

Package: garchf (via r-universe)

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Type Package

Title Forecasting with GARCH

Version 0.1.0

Description Forecasting with GARCH, using an interface like the forecast package's.

Imports forecast, rugarch

Suggests MASS, fpp2, ahead, crossvalidation

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 condvolf

Model-agnostic statistical probabilistic forecasting with conditional volatility

Description

Combines a mean model and a volatility model to produce probabilistic forecasts. Uncertainty is propagated from two sources: simulated innovation draws and simulated volatility paths derived from the variance model's residuals via bootstrap.

Usage

```
condvolf(
  y,
  h = 10L,
  mean_model = forecast::thetaf,
  sigma_model = forecast::thetaf,
  innovation = c("empirical", "gaussian", "student"),
  level = 95,
  B = 1000L,
  bootstrap_vol = TRUE,
  conformal = FALSE,
  cal_frac = 0.5,
  residuals_model = forecast::thetaf,
  ...
)
```

Arguments

y	A numeric vector or time series of class ts.
h	Forecasting horizon.
mean_model	Function to fit the mean model (default: forecast::auto.arima). Options include forecast::auto.arima, forecast::ets, forecast::thetaf, or any custom function that returns an object with forecast method.
sigma_model	Function to fit the variance model on squared residuals (default: forecast::auto.arima).
innovation	Distribution for standardized innovations. One of "gaussian" (fastest, often sufficient), "student" (heavy-tailed), or "empirical" (non-parametric, most flexible). Only used for conformal=FALSE
level	Confidence level for prediction intervals (default: 95).
B	Number of simulated paths (default: 2000).
bootstrap_vol	Whether to bootstrap volatility residuals (default: TRUE). If FALSE, uses parametric normal approximation for volatility.

<code>conformal</code>	Logical. If TRUE, enables split conformal prediction to calibrate interval width. The series is split into a training portion and a calibration portion (controlled by <code>cal_frac</code>); a base <code>condvolf</code> fit on the training portion produces calibration residuals, which are normalized by a volatility model, resampled via moving block bootstrap, re-inflated by the forecast volatility, and added to the full-data point forecast. This typically yields tighter and better-calibrated intervals than the raw simulation approach. Default is FALSE.
<code>cal_frac</code>	Numeric in $(0, 1)$. Fraction of observations reserved for the calibration set when <code>conformal = TRUE</code> . The calibration set is always at least <code>h</code> observations long regardless of <code>cal_frac</code> . Larger values improve calibration stability at the cost of a shorter training set for the base model. Default is 0.5.
<code>residuals_model</code>	A function from package <code>forecast</code> , the model adjusted to residuals for 2-stage forecasting, and only for <code>conformal=TRUE</code>
<code>...</code>	Additional arguments passed to <code>mean_model</code> and <code>sigma_model</code> .

Value

A forecast object with components:

- `x` Original time series.
- `mean` Point forecast (mean of simulations).
- `lower` Lower prediction interval bound.
- `upper` Upper prediction interval bound.
- `level` Confidence level.
- `sims` Matrix of simulated forecast paths ($h \times B$).
- `model` List containing mean and `sigma` model fits.
- `residuals` Residuals from the mean model.
- `standardized_residuals` Scaled standardized residuals used for innovation fitting.
- `method` Character string describing the method used.

Examples

```
library(forecast)

# Basic usage with Google stock data
y <- fpp2::goog200

# Gaussian innovations (fastest, often sufficient)
fc1 <- condvolf(y, h = 20, innovation = "gaussian")
plot(fc1)

# Student-t innovations (heavy-tailed)
fc2 <- condvolf(y, h = 20, innovation = "student")
plot(fc2)

# Empirical innovations (non-parametric)
```

```

fc3 <- condvolf(y, h = 20, innovation = "empirical")
plot(fc3)

# Using different mean and volatility models
fc4 <- condvolf(y, h = 20,
               mean_model = forecast::thetaf,
               sigma_model = forecast::ets,
               innovation = "gaussian")

# Compare prediction intervals
par(mfrow = c(2, 2))
plot(fc1, main = "Gaussian innovations")
plot(fc2, main = "Student-t innovations")
plot(fc3, main = "Empirical innovations")
plot(fc4, main = "Different mean and volatility models")
par(mfrow = c(1, 1))

```

xgarchf

GARCH forecasting wrapper returning a forecast object

Description

Similar spirit to `forecast::thetaf()`, but using rugarch models.

Usage

```

xgarchf(
  y,
  h = 10L,
  model = c("eGARCH", "sGARCH", "gjrGARCH", "apARCH", "iGARCH"),
  armaOrder = c(1, 1),
  garchOrder = c(1, 1),
  distribution.model = c("norm", "std", "ged"),
  level = c(80, 95),
  ...
)

```

Arguments

<code>y</code>	numeric vector or ts
<code>h</code>	forecast horizon
<code>model</code>	mean model ("sGARCH", "eGARCH", "gjrGARCH", ...)
<code>armaOrder</code>	ARMA order for the conditional mean
<code>garchOrder</code>	GARCH order
<code>distribution.model</code>	conditional distribution
<code>level</code>	confidence level(s)
<code>...</code>	Additional parameters passed to <code>rugarch::ugarchfit</code>

Value

object of class forecast

Examples

```
## Not run:  
y <- fpp2::goog200  
fit <- xgarchf(y, h = 20, model = "eGARCH")  
plot(fit)  
  
## End(Not run)
```

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