

LEVEL 3 - TASK 2: RESTAURANT REVIEWS

--2:1 Analyze the text reviews to identify the most common positive and negative keywords.

--2:2 Calculate the average length of reviews and explore if there is a relationship between review length and rating.

2:1 ANALYZE THE TEXT REVIEWS TO IDENTIFY THE MOST COMMON POSITIVE AND NEGATIVE KEYWORDS.

```
#import libraries
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import plotly.express as px

#import data
dataset= pd.read_csv("dataset.csv")

#check data
dataset.head(10)
```

	Restaurant ID	Restaurant Name	Country
0	6317637	Le Petit Souffle	
1	6304287	Izakaya Kikufuji	
2	6300002	Heat - Edsa Shangri-La	
3	6318506	Ooma	
4	6314302	Sambo Kojin	
5	18189371	Din Tai Fung	
6	6300781	Buffet 101	
7	6301290	Vikings	
8	6300010	Spiral - Sofitel Philippine Plaza Manila	
9	6314987	Locavore	

	City	Address
0	Makati City	Third Floor, Century City Mall, Kalayaan Avenu...

1	Makati City	Little Tokyo, 2277 Chino Roces Avenue, Legaspi...
2	Mandaluyong City	Edsa Shangri-La, 1 Garden Way, Ortigas, Mandal...
3	Mandaluyong City	Third Floor, Mega Fashion Hall, SM Megamall, 0...
4	Mandaluyong City	Third Floor, Mega Atrium, SM Megamall, Ortigas...
5	Mandaluyong City	Ground Floor, Mega Fashion Hall, SM Megamall, ...
6	Pasay City	Building K, SM By The Bay, Sunset Boulevard, M...
7	Pasay City	Building B, By The Bay, Seaside Boulevard, Mal...
8	Pasay City	Plaza Level, Sofitel Philippine Plaza Manila, ...
9	Pasig City	Brixton Technology Center, 10 Brixton Street, ...

	Locality \
0	Century City Mall, Poblacion, Makati City
1	Little Tokyo, Legaspi Village, Makati City
2	Edsa Shangri-La, Ortigas, Mandaluyong City
3	SM Megamall, Ortigas, Mandaluyong City
4	SM Megamall, Ortigas, Mandaluyong City
5	SM Megamall, Ortigas, Mandaluyong City
6	SM by the Bay, Mall of Asia Complex, Pasay City
7	SM by the Bay, Mall of Asia Complex, Pasay City
8	Sofitel Philippine Plaza Manila, Pasay City
9	Kapitolyo

	Locality Verbose	Longitude
Latitude \		
0	Century City Mall, Poblacion, Makati City, Mak...	121.027535
	14.565443	
1	Little Tokyo, Legaspi Village, Makati City, Ma...	121.014101
	14.553708	
2	Edsa Shangri-La, Ortigas, Mandaluyong City, Ma...	121.056831
	14.581404	
3	SM Megamall, Ortigas, Mandaluyong City, Mandal...	121.056475
	14.585318	
4	SM Megamall, Ortigas, Mandaluyong City, Mandal...	121.057508
	14.584450	
5	SM Megamall, Ortigas, Mandaluyong City, Mandal...	121.056314
	14.583764	
6	SM by the Bay, Mall of Asia Complex, Pasay Cit...	120.979667
	14.531333	
7	SM by the Bay, Mall of Asia Complex, Pasay Cit...	120.979333
	14.540000	
8	Sofitel Philippine Plaza Manila, Pasay City, P...	120.980090

14.552990

9

Kapitolyo, Pasig City 121.056532

14.572041

	Cuisines	...	Currency	\
0	French, Japanese, Desserts	...	Botswana Pula(P)	
1	Japanese	...	Botswana Pula(P)	
2	Seafood, Asian, Filipino, Indian	...	Botswana Pula(P)	
3	Japanese, Sushi	...	Botswana Pula(P)	
4	Japanese, Korean	...	Botswana Pula(P)	
5	Chinese	...	Botswana Pula(P)	
6	Asian, European	...	Botswana Pula(P)	
7	Seafood, Filipino, Asian, European	...	Botswana Pula(P)	
8	European, Asian, Indian	...	Botswana Pula(P)	
9	Filipino	...	Botswana Pula(P)	

	Has Table booking	Has Online delivery	Is delivering now	\
0	Yes	No	No	
1	Yes	No	No	
2	Yes	No	No	
3	No	No	No	
4	Yes	No	No	
5	No	No	No	
6	Yes	No	No	
7	Yes	No	No	
8	Yes	No	No	
9	Yes	No	No	

	Switch to order menu	Price range	Aggregate rating	Rating color	\
0	No	3	4.8	Dark Green	
1	No	3	4.5	Dark Green	
2	No	4	4.4	Green	
3	No	4	4.9	Dark Green	
4	No	4	4.8	Dark Green	
5	No	3	4.4	Green	
6	No	4	4.0	Green	
7	No	4	4.2	Green	
8	No	4	4.9	Dark Green	
9	No	3	4.8	Dark Green	

	Rating text	Votes
0	Excellent	314
1	Excellent	591
2	Very Good	270
3	Excellent	365
4	Excellent	229
5	Very Good	336
6	Very Good	520
7	Very Good	677
8	Excellent	621

```
9    Excellent    532
```

```
[10 rows x 21 columns]
```

```
#check database shape
```

```
dataset.shape
```

```
(9551, 21)
```

```
#check dataset information
```

```
dataset.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 9551 entries, 0 to 9550
```

```
Data columns (total 21 columns):
```

#	Column	Non-Null Count	Dtype
0	Restaurant ID	9551 non-null	int64
1	Restaurant Name	9551 non-null	object
2	Country Code	9551 non-null	int64
3	City	9551 non-null	object
4	Address	9551 non-null	object
5	Locality	9551 non-null	object
6	Locality Verbose	9551 non-null	object
7	Longitude	9551 non-null	float64
8	Latitude	9551 non-null	float64
9	Cuisines	9542 non-null	object
10	Average Cost for two	9551 non-null	int64
11	Currency	9551 non-null	object
12	Has Table booking	9551 non-null	object
13	Has Online delivery	9551 non-null	object
14	Is delivering now	9551 non-null	object
15	Switch to order menu	9551 non-null	object
16	Price range	9551 non-null	int64
17	Aggregate rating	9551 non-null	float64
18	Rating color	9551 non-null	object
19	Rating text	9551 non-null	object
20	Votes	9551 non-null	int64

```
dtypes: float64(3), int64(5), object(13)
```

```
memory usage: 1.5+ MB
```

```
#check dataset column names
```

```
dataset.columns
```

```
Index(['Restaurant ID', 'Restaurant Name', 'Country Code', 'City',  
      'Address',  
      'Locality', 'Locality Verbose', 'Longitude', 'Latitude',  
      'Cuisines',  
      'Average Cost for two', 'Currency', 'Has Table booking',  
      'Has Online delivery', 'Is delivering now', 'Switch to order  
menu',
```

```

    'Price range', 'Aggregate rating', 'Rating color', 'Rating
text',
    'Votes'],
    dtype='object')

```

Data Preprocessing

#check for null values

```
pd.isnull(dataset).sum()
```

```

Restaurant ID          0
Restaurant Name        0
Country Code          0
City                  0
Address               0
Locality              0
Locality Verbose      0
Longitude             0
Latitude              0
Cuisines               9
Average Cost for two  0
Currency              0
Has Table booking     0
Has Online delivery   0
Is delivering now     0
Switch to order menu  0
Price range           0
Aggregate rating      0
Rating color          0
Rating text           0
Votes                 0
dtype: int64

```

#drop all null values

```
dataset.dropna(inplace=True)
```

#check database

```
dataset.shape
```

```
(9542, 21)
```

```
dataset.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
Index: 9542 entries, 0 to 9550
```

```
Data columns (total 21 columns):
```

#	Column	Non-Null Count	Dtype
0	Restaurant ID	9542 non-null	int64
1	Restaurant Name	9542 non-null	object

2	Country Code	9542	non-null	int64
3	City	9542	non-null	object
4	Address	9542	non-null	object
5	Locality	9542	non-null	object
6	Locality Verbose	9542	non-null	object
7	Longitude	9542	non-null	float64
8	Latitude	9542	non-null	float64
9	Cuisines	9542	non-null	object
10	Average Cost for two	9542	non-null	int64
11	Currency	9542	non-null	object
12	Has Table booking	9542	non-null	object
13	Has Online delivery	9542	non-null	object
14	Is delivering now	9542	non-null	object
15	Switch to order menu	9542	non-null	object
16	Price range	9542	non-null	int64
17	Aggregate rating	9542	non-null	float64
18	Rating color	9542	non-null	object
19	Rating text	9542	non-null	object
20	Votes	9542	non-null	int64

dtypes: float64(3), int64(5), object(13)

memory usage: 1.6+ MB

#check description of data

dataset[['Average Cost for two', 'Price range', 'Aggregate rating', 'Votes']].describe()

