

## group 22

## motivation

*Chapter One*

*My first glimpse of Rosemary Drive came just before dusk, when the last of the day's sunlight dyed the old stones a rose gold. It was vast and desolating, bay-ogee bay of Elizabethan windows, with one ornate twisted tower pointing near its opposite end. Everything I knew about my mother shrank in that instant. I recognized it, of course, from her fantastical paintings; I recognized the*

*kindle*

the dataset we used is a public-available set with about 1600 entries from reviews – from the amazon-designed e-reader kindle and the amazon echo. because of the number of data we were were interested, to get a realistic overview.



github-repository  
+ dataset-link

## approach

main goal was to gain information about the language used in various forms of reviews, e.g. if there is a difference in positive or negative reviews.



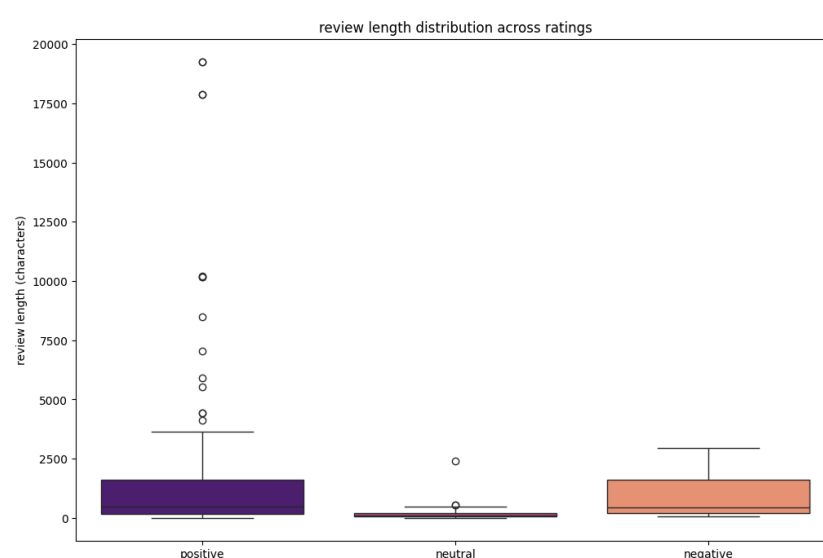
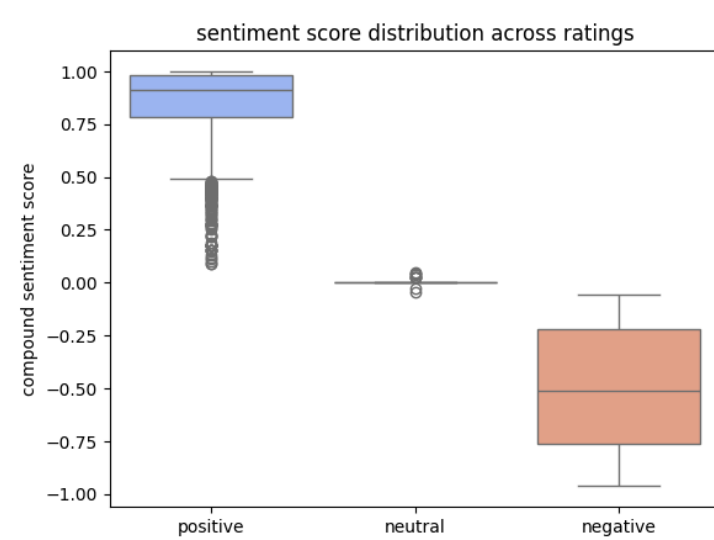
2

because of our interest in the dataset, we formed the research-question as following

***How do sentiment vary across positive, neutral, and negative reviews and what insights can be drawn from the sentiment analysis and the interaction between review helpfulness and product price?***

1

Sentiment	Count
positive	1421
negative sentiment	122
neutral	54

[illegible]

6



- a. majority of reviews (1,421) are positive, only 122 negative and 54 neutral reviews. suggestion: customers generally tend to review favorable experiences
- b. strong alignment between higher ratings and positive sentiments. however, negative ratings correspond with lower counts -> consistent correlation between ratings and sentiment
- c. scatterplot reveals that lower to mid-price range (approximately \$50-\$100) tend to receive more helpful reviews. high-priced products have fewer such reviews. indicates that perceived value may influence consumer engagement and helpfulness ratings.
- d. positive and negative reviews are generally longer than neutral reviews. suggests that dissatisfied customers tend to provide more detailed feedback, possibly to justify their praise or criticisms. especially positive reviews have some very long outliers.



afoma steline otikpa, akanksha mishra, florian oberbichler,  
hermann otto josef oratsch, matthias berger