```
from keras.models import Sequential
from keras.layers import Dense
from keras.optimizers import Adam

import numpy as np
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score, mean_squared_error
```

```
WARNING:tensorflow:From c:\Users\ADMIN\AppData\Local\Programs\Python\Python311\Lib\site-
        packages\keras\src\losses.py:2976: The name tf.losses.sparse_softmax_cross_entropy is de
        precated. Please use tf.compat.v1.losses.sparse_softmax_cross_entropy instead.
In [2]:
         dataset = pd.read csv('DelhiAQI.csv')
In [3]:
         dataSet = dataset.dropna()
         for i in dataSet.columns:
             if i != 'AQI':
                 dataSet[i] = (dataSet[i] - dataSet[i].mean()) / (dataSet[i].std())
         x = dataSet.drop('AQI', axis = 1)
         y = dataSet['AQI']
In [4]:
         # Add a fully connected layer with 32 neurons with sigmoid activation and glorot unifol
         # Add a fully connected layer layer with 16 neurons, relu activation and he uniform as
         # Add a fully a connected layer with ] neuron, relu activation function and he uniform
         # Use Adam optimizer with batch size 16, learning rate 0.01 and epochs set to 20.
         def model1(x_train, x_val, y_train, y_val):
             model = Sequential()
             model.add(Dense(32, activation = 'sigmoid', kernel_initializer = 'glorot_uniform'))
             model.add(Dense(16, activation = 'relu', kernel_initializer = 'he_uniform'))
             model.add(Dense(1, activation = 'relu', kernel_initializer = 'he_uniform'))
             model.compile(optimizer = Adam(learning_rate = 0.01), loss = 'mean_squared_error')
             model.fit(x_train, y_train, batch_size = 16, epochs = 20, validation_data = (x_val,
             return model
         count = 5
         avg mse = 0
         for i in range(count):
             x_train, x_test, y_train, y_test = train_test_split(x, y, test_size = 0.2, random_s
             x_train, x_val, y_train, y_val = train_test_split(x_train, y_train, test_size = 0.2
             model = model1(x_train, x_val, y_train, y_val)
             y_pred = model.predict(x_test)
             y_pred = np.round(y_pred)
             mse = mean_squared_error(y_test, y_pred)
             avg_mse += mse
             print('MSE:', mse)
         print('Average MSE:', avg_mse/count)
```

WARNING:tensorflow:From c:\Users\ADMIN\AppData\Local\Programs\Python\Python311\Lib\site-packages\keras\src\backend.py:873: The name tf.get_default_graph is deprecated. Please u se tf.compat.v1.get_default_graph instead.

