

# Thesis\_R\_Code

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## Loading Libraries

```
library(rugarch)
```

```
## Loading required package: parallel
```

```
##
```

```
## Attaching package: 'rugarch'
```

```
## The following object is masked from 'package:stats':
```

```
##
```

```
##      sigma
```

```
library(readr)
```

```
library(forecast)
```

```
## Registered S3 method overwritten by 'quantmod':
```

```
##   method      from
```

```
## as.zoo.data.frame zoo
```

```
library(tseries)
```

```
library(garchx)
```

```
## Loading required package: zoo
```

```
##
```

```
## Attaching package: 'zoo'
```

```
## The following objects are masked from 'package:base':
```

```
##
```

```
##      as.Date, as.Date.numeric
```

```
library(TSA)
```

```
## Registered S3 methods overwritten by 'TSA':
```

```
##   method      from
```

```
## fitted.Arima forecast
```

```
## plot.Arima   forecast
```

```
##
```

```
## Attaching package: 'TSA'
```

```
## The following object is masked from 'package:readr':
```

```
##
```

```
##      spec
```

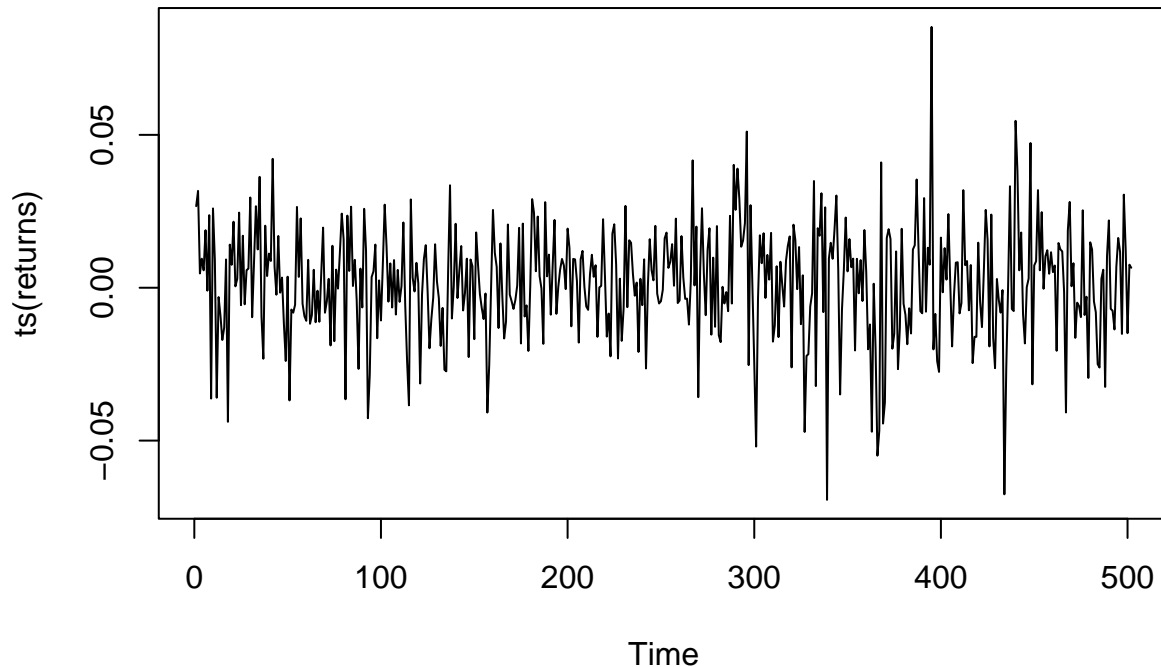
```
## The following objects are masked from 'package:stats':
```

```
##
```

```
##      acf, arima
## The following object is masked from 'package:utils':
##
##      tar
```

## Loading Data and Calculating Returns

```
stock_data <- read.csv("/Users/pranavtavildar/Desktop/Senior-Thesis/data/CVX_stock_data.csv")
returns <- diff(log(stock_data$Close))
plot(ts(returns))
```



Here we use `auto.arima` to select the best arima model

```
arma_model <- auto.arima(returns, max.D = 0, max.d = 0)
arma_model
```

```
## Series: returns
## ARIMA(0,0,0) with non-zero mean
##
## Coefficients:
##      mean
##      0.0015
## s.e.  0.0008
##
## sigma^2 = 0.000339: log likelihood = 1293.56
## AIC=-2583.12   AICc=-2583.1   BIC=-2574.68
```

