

# Package ‘RobGARCHBoot’

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**Type** Package

**Title** Robust Bootstrap Forecast Densities for GARCH Models

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**Description** Bootstrap forecast densities for GARCH (Generalized Autoregressive Conditional Heteroskedastic) returns and volatilities using the robust residual-based bootstrap procedure of Trucios, Hotta and Ruiz (2017) <DOI:10.1080/00949655.2017.1359601>.

**Encoding** UTF-8

**License** GPL (>= 2)

**Depends** R (>= 3.5.0)

**Imports** Rcpp (>= 1.0.1)

**LinkingTo** Rcpp, RcppArmadillo

**LazyData** true

**RoxygenNote** 6.1.1

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RobGARCHBoot-package    *Robust Bootstrap Forecast Densities for GARCH Models*

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### Description

Bootstrap forecast densities for returns and volatilities using the robust residual-based bootstrap procedure of Trucíos et al. (2017). The package also includes the robust GARCH (Generalized Autoregressive Conditional Heteroskedastic) estimator of Boudt et al. (2013) with the modification introduced by Trucíos et al. (2017).

### Details

This package provides a robust bootstrap procedure to obtain forecast densities for both return and volatilities in a GARCH context. The forecast densities are useful to obtain forecast intervals as well as to estimate risk measures such as Value-at-Risk (VaR). Additionally, we also provide the robust GARCH estimator of Boudt et al. (2013) with the modification introduced by Trucíos et al. (2017). This procedure showed good finite sample properties in both Monte Carlo experiments and empirical data. For a recent implementation of this procedure see Trucíos (2019).

### Author(s)

Carlos Trucíos <ctrucios@gmail.com>

### References

Boudt, Kris, Jon Danielsson, and Sébastien Laurent. Robust forecasting of dynamic conditional correlation GARCH models. *International Journal of Forecasting* 29.2 (2013): 244-257.

Trucíos, Carlos, Luiz K. Hotta, and Esther Ruiz. Robust bootstrap forecast densities for GARCH returns and volatilities. *Journal of Statistical Computation and Simulation* 87.16 (2017): 3152-3174.

Trucíos, Carlos. Forecasting Bitcoin risk measures: A robust approach. *International Journal of Forecasting* 35.3 (2019): 836-847.

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fitted_Vol	<i>Estimated Volatility</i>
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### Description

Using the robust estimated parameters of Boudt et al. (2013) and Trucíos et al. (2017), we obtain the estimated volatility.

### Usage

```
fitted_Vol(theta,r)
```

### Arguments

theta	Vector of robust estimated parameters obtained from ROBUSTGARCH function.
r	Vector of time series returns.

**Details**

More details can be found in Boudt et al. (2013) and Trucíos et al. (2017).

**Value**

The function returns the estimated volatility from 1 to T+1.

**Author(s)**

Carlos Trucíos

**References**

Boudt, Kris, Jon Danielsson, and Sébastien Laurent. Robust forecasting of dynamic conditional correlation GARCH models. *International Journal of Forecasting* 29.2 (2013): 244-257.

Trucíos, Carlos, Luiz K. Hotta, and Esther Ruiz. Robust bootstrap forecast densities for GARCH returns and volatilities. *Journal of Statistical Computation and Simulation* 87.16 (2017): 3152-3174.

**Examples**

```
# Using the Bitcoin daily returns, we estimate the parameter of the GARCH model in a robust way
param = ROBUSTGARCH(returnsexample)
# With the estimated parameters, we estimate the volatility in a robust way
vol = fitted_Vol(param, returnsexample)
```

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returnsexample	<i>Time series returns for illustrative purposes</i>
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**Description**

Cryptocurrencies report large returns over time. In this sense and with illustrative purposes, we use Bitcoin daily returns from July 2014 to February 2017.

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RobGARCHBoot	<i>Robust GARCH bootstrap procedure</i>
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**Description**

Robust GARCH (Generalized Autoregressive Conditional Heteroskedastic) Bootstrap procedure of Trucíos et al. (2017)

**Usage**

```
RobGARCHBoot(data, n.boot = 1000, n.ahead = 1)
```

**Arguments**

data	Vector of time series returns.
n.boot	Number of bootstrap replications. By default n.boot = 1000
n.ahead	Numbers of steps-ahead. By default n.ahead = 1

**Details**

More details can be found in Trucíos et al. (2017), Hotta and Trucíos (2018) and Trucíos (2019).

**Value**

The function returns two lists with the empirical H-steps-ahead bootstrap densities for returns and squared volatilities.

**Author(s)**

Carlos Trucíos

**References**

Hotta, Luiz Koodi, and Carlos Trucíos. Inference in (M)GARCH models in the presence of additive outliers: Specification, estimation, and prediction. *Advances in Mathematics and Applications*. Springer, Cham, 2018. 179-202.

Trucíos, Carlos, Luiz K. Hotta, and Esther Ruiz. Robust bootstrap forecast densities for GARCH returns and volatilities. *Journal of Statistical Computation and Simulation* 87.16 (2017): 3152-3174.

Trucíos, Carlos. Forecasting Bitcoin risk measures: A robust approach. *International Journal of Forecasting* 35.3 (2019): 836-847.

**Examples**

```
# Robust bootstrap forecast densities for returns and volatilities
boot = RobGARCHBoot(returnsexample, n.boot = 1000, n.ahead = 1)

# Obtaining the forecast intervals for returns (95%)
quantile(boot[[1]], prob = c(0.025, 0.975))
# Obtaining the forecast intervals for volatilities (95%)
quantile(boot[[2]], prob = c(0.025, 0.975))

# Risk measures can also be obtained
VaR1 = quantile(boot[[1]], prob = 0.01)
```

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ROBUSTGARCH

*Robust GARCH Estimator*

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**Description**

Robust GARCH (Generalized Autoregressive Conditional Heteroskedastic) estimator of Boudt et al. (2013) with the modification introduced by Trucíos et al. (2017).

**Usage**

ROBUSTGARCH(y)

**Arguments**

y                      Vector of time series returns.

**Details**

More details can be found in Boudt et al. (2013) and Trucíos et at. (2017).

**Value**

The function returns the estimated parameters.

**Author(s)**

Carlos Trucíos

**References**

Boudt, Kris, Jon Danielsson, and Sébastien Laurent. Robust forecasting of dynamic conditional correlation GARCH models. *International Journal of Forecasting* 29.2 (2013): 244-257.

Trucíos, Carlos, Luiz K. Hotta, and Esther Ruiz. Robust bootstrap forecast densities for GARCH returns and volatilities. *Journal of Statistical Computation and Simulation* 87.16 (2017): 3152-3174.

**Examples**

```
# Estimating the parameters of the GARCH model in a robust way.
param = ROBUSTGARCH(returnsexample*100)
param
```

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ROBUSTGARCHloss\_RCPP    *Loss function used in GARCH robust estimation.*

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**Description**

Loss function used in GARCH (Generalized Autoregressive Conditional Heteroskedastic) robust estimation.

**Usage**

```
ROBUSTGARCHloss_RCPP(theta, r, sigma2)
```

**Arguments**

theta	Vector of robust estimated (or initial values) parameters obtained from ROBUSTGARCH function.
r	Vector of time series returns.
sigma2	robust squared volatility estimation (or initial value of squared volatility)

**Details**

This functions is used in the robust estimation. We can use it to evaluate the value of the loss function using several values of the vector parameters (theta)

**Value**

Returns the value of the loss function

**Author(s)**

Carlos Trucíos

**References**

Boudt, Kris, Jon Danielsson, and Sébastien Laurent. Robust forecasting of dynamic conditional correlation GARCH models. *International Journal of Forecasting* 29.2 (2013): 244-257.

Trucíos, Carlos, Luiz K. Hotta, and Esther Ruiz. Robust bootstrap forecast densities for GARCH returns and volatilities. *Journal of Statistical Computation and Simulation* 87.16 (2017): 3152-3174.

**Examples**

```
# Using the Bitcoin daily returns, we estimate the parameter of the GARCH model in a robust way
param = ROBUSTGARCH(returnsexample)
# We can evaluate the loss function using the estimated parameters
ROBUSTGARCHloss_RCPP(param[2:3], returnsexample, param[1]/(1-param[2]-param[3]))
```

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