

Assignment 1

Importing the libraries

```
In [1]: import numpy as np
import pandas as pd
import tensorflow as tf
```

```
In [2]: tf.__version__
```

```
Out[2]: '2.18.0'
```

Part 1 - Data Preprocessing

Importing the dataset

```
In [3]: ds = pd.read_csv('boston_housing.csv')
ds
```

```
Out[3]:
```

	CRIM	ZN	INDUS	CHAS	NOX	RM	AGE	DIS	RAD	TAX	PTRATIO	
0	0.00632	18.0	2.31	0	0.538	6.575	65.2	4.0900	1	296	15.3	396
1	0.02731	0.0	7.07	0	0.469	6.421	78.9	4.9671	2	242	17.8	396
2	0.02729	0.0	7.07	0	0.469	7.185	61.1	4.9671	2	242	17.8	396
3	0.03237	0.0	2.18	0	0.458	6.998	45.8	6.0622	3	222	18.7	394
4	0.06905	0.0	2.18	0	0.458	7.147	54.2	6.0622	3	222	18.7	396
...
501	0.06263	0.0	11.93	0	0.573	6.593	69.1	2.4786	1	273	21.0	397
502	0.04527	0.0	11.93	0	0.573	6.120	76.7	2.2875	1	273	21.0	396
503	0.06076	0.0	11.93	0	0.573	6.976	91.0	2.1675	1	273	21.0	396
504	0.10959	0.0	11.93	0	0.573	6.794	89.3	2.3889	1	273	21.0	395
505	0.04741	0.0	11.93	0	0.573	6.030	80.8	2.5050	1	273	21.0	396

506 rows × 14 columns



```
In [4]: X = ds.iloc[:, :-1].values
y = ds.iloc[:, -1].values
```

Splitting the dataset into the Training set and Test set

```
In [5]: from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.2, random_state=42)
```

Feature Scaling

```
In [6]: from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
X_train = sc.fit_transform(X_train)
X_test = sc.transform(X_test)
```

Part 2 - Building the ANN

Initializing the ANN

```
In [7]: ann = tf.keras.models.Sequential()
```

Adding the input layer and the first hidden layer

```
In [8]: ann.add(tf.keras.layers.Dense(units=150, activation='relu'))
```

Adding the second hidden layer

```
In [9]: ann.add(tf.keras.layers.Dense(units=100, activation='relu'))
```

```
In [10]: ann.add(tf.keras.layers.Dense(units=50, activation='relu'))
```

Adding the output layer

```
In [11]: ann.add(tf.keras.layers.Dense(units=1, activation='relu'))
```

Part 3 - Training the ANN































Compiling the ANN































```
In [12]: ann.compile(optimizer = 'adam', loss = 'mse', metrics = ['mae'])
```

Training the ANN on the Training set

```
In [13]: ann.fit(X_train, y_train, batch_size = 32, epochs = 100)
```

