

# Robo-Reviews

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# Project Overview

**RoboReviews** is a project designed to analyze customer reviews using AI to generate valuable insights for product improvement and consumer guidance.

## Main Goals:

1. **Sentiment Analysis:** Classify reviews as positive, negative, or neutral to help understand customer sentiment.
2. **Product Clustering:** Cluster products into 4-6 meaningful categories for better market understanding.
3. **Review Summarization:** Use generative AI to summarize reviews and recommend the top products for each category.



# RoboReviews:

Turning customer reviews into... Something, eventually...

Why This Is Still Awesome:

- Despite the challenges, the groundwork is in place. Every good project starts somewhere
- If nothing else, I've learned a whole lot during this project. Specially how not to build a model or two.
- Plus, the preprocessing is almost (sadly) a project on its own



# What is your story?

## **Model 1: Sentiment Analysis**

This one's almost functional! It can tell a happy customer from an angry one... most of the time. Let's call it a 'work-in-progress with potential.

## **Model 2: Product Category Clustering**

Technically, there's code. Realistically, it's just a very confused collection of products that refuses to behave. Consider it a philosophical experiment in chaos.

## **Model 3: Generative AI for Summarization**

Barely started, but the ideas are solid. If ideas could be considered finished models, then I'd have a masterpiece!

# Demo

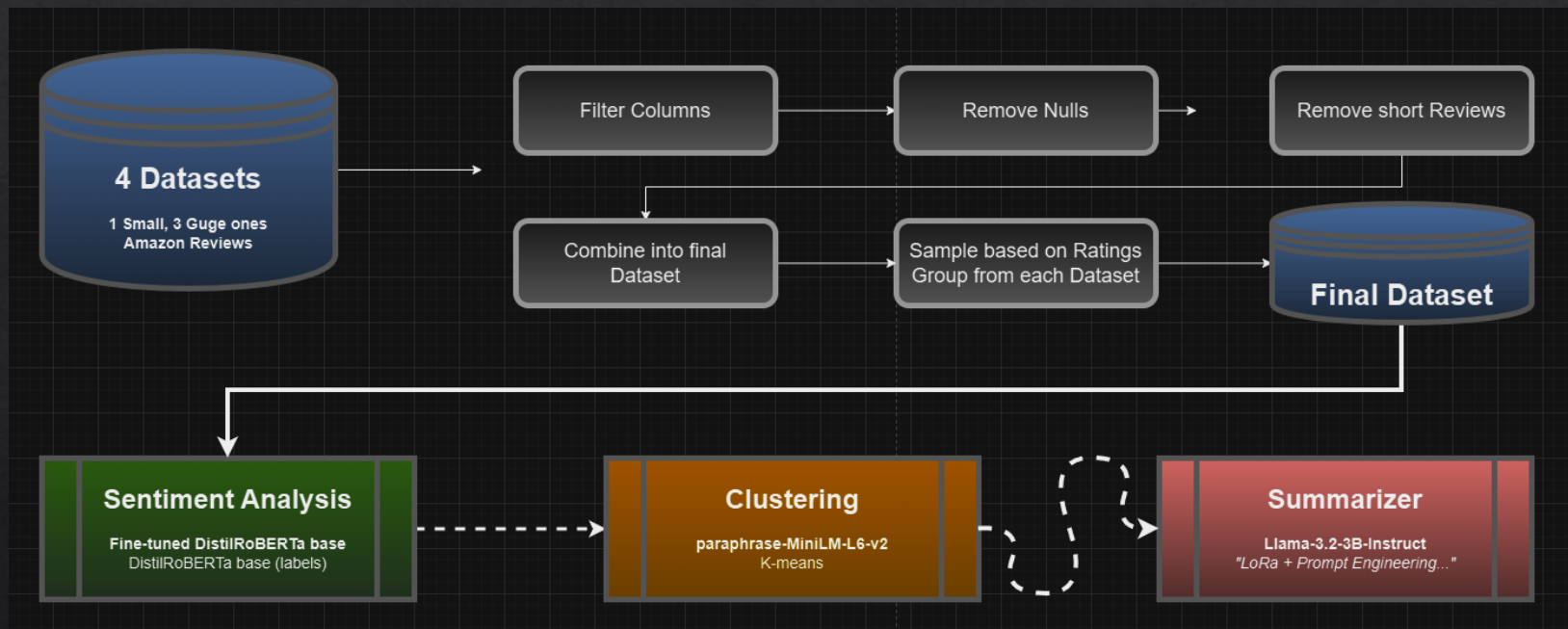
- Add a short demo of how your product can be used
- And/Or add an example of a chat review based on your product

# Introduction

- Explain the real world problem this solves
- Leave a point of connection to your methodology

Now that you have seen our super cool product and know why this matters, let me show how we did it!

# Methods





# Model 1: Sentiment Analysis

The goal was to classify customer reviews into positive, negative, or neutral categories.

## Problems:

- The review ratings are not always consistent with the sentiment.
- Lack of True sentiment labels for training the model.

## Approach:

- Generate labels with a simpler approach to then train and evaluate the model.
- **VADER**: Initially used, but it struggled with nuanced sentiment.
- **DistilRoBERTa**: Switched to using the base model before training to generate the labels.



# Model 1 - Evaluation

- VADER

Epoch	Training Loss	Validation Loss	Accuracy
0	0.652000	0.585210	0.750366
2	0.456400	0.451295	0.846159
4	0.357300	0.418835	0.863376
6	0.286000	0.424632	0.863281
8	0.248900	0.436297	0.876551

- DistilRoBERTa

Epoch	Training Loss	Validation Loss	Accuracy
0	0.507100	0.466671	0.821113
2	0.249100	0.232423	0.928140
4	0.167100	0.215747	0.931417
6	0.131000	0.214208	0.935364
8	0.110800	0.227629	0.934504

True label	NEGATIVE	NEUTRAL	POSITIVE
NEGATIVE	5666	714	118
NEUTRAL	753	3213	384
POSITIVE	481	1912	18181
		Predicted label	

# Takeaway

## Project Recap:

- Model 1:  
Functional but not Perfect! Managed to classify reviews into sentiment categories.
- Model 2:  
Technically built, practically a bit of a mess (clustering didn't quite work out as planned).
- Model 3:  
Barely started, but I laid some foundations...



# Takeaway

## Lessons Learned:

- Spent way too much time on data preprocessing, which the Project was definitely not about.
- Got distracted by investigation, learning better ways of doing things and side experiments instead of focusing on the project goals.
- **Main Lesson:** Finish first, then optimize.

## Final Note:

- Despite running out of time, I learned a lot about both what to do and what *not* to do in projects like this.

Repository: [Grimngor/RoboReviews \(github.com\)](https://github.com/Grimngor/RoboReviews)