

## 一、本周研究内容

### 研究内容：

1. 研究打标方案：调研了相关文献，针对本项目的 COVID19 的 Twitter 数据集提出自己的方案。
2. 实验代码进展：预处理部分更新，text-mining 部分更新。

## 二、项目实施当前状态

### 项目进度实施情况：

在大数据和自然语言处理部分，建立了 Twitter 数据集，对 Twitter 数据集进行了预处理，提出了打标方案，以及对部分数据进行了打标。

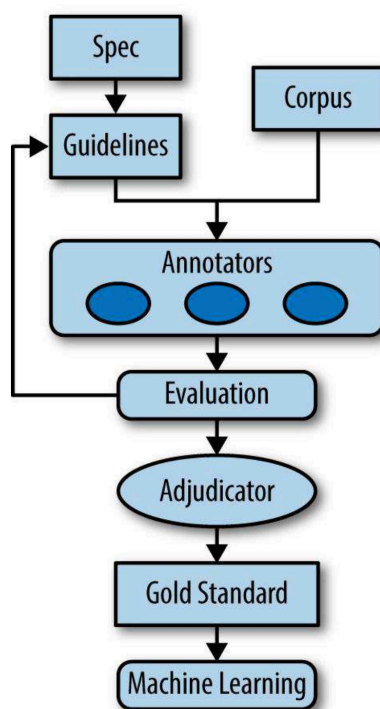
### 项目整体进度完成情况

大数据和自然语言处理部分：下周内容是写一个 baseline 方法，然后评估打标的质量。

## 三、本周成果

### 1. 打标的相关文献：

通过阅读教材“Natural Language Annotation for Machine Learning” [1] 学习了 NLP 数据的打标流程，高质量的打标是要通过不断的迭代以及调整 guidelines。



图一：NLP 数据打标流程

通过阅读文献 [2]，作者在文中建立了一个与本项目类似的系统，但使用 twitter 是用来判别是否 health-related，作者手动打标是根据他们设定的 1026 个与健康有关的关键词，然后针对每个关键词手动标注 10 个有关的 tweets。

## 2. 针对本项目提出的打标方案：

**Specifications:** The goal of the project is to build a Epidemic Diseases Early-warning System through COVID-19 Twitter Mining and Text Classification. As many users use Twitter to talk about public health topics and sometimes they also share information about the self-reporting of an illness (such as COVID19). We label the tweet data '1' if the tweet can be treated as a COVID-19 self-reporting, otherwise we label '0' .

### Guidelines:

- According to WHO official website [3] : COVID-19 Symptoms including *fever, dry cough, tiredness, aches, pains, nasal congestion, sore throat, diarrhea*. Hence the first guideline is to label tweet data '1' if it contains such keywords in health related context, but not news report.
- Other similar expressions about these symptoms.
- Twitter that indicated the user or her/his family member or friend has been tested positive.
- ...

## 3. 从打标数据中找到的一些例子：

- RT @CriticalCezanne: He had fever of 39 degrees for 4 consecutive days so went to hospital to get medicine. However, hospital had no medicine...
- Coronavirus outbreak: 3 still quarantined at hospital

## 4. 打标中要注意的问题以及局限和难点：

- 如果标注的噪声太大或者标签边界太过模糊（**大量标注错误，或标注规则写的太松、太模糊**，导致人都分不清某几个类别之间的区别），很可能再复杂的模型都在这份数据集上无法收敛。
- 如果标签与内容有非常直接的映射关系（**类别太过具体或标注规则写的太死**），例如只有出现某些关键词才可以贴为某个标签。则会导致一个非常简单的模型都会表现非常好，那这个模型学到的知识基本是没有什么实际意义的，没有了使用数据驱动的机器学习和深度学习的需要。

## 5. 代码进展：

### 5.1 原始数据集一共有 **2576357** 条 tweets。

```
[2]: pd.set_option('display.max_colwidth', -1)

dfs = glob.glob('*.csv')

result = pd.concat([pd.read_csv(df) for df in dfs], ignore_index=True)

...

[3]: result.shape

[3]: (2576357, 40)
```

图二：原始数据集

## 5.2 原始数据集中 twitter 为英文的一共有 1667122 条 tweets。

```
[34]: annotate_df = df[['id_str', 'full_text']].copy()

[35]: annotate_df = pd.DataFrame(annotate_df)
      annotate_df["label"] = np.nan
      annotate_df.shape

[35]: (1667122, 3)
```

