

# Package ‘HMMCont’

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**Type** Package

**Title** Hidden Markov Model for Continuous Observations Processes

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**Description** The package includes the functions designed to analyse continuous observations processes with the Hidden Markov Model approach. They include Baum-Welch and Viterbi algorithms and additional visualisation functions. The observations are assumed to have Gaussian distribution and to be weakly stationary processes. The package was created for analyses of financial time series, but can also be applied to any continuous observations processes.

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**Description**

The package includes the functions designed to analyse continuous observations processes with the Hidden Markov Model approach. They include Baum-Welch and Viterbi algorithms and additional visualisation functions. The observations are assumed to have Gaussian distribution and to be weakly stationary processes. The package was created for analyses of financial time series, but can also be applied to any continuous observations processes.

**Details**

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License: GPL 3

A Hidden Markov Model (HMM) is a statistical model in which the process being modelled is assumed to be a Markov process with unobserved, i.e. hidden states. This unobserved Markov process can be revealed from an observable process that is dependent on the states of the underlying Markov process. The HMMCont package compiles the functions that can analyse the continuous observable processes (i.e. continuous in space, discrete in time) and identify the underlying two-states Markov processes. The observable process should be weakly stationary (e.g. in case of financial time series the returns, but not the prices should be analysed). The state-dependent probabilities of the observations are modelled with Gaussian probability density functions (Rabiner, 1989). The implemented analysis procedure includes: (i) setting the initial model parameters and loading the data (function `hmmsetcont`), repeated execution of the Baum-Welch algorithm (function `baumwelchcont`), and execution of the Viterbi algorithm (`viterbicont`). The function `baumwelchcont` allows to control the model parameters after each Baum-Welch iteration, and accumulates the information on the model evolution. The model object can be analysed with tailored `print`, `summary`, and `plot` functions (S3 methods). For details on HMMs see the publications by Viterbi (1967), Baum et al (1970), and Rabiner (1989).

**Author(s)**

Mikhail A. Beketov

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**References**

Baum, L.E., Petrie, T., Soules, G., and Weiss, N. 1970. A maximization technique occurring in the statistical analysis of probabilistic functions of Markov chains. *The Annals of Mathematical Statistics*. 41: 164-171.



**Value**

An object of the class `ContObservHMM` (see section on the function `hmmsetcont`). After sufficient number of iterations the object can be used to derive the Markov states sequence by the Viterbi algorithm (function `viterbicont`). The object can be analysed with the class-specific functions `print`, `summary`, and `plot`.

**Author(s)**

Mikhail A. Beketov

**References**

Baum, L.E., Petrie, T., Soules, G., and Weiss, N. 1970. A maximization technique occurring in the statistical analysis of probabilistic functions of Markov chains. *The Annals of Mathematical Statistics*. 41: 164-171.

Rabiner, L.R. 1989. A tutorial on hidden Markov models and selected applications in speech recognition. *Proceedings of the IEEE*. 77: 257-286.

**See Also**

Functions: [hmmsetcont](#), [viterbicont](#), and [statesDistributionsPlot](#).

**Examples**

```
Returns<-logreturns(Prices) # Getting a stationary process
Returns<-Returns*10 # Scaling the values
hmm<-hmmsetcont>Returns) # Creating a HMM object
print(hmm) # Checking the initial parameters

hmm<-baumwelchcont(hmm) # First iteration
print(hmm) # Inspecting

for(i in 1:5){hmm<-baumwelchcont(hmm)} # Subsequent iterations
print(hmm) # Inspecting

hmmcomplete<-viterbicont(hmm) # Viterbi execution
par(mfrow=c(2,1))
plot(hmmcomplete, Prices, ylabel="Price")
plot(hmmcomplete, ylabel="Returns") # the revealed
# Markov chain and the observations are plotted
```

**Description**

The function simulates (i) two observation processes that correspond to the distributions of the two states in an HMM, (ii) the underlying Markov process, and (iii) an observation process that correspond to an HMM in terms of both the underlying Markov and observations processes. The function uses the last-iteration parameters, when the HMM-object was modified by function `baumwelchcont` previously.

**Usage**

```
hmmcontSimul(hmm, n)
```

**Arguments**

<code>hmm</code>	An object of the class <code>ContObservHMM</code>
<code>n</code>	Number of observations to be simulated.

**Value**

The function returns an object of the class `SimulContHMM` that is a list comprising the two observations processes (each corresponds to the distribution of one of the two states), the Markov chain (i.e. the underlying process consisting of two states), and the observation process that correspond to that Markov chain and the two distributions. Hence, the latter is the process simulated according a given HMM.

**Author(s)**

Mikhail A. Beketov

**See Also**

Functions: [hmmsetcont](#), [baumwelchcont](#), and [viterbicont](#).

**Examples**

```
Returns<-(logreturns(Prices))*10
hmm<-hmmsetcont>Returns)
for(i in 1:6){hmm<-baumwelchcont(hmm)}
hmmcomplete<-viterbicont(hmm)

sim<-hmmcontSimul(hmmcomplete, n=100) # simulating the processes

plot(sim$StateProcess1, type="l", ylab="State 1 Process")
plot(sim$StateProcess2, type="l", ylab="State 2 Process")
plot(sim$MarkovChain, type="l", ylab="Markov chain")
plot(sim$SimulatedObservation, type="l", ylab="Full HMM Process")
```

hmmsetcont

*Setting an initial HMM object***Description**

The function sets an initial Hidden Markov Model object with initial set of model parameters. It returns the object of class ContObservHMM that can be analysed with Baum-Welch (function baumwelchcont) and Viterbi algorithms (viterbicont).

**Usage**

```
hmmsetcont(Observations, Pi1 = 0.5, Pi2 = 0.5, A11 = 0.7, A12 = 0.3,
A21 = 0.3, A22 = 0.7, Mu1 = 5, Mu2 = (-5), Var1 = 10, Var2 = 10)
```

```
## S3 method for class 'ContObservHMM'
print(x, ...)
## S3 method for class 'ContObservHMM'
summary(object, ...)
## S3 method for class 'ContObservHMM'
plot(x, Series=x$Observations,
ylabel="Observation series", xlabel="Time", ...)
```

**Arguments**

Observations	Vector of observations (class "numeric"), a weakly stationary process (e.g. returns time series).
Pi1	Initial probability of state 1.
Pi2	Initial probability of state 2.
A11	Initial transition probability from state 1 to state 1.
A12	Initial transition probability from state 1 to state 2.
A21	Initial transition probability from state 2 to state 1.
A22	Initial transition probability from state 2 to state 2.
Mu1	Initial mean for Gaussian PDF for state 1.
Mu2	Initial mean for Gaussian PDF for state 2.
Var1	Initial variance for Gaussian PDF for state 1.
Var2	Initial variance for Gaussian PDF for state 2.
x	An object returned by the function hmmsetcont.
object	An object returned by the function hmmsetcont.
Series	Observations time series to be plotted along the Markov states.
ylabel	Y axis label.
xlabel	X axis label.
...	Not used.

**Value**

The function returns an object of the class `ContObservHMM` that is a list comprising the observations, tables accumulating the model parameters and results after each Baum-Welch iterations (i.e. after each execution of the function `baumwelchcont`), table for the state sequence derived by the Viterbi algorithm (function `viterbicont`), and table of the b-probabilities. The object can be analysed with the class-specific functions `print`, `summary`, and `plot`.

**Author(s)**

Mikhail A. Beketov

**See Also**

Functions: [baumwelchcont](#), [viterbicont](#), and [statesDistributionsPlot](#).

**Examples**

```
Returns<-logreturns(Prices) # Getting a stationary process
Returns<-Returns*10 # Scaling the values

hmm<-hmmsetcont>Returns) # Creating a HMM object
print(hmm) # Checking the initial parameters

for(i in 1:6){hmm<-baumwelchcont(hmm)} # Baum-Welch is
# executed 6 times and results are accumulated
hmmcomplete<-viterbicont(hmm) # Viterbi execution
print(hmm) # Checking the accumulated parameters
summary(hmm) # Getting more detailed information
par(mfrow=c(2,1))
plot(hmmcomplete, Prices, ylabel="Price")
plot(hmmcomplete, ylabel="Returns") # the revealed
# Markov chain and the observations are plotted
```

---

logreturns

*Calculating Log-returns*

---

**Description**

Simple function that calculates log-returns from prices time series.

**Usage**

```
logreturns(x)
```

**Arguments**

x                      Vector of prices or an index values (class "numeric").

**Value**

Vector of log-returns (class "numeric").

**Author(s)**

Mikhail A. Beketov

**See Also**

Functions: [hmmsetcont](#), [baumwelchcont](#), [viterbicont](#), and [statesDistributionsPlot](#).

**Examples**

```
Returns<-logreturns(Prices)
par(mfrow=c(2,1))
plot(Prices, type="l")
plot>Returns, type="l")
```

---

Prices

*A dummy data set of prices.*

---

**Description**

A dummy data set of prices or a non-stationary random process.

**Usage**

```
data(Prices)
```

**Format**

The format is: num [1:168] 1394 1366 1499 1452 1421 ...

**Source**

A dummy data set generated by the author.

**Examples**

```
plot(Prices, type="l")
```



---

`statesDistributionsPlot`*Probability Density Functions of the States*

---

**Description**

The function plots the Gaussian probability density functions from the means and variances of the whole data set, the two sub-sets corresponding to the two Markov chain states, and additionally from the HMM model (i.e. the means and variances taken from the last Baum-Welch iteration).

**Usage**

```
statesDistributionsPlot(hmm, sc = 1)
```

**Arguments**

<code>hmm</code>	An object of the class <code>ContObservHMM</code> .
<code>sc</code>	Scaling factor used when the initial HMM-object was set.

**Value**

Plot of the probability density functions.

**Author(s)**

Mikhail A. Beketov

**See Also**

Functions: [hmmsetcont](#), [baumwelchcont](#), and [viterbicont](#).

**Examples**

```
Returns<-logreturns(Prices) # Getting a stationary process
Returns<-Returns*10 # Scaling the values
hmm<-hmmsetcont>Returns) # Creating a HMM object
for(i in 1:6){hmm<-baumwelchcont(hmm)} # Baum-Welch is
# executed 6 times and results are accumulated
hmmcomplete<-viterbicont(hmm) # Viterbi execution

statesDistributionsPlot(hmmcomplete, sc=10) # PDFs of
# the whole data set and two states are plotted
```

---

viterbicont

*Viterbi Algorithm*

---

### Description

The function performs Viterbi algorithm (Viterbi, 1967). It can be applied to a ContObservHMM object after sufficient number of Baum-welch iterations (function `baumwelchcont`).

### Usage

```
viterbicont(hmm)
```

### Arguments

`hmm` An object of the class ContObservHMM.

### Value

An object of the class ContObservHMM (see section on the function `hmmsetcont`). The object can be analysed with the class-specific functions `print`, `summary`, and `plot`.

### Author(s)

Mikhail A. Beketov

### References

Viterbi, A.J. 1967. Error bounds for convolutional codes and an asymptotically optimum decoding algorithm. IEEE Transactions on Information Theory. 13: 260-269.

### See Also

Functions: [hmmsetcont](#), [baumwelchcont](#), and [statesDistributionsPlot](#).

### Examples

```
Returns<-logreturns(Prices) # Getting a stationary process
Returns<-Returns*10 # Scaling the values
hmm<-hmmsetcont>Returns) # Creating a HMM object
for(i in 1:6){hmm<-baumwelchcont(hmm)} # Baum-Welch is
# executed 6 times and results are accumulated

hmmcomplete<-viterbicont(hmm) # Viterbi execution

par(mfrow=c(2,1))
plot(hmmcomplete, Prices, ylabel="Price")
plot(hmmcomplete, ylabel="Returns") # the revealed
# Markov chain and the observations are plotted
```

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