

Restaurants Rating Prediction using Machine Learning Algorithms



Vicky Malik, S.Prasad Babu Vagolu, Sunil Chandolu

Abstract: Restaurant Rating has become the most commonly used parameter for judging a restaurant for any individual. A lot of research has been done on different restaurants and the quality of food it serves. Rating of a restaurant depends on factors like reviews, area situated, average cost for two people, votes, cuisines and the type of restaurant. The project aim is to find out the relationship between the dependent and independent variable. Proposed project is a Machine Learning Regression problem which uses Restaurant Rating dataset. Based on various attributes like the food, quality, prize ambience of the restaurant it predicts the Restaurant Rating.

Keywords: Restaurant Rating, Random Forest Algorithm, Linear Regression, Machine Learning Algorithm.

I. INTRODUCTION

In today's digitized modern world, popularity of food apps is increasing due to its functionality to view, book and order for food by a few clicks on the phone for their favorite restaurant or cafes, by surveying the user ratings and reviews of the previously visited customers. Restaurant Rating also provides columns for writing classified user reviews. Such sort of substance provided by web is named as client produced content. Client created content contains a great deal of significant and essential data about the food items and restaurant administrations. Since there is no control on the nature of this substance on the web and thus, these elevate fraudsters to compose counterfeit surveys to defame the restaurant administrations, to provide misguiding reviews, to generate irrelevant content regardless of the product or service, to advertise unrelated content, etc. These phony surveys anticipate clients and associations achieving genuine decisions about the product, services, and amenities of the restaurants or cafes. In this case, Review Analysis has become vital to generate authenticated and unbiased reviews which help in avoiding fraudulent activities used to promote business by publishing fake reviews.

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Hereby in this paper we focus on mining customer reviews, authenticate them, classify them into positive and negative reviews, and find worthiness of the product.

Different machine learning algorithms like SVM, Linear regression, Decision Tree, Random Forest can be used to predict the ratings of the restaurants.

II. DATA SET DESCRIPTION

This is a kaggle dataset.

(https://www.kaggle.com/himanshupoddar/zomato-bangalor e- restaurants).

It Represents information of Restaurants in the City of Bangalore. It contains 17 Columns and 51,000 Rows.

The dataset has the following attributes such as: Restaurant Name, Restaurant ID, City, Address, Cuisines, Cost for two people, has table booking, Has online delivery, Is delivering now, Switch to order menu, Prize range, Aggregate rating, Rating color, Rating text and votes.

So, for the restaurant to have a higher rating the customer rating plays an important role, and if the rating of the restaurant is higher it will also bring new customers to restaurants. The customer relationship plays a very important role for the success and profit in business.

A	В	C	0	E	F	G	Н	1	J	K	L	M	N	0	P	Q
1 url	address	name	online_or	r book_tab	rate	votes	phone	location	rest_type	dish_like	couisines	approx_c	creviews	menu_i	te listed_	in(flisted_in(city
2 https://	w 942, 21st	Malsa	Yes	Yes	4.1/5	77	080	Banashan	Casual Dir	Pasta, Lu	n North Ind	800	('Rated	1.[]	Buffet	Banashankar
3 https://	w 2nd Floor	r, Spice Elep	Yes	No	4.1/5	78	080 41714	Banashan	Casual Dir	Mamos,	Li Chinese,	800	[('Rated	1[]	Buffet	Banashankar
4 https://	w 1112, Ne	r San Churr	Yes	No	3.8/5	918	+91 96634	Banashan	Cafe, Casu	Churros,	CCafe, Mex	800	[('Rated	3.[]	Buffet	Banashankar
5 https://	w 1st Floor,	. Addhuri U	No	No	3.7/5	8	+91 96200	Banashan	Quick Bite	Masala D	c South Ind	300	[('Rated	1.[]	Buffet	Banashankar
6 https://	w 10, 3rd Fl	c Grand Vill	No	No	3.8/5	160	+91	Basavana	Casual Dir	Panipuri	North Ind	600	('Rated	1.[]	Buffet	Banashankar
7 https://	w 37, 5-1, 4	t Timepass	Yes	No	3.8/5	286	+91	Basavana	Casual Dir	Onion Ri	n North Ind	600	[('Rated	3.[]	Buffet	Banashankar
8 https://	/w 19/1, Ne	n Rosewood	No	No	3.6/5	1	+91	Mysore R	Casual Dir	ning	North Ind	800	[('Rated !	5.[]	Buffet	Banashankar
9 https://	w 2469, 3rd	Onesta	Yes	Yes	4.6/5	2556	080	Banashan	Casual Dir	Farmhou	is Pizza, Caf	600	[('Rated !	5.[]	Cafes	Banashankar
10 https://	w 1, 30th M	a Penthous	Yes	No	4.0/5	324	+91	Banashan	Cafe	Pizza, Mo	c Cafe, Itali	700	('Rated	3.0', "RAT	ED\n I ha	d been to this pla
11 e are a	bit things te	r service w	('Rated 4	. 'RATED\n	\nTop fl	oc no outdo	though th	a nice pla	('Rated 1.	'RATED\	n we had n	it turned	('Rated 3	. "RATE	D\ parkin	g fa it was litt []
12 https://	/w 2470, 21 I	V Smaczneg	Yes	No	4.2/5	504	+91	Banashan	Cafe	Waffles,	F Cafe, Mex	550	('Rated	1[]	Cafes	Banashankar
13 https://	w 12,29 Ne	a CafĀfĀfĀ,	Yes	No	4.1/5	400	080	Banashan	Cafe	Waffles,	FCafe	500	[('Rated	1.[]	Cafes	Banashankar
4 https://	w 941, 3rd F	Cafe Shuf	Yes	Yes	4.2/5	150	+91 97421	Banashan	Cafe	Mocktail	s, Cafe, Itali	600	('Rated	1.[]	Cafes	Banashankar
5 https://	w 6th Block	, The Coffe	Yes	Yes	4.2/5	16	+91 97316	Banashan	Cafe	Coffee, S	cafe, Chir	500	[('Rated	1[]	Cafes	Banashankar
l6 https://	w 111, Sapp	d Caf-Eleve	No	No	4.0/5	424	080 49577	Banashan	Cafe	Sandwid	h, Cafe, Con	450	[('Rated	2.[]	Cafes	Banashankar
17 https://	w 1112, Ne	r San Churr	Yes	No	3.8/5	918	+91 96634	Banashan	Cafe, Casi	Churros,	CCafe, Mex	800	[('Rated	3.[]	Cafes	Banashankar
18 https://	w 2303, 21s	t Cafe Viva	Yes	No	3.8/5	9	080	Banashan	Cafe	Garlic Br	e Cafe	650	('Rated	2.[]	Cafes	Banashankar
19 https://	w 241, 4th F	Catch-up-	Yes	No	3.9/5	133	+91	Banashan	Cafe	Mamas,	N Cafe, Fast	800	[('Rated	L[]	Cafes	Banashankar
0 https://	w 405, 24th	(Kirthi's Bir	Yes	No	3.8/5	14	080	Banashan	Cafe	Pasta, Ge	el Chinese,	700	[('Rated	3.[]	Cafes	Banashankar
1 https://	w 504, CJ V	e T3H Cafe	No	No	3.9/5	99	+91 88847	Banashan	Cafe	Cheese I	M Cafe, Itali	300	[('Rated	1.[]	Cafes	Banashankar
22 https://	w 47, 48 &4	9 360 Atoms	Yes	No	3.1/5	13	+91 98805	Banashan	Cafe		Cafe, Chir	400	('Rated	5.[]	Cafes	Banashankar
23 https://	w 146, 50 ft	The Vinta	Yes	No	3.0/5	6	+91	Banashan	Cafe	Burgers,	C Cafe, Frei	400	[('Rated	2.[]	Cafes	Banashankar
4 https://	w 3353, 2nd	Woodee F	Yes	No	3.7/5	180	+91 74068	Banashan	Cafe	Pizza, Ga	r Cafe, Pizz	500	[('Rated	3.[]	Cafes	Banashankar
25 https://	w SRF Com	o Cafe Coffe	No	No	3.6/5	20	080 32486	Banashan	Cafe		Cafe, Fast	900	[('Rated	1.[]	Cafes	Banashankar

Pre Processing

The Dataset contained 17 Attributes.



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- Records with null values were dropped from ratings columns and were replaced in the other columns with a numerical value.
- Values in the 'Rating' column were changed. The '/5' string was deleted. For eg. If the rating of a restaurant was 3.5/5, it was changed to 3.5.
- Using LabelEncoding from sklearn library, encoding was done on columns like

book_table,online_order,rest_type,listed_in(city).

III. FEATURE SELECTION

We did not use any feature selection algorithms but eliminated some columns due to available domain knowledge and thorough study of the system.

Dropped columns mentioned below:

- URL
- Address
- Dish liked
- Phone
- Menu
- Review_list
- Location
- Cuisine

Some of these columns may look like they are important but all of the same information could be found in other columns with lesser complexity.

The Columns being used are as follows:

- Name
- Online_order
- Book_table
- Votes
- Rest type
- Approx. cost of two people
- Listed_in(type)
- Listed_in(city)

IV. EXPLORATORY DATA ANALYSIS

A lot of effort went into the EDA as it gives us a detailed knowledge of our data.

Exploratory Data Analysis (EDA) is an approach/philosophy for data analysis that employs a variety of techniques (mostly graphical) to

- Maximize insight into a data set;
- Uncover underlying structure;
- Extract important variables;
- Detect outliers and anomalies;
- Test underlying assumptions;
- Develop parsimonious models; and
- Determine optimal factor settings.

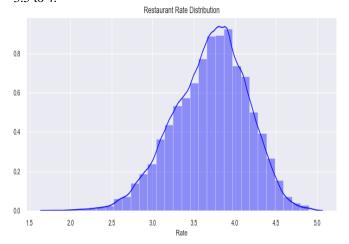
A.Restaurant Rate Distribution

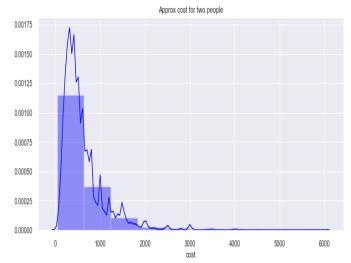
We can see that the number of restaurants with the rating between 3.5 and 4 are the highest. We will look into its dependencies further.

B.Approximate Cost of two people

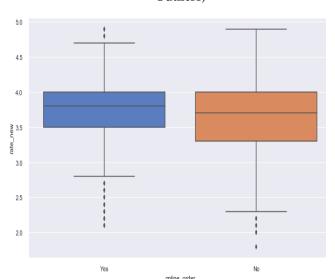
This is a graph for the 'Approximate cost of 2 people' for dining in a restaurant. Restaurants with this cost below 1000 Rupees are more.

This box plot helps us look into the outliers. We can also see that online ordering service also affects the rating. Restaurants with online ordering service have a rating from 3.5 to 4.





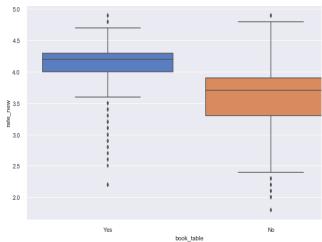
C. Online ordering with respect to Rating(Finding Outliers)



D. Booking table with respect to rating (Finding Outliers)

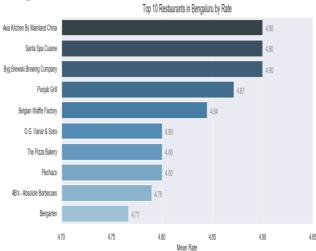






This box plot also helps us look into the outliers. This box plot is regarding how table booking availability is seen in restaurants with rating over 4.

F. Top Rated Restaurants



V. REVIEW CRITERIA

This graph just showcases the best restaurants in Bangalore along with their rating.

G. Cost and Rate Distribution according to online ordering and booking table

A very important scatter plot shows the correspondence between the cost, online ordering, bookings and rating of the restaurant.

4.1. Key Findings

	Votes	approx_cost(for two people)	Rating
online_order			
No	367.992471	716.025190	3.658071
Yes	343.228663	544.365434	3.722440

	Votes	approx_cost(for two people)	Rating
Book_table			
No	204.580566	482.404625	3.620801
Yes	1171.342957	1276.491117	4.143464

VI. RESULTS

Algorithms	Accuracy
Linear Regression	30%
KNN	44%
Support Vector Machine	43%
Decision Tree	69%
Random Forest	81%

In this model, we have considered various restaurants records with features like the name, average cost, locality, whether it accepts online order, can we book a table, type of restaurant. This model will help business owners predict their rating on the parameters considered in our model and improve the customer experience.

Different algorithms were used but in the end the final model is selected on Random Forest Algorithm which gives the

highest accuracy compared to others. Cost and Rate Distribution by Book Table Cost and Rate Distribution by Online Order

	Cost and	Rate Distri	d noisud	y Online C	Jraer	Cost	and Rate Di	stribution	Dy Book	lable
6000	on • Ye • No			•	6000	•	book_table Yes No		•	
5000					5000					
				•					٠	
4000					4000				:	
two people				·	two people				•	
approx_cost(ror two people)					approx_cost(for two people)					
2000		:			2000					
1000					1000	•				
0					0					
	2.0	2.5 3.0 rate	3.5 4 e_new	4.0 4.5	5.0	2.0		3.5 ate_new	4.0 4.5	5.0

VII. CONCLUSIONS

This project performs both multinomial classification in terms of rating prediction and binary classification in terms of popularity change prediction. In this model, we have considered various restaurants records with features like the name, average cost, locality, whether it accepts online order, can we book a table, type of restaurant.

This model will help business owners predict their rating on the parameters considered in our model and improve the customer experience.

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This paper studies a number of features about existing restaurants of different areas in a city and analyses them to predict rating of the restaurant. This makes it an important aspect to be considered, before making a dining decision. Such analysis is essential part of planning before establishing a venture like that of a restaurant.

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