# supplementry

December 6, 2020

#### 1 load the data

```
[1]: import pandas as pd
    from sklearn.metrics import r2_score
    import numpy as np

[2]: df = pd.read_csv('data_thesis.csv')

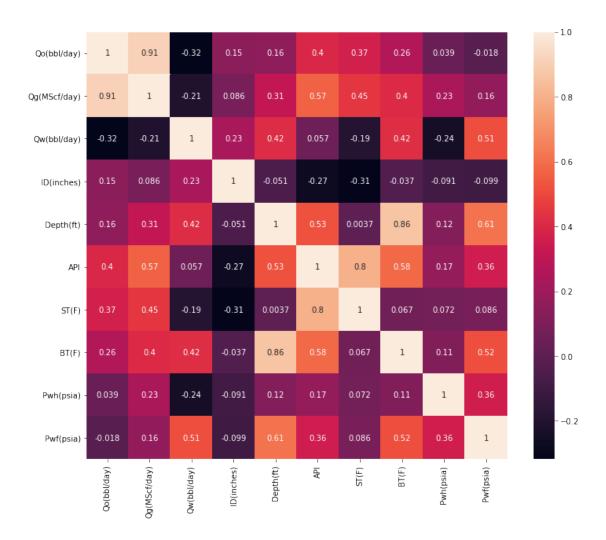
[3]: df=df.dropna()
    df.to_csv("my_description.csv")

[4]: table=df.head()
    table.to_csv("data.csv")
```

### 2 Explore the data

```
[5]: import seaborn as sns import matplotlib.pyplot as plt %matplotlib inline
```

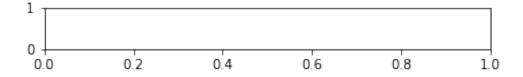
```
[6]: plt.figure(figsize=(12,10))
sns.heatmap(df.corr(),annot=True)
plt.savefig('heatmap.png', dpi=300)
```

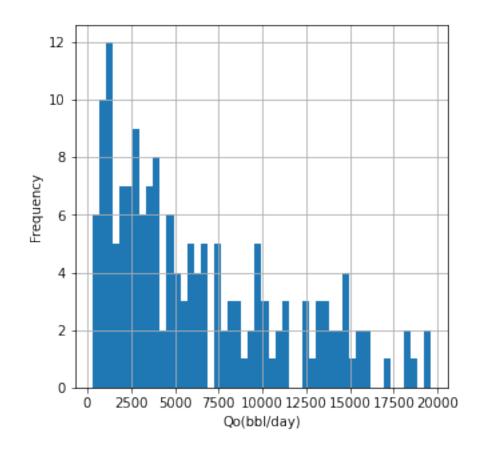


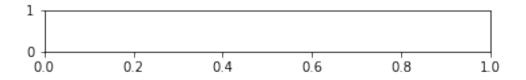
## 3 Training and testing the data

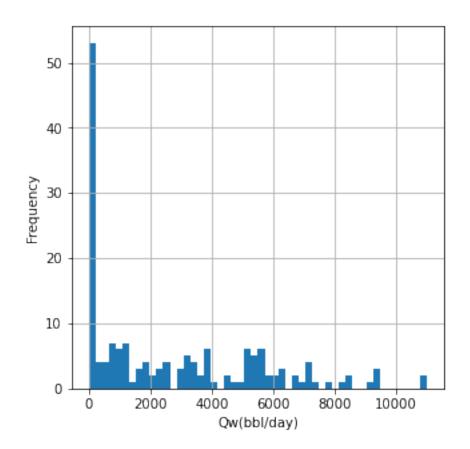
```
[12]: X_test.shape
[12]: (41, 9)
     4 Feature selection
[13]: # reduce the number of the input data by using filtter by correlation method
      def correlation(dataset,threshold):
          col corr = set()
          corr_matrix = dataset.corr()
          for i in range(len(corr_matrix.columns)):
              for j in range(i):
                  if abs(corr_matrix.iloc[i,j]) > threshold:
                      colname= corr_matrix.columns[i]
                      col corr.add(colname)
          return col_corr
[14]: corr_features = correlation(X_train,0.8)
      len(set(corr_features))
[14]: 3
[15]: corr_features
[15]: {'BT(F)', 'Qg(MScf/day)', 'ST(F)'}
[16]: #drop the corr features from the X_test and X-train
      X_train.drop(labels=corr_features,axis=1,inplace=True)
      X_test.drop(labels=corr_features,axis=1,inplace=True)
      X_train.shape,X_test.shape
[16]: ((165, 6), (41, 6))
[17]: df_train=pd.DataFrame(X_train)
[18]: df_test=pd.DataFrame(X_test)
[19]: table=y_train.describe()
      table.to_csv("train_out.csv")
[20]: table2=y test.describe()
      table2.to_csv("test_out_.csv")
[21]: X_test.skew()
```

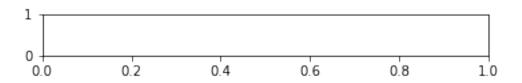
```
[21]: Qo(bbl/day)
                     0.920679
     Qw(bbl/day)
                     0.731498
      ID(inches)
                    -0.829107
     Depth(ft)
                    -1.615509
     API
                     0.361125
     Pwh(psia)
                     1.538948
      dtype: float64
[81]: # Plot the histogram of train daat
      plt.close()
      fig=plt.figure()
      for i in range (len(dtrain.columns)):
       plt.subplot(6,1,i+1)
       plt.figure(figsize = (5, 5))
       plt.hist(dtrain[dtrain.columns[i]],bins = 50)
       plt.xlabel(dtrain.columns[i])
       plt.ylabel("Frequency")
       plt.grid()
        #plt.savefig("train dis_%.png" %i)
        fig.savefig('train dis'+str(i)+'.png')
       plt.show();
```

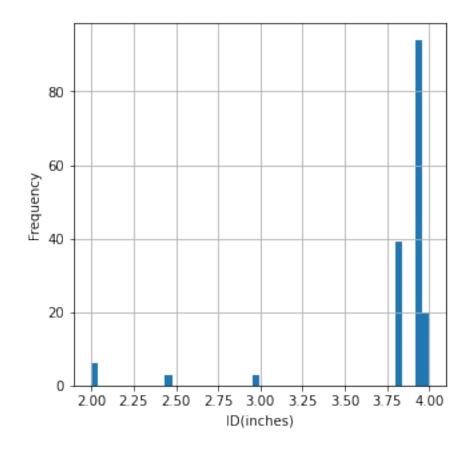


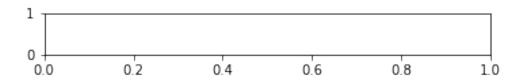


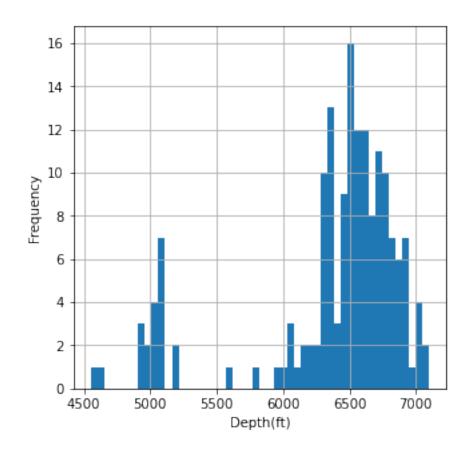


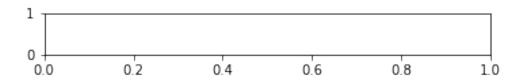


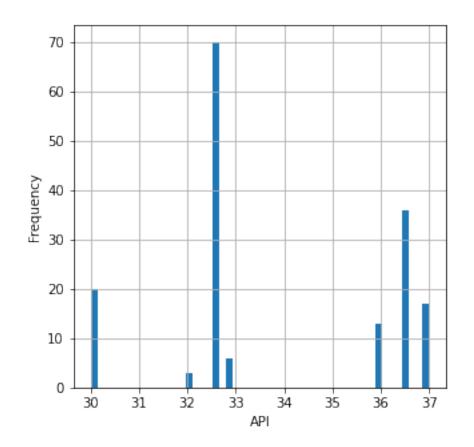


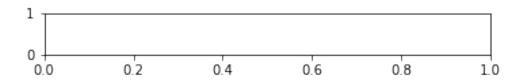


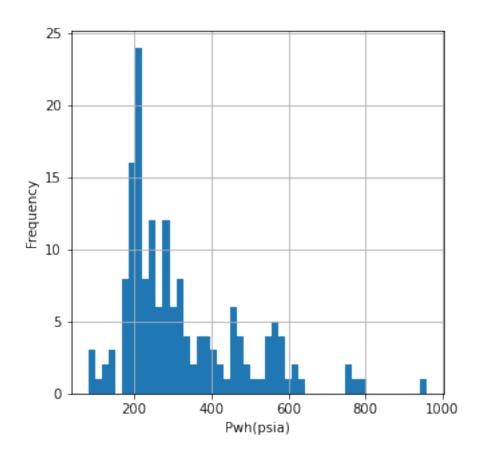






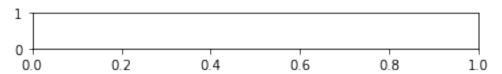


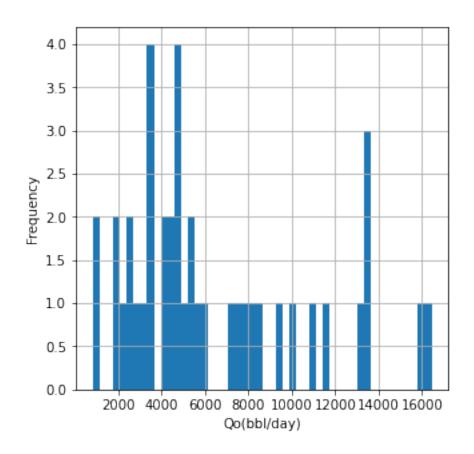




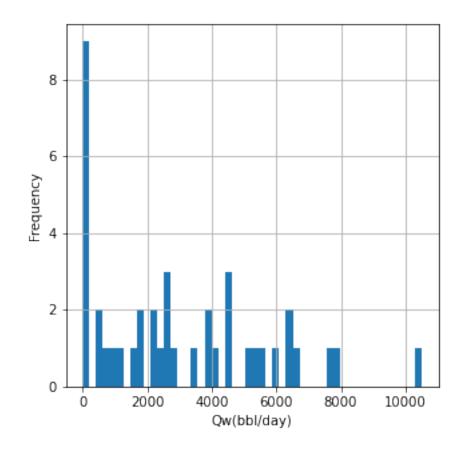
```
[376]: # Plot the histogram of test daat
for i in range (len(dtest.columns)):
    plt.subplot(7,1,i+1)
    plt.figure(figsize = (5, 5))
    plt.hist(dtest[dtest.columns[i]],bins = 50)

    plt.xlabel(dt.columns[i])
    plt.ylabel("Frequency")
    plt.grid()
    plt.show();
```

