

parallel_rss_model_all_data_fine-Copy1

September 16, 2024

```
[1]: import h5py
import numpy as np
import tensorflow as tf
import matplotlib.pyplot as plt
from tensorflow.python.framework.ops import EagerTensor
from tensorflow.python.ops.resource_variable_ops import ResourceVariable
import time
import pandas as pd
import scipy.io
import math
import sklearn
import sklearn.datasets
import tensorflow.keras as keras
from keras.models import Sequential
from keras.layers import Dense
from keras_visualizer import visualizer

dfx = pd.read_csv(r'D:\Ashish Google Drive\Seventh_
↳Sem\PhD\4th_paper_ml\python\Final_dataset_fine\data_output_all_combined_fine.
↳txt')
dfy = pd.read_csv(r'D:\Ashish Google Drive\Seventh_
↳Sem\PhD\4th_paper_ml\python\Final_dataset_fine\data_input_all_combined_fine.
↳txt')

#print(dfx)
#print(dfy)

[2]: from sklearn.model_selection import train_test_split
X_train,X_test = train_test_split(dfx,test_size=0.3,random_state=0)

[3]: from sklearn.model_selection import train_test_split
Y_train,Y_test = train_test_split(dfy,test_size=0.3,random_state=0)

[4]: from tensorflow.keras.layers.experimental import preprocessing
from tensorflow.keras import layers

#train
X_train.head()
```

```

X_train_norm_features = X_train.copy()
X_train_norm_features = np.array(X_train_norm_features)

Y_train_norm_labels = Y_train.copy()
Y_train_norm_labels = np.array(Y_train_norm_labels)

#test
X_test_norm_features = X_test.copy()
X_test_norm_features = np.array(X_test_norm_features)

Y_test_norm_labels = Y_test.copy()
Y_test_norm_labels = np.array(Y_test_norm_labels)

```

```

X_train_norm.copy()          -39.982 -47.84 -33.913 -9.368 -2.968 22.978
10866346    -4.508 -9.588  11.907  0.741 12.987 10.707
8350807     18.353 12.409 -23.351 -27.215 23.637 10.057
11962099    -2.786 16.327 -12.920  6.053 31.695 35.432
7449500      7.200 19.850 -24.389 -4.811 15.891 20.969
6489633    -12.492 -6.499 -34.726 -27.319 -11.180 -12.544
...
10638075   -31.586 -30.872  16.176  8.634  0.142  0.626
5157699     8.381 18.882 -30.882 -8.957  1.107 17.660
2215104    -14.904 -28.782  -2.804 -4.793 15.434 24.051
1484405     7.594 -9.104  -5.600 -4.577 15.655 20.647
8325804     10.388 13.524 -35.457 -30.695 10.906  6.617

```

[8583239 rows x 6 columns]

```

X_train_norm_features [[ -4.508 -9.588 11.907  0.741 12.987 10.707]
 [ 18.353 12.409 -23.351 -27.215 23.637 10.057]
 [ -2.786 16.327 -12.92  6.053 31.695 35.432]
 ...
 [-14.904 -28.782 -2.804 -4.793 15.434 24.051]
 [  7.594 -9.104 -5.6   -4.577 15.655 20.647]
 [ 10.388 13.524 -35.457 -30.695 10.906  6.617]]
Y_train_norm_labels [[ 4 -14  7 -4 -2  4]
 [16 18  6  6 -10  6]
 [19 10 19 -8 -10 -10]
 ...
 [-10 -20  0 -8 -10  0]
 [-14 -6  8 -2 -10  2]
 [16 16 -2  6 -10  0]]

```

```

Y_test_norm.copy()          -20 -20.1 -20.2 -10 -10.1 -10.2
9460651    -17    -20    -11  2     0    -6
10521359    -2     -2    -20  -4     2   -10
3652691     -4    -10    18  2     8    -6
8083701     16     -8    18  4    -10   -6
7591956     14    -12     0  8     0    4

```

```

...      ...      ...      ...      ...      ...
5988974      6      4      -14      8      4      0
6289334      8      -10      0      -4      -4      0
5079849      2      0      16      6      0      -10
9916060     -11      -8      13     -10      2      -8
8897826      20     -14      8      6      -6      4

```

[3678532 rows x 6 columns]

```

X_test_norm_features      -39.982  -47.84  -33.913  -9.368  -2.968  22.978
9460651    -10.292 -10.560   -5.794  -1.618 -23.774   3.227
10521359   -37.162 -17.353  -18.699   0.022 -22.304  -5.684
3652691      5.551  16.327   22.268  23.904  -7.799   5.850
8083701     13.681  24.510  -11.623 -10.161  17.583  31.059
7591956      8.830  10.023   0.288 -20.925  -2.338  -2.972

...      ...      ...      ...      ...      ...
5988974      0.626   6.008  -12.031 -18.507 -27.711 -29.157
6289334    -12.311 -10.214   -2.112  -6.380  10.024  11.437
5079849     12.372  27.465   0.925  11.914  -4.663  12.482
9916060    -16.776  -9.643   12.225  27.693   9.456  21.676
8897826     10.246  11.354   -1.845 -22.265  15.171  15.288

