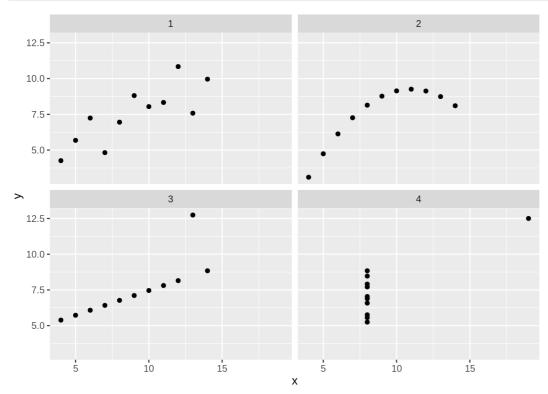
Task 1

Anscombe's data set

· Scatter plot facetted by set



· Summary calculation (mean, sd) grouped by set

```
setNames(aggregate(data[, 1:2], list(data$set), mean), c('set', 'mean_x', 'mean_y'))
```

```
## set mean_x mean_y
## 1 1 9 7,500909
## 2 2 9 7,500909
## 3 3 9 7,500000
## 4 4 9 7,500909
```

```
setNames(aggregate(data[, 1:2], list('Sd' = data$set), sd), c('set', 'sd_x', 'sd_y'))
```

```
## set sd_x sd_y

## 1 1 3,316625 2,031568

## 2 2 3,316625 2,031657

## 3 3 3,316625 2,030424

## 4 4 3,316625 2,030579
```

• Pearson's correlation by set, and non-parametric, and p-value

```
## # A tibble: 4 x 4
## set cor_pearson cor_kendall cor_spearmen
           <dbl>
                     <dbl>
                              <dbl>
## <int>
## 1
           0.816
                    0.636
                              0.818
## 2 2
           0.816
                    0.564
                              0.691
           0.816
                    0.964
                              0.991
## 3
      3
## 4
           0.817
                    0.426
                              0.5
```

```
data %>% group_by(set) %>% summarise(p_pearson = cor.test(x,y, method = 'pearson')$p.value,

p_kendall = cor.test(x,y, method = 'kendall')$p.value,

p_spearmen = cor.test(x,y, method = 'spearman')$p.value)
```

```
## # A tibble: 4 x 4

## set p_pearson p_kendall p_spearmen

## <int> <dbl> <dbl> <dbl> <dbl>
## 1 1 0.00217 0.00571 0.00373

## 2 2 0.00218 0.0165 0.0231

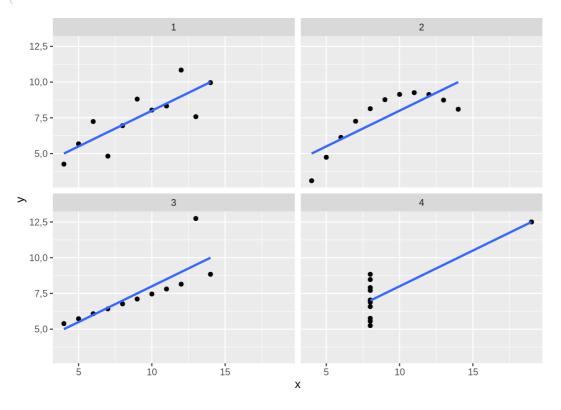
## 3 3 0.00218 0.00000551 0

## 4 4 0.00216 0.114 0.117
```

• Add geom_smooth() to the plot

```
ggplot(data, aes(x, y)) +
geom_point() +
geom_smooth(method = lm, se = FALSE) +
facet_wrap(~ set)
```

```
## `geom_smooth()` using formula 'y \sim x'
```



Air Quality data set

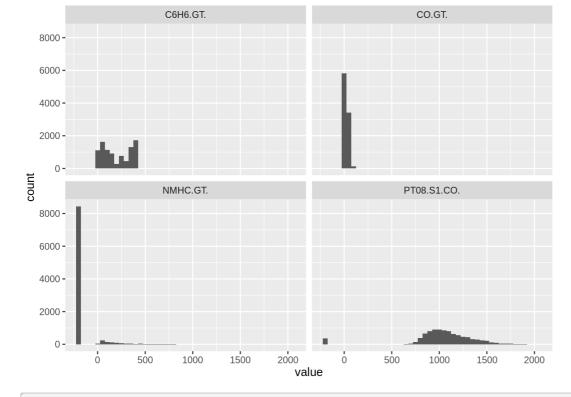
```
airq_data <- read.csv('/home/marina/Загрузки/AirQualityUCl.csv', sep = ';')
```

• Deleting NA columns and rows. Averaged concentration CO, Averaged Benzene concentration, Temperature, Humidity are factor columns in data, so it will be better to convert them into numeric. Actually, I don't need date and time for futher analysis, so I will slice the data.

