









A brown dog, possibly a Weimaraner, is shown in profile, sitting in a garden. The dog has a white collar and is looking towards the right. The background is a blurred garden scene with green grass and some purple flowers.

# Central Garden & Pet Capstone Proposal

# Agenda

-  1. Problem Statement
-  2. Project Plan
-  3. Data Understanding
-  4. Objective/Scope
-  5. Metric/Success Criteria
-  6. Methodology
-  7. Demo
-  8. Way Forward

# Problem Statement

**Background:** The client, Central Garden & Pet (Sponsor), wants to develop a reproducible analytical procedure to utilize customer reviews to extract useful insights.

## Problems to solve:

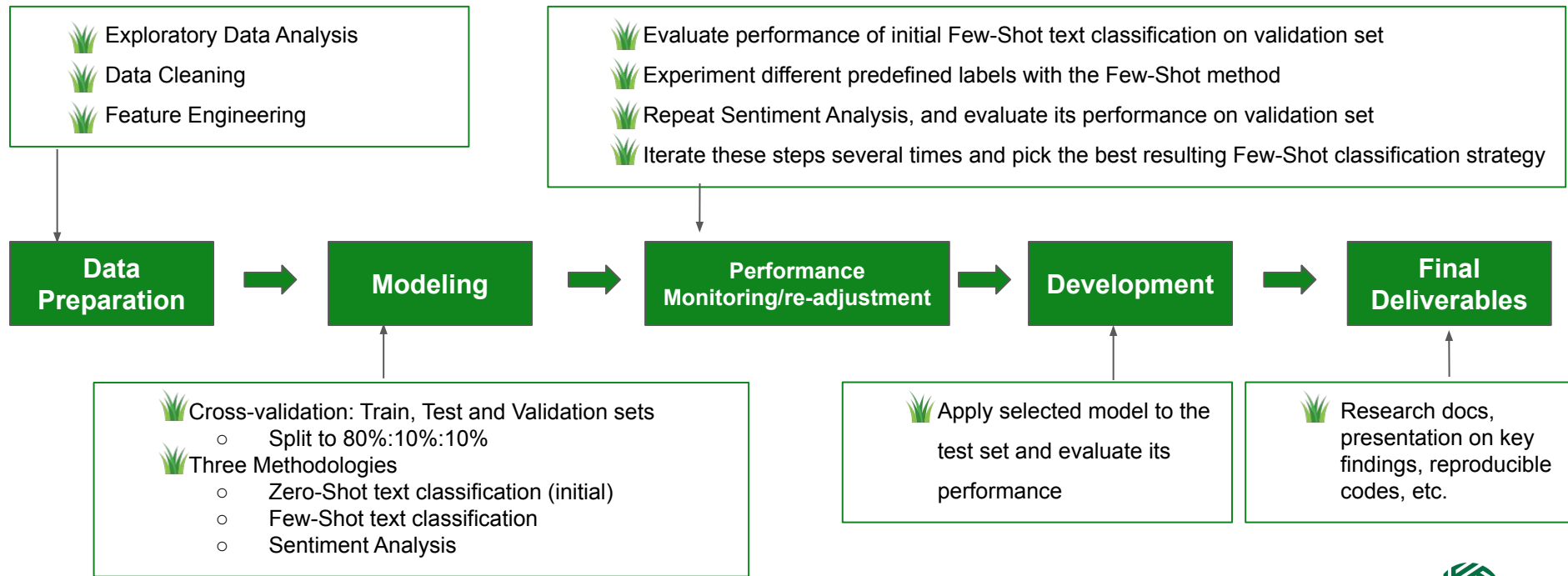


How to identify the overall sentiment of customers for insights and turn them into more efficient and targeted strategies



What factors/aspects drive those sentiments? - product, delivery, and so on.

# Project Plan



*(Iterate steps from EDA to Deployment for each brand and each retailer)*

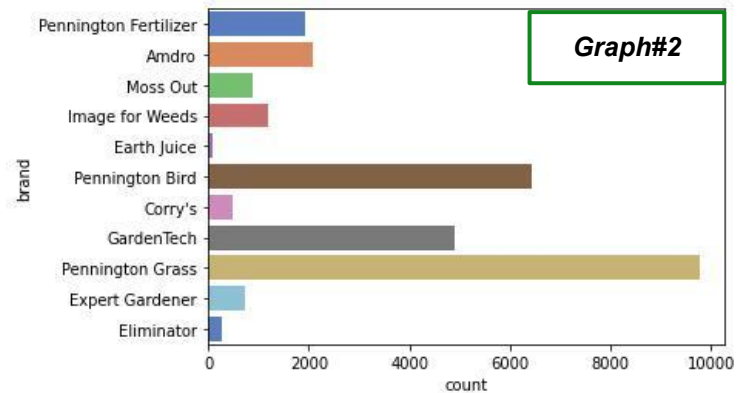
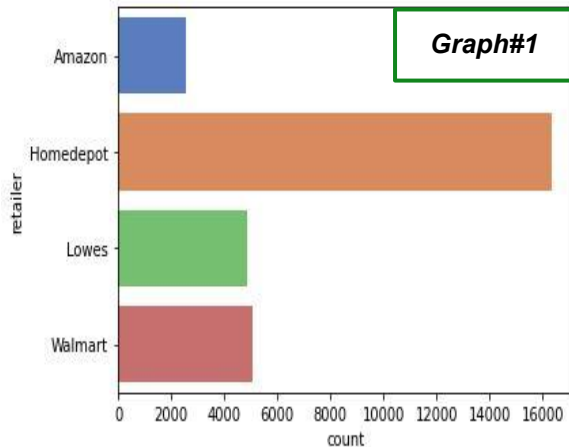
# Data Understanding

## Exploratory Data Analysis:

The review data was collected from **4 retailers** and **11 brands** from **2021-06-01** to **2022-06-30**  
Total of **28,786** reviews.

Home depot was the retailer with the most customer reviews. (See **Graph#1**)

Pennington Grass was the brand with the most customer reviews. (See **Graph#2**)





# Objective/Scope

## Overall Sentiment

Investigating the overall sentiment of customer satisfaction and identifying contributing factors for the sentiment

## Key Factors

Factors could including **price, delivery speed, convenience, product effectiveness, quality, unique selling propositions, marketing outreach**, etc

## Further Strategies

Investigating supporting insights for further strategies for **individual brands**

# Metrics/Success Criteria

## Benchmark Measurement Questions:

- Does **high price** link to negative reviews?
- Does **efficiency** of the product contribute to positive or negative reviews?
- Do products that customers find **easy to use** have positive reviews?
- Does the **effectiveness** of the products contribute to positive reviews?

**Intrinsic Metrics**- Focus on the performance of the models

- **Topic Coverage**
- **Prediction Error**

**Extrinsic Metrics**- Focus on the performance of the final outcomes

- **Goals Coverage**

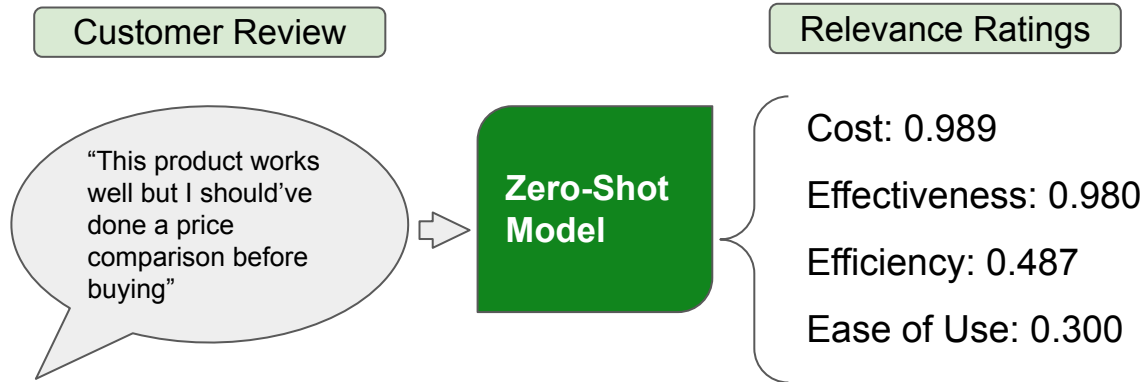
# Methodology: Zero-Shot Classification

## What is Zero-Shot Classification?

Developed by Facebook software engineers, the Zero-Shot Classification is a Machine Learning technique with human-like flexibility and efficiency for customer review analysis. The model can classify any text with given labels without any prior data.

## How does it work?

We created four basic category labels for classification - **cost, effectiveness, efficiency, and ease of use**. Each review will get a probability rating for each category depending on the semantics of the review.

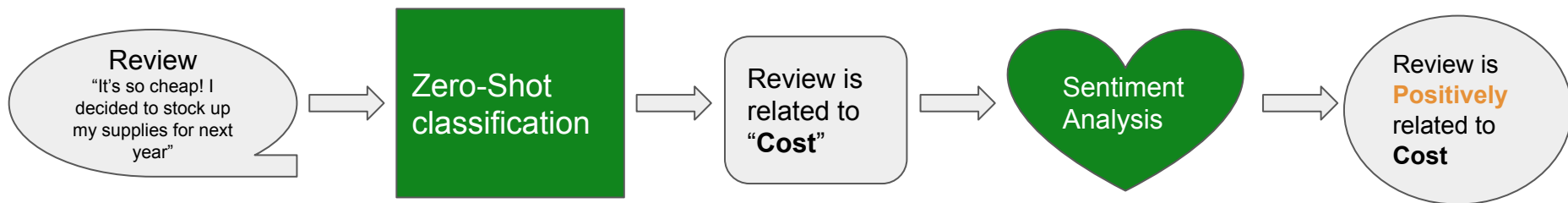




# Methodology: Sentiment Analysis

## Sentiment Analysis to determine the sentiments among the categories

Now that we have categorized the topics of each review under the predefined categories, we are interested in finding out the positive and negative sentiments associated with each topic.



Combining the Zero-Shot classification with Sentiment analysis we can intelligently label each of the reviews to be related to one or more of **price**, **delivery**, **effectiveness**, and/or **ease-of-use** with information on its sentiment.

	review_lower	cost_m	efficient_m	effective_m	ease of use_m	sentiment
this works great for keeping the brass from gr...	0.251741	0.998199	0.999644	0.376084	Positive	
waiting for it to work. will write more later	0.325730	0.027556	0.012378	0.095561	Positive	
[this review was collected as part of a promot...	0.531525	0.998460	0.999477	0.999256	Positive	
i received product reimbursement in exchange f...	0.716651	0.968388	0.996363	0.774749	Positive	

# Demo: Zero-Shot Classification

## Review 1:

grass grows fast and a lot of weeds. this works excellent in killing the weeds out. it was easy to use just hook it to a hose and start spray.

cost_m	efficient_m	effective_m	ease of use_m
0.531525	0.99846	0.999477	0.999256

## Review 2:

'waiting for it to work. will write more later

cost_m	efficient_m	effective_m	ease of use_m
0.32573	0.027556	0.012378	0.095561

## Observation:

Zero shot classification does well in identifying an effect and labeling the observations correctly.

# Demo: Zero-Shot Classification (shortcomings)

## Review 3:

it killed a few bugs, but it rained the next day and we say some bugs remained alive.

cost_m	efficient_m	effective_m	ease of use_m
0.653789	0.245608	0.780158	0.02614

## Review 4:

great product - all the birds love it!

cost_m	efficient_m	effective_m	ease of use_m
0.039136	0.873667	0.993279	0.655259

## Observation:



Zero shot classification model sometimes gets confused, especially if the reviews are short.



Need to set a cutoff threshold.

# Demo: Zero-Shot Classification (shortcomings)

## Review 3:

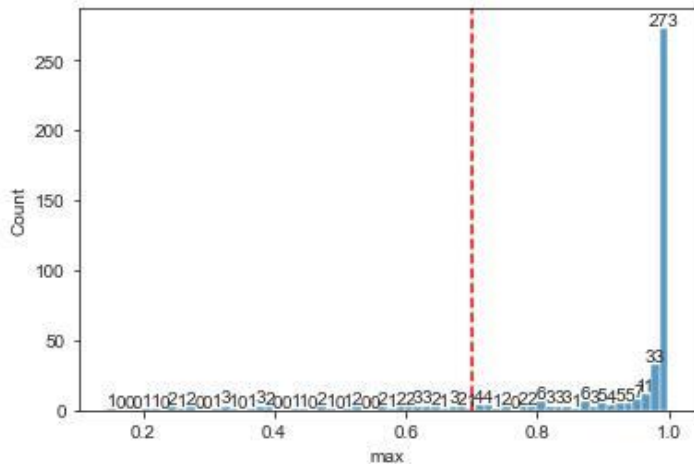
it killed a few bugs, but it rained the next day and we say some bugs remained alive.

cost_m	efficient_m	effective_m	ease of use_m
0.653789	0.245608	0.780158	0.02614

## Review 4:

great product - all the birds love it!

cost_m	efficient_m	effective_m	ease of use_m
0.039136	0.873667	0.993279	0.655259



## Result:

Zero-Shot classification resulted in 89% of the data being classified into the labels with higher than 0.7 probability level.




