```
In [1]:
        import pandas as pd
        import seaborn as sns
         import matplotlib.pyplot as plt
        from sklearn.preprocessing import StandardScaler
        from sklearn.model_selection import train_test_split
        from sklearn.linear_model import LogisticRegression
        from sklearn.metrics import accuracy_score, classification_report
        from sklearn.metrics import confusion_matrix
         import warnings
        warnings.filterwarnings('ignore')
        df = pd.read_csv(r'D:\CSE ENGG NOTES\6th SEMESTER\ML LAB\Datasets\heart.csv')
In [2]:
        df.head()
In [3]:
Out[3]:
                         trestbps
                                   chol fbs
                                            restecg
                                                     thalach
                                                              exang
                                                                     oldpeak slope
                                                                                         thal
            age sex
                      ср
                                                                                     ca
         0
                                          0
                                                         168
                                                                                      2
                                                                                           3
             52
                   1
                       0
                              125
                                   212
                                                   1
                                                                  0
                                                                          1.0
                                                                                  2
                                   203
                                                  0
                                                         155
                                                                                      0
                                                                                           3
         1
             53
                   1
                       0
                              140
                                          1
                                                                  1
                                                                          3.1
                                                                                  0
         2
             70
                   1
                       0
                              145
                                   174
                                          0
                                                   1
                                                         125
                                                                  1
                                                                          2.6
                                                                                      0
                                                                                           3
                                                                                  0
                                                         161
                                                                          0.0
                                                                                           3
         3
             61
                   1
                       0
                              148
                                   203
                                          0
                                                                  0
                                                                                  2
                                                                                      1
                                                   1
                                                         106
                                                                  0
                                                                          1.9
                                                                                      3
                                                                                           2
             62
                   0
                       0
                              138
                                   294
                                          1
                                                                                  1
In [4]:
        df.shape
Out[4]: (1025, 14)
In [5]: df.info()
       <class 'pandas.core.frame.DataFrame'>
       RangeIndex: 1025 entries, 0 to 1024
       Data columns (total 14 columns):
        #
            Column
                      Non-Null Count Dtype
       ---
            ----
        0
                       1025 non-null
                                       int64
            age
                       1025 non-null
        1
            sex
                                       int64
        2
                      1025 non-null
                                       int64
            ср
        3
            trestbps 1025 non-null
                                       int64
        4
            chol
                      1025 non-null
                                       int64
        5
            fbs
                      1025 non-null
                                       int64
                      1025 non-null
                                       int64
        6
            restecg
        7
            thalach
                       1025 non-null
                                       int64
                                       int64
        8
            exang
                      1025 non-null
        9
            oldpeak
                      1025 non-null
                                       float64
                                       int64
        10 slope
                       1025 non-null
        11
                       1025 non-null
                                       int64
           ca
        12 thal
                       1025 non-null
                                       int64
        13 target
                      1025 non-null
                                       int64
       dtypes: float64(1), int64(13)
       memory usage: 112.2 KB
In [6]: df.isnull().sum()
```

```
Out[6]: age
                   0
        sex
                   0
        ср
                   0
        trestbps 0
        chol
        fbs
                   0
        restecg
                   0
        thalach
                   0
                   0
        exang
        oldpeak
                   0
        slope
                   0
        ca
                   0
                   0
        thal
                   0
        target
        dtype: int64
```

In [7]: df.describe()

min

max

29.000000

77.000000

Out[7]:		age	sex	ср	trestbps	chol	fbs
	count	1025.000000	1025.000000	1025.000000	1025.000000	1025.00000	1025.000000
	mean	54.434146	0.695610	0.942439	131.611707	246.00000	0.149268
	std	9.072290	0.460373	1.029641	17.516718	51.59251	0.356527

25% 48.000000 0.000000 0.000000 120.000000 211.00000 0.000000 **50**% 56.000000 1.000000 1.000000 130.000000 240.00000 0.000000 75% 61.000000 1.000000 2.000000 140.000000 275.00000 0.000000

3.000000

0.000000

94.000000

200.000000

126.00000

564.00000

4

```
In [8]: plt.figure(figsize=(5,5))
    sns.countplot(x='target', data=df)
    plt.title("Heart Disease Distribution")
    plt.show()
```

