# Sentiment Classification on UGC data

### 1. Text Processing

Apply the following procedures on text data.

- All lower-case
- Remove url
- Expand short forms
- Remove punctuations
- Tokenization
- Remove numbers
- Remove stop words

## 2. Model Selection

## 2.1 Label Encoding

Encode all labels in train and test set with numbers.

Label	Encoded Label
Negative	0
Neutral	1
Positive	2

### 2.2 Feature Engineer

Use TF-IDF to create features, ser min\_df to 5. The model generates 7015 features. Reduce feature dimension using SVD to 100.

### 2.3 Modeling and Evaluation

Experiment with 6 models and the results are showed below.

	model_name	accuracy_score	precision_score	recall_score	f1_score
2	Random Forest	0.609339	0.597	0.59635	0.590595
3	AdaBoost	0.566426	0.551304	0.552421	0.545379
5	K Nearest Neighbor	0.54949	0.552702	0.544515	0.540771
1	Decsision Tree	0.479448	0.470986	0.471049	0.470777
4	Gaussian Naive Bayes	0.480434	0.507912	0.476016	0.463437
0	Dummy	0.349227	0.345812	0.346062	0.345819

The best model in terms of overall accuracy is random forest. Tune the hyperparameter using 5-f old-cross validation and refit the model using be st hyperparameters. The result is showed below.

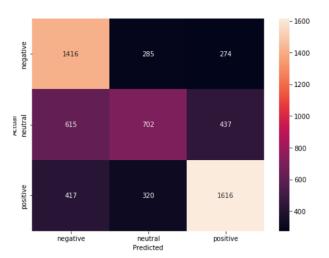
Metric	Score
Accuracy	0.6140
Precision	0.6033
Recall	0.6013
F1	0.5695

#### 3. Result Evaluation

As showed below, positive sentiments are best classified, neutral sentiments are worst classified. There are two potential reasons.

- Data Imbalance: the number of positive and negative instances is much larger than the number of neutral instances. This can cause the model to favor the majority class and perform poorly on the minority class.
- Neutral sentiments often lack clear positive or negative cues, making them harder to classify.

Label	Precision	Recall	
Negative	0.5784	0.7169	
Neutral	0.5371	0.4002	
Positive	0.6945	0.6868	



### 4. Error Analysis

There are 6 types of errors.

## 4.1 Negative Classified as Neutral

	cleaned_text	text
35	itchy miserable	I'm itchy and miserable I
41	cant sleep tooth aching	cant sleep my tooth is aching .
48	started think citi really deep gon na survive turmoil gon na next aig	started to think that Citi is in really deep s & $^t$ . Are they gonna survive the turmoil or are they gonna be the next AIG ?
77	needs someone explain lambda calculus	needs someone to explain lambda calculus to him I :(
89	burning cash chrysler gm stop financial tsunami bailout means taking handout	Are YOU burning more cash \$ than Chrysler and GM ? Stop the financial tsunami . Where " bailout " means taking a handout !
90	insects infected spinach plant	insects have infected my spinach plant :(
95	history exam studying ugh	History exam studying ugh
137	unfortunate stimulus plan put place twice help gm back american people led inevitable	It's unfortunate that after the Stimulus plan was put in place twice to help GM on the back of the American people has led to the inevitable
160	recovering surgerywishing julesrenner	Recovering from surgery, wishing @julesrenner was here :(
212	naive bayes using em text classification really frustrating	Naive Bayes using EM for Text Classification . Really Frustrating

 Floppy spelling: For example, misspell "can't" as "cant". So the negative "not" is not identified.

- Should not remove some punctuations like
  "!", ':(', ':)' . "!" expresses excitement.
  ':(' expresses negative feelings and ':)'
  expresses positive feelings. Should replace
  them with their meaning in words.
- The **TF-IDF model** considers the frequency and importance of each word in a corpus, so if the overall context and usage of the words in the corpus do not indicate a negative sentiment, the model may not classify the sentence as negative. This explanation applies to all errors. Data balance is important.

### 4.2 Negative Classified as Positive



- Appearance of **positive words** in negative sentence: e.g. firmly believe, dear, cheering, glad
- Wrong label: "Lebron is a Beast, but I'm still cheering 4 the A..til the end." It looks positive to me.

### 4.3 Neutral Classified as Negative



 Using punctuations without space is a problem. After removing punctuations, multiple words are joined

#### 4.4 Neutral Classified as Positive



Appearance of **positive words** in neutral sentence: e.g. good, play

### 4.5 Positive Classified as Negative



- **Multiple emotions** within one text
- Should not remove **emoji**
- **Wrong label:** e.g. 165, 200 are hard to tell if they are positive or sarcasm.

#### 4.6 Positive Classified as Neutral



Should not remove emoji and exclamations.

### 5. Limitation

Did not experiment any deep learning models.

### **Referencec:**

 $\frac{https://towardsdatascience.com/multi-class-text-classification-with-scikit-learn-}{12f1e60e0a9f}$ 

https://medium.com/@robert.salgado/multicla ss-text-classification-from-start-to-finishf616a8642538

YouTube channel: NormalizedNerd