OPIM 5604 - Predictive Modeling

Project 1

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Summary:

Our approach involved several key steps. Initially, we examined the columns and their values to identify any meaningful patterns. We also transformed categorical variables into continuous ones and introduced new columns to enhance the dataset. Subsequently, we conducted a thorough assessment for missing values, removing columns with a high volume of missing data. To address these missing values, we employed an imputation method based on averages. We further identified and eliminated outliers from the dataset to enhance data quality. Finally, we performed Principal Component Analysis (PCA) on the refined dataset, allowing us to extract meaningful information while reducing data complexity and the number of columns significantly.

Columns and Explanations:

Id: Excluded as IDs are unique identifiers and are irrelevant to price.

listing_url: Excluded as each URL is unique and doesn't contribute as a relevant factor in affecting price.

scrape_id: Excluded as IDs are unique identifiers and are irrelevant to price.

last_scraped: This column provides information that is once again irrelevant to the price. Hence, it's excluded.

source: This column contains information that is similar to the last_scraped column. It's information that can't be used to predict price. Hence, this column is also excluded.

name: This column contains information that doesn't follow a certain pattern or standard. It's very vague information and can't be used to evaluate price. Hence, it's excluded.

description: A very subjective information that is almost unique as it doesn't follow any established pattern or standard. This information column cannot be quantified and had to be excluded.

neighborhood_overview: Very similar scenario to the description column. Not quantifiable and excluded.

picture_url: Excluded as each URL is unique and doesn't contribute as a relevant factor in affecting price.

host id: Excluded as IDs are unique identifiers and are irrelevant to price.

host_url: Excluded as each URL is unique and doesn't contribute as a relevant factor in affecting price. Excluded as each URL is unique and doesn't contribute as a relevant factor in affecting price. There are URLs repeating multiple times when a host has multiple listings, but it still remains irrelevant to price.

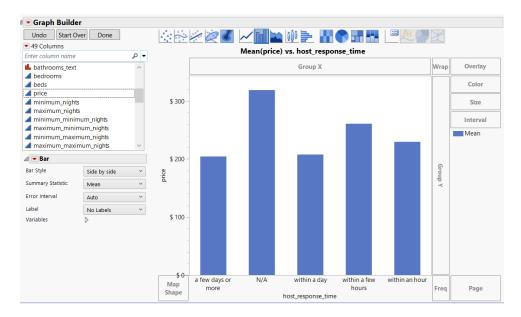
host_name: Excluded as a name is a unique identifier and isn't relevant to predicting price. **host_since:** This column provides information that is irrelevant to the price. Hence, it's excluded.

Host_location: Excluded the column as the location of the host does not relate with the price variable. Also it has 2344 missing values.

Host_about: Excluded as it's not related to price. Host_about has 4810 missing values which can not be imputed as the data is categorical.

Summary Statisti	rs						
Julilliary Statisti	C3						
13 Columns Clear Select	Distribution						
Columns	N	N Missing	N Categories	Min	Max	Mean	Std Dev
host_location	9964	2344	302				
host_about	7430	4878	2985				
host_response_time	12308	0	5				
host_response_rate	12308	0	59				
host_acceptance_rate	12308	0	95				
host_is_superhost	11946	362	2				
host_thumbnail_url	12308	0	5772				
host_picture_url	12308	0	5772				
host_neighbourhood	7498	4810	57				
host_listings_count	12308	0		1	2554	26.818898278	87.359141901
host_total_listings_count	12308	0		1	5305	38.333929152	141.51208454
host_verifications	12308	0	6				
host_has_profile_pic	12308	0	2				

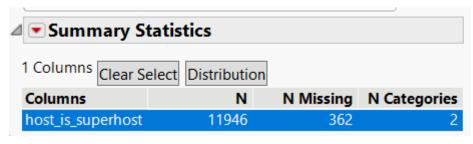
Host_response_time: Kept as depending on how quick a host responds on the inquiries or bookings from a potential guest.

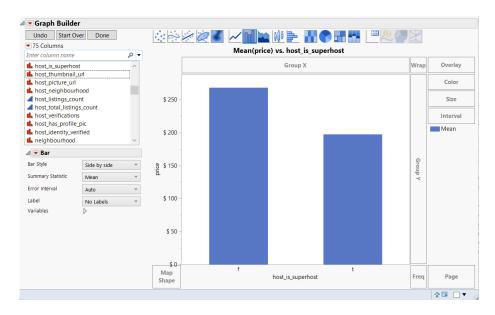


Host_response_rate: Keeping this column as higher response rates might command different price levels based on different hosts.

Host acceptance rate: Keeping this column as higher acceptance rates

Host_is_superhost: This column is directly related to the price. We could add an indicator column to convert this categorical value to a continuous in order to analyze. This column has only 362 missing values which is about 3%. So we can exclude and hide those rows. Imputing the data on superhost is not ideal as it can directly affect the prices variable.





Is_superhost: We have created this new indicator column which represents true for the host is superhost.

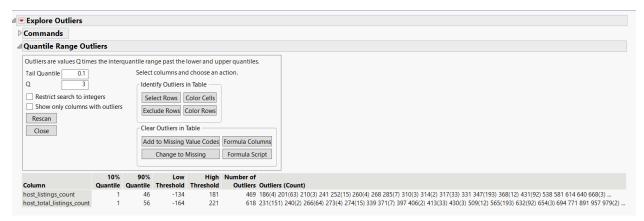
Host thumbnail url: Excluding as it is not related to the price.

Host picture url: Excluding as it is not related to the price.

Host neighbourhood: Excluding as it is not related to the price.

Host listing count: Keep the column as the number of listings can affect the price variable.

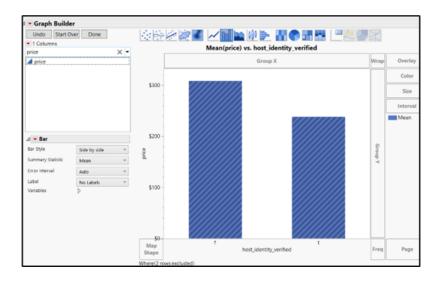
Host_total_listing_count: Keep the column as the total number of listings have relation with the price variable.



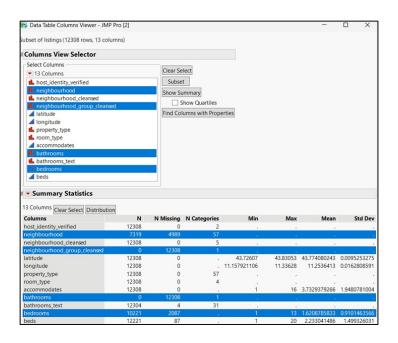
Host_verifications: Not relevant to the price. It does not affect the price variable.

Host_has_profile pic: Excluding as it is not related to the price.

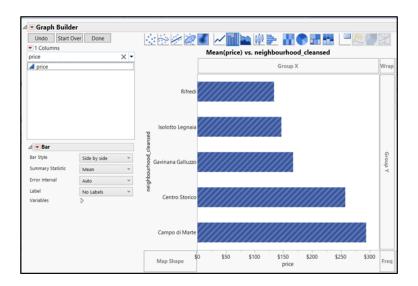
Host identity verified: No modification as it has direct relation with price.



Neighborhood: Excluded this column as it has 4989 missing values which cannot be imputed as data is categorical.



neighbourhood_cleansed: Kept as is as direct relation with price and no missing values.



Neighborhood group cleansed

Excluded this column as it has 0 available records.

Latitude: Kept as location of a place affects price.

Longitude: Kept as location of a place affects price.

property type: Kept as type of property should have an impact on price.

room type: Kept as room's type should have a direct impact on price.

Accommodates: Kept as capacity of property should have an impact on price

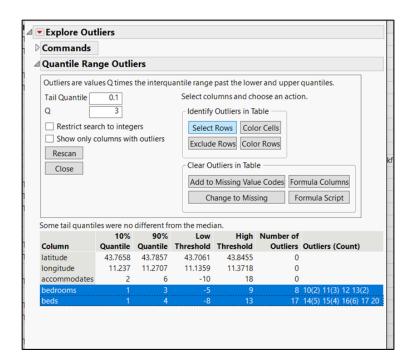
Bathrooms: Excluded this column as it has 0 available records.

Bathrooms_text: Kept as number and type of bathrooms should have an impact on price of property.

Bedrooms:

Imputed the data with average values, as 2087 values were missing.

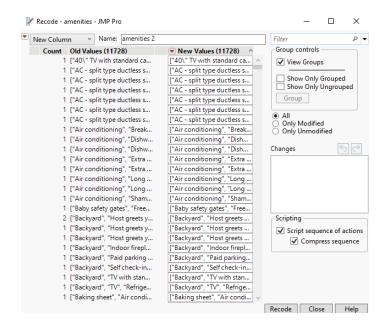
There are 8 outliers as well, but not imputing its data as the properties with more number of bedrooms have higher prices which is understandable and enriches data quality for real life predictions.



Beds: Imputed the data with average values as 87 values were missing.

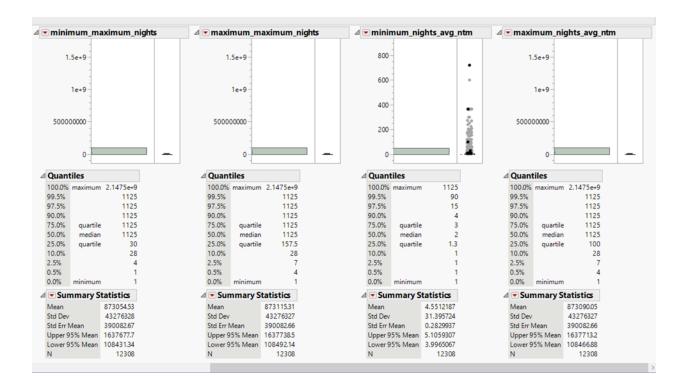
There are 17 outliers as well, but not imputing its data as it seems the properties with more no of beds have higher prices which is understandable and enriches data quality for real life prediction.

Amenities: Removing the variable a this is a categorical variable and it has no values which can be grouped together to make any meaningful pattern. Attached the screenshot.



Finding Outlier in the below columns:





Price: This is our target variable, we will not make any change to this column.

Minimum_nights: The data in the given column is not making any meaningful pattern with the price hence excluding it.

Maximum_nights: The data in the given column is not making any meaningful pattern with the price hence excluding it.

Minimum_minimum_nights: The data in the given column is not making any meaningful pattern with the price hence excluding it.

Maximum_minimum_nights: The data in the given column is not making any meaningful pattern with the price hence excluding it.

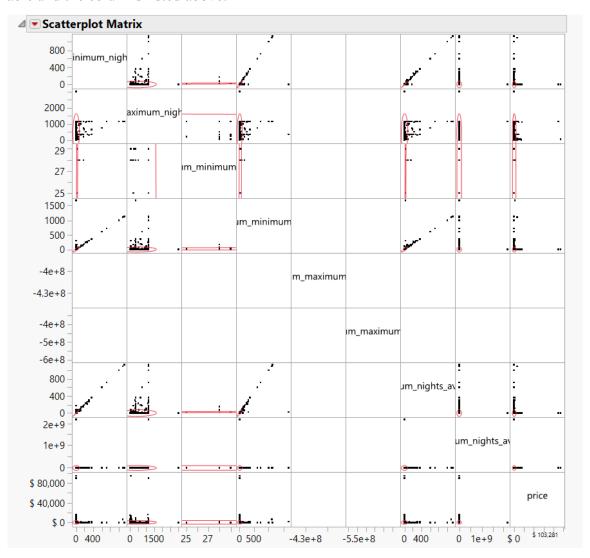
Minimum_maximum_nights: The data in the given column is not making any meaningful pattern with the price hence excluding it.

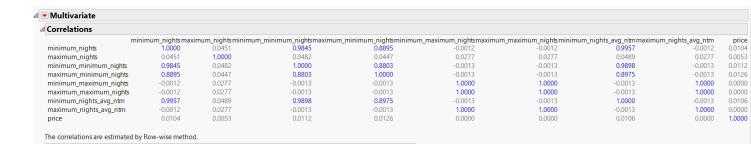
Maximum_maximum_nights: The data in the given column is not making any meaningful pattern with the price hence excluding it.

Minimum_nights_avg_ntm: The data in the given column is not making any meaningful pattern with the price hence excluding it.

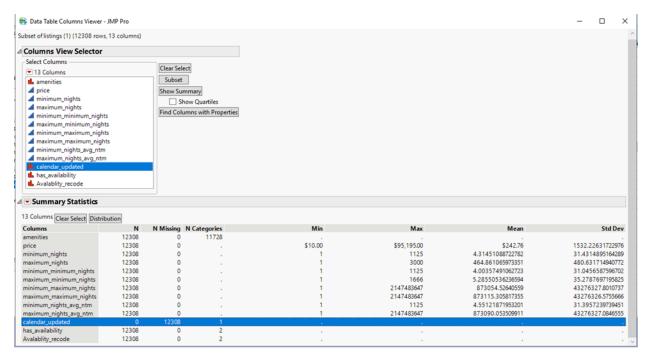
Maximum_nights_avg_ntm: The data in the given column is not making any meaningful pattern with the price hence excluding it.

I have attached the below screenshot, which shows that there is no relation between the target variable and the columns listed above.

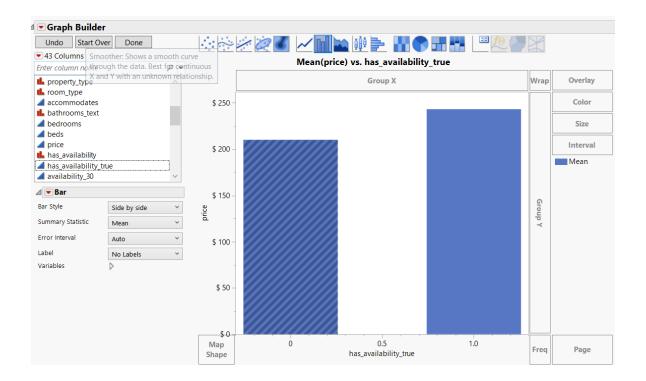




Calendar_updated : All Missing Values in this column, hence Excluding it from the data file. Attached screenshot.



Has_availability: Created Indicator column for the values t and f, and created new Column has_availablity_true



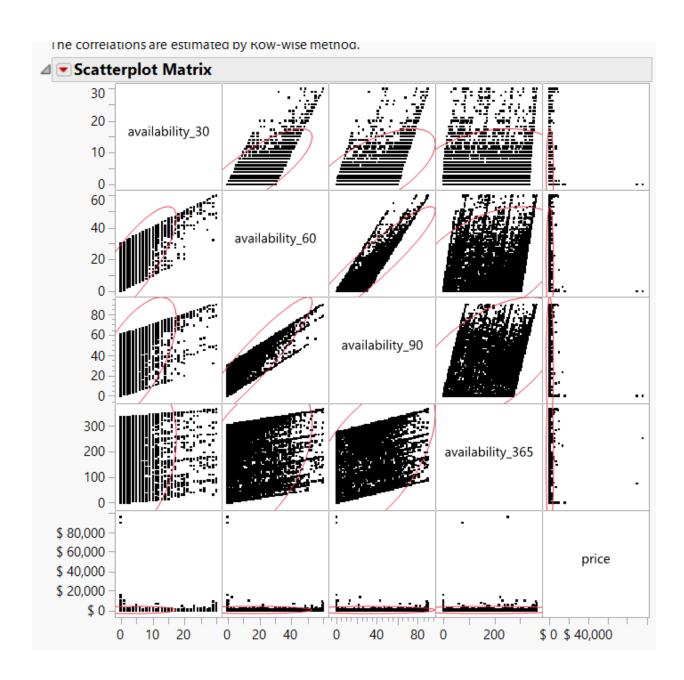
Availability_30: Has no relationship with price variables. That is why excluding it from the data set

Availability_60: Has no relationship with price variables. That is why excluding it from the data set.

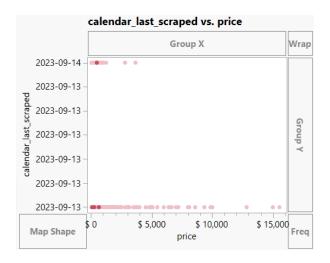
Availablility_90: Has no relationship with price variables. That is why excluding it from the data set

Availablility 365: Has no relationship with price variables.

All the available columns have no relation with price variables. That's why excluding them.

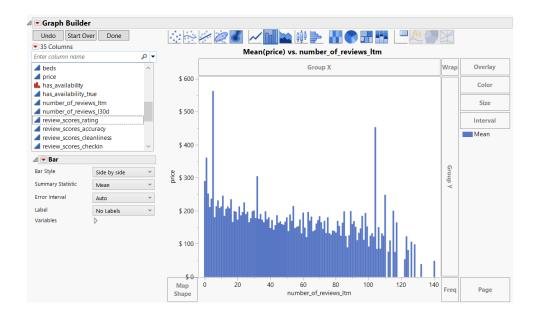


Calendar last scraped: This column provides us little to no information. Because there's only two dates when the calendar was last scraped. Hence can be eliminated.

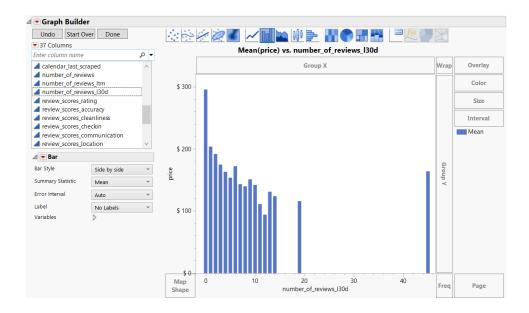


Number of reviews: There is no meaningful pattern between price and number of reviews.

Number_of_reviews_Itm: There exists a relationship with price.



number_of_reviews_I30d: It shows relation with price. We can see as the review increases the price decreases.



First_review: We are excluding this column because there's no correlation with price.

last review: We are excluding this column because there's no correlation with price.

Review scores rating: Imputed the data on the missing values.

Review scores accuracy: Imputed the data on the missing values.

Column	Number Missing
availability_30	0
availability_60	0
availability_90	0
availability_365	0
calendar_last_scraped	0
number_of_reviews	0
number_of_reviews_ltm	0
number_of_reviews_I30d	0
first_review	1499
last_review	1499
review_scores_rating	1499
review_scores_accuracy	1533

Imputation would fit well in this case since only 1499 or 1533 rows of null values are present in each column out of 12,308

rows.

△ Imputation Report	•	
Undo		
3032 missing values were replaced by le shrinkage estimate was used, with off-d of 0.99763. 1533 rows and 2 columns w missing value patterns across columns. I blue.	iagonals scaled by a facto ere affected. There were 2	r
△ Missing Columns		
☐ Show only columns with missing	Column	Number Missing
Close	review_scores_rating	0
Select columns and choose an action.	review_scores_accuracy	0
Select Rows Color Cells		
Exclude Rows Color Rows		

Column Name: review_scores_cleanliness

Comments:

N Missing: 1533; Missing Data - Decided to impute due to <12% of rows

Outliers: 63; Outlier - Exclude outliers as this will affect price

Column Name: review_scores_checkin

N Missing: 1534

Outliers: 74

Comments:

Missing Data - Decided to impute due to <12% of rows

Outlier - Exclude outliers as this will affect price

Column Name: review_scores_communication

N Missing: 1534

Outliers: 85

Comments:

Missing Data - Decided to impute due to <12% of rows

Outlier - Exclude outliers as this will affect price

Column Name: review scores location

N Missing: 1534

Outliers: 66

Comments:

Missing Data - Decided to impute due to <12% of rows

Outlier - Exclude outliers as this will affect price

Column Name: review scores value

N Missing: 1534

Outliers: 47

Comments:

Missing Data - Decided to impute due to <12% of rows

Outlier - Exclude outliers as this will affect price

After imputing missing data:

Missing Columns								
☐ Show only columns with missing	Column	Number Missing						
Close	review_scores_cleanliness	0						
	review_scores_checkin	0						
Select columns and choose an action.	review_scores_communication	0						
Select Rows Color Cells	review_scores_location	0						
Fuelude Device Color Berry	review_scores_value	0						
Exclude Rows Color Rows								

Column Name: license

N Missing: 10484

Outliers: #N/A

Comments:

Should exclude this column due to missing values and no effect on price

Column Name: instant bookable

N Missing: 0
Outliers: #N/A

Comments:

Add an indicator column 'Instant_bookable_true to have 0 and 1. This was done to create a continuous variable associated with this character value type to be able to analyze.

Column Name: instant bookable true

Comments:

Add an indicator column 'Instant_bookable_true to have 0 and 1. This was done to create a continuous variable associated with this character value type to be able to analyze.

Column Name: calculated_host_listings_count

N Missing: 0 Outliers: 477

Comments:

Outliers - Could be possible to have hosts with a large amount of listings, this should not affect the price model

Decision - Exclude from price model, as this should not affect price"

Column Name: calculated host listings count entire homes

N Missing: 0 Outliers: 477

Comments:

Outliers - Could be possible to have hosts with a large amount of listings, this should not affect the price model

Column Name: calculated_host_listings_count_private_rooms

N Missing: 0
Outliers: 45

Comments:

Outliers - Decided to leave alone due to the possibility of there being this amount of private rooms. The amount of outliers should not affect overall price model

Column Name: calculated_host_listings_count_shared_rooms

N Missing: 0
Outliers: 11

Comments:

Outliers - Decided to leave alone due to 11 shared rooms makes sense in a house

Column Name: reviews_per_month

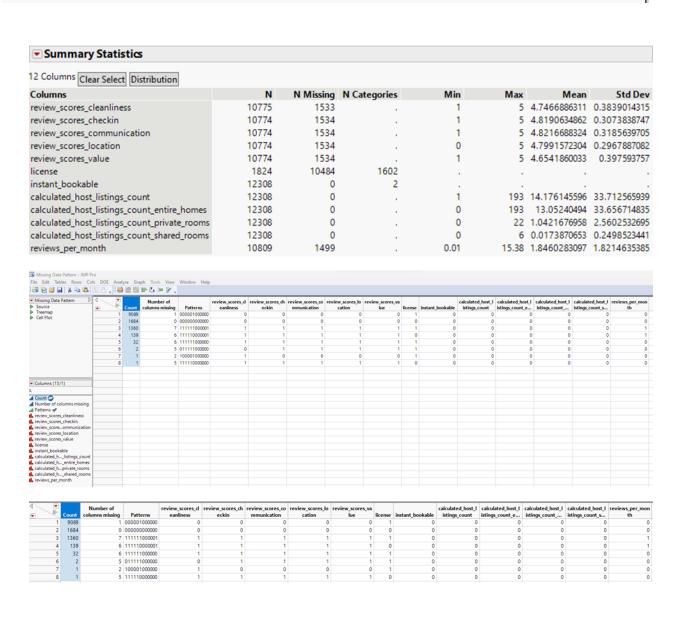
N Missing: 1499

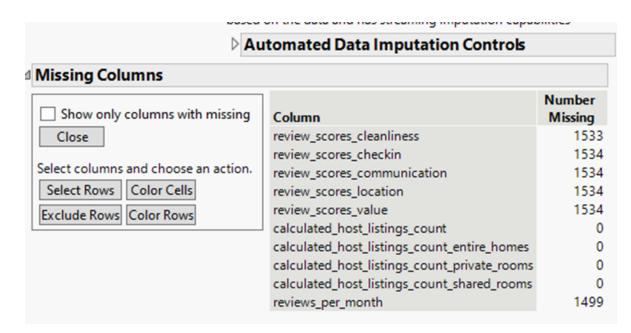
Outliers: 0

Comments:

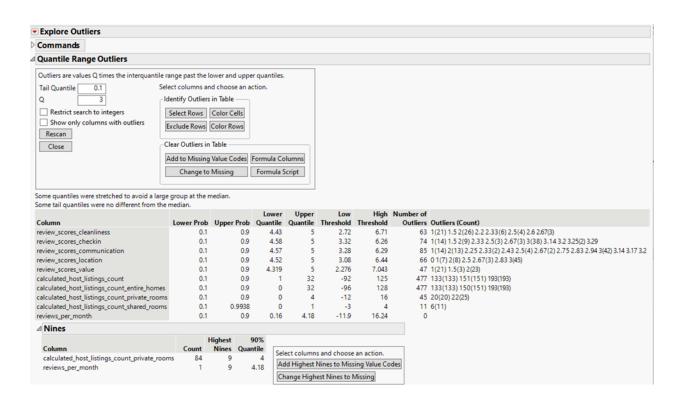
Missing Data - Decided to impute due to <12% of rows

■ Missing Columns								
Show only columns with missing	Column	Number Missing						
Close	reviews_per_month	0						
Select columns and choose an action. Select Rows Color Cells Exclude Rows Color Rows								





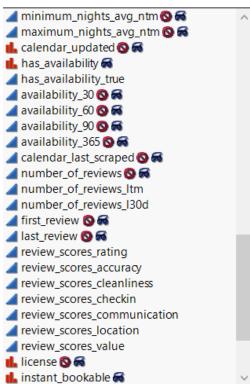
Outliers:



Total 30 columns we have finalized at the end in which 24 are continuous and 6 are categorical.

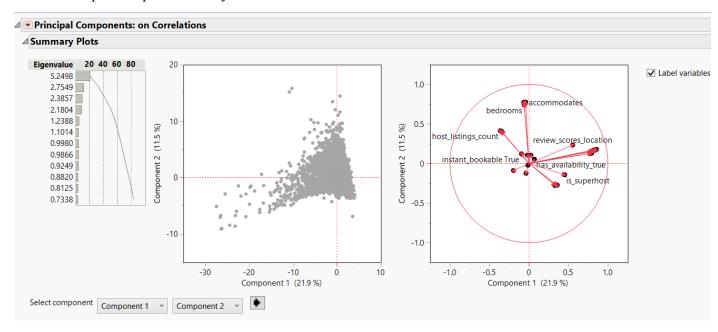








PCA: Principal component Analysis has been done on the 24 continuous columns of the data set.



We can see that 95% of accuracy can be achieved by reducing the column to 18

Eigenva	lues			
Number	Eigenvalue	Percent	20 40 60 80	Cum Percent
1	5.2498	21.874		21.874
2	2.7549	11.479		33.353
3	2.3857	9.940		43.293
4	2.1804	9.085		52.378
5	1.2388	5.162		57.540
6	1.1014	4.589	\	62.129
7	0.9980	4.158		66.287
8	0.9866	4.111		70.398
9	0.9249	3.854	\	74.252
10	0.8820	3.675	\	77.927
11	0.8125	3.385		81.312
12	0.7338	3.057		84.370
13	0.6375	2.656		87.026
14	0.5689	2.371		89.397
15	0.4666	1.944		91.341
16	0.4029	1.679		93.019
17	0.3263	1.360		94.379
18	0.2888	1.203		95.582
19	0.2608	1.087		96.669
20	0.2377	0.991		97.659
21	0.2088	0.870		98.529
22	0.1825	0.760		99.290
23	0.1270	0.529		99.819
24	0.0434	0.181		100.000

Eigenvectors													
	Prin1	Prin2	Prin3	Prin4	Prin5	Prin6	Prin7	Prin8	Prin9	Prin10	Prin11	Prin12	Prin13
is_superhost	0.19789	-0.08619	0.15152	-0.16114			-0.07788		-0.07662	0.34614	-0.25522	0.78344	0.15816
host_listings_count	-0.15826	0.24936	0.33462	0.37622	0.13264	-0.10445	0.07226			0.12925	-0.11360	0.04232	
host_total_listings_count	-0.14953	0.24204	0.32906	0.37462	0.13845	-0.11957	0.07585			0.14925	-0.12876	0.04719	
latitude		-0.07476	-0.07738		0.66064	0.04530	-0.04792	0.15392	0.23014	0.21434	0.50929		0.29856
Iongitude		0.06348			-0.52796	0.07231	0.34500	-0.23655	0.33866	0.44285	0.46966		
accommodates		0.46750		-0.33239			-0.03431			-0.04754			
bedrooms	-0.02584	0.44602		-0.32839	0.06270	0.03224	0.03466	-0.04898	0.07825				
beds		0.46694		-0.33616	0.08220	0.03451		-0.06221			-0.04591		
price		0.06308	-0.02446		-0.07667	0.14704	0.55279	0.79370	-0.14655	0.05654	-0.06780		
has_availability_true			0.13283		-0.14115	0.50896	-0.54030	0.24978	0.14366	0.42087	-0.19981	-0.32730	
number_of_reviews_ltm	0.15643	-0.16438	0.44514	-0.24986			0.08253					-0.02594	
number_of_reviews_I30d	0.14242	-0.16677	0.44218	-0.22614	0.05483	-0.05924	0.09602		0.05039	-0.02304		-0.16228	-0.01994
review_scores_rating	0.35812	0.10239		0.12740	0.06809					0.04457	0.03931		-0.41358
review_scores_accuracy	0.36508	0.10121		0.14046						0.01744	0.06079		-0.36942
review_scores_cleanliness	0.34232	0.09032	-0.02984	0.12710	0.06714						0.02454	0.04200	-0.22557
review_scores_checkin	0.33818	0.07635	-0.07684	0.10724	0.04562	-0.01774	0.02114	0.01514				-0.05006	0.33853
review_scores_communication	0.34695	0.07918	-0.05781	0.10533	0.04649		0.02484					-0.04639	0.34127
review_scores_location	0.24221	0.14048	0.05534	0.11307	-0.28524			-0.07212		-0.25646	-0.09379	-0.14800	0.53359
review_scores_value	0.37385	0.10759	-0.03939	0.10994	0.04698					-0.03164			-0.03947
instant_bookable True	-0.04137	0.07353	0.26110	0.06792	-0.16256	0.39212	-0.24427	0.14529	-0.12848	-0.46031	0.48667	0.36338	
calculated_host_listings_count_entire_homes	-0.14937	0.24242	0.27378	0.26343		-0.10385	-0.06923		-0.16764		0.15647	-0.07387	
calculated_host_listings_count_private_rooms	-0.08610			0.15117	0.16840	0.48121	0.27779	-0.15510	0.58166	-0.30199	-0.31011	0.13512	
calculated_host_listings_count_shared_rooms		-0.01334			0.20802	0.52888	0.28876	-0.40563	-0.61707	0.18787		-0.09671	
reviews_per_month	0.14897	-0.16515	0.42057	-0.21882	0.06920	-0.03341	0.12925		0.05252	-0.08537	0.07270	-0.22753	-0.06480

Prin14	Prin15	Prin16	Prin17	Prin18	Prin19	Prin20	Prin21	Prin22	Prin23	Prin24
0.10164	0.12885	-0.03702	0.12339	0.02196	0.14106	0.05252	-0.02214	0.00934	0.03022	
-0.06382	-0.24551	-0.03577	-0.02129	0.01235	-0.00184	-0.01031	0.00391	0.01387		-0.72484
-0.07389	-0.32766	-0.04822	-0.02886	0.03010		-0.01445	0.01115	-0.02313	0.00587	0.68585
0.26397	-0.01848	-0.06171		-0.01411	-0.00504	0.01591	0.02322	0.01446		
-0.03416	-0.01402	0.03002	-0.01009	-0.02666	0.00160				0.00109	0.00176
-0.01232	-0.05109	0.00140	-0.04139	-0.48386		0.03379	-0.06523	0.64648		0.02306
-0.02115	0.01168	0.00772	0.09942	0.80443	-0.03866	-0.03131	0.04124	0.11746	-0.00297	-0.00394
-0.01810	0.02301	-0.01076	-0.05668	-0.30206	0.05219	0.00487	0.04961	-0.74113		-0.01314
0.03569		0.02021	-0.00436	-0.00934			-0.01048	0.00264		
-0.01484	-0.01057		0.02061	-0.00224			-0.01452		0.01382	0.00071
0.03522	0.04040		-0.13708	-0.02499	-0.77751	-0.18903	0.10805	-0.02129	-0.03571	-0.00131
-0.03425	0.08645	0.05011	-0.61883	0.10726	0.49154	0.10036	-0.01948	0.03491	0.01409	0.00224
0.10340	0.06361	-0.42785	-0.01137	-0.00287	0.03933		0.11844	0.00964	-0.67123	
0.04317	0.07640	-0.36946	-0.03194	0.00951	-0.04190	0.01628	0.04542		0.73451	
0.16722	-0.12963	0.72239	0.02894	-0.01878	0.01051	0.14240	0.45748	0.01128		-0.00244
-0.51228	0.08541	-0.03275	-0.03041	0.01616	-0.19180	0.65213	-0.06839	-0.00821	-0.05580	
-0.42606	0.11824	0.01168	0.04716	-0.04830	0.17382	-0.67091	0.22947	0.04756		0.00128
0.59769	-0.10869	-0.20010		0.00779	0.04772	0.07826	0.12884	0.00648	0.02161	-0.00034
0.13592	-0.08846	0.28543	-0.04646	0.03683	-0.05614	-0.19117	-0.81893	-0.06938	-0.05415	-0.00172
-0.15288	-0.16808	-0.02993	-0.04406	0.04830	0.01291	-0.01081	-0.01124	-0.06035	-0.00955	0.00381
0.13002	0.80019	0.14622	0.11692	-0.01166	-0.01126	0.03241	-0.04878		-0.01259	0.05420
-0.00214	0.23425	0.02411	0.01880	-0.04288	0.01018	0.00497		0.05372		0.02244
0.00848	-0.01644		-0.01845		0.00091		0.00284	0.03924	0.00197	
-0.06261	-0.10695	-0.01498	0.73665	-0.06347	0.23115	0.09289	-0.07389	-0.02662	0.01857	