0.000000

1.000000

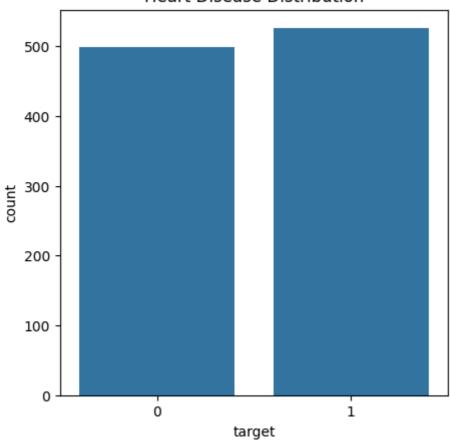
10

0.000000

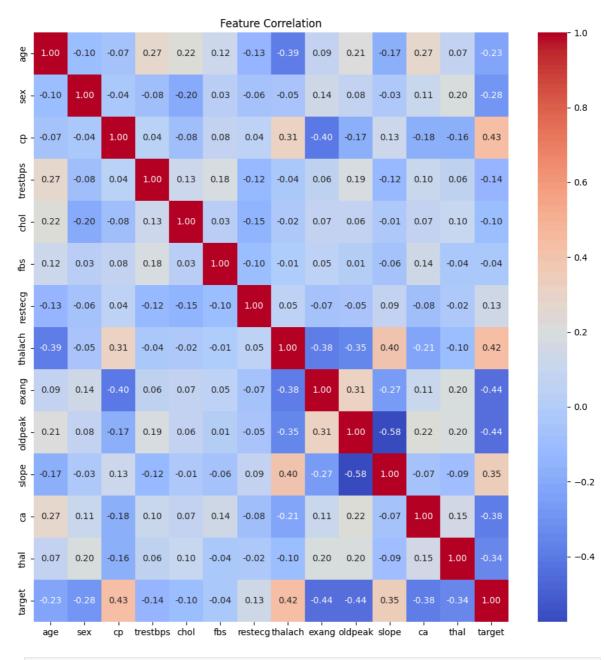
1.000000

5/9/25, 12:15 PM Heart Disease Detection

Heart Disease Distribution



```
In [9]: plt.figure(figsize=(12,12))
    sns.heatmap(df.corr(), annot=True, cmap='coolwarm', fmt='.2f')
    plt.title("Feature Correlation")
    plt.show()
```



```
In [10]:
        df = pd.get dummies(df, columns=['cp'], drop first=True)
         scaler = StandardScaler()
In [11]:
         scaled_features = scaler.fit_transform(df.drop('target', axis=1))
         df_scaled = pd.DataFrame(scaled_features, columns=df.drop('target', axis=1).colu
         df_scaled['target'] = df['target']
In [12]:
        X = df_scaled.drop('target', axis=1)
         y = df scaled['target']
         X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_
         print(f"Training data shape: {X_train.shape}")
         print(f"Test data shape: {X_test.shape}")
        Training data shape: (820, 15)
        Test data shape: (205, 15)
In [13]: model = LogisticRegression()
```

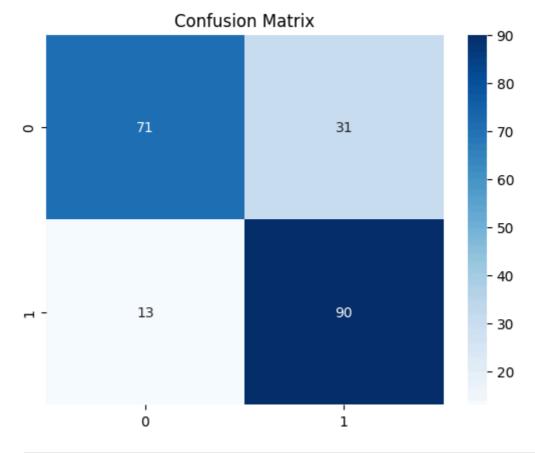
model.fit(X_train, y_train)

```
Out[13]: v LogisticRegression ()
```

```
In [14]: y_pred = model.predict(X_test)
print(f"Accuracy: {accuracy_score(y_test, y_pred)*100:.2f}%")
```

Accuracy: 78.54%

```
In [15]: conf_matrix = confusion_matrix(y_test, y_pred)
    sns.heatmap(conf_matrix, annot=True, fmt='d', cmap='Blues')
    plt.title("Confusion Matrix")
    plt.show()
```



In []: