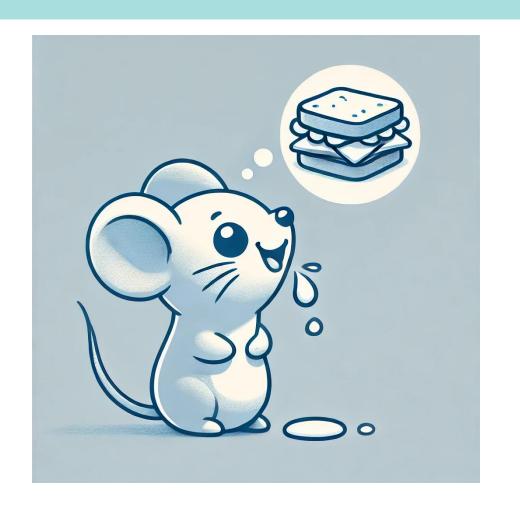
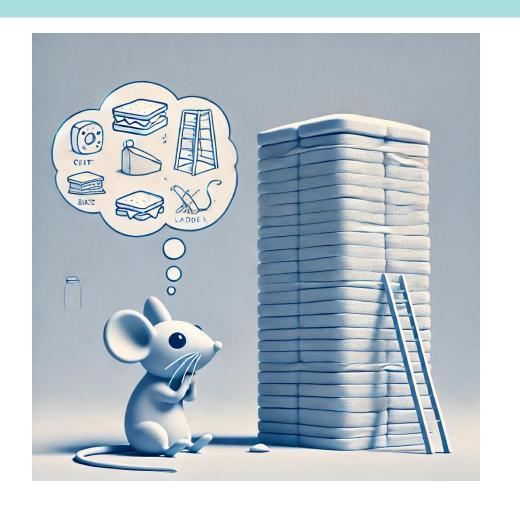
Iron Rat race

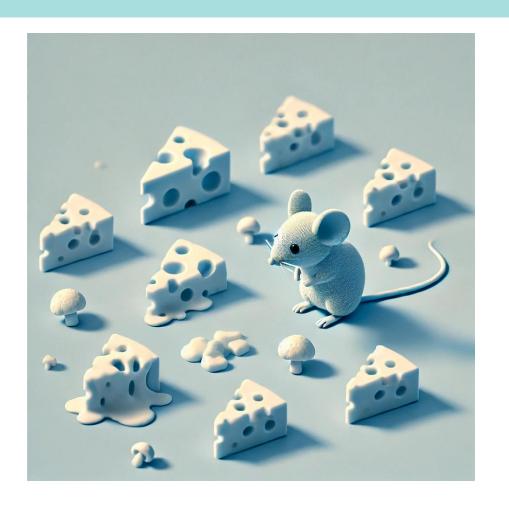
LUIS

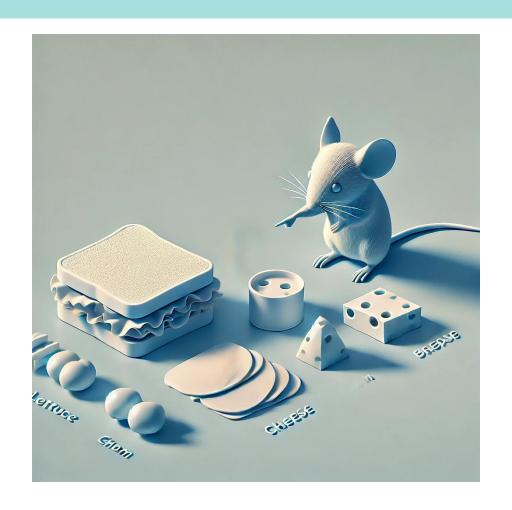




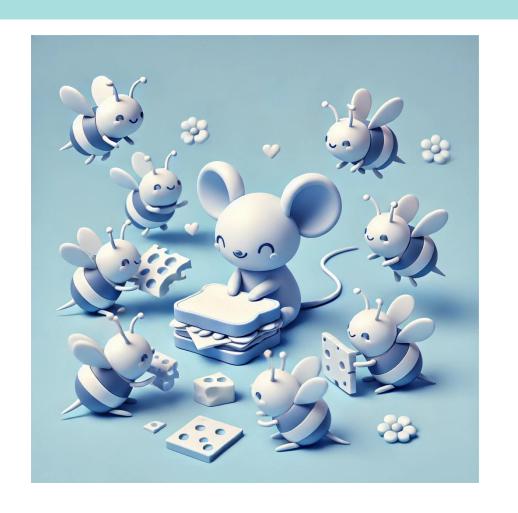




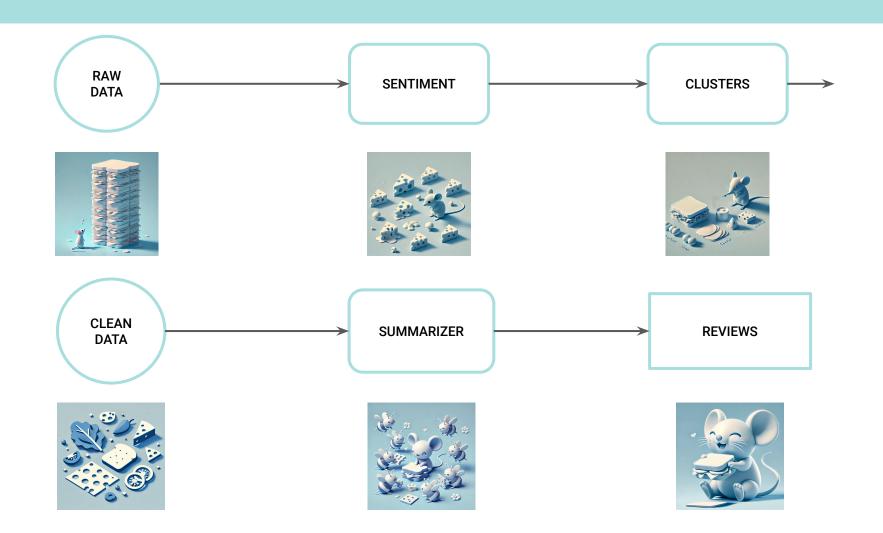












DATA PREPROCESSING



DATA PREPROCESSING



- Small dataset (1429.csv)
- Transformers: robust to noise
- No need to lowercase or remove punctuation
- Text without redundant special characters
- Concatenation of title and text columns
- Removed NaNs





