ZOMATO'S EXPLORATORY DATA ANALYSIS

Importing the Dataset

dataframe = importZomatoData("zomato.csv")

dataframe = 9551×21 table

	RestaurantID	RestaurantName	CountryCode	City	Address	Locality
1	6317637	"Le Petit Souffle"	162	Makati City	"Third Floor	Century City
2	6304287	"Izakaya Kikufuji"	162	Makati City	"Little Toky	Little Tokyo,
3	6300002	"Heat - Edsa Shangri-La"	162	Mandaluyon.	."Edsa Shangr	Edsa Shangri
4	6318506	"Ooma"	162	Mandaluyon.	."Third Floor	SM Megamall,
5	6314302	"Sambo Kojin"	162	Mandaluyon.	."Third Floor	SM Megamall,
6	18189371	"Din Tai Fung"	162	Mandaluyon.	."Ground Floo	SM Megamall,
7	6300781	"Buffet 101"	162	Pasay City	"Building K,	SM by the Bay
8	6301290	"Vikings"	162	Pasay City	"Building B,	SM by the Bay
9	6300010	"Spiral - Sofitel Ph	162	Pasay City	"Plaza Level	Sofitel Phili
10	6314987	"Locavore"	162	Pasig City	"Brixton Tec	Kapitolyo
11	6309903	"Silantro Fil-Mex"	162	Pasig City	"75 East Cap	Kapitolyo
12	6309455	"Mad Mark's Creamery	162	Pasig City	"23 East Cap	Kapitolyo
13	6318433	"Silantro Fil-Mex"	162	Quezon City	"Second Floo	UP Town Cente
14	6310470	"Guevarra's"	162	San Juan C	. "387 P. Guev	Addition Hills

Converting into Categorical Arrays

dataframe.RestaurantName = categorical(dataframe.RestaurantName);
dataframe.Cuisines = categorical(dataframe.Cuisines);
dataframe

dataframe = 9551×21 table

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City with the Highest Number of Orders

City_high_num_Order = groupcounts(dataframe, "City");
City_high_num_Order = sortrows(City_high_num_Order,'GroupCount','descend')

City_high_num_Order = 141×3 table

	City	GroupCount	Percent
1	New Delhi	5473	57.3029
2	Gurgaon	1118	11.7056
3	Noida	1080	11.3077
4	Faridabad	251	2.6280
5	Ghaziabad	25	0.2618
6	Ahmedabad	21	0.2199
7	Amritsar	21	0.2199
8	Bhubaneshw	ar 21	0.2199
9	Guwahati	21	0.2199
10	Lucknow	21	0.2199
11	Abu Dhabi	20	0.2094
12	Agra	20	0.2094
13	Albany	20	0.2094
14	Allahabad	20	0.2094

```
dataframe_NewDelhi = dataframe(dataframe.City == 'New Delhi',:)
```

dataframe_NewDelhi = 5473×21 table

. . .

	RestaurantID	RestaurantName	CountryCode	City	Address	Locality
1	18287358	Food Cloud	1	New Delhi	"Aaya Nagar,	Aaya Nagar
2	18216944	Burger.in	1	New Delhi	"84, Near Ho	Adchini
3	313333	Days of the Raj	1	New Delhi	"81/3, 1st F	Adchini
4	18384127	Dilli Ka Dhaba	1	New Delhi	"66 A, Groun	Adchini
5	582	Govardhan	1	New Delhi	"84, Adjacen	Adchini
6	18414465	Mezbaan Grills	1	New Delhi	"A- 96, Shri	Adchini
7	304243	Say Cheese	1	New Delhi	"88/3, Adchi	Adchini
В	3554	Southy	1	New Delhi	"88/4, Adchi	Adchini
9	18369872	Monosoz	1	New Delhi	"Sri Aurobin	Adchini
10	948	Waves	1	New Delhi	"A-4, Sarvod	Adchini
11	2853	Delhi Darbar	1	New Delhi	"84, Main Ro	Adchini
12	18433900	Chateau	1	New Delhi	"84, Aurobin	Adchini
13	18425159	Nariyal Cafe	1	New Delhi	"A-4, 3rd Fo	Adchini
14	310958	Rustom's Parsi Bhonu	1	New Delhi	"94-A/B, Adc	Adchini

According to the data, we can say that, *NEW DELHI* is the city with the highest number of orders (*roughly around 57.3029%*)

Most Ordered Cuisines from the city with highest number of orders

```
Most_Ordered_Cuisines = groupcounts(dataframe_NewDelhi, "Cuisines");
Most_Ordered_Cuisines = sortrows(Most_Ordered_Cuisines, 'GroupCount', 'descend')
```

Most Ordered Cuisines = 892×3 table

	Cuisines	GroupCount	Percent
1	North Indian	658	12.0227
2	North Indian,	284	5.1891
3	Fast Food	242	4.4217
4	Chinese	228	4.1659
5	North Indian,	207	3.7822
6	Cafe	158	2.8869
7	Street Food	123	2.2474
8	Bakery	122	2.2291
9	North Indian,	120	2.1926
10	Bakery, Desse	117	2.1378