```

[3678532 rows x 6 columns]

```
[5]: import keras_tuner
```

Using TensorFlow backend

```

[20]: def build_regressor(hp):
    model = keras.Sequential(
        [
            layers.Dense(units=hp.Int("units", 6, 256, 32), activation="tanh"),
            layers.Dense(units=6),
        ]
    )
    model.compile(
        optimizer="adam",
        loss="mean_squared_error",
        # Objective is one of the metrics.
        metrics=[keras.metrics.MeanAbsoluteError()],
    )
    return model

tuner = keras_tuner.RandomSearch(
    hypermodel=build_regressor,
    objective=keras_tuner.Objective("val_mean_absolute_error", direction="min"),
    max_trials=6,
    overwrite=True,

```

```

    directory="./",
    project_name="built_in_metrics_all_fine_2",
)

tuner.search(X_train_norm_features, Y_train_norm_labels, validation_split = 0.
    ↪15, epochs=10)

tuner.results_summary()

```

Trial 6 Complete [00h 44m 52s]

val_mean_absolute_error: 0.07545071840286255

Best val_mean_absolute_error So Far: 0.057804595679044724

Total elapsed time: 04h 21m 38s

Results summary

Results in ./built_in_metrics_all_fine_1

Showing 10 best trials

Objective(name="val_mean_absolute_error", direction="min")

Trial 4 summary

Hyperparameters:

units: 230

Score: 0.057804595679044724

Trial 1 summary

Hyperparameters:

units: 166

Score: 0.06618934124708176

Trial 3 summary

Hyperparameters:

units: 134

Score: 0.07078669220209122

Trial 5 summary

Hyperparameters:

units: 102

Score: 0.07545071840286255

Trial 0 summary

Hyperparameters:

units: 70

Score: 0.07765627652406693

Trial 2 summary

Hyperparameters:

units: 6

Score: 0.551080584526062

```
[21]: tuner.search_space_summary()
```

```
Search space summary
Default search space size: 1
units (Int)
{'default': None, 'conditions': [], 'min_value': 6, 'max_value': 256, 'step':
32, 'sampling': 'linear'}
```

```
[22]: # Get the top 2 models.
models = tuner.get_best_models(num_models=2)
best_model = models[0]
# Build the model.
# Needed for `Sequential` without specified `input_shape`.
best_model.build(input_shape=(None, 6))
best_model.summary()
```

```
Model: "sequential"
```

Layer (type)	Output Shape	Param #
dense (Dense)	(None, 230)	1610
dense_1 (Dense)	(None, 6)	1386

=====
Total params: 2996 (11.70 KB)
Trainable params: 2996 (11.70 KB)
Non-trainable params: 0 (0.00 Byte)
=====

```
[23]: tuner.results_summary()
```

```
Results summary
Results in ./built_in_metrics_all_fine_1
Showing 10 best trials
Objective(name="val_mean_absolute_error", direction="min")
```

```
Trial 4 summary
Hyperparameters:
units: 230
Score: 0.057804595679044724
```

```
Trial 1 summary
Hyperparameters:
units: 166
Score: 0.06618934124708176
```

```
Trial 3 summary
Hyperparameters:
```

units: 134
Score: 0.07078669220209122

Trial 5 summary
Hyperparameters:
units: 102
Score: 0.07545071840286255

Trial 0 summary
Hyperparameters:
units: 70
Score: 0.07765627652406693

Trial 2 summary
Hyperparameters:
units: 6
Score: 0.551080584526062

```
[24]: best_model.fit(X_train_norm_features, Y_train_norm_labels, validation_split = 0.  
    ↪15, epochs=10)
```

Epoch 1/10
227993/227993 [=====] - 251s 1ms/step - loss: 0.0089 -
mean_absolute_error: 0.0679 - val_loss: 0.0130 - val_mean_absolute_error: 0.0800
Epoch 2/10
227993/227993 [=====] - 243s 1ms/step - loss: 0.0089 -
mean_absolute_error: 0.0677 - val_loss: 0.0122 - val_mean_absolute_error: 0.0760
Epoch 3/10
227993/227993 [=====] - 247s 1ms/step - loss: 0.0088 -
mean_absolute_error: 0.0675 - val_loss: 0.0116 - val_mean_absolute_error: 0.0788
Epoch 4/10
227993/227993 [=====] - 245s 1ms/step - loss: 0.0088 -
mean_absolute_error: 0.0675 - val_loss: 0.0070 - val_mean_absolute_error: 0.0604
Epoch 5/10
227993/227993 [=====] - 246s 1ms/step - loss: 0.0088 -
mean_absolute_error: 0.0674 - val_loss: 0.0096 - val_mean_absolute_error: 0.0702
Epoch 6/10
227993/227993 [=====] - 249s 1ms/step - loss: 0.0088 -
mean_absolute_error: 0.0676 - val_loss: 0.0092 - val_mean_absolute_error: 0.0688
Epoch 7/10
227993/227993 [=====] - 247s 1ms/step - loss: 0.0088 -
mean_absolute_error: 0.0676 - val_loss: 0.0108 - val_mean_absolute_error: 0.0757
Epoch 8/10
227993/227993 [=====] - 249s 1ms/step - loss: 0.0088 -
mean_absolute_error: 0.0676 - val_loss: 0.0088 - val_mean_absolute_error: 0.0658
Epoch 9/10
227993/227993 [=====] - 249s 1ms/step - loss: 0.0087 -
mean_absolute_error: 0.0674 - val_loss: 0.0100 - val_mean_absolute_error: 0.0741

```
227993/227993 [=====] - 249s 1ms/step - loss: 0.0087 -  
mean_absolute_error: 0.0673 - val_loss: 0.0090 - val_mean_absolute_error: 0.0692
```

```
[13]: #plt.plot(best_model['loss'], label='train')
      #plt.plot(best_model['val_loss'], label='test')
      #plt.legend()
      #plt.title('Training History'),
      #plt.xlabel('Epoch'),
      #plt.ylabel('Validation Loss')
      #plt.show()
```

Saved model to disk

```
C:\Users\Hp\AppData\Local\Programs\Python\Python311\Lib\site-
packages\keras\src\engine\training.py:3079: UserWarning: You are saving your
model as an HDF5 file via `model.save()`. This file format is considered legacy.
We recommend using instead the native Keras format, e.g.
`model.save('my_model.keras')`.
    saving_api.save_model(
```

7

```
#plt(history)
```

```
268227/268227 [=====] - 199s 740us/step
[[ 3.98273325e+00 -1.41698942e+01  7.02920389e+00 -3.96829557e+00
  -2.04198885e+00  3.97841716e+00]
 [ 1.58094006e+01  1.75955372e+01  6.18313789e+00  5.94216681e+00
  -9.92908955e+00  5.88854694e+00]
 [ 1.90149002e+01  9.83222294e+00  1.89996948e+01 -8.09838676e+00
  -9.99605465e+00 -9.98972225e+00]
 ...
 [-9.97579575e+00 -1.99920197e+01  1.61714107e-02 -7.87847996e+00
  -1.00445576e+01  8.23003054e-02]
 [-1.39123783e+01 -6.04278421e+00  8.06994724e+00 -1.94951534e+00
  -1.00191841e+01  1.92616343e+00]
 [ 1.58520432e+01  1.57699795e+01 -1.72332728e+00  5.98136711e+00
  -9.94271183e+00  4.75108624e-04]]
(8583239, 6)
[[ 4 -14  7 -4 -2  4]
 [ 16 18  6  6 -10  6]
 [ 19 10 19 -8 -10 -10]
 ...
 [-10 -20  0 -8 -10  0]
 [-14 -6  8 -2 -10  2]
 [ 16 16 -2  6 -10  0]]
<class 'numpy.ndarray'>
(8583239, 6)
(8583239, 6)
[[-1.72667503e-02 -1.69894218e-01  2.92038918e-02  3.17044258e-02
  -4.19888496e-02 -2.15828419e-02]
 [-1.90599442e-01 -4.04462814e-01  1.83137894e-01 -5.78331947e-02
  7.09104538e-02 -1.11453056e-01]
 [ 1.49002075e-02 -1.67777061e-01 -3.05175781e-04 -9.83867645e-02
  3.94535065e-03  1.02777481e-02]
 ...
 [ 2.42042542e-02  7.98034668e-03  1.61714107e-02  1.21520042e-01
  -4.45575714e-02  8.23003054e-02]
 [ 8.76216888e-02 -4.27842140e-02  6.99472427e-02  5.04846573e-02
  -1.91841125e-02 -7.38365650e-02]
 [-1.47956848e-01 -2.30020523e-01  2.76672721e-01 -1.86328888e-02
  5.72881699e-02  4.75108624e-04]]
0.009009736494002307
```

```
[ ]: predict12a = best_model.predict(X_test_norm_features)
      #predict_train_X = norm_model8.predict(X_train_norm_features)
      #print("predict_train_X",predict_train_X)
      #denorm_predict_train_X = np.multiply(y_norm_train,predict_train_X)
      #print("denorm_predict_train_X",denorm_predict_train_X)
```



```

print(predict12a)

print(np.shape(predict12a))
print(Y_test_norm_labels)
print(type(Y_test_norm_labels))
print(np.shape(Y_test_norm_labels))
#Y_train_norm_labels = Y_train_norm_labels.reshape(137,1)
error12a = predict12a-Y_test_norm_labels
print(np.shape(Y_test_norm_labels))
print(error12a)
print(np.shape(error12a))
print(np.mean(np.square(np.abs(error12a))))
#plt(history)

