```
In [40]: import pandas as pd import numpy as np
```

```
In [41]: df = pd.read_csv("Boston.csv")
df
```

Out[41]:

	Unnamed: 0	crim	zn	indus	chas	nox	rm	age	dis	rad	tax	ptratio	black
0	1	0.00632	18.0	2.31	0	0.538	6.575	65.2	4.0900	1	296	15.3	396.90
1	2	0.02731	0.0	7.07	0	0.469	6.421	78.9	4.9671	2	242	17.8	396.90
2	3	0.02729	0.0	7.07	0	0.469	7.185	61.1	4.9671	2	242	17.8	392.83
3	4	0.03237	0.0	2.18	0	0.458	6.998	45.8	6.0622	3	222	18.7	394.63
4	5	0.06905	0.0	2.18	0	0.458	7.147	54.2	6.0622	3	222	18.7	396.90
501	502	0.06263	0.0	11.93	0	0.573	6.593	69.1	2.4786	1	273	21.0	391.99
502	503	0.04527	0.0	11.93	0	0.573	6.120	76.7	2.2875	1	273	21.0	396.90
503	504	0.06076	0.0	11.93	0	0.573	6.976	91.0	2.1675	1	273	21.0	396.90
504	505	0.10959	0.0	11.93	0	0.573	6.794	89.3	2.3889	1	273	21.0	393.45
505	506	0.04741	0.0	11.93	0	0.573	6.030	80.8	2.5050	1	273	21.0	396.90

506 rows × 15 columns

,

In [42]: df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 506 entries, 0 to 505
Data columns (total 15 columns):

#	Column	Non-Null Count	Dtype
0	Unnamed: 0	506 non-null	int64
1	crim	506 non-null	float64
2	zn	506 non-null	float64
3	indus	506 non-null	float64
4	chas	506 non-null	int64
5	nox	506 non-null	float64
6	rm	506 non-null	float64
7	age	506 non-null	float64
8	dis	506 non-null	float64
9	rad	506 non-null	int64
10	tax	506 non-null	int64
11	ptratio	506 non-null	float64
12	black	506 non-null	float64
13	lstat	506 non-null	float64
14	medv	506 non-null	float64

dtypes: float64(11), int64(4)

memory usage: 59.4 KB

```
In [43]: | df.isnull().sum()
Out[43]: Unnamed: 0
                        0
         crim
                        0
         zn
                        0
         indus
                        0
         chas
                        0
         nox
                        0
         rm
                        0
         age
                        0
         dis
         rad
                        0
         tax
                        0
                        0
         ptratio
                        0
         black
         1stat
                        0
         medv
         dtype: int64
In [44]: x = df.drop(['medv', 'Unnamed: 0'], axis = 1)
         y = df['medv']
In [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_st
         ate = 43)
In [46]: | from sklearn.preprocessing import StandardScaler
         scaler = StandardScaler()
         X_train_scaled = scaler.fit_transform(x_train)
         X_test_scaled = scaler.transform(x_test)
In [47]:
         import tensorflow
         from tensorflow import keras
         from tensorflow.keras import Sequential
         from tensorflow.keras.layers import Dense
In [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'))
```

In [49]: model.summary()

Model: "sequential_7"

Layer (type)	Output Shape	Param #
dense_23 (Dense)	(None, 7)	98
dense_24 (Dense)	(None, 5)	40
dense_25 (Dense)	(None, 3)	18
dense_26 (Dense)	(None, 1)	4

Total params: 160 (640.00 Byte)
Trainable params: 160 (640.00 Byte)
Non-trainable params: 0 (0.00 Byte)

```
In [50]: model.compile(loss = tensorflow.keras.losses.Huber(delta=1.0),optimizer = 'Ada
m')
```

