

Milestone Project 5



Customer Sentiment Analysis of Iphone 15 (128 GB)

Using

Python

By

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Customer Sentiment Analysis Report for iPhone 15 128GB (Flipkart Reviews)

Overview of Data Collection and Cleaning Process

The customer sentiment analysis was conducted on reviews for the Apple iPhone 15 (128GB, Black) model, scraped from Flipkart. **A total of 320 reviews, along with corresponding customer names and ratings, were collected using web scraping techniques involving Selenium and BeautifulSoup, please find below snippets for reference.**

```
from bs4 import BeautifulSoup
from selenium import webdriver
from selenium.webdriver.common.by import By
from selenium.webdriver.support.ui import WebDriverWait
from selenium.webdriver.support import expected_conditions as EC

# Initialize lists to store data
names = []
ratings = []
reviews = []
review_count = 0

# Flipkart product reviews URL
url = "https://www.flipkart.com/apple-iphone-15-black-128-gb/product-reviews/itm6ac6485515ae4?pid=MOBGTAGPTB3VS24W&lid=LSTMOBGTAGPTB3VS24WKFODHL&marke

# Start Chrome WebDriver
driver = webdriver.Chrome()
driver.get(url)
wait = WebDriverWait(driver, 20) # Set up an explicit wait

# Scrape multiple pages (adjust range as needed)
for page in range(1, 8000): # Increase the range to get 320 reviews
    soup = BeautifulSoup(driver.page_source, "html.parser")

    # Extract Names
    name_elements = soup.find_all("p", {"class": "_2NsDsF AwS1CA"})
    extracted_names = [name.text.strip() for name in name_elements]

    # Extract Ratings
    rating_elements = soup.find_all("div", {"class": "XQDdHH Ga3i8K"})
    extracted_ratings = [rating.text.strip() for rating in rating_elements]

    # Extract Reviews
    review_elements = soup.find_all("div", {"class": "ZmyHeo"})
    extracted_reviews = [review.text.strip() for review in review_elements]

    # Ensure equal lengths of all lists
    min_length = min(len(extracted_names), len(extracted_ratings), len(extracted_reviews))

    names.extend(extracted_names[:min_length])
    ratings.extend(extracted_ratings[:min_length])
    reviews.extend(extracted_reviews[:min_length])
    review_count += min_length # Update total review count

# Stop if 320 reviews are collected
if review_count == 320:
    break

# Click the "Next" button to go to the next page
try:
    next_button = wait.until(EC.element_to_be_clickable((By.XPATH, "/html/body/div/div/div[3]/div/div[1]/div[2]/div[2]/div")))
    next_button.click()
except Exception as e:
    print("Next button not found or last page reached.")
    break # Stop if no next button is found

# Close the browser
driver.quit()

# Print final results
print(f"\nFinal Review Count: {len(reviews)}")
print(f"Length of Names: {len(names)}")
print(f"Length of Ratings: {len(ratings)}")
print(f"Length of Reviews: {len(reviews)}")
```

Final Review Count: 320
Length of Names: 320
Length of Ratings: 320
Length of Reviews: 320

Importing data using pandas

```
import pandas as pd
df = pd.DataFrame({"Names":names, "Ratings":ratings, "Reviews":reviews})
df
```

	Names	Ratings	Reviews
0	bijaya mohanty	5	Just go for it.Amazing one.Beautiful camera wi...
1	Ajin V	5	High quality camera 📷 READ MORE
2	Flipkart Customer	5	Awesome photography experience. Battery backup...
3	Rishabh Jha	5	Awesome 🔥🔥👍 READ MORE
4	Mousam Guha Roy	4	Very niceREAD MORE
...
315	Prithivi Boruah	5	Camera Quality Is Improved Loving ItREAD MORE
316	Nikhil Kumar	5	Switch from OnePlus to iPhone I am stunned wit...
317	Akshay Meena	5	So beautiful, so elegant, just a vowww 📷❤ READ ...
318	Arunji Govindaraju	5	Awesome product very happy to hold this. Bette...
319	Sheetla Prasad Maurya	4	Best mobile phoneCamera quality is very nice B...

320 rows × 3 columns

Data cleaning and preprocessing steps included:

- Converting all review text to lowercase.

```
df["Reviews"] = df["Reviews"].str.lower()
df
```

	Names	Ratings	Reviews
0	Ajin V	5	high quality camera 📷 read more
1	Mousam Guha Roy	4	very niceread more
2	bijaya mohanty	5	just go for it.amazing one.beautiful camera wi...
3	Nikhil Kumar	5	switch from oneplus to iphone i am stunned wit...
4	Prithivi Boruah	5	camera quality is improved loving itread more
...
305	Akshay Meena	5	so beautiful, so elegant, just a vowww 📷❤ read ...
306	Flipkart Customer	5	awesome photography experience. battery backup...
307	Saurabh Gothwad	5	simply premium.read more
308	Sheetla Prasad Maurya	4	best mobile phonecamera quality is very nice b...
309	Rahul Shedge	5	totally happy!camera 5battery 5 display 5desig...

310 rows × 3 columns

- Downloading necessary NLTK (Natural Language Toolkit) resources such as punkt, stopwords, and wordnet for text processing. Handling missing values in the review column by converting them to strings and replacing NaN values with empty strings. Tokenization of sentences within the reviews.

```
import nltk
nltk.download('punkt')
nltk.download('punkt_tab')
```

```
[nltk_data] Downloading package punkt to
[nltk_data] C:\Users\sande\AppData\Roaming\nltk_data...
[nltk_data] Package punkt is already up-to-date!
[nltk_data] Downloading package punkt_tab to
[nltk_data] C:\Users\sande\AppData\Roaming\nltk_data...
[nltk_data] Unzipping tokenizers\punkt_tab.zip.
```

True

```
import nltk
from nltk.corpus import stopwords
from nltk.tokenize import sent_tokenize, word_tokenize
from textblob import TextBlob
import string
import pandas as pd

# Download necessary NLTK resources
nltk.download('stopwords')
nltk.download('punkt')
nltk.download('wordnet')

# Ensure 'Reviews' column exists and handle missing values
if 'Reviews' in df.columns:
    df['Reviews'] = df['Reviews'].astype(str).fillna('') # Convert to string and handle NaN values
    df["Reviews_t"] = df["Reviews"].apply(lambda x: sent_tokenize(x)) # Apply sentence tokenization
else:
    print("Column 'Reviews' not found in DataFrame!")
```

Sentiment Analysis Results

Sentiment analysis was performed using TextBlob, which assigned a polarity score to each review. Reviews with a polarity score of 0.1 or greater were classified as "Positive," while those below 0.1 were classified as "Negative".

The distribution of sentiments is as follows:

- A significant majority of reviews were classified as **Positive**, indicating high overall customer satisfaction.
- A smaller proportion of reviews were classified as **Negative**.

The average sentiment polarity generally correlates with the star ratings, with higher ratings (e.g., 4 and 5 stars) corresponding to more positive sentiment scores.

```

from textblob import TextBlob

# Function to get sentiment polarity
def get_sentiment_polarity(text):
    return TextBlob(text).sentiment.polarity

# Function to classify sentiment
def classify_sentiment(polarity):
    return "Positive" if polarity >= 0.1 else "Negative"

# Ensure 'Reviews' column exists and is converted to string
if 'Reviews' in df.columns:
    df['Reviews'] = df['Reviews'].astype(str).fillna('')

    # Compute sentiment polarity
    df["Polarity"] = df["Reviews"].apply(get_sentiment_polarity)

    # Classify sentiment
    df["Sentiment"] = df["Polarity"].apply(classify_sentiment)

print(df[["Reviews", "Polarity", "Sentiment"]].head(10)) # Display sample output
else:
    print("Column 'Reviews' not found in DataFrame!")

```

```

      Reviews  Polarity Sentiment
0  Just go for it.Amazing one.Beautiful camera wi...  0.344444  Positive
1      High quality camera READ MORE  0.330000  Positive
2  Awesome photography experience. Battery backup...  0.675000  Positive
3      Awesome READ MORE  0.750000  Positive
4      Very niceREAD MORE  0.350000  Positive
5      Camera Quality Is Improved Loving ItREAD MORE  0.550000  Positive
6  Switch from OnePlus to iPhone I am stunned wit...  0.750000  Positive
7  So beautiful, so elegant, just a WOWWWW READ ...  0.616667  Positive
8  Awesome product very happy to hold this. Bette...  0.573333  Positive
9  Best mobile phoneCamera quality is very nice B...  0.718000  Positive

```

Insights

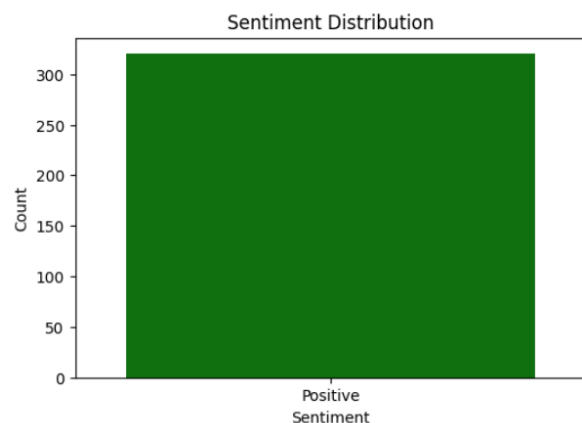
```

import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from wordcloud import WordCloud
import numpy as np

# Ensure Ratings column exists and convert to numeric
df['Ratings'] = pd.to_numeric(df['Ratings'], errors='coerce')

## Sentiment Distribution
plt.figure(figsize=(6, 4))
sns.barplot(data=sentiment_counts, x="Sentiment", y="Count", hue="Sentiment", palette={"Positive": "green", "Negative": "red"}, legend=False)
plt.xlabel("Sentiment")
plt.ylabel("Count")
plt.title("Sentiment Distribution")
plt.show()

```



Key trends and common themes from the sentiment analysis include:

- **Positive Highlights:**
 - **Camera Quality:** Many positive reviews frequently mention the "camera" and its "quality," often described as "amazing" and "awesome."

tioned words in the pos

Correlation between Ratings and Polarity:

	Ratings	Polarity
Ratings	1.000000	0.099006
Polarity	0.099006	1.000000

```

", "it", "of", "this", "that", "on",
, string.punctuation))
if w not in stopwords]))

```

```
from wordcloud import WordCloud
from sklearn.feature_extraction.text import CountVectorizer
import string

stopwords = set(["the", "is", "was", "very", "and", "to", "for", "with", "in", "it", "of", "this", "that", "on", "as"])
df['cleaned'] = df['Reviews'].str.lower().str.translate(str.maketrans('', '', string.punctuation))
df['cleaned'] = df['cleaned'].apply(lambda x: ' '.join([w for w in x.split() if w not in stopwords]))

vectorizer = CountVectorizer(max_features=100)
X = vectorizer.fit_transform(df['cleaned'])
word_freq = pd.Series(X.toarray().sum(axis=0), index=vectorizer.get_feature_names_out())
wordcloud = WordCloud(width=800, height=400).generate_from_frequencies(word_freq)

plt.imshow(wordcloud, interpolation='bilinear')
plt.axis("off")
plt.title("Customer Review Word Cloud")
plt.show()
```

[illegible]

Review Length Analysis

```
import matplotlib.pyplot as plt
import seaborn as sns

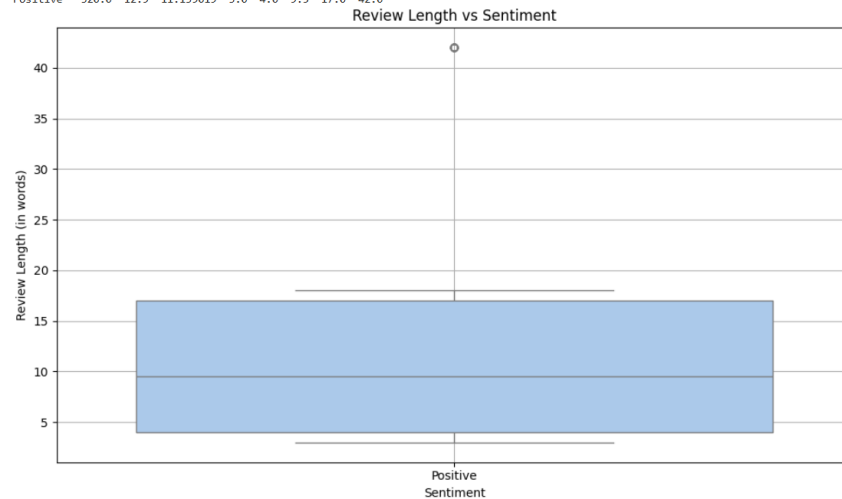
df['Review_Length'] = df['Reviews'].apply(lambda x: len(str(x).split())) #Calculating Review Length

length_stats = df.groupby('Sentiment')['Review_Length'].describe() #descriptive statistics
print("Review Length Stats by Sentiment:\n", length_stats)

# Plotting review length distribution per sentiment
plt.figure(figsize=(10, 6))
sns.boxplot(data=df, x='Sentiment', y='Review_Length', palette='pastel')
plt.title('Review Length vs Sentiment')
plt.xlabel('Sentiment')
plt.ylabel('Review Length (in words)')
plt.grid(True)
plt.tight_layout()
plt.show()
```

