

1. Libraries and settings

```
import pandas as pd
import numpy as np
import math
import sklearn
import sklearn.preprocessing
from sklearn import metrics
from sklearn.metrics import classification_report
import seaborn as sns
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()
```

#看資料

```
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:

"""

#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(df) - time_steps, step):#

segments.append(df.values[i: i + time_steps])

#以當期四種特徵預測下一期收盤價

rate = (df.open.values[i + time_steps]-df.open.values[i + time_steps-1])/df.open.values[i + time_steps-1]

temp = rate

if temp < 0:

if temp <= -0.2:

label = 0

elif temp <= -0.1:

label = 1

elif temp < 0:

label = 2

else:

if temp == 0:

label = 3

elif temp <= 0.1:

label = 4

elif temp <= 0.2:

label = 5

elif temp > 0.2:

label = 6

labels.append([label])

```

test_set_size =
np.round(test_set_size_percentage/100*np.asarray(segments).shape[0])
train_set_size = int(np.asarray(segments).shape[0] - (test_set_size));
print(train_set_size)
#     segments = np.array(segments);
    reshaped_segments_train = np.asarray(segments[:train_set_size], dtype=
np.float32).reshape(-1, time_steps, N_FEATURES)
    reshaped_segments_test = np.asarray(segments[train_set_size:], dtype=
np.float32).reshape(-1, time_steps, N_FEATURES)
    labels_train = np.asarray(labels[:train_set_size])
    labels_test = np.asarray(labels[train_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:]#(1394, 19, 4)
#     y_train = labels[:train_set_size,-1:]#(1394, 4)
#     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 = labels[train_set_size:-1:]
    return reshaped_segments_train, labels_train,
    reshaped_segments_test, labels_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_test.shape = ',x_test.shape)
print('y_test.shape = ', y_test.shape)
num_classes = 7
y_train_onehot = np_utils.to_categorical(y_train, num_classes)
print(f"y_train_onehot:{y_train_onehot.shape}")
y_test_onehot = np_utils.to_categorical(y_test, num_classes)
print(f"y_test_onehot:{y_test_onehot.shape}")
#繪刪除特徵後圖形
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
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

```

# Compatibility layer between Python 2 and Python 3
from __future__ import print_function
from matplotlib import pyplot as plt
import numpy as np
import pandas as pd
import seaborn as sns

```

```

from scipy import stats

from sklearn import metrics
from sklearn.metrics import classification_report
from sklearn import preprocessing

import keras
from keras.utils import np_utils
from keras.models import Sequential
from keras.layers import Dense, Dropout, Activation, Flatten, LSTM, TimeDistributed,
RepeatVector, GRU
from keras.layers import SimpleRNN, Activation, Dense, RNN
from keras.layers.normalization import BatchNormalization
from keras.optimizers import Adam
from keras.callbacks import EarlyStopping, ModelCheckpoint
import matplotlib.pyplot as plt
%matplotlib inline
from keras.callbacks import ReduceLROnPlateau

learning_rate_function = ReduceLROnPlateau(monitor='val_acc',
                                           patience=3, #準確率重複 3 次
                                           verbose=1,
                                           factor=0.5, #準確率乘上
                                           min_lr=0.00001)

就要減少

factor 設成下一個 learning_rate

model_gru = Sequential()
## SimpleRNN 注意他的 input 格式!!(先變成三維，把第二維、三維放在這邊)
# model_lstm.add(SimpleRNN(units=128, input_shape=(x_train.shape[1],
x_train.shape[2])))
# model_gru.add(LSTM(128, input_length= x_train.shape[1], input_dim=
x_train.shape[2],return_sequences=False))
model_gru.add(GRU(units=128, input_shape=( x_train.shape[1], x_train.shape[2])))
#units 隱藏層神經元個數
model_gru.add(Dropout(0.1))

```

```
model_gru.add(Dense(64, activation = 'relu'))
model_gru.add(Dropout(0.1))
model_gru.add(Dense(16, activation = 'relu'))#三層隱藏層
model_gru.add(Dropout(0.1))
model_gru.add(Dense(num_classes, activation='softmax'))
#num_classes 不能改
model_gru.compile(loss='categorical_crossentropy',
                  optimizer='adam', metrics=['accuracy'])
model_gru.summary()
```

#開始訓練

```
print("\n--- Fit the model ---\n")
```

```
train_history = model_gru.fit(x=x_train, y= y_train_onehot, validation_split=0.1,
epochs=550,
batch_size=10,callbacks=[learning_rate_function],verbose=2)#callbacks=[learning_rate_function], verbose=2)
```

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

```
get_ipython().magic('matplotlib inline')
```

IPython 有一組預先定義好的所謂的魔法函數（ Magic Functions ），你可以通過命令列的語法形式來訪問它們。

#繪圖

#訓練驗證圖

```
# summarize history for accuracy and loss
plt.figure(figsize=(6, 4))
plt.plot(history.history['acc'], "g--", label="Accuracy of training data")
plt.plot(history.history['val_acc'], "g", label="Accuracy of validation data")
plt.plot(history.history['loss'], "r--", label="Loss of training data")
plt.plot(history.history['val_loss'], "r", label="Loss of validation data")
plt.title('Model Accuracy and Loss')
plt.ylabel('Accuracy and Loss')
plt.xlabel('Training Epoch')
plt.ylim(0)
```

```
plt.legend()
plt.show()
```

#正確誤差圖

```
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, "acc", "val_acc") ## 訓練正確率圖

show_train_history(train_history, "loss", "val_loss") ## 訓練誤差圖
```

#評估測試準確度

```
score = model_gru.evaluate(x_test, y_test_onehot, verbose=1)

print("\nAccuracy on test data: %0.2f" % score[1])
print("\nLoss on test data: %0.2f" % score[0])
```

#混沌矩陣

```
# %%

print("\n--- Confusion matrix for test data ---\n")
y_pred_test = model_gru.predict(x_test)
# Take the class with the highest probability from the test predictions
max_y_pred_test = np.argmax(y_pred_test, axis=1)
max_y_test = np.argmax(y_test_onehot, axis=1)

show_confusion_matrix(max_y_test, max_y_pred_test)

# %%

print("\n--- Classification report for test data ---\n")
```



```
print(classification_report(max_y_test, max_y_pred_test))
```