

Business Problem Statement:

- So these people need food categorical availability to take consideration with respect to worldwide tourist/visitor in the city.
- So successful tourist agency should suggest visitor the best restaurants according to their food choices or they should give them proper suggestions.

Audience:

 Audience for this service are the Tourists, Hotel or Restaurant owners & Native People.

Data Sources:

- GeoCoder/Google Geolocation APIs
- Data Requirement:

For this project, I will be utilizing the Foursquare API to pull the following location data on restaurants in Munich, Germany:

- Venue Name
- Venue ID
- Venue Location
- Venue Category
- Count of Likes

First we will find out likes for the restaurant:

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		Piuni	LCII_	venue	s[to	rai i	IKES_	Lac j=r	nuni	con_venues.apply(condi	cions, a	3X15=1)					
	[161]:	Muni	Munich_venues														
	[161]:							nam	e	id	l	categories	lat	Ing	total likes		
		0				•	М	arienplat	tz	4ade0ccef964a520246921e3	;	Plaza	48.137177	11.575561	4266		
		4				Ch	ocolate	rie Belug	a	4c65358a94160f47eab52e17	,	Café	48.135750	11.575776	93		
		7			Biergar	ten an	ı Viktua	alienmark	kt 4	4bbc6329afe1b7136d4d304b		Beer Garden	48.135194	11.576365	368		
		9						Kilian	1S 4	4b58d627f964a520d66d28e3		Irish Pub	48.138720	11.574556	296		
		10					Rir	ndermark	kt	4b2aa139f964a5209dac24e3		Plaza	48.136099	11.573915	102		
		11					Lo	ouis Hote	el 4	4b03b59df964a5201d5022e3		Hotel	48.135792	11.575963	55		
		12				An	dechse	r am Don	m	4ade0ca0f964a5202a6821e3		Bavarian Restaurant	48.138302	11.573778	423		
		14	1	Nürnbei	rger Br	atwurs	t Glöck	l am Don	m -	4b294373f964a520ed9b24e3		German Restaurant	48.138338	11.574180	213		
		16	ZW	ICKL G	astlich	ceit an	ı Viktua	alienmark	kt	50143667e4b0c79e483256fd		German Restaurant	48.135240	11.577249	135		
		17			B.	AR AL	PINA in	Schuste	er	5beafff4db1d81002cc32bb8		Coffee Shop	48.136362	11.574146	10		
		21					Lege	r am Don	m	5aa40abaa22db77cd6026bf3		Restaurant	48.138262	11.572932	16		
		22				Augu	stiner K	Closterwir	rt !	5238825911d2715c80955a54		German Restaurant	48.138649	11.572527	383		
		23		I	Kaffeer	östere	i Viktua	alienmark	kt !	54b93632498e7c86d18459ac		Coffee Shop	48.134732	11.576193	90		
		24						Hirme	er	4ade0d1ef964a520e26a21e3		Men's Store	48.138023	11.572046	242		
		25				Sci	nneider	Bräuhau	IS	4b44c33ef964a52078fb25e3		Bavarian Restaurant	48.136378	11.578446	640		
		27						Sabab	a 4	4b8bb1b6f964a52051a732e3		Falafel Restaurant	48.134748	11.577250	217		
		28				LUX	Restaur	rant & Ba	ar	4ff06f73f136e5332508001d		Restaurant	48.137193	11.578747	110		
		29						Ratskelle	er	4b6dcecdf964a520c3912ce3		Bavarian Restaurant	48.137583	11.576337	544		
		30						Gran	0	4b5f1e69f964a52022a729e3		Pizza Place	48.135090	11.573765	92		

Exploratory Data Analysis:

Likes as per the restaurants:

```
url_list = []
like_list = []
json_list = []

for i in venue_id_list:
    venue_url = 'https://api.foursquare.com/v2/venues/{}/likes?client_id={}&client_secret={}&v={}'.format(i, CLIENT_ID, CLIENT_SE url_list.append(venue_url)

for link in url_list:
    result = requests.get(link).json()
    likes = result['response']['likes']['count']
    like_list.append(likes)
print(like_list)
```