Epoch 1/100

13/13		2s	3ms/step	-	loss: 563.7682	-	mae: 21.9091
Epoch 2/100							
13/13		0s	2ms/step	-	loss: 487.1405	-	mae: 20.1474
Epoch 3/100							
13/13		0s	2ms/step	-	loss: 341.0410	-	mae: 16.0792
Epoch 4/100							
13/13		0s	2ms/step	-	loss: 111.9793	-	mae: 8.5422
Epoch 5/100							
13/13		0s	2ms/step	-	loss: 56.9797	-	mae: 5.7580
Epoch 6/100							
13/13		0s	2ms/step	-	loss: 31.1616	-	mae: 4.0791
Epoch 7/100							
13/13		0s	2ms/step	-	loss: 22.3824	-	mae: 3.4530
Epoch 8/100							
13/13		0s	2ms/step	-	loss: 20.5991	-	mae: 3.2987
Epoch 9/100							
13/13		0s	2ms/step	-	loss: 19.8079	-	mae: 3.2045
Epoch 10/100							
13/13		0s	2ms/step	-	loss: 23.2158	-	mae: 3.2876
Epoch 11/100							
13/13		0s	2ms/step	-	loss: 14.2159	-	mae: 2.6943
Epoch 12/100							
13/13		0s	2ms/step	-	loss: 13.4323	-	mae: 2.6931
Epoch 13/100							
13/13		0s	2ms/step	-	loss: 11.0760	-	mae: 2.4843
Epoch 14/100							
13/13		0s	2ms/step	-	loss: 14.3434	-	mae: 2.5998
Epoch 15/100							
13/13		0s	2ms/step	-	loss: 11.7372	-	mae: 2.5316
Epoch 16/100							
13/13		0s	2ms/step	-	loss: 10.7616	-	mae: 2.3284
Epoch 17/100							
13/13		0s	2ms/step	-	loss: 10.6445	-	mae: 2.3264
Epoch 18/100							
13/13		0s	2ms/step	-	loss: 9.4197	-	mae: 2.2221
Epoch 19/100							
13/13		0s	2ms/step	-	loss: 10.1069	-	mae: 2.4366
Epoch 20/100							
13/13		0s	2ms/step	-	loss: 10.1548	-	mae: 2.3020
Epoch 21/100							
13/13		0s	2ms/step	-	loss: 8.2259	-	mae: 2.1241
Epoch 22/100							
13/13		0s	3ms/step	-	loss: 9.3914	-	mae: 2.2269
Epoch 23/100							
13/13		0s	2ms/step	-	loss: 9.7292	-	mae: 2.1913
Epoch 24/100							
13/13		0s	2ms/step	-	loss: 8.8796	-	mae: 2.1650
Epoch 25/100							
13/13		0s	2ms/step	-	loss: 8.1743	-	mae: 2.0516
Epoch 26/100							
13/13		0s	2ms/step	-	loss: 7.7929	-	mae: 2.0550
Epoch 27/100							
13/13		0s	2ms/step	-	loss: 8.4374	-	mae: 2.0709
Epoch 28/100							
13/13		0s	2ms/step	-	loss: 7.1634	-	mae: 1.9542
Epoch 29/100							
13/13		0s	2ms/step	-	loss: 6.8790	-	mae: 1.9721
Epoch 30/100							
13/13		0s	3ms/step	-	loss: 6.9440	-	mae: 1.9513
Epoch 31/100							

