

part1

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```
# Importing Libraries
```

```
library(readr)  
library(ggplot2)  
library(dplyr)
```

```
##
```

```
## Attaching package: 'dplyr'
```

```
## The following objects are masked from 'package:stats':
```

```
##
```

```
## filter, lag
```

```
## The following objects are masked from 'package:base':
```

```
##
```

```
## intersect, setdiff, setequal, union
```

```
library(tidyr)
```

```
library(maps)
```

```
## Warning: package 'maps' was built under R version 4.3.3
```

```
library(mapdata)
```

```
library(corrplot)
```

```
## Warning: package 'corrplot' was built under R version 4.3.3
```

```
## corrplot 0.95 loaded
```

```
## Task 1: DATA EXPLORATION AND PREPROCESSING
```

```
# Load Dataset
```

```
data <- read.csv("~/Desktop/data_analysis_restaurant_data/Dataset.csv")
```

```
# View top 10 rows of the dataset
```

```
head(data, 10)
```

##	Restaurant.ID	Restaurant.Name	Country.Code
## 1	6317637	Le Petit Souffle	162
## 2	6304287	Izakaya Kikufuji	162
## 3	6300002	Heat - Edsa Shangri-La	162
## 4	6318506	Ooma	162
## 5	6314302	Sambo Kojin	162
## 6	18189371	Din Tai Fung	162
## 7	6300781	Buffet 101	162
## 8	6301290	Vikings	162
## 9	6300010	Spiral - Sofitel Philippine Plaza Manila	162
## 10	6314987	Locavore	162
##	City		
## 1	Makati City		
## 2	Makati City		
## 3	Mandaluyong City		
## 4	Mandaluyong City		
## 5	Mandaluyong City		
## 6	Mandaluyong City		
## 7	Pasay City		
## 8	Pasay City		
## 9	Pasay City		
## 10	Pasig City		
##		Address	
## 1		Third Floor, Century City Mall, Kalayaan Avenue, Poblacion, Makati City	
## 2		Little Tokyo, 2277 Chino Roces Avenue, Legaspi Village, Makati City	
## 3		Edsa Shangri-La, 1 Garden Way, Ortigas, Mandaluyong City	
## 4		Third Floor, Mega Fashion Hall, SM Megamall, Ortigas, Mandaluyong City	
## 5		Third Floor, Mega Atrium, SM Megamall, Ortigas, Mandaluyong City	
## 6		Ground Floor, Mega Fashion Hall, SM Megamall, Ortigas, Mandaluyong City	
## 7		Building K, SM By The Bay, Sunset Boulevard, Mall of Asia Complex (MOA), Pasay City	
## 8		Building B, By The Bay, Seaside Boulevard, Mall of Asia Complex (MOA), Pasay City	
## 9		Plaza Level, Sofitel Philippine Plaza Manila, CCP Complex, Pasay City	
## 10		Brixton Technology Center, 10 Brixton Street, Kapitolyo, Pasig City	
##		Locality	
## 1		Century City Mall, Poblacion, Makati City	
## 2		Little Tokyo, Legaspi Village, Makati City	
## 3		Edsa Shangri-La, Ortigas, Mandaluyong City	
## 4		SM Megamall, Ortigas, Mandaluyong City	
## 5		SM Megamall, Ortigas, Mandaluyong City	
## 6		SM Megamall, Ortigas, Mandaluyong City	
## 7		SM by the Bay, Mall of Asia Complex, Pasay City	
## 8		SM by the Bay, Mall of Asia Complex, Pasay City	
## 9		Sofitel Philippine Plaza Manila, Pasay City	
## 10		Kapitolyo	
##		Locality.Verbose Longitude	
## 1		Century City Mall, Poblacion, Makati City, Makati City	121.0275
## 2		Little Tokyo, Legaspi Village, Makati City, Makati City	121.0141
## 3		Edsa Shangri-La, Ortigas, Mandaluyong City, Mandaluyong City	121.0568
## 4		SM Megamall, Ortigas, Mandaluyong City, Mandaluyong City	121.0565
## 5		SM Megamall, Ortigas, Mandaluyong City, Mandaluyong City	121.0575
## 6		SM Megamall, Ortigas, Mandaluyong City, Mandaluyong City	121.0563
## 7		SM by the Bay, Mall of Asia Complex, Pasay City, Pasay City	120.9797
## 8		SM by the Bay, Mall of Asia Complex, Pasay City, Pasay City	120.9793
## 9		Sofitel Philippine Plaza Manila, Pasay City, Pasay City	120.9801