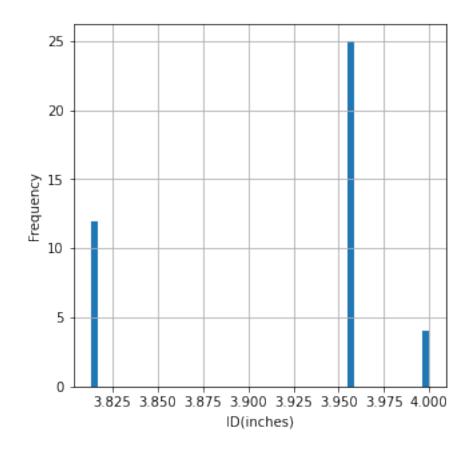




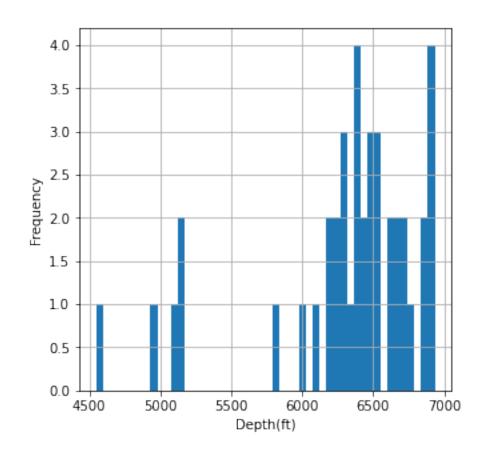




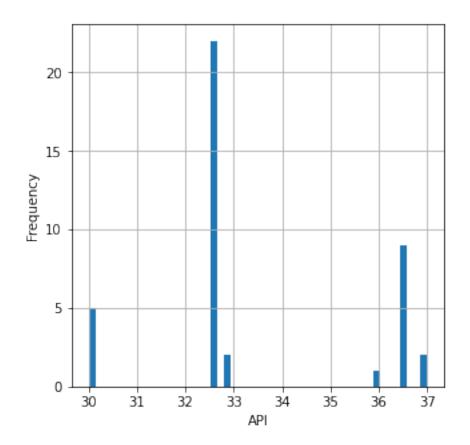




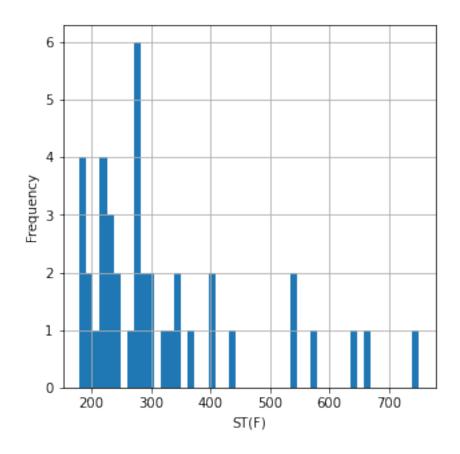












#### 5 Scaling the input dataset

```
[129]: from tensorflow.keras.models import Sequential
       from tensorflow.keras.layers import Dense, Activation
 []: # Improving the ANN
       # Dropout Regularization to reduce overfitting if needed
       # Tuning the ANN
       from tensorflow.keras.wrappers.scikit_learn import KerasRegressor
       from sklearn.model selection import GridSearchCV
       from tensorflow.keras.models import Sequential
       from tensorflow.keras.layers import Dense
       def build_regressor(optimizer):
          regressor = Sequential()
          regressor.add(Dense(units = 6, kernel_initializer = 'uniform', activation = ___
       →'relu', input_dim = 6))
          regressor.add(Dense(units = 6, kernel_initializer = 'uniform', activation = u

¬'relu'))
          regressor.add(Dense(units = 1, kernel_initializer = 'uniform', activation = u
       regressor.compile(optimizer = optimizer, loss = 'mse')
          return regressor
       regressor = KerasRegressor(build_fn = build_regressor)
       parameters = {'batch_size': [25, 32],
                     'epochs': [100, 1000],
                     'optimizer': ['adam', 'rmsprop']}
       grid_search = GridSearchCV(estimator = regressor,
                                  param_grid = parameters,
                                  cv = 10)
       grid_search = grid_search.fit(X_train, y_train)
       best_parameters = grid_search.best_params_
       best_accuracy = grid_search.best_score_
[130]: model = Sequential()
       model.add(Dense(50,activation='relu'))
      model.add(Dense(50,activation='relu'))
       model.add(Dense(50,activation='relu'))
       model.add(Dense(50,activation='relu'))
       model.add(Dense(50,activation='relu'))
       model.add(Dense(50,activation='relu'))
       model.add(Dense(50,activation='relu'))
       model.add(Dense(50,activation='relu'))
       model.add(Dense(50,activation='relu'))
       # Final output node for prediction
       model.add(Dense(1))
       model.compile(optimizer='adam',loss='mse')
```

#### [131]: model.fit(X\_train,y\_train,epochs=1000)

```
Epoch 1/1000
Epoch 2/1000
Epoch 3/1000
Epoch 4/1000
Epoch 5/1000
Epoch 6/1000
Epoch 7/1000
Epoch 8/1000
6/6 [=========== ] - 0s 1ms/step - loss: 5373073.0000
Epoch 9/1000
Epoch 10/1000
Epoch 11/1000
Epoch 12/1000
Epoch 13/1000
Epoch 14/1000
Epoch 15/1000
Epoch 16/1000
Epoch 17/1000
Epoch 18/1000
Epoch 19/1000
Epoch 20/1000
6/6 [============ ] - Os 1ms/step - loss: 97426.5156
Epoch 21/1000
Epoch 22/1000
Epoch 23/1000
```

```
Epoch 24/1000
Epoch 25/1000
Epoch 26/1000
6/6 [=========== ] - Os 2ms/step - loss: 63867.2188
Epoch 27/1000
Epoch 28/1000
6/6 [============== ] - Os 2ms/step - loss: 57784.1953
Epoch 29/1000
6/6 [============ ] - Os 1ms/step - loss: 53615.4062
Epoch 30/1000
Epoch 31/1000
6/6 [============ ] - Os 1ms/step - loss: 47304.8789
Epoch 32/1000
6/6 [============ ] - Os 1ms/step - loss: 49084.9375
Epoch 33/1000
Epoch 34/1000
Epoch 35/1000
6/6 [========== ] - Os 1ms/step - loss: 40828.7500
Epoch 36/1000
6/6 [============ ] - Os 1ms/step - loss: 39273.5469
Epoch 37/1000
6/6 [=========== ] - Os 1ms/step - loss: 38214.8984
Epoch 38/1000
6/6 [============ ] - Os 2ms/step - loss: 39311.7109
Epoch 39/1000
6/6 [============ ] - Os 1ms/step - loss: 35443.8477
Epoch 40/1000
Epoch 41/1000
Epoch 42/1000
6/6 [============== ] - Os 1ms/step - loss: 33013.9141
Epoch 43/1000
6/6 [============== ] - Os 1ms/step - loss: 32473.5273
Epoch 44/1000
6/6 [============ ] - Os 1ms/step - loss: 33028.4531
Epoch 45/1000
6/6 [============ ] - Os 1ms/step - loss: 31950.5312
Epoch 46/1000
6/6 [=========== ] - Os 1ms/step - loss: 30924.0918
Epoch 47/1000
6/6 [============ ] - Os 1ms/step - loss: 30972.9941
```

```
Epoch 48/1000
Epoch 49/1000
Epoch 50/1000
6/6 [============== ] - Os 1ms/step - loss: 29127.8086
Epoch 51/1000
Epoch 52/1000
6/6 [============== ] - Os 1ms/step - loss: 28634.6602
Epoch 53/1000
Epoch 54/1000
Epoch 55/1000
6/6 [============ ] - Os 1ms/step - loss: 27147.0723
Epoch 56/1000
6/6 [=========== ] - Os 1ms/step - loss: 28211.6445
Epoch 57/1000
Epoch 58/1000
Epoch 59/1000
Epoch 60/1000
6/6 [============ ] - Os 2ms/step - loss: 25388.8496
Epoch 61/1000
6/6 [============ ] - Os 1ms/step - loss: 27313.6035
Epoch 62/1000
6/6 [=========== ] - Os 1ms/step - loss: 27380.0508
Epoch 63/1000
Epoch 64/1000
Epoch 65/1000
Epoch 66/1000
6/6 [============= ] - Os 1ms/step - loss: 24086.1289
Epoch 67/1000
6/6 [============== ] - Os 1ms/step - loss: 24249.6875
Epoch 68/1000
6/6 [============ ] - Os 1ms/step - loss: 25331.6387
Epoch 69/1000
6/6 [============ ] - Os 1ms/step - loss: 23201.3125
Epoch 70/1000
Epoch 71/1000
6/6 [=========== ] - Os 1ms/step - loss: 23144.9570
```

```
Epoch 72/1000
Epoch 73/1000
Epoch 74/1000
Epoch 75/1000
6/6 [============== ] - Os 1ms/step - loss: 21646.3203
Epoch 76/1000
6/6 [============== ] - Os 1ms/step - loss: 21727.8008
Epoch 77/1000
6/6 [=========== ] - Os 1ms/step - loss: 22681.2930
Epoch 78/1000
6/6 [============ ] - Os 1ms/step - loss: 23646.9629
Epoch 79/1000
6/6 [============ ] - Os 1ms/step - loss: 23193.0625
Epoch 80/1000
6/6 [=========== ] - Os 1ms/step - loss: 21310.2090
Epoch 81/1000
Epoch 82/1000
Epoch 83/1000
6/6 [============= - - os 1ms/step - loss: 20785.9414
Epoch 84/1000
6/6 [============ ] - Os 1ms/step - loss: 20103.4570
Epoch 85/1000
6/6 [=========== ] - Os 1ms/step - loss: 20211.8594
Epoch 86/1000
6/6 [============ ] - Os 1ms/step - loss: 20216.7812
Epoch 87/1000
6/6 [============ ] - Os 1ms/step - loss: 20903.5059
Epoch 88/1000
Epoch 89/1000
Epoch 90/1000
6/6 [============= ] - Os 1ms/step - loss: 20078.9668
Epoch 91/1000
6/6 [============== ] - Os 1ms/step - loss: 22132.1855
Epoch 92/1000
6/6 [============ ] - Os 1ms/step - loss: 22425.7129
Epoch 93/1000
6/6 [============ ] - Os 1ms/step - loss: 19600.2402
Epoch 94/1000
6/6 [============ ] - Os 1ms/step - loss: 19813.4746
Epoch 95/1000
6/6 [=========== ] - Os 1ms/step - loss: 18837.2090
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```
Epoch 96/1000
Epoch 97/1000
Epoch 98/1000
Epoch 99/1000
Epoch 100/1000
6/6 [============== ] - Os 1ms/step - loss: 20445.5781
Epoch 101/1000
6/6 [============ ] - Os 1ms/step - loss: 18969.4980
Epoch 102/1000
6/6 [============ ] - Os 1ms/step - loss: 19024.1465
Epoch 103/1000
6/6 [============ ] - Os 1ms/step - loss: 18032.9355
Epoch 104/1000
Epoch 105/1000
Epoch 106/1000
Epoch 107/1000
6/6 [============= ] - Os 997us/step - loss: 17760.6035
Epoch 108/1000
6/6 [============ ] - Os 1ms/step - loss: 18352.8281
Epoch 109/1000
6/6 [============ ] - Os 1ms/step - loss: 18908.9297
Epoch 110/1000
Epoch 111/1000
Epoch 112/1000
Epoch 113/1000
6/6 [============== - - os 1ms/step - loss: 17604.7656
Epoch 114/1000
6/6 [============== ] - Os 1ms/step - loss: 18865.7070
Epoch 115/1000
6/6 [============== ] - Os 1ms/step - loss: 21200.8965
Epoch 116/1000
6/6 [============ ] - Os 1ms/step - loss: 19763.0840
Epoch 117/1000
Epoch 118/1000
6/6 [============ ] - Os 1ms/step - loss: 18487.6875
Epoch 119/1000
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Epoch 120/1000
Epoch 121/1000
Epoch 122/1000
Epoch 123/1000
6/6 [============== ] - Os 1ms/step - loss: 18072.9707
Epoch 124/1000
6/6 [=============== ] - Os 1ms/step - loss: 17784.4102
Epoch 125/1000
6/6 [============ ] - Os 1ms/step - loss: 17423.9023
Epoch 126/1000
Epoch 127/1000
6/6 [============ ] - Os 1ms/step - loss: 17781.8848
Epoch 128/1000
6/6 [=========== ] - Os 1ms/step - loss: 20123.9883
Epoch 129/1000
Epoch 130/1000
Epoch 131/1000
6/6 [============= ] - Os 997us/step - loss: 19732.2676
Epoch 132/1000
6/6 [============ ] - Os 1ms/step - loss: 17635.1523
Epoch 133/1000
Epoch 134/1000
6/6 [============ ] - Os 1ms/step - loss: 16479.9492
Epoch 135/1000
Epoch 136/1000
Epoch 137/1000
Epoch 138/1000
6/6 [============== ] - Os 1ms/step - loss: 17263.6309
Epoch 139/1000
6/6 [============== ] - Os 1ms/step - loss: 17013.9648
Epoch 140/1000
6/6 [============ ] - Os 1ms/step - loss: 22396.5234
Epoch 141/1000
6/6 [=========== ] - Os 1ms/step - loss: 22505.2070
Epoch 142/1000
6/6 [============ ] - Os 1ms/step - loss: 22085.3105
Epoch 143/1000
6/6 [============ ] - Os 1ms/step - loss: 20141.1484
```

