

ZOMATO

STRATEGIC GOAL

- *Improve customer experience.
- *Focused item promotions.
- *Business opportunities for expansion.

OBJECTIVE

- *Recommend top ranked restaurants to the customer
- *Recommend the restaurants to the customer for their menu choices
- *Analysis of menuitems across various location and recommend less/more competitive location for promotion of menu items
- *Recommend the location needs on menu item choices, restarurant choices, ... etc

User and Data Customers

Customers, Restaurant Managers, Business unit Heads

Explaratory Data Analysis

Field Definition

Each row represents the restaurant of Bangalore city in which it shows the following information about the restaurants by column wise. Given below,

URL- Website of restaurants.

Address- Address of restaurants.

Name - Name of the restaurants.

Online_order - Online_order facility present or not.

Book_table - Booking table facility present or not.

Rate - Rating of the restaurants by people.

Votes - Votes given by people for the restaurants.

Phone - Contact number of the restaurants.

Location - Location of the restaurants.

Rest_type - Type of restaurants.

Dish_liked - Dishes which are best in the restaurants.

Cuisines - Dishes by area wise whether it is north indian or south indian etc.,

Approx_cost(for two people) - Approximate cost for two people in the restaurants.

Reviews_list - Reviews given by the people for the restaurants.

Menu_items - Dishes present in the restaurants.

Listed_in(type) - Type of Dish present in the restaurants.

Listed_in(city) - City of the restaurant.

Data Cleanup

Identified the fields that are not needed for the analysis to meet the objectives. Removed Fields: URL, Address, Phone, Review list and Menu items.

Removed duplicate restaurant records. Removed empty records.

After Cleanup

IMPORTING PACKAGES

```
In [40]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
plt.style.use('ggplot')
```

IMPORTING DATASET

```
In [41]: df = pd.read_excel('Downloads\zom.xlsx')
```

```
In [42]: df.isnull().sum()
```

```
Out[42]: Unnamed: 0      0
address      0
name         0
online_order  0
book_table   0
rate         0
location     0
rest_type    0
dish_liked   0
cuisines     0
approx_cost(for two people)  0
meal_type    0
city         0
votes        0
dtype: int64
```

Describing the data

```
In [43]: df.describe()
```

```
Out[43]:
```

| | Unnamed: 0 | approx_cost(for two people) | votes |
|-------|--------------|-----------------------------|--------------|
| count | 19233.000000 | 19233.000000 | 19233.000000 |
| mean | 9616.000000 | 708.809858 | 420.982582 |
| std | 5552.233199 | 503.524302 | 699.775971 |
| min | 0.000000 | 40.000000 | 0.000000 |
| 25% | 4808.000000 | 400.000000 | 94.000000 |
| 50% | 9616.000000 | 600.000000 | 199.000000 |
| 75% | 14424.000000 | 800.000000 | 462.000000 |
| max | 19232.000000 | 6000.000000 | 16345.000000 |

In [44]: df.info

```
Out[44]: <bound method DataFrame.info of          Unnamed: 0
address \
0          0  # 31, 7th Cross, Opposite Canara Bank, Domlur,...
1          1  # 31, 7th Cross, Opposite Canara Bank, Domlur,...
2          2  # 31, 7th Cross, Opposite Canara Bank, Domlur,...
3          3  # 31, 7th Cross, Opposite Canara Bank, Domlur,...
4          4  #10061 Tower 10, Prestige Shanti Niketan., Whi...
...          ...
19228      19228  Zone By The Park, 26/A, Electronic City, Banga...
19229      19229  Zone by The Park, 26/A, Phase 1, Electronic Ci...
19230      19230  Zone by The Park, 26/A, Phase 1, Electronic Ci...
19231      19231  c/o Foodism, 2nd A Main Road, 445/31, 7th Cros...
19232      19232  c/o Foodism, 2nd A Main Road, 445/31, 7th Cros...

          name online_order book_table  rate \
0          Vinny's          Yes         No  4.1/5
1          Vinny's          Yes         No  4.1/5
2          Vinny's          Yes         No  4.1/5
3          Vinny's          Yes         No  4.1/5
4      Cupcake Couture          No         No  3.8/5
...          ...          ...         ...  ...
19228          Z-Bar          No         Yes  4.1/5
19229      Bazaar - Zone By The Park          No         Yes  4.0/5
19230      Bazaar - Zone By The Park          No         Yes  4.0/5
19231  Artinci Artisanal Ice Cream          Yes         No  4.2/5
19232  Artinci Artisanal Ice Cream          Yes         No  4.2/5

          location      rest_type \
0          Domlur  Casual Dining
1          Domlur  Casual Dining
2          Domlur  Casual Dining
3          Domlur  Casual Dining
4      Whitefield      Delivery
...          ...          ...
19228  Electronic City          Bar
19229  Electronic City  Casual Dining
19230  Electronic City  Casual Dining
19231      Marathahalli  Dessert Parlor
19232      Marathahalli  Dessert Parlor

          dish_liked \
0      Pizza, Pasta, Mocktails, Tiramisu, Garlic Brea...
1      Pizza, Pasta, Mocktails, Tiramisu, Garlic Brea...
2      Pizza, Pasta, Mocktails, Tiramisu, Garlic Brea...
3      Pizza, Pasta, Mocktails, Tiramisu, Garlic Brea...
4          Cup Cake, Chocolate Truffle
...          ...
19228  Cocktails, Tandoori Chicken, Wine, Ice Tea, Fr...
19229  Salads, Chicken Tikka, Roti, Gulab Jamun, Lunc...
19230  Salads, Chicken Tikka, Roti, Gulab Jamun, Lunc...
19231          Coffee, Vanilla Ice Cream, Almond Ice Cream
19232          Coffee, Vanilla Ice Cream, Almond Ice Cream
```

```
          cuisines \
0      Burger, Desserts, Italian, Pizza, Salad
```

```

1      Burger, Desserts, Italian, Pizza, Salad
2      Burger, Desserts, Italian, Pizza, Salad
3      Burger, Desserts, Italian, Pizza, Salad
4      Desserts
...
19228      Finger Food, North Indian
19229      Continental, South Indian, North Indian, Asian
19230      Continental, South Indian, North Indian, Asian
19231      Desserts, Ice Cream
19232      Desserts, Ice Cream

```

| | approx_cost(for two people) | meal_type | city \ |
|-------|-----------------------------|---------------|-----------------------|
| 0 | 600 | Delivery | Koramangala 6th Block |
| 1 | 600 | Delivery | Old Airport Road |
| 2 | 600 | Dine-out | Indiranagar |
| 3 | 600 | Dine-out | Old Airport Road |
| 4 | 300 | Delivery | Whitefield |
| ... | ... | ... | ... |
| 19228 | 1500 | Pubs and bars | Electronic City |
| 19229 | 1400 | Buffet | Electronic City |
| 19230 | 1400 | Dine-out | Electronic City |
| 19231 | 300 | Delivery | Marathahalli |
| 19232 | 300 | Desserts | Marathahalli |

| | votes |
|-------|-------|
| 0 | 366 |
| 1 | 367 |
| 2 | 360 |
| 3 | 366 |
| 4 | 40 |
| ... | ... |
| 19228 | 410 |
| 19229 | 415 |
| 19230 | 415 |
| 19231 | 44 |
| 19232 | 44 |

[19233 rows x 14 columns]>

Columns present in the Data

```
In [45]: df.columns
```

```
Out[45]: Index(['Unnamed: 0', 'address', 'name', 'online_order', 'book_table', 'rate',
               'location', 'rest_type', 'dish_liked', 'cuisines',
               'approx_cost(for two people)', 'meal_type', 'city', 'votes'],
              dtype='object')
```

