```
In [1]:
          from keras.models import Sequential
          from keras.layers import Dense
          from keras.optimizers import Adam
          import pandas as pd
          import numpy as np
          from sklearn.model selection import train test split
          from sklearn.metrics import accuracy_score, mean_squared_error
         WARNING:tensorflow:From C:\Users\Laptop\AppData\Local\Packages\PythonSoftwareFoundation.
         Python.3.11_qbz5n2kfra8p0\LocalCache\local-packages\Python311\site-packages\keras\src\lo
         sses.py:2976: The name tf.losses.sparse_softmax_cross_entropy is deprecated. Please use
         tf.compat.v1.losses.sparse_softmax_cross_entropy instead.
In [10]:
          dataSet = pd.read_csv('DelhiAQI.csv')
 In [3]:
          print(dataSet.head())
          print(dataSet.shape)
             PM2.5
                     PM10
                               NO
                                     NO2
                                             NOx
                                                    NH3
                                                            CO SO2
                                                                        03 Benzene \
         0 454.58 935.18
                            81.52 41.78 187.66
                                                  27.54
                                                          9.29 3.41 54.94
                                                                              25.24
         1 440.44 935.18
                           70.80 43.46 176.83
                                                  27.72 13.28 3.88 50.53
                                                                               23.10
         2 409.09 935.18 132.46 41.19 141.02
                                                                              19.04
                                                  28.94 29.67 2.83 19.33
         3 436.12 935.18
                            84.78
                                   39.55 102.84
                                                  29.30
                                                         21.76 4.33 20.08
                                                                              13.99
         4 415.88 976.99
                            60.24 37.41 80.12 30.84 26.19 6.17 16.00
                                                                              11.14
            Toluene Xylene AQI
         0
              58.57
                    13.80 653
              49.37
                    15.63 645
         1
                    17.18 532
         2
              38.94
         3
              27.53
                    16.82 561
              21.99
                     14.29 567
         (15000, 13)
In [11]:
          # normalizing the data
          dataSetNorm1 = dataSet.dropna()
          for i in dataSetNorm1.columns:
              if i != 'AQI':
                  # use the median and std of the training data to normalize (Z-score scaling)
                  dataSetNorm1[i] = (dataSetNorm1[i] - dataSetNorm1[i].mean()) / dataSetNorm1[i].
          X = dataSetNorm1.drop('AQI', axis=1)
          y = dataSetNorm1['AQI']
In [12]:
          # Implement deep multilayer perceptron neural network Model-I : Add a fully connected l
          # and glorot uniform kernel initializer. Add a fully connected layer layer with 16
          # neurons, relu activation and he uniform as kernel initializer. Add a fully
          # connected layer with 1 neuron, relu activation function and he uniform as
          # kernel initializer. Use Adam optimizer with batch size 16, learning rate 0.01
          # and epochs set to 20.
          # Test using root mean squared error as loss function.
          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', batch_
    model.fit(X_train, y_train, validation_data=(X_val, y_val))
    return model
count = 5
avg_mse = 0
for i in range(count):
    # Split data to 60% training 20% validation and 20% testing
   X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_sta
   X_train, X_val, y_train, y_val = train_test_split(X_train, y_train, test_size=0.25,
   model = model1(X_train, X_val, y_train, y_val)
   y_pred = model.predict(X_test)
   print('Model:', i+1)
   y_pred = np.round(y_pred)
   mse = mean_squared_error(y_test, y_pred)
    print('Mean Squared Error:', mse)
    print("----")
    avg_mse += mse
print('Average Mean Squared Error:', avg_mse/count)
```

```
Epoch 1/20
69.8613
Epoch 2/20
002.4209
Epoch 3/20
3019.6687
Epoch 4/20
909.5486
Epoch 5/20
874,4570
Epoch 6/20
827.1260
Epoch 7/20
841.1707
Epoch 8/20
938.6287
Epoch 9/20
818.6968
Epoch 10/20
849.4070
Epoch 11/20
```

```
806.5847
Epoch 12/20
3.4207
Epoch 13/20
695.7910
Epoch 14/20
7,2566
Epoch 15/20
696.4504
Epoch 16/20
701.8970
Epoch 17/20
Epoch 18/20
2.7073
Epoch 19/20
636.5217
Epoch 20/20
1.7043
94/94 [=======] - 0s 635us/step
Model: 1
Mean Squared Error: 2833.144
-----
Epoch 1/20
91.4583
Epoch 2/20
6.0957
Epoch 3/20
5.0986
Epoch 4/20
2.5991
Epoch 5/20
864.8022
Epoch 6/20
920.1394
Epoch 7/20
788.0869
Epoch 8/20
758.7078
Epoch 9/20
720.4287
```

```
Epoch 10/20
730.7773
Epoch 11/20
701.7969
Epoch 12/20
755.0935
Epoch 13/20
6.9539
Epoch 14/20
3.2146
Epoch 15/20
9.4521
Epoch 16/20
7.4136
Epoch 17/20
709.9001
Epoch 18/20
634.8906
Epoch 19/20
5.9331
Epoch 20/20
661.7500
94/94 [======= ] - 0s 617us/step
Model: 2
Mean Squared Error: 2813.283
Epoch 1/20
04.3887
Epoch 2/20
072.2812
Epoch 3/20
991.6814
Epoch 4/20
094.5723
Epoch 5/20
924.0828
Epoch 6/20
3.2686
Epoch 7/20
839.5742
Epoch 8/20
```

```
874.7520
Epoch 9/20
8.4583
Epoch 10/20
3.4084
Epoch 11/20
750.9062
Epoch 12/20
744.0894
Epoch 13/20
6.3794
Epoch 14/20
744.6260
Epoch 15/20
5.9426
Epoch 16/20
649.8630
Epoch 17/20
684.5811
Epoch 18/20
667.4841
Epoch 19/20
2.9536
Epoch 20/20
4.6270
94/94 [=======] - 0s 650us/step
Model: 3
Mean Squared Error: 2827.5693333333334
Epoch 1/20
30.0537
Epoch 2/20
217.1248
Epoch 3/20
996.3950
Epoch 4/20
906.5532
Epoch 5/20
919.6509
Epoch 6/20
891.7434
Epoch 7/20
```

```
827.2297
Epoch 8/20
803.2207
Epoch 9/20
754.5618
Epoch 10/20
738.8167
Epoch 11/20
1.6121
Epoch 12/20
695.4036
Epoch 13/20
663.1221
Epoch 14/20
667.6621
Epoch 15/20
693.4387
Epoch 16/20
651.6179
Epoch 17/20
659.5645
Epoch 18/20
6.0557
Epoch 19/20
3.6868
Epoch 20/20
595.0212
94/94 [=======] - 0s 675us/step
Model: 4
Mean Squared Error: 2814.11
Epoch 1/20
44.6516
Epoch 2/20
5.6023
Epoch 3/20
953.8066
Epoch 4/20
932.8113
Epoch 5/20
826.3479
```

```
Epoch 6/20
563/563 [===========] - 1s 978us/step - loss: 2958.3760 - val_loss: 3
035.8328
Epoch 7/20
785.8027
828.3806
Epoch 9/20
748.5144
Epoch 10/20
772.7317
Epoch 11/20
740.9192
Epoch 12/20
1.5090
Epoch 13/20
751.8921
Epoch 14/20
642.9124
Epoch 15/20
636.0046
Epoch 16/20
669.4751
Epoch 17/20
659.5439
Epoch 18/20
629.3364
Epoch 19/20
617.2151
Epoch 20/20
633.6499
Model: 5
Mean Squared Error: 2828.3056666666666
-----
Average Mean Squared Error: 2823.2824
```

```
In [21]:
```

```
# Model-2: Add a fully connected layer with 32 neurons with sigmoid activation # and glorot uniform kernel initializer. Add a fully connected layer layer with 8 # neurons, sigmoid activation and glorot normal as kernel initializer. Add a fully # connected layer with 1 neuron, relu activation function and he uniform as # kernel initializer. 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 model it using a Support Vector regressor.
```

```
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, y_val)
    intermediate_layer_model = Model(inputs=model.input, outputs=model.layers[-2].output
    intermediate output = intermediate layer model.predict(X train)
    svr = SVR()
    svr.fit(intermediate_output, y_train)
    return model, svr
count = 5
avg_mse = 0
for i in range(count):
    # Split data to 60% training 20% validation and 20% testing
    X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_sta
    X_train, X_val, y_train, y_val = train_test_split(X_train, y_train, test_size=0.25,
    model, svr = model2(X_train, X_val, y_train, y_val)
    intermediate_output = model.layers[-2].output
    intermediate_layer_model = Model(inputs=model.input, outputs=intermediate_output)
    intermediate_output = intermediate_layer_model.predict(X_test)
    y_pred = svr.predict(intermediate_output)
    print('Model:', i+1)
    y_pred = np.round(y_pred)
    mse = mean_squared_error(y_test, y_pred)
    print('Mean Squared Error:', mse)
    print("----")
    avg_mse += mse
print('Average Mean Squared Error:', avg_mse/count)
Epoch 1/20
```

```
s: 87336.4844
Epoch 2/20
s: 87336.4844
Epoch 3/20
s: 87336.4844
Epoch 4/20
s: 87336.4844
Epoch 5/20
```

```
s: 87336.4844
Epoch 6/20
s: 87336.4844
Epoch 7/20
s: 87336.4844
Epoch 8/20
s: 87336.4844
Epoch 9/20
s: 87336.4844
Epoch 10/20
s: 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
s: 87336.4844
Epoch 17/20
s: 87336.4844
Epoch 18/20
s: 87336.4844
Epoch 19/20
s: 87336.4844
Epoch 20/20
s: 87336.4844
282/282 [========== ] - 0s 572us/step
94/94 [======== ] - 0s 611us/step
Model: 1
Mean Squared Error: 4731.06933333333
Epoch 1/20
s: 46958.6758
Epoch 2/20
s: 24068.2637
Epoch 3/20
13300.7422
```

```
Epoch 4/20
7595.8467
Epoch 5/20
5031.7549
Epoch 6/20
4021.4944
Epoch 7/20
3598.6252
Epoch 8/20
3362.4221
Epoch 9/20
3253.7393
Epoch 10/20
3097.0500
Epoch 11/20
3105.1289
Epoch 12/20
2992,4409
Epoch 13/20
2935.9104
Epoch 14/20
2862.7556
Epoch 15/20
2859.9844
Epoch 16/20
841.6377
Epoch 17/20
776.0459
Epoch 18/20
738.9321
Epoch 19/20
2724.4058
Epoch 20/20
2679.6694
282/282 [============ ] - 0s 567us/step
94/94 [=======] - 0s 575us/step
Model: 2
Mean Squared Error: 2957.198
Epoch 1/20
50201.3008
Epoch 2/20
```

```
s: 25001.9492
Epoch 3/20
s: 13616.2734
Epoch 4/20
s: 7670.9629
Epoch 5/20
5105.7202
Epoch 6/20
4027.2471
Epoch 7/20
3605.2471
Epoch 8/20
3276.8850
Epoch 9/20
3120.7939
Epoch 10/20
3092.0896
Epoch 11/20
2914.6960
Epoch 12/20
2835.3433
Epoch 13/20
2873.0576
Epoch 14/20
2813.1252
Epoch 15/20
2767.7368
Epoch 16/20
2732.5759
Epoch 17/20
2759.2939
Epoch 18/20
2698.6003
Epoch 19/20
2675.0581
Epoch 20/20
2630.8638
282/282 [============ ] - 0s 629us/step
94/94 [======== ] - 0s 627us/step
Model: 3
Mean Squared Error: 2919.4283333333333
```

-----Epoch 1/20 s: 87336.4844 Epoch 2/20 s: 87336.4844 Epoch 3/20 s: 87336.4844 Epoch 4/20 s: 87336.4844 Epoch 5/20 s: 87336.4844 Epoch 6/20 s: 87336.4844 Epoch 7/20 s: 87336.4844 Epoch 8/20 s: 87336.4844 Epoch 9/20 s: 87336.4844 Epoch 10/20 s: 87336.4844 Epoch 11/20 s: 87336.4844 Epoch 12/20 s: 87336.4844 Epoch 13/20 s: 87336.4844 Epoch 14/20 s: 87336.4844 Epoch 15/20 s: 87336.4844 Epoch 16/20 s: 87336.4844 Epoch 17/20 s: 87336.4844 Epoch 18/20 s: 87336.4844 Epoch 19/20 1125/1125 [=============] - 1s 986us/step - loss: 88622.5469 - val_los s: 87336.4844 Epoch 20/20

```
87336.4844
282/282 [========= ] - 0s 596us/step
94/94 [=======] - 0s 584us/step
Model: 4
Mean Squared Error: 4255.151
Epoch 1/20
45830.1875
Epoch 2/20
s: 23672.1777
Epoch 3/20
s: 13199.2109
Epoch 4/20
s: 7474.0571
Epoch 5/20
4997.7593
Epoch 6/20
4051.9299
Epoch 7/20
3609.6589
Epoch 8/20
3407.7026
Epoch 9/20
3184.5012
Epoch 10/20
3058.9360
Epoch 11/20
2947.8428
Epoch 12/20
2838.8621
Epoch 13/20
2844.7798
Epoch 14/20
2770.3174
Epoch 15/20
2765.5266
Epoch 16/20
2786.7683
Epoch 17/20
2680.2156
Epoch 18/20
2682.2791
```

```
Epoch 19/20
        2644.6187
        Epoch 20/20
        2610.8401
        282/282 [============ ] - 0s 562us/step
        94/94 [======= ] - 0s 591us/step
        Model: 5
        Mean Squared Error: 2925.88733333333
        Average Mean Squared Error: 3557.7468
In [22]:
         # Model-3: Extract the deep features from Model-I (from 2nd layer) and Model-
         # 2 (from 2nd layer) stack the features horizontally and model it using a Support
         # Vector Regressor.
         def model3(X_train, X_val, y_train, y_val):
             model_1 = model1(X_train, X_val, y_train, y_val)
             model_2, _ = model2(X_train, X_val, y_train, y_val)
             intermediate_output1 = model_1.layers[-2].output
             intermediate_output2 = model_2.layers[-2].output
             intermediate_layer_model_1 = Model(inputs=model_1.input, outputs=intermediate_output)
             intermediate_output1 = intermediate_layer_model_1.predict(X_train)
             intermediate_layer_model_2 = Model(inputs=model_2.input, outputs=intermediate_output)
             intermediate_output2 = intermediate_layer_model_2.predict(X_train)
             intermediate_output = np.hstack((intermediate_output1, intermediate_output2))
             svr = SVR()
             svr.fit(intermediate_output, y_train)
             return model_1, model_2, svr
         count = 5
         avg_mse = 0
         for i in range(count):
             # Split data to 60% training 20% validation and 20% testing
            X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_sta
            X_train, X_val, y_train, y_val = train_test_split(X_train, y_train, test_size=0.25,
            model_1, model_2, svr = model3(X_train, X_val, y_train, y_val)
             intermediate_output1 = model_1.layers[-2].output
             intermediate_output2 = model_2.layers[-2].output
             intermediate_layer_model_1 = Model(inputs=model_1.input, outputs=intermediate_output)
             intermediate_output1 = intermediate_layer_model_1.predict(X_test)
             intermediate_layer_model_2 = Model(inputs=model_2.input, outputs=intermediate_output)
             intermediate_output2 = intermediate_layer_model_2.predict(X_test)
```

```
intermediate_output = np.hstack((intermediate_output1, intermediate_output2))
   y_pred = svr.predict(intermediate_output)
   print('Model:', i+1)
   y pred = np.round(y pred)
   mse = mean_squared_error(y_test, y_pred)
   print('Mean Squared Error:', mse)
   print("----")
   avg_mse += mse
print('Average Mean Squared Error:', avg_mse/count)
```

```
Epoch 1/20
44.1050
Epoch 2/20
050.7461
Epoch 3/20
954.7461
Epoch 4/20
009.6094
Epoch 5/20
870.3667
Epoch 6/20
916.1250
Epoch 7/20
819.6453
Epoch 8/20
790.2700
Epoch 9/20
809.4346
Epoch 10/20
740.3032
Epoch 11/20
711.7197
Epoch 12/20
716.6248
Epoch 13/20
693.9026
Epoch 14/20
767.4211
Epoch 15/20
```

```
674.4001
Epoch 16/20
688.4128
Epoch 17/20
674.8105
Epoch 18/20
701.1467
Epoch 19/20
639.5505
Epoch 20/20
638.8616
Epoch 1/20
s: 87336.4844
Epoch 2/20
s: 87336.4844
Epoch 3/20
s: 87336.4844
Epoch 4/20
s: 87336.4844
Epoch 5/20
s: 87336.4844
Epoch 6/20
s: 87336.4844
Epoch 7/20
s: 87336.4844
Epoch 8/20
87336.4844
Epoch 9/20
s: 87336.4844
Epoch 10/20
s: 87336.4844
Epoch 11/20
s: 87336.4844
Epoch 12/20
87336.4844
Epoch 13/20
s: 87336.4844
Epoch 14/20
s: 87336.4844
Epoch 15/20
```

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s: 87336.4844
Epoch 16/20
s: 87336.4844
Epoch 17/20
s: 87336.4844
Epoch 18/20
87336.4844
Epoch 19/20
s: 87336.4844
Epoch 20/20
s: 87336.4844
282/282 [========= ] - 0s 741us/step
282/282 [========== ] - 0s 566us/step
282/282 [========== ] - 0s 563us/step
94/94 [=======] - 0s 587us/step
94/94 [======= ] - 0s 600us/step
Model: 1
Mean Squared Error: 2954.1633333333334
_____
Epoch 1/20
86.7427
Epoch 2/20
6.8157
Epoch 3/20
047.5601
Epoch 4/20
951.2502
Epoch 5/20
6.9316
Epoch 6/20
820.8374
Epoch 7/20
7.8354
Epoch 8/20
811.5676
Epoch 9/20
772.6782
Epoch 10/20
767.0742
Epoch 11/20
770.4902
Epoch 12/20
3.3662
```

```
Epoch 13/20
681.4985
Epoch 14/20
737.4854
Epoch 15/20
677.8940
Epoch 16/20
679.7388
Epoch 17/20
651.6670
Epoch 18/20
663.7329
Epoch 19/20
0.6167
Epoch 20/20
2.4475
Epoch 1/20
s: 50208.4688
Epoch 2/20
s: 28356.7012
Epoch 3/20
s: 16109.8799
Epoch 4/20
s: 9615.3115
Epoch 5/20
6125.2231
Epoch 6/20
4522.5806
Epoch 7/20
3756.8506
Epoch 8/20
3498.5200
Epoch 9/20
3288.3403
Epoch 10/20
3183.3677
Epoch 11/20
3096.9766
Epoch 12/20
2979.8027
```

```
Epoch 13/20
2975.9221
Epoch 14/20
2877.1814
Epoch 15/20
2826.5002
Epoch 16/20
2901.6204
Epoch 17/20
2822.2219
Epoch 18/20
2794.9041
Epoch 19/20
2760.5459
Epoch 20/20
2680.1074
282/282 [============ ] - 0s 558us/step
282/282 [========= ] - 0s 564us/step
282/282 [========= ] - 0s 575us/step
94/94 [=======] - 0s 601us/step
94/94 [======= ] - 0s 568us/step
Model: 2
Mean Squared Error: 2941.9313333333334
Epoch 1/20
22.3120
Epoch 2/20
056.3147
Epoch 3/20
2.8611
Epoch 4/20
8.4043
Epoch 5/20
4.7302
Epoch 6/20
822.2686
Epoch 7/20
7.1992
Epoch 8/20
829.3567
Epoch 9/20
9.7109
Epoch 10/20
```

```
746.7163
Epoch 11/20
9.6951
Epoch 12/20
685.9041
Epoch 13/20
697.0869
Epoch 14/20
749.1265
Epoch 15/20
666.6025
Epoch 16/20
670.0474
Epoch 17/20
7.7393
Epoch 18/20
681.0649
Epoch 19/20
653.2661
Epoch 20/20
1.2068
Epoch 1/20
s: 87336.4844
Epoch 2/20
s: 87336.4844
Epoch 3/20
s: 87336.4844
Epoch 4/20
s: 87336.4844
Epoch 5/20
s: 87336.4844
Epoch 6/20
s: 87336.4844
Epoch 7/20
s: 87336.4844
Epoch 8/20
s: 87336.4844
Epoch 9/20
s: 87336.4844
Epoch 10/20
```

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s: 87336.4844
Epoch 11/20
s: 87336.4844
Epoch 12/20
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Epoch 13/20
s: 87336.4844
Epoch 14/20
s: 87336.4844
Epoch 15/20
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Epoch 16/20
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Epoch 17/20
s: 87336.4844
Epoch 18/20
s: 87336.4844
Epoch 19/20
s: 87336.4844
Epoch 20/20
s: 87336.4844
282/282 [============ ] - 0s 568us/step
282/282 [========= ] - 0s 577us/step
282/282 [========== ] - 0s 602us/step
94/94 [=======] - 0s 547us/step
94/94 [=======] - 0s 606us/step
Model: 3
Mean Squared Error: 2911.193
_____
Epoch 1/20
33.7229
Epoch 2/20
052.9551
Epoch 3/20
059.5913
Epoch 4/20
909.4988
Epoch 5/20
920.9871
Epoch 6/20
851.5503
Epoch 7/20
```

```
812.1689
Epoch 8/20
0.5566
Epoch 9/20
792.9456
Epoch 10/20
780.5901
Epoch 11/20
755.5859
Epoch 12/20
706.9338
Epoch 13/20
689.1570
Epoch 14/20
703.7961
Epoch 15/20
707.7153
Epoch 16/20
646.6394
Epoch 17/20
629.0342
Epoch 18/20
662.9802
Epoch 19/20
650.8433
Epoch 20/20
653.0955
Epoch 1/20
s: 46109.1953
Epoch 2/20
s: 23958.7949
Epoch 3/20
s: 13278.2842
Epoch 4/20
s: 7512.2915
Epoch 5/20
5047.6694
Epoch 6/20
1125/1125 [============] - 1s 915us/step - loss: 4686.8350 - val_loss:
4007.0200
Epoch 7/20
```

```
3594.4431
Epoch 8/20
3369.2991
Epoch 9/20
3182.1599
Epoch 10/20
3012.5601
Epoch 11/20
2962.7993
Epoch 12/20
2888.3547
Epoch 13/20
2851.3584
Epoch 14/20
2879.9460
Epoch 15/20
2831.8049
Epoch 16/20
2843.3196
Epoch 17/20
2764.3765
Epoch 18/20
2740.7009
Epoch 19/20
2986.2676
Epoch 20/20
2740.8047
282/282 [========== ] - 0s 573us/step
282/282 [========== ] - 0s 534us/step
282/282 [========= ] - 1s 579us/step
94/94 [======= ] - 0s 606us/step
94/94 [=======] - 0s 579us/step
Model: 4
Mean Squared Error: 2947.3006666666665
Epoch 1/20
64.4871
Epoch 2/20
028.2126
Epoch 3/20
979.6736
Epoch 4/20
1.9377
```

```
Epoch 5/20
6.5117
Epoch 6/20
802.6816
Epoch 7/20
9.5337
Epoch 8/20
0.2520
Epoch 9/20
918.4690
Epoch 10/20
835.1299
Epoch 11/20
711.3501
Epoch 12/20
682.3149
Epoch 13/20
788.6262
Epoch 14/20
5.6934
Epoch 15/20
4.6711
Epoch 16/20
6.2200
Epoch 17/20
1.9294
Epoch 18/20
9.1372
Epoch 19/20
3.7534
Epoch 20/20
2.5774
Epoch 1/20
49615.8359
Epoch 2/20
s: 24438.2070
Epoch 3/20
s: 13426.3096
Epoch 4/20
s: 7626.0527
```

```
Epoch 5/20
5053.8169
Epoch 6/20
4090.6057
Epoch 7/20
3613.6758
Epoch 8/20
3331.2336
Epoch 9/20
3157.3040
Epoch 10/20
980.0500
Epoch 11/20
922.2612
Epoch 12/20
861.4727
Epoch 13/20
915.9761
Epoch 14/20
2785.9866
Epoch 15/20
2766.4812
Epoch 16/20
2797.2014
Epoch 17/20
2775.4104
Epoch 18/20
2726.1853
Epoch 19/20
2686.8391
Epoch 20/20
2721.8665
282/282 [=========== ] - 0s 571us/step
282/282 [========= ] - 0s 565us/step
282/282 [========== ] - 0s 597us/step
94/94 [=======] - 0s 626us/step
94/94 [======== ] - 0s 649us/step
Model: 5
Mean Squared Error: 2959.484333333333
Average Mean Squared Error: 2942.814533333333
```

```
In [23]:
          # Model-4: Extract the deep features from Model-I and Model-2 stack the
          # features horizontally, reduce the dimension to either 8, IO or 12 using
          # principal component analysis (PCA) and model the reduced features using a
          # Random Forest classifier. Identify the best number of reduced components of
          # PCA.
          from sklearn.decomposition import PCA
          from sklearn.ensemble import RandomForestRegressor
          def model4(X_train, X_val, y_train, y_val):
              model_1 = model1(X_train, X_val, y_train, y_val)
              model_2, _ = model2(X_train, X_val, y_train, y_val)
              intermediate_output1 = model_1.layers[-2].output
              intermediate_output2 = model_2.layers[-2].output
              intermediate_layer_model_1 = Model(inputs=model_1.input, outputs=intermediate_output)
              intermediate_output1 = intermediate_layer_model_1.predict(X_train)
              intermediate_layer_model_2 = Model(inputs=model_2.input, outputs=intermediate_output)
              intermediate_output2 = intermediate_layer_model_2.predict(X_train)
              intermediate_output = np.hstack((intermediate_output1, intermediate_output2))
              pca = PCA(n_components=8)
              pca.fit(intermediate_output)
              intermediate_output = pca.transform(intermediate_output)
              rf = RandomForestRegressor()
              rf.fit(intermediate_output, y_train)
              return model_1, model_2, pca, rf
          count = 5
          avg_mse = 0
          for i in range(count):
              # Split data to 60% training 20% validation and 20% testing
              X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_sta
              X_train, X_val, y_train, y_val = train_test_split(X_train, y_train, test_size=0.25,
              model_1, model_2, pca, rf = model4(X_train, X_val, y_train, y_val)
              intermediate_output1 = model_1.layers[-2].output
              intermediate_output2 = model_2.layers[-2].output
