Al Based Diabetes Prediction System

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PROJECT: Al Based Diabetes Prediction System

PROBLEM DEFINITION:

The problem is to build an Al-powered diabetes prediction system that uses machine learning algorithms to analyze medical data and predict the likelihood of an individual developing diabetes. The system aims to provide early risk assessment and personalized preventive measures, allowing individuals to take proactive actions to manage their health. Early prediction of diabetes and prediabetes can reduce treatment cost and improve intervention. The development of (pre)diabetes is associated with various health conditions that can be monitored by routine health checkups. This study aimed to develop amachine learning-based model for predicting (pre)diabetes.

DESIGN THINKING:

- 1. Data Collection: We need a dataset containing medical features such as glucose levels, blood pressure, BMI, etc., along with information about whether the individual has diabetes or not.
- 2. Data Preprocessing: The medical data needs to be cleaned, normalized, and prepared for training machine learning models.
- 3. Feature Selection: We will select relevant features that can impact diabetes risk prediction.
- 4. Model Selection: We can experiment with various machine learning algorithms like Logistic Regression, Random Forest, and Gradient Boosting.

- 5. Evaluation: We will evaluate the model's performance using metrics like accuracy, precision, recall, F1-score, and ROC-AUC.
- 6. Iterative Improvement: We will fine-tune the model parameters and explore techniques like feature engineering to enhance prediction accuracy.

PROGRAM:

```
import pandas as pd
import numpy as np
from sklearn.preprocessing import StandardScaler
from sklearn.linear_model import LogisticRegression
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score, confusion_matrix
import matplotlib.pyplot as plt
import seaborn as sns
import pandas as pd
data = pd.read_csv('/kaggle/input/diabetes-data-set/diabetes.csv')
data.head()
data.describe()
data.isnull().sum()
```

```
data['BMI'] = data['BMI'].replace(0,data['BMI'].mean())
data['BloodPressure'] =
data['BloodPressure'].replace(0,data['BloodPressure'].mean())
data['Glucose'] = data['Glucose'].replace(0,data['Glucose'].mean())
data['Insulin'] = data['Insulin'].replace(0,data['Insulin'].mean())
data['SkinThickness'] =
data['SkinThickness'].replace(0,data['SkinThickness'].mean())
import matplotlib.pyplot as plt
import seaborn as sns
fig, ax = plt.subplots(figsize=(15,10))
sns.boxplot(data=data, width= 0.5,ax=ax, fliersize=3)
X = data.drop(columns = ['Outcome'])
y = data['Outcome']
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test =
train_test_split(X,y,test_size=0.25,random_state=0)
X_train.shape, X_test.shape
```

```
import pickle
##standard Scaling- Standardization
def scaler_standard(X_train, X_test):
  #scaling the data
  scaler = StandardScaler()
  X_train_scaled = scaler.fit_transform(X_train)
  X_{test\_scaled} = scaler.transform(X_{test})
  #saving the model
  file = open('standardScalar.pkl','wb')
  pickle.dump(scaler,file)
  file.close()
  return X_train_scaled, X_test_scaled
X\_train\_scaled, \ X\_test\_scaled = scaler\_standard(X\_train, \ X\_test)
X_train_scaled
```

log_reg = LogisticRegression()

log_reg.fit(X_train_scaled,y_train)