# SENTIMENT.

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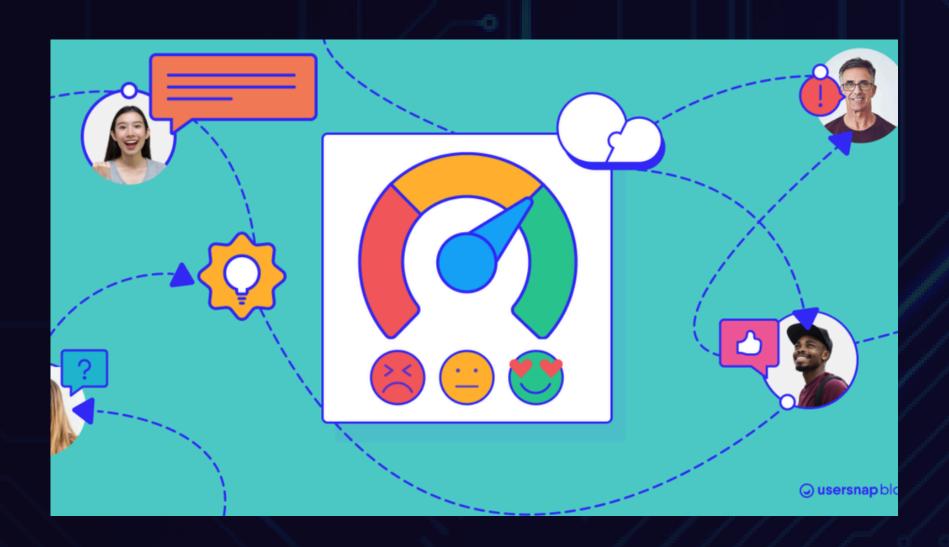
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# WHAT IS IT

Sentiment analysis or opinion mining is the use of natural language processing and text analysis that allows the user to take a text and feed it into an AI and through machine learning, output what the AI thinks is the most likely sentiment with the most commonly used sentiments being Positive and Negative.

Look for open-sourced models that have already been pre-trained with data as they have been trained with more comprehensive datasets

01

02

Make modifications to their script to be able to calculate for the confidence\_score and add it to the dataframe

03

Test run it against dataset Sentiment Analysis Dataset found on Kaggle for a rough gauge of the results(Actual cases were not included as a result the results of the model will be skewed)



Twitter-roBERTa-base for Sentiment Analysis

- Data trained on tweets
  - People are more free with emotions normally not as binary responses allowing for more neutral entries
- In comparison to TweetNLP, resulted in higher accuracy



**Hugging Face** 

Q Search models, datasets, users...

cardiffnlp/twitter-roberta-base-sentiment-latest <a>[</a>



01

Accuracy only reached 71% potentially due to the model being slightly out of date or due to the dataset including characters that is being processed incorrectly.

02

CSV output was correctly created through the model used accompanied with the confidence score.

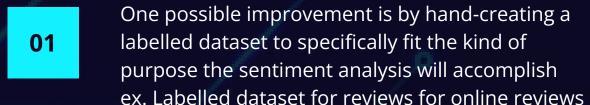
#### Accuracy: 71.00%

0.9
0.99
0.93
0.98
0.97
0.99



## POSSIBLE FUTURE

### **IMPROVEMENTS**



Better text processing to prevent any errors which may have occurred due to the way it was imported 02

A better way that I wanted to attempt was through 03 the use of polarity instead between positive and negative. It currently outputs the likelihood of a tweet being positive, negative and neutral. Removing neutral as a probability and allocating a range in polarity would be better to distinguish the emotion as to which it leans towards and allow for a more detailed analysis as to the level of intensity of the sentiment

The creation of a model that could better understand more complex texts such as multiple sentences should also be looked into

04

