# AI 作業二

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## Colab 連結:

Text Generation: <a href="https://colab.research.google.com/drive/1CHxE3jOds2XGrC\_ys8ede\_zg">https://colab.research.google.com/drive/1CHxE3jOds2XGrC\_ys8ede\_zg</a> <a href="https://colab.research.google.com/drive/1CHxE3jOds2XGrC\_ys8ede\_zg">https://colab.research.google.com/drive/1CHxE3jOds2XGrC\_ys8ede\_zg</a> <a href="https://colab.research.google.com/drive/1CHxE3jOds2XGrC">https://colab.research.google.com/drive/1CHxE3jOds2XGrC\_ys8ede\_zg</a> <a href="https://colab.research.google.com/drive/1CHxE3jOds2XGrC">https://colab.research.google.com/drive/1CHxE3jOds2XGrC</a> ys8ede\_zg</a>

Stock RNN: <a href="https://colab.research.google.com/drive/1vyoRE1NIKGiso6tF1gskCb2aCF\_D">https://colab.research.google.com/drive/1vyoRE1NIKGiso6tF1gskCb2aCF\_D</a>
5D-r?usp=sharing

## 評分標準:

## 作業繳交

- 1. 完成訓練後保存output結果,更改\*\*檔名學號\*\*,左上角 檔案->下載 成ipynb檔。
- 2. 交一個\*\*pdf\*\*檔,裡面需要附上\*\*你的作業colab連結(設為可編輯、所有人觀看)\*\*、姓名學號年級、過程和執行結果的截圖,並說明你程式撰寫的過程、本次作業心得(字數不限)。
- 3. 上述三個檔案繳交至ee-class。(Text Generation、Stock、PDF)

#### 算分標準

Text Generation(20) + Stock RNN(30) + Predict function(20分) + 30分文件

- 1. [Text Generation] 換個其他資料集來訓練,完成就20分
- 2. Stock RNN完成模型就30分
- 3. predit function有寫出來就20分(第七段)
- 4.30分是文件分數

### **Text Generation**

# 過程截圖:

剩餘92167個字

· 2. 取得資料集

```
[] # 作業之一就是試試看其他本小說
   book = ""
with open("content/You_Are_the_Apple_of_My_Eye_txt","r",encoding="utf8") as file:
for line in file:
book == line
   book_length = len(book)
unique_words = set(book)
print(*(郵座年, 表們一起追的支流) 共有(book_length)字詞)
print(*金子 (len(unique_words)) 個漢一集二的字(含葉貼符號)\m^)
print(book[0:800])
   《分漢爾德》與咸敬
內容頭介
「歷時主構思、一年萬德、十個月的抽攝與後數。電影雷力走出了小說,躍上了大線幕。我也終於學會,不用傷口,就配住青春裡最重要的事。這場戰鬥獻始你們,一路支持我的讀者。獻始,我的女孩,」——九耙刀
男孩用電影打造了時光機,只為了再一次與政務相急。
8/19,帶著你的命中就定,走進電影院,再一次收藏青春
現在,我是跟妳設~~~
現在,我是跟妳設~~~
初風誘径?
初開誘径?
  [] # 計算字數統計
        words_count = {}
        for w in book:
           if w in words_count:
                 words_count[w] += 1
            else:
                 words_count[w] = 1
       words_count = sorted(words_count.items(), key=lambda x:x[1])
  [] stop\_word = 8
       unique_words = [w_tup[0] for w_tup in words_count if w_tup[1]>stop_word]
        print(f"去除次數小於{stop_word}的文字剩餘 : {len(unique_words)}")
        去除次數小於8的文字剩餘: 1137
  [] print(f"原本《那些年,我們一起追的女孩》共有 {book_length} 字詞")
        print(f"去除不常出現的文字後")
        book \ = \ [w \ for \ w \ in \ book \ if \ w \ in \ unique\_words]
        print(f"剩餘{len(book)}個字")
        原本《那些年,我們一起追的女孩》共有 97045 字詞
        去除不常出現的文字後
```

```
68/68 [=======] - 23s 337ms/step - loss: 4.8978
  Epoch 6/20
С⇒
  Epoch 7/20
  68/68 [====
                ========= ] - 24s 349ms/step - loss: 4.5443
  Epoch 8/20
  68/68 [============] - 24s 348ms/step - loss: 4.3714
  Epoch 9/20
  Epoch 10/20
  68/68 [====
                ========] - 25s 363ms/step - 1oss: 4.0150
  Epoch 11/20
  68/68 [===========] - 24s 352ms/step - loss: 3.8066
  Epoch 12/20
  68/68 [====
              Epoch 13/20
  68/68 [============= ] - 24s 351ms/step - loss: 3.3124
  Epoch 14/20
  68/68 [============ ] - 24s 347ms/step - loss: 3.0044
  Epoch 15/20
  68/68 [===========] - 24s 350ms/step - loss: 2.6437
  Epoch 16/20
  68/68 [===========] - 24s 353ms/step - loss: 2.2235
  Epoch 17/20
  68/68 [=======] - 24s 348ms/step - loss: 1.7619
  Epoch 18/20
  Enoch 19/20
  68/68 [============] - 24s 351ms/step - loss: 0.8481
  Epoch 20/20
  68/68 [===========] - 24s 351ms/step - 1oss: 0.5273
```

## 執行結果:

```
[3] Sist_seq = "$$\frac{\pi_{\text{seq}}}{\text{constant}} = \text{init_seq_ind(-seq_lent)} \ for \times in \text{init_seq_lent(-seq_lent)} \ generateFords(input, 1000)

$\text{generateFords(input, 1000)} \ \text{generateFords(input, 1000)} \ \t
```