13/13		0s	2ms/step	-	loss: 6.5877	-	mae: 1.9434
Epoch 32/100							
13/13		0s	2ms/step	-	loss: 7.2339	-	mae: 2.0440
Epoch 33/100							
13/13		0s	1ms/step	-	loss: 7.5452	-	mae: 1.9965
Epoch 34/100							
13/13		0s	2ms/step	-	loss: 6.3850	-	mae: 1.8628
Epoch 35/100							
13/13		0s	2ms/step	-	loss: 6.4143	-	mae: 1.9188
Epoch 36/100							
13/13		0s	2ms/step	-	loss: 6.4348	-	mae: 1.9004
Epoch 37/100							
13/13		0s	2ms/step	-	loss: 7.0753	-	mae: 1.9538
Epoch 38/100							
13/13		0s	2ms/step	-	loss: 6.8706	-	mae: 1.9252
Epoch 39/100							
13/13		0s	2ms/step	-	loss: 7.5215	-	mae: 2.0026
Epoch 40/100							
13/13		0s	2ms/step	-	loss: 6.2592	-	mae: 1.8697
Epoch 41/100							
13/13		0s	2ms/step	-	loss: 6.3253	-	mae: 1.8757
Epoch 42/100							
13/13		0s	2ms/step	-	loss: 5.8362	-	mae: 1.7908
Epoch 43/100							
13/13		0s	2ms/step	-	loss: 6.2725	-	mae: 1.8872
Epoch 44/100							
13/13		0s	2ms/step	-	loss: 5.8477	-	mae: 1.8701
Epoch 45/100							
13/13		0s	2ms/step	-	loss: 5.5362	-	mae: 1.7230
Epoch 46/100							
13/13		0s	2ms/step	-	loss: 6.2573	-	mae: 1.8399
Epoch 47/100							
13/13		0s	2ms/step	-	loss: 5.8049	-	mae: 1.7763
Epoch 48/100							
13/13		0s	2ms/step	-	loss: 5.7743	-	mae: 1.7914
Epoch 49/100							
13/13		0s	2ms/step	-	loss: 5.0485	-	mae: 1.7178
Epoch 50/100							
13/13		0s	2ms/step	-	loss: 5.5995	-	mae: 1.7030
Epoch 51/100							
13/13		0s	2ms/step	-	loss: 5.8756	-	mae: 1.7738
Epoch 52/100							
13/13		0s	1ms/step	-	loss: 4.8289	-	mae: 1.6777
Epoch 53/100							
13/13		0s	2ms/step	-	loss: 4.8117	-	mae: 1.6290
Epoch 54/100							
13/13		0s	2ms/step	-	loss: 5.2025	-	mae: 1.7635
Epoch 55/100							
13/13		0s	2ms/step	-	loss: 5.4743	-	mae: 1.7728
Epoch 56/100							
13/13		0s	2ms/step	-	loss: 5.3838	-	mae: 1.7171
Epoch 57/100							
13/13		0s	2ms/step	-	loss: 4.8033	-	mae: 1.6608
Epoch 58/100							
13/13		0s	2ms/step	-	loss: 4.3749	-	mae: 1.6105
Epoch 59/100							
13/13		0s	2ms/step	-	loss: 4.6322	-	mae: 1.6517
Epoch 60/100							
13/13		0s	2ms/step	-	loss: 4.5509	-	mae: 1.5971
Epoch 61/100							

