


# Data Description

## Zomato Restaurant Dataset

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This dataset is all about restaurants listed on Zomato in India.  Each row represents a restaurant, and we get to see its location, cost, delivery options, user ratings, and more. Our goal? To figure out what makes a restaurant *successful* — based on the ratings it receives!

### What's in the Dataset?

There are thousands of restaurants across different Indian cities, and for each one, we have:

Feature	What it Means
Longitude	East-west GPS coordinate of the restaurant
Latitude	North-south GPS coordinate of the restaurant
Average Cost for two	Average meal price for two people
Has Table booking	Whether it accepts table bookings (Yes/No → 1/0)
Has Online delivery	Whether it offers online food delivery (Yes/No → 1/0)
Price range	Pricing tier assigned by Zomato (1 = low, 4 = high)

Aggregate rating	Average user rating of the restaurant (0.0 to 5.0)
Votes	Number of user votes submitted on Zomato
Successful	Target variable — 1 if rating > 4.0, else 0 (indicates popularity/success)
alcohol	Alcohol percentage
quality	A score from 0 to 10 given to each wine (this is what we want to predict!)

## What Cleaning Did We Do?

### Filtered to Indian Restaurants:

We only kept rows where `Country Code == 1`, so we're looking at Indian data.

### Dropped Unnecessary Columns:

Removed IDs, verbose addresses, and features not useful for modeling.

### Handled Missing Data:

Dropped rows where the rating (`Aggregate rating`) was missing.

### Encoded Categorical Variables:

Converted "Yes"/"No" columns like `Has Table booking` and `Has Online delivery` into 1s and 0s.

### Outlier Treatment:

For numeric features like cost and votes, we used **IQR** (interquartile range) to remove extreme outliers.

## Fixed Skewed Data:

We applied **log transformation** to highly skewed features — this makes models work better.

## What Are We Trying to Predict?

We introduced a new column: **Successful**

- **1** → The restaurant is successful (rating > 4.0)
- **0** → Otherwise

This turns our task into a **classification problem** — predicting which restaurants are likely to be successful on Zomato.

## What Did We Learn from the Data?

- Restaurants with **more votes and higher price ranges** often have better ratings ★
- **Online delivery** and **table booking** aren't always linked with higher success
- Some locations (by **Latitude/Longitude**) show clusters of higher ratings
- We used a **heatmap** to visualize how all features relate to each other