## assignment-day-15

## February 9, 2024

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[40]:
      import pandas as pd
      import numpy as np
           = pd.read_csv("Boston.csv")
[41]: df
      df
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                            393.45
                                      6.48
                                            22.0
                     21.0
      505
            273
                     21.0
                           396.90
                                      7.88
                                           11.9
      [506 rows x 15 columns]
[42]:
      df.info()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 506 entries, 0 to 505

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#
          Column
                       Non-Null Count
                                       Dtype
      0
          Unnamed: 0 506 non-null
                                        int64
      1
          crim
                       506 non-null
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          chas
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          lstat
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                                        float64
      14 medv
                       506 non-null
                                        float64
     dtypes: float64(11), int64(4)
     memory usage: 59.4 KB
[43]: df.isnull().sum()
[43]: Unnamed: 0
                     0
      crim
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      indus
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      lstat
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      medv
      dtype: int64
[44]: x = df.drop(['medv', 'Unnamed: 0'], axis = 1)
      y = df['medv']
[45]: from sklearn.model_selection import train_test_split
      x_train,x_test,y_train,y_test = train_test_split(x,y,test_size = .2_
       →,random_state = 43)
```

Data columns (total 15 columns):

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[46]: from sklearn.preprocessing import StandardScaler
     scaler = StandardScaler()
     X_train_scaled = scaler.fit_transform(x_train)
     X_test_scaled = scaler.transform(x_test)
[47]: import tensorflow
     from tensorflow import keras
     from tensorflow.keras import Sequential
     from tensorflow.keras.layers import Dense
[48]: model= Sequential()
     model.add(Dense(7,activation = 'relu',input_dim = 13))
     model.add(Dense(5,activation = 'relu'))
     model.add(Dense(3,activation = 'linear'))
     model.add(Dense(1,activation = 'linear'))
[49]: model.summary()
    Model: "sequential_7"
     Layer (type)
                            Output Shape
    ______
                            (None, 7)
     dense_23 (Dense)
     dense_24 (Dense)
                            (None, 5)
                                                  40
     dense_25 (Dense)
                            (None, 3)
                                                  18
     dense_26 (Dense)
                            (None, 1)
    Total params: 160 (640.00 Byte)
    Trainable params: 160 (640.00 Byte)
    Non-trainable params: 0 (0.00 Byte)
[50]: model.compile(loss = tensorflow.keras.losses.Huber(delta=1.0),optimizer = ____
     [54]: history = model.fit(X_train_scaled,y_train,epochs = 200,validation_split= 0.2)
    Epoch 1/200
    2.4191
    Epoch 2/200
    2.4311
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Epoch 3/200
2,4086
Epoch 4/200
2.4020
Epoch 5/200
2.4224
Epoch 6/200
2.4471
Epoch 7/200
2.4394
Epoch 8/200
2.4137
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2.3590
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Epoch 14/200
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Epoch 19/200
2.3919
Epoch 20/200
2.4302
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2.2502
Epoch 32/200
2.2340
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2.2607
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Epoch 35/200
2.3023
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2.2759
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2.2675
Epoch 38/200
2.2721
Epoch 39/200
2.2683
Epoch 40/200
2.2569
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Epoch 42/200
2.2210
Epoch 43/200
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2.2058
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Epoch 51/200
2.1986
Epoch 52/200
2.2029
Epoch 53/200
2.2205
Epoch 54/200
2.2404
Epoch 55/200
2.2165
Epoch 56/200
2.1920
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2.2027
Epoch 58/200
2.1954
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2.1716
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2.1008
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 Epoch 200/200
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[55]: |y_pred = model.predict(X_test_scaled)
 4/4 [======= ] - Os 3ms/step
[59]: from sklearn.metrics import r2_score
  r2_score(y_test,y_pred)
```

[59]: 0.8430391976125973