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Epoch 144/1000
Epoch 145/1000
Epoch 146/1000
Epoch 147/1000
6/6 [============== ] - Os 1ms/step - loss: 15363.7773
Epoch 148/1000
Epoch 149/1000
6/6 [============ ] - Os 1ms/step - loss: 15483.8867
Epoch 150/1000
6/6 [============ ] - Os 1ms/step - loss: 15891.8545
Epoch 151/1000
6/6 [============= ] - 0s 997us/step - loss: 16068.6016
Epoch 152/1000
6/6 [============ ] - Os 1ms/step - loss: 15556.2939
Epoch 153/1000
- loss: 15509.9443
Epoch 154/1000
6/6 [============== ] - Os 1ms/step - loss: 15357.3076
Epoch 155/1000
Epoch 156/1000
6/6 [============ ] - Os 1ms/step - loss: 15465.8242
Epoch 157/1000
6/6 [============ ] - Os 1ms/step - loss: 15093.8496
Epoch 158/1000
6/6 [============ ] - Os 1ms/step - loss: 15582.2275
Epoch 159/1000
6/6 [============ ] - Os 1ms/step - loss: 16281.1104
Epoch 160/1000
Epoch 161/1000
Epoch 162/1000
Epoch 163/1000
Epoch 164/1000
6/6 [============ ] - Os 1ms/step - loss: 15228.3027
Epoch 165/1000
6/6 [============ ] - Os 1ms/step - loss: 15359.9941
Epoch 166/1000
6/6 [============ ] - Os 1ms/step - loss: 14857.3027
Epoch 167/1000
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6/6 [============ ] - Os 1ms/step - loss: 15554.4199
Epoch 168/1000
Epoch 169/1000
Epoch 170/1000
Epoch 171/1000
6/6 [=========== ] - Os 1ms/step - loss: 16257.2305
Epoch 172/1000
6/6 [============ ] - Os 1ms/step - loss: 15468.8291
Epoch 173/1000
6/6 [============ ] - Os 1ms/step - loss: 14815.4619
Epoch 174/1000
6/6 [============ ] - Os 1ms/step - loss: 15647.1572
Epoch 175/1000
Epoch 176/1000
Epoch 177/1000
6/6 [=========== ] - Os 1ms/step - loss: 14808.6045
Epoch 178/1000
6/6 [============== ] - Os 1ms/step - loss: 14499.8818
Epoch 179/1000
Epoch 180/1000
6/6 [============ ] - Os 1ms/step - loss: 15713.4717
Epoch 181/1000
6/6 [=========== ] - Os 1ms/step - loss: 15961.6924
Epoch 182/1000
6/6 [============ ] - Os 1ms/step - loss: 14144.7744
Epoch 183/1000
6/6 [============ ] - Os 1ms/step - loss: 14452.3564
Epoch 184/1000
Epoch 185/1000
Epoch 186/1000
Epoch 187/1000
Epoch 188/1000
Epoch 189/1000
6/6 [============ ] - Os 1ms/step - loss: 14983.9258
Epoch 190/1000
6/6 [============ ] - Os 1ms/step - loss: 16976.7207
Epoch 191/1000
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6/6 [============ ] - Os 1ms/step - loss: 15730.7285
Epoch 192/1000
6/6 [============ ] - Os 1ms/step - loss: 14595.5020
Epoch 193/1000
Epoch 194/1000
Epoch 195/1000
6/6 [=========== ] - Os 1ms/step - loss: 13747.7227
Epoch 196/1000
6/6 [============ ] - Os 1ms/step - loss: 14189.1621
Epoch 197/1000
6/6 [============ ] - Os 1ms/step - loss: 16134.8789
Epoch 198/1000
6/6 [============ ] - Os 1ms/step - loss: 14624.0352
Epoch 199/1000
6/6 [============ ] - Os 997us/step - loss: 14015.4238
Epoch 200/1000
Epoch 201/1000
Epoch 202/1000
6/6 [============== ] - Os 1ms/step - loss: 14729.4805
Epoch 203/1000
6/6 [============== ] - Os 1ms/step - loss: 15205.2012
Epoch 204/1000
Epoch 205/1000
6/6 [============ ] - Os 1ms/step - loss: 14471.7754
Epoch 206/1000
Epoch 207/1000
Epoch 208/1000
Epoch 209/1000
Epoch 210/1000
Epoch 211/1000
Epoch 212/1000
6/6 [============ ] - Os 1ms/step - loss: 14201.7715
Epoch 213/1000
6/6 [============ ] - Os 1ms/step - loss: 13428.9336
Epoch 214/1000
6/6 [============ ] - Os 1ms/step - loss: 13935.3359
Epoch 215/1000
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6/6 [============ ] - Os 1ms/step - loss: 13672.4062
Epoch 216/1000
Epoch 217/1000
Epoch 218/1000
Epoch 219/1000
6/6 [=========== ] - Os 1ms/step - loss: 15941.3154
Epoch 220/1000
6/6 [============ ] - Os 1ms/step - loss: 14796.3955
Epoch 221/1000
6/6 [============ ] - Os 1ms/step - loss: 13321.6807
Epoch 222/1000
6/6 [============ ] - Os 1ms/step - loss: 12971.8545
Epoch 223/1000
Epoch 224/1000
Epoch 225/1000
Epoch 226/1000
Epoch 227/1000
6/6 [============== ] - Os 1ms/step - loss: 12812.0244
Epoch 228/1000
Epoch 229/1000
6/6 [============ ] - Os 1ms/step - loss: 13841.3848
Epoch 230/1000
Epoch 231/1000
6/6 [============ ] - Os 1ms/step - loss: 13740.1182
Epoch 232/1000
Epoch 233/1000
6/6 [============ ] - 0s 997us/step - loss: 15386.0088
Epoch 234/1000
Epoch 235/1000
Epoch 236/1000
Epoch 237/1000
Epoch 238/1000
6/6 [============ ] - Os 1ms/step - loss: 13521.4609
Epoch 239/1000
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6/6 [============ ] - Os 1ms/step - loss: 13107.1641
Epoch 240/1000
6/6 [============ ] - Os 2ms/step - loss: 12862.3398
Epoch 241/1000
Epoch 242/1000
Epoch 243/1000
6/6 [=========== ] - Os 2ms/step - loss: 12806.1562
Epoch 244/1000
6/6 [=========== ] - Os 1ms/step - loss: 12990.0186
Epoch 245/1000
Epoch 246/1000
6/6 [============ ] - Os 1ms/step - loss: 14711.0107
Epoch 247/1000
6/6 [============ ] - Os 1ms/step - loss: 13541.1602
Epoch 248/1000
Epoch 249/1000
Epoch 250/1000
6/6 [============== ] - Os 1ms/step - loss: 15250.8438
Epoch 251/1000
Epoch 252/1000
6/6 [============ ] - Os 1ms/step - loss: 12643.0957
Epoch 253/1000
Epoch 254/1000
6/6 [============ ] - Os 1ms/step - loss: 14756.5439
Epoch 255/1000
Epoch 256/1000
Epoch 257/1000
Epoch 258/1000
Epoch 259/1000
Epoch 260/1000
Epoch 261/1000
Epoch 262/1000
Epoch 263/1000
```

```
6/6 [============ ] - Os 1ms/step - loss: 12953.9287
Epoch 264/1000
6/6 [============ ] - Os 1ms/step - loss: 12620.7139
Epoch 265/1000
Epoch 266/1000
Epoch 267/1000
Epoch 268/1000
6/6 [============ ] - Os 1ms/step - loss: 13578.5088
Epoch 269/1000
6/6 [============ ] - Os 1ms/step - loss: 12078.8975
Epoch 270/1000
Epoch 271/1000
6/6 [============ ] - Os 1ms/step - loss: 11850.7188
Epoch 272/1000
Epoch 273/1000
6/6 [=========== ] - Os 1ms/step - loss: 11484.0029
Epoch 274/1000
Epoch 275/1000
Epoch 276/1000
6/6 [============ ] - Os 1ms/step - loss: 14256.4229
Epoch 277/1000
6/6 [=========== ] - Os 1ms/step - loss: 14908.5938
Epoch 278/1000
6/6 [=========== ] - Os 1ms/step - loss: 15308.1270
Epoch 279/1000
6/6 [============ ] - Os 1ms/step - loss: 14467.4365
Epoch 280/1000
Epoch 281/1000
Epoch 282/1000
Epoch 283/1000
Epoch 284/1000
Epoch 285/1000
loss: 13269.2725
Epoch 286/1000
6/6 [============ ] - Os 1ms/step - loss: 12100.6533
```

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Epoch 287/1000
Epoch 288/1000
Epoch 289/1000
Epoch 290/1000
6/6 [============== ] - Os 1ms/step - loss: 13040.8271
Epoch 291/1000
6/6 [============== ] - Os 1ms/step - loss: 12232.6133
Epoch 292/1000
6/6 [============ ] - Os 1ms/step - loss: 14050.2988
Epoch 293/1000
Epoch 294/1000
Epoch 295/1000
Epoch 296/1000
Epoch 297/1000
Epoch 298/1000
Epoch 299/1000
Epoch 300/1000
6/6 [============ ] - Os 1ms/step - loss: 11357.9746
Epoch 301/1000
Epoch 302/1000
Epoch 303/1000
Epoch 304/1000
Epoch 305/1000
6/6 [============== ] - Os 1ms/step - loss: 13532.8213
Epoch 306/1000
Epoch 307/1000
6/6 [============ ] - Os 1ms/step - loss: 11398.4990
Epoch 308/1000
6/6 [============ ] - Os 1ms/step - loss: 11227.1035
Epoch 309/1000
Epoch 310/1000
6/6 [============ ] - Os 1ms/step - loss: 11647.1221
```

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Epoch 311/1000
Epoch 312/1000
Epoch 313/1000
6/6 [============== ] - Os 1ms/step - loss: 11086.9844
Epoch 314/1000
Epoch 315/1000
Epoch 316/1000
Epoch 317/1000
Epoch 318/1000
6/6 [=========== ] - Os 1ms/step - loss: 11272.1904
Epoch 319/1000
6/6 [============ ] - Os 1ms/step - loss: 11356.4668
Epoch 320/1000
6/6 [============== ] - Os 997us/step - loss: 11952.4814
Epoch 321/1000
Epoch 322/1000
6/6 [============= ] - Os 997us/step - loss: 10417.7998
Epoch 323/1000
6/6 [============ ] - Os 1ms/step - loss: 15771.0166
Epoch 324/1000
6/6 [============ ] - Os 1ms/step - loss: 12541.9609
Epoch 325/1000
6/6 [=========== ] - Os 1ms/step - loss: 10244.7324
Epoch 326/1000
Epoch 327/1000
Epoch 328/1000
6/6 [============= - - os 1ms/step - loss: 11035.5244
Epoch 329/1000
Epoch 330/1000
Epoch 331/1000
Epoch 332/1000
6/6 [============ ] - Os 1ms/step - loss: 10472.4639
Epoch 333/1000
6/6 [============ ] - Os 3ms/step - loss: 10859.7109
Epoch 334/1000
6/6 [=========== ] - Os 1ms/step - loss: 10340.2324
```

```
Epoch 335/1000
6/6 [============ ] - Os 1ms/step - loss: 10599.7012
Epoch 336/1000
Epoch 337/1000
6/6 [========== ] - Os 1ms/step - loss: 10259.2520
Epoch 338/1000
6/6 [============== ] - Os 1ms/step - loss: 10264.0527
Epoch 339/1000
6/6 [============== ] - Os 1ms/step - loss: 11851.4746
Epoch 340/1000
6/6 [============ ] - Os 1ms/step - loss: 12217.6367
Epoch 341/1000
6/6 [============ ] - Os 1ms/step - loss: 11016.7588
Epoch 342/1000
6/6 [============ ] - Os 1ms/step - loss: 12512.7598
Epoch 343/1000
6/6 [============ ] - Os 1ms/step - loss: 12854.5791
Epoch 344/1000
Epoch 345/1000
Epoch 346/1000
Epoch 347/1000
6/6 [============ ] - Os 1ms/step - loss: 12854.4209
Epoch 348/1000
loss: 12150.0645
Epoch 349/1000
6/6 [============ ] - Os 1ms/step - loss: 11523.0645
Epoch 350/1000
6/6 [============ ] - Os 1ms/step - loss: 10361.3135
Epoch 351/1000
Epoch 352/1000
Epoch 353/1000
Epoch 354/1000
Epoch 355/1000
Epoch 356/1000
Epoch 357/1000
6/6 [============ ] - 0s 1ms/step - loss: 9397.1104
Epoch 358/1000
```

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6/6 [============ ] - Os 1ms/step - loss: 10037.5840
Epoch 359/1000
Epoch 360/1000
Epoch 361/1000
Epoch 362/1000
Epoch 363/1000
Epoch 364/1000
6/6 [============ ] - 0s 1ms/step - loss: 9567.1973
Epoch 365/1000
6/6 [============ ] - 0s 1ms/step - loss: 9439.9668
Epoch 366/1000
Epoch 367/1000
Epoch 368/1000
Epoch 369/1000
6/6 [============== ] - Os 1ms/step - loss: 10266.2031
Epoch 370/1000
6/6 [============== ] - Os 1ms/step - loss: 13758.7393
Epoch 371/1000
6/6 [============ ] - Os 1ms/step - loss: 12947.1641
Epoch 372/1000
Epoch 373/1000
Epoch 374/1000
Epoch 375/1000
Epoch 376/1000
Epoch 377/1000
Epoch 378/1000
Epoch 379/1000
6/6 [============= ] - 0s 1ms/step - loss: 9035.4141
Epoch 380/1000
6/6 [============ ] - 0s 1ms/step - loss: 9171.3496
Epoch 381/1000
loss: 9025.9502
```

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Epoch 382/1000
6/6 [============ ] - 0s 1ms/step - loss: 8843.3496
Epoch 383/1000
Epoch 384/1000
6/6 [============= ] - Os 1ms/step - loss: 9243.1758
Epoch 385/1000
6/6 [============= ] - 0s 1ms/step - loss: 9393.1465
Epoch 386/1000
6/6 [============= ] - 0s 1ms/step - loss: 9860.5322
Epoch 387/1000
6/6 [============ ] - Os 1ms/step - loss: 12479.7393
Epoch 388/1000
Epoch 389/1000
6/6 [============ ] - 0s 1ms/step - loss: 8605.0762
Epoch 390/1000
6/6 [============ ] - 0s 1ms/step - loss: 9438.4609
Epoch 391/1000
6/6 [============== ] - Os 1ms/step - loss: 8579.9541
Epoch 392/1000
Epoch 393/1000
Epoch 394/1000
6/6 [============ ] - 0s 1ms/step - loss: 9044.2607
Epoch 395/1000
Epoch 396/1000
6/6 [============ ] - 0s 1ms/step - loss: 9331.6846
Epoch 397/1000
Epoch 398/1000
Epoch 399/1000
Epoch 400/1000
Epoch 401/1000
6/6 [============= ] - 0s 1ms/step - loss: 9576.9072
Epoch 402/1000
6/6 [============ ] - Os 1ms/step - loss: 13493.5518
Epoch 403/1000
Epoch 404/1000
6/6 [============ ] - Os 1ms/step - loss: 12049.3291
Epoch 405/1000
6/6 [============ ] - Os 1ms/step - loss: 11945.1221
```

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Epoch 406/1000
6/6 [=========== ] - Os 1ms/step - loss: 16662.9980
Epoch 407/1000
Epoch 408/1000
Epoch 409/1000
6/6 [============== ] - Os 1ms/step - loss: 17075.2891
Epoch 410/1000
6/6 [============== ] - Os 1ms/step - loss: 11638.3115
Epoch 411/1000
6/6 [============ ] - Os 1ms/step - loss: 12642.1895
Epoch 412/1000
Epoch 413/1000
6/6 [============ ] - Os 1ms/step - loss: 10119.7266
Epoch 414/1000
Epoch 415/1000
Epoch 416/1000
6/6 [============== ] - Os 1ms/step - loss: 8492.9443
Epoch 417/1000
Epoch 418/1000
Epoch 419/1000
Epoch 420/1000
Epoch 421/1000
Epoch 422/1000
Epoch 423/1000
6/6 [============== ] - Os 1ms/step - loss: 8261.5957
Epoch 424/1000
6/6 [============= ] - 0s 1ms/step - loss: 7983.0923
Epoch 425/1000
6/6 [============= ] - 0s 1ms/step - loss: 8329.6338
Epoch 426/1000
Epoch 427/1000
Epoch 428/1000
6/6 [============ ] - 0s 1ms/step - loss: 8295.8564
Epoch 429/1000
```

```
Epoch 430/1000
6/6 [============= ] - 0s 1ms/step - loss: 8648.2637
Epoch 431/1000
Epoch 432/1000
Epoch 433/1000
6/6 [============= ] - 0s 1ms/step - loss: 9598.7656
Epoch 434/1000
6/6 [============= ] - 0s 1ms/step - loss: 8403.4141
Epoch 435/1000
6/6 [=========== ] - Os 1ms/step - loss: 8823.9443
Epoch 436/1000
6/6 [============ ] - 0s 1ms/step - loss: 8181.0479
Epoch 437/1000
Epoch 438/1000
Epoch 439/1000
6/6 [============== ] - Os 1ms/step - loss: 9755.5322
Epoch 440/1000
Epoch 441/1000
6/6 [============= - - os 1ms/step - loss: 10845.9082
Epoch 442/1000
6/6 [============ ] - Os 1ms/step - loss: 10452.5010
Epoch 443/1000
Epoch 444/1000
Epoch 445/1000
6/6 [=========== ] - Os 997us/step - loss: 7741.6646
Epoch 446/1000
Epoch 447/1000
6/6 [============== ] - Os 1ms/step - loss: 7910.3696
Epoch 448/1000
6/6 [============== ] - 0s 1ms/step - loss: 8104.8145
Epoch 449/1000
Epoch 450/1000
6/6 [=========== ] - Os 1ms/step - loss: 9204.8428
Epoch 451/1000
Epoch 452/1000
6/6 [============ ] - Os 1ms/step - loss: 10509.7012
Epoch 453/1000
6/6 [=========== ] - 0s 1ms/step - loss: 8778.6846
```

```
Epoch 454/1000
Epoch 455/1000
Epoch 456/1000
6/6 [============== ] - Os 1ms/step - loss: 8037.2280
Epoch 457/1000
6/6 [============== ] - Os 1ms/step - loss: 10377.8486
Epoch 458/1000
Epoch 459/1000
6/6 [=========== ] - Os 1ms/step - loss: 8593.9678
Epoch 460/1000
6/6 [=========== ] - Os 1ms/step - loss: 7608.7798
Epoch 461/1000
Epoch 462/1000
Epoch 463/1000
6/6 [============== ] - Os 1ms/step - loss: 7904.4536
Epoch 464/1000
6/6 [============= ] - Os 1ms/step - loss: 7399.3242
Epoch 465/1000
Epoch 466/1000
Epoch 467/1000
6/6 [============= ] - 0s 1ms/step - loss: 8184.0581
Epoch 468/1000
Epoch 469/1000
Epoch 470/1000
Epoch 471/1000
Epoch 472/1000
Epoch 473/1000
6/6 [============== ] - Os 1ms/step - loss: 8537.2910
Epoch 474/1000
6/6 [=========== ] - Os 2ms/step - loss: 7983.4688
Epoch 475/1000
6/6 [============= ] - 0s 1ms/step - loss: 9141.1768
Epoch 476/1000
6/6 [============= ] - 0s 1ms/step - loss: 8800.8174
Epoch 477/1000
6/6 [=========== ] - 0s 1ms/step - loss: 8254.7461
```

```
Epoch 478/1000
Epoch 479/1000
Epoch 480/1000
6/6 [============== ] - Os 1ms/step - loss: 7137.6733
Epoch 481/1000
Epoch 482/1000
6/6 [============== ] - 0s 1ms/step - loss: 7586.7969
Epoch 483/1000
6/6 [============ ] - Os 1ms/step - loss: 8525.4912
Epoch 484/1000
6/6 [=================== ] - Os 1ms/step - loss: 7474.1318
Epoch 485/1000
6/6 [=========== ] - 0s 1ms/step - loss: 7473.6343
Epoch 486/1000
Epoch 487/1000
6/6 [============= ] - Os 1ms/step - loss: 6642.5864
Epoch 488/1000
Epoch 489/1000
Epoch 490/1000
6/6 [============= - 0s 1ms/step - loss: 7844.5332
Epoch 491/1000
Epoch 492/1000
6/6 [============ ] - 0s 1ms/step - loss: 7032.7227
Epoch 493/1000
Epoch 494/1000
Epoch 495/1000
Epoch 496/1000
6/6 [============= ] - 0s 1ms/step - loss: 7376.6426
Epoch 497/1000
6/6 [============= ] - 0s 1ms/step - loss: 8734.7715
Epoch 498/1000
6/6 [=========== ] - Os 1ms/step - loss: 7669.9258
Epoch 499/1000
Epoch 500/1000
Epoch 501/1000
6/6 [=========== ] - 0s 1ms/step - loss: 6682.8540
```

```
Epoch 502/1000
Epoch 503/1000
Epoch 504/1000
6/6 [============= ] - Os 1ms/step - loss: 6684.3984
Epoch 505/1000
6/6 [=============== ] - Os 1ms/step - loss: 17611.3691
Epoch 506/1000
6/6 [============== ] - Os 1ms/step - loss: 12979.6367
Epoch 507/1000
6/6 [============ ] - Os 1ms/step - loss: 10141.7988
Epoch 508/1000
Epoch 509/1000
6/6 [============ ] - Os 1ms/step - loss: 13067.8262
Epoch 510/1000
6/6 [============ ] - 0s 1ms/step - loss: 8841.4951
Epoch 511/1000
6/6 [============== ] - Os 1ms/step - loss: 6892.3706
Epoch 512/1000
6/6 [============== ] - Os 1ms/step - loss: 8233.1201
Epoch 513/1000
Epoch 514/1000
6/6 [=========== ] - 0s 1ms/step - loss: 7748.8613
Epoch 515/1000
6/6 [============ ] - 0s 1ms/step - loss: 7726.7900
Epoch 516/1000
Epoch 517/1000
6/6 [=========== ] - Os 1ms/step - loss: 6125.9624
Epoch 518/1000
Epoch 519/1000
6/6 [============= ] - Os 1ms/step - loss: 6303.7070
Epoch 520/1000
Epoch 521/1000
Epoch 522/1000
6/6 [=========== ] - Os 1ms/step - loss: 6800.2197
Epoch 523/1000
6/6 [============ ] - 0s 1ms/step - loss: 6073.4482
Epoch 524/1000
Epoch 525/1000
```

```
Epoch 526/1000
Epoch 527/1000
Epoch 528/1000
6/6 [============= ] - 0s 1ms/step - loss: 6673.7915
Epoch 529/1000
Epoch 530/1000
Epoch 531/1000
6/6 [=========== ] - Os 1ms/step - loss: 6374.0752
Epoch 532/1000
Epoch 533/1000
6/6 [============ ] - 0s 1ms/step - loss: 9132.9609
Epoch 534/1000
Epoch 535/1000
6/6 [============== ] - Os 1ms/step - loss: 8762.4961
Epoch 536/1000
6/6 [============== ] - Os 1ms/step - loss: 7831.4351
Epoch 537/1000
Epoch 538/1000
6/6 [=========== ] - 0s 1ms/step - loss: 7161.4976
Epoch 539/1000
Epoch 540/1000
Epoch 541/1000
Epoch 542/1000
Epoch 543/1000
6/6 [============== ] - Os 1ms/step - loss: 5609.2964
Epoch 544/1000
6/6 [============= ] - 0s 1ms/step - loss: 5919.0527
Epoch 545/1000
6/6 [============= ] - 0s 1ms/step - loss: 5552.4097
Epoch 546/1000
Epoch 547/1000
Epoch 548/1000
6/6 [============= ] - 0s 1ms/step - loss: 5639.3862
Epoch 549/1000
6/6 [=========== ] - 0s 1ms/step - loss: 5573.2949
```

```
Epoch 550/1000
Epoch 551/1000
Epoch 552/1000
6/6 [============== ] - Os 1ms/step - loss: 6632.3667
Epoch 553/1000
6/6 [============= ] - 0s 1ms/step - loss: 8119.4614
Epoch 554/1000
Epoch 555/1000
6/6 [============ ] - 0s 1ms/step - loss: 5544.2334
Epoch 556/1000
Epoch 557/1000
Epoch 558/1000
Epoch 559/1000
6/6 [============== ] - Os 1ms/step - loss: 6198.1646
Epoch 560/1000
6/6 [============== ] - Os 1ms/step - loss: 9117.0732
Epoch 561/1000
Epoch 562/1000
6/6 [============ ] - 0s 1ms/step - loss: 8132.1221
Epoch 563/1000
Epoch 564/1000
Epoch 565/1000
Epoch 566/1000
Epoch 567/1000
Epoch 568/1000
6/6 [============= ] - 0s 1ms/step - loss: 7119.1372
Epoch 569/1000
6/6 [============== ] - Os 1ms/step - loss: 7909.9106
Epoch 570/1000
6/6 [=========== ] - Os 1ms/step - loss: 5620.6465
Epoch 571/1000
Epoch 572/1000
6/6 [=========== ] - 0s 1ms/step - loss: 7038.8608
Epoch 573/1000
6/6 [=========== ] - 0s 1ms/step - loss: 6423.9282
```

```
Epoch 574/1000
Epoch 575/1000
Epoch 576/1000
6/6 [============== ] - 0s 1ms/step - loss: 7117.5742
Epoch 577/1000
6/6 [============== ] - 0s 1ms/step - loss: 5984.2793
Epoch 578/1000
6/6 [============= ] - 0s 1ms/step - loss: 5302.0396
Epoch 579/1000
Epoch 580/1000
6/6 [=========== ] - Os 1ms/step - loss: 6530.0532
Epoch 581/1000
Epoch 582/1000
Epoch 583/1000
Epoch 584/1000
Epoch 585/1000
6/6 [============ ] - Os 997us/step - loss: 5610.0581
Epoch 586/1000
Epoch 587/1000
6/6 [============ ] - 0s 1ms/step - loss: 5412.4360
Epoch 588/1000
Epoch 589/1000
6/6 [=========== ] - Os 997us/step - loss: 6639.7280
Epoch 590/1000
Epoch 591/1000
Epoch 592/1000
6/6 [============== ] - 0s 1ms/step - loss: 5144.1509
Epoch 593/1000
Epoch 594/1000
Epoch 595/1000
Epoch 596/1000
Epoch 597/1000
6/6 [=========== ] - 0s 1ms/step - loss: 6089.7749
```

```
Epoch 598/1000
Epoch 599/1000
Epoch 600/1000
6/6 [============== ] - Os 1ms/step - loss: 5250.3960
Epoch 601/1000
Epoch 602/1000
6/6 [============= ] - 0s 1ms/step - loss: 6419.3740
Epoch 603/1000
6/6 [=========== ] - Os 1ms/step - loss: 6454.0903
Epoch 604/1000
6/6 [=========== ] - Os 1ms/step - loss: 5344.6514
Epoch 605/1000
6/6 [=========== ] - 0s 1ms/step - loss: 5086.3003
Epoch 606/1000
6/6 [============= ] - 0s 1ms/step - loss: 5041.3311
Epoch 607/1000
6/6 [============== ] - Os 1ms/step - loss: 5077.2104
Epoch 608/1000
Epoch 609/1000
Epoch 610/1000
6/6 [============= ] - 0s 1ms/step - loss: 5145.3081
Epoch 611/1000
Epoch 612/1000
Epoch 613/1000
6/6 [============ ] - Os 997us/step - loss: 5583.1143
Epoch 614/1000
Epoch 615/1000
Epoch 616/1000
6/6 [============= ] - 0s 1ms/step - loss: 5412.9155
Epoch 617/1000
Epoch 618/1000
6/6 [=========== ] - Os 1ms/step - loss: 5712.8740
Epoch 619/1000
Epoch 620/1000
Epoch 621/1000
```

```
Epoch 622/1000
6/6 [============= ] - 0s 1ms/step - loss: 5189.8545
Epoch 623/1000
Epoch 624/1000
6/6 [============== ] - Os 1ms/step - loss: 5771.2529
Epoch 625/1000
6/6 [============= ] - 0s 1ms/step - loss: 7809.4062
Epoch 626/1000
6/6 [============= ] - 0s 1ms/step - loss: 5468.7236
Epoch 627/1000
6/6 [=========== ] - Os 1ms/step - loss: 6217.2197
Epoch 628/1000
6/6 [=========== ] - Os 1ms/step - loss: 5269.7764
Epoch 629/1000
6/6 [============= ] - 0s 1ms/step - loss: 5077.7427
Epoch 630/1000
6/6 [============ ] - 0s 1ms/step - loss: 5089.8584
Epoch 631/1000
6/6 [============== ] - Os 1ms/step - loss: 4904.6162
Epoch 632/1000
6/6 [============== ] - Os 1ms/step - loss: 5526.6260
Epoch 633/1000
Epoch 634/1000
Epoch 635/1000
Epoch 636/1000
Epoch 637/1000
Epoch 638/1000
Epoch 639/1000
Epoch 640/1000
6/6 [============= ] - 0s 1ms/step - loss: 5122.0874
Epoch 641/1000
Epoch 642/1000
6/6 [=========== ] - Os 1ms/step - loss: 4813.9111
Epoch 643/1000
Epoch 644/1000
Epoch 645/1000
```

```
Epoch 646/1000
Epoch 647/1000
Epoch 648/1000
6/6 [============== ] - Os 1ms/step - loss: 5532.6099
Epoch 649/1000
6/6 [============= ] - 0s 1ms/step - loss: 6079.3574
Epoch 650/1000
6/6 [============= ] - 0s 1ms/step - loss: 5168.1309
Epoch 651/1000
Epoch 652/1000
Epoch 653/1000
6/6 [============ ] - 0s 1ms/step - loss: 5430.4937
Epoch 654/1000
6/6 [============= ] - 0s 1ms/step - loss: 5269.2188
Epoch 655/1000
6/6 [============== ] - Os 1ms/step - loss: 5628.1768
Epoch 656/1000
6/6 [============== ] - Os 1ms/step - loss: 5218.8555
Epoch 657/1000
Epoch 658/1000
Epoch 659/1000
Epoch 660/1000
Epoch 661/1000
6/6 [=========== ] - Os 997us/step - loss: 4757.7441
Epoch 662/1000
Epoch 663/1000
Epoch 664/1000
6/6 [============= ] - 0s 1ms/step - loss: 4693.3184
Epoch 665/1000
6/6 [============== ] - Os 1ms/step - loss: 5080.8579
Epoch 666/1000
6/6 [=========== ] - Os 1ms/step - loss: 5091.6309
Epoch 667/1000
6/6 [============ ] - 0s 1ms/step - loss: 5146.4668
Epoch 668/1000
Epoch 669/1000
6/6 [=========== ] - 0s 1ms/step - loss: 5228.7329
```

```
Epoch 670/1000
6/6 [============= ] - 0s 1ms/step - loss: 4639.5552
Epoch 671/1000
Epoch 672/1000
6/6 [============== ] - Os 1ms/step - loss: 5606.2017
Epoch 673/1000
6/6 [============= ] - 0s 1ms/step - loss: 4731.5054
Epoch 674/1000
Epoch 675/1000
6/6 [=========== ] - Os 1ms/step - loss: 5549.3623
Epoch 676/1000
6/6 [=========== ] - Os 1ms/step - loss: 4888.0845
Epoch 677/1000
Epoch 678/1000
Epoch 679/1000
Epoch 680/1000
Epoch 681/1000
Epoch 682/1000
Epoch 683/1000
Epoch 684/1000
6/6 [============ ] - 0s 1ms/step - loss: 7063.5244
Epoch 685/1000
6/6 [=========== ] - Os 997us/step - loss: 6008.0200
Epoch 686/1000
Epoch 687/1000
Epoch 688/1000
Epoch 689/1000
6/6 [============= ] - 0s 1ms/step - loss: 5003.5127
Epoch 690/1000
Epoch 691/1000
Epoch 692/1000
Epoch 693/1000
```

```
Epoch 694/1000
Epoch 695/1000
Epoch 696/1000
6/6 [============== ] - Os 1ms/step - loss: 5066.5210
Epoch 697/1000
Epoch 698/1000
Epoch 699/1000
Epoch 700/1000
6/6 [=========== ] - Os 1ms/step - loss: 5345.4673
Epoch 701/1000
6/6 [============ ] - 0s 1ms/step - loss: 5041.3613
Epoch 702/1000
Epoch 703/1000
Epoch 704/1000
Epoch 705/1000
Epoch 706/1000
Epoch 707/1000
Epoch 708/1000
Epoch 709/1000
6/6 [=========== ] - Os 997us/step - loss: 7121.0259
Epoch 710/1000
Epoch 711/1000
Epoch 712/1000
6/6 [============= ] - 0s 1ms/step - loss: 4505.7759
Epoch 713/1000
6/6 [============= ] - 0s 1ms/step - loss: 4617.8809
Epoch 714/1000
Epoch 715/1000
6/6 [============ ] - 0s 1ms/step - loss: 7029.2842
Epoch 716/1000
Epoch 717/1000
```

```
Epoch 718/1000
6/6 [=========== ] - 0s 1ms/step - loss: 4657.6113
Epoch 719/1000
Epoch 720/1000
Epoch 721/1000
6/6 [============= ] - 0s 1ms/step - loss: 5301.2344
Epoch 722/1000
6/6 [============= ] - 0s 1ms/step - loss: 4779.8887
Epoch 723/1000
6/6 [=========== ] - Os 1ms/step - loss: 4815.9224
Epoch 724/1000
Epoch 725/1000
Epoch 726/1000
Epoch 727/1000
Epoch 728/1000
Epoch 729/1000
Epoch 730/1000
Epoch 731/1000
Epoch 732/1000
6/6 [============ ] - 0s 1ms/step - loss: 7161.2407
Epoch 733/1000
Epoch 734/1000
Epoch 735/1000
Epoch 736/1000
6/6 [============= ] - 0s 1ms/step - loss: 5482.0898
Epoch 737/1000
Epoch 738/1000
6/6 [=========== ] - Os 1ms/step - loss: 7013.8198
Epoch 739/1000
Epoch 740/1000
Epoch 741/1000
6/6 [=========== ] - 0s 1ms/step - loss: 4734.4580
```

```
Epoch 742/1000
Epoch 743/1000
Epoch 744/1000
6/6 [============== ] - Os 1ms/step - loss: 6051.2070
Epoch 745/1000
6/6 [============== ] - 0s 1ms/step - loss: 6671.3721
Epoch 746/1000
6/6 [============= ] - 0s 1ms/step - loss: 8920.5625
Epoch 747/1000
6/6 [============ ] - Os 1ms/step - loss: 10742.9873
Epoch 748/1000
6/6 [=========== ] - Os 1ms/step - loss: 7759.6045
Epoch 749/1000
6/6 [============ ] - 0s 1ms/step - loss: 5419.2153
Epoch 750/1000
Epoch 751/1000
Epoch 752/1000
Epoch 753/1000
Epoch 754/1000
Epoch 755/1000
6/6 [============ ] - 0s 1ms/step - loss: 5278.7461
Epoch 756/1000
6/6 [============ ] - 0s 1ms/step - loss: 5345.9438
Epoch 757/1000
6/6 [============ ] - Os 997us/step - loss: 6948.5273
Epoch 758/1000
Epoch 759/1000
6/6 [============== ] - Os 1ms/step - loss: 5286.7529
Epoch 760/1000
6/6 [============= ] - 0s 1ms/step - loss: 4190.5938
Epoch 761/1000
6/6 [============= ] - 0s 1ms/step - loss: 4186.5659
Epoch 762/1000
6/6 [=========== ] - Os 1ms/step - loss: 4473.8564
Epoch 763/1000
Epoch 764/1000
Epoch 765/1000
6/6 [=========== ] - 0s 1ms/step - loss: 4695.5156
```

```
Epoch 766/1000
6/6 [============ ] - 0s 1ms/step - loss: 4951.2427
Epoch 767/1000
Epoch 768/1000
6/6 [============== ] - Os 1ms/step - loss: 4453.9258
Epoch 769/1000
6/6 [============= ] - 0s 1ms/step - loss: 4411.0376
Epoch 770/1000
6/6 [============= - 0s 1ms/step - loss: 4832.7358
Epoch 771/1000
Epoch 772/1000
6/6 [=========== ] - Os 1ms/step - loss: 4272.0186
Epoch 773/1000
Epoch 774/1000
6/6 [=========== ] - 0s 1ms/step - loss: 4943.2803
Epoch 775/1000
6/6 [============== ] - Os 1ms/step - loss: 4219.0269
Epoch 776/1000
6/6 [============== ] - Os 1ms/step - loss: 6118.5186
Epoch 777/1000
6/6 [=========== - - 0s 1ms/step - loss: 5917.2158
Epoch 778/1000
6/6 [============ ] - 0s 1ms/step - loss: 5842.4941
Epoch 779/1000
Epoch 780/1000
Epoch 781/1000
Epoch 782/1000
Epoch 783/1000
Epoch 784/1000
Epoch 785/1000
Epoch 786/1000
Epoch 787/1000
Epoch 788/1000
6/6 [=========== ] - 0s 1ms/step - loss: 7067.6206
Epoch 789/1000
6/6 [============ ] - 0s 1ms/step - loss: 4622.6377
```

```
Epoch 790/1000
6/6 [============ ] - 0s 1ms/step - loss: 4462.8027
Epoch 791/1000
Epoch 792/1000
Epoch 793/1000
6/6 [============= ] - 0s 1ms/step - loss: 4379.1143
Epoch 794/1000
Epoch 795/1000
6/6 [=========== ] - Os 1ms/step - loss: 4485.9233
Epoch 796/1000
6/6 [=========== - Os 1ms/step - loss: 6860.6235
Epoch 797/1000
Epoch 798/1000
Epoch 799/1000
6/6 [============== ] - Os 1ms/step - loss: 5952.7690
Epoch 800/1000
6/6 [============== ] - Os 1ms/step - loss: 6361.3979
Epoch 801/1000
Epoch 802/1000
6/6 [=========== ] - 0s 1ms/step - loss: 7405.6982
Epoch 803/1000
Epoch 804/1000
6/6 [============ ] - 0s 1ms/step - loss: 5578.6040
Epoch 805/1000
6/6 [=========== ] - Os 997us/step - loss: 5941.5498
Epoch 806/1000
Epoch 807/1000
6/6 [============== ] - Os 1ms/step - loss: 5545.6963
Epoch 808/1000
6/6 [============= ] - 0s 1ms/step - loss: 4599.0659
Epoch 809/1000
6/6 [============= ] - 0s 1ms/step - loss: 4662.7656
Epoch 810/1000
6/6 [=========== ] - Os 1ms/step - loss: 5201.0708
Epoch 811/1000
Epoch 812/1000
Epoch 813/1000
```

```
Epoch 814/1000
Epoch 815/1000
Epoch 816/1000
6/6 [============== ] - Os 1ms/step - loss: 5895.3706
Epoch 817/1000
6/6 [============== ] - 0s 1ms/step - loss: 5763.7729
Epoch 818/1000
6/6 [============= ] - 0s 1ms/step - loss: 5845.7197
Epoch 819/1000
6/6 [=========== ] - Os 1ms/step - loss: 4939.8843
Epoch 820/1000
Epoch 821/1000
Epoch 822/1000
Epoch 823/1000
6/6 [============== ] - Os 1ms/step - loss: 4176.1777
Epoch 824/1000
6/6 [============== ] - Os 1ms/step - loss: 4239.1372
Epoch 825/1000
Epoch 826/1000
6/6 [============ ] - 0s 1ms/step - loss: 4167.7305
Epoch 827/1000
6/6 [=========== ] - 0s 1ms/step - loss: 4020.5583
Epoch 828/1000
Epoch 829/1000
Epoch 830/1000
Epoch 831/1000
Epoch 832/1000
Epoch 833/1000
Epoch 834/1000
Epoch 835/1000
Epoch 836/1000
6/6 [============ - 0s 1ms/step - loss: 5966.1582
Epoch 837/1000
```

```
Epoch 838/1000
Epoch 839/1000
Epoch 840/1000
6/6 [============= ] - Os 1ms/step - loss: 4294.5757
Epoch 841/1000
6/6 [============= ] - 0s 1ms/step - loss: 4640.5078
Epoch 842/1000
6/6 [============= ] - 0s 1ms/step - loss: 7136.1899
Epoch 843/1000
6/6 [=========== ] - Os 1ms/step - loss: 4480.5718
Epoch 844/1000
6/6 [=========== ] - 0s 1ms/step - loss: 4198.8413
Epoch 845/1000
Epoch 846/1000
6/6 [============ ] - 0s 1ms/step - loss: 5236.9639
Epoch 847/1000
6/6 [============= ] - Os 1ms/step - loss: 4375.5303
Epoch 848/1000
6/6 [============== ] - Os 1ms/step - loss: 3871.9561
Epoch 849/1000
Epoch 850/1000
6/6 [============= ] - 0s 1ms/step - loss: 4670.1597
Epoch 851/1000
Epoch 852/1000
6/6 [============ ] - 0s 1ms/step - loss: 4055.8564
Epoch 853/1000
Epoch 854/1000
Epoch 855/1000
6/6 [============== ] - Os 1ms/step - loss: 5836.2783
Epoch 856/1000
6/6 [============= ] - 0s 1ms/step - loss: 4898.4458
Epoch 857/1000
6/6 [============= ] - 0s 1ms/step - loss: 4189.0317
Epoch 858/1000
6/6 [=========== ] - Os 1ms/step - loss: 4229.8032
Epoch 859/1000
Epoch 860/1000
6/6 [============ ] - 0s 1ms/step - loss: 3968.3008
Epoch 861/1000
6/6 [=========== ] - 0s 1ms/step - loss: 4191.9380
```

```
Epoch 862/1000
Epoch 863/1000
Epoch 864/1000
Epoch 865/1000
6/6 [============== ] - 0s 1ms/step - loss: 4288.9517
Epoch 866/1000
6/6 [============= ] - 0s 1ms/step - loss: 5346.4932
Epoch 867/1000
6/6 [=========== ] - Os 1ms/step - loss: 5302.3892
Epoch 868/1000
6/6 [=========== ] - Os 1ms/step - loss: 4043.3682
Epoch 869/1000
6/6 [============ ] - 0s 1ms/step - loss: 4150.2920
Epoch 870/1000
Epoch 871/1000
6/6 [============== ] - Os 1ms/step - loss: 5056.9316
Epoch 872/1000
Epoch 873/1000
Epoch 874/1000
Epoch 875/1000
Epoch 876/1000
Epoch 877/1000
6/6 [=========== ] - Os 1ms/step - loss: 5440.8369
Epoch 878/1000
Epoch 879/1000
6/6 [============== ] - Os 1ms/step - loss: 4424.7788
Epoch 880/1000
6/6 [============= ] - 0s 1ms/step - loss: 4174.9849
Epoch 881/1000
Epoch 882/1000
6/6 [=========== ] - Os 1ms/step - loss: 4231.8042
Epoch 883/1000
Epoch 884/1000
6/6 [============ ] - 0s 1ms/step - loss: 4487.2339
Epoch 885/1000
```

```
Epoch 886/1000
Epoch 887/1000
Epoch 888/1000
Epoch 889/1000
Epoch 890/1000
6/6 [============= ] - 0s 1ms/step - loss: 4266.0850
Epoch 891/1000
6/6 [=========== ] - Os 1ms/step - loss: 4578.7339
Epoch 892/1000
Epoch 893/1000
6/6 [============ ] - 0s 1ms/step - loss: 4030.5740
Epoch 894/1000
6/6 [============ ] - 0s 1ms/step - loss: 4469.2915
Epoch 895/1000
6/6 [============== ] - Os 1ms/step - loss: 5141.6846
Epoch 896/1000
Epoch 897/1000
Epoch 898/1000
6/6 [============ ] - 0s 1ms/step - loss: 3921.6245
Epoch 899/1000
Epoch 900/1000
Epoch 901/1000
Epoch 902/1000
Epoch 903/1000
Epoch 904/1000
6/6 [============= ] - 0s 1ms/step - loss: 4177.2437
Epoch 905/1000
6/6 [============= ] - 0s 1ms/step - loss: 4000.7993
Epoch 906/1000
Epoch 907/1000
6/6 [============ ] - 0s 1ms/step - loss: 4461.6904
Epoch 908/1000
6/6 [============ ] - 0s 1ms/step - loss: 4044.1470
Epoch 909/1000
6/6 [=========== ] - 0s 1ms/step - loss: 3943.0125
```

```
Epoch 910/1000
6/6 [============ ] - 0s 1ms/step - loss: 5530.7002
Epoch 911/1000
Epoch 912/1000
Epoch 913/1000
6/6 [============= ] - 0s 1ms/step - loss: 6839.5298
Epoch 914/1000
6/6 [============= ] - 0s 1ms/step - loss: 5893.9434
Epoch 915/1000
Epoch 916/1000
6/6 [============ ] - 0s 1ms/step - loss: 4138.8022
Epoch 917/1000
Epoch 918/1000
loss: 7841.9067
Epoch 919/1000
Epoch 920/1000
6/6 [============== ] - Os 1ms/step - loss: 4501.6372
Epoch 921/1000
6/6 [============== ] - 0s 1ms/step - loss: 5010.3721
Epoch 922/1000
Epoch 923/1000
6/6 [============ ] - 0s 1ms/step - loss: 4189.9009
Epoch 924/1000
Epoch 925/1000
Epoch 926/1000
Epoch 927/1000
6/6 [============== ] - Os 1ms/step - loss: 4279.0869
Epoch 928/1000
Epoch 929/1000
Epoch 930/1000
6/6 [============ ] - 0s 1ms/step - loss: 7769.2979
Epoch 931/1000
6/6 [============ ] - 0s 1ms/step - loss: 5280.3242
Epoch 932/1000
Epoch 933/1000
```

```
Epoch 934/1000
Epoch 935/1000
Epoch 936/1000
6/6 [============= ] - Os 1ms/step - loss: 4280.1963
Epoch 937/1000
Epoch 938/1000
6/6 [============ ] - 0s 1ms/step - loss: 4612.5010
Epoch 939/1000
6/6 [============ ] - 0s 1ms/step - loss: 4469.4043
Epoch 940/1000
6/6 [=========== ] - 0s 1ms/step - loss: 4832.6353
Epoch 941/1000
Epoch 942/1000
Epoch 943/1000
6/6 [============= ] - Os 1ms/step - loss: 4548.0220
Epoch 944/1000
6/6 [============== ] - Os 1ms/step - loss: 7044.5767
Epoch 945/1000
6/6 [============= ] - 0s 1ms/step - loss: 4329.9844
Epoch 946/1000
6/6 [============ ] - 0s 1ms/step - loss: 3919.6667
Epoch 947/1000
Epoch 948/1000
Epoch 949/1000
Epoch 950/1000
6/6 [=========== - - 0s 1ms/step - loss: 4999.4072
Epoch 951/1000
Epoch 952/1000
Epoch 953/1000
Epoch 954/1000
Epoch 955/1000
Epoch 956/1000
Epoch 957/1000
```

```
Epoch 958/1000
6/6 [============ ] - 0s 1ms/step - loss: 5631.0825
Epoch 959/1000
Epoch 960/1000
6/6 [============== ] - Os 1ms/step - loss: 4687.3828
Epoch 961/1000
Epoch 962/1000
6/6 [============ ] - 0s 1ms/step - loss: 4002.3069
Epoch 963/1000
Epoch 964/1000
6/6 [============ ] - 0s 1ms/step - loss: 3797.7795
Epoch 965/1000
Epoch 966/1000
Epoch 967/1000
6/6 [============= ] - Os 1ms/step - loss: 3890.4924
Epoch 968/1000
6/6 [============== ] - Os 1ms/step - loss: 3847.0867
Epoch 969/1000
6/6 [============= ] - 0s 1ms/step - loss: 4687.7559
Epoch 970/1000
Epoch 971/1000
Epoch 972/1000
6/6 [============ ] - 0s 1ms/step - loss: 5319.4248
Epoch 973/1000
Epoch 974/1000
Epoch 975/1000
6/6 [============== ] - Os 1ms/step - loss: 4606.7139
Epoch 976/1000
Epoch 977/1000
Epoch 978/1000
6/6 [============ ] - 0s 1ms/step - loss: 5039.9160
Epoch 979/1000
Epoch 980/1000
Epoch 981/1000
```

