

Wielowymiarowe Modele Ekonometrii Finansowej

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Część I

Bazy Danych

- stooq.pl
 - finance.yahoo.com
 - datamarket.com
 - bossa
 - econstats
-

Zadanie 1

- Ściągnij plik notowań dla Bitcoin USD (BTC-USD) z okresu od 05-03-2016 do 05-03-2021
- Upewnij się, że dane w kolumnie są typu liczbowego
- Narysuj wykres danych
- Dodatkowe Informacje: finance.yahoo; pracujemy w Excelu

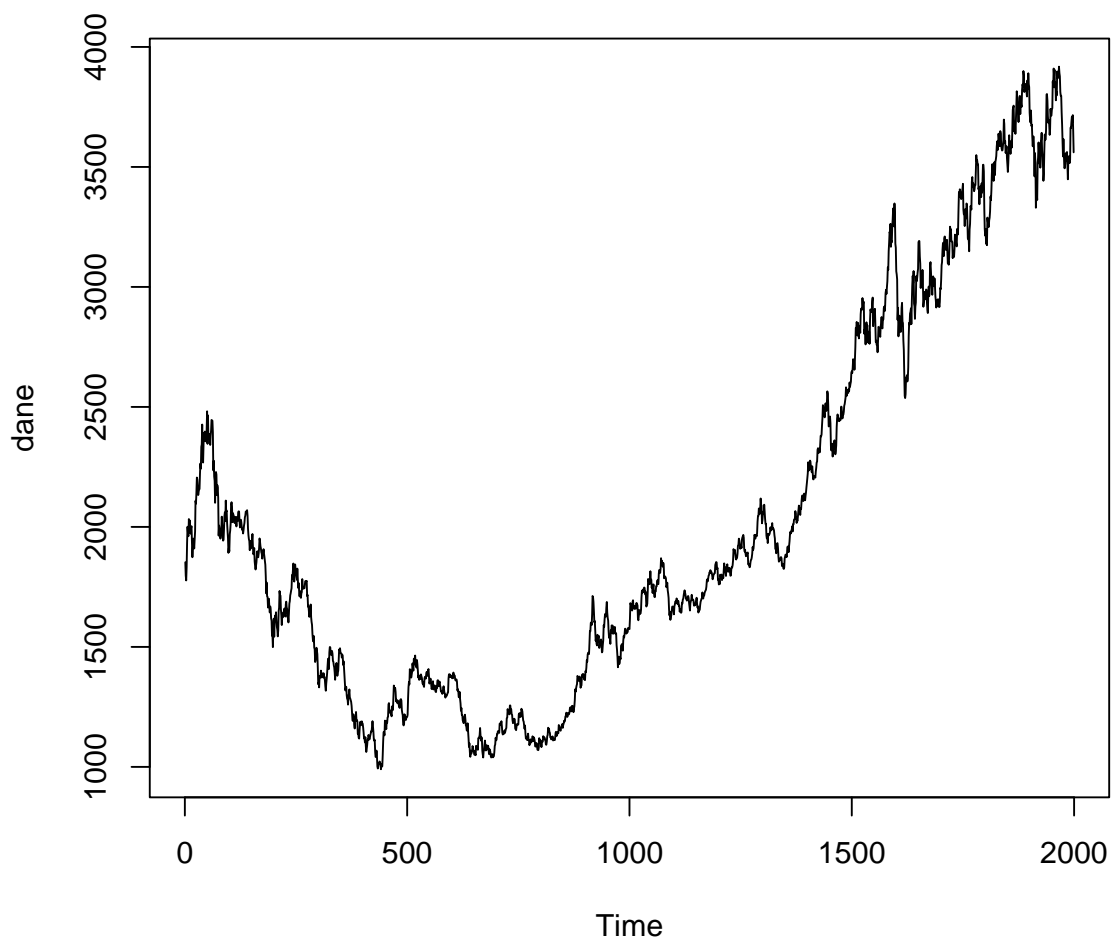
Zadanie 2

- Załaduj plik danych Wig20_2000do2007.csv
- Oblicz logarytmiczne stopy zwrotu (pracujemy na kursach zamknięcia)
- Narysuj ceny i logarytmiczne stopy zwrotu (jeden ekran 2 wykresy)
- Oblicz i zinterpretuj podstawowe statystyki opisowe
- Dlaczego często modelujemy stopy zwrotu, a nie ceny?
- Narysuj i podaj interpretacje ACF dla cen i stóp zwrotu oraz ich kwadratów
- Oceń występowanie efektu ARCH
 - wnioskowanie na podstawie rysunków
 - testy statystyczne

Wskazówki:

```
setwd()
dane<- read.csv2(file="Wig20_2000do2007.csv",dec=",")
dane<-ts(dane[,4])
lnrdane<- #logarytmiczne stopy zwrotu
library(psych)# wykorzystaj describe() do obliczenia statystyk opisowych
acf(?)#Autocorrelation Function
library(TSA)#zmiana pakietu do rysowania acf
TSA::acf()#Narysuj bez opóźnienia w 0
Box.test() # wersja Ljung; lag= 20 lub do dyskusji
McLeod.Li.test()
library(MTS)
archTest()
```

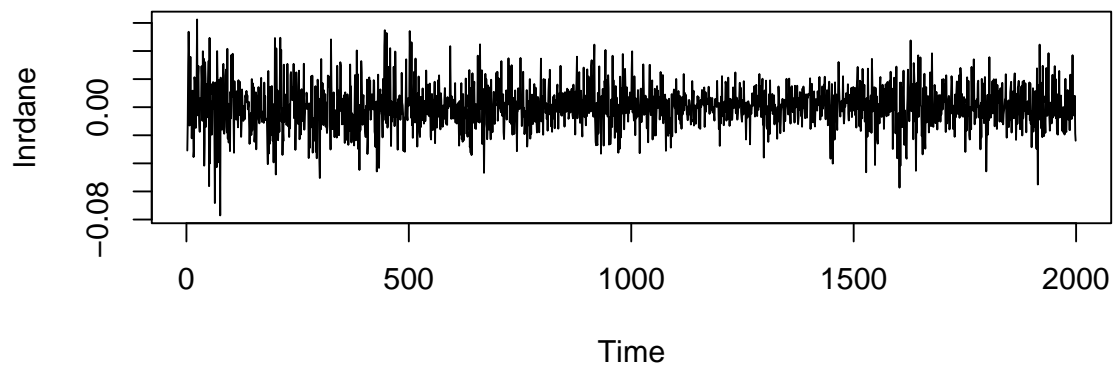
Rozwiązanie zadania 2



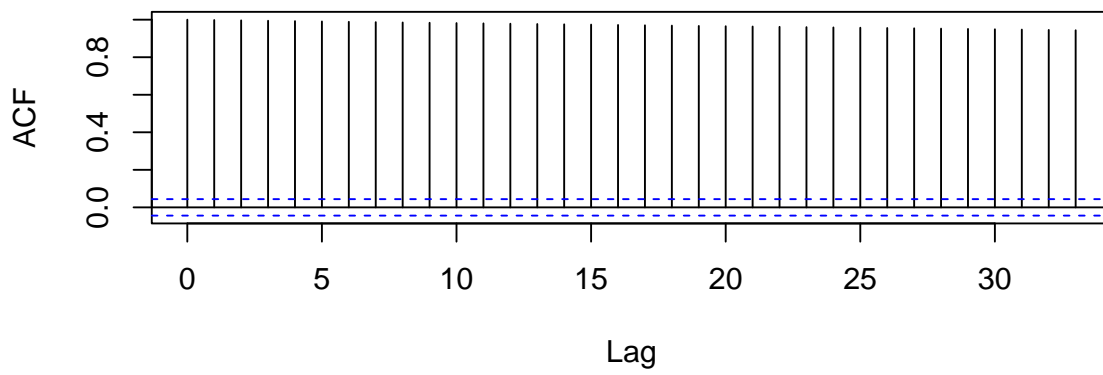
```
##                               X1
## vars      1.0000000
## n        1999.000000
## mean     2037.1221411
## sd       811.8572246
## median   1801.4300000
## trimmed  1953.9834791
## mad      740.3363100
## min      990.2300000
## max     3917.8700000
## range    2927.6400000
## skew      0.7728156
## kurtosis  -0.6143705
## se       18.1582195

##                               X1
## vars      1.0000000e+00
## n        1.9980000e+03
## mean     3.269534e-04
```

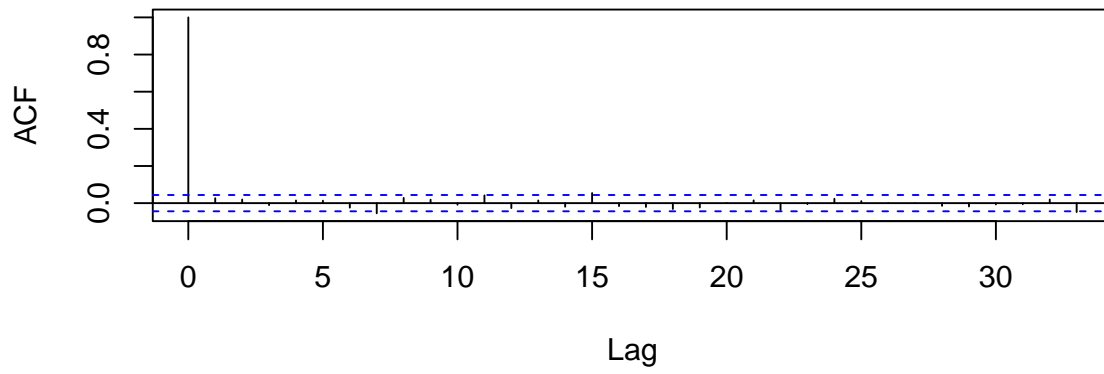
```
## sd      1.554302e-02
## median  2.455809e-04
## trimmed 2.507795e-04
## mad     1.275988e-02
## min     -7.705723e-02
## max     6.246071e-02
## range   1.395179e-01
## skew    8.626955e-04
## kurtosis 1.293533e+00
## se      3.477265e-04
```



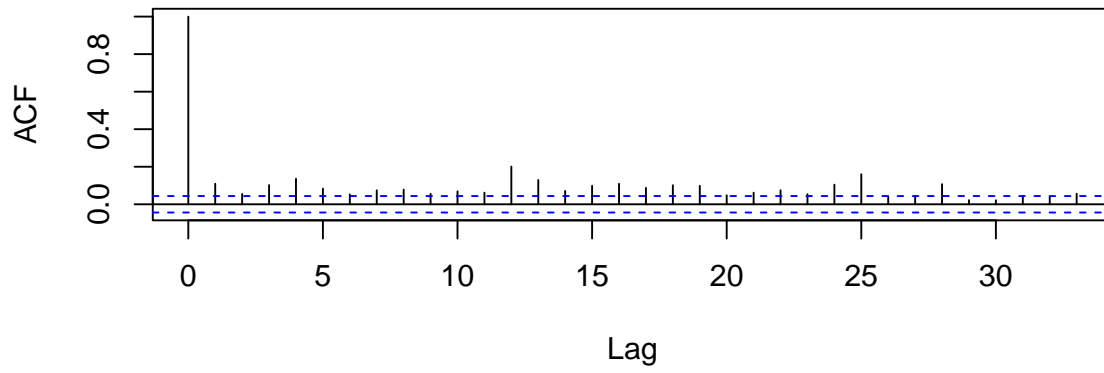
Series dane

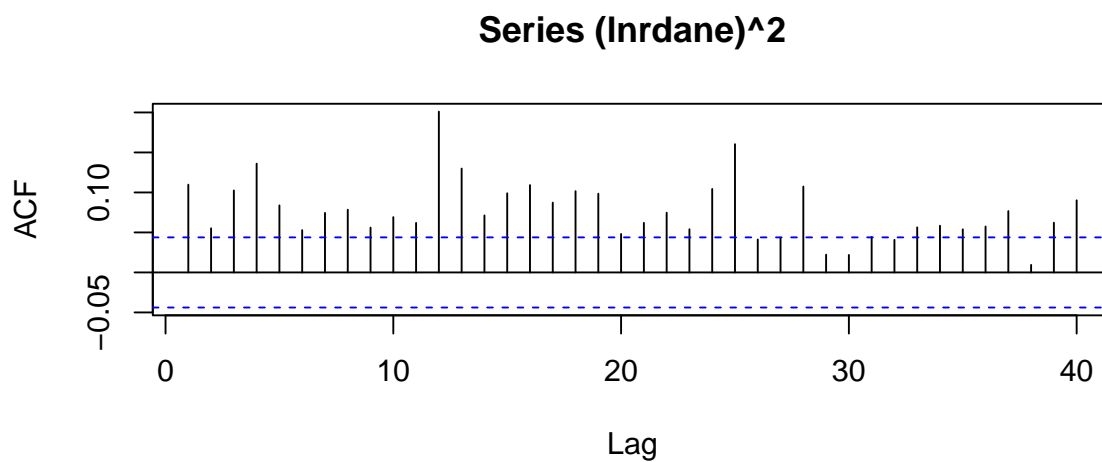
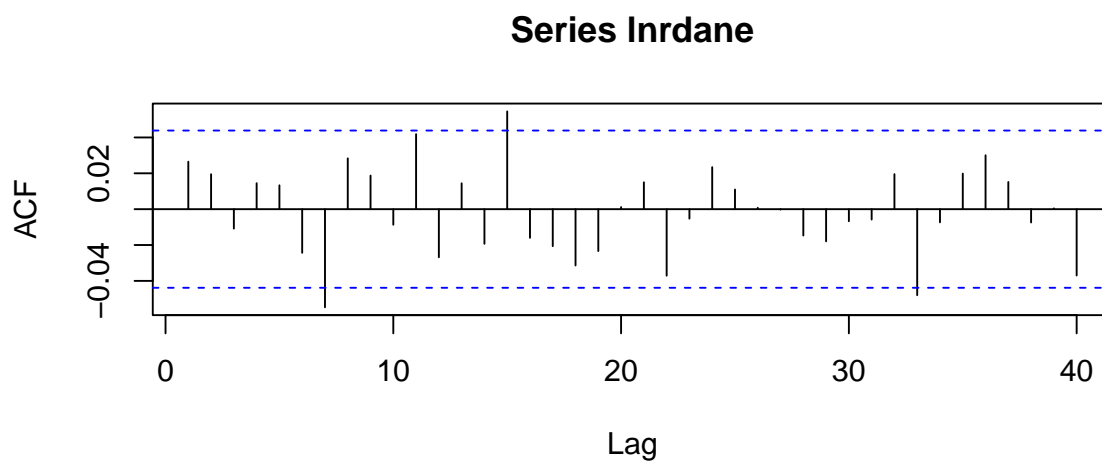


Series Inrdane

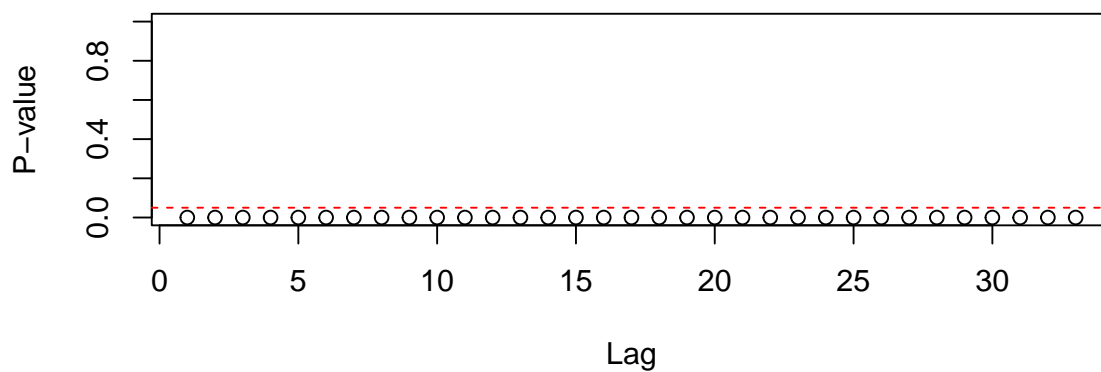
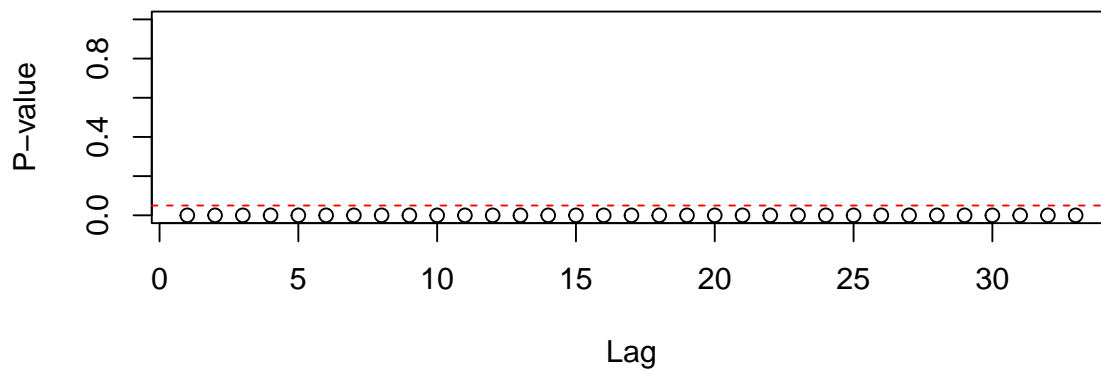


Series Inrdane^2





```
##
## Box-Ljung test
##
## data: lnrdane
## X-squared = 29.463, df = 20, p-value = 0.07904
##
## Box-Ljung test
##
## data: lnrdane^2
## X-squared = 384.98, df = 20, p-value < 2.2e-16
## $p.values
## [1] 9.155079e-07 2.768742e-07 4.316858e-11 0.000000e+00 0.000000e+00
## [6] 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00
## [11] 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00
## [16] 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00
## [21] 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00
## [26] 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00
## [31] 0.000000e+00 0.000000e+00 0.000000e+00
```



```
## $p.values
## [1] 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0

## Q(m) of squared series(LM test):
## Test statistic: 147.4539 p-value: 0
## Rank-based Test:
## Test statistic: 133.5223 p-value: 0
```

Jednowymiarowe modele GARCH

Zadanie 3

- Wczytaj plik DAX_2006_2016_ost.xlsx
- Oceń występowanie efektu ARCH
- Dopasuj modele GARCH do danych:
 - ARMA-sGARCH z warunkowym rozkładem normalnym
 - ARMA-sGARCH z war. rozkładem studenta
 - ARMA-GJR-GARCH z war. skośnym rozkładem studenta
 - Ocen który model jest lepszy na podstawie kryteriów informacyjnych
 - Analiza estymatorów parametrów

- Oblicz/Narysuj:
 - wykres surowych reszt
 - wykres wariancji warunkowych
 - wykres odchylen warunkowych
- Wykonaj diagnostykę reszt
- Dla najlepszego modelu wykonaj prognozę

Wskazówki: Introduction to rugarch

```
library(rugarch)
library(readxl)
DAX_2006_2016_ost <- read_excel("Dane/DAX_2006_2016_ost.xlsx")
y<-ts(DAX_2006_2016_ost[,2])
yGARCH_spec <- ugarchspec(variance.model=list(model=?, garchOrder=?), mean.model = list(armaOrder
= ?, include.mean = TRUE), distribution.model=?)
yGARCH_fit<-ugarchfit(data = y, spec = yGARCH_spec, solver = "hybrid")
plot(yGARCH_fit)#brak rysunku w rozwiązaniu

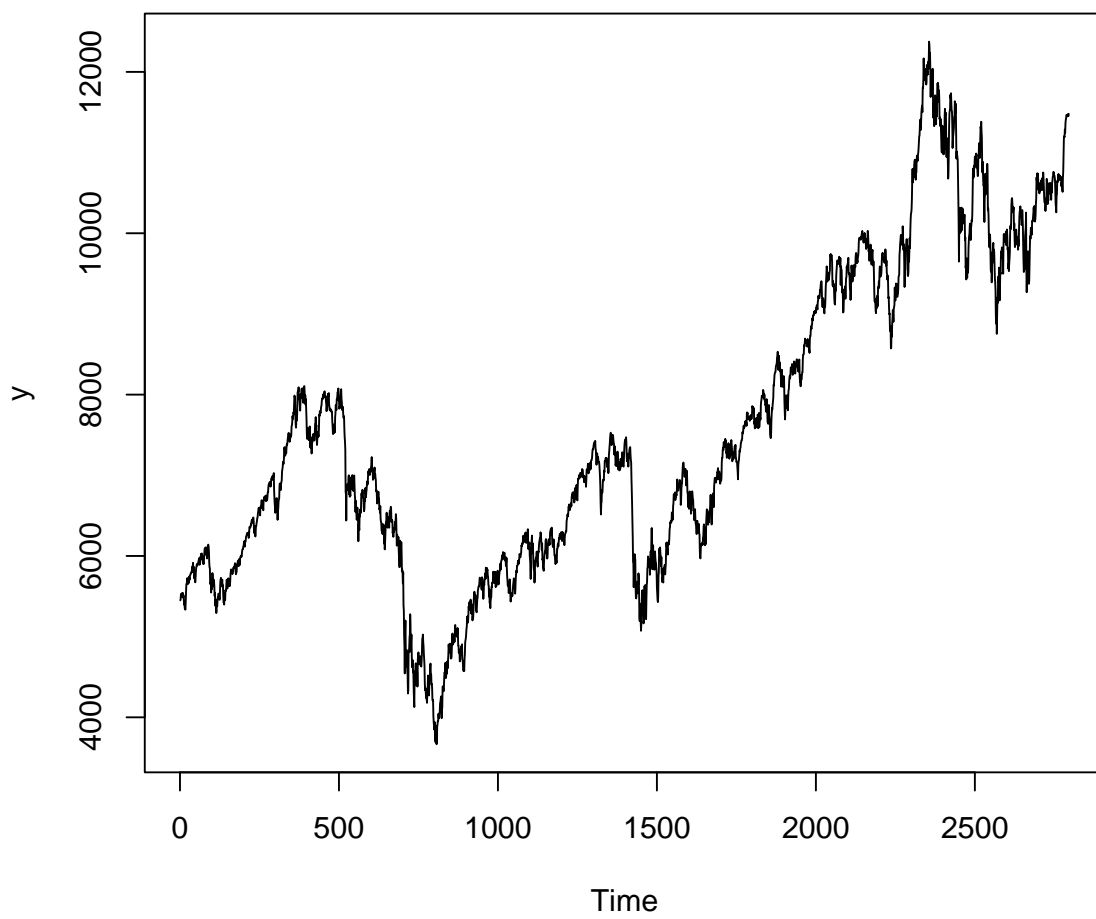
infocriteria()
names(yGARCH_fit@model)
names(yGARCH_fit@fit)
yGARCH_fit@fit$coef
plot(yGARCH_fit@fit$?,type="l")
```

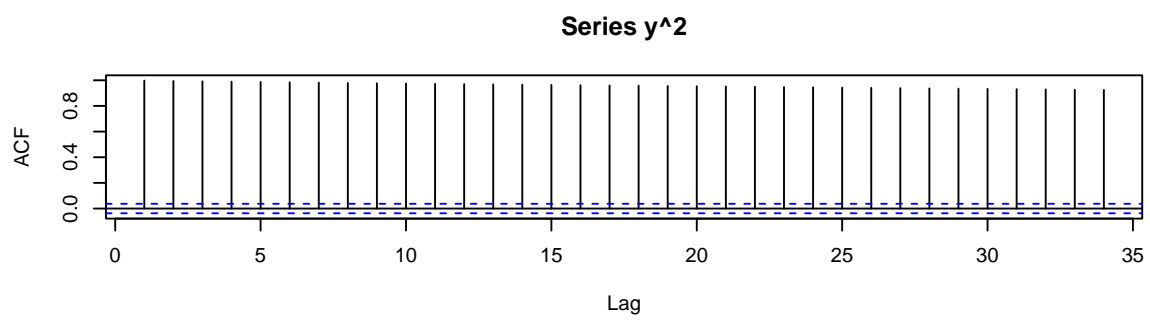
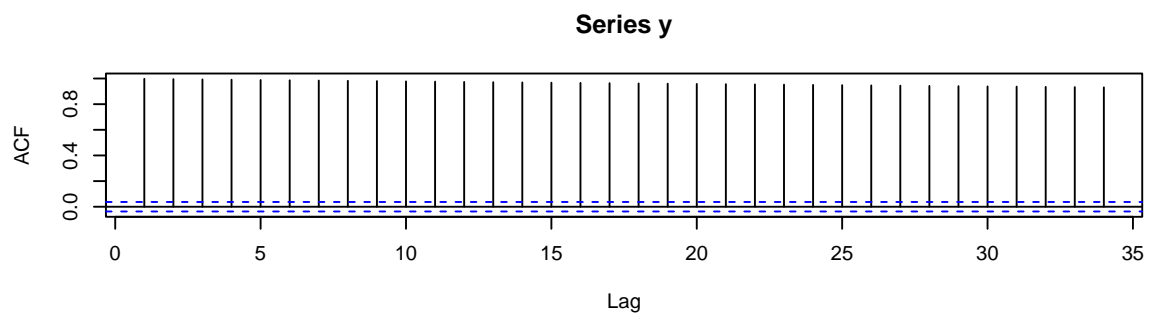
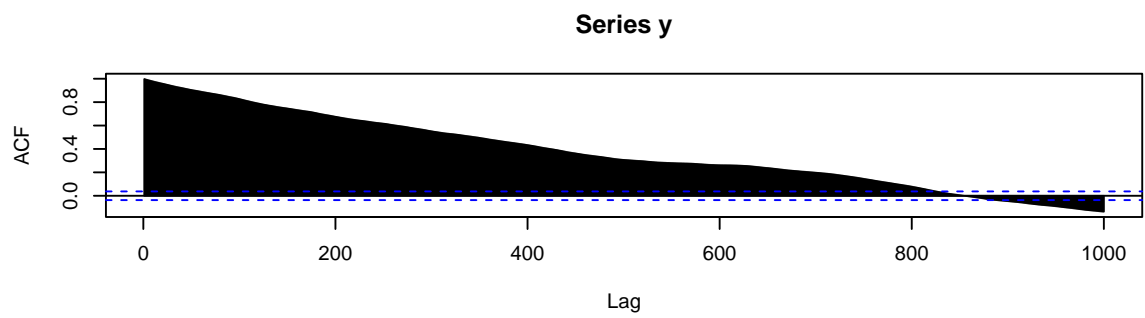
Wyznaczmy standaryzowane reszty z dopasowania:

```
library(car)
res_n<-fit_n@residuals/fit_n@sigma.t#Garch residuals (normal)
res_t<-fit_t@residuals/fit_t@sigma.t#Garch residuals (student)
plot(res_n-res_t)
qqPlot(res_n,dist=?)
qqPlot(res_t,dist=?)
```

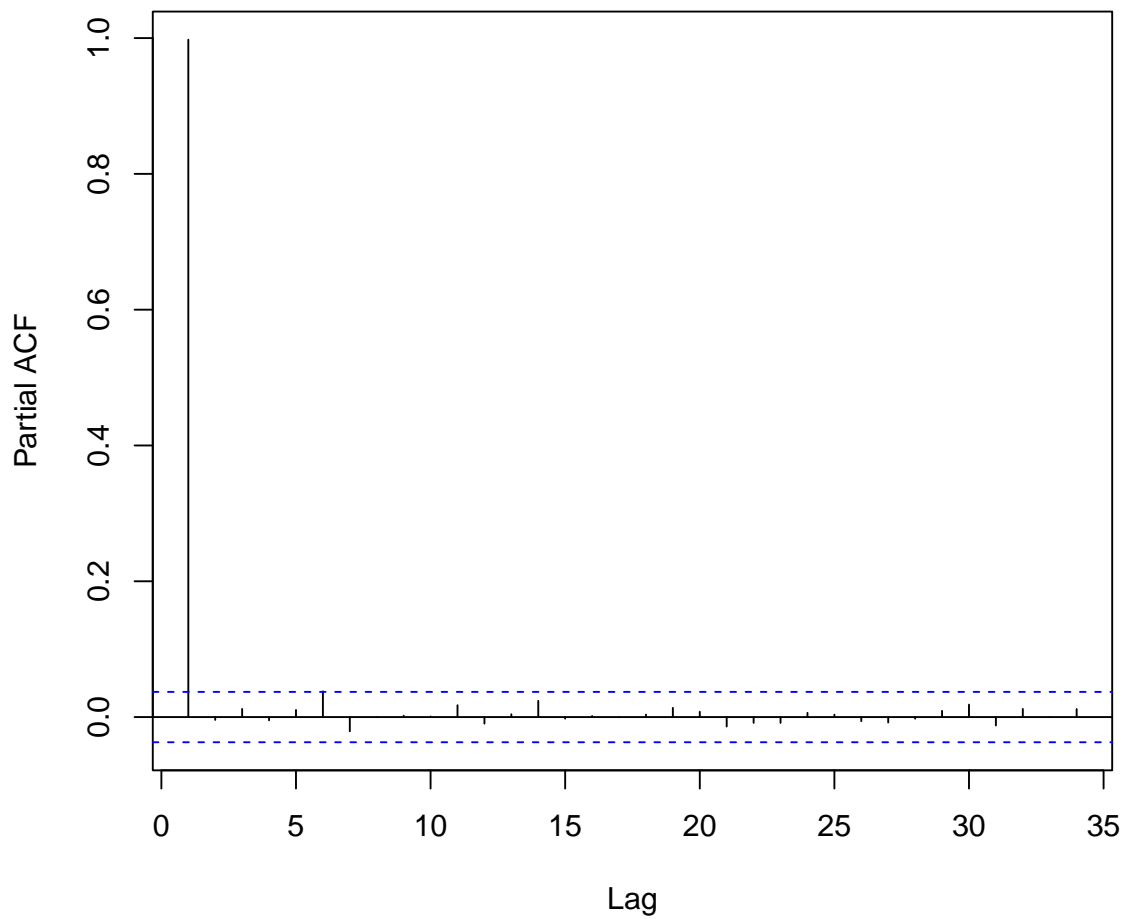
Forecast:

```
yGARCH_fit<-ugarchfit(data = y, spec = yGARCH_spec, out.sample=?, solver = "hybrid")
forecast <-ugarchforecast(yGARCH_fit, n.ahead=?, n.roll = ?,out.sample=?)
par(mfrow = c(1, 1))
plot(forecast, which = "all")
```





Series y



```
##
## Box-Ljung test
##
## data: y
## X-squared = 13806, df = 5, p-value < 2.2e-16
##
## Box-Ljung test
##
## data: y^2
## X-squared = 13782, df = 5, p-value < 2.2e-16
##
## Akaike      12.07239
## Bayes      12.08301
## Shibata    12.07238
## Hannan-Quinn 12.07622
##
## *-----*
```

```

## *          GARCH Model Fit          *
## *-----*
##
## Conditional Variance Dynamics
## -----
## GARCH Model   : sGARCH(1,1)
## Mean Model    : ARFIMA(1,0,0)
## Distribution   : norm
##
## Optimal Parameters
## -----
##           Estimate  Std. Error   t value Pr(>|t|)
## mu       1.6742e+04  4.2281e+03    3.9596  7.5e-05
## ar1       9.9954e-01  2.5300e-04 3958.2034  0.0e+00
## omega     8.0396e+03  2.8025e+02   28.6874  0.0e+00
## alpha1    2.2229e-01  2.9352e-02    7.5733  0.0e+00
## beta1     0.0000e+00  6.0500e-04    0.0000  1.0e+00
##
## Robust Standard Errors:
##           Estimate  Std. Error   t value Pr(>|t|)
## mu       1.6742e+04  6.1975e+03    2.7014  0.006906
## ar1       9.9954e-01  2.6300e-04 3804.5277  0.000000
## omega     8.0396e+03  6.0813e+02   13.2203  0.000000
## alpha1    2.2229e-01  4.8001e-02    4.6309  0.000004
## beta1     0.0000e+00  3.7900e-04    0.0000  1.000000
##
## LogLikelihood : -16866.16
##
## Information Criteria
## -----
##
## Akaike          12.072
## Bayes           12.083
## Shibata         12.072
## Hannan-Quinn    12.076
##
## Weighted Ljung-Box Test on Standardized Residuals
## -----
##                               statistic p-value
## Lag[1]                                0.02936  0.8639
## Lag[2*(p+q)+(p+q)-1] [2]    0.18657  0.9991
## Lag[4*(p+q)+(p+q)-1] [5]    1.15499  0.9210
## d.o.f=1
## H0 : No serial correlation
##
## Weighted Ljung-Box Test on Standardized Squared Residuals
## -----
##                               statistic p-value
## Lag[1]                                0.0006711  0.9793
## Lag[2*(p+q)+(p+q)-1] [5]    0.0007787  1.0000
## Lag[4*(p+q)+(p+q)-1] [9]    0.0008802  1.0000
## d.o.f=2
##
## Weighted ARCH LM Tests

```

```

## -----
##           Statistic Shape Scale P-Value
## ARCH Lag[3] 6.448e-05 0.500 2.000 0.9936
## ARCH Lag[5] 1.479e-04 1.440 1.667 1.0000
## ARCH Lag[7] 1.883e-04 2.315 1.543 1.0000
##
## Nyblom stability test
## -----
## Joint Statistic: 8.8721
## Individual Statistics:
## mu      0.8584
## ar1     0.2061
## omega   5.9637
## alpha1  1.7499
## beta1   0.9699
##
## Asymptotic Critical Values (10% 5% 1%)
## Joint Statistic:      1.28 1.47 1.88
## Individual Statistic: 0.35 0.47 0.75
##
## Sign Bias Test
## -----
##           t-value      prob sig
## Sign Bias      1.3238 0.18567
## Negative Sign Bias 0.2768 0.78194
## Positive Sign Bias 1.3043 0.19225
## Joint Effect      8.0857 0.04427 **
##
##
## Adjusted Pearson Goodness-of-Fit Test:
## -----
##   group statistic p-value(g-1)
## 1    20      197.5    1.051e-31
## 2    30      216.1    1.662e-30
## 3    40      235.1    7.525e-30
## 4    50      249.6    1.124e-28
##
##
## Elapsed time : 19.9334

## [1] "modelinc" "modeldesc" "modeldata" "pars" "start.pars"
## [6] "fixed.pars" "maxOrder" "pos.matrix" "fmodel" "pidx"
## [11] "n.start"

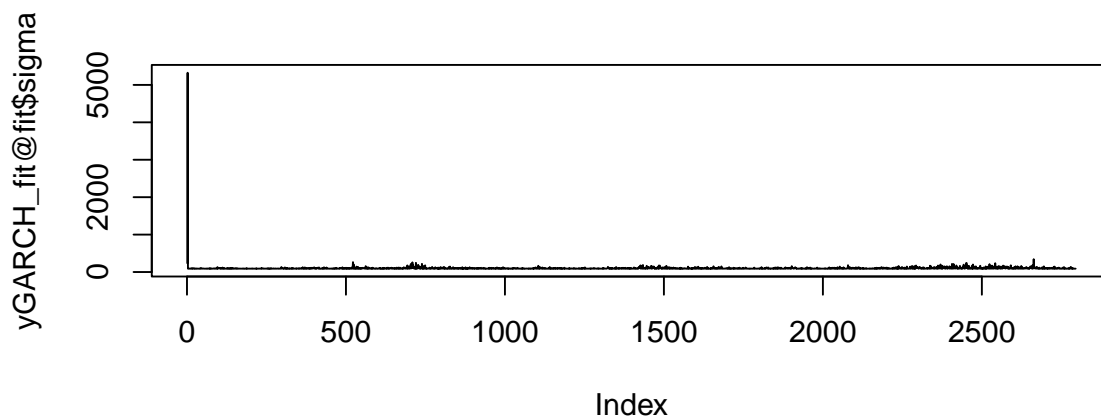
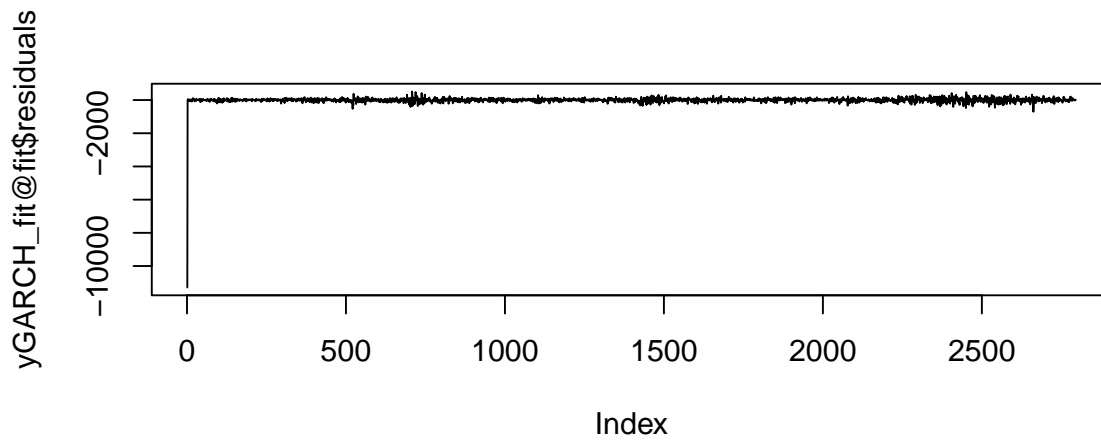
## [1] "hessian" "cvar" "var" "sigma"
## [5] "condH" "z" "LLH" "log.likelihoods"
## [9] "residuals" "coef" "robust.cvar" "A"
## [13] "B" "scores" "se.coef" "tval"
## [17] "matcoef" "robust.se.coef" "robust.tval" "robust.matcoef"
## [21] "fitted.values" "convergence" "kappa" "persistence"
## [25] "timer" "ipars" "solver"

##           mu      ar1      omega      alpha1      beta1
## 1.674172e+04 9.995421e-01 8.039603e+03 2.222892e-01 3.177549e-11

##           ar1

```

```
## 0.9995421
```



```
##
## Akaike      11.78787
## Bayes      11.80486
## Shibata    11.78785
## Hannan-Quinn 11.79400
##
## *-----*
## *          GARCH Model Fit          *
## *-----*
##
## Conditional Variance Dynamics
## -----
## GARCH Model : gjrGARCH(1,1)
## Mean Model  : ARFIMA(1,0,0)
## Distribution : sstd
##
## Optimal Parameters
```

```

## -----
##           Estimate Std. Error   t value Pr(>|t|)
## mu      5.4317e+03   73.640292   73.75999 0.000000
## ar1     1.0000e+00    0.000567 1764.98629 0.000000
## omega   1.5917e+02   38.559165    4.12803 0.000037
## alpha1  3.2780e-03    0.009254    0.35419 0.723194
## beta1   9.0430e-01    0.012575   71.91545 0.000000
## gamma1  1.6586e-01    0.025291    6.55801 0.000000
## skew    8.8512e-01    0.021638   40.90579 0.000000
## shape   7.5736e+00    1.109048    6.82895 0.000000
##
## Robust Standard Errors:
##           Estimate Std. Error   t value Pr(>|t|)
## mu      5.4317e+03   13.446578  403.9472 0.000000
## ar1     1.0000e+00    0.000646 1549.0653 0.000000
## omega   1.5917e+02   44.370446    3.5874 0.000334
## alpha1  3.2780e-03    0.011440    0.2865 0.774493
## beta1   9.0430e-01    0.013995   64.6182 0.000000
## gamma1  1.6586e-01    0.029843    5.5579 0.000000
## skew    8.8512e-01    0.017379   50.9300 0.000000
## shape   7.5736e+00    1.057408    7.1625 0.000000
##
## LogLikelihood : -16465.55
##
## Information Criteria
## -----
##
## Akaike          11.788
## Bayes           11.805
## Shibata         11.788
## Hannan-Quinn   11.794
##
## Weighted Ljung-Box Test on Standardized Residuals
## -----
##                               statistic p-value
## Lag[1]                      0.2724  0.6017
## Lag[2*(p+q)+(p+q)-1] [2]    0.3275  0.9920
## Lag[4*(p+q)+(p+q)-1] [5]    2.3731  0.6028
## d.o.f=1
## H0 : No serial correlation
##
## Weighted Ljung-Box Test on Standardized Squared Residuals
## -----
##                               statistic p-value
## Lag[1]                      1.881  0.1702
## Lag[2*(p+q)+(p+q)-1] [5]    2.927  0.4206
## Lag[4*(p+q)+(p+q)-1] [9]    3.554  0.6650
## d.o.f=2
##
## Weighted ARCH LM Tests
## -----
##           Statistic Shape Scale P-Value
## ARCH Lag[3]    0.9443 0.500 2.000  0.3312
## ARCH Lag[5]    1.1024 1.440 1.667  0.7028

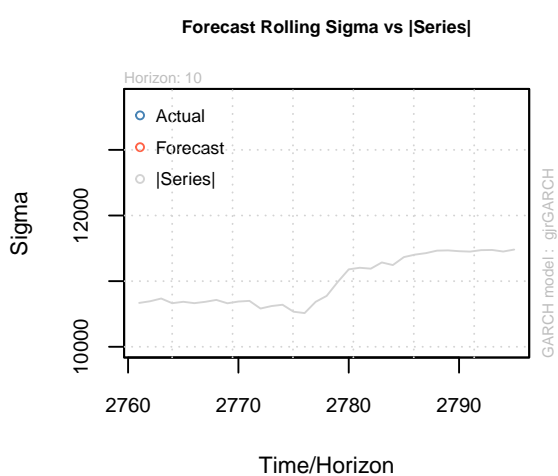
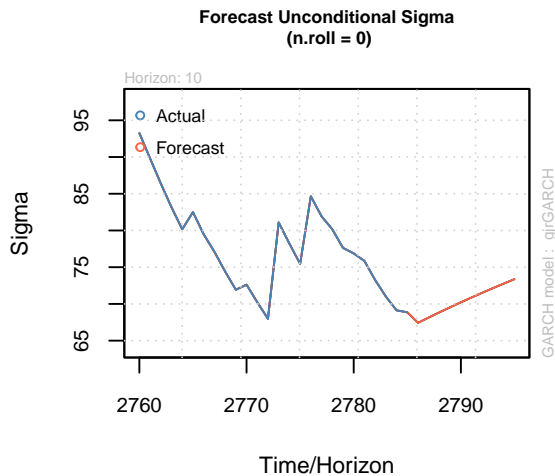
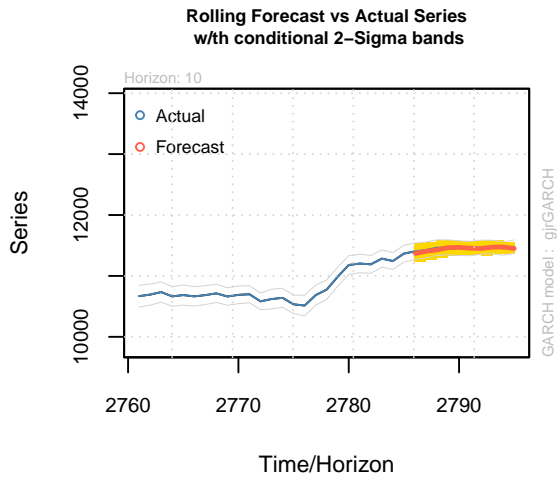
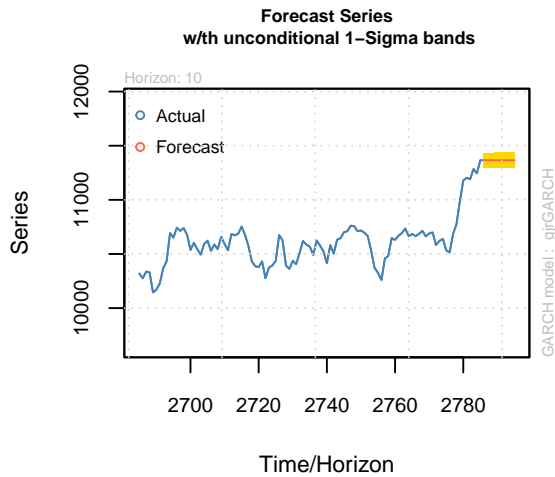
```

```

## ARCH Lag[7]      1.2625 2.315 1.543 0.8675
##
## Nyblom stability test
## -----
## Joint Statistic: 2.9408
## Individual Statistics:
## mu      0.00141
## ar1     0.44107
## omega   0.72881
## alpha1  0.85825
## beta1   0.77170
## gamma1  0.68595
## skew    0.25382
## shape   0.07983
##
## Asymptotic Critical Values (10% 5% 1%)
## Joint Statistic:      1.89 2.11 2.59
## Individual Statistic: 0.35 0.47 0.75
##
## Sign Bias Test
## -----
##              t-value   prob sig
## Sign Bias      1.029 0.30367
## Negative Sign Bias 1.237 0.21616
## Positive Sign Bias 1.704 0.08855 *
## Joint Effect    7.552 0.05624 *
##
##
## Adjusted Pearson Goodness-of-Fit Test:
## -----
##   group statistic p-value(g-1)
## 1    20      29.77    0.05486
## 2    30      40.45    0.07691
## 3    40      57.45    0.02862
## 4    50      62.26    0.09663
##
##
## Elapsed time : 1.801423
##
## *-----*
## *      GARCH Model Forecast      *
## *-----*
## Model: gjrGARCH
## Horizon: 10
## Roll Steps: 10
## Out of Sample: 10
##
## 0-roll forecast [T0=2785-01-01]:
##      Series Sigma
## T+1    11366 67.43
## T+2    11366 68.15
## T+3    11366 68.85
## T+4    11366 69.54

```

```
## T+5    11366 70.22
## T+6    11366 70.88
## T+7    11366 71.52
## T+8    11366 72.15
## T+9    11366 72.77
## T+10   11366 73.37
```



```
##
## *-----*
## *      GARCH Model Forecast      *
## *-----*
## Model: gjrGARCH
## Horizon: 10
## Roll Steps: 10
## Out of Sample: 10
##
## 0-roll forecast [T0=2785-01-01]:
##      Series Sigma
## T+1    11366 67.43
```


##	T+2	11366	68.15
##	T+3	11366	68.85
##	T+4	11366	69.54
##	T+5	11366	70.22
##	T+6	11366	70.88
##	T+7	11366	71.52
##	T+8	11366	72.15
##	T+9	11366	72.77
##	T+10	11366	73.37

Zadanie 4 - zadanie domowe

Dla wybranych szeregów czasowych wykonaj powyższe polecenia