13/13	0s	2ms/step	-	loss: 4.6773	-	mae: 1.6010
Epoch 62/100						
13/13	0s	2ms/step	-	loss: 4.3954	-	mae: 1.5994
Epoch 63/100						
13/13	0s	2ms/step	-	loss: 5.4361	-	mae: 1.7930
Epoch 64/100						
13/13	0s	2ms/step	-	loss: 5.0977	-	mae: 1.7404
Epoch 65/100						
13/13	0s	2ms/step	-	loss: 4.5783	-	mae: 1.5902
Epoch 66/100						
13/13	0s	2ms/step	-	loss: 4.8521	-	mae: 1.7237
Epoch 67/100						
13/13	0s	2ms/step	-	loss: 4.8779	-	mae: 1.7235
Epoch 68/100						
13/13	0s	2ms/step	-	loss: 4.5348	-	mae: 1.6140
Epoch 69/100						
13/13	0s	3ms/step	-	loss: 4.7998	-	mae: 1.6934
Epoch 70/100						
13/13	0s	2ms/step	-	loss: 5.0253	-	mae: 1.6833
Epoch 71/100						
13/13	0s	2ms/step	-	loss: 3.7800	-	mae: 1.5010
Epoch 72/100						
13/13	0s	2ms/step	-	loss: 4.2671	-	mae: 1.5452
Epoch 73/100						
13/13	0s	2ms/step	-	loss: 3.8306	-	mae: 1.4788
Epoch 74/100						
13/13	0s	2ms/step	-	loss: 4.0283	-	mae: 1.5410
Epoch 75/100						
13/13	0s	2ms/step	-	loss: 3.6946	-	mae: 1.4680
Epoch 76/100						
13/13	0s	2ms/step	-	loss: 3.4699	-	mae: 1.4643
Epoch 77/100						
13/13	0s	2ms/step	-	loss: 4.1583	-	mae: 1.5345
Epoch 78/100						
13/13	0s	2ms/step	-	loss: 3.2960	-	mae: 1.4199
Epoch 79/100						
13/13	0s	2ms/step	-	loss: 3.8190	-	mae: 1.4887
Epoch 80/100						
13/13	0s	2ms/step	-	loss: 3.9262	-	mae: 1.4995
Epoch 81/100						
13/13	0s	2ms/step	-	loss: 3.6127	-	mae: 1.4592
Epoch 82/100						
13/13	0s	2ms/step	-	loss: 3.7006	-	mae: 1.4414
Epoch 83/100						
13/13	0s	2ms/step	-	loss: 3.3318	-	mae: 1.3775
Epoch 84/100						
13/13	0s	1ms/step	-	loss: 3.5521	-	mae: 1.4353
Epoch 85/100						
13/13	0s	2ms/step	-	loss: 3.3651	-	mae: 1.4300
Epoch 86/100						
13/13	0s	3ms/step	-	loss: 3.6556	-	mae: 1.4559
Epoch 87/100						
13/13	0s	2ms/step	-	loss: 3.2425	-	mae: 1.4052
Epoch 88/100						
13/13	0s	2ms/step	-	loss: 3.8450	-	mae: 1.5218
Epoch 89/100						
13/13	0s	2ms/step	-	loss: 3.7119	-	mae: 1.4949
Epoch 90/100						
13/13	0s	2ms/step	-	loss: 2.8289	-	mae: 1.3033
Epoch 91/100						

```
13/13 ————— 0s 2ms/step - loss: 2.5812 - mae: 1.2433
Epoch 92/100
13/13 ————— 0s 2ms/step - loss: 3.2835 - mae: 1.3468
Epoch 93/100
13/13 ————— 0s 2ms/step - loss: 3.5620 - mae: 1.4550
Epoch 94/100
13/13 ————— 0s 2ms/step - loss: 2.8241 - mae: 1.2703
Epoch 95/100
13/13 ————— 0s 2ms/step - loss: 2.9526 - mae: 1.3182
Epoch 96/100
13/13 ————— 0s 2ms/step - loss: 2.6215 - mae: 1.2318
Epoch 97/100
13/13 ————— 0s 2ms/step - loss: 3.9691 - mae: 1.5211
Epoch 98/100
13/13 ————— 0s 2ms/step - loss: 2.9109 - mae: 1.2978
Epoch 99/100
13/13 ————— 0s 2ms/step - loss: 2.7802 - mae: 1.2513
Epoch 100/100
13/13 ————— 0s 2ms/step - loss: 2.6324 - mae: 1.2657
```

Out[13]: <keras.src.callbacks.history.History at 0x16cb334da10>

Part 4 - Making the predictions and evaluating the model

Predicting the Test set results

```
In [14]: y_pred=ann.predict(X_test)
```

```
4/4 ————— 0s 14ms/step
```

```
In [15]: from sklearn.metrics import r2_score
r2_score(y_test,y_pred)
```

Out[15]: 0.7992892662530684

```
In [ ]:
```

Assignment 2

```
In [108... from keras.datasets import imdb
from keras.preprocessing.sequence import pad_sequences
from keras.layers import Dense
from keras.models import Sequential
import numpy as np

In [109... (x_train,y_train),(x_test,y_test)=imdb.load_data(num_words=1000)

In [110... x_train, y_train = np.array(x_train), np.array(y_train)

In [111... max_len=250

In [112... x_train=pad_sequences(x_train,maxlen=max_len)
x_test=pad_sequences(x_test,maxlen=max_len)

In [113... model=Sequential()

In [114... model.add(Dense(units=100,activation='relu'))


In [115... model.add(Dense(units=50,activation='relu'))


In [116... # model.add(Dense(units=50,activation='relu'))


In [117... model.add(Dense(units=1,activation='sigmoid'))


In [118... model.compile(optimizer='adam',loss='binary_crossentropy',metrics=['accuracy'])


In [119... model.fit(x_train,y_train,epochs=30,batch_size=64)
```