In [54]: history = model.fit(X_train_scaled,y_train,epochs = 200,validation_split= 0.2)

```
Epoch 1/200
ss: 2.4191
Epoch 2/200
11/11 [================ ] - 0s 9ms/step - loss: 1.9841 - val_los
s: 2.4311
Epoch 3/200
ss: 2.4086
Epoch 4/200
ss: 2.4020
Epoch 5/200
ss: 2.4224
Epoch 6/200
11/11 [================ ] - 0s 11ms/step - loss: 1.9641 - val lo
ss: 2.4471
Epoch 7/200
11/11 [================= ] - 0s 17ms/step - loss: 1.9573 - val lo
ss: 2.4394
Epoch 8/200
11/11 [================ ] - 0s 11ms/step - loss: 1.9515 - val lo
ss: 2.4137
Epoch 9/200
11/11 [============= ] - 0s 6ms/step - loss: 1.9480 - val_los
s: 2.4099
Epoch 10/200
11/11 [================= ] - 0s 5ms/step - loss: 1.9439 - val los
s: 2.3734
Epoch 11/200
11/11 [================= ] - 0s 7ms/step - loss: 1.9376 - val los
s: 2.3590
Epoch 12/200
s: 2.3328
Epoch 13/200
s: 2.3260
Epoch 14/200
s: 2.3905
Epoch 15/200
s: 2.4635
Epoch 16/200
s: 2.3742
Epoch 17/200
s: 2.3539
Epoch 18/200
11/11 [================ ] - 0s 7ms/step - loss: 1.9300 - val_los
s: 2.3331
Epoch 19/200
s: 2.3919
```

```
Epoch 20/200
s: 2.4302
Epoch 21/200
11/11 [================= ] - 0s 7ms/step - loss: 1.9054 - val_los
s: 2.3829
Epoch 22/200
s: 2.3790
Epoch 23/200
s: 2.4003
Epoch 24/200
11/11 [================== ] - 0s 6ms/step - loss: 1.9050 - val los
s: 2.3509
Epoch 25/200
s: 2.3021
Epoch 26/200
11/11 [================= ] - 0s 6ms/step - loss: 1.8855 - val los
s: 2.2942
Epoch 27/200
11/11 [================= ] - 0s 6ms/step - loss: 1.8816 - val los
s: 2.2991
Epoch 28/200
11/11 [================ ] - 0s 7ms/step - loss: 1.8761 - val los
s: 2.2744
Epoch 29/200
11/11 [================ ] - 0s 7ms/step - loss: 1.8776 - val los
s: 2.2540
Epoch 30/200
s: 2.2651
Epoch 31/200
11/11 [================= ] - 0s 7ms/step - loss: 1.8703 - val los
s: 2.2502
Epoch 32/200
s: 2.2340
Epoch 33/200
s: 2.2607
Epoch 34/200
s: 2.2901
Epoch 35/200
11/11 [================== ] - 0s 7ms/step - loss: 1.8607 - val_los
s: 2.3023
Epoch 36/200
s: 2.2759
Epoch 37/200
s: 2.2675
Epoch 38/200
s: 2.2721
```

```
Epoch 39/200
s: 2.2683
Epoch 40/200
s: 2.2569
Epoch 41/200
s: 2.2505
Epoch 42/200
11/11 [=============== ] - 0s 7ms/step - loss: 1.8379 - val_los
s: 2.2210
Epoch 43/200
11/11 [================= ] - 0s 7ms/step - loss: 1.8421 - val los
s: 2.2391
Epoch 44/200
11/11 [================= ] - 0s 8ms/step - loss: 1.8337 - val los
s: 2.2020
Epoch 45/200
11/11 [================ ] - 0s 6ms/step - loss: 1.8413 - val los
s: 2.2053
Epoch 46/200
11/11 [================= ] - 0s 7ms/step - loss: 1.8268 - val los
s: 2.2309
Epoch 47/200
11/11 [================ ] - 0s 7ms/step - loss: 1.8244 - val los
s: 2.2684
Epoch 48/200
s: 2.2709
Epoch 49/200
s: 2.2304
Epoch 50/200
11/11 [================ ] - 0s 7ms/step - loss: 1.8194 - val los
s: 2.2058
Epoch 51/200
11/11 [================= ] - 0s 5ms/step - loss: 1.8151 - val los
s: 2.1986
Epoch 52/200
s: 2.2029
Epoch 53/200
s: 2.2205
Epoch 54/200
11/11 [================= ] - 0s 7ms/step - loss: 1.8103 - val_los
s: 2.2404
Epoch 55/200
s: 2.2165
Epoch 56/200
11/11 [================= ] - 0s 7ms/step - loss: 1.8017 - val_los
s: 2.1920
