

Programming for Bioinformatics (SECB3203)

PROJECT PROPOSAL

Title: Diabetic Prediction using Machine Learning

Group 18

Group Members	Matric Number
HARCHANA A/P ARULAPPAN	A21EC0028
MALAVIKA A/P BASKARAN	B22EC0069

Table of Contents

1.0 Importing Dataset	2
2.0 Check for Missing Values	3
3.0 Check missing values in each column	4
4.0 Check for duplicate rows	. 5
5.0 Data Formatting	. 5
6.0 Data Normalization	. 6
7.0 Data Binning	. 7

1.0 Importing Dataset

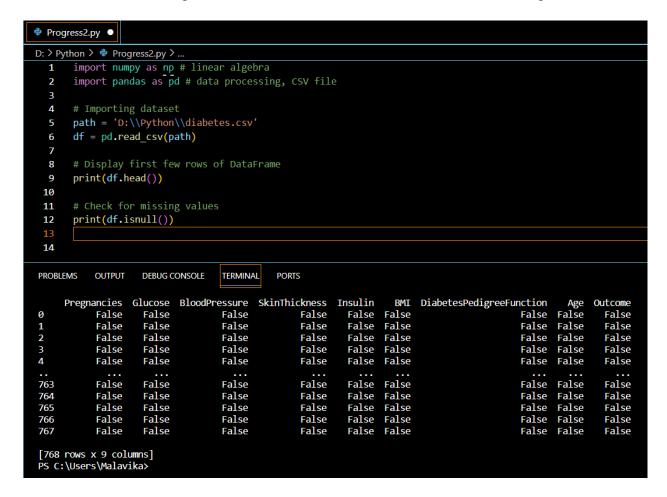
Our dataset from kaggle: https://www.kaggle.com/datasets/uciml/pima-indians-diabetes-database

We import the diabetes.csv file from the local directory in Python and display the result to the first 5 rows.

```
Progress2.py
D: > Python > 💠 Progress2.py > ...
        import numpy as np # linear algebra
        import pandas as pd # data processing, CSV file
        # Importing dataset
        path = 'D:\\Python\\diabetes.csv'
        df = pd.read_csv(path)
        # Display first few rows of DataFrame
        print(df.head())
  10
PROBLEMS
            OUTPUT
                      DEBUG CONSOLE
                                        TERMINAL
                                                   PORTS
PS C:\Users\Malavika> & D:/Python/python.exe d:/Python/Progress2.py
Pregnancies Glucose BloodPressure SkinThickness Insulin BMI
                                                                                                           Age
50
                                                                              DiabetesPedigreeFunction
                                                                                                                 Outcome
                       148
                                         72
                                                                    0
                                                                       33.6
                                                                                                   0.627
 1
2
3
               1
                        85
                                         66
                                                         29
                                                                                                   0.351
                                                                                                            31
                                                                                                                       0
                                                                    0
                                                                       26.6
                                                          0
                                                                                                                       1
               8
                       183
                                         64
                                                                    0
                                                                       23.3
                                                                                                   0.672
                                                                                                            32
               1
                                                         23
                                                                                                            21
                        89
                                         66
                                                                   94
                                                                       28.1
                                                                                                   0.167
                       137
                                                                  168
                                                                       43.1
                                                                                                    2.288
                                                                                                            33
                                                                                                                       1
 PS C:\Users\Malavika>
```

2.0 Check for Missing Values

Next, we checked missing values in our dataset and found that there is no missing values.



3.0 Check missing values in each column

Then, we checked missing values in each column.

```
Progress2.py X
D: > Python > ♥ Progress2.py > ...
 13
       # Check missing values in each columns
 14
       print(df.isnull().sum())
 15
 16
 17
PROBLEMS
           OUTPUT
                     DEBUG CONSOLE
                                     TERMINAL
                                                PORTS
Pregnancies
                             0
Glucose
                             0
BloodPressure
                             0
SkinThickness
                             0
Insulin
                             0
BMI
                             0
DiabetesPedigreeFunction
                             0
Age
                              0
Outcome
                             0
dtype: int64
```

4.0 Check for duplicate rows

Then, we checked for any duplicate rows in the dataset and we found that there is no duplicate rows.

```
D: > Python > Progress2.py > ...

10

17  # Check for duplicate rows

18  duplicate_rows = df[df.duplicated()]

19  print(duplicate_rows)

20

21

PS C:\Users\Malavika> & D:/Python/python.exe d:/Python/Progress2.py

Empty DataFrame
Columns: [Pregnancies, Glucose, BloodPressure, SkinThickness, Insulin, BMI, DiabetesPedigreeFunction, Age, Outcome]
Index: []
```

