

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score, classification_report,
confusion_matrix

# 1. Load the Data
# -----
try:
    df = pd.read_csv('heart_disease_data.csv')
    print("Data loaded successfully.")
    print(f"Dataset Shape: {df.shape}")
except FileNotFoundError:
    print("Error: The file 'heart_disease_data.csv' was not found.")
    exit()

# Display the first 5 rows
print("\n--- First 5 Rows ---")
print(df.head())

# Display basic statistical details
print("\n--- Data Description ---")
print(df.describe())

# 2. Data Visualization
# -----
# Set the style for seaborn
sns.set()

# Plot the distribution of the 'target' variable
plt.figure(figsize=(6, 4))
sns.countplot(x='target', data=df)
plt.title('Distribution of Heart Disease Presence (Target)')
plt.xlabel('Target (0 = Healthy, 1 = Disease)')
plt.ylabel('Count')
plt.savefig('target_distribution.png')
plt.show()

# Plot a Correlation Heatmap
plt.figure(figsize=(10, 8))
sns.heatmap(df.corr(), annot=True, cmap='coolwarm', fmt='.1f')
plt.title('Correlation Matrix')
plt.savefig('correlation_matrix.png')
plt.show()

# 3. Data Preprocessing & Splitting
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# Separating features (X) and target label (Y)
X = df.drop(columns='target', axis=1)
Y = df['target']

# Split data into Training and Testing sets (80% Train, 20% Test)
X_train, X_test, Y_train, Y_test = train_test_split(X, Y,
test_size=0.2, stratify=Y, random_state=2)

print(f"\nTraining Set Size: {X_train.shape}")
print(f"Testing Set Size: {X_test.shape}")

# 4. Model Training (Logistic Regression)
# -----
# Logistic Regression is used for binary classification problems
model = LogisticRegression(max_iter=1000)
model.fit(X_train, Y_train)

# 5. Model Evaluation
# -----
# Accuracy on training data
X_train_prediction = model.predict(X_train)
training_data_accuracy = accuracy_score(X_train_prediction, Y_train)

# Accuracy on test data
X_test_prediction = model.predict(X_test)
test_data_accuracy = accuracy_score(X_test_prediction, Y_test)

print("\n--- Model Performance ---")
print(f"Accuracy on Training data: {training_data_accuracy:.2%}")
print(f"Accuracy on Test data: {test_data_accuracy:.2%}")

print("\n--- Classification Report ---")
print(classification_report(Y_test, X_test_prediction))

print("\n--- Confusion Matrix ---")
print(confusion_matrix(Y_test, X_test_prediction))

Data loaded successfully.
Dataset Shape: (303, 14)

--- First 5 Rows ---
   age  sex  cp  trestbps  chol  fbs  ...  exang  oldpeak  slope  ca
thal  target
0    63     1    3        145   233    1  ...      0       2.3     0     0
1      1
1    37     1    2        130   250    0  ...      0       3.5     0     0
2      1
2    41     0    1        130   204    0  ...      0       1.4     2     0
2      1
3    56     1    1        120   236    0  ...      0       0.8     2     0

