

**Sentiment  
Discriminator**

TRIP ADVISOR

**Intensifier**

How to filter Malicious comments!

Neutralizer

AMAZON

Evaluation

Double

Negation

**NLTK**

Conjunction

filtering

YOUTUBE

python

Movie Reviews

CS372 Team Project

# Filtering Malicious Comments with NLTK

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# 01 | Introduction



“ lot of malicious comments  
→ Rating & Filtering comments with NLTK ”

# 01 | Introduction

## | Data

### 1. Collect data with ratings

- e.g. product reviews, book/movie reviews.
- 5 datasets:  
Amazon, Mendeley movie reviews,  
Trip advisor, Book reviews  
→ more than 1 million reviews

### 2. Youtube Crawling

- For evaluation

## | External Model & Data

### 1. vader\_lexicon.txt

- SentiWordnet's low accuracy  
(especially for comments)

### 2. autocorrect python package

- Lots of typos in comments
- Spell check each comment

## 02 | Method

### Mode

9 aspects to increase the performance of our results



#### Intensifiers

Increase score of words that follow intensifiers  
e.g. really, very



#### Neutralizers

Opposite of intensifier  
e.g. seldom, sparsely



#### Uppercase

If all letters are capitalized, amplify score  
e.g. I REALLY LIKE THIS



#### Threshold

Exclusion of words with a score below certain level (for noise removal)



#### Emphasis

Give an amplified score for the first and last sentences of the review



#### Conjunction

Put weight on sentences that start with a conjunction.



#### Exclamation

Magnify score if there are many exclamation marks  
e.g. love it!!!



#### Negative words

If a word follows a negative descriptor, reverse the sentiment



#### No/not phrases

Reverse phrases' sentiment if they start with 'not' or 'no'

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## 02 | Method

### | Scoring Algorithm

$(\text{Review Score}) = \sum (\text{Sentence Score}) * (\text{Sentence Sentiment Importance})$

$(\text{Sentence Score}) = \sum (\text{Word Score}) * (\text{Word Sentiment Modifiers})$

So, what we could discover... **IT PERFORMS WELL OVERALL!**

The main key was

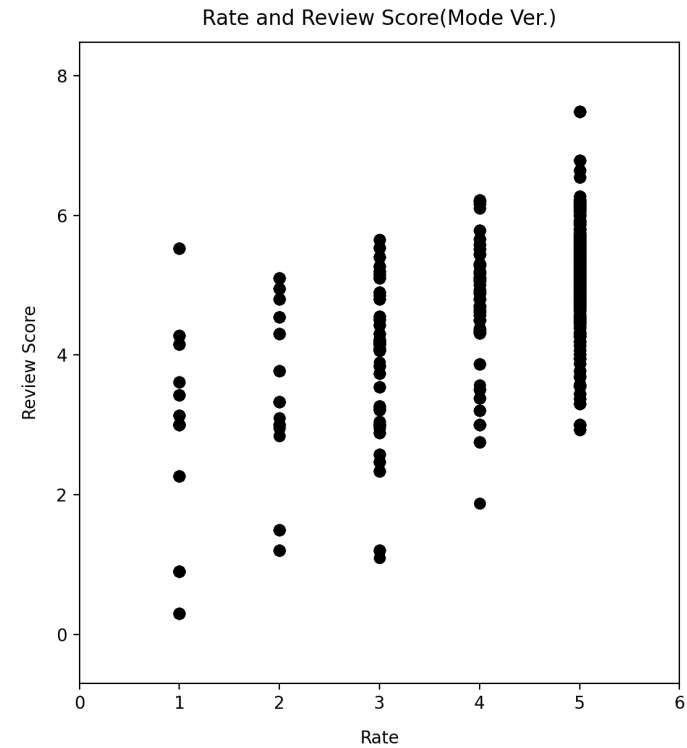
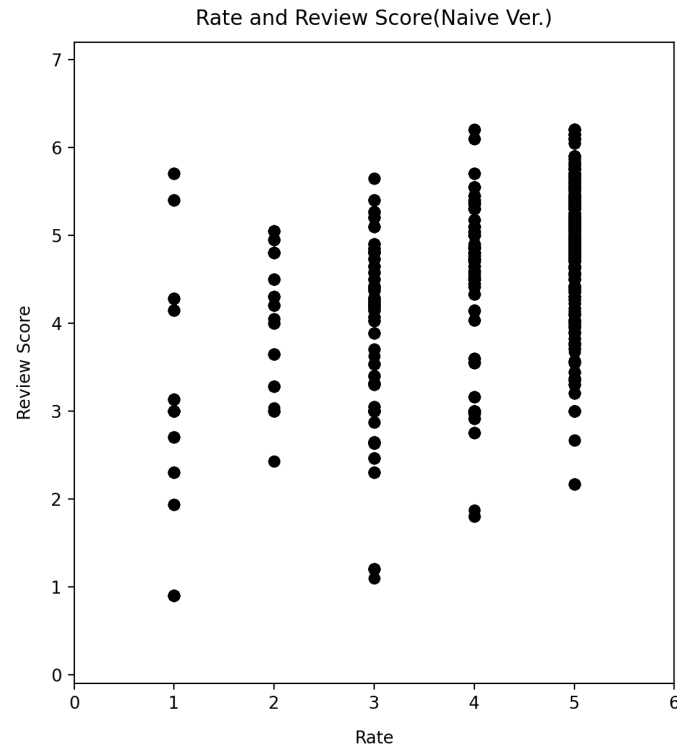
First, amplify the positive/negative sentiments extracted.