图三：英文数据集

## 5.3 手动打标可交互代码实现。

```
[*]: manually_label('annotate_df_1.pickle')

Is this tweet content related to COVID-19 outbreak? Type 1 if yes. Type 0 if no.
Progress: 1/99

RT @OriginalDWoods: A dog has tested positive for the Coronavirus. White people about to find a cure ASAP now
1
```

图四：手动打标

## 5.4 文本预处理中，通过分析文本，发现包含 'rt' 比较多，在去停词 list 增加了 'rt' 。

### 3.4 Remove stop words

```
[57]: stopwords = nltk.corpus.stopwords.words('english')
      ## add 'rt' to stop word list
      stopwords.append('rt')

      def remove_stopwords(tokenized_list):
          text = [word for word in tokenized_list if word not in stopwords]# To remove all stopwords
          return text

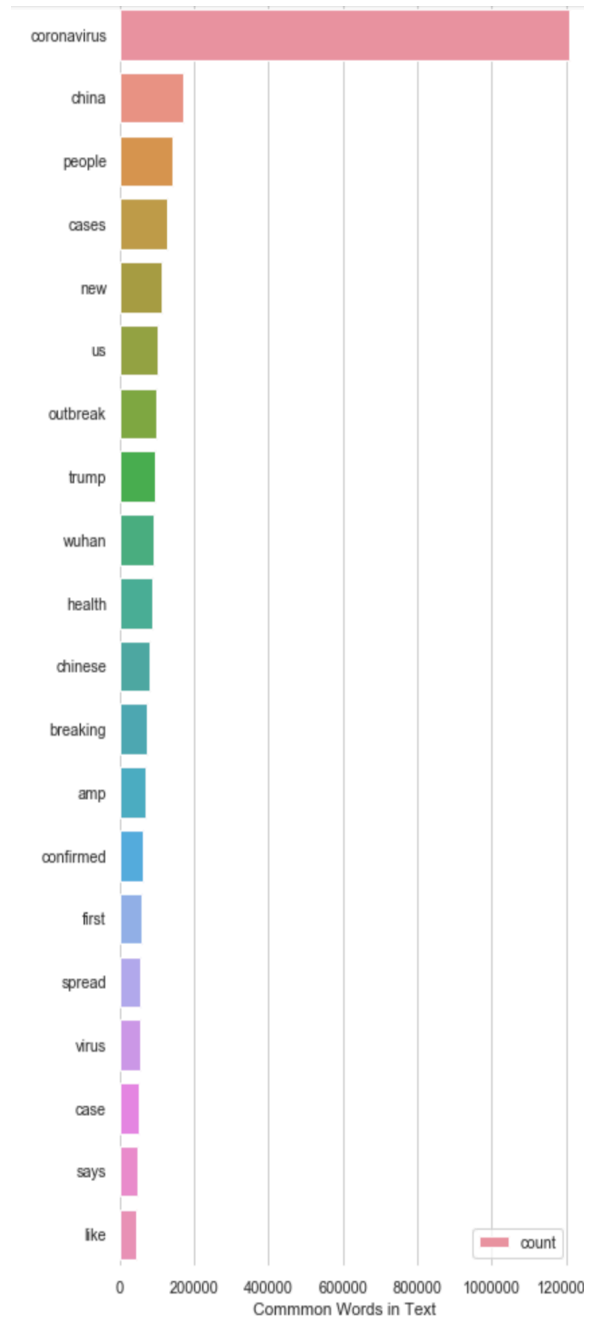
[58]: print(stopwords)

['i', 'me', 'my', 'myself', 'we', 'our', 'ours', 'ourselves', 'you', "you're", "you've", "you'll", "you'd", 'your', 'yours', 'yourself', 'yourselves', 'he', 'him', 'his', 'himself', 'she', "she's", 'her', 'hers', 'herself', 'it', "it's", 'its', 'itself', 'they', 'them', 'their', 'theirs', 'the', 'mselves', 'what', 'which', 'who', 'whom', 'this', 'that', "that'll", 'these', 'those', 'am', 'is', 'are', 'was', 'were', 'be', 'been', 'being', 'have', 'has', 'had', 'having', 'do', 'does', 'did', 'doing', 'a', 'an', 'the', 'and', 'but', 'if', 'or', 'because', 'as', 'until', 'while', 'of', 'at', 'by', 'for', 'with', 'about', 'against', 'between', 'into', 'through', 'during', 'before', 'after', 'above', 'below', 'to', 'from', 'up', 'down', 'in', 'on', 'off', 'over', 'under', 'again', 'further', 'then', 'once', 'here', 'there', 'when', 'where', 'why', 'how', 'all', 'any', 'both', 'each', 'few', 'more', 'most', 'other', 'some', 'such', 'no', 'nor', 'not', 'only', 'own', 'same', 'so', 'than', 'too', 'very', 's', 't', 'ca', 'n', 'will', 'just', 'don', "don't", 'should', "should've", 'now', 'd', 'll', 'm', 'o', 're', 've', 'y', 'ain', 'aren', 'aren't', 'couldn', "couldn't", 'didn', "didn't", 'doesn', "doesn't", 'hadn', "hadn't", 'hasn', "hasn't", 'haven', "haven't", 'isn', "isn't", 'ma', 'mightn', "mightn't", 'mustn', "mustn't", 'needn', "needn't", 'shan', "shan't", 'shouldn', "shouldn't", 'wasn', "wasn't", 'weren', "weren't", 'won', "won't", 'wouldn', "wouldn't", 'rt']
```

