# QMB Exercise 2 - Estimation and Testing

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#### Introduction

The following report is based on the QMB Exercise 2 - Estimation and Testing. The task description pdf file is bis\_ex2-EstimationTesting-20150429.pdf

### Requirements

Please make sure that you the following packages loaded in your workspace.

```
library("dplyr")
library("ggplot2")
library("ggExtra")
library("gridExtra")
library("moments")
```

#### Data Set

Please make sure you have the file housingrents.csv in the subdirectory Data in your workspace.

```
housingrents <- read.csv("./Data/housingrents.csv",sep=";")
```

#### **Data Processing**

For analysis purposes it is necessary to convert the rooms and NRE variable to a factor. Furthermore a new variable rps (rent per square meter) is created. Additionally the 2 sub datasets are created for NRE respectively non-NRE appartments.

```
housingrents <- mutate(housingrents, rooms = factor(rooms),
   nre = factor(nre,levels=c(0,1),labels=c("no","yes")))
housingrents <- mutate(housingrents,rps = rent/area)
nrehousing <- filter(housingrents,nre=="yes")
nonnrehousing <- filter(housingrents,nre=="no")</pre>
```

# Task 1

. . .

# Conclusion

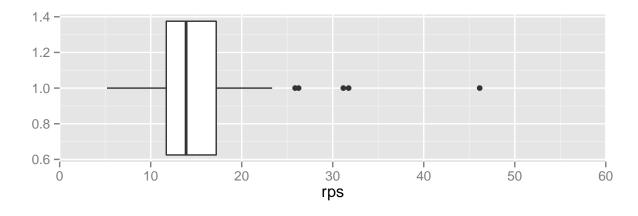
...

#### Task 2

Task 2 checks normality of variable rent per square (rps). Additionally t-tests are conducted

**a**)

In this sub task the distribution of the variable rps for NRE respectively non-NRE appartments are checked for normality. Figure 1 shows that the distribution is right skewed with several outliers on the higher rps range. Skewness value is 1.9353555. Figure 2 shows a qq plot. The plot does also show the outliers and the non-normality of the distribution.



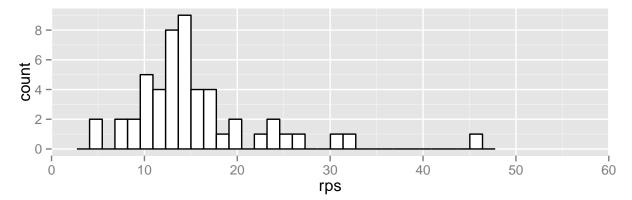


Figure 1: Boxplot and Histogramm for variable rps for NRE appartments

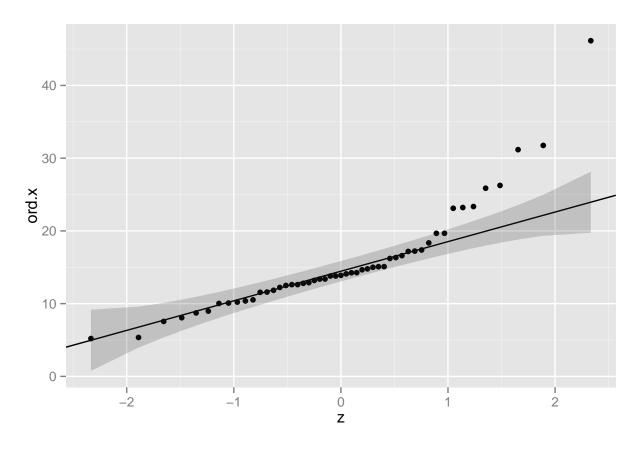
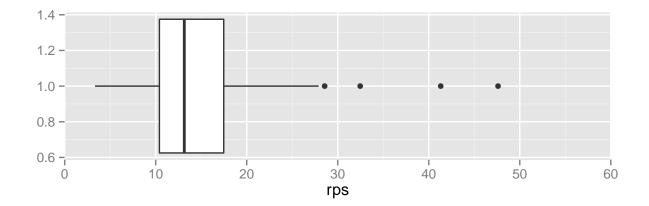
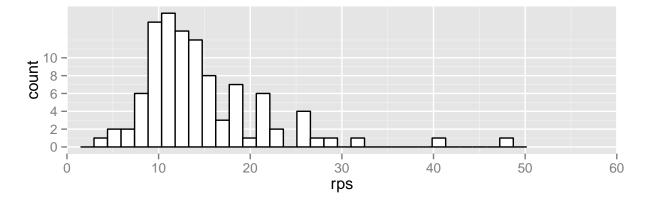


Figure 2: QQ Plot for variable rps for NRE appartments

Check normal distribution of rps variable for non-NRE appartments: Figure 3 shows that the distribution is right skewed with several outliers on the higher rps range. Skewness value is 1.9708836. Figure 4 shows a qq plot. The plot does also show the outliers and the non-normality of the distribution.





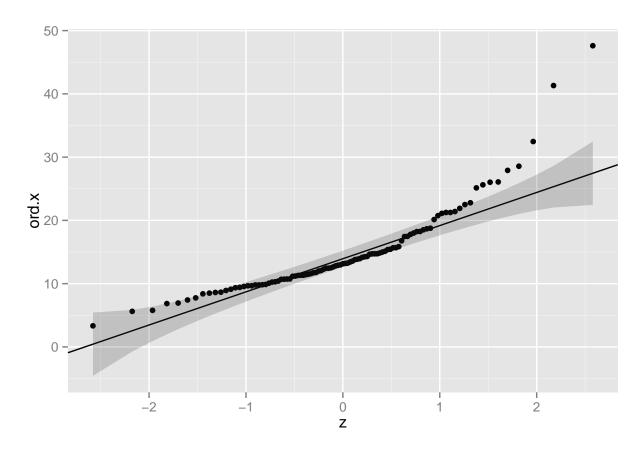


Figure 3: QQ Plot for variable rps for non-NRE appartments

b)

In this section a two-sided Student's t-test is conducted in order to check the following hypothesis:

 $H_0 = Mean$  diference of the variable rps for NRE and non-NRE appartments is equal 0.

 $H_a = Mean$  diference of the variable rps for NRE and non-NRE appartments is not equal 0.

In the first t-test equal variance is not assumed. The p-value is 0.5208345 and much higher than 0.05. The mean difference is 0.7852164 and lies in the 95% confidence interval of the estimated population mean of -3.2036613 and 1.6332286 . Therefore we fail to reject  $H_0$ .

t.test(housingrents\$rps~housingrents\$nre,alternative = "two.sided", mu=0, var.equal = FALSE)

```
##
## Welch Two Sample t-test
##
## data: housingrents$rps by housingrents$nre
## t = -0.64441, df = 96.874, p-value = 0.5208
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -3.203661 1.633229
## sample estimates:
## mean in group no mean in group yes
## 14.77912 15.56434
```

In the second t-test equal variance is assumed. The p-value is 0.5146434 and much higher than 0.05. The mean difference is 0.7852164 and lies in the 95% confidence interval of the estimated population mean of -3.160557 and 1.5901243. Therefore we fail to reject  $H_0$ .

t.test(housingrents\$rps~housingrents\$nre,alternative = "two.sided", mu=0, var.equal = TRUE)

```
##
## Two Sample t-test
##
## data: housingrents$rps by housingrents$nre
## t = -0.65318, df = 150, p-value = 0.5146
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -3.160557 1.590124
## sample estimates:
## mean in group no mean in group yes
## 14.77912 15.56434
```

**c**)

In this section a one-sided Student's t-test is conducted in order to check the following hypothesis:

 $H_0 = \mathrm{Mean}$  difference of the variable rps for NRE and non-NRE appartments is equal 0.

 $H_a = \text{Mean diference of the variable rps for NRE}$  and non-NRE appartments is greater 0.

In the second t-test equal variance is assumed. The p-value is 0.7395827 and much higher than 0.05. The mean difference is 0.7852164 and lies in the 95% confidence interval of the estimated population mean of -2.8088392 and Infinity. Therefore we fail to reject  $H_0$ .

t.test(housingrents\$rps~housingrents\$nre,alternative = "greater", mu=0, var.equal = FALSE)

#### Task 3

 $http://ww2.coastal.edu/kingw/statistics/R-tutorials/independ.html\ http://geneseo.edu/~bearden/socl211/chisquareweb/chisquare.html$ 

```
housingrentTbl <- xtabs(~rooms+nre, data=housingrents)
housingrentChi <- chisq.test(housingrentTbl,simulate.p.value=TRUE)
housingrentChi

##
## Pearson's Chi-squared test with simulated p-value (based on 2000
## replicates)
##
## data: housingrentTbl
## X-squared = 26.749, df = NA, p-value = 0.0004998
```

#### housingrentChi\$observed

```
##
       nre
## rooms no yes
##
       1 18
             5
##
       2 23
              6
       3 32
##
            5
##
       4 14 18
##
       5 9 15
##
       6 5
```

#### housingrentChi\$expected

```
##
       nre
## rooms
               no
                        yes
##
      1 15.282895 7.717105
##
      2 19.269737 9.730263
##
      3 24.585526 12.414474
##
      4 21.263158 10.736842
##
      5 15.947368 8.052632
##
      6 4.651316 2.348684
```

#### housingrentChi\$resid

```
##
       nre
## rooms
                no
                          yes
##
      1 0.6950302 -0.9780910
##
      2 0.8497704 -1.1958513
##
      3 1.4953421 -2.1043413
##
      4 -1.5751126 2.2165994
##
      5 -1.7397058 2.4482255
##
      6 0.1616756 -0.2275202
```

### resid(housingrentChi)

```
## nre
## rooms no yes
## 1 0.6950302 -0.9780910
## 2 0.8497704 -1.1958513
## 3 1.4953421 -2.1043413
## 4 -1.5751126 2.2165994
## 5 -1.7397058 2.4482255
## 6 0.1616756 -0.2275202
```

# Conclusion

. . .

# **Appendix**

#### **Functions**

Plot qqplot with ggplot

```
gg_qq <- function(x, distribution = "norm", ..., line.estimate = NULL, conf = 0.95,
                   labels = names(x)){
        q.function <- eval(parse(text = paste0("q", distribution)))</pre>
        d.function <- eval(parse(text = paste0("d", distribution)))</pre>
        x <- na.omit(x)
        ord <- order(x)
        n <- length(x)
        P <- ppoints(length(x))</pre>
        df <- data.frame(ord.x = x[ord], z = q.function(P, ...))</pre>
        if(is.null(line.estimate)){
                 Q.x \leftarrow quantile(df$ord.x, c(0.25, 0.75))
                 Q.z \leftarrow q.function(c(0.25, 0.75), ...)
                 b <- diff(Q.x)/diff(Q.z)</pre>
                 coef \leftarrow c(Q.x[1] - b * Q.z[1], b)
        } else {
                 coef <- coef(line.estimate(ord.x ~ z))</pre>
        }
        zz \leftarrow qnorm(1 - (1 - conf)/2)
        SE <- (coef[2]/d.function(df$z)) * sqrt(P * (1 - P)/n)
        fit.value <- coef[1] + coef[2] * df$z
        df$upper <- fit.value + zz * SE</pre>
        df$lower <- fit.value - zz * SE</pre>
        if(!is.null(labels)){
                 df$label <- ifelse(df$ord.x > df$upper |
                                               df$ord.x < df$lower, labels[ord],"")</pre>
        }
        p <- ggplot(df, aes(x=z, y=ord.x)) +</pre>
                 geom_point() +
                 geom_abline(intercept = coef[1], slope = coef[2]) +
                 geom_ribbon(aes(ymin = lower, ymax = upper), alpha=0.2)
        if(!is.null(labels)) p <- p + geom_text( aes(label = label))</pre>
        return(p)
```