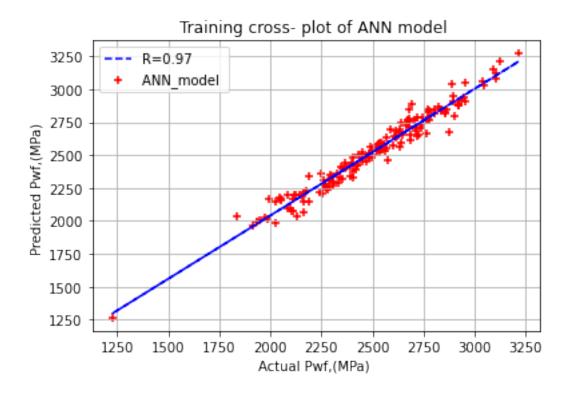
```
Epoch 982/1000
  6/6 [========== ] - 0s 1ms/step - loss: 3898.7292
  Epoch 983/1000
  Epoch 984/1000
  Epoch 985/1000
  Epoch 986/1000
  6/6 [============ ] - 0s 1ms/step - loss: 3815.9814
  Epoch 987/1000
  Epoch 988/1000
  6/6 [============ ] - 0s 1ms/step - loss: 3853.8250
  Epoch 989/1000
  Epoch 990/1000
  Epoch 991/1000
  Epoch 992/1000
  6/6 [============= ] - Os 1ms/step - loss: 3752.2451
  Epoch 993/1000
  6/6 [=========== - - 0s 1ms/step - loss: 3770.3982
  Epoch 994/1000
  6/6 [============ ] - 0s 1ms/step - loss: 3736.6614
  Epoch 995/1000
  Epoch 996/1000
  6/6 [============ ] - 0s 1ms/step - loss: 3686.4458
  Epoch 997/1000
  6/6 [============= ] - 0s 1ms/step - loss: 3645.4326
  Epoch 998/1000
  Epoch 999/1000
  Epoch 1000/1000
  [131]: <tensorflow.python.keras.callbacks.History at 0x1aa30bdfbb0>
  sns.lineplot(x=range(len(loss)),y=loss) plt.title("Training Loss per Epoch");
```