# Demo: Zero-Shot Classification (shortcomings)

## Review 5:

nice quality sunflower seed, the cardinals and squirrels love it.

cost_m	efficient_m	effective_m	ease of use_m
0.232189	0.891521	0.993074	0.289773

## Observation:

-  Zero shot classification mislabels some reviews.
-  Fine-tuning the classification model.
-  Setting a minimum word count.

# Sample Deliverables

## Homedepot

	Cost	Efficient	Effective	Ease of Use
Positive	34.38	84.38	76.02	87.91
Neutral	17.19	8.75	13.27	2.2
Negative	48.44	6.88	10.71	2.2

## Output:

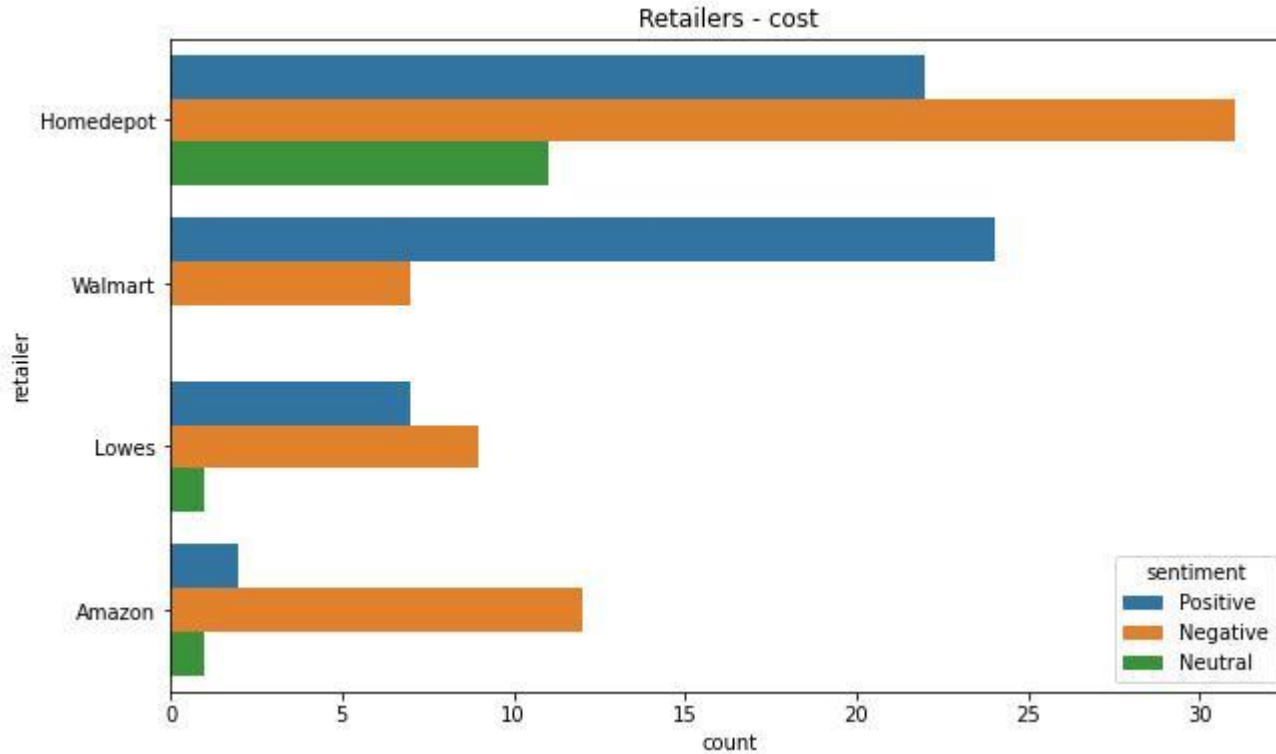
Dashboard view of the heat-map of the types of reviews and sentiment based on the retailers.

## Amazon

	Cost	Efficient	Effective	Ease of Use
Positive	13.33	50	45.45	62.5
Neutral	6.67	22.22	22.73	37.5
Negative	80	27.78	31.82	0



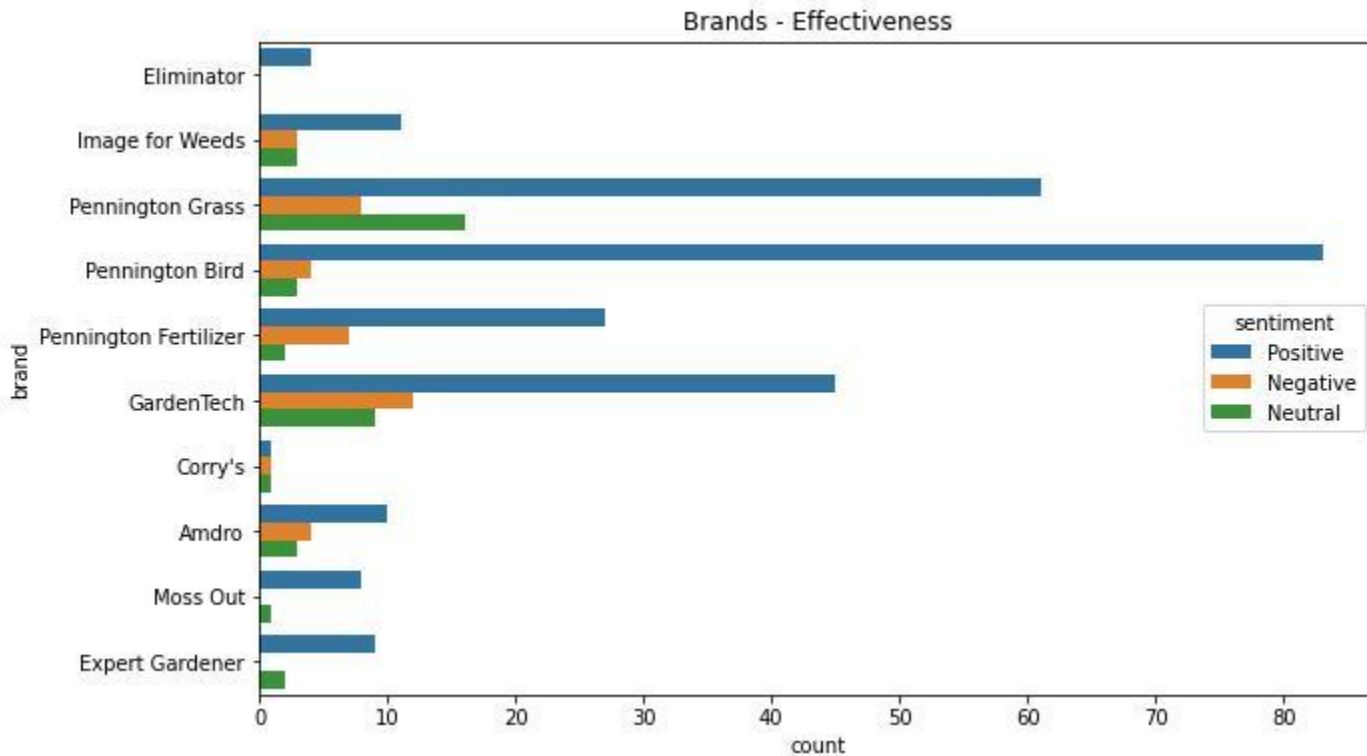
# Sample Deliverables



Output:

Distribution of the reviews for each **retailers** and their sentiments.

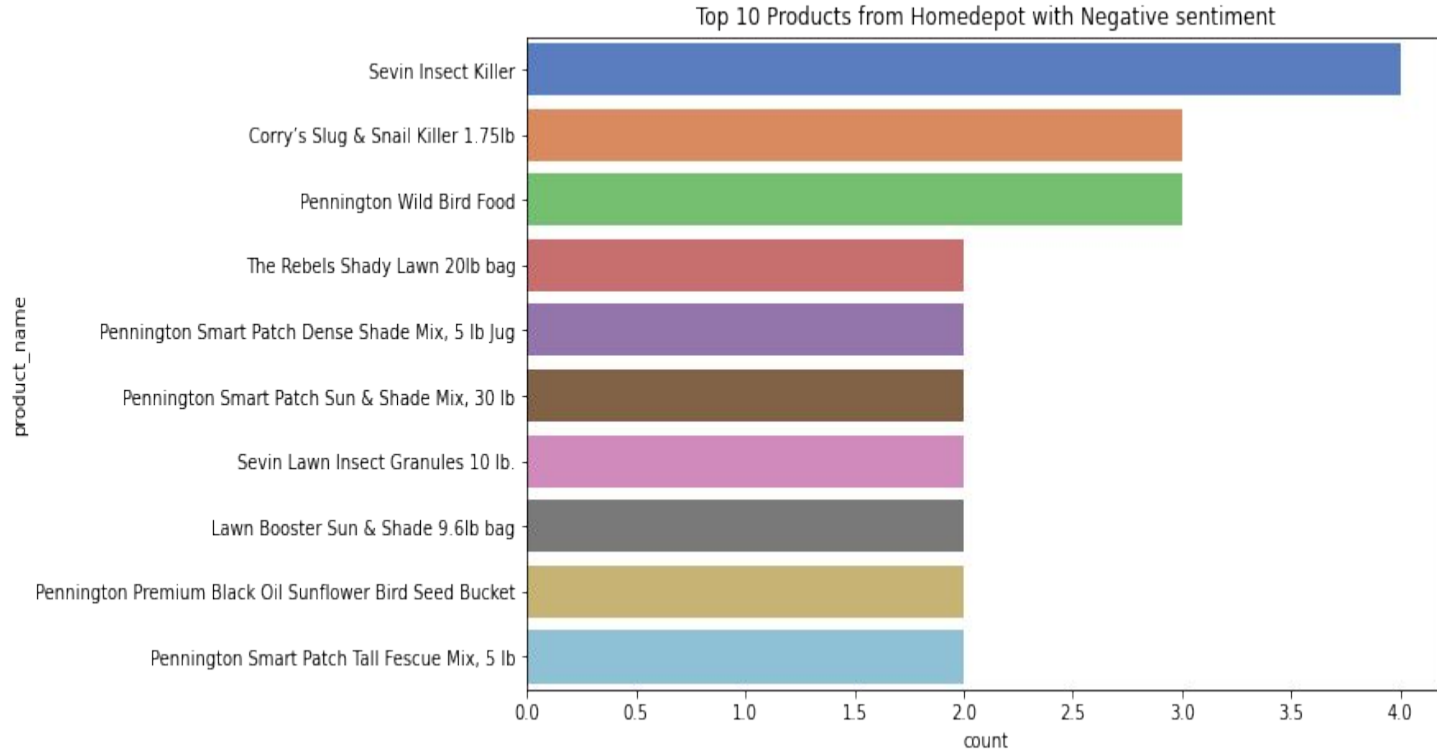
# Sample Deliverables



## Output:

Distribution of the reviews for each **brands** and their sentiments.

# Sample Deliverables



Output:

List of **products** with the most negative reviews.

# Way Forward



Fine tuning the zero shot model to improve classification.  
(currently 88% of the data were classified with higher than 0.7 confidence level)



Additional analysis to produce intelligent information and visualizations



Deep dive analysis on products with negative reviews to identify gaps.

