

Marketing Project

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June 14, 2019

```
options(repos=structure(c(CRAN="http://cran.utstat.utoronto.ca/")))
install.packages("magrittr")

## package 'magrittr' successfully unpacked and MD5 sums checked
##
## The downloaded binary packages are in
## C:\Users\soudik\AppData\Local\Temp\Rtmp082IkN\downloaded_packages

install.packages("tidyverse")

## package 'tidyverse' successfully unpacked and MD5 sums checked
##
## The downloaded binary packages are in
## C:\Users\soudik\AppData\Local\Temp\Rtmp082IkN\downloaded_packages

install.packages("ggplot2")

## package 'ggplot2' successfully unpacked and MD5 sums checked
##
## The downloaded binary packages are in
## C:\Users\soudik\AppData\Local\Temp\Rtmp082IkN\downloaded_packages

library(magrittr)

## Warning: package 'magrittr' was built under R version 3.5.3

library(tidyverse)

## Warning: package 'tidyverse' was built under R version 3.5.3

## -- Attaching packages -----
## ----- tidyverse 1.2.1 -----

## v ggplot2 3.2.1      v purrr  0.3.2
## v tibble  2.1.3      v dplyr  0.8.3
## v tidyr   1.0.0      v stringr 1.4.0
## v readr   1.3.1      v forcats 0.4.0

## Warning: package 'ggplot2' was built under R version 3.5.3
## Warning: package 'tibble' was built under R version 3.5.3
## Warning: package 'tidyr' was built under R version 3.5.3
## Warning: package 'readr' was built under R version 3.5.3
```

```
## Warning: package 'purrr' was built under R version 3.5.3
## Warning: package 'dplyr' was built under R version 3.5.3
## Warning: package 'stringr' was built under R version 3.5.3
## Warning: package 'forcats' was built under R version 3.5.3
```

```
## -- Conflicts -----
----- tidyverse_conflicts() --
## x tidyr::extract() masks magrittr::extract()
## x dplyr::filter() masks stats::filter()
## x dplyr::lag() masks stats::lag()
## x purrr::set_names() masks magrittr::set_names()
```

```
library(ggplot2)
```

Importing the file

```
mydata<-read.csv("K:/_Staff/souidik/Bookings_data.csv", header =TRUE,
stringsAsFactors = FALSE )
```

Summary of the file

```
head(mydata)
```

```
##   ID ADR  hotel_id city_id star_rating accomadation_type_name chain_hotel
## 1  1  71   297,388   9,395         2.5             Hotel non-chain
## 2  2  77   298,322   9,395         3.0             Hotel non-chain
## 3  3 154  2,313,076   9,395         5.0             Hotel chain
## 4  4 127  2,240,838   9,395         3.5             Hotel non-chain
## 5  5 115  2,240,838   9,395         3.5             Hotel non-chain
## 6  6  82   331,350   9,395         3.0             Hotel non-chain
##   booking_date checkin_date checkout_date City time_until_checkin
## 1  02/08/2016   01/10/2016   02/10/2016   A              60
## 2  02/08/2016   01/10/2016   02/10/2016   A              60
## 3  02/08/2016   01/10/2016   02/10/2016   A              60
## 4  04/08/2016   02/10/2016   03/10/2016   A              59
## 5  04/08/2016   02/10/2016   03/10/2016   A              59
## 6  04/08/2016   03/10/2016   05/10/2016   A              60
##   Lengh_of_stay RevPR booking_weekend checkin_weekend dow_check_in
## 1             1    71      Week-Day      Week-End      Saturday
## 2             1    77      Week-Day      Week-End      Saturday
## 3             1   154      Week-Day      Week-End      Saturday
## 4             1   127      Week-Day      Week-End      Sunday
## 5             1   115      Week-Day      Week-End      Sunday
## 6             2   163      Week-Day      Week-Day      Monday
```

```
summary(mydata$ADR)
```

```
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##    4.0   63.0   114.0  148.1  192.0  3157.0
```

```
str(mydata)
```

```
## 'data.frame':    49064 obs. of  17 variables:
## $ ID              : chr  "1" "2" "3" "4" ...
## $ ADR              : int   71 77 154 127 115 82 77 168 48 82 ...
## $ hotel_id         : chr  "297,388" "298,322" "2,313,076"
##                    "2,240,838" ...
## $ city_id          : chr  "9,395" "9,395" "9,395" "9,395" ...
## $ star_rating       : num   2.5 3 5 3.5 3.5 3 3 5 2 3 ...
## $ accommodation_type_name: chr  "Hotel" "Hotel" "Hotel" "Hotel" ...
## $ chain_hotel       : chr  "non-chain" "non-chain" "chain" "non-
##                    chain" ...
## $ booking_date      : chr  "02/08/2016" "02/08/2016" "02/08/2016"
##                    "04/08/2016" ...
## $ checkin_date      : chr  "01/10/2016" "01/10/2016" "01/10/2016"
##                    "02/10/2016" ...
## $ checkout_date     : chr  "02/10/2016" "02/10/2016" "02/10/2016"
##                    "03/10/2016" ...
## $ City              : chr  "A" "A" "A" "A" ...
## $ time_until_checkin : int   60 60 60 59 59 60 60 59 58 57 ...
## $ Lenght_of_stay    : int    1 1 1 1 1 2 1 3 1 3 ...
## $ RevPR             : int   71 77 154 127 115 163 77 505 48 245 ...
## $ booking_weekend   : chr  "Week-Day" "Week-Day" "Week-Day" "Week-
##                    Day" ...
## $ checkin_weekend    : chr  "Week-End" "Week-End" "Week-End" "Week-
##                    End" ...
## $ dow_check_in      : chr  "Saturday" "Saturday" "Saturday" "Sunday"
##                    ...
```

Formatting some variables

```
mydata$ADR<-as.numeric(mydata$ADR)
mydata$time_until_checkin<-as.numeric(mydata$time_until_checkin)
city<-as.factor(mydata$City)
checkin_date<-as.Date(mydata$checkin_date)
booking_date<-as.Date(mydata$booking_date)
checkout_date<-as.Date(mydata$checkout_date)
```

Cheking format of Data variables

```
class(checkin_date)

## [1] "Date"

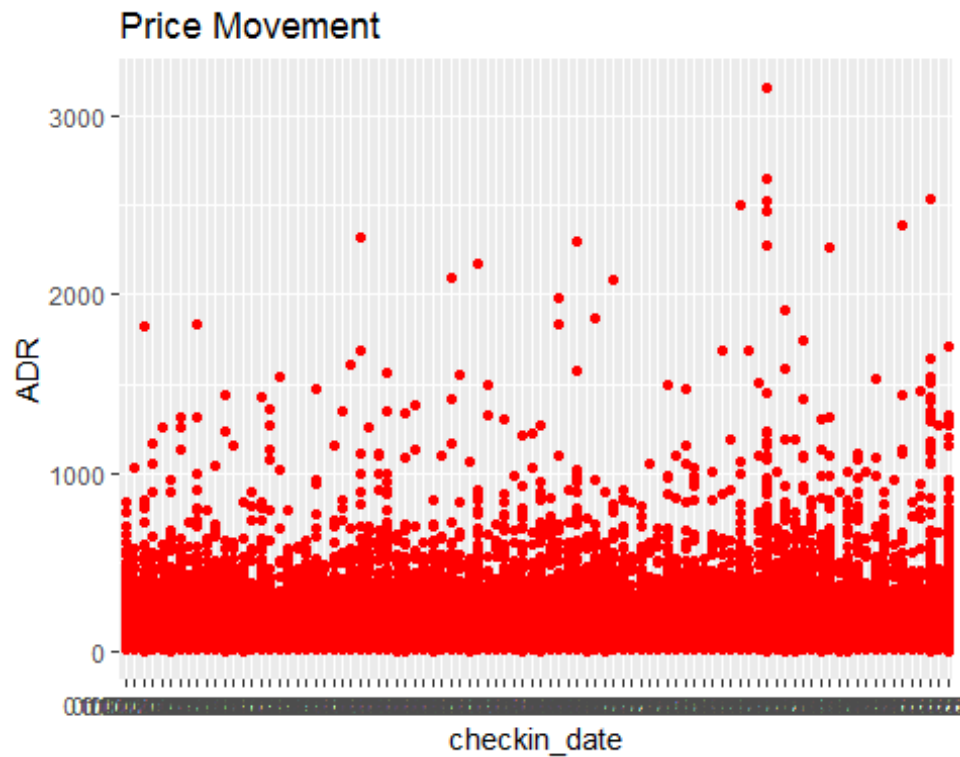
class(checkin_date)

## [1] "Date"

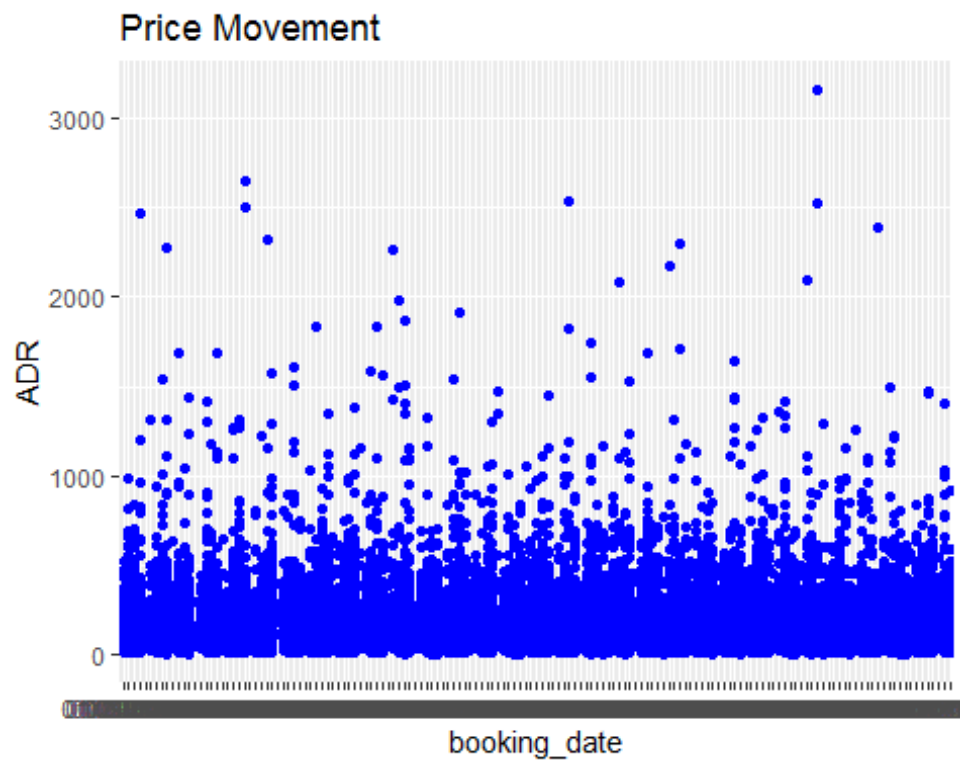
class(checkin_date)

## [1] "Date"

ggplot2::ggplot(data=mydata, aes(x = checkin_date, y = ADR)) +
  geom_point(color = "red") +
  labs(title = "Price Movement")
```



```
ggplot2::ggplot(data=mydata, aes(x = booking_date, y = ADR)) +  
  geom_point(color = "blue") +  
  labs(title = "Price Movement")
```



Starting to build a demand function for hotels

```
demand<-mydata %>%  
  dplyr::group_by(ADR) %>%  
  dplyr::summarise(number = n())
```

```
summary(demand)
```

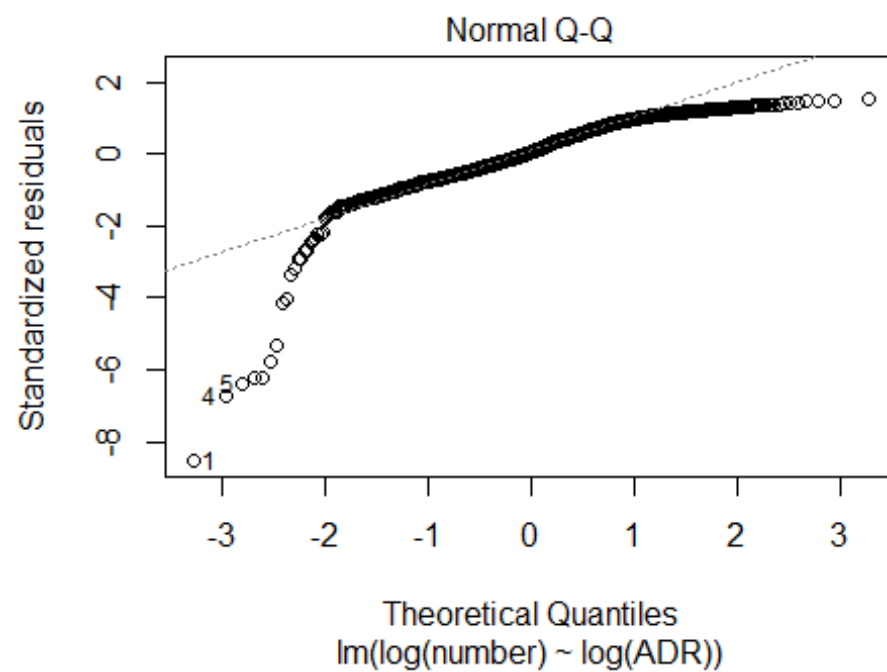
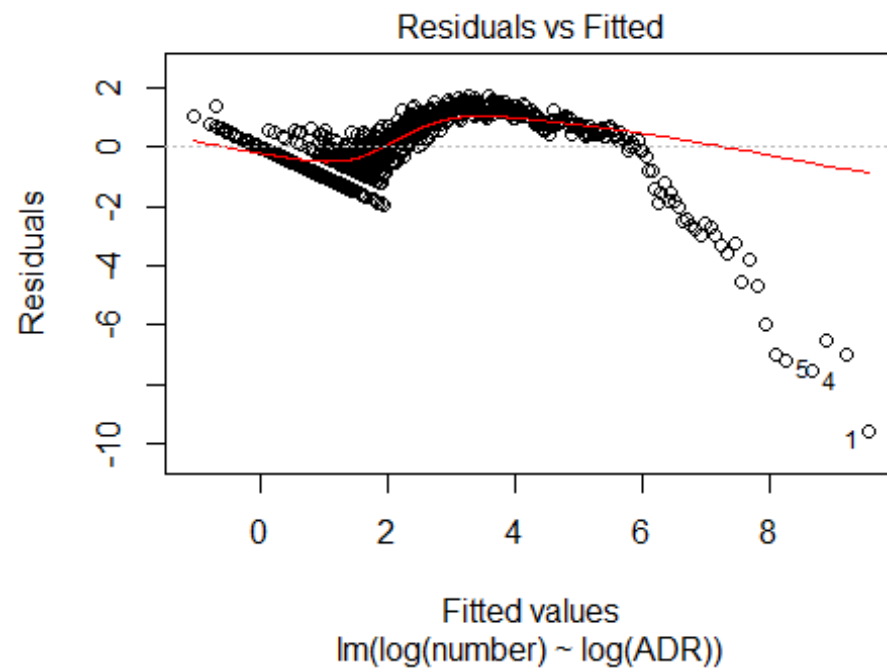
```
##      ADR      number  
##  Min.   :  4.0   Min.   :  1.00  
## 1st Qu.: 241.8   1st Qu.:  2.00  
##  Median : 478.5   Median :  7.00  
##   Mean  : 561.9   Mean    : 51.76  
## 3rd Qu.: 740.2   3rd Qu.: 62.00  
##   Max.  :3157.0   Max.    :483.00
```

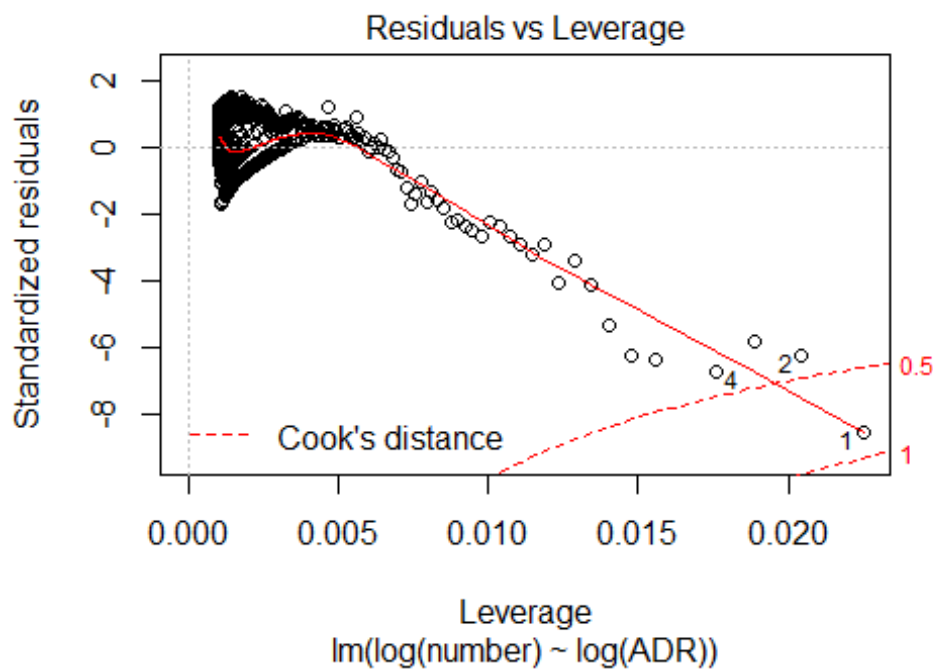
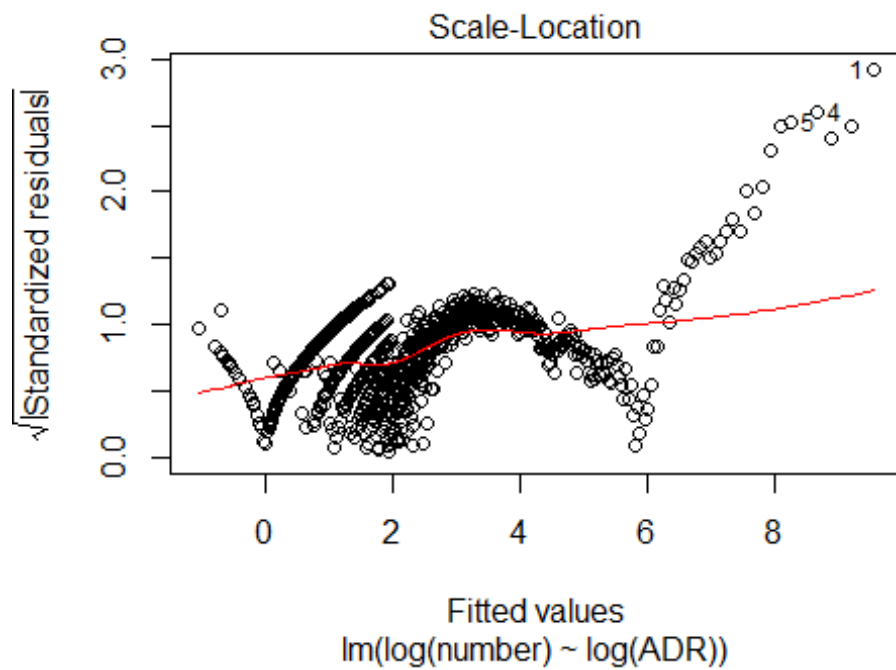
The demand function

```
demand_func<-lm(log(number)~log(ADR), data = demand)  
summary(demand_func)
```

```
##  
## Call:  
## lm(formula = log(number) ~ log(ADR), data = demand)  
##  
## Residuals:  
##      Min       1Q   Median       3Q      Max   
## -9.5515 -0.6202  0.0500  0.8261  1.7125   
##  
## Coefficients:  
##              Estimate Std. Error t value Pr(>|t|)      
## (Intercept) 11.75560    0.21944   53.57  <2e-16 ***  
## log(ADR)     -1.58991    0.03632  -43.78  <2e-16 ***  
## ---  
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1  
##  
## Residual standard error: 1.133 on 946 degrees of freedom  
## Multiple R-squared:  0.6695, Adjusted R-squared:  0.6692   
## F-statistic: 1917 on 1 and 946 DF,  p-value: < 2.2e-16
```

```
plot(demand_func)
```





Dedmand function by city

```
demand_by_city <- mydata %>%
  group_by(City, ADR) %>%
```

```

      tally()
head(demand_by_city)

## # A tibble: 6 x 3
## # Groups:   City [1]
##   City     ADR     n
##   <chr> <dbl> <int>
## 1 A         4         1
## 2 A         5         9
## 3 A         6        11
## 4 A         7         2
## 5 A         9         3
## 6 A        10         1

```

Dedmand function by city Model

```

demand_by_city%>%
  filter(City == "A")%>%
  lm(log(n)~log(ADR), data =.)%>%
  summary()

##
## Call:
## lm(formula = log(n) ~ log(ADR), data = .)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -7.7534 -0.8153  0.0961  1.1376  2.1003
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   9.63815     0.35807   26.92  <2e-16 ***
## log(ADR)      -1.35959     0.06761  -20.11  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.341 on 460 degrees of freedom
## Multiple R-squared:  0.4678, Adjusted R-squared:  0.4667
## F-statistic: 404.4 on 1 and 460 DF,  p-value: < 2.2e-16

demand_by_city%>%
  filter(City == "B")%>%
  lm(log(n)~log(ADR), data =.)%>%
  summary()

##
## Call:
## lm(formula = log(n) ~ log(ADR), data = .)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -5.1640 -0.5023  0.0563  0.5979  1.5346

```



```
##
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)
## (Intercept)  7.27585    0.21152   34.40  <2e-16 ***
## log(ADR)     -1.08529    0.03857  -28.14  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.8403 on 477 degrees of freedom
## Multiple R-squared:  0.624, Adjusted R-squared:  0.6232
## F-statistic: 791.6 on 1 and 477 DF,  p-value: < 2.2e-16

demand_by_city%>%
  filter(City == "D")%>%
  lm(log(n)~log(ADR), data =.)%>%
  summary()

##
## Call:
## lm(formula = log(n) ~ log(ADR), data = .)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.7375 -0.7443 -0.0474  0.9692  2.2710
##
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)
## (Intercept)  7.29191    0.27238   26.77  <2e-16 ***
## log(ADR)     -0.94327    0.04719  -19.99  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.105 on 667 degrees of freedom
## Multiple R-squared:  0.3746, Adjusted R-squared:  0.3737
## F-statistic: 399.6 on 1 and 667 DF,  p-value: < 2.2e-16

demand_by_city%>%
  filter(City == "E")%>%
  lm(log(n)~log(ADR), data =.)%>%
  summary()

##
## Call:
## lm(formula = log(n) ~ log(ADR), data = .)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.4376 -0.4492  0.0231  0.5719  1.6437
##
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)
```

```
## (Intercept)  7.2090      0.2217   32.52   <2e-16 ***
## log(ADR)     -1.0501      0.0401  -26.19   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.8308 on 502 degrees of freedom
## Multiple R-squared:  0.5773, Adjusted R-squared:  0.5765
## F-statistic: 685.7 on 1 and 502 DF,  p-value: < 2.2e-16
```