Package 'MSGARCH'

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Description The MSGARCH package offer functionalities to fit (by Maximum Likelihood or Bayesian), simulate, and forecast various Markov-Switching GARCH processes.					
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LinkingTo Rcpp, RcppArmadillo					
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MSGARCH-package					
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Description

The R package MSGARCH aims to provide a comprehensive set of functionalities for Markov-switching GARCH processes, including fitting, filtering, forecasting, and simulating. Other functions related to Value-at-Risk, Expected-Shortfall, and conditional distributions are also available. The main functions of the package are coded in C++ with Rcpp (Eddelbuettel and Francois, 2011) and RcppArmadillo (Eddelbuettel and Sanderson, 2014). In the R package MSGARCH there is no equation for the mean as in the R package rugarch (Ghalanos, 2015). This means that we assume that before modeling, the user has filter the mean from their time series.

We provided a variety of single-regime GARCH process and regime-switching process as well as many conditional distributions. This allows for a rich modeling environment for Markov-switching GARCH model. This flexibility can come with problematic since the user must take care when optimizing since the optimizer is not always guaranteed to converge. As option, we provide the kernel method which allows the user to develop his own optimization process if none of the optimization options that the package provides is adequate.

The authors acknowledge Google for financial support via the Google Summer of Code 2016 project "MSGARCH"; see https://summerofcode.withgoogle.com/projects/#6497774455488512, the International Institute of Forecasting and Industrielle-Alliance.

References

Eddelbuettel, D. & Francois, R. (2011). Rcpp: Seamless R and C++ Integration. *Journal of Statistical Software*, 40, pp. 1-18, http://www.jstatsoft.org/v40/i08/.

Eddelbuettel, D. & Sanderson, C. (2014). RcppArmadillo: Accelerating R with High-Performance C++ Linear Algebra. *Computational Statistics & Data Analysis*, 71, pp. 1054-1063, http://dx.doi.org/10.1016/j.csda.2013.02.005.

Ghalanos, A. (2015). rugarch: Univariate GARCH Models. https://cran.r-project.org/web/packages/rugarch/.

AIC 3

AIC

Compute Akaike information criterion (AIC).

Description

Compute Akaike information criterion (AIC).

Usage

AIC(fit)

Arguments

fit

Fit object of type MSGARCH_MLE_FIT created with fit.mle or MSGARCH_BAY_FIT created with fit.bayes.

Details

If a matrix of MCMC posterior draws estimates is given, the AIC on the posterior mean is calculated.

Value

AIC value.

References

Akaike, H. (1974). A New Look at the Statistical Model Identification. *IEEE Transactions on Automatic Control*, 19, pp. 716-723.

Examples

BIC

Compute Bayesian information criterion (BIC).

Description

Compute Bayesian information criterion (BIC).

Usage

BIC(fit)

4 cdf

Arguments

Fit object of type MSGARCH_MLE_FIT created with fit.mle or MSGARCH_BAY_FIT created with fit.bayes

Details

If a matrix of MCMC posterior draws estimates is given, the BIC on the posterior mean is calculated.

Value

BIC value.

References

Schwarz, G. (1978). Estimating the dimension of a model. Annals of Statistics, 6, pp. 461-464.

Examples

cdf

Cumulative density function at T + 1.

Description

Method returning the cumulative density of a vector of points.

Usage

```
cdf(spec, x, theta, y, log = TRUE)
```

Arguments

spec	Model specification of class MSGARCH_SPEC created with create.spec.
x	Vector (of size N) of point to be evaluated.
theta	Vector (of size d) or matrix (of size M x d) of parameter estimates.
У	Vector (of size T) of observations.
log	Boolean indicating if the log cumulative is returned. (default: log = TRUE)

Details

If a matrix of parameter estimates is given, each parameter estimates is evaluated individually. The cdf method uses the last variance estimate by filtering.

create.spec 5

Value

Cumulative density or log-density of the points x (vector of size N or matrix of size M x N).

Examples

create.spec

Specification creation

Description

Function for creating a variance specification before fitting and using the R package MSGARCH functionalities.

Usage

```
create.spec(model = c("sGARCH", "sGARCH"), distribution = c("norm", "norm"),
  do.skew = c(FALSE, FALSE), do.mix = FALSE, do.shape.ind = FALSE)
```

Arguments

model	Vector (of size K) containing the variance model specifications. Valid models are "sGARCH", "eGARCH", "gjrGARCH", "tGARCH", and "GAS". (default: model = c("sGARCH", "sGARCH"))
distribution	Vector (of size K) of conditional densities. Valid distribution are "norm", "std", and "ged". The vector must be of the same length as the models vector. (default: distribution = $c("norm", "norm")$)
do.skew	Vector (of size K) of boolean indicating if the conditional density is skewed. The vector must be of the same length as the distributions vector. (default: do.skew = c(FALSE, FALSE))
do.mix	Boolean indicating if the specification is a mixture type. If the argument is TRUE, a Mixture of GARCH is created, while if the argument is FALSE, a Markov-Switching GARCH is created. (default: do.mix = FALSE)
do.shape.ind	Boolean indicating if the distribution are Regime-Independent. If the argument is TRUE, all distributions are the same and the distribution parameters does not dependent on the regime in which the distribution is attributed to. If the argument is TRUE, all distributions in the distribution argument and all skew argument

must be the same. (default: do.shape.ind = FALSE)

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Details

The Markov-Switching specification created is based on the Haas et al. (2004a) MSGARCH specification. It is a MSGARCH model that is separated in K single-regimes specification which are updated in parallel. Under this specification, the conditional variance is a function of the past data and the current state. The Mixture of GARCH option is based on the Haas et al. (2004b). A Mixture of GARCH is a mixture of distribution where the variance process of each distribution is a single-regime process.

Value

A list of class MSGARCH_SPEC containing variables related to the created specification. The list contains:

Variables:

- theta0 : Vector (of size d) of default parameters.
- is.mix: Boolean indicating if the specification is a mixture.
- is.shape.ind: Boolean indicating if the distribution parameter are regime-independent.
- K: Number of regimes.
- sigma0: Default parameters variance-covariance matrix (of size K x K) used during Bayesian esimation.
- lower: Vector (of size d) of lower parameters bound.
- upper: Vector (of size d) of upper parameters bound.
- ineqlb: Vector (of size d) of lower inequality function bound.
- inequb: Vector (of size d) of upper inequality function bound.
- n.params : Vector (of size K) of the total number of parameters by regime including distribution parameters.
- n.params.vol : Vector (of size K) of the total number of parameters by regime excluding distribuion parameters.
- do.init: Boolean indicating the default do.init argument.
- label: Vector (of size d) of parameters label.
- name : Vector (of size K) of specification name.
- func : List of R functions internaly used.
- rcpp. func: List of Rcpp functions internaly used.

The MSGARCH_SPEC class possesses these methods:

- sim: Simulation method.
- ht : Conditional variance in each regime.
- kernel: Kernel method.
- unc.vol: Unconditional variance in each regime.
- pred : Predictive density method.
- pit: Probability Integral Transform at T + 1.
- risk: Value-at-Risk And Expected-Shortfall methods.
- rnd : Simulation method at T + 1.
- pdf: Probability density function at T + 1.

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- cdf : Cumulative density function at T + 1.
- Pstate: State probabilities filtering method.
- Plast : State probabilities at T + 1.
- fit.mle: Maximum Likelihood estimation.
- fit.bayes: Bayesian estimation.

References

Bollerslev, T. (1986). Generalized Autoregressive Conditional Heteroskedasticity. *Journal of Econometrics*, 31, pp. 307-327.

Creal, D. Koopman, S. J. & Lucas, A. (2013). Generalized Autoregressive Score Models with Applications. *Journal of Applied Econometrics*, 28, pp. 777-795.

Fernandez, C. & Steel, M. F. (1998). On Bayesian Modeling of Fat Tails and Skewness. *Journal of the American Statistical Association*, 93, pp. 359-371.

Glosten, L. R. Jagannathan, R. & Runkle, D. E. (1993). On the Relation Between the Expected Value and the Volatility of the Nominal Excess Return on Stocks. *Journal of Finance*, 48, pp. 1779-1801.

Haas, M. Mittnik, S. & Paolella, M. S. (2004a). A New Approach to Markov-Switching GARCH Models. *Journal of Financial Econometrics*, 2, pp. 493-530.

Haas, M. Mittnik, S. & Paolella, M. S. (2004b). Mixed Normal Conditional Heteroskedasticity. *Journal of Financial Econometrics*, 2, pp. 211-250.

Nelson, D. B. (1991). Conditional Heteroskedasticity in Asset Returns: A New Approach. *Econometrica*, 59, pp. 347-370.

Zakoian, J.-M. (1994). Threshold Heteroskedastic Models Journal of Economic. *Dynamics and Control*, 18, pp. 931-955.

Examples

DIC

Compute Deviance Information Criterion (DIC).

Description

Compute Deviance Information Criterion (DIC).

Usage

DIC(fit)

Arguments

fit

Fit object of type MSGARCH_BAY_FIT created with fit.bayes.

fit.bayes

Value

A list containing six variables:

- DIC: Deviance Information Criterion.
- IC: Bayesian Predictive Information Criterion.
- pD : Effective number of parameters (pD = Dbar Dhat)
- pV : Effective number of parameters (pV = var(D)/2)
- D.bar: Expected value of the deviance over the posterior
- D. hat: Deviance at the mean posterior estimate

References

Gelman, A. Carlin, J. B. Stern, H. S. & Rubin, D. B. (2003). Bayesian Data Analysis. *Chapman and Hall/CRC*

Examples

fit.bayes

Bayesian estimation.

Description

Method that performs Bayesian estimation of a MSGARCH_SPEC object on a set of observations.

Usage

```
fit.bayes(spec, y, ctr = list())
```

Arguments

spec Model specification of class MSGARCH_SPEC created with create.spec.

y Vector (of size T) of observations.

ctr A list of control parameters.

The control parameters have three components:

• N. burn (integer >= 0): Number of discarded draws. (default: N. burn = 1000)

- N.mcmc (integer > 0): Number of draws. (default: N.mcmc = 5000)
- N. thin (integer > 0): Thinning factor (every N. thin draws are kept). (default: N. thin = 10)

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Details

The total number of draws is equal to N.burn + N.mcmc. The Bayesian estimation uses the R package adaptMCMC (Andreas, 2012) which implements the adaptive sampler of Vihola (2012).

Value

A list of class MSGARCH_BAY_FIT containing four variables:

- theta: The Bayesian chain (matrix from the R package coda (Plummer et al., 2006) of size N.mcmc / N.thin).
- accept : Acceptation rate of the sampler.
- y : Initial vector (of size T) of observations.
- spec : Initial specification.

The MSGARCH_BAY_FIT possess this method:

- AIC: Compute Akaike information criterion (AIC).
- BIC: Compute Bayesian information criterion (BIC).
- DIC: Compute Deviance Information Criterion (DIC).

References

Andreas, S. (2012). adaptMCMC: Implementation of a Generic Adaptive Monte Carlo Markov Chain Sampler. https://cran.r-project.org/web/packages/adaptMCMC/.

Metropolis, N.; Rosenbluth, A. W.; Rosenbluth, M. N.; Teller, A. H. & Teller, E. (1953). Equation of State Calculations by Fast Computing Machines. *Journal of Chemical Physics*, 21, pp. 1087-1092.

Plummer, M. Best, N. Cowles, K. & Vines, K. (2006). CODA: Convergence Diagnosis and Output Analysis for MCMC. *R News*, 6, pp.7-11. https://cran.r-project.org/web/packages/coda/.

Vihola, M. (2012). Robust Adaptive Metropolis Algorithm with Coerced Acceptance Rate. *Statistics and Computing*, 22, pp. 997-1008.

10 fit.mle

fit.mle

ML estimation.

Description

Method that performs Maximum Likelihood estimation of a MSGARCH_SPEC object on a set of observations.

Usage

```
fit.mle(spec, y, ctr = list())
```

Arguments

spec Model specification created with create.spec.

y Vector (of size T) of observations.

ctr List of control parameters. The control parameters have two components to it:

- theta0: Starting parameters (vector of size d). If no starting parameters is provided, the default starting parameters of the specification are used.
- do.init: Boolean indicating if there is a pre-optimization with the R package DEoptim (Ardia et al., 2011). (default: do.init = FALSE)

Details

The Maximum likelihood estimation uses the R package nloptr (Johnson, 2014) for main optimizer while it uses the R package DEoptim when do.init = TRUE.

Value

A list of class MSGARCH_MLE_FIT containing five variables:

- theta: Optimal parameters (vector of size d).
- 1_likelihood: log-likelihood of y given the optimal parameters.
- spec : Initial specification.
- is.init: Indicating if estimation was made with do.init option.
- y : Initial vector (of size T) of observations..

The MSGARCH_MLE_FIT possess these methods:

- AIC: Compute Akaike information criterion (AIC).
- BIC: Compute Bayesian information criterion (BIC).

References

Ardia, D.; Mullen, K. M.; Peterson, B. G. & Ulrich, J. (2015). DEoptim: Differential Evolution in R. https://cran.r-project.org/web/packages/DEoptim/.

Johnson, S. G. (2014). The NLopt Nonlinear-Optimization. https://cran.r-project.org/web/packages/NLopt/.

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Examples

ht

Conditional variance in each regime.

Description

Method returning the conditional variance of each regime.

Usage

```
ht(spec, theta, y)
```

Arguments

spec	$Model\ specification\ of\ class\ {\tt MSGARCH_SPEC}\ created\ with\ {\tt create.spec}.$
theta	Vector (of size d) or matrix (of size M x d) of parameter estimates.
у	Vector (of size T) of observations.

Details

If a matrix of parameter estimates is given, each parameter estimates is evaluated individually.

Value

Condititional variance time serie (array of size T + 1 x M x K) for each regime.

12 kernel

kernel	Kernel function.	
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Description

Method returning the kernel value of a vector of observations.

Usage

```
kernel(spec, theta, y, log = TRUE)
```

Arguments

spec	$Model\ specification\ of\ class\ {\tt MSGARCH_SPEC}\ created\ with\ {\tt create.spec}.$
theta	Vector (of size d) or matrix (of size M x d) of parameter estimates.
у	Vector (of size T) of observations.
log	Boolean indicating if the log kernel is returned. (default: log = TRUE)

Details

If a matrix of parameter estimates is given, each parameter estimates is evaluated individually. The kernel is a combination of the prior and the likelihood function. The kernel is equal to $prior(\theta) + L(y|\theta)$ where L is the likelihood of y given the parameter θ . When doing optimization, the goal is to minimize the negative log-kernel.

• Details on the prior

The prior is different for each specification. It ensures that the θ makes the conditional variance process stationary, positive, and that it respect that the sums of the probabilities in the case of a multiple-regime models are all equal to 1. If any of these three conditions is not respected the prior return -1e10, meaning that the optimizer or sampler will know that θ is not a good candidate.

Value

Kernel or log-kernel value (scalar or vector of size M) of the vector of observations.

References

Hamilton, J. D. (1989) A New Approach to the Economic Analysis of Nonstationary Time Series and the Business Cycle. *Econometrica*, 57, pp.357-38

pdf 13

pdf	Probability density function at $T + 1$.	
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Description

Method returning the probability density of a vector of points.

Usage

```
pdf(spec, x, theta, y, log = TRUE)
```

Arguments

spec	$Model\ specification\ of\ class\ {\tt MSGARCH_SPEC}\ created\ with\ {\tt create.spec}.$
Х	Vector (of size N) of point to be evaluated
theta	Vector (of size d) or matrix (of size M x d) of parameter estimates.
у	Vector (of size T) of observations.
log	Boolean indicating if the log-density is returned. (default: log = TRUE)

Details

If a matrix of parameter estimates is given, each parameter estimates is evaluated individually. The pdf method uses the last variance estimate by filtering.

Value

Probability density or log-density of the points x (vector of size N or matrix of size M x N).

14 pit

Description

Method returning the predictive Probability integral transform (PIT).

Usage

```
pit(spec, x, theta, y, do.norm = FALSE)
```

Arguments

spec	Model specification of class MSGARCH_SPEC created with create.spec.
x	Vector (of size N) of point to be evaluated
theta	Vector (of size d) or matrix (of size M x d) of parameter estimates.
У	Vector (of size T) of observations.
do.norm	Boolean indicating if the PIT value are transforms into standard Normal variate. (do.norm = FALSE)

Details

If a matrix of MCMC posterior draws estimates is given, the Bayesian Probability integral transform is calculated. The do.norm argument transforms the PIT value into Normal variate so that normality test can be done.

Value

Probability integral transform of the points x or Normal variate derived from the Probability integral transform of x (vector of size N).

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Plast

State probabilities at T + 1.

Description

Method returning the state probabilities at T + 1.

Usage

```
Plast(spec, theta, y)
```

Arguments

spec Model specification of class MSGARCH_SPEC created with create.spec.
theta Vector (of size d) or matrix (of size M x d) of parameter estimates.

y Vector (of size T) of observations.

Details

If a matrix of parameter estimates is given, each parameter estimates is evaluated individually.

Value

State probabilities at T + 1 (matrix of size M x K).

Examples

pred

Predictive density function.

Description

Method returning the predictive probability density of a vector of points.

Usage

```
pred(spec, x, theta, y, log = TRUE)
```

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Arguments

spec	Model specification of class MSGARCH_SPEC created with create.spec.
Х	Vector (of size N) of point to be evaluated
theta	Vector (of size d) or matrix (of size M x d) of parameter estimates.
у	Vector (of size T) of observations.
log	Boolean indicating if the log-density is returned. (default: log = TRUE)

Details

If a matrix of MCMC posterior draws estimates is given, the Bayesian predictive density is calculated.

Value

Predictive density or log-density of x (vector of size N).

Examples

Pstate

State probabilities filtering function.

Description

Method returning the filtered state probabilities.

Usage

```
Pstate(spec, theta, y)
```

Arguments

spec	Model specification of class MSGARCH_SPEC created with create.spec.
theta	Vector (of size d) or matrix (of size M x d) of parameter estimates.
У	Vector (of size T) of observations.

Details

If a matrix of parameter estimates is given, each parameter estimates is evaluated individually.

risk 17

Value

Filtered state probabilities (array of size T x M x K).

Examples

risk

Value-at-Risk And Expected-shortfall functions.

Description

Method returning the Value-at-Risk and Expected-shortfall.

Usage

```
risk(spec, theta, y, level = c(0.95, 0.99), ES = TRUE)
```

Arguments

spec	Model specification of class MSGARCH_SPEC created with create, spec.
theta	Vector (of size d) or matrix (of size M x d) of parameter estimates.
У	Vector (of size T) of observations.
level	Vector (of size A) of Value-at-risk and Expected-shortfall levels. (default: level = c(0.95, 0.99))
ES	Boolean indicating if Expected-shortfall is also calculated. (default: ES = TRUE)

Details

If a matrix of MCMC posterior draws estimates is given, the Bayesian Value-at-Risk and Expected-shortfall are calculated.

Value

A list containing one or two components:

- VaR: Value-at-Risk at the choosen levels (vector of size A).
- ES: Expected-shortfall at the choosen levels (vector of size A).

18 rnd

Description

Method returning random draws at T + 1.

Usage

```
rnd(spec, n, theta, y = vector("double", 0), do.state = FALSE)
```

Arguments

spec	Model specification of class MSGARCH_SPEC created with create.spec.
n	Number of random draws to be generated.
theta	Vector (of size d) or matrix (of size M x d) of parameter estimates.
у	Vector (of size T) of observations.
do.state	Boolean indicating if the simulated state are also output. (default: do.state = FALSE)

Details

If a matrix of parameter estimates is given, each parameter estimates is evaluated individually.

Value

A list containing one or two components:

- draws: vector (of size n) or matrix (of size M x n) of simulated draws at T + 1.
- state: vector (of size n) or matrix (of size $M \times n$) of simulated states at T + 1. The state value appear only if do.state = TRUE.

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sim Simulation function.	sim	Simulation function.	
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Description

Method returning a simulated process.

Usage

```
sim(spec, n, theta, burnin = 500, do.state = FALSE)
```

Arguments

spec	Model specification of class MSGARCH_SPEC created with create.spec.
n	Simulation length.
theta	Vector (of size d) or matrix (of size M x d) of parameter estimates.
burnin	(integer \geq 0) Burnin period discarded (first simulation draws). (default: burnin = 500)
do.state	Boolean indicating if the simulated state are also output. (default: log = TRUE)

Details

If a matrix of parameter estimates is given, each parameter estimates is evaluated individually.

Value

A list of class MSGARCH_SIM containing one or two components.

- draws: vector (of size n) or matrix (of size M x n) of simulated draws.
- state: vector (of size n) or matrix (of size M x n) of simulated states. The state value appear only if do.state = TRUE.

20 unc.vol

sp500ret

Standard and poors 500 closing Value log return

Description

The S&P500 index closing value log return from 1987-03-10 to 2009-01-30 from Yahoo Finance https://ca.finance.yahoo.com/.

Usage

```
data("sp500ret")
```

Format

Vector containing 5523 observations.

Source

Yahoo Finance https://ca.finance.yahoo.com/

unc.vol

Unconditional variance of each regime.

Description

Method returning the unconditional variance of the process in each state.

Usage

```
unc.vol(spec, theta)
```

Arguments

spec Model specification of class MSGARCH_SPEC created with create.spec. theta Vector (of size d) or matrix (of size M x d) of parameter estimates.

Details

If a matrix of parameter estimates is given, each parameter estimates is evaluated individually.

Value

Unconditional variance (vector of size K or matrix of size M x K) of each regime.

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