Second, reverse the sentiments according to their contexts.



## 03 | Result

### How do these modes affect the results?



## 03 | Result

### Results of individual reviews

These run large. I wear a 9.5 exactly in every shoe, but these I bought a 9, and the toe area is pretty wide plus plenty long even with socks. I love them but glad I was able to try them on before purchasing.

score\_naive: 4.80, score\_mode: 5.09, answer: 5.00

Absolutely love them. Can even wear them without my insoles.

score\_naive: 5.70, score\_mode: 6.98, answer: 5.00

Love these! I have 3 pairs...they're so comfortable - I'm on my feet all day & my feet actually don't hurt at the end of the day when I wear these :)  
I've learned to get these 1/2 size smaller....I'm normally a size 7 and the 6 1/2 fit perfectly

score\_naive: 4.13, score\_mode: 5.05, answer: 5.00

### F-score

- Amazon review dataset

True Positive: 0.877 / False Positive: 0.074 /

False Negative: 0.049 / **F-score: 0.934**

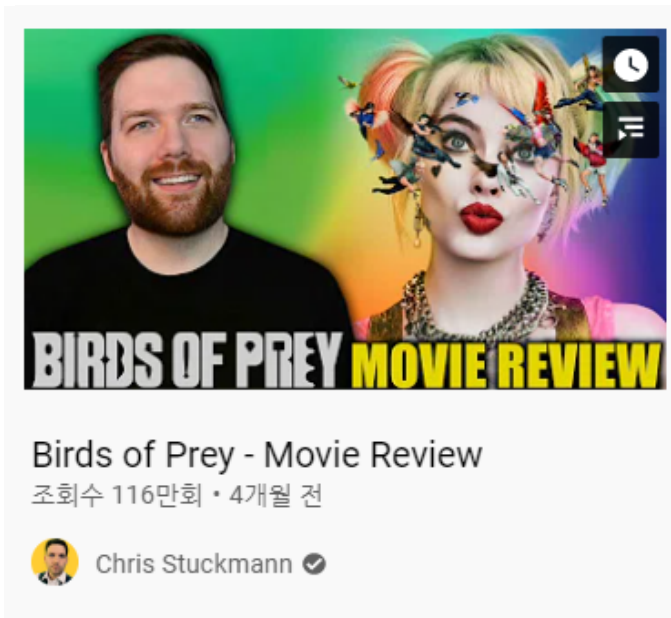
- TripAdvisor review dataset

True Positive: 0.717 / False Positive: 0.110 /

False Negative: 0.174 / **F-score: 0.835**

## 04 | Application

### Analyzing YouTube comments



```
david@Natrium ~/Documents/cs372/CS372-NLP master python main.py
##### FILTERED OUT 50 REVIEWS WITH THE LOWEST SCORE #####

Review Text: In Deadpool we side with a fucking killer!!!!!!!!!!!!
Score Mode: -1.62

Review Text: This movie is so fucking bad
Score Mode: -0.50

Review Text: Killed her... husband...whackedd emm!!!
Score Mode: -0.50

Review Text: KyDogg Batmanfan30 I would rather watch suicide squad 🦋🚫
Score Mode: -0.50

Review Text: movie was so bad
Score Mode: -0.50

Review Text: Why would they kill him off before he really even becomes black mask, even just killing him in general was very dumb
Score Mode: -0.44

Review Text: Saw this movie for the first time last night. What a misandristic piece of bullshit.
I fucking hate modern feminism...
Score Mode: -0.39

Review Text: Unfunny dead pool
Score Mode: -0.30

Review Text: If you took a shot Everytime you cringed watching this movie you would be dead from alcohol overdose
Score Mode: -0.30

Review Text: Deadpool 1 > Dead Pool 2 > Iron Man 3 > Sh*t > Birds of Prey
Score Mode: -0.30

Review Text: u mean the crossbow killer?
Score Mode: -0.30

Review Text: I hated this movie.
Score Mode: -0.20
```

## 05 | Further Improvement & Suggestion

### How to improve quality?

- **Using big data**

Core of the sentiment discriminator is the scoring of each words.

→ use big data to search for commonly used expressions → score better

- **Use different models depending on the type of comments**

i.e. different models for comments on movies, news, music videos, mukbang, etc.

- **Method for discrete rating**

Current implementation: infinite range → 1 to 5 discrete rating (like common reviews)

### Where to use this?

- Filtering malicious comments

- YouTube channels can obtain an overall numeric mood of comments

- ...

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**THANK YOU :)**