# Appendix

# Heatmap of all the retailers

## Homedepot

	Cost	Efficient	Effective	Ease of Use
Positive	34.38	84.38	76.02	87.91
Neutral	17.19	8.75	13.27	2.2
Negative	48.44	6.88	10.71	2.2

## Lowes

	Cost	Efficient	Effective	Ease of Use
Positive	41.18	81.25	74.07	87.88
Neutral	5.88	8.33	9.26	6.06
Negative	52.94	10.42	16.67	6.06

## Walmart

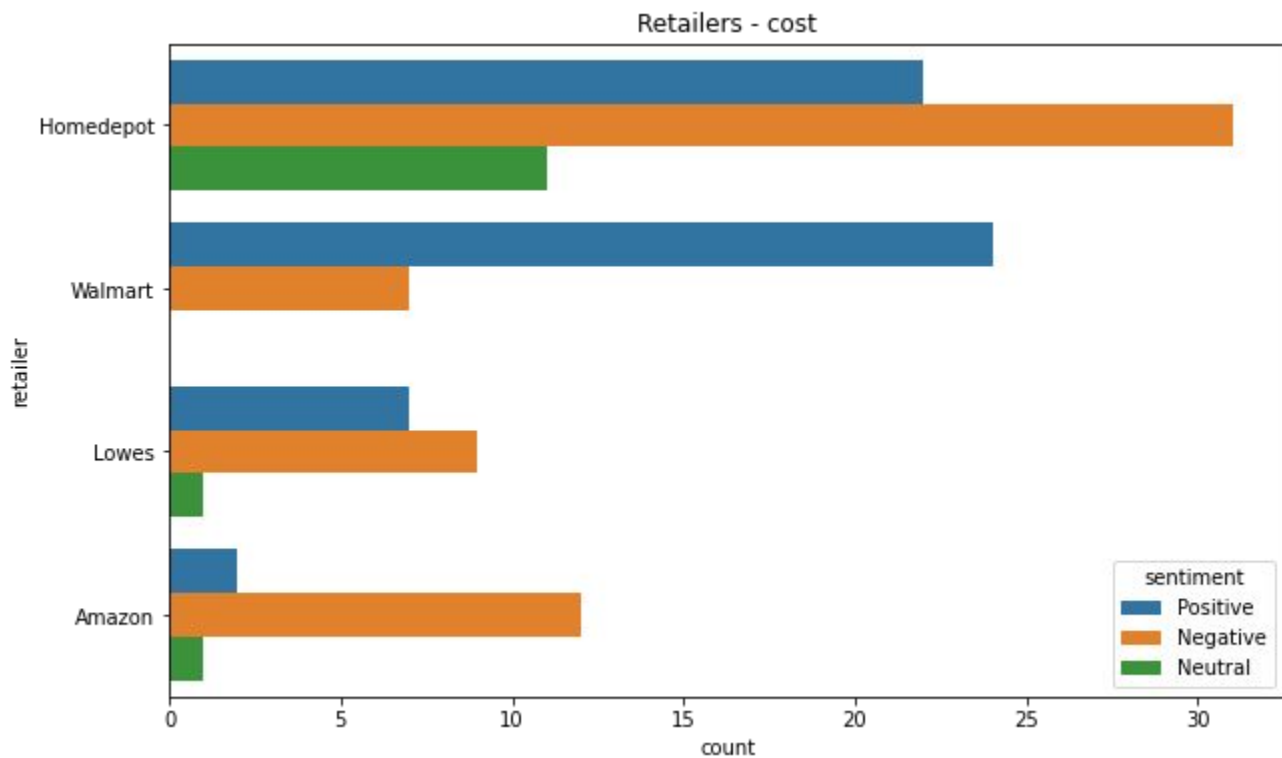
	Cost	Efficient	Effective	Ease of Use
Positive	77.42	93.1	90.91	96.88
Neutral	0	5.17	6.06	3.125
Negative	22.58	1.72	3.03	0

## Amazon

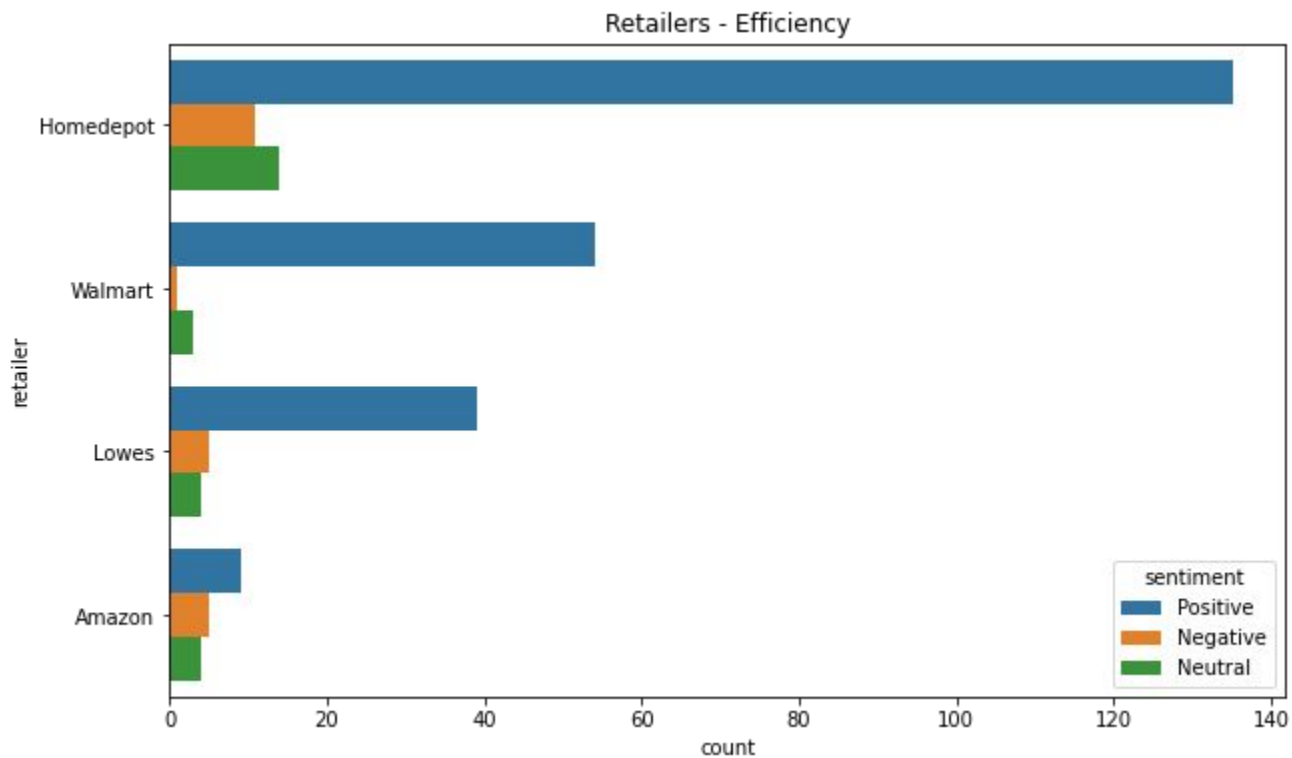
	Cost	Efficient	Effective	Ease of Use
Positive	13.33	50	45.45	62.5
Neutral	6.67	22.22	22.73	37.5
Negative	80	27.78	31.82	0



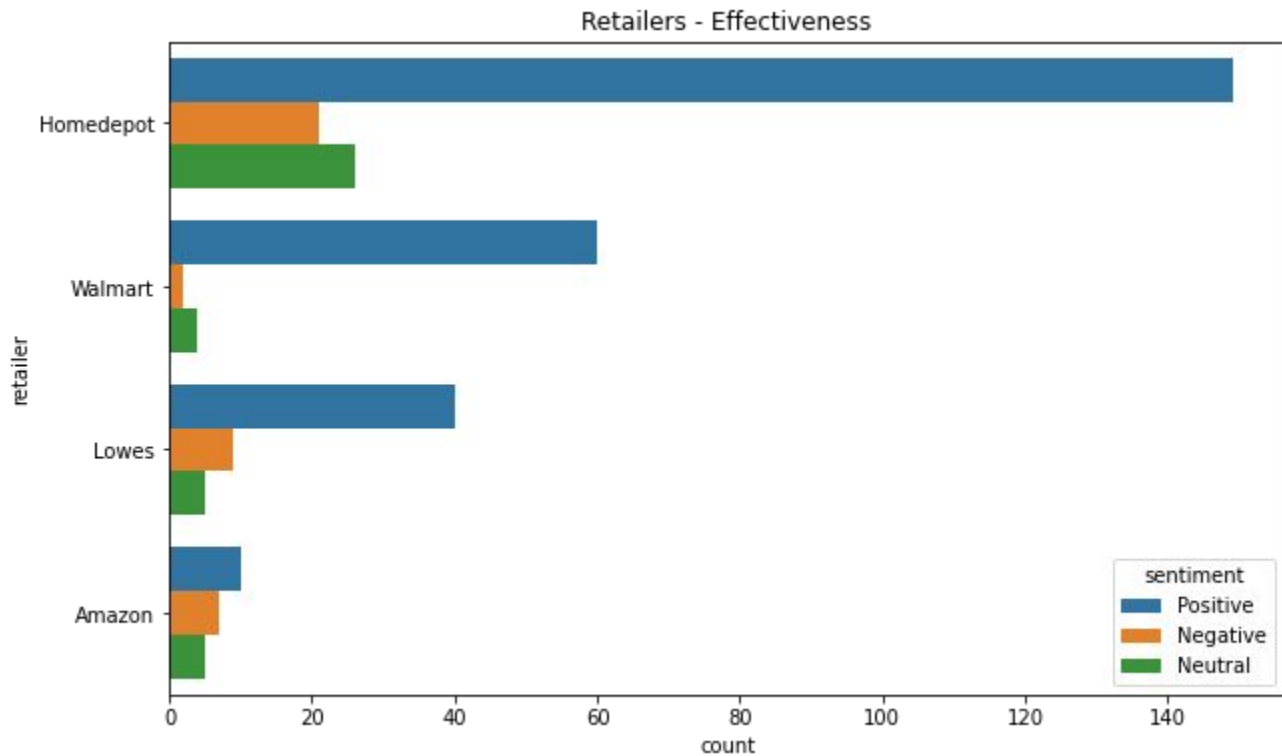
# Sentiment Histogram of the Retailers by labels



