# Analyzing Sentiments of Fine Food Reviews on Amazon

Tarani Rupa Neelapu

COMPSCI 715 Programming for ML



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#### Introduction

#### What?

Analyze customer sentiments on Amazon fine food reviews using VADER and Logistic Regression

#### Why?

- To understand customer feedback for product improvements
- Performing sentiment analysis will help product owners understand customer satisfaction and dissatisfaction

#### How?

- Use **VADER** Sentiment Analysis.
- Use **Logistic Regression** Model.
- Compare the results by both models.

### Methodolgy



Step 1: Data Collection



Step 2: Data preprocessing



Step 3: Basic Exploratory Data Analysis



Step 4: Applying VADER Sentiment Aanalysis



Step 5: Applying Logistic Regression Model



Step 6: Evaluating Results

### VADER Sentiment Analysis

- VADER (Valence Aware Dictionary and sEntiment Reasoner) is a rule-based sentiment analysis tool specifically designed to analyze social media texts.
- It is a pre-trained sentiment analysis model that provides sentiment scores for given text
- VADER has 4 categories positive, negative, neutral and compound
- Compound is a calculation of positive, negative and neutral score ranging between -1 to 1
- VADER is easily available as a part of the NLTK (Natural Language Toolkit) library in python.

### Results by VADER Sentiment Analysis

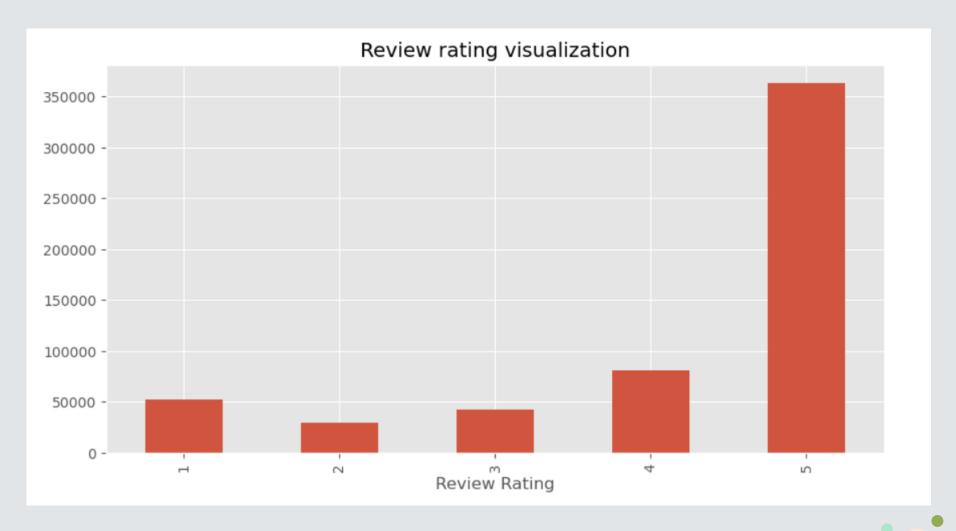


Figure 1: Exploratory Data Analysis

### Results by VADER Sentiment Analysis

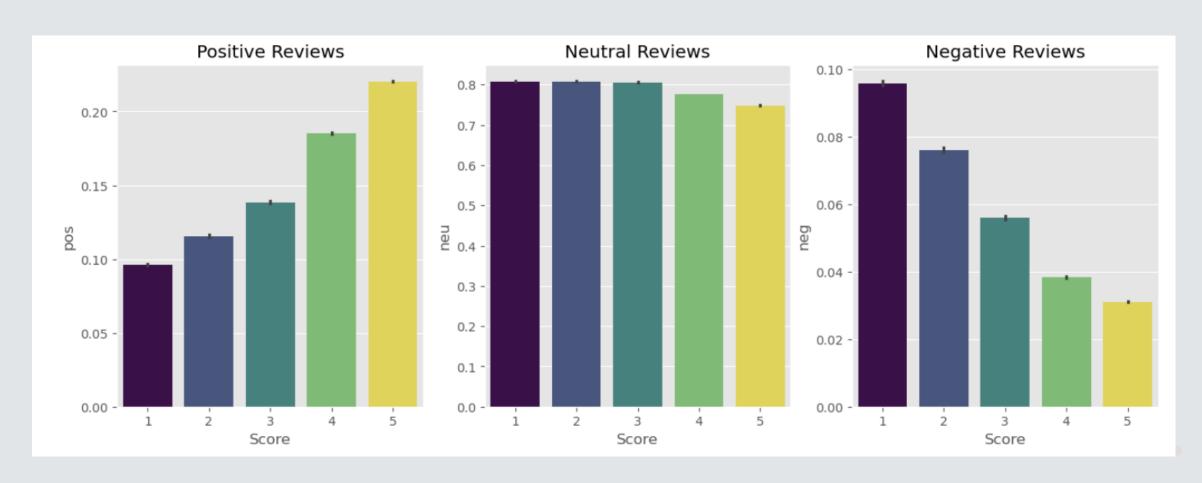


Figure 2: Customer Rating vs VADER Sentiment Score

### Results by VADER Sentiment Analysis

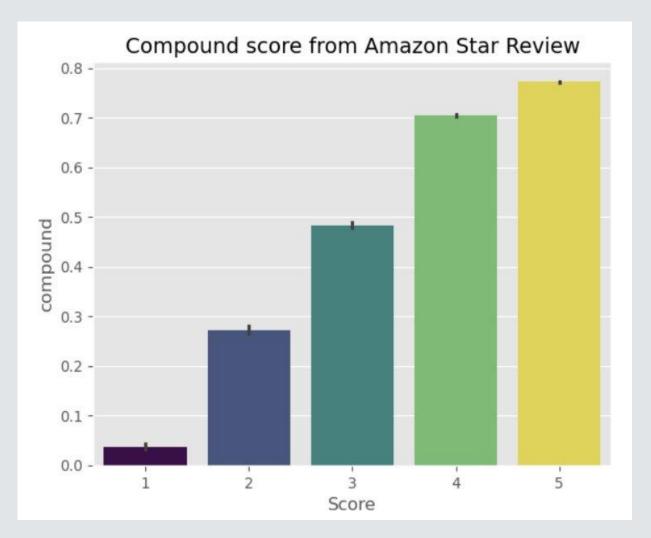


Figure 3: Customer Rating vs VADER Compund score

#### Results:

#### VADER Sentiment Analysis Accuracy:

Accuracy for VADER model: 0.7970

#### Logistic Regression Results:

Classification Report:					
	precision	recall	f1-score	support	
0.0	0.84	0.68	0.75	16407	
1.0	0.94	0.98	0.96	88756	
accuracy			0.93	105163	
macro avg	0.89	0.83	0.86	105163	
weighted avg	0.93	0.93	0.93	105163	
Confusion Matrix:					
[[11232 5175]					
[ 2209 86547]]					
Accuracy Score: 0.9297851906088643					

