

#### Introduction

• On average, approximately 60,000 people globally died from natural disasters each year

• Low frequency, high impact events such as earthquakes cause significant but preventable loss of human life

- Why Twitter?
  - o real-time microblogging platform
  - 186 million daily active users as of 2020
  - o 150 million users worldwide, 36 million from US

# Why Bother?

- Total \$210 billion dollars in losses world-wide in 2020
  - Up 26% since 2019
  - C
- Preventable loss of human life
  - Last year's natural disasters claimed approximately 8,200 lives
- Fraudulent insurance claims
- Twitter is being used over 50 countries



# Objectives

Detect natural disaster tweets

Detect natural disaster images from attached images

## Data sources and Methodology

• Pre-labeled tweet dataset from Kaggle

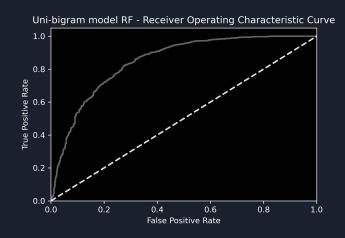
- Python
  - NLTK to tokenize and lemmatize tweets
  - Pandas for handling data in dataframes and Numpy for transforming data.
  - wordcloud
  - Sci-kit learn:
    - Bag-of-Words and TF-IDF to vectorize corpus
    - Supervised classifiers: Random Forest, Logistic Regression, Gradient Boost



### Results

- Best model based on accuracy
  - RandomForest classifier
  - ~78% accuracy
  - o AUC ~0.85

- Worst model
  - Unigram BoW with Gradient-Boost classifier
  - o ~65%
  - o AUC ~0.85



#### Limitations

• Class imbalance

- Bias introduced during pre-processing
  - o Suicide, war were excluded
  - o Locations and keywords were excluded

### **Future Directions**

• Include location and hashtags/keywords in modeling

• Improve accuracy of model (95% target)

• Word embeddings and deep learning (LSTM)

• Mobile app that tracks and sends email to subscriber

# Fin