

Text Mining

Natural Language Toolkit

<https://www.nltk.org/>

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In [ ]: # Load up libraries
import pandas as pd
import string
import nltk
from nltk.corpus import stopwords
from nltk.tokenize import word_tokenize
from nltk.stem import WordNetLemmatizer
from nltk.sentiment import SentimentIntensityAnalyzer

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

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In [ ]: # Load the dataframe
review = pd.read_csv("bartini_reviews.csv")

# remove na values

review = review.dropna()

# change Date to date value

review["Date"] = pd.to_datetime(review["Date"])
review
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Tokenize

Splitting the text into individual words or "tokens"

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In [ ]: review['tokens'] = review['Review'].apply(word_tokenize)
review
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Stopwords

Removing common words that do not carry significant meaning

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In [ ]: stop_words = set(stopwords.words('english'))
review['tokens'] = review['tokens'].apply(lambda x: [word for word in x if word.lower() not in stop_words])
review
```

Remove Punctuation

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In [ ]: review['tokens'] = review['tokens'].apply(lambda x: [word for word in x if word not in string.punctuation])
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Lemmatization

Reducing words to their base or root form

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In [ ]: lemmatizer = WordNetLemmatizer()
review['tokens'] = review['tokens'].apply(lambda x: [lemmatizer.lemmatize(word) for word in x])
```

Basic Analysis

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In [ ]: !pip install textblob
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In [ ]: # Load some Libraries

from collections import Counter
import matplotlib.pyplot as plt
from textblob import TextBlob
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Word Frequency Analysis

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In [ ]: word_freq = Counter(word for sublist in review['tokens'] for word in sublist)
top_words = word_freq.most_common(10)
print("Top 10 frequent words:", top_words)
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In [ ]: # visualize

# Visualization of Word Frequencies
plt.bar(*zip(*top_words))
plt.xlabel('Words')
plt.ylabel('Frequency')
plt.title('Top 10 Frequent Words')
plt.xticks(rotation=45)
plt.show()
```

Vader Sentiment Analyzer

<https://vadersentiment.readthedocs.io/>

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In [ ]: # Initialize the VADER sentiment analyzer
sid = SentimentIntensityAnalyzer()
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In [ ]: # Function to analyze sentiment and return sentiment label
def analyze_sentiment(tokens):
    text = ' '.join(tokens)
    scores = sid.polarity_scores(text)
    if scores['compound'] >= 0.05:
        return "Positive"
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elif scores['compound'] <= -0.05:
    return "Negative"
else:
    return "Neutral"

```

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In [ ]: # Apply sentiment analysis to each row of the DataFrame
review['Opinion'] = review['tokens'].apply(analyze_sentiment)

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In [ ]: # Print the DataFrame with sentiment analysis results
print(review[['Review', 'Opinion']])

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In [ ]: #bar plot for opinion

import seaborn as sns

sns.countplot(data = review, x = "Opinion")

# Add title and labels
plt.title('Opinion Sentiment')
plt.xlabel('')
plt.ylabel('')

# Show plot
plt.show()

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In [ ]: # Apply sentiment analysis to each row of the DataFrame
review['Opinion'] = review['tokens'].apply(analyze_sentiment)

# Aggregate sentiment over time
review['month'] = review['Date'].dt.to_period('M')
sentiment_over_time = review.groupby(['month', 'Opinion']).size().unstack(fill_value=0).reset_index()

# Convert period to datetime
sentiment_over_time['month'] = sentiment_over_time['month'].dt.to_timestamp()

# Melt the DataFrame for seaborn compatibility
sentiment_melted = sentiment_over_time.melt(id_vars='month', value_vars=['Positive', 'Negative', 'Neutral'])

# Plot the sentiment trends over time using seaborn
plt.figure(figsize=(12, 6))
sns.scatterplot(data=sentiment_melted, x='month', y='Count', hue='Sentiment', style='Sentiment',
plt.title('Sentiment Analysis Over Time')
plt.xlabel('Month')
plt.ylabel('Number of Reviews')
plt.xticks(rotation=45)
plt.grid(True)
plt.show()

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In [ ]:

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