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#Importing Libraries
In [ ]:
         import numpy as np
         import matplotlib.pyplot as plt
         import pandas as pd
         from sklearn.preprocessing import LabelEncoder
         from sklearn.utils import shuffle
         #Constants
         #After functional day and 0 rows removal
         TRAIN SIZE = 6772
         TEST SIZE = 1693
         ACTIVATION_F = 'tanh'
         #Import Training Set
         df = pd.read_csv('SeoulBikeData.csv',engine='python')
         dummies = pd.get dummies(df.Seasons)
         df= pd.concat([df,dummies],axis='columns')
         df= df.drop(['Seasons','Winter'], axis='columns')
         le= LabelEncoder()
         dfle = df
         df.Holiday=le.fit transform(dfle.Holiday)
         df['Functioning Day']=le.fit transform(dfle['Functioning Day'])
         df=shuffle(df)
         df=df[df['Functioning Day'] == 1]
         df = df.drop(['Functioning Day'], axis='columns')
         training set df= df.iloc[:TRAIN SIZE, 1:]
         training set= df.iloc[:TRAIN SIZE, 1:].values
         y_set= df.iloc[:TRAIN_SIZE, 1].values
         #Feature Scaling
         from sklearn.preprocessing import MinMaxScaler
         sc= MinMaxScaler(feature range = (0,1))
         training set scaled = sc.fit transform(training set)
         sc2= MinMaxScaler(feature_range = (0,1))
         y_set=y_set.reshape(-1,1)
         y set scaled = sc2.fit transform(y set)
         X train=np.array(training set scaled[:,1:])
         Y_train=np.array(training_set_scaled[:,0])
         #reshaping
         X_train = np.reshape(X_train, (X_train.shape[0], 1, 13))
         # Importing the Keras libraries and packages
         !pip install -q -U keras-tuner
         from keras.models import Sequential
         from keras.layers import Dense
         from keras.layers import LSTM
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from keras.layers import Dropout
from keras import backend as K
from kerastuner.tuners import RandomSearch
import keras
def coeff_determination(y_true, y_pred):
    SS res = K.sum(K.square( y true-y pred ))
    SS_tot = K.sum(K.square( y_true - K.mean(y_true) ) )
    return ( 1 - SS_res/(SS_tot + K.epsilon()) )
def build_model(hp):
    regressor = Sequential()
    regressor.add(LSTM(units=hp.Int('units ', min value=32, max value=256, step=32), ac
    for i in range(hp.Int('num layers', 0, 20)):
        regressor.add(LSTM(units=hp.Int('units_',
                                             min_value=32,
                                             max value=256,
                                             step=32),
                               activation=ACTIVATION F, return sequences=True,))
        regressor.add(Dropout(0.1))
    regressor.add(LSTM(units=hp.Int('units_',min_value=32,max_value=256,step=32),activa
    regressor.add(Dropout(0.1))
    regressor.add(Dense(units=1))
    regressor.compile(
        optimizer=keras.optimizers.RMSprop(
            hp.Choice('learning_rate', [1e-2, 1e-3, 1e-4])),
        loss='mean squared error',
    return regressor
tuner = RandomSearch(
    build model,
    objective='val_loss',
    max_trials=10,
    executions per trial=2,
    directory='project',
    project name='INNS')
tuner.search_space_summary()
tuner.search(X train, Y train,
             epochs=100,
             validation split=0.2)
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In [ ]: tuner.results_summary()
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