



**Green University of Bangladesh**  
**Department of Computer Science and Engineering (CSE)**  
Faculty of Sciences and Engineering  
Semester: (Spring, Year:2025), B.Sc. in CSE (Day)

**Report Title: Diabetes Prediction using Linear Regression.**

**CourseTitle: Machine Learning Lab**  
**Course Code: CSE-412**                      **Section:221-D13**

**Student Details**

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**Submission Date: 10.07.2025**

**[For Teachers use only: **Don't Write Anything inside this box**]**

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<b>Comments:.....</b>	<b>Date:.....</b>

# TITLE OF THE LAB REPORT

Diabetes Prediction using Linear Regression

## 2. OBJECTIVES

The objective of learning linear regression is to develop an understanding of a fundamental statistical and machine learning technique used for predictive modeling and understanding the relationships between variables. Linear regression is a simple yet powerful method used in various fields, including statistics, economics, finance, and machine learning. Here are the primary objectives of learning linear regression:

- Understanding the Basics: Learn the fundamental concepts of linear regression, including the terminology (dependent and independent variables, coefficients, intercept, etc.) and the mathematical representation of linear regression models.
- Model Building: Learn how to build a linear regression model by selecting appropriate independent variables (features) and estimating coefficients that best fit the data.
- Interpretation: Develop the ability to interpret the coefficients of a linear regression model. Understand how changes in the independent variables affect the dependent variable.

## IMPLEMENTATION

### Code:

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
```

```
!git clone
https://github.com/ShawonTech/Machine-Learning-Lab/tree/main

dataset = pd.read_csv('/content/drive/MyDrive/diabetes.csv')

dataset.head()

dataset.plot(x='Glucose', y='BloodPressure', style='o')

plt.title('Glucose vs BloodPressure')

plt.xlabel('Glucose')

plt.ylabel('BloodPressure')

plt.show()

cols_with_zero_invalid = ['Glucose', 'BloodPressure',
                           'SkinThickness', 'Insulin', 'BMI']

dataset[cols_with_zero_invalid] =
dataset[cols_with_zero_invalid].replace(0, np.nan)

dataset[cols_with_zero_invalid] =
dataset[cols_with_zero_invalid].fillna(dataset[cols_with_zero_invalid]
].mean())

dataset.loc[0, 'Glucose'] = dataset['Glucose'].max()

dataset.loc[0, 'Glucose'] = dataset['Glucose'].max()

min_age = dataset['Age'].min()

min_glucose = dataset['Glucose'].min()

dataset.loc[dataset['Age'] == min_age, 'Glucose'] = min_glucose
```

```
X = dataset.drop('Outcome', axis=1).values
y = dataset['Outcome'].values

from sklearn.linear_model import LinearRegression
regressor = LinearRegression ()
regressor.fit(X_train , y_train)

y_pred_continuous = regressor.predict(X_test)
y_pred = np.round(y_pred_continuous).astype(int)

from sklearn import metrics

print('Mean Squared Error:', metrics.mean_squared_error(y_test ,
y_pred))

from sklearn.metrics import accuracy_score, confusion_matrix,
precision_score, recall_score, f1_score

accuracy = accuracy_score(y_test, y_pred)
conf_matrix = confusion_matrix(y_test, y_pred)
precision = precision_score(y_test, y_pred)
recall = recall_score(y_test, y_pred)
f1 = f1_score(y_test, y_pred)

print(f'Accuracy: {accuracy:.4f}')
print('Confusion Matrix:')
print(conf_matrix)
print(f'Precision: {precision:.4f}')
print(f'Recall: {recall:.4f}')
print(f'F1 Score: {f1:.4f}')
```

# OUTPUT

labreport.ipynb

File Edit View Insert Runtime Tools Help

Commands + Code + Text Run all

Files

- drive
- sample\_data

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
```

```
[3] dataset = pd.read_csv('/content/drive/MyDrive/diabetes.csv')
dataset.head()
```

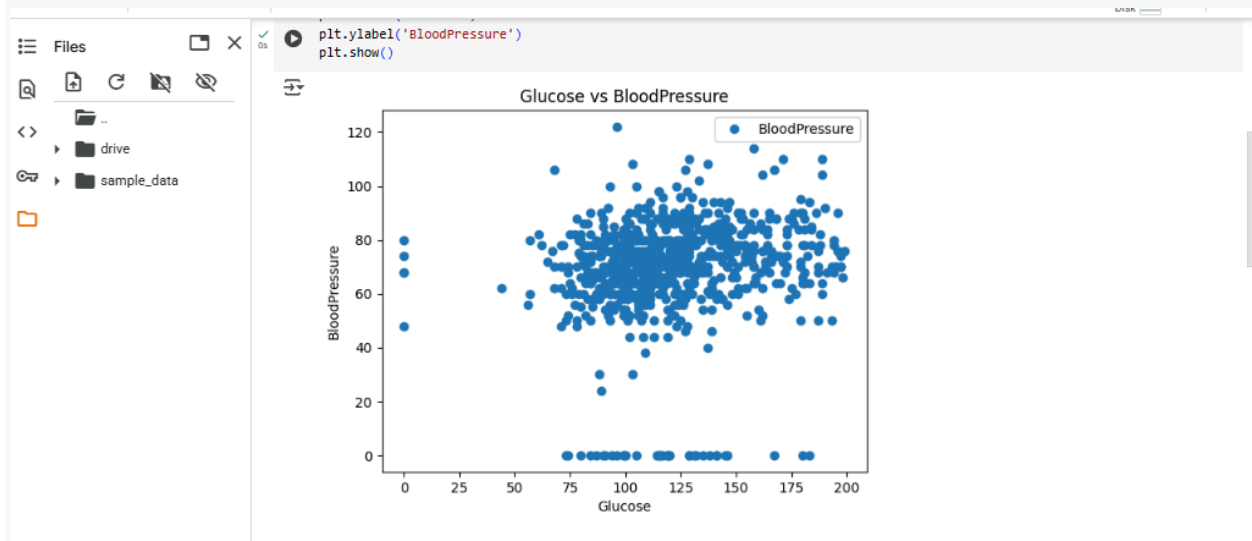
	Pregnancies	Glucose	BloodPressure	skinThickness	Insulin	BMI	DiabetesPedigreeFunction	Age	Outcome
0	6	148	72	35	0	33.6	0.627	50	1
1	1	85	66	29	0	26.6	0.351	31	0
2	8	183	64	0	0	23.3	0.672	32	1
3	1	89	66	23	94	28.1	0.167	21	0
4	0	137	40	35	168	43.1	2.288	33	1

Next steps: [Generate code with dataset](#) [View recommended plots](#) [New interactive sheet](#)

```
[4] dataset.plot(x='Glucose', y='BloodPressure', style='o')
plt.title('Glucose vs BloodPressure')
plt.xlabel('Glucose')
plt.ylabel('BloodPressure')
plt.show()
```

Glucose vs BloodPressure

8:18 AM Python 3



```
[22] from sklearn.linear_model import LinearRegression
regressor = LinearRegression()
regressor.fit(X_train, y_train)

[23] y_pred_continuous = regressor.predict(X_test)
y_pred = np.round(y_pred_continuous).astype(int)
```

```
from sklearn.metrics import accuracy_score, confusion_matrix, precision_score, recall_score, f1_score

accuracy = accuracy_score(y_test, y_pred)
conf_matrix = confusion_matrix(y_test, y_pred)
precision = precision_score(y_test, y_pred)
recall = recall_score(y_test, y_pred)
f1 = f1_score(y_test, y_pred)

print(f'Accuracy: {accuracy:.4f}')
print('Confusion Matrix:')
print(conf_matrix)
print(f'Precision: {precision:.4f}')
print(f'Recall: {recall:.4f}')
print(f'F1 Score: {f1:.4f}')
```

Accuracy: 0.8117  
Confusion Matrix:  
[[97 10]  
 [19 28]]  
Precision: 0.7368  
Recall: 0.5957  
F1 Score: 0.6588

## Analysis and Discussion

Linear Regression can be used for binary prediction in a very basic way, but it is not ideal for classification problems.

For a real-world diabetes prediction model, it's better to use Logistic Regression or other classification algorithms that are built for this purpose

Despite limitations, the experiment provides valuable insights into model behavior and importance of preprocessing.

**GitHub Link:**

<https://github.com/ShawonTech/Machine-Learning-Lab/tree/main>