#### Results:

Top 10 I	Performing Products (Based on Sentiment):
	ProductId sentiment \
156	B0036VM05I 1.0
35633	B0036UQTQA 1.0
36332	B0036UWPJ0 1.0
105943	B0036VE3V2 1.0
228469	B0036V7K4E 1.0
229039	B0036VFX5M 1.0
309833	B0036UV8RU 1.0
487525	B001XWR35Y 1.0
539149	B0036VFTMO 1.0
544019	B0036UUY9I 1.0
	Summary
156	Great Deal
35633	New label, same great stuff
36332	Works really well
105943	My Dog Thinks They Are The Best!
228469	You can't buy cheddar like this at the grocery
229039	Not for everyone but I LOVE it!
309833	I love the coffee, but:
487525	Best Risotto in the World
539149	Dog + cats LOVE it!
544019	YUMMMMY

```
Worst 10 Performing Products (Based on Sentiment):
         ProductId sentiment \
93743
                          0.0
        B00514BIQK
173990
        B005116LQU
                          0.0
210927
        B0025VF8TK
                          0.0
226725
        B0025U00DC
                          0.0
312697
        B004TMAAM8
                          0.0
369755
       B0015DEH8C
                          0.0
369936
       B0015DEH8W
                          0.0
396416
       B0025VP9I0
                          0.0
456925
        B005116KK2
                          0.0
525741 B005116LMO
                          0.0
                                                   Summary
93743
                           WARNING to SENSITIVE Celiacs!!
                                   Lowell Foods Dew Sugar
173990
210927
        Doesn't look like a salmon, doesn't taste like...
226725
                                     Terrible After Taste
312697
                                            Harm no Charm
369755
                             Do not order... stale candy!
369936
           Sent an open package containing only 28 pieces
396416
                        item not as shown or name ordered
456925
                                     Worst pork pate ever
525741
                                Disappointed with quality
```

#### Conclusion and Future work

Which model performed better and what could be the reason?

How can this work be used in future for better sentiment analysis?

## Thank you