

# Data Analysis of Rental Pricing in Klang Valley

## Rental Pricing Trend within Klang Valley

Dataset was taken from Kaggle, webscraped from Mudah.com by a public user. This project aims to provide visual insights on Klang Valley's rental trend with given category (provided in Mudah website) such as location, size, number of bedrooms, parking or furnished. Chosen this project as I am also actively looking for a new unit to move in Klang Valley, having this insights will help me compare my options, evaluating them within the actual market trend.

### Installing data packages and library

```
options(repos = "https://cloud.r-project.org")
```

```
install.packages("stringr")
```

```
## Installing package into 'C:/Users/1/AppData/Local/R/win-library/4.3'  
## (as 'lib' is unspecified)
```

```
## package 'stringr' successfully unpacked and MD5 sums checked  
##  
## The downloaded binary packages are in  
## C:\Users\1\AppData\Local\Temp\RtmpA1Y9F6\downloaded_packages
```

```
library(stringr)
```

```
## Warning: package 'stringr' was built under R version 4.3.1
```

```
install.packages("ggplot2")
```

```
## Installing package into 'C:/Users/1/AppData/Local/R/win-library/4.3'  
## (as 'lib' is unspecified)
```

```
## package 'ggplot2' successfully unpacked and MD5 sums checked  
##  
## The downloaded binary packages are in  
## C:\Users\1\AppData\Local\Temp\RtmpA1Y9F6\downloaded_packages
```

```
library(ggplot2)
```

```
## Warning: package 'ggplot2' was built under R version 4.3.1
```

```
install.packages("tidyverse")
```

```
## Installing package into 'C:/Users/1/AppData/Local/R/win-library/4.3'  
## (as 'lib' is unspecified)
```

```
## package 'tidyverse' successfully unpacked and MD5 sums checked  
##  
## The downloaded binary packages are in  
## C:\Users\1\AppData\Local\Temp\RtmpAlY9F6\downloaded_packages
```

```
library(tidyverse)
```

```
## Warning: package 'tidyverse' was built under R version 4.3.1
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --  
## v dplyr      1.1.2      v readr      2.1.4  
## v forcats    1.0.0      v tibble    3.2.1  
## v lubridate  1.9.2      v tidyr     1.3.0  
## v purrr      1.0.1
```

```
## -- Conflicts ----- tidyverse_conflicts() --  
## x dplyr::filter() masks stats::filter()  
## x dplyr::lag()     masks stats::lag()  
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

```
install.packages("dplyr")
```

```
## Warning: package 'dplyr' is in use and will not be installed
```

```
library(dplyr)  
library(readr)  
install.packages("wordcloud")
```

```
## Installing package into 'C:/Users/1/AppData/Local/R/win-library/4.3'  
## (as 'lib' is unspecified)
```

```
## package 'wordcloud' successfully unpacked and MD5 sums checked
```

```
## Warning: cannot remove prior installation of package 'wordcloud'
```

```
## Warning in file.copy(savedcopy, lib, recursive = TRUE): problem copying  
## C:\Users\1\AppData\Local\R\win-library\4.3\00LOCK\wordcloud\libs\x64\wordcloud.dll  
## to C:\Users\1\AppData\Local\R\win-library\4.3\wordcloud\libs\x64\wordcloud.dll:  
## Permission denied
```

```
## Warning: restored 'wordcloud'
```

```
##  
## The downloaded binary packages are in  
## C:\Users\1\AppData\Local\Temp\RtmpAlY9F6\downloaded_packages
```

```
library(wordcloud)
```

```
## Warning: package 'wordcloud' was built under R version 4.3.1
```

```
## Loading required package: RColorBrewer
```

```
install.packages("broom")
```

```
## Installing package into 'C:/Users/1/AppData/Local/R/win-library/4.3'  
## (as 'lib' is unspecified)
```

```
## package 'broom' successfully unpacked and MD5 sums checked  
##  
## The downloaded binary packages are in  
## C:\Users\1\AppData\Local\Temp\RtmpAly9F6\downloaded_packages
```

```
library(broom)
```

```
## Warning: package 'broom' was built under R version 4.3.1
```

```
rental_df = read.csv("Rental Pricing.csv")
```

## Data Exploration

To describe the whole dataset (rows and columns), also checking on how many empty cells are within the dataset

```
colnames(rental_df)
```

```
## [1] "ads_id"           "property_type"    "prop_name"  
## [4] "completion_year" "monthly_rent"     "district"  
## [7] "parking"          "bathroom"         "size"  
## [10] "furnished"        "facilities"       "additional_facilities"  
## [13] "region"
```

```
str(rental_df)
```

```
## 'data.frame': 19991 obs. of 13 variables:  
## $ ads_id : int 100323185 100203973 100323128 100191767 97022692 100322897 100322962 ...  
## $ property_type : chr "Condominium" "Condominium" "Apartment" "Apartment" ...  
## $ prop_name : chr "The Hipster @ Taman Desa" "Segar Courts" "Pangsapuri Teratak Muhibbal ...  
## $ completion_year : int 2022 NA NA 2020 NA NA NA 2018 2014 NA ...  
## $ monthly_rent : num 4200 2300 1000 1700 1299 ...  
## $ district : chr "Taman Desa" "Cheras" "Taman Desa" "Sentul" ...  
## $ parking : int 2 1 NA 1 1 1 2 1 1 1 ...  
## $ bathroom : int 6 2 2 2 1 2 2 1 1 2 ...  
## $ size : int 1842 1170 650 743 494 884 982 700 750 862 ...  
## $ furnished : chr "Fully Furnished" "Partially Furnished" "Fully Furnished" "Partially ...  
## $ facilities : chr "Minimart, Gymnasium, Security, Playground, Swimming Pool, Parking, L ...  
## $ additional_facilities: chr "Air-Cond, Cooking Allowed, Washing Machine" "Air-Cond, Cooking Allow ...  
## $ region : chr "Kuala Lumpur" "Kuala Lumpur" "Kuala Lumpur" "Kuala Lumpur" ...
```

```
head(rental_df)
```

```
##      ads_id      property_type      prop_name completion_year
## 1 100323185      Condominium      The Hipster @ Taman Desa      2022
## 2 100203973      Condominium      Segar Courts      NA
## 3 100323128      Apartment Pangsapuri Teratak Muhibbah 2      NA
## 4 100191767      Apartment      Sentul Point Suite Apartment      2020
## 5 97022692 Service Residence      Arte Mont Kiara      NA
## 6 100322897      Apartment      Residensi Vista Wirajaya      NA
##      monthly_rent      district parking bathroom size      furnished
## 1          4200      Taman Desa      2          6 1842      Fully Furnished
## 2          2300      Cheras      1          2 1170      Partially Furnished
## 3          1000      Taman Desa      NA          2 650      Fully Furnished
## 4          1700      Sentul      1          2 743      Partially Furnished
## 5          1299      Mont Kiara      1          1 494      Not Furnished
## 6          1500      Setapak      1          2 884      Partially Furnished
##
## 1          Minimart, Gymnasium, Security, Playground, Swimming Pool, Parking, Lift, Barbeque area
## 2          Playground, Parking, Barbeque area, Security, Jogging Track, Swimming Pool
## 3          Minimart, Gymnasium, Security, Playground, Swimming Pool, Parking, Lift, Barbeque area
## 4          Parking, Playground, Swimming Pool, Squash Court, Squash Court, Swimming Pool
## 5 Parking, Security, Lift, Swimming Pool, Playground, Gymnasium, Barbeque area, Minimart, Multipurpose
## 6          Parking, Security, Lift, Swimming Pool, Gymnasium, Barbeque area, Minimart, Multipurpose
##
##      additional_facilities      region
## 1      Air-Cond, Cooking Allowed, Washing Machine Kuala Lumpur
## 2      Air-Cond, Cooking Allowed, Near KTM/LRT Kuala Lumpur
## 3          Cooking Allowed, Near KTM/LRT, Washing Machine Kuala Lumpur
## 4      Air-Cond, Cooking Allowed, Near KTM/LRT, Washing Machine Kuala Lumpur
## 5          Air-Cond, Cooking Allowed, Near KTM/LRT Kuala Lumpur
## 6          Cooking Allowed, Near KTM/LRT Kuala Lumpur
```

```
emptycell = colSums(is.na(rental_df))
print(emptycell)
```

```
##      ads_id      property_type      prop_name
##          0          0          0
##      completion_year      monthly_rent      district
##          9185          2          0
##      parking      bathroom      size
##          5702          6          0
##      furnished      facilities additional_facilities
##          0          0          0
##      region
##          0
```

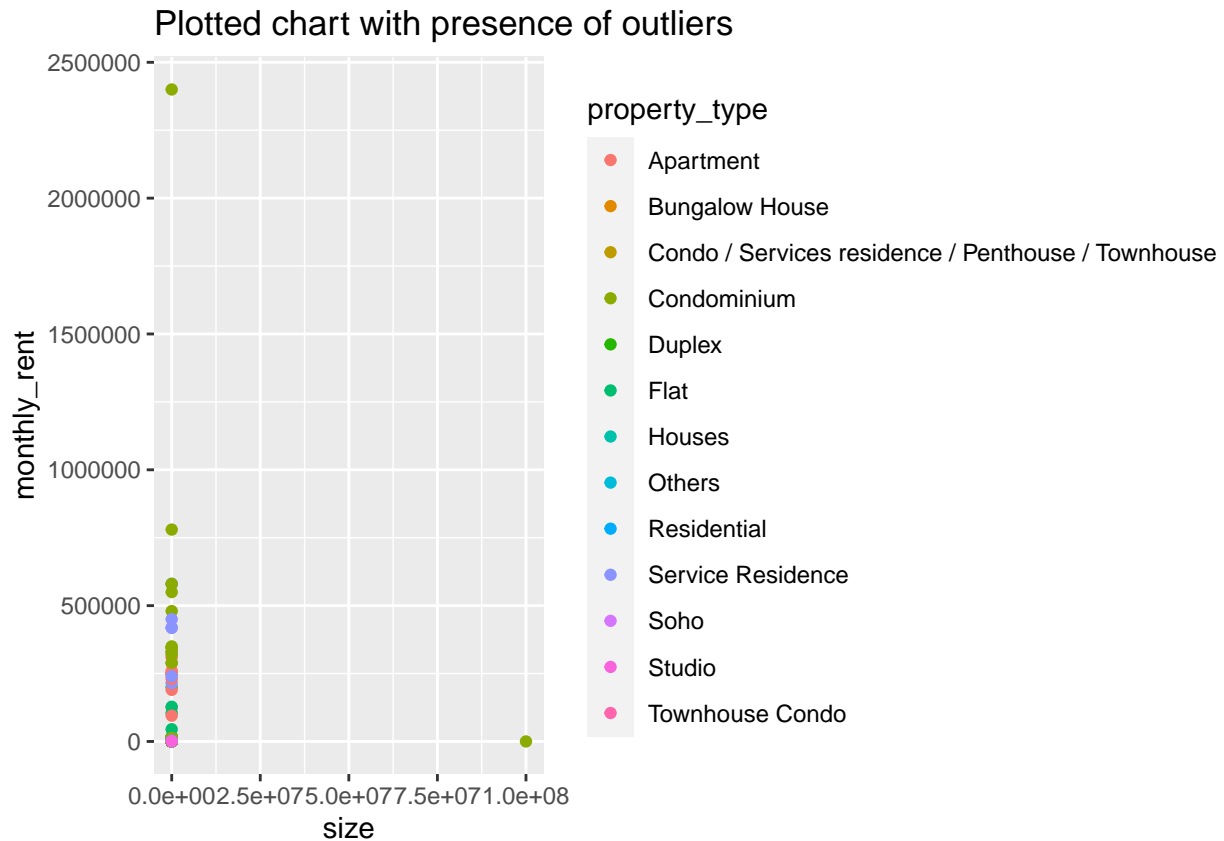
Dataset is already good as it is. Empty cells in column 'parking' and 'completion\_year' will be ignored as it will not be used as much in the upcoming analysis.

## Data Preparation

**Checking for numerical outliers in the dataset** ### Checking for outliers and important variable column of interest; size and monthly rent. Plotting graph with outliers will make our plots skewed.

```
ggplot(data = rental_df) +
  geom_point(mapping = aes(x = size, y = monthly_rent, color = property_type)) +
  labs(title = "Plotted chart with presence of outliers")
```

```
## Warning: Removed 2 rows containing missing values ('geom_point()').
```



```
rentsummary = summary(rental_df$monthly_rent)
q1 = rentsummary["1st Qu."]
median = rentsummary['Median']
q3 = rentsummary["3rd Qu."]

lower_limit = q1 - 1.5 * IQR(rental_df$monthly_rent, na.rm = TRUE)
upper_limit = q3 + 1.5 * IQR(rental_df$monthly_rent, na.rm = TRUE)

print(lower_limit)
```

```
## 1st Qu.
##      50
```

```
print(upper_limit)
```

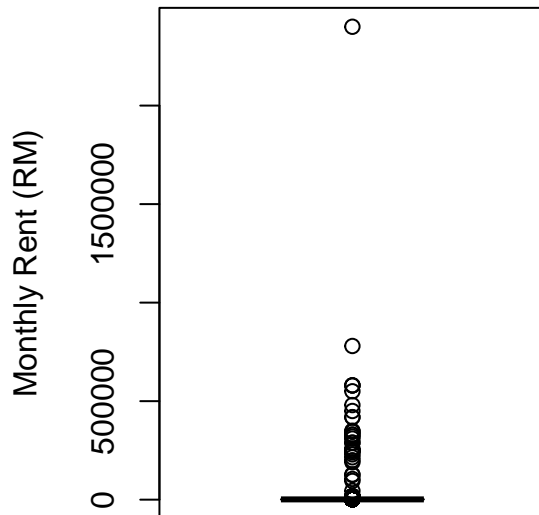
```
## 3rd Qu.
##     2850
```

```

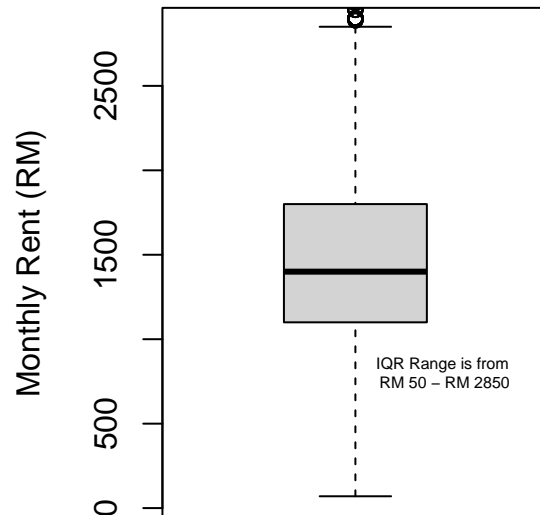
par(mfrow = c(1, 2))
boxplot(rental_df$monthly_rent, main = "Boxplot of Monthly Rent", ylab = "Monthly Rent (RM)", cex.main = 1.5)
boxplot(rental_df$monthly_rent, main = "Boxplot of Monthly Rent (outliers removed)", ylab = "Monthly Rent (RM)", cex.main = 1.5)
text(x = 1.25, y = 800, labels = "IQR Range is from RM 50 - RM 2850", cex = 0.5)

```

Boxplot of Monthly Rent



Boxplot of Monthly Rent (outliers removed)



```

sizesummary = summary(rental_df$size)
q1_size = sizesummary["1st Qu."]
median_size = sizesummary['Median']
q3_size = sizesummary["3rd Qu."]

size_lowerlimit = q1 - 1.5 * IQR(rental_df$size, na.rm = TRUE)
size_upperlimit = q3 + 1.5 * IQR(rental_df$size, na.rm = TRUE)

print(size_lowerlimit)

```

```

## 1st Qu.
##      659

```

```

print(size_upperlimit)

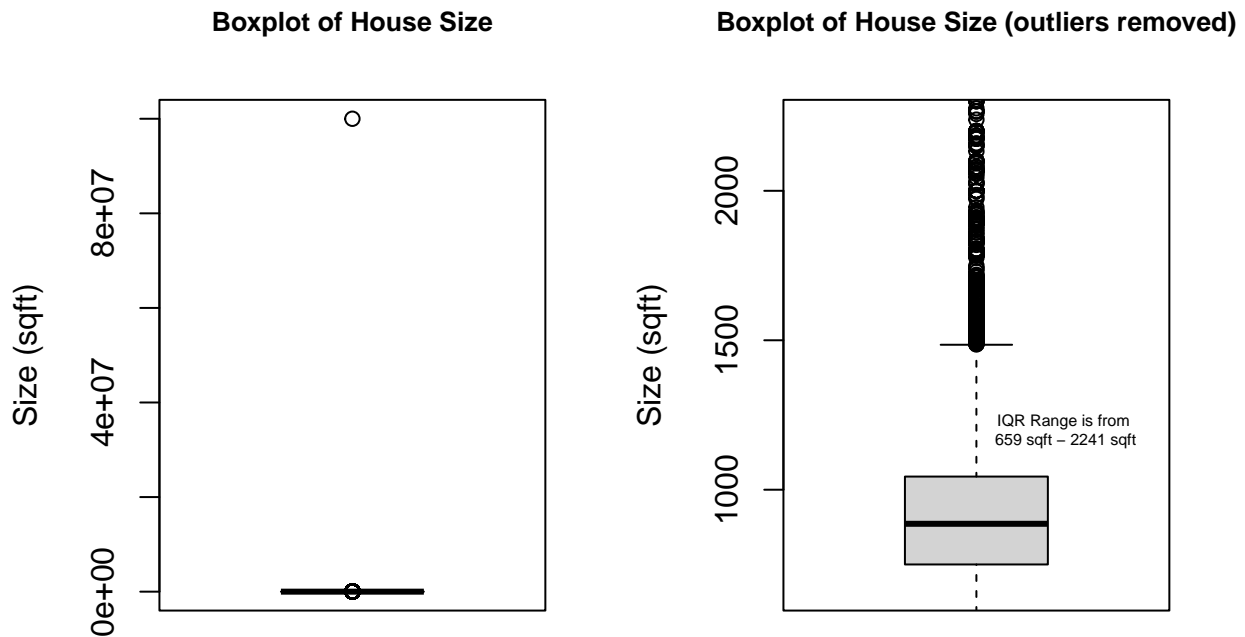
```

```

## 3rd Qu.
##     2241

```

```
par(mfrow = c(1, 2))
boxplot(rental_df$size, main = "Boxplot of House Size", ylab = "Size (sqft)", cex.main = 0.8)
boxplot(rental_df$size, main = "Boxplot of House Size (outliers removed)", ylab = "Size (sqft)", ylim =
text(x = 1.25, y = 1200, labels = "IQR Range is from \n659 sqft - 2241 sqft", cex = 0.5)
```



Creating new dataframe after removing outliers in our numerical column: monthly rent and size.

```
filteredrental = rental_df %>%
  filter(monthly_rent <= 2850 & size <= 2241 & monthly_rent > 100 & size > 100)

dim(filteredrental)
```

```
## [1] 18626    13
```

```
summary(filteredrental)
```

```
##      ads_id      property_type      prop_name      completion_year
##  Min.   : 16525511 Length:18626      Length:18626      Min.   :1977
## 1st Qu.: 99850470 Class :character Class :character 1st Qu.:2013
## Median :100228333 Mode  :character Mode  :character Median :2017
## Mean   : 99723326                      Mean   :2015
## 3rd Qu.:100627338                      3rd Qu.:2020
## Max.   :100854617                      Max.   :2025
##                                     NA's   :8693
```

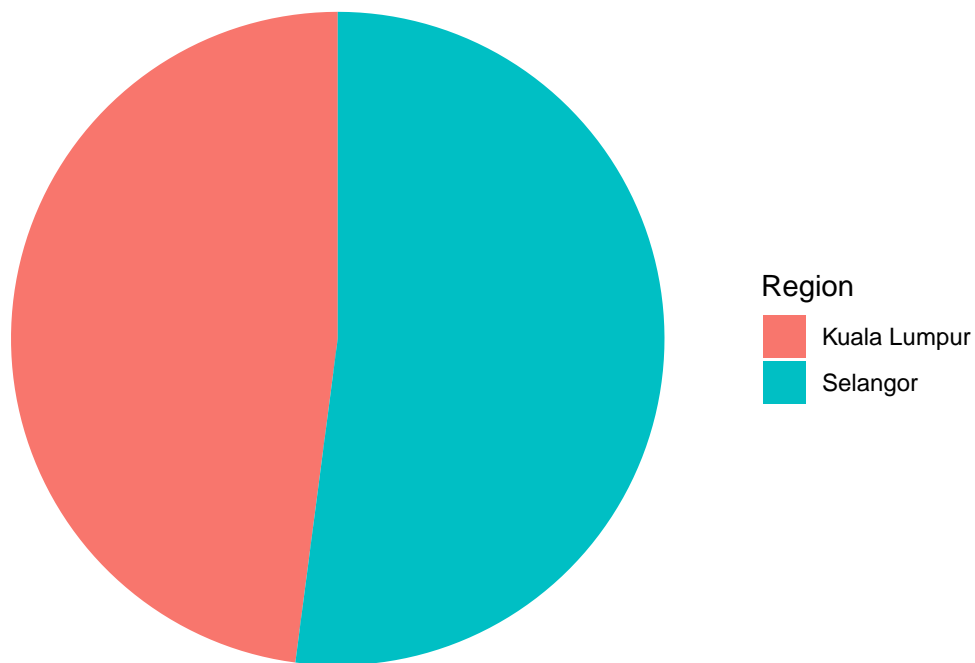
```
##   monthly_rent    district          parking          bathroom
##   Min.      : 120    Length:18626    Min.      : 1.000    Min.      :1.000
##   1st Qu.:1100    Class :character    1st Qu.: 1.000    1st Qu.:2.000
##   Median :1400    Mode  :character    Median : 1.000    Median :2.000
##   Mean   :1447                    Mean   : 1.391    Mean   :1.849
##   3rd Qu.:1750                    3rd Qu.: 2.000    3rd Qu.:2.000
##   Max.    :2850                    Max.    :10.000    Max.    :8.000
##                                     NA's    :5356    NA's    :2
##   size          furnished          facilities          additional_facilities
##   Min.      : 104.0    Length:18626    Length:18626    Length:18626
##   1st Qu.: 737.0    Class :character    Class :character    Class :character
##   Median : 867.0    Mode  :character    Mode  :character    Mode  :character
##   Mean   : 884.6
##   3rd Qu.:1012.0
##   Max.    :2238.0
##
##   region
##   Length:18626
##   Class :character
##   Mode  :character
##
##
##
```

**Data Visualization** To easily illustrate trends in our dataset.

```
ggplot(data = filteredrental, aes(x = "", fill = region)) +
  geom_bar(width = 1) +
  coord_polar("y", start = 0) +
  theme_void() +
  labs(title = "Distribution of rental property availability in KL and Selangor", fill = "Region")
```



## Distribution of rental property availability in KL and Selangor



```
facility_lists = strsplit(filteredrental$facilities, ",")

# Create a list to store the frequency counts
facility_counts = vector("list", length = length(facility_lists))

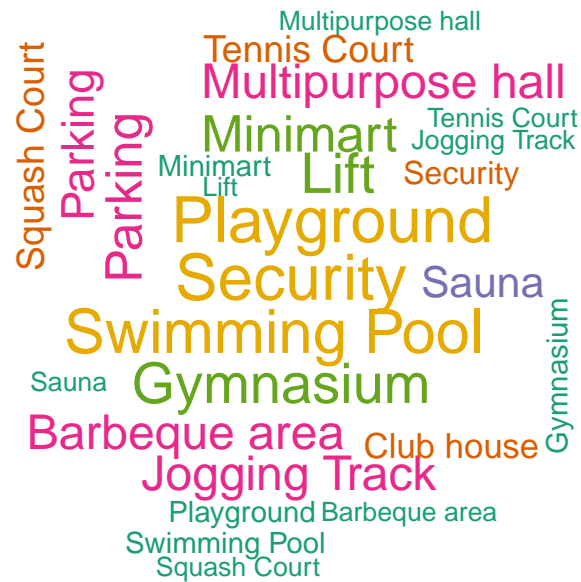
# Count the frequency of each variable in the 'facilities' column
for (i in seq_along(facility_lists)) {
  facility_counts[[i]] = table(facility_lists[[i]])
}

facility_characters = unlist(facility_lists)

# Count the frequency of each character
character_frequency = table(facility_characters)

facility_df = as.data.frame(character_frequency) %>%
  filter(!(facility_characters %in% c(11,10,6))) %>%
  arrange(desc(Freq))

wordcloud(words = facility_df$facility_characters, freq = facility_df$Freq,
  scale = c(2, 0.8), min.freq = 100, random.order = FALSE,
  rot.per = 0.2, colors = brewer.pal(6, "Dark2"),
  main = "Distribution of facilities mentioned")
```

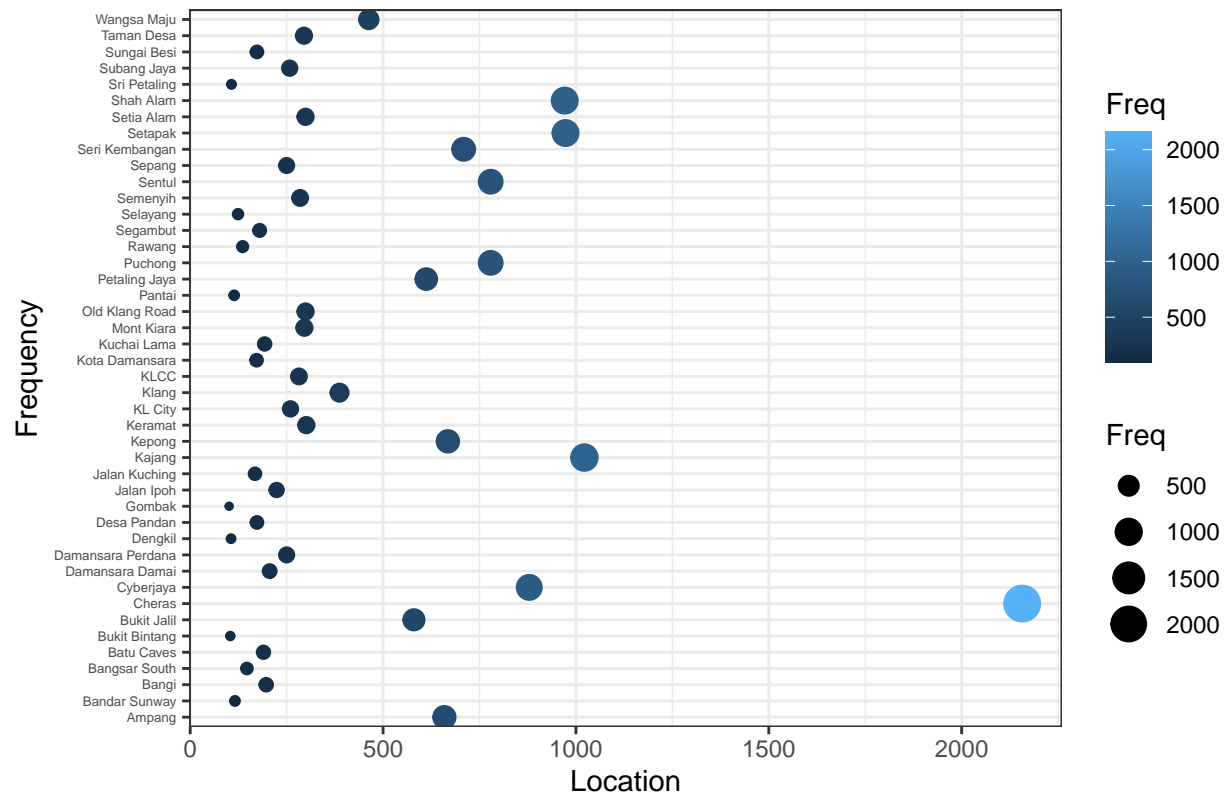


Most listed rentals in Klang Valley are frequently associated with these facilities; security, playground, swimming pool and so on.

```
location_count = (table(rental_df$district)) %>%
  as.data.frame() %>%
  arrange(desc(Freq))
```

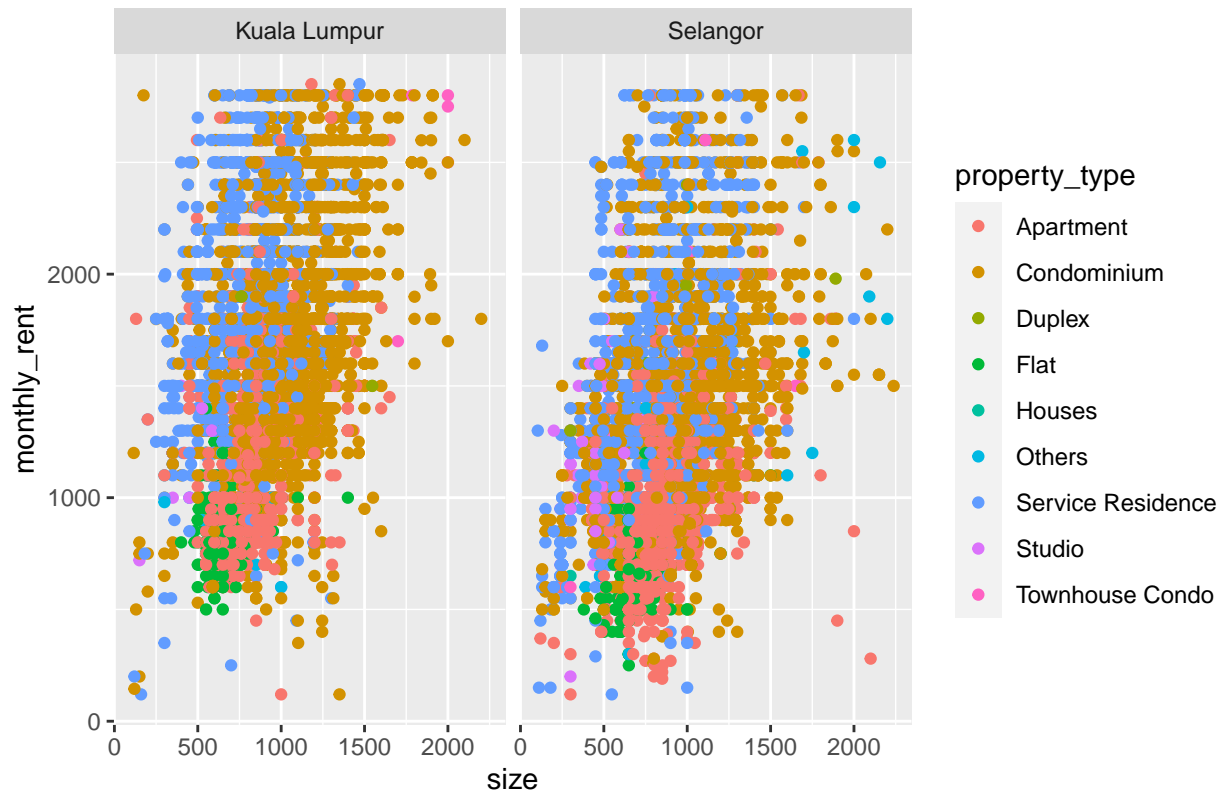
```
location_count %>%
  filter(Freq >= 100) %>%
  ggplot(aes(x = Freq, y = Var1, size = Freq, color = Freq)) +
  geom_point() +
  labs(title = "Distribution of Rental Availability according to District", x = "Location", y = "Frequency") +
  theme_bw() +
  theme(axis.text.y = element_text(size = 5))
```

Distribution of Rental Availability according to District



```
ggplot (data = filteredrental) +
  geom_point(mapping = aes(x = size, y = monthly_rent, color = property_type)) +
  facet_wrap(~region) +
  labs(title = "Distribution of rental property according to Property Type")
```

## Distribution of rental property according to Property Type



Removing property type listed with less than 50, to visually improving the plot and analysis.

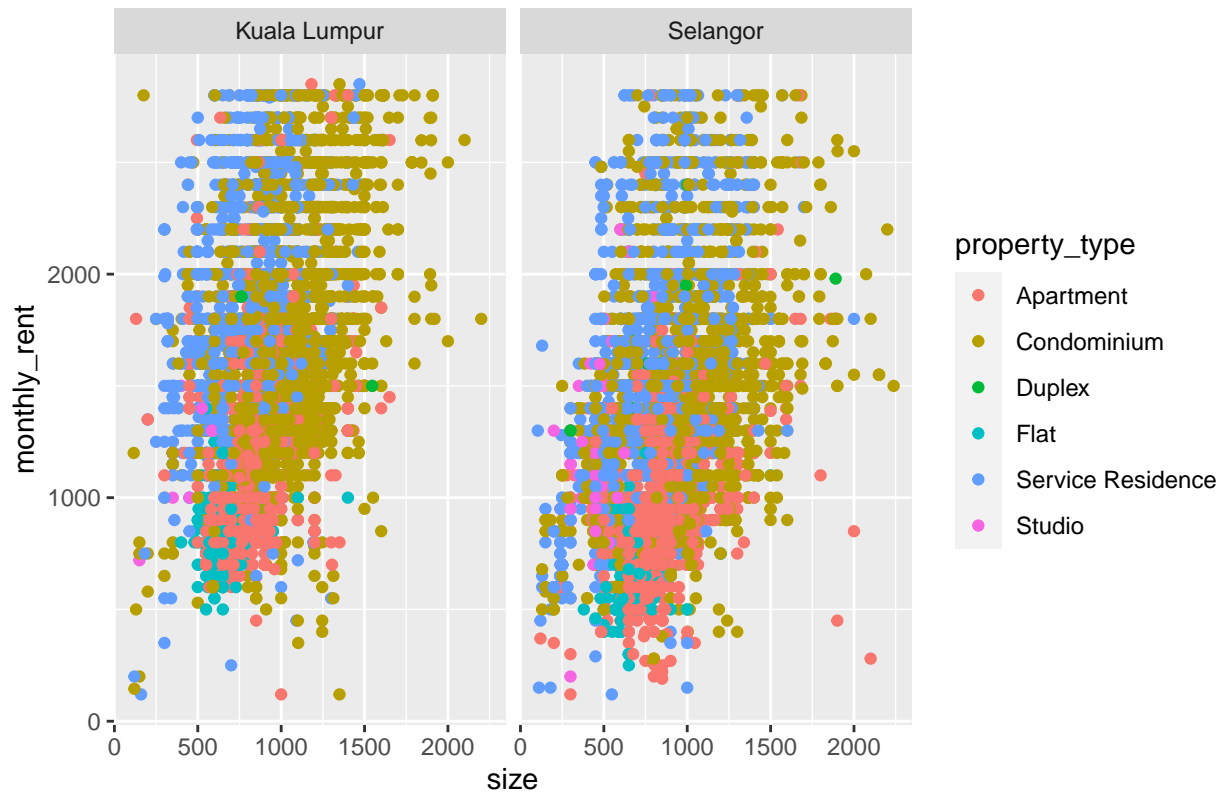
```
table(filteredrental$property_type)
```

```
##
##      Apartment      Condominium      Duplex      Flat
##      5194          7681          69          577
##      Houses      Others Service Residence      Studio
##      1          78          4808          187
##      Townhouse Condo
##      31
```

```
apt_rental = filteredrental %>%
  filter(property_type %in% c("Apartment", "Condominium", "Service Residence", "Flat", "Studio", "Duplex"))

ggplot (data = apt_rental) +
  geom_point(mapping = aes(x = size, y = monthly_rent, color = property_type)) +
  facet_wrap(~region) +
  labs(title = "Distribution of rental property according to Property Type")
```

## Distribution of rental property according to Property Type



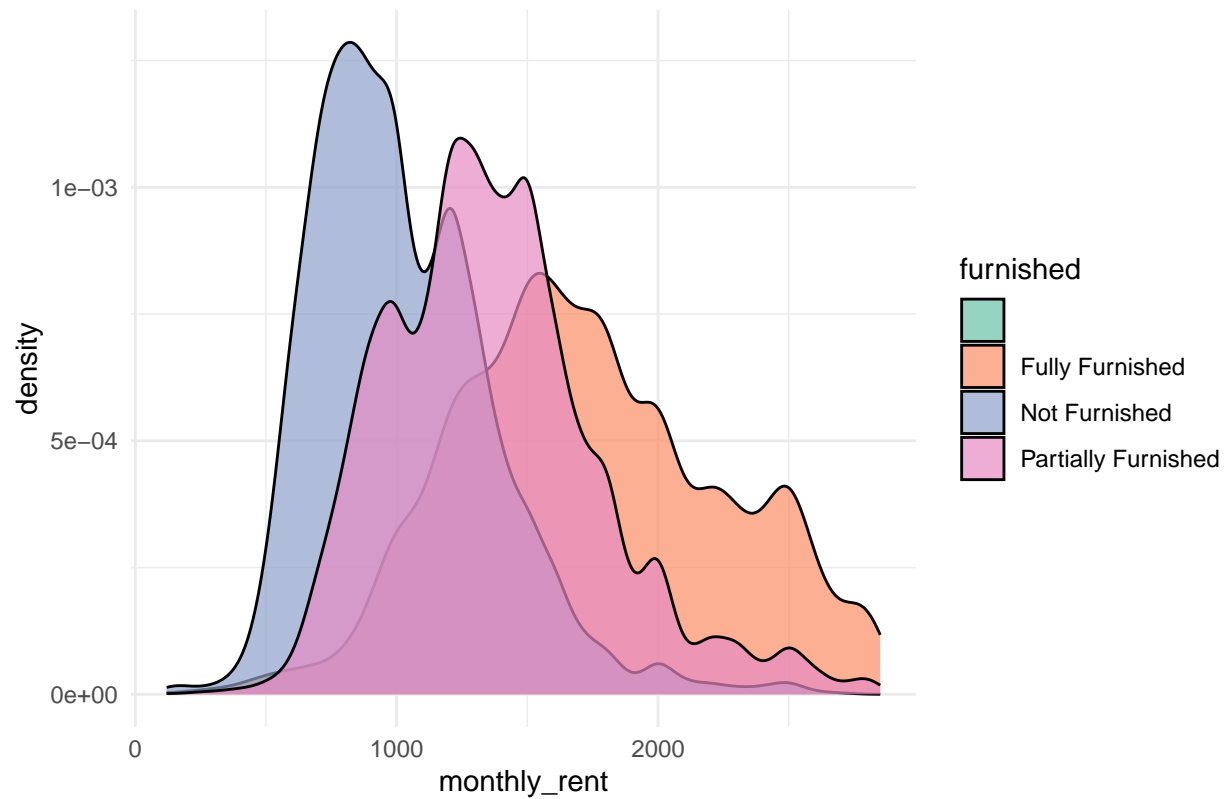
Analysing rental dependency on variable of interest.

```
ggplot(filteredrental, aes(x = monthly_rent, fill = furnished)) +
  geom_density(alpha = 0.7) +
  scale_fill_brewer(palette = "Set2") +
  labs(title = "Rental dependency on Furnished Condition") +
  theme_minimal()
```

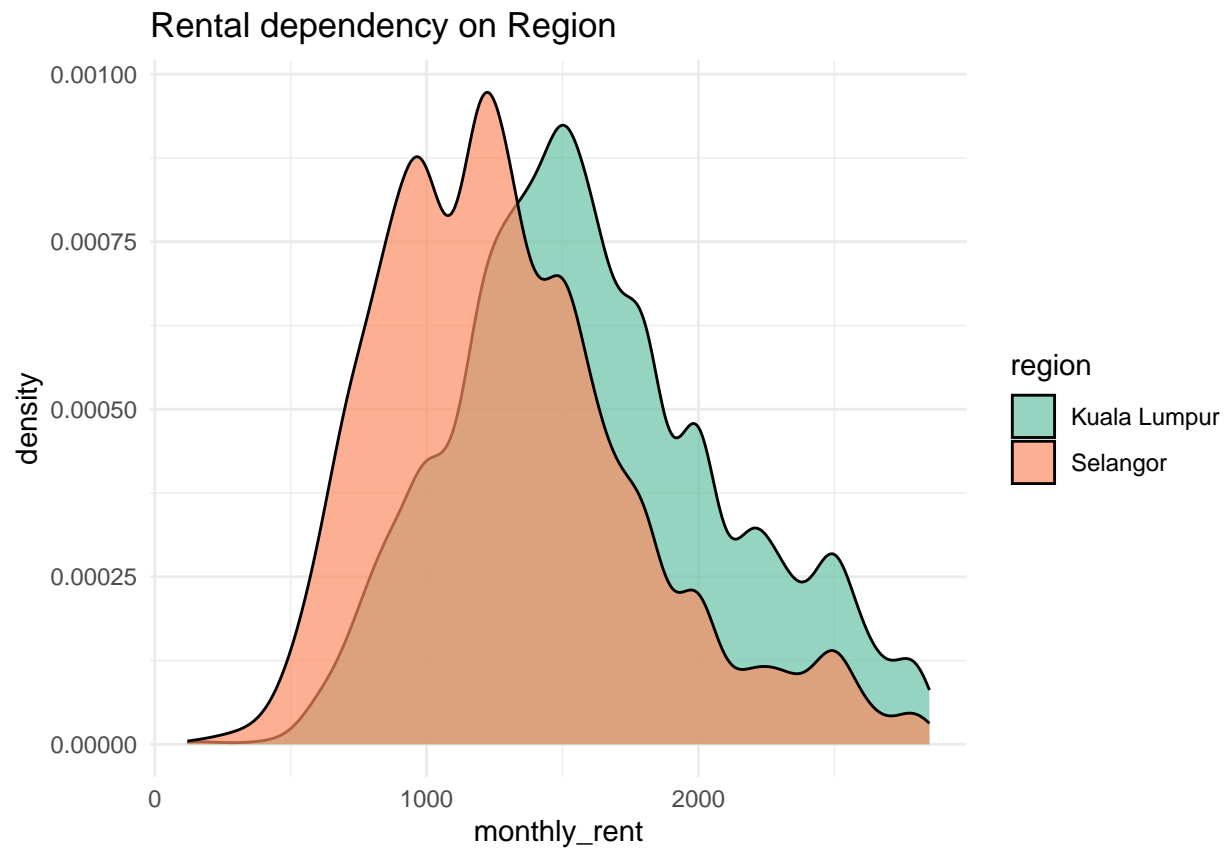
```
## Warning: Groups with fewer than two data points have been dropped.
```

```
## Warning in max(ids, na.rm = TRUE): no non-missing arguments to max; returning
## -Inf
```

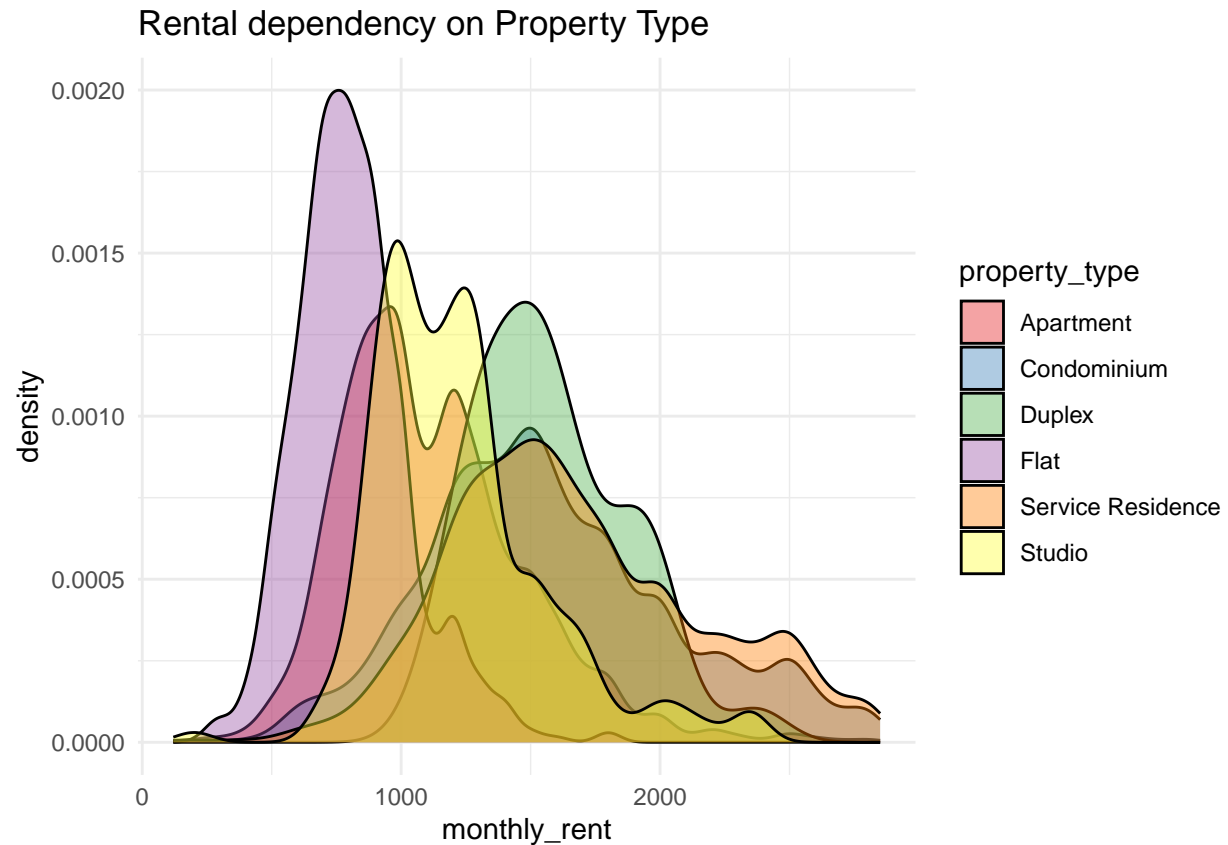
Rental dependency on Furnished Condition



```
ggplot(filteredrental, aes(x = monthly_rent, fill = region)) +  
  geom_density(alpha = 0.7) +  
  scale_fill_brewer(palette = "Set2") +  
  labs(title = "Rental dependency on Region") +  
  theme_minimal()
```



```
ggplot(apt_rental, aes(x = monthly_rent, fill = property_type)) +  
  geom_density(alpha = 0.4) +  
  scale_fill_brewer(palette = "Set1") +  
  labs(title = "Rental dependency on Property Type") +  
  theme_minimal()
```



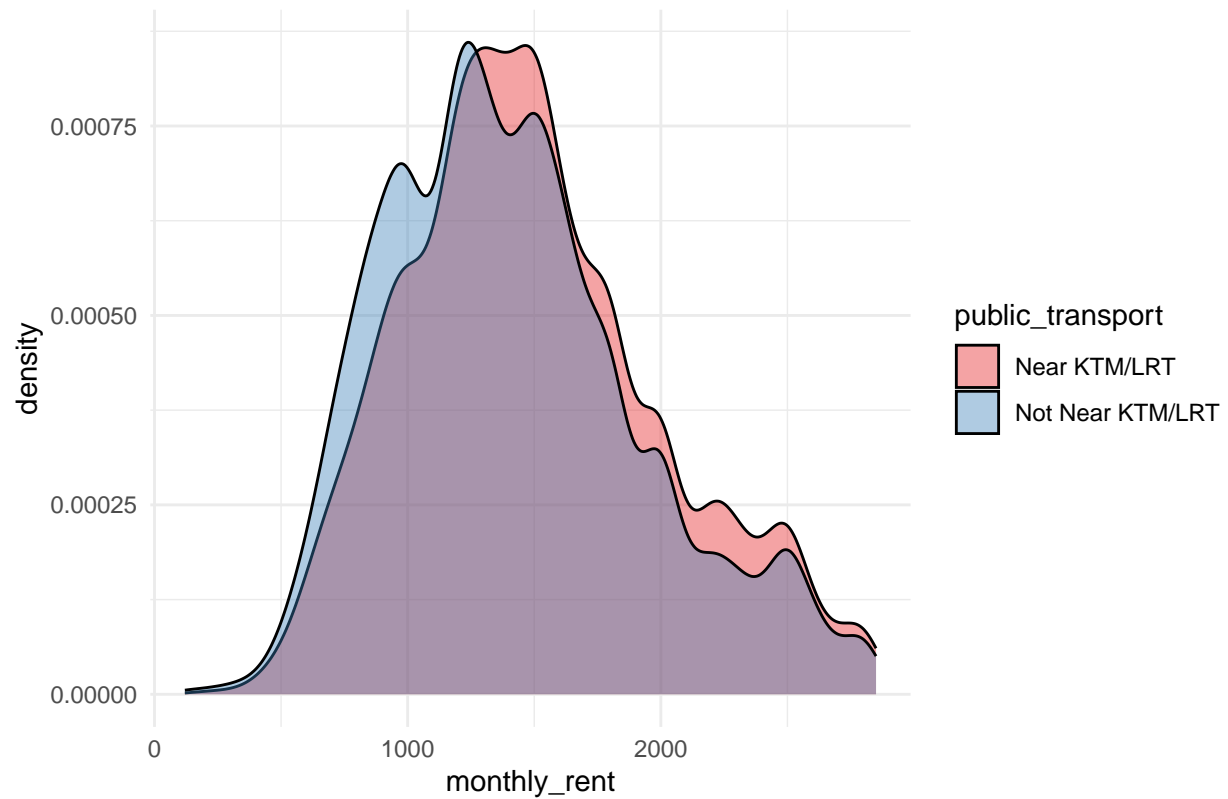
Extracting “Near KTM/LRT” information into a new column, added onto dataframe.

```
rental_publictransport = filteredrental %>%
  mutate(
    public_transport = str_extract(additional_facilities, "Near KTM/LRT"),
    public_transport = replace(public_transport, is.na(public_transport), "Not Near KTM/LRT"))

ggplot(rental_publictransport, aes(x = monthly_rent, fill = public_transport)) +
  geom_density(alpha = 0.4) +
  scale_fill_brewer(palette = "Set1") +
  labs(title = "Rental dependency on Distance to Public Transport") +
  theme_minimal()
```



## Rental dependency on Distance to Public Transport



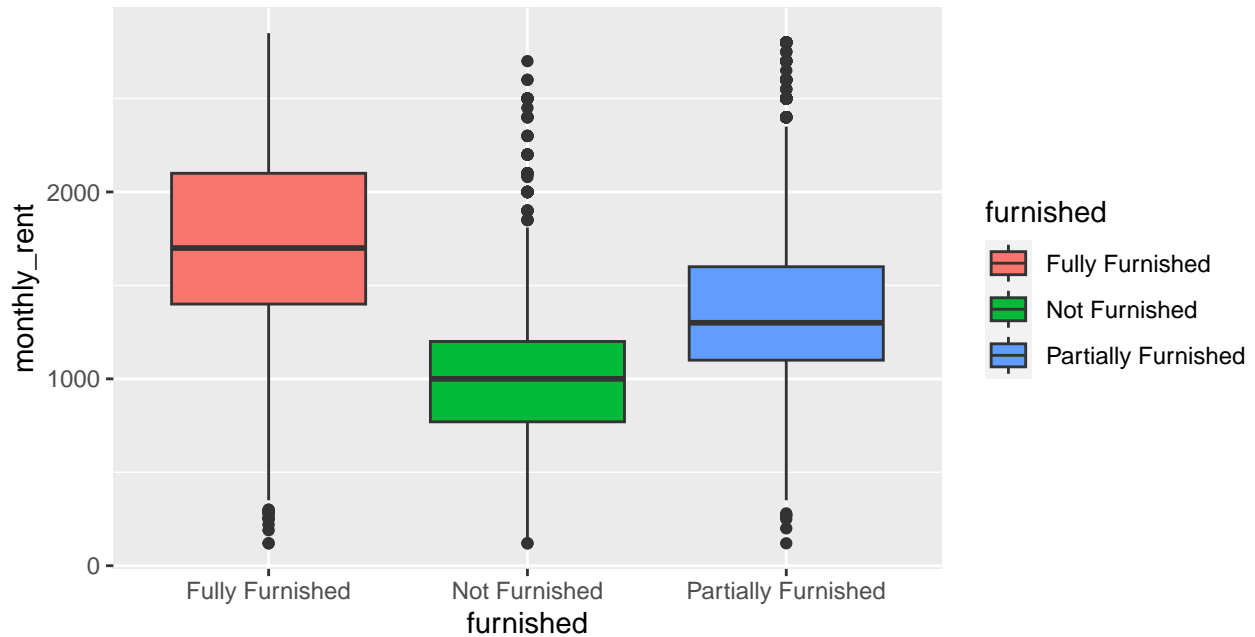
```
filteredrental %>%  
  filter(!is.na(furnished) & furnished != "") %>%  
  select(furnished, monthly_rent) %>%  
  ggplot(aes(x = furnished, y = monthly_rent, fill = furnished)) +  
  geom_boxplot() +  
  labs(title = "Boxplot of Monthly Rent with Furnished Condition",  
        subtitle = "Median rent for Fully Furnished = RM 1700 \nMedian rent for Not Furnished = RM 1000",  
        theme(plot.margin = margin(b=30))
```

## Boxplot of Monthly Rent with Furnished Condition

Median rent for Fully Furnished = RM 1700

Median rent for Not Furnished = RM 1000

Median rent for Partially Furnished = RM 1300

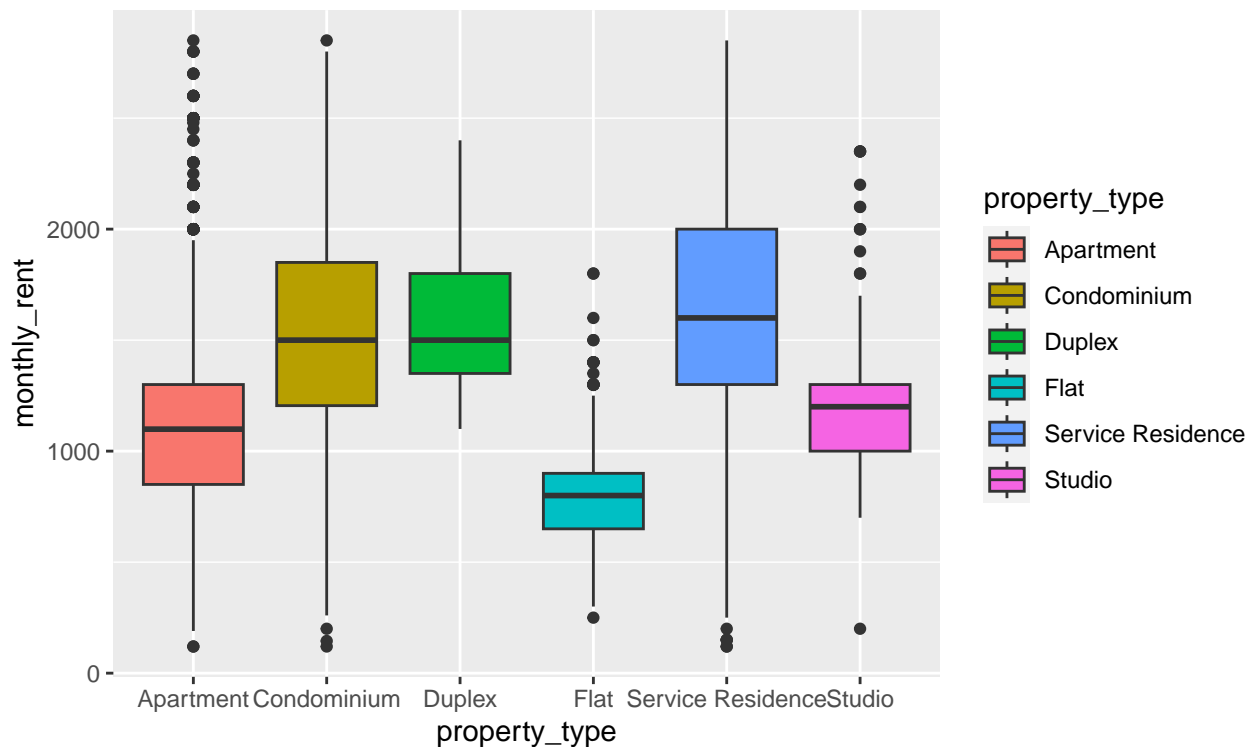


```
median_data = filteredrental %>%
  filter(!is.na(furnished) & furnished != "") %>%
  group_by(furnished) %>%
  summarize(median_rent = median(monthly_rent))
print(median_data)
```

```
## # A tibble: 3 x 2
##   furnished      median_rent
##   <chr>          <dbl>
## 1 Fully Furnished      1700
## 2 Not Furnished        1000
## 3 Partially Furnished  1300
```

```
apt_rental %>%
  select(property_type, monthly_rent) %>%
  ggplot(aes(x = property_type, y = monthly_rent, fill = property_type)) +
  geom_boxplot() +
  labs(title = "Boxplot of Monthly Rent with Property Type") +
  theme(plot.margin = margin(b=30))
```

Boxplot of Monthly Rent with Property Type



```
propertymedian_data = apt_rental %>%
  group_by(property_type) %>%
  summarize(propertymedian_rent = median(monthly_rent))
print(propertymedian_data)
```

```
## # A tibble: 6 x 2
##   property_type    propertymedian_rent
##   <chr>            <dbl>
## 1 Apartment         1099
## 2 Condominium       1500
## 3 Duplex            1500
## 4 Flat              800
## 5 Service Residence 1600
## 6 Studio            1200
```

**Insights gained from all the visuals generated** As a potential tenant, I'd always consider an affordable rent of below RM 1.5K as my budget, before deciding on a house. From this analysis, RM1.5K per month offers a lot of potential unit such as: **1)** Apartment, flat, and condominium shows a median of monthly rent close to RM 1.5k. However, service residence can be considered out of the budget. **2)** Not furnished and partially furnished rental are still within RM 1.5k budget, however a fully furnished unit median is around RM1.7k **3)** Monthly rental for units in KL are slightly higher than units in Selangor. **4)** Vicinity of public transport, whether it is near KTM/LRT or not, does not significantly affect monthly rental as much. **5)** Will proceed on using a machine learning model to explore more on variables correlation and feature importance.

## Model prediction using Random Forest

```
ohe_df = filteredrental %>%
  filter(property_type %in% c("Apartment", "Condominium", "Service Residence", "Flat", "Studio", "Duplex"))
  mutate(
    public_transport = str_extract(additional_facilities, "Near KTM/LRT"),
    public_transport = replace(public_transport, is.na(public_transport), "Not Near KTM/LRT")) %>%
  filter(complete.cases(.))

install.packages("fastDummies")
```

```
## Installing package into 'C:/Users/1/AppData/Local/R/win-library/4.3'
## (as 'lib' is unspecified)
```

```
## package 'fastDummies' successfully unpacked and MD5 sums checked
##
## The downloaded binary packages are in
## C:\Users\1\AppData\Local\Temp\RtmpALY9F6\downloaded_packages
```

```
library(fastDummies)
```

```
## Warning: package 'fastDummies' was built under R version 4.3.1
```

```
## Thank you for using fastDummies!
```

```
## To acknowledge our work, please cite the package:
```

```
## Kaplan, J. & Schlegel, B. (2023). fastDummies: Fast Creation of Dummy (Binary) Columns and Rows from
```

```
ohe_df = dummy_columns(ohe_df, select_column = c("property_type", "furnished", "region", "public_transportation"))
select(-'ads_id')
```

Removing non-numeric column

```
ohe_numerical = list()
for (col in names(ohe_df)) {
  if (is.numeric(ohe_df[[col]]))
    ohe_numerical[[col]] = ohe_df[[col]]
}
ohe_num_df = data.frame(ohe_numerical)
```

```
cor(ohe_num_df)
```

```
##
## completion_year  completion_year  monthly_rent  parking
## completion_year      1.000000000  0.156476519  0.302906721
## monthly_rent         0.156476519  1.000000000  0.236038936
## parking              0.302906721  0.236038936  1.000000000
## bathroom            -0.055847337  0.243541825  0.308305121
## size                 -0.212728460  0.336251870  0.326509749
```

## property_type_Apartment	-0.077835651	-0.228439517	-0.090615866
## property_type_Condominium	-0.213141725	0.090509125	0.064336949
## property_type_Duplex	-0.013414094	-0.007075511	-0.024944866
## property_type_Flat	-0.037364657	-0.053604933	-0.008434526
## property_type_Service.Residence	0.285258181	0.091137882	0.016522652
## property_type_Studio	0.003068963	-0.073679369	-0.070723932
## furnished_Fully.Furnished	-0.030513273	0.399213816	-0.037552978
## furnished_Not.Furnished	-0.019914078	-0.301919267	-0.053978944
## furnished_Partially.Furnished	0.043506532	-0.212670940	0.072224913
## region_Kuala.Lumpur	0.026772980	0.355657928	-0.117887470
## region_Selangor	-0.026772980	-0.355657928	0.117887470
## public_transport_Near.KTM.LRT	0.061270312	0.045134140	-0.012382742
## public_transport_Not.Near.KTM.LRT	-0.061270312	-0.045134140	0.012382742
##	bathroom	size	
## completion_year	-5.584734e-02	-0.21272846	
## monthly_rent	2.435418e-01	0.33625187	
## parking	3.083051e-01	0.32650975	
## bathroom	1.000000e+00	0.68705259	
## size	6.870526e-01	1.00000000	
## property_type_Apartment	5.399443e-02	-0.05873774	
## property_type_Condominium	1.830820e-01	0.32423788	
## property_type_Duplex	-6.235472e-05	-0.02073657	
## property_type_Flat	-1.981827e-02	-0.02961620	
## property_type_Service.Residence	-1.958776e-01	-0.26250571	
## property_type_Studio	-1.732475e-01	-0.15690403	
## furnished_Fully.Furnished	-1.665083e-01	-0.11432207	
## furnished_Not.Furnished	2.501572e-02	-0.02912032	
## furnished_Partially.Furnished	1.526699e-01	0.13416720	
## region_Kuala.Lumpur	1.723227e-01	0.13232738	
## region_Selangor	-1.723227e-01	-0.13232738	
## public_transport_Near.KTM.LRT	5.954890e-03	-0.03088843	
## public_transport_Not.Near.KTM.LRT	-5.954890e-03	0.03088843	
##	property_type_Apartment		
## completion_year	-0.07783565		
## monthly_rent	-0.22843952		
## parking	-0.09061587		
## bathroom	0.05399443		
## size	-0.05873774		
## property_type_Apartment	1.00000000		
## property_type_Condominium	-0.40806530		
## property_type_Duplex	-0.02324636		
## property_type_Flat	-0.01228613		
## property_type_Service.Residence	-0.28973753		
## property_type_Studio	-0.03957417		
## furnished_Fully.Furnished	-0.14041567		
## furnished_Not.Furnished	0.16249386		
## furnished_Partially.Furnished	0.03911515		
## region_Kuala.Lumpur	-0.03012189		
## region_Selangor	0.03012189		
## public_transport_Near.KTM.LRT	-0.01910161		
## public_transport_Not.Near.KTM.LRT	0.01910161		
##	property_type_Condominium		
## completion_year	-0.21314172		
## monthly_rent	0.09050912		

## parking	0.06433695
## bathroom	0.18308203
## size	0.32423788
## property_type_Apartment	-0.40806530
## property_type_Condominium	1.00000000
## property_type_Duplex	-0.05832651
## property_type_Flat	-0.03082664
## property_type_Service.Residence	-0.72696903
## property_type_Studio	-0.09929399
## furnished_Fully.Furnished	0.01750938
## furnished_Not.Furnished	-0.05456261
## furnished_Partially.Furnished	0.01686493
## region_Kuala.Lumpur	0.05064233
## region_Selangor	-0.05064233
## public_transport_Near.KTM.LRT	-0.06424446
## public_transport_Not.Near.KTM.LRT	0.06424446
##	
## completion_year	property_type_Duplex property_type_Flat -1.341409e-02 -0.037364657
## monthly_rent	-7.075511e-03 -0.053604933
## parking	-2.494487e-02 -0.008434526
## bathroom	-6.235472e-05 -0.019818267
## size	-2.073657e-02 -0.029616200
## property_type_Apartment	-2.324636e-02 -0.012286127
## property_type_Condominium	-5.832651e-02 -0.030826638
## property_type_Duplex	1.000000e+00 -0.001756109
## property_type_Flat	-1.756109e-03 1.000000000
## property_type_Service.Residence	-4.141342e-02 -0.021887757
## property_type_Studio	-5.656505e-03 -0.002989567
## furnished_Fully.Furnished	-8.326850e-03 -0.020093195
## furnished_Not.Furnished	1.662659e-02 0.044979128
## furnished_Partially.Furnished	-2.111596e-03 -0.008174934
## region_Kuala.Lumpur	9.632494e-03 -0.031487338
## region_Selangor	-9.632494e-03 0.031487338
## public_transport_Near.KTM.LRT	-9.750312e-03 -0.007282756
## public_transport_Not.Near.KTM.LRT	9.750312e-03 0.007282756
##	
## completion_year	property_type_Service.Residence 0.28525818
## monthly_rent	0.09113788
## parking	0.01652265
## bathroom	-0.19587765
## size	-0.26250571
## property_type_Apartment	-0.28973753
## property_type_Condominium	-0.72696903
## property_type_Duplex	-0.04141342
## property_type_Flat	-0.02188776
## property_type_Service.Residence	1.00000000
## property_type_Studio	-0.07050145
## furnished_Fully.Furnished	0.07652797
## furnished_Not.Furnished	-0.06189621
## furnished_Partially.Furnished	-0.03822053
## region_Kuala.Lumpur	-0.01252240
## region_Selangor	0.01252240
## public_transport_Near.KTM.LRT	0.08798382
## public_transport_Not.Near.KTM.LRT	-0.08798382

##	property_type_Studio
## completion_year	0.003068963
## monthly_rent	-0.073679369
## parking	-0.070723932
## bathroom	-0.173247469
## size	-0.156904033
## property_type_Apartment	-0.039574170
## property_type_Condominium	-0.099293992
## property_type_Duplex	-0.005656505
## property_type_Flat	-0.002989567
## property_type_Service.Residence	-0.070501451
## property_type_Studio	1.000000000
## furnished_Fully.Furnished	0.049162540
## furnished_Not.Furnished	-0.021389019
## furnished_Partially.Furnished	-0.036200358
## region_Kuala.Lumpur	-0.087785523
## region_Selangor	0.087785523
## public_transport_Near.KTM.LRT	-0.022267228
## public_transport_Not.Near.KTM.LRT	0.022267228
##	furnished_Fully.Furnished
## completion_year	-0.03051327
## monthly_rent	0.39921382
## parking	-0.03755298
## bathroom	-0.16650829
## size	-0.11432207
## property_type_Apartment	-0.14041567
## property_type_Condominium	0.01750938
## property_type_Duplex	-0.00832685
## property_type_Flat	-0.02009319
## property_type_Service.Residence	0.07652797
## property_type_Studio	0.04916254
## furnished_Fully.Furnished	1.00000000
## furnished_Not.Furnished	-0.33215999
## furnished_Partially.Furnished	-0.80157208
## region_Kuala.Lumpur	0.02560502
## region_Selangor	-0.02560502
## public_transport_Near.KTM.LRT	0.05292138
## public_transport_Not.Near.KTM.LRT	-0.05292138
##	furnished_Not.Furnished
## completion_year	-0.01991408
## monthly_rent	-0.30191927
## parking	-0.05397894
## bathroom	0.02501572
## size	-0.02912032
## property_type_Apartment	0.16249386
## property_type_Condominium	-0.05456261
## property_type_Duplex	0.01662659
## property_type_Flat	0.04497913
## property_type_Service.Residence	-0.06189621
## property_type_Studio	-0.02138902
## furnished_Fully.Furnished	-0.33215999
## furnished_Not.Furnished	1.00000000
## furnished_Partially.Furnished	-0.29770116
## region_Kuala.Lumpur	-0.06034122

## region_Selangor	0.06034122
## public_transport_Near.KTM.LRT	-0.08211101
## public_transport_Not.Near.KTM.LRT	0.08211101
##	furnished_Partially.Furnished
## completion_year	0.043506532
## monthly_rent	-0.212670940
## parking	0.072224913
## bathroom	0.152669936
## size	0.134167196
## property_type_Apartment	0.039115154
## property_type_Condominium	0.016864935
## property_type_Duplex	-0.002111596
## property_type_Flat	-0.008174934
## property_type_Service.Residence	-0.038220534
## property_type_Studio	-0.036200358
## furnished_Fully.Furnished	-0.801572078
## furnished_Not.Furnished	-0.297701158
## furnished_Partially.Furnished	1.000000000
## region_Kuala.Lumpur	0.012334133
## region_Selangor	-0.012334133
## public_transport_Near.KTM.LRT	-0.001513808
## public_transport_Not.Near.KTM.LRT	0.001513808
##	region_Kuala.Lumpur region_Selangor
## completion_year	0.026772980 -0.026772980
## monthly_rent	0.355657928 -0.355657928
## parking	-0.117887470 0.117887470
## bathroom	0.172322725 -0.172322725
## size	0.132327378 -0.132327378
## property_type_Apartment	-0.030121892 0.030121892
## property_type_Condominium	0.050642331 -0.050642331
## property_type_Duplex	0.009632494 -0.009632494
## property_type_Flat	-0.031487338 0.031487338
## property_type_Service.Residence	-0.012522397 0.012522397
## property_type_Studio	-0.087785523 0.087785523
## furnished_Fully.Furnished	0.025605022 -0.025605022
## furnished_Not.Furnished	-0.060341223 0.060341223
## furnished_Partially.Furnished	0.012334133 -0.012334133
## region_Kuala.Lumpur	1.000000000 -1.000000000
## region_Selangor	-1.000000000 1.000000000
## public_transport_Near.KTM.LRT	0.140421382 -0.140421382
## public_transport_Not.Near.KTM.LRT	-0.140421382 0.140421382
##	public_transport_Near.KTM.LRT
## completion_year	0.061270312
## monthly_rent	0.045134140
## parking	-0.012382742
## bathroom	0.005954890
## size	-0.030888429
## property_type_Apartment	-0.019101610
## property_type_Condominium	-0.064244459
## property_type_Duplex	-0.009750312
## property_type_Flat	-0.007282756
## property_type_Service.Residence	0.087983822
## property_type_Studio	-0.022267228
## furnished_Fully.Furnished	0.052921381



```

## furnished_Not.Furnished -0.082111007
## furnished_Partially.Furnished -0.001513808
## region_Kuala.Lumpur 0.140421382
## region_Selangor -0.140421382
## public_transport_Near.KTM.LRT 1.000000000
## public_transport_Not.Near.KTM.LRT -1.000000000
## public_transport_Not.Near.KTM.LRT
## completion_year -0.061270312
## monthly_rent -0.045134140
## parking 0.012382742
## bathroom -0.005954890
## size 0.030888429
## property_type_Apartment 0.019101610
## property_type_Condominium 0.064244459
## property_type_Duplex 0.009750312
## property_type_Flat 0.007282756
## property_type_Service.Residence -0.087983822
## property_type_Studio 0.022267228
## furnished_Fully.Furnished -0.052921381
## furnished_Not.Furnished 0.082111007
## furnished_Partially.Furnished 0.001513808
## region_Kuala.Lumpur -0.140421382
## region_Selangor 0.140421382
## public_transport_Near.KTM.LRT -1.000000000
## public_transport_Not.Near.KTM.LRT 1.000000000

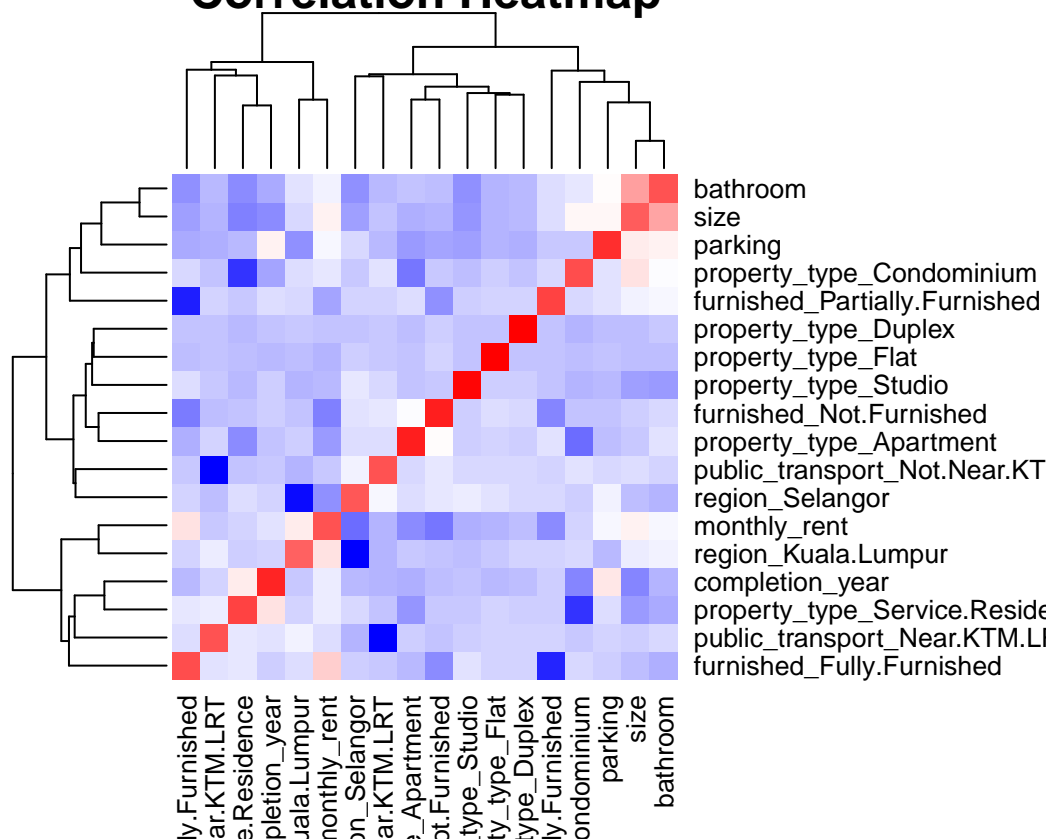
```

```

correlation_table = cor(ohe_num_df)
cor_df = data.frame(variable = colnames(ohe_num_df), correlation = correlation_table)
heatmap(correlation_table,
        col = colorRampPalette(c("blue", "white", "red"))(100),
        main = "Correlation Heatmap",
        xlab = "",
        ylab = "")

```

## Correlation Heatmap



Train model

```
install.packages("randomForest")
```

```
## Installing package into 'C:/Users/1/AppData/Local/R/win-library/4.3'
## (as 'lib' is unspecified)
```

```
## package 'randomForest' successfully unpacked and MD5 sums checked
```

```
## Warning: cannot remove prior installation of package 'randomForest'
```

```
## Warning in file.copy(savedcopy, lib, recursive = TRUE): problem copying
## C:\Users\1\AppData\Local\R\win-library\4.3\00LOCK\randomForest\libs\x64\randomForest.dll
## to
## C:\Users\1\AppData\Local\R\win-library\4.3\randomForest\libs\x64\randomForest.dll:
## Permission denied
```

```
## Warning: restored 'randomForest'
```

```
##
## The downloaded binary packages are in
## C:\Users\1\AppData\Local\Temp\RtmpAly9F6\downloaded_packages
```

```
library(randomForest)
```

```
## Warning: package 'randomForest' was built under R version 4.3.1
```

```
## randomForest 4.7-1.1
```

```
## Type rfNews() to see new features/changes/bug fixes.
```

```
##
```

```
## Attaching package: 'randomForest'
```

```
## The following object is masked from 'package:dplyr':
```

```
##
```

```
##      combine
```

```
## The following object is masked from 'package:ggplot2':
```

```
##
```

```
##      margin
```

```
set.seed(123)
```

```
train_index = sample(1:nrow(ohe_num_df), 0.8 * nrow(ohe_num_df))
```

```
train_data = ohe_num_df[train_index, ]
```

```
test_data = ohe_num_df[-train_index, ]
```

```
# Define the target variable and predictors
```

```
target_variable = "monthly_rent"
```

```
predictors = setdiff(names(ohe_num_df), target_variable)
```

```
# Train a Random Forest model
```

```
rental_model = randomForest(formula(paste(target_variable, "~", paste(predictors, collapse = "+"))),  
                             data = train_data)
```

```
# Print the model summary
```

```
print(rental_model)
```

```
##
```

```
## Call:
```

```
## randomForest(formula = formula(paste(target_variable, "~", paste(predictors, collapse = "+")))
```

```
##           Type of random forest: regression
```

```
##           Number of trees: 500
```

```
## No. of variables tried at each split: 5
```

```
##
```

```
##           Mean of squared residuals: 84714.21
```

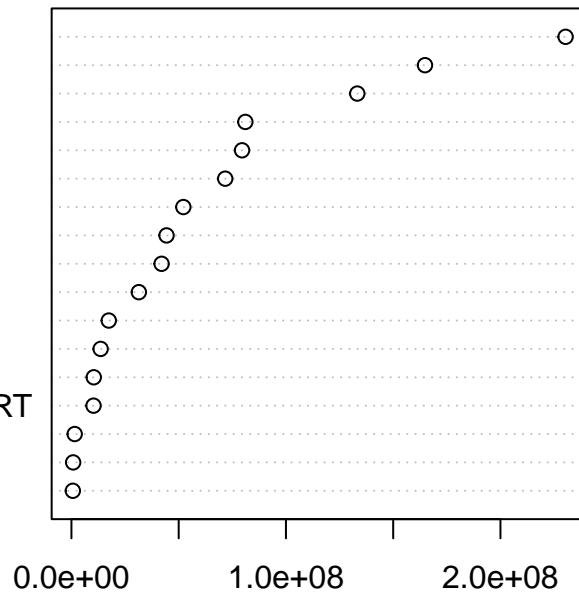
```
##           % Var explained: 63.45
```

Evaluating feature importance on targeted variable of 'monthly rent'

```
varImpPlot(rental_model)
```

## rental\_model

size  
furnished\_Fully.Furnished  
completion\_year  
region\_Selangor  
region\_Kuala.Lumpur  
parking  
furnished\_Not.Furnished  
bathroom  
furnished\_Partially.Furnished  
property\_type\_Apartment  
property\_type\_Service.Residence  
property\_type\_Condominium  
public\_transport\_Near.KTM.LRT  
public\_transport\_Not.Near.KTM.LRT  
property\_type\_Studio  
property\_type\_Duplex  
property\_type\_Flat



IncNodePurity

```
prediction = predict(rental_model, newdata = test_data)

actual = test_data$monthly_rent
ss_total = sum((actual - mean(actual))^2)
ss_residual = sum((actual - prediction)^2)
r_squared = 1 - (ss_residual / ss_total)

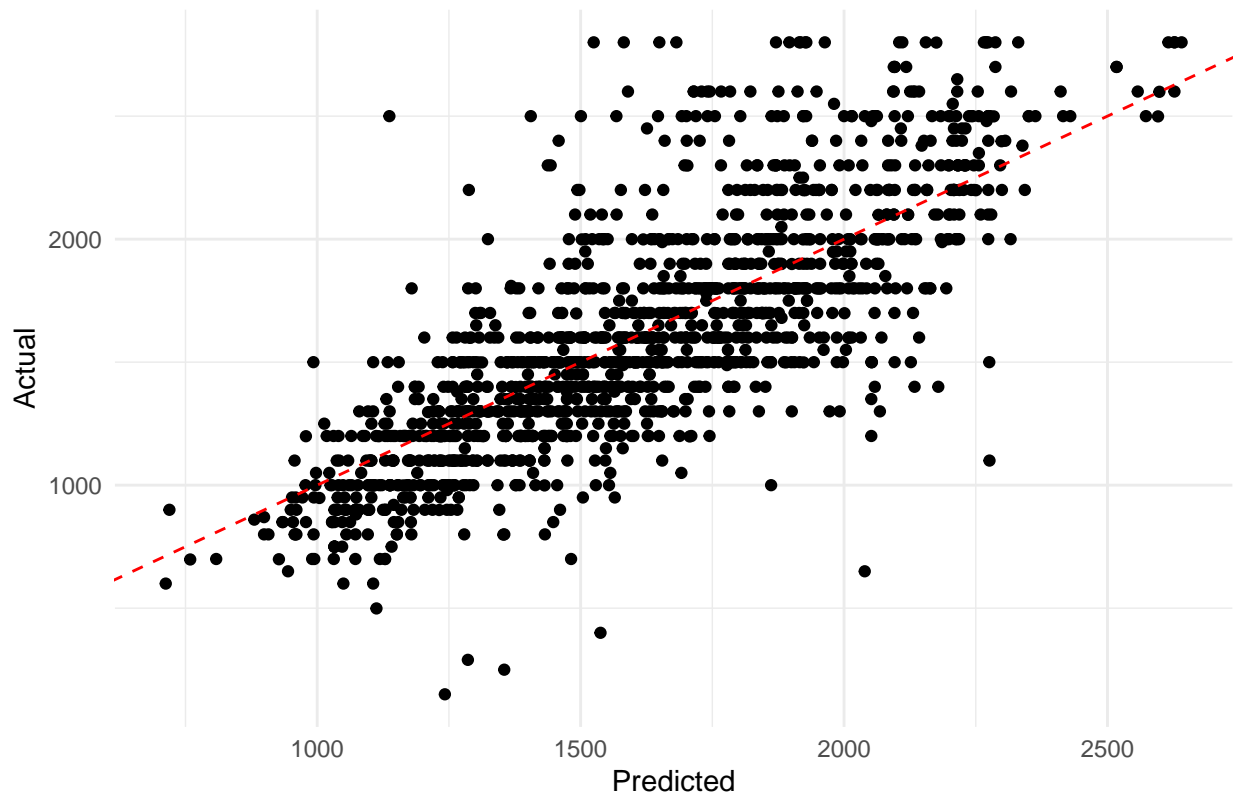
print(r_squared)
```

```
## [1] 0.6225412
```

```
# Create a data frame for plotting
prediction_df = data.frame(Actual = actual, Predicted = prediction)

# Create Scatter Plot
ggplot(prediction_df, aes(x = Predicted, y = Actual)) +
  geom_point() +
  geom_abline(intercept = 0, slope = 1, color = "red", linetype = "dashed") + # Add diagonal line
  labs(title = "Predicted vs Actual Values", x = "Predicted", y = "Actual") +
  theme_minimal()
```

## Predicted vs Actual Values



```
head(prediction_df)
```

```
##      Actual Predicted
## 4      2000    2031.451
## 10     1199    1473.708
## 11     1100    1295.426
## 12     2500    2166.986
## 14     1400    1726.554
## 15     1800    1474.362
```

## Conclusion

Random Forest model r-squared value is at 0.6225412, which only represent 60% of accuracy instead of 100%. Model may be needed to be revised, improved or undergo hyperparameter tuning (of which I'm still exploring). As for now, this model has illustrated that top 3 variables that highly affected monthly rent in Klang Valley is the unit's size, it's furnished condition and the year it's build. Personally, I always thought that public transport facilities (whether its within/outside vicinity) will highly affect rental price, but my analysis shows the contrast. This also shows that Klang Valley is definitely a city of cars, and most its potential residents are car owner, hence why it doesnt affect rental price.