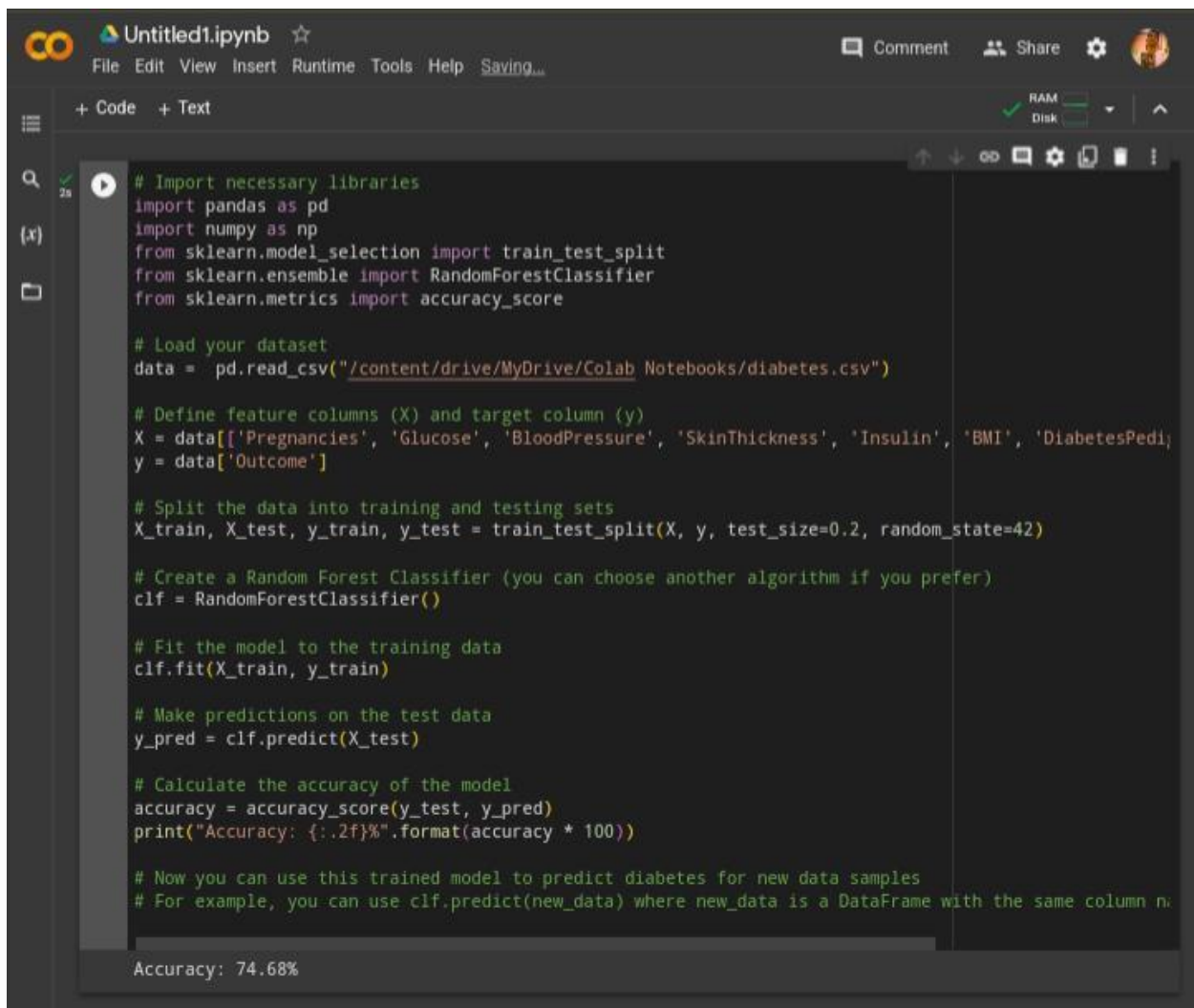


AI based diabetes prediction system

Overview

This Python code is designed to create an AI-based diabetes prediction system using a dataset with the following columns: 'Pregnancies', 'Glucose', 'BloodPressure', 'SkinThickness', 'Insulin', 'BMI', 'DiabetesPedigreeFunction', 'Age', and 'Outcome'. It employs the Random Forest Classifier for prediction.



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# Import necessary libraries
import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score

# Load your dataset
data = pd.read_csv("/content/drive/MyDrive/Colab Notebooks/diabetes.csv")

# Define feature columns (X) and target column (y)
X = data[['Pregnancies', 'Glucose', 'BloodPressure', 'SkinThickness', 'Insulin', 'BMI', 'DiabetesPedi
y = data['Outcome']

# Split the data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)

# Create a Random Forest Classifier (you can choose another algorithm if you prefer)
clf = RandomForestClassifier()

# Fit the model to the training data
clf.fit(X_train, y_train)

# Make predictions on the test data
y_pred = clf.predict(X_test)

# Calculate the accuracy of the model
accuracy = accuracy_score(y_test, y_pred)
print("Accuracy: {:.2f}%".format(accuracy * 100))

# Now you can use this trained model to predict diabetes for new data samples
# For example, you can use clf.predict(new_data) where new_data is a DataFrame with the same column n