WARNING:tensorflow:From c:\Users\ADMIN\AppData\Local\Programs\Python\Python311\Lib\site-packages\keras\src\utils\tf_utils.py:492: The name tf.ragged.RaggedTensorValue is deprec ated. Please use tf.compat.v1.ragged.RaggedTensorValue instead.

```
03.0957
Epoch 2/20
1.1687
Epoch 3/20
7.6670
Epoch 4/20
4.0859
Epoch 5/20
9.5879
Epoch 6/20
1.4678
Epoch 7/20
1.6914
Epoch 8/20
3.0593
Epoch 9/20
4.9087
Epoch 10/20
7.3730
Epoch 11/20
3.4844
Epoch 12/20
1.3958
Epoch 13/20
8.8076
Epoch 14/20
7.2649
Epoch 15/20
9.7886
Epoch 16/20
5.6965
Epoch 17/20
5.4182
Epoch 18/20
0.6631
Epoch 19/20
4.9155
```

```
Epoch 20/20
2.1331
94/94 [======= ] - 0s 882us/step
MSE: 2877.098666666668
Epoch 1/20
72.9368
Epoch 2/20
1.9680
Epoch 3/20
7.1572
Epoch 4/20
0.8904
Epoch 5/20
Epoch 6/20
6.2839
Epoch 7/20
5.2866
Epoch 8/20
6.5703
Epoch 9/20
5.6357
Epoch 10/20
1.9019
Epoch 11/20
5.1326
Epoch 12/20
6.5164
Epoch 13/20
2.7251
Epoch 14/20
3.0427
Epoch 15/20
5.0994
Epoch 16/20
6.8418
Epoch 17/20
5.3547
Epoch 18/20
7.4404
Epoch 19/20
```

```
0.8010
Epoch 20/20
0.2537
94/94 [======= ] - 0s 1ms/step
MSE: 2835.430666666666
Epoch 1/20
87.7444
Epoch 2/20
9.0623
Epoch 3/20
5.0273
Epoch 4/20
3.3496
Epoch 5/20
2.6743
Epoch 6/20
9.9751
Epoch 7/20
3.4456
Epoch 8/20
9.4553
Epoch 9/20
4.9788
Epoch 10/20
9.1509
Epoch 11/20
1.2097
Epoch 12/20
1.1367
Epoch 13/20
8.2256
Epoch 14/20
3.4185
Epoch 15/20
0.8801
Epoch 16/20
3.0732
Epoch 17/20
2.1396
Epoch 18/20
```

```
2.0293
Epoch 19/20
8.2400
Epoch 20/20
7.8345
94/94 [======= ] - 0s 1ms/step
MSE: 2847.651666666666
Epoch 1/20
97.6963
Epoch 2/20
8.7739
Epoch 3/20
1.3760
Epoch 4/20
0.9177
Epoch 5/20
7.6963
Epoch 6/20
3.5063
Epoch 7/20
5.0623
Epoch 8/20
2.3604
Epoch 9/20
4.7356
Epoch 10/20
2.3997
Epoch 11/20
1.0684
Epoch 12/20
1.1499
Epoch 13/20
2.0781
Epoch 14/20
6.4431
Epoch 15/20
5.2974
Epoch 16/20
0.9666
Epoch 17/20
4.9619
```

```
Epoch 18/20
3.3359
Epoch 19/20
0.2034
Epoch 20/20
6.3879
94/94 [======== ] - 0s 947us/step
MSE: 2808.517666666666
Epoch 1/20
44.0952
Epoch 2/20
7.1943
Epoch 3/20
Epoch 4/20
7.9487
Epoch 5/20
7.1772
Epoch 6/20
4.7903
Epoch 7/20
2.8643
Epoch 8/20
2.8796
Epoch 9/20
3.2529
Epoch 10/20
4.3367
Epoch 11/20
1.3489
Epoch 12/20
5.8213
Epoch 13/20
2.5522
Epoch 14/20
3.1187
Epoch 15/20
7.6545
Epoch 16/20
7.4702
Epoch 17/20
```

```
8.0186
      Epoch 18/20
      8.8293
      Epoch 19/20
      6.7502
      Epoch 20/20
      0.2629
      94/94 [======== ] - 0s 1ms/step
      MSE: 2801.200666666666
      Average MSE: 2833.979866666667
In [7]:
       # Add a fully connected layer with 32 neurons with sigmoid activation and glorot unifor
       # Add a fully connected layer layer with 8 a neurons, sigmoid activation and glorot nor
       # Add a fully a connected layer ith neuron, relu activation function and he uniform as
       # Use Adam optimizer with batch size 8. Learning rate 0.01 and epochs set to 20.
       # Extract the features from second last fully connected layer (having 8 neurons) and mo
       from keras.models import Model
       from sklearn.svm import SVR
       def model2(x_train, x_val, y_train, y_val):
          model = Sequential()
          model.add(Dense(32, activation = 'sigmoid', kernel_initializer = 'glorot_uniform'))
          model.add(Dense(8, activation = 'sigmoid', kernel_initializer = 'glorot_normal'))
          model.add(Dense(1, activation = 'relu', kernel_initializer = 'he_uniform'))
          model.compile(optimizer = Adam(learning_rate = 0.01), loss = 'mean_squared_error')
          model.fit(x_train, y_train, batch_size = 8, epochs = 20, validation_data = (x_val,
          model = Model(inputs = model.input, outputs = model.get_layer(index = 2).output)
          return model
       count = 5
       avg mse = 0
       for i in range(count):
          x_train, x_test, y_train, y_test = train_test_split(x, y, test_size = 0.2, random_s
          x_train, x_val, y_train, y_val = train_test_split(x_train, y_train, test_size = 0.2
          model = model2(x_train, x_val, y_train, y_val)
          x_train = model.predict(x_train)
          x val = model.predict(x_val)
          x_test = model.predict(x_test)
          model = SVR()
          model.fit(x_train, y_train)
          y_pred = model.predict(x_test)
          y_pred = np.round(y_pred)
          mse = mean_squared_error(y_test, y_pred)
          avg mse += mse
          print('MSE:', mse)
       print('Average MSE:', avg_mse/count)
      Epoch 1/20
      50657.1992
```

```
16134.6904
Epoch 4/20
9656.1211
Epoch 5/20
164.1860
Epoch 6/20
480.5068
Epoch 7/20
808.3450
Epoch 8/20
491.0483
Epoch 9/20
400.2173
Epoch 10/20
165.8943
Epoch 11/20
063.8867
Epoch 12/20
004.0940
Epoch 13/20
910.4946
Epoch 14/20
819.6416
Epoch 15/20
835.8118
Epoch 16/20
823.4824
Epoch 17/20
724.5962
Epoch 18/20
708.2913
Epoch 19/20
685.8408
Epoch 20/20
694.1152
282/282 [========= ] - 1s 2ms/step
94/94 [========] - 0s 2ms/step
94/94 [=======] - 0s 2ms/step
MSE: 2948.92066666667
Epoch 1/20
50232.2578
```

```
Epoch 2/20
28345.7617
Epoch 3/20
15929.0508
Epoch 4/20
8151.7500
Epoch 5/20
242.5938
Epoch 6/20
058.0747
Epoch 7/20
634.0183
Epoch 8/20
336.7974
Epoch 9/20
135.6699
Epoch 10/20
991.7917
Epoch 11/20
912.6824
Epoch 12/20
811.8821
Epoch 13/20
868.7400
Epoch 14/20
761.2046
Epoch 15/20
824.2510
Epoch 16/20
762.8257
Epoch 17/20
633.4331
Epoch 18/20
685.6045
Epoch 19/20
608.6250
Epoch 20/20
670.0762
282/282 [========= ] - 1s 2ms/step
94/94 [=======] - 0s 2ms/step
94/94 [======== ] - 0s 2ms/step
```

```
MSE: 2927.458
Epoch 1/20