              intermediate layer model 1 = Model(inputs=model 1.input, outputs=intermediate output)
              intermediate_output1 = intermediate_layer_model_1.predict(X_test)
              intermediate_layer_model_2 = Model(inputs=model_2.input, outputs=intermediate output)
              intermediate_output2 = intermediate_layer_model_2.predict(X_test)
              intermediate_output = np.hstack((intermediate_output1, intermediate_output2))
              intermediate_output = pca.transform(intermediate_output)
              y_pred = rf.predict(intermediate_output)
```

```
print('Model:', i+1)
y_pred = np.round(y_pred)
mse = mean_squared_error(y_test, y_pred)

print('Mean Squared Error:', mse)
print("-----")

avg_mse += mse

print('Average Mean Squared Error:', avg_mse/count)
```

```
Epoch 1/20
43.3008
Epoch 2/20
031.9114
Epoch 3/20
995.2710
Epoch 4/20
928.5117
Epoch 5/20
Epoch 6/20
833.1746
Epoch 7/20
8.6130
Epoch 8/20
784.0286
Epoch 9/20
8.1917
Epoch 10/20
7.2976
Epoch 11/20
732.3027
Epoch 12/20
2.3333
Epoch 13/20
709.0298
Epoch 14/20
666.6453
Epoch 15/20
2.7349
Epoch 16/20
715.1587
```

```
Epoch 17/20
7.8647
Epoch 18/20
645.9009
Epoch 19/20
649.5842
Epoch 20/20
859.0193
Epoch 1/20
s: 87336.4844
Epoch 2/20
s: 87336.4844
Epoch 3/20
s: 87336.4844
Epoch 4/20
s: 87336.4844
Epoch 5/20
s: 87336.4844
Epoch 6/20
s: 87336.4844
Epoch 7/20
s: 87336.4844
Epoch 8/20
s: 87336.4844
Epoch 9/20
s: 87336.4844
Epoch 10/20
s: 87336.4844
Epoch 11/20
s: 87336.4844
Epoch 12/20
s: 87336.4844
Epoch 13/20
s: 87336.4844
Epoch 14/20
s: 87336.4844
Epoch 15/20
s: 87336.4844
Epoch 16/20
s: 87336.4844
```

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Epoch 17/20
s: 87336.4844
Epoch 18/20
s: 87336.4844
Epoch 19/20
s: 87336.4844
Epoch 20/20
s: 87336.4844
282/282 [========== ] - 0s 563us/step
282/282 [========= ] - 0s 587us/step
282/282 [========= ] - 0s 609us/step
94/94 [=======] - 0s 573us/step
94/94 [======= ] - 0s 600us/step
Model: 1
Mean Squared Error: 2751.744
-----
Epoch 1/20
53.9331
Epoch 2/20
9.6631
Epoch 3/20
025.1003
Epoch 4/20
891.5806
Epoch 5/20
852.5474
Epoch 6/20
826.4072
Epoch 7/20
830.2637
Epoch 8/20
5.2571
Epoch 9/20
747.7722
Epoch 10/20
749.9941
Epoch 11/20
705.0662
Epoch 12/20
678.4570
Epoch 13/20
675.4189
Epoch 14/20
```

```
8.2666
Epoch 15/20
4.7649
Epoch 16/20
2.3501
Epoch 17/20
5.0906
Epoch 18/20
599.9062
Epoch 19/20
661.2976
Epoch 20/20
613.2417
Epoch 1/20
s: 49688.4844
Epoch 2/20
s: 27955.9316
Epoch 3/20
s: 15935.0078
Epoch 4/20
s: 9551.4482
Epoch 5/20
6095.4238
Epoch 6/20
4484.7124
Epoch 7/20
3864.6760
Epoch 8/20
3517.2686
Epoch 9/20
3270.2141
Epoch 10/20
3150.5591
Epoch 11/20
3176.7234
Epoch 12/20
3000.8875
Epoch 13/20
2996.8259
Epoch 14/20
```

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2951.4600
Epoch 15/20
2788.2395
Epoch 16/20
2825.8760
Epoch 17/20
2790.7009
Epoch 18/20
2859.9373
Epoch 19/20
2799.7625
Epoch 20/20
282/282 [============ ] - 0s 568us/step
282/282 [========== ] - 0s 552us/step
282/282 [========== ] - 0s 558us/step
94/94 [=======] - 0s 603us/step
94/94 [=======] - 0s 583us/step
Model: 2
Mean Squared Error: 2708.8306666666667
______
Epoch 1/20
94.4080
Epoch 2/20
018.0964
Epoch 3/20
983.7410
Epoch 4/20
087.4863
Epoch 5/20
887.9866
Epoch 6/20
833.9929
Epoch 7/20
886.5530
Epoch 8/20
806.2671
Epoch 9/20
784.0635
Epoch 10/20
790.7532
Epoch 11/20
```

```
754.3323
Epoch 12/20
812.0967
Epoch 13/20
677.5107
Epoch 14/20
780.1484
Epoch 15/20
711.6396
Epoch 16/20
646.1152
Epoch 17/20
636.1648
Epoch 18/20
563/563 [===========] - 1s 963us/step - loss: 2666.8840 - val_loss: 2
805.2483
Epoch 19/20
765.0342
Epoch 20/20
0.6350
Epoch 1/20
s: 54774.1289
Epoch 2/20
s: 33750.4648
Epoch 3/20
s: 20420.0059
Epoch 4/20
s: 13038.4492
Epoch 5/20
s: 8117.9341
Epoch 6/20
5639.3013
Epoch 7/20
1125/1125 [=============] - 1s 918us/step - loss: 5196.1655 - val_loss:
4349.7949
Epoch 8/20
3728.5803
Epoch 9/20
3450.0049
Epoch 10/20
3251.2944
Epoch 11/20
```

```
3198.0391
Epoch 12/20
3018.9136
Epoch 13/20
2963.7300
Epoch 14/20
2920.4534
Epoch 15/20
2963.5071
Epoch 16/20
2921.3130
Epoch 17/20
866.0356
Epoch 18/20
2777.5452
Epoch 19/20
2813.3269
Epoch 20/20
2736.0710
282/282 [=========] - 0s 573us/step
282/282 [============ ] - 0s 570us/step
282/282 [========= ] - 0s 586us/step
94/94 [========] - 0s 1ms/step
94/94 [======== ] - 0s 1ms/step
Model: 3
Mean Squared Error: 2712.294333333333
-----
Epoch 1/20
54.7388
Epoch 2/20
5.0730
Epoch 3/20
4.5784
Epoch 4/20
4.9746
Epoch 5/20
5.6060
Epoch 6/20
4.4580
Epoch 7/20
8.7781
Epoch 8/20
0.5547
```

```
Epoch 9/20
5.7690
Epoch 10/20
9.7546
Epoch 11/20
9.2507
Epoch 12/20
4.3044
Epoch 13/20
4.1934
Epoch 14/20
3.1675
Epoch 15/20
6.7009
Epoch 16/20
9.2913
Epoch 17/20
2.8787
Epoch 18/20
1.9138
Epoch 19/20
1.8518
Epoch 20/20
3.9724
Epoch 1/20
46315.3281
Epoch 2/20
24182.8926
Epoch 3/20
13376.8877
Epoch 4/20
7606.8179
Epoch 5/20
092.2881
Epoch 6/20
032.4106
Epoch 7/20
537.9446
Epoch 8/20
312.3313
```

```
Epoch 9/20
136.4500
Epoch 10/20
026.4426
Epoch 11/20
995.3538
Epoch 12/20
923.5491
Epoch 13/20
813.4482
Epoch 14/20
821.4646
Epoch 15/20
794.4688
Epoch 16/20
791.9104
Epoch 17/20
784.4641
Epoch 18/20
788.6636
Epoch 19/20
839.7964
Epoch 20/20
734.8628
282/282 [========= ] - 0s 1ms/step
282/282 [========= ] - 1s 2ms/step
94/94 [=======] - 0s 2ms/step
94/94 [========] - 0s 1ms/step
Model: 4
Mean Squared Error: 2658.715666666665
Epoch 1/20
336.4453
Epoch 2/20
336.4453
Epoch 3/20
336.4453
Epoch 4/20
336.4453
Epoch 5/20
336.4453
Epoch 6/20
```

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336.4453
Epoch 7/20
336.4453
Epoch 8/20
336.4453
Epoch 9/20
336.4453
Epoch 10/20
336.4453
Epoch 11/20
336.4453
Epoch 12/20
336.4453
Epoch 13/20
336.4453
Epoch 14/20
336.4453
Epoch 15/20
336.4453
Epoch 16/20
336.4453
Epoch 17/20
336.4453
Epoch 18/20
336.4453
Epoch 19/20
336.4453
Epoch 20/20
336.4453
Epoch 1/20
46148.9766
Epoch 2/20
23976.4805
Epoch 3/20
13314.9199
Epoch 4/20
7528.9795
Epoch 5/20
044.1284
Epoch 6/20
```

```
011.5740
Epoch 7/20
611.7720
Epoch 8/20
284.8816
Epoch 9/20
3188.5408
Epoch 10/20
078.6143
Epoch 11/20
017.3213
Epoch 12/20
999.6113
Epoch 13/20
959.0901
Epoch 14/20
863.7153
Epoch 15/20
894.5596
Epoch 16/20
855.2156
Epoch 17/20
750.3057
Epoch 18/20
801.4744
Epoch 19/20
751.8027
Epoch 20/20
725.7539
282/282 [========== ] - 0s 1ms/step
282/282 [========= ] - 1s 2ms/step
94/94 [=======] - 0s 1ms/step
94/94 [========] - 0s 2ms/step
Model: 5
Mean Squared Error: 2755.363
-----
Average Mean Squared Error: 2717.389533333333
ONo 5
```

For model1: Average Mean Squared Error: 2823.2824

For model2: Average Mean Squared Error: 3557.7468

For model3: Average Mean Squared Error: 2942.814533333333

For model4: Average Mean Squared Error: 2717.389533333333

so by seeing this avg mean square error we can conclude model 4 is best but when we also consider time to train model 1 is faster