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2      1  
4    57    0    0     120   354    0    ...     1     0.6    2    0  
2      1
```

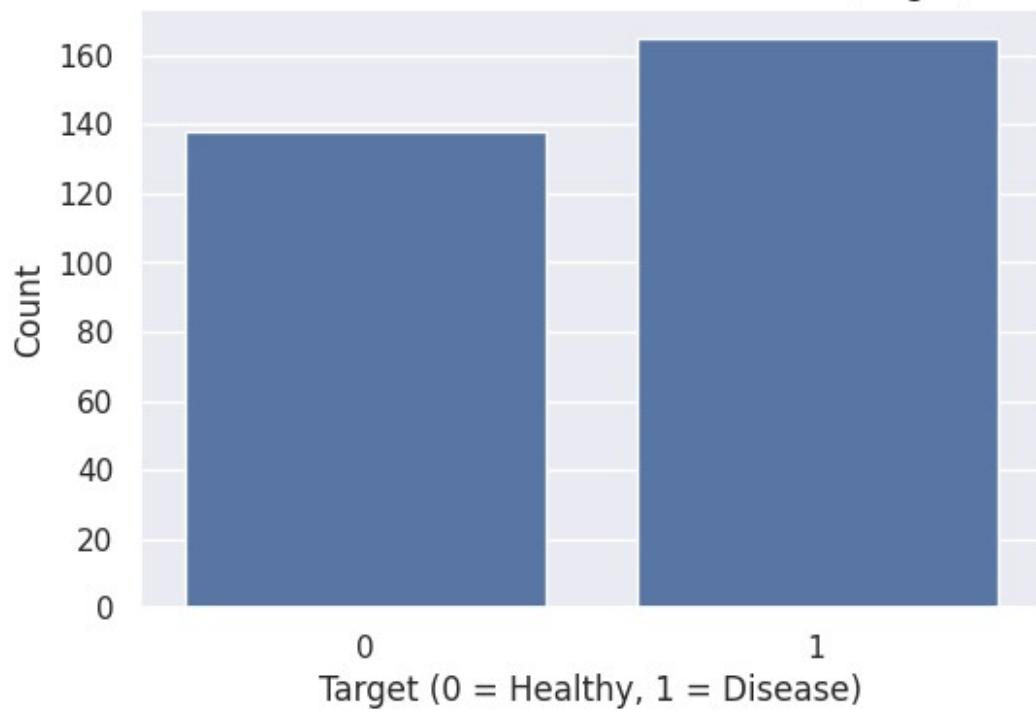
[5 rows x 14 columns]

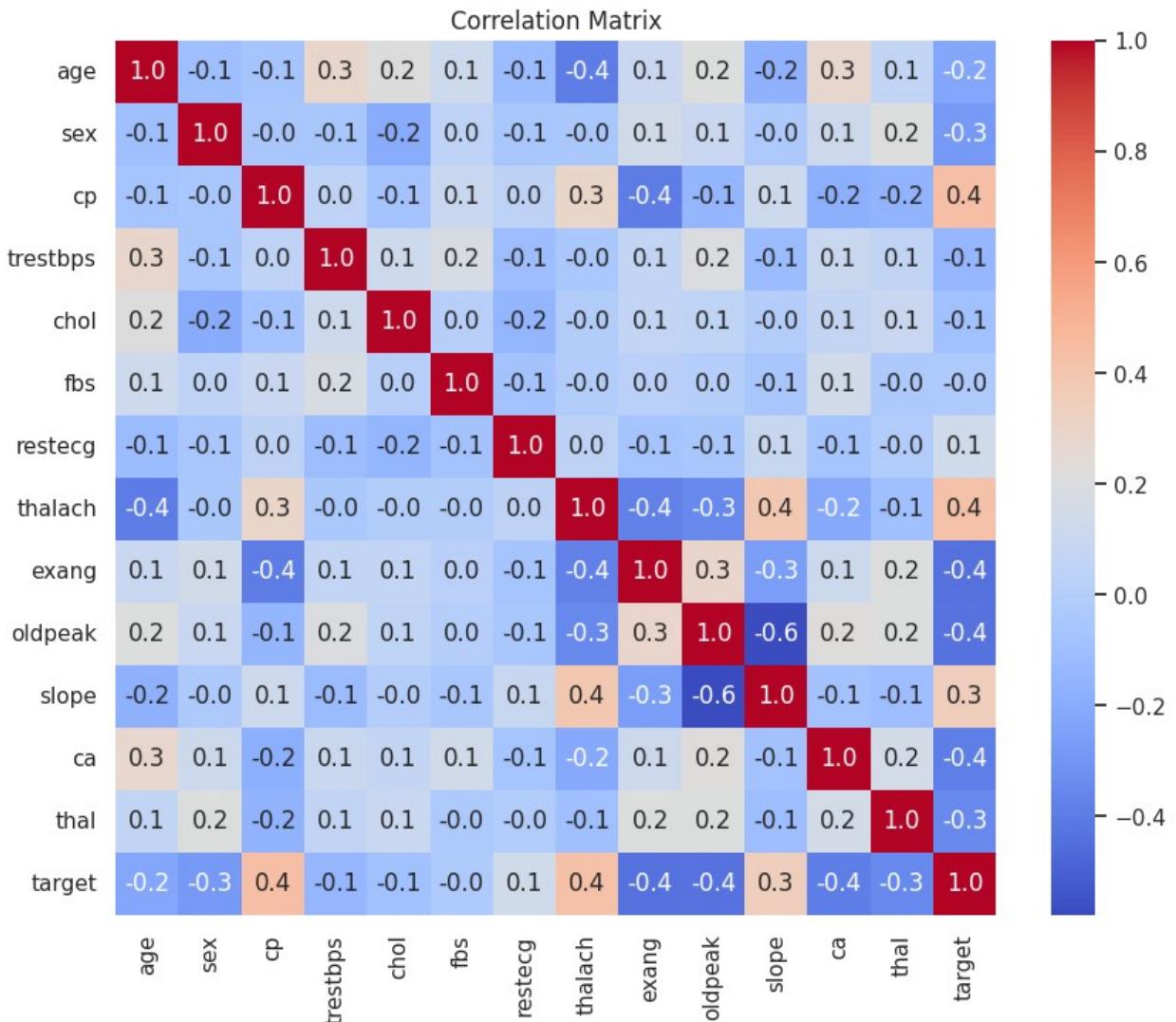
--- Data Description ---

	age	sex	cp	...	ca	thal
target						
count	303.000000	303.000000	303.000000	...	303.000000	303.000000
303.000000						
mean	54.366337	0.683168	0.966997	...	0.729373	2.313531
0.544554						
std	9.082101	0.466011	1.032052	...	1.022606	0.612277
0.498835						
min	29.000000	0.000000	0.000000	...	0.000000	0.000000
0.000000						
25%	47.500000	0.000000	0.000000	...	0.000000	2.000000
0.000000						
50%	55.000000	1.000000	1.000000	...	0.000000	2.000000
1.000000						
75%	61.000000	1.000000	2.000000	...	1.000000	3.000000
1.000000						
max	77.000000	1.000000	3.000000	...	4.000000	3.000000
1.000000						

[8 rows x 14 columns]

Distribution of Heart Disease Presence (Target)





Training Set Size: (242, 13)
Testing Set Size: (61, 13)

--- Model Performance ---
Accuracy on Training data: 85.54%
Accuracy on Test data: 80.33%

--- Classification Report ---

	precision	recall	f1-score	support
0	0.79	0.79	0.79	28
1	0.82	0.82	0.82	33
accuracy			0.80	61
macro avg	0.80	0.80	0.80	61
weighted avg	0.80	0.80	0.80	61

```
--- Confusion Matrix ---  
[[22  6]  
 [ 6 27]]
```