Assignment 1

Problem Statement:

Linear regression by using Deep Neural network: Implement Boston housing price prediction problem by Linear regression using Deep Neural network. Use Boston House price prediction dataset.

```
import tensorflow as tf
from tensorflow.keras.datasets import boston housing
from sklearn import preprocessing
import statistics
(train x,train y),(test x,test y)=boston housing.load data()
Downloading data from https://storage.googleapis.com/tensorflow/tf-
keras-datasets/boston housing.npz
57026/57026 [============ ] - Os 9us/step
print("Train shape", train x.shape)
print("Test shape",test_x.shape)
print("Actual train output", train x.shape)
print("Actual test output", test y.shape)
Train shape (404, 13)
Test shape (102, 13)
Actual train output (404, 13)
Actual test output (102,)
print(train_x[0])
train y[0]
[ 1.23247
            0.
                      8.14
                               0.
                                          0.538
                                                   6.142
                                                            91.7
  3.9769
                    307.
                               21.
            4.
                                        396.9
                                                  18.72
15.2
train x = preprocessing.normalize(train x)
test x = preprocessing.normalize(test x)
train_x[0]
array([0.0024119 , 0. , 0.01592969, 0. , 0.00105285,
       0.01201967, 0.17945359, 0.00778265, 0.00782786, 0.6007879 ,
       0.04109624, 0.77671895, 0.03663436])
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import *
```

```
def HousePricePredictionModel():
   model = Sequential()
   model.add(Dense(128,activation = 'relu' ,
input shape=(train x[0].shape)))
   model.add(Dense(64,activation = 'relu'))
   model.add(Dense(32,activation = 'relu'))
   model.add(Dense(1))
   model.compile(optimizer = 'rmsprop',loss = 'mse', metrics =
['mae'])
   return model
import numpy as np
k = 4
num val samples = len(train x)
num epochs = 100
all scores = []
model = HousePricePredictionModel()
history = model.fit(x = train_x , y = train_y , epochs = num_epochs ,
batch_size = 1 , verbose = 1 , validation_data = (test_x,test_y))
Epoch 1/100
WARNING:tensorflow:From C:\Users\Dell\AppData\Local\Programs\Python\
Python310\lib\site-packages\keras\src\utils\tf utils.py:492: The name
tf.ragged.RaggedTensorValue is deprecated. Please use
tf.compat.v1.ragged.RaggedTensorValue instead.
WARNING:tensorflow:From C:\Users\Dell\AppData\Local\Programs\Python\
Python310\lib\site-packages\keras\src\engine\base layer utils.py:384:
The name tf.executing eagerly outside functions is deprecated. Please
use tf.compat.v1.executing eagerly outside functions instead.
128.0591 - mae: 7.8230 - val loss: 68.9412 - val mae: 6.3187
Epoch 2/100
- mae: 5.8474 - val loss: 60.7842 - val mae: 5.6364
Epoch 3/100
- mae: 5.5251 - val loss: 56.2599 - val mae: 5.6630
Epoch 4/100
- mae: 5.4064 - val loss: 57.8930 - val mae: 5.8615
Epoch 5/100
- mae: 5.3897 - val loss: 63.9455 - val mae: 5.6201
Epoch 6/100
- mae: 5.1313 - val loss: 54.2596 - val mae: 5.2990
Epoch 7/100
```

```
- mae: 5.0763 - val loss: 61.4212 - val mae: 5.4867
Epoch 8/100
- mae: 4.9398 - val loss: 50.7711 - val mae: 5.3108
Epoch 9/100
404/404 [============= ] - 1s 3ms/step - loss: 52.0195
- mae: 4.9370 - val_loss: 48.9594 - val mae: 4.9985
Epoch 10/100
- mae: 4.7568 - val loss: 48.4572 - val mae: 5.0687
Epoch 11/100
- mae: 4.7740 - val loss: 52.3460 - val_mae: 5.0752
Epoch 12/100
- mae: 4.6680 - val loss: 43.8154 - val mae: 4.7239
Epoch 13/100
- mae: 4.5930 - val loss: 42.5392 - val mae: 4.5271
Epoch 14/100
- mae: 4.4959 - val loss: 59.2844 - val mae: 5.4363
Epoch 15/100
- mae: 4.5453 - val loss: 39.0927 - val mae: 4.3184
Epoch 16/100
- mae: 4.4501 - val loss: 39.6468 - val mae: 4.3140
Epoch 17/100
- mae: 4.3915 - val loss: 36.1830 - val mae: 4.3275
Epoch 18/100
- mae: 4.3241 - val loss: 34.2512 - val mae: 4.0915
Epoch 19/100
- mae: 4.1357 - val_loss: 46.1434 - val mae: 4.6565
Epoch 20/100
- mae: 4.1835 - val loss: 32.9330 - val mae: 4.1309
Epoch 21/100
- mae: 4.0955 - val loss: 32.8552 - val mae: 4.2409
Epoch 22/100
- mae: 4.0744 - val loss: 30.1412 - val_mae: 3.8964
Epoch 23/100
404/404 [============= ] - 2s 4ms/step - loss: 33.9822
```

```
- mae: 4.0255 - val loss: 36.2702 - val mae: 4.1493
Epoch 24/100
- mae: 4.0761 - val_loss: 29.1224 - val_mae: 3.8251
Epoch 25/100
- mae: 3.9159 - val loss: 48.7245 - val mae: 4.8987
Epoch 26/100
- mae: 3.8292 - val loss: 29.8893 - val mae: 4.0170
Epoch 27/100
- mae: 3.9341 - val_loss: 35.7099 - val_mae: 4.0735
Epoch 28/100
- mae: 3.7983 - val loss: 31.9667 - val_mae: 3.8874
Epoch 29/100
- mae: 3.9637 - val loss: 30.6505 - val mae: 3.7979
Epoch 30/100
- mae: 3.8146 - val loss: 28.2646 - val mae: 3.7082
Epoch 31/100
- mae: 3.7619 - val loss: 28.4177 - val mae: 4.1479
Epoch 32/100
- mae: 3.6074 - val loss: 25.4591 - val mae: 3.6637
Epoch 33/100
- mae: 3.6441 - val loss: 35.8763 - val mae: 4.1622
Epoch 34/100
- mae: 3.6945 - val loss: 25.6150 - val mae: 3.5820
Epoch 35/100
- mae: 3.5590 - val loss: 35.5714 - val mae: 4.1870
Epoch 36/100
- mae: 3.5287 - val loss: 24.4494 - val mae: 3.5551
Epoch 37/100
- mae: 3.5034 - val loss: 25.8649 - val_mae: 3.6792
Epoch 38/100
- mae: 3.5263 - val_loss: 26.6710 - val_mae: 3.6159
Epoch 39/100
- mae: 3.4885 - val loss: 36.5503 - val mae: 4.2475
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Epoch 40/100
- mae: 3.4193 - val loss: 28.6207 - val mae: 4.2058
Epoch 41/100
- mae: 3.3916 - val loss: 24.3072 - val mae: 3.5271
Epoch 42/100
- mae: 3.3257 - val loss: 25.3539 - val mae: 3.6888
Epoch 43/100
- mae: 3.3876 - val loss: 25.5542 - val mae: 3.5290
Epoch 44/100
- mae: 3.2976 - val loss: 30.2846 - val mae: 3.8610
Epoch 45/100
- mae: 3.4451 - val loss: 25.9018 - val mae: 3.5374
Epoch 46/100
- mae: 3.3189 - val loss: 25.8129 - val mae: 3.7734
Epoch 47/100
- mae: 3.1198 - val_loss: 26.9875 - val_mae: 3.6932
Epoch 48/100
404/404 [============== ] - 1s 3ms/step - loss: 20.6364
- mae: 3.3685 - val_loss: 31.1392 - val_mae: 3.9131
Epoch 49/100
- mae: 3.2307 - val loss: 28.7727 - val mae: 4.0031
Epoch 50/100
- mae: 3.2292 - val loss: 26.3803 - val mae: 3.5611
Epoch 51/100
- mae: 3.1358 - val loss: 28.8821 - val mae: 3.7873
Epoch 52/100
404/404 [=============] - 1s 3ms/step - loss: 20.0229
- mae: 3.1643 - val loss: 33.9240 - val mae: 4.1435
Epoch 53/100
- mae: 3.1406 - val loss: 26.4216 - val mae: 3.6331
Epoch 54/100
- mae: 3.1805 - val loss: 28.4030 - val mae: 3.9584
Epoch 55/100
- mae: 3.1775 - val_loss: 27.2741 - val_mae: 3.8400
Epoch 56/100
```

```
- mae: 3.0941 - val loss: 26.6840 - val mae: 3.5288
Epoch 57/100
- mae: 3.1427 - val loss: 38.9547 - val mae: 4.4885
Epoch 58/100
- mae: 3.1791 - val loss: 33.5011 - val mae: 4.0668
Epoch 59/100
- mae: 3.0623 - val loss: 26.5348 - val mae: 3.5416
Epoch 60/100
- mae: 3.1293 - val loss: 34.6186 - val mae: 4.1526
Epoch 61/100
- mae: 3.1269 - val loss: 27.8989 - val mae: 3.6327
Epoch 62/100
- mae: 3.0574 - val loss: 28.7944 - val mae: 3.6100
Epoch 63/100
- mae: 3.1885 - val loss: 30.5573 - val mae: 3.7372
Epoch 64/100
- mae: 3.1386 - val loss: 34.5406 - val mae: 4.2747
Epoch 65/100
- mae: 3.0581 - val loss: 31.2109 - val mae: 3.8233
Epoch 66/100
- mae: 3.0429 - val loss: 26.7627 - val mae: 3.8113
Epoch 67/100
- mae: 3.1240 - val loss: 31.4267 - val mae: 3.7336
Epoch 68/100
- mae: 2.9807 - val_loss: 28.8556 - val mae: 3.6217
Epoch 69/100
- mae: 3.1664 - val loss: 31.1792 - val mae: 3.8508
Epoch 70/100
- mae: 3.0782 - val loss: 29.9481 - val mae: 3.8888
Epoch 71/100
- mae: 3.0550 - val loss: 32.5425 - val_mae: 3.8812
Epoch 72/100
404/404 [============= ] - 1s 4ms/step - loss: 18.5199
```

```
- mae: 3.0208 - val loss: 32.9388 - val mae: 3.8158
Epoch 73/100
- mae: 2.9963 - val loss: 30.9806 - val mae: 3.7196
Epoch 74/100
- mae: 2.9822 - val loss: 32.0714 - val mae: 3.8845
Epoch 75/100
- mae: 3.0460 - val loss: 29.4708 - val mae: 3.7798
Epoch 76/100
- mae: 2.8880 - val_loss: 31.2977 - val_mae: 3.8956
Epoch 77/100
- mae: 2.9086 - val loss: 35.8937 - val_mae: 4.6390
Epoch 78/100
- mae: 2.9369 - val loss: 36.4089 - val mae: 4.3208
Epoch 79/100
- mae: 3.0158 - val loss: 32.0566 - val mae: 3.8074
Epoch 80/100
- mae: 2.9510 - val loss: 28.3061 - val mae: 3.7662
Epoch 81/100
- mae: 2.9636 - val loss: 37.3824 - val mae: 4.1582
Epoch 82/100
- mae: 2.9911 - val loss: 34.4895 - val mae: 3.9884
Epoch 83/100
- mae: 3.0692 - val loss: 28.7954 - val mae: 3.6450
Epoch 84/100
- mae: 2.9391 - val loss: 29.6099 - val mae: 3.7594
Epoch 85/100
- mae: 3.0009 - val loss: 52.9308 - val mae: 5.4933
Epoch 86/100
- mae: 2.9237 - val_loss: 36.6663 - val_mae: 4.4242
Epoch 87/100
- mae: 2.8675 - val_loss: 29.6442 - val_mae: 3.5867
Epoch 88/100
- mae: 3.0148 - val loss: 30.8007 - val mae: 3.7223
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Epoch 89/100
- mae: 2.9349 - val loss: 30.4669 - val mae: 3.6414
Epoch 90/100
404/404 [============= ] - 1s 3ms/step - loss: 17.6201
- mae: 2.9095 - val loss: 29.1059 - val mae: 3.6205
Epoch 91/100
- mae: 2.9661 - val loss: 29.8378 - val mae: 3.9131
Epoch 92/100
- mae: 2.9467 - val loss: 31.0635 - val mae: 3.7694
Epoch 93/100
- mae: 2.9177 - val loss: 26.6914 - val mae: 3.5139
Epoch 94/100
- mae: 2.9064 - val loss: 26.1249 - val mae: 3.4292
Epoch 95/100
- mae: 2.9948 - val loss: 31.4276 - val mae: 3.8287
Epoch 96/100
404/404 [============== ] - 1s 3ms/step - loss: 17.1544
- mae: 2.9307 - val loss: 27.6774 - val mae: 3.5788
Epoch 97/100
- mae: 2.9563 - val_loss: 26.9573 - val_mae: 3.4660
Epoch 98/100
- mae: 2.8757 - val loss: 35.3300 - val mae: 4.0154
Epoch 99/100
- mae: 2.9171 - val loss: 30.8429 - val mae: 4.0270
Epoch 100/100
- mae: 2.9354 - val loss: 26.4961 - val mae: 3.5876
y pred = model.predict(test x)
from IPython.display import display
import pandas as pd
import numpy as np
res = np.array(y pred)
result = res.ravel()
dict = {'y predicted' : result,
     'y actual' : test y}
df = pd.DataFrame(dict)
```

```
# displaying the DataFrame
display(df)
4/4 [=======] - 0s 3ms/step
    y_predicted y_actual
0
       9.804584
                   7.2
      18.540216
1
                   18.8
2
      23.167807
                   19.0
3
                   27.0
      35.055805
4
                   22.2
      25.245564
                    . . .
      45.174702
97
                   21.9
98
      24.418425
                   24.1
99
      53.045532
                   50.0
100
      27.694336
                   26.7
101 18.541021 25.0
[102 rows x 2 columns]
```