Epoch 57/200
11/11 [================ ] - 0s 6ms/step - loss: 1.7976 - val_los
s: 2.2027
```

```
Epoch 58/200
s: 2.1954
Epoch 59/200
11/11 [================ ] - 0s 7ms/step - loss: 1.7911 - val_los
s: 2.1716
Epoch 60/200
s: 2.1606
Epoch 61/200
s: 2.1705
Epoch 62/200
11/11 [================== ] - 0s 5ms/step - loss: 1.7906 - val los
s: 2.1880
Epoch 63/200
11/11 [================== ] - 0s 5ms/step - loss: 1.7835 - val los
s: 2.1547
Epoch 64/200
11/11 [================== ] - 0s 6ms/step - loss: 1.8005 - val los
s: 2.1377
Epoch 65/200
11/11 [================== ] - 0s 6ms/step - loss: 1.7862 - val los
s: 2.1508
Epoch 66/200
11/11 [================= ] - 0s 5ms/step - loss: 1.7754 - val los
s: 2.1463
Epoch 67/200
11/11 [================ ] - 0s 7ms/step - loss: 1.7790 - val los
s: 2.1676
Epoch 68/200
11/11 [================= ] - 0s 7ms/step - loss: 1.7707 - val los
s: 2.1529
Epoch 69/200
11/11 [================ ] - 0s 5ms/step - loss: 1.7711 - val los
s: 2.1733
Epoch 70/200
s: 2.2022
Epoch 71/200
s: 2.1914
Epoch 72/200
s: 2.2044
Epoch 73/200
11/11 [================= ] - 0s 5ms/step - loss: 1.7766 - val_los
s: 2.1868
Epoch 74/200
s: 2.1736
Epoch 75/200
11/11 [================== ] - 0s 7ms/step - loss: 1.7620 - val_los
s: 2.1443
Epoch 76/200
11/11 [================== ] - 0s 7ms/step - loss: 1.7588 - val_los
s: 2.1630
```

```
Epoch 77/200
s: 2.1507
Epoch 78/200
s: 2.1836
Epoch 79/200
s: 2.1765
Epoch 80/200
11/11 [=============== ] - 0s 5ms/step - loss: 1.7433 - val_los
s: 2.1692
Epoch 81/200
11/11 [================= ] - 0s 7ms/step - loss: 1.7600 - val los
s: 2.1570
Epoch 82/200
11/11 [================== ] - 0s 6ms/step - loss: 1.7500 - val los
s: 2.1754
Epoch 83/200
11/11 [================ ] - 0s 5ms/step - loss: 1.7408 - val los
s: 2.1724
Epoch 84/200
11/11 [================= ] - 0s 7ms/step - loss: 1.7416 - val los
s: 2.1692
Epoch 85/200
11/11 [================ ] - 0s 7ms/step - loss: 1.7374 - val los
s: 2.1478
Epoch 86/200
11/11 [================ ] - 0s 7ms/step - loss: 1.7371 - val los
s: 2.1407
Epoch 87/200
s: 2.1414
Epoch 88/200
11/11 [================= ] - 0s 7ms/step - loss: 1.7358 - val los
s: 2.1770
Epoch 89/200
s: 2.1438
Epoch 90/200
s: 2.1447
Epoch 91/200
s: 2.1555
Epoch 92/200
11/11 [================== ] - 0s 8ms/step - loss: 1.7237 - val_los
s: 2.1625
Epoch 93/200
11/11 [================== ] - 0s 9ms/step - loss: 1.7188 - val_los
s: 2.1713
Epoch 94/200
s: 2.1652
Epoch 95/200
11/11 [=============== ] - 0s 9ms/step - loss: 1.7174 - val_los
s: 2.1894
```

```
Epoch 96/200
11/11 [================= ] - 0s 9ms/step - loss: 1.7202 - val los
s: 2.1852
Epoch 97/200
ss: 2.1730
Epoch 98/200
ss: 2.1705
Epoch 99/200
11/11 [=================== ] - 0s 9ms/step - loss: 1.7141 - val_los
s: 2.1759
Epoch 100/200
11/11 [================= ] - 0s 8ms/step - loss: 1.7157 - val los
s: 2.1685
Epoch 101/200
11/11 [================= ] - 0s 8ms/step - loss: 1.7101 - val los
s: 2.1421
Epoch 102/200
11/11 [================= ] - 0s 8ms/step - loss: 1.7049 - val los
s: 2.1321
Epoch 103/200
11/11 [================= ] - 0s 9ms/step - loss: 1.7119 - val los
s: 2.1320
Epoch 104/200