Star-rating analysis

- Assigned **sentiment** to stars (4 & 5 positive, 3 neutral, 1 & 2 negative)
- **Imbalanced**: positive ratings overrepresented (>90%)
- Tried to generate **synthetic** reviews: didn't work
- Finally did not under- or oversample: could affect the clusterizer and summarizer

Customer rating not fully reliable

- Zero-shot classification with facebook/bart-large-mnli
- Analized results and cleaned errors (0.32%)



Evaluation

- **Accuracy**: 0.93%

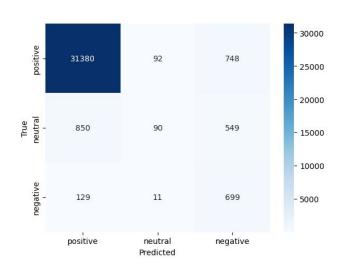
- **Precision:** 0.93%

- **Recall:** 0.93%

- **F1 Score**: 0.92%

Confusion Matrix

- Perform best on the positive class
- Neutral class most challenging to classify
- Good at identifying negative instances, with some confusion with positives





• First attempt with distilbert-base-uncased-finetuned-sst-2-english

• Saved csv file with predicted labels

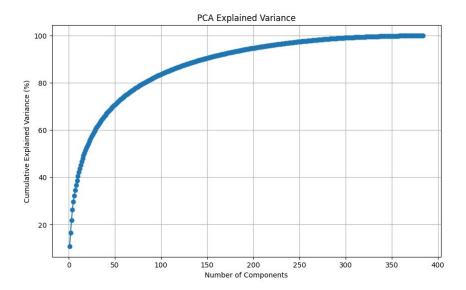




- Embed text with Transformers (sentence-transformers/all-MiniLM-L6-v2)
- Text to numeric representations & capturing semantic meaning
- **Cosine** similarity
- Measures the cosine of the angle between two vectors
- Dimensionality Reduction (PCA)
- Reduces dimensionality while preserving variance
- Efficient, preserves global structure



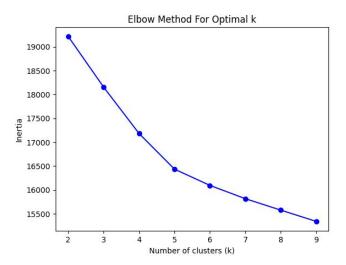
- Cumulative explained variance
- Obtain number of components for PCA (125)





K-Means

- Elbow Method for optimal number of clusters (5)





K-Means

- Analise cluster content
- Identify dominant product categories
- Common words per cluster



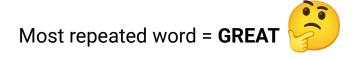
Cluster 0 = Tablet (Tablet)

Cluster 1 = Smart Speaker (Echo, Alexa)

Cluster 2 = E-book (Kindle)

Cluster 3 = Smart TV (TV, Fire)

Cluster 4 = Generic Love (Great, Product, Love)





- Assign labels to cluster
- Silhouette Coefficient (cluster dispersion) = 0.10
- Between -1 and 1
- Not misclassified but neighbouring clusters nearby
- DB Index (avg similarity of each cluster and its most similar one) = 2.67
- Between 0 and 1 = better clustering
- Suggest poor clustering



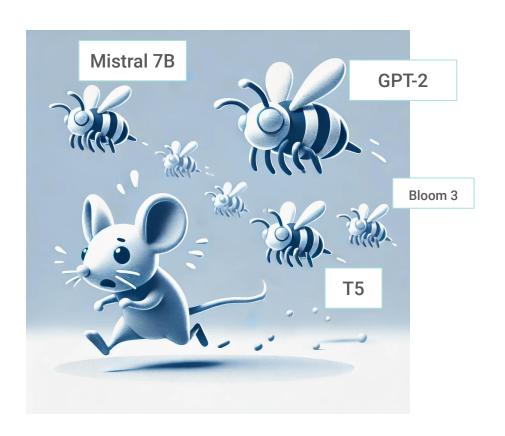
- Removed words with Regex to improve metrics
- Great, product, love
- Same clusters, no improvement
- 1st try clustering with product title
- Loses the point of the task













- Removed the "Generic Love" category & concatenated reviews + cluster + sentiment
- Mistral 7B, T5, Bloom 3
- Didn't manage to fully implement them
- GPT-2
- High training (2) and validation (2.31) loss
- Fine tuning: training arguments & prompt
- None acceptable results



RECAP & TAKEAWAYS

- Small dataset
- Sentiment with facebook/bart-large-mnli
- Clustering with sentence-transformers/all-MiniLM-L6-v2 and K-Means
- **Summarizer** didn't success (*GPT-2, T5, Mistral 7b, Bloom 3*)

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- Find the balance between research and implement/test/fine-tune models
- Clean and balance the data as much as possible

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