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# 1. Column Summary

Project 1 was started by undertaking an overview of all the columns which were available in the "Listings" JMP file for Western Australia. There was a total of 75 columns and 10,485 rows.

Upon going through the data set, the columns below summarize the project's initial findings. In accordance with the original data set, the table outlines:

- (1) Column number
- (2) Column name
- (3) Activity-
  - Retain the column.
  - Hide and exclude the column.
  - Modify the column, where a new column is created, and the original column is hidden and excluded
- (4) Column description and rationale for the activity.

Table 1: Column Summary

Col No.	Column name	Activity	Column Description and Rationale
1	ID	Retained	This column indicates the ID of the individual property or Airbnb listing. It is a unique key and differentiates every single row from the others. Therefore the ID column is retained.
2	Listing_url	Hidden and excluded	This column indicates the url of the listing specified by the ID. There is no influence of listing url and hence is a non-useful data set to be considered as a predictor variable for the target. Hence it has been hidden and excluded from the analysis.
3	Scrape_id	Hidden and excluded	This column indicates the ID for when the scraping happened. Given there is no time reference to be put to the data, there is no impact of the scraping on the target variable; scrape ID will not influence the target and has been hidden and excluded.
4	Last_scraped	Hidden and excluded	This column indicates the date the data was last scrapped on. As the time references on the data set as mentioned above are not required, there is no need for the column and hence has been hidden and excluded from the analysis.

Col No.	Column name	Activity	Column Description and Rationale
5	Source	Hidden and excluded	This column indicates the source for scraping the data. There is no influence on the target variable, based on the source of scrapping, hence the column has been hidden and excluded.
6	Name	Hidden and excluded	This column describes the listing in slight detail and is a non - useful data set as a predictor variable to determine prices of properties as there is no influence this description can have on figuring property prices.
7	Description	Hidden and excluded	This column describes in greater detail about the property listing. It indicates the location, nearby locality details to figure out the place and can be used to trace the listing in question. However, it is of no use to find the prices of the listing and hence is being hidden and excluded from the analysis.
8	Neighborhood _overview	Hidden and excluded	This column describes the neighborhood where the property is. Better neighborhoods can help charge better prices for the property. The information from column "neighborhood cleansed" will be a better determinant with clear categories of the neighborhood. Hence, this column has been excluded and hidden from the analysis.
9	Picture_url	Hidden and excluded	This column describes the url of the picture of the property in question. As the prices cannot be determined by picture url and does not serve as a predictor, therefore it is excluded from the data set.
10	Host_id	Retained	This column indicates a unique ID for every single host. The column also has the same hosts for different properties. This predictor variable can help in the analysis of prices, if a particular host charges higher prices on average. Thus, it can help in modeling the target variable. Therefore, the column has been retained.
11	Host_url	Hidden and excluded	This column describes the url of the host. As url cannot help in building the model to determine prices, it needs to be dropped as it only gives the url of the host and is immaterial to the target variable.

Col No.	Column name	Activity	Column Description and Rationale
12	Host_name	Hidden and excluded	This column indicates the name of the host. As for distinction, the host ID is used already, therefore the host name is hidden and excluded to reduce the complexity.
13	host_since	Hidden and excluded	This column indicates the time since the host has been a "host" and hence has no influence on the price of the property. Hence, it is a non useful data column and is being hidden and excluded from the analysis.
14	host_location	Hidden and excluded	This column indicates the location of the host of the property in question. For determining the price, it does not matter which part of the world the host is located in and hence is being dropped from the data set.
15	host_about	Hidden and excluded	This column indicates a bit about the host. The host describes who they are and what they like. It gives no details about the property and does not help in determining the target variable. Hence, it is being dropped from the data set.
16	host_response_ time	Modified	This column describes the time it takes for the host to respond to a customer request. It is categorized into different time references - under an hour, in a day, etc and can be a good data point to build user ratings and determine prices. Hence, it is being retained in the data set.
17	host_response_ rate	Retained	This column describes the response rate of the host, if the host has been able to respond back to the customer 100% or any lesser. This can be a good predictor variable to understand how good the services of a particular listing are and succeedingly can help in determining prices. Therefore, it is being retained for analysis.
18	host_acceptanc e_rate	Hidden and excluded	This column indicates the acceptance rate of the customer requests and is at 100% or lesser. The acceptance rate if higher or lower cannot have an influence on the price a particular property may charge as it purely depends on the traffic inflow. Acceptance might be low during peak traffic and still the property might charge low or high based on their service. Hence, the column has been dropped from the analysis.

Col No.	Column name	Activity	Column Description and Rationale
19	host_is_superh ost	Hidden and excluded	This column was hidden and excluded as it is a formula used by Airbnb to determine this. It seems to be a weighted average of the previous two columns, host_response_time and host_response_rate.
20	host_thumbnail _url	Hidden and excluded	This column provides an image of the host's profile picture on Airbnb. However, this column was hidden and excluded as it does not provide additional information or insight in determining the target variable, Price.
21	host_picture_ur	Hidden and excluded	This column, similar to host_thumbnail_url, provides an image of the host's profile picture on Airbnb. However, this column was also hidden and excluded as it does not provide additional information or insight in determining the target variable, Price.
22	host_neigbourh ood	Hidden and excluded	This column provides the host's neighborhood. It was hidden and excluded as more than 95% of the values are missing. Furthermore, this is not a useful column in predicting the target variable.
23	host_listings_c ount	Hidden and excluded	This column provides the number of host listings on the Airbnb website. However, this column was hidden and excluded as it does not provide additional information or insight in determining the target variable.
24	host_total_listi ngs_count	Hidden and excluded	Similar to the column, host_listings_count, this column provides the total number of host listings on the Airbnb website. However, this column was also hidden and excluded as it does not provide additional information or insight in determining the target variable.
25	host_verificatio ns	Hidden and excluded	This column provides the method used to verify the host, combinations of email and phone are seen in this column. However, this column was hidden and excluded as it contains redundant information that will be seen in column host_identity_verified. It is not imperative to know what method was used to verify the host as long as the host is verified and thereby legitimate.

Col No.	Column name	Activity	Column Description and Rationale
26	host_has_profil e_pic	Hidden and excluded	This column provides if the host has a profile picture or not on the Airbnb website. This column was hidden and excluded as it does not provide additional information or insight in determining the target variable.
27	host_identity_v erified	Modified	This column provides information regarding if the host is verified or not. This column was modified using dummy variables/indicator columns. The original column consists of true or false values, which is two dimensional. Thereby, a new column Host_Identity_Verified_True, is created that consists of 1s or 0s, that depicts if the host is verified or not. Accordingly, if a 0 is seen in the new modified column, then the host is not verified.
28	neighborhood	Hidden and excluded	This column provides the neighborhood that the property resides in. It was hidden and excluded as over 30% of the values are missing. Moreover, the next column, neighborhood_cleansed, depicts the same information in a more concise manner. This column also consists of redundant information as well.
29	neighbourhood _cleansed	Retained	This column provides the neighborhood the property is located in. It has no missing values and is retained. It gives us information regarding the neighborhood and location, which is the most important factor in determining real estate prices. This is a very important predictor variable in determining the target variable.
30	neighbourhood _group_cleanse d	Hidden and excluded	This column is attempting to group neighborhoods together. However, it only consists of missing values. Therefore, it is a column that was hidden and excluded as there is no information that can be extrapolated.
31	Latitude	Hidden and excluded	This column provides information regarding the latitude of the property using WGS84. This column was hidden and excluded, as it is redundant information and not required to predict the target variable.

Col No.	Column name	Activity	Column Description and Rationale
32	Longitude	Hidden and excluded	This column provides information regarding the longitude of the property using WGS84. This column was hidden and excluded, as it is redundant information and not required to predict the target variable.
33	property_type	Hidden and excluded	This column provides information about the property itself. It provides information regarding what the host is trying to rent, e.g. a private room, the entire home/apartment, etc. While it is very important in determining price, the column is hidden and excluded as column room_type provides a more concise representation of the same information.
34	room_type	Retained	This column provides information regarding the host's intention on how they want to rent out their property. It consists of 4 values: shared room, private room, hotel room, and entire home/apartment. This is a very important column in determining the target variable, as it generally gives us insight regarding the size of the property being rented out. Therefore, this column is retained.
35	accommodates	Retained	This column provides information regarding the number of people that the host would accept for the property. This column is also retained, as it generally gives us insight into the size of the property, in combination with the column room_type. This will be a very important variable in predicting the target variable.
36	bathrooms	Hidden and excluded	This column provides information regarding the number of bathrooms in the property. Unfortunately, while important, most of the data was missing, so it had to be hidden and excluded. However, column <code>bathroom_texts</code> captures this information in a more concise manner, making the column bathrooms redundant as well.

Col No.	Column name	Activity	Column Description and Rationale
37	bathroom_texts	Modified	This column describes the total number of bathrooms in string. Certain rows indicate whether bathrooms are shared or private. This column has been modified and re-coded into a numeric column 'Total Bathrooms', which indicates the total number of bathrooms in the property. The information regarding the type of bathroom (private, shared) was available only for less than 10% of the data and hence was dropped.
38	bedrooms	Retained	This integer type column describes the total number of bedrooms in the listed property. It has been retained given its relation to the target variable. Higher bedrooms would generally indicate higher prices.
39	beds	Retained	This integer type column describes the total number of beds available in the listed property. It has been retained given its relation to the target variable price. Higher number of beds would generally indicate higher prices.
40	amenities	Hidden and excluded	This column had indecipherable data. Given any treatment, it was not possible to incorporate the data into the data set. Therefore, the column was excluded and hidden.
41	price	Retained	This column shows the daily price in Australian Dollars. This serves as the target variable of the data set and has been retained.
42	minimum_nigh ts	Modified	This column describes the minimum number of nights required to accept a booking as listed by the host. Given its importance in determining the price for the listing, this column was retained.
43	maximum_nigh ts	Retained	This column describes the maximum number of nights allowed for a customer to request in a booking as listed by the host. Given its importance in determining the price for the listing, this column was retained.

Col No.	Column name	Activity	Column Description and Rationale
44	minimum_mini mum_nights	Hidden and excluded	This is a calculated column showing the smallest value of minimum nights. Due to the ambiguity of the calculation involved in deriving the column, it was hidden and excluded. Also, the column does not have an impact in determining the target variable given the columns: minimum and maximum nights.
45	maximum_mini mum_nights	Hidden and excluded	This is a calculated column showing the largest value of minimum nights. Due to the ambiguity of the calculation involved in deriving the column, it was hidden and excluded. Also, the column does not have an impact in determining the target variable given the columns: minimum and maximum nights.
46	minimum_maxi mum_nights	Hidden and excluded	This is a calculated column showing the smallest value of maximum nights. Due to the ambiguity of the calculation involved in deriving the column, it was hidden and excluded. Also, the column does not have an impact in determining the target variable given the columns: minimum and maximum nights.
47	maximum_max imum_nights	Hidden and excluded	This is a calculated column showing the largest value of maximum nights. Due to the ambiguity of the calculation involved in deriving the column, it was hidden and excluded. Also, the column does not have an impact in determining the target variable given the columns: minimum and maximum nights.
48	minimum_nigh ts_avg_ntm	Hidden and excluded	This calculated column shows the average number of minimum nights. As its inclusion in the data set will lead to data redundancy with the inclusion of the original column 'minimum_nights', this column was hidden and excluded.
49	maximum_nigh ts_avg_ntm	Hidden and excluded	This calculated column shows the average number of maximum nights. As its inclusion in the data set will lead to data redundancy with the inclusion of the original column 'maximum_nights', this column was hidden and excluded.
50	calendar_updat ed	Hidden and excluded	This column was blank and fully comprised of missing values. Therefore, it was hidden and excluded.

Col No.	Column name	Activity	Column Description and Rationale
51	has_availability	Hidden and excluded	This column indicates TRUE/FALSE values to show whether the listing has availability. Given the similarity in data points, this column was hidden and excluded. More than 99% of data indicates the same response - TRUE and hence has no additional insight from the column.
52	availability_30	Hidden and excluded	This column indicated the availability of each listing 30 days in the future from the date the data was last scrapped. The column was hidden and excluded as a reference on the occupancy of the property and has information overlap with column availability_365 which is being retained.
53	availability_60	Hidden and excluded	This column indicated the availability of each listing 60 days in the future from the date that the data was last scrapped. The column was hidden and excluded as a reference on the occupancy of the property and has information overlap with column availability_365 which is being retained.
54	availability_90	Hidden and excluded	This column indicated the availability of each listing 90 days in the future from the date that the data was last scrapped. The column was hidden and excluded as a reference on the occupancy of the property and has information overlap with column availability_365 which is being retained.
55	availability_36 5	Retained	This column indicates the availability of each listing 365 days in the future from the date the data was last scrapped. A property's high occupancy rate indicates that the rates will be high. Hence, it will provide a more accurate estimate of the occupancy and will influence the price of the property. The column is retained given its relationship with the target variable.
56	calendar_last_s crapped	Retained	The calendar_last_scrapped column will serve as a baseline date to check the availability of each listing in the future. This column will be used by the availability_365 column to check the availability of each listing 365 days in the future. Therefore, this column is retained.

Col No.	Column name	Activity	Column Description and Rationale
57	number_of_rev iews	Modified	This column indicates the number of reviews a listing has and this variable will impact the price of the property. The column is modified by using SHASH transformation to normalize the distribution. ;The new column is called SHASH Transformed to Normal number_of_reviews.
58	number_of_rev iews_ltm	Hidden and Excluded	This column indicates the number of reviews the listing has had in the last 12 months. The column is a calculated and redundant column. The number_of_reviews column already contains all the reviews a property has. Therefore, it was hidden and excluded.
59	number_of_rev iews_130d	Hidden and Excluded	This column indicates the number of reviews the listing has had in the last 30 days. The column is a calculated and redundant column. The number_of_reviews column already contains all the reviews a property has. Therefore, it was hidden and excluded.
60	first_review	Hidden and Excluded	This column indicates the date of the first/oldest review. The first_review column does not have any impact on the price of a listing; so it was hidden and excluded.
61	last_review	Hidden and Excluded	This column indicates the date of the last/newest review. The first_review column does not have any impact on the price of a listing; so it was hidden and excluded.
62	review_score_r ating	Modified	This column indicates the review score rating. There are 7 columns in the table which have different types of review data and to reduce the dimensionality, PCA was applied on those 7 columns. After reviewing the PCA, most of the data i.e 91.29% was present within the first 4 PCA's.The new columns formed in the table are Review_Scores_PCA1, Review_Scores_PCA2, Review_Scores_PCA3 and Review_Scores_PCA4. Therefore, the review_score_rating is hidden and excluded.

Col No.	Column name	Activity	Column Description and Rationale
63	review_score_a ccuracy	Modified	This column indicates the review score accuracy. There are 7 columns in the table which have different types of review data and to reduce the dimensionality, PCA was applied on those 7 columns. After reviewing the PCA, most of the data i.e 91.29% was present within the first 4 PCA's. The new columns formed in the table are Review_Scores_PCA1, Review_Scores_PCA2, Review_Scores_PCA3 and Review_Scores_PCA4. Therefore, the review_score_accuracy is hidden and excluded.
64	review_score_c leanliness	Modified	This column indicates the review score cleanliness. There are 7 columns in the table which have different types of review data and to reduce the dimensionality, PCA was applied on those 7 columns. After reviewing the PCA, most of the data i.e 91.29% was present within the first 4 PCA's. The new columns formed in the table are Review_Scores_PCA1, Review_Scores_PCA2, Review_Scores_PCA3 and Review_Scores_PCA4. Therefore, the review_score_cleanliness is hidden and excluded.
65	review_score_c heckin	Modified	This column indicates the review score check in. There are 7 columns in the table which have different types of review data and to reduce the dimensionality, PCA was applied on those 7 columns. After reviewing the PCA, most of the data i.e 91.29% was present within the first 4 PCA's. The new columns formed in the table are Review_Scores_PCA1, Review_Scores_PCA2, Review_Scores_PCA3 and Review_Scores_PCA4. Therefore, the review_score_checkin is hidden and excluded.
66	review_score_c ommunication	Modified	This column indicates the review score communication. There are 7 columns in the table which have different types of review data and to reduce the dimensionality, PCA was applied on those 7 columns. After reviewing the PCA, most of the data i.e 91.29% was present within the first 4 PCA's. The new columns formed in the table are Review_Scores_PCA1, Review_Scores_PCA2, Review_Scores_PCA3 and Review_Scores_PCA4. Therefore the review_score_communication is hidden and excluded.

Col No.	Column name	Activity	Column Description and Rationale
67	review_score_l ocation	Modified	This column indicates the review score location. There are 7 columns in the table which have different types of review data and to reduce the dimensionality, PCA was applied on those 7 columns. After reviewing the PCA, most of the data i.e 91.29% was present within the first 4 PCA's. The new columns formed in the table are Review_Scores_PCA1, Review_Scores_PCA2, Review_Scores_PCA3 and Review_Scores_PCA4. Therefore, the review_score_location is hidden and excluded.
68	review_score_v alue	Modified	This column indicates the review score value. There are 7 columns in the table which have different types of review data and to reduce the dimensionality, PCA was applied on those 7 columns. After reviewing the PCA, most of the data i.e 91.29% was present within the first 4 PCA's.The new columns formed in the table are Review_Scores_PCA1, Review_Scores_PCA2, Review_Scores_PCA3 and Review_Scores_PCA4. Therefore, the review_score_value is hidden and excluded.
69	license	Hidden and Excluded	This column indicates the license/permit/registration number. The license column is hidden and excluded as most of the data was missing in the column.
70	Instant_bookab le	Hidden and Excluded	This column indicates whether the guest can automatically book the listing without the host accepting their booking request. The target variable is not impacted by whether a reservation can be made automatically or not. So, the instant_bookable column has no impact on the target variable. Therefore, it is hidden and excluded.
71	Calculated_hos t_listings_count	Hidden and Excluded	This column indicates the number of listings that the host has in the current scrape based across city/region geography. The target variable is not impacted by the number of properties a host has listed. Therefore, this column is hidden and excluded.
72	Calculated_hos t_listings_count _entire_homes	Hidden and Excluded	This column indicates the number of entire home/apartment listings that the host has in the current scrape based across the city/region geography. The target variable of a listing is not impacted by the number of home/apartment that a host has listed. Therefore, this column is hidden and excluded.

Col No.	Column name	Activity	Column Description and Rationale
73	Calculated_hos t_listings_count _private_rooms	Hidden and Excluded	This column indicates the number of private room listings the host has in the current scrape, in the city/region geography. The target variable is not impacted by the number of private rooms a host has listed. Therefore, this column is hidden and excluded.
74	Calculated_hos t_listings_count _shared_rooms	Hidden and Excluded	This column indicates the number of shared room listings that the host has in the current scrape, across the city/region geography. The target variable will not be impacted by the number of shared rooms a host has listed. Therefore, this column is hidden and excluded.
75	Reviews_per_ month	Hidden and Excluded	This column indicates the number of reviews that a listing has per month. The data present in this column is redundant and therefore, it is hidden and excluded.

# 2. Data Preprocessing:

In order to build a model, dataset 'Listings' was pre-processed using a range of data preprocessing techniques to reduce the complexity of the dataset and clean the data set. The following steps and techniques were used:

- 2.1 Hide and Exclude
- 2.2 Binning, Indicator Columns, and Re-code
- 2.3 Missing Values
- 2.4 Outliers
- 2.5 Reducing dimensionality

The detailed treatment of the data set has been outlined in the below sections along with appropriate screenshots to support the process.

#### 2.1 Hide and exclude

All columns were reviewed as mentioned in Section 3 Column Summary to ensure well-defined and appropriate columns were identified and included in the data set. All columns which were found to not be relevant or would not help with the prediction of the target variable were hidden and excluded.

A total of 42 columns were hidden and excluded from the data set. The remaining 33 columns were then further worked upon in the next section.

Figure 1 shows an example of column 'Amenities' which included non - decipherable data sets when converted into columns for data processing. As a result, this column was hidden and excluded from the data set.

| Street | S

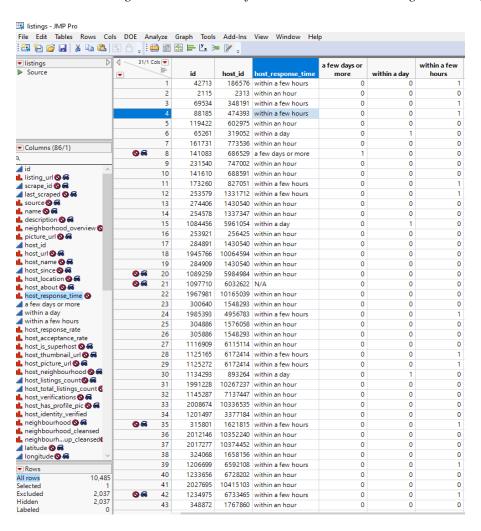
Figure 1: Screenshot of column Amenities

#### 2.2 Binning, Indicator Columns, and Re-code

A total of 3 columns were assessed and treated in this section. These are:

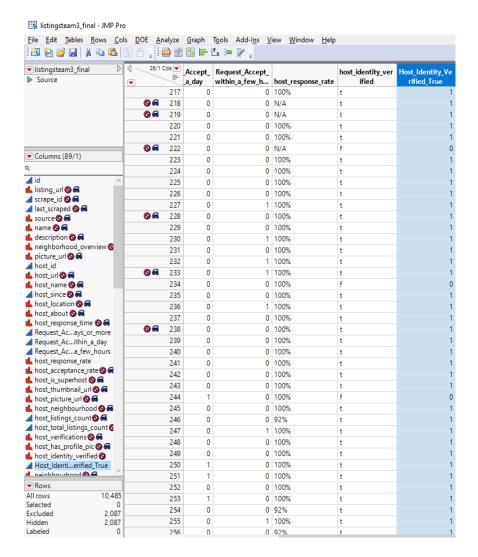
- (1) host\_response\_time
- (2) host\_identity\_verified
- (3) bathroom\_text

Figure 2: Screenshot of indicator columns breaking down host\_response\_time



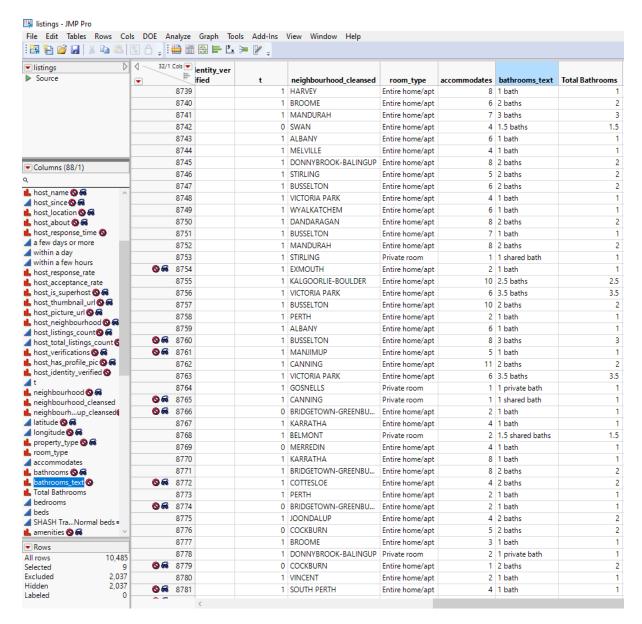
Here, the categorical variable, **Host\_response\_time**, is converted into a numeric variable by creating indicator columns. There was a total of 4 categories and by creating indicator columns, the columns were reduced to 3.

Figure 3: Screenshot of creating indicator columns for Host\_identity\_verified



For the categorical variable, **host\_identity\_verified**, a new indicator column was created called **Host\_Identity\_Verified\_True**. The original column had a dimensionality of 2. Therefore, by creating a new indicator column, the total number of columns is reduced to 1, while retaining all the information.

Figure 4: Screenshot of re-coding bathroom\_text and conversion into numerical column



Finally, in Figure 4, to break down the content of the **bathroom\_text** column, a re-coding of the data was selected. A new column was created which was kept numeric in nature. Upon re-coding the content, this column was called **Total Bathrooms**. It reflects the total number of bathrooms available on the listed property.

With this treatment, the most essential information about bathrooms was retained i.e. the number of bathrooms. Partial information such as whether the bathroom is shared or private was excluded given that less than 10% of the data provided such information.

### 2.3 Missing Values

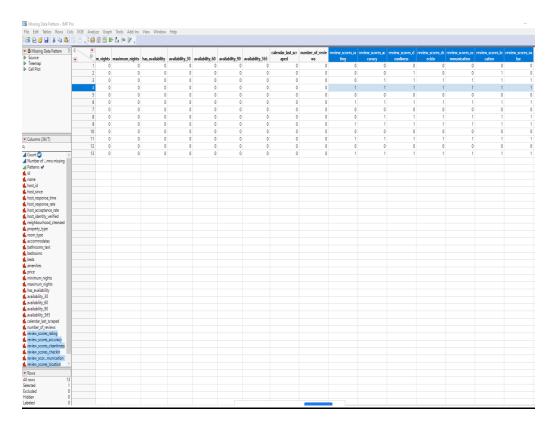
### Hidden and exclusion

In order to further fix abnormalities in the data set, missing values in each column were identified and explored using the 'Missing Data Pattern' function in JMP. Figure 5 shows the missing data pattern of the data.

Missing Data Pattern - JMP Pro File Edit Tables Rows Cols DOE Analyze Graph Tools Add-Ins View Window Help : 📇 🚰 📂 🔒 | 🔉 🖦 🖎 | 🕾 A 🛫 : 🖶 🛗 🛗 🖽 📂 📝 **▼** ➤ Source
➤ Treemap
➤ Cell Plot Count Patterns 9154 6 0000000000000000000000000000111111 989 7 000000000000000000000000001111111 59 8 000000000000001000000000001111111 201 7 000000000000010000000000000111111 8 000000000000010000000000001111111 2 00000000000011000000000000000000 ▼ Columns (36/7) 11 9 0000000000001100000000001111111 12 ✓ Count <a> 13 8 000000000000100000000000001111111 → Patterns � i id 🔥 name host\_id host\_response\_time
host\_response\_rate host\_acceptance\_rate
host\_identity\_verified
neighbourhood\_cleansed nroperty\_type accommodates
bathrooms\_text d bedroomsd beds 📠 amenities minimum\_nights
maximum\_nights
has\_availability
availability\_30 availability\_60 availability\_90
availability\_90
availability\_365
calendar\_last\_scraped
number\_of\_reviews review\_scores\_rating
treview\_scores\_accuracy
treview\_scores\_cleanliness review\_scores\_checkin review\_scores\_location

Rows All rows Labeled

Figure 5: Screenshot of identified missing value patterns



As shown in the above figure, 1331 rows were hidden and excluded from the data set. These are all the rows that are identified to have 1 or more missing values in each row. The remaining 9154 rows were used for the onward pre-processing of the dataset.

Some of the major columns with missing values have been outlined below:

- Review scores for different parameters (e.g., accuracy, cleanliness, etc.) could not be imputed as the missing values for each score appeared to be linked to a customer's rating. Given that the formula used for these columns was unclear, the decision to hide and exclude the identified missing values (989 rows) was made.
- For **bedrooms**, 201 rows of missing values were found. These values could not be imputed as the number of bedrooms can vary and is dependent on other predictor variables, such as "accommodates" or "beds". Hence, the column cannot be imputed based on the column. The missing rows were hidden and excluded from the data set.
- **Bathrooms\_text** was identified to have 4 missing values. As the data set cannot derive which type of bath is included in each listing, the missing values were hidden and excluded.
- **Beds** were found to have 168 missing values. As it cannot be derived from any of the available variables, the missing values were hidden and excluded.

#### Check for N/A values in the data set

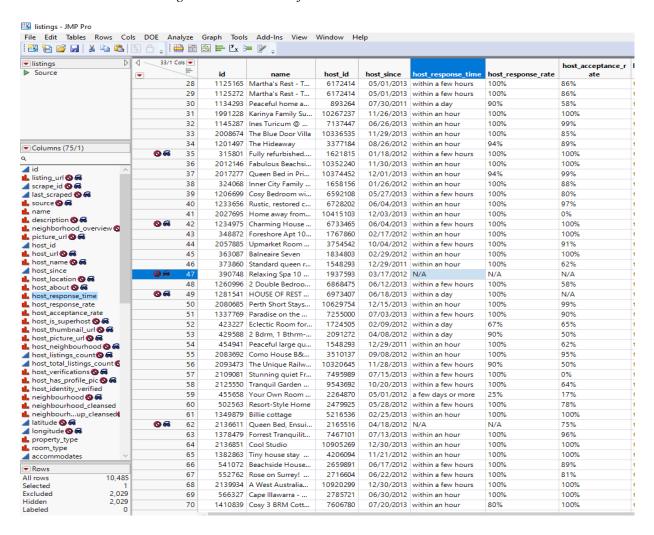
Given that there is a possibility of incorrect values being keyed into the data set, a quick check of the distribution of the columns was undertaken to check for such values. It was found that certain columns have values inputted as N/As.

For the **host\_response\_time** column, N/A rows were removed as that would not help to,, determine the target variable. Further, this column cannot be imputed using the data as the response time is specific to each property. Lastly, there is no reference to determine what is the response rate for a particular property. Hence, all rows with N/As were removed.

A similar treatment was given to columns **response\_rate** and **acceptance\_rate**. In total, these columns have 900 N/A values and cannot be predicted based on other values. Hence, these values were hidden and excluded.

Figure 6 shows the N/A inputs that were found in the data set.

Figure 6: Screenshot of hidden and excluded N/A rows



#### 2.4 Outliers

#### 2.4.1 Multivariate outlier detection:

Upon identifying the rows and columns to be hidden and excluded in the preliminary analysis, the outlier function was used to identify and explore outliers across the variables. The multivariate outlier detection was used with a tail of 0.1 and a quantile interval of 3 to find the outliers as in the figure below.

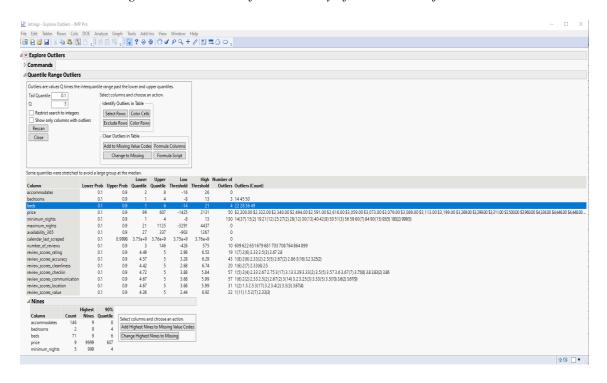
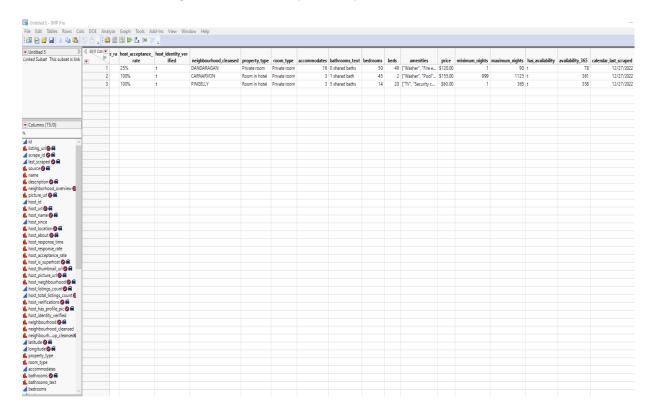


Figure 7: Screenshot of the summary of outliers identified

For **bedrooms**, three rows were identified to feature outliers and two rows were hidden and excluded as a result. Upon further examination, Row 2 was retained as the host could have limited the number of guests to a smaller count to avoid dealing with large parties staying in their listing. Rows 1 and 3 have extreme data points which paint an unrealistic picture. This can be observed in the figure below.

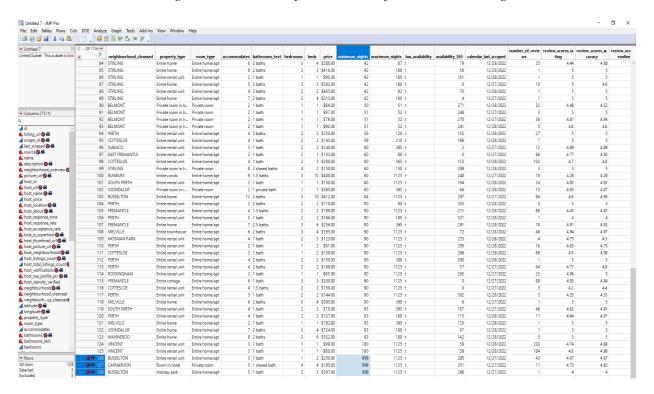
Figure 8: Screenshot of outliers found in 'Bedrooms'



**Beds** were found to have 4 outliers. These values were obscenely high and impractical. Therefore, they were hidden and excluded from the data set.

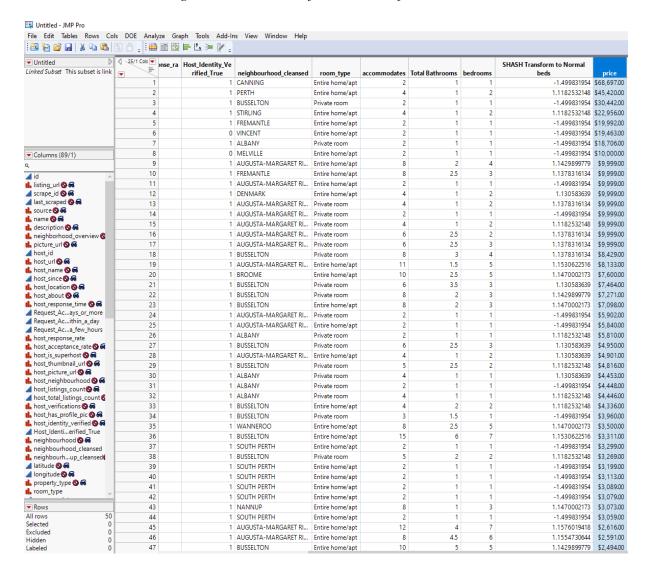
**Minimum\_nights** was found to have 128 outliers. However, only 3 values were considered outliers as they were very high in value (i.e. 999). A minimum stay of 999 days is not practical and is thereby hidden and excluded. No further action was taken on the remaining outliers as it was assumed that the values could have been set with guests staying for a longer minimum period of time.

Figure 9: Screenshot of outliers identified in minimum\_nights



For **Price**, 50 rows were found to be outliers in the data set. Given the target variable has an influence on the data set, the outliers were treated by hiding and excluding them. Extreme values of price indicate the listing can be considered as luxury properties which in turn can be used in another model.

Figure 10: Screenshot of outliers identified in Price



For number\_of\_reviews and review\_scores for different parameters (ex. accuracy, cleanliness, etc), the values have been retained with no change, as the range of these variables can be high or low numbers, as long as it is non-negative and lower than 5, which was verified.

### 2.4.2 Mahalanobis distance:

Additional outliers were identified using the Scatterplot matrix and derived from the Mahalanobis distances.

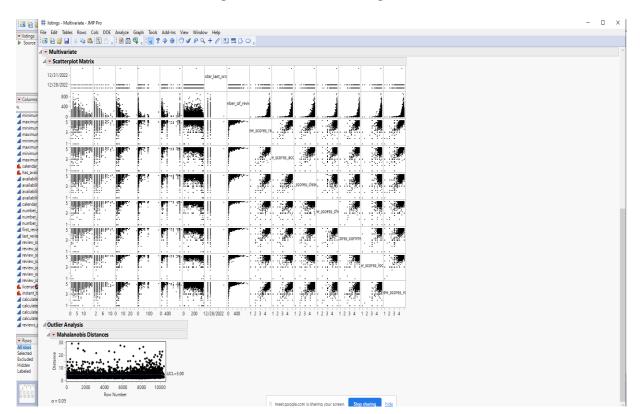
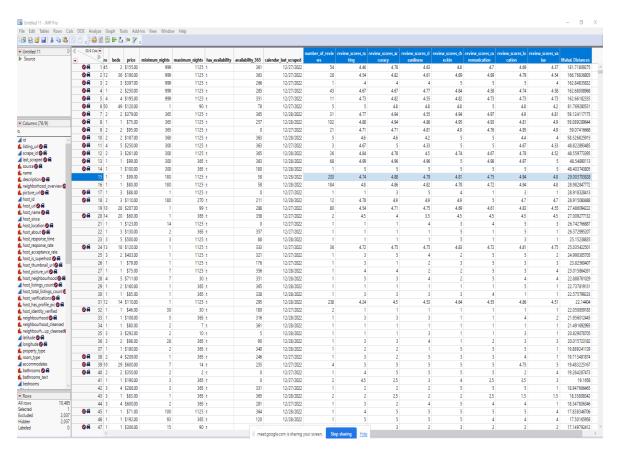


Figure 11: Mahalanobis Diagram

Figure 12: Outliers identified by Mahalanobis Diagram



Looking at the Mahalanobis Distances in descending order, there were many rows found to be away from the Upper Control Limit as can be seen in Figure 12. A decision was made to exclude the first 15 rows, as the distance values were unusually higher than the remaining data.

#### 2.4.3 Transformation of variables:

To further clean the columns with outliers, the histogram distribution was viewed for all continuous variables. **Bedrooms, minimum\_nights, and number\_of\_reviews** were identified to have a lot of outliers. The majority of the data lay beyond the whiskers of the box plot. Therefore, these three variables were transformed to smooth out the data and capture the points, which were currently identified as outliers.

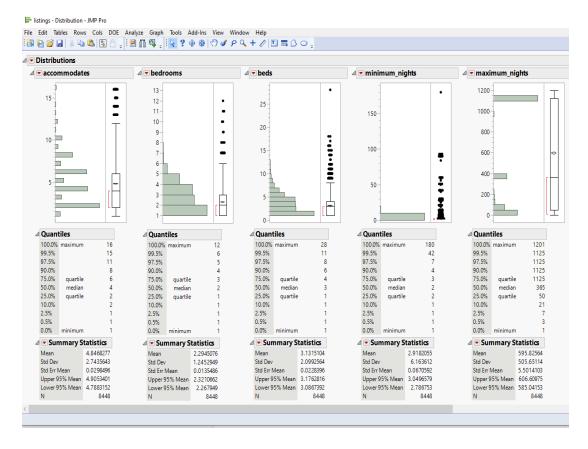


Figure 13: Screenshot of identified variables with outliers prior to transformation

After applying the 'fit all' distribution command on the 3 identified variables, the SHASH distribution was identified to be the best distribution which transforms the data closest to the normal distribution. They were found to have the lowest AICC and the resultant distribution has smoothened the data as can be seen in the figure below.

File Edit Tables Rows Cols DOE Analyze Graph Tools Add-Ins View Window Help Normal minimum\_nights Normal number\_of\_reviews 0.5 0.5 0 0 -0.5 -0.5 0 -1.5 -1.5 ⊿ Ouantiles 100.0% maximum **⊿** Quantiles **⊿** Quantiles 1.1775675 1.1595082 100.0% maximum 4.3753427 1.4454852 100.0% maximum 97.5% 1.1530623 99.5% 97.5% 1.4281265 99.5% 97.5% 2.7364557 1.4041816 1.9178663 75.0% 50.0% quartile 1.1378316 90.0% 75.0% 1.3937158 90.0% 75.0% 1.2394411 1.1305836 1.385833 quartile 0.6888184 quartile 25.0% quartile -1.499832 -0.066127 50.0% 0.0885309 25.0% -0.066127 25.0% -0.533377 quartile quartile 2.5% -1.499832 10.0% -1.624462 10.0% -1.412227 -1.624462 2.5% -2.135838 2.5% minimum -1.499832 0.0% -1.624462 ✓ Summary Statistics 0.0% minimum -1.624462 0.0% minimum -2.135838 Mean 0.3835374 ✓ Summary Statistics ✓ Summary Statistics Std Dev Std Err Mean 1.1892812 0.0559974 Mean 0.0405361 Std Dev 1.105573 Std Dev 0.9932327 Upper 95% Mean 0.4089014 Lower 95% Mean 0.3581734 Std Err Mean Std Err Mean Upper 95% Mean 0.0795762 Lower 95% Mean 0.0324187 Upper 95% Mean 0.061719 Lower 95% Mean 0.0193533

N

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Figure 14: Screenshot of identified variables with outliers post-transformation

#### 2.5 Reducing Dimensionality:

The data set has 7 columns that contain reviews on different parameters: rating, accuracy, cleanliness, checkin, communication, location, and value. These 7 columns have similar information and a decision was made to reduce the dimensionality. Therefore, the Principal Components Analysis technique was applied. Reviewing the eigenvalues, most of the data i.e. 91.29% is present within the first 4 PCAs. So, these 4 PCAs were saved to the dataset. The new columns created in the dataset are **Review\_Scores\_PCA1**, **Review\_Scores\_PCA3**, and **Review\_Scores\_PCA4**.

Eigenvalue 20 40 60 80 15 ✓ Label variables 4.8886 1.0 0.6782 10 0.5283 0.2952 0.2688 review scores location 0.5 8 (% 69.6) 0.2017 69.6) review\_scores\_checkin 0.1393 review scores communication 0 review\_scores\_rating review\_scores\_cleanlines -0.5 -10 -1.0 -15 -40 -20 10 -0.5 Ó Component 1 (69.8 %) Component 1 (69.8 %) Select component Component 1 V Component 2 ∠ Eigenvalues 20 40 60 80 Number Eigenvalue Percent Cum Percent 4.8886 69.837 69.837 0.6782 9.688 79.526 0.5283 7.548 87.073 0.2952 4.217 91.290 95.129 0.2688 3.839 0.2017 2.881 98.011 0.1393 1.989 100.000 ∠ Eigenvectors Prin1 Prin2 Prin3 Prin4 Prin5 Prin7 0.41735 -0.23967 -0.86274 review\_scores\_rating 0.41119 -0.14223 -0.79558 0.38281 review\_scores\_accuracy review\_scores\_cleanliness 0.37726 -0.41814 0.19295 0.64262 0.39406 0.24870 -0.57820 0.35674 0.49507 review scores checkin 0.35210 0.40905 review scores communication 0.38259 -0.42796 -0.34078 -0.64160 0.32048 review\_scores\_location 0.30353 0.69411 0.64658 review\_scores\_value 0.38980 -0.24735 0.15043 -0.58266 0.54938 0.29194 0.19419 **☆**□ ▼ evaluations done

Figure 15: Screenshot of identified variables with outliers post-transformation

Figure 16: Principal Component Analysis

		x > y	<b>-</b>								
▼ listingsteam3_final ▷ ▶ Source	√ 26/0 Cols ▼ Normal	price	SHASH Transform to Normal minimum nights	maximum nights	availability 365	calendar last scraped	SHASH Transform to Normal number of reviews	Reivew Scores PCA 1	Reivew Scores PCA 2	Reivew Scores PCA 3	Reivew Scores PCA
	1 99831954		-1.624462172	730	360	12/27/2022	-0.304766071	-1.443540968	0.162292797	-0.732511153	-0.5691839
	2 30583639	\$149.00	-1.624462172	30	77	12/28/2022	0.9119017915	1,5435908955	-0.149768457	0.1053247586	0.00752210
	3 99831954	\$112.00	1.3858329728	112	187	12/28/2022	-0.47942966	0.0425256434	-0.343048026		-0.152584
	4 99831954	\$75.00	-0.066126824	7	96	12/28/2022	1.6479781839	1.4856976033	-0.139553104	0.0461557562	-0.078977
	5 99831954		-0.066126824	180	323	12/28/2022	0.0688229795	1,4137735323	0.0181383869		-0.105292
	6 99831954		-0.066126824	20	128	12/28/2022	2.0558727315	-3.380573862	-0.320516923		-1.165510
	7 99831954		-1.624462172	1125	347	12/28/2022	1.3712272309	1.126989776	-0.191369493		-0.060165
Columns (89/0)	99831954 <b>©</b> 6 8 99831954		-0.066126824	90	365	12/27/2022	-2.958107805	1.120909770			-0.000103
					290						0.400070
id ^	9 54730644		-0.066126824	27		12/27/2022	1.3260418182	0.9167334256	0.2982538197		-0.198278
listing_url 🛇 🚳	10 30583639		1.3983405912	365	290	12/27/2022	-0.7363321	0.3039764466	-0.471643803	0.4361906246	0.7656599
scrape_id 🛇 🕷	11 02827453		-0.066126824	1125	321	12/28/2022	0.3468523135	0.273936654	0.2453880546		-0.597877
last_scraped 🔇 🚳	12 99831954		1.3983405912	21	298	12/28/2022	0.1262974587	0.7747941061	-1.056649972		0.0203143
source 🛇 🚳	13 99831954	\$57.00	-0.066126824	30	7	12/28/2022	1.5195699552	1.2835396365	-0.084444371	0.3158606318	-0.080381
name 🛇 📾	14 783 16134	\$192.00	1.4142297929	120	324	12/27/2022	-1.696329167	0.0909456981	1.6060594825	-0.5577821	-1.215909
description 🛇 🚳 neighborhood overview 🔇	15 99831954	\$190.00	-1.624462172	1125	336	12/27/2022	1.6193566123	-0.069239743	0.7042672038	0.0159555654	-0.026271
picture url	16 99831954	\$80.00	1.424000168	1125	343	12/28/2022	-0.593178028	1.439832238	0.1340564491	0.2281352549	-0.04956
host_id	17 99831954	\$52.00	-0.066126824	30	13	12/28/2022	2.0159279874	1.2649976697	-0.054162838	0.1397815414	-0.07097
host url 🛇 📾	18 82532148	\$127.00	1.3858329728	1125	5	12/28/2022	0.9036383212	0.3411081614	-0.372384377	-0.139726906	-0.21013
host_name 🛇 🚳	19 99831954	\$53.00	-0.066126824	30	12	12/28/2022	2.0707289417	1.2892941028	-0.143572338	0.1494353405	-0.174852
host_since 🛇 😭	<b>◎</b> € 20 99831954		-1.624462172	999	320	12/28/2022	-2.958107805				
host_location 🛇 🚳	<b>◎</b>		-0.066126824	14	317	12/27/2022	-1.696329167	1.2813240882	0.2866859124	0.0510322133	0.8117834
host_about 🛇 🚜	22 78316134		1.3858329728	1125	331	12/27/2022	0.3738896171	0.5322515379	-0.5701767	-0.397539764	-0.078311
host_response_time 🛇 🕷			1.4041816223	180	363	12/28/2022	-0.927647998				-0.07631
Request_Acays_or_more Request Acithin a day	23 99831954				304			0.9817345061	-0.537835244		
Request_Aca_few_hours	24 99831954		-0.066126824	1125		12/28/2022	0.4624861689	1.244230858	-0.568623233	-0.12213042	-0.025104
host response rate	25 783 16134		1.3937157675	1125	104	12/28/2022	-0.234483249	1.0012001172	0.5536388563	0.0122533498	-0.412643
host acceptance rate &	26 99831954		1.4041816223	180	353	12/27/2022	-0.47942966	-3.535467495	0.0213904196		-0.943415
host_is_superhost 🛇 🙈	27 99831954		1.3983405912	1125	317	12/27/2022	0.4863154359	-1.521459509	-0.257042597	-1.309110896	-1.36375
host_thumbnail_url 🛇 🚳	28 99831954		-0.066126824	14	281	12/28/2022	0.5209990708	0.0465927499	-0.582868976		0.03619
host_picture_url 🛇 🕷	29 99831954	\$100.00	-0.066126824	14	302	12/28/2022	0.2751340357	-0.875721544	-0.424684637	-0.453556559	0.0100178
host_neighbourhood 🛇 😭	30 99831954	\$70.00	-0.066126824	1125	354	12/27/2022	0.027541921	-1.134898633	-1.304890925	-0.780287822	-0.185659
host_listings_count	31 783 16134	\$293.00	1.3983405912	1125	265	12/27/2022	-1.412227025	1.8724793941	-0.08843743	0.2791741107	-0.071852
host_verifications 🛇 🚳	32 30583639	\$224.00	-1.624462172	30	0	12/27/2022	1.4465919801	-0.300672597	0.4187316089	-0.033116535	0.0080742
host has profile pic 🛇 🚳	33 99831954	\$160.00	1.3937157675	1125	271	12/28/2022	0.543470648	-1.483106602	0.1082979576	-0.7910769	-0.196145
host identity verified	34 99831954	\$137.00	-1.624462172	1125	241	12/27/2022	4.3753426969	0.2606175879	-0.518997824	-0.260679002	0.0618273
Host_Identierified_True	<b>⊘</b> € 35	\$75.00	1.3858329728	1125	165	12/28/2022	0.0688229795	1.0749651967	-0.165514251	0.1691869023	0.0360302
neighbourhood 🛇 🚳	36 70002173		-0.066126824	1125	302	12/28/2022	0.4001068035	0.648055593	-0.046727851	0.3413919654	-0.077261
neighbourhood_cleansed	37 99831954		-0.066126824	1125	345	12/28/2022	0.7459258439	-1,772710176	-0.35962993		-0.805916
neighbourhup_cleansed	38 30583639		1.3983405912	60	26	12/28/2022	-1.696329167	1.8724793941	-0.08843743		-0.071852
latitude 🛇 🚳 🔍	39 99831954		-1.624462172	30	20	12/28/2022	-0.927647998	-4.713291274	-1.673889233	0.0291479568	-0.071852
Rows	40 2050200			14							
l rows 10,485			-0.066126824		335	12/27/2022	0.7459258439	-0.373165596	1.1941310521	-0.34750955	-0.815092
lected 0	41 99831954		-0.066126824	1125	48	12/28/2022	-0.824195977	1.8724793941	-0.08843743		-0.07185
xcluded 2,087 lidden 2,087	<b>◎ 6</b> 42 30583639		1.3858329728	30	298	12/28/2022	-2.958107805				
abeled 2,087	43 99831954	\$147.00	-1.624462172	1125	57	12/27/2022	2,9779601374	1.0604539664	-0.13320605	0.0194372296	-0.051413

# 3. Partitioning the data

Post pre-processing the data, cleansing the outliers and missing data values, Figure 17 shows the distribution across different columns.

Flistings - Distribution - JMP Pro File Edit Tables Rows Cols DOE Analyze Graph Tools Add-Ins View Window Help ✓ bedrooms 13 1200 15-12 25 11 • 1000 150 10 20 9 800 10 8 100 • 600 ÷ 6 Į 50 3 200 Ė 2 I **△** Quantiles ∠ Quantiles ∠ Quantiles ⊿ Quantiles ⊿ Quantiles 100.0% maximum 16 100.0% maximum 12 100.0% maximum 28 100.0% maximum 180 100.0% maximum 1201 99.5% 99.5% 99.5% 97.5% 97.5% 97.5% 1125 97.5% 90.0% 90.0% 90.0% 90.0% 1125 90.0% quartile 75.0% 75.0% quartile 75.0% 75.0% 1125 75.0% median 50.0% 50.0% 50.0% 50.0% median 50.0% 365 25.0% quartile 25.0% 25.0% 25.0% 25.0% 50 10.0% 10.0% 10.0% 10.0% 10.0% 21 2.5% 2.5% 2.5% 2.5% 2.5% 0.5% 0.5% 0.5% 0.5% 0.5% 0.0% 0.0% minimum 0.0% minimum 0.0% minimum minimum 0.0% minimum ✓ Summary Statistics ✓ Summary Statistics 4.8468277 3.1315104 2.9182055 595.82564 Mean 2.2945076 Mean Mean Mean Mean Std Dev Std Dev 6.163612 Std Dev 505.65114 1.2452949 Std Dev Std Dev 5.5014103 Std Err Mean 0.0298496 Std Err Mean 0.0135486 Std Err Mean Std Err Mean Std Err Mean Upper 95% Mean 4.9053401 Upper 95% Mean 2.3210662 Upper 95% Mean 3.1762816 Upper 95% Mean 3.0496579 Upper 95% Mean 606.60975 Lower 95% Mean 4.7883152 Lower 95% Mean 2.786753 Lower 95% Mean Lower 95% Mean 3.0867392

Figure 17: Screenshot of identified variables with outliers prior to transformation

With the refined dataset, data can now be partitioned into 3 sets: Training, Validation, and test data. This will help to build models, cross-check, and validate the resultant model with the validation data set and finally, predict based on the test data.

The following split between the 3 partitions has been executed:

Training: 60%Validation: 20%Test: 20%

Figure 18 shows the final data set, which was created upon cleaning the data

<u>File Edit Tables Rows Cols DOE</u> 🚇 😜 🥳 🔒 🔉 😘 i 🗎 🖺 🖺 🗠 🏏 26/1 Cols Vormal SHASH Transform to Normal listingsteam3\_final SHASH Transform to Normal 
 nights
 availability\_365
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 730
 360
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 ber\_of\_reviews -0.30476607 v\_Scores\_PCA\_1 Reivew\_Scores\_PCA\_2 Reive Scores PCA 3 Reivew Scores PCA 4 Mahal, Distances Validation price 1 99831954 \$94.00 ights -1.624462172 2.9012817379 Traini 2.1346995203 Traini 2 30583639 \$149.00 1.5435908955 0.105324758 99831954 \$112.00 1.3858329728 12/28/2022 -0.47942966 0.0425256434 -0.343048026 -0.50886136 -0.15258441 2.1231811947 Training 1.6479781839 0.0688229795 4 99831954 \$75.00 -0.066126824 4856976033 -0.139553104 0.046155756 -0.078977254 2.7827847495 Test 6 99831954 \$70.00 -0.066126824 128 12/28/2022 2.0558727315 -3.380573862 -0.320516923 -0.501667594 -1.165510294 4.6385162872 Training ▼ Columns (89/1) 7 00831054 \$85 N -1 624462172 1125 12/28/2022 1.3712272309 1.126989776 -0.191369493 0.1173629084 -0.060165137 2.625621173 Test 8 99831954 \$195.00 -2.958107805 id

listing\_url 

scrape\_id 

scrape\_id 

scrape\_id 

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last\_scraped 

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laname 

laname 0.9167334256 0.2982538197 0.018804594 -0.198278122 9 54730644 \$277.00 -0.066126824 12/27/2022 1.3260418182 5.6910888901 Training 10 30583639 \$344.00 1 3983405912 -0.7363321 0.3039764466 0.436190624 2 971426969 Traini 12 99831954 \$49.00 1.3983405912 298 12/28/2022 0.1262974587 0.7747941061 -1.056649972 -0.68706007 0.0203143259 2.6627446857 Validation 3.0301942432 Training 13 99831954 \$57.00 -0.066126824 12/28/2022 1.5195699552 1.2835396365 -0.08444437 0.3158606318 -0.08038181 1.4142297929 15 99831954 \$190.00 -1.624462172 1125 336 12/27/2022 1.6193566123 -0.069239743 0.7042672038 0.015955565 -0.026271262 3.326449673 Training 16 99831954 \$80.00 1.424000168 1125 343 12/28/2022 .n 593178028 1,439832238 0.1340564491 0.2281352549 -0.049561095 5.3630072063 Training 2.0159279874 -0.070973686 host url nost\_iname () 66 nost\_since () 66 nost\_location () 66 nost\_about () 6 18 82532148 \$127.00 1.3858329728 12/28/2022 0.9036383212 0.3411081614 0.37238437 -0.139726906 -0.210132038 2.4634413763 Training 19 99831954 \$53.00 -0.066126824 12/28/2022 2.0707289417 1.2892941028 -0.143572338 0.1494353405 -0.174852251 3.9068294442 Valida 320 317 0.0510322133 **⊚** € 21 99831954 \$199.00 -0.066126824 12/27/2022 -1.696329167 1.2813240882 0.2866859124 0.8117834579 ost response time 🛇 🕏 3.9458950235 Training 22 78316134 \$550,00 1.3858329728 1125 12/27/2022 0.3738896171 0.5322515379 -0.5701767 -0.397539764 -0.078311848 -0.537835244 uest\_Ac...a\_few\_hours 24 99831954 \$65.00 -0.066126824 1125 12/28/2022 0.4624861689 1.244230858 -0.568623233 -0.12213042 -0.025104134 2.2668522707 Validatio 25 78316134 \$330.00 1 3937157675 1125 12/28/2022 -0.234483249 1.0012001172 0.5536388563 0.012253349 -0.41264366 3 6246109627 Validation ost acceptance rate of a cost, is superhost of a cost, thumbnail, url of a cost, thumbnail, url of a cost, picture, url of a cost, listings, count of cost, url fictions of a cost, pas, profile, pic of a cost, pas, profile, pic of a cost, dentity, verified of lost, Identiu-erified True 27 99831954 \$45.00 1.3983405912 0.4863154359 -1.521459509 -0.25704259 -1.30911089 -1.36375833 4.1088880479 Training 28 99831954 \$95.00 -0.066126824 12/28/2022 0.5209990708 0.0465927499 -0.582868976 -0.577839129 0.03619636 2.5020665133 Train -0.45355655 354 30 99831954 \$70.00 -0.066126824 1125 0.027541921 -1.304890925 -0.780287822 -0.185659418 3.9740793112 Training 31 78316134 \$293.00 1.3983405912 1125 265 12/27/2022 -1,412227025 1.8724793941 -0.08843743 0.2791741107 -0.071852428 2.5768497707 Traini 1.3937157675 33 99831954 \$160.00 12/28/2022 0.543470648 -1.483106602 0.1082979576 -0.7910769 -0.196145983 2.5394125513 Test 34 99831954 \$137.00 -1.624462172 1125 4.3753426969 0.2606175879 -0.518997824 -0.260679007 0.0618273988 11.668382199 Test

Figure 18: Screenshot of the partition done on the pre-processed data set

### 4. Conclusion

2,087 2,087 0 36 70002173 \$367.00

37 99831954 \$77.00

39 99831954 \$50.00

40 30583639 \$168.00

-0.066126824

-0.066126824

-1.624462172

-0.066126824

1125

1125

26 223

The initial Western Australia data set contained 10485 rows x 75 columns. Through pre-processing, the data set has been cleansed and reduced to 8398 rows x 26 columns. The same has been achieved by reducing data dimensionality, excluding the missing values, addressing all N/As in the data set, smoothening the outliers (*transformations*), which will have an impact on determining the target variable, and lastly addressing outliers, if any (*through Mahalanobis / Multivariate*, etc.). Now, the data set is to be modeled and run through the partitions to test on the test data.

12/28/2022

12/28/2022

12/28/2022

0.4001068035

0.7459258439

-0.927647998

0.7459258439

0.648055593

-1.772710176

-0.373165596

Disclaimer: The work contained and presented here is our work and our work alone.

\* \* \* \* \*

0.3413919654

-0.335141288

.279174110

0.029147956

-0.046727851

-0.35962993

1.673889233

1.1941310521

2.6783417555 Training

2.8583579018 Training

4.2763748106 Test

-0.077261306

-0.805916276

-0.685349273