Зачетная работа. Вариант 3

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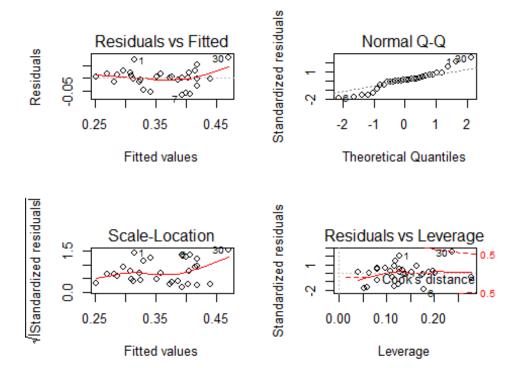
Номер 1.

```
library(Ecdat)
## Loading required package: Ecfun
##
## Attaching package: 'Ecfun'
## The following object is masked from 'package:base':
##
##
       sign
##
## Attaching package: 'Ecdat'
## The following object is masked from 'package:datasets':
##
##
       Orange
library(MASS)
##
## Attaching package: 'MASS'
## The following object is masked from 'package:Ecdat':
##
##
       SP500
library(lmtest)
## Loading required package: zoo
##
## Attaching package: 'zoo'
## The following objects are masked from 'package:base':
##
##
       as.Date, as.Date.numeric
library(orcutt)
library(sandwich)
library(AER)
## Loading required package: car
## Loading required package: carData
```

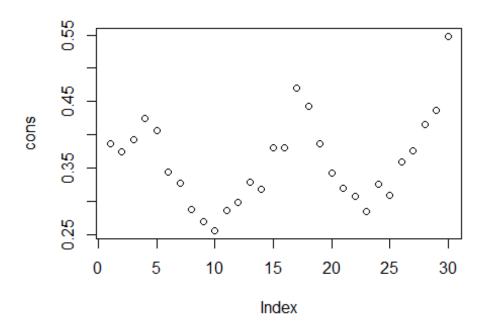
```
##
## Attaching package: 'carData'
## The following object is masked from 'package:Ecdat':
##
##
       Mroz
## Loading required package: survival
Пункт 1)
data(Icecream)
model <- lm(cons ~ ., data = Icecream)</pre>
summary(model)
##
## Call:
## lm(formula = cons ~ ., data = Icecream)
## Residuals:
##
        Min
                   1Q
                         Median
                                                Max
                                       30
## -0.065302 -0.011873 0.002737 0.015953 0.078986
##
## Coefficients:
##
                 Estimate Std. Error t value Pr(>|t|)
## (Intercept) 0.1973151 0.2702162 0.730 0.47179
## income
               0.0033078 0.0011714 2.824 0.00899 **
## price
             -1.0444140 0.8343573 -1.252 0.22180
               0.0034584 0.0004455 7.762 3.1e-08 ***
## temp
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '* 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.03683 on 26 degrees of freedom
## Multiple R-squared: 0.719, Adjusted R-squared: 0.6866
## F-statistic: 22.17 on 3 and 26 DF, p-value: 2.451e-07
```

par(mfrow = c(2,2))

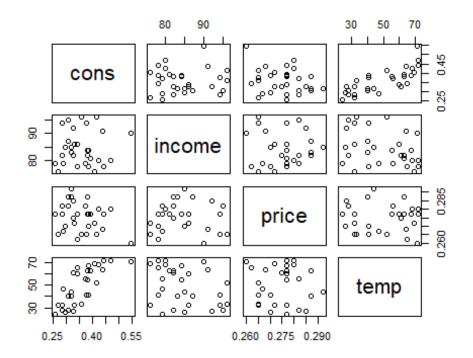
plot(model)



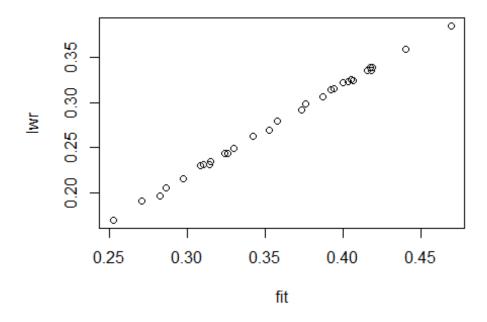
Пункт 2)
dataIce <- data.frame(Icecream)
confs <- predict(model, dataIce, interval = "confidence")
preds <- predict(model, dataIce, interval = "prediction")
plot(cons ~ 1, data = Icecream)



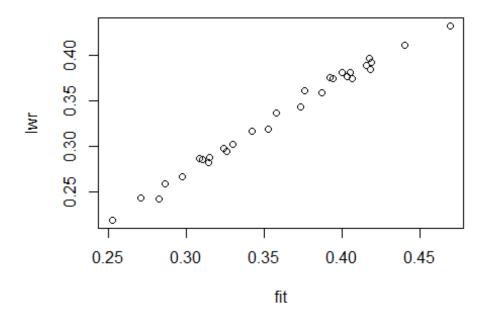
plot(dataIce)



plot(preds)



plot(confs)



Пункт 3)

```
bgtest(model, order = 2)
##
## Breusch-Godfrey test for serial correlation of order up to 2
##
## data: model
## LM test = 4.4872, df = 2, p-value = 0.1061
```

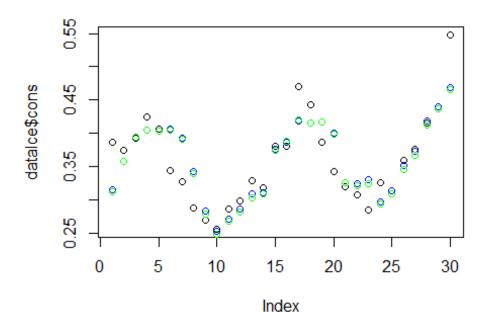
Пункт 4)

```
model_orcutt = cochrane.orcutt(model, max.iter = 1000)
model_orcutt
## Cochrane-orcutt estimation for first order autocorrelation
##
## Call:
## lm(formula = cons ~ ., data = Icecream)
    number of interaction: 11
##
##
    rho 0.400926
##
## Durbin-Watson statistic
## (original):
                  1.02117 , p-value: 3.024e-04
## (transformed): 1.54884 , p-value: 5.061e-02
##
##
   coefficients:
## (Intercept)
                    income
                                 price
                                               temp
      0.157148
                  0.003203
                             -0.892396
                                           0.003558
```

```
coefficients(model)
## (Intercept)
                    income
                                 price
                                              temp
## 0.19731507 0.00330776 -1.04441399 0.00345843
Пункт 5)
NeweyWest(model)
##
                 (Intercept)
                                    income
                                                   price
                                                                  temp
## (Intercept) 1.030769e-01 -2.496457e-04 -0.3015182861 7.269565e-05
## income
               -2.496457e-04 8.769416e-07 0.0006296449 -6.511207e-08
## price
               -3.015183e-01 6.296449e-04 0.9230633344 -2.800484e-04
                7.269565e-05 -6.511207e-08 -0.0002800484 2.931894e-07
## temp
coefficients(model)
## (Intercept)
                    income
                                 price
                                              temp
## 0.19731507
                0.00330776 -1.04441399 0.00345843
pt(-1.0444140 / sqrt(0.9230633344), 30-3-1)*2
## [1] 0.286982
```

Пунккт 6)

```
preds <- predict(model, dataIce)
predsOrc <- predict(model_orcutt, dataIce)
plot(dataIce$cons)
points(preds, col = "blue")
points(predsOrc, col = "green")</pre>
```



Номер 2

Пунккт 1)

```
data(HousePrices)
dataHouse <- data.frame(HousePrices)</pre>
dataHouse$price = log(dataHouse$price)
dataHouse$lotsize = log(dataHouse$lotsize)
min.model = lm(price ~ 1, data=HousePrices)
biggest <- formula(lm(price~., HousePrices))</pre>
summary(HousePrices)
##
        price
                         lotsize
                                         bedrooms
                                                        bathrooms
   Min.
          : 25000
                     Min.
                           : 1650
                                      Min.
##
                                             :1.000
                                                      Min.
                                                              :1.000
   1st Qu.: 49125
                     1st Qu.: 3600
                                      1st Qu.:2.000
                                                      1st Qu.:1.000
##
##
   Median : 62000
                     Median : 4600
                                      Median :3.000
                                                      Median :1.000
##
   Mean
           : 68122
                     Mean
                             : 5150
                                      Mean
                                             :2.965
                                                      Mean
                                                              :1.286
    3rd Ou.: 82000
                     3rd Qu.: 6360
                                      3rd Ou.:3.000
                                                       3rd Qu.:2.000
##
##
    Max.
           :190000
                     Max.
                            :16200
                                      Max.
                                             :6.000
                                                      Max.
                                                              :4.000
##
                    driveway recreation fullbase gasheat
       stories
                                                               aircon
                    no : 77
##
   Min.
           :1.000
                              no:449
                                          no :355
                                                    no:521
                                                               no:373
##
   1st Qu.:1.000
                    yes:469
                              yes: 97
                                                    yes: 25
                                          yes:191
                                                               yes:173
   Median :2.000
##
##
   Mean
           :1.808
   3rd Qu.:2.000
##
##
           :4.000
   Max.
##
                     prefer
        garage
##
   Min.
           :0.0000
                     no:418
   1st Ou.:0.0000
                     ves:128
   Median :0.0000
##
##
   Mean
           :0.6923
## 3rd Qu.:1.0000
## Max. :3.0000
```

Акаике:

```
lm_a = stepAIC(min.model, direction = "forward", k=2, scope=biggest, trace = FAL
SE)
lm a
##
## Call:
## lm(formula = price ~ lotsize + bathrooms + aircon + stories +
       prefer + garage + fullbase + gasheat + driveway + recreation +
##
##
       bedrooms, data = HousePrices)
##
## Coefficients:
                                                      airconyes
##
     (Intercept)
                         lotsize
                                      bathrooms
                                                                        stories
##
       -4038.350
                           3.546
                                      14335.558
                                                      12632.890
                                                                       6556.946
##
       preferves
                                    fullbaseyes
                                                     gasheatyes
                                                                    drivewayyes
                          garage
##
                                                      12831.406
                                                                       6687.779
        9369.513
                        4244.829
                                       5452.386
## recreationyes
                        bedrooms
        4511.284
                        1832.003
##
```

```
Шварц:
```

```
lm b = stepAIC(min.model, direction = "forward", k=log(nrow(HousePrices)), scope
=biggest, trace = FALSE)
lm_b
##
## Call:
## lm(formula = price ~ lotsize + bathrooms + aircon + stories +
       prefer + garage + fullbase + gasheat + driveway, data = HousePrices)
##
## Coefficients:
                    lotsize
                               bathrooms
## (Intercept)
                                            airconyes
                                                           stories
                                                                      preferyes
##
     -1123.144
                      3.666
                               15072.868
                                            12875.657
                                                          7241,264
                                                                       9595.888
##
        garage fullbaseyes
                             gasheatyes drivewayyes
     4265.862
                  7134.099 12954.080
##
                                             6428.566
```

Пункт 2)

Акаике:

```
summary(lm_a)$r.squared
## [1] 0.6731236
```

Шварц:

```
summary(lm_b)$r.squared
## [1] 0.6679045
```

Пункт 3)

Акаике:

```
resettest(lm_a)

##

## RESET test

##

## data: lm_a

## RESET = 13.481, df1 = 2, df2 = 532, p-value = 1.944e-06
```

Шварц:

```
resettest(lm_b)
##
## RESET test
##
## data: lm_b
## RESET = 11.2, df1 = 2, df2 = 534, p-value = 1.719e-05
```

В обоих случаях гипотеза о том, что в модели нет отсутствующих переменных или модель линейная, отвергается.