Classifying Real and Fake Disaster Tweets

Final Capstone
By Malak Mosly

Problem Statement

- It is critical to respond to disasters (fires, shootings, etc.) in a timely manner.
- Social media sites such as Twitter are a very effective way to report disasters in real-time.
- The goal of this project is to create a model that can quickly identify a disaster tweet.

Who can benefit from this model?

- ❏ News outlets
- ❏ Disaster relief organizations
- ❏ Grassroots networks
- ❏ Politicians
- ❏ Law enforcement agencies
- ❏ Fire departments, hospitals, etc.
- ❏ The general public

Dataset (Kaggle Competitions)

- Consisted of 7,613 rows of tweets.
- Columns:
  - ID
  - Keyword
  - Location
  - Text
  - Target

Project Steps

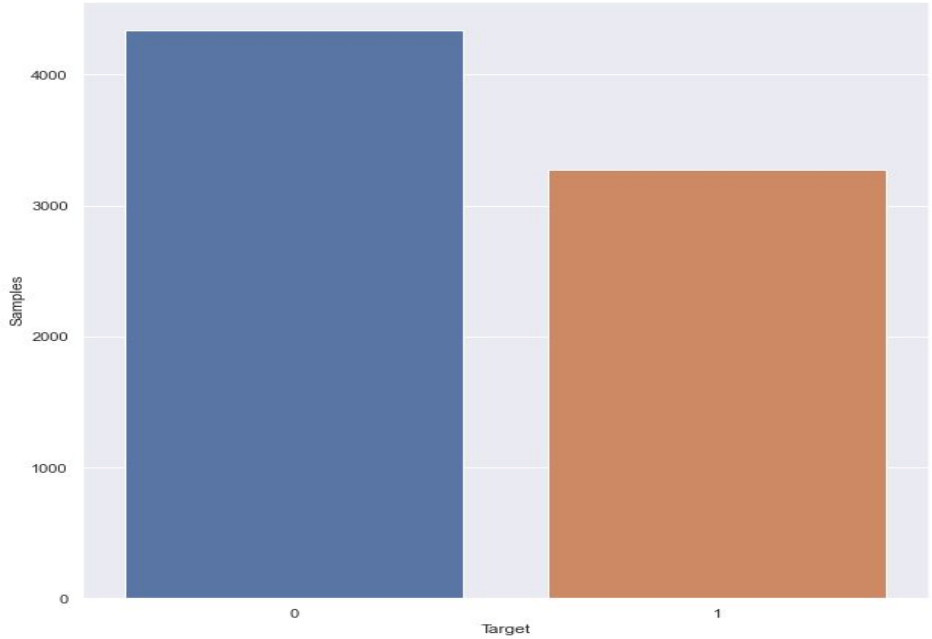
- ❏ Data wrangling and cleaning
- ❏ Exploratory data analysis
- ❏ Preprocessing and training
- ❏ Modeling

Data wrangling and cleaning

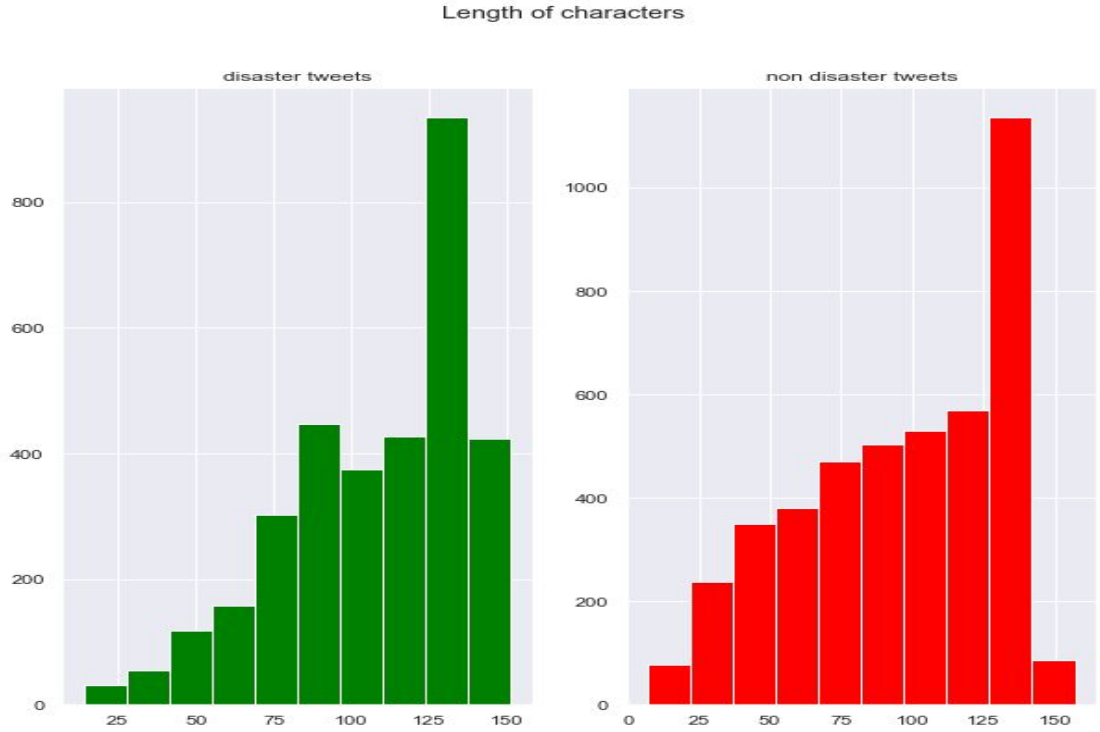
- Dataset was explored (shape, info, value counts)
- Text cleaning was the main focus of this step:
  - Punctuation removal
  - Emoji removal
  - Html link removal
  - Numbers removed
  - Square brackets removed
  - All text made lowercase
  - Stop word removal

