# Intel Core i7-12700K Processor Sentiment Analysis

Analyzing Customer Sentiments from Online Reviews

Presented by Team Techashu

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# Problem Statement

- ❖ **Problem:** With the increasing number of online reviews for Intel products, it's challenging to manually analyze customer sentiments and extract meaningful insights.
- \* Impact: Understanding customer sentiments is crucial for Intel to improve product quality, customer satisfaction, and market strategies.
- ❖ Goal: Develop an automated system to analyze sentiments from online reviews, providing actionable insights for Intel.



# Manual Sentiment Analysis System

- ❖ Innovation: Utilizes machine learning and natural language processing to analyze and classify sentiments from online reviews.
- **Efficiency:** Reduces manual effort and time required to understand customer feedback.
- \* Insights: Provides comprehensive insights into customer opinions, highlighting strengths and areas for improvement.

# Key Features of the Sentiment Analysis System

- \* Manual Sentiment Analysis: Downloading HTML pages of reviews and extracting data from them.
- **Sentiment Classification:** Categorizes reviews into positive, negative, and neutral sentiments.
- **Keyword Extraction:** Identifies common keywords and phrases in reviews.
- **Insights Generation:** Provides valuable insights into customer opinions.

# Process flow

01

### **Data Collection**

- Task: Download HTML pages of Amazon and Best Buy reviews for the Intel Core i7-12700K processor.
- Method: Use web scraping tools to collect reviews from multiple country-specific Amazon sites and Best Buy.

02

### Data Extraction

- Task: Extract relevant information (Content, Rating, Name, Date) from the downloaded HTML pages.
- Method: Utilize BeautifulSoup to parse the HTML and extract review data, saving it to a CSV file.

03

### **Data Translation**

- **Task:** Translate non-English reviews to English.
- Method: Use the 'langdetect' ibrary to detect the language and Google Translate API for translation, saving results to 'translated.csv'.

04

### **Data Cleaning**

- Task: Clean the translated reviews data.
- Method: Remove duplicates, extract numeric ratings, filter unreadable text, and keep only relevant columns, saving cleaned data to 'cleaned\_reviews.csv'.

05

### Sentiment Analysis, Evaluation and Visualization

- Task: Perform sentiment analysis on the cleaned reviews.
- Method: Train and apply machine learning models to classify sentiments, generate visualizations, and evaluate model performance.

# Methodology

### **❖** Data Collection

• Sources: Collected user reviews from various platforms such as e-commerce websites (e.g., Amazon, Bestbuy)

### Sample Size

Gathered a total of 737 reviews to ensure a comprehensive analysis.

### Data Preprocessing

- · Cleaning: Removed any duplicate reviews, non-English text, and irrelevant information such as advertisements or spam.
- Tokenization: Broke down the text into individual words or tokens to facilitate analysis.

### Sentiment Analysis Tools

- NLP Libraries: Used Natural Language Processing (NLP) libraries such as NLTK to perform sentiment analysis.
- Sentiment Scoring: Each review was assigned a sentiment score (e.g., positive, negative, neutral) based on the analysis.

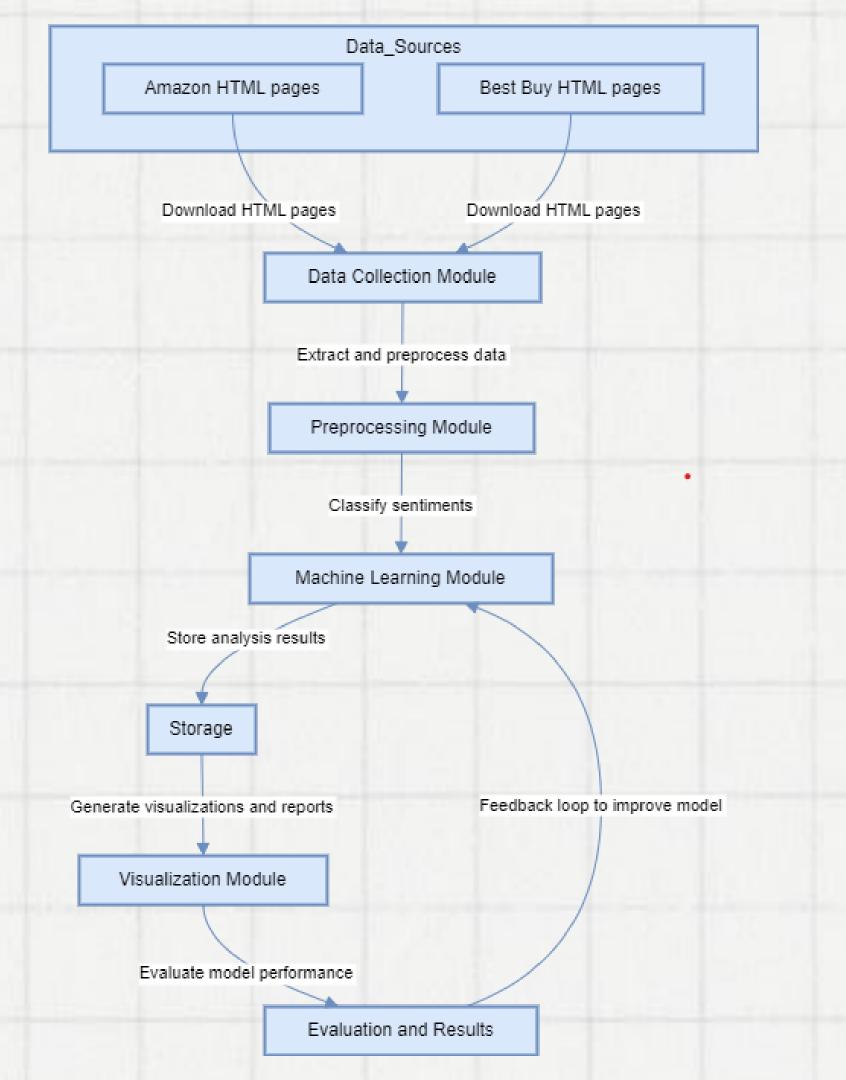
### Sentiment Classification

- Threshold Setting: Defined thresholds for categorizing reviews into positive, negative, and neutral. For instance, scores above 0.5 were classified as positive, below -0.5 as negative, and in between as neutral.
- Manual Validation: Conducted a manual review of a subset of reviews to validate the accuracy of the sentiment analysis.