54451.3164
Epoch 2/20
33438.0898
Epoch 3/20
20211.9902
Epoch 4/20
12976.6123
Epoch 5/20
7988.6826
Epoch 6/20
533.5049
Epoch 7/20
1125/1125 [============] - 3s 2ms/step - loss: 5141.5225 - val_loss: 4
310.1387
Epoch 8/20
774.9807
Epoch 9/20
497.0000
Epoch 10/20
334.8884
Epoch 11/20
147.7681
Epoch 12/20
037.4814
Epoch 13/20
945.1704
Epoch 14/20
954.4187
Epoch 15/20
865.0393
Epoch 16/20
882.7854
Epoch 17/20
796.6370
Epoch 18/20
758.7383
Epoch 19/20
765.8882
Epoch 20/20
```

```
769.4978
94/94 [=======] - 0s 1ms/step
94/94 [======= ] - 0s 1ms/step
MSE: 2966.337666666667
Epoch 1/20
50210.9766
Epoch 2/20
27811.3047
Epoch 3/20
15842.2988
Epoch 4/20
9330.0820
Epoch 5/20
004.6406
Epoch 6/20
437.7559
Epoch 7/20
806.6187
Epoch 8/20
428.4897
Epoch 9/20
288.4534
Epoch 10/20
178.8091
Epoch 11/20
037.0840
Epoch 12/20
948.4126
Epoch 13/20
959.2700
Epoch 14/20
874.8994
Epoch 15/20
917.8752
Epoch 16/20
868.1797
Epoch 17/20
750.0330
Epoch 18/20
780.0659
Epoch 19/20
```

```
726.6799
Epoch 20/20
727.4819
282/282 [========== ] - 1s 2ms/step
94/94 [======== ] - 0s 2ms/step
94/94 [=======] - 0s 3ms/step
MSE: 2992.367333333333
Epoch 1/20
87336.4844
Epoch 2/20
87336.4844
Epoch 3/20
87336.4844
Epoch 4/20
87336.4844
Epoch 5/20
87336.4844
Epoch 6/20
87336.4844
Epoch 7/20
87336.4844
Epoch 8/20
87336.4844
Epoch 9/20
87336.4844
Epoch 10/20
87336.4844
Epoch 11/20
87336.4844
Epoch 12/20
87336.4844
Epoch 13/20
87336.4844
Epoch 14/20
87336.4844
Epoch 15/20
87336.4844
Epoch 16/20
87336.4844
Epoch 17/20
87336.4844
```

```
# Extract the deep features from Model- 1 (from 2nd layer) and Model- 2 (from 2nd layer
count = 5
avg_mse = 0
for i in range(count):
    x_train, x_test, y_train, y_test = train_test_split(x, y, test_size = 0.2, random_s
    x_train, x_val, y_train, y_val = train_test_split(x_train, y_train, test_size = 0.2
    model_1 = model1(x_train, x_val, y_train, y_val)
    model_2 = model2(x_train, x_val, y_train, y_val)
    x_train1 = model_1.predict(x_train)
   x_val1 = model_1.predict(x_val)
   x_test1 = model_1.predict(x_test)
   x_train2 = model_2.predict(x_train)
   x_val2 = model_2.predict(x_val)
   x_test2 = model_2.predict(x_test)
   x_train = np.hstack((x_train1, x_train2))
   x_{val} = np.hstack((x_{val1}, x_{val2}))
   x_test = np.hstack((x_test1, x_test2))
   model = SVR()
   model.fit(x_train, y_train)
   y_pred = model.predict(x_test)
   y_pred = np.round(y_pred)
   mse = mean_squared_error(y_test, y_pred)
    avg_mse += mse
    print('MSE:', mse)
print('Average MSE:', avg_mse/count)
```

```
Epoch 1/20
13.5969
Epoch 2/20
3.9937
Epoch 3/20
5.3091
Epoch 4/20
Epoch 5/20
5.5674
Epoch 6/20
```

```
9.1362
Epoch 7/20
1.2363
Epoch 8/20
7.0530
Epoch 9/20
9.2288
Epoch 10/20
5.5522
Epoch 11/20
3.9229
Epoch 12/20
7.5344
Epoch 13/20
4.5386
Epoch 14/20
3.0515
Epoch 15/20
7.9387
Epoch 16/20
7.5518
Epoch 17/20
2.5439
Epoch 18/20
4.1973
Epoch 19/20
8.4656
Epoch 20/20
6.9724
Epoch 1/20
50073.3242
Epoch 2/20
28332.6289
Epoch 3/20
16141.3164
Epoch 4/20
9785.9482