```

## 10                               Kapitolyo, Pasig City 121.0565
## Latitude                        Cuisines Average.Cost.for.two
## 1 14.56544      French, Japanese, Desserts      1100
## 2 14.55371                        Japanese      1200
## 3 14.58140      Seafood, Asian, Filipino, Indian      4000
## 4 14.58532                        Japanese, Sushi      1500
## 5 14.58445                        Japanese, Korean      1500
## 6 14.58376                        Chinese      1000
## 7 14.53133                        Asian, European      2000
## 8 14.54000      Seafood, Filipino, Asian, European      2000
## 9 14.55299                        European, Asian, Indian      6000
## 10 14.57204                        Filipino      1100
## Currency Has.Table.booking Has.Online.delivery Is.delivering.now
## 1 Botswana Pula(P)      Yes      No      No
## 2 Botswana Pula(P)      Yes      No      No
## 3 Botswana Pula(P)      Yes      No      No
## 4 Botswana Pula(P)      No      No      No
## 5 Botswana Pula(P)      Yes      No      No
## 6 Botswana Pula(P)      No      No      No
## 7 Botswana Pula(P)      Yes      No      No
## 8 Botswana Pula(P)      Yes      No      No
## 9 Botswana Pula(P)      Yes      No      No
## 10 Botswana Pula(P)      Yes      No      No
## Switch.to.order.menu Price.range Aggregate.rating Rating.color Rating.text
## 1      No      3      4.8      Dark Green      Excellent
## 2      No      3      4.5      Dark Green      Excellent
## 3      No      4      4.4      Green      Very Good
## 4      No      4      4.9      Dark Green      Excellent
## 5      No      4      4.8      Dark Green      Excellent
## 6      No      3      4.4      Green      Very Good
## 7      No      4      4.0      Green      Very Good
## 8      No      4      4.2      Green      Very Good
## 9      No      4      4.9      Dark Green      Excellent
## 10     No      3      4.8      Dark Green      Excellent
## Votes
## 1      314
## 2      591
## 3      270
## 4      365
## 5      229
## 6      336
## 7      520
## 8      677
## 9      621
## 10     532

```

```
# Explore the dataset and identify the number of rows and columns
```

```
# Checking number of rows and columns of the dataset
```

```
cat("Number of rows:", nrow(data), "\n")
```

```
## Number of rows: 9551
```

```
cat("Number of columns:", ncol(data), "\n")
```

```
## Number of columns: 21
```

```
# Dataset Duplicate Value Count  
dup <- sum(duplicated(data))  
cat("Number of duplicate rows:", dup)
```

```
## Number of duplicate rows: 0
```

```
# Check for missing values in each column and handle them accordingly  
  
# Check for missing values  
missing_values <- sum(is.na(data))  
  
# Check for empty values  
empty_values <- sum(data == "")  
  
cat("Missing values count:", missing_values, "\n")
```

```
## Missing values count: 0
```

```
cat("Empty values count:", empty_values, "\n")
```

```
## Empty values count: 9
```

```
# There are 9 empty values, let's find out which column/columns has it  
empty_values_count <- colSums(data == "")  
cat("Empty Values Count:\n")
```

```
## Empty Values Count:
```

```
print(empty_values_count)
```

```
##      Restaurant.ID      Restaurant.Name      Country.Code  
##           0           0           0  
##           City           Address           Locality  
##           0           0           0  
##      Locality.Verbose      Longitude      Latitude  
##           0           0           0  
##           Cuisines Average.Cost.for.two      Currency  
##           9           0           0  
##      Has.Table.booking Has.Online.delivery Is.delivering.now  
##           0           0           0  
## Switch.to.order.menu      Price.range      Aggregate.rating  
##           0           0           0  
##           Rating.color      Rating.text      Votes  
##           0           0           0
```

```
# The Cuisines column has 9 empty values. Since it's not many, let's remove these rows
data <- data[!(data$Cuisines == ""), , drop = FALSE]
```

```
# Check for empty values after Removing
empty_values <- sum(data == "")
cat("Empty values count:", empty_values, "\n")
```

```
## Empty values count: 0
```

```
# Display basic information about the dataset to check various data types
str(data)
```

```
## 'data.frame': 9542 obs. of 21 variables:
## $ Restaurant.ID : int 6317637 6304287 6300002 6318506 6314302 18189371 6300781 6301290 6300002 ...
## $ Restaurant.Name : chr "Le Petit Souffle" "Izakaya Kikufuji" "Heat - Edsa Shangri-La" "Ooma" ...
## $ Country.Code : int 162 162 162 162 162 162 162 162 162 ...
## $ City : chr "Makati City" "Makati City" "Mandaluyong City" "Mandaluyong City" ...
## $ Address : chr "Third Floor, Century City Mall, Kalayaan Avenue, Poblacion, Makati City" ...
## $ Locality : chr "Century City Mall, Poblacion, Makati City" "Little Tokyo, Legaspi Village" ...
## $ Locality.Verbose : chr "Century City Mall, Poblacion, Makati City, Makati City" "Little Tokyo, Legaspi Village, Legaspi Village" ...
## $ Longitude : num 121 121 121 121 121 ...
## $ Latitude : num 14.6 14.6 14.6 14.6 14.6 ...
## $ Cuisines : chr "French, Japanese, Desserts" "Japanese" "Seafood, Asian, Filipino, Indian" ...
## $ Average.Cost.for.two : int 1100 1200 4000 1500 1500 1000 2000 2000 6000 1100 ...
## $ Currency : chr "Botswana Pula(P)" "Botswana Pula(P)" "Botswana Pula(P)" "Botswana Pula(P)" ...
## $ Has.Table.booking : chr "Yes" "Yes" "Yes" "No" ...
## $ Has.Online.delivery : chr "No" "No" "No" "No" ...
## $ Is.delivering.now : chr "No" "No" "No" "No" ...
## $ Switch.to.order.menu : chr "No" "No" "No" "No" ...
## $ Price.range : int 3 3 4 4 4 3 4 4 4 3 ...
## $ Aggregate.rating : num 4.8 4.5 4.4 4.9 4.8 4.4 4 4.2 4.9 4.8 ...
## $ Rating.color : chr "Dark Green" "Dark Green" "Green" "Dark Green" ...
## $ Rating.text : chr "Excellent" "Excellent" "Very Good" "Excellent" ...
## $ Votes : int 314 591 270 365 229 336 520 677 621 532 ...
```

```
# Analyze the distribution of the target variable ("Aggregate rating") and identify any class imbalance.
```

```
# Distribution of the target variable ("Aggregate rating")
target_counts <- table(data$'Aggregate rating')
```

```
# Print the distribution
print("Distribution of target variable:")
```

```
## [1] "Distribution of target variable:"
```

```
print(target_counts)
```

```
## < table of extent 0 >
```

```
# Check if the distribution is balanced
is_balanced <- all(target_counts >= mean(target_counts))
if (is_balanced) {
  print("The distribution of the target variable is balanced.")
} else {
  print("The distribution of the target variable is imbalanced.")
}
```

```
## [1] "The distribution of the target variable is balanced."
```

Task 2: DESCRIPTIVE ANALYSIS

Basic statistical measures (mean, median, standard deviation, etc.) for numerical columns

Select Numerical Columns

```
numeric_columns <- data[, sapply(data, is.numeric)]
```

Calculate basic statistical measures using summary()

```
summary_stats <- summary(numeric_columns)
print(summary_stats)
```

```
## Restaurant.ID      Country.Code      Longitude      Latitude
## Min.   :      53    Min.   : 1.00    Min.   : -157.95  Min.   : -41.33
## 1st Qu.: 301931    1st Qu.: 1.00    1st Qu.:  77.08    1st Qu.:  28.48
## Median : 6002726    Median : 1.00    Median :  77.19    Median :  28.57
## Mean   : 9043301    Mean   : 18.18    Mean   :  64.28    Mean   :  25.85
## 3rd Qu.:18352604    3rd Qu.: 1.00    3rd Qu.:  77.28    3rd Qu.:  28.64
## Max.   :18500652    Max.   :216.00    Max.   : 174.83    Max.   :  55.98
## Average.Cost.for.two Price.range      Aggregate.rating      Votes
## Min.   :      0      Min.   :1.000    Min.   :0.000    Min.   :   0.0
## 1st Qu.:  250      1st Qu.:1.000    1st Qu.:2.500    1st Qu.:   5.0
## Median :  400      Median :2.000    Median :3.200    Median :  31.0
## Mean   : 1200      Mean   :1.805    Mean   :2.665    Mean   : 156.8
## 3rd Qu.:  700      3rd Qu.:2.000    3rd Qu.:3.700    3rd Qu.: 130.0
## Max.   :800000      Max.   :4.000    Max.   :4.900    Max.   :10934.0
```

Calculate standard deviation for numerical columns

```
sds <- sapply(data[, sapply(data, is.numeric)], sd, na.rm = TRUE)
```

```
print("Standard deviation for numerical columns:")
```

```
## [1] "Standard deviation for numerical columns:"
```

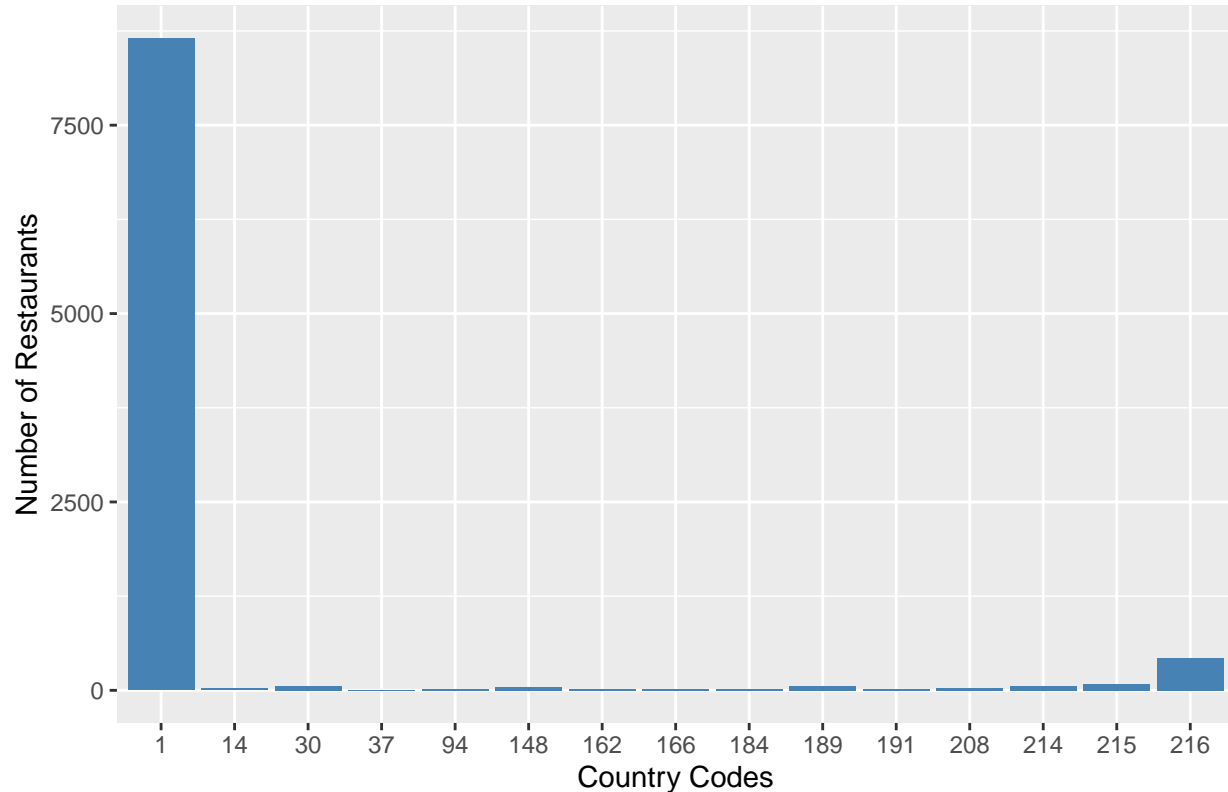
```
print(sds)
```

```
## Restaurant.ID      Country.Code      Longitude
## 8.791967e+06      5.645160e+01      4.119760e+01
## Latitude Average.Cost.for.two      Price.range
## 1.101009e+01      1.612874e+04      9.055631e-01
## Aggregate.rating      Votes
## 1.516588e+00      4.302033e+02
```

```
# The Distribution of Categorical Variables like 'Country Code', 'City', and 'Cuisines'

# Create count plot for Country Code
ggplot(data, aes(x = factor(Country.Code))) +
  geom_bar(fill = "steelblue") +
  labs(title = "Distribution of Restaurants by Country Codes",
       x = "Country Codes", y = "Number of Restaurants")
```

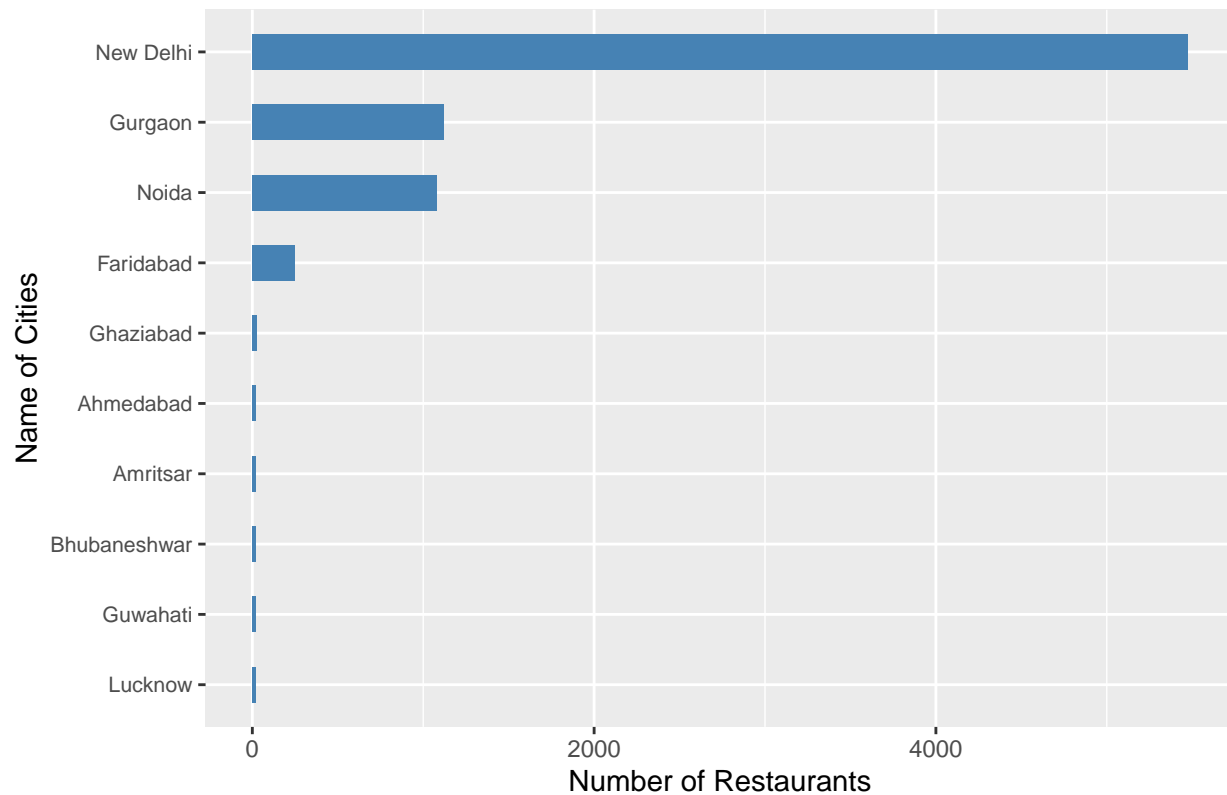
Distribution of Restaurants by Country Codes



```
# Create a subset of the data containing only the top 10 cities
top_10_cities <- head(names(sort(table(data$City), decreasing = TRUE)), 10)
data_top_10_cities <- data[data$City %in% top_10_cities, ]

# Create count plot for top 10 cities (horizontal bar plot)
ggplot(data = data_top_10_cities, aes(y = factor(City, levels = rev(top_10_cities)))) +
  geom_bar(fill = "steelblue", width = 0.5, stat = "count") +
  labs(title = "Top 10 Cities with Highest Number of Restaurants",
       x = "Number of Restaurants", y = "Name of Cities") +
  theme(axis.text.y = element_text(size = 8))
```

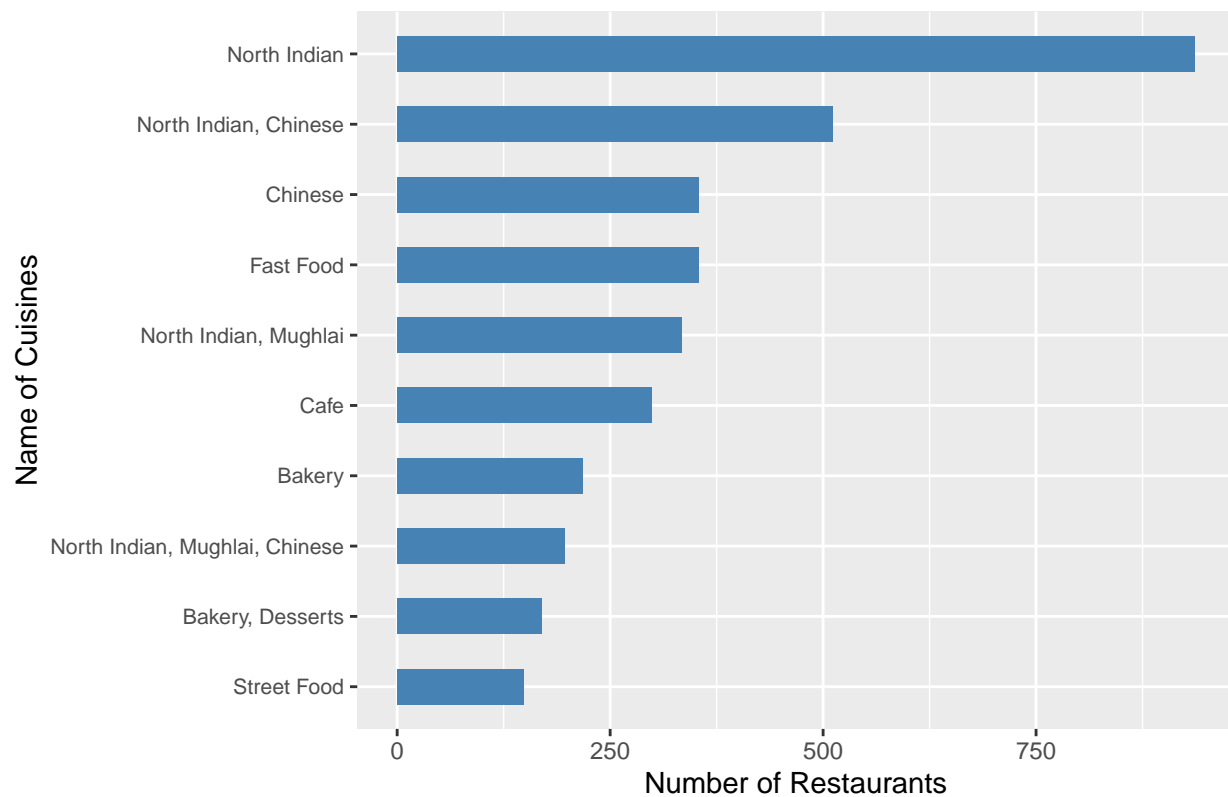
Top 10 Cities with Highest Number of Restaurants



```
# Subset the data to include only the top 10 cuisines
top_10_cuisines <- head(names(sort(table(data$Cuisines), decreasing = TRUE)), 10)
data_top_10 <- data[data$Cuisines %in% top_10_cuisines, ]

# Create count plot for top 10 cuisines (horizontal bar plot)
ggplot(data = data_top_10, aes(y = factor(Cuisines, levels = rev(top_10_cuisines)))) +
  geom_bar(fill = "steelblue", width = 0.5, stat = "count") +
  labs(title = "Top 10 Cuisines with Highest Number of Restaurants",
       x = "Number of Restaurants", y = "Name of Cuisines") +
  theme(axis.text.y = element_text(size = 8))
```


Top 10 Cuisines with Highest Number of Restaurants



```
# The top cuisines and cities with the highest number of restaurants

# Identify the top 10 cuisines and their counts
top_cuisines <- head(sort(table(data$Cuisines), decreasing = TRUE), 10)

# Create a dataframe with cuisine names and counts
top_cuisines_df <- data.frame(Cuisine = names(top_cuisines), Count = as.numeric(top_cuisines))

# Display the dataframe
print("Top 10 Cuisines with the Highest Number of Restaurants:")
```

```
## [1] "Top 10 Cuisines with the Highest Number of Restaurants:"
```

```
print(top_cuisines_df)
```

```
##           Cuisine Count
## 1      North Indian  936
## 2 North Indian, Chinese  511
## 3           Chinese  354
## 4          Fast Food  354
## 5 North Indian, Mughlai  334
## 6             Cafe  299
## 7            Bakery  218
## 8 North Indian, Mughlai, Chinese  197
## 9      Bakery, Desserts  170
```

```
## 10 Street Food 149
```

```
# Identify the top 10 city and their counts
top_city <- head(sort(table(data$City), decreasing = TRUE), 10)

# Create a dataframe with city names and counts
top_city_df <- data.frame(City = names(top_city), Count = as.numeric(top_city))

# Display the dataframe
print("Top 10 Cities with the Highest Number of Restaurants:")
```

```
## [1] "Top 10 Cities with the Highest Number of Restaurants:"
```

```
print(top_city_df)
```

```
##      City Count
## 1 New Delhi 5473
## 2 Gurgaon 1118
## 3 Noida 1080
## 4 Faridabad 251
## 5 Ghaziabad 25
## 6 Ahmedabad 21
## 7 Amritsar 21
## 8 Bhubaneswar 21
## 9 Guwahati 21
## 10 Lucknow 21
```

Task 3: GEOSPATIAL ANALYSIS

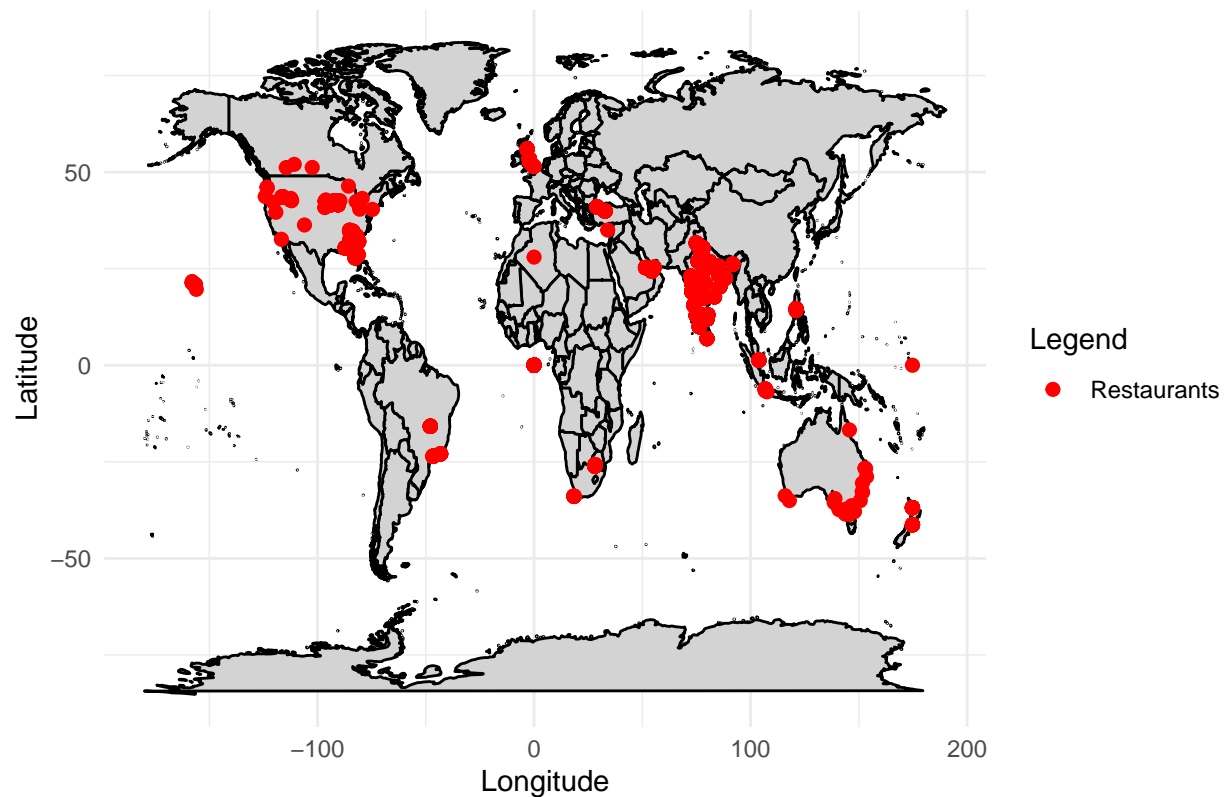
```
# Visualize the locations of restaurants on a map
```

```
# Create a map of the world
world_map <- map_data("world")
```

```
# Plot restaurant locations on the map
```

```
ggplot() +
  geom_polygon(data = world_map, aes(x = long, y = lat, group = group), fill = "lightgrey", color = "black") +
  geom_point(data = data, aes(x = Longitude, y = Latitude, color = "Restaurants"), size = 2) +
  scale_color_manual(name = "Legend", values = c(Restaurants = "red")) +
  labs(title = "Restaurant Locations on World Map", x = "Longitude", y = "Latitude") +
  theme_minimal()
```

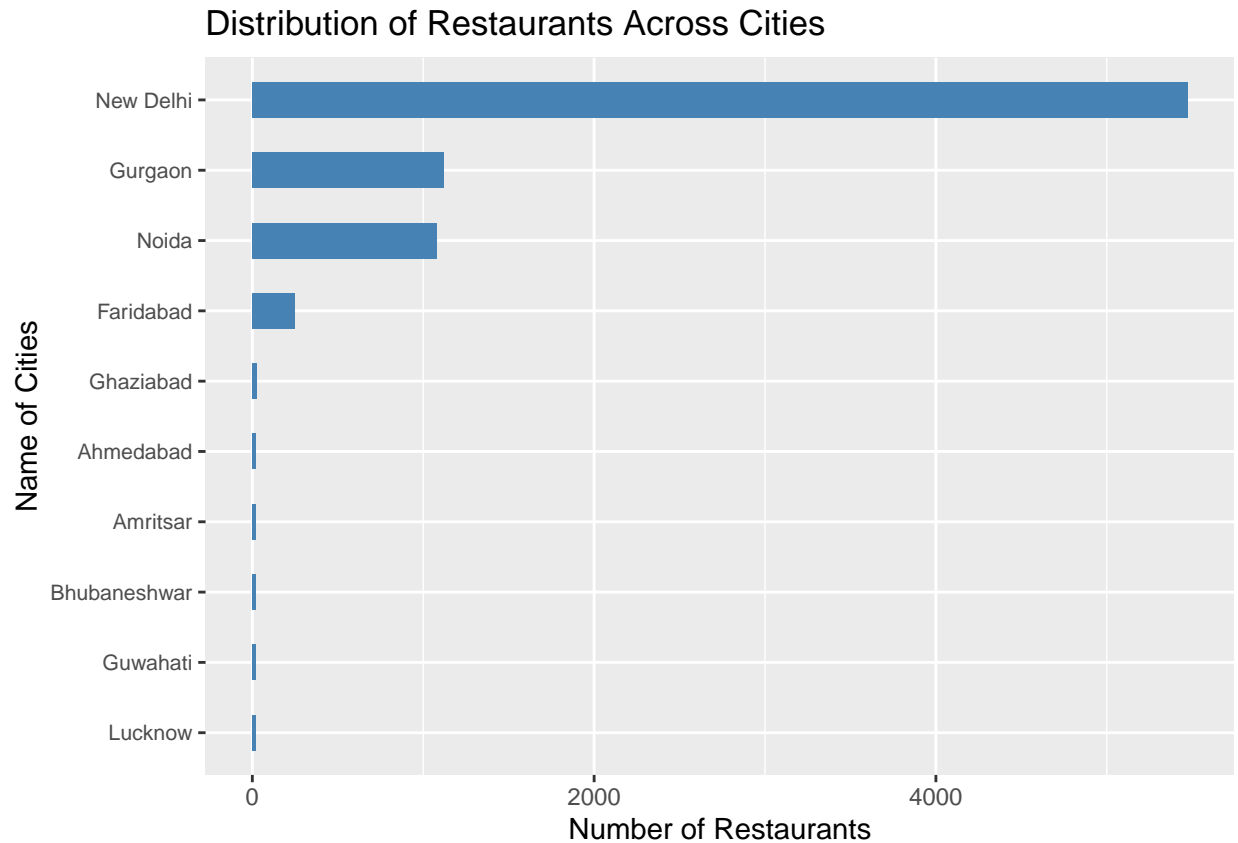
Restaurant Locations on World Map



```
# Analyze the distribution of restaurants across different cities or countries

# Create a subset of the data containing only the top 10 cities
top_10_cities <- head(names(sort(table(data$City), decreasing = TRUE)), 10)
data_top_10_cities <- data[data$City %in% top_10_cities, ]

# Create a plot of the distribution of restaurants across cities
ggplot(data = data_top_10_cities, aes(y = factor(City, levels = rev(top_10_cities)))) +
  geom_bar(fill = "steelblue", width = 0.5, stat = "count") +
  labs(title = "Distribution of Restaurants Across Cities",
       x = "Number of Restaurants", y = "Name of Cities") +
  theme(axis.text.y = element_text(size = 8))
```



Determine if there is any correlation between the restaurant's location and its rating

Calculate the correlation matrix

```
correlation_matrix <- cor(data[c("Latitude", "Longitude", "Aggregate.rating")])
```

Create a heatmap to visualize the correlation

```
corrplot(correlation_matrix, method="color", col=colorRampPalette(c("blue", "white", "red"))(20), type=
  order="hclust", tl.col="black", tl.srt=45, title="Correlation Between Restaurant's Location and",
  mar=c(0, 0, 3, 1))
```

Correlation Between Restaurant's Location and Rating