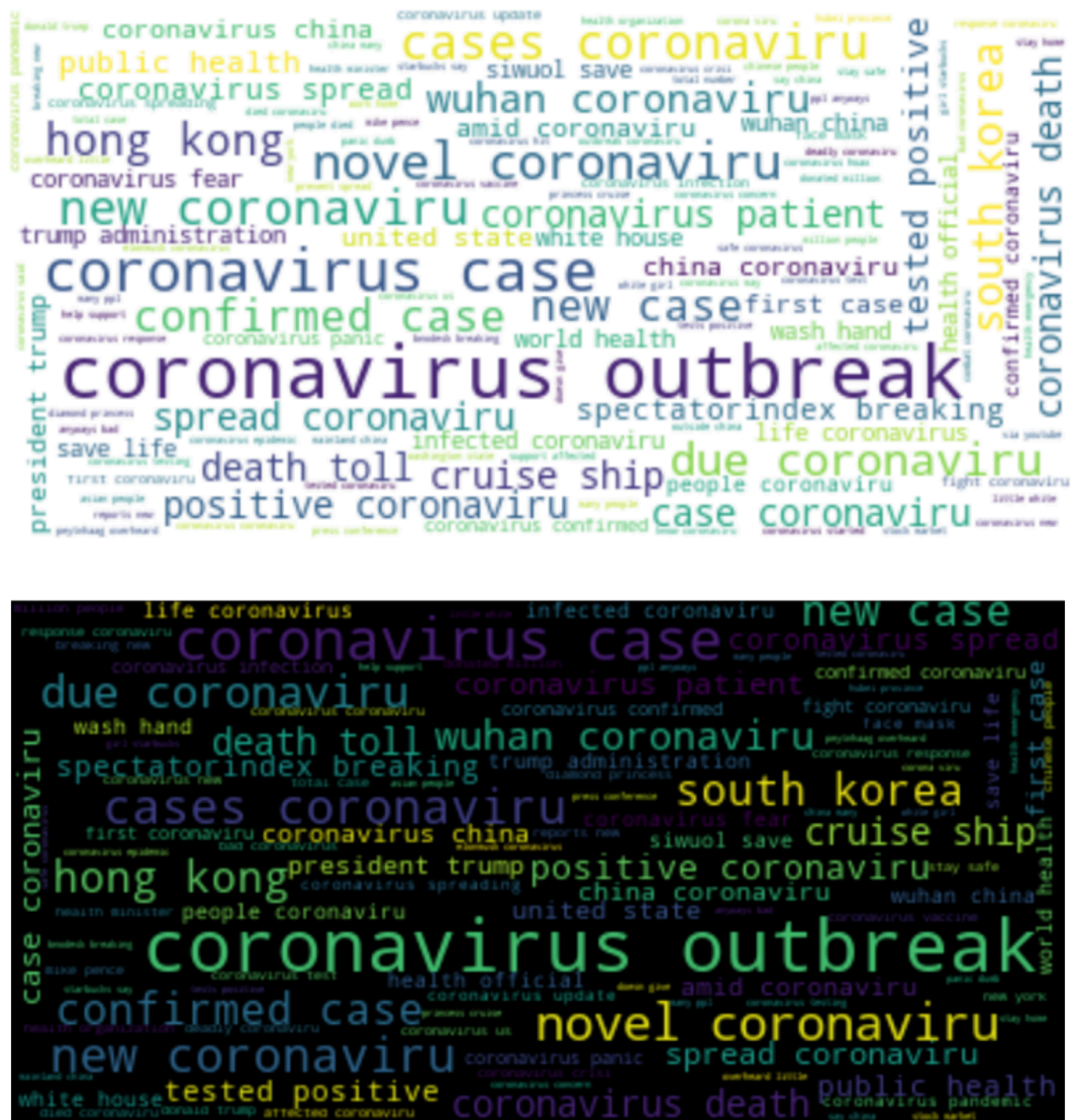
图五：去停词

## 5.5 Tweet Text mining 部分，找出了在文本中最常见的 top20 词，绘制了 table, horizontal chart 和 word cloud。

|    | Common_words | count   |
|----|--------------|---------|
| 0  | coronavirus  | 1207368 |
| 1  | china        | 172127  |
| 2  | people       | 142837  |
| 3  | cases        | 127516  |
| 4  | new          | 111619  |
| 5  | us           | 100599  |
| 6  | outbreak     | 97945   |
| 7  | trump        | 95318   |
| 8  | wuhan        | 91527   |
| 9  | health       | 86608   |
| 10 | chinese      | 81514   |
| 11 | breaking     | 71636   |
| 12 | amp          | 68597   |
| 13 | confirmed    | 61611   |
| 14 | first        | 57478   |
| 15 | spread       | 55880   |
| 16 | virus        | 55676   |
| 17 | case         | 52054   |
| 18 | says         | 47603   |
| 19 | like         | 45826   |



图六： COVID19 数据集中 Top 20 常见词



图七：用 COVID19 数据集生成词云

## 6. 发现的问题：原始数据集中无法正确解码 emoji 的表情，造成乱码以及文本缺失。

RT @EmeraldJohnson: Thank you George Mitt, Dean and Emily Simon for ensuring car manufacturing jobs remain in and paying commensurate wages  
RT @nypost: First case of coronavirus confirmed in Manhattan <https://t.co/DHZ5uROtAa> <https://t.co/6NfAtbDFT6>  
RT @thehowie: Thread: #COVID19 #Coronavirus updates & data.  
RT @tonyperson2: How worried are you by the coronavirus??  
RT @IbrahimLudwick: Somebody who got the coronavirus in China got a lung transplant to save their miserable life.  
RT @AKasingye: When friends bump into each. Even #coronaVirus can,Äôt come into your way. @DoreenNasaasira. Photo credit: David Lubz @933kfm,Ä¶  
#Covid19usa is highlighting problems in #health systems. | Carl Gibson <https://t.co/zlWCITL2tc>  
How can philanthropy plan for, respond to, and support communities affected by the #COVID-19 #Coronavirus? Join @funds4disaster this Thursday to learn at  
RT @Anna\_Rothschild: To my friends who are reporting on coronavirus: are there advisories yet for people with severe asthma? I've checked a,Ä¶

图八：Raw data 乱码问题

## 四、下周计划

1. 尝试解决发现的数据集无法正确解码的问题。
2. 尽快标注好数据集，使用小规模数据集或使用 semi-supervised learning 的方法。
3. 写一个 baseline 方法，然后对标注好的数据集进行迭代，测试在数据集上的表现，来评估打标的质量。

## Reference

[1] Pustejovsky, J. (2013). *Natural Language Annotation for Machine Learning*. OREILLY.

[2] Şerban, O., Thapen, N., Maginnis, B., Hankin, C. and Foot, V., 2019. Real-time processing of social media with SENTINEL: A syndromic surveillance system incorporating deep learning for health classification. *Information Processing & Management*, 56(3), pp.1166-1184.

[3] Q&A on coronaviruses (COVID-19). (2020). Retrieved 26 April 2020, from <https://www.who.int/news-room/q-a-detail/q-a-coronaviruses>