Epoch 1/30
391/391  3s 2ms/step - accuracy: 0.5034 - loss: 18.6443


Epoch 2/30
391/391  1s 2ms/step - accuracy: 0.5186 - loss: 1.1738


Epoch 3/30
391/391  1s 2ms/step - accuracy: 0.5218 - loss: 0.7723


Epoch 4/30
391/391  1s 2ms/step - accuracy: 0.5329 - loss: 0.7365


Epoch 5/30
391/391  1s 2ms/step - accuracy: 0.5301 - loss: 0.7327


Epoch 6/30
391/391  1s 2ms/step - accuracy: 0.5347 - loss: 0.7282


Epoch 7/30
391/391  1s 2ms/step - accuracy: 0.5359 - loss: 0.7298


Epoch 8/30
391/391  1s 2ms/step - accuracy: 0.5483 - loss: 0.6916


Epoch 9/30
391/391  1s 2ms/step - accuracy: 0.5565 - loss: 0.7049


Epoch 10/30
391/391  1s 2ms/step - accuracy: 0.5803 - loss: 0.6753


Epoch 11/30
391/391  1s 2ms/step - accuracy: 0.5865 - loss: 0.6689


Epoch 12/30
391/391  1s 2ms/step - accuracy: 0.6154 - loss: 0.6448


Epoch 13/30
391/391  1s 2ms/step - accuracy: 0.6274 - loss: 0.6279


Epoch 14/30
391/391  1s 2ms/step - accuracy: 0.6563 - loss: 0.5957


Epoch 15/30
391/391  1s 2ms/step - accuracy: 0.6797 - loss: 0.5732


Epoch 16/30
391/391  1s 2ms/step - accuracy: 0.7063 - loss: 0.5389


Epoch 17/30
391/391  1s 2ms/step - accuracy: 0.7208 - loss: 0.5240


Epoch 18/30
391/391  1s 2ms/step - accuracy: 0.7506 - loss: 0.4859


Epoch 19/30
391/391  1s 2ms/step - accuracy: 0.7649 - loss: 0.4642


Epoch 20/30
391/391  1s 2ms/step - accuracy: 0.7861 - loss: 0.4308


Epoch 21/30
391/391  1s 2ms/step - accuracy: 0.7950 - loss: 0.4158


Epoch 22/30
391/391  1s 2ms/step - accuracy: 0.8183 - loss: 0.3790


Epoch 23/30
391/391  1s 2ms/step - accuracy: 0.8315 - loss: 0.3568


Epoch 24/30
391/391  1s 2ms/step - accuracy: 0.8469 - loss: 0.3323


Epoch 25/30
391/391  1s 2ms/step - accuracy: 0.8623 - loss: 0.3051

Epoch 26/30
391/391  1s 2ms/step - accuracy: 0.8697 - loss: 0.2922

Epoch 27/30
391/391  1s 2ms/step - accuracy: 0.8810 - loss: 0.2718

Epoch 28/30
391/391  1s 2ms/step - accuracy: 0.8907 - loss: 0.2521

Epoch 29/30
391/391  1s 2ms/step - accuracy: 0.8972 - loss: 0.2422

Epoch 30/30
391/391  1s 2ms/step - accuracy: 0.9033 - loss: 0.2225

Out[119... <keras.src.callbacks.history.History at 0x27b9f214e10>

In [120... `loss,acc=model.evaluate(x_test,y_test,batch_size=128)`

196/196  **1s** 2ms/step - accuracy: 0.5088 - loss: 2.1961

In []:

Assignment 3

```
In [38]: from tensorflow.keras.datasets import fashion_mnist
import tensorflow as tf
from tensorflow.keras.utils import to_categorical
```

```
In [2]: (x_train,y_train),(x_test,y_test)=fashion_mnist.load_data()
```

Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datasets/train-labels-idx1-ubyte.gz

