

Deep Learning Project: Creating Artificial Neural Network using Tensorflow & Keras

Import the required libraries

```
In [1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

```
In [2]: import tensorflow as tf
```

```
In [3]: tf.__version__
```

```
Out[3]: '2.0.0'
```

```
In [50]: %matplotlib inline
```

Read Dataset to a Pandas DataFrame

```
In [5]: df = pd.read_csv('E:\\Tensorflow-Projects\\Training-Data\\TF_2_Notebooks_and_Data\\DATA
```

```
In [6]: df.head()
```

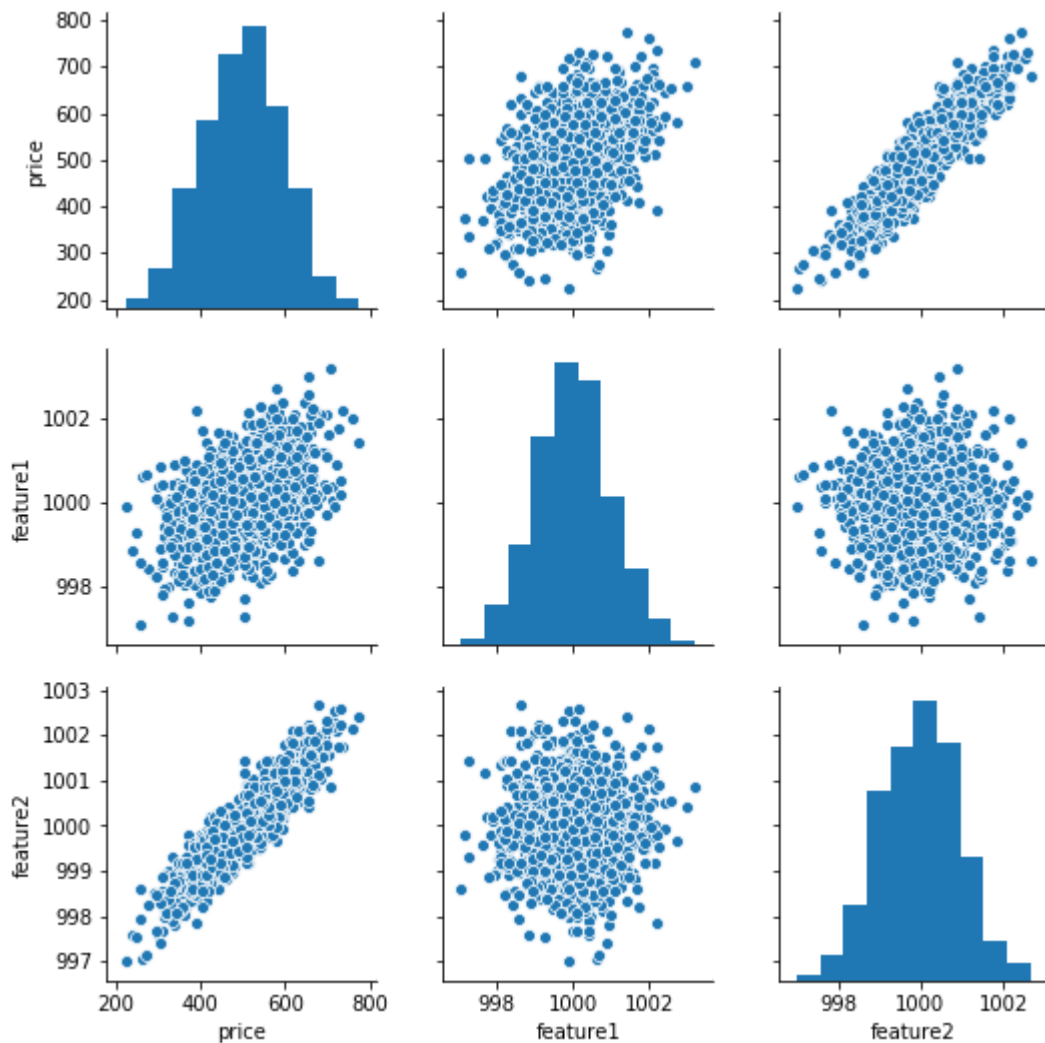
```
Out[6]:
```

	price	feature1	feature2
0	461.527929	999.787558	999.766096
1	548.130011	998.861615	1001.042403
2	410.297162	1000.070267	998.844015
3	540.382220	999.952251	1000.440940
4	546.024553	1000.446011	1000.338531

Visualize Data using Seaborn