### **❖** Data Visualization

- Charts and Graphs: Created visual representations such as pie charts, bar graphs, and word clouds to illustrate the sentiment distribution and key themes.
- Highlighting Key Themes: Identified and highlighted the most common positive and negative aspects mentioned in the reviews

# Architecture Diagram



# **Technologies and Tools**

- Programming Languages: Python
- Libraries: pandas, numpy, BeautifulSoup, scikit-learn, imbalanced-learn, googletrans, langdetect, nltk, wordcloud
- Machine Learning Models: Logistic Regression
- Visualization: Matplotlib, Seaborn



## Sentimental Analysis Breakdown

### **Positive Feedback**

### Performance

- High praise for speed and multi-core capabilities.
- Excellent gaming and professional application performance.
- The multi-core performance of this processor make this outstanding.

### Value for Money

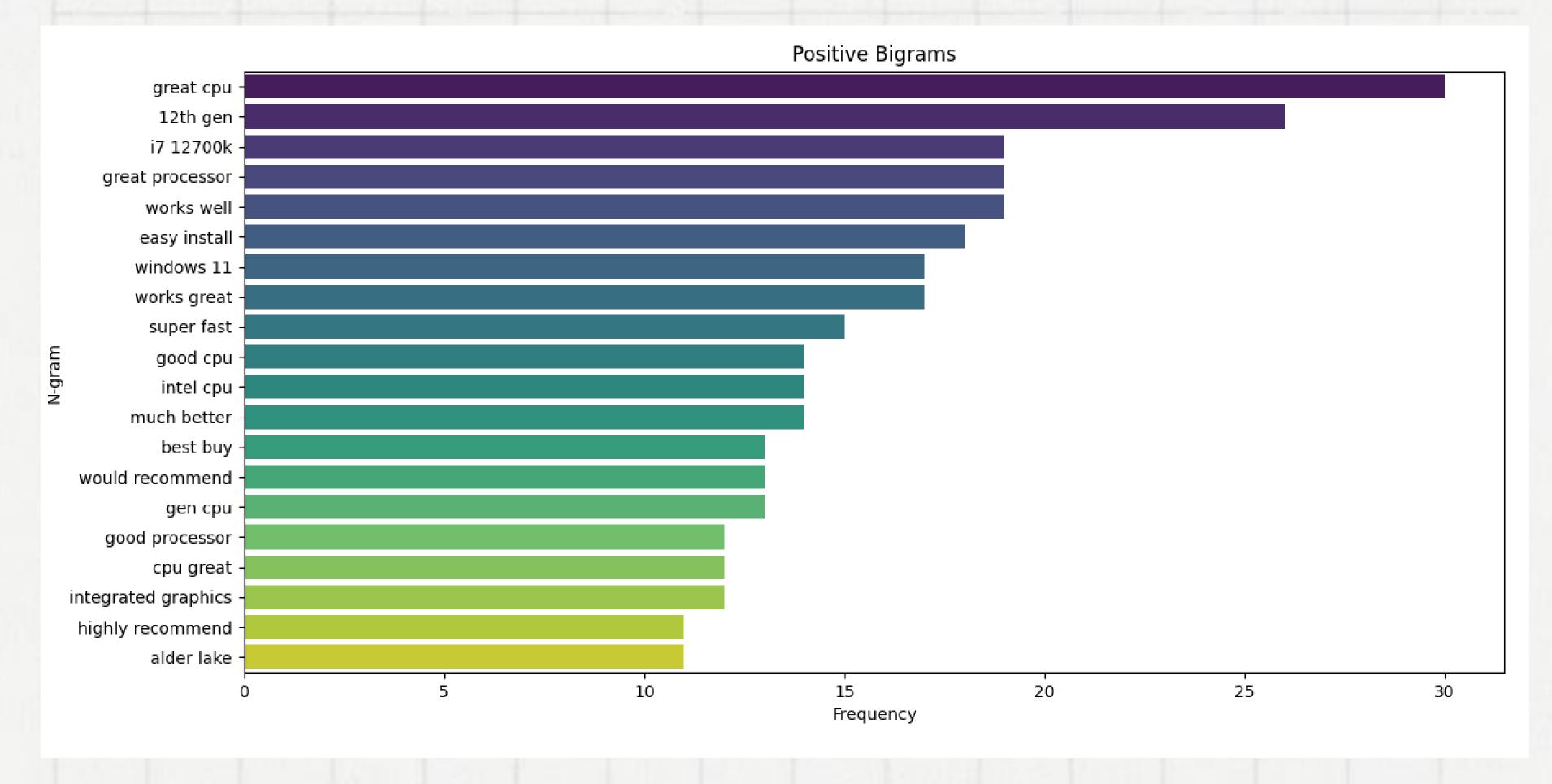
- Perceived as offering great value compared to other high-end processors.
- Many users felt it offered good value relative to its capabilities

### User Experience

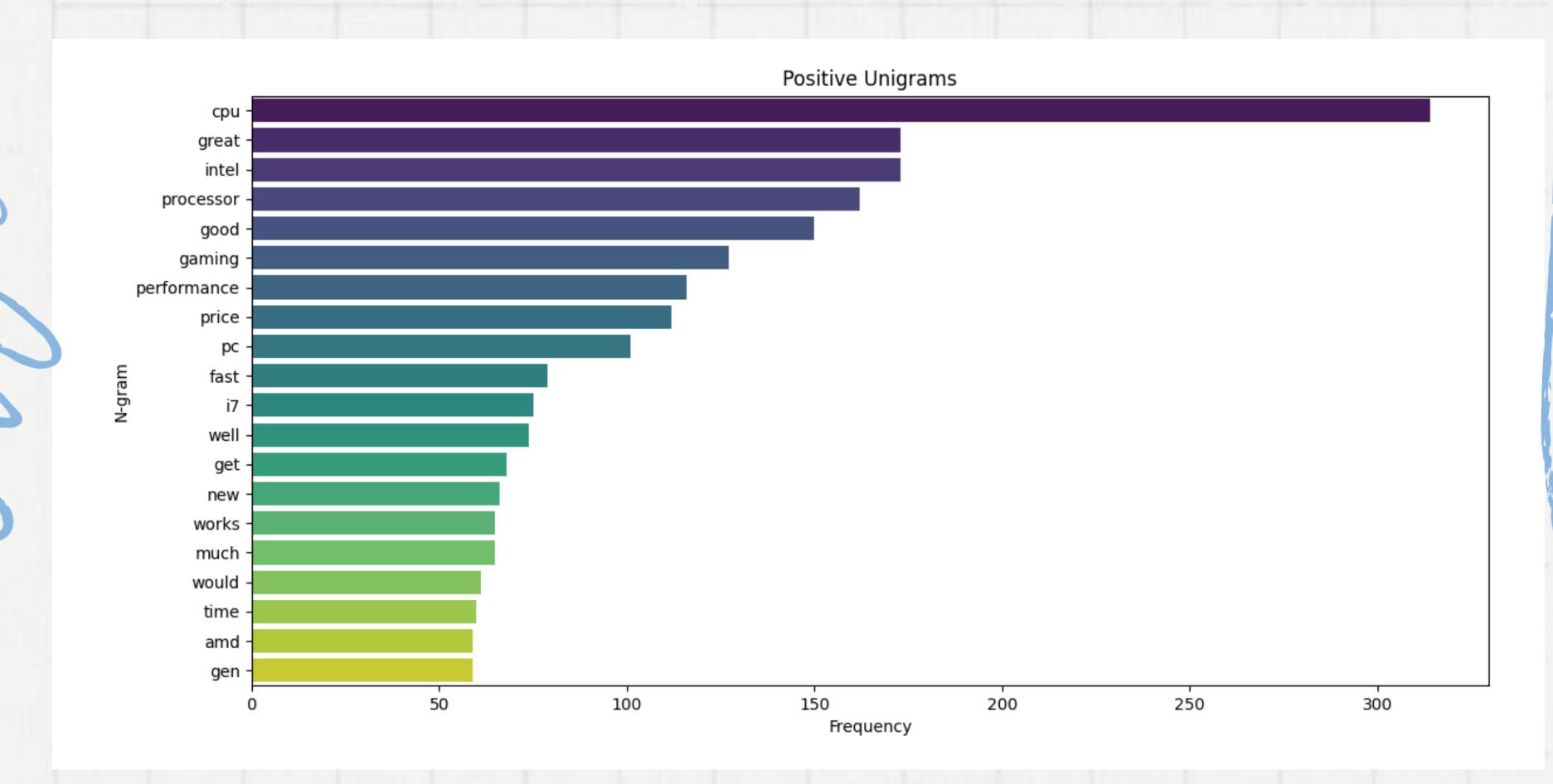
- Easy installation and compatibility with various systems.
- · High overall satisfaction.



### Visual Representation Of Most Frequent Words Used In The Reviews Of The User In The Form Of Bigrams



• Visual Representation Of Most Frequent Words Used In The Reviews Of The User In The Form Of Unigrams