	Average Cost for two	Price range	Aggregate rating
Votes			
count	9542.000000	9542.000000	9542.000000
9542.000000			
mean	1200.326137	1.804968	2.665238
156.772060			
std	16128.743876	0.905563	1.516588
430.203324			
min	0.000000	1.000000	0.000000
0.000000			
25%	250.000000	1.000000	2.500000
5.000000			
50%	400.000000	2.000000	3.200000
31.000000			
75%	700.000000	2.000000	3.700000
130.000000			
max	800000.000000	4.000000	4.900000
10934.000000			

pip install wordcloud

Requirement already satisfied: wordcloud in c:\users\dimpi\anaconda3\lib\site-packages (1.9.4)

Requirement already satisfied: numpy>=1.6.1 in c:\users\dimpi\

```
anaconda3\lib\site-packages (from wordcloud) (1.26.4)
Requirement already satisfied: pillow in c:\users\dimpi\anaconda3\lib\
site-packages (from wordcloud) (10.4.0)
Requirement already satisfied: matplotlib in c:\users\dimpi\anaconda3\
lib\site-packages (from wordcloud) (3.9.2)
Requirement already satisfied: contourpy>=1.0.1 in c:\users\dimpi\
anaconda3\lib\site-packages (from matplotlib->wordcloud) (1.2.0)
Requirement already satisfied: cycler>=0.10 in c:\users\dimpi\
anaconda3\lib\site-packages (from matplotlib->wordcloud) (0.11.0)
Requirement already satisfied: fonttools>=4.22.0 in c:\users\dimpi\
anaconda3\lib\site-packages (from matplotlib->wordcloud) (4.51.0)
Requirement already satisfied: kiwisolver>=1.3.1 in c:\users\dimpi\
anaconda3\lib\site-packages (from matplotlib->wordcloud) (1.4.4)
Requirement already satisfied: packaging>=20.0 in c:\users\dimpi\
anaconda3\lib\site-packages (from matplotlib->wordcloud) (24.1)
Requirement already satisfied: pyparsing>=2.3.1 in c:\users\dimpi\
anaconda3\lib\site-packages (from matplotlib->wordcloud) (3.1.2)
Requirement already satisfied: python-dateutil>=2.7 in c:\users\dimpi\
anaconda3\lib\site-packages (from matplotlib->wordcloud) (2.9.0.post0)
Requirement already satisfied: six>=1.5 in c:\users\dimpi\anaconda3\
lib\site-packages (from python-dateutil>=2.7->matplotlib->wordcloud)
(1.16.0)
Note: you may need to restart the kernel to use updated packages.
```

```
pip install Counter
```

```
Requirement already satisfied: Counter in c:\users\dimpi\anaconda3\
lib\site-packages (1.0.0)
Note: you may need to restart the kernel to use updated packages.
```

```
from wordcloud import WordCloud
from collections import Counter
```

```
#categorise reviews into positive,negative or neutral
positive_reviews = dataset[dataset["Rating
text"].isin(["Excellent","Very Good", "Good"])]
neutral_reviews = dataset[dataset["Rating text"]=="Average"]
negative_reviews = dataset[dataset["Rating
text"].isin(["Poor","Terrible"])]
```

```
#extract words from rating text
positive_words = " ".join(positive_reviews["Rating
text"]).lower().split()
neutral_words = " ".join(neutral_reviews["Rating
text"]).lower().split()
negative_words = " ".join(negative_reviews["Rating
text"]).lower().split()
```

```
#get the most common words
positive_common = Counter(positive_words).most_common(10)
```

```

neutral_common = Counter(neutral_words).most_common(10)
negative_common = Counter(negative_words).most_common(10)

#generate word clouds
fig, axes = plt.subplots(1,3,figsize=(18,6))
wordcloud_positive =
WordCloud(background_color="white").generate("".join(positive_words))
wordcloud_neutral =
WordCloud(background_color="white").generate("".join(neutral_words))
wordcloud_negative =
WordCloud(background_color="white").generate("".join(negative_words))

axes[0].imshow(wordcloud_positive, interpolation="bilinear")
axes[0].set_title("Positive Reviews", fontsize=14)
axes[0].axis("off")

axes[1].imshow(wordcloud_neutral, interpolation="bilinear")
axes[1].set_title("Neutral Reviews", fontsize=14)
axes[1].axis("off")

axes[2].imshow(wordcloud_negative, interpolation="bilinear")
axes[2].set_title("Negative Reviews", fontsize=14)
axes[2].axis("off")

plt.show()

#save the visualization
fig.savefig("wordcloud_visualization.png",
dpi=300,bbox_inches='tight')

#display common words in each category
positive_common, neutral_common, negative_common

```

Positive Reviews

Neutral Reviews

Negative Reviews

```

([('good', 3174), ('very', 1078), ('excellent', 300)],
 [('average', 3734)],
 [('poor', 186)])

```

2:2 CALCULATE THE AVERAGE LENGTH OF REVIEWS AND EXPLORE IF THERE IS A RELATIONSHIP BETWEEN REVIEW LENGTH AND RATING.