```
In [7]: sns.pairplot(df)
```

```
Out[7]: <seaborn.axisgrid.PairGrid at 0x2c25e1a7888>
```



Split dataset into training & testing data using Skit-Learn

```
In [8]: from sklearn.model_selection import train_test_split
```

```
In [9]: X = df[['feature1', 'feature2']].values
```

```
In [10]: y = df['price'].values
```

```
In [11]: X
```

```
Out[11]: array([[ 999.78755752,  999.7660962 ],
 [ 998.86161491, 1001.04240315],
 [1000.07026691,  998.84401463],
 ...,
 [1001.45164617,  998.84760554],
 [1000.77102275,  998.56285086],
 [ 999.2322436 , 1001.45140713]])
```

```
In [12]: y
```

```
Out[12]: array([461.52792939, 548.13001146, 410.29716167, 540.38221981,
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 362.12270299, 433.41064026, 562.00257647, 637.30962074,
```

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```

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657.73425067, 518.35614426, 481.86280607, 476.52607826,
457.31318609, 456.72099249, 403.31557562, 599.36709348])

```

```
In [13]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=4
```

```
In [14]: X_train.shape
```

```
Out[14]: (700, 2)
```

```
In [15]: X_test.shape
```

```
Out[15]: (300, 2)
```

```
In [16]: y_train.shape
```

Out[16]: (700,)

In [17]: `y_test.shape`

Out[17]: (300,)

Normalize data and do model fitting

In [18]: `from sklearn.preprocessing import MinMaxScaler`

In [19]: `scaler = MinMaxScaler()`

In [20]: `scaler.fit(X_train)`

Out[20]: MinMaxScaler(copy=True, feature_range=(0, 1))

In [21]: `X_train = scaler.transform(X_train)`

In [22]: `X_train`

Out[22]: array([[0.74046017, 0.32583248],
 [0.43166001, 0.2555088],
 [0.18468554, 0.70500664],
 ...,
 [0.54913363, 0.79933822],
 [0.2834197 , 0.38818708],
 [0.56282703, 0.42371827]])

In [23]: `#scaler.fit(X_test)`

In [24]: `X_test = scaler.transform(X_test)`

In [25]: `X_test`

Out[25]: array([[0.39533339, 0.38540671],
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```
In [26]: X_train.max()
```

```
Out[26]: 1.0
```

```
In [27]: X_train.min()
```

```
Out[27]: 0.0
```

Create ANN using Tensorflow & Keras

```
In [28]: from tensorflow.keras.models import Sequential
         from tensorflow.keras.layers import Dense
```

```
In [29]: #model = Sequential([Dense(4, activation = 'relu'), Dense(2, activation = 'relu'), Dens
```

```
In [30]: model = Sequential()

         model.add(Dense(4, activation = 'relu'))
         model.add(Dense(4, activation = 'relu'))
         model.add(Dense(4, activation = 'relu'))

         model.add(Dense(1))

         model.compile(optimizer = 'rmsprop', loss = 'mse')
```

```
In [31]: model.fit(X_train, y_train, epochs=250)
```

```
Train on 700 samples
Epoch 1/250
700/700 [=====] - 1s 2ms/sample - loss: 256557.7725
Epoch 2/250
700/700 [=====] - 0s 63us/sample - loss: 256421.7502
Epoch 3/250
700/700 [=====] - 0s 70us/sample - loss: 256282.3427
Epoch 4/250
700/700 [=====] - 0s 68us/sample - loss: 256125.5980
Epoch 5/250
700/700 [=====] - 0s 58us/sample - loss: 255951.8255
Epoch 6/250
700/700 [=====] - 0s 58us/sample - loss: 255759.1884
Epoch 7/250
700/700 [=====] - 0s 54us/sample - loss: 255546.2038
Epoch 8/250
700/700 [=====] - 0s 45us/sample - loss: 255311.1220
```

```
Epoch 9/250
700/700 [=====] - 0s 55us/sample - loss: 255054.3161
Epoch 10/250
700/700 [=====] - 0s 57us/sample - loss: 254774.2131
Epoch 11/250
700/700 [=====] - 0s 58us/sample - loss: 254466.6479
Epoch 12/250
700/700 [=====] - 0s 71us/sample - loss: 254133.0237
Epoch 13/250
700/700 [=====] - 0s 63us/sample - loss: 253771.3095
Epoch 14/250
700/700 [=====] - 0s 63us/sample - loss: 253377.8091
Epoch 15/250
700/700 [=====] - 0s 61us/sample - loss: 252953.4549
Epoch 16/250
700/700 [=====] - 0s 70us/sample - loss: 252494.2045
Epoch 17/250
700/700 [=====] - 0s 63us/sample - loss: 252001.2207
Epoch 18/250
700/700 [=====] - 0s 66us/sample - loss: 251471.0077
Epoch 19/250
700/700 [=====] - 0s 61us/sample - loss: 250901.0522
Epoch 20/250
700/700 [=====] - 0s 96us/sample - loss: 250289.2633
Epoch 21/250
700/700 [=====] - 0s 85us/sample - loss: 249637.7737
Epoch 22/250
700/700 [=====] - 0s 72us/sample - loss: 248938.7877
Epoch 23/250
700/700 [=====] - 0s 69us/sample - loss: 248195.8648
Epoch 24/250
700/700 [=====] - 0s 56us/sample - loss: 247400.5395
Epoch 25/250
700/700 [=====] - 0s 59us/sample - loss: 246564.5959
Epoch 26/250
700/700 [=====] - 0s 52us/sample - loss: 245672.1789
Epoch 27/250
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Epoch 28/250
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Epoch 29/250
700/700 [=====] - 0s 58us/sample - loss: 242654.9055
Epoch 30/250
700/700 [=====] - 0s 46us/sample - loss: 241534.6576
Epoch 31/250
700/700 [=====] - 0s 54us/sample - loss: 240356.4185
Epoch 32/250
700/700 [=====] - 0s 57us/sample - loss: 239109.3000
Epoch 33/250
700/700 [=====] - 0s 57us/sample - loss: 237798.5691
Epoch 34/250
700/700 [=====] - 0s 43us/sample - loss: 236417.4737
Epoch 35/250
700/700 [=====] - 0s 43us/sample - loss: 234973.8300
Epoch 36/250
700/700 [=====] - 0s 57us/sample - loss: 233453.2707
Epoch 37/250
700/700 [=====] - 0s 57us/sample - loss: 231854.1284
Epoch 38/250
700/700 [=====] - 0s 49us/sample - loss: 230185.6584
Epoch 39/250
700/700 [=====] - 0s 43us/sample - loss: 228448.7166
Epoch 40/250
700/700 [=====] - 0s 58us/sample - loss: 226619.8066
Epoch 41/250
```

```
700/700 [=====] - 0s 57us/sample - loss: 224716.3676
Epoch 42/250
700/700 [=====] - 0s 43us/sample - loss: 222728.6550
Epoch 43/250
700/700 [=====] - 0s 57us/sample - loss: 220667.4054
Epoch 44/250
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Epoch 45/250
700/700 [=====] - 0s 124us/sample - loss: 216264.4286
Epoch 46/250
700/700 [=====] - 0s 138us/sample - loss: 213943.0517
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Epoch 53/250
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Epoch 54/250
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Epoch 55/250
700/700 [=====] - 0s 62us/sample - loss: 188897.8037
Epoch 56/250
700/700 [=====] - 0s 60us/sample - loss: 185654.2579
Epoch 57/250
700/700 [=====] - 0s 67us/sample - loss: 182302.8646
Epoch 58/250
700/700 [=====] - 0s 66us/sample - loss: 178850.4992
Epoch 59/250
700/700 [=====] - 0s 66us/sample - loss: 175318.4883
Epoch 60/250
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Epoch 61/250
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Epoch 64/250
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Epoch 66/250
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Epoch 68/250
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Epoch 69/250
700/700 [=====] - 0s 64us/sample - loss: 135178.3969
Epoch 70/250
700/700 [=====] - 0s 69us/sample - loss: 130748.2560
Epoch 71/250
700/700 [=====] - 0s 66us/sample - loss: 126237.5917
Epoch 72/250
700/700 [=====] - 0s 64us/sample - loss: 121687.3222
Epoch 73/250
700/700 [=====] - 0s 100us/sample - loss: 117068.3897
```

```
Epoch 74/250
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Epoch 75/250
700/700 [=====] - 0s 66us/sample - loss: 107700.2043
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700/700 [=====] - 0s 59us/sample - loss: 102950.7469
Epoch 77/250
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700/700 [=====] - 0s 66us/sample - loss: 93366.1744
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Epoch 90/250
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Epoch 91/250
700/700 [=====] - 0s 69us/sample - loss: 33806.4883
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700/700 [=====] - 0s 63us/sample - loss: 13243.1776
Epoch 98/250
700/700 [=====] - 0s 59us/sample - loss: 10669.7509
Epoch 99/250
700/700 [=====] - 0s 61us/sample - loss: 8372.6360
Epoch 100/250
700/700 [=====] - 0s 61us/sample - loss: 6382.3847
Epoch 101/250
700/700 [=====] - 0s 63us/sample - loss: 4751.8595
Epoch 102/250
700/700 [=====] - 0s 64us/sample - loss: 3445.9074
Epoch 103/250
700/700 [=====] - 0s 61us/sample - loss: 2497.4809
Epoch 104/250
700/700 [=====] - 0s 61us/sample - loss: 1887.7772
Epoch 105/250
700/700 [=====] - 0s 58us/sample - loss: 1597.8615
Epoch 106/250
```

```
700/700 [=====] - 0s 64us/sample - loss: 1516.3802
Epoch 107/250
700/700 [=====] - 0s 57us/sample - loss: 1493.9350
Epoch 108/250
700/700 [=====] - 0s 60us/sample - loss: 1475.7048
Epoch 109/250
700/700 [=====] - 0s 61us/sample - loss: 1453.5903
Epoch 110/250
700/700 [=====] - 0s 58us/sample - loss: 1427.2151
Epoch 111/250
700/700 [=====] - 0s 69us/sample - loss: 1405.2012
Epoch 112/250
700/700 [=====] - 0s 69us/sample - loss: 1382.1222
Epoch 113/250
700/700 [=====] - 0s 59us/sample - loss: 1360.5543
Epoch 114/250
700/700 [=====] - 0s 60us/sample - loss: 1337.5142
Epoch 115/250
700/700 [=====] - 0s 59us/sample - loss: 1314.7563
Epoch 116/250
700/700 [=====] - 0s 60us/sample - loss: 1293.2253
Epoch 117/250
700/700 [=====] - 0s 64us/sample - loss: 1271.5510
Epoch 118/250
700/700 [=====] - 0s 60us/sample - loss: 1251.4214
Epoch 119/250
700/700 [=====] - 0s 61us/sample - loss: 1229.1171
Epoch 120/250
700/700 [=====] - 0s 60us/sample - loss: 1208.7758
Epoch 121/250
700/700 [=====] - 0s 60us/sample - loss: 1187.9427
Epoch 122/250
700/700 [=====] - 0s 63us/sample - loss: 1165.2395
Epoch 123/250
700/700 [=====] - 0s 63us/sample - loss: 1145.2062
Epoch 124/250
700/700 [=====] - 0s 61us/sample - loss: 1126.4923
Epoch 125/250
700/700 [=====] - 0s 60us/sample - loss: 1106.6625
Epoch 126/250
700/700 [=====] - 0s 64us/sample - loss: 1086.2521
Epoch 127/250
700/700 [=====] - 0s 61us/sample - loss: 1066.8931
Epoch 128/250
700/700 [=====] - 0s 70us/sample - loss: 1046.8664
Epoch 129/250
700/700 [=====] - 0s 61us/sample - loss: 1027.4995
Epoch 130/250
700/700 [=====] - 0s 63us/sample - loss: 1006.7457
Epoch 131/250
700/700 [=====] - 0s 58us/sample - loss: 986.1231
Epoch 132/250
700/700 [=====] - 0s 63us/sample - loss: 966.0200
Epoch 133/250
700/700 [=====] - 0s 61us/sample - loss: 946.2505
Epoch 134/250
700/700 [=====] - 0s 63us/sample - loss: 926.6428
Epoch 135/250
700/700 [=====] - 0s 63us/sample - loss: 906.2352
Epoch 136/250
700/700 [=====] - 0s 60us/sample - loss: 886.0105
Epoch 137/250
700/700 [=====] - 0s 63us/sample - loss: 869.6947
Epoch 138/250
700/700 [=====] - 0s 64us/sample - loss: 851.0990
```



```
Epoch 139/250
700/700 [=====] - 0s 63us/sample - loss: 834.8774
Epoch 140/250
700/700 [=====] - 0s 63us/sample - loss: 817.4577
Epoch 141/250
700/700 [=====] - 0s 69us/sample - loss: 798.3013
Epoch 142/250
700/700 [=====] - 0s 64us/sample - loss: 781.5522
Epoch 143/250
700/700 [=====] - 0s 66us/sample - loss: 762.3143
Epoch 144/250
700/700 [=====] - 0s 64us/sample - loss: 743.4881
Epoch 145/250
700/700 [=====] - 0s 66us/sample - loss: 724.0567
Epoch 146/250
700/700 [=====] - 0s 63us/sample - loss: 705.3350
Epoch 147/250
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Epoch 148/250
700/700 [=====] - 0s 66us/sample - loss: 667.0976
Epoch 149/250
700/700 [=====] - 0s 60us/sample - loss: 655.9895
Epoch 150/250
700/700 [=====] - 0s 69us/sample - loss: 635.5615
Epoch 151/250
700/700 [=====] - 0s 74us/sample - loss: 620.9428
Epoch 152/250
700/700 [=====] - 0s 71us/sample - loss: 606.0777
Epoch 153/250
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Epoch 154/250
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Epoch 155/250
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Epoch 156/250
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Epoch 157/250
700/700 [=====] - 0s 70us/sample - loss: 524.0872
Epoch 158/250
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Epoch 159/250
700/700 [=====] - 0s 64us/sample - loss: 493.7547
Epoch 160/250
700/700 [=====] - 0s 63us/sample - loss: 478.8888
Epoch 161/250
700/700 [=====] - 0s 76us/sample - loss: 463.3976
Epoch 162/250
700/700 [=====] - 0s 64us/sample - loss: 449.5318
Epoch 163/250
700/700 [=====] - 0s 64us/sample - loss: 435.2809
Epoch 164/250
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Epoch 165/250
700/700 [=====] - 0s 61us/sample - loss: 408.0466
Epoch 166/250
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Epoch 167/250
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Epoch 168/250
700/700 [=====] - 0s 63us/sample - loss: 367.9875
Epoch 169/250
700/700 [=====] - 0s 63us/sample - loss: 354.1670
Epoch 170/250
700/700 [=====] - 0s 63us/sample - loss: 340.1690
Epoch 171/250
```

```
700/700 [=====] - 0s 64us/sample - loss: 328.7966
Epoch 172/250
700/700 [=====] - 0s 66us/sample - loss: 317.7703
Epoch 173/250
700/700 [=====] - 0s 69us/sample - loss: 306.8613
Epoch 174/250
700/700 [=====] - 0s 67us/sample - loss: 296.0338
Epoch 175/250
700/700 [=====] - 0s 67us/sample - loss: 284.8212
Epoch 176/250
700/700 [=====] - 0s 60us/sample - loss: 273.0503
Epoch 177/250
700/700 [=====] - 0s 66us/sample - loss: 262.2324
Epoch 178/250
700/700 [=====] - 0s 61us/sample - loss: 251.7230
Epoch 179/250
700/700 [=====] - 0s 61us/sample - loss: 240.7198
Epoch 180/250
700/700 [=====] - 0s 61us/sample - loss: 230.6565
Epoch 181/250
700/700 [=====] - 0s 64us/sample - loss: 219.9403
Epoch 182/250
700/700 [=====] - 0s 63us/sample - loss: 211.2201
Epoch 183/250
700/700 [=====] - 0s 63us/sample - loss: 201.5473
Epoch 184/250
700/700 [=====] - 0s 60us/sample - loss: 192.8432
Epoch 185/250
700/700 [=====] - 0s 61us/sample - loss: 184.2710
Epoch 186/250
700/700 [=====] - 0s 57us/sample - loss: 175.3587
Epoch 187/250
700/700 [=====] - 0s 55us/sample - loss: 166.2722
Epoch 188/250
700/700 [=====] - 0s 61us/sample - loss: 157.3738
Epoch 189/250
700/700 [=====] - 0s 59us/sample - loss: 149.2969
Epoch 190/250
700/700 [=====] - 0s 61us/sample - loss: 141.6626
Epoch 191/250
700/700 [=====] - 0s 59us/sample - loss: 134.6339
Epoch 192/250
700/700 [=====] - 0s 57us/sample - loss: 128.4960
Epoch 193/250
700/700 [=====] - 0s 69us/sample - loss: 121.8767
Epoch 194/250
700/700 [=====] - 0s 60us/sample - loss: 116.2836
Epoch 195/250
700/700 [=====] - 0s 61us/sample - loss: 109.2472
Epoch 196/250
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Epoch 197/250
700/700 [=====] - 0s 60us/sample - loss: 98.8897
Epoch 198/250
700/700 [=====] - 0s 60us/sample - loss: 93.7661
Epoch 199/250
700/700 [=====] - 0s 61us/sample - loss: 88.4047
Epoch 200/250
700/700 [=====] - 0s 60us/sample - loss: 84.2646
Epoch 201/250
700/700 [=====] - 0s 61us/sample - loss: 79.5805
Epoch 202/250
700/700 [=====] - 0s 61us/sample - loss: 74.9456
Epoch 203/250
700/700 [=====] - 0s 61us/sample - loss: 70.9472
```

```
Epoch 204/250
700/700 [=====] - 0s 61us/sample - loss: 67.1021
Epoch 205/250
700/700 [=====] - 0s 61us/sample - loss: 63.1922
Epoch 206/250
700/700 [=====] - 0s 59us/sample - loss: 59.8186
Epoch 207/250
700/700 [=====] - 0s 63us/sample - loss: 56.8748
Epoch 208/250
700/700 [=====] - 0s 64us/sample - loss: 53.9034
Epoch 209/250
700/700 [=====] - 0s 63us/sample - loss: 50.6025
Epoch 210/250
700/700 [=====] - 0s 61us/sample - loss: 47.8103
Epoch 211/250
700/700 [=====] - 0s 63us/sample - loss: 45.3179
Epoch 212/250
700/700 [=====] - 0s 60us/sample - loss: 42.6483
Epoch 213/250
700/700 [=====] - 0s 61us/sample - loss: 40.1013
Epoch 214/250
700/700 [=====] - 0s 61us/sample - loss: 38.2661
Epoch 215/250
700/700 [=====] - 0s 67us/sample - loss: 36.5119
Epoch 216/250
700/700 [=====] - 0s 64us/sample - loss: 34.6081
Epoch 217/250
700/700 [=====] - 0s 66us/sample - loss: 33.7661
Epoch 218/250
700/700 [=====] - 0s 60us/sample - loss: 32.7889
Epoch 219/250
700/700 [=====] - 0s 66us/sample - loss: 31.6185
Epoch 220/250
700/700 [=====] - 0s 66us/sample - loss: 30.3277
Epoch 221/250
700/700 [=====] - 0s 66us/sample - loss: 29.4205
Epoch 222/250
700/700 [=====] - 0s 66us/sample - loss: 28.9027
Epoch 223/250
700/700 [=====] - 0s 60us/sample - loss: 28.1476
Epoch 224/250
700/700 [=====] - 0s 66us/sample - loss: 27.8418
Epoch 225/250
700/700 [=====] - 0s 66us/sample - loss: 27.0312
Epoch 226/250
700/700 [=====] - 0s 60us/sample - loss: 26.8154
Epoch 227/250
700/700 [=====] - 0s 60us/sample - loss: 26.2801
Epoch 228/250
700/700 [=====] - 0s 61us/sample - loss: 26.0723
Epoch 229/250
700/700 [=====] - 0s 61us/sample - loss: 25.6823
Epoch 230/250
700/700 [=====] - 0s 61us/sample - loss: 25.4368
Epoch 231/250
700/700 [=====] - 0s 63us/sample - loss: 25.2948
Epoch 232/250
700/700 [=====] - 0s 63us/sample - loss: 24.8572
Epoch 233/250
700/700 [=====] - 0s 61us/sample - loss: 25.3475
Epoch 234/250
700/700 [=====] - 0s 66us/sample - loss: 24.7605
Epoch 235/250
700/700 [=====] - 0s 63us/sample - loss: 25.0212
Epoch 236/250
```

```

700/700 [=====] - 0s 64us/sample - loss: 24.3752
Epoch 237/250
700/700 [=====] - 0s 64us/sample - loss: 24.8712
Epoch 238/250
700/700 [=====] - 0s 63us/sample - loss: 24.5973
Epoch 239/250
700/700 [=====] - 0s 60us/sample - loss: 24.1756
Epoch 240/250
700/700 [=====] - 0s 64us/sample - loss: 24.3818
Epoch 241/250
700/700 [=====] - 0s 63us/sample - loss: 24.3913
Epoch 242/250
700/700 [=====] - 0s 60us/sample - loss: 24.4098
Epoch 243/250
700/700 [=====] - 0s 63us/sample - loss: 24.5523
Epoch 244/250
700/700 [=====] - 0s 61us/sample - loss: 24.4174
Epoch 245/250
700/700 [=====] - 0s 61us/sample - loss: 24.4073
Epoch 246/250
700/700 [=====] - 0s 69us/sample - loss: 24.2050
Epoch 247/250
700/700 [=====] - 0s 67us/sample - loss: 24.2827
Epoch 248/250
700/700 [=====] - 0s 64us/sample - loss: 24.2046
Epoch 249/250
700/700 [=====] - 0s 71us/sample - loss: 24.3464
Epoch 250/250
700/700 [=====] - 0s 90us/sample - loss: 23.8913

```

Out[31]: <tensorflow.python.keras.callbacks.History at 0x2c266e5bc08>

Plot Loss Data

In [32]: `model.history.history`

Out[32]:

```

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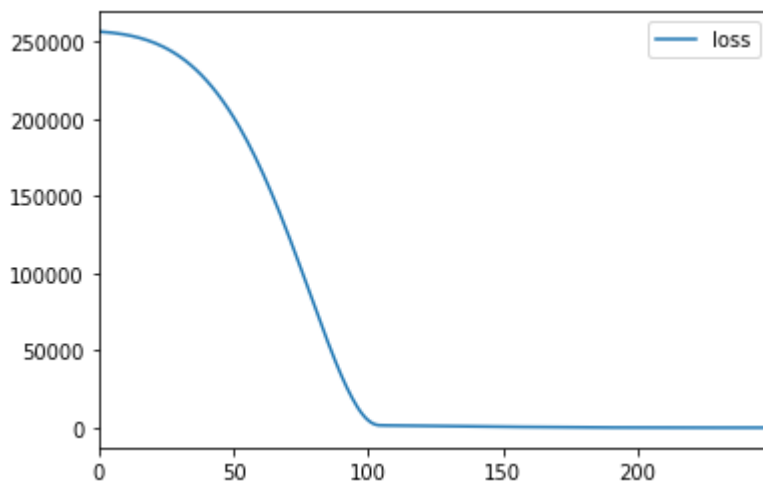
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```

```
In [33]: loss_df = pd.DataFrame(model.history.history)
```

```
In [34]: loss_df.plot()
```

```
Out[34]: <matplotlib.axes._subplots.AxesSubplot at 0x2c266e78b48>
```



Evaluate Model & do Predictions

```
In [35]: model.evaluate(X_train, y_train, verbose = 0)
```

```
Out[35]: 24.829104952130997
```

```
In [36]: model.evaluate(X_test, y_test, verbose=0)
```

```
Out[36]: 25.34867889404297
```



```
In [37]: test_predictions = model.predict(X_test)
```

```
In [38]: test_predictions
```

```
Out[38]: array([[404.88785],
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```

```
In [39]: test_predictions = pd.Series(test_predictions.reshape(300,))
```

Create a DataFrame having Actual & Predicted value

```
In [40]: prediction_df = pd.DataFrame(y_test, columns=['Test True Y'])
```

```
In [41]: prediction_df = pd.concat([prediction_df, test_predictions], axis=1)
```