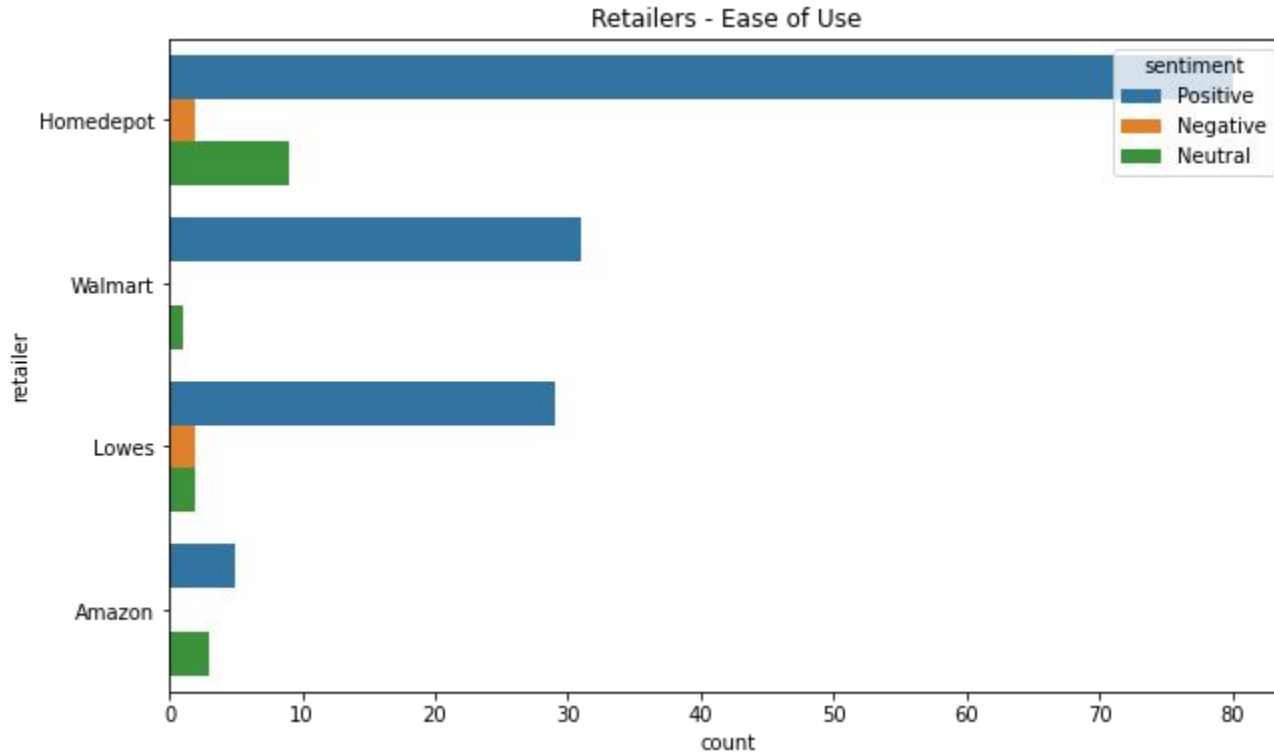
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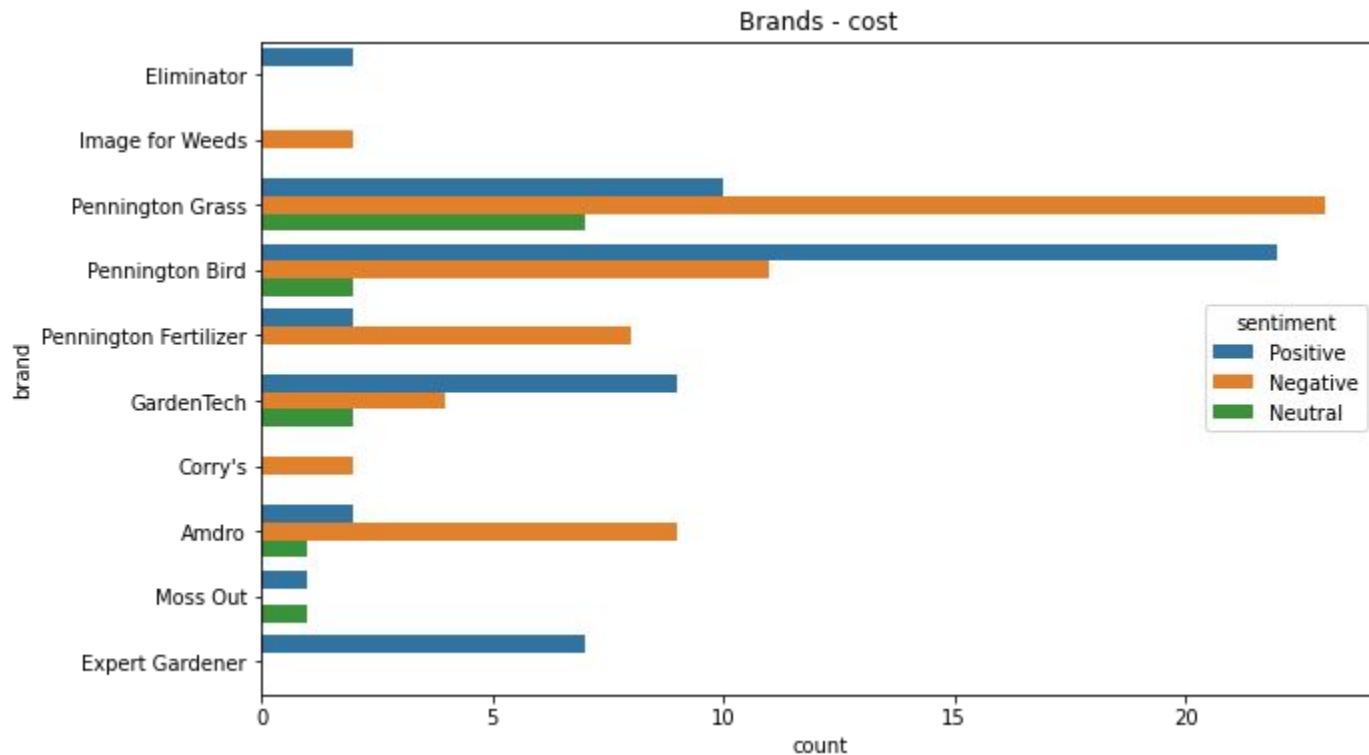
# Sentiment Histogram of the Retailers by labels



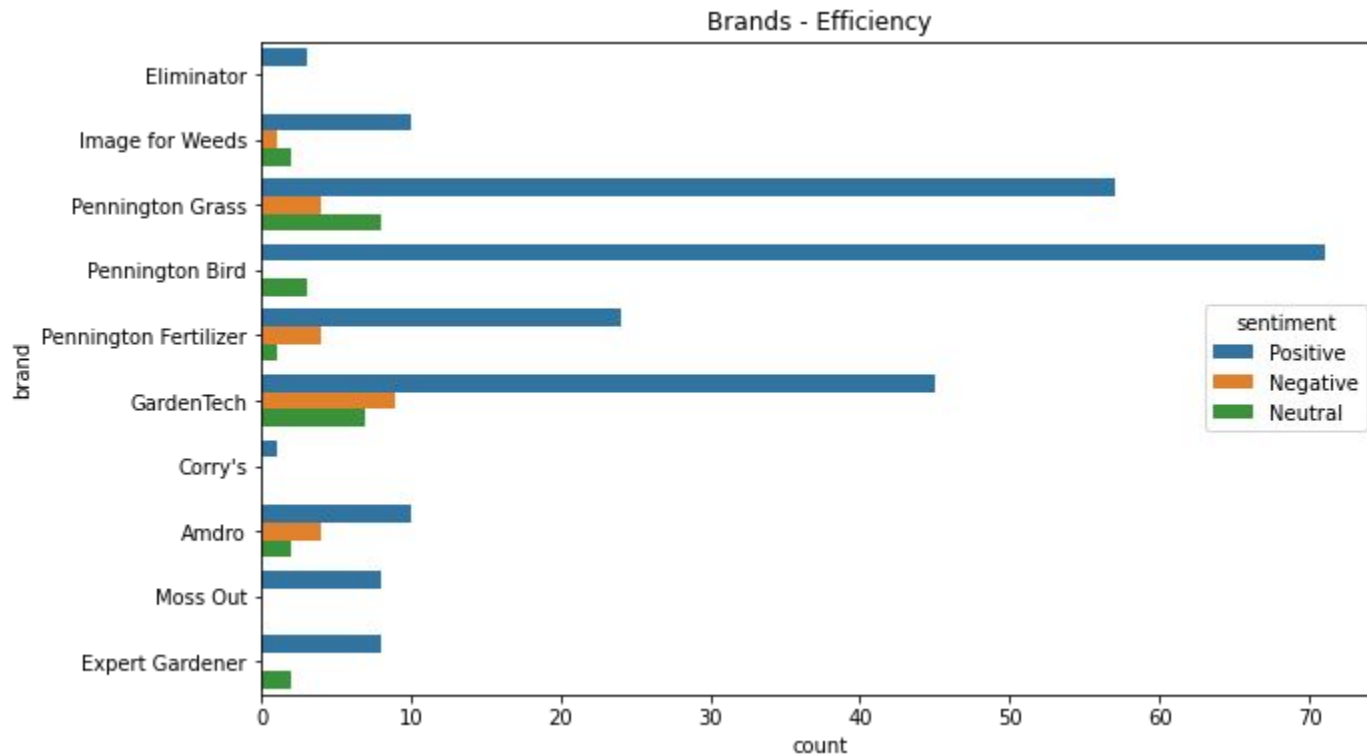
# Sentiment Histogram of the Retailers by labels