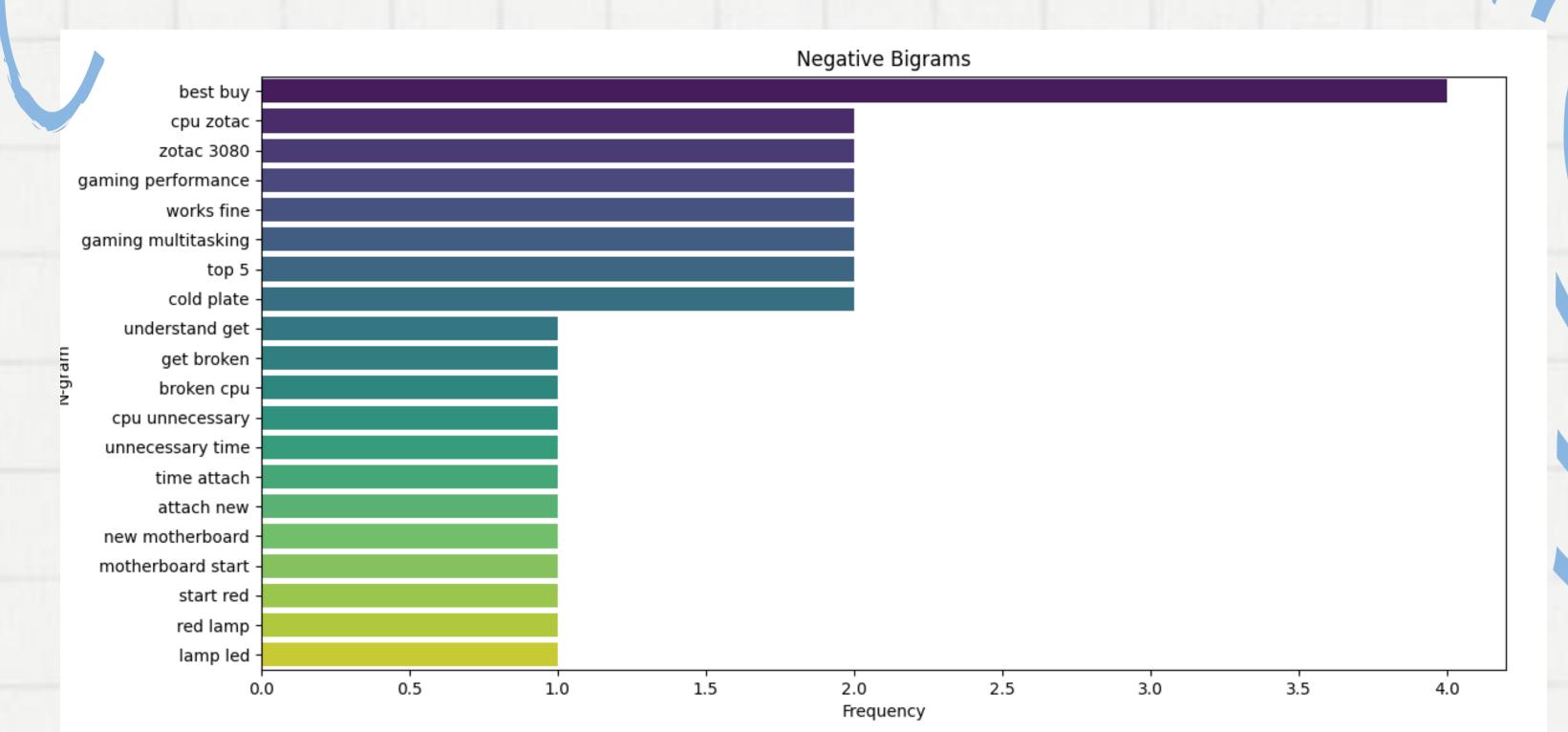
### **Negative Feedback**

### **❖ Heat Generation**

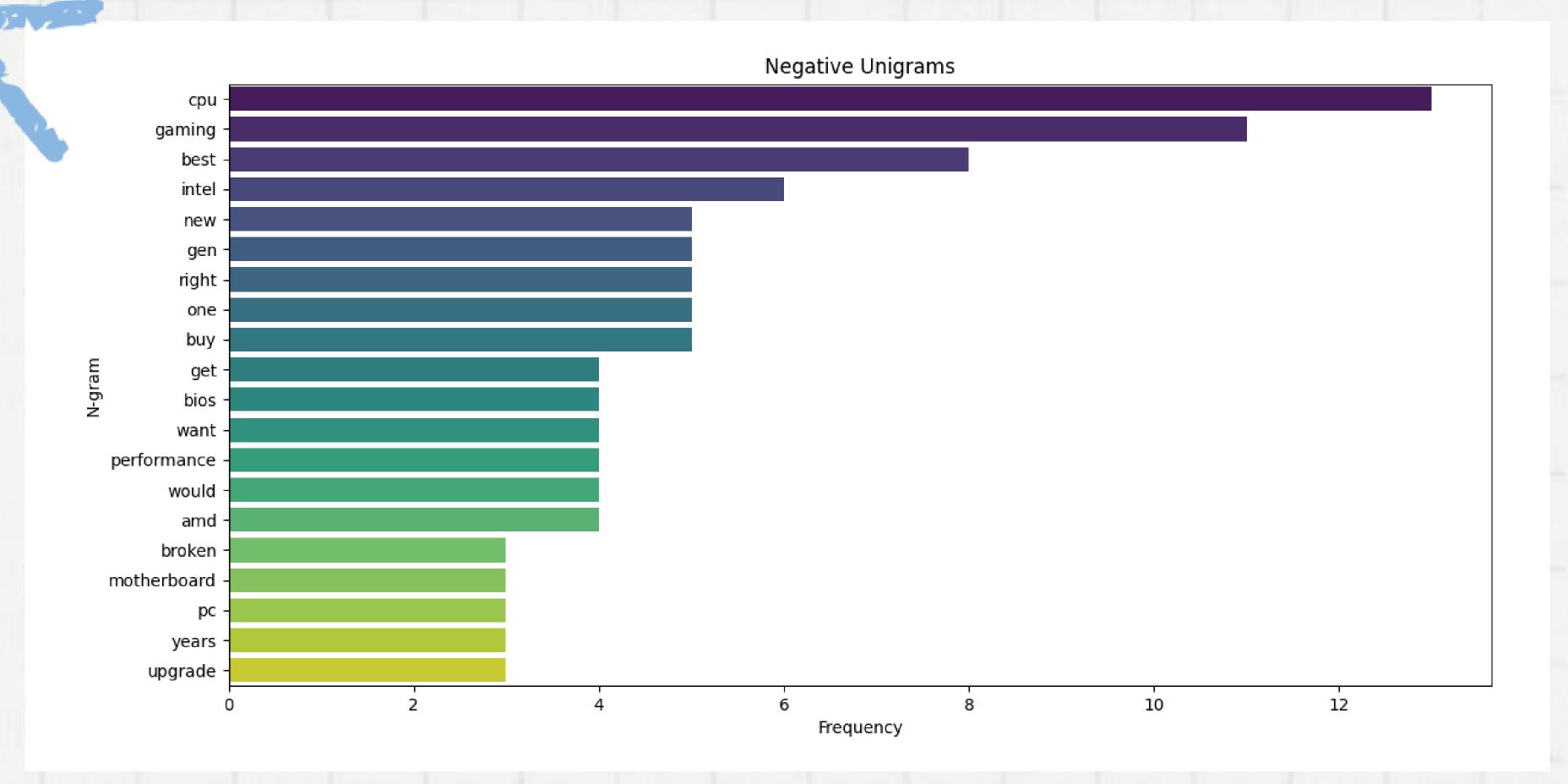
- Reports of running hot under load.
- Necessity for additional cooling solutions.

### **♦ Other Issues**

- Occasional faulty units and customer service complaints.
- Visual Representation Of Most Frequent Words Used In The Reviews Of The User In The Form Of Bigrams



Visual Representation Of Most Frequent Words Used In The Reviews Of The User In The Form Of Unigrams