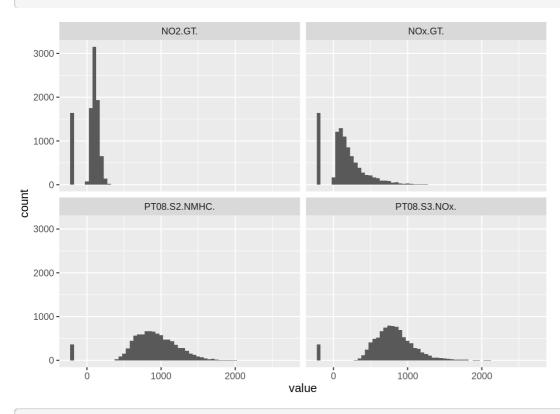
```
airq_data <- airq_data %>% select_if(~sum(!is.na(.)) > 0) %>% drop_na() airq_data[c('CO.GT.', 'C6H6.GT.', 'T', 'RH', 'AH')] <- sapply(airq_data[c('CO.GT.', 'C6H6.GT.', 'T', 'RH', 'AH')], as.numeric) airq_data <- airq_data[, c(3:15)]
```

• Exploring the variables (I make it in a 3 parts to better visualizing)

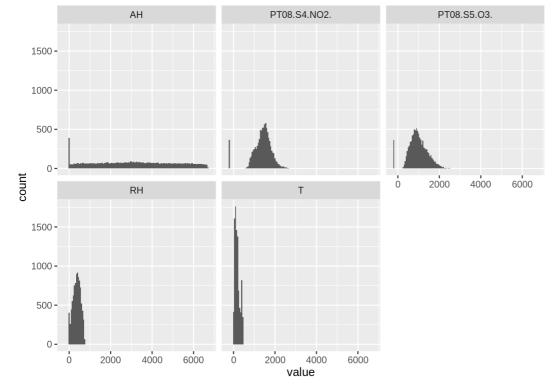
```
ggplot(gather(airq_data[, c(1:4)], cols, value), aes(x = value)) + geom_histogram(binwidth = 50) + facet_wrap(.~cols)
```



$$\begin{split} & ggplot(gather(airq_data[, c(5:8)], cols, value), aes(x = value)) + \\ & geom_histogram(binwidth = 50) + facet_wrap(.\sim cols) \end{split}$$



$$\begin{split} & ggplot(gather(airq_data[,\,c(9:13)],\,cols,\,value),\,aes(x=value)) + \\ & geom_histogram(binwidth=50) + facet_wrap(.\sim\!cols) \end{split}$$



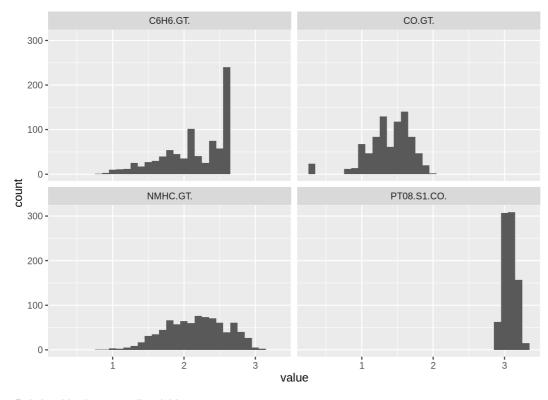
Ok, so it looks like a lot of variables have outliers and all of them in a different scale range So, as for me, it will be better to log-transform the data

· Log-transformation

```
airq_data <- log10(airq_data) %>% drop_na()
```

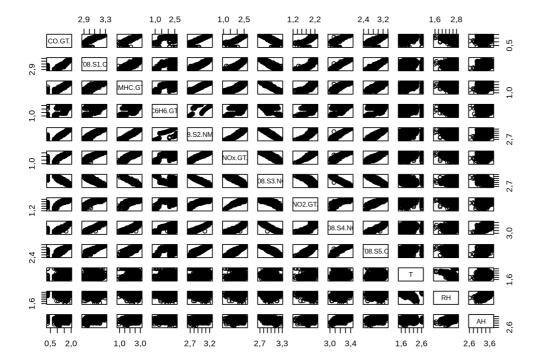
So, let's look now as an example for the first 4 columns:

```
\begin{split} & ggplot(gather(airq\_data[, c(1:4)], cols, value), aes(x = value)) + \\ & geom\_histogram(binwidth = 0.1) + facet\_wrap(.\sim cols) \end{split}
```