Shape of the Data

```
In [46]: df.shape
```

```
Out[46]: (19233, 14)
```

Count of the Data

In [47]: df.count

Out[47]: <bound method DataFrame.count of Unnamed: 0

```

address \
0          0 # 31, 7th Cross, Opposite Canara Bank, Domlur,...
1          1 # 31, 7th Cross, Opposite Canara Bank, Domlur,...
2          2 # 31, 7th Cross, Opposite Canara Bank, Domlur,...
3          3 # 31, 7th Cross, Opposite Canara Bank, Domlur,...
4          4 #10061 Tower 10, Prestige Shanti Niketan., Whi...
...
19228      19228 Zone By The Park, 26/A, Electronic City, Banga...
19229      19229 Zone by The Park, 26/A, Phase 1, Electronic Ci...
19230      19230 Zone by The Park, 26/A, Phase 1, Electronic Ci...
19231      19231 c/o Foodism, 2nd A Main Road, 445/31, 7th Cros...
19232      19232 c/o Foodism, 2nd A Main Road, 445/31, 7th Cros...

name online_order book_table rate \
0          Vinny's          Yes          No 4.1/5
1          Vinny's          Yes          No 4.1/5
2          Vinny's          Yes          No 4.1/5
3          Vinny's          Yes          No 4.1/5
4          Cupcake Couture          No          No 3.8/5
...
19228          Z-Bar          No          Yes 4.1/5
19229      Bazaar - Zone By The Park          No          Yes 4.0/5
19230      Bazaar - Zone By The Park          No          Yes 4.0/5
19231      Artinci Artisanal Ice Cream          Yes          No 4.2/5
19232      Artinci Artisanal Ice Cream          Yes          No 4.2/5

location rest_type \
0          Domlur      Casual Dining
1          Domlur      Casual Dining
2          Domlur      Casual Dining
3          Domlur      Casual Dining
4          Whitefield      Delivery
...
19228      Electronic City          Bar
19229      Electronic City      Casual Dining
19230      Electronic City      Casual Dining
19231          Marathahalli      Dessert Parlor
19232          Marathahalli      Dessert Parlor

dish_liked \
0          Pizza, Pasta, Mocktails, Tiramisu, Garlic Brea...
1          Pizza, Pasta, Mocktails, Tiramisu, Garlic Brea...
2          Pizza, Pasta, Mocktails, Tiramisu, Garlic Brea...
3          Pizza, Pasta, Mocktails, Tiramisu, Garlic Brea...
4          Cup Cake, Chocolate Truffle
...
19228      Cocktails, Tandoori Chicken, Wine, Ice Tea, Fr...
19229      Salads, Chicken Tikka, Roti, Gulab Jamun, Lunc...
19230      Salads, Chicken Tikka, Roti, Gulab Jamun, Lunc...
19231          Coffee, Vanilla Ice Cream, Almond Ice Cream
19232          Coffee, Vanilla Ice Cream, Almond Ice Cream

```

```

cuisines \
0          Burger, Desserts, Italian, Pizza, Salad

```

```

1      Burger, Desserts, Italian, Pizza, Salad
2      Burger, Desserts, Italian, Pizza, Salad
3      Burger, Desserts, Italian, Pizza, Salad
4      Desserts
...
19228      Finger Food, North Indian
19229      Continental, South Indian, North Indian, Asian
19230      Continental, South Indian, North Indian, Asian
19231      Desserts, Ice Cream
19232      Desserts, Ice Cream

```

```

      approx_cost(for two people)      meal_type      city \
0      600      Delivery      Koramangala 6th Block
1      600      Delivery      Old Airport Road
2      600      Dine-out      Indiranagar
3      600      Dine-out      Old Airport Road
4      300      Delivery      Whitefield
...
19228      1500      Pubs and bars      Electronic City
19229      1400      Buffet      Electronic City
19230      1400      Dine-out      Electronic City
19231      300      Delivery      Marathahalli
19232      300      Desserts      Marathahalli

```

```

      votes
0      366
1      367
2      360
3      366
4      40
...
19228      410
19229      415
19230      415
19231      44
19232      44

```

```
[19233 rows x 14 columns]>
```

Datatype of each column


```
In [48]: df.dtypes
```

```
Out[48]: Unnamed: 0          int64  
address          object  
name             object  
online_order     object  
book_table       object  
rate            object  
location         object  
rest_type        object  
dish_liked       object  
cuisines         object  
approx_cost(for two people)  int64  
meal_type        object  
city            object  
votes           int64  
dtype: object
```

Count of the ratings

Removing the null values present in the rating column.

```
In [49]: df['rate'].value_counts()
df['rate'].isnull().sum()
df['rate'] = df['rate'].apply(lambda x: str(x).split('/')[0])
df['rate'].value_counts()
```

```
Out[49]: 3.9      3012
4.0      2629
4.1      2463
3.8      2288
4.2      1698
3.7      1448
4.3      1148
3.6       762
4.4       598
3.5       424
3.4       320
2.8       286
2.9       261
4.5       259
3.1       237
3.0       236
3.2       209
3.3       190
2.7       162
NEW       135
4.6       119
2.6       112
4.7        53
2.5        44
2.4        44
2.3        33
2.2        23
2.1        11
2.0        11
4.8        10
4.9         8
Name: rate, dtype: int64
```

```

In [50]: g23 = list(df['rate'])
for i in range(0, len(g23)):
    if g23[i] == 'nan':
        g23[i] = 'unrated'
    elif g23[i] == '-':
        g23[i] = 'unrated'
    elif g23[i] == 'NEW':
        g23[i] = 'unrated'

df['rate'] = g23

df['rate'].value_counts()

g23 = list(df['rate'])
for i in range(0, len(g23)):
    if g23[i] == 'unrated':
        g23[i] = None
    else :
        g23[i] = float(g23[i])
df['rate'] = g23
df['rate'].value_counts()

```

```

Out[50]: 3.9    3012
         4.0    2629
         4.1    2463
         3.8    2288
         4.2    1698
         3.7    1448
         4.3    1148
         3.6     762
         4.4     598
         3.5     424
         3.4     320
         2.8     286
         2.9     261
         4.5     259
         3.1     237
         3.0     236
         3.2     209
         3.3     190
         2.7     162
         4.6     119
         2.6     112
         4.7      53
         2.5      44
         2.4      44
         2.3      33
         2.2      23
         2.0      11
         2.1      11
         4.8      10
         4.9       8
         Name: rate, dtype: int64

```

Count of city

```
In [51]: df['city'].value_counts()
```

```
Out[51]: BTM 1225
Koramangala 7th Block 1132
Koramangala 4th Block 1122
Koramangala 5th Block 1047
Koramangala 6th Block 979
Jayanagar 925
Indiranagar 879
Church Street 802
Brigade Road 779
MG Road 734
JP Nagar 689
Lavelle Road 684
Residency Road 625
HSR 611
Old Airport Road 599
Basavanagudi 562
Brookefield 522
Whitefield 517
Marathahalli 496
Bannerghatta Road 482
Frazer Town 477
Kalyan Nagar 453
Malleshwaram 436
Kammanahalli 434
Bellandur 424
Sarjapur Road 404
Banashankari 347
Rajajinagar 314
Electronic City 288
New BEL Road 245
Name: city, dtype: int64
```

```
In [52]: df['location'].isnull().sum()
```

```
Out[52]: 0
```

```
In [53]: len(df['city'].value_counts())
```

```
Out[53]: 30
```

UNIVARIATE ANALYSIS

```
In [54]: # plotting number of restaurants that take online_order
sns.set_context("paper", font_scale = 2, rc = {"font.size": 20, "axes.titlesize":
sns.catplot(data = df, kind = 'count', x = 'online_order')
plt.title('Number of restaurants that take order online')
plt.show()
```

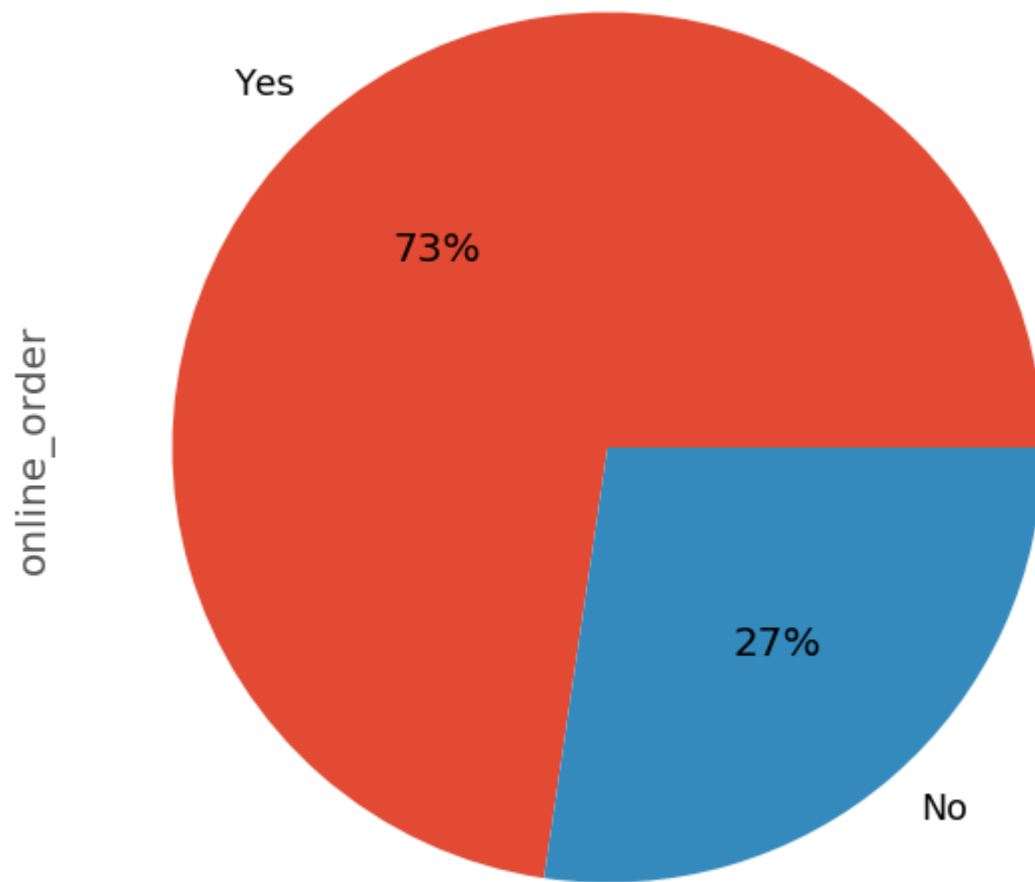
Number of restaurants that take order online



Type *Markdown* and LaTeX: α^2

```
In [55]: df['online_order'].value_counts().plot(kind='pie', figsize=(10,10), autopct='%1.0
```

```
Out[55]: <matplotlib.axes._subplots.AxesSubplot at 0x1e418378208>
```



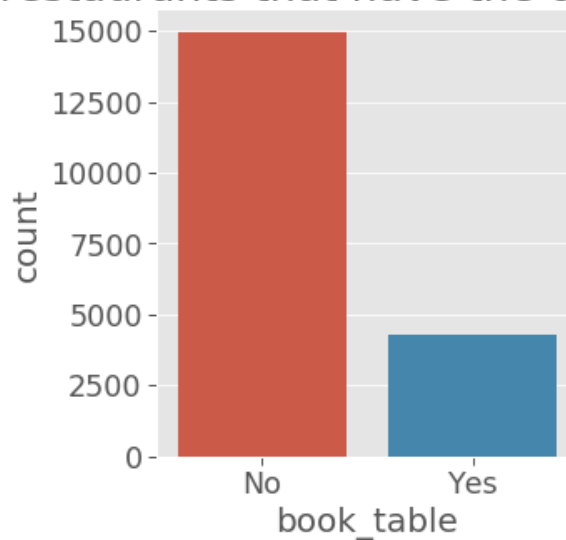
Conclusion

The IT field has the most value in developing an application for the hotels which are not taking online orders.

and mostly Bangalore people takes online order.

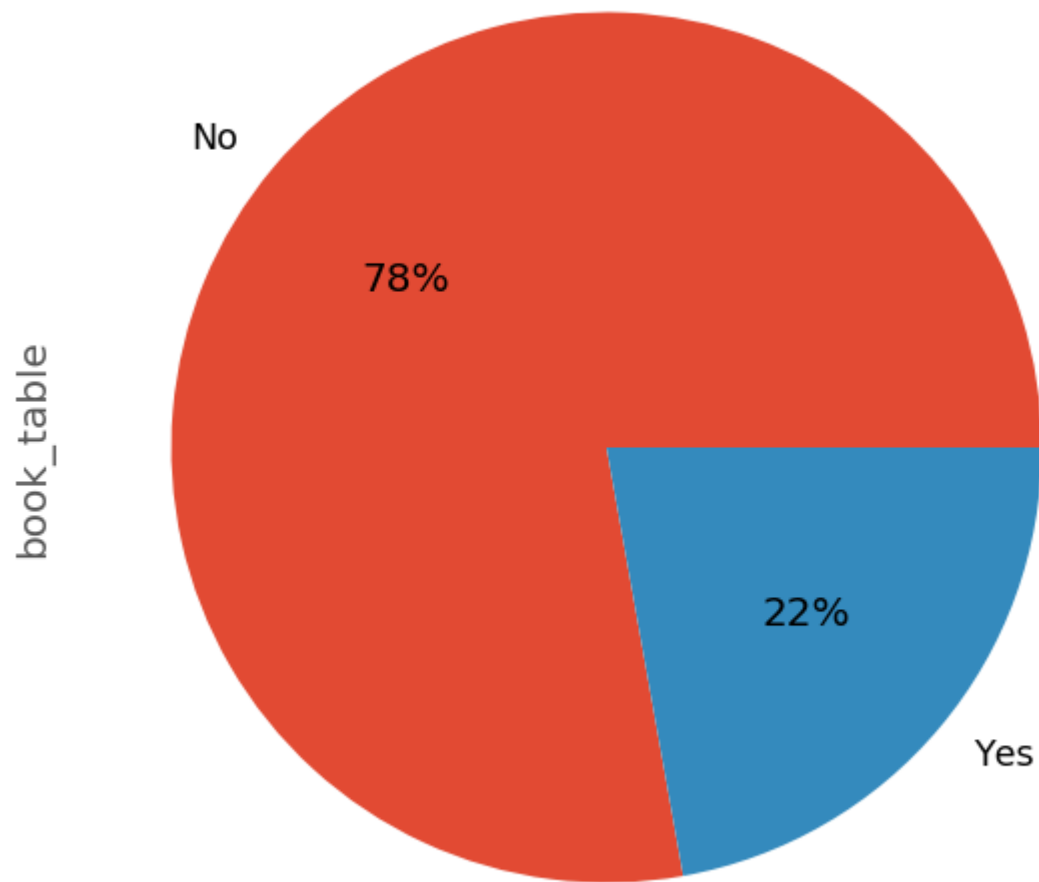
```
In [56]: sns.set_context("paper", font_scale = 2, rc = {"font.size": 20, "axes.titlesize":  
sns.catplot(data = df, kind = 'count', x = 'book_table')  
plt.title('Number of restaurants that have the option to book table')  
plt.show()
```

Number of restaurants that have the option to book table



```
In [57]: df['book_table'].value_counts().plot(kind='pie', figsize=(10,10),autopct='%1.0f%%')
```

```
Out[57]: <matplotlib.axes._subplots.AxesSubplot at 0x1e415de1a48>
```

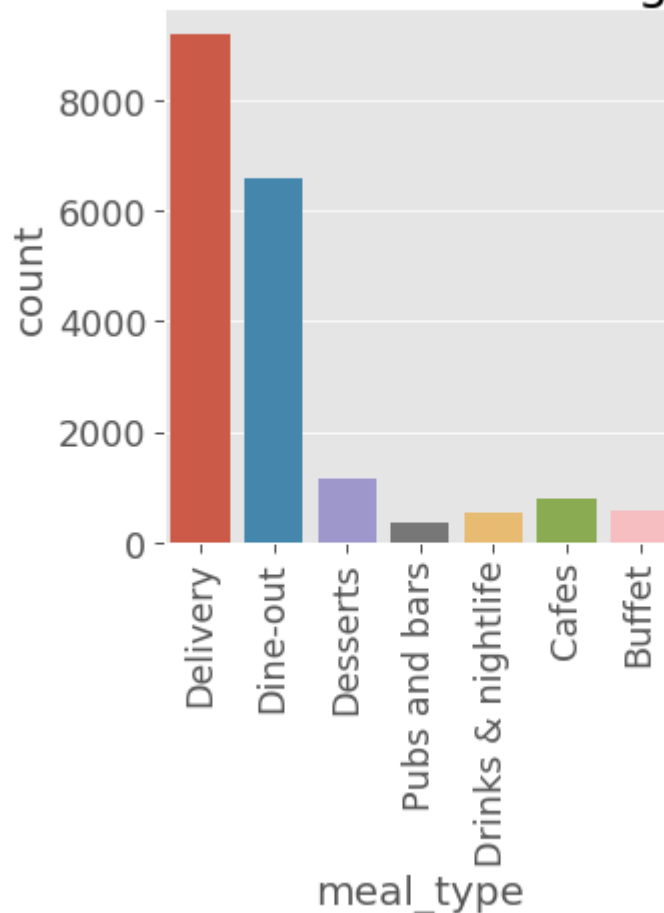


Conclusion

We see that very high number of restaurants does not offer the service to book table, and very less do. This means that people at Bangalore prefer to eat at their homes or prefer fast food (snacks — quick bites)

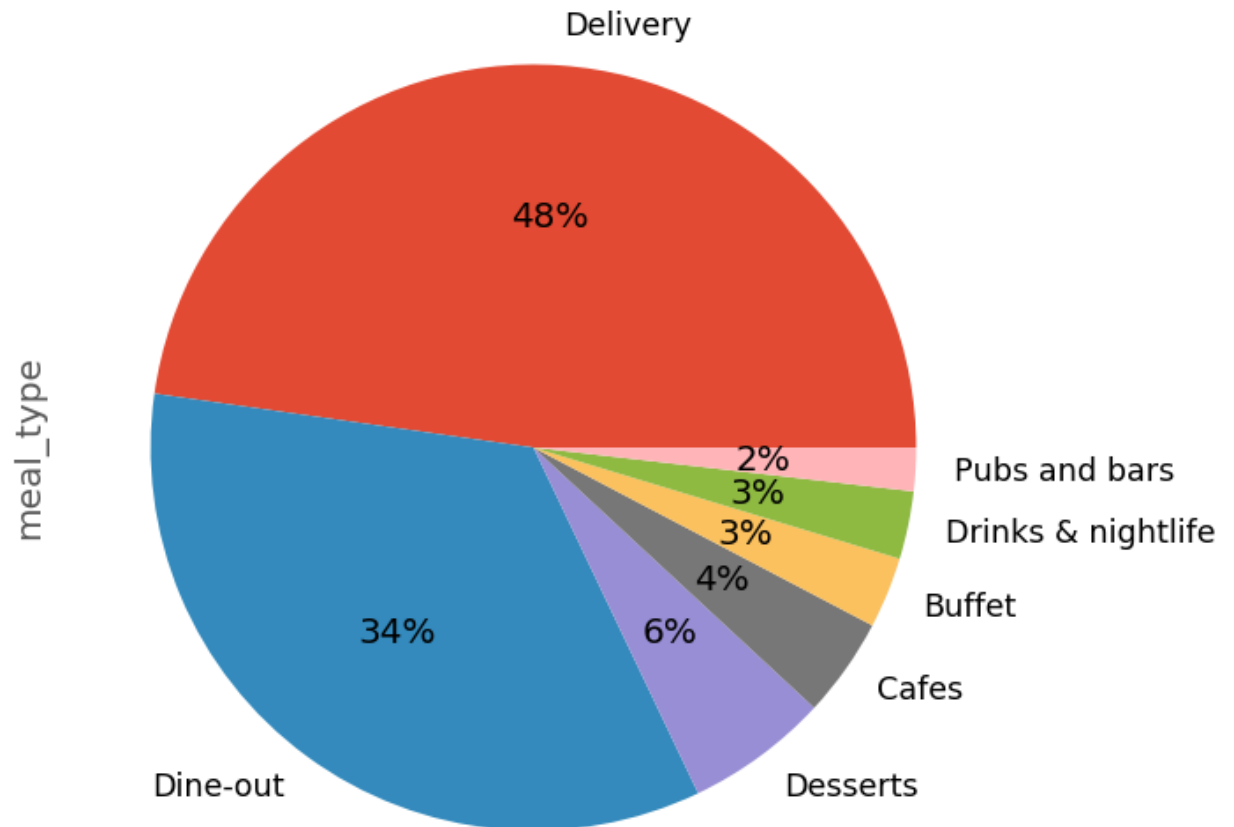
```
In [58]: sns.set_context("paper", font_scale = 2, rc = {"font.size": 20, "axes.titlesize":  
sns.catplot(data = df, kind = 'count', x = 'meal_type')  
plt.title('Number of restaurants according to meal type')  
plt.xticks(rotation=90)  
plt.show()
```

Number of restaurants according to meal type



```
In [59]: df['meal_type'].value_counts().plot(kind='pie', figsize=(10,10), autopct='%1.0f%%')
```

```
Out[59]: <matplotlib.axes._subplots.AxesSubplot at 0x1e4198e2908>
```



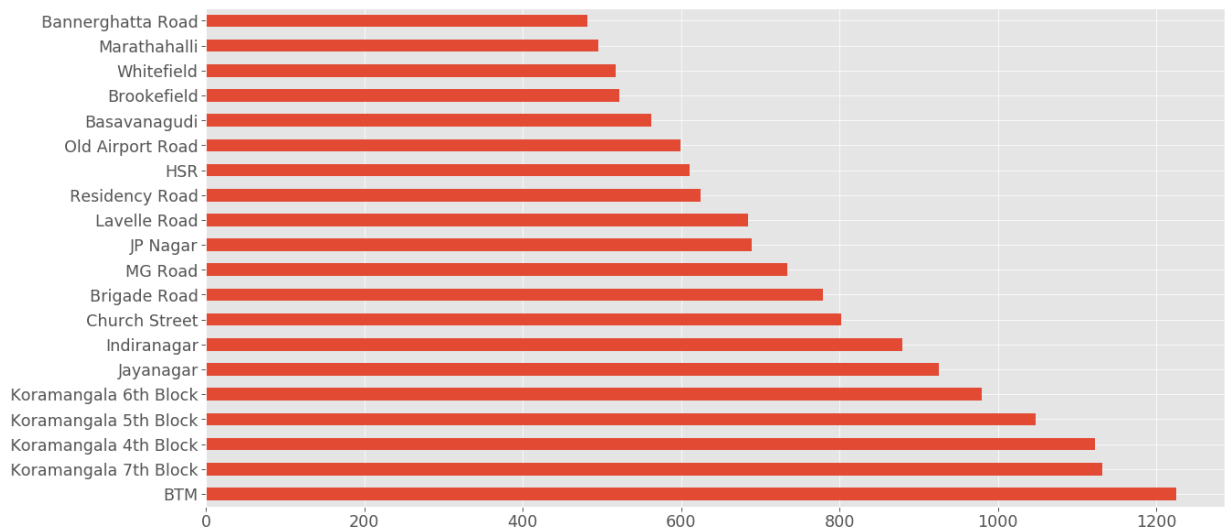
Conclusion

***We see that the majority of restaurants offer delivery, which might account for the above conclusion that people at Bangalore prefer to eat at home.**

***Pubs and Bars and Buffet are offered by very less number of restaurants.**

```
In [60]: df['city'].value_counts().head(20).plot(kind='barh', figsize=(20,10))
```

```
Out[60]: <matplotlib.axes._subplots.AxesSubplot at 0x1e419812a48>
```



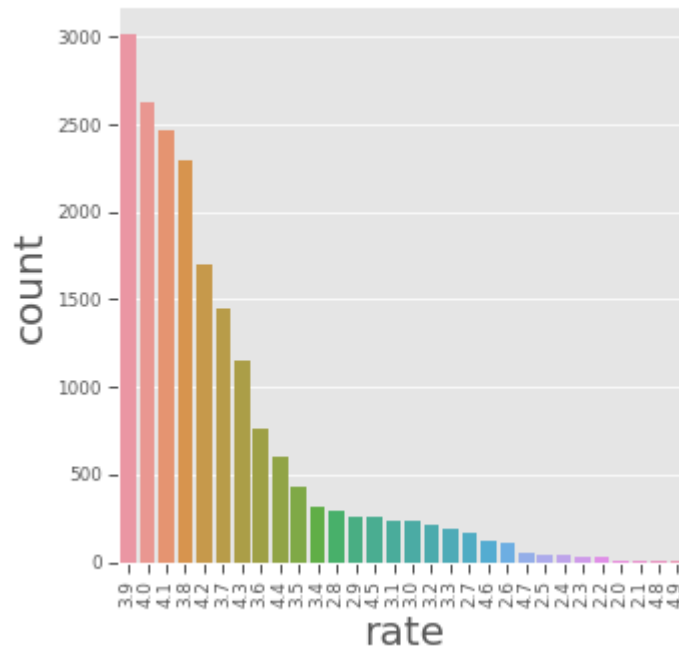
Conclucision

***BTM has the highest number of restaurants, followed by Koramangala 7th Block. New BEL Road has the least number of restaurants, followed by Banashankari.**

***It seems that the main foodies live in BTM and Koramangala.**

```
In [61]: sns.set_context("paper", font_scale = 1, rc = {"font.size": 20, "axes.titlesize":
g23 = sns.catplot(data = df, kind = 'count', x = 'rate', order = df['rate'].value
plt.title('Number of restaurants for each rating')
g23.set_xticklabels(rotation = 90)
plt.show()
```

Number of restaurants for each rating



Conclucision

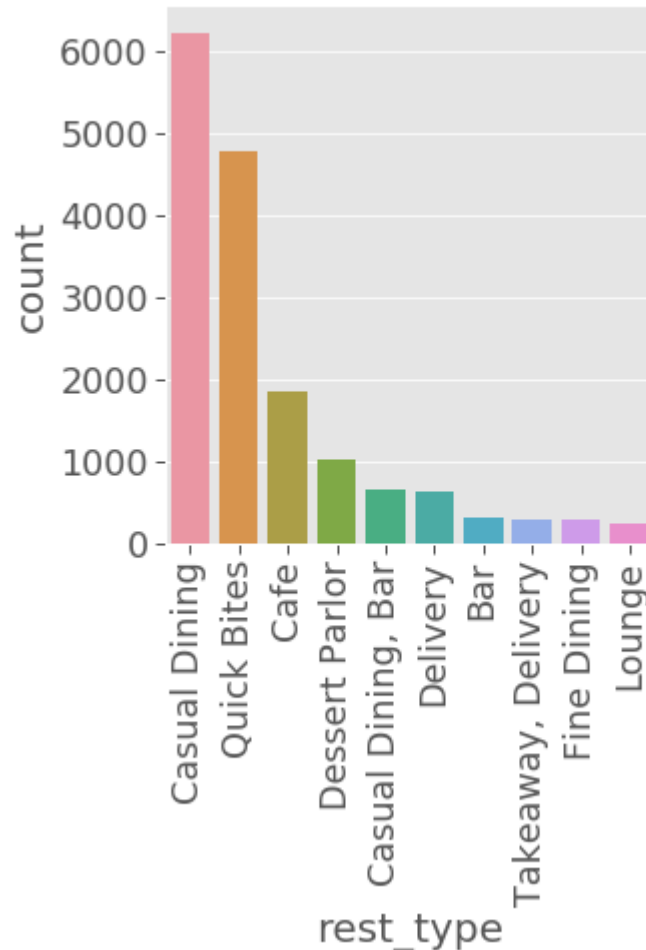
***The majority of restaurants are rated 3.9(out of 5), followed by 3.8 and 3.7, which are decent ratings.**

***This means that most of the restaurants at Bangalore are liked by the citizens and hence rated above average**

***This shows that due to high rise in the number of restaurants in Bangalore, the competition is at its best, as every restaurant is trying to excel in quality and services to increase the number of customers, and hence enhance profit**

```
In [62]: sns.set_context("paper", font_scale = 2, rc = {"font.size": 20, "axes.titlesize":  
b = sns.catplot(data = df, kind = 'count', x = 'rest_type', order = df['rest_type'  
plt.title('Number of restaurants for each type')  
b.set_xticklabels(rotation = 90)  
plt.show()
```

Number of restaurants for each type



Conclucision

***We see that, the most number of restaurants are of the type casual dining.**

***From the above plot it is clear that people at Bangalore prefer 'quick bites' more often.**

***This supports our above conclusion, which was drawn from the count plot of book_table feature, that citizens of Bangalore prefer fast food-quick bites more often, this is why many restaurants do not offer table booking facility**

```
In [63]: df['rest_type'].value_counts()
```

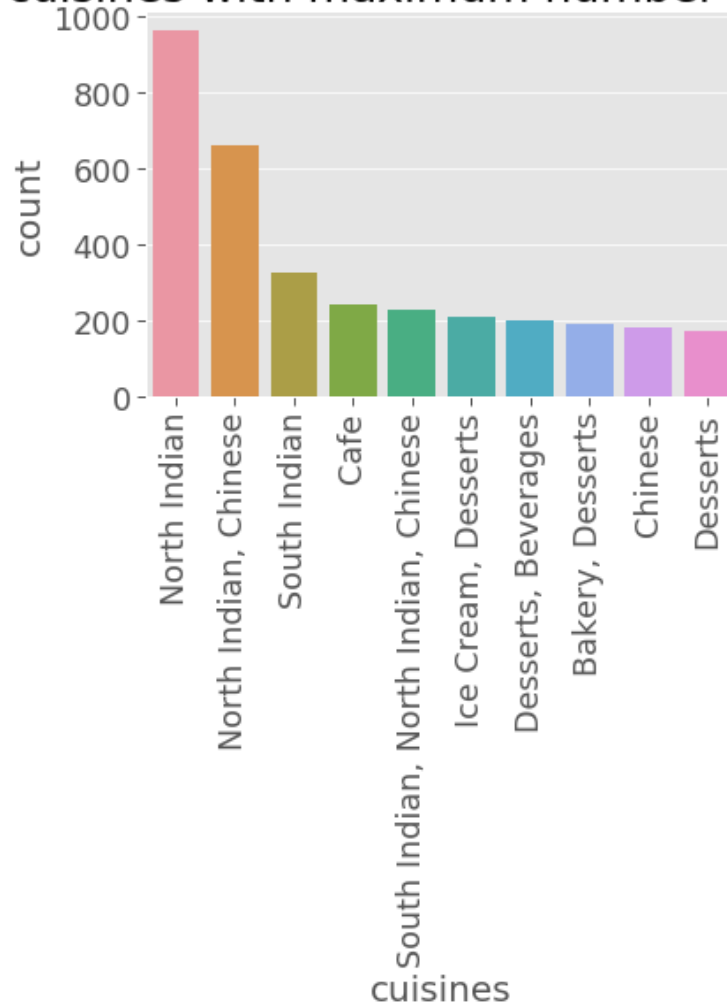
```
Out[63]: Casual Dining          6221
Quick Bites          4779
Cafe                 1839
Dessert Parlor       1005
Casual Dining, Bar    641
...
Dessert Parlor, Kiosk      2
Microbrewery, Bar         2
Cafe, Lounge              2
Dhaba                     1
Food Court, Casual Dining  1
Name: rest_type, Length: 73, dtype: int64
```

```
In [64]: df['cuisines'].value_counts()
```

```
Out[64]: North Indian          964
North Indian, Chinese          658
South Indian                   325
Cafe                          243
South Indian, North Indian, Chinese 227
...
Chinese, Vietnamese, Thai, Malaysian      1
Coffee, South Indian                      1
Biryani, Andhra, North Indian, Seafood     1
Biryani, Kebab, Continental               1
Cafe, Healthy Food, North Indian, Biryani, Continental, Desserts 1
Name: cuisines, Length: 1602, dtype: int64
```

```
In [65]: b = sns.countplot(data = df, x = 'cuisines', order = df.cuisines.value_counts().\n      b.set_xticklabels(b.get_xticklabels(),rotation = 90)\n      plt.title('Top 10 cuisines with maximum number of restaurants')\n      plt.show()
```

Top 10 cuisines with maximum number of restaurants

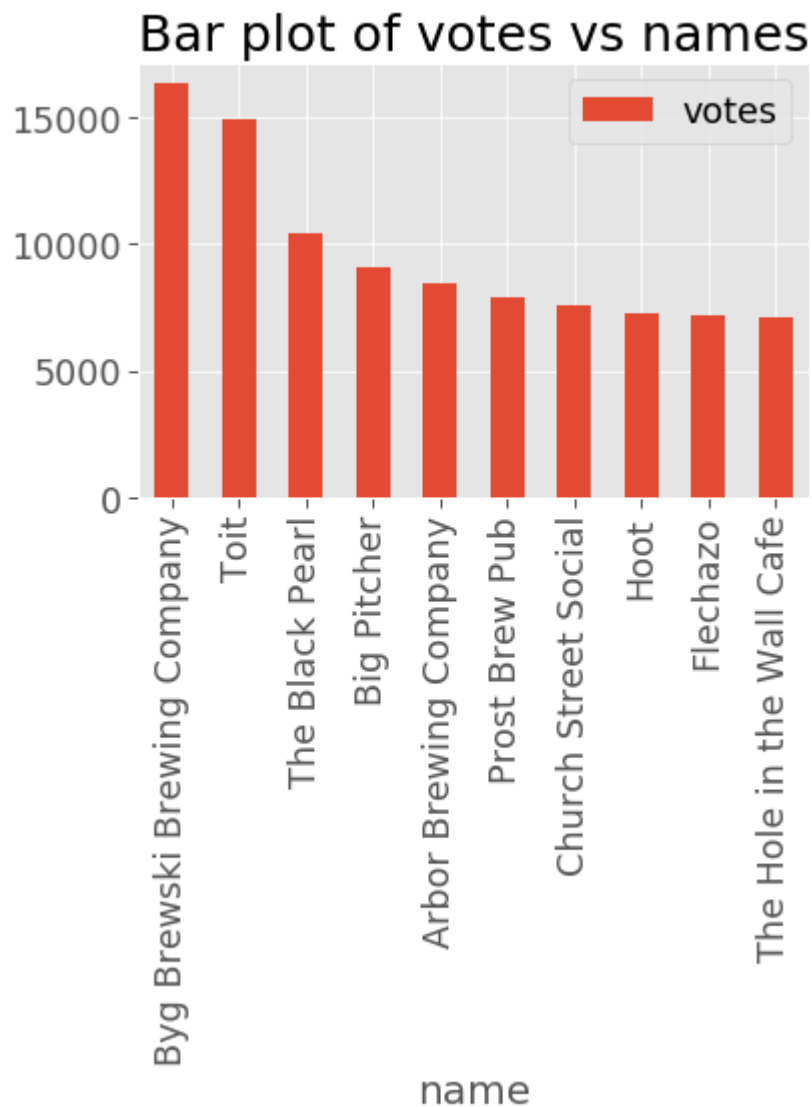


Conlcusion

*We see that North Indian cuisine is offered by most number of restaurants.

*This is worth discussing, as Bangalore is situated in South India and yet people there prefer to eat North Indian cuisine

```
In [66]: df[['votes', 'name']].groupby(['name']).median().sort_values("votes", ascending :  
plt.title('Bar plot of votes vs names')  
plt.show()
```



Conlcusion

*We see that Byg Brewski Brewing Company restaurant has the maximum number of upvote.

*We see that this restaurant has outstanding rating, 4.9 out of 5.0.

*The restaurant is a Microbrewery type of restaurant, and offers the above mentioned cuisines.

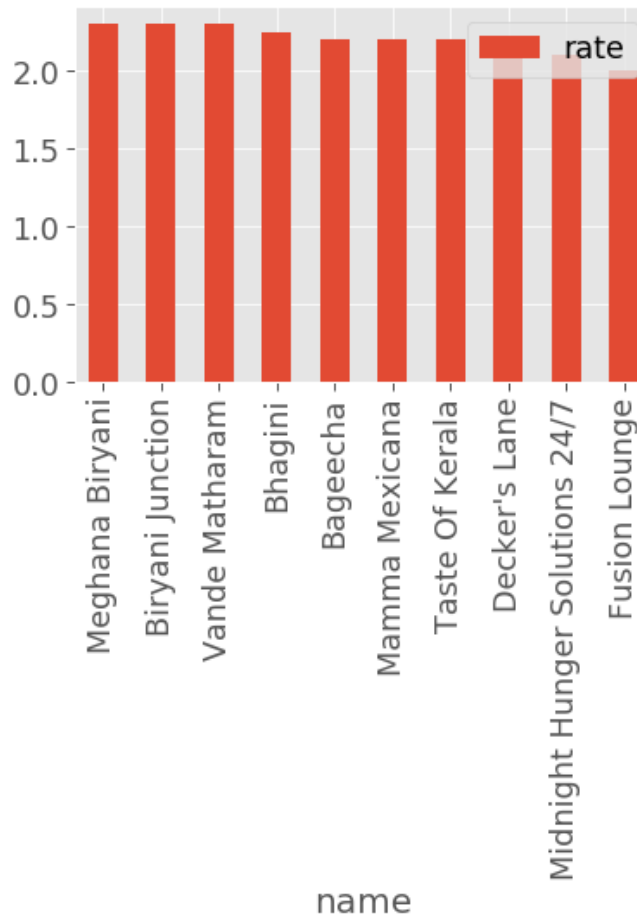
*The dishes liked by the people are also mentioned.

*We see that the meal type for different branches of the restaurant include Delivery, Dine-out and Drinks & nightlife, this means that Byg Brewski Brewing Company restaurant is one of those strong competitors in these types of meals.

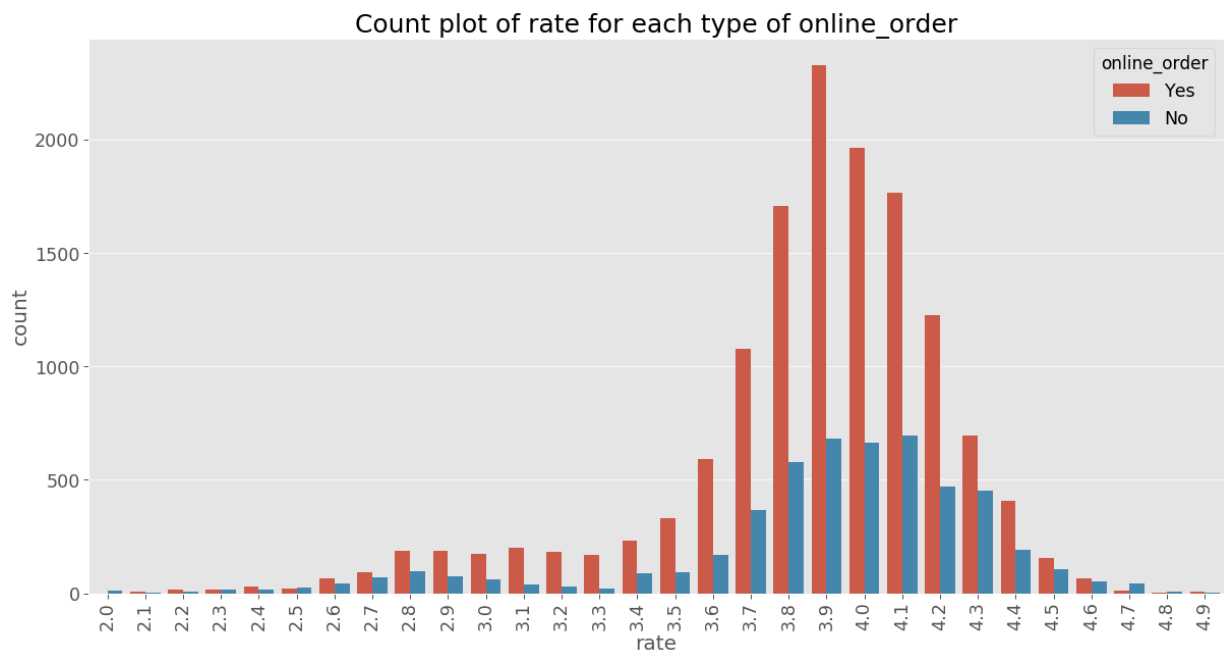
```
In [67]: b=[]
a = df.iloc[:, :].values
for i in range(0, len(a)):
    if a[i][0] == 'Byg Brewski Brewing Company' or a[i][0] == 'Byg Brewski Brewin
        b.append(list(a[i]))
```

```
In [68]: a = pd.DataFrame(df['rate'])
a['name'] = df['name']
a = a.dropna(axis = 0, how = 'any')
a[['rate', 'name']].groupby(['name']).median().sort_values("rate", ascending = F)
plt.title('Bar plot of rate vs names for 10 least rated restaurants')
plt.show()
```

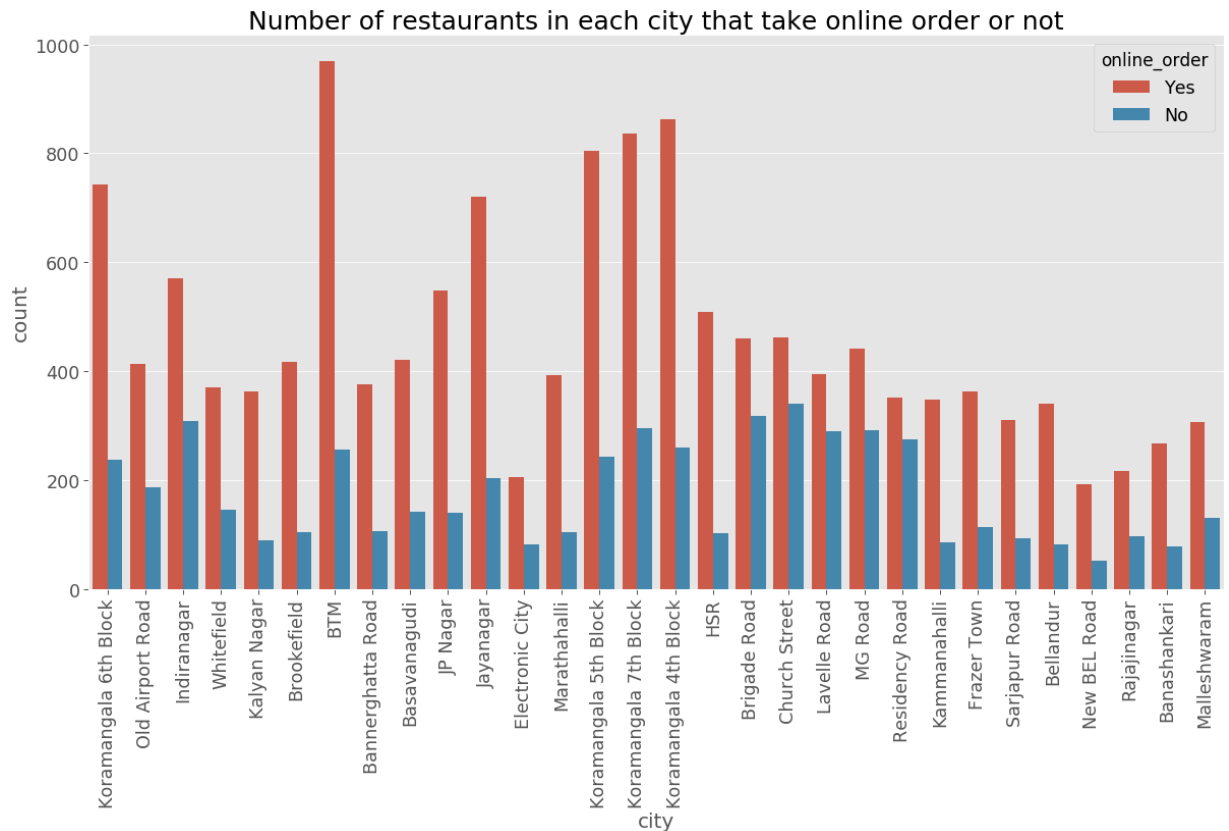
Bar plot of rate vs names for 10 least rated restaurants



```
In [69]: plt.figure(figsize=(20,10))
b=sns.countplot(data = df, hue = 'online_order', x = 'rate')
plt.title('Count plot of rate for each type of online_order')
plt.xticks(rotation = 90)
plt.show()
```



```
In [70]: plt.figure(figsize=(20,10))
sns.set_context("paper", font_scale = 2, rc = {"font.size": 20,"axes.titlesize":
b = sns.countplot(data = df, x = 'city', hue = 'online_order')
plt.title('Number of restaurants in each city that take online order or not')
b.set_xticklabels(b.get_xticklabels(),rotation = 90)
plt.show()
```



We see that BTM has the highest number of restaurants where you can or can not book table.

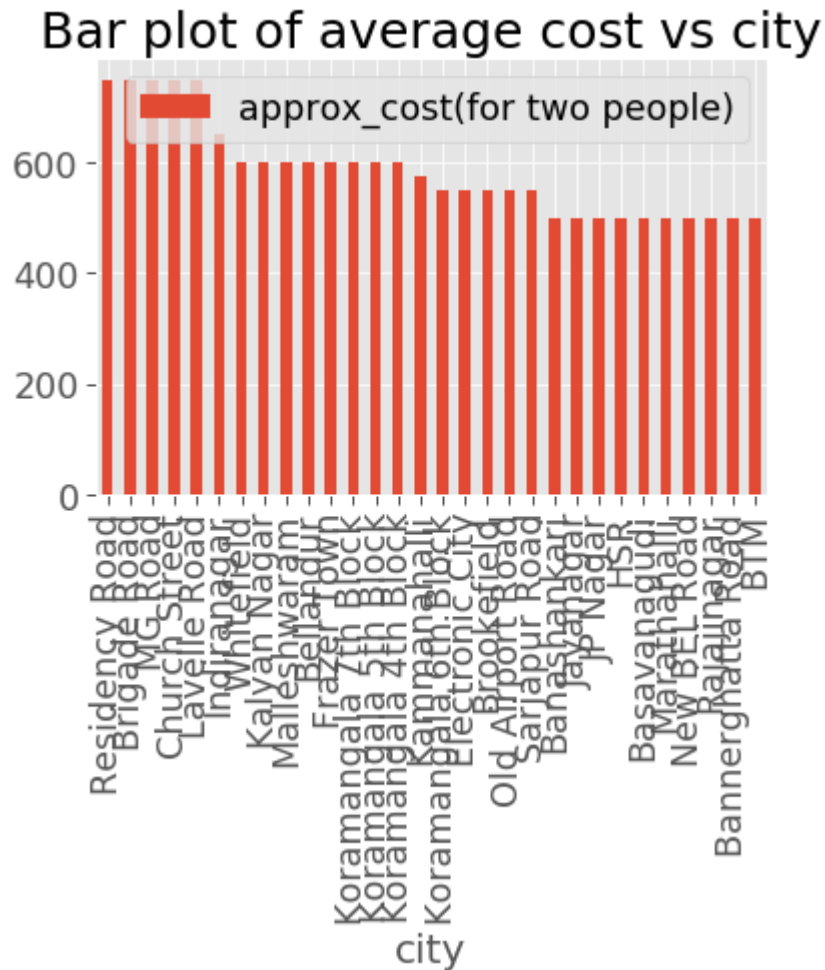
```
In [71]: sns.set_context("paper", font_scale = 2, rc = {"font.size": 20,"axes.titlesize":
b = sns.countplot(data = df, x = 'online_order', hue = 'book_table')
plt.title('Number of restaurants in online order in which you can book a table or not')
b.set_xticklabels(b.get_xticklabels(),rotation = 90)
plt.show()
```

Number of restaurants in online order in which you can book a table or not



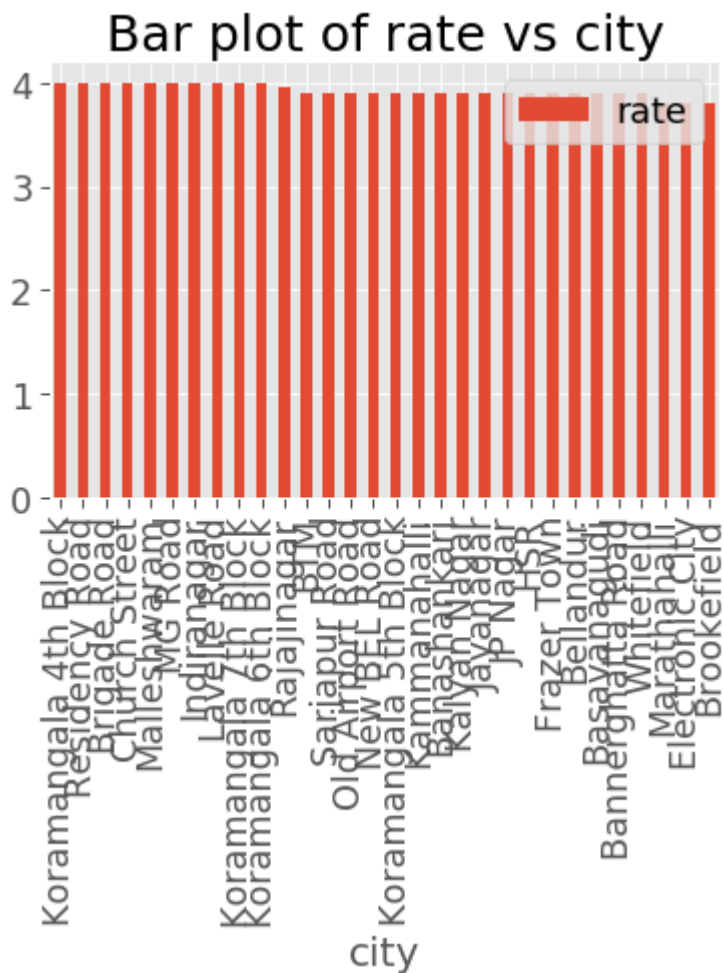
```
In [72]: plt.figure(figsize=(20,10))
df[['approx_cost(for two people)', 'city']].groupby(['city']).median().sort_values()
plt.title('Bar plot of average cost vs city')
plt.show()
```

<Figure size 1440x720 with 0 Axes>



```
In [73]: plt.figure(figsize=(20,10))
df[['rate', 'city']].groupby(['city']).median().sort_values("rate", ascending = True)
plt.title('Bar plot of rate vs city')
plt.show()
```

<Figure size 1440x720 with 0 Axes>



```

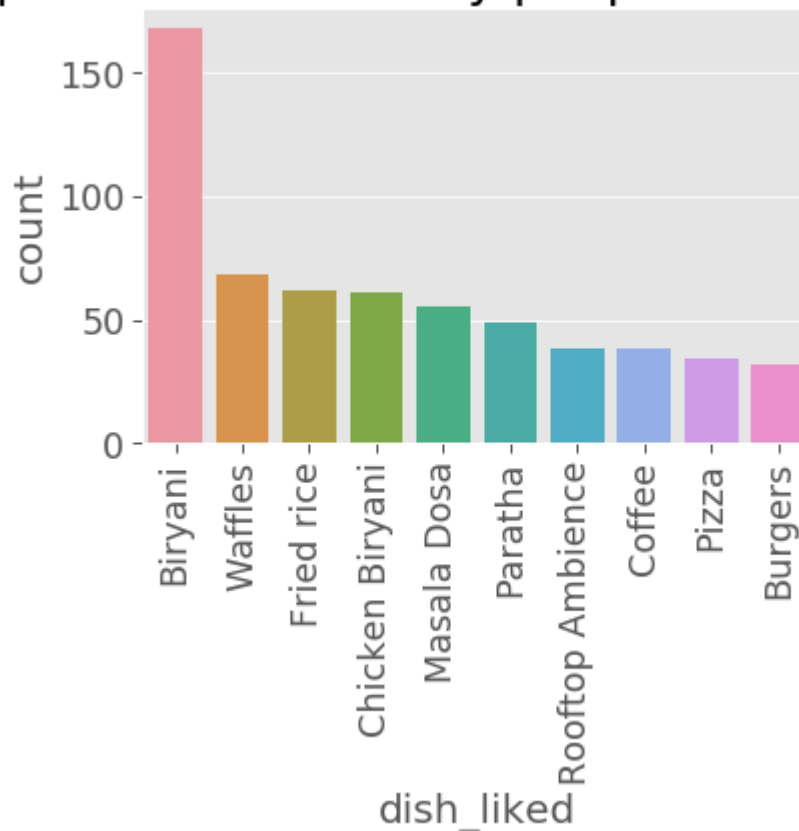
In [74]: a = df.iloc[:, :].values
for i in range(0, len(a)):
    if a[i, 6] == 'Fried rice':
        print(a[i, 6])
        a[i, 6] = None

a = pd.DataFrame(a)
a.columns = df.columns

sns.set_context("paper", font_scale = 2, rc = {"font.size": 20, "axes.titlesize":
b = sns.countplot(data = a, x = 'dish_liked', order = a.dish_liked.value_counts(
b.set_xticklabels(b.get_xticklabels(), rotation = 90)
plt.title('Top 10 dishes liked by people of Bangalore')
plt.show()

```

Top 10 dishes liked by people of Bangalore



Conclution

***We see that people at Bangalore love Biryani the most, then comes Chicken Biryani and then Fried rice**

In []:

In []: