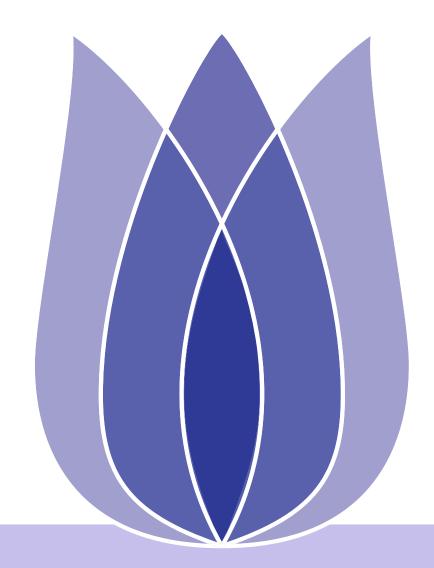
Disaster Tweets Prediction Using BERT

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- Twitter has become an important communication channel in times of emergency.
- But, it's not always clear whether a person's words are actually announcing a disaster.

LOOK AT THE SKY LAST NIGHT IT WAS ABLAZE!

aic (1)

- The author explicitly uses the word "ABLAZE" which is related to disaster.
- But it is an exaggerated expression.





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Given a set of labeled data which we will use to train a classifier and use it to predict whether a tweet is about disaster or not.

- The training set was collected from Twitter
- It has been labeled manually.
- A binary classification problem.
- F1 score is evaluation metric.





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Related Work - Text Classification

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- Existing Methods Rule-Based Methods
 - ◆ Rule-based methods classify text into different categories using a set of pre-defined rules.

Disadvantages

- Require a deep domain knowledge.
- Require a lot of manpower and time.
- When facing a new problem,
 previous rules may become useless.

Advantages

- ◆ Fast
- ◆ Easy
- ◆ Interpretable





Related Work - Text Classification

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■ Existing Methods - Traditional Machine Learning (Statistical methods)

Naïve Bayes, support vector machines, hidden Markov model, random forests...

Disadvantages

- Reliance on the handcrafted features.
- Cannot take full advantage of large training data because the features are pre-defined.

Advantages

 More accurate than rule-based methids.



Related Work - Text Classification

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- Existing Methods Deep Learning
 - Convolutional Neural Network, Long Short-Term Memory...

Disadvantages

- Reliance on large amount of training data.
- Weak Interpretability.

Advantages

- Capture deep contexual features.
- Greatly improved accuracy.





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- How to capture deep features?
 - Models should have the ability to capture deep features.
 - ◆ Generalization of models need to be improved.





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- Limited dataset.
 - Cannot train a model from scratch.
 - Although dataset is small, performance of model still needs to meet the requirements.





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Table 1: Data structure

Term	Example
id	210
keyword	airplain accident
location	Eagle Pass, Texas
text	A Cessna airplane accident in Mexico
label	1



Data Statistics

Prob	olem	Defin	iition

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Number of each label in training set.

Table 2: Number of each label

label	Number
1	3721
0	3892
total	7613





Data Statistics

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- Character length of text in training set.
 - max: 157 min: 7

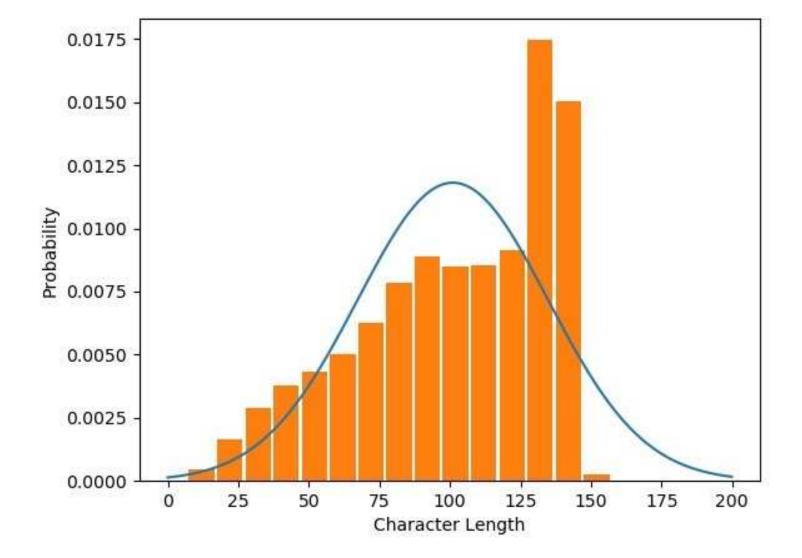


Figure 1: Distribution of Character Length





Data Statistics

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- Token length of text in training set.
 - max: 84 min: 3

