

# untitled2

March 28, 2025

```
[1]: #NAME: SHIVANI GADKARI  
#ROLL NO.: 13342
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```
[4]: import pandas as pd  
from sklearn.model_selection import train_test_split  
from sklearn.linear_model import LogisticRegression  
from sklearn.preprocessing import StandardScaler  
from sklearn.metrics import classification_report, accuracy_score
```

```
[5]: df = pd.read_csv('diabetes.csv')
```

```
[6]: df
```

```
[6]:      Pregnancies  Glucose  BloodPressure  SkinThickness  Insulin   BMI  \  
0              6     148             72           35         0  33.6  
1              1      85             66           29         0  26.6  
2              8     183             64            0         0  23.3  
3              1      89             66           23        94  28.1  
4              0     137             40           35       168  43.1  
..          ...      ...             ...           ...         ...  ...  
763            10     101             76           48       180  32.9  
764              2     122             70           27         0  36.8  
765              5     121             72           23       112  26.2  
766              1     126             60            0         0  30.1  
767              1      93             70           31         0  30.4
```

```
      DiabetesPedigreeFunction  Age  Outcome  
0              0.627     50         1  
1              0.351     31         0  
2              0.672     32         1  
3              0.167     21         0  
4              2.288     33         1  
..          ...      ...             ...  
763            0.171     63         0  
764            0.340     27         0  
765            0.245     30         0  
766            0.349     47         1
```

```
767          0.315    23          0
```

```
[768 rows x 9 columns]
```

```
[7]: df.shape
```

```
[7]: (768, 9)
```

```
[8]: df.head
```

```
[8]: <bound method NDFrame.head of
SkinThickness  Insulin  BMI  \
0              6     148      72      35      0  33.6
1              1      85      66      29      0  26.6
2              8     183      64       0      0  23.3
3              1      89      66      23     94  28.1
4              0     137      40      35    168  43.1
..          ...     ...      ...      ...      ...
763            10     101      76      48    180  32.9
764             2     122      70      27      0  36.8
765             5     121      72      23    112  26.2
766             1     126      60       0      0  30.1
767             1      93      70      31      0  30.4
```

```
DiabetesPedigreeFunction  Age  Outcome
0              0.627    50          1
1              0.351    31          0
2              0.672    32          1
3              0.167    21          0
4              2.288    33          1
..          ...     ...      ...
763            0.171    63          0
764            0.340    27          0
765            0.245    30          0
766            0.349    47          1
767            0.315    23          0
```

```
[768 rows x 9 columns]>
```

```
[10]: df.isnull().sum()
```

```
[10]: Pregnancies      0
Glucose              0
BloodPressure        0
SkinThickness        0
Insulin              0
BMI                  0
```

```
DiabetesPedigreeFunction    0
Age                        0
Outcome                    0
dtype: int64
```

```
[12]: X = df.drop(columns=['Outcome'])
      y = df['Outcome']
```

```
[13]: #Defining X_train, X_test, etc
      from sklearn.model_selection import train_test_split
      X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
      ↪ random_state=42)
```

```
[14]: #Model is trained using GaussianNB:
      from sklearn.naive_bayes import GaussianNB
      gaussian = GaussianNB()
      gaussian.fit(X_train, y_train)
```

```
[14]: GaussianNB()
```

```
[23]: #Making predictions:
      y_pred_all = model.predict(X_test_scaled)
      probabilities_all = model.predict_proba(X_test_scaled)
```

```
[24]: #Printing prediction and probabilities:
      print("\nPredictions for all test samples:")
      for i, (prediction, prob) in enumerate(zip(y_pred_all, probabilities_all)): #
      ↪ all predictcts probablites for each sample
          print(f"Sample {i+1}: Predicted Outcome = {prediction}, P(No) = {prob[0]:.
      ↪ 4f}, P(Yes) = {prob[1]:.4f}")
```

Predictions for all test samples:

```
Sample 1: Predicted Outcome = 0, P(No) = 0.7226, P(Yes) = 0.2774
Sample 2: Predicted Outcome = 0, P(No) = 0.8140, P(Yes) = 0.1860
Sample 3: Predicted Outcome = 0, P(No) = 0.8853, P(Yes) = 0.1147
Sample 4: Predicted Outcome = 0, P(No) = 0.8286, P(Yes) = 0.1714
Sample 5: Predicted Outcome = 0, P(No) = 0.5316, P(Yes) = 0.4684
Sample 6: Predicted Outcome = 0, P(No) = 0.5601, P(Yes) = 0.4399
Sample 7: Predicted Outcome = 0, P(No) = 0.9838, P(Yes) = 0.0162
Sample 8: Predicted Outcome = 1, P(No) = 0.3404, P(Yes) = 0.6596
Sample 9: Predicted Outcome = 1, P(No) = 0.4638, P(Yes) = 0.5362
Sample 10: Predicted Outcome = 1, P(No) = 0.2294, P(Yes) = 0.7706
Sample 11: Predicted Outcome = 0, P(No) = 0.7461, P(Yes) = 0.2539
Sample 12: Predicted Outcome = 1, P(No) = 0.1086, P(Yes) = 0.8914
Sample 13: Predicted Outcome = 0, P(No) = 0.6671, P(Yes) = 0.3329
Sample 14: Predicted Outcome = 0, P(No) = 0.6871, P(Yes) = 0.3129
```

Sample 15: Predicted Outcome = 0, P(No) = 0.9147, P(Yes) = 0.0853  
 Sample 16: Predicted Outcome = 0, P(No) = 0.6137, P(Yes) = 0.3863  
 Sample 17: Predicted Outcome = 0, P(No) = 0.8606, P(Yes) = 0.1394  
 Sample 18: Predicted Outcome = 0, P(No) = 0.9235, P(Yes) = 0.0765  
 Sample 19: Predicted Outcome = 1, P(No) = 0.1422, P(Yes) = 0.8578  
 Sample 20: Predicted Outcome = 1, P(No) = 0.4448, P(Yes) = 0.5552  
 Sample 21: Predicted Outcome = 0, P(No) = 0.7916, P(Yes) = 0.2084  
 Sample 22: Predicted Outcome = 0, P(No) = 0.9209, P(Yes) = 0.0791  
 Sample 23: Predicted Outcome = 1, P(No) = 0.4688, P(Yes) = 0.5312  
 Sample 24: Predicted Outcome = 0, P(No) = 0.9046, P(Yes) = 0.0954  
 Sample 25: Predicted Outcome = 1, P(No) = 0.4650, P(Yes) = 0.5350  
 Sample 26: Predicted Outcome = 1, P(No) = 0.1201, P(Yes) = 0.8799  
 Sample 27: Predicted Outcome = 0, P(No) = 0.8755, P(Yes) = 0.1245  
 Sample 28: Predicted Outcome = 0, P(No) = 0.9689, P(Yes) = 0.0311  
 Sample 29: Predicted Outcome = 0, P(No) = 0.7520, P(Yes) = 0.2480  
 Sample 30: Predicted Outcome = 0, P(No) = 0.8805, P(Yes) = 0.1195  
 Sample 31: Predicted Outcome = 1, P(No) = 0.0928, P(Yes) = 0.9072  
 Sample 32: Predicted Outcome = 1, P(No) = 0.1327, P(Yes) = 0.8673  
 Sample 33: Predicted Outcome = 1, P(No) = 0.2423, P(Yes) = 0.7577  
 Sample 34: Predicted Outcome = 1, P(No) = 0.1670, P(Yes) = 0.8330  
 Sample 35: Predicted Outcome = 1, P(No) = 0.3795, P(Yes) = 0.6205  
 Sample 36: Predicted Outcome = 1, P(No) = 0.3141, P(Yes) = 0.6859  
 Sample 37: Predicted Outcome = 1, P(No) = 0.0346, P(Yes) = 0.9654  
 Sample 38: Predicted Outcome = 0, P(No) = 0.7502, P(Yes) = 0.2498  
 Sample 39: Predicted Outcome = 1, P(No) = 0.4913, P(Yes) = 0.5087  
 Sample 40: Predicted Outcome = 1, P(No) = 0.2764, P(Yes) = 0.7236  
 Sample 41: Predicted Outcome = 0, P(No) = 0.9288, P(Yes) = 0.0712  
 Sample 42: Predicted Outcome = 1, P(No) = 0.4057, P(Yes) = 0.5943  
 Sample 43: Predicted Outcome = 1, P(No) = 0.4182, P(Yes) = 0.5818  
 Sample 44: Predicted Outcome = 0, P(No) = 0.6694, P(Yes) = 0.3306  
 Sample 45: Predicted Outcome = 0, P(No) = 0.9712, P(Yes) = 0.0288  
 Sample 46: Predicted Outcome = 1, P(No) = 0.4930, P(Yes) = 0.5070  
 Sample 47: Predicted Outcome = 1, P(No) = 0.3498, P(Yes) = 0.6502  
 Sample 48: Predicted Outcome = 0, P(No) = 0.7749, P(Yes) = 0.2251  
 Sample 49: Predicted Outcome = 0, P(No) = 0.6303, P(Yes) = 0.3697  
 Sample 50: Predicted Outcome = 1, P(No) = 0.0442, P(Yes) = 0.9558  
 Sample 51: Predicted Outcome = 0, P(No) = 0.9504, P(Yes) = 0.0496  
 Sample 52: Predicted Outcome = 1, P(No) = 0.3362, P(Yes) = 0.6638  
 Sample 53: Predicted Outcome = 1, P(No) = 0.1903, P(Yes) = 0.8097  
 Sample 54: Predicted Outcome = 0, P(No) = 0.7519, P(Yes) = 0.2481  
 Sample 55: Predicted Outcome = 0, P(No) = 0.9057, P(Yes) = 0.0943  
 Sample 56: Predicted Outcome = 0, P(No) = 0.9573, P(Yes) = 0.0427  
 Sample 57: Predicted Outcome = 1, P(No) = 0.2290, P(Yes) = 0.7710  
 Sample 58: Predicted Outcome = 0, P(No) = 0.9940, P(Yes) = 0.0060  
 Sample 59: Predicted Outcome = 0, P(No) = 0.5870, P(Yes) = 0.4130  
 Sample 60: Predicted Outcome = 1, P(No) = 0.2525, P(Yes) = 0.7475  
 Sample 61: Predicted Outcome = 1, P(No) = 0.2643, P(Yes) = 0.7357  
 Sample 62: Predicted Outcome = 0, P(No) = 0.6519, P(Yes) = 0.3481

Sample 63: Predicted Outcome = 0, P(No) = 0.7994, P(Yes) = 0.2006  
 Sample 64: Predicted Outcome = 0, P(No) = 0.7839, P(Yes) = 0.2161  
 Sample 65: Predicted Outcome = 0, P(No) = 0.9221, P(Yes) = 0.0779  
 Sample 66: Predicted Outcome = 1, P(No) = 0.3828, P(Yes) = 0.6172  
 Sample 67: Predicted Outcome = 0, P(No) = 0.9480, P(Yes) = 0.0520  
 Sample 68: Predicted Outcome = 1, P(No) = 0.2740, P(Yes) = 0.7260  
 Sample 69: Predicted Outcome = 0, P(No) = 0.9610, P(Yes) = 0.0390  
 Sample 70: Predicted Outcome = 1, P(No) = 0.2848, P(Yes) = 0.7152  
 Sample 71: Predicted Outcome = 1, P(No) = 0.3241, P(Yes) = 0.6759  
 Sample 72: Predicted Outcome = 0, P(No) = 0.9280, P(Yes) = 0.0720  
 Sample 73: Predicted Outcome = 0, P(No) = 0.8171, P(Yes) = 0.1829  
 Sample 74: Predicted Outcome = 0, P(No) = 0.8851, P(Yes) = 0.1149  
 Sample 75: Predicted Outcome = 0, P(No) = 0.9072, P(Yes) = 0.0928  
 Sample 76: Predicted Outcome = 1, P(No) = 0.4851, P(Yes) = 0.5149  
 Sample 77: Predicted Outcome = 0, P(No) = 0.8318, P(Yes) = 0.1682  
 Sample 78: Predicted Outcome = 0, P(No) = 0.8567, P(Yes) = 0.1433  
 Sample 79: Predicted Outcome = 0, P(No) = 0.8659, P(Yes) = 0.1341  
 Sample 80: Predicted Outcome = 0, P(No) = 0.7662, P(Yes) = 0.2338  
 Sample 81: Predicted Outcome = 1, P(No) = 0.3496, P(Yes) = 0.6504  
 Sample 82: Predicted Outcome = 0, P(No) = 0.8517, P(Yes) = 0.1483  
 Sample 83: Predicted Outcome = 0, P(No) = 0.9369, P(Yes) = 0.0631  
 Sample 84: Predicted Outcome = 0, P(No) = 0.6293, P(Yes) = 0.3707  
 Sample 85: Predicted Outcome = 0, P(No) = 0.7380, P(Yes) = 0.2620  
 Sample 86: Predicted Outcome = 1, P(No) = 0.1697, P(Yes) = 0.8303  
 Sample 87: Predicted Outcome = 1, P(No) = 0.1032, P(Yes) = 0.8968  
 Sample 88: Predicted Outcome = 0, P(No) = 0.6902, P(Yes) = 0.3098  
 Sample 89: Predicted Outcome = 0, P(No) = 0.8759, P(Yes) = 0.1241  
 Sample 90: Predicted Outcome = 0, P(No) = 0.9142, P(Yes) = 0.0858  
 Sample 91: Predicted Outcome = 0, P(No) = 0.9324, P(Yes) = 0.0676  
 Sample 92: Predicted Outcome = 0, P(No) = 0.7634, P(Yes) = 0.2366  
 Sample 93: Predicted Outcome = 0, P(No) = 0.9956, P(Yes) = 0.0044  
 Sample 94: Predicted Outcome = 1, P(No) = 0.4601, P(Yes) = 0.5399  
 Sample 95: Predicted Outcome = 1, P(No) = 0.4874, P(Yes) = 0.5126  
 Sample 96: Predicted Outcome = 1, P(No) = 0.3556, P(Yes) = 0.6444  
 Sample 97: Predicted Outcome = 0, P(No) = 0.6315, P(Yes) = 0.3685  
 Sample 98: Predicted Outcome = 0, P(No) = 0.8686, P(Yes) = 0.1314  
 Sample 99: Predicted Outcome = 1, P(No) = 0.3873, P(Yes) = 0.6127  
 Sample 100: Predicted Outcome = 0, P(No) = 0.9150, P(Yes) = 0.0850  
 Sample 101: Predicted Outcome = 1, P(No) = 0.2838, P(Yes) = 0.7162  
 Sample 102: Predicted Outcome = 0, P(No) = 0.9357, P(Yes) = 0.0643  
 Sample 103: Predicted Outcome = 1, P(No) = 0.2236, P(Yes) = 0.7764  
 Sample 104: Predicted Outcome = 0, P(No) = 0.5024, P(Yes) = 0.4976  
 Sample 105: Predicted Outcome = 1, P(No) = 0.3566, P(Yes) = 0.6434  
 Sample 106: Predicted Outcome = 0, P(No) = 0.7817, P(Yes) = 0.2183  
 Sample 107: Predicted Outcome = 0, P(No) = 0.7393, P(Yes) = 0.2607  
 Sample 108: Predicted Outcome = 1, P(No) = 0.2634, P(Yes) = 0.7366  
 Sample 109: Predicted Outcome = 0, P(No) = 0.8639, P(Yes) = 0.1361  
 Sample 110: Predicted Outcome = 1, P(No) = 0.4573, P(Yes) = 0.5427

Sample 111: Predicted Outcome = 0, P(No) = 0.8995, P(Yes) = 0.1005  
 Sample 112: Predicted Outcome = 0, P(No) = 0.6827, P(Yes) = 0.3173  
 Sample 113: Predicted Outcome = 0, P(No) = 0.9792, P(Yes) = 0.0208  
 Sample 114: Predicted Outcome = 1, P(No) = 0.2718, P(Yes) = 0.7282  
 Sample 115: Predicted Outcome = 0, P(No) = 0.8227, P(Yes) = 0.1773  
 Sample 116: Predicted Outcome = 0, P(No) = 0.6631, P(Yes) = 0.3369  
 Sample 117: Predicted Outcome = 1, P(No) = 0.2327, P(Yes) = 0.7673  
 Sample 118: Predicted Outcome = 0, P(No) = 0.7785, P(Yes) = 0.2215  
 Sample 119: Predicted Outcome = 0, P(No) = 0.9340, P(Yes) = 0.0660  
 Sample 120: Predicted Outcome = 1, P(No) = 0.4278, P(Yes) = 0.5722  
 Sample 121: Predicted Outcome = 0, P(No) = 0.9406, P(Yes) = 0.0594  
 Sample 122: Predicted Outcome = 0, P(No) = 0.7023, P(Yes) = 0.2977  
 Sample 123: Predicted Outcome = 0, P(No) = 0.7601, P(Yes) = 0.2399  
 Sample 124: Predicted Outcome = 0, P(No) = 0.9167, P(Yes) = 0.0833  
 Sample 125: Predicted Outcome = 0, P(No) = 0.7383, P(Yes) = 0.2617  
 Sample 126: Predicted Outcome = 0, P(No) = 0.5830, P(Yes) = 0.4170  
 Sample 127: Predicted Outcome = 0, P(No) = 0.9584, P(Yes) = 0.0416  
 Sample 128: Predicted Outcome = 1, P(No) = 0.1257, P(Yes) = 0.8743  
 Sample 129: Predicted Outcome = 1, P(No) = 0.0310, P(Yes) = 0.9690  
 Sample 130: Predicted Outcome = 1, P(No) = 0.2722, P(Yes) = 0.7278  
 Sample 131: Predicted Outcome = 1, P(No) = 0.2990, P(Yes) = 0.7010  
 Sample 132: Predicted Outcome = 1, P(No) = 0.1497, P(Yes) = 0.8503  
 Sample 133: Predicted Outcome = 0, P(No) = 0.9101, P(Yes) = 0.0899  
 Sample 134: Predicted Outcome = 0, P(No) = 0.5818, P(Yes) = 0.4182  
 Sample 135: Predicted Outcome = 1, P(No) = 0.1852, P(Yes) = 0.8148  
 Sample 136: Predicted Outcome = 0, P(No) = 0.8877, P(Yes) = 0.1123  
 Sample 137: Predicted Outcome = 0, P(No) = 0.8439, P(Yes) = 0.1561  
 Sample 138: Predicted Outcome = 1, P(No) = 0.1501, P(Yes) = 0.8499  
 Sample 139: Predicted Outcome = 1, P(No) = 0.2008, P(Yes) = 0.7992  
 Sample 140: Predicted Outcome = 0, P(No) = 0.9863, P(Yes) = 0.0137  
 Sample 141: Predicted Outcome = 0, P(No) = 0.9041, P(Yes) = 0.0959  
 Sample 142: Predicted Outcome = 0, P(No) = 0.9592, P(Yes) = 0.0408  
 Sample 143: Predicted Outcome = 0, P(No) = 0.7986, P(Yes) = 0.2014  
 Sample 144: Predicted Outcome = 0, P(No) = 0.6218, P(Yes) = 0.3782  
 Sample 145: Predicted Outcome = 0, P(No) = 0.8754, P(Yes) = 0.1246  
 Sample 146: Predicted Outcome = 0, P(No) = 0.7079, P(Yes) = 0.2921  
 Sample 147: Predicted Outcome = 0, P(No) = 0.8504, P(Yes) = 0.1496  
 Sample 148: Predicted Outcome = 0, P(No) = 0.9778, P(Yes) = 0.0222  
 Sample 149: Predicted Outcome = 0, P(No) = 0.5766, P(Yes) = 0.4234  
 Sample 150: Predicted Outcome = 1, P(No) = 0.2542, P(Yes) = 0.7458  
 Sample 151: Predicted Outcome = 0, P(No) = 0.8739, P(Yes) = 0.1261  
 Sample 152: Predicted Outcome = 0, P(No) = 0.5135, P(Yes) = 0.4865  
 Sample 153: Predicted Outcome = 0, P(No) = 0.7735, P(Yes) = 0.2265  
 Sample 154: Predicted Outcome = 0, P(No) = 0.7951, P(Yes) = 0.2049

```

[25]: #Model Evaluation:
      print("\nModel Evaluation:")
  
```

```
print("Accuracy:", accuracy_score(y_test, y_pred_all))
```

Model Evaluation:

Accuracy: 0.7532467532467533

```
[27]: #Classification Report:
print("\nClassification Report:") # Genrates precision , recall,,f-1 score(mean
    ↪ of precison & recall) and
                                     # support(no of occurence of each sample in
    ↪ dataset)
print(classification_report(y_test, y_pred_all))
```

Classification Report:

	precision	recall	f1-score	support
0	0.81	0.80	0.81	99
1	0.65	0.67	0.66	55
accuracy			0.75	154
macro avg	0.73	0.74	0.73	154
weighted avg	0.76	0.75	0.75	154

```
[28]: #creating sample datapoints:
sample_data = pd.DataFrame({
    'Pregnancies': [1],
    'Glucose': [85],
    'BloodPressure': [66],
    'SkinThickness': [29],
    'Insulin': [0],
    'BMI': [26.6],
    'DiabetesPedigreeFunction': [0.351],
    'Age': [50]
})
```

```
[29]: #Scalimng sample data:
sample_scaled = scaler.transform(sample_data)
```

```
[30]: #rediction making for sample data:
sample_prediction = model.predict(sample_scaled)
```

```
[31]: #Getting prediction probablity for it:
sample_probability = model.predict_proba(sample_scaled)
```

```
[32]: #Printing Prediction & Probability:
print("\nPrediction for the sample data:")
print(f"Predicted Outcome: {sample_prediction[0]}")
print(f"Probabilities- P(No): {sample_probability[0][0]:.4f}, P(Yes):_
↪{sample_probability[0][1]:.4f}")
```

```
Prediction for the sample data:
Predicted Outcome: 0
Probabilities- P(No): 0.8856, P(Yes): 0.1144
```

```
[ ]:
```