

# 災難推文辨識

## 第十組


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# 動機和目的

在Kaggle上，看到了這一個競賽，覺得有趣且實作出來是有應用價值的。

智能手機的普及使得人們能夠即時地發布他們觀察到的災難和緊急情況。因此，希望以編程方式來時時監測 Twitter 上的推文，讓救災組織可以在第一時間對發文的地點進行施救。

 Getting Started Prediction Competition

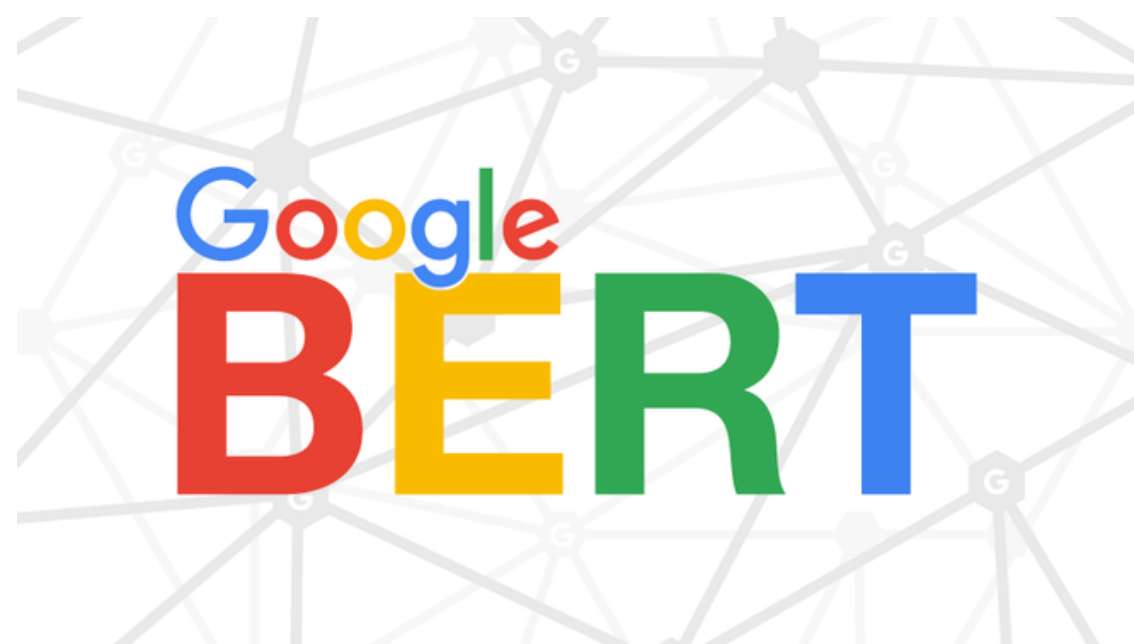
## Natural Language Processing with Disaster Tweets

Predict which Tweets are about real disasters and which ones are not



Kaggle · 1,135 teams · Ongoing

# 用到的技術



  
PyTorch

spaCy

# 實作過程

- 讀取資料集
- 資料預處理
- 訓練模型
- 預測並輸出結果

# 資料集

Each sample in the train and test set has the following information:

- The `text` of a tweet
- A `keyword` from that tweet (although this may be blank!)
- The `location` the tweet was sent from (may also be blank)

## 讀取資料集

```
data = pd.read_csv('nlp-getting-started/train.csv')  
test_data = pd.read_csv('nlp-getting-started/test.csv')
```

# 資料預處理

```
def keyword_preprocess(text):  
    """移除 '%20'"""  
    if pd.notnull(text):  
        text = text.replace("%20", " ")  
    else:  
        text = ''  
    return text  
  
def remove_url(text):  
    url_pattern = re.compile(r'https?://t\.co/[\s]*')  
    new_text = url_pattern.sub('', text)  
    return new_text  
  
def remove_at(text):  
    at_pattern = re.compile(r'@[\s]*')  
    new_text = at_pattern.sub('', text)  
    return new_text  
  
def text_preprocess(text):  
    """移除 url、@xxx"""  
    text = remove_url(text)  
    text = remove_at(text)  
    return text
```

*# combine keyword and text*

```
data['keyword_text'] = data.apply(lambda row: row['keyword'] + ' ' + row['text'], axis=1)  
test_data['keyword_text'] = test_data.apply(lambda row: row['keyword'] + ' ' + row['text'], axis=1)
```

將 keyword 標籤內有包含 '%20' 的替換掉

移除網址

移除@someone

將 keyword 加入 text(tweet)

# 載入BERT模型

初始化一個 BERT tokenizer ，用於將文本轉換成 BERT 模型所需的輸入格式

載入預訓練模型BERT，用於訓練一個基於BERT模型的序列分類器

```
checkpoint = "bert-base-uncased"
tokenizer = AutoTokenizer.from_pretrained(checkpoint)

def tokenize_function(example):
    return tokenizer(example["text"], truncation=True)

tokenized_train_dataset = train_dataset.map(tokenize_function, batched=True)
tokenized_test_dataset = test_dataset.map(tokenize_function, batched=True)

# use dynamic padding
data_collator = DataCollatorWithPadding(tokenizer=tokenizer)

model = AutoModelForSequenceClassification.from_pretrained(checkpoint, num_labels=2)
```

# 調整訓練參數並輸出結果

```
training_args = TrainingArguments(  
    "test-trainer",  
    report_to='none',  
    num_train_epochs=2,  
    save_strategy = "epoch"  
)  
  
trainer = Trainer(  
    model,  
    training_args,  
    train_dataset=tokenized_train_dataset,  
    data_collator=data_collator,  
    tokenizer=tokenizer,  
)  
  
trainer.train()  
  
predictions = trainer.predict(tokenized_test_dataset)  
preds = np.argmax(predictions.predictions, axis=-1)  
  
submission = pd.DataFrame({'id':test_data['id'],'target':preds})  
submission.to_csv('nlp-getting-started/submission1.csv', index=False)
```



# 提交結果

有預處理



**submission1.csv**

Complete · 19h ago

**0.83634**

沒有預處理



**submission\_test.csv**

Complete · now

**0.84063**

# 找出受災地區

```
import en_core_web_sm
nlp = en_core_web_sm.load()
def location_detect(text):
    doc = nlp(text)
    data = [(X.text, X.label_) for X in doc.ents]
    for word, pos in data:
        if pos == 'GPE':
            return(word)
test_data_location['location'] = test_data_location['text'].apply(location_detect)
# print(test_data_location['location'])
result = pd.DataFrame({'text':test_data_location['text'], 'target':preds, 'location':test_data_location['location']})
result = result[result['target'] == 1]
result = result[result['location'].notnull()]
print(result)
```

使用了spaCy套件的en\_core\_web\_sm模型  
，檢測每條貼文中的地點

	location
4	China
15	Birmingham
34	Lewes
36	Legal
52	Hiroshima
...	...
3238	Wreckage
3239	Wreckage
3254	Alabama
3257	the Village of Rajman
3260	Chicago

[441 rows x 3 columns]

**End**