\blacksquare

th

0

1

0.627

0.351

0.672

0.167

2.288

New interactive sheet

50

31

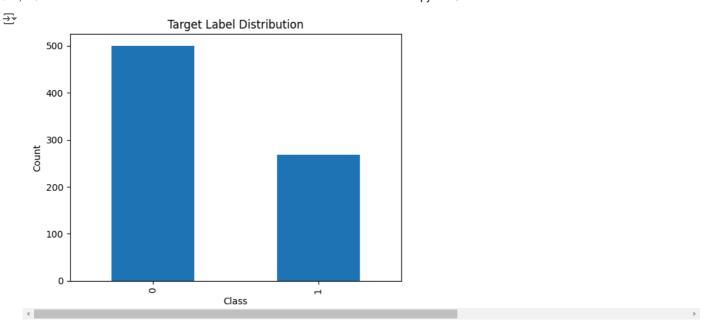
32

21

33

plt.xlabel("Class") plt.ylabel("Count") plt.show()

```
import numpy as np
import pandas as pd
from sklearn.preprocessing import StandardScaler
from sklearn.model_selection import train_test_split
from sklearn import svm
from sklearn.metrics import accuracy_score
import matplotlib.pyplot as plt
diabetese_dataset = pd.read_csv('/content/diabetes.csv')
diabetese_dataset.head()
₹
        Pregnancies Glucose BloodPressure SkinThickness Insulin BMI DiabetesPedigreeFunction Age Outcome
      0
                                         72
                   6
                         148
                                                        35
                                                                  0 33.6
                                                        29
      1
                   1
                          85
                                         66
                                                                  0 26.6
      2
                  8
                         183
                                         64
                                                         0
                                                                  0 23.3
                          89
                                         66
                                                        23
                                                                 94 28.1
      4
                   0
                         137
                                         40
                                                        35
                                                                168 43.1
             Generate code with diabetese_dataset
                                                     View recommended plots
 Next steps:
diabetese dataset.shape
→ (768, 9)
# Check the unique values in the last column (target column)
unique_values = diabetese_dataset['Outcome'].unique()
print("Unique target values:", unique_values)
→ Unique target values: [1 0]
# Check data type and number of unique values in the target column
target = diabetese_dataset['Outcome']
print("Data type of target:", target.dtype)
print("Number of unique values:", target.nunique())
# Determine if classification or regression
if target.nunique() < 20 and target.dtype == 'object':</pre>
   print("This is a classification problem.")
   print("This is a regression problem.")
    Data type of target: int64
     Number of unique values: 2
     This is a regression problem.
# Plot the distribution of target labels
target.value_counts().plot(kind='bar')
plt.title("Target Label Distribution")
```



diabetese_dataset["Outcome"].value_counts()



diabetese_dataset.groupby('Outcome').mean()

₹		Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	DiabetesPedigreeFunction	Age	
	Outcome									11.
	0	3.298000	109.980000	68.184000	19.664000	68.792000	30.304200	0.429734	31.190000	
	1	4.865672	141.257463	70.824627	22.164179	100.335821	35.142537	0.550500	37.067164	
	4									>

Separating the data and variables

X = diabetese_dataset.drop(columns='Outcome', axis = 1)

Y = diabetese_dataset['Outcome']

print(X)

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

[768 rows x 8 columns]

print(Y)

```
0
    2
           1
    3
           0
    4
           1
    763
           0
    764
           a
    765
           0
    766
           1
    767
           0
    Name: Outcome, Length: 768, dtype: int64
# Data Standardization
scaler = StandardScaler()
scaler.fit(X)
     ▼ StandardScaler ① ?
     StandardScaler()
standardized_data = scaler.transform(X)
display(standardized_data)
⇒ array([[ 0.63994726, 0.84832379, 0.14964075, ..., 0.20401277,
             0.46849198, 1.4259954 ],
            [-0.84488505, -1.12339636, -0.16054575, \ldots, -0.68442195,
           -0.36506078, -0.19067191],
[ 1.23388019, 1.94372388, -0.26394125, ..., -1.10325546,
             0.60439732, -0.10558415],
           [ 0.3429808 , 0.00330087, 0.14964075, ..., -0.73518964, -0.68519336, -0.27575966], [-0.84488505, 0.1597866 , -0.47073225, ..., -0.24020459,
            -0.37110101, 1.17073215],
           [-0.84488505, -0.8730192], 0.04624525, ..., -0.20212881, -0.47378505 -0.87137393111
             0 47378505
    4
\ensuremath{\text{\#}} again we separate the data after standardization
X = standardized_data
Y = diabetese_dataset['Outcome']
print(X)
print(Y)
→ [[ 0.63994726 0.84832379 0.14964075 ... 0.20401277 0.46849198
       1,4259954 1
      [-0.84488505 -1.12339636 -0.16054575 ... -0.68442195 -0.36506078
      -0.19067191]
      -0.10558415]
     -0.27575966]
      1.17073215]
     -0.87137393]]
    0
           1
    1
           0
    2
           1
    3
           0
    4
           1
    763
           0
    764
    765
           0
    766
           1
    767
           0
    Name: Outcome, Length: 768, dtype: int64
# train test split
X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.2, stratify=Y, random_state=2)
\label{eq:continuous_print}  \text{print}(\texttt{"X=:", X.shape, "} \texttt{``nX\_train=:", X\_train.shape, "} \texttt{``nX\_test=:", X\_test.shape)} 
→ X= : (768, 8)
    X_train= : (614, 8)
    X_test= : (154, 8)
```

```
# Training the model
classifier = svm.SVC(kernel='linear')
# training the SVM
classifier.fit(X_train, Y_train)
            SVC
                    i ?
     SVC(kernel='linear')
     4
# Evaluating the model
X_train_prediction = classifier.predict(X_train)
training\_data\_accuracy = accuracy\_score(X\_train\_prediction, Y\_train)
print("Accuracy Score of the Training data : ", training_data_accuracy)
Accuracy Score of the Training data: 0.7866449511400652
X_test_prediction = classifier.predict(X_test)
tests\_data\_accuracy = accuracy\_score(X\_test\_prediction, Y\_test)
print("Accuracy Score of the Test data : ", tests_data_accuracy)
Accuracy Score of the Test data: 0.7727272727272727
# making the predictive system
# input_data = (4,110,92,0,0,37.6,0.191,30)
input_data = (1,189,60,23,846,30.1,0.398,59)
# convert to numpy array
input_data_as_numpy_array = np.asarray(input_data)
# reshape the array
input_data_reshaped = input_data_as_numpy_array.reshape(1, -1)
# standardize the data
std_data = scaler.transform(input_data_reshaped)
print(std_data)
prediction = classifier.predict(std_data)
print(prediction)
if (prediction[0] == 0):
 print("The Person is not diabetic")
else :
 print("The Person is diabetic")
 [[-0.84488505 2.13150675 -0.47073225 0.15453319 6.65283938 -0.24020459
       -0.2231152 2.19178518]]
     [1]
     The Person is diabetic
     /usr/local/lib/python3.10/dist-packages/sklearn/base.py:493: UserWarning: X does not have valid feature names, but StandardScaler wa
      warnings.warn(
```

Start coding or generate with AI.