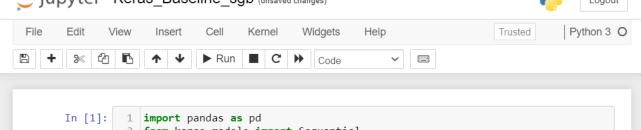


Logout



```
2 from keras.models import Sequential
          3 from keras.layers import Dense, LayerNormalization , Dropout, LSTM, Embeddin
          4 from keras.layers.advanced activations import PReLU, ReLU
          5 | from sklearn.model_selection import train_test_split
          6 from tensorflow import keras
           7 import matplotlib.pyplot as plt
          8 #from sklearn.model selection import train test split
          9 from sklearn.metrics import mean_squared_error
          10 import tensorflow as tf
          11 import numpy as np
          1 df = pd.read_csv("row_size25_vector_size8000.csv")
In [2]:
In [3]:
          1 X = df.iloc[:,7:157]
           2 #X = df[['accelerometer_reading_x_0','accelerometer_reading_y_0','accelerome
          3 #t = df[['linear_position_x','linear_position_y','linear_position_z','angula
4 #t = df[['phi_change','theta_change','psi_change']]
           5 t = df[['phi_change']]
In [4]:
          1 t.shape
Out[4]: (8000, 1)
In [5]:
          1 in_dim = X.shape[1]
           2 out_dim = t.shape[1]
In [6]:
          1 in dim
Out[6]: 150
          1 out_dim
In [7]:
Out[7]: 1
In [8]:
          1 | X_train, X_test, t_train, t_test = train_test_split(X, t, test_size=0.2)
In [9]:
          1 t_train.shape
Out[9]: (6400, 1)
In [10]:
             model = Sequential()
             model.add(Dense(256,input_dim=in_dim, activation="elu"))
          3 model.add(ReLU())
          4 model.add(Dropout(.32))
          5 model.add(Dense(128, activation="elu"))
          6 model.add(ReLU())
             model.add(LayerNormalization ())
          8 model.add(Dropout(.25))
          9 model.add(Dense(64, activation="elu"))
          10 model.add(ReLU())
          11 model.add(LayerNormalization ())
          12 model.add(Dropout(.1))
          13 model.add(Dense(32, activation="elu"))
          14 model.add(ReLU())
          15 model.add(LayerNormalization ())
             model.add(Dense(out dim,activation="elu"))
          17 model.compile(loss="mse", optimizer="sgd")
        4 11 (1/21 1 1 1 1 1 25 1 1 1 1 40 1
```

```
In [III]: | 1 | mode1.flt(X_train, t_train, epocns=/5, batcn_size=12, verbose=2)
           Epoch 67/75
           534/534 - 1s - loss: 3.5954e-05
           Epoch 68/75
           534/534 - 1s - loss: 3.3395e-05
           Epoch 69/75
           534/534 - 1s - loss: 2.7003e-05
           Epoch 70/75
           534/534 - 1s - loss: 3.2815e-05
           Epoch 71/75
           534/534 - 2s - loss: 2.8905e-05
           Epoch 72/75
           534/534 - 2s - loss: 3.0152e-05
           Epoch 73/75
           534/534 - 2s - loss: 2.9395e-05
           Epoch 74/75
           534/534 - 1s - loss: 2.5979e-05
           Epoch 75/75
           534/534 - 1s - loss: 2.0701e-05
Out[11]: <tensorflow.python.keras.callbacks.History at 0x21d9ffdc580>
In [12]:
            1 ypred = model.predict(X_test)
               print("y1 MSE: ", mean_squared_error(t_test.iloc[:, 0], ypred[:,0]))
#print("y2 MSE: ", mean_squared_error(t_test.iloc[:, 1], ypred[:,1])
            #print("y2 MSE: ", mean_squared_error(t_test.iloc[:, 1], ypred[:,1]))
#print("y3 MSE: ", mean_squared_error(t_test.iloc[:, 2], ypred[:,2]))
          y1 MSE: 1.1785877934678795e-06
In [13]:
            1 x_ax = range(len(X_test))
               plt.scatter(x_ax, t_test.iloc[:, 0], s=6, label="y1-test")
plt.scatter(x_ax, ypred[:,0], label="y1-pred",c="red",alpha = 0.1)
             6 plt.legend()
                plt.show()
             0.010
                                                              y1-test
                                                              y1-pred
             0.005
             0.000
            -0.005
            -0.010
            -0.015
                               400
                                           800
                                               1000 1200 1400 1600
                    0
                         200
                                     600
In [14]:
            1 x_ax = range(len(X_test))
             3 y_test_index = np.argsort(t_test.iloc[:, 0], axis=0).to_numpy()
            5 f = plt.figure()
               plt.scatter(x_ax, t_test.iloc[y_test_index], s=6, label="y_test")
                plt.scatter(x_ax, ypred[y_test_index], s=6, label="pred",c="orange", alpha=0
               #plt.ylim(t_test.iloc[y_test_index[0]].to_numpy()[0])
            9 plt.legend()
            10 plt.show()
           11
           12 f.savefig("foo.pdf", bbox_inches='tight')
             0.010
                        y test
                        pred
             0.005
             0.000
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