```
# Libraries and settings
import pandas as pd
import numpy as np
import math
import sklearn
import sklearn.preprocessing
import datetime
import os
import matplotlib.pyplot as plt
import tensorflow as tf
import matplotlib.pyplot as plt
import keras
from keras.models import Sequential
from keras.layers import Dense, Dropout, Flatten, Reshape, GlobalAveragePooling1D
from keras.layers import Conv2D, MaxPooling2D, Conv1D, MaxPooling1D
from keras.utils import np utils
#display parent directory and working directory
print(os.path.dirname(os.getcwd())+':', os.listdir(os.path.dirname(os.getcwd())));
print(os.getcwd()+':', os.listdir(os.getcwd()));
# Analyze data
df = pd.read_csv("../../prices-split-adjusted.csv", index_col = 0)
print(df.info())
print(df.head())
print(df.values.shape)
# number of different stocks
print('\nnumber of different stocks: ', len(list(set(df.symbol))))
print(list(set(df.symbol))[:10])
df.tail()
df.describe()
#plot data
plt.figure(figsize=(15, 5));
plt.subplot(1,2,1);
plt.plot(df[df.symbol == 'EQIX'].open.values, color='red', label='open')
plt.plot(df[df.symbol == 'EQIX'].close.values, color='green', label='close')
plt.plot(df[df.symbol == 'EQIX'].low.values, color='blue', label='low')
plt.plot(df[df.symbol == 'EQIX'].high.values, color='black', label='high')
plt.title('stock price')
plt.xlabel('time [days]')
```

```
plt.ylabel('price')
plt.legend(loc='best')
#plt.show()
plt.subplot(1,2,2);
plt.plot(df[df.symbol == 'EQIX'].volume.values, color='black', label='volume')
plt.title('stock volume')
plt.xlabel('time [days]')
plt.ylabel('volume')
plt.legend(loc='best');
# Manipulate data
#- choose a specific stock
#- drop feature: volume
#- normalize stock data
#- create train and test data sets
def feature normalize(train):
    train_norm = train.apply(lambda x: (x - np.min(x)) / (np.max(x) - np.min(x))) #
標準化(介於 0~1 之間)
    return train norm
## 很重要 切割視窗
def create segments and labels(df, time steps, step):#, label name):
    This function receives a dataframe and returns the reshaped segments
    of x,y,z acceleration as well as the corresponding labels
    Args:
         df: Dataframe in the expected format
        time steps: Integer value of the length of a segment that is created
    Returns:
        reshaped segments
        labels:
    111111
#圖畫中的 overlap 越高,代表資料中的相關性越強
#圖中80筆資料一次跳40筆,代表其並非相關性高
#feature 有四個
    N FEATURES = 4
#選擇測試切出 20%
```

```
test_set_size_percentage = 20
    segments = []
    labels = []
    data raw = df.as matrix()
#創造時間窗,將所有選擇特徵一起切割視窗
    for i in range(0, len(data raw) - time steps, step):#
         segments.append(data raw[i: i + time steps])
    segments = np.array(segments);
    test set size =
int(np.round(test set size percentage/100*segments.shape[0]));
    train set size = segments.shape[0] - (test set size);
#以訓練資料占比分割訓練測試集,並以視窗最後一筆資料當作預測值
    x train = segments[:train set size,:-1,:]
    y_train = segments[:train_set_size,-1,:]
#
       x_valid = data[train_set_size:train_set_size+valid_set_size,:-1,:]
#
       y valid = data[train set size:train set size+valid set size,-1,:]
    x_test = segments[train_set_size:,:-1,:]
    y test = segments[train set size:,-1,:]
    return [x_train, y_train, x_test, y_test]
#
       return [x_train, y_train, x_valid, y_valid, x_test, y_test]
# choose one stock & drop volume
df stock = df[df.symbol == 'EQIX'].copy()
df stock.drop(['symbol'],1,inplace=True)
df stock.drop(['volume'],1,inplace=True)
cols = list(df stock.columns.values)
print('df stock.columns.values = ', cols)
# normalize stock
df_stock_norm = df_stock.copy()
df_stock_norm = feature_normalize(df_stock_norm)
```

```
# create train, test data
time_steps = 20# choose sequence length
step = 5
x train, y train, x test, y test = create segments and labels(df stock norm,
time steps, step)
print('x_train.shape = ',x_train.shape)
print('y train.shape = ', y train.shape)
# print('x valid.shape = ',x valid.shape)
# print('y valid.shape = ', y valid.shape)
print('x test.shape = ', x test.shape)
print('y_test.shape = ',y_test.shape)
df stock norm.values.shape
plt.plot(df stock norm.open.values, color='red', label='open')
plt.plot(df stock norm.close.values, color='green', label='close')
plt.plot(df stock norm.low.values, color='blue', label='low')
plt.plot(df stock norm.high.values, color='black', label='high')
#plt.plot(df_stock_norm.volume.values, color='gray', label='volume')
plt.title('stock')
plt.xlabel('time [days]')
plt.ylabel('normalized price/volume')
plt.legend(loc='best')
plt.show()
#reshape 資料
num time periods, num sensors = x train.shape[1], x train.shape[2]
input shape = (num time periods*num sensors)
                                                    ## 80*3 每一筆資料 80(時
間窗) 3 個變數( xyz)
x train reshape = x train.reshape(x train.shape[0], input shape).astype('float32')
print(f"x train reshape.shape:{x train reshape.shape}")
x test reshape = x test.reshape(x test.shape[0], input shape).astype('float32')
print(f"x test reshape.shape:{x test reshape.shape}")
```

#建立模型 ## units 可自行調整,設越大訓練越久 ## kernel_initializer:函數初始值 ## activation:激勵函數可自行挑選 # units 需設為 13,否則會報錯 model_mlp = Sequential() model_mlp.add(Dense(units=256, input_dim = 76, kernel_initializer="normal", activation="softmax")) model_mlp.add(Dense(units=100, kernel_initializer="normal", activation="relu")) model_mlp.add(Dense(units=4, kernel_initializer="normal", activation="relu")) print(model_mlp.summary())

#開始訓練

```
# model.compile(loss="categorical_crossentropy", optimizer="Adam", metrics=["accuracy"])
model_mlp.compile(loss="MSE", optimizer="Adam", metrics=["mse"])
# epochs:訓練幾次
train_history = model_mlp.fit(x = x_train_reshape, y = y_train, validation_split=0.18, epochs=50, batch_size=30, verbose=2)
```

#繪圖

```
import matplotlib.pyplot as plt
def show_train_history(train_history, train, validation):
    plt.plot(train_history.history[train])
    plt.plot(train_history.history[validation])
    plt.title("Train History")
    plt.ylabel(train)
    plt.xlabel('Epoch')
    plt.show()
```

show_train_history(train_history, "mean_squared_error", "val_loss") ## 訓練正確率圖

```
x_test_reshape = x_test.reshape(x_test.shape[0], input_shape).astype('float32')
print(f"x_test_reshape.shape:{x_test_reshape.shape}")
scores = model_mlp.evaluate(x_test_reshape, y_test)
print(f"MSE:{scores[0]}")
```