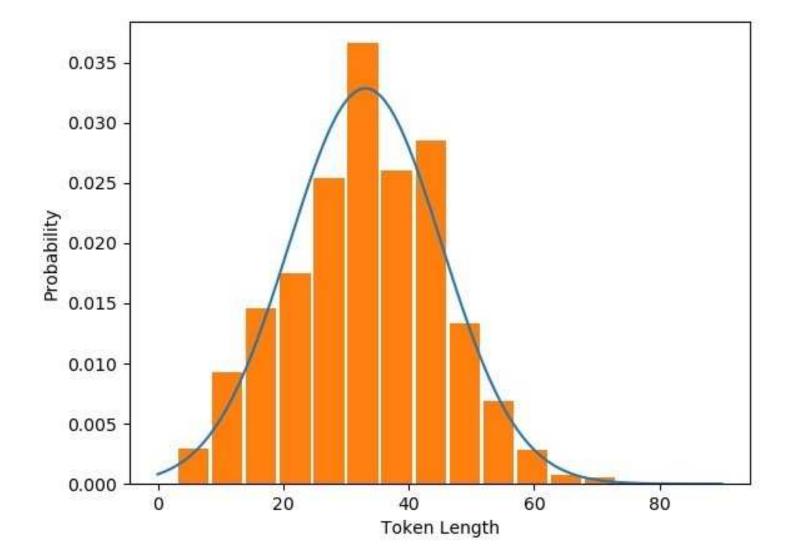


Figure 2: Distribution of Token Length



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Input Format

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- Original text:
 - Three people died from the heat wave so far.
- Input format:
 - ♦ token vector:

```
[ 101, 2093, 2111, 2351, 2013, 1996, 3684, 4400, 2061, 2521, 102, 0, 0, 0, ... ]
```

• mask vector:

$$[1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 0, \dots]$$

• segment vector:





Model Detils

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Table 3: Model details

Model	Layer	Hidden	Attention	Mask	Do lower
bert-base-cased	12	768	12	Token	False
bert-base-uncased	12	768	12	Token	True
bert-large-cased	24	1024	16	Token	False
bert-large-uncased	24	1024	16	Token	True
bert-large-wwm-cased	24	1024	16	Span	False
bert-large-wwm-uncased	24	1024	16	Span	True



Model Detils

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Table 4: Training setup

Name	Value	
Token length	256	
Dropout rate	0.1	
Optimizer	Adam	
Learning rate	5e-5, 3e-5, 2e-5	
eta_1	0.9	
eta_2	0.999	
Train: Validation	8: 2	
Batch size	16	
Number of epochs	3	



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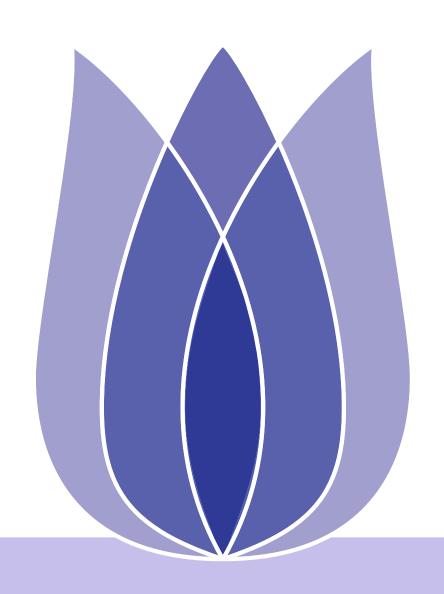
- $F_1 \ score = \frac{2 * precision * recall}{precision + recall}$
- Rank: 21/887 (2.3 %)

Table 5: Results

Model	$F_1 score$
bert-base-cased	0.825
bert-base-uncased	0.831
bert-large-cased	0.830
bert-large-uncased	0.848
bert-large-wwm-cased	0.828
bert-large-wwm-uncased	0.825



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