NFW 9/28/24, 1:08 PM

```
In [1]: # Step 1 - Import all the libraries
        import pandas as pd
        import matplotlib.pyplot as plt
        import seaborn as sns
        import nltk
        from nltk.stem.porter import PorterStemmer
        nltk.download('stopwords')
        from nltk.corpus import stopwords
        STOPWORDS = set(stopwords.words('english'))
        from sklearn.model_selection import train_test_split
        from sklearn.preprocessing import MinMaxScaler
        from sklearn.feature_extraction.text import CountVectorizer
        from sklearn.model_selection import cross_val_score
        from sklearn.model_selection import GridSearchCV
        from sklearn.model_selection import StratifiedKFold
        from sklearn.metrics import accuracy_score
        from wordcloud import WordCloud
        import pickle
        import re
        from sklearn.metrics import confusion_matrix,accuracy_score,classification_report
        from sklearn.feature_extraction.text import TfidfVectorizer
        from nltk.stem import PorterStemmer
        import regex as re
        import string
        from nltk.corpus import stopwords
        from nltk.stem import WordNetLemmatizer
       [nltk_data] Downloading package stopwords to
                     C:\Users\swaro\AppData\Roaming\nltk_data...
       [nltk data]
       [nltk_data] Package stopwords is already up-to-date!
```

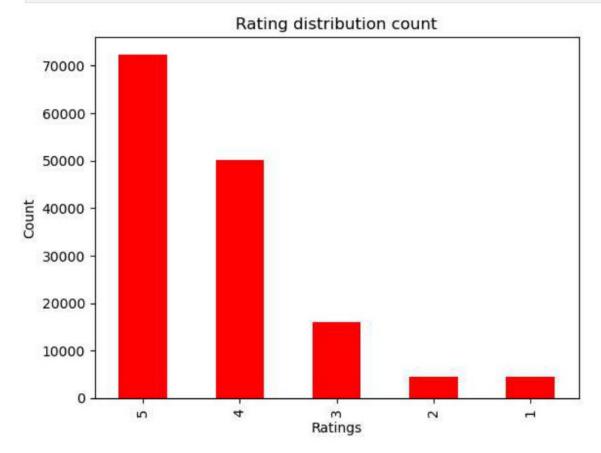
```
In [2]: %pip install wordcloud
```

```
wordcloud) (1.26.4)
       Requirement already satisfied: pillow in c:\inter .ai\lib\site-packages (from wordc
       loud) (10.3.0)
       Requirement already satisfied: matplotlib in c:\interr .ai\lib\site-packages (from w
       ordcloud) (3.8.4)
       Requirement already satisfied: contourpy>=1.0.1 in c:\interr .ai\lib\site-packages
       (from matplotlib->wordcloud) (1.2.0)
       Requirement already satisfied: cycler>=0.10 in c:\interr .ai\lib\site-packages (from
       matplotlib->wordcloud) (0.11.0)
       Requirement already satisfied: fonttools>=4.22.0 in c:\interr .ai\lib\site-packages
       (from matplotlib->wordcloud) (4.51.0)
       Requirement already satisfied: kiwisolver>=1.3.1 in c:\interr .ai\lib\site-packages
       (from matplotlib->wordcloud) (1.4.4)
       Requirement already satisfied: packaging>=20.0 in c:\interr .ai\lib\site-packages (f
       rom matplotlib->wordcloud) (23.2)
       Requirement already satisfied: pyparsing>=2.3.1 in c:\interr .ai\lib\site-packages
       (from matplotlib->wordcloud) (3.0.9)
       Requirement already satisfied: python-dateutil>=2.7 in c:\interr .ai\lib\site-packag
       es (from matplotlib->wordcloud) (2.9.0.post0)
       Requirement already satisfied: six>=1.5 in c:\interr .ai\lib\site-packages (from pyt
       hon-dateutil>=2.7->matplotlib->wordcloud) (1.16.0)
       Note: you may need to restart the kernel to use updated packages.
In [3]: Data=pd.read_csv(r"C:\Users\swaro\OneDrive\Desktop\data science swaroop ky 1\Projec
        print(Data)
               rating_review
                                                                    review full
       0
                           5 Totally in love with the Auro of the place, re...
       1
                           5 I went this bar 8 days regularly with my husba...
       2
                           5 We were few friends and was a birthday celebra...
       3
                           5 Fatjar Cafe and Market is the perfect place fo...
       4
                           5 Hey Guys, if you are craving for pizza and sea...
       . . .
       147576
                          5 Near by airport very calm and cool environment...
                          5 My favourite place to stay. Great service. Ash...
       147577
       147578
                          5 good food with nice decoration, drinks list al...
       147579
                           4 Near to airport .it is fine property. Staff i...
       147580
                           5 Amazing food .. Excellent ambience .. Great ...
       [147581 rows x 2 columns]
In [4]: # step 2
        Data.shape
Out[4]: (147581, 2)
In [5]: # Coloumns
        Data.columns
Out[5]: Index(['rating_review', 'review_full'], dtype='object')
In [6]: #Check for null values
        Data.isnull().sum()
```

Requirement already satisfied: wordcloud in c:\interr .ai\lib\site-packages (1.9.3)
Requirement already satisfied: numpy>=1.6.1 in c:\interr .ai\lib\site-packages (from

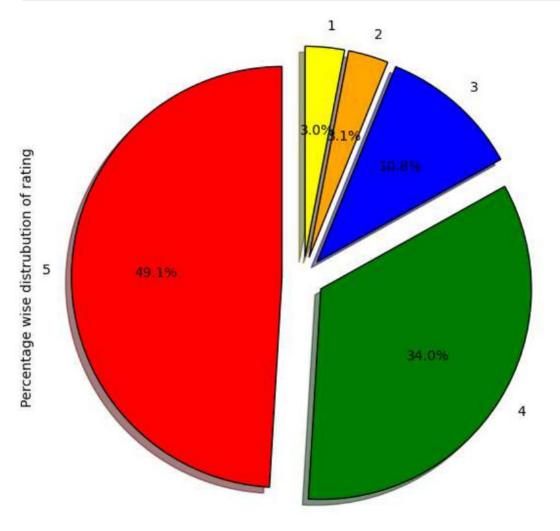
```
Out[6]: rating_review
             review_full
                               2
             dtype: int64
   In [7]: #We will drop the null record
            Data.dropna(inplace=True)
   In [9]: # shape after dropping null values
            Data.shape
   Out[9]: (147579, 2)
  In [10]: #Creating a new column 'length' that will contain the length of the string in 'veri
            Data['length'] = Data['review_full'].apply(len)
            Data.head()
  Out[10]:
                rating_review
                                                              review_full length
             0
                           5
                                  Totally in love with the Auro of the place, re...
                                                                             720
             1
                           5
                                I went this bar 8 days regularly with my husba...
                                                                             202
             2
                           5 We were few friends and was a birthday celebra...
                                                                             144
                           5
                                 Fatjar Cafe and Market is the perfect place fo...
                                                                             435
             3
             4
                           5
                                Hey Guys, if you are craving for pizza and sea...
                                                                             533
Datatypes of the features
  In [11]: # step 3
            Data.dtypes
  Out[11]: rating_review
                                int64
             review_full
                               object
             length
                                int64
             dtype: object
  In [13]: #Step 4 #Distinct values of 'rating' and its count
            print(f"Rating value count: \n{Data["rating_review"].value_counts()}")
           Rating value count:
           rating_review
           5
                72389
           4
                50248
           3
                15935
           2
                 4552
                 4455
           1
           Name: count, dtype: int64
  In [14]: #Bar plot to visualize the total counts of each rating
            Data["rating_review"].value_counts().plot.bar(color = 'red')
             plt.title('Rating distribution count')
            plt.xlabel('Ratings')
```

```
plt.ylabel('Count')
plt.show()
```



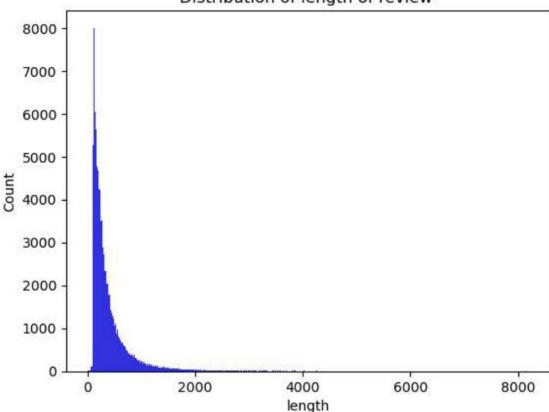
```
In [15]: #Finding the percentage distribution of each rating
         print(f"Rating value count - percentage distribution: \n{round(Data['rating_review'
        Rating value count - percentage distribution:
        rating_review
        5
             49.05
        4
             34.05
        3
             10.80
              3.08
        2
              3.02
        1
        Name: count, dtype: float64
In [16]: fig = plt.figure(figsize=(7,7))
         colors = ('red', 'green', 'blue', 'orange', 'yellow')
         wp = {'linewidth':1, "edgecolor":'black'}
         tags = Data["rating_review"].value_counts()/Data.shape[0]
         explode=(0.1,0.1,0.1,0.1,0.1)
         tags.plot(kind='pie', autopct="%1.1f%%", shadow=True, colors=colors, startangle=90,
         from io import BytesIO
```

```
graph = BytesIO()
fig.savefig(graph, format="png")
```



```
In [17]: # Analyzing 'review_full' column
         Data[ "length"].describe()
Out[17]: count
                   147579.000000
                      404.745445
         mean
          std
                      463.877369
         min
                        6.000000
          25%
                      163.000000
          50%
                      263.000000
         75%
                      462.000000
                     8192.000000
         max
         Name: length, dtype: float64
In [18]: sns.histplot(Data['length'],color='blue').set(title='Distribution of length of revi
Out[18]: [Text(0.5, 1.0, 'Distribution of length of review ')]
```

Distribution of length of review



```
In [19]: # Combine all reviews
    reviews = " ".join([review for review in Data['review_full']])

# Initialize wordcloud object
    wc = WordCloud(background_color='white', max_words=50)

# Generate and plot wordcloud
    plt.figure(figsize=(10,10))
    plt.imshow(wc.generate(reviews))
    plt.title('Wordcloud for all reviews', fontsize=10)
    plt.axis('off')
    plt.show()
```



```
In [20]: neg_reviews = " ".join([review for review in Data[Data['review_full'].str.contains(
         neg_reviews = neg_reviews.lower().split()
         # Combine all positive reviews into a single string
         pos_reviews = " ".join([review for review in Data[~Data['review_full'].str.contains
         pos_reviews = pos_reviews.lower().split()
In [21]: # Assuming neg_reviews and pos_reviews are lists of words from the reviews
         # Convert Lists to sets for unique words
         set_neg_reviews = set(neg_reviews)
         set_pos_reviews = set(pos_reviews)
         # Find unique negative and positive words using set difference
         unique_negative = set_neg_reviews - set_pos_reviews
         unique_positive = set_pos_reviews - set_neg_reviews
         # Join the unique words back into strings
         unique_negative = " ".join(unique_negative)
         unique_positive = " ".join(unique_positive)
In [22]: wc = WordCloud(background_color='white', max_words=50)
         # Generate and plot wordcloud
         plt.figure(figsize=(10,10))
         plt.imshow(wc.generate(unique_negative))
         plt.title('Wordcloud for negative reviews', fontsize=10)
         plt.axis('off')
         plt.show()
```



```
In [23]: wc = WordCloud(background_color='white', max_words=50)

# Generate and plot wordcloud
plt.figure(figsize=(10,10))
plt.imshow(wc.generate(unique_positive))
plt.title('Wordcloud for positive reviews', fontsize=10)
plt.axis('off')
plt.show()
```

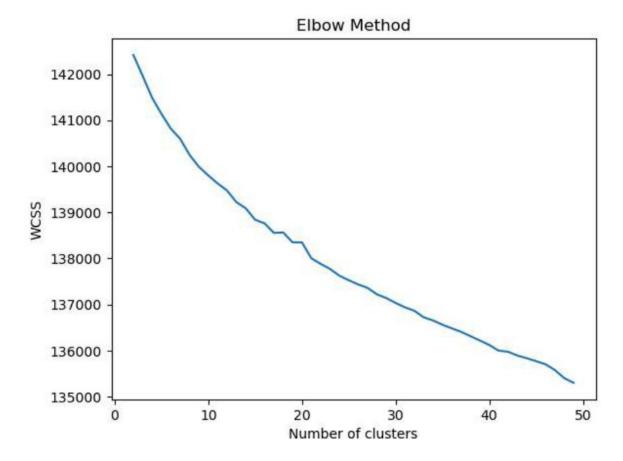


Preprocessing and Modelling

```
In [24]: from nltk.stem import WordNetLemmatizer
# Initialize Lemmatizer
lemmatizer = WordNetLemmatizer()
stop_words = set(stopwords.words('english'))

def clean_text(text):
```

```
# Lowercase
             text = text.lower()
             # Remove punctuation
             text = re.sub(r'[^\w\s]', '', text)
             # Remove stop words and Lemmatize
             text = ' '.join(lemmatizer.lemmatize(word) for word in text.split() if word not
             return text
         # Apply cleaning function
         Data['cleaned_reviews'] = Data['review_full'].apply(clean_text)
In [25]: from sklearn.feature_extraction.text import TfidfVectorizer
         # Initialize TF-IDF Vectorizer
         tfidf_vectorizer = TfidfVectorizer(max_features=5000) # Limit features for perform
         tfidf_matrix = tfidf_vectorizer.fit_transform(Data['cleaned_reviews'])
In [26]: from sklearn.cluster import KMeans
         WCSS = []
         for i in range(2, 50): # Adjusting the range to be more efficient
             kmeans = KMeans(n_clusters=i, random_state=42)
             kmeans.fit(tfidf_matrix)
             wcss.append(kmeans.inertia_)
In [27]: plt.plot(range(2, 50), wcss) # Adjusting the range accordingly
         plt.title('Elbow Method')
         plt.xlabel('Number of clusters')
         plt.ylabel('WCSS')
         plt.show()
```



```
In [28]:
          # Fit KMeans with a chosen number of clusters, e.g., 5
          optimal clusters = 20
          kmeans = KMeans(n_clusters=optimal_clusters, random_state=42)
          Data['cluster'] = kmeans.fit_predict(tfidf_matrix)
In [29]: # Get feature names
         feature_names = tfidf_vectorizer.get_feature_names_out()
         # Function to get top words for each cluster
         def get_top_words_per_cluster(model, n_words=10):
             order_centroids = model.cluster_centers_.argsort()[:, ::-1]
             top_words = {}
             for i in range(optimal clusters):
                 top_words[i] = [feature_names[ind] for ind in order_centroids[i, :n_words]]
             return top_words
         top_words = get_top_words_per_cluster(kmeans)
         # Display top words for each cluster
         for cluster, words in top_words.items():
```

print(f"Cluster {cluster}: {', '.join(words)}")

NFW 9/28/24, 1:08 PM

```
Cluster 0: price, reasonable, food, good, place, service, quality, restaurant, high,
great
Cluster 1: food, service, great, good, one, time, taste, place, staff, dish
Cluster 2: excellent, service, food, great, good, staff, restaurant, place, ambienc
Cluster 3: amazing, food, place, great, service, staff, experience, ambience, reall
y, visit
Cluster 4: quick, service, food, good, place, bite, great, nice, staff, restaurant
Cluster 5: good, food, service, nice, place, ambience, also, really, staff, quality
Cluster 6: place, food, great, visit, good, nice, friend, one, must, best
Cluster 7: south, indian, food, dosa, place, good, authentic, restaurant, delhi, nor
Cluster 8: pizza, pasta, good, place, cheese, domino, crust, italian, food, great
Cluster 9: restaurant, food, delhi, good, service, one, indian, great, hotel, best
Cluster 10: music, place, good, live, food, great, nice, loud, ambience, drink
Cluster 11: beer, good, place, food, nice, great, service, bar, one, menu
Cluster 12: chicken, butter, place, good, food, dish, tikka, ordered, try, one
Cluster 13: biryani, mutton, kebab, chicken, karims, place, food, good, taste, try
Cluster 14: dal, paneer, makhani, bukhara, naan, food, good, tikka, makhni, place
Cluster 15: veg, non, food, good, place, restaurant, dish, also, vegetarian, taste
Cluster 16: burger, good, place, fry, shake, chicken, food, try, one, taste
Cluster 17: drink, bar, place, good, great, food, cocktail, nice, service, friend
Cluster 18: breakfast, buffet, good, service, staff, great, food, hotel, dinner, nic
Cluster 19: thanks, mr, special, service, great, food, team, staff, hospitality, goo
Requirement already satisfied: textblob in c:\interr .ai\lib\site-packages (0.18.0.p
```

In [30]: pip install textblob

Requirement already satisfied: nltk>=3.8 in c:\interr .ai\lib\site-packages (from te xtblob) (3.8.1)

Requirement already satisfied: click in c:\interr .ai\lib\site-packages (from nltk>= 3.8->textblob) (8.1.7)

Requirement already satisfied: joblib in c:\inter .ai\lib\site-packages (from nltk> =3.8->textblob) (1.4.2)

Requirement already satisfied: regex>=2021.8.3 in c:\interr .ai\lib\site-packages (f rom nltk>=3.8->textblob) (2023.10.3)

Requirement already satisfied: tqdm in c:\interr .ai\lib\site-packages (from nltk>= 3.8->textblob) (4.66.4)

Requirement already satisfied: colorama in c:\interr .ai\lib\site-packages (from cli ck->nltk>=3.8->textblob) (0.4.6)

Note: you may need to restart the kernel to use updated packages.

```
In [31]: from textblob import TextBlob
         # Function to calculate sentiment polarity
         def get_sentiment(text):
             return TextBlob(text).sentiment.polarity
         # Apply sentiment analysis on cleaned reviews
         Data['sentiment'] = Data['cleaned_reviews'].apply(get_sentiment)
         # Calculate average sentiment for each cluster
         cluster_sentiment = Data.groupby('cluster')['sentiment'].mean().reset_index()
```

```
# Define a function to label clusters based on average sentiment
         def label_cluster(row):
            if row['sentiment'] > 0:
                return 'Positive'
             elif row['sentiment'] < 0:</pre>
                 return 'Negative'
             else:
                 return 'Neutral'
         # Apply the Labeling function
         cluster_sentiment['label'] = cluster_sentiment.apply(label_cluster, axis=1)
         # Display the average sentiment and labels for each cluster
         print(cluster_sentiment)
           cluster sentiment
                                 label
                 0 0.315965 Positive
                 1 0.285559 Positive
       1
       2
                 2 0.577745 Positive
       3
                 3 0.475964 Positive
       4
                4 0.360778 Positive
       5
                5 0.481366 Positive
                6 0.385212 Positive
       6
                7 0.394815 Positive
       7
       8
                8 0.325297 Positive
       9
                9 0.317628 Positive
       10
              10 0.384022 Positive
               11 0.277481 Positive
       11
       12
               12 0.129191 Positive
       13
               13 0.266642 Positive
               14 0.343193 Positive
       14
               15 0.356666 Positive
       15
                16 0.293418 Positive
       16
                17 0.354854 Positive
       17
       18
               18 0.420457 Positive
       19
                19 0.471944 Positive
In [32]: # Function to calculate sentiment polarity
         def get_sentiment(text):
             return TextBlob(text).sentiment.polarity
         # Apply sentiment analysis on cleaned reviews
         Data['sentiment'] = Data['cleaned_reviews'].apply(get_sentiment)
         # Filter for negative reviews
         negative_reviews = Data[Data['sentiment'] < 0]</pre>
         # Calculate average sentiment for each cluster with negative reviews
         cluster_sentiment = negative_reviews.groupby('cluster')['sentiment'].mean().reset_i
         # Define a function to label clusters based on average sentiment
         def label cluster(row):
             if row['sentiment'] < 0:</pre>
                return 'Negative'
             else:
```

```
# Apply the labeling function
         cluster_sentiment['label'] = cluster_sentiment.apply(label_cluster, axis=1)
         # Display the average sentiment and labels for each cluster
         print(cluster_sentiment)
           cluster sentiment
                                  label
                 0 -0.145982 Negative
       0
       1
                 1 -0.172655 Negative
       2
                 2 -0.122959 Negative
                 3 -0.083245 Negative
       3
                4 -0.094006 Negative
       4
                 5 -0.147875 Negative
       5
       6
                 6 -0.168441 Negative
       7
                7 -0.144348 Negative
                8 -0.153509 Negative
       8
                9 -0.158714 Negative
       9
       10
                10 -0.134783 Negative
       11
                11 -0.127412 Negative
       12
               12 -0.154380 Negative
               13 -0.149298 Negative
       13
                14 -0.149447 Negative
       14
       15
                15 -0.132757 Negative
       16
               16 -0.166638 Negative
       17
               17 -0.156612 Negative
               18 -0.156940 Negative
       18
       19
                19 -0.170526 Negative
In [33]: # Function to get top positive and negative reviews for each cluster
         def get_top_reviews_per_cluster(data, cluster_labels):
             top_reviews = {}
             for label in ['Positive', 'Negative']:
                 top_reviews[label] = {}
                 for cluster in data['cluster'].unique():
                     cluster_data = data[data['cluster'] == cluster]
                     if label == 'Positive':
                         top_reviews[label][cluster] = cluster_data[cluster_data['sentiment'
                     else:
                         top_reviews[label][cluster] = cluster_data[cluster_data['sentiment']
             return top_reviews
         # Get top positive and negative reviews for each cluster
         top_reviews = get_top_reviews_per_cluster(Data, cluster_sentiment)
         # Display the top positive and negative reviews
         for label, cluster_reviews in top_reviews.items():
             print(f"Top {label} Reviews:")
             for cluster, reviews in cluster_reviews.items():
                 print(f"Cluster {cluster}:")
                 for review in reviews:
                     print(review)
                 print()
```

return 'Not Applicable' # Since we're only focusing on negative reviews

Top Positive Reviews:

Cluster 6:

Totally in love with the Auro of the place, really beautiful and quite fancy at the same time. The ambience is very pure and gives a sense of positivity throughout. Out door and indoor interior are quite quaint and cute. Love the open kitchen idea and t here whole marketplace ideology. Due to coronovirus they specifically use disposable cutlery to keep the pandemic in mind taking all the precautionary measures from the beginning of the place with the mask on their staff and using good sanitisation. The food is really amazing specially the pizza straight from the oven and the hummus and pita bread are quite delicious too. If you're looking for a classy yet soothing Ital ian place in Delhi, Fatjar is a go to for you!

Fatjar Cafe and Market is the perfect place for a casual lunch with your loved ones. The Ambience is delightful and the food is totally the best thing about the place! Y Ou can purchase your choosing of Olives and a lot more directly from the place. They are taking all the prevention measures for the spread of COVID-19 and I can say this without hesitation that it is by far the most safest place felt after looking at the ir actions **

If you are looking for a varied selection of delicious, high quality dishes, this is your place. We spent three weeks traveling around India, and our best meals of the trip were unquestionably found here. Great place if you are tiring of the typical Indian dishes and are lusting for some ethnic variety. Oh, and do yourself a favor: You MUST get the Sticky Date Pudding with Vanilla Bean Ice Cream. It may be the best dessert we have ever had.

was staying near by so visited this place. huge place with loads of options to eat a nd has a small deli too. ordered for the pesto mozzarella pesto sandwich which turne d out to be very good. shot of espresso was good too. if budget isn't a problem, this place is not to be missed for the quality of the ingredients they use.

Fat Jar is one of the most hygienic and beautiful places in Delhi NCR.Pizza and coff ee are so good here. I would also recommend the in house tea which is served hot and is delicous.

Cluster 17:

I went this bar 8 days regularly with my husband. we are fully satisfied by the ser vices. Staff is very good (vitoni and amen) they both served us daily . Will surely visit again. Highly recommended ♥

We had drinks during happy hour, good selection & excellent service at the bar. We d idn't eat but the food looked really good.

Great service with quite polite and generous staff. Must hangout at hanger. I was se rved by Mr. Shubham very cooperative and friendly person. The place is very neat and clean and follows all rules of safety and cleanliness. Great food and great Drinks. Cheers!

Beautiful and hygienic bar with great service by Ashwin... A smart guy with lots of talent... Love to visit again!!!

The hotel and staff are lovely and extremely attentive. The hot drinks and snacks ar e very nice indeed. Had a lovely stay in the hotel and I was served by a very lovel y staff member by the name of Laxmi. Thsank you

Cluster 5:

We were few friends and was a birthday celebration. The food was so good in taste an d it was really fresh. We all loved it and highly recomment.

Emmanuel is a good stirrer and shaker. They taste great and he is very knowledgeabl e. A good place to relax.

Superb , fantastic , good service..food was tasty , Tarun and Akash was really good in service and very polite too...

Very good service as well as food....Nice experience....Staff is very supportive and cooperative...

Really good atmosphere and good service from the staff. Especially surendra, really helpful through out the match even at such a late time.

Cluster 8:

Hey Guys, if you are craving for pizza and searching for it then you should visit th is cafe. Yes, I highly recommend you to visit this FatJar Cafe because the cafe offe rs scrumptious pizza. I have ordered Veggie Pizza which was really awesome. Apart fr om that, I have also ordered Cappuccino and one soft drink. If I talk about the ambi ance then no doubt, because it is absolutely pretty. From my side I'll give 5/5 for this cafe. It is completely amazing. I had a great experience with this cafe. Here a re some pictures have a look

Kailash colony is brimming with small cafes now. I happened to visit this beauty for lunch and was amazed with the 1st impression it made. We started with Falafel & Pit a and Margherita Pizza. Safe to say, Margherita pizza was the best choice of the day. Its made freshly in front of you in the Wood Fired oven. The taste was exquisite and I would recommend this to anybody who plans to visit this outlet. I could come here just to have that Pizza. Hummus, Pita and Falafel were a good side dish while we ordered the next couple of dishes. We moved on to order Fussili Alfredo. This white sauce fussili pasta was good but could not be compared to the pizza we had earlier. In the end, we went for Frozen Yoghurt with Honey. It is recommended to have it with honey, which will do wonders to the frozen yoghurt! This place turned out to be quite peaceful and the plants enhanced its beauty.

Kailash Colony is the place I always head to when I'm not in the mood for GK & Nehru Place's hustle bustle. This time around we planned to visit Fatjar Café and Market w hich is right on the road and was very easy to find. Fatjar's ambience is very ple asing to the eyes and the light colours used makes it look very airy. The use of pla nts further makes this place look more beautiful. Most of the cooking is done in fro nt of you with a semi open kitchen and their food fire oven looks amazing too. The p roducts on display establishes the word "market" in the name perfectly. The varied s eating options are very comfortable and a small outdoor seating is great for sunny w inter lunches. The parking though is limited in front of the café. We started our meal with some "Falafel & Pita" which had some moist falafels which were nicely frie d and not at all oily. The pita breads were very fresh and so was the hummus. You co uld literally tell that that just after your first bite. The consistency of the humm us was noteworthy too. Next we ordered some humble "Truffle Fries" which were very c rispy and a pinch of truffle to top them made them taste even better. For the main course, we ordered a "Fussili Alfredo" which was very creamy and the pasta was cooke d to perfection. The only thing they need to look at is that usually the pasta is su pposed to be served along with a bread which they don't. The highlight of our meal w as simply their "Margherita Pizza" which had a brilliant base made out of the freshe st dough you could think of. The cheese used to top the pizza looked very premium an d the balance between the sauces and the cheese was tremendous too. The presentation deserves a special mention as it wasn't the usual pizza stand they serve their pizza on but is more a raised steel stand which again was a nice touch. To accompany our meal, we ordered a "Berries Shakes" which was very thick and was prepared using real berries instead of the preservatives filled syrups. Though the sweetness was a bit t oo low for our liking and a little more sweetness would have made that shake perfec We ended our meal with a "Frozen Yoghurt with Honey" which had two scoops of fr ozen yoghurt which had some ice crystals too and once you top it up with honey, the combination really does wonders to the taste. The service was top notch and the st ewards were very considerate about what we were ordering and didn't shy away from re commending their favourites and also explaining the dishes on the menu. Overall, Fat jar simply amazed us with its varied food options on the menu and its serene ambienc e. The place looked fantastic and is perfect for a romantic date or co working or if you are looking forward to a quiet and a hearty meal.

The service was great. this young gentlemen Suraj gave us a awesome service. The piz