11/11 [================= ] - 0s 9ms/step - loss: 1.7051 - val los
s: 2.1490
Epoch 105/200
11/11 [================= ] - 0s 9ms/step - loss: 1.7053 - val los
s: 2.1535
Epoch 106/200
s: 2.1325
Epoch 107/200
11/11 [================ ] - 0s 9ms/step - loss: 1.6973 - val los
s: 2.1375
Epoch 108/200
ss: 2.1249
Epoch 109/200
s: 2.1008
Epoch 110/200
s: 2.1100
Epoch 111/200
s: 2.1177
Epoch 112/200
s: 2.1139
Epoch 113/200
11/11 [================ ] - 0s 5ms/step - loss: 1.6908 - val_los
s: 2.1299
Epoch 114/200
s: 2.1284
```

```
Epoch 115/200
11/11 [================= ] - 0s 7ms/step - loss: 1.7048 - val los
s: 2.1302
Epoch 116/200
s: 2.1313
Epoch 117/200
s: 2.1507
Epoch 118/200
s: 2.1576
Epoch 119/200
s: 2.1511
Epoch 120/200
11/11 [================== ] - 0s 5ms/step - loss: 1.6763 - val los
s: 2.1453
Epoch 121/200
11/11 [================= ] - 0s 5ms/step - loss: 1.6779 - val los
s: 2.1465
Epoch 122/200
11/11 [================== ] - 0s 7ms/step - loss: 1.6788 - val los
s: 2.1506
Epoch 123/200
11/11 [================= ] - 0s 5ms/step - loss: 1.6738 - val los
s: 2.1464
Epoch 124/200
11/11 [================= ] - 0s 6ms/step - loss: 1.6652 - val los
s: 2.1281
Epoch 125/200
11/11 [================== ] - 0s 7ms/step - loss: 1.6683 - val los
s: 2.1215
Epoch 126/200
11/11 [================ ] - 0s 7ms/step - loss: 1.6715 - val los
s: 2.1139
Epoch 127/200
11/11 [================= ] - 0s 7ms/step - loss: 1.6796 - val los
s: 2.1505
Epoch 128/200
s: 2.1057
Epoch 129/200
s: 2.1036
Epoch 130/200
11/11 [================== ] - 0s 7ms/step - loss: 1.6716 - val_los
s: 2.1399
Epoch 131/200
s: 2.1282
Epoch 132/200
s: 2.1360
Epoch 133/200
s: 2.1511
```

```
Epoch 134/200
s: 2.1273
Epoch 135/200
s: 2.1116
Epoch 136/200
s: 2.1312
Epoch 137/200
s: 2.1269
Epoch 138/200
11/11 [================== ] - 0s 6ms/step - loss: 1.6523 - val los
s: 2.1191
Epoch 139/200
11/11 [================== ] - 0s 5ms/step - loss: 1.6464 - val los
s: 2.1418
Epoch 140/200
11/11 [================= ] - 0s 6ms/step - loss: 1.6502 - val los
s: 2.1443
Epoch 141/200
11/11 [================== ] - 0s 5ms/step - loss: 1.6491 - val los
s: 2.1580
Epoch 142/200
11/11 [================== ] - 0s 5ms/step - loss: 1.6464 - val los
s: 2.1440
Epoch 143/200
11/11 [================= ] - 0s 6ms/step - loss: 1.6407 - val los
s: 2.1546
Epoch 144/200
s: 2.1753
Epoch 145/200
11/11 [================ ] - 0s 7ms/step - loss: 1.6471 - val los
s: 2.1895
Epoch 146/200
s: 2.1691
Epoch 147/200
s: 2.1762
Epoch 148/200
s: 2.1589
Epoch 149/200
s: 2.1786
Epoch 150/200
s: 2.1809
Epoch 151/200
11/11 [================= ] - 0s 7ms/step - loss: 1.6332 - val_los
s: 2.1625
Epoch 152/200
s: 2.1628
```

```
Epoch 153/200
s: 2.1462
Epoch 154/200
s: 2.1557
Epoch 155/200
s: 2.1700
Epoch 156/200
s: 2.1637
Epoch 157/200
11/11 [================== ] - 0s 5ms/step - loss: 1.6236 - val los
s: 2.1441
Epoch 158/200
11/11 [================= ] - 0s 6ms/step - loss: 1.6246 - val los
s: 2.1296
Epoch 159/200
11/11 [================ ] - 0s 6ms/step - loss: 1.6239 - val los
s: 2.1302
Epoch 160/200
11/11 [================= ] - 0s 7ms/step - loss: 1.6209 - val los
s: 2.1369
