$TI35_DSBDA_5th_HeartDisease$

January 30, 2025

```
[1]:
     import pandas as pd
     df=pd.read_csv("/home/bcl07/heart_disease.csv")
[4]:
    df
[4]:
                             trestbps
                                         chol
                                                fbs
                                                                                   oldpeak \
                  sex
                                                      restecg
                                                                thalach
                                                                           exang
            age
                        ср
                                                                     168
                                                                               0
                                                                                        1.0
     0
             52
                    1
                         0
                                   125
                                          212
                                                  0
                                                             1
     1
              53
                    1
                         0
                                   140
                                          203
                                                  1
                                                             0
                                                                     155
                                                                                1
                                                                                        3.1
     2
              70
                                                                     125
                                                                                        2.6
                     1
                         0
                                   145
                                          174
                                                  0
                                                             1
                                                                                1
     3
              61
                     1
                         0
                                   148
                                          203
                                                  0
                                                             1
                                                                     161
                                                                                0
                                                                                        0.0
     4
              62
                    0
                         0
                                   138
                                          294
                                                  1
                                                             1
                                                                     106
                                                                                0
                                                                                        1.9
     1020
              59
                    1
                         1
                                   140
                                          221
                                                  0
                                                             1
                                                                     164
                                                                                1
                                                                                        0.0
     1021
                                                                     141
                                                                                1
                                                                                        2.8
              60
                     1
                         0
                                   125
                                          258
                                                  0
                                                             0
     1022
              47
                         0
                                          275
                                                  0
                                                             0
                                                                     118
                                                                                1
                                                                                        1.0
                     1
                                   110
     1023
                                                             0
                                                                                0
                                                                                        0.0
              50
                    0
                         0
                                   110
                                          254
                                                  0
                                                                     159
     1024
              54
                     1
                         0
                                   120
                                          188
                                                  0
                                                             1
                                                                     113
                                                                                0
                                                                                        1.4
            slope
                    ca
                         thal
                                target
     0
                 2
                      2
                             3
                                      0
     1
                 0
                      0
                             3
                                      0
     2
                             3
                                      0
                 0
                      0
     3
                 2
                      1
                             3
                                      0
     4
                 1
                      3
                             2
                                      0
                             2
     1020
                 2
                      0
                                      1
     1021
                 1
                      1
                             3
                                      0
     1022
                      1
                             2
                                      0
                 1
     1023
                 2
                             2
                      0
                                      1
     1024
                             3
                                      0
                 1
                      1
     [1025 rows x 14 columns]
[5]: df.columns
```

```
'exang', 'oldpeak', 'slope', 'ca', 'thal', 'target'],
           dtype='object')
    df.isnull().sum()
[6]: age
                  0
                  0
     sex
                  0
     ср
     trestbps
                  0
     chol
                  0
     fbs
                  0
                  0
     restecg
     thalach
                  0
     exang
                  0
     oldpeak
                  0
     slope
                  0
     ca
                  0
     thal
                  0
                  0
     target
     dtype: int64
[8]:
    df=df.drop_duplicates()
[9]:
     df.describe()
[9]:
                                                    trestbps
                                                                      chol
                                                                                    fbs
                                              ср
                   age
                                sex
                                     302.000000
                                                  302.000000
                                                               302.000000
                                                                            302.000000
            302.00000
                        302.000000
     count
                                       0.963576
             54.42053
                                                  131.602649
                                                               246.500000
                                                                              0.149007
     mean
                           0.682119
     std
               9.04797
                           0.466426
                                        1.032044
                                                    17.563394
                                                                51.753489
                                                                              0.356686
     min
              29.00000
                           0.000000
                                        0.000000
                                                    94.000000
                                                               126.000000
                                                                              0.000000
     25%
             48.00000
                           0.000000
                                        0.000000
                                                  120.000000
                                                               211.000000
                                                                              0.000000
     50%
                                                  130.000000
             55.50000
                           1.000000
                                        1.000000
                                                               240.500000
                                                                              0.000000
     75%
             61.00000
                           1.000000
                                        2.000000
                                                  140.000000
                                                               274.750000
                                                                              0.000000
     max
              77.00000
                           1.000000
                                        3.000000
                                                  200.000000
                                                               564.000000
                                                                               1.000000
                restecg
                             thalach
                                                       oldpeak
                                                                      slope
                                            exang
                                                                                      ca
     count
            302.000000
                         302.000000
                                       302.000000
                                                   302.000000
                                                                302.000000
                                                                             302.000000
               0.526490
                         149.569536
                                         0.327815
                                                      1.043046
                                                                   1.397351
                                                                                0.718543
     mean
                           22.903527
                                        0.470196
     std
               0.526027
                                                      1.161452
                                                                   0.616274
                                                                                1.006748
     min
               0.000000
                          71.000000
                                        0.00000
                                                      0.000000
                                                                   0.000000
                                                                                0.00000
     25%
               0.000000
                         133.250000
                                        0.00000
                                                      0.000000
                                                                                0.00000
                                                                   1.000000
     50%
               1.000000
                         152.500000
                                        0.00000
                                                      0.800000
                                                                   1.000000
                                                                                0.00000
     75%
               1.000000
                         166.000000
                                         1.000000
                                                      1.600000
                                                                   2.000000
                                                                                1.000000
               2.000000
                         202.000000
                                         1.000000
                                                      6.200000
     max
                                                                   2.000000
                                                                                4.000000
                   thal
                              target
```