## 6 Compare final evaluation (MSE) on training set and test set

```
[134]: #test_predictions
[135]: from sklearn.metrics import
        →mean_squared_error,mean_absolute_error,explained_variance_score
[136]: predictions_test = model.predict(X_test)
[137]: predictions_train = model.predict(X_train)
[138]: # Create a scatterplot of the real test values versus the predicted values.
       RFtest=plt.
       →scatter(y_train,predictions_train,color='red',marker='+',label='ANN_model ')
       plt.xlabel(('Actual Pwf,(MPa) '))
       plt.ylabel(('Predicted Pwf,(MPa) '))
       R = r2_score(y_train, predictions_train)
       plt.title('Training cross- plot of ANN model ')
       plt.grid()
       print(R)
       m, b = np.polyfit(y_train, predictions_train, 1)
       plt.plot(y_train, m*y_train+b,linestyle='--',c='blue',label='R=0.97')
       plt.legend()
       #plt.savefig('ANN-test')
```

## 0.9561095299937702

[138]: <matplotlib.legend.Legend at 0x1aa2d276df0>

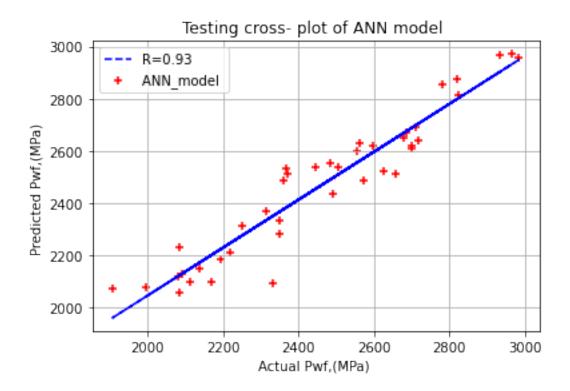


```
[160]: predictions_test=pd.DataFrame(predictions_test)
       predictions_train=pd.DataFrame(predictions_train)
       test=pd.DataFrame(y test)
       train=pd.DataFrame(y_train)
[161]: # Create a scatterplot of the real test values versus the predicted values.
       RFtest=plt.

→scatter(y_test,predictions_test,color='red',marker='+',label='ANN_model ')
       plt.xlabel(('Actual Pwf,(MPa) '))
       plt.ylabel(('Predicted Pwf,(MPa) '))
       R = r2_score(y_test, predictions_test)
       plt.title('Testing cross- plot of ANN model ')
       plt.grid()
       print(R)
       m, b = np.polyfit(y_test, predictions_test, 1)
       plt.plot(y_test, m*y_test+b,linestyle='--',c='blue',label='R=0.93')
       plt.legend()
       #plt.savefig('ANN-test')
```

## 0.9080054721196551

[161]: <matplotlib.legend.Legend at 0x1aa2eff8af0>



```
[162]: from sklearn import metrics

print('MAE:', metrics.mean_absolute_error(y_test, predictions_test))

print('MSE:', metrics.mean_squared_error(y_test, predictions_test))

print('RMSE:', np.sqrt(metrics.mean_squared_error(y_test, predictions_test)))
y_test.mean()
```

MAE: 66.71120545922255 MSE: 7275.450008222243 RMSE: 85.29624850028425

[162]: 2456.9756097560976

```
[167]: print('MAE:', metrics.mean_absolute_error(y_train, predictions_train))
print('MSE:', metrics.mean_squared_error(y_train, predictions_train))
print('RMSE:', np.sqrt(metrics.mean_squared_error(y_train, predictions_train)))
MAPE=np.mean((train - predictions_train) / train ) * 100
MAPE
```

MAE: 49.859710878314395 MSE: 4102.350726502411 RMSE: 64.04959583402858

```
[167]: 0
                   NaN
      Pwf(psia)
                   NaN
       dtype: float64
[44]: from sklearn.tree import DecisionTreeRegressor
       from sklearn.ensemble import RandomForestRegressor
       from sklearn.neighbors import KNeighborsRegressor
       from sklearn.svm import SVC
[45]: from sklearn.model selection import GridSearchCV
       k range = list(range(1,40))
       weight_options = ["uniform", "distance"]
       param_grid = dict(n_neighbors = k_range, weights = weight_options)
       knn = KNeighborsRegressor()
       grid = GridSearchCV(knn, param_grid, cv = 10)
       grid.fit(X,y)
       print (grid.best_params_)
      {'n_neighbors': 6, 'weights': 'distance'}
  []:
[46]: #first, initialize the classificators
       #tree = DecisionTreeRegressor( splitter = 'random', max_leaf_nodes = 10,__
       \rightarrow min_samples_leaf = 5, max_depth= 5)
       tree= DecisionTreeRegressor(random_state=24) # using the random state for_
       \rightarrow reproducibility
       forest= RandomForestRegressor(random state=48)
       knn= KNeighborsRegressor(6, weights = "distance")
       svm= SVC(random_state=24)
[59]: models= [tree, forest, knn]
       from sklearn import metrics
       from sklearn.metrics import accuracy_score
       from sklearn.metrics import r2_score
       for model in models:
           model.fit(X_train, y_train) # fit the model
           y_pred= model.predict(X_test)# then predict on the test set
           y_pred_t=model.predict(X_train)
           MAE=metrics.mean_absolute_error(y_train, y_pred_t)
```

```
MSE= metrics.mean_squared_error(y_test, y_pred)
MAPE=np.mean(np.abs((y_train - y_pred_t) /y_train )) * 100
R = r2_score(y_test, y_pred)
RMSE= np.sqrt(metrics.mean_squared_error(y_test, y_pred))
print(f"The MAE of model {type(model).__name__} is {MAE:.2f}")
print(f"The MSE of model {type(model).__name__} is {RMSE:.2f}")
print(f"The RMSE of model {type(model).__name__} is {RMSE:.2f}")
print(f"The MAPE of model {type(model).__name__} is {MAPE:.2f}")
print(f"The R of model {type(model).__name__} is {R:.2f}")
print(f"The R of model {type(model).__name__} is {R:.2f}")
```

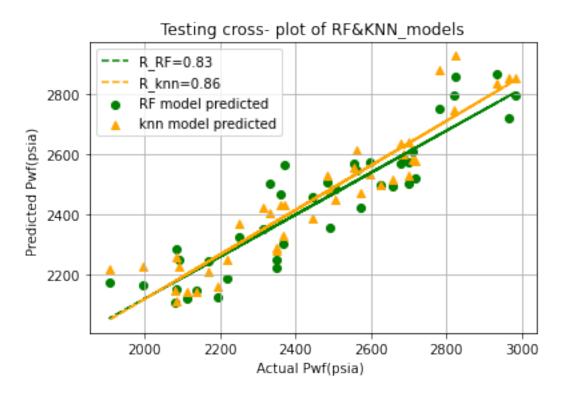
```
The MAE of model DecisionTreeRegressor is 0.00
The MSE of model DecisionTreeRegressor is 20260.34
The RMSE of model DecisionTreeRegressor is 142.34
The MAPE of model DecisionTreeRegressor is 0.00
The R of model DecisionTreeRegressor is 0.74

The MAE of model RandomForestRegressor is 39.43
The MSE of model RandomForestRegressor is 14893.91
The RMSE of model RandomForestRegressor is 122.04
The MAPE of model RandomForestRegressor is 1.69
The R of model RandomForestRegressor is 0.81

The MAE of model KNeighborsRegressor is 11307.00
The RMSE of model KNeighborsRegressor is 106.33
The MAPE of model KNeighborsRegressor is 0.00
The R of model KNeighborsRegressor is 0.00
```

```
R2 = r2_score(y_test,y_pred_F )
R3 = r2_score(y_test,y_pred_knn )
plt.xlabel(('Actual Pwf(psia) '))
plt.ylabel(('Predicted Pwf(psia) '))
print(R)
print(R2)
print(R3)
m, b = np.polyfit(y_test, y_pred_DT, 1)
m2, b2 = np.polyfit(y_test, y_pred_F, 1)
m3, b3 = np.polyfit(y_test, y_pred_knn, 1)
\#plt.plot(y\_test, m*y\_test+b, linestyle='--', c='red', label="R_DT=0.74")
plt.plot(y_test, m2*y_test+b2,linestyle='--',c='green',label="R_RF=0.83")
plt.plot(y_test, m3*y_test+b3,linestyle='--',c='orange',label="R_knn=0.86")
#plt.plot(y test, m*y test+b, linestyle='--',c='red', label="R_DT=0.74")
plt.legend()
plt.title('Testing cross- plot of RF&KNN_models')
plt.grid()
#plt.savefig('Machine-test')
```

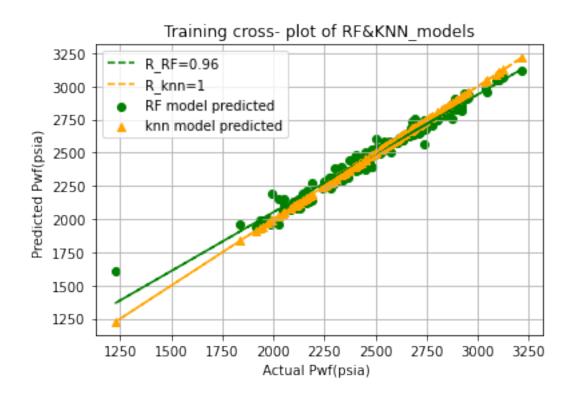
- 0.7438178331903849
- 0.8116737438852756
- 0.8570285039387008



```
[49]: models= [tree, forest, knn]
      from sklearn import metrics
      from sklearn.metrics import r2_score
      tree_m=tree.fit(X_train, y_train)
      forest_m=forest.fit(X_train, y_train)
      knn_m=knn.fit(X_train, y_train)
      y_pred_DT= tree_m.predict(X_train)
      y_pred_F= forest_m.predict(X_train)
      y_pred_knn= knn_m.predict(X_train)
      #plt.scatter(y_test,y_pred_DT,color='red',marker='*',label="DT model predicted")
      plt.scatter(y_train,y_pred_F,color='green',marker='o',label="RF model_\( \)
      →predicted")
      plt.scatter(y_train,y_pred_knn,color='orange',marker='^',label="knn model__
      →predicted")
      R = r2_score(y_train,y_pred_DT )
      R2 = r2_score(y_train,y_pred_F )
      R3 = r2_score(y_train,y_pred_knn )
      plt.xlabel(('Actual Pwf(psia) '))
      plt.ylabel(('Predicted Pwf(psia) '))
      print(R)
      print(R2)
      print(R3)
      m, b = np.polyfit(y_train, y_pred_DT, 1)
      m2, b2 = np.polyfit(y_train, y_pred_F, 1)
      m3, b3 = np.polyfit(y_train, y_pred_knn, 1)
      \#plt.plot(y\_test, m*y\_test+b, linestyle='--', c='red', label="R_DT=0.74")
      plt.plot(y_train, m2*y_train+b2,linestyle='--',c='green',label="R_RF=0.96")
      plt.plot(y_train, m3*y_train+b3,linestyle='--',c='orange',label="R_knn=1")
      plt.legend()
      plt.title('Training cross- plot of RF&KNN models')
      plt.grid()
      #plt.savefig('machine_train')
```

1.0

0.9642425846470991



[]: