

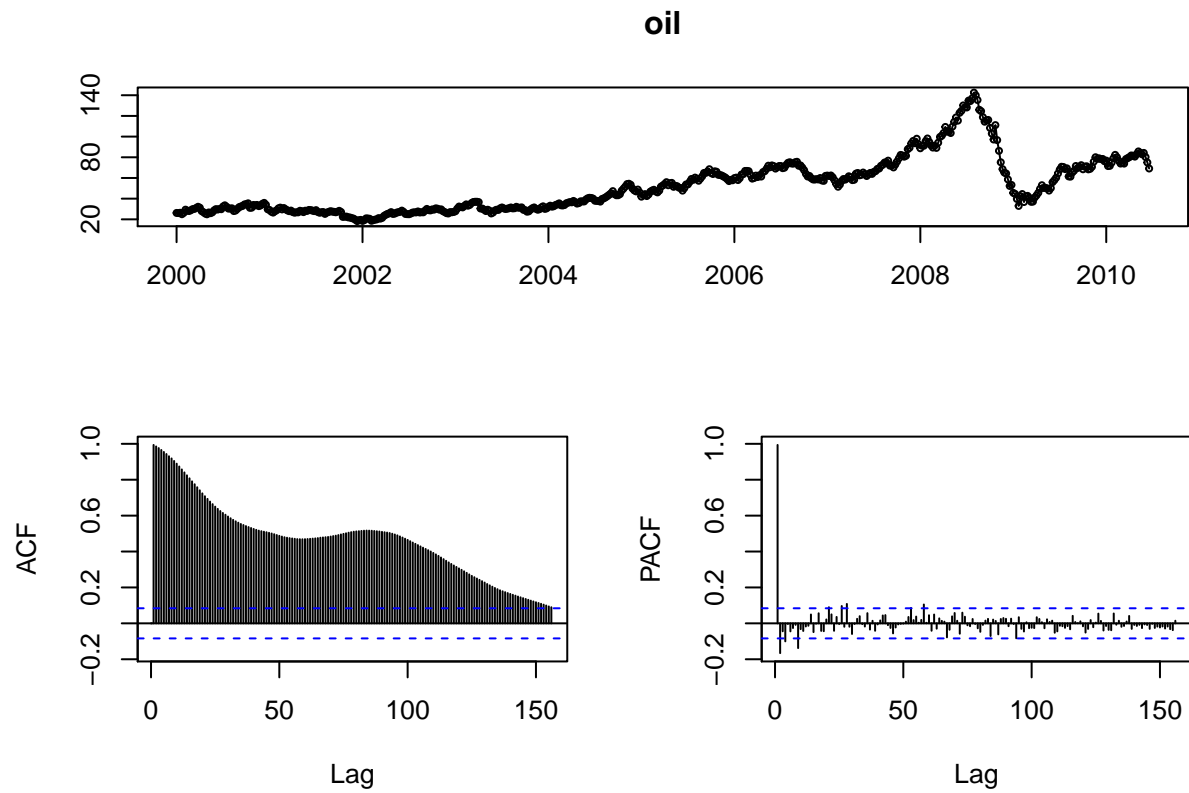
STAT 621 HW 7 GARCH

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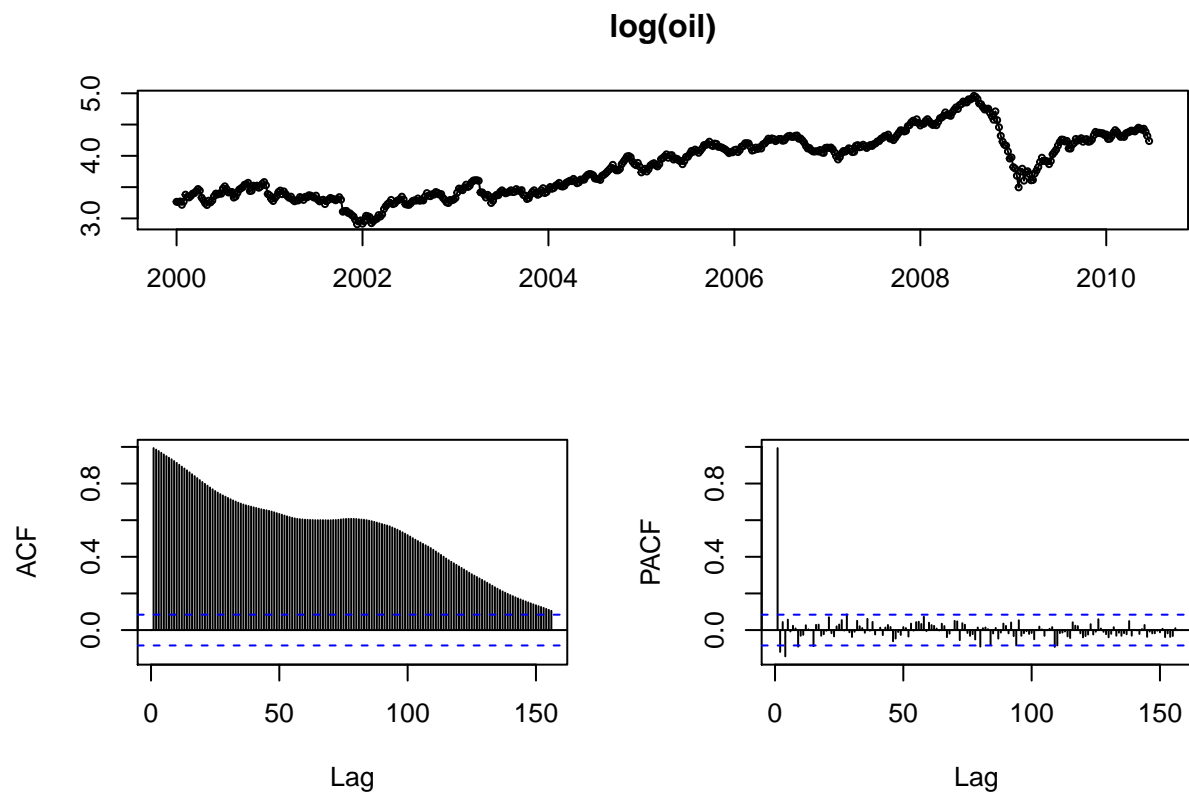
4/7/2018 (Due)

Q1

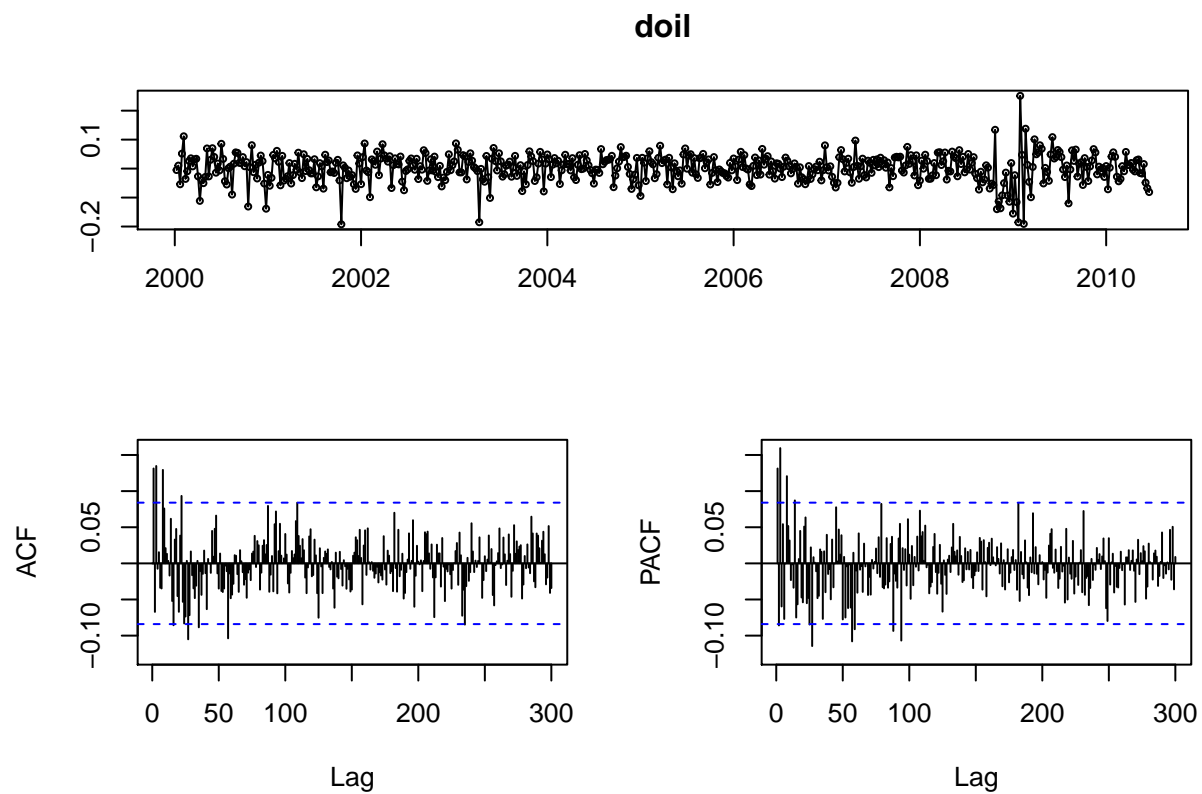
```
tsdisplay(oil)
```



```
tsdisplay(log(oil))
```

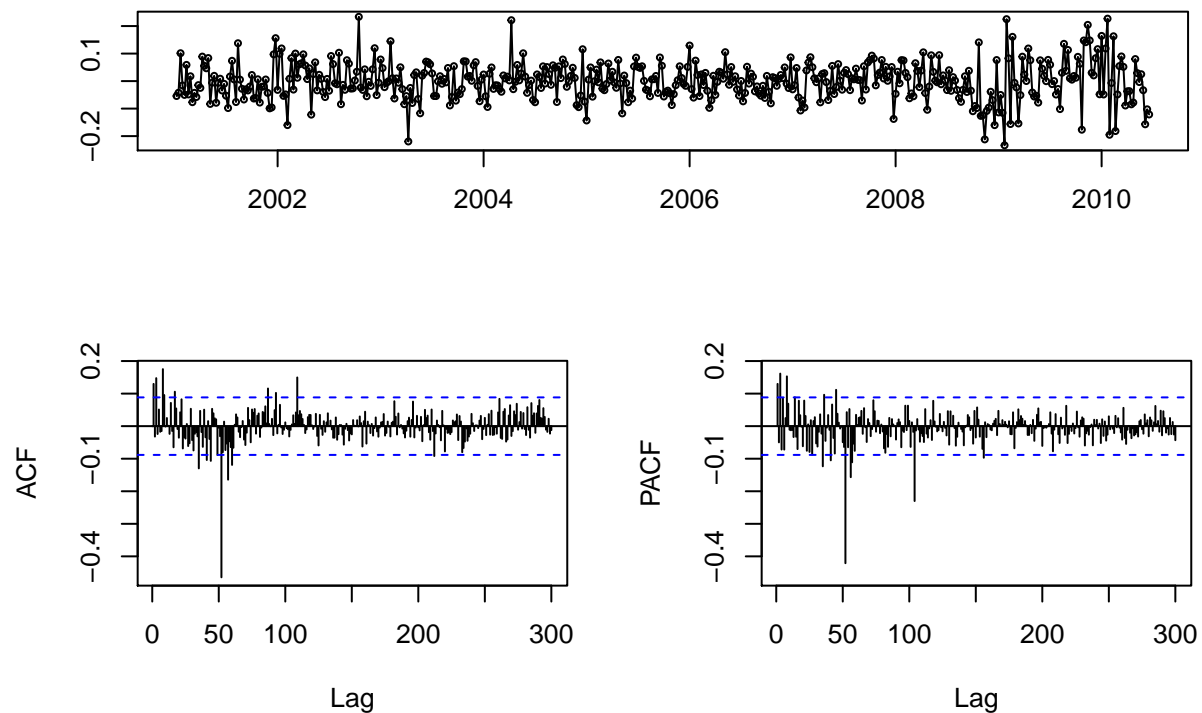


```
doil=diff(log(oil))  
tsdisplay(doil,lag.max = 300)
```



```
doil_52=diff(doil,52)    # Since it is weekly data  
tsdisplay(doil_52,lag.max = 300)
```

doil_52



```
auto.arima(log(oil))
```

```
## Series: log(oil)
## ARIMA(3,1,0)(1,0,1)[52]
##
## Coefficients:
##          ar1      ar2      ar3      sar1      sma1
##          0.1586 -0.1105  0.1636  0.1446 -0.1952
## s.e.    0.0424   0.0428  0.0425  0.8856  0.8746
##
## sigma^2 estimated as 0.002111:  log likelihood=906.13
## AIC=-1800.27   AICc=-1800.11   BIC=-1774.48
```

Auto.Arima shows it might be a [3, 1, 0][1, 0, 1][52] Sarima model.

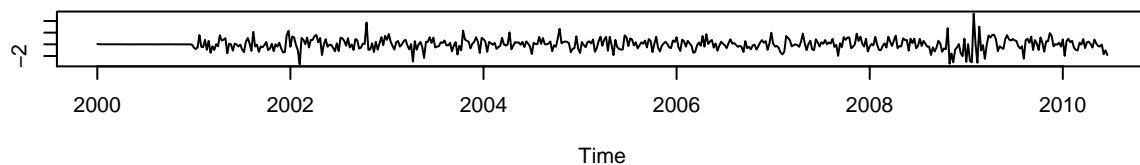
```
sarima(log(oil),p=3,d=1,q=0,P=0,D=1,Q=1,S=52,xreg=1:545,no.constant=T, details=T)
```

```
## initial  value -2.690135
## iter    2 value -2.887661
## iter    3 value -2.920082
## iter    4 value -2.926528
## iter    5 value -2.930528
## iter    6 value -2.931395
## iter    7 value -2.931449
## iter    8 value -2.931449
## iter    8 value -2.931449
## iter    8 value -2.931449
## final    value -2.931449
```

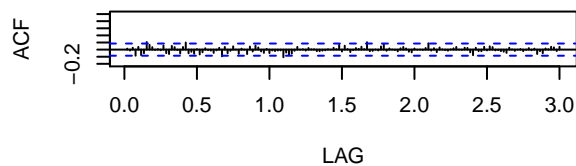
```
## converged
## initial value -2.946806
## iter 2 value -2.959379
## iter 3 value -2.960304
## iter 4 value -2.960508
## iter 5 value -2.960561
## iter 6 value -2.960575
## iter 7 value -2.960578
## iter 8 value -2.960578
## iter 8 value -2.960578
## iter 8 value -2.960578
## final value -2.960578
## converged
```

Model: (3,1,0) (0,1,1) [52]

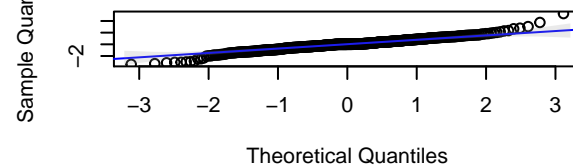
Standardized Residuals



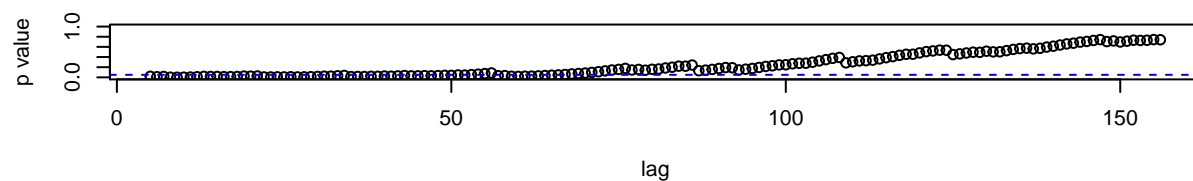
ACF of Residuals



Normal Q-Q Plot of Std Residuals



p values for Ljung-Box statistic

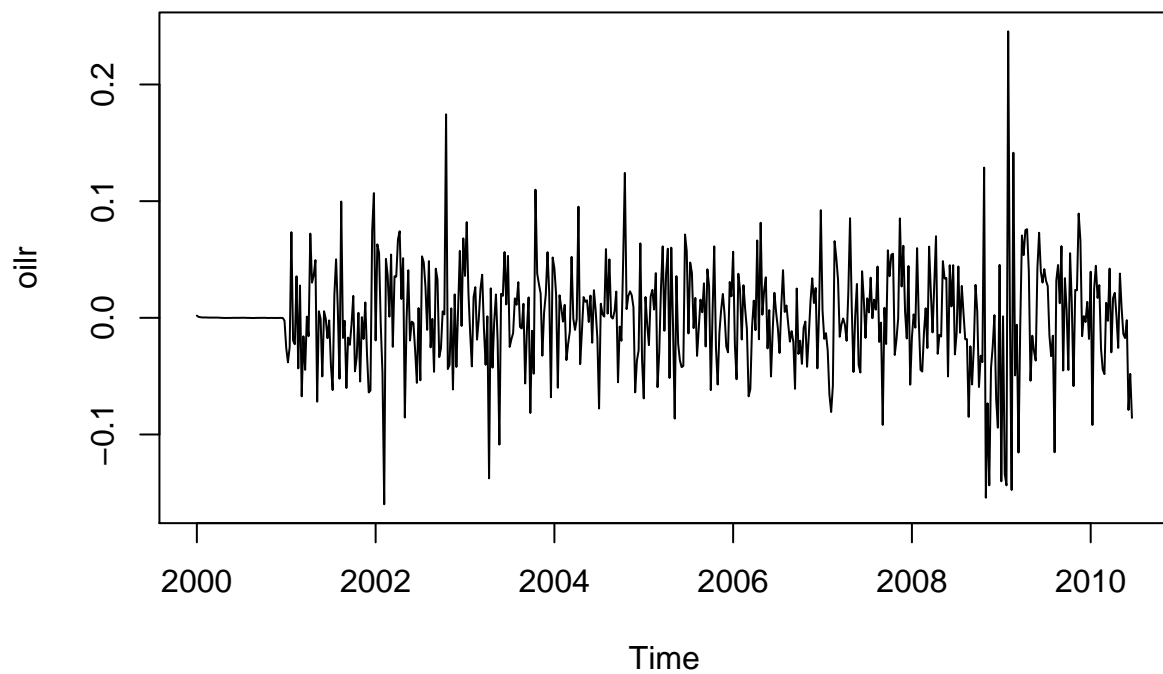


```
## $fit
##
## Call:
## stats::arima(x = xdata, order = c(p, d, q), seasonal = list(order = c(P, D,
## Q), period = S), xreg = xreg, optim.control = list(trace = trc, REPORT = 1,
## reltol = tol))
##
## Coefficients:
##          ar1          ar2          ar3          sma1          xreg
##          0.1506    -0.0993    0.1535    -0.9604    0.0112
## s.e.    0.0447     0.0451    0.0448     0.2199    5.4448
##
```

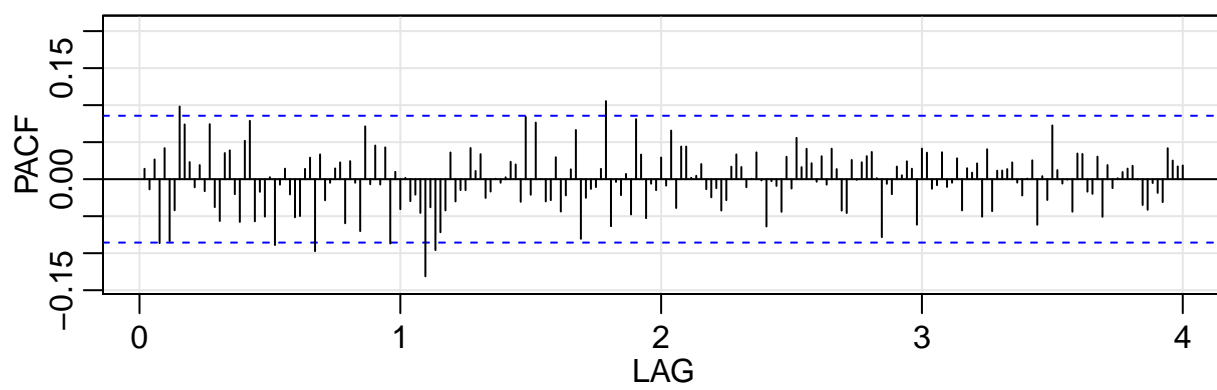
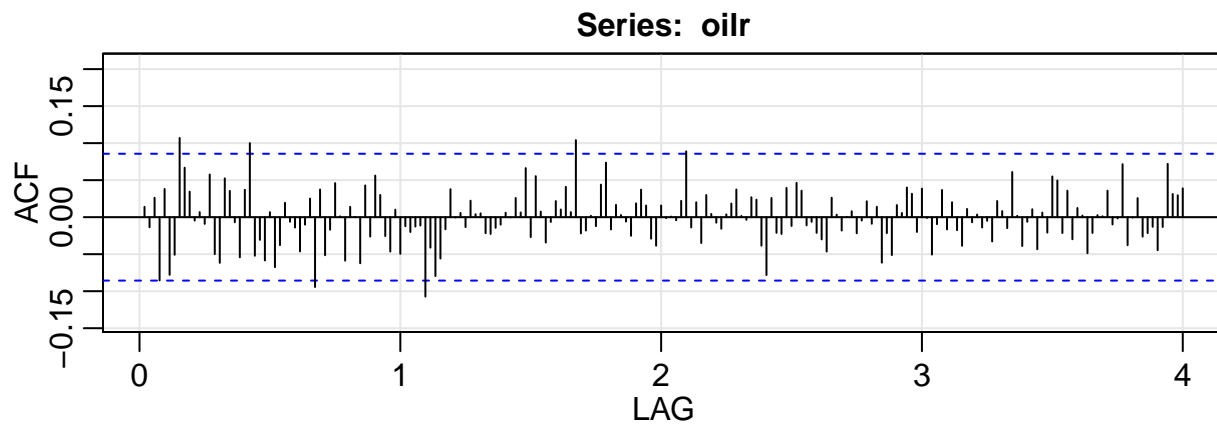
```
## sigma^2 estimated as 0.002172: log likelihood = 758.49, aic = -1504.97
##
## $degrees_of_freedom
## [1] 487
##
## $ttable
##      Estimate      SE t.value p.value
## ar1      0.1506 0.0447  3.3727 0.0008
## ar2     -0.0993 0.0451 -2.2022 0.0281
## ar3      0.1535 0.0448  3.4244 0.0007
## sma1    -0.9604 0.2199 -4.3674 0.0000
## xreg      0.0112 5.4448  0.0021 0.9984
##
## $AIC
## [1] -5.113762
##
## $AICc
## [1] -5.109806
##
## $BIC
## [1] -6.074305
```

The normal Q-Q Plot of Std Residuals suggests it might fit a (G)arch model.

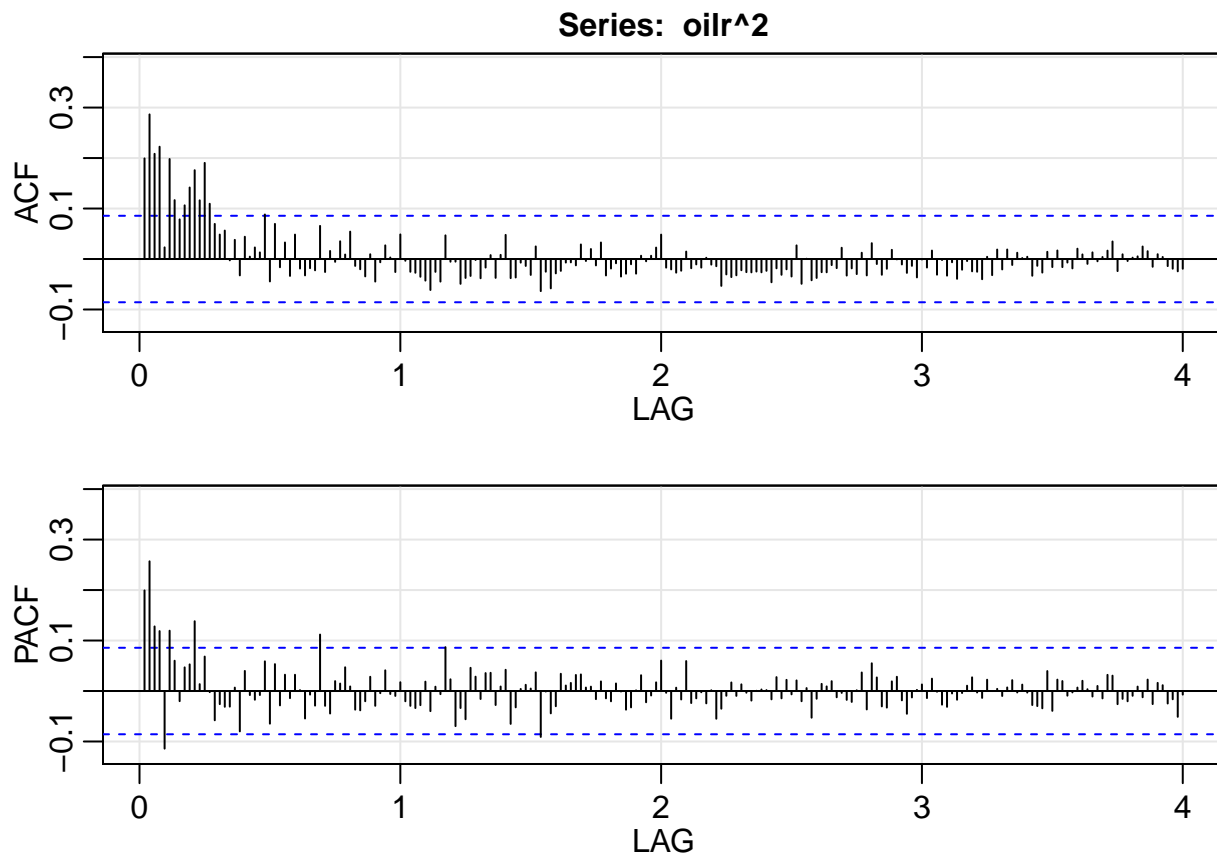
```
oilr <- resid(sarima(log(oil),p=3,d=1,q=0,P=0,D=1,Q=1,S=52,xreg=1:545,no.constant=T, details=F)$fit )
invisible(plot(oilr))           # plot residuals
```



```
invisible(acf2(oilr))           # P/ACF suggest noise
```



```
invisible(acf2(oilr^2))      # some correlation
```

The $residual^2$ ($oilr^2$) suggests it could be a GARCH(2,0) model.

```
summary(fit <- garchFit(~garch(2,0), data=oilr)) # alpha1 and alpha2 are significant
```

```
##
## Series Initialization:
## ARMA Model:          arma
## Formula Mean:        ~ arma(0, 0)
## GARCH Model:         garch
## Formula Variance:    ~ garch(2, 0)
## ARMA Order:          0 0
## Max ARMA Order:      0
## GARCH Order:         2 0
## Max GARCH Order:     2
## Maximum Order:       2
## Conditional Dist:    norm
## h.start:             3
## llh.start:           1
## Length of Series:    545
## Recursion Init:      mci
## Series Scale:        0.04433526
##
## Parameter Initialization:
## Initial Parameters:   $params
## Limits of Transformations: $U, $V
## Which Parameters are Fixed? $includes
## Parameter Matrix:
```

```

##          U          V      params includes
## mu      -0.02603469  0.02603469 0.002603469    TRUE
## omega   0.00000100 100.00000000 0.100000000    TRUE
## alpha1  0.00000001  0.99999999 0.050000000    TRUE
## alpha2  0.00000001  0.99999999 0.050000000    TRUE
## gamma1 -0.99999999  0.99999999 0.100000000    FALSE
## gamma2 -0.99999999  0.99999999 0.100000000    FALSE
## delta   0.00000000  2.00000000 2.000000000    FALSE
## skew    0.10000000 10.00000000 1.000000000    FALSE
## shape   1.00000000 10.00000000 4.000000000    FALSE
## Index List of Parameters to be Optimized:
## mu omega alpha1 alpha2
## 1 2 3 4
## Persistence: 0.1
##
## --- START OF TRACE ---
## Selected Algorithm: nlminb
##
## R coded nlminb Solver:
##
## 0: 1482.7699: 0.00260347 0.100000 0.0500000 0.0500000
## 1: 779.90516: 0.00260355 0.945127 0.423843 0.432102
## 2: 771.88002: 0.00260648 1.20391 0.0910835 0.163212
## 3: 769.11388: 0.00260794 1.18734 0.119879 0.0733098
## 4: 762.74041: 0.00260800 1.09856 0.0867253 0.0589984
## 5: 747.87045: 0.00260830 0.725400 0.0620264 0.143409
## 6: 746.08512: 0.00260847 0.724503 0.154185 0.167333
## 7: 745.72588: 0.00261575 0.637501 0.158162 0.128948
## 8: 745.21441: 0.00273234 0.649692 0.150425 0.182793
## 9: 745.18880: 0.00277573 0.647807 0.165099 0.172259
## 10: 745.18374: 0.00283920 0.647803 0.161565 0.171343
## 11: 745.18082: 0.00299634 0.648850 0.159509 0.171564
## 12: 745.17489: 0.00341365 0.649680 0.157334 0.171856
## 13: 745.15506: 0.00508302 0.651496 0.152632 0.172594
## 14: 745.11271: 0.00961196 0.653843 0.145909 0.174083
## 15: 745.07001: 0.0141409 0.654312 0.143765 0.175124
## 16: 744.95071: 0.0260347 0.647161 0.157548 0.176288
## 17: 744.95002: 0.0260347 0.645883 0.159178 0.175968
## 18: 744.95000: 0.0260347 0.645819 0.159547 0.175910
## 19: 744.95000: 0.0260347 0.645801 0.159545 0.175923
##
## Final Estimate of the Negative LLH:
## LLH: -953.2563 norm LLH: -1.749094
## mu omega alpha1 alpha2
## 0.001154255 0.001269397 0.159544520 0.175923064
##
## R-optimhess Difference Approximated Hessian Matrix:
## mu omega alpha1 alpha2
## mu -335579.5811 -170739.51 -248.8301 908.3523
## omega -170739.5058 -103232438.07 -92587.0529 -90981.7440
## alpha1 -248.8301 -92587.05 -404.7351 -119.1043
## alpha2 908.3523 -90981.74 -119.1043 -356.4917
## attr("time")

```

```

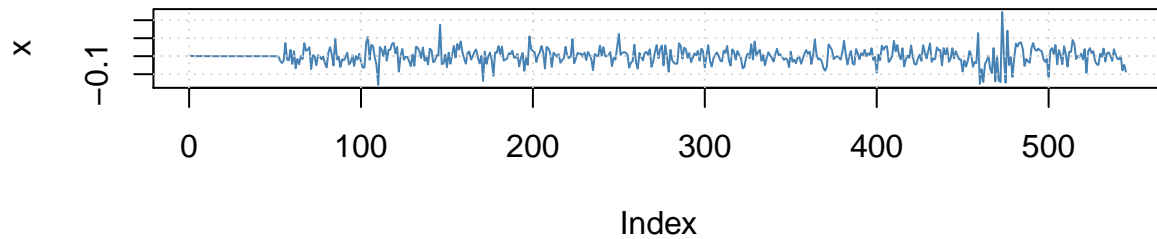
## Time difference of 0.02005816 secs
##
## --- END OF TRACE ---
##
##
## Time to Estimate Parameters:
## Time difference of 0.06942415 secs
##
## Title:
## GARCH Modelling
##
## Call:
## garchFit(formula = ~garch(2, 0), data = oilr)
##
## Mean and Variance Equation:
## data ~ garch(2, 0)
## <environment: 0x000000001e48ce30>
## [data = oilr]
##
## Conditional Distribution:
## norm
##
## Coefficient(s):
##      mu      omega    alpha1    alpha2
## 0.0011543 0.0012694 0.1595445 0.1759231
##
## Std. Errors:
## based on Hessian
##
## Error Analysis:
##      Estimate Std. Error t value Pr(>|t|)
## mu      0.0011543 0.0017380 0.664 0.50661
## omega 0.0012694 0.0001203 10.554 < 2e-16 ***
## alpha1 0.1595445 0.0562160 2.838 0.00454 **
## alpha2 0.1759231 0.0610256 2.883 0.00394 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Log Likelihood:
## 953.2563    normalized: 1.749094
##
## Description:
## Sat Apr 07 10:34:35 2018 by user: yydab
##
##
## Standardised Residuals Tests:
##
##      Statistic p-Value
## Jarque-Bera Test R Chi^2 66.44128 3.774758e-15
## Shapiro-Wilk Test R W 0.9814722 2.081421e-06
## Ljung-Box Test R Q(10) 14.81414 0.1389862
## Ljung-Box Test R Q(15) 19.72929 0.1825653
## Ljung-Box Test R Q(20) 23.57797 0.2613134
## Ljung-Box Test R^2 Q(10) 19.19412 0.03786519
## Ljung-Box Test R^2 Q(15) 22.86526 0.0870429

```

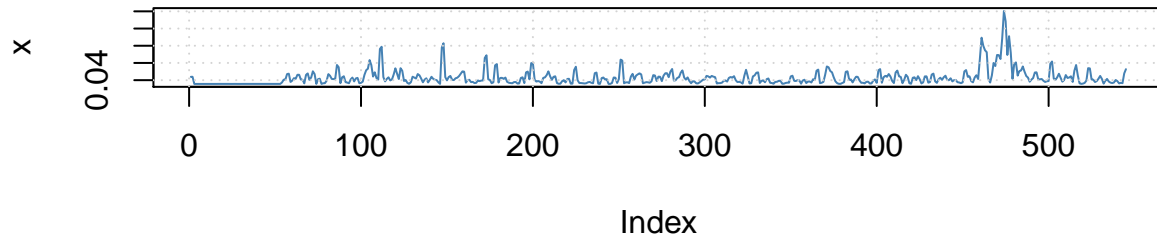
```
## Ljung-Box Test      R^2  Q(20)  25.29213  0.1904827
## LM Arch Test       R    TR^2   18.70875  0.0958038
##
## Information Criterion Statistics:
##      AIC      BIC      SIC      HQIC
## -3.483510 -3.451944 -3.483616 -3.471169
```

```
par(mfrow=2:1)
plot(fit, which=1:2)    # plot data and root volatility
```