[5]: Index(['age', 'sex', 'cp', 'trestbps', 'chol', 'fbs', 'restecg', 'thalach',

```
302.000000 302.000000
count
         2.314570
                     0.543046
mean
std
         0.613026
                     0.498970
min
         0.000000
                     0.000000
25%
         2.000000
                     0.000000
50%
         2.000000
                     1.000000
75%
         3.000000
                     1.000000
max
         3.000000
                     1.000000
```

[10]: df.info()

<class 'pandas.core.frame.DataFrame'>

Index: 302 entries, 0 to 878
Data columns (total 14 columns):

#	Column	Non-Null Count	Dtype
0	age	302 non-null	int64
1	sex	302 non-null	int64
2	ср	302 non-null	int64
3	trestbps	302 non-null	int64
4	chol	302 non-null	int64
5	fbs	302 non-null	int64
6	restecg	302 non-null	int64
7	thalach	302 non-null	int64
8	exang	302 non-null	int64
9	oldpeak	302 non-null	float64
10	slope	302 non-null	int64
11	ca	302 non-null	int64
12	thal	302 non-null	int64
13	target	302 non-null	int64

dtypes: float64(1), int64(13)

memory usage: 35.4 KB

[11]: df.isna().sum()

[11]: age 0 sex 0 0 ср trestbps 0 chol 0 fbs 0 0 restecg thalach 0 exang 0 oldpeak 0 slope 0 ca 0

```
dtype: int64
[12]: df.head()
[12]:
                        trestbps chol
                                         fbs
                                               restecg
                                                        thalach exang
                                                                          oldpeak slope \
         age
              sex
                    ср
      0
          52
                 1
                     0
                              125
                                    212
                                            0
                                                     1
                                                             168
                                                                       0
                                                                              1.0
                                                                                        2
      1
          53
                     0
                              140
                                    203
                                            1
                                                     0
                                                             155
                                                                       1
                                                                              3.1
                                                                                        0
                 1
      2
          70
                 1
                     0
                              145
                                    174
                                            0
                                                     1
                                                             125
                                                                       1
                                                                              2.6
                                                                                        0
                                                                              0.0
                                                                                        2
      3
          61
                     0
                              148
                                    203
                                            0
                                                     1
                                                             161
                                                                       0
                 1
      4
                     0
          62
                 0
                              138
                                    294
                                            1
                                                     1
                                                             106
                                                                       0
                                                                              1.9
                                                                                        1
             thal
                    target
         ca
      0
          2
                 3
                         0
                         0
      1
          0
                 3
      2
          0
                 3
                         0
      3
          1
                 3
                         0
      4
          3
                 2
                         0
[13]: df.fbs.unique()
[13]: array([0, 1])
[16]: subSet1 = df[['age','cp','chol','thal']]
[18]: subSet2 = df[['exang','slope','target']]
[19]: merged_df = subSet1.merge(right=subSet2,how='cross')
      merged_df.head()
[19]:
                         thal
                                exang
                                       slope
                                               target
         age
              ср
                   chol
      0
          52
                0
                    212
                             3
                                    0
                                            2
                                                    0
      1
          52
                0
                    212
                             3
                                    1
                                            0
                                                    0
      2
          52
                    212
                             3
                                    1
                                            0
                                                    0
                0
                             3
                                            2
      3
          52
                0
                    212
                                    0
                                                    0
          52
                                    0
                                                    0
                    212
                             3
                                            1
[20]: df.columns
[20]: Index(['age', 'sex', 'cp', 'trestbps', 'chol', 'fbs', 'restecg', 'thalach',
              'exang', 'oldpeak', 'slope', 'ca', 'thal', 'target'],
             dtype='object')
[21]: def remove_outliers(column):
          Q1 = column.quantile(0.25)
          Q3 = column.quantile(0.75)
```