### Stock RNN

# 過程截圖:

# 模型

```
[] # todo
input_shape = [BATCH_SIZE, seq_len ,1 ]
output_shape = [BATCH_SIZE, seq_len ,1 ]
keras.backend.clear_session()
model = tf.keras.Sequential()

model.add(tf.keras.layers.LSTM(units = 256,input_shape=(x_train.shape[1], x_train.shape[2]),return_sequences=True))
# print(input_shape)
model.add(tf.keras.layers.LSTM(units = 256,return_sequences=True))
model.add(tf.keras.layers.ISTM(units = 256,return_sequences=True))
model.add(tf.keras.layers.ISTM(units = 256,return_sequences=True))
model.add(tf.keras.layers.DimeDistributed(tf.keras.layers.Dense(1)))

model.add(tf.keras.layers.DimeDistributed(tf.keras.layers.Dense(1)))

# model.add(tf.keras.layers.Flatten())
model.add(tf.keras.layers.Dense(1))

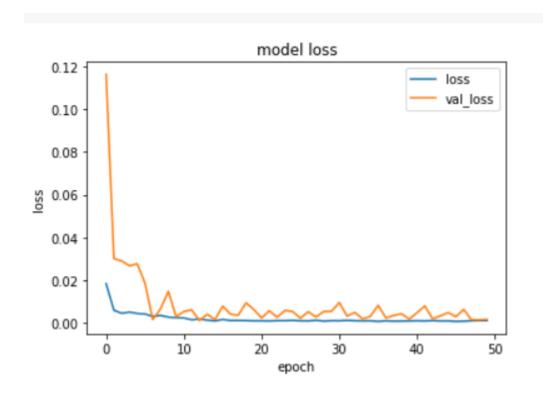
# model.add(tf.keras.layers.Dense(1))

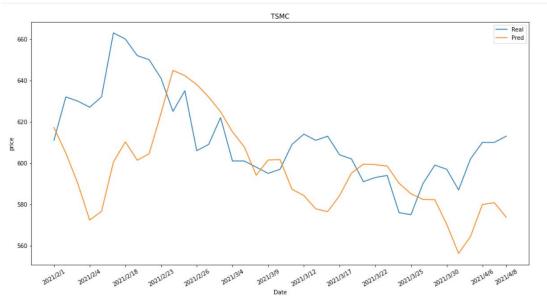
# model.compile(loss="mse", optimizer="adam")
# model.build(input_shape)
model.summary()
```

# 訓練計畫

```
Epoch 33/50
[]
    11/11 [=====
                         =======] - Os 9ms/step - loss: 9.9146e-04 - val_loss: 0.0049
    Epoch 34/50
                        ========] - Os 9ms/step - loss: 9.3336e-04 - val_loss: 0.0019
    11/11 [=====
    Epoch 35/50
    11/11 [=====
                        ========] - Os 10ms/step - loss: 9.6181e-04 - val_loss: 0.0031
    Epoch 36/50
    11/11 [=====
                 Epoch 37/50
    11/11 [=====
                        =========] - Os 10ms/step - loss: 9.6148e-04 - val_loss: 0.0023
    Epoch 38/50
    11/11 Г=====
                        =======] - Os 10ms/step - 1oss: 8.0041e-04 - val_loss: 0.0035
    Epoch 39/50
    11/11 [=====
                        =======] - Os 9ms/step - loss: 8.5752e-04 - val_loss: 0.0043
    Epoch 40/50
    11/11 [=====
                         =======] - Os 10ms/step - loss: 9.0216e-04 - val_loss: 0.0018
    Epoch 41/50
                 =========] - Os 9ms/step - loss: 9.7620e-04 - val_loss: 0.0048
    11/11 [======
    Epoch 42/50
    11/11 [=====
                        ========] - Os 10ms/step - loss: 8.9917e-04 - val_loss: 0.0079
    Epoch 43/50
    11/11 [====
                          =======] - Os 10ms/step - loss: 0.0011 - val_loss: 0.0019
    Epoch 44/50
    11/11 「=====
                        =======] - Os 9ms/step - loss: 8.9099e-04 - val_loss: 0.0033
    Epoch 45/50
    11/11 [=====
                       ========] - Os 8ms/step - loss: 9.2151e-04 - val_loss: 0.0048
    Epoch 46/50
    11/11 [=====
                      Epoch 47/50
    11/11 [=====
                       ======== ] - Os 9ms/step - loss: 7.8482e-04 - val loss: 0.0063
    Epoch 48/50
    11/11 [=====
                         =======] - Os 8ms/step - loss: 0.0010 - val_loss: 0.0015
    Epoch 49/50
    11/11 [=====
                    -----] - 0s 8ms/step - loss: 0.0011 - val_loss: 0.0014
    Epoch 50/50
    11/11 [============] - Os 9ms/step - loss: 0.0011 - val_loss: 0.0016
```

# 執行結果:





# 作業心得:

文字生成器的部分我就是很水的找了九把刀的小說來用,助教連爬 蟲程式都給我們真的是太佛心了,另外,覺得文本生成器很酷,雖 然他跑出來的文本有點不符合人寫的文章,但是像是小說會有很多 「對話」,他都有把那個模式給模仿出來。

Stock RNN 基本是參考文字生成器的 model 寫的,並使用 LSTM。股票預測真的是一個很有用的東西耶,我不敢想像如果我真的能寫出一個預測很準的,我就直接靠股票賺錢就好了耶,多輕鬆愜意。

這次的作業讓我覺得更有實用價值,就像股票預測,是真的有可能 拿來運用賺錢的,然後,我覺得這些學到的東西,將來在寫專題的 時候一定很有幫助。