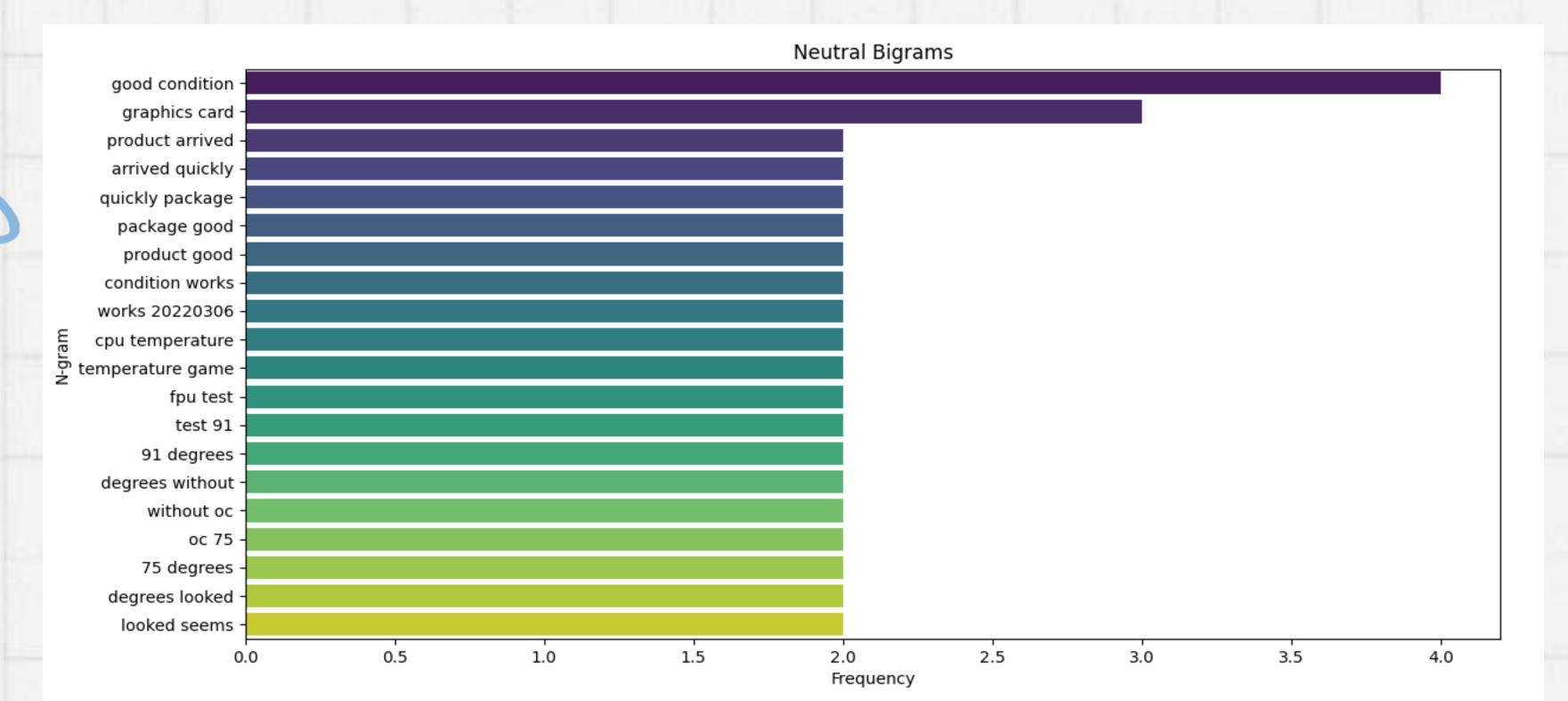
### **Neutral Feedback**

### **❖** Balanced Views

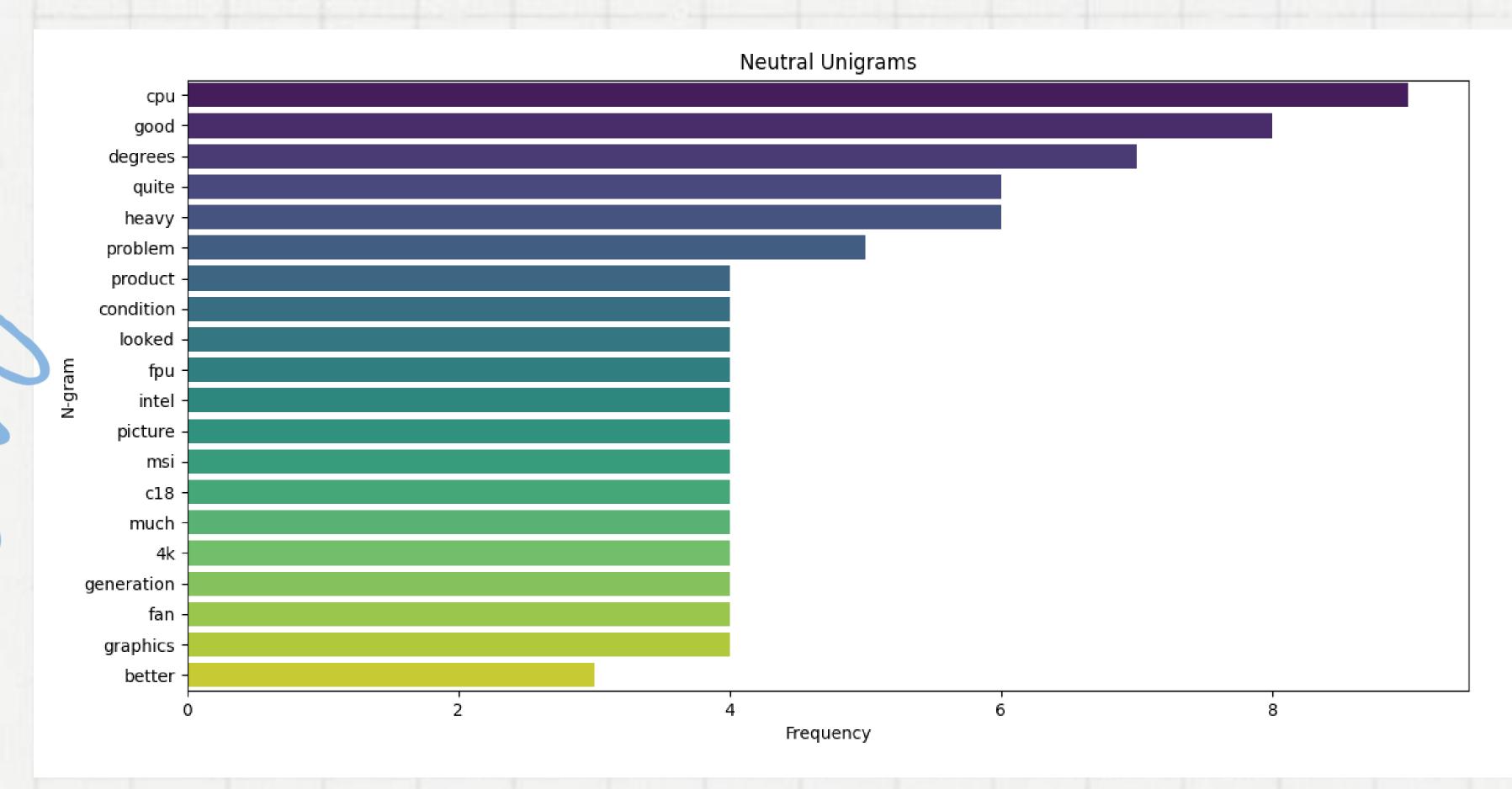
**Common Themes** 

• Mixed feedback, noting both strengths and weaknesses.

- Performance praised but heat management and cost
- Visual Representation Of Most Frequent Words Used In The Reviews Of The User In The Form Of Bigrams

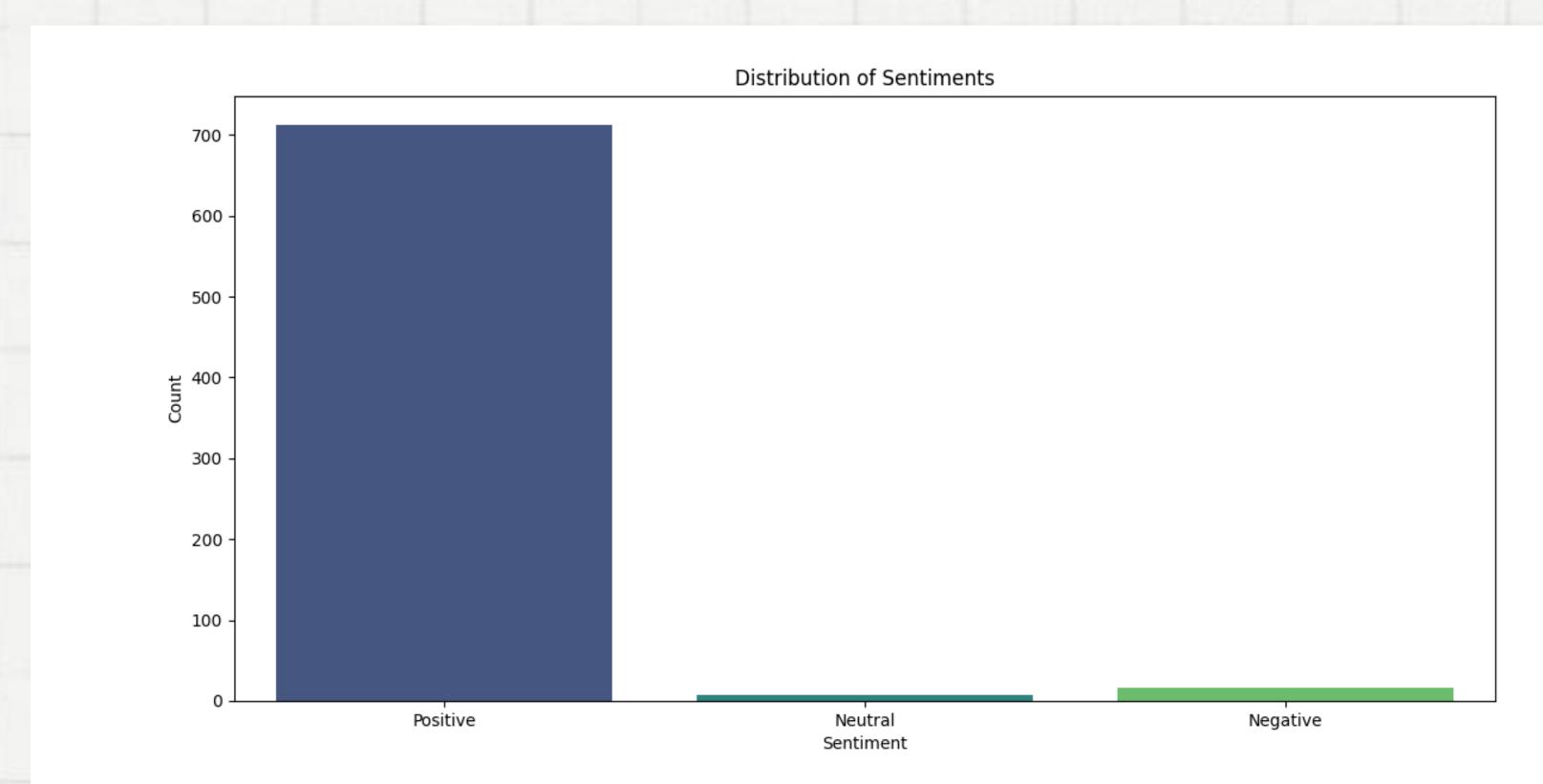


### Visual Representation Of Most Frequent Words Used In The Reviews Of The User In The Form Of Unigrams



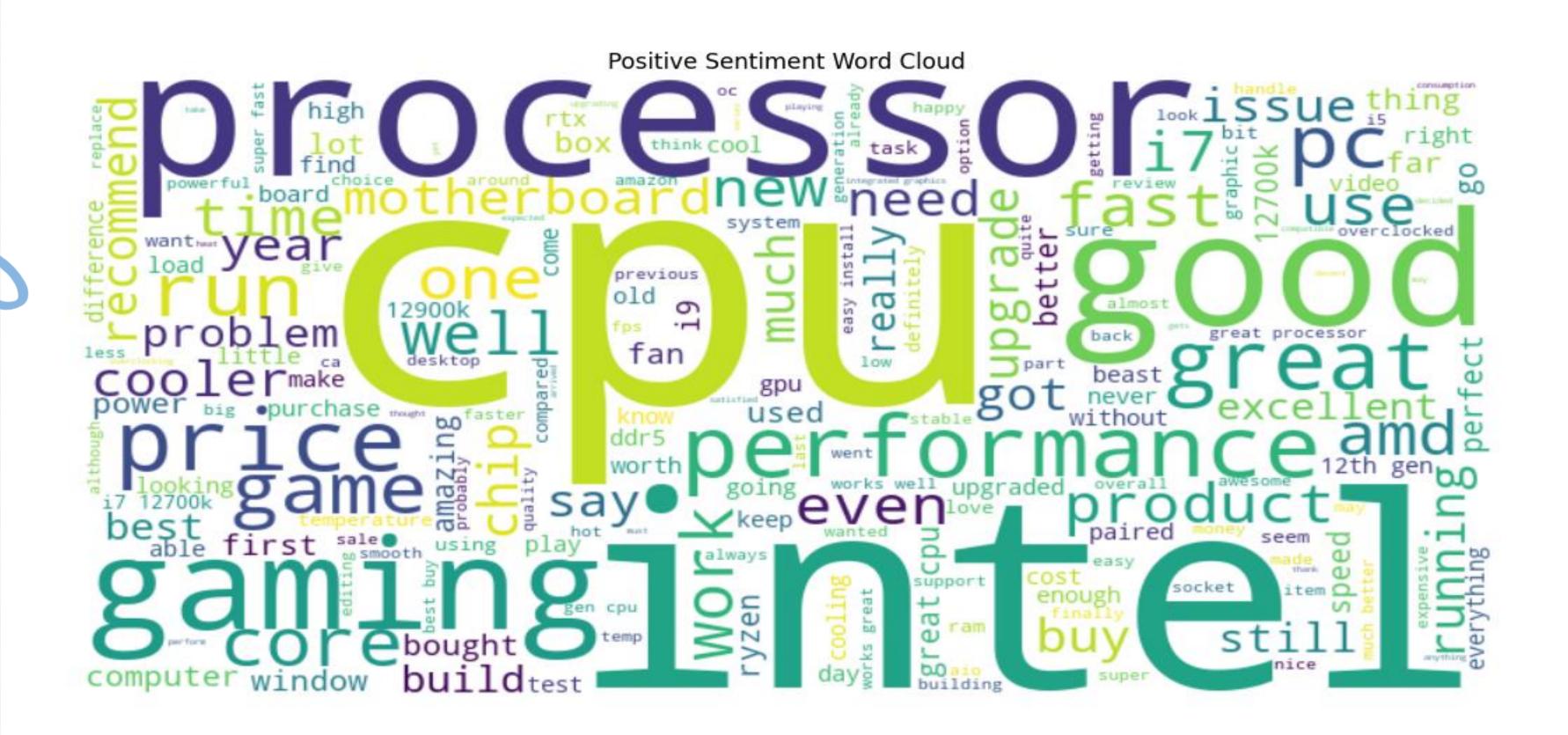
### **Visual Representations**

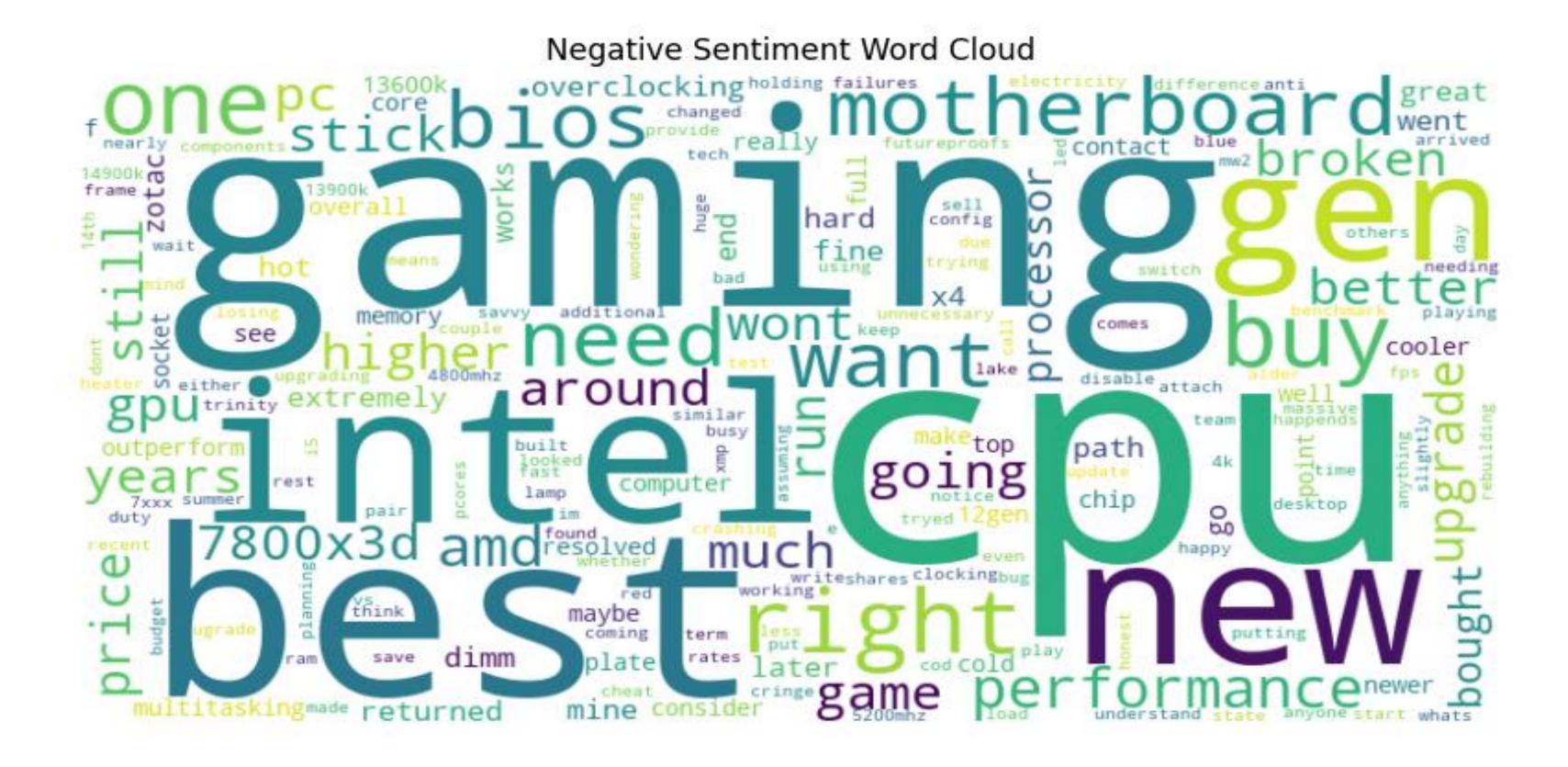
• Distributions Of Positive, Negative And Neutral Reviews