Epoch 161/200
11/11 [================ ] - 0s 7ms/step - loss: 1.6258 - val los
s: 2.1310
Epoch 162/200
11/11 [================= ] - 0s 6ms/step - loss: 1.6208 - val los
s: 2.1630
Epoch 163/200
11/11 [================ ] - 0s 7ms/step - loss: 1.6203 - val los
s: 2.1656
Epoch 164/200
11/11 [================= ] - 0s 6ms/step - loss: 1.6228 - val los
s: 2.1378
Epoch 165/200
s: 2.1388
Epoch 166/200
s: 2.1645
Epoch 167/200
s: 2.1748
Epoch 168/200
11/11 [================== ] - 0s 7ms/step - loss: 1.6149 - val_los
s: 2.1718
Epoch 169/200
s: 2.1502
Epoch 170/200
11/11 [=============== ] - 0s 6ms/step - loss: 1.6137 - val_los
s: 2.1580
Epoch 171/200
s: 2.1389
```

```
Epoch 172/200
s: 2.1479
Epoch 173/200
s: 2.1886
Epoch 174/200
s: 2.1635
Epoch 175/200
s: 2.1460
Epoch 176/200
11/11 [================== ] - 0s 6ms/step - loss: 1.6021 - val los
s: 2.1452
Epoch 177/200
11/11 [================= ] - 0s 7ms/step - loss: 1.6230 - val los
s: 2.1626
Epoch 178/200
11/11 [================ ] - 0s 7ms/step - loss: 1.6021 - val los
s: 2.1383
Epoch 179/200
11/11 [================= ] - 0s 7ms/step - loss: 1.6021 - val los
s: 2.1598
Epoch 180/200
11/11 [================= ] - 0s 6ms/step - loss: 1.6007 - val los
s: 2.1421
Epoch 181/200
11/11 [================= ] - 0s 6ms/step - loss: 1.6005 - val los
s: 2.1322
Epoch 182/200
11/11 [================ ] - 0s 7ms/step - loss: 1.6027 - val los
s: 2.1462
Epoch 183/200
11/11 [================ ] - 0s 6ms/step - loss: 1.6125 - val los
s: 2.1515
Epoch 184/200
s: 2.1123
Epoch 185/200
s: 2.1473
Epoch 186/200
s: 2.1729
Epoch 187/200
11/11 [================== ] - 0s 5ms/step - loss: 1.6032 - val_los
s: 2.1841
Epoch 188/200
s: 2.1538
Epoch 189/200
11/11 [================== ] - 0s 6ms/step - loss: 1.5915 - val_los
s: 2.1568
Epoch 190/200
s: 2.1541
```

```
Epoch 191/200
        11/11 [=============== ] - 0s 7ms/step - loss: 1.5857 - val los
        s: 2.1541
        Epoch 192/200
        11/11 [================ ] - 0s 8ms/step - loss: 1.6003 - val_los
        s: 2.1761
        Epoch 193/200
        11/11 [=============== ] - 0s 6ms/step - loss: 1.5893 - val los
        s: 2.1850
        Epoch 194/200
        11/11 [================ ] - 0s 6ms/step - loss: 1.5981 - val_los
        s: 2.1704
        Epoch 195/200
        11/11 [================== ] - 0s 7ms/step - loss: 1.5938 - val los
        s: 2.1349
        Epoch 196/200
        11/11 [============== ] - 0s 7ms/step - loss: 1.5943 - val los
        s: 2.1458
        Epoch 197/200
        11/11 [================ ] - 0s 8ms/step - loss: 1.5844 - val los
        s: 2.1813
        Epoch 198/200
        11/11 [============== ] - 0s 6ms/step - loss: 1.5853 - val los
        s: 2.1544
        Epoch 199/200
        11/11 [================== ] - 0s 6ms/step - loss: 1.5849 - val los
        s: 2.1598
        Epoch 200/200
        11/11 [================= ] - 0s 7ms/step - loss: 1.5794 - val los
        s: 2.1593
In [55]: | y_pred = model.predict(X_test_scaled)
        4/4 [======== ] - 0s 3ms/step
In [59]: from sklearn.metrics import r2 score
        r2_score(y_test,y_pred)
```

Out[59]: 0.8430391976125973