thal

target

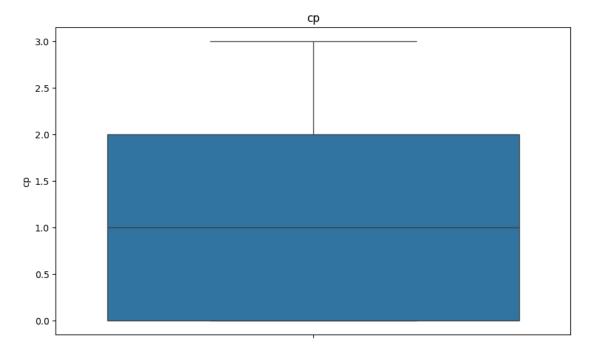
```
IQR = Q3 - Q1
          threshold = 1.5 * IQR
          outlier_mask = (column < Q1 - threshold) | (column > Q3 + threshold)
          return column[~outlier_mask]
[23]: col_name = ['cp', 'thal', 'exang', 'oldpeak', 'slope', 'ca']
      for col in col name:
          df[col] = remove_outliers(df[col])
     /tmp/ipykernel 10564/1228815343.py:3: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame.
     Try using .loc[row_indexer,col_indexer] = value instead
     See the caveats in the documentation: https://pandas.pydata.org/pandas-
     docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
       df[col] = remove_outliers(df[col])
     /tmp/ipykernel_10564/1228815343.py:3: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame.
     Try using .loc[row_indexer,col_indexer] = value instead
     See the caveats in the documentation: https://pandas.pydata.org/pandas-
     docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
       df[col] = remove_outliers(df[col])
     /tmp/ipykernel_10564/1228815343.py:3: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame.
     Try using .loc[row_indexer,col_indexer] = value instead
     See the caveats in the documentation: https://pandas.pydata.org/pandas-
     docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
       df[col] = remove_outliers(df[col])
     /tmp/ipykernel_10564/1228815343.py:3: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame.
     Try using .loc[row_indexer,col_indexer] = value instead
     See the caveats in the documentation: https://pandas.pydata.org/pandas-
     docs/stable/user guide/indexing.html#returning-a-view-versus-a-copy
       df[col] = remove_outliers(df[col])
     /tmp/ipykernel 10564/1228815343.py:3: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame.
     Try using .loc[row_indexer,col_indexer] = value instead
     See the caveats in the documentation: https://pandas.pydata.org/pandas-
     docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
       df[col] = remove_outliers(df[col])
     /tmp/ipykernel_10564/1228815343.py:3: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame.
     Try using .loc[row_indexer,col_indexer] = value instead
```

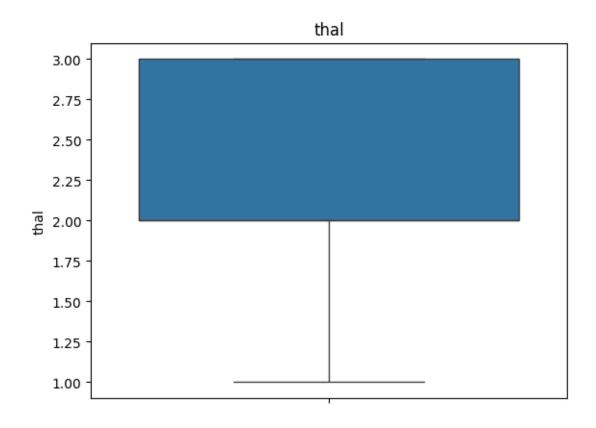
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy df[col] = remove_outliers(df[col])

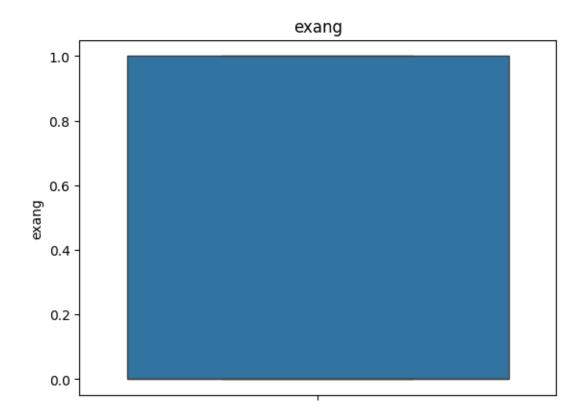
```
[28]: from sklearn.model_selection import train_test_split import matplotlib.pyplot as plt import seaborn as sns from sklearn.preprocessing import LabelEncoder from sklearn.metrics import accuracy_score,confusion_matrix from sklearn.linear_model import LogisticRegression import seaborn as sns import matplotlib.pyplot as plt
```

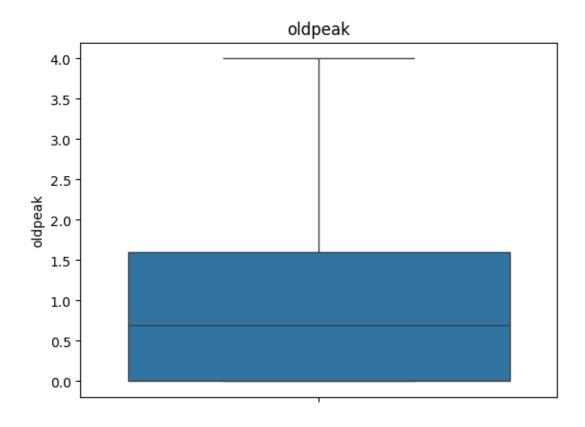
```
[29]: plt.figure(figsize=(10, 6)) # Adjust the figure size if needed

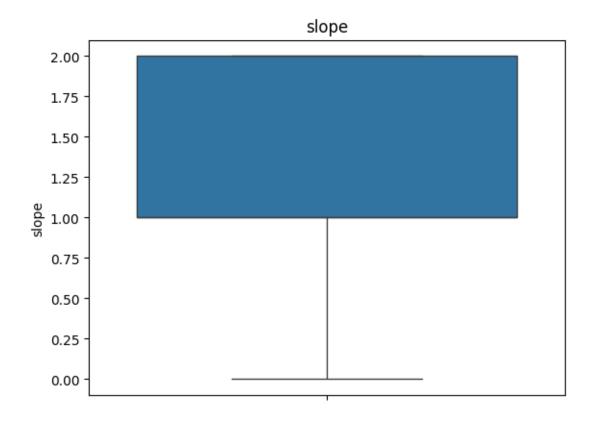
for col in col_name:
    sns.boxplot(data=df[col])
    plt.title(col)
    plt.show()
```

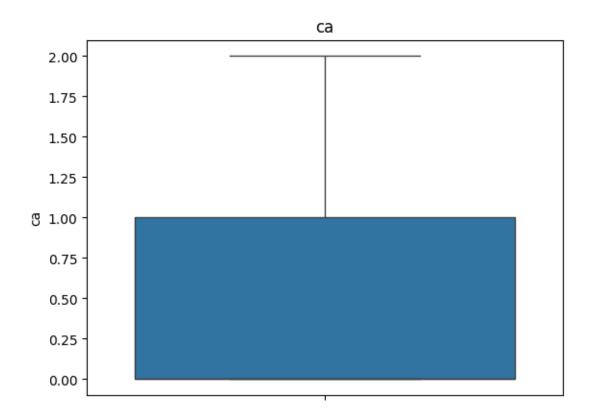








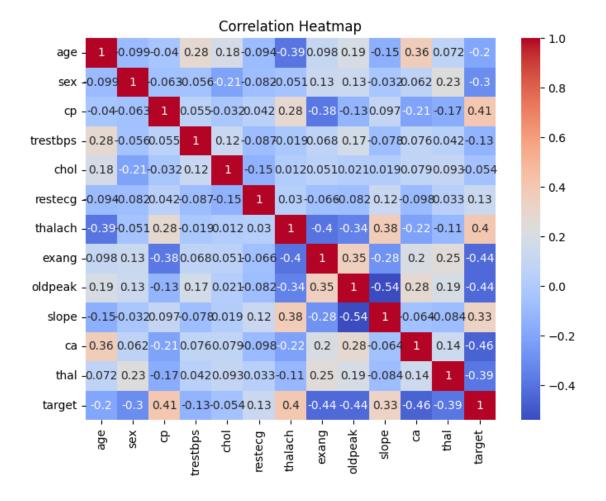




```
[30]: df = df.dropna()
[31]: df.isna().sum()
[31]: age
      sex
                  0
                  0
      ср
      trestbps
                  0
                  0
      chol
      fbs
                  0
      restecg
                  0
     thalach
                  0
      exang
                  0
      oldpeak
                  0
      slope
                  0
                  0
      ca
      thal
                  0
                  0
      target
      dtype: int64
[32]: df = df.drop('fbs',axis=1)
[34]: correlations = df.corr()['target'].drop('target')
      # Print correlations
      print("Correlation with the Target:")
      print(correlations)
      print()
      # Plot correlation heatmap
      plt.figure(figsize=(8, 6))
      sns.heatmap(df.corr(), annot=True, cmap='coolwarm')
      plt.title('Correlation Heatmap')
      plt.show()
     Correlation with the Target:
                 -0.199970
     age
     sex
                 -0.300311
                 0.408985
     ср
     trestbps
                -0.132882
     chol
                 -0.053834
                 0.125710
     restecg
     thalach
                 0.398870
                 -0.435511
     exang
     oldpeak
                 -0.436247
```

slope 0.327420 ca -0.459629 thal -0.389514

Name: target, dtype: float64



```
[37]: x = df[['cp','thal','exang','oldpeak','slope','ca']]
y = df.target
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,random_state=0)
x_train.shape,x_test.shape,y_train.shape,y_test.shape
[37]: ((219, 6), (55, 6), (219,), (55,))
[38]: from sklearn.preprocessing import StandardScaler
[39]: scaler = StandardScaler()
```

```
[40]: x_train_scaled = scaler.fit_transform(x_train)
      x_test_scaled = scaler.transform(x_test)
[42]: import numpy as np
[43]: y_train= np.array(y_train).reshape(-1, 1)
      y_test= np.array(y_test).reshape(-1, 1)
[44]: y_train.shape
[44]: (219, 1)
[45]: model = LogisticRegression()
      model.fit(x_train_scaled, y_train)
      # Make predictions on the test set
      y_pred = model.predict(x_test_scaled)
      # Evaluate the model's accuracy
      accuracy = accuracy_score(y_test, y_pred)
      print("Accuracy:", accuracy)
     Accuracy: 0.81818181818182
     /home/bcl07/.local/lib/python3.8/site-packages/sklearn/utils/validation.py:1183:
     DataConversionWarning: A column-vector y was passed when a 1d array was
     expected. Please change the shape of y to (n_samples, ), for example using
     ravel().
       y = column_or_1d(y, warn=True)
[46]: #Classification model using Decision Tree
      from sklearn.tree import DecisionTreeClassifier
      tc=DecisionTreeClassifier(criterion='entropy')
      tc.fit(x_train_scaled,y_train)
      y_pred=tc.predict(x_test_scaled)
      print("Training Accuracy Score :",accuracy_score(y_pred,y_test))
      print("Training Confusion Matrix :",confusion_matrix(y_pred,y_test))
     Training Accuracy Score : 0.7454545454545455
     Training Confusion Matrix : [[21 5]
      [ 9 20]]
 []:
```