Exploratory data analysis (I)

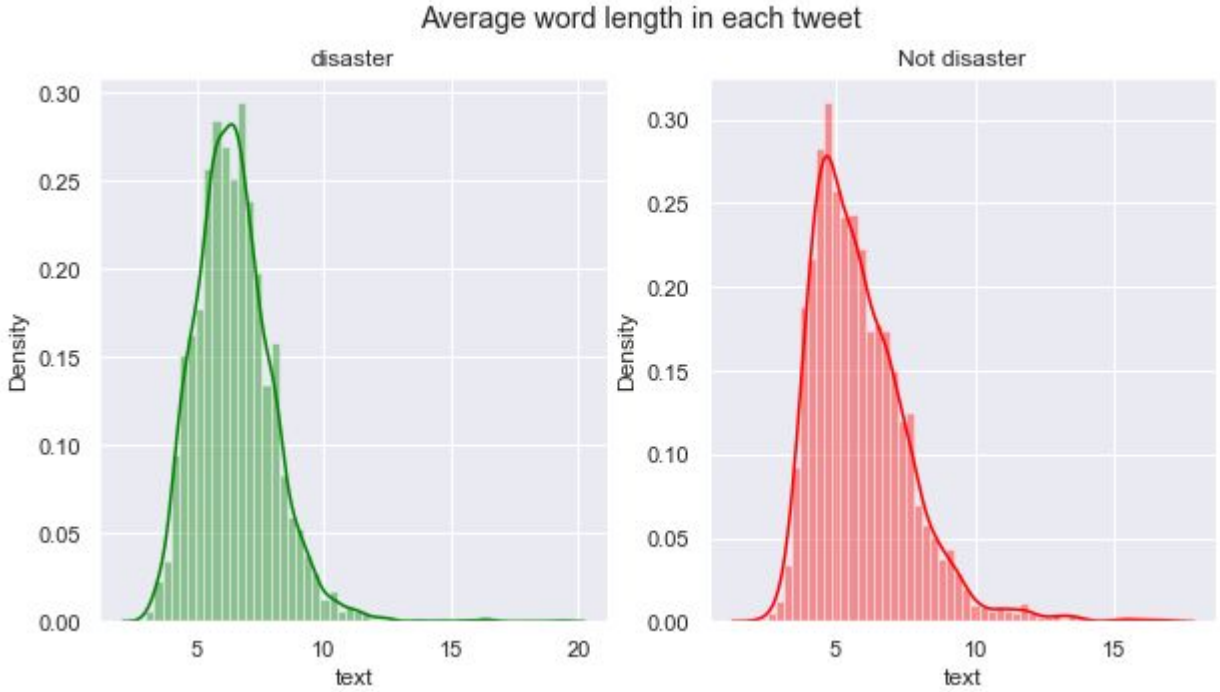
Distributions of target column



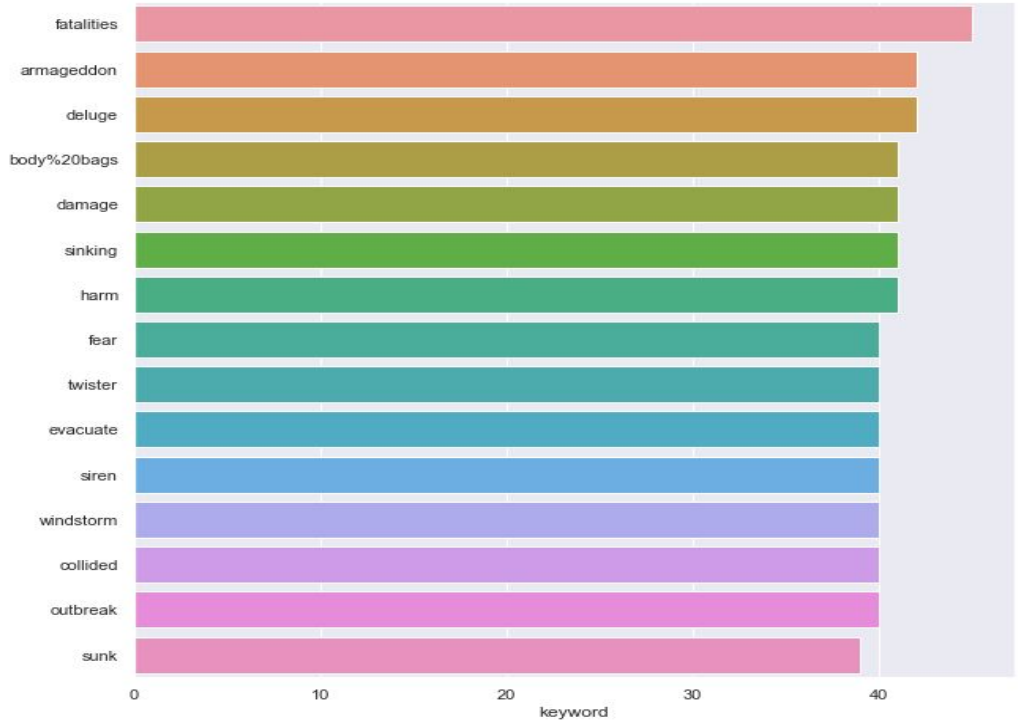
Exploratory data analysis (II)



Exploratory data analysis (III)



Exploratory data analysis (IV)



Preprocessing and training

- Defining X and y:
  - X → text column (predictor variable)
  - Y → target column (target variable, binary)
- Used CountVectorizer() on text column.
- Trian_test_split used to create training and testing datasets.
- TF-IDF transformation on text data.

Modeling (I)

- Three supervised models chosen:
  - Naive Bayes
  - Support-vector machines (SVM)
  - Logistic Regression