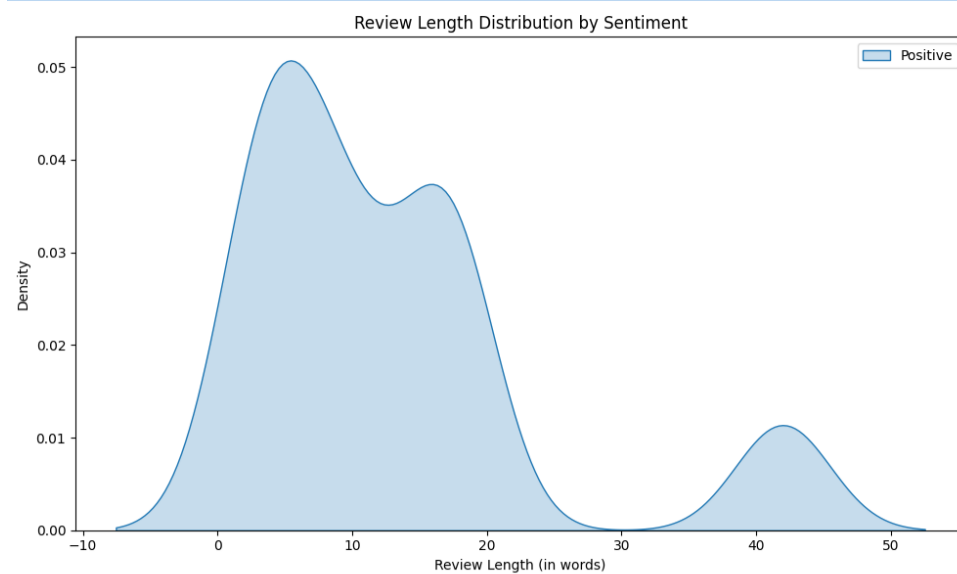
Review Length Stats by Sentiment:

	count	mean	std	min	25%	50%	75%	max
Sentiment								
Positive	320.0	12.9	11.139019	3.0	4.0	9.5	17.0	42.0



KDE plot

```
plt.figure(figsize=(10, 6))
for sentiment in df['Sentiment'].unique():
    sns.kdeplot(df[df['Sentiment'] == sentiment]['Review_Length'], label=sentiment, fill=True)
plt.title('Review Length Distribution by Sentiment')
plt.xlabel('Review Length (in words)')
plt.ylabel('Density')
plt.legend()
plt.tight_layout()
plt.show()
```



Recommendations

Based on the customer sentiment analysis, the following recommendations are suggested for the iPhone 15 128GB model and Flipkart's marketing strategies:

- **For Apple (iPhone 15 128GB Model):**
 - **Address Heating Concerns:** Investigate and optimize software or hardware to mitigate reported "heating" issues to improve overall user experience for the few negative reviews.
 - **Continue Camera Innovation:** The camera is a major positive driver; continued focus on enhancing camera features and quality will likely maintain high customer satisfaction.
 - **Monitor Performance for Edge Cases:** While generally positive, address specific "poor" performance feedback to ensure consistent quality across all devices.
- **For Flipkart (Marketing Focus):**
 - **Highlight Camera, Battery, and Display:** Emphasize the "amazing camera quality," "good battery backup," and "display" in marketing campaigns, as these are strong positive selling points identified by customers.
 - **Leverage Positive Testimonials:** Showcase reviews that specifically praise these features to build trust and attract potential buyers.
 - **Customer Support for Issues:** For the few negative reviews, Flipkart could highlight their robust return policy or customer service for addressing product issues, which may alleviate customer concerns about "heating" or other problems.