

Package ‘ldhmm’

April 13, 2017

Type Package

Title Hidden Markov Model for Return Time-Series Based on Lambda Distribution

Version 0.1.0

Date 2017-04-13

Author Stephen H-T. Lihn [aut, cre]

Maintainer Stephen H-T. Lihn <stevelihn@gmail.com>

Description Hidden Markov Model (HMM) based on symmetric lambda distribution framework is implemented for the study of return time-series in the financial market. Major features in the S&P500 index, such as regime identification, volatility clustering, and anti-correlation between return and volatility, can be extracted from HMM cleanly. Univariate symmetric lambda distribution is essentially a location-scale family of power-exponential distribution. Such distribution is suitable for describing highly leptokurtic time series obtained from the financial market. It provides a theoretically solid foundation to explore such data where the normal distribution is not adequate. The HMM implementation follows closely the book: “Hidden Markov Models for Time Series”, by Zucchini, MacDonald, Langrock (2016).

Depends R (>= 3.3.1)

Imports stats, utils, ecd, xts, zoo, moments, parallel, graphics, methods

Suggests knitr, testthat, depmixS4, roxygen2, scales, shape

License Artistic-2.0

Encoding latin1

LazyData true

RoxygenNote 5.0.1

Collate 'ldhmm-calc_stats_from_obs.R' 'ldhmm-numericOrNull-class.R'
'ldhmm-package.R' 'ldhmm-class.R' 'ldhmm-conditional_prob.R'
'ldhmm-constructor.R' 'ldhmm-decoding.R'
'ldhmm-forecast_prob.R' 'ldhmm-forecast_state.R'
'ldhmm-ld_stats.R' 'ldhmm-log_forward.R' 'ldhmm-mle.R'
'ldhmm-mlk.R' 'ldhmm-n2w.R' 'ldhmm-pseudo_residuals.R'
'ldhmm-state_ld.R' 'ldhmm-state_pdf.R' 'ldhmm-ts_abs_acf.R'
'ldhmm-viterbi.R' 'ldhmm-w2n.R' 'ldhmm.ts_log_rtn.R'

NeedsCompilation no

R topics documented:

ldhmm-package	2
ldhmm	3
ldhmm-class	4
ldhmm.calc_stats_from_obs	4
ldhmm.conditional_prob	5
ldhmm.decoding	6
ldhmm.forecast_prob	6
ldhmm.forecast_state	7
ldhmm.ld_stats	7
ldhmm.log_forward	8
ldhmm.mle	8
ldhmm.mllk	9
ldhmm.n2w	10
ldhmm.pseudo_residuals	10
ldhmm.state_ld	11
ldhmm.state_pdf	12
ldhmm.ts_abs_acf	12
ldhmm.ts_log_rtn	13
ldhmm.viterbi	13
ldhmm.w2n	14
numericOrNull-class	14

Index	15
--------------	-----------

ldhmm-package	<i>ldhmm: A package for HMM using lambda distribution.</i>
---------------	--

Description

The ldhmm package provides the core class and functions to calculate Hidden Markov Model (HMM) using lambda distribution framework. The main goal is to provide a theoretically solid foundation to explore the return time-series in the financial market, where the normal distribution is not adequate due to the leptokurtic nature of the data. Major features in the S&P 500 index, such as regime identification, volatility clustering, and anti-correlation between return and volatility, can be extracted from HMM cleanly. Univariate symmetric lambda distribution is essentially a location-scale family of power-exponential distribution. Such distribution is suitable for describing highly leptokurtic time series obtained from the financial market.

Details

The main change compared to a normal-distribution based HMM is to add the third paramter lambda to describe the kurtosis level of the distribution. When lambda is one, the model converges back to a normal-distribution based HMM (e.g. using depmixS4 package). The ability to optimize kurtosis brings the model output to be more consistent with the data. In particular, for daily data, the level of kurtosis is quite high. This puts the normal distribution in great disadvantage. This problem is solved by using the lambda distribution.

Author(s)

Stephen H-T. Lihn

References

Walter Zucchini, Iain L. MacDonald, Roland Langrock (2016). "Hidden Markov Models for Time Series, An Introduction Using R." Second Edition. CRC Press.

ldhmm	<i>Constructor of ldhmm class</i>
-------	-----------------------------------

Description

Construct an ldhmm class by providing the required parameters.

Usage

```
ldhmm(m, param, gamma, delta = NULL, stationary = TRUE)
```

Arguments

m	numeric, number of states
param	matrix, the ecd parameters of states.
gamma	matrix, the transition probability matrix, must be m by m.
delta	numeric, the initial distribution for each state, default is NULL.
stationary	logical, specify whether the initial distribution is stationary or not, default is TRUE.

Value

An object of ldhmm class

Author(s)

Stephen H. Lihn

Examples

```
param0 <- matrix(c(0.003, 0.02, 1, -0.006, 0.03, 1.3), 2, 3, byrow=TRUE)
gamma0 <- matrix(c(0.9, 0.1, 0.1, 0.9), 2, 2, byrow=TRUE)
d <- ldhmm(m=2, param=param0, gamma=gamma0)
```

ldhmm-class

*The ldhmm class***Description**

This S4 class is the major object class for ldhmm package

Slots

call The match.call slot

m numeric, length 1, number of states

param.nbr numeric, number of parameters (2 or 3) for each ecld object

param matrix, natural parameters for ecld objects, size of states times param.nbr. Each row can be 2-parameter sequences, or 3-parameter sequences. Three-parameter unit (μ , σ , λ) forms an ecld object representing a leptokurtic symmetric lambda distribution. On the other hand, to provide compatibility to a normal distribution HMM, two-parameter unit (μ , σ) forms an ecld object with $\lambda=1$.

gamma matrix, the transition probability matrix, must be m by m.

delta numeric, the initial distribution for each state, default is NULL.

stationary logical, specify whether the initial distribution is stationary or not, default is TRUE.

mle.optimizer character, the MLE optimizer. Currently it is just set to "nlm".

return.code numeric, the return code from the MLE optimizer.

iterations numeric, number of iterations MLE optimizer takes.

mllk numeric, the final mllk value.

AIC numeric, the final AIC.

BIC numeric, the final BIC.

observations numeric, stores the observations post optimization

states.prob matrix, stores the state probabilities post optimization

states.local numeric, stores the local decoding states post optimization

states.global numeric, stores the global decoding states post optimization (Viterbi)

states.local.stats matrix, stores the statistics of local states post optimization

states.global.stats matrix, stores the statistics of global states post optimization

ldhmm.calc_stats_from_obs

*Computing the statistics for each state***Description**

This utility computes the statistics (mean, sd, kurtosis, length) for each state. It can be based on the local or global decoding result. The concept of asymptotic statistics can be applied by which the largest N observations (in absolute term) can be dropped to avoid distortion from outliers. It is assumed the object already has come with filled data in observations, states.prob, states.local, states.global slots.

Usage

```
ldhmm.calc_stats_from_obs(object, drop = 0, use.local = TRUE)
```

Arguments

object	an ldhmm object
drop	numeric, an integer to drop the largest N observations, default is zero.
use.local	logical, use local decoding result, default is TRUE. Otherwise, use global decoding result.

Value

an ldhmm object containing results of decoding

Author(s)

Stephen H. Lihn

ldhmm.conditional_prob

Computing the conditional probabilities

Description

This utility computes the conditional probabilities that observation at time t equals xc, given all observations other than that at time t being the same.

Usage

```
ldhmm.conditional_prob(object, x, xc)
```

Arguments

object	an ldhmm object
x	numeric, the observations.
xc	numeric, the conditional observations.

Value

matrix of probabilities, size of xc times size of x.

Author(s)

Stephen H. Lihn

ldhmm.decoding	<i>Computing the minus log-likelihood (MLLK)</i>
----------------	--

Description

This utility computes the state probabilities, uses local and global decoding to calculate the states. The results are saved to the returned ldhmm object.

Usage

```
ldhmm.decoding(object, x)
```

Arguments

object	an ldhmm object
x	numeric, the observations.

Value

an ldhmm object containing results of decoding

Author(s)

Stephen H. Lihn

ldhmm.forecast_prob	<i>Computing the forecast probability distribution</i>
---------------------	--

Description

This utility computes the forecast probability distribution (Zucchini, 5.3)

Usage

```
ldhmm.forecast_prob(object, x, xf, h = 1)
```

Arguments

object	an ldhmm object
x	numeric, the observations.
xf	numeric, the future observations to be forecasted.
h	integer, time steps to forecast.

Value

matrix of probabilities, size of h times size of xf.

Author(s)

Stephen H. Lihn

`ldhmm.forecast_state` *Computing the state forecast*

Description

This utility computes the state forecast.

Usage

```
ldhmm.forecast_state(object, x, h = 1)
```

Arguments

<code>object</code>	an ldhmm object
<code>x</code>	numeric, the observations.
<code>h</code>	integer, time steps to forecast.

Value

matrix of probabilities per state (even if $h=1$), number of states times size of h

Author(s)

Stephen H. Lihn

`ldhmm.ld_stats` *Computes the theoretical statistics per state*

Description

This utility computes the statistics (mean, sd, kurtosis) based on the lambda distribution. This is used to compare to the statistics from observations for each state.

Usage

```
ldhmm.ld_stats(object)
```

Arguments

<code>object</code>	an ldhmm object
---------------------	-----------------

Value

a matrix of statistics for each state, size of states times 3

Author(s)

Stephen H. Lihn

ldhmm.log_forward	<i>Computing the log forward and backward probabilities</i>
-------------------	---

Description

This utility computes the logarithms of the forward and backward probabilities, aka alpha and beta. The logarithm keeps the computation away from floating point under/over-flow. (Zucchini, 5.4)

Usage

```
ldhmm.log_forward(object, x)
```

```
ldhmm.log_backward(object, x)
```

Arguments

object	an ldhmm object
x	numeric, the observations.

Value

numeric, the log probabilities

Author(s)

Stephen H. Lihn

ldhmm.mle	<i>Computing the MLEs</i>
-----------	---------------------------

Description

Computing the MLEs using nlm package

Usage

```
ldhmm.mle(object, x, print.level = 0, iterlim = 1000, ...)
```

Arguments

object	an ldhmm object that can supply m, param.nbr and stationary.
x	numeric, the observations.
print.level	numeric, this argument determines the level of printing which is done during the minimization process. The default value of 0 means that no printing occurs, a value of 1 means that initial and final details are printed and a value of 2 means that full tracing information is printed.
iterlim	numeric, a positive integer specifying the maximum number of iterations to be performed before the program is terminated.
...	additional parameters passed to the MLE optimizer

Value

an ldhmm object containing results of MLE optimization

Author(s)

Stephen H. Lihn

Examples

```
## Not run:
param0 <- matrix(c(0.003, 0.02, 1, -0.006, 0.03, 1.3), 2, 3, byrow=TRUE)
gamma0 <- matrix(c(0.9, 0.1, 0.1, 0.9), 2, 2, byrow=TRUE)
h <- ldhmm(m=2, param=param0, gamma=gamma0)
spx <- ldhmm.ts_log_rtn()
ldhmm.mle(h, spx$x)

## End(Not run)
```

ldhmm.mllk

Computing the minus log-likelihood (MLLK)

Description

This utility computes the MLLK. It is typically invoked by the MLE optimizer. (Zucchini, 3.2)

Usage

```
ldhmm.mllk(par.vector, x, ldhmm, mllk.print.level = 0)
```

Arguments

par.vector	numeric, the working parameter vector
x	numeric, the observations.
ldhmm	an input ldhmm object to provide static reference, such as m, param.nbr, stationary.
mllk.print.level	numeric, this argument determines the level of printing which is done during the minimization process. The default value of 0 means that no printing occurs, a value of 1 or greater means some tracing information is printed.

Value

an ldhmm object containing results of MLE optimization

Author(s)

Stephen H. Lihn

ldhmm.n2w

Transforming natural parameters to a linear working parameter array

Description

This utility linearizes the natural parameters and transforms the constrained parameters to unconstrained parameters. (Zucchini, 3.3.1)

Usage

```
ldhmm.n2w(object)
```

Arguments

object an ldhmm object

Value

numeric, linear working parameter array

Author(s)

Stephen H. Lihn

Examples

```
param0 <- matrix(c(0.003, 0.02, 1, -0.006, 0.03, 1.3), 2, 3, byrow=TRUE)
gamma0 <- matrix(c(0.9, 0.1, 0.1, 0.9), 2, 2, byrow=TRUE)
d <- ldhmm(m=2, param=param0, gamma=gamma0)
v <- ldhmm.n2w(d)
```

ldhmm.pseudo_residuals

Computing pseudo-residuals

Description

This utility computes pseudo-residuals. (Zucchini, 6.2)

Usage

```
ldhmm.pseudo_residuals(object, x, xc.length = 1000)
```

Arguments

object an ldhmm object
x numeric, the observations.
xc.length a positive integer specifying the length of xc when calculating conditional probabilities, default is 1000.

Value

a vector of normal quantiles

Author(s)

Stephen H. Lihn

Examples

```
## Not run:
sr <- ldhmm.pseudo_residuals(object, x)
hist(sr)
acf(sr)
qqnorm(sr, cex=0.5)
L <- seq(-3,3,length.out=100)
lines(L,L,col="red",lwd=2, lty=2)

## End(Not run)
```

ldhmm.state_ld

Constructing the ecld objects per state

Description

This utility constructs the ecld objects per state and return them in a list of easy query.

Usage

```
ldhmm.state_ld(object, state = NULL)
```

Arguments

object	an ldhmm object
state	numeric, the states.

Value

a list of ecld objects

Author(s)

Stephen H. Lihn

ldhmm.state_pdf

Computing the PDF per state given the observations

Description

Computing the PDF per state given the observations. Only one of state or x can be a vector per call.

Usage

```
ldhmm.state_pdf(object, state, x)
```

Arguments

object	an ldhmm object
state	numeric, the states.
x	numeric, the observations.

Value

a vector or matrix of PDF. The dimension of matrix is state times x

Author(s)

Stephen H. Lihn

ldhmm.ts_abs_acf

Computing ACF of the absolute value of a time series

Description

This utility computes the ACF of the absolute value of a time series as a proxy of the auto-correlation of the volatility. It allows to drop the largest N outliers so that they would not skew the ACF calculation.

Usage

```
ldhmm.ts_abs_acf(x, drop = 0, lag.max = 100)
```

Arguments

x	numeric, the observations.
drop	a positive integer, specifying number of outliers to be dropped.
lag.max	a positive integer, specifying number of lags to be computed.

Value

a vector of ACF

Author(s)

Stephen H. Lihn

ldhmm.ts_log_rtn	<i>Get log-returns from historic prices of an index</i>
------------------	---

Description

This utility returns the dates and log-returns of an index available in ecd package.

Usage

```
ldhmm.ts_log_rtn(symbol = "spx", start.date = "1950-01-01",
  end.date = "2015-12-31", on = "weeks")
```

Arguments

symbol	character, specify the symbol of the index, default is spx.
start.date, end.date	character, specify the date range in ISO-format, default is from 1950-01-01 to 2016-12-31.
on	character, specify the interval, days, weeks, months. Default is weeks.

Value

list of three vectors: d is the dates and x is log-returns and p is prices

Author(s)

Stephen H. Lihn

Examples

```
a <- ldhmm.ts_log_rtn()
```

ldhmm.viterbi	<i>Computing the global decoding by the Viterbi algorithm</i>
---------------	---

Description

This utility computes the global decoding by the Viterbi algorithm.

Usage

```
ldhmm.viterbi(object, x)
```

Arguments

object	an ldhmm object
x	numeric, the observations.

Value

a vector of states

Author(s)

Stephen H. Lihn

`ldhmm.w2n`

Transforming working parameter array to natural parameters

Description

This utility transforms the working parameter array back to the vectors and matrix of the constrained parameters. (Zucchini, 3.3.1)

Usage

```
ldhmm.w2n(object, par.vector)
```

Arguments

<code>object</code>	an <code>ldhmm</code> object that can supply <code>m</code> , <code>param.nbr</code> and <code>stationary</code> .
<code>par.vector</code>	numeric, linear working parameter array. See <code>ldhmm.n2w</code> .

Value

an `ldhmm` object

Author(s)

Stephen H. Lihn

`numericOrNull-class`

The numericOrNull class

Description

The S4 class union of numeric and NULL, primarily used for delta

Index

- *Topic **acf**
 - ldhmm.ts_abs_acf, [12](#)
- *Topic **class**
 - ldhmm-class, [4](#)
- *Topic **constructor**
 - ldhmm, [3](#)
 - ldhmm-class, [4](#)
- *Topic **data**
 - ldhmm.ts_log_rtn, [13](#)
- *Topic **forecast**
 - ldhmm.forecast_prob, [6](#)
 - ldhmm.forecast_state, [7](#)
- *Topic **mle**
 - ldhmm.mle, [8](#)
- *Topic **mllk**
 - ldhmm.calc_stats_from_obs, [4](#)
 - ldhmm.decoding, [6](#)
 - ldhmm.mllk, [9](#)
- *Topic **parameter**
 - ldhmm.n2w, [10](#)
 - ldhmm.w2n, [14](#)
- *Topic **pdf**
 - ldhmm.conditional_prob, [5](#)
 - ldhmm.ld_stats, [7](#)
 - ldhmm.log_forward, [8](#)
 - ldhmm.state_ld, [11](#)
 - ldhmm.state_pdf, [12](#)
- *Topic **residuals**
 - ldhmm.pseudo_residuals, [10](#)
- *Topic **viterbi**
 - ldhmm.viterbi, [13](#)

- ldhmm, [3](#)
- ldhmm-class, [4](#)
- ldhmm-package, [2](#)
- ldhmm.calc_stats_from_obs, [4](#)
- ldhmm.conditional_prob, [5](#)
- ldhmm.decoding, [6](#)
- ldhmm.forecast_prob, [6](#)
- ldhmm.forecast_state, [7](#)
- ldhmm.ld_stats, [7](#)
- ldhmm.log_backward (ldhmm.log_forward),
[8](#)
- ldhmm.log_forward, [8](#)
- ldhmm.mle, [8](#)
- ldhmm.mllk, [9](#)
- ldhmm.n2w, [10](#)
- ldhmm.pseudo_residuals, [10](#)
- ldhmm.state_ld, [11](#)
- ldhmm.state_pdf, [12](#)
- ldhmm.ts_abs_acf, [12](#)
- ldhmm.ts_log_rtn, [13](#)
- ldhmm.viterbi, [13](#)
- ldhmm.w2n, [14](#)
- numericOrNull-class, [14](#)