Accuracy: 74.68%
```

Prerequisites

Before using this code, ensure that you have the following libraries installed:

- pandas

- numpy

- scikit-learn

1. Load Dataset

**- The code begins by loading your dataset from a CSV file. Replace
"your_dataset.csv" with the path to your dataset file.**

Load your dataset

Dataset_path = "/content/drive/MyDrive/Colab Notebooks/diabetes.csv"

Data = pd.read_csv(dataset_path)

2. Define Feature and Target Columns

**- Specify the feature columns (X) and the target column (y) based on your
dataset's column names.**

Define feature columns (X) and target column (y)

**X = data[['Pregnancies', 'Glucose', 'BloodPressure', 'SkinThickness', 'Insulin',
'BMI', 'DiabetesPedigreeFunction', 'Age']]**

Y = data['Outcome']

3. Split Data

- The data is split into training and testing sets using `train test split`. Adjust the `test size` and `random state` parameters as needed.

Split the data into training and testing sets

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)

4. Model Creation

- A Random Forest Classifier is used for building the prediction model. You can replace it with another machine learning algorithm of your choice.

Create a Random Forest Classifier (you can choose another algorithm if you prefer)

Clf = RandomForestClassifier()

5. Model Training

- The model is trained using the training data with `clf.fit(X_train, y_train)`.

Fit the model to the training data

Clf.fit(X_train, y_train)

6. Make Predictions

- Predictions are made on the test data with `clf.predict(X_test)`.

Make predictions on the test data

`Y_pred = clf.predict(X_test)`

7. Evaluate Model Performance

- The accuracy of the model is calculated and displayed. You can add more evaluation metrics as needed.

8. Prediction for New Data

- You can use the trained model to predict diabetes for new data samples by calling `clf.predict(new_data)` where `new_data` is a DataFrame with the same column names.

Calculate the accuracy of the model

`Accuracy = accuracy_score(y_test, y_pred)`

`Print("Accuracy: {:.2f}%".format(accuracy * 100))`

Create a simple bar plot to visualize accuracy

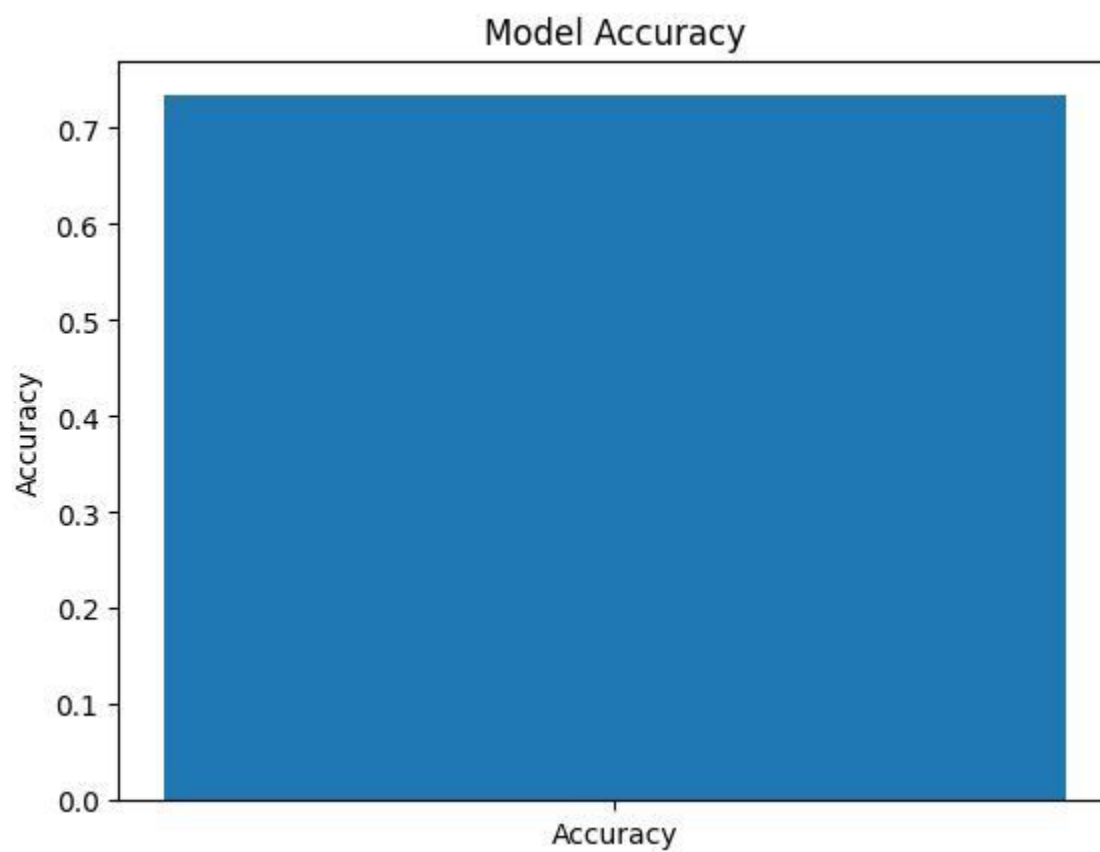
`Plt.bar(['Accuracy'], [accuracy])`

Plt.title('Model Accuracy')

Plt.ylabel('Accuracy')

Plt.show()

Output :



Accuracy :73.38%

Conclusion:

This code is a simplified example. In real-world applications, more comprehensive data preprocessing and model optimization would be necessary for building a robust diabetes prediction system.