

Notebook 1

Code ▾

Nguyen Tran

Data: LondonAirBNB (https://www.kaggle.com/datasets/jinxzed/londonairbnb?select=Uphaar_listing.csv)

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```
library(readr)
library(caret)
```

```
Loading required package: ggplot2
Loading required package: lattice
Registered S3 method overwritten by 'data.table':
  method      from
print.data.table
```

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```
library(tree)
```

Cleaning Data

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```
df <- read.csv("Uphaar_listing.csv")
head(df)
```

	X	price_x	experiences_offered	host_is_superhost	host_listings_count	host_has_profi
	<int>	<int>	<chr>	<chr>	<dbl>	<chr>
1	0	88	family	f	3	t
2	1	65	business	f	4	t
3	2	100	romantic	f	1	t
4	3	300	none	f	18	t
5	4	150	business	f	3	t
6	5	29	none	t	3	t

6 rows | 1-7 of 42 columns

Dropping 'X', "city", "host_location", "street", "host_since", "is_business_travel_ready" and "requires_license" and "id"

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```
exclude <- names(df) %in% c("X", "city", "host_location", "street", "host_since", "is_business_travel_ready", "requires_license", "id")  
df <- df[!exclude]
```

There is a need for transforming and cleaning this dataset. There are 41 columns, going through and figuring out which ones are factors

[Hide](#)

```
convert.to.numeric <- function(vector) {  
  vector <- as.numeric(as.factor(vector))-1  
}
```

Looking at the summary of dataframe to consider which ones need to be factorized.

[Hide](#)

```
summary(df)
```

```

price_x      experiences_offered host_is_superhost host_listings_count host_has_profile_p
ic host_identity_verified is_location_exact property_type
Min. : 0.0 Length:83709 Length:83709 Min. : 0.00 Length:83709
Length:83709 Length:83709 Length:83709
1st Qu.: 46.0 Class :character Class :character 1st Qu.: 1.00 Class :character
Class :character Class :character Class :character
Median : 81.0 Mode :character Mode :character Median : 1.00 Mode :character
Mode :character Mode :character Mode :character
Mean : 127.5 Mean : 25.97
3rd Qu.: 139.0 3rd Qu.: 4.00
Max. :19970.0 Max. :3142.00
NA's :11

room_type_y      accommodates      bathrooms      bedrooms      beds      bed_type
e      guests_included      extra_people      minimum_nights_y
Length:83709 Min. : 1.000 Min. : 0.000 Min. : 0.000 Min. : 0.00 Length:83
709 Min. : 1.000 Min. : 0.000 Min. : 1.000
Class :character 1st Qu.: 2.000 1st Qu.: 1.000 1st Qu.: 1.000 1st Qu.: 1.00 Class :ch
aracter 1st Qu.: 1.000 1st Qu.: 0.000 1st Qu.: 1.000
Mode :character Median : 2.000 Median : 1.000 Median : 1.000 Median : 1.00 Mode :ch
aracter Median : 1.000 Median : 0.000 Median : 2.000
Mean : 3.173 Mean : 1.308 Mean : 1.409 Mean : 1.74
Mean : 1.607 Mean : 7.393 Mean : 4.632
3rd Qu.: 4.000 3rd Qu.: 1.500 3rd Qu.: 2.000 3rd Qu.: 2.00
3rd Qu.: 2.000 3rd Qu.: 10.000 3rd Qu.: 3.000
Max. :32.000 Max. :35.000 Max. :50.000 Max. :50.00
Max. :46.000 Max. :247.000 Max. :1125.000
NA's :133 NA's :164 NA's :984
maximum_nights      availability_30      availability_60      availability_90      availability_365_y      number_
of_reviews_y      number_of_reviews_ltm      instant_bookable      cancellation_policy
Min. : 1 Min. : 0.000 Min. : 0.00 Min. : 0.00 Min. : 0.0 Min. : 0.0 Min. :
: 0.0 Min. : 0.00 Length:83709 Length:83709
1st Qu.: 31 1st Qu.: 0.000 1st Qu.: 0.00 1st Qu.: 0.00 1st Qu.: 0.0 1st Q
u.: 0.0 1st Qu.: 0.00 Class :character Class :character
Median : 1125 Median : 0.000 Median : 7.00 Median :16.00 Median : 69.0 Median
: 4.0 Median : 1.00 Mode :character Mode :character
Mean : 19156 Mean : 4.907 Mean :18.48 Mean :32.51 Mean :117.4 Mean
: 16.9 Mean : 4.75
3rd Qu.: 1125 3rd Qu.:11.000 3rd Qu.:40.00 3rd Qu.:70.00 3rd Qu.:249.0 3rd Q
u.: 17.0 3rd Qu.: 5.00
Max. :999999999 Max. :30.000 Max. :60.00 Max. :90.00 Max. :365.0 Max.
:775.0 Max. :334.00

require_guest_profile_picture      require_guest_phone_verification      calculated_host_listings_count_y
calculated_host_listings_count_entire_homes
Length:83709 Length:83709 Min. : 1.00
Min. : 0.00
Class :character Class :character 1st Qu.: 1.00
1st Qu.: 0.00
Mode :character Mode :character Median : 1.00
Median : 1.00
Mean : 16.76 Mean : 19.59

```

3rd Qu.: 2.00	3rd Qu.: 4.00
Max. :919.00	Max. :921.00

calculated_host_listings_count_private_rooms	calculated_host_listings_count_shared_rooms	security_deposit	cleaning_fee
Min. : 0.000	Min. : 0.00000		
h:83709	Length:83709		
1st Qu.: 0.000	1st Qu.: 0.00000		
:character	Class :character		
Median : 1.000	Median : 0.00000		
:character	Mode :character		
Mean : 2.166	Mean : 0.05133		
3rd Qu.: 1.000	3rd Qu.: 0.00000		
Max. :223.000	Max. :18.00000		

Encoding the following variables for data exploration

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```
encodables <- c("experiences_offered","host_is_superhost","host_has_profile_pic","host_identity_verified",
               "is_location_exact","property_type","room_type_y","bed_type",
               "instant_bookable","cancellation_policy","require_guest_profile_picture",
               "require_guest_phone_verification")

for(col in encodables) {
  df[,col] <- convert.to.numeric(df[,col])
}
```

security_deposit and cleaning_fee are characters currently, representing currency, parsing the numbers out of the column.

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```
df$security_deposit <- parse_number(df$security_deposit)
df$cleaning_fee <- parse_number(df$cleaning_fee)
```

Exploring NA's in data to see if these can get replaced by the mean of the column.

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```
df$host_listings_count[is.na(df$host_listings_count)] <- mean(df$host_listings_count,na.rm=TRUE)
df$bathrooms[is.na(df$bathrooms)] <- mean(df$bathrooms,na.rm=TRUE)
df$bedrooms[is.na(df$bedrooms)] <- mean(df$bedrooms,na.rm=TRUE)
df$beds[is.na(df$beds)] <- mean(df$beds,na.rm=TRUE)
df$security_deposit[is.na(df$security_deposit)] <- mean(df$security_deposit,na.rm=TRUE)
df$cleaning_fee[is.na(df$cleaning_fee)] <- mean(df$cleaning_fee,na.rm=TRUE)
```

head(df)

	price_x <int>	experiences_offered <dbl>	host_is_superhost <dbl>	host_listings_count <dbl>	host_has_prof
1	88	1	1	3	
2	65	0	1	4	
3	100	3	1	1	
4	300	2	1	18	
5	150	0	1	3	
6	29	2	2	3	

6 rows | 1-6 of 34 columns

Train / Test Split

Splitting dataset into the training and test (80%/20%)

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set.seed(123) # reproducibility
i <- sample(1:nrow(df), nrow(df)*.80, replace=FALSE)
train <- df[i,]
test <- df[-i,]

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train

	price_x <int>	experiences_offered <dbl>	host_is_superhost <dbl>	host_listings_count <dbl>	host_has_prof
51663	70	2	1	1	
57870	121	2	2	1	
2986	32	2	1	1	
29925	45	2	1	1	
68293	100	2	1	1	
62555	150	2	1	11	
45404	35	2	2	1	
65161	115	2	1	3	

	price_x <int>	experiences_offered <dbl>	host_is_superhost <dbl>	host_listings_count <dbl>	host_has
46435	150	2	2	1	
9642	95	2	1	7	

1-10 of 66,967 rows | 1-6 of 34 columns

Previous123456...100Next



Exploring Data

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```
suppressWarnings(cor(train))
```



		price_x	experiences_offered	host_is_superhost
host_listings_count	host_has_profile_pic	host_identity_verified		
price_x		1.0000000000	-0.0067421142	-0.023725338
0.1663967363	-0.0023581398	-0.042989922		
experiences_offered		-0.0067421142	1.0000000000	-0.013591841
0.0037882267	0.0001125073	-0.022259940		
host_is_superhost		-0.0237253378	-0.0135918415	1.0000000000
-0.0610369289	0.0254434617	0.075619634		
host_listings_count		0.1663967363	0.0037882267	-0.061036929
1.0000000000	0.0094878746	-0.077264590		
host_has_profile_pic		-0.0023581398	0.0001125073	0.025443462
0.0094878746	1.0000000000	0.049873378		
host_identity_verified		-0.0429899216	-0.0222599405	0.075619634
-0.0772645895	0.0498733775	1.0000000000		
is_location_exact		-0.0018161181	-0.0065166529	0.041912977
-0.0837713851	-0.0032147661	0.014943387		
property_type		0.0183285892	-0.0163809924	0.057880494
0.0154471708	-0.0042187036	-0.023645723		
room_type_y		-0.1878753447	0.0216824568	0.053709304
-0.1088322567	-0.0211390195	-0.020696571		
accommodates		0.2626851733	-0.0302762599	-0.032092339
0.1451887326	0.0173942705	0.007558997		
bathrooms		0.1997602983	-0.0192410770	-0.019169887
0.1031078657	0.0051406485	-0.007301477		
bedrooms		0.2263977986	-0.0316807318	-0.021047035
0.1470528796	0.0089424772	0.023539282		
beds		0.2050605423	-0.0323056777	-0.028332863
0.1360640398	0.0096262958	-0.004299932		
bed_type		0.0167297082	-0.0108997642	-0.004371366
0.0118860953	-0.0045811445	-0.028053682		
guests_included		0.1873605144	-0.0234546178	-0.006942722
0.2358262620	0.0106906427	-0.012507487		
extra_people		0.0311587403	-0.0169632430	0.073398571
-0.0510251666	0.0021022809	0.029930218		
minimum_nights_y		0.0366950233	-0.0069164681	-0.018447265
-0.0063416174	0.0054119706	0.022739744		
maximum_nights		-0.0008063852	0.0001244833	-0.001693518
-0.0005758112	0.0002478717	0.005226243		
availability_30		0.0459545355	-0.0155221317	0.077522761
-0.0440034506	-0.0146190805	-0.051512161		
availability_60		0.0640443467	-0.0163488979	0.074222302
-0.0409143392	-0.0123863797	-0.054309446		
availability_90		0.0665370890	-0.0158674982	0.077651153
-0.0412637357	-0.0110135333	-0.056058465		
availability_365_y		0.0900999948	-0.0249889188	0.052345237
-0.0294376932	-0.0091428997	-0.040989984		
number_of_reviews_y		-0.0548052080	-0.0115775966	0.281531438
-0.0539368607	0.0179156794	0.117802987		
number_of_reviews_ltm		-0.0450640008	0.0041920273	0.311983592
-0.0513843938	0.0193122782	0.002977735		
instant_bookable		0.0592326998	0.0143609603	-0.049372508
0.1345013216	-0.0108195040	-0.158831547		

cancellation_policy				0.0579761799	-0.0261248513	0.087062193
0.1351874698	0.0214099724			0.069122885		
require_guest_profile_picture				-0.0104521136	-0.0392836555	0.045209655
-0.0124641879	0.0061048773			0.062099885		
require_guest_phone_verification				0.0014957474	-0.0326306459	0.032509003
0.0136455316	0.0078166858			0.080090033		
calculated_host_listings_count_y				0.1502532651	0.0036273812	-0.063426146
0.8947683689	0.0107833787			-0.074109121		
				is_location_exact	property_type	room_type_y accom
modates	bathrooms	bedrooms	beds	bed_type	guests_included	
price_x				-0.0018161181	0.018328589	-0.187875345 0.262
6851733	0.199760298	0.226397799	0.205060542	1.672971e-02	0.187360514	
experiences_offered				-0.0065166529	-0.016380992	0.021682457 -0.030
2762599	-0.019241077	-0.031680732	-0.032305678	-1.089976e-02	-0.023454618	
host_is_superhost				0.0419129775	0.057880494	0.053709304 -0.032
0923392	-0.019169887	-0.021047035	-0.028332863	-4.371366e-03	-0.006942722	
host_listings_count				-0.0837713851	0.015447171	-0.108832257 0.145
1887326	0.103107866	0.147052880	0.136064040	1.188610e-02	0.235826262	
host_has_profile_pic				-0.0032147661	-0.004218704	-0.021139019 0.017
3942705	0.005140649	0.008942477	0.009626296	-4.581145e-03	0.010690643	
host_identity_verified				0.0149433874	-0.023645723	-0.020696571 0.007
5589971	-0.007301477	0.023539282	-0.004299932	-2.805368e-02	-0.012507487	
is_location_exact				1.0000000000	0.073300680	-0.017510945 0.013
0582370	0.029325593	0.023477754	0.013535443	-9.019503e-04	-0.018094455	
property_type				0.0733006803	1.000000000	0.217591182 0.065
5431292	0.205495227	0.163531988	0.121286614	2.697069e-03	0.026219218	
room_type_y				-0.0175109451	0.217591182	1.000000000 -0.537
6147146	-0.154993532	-0.352460682	-0.372537508	-3.539304e-02	-0.303098020	
accommodates				0.0130582370	0.065543129	-0.537614715 1.000
0000000	0.486517088	0.751759714	0.794293606	3.061522e-02	0.529051084	
bathrooms				0.0293255929	0.205495227	-0.154993532 0.486
5170883	1.000000000	0.593577019	0.495174168	2.272909e-02	0.274158751	
bedrooms				0.0234777542	0.163531988	-0.352460682 0.751
7597142	0.593577019	1.000000000	0.722143064	3.219163e-02	0.416555791	
beds				0.0135354433	0.121286614	-0.372537508 0.794
2936057	0.495174168	0.722143064	1.000000000	3.283102e-02	0.442031176	
bed_type				-0.0009019503	0.002697069	-0.035393038 0.030
6152239	0.022729087	0.032191634	0.032831022	1.000000e+00	0.021862918	
guests_included				-0.0180944550	0.026219218	-0.303098020 0.529
0510842	0.274158751	0.416555791	0.442031176	2.186292e-02	1.000000000	
extra_people				-0.0052580198	0.019859595	-0.042687446 0.116
1533701	0.051567523	0.072810208	0.088622615	2.338228e-03	0.268640298	
minimum_nights_y				-0.0058476800	-0.013168057	-0.029519958 -0.000
3416719	0.006846080	0.009880198	0.002098955	1.632088e-03	-0.003270478	
maximum_nights				0.0025683656	0.005012831	0.004435782 -0.004
2590723	-0.001901828	-0.001730709	-0.005275439	3.000275e-04	-0.001751121	
availability_30				-0.0045418263	0.061236012	0.094262092 -0.008
2529385	0.009033050	-0.040767020	-0.008478052	4.823172e-03	0.009825278	
availability_60				-0.0039009298	0.060674046	0.073750964 0.014
2044100	0.023713339	-0.027597698	0.011346975	6.859145e-03	0.025689867	
availability_90				-0.0025626508	0.061408211	0.072191786 0.015
8394651	0.023884377	-0.030254131	0.012421278	7.013615e-03	0.026729806	

availability_365_y				0.0034213109	0.068719825	0.018553924	0.070
3291183	0.041432897	-0.005018585	0.054693812	1.617300e-03	0.066291758		
number_of_reviews_y				0.0299768328	0.018618854	0.090696747	-0.032
6508379	-0.037129423	-0.061823189	-0.033886359	-2.561217e-02	0.013690664		
number_of_reviews_ltm				0.0282028406	0.016109059	0.079983806	-0.015
5876843	-0.027195271	-0.062819675	-0.014822458	1.495725e-03	0.022287644		
instant_bookable				-0.0151079032	0.010826657	0.025161029	0.024
8734292	-0.004002623	-0.047745818	0.016869953	3.694010e-02	0.038473629		
cancellation_policy				-0.0097427025	-0.055946095	-0.239535227	0.205
1923609	0.071748526	0.123627874	0.144365997	1.207120e-02	0.200026045		
require_guest_profile_picture				0.0235835745	0.016214402	0.008370418	-0.011
0124247	-0.007716830	-0.007968078	-0.009644394	-8.713657e-03	-0.003792267		
require_guest_phone_verification				0.0171529920	0.004338892	-0.024856160	0.023
1053491	0.001157264	0.012310992	0.014271905	-3.282141e-03	0.054040220		
calculated_host_listings_count_y				-0.0863272705	0.023986955	-0.122102038	0.167
5737308	0.118659835	0.167682069	0.152632531	1.286130e-02	0.265446265		
extra_people minimum_nights_y maximum_nights availa							
bility_30	availability_60	availability_90	availability_365_y				
price_x				0.031158740	0.0366950233	-8.063852e-04	0.
045954535	0.064044347	0.066537089		0.090099995			
experiences_offered				-0.016963243	-0.0069164681	1.244833e-04	-0.
015522132	-0.016348898	-0.015867498		-0.024988919			
host_is_superhost				0.073398571	-0.0184472653	-1.693518e-03	0.
077522761	0.074222302	0.077651153		0.052345237			
host_listings_count				-0.051025167	-0.0063416174	-5.758112e-04	-0.
044003451	-0.040914339	-0.041263736		-0.029437693			
host_has_profile_pic				0.002102281	0.0054119706	2.478717e-04	-0.
014619081	-0.012386380	-0.011013533		-0.009142900			
host_identity_verified				0.029930218	0.0227397441	5.226243e-03	-0.
051512161	-0.054309446	-0.056058465		-0.040989984			
is_location_exact				-0.005258020	-0.0058476800	2.568366e-03	-0.
004541826	-0.003900930	-0.002562651		0.003421311			
property_type				0.019859595	-0.0131680567	5.012831e-03	0.
061236012	0.060674046	0.061408211		0.068719825			
room_type_y				-0.042687446	-0.0295199576	4.435782e-03	0.
094262092	0.073750964	0.072191786		0.018553924			
accommodates				0.116153370	-0.0003416719	-4.259072e-03	-0.
008252939	0.014204410	0.015839465		0.070329118			
bathrooms				0.051567523	0.0068460801	-1.901828e-03	0.
009033050	0.023713339	0.023884377		0.041432897			
bedrooms				0.072810208	0.0098801978	-1.730709e-03	-0.
040767020	-0.027597698	-0.030254131		-0.005018585			
beds				0.088622615	0.0020989551	-5.275439e-03	-0.
008478052	0.011346975	0.012421278		0.054693812			
bed_type				0.002338228	0.0016320881	3.000275e-04	0.
004823172	0.006859145	0.007013615		0.001617300			
guests_included				0.268640298	-0.0032704778	-1.751121e-03	0.
009825278	0.025689867	0.026729806		0.066291758			
extra_people				1.000000000	-0.0062577761	-2.066431e-03	0.
079421716	0.086174072	0.087381977		0.084788485			
minimum_nights_y				-0.006257776	1.000000000	-7.061842e-04	0.
027565507	0.033597746	0.034279107		0.041923314			

maximum_nights			-0.002066431	-0.0007061842	1.000000e+00	0.
005698776	0.004900163	0.004738915	0.006478851			
availability_30			0.079421716	0.0275655065	5.698776e-03	1.
000000000	0.954985675	0.925042654	0.662534897			
availability_60			0.086174072	0.0335977457	4.900163e-03	0.
954985675	1.000000000	0.986993963	0.716551724			
availability_90			0.087381977	0.0342791073	4.738915e-03	0.
925042654	0.986993963	1.000000000	0.743495929			
availability_365_y			0.084788485	0.0419233143	6.478851e-03	0.
662534897	0.716551724	0.743495929	1.000000000			
number_of_reviews_y			0.067322565	-0.0301911036	1.396820e-02	0.
076568736	0.074005708	0.079618657	0.107476559			
number_of_reviews_ltm			0.038047263	-0.0462768680	7.263806e-03	0.
110842422	0.112603996	0.120466347	0.111328731			
instant_bookable			-0.034836777	-0.0168357537	4.658295e-03	0.
039961891	0.045553073	0.047545922	0.068907068			
cancellation_policy			0.117390401	0.0127942271	3.874040e-03	0.
049093767	0.063751498	0.067920615	0.106420077			
require_guest_profile_picture			0.026003088	0.0070092653	-3.999833e-04	0.
035470324	0.037767741	0.038265740	0.051954247			
require_guest_phone_verification			0.041433971	0.0030292926	-5.045750e-04	0.
033773665	0.039225006	0.041211139	0.065706532			
calculated_host_listings_count_y			-0.054782179	-0.0049880692	-6.375886e-04	-0.
046796651	-0.043815834	-0.045836557	-0.024873745			
number_of_reviews_y number_of_reviews_ltm instant_b						
ookable	cancellation_policy	require_guest_profile_picture				
price_x			-0.0548052080	-0.045064001	0.05	
9232700	0.057976180		-0.0104521136			
experiences_offered			-0.0115775966	0.004192027	0.01	
4360960	-0.026124851		-0.0392836555			
host_is_superhost			0.2815314380	0.311983592	-0.04	
9372508	0.087062193		0.0452096549			
host_listings_count			-0.0539368607	-0.051384394	0.13	
4501322	0.135187470		-0.0124641879			
host_has_profile_pic			0.0179156794	0.019312278	-0.01	
0819504	0.021409972		0.0061048773			
host_identity_verified			0.1178029867	0.002977735	-0.15	
8831547	0.069122885		0.0620998852			
is_location_exact			0.0299768328	0.028202841	-0.01	
5107903	-0.009742702		0.0235835745			
property_type			0.0186188542	0.016109059	0.01	
0826657	-0.055946095		0.0162144022			
room_type_y			0.0906967471	0.079983806	0.02	
5161029	-0.239535227		0.0083704177			
accommodates			-0.0326508379	-0.015587684	0.02	
4873429	0.205192361		-0.0110124247			
bathrooms			-0.0371294230	-0.027195271	-0.00	
4002623	0.071748526		-0.0077168301			
bedrooms			-0.0618231895	-0.062819675	-0.04	
7745818	0.123627874		-0.0079680784			
beds			-0.0338863591	-0.014822458	0.01	
6869953	0.144365997		-0.0096443943			

bed_type		-0.0256121679	0.001495725	0.03
6940096	0.012071199	-0.0087136571		
guests_included		0.0136906635	0.022287644	0.03
8473629	0.200026045	-0.0037922665		
extra_people		0.0673225654	0.038047263	-0.03
4836777	0.117390401	0.0260030883		
minimum_nights_y		-0.0301911036	-0.046276868	-0.01
6835754	0.012794227	0.0070092653		
maximum_nights		0.0139681972	0.007263806	0.00
4658295	0.003874040	-0.0003999833		
availability_30		0.0765687355	0.110842422	0.03
9961891	0.049093767	0.0354703240		
availability_60		0.0740057085	0.112603996	0.04
5553073	0.063751498	0.0377677412		
availability_90		0.0796186571	0.120466347	0.04
7545922	0.067920615	0.0382657402		
availability_365_y		0.1074765591	0.111328731	0.06
8907068	0.106420077	0.0519542474		
number_of_reviews_y		1.0000000000	0.692967926	0.00
6215966	0.152924287	0.1114144316		
number_of_reviews_ltm		0.6929679255	1.0000000000	0.09
4789048	0.138186612	0.0222186523		
instant_bookable		0.0062159658	0.094789048	1.00
0000000	-0.030585996	-0.0375870505		
cancellation_policy		0.1529242872	0.138186612	-0.03
0585996	1.0000000000	0.0441067037		
require_guest_profile_picture		0.1114144316	0.022218652	-0.03
7587050	0.044106704	1.0000000000		
require_guest_phone_verification		0.1078535126	0.022100630	-0.01
0574076	0.053326267	0.6651862210		
calculated_host_listings_count_y		-0.0557015529	-0.051839822	0.13
2807680	0.138198851	-0.0133733088		
require_guest_phone_verification calculated_host_listings_count_y calculated_host_listings_count_entire_homes				
price_x			0.001495747	
0.1502532651		0.1547197068		
experiences_offered			-0.032630646	
0.0036273812		0.0031979067		
host_is_superhost			0.032509003	
-0.0634261458		-0.0581317082		
host_listings_count			0.013645532	
0.8947683689		0.8879796742		
host_has_profile_pic			0.007816686	
0.0107833787		0.0102653869		
host_identity_verified			0.080090033	
-0.0741091212		-0.0656839372		
is_location_exact			0.017152992	
-0.0863272705		-0.0771443039		
property_type			0.004338892	
0.0239869549		0.0187570718		
room_type_y			-0.024856160	
-0.1221020383		-0.1409824965		

accommodates		0.023105349
0.1675737308	0.1727063060	
bathrooms		0.001157264
0.1186598350	0.1232185973	
bedrooms		0.012310992
0.1676820689	0.1777484996	
beds		0.014271905
0.1526325309	0.1554511478	
bed_type		-0.003282141
0.0128612953	0.0117577245	
guests_included		0.054040220
0.2654462649	0.2671990731	
extra_people		0.041433971
-0.0547821787	-0.0558449391	
minimum_nights_y		0.003029293
-0.0049880692	-0.0044028222	
maximum_nights		-0.000504575
-0.0006375886	-0.0006556852	
availability_30		0.033773665
-0.0467966512	-0.0620408910	
availability_60		0.039225006
-0.0438158341	-0.0611169691	
availability_90		0.041211139
-0.0458365570	-0.0637956369	
availability_365_y		0.065706532
-0.0248737450	-0.0485108688	
number_of_reviews_y		0.107853513
-0.0557015529	-0.0551362250	
number_of_reviews_ltm		0.022100630
-0.0518398220	-0.0536698062	
instant_bookable		-0.010574076
0.1328076803	0.1295002736	
cancellation_policy		0.053326267
0.1381988509	0.1310447785	
require_guest_profile_picture		0.665186221
-0.0133733088	-0.0131087550	
require_guest_phone_verification		1.000000000
0.0235807616	0.0251866599	
calculated_host_listings_count_y		0.023580762
1.0000000000	0.9880298464	
	calculated_host_listings_count_private_rooms	calculated_host_listings_count_shared_rooms
security_deposit	cleaning_fee	
price_x		-0.0228103656
2.909236e-02	1.345538e-01	0.247801934
experiences_offered		0.0020984474
3.314690e-03	-2.503041e-02	-0.013506432
host_is_superhost		-0.0278330839
-2.782984e-02	2.595129e-03	-0.053853458
host_listings_count		0.1063719011
-1.388643e-03	1.017570e-01	0.400934999
host_has_profile_pic		0.0026636636
3.572832e-03	-5.068144e-03	0.009669859

host_identity_verified				-0.0473734478
-2.592674e-02	4.078645e-02	-0.052788785		
is_location_exact				-0.0516114357
3.713586e-03	-5.860764e-03	-0.012274697		
property_type				0.0019882620
2.961655e-02	2.545005e-02	0.018997741		
room_type_y				0.1303234612
9.589852e-02	-1.500926e-01	-0.359718845		
accommodates				-0.0164297090
2.219670e-02	1.760922e-01	0.450702969		
bathrooms				-0.0113070672
7.196242e-02	1.686900e-01	0.336216655		
bedrooms				-0.0202891303
-3.110982e-02	1.932609e-01	0.444869230		
beds				-0.0023715701
9.398493e-02	1.559637e-01	0.387562811		
bed_type				0.0069817699
8.560413e-05	2.312033e-03	0.022767270		
guests_included				0.0312538843
-2.160267e-02	1.086775e-01	0.384347320		
extra_people				0.0120475167
-2.301364e-03	6.366417e-02	0.049926735		
minimum_nights_y				-0.0058882417
-8.208684e-03	5.598363e-02	0.037140531		
maximum_nights				0.0002487684
-2.938229e-04	2.586969e-06	-0.003971972		
availability_30				0.0881804032
1.582381e-02	1.417277e-02	-0.016071589		
availability_60				0.0912516903
3.194194e-02	2.238228e-02	-0.001001482		
availability_90				0.0920610699
3.785224e-02	2.198503e-02	-0.002434899		
availability_365_y				0.1244479326
5.511872e-02	3.242523e-02	0.014480750		
number_of_reviews_y				0.0022693517
1.882165e-04	-4.831156e-02	-0.096702746		
number_of_reviews_ltm				0.0162194111
1.925436e-02	-7.482439e-02	-0.095994666		
instant_bookable				-0.0070298173
3.681745e-02	-5.234460e-02	0.020718684		
cancellation_policy				0.0677996162
-2.862394e-02	7.828071e-02	0.126077038		
require_guest_profile_picture				0.0011551373
-6.267512e-03	2.492514e-02	-0.016174696		
require_guest_phone_verification				-0.0067671540
-8.284440e-03	3.939727e-02	0.026755340		
calculated_host_listings_count_y				0.1464385574
3.801238e-03	1.220339e-01	0.446507846		

[reached getopt("max.print") -- omitted 5 rows]

It appears the most correlated variables are: "host_listings_count", "room_type_y", "accomodates", "bathrooms", "bedrooms", "beds", "guests_included", "calculated_host_listings_count_y", "calculated_host_listings_count_entire_homes", "security_deposit", "cleaning_fee" relative to the target price_x.

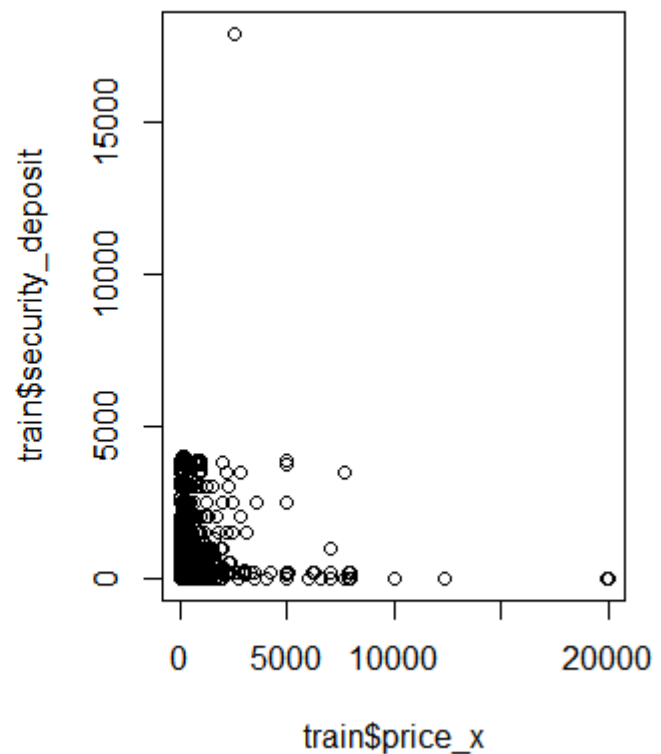
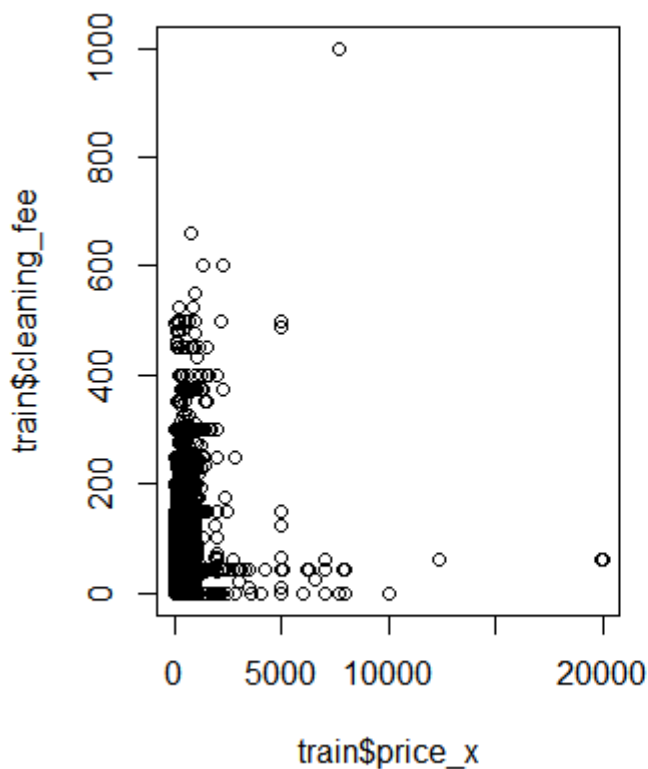
It appears cleaning fees and security deposits are associated with lower range prices for the air bnb.

[Hide](#)

```
par(mfrow=c(1,2))
plot(train$price_x, train$cleaning_fee)
plot(train$price_x, train$security_deposit)
```

[Hide](#)

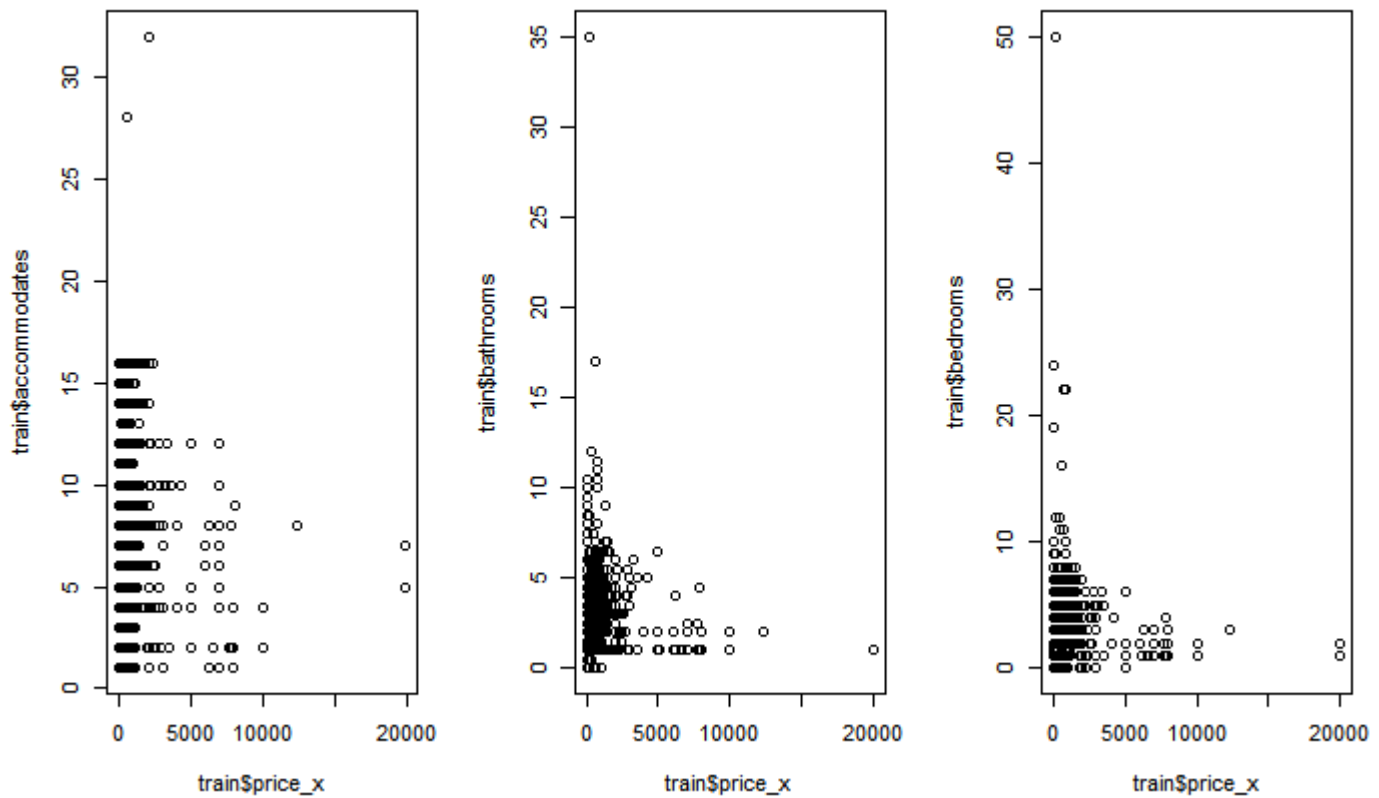
```
par(mfrow=c(1,2))
```


[Hide](#)

```
par(mfrow=c(1,3))
plot(train$price_x, train$accommodates)
plot(train$price_x, train$bathrooms)
```

[Hide](#)

```
plot(train$price_x, train$bedrooms)
par(mfrow=c(1,3))
```



It appears most of the data tends to cluster around a relatively low price point per night, accommodating 15 or less, with around 5 or less bathrooms and less than 10 bedrooms. Interestingly, there are a few outliers, unbelievable ones really. Such as the mysteriously low-priced per night stay at a place with 50 bedrooms and around 35 bathrooms, accomodating over 30 people. I'd like to have that for myself to be honest haha. What a deal!

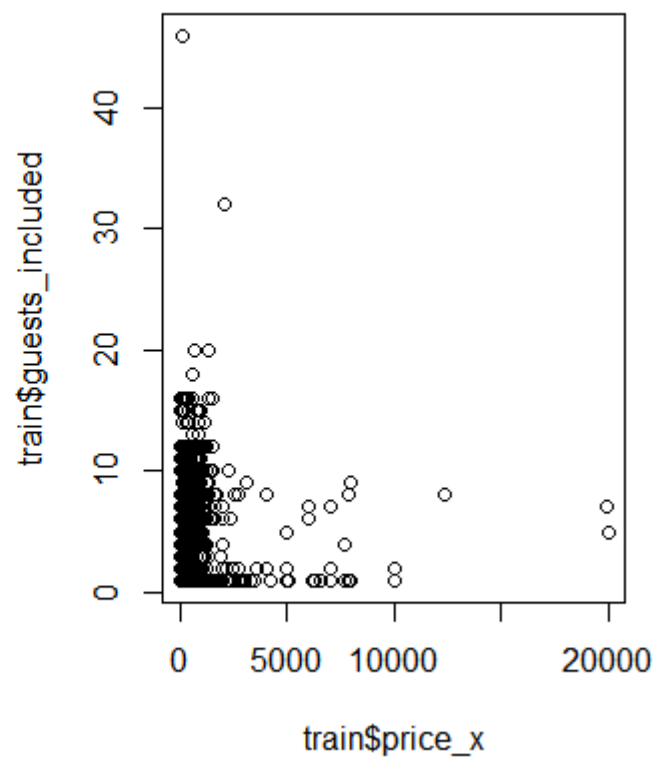
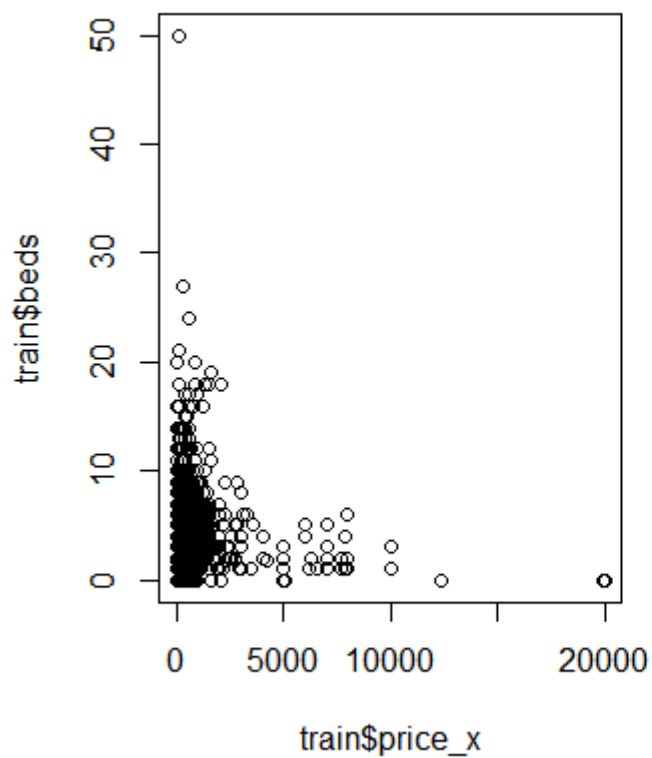
Nonetheless, let's consider more datapoints to get a better view of what the data is telling us.

[Hide](#)

```
par(mfrow=c(1,2))
plot(train$price_x, train$beds)
plot(train$price_x, train$guests_included)
```

[Hide](#)

```
par(mfrow=c(1,2))
```



So it appears other points tend to cluster in those same areas suggesting the same as implied before.

Linear Regression

[Hide](#)

```
mod <- lm(price_x ~ host_listings_count + accommodates + bathrooms + bedrooms + beds + guests_included, data=train)
summary(mod)
```


Call:

```
lm(formula = price_x ~ host_listings_count + accommodates + bathrooms +
    bedrooms + beds + guests_included, data = train)
```

Residuals:

Min	1Q	Median	3Q	Max
-1373.9	-49.3	-23.9	8.2	19759.4

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	-19.563239	2.440795	-8.015	1.12e-15	***
host_listings_count	0.228032	0.007092	32.152	< 2e-16	***
accommodates	27.788278	0.943777	29.444	< 2e-16	***
bathrooms	38.502405	1.993199	19.317	< 2e-16	***
bedrooms	6.562241	1.813960	3.618	0.000298	***
beds	-11.136999	1.350894	-8.244	< 2e-16	***
guests_included	7.778748	0.876780	8.872	< 2e-16	***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 258.5 on 66960 degrees of freedom

Multiple R-squared: 0.09379, Adjusted R-squared: 0.09371

F-statistic: 1155 on 6 and 66960 DF, p-value: < 2.2e-16

It appears all of the independent variables are statistically significant, yet the R^2 is pretty low, nearing 9%. Let's see how well the model is predicting with respect to correlation and mean squared error.

Hide

```
pred <- predict(mod, newdata=test)
cor_lm <- cor(pred, test$price_x)
mse_lm <- mean((pred-test$price_x)^2)
print(paste("cor=", cor_lm))
```

```
[1] "cor= 0.272380549096964"
```

Hide

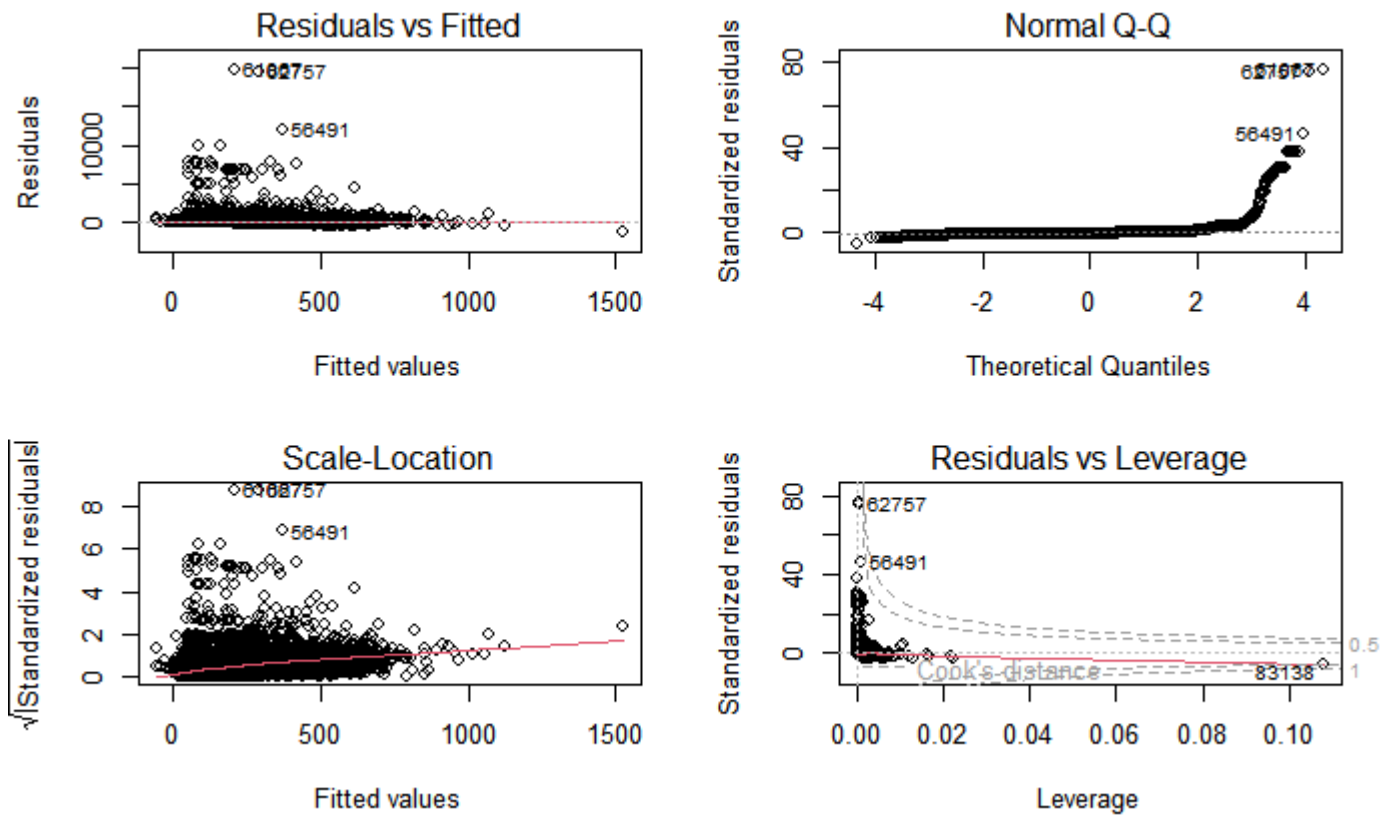
```
print(paste("mse=", mse_lm))
```

```
[1] "mse= 95870.6811589135"
```

Darn. The correlation is low and the error is high. Let's inspect the model performance using residuals plot.

Hide

```
par(mfrow=c(2,2))
plot(mod)
par(mfrow=c(2,2))
```



It appears there's a good number of outliers sitting aside the line in the residuals vs fitted plot, the variance does not appear to be constant, as an assumption of linear model. In the scale-location plot this is seen as well, most of the points appear to cluster around 1 and extend to around 2, however, the number of outliers seem to point at a non-linear relationship being present in the data. In addition the normal q-q points that while most of the data fits to what appears to be linear, the end tail towards the upper theoretical quantile seems to skew away. It appears some of those outliers have high residual, indicating their poor fit by the model.

So it doesn't look like the relationship between these points and the data can be prescribed to a linear regression, let's explore the other kinds of regression.

KNN Regression

We will explore KNN regression using the same variables used in the linear regression: price_x, host_listings_count, accommodates, bathrooms, bedrooms, beds, and guests_included

[Hide](#)

```
set.seed(124) # reproducibility
data <- subset(df, select = c(price_x, host_listings_count, accommodates, bathrooms, bedrooms, beds, guests_included))
i <- sample(1:nrow(data), nrow(data)*.80, replace=FALSE)
train <- data[i,]
test <- data[-i,]
```

Clustering algorithms work best when the data is scaled. Let's scale the data.

[Hide](#)

```
means <- sapply(train, mean)
stdevs <- sapply(train, sd)
train_scaled <- scale(train, center=means, scale=stdevs)
test_scaled <- scale(test, center=means, scale=stdevs)
```

Let's test various values of k to find the best value for k.

[Hide](#)

```
cor_k <- rep(0,20)
mse_k <- rep(0,20)
i <- 1
for(k in seq(1,39,2)) {
  fit_k <- knnreg(train_scaled, train$price_x, k=k)
  pred_k <- predict(fit_k, test_scaled)
  cor_k[i] <- cor(pred_k, test$price_x)
  mse_k[i] <- mean((pred_k - test$price_x)^2)
  print(paste("k=",k,cor_k[i],mse_k[i]))
  i <- i+1
}
```

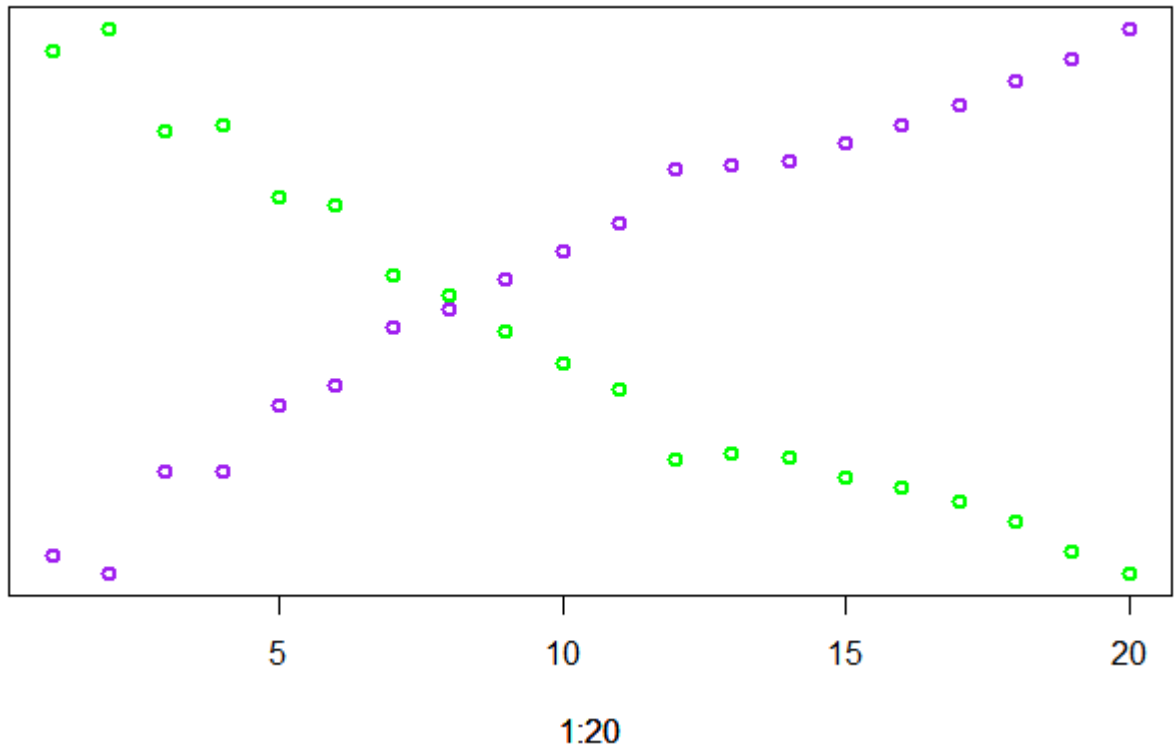
```
[1] "k= 1 0.99289845218099 1335.26132077423"
[1] "k= 3 0.994586889110888 1021.90929102559"
[1] "k= 5 0.987103422856028 2780.47789242571"
[1] "k= 7 0.987463378553037 2771.71592311804"
[1] "k= 9 0.98228937076861 3910.34731044692"
[1] "k= 11 0.981623168707515 4244.56988157346"
[1] "k= 13 0.976617724836065 5249.4019821889"
[1] "k= 15 0.975091605365138 5575.93896250606"
[1] "k= 17 0.972489561104916 6066.98821601118"
[1] "k= 19 0.970156297423221 6564.80508932371"
[1] "k= 21 0.968159141080858 7040.79195210361"
[1] "k= 23 0.963122972556195 7972.64791578296"
[1] "k= 25 0.9635670752405 8032.37822502446"
[1] "k= 27 0.963222969825831 8113.40101313052"
[1] "k= 29 0.961791341311107 8428.0823145298"
[1] "k= 31 0.961104760727688 8750.53111666039"
[1] "k= 33 0.960091948584439 9094.63946673805"
[1] "k= 35 0.958517368092966 9501.31780704318"
[1] "k= 37 0.956327483006558 9889.88697467776"
[1] "k= 39 0.954798489339322 10404.5673664586"
```

[Hide](#)

```
plot(1:20, cor_k, lwd=2, col='green', ylab="", yaxt='n')
par(new=TRUE)
```

[Hide](#)

```
plot(1:20, mse_k, lwd=2, col='purple', ylab="", labels=FALSE, yaxt='n')
```



It appears when $k=3$ we got the lowest MSE and the highest correlation, just to reiterate, let's fit a knn regression with $k=3$ and see our results.

[Hide](#)

```
fit <- knnreg(train_scaled, train$price_x, k=3)
pred <- predict(fit, test_scaled)
cor_knn <- cor(pred, test$price_x)
mse_knn <- mean((pred-test$price_x)^2)
print(paste("cor=", cor_knn))
```

```
[1] "cor= 0.994586889110888"
```

[Hide](#)

```
print(paste("mse=", mse_knn))
```

```
[1] "mse= 1021.90929102559"
```

In so far, KNN regression appears to be heavily outperforming linear regression. Let's see if decision tree regression can improve it further.

Decision Tree Regression

We will use the same columns used in the other two regressions.

[Hide](#)

```
set.seed(125) # reproducibility
i <- sample(1:nrow(data), nrow(data)*0.80, replace=FALSE)
train <- data[i,]
test <- data[-i,]
```

Hide

```
tree1 <- tree(price_x~.,data=train)
summary(tree1)
```

Regression tree:
tree(formula = price_x ~ ., data = train)
Variables actually used in tree construction:
[1] "bedrooms" "bathrooms"
Number of terminal nodes: 3
Residual mean deviance: 72110 = 4.829e+09 / 66960
Distribution of residuals:

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
-440.500	-56.150	-32.800	0.000	7.853	19840.000

Hide

```
pred <- predict(tree1,newdata=test)
cor_tree <- cor(pred, test$price_x)
mse_tree <- mean((pred-test$price_x)^2)
print(paste("cor=",cor_tree))
```

```
[1] "cor= 0.216114372156003"
```

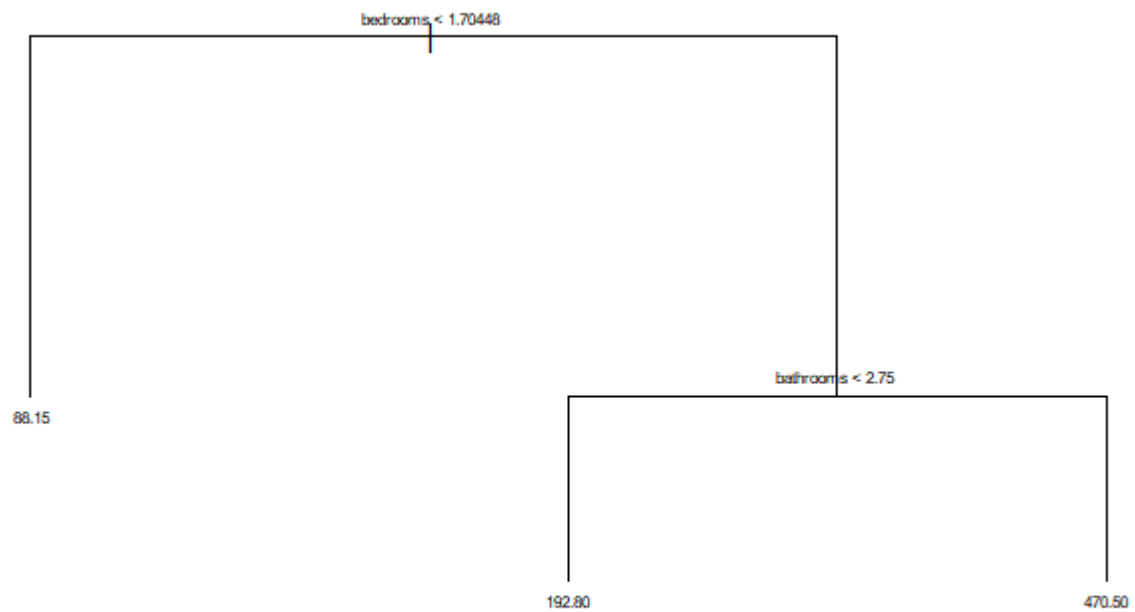
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```
print(paste("mse=",mse_tree))
```

```
[1] "mse= 85121.8317102931"
```

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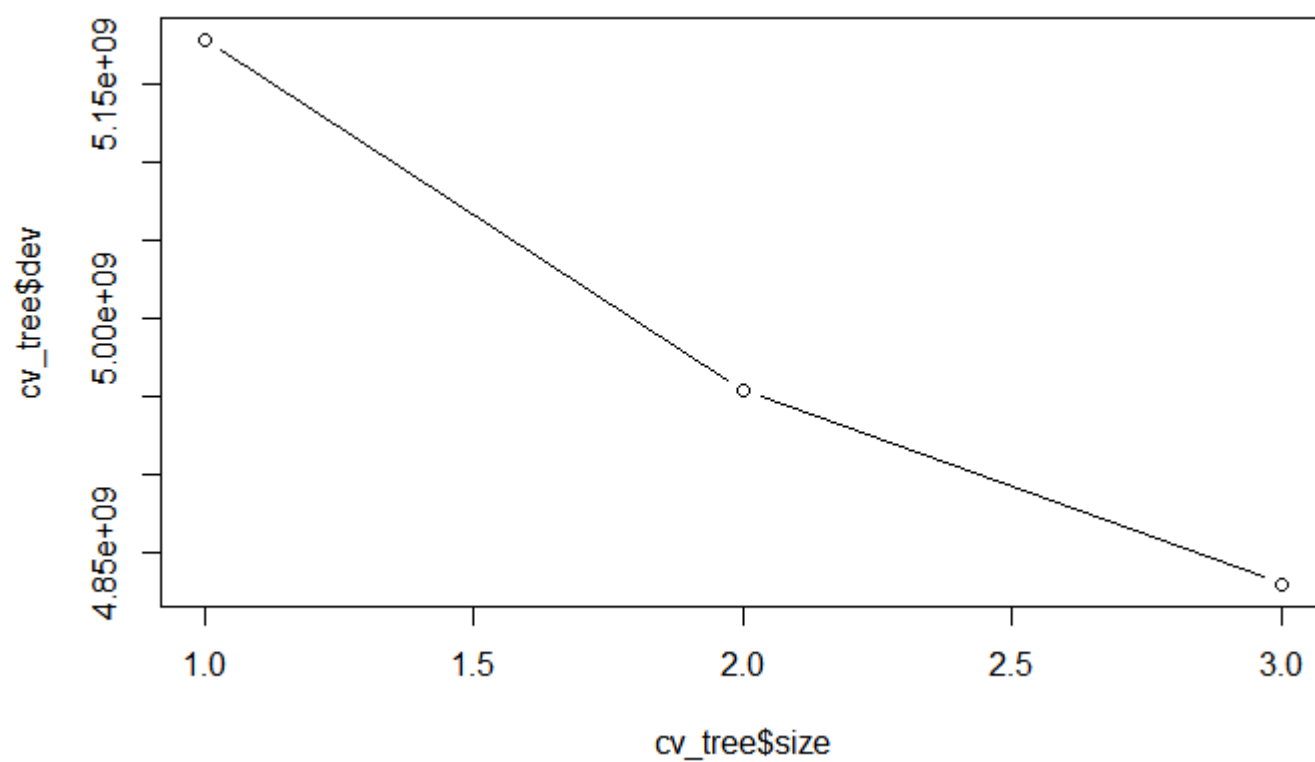
```
plot(tree1)
text(tree1, cex=0.5, pretty=0)
```



The tree only has 3 nodes, it's deviance is quite high, let's inspect cross validation.

[Hide](#)

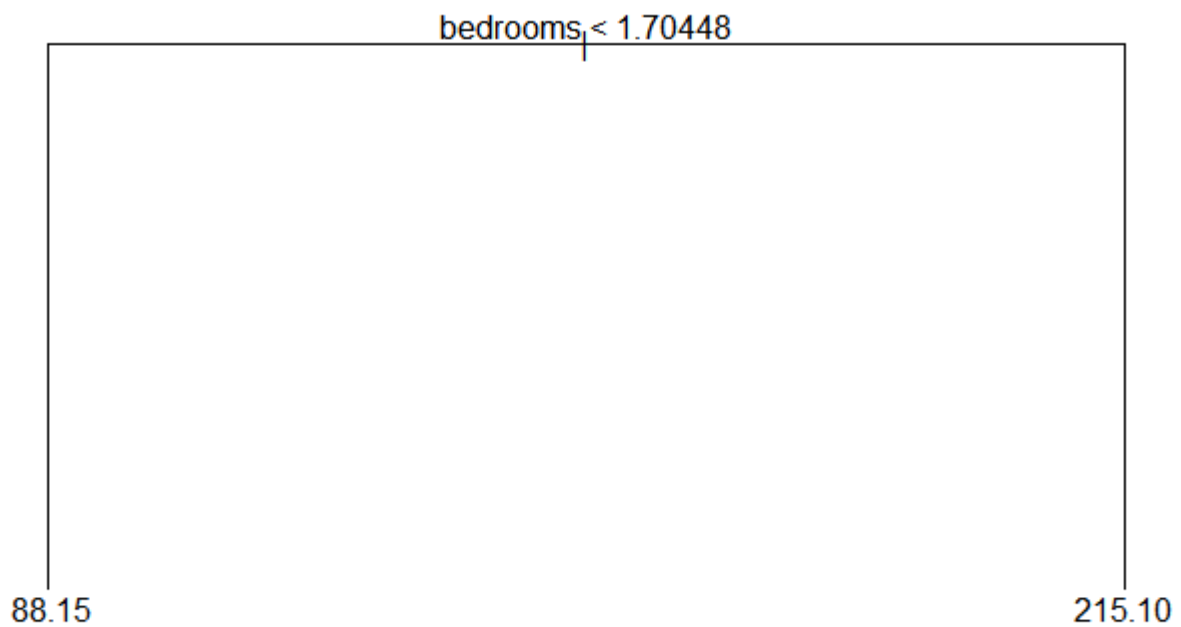
```
cv_tree <- cv.tree(tree1)
plot(cv_tree$size, cv_tree$dev, type='b')
```



The middle ground appears to be at 2 nodes, pruning the tree to 2 terminal nodes, then plotting

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```
tree_pruned <- prune.tree(tree1, best=2)
plot(tree_pruned)
text(tree_pruned, pretty=0)
```



The correlation and mse for the original tree were the worst in so far. Measuring for the pruned tree, let's see if it improves.

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```

pred_pruned <- predict(tree_pruned, newdata=test)
cor_pruned <- cor(pred_pruned, test$price_x)
mse_pruned <- mean((pred_pruned-test$price_x)^2)
print(paste("corr=",cor_pruned))

```

```
[1] "corr= 0.184066304579861"
```

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```
print(paste("mse=",mse_pruned))
```

```
[1] "mse= 86212.5684800797"
```

Nope, apparently not. The correlation went down and the mse raised. This was the worst result of all the regressions.

Analysis

Out of the three models explored in this notebook: linear regression, knn regression, and decision tree regression, the most performant was knn regression. There was apparent clustering as a result of data exploration. We could clearly see clustering relationships between the target and multiple predictors, showing a similar relationship. The goodness of fit for the linear regression was poor as there were many outliers present in the data. The knn

clustering algorithm helped reduce the group size to just 3 clusters, while minimizing the mean absolute error and maximizing the correlation. The decision tree regression wasn't as helpful in this regard as a highly deviant tree of only three nodes was constructed, leaving the middle ground of two terminal nodes to be as deviant.