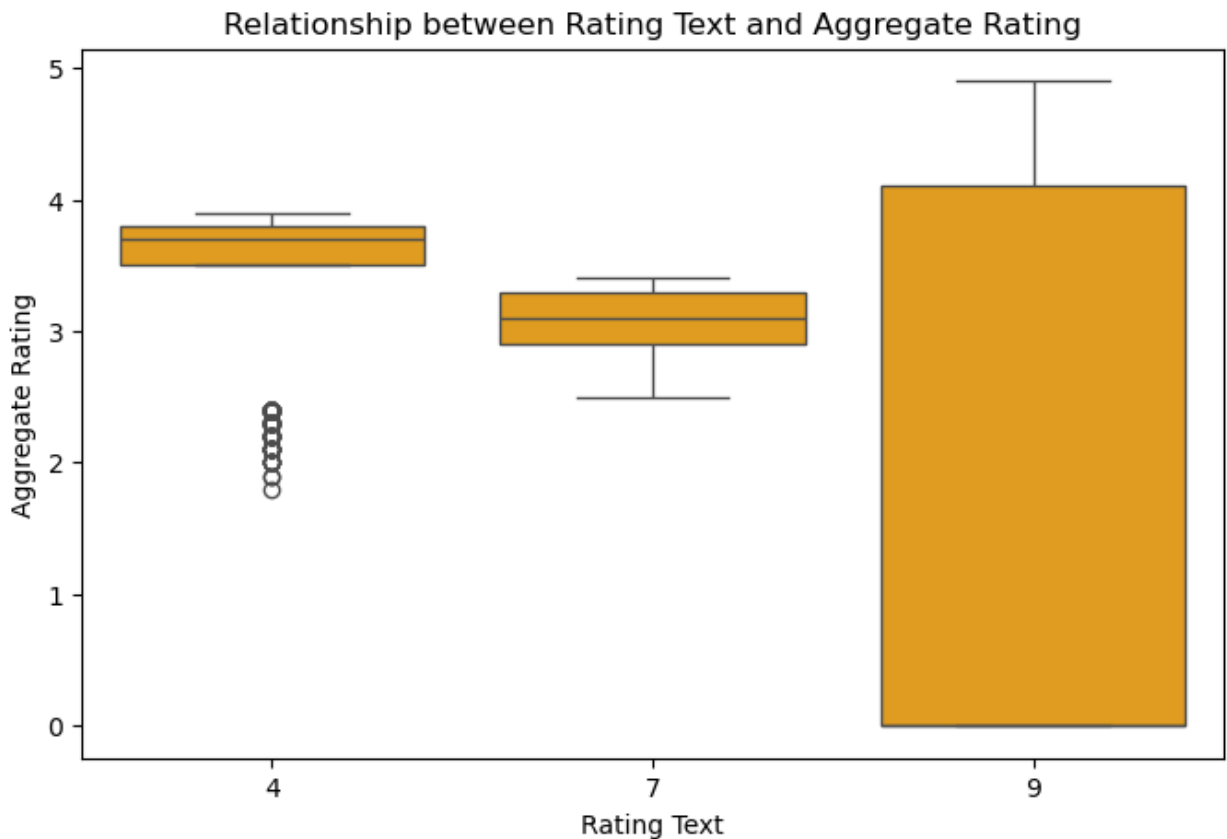

```

#visualize the relationship between rating text and aggregate rating
plt.figure(figsize=(8,5))
sns.boxplot(x="Rating text", y="Aggregate rating",
data=dataset,color="orange")
plt.title("Relationship between Rating Text and Aggregate Rating")
plt.xlabel("Rating Text")
plt.ylabel("Aggregate Rating")
plt.show()

#calculate correlation between rating text and aggregate rating
correlation= dataset["Rating text"].corr(dataset["Aggregate rating"])

#return correlation value
correlation

```



-0.47908961921650234