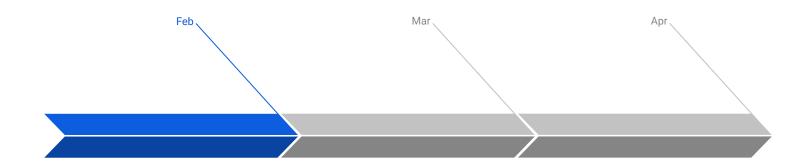
## **Disaster Tweets**

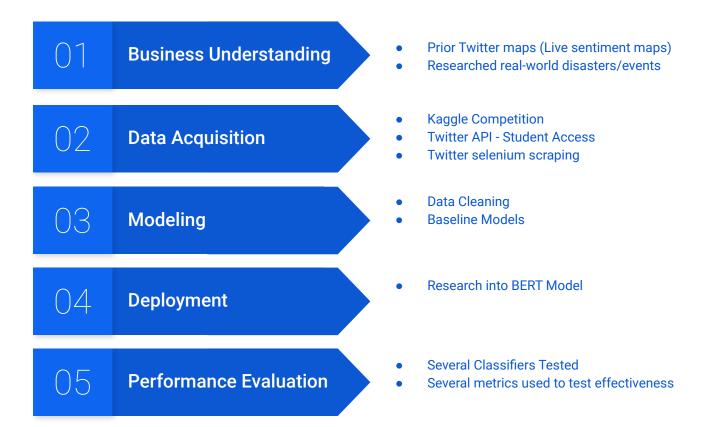
Week 5 Check-in

### **Timeline**



Baseline	Improved	Final
Working "simple" model	NLP / Attention	Interactive dashboard
POC Twitter scraping setup	BERT	Clustering mechanism

## **Business Lifecycle Progress**



#### **Twitter Scraping**

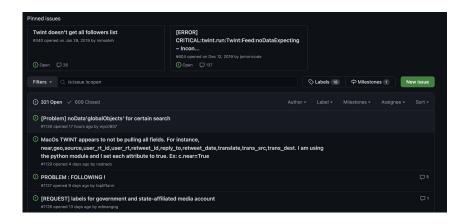
#### Scrapers broken

Legacy Web UI removed

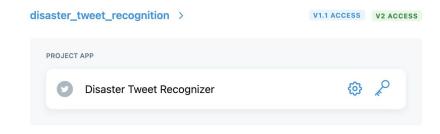
#### **Twitter API**

Historical tweet access

PhD student, research team, etc.

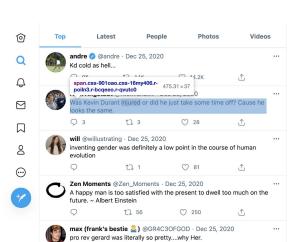




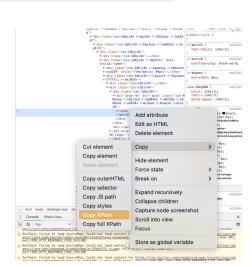


## **Twitter Scraping**

- Selenium-based
  - Browser window
  - XPaths
- Limitations
  - Maximum number of tweets/search
  - Same tweets every time



1



lang:en until:2020-12-26 since:2020-12-24 -filter:links -

# Twitter Scraping

Beirut: 100 tweets

Nashville: 83 tweets

Brunswick Co: 286 tweets

#### **Scraping API**

We scrape tweets on an "event" basis. This means that, given a real-world disaster, we can scrape tweets from around that time in the city itself and several other "control" cities.

#### **Output Format**

To get the output as a pandas dataframe, call nashville\_bombing.toPandas()

The table will look like

key	date	contents	city	city_in_disaster
0	2020-12- 25T15:41:14.000Z	Tweet (presumably) about the event	nashville	1
1	2020-12- 25T21:11:54.000Z	Probably not a disaster tweet	los angeles	0

#### **Performance Evaluation**

		Classifier	AUC	Accuracy	F1 Score
0	Logistic	Regression	0.777215	0.782563	0.746634
3		SVM	0.779095	0.796218	0.734247
5		Perceptron	0.754251	0.757878	0.723123
1	Ra	andom Forest	0.756242	0.778361	0.696839
2		AdaBoost	0.733616	0.745798	0.686528
4	Gradient Boosting	Classifier	0.720549	0.750525	0.630350

## **Next Steps: Looking toward Phase 2...**

- Have researched NLP models and specifically BERT in Phase 1
  - Continue to build knowledge base as project moves forward
- In process of configuring environment
  - Getting appropriate tools/libraries set-up
  - Start experimenting with different models