	Cuisines	GroupCount	Percent
11	Chinese, Fast	99	1.8089
12	Pizza, Fast F	92	1.6810
13	Mithai, Stree	90	1.6444
14	Mughlai	86	1.5714
	:		

Top 3 most ordered cuisines from New Delhi are: -

- 1. North Indian (approx. 12.02%)
- 2. A combination of North Indian and Chinese (approx 5.18%)
- 3. Fast Food (approx. 4.42)

Highest Rated Locality

Highest_Rated_Locality = groupcounts(dataframe_NewDelhi, ["Locality", "Cuisines" "AggregateRate Highest_Rated_Locality = sortrows(Highest_Rated_Locality, ['AggregateRating'],['descend'])

Highest_Rated_Locality = 5189×6 table

	Locality	Cuisines	AggregateRating	Votes	GroupCount	Percent
1	Connaught Place	Ice Cream	4.9000	2620	1	0.0183
2	Friends Colony	Modern Indian	4.9000	1934	1	0.0183
3	Janpath	Modern Indian	4.9000	408	1	0.0183
4	Rajouri Garden	North Indian	4.8000	83	1	0.0183
5	Connaught Place	Continental,	4.7000	52	1	0.0183
6	Connaught Place	North Indian	4.7000	242	1	0.0183
7	Rajouri Garden	Ice Cream	4.7000	474	1	0.0183
8	Satyaniketan	Cafe, Contine	4.7000	1563	1	0.0183
9	Select Citywa	Asian, Chines	4.7000	268	1	0.0183
10	Delhi Univers	Cafe, Contine	4.6000	1071	1	0.0183
11	Greater Kaila	Modern Indian	4.6000	304	1	0.0183
12	Khan Market	Italian, Cont	4.6000	1569	1	0.0183
13	Punjabi Bagh	North Indian,	4.6000	61	1	0.0183
14	Rajouri Garden	American, Asi	4.6000	1691	1	0.0183

Connaught Place has the highest Aggregate Rating of 4.9

Highest Votes Locality

Highest_Votes_Locality = sortrows(Highest_Rated_Locality,'Votes','descend')

Highest Votes Locality = 5189×6 table

	Locality	Cuisines	AggregateRating	Votes	GroupCount	Percent
1	Hauz Khas Vil	Continental,	4.3000	7931	1	0.0183
2	Connaught Place	South Indian	4.3000	5172	1	0.0183
3	Khan Market	Italian, Cont	4.5000	4986	1	0.0183
4	Connaught Place	American, Con	3.7000	4914	1	0.0183
5	Jama Masjid	Mughlai, Nort	4	4689	1	0.0183
6	Pandara Road	North Indian,	4.4000	4373	1	0.0183
7	Delhi Univers	Cafe, Mexican	4.3000	4085	1	0.0183
8	Vijay Nagar	Cafe, Italian	4.3000	3986	1	0.0183
9	Hauz Khas Vil	American, Nor	3.7000	3697	1	0.0183
10	Connaught Place	Bakery, Fast	4.3000	3591	1	0.0183
11	Safdarjung	North Indian,	4.3000	3530	1	0.0183
12	Lodhi Colony	American, Fas	4.1000	3495	1	0.0183
13	Connaught Place	North Indian,	3.9000	3413	1	0.0183
14	Satyaniketan	Cafe, Fast Fo	4.2000	3311	1	0.0183

Hauz Khas Village has the highest number of *votes* approx <u>7931</u> with an <u>Aggregate Rating</u> of <u>4.3</u>

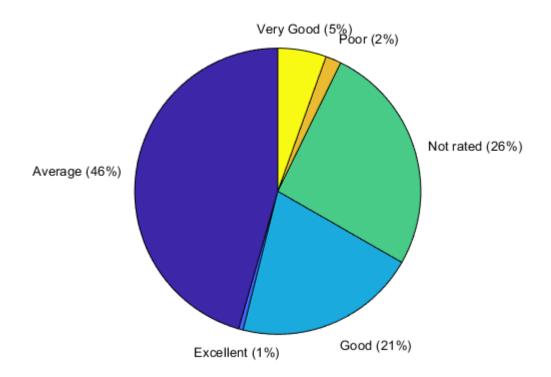
Most Rated Text

```
Most_Rated_Text = groupcounts(dataframe_NewDelhi, "RatingText");
Most_Rated_Text = Most_Rated_Text(~ismissing(Most_Rated_Text.Percent),:);
Most_Rated_Text = sortrows(Most_Rated_Text,'Percent','descend')
```

 $Most_Rated_Text = 6 \times 3 table$

	RatingText	GroupCount	Percent
1	Average	2495	45.5874
2	Not rated	1425	26.0369
3	Good	1128	20.6103
4	Very Good	300	5.4815
5	Poor	97	1.7723
6	Excellent	28	0.5116

pie(dataframe_NewDelhi.RatingText)



 $\underline{{\it "AVERAGE"}}$ is the most rated text in ${\it New Delhi}$