# Sentiment Histogram of the Brands by labels

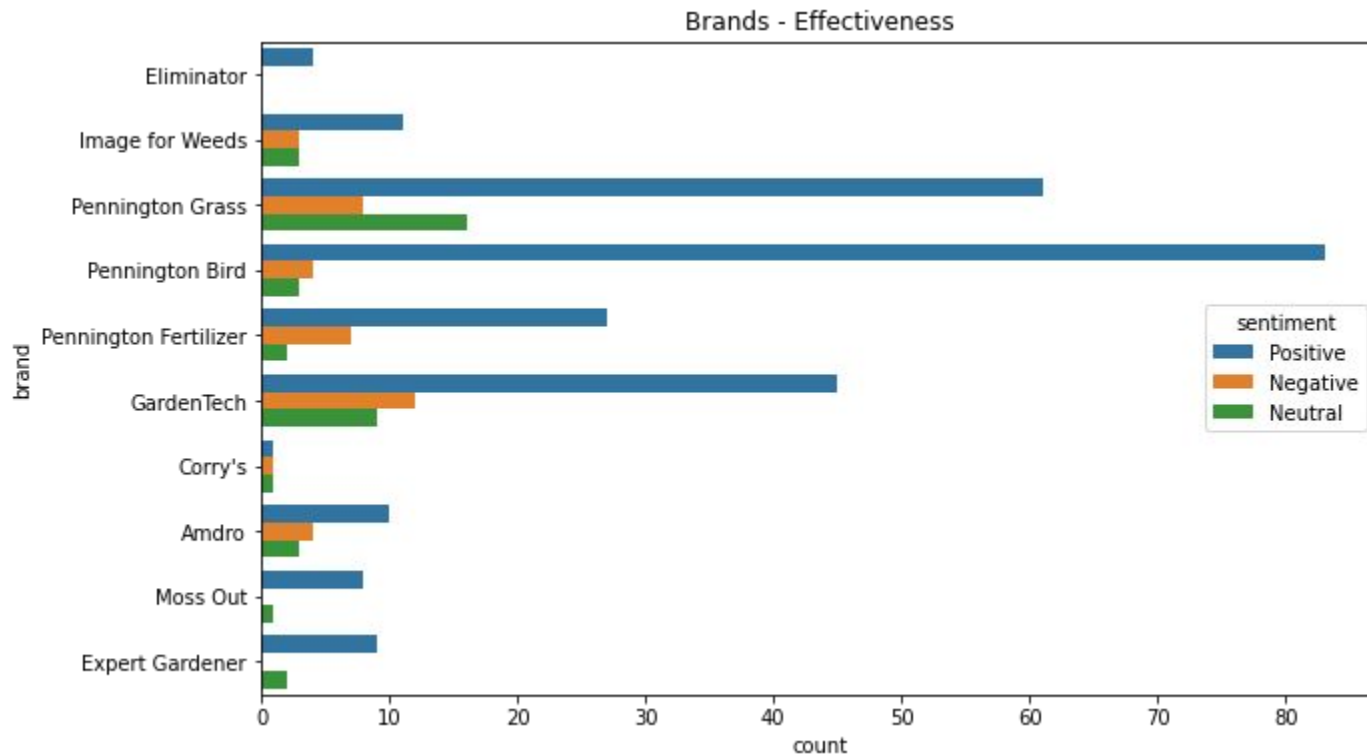


# Sentiment Histogram of the Brands by labels

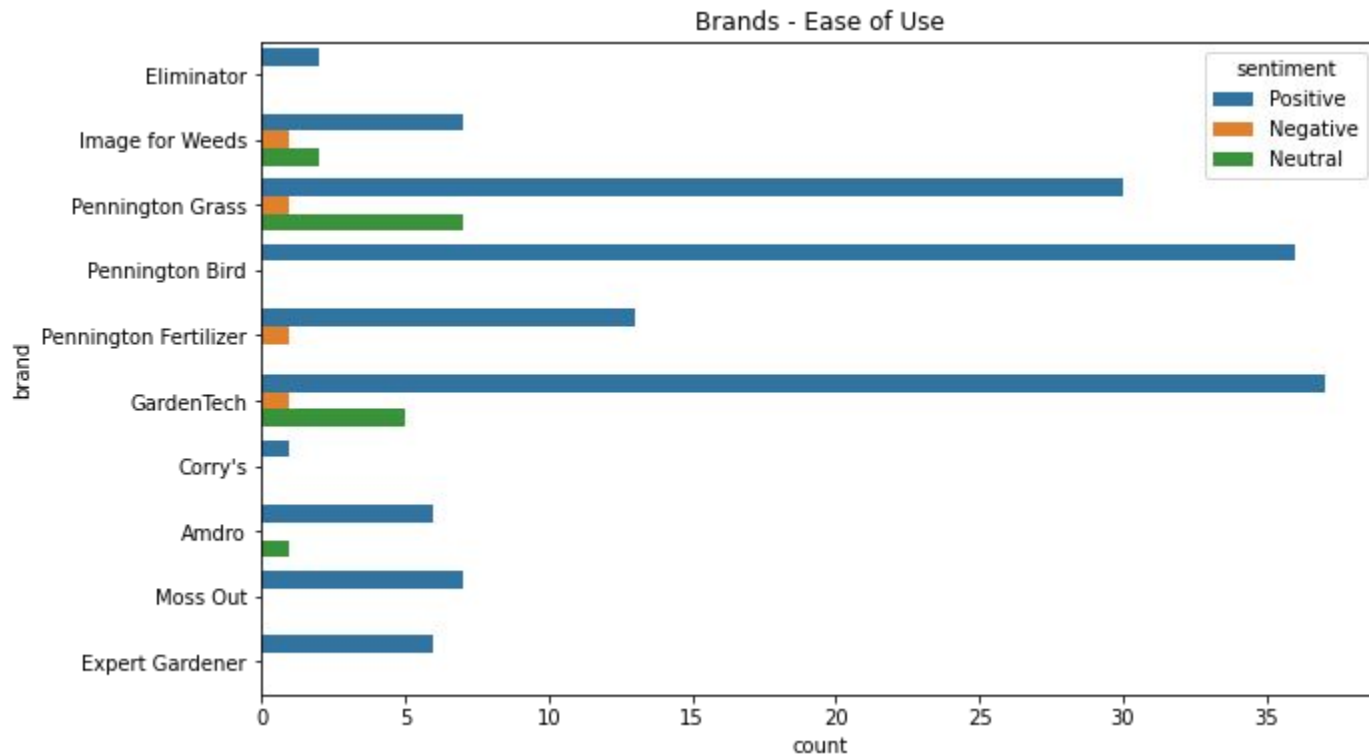




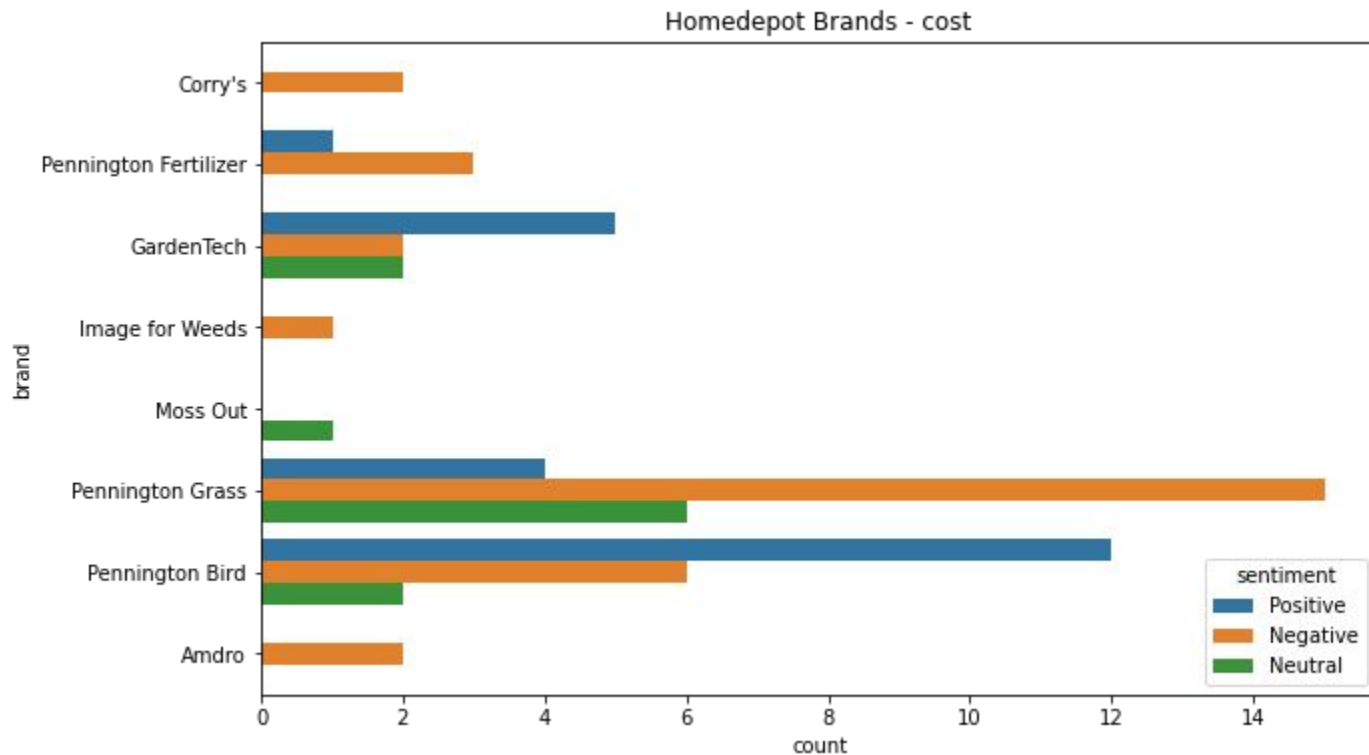
# Sentiment Histogram of the Brands by labels



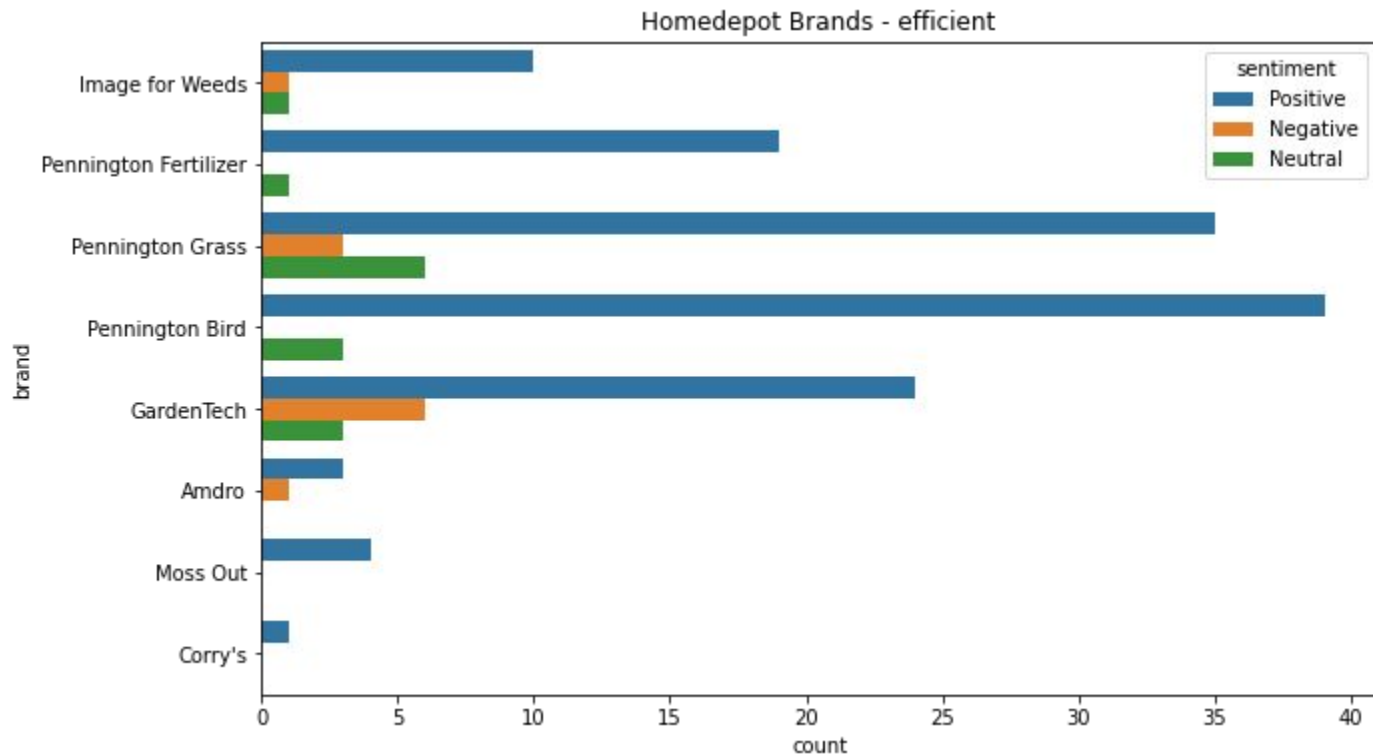
# Sentiment Histogram of the Brands by labels



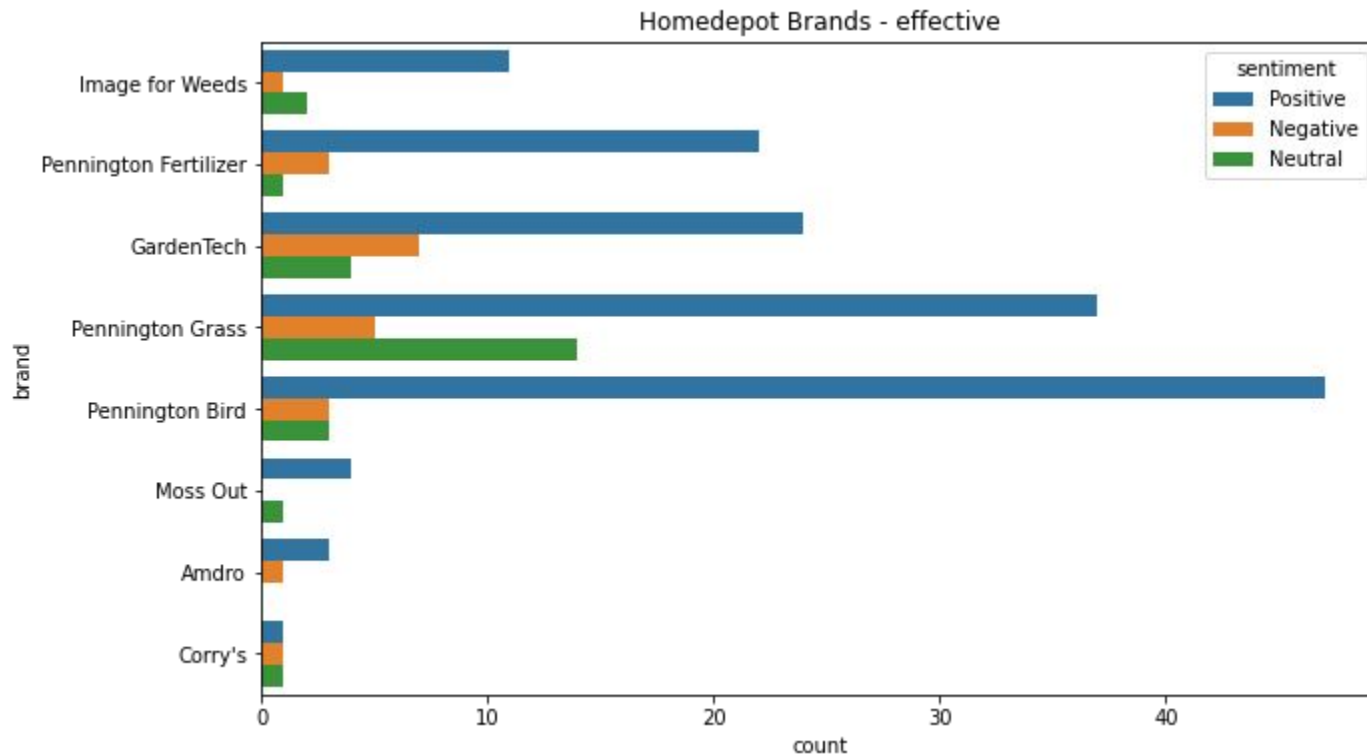
# Sentiment Histogram of Homedepot by Brands - Cost



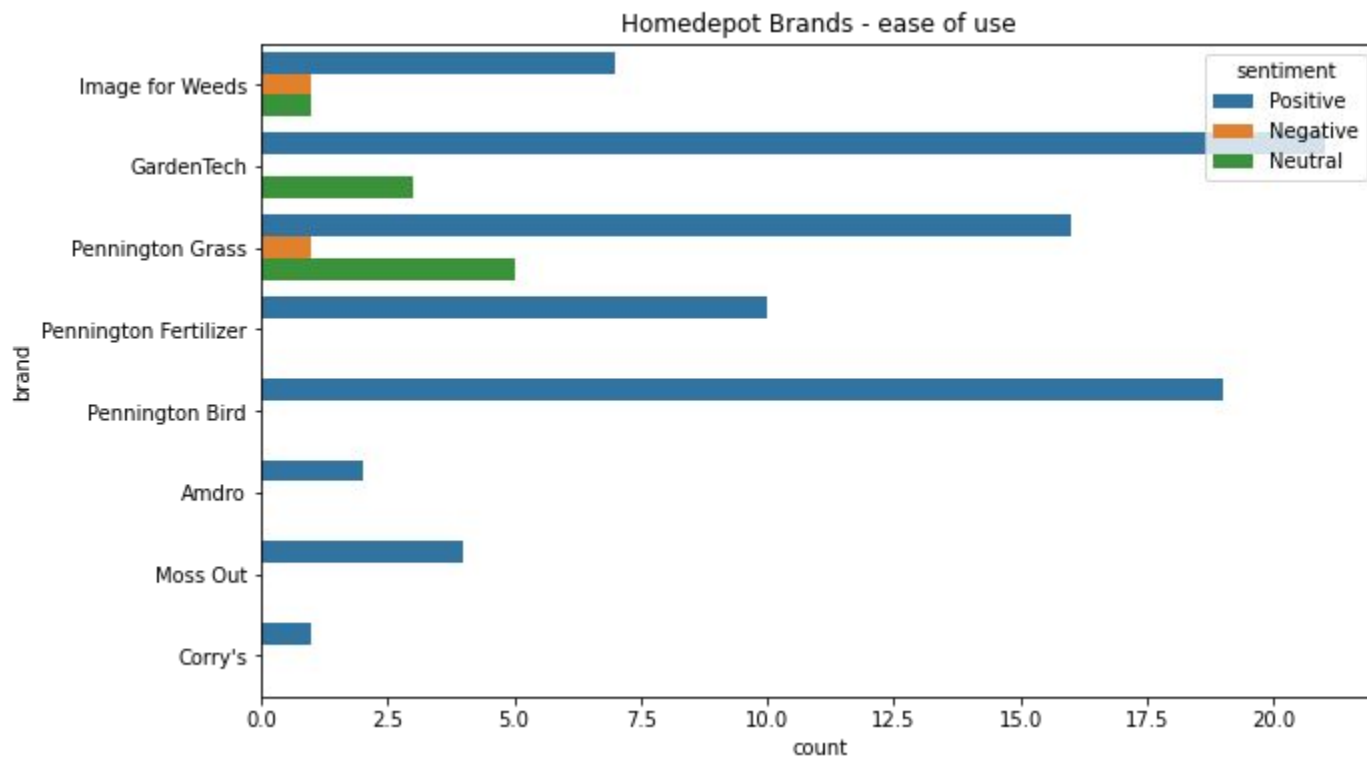
# Sentiment Histogram of Homedepot by Brands - Efficient



# Sentiment Histogram of Homedepot by Brands - Effective

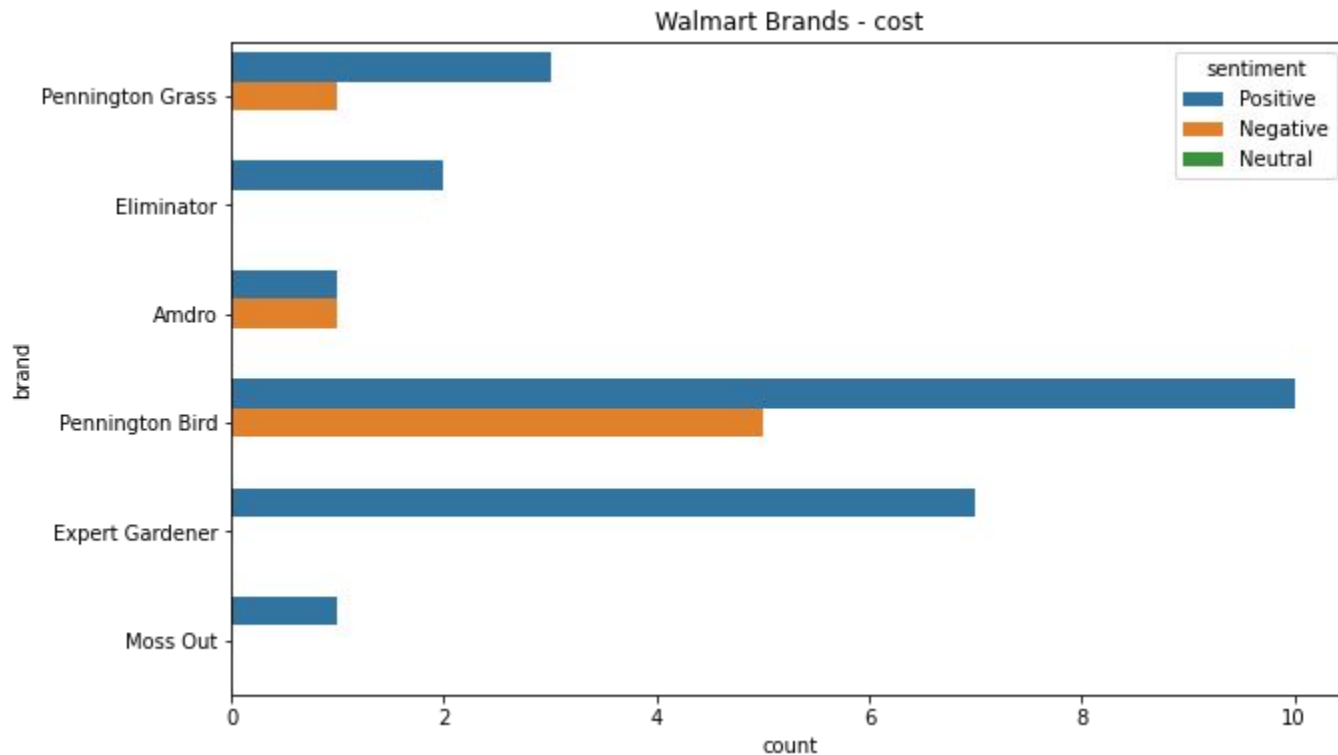


# Sentiment Histogram of Homedepot by Brands - Ease of Use

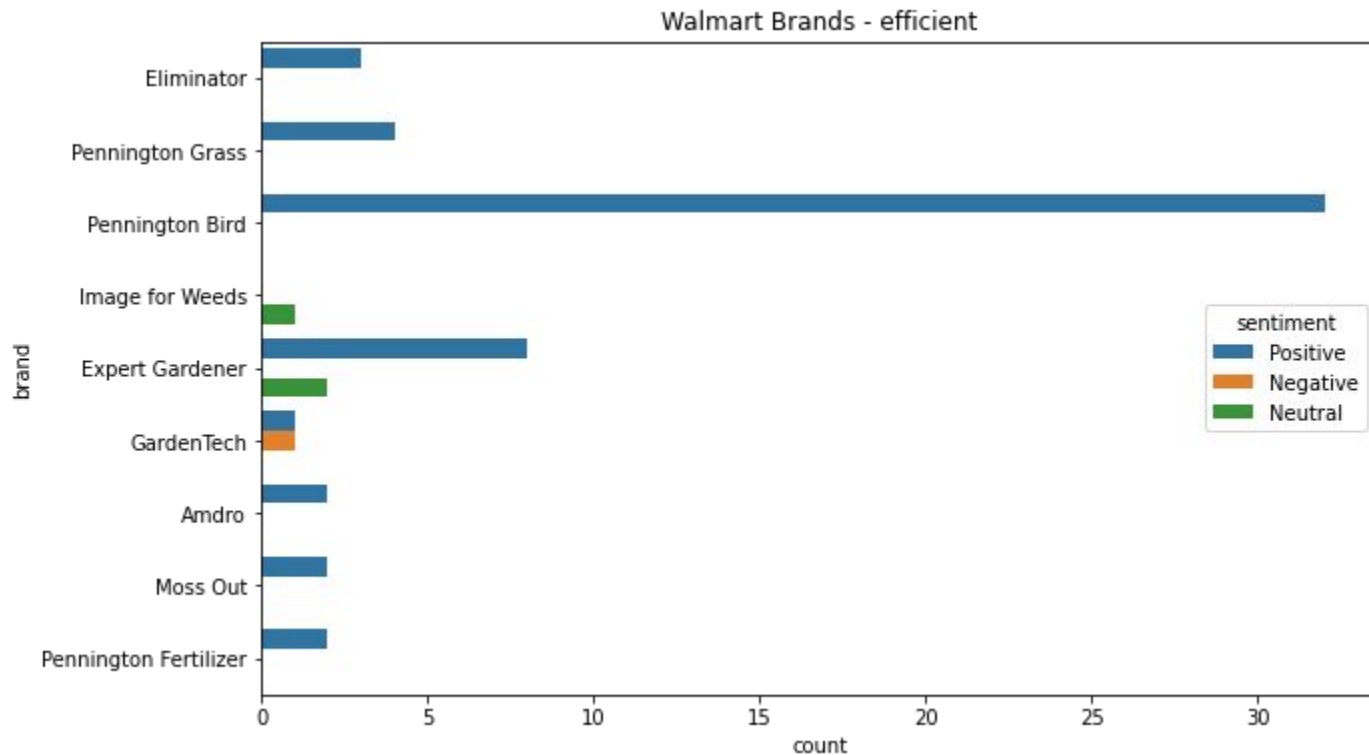




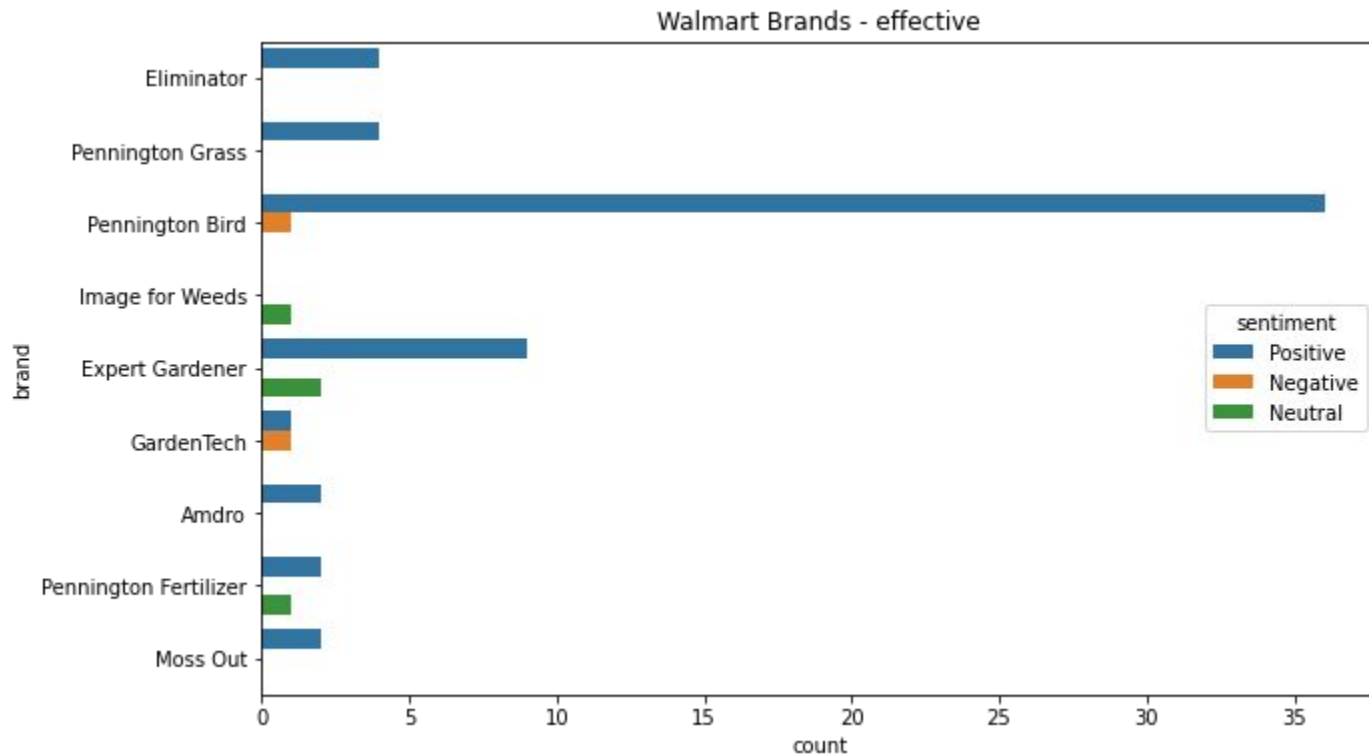
# Sentiment Histogram of Walmart by Brands - Cost



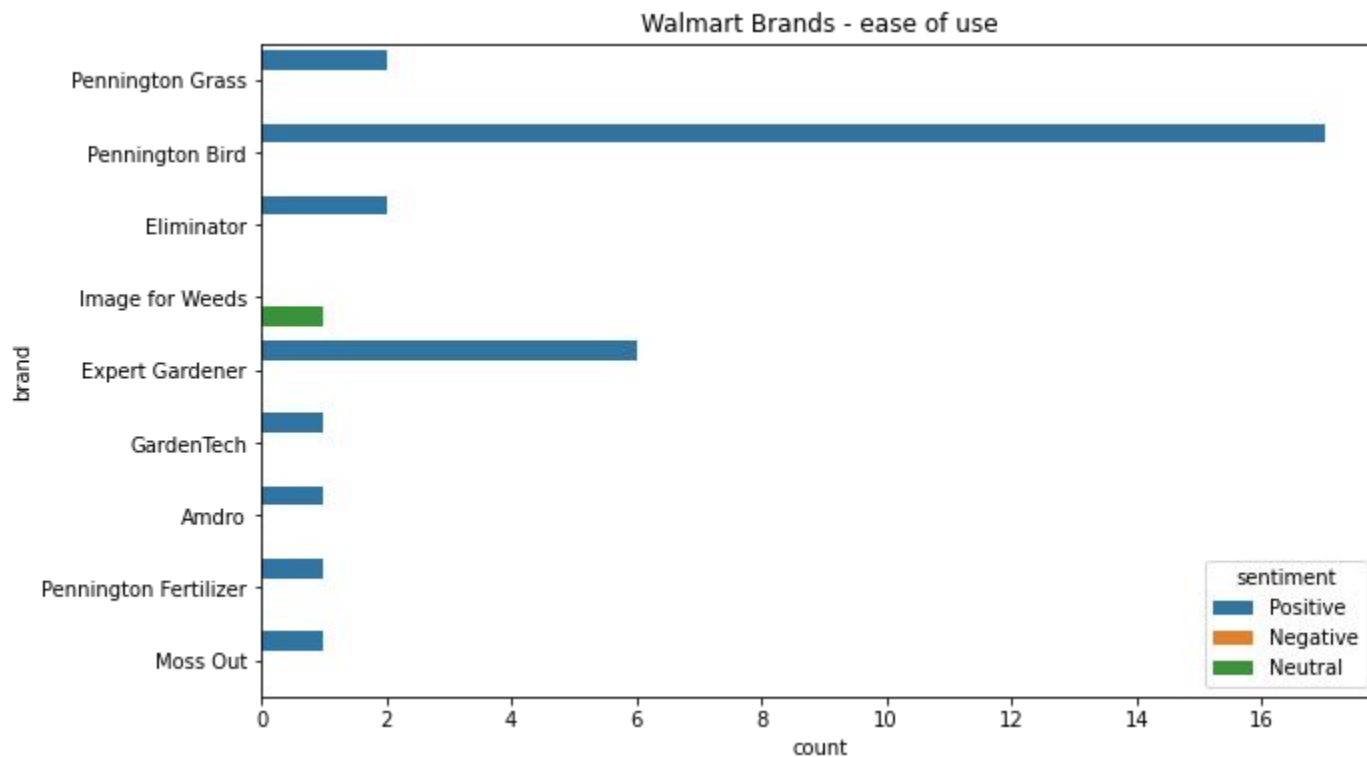
# Sentiment Histogram of Walmart by Brands - Efficient



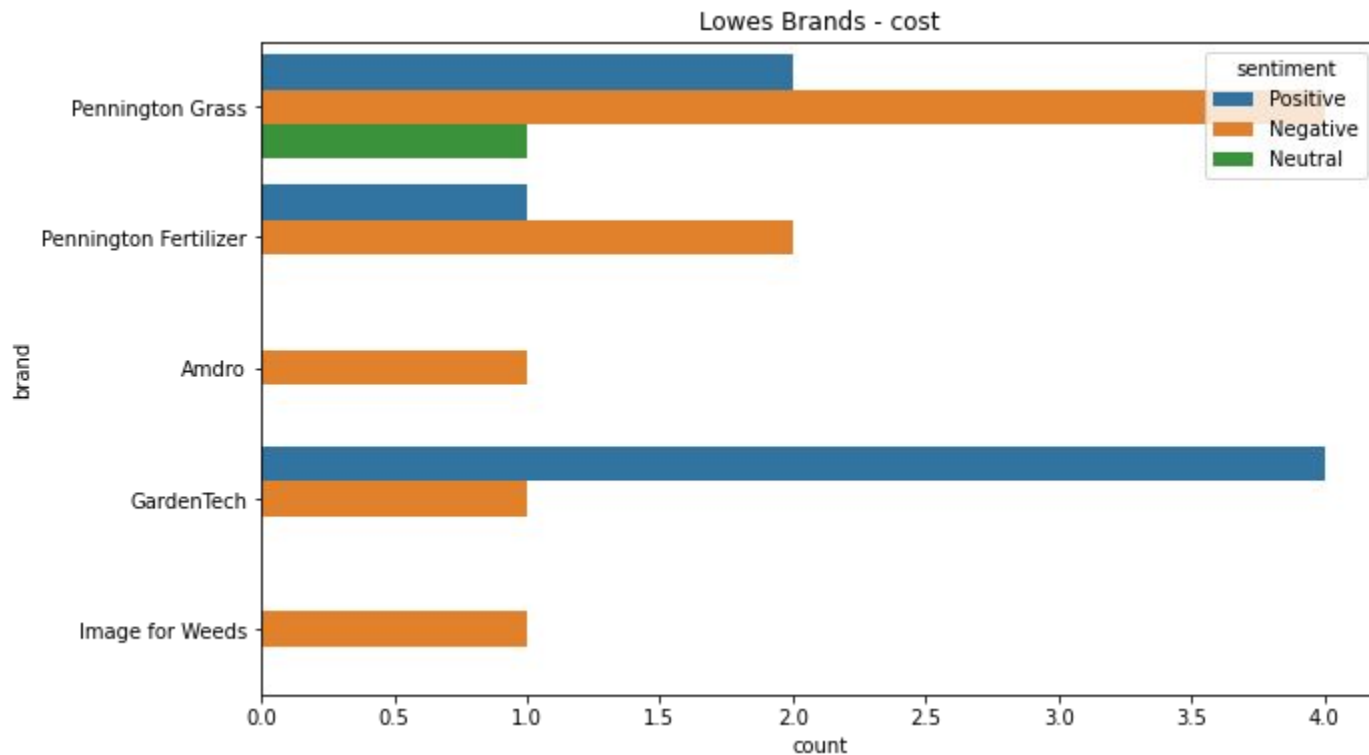
# Sentiment Histogram of Walmart by Brands - Effective



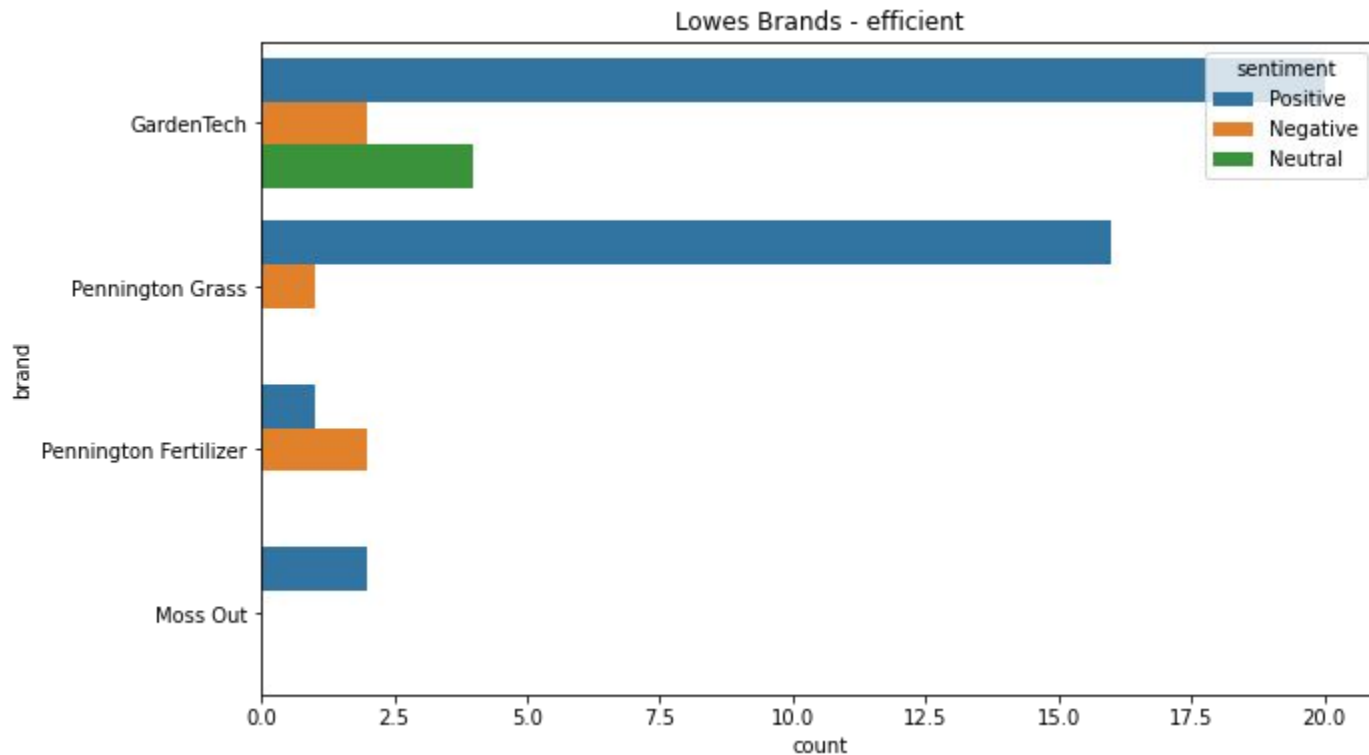
# Sentiment Histogram of Walmart by Brands - Ease of Use



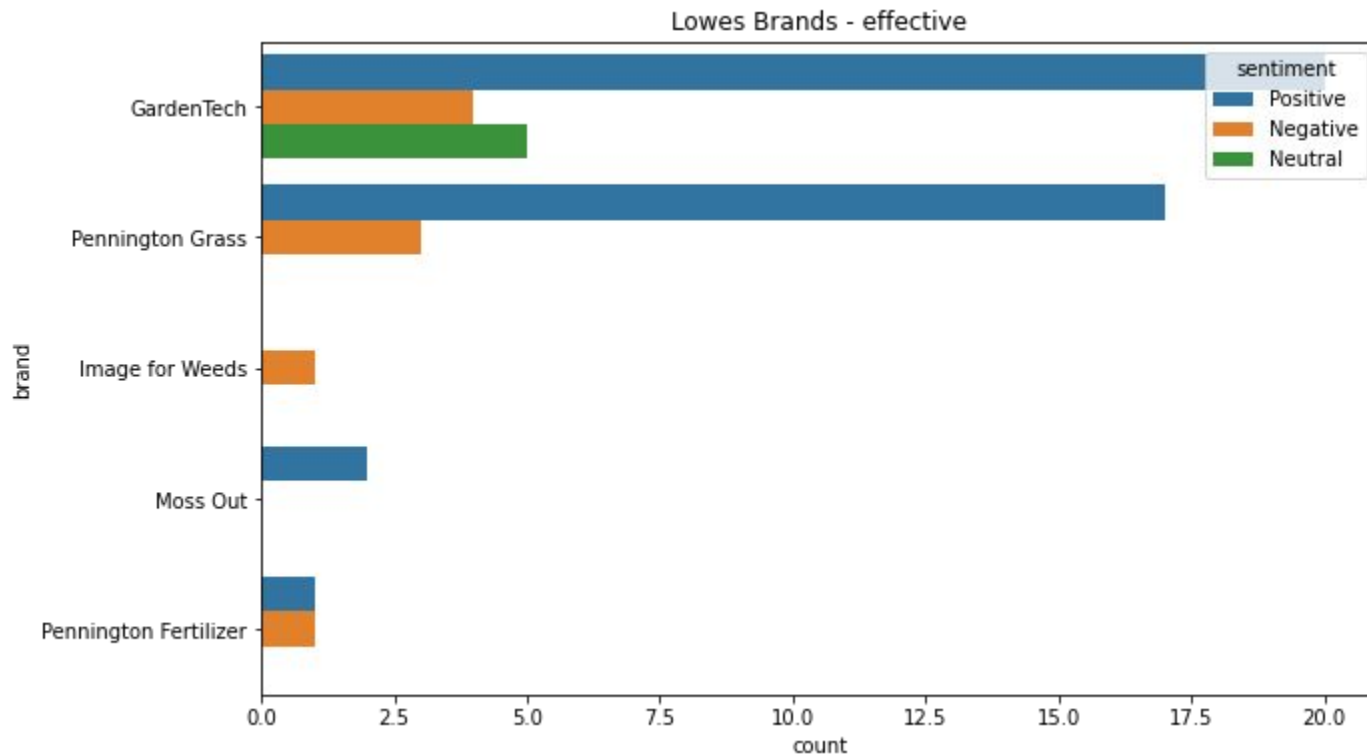
# Sentiment Histogram of Lowes by Brands - Cost



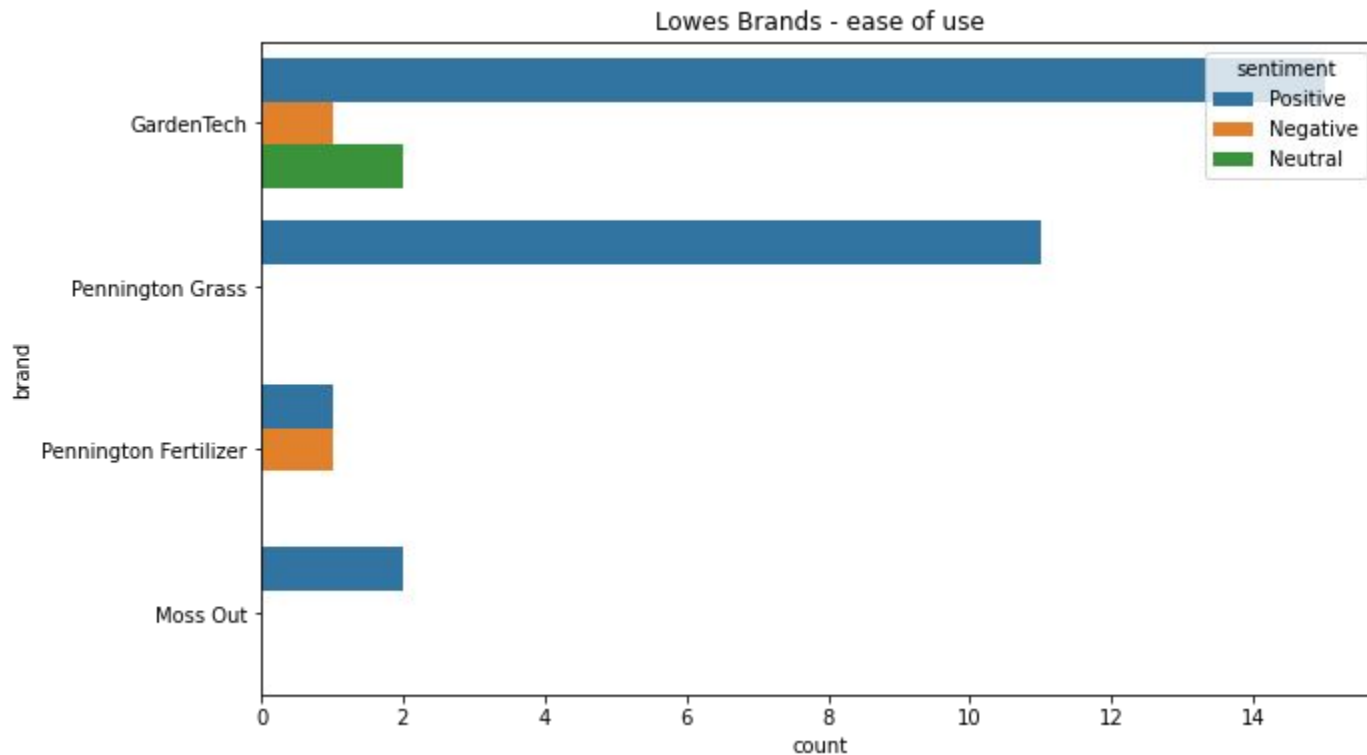
# Sentiment Histogram of Lowes by Brands - Efficient



# Sentiment Histogram of Lowes by Brands - Effective

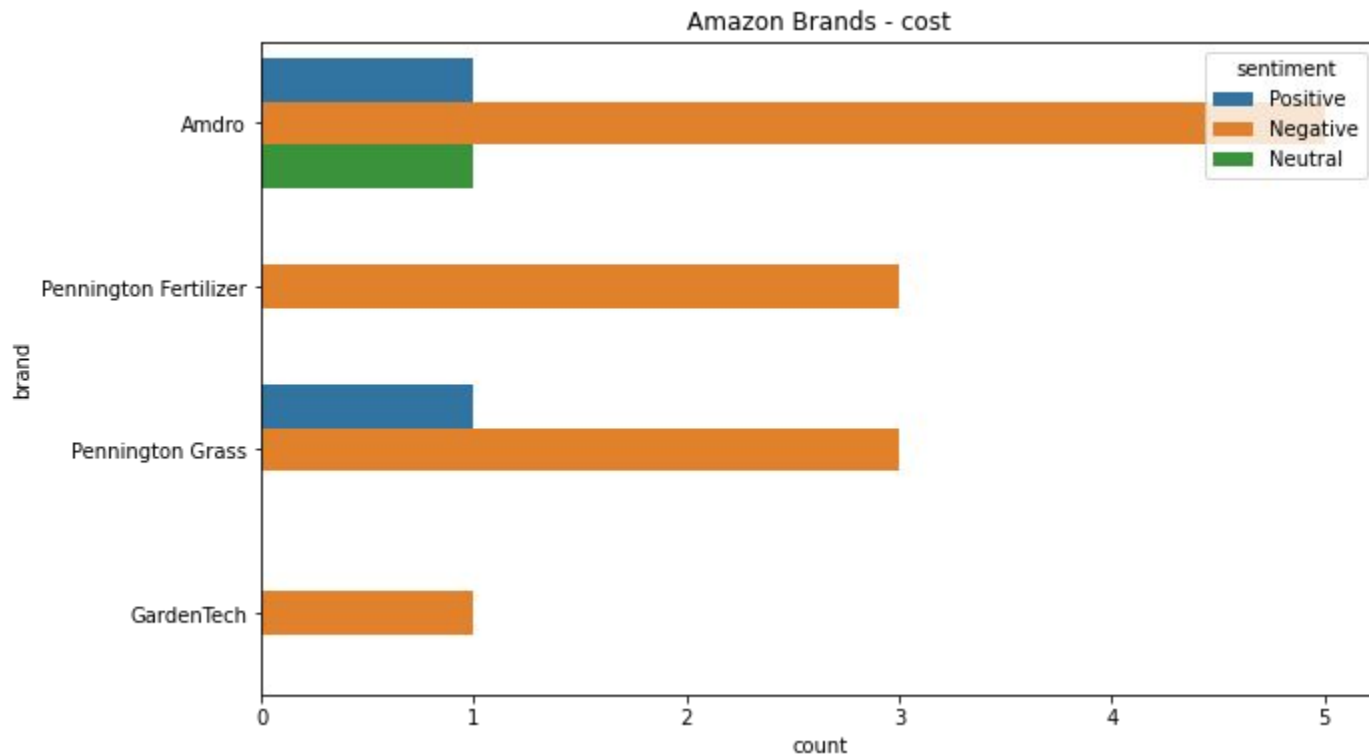


# Sentiment Histogram of Lowes by Brands - Ease of Use

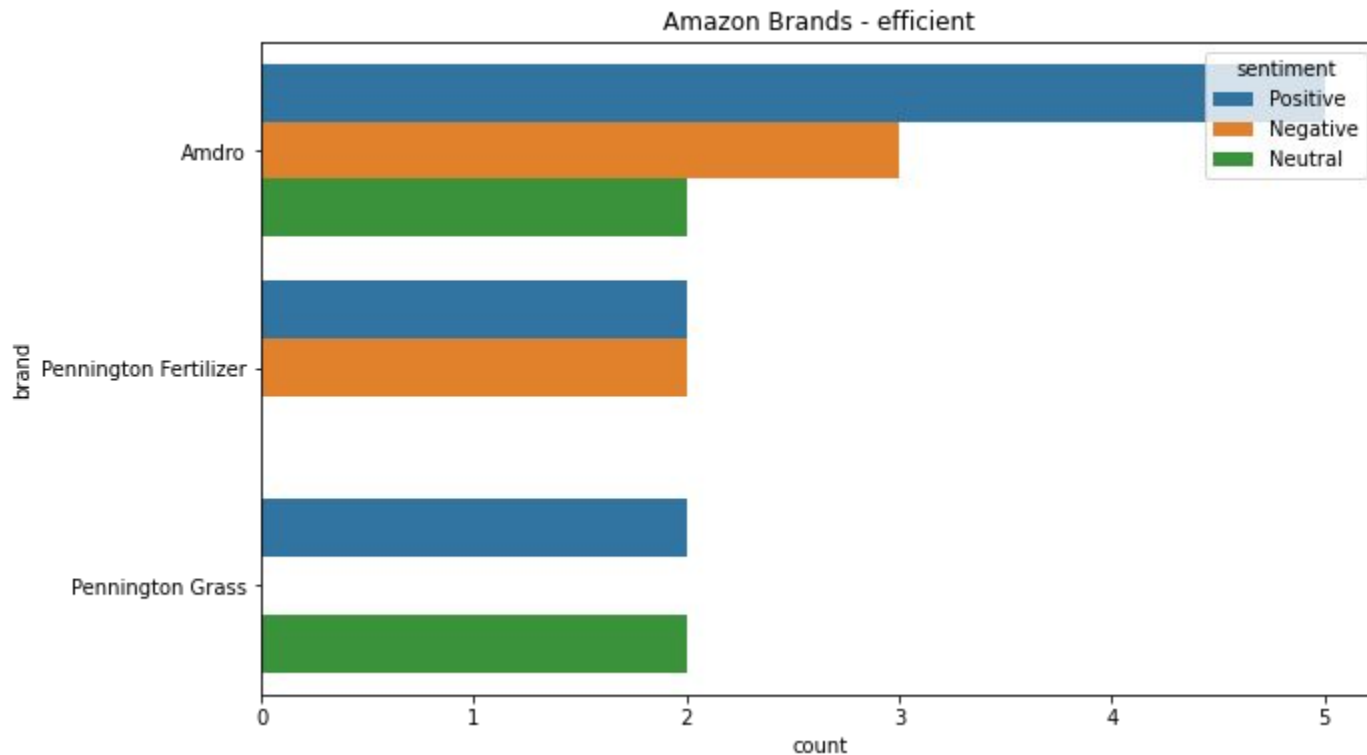




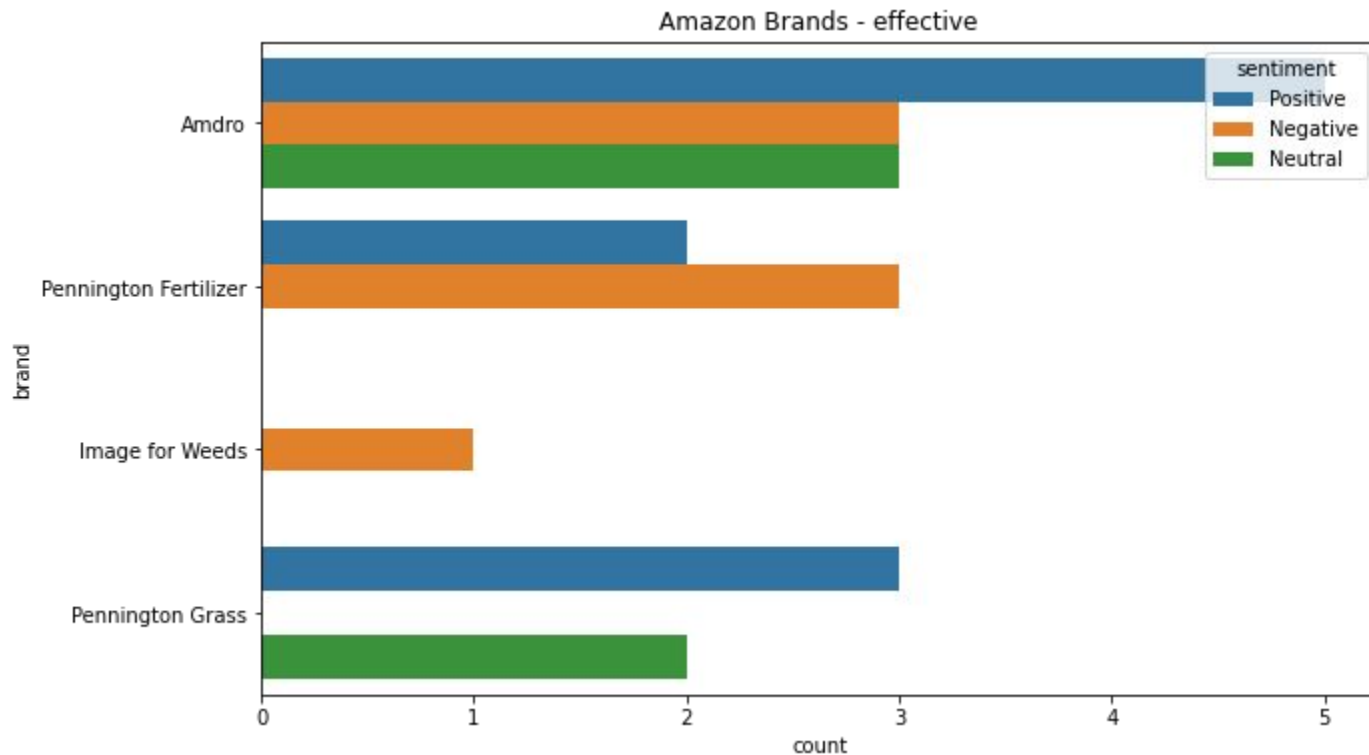
# Sentiment Histogram of Amazon by Brands - Cost



# Sentiment Histogram of Amazon by Brands - Efficient



# Sentiment Histogram of Amazon by Brands - Effective



# Sentiment Histogram of Amazon by Brands - Ease of Use