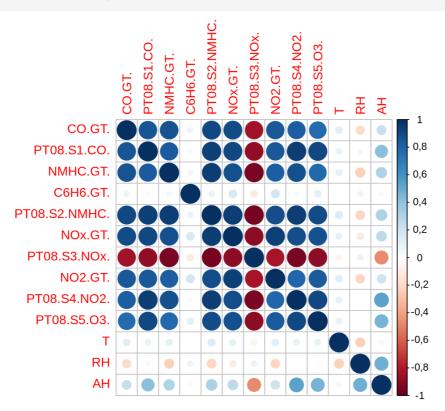


• Relationships between all variables

pairs(airq_data)



cor_data <- cor(airq_data)
corrplot(cor_data, method = 'circle')

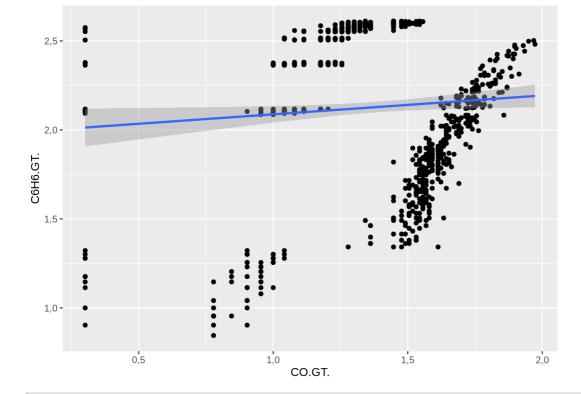


We can see that C6H6.GT. doesn't have a nice correlation with other variables. An if we check the assumptions for each variable we will see, that mostly all of them are not adhered.

Example:

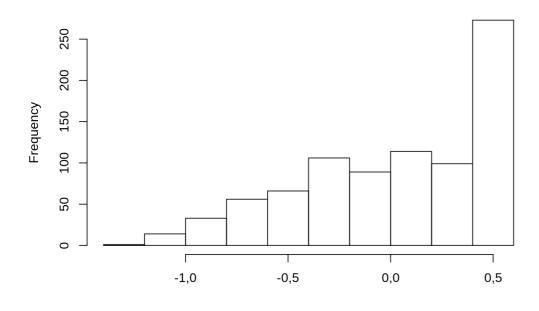
```
ggplot(airq_data, aes(x = CO.GT., y = C6H6.GT. )) +
geom_point() +
geom_smooth(method = 'lm')
```

$geom_smooth()$ using formula $y \sim x'$

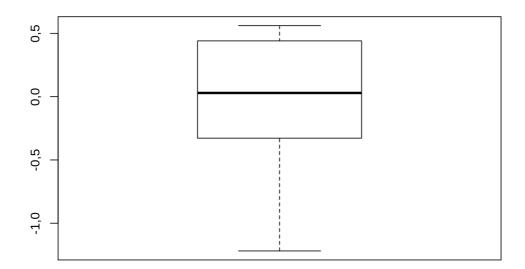


model_CO.GT. <- airq_data %>% lm(data = ., C6H6.GT. ~ CO.GT.) residuals(model_CO.GT.) %>% hist()

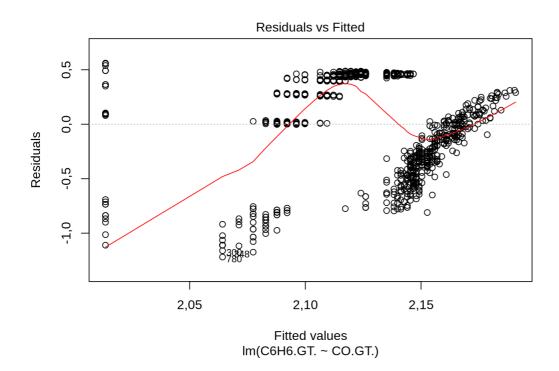
$\ \, \hbox{Histogram of} \,\, .$

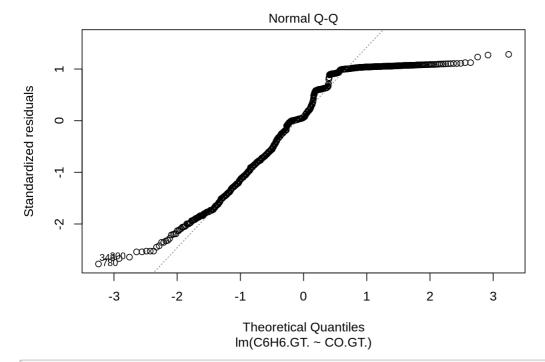


residuals(model_CO.GT.) %>% boxplot()



 $plot(model_CO.GT., which = c(1,2))$





```
summary(model_CO.GT.)
```

```
##
## Call:
## Im(formula = C6H6.GT. ~ CO.GT., data = .)
##
## Residuals:
## Min 1Q Median 3Q Max
## -1,21911 -0,32771 0,02954 0,44132 0,56155
##
## Coefficients:
```

Warning in printCoefmat(coefs, digits = digits, signif.stars = signif.stars, : в ## результате преобразования созданы NA

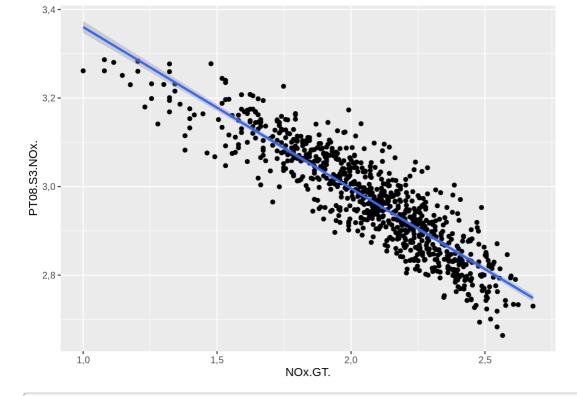
```
## Estimate Std. Error t value Pr(>|t|)
## (Intercept) 1,98172 0,06817 29,069 <2e-16 ***
## CO.GT. 0,10600 0,04815 2,202 0,028 *
## ---
## Signif. codes: 0 '***' 0,001 '**' 0,01 '*' 0,05 '.' 0,1 ' ' 1
##
## Residual standard error: 0,4408 on 849 degrees of freedom
## Multiple R-squared: 0,005677, Adjusted R-squared: 0,004506
## F-statistic: 4,847 on 1 and 849 DF, p-value: 0,02796
```

So I decided to take a PT08.S3.NOx. as a predictor, because for me it seems a little bit better. I checked the assumptions for each variable and took 3 best:

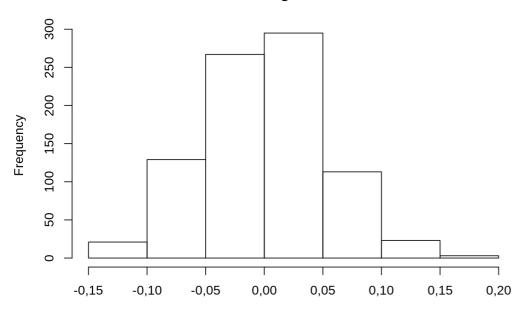
1. NOx.GT.

```
ggplot(airq_data, aes(x = NOx.GT., y = PT08.S3.NOx. )) +
geom_point() +
geom_smooth(method = 'Im')
```

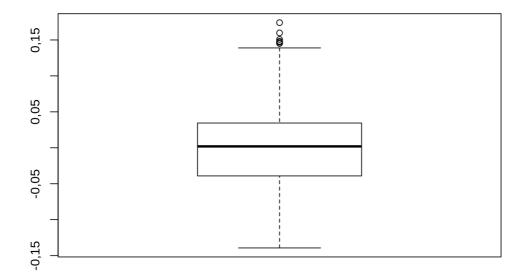
```
## `geom_smooth()` using formula 'y ~ x'
```



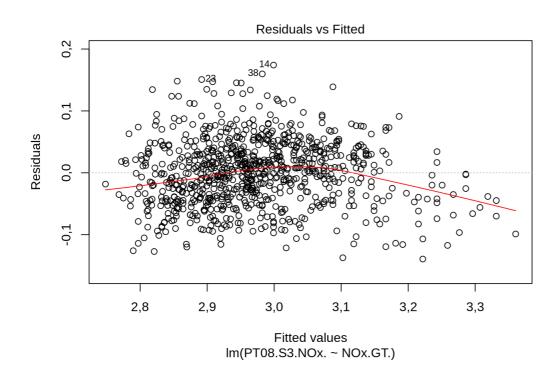


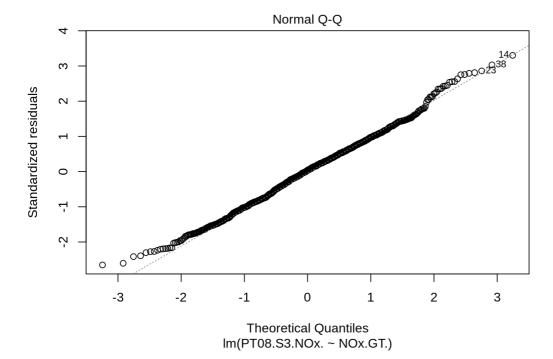


residuals(model_NOx.GT.) %>% boxplot()



 $plot(model_NOx.GT., which = c(1,2))$





```
summary(model_NOx.GT.)
```

```
## Call:
## Im(formula = PT08.S3.NOx. ~ NOx.GT., data = .)
##
## Residuals:
## Min 1Q Median 3Q Max
## -0,139384 -0,039168 0,001901 0,034434 0,174095
##
## Coefficients:
```

```
## Warning in printCoefmat(coefs, digits = digits, signif.stars = signif.stars, : в
## результате преобразования созданы NA
```

```
## Estimate Std. Error t value Pr(>|t|)
## (Intercept) 3,724907 0,013022 286,04 <2e-16 ***
## NOx.GT. -0,364507 0,006246 -58,36 <2e-16 ***
## ---
## Signif. codes: 0 '***' 0,001 '**' 0,01 '*' 0,05 '.' 0,1 ' ' 1
##
## Residual standard error: 0,05276 on 849 degrees of freedom
## Multiple R-squared: 0,8005, Adjusted R-squared: 0,8002
## F-statistic: 3406 on 1 and 849 DF, p-value: < 2,2e-16
```

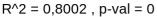
Prediction:

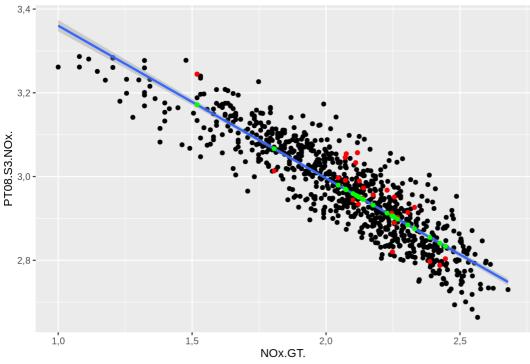
```
test_subset_NOx.GT. <- airq_data[which(row.names(airq_data) %in% sample(row.names(airq_data), 25, replace = FALSE)), c(6,7)] test_NOx.GT. <- data.frame(NOx.GT. = test_subset_NOx.GT.$NOx.GT.) test_subset_NOx.GT.$pred_PT08.S3.NOx. <- predict(model_NOx.GT., newdata = test_NOx.GT.) colnames(test_subset_NOx.GT.) <- c('real_NOx.GT.', 'real_PT08.S3.NOx.', 'pred_PT08.S3.NOx.') head(test_subset_NOx.GT.)
```

```
##
    real_NOx.GT. real_PT08.S3.NOx. pred_PT08.S3.NOx.
## 3
       2,117271
                     3,056905
                                  2,953147
## 16
       2,110590
                     3,033021
                                   2,955582
                                   2,884590
## 50
        2,305351
                     2,914343
                     2,926342
## 77
        2,330414
                                   2,875455
## 142
       2,255273
                      2,950851
                                   2,902844
## 292
        2,255273
                      2,889302
                                    2,902844
```

```
R <- round(summary(model_NOx.GT.)$adj.r.squared, digits = 4)
p <- round(summary(model_NOx.GT.)$coefficients[2,4], digits = 3)
titl <- paste('R^2 =', as.character(R),', p-val =', as.character(p))
ggplot() +
geom_point(data = airq_data, aes(NOx.GT., PT08.S3.NOx.)) +
geom_smooth(data = airq_data, aes(NOx.GT., PT08.S3.NOx.), method = 'Im') +
geom_point(data = test_subset_NOx.GT., aes(real_NOx.GT., real_PT08.S3.NOx.), color = 'red') +
geom_point(data = test_subset_NOx.GT., aes(real_NOx.GT., pred_PT08.S3.NOx.), color = 'green') +
labs(title = titl)
```

`geom_smooth()` using formula 'y \sim x'

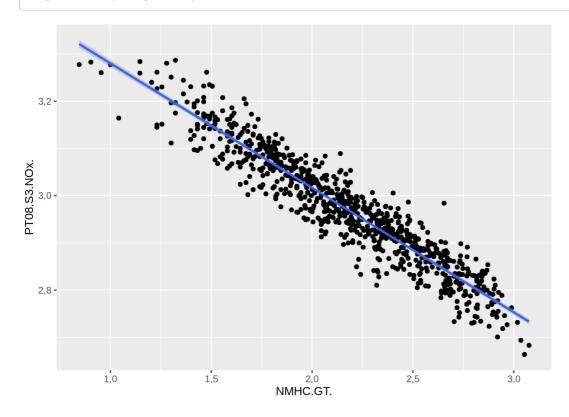




2. NMHC.GT.

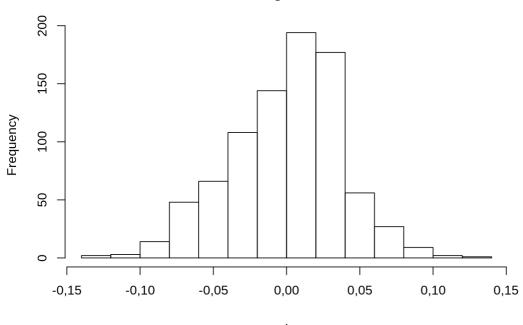
```
ggplot(airq_data, aes(x = NMHC.GT., y = PT08.S3.NOx.)) +
geom_point() +
geom_smooth(method = 'lm')
```

$geom_smooth()$ using formula $y \sim x'$

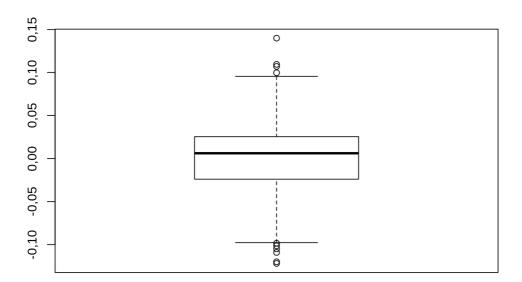


 $model_NMHC.GT. <- airq_data \%>\% \ lm(data = ., PT08.S3.NOx.^ NMHC.GT.) \\ residuals(model_NMHC.GT.) \%>\% \ hist()$



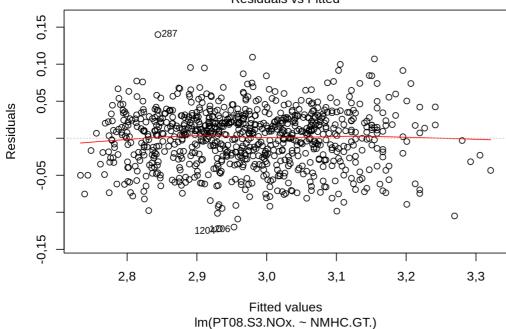


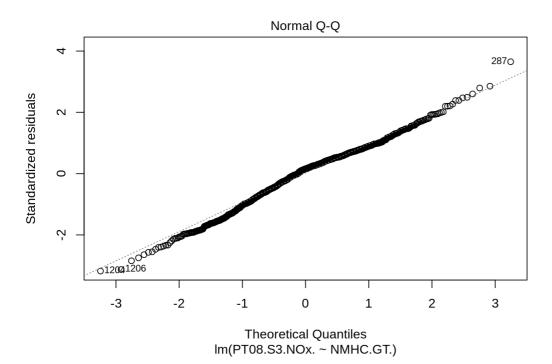
residuals(model_NMHC.GT.) %>% boxplot()



 $plot(model_NMHC.GT., which = c(1,2))$

Residuals vs Fitted





summary(model_NMHC.GT.)

```
##
## Call:
## Im(formula = PT08.S3.NOx. ~ NMHC.GT., data = .)
##
## Residuals:
## Min 1Q Median 3Q Max
## -0,121954 -0,023971 0,006133 0,025483 0,139954
##
## Coefficients:
```

```
## Warning in printCoefmat(coefs, digits = digits, signif.stars = signif.stars, : в
## результате преобразования созданы NA
```

```
## Estimate Std. Error t value Pr(>|t|)
## (Intercept) 3,543874 0,006867 516,0 <2e-16 ***
## NMHC.GT. -0,263641 0,003109 -84,8 <2e-16 ***
## ---
## Signif. codes: 0 '***' 0,001 '**' 0,05 '.' 0,1 ' ' 1
##
## Residual standard error: 0,03838 on 849 degrees of freedom
## Multiple R-squared: 0,8944, Adjusted R-squared: 0,8943
## F-statistic: 7191 on 1 and 849 DF, p-value: < 2,2e-16
```

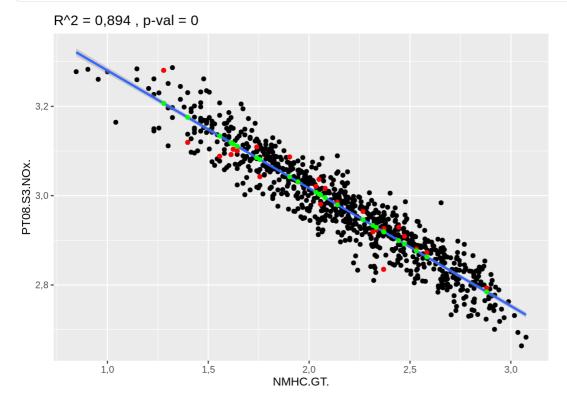
Prediction:

```
test_subset_NMHC.GT. <- airq_data[which(row.names(airq_data) %in% sample(row.names(airq_data), 25, replace = FALSE)), c(3,7)] test_NMHC.GT. <- data.frame(NMHC.GT. = test_subset_NMHC.GT.) test_subset_NMHC.GT.$pred_PT08.S3.NOx. <- predict(model_NMHC.GT., newdata = test_NMHC.GT.) colnames(test_subset_NMHC.GT.) <- c('real_NMHC.GT.', 'real_PT08.S3.NOx.', 'pred_PT08.S3.NOx.') head(test_subset_NMHC.GT.)
```

```
real_NMHC.GT. real_PT08.S3.NOx. pred_PT08.S3.NOx.
## 46
      2,267172 2,964731 2,946153
     2,332438 2,923762
                              2.928946
## 53
      2,049218 3,036629
                              3,003615
## 69
## 70
      2,033424 3,020775
                              3,007779
## 87
       1,556303
                  3.088136
                              3,133568
## 116 2,315970
                   2,919078
                               2,933288
```

```
R <- round(summary(model_NMHC.GT.)$adj.r.squared, digits = 3)
p <- round(summary(model_NMHC.GT.)$coefficients[2,4], digits = 3)
titl <- paste('R^2 =', as.character(R),', p-val =', as.character(p))
ggplot() +
geom_point(data = airq_data, aes(NMHC.GT., PT08.S3.NOx.)) +
geom_smooth(data = airq_data, aes(NMHC.GT., PT08.S3.NOx.), method = 'Im') +
geom_point(data = test_subset_NMHC.GT., aes(real_NMHC.GT., real_PT08.S3.NOx.), color = 'red') +
geom_point(data = test_subset_NMHC.GT., aes(real_NMHC.GT., pred_PT08.S3.NOx.), color = 'green') +
labs(title = titl)
```

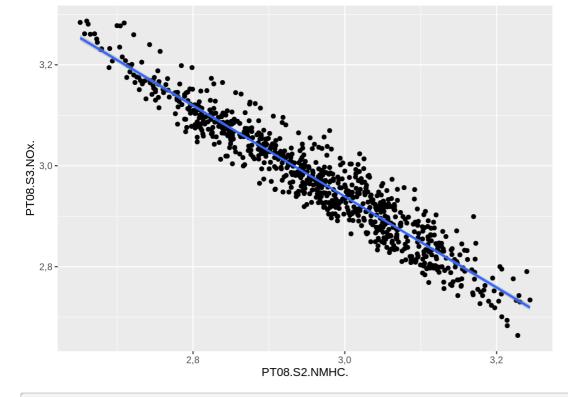
`geom_smooth()` using formula 'y ~ x'



3. PT08.S2.NMHC.

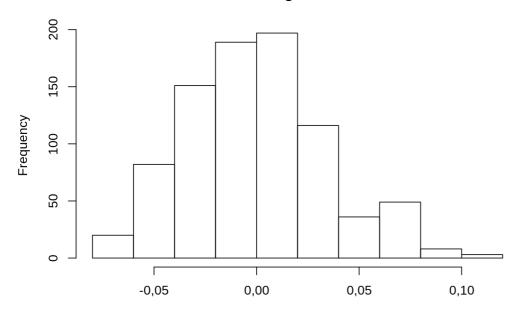
```
ggplot(airq_data, aes(x = PT08.S2.NMHC., y = PT08.S3.NOx. )) + geom_point() + geom_smooth(method = 'Im')
```

```
## `geom_smooth()` using formula 'y ~ x'
```

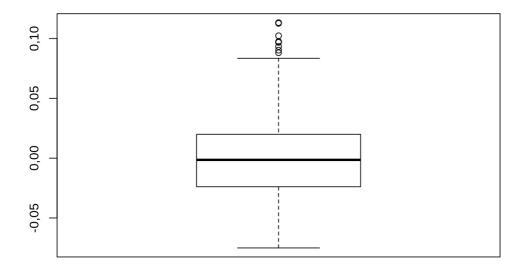


 $model_PT08.S2.NMHC. <- airq_data ~ \%>\% ~ lm(data = ., ~ PT08.S3.NOx. ~ PT08.S2.NMHC.) \\ residuals(model_PT08.S2.NMHC.) ~ \%>\% ~ hist()$

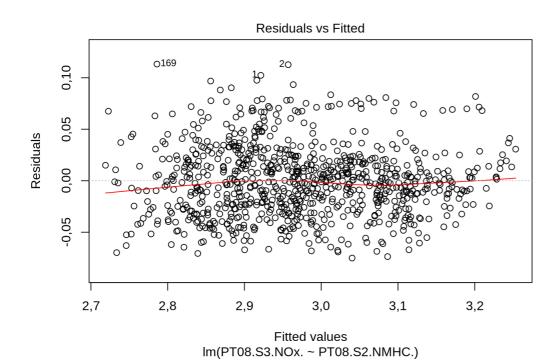
Histogram of .



residuals(model_PT08.S2.NMHC.) %>% boxplot()



 $plot(model_PT08.S2.NMHC., which = c(1,2))$



Im(PT08.S3.NOx. ~ PT08.S2.NMHC.)

```
summary(model_PT08.S2.NMHC.)
```

```
## Call:
## Im(formula = PT08.S3.NOx. ~ PT08.S2.NMHC., data = .)
##
## Residuals:
## Min 1Q Median 3Q Max
## -0,075066 -0,023927 -0,001433 0,019915 0,113269
##
## Coefficients:
```

```
## Warning in printCoefmat(coefs, digits = digits, signif.stars = signif.stars, : в ## результате преобразования созданы NA
```

```
## Estimate Std. Error t value Pr(>|t|)
## (Intercept) 5,64335 0,02773 203,53 <2e-16 ***
## PT08.S2.NMHC. -0,90146 0,00935 -96,42 <2e-16 ***
## ---
## Signif. codes: 0 '***' 0,001 '**' 0,01 '*' 0,05 '.' 0,1 ' ' 1
##
## Residual standard error: 0,03417 on 849 degrees of freedom
## Multiple R-squared: 0,9163, Adjusted R-squared: 0,9162
## F-statistic: 9296 on 1 and 849 DF, p-value: < 2,2e-16
```

Prediction:

```
test_subset_PT08.S2.NMHC. <- airq_data[which(row.names(airq_data) %in% sample(row.names(airq_data), 25, replace = FALSE)), c(5,7)] test_PT08.S2.NMHC. <- data.frame(PT08.S2.NMHC. = test_subset_PT08.S2.NMHC.) test_subset_PT08.S2.NMHC.$pred_PT08.S3.NOx. <- predict(model_PT08.S2.NMHC., newdata = test_PT08.S2.NMHC.) colnames(test_subset_PT08.S2.NMHC.) <- c('real_PT08.S2.NMHC.', 'real_PT08.S3.NOx.', 'pred_PT08.S3.NOx.') head(test_subset_PT08.S2.NMHC.)
```

```
##
    real_PT08.S2.NMHC. real_PT08.S3.NOx. pred_PT08.S3.NOx.
## 76
           3,095866
                         2,881955
                                       2,852539
## 86
           2,788875
                         3,141136
                                       3,129280
## 89
           2,965202
                         2,978637
                                       2,970328
## 171
            3,221936
                         2,775974
                                       2,738892
## 180
            2,794488
                         3,111599
                                       3,124220
## 436
            2,825426
                          3,100715
                                       3,096331
```

```
R <- round(summary(model_PT08.S2.NMHC.)$adj.r.squared, digits = 3)
p <- round(summary(model_PT08.S2.NMHC.)$coefficients[2,4], digits = 3)
titl <- paste('R^2 =', as.character(R),', p-val =', as.character(p))
ggplot() +
geom_point(data = airq_data, aes(PT08.S2.NMHC., PT08.S3.NOx.)) +
geom_smooth(data = airq_data, aes(PT08.S2.NMHC., PT08.S3.NOx.), method = 'lm') +
geom_point(data = test_subset_PT08.S2.NMHC., aes(real_PT08.S2.NMHC., real_PT08.S3.NOx.), color = 'red') +
geom_point(data = test_subset_PT08.S2.NMHC., aes(real_PT08.S2.NMHC., pred_PT08.S3.NOx.), color = 'green') +
labs(title = titl)
```

`geom_smooth()` using formula 'y \sim x'

$R^2 = 0.916$, p-val = 0