29515/29515 ————— 0s 3us/step

Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datasets/train-images-idx3-ubyte.gz

26421880/26421880 ————— 13s 1us/step

Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datasets/t10k-labels-idx1-ubyte.gz

5148/5148 ————— 0s 1us/step

Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datasets/t10k-images-idx3-ubyte.gz

4422102/4422102 ————— 2s 0us/step

```
In [5]: x_train=x_train/255.0
x_test=x_test/255.0
```

```
In [ ]: x_train=x_train.reshape(-1,28,28,1)
x_test=x_test.reshape(-1,28,28,1)
```

```
In [39]: # One-hot encode the Labels
y_train = to_categorical(y_train,10)
y_test = to_categorical(y_test,10)
```

```
In [41]: from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Conv2D, MaxPooling2D, Flatten, Dense, Dropout

model = Sequential([
    Conv2D(32, (3, 3), activation='relu', input_shape=(28, 28, 1)),
    MaxPooling2D(2, 2),

    Conv2D(64, (3, 3), activation='relu'),
    MaxPooling2D(2, 2),

    Flatten(),
    Dropout(0.5),
    Dense(128, activation='relu'),
    Dense(10, activation='softmax')
])
```


C:\Users\acer\AppData\Roaming\Python\Python311\site-packages\keras\src\layers\convolutional\base_conv.py:107: UserWarning: Do not pass an `input_shape`/`input_dim` argument to a layer. When using Sequential models, prefer using an `Input(shape)` object as the first layer in the model instead.

```
super().__init__(activity_regularizer=activity_regularizer, **kwargs)
```


```
In [43]: model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accu
```

```
# Train
model.fit(x_train, y_train, epochs=10, batch_size=32, validation_data=(x_test, y_test))
```


Epoch 1/10

1875/1875  **19s** 9ms/step - accuracy: 0.8910 - loss: 0.2891 - val_accuracy: 0.8981 - val_loss: 0.2781


Epoch 2/10

1875/1875  **18s** 9ms/step - accuracy: 0.9017 - loss: 0.2636 - val_accuracy: 0.8910 - val_loss: 0.2799


Epoch 3/10

1875/1875  **17s** 9ms/step - accuracy: 0.9058 - loss: 0.2491 - val_accuracy: 0.9002 - val_loss: 0.2591


Epoch 4/10

1875/1875  **20s** 10ms/step - accuracy: 0.9126 - loss: 0.2321 - val_accuracy: 0.9128 - val_loss: 0.2381


Epoch 5/10

1875/1875  **17s** 9ms/step - accuracy: 0.9163 - loss: 0.2226 - val_accuracy: 0.9124 - val_loss: 0.2404


Epoch 6/10

1875/1875  **17s** 9ms/step - accuracy: 0.9201 - loss: 0.2106 - val_accuracy: 0.9099 - val_loss: 0.2392


Epoch 7/10

1875/1875  **17s** 9ms/step - accuracy: 0.9213 - loss: 0.2040 - val_accuracy: 0.9140 - val_loss: 0.2421


Epoch 8/10

1875/1875  **17s** 9ms/step - accuracy: 0.9233 - loss: 0.1984 - val_accuracy: 0.9119 - val_loss: 0.2370

Epoch 9/10


1875/1875  **17s** 9ms/step - accuracy: 0.9268 - loss: 0.1912 - val_accuracy: 0.9122 - val_loss: 0.2408

Epoch 10/10

1875/1875  **18s** 10ms/step - accuracy: 0.9308 - loss: 0.1795 - val_accuracy: 0.9134 - val_loss: 0.2407

Out[43]: <keras.src.callbacks.history.History at 0x21c28b8ec90>

```
In [44]: test_loss, test_accuracy = model.evaluate(x_test, y_test)
          print(f"Test Accuracy: {test_accuracy:.4f}")
```

313/313  **2s** 5ms/step - accuracy: 0.9144 - loss: 0.2440
Test Accuracy: 0.9134

In []: