

INTC GARCH comparison

- Data: 1973 to 2009 monthly returns
- Build low order GARCH model
- Demonstrate empirical analysis of GARCH processes
- Compare different GARCH models
- Show predictions

GARCH(1,1) with normal innovations

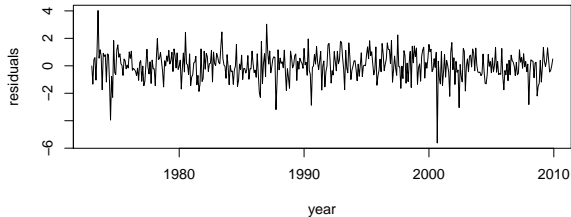
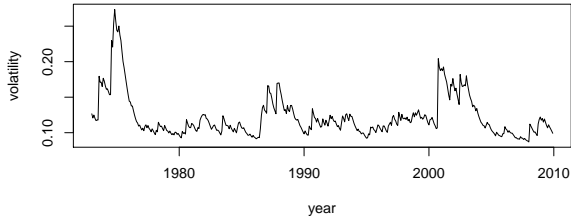
- The fitted model is

$$r_t = 0.0113 + a_t, a_t = \sigma_t \epsilon_t$$

$$\sigma_t^2 = 0.00092 + 0.086a_{t-1}^2 + 0.853\sigma_{t-1}^2.$$

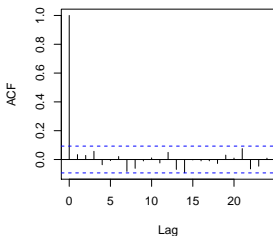
- All the estimates are significant, AIC=-1.3889.
- The unconditional variance is $0.000919/(1 - 0.0864 - 0.853) = 0.0152$. Sample variance is 0.0161.

Time plot of volatility series and standardized residuals

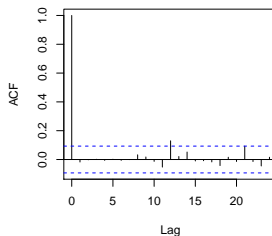


ACF and PACF of the residuals and residuals squares

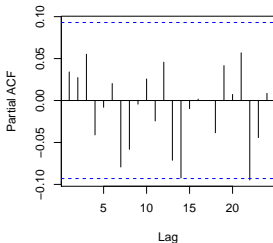
Series residual



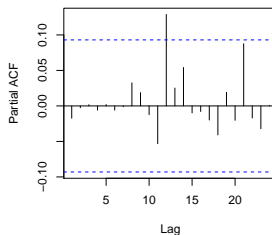
Series residual²



Series residual

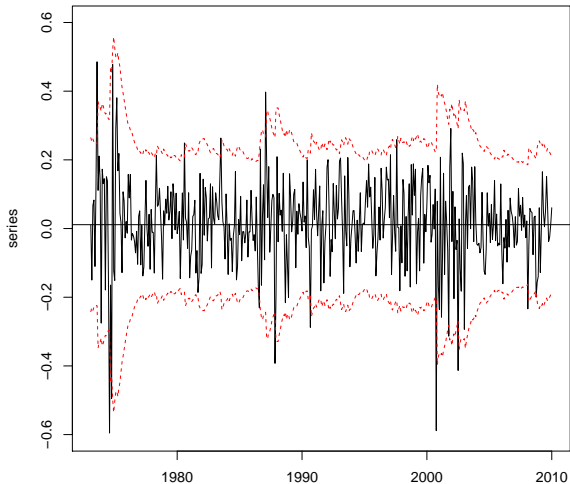


Series residual²

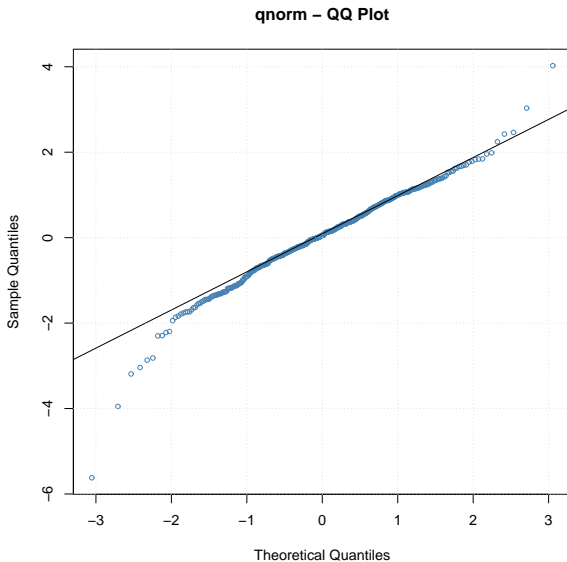


Predictive interval plot

This is the time plot of the 95% predictive intervals.



The QQ plots



GARCH(1,1) with t innovations

- From the qq plot it seems that innovations do not follow normal distribution.
- The fitted model with a t innovation is

$$r_t = 0.0165 + a_t, a_t = \sigma_t \epsilon_t, \epsilon_t \sim t_{6.77}.$$

$$\sigma_t^2 = 0.00116 + 0.1059a_{t-1}^2 + 0.817\sigma_{t-1}^2.$$

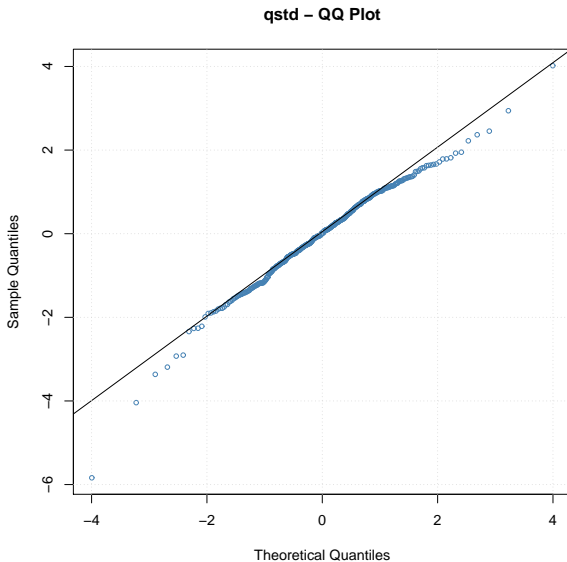
- All the estimates are significant, AIC=-1.447.
- The unconditional variance is $0.00116/(1 - 0.1059 - 0.817) = 0.015$.
- The fitted model with a skewed t innovation is

$$r_t = 0.0133 + a_t, a_t = \sigma_t \epsilon_t, \epsilon_t \sim t_{0.87, 7.23}.$$

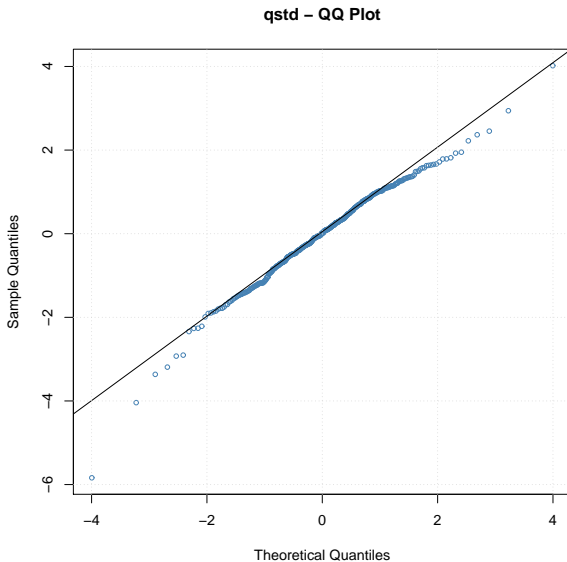
$$\sigma_t^2 = 0.00116 + 0.1049a_{t-1}^2 + 0.8178\sigma_{t-1}^2.$$

- All the estimates are significant, AIC=-1.4509.

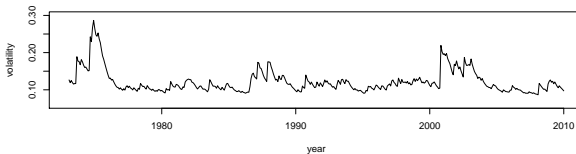
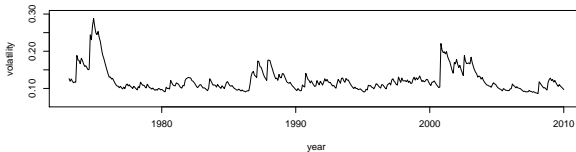
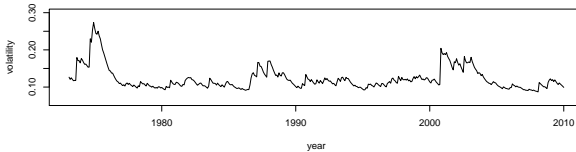
The QQ plots



The QQ plots



The time plots of volatilities



Forecast comparisons

model	1	2	3	4	5
normal	0.0975	0.0993	0.1009	0.1023	0.1037
t	0.0951	0.0975	0.0997	0.1016	0.1034
skewed t	0.0954	0.0979	0.1000	0.1019	0.1037

Table: Volatility forecasts for monthly log returns of intel stock

```

library(fGarch)
da=read.table("m-intcsp7309.txt", header=T)
intc=log(da$intc+1)
m1=garchFit(~1+garch(1,1), data=intc,trace=F)

summary(m1)
garchFit(formula = ~1 + garch(1, 1), data = intc, trace = F)
Error Analysis:
      Estimate Std. Error  t value Pr(>|t|)
mu      0.0112657   0.0053931    2.089  0.03672 *
omega   0.0009190   0.0003888    2.364  0.01808 *
alpha1  0.0864383   0.0265439    3.256  0.00113 **
beta1   0.8525855   0.0394322   21.622 < 2e-16 ***
---
Signif. codes:  0 *** 0.001 ** 0.01 * 0.05 . 0.1 1

Log Likelihood:
312.3307    normalized: 0.7034475
Standardised Residuals Tests:

      Statistic p-Value
Jarque-Bera Test  R    Chi^2 174.904 0
Shapiro-Wilk Test R    W      0.9709618 1.030402e-07
Ljung-Box Test    R    Q(10) 8.016844 0.6271916
Ljung-Box Test    R    Q(15) 15.5006 0.4159946
Ljung-Box Test    R    Q(20) 16.41549 0.6905368
Ljung-Box Test    R^2  Q(10) 0.8746345 0.9999072
Ljung-Box Test    R^2  Q(15) 11.35935 0.7267295
Ljung-Box Test    R^2  Q(20) 12.55994 0.8954573
LM Arch Test      R    TR^2 10.51401 0.5709617
Information Criterion Statistics:
      AIC      BIC      SIC      HQIC
-1.388877 -1.351978 -1.389037 -1.374326

```

```

v1=volatility(m1)
residual=residuals(m1, standardize=T)
vol=ts(v1, frequency=12, start=c(1973,1))
res=ts(residual, frequency=12, start=c(1973,1))
par(mfcol=c(2,1))
plot(vol,xlab='year',ylab='volatility',type='l')
plot(res,xlab='year',ylab='residuals',type='l')
par(mfcol=c(2,2))
acf(residual, lag=24)
pacf(residual, lag=24)
acf(residual^2, lag=24)
pacf(residual^2, lag=24)
#obtain plots of predictive intervals
par(mfcol=c(1,1))
upp=0.01126568+2*v1
low=0.01126568-2*v1
tdx=c(1:444)/12+1973
plot(tdx, intc, xlab='year',ylab='series',type='l', ylim=c(-0.6, 0.6))
lines(tdx, upp, lty=2, col='red')
lines(tdx, low, lty=2, col='red')
abline(h=c(0.01126568))
m2=garchFit(~1+garch(1,1), data=intc,trace=F,cond.dist='std')

summary(m2)

Title:
  GARCH Modelling

Call:
  garchFit(formula = ~1 + garch(1, 1), data = intc, cond.dist = "std",
    trace = F)

```

```

      Estimate Std. Error t value Pr(>|t|)
mu      0.0165075  0.0051031   3.235 0.001217 **
omega   0.0011576  0.0005782   2.002 0.045286 *
alpha1  0.1059030  0.0372047   2.846 0.004420 **
beta1   0.8171313  0.0580141  14.085 < 2e-16 ***
shape   6.7723503  1.8572367   3.646 0.000266 ***
---

```

```
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
Standardised Residuals Tests:
```

			Statistic	p-Value
Jarque-Bera Test	R	Chi ²	203.4933	0
Shapiro-Wilk Test	R	W	0.9687606	3.970356e-08
Ljung-Box Test	R	Q(10)	7.877778	0.6407741
Ljung-Box Test	R	Q(15)	15.5522	0.4124197
Ljung-Box Test	R	Q(20)	16.50475	0.6848581
Ljung-Box Test	R ²	Q(10)	1.066054	0.9997694
Ljung-Box Test	R ²	Q(15)	11.49875	0.7165045
Ljung-Box Test	R ²	Q(20)	12.61496	0.8932865
LM Arch Test	R	TR ²	10.80739	0.5454935

```
Information Criterion Statistics:
```

AIC	BIC	SIC	HQIC
-1.446966	-1.400841	-1.447215	-1.428776

```

v2=volatility(m2)
m3=garchFit(~1+garch(1,1), data=intc,trace=F,cond.dist='sstd')
summary(m3)
garchFit(formula = ~1 + garch(1, 1), data = intc, cond.dist = "sstd",
  trace = F)

```

```

      Estimate Std. Error t value Pr(>|t|)
mu      0.0133343   0.0053430    2.496 0.012572 *
omega   0.0011621   0.0005587    2.080 0.037519 *
alpha1  0.1049289   0.0358860    2.924 0.003456 **
beta1   0.8177875   0.0559863   14.607 < 2e-16 ***
skew    0.8717220   0.0629129   13.856 < 2e-16 ***
shape   7.2344224   2.1018054    3.442 0.000577 ***

Standardised Residuals Tests:

      Statistic p-Value
Jarque-Bera Test  R      Chi^2  195.2178  0
Shapiro-Wilk Test R      W      0.969251  4.893319e-08
Ljung-Box Test   R      Q(10)  7.882126  0.6403496
Ljung-Box Test   R      Q(15)  15.62496   0.4074054
Ljung-Box Test   R      Q(20)  16.5774   0.6802193
Ljung-Box Test   R^2     Q(10)  1.078429  0.9997569
Ljung-Box Test   R^2     Q(15)  11.95155  0.6826924
Ljung-Box Test   R^2     Q(20)  13.03792  0.8757513
LM Arch Test     R      TR^2   11.18826  0.5128574

Information Criterion Statistics:
      AIC      BIC      SIC      HQIC
-1.450899 -1.395550 -1.451257 -1.429071

```

```

v3=volatility(m3)
par(mfcol=c(3,1))
plot(tdx,v1,xlab='year',ylab='volatility',type='l', ylim=c(0.06, 0.3))
plot(tdx,v2,xlab='year',ylab='volatility',type='l', ylim=c(0.06, 0.3))
plot(tdx,v3,xlab='year',ylab='volatility',type='l', ylim=c(0.06, 0.3))
cor(cbind(v1,v2,v3))

```

```
x=intc[length(intc)]
at=(x-0.01126568)
f=0.0009190+0.0864383*at^2+0.8525855*(v1[length(v1)])^2

sqrt(f)

[1] 0.09754446

predict(m1,5)
  meanForecast  meanError standardDeviation
1  0.01126568  0.09754454      0.09754454
2  0.01126568  0.09926616      0.09926616
3  0.01126568  0.10085604      0.10085604
4  0.01126568  0.10232650      0.10232650
5  0.01126568  0.10368830      0.10368830
```