

## Assignment 3: Covid-19 Fact Checking

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#### **Datasets**

- Very dirty/noisy
  - Emojis
  - Twitter Hyperlinks -> Unique
  - Different Languages
  - Grammatical
    - Slang
    - Plurals vs singular, ex: case vs cases
  - Numbers
    - How do you determine if a number is relevant or correct
  - Stopwords
- These all affect the probability

## Regular vocab vs Filtered Vocab

- Sample Length:
  - Regular:

■ Yes: 3316

■ No: 1640

o Filtered:

■ Yes: 1113

■ No: 732

Takes out unique elements

 Leads to a better model as it removes any non unique word. Boost the conditional weight of all words with frequency >= 2

# Summary of Model Performances

	Accuracy	F1-SCORE		Recall		Precision	
MODEL		YES	NO	YES	NO	YES	NO
NB - Regular	0.67	0.77	0.44	0.91	0.32	0.67	0.70
NB - Filtered	0.75	0.82	0.59	0.94	0.46	0.72	0.83
LSTM (averages)	0.73	0.81	0.55	0.90	0.45	0.73	0.77

#### Extra: Sanitizer

#### Cleans strings of:

- emoji
- &amp
- url
- punctuation

#### Sample:

"day 6 of quarantine: my dad gave a talk to the cats about covid-19

https://t.co/mLdFXzcE2P"

"day 6 of quarantine my dad gave a talk to the cats about covid19"

(Only really works for english; accents and non latin/roman characters are lost)

### Performance

	Accuracy	F1-SCORE		Recall		Precision	
MODEL		YES	NO	YES	NO	YES	NO
Sanitized - Unfiltered	0.745	0.811	0.611	0.909	0.5	0.732	0.786
Unsanitized - Filtered	0.75	0.82	0.59	0.94	0.46	0.72	0.83

<sup>&</sup>quot;More than 2" Filter acts like a less complex sanitizer