```

```

[ ]: model_json = best_model.to_json()
with open("all_model.json", "w") as json_file:
    json_file.write(model_json)
# serialize weights to HDF5
best_model.save_weights("all_model.h5")
print("Saved model to disk")

```

```

[ ]: json_file = open('all_model.json', 'r')
loaded_model_json = json_file.read()
json_file.close()

```

```

[ ]: best_model.save('all_model.keras')

```

```

[13]: new_model = tf.keras.models.load_model('all_model_1.keras')

```

```

[14]: new_model.summary()

```

Model: "sequential"

Layer (type)	Output Shape	Param #
dense (Dense)	(None, 230)	1610
dense_1 (Dense)	(None, 6)	1386

=====
 Total params: 2996 (11.70 KB)
 Trainable params: 2996 (11.70 KB)
 Non-trainable params: 0 (0.00 Byte)
 =====

```

[ ]: best_model.save('all_model.h5')

```

```

[4]: new_model1 = tf.keras.models.load_model('all_model_1.keras', compile=False)

```

```
[5]: new_model1.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
dense (Dense)	(None, 230)	1610
dense_1 (Dense)	(None, 6)	1386

=====
 Total params: 2996 (11.70 KB)
 Trainable params: 2996 (11.70 KB)
 Non-trainable params: 0 (0.00 Byte)
 =====

```
[31]: dfx_check = pd.read_csv(r'D:\Ashish Google Drive\Seventh_
    ↪Sem\PhD\4th_paper_ml\python\DataSets\data_output_3_2.txt')
dfy_check = pd.read_csv(r'D:\Ashish Google Drive\Seventh_
    ↪Sem\PhD\4th_paper_ml\python\DataSets\data_input_3_2.txt')
```

```
[40]: from sklearn.model_selection import train_test_split
X_train,X_test = train_test_split(dfx_check,test_size=0.3,random_state=0)

from sklearn.model_selection import train_test_split
Y_train,Y_test = train_test_split(dfy_check,test_size=0.3,random_state=0)
```

```
[41]: from tensorflow.keras.layers.experimental import preprocessing
from tensorflow.keras import layers
X_train.head()
X_train_norm_features = X_train.copy()
print("X_train_norm.copy()",X_train.copy())
Y_train_norm_labels = Y_train.copy()
Y_train_norm_labels = np.array(Y_train_norm_labels)
X_train_norm_features = np.array(X_train_norm_features)
print("X_train_norm_features",X_train_norm_features)
print("Y_train_norm_labels",Y_train_norm_labels)
X_test_norm_features = X_test.copy()
Y_test_norm_labels = Y_test.copy()
print("Y_test_norm.copy()",Y_test.copy())
#Y_test_norm_labels = Y_test_norm_labels.to_numpy()
X_test_norm_features = np.array(X_test_norm_features)
Y_test_norm_labels = np.array(Y_test_norm_labels)
#X_test_norm_features = X_test.copy()
print("X_test_norm_features",X_test)
```

```
X_train_norm.copy()      0  0.1  0.2  0.3  0.4  0.5
6   7   0   0   0   0   0
```

```

13 14 0 0 0 0 0
4 5 0 0 0 0 0
2 3 0 0 0 0 0
5 6 0 0 0 0 0
14 15 0 0 0 0 0
9 10 0 0 0 0 0
7 8 0 0 0 0 0
16 17 0 0 0 0 0
11 12 0 0 0 0 0
3 4 0 0 0 0 0
0 1 0 0 0 0 0
15 16 0 0 0 0 0
12 13 0 0 0 0 0
X_train_norm_features [[ 7 0 0 0 0 0]
[14 0 0 0 0 0]
[ 5 0 0 0 0 0]
[ 3 0 0 0 0 0]
[ 6 0 0 0 0 0]
[15 0 0 0 0 0]
[10 0 0 0 0 0]
[ 8 0 0 0 0 0]
[17 0 0 0 0 0]
[12 0 0 0 0 0]
[ 4 0 0 0 0 0]
[ 1 0 0 0 0 0]
[16 0 0 0 0 0]
[13 0 0 0 0 0]]
Y_train_norm_labels [[ 7 7 19 2 4 4]
[ -5 1 -5 -4 8 6]
[ 1 1 10 8 -6 6]
...
[ 13 -8 -11 -8 -10 0]
[ 1 4 -20 -2 -10 2]
[-17 1 13 4 -8 -6]]
Y_test_norm.copy() -20 -20.1 -20.2 -10 -10.1 -10.2
2558706 19 -20 10 4 -10 4
2208349 13 -11 10 -4 0 -10
1426390 1 -11 16 -4 8 -8
1621354 4 -11 13 -4 0 0
1406908 1 -14 -2 8 -10 8
... ...
2015101 10 -11 19 -8 -10 -6
953357 -8 16 -17 -4 0 6
1519020 1 10 1 -10 -6 -8
633929 -11 -11 -11 8 -4 -10
1313148 -2 7 13 -8 -2 8

[823200 rows x 6 columns]

```

X_test_norm_features		0	0.1	0.2	0.3	0.4	0.5
18	19	0	0	0	0	0	
1	2	0	0	0	0	0	
19	20	0	0	0	0	0	
8	9	0	0	0	0	0	
10	11	0	0	0	0	0	
17	18	0	0	0	0	0	

```
[34]: predict12 = new_model1.predict(X_train_norm_features)
      #predict_train_X = norm_model8.predict(X_train_norm_features)
      #print("predict_train_X",predict_train_X)
      #denorm_predict_train_X = np.multiply(y_norm_train,predict_train_X)
      #print("denorm_predict_train_X",denorm_predict_train_X)
      print(predict12)
      print(np.shape(predict12))
      print(Y_train_norm_labels)
      print(type(Y_train_norm_labels))
      print(np.shape(Y_train_norm_labels))
      #Y_train_norm_labels = Y_train_norm_labels.reshape(137,1)
      error12 = predict12-Y_train_norm_labels
      print(np.shape(Y_train_norm_labels))
      print(error12)
      #print(np.shape(error8))
      print(np.mean(np.square(np.abs(error12))))
      #plt(history)
```

```
60025/60025 [=====] - 49s 821us/step
[[ 7.045394   6.967816  19.077456   1.9879396   4.0535655
   3.9502766 ]
 [-5.2469463   1.0126202  -5.0328045  -3.9583404   7.935346
   6.0668917 ]
 [ 0.87774295   0.9591308  10.026771   7.989295  -6.043178
   5.896037 ]
 ...
 [ 12.998318  -7.9460073 -10.965882  -7.8988934 -10.039447
   0.06163019]
 [ 0.8453657   3.9354253 -20.042774  -1.8691561 -10.124183
   1.9683435 ]
 [-16.859388   0.88292783  12.974099   4.003473  -8.045867
  -6.059412  ]]
(1920799, 6)
[[ 7   7  19   2   4   4]
 [-5   1  -5  -4   8   6]
 [ 1   1  10   8  -6   6]
 ...
 [ 13  -8 -11  -8 -10   0]
 [ 1   4 -20  -2 -10   2]
 [-17  1  13   4  -8  -6]]
```

```

<class 'numpy.ndarray'>
(1920799, 6)
(1920799, 6)
[[ 0.04539394 -0.03218412  0.07745552 -0.0120604   0.0535655  -0.04972339]
 [-0.24694633  0.01262021 -0.03280449  0.04165959 -0.06465387  0.06689167]
 [-0.12225705 -0.04086918  0.02677059 -0.01070499 -0.04317808 -0.1039629 ]
 ...
 [-0.00168228  0.05399275  0.0341177   0.10110664 -0.03944683  0.06163019]
 [-0.1546343  -0.06457472 -0.0427742   0.13084388 -0.1241827  -0.0316565 ]
 [ 0.14061165 -0.11707217 -0.02590084  0.00347281 -0.04586697 -0.059412  ]]
0.008920830023362894

```

```

[35]: predict12a = new_model1.predict(X_test_norm_features)
      #predict_train_X = norm_model8.predict(X_train_norm_features)
      #print("predict_train_X",predict_train_X)
      #denorm_predict_train_X = np.multiply(y_norm_train,predict_train_X)
      #print("denorm_predict_train_X", denorm_predict_train_X)
      print(predict12a)

      print(np.shape(predict12a))
      print(Y_test_norm_labels)
      print(type(Y_test_norm_labels))
      print(np.shape(Y_test_norm_labels))
      #Y_train_norm_labels = Y_train_norm_labels.reshape(137,1)
      error12a = predict12a-Y_test_norm_labels
      print(np.shape(Y_test_norm_labels))
      print(error12a)
      print(np.shape(error12a))
      print(np.mean(np.square(np.abs(error12a))))
      #plt(history)

```

```

25725/25725 [=====] - 19s 752us/step
[[ 18.819689  -19.907509   10.01861    4.0113683  -10.062325
    3.937744 ]
 [ 13.080879  -11.196361   10.00459   -4.0012503  -0.09037608
   -9.99972 ]
 [  1.1062601 -11.170524   16.086462  -4.0443482    7.942209
   -7.8918486 ]
 ...
 [  1.0602305   9.702753    1.1198127  -9.98101   -6.000906
   -8.060306 ]
 [-11.080045  -10.91409   -11.012386    8.105999   -3.8805215
  -10.114193 ]
 [ -2.1043534   6.87991    13.081214  -8.032471   -2.1050594
    7.9799547 ]]
(823200, 6)
[[ 19 -20  10   4 -10   4]
 [ 13 -11  10  -4   0 -10]

```

```

[ 1 -11 16 -4 8 -8]
...
[ 1 10 1 -10 -6 -8]
[-11 -11 -11 8 -4 -10]
[ -2 7 13 -8 -2 8]]
<class 'numpy.ndarray'>
(823200, 6)
(823200, 6)
[[-1.80311203e-01 9.24911499e-02 1.86100006e-02 1.13682747e-02
 -6.23245239e-02 -6.22560978e-02]
 [ 8.08792114e-02 -1.96360588e-01 4.59003448e-03 -1.25026703e-03
 -9.03760791e-02 2.80380249e-04]
 [ 1.06260061e-01 -1.70523643e-01 8.64620209e-02 -4.43482399e-02
 -5.77912331e-02 1.08151436e-01]
 ...
 [ 6.02304935e-02 -2.97246933e-01 1.19812727e-01 1.89895630e-02
 -9.05990601e-04 -6.03055954e-02]
 [-8.00447464e-02 8.59098434e-02 -1.23863220e-02 1.05998993e-01
 1.19478464e-01 -1.14192963e-01]
 [-1.04353428e-01 -1.20090008e-01 8.12139511e-02 -3.24707031e-02
 -1.05059385e-01 -2.00452805e-02]]
(823200, 6)
0.008934312423057348

```

```

[23]: plt.plot(new_model1.history['loss'])
plt.plot(new_model1.history['val_loss'])
plt.title('model loss')
plt.ylabel('loss')
plt.xlabel('epoch')
plt.legend(['train', 'val'], loc='upper left')
plt.show()

```

```

-----
TypeError                                Traceback (most recent call last)
Cell In[23], line 1
----> 1 plt.plot(new_model1.history['loss'])
      2 plt.plot(new_model1.history['val_loss'])
      3 plt.title('model loss')

TypeError: 'History' object is not subscriptable

```

```

[6]: j_val= pd.read_csv(r'D:\Ashish Google Drive\Seventh_
↳Sem\PhD\4th_paper_ml\python\validate\joint1.txt')

```

```

[8]: from tensorflow.keras.layers.experimental import preprocessing
from tensorflow.keras import layers

```

```

j_val.head()
j_val_norm_features = j_val.copy()
#print("j_val_norm.copy()", j_val.copy())
j_val_norm_features = np.array(j_val_norm_features)
print("j_val_norm_features", j_val_norm_features)
print("j_val_shape", np.shape(j_val))

```

```

j_val_norm_features [[ 1  0  0  0  0  0]
 [ 2  0  0  0  0  0]
 [ 3  0  0  0  0  0]
 [ 4  0  0  0  0  0]
 [ 5  0  0  0  0  0]
 [ 6  0  0  0  0  0]
 [ 7  0  0  0  0  0]
 [ 8  0  0  0  0  0]
 [ 9  0  0  0  0  0]
 [10  0  0  0  0  0]
 [11  0  0  0  0  0]
 [12  0  0  0  0  0]
 [13  0  0  0  0  0]
 [14  0  0  0  0  0]
 [15  0  0  0  0  0]
 [16  0  0  0  0  0]
 [17  0  0  0  0  0]
 [18  0  0  0  0  0]
 [19  0  0  0  0  0]
 [20  0  0  0  0  0]]
j_val_shape (20, 6)

```

```
[9]: predict12a = new_model1.predict(j_val_norm_features)
```

```

print(predict12a)

print(np.shape(predict12a))

#plt(history)

```

```

1/1 [=====] - 0s 170ms/step
[[ -1.1244261    0.3311052    2.3444054    0.13440591  -0.15608305
    0.22917116]
 [ -2.1868036    0.5127536    2.5655348    0.29968756  -0.30188364
    0.4412458 ]
 [ -3.2855194    0.6929818    2.7781472    0.46150094  -0.45086998
    0.65019524]
 [ -4.364477     0.87604487    2.9793143    0.61728126  -0.5937211
    0.86381125]
 [ -5.4284225    1.05996       3.1717136    0.77157027  -0.7301032
    1.0855012 ]

```