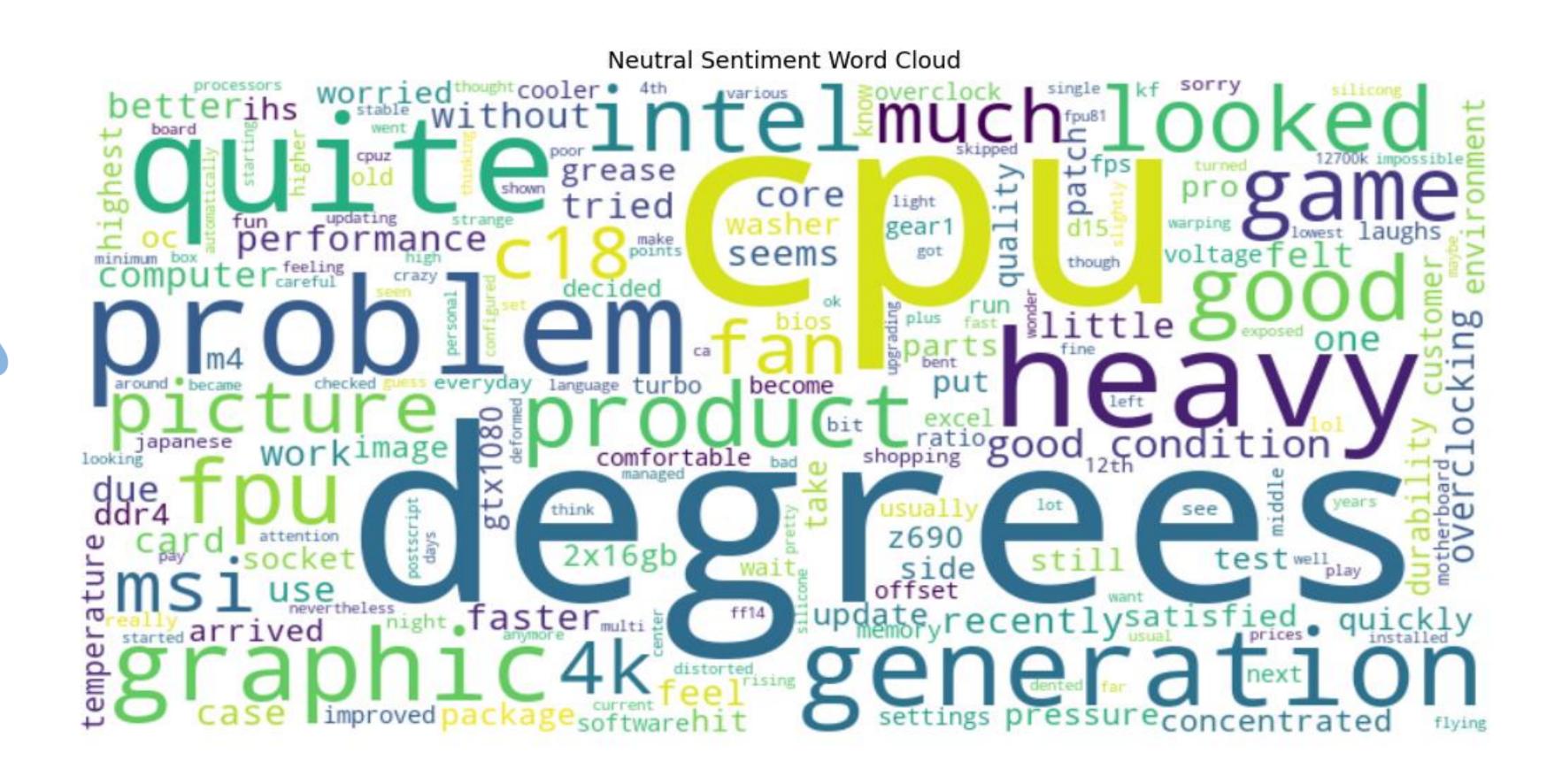
### Visualising Using Word Cloud

### **Positive Review Word Cloud**



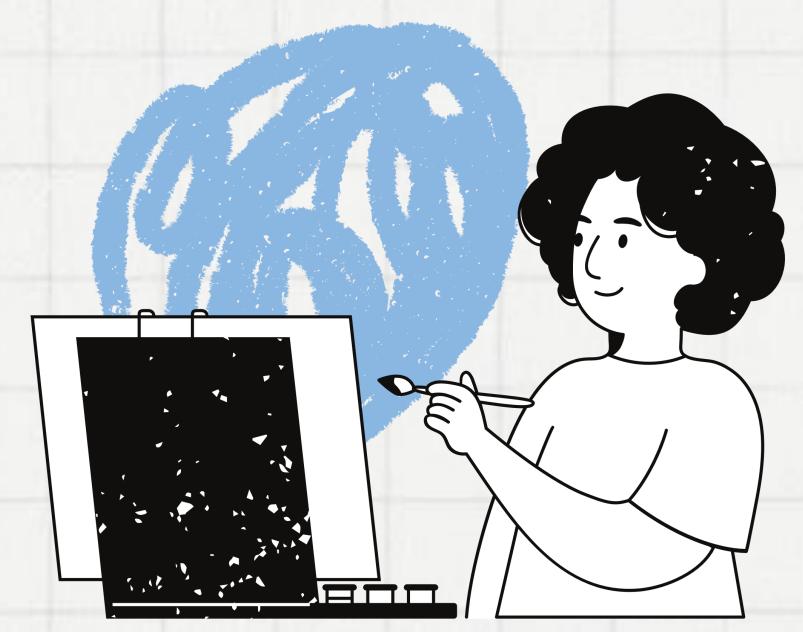


### **Neutral Review Word Cloud**



# Team members and their contribution:

- 1. Priyanshu Kumar (Team Lead):
  Sentiment analysis, model training, visualization, and reporting
- 2. Mayank Vaibhav: Data Scraping,
  Data Collection
- 3. Harshit Raj: Data Translation and Cleaning
- 4. Amit Kumar: Resources Collection





# **Conclusion and Future Work**

The sentiment analysis project for the Intel Core i7-12700K processor provides valuable insights into customer opinions and satisfaction levels across different regions. Through a comprehensive workflow involving data collection, translation, cleaning, and analysis, we were able to accurately categorize customer reviews into positive, neutral, and negative sentiments.

- **Positive Feedback:** A significant portion of reviews were positive, indicating high customer satisfaction with the Intel Core i7-12700K processor's performance and features.
- **Common Themes:** The most frequent positive comments highlighted the processor's speed, efficiency, and value for money. Negative comments were often related to pricing and occasional technical issues.
- Model Performance: The Logistic Regression model, optimized using GridSearchCV, demonstrated satisfactory accuracy in predicting review sentiments. The use of SMOTE effectively addressed class imbalance, enhancing the model's performance.
- Insights for Improvement: The analysis provides actionable insights for Intel to address common customer concerns and improve future product iterations.

# Thank you very much!