performing one.

1. Logistic Regression:

Accuracy = 76% Precision = 74% Recall = 71% F1 Score = 72%

2. K-Nearest Neighbors (KNNs):

Accuracy = 85% Precision = 84% Recall = 82% F1 Score = 83%

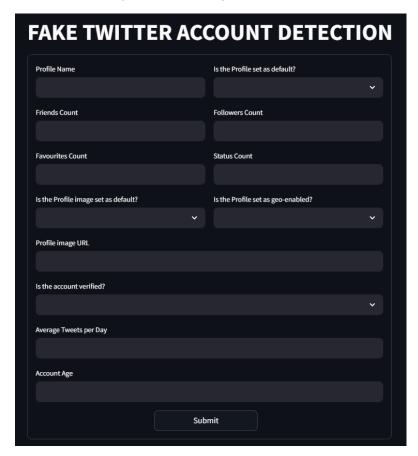
3. Decision Trees:

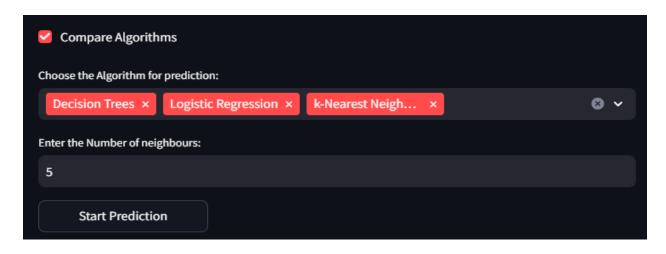
Accuracy = 83% Precision = 80% Recall = 81% F1 Score = 81%

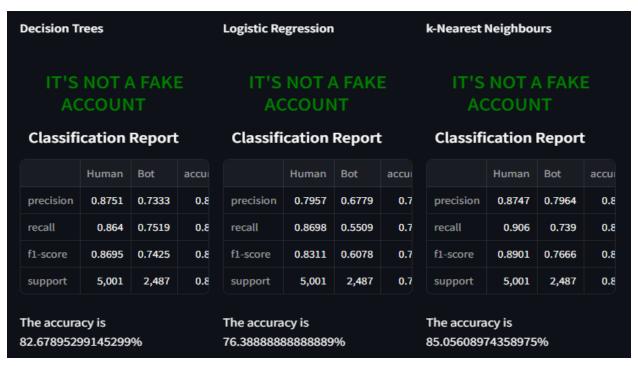
K-Nearest Neighbors emerged as the best-performing model based on these metrics.

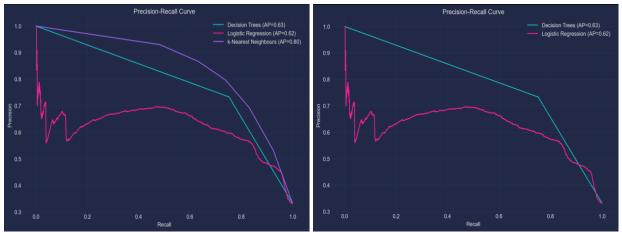
USER-FRIENDLY INTERFACE

A user-friendly interface was created using Streamlit, allowing users to compare the performance of multiple algorithms and predict the likelihood of an account being fraudulent using an unknown sample. This interface enhances the usability and accessibility of the model.

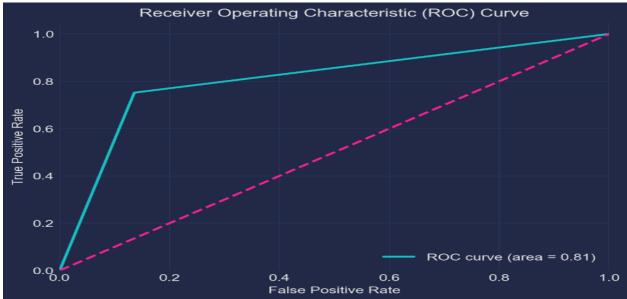














CONCLUSION

The project successfully demonstrated the application of machine learning algorithms to detect fraudulent Twitter accounts. Future work could explore more advanced algorithms and larger datasets for improved performance.