Here, we are doing Garch where the order is (1,1) and we apply it to the arima model selected by auto.arima

```
garch_spec <- ugarchspec(variance.model = list(model = "sGARCH", garchOrder = c(1,1)), mean.model = list(model = "ARFIMA", arfimaOrder = c(0,0,0)), riskMetrics = list())
garch_model2 <- ugarchfit(garch_spec, data = returns)
summary(garch_model2)
```

```
##      Length      Class      Mode
##      1 uGARCHfit      S4
```

```
show(garch_model2)
```

```
##
## *-----*
## *          GARCH Model Fit          *
## *-----*
##
## Conditional Variance Dynamics
## -----
## GARCH Model   : sGARCH(1,1)
## Mean Model    : ARFIMA(0,0,0)
## Distribution   : norm
##
## Optimal Parameters
## -----
##      Estimate  Std. Error  t value Pr(>|t|)
## mu      0.001503    0.000786   1.91115 0.055985
## omega    0.000003    0.000004   0.82203 0.411059
## alpha1   0.024261    0.011361   2.13550 0.032721
## beta1    0.966171    0.014672  65.85297 0.000000
##
## Robust Standard Errors:
##      Estimate  Std. Error  t value Pr(>|t|)
## mu      0.001503    0.000786   1.91204 0.055872
## omega    0.000003    0.000016   0.20481 0.837723
## alpha1   0.024261    0.027300   0.88868 0.374173
## beta1    0.966171    0.046123  20.94758 0.000000
##
## LogLikelihood : 1300.803
##
## Information Criteria
## -----
##
## Akaike          -5.1665
## Bayes           -5.1329
## Shibata         -5.1667
## Hannan-Quinn    -5.1534
##
## Weighted Ljung-Box Test on Standardized Residuals
## -----
##      statistic p-value
## Lag[1]          0.4181 0.5179
## Lag[2*(p+q)+(p+q)-1] [2] 0.4548 0.7152
## Lag[4*(p+q)+(p+q)-1] [5] 1.3065 0.7873
## d.o.f=0
```

```

## H0 : No serial correlation
##
## Weighted Ljung-Box Test on Standardized Squared Residuals
## -----
##
##               statistic p-value
## Lag[1]                0.1824  0.6693
## Lag[2*(p+q)+(p+q)-1][5]  1.7904  0.6687
## Lag[4*(p+q)+(p+q)-1][9]  3.8059  0.6219
## d.o.f=2
##
## Weighted ARCH LM Tests
## -----
##
##      Statistic Shape Scale P-Value
## ARCH Lag[3]      2.137 0.500 2.000  0.1438
## ARCH Lag[5]      2.913 1.440 1.667  0.3026
## ARCH Lag[7]      4.259 2.315 1.543  0.3108
##
## Nyblom stability test
## -----
## Joint Statistic:  11.883
## Individual Statistics:
## mu      0.05534
## omega   1.16577
## alpha1  0.18610
## beta1   0.18235
##
## Asymptotic Critical Values (10% 5% 1%)
## Joint Statistic:      1.07 1.24 1.6
## Individual Statistic:  0.35 0.47 0.75
##
## Sign Bias Test
## -----
##
##      t-value   prob sig
## Sign Bias      0.7048 0.4813
## Negative Sign Bias 0.2108 0.8331
## Positive Sign Bias 0.6407 0.5220
## Joint Effect    0.5902 0.8987
##
##
## Adjusted Pearson Goodness-of-Fit Test:
## -----
##      group statistic p-value(g-1)
## 1      20      33.46      0.02127
## 2      30      40.99      0.06903
## 3      40      54.33      0.05226
## 4      50      70.11      0.02552
##
##
## Elapsed time : 0.08320093

```

t value of beta 1 proves that the p value is non significant

## Alternative Garch implementation gives us the same result

```
garch_model <- garch(returns)
```

```
##
## ***** ESTIMATION WITH ANALYTICAL GRADIENT *****
##
##
##      I      INITIAL X(I)      D(I)
##
##      1      3.050993e-04      1.000e+00
##      2      5.000000e-02      1.000e+00
##      3      5.000000e-02      1.000e+00
##
##      IT      NF      F      RELDF      PRELDF      RELDX      STPPAR      D*STEP      NPRELDF
##      0      1 -1.752e+03
##      1      8 -1.752e+03  1.17e-08  6.43e-07  1.0e-05  1.1e+09  1.0e-06  3.62e+02
##      2      9 -1.752e+03  1.70e-07  2.06e-07  5.0e-06  2.1e+00  5.0e-07  9.03e-03
##      3     10 -1.752e+03  1.52e-09  4.40e-09  1.0e-05  2.0e+00  1.0e-06  9.03e-03
##      4     13 -1.752e+03  3.98e-08  4.72e-08  1.7e-04  4.3e+00  1.7e-05  9.02e-03
##      5     18 -1.752e+03  1.38e-05  2.76e-05  7.9e-02  2.0e+00  8.6e-03  8.92e-03
##      6     22 -1.752e+03  3.24e-04  2.98e-05  7.6e-01  0.0e+00  3.2e-01  2.98e-05
##      7     24 -1.753e+03  3.28e-04  2.37e-04  1.5e-01  2.0e+00  1.3e-01  3.25e-02
##      8     26 -1.755e+03  1.11e-03  7.85e-04  2.0e-01  2.0e+00  2.6e-01  5.18e+00
##      9     28 -1.755e+03  1.03e-04  2.96e-04  3.3e-02  2.0e+00  5.2e-02  8.42e+01
##     10     30 -1.755e+03  8.86e-05  2.31e-04  1.4e-02  2.0e+00  2.4e-02  5.75e+01
##     11     31 -1.755e+03  6.85e-05  9.94e-05  1.4e-02  2.0e+00  2.4e-02  2.40e-01
##     12     32 -1.755e+03  5.91e-07  1.96e-05  1.4e-02  2.0e+00  2.4e-02  2.00e-01
##     13     34 -1.755e+03  7.46e-06  7.64e-06  7.1e-03  2.0e+00  1.2e-02  6.51e-01
##     14     40 -1.755e+03  1.39e-07  4.18e-06  2.3e-07  5.1e+00  3.8e-07  3.74e-02
##     15     41 -1.755e+03  9.96e-07  8.51e-07  1.0e-07  2.0e+00  1.9e-07  1.40e-03
##     16     52 -1.755e+03  3.08e-05  1.11e-05  2.9e-02  1.9e+00  5.0e-02  1.61e-03
##     17     54 -1.755e+03  2.09e-05  1.52e-05  5.6e-03  2.0e+00  1.0e-02  6.91e-01
##     18     55 -1.755e+03  2.97e-06  3.58e-05  1.1e-02  2.0e+00  2.0e-02  2.32e+02
##     19     61 -1.755e+03  6.46e-06  1.16e-05  1.1e-07  1.8e+01  2.0e-07  7.03e-03
##     20     69 -1.755e+03  1.33e-05  1.93e-05  1.8e-03  2.0e+00  3.3e-03  1.53e-02
##     21     72 -1.756e+03  2.46e-04  2.33e-04  6.3e-03  1.2e+00  1.3e-02  2.89e-03
##     22     73 -1.756e+03  4.45e-05  4.79e-04  1.2e-02  2.0e+00  2.6e-02  1.02e+00
##     23     74 -1.757e+03  4.49e-04  8.23e-04  4.7e-03  0.0e+00  1.2e-02  8.23e-04
##     24     76 -1.757e+03  1.07e-04  4.04e-04  1.7e-03  1.4e+00  3.2e-03  4.97e-04
##     25     77 -1.757e+03  1.46e-04  1.19e-04  1.3e-03  1.5e+00  3.2e-03  1.35e-04
##     26     79 -1.757e+03  6.37e-05  1.15e-04  5.0e-03  3.6e-01  1.1e-02  1.24e-04
##     27     80 -1.757e+03  3.71e-05  4.35e-05  1.3e-03  0.0e+00  3.0e-03  4.35e-05
##     28     81 -1.757e+03  3.06e-06  3.62e-06  8.5e-04  0.0e+00  1.7e-03  3.62e-06
##     29     82 -1.757e+03  2.48e-07  3.48e-07  1.3e-04  0.0e+00  2.6e-04  3.48e-07
##     30     83 -1.757e+03  1.29e-08  9.34e-09  1.2e-05  0.0e+00  2.5e-05  9.34e-09
##     31     84 -1.757e+03  1.62e-09  1.25e-09  9.9e-06  0.0e+00  2.4e-05  1.25e-09
##     32     85 -1.757e+03 -2.67e-10  6.95e-12  1.2e-06  0.0e+00  2.5e-06  6.95e-12
##
## ***** RELATIVE FUNCTION CONVERGENCE *****
##
##      FUNCTION      -1.757344e+03      RELDX      1.152e-06
##      FUNC. EVALS      85      GRAD. EVALS      32
##      PRELDF      6.954e-12      NPRELDF      6.954e-12
```

```
##
##      I      FINAL X(I)      D(I)      G(I)
##
##      1      3.326084e-06      1.000e+00      -1.007e+00
##      2      2.427247e-02      1.000e+00      1.080e-03
##      3      9.658585e-01      1.000e+00      1.137e-02
```

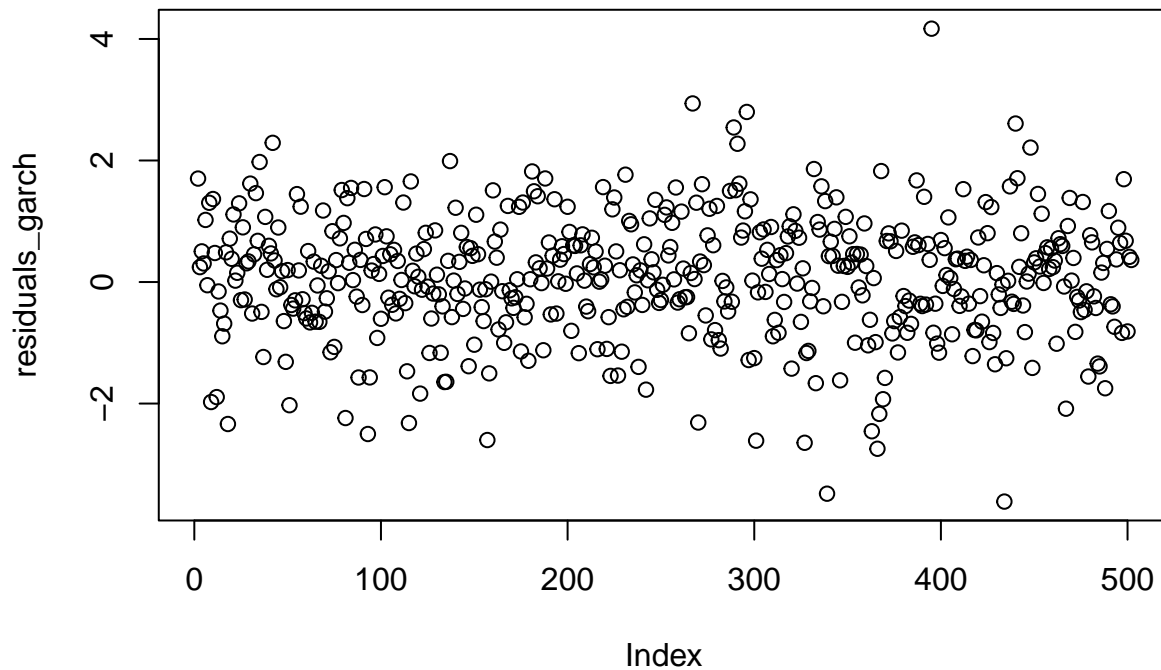
```
garch_model
```

```
##
## Call:
## garch(x = returns)
##
## Coefficient(s):
##      a0      a1      b1
## 3.326e-06 2.427e-02 9.659e-01
```

```
summary(arma_model)
```

```
## Series: returns
## ARIMA(0,0,0) with non-zero mean
##
## Coefficients:
##      mean
##      0.0015
## s.e. 0.0008
##
## sigma^2 = 0.000339: log likelihood = 1293.56
## AIC=-2583.12 AICc=-2583.1 BIC=-2574.68
##
## Training set error measures:
##      ME      RMSE      MAE      MPE      MAPE      MASE
## Training set 7.762973e-19 0.01839358 0.01405025 86.25959 115.9893 0.7069355
##      ACF1
## Training set 0.03172501
```

```
residuals_garch <- residuals(garch_model)
plot(residuals_garch)
```



### Adding Exogenous Variable (sentiment score)

```
sentiment_df <- read.csv("/Users/pranavtavildar/Desktop/Senior-Thesis/data/daily_sentiments.csv", header = TRUE)
str(sentiment_df)
```

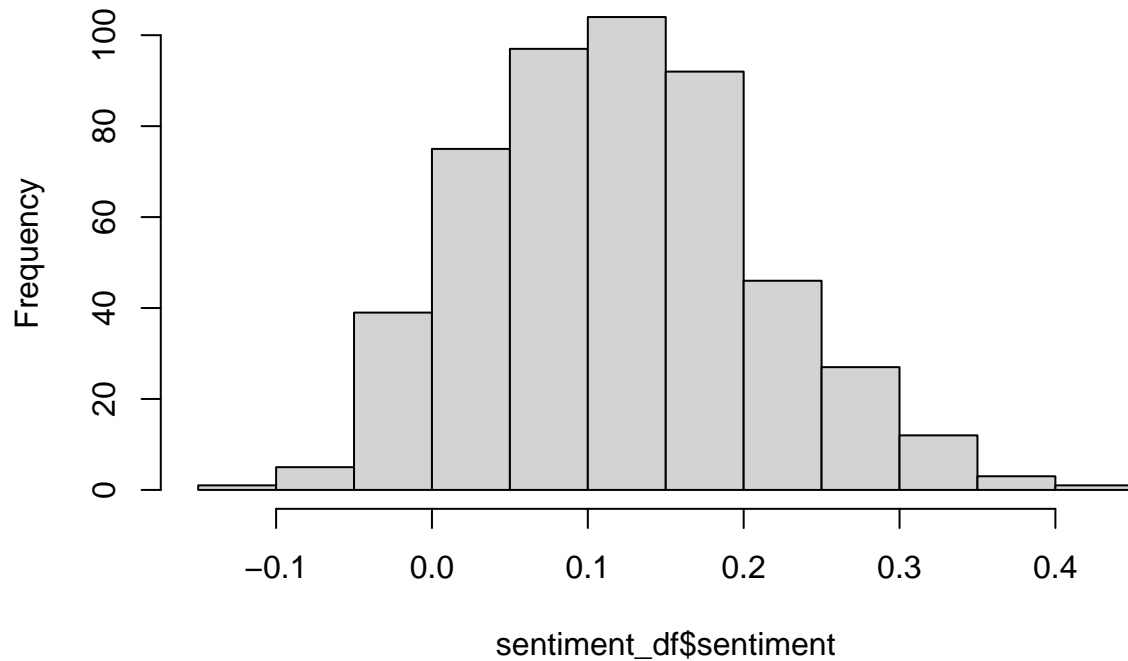
```
## 'data.frame':    503 obs. of  2 variables:
## $ Date      : chr  "2021-01-04" "2021-01-05" "2021-01-06" "2021-01-07" ...
## $ sentiment: num  0.1328 0.1231 0.0734 0.0328 0.0576 ...
```

```
str(returns)
```

```
## num [1:502] 0.02667 0.03168 0.00467 0.00949 0.00569 ...
```

```
sentiment_df = sentiment_df[-1,]
hist(sentiment_df$sentiment)
```

## Histogram of sentiment\_df\$sentiment



```
output_path <- "/Users/pranavtavildar/Desktop/Senior-Thesis/FrontiersPoster/FrontiersPoster_files/figure"
png(file = output_path)
hist(sentiment_df$sentiment, main = "Sentiment Distribution", xlab = "Sentiment Score", ylab = "Frequency",
     dev.off())

## pdf
## 2
```

## Fitting Arima Model to the new data

```
arma_model <- auto.arima(returns, max.D = 0, max.d = 0)
arma_model

## Series: returns
## ARIMA(0,0,0) with non-zero mean
##
## Coefficients:
##      mean
##      0.0015
## s.e.  0.0008
##
## sigma^2 = 0.000339: log likelihood = 1293.56
## AIC=-2583.12   AICc=-2583.1   BIC=-2574.68

sentiment_arma_model <- auto.arima(sentiment_df$sentiment, max.D = 0, max.d = 0)
sentiment_arma_model

## Series: sentiment_df$sentiment
## ARIMA(0,0,0) with non-zero mean
```



```
##
## Coefficients:
##      mean
##      0.119
## s.e.  0.004
##
## sigma^2 = 0.00785:  log likelihood = 504.85
## AIC=-1005.7   AICc=-1005.67   BIC=-997.26
```

## Adjusting Scale for Garchx

```
adj_sentiment <- sentiment_df$sentiment/100
```

## Running arimax and Garchx

```
###arimax
exarma_model <- arimax(x = returns, xreg = sentiment_df$sentiment, order= c(1,0,1))
exarma_model
```

```
##
## Call:
## arimax(x = returns, order = c(1, 0, 1), xreg = sentiment_df$sentiment)
##
## Coefficients:
##      ar1      ma1 intercept      xreg
##      0.6004 -0.5787      0.0008  0.0059
## s.e.  0.5483  0.5661      0.0014  0.0093
##
## sigma^2 estimated as 0.0003378:  log likelihood = 1293.98,  aic = -2579.95
```

```
###garchx
centered_returns <- scale(returns, center = TRUE, scale = FALSE)
garchx_model <- garchx(centered_returns, xreg=adj_sentiment, order = c(1,1), initial.values = c(coef(ga
garchx_model
```

```
## Warning in sqrt(diag(vcovmat)): NaNs produced
```

```
##
## Date: Tue Apr 11 12:00:53 2023
## Method: normal ML
## Coefficient covariance: ordinary
## Message (nlminb): false convergence (8)
## No. of observations (fitted): 501
## Sample: 1 to 502
##
##      intercept      arch1      garch1      xreg1
## Estimate:  3.309414e-06 0.02427247 0.9658585 9.931742e-11
## Std. Error:           NaN           NaN           NaN 1.542782e-03
##
## Log-likelihood: 1298.663
```

## Repeating Process with Weighted Sentiment Score

```

### weighted sentiment score
weighted_df <- read.csv("/Users/pranavtavildar/Desktop/Senior-Thesis/data/weighted_daily_sentiments.csv")
str(weighted_df)

## 'data.frame': 503 obs. of 3 variables:
## $ Date : chr "2021-01-04" "2021-01-05" "2021-01-06" "2021-01-07" ...
## $ weighted_sentiment: num 0.18486 0.37177 0.22083 0.03277 0.00859 ...
## $ adjusted_likes : int 18 99 18 10 67 12 30 56 74 15 ...

str(returns)

## num [1:502] 0.02667 0.03168 0.00467 0.00949 0.00569 ...

weighted_df = weighted_df[-1,]

###arimax
exarma_model <- arimax(x = returns, xreg = weighted_df$weighted_sentiment)
exarma_model

##
## Call:
## arimax(x = returns, xreg = weighted_df$weighted_sentiment)
##
## Coefficients:
## intercept xreg
## 0.0023 -0.0069
## s.e. 0.0011 0.0063
##
## sigma^2 estimated as 0.0003375: log likelihood = 1294.16, aic = -2584.32

###garchx
garchx_model <- garchx(returns, xreg=sentiment_df$weighted_sentiment)
summary(garchx_model)

##
## Length Class Mode
## date 1 -none- character
## sys.call 3 -none- call
## y.name 1 -none- character
## y.n 1 -none- numeric
## recursion.n 1 -none- numeric
## y.coredata 502 -none- numeric
## y.index 502 -none- numeric
## y2 502 -none- numeric
## y2mean 1 -none- numeric
## order 3 -none- numeric
## archK 1 -none- numeric
## arch 1 -none- numeric
## archOrder 1 -none- numeric
## garchK 1 -none- numeric
## garch 1 -none- numeric
## garchOrder 1 -none- numeric
## asymK 1 -none- numeric
## asymOrder 1 -none- numeric
## xregK 1 -none- numeric
## maxpqr 1 -none- numeric
## maxpqrpluss1 1 -none- numeric

```

## archIndx	1	-none- numeric
## garchIndx	1	-none- numeric
## asymIndx	1	-none- numeric
## xregIndx	1	-none- numeric
## backcast.values	1	-none- numeric
## y2short	501	-none- numeric
## y2matrix	502	-none- numeric
## sigma2	502	-none- numeric
## initial.values	3	-none- numeric
## upper	1	-none- numeric
## lower	1	-none- numeric
## control	0	-none- list
## hessian.control	0	-none- list
## solve.tol	1	-none- numeric
## c.code	1	-none- logical
## sigma2.min	1	-none- numeric
## objective.fun	1	-none- numeric
## penalty.value	1	-none- numeric
## par	3	-none- numeric
## objective	1	-none- numeric
## convergence	1	-none- numeric
## iterations	1	-none- numeric
## evaluations	2	-none- numeric
## message	1	-none- character
## fitted	501	zoo numeric
## residuals	501	zoo numeric
## hessian	9	-none- numeric
## vcov	9	-none- numeric