- Models were cross-validated and hyperparameter tuning done using GridSearchCV.

Modeling (II)

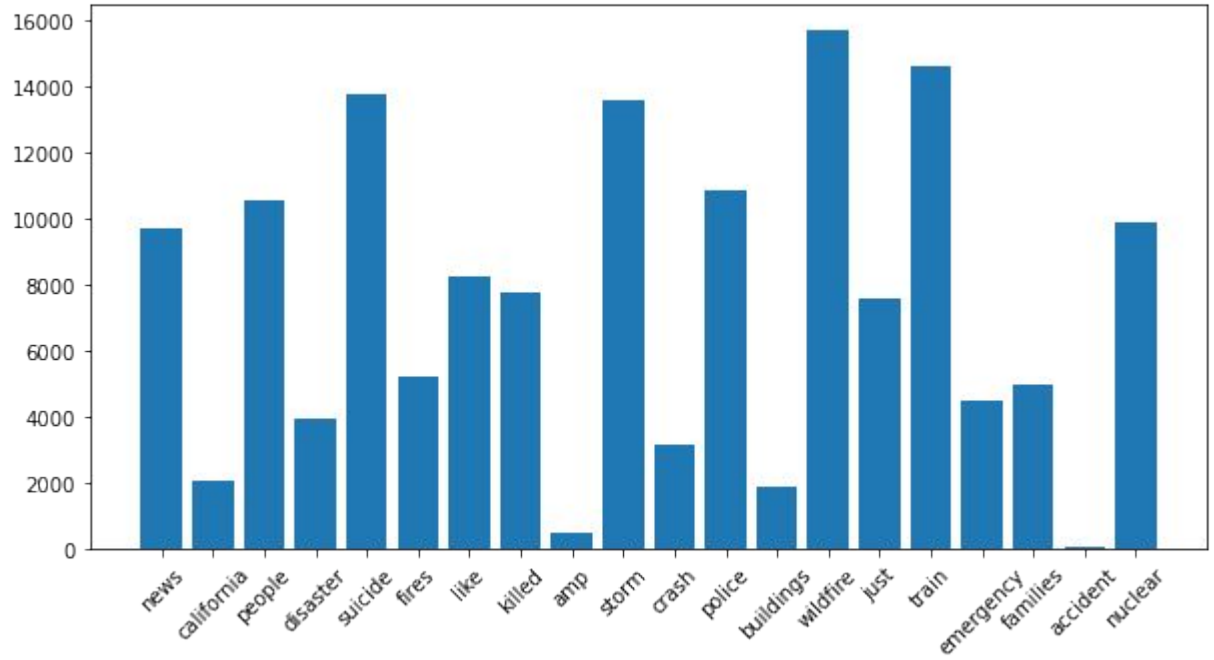
- Performance assessment metrics:
  - Classification report
  - Confusion matrix
  - Cross validation scores
  - Balanced accuracy scores
- Best performing model was Naive Bayes, with 80.22% testing accuracy.

Modeling (III)

Summary of performances:

Model	Testing Accuracy
Naive Bayes	80.22%
Logistic Regression	79.78%
SVM	78.95%

Most important features



Future research recommendations

- Expand model to note location and time of disaster
- Expand to all types of social media
- Model can go beyond fake/real disaster classification