[4266, 93, 368, 296, 102, 55, 423, 213, 135, 10, 16, 383, 90, 242, 640, 217, 110, 544, 92, 8, 105, 102, 88, 181, 10, 13, 353, 51, 9, 17, 174, 73, 25, 67, 9, 141, 12, 37, 286, 169, 56, 6, 568, 44, 10, 254, 68, 958, 9, 3755, 161, 50, 145, 44, 328, 124, 40, 220, 70, 234, 163, 13, 324, 65, 417, 12, 231, 280]

Exploratory Data Analysis

 Following are the maximum, minimum, median & mean values of likes for all the restaurants.

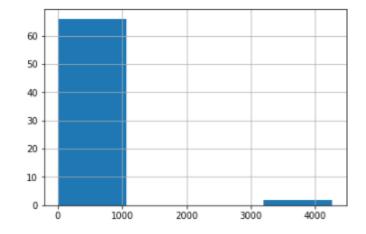
```
# now let's bin total likes
print(Munich_venues['total likes'].max())
print(Munich_venues['total likes'].min())
print(Munich_venues['total likes'].median())
print(Munich_venues['total likes'].mean())

4266
6
103.5
277.55882352941177
```

Exploratory Data Analysis

Histogram

```
import matplotlib.pyplot as plt
Munich_venues['total likes'].hist(bins=4)
plt.show()
```



Clusters-1: In this cluster we have found out restaurants with great ratings

	name	id	categories	lat	Ing	likes	likes_cat	categories_new	label
0	Marienplatz	4ade0ccef964a520246921e3	Plaza	48.137177	11.575561	4266	great	other	1
4	Chocolaterie Beluga	4c65358a94160f47eab52e17	Café	48.135750	11.575776	93	great	other	1
7	Biergarten am Viktualienmarkt	4bbc6329afe1b7136d4d304b	Beer Garden	48.135194	11.576365	368	great	bars_breverages	1
9	Kilians	4b58d627f964a520d66d28e3	Irish Pub	48.138720	11.574556	296	great	bars_breverages	1
10	Rindermarkt	4b2aa139f964a5209dac24e3	Plaza	48.136099	11.573915	102	great	other	1
12	Andechser am Dom	4ade0ca0f964a5202a6821e3	Bavarian Restaurant	48.138302	11.573778	423	great	None	1
23	Kaffeerösterei Viktualienmarkt	54b93632498e7c86d18459ac	Coffee Shop	48.134732	11.576193	90	great	other	1
24	Hirmer	4ade0d1ef964a520e26a21e3	Men's Store	48.138023	11.572046	242	great	None	1
25	Schneider Bräuhaus	4b44c33ef964a52078fb25e3	Bavarian Restaurant	48.136378	11.578446	640	great	None	1
27	Sababa	4b8bb1b6f964a52051a732e3	Falafel Restaurant	48.134748	11.577250	217	great	mex southam food	1
29	Ratskeller	4b6dcecdf964a520c3912ce3	Bavarian Restaurant	48.137583	11.576337	544	great	None	1
35	Schmalznudel - Café Frischhut	4c56c1852308be9afe84576c	Café	48.134858	11.574783	102	great	other	1
36	NORDSEE	4b72ab38f964a52001832de3	Seafood Restaurant	48.135265	11.576559	88	great	italian food	1
37	Hans im Glück - Burgergrill	5602ea37498e7034baa6958e	Burger Joint	48.136214	11.578369	181	great	mex southam food	1
41	Bar Centrale	4ade0ca8f964a5205b6821e3	Café	48.136821	11.579873	353	great	other	1
55	Max-Joseph-Platz	4be9807c1838952165b90acf	Plaza	48.139817	11.577886	141	great	other	1
59	Ohlulia	51795ac7498ee4f33h7d751d	Italian Restaurant	48.136359	11.571843	286	oreat	italian food	1

<u>Clusters-2</u>:In this cluster we have found out restaurants with poor ratings

	name	id	categories	lat	Ing	total likes	total likes_cat	categories_new	label
17	BAR ALPINA im Schuster	5beafff4db1d81002cc32bb8	Coffee Shop	48.136362	11.574146	10	poor	other	2
21	Leger am Dom	5aa40abaa22db77cd6026bf3	Restaurant	48.138262	11.572932	16	poor	euro_indian_food	2
32	Due Passi	4b9a25e4f964a52052a135e3	Italian Restaurant	48.137194	11.578709	8	poor	italian food	2
39	Elly Seidl	4ade0d26f964a5200d6b21e3	Candy Store	48.139410	11.575512	10	poor	other	2
40	Schenk's frischgepresste Säfte	4c31d79ded37a593c7dd6a03	Juice Bar	48.134753	11.576494	13	poor	None	2
44	La Piazza	597c97405455b250bd3c1700	Italian Restaurant	48.134408	11.574506	9	poor	italian food	2
45	Grapes Weinbar	585ae0fe45c3ed1e7dd99fa5	Wine Bar	48.136930	11.579391	17	poor	bars_breverages	2
53	Caspar Plautz	5a02f12a9411f217c4079100	Snack Place	48.134683	11.576058	9	poor	mex southam food	2
56	Hotel am Markt	4ade0c94f964a520ec6721e3	Hotel	48.135427	11.577078	12	poor	other	2
66	MONACO Café & franzmünchinger SHOP	59662f37b8fd9d4e19b75f9d	Coffee Shop	48.137049	11.576329	6	poor	other	2
69	Wir Machen Cupcakes	4cf3f532b3662c0f02a87b30	Cupcake Shop	48.133826	11.575904	44	below avg	other	2
70	Derag Livinghotel Am Viktualienmarkt	4ee3aec2722ea7b94470047d	Hotel	48.134218	11.575501	10	poor	other	2
75	Luigino's Feinkost	4de37f417d8b2547eb1c6760	Sandwich Place	48.134584	11.576376	9	poor	italian food	2
81	Café Nymphenburger Sekt	4cefde810acea35db5e7f0ae	Café	48.135105	11.575326	44	below avg	other	2

<u>Clusters-3</u>:In this cluster we have found out restaurants with average ratings

	name	id	categories	lat	Ing	total likes	total likes_cat	categories_new	label
11	Louis Hotel	4b03b59df964a5201d5022e3	Hotel	48.135792	11.575963	55	avg avg	other	3
43	Cortiina Hotel	4b55ecd5f964a520c8f627e3	Hotel	48.136804	11.579450	51	avg avg	other	3
48	Platzi Hotel	4b4253fff964a520add125e3	Hotel	48.137823	11.578919	73	avg avg	other	3
52	Bite Delite	56e15a45498eae4657e9617c	Café	48.139996	11.575072	67	avg avg	other	3
65	Ned Kelly's Australian Bar	4b0ec98af964a520075b23e3	Pub	48.138632	11.574618	56	avg avg	bars_breverages	3
72	Münchner Suppenküche	4b48907df964a520734f26e3	Soup Place	48.134760	11.576290	68	avg avg	other	3
78	Bar Tabacco	4b74556cf964a520ddd52de3	Cocktail Bar	48.139715	11.572672	50	avg avg	bars_breverages	3
87	Cafe Fräulein	4c5f2bbe832fa59389eaecd3	Café	48.134487	11.578152	70	avg avg	other	3
92	El Gaucho	540a202b498ed28daa9363b3	Argentinian Restaurant	48.134512	11.577664	65	avg avg	american food	3

Conclusion

- In this project we have find out restaurants with the great, average & poor ratings.
- User can categorize restaurant as per his/her need as we have also created categories.e.g.bar,euro indian food.