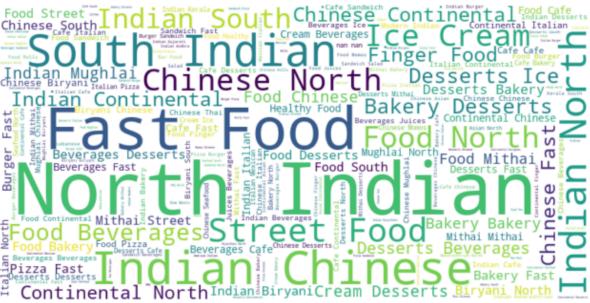
```
In [1]: import numpy as np
        import pandas as pd
        import matplotlib.pyplot as plt
        import seaborn as sns
        from textblob import TextBlob #to understand natural Language processing
        from wordcloud import WordCloud #to create wordcloud
        df = pd.read_csv("food_dely.csv")
        df.info()
      <class 'pandas.core.frame.DataFrame'>
      RangeIndex: 211944 entries, 0 to 211943
      Data columns (total 26 columns):
          Column
                                Non-Null Count
                                                Dtype
       --- -----
                                -----
           res_id
                                211944 non-null int64
       1
           name
                                211944 non-null object
       2
           establishment
                                207117 non-null object
       3
           url
                                211944 non-null object
       4
           address
                               211810 non-null object
       5
           city
                               211944 non-null object
       6
           city_id
                                211944 non-null int64
       7
           locality
                              211944 non-null
                                                object
       8
           latitude
                               211944 non-null float64
       9
           longitude
                              211944 non-null float64
       10 zipcode
                                48757 non-null
                                                object
       11 country_id
                                211944 non-null int64
       12 locality_verbose
                              211944 non-null object
       13 cuisines
                                210553 non-null
                                                object
       14 timings
                                208070 non-null object
       15 average_cost_for_two 211944 non-null int64
       16 price_range
                             211944 non-null int64
       17 currency
                               211944 non-null object
       18 highlights
                                209875 non-null object
       19 aggregate_rating
                              211944 non-null float64
       20 rating_text
                                211944 non-null
                                                object
       21 votes
                                211944 non-null int64
       22 photo_count
                                211944 non-null int64
                                211896 non-null float64
       23 opentable_support
       24 delivery
                                211944 non-null int64
                                211944 non-null int64
       25 takeaway
      dtypes: float64(4), int64(9), object(13)
      memory usage: 42.0+ MB
In [2]: df.head(2)
```