5.0 Data Formatting

Display data types in each column.

```
Progress2.py •

D: > Python > Progress2.py > ...

20
21  # Display data types
22  print(df.dtypes)
23
24
25
```

```
PS C:\Users\Malavika> & D:/Python/python.exe d:/Python/Progress2.py
Pregnancies
                               int64
Glucose
BloodPressure
                               int64
SkinThickness
                               int64
Insulin
                               int64
                             float64
BMT
DiabetesPedigreeFunction
                             float64
                               int64
Age
Outcome
                               int64
dtype: object
```

6.0 Data Normalization

For data normalization, we use Z-score to measure how many standard deviations a data point is from the mean of a dataset.

```
Progress2.py
D: > Python > Progress2.py > ...
       numeric_data = df.select_dtypes(include=['float64', 'int64'])
 28
       # Data Normalization (Z-Score)
 29
       scaler = StandardScaler()
       df_normalized = pd.DataFrame(scaler.fit_transform(numeric_data), columns = df.columns)
 30
 31
       print("Normalized DataFrame: ")
       print(df_normalized)
 33
 34
PROBLEMS
          OUTPUT
                   DEBUG CONSOLE
                                   TERMINAL
                                             PORTS
PS C:\Users\Malavika> & D:/Python/python.exe d:/Python/Progress2.py
Normalized DataFrame:
     Pregnancies Glucose BloodPressure SkinThickness
                                                                        BMI DiabetesPedigreeFunction
                                                          Insulin
                                                                                                                  Outcome
                                                                                                            Age
        0.639947 0.848324
                                0.149641
                                                0.907270 -0.692891 0.204013
                                                                                             0.468492 1.425995
                                                                                                                 1.365896
                                                0.530902 -0.692891 -0.684422
1
2
3
4
       -0.844885 -1.123396
                                -0.160546
                                                                                             -0.365061 -0.190672 -0.732120
                                                                                             0.604397 -0.105584 1.365896
                                -0.263941
        1.233880 1.943724
                                               -1.288212 -0.692891 -1.103255
                                -0.160546
       -0.844885 -0.998208
                                                0.154533 0.123302 -0.494043
                                                                                             -0.920763 -1.041549 -0.732120
       -1.141852 0.504055
                                -1.504687
                                                0.907270
                                                         0.765836
                                                                   1.409746
                                                                                             5.484909 -0.020496
                                                                                                                 1.365896
        1.827813 -0.622642
                                                                                            -0.908682
763
                                 0.356432
                                                1.722735
                                                         0.870031
                                                                   0.115169
                                                                                                       2.532136 -0.732120
764
       -0.547919 0.034598
                                 0.046245
                                                0.405445 -0.692891 0.610154
                                                                                            -0.398282 -0.531023 -0.732120
                                 0.149641
                                                0.154533 0.279594 -0.735190
765
        0.342981 0.003301
                                                                                            -0.685193 -0.275760 -0.732120
       -0.844885 0.159787
                                -0.470732
                                               -1.288212 -0.692891 -0.240205
                                                                                            -0.371101 1.170732 1.365896
766
       -0.844885 -0.873019
                                 0.046245
                                                0.656358 -0.692891 -0.202129
                                                                                            -0.473785 -0.871374 -0.732120
[768 rows x 9 columns]
PS C:\Users\Malavika>
```

7.0 Data Binning

We use age group for data bining.

```
# Binning the 'age' columns
        bins = [20, 30, 40, 50, 60, 70, 80, 90]
        labels = ['20-30', '31-40', '41-50', '51-60', '61-70', '71-80', '81-90']
df['age_grouop'] = pd.cut(df['Age'], bins=bins, labels=labels, right=False)
 37
 38
 39
        print(df['age_grouop'])
 40
PROBLEMS
              OUTPUT
                         DEBUG CONSOLE
                                             TERMINAL
                                                          PORTS
PS C:\Users\Malavika> & D:/Python/python.exe d:/Python/Progress2.py
0
         31-40
         31-40
3
4
         20-30
         31-40
763
         61-70
764
         20-30
765
         31-40
         41-50
766
767
         20-30
Name: age_grouop, Length: 768, dtype: category
Categories (7, object): ['20-30' < '31-40' < '41-50' < '51-60' < '61-70' < '71-80' < '81-90']
PS C:\Users\Malavika> [
```