Time Series

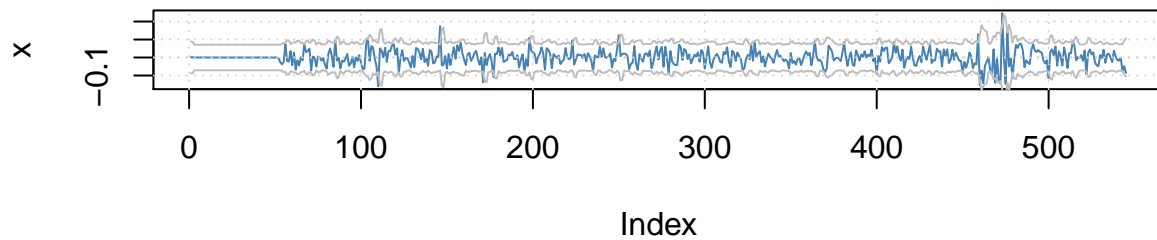


Conditional SD

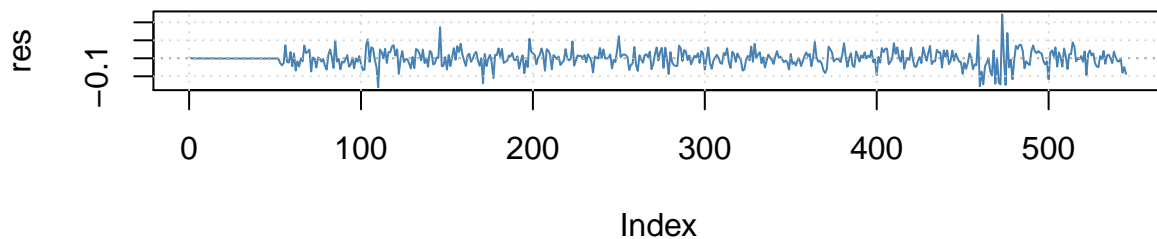


```
plot(fit, which=c(3,7)) # Series with 2 Conditional SD; Standardized Residuals
```

Series with 2 Conditional SD Superimposed



Residuals

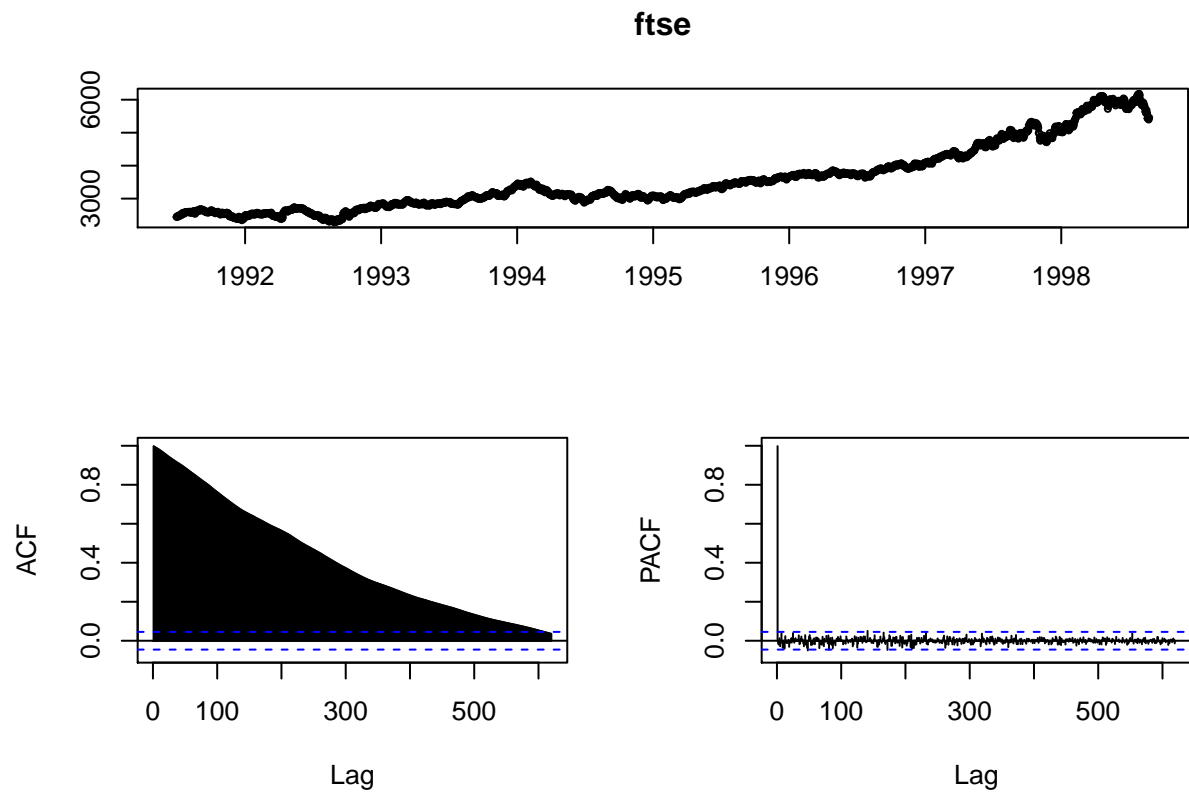


Q1 Conclusion:

First I use `auto.arima` to detect the model and then fit the `sarima` model. The normal Q-Q Plot of Std Residuals suggests it might fit a (G)arch model. After trying different garch combination, `Garch(2,0)` fit well. Both `alpha1` and `alpha2` are significant. I suggest it is a $SARIMA[3, 1, 0] \times [0, 1, 1]_{52} + GARCH(2, 0)$ model.

Q2 Eu Stock Markets

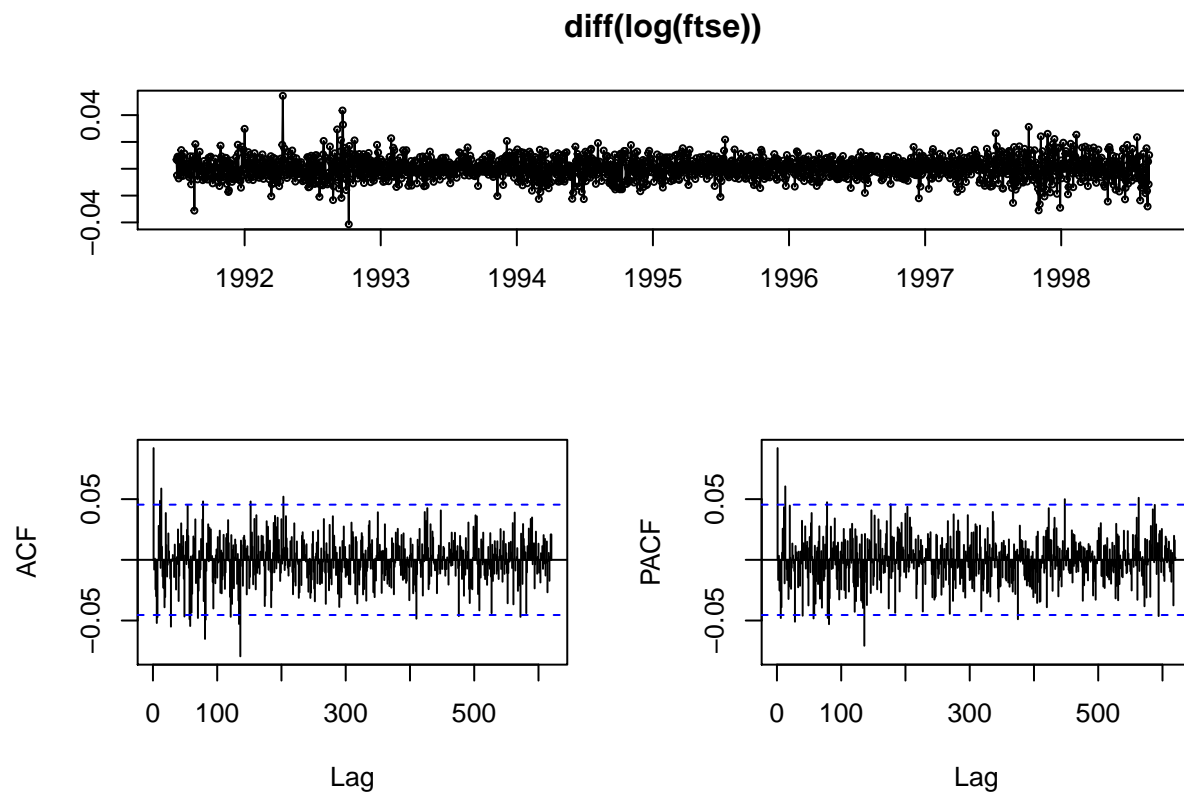
```
ftse<-EuStockMarkets[, "FTSE"]  
tsdisplay(ftse)
```



Transfer the data as return

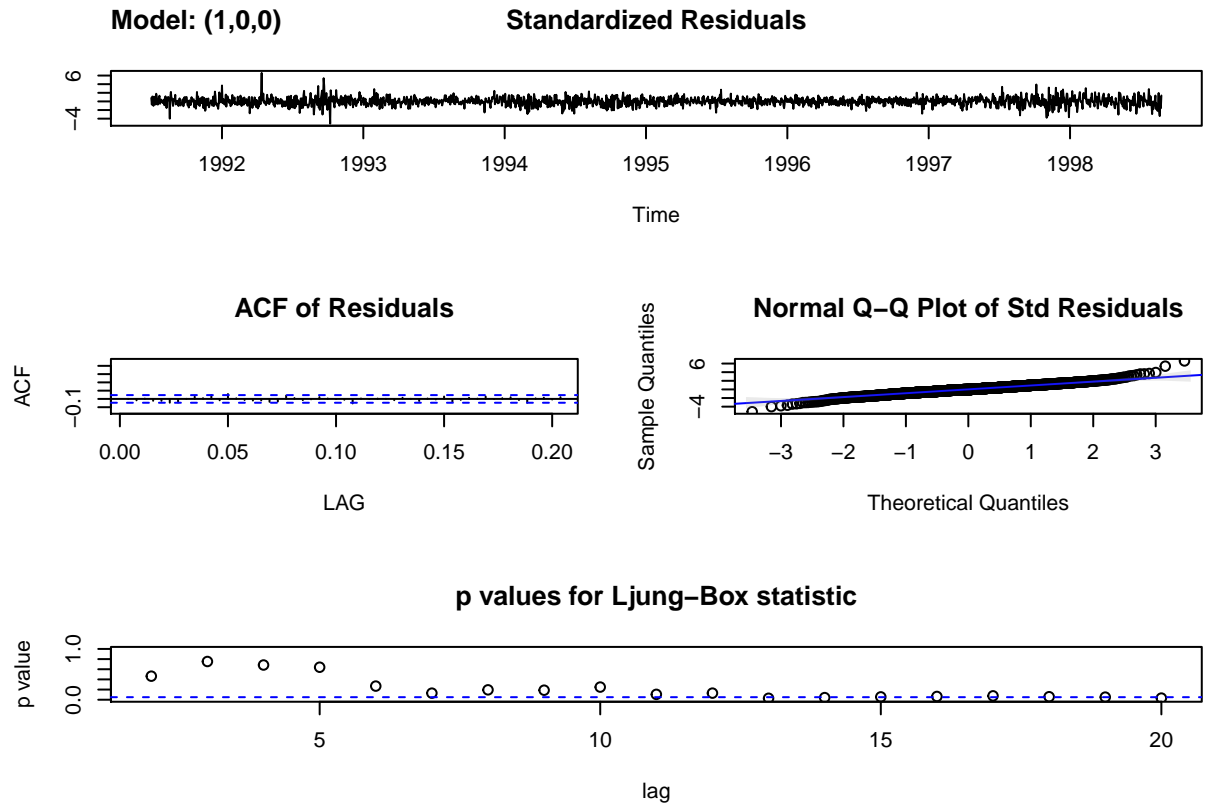
$r_t = \frac{x_t - x_{t-1}}{x_{t-1}} \approx \log(r_t) = \log\left(\frac{x_t}{x_{t-1}}\right)$, so we can use $\text{diff}(\log(\text{data}))$ as a proxy of growth rate.

```
tsdisplay(diff(log(ftse)))
```



```
sarima(diff(log(ftse)),1,0,0)
```

```
## initial value -4.833783
## iter 2 value -4.838040
## iter 2 value -4.838040
## iter 2 value -4.838040
## final value -4.838040
## converged
## initial value -4.838136
## iter 1 value -4.838136
## final value -4.838136
## converged
```



```
## $fit
##
## Call:
## stats::arima(x = xdata, order = c(p, d, q), seasonal = list(order = c(P, D,
##     Q), period = S), xreg = xmean, include.mean = FALSE, optim.control = list(trace = trc,
##     REPORT = 1, reltol = tol))
##
## Coefficients:
##      ar1  xmean
##    0.0921 4e-04
## s.e. 0.0231 2e-04
##
## sigma^2 estimated as 6.275e-05:  log likelihood = 6356.29,  aic = -12706.58
##
## $degrees_of_freedom
## [1] 1857
##
## $ttable
##      Estimate      SE t.value p.value
## ar1      0.0921 0.0231  3.9865  0.0001
## xmean     0.0004 0.0002  2.1190  0.0342
##
## $AIC
## [1] -8.674125
##
## $AICc
```

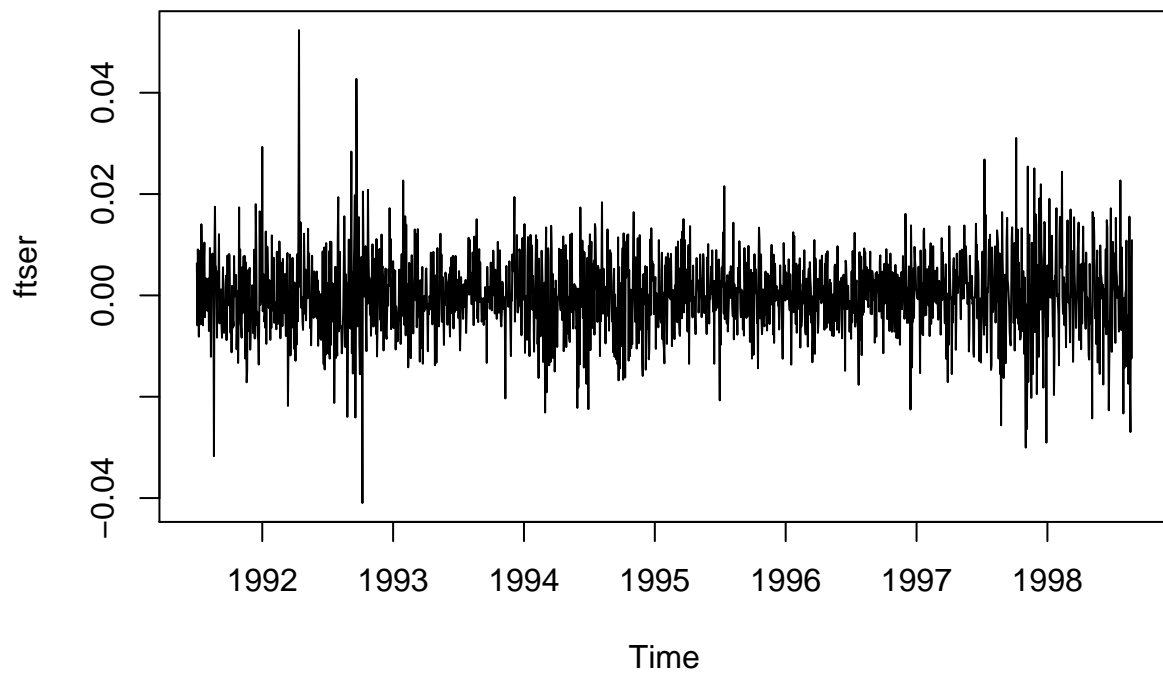


```
## [1] -8.673042
##
## $BIC
## [1] -9.668178
```

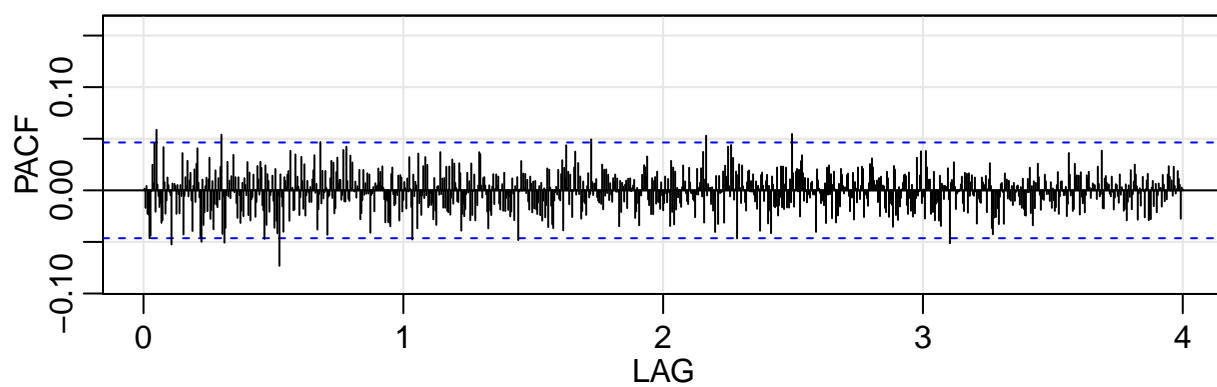
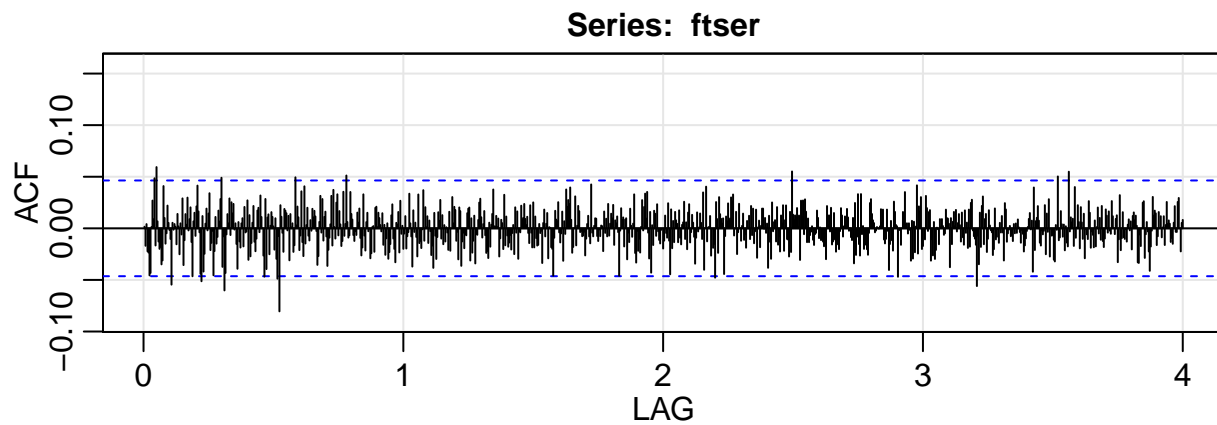
Get the residuals from an AR(1) fit

```
ftser <- resid(sarima(diff(log(ftse)), 1,0,0, details=FALSE)$fit )

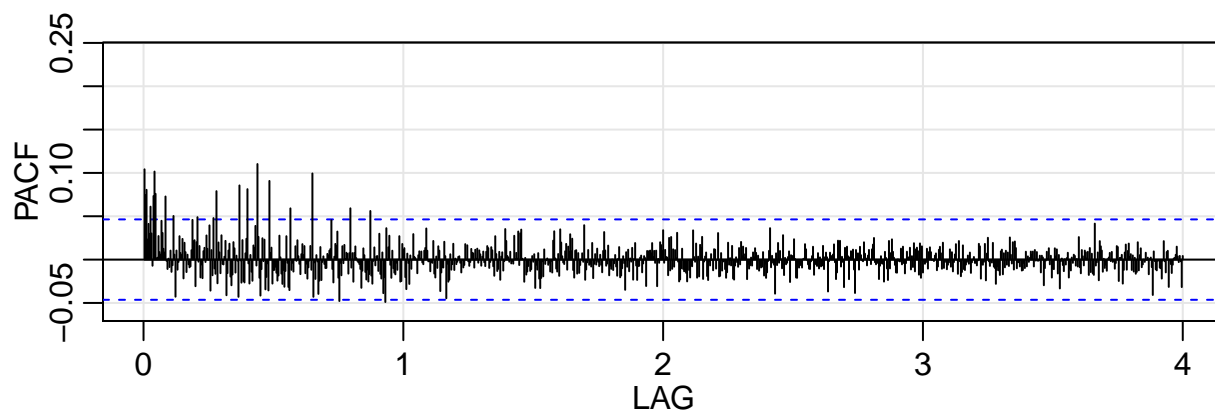
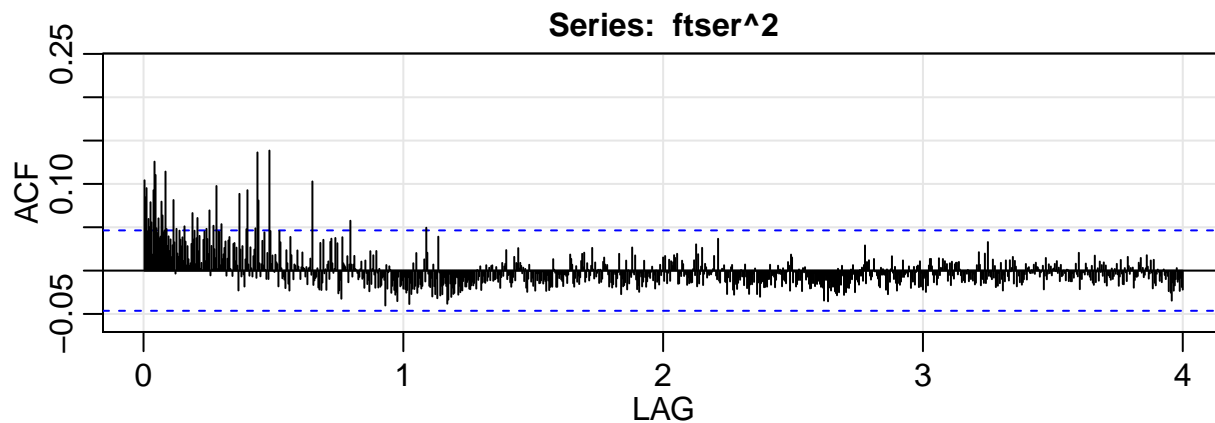
# Looking at residuals to growth rate
invisible(plot(ftser))           # plot residuals
```



```
invisible(acf2(ftser))           # P/ACF suggest noise
```



```
invisible(acf2(ftser^2))           # some correlation
```



```
# Let's try fitting an ARCH model
summary(fit <- garchFit(~garch(1,1), data=ftser))
```

```
##
## Series Initialization:
## ARMA Model:          arma
## Formula Mean:        ~ arma(0, 0)
## GARCH Model:         garch
## Formula Variance:    ~ garch(1, 1)
## ARMA Order:          0 0
## Max ARMA Order:      0
## GARCH Order:         1 1
## Max GARCH Order:     1
## Maximum Order:       1
## Conditional Dist:    norm
## h.start:             2
## llh.start:           1
## Length of Series:    1859
## Recursion Init:      mci
## Series Scale:        0.007923918
##
## Parameter Initialization:
## Initial Parameters:   $params
## Limits of Transformations: $U, $V
## Which Parameters are Fixed? $includes
## Parameter Matrix:
```

```

##          U          V      params includes
## mu      -0.00104750 1.0475e-03 0.00010475      TRUE
## omega    0.00000100 1.0000e+02 0.10000000      TRUE
## alpha1   0.00000001 1.0000e+00 0.10000000      TRUE
## gamma1  -0.99999999 1.0000e+00 0.10000000      FALSE
## beta1    0.00000001 1.0000e+00 0.80000000      TRUE
## delta    0.00000000 2.0000e+00 2.00000000      FALSE
## skew     0.10000000 1.0000e+01 1.00000000      FALSE
## shape    1.00000000 1.0000e+01 4.00000000      FALSE
## Index List of Parameters to be Optimized:
## mu omega alpha1 beta1
## 1 2 3 5
## Persistence: 0.9
##
## --- START OF TRACE ---
## Selected Algorithm: nlminb
##
## R coded nlminb Solver:
##
## 0: 2574.3908: 0.000104750 0.100000 0.100000 0.800000
## 1: 2573.6072: 0.000104750 0.0944825 0.104466 0.807772
## 2: 2572.5218: 0.000104750 0.0844104 0.104407 0.810830
## 3: 2571.5614: 0.000104750 0.0791065 0.107162 0.819495
## 4: 2569.9584: 0.000104750 0.0630579 0.105506 0.833019
## 5: 2565.1180: 0.000104750 0.0452323 0.0814825 0.875602
## 6: 2565.0303: 0.000104750 0.0439424 0.0808264 0.876184
## 7: 2564.9267: 0.000104751 0.0443359 0.0803625 0.877620
## 8: 2564.8015: 0.000104754 0.0436324 0.0776741 0.879037
## 9: 2564.7338: 0.000104757 0.0446117 0.0763941 0.881014
## 10: 2564.6639: 0.000104752 0.0417001 0.0804014 0.879793
## 11: 2564.4843: 0.000104752 0.0403824 0.0793960 0.884618
## 12: 2563.9799: 0.000104757 0.0387602 0.0752936 0.887181
## 13: 2563.5540: 0.000104766 0.0377754 0.0690680 0.895204
## 14: 2563.5508: 0.000104770 0.0360319 0.0667081 0.895864
## 15: 2563.3094: 0.000104778 0.0364159 0.0671538 0.897245
## 16: 2563.1482: 0.000104793 0.0336686 0.0674558 0.898423
## 17: 2562.8796: 0.000104816 0.0335808 0.0637604 0.903162
## 18: 2562.4372: 0.000104902 0.0263033 0.0602135 0.912023
## 19: 2562.2042: 0.000104926 0.0256757 0.0652918 0.911840
## 20: 2561.9950: 0.000104934 0.0264412 0.0608002 0.914187
## 21: 2561.9726: 0.000104934 0.0259773 0.0606038 0.914092
## 22: 2561.9455: 0.000104939 0.0259180 0.0606823 0.914593
## 23: 2561.8998: 0.000104954 0.0252346 0.0602931 0.915235
## 24: 2561.1735: 0.000105488 0.0155976 0.0471259 0.937782
## 25: 2561.1712: 0.000105550 0.0147333 0.0481634 0.938394
## 26: 2561.1567: 0.000105585 0.0146077 0.0475630 0.938782
## 27: 2561.1502: 0.000105615 0.0147452 0.0469817 0.939222
## 28: 2561.1472: 0.000105795 0.0142323 0.0460085 0.940641
## 29: 2561.1472: 0.000105969 0.0142799 0.0460050 0.940580
## 30: 2561.1472: 0.000106108 0.0142711 0.0460065 0.940593
## 31: 2561.1472: 0.000106367 0.0142681 0.0460077 0.940597
## 32: 2561.1472: 0.000108803 0.0142542 0.0460129 0.940613
## 33: 2561.1471: 0.000114120 0.0142385 0.0460189 0.940632

```

```

## 34:      2561.1470: 0.000129948 0.0142117 0.0460296 0.940664
## 35:      2561.1466: 0.000174644 0.0141665 0.0460488 0.940716
## 36:      2561.1457: 0.000289302 0.0140982 0.0460808 0.940793
## 37:      2561.1434: 0.000575009 0.0140010 0.0461342 0.940897
## 38:      2561.1366: 0.00104750 0.0140138 0.0461571 0.940861
## 39:      2561.1334: 0.00104750 0.0142079 0.0460331 0.940665
## 40:      2561.1332: 0.00104750 0.0142621 0.0459968 0.940612
## 41:      2561.1332: 0.00104750 0.0142655 0.0460024 0.940603
##
## Final Estimate of the Negative LLH:
## LLH: -6432.466      norm LLH: -3.460175
##      mu      omega      alpha1      beta1
## 8.300307e-06 8.957092e-07 4.600240e-02 9.406031e-01
##
## R-optimhess Difference Approximated Hessian Matrix:
##      mu      omega      alpha1      beta1
## mu      -3.589849e+07 -1.360249e+08 -4.894673e+02 5.080255e+01
## omega -1.360249e+08 -1.012940e+14 -4.264539e+09 -5.089170e+09
## alpha1 -4.894673e+02 -4.264539e+09 -2.292469e+05 -2.457651e+05
## beta1 5.080255e+01 -5.089170e+09 -2.457651e+05 -2.789727e+05
## attr("time")
## Time difference of 0.03125501 secs
##
## --- END OF TRACE ---
##
##
## Time to Estimate Parameters:
## Time difference of 0.211838 secs
##
## Title:
## GARCH Modelling
##
## Call:
## garchFit(formula = ~garch(1, 1), data = ftser)
##
## Mean and Variance Equation:
## data ~ garch(1, 1)
## <environment: 0x000000001f930818>
## [data = ftser]
##
## Conditional Distribution:
## norm
##
## Coefficient(s):
##      mu      omega      alpha1      beta1
## 8.3003e-06 8.9571e-07 4.6002e-02 9.4060e-01
##
## Std. Errors:
## based on Hessian
##
## Error Analysis:
##      Estimate Std. Error t value Pr(>|t|)
## mu      8.300e-06 1.669e-04 0.050 0.96034
## omega 8.957e-07 4.616e-07 1.941 0.05231 .

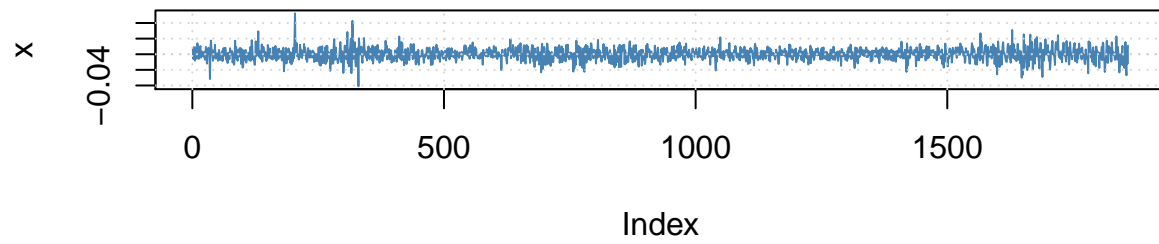
```

```
## alpha1 4.600e-02  1.189e-02   3.868  0.00011 ***
## beta1  9.406e-01  1.737e-02  54.135  < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Log Likelihood:
## 6432.466    normalized:  3.460175
##
## Description:
## Sat Apr 07 10:34:36 2018 by user: yydab
##
##
## Standardised Residuals Tests:
##
##               Statistic p-Value
## Jarque-Bera Test   R    Chi^2 175.0399 0
## Shapiro-Wilk Test  R     W    0.9901225 6.229921e-10
## Ljung-Box Test     R    Q(10) 7.392143 0.6879777
## Ljung-Box Test     R    Q(15) 16.34177 0.359708
## Ljung-Box Test     R    Q(20) 23.31753 0.2734735
## Ljung-Box Test     R^2  Q(10) 4.415861 0.9266432
## Ljung-Box Test     R^2  Q(15) 8.447815 0.904578
## Ljung-Box Test     R^2  Q(20) 11.53827 0.9310524
## LM Arch Test       R    TR^2  6.353908 0.8972016
##
## Information Criterion Statistics:
##      AIC      BIC      SIC      HQIC
## -6.916047 -6.904153 -6.916057 -6.911664
```

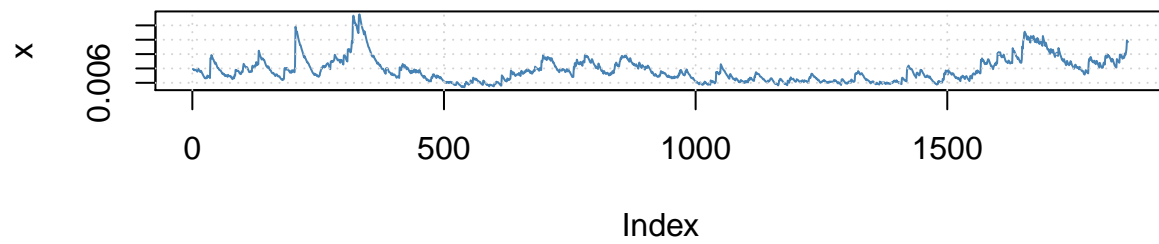
alpha1 and beta1 are significant. I also tried other different combinations but Garch(1,1) is the bes

```
par(mfrow=2:1)
plot(fit, which=1:2)    # plot data and root volatility
```

Time Series

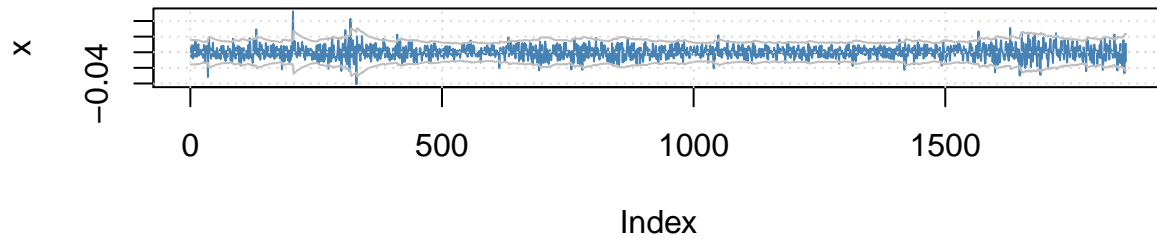


Conditional SD

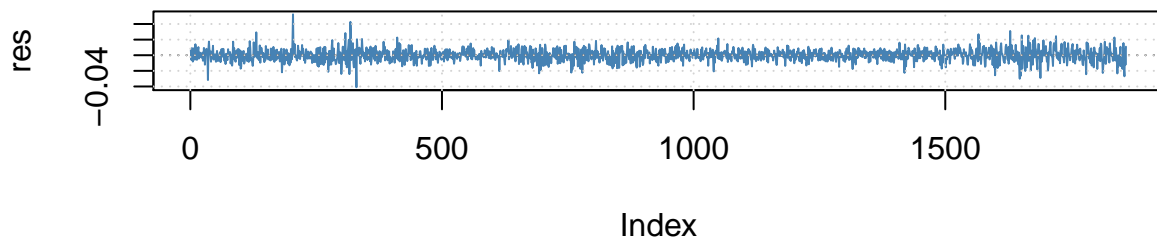


```
plot(fit, which=c(3,7)) # Series with 2 Conditonal SD; Standardized Residuals
```

Series with 2 Conditional SD Superimposed



Residuals



```
# ARMA with ARCH errors
ftsed1 <- diff(log(ftse))
fit2<- garchFit(~arma(1,0)+garch(1,1), data=ftsed1)
```

```
##
## Series Initialization:
## ARMA Model:          arma
## Formula Mean:        ~ arma(1, 0)
## GARCH Model:         garch
## Formula Variance:    ~ garch(1, 1)
## ARMA Order:          1 0
## Max ARMA Order:      1
## GARCH Order:         1 1
## Max GARCH Order:     1
## Maximum Order:       1
## Conditional Dist:    norm
## h.start:             2
## llh.start:           1
## Length of Series:    1859
## Recursion Init:      mci
## Series Scale:        0.007957728
##
## Parameter Initialization:
## Initial Parameters:   $params
## Limits of Transformations: $U, $V
## Which Parameters are Fixed? $includes
```



```

## Parameter Matrix:
##          U          V      params includes
##      mu      -0.54284978  0.5428498 0.05439580      TRUE
##      ar1      -0.99999999  1.0000000 0.09208644      TRUE
##      omega     0.00000100 100.0000000 0.10000000      TRUE
##      alpha1    0.00000001  1.0000000 0.10000000      TRUE
##      gamma1   -0.99999999  1.0000000 0.10000000      FALSE
##      beta1     0.00000001  1.0000000 0.80000000      TRUE
##      delta     0.00000000  2.0000000 2.00000000      FALSE
##      skew      0.10000000 10.0000000 1.00000000      FALSE
##      shape     1.00000000 10.0000000 4.00000000      FALSE
## Index List of Parameters to be Optimized:
##      mu      ar1      omega      alpha1      beta1
##        1        2        3        4        6
## Persistence:                0.9
##
##
## --- START OF TRACE ---
## Selected Algorithm: nlminb
##
## R coded nlminb Solver:
##
## 0:      2566.1950: 0.0543958 0.0920864 0.100000 0.100000 0.800000
## 1:      2566.0010: 0.0543899 0.0959064 0.0318122 0.125884 0.850195
## 2:      2564.3364: 0.0543899 0.0958215 0.0365554 0.126030 0.853242
## 3:      2563.4066: 0.0543900 0.0956233 0.0364293 0.120600 0.851734
## 4:      2561.7230: 0.0543904 0.0950928 0.0461914 0.114993 0.851255
## 5:      2560.3734: 0.0543910 0.0938305 0.0490588 0.0927912 0.853762
## 6:      2557.5021: 0.0543902 0.0934531 0.0439034 0.0875225 0.875082
## 7:      2555.4819: 0.0543896 0.0930316 0.0317217 0.0785271 0.891801
## 8:      2554.4767: 0.0543900 0.0919706 0.0231445 0.0703703 0.910979
## 9:      2554.1159: 0.0543978 0.0876211 0.0193378 0.0534546 0.924743
## 10:     2553.9889: 0.0544040 0.0853049 0.0232704 0.0517306 0.927731
## 11:     2553.0974: 0.0544101 0.0852014 0.0204800 0.0523369 0.927826
## 12:     2552.9963: 0.0544161 0.0865829 0.0185735 0.0527493 0.929394
## 13:     2552.9935: 0.0544161 0.0865826 0.0186155 0.0527925 0.929499
## 14:     2552.9916: 0.0544162 0.0865817 0.0185137 0.0527372 0.929533
## 15:     2552.9864: 0.0544172 0.0865811 0.0184903 0.0526643 0.929761
## 16:     2552.9035: 0.0544482 0.0865807 0.0171050 0.0495520 0.933884
## 17:     2552.9000: 0.0544482 0.0865796 0.0171353 0.0496185 0.933967
## 18:     2552.8977: 0.0544482 0.0865771 0.0170297 0.0496299 0.933996
## 19:     2552.8957: 0.0544490 0.0864216 0.0169992 0.0497016 0.934130
## 20:     2552.8925: 0.0544513 0.0860048 0.0168762 0.0496536 0.934164
## 21:     2552.8899: 0.0544595 0.0851692 0.0167956 0.0496669 0.934385
## 22:     2552.8810: 0.0545216 0.0864312 0.0164811 0.0495887 0.934702
## 23:     2552.8763: 0.0545668 0.0849468 0.0163000 0.0494720 0.935104
## 24:     2552.8758: 0.0545668 0.0849518 0.0162466 0.0494340 0.935096
## 25:     2552.8752: 0.0545683 0.0849917 0.0162586 0.0494441 0.935139
## 26:     2552.8745: 0.0545717 0.0850821 0.0162045 0.0494059 0.935170
## 27:     2552.8735: 0.0545811 0.0852748 0.0162106 0.0494125 0.935226
## 28:     2552.8722: 0.0546055 0.0855368 0.0161450 0.0493602 0.935281
## 29:     2552.8553: 0.0557378 0.0821543 0.0150985 0.0479871 0.937683
## 30:     2552.8383: 0.0568791 0.0847432 0.0142464 0.0469280 0.939855
## 31:     2552.8306: 0.0569453 0.0853137 0.0141008 0.0458495 0.940710

```

```

## 32:      2552.8295: 0.0568405 0.0853181 0.0141379 0.0458615 0.940783
## 33:      2552.8292: 0.0567357 0.0853385 0.0141025 0.0458499 0.940803
## 34:      2552.8290: 0.0565261 0.0853918 0.0140620 0.0458481 0.940857
## 35:      2552.8289: 0.0563637 0.0856447 0.0140951 0.0458991 0.940763
## 36:      2552.8289: 0.0563906 0.0856440 0.0140908 0.0459091 0.940761
## 37:      2552.8289: 0.0563930 0.0855965 0.0140854 0.0458900 0.940788
## 38:      2552.8289: 0.0563935 0.0856158 0.0140874 0.0458984 0.940776
## 39:      2552.8289: 0.0563934 0.0856160 0.0140874 0.0458984 0.940776
##
## Final Estimate of the Negative LLH:
## LLH: -6432.855      norm LLH: -3.460385
##      mu      ar1      omega      alpha1      beta1
## 4.487632e-04 8.561600e-02 8.920910e-07 4.589840e-02 9.407760e-01
##
## R-optimhess Difference Approximated Hessian Matrix:
##      mu      ar1      omega      alpha1
## mu      -35886704.462 -15873.9144 4.286244e+07 5.265352e+03
## ar1      -15873.914 -1744.9895 -2.793655e+06 -2.026047e+02
## omega 42862440.966 -2793654.9294 -1.019452e+14 -4.287758e+09
## alpha1 5265.352 -202.6047 -4.287758e+09 -2.304051e+05
## beta1 8187.296 -306.7414 -5.118538e+09 -2.470751e+05
##      beta1
## mu      8.187296e+03
## ar1      -3.067414e+02
## omega -5.118538e+09
## alpha1 -2.470751e+05
## beta1 -2.805122e+05
## attr("time")
## Time difference of 0.04688382 secs
##
## --- END OF TRACE ---
##
## Time to Estimate Parameters:
## Time difference of 0.346226 secs

```

```
summary(fit2)
```

```

##
## Title:
## GARCH Modelling
##
## Call:
## garchFit(formula = ~arma(1, 0) + garch(1, 1), data = ftsed1)
##
## Mean and Variance Equation:
## data ~ arma(1, 0) + garch(1, 1)
## <environment: 0x000000001a213a90>
## [data = ftsed1]
##
## Conditional Distribution:
## norm
##
## Coefficient(s):
##      mu      ar1      omega      alpha1      beta1

```

```
## 4.4876e-04 8.5616e-02 8.9209e-07 4.5898e-02 9.4078e-01
##
## Std. Errors:
## based on Hessian
##
## Error Analysis:
##      Estimate Std. Error t value Pr(>|t|)
## mu      4.488e-04 1.673e-04 2.683 0.007304 **
## ar1     8.562e-02 2.402e-02 3.565 0.000364 ***
## omega   8.921e-07 4.599e-07 1.940 0.052421 .
## alpha1  4.590e-02 1.189e-02 3.860 0.000113 ***
## beta1   9.408e-01 1.736e-02 54.207 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Log Likelihood:
## 6432.855      normalized: 3.460385
##
## Description:
## Sat Apr 07 10:34:37 2018 by user: yydab
##
##
## Standardised Residuals Tests:
##
##      Statistic p-Value
## Jarque-Bera Test  R    Chi^2 178.8086 0
## Shapiro-Wilk Test  R    W      0.9900033 5.151592e-10
## Ljung-Box Test     R    Q(10) 7.513918 0.6761988
## Ljung-Box Test     R    Q(15) 16.53416 0.3474659
## Ljung-Box Test     R    Q(20) 23.55409 0.2624131
## Ljung-Box Test     R^2 Q(10) 4.468907 0.9237251
## Ljung-Box Test     R^2 Q(15) 8.535124 0.9005445
## Ljung-Box Test     R^2 Q(20) 11.61467 0.9287007
## LM Arch Test       R    TR^2 6.422273 0.8933186
##
## Information Criterion Statistics:
##      AIC      BIC      SIC      HQIC
## -6.915390 -6.900523 -6.915405 -6.909911
```

Skew Normal fit

Note the null hypothesis of normality was rejected. Try a skew-normal. Skew parameter is significant under the model.

```
fit3<- garchFit(~arma(1,0)+garch(1,1), cond.dist="snorm", data=ftsed1)
```

```
##
## Series Initialization:
## ARMA Model:          arma
## Formula Mean:        ~ arma(1, 0)
## GARCH Model:         garch
## Formula Variance:    ~ garch(1, 1)
## ARMA Order:          1 0
## Max ARMA Order:      1
## GARCH Order:         1 1
```

```

## Max GARCH Order:          1
## Maximum Order:            1
## Conditional Dist:         snorm
## h.start:                  2
## llh.start:                1
## Length of Series:         1859
## Recursion Init:           mci
## Series Scale:              0.007957728
##
## Parameter Initialization:
## Initial Parameters:        $params
## Limits of Transformations: $U, $V
## Which Parameters are Fixed? $includes
## Parameter Matrix:
##           U           V      params includes
## mu      -0.54284978    0.5428498 0.05439580      TRUE
## ar1      -0.99999999    1.0000000 0.09208644      TRUE
## omega    0.00000100 100.0000000 0.10000000      TRUE
## alpha1   0.00000001    1.0000000 0.10000000      TRUE
## gamma1  -0.99999999    1.0000000 0.10000000     FALSE
## beta1    0.00000001    1.0000000 0.80000000      TRUE
## delta    0.00000000    2.0000000 2.00000000     FALSE
## skew     0.10000000   10.0000000 1.00000000      TRUE
## shape    1.00000000   10.0000000 4.00000000     FALSE
## Index List of Parameters to be Optimized:
## mu      ar1  omega alpha1  beta1  skew
## 1        2    3      4      6     8
## Persistence:              0.9
##
##
## --- START OF TRACE ---
## Selected Algorithm: nlminb
##
## R coded nlminb Solver:
##
## 0:      2566.1950: 0.0543958 0.0920864 0.100000 0.100000 0.800000 1.00000
## 1:      2565.7404: 0.0543899 0.0958992 0.0319416 0.125835 0.850100 0.989762
## 2:      2564.1258: 0.0543899 0.0958074 0.0366237 0.125963 0.853119 0.989630
## 3:      2563.2169: 0.0543899 0.0955971 0.0364674 0.120599 0.851638 0.989414
## 4:      2561.5649: 0.0543902 0.0950504 0.0460765 0.114995 0.851265 0.989800
## 5:      2560.1512: 0.0543901 0.0937224 0.0485003 0.0930829 0.854306 0.989218
## 6:      2557.3570: 0.0543888 0.0932693 0.0433019 0.0879689 0.875222 0.986669
## 7:      2555.4134: 0.0543874 0.0927596 0.0312823 0.0792111 0.891660 0.984285
## 8:      2554.4471: 0.0543868 0.0915242 0.0229554 0.0711154 0.910634 0.983317
## 9:      2553.8989: 0.0543903 0.0870521 0.0196704 0.0547145 0.923465 0.989053
## 10:     2553.8187: 0.0543936 0.0849822 0.0234363 0.0529829 0.926290 0.990449
## 11:     2553.0495: 0.0543975 0.0850825 0.0209323 0.0530952 0.926559 0.989233
## 12:     2552.9468: 0.0544011 0.0859585 0.0189423 0.0535273 0.928098 0.988483
## 13:     2552.9370: 0.0544012 0.0859587 0.0190349 0.0536110 0.928237 0.988481
## 14:     2552.9295: 0.0544013 0.0859596 0.0186169 0.0534743 0.928583 0.988455
## 15:     2552.9033: 0.0544050 0.0863034 0.0184835 0.0529983 0.929451 0.988090
## 16:     2552.8194: 0.0544331 0.0852157 0.0168305 0.0493247 0.934259 0.988112
## 17:     2552.8133: 0.0544332 0.0852175 0.0169451 0.0494799 0.934434 0.988112
## 18:     2552.8070: 0.0544350 0.0851515 0.0167918 0.0494264 0.934451 0.988301

```

```

## 19:      2552.8023: 0.0544392 0.0851156 0.0165937 0.0495597 0.934771 0.988625
## 20:      2552.7936: 0.0544592 0.0858372 0.0162378 0.0493846 0.935061 0.989059
## 21:      2552.7897: 0.0544722 0.0851912 0.0161750 0.0494481 0.935206 0.988295
## 22:      2552.7853: 0.0545191 0.0862955 0.0159006 0.0492291 0.935691 0.989718
## 23:      2552.7850: 0.0545192 0.0862879 0.0158973 0.0492219 0.935757 0.989706
## 24:      2552.7844: 0.0545207 0.0862492 0.0158594 0.0491954 0.935768 0.989697
## 25:      2552.7836: 0.0545238 0.0861724 0.0158404 0.0491802 0.935858 0.989674
## 26:      2552.7822: 0.0545326 0.0860039 0.0157587 0.0491377 0.935931 0.989602
## 27:      2552.7789: 0.0545704 0.0858637 0.0155922 0.0491011 0.936229 0.988794
## 28:      2552.7783: 0.0545957 0.0850777 0.0154825 0.0489775 0.936353 0.989421
## 29:      2552.7763: 0.0546524 0.0853367 0.0155060 0.0489444 0.936436 0.989097
## 30:      2552.7760: 0.0546685 0.0857294 0.0154983 0.0489012 0.936411 0.988701
## 31:      2552.7742: 0.0547309 0.0853284 0.0154658 0.0488091 0.936588 0.988978
## 32:      2552.7729: 0.0548601 0.0846993 0.0153020 0.0484937 0.936928 0.989305
## 33:      2552.7680: 0.0549863 0.0852679 0.0152988 0.0483860 0.937149 0.988517
## 34:      2552.7678: 0.0550758 0.0856528 0.0151257 0.0480377 0.937502 0.987956
## 35:      2552.7634: 0.0551752 0.0855248 0.0150756 0.0479896 0.937767 0.987817
## 36:      2552.7620: 0.0551878 0.0843222 0.0148750 0.0477338 0.938209 0.989105
## 37:      2552.7562: 0.0564549 0.0847928 0.0142522 0.0465926 0.939880 0.988665
## 38:      2552.7552: 0.0560997 0.0850601 0.0142849 0.0463994 0.940078 0.988969
## 39:      2552.7552: 0.0560997 0.0850642 0.0142424 0.0464500 0.940100 0.988961
## 40:      2552.7551: 0.0560959 0.0850638 0.0142345 0.0464447 0.940097 0.988957
## 41:      2552.7550: 0.0559766 0.0850424 0.0142120 0.0464184 0.940149 0.988807
## 42:      2552.7550: 0.0558905 0.0851870 0.0142027 0.0464280 0.940152 0.988672
## 43:      2552.7550: 0.0559027 0.0851619 0.0142067 0.0464318 0.940143 0.988637
## 44:      2552.7550: 0.0559037 0.0851506 0.0142059 0.0464290 0.940146 0.988650
##
## Final Estimate of the Negative LLH:
## LLH: -6432.929      norm LLH: -3.460425
##      mu      ar1      omega      alpha1      beta1
## 4.448661e-04 8.515062e-02 8.995943e-07 4.642901e-02 9.401463e-01
##      skew
## 9.886496e-01
##
## R-optimhess Difference Approximated Hessian Matrix:
##      mu      ar1      omega      alpha1
## mu      -35916127.30 -1.556910e+04 -6.694879e+08 -2.719471e+04
## ar1      -15569.10 -1.744539e+03 -2.667938e+06 -2.050571e+02
## omega    -669487907.87 -2.667938e+06 -9.998224e+13 -4.199581e+09
## alpha1    -27194.71 -2.050571e+02 -4.199581e+09 -2.255025e+05
## beta1     -28718.83 -3.037074e+02 -5.016185e+09 -2.419136e+05
## skew      12576.34 8.018133e+01 1.338307e+07 -3.630815e+01
##      beta1      skew
## mu      -2.871883e+04 1.257634e+04
## ar1     -3.037074e+02 8.018133e+01
## omega   -5.016185e+09 1.338307e+07
## alpha1  -2.419136e+05 -3.630815e+01
## beta1   -2.748331e+05 4.510115e+02
## skew    4.510115e+02 -1.183642e+03
## attr("time")
## Time difference of 0.08070016 secs
##
## --- END OF TRACE ---
##

```

```
##
## Time to Estimate Parameters:
## Time difference of 0.3591251 secs
summary(fit3)

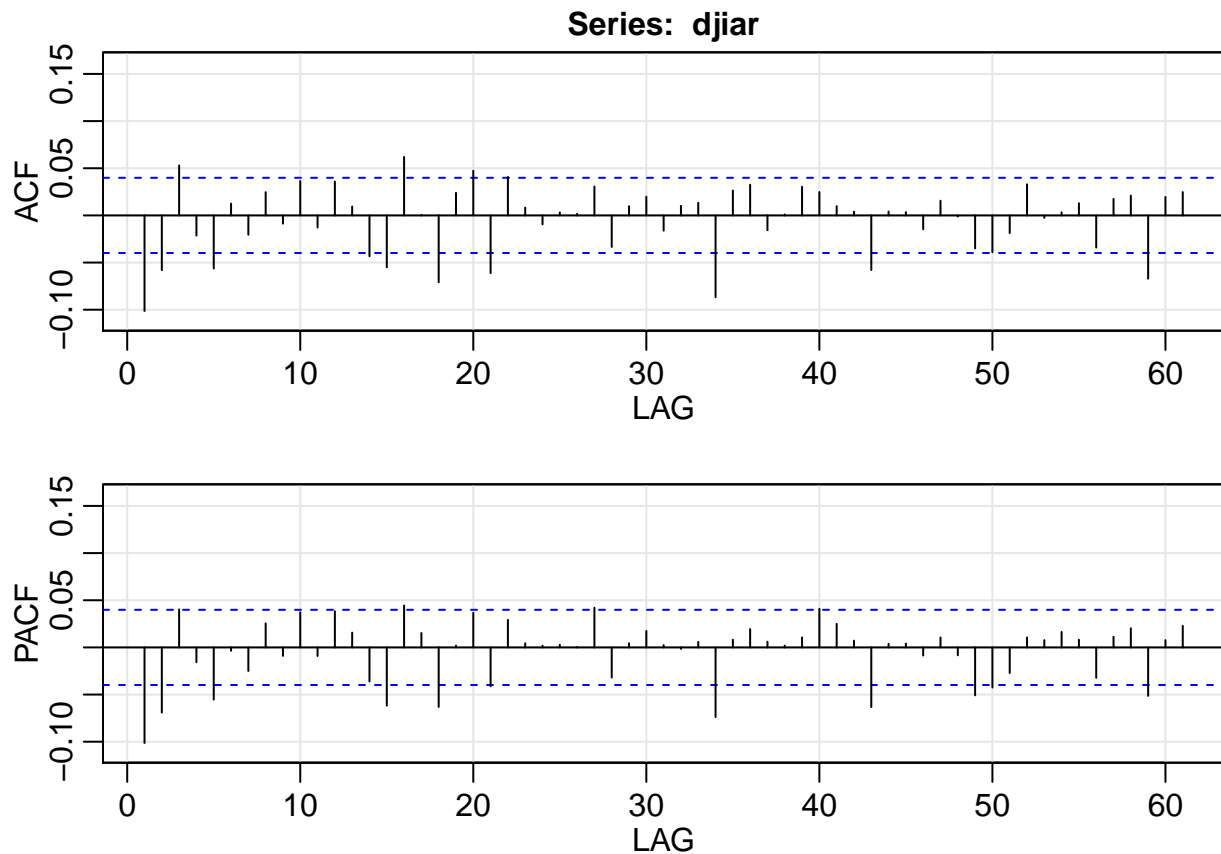
##
## Title:
## GARCH Modelling
##
## Call:
## garchFit(formula = ~arma(1, 0) + garch(1, 1), data = ftsed1,
## cond.dist = "snorm")
##
## Mean and Variance Equation:
## data ~ arma(1, 0) + garch(1, 1)
## <environment: 0x000000001d359448>
## [data = ftsed1]
##
## Conditional Distribution:
## snorm
##
## Coefficient(s):
##      mu      ar1      omega      alpha1      beta1      skew
## 4.4487e-04 8.5151e-02 8.9959e-07 4.6429e-02 9.4015e-01 9.8865e-01
##
## Std. Errors:
## based on Hessian
##
## Error Analysis:
##      Estimate Std. Error t value Pr(>|t|)
## mu      4.449e-04 1.675e-04   2.656 0.007911 **
## ar1      8.515e-02 2.405e-02   3.541 0.000398 ***
## omega    8.996e-07 4.608e-07   1.952 0.050929 .
## alpha1   4.643e-02 1.199e-02   3.872 0.000108 ***
## beta1    9.401e-01 1.743e-02  53.951 < 2e-16 ***
## skew     9.886e-01 2.937e-02  33.665 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Log Likelihood:
## 6432.929      normalized: 3.460425
##
## Description:
## Sat Apr 07 10:34:37 2018 by user: yydab
##
## Standardised Residuals Tests:
##
##      Jarque-Bera Test  R      Chi^2 178.3195 0
##      Shapiro-Wilk Test R      W      0.9900119 5.22203e-10
##      Ljung-Box Test   R      Q(10) 7.512191 0.6763662
##      Ljung-Box Test   R      Q(15) 16.51273 0.348817
##      Ljung-Box Test   R      Q(20) 23.53262 0.2634044
##      Ljung-Box Test   R^2 Q(10) 4.473355 0.9234776
```

```
## Ljung-Box Test      R^2  Q(15)  8.566279  0.8990826
## Ljung-Box Test      R^2  Q(20) 11.64354  0.9277988
## LM Arch Test        R    TR^2   6.419559  0.8934741
##
## Information Criterion Statistics:
##      AIC      BIC      SIC      HQIC
## -6.914394 -6.896553 -6.914415 -6.907819
```

Conclusion: the model $ARMA(1,0) + GARCH(1,1)$ with skew normal distribution fit the data.

Q3

```
djiar = diff(log(djia$Close))[-1]
acf2(djiar)      # exhibits some autocorrelation
```



```
##      ACF  PACF
## [1,] -0.10 -0.10
## [2,] -0.06 -0.07
## [3,]  0.05  0.04
## [4,] -0.02 -0.02
## [5,] -0.06 -0.06
## [6,]  0.01  0.00
## [7,] -0.02 -0.02
## [8,]  0.02  0.03
```

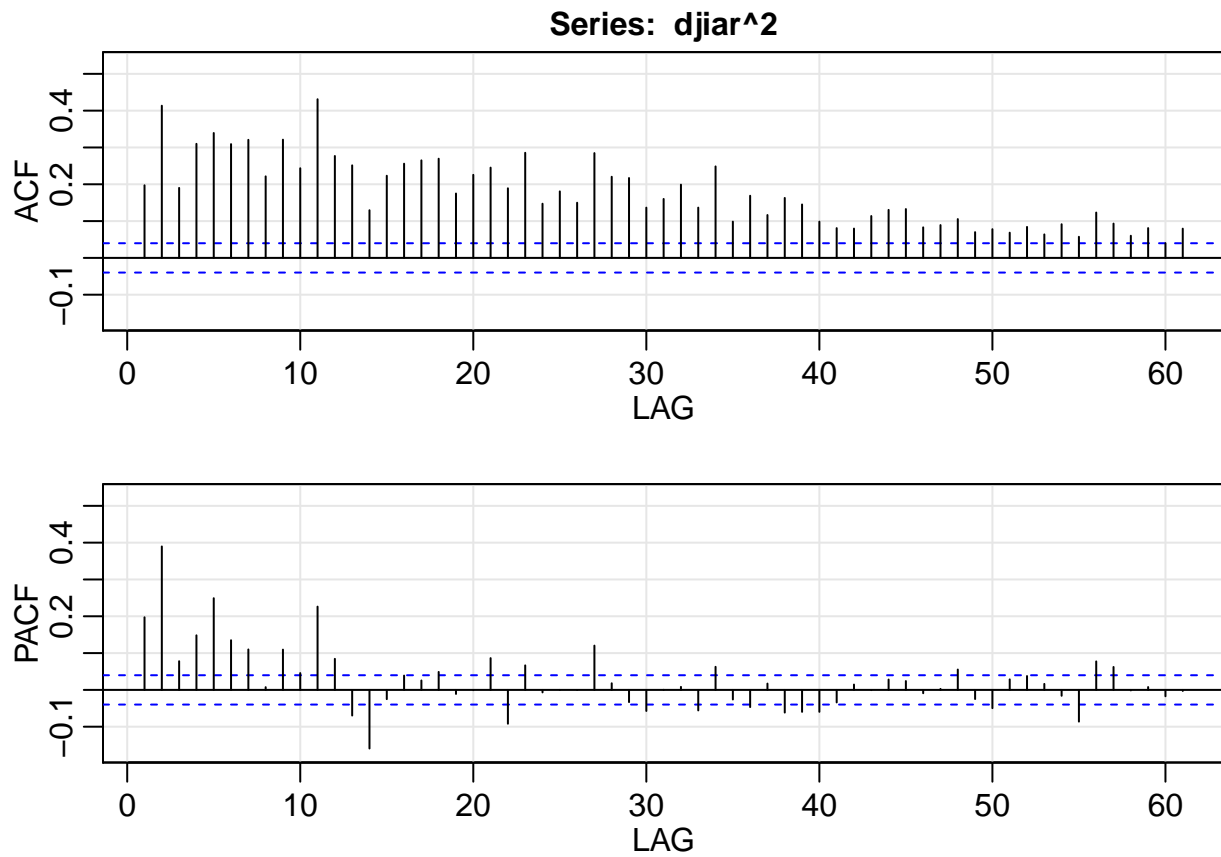
```

## [9,] -0.01 -0.01
## [10,] 0.04 0.04
## [11,] -0.01 -0.01
## [12,] 0.04 0.04
## [13,] 0.01 0.02
## [14,] -0.04 -0.04
## [15,] -0.06 -0.06
## [16,] 0.06 0.04
## [17,] 0.00 0.02
## [18,] -0.07 -0.06
## [19,] 0.02 0.00
## [20,] 0.05 0.04
## [21,] -0.06 -0.04
## [22,] 0.04 0.03
## [23,] 0.01 0.00
## [24,] -0.01 0.00
## [25,] 0.00 0.00
## [26,] 0.00 0.00
## [27,] 0.03 0.04
## [28,] -0.03 -0.03
## [29,] 0.01 0.00
## [30,] 0.02 0.02
## [31,] -0.02 0.00
## [32,] 0.01 0.00
## [33,] 0.01 0.01
## [34,] -0.09 -0.07
## [35,] 0.03 0.01
## [36,] 0.03 0.02
## [37,] -0.02 0.01
## [38,] 0.00 0.00
## [39,] 0.03 0.01
## [40,] 0.02 0.04
## [41,] 0.01 0.02
## [42,] 0.00 0.01
## [43,] -0.06 -0.06
## [44,] 0.00 0.00
## [45,] 0.00 0.00
## [46,] -0.01 -0.01
## [47,] 0.02 0.01
## [48,] 0.00 -0.01
## [49,] -0.04 -0.05
## [50,] -0.04 -0.04
## [51,] -0.02 -0.03
## [52,] 0.03 0.01
## [53,] 0.00 0.01
## [54,] 0.00 0.02
## [55,] 0.01 0.01
## [56,] -0.03 -0.03
## [57,] 0.02 0.01
## [58,] 0.02 0.02
## [59,] -0.07 -0.05
## [60,] 0.02 0.01
## [61,] 0.02 0.02

```



```
acf2(djia2) # oozes autocorrelation
```



##		ACF	PACF
##	[1,]	0.20	0.20
##	[2,]	0.41	0.39
##	[3,]	0.19	0.08
##	[4,]	0.31	0.15
##	[5,]	0.34	0.25
##	[6,]	0.31	0.13
##	[7,]	0.32	0.11
##	[8,]	0.22	0.01
##	[9,]	0.32	0.11
##	[10,]	0.24	0.05
##	[11,]	0.43	0.23
##	[12,]	0.28	0.08
##	[13,]	0.25	-0.07
##	[14,]	0.13	-0.16
##	[15,]	0.22	-0.03
##	[16,]	0.26	0.04
##	[17,]	0.27	0.03
##	[18,]	0.27	0.05
##	[19,]	0.17	-0.01
##	[20,]	0.23	0.00
##	[21,]	0.25	0.09
##	[22,]	0.19	-0.09
##	[23,]	0.29	0.07

```
## [24,] 0.15 -0.01
## [25,] 0.18 0.00
## [26,] 0.15 0.00
## [27,] 0.28 0.12
## [28,] 0.22 0.02
## [29,] 0.22 -0.03
## [30,] 0.14 -0.06
## [31,] 0.16 0.00
## [32,] 0.20 0.01
## [33,] 0.14 -0.06
## [34,] 0.25 0.06
## [35,] 0.10 -0.03
## [36,] 0.17 -0.05
## [37,] 0.12 0.02
## [38,] 0.16 -0.06
## [39,] 0.15 -0.06
## [40,] 0.10 -0.06
## [41,] 0.08 -0.03
## [42,] 0.08 0.01
## [43,] 0.11 0.00
## [44,] 0.13 0.03
## [45,] 0.13 0.02
## [46,] 0.08 -0.01
## [47,] 0.09 0.00
## [48,] 0.11 0.06
## [49,] 0.07 -0.02
## [50,] 0.08 -0.05
## [51,] 0.07 0.03
## [52,] 0.08 0.04
## [53,] 0.06 0.02
## [54,] 0.09 -0.02
## [55,] 0.06 -0.09
## [56,] 0.12 0.08
## [57,] 0.09 0.06
## [58,] 0.06 0.00
## [59,] 0.08 0.01
## [60,] 0.04 -0.02
## [61,] 0.08 0.00
```

```
# GARCH fit
summary(fit_31<- garchFit(~arma(1,0)+garch(1,1), data=djia, cond.dist='std'))
```

```
##
## Series Initialization:
## ARMA Model: arma
## Formula Mean: ~ arma(1, 0)
## GARCH Model: garch
## Formula Variance: ~ garch(1, 1)
## ARMA Order: 1 0
## Max ARMA Order: 1
## GARCH Order: 1 1
## Max GARCH Order: 1
## Maximum Order: 1
## Conditional Dist: std
## h.start: 2
```

```

## llh.start:          1
## Length of Series:   2517
## Recursion Init:     mci
## Series Scale:       0.01210097
##
## Parameter Initialization:
## Initial Parameters:  $params
## Limits of Transformations:  $U, $V
## Which Parameters are Fixed? $includes
## Parameter Matrix:
##           U           V      params includes
## mu      -0.15336279   0.1533628  0.01533395    TRUE
## ar1     -0.99999999   1.0000000 -0.10129752    TRUE
## omega    0.00000100 100.0000000  0.10000000    TRUE
## alpha1   0.00000001  1.0000000  0.10000000    TRUE
## gamma1  -0.99999999   1.0000000  0.10000000    FALSE
## beta1    0.00000001  1.0000000  0.80000000    TRUE
## delta    0.00000000  2.0000000  2.00000000    FALSE
## skew     0.10000000 10.0000000  1.00000000    FALSE
## shape    1.00000000 10.0000000  4.00000000    TRUE
## Index List of Parameters to be Optimized:
## mu  ar1  omega  alpha1  beta1  shape
##  1    2    3      4      6      9
## Persistence:          0.9
##
##
## --- START OF TRACE ---
## Selected Algorithm: nlminb
##
## R coded nlminb Solver:
##
## 0:      2966.5649: 0.0153339 -0.101298 0.100000 0.100000 0.800000 4.00000
## 1:      2944.7773: 0.0153351 -0.0991717 0.0803851 0.105688 0.794226 3.99986
## 2:      2910.9444: 0.0153425 -0.0864391 0.0198839 0.162519 0.809277 4.00053
## 3:      2891.9987: 0.0153434 -0.0855698 0.0350001 0.168391 0.817706 4.00077
## 4:      2882.6364: 0.0153798 -0.0489486 0.0211395 0.164589 0.840541 4.00286
## 5:      2881.9304: 0.0153823 -0.0484851 0.0204244 0.166791 0.845392 4.00314
## 6:      2881.5678: 0.0153882 -0.0491681 0.0164060 0.164252 0.847791 4.00365
## 7:      2880.9562: 0.0153959 -0.0493829 0.0174447 0.161530 0.852258 4.00434
## 8:      2880.3922: 0.0154144 -0.0492493 0.0148742 0.153976 0.859252 4.00602
## 9:      2880.0788: 0.0154929 -0.0464297 0.0155698 0.139117 0.871707 4.01357
## 10:     2879.5719: 0.0156667 -0.0502107 0.0125331 0.137634 0.873099 4.03121
## 11:     2879.2754: 0.0158465 -0.0534691 0.0130134 0.140284 0.873078 4.04888
## 12:     2878.9979: 0.0160179 -0.0464745 0.0137975 0.139197 0.871470 4.06589
## 13:     2878.9707: 0.0160186 -0.0464112 0.0132405 0.139289 0.871633 4.06595
## 14:     2878.9510: 0.0160193 -0.0463401 0.0134262 0.139559 0.872118 4.06603
## 15:     2878.9240: 0.0160288 -0.0462943 0.0129595 0.139571 0.872158 4.06693
## 16:     2878.8886: 0.0160494 -0.0462517 0.0131197 0.139735 0.872456 4.06886
## 17:     2872.3653: 0.0239833 -0.0535089 0.0143767 0.137582 0.858370 4.81181
## 18:     2871.5453: 0.0252884 -0.0550775 0.0137416 0.122667 0.870984 4.83396
## 19:     2871.3687: 0.0252889 -0.0548939 0.0140394 0.123736 0.872173 4.83400
## 20:     2871.2367: 0.0253052 -0.0547681 0.0128189 0.123951 0.872237 4.83405
## 21:     2871.1340: 0.0253530 -0.0546310 0.0132252 0.124568 0.872883 4.83413
## 22:     2871.0366: 0.0254525 -0.0544735 0.0123989 0.124866 0.872942 4.83425

```

```

## 23:      2870.9225: 0.0256529 -0.0543024 0.0127160 0.125465 0.873271 4.83439
## 24:      2865.2785: 0.0596248 -0.0456160 0.0160045 0.157797 0.842171 4.84608
## 25:      2862.9397: 0.0677622 -0.0434674 0.0102271 0.136586 0.867384 5.04667
## 26:      2862.1051: 0.0743737 -0.0636105 0.0112956 0.123954 0.872267 5.35478
## 27:      2861.7101: 0.0719553 -0.0554743 0.0110194 0.124369 0.870600 5.66110
## 28:      2861.6164: 0.0709488 -0.0545210 0.0111056 0.125771 0.869190 5.85080
## 29:      2861.6013: 0.0708292 -0.0551561 0.0109174 0.124128 0.870274 5.97092
## 30:      2861.6001: 0.0709556 -0.0553034 0.0110181 0.124507 0.869952 5.97943
## 31:      2861.6000: 0.0709486 -0.0553165 0.0109951 0.124451 0.870011 5.97873
## 32:      2861.6000: 0.0709475 -0.0553146 0.0109959 0.124444 0.870013 5.97878
##
## Final Estimate of the Negative LLH:
## LLH: -8249.619      norm LLH: -3.27756
##      mu      ar1      omega      alpha1      beta1
## 8.585338e-04 -5.531459e-02 1.610165e-06 1.244440e-01 8.700129e-01
##      shape
## 5.978777e+00
##
## R-optimhess Difference Approximated Hessian Matrix:
##      mu      ar1      omega      alpha1
## mu      -4.791648e+07 -4.661855e+04 -1.205649e+09 -3.466882e+04
## ar1      -4.661855e+04 -2.490883e+03 -1.234056e+06 -8.201170e+00
## omega     -1.205649e+09 -1.234056e+06 -1.703959e+13 -5.515908e+08
## alpha1    -3.466882e+04 -8.201170e+00 -5.515908e+08 -3.611267e+04
## beta1     -9.149142e+04 -9.743995e+01 -8.452232e+08 -4.469044e+04
## shape     -9.970359e+02 8.560505e-02 -3.050355e+06 -1.832751e+02
##      beta1      shape
## mu      -9.149142e+04 -9.970359e+02
## ar1      -9.743995e+01 8.560505e-02
## omega     -8.452232e+08 -3.050355e+06
## alpha1    -4.469044e+04 -1.832751e+02
## beta1     -6.270066e+04 -2.340750e+02
## shape     -2.340750e+02 -2.547444e+00
## attr("time")
## Time difference of 0.1431739 secs
##
## --- END OF TRACE ---
##
##
## Time to Estimate Parameters:
## Time difference of 0.4600739 secs
##
## Title:
## GARCH Modelling
##
## Call:
## garchFit(formula = ~arma(1, 0) + garch(1, 1), data = djiar, cond.dist = "std")
##
## Mean and Variance Equation:
## data ~ arma(1, 0) + garch(1, 1)
## <environment: 0x000000001eea62a8>
## [data = djiar]
##
## Conditional Distribution:

```

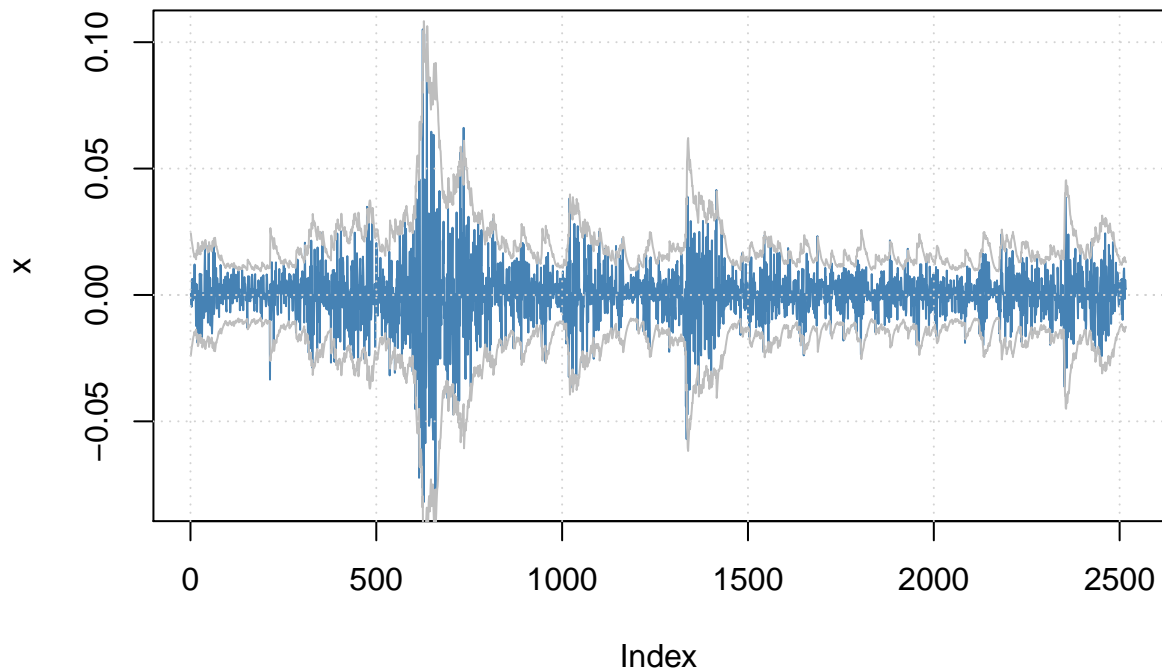
```

## std
##
## Coefficient(s):
##      mu      ar1      omega      alpha1      beta1
## 8.5853e-04 -5.5315e-02 1.6102e-06 1.2444e-01 8.7001e-01
##      shape
## 5.9788e+00
##
## Std. Errors:
## based on Hessian
##
## Error Analysis:
##      Estimate Std. Error t value Pr(>|t|)
## mu      8.585e-04 1.470e-04 5.842 5.16e-09 ***
## ar1     -5.531e-02 2.023e-02 -2.735 0.006239 **
## omega   1.610e-06 4.459e-07 3.611 0.000305 ***
## alpha1  1.244e-01 1.660e-02 7.497 6.55e-14 ***
## beta1   8.700e-01 1.526e-02 57.022 < 2e-16 ***
## shape   5.979e+00 7.917e-01 7.552 4.31e-14 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Log Likelihood:
## 8249.619      normalized: 3.27756
##
## Description:
## Sat Apr 07 10:34:38 2018 by user: yydab
##
##
## Standardised Residuals Tests:
##
##      Statistic p-Value
## Jarque-Bera Test R Chi^2 310.0055 0
## Shapiro-Wilk Test R W 0.9820294 0
## Ljung-Box Test R Q(10) 16.82241 0.07838716
## Ljung-Box Test R Q(15) 26.44807 0.03356859
## Ljung-Box Test R Q(20) 28.71095 0.09360887
## Ljung-Box Test R^2 Q(10) 15.36776 0.119218
## Ljung-Box Test R^2 Q(15) 19.13661 0.2076099
## Ljung-Box Test R^2 Q(20) 22.9289 0.2922993
## LM Arch Test R TR^2 15.0399 0.2392619
##
## Information Criterion Statistics:
##      AIC      BIC      SIC      HQIC
## -6.550353 -6.536453 -6.550364 -6.545309

```

```
plot(fit_31,which=3)
```

Series with 2 Conditional SD Superimposed



APARCH fit

Suppose y_t is APARCH noise with conditional variance.

```
summary(fit_32<- garchFit(~arma(1,0)+aparch(1,1), data=djia, cond.dist='std'))
```

```
##
## Series Initialization:
## ARMA Model:          arma
## Formula Mean:        ~ arma(1, 0)
## GARCH Model:         aparch
## Formula Variance:    ~ aparch(1, 1)
## ARMA Order:          1 0
## Max ARMA Order:      1
## GARCH Order:         1 1
## Max GARCH Order:     1
## Maximum Order:       1
## Conditional Dist:    std
## h.start:             2
## llh.start:           1
## Length of Series:    2517
## Recursion Init:      mci
## Series Scale:        0.01210097
##
## Parameter Initialization:
```

```

## Initial Parameters:          $params
## Limits of Transformations:  $U, $V
## Which Parameters are Fixed? $includes
## Parameter Matrix:
##           U           V      params includes
## mu      -0.15336279   0.1533628  0.01533395    TRUE
## ar1      -0.99999999   1.0000000 -0.10129752    TRUE
## omega    0.00000100  100.0000000  0.10000000    TRUE
## alpha1   0.00000001   1.0000000  0.10000000    TRUE
## gamma1  -0.99999999   1.0000000  0.10000000    TRUE
## beta1    0.00000001   1.0000000  0.80000000    TRUE
## delta    0.00000000   2.0000000  2.00000000    TRUE
## skew     0.10000000  10.0000000  1.00000000    FALSE
## shape    1.00000000  10.0000000  4.00000000    TRUE
## Index List of Parameters to be Optimized:
## mu      ar1  omega alpha1 gamma1  beta1  delta  shape
## 1        2      3      4      5      6      7      9
## Persistence:          0.901
##
##
## --- START OF TRACE ---
## Selected Algorithm: nlminb
##
## R coded nlminb Solver:
##
## 0:      2956.2216: 0.0153339 -0.101298 0.100000 0.100000 0.100000 0.800000 2.00000 4.00000
## 1:      2933.7102: 0.0153349 -0.0994951 0.0816702 0.105086 0.101441 0.794372 1.99989 3.99987
## 2:      2915.7665: 0.0153364 -0.0969033 0.0659019 0.116673 0.103734 0.795785 2.00000 3.99997
## 3:      2868.4754: 0.0153445 -0.0848309 0.0240064 0.174770 0.115713 0.826263 2.00000 4.00127
## 4:      2866.0092: 0.0153484 -0.0806185 0.0223796 0.179099 0.121370 0.835422 2.00000 4.00179
## 5:      2865.7443: 0.0153525 -0.0768099 0.0131353 0.176107 0.127199 0.838908 1.99992 4.00228
## 6:      2863.1450: 0.0153538 -0.0756692 0.0180404 0.176968 0.129004 0.841876 2.00000 4.00245
## 7:      2856.2636: 0.0153873 -0.0477472 0.00967343 0.170088 0.175541 0.860597 1.99995 4.00606
## 8:      2853.4395: 0.0154246 -0.0297309 0.0251948 0.165528 0.224263 0.839295 1.99939 4.00933
## 9:      2853.3035: 0.0154254 -0.0301564 0.0152622 0.163962 0.225414 0.836633 1.99923 4.00939
## 10:     2850.4729: 0.0154259 -0.0302849 0.0196607 0.165302 0.225960 0.839073 1.99934 4.00948
## 11:     2849.5507: 0.0154305 -0.0320034 0.0140574 0.167395 0.231771 0.845156 1.99944 4.01012
## 12:     2847.7408: 0.0154470 -0.0377746 0.0183930 0.169017 0.251279 0.844831 1.99918 4.01197
## 13:     2845.4604: 0.0154649 -0.0404007 0.0166508 0.170044 0.271510 0.841014 1.99879 4.01390
## 14:     2840.6097: 0.0155659 -0.0398960 0.0220312 0.179337 0.362238 0.827889 1.99564 4.02504
## 15:     2840.2159: 0.0155666 -0.0399220 0.0175565 0.177807 0.362739 0.826608 1.99550 4.02510
## 16:     2839.6368: 0.0155818 -0.0427577 0.0195156 0.177280 0.363341 0.828508 1.99333 4.02667
## 17:     2839.2986: 0.0156224 -0.0472413 0.0180222 0.174695 0.365336 0.830519 1.98742 4.03092
## 18:     2838.3755: 0.0158110 -0.0320240 0.0189810 0.174719 0.374079 0.830619 1.96003 4.05076
## 19:     2838.1806: 0.0158142 -0.0328761 0.0200414 0.172691 0.376493 0.833566 1.95990 4.05118
## 20:     2837.8264: 0.0158349 -0.0334367 0.0184927 0.171928 0.377234 0.833250 1.95678 4.05345
## 21:     2837.5310: 0.0158776 -0.0347303 0.0194491 0.170603 0.379522 0.834980 1.95050 4.05826
## 22:     2837.0525: 0.0159673 -0.0366418 0.0184466 0.169757 0.381820 0.835073 1.93697 4.06843
## 23:     2831.2211: 0.0181618 -0.0743039 0.0230404 0.163104 0.433944 0.843617 1.61079 4.32024
## 24:     2817.9701: 0.0238342 -0.0395263 0.0222446 0.116086 0.625292 0.887192 0.952834 4.96894
## 25:     2816.4431: 0.0238343 -0.0395438 0.0234501 0.116900 0.625343 0.888222 0.952877 4.96895
## 26:     2816.0968: 0.0238346 -0.0397073 0.0221922 0.118042 0.625724 0.888557 0.952974 4.96899
## 27:     2815.9413: 0.0238547 -0.0398291 0.0225848 0.118850 0.626312 0.888738 0.955815 4.97031
## 28:     2815.7521: 0.0238673 -0.0399496 0.0221607 0.119049 0.626607 0.888251 0.959122 4.97106

```

```

## 29:      2815.5147: 0.0238935 -0.0402344 0.0221744 0.120186 0.627325 0.888144 0.965762 4.97248
## 30:      2812.9395: 0.0247901 -0.0462755 0.0254998 0.131035 0.642407 0.870254 1.18463 5.01118
## 31:      2810.0292: 0.0275158 -0.0436698 0.0253200 0.137236 0.714729 0.865646 1.08442 5.09245
## 32:      2806.6352: 0.0307215 -0.0488468 0.0202554 0.123557 0.752027 0.880103 1.17165 5.11226
## 33:      2801.0388: 0.0434902 -0.0442804 0.0227921 0.122527 0.941344 0.881833 1.00718 5.42913
## 34:      2799.4186: 0.0454608 -0.0467090 0.0206287 0.110682 0.974697 0.890429 1.03440 5.52098
## 35:      2799.3567: 0.0454608 -0.0466941 0.0213348 0.110090 0.974750 0.890322 1.03440 5.52100
## 36:      2799.3092: 0.0454567 -0.0466982 0.0211179 0.109861 0.974937 0.890133 1.03430 5.52178
## 37:      2799.2864: 0.0454480 -0.0467008 0.0213211 0.109735 0.975290 0.890215 1.03417 5.52349
## 38:      2799.2562: 0.0454309 -0.0467111 0.0212701 0.109485 0.975963 0.890174 1.03390 5.52695
## 39:      2797.2264: 0.0440041 -0.0503513 0.0197231 0.0974177 1.00000 0.896930 1.04374 6.72117
## 40:      2796.9414: 0.0436647 -0.0468928 0.0199020 0.0999479 1.00000 0.893330 1.08177 6.93829
## 41:      2796.8934: 0.0439925 -0.0477799 0.0210184 0.0986150 1.00000 0.893827 1.06058 7.10174
## 42:      2796.8503: 0.0433479 -0.0484091 0.0202761 0.0980377 1.00000 0.894290 1.07211 7.22780
## 43:      2796.8478: 0.0433159 -0.0482204 0.0202479 0.0980887 1.00000 0.894301 1.07263 7.26082
## 44:      2796.8472: 0.0432654 -0.0481866 0.0202579 0.0980839 1.00000 0.894407 1.07101 7.28539
## 45:      2796.8472: 0.0432532 -0.0481827 0.0202595 0.0980866 1.00000 0.894447 1.07040 7.28659
## 46:      2796.8472: 0.0432526 -0.0481838 0.0202596 0.0980896 1.00000 0.894456 1.07023 7.28591
## 47:      2796.8472: 0.0432526 -0.0481839 0.0202594 0.0980893 1.00000 0.894457 1.07023 7.28576
##
## Final Estimate of the Negative LLH:
## LLH: -8311.583      norm LLH: -3.302178
##      mu      ar1      omega      alpha1      gamma1
## 0.0005233986 -0.0481839398 0.0001798026 0.0980892614 0.9999999900
##      beta1      delta      shape
## 0.8944566891 1.0702336184 7.2857627317
##
## R-optimhess Difference Approximated Hessian Matrix:
##      mu      ar1      omega      alpha1
## mu      -9.257268e+07 -1.304497e+05 -6.612990e+08 -2.435875e+06
## ar1      -1.304497e+05 -2.960736e+03 -7.708000e+05 -1.492695e+03
## omega     -6.612990e+08 -7.708000e+05 -9.933032e+09 -3.284959e+07
## alpha1    -2.435875e+06 -1.492695e+03 -3.284959e+07 -1.662177e+05
## gamma1    -1.277035e+05 -7.809040e+01 -1.721819e+06 -8.715942e+03
## beta1     -3.642711e+06 -3.791572e+03 -5.191596e+07 -2.185971e+05
## delta     -2.481012e+05 -2.643131e+02 -3.568315e+06 -1.497209e+04
## shape     -4.448811e+03 -2.850991e+00 -4.321053e+04 -2.197851e+02
##      gamma1      beta1      delta      shape
## mu      -1.277035e+05 -3.642711e+06 -248101.2220 -4448.811095
## ar1      -7.809040e+01 -3.791572e+03 -264.3131 -2.850991
## omega     -1.721819e+06 -5.191596e+07 -3568314.8658 -43210.531832
## alpha1    -8.715942e+03 -2.185971e+05 -14972.0922 -219.785097
## gamma1    -9.621456e+03 -1.146450e+04 -783.1003 -11.538915
## beta1     -1.146450e+04 -3.201325e+05 -21535.8361 -285.458078
## delta     -7.831003e+02 -2.153584e+04 -1520.6731 -19.344100
## shape     -1.153892e+01 -2.854581e+02 -19.3441 -1.118540
## attr("time")
## Time difference of 0.5037329 secs
##
## --- END OF TRACE ---
##
##
## Time to Estimate Parameters:
## Time difference of 1.38564 secs

```



```

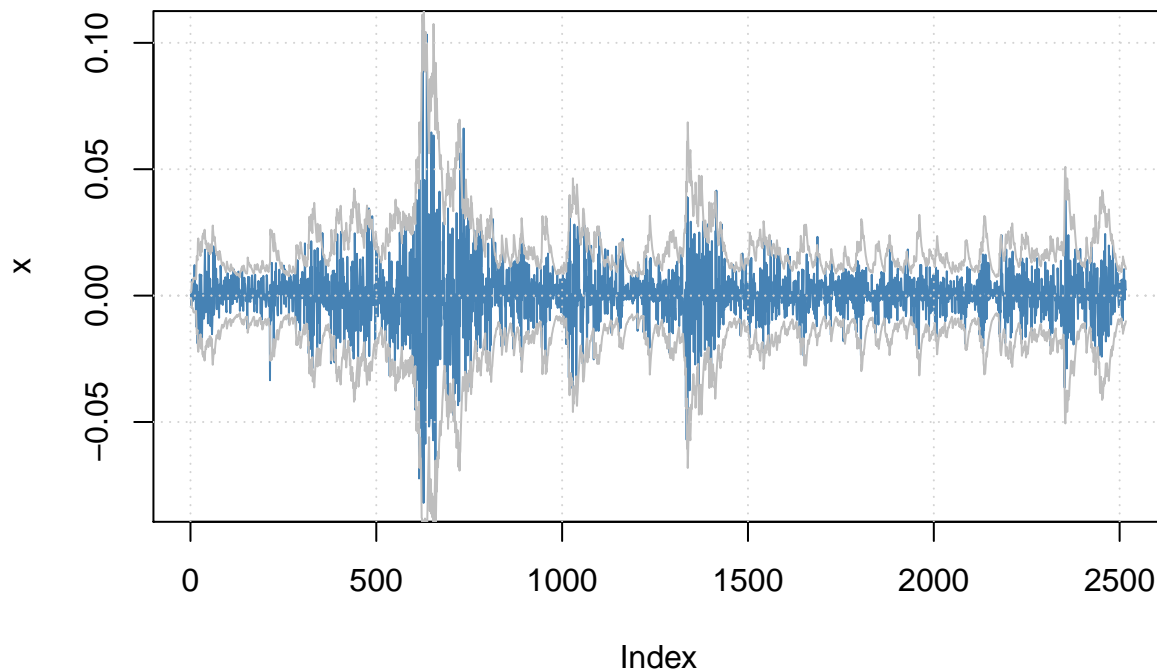
##
## Title:
## GARCH Modelling
##
## Call:
## garchFit(formula = ~arma(1, 0) + aparch(1, 1), data = djiar,
## cond.dist = "std")
##
## Mean and Variance Equation:
## data ~ arma(1, 0) + aparch(1, 1)
## <environment: 0x000000001c77b008>
## [data = djiar]
##
## Conditional Distribution:
## std
##
## Coefficient(s):
##      mu      ar1      omega      alpha1      gamma1      beta1
## 0.0005234 -0.0481839 0.0001798 0.0980893 1.0000000 0.8944567
##      delta      shape
## 1.0702336 7.2857627
##
## Std. Errors:
## based on Hessian
##
## Error Analysis:
##      Estimate Std. Error t value Pr(>|t|)
## mu      5.234e-04 1.525e-04 3.432 0.000598 ***
## ar1     -4.818e-02 1.934e-02 -2.491 0.012727 *
## omega   1.798e-04 3.443e-05 5.222 1.77e-07 ***
## alpha1  9.809e-02 1.030e-02 9.525 < 2e-16 ***
## gamma1  1.000e+00 1.045e-02 95.731 < 2e-16 ***
## beta1   8.945e-01 1.049e-02 85.280 < 2e-16 ***
## delta   1.070e+00 1.350e-01 7.928 2.22e-15 ***
## shape   7.286e+00 1.123e+00 6.489 8.61e-11 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Log Likelihood:
## 8311.583 normalized: 3.302178
##
## Description:
## Sat Apr 07 10:34:39 2018 by user: yydab
##
##
## Standardised Residuals Tests:
##
##      Statistic p-Value
## Jarque-Bera Test R Chi^2 245.1569 0
## Shapiro-Wilk Test R W 0.9830579 0
## Ljung-Box Test R Q(10) 15.59584 0.1118014
## Ljung-Box Test R Q(15) 26.45095 0.03354158
## Ljung-Box Test R Q(20) 30.17074 0.06713391
## Ljung-Box Test R^2 Q(10) 19.17687 0.0380728
## Ljung-Box Test R^2 Q(15) 30.46672 0.01034547

```

```
## Ljung-Box Test      R^2  Q(20)  35.36462  0.01824615
## LM Arch Test       R    TR^2   29.57741  0.003231586
##
## Information Criterion Statistics:
##      AIC      BIC      SIC      HQIC
## -6.598000 -6.579468 -6.598020 -6.591274
```

```
plot(fit_32,which=3)
```

Series with 2 Conditional SD Superimposed



Conclusion: The distribution of the noise is not normal. But in the fGarch package there are various distributions to fit the data. May use skew distribution or some other distributions. Compare model fits – essentially the same but skew coefficient is significant. I would choose the model with skew distribution.

```
fit_33<- garchFit(~arma(1,0)+garch(1,1), data=djia, cond.dist="snorm")
```

```
##
## Series Initialization:
## ARMA Model:          arma
## Formula Mean:        ~ arma(1, 0)
## GARCH Model:         garch
## Formula Variance:    ~ garch(1, 1)
## ARMA Order:          1 0
## Max ARMA Order:      1
## GARCH Order:         1 1
## Max GARCH Order:     1
```

```

## Maximum Order:          1
## Conditional Dist:       snorm
## h.start:               2
## llh.start:             1
## Length of Series:      2517
## Recursion Init:        mci
## Series Scale:          0.01210097
##
## Parameter Initialization:
## Initial Parameters:     $params
## Limits of Transformations: $U, $V
## Which Parameters are Fixed? $includes
## Parameter Matrix:
##           U           V      params includes
## mu      -0.15336279   0.1533628  0.01533395      TRUE
## ar1     -0.99999999   1.0000000 -0.10129752      TRUE
## omega    0.00000100 100.0000000  0.10000000      TRUE
## alpha1   0.00000001  1.0000000  0.10000000      TRUE
## gamma1  -0.99999999   1.0000000  0.10000000     FALSE
## beta1    0.00000001  1.0000000  0.80000000      TRUE
## delta    0.00000000  2.0000000  2.00000000     FALSE
## skew     0.10000000 10.0000000  1.00000000      TRUE
## shape    1.00000000 10.0000000  4.00000000     FALSE
## Index List of Parameters to be Optimized:
## mu  ar1  omega alpha1  beta1  skew
##  1   2    3     4      6     8
## Persistence:          0.9
##
##
## --- START OF TRACE ---
## Selected Algorithm: nlminb
##
## R coded nlminb Solver:
##
## 0:    3064.7864: 0.0153339 -0.101298 0.100000 0.100000 0.800000 1.00000
## 1:    2998.2884: 0.0153343 -0.100527 0.0719804 0.0985390 0.786107 0.997935
## 2:    2958.0939: 0.0153351 -0.0987608 0.0475087 0.117064 0.790039 0.992986
## 3:    2941.5371: 0.0153363 -0.0967019 0.0453166 0.138342 0.806616 0.987375
## 4:    2935.0495: 0.0153391 -0.0924070 0.0213701 0.132931 0.812846 0.976980
## 5:    2914.5879: 0.0153441 -0.0851296 0.0291722 0.133128 0.830661 0.958631
## 6:    2895.7673: 0.0153589 -0.0672051 0.0123449 0.116131 0.868972 0.909676
## 7:    2895.0876: 0.0153708 -0.0619546 0.0186594 0.119789 0.858316 0.881580
## 8:    2894.4638: 0.0153729 -0.0621146 0.0147645 0.119450 0.856359 0.879298
## 9:    2893.1783: 0.0153819 -0.0653354 0.0158836 0.119196 0.859716 0.878265
## 10:   2892.8715: 0.0153955 -0.0678638 0.0141583 0.118022 0.863048 0.876952
## 11:   2892.4419: 0.0154354 -0.0649186 0.0149319 0.117933 0.863625 0.867946
## 12:   2892.4191: 0.0154938 -0.0591800 0.0141979 0.118016 0.862258 0.861047
## 13:   2892.1396: 0.0155486 -0.0590475 0.0143136 0.116923 0.865268 0.862195
## 14:   2892.0677: 0.0155834 -0.0630643 0.0137028 0.116263 0.866292 0.861082
## 15:   2892.0315: 0.0156557 -0.0636925 0.0138591 0.115788 0.867453 0.860665
## 16:   2891.9929: 0.0157152 -0.0610225 0.0133482 0.115314 0.868229 0.861692
## 17:   2891.9493: 0.0158913 -0.0645879 0.0136067 0.115605 0.867053 0.856624
## 18:   2891.9359: 0.0158914 -0.0645752 0.0137544 0.115711 0.867205 0.856614
## 19:   2891.9280: 0.0158916 -0.0645492 0.0135606 0.115780 0.867316 0.856592

```

```

## 20:      2891.9221: 0.0158979 -0.0644900 0.0136587 0.115877 0.867497 0.856533
## 21:      2891.9129: 0.0159121 -0.0643880 0.0135018 0.115850 0.867527 0.856431
## 22:      2891.9019: 0.0159407 -0.0641853 0.0135465 0.115876 0.867707 0.856234
## 23:      2891.7289: 0.0170988 -0.0575362 0.0134040 0.113732 0.869100 0.849695
## 24:      2890.9588: 0.0244146 -0.0765944 0.0120269 0.105468 0.877786 0.845200
## 25:      2890.0556: 0.0317358 -0.0811964 0.0134596 0.114641 0.869856 0.840726
## 26:      2889.4834: 0.0390587 -0.0808913 0.0128869 0.117402 0.863991 0.847025
## 27:      2889.0193: 0.0418075 -0.0760610 0.0146025 0.113896 0.865574 0.859131
## 28:      2888.7351: 0.0445616 -0.0744274 0.0137870 0.119007 0.865300 0.853164
## 29:      2888.4421: 0.0473166 -0.0700594 0.0127260 0.118278 0.866010 0.857754
## 30:      2888.3918: 0.0500732 -0.0685051 0.0130424 0.118626 0.865062 0.857219
## 31:      2888.3834: 0.0499739 -0.0667871 0.0129414 0.117479 0.866280 0.857178
## 32:      2888.3828: 0.0500362 -0.0667716 0.0130134 0.117249 0.866158 0.855841
## 33:      2888.3820: 0.0501018 -0.0667571 0.0129964 0.117371 0.866167 0.856486
## 34:      2888.3820: 0.0500955 -0.0667593 0.0129952 0.117369 0.866171 0.856481
##
## Final Estimate of the Negative LLH:
## LLH: -8222.837      norm LLH: -3.26692
##      mu      ar1      omega      alpha1      beta1
## 6.062042e-04 -6.675929e-02 1.902939e-06 1.173689e-01 8.661709e-01
##      skew
## 8.564812e-01
##
## R-optimhess Difference Approximated Hessian Matrix:
##      mu      ar1      omega      alpha1
## mu      -4.350147e+07 -2.897963e+04 -5.208178e+09 -1.921042e+05
## ar1      -2.897963e+04 -2.215986e+03 -1.394565e+06 1.160171e+02
## omega     -5.208178e+09 -1.394565e+06 -2.566138e+13 -8.569594e+08
## alpha1    -1.921042e+05 1.160171e+02 -8.569594e+08 -5.853531e+04
## beta1     -3.277218e+05 -1.760735e-01 -1.250712e+09 -6.810566e+04
## skew      1.578738e+04 2.235437e+02 7.745196e+06 -5.306309e+02
##      beta1      skew
## mu      -3.277218e+05 15787.3836
## ar1      -1.760735e-01 223.5437
## omega     -1.250712e+09 7745196.3326
## alpha1    -6.810566e+04 -530.6309
## beta1     -9.028858e+04 -470.9147
## skew      -4.709147e+02 -2274.4003
## attr("time")
## Time difference of 0.1546512 secs
##
## --- END OF TRACE ---
##
## Time to Estimate Parameters:
## Time difference of 0.495188 secs
summary(fit_33)

##
## Title:
## GARCH Modelling
##
## Call:
## garchFit(formula = ~arma(1, 0) + garch(1, 1), data = djiar, cond.dist = "snorm")

```

```
##
## Mean and Variance Equation:
## data ~ arma(1, 0) + garch(1, 1)
## <environment: 0x0000000021efe940>
## [data = djiar]
##
## Conditional Distribution:
## snorm
##
## Coefficient(s):
##      mu      ar1      omega      alpha1      beta1
## 6.0620e-04 -6.6759e-02 1.9029e-06 1.1737e-01 8.6617e-01
##      skew
## 8.5648e-01
##
## Std. Errors:
## based on Hessian
##
## Error Analysis:
##      Estimate Std. Error t value Pr(>|t|)
## mu      6.062e-04 1.553e-04 3.903 9.50e-05 ***
## ar1     -6.676e-02 2.144e-02 -3.114 0.00185 **
## omega   1.903e-06 3.720e-07 5.115 3.13e-07 ***
## alpha1  1.174e-01 1.269e-02 9.248 < 2e-16 ***
## beta1   8.662e-01 1.286e-02 67.367 < 2e-16 ***
## skew    8.565e-01 2.122e-02 40.365 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Log Likelihood:
## 8222.837    normalized: 3.26692
##
## Description:
## Sat Apr 07 10:34:40 2018 by user: yydab
##
##
## Standardised Residuals Tests:
##
##      Statistic p-Value
## Jarque-Bera Test R Chi^2 282.4382 0
## Shapiro-Wilk Test R W 0.9823857 0
## Ljung-Box Test R Q(10) 16.69802 0.08131888
## Ljung-Box Test R Q(15) 26.25585 0.03541792
## Ljung-Box Test R Q(20) 28.5981 0.09597972
## Ljung-Box Test R^2 Q(10) 17.12103 0.0717285
## Ljung-Box Test R^2 Q(15) 20.22023 0.1636462
## Ljung-Box Test R^2 Q(20) 23.47313 0.2661642
## LM Arch Test R TR^2 16.67805 0.1621144
##
## Information Criterion Statistics:
##      AIC      BIC      SIC      HQIC
## -6.529072 -6.515173 -6.529083 -6.524028
fit_31@fit$ics
```

```
##      AIC      BIC      SIC      HQIC
```

```
## -6.550353 -6.536453 -6.550364 -6.545309
```

```
fit_32@fit$ics
```

```
##      AIC      BIC      SIC      HQIC
```

```
## -6.598000 -6.579468 -6.598020 -6.591274
```

```
fit_33@fit$ics
```

```
##      AIC      BIC      SIC      HQIC
```

```
## -6.529072 -6.515173 -6.529083 -6.524028
```

```
# Forecasting from ARMA+GARCH models
```

```
#invisible(predict(fit_31,n.head=5,nx=200,plot=TRUE))
```

```
#invisible(predict(fit_32,n.head=5,nx=200,plot=TRUE))
```

```
invisible(predict(fit_33,n.head=5,nx=200,plot=TRUE))
```

Prediction with confidence intervals