Out[2]:	I	res_id	name	establishmen	t		url	address	c
	0 3400299 Bikanervala Quick Bites https://www.zomato.com/agra/bikanervala khanda						/bikanervala- khanda	Kalyani Point, Near Tulsi Cinema, Bypass Road,	A
	1 34	00005	Mama Chicken Mama Franky House	Quick Bite	https://v s	vww.zomato.com/ chicl	'agra/mama- ken-mama	Main Market, Sadar Bazaar, Agra Cantt, Agra	A _!
	2 rows	× 26 c	olumns						
	4								
In [3]:	<pre>df.shape # Len(df) Or df.shape[0]</pre>								
Out[3]:	(211944, 26)								
In [4]:	<pre>df.describe() #Statistical analysis of numeric datatype columns.</pre>								
Out[4]:			res_id	city_id	latitude	longitude	country_id	average_	COS
	count	2.119	9440e+05 2	11944.000000	211944.000000	211944.000000	211944.0	21	119
	mean	1.349	9411e+07	4746.785434	21.499475	77.615276	1.0		5
	std	7.883	3722e+06						
	min		77220100	5568.766386	22.781261	7.500104	0.0		6
	***************************************	5.000	0000e+01	1.000000	22.781261 0.000000	7.500104 0.000000	0.0		6
	25%								2
		3.301	0000e+01	1.000000	0.000000	0.000000	1.0		
	25% 50% 75%	3.301 1.869 1.881	0000e+01 027e+06 0573e+07 297e+07	1.000000	0.000000 15.496071 22.514181 26.841214	0.000000 74.877961	1.0 1.0 1.0 1.0		2
	25% 50% 75%	3.301 1.869 1.881	0000e+01 027e+06 0573e+07 297e+07	1.000000 11.000000 34.000000	0.000000 15.496071 22.514181	0.000000 74.877961 77.425971	1.0 1.0 1.0	3	2
	25% 50% 75%	3.301 1.869 1.881	0000e+01 027e+06 0573e+07 297e+07	1.000000 11.000000 34.000000 11306.000000	0.000000 15.496071 22.514181 26.841214	0.000000 74.877961 77.425971 80.219323	1.0 1.0 1.0 1.0	3	2 4 7
In [5]:	25% 50% 75% max	3.301 1.869 1.881 1.915	0000e+01 027e+06 0573e+07 297e+07	1.000000 11.000000 34.000000 11306.000000 11354.000000	0.000000 15.496071 22.514181 26.841214	0.000000 74.877961 77.425971 80.219323	1.0 1.0 1.0 1.0	3	2 4 7
<pre>In [5]: Out[5]:</pre>	25% 50% 75% max df.du True False	3.301 1.869 1.881 1.915	0000e+01 027e+06 0573e+07 297e+07 5979e+07	1.000000 11.000000 34.000000 11306.000000 11354.000000	0.000000 15.496071 22.514181 26.841214	0.000000 74.877961 77.425971 80.219323	1.0 1.0 1.0 1.0	3	2 4 7

```
df.duplicated().value_counts()
 Out[6]: False
                   60417
         Name: count, dtype: int64
 In [7]: df.isna().sum()
 Out[7]: res_id
                                      0
                                      0
         name
          establishment
                                   1920
          url
          address
                                     18
          city
                                      0
          city_id
                                      0
         locality
                                      0
          latitude
                                      0
          longitude
                                      0
          zipcode
                                  47869
          country id
                                      0
          locality_verbose
                                      0
          cuisines
                                    470
                                   1070
         timings
                                      0
          average_cost_for_two
         price_range
                                      0
                                      0
          currency
                                    743
         highlights
                                      0
          aggregate_rating
          rating_text
                                      0
          votes
                                      0
                                      0
          photo count
          opentable support
                                     19
          delivery
                                      0
         takeaway
                                      0
         dtype: int64
 In [8]: df["res_id"] = df["res_id"].drop_duplicates().astype(int) #replace the duplicates
         df["res_id"] = df["res_id"].replace(np.nan, "Not Registered")
 In [9]: df["establishment"] = df["establishment"].replace({np.nan : df["establishment"].mod
         df["address"] = df["address"].fillna("Unknown")
In [10]: | df["zipcode"] = df["zipcode"].where(df["zipcode"].str.len()==6, "Not Provided")
In [11]: df["country_id"].value_counts()
Out[11]: country_id
               60417
          Name: count, dtype: int64
In [12]: df["cuisines"] = df["cuisines"].astype(str) #converts None, np.nan(NaN) values to
         all_cuisines = " ".join(df["cuisines"])
         career = WordCloud(width=800, height=400, background_color="White").generate(all_cu
```

```
plt.figure(figsize=(8,6))
plt.imshow(career, interpolation="bilinear")
plt.title("Wordcloud of cuisines")
plt.axis("off")
plt.show()
```

Wordcloud of cuisines



```
In [13]: df["cuisines"] = df["cuisines"].replace("nan", "North Indian, South Indian, Chinese
In [14]:
         df["timings"] = df["timings"].fillna("10 AM to 10 PM")
In [15]:
         df["average_cost_for_two"].max(), df["average_cost_for_two"].min()
Out[15]: (30000, 0)
         df.loc[df["average_cost_for_two"]<100, "average_cost_for_two"] = 100</pre>
         df.loc[df["average_cost_for_two"]>5000, "average_cost_for_two"] = 5000
In [17]: df["price_range"].value_counts()
Out[17]: price_range
               30681
          1
               18378
          2
          3
                8300
                3058
          Name: count, dtype: int64
In [18]: df["highlights"].isna().sum()
Out[18]: np.int64(743)
         df["highlights"] = df["highlights"].astype(str)
In [19]:
         all_highlights = " ".join(df["highlights"])
```

```
career = WordCloud(width=800, height=400, background_color="White").generate(all_hi
plt.figure(figsize=(8,6))
plt.imshow(career, interpolation="bilinear")
plt.title("Wordcloud of all highlights")
plt.axis("off")
plt.show()
```

Wordcloud of all highlights



```
In [20]: df["highlights"] = df["highlights"].replace("nan", " Indoor Seating, Takeaway Avail
In [21]: df["aggregate_rating"].max(), df["aggregate_rating"].min()
Out[21]: (4.9, 0.0)
In [22]: df["rating_text"].value_counts()
```

```
Out[22]: rating_text
          Good
                           17569
          Average
                           16782
          Very Good
                           12714
          Not rated
                           10160
          Excellent
                             2065
          Poor
                              590
                               56
          Çok iyi
          Sangat Baik
                               44
          Muito Bom
                               43
          Excelente
                               34
          Muy Bueno
                               33
                               29
          Bardzo dobrze
          Bom
                               26
          Skvělé
                               24
          Baik
                               24
          Velmi dobré
                               22
          İyi
                               19
          Harika
                               18
          Ottimo
                               17
          Veľmi dobré
                               16
          Buono
                               14
          Terbaik
                               14
          Skvělá volba
                               13
          Dobré
                               12
          Bueno
                               11
          Dobrze
                                9
          Wybitnie
                                8
                                8
          Eccellente
                                7
          Vynikajúce
          Průměr
                                6
                                5
          Muito bom
          Promedio
                                5
                                5
          Média
          Ortalama
                                3
                                3
          Średnio
          Media
                                3
                                3
          Priemer
          Biasa
                                2
          Scarso
          Name: count, dtype: int64
In [23]: df["votes"].mean(), df["votes"].mode()
Out[23]: (np.float64(261.4960524355728),
           Name: votes, dtype: int64)
In [24]: df["photo_count"].max(), df["photo_count"].min()
Out[24]: (17702, 0)
In [25]: df["opentable_support"] = df["opentable_support"].fillna(0)
          df["opentable_support"].value_counts()
```