```
In [42]: prediction_df.columns = ['Test True Y', 'Model Predictions']
```

```
In [43]: prediction_df
```

```
Out[43]:
```

	Test True Y	Model Predictions
0	402.296319	404.887848

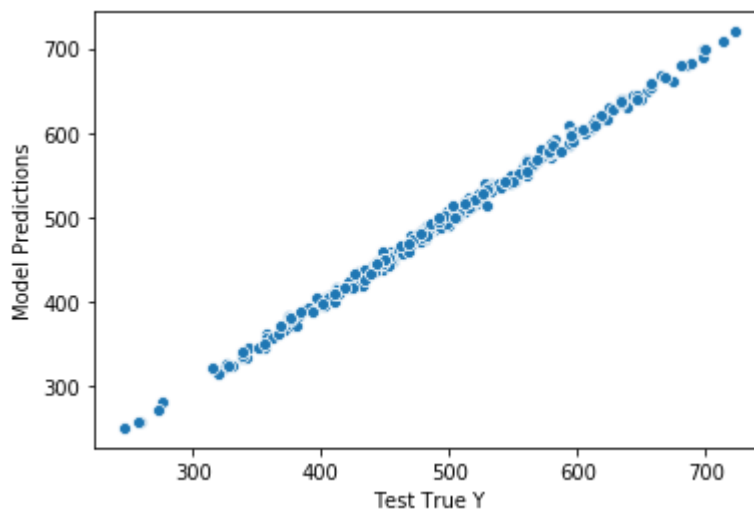
	Test True Y	Model Predictions
1	624.156198	622.522888
2	582.455066	591.146301
3	578.588606	571.484802
4	371.224104	366.573822
...
295	525.704657	528.406067
296	502.909473	514.918274
297	612.727910	608.645691
298	417.569725	416.402802
299	410.538250	410.554047

300 rows × 2 columns

Plot Actual vs Predicted results

```
In [44]: sns.scatterplot(x='Test True Y', y='Model Predictions', data=prediction_df)
```

```
Out[44]: <matplotlib.axes._subplots.AxesSubplot at 0x2c268900fc8>
```



Determine Model Evaluation parameters

```
In [45]: from sklearn import metrics
```

```
In [46]: metrics.mean_absolute_error(prediction_df['Test True Y'], prediction_df['Model Prediction'])
```

```
Out[46]: 4.035157520814229
```

```
In [47]: metrics.mean_squared_error(prediction_df['Test True Y'], prediction_df['Model Prediction'])
```

```
Out[47]: 25.348683058822864
```

```
In [48]: df.describe()
```

```
Out[48]:
```

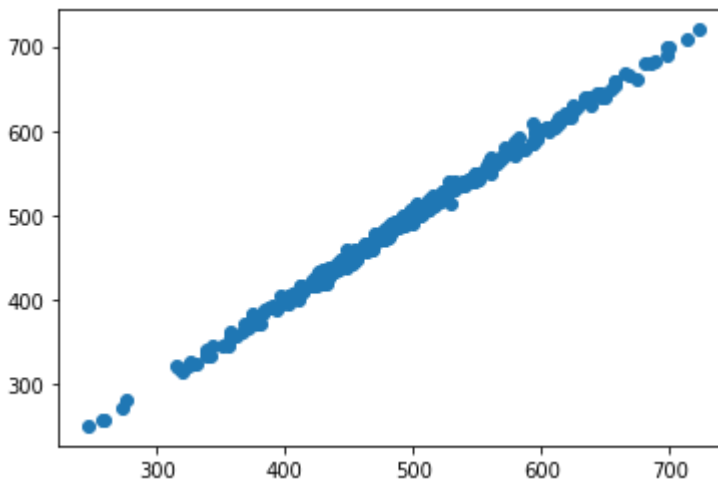
	price	feature1	feature2
count	1000.000000	1000.000000	1000.000000
mean	498.673029	1000.014171	999.979847
std	93.785431	0.974018	0.948330
min	223.346793	997.058347	996.995651
25%	433.025732	999.332068	999.316106
50%	502.382117	1000.009915	1000.002243
75%	564.921588	1000.637580	1000.645380
max	774.407854	1003.207934	1002.666308

```
In [49]: np.sqrt(metrics.mean_squared_error(prediction_df['Test True Y'],prediction_df['Model Pr
```

```
Out[49]: 5.034747566544212
```

```
In [53]: plt.scatter(prediction_df['Test True Y'],prediction_df['Model Predictions'])
```

```
Out[53]: <matplotlib.collections.PathCollection at 0x2c269fa6e08>
```



Predicting against a random data and saving Model for future use

```
In [60]: new_data = [[998, 1000]]
```

```
In [61]: new_data = scaler.transform(new_data)
```

```
In [62]: model.predict(new_data)
```

```
Out[62]: array([[419.4767]], dtype=float32)
```

```
In [63]: from tensorflow.keras.models import load_model
```

```
In [64]: model.save('ann-predictor.h5')
```

```
In [72]: #future_model = load_model('ann-predictor.h5')
```

```
In [ ]:
```