

AI BASED DIABETES PREDICTION SYSTEM



**J.SEBASTIN SELVIN ERAUD
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RVS COLLEGE OF
ENGINEERING**

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```

        IMPORT DATASET
        DF = PD.READ_CSV("../INPUT/PIMA-
        INDIANS-DIABETES-
        DATABASE/DIABETES.CSV")
# GET FAMILIER WITH DATASET STRUCTURE
        DF.INFO()

```

```

class 'pandas.core.frame.DataFrame'>
RangeIndex: 768 entries, 0 to 767
Data columns (total 9 columns):
   Column                Non-Null Count  Dtype
-  -----
   Pregnancies           768 non-null    int64
   Glucose               768 non-null    int64
   BloodPressure         768 non-null    int64
   SkinThickness         768 non-null    int64
   Insulin               768 non-null    int64
   BMI                   768 non-null    float64
   DiabetesPedigreeFunction 768 non-null    float64
   Age                   768 non-null    int64
   Outcome               768 non-null    int64
dtypes: float64(2), int64(7)
memory usage: 54.1 KB

```

```
DF['GLUCOSE'] = DF['GLUCOSE'].REPLACE(0,  
    DF['GLUCOSE'].MEAN())  
# CORRECTING MISSING VALUES IN BLOOD  
    PRESSURE  
    DF['BLOODPRESSURE'] =  
    DF['BLOODPRESSURE'].REPLACE(0,  
    DF['BLOODPRESSURE'].MEAN()) # THERE  
ARE 35 RECORDS WITH 0 BLOODPRESSURE  
    IN DATASET  
# CORRECTING MISSING VALUES IN BMI  
    DF['BMI'] = DF['BMI'].REPLACE(0,  
    DF['BMI'].MEDIAN())
```

```
# Data Transformation  
q = QuantileTransformer()  
X = q.fit_transform  
transformedDF = q.transform(X)  
transformedDF = pd.DataFrame(X)  
transformedDF.columns =  
['Pregnancies', 'Glucose',  
'BloodPressure', 'SkinThickness',  
'Insulin', 'BMI',  
'DiabetesPedigreeFunction', 'Age',  
'Outcome']  
# Show top 5 rows  
transformedDF.head
```

```
import pandas as pd  
from sklearn.model_selection  
import train_test_split  
from sklearn.preprocessing  
import StandardScaler,  
LabelEncoder
```

```
# Load the dataset  
data =  
pd.read_csv('your_dataset.csv')
```

```
# Data Exploration (Optional)  
# You can explore
```

Data Cleaning

```
# Handle missing values and remove  
duplicates if necessary
```

Data Preprocessing

**# For example, encoding
categorical variables**

label_encoder =

LabelEncoder()

**data['categorical_column'] =
label_encoder**

**X = data.drop('target_column',
axis=1) # Features**

**y = data['target_column'] #
Target variable**