```

[ -6.522745    1.2406223    3.3603013    0.93041736   -0.86720294
  1.3115559 ]
[ -7.630296    1.4210577    3.5460904    1.0925899    -1.0056775
  1.5364754 ]
[ -8.713226    1.6070437    3.7269533    1.2536659    -1.1422923
  1.7578521 ]
[ -9.816599    1.797044    3.9023683    1.4152267    -1.2814488
  1.9771683 ]
[ -10.975772   1.9885356    4.0732584    1.5798028    -1.424406
  2.2005541 ]
[ -12.114705   2.1864862    4.239405    1.7431915    -1.5591667
  2.4352508 ]
[ -13.194851   2.3922246    4.400227    1.9034963    -1.6822653
  2.678908 ]
[ -14.302092   2.597243    4.5557265    2.06641     -1.8087406
  2.9195635 ]
[ -15.467853   2.7984676    4.7052474    2.23222     -1.9442332
  3.1522837 ]
[ -16.607662   3.0032256    4.8464      2.392235    -2.0747974
  3.386411 ]
[ -17.724266   3.2114024    4.9749618    2.545282    -2.1966677
  3.6309931 ]
[ -18.900156   3.4145553    5.0896344    2.6989145    -2.3201647
  3.8837597 ]
[ -20.0862     3.6130273    5.196701     2.8535326    -2.4427636
  4.1367493 ]
[ -21.23248    3.8084521    5.298637     3.0073874    -2.561035
  4.383404 ]
[ -22.44708    3.9939146    5.3856797    3.1640573    -2.6859055
  4.6256094 ]]
(20, 6)

```

```

[10]: df = pd.DataFrame(predict12a)

print(df)

```

	0	1	2	3	4	5
0	-1.124426	0.331105	2.344405	0.134406	-0.156083	0.229171
1	-2.186804	0.512754	2.565535	0.299688	-0.301884	0.441246
2	-3.285519	0.692982	2.778147	0.461501	-0.450870	0.650195
3	-4.364477	0.876045	2.979314	0.617281	-0.593721	0.863811
4	-5.428422	1.059960	3.171714	0.771570	-0.730103	1.085501
5	-6.522745	1.240622	3.360301	0.930417	-0.867203	1.311556
6	-7.630296	1.421058	3.546090	1.092590	-1.005677	1.536475
7	-8.713226	1.607044	3.726953	1.253666	-1.142292	1.757852
8	-9.816599	1.797044	3.902368	1.415227	-1.281449	1.977168
9	-10.975772	1.988536	4.073258	1.579803	-1.424406	2.200554
10	-12.114705	2.186486	4.239405	1.743191	-1.559167	2.435251


```

11 -13.194851  2.392225  4.400227  1.903496 -1.682265  2.678908
12 -14.302092  2.597243  4.555727  2.066410 -1.808741  2.919564
13 -15.467853  2.798468  4.705247  2.232220 -1.944233  3.152284
14 -16.607662  3.003226  4.846400  2.392235 -2.074797  3.386411
15 -17.724266  3.211402  4.974962  2.545282 -2.196668  3.630993
16 -18.900156  3.414555  5.089634  2.698915 -2.320165  3.883760
17 -20.086201  3.613027  5.196701  2.853533 -2.442764  4.136749
18 -21.232479  3.808452  5.298637  3.007387 -2.561035  4.383404
19 -22.447081  3.993915  5.385680  3.164057 -2.685905  4.625609

```

```
[11]: df.to_csv('val.txt', index=False)
```

```
[12]: x =df.to_numpy()
      print(x)
```

```

[[ -1.1244261    0.3311052    2.3444054    0.13440591  -0.15608305
   0.22917116]
 [ -2.1868036    0.5127536    2.5655348    0.29968756  -0.30188364
   0.4412458 ]
 [ -3.2855194    0.6929818    2.7781472    0.46150094  -0.45086998
   0.65019524]
 [ -4.364477     0.87604487    2.9793143    0.61728126  -0.5937211
   0.86381125]
 [ -5.4284225    1.05996     3.1717136    0.77157027  -0.7301032
   1.0855012 ]
 [ -6.522745     1.2406223    3.3603013    0.93041736  -0.86720294
   1.3115559 ]
 [ -7.630296     1.4210577    3.5460904    1.0925899   -1.0056775
   1.5364754 ]
 [ -8.713226     1.6070437    3.7269533    1.2536659   -1.1422923
   1.7578521 ]
 [ -9.816599     1.797044     3.9023683    1.4152267   -1.2814488
   1.9771683 ]
 [-10.975772     1.9885356    4.0732584    1.5798028   -1.424406
   2.2005541 ]
 [-12.114705     2.1864862    4.239405     1.7431915   -1.5591667
   2.4352508 ]
 [-13.194851     2.3922246    4.400227     1.9034963   -1.6822653
   2.678908 ]
 [-14.302092     2.597243     4.5557265    2.06641     -1.8087406
   2.9195635 ]
 [-15.467853     2.7984676    4.7052474    2.23222     -1.9442332
   3.1522837 ]
 [-16.607662     3.0032256    4.8464       2.392235    -2.0747974
   3.386411 ]
 [-17.724266     3.2114024    4.9749618    2.545282    -2.1966677
   3.6309931 ]
 [-18.900156     3.4145553    5.0896344    2.6989145    -2.3201647

```

