

1. 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()));
```

2. 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()
```

#3.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');
```

3. 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:

```
    """
```

#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

```
print("\n--- Create neural network model ---\n")
```

```
input_shape = (x_train.shape[1], x_train.shape[2], 1) ## 定義 CNN 的輸入維度!! (10*10)
```

```
# 建立簡單的線性執行的模型
```

```
# model.add(Convolution2D(32, (3, 3), activation='relu', input_shape=(1,28,28), data_format='channels_first'))
```

```
model_cnn2d = Sequential()
```

```
# 建立卷積層，filter=32,即 output space 的深度, Kernal Size: 3x3, activation function 採用 relu
```

```
model_cnn2d.add(Conv2D(32, kernel_size=(3, 3), activation='relu', input_shape=input_shape)) ## 注意喔!! 這一邊放的是 (10,10,1) 的型態
```

```
# 建立卷積層，filter=64,即 output size, Kernal Size: 3x3, activation function 採用 relu
```

```
# model_cnn2d.add(Conv2D(64, (3, 3), activation='relu'))
```

```
# 建立池化層，池化大小=2x2，取最大值
```

```
model_cnn2d.add(MaxPooling2D(pool_size=(2, 2)))
```

```
# Dropout 層隨機斷開輸入神經元，用於防止過度擬合，斷開比例:0.25
```

```
model_cnn2d.add(Dropout(0.25))
```

```
# Flatten 層把多維的輸入一維化，常用在從卷積層到全連接層的過渡。
```

```
model_cnn2d.add(Flatten())
```

```
# 全連接層: 128 個 output
```

```
model_cnn2d.add(Dense(128, activation='relu'))
```

```
# Dropout 層隨機斷開輸入神經元，用於防止過度擬合，斷開比例:0.5
```

```
model_cnn2d.add(Dropout(0.5))
```

```
# 使用全連接層 softmax activation function，將結果分類
```

```
# model.add(Dense(num_classes, activation='softmax'))
```

```
model_cnn2d.add(Dense(units=64, kernel_initializer="normal", activation="softmax"))
```

```
model_cnn2d.add(Dense(units=4, kernel_initializer="normal", activation="softmax"))
```

```
# 編譯: 選擇損失函數、優化方法及成效衡量方式
```

```
# model.compile(loss=keras.losses.categorical_crossentropy, optimizer=keras.optimizers.Adadelta(), metrics=['accuracy'])
```

```
model_cnn2d.summary()
# x_train.reshape(x_train.shape[0], input_shape).astype('float32')
x_train_2D =
x_train.reshape(x_train.shape[0],x_train.shape[1],x_train.shape[2],1).astype('float32'
)
x_test_2D =
x_test.reshape(x_test.shape[0],x_test.shape[1],x_test.shape[2],1).astype('float32')
```

#開始訓練

```
BATCH_SIZE = 400
EPOCHS = 50
# training stops early
callbacks_list = [
    keras.callbacks.ModelCheckpoint(
        filepath='best_model.{epoch:02d}-{val_loss:.2f}.h5',
        monitor='val_loss', save_best_only=True),
    keras.callbacks.EarlyStopping(monitor='acc', patience=1)
]

model_cnn2d.compile(loss='MSE',
                    optimizer='adam', metrics=['MSE'])

history = model_cnn2d.fit(x_train_2D, y_train, batch_size=BATCH_SIZE,
epochs=EPOCHS, callbacks=callbacks_list, validation_split=0.2, verbose=1)

# %%

print("\n--- Learning curve of model training ---\n")
```

#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
```

```
scores = model_cnn2d.evaluate(x_test_2D, y_test)
print(f"MSE:{scores[0]}")
```