Subject Code: BUAN 6337.002 - Predictive Analytics using SAS

A Group Project Report on

ZOMATO PREDICTIVE ANALYSIS

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It has been a great opportunity to gain a lot of experience in real time projects, followed by the knowledge of how to design and analyse them. For that we would like to thank our professor, **Dr Shervin Tehrani**, who made it possible. We would like to acknowledge him for his efforts in providing us with useful information and for making the path clear.

INTRODUCTION

Background:

Zomato is an Indian restaurant search and discovery service founded in 2008 by Deepinder Goyal and Pankaj Chaddah. It currently operates in 24 countries. It provides information and reviews of restaurants, including images of menus where the restaurant does not have its own website and also online delivery.

In 2011, Zomato launched in Bengaluru, Pune, Chennai, Hyderabad and other cities. In September 2012, Zomato expanded overseas to the UAE, Sri Lanka, Qatar, UK, the Philippines, and South Africa. In April 2014, Zomato launched its services in Portugal, followed by launches in Canada, Lebanon and Ireland the same year. The acquisition of Seattle-based food portal Urbanspoon marked the firm's entry into the United States, Canada and Australia, and brought it into direct competition with Yelp, Zagat and OpenTable.

Objective:

The objective of our project or the research question which we were considering are as follows:

- What factors drive the customers to vote for the restaurants on Zomato?
- How people rate a restaurant based on certain parameters?

As, we can see that Zomato takes input from the customer to rate the restaurants which can be further used by other customers. So, we are using the inputs given by the data to find the factors which affect how the customers rate a restaurant.

Data Description

The data which was given to us can be summarised as follows:

- Initially we had 9,551 Restaurants and 21 attributes.
- The total number of countries were 15 in the dataset.
- 8,649 restaurants belong to India and 899 restaurants from other 14 countries of the globe
- Hence, we chose India for our major analysis of the restaurants.

Data Fields:

Variable Name	Description
Restaurant Id	Unique identifier for each restaurant in dataset
Restaurant Name	Restaurant names
Country Code	Country code in which the restaurant is situated
City	City names in which the restaurant is situated
Address	Address of the restaurant
Locality	Locality in which the restaurant is situated
Longitude	Latitude of the restaurant is situated.
Cuisines	Different types of cuisines provided by the restaurant
Average Cost for two	Average cost for two people visiting the restaurant.
Currency	Currency in which the average cost is mentioned
Has Table booking	Binary variable for table booking.
Has Online delivery	Binary variable if the restaurant has online delivery or not.
Price range	Expected price range for two people in the restaurant
Aggregate Rating	Average rating of the restaurant
Votes	Number of votes received by the votes

Table 1: Data Description

Data Pre-Processing & Exploratory Data Analysis

- The Dataset had 9551 records, out of which there were in 9 missing values in the Average
 Cost for Two columns and 1093 restaurants that received no votes, these rows were
 dropped before we began our exploratory analysis.
- The country codes were changed to the country name using the country codes file that
 was provided along with the dataset. The original Dataset had 75 unique cuisines which
 were classified into 15 primary cuisines.
- Since, India had the maximum number of restaurants (8649), it was selected for analysis and grouped into 'New Delhi' and 'Others'.
- New Delhi had 5,743 restaurants while all other remaining were in other cities of India.
- New Variables like Number of Cuisines offered by the restaurant and Major Cuisines were synthesized.



Figure 1: Location of Countries

 From Figure 1, we observe that the dataset had 15 countries, out of which New Zealand had no data. Since India had the greatest number of restaurants (8649) while all others countries combined had a 899 restaurants. As a result, India was selected for further analysis.

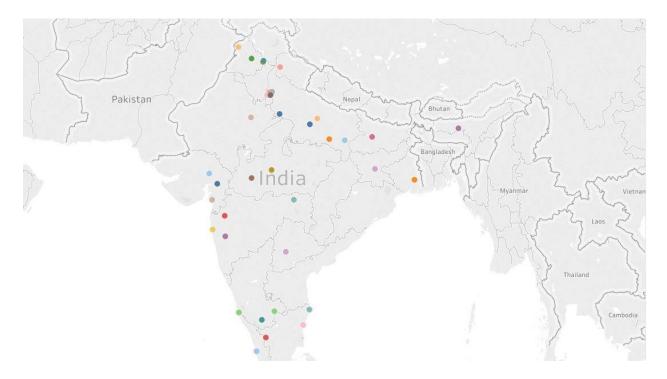


Figure 2: Location of Cities within India

• Figure 2 shows us the different location of cities within India. A total of 43 distinct cities in India were present within the dataset, however since 5,473 Restaurants were present in the capital New Delhi while the rest 3,170 were located in the other 42 cities of India, so were grouped the cities column into 'New Delhi' and 'Others'.



BUAN 6337.002 – GROUP PROJECT – ZOMATO PREDICTIVE ANALYSIS Figure 3.a: Average of Aggregate Ratings for Major Cuisines

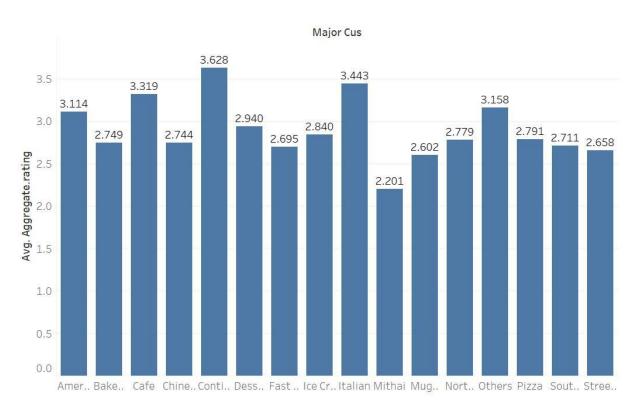
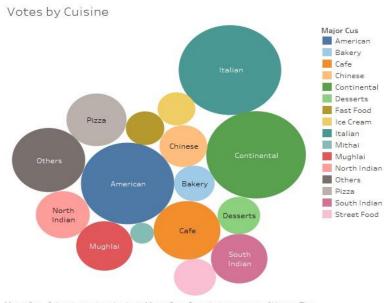


Figure 3.b: Average Aggregate Ratings for Major Cuisines

- Figure 3.a shows the average of Aggregate ratings for Major Cuisines. We see that American, Continental and Italian have higher aggregate ratings in comparison to the other cuisines which is a little surprising considering that most of the restaurants are situated in New Delhi from where the North Indian cuisine originates.
- From Figure 3.b we see that the average Aggregate ratings for Continental is 3.628 followed by Italian (3.443) and American (3.114). North Indian, the traditional food of New Delhi where the maximum restaurants are located has an average aggregate rating of 2.779.



Major Cus. Color shows details about Major Cus. Size shows average of Votes. The

Figure 4.a: Average of votes for Major Cuisines

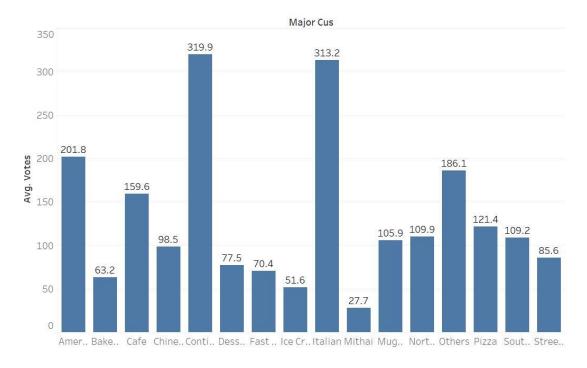


Figure 4.b: Average of votes for Major Cuisines

- Figure 4.a shows the average of Votes for Major Cuisines. We see that American, Continental and Italian have higher average votes in comparison to the other cuisines which is a little surprising considering that most of the restaurants are situated in New Delhi from where the North Indian cuisine originates.
- From Figure 4.b we see that the average votes for Continental is 319.9 followed by Italian (313.2) and American (201.8). North Indian, the traditional food of New Delhi where the maximum restaurants are located has an average vote of 109.9.

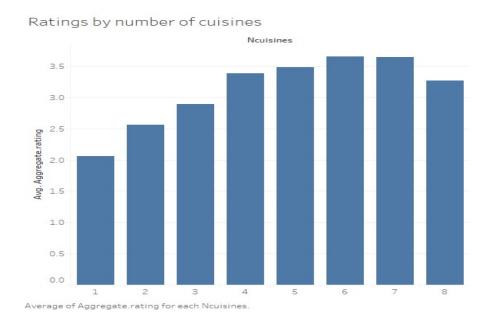


Figure 5: Average Aggregate Ratings according to number of cuisines offered by the restaurant.

• From Figure 5, we observe that the average aggregate ratings increase as the number of cuisines offered by the restaurant increase, however after 6 cuisines, the average aggregate ratings experience a drop.

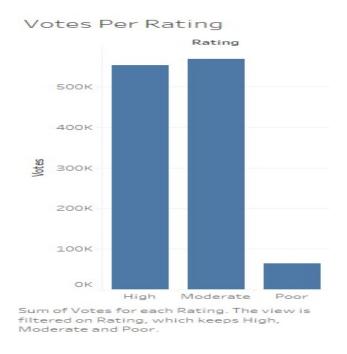


Figure 6: Sum of Votes for Each Rating

• From figure 6, we observe that High and Moderate categories have the most votes.

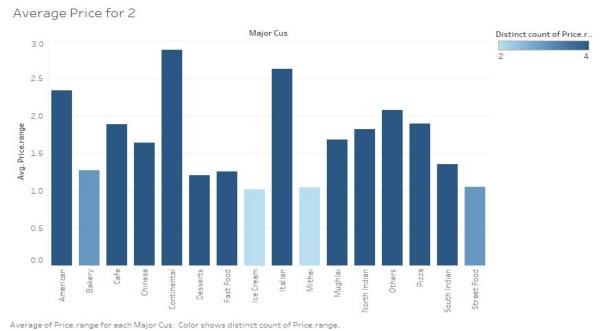


Figure 7: Average Price For Each Major Cuisine

• Figure 7 shows that the continental has the highest average price range followed by Italian and American. The colour coding shows the distinct count of the price range with the lighest being 2 and darkest being 4.

Model and Analysis

We used Linear Regression, LASSO, Forward, Backward and stepwise selection for votes.
 For finding out how people rate a restaurant based on certain parameters, we used logit and probit model in SAS

Linear Regression:

The GLM Procedure							
	Class Level Information						
Class	Levels	Values					
MajorCus	16	American Bakery Cafe Chinese Continental Desserts Fast Food Ice Cream Italian Mithai Mughlai North Indian Others Pizza Street Food South Indian					
City	2	New Delh Others					
Has_Online_delivery	2	Yes No					
Has_Table_booking	2	Yes No					
Price_range	4	2341					

Parameter	Estimate		Standard Error	t Value	Pr > jt
Intercept	-31.8909646	В	30.2848184	-1.05	0.2924
Price_range 2	84.8017998	В	30.7082289	2.76	0.0058
Price_range 3	467.0930047	В	101.9364496	4.58	<.000
Price_range 4	240.6158144	В	219.0375843	1.10	0.2720
Price_range 1	0.0000000	В		9	
MajorCus American	-29.8175059	В	131.8689081	-0.23	0.821
MajorCus Bakery	-14.7624795	В	34.2786299	-0.43	0.666
MajorCus Cafe	10.8668961	В	36.1489382	0.30	0.763
MajorCus Chinese	-14.6584027	В	34.8809085	-0.42	0.674
MajorCus Continental	-22.9393130	В	68.2709752	-0.34	0.7369
MajorCus Desserts	38.7421217	В	49.6972769	0.78	0.435
MajorCus Fast Food	14.3077491	В	34.3481825	0.42	0.6770
MajorCus Ice Cream	-21.8998503	В	44.8143073	-0.49	0.625
MajorCus Italian	103.3281509	В	76.2321372	1.36	0.175
MajorCus Mithai	-25.7370900	В	44.5025644	-0.58	0.563
MajorCus Mughlai	30.1853464	В	56.3518354	0.54	0.592
MaiorCus North Indian	-20.1010493	В	30.9681489	-0.65	0.516
MajorCus Others	10.4523476	В	33.5077813	0.31	0.755
MaiorCus Pizza	-10 5869522	В	58.3383736	-0.18	0.856
MajorCus Street Food	-30.7282721	В	47.8577811	-0.64	0.520
MajorCus South Indian	0.0000000	В	17.0077011	0.01	0.020
City New Delh	14.3725817	В	31.5973537	0.45	0.6493
City Others	0.0000000	В	31.3373337	0.40	0.040
Has Online delivery Yes	61.5362765	В	5.8462725	10.53	<.000
Has Online delivery No	0.0000000	В	3.0402723	10.55	1.000
Has Table booking Yes	29.7382557	В	9.7221175	3.06	0.002
Has Table booking No	0.0000000	В	9.7221173		0.002
ncuisines	62.1643127	D	8.0653140	7.71	<.000
	-81.0283076	300	17.9447150	-4.52	<.000
ncuisines_log	100.00000000000000000000000000000000000	-	100000000000000000000000000000000000000	10.10.000	None Person
MajorCus*Price_range American 2	-74.1636831	В	131.6749452	-0.56	0.5733
MajorCus*Price_range American 3	-245.5366443	В	168.5513107	-1.46	0.145
MajorCus*Price_range American 4	287.4132026	В	260.2512606	1.10	0.269
MajorCus*Price_range American 1	0.0000000	В	(0)		
MajorCus*Price_range Bakery 2	-42.5101803	В	38.4336089	-1.11	0.268
MajorCus*Price_range Bakery 3	-428.8489007	В	119.0330236	-3.60	0.000
MajorCus*Price_range Bakery 1	0.0000000	В	100		
MajorCus*Price_range Cafe 2	2.0681446	В	37.6946508	0.05	0.9562
MajorCus*Price_range Cafe 3	-272.5184685	В	106.1852194	-2.57	0.010
MajorCus*Price_range Cafe 4	-159.1174760	В	231.5063592	-0.69	0.4919
MajorCus*Price_range Cafe 1	0.0000000	В			
MajorCus*Price_range Chinese 2	-57.9747022	В	36.1183694	-1.61	0.108
MajorCus*Price_range Chinese 3	-251.4000809	В	104.8203691	-2.40	0.016
MajorCus*Price_range Chinese 4	-133.2889062	В	223.2014744	-0.60	0.550
MajorCus*Price_range Chinese 1	0.0000000	В			

Parameter	Estimate		Standard Error	t Value	Pr>
Intercept	-31.8909646	В	30.2848184	-1.05	0.292
Price_range 2	84.8017998	В	30.7082289	2.76	0.005
Price_range 3	467.0930047	В	101.9364496	4.58	<.000
Price_range 4	240.6158144	В	219.0375843	1.10	0.272
Price_range 1	0.0000000	В			
MajorCus American	-29.8175059	В	131.8689081	-0.23	0.821
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MajorCus Mithai	-25.7370900	В	44.5025644	-0.58	0.563
MajorCus Mughlai	30.1853464	В	56.3518354	0.54	0.592
MajorCus North Indian	-20.1010493	В	30.9681489	-0.65	0.516
MajorCus Others	10.4523476	В	33.5077813	0.31	0.755
MaiorCus Pizza	-10.5869522	В	58.3383736	-0.18	0.856
MajorCus Street Food	-30.7282721	В	47.8577811	-0.64	0.520
MajorCus South Indian	0.0000000	В			
City New Delh	14.3725817	В	31 5973537	0.45	0.649
City Others	0.0000000	В	01.0010001	0.10	0.010
Has Online delivery Yes	61.5362765	В	5.8462725	10.53	<.000
Has Online delivery No	0.0000000	В			
Has Table booking Yes	29.7382557	В	9.7221175	3.06	0.002
Has Table booking No	0.0000000	В	0.1221113	3.00	0.002
ncuisines	62.1643127		8.0653140	7.71	<.000
ncuisines log	-81.0283076		17.9447150	-4.52	<.000
MajorCus*Price range American 2	-74.1636831	В	131.6749452	-0.56	0.573
MajorCus*Price_range American 2	-245.5366443	В	168.5513107	-1.46	0.145
MajorCus*Price_range American 4	287 4132026	B	260 2512606	1 10	0.145
	0.0000000	В	200.2512000	1.10	0.208
MajorCus*Price_range American 1	1010000000000	В	20 4220000	4.44	0.200
MajorCus*Price_range Bakery 2	-42.5101803 -428.8489007	В	38.4336089 119.0330236	-1.11 -3.60	0.268
MajorCus*Price_range Bakery 3			119.0330236	-3.60	0.000
MajorCus*Price_range Bakery 1	0.0000000	В	07.00.05***		0.05
MajorCus*Price_range Cafe 2	2.0681446	В	37.6946508	0.05	0.956
MajorCus*Price_range Cafe 3	-272.5184685	В	106.1852194	-2.57	0.010
MajorCus*Price_range Cafe 4	-159.1174760	В	231.5063592	-0.69	0.491
MajorCus*Price_range Cafe 1	0.0000000	В			
MajorCus*Price_range Chinese 2	-57.9747022	В	36.1183694	-1.61	0.108
MajorCus*Price_range Chinese 3	-251.4000809	В	104.8203691	-2.40	0.016
MajorCus*Price_range Chinese 4	-133.2889062	В	223.2014744	-0.60	0.550
MajorCus*Price_range Chinese 1	0.0000000	В		-	

MajorCus*City American New Delh	-9.8243682	В	50.4831607	-0.19	0.8457
MajorCus*City American Others	0.0000000	В			
MajorCus*City Bakery New Delh	-1.4283496	В	37.5014333	-0.04	0.9696
MajorCus*City Bakery Others	0.0000000	В			
MajorCus*City Cafe New Delh	-17.5106656	В	37.0332509	-0.47	0.6363
MajorCus*City Cafe Others	0.0000000	В			
MajorCus*City Chinese New Delh	-18.5603857	В	36.6504119	-0.51	0.6126
MajorCus*City Chinese Others	0.0000000	В			
MajorCus*City Continental New Delh	-89.6803026	В	43.3052958	-2.07	0.0384
MajorCus*City Continental Others	0.0000000	В			
MajorCus*City Desserts New Delh	-61.8169396	В	54.6548949	-1.13	0.2581
MajorCus*City Desserts Others	0.0000000	В			
MajorCus*City Fast Food New Delh	-32.5494684	В	37.2703233	-0.87	0.3825
MajorCus*City Fast Food Others	0.0000000	В			
MajorCus*City Ice Cream New Delh	3.5185075	В	50.7443788	0.07	0.9447
MajorCus*City Ice Cream Others	0.0000000	В			
MajorCus*City Italian New Delh	-7.4761634	В	47.1507095	-0.16	0.8740
MajorCus*City Italian Others	0.0000000	В			
MajorCus*City Mithai New Delh	-8.8009357	В	48.7529861	-0.18	0.8567
MajorCus*City Mithai Others	0.0000000	В			
MajorCus*City Mughlai New Delh	-38.5949928	В	52.6380415	-0.73	0.4635
MajorCus*City Mughlai Others	0.0000000	В			
MajorCus*City North Indian New Delh	-19.1574304	В	32.8723615	-0.58	0.5601
MajorCus*City North Indian Others	0.0000000	В			
MajorCus*City Others New Delh	-44.8868105	В	35.0546319	-1.28	0.2004
MajorCus*City Others Others	0.0000000	В			
MajorCus*City Pizza New Delh	-38.2789994	В	46.2889002	-0.83	0.4083
MajorCus*City Pizza Others	0.0000000	В	18		
MajorCus*City Street Food New Delh	47.5395168	В	51.7920335	0.92	0.3587
MajorCus*City Street Food Others	0.0000000	В			
MajorCus*City South Indian New Delh	0.0000000	В			
MajorCus*City South Indian Others	0.0000000	В		- 6	

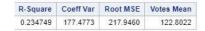
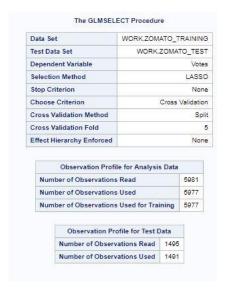
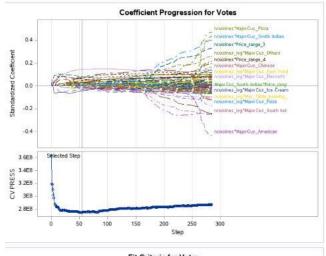


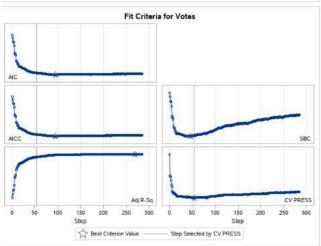
Figure 8: Linear Regression

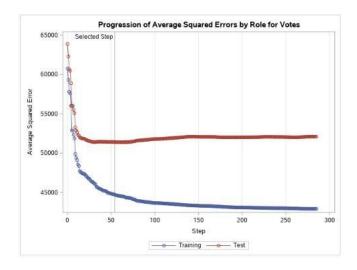
- From Figure 8, we observe that price has an important effect on votes, as the price increases, the votes also increase. Votes increase by the biggest margin for price range 3.
- The votes decrease when the cuisine is North Indian. The interaction between Street Food and Price range 2 have a positive impact on votes.
- If online delivery is available, then voting is bound to increase.
- We have taken log of cuisines to show that very large number of cuisines offered by a restaurant can lead to a fall in votes. Votes with the number of cuisines offered increase up to a certain extent and then decrease.
- If table booking is available, then the votes increase.
- We also observe if that if the city is New Delhi, there is a decline in votes.
- The R-square is very low(0.23), we will use LASSO, forward, backward and best subset selection in order to better explain the variance in the model.

LASSO:









Root MSE	212.35091
Dependent Mean	122.69834
R-Square	0.2641
Adj R-Sq	0.2582
AIC	70080
AICC	70081
SBC	64429
A SE (Train)	44723
ASE (Test)	51405
CV PRESS	275021172

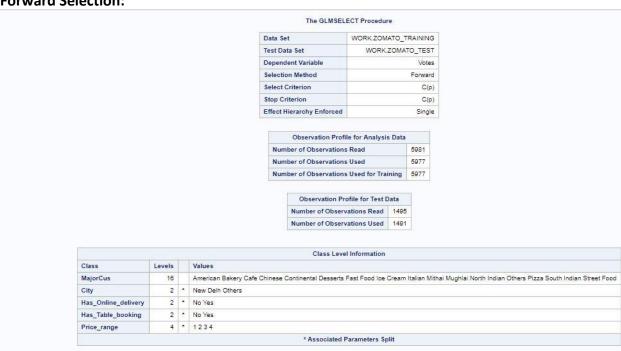
Parameter	DF	Estimate
Intercept	1	86.641759
Price_range_1	1	-21.796978
Price_range_3	1	5.993878
MajorCus_American*Price_range_4	1	188.685505
MajorCus_Bakery*Price_range_3	1	-85.601654
MajorCus_Cafe*Price_range_2	1	10.453361
MajorCus_Cafe*Price_range_3	- 1	-2.712683
MajorCus_Continental*Price_range_3	1	25.066307
MajorCus_Continental*Price_range_4	1	65.997969
MajorCus_Desserts*Price_range_3	140	-37.190825
MajorCus_Italian*Price_range_3	1	80.608659
MajorCus_Italian*Price_range_4	1	-18.532092
MajorCus_North Indian*Price_range_1	1	-5.088000
MajorCus_North Indian*Price_range_2	1	-19.620385
MajorCus_North Indian*Price_range_4	- 1	2.228266
MajorCus_Others*Price_range_2	1	26.018843
MajorCus_Pizza*Price_range_3	1	178.182507
MajorCus_South Indian*Price_range_3	1	182.418641
MajorCus_Street Food*Price_range_2	1	75.003782
City_Others*Price_range_3	1	5.429686
MajorCus_Continental*City_Others	- 14	20.452054
MajorCus_Others*City_Others	1	6.462069
MajorCus Street Food*City New Delh	1	3.274310

Has_Online_delivery_No	1	-6.787581
Has_Online_delivery_No*Price_range_1	1	-16.436438
Has_Online_delivery_No*Price_range_3	1	60.398286
Has_Online_delivery_Yes*Price_range_4	1	236.479368
MajorCus_Cafe*Has_Online_delivery_Yes	1	27.183810
MajorCus_Chinese*Has_Online_delivery_No	1	-1.506727
MajorCus_Continental*Has_Online_delivery_Yes	1	25.867679
MajorCus_Italian*Has_Online_delivery_No	1	40.363956
MajorCus_Pizza*Has_Online_delivery_Yes	1	25.273445
MajorCus_South Indian*Has_Online_delivery_Yes	1	42.138325
MajorCus_Street Food*Has_Online_delivery_Yes	1	24.979851
Has_Table_booking_No*Price_range_3	1	69.353512
Has_Table_booking_No*Price_range_4	1	92.413816
MajorCus_American*Has_Table_booking_Yes	1	72.605654
MajorCus_Italian*Has_Table_booking_Yes	1	87.338286
MajorCus_Others*Has_Table_booking_Yes	1	46.014390
City_Others*Has_Table_booking_Yes	1	50.467180
Has_Online_delivery_Yes*Has_Table_booking_Yes	1	155.438988
ncuisines*Price_range_1	1	-5.424482
ncuisines*Price_range_3	1	18.679749
ncuisines*Price_range_4	1	25.475098
ncuisines*Has_Online_delivery_Yes	1	5.005358
ncuisines_log*MajorCus_Cafe	1	31.118553
ncuisines_log*MajorCus_Continental	1	33.243635
ncuisines_log*MajorCus_North Indian	1	-5.914380
ncuisines*ncuisines_	1	8.836236

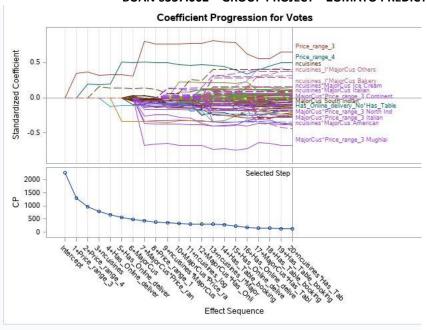
Figure 9: LASSO

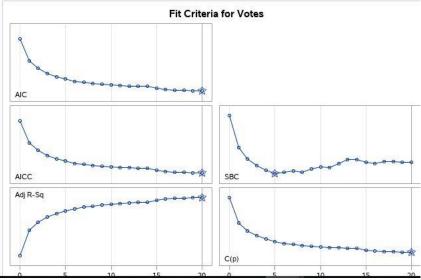
- From the model, we can see that the LASSO has done the selection steps 285 times to find the best effects on votes.
- The R -square value has improved to 0.26 in comparison to the Linear regression Model.
- The AIC is 70080
- In LASSO, people tend to vote for high priced restaurant, the result is same as linear regression but for North Indian, Italian and dessert, high priced restaurants tend to get less votes than lower priced range restaurant.
- If online delivery is available, then voting is bound to increase.
- The R-square is still very low(0.26), we will use forward , backward and best subset selection in order to better explain the variance in the model.

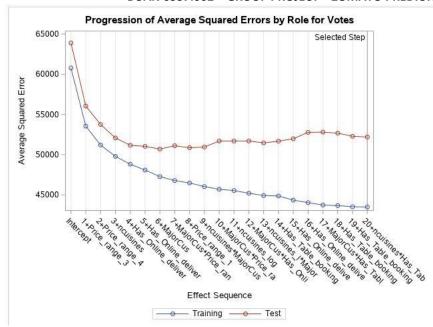
Forward Selection:



	Fo	orward Sele	ection Summ	ary			
Step	Effect Entered		Number Effects In	Number Parms In	СР	ASE	Test A Si
0	Intercept		1	1	2268.1663	60777.4876	63888.547
1	Price_range_3		2	2	1290.0690	53551.1566	56036.564
2	Price_range_4		3	3	972.7389	51196.7111	53753.541
3	ncuisines		4	4	784.2051	49791.8915	52076.047
4	Has_Online_delivery_No		5	5	654.3176	48819.4746	51169.049
5	Has_Online_delivery_No*Price_range_4		6	6	557.4723	48090.6813	51032.441
6	MajorCus		7	21	477.1219	47277.0589	50700.359
7	MajorCus*Price_range_3		8	33	434.9069	46788.8507	51109.137
8	Price_range_1		9	34	395.3641	46482.5524	50864.291
9	ncuisines*MajorCus		10	49	365.2565	46039.3749	50944.407
10	MajorCus*Price_range_4		11	59	338.8403	45697.1449	51695.072
11	ncuisines_log		12	60	319.8738	45542,5574	51690.003
12	MajorCus*Has_Online_delivery_No		13	75	304.0348	45204.5829	51698.573
13	ncuisines_I*MajorCus		14	90	296.0837	44924.7662	51464.077
14	Has_Table_booking_Yes		15	91	290.3098	44867.4485	51672.354
15	Has_Online_delivery_No*Has_Table_book	king_Yes	16	92	222.7983	44354.9356	51979.671
16	Has_Online_delivery_No*Price_range_3		17	93	181,9226	44038.8097	52770.491
17	MajorCus*Has_Table_booking_Yes		18	103	161.0346	43737.3389	52803.662
18	Has_Table_booking_Yes*Price_range_4		19	104	154.3013	43672.9481	52671.430
19	Has_Table_booking_Yes*Price_range_3		20	105	136.8499	43529.5313	52285.395
20	ncuisines*Has_Table_booking_Yes		21	106	134.4076*	43496.7780	52193.211
	*	Optimal Va	lue of Criteri	ion			
	Selection stoppe	ed at a local	minimum of	the C(p) cri	terion.		
	Stop Details						
	Candidate For E	ffect	Candidate C(p)		npare		
	Entry C	ity_Others	134,9338	> 134.	4076		







BUAN 6337.002 - GROUP PROJECT - ZOMATO PREDICTIVE ANALYSIS

Parameter Estimates									
Parameter	DF	Estimate	Standard Error	t Value	Pr > t				
Intercept	1	148.908561	86.369471	1.72	0.0847				
Price_range_1	1	-51.469530	7.455290	-6.90	<.0001				
Price_range_3	1	462.667554	120.787976	3.83	0.0001				
Price_range_4	1	575.155524	137.409975	4.19	<.0001				
MajorCus American	1	-34.941508	120.744719	-0.29	0.7723				
MajorCus Bakery	1	-60.971982	98.613747	-0.62	0.5364				
MajorCus Cafe	1	-94.942178	94.711492	-1.00	0.3162				
MajorCus Chinese	1	-123.414327	94.071024	-1.31	0.1896				
MajorCus Continental	1	-123.537602	99.525869	-1.24	0.2146				
MajorCus Desserts	1	8.372418	130.242222	0.06	0.9487				
MajorCus Fast Food	1	-0.220602	100.198128	-0.00	0.9982				
MajorCus Ice Cream	1	-207.563848	160.890573	-1.29	0.1971				
MajorCus Italian	1	-143.034921	103.284192	-1.38	0.1661				
MajorCus Mithai	1	38.769054	139.842556	0.26	0.7926				
MajorCus Mughlai	1	15.325153	135.408035	0.11	0.9099				
MajorCus North Indian	1	-112.959012	87.485357	-1.29	0.1967				
MajorCus Others	1	-11.919086	90.507624	-0.13	0.8952				
MajorCus Pizza	1	-424.058172	180.945891	-2.34	0.0191				
MajorCus South Indian	1	-178.080639	115.411435	-1.54	0.1229				
MajorCus Street Food	0	0		-					
MajorCus*Price_range_3 American	1	-409.984851	139.883556	-2.93	0.0034				
MajorCus*Price_range_3 Bakery	1	-597.949402	141.131915	-4.24	<.0001				
MajorCus*Price_range_3 Cafe	1	-452.941516	122.998395	-3.68	0.0002				
MajorCus*Price_range_3 Chinese	1	-310.630857	122.828685	-2.53	0.0115				
MajorCus*Price_range_3 Continental	1	-145.534779	128.758772	-1.15	0.2510				
MajorCus*Price_range_3 Desserts	1	-593.121265	181.211040	-3.27	0.0011				
MajorCus*Price_range_3 Fast Food	1	-155.290889	154.998377	-1.00	0.3164				
MajorCus*Price_range_3 Italian	1	-277.568200	127.123975	-2.18	0.0290				
MajorCus*Price_range_3 Mughlai	1	-469.840636	138.487028	-3.39	0.0007				
MajorCus*Price_range_3 North Indian	1	-296.326852	119.779325	-2.47	0.0134				
MajorCus*Price_range_3 Others	1	-361.131460	121.458542	-2.97	0.0030				
MajorCus*Price_range_3 Pizza	1	-147.683988	152.259133	-0.97	0.3321				
MajorCus*Price_range_3 South Indian	0	0							
MajorCus*Price_range_4 American	1	104.769057	167.743588	0.62	0.5323				
MajorCus*Price_range_4 Cafe	1	-278.286232	158.193488	-1.76	0.0786				
MajorCus*Price_range_4 Chinese	1	-71.224748	140.894873	-0.51	0.6132				
MajorCus*Price_range_4 Continental	1	179.706350	140.434608	1.28	0.2007				
MajorCus*Price_range_4 Desserts	1	-354.557168	250.267125	-1.42	0.1566				
MajorCus*Price_range_4 Fast Food	1	18.082585	248.938203	0.07	0.9421				
MajorCus*Price_range_4 Italian	1	-330.845932	146.564131	-2.26	0.0240				
MajorCus*Price_range_4 Mughlai	1	-220.730809	166.343173	-1.33	0.1846				
MajorCus*Price_range_4 North Indian	1	-19.507774	132.897806	-0.15	0.8833				
MajorCus*Price_range_4 Others	1	-135.389527	134.477246	-1.01	0.3141				
MajorCus*Price_range_4 Pizza	0	0	20						
Has Online delivery No	1	-120.577516	49.877767	-2.42	0.0157				

Root MSE	210.43314
Dependent Mean	122.69834
R-Square	0.2843
Adj R-Sq	0.2715
AIC	70028
AICC	70032
BIC	64054
C(p)	134.40761
SBC	64759
ASE (Train)	43497
ASE (Test)	52193

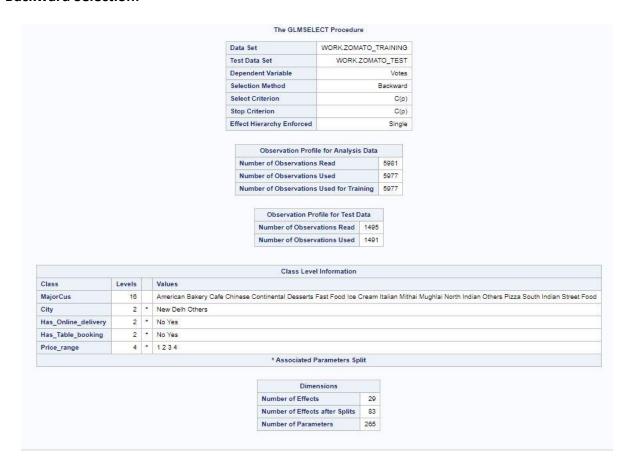
Has_Online_delivery_No*Price_range_3	1	98.426136	23.107017	4.26	<.0001
Has_Online_delivery_No*Price_range_4	1	-281.894279	47.919472	-5.88	<.0001
MajorCus*Has_Online_delivery_No American	1	116.220545	70,171624	1.66	0.0977
MajorCus*Has_Online_delivery_No Bakery	1	82.862289	55.251756	1.50	0.1337
MajorCus*Has_Online_delivery_No Cafe	1	48.239802	55.234630	0.87	0.3825
MajorCus*Has_Online_delivery_No Chinese	1	73.272378	53.902550	1.36	0.1741
MajorCus*Has_Online_delivery_No Continental	1	-50.697334	63.724719	-0.80	0.4263
MajorCus*Has_Online_delivery_No Desserts	1	55.823947	71.168112	0.78	0.4328
MajorCus*Has_Online_delivery_No Fast Food	1	111.254470	54.201717	2.05	0.0402
MajorCus*Has_Online_delivery_No Ice Cream	1	80.006922	64.207867	1.25	0.2128
MajorCus*Has_Online_delivery_No Italian	1	209.635250	65.767575	3.19	0.0014
MajorCus*Has_Online_delivery_No Mithai	1	-11.111247	105.540863	-0.11	0.9162
MajorCus*Has_Online_delivery_No Mughlai	1	79.500593	66.048807	1.20	0.2288
MajorCus*Has_Online_delivery_No North Indian	1	92.088094	51.099334	1.80	0.0716
MajorCus*Has_Online_delivery_No Others	1	72.698899	53.118801	1.37	0.1712
MajorCus*Has_Online_delivery_No Pizza	1	9.245537	63.788507	0.14	0.8848
MajorCus*Has_Online_delivery_No South Indian	1	5.404988	63.576868	0.09	0.9323
MajorCus*Has_Online_delivery_No Street Food	0	0			
Has_Table_booking_Yes	1	259.671543	110.173821	2.36	0.0188
Has_Table_booking_Yes*Price_range_3	1	-111.501094	24.542621	-4.54	<.000
Has_Table_booking_Yes*Price_range_4	1	-164.590254	35.800736	-4.60	<.000
MajorCus*Has_Table_booking_Yes American	1	117.995357	135.603949	0.87	0.3843
MajorCus*Has_Table_booking_Yes Cafe	1	-90.841520	115.969650	-0.78	0.4338
MajorCus*Has_Table_booking_Yes Chinese	1	-130.978115	112,198881	-1.17	0.243
MajorCus*Has_Table_booking_Yes Continental	1	-191.872678	114.782412	-1.67	0.0947
MajorCus*Has_Table_booking_Yes Fast Food	1	-231.256082	168.207277	-1.37	0.1692
MajorCus*Has_Table_booking_Yes Italian	1	93.225705	117.199925	0.80	0.4264
MajorCus*Has_Table_booking_Yes Mughlai	1	-88.412261	133.722112	-0.66	0.5088
MajorCus*Has_Table_booking_Yes North Indian	1	-117.805400	109.230570	-1.08	0.2809
MajorCus*Has_Table_booking_Yes Others	1	0.923707	111.500381	0.01	0.9934
MajorCus*Has_Table_booking_Yes Pizza	1	-187.872143	148.761005	-1.26	0.2067
MajorCus*Has_Table_booking_Yes South Indian	0	0	*		
Has_Online_delivery_No*Has_Table_booking_Yes	1	-173.150359	23.563666	-7.35	<.000
ncuisines	1	90.073886	71.386071	1.26	0.207
ncuisines*MajorCus American	1	-237.014465	143.722607	-1.65	0.0992
ncuisines*MajorCus Bakery	1	-86.470197	84.578137	-1.02	0.3066
ncuisines*MajorCus Cafe	1	10.883700	79.743676	0.14	0.8914
ncuisines*MajorCus Chinese	1	5.994808	80.322571	0.07	0.9408
ncuisines*MajorCus Continental	1	-31.844404	82.164091	-0.39	0.6983
ncuisines*MajorCus Desserts	1	-94.534697	114.716119	-0.82	0.4099
ncuisines*MajorCus Fast Food	1	-125.782629	87.777264	-1.43	0.1519
ncuisines*MajorCus Ice Cream	1	87.067353	156.466183	0.56	0.5779
ncuisines*MajorCus Italian	1	12.126001	87.796611	0.14	0.8902
ncuisines*MajorCus Mithai	1	-75.156883	84.941947	-0.88	0.3763
ncuisines*MajorCus Mughlai	1	-132.334413	128.983326	-1.03	0.3049
ncuisines*MajorCus North Indian	1	-20.744881	72.690245	-0.29	0.775
ncuisines*MajorCus Others	1	-119.066103	75.972107	-1.57	0.117
ncuisines*MajorCus Pizza	1	205 087528	178 116661	1.66	0.0986

ncuisines*MajorCus Pizza	1	295,987536	178.116661	1.66	0.0966
ncuisines*MajorCus South Indian	1	190.131386	101.359847	1.88	0.0607
ncuisines*MajorCus Street Food	0	0	19	*	
ncuisines*Has_Table_booking_Yes	1	15.886124	7.545983	2.10	0.0355
ncuisines_log	1	-167.819061	144.864725	-1.16	0.2467
ncuisines_I*MajorCus American	1	595.821520	367.852579	1.62	0.1053
ncuisines_I*MajorCus Bakery	.1	223.812126	169.745315	1.32	0.1874
ncuisines_I*MajorCus Cafe	.1	64.314986	163.992189	0.39	0.6949
ncuisines_I*MajorCus Chinese	- 1	-2.888581	162.561672	-0.02	0.9858
ncuisines_I*MajorCus Continental	1	186.247197	187.449901	0.99	0.3205
ncuisines_I*MajorCus Desserts	1	193.340529	225.807407	0.86	0.3919
ncuisines_I*MajorCus Fast Food	1	230.266982	172.417802	1.34	0.1818
ncuisines_I*MajorCus Ice Cream	1	-95.588233	279.322147	-0.34	0.7322
ncuisines_I*MajorCus Italian	1	-31.152932	196.665145	-0.16	0.8741
ncuisines_I*MajorCus Mithai	1	150.894698	180.904213	0.83	0.4043
ncuisines_l*MajorCus Mughlai	1	285.628694	244.005949	1.17	0.2418
ncuisines_I*MajorCus North Indian	1	28.648057	148.321063	0.19	0.8468
ncuisines_I*MajorCus Others	1	288.306559	155.547866	1.85	0.0639
ncuisines_I*MajorCus Pizza	1	-275.567786	308.043414	-0.89	0.3711
ncuisines_I*MajorCus South Indian	1	-376.706208	201.378628	-1.87	0.0814
ncuisines_I*MajorCus Street Food	0	0	85		

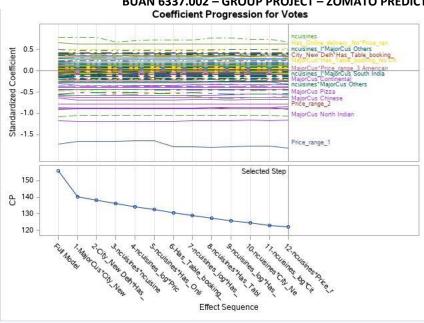
Figure 10: Forward Selection

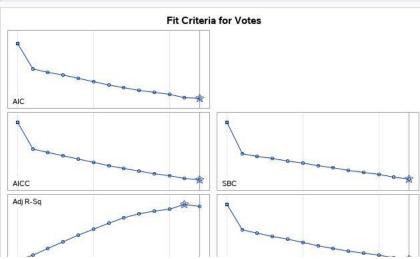
- We observe that the R-square improves to 0.28 in the case of Forward Selection which is better than both linear regression and LASSO.
- The AIC is 70028 which is lower than that of LASSO, so a better fit than LASSO can be implied.
- In Forward selection also shows when the level of price increase, the possibility of getting votes increase.
- If online delivery is available, then voting is bound to increase but people tend to not give votes for price range 3 and 4 if they provide online delivery. This is different from using LASSO.
- The R-square is still very low(0.28), we will use backward and best subset selection in order to better explain the variance in the model.

Backward Selection:



	Backward Se		-			
Step	Effect Removed	Number Effects In	Number Parme In	СР	ASE	Test A S
0		83	156	156.0000	42918.6727	52098.724
	nculsines_log*Has_Table_booking_Yes	82	156	156,0000	42918.6727	52096.724
	nculsines log*Has Online delivery Yes	81	156	156.0000	42918.6727	52098.724
	nculsines log*City_Others	80	156	156.0000	42918.6727	52096.724
	nculsines log*Price range 4	79	156	158.0000	42918.6727	52096.724
	nculsines*Has Table booking Yes	78	158	156,0000	42918.6727	52096.724
	nculsines*Has Online delivery Yes	77	156	156.0000	42918.6727	52098.724
	nculsines*City Others	76	156	156,0000	42918,6727	52096.724
	nculsines*Price range 4	75	156	156,0000	42918.6727	52096.724
	Has Online delivery Yes*Has Table booking Yes	74	156	156,0000	42918.6727	52098.724
	Has Online delivery Yes*Has Table booking No	73	156	156,0000	42918.6727	52098 724
	Has Online delivery No*Has Table booking Yes	72	156	156.0000	42918.6727	52096.724
	City Others*Has Table booking Yes	71	156	158,0000	42918.6727	52096,724
	City Others*Has Table booking No	70	156	158,0000	42918.6727	52096.724
	The state of the s		100	100.000		DE CONTRACTOR EST
	City_New Delh*Has_Table_booking_Yes	69	156 156	156,0000	42918.6727 42918.6727	52098.724
	MajorCus*Has_Table_booking_Yes	68				
	Has_Table_booking_Yes*Price_range_4	67	156	156.0000	42918.6727	52096.724
	Haa_Table_booking_Yes*Price_range_3	66	156	156.0000	42918.6727	52096.724
	Has_Table_booking_Yes*Price_range_2	65	156	156.0000	42918.6727	52098.724
	Has_Table_booking_Yes*Price_range_1	64	156	156.0000	42918.6727	52098,724
	Haa_Table_booking_No*Price_range_4	63	156	156.0000	42918.6727	52098.724
	Has_Table_booking_Yes	62	156	156,0000	42918.6727	52098.724
	City_Others*Has_Online_delivery_Yes	61	156	156.0000	42918.6727	52098.724
	City_Others*Has_Online_delivery_No	60	158	156.0000	42918.6727	52096.724
	City_New Delh*Has_Online_delivery_Yes	59	156	156.0000	42918.6727	52096.724
	MajorCue*Has_Online_delivery_Yes	58	158	156.0000	42918.6727	52096.724
	Haa_Online_delivery_Yes*Price_range_4	57	156	156.0000	42918.6727	52098.724
	Has_Online_delivery_Yes*Price_range_3	56	156	156.0000	42918.6727	52098.724
	Has_Online_delivery_Yes*Price_range_2	55	156	156.0000	42918.6727	52098.724
	Has_Online_delivery_Yes*Price_range_1	54	158	156,0000	42918.6727	52096.724
	Has Online delivery No*Price range 4	53	156	156.0000	42918.6727	52098.724
	Has Online delivery Yes	52	156	156.0000	42918.6727	52096.724
	MajorCus*City Others	51	156	156,0000	42918.6727	52098.724
	City Others*Price range 4	50	158	156,0000	42918.6727	52098.724
	City Others*Price_range_3	49	156	156.0000	42918.6727	52098.724
	City Others*Price range 2	48	156	156.0000	42918,6727	52098.724
	City_Others*Price_range_1	47	156	158.0000	42918.6727	52098.724
	City New Delh*Price range 4	46	156	156,0000	42918.6727	52096,724
	City Others	45	156	156.0000	42918.6727	52098.724
	MajorCus*Price_range_4	44	156	158.0000	42918.6727	52098 724
	Price range 4	43	156	156.0000	42918.8727	52096 724
1	MajorCus*City New Delh	42	121	140 2432	43023.6888	52008.853
2	City New Delh*Has Online delivery No	42	140	138.2524	43023.0600	52011.606
3	nculsines*nculsines	41	139	138.2524	43023,7570	52017.419
4	nculaines log*Price range 3	39	139	134,2964	43023.9106	52017.419
5		39	138	134.2964	43024.0815	52038.236
6	nculaines*Has_Online_delivery_No	38	137	132.3767	43024.6735	52038.238
6	Has_Table_booking_No*Price_range_1	100	136		300000000000000000000000000000000000000	000000000000000000000000000000000000000
	nculainea_log*Has_Online_delivery_No	36	100	128.7845	43027.6798	52000.260
8	nculaines*Has_Table_booking_No	35	134	127.1451	43030.3390	52020.031
9	nculsines_log*Has_Table_booking_No	34	133	125.7667	43034.9219	52031.007
10	nculsines*City_New Delh	33	132	124.4513	43039.9692	52004.685
11	nculsines_log^City_New Delh	32	131	122.7695	43042.3154	52027.296
12	nculaines*Price_range_3	31	130	122.11911	43052,2658	51916.220





Root MSE	209.78435
Dependent Mean	122.69634
R-Square	0.2916
Adj R-Sq	0.2760
AIC	70015
AICC	70021
BIC	64044
C(p)	122,11905
SBC	64906
ASE (Train)	43052
A SE (Teat)	51916

			Standard		
Parameter	DF	Estimate	Error	t Value	Pr >
Intercept	1	1155.239388	237.312267	4.87	<.000
Price_range_1	1	900.383744	222.085324	-4.05	<.000
Price_range_2	1	-465.403226	203.853596	-2.28	0.022
Price_range_3	1	-221.406420	164.647288	-1.34	0.178
MajorCue American	1	-171.231965	257.697019	-0.66	0.506
MajorCue Bakery	1	-841.330120	189,685305	-4.44	<.000
MajorCue Cafe	1	-864.901884	259.509862	-3.33	0.000
MajorCus Chinese	1	-885.697057	242.800579	-2.74	0.006
MajorCus Continental	1	-472.213723	241.893346	-1.95	0.051
MajorCus Desserts	1	-785.294113	329.245317	-2.32	0.020
MajorCus Fast Food	1	-517.402588	345,638715	-1.50	0.134
MajorCus Ice Cream	1	-389.650611	268.331030	-1.38	0.168
MajorCus Italian	- 1	-721.427435	246.278440	-2.93	0.003
MajorCus Mithal	1	-103.218332	197.617574	-0.52	0.601
MajorCus Mughial	1	-632.623096	278.422668	-2.27	0.023
MajorCus North Indian	1	-600.489661	235.763406	-2.55	0.010
MajorCus Others	1	-516.490483	237.537067	-2.17	0.029
MajorCue Pizza	1	972.917783	283.431049	-3.43	0.000
MajorCus South Indian	1	-350.864920	140,795285	-2.49	0.012
MajorCus Street Food	0	0		- 37	-
MajorCus*Price_range_1 American	1	422.705036	353.430979	1.20	0.231
MajorCue*Price range 1 Bakery	1	786,078384	162.989483	4.82	<.000
MajorCue*Price range 1 Cafe	1	719.294592	235.535997	3.05	0.002
MajorCus*Price range 1 Chinese	1	470.134114	222.065368	2.12	0.034
MajorCus*Price_range_1 Continental	1	210.074876	229,117145	0.92	0.359
MajorCue*Price range 1 Desserts	1	789.948105	304.690132	2.59	0.009
MajorCus*Price range 1 Fast Food	1	341 429551	302 234465	1.13	0.258
MajorCus*Price range 1 ice Cream	1	161 486085	227.496814	0.71	0.477
MajorCus*Price range 1 Italian	1	808.777116	234.784600	3.44	0.000
MajorCue*Price_range_1 Mithal	1	154.709134	117.084218	1.32	0.186
MajorCus*Price_range_1 Mughial	1	584.435939	240.133172	2.35	0.018
MajorCus*Price_range_1 Mugniai MajorCus*Price_range_1 North Indian	1	409:874159	216.704482	1.89	0.058
MajorCus*Price_range_1 Others	1	516.205336	218.421303	2.36	0.008
MajorCue*Price_range_1 Others MajorCue*Price_range_1 Pizza	1	303.434383	176.682752	1.72	0.088
Majorcus*Price_range_1 Pizza Majorcus*Price_range_1 South Indian	1	165.672169	86.806928	1.72	0.066
MajorCus*Price_range_1 south indian MajorCus*Price_range_1 Street Food	0	165.672169	190.000928	1.91	0.056
	1	81.934843	226.959940	0.00	0.718
MajorCus*Price_range_2 American	1	81,934843 594,371485	226.959940 144.302345	0.36	<.000
MajorCue*Price_range_2 Bakery	1	594,371485	144.302345	2.45	<.000
MajorCue*Price_range_2 Cafe	1 1/2				
MajorCus*Price_range_2 Chinese	1	266.260394	207.742512	1.28	0.200
MajorCus*Price_range_2 Continental	1	47.492669	207.975292	0.23	0.819
MajorCue*Price_range_2 Desserts	1	704.338472	299.424483	2.35	0.018
MajorCus*Price_range_2 Fast Food	1	149.045756	292.904051	0.51	0.610
MajorCus*Price_range_2 Ice Cream	0	0		- 3	
MajorCus*Price_range_2 Italian	1	498.989612	213,085890	2.34	0.019
MajorCue*Price_range_2 Mithal	0	0		-	
MajorCue*Price_range_2 Mughlal	1	411.461645	225.929675	1.82	0.068
MajorCus*Price_range_2 North Indian	1	212.182163	201.978859	1.05	0.293

Programme and the second secon			the second secon		
MajorCus*Price_range_2 Others	1	377.790037	204.105252	1.85	0.0642
MajorCue*Price_range_2 Pizza	1	141.522181	153.269943	0.92	0.3559
MajorCus*Price_range_2 South Indian	0	0			
MajorCus*Price_range_2 Street Food	0	0			
MajorCus*Price_range_3 American	1	-304,214866	183,523510	-1.66	0.0974
MajorCus*Price_range_3 Bakery	0	0	-		
MajorCus*Price range 3 Cafe	1	89.354108	187.388864	0.48	0.6335
MajorCus*Price_range_3 Chinese	-1	-24.115218	169.469303	-0.14	0.8868
MajorCus*Price range 3 Continental	1	-114.660782	166.085331	-0.69	0.4900
MajorCus*Price_range_3 Desserts	1	20.488322	299.420819	0.07	0.9454
MajorCus*Price_range_3 Fast Food	1	-28.927303	283.649930	-0.10	0.9188
MajorCus*Price range 3 Italian	1	254.293449	170.530072	1.49	0.1360
MajorCus*Price_range_3 Mughial	1	-60.184050	190.523848	-0.32	0.7521
MajorCue*Price_range_3 North Indian	1	-67.911949	162.314724	-0.42	0.6757
MajorCus*Price range 3 Others	1	-4.598252	163.562931	-0.03	0.9776
MajorCus*Price_range_3 Pizza	0	0	15		-
MajorCus*Price range 3 South Indian	0	0		- 61	10
City New Delh	1	-6.539940	29.394178	-0.22	0.8239
City New Delh*Price range 1	1	-105.985474	33.804319	3.14	0.0017
City New Delh*Price range 2	1	-111,455320	33,100786	-3.37	0.0008
City New Delh*Price range 3	1	-66.635238	32.704021	-2.04	0.0416
Has_Online_delivery_No	1	-562.221935	67.441517	8.34	<.0001
Has Online delivery No*Price range 1	1	263,080549	49.956950	5.27	<.0001
Has Online delivery No*Price range 2	1	263,485243	48.343671	5.45	<.0001
Has_Online_delivery_No*Price_range_3	1	371.771055	45.985639	8.08	<.0001
MajorCue*Has_Online_delivery_No American	1	105,119801	71.589068	1 47	0.1419
MajorCus*Has Online delivery No Bakery	1	76.172233	55.254000	1.38	0.1681
MajorCus*Has Online delivery No Cafe	1	41.975471	56.003831	0.75	0.4538
MajorCus*Has Online delivery No Chinese	1	67.179058	54 445555	1.23	0.2173
MajorCue*Has_Online_delivery_No Continental	1	-58.652778	84.358809	-0.91	0.3822
Majorcus*Has Online delivery No Desserts	1	42.308948	71.196375	0.59	0.5624
MajorCus*Has Online delivery No Fast Food	1	104,110861	54.491227	1.91	0.0561
MajorCus*Has Online delivery No ice Cream	1	74.982791	64.317403	1.17	0.2437
MajorCus*Has_Online_delivery_No Italian	1	191.557917	67.761547	2.83	0.0047
MajorCus*Has_Online_delivery_No Mithal	1	-22.236979	105.762149	-0.21	0.8335
MajorCus*Has Online delivery No Mughial	1	76.131257	67.435324	1.13	0.2590
MajorCus*Has Online delivery No North Indian	1	83.221045	51,908671	1.60	0.1089
MajorCus*Has Online delivery No Others	1	84 357059	54 107784	1.56	0.1190
MajorCus*Has_Online_delivery_No Pizza	1	6.578733	85.578964	0.10	0.9201
MajorCus*Has Online delivery No South Indian	1	7.661612	65.090331	0.10	0.9063
MajorCus*Has_Online_delivery_No Street Food	0	0	00.030031	36.12	0.3000
Haa_Table_booking_No	1	-160 983592	115.566659	-1.39	0.1837
Has_Table_booking_No*Price_range_2	1	-200.913907	37.744996	-5.32	<.0001
Has Table booking No*Price range 3	1	-74.697271	35.177300	-2.12	0.0338
MajorCus*Has_Table_booking_No American	1	-133.138288	136.453236	-0.98	0.3293
MajorCus*Has Table booking No Bakery	0	-133.130200	130.4002.30	10.00	0,3253
MajorCus*Has_Table_booking_No Cafe	1	89.691362	116.046886	0.60	0.5482
MajorCus*Has Table booking No Chinese	1	106.367308	112.542644	0.95	0.3446
MajorCus*Has Table booking No Continental	1	156.847358	114.914220	1.36	0.1723
MajorCus*Has_Table_booking_No Desserts	0	100.047.000	(14.9)4220	1.30	0.1723
	1	187,229525	168 356524		0.2661
MajorCus*Has_Table_booking_No Fast Food	0	187.229020	166.356524	1.11	0.2661
MajorCue*Has_Table_booking_No ice Cream	1	200000000000000000000000000000000000000	147 0047040	-	40400
MajorCue*Has_Table_booking_No Italian	1	-109.737010	117.361716	-0.94	0.3498
MajorCus*Has_Table_booking_No Mithal MajorCus*Has Table booking No Mughial	0	73.431450	133.776917	0.55	0.5831
	100		133.776917	0.55	Company (
	1	1,00,100,1100	400 24045	0.00	
MajorCus*Has_Table_booking_No North Indian	1	96.108975	109.442175	0.88	0.3799
MajorCus*Has_Table_booking_No North Indian MajorCus*Has_Table_booking_No Others	1	96.108975 -15.974299	111.698675	-0.14	0.8863
MajorCue*Hae_Table_booking_No North Indian MajorCue*Hae_Table_booking_No Others MajorCue*Hae_Table_booking_No Pizza	1 1	96.108975 -15.974299 226.761323			
MajorCus*Has_Table_booking_No North Indian MajorCus*Has_Table_booking_No Others MajorCus*Has_Table_booking_No Pizza MajorCus*Has_Table_booking_No South Indian	1 1 1 0	96.108975 -15.974299 226.761323 0	111.698675	-0.14	0.8863
MajorCus*Has_Table_booking_No North Indian MajorCus*Has_Table_booking_No Cithers MajorCus*Has_Table_booking_No Pizza MajorCus*Has_Table_booking_No South Indian MajorCus*Has_Table_booking_No Street Food	1 1 1 0	96.106975 -15.974299 226.761323 0 0	111.698675 148.799840	-0.14 1.52	0.8863
MajorCus*Has_Table_booking_No North Indian MajorCus*Has_Table_booking_No Others MajorCus*Has_Table_booking_No Pizza MajorCus*Has_Table_booking_No South Indian	1 1 1 0	96.108975 -15.974299 226.761323 0	111.698675	-0.14	0.8863

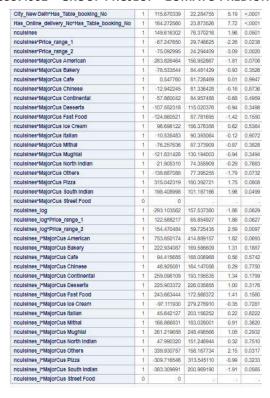
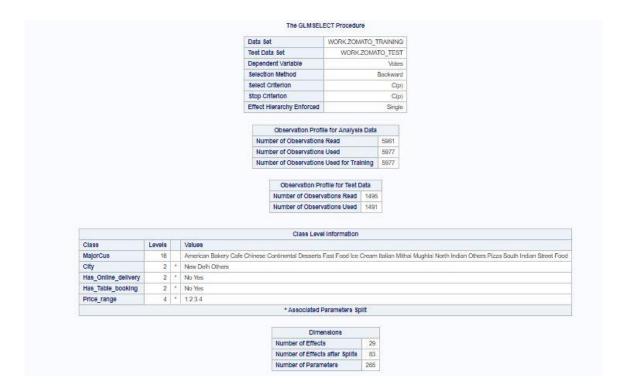


Figure 11: Backward Selection

- In case of backward selection, we observe that the R square is 0.27 which is lower than that of forward selection.
- The AIC of backward selection is better than forward and LASSO with a value of 70015 which implies a better fit.
- In Backward selection, price range 4 restaurant have higher possibility of getting votes.
- If online delivery is available, then voting is bound to increase.
- The R-square is lower than forward selection (0.28), we will use stepwise selection and best subset selection in order to better explain the variance in the model.

Stepwise Selection:

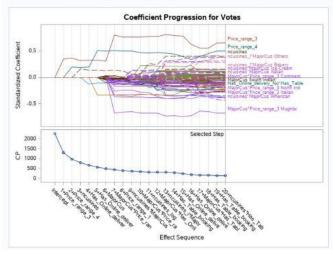


BUAN 6337.002 - GROUP PROJECT - ZOMATO PREDICTIVE ANALYSIS

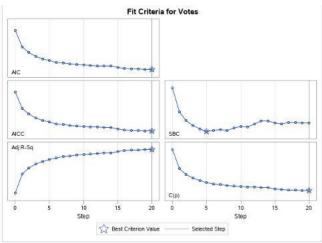
Step	Effect Entered	Removed Removed	Number Effects in	Number Parms in	СР	ASE	Teet A SE
0	Intercept		1	- 1	2268.1663	60777.4876	63888.5477
1	Price_range_3		2	2	1290.0890	53551.1566	56036.5642
2	Price_range_4		3	3	972,7389	51196.7111	53753.5413
3	nculsines		4	4	784.2051	49791.8915	52076.0479
4	Has_Online_delivery_No		5	. 5	654.3176	48819.4746	51169.049
5	Has_Online_delivery_No*Price_range_4		6	6	557.4723	48090.6813	51032.441
6	MajorCue		7	21	477.1219	47277.0589	50700.359
7	MajorCus*Price_range_3		8	33	434.9069	46788.8507	51109.137
8	Price_range_1		9	34	395.3841	48482.5524	50864.291
9	nculsines*MajorCus		10	49	385.2565	46039.3749	50944.407
10	MajorCus*Price_range_4		11	59	338.8403	45697.1449	51695.072
11	nculaines_log		12	60	319.8738	45542.5574	51690.003
12	MajorCus*Has_Online_delivery_No		13	75	304.0348	45204.5829	51696.573
13	nculsines_I*MajorCus		14	90	296.0837	44924.7662	51464.077
14	Has_Table_booking_Yes		15	91	290.3098	44867.4485	51872.354
15	Has_Online_delivery_No*Has_Table_booking_Yes		16	92	222.7983	44354.9358	51979.671
16	Has_Online_delivery_No*Price_range_3		17	93	181.9226	44038.8097	52770.491
17	MajorCus*Has_Table_booking_Yes		18	103	161.0346	43737.3389	52803.662
18	Has_Table_booking_Yes*Price_range_4		19	104	154.3013	43872.9481	52671.430
19	Has_Table_booking_Yes*Price_range_3		20	105	136.8499	43529.5313	52285.395
20	nculaines*Has_Table_booking_Yes		21	106	134.4076*	43496.7780	52193.211

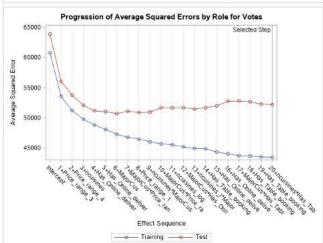
Selection stopped at a local minimum of the C(p) criterion.

Stop Details							
Candidate For	Effect	Candidate C(p)		Compare C(p)			
Entry	City_Others	134.9338	>	134.4076			
Removal	nouisines*htas_Table_booking_Yes	136.8499	>	134.4076			



Dependent Mean	122.69634
R-Square	0.2843
Adj R-Sq	0.2715
AIC	70028
AICC	70032
BIC	64054
C(p)	134,40761
SBC	84759
ASE (Train)	43497
ASE (Test)	52193





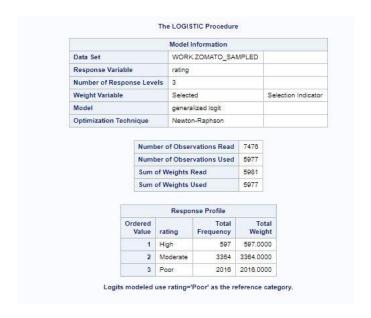
	1	natea	Standard		177
Parameter	DF	Estimate	Error	t Value	Pr>
Intercept	1	148.908561	86.389471	1.72	0.084
Price_range_1	- 1	-51.469530	7.455290	-6.90	<.000
Price_range_3	1	462.667554	120.787976	3.83	0.000
Price_range_4	1	575.155524	137,409975	4.19	<.000
MajorCue American	1	-34.941508	120.744719	-0.29	0.772
MajorCue Bakery	- 1	-60.971982	98.613747	-0.62	0.536
MajorCue Cafe	1	-94.942178	94.711492	-1.00	0.316
MajorCus Chinese	1	-123.414327	94.071024	-1.31	0.189
MajorCus Continental	1	-123.537602	99.525869	-1.24	0.214
MajorCue Desserts	- 1	8.372418	130.242222	0.08	0.948
MajorCus Fast Food	31	-0.220602	100.198128	-0.00	0.998
MajorCue Ice Cream	1	-207.563848	160.890573	-1.29	0.197
MajorCue Italian	1	-143.034921	103,284192	-1.38	0.166
MajorCus Mithal	- 1	36.769054	139.842556	0.26	0.792
MajorCue Mughial	- 31	15.325153	135.408035	0.11	0.909
MajorCus North Indian	1	-112.959012	87.485357	-1.29	0.196
MajorCus Others	1	-11.919086	90.507624	-0.13	0.896
MajorCue Pizza	1	-424.058172	180.945891	-2.34	0.019
MajorCue South Indian	- 1	-178.080639	115.411435	1.54	0.123
MajorCue Street Food	0	0			-
MajorCus*Price_range_3 American	1	-409.964851	139.883556	-2.93	0.000
MajorCue*Price range 3 Bakery	1	-597 949402	141.131915	-4.24	<.000
MajorCue*Price range 3 Cafe	- 1	-452.941516	122.998395	-3.68	0.000
MajorCus*Price range 3 Chinese	1	-310.630657	122.828685	-2.53	0.01
MajorCus*Price range 3 Continental	1	-145.534779	126.758772	-1.15	0.251
MajorCus*Price range 3 Desserts	1	-583.121285	181,211040	3.27	0.001
MajorCue*Price range 3 Fast Food	- 1	-155,290889	154.996377	-1.00	0.316
MajorCus*Price range 3 Italian	1	-277.588200	127.123975	-2.18	0.029
MajorCus*Price range 3 Mughial	1	469.840636	138.487028	-3.39	0.000
MajorCue*Price range 3 North Indian	1	-296.326852	119.779325	-2.47	0.013
MajorCus*Price range 3 Others	1	-361.131460	121,458542	-2.97	0.000
MajorCus*Price_range_3 Pizza	1	-147.683988	152.259133	0.97	0.333
MajorCus*Price range 3 South Indian	0	-147.000000	102,200100	0.27	0.556
MajorCue*Price range 4 American	1	104.769057	167.743588	0.62	0.530
MajorCus*Price_range_4 American	- 1	278.286232	158.193488	1.78	0.032
	1	-71.224748	140.894873	-0.51	0.613
MajorCus*Price_range_4 Chinese MajorCus*Price_range_4 Continental	1	179.706350	140.434608	1.28	0.200
MajorCus*Price_range_4 Continental	1	-354.557168	250.287125	1.42	0.156
MajorCus*Price_range_4 Desserts MajorCus*Price_range_4 Fast Food	1	18.082585	248.938203	0.07	0.150
10 To	-	-330.845932	146.584131	-2.28	0.024
MajorCus*Price_range_4 Italian	1	-330,845932	166.343173	-2.26	0.024
MajorCue*Price_range_4 Mughlal	1		100000000000000000000000000000000000000		
MajorCus*Price_range_4 North Indian	1	-19.507774	132.897806	-0.15	0.883
MajorCus*Price_range_4 Others	1	-135.389527	134.477246	-1.01	0.314
MajorCus*Price_range_4 Pizza	0	0	40.0000	3	
Has_Online_delivery_No	1	-120.577516	49.877767	-2.42	0.015
Has_Online_delivery_No*Price_range_3	. 1	98.426136	23.107017	4.26	<.000
Has_Online_delivery_No*Price_range_4	- 1	-281.894279	47.919472	-5.88	<.000
MajorCus*Has_Online_delivery_No American	1	116.220545	70.171624	1.66	0.097
MajorCus*Has_Online_delivery_No Bakery	1	82.862289	55,251756	1.50	0.133
MajorCus*Has Online delivery No Cafe	- 1	48.239802	55.234630	0.87	0.382

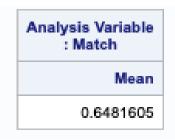
MajorCus*Has_Online_delivery_No Chinese	1	73.272378	53.902550	1.36	0.1741
MajorCue*Has Online delivery No Continental	1	-50.697334	63.724719	-0.80	0.4263
MajorCus*Has_Online_delivery_No Desserts	1	55.823947	71.168112	0.78	0.4328
	1	111.254470	54.201717	2.05	0.0402
MajorCus*Has_Online_delivery_No Fast Food					
MajorCus*Has_Online_delivery_No ice Cream	1	80.006922	64.207867	1.25	0.2128
MajorCus*Has_Online_delivery_No Italian	1	209.635250	65.767575	3, 19	0.0014
MajorCus*Has_Online_delivery_No Mithal	1	-11.111247	105.540663	-0.11	0.9162
MajorCus*Has_Online_delivery_No Mughial	1	79.500593	66.048807	1.20	0.2288
MajorCue*Has_Online_delivery_No North Indian	1	92.088094	51.099334	1.80	0.0716
MajorCue*Has_Online_delivery_No Others	1	72.698899	53.118801	1.37	0.1712
MajorCue*Hae_Online_delivery_No Pizza	1	9.245537	83.788507	0.14	0.8848
MajorCus*Has_Online_delivery_No South Indian	1	5.404988	63.576868	0.09	0.9323
MajorCus*Has Online delivery No Street Food	0	0			
Has Table booking Yes	- 1	259.671543	110.173821	2.36	0.0185
Has_Table_booking_Yes*Price_range_3	1	-111.501094	24.542621	-4.54	<.0001
Has_Table_booking_Yes*Price_range_4	1	-164 590254	35.800736	-4.60	< 0001
MajorCue*Hae Table booking Yee American	1	117.996357	135.603949	0.87	0.3843
MajorCue*Hae Table booking Yee Cafe	1	-90.841520	115.989650	-0.78	0.4335
	1	-130.978115	112.198881		0.4333
MajorCustiles Table booking Yes Chinese				-1.17	
MajorCus*Has_Table_booking_Yes Continental	1	-191.872678	114.782412	-1.67	0.0947
MajorCue*Has_Table_booking_Yee Fast Food	1	-231.258082	168.207277	-1.37	0.1692
MajorCue*Has_Table_booking_Yes Italian	1	93.225705	117.199925	0.80	0.4264
MajorCue*Hae_Table_booking_Yee Mughtal	1	-88.412261	133.722112	-0.66	0.5085
MajorCus*Has_Table_booking_Yee North Indian	1	-117.805400	109.230570	-1.08	0.2809
MajorCus*Has_Table_booking_Yes Others	1	0.923707	111.500381	0.01	0.9934
MajorCue*Hae_Table_booking_Yee Pizza	1	-187.872143	148.761005	-1.26	0.2067
MajorCue*Hae_Table_booking_Yee South Indian	0	0	(A)	54	
Has_Online_delivery_No*Has_Table_booking_Yes	1	-173.150359	23.563666	-7.35	<.0001
nculsines	1	90.073888	71.386071	1.26	0.2071
nculsines*MajorCus American	- 1	-237.014485	143.722607	-1.65	0.0992
nculsines MajorCus Bakery	1	-86.470197	84.578137	-1.02	0.3066
nculeinee*MajorCue Cafe	1	10.883700	79.743878	0.14	0.8914
nculsines*MajorCus Chinese	1	5.994808	80.322571	0.07	0.9405
nculsines Major Cus Continental	1	-31.844404	82.164091	-0.39	0.6983
nculaines Major Cus Desserts	1	-94.534697	114,716119	-0.82	0.4099
nculsines Major Cus Desserts nculsines Major Cus Fast Food	1	-125.782629	87,777264	-1.43	0.4099
11 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1					
nculsines*MajorCus ice Cream	1	87.067353	156.466183	0.56	0.5779
nculsines*MajorCus Italian	1	12.126001	87.798611	0.14	0.8902
nculsines MajorCus Mithal	1	-75.156883	84.941947	-0.88	0.3763
nculsines MajorCus Mughial	1	-132.334413	128.983326	-1.03	0.3049
nculsines*MajorCus North Indian	1	-20.744881	72.690245	-0.29	0.7754
nculsines*MajorCus Others	1	-119.086103	75.972107	-1.57	0.1171
nculsines*MajorCus Pizza	1	295.967536	178.118681	1,66	0.0966
nculaines*MajorCus South Indian	1	190.131386	101,359647	1.88	0.0607
nculsines*MajorCus Street Food	0	0			
nculsines*Has_Table_booking_Yes	1	15.886124	7.545983	2.10	0.0355
nculsines_log	1	-167.819061	144.884725	-1,16	0.2467
nculsines_PMajorCus American	1	595.821520	367.852579	1.62	0.1053
nculsines_PMajorCus Bakery	1	223.812126	169.745315	1.32	0.1874
nculsines_I*MajorCus Cafe	1	64.314988	163,992189	0.39	0.6949
nculsines_I*MajorCus Chinese	1	-2.888581	162.561672	-0.02	0.9858
nculsines PMalorCus Continental	1	186.247197	187.449901	0.99	0.3205
nculsines PMajorCus Desserts	1	193.340529	225.807407	0.86	0.3919
nculsines PMajorCus Fast Food	1	230.268982	172.417802	1.34	0.1818
nculsines PMajorCus ice Cream	1	-95.588233	279 322147	-0.34	0.7322
	1	-31.152932	196.665145	-0.34	0.7322
nculsines_PMajorCus Italian					
nculsines_i*MajorCus Mithal	1	150.894698	180.904213	0.83	0.4043
nculsines_i*MajorCus Mughial	1	285.628694	244.005949	1.17	0.2418
nculsines_PMajorCus North Indian	1	28.648057	148.321063	0.19	0.8468
nculsines_PMajorCus Others	1	288.308559	155.547888	1.85	0.0639
nculsines_PMajorCus Pizza	1	-275.567788	308.043414	-0.89	0.3711
nculsines_i^MajorCus South Indian	1.	-376.706208	201.378628	-1.87	0.0614
nculsines_PMajorCus Street Food	0	0		184	- 2

Figure 11: Stepwise Selection

- The stepwise selection has R square value of 0.27 which is higher than that of LASSO however lower than the Forward selection.
- In terms of fit, the AIC is 70028 which is equivalent to the forward selection model.
- The price range result showing the same pattern as before, when the price range is low especially in range 1, the possibility of getting votes is low.
- If online delivery is available, then voting is bound to increase.
- Forward selection has the highest R square for now.

Logit Model:





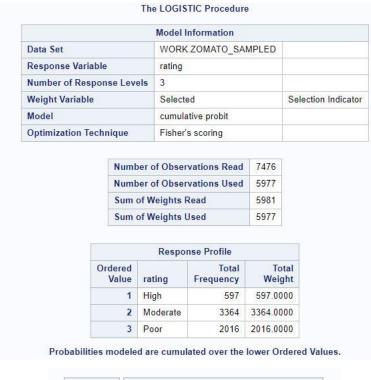
BUAN 6337.002 - GROUP PROJECT - ZOMATO PREDICTIVE ANALYSIS

Frequency Percent Row Pct Col Pct	Table of rating by I_rating					
		I_rating(Into: rating)				
	rating	High	Moderate	Poor	Total	
	High	35	116	1	152	
	_	2.34	7.76	0.07	10.17	
		23.03	76.32	0.66		
		68.63	12.06	0.21		
	Moderate	15	665	212	892	
		1.00	44.48	14.18	59.67	
		1.68	74.55	23.77		
		29.41	69.13	43.98		
	Poor	1	181	269	451	
		0.07	12.11	17.99	30.17	
		0.22	40.13	59.65		
		1.96	18.81	55.81		
	Total	51	962	482	1495	
		3.41	64.35	32.24	100.00	

Figure 12: Logit Model

- From the confusion Matrix in Figure 12 we observe that High is correctly predicted 23.03% times, Moderate is correctly predicted 74.55% times while poor is predicted correctly 59.5% times.
- The overall accuracy of the model is 64.81%.

Probit Model:





Frequency	Table of rating by I_rating				
Percent Row Pct		I_rating(Into: rating)			
Col Pct	rating	High	Moderate	Poor	Total
	High	35	117	0	152
		2.34	7.83	0.00	10.17
		23.03	76.97	0.00	
		58.33	10.66	0.00	
	Moderate	24	714	154	892
		1.61	47.76	10.30	59.67
		2.69	80.04	17.26	
		40.00	65.03	45.70	
	Poor	1	267	183	451
		0.07	17.86	12.24	30.17
		0.22	59.20	40.58	
		1.67	24.32	54.30	
	Total	60	1098	337	1495
		4.01	73.44	22.54	100.00

Figure 13:Probit Model

- From the confusion Matrix in Figure 13 we observe that High is correctly predicted 23.03% times, Moderate is correctly predicted 80.04% times while poor is predicted correctly 40.58.% times.
- The overall accuracy of the model is 62.34%.

Findings and Managerial Implications

Model	R-Sq	AIC
Linear Regression	0.23	-
LASSO	0.26	70080
Forward	0.28	70028
Backward	0.27	70015
Stepwise	0.27	70028

Table 2: Results for linear models for Votes

Model	%correct High	%correct Moderate	%correct Low	Overall Accuracy
Logit	23.03	74.55	59.65	64.8%
Probit	23.03	80.04	40.58	62.32%

Table 3: Results for classification of ratings

Based on the descriptive analysis, opening a Italian or continental restaurants would get more votes and higher rating compare to other cuisine. Italian and continental restaurants are fall at price range 4 restaurants.

Moreover, Forward selection is the best model among the five models. Based on Forward selection model, if the restaurants wants to get more voting. First, the higher priced restaurants, the possibility of getting votes is higher. Second, online delivery service will increase the possibility of getting more votes, but not work on the price rang 4 restaurants. If The luxury restaurants focus on getting more votes, they do not need to spend money on provide online delivery. Third, providing table booking increase the possibility of getting more votes but this also not work on restaurant in price range 3 and 4. This result shows if the restaurants are in lower price range, they should provide table booking in order to increase the possibility of getting more votes.

In Summary, based on the dataset, open a Italian or continental restaurants would have the highest possibility of getting higher rating and more votes and since they are high price range restaurants, the owners do not need to spend money on providing delivery service and table booking.

Conclusion

90 % of the people buying decisions are based on online review. Having many good ratings is a big opportunity to drive more traffic to your business (Vermaat, 2018). Only have good rating is not enough to attract more customers, you have to have huge quantity reviews to make people believe in restaurant consistency.

From Zomato dataset, we found out the price range affect the possibility for people to rate the restaurant. Restaurants owners should adjust their service based on their food price to get more rating. For the price range 1, people do not bother or feel obligate to rate the restaurants but for the restaurant people spend a lot of money on, they expect more than price range 1 restaurants so they feel more obligate to rate the restaurants. However, they can provide online delivery and table booking to increase the possibility of getting more reviews. Second, we found out in price range 4 restaurants, people do not care about if they have online delivery. Customer service and restaurant atmosphere are a huge factors for people paying more money, people do not order take out at luxury restaurants which lead to the result luxury restaurants do not need to hire people to do delivery.

There are a lot of factors affect restaurants rating and the motivation to rate. The dataset does not include many other factors. According to Wakefield (2004) who identify five important factors for restaurants which are customers interaction, technological investments, capacity, quality of physical facilities. Satisfied customer service increases the possibility of rating them online (Vermaat, 2018).in the future dataset, these factors can be added in the survey to come up with a better outcome.

References

- 1. Gareth James, Daniela Witten, Trevor Hastie and Robert Tibshirani(2009). An Introduction to Statistical Learning. Springer
- 2. Zomato Restaurants Data retrieved from https://www.kaggle.com/shrutimehta/zomato-restaurants-data
- 3. Budhwar, K. (2004). Understand the sucess factors for indepedent restaurants in Delhi.
- 4. Vermaat, D. (2018). 11 tips to encourage customers to leave reviews for your business online. Retrieved from loopyloyalty: https://blog.loopyloyalty.com
- 5. Wakefield, K. (2004). Research Reveals Five Most Important Success Factors For Pizza. Retrieved from Pizza marketing Quarterly.

Appendix

SAS Codes

```
LIBNAME PRO 'C:\Users\vdr180000\Desktop';
/* This imports the csv dataset into SAS. */
PROC IMPORT OUT= zomato in
      DATAFILE= "C:\Users\vdr180000\Desktop\zomato in.csv"
      DBMS=CSV REPLACE;
  GETNAMES=YES;
  DATAROW=2;
RUN:
/* generating the working dataset in Work library */
data zomato;
set zomato in;
run;
/* Defining log */
data zomato;
set zomato in;
ncuisines log=log(ncuisines);
run;
/*splitting*/
proc surveyselect data=zomato out=zomato sampled outall samprate=0.8 seed=2;
run;
data zomato training zomato test;
set zomato sampled;
if selected then output zomato training; /* Tell SAS that only keep the 80% selected one in
sample. The rest
will be in test data */
else output zomato test;
run;
/*linear regression*/
ods graphics on;
proc glm data =zomato sampled;
class MajorCus(ref='South Indian') City(ref='Others') Has Online delivery(ref='No')
Has Table booking(ref='No') Price range(ref='1');
model votes = price range MajorCus City Has Online delivery Has Table booking ncuisines
ncuisines_log price_range*Majorcus MajorCus*City /solution;
run;
```

```
ods graphics off;
/*ASE in train vs. test data */
/* Stepwise with LASSO regression with AIC as criteria */
proc glmselect data=zomato training testdata=zomato test seed = 2 plots=all;
class MajorCus City(split) Has Online delivery(split) Has Table booking(split)
Price range(split);
model Votes=
Price range | MajorCus | City | Has Online delivery | Has Table booking | ncuisines | ncuisines | log
@2
/selection=lasso(choose=cv stop=none) hierarchy=single cvmethod=split(5) showpvalues;
performance buildsscp=incremental;
run;
/*ASE in train vs. test data */
/* Selection with forard, backward and stepwise */
proc glmselect data=zomato training testdata=zomato test seed = 2 plots=all;
class MajorCus City(split) Has Online delivery(split) Has Table booking(split)
Price range(split);
model Votes=
Price range | MajorCus | City | Has Online delivery | Has Table booking | ncuisines | ncuisines log
@2
/selection=forward(select=cp) hierarchy=single showpvalues;
performance buildsscp=incremental;
run;
proc glmselect data=zomato training testdata=zomato test seed = 2 plots=all;
class MajorCus City(split) Has Online delivery(split) Has Table booking(split)
Price range(split);
model Votes=
Price range | MajorCus | City | Has Online delivery | Has Table booking | ncuisines | ncuisines | log
@2
/selection=backward(select=cp) hierarchy=single showpvalues;
performance buildsscp=incremental;
run;
proc glmselect data=zomato training testdata=zomato test seed = 2 plots=all;
class MajorCus City(split) Has Online delivery(split) Has Table booking(split)
Price range(split);
model Votes=
Price range | MajorCus | City | Has Online delivery | Has Table booking | ncuisines | ncuisines | log
@2
/selection=stepwise(select=cp) hierarchy=single showpvalues;
performance buildsscp=incremental;
```

```
run;
/*logistic regression*/
proc logistic data=zomato sampled;
class MajorCus(ref='South Indian') City(ref='Others') Has Online delivery(ref='No')
Has Table booking(ref='No') Price range(ref='1');
model rating(event='High') = Price range City MajorCus Has Online delivery
Has Table booking neuisines neuisines log Votes/link=glogit;
weight selected;
output out=preds predprobs=individual;/*only training sample is used for estimation, since
selected = 0 for test sample */
score data=zomato test out=zomatoPred;
run;
proc freq data=zomatoPred;
    table rating*I rating / out=CellCounts;
    run;
   data CellCounts;
    set CellCounts;
    Match=0:
    if rating=I rating then Match=1;
    run;
   proc means data=CellCounts mean;
    freq count;
    var Match;
    run;
/*probit model*/
proc logistic data=zomato sampled;
class MajorCus(ref='South Indian') City(ref='Others') Has Online delivery(ref='No')
Has Table booking(ref='No') Price range(ref='1');
model rating(event='High') = Price range City MajorCus Has Online delivery
Has Table booking ncuisines ncuisines_log Votes/link=probit;
weight selected;
output out=preds1 predprobs=individual;/*only training sample is used for estimation, since
selected = 0 for test sample */
score data=zomato test out=zomatoPred1;
run;
proc freq data=zomatoPred1;
    table rating*I rating / out=CellCounts1;
    run;
   data CellCounts1;
    set CellCounts1;
    Match=0;
    if rating=I rating then Match=1;
```

```
run;
proc means data=CellCounts1 mean;
freq count;
var Match;
run;
```