Epoch 5/20
145.6099
Epoch 6/20
```

```
532.0371
Epoch 7/20
939.8396
Epoch 8/20
547.3679
Epoch 9/20
295.5627
Epoch 10/20
171.7029
Epoch 11/20
067.6853
Epoch 12/20
973.3848
Epoch 13/20
949.5757
Epoch 14/20
847.2585
Epoch 15/20
758.5303
Epoch 16/20
766.0891
Epoch 17/20
769.8650
Epoch 18/20
828.3940
Epoch 19/20
719.2419
Epoch 20/20
701.8008
282/282 [========== - 1s 2ms/step
94/94 [======== ] - 0s 2ms/step
94/94 [========] - 0s 2ms/step
282/282 [========= ] - 1s 3ms/step
94/94 [=======] - 0s 2ms/step
94/94 [========] - 0s 2ms/step
MSE: 2840.039666666666
Epoch 1/20
01.6831
Epoch 2/20
9.7170
Epoch 3/20
4.9854
Epoch 4/20
```

```
6.0496
Epoch 5/20
7.4578
Epoch 6/20
8.7346
Epoch 7/20
5.7668
Epoch 8/20
6.8357
Epoch 9/20
5.4473
Epoch 10/20
Epoch 11/20
44.0706
Epoch 12/20
6.5364
Epoch 13/20
3.5859
Epoch 14/20
4.2368
Epoch 15/20
4.7844
Epoch 16/20
2.9443
Epoch 17/20
7.6309
Epoch 18/20
9.4700
Epoch 19/20
0.7129
Epoch 20/20
3.0806
Epoch 1/20
42637.1055
Epoch 2/20
20526.3418
Epoch 3/20
10737.6436
Epoch 4/20
```

```
256.4331
Epoch 5/20
428.6177
Epoch 6/20
718.5066
Epoch 7/20
404.4971
Epoch 8/20
252.9097
Epoch 9/20
153.7886
Epoch 10/20
030.5603
Epoch 11/20
2974.1443
Epoch 12/20
903.0071
Epoch 13/20
894.6982
Epoch 14/20
844.1667
Epoch 15/20
822.2900
Epoch 16/20
832.5339
Epoch 17/20
755.4551
Epoch 18/20
764.3818
Epoch 19/20
762.7161
Epoch 20/20
741.1724
282/282 [========= ] - 1s 2ms/step
94/94 [=======] - Os 2ms/step
94/94 [======== ] - 0s 2ms/step
282/282 [========== ] - 1s 1ms/step
94/94 [======== ] - 0s 1ms/step
94/94 [=======] - 0s 1ms/step
MSE: 2869.948666666667
Epoch 1/20
15.7097
```

```
Epoch 2/20
2.0037
Epoch 3/20
2.6077
Epoch 4/20
4.2207
Epoch 5/20
2.1860
Epoch 6/20
7.2908
Epoch 7/20
4.4041
Epoch 8/20
1.2263
Epoch 9/20
2.6174
Epoch 10/20
9.8567
Epoch 11/20
4.1924
Epoch 12/20
0.1736
Epoch 13/20
6.6572
Epoch 14/20
3.9666
Epoch 15/20
5.3772
Epoch 16/20
5.0132
Epoch 17/20
0.3379
Epoch 18/20
9.6577
Epoch 19/20
3.9912
Epoch 20/20
0.4131
Epoch 1/20
87336.4844
```

```
Epoch 2/20
87336.4844
Epoch 3/20
87336.4844
Epoch 4/20
87336.4844
Epoch 5/20
87336.4844
Epoch 6/20
87336.4844
Epoch 7/20
87336.4844
Epoch 8/20
87336.4844
Epoch 9/20
87336.4844
Epoch 10/20
87336.4844
Epoch 11/20
87336.4844
Epoch 12/20
87336.4844
Epoch 13/20
87336.4844
Epoch 14/20
87336.4844
Epoch 15/20
87336.4844
Epoch 16/20
87336.4844
Epoch 17/20
87336.4844
Epoch 18/20
87336.4844
Epoch 19/20
87336.4844
Epoch 20/20
87336.4844
282/282 [========== ] - 1s 2ms/step
94/94 [======= ] - 0s 1ms/step
94/94 [======== ] - 0s 2ms/step
```

```
282/282 [========= ] - 0s 1ms/step
94/94 [======= ] - 0s 1ms/step
94/94 [======= ] - 0s 1ms/step
MSE: 2919.423
Epoch 1/20
17.0916
Epoch 2/20
0.4141
Epoch 3/20
9.7646
Epoch 4/20
8.3823
Epoch 5/20
3.3201
Epoch 6/20
0.1990
Epoch 7/20
4.3684
Epoch 8/20
6.0823
Epoch 9/20
3.4463
Epoch 10/20
8.5557
Epoch 11/20
5.3506
Epoch 12/20
0.4287
Epoch 13/20
7.0083
Epoch 14/20
8.8372
Epoch 15/20
1.1765
Epoch 16/20
4.1938
Epoch 17/20
3.0208
Epoch 18/20
3.4648
Epoch 19/20
```

```
7.9634
Epoch 20/20
7.2178
Epoch 1/20
46309.5234
Epoch 2/20
24107.1582
Epoch 3/20
13370.1377
Epoch 4/20
7599.3828
Epoch 5/20
082.8848
Epoch 6/20
1125/1125 [===========] - 3s 3ms/step - loss: 4697.3584 - val_loss: 3
989.8762
Epoch 7/20
577.7024
Epoch 8/20
350.4854
Epoch 9/20
198.1641
Epoch 10/20
061.4373
Epoch 11/20
961.3354
Epoch 12/20
947.3132
Epoch 13/20
867.0073
Epoch 14/20
020.6851
Epoch 15/20
816.6604
Epoch 16/20
822.6982
Epoch 17/20
827.4006
Epoch 18/20
799.6377
Epoch 19/20
```

```
779.2388
Epoch 20/20
798.0347
282/282 [========== ] - 1s 2ms/step
94/94 [=======] - 0s 1ms/step
94/94 [=======] - 0s 2ms/step
282/282 [========= ] - 0s 1ms/step
94/94 [=======] - 0s 1ms/step
94/94 [========] - 0s 2ms/step
MSE: 2875.579666666665
Epoch 1/20
5.0298
Epoch 2/20
7.3196
Epoch 3/20
Epoch 4/20
7.2114
Epoch 5/20
4.9131
Epoch 6/20
3.5000
Epoch 7/20
3.4526
Epoch 8/20
8.7957
Epoch 9/20
7.4260
Epoch 10/20
0.7043
Epoch 11/20
5.7546
Epoch 12/20
8.8145
Epoch 13/20
4.5229
Epoch 14/20
9.4802
Epoch 15/20
9.0454
Epoch 16/20
5.9275
Epoch 17/20
```

```
6.0601
Epoch 18/20
1.4763
Epoch 19/20
2.4824
Epoch 20/20
4.2805
Epoch 1/20
46533.6367
Epoch 2/20
23986.9902
Epoch 3/20
13303.3682
Epoch 4/20
7537.4219
Epoch 5/20
193.8081
Epoch 6/20
076.9893
Epoch 7/20
533.1426
Epoch 8/20
321.0791
Epoch 9/20
166.3474
Epoch 10/20
030.4473
Epoch 11/20
970.6677
Epoch 12/20
945.1526
Epoch 13/20
870.8040
Epoch 14/20
861.6777
Epoch 15/20
817.0906
Epoch 16/20
759.7041
Epoch 17/20
```

```
748.5825
       Epoch 18/20
       769.0605
       Epoch 19/20
       780.4387
       Epoch 20/20
       694.2515
       94/94 [========] - 0s 1ms/step
       94/94 [========] - 0s 1ms/step
       282/282 [========= ] - 0s 1ms/step
       94/94 [========] - 0s 1ms/step
       94/94 [======== ] - 0s 1ms/step
       MSE: 2820.66333333333334
       Average MSE: 2865.130866666667
In [16]:
       # Extract the deep features from Model-1 (from 2nd layer) and Model- 2 (from 2nd layer)
       from sklearn.decomposition import PCA
       count = 5
       avg_mse = 0
       for i in range(count):
          x_train, x_test, y_train, y_test = train_test_split(x, y, test_size = 0.2, random_s
           x_train, x_val, y_train, y_val = train_test_split(x_train, y_train, test_size = 0.2
           model_1 = model1(x_train, x_val, y_train, y_val)
           model_2 = model2(x_train, x_val, y_train, y_val)
           model_1 = Model(inputs = model_1.input, outputs = model_1.get_layer(index = 2).outp
          model_2 = Model(inputs = model_2.input, outputs = model_2.get_layer(index = 2).outp
          x_train1 = model_1.predict(x_train)
          x_val1 = model_1.predict(x_val)
          x test1 = model 1.predict(x test)
          x_train2 = model_2.predict(x_train)
          x_{val2} = model_{2.predict(x_{val})}
          x_test2 = model_2.predict(x_test)
          x_train = np.hstack((x_train1, x_train2))
          x_val = np.hstack((x_val1, x_val2))
          x_test = np.hstack((x_test1, x_test2))
          mse = []
          pca = PCA(n\_components = 8)
          pca.fit(x_train)
          x_train_pca = pca.transform(x_train)
          x_val_pca = pca.transform(x_val)
          x_test_pca = pca.transform(x_test)
           model = SVR()
           model.fit(x_train_pca, y_train)
          y pred = model.predict(x_test_pca)
           y_pred = np.round(y_pred)
           mse.append(mean_squared_error(y_test, y_pred))
       print('MSE for 8 components:', mse)
```

```
Epoch 1/20
29.8757
Epoch 2/20
5.5012
Epoch 3/20
5.2224
Epoch 4/20
9.1846
Epoch 5/20
4.3613
Epoch 6/20
5.7170
Epoch 7/20
9.2092
Epoch 8/20
7.1870
Epoch 9/20
7.6816
Epoch 10/20
3.0334
Epoch 11/20
3.6438
Epoch 12/20
3.6799
Epoch 13/20
8.1621
Epoch 14/20
6.6042
Epoch 15/20
7.4451
Epoch 16/20
2.1758
Epoch 17/20
1.0723
Epoch 18/20
9.4424
Epoch 19/20
6.2637
Epoch 20/20
1.5098
```

```
Epoch 1/20
55767.7383
Epoch 2/20
33730.2070
Epoch 3/20
20332.0449
Epoch 4/20
12980.3320
Epoch 5/20
8014.5767
Epoch 6/20
538.8364
Epoch 7/20
350.5479
Epoch 8/20
748.3669
Epoch 9/20
508.8459
Epoch 10/20
230.0811
Epoch 11/20
098.3623
Epoch 12/20
010.2346
Epoch 13/20
949.6387
Epoch 14/20
903.2646
Epoch 15/20
927.9819
Epoch 16/20
833.6243
Epoch 17/20
849.8396
Epoch 18/20
768.7876
Epoch 19/20
741.6062
Epoch 20/20
762.8892
```

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282/282 [========= ] - 0s 1ms/step
94/94 [======= ] - 0s 1ms/step
94/94 [======= ] - 0s 1ms/step
282/282 [========= ] - 0s 1ms/step
94/94 [=======] - 0s 1ms/step
94/94 [=======] - 0s 1ms/step
72.1624
Epoch 2/20
5.4351
Epoch 3/20
6.1333
Epoch 4/20
0.6506
Epoch 5/20
4.8887
Epoch 6/20
2.9463
Epoch 7/20
6.4238
Epoch 8/20
2.5300
Epoch 9/20
0.3047
Epoch 10/20
1.9727
Epoch 11/20
3.1165
Epoch 12/20
8.5432
Epoch 13/20
3.9263
Epoch 14/20
2.0667
Epoch 15/20
8.5674
Epoch 16/20
1.5762
Epoch 17/20
1.9497
Epoch 18/20
0.6140
```

```
Epoch 19/20
5.0562
Epoch 20/20
9.8953
Epoch 1/20
46645.7773
Epoch 2/20
24315.4727
Epoch 3/20
13474.3652
Epoch 4/20
7652.5371
Epoch 5/20
130.5205
Epoch 6/20
028.8154
Epoch 7/20
623.7336
Epoch 8/20
398.1074
Epoch 9/20
174.8022
Epoch 10/20
070.0427
Epoch 11/20
016.9480
Epoch 12/20
030.9163
Epoch 13/20
865.9043
Epoch 14/20
842.9380
Epoch 15/20
828.6106
Epoch 16/20
879.3540
Epoch 17/20
784.2910
Epoch 18/20
742.8259
```

```
Epoch 19/20
760.0767
Epoch 20/20
854.3057
94/94 [======= ] - 0s 2ms/step
94/94 [=======] - 0s 2ms/step
282/282 [=========] - 1s 2ms/step
94/94 [=======] - 0s 2ms/step
94/94 [======== ] - 0s 2ms/step
Epoch 1/20
03.7708
Epoch 2/20
0.8459
Epoch 3/20
0.2644
Epoch 4/20
4.6689
Epoch 5/20
7.6768
Epoch 6/20
0.8369
Epoch 7/20
1.3516
Epoch 8/20
8.7214
Epoch 9/20
4.2766
Epoch 10/20
9.1484
Epoch 11/20
1.8650
Epoch 12/20
7.2009
Epoch 13/20
8.1479
Epoch 14/20
4.0088
Epoch 15/20
9.0186
Epoch 16/20
6.7603
```

```
Epoch 17/20
9.4846
Epoch 18/20
6.1956
Epoch 19/20
5.4048
Epoch 20/20
6.3943
Epoch 1/20
52806.8281
Epoch 2/20
28430.0449
Epoch 3/20
16118.9307
Epoch 4/20
9609.5166
Epoch 5/20
150.6899
Epoch 6/20
497.0005
Epoch 7/20
792.8979
Epoch 8/20
478.3047
Epoch 9/20
241.9934
Epoch 10/20
095.3406
Epoch 11/20
991.7507
Epoch 12/20
896.9036
Epoch 13/20
862.7734
Epoch 14/20
804.3186
Epoch 15/20
774.5906
Epoch 16/20
728.1887
```

```
Epoch 17/20
713.1819
Epoch 18/20
735.7505
Epoch 19/20
667.7285
Epoch 20/20
662.4702
282/282 [========== ] - 1s 2ms/step
94/94 [=======] - 0s 2ms/step
94/94 [=======] - 0s 2ms/step
282/282 [========== ] - 1s 2ms/step
94/94 [======= ] - 0s 2ms/step
94/94 [=======] - 0s 1ms/step
Epoch 1/20
65.1987
Epoch 2/20
8.2544
Epoch 3/20
9.4382
Epoch 4/20
8.3701
Epoch 5/20
6.9973
Epoch 6/20
1.9573
Epoch 7/20
7.4802
Epoch 8/20
7,4089
Epoch 9/20
2.8904
Epoch 10/20
2.7810
Epoch 11/20
8.9800
Epoch 12/20
2.4199
Epoch 13/20
5.1028
Epoch 14/20
6.3879
```

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Epoch 15/20
8.7227
Epoch 16/20
8.3274
Epoch 17/20
5.3335
Epoch 18/20
7.4910
Epoch 19/20
1.8997
Epoch 20/20
8.8745
Epoch 1/20
87336.4844
Epoch 2/20
87336.4844
Epoch 3/20
87336.4844
Epoch 4/20
87336.4844
Epoch 5/20
87336.4844
Epoch 6/20
87336.4844
Epoch 7/20
87336.4844
Epoch 8/20
87336.4844
Epoch 9/20
87336.4844
Epoch 10/20
87336.4844
Epoch 11/20
87336.4844
Epoch 12/20
87336.4844
Epoch 13/20
87336.4844
Epoch 14/20
87336.4844
```

```
Epoch 15/20
87336.4844
Epoch 16/20
87336.4844
Epoch 17/20
87336.4844
Epoch 18/20
87336.4844
Epoch 19/20
87336.4844
Epoch 20/20
87336.4844
282/282 [=========] - 1s 2ms/step
94/94 [======== ] - 0s 2ms/step
94/94 [======== ] - 0s 2ms/step
282/282 [========== ] - 1s 2ms/step
94/94 [=======] - 0s 2ms/step
94/94 [======= ] - 0s 2ms/step
Epoch 1/20
72.3904
Epoch 2/20
2.5969
Epoch 3/20
7.9236
Epoch 4/20
9.7576
Epoch 5/20
6.0623
Epoch 6/20
8.4106
Epoch 7/20
5.6038
Epoch 8/20
6.7476
Epoch 9/20
1.4663
Epoch 10/20
5.2534
Epoch 11/20
5.2561
Epoch 12/20
7.6079
```

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Epoch 13/20
5.6641
Epoch 14/20
3.5813
Epoch 15/20
7.9094
Epoch 16/20
1.7388
Epoch 17/20
9.0317
Epoch 18/20
2.4094
Epoch 19/20
2.1584
Epoch 20/20
6.2744
Epoch 1/20
46614.2461
Epoch 2/20
24407.0488
Epoch 3/20
13506.9482
Epoch 4/20
7690.3521
Epoch 5/20
086.7085
Epoch 6/20
070.4006
Epoch 7/20
614.0610
Epoch 8/20
331.8704
Epoch 9/20
167.6968
Epoch 10/20
195.1526
Epoch 11/20
991.8018
Epoch 12/20
926.5940
```

```
Epoch 13/20
   942.6768
   Epoch 14/20
   853.0276
   Epoch 15/20
   828.3384
   Epoch 16/20
   851.3240
   Epoch 17/20
   805.1057
   Epoch 18/20
   729.4609
   Epoch 19/20
   853.1448
   Epoch 20/20
   782.6743
   94/94 [======== ] - 0s 2ms/step
   94/94 [========] - 0s 2ms/step
   282/282 [========= ] - 1s 2ms/step
   94/94 [========] - 0s 2ms/step
   94/94 [======== ] - 0s 2ms/step
   MSE for 8 components: [2963.648333333333]
In [ ]:
   # Draw conclusions on the best model among the above four models for predicting the Dell
    Average MSE for model 1: 2833.979866666667
    Average MSE for model 2 : 4231.987533333333
    Average MSE for model 3 : 2865.130866666667
    Average MSE for model 4: 2963.648333333333
    The best model is Model 1.
```