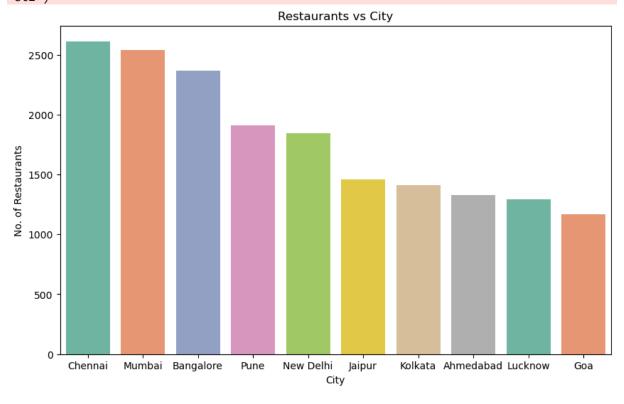
```
Out[25]: opentable support
         0.0
                60417
         Name: count, dtype: int64
In [26]: df["delivery"].value_counts()
Out[26]: delivery
               41267
         -1
          1
               18806
          0
                 344
         Name: count, dtype: int64
In [27]: df.info()
       <class 'pandas.core.frame.DataFrame'>
       Index: 60417 entries, 0 to 211942
       Data columns (total 26 columns):
            Column
                                 Non-Null Count Dtype
       ---
            -----
                                 _____
            res_id
                                 60417 non-null object
        0
        1
            name
                                 60417 non-null object
            establishment
                                 60417 non-null object
                                 60417 non-null object
        3
            url
        4
            address
                                 60417 non-null object
        5
                                 60417 non-null object
            city
        6
            city_id
                                 60417 non-null int64
        7
            locality
                                 60417 non-null object
                                 60417 non-null float64
        8
            latitude
        9
            longitude
                               60417 non-null float64
        10 zipcode
                                 60417 non-null object
        11 country_id
                                 60417 non-null int64
        12 locality_verbose
                               60417 non-null object
        13 cuisines
                                 60417 non-null object
        14 timings
                                 60417 non-null object
        15 average_cost_for_two 60417 non-null int64
        16 price_range
                                 60417 non-null int64
        17 currency
                                 60417 non-null object
        18 highlights
                               60417 non-null object
        19 aggregate_rating
                                 60417 non-null float64
        20 rating_text
                                 60417 non-null object
        21 votes
                                 60417 non-null int64
        22 photo_count
                                 60417 non-null int64
        23 opentable_support
                                 60417 non-null float64
        24 delivery
                                 60417 non-null int64
        25 takeaway
                                 60417 non-null int64
       dtypes: float64(4), int64(8), object(14)
       memory usage: 12.4+ MB
         career = df["city"].value_counts().sort_values(ascending=False).head(10)
In [28]:
         career.values
Out[28]: array([2612, 2538, 2365, 1911, 1847, 1456, 1413, 1329, 1290, 1169])
In [29]: #Data visualization of top 10 cities having max. number restaurants.
         career = df["city"].value_counts().sort_values(ascending=False).head(10)
```

```
plt.figure(figsize=(10,6))
sns.barplot(x=career.index, y=career.values, data=pd.DataFrame(career), palette="Se
plt.title("Restaurants vs City")
plt.xlabel("City")
plt.ylabel("No. of Restaurants")
plt.show()
```

C:\Users\Sunder Singh Tulera\AppData\Local\Temp\ipykernel_13836\3189720581.py:6: Fut
ureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.1 4.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

sns.barplot(x=career.index, y=career.values, data=pd.DataFrame(career), palette="S
et2")



```
In [30]: #Data visualization based on Average Cost for 2 Person.

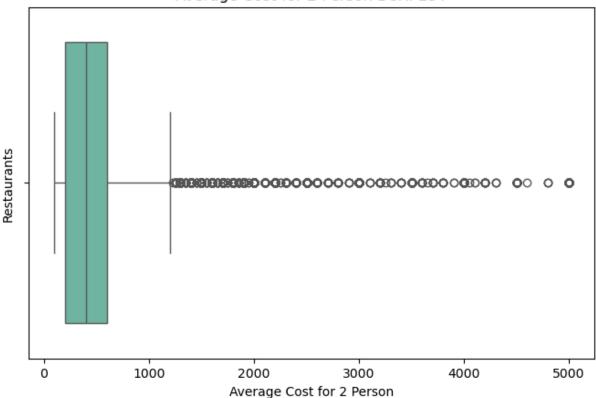
plt.figure(figsize=(8,5))
sns.boxplot(x="average_cost_for_two", data=df, palette="Set2")
plt.title("Average Cost for 2 Person BOXPLOT")
plt.xlabel("Average Cost for 2 Person")
plt.ylabel("Restaurants")
plt.show()
```

C:\Users\Sunder Singh Tulera\AppData\Local\Temp\ipykernel_13836\1124723171.py:4: Fut
ureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.1 4.0. Assign the `y` variable to `hue` and set `legend=False` for the same effect.

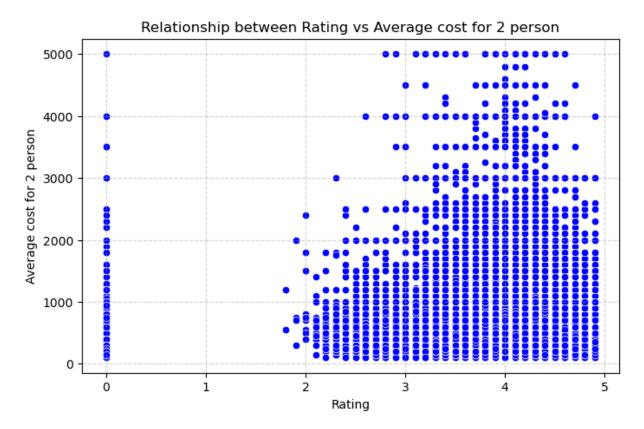
sns.boxplot(x="average_cost_for_two", data=df, palette="Set2")

Average Cost for 2 Person BOXPLOT



```
In [31]: #Data visualization of relationship between rating vs average cost for 2 person.

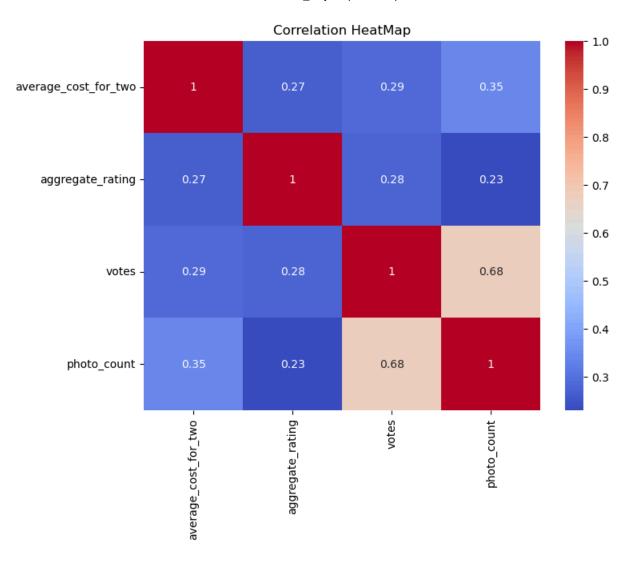
plt.figure(figsize=(8,5))
    sns.scatterplot(x="aggregate_rating", y="average_cost_for_two", data=df, color="blu plt.grid(True, linestyle="--", alpha=1)
    plt.title("Relationship between Rating vs Average cost for 2 person")
    plt.grid(True, linestyle = "--", alpha = 0.5)
    plt.xlabel("Rating")
    plt.ylabel("Average cost for 2 person")
    plt.show()
```



```
In [32]: #Correlation between average_cost_for_two, aggregate_rating, votes & photo_count.

data = df[["average_cost_for_two", "aggregate_rating", "votes", "photo_count"]]

plt.figure(figsize=(8,6))
sns.heatmap(data.corr(), cmap="coolwarm", annot=True)
plt.title("Correlation HeatMap")
plt.show()
```



```
In [33]: #Create a column region.
         med_lat = df["latitude"].median()
         med_lon = df["longitude"].median()
         def check_region(lat, lon):
              if lat==np.nan or lon==np.nan:
                  return "Unknown"
              if lat>=med_lat and lon>=med_lon:
                  return "NE"
              if lat>=med_lat and lon<=med_lon:</pre>
                  return "NW"
              if lat<=med_lat and lon<=med_lon:</pre>
                  return "SW"
              else:
                  return "SE"
         df["region"] = df.apply(lambda row : check_region(row["latitude"], row["longitude"]
         df["region"].value_counts()
```

```
Out[33]: region
    NE    16703
    SW    16702
    NW    13506
    SE    13506
    Name: count, dtype: int64

In []: df.to_csv("Mini_Project.csv", index=False)
In []:
```