MAJOR PROJECT

BY: SWASTIK MOHAPATRA

Aim:

we have been given a "diabetes.csv" data file . This dataset is originally from the National Institute of Diabetes and Digestive and Kidney Diseases. The objective of the dataset is to diagnostically predict whether or not a patient has diabetes, based on certain diagnostic measurements included in the dataset. Several constraints were placed on the selection of these instances from a larger database. In particular, all patients here are females at least 21 years old of Pima Indian heritage.

And hence we need to apply at minimum 2 classification algorithms and compare their respective accuracies.

Preview: this project classifying whether a person has diabetes or not considering the different data columns given in the csv file that where:

"pregnancies", "glucose", "blood pressure", "skin thickness", "insulin", "BMI", "diabetes pedigree function" and "age".

We tend to find out the relation and apply 4 machine learning algorithms to the following dataset in order to predict their respective accuracies and hence point out the best prediction algorithm among them.

Tools used:

There are a number of tools used for completing this major project some of them being as follows:

Pandas – for implementing fataframe structures and using it to store the values of the given csv files (eg: pd.read_csv("file_path"))

Matplotlib – for implementing graphical representation of the results as well as the analysis as it helps us to get a clear overview and errors in the data.

Sklearn – this library seems to help us a lot from data pre processing for generating usable data to dividing test and train data to implementing the 4 classification algorithms it helps us access pre defined classes for classifying data.

Python – is the programming language we have used to write the code implementing the machine learning concepts.

```
[] import pandas as pd
  import matplotlib.pyplot as plt
  import seaborn as sns
  from sklearn.model_selection import train_test_split
  from sklearn.preprocessing import StandardScaler
  from sklearn.linear_model import LogisticRegression
  from sklearn.tree import DecisionTreeClassifier
  from sklearn.ensemble import RandomForestClassifier
  from sklearn.svm import SVC
  from sklearn.metrics import accuracy_score
```

Pre-processing and understanding:

We analyse the data before applying the algorithms to the dataset for machine learning. While analysing I plotted histograms to possibly guess important factors.

We also got abnormal values where BMI, glucose and skin thickness were 0 and insulin being 0 wasn't an uncommon phenomenon.

```
(df== 0).sum()
Pregnancies
                                111
   Glucose
                                  5
   BloodPressure
                                 35
    SkinThickness
                                227
    Insulin
                                374
    BMI
                                 11
   DiabetesPedigreeFunction
                                  0
                                  0
    Outcome
                                500
    dtype: int64
```

```
df_refined=df[(df.Glucose!=0) & (df.BloodPressure!=0) & (df.BMI!=0)]
#data cleaning
df_refined.shape

(724, 9)
```

so we cleaned the data accordingly and then divided the it into train and test parts further scaling them relatively using StandardScaler.

Algorithms:

In this project I have implemented 4 classification algorithm:

- Logistic regression: the logistic model is used to model the probability of a certain class or event existing such as pass/fail, win/lose, alive/dead or healthy/sick. Basically it points the data points in a 2 dimensional graph and divides it into 2 classes like here being diabetic or not.
- 2. Decision tree: A decision tree is a decision support tool that uses a tree-like model of decisions and their possible consequences, including chance event outcomes, resource costs, and utility.it creates conditional clauses in a tree like format with a specific priority order helping us to classify the given dataset.
- 3. Random forest: Random forests or random decision forests are an ensemble learning method for classification, regression and other tasks that operates by constructing a multitude of decision trees at training time. For classification tasks, the output of the random forest is the class selected by most trees.

4. Support vector machines: In machine learning, support-vector machines are supervised learning models with associated learning algorithms that analyse data for classification and regression analysis. Here we have classified the data into 2 parts.

```
[61] lr=LogisticRegression(random_state=0)
    dct=DecisionTreeClassifier(random_state=0)
    rfc=RandomForestClassifier(random_state=0)
    sv=SVC(random_state=0,kernel='rbf')
    #TRIED CHANGING KERNEL VALUES BUT GOT EVEN LOWER ACCURACY
    x=df_refined.iloc[:,0:8].values
    y=df_refined.iloc[:,8].values
    print(x)
    print(y)
```

CONCLUSION:

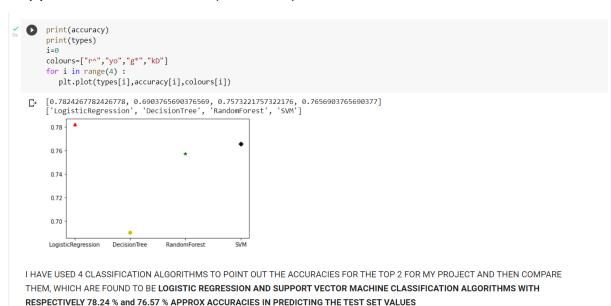
We applied the 4 classification algorithms for the given diabetes dataset and found their respective accuracies which are :

Logistic regression: 78.24 %

Decision tree: 69.04 %

Random forests: 75.73 %

Support vector machines(classifier): 76.57 %



(THE .ipynb file has bin given in the zipped folder)

NAME-SWASTIK MOHAPATRA

STREAM- COMPUTER ENGINEERING

YEAR OF STUDY- 1st

COLLEGE- BHARATI VIDYAPEETH DEEMED TO UNIVERSITY BE COLLEGE OF ENGINEERING, PUNE.

College prn : 2014110710