```

    3.8837597 ]
[-20.0862      3.6130273    5.196701    2.8535326   -2.4427636
    4.1367493 ]
[-21.23248    3.8084521    5.298637    3.0073874   -2.561035
    4.383404 ]
[-22.44708    3.9939146    5.3856797    3.1640573   -2.6859055
    4.6256094 ]]

```

```
[13]: np.save("data_test", x)
```

```
[29]: with open("val.txt", 'r') as data_file:
        for line in data_file:
            data = line.split()
            res = [s.strip() for s in data[0].split(',')]
            print(res[0], res[1], res[2], res[3], res[4], res[5])
```

```

0 1 2 3 4 5
-1.1244261 0.3311052 2.3444054 0.13440591 -0.15608305 0.22917116
-2.1868036 0.5127536 2.5655348 0.29968756 -0.30188364 0.4412458
-3.2855194 0.6929818 2.7781472 0.46150094 -0.45086998 0.65019524
-4.364477 0.87604487 2.9793143 0.61728126 -0.5937211 0.86381125
-5.4284225 1.05996 3.1717136 0.77157027 -0.7301032 1.0855012
-6.522745 1.2406223 3.3603013 0.93041736 -0.86720294 1.3115559
-7.630296 1.4210577 3.5460904 1.0925899 -1.0056775 1.5364754
-8.713226 1.6070437 3.7269533 1.2536659 -1.1422923 1.7578521
-9.816599 1.797044 3.9023683 1.4152267 -1.2814488 1.9771683
-10.975772 1.9885356 4.0732584 1.5798028 -1.424406 2.2005541
-12.114705 2.1864862 4.239405 1.7431915 -1.5591667 2.4352508
-13.194851 2.3922246 4.400227 1.9034963 -1.6822653 2.678908
-14.302092 2.597243 4.5557265 2.06641 -1.8087406 2.9195635
-15.467853 2.7984676 4.7052474 2.23222 -1.9442332 3.1522837
-16.607662 3.0032256 4.8464 2.392235 -2.0747974 3.386411
-17.724266 3.2114024 4.9749618 2.545282 -2.1966677 3.6309931
-18.900156 3.4145553 5.0896344 2.6989145 -2.3201647 3.8837597
-20.0862 3.6130273 5.196701 2.8535326 -2.4427636 4.1367493
-21.23248 3.8084521 5.298637 3.0073874 -2.561035 4.383404
-22.44708 3.9939146 5.3856797 3.1640573 -2.6859055 4.6256094

```

```
[37]: from ikin import *
rbt = Robot()
with open("val.txt", 'r') as data_file:
    for line in data_file:
        data = line.split()
        #print(data[0])
        res = [s.strip() for s in data[0].split(',')]
        #print(res[0], res[1], res[2], res[3], res[4], res[5])
        #print(float(res[0]),
        ↪ float(res[1]), float(res[2]), float(res[3]), float(res[4]), float(res[5]))
```

```
print(rbt.goto(float(res[0]),  
↪float(res[1]),float(res[2]),float(res[3]),float(res[4]),float(res[5])))
```

```
[8.826, 3.44, 7.734, -2.517, -4.968, -13.319]  
[1.074, 0.01, 0.019, -0.015, -0.007, 0.023]  
[2.072, 0.017, 0.009, -0.019, -0.012, 0.024]  
[3.067, 0.011, -0.009, -0.013, -0.023, 0.035]  
[4.041, -0.002, -0.019, -0.008, -0.028, 0.037]  
[5.009, -0.016, -0.017, -0.008, -0.03, 0.024]  
[5.996, -0.029, -0.015, -0.009, -0.04, 0.012]  
[6.991, -0.032, -0.018, -0.01, -0.052, 0.006]  
[7.974, -0.024, -0.026, -0.015, -0.058, 0.004]  
[8.962, -0.014, -0.044, -0.016, -0.067, 0.01]  
[9.976, -0.009, -0.064, -0.009, -0.086, 0.02]  
[10.991, -0.001, -0.068, -0.008, -0.095, 0.014]  
[11.987, 0.017, -0.056, -0.018, -0.087, -0.013]  
[12.994, 0.04, -0.053, -0.026, -0.087, -0.03]  
[14.018, 0.061, -0.069, -0.023, -0.1, -0.023]  
[15.023, 0.081, -0.081, -0.021, -0.104, -0.02]  
[16.013, 0.089, -0.077, -0.024, -0.1, -0.037]  
[17.026, 0.079, -0.07, -0.022, -0.108, -0.056]  
[18.043, 0.072, -0.064, -0.021, -0.122, -0.073]  
[19.038, 0.079, -0.061, -0.025, -0.132, -0.087]  
[20.053, 0.077, -0.07, -0.018, -0.165, -0.087]
```

[]: