## practical-exam-06

May 23, 2023

```
[]: from google.colab import drive drive.mount('/content/drive')
```

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force\_remount=True).

## 1 Problem Statement 6

Create a Linear Regression Model using Python to predict home prices using the Boston Housing Dataset. The Boston Housing dataset contains information about various houses in Boston through different parameters. There are 506 samples and 14 feature variables in this dataset.

```
[]: import pandas as pd
[160]: data = pd.read_csv('/content/drive/MyDrive/Colab Notebooks/exam_datasets/6.
         ⇔housing_data.csv')
      <IPython.core.display.HTML object>
[165]: data.dropna()
      <IPython.core.display.HTML object>
[165]:
               CRIM
                        ZN
                            INDUS
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                                                                        RAD
                                                         65.2
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       0
            0.00632
                      18.0
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       1
            0.02731
                       0.0
                             7.07
                                     0.0
                                          0.469
                                                  6.421
                                                         78.9
                                                                4.9671
                                                                             242
       2
            0.02729
                       0.0
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                                          0.469
                                                  7.185
                                                         61.1
                                                                4.9671
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                                                         45.8
       3
            0.03237
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                             2.18
                                     0.0
                                          0.458
                                                  6.998
                                                                6.0622
                                                                          3
                                                                             222
            0.02985
                                          0.458
                                                         58.7
                                                                6.0622
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       5
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                             9.69
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                                                  5.569
                                                         73.5
                                                                2.3999
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           0.22438
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            PTRATIO
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               15.3
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```

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1
              17.8 396.90
                              9.14 21.6
       2
               17.8 392.83
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                              2.94
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              18.7
                    394.12
                              5.21
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               19.2 395.77 15.10
                                   17.5
       500
              19.2 396.90
                            14.33
                                    16.8
       502
              21.0 396.90
                              9.08 20.6
       503
              21.0 396.90
                              5.64
                                    23.9
       504
              21.0 393.45
                              6.48 22.0
       [394 rows x 14 columns]
[166]: data.info()
      <IPython.core.display.HTML object>
      <class 'pandas.core.frame.DataFrame'>
      RangeIndex: 506 entries, 0 to 505
      Data columns (total 14 columns):
                    Non-Null Count Dtype
           Column
           _____
                    _____
       0
           CRIM
                    486 non-null
                                    float64
       1
           ZN
                    486 non-null
                                    float64
       2
           INDUS
                    486 non-null
                                    float64
       3
           CHAS
                    486 non-null
                                    float64
       4
           NOX
                    506 non-null
                                    float64
       5
           RM
                    506 non-null
                                    float64
       6
           AGE
                    486 non-null
                                    float64
       7
           DIS
                    506 non-null
                                    float64
       8
           RAD
                    506 non-null
                                    int64
       9
           TAX
                    506 non-null
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          PTRATIO
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       11 B
       12
          LSTAT
                    486 non-null
                                    float64
          MEDV
                    506 non-null
                                    float64
      dtypes: float64(12), int64(2)
      memory usage: 55.5 KB
[173]: from sklearn.model_selection import train_test_split
       from sklearn.linear_model import LinearRegression
       from sklearn.metrics import confusion_matrix, accuracy_score, precision_score,
        ⊶recall_score
      <IPython.core.display.HTML object>
[180]: data = data.dropna()
```

X = data.drop('MEDV', axis=1)

```
y = data['MEDV']
       X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2)
       lr = LinearRegression()
       lr.fit(X_train, y_train)
       y_pred = lr.predict(X_test)
      <IPython.core.display.HTML object>
[181]: r2 = lr.score(X_test, y_test)
      print(f'R^2 score: {r2:.2f}')
      <IPython.core.display.HTML object>
      R^2 score: 0.79
[182]: import matplotlib.pyplot as plt
       # Plot the predicted and actual home prices
       plt.scatter(y_test, y_pred)
       plt.xlabel('Actual Home Prices')
       plt.ylabel('Predicted Home Prices')
       plt.title('Actual vs Predicted Home Prices')
       plt.show()
```

<IPython.core.display.HTML object>

