

# Package ‘seqHMM’

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**Title** Mixture Hidden Markov Models for Social Sequence Data and Other Multivariate, Multichannel Categorical Time Series

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**Description** Designed for fitting hidden (latent) Markov models and mixture hidden Markov models for social sequence data and other categorical time series. Also some more restricted versions of these type of models are available: Markov models, mixture Markov models, and latent class models. The package supports models for one or multiple subjects with one or multiple parallel sequences (channels). External covariates can be added to explain cluster membership in mixture models. The package provides functions for evaluating and comparing models, as well as functions for visualizing of multichannel sequence data and hidden Markov models. Models are estimated using maximum likelihood via the EM algorithm and/or direct numerical maximization with analytical gradients. All main algorithms are written in C++ with support for parallel computation. Documentation is available via several vignettes in this page, and the paper by Helske and Helske (2019, <[doi:10.18637/jss.v088.i03](https://doi.org/10.18637/jss.v088.i03)>).

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**Suggests** MASS, nnet, knitr, testthat (>= 3.0.0), covr

**License** GPL (>= 2)

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## R topics documented:

biofam3c	3
build_hmm	4
build_lcm	8
build_mhmm	12
build_mm	18
build_mmm	19
cluster_names	22
cluster_names<-	22
colorpalette	23
estimate_coef	24
fit_model	24
forward_backward	36
gridplot	37
hidden_paths	40
hmm_biofam	41
hmm_mvad	43
logLik.hmm	44
logLik.mhmm	45
mc_to_sc	46
mc_to_sc_data	47
mhmm_biofam	49
mhmm_mvad	52
mssplot	54
plot.hmm	58
plot.mhmm	63
plot.ssp	67
plot_colors	68
posterior_probs	69
print.hmm	69
separate_mhmm	70
seqdef	71
seqHMM	71
seqHMM-deprecated	72
simulate_hmm	74
simulate_initial_probs	76
simulate_mhmm	76
ssp	79
ssplot	84
state_names	88

state\_names<- . . . . . 89  
summary.mhmm . . . . . 90  
trim\_model . . . . . 91  
vcov.mhmm . . . . . 92

**Index 94**

biofam3c *Three-channel biofam data*

**Description**

Biofam data from the TraMineR package converted into three channels.

**Format**

A list including three sequence data sets for 2000 individuals with 16 state variables, and a separate data frame with 1 id variable, 8 covariates, and 2 weight variables.

**Details**

This data is constructed from the [biofam](#) data in the TraMineR package. Here the original state sequences are converted into three separate data sets: children, married, and left. These include the corresponding life states from age 15 to 30: childless or (having) children; single, married, or divorced; and (living) with parents or left home.

Note that the divorced state does not give information on parenthood or residence, so a guess is made based on preceding states.

The fourth data frame covariates is a collection of additional variables from the original data:

idhous	id
sex	sex
birthyr	birth year
nat_1_02	first nationality
plingu02	language of questionnaire
p02r01	religion
p02r04	religious participation
cspfaj	father's social status
cspmoj	mother's social status
wp00tbgp	weights inflating to the Swiss population
wp00tbgs	weights respecting sample size

The data is loaded by calling `data(biofam3c)`. It was built using following code:

```
data("biofam" , package = "TraMineR")
biofam3c <- with(biofam, {

## Building one channel per type of event left, children or married
```

```

bf <- as.matrix(biofam[, 10:25])
children <- bf == 4 | bf == 5 | bf == 6
married <- bf == 2 | bf == 3 | bf == 6
left <- bf == 1 | bf == 3 | bf == 5 | bf == 6 | bf == 7

children[children == TRUE] <- "children"
children[children == FALSE] <- "childless"
# Divorced parents
div <- bf[(rowSums(bf == 7) > 0 & rowSums(bf == 5) > 0) |
          (rowSums(bf == 7) > 0 & rowSums(bf == 6) > 0),]
children[rownames(bf) %in% rownames(div) & bf == 7] <- "children"

married[married == TRUE] <- "married"
married[married == FALSE] <- "single"
married[bf == 7] <- "divorced"

left[left == TRUE] <- "left home"
left[left == FALSE] <- "with parents"
# Divorced living with parents (before divorce)
wp <- bf[(rowSums(bf == 7) > 0 & rowSums(bf == 2) > 0 &
          rowSums(bf == 3) == 0 & rowSums(bf == 5) == 0 &
          rowSums(bf == 6) == 0) |
          (rowSums(bf == 7) > 0 & rowSums(bf == 4) > 0 &
          rowSums(bf == 3) == 0 & rowSums(bf == 5) == 0 &
          rowSums(bf == 6) == 0), ]
left[rownames(bf) %in% rownames(wp) & bf == 7] <- "with parents"

list("children" = children, "married" = married, "left" = left,
     "covariates" = biofam[, c(1:9, 26:27)])
})

```

### Source

biofam data constructed from the Swiss Household Panel <https://forscenter.ch/projects/swiss-household-panel/>

### References

Müller, N. S., M. Studer, G. Ritschard (2007). Classification de parcours de vie à l'aide de l'optimal matching. In *XIVe Rencontre de la Société francophone de classification (SFC 2007), Paris, 5 - 7 septembre 2007*, pp. 157–160.

---

build\_hmm

*Build a Hidden Markov Model*

---

### Description

Function build\_hmm constructs a hidden Markov model object of class hmm.

**Usage**

```

build_hmm(
  observations,
  n_states,
  transition_probs,
  emission_probs,
  initial_probs,
  state_names = NULL,
  channel_names = NULL,
  ...
)

```

**Arguments**

observations	An stslst object (see <a href="#">seqdef</a> ) containing the sequences, or a list of such objects (one for each channel).
n_states	A scalar giving the number of hidden states. Not used if starting values for model parameters are given with <code>initial_probs</code> , <code>transition_probs</code> , or <code>emission_probs</code> .
transition_probs	A matrix of transition probabilities.
emission_probs	A matrix of emission probabilities or a list of such objects (one for each channel). Emission probabilities should follow the ordering of the alphabet of observations ( <code>alphabet(observations)</code> ), returned as <code>symbol_names</code> .
initial_probs	A vector of initial state probabilities.
state_names	A list of optional labels for the hidden states. If <code>NULL</code> , the state names are taken from the row names of the transition matrix. If this is also <code>NULL</code> , numbered states are used.
channel_names	A vector of optional names for the channels.
...	Additional arguments to <code>simulate_transition_probs</code> .

**Details**

The returned model contains some attributes such as `nobs` and `df`, which define the number of observations in the model and the number of estimable model parameters, used in computing BIC. When computing `nobs` for a multichannel model with  $C$  channels, each observed value in a single channel amounts to  $1/C$  observation, i.e. a fully observed time point for a single sequence amounts to one observation. For the degrees of freedom `df`, zero probabilities of the initial model are defined as structural zeroes.

**Value**

Object of class `hmm` with the following elements:

observations	State sequence object or a list of such objects containing the data.
transition_probs	A matrix of transition probabilities.
emission_probs	A matrix or a list of matrices of emission probabilities.

initial\_probs A vector of initial probabilities.  
 state\_names Names for hidden states.  
 symbol\_names Names for observed states.  
 channel\_names Names for channels of sequence data.  
 length\_of\_sequences (Maximum) length of sequences.  
 n\_sequences Number of sequences.  
 n\_symbols Number of observed states (in each channel).  
 n\_states Number of hidden states.  
 n\_channels Number of channels.

### See Also

[fit\\_model](#) for estimating model parameters; and [plot.hmm](#) for plotting hmm objects.

### Examples

```

# Single-channel data

data("mvad", package = "TraMineR")

mvad_alphabet <- c(
  "employment", "FE", "HE", "joblessness", "school",
  "training"
)
mvad_labels <- c(
  "employment", "further education", "higher education",
  "joblessness", "school", "training"
)
mvad_scodes <- c("EM", "FE", "HE", "JL", "SC", "TR")
mvad_seq <- seqdef(mvad, 17:86,
  alphabet = mvad_alphabet, states = mvad_scodes,
  labels = mvad_labels, xtstep = 6
)

# Initializing an HMM with 4 hidden states, random starting values
init_hmm_mvad1 <- build_hmm(observations = mvad_seq, n_states = 4)

# Starting values for the emission matrix
emiss <- matrix(NA, nrow = 4, ncol = 6)
emiss[1, ] <- seqstatf(mvad_seq[, 1:12])[, 2] + 1
emiss[2, ] <- seqstatf(mvad_seq[, 13:24])[, 2] + 1
emiss[3, ] <- seqstatf(mvad_seq[, 25:48])[, 2] + 1
emiss[4, ] <- seqstatf(mvad_seq[, 49:70])[, 2] + 1
emiss <- emiss / rowSums(emiss)

# Starting values for the transition matrix

tr <- matrix(

```

```

    c(
      0.80, 0.10, 0.05, 0.05,
      0.05, 0.80, 0.10, 0.05,
      0.05, 0.05, 0.80, 0.10,
      0.05, 0.05, 0.10, 0.80
    ),
    nrow = 4, ncol = 4, byrow = TRUE
  )

# Starting values for initial state probabilities
init <- c(0.3, 0.3, 0.2, 0.2)

# HMM with own starting values
init_hmm_mvad2 <- build_hmm(
  observations = mvad_seq, transition_probs = tr,
  emission_probs = emiss, initial_probs = init
)

#####

# Multichannel data

# Three-state three-channel hidden Markov model
# See ?hmm_biofam for a five-state version

data("biofam3c")

# Building sequence objects
marr_seq <- seqdef(biofam3c$marrried,
  start = 15,
  alphabet = c("single", "married", "divorced")
)
child_seq <- seqdef(biofam3c$children,
  start = 15,
  alphabet = c("childless", "children")
)
left_seq <- seqdef(biofam3c$left,
  start = 15,
  alphabet = c("with parents", "left home")
)

# Define colors
attr(marr_seq, "cpal") <- c("violetred2", "darkgoldenrod2", "darkmagenta")
attr(child_seq, "cpal") <- c("darkseagreen1", "coral3")
attr(left_seq, "cpal") <- c("lightblue", "red3")

# Left-to-right HMM with 3 hidden states and random starting values
set.seed(1010)
init_hmm_bf1 <- build_hmm(
  observations = list(marr_seq, child_seq, left_seq),
  n_states = 3, left_right = TRUE, diag_c = 2
)

```

```

# Starting values for emission matrices

emiss_marr <- matrix(NA, nrow = 3, ncol = 3)
emiss_marr[1, ] <- seqstatf(marr_seq[, 1:5])[, 2] + 1
emiss_marr[2, ] <- seqstatf(marr_seq[, 6:10])[, 2] + 1
emiss_marr[3, ] <- seqstatf(marr_seq[, 11:16])[, 2] + 1
emiss_marr <- emiss_marr / rowSums(emiss_marr)

emiss_child <- matrix(NA, nrow = 3, ncol = 2)
emiss_child[1, ] <- seqstatf(child_seq[, 1:5])[, 2] + 1
emiss_child[2, ] <- seqstatf(child_seq[, 6:10])[, 2] + 1
emiss_child[3, ] <- seqstatf(child_seq[, 11:16])[, 2] + 1
emiss_child <- emiss_child / rowSums(emiss_child)

emiss_left <- matrix(NA, nrow = 3, ncol = 2)
emiss_left[1, ] <- seqstatf(left_seq[, 1:5])[, 2] + 1
emiss_left[2, ] <- seqstatf(left_seq[, 6:10])[, 2] + 1
emiss_left[3, ] <- seqstatf(left_seq[, 11:16])[, 2] + 1
emiss_left <- emiss_left / rowSums(emiss_left)

# Starting values for transition matrix
trans <- matrix(
  c(
    0.9, 0.07, 0.03,
    0, 0.9, 0.1,
    0, 0, 1
  ),
  nrow = 3, ncol = 3, byrow = TRUE
)

# Starting values for initial state probabilities
inits <- c(0.9, 0.09, 0.01)

# HMM with own starting values
init_hmm_bf2 <- build_hmm(
  observations = list(marr_seq, child_seq, left_seq),
  transition_probs = trans,
  emission_probs = list(emiss_marr, emiss_child, emiss_left),
  initial_probs = inits
)

```

---

 build\_lcm

*Build a Latent Class Model*


---

### Description

Function `build_lcm` is a shortcut for constructing a latent class model as a restricted case of an `mhmm` object.



**Usage**

```

build_lcm(
  observations,
  n_clusters,
  emission_probs,
  formula = NULL,
  data = NULL,
  coefficients = NULL,
  cluster_names = NULL,
  channel_names = NULL
)

```

**Arguments**

- |                |   |
|----------------|---|
| observations   | An <code>stslst</code> object (see <a href="#">seqdef</a> ) containing the sequences, or a list of such objects (one for each channel).   |
| n_clusters     | A scalar giving the number of clusters/submodels (not used if starting values for model parameters are given with <code>emission_probs</code> ).  |
| emission_probs | A matrix containing emission probabilities for each class by rows, or in case of multichannel data a list of such matrices. Note that the matrices must have dimensions $k \times s$ where $k$ is the number of latent classes and $s$ is the number of unique symbols (observed states) in the data. Emission probabilities should follow the ordering of the alphabet of observations ( <code>alphabet(observations)</code> ), returned as <code>symbol_names</code> ). |
| formula        | Optional formula of class <code>formula</code> for the mixture probabilities. Left side omitted.  |
| data           | A data frame containing the variables used in the formula. Ignored if no formula is provided.   |
| coefficients   | An optional $k \times l$ matrix of regression coefficients for time-constant covariates for mixture probabilities, where $l$ is the number of clusters and $k$ is the number of covariates. A logit-link is used for mixture probabilities. The first column is set to zero.  |
| cluster_names  | A vector of optional names for the classes/clusters.  |
| channel_names  | A vector of optional names for the channels.  |

**Value**

Object of class `mhmm` with the following elements:

- |                  |  |
|------------------|--|
| observations     | State sequence object or a list of such containing the data.                                 |
| transition_probs | A matrix of transition probabilities.  |
| emission_probs   | A matrix or a list of matrices of emission probabilities.                                    |
| initial_probs    | A vector of initial probabilities.   |
| coefficients     | A matrix of parameter coefficients for covariates (covariates in rows, clusters in columns). |

X Covariate values for each subject.  
 cluster\_names Names for clusters.  
 state\_names Names for hidden states.  
 symbol\_names Names for observed states.  
 channel\_names Names for channels of sequence data  
 length\_of\_sequences (Maximum) length of sequences.  
 n\_sequences Number of sequences.  
 n\_symbols Number of observed states (in each channel).  
 n\_states Number of hidden states.  
 n\_channels Number of channels.  
 n\_covariates Number of covariates.  
 n\_clusters Number of clusters.

### See Also

[fit\\_model](#) for estimating model parameters; [summary.mhmm](#) for a summary of a mixture model; [separate\\_mhmm](#) for organizing an mhmm object into a list of separate hmm objects; and [plot.mhmm](#) for plotting mixture models.

### Examples

```

# Simulate observations from two classes
set.seed(123)
obs <- seqdef(rbind(
  matrix(sample(letters[1:3], 500, TRUE, prob = c(0.1, 0.6, 0.3)), 50, 10),
  matrix(sample(letters[1:3], 200, TRUE, prob = c(0.4, 0.4, 0.2)), 20, 10)
))

# Initialize the model
set.seed(9087)
model <- build_lcm(obs, n_clusters = 2)

# Estimate model parameters
fit <- fit_model(model)

# How many of the observations were correctly classified:
sum(summary(fit$model)$most_probable_cluster == rep(c("Class 2", "Class 1"), times = c(500, 200)))

#####
## Not run:
# LCM for longitudinal data

# Define sequence data
data("mvad", package = "TraMineR")
mvad_alphabet <- c(
  "employment", "FE", "HE", "joblessness", "school",
  "training"
)

```

```

mvad_labels <- c(
  "employment", "further education", "higher education",
  "joblessness", "school", "training"
)
mvad_scodes <- c("EM", "FE", "HE", "JL", "SC", "TR")
mvad_seq <- seqdef(mvad, 17:86,
  alphabet = mvad_alphabet, states = mvad_scodes,
  labels = mvad_labels, xtstep = 6
)

# Initialize the LCM with random starting values
set.seed(7654)
init_lcm_mvad1 <- build_lcm(
  observations = mvad_seq,
  n_clusters = 2, formula = ~male, data = mvad
)

# Own starting values for emission probabilities
emiss <- rbind(rep(1 / 6, 6), rep(1 / 6, 6))

# LCM with own starting values
init_lcm_mvad2 <- build_lcm(
  observations = mvad_seq,
  emission_probs = emiss, formula = ~male, data = mvad
)

# Estimate model parameters (EM algorithm with random restarts)
lcm_mvad <- fit_model(init_lcm_mvad1,
  control_em = list(restart = list(times = 5))
)$model

# Plot the LCM
plot(lcm_mvad, interactive = FALSE, ncol = 2)

#####

# Binomial regression (comparison to glm)

require("MASS")
data("birthwt")

model <- build_lcm(
  observations = seqdef(birthwt$low), emission_probs = diag(2),
  formula = ~ age + lwt + smoke + ht, data = birthwt
)
fit <- fit_model(model)
summary(fit$model)
summary(glm(low ~ age + lwt + smoke + ht, binomial, data = birthwt))

# Multinomial regression (comparison to multinom)

require("nnet")

```

```

set.seed(123)
n <- 100
X <- cbind(1, x1 = runif(n, 0, 1), x2 = runif(n, 0, 1))
coefs <- cbind(0, c(-2, 5, -2), c(0, -2, 2))
pr <- exp(X %*% coefs) + rnorm(n * 3)
pr <- pr / rowSums(pr)
y <- apply(pr, 1, which.max)
table(y)

model <- build_lcm(
  observations = seqdef(y), emission_probs = diag(3),
  formula = ~ x1 + x2, data = data.frame(X[, -1])
)
fit <- fit_model(model)
summary(fit$model)
summary(multinom(y ~ x1 + x2, data = data.frame(X[, -1])))

## End(Not run)

```

---

 build\_mhmm

*Build a Mixture Hidden Markov Model*


---

## Description

Function `build_mhmm` constructs a mixture hidden Markov model object of class `mhmm`.

## Usage

```

build_mhmm(
  observations,
  n_states,
  transition_probs,
  emission_probs,
  initial_probs,
  formula = NULL,
  data = NULL,
  coefficients = NULL,
  cluster_names = NULL,
  state_names = NULL,
  channel_names = NULL,
  ...
)

```

## Arguments

`observations` An `stslst` object (see [seqdef](#)) containing the sequences, or a list of such objects (one for each channel).

n_states	A numerical vector giving the number of hidden states in each submodel (not used if starting values for model parameters are given with initial_probs, transition_probs, or emission_probs).
transition_probs	A list of matrices of transition probabilities for the submodel of each cluster.
emission_probs	A list which contains matrices of emission probabilities or a list of such objects (one for each channel) for the submodel of each cluster. Note that the matrices must have dimensions $m \times s$ where $m$ is the number of hidden states and $s$ is the number of unique symbols (observed states) in the data. Emission probabilities should follow the ordering of the alphabet of observations (alphabet(observations), returned as symbol_names).
initial_probs	A list which contains vectors of initial state probabilities for the submodel of each cluster.
formula	Optional formula of class <code>formula</code> for the mixture probabilities. Left side omitted.
data	A data frame containing the variables used in the formula. Ignored if no formula is provided.
coefficients	An optional $k \times l$ matrix of regression coefficients for time-constant covariates for mixture probabilities, where $l$ is the number of clusters and $k$ is the number of covariates. A logit-link is used for mixture probabilities. The first column is set to zero.
cluster_names	A vector of optional names for the clusters.
state_names	A list of optional labels for the hidden states. If NULL, the state names are taken as row names of transition matrices. If this is also NULL, numbered states are used.
channel_names	A vector of optional names for the channels.
...	Additional arguments to <code>simulate_transition_probs</code> .

### Details

The returned model contains some attributes such as `nobs` and `df`, which define the number of observations in the model and the number of estimable model parameters, used in computing BIC. When computing `nobs` for a multichannel model with  $C$  channels, each observed value in a single channel amounts to  $1/C$  observation, i.e. a fully observed time point for a single sequence amounts to one observation. For the degrees of freedom `df`, zero probabilities of the initial model are defined as structural zeroes.

### Value

Object of class `mhmm` with following elements:

`observations` State sequence object or a list of such containing the data.  
`transition_probs` A matrix of transition probabilities.  
`emission_probs` A matrix or a list of matrices of emission probabilities.  
`initial_probs` A vector of initial probabilities.

coefficients A matrix of parameter coefficients for covariates (covariates in rows, clusters in columns).

X Covariate values for each subject.

cluster\_names Names for clusters.

state\_names Names for hidden states.

symbol\_names Names for observed states.

channel\_names Names for channels of sequence data

length\_of\_sequences (Maximum) length of sequences.

n\_sequences Number of sequences.

n\_symbols Number of observed states (in each channel).

n\_states Number of hidden states.

n\_channels Number of channels.

n\_covariates Number of covariates.

n\_clusters Number of clusters.

## References

Helske S. and Helske J. (2019). Mixture Hidden Markov Models for Sequence Data: The seqHMM Package in R, *Journal of Statistical Software*, 88(3), 1-32. doi:10.18637/jss.v088.i03

## See Also

[fit\\_model](#) for fitting mixture Hidden Markov models; [summary.mhmm](#) for a summary of a MHMM; [separate\\_mhmm](#) for reorganizing a MHMM into a list of separate hidden Markov models; and [plot.mhmm](#) for plotting mhmm objects.

## Examples

```
data("biofam3c")

## Building sequence objects
marr_seq <- seqdef(biofam3c$married,
  start = 15,
  alphabet = c("single", "married", "divorced")
)
child_seq <- seqdef(biofam3c$children,
  start = 15,
  alphabet = c("childless", "children")
)
left_seq <- seqdef(biofam3c$left,
  start = 15,
  alphabet = c("with parents", "left home")
)

## Choosing colors
attr(marr_seq, "cpal") <- c("#AB82FF", "#E6AB02", "#E7298A")
```

```
attr(child_seq, "cpal") <- c("#66C2A5", "#FC8D62")
attr(left_seq, "cpal") <- c("#A6CEE3", "#E31A1C")

## MHMM with random starting values, no covariates
set.seed(468)
init_mhmm_bf1 <- build_mhmm(
  observations = list(marr_seq, child_seq, left_seq),
  n_states = c(4, 4, 6),
  channel_names = c("Marriage", "Parenthood", "Residence")
)

## Starting values for emission probabilities

# Cluster 1
B1_marr <- matrix(
  c(
    0.8, 0.1, 0.1, # High probability for single
    0.8, 0.1, 0.1,
    0.3, 0.6, 0.1, # High probability for married
    0.3, 0.3, 0.4
  ), # High probability for divorced
  nrow = 4, ncol = 3, byrow = TRUE
)

B1_child <- matrix(
  c(
    0.9, 0.1, # High probability for childless
    0.9, 0.1,
    0.9, 0.1,
    0.9, 0.1
  ),
  nrow = 4, ncol = 2, byrow = TRUE
)

B1_left <- matrix(
  c(
    0.9, 0.1, # High probability for living with parents
    0.1, 0.9, # High probability for having left home
    0.1, 0.9,
    0.1, 0.9
  ),
  nrow = 4, ncol = 2, byrow = TRUE
)

# Cluster 2
B2_marr <- matrix(
  c(
    0.8, 0.1, 0.1, # High probability for single
    0.8, 0.1, 0.1,
    0.1, 0.8, 0.1, # High probability for married
    0.7, 0.2, 0.1
  )
)
```

```

),
nrow = 4, ncol = 3, byrow = TRUE
)

B2_child <- matrix(
  c(
    0.9, 0.1, # High probability for childless
    0.9, 0.1,
    0.9, 0.1,
    0.1, 0.9
  ),
  nrow = 4, ncol = 2, byrow = TRUE
)

B2_left <- matrix(
  c(
    0.9, 0.1, # High probability for living with parents
    0.1, 0.9,
    0.1, 0.9,
    0.1, 0.9
  ),
  nrow = 4, ncol = 2, byrow = TRUE
)

# Cluster 3
B3_marr <- matrix(
  c(
    0.8, 0.1, 0.1, # High probability for single
    0.8, 0.1, 0.1,
    0.8, 0.1, 0.1,
    0.1, 0.8, 0.1, # High probability for married
    0.3, 0.4, 0.3,
    0.1, 0.1, 0.8
  ), # High probability for divorced
  nrow = 6, ncol = 3, byrow = TRUE
)

B3_child <- matrix(
  c(
    0.9, 0.1, # High probability for childless
    0.9, 0.1,
    0.5, 0.5,
    0.5, 0.5,
    0.5, 0.5,
    0.1, 0.9
  ),
  nrow = 6, ncol = 2, byrow = TRUE
)

B3_left <- matrix(
  c(
    0.9, 0.1, # High probability for living with parents

```



```

    0.1, 0.9,
    0.5, 0.5,
    0.5, 0.5,
    0.1, 0.9,
    0.1, 0.9
  ),
  nrow = 6, ncol = 2, byrow = TRUE
)

# Starting values for transition matrices
A1 <- matrix(
  c(
    0.80, 0.16, 0.03, 0.01,
    0, 0.90, 0.07, 0.03,
    0, 0, 0.90, 0.10,
    0, 0, 0, 1
  ),
  nrow = 4, ncol = 4, byrow = TRUE
)

A2 <- matrix(
  c(
    0.80, 0.10, 0.05, 0.03, 0.01, 0.01,
    0, 0.70, 0.10, 0.10, 0.05, 0.05,
    0, 0, 0.85, 0.01, 0.10, 0.04,
    0, 0, 0, 0.90, 0.05, 0.05,
    0, 0, 0, 0, 0.90, 0.10,
    0, 0, 0, 0, 0, 1
  ),
  nrow = 6, ncol = 6, byrow = TRUE
)

# Starting values for initial state probabilities
initial_probs1 <- c(0.9, 0.07, 0.02, 0.01)
initial_probs2 <- c(0.9, 0.04, 0.03, 0.01, 0.01, 0.01)

# Birth cohort
biofam3c$covariates$cohort <- cut(biofam3c$covariates$birthyr, c(1908, 1935, 1945, 1957))
biofam3c$covariates$cohort <- factor(
  biofam3c$covariates$cohort,
  labels = c("1909-1935", "1936-1945", "1946-1957")
)

## MHMM with own starting values and covariates
init_mhmm_bf2 <- build_mhmm(
  observations = list(marr_seq, child_seq, left_seq),
  initial_probs = list(initial_probs1, initial_probs2),
  transition_probs = list(A1, A1, A2),
  emission_probs = list(
    list(B1_marr, B1_child, B1_left),
    list(B2_marr, B2_child, B2_left),
    list(B3_marr, B3_child, B3_left)
  ),
)
```

```

formula = ~ sex + cohort, data = biofam3c$covariates,
cluster_names = c("Cluster 1", "Cluster 2", "Cluster 3"),
channel_names = c("Marriage", "Parenthood", "Residence"),
state_names = list(
  paste("State", 1:4), paste("State", 1:4),
  paste("State", 1:6)
)
)

```

---

 build\_mm

*Build a Markov Model*


---

### Description

Function `build_mm` builds and automatically estimates a Markov model. It is also a shortcut for constructing a Markov model as a restricted case of an `hmm` object.

### Usage

```
build_mm(observations)
```

### Arguments

`observations` An `stslst` object (see [seqdef](#)) containing the sequences.

### Details

Unlike the other build functions in `seqHMM`, the `build_mm` function automatically estimates the model parameters. In case of no missing values, initial and transition probabilities are directly estimated from the observed initial state probabilities and transition counts. In case of missing values, the EM algorithm is run once.

Note that it is possible that the data contains a symbol from which there are no transitions anywhere (even to itself), which would lead to a row in transition matrix full of zeros. In this case the ‘`build_mm`’ (as well as the EM algorithm) assumes that the the state is absorbing in a way that probability of staying in this state is 1.

### Value

Object of class `hmm` with following elements:

`observations` State sequence object or a list of such containing the data.

`transition_probs` A matrix of transition probabilities.

`emission_probs` A matrix or a list of matrices of emission probabilities.

`initial_probs` A vector of initial probabilities.

`state_names` Names for hidden states.

`symbol_names` Names for observed states.

channel\_names Names for channels of sequence data.  
length\_of\_sequences (Maximum) length of sequences.  
n\_sequences Number of sequences.  
n\_symbols Number of observed states (in each channel).  
n\_states Number of hidden states.  
n\_channels Number of channels.

### See Also

[plot.hmm](#) for plotting the model.

### Examples

```
# Construct sequence data
data("mvad", package = "TraMineR")

mvad_alphabet <-
  c("employment", "FE", "HE", "joblessness", "school", "training")
mvad_labels <- c(
  "employment", "further education", "higher education",
  "joblessness", "school", "training"
)
mvad_scodes <- c("EM", "FE", "HE", "JL", "SC", "TR")
mvad_seq <- seqdef(mvad, 17:86,
  alphabet = mvad_alphabet,
  states = mvad_scodes, labels = mvad_labels, xtstep = 6
)

# Define a color palette for the sequence data
attr(mvad_seq, "cpal") <- colorpalette[[6]]

# Estimate the Markov model
mm_mvad <- build_mm(observations = mvad_seq)
```

---

build\_mmm

*Build a Mixture Markov Model*

---

### Description

Function `build_mmm` is a shortcut for constructing a mixture Markov model as a restricted case of an `mhmm` object.

**Usage**

```

build_mmm(
  observations,
  n_clusters,
  transition_probs,
  initial_probs,
  formula = NULL,
  data = NULL,
  coefficients = NULL,
  cluster_names = NULL,
  ...
)

```

**Arguments**

observations	An stslist object (see <a href="#">seqdef</a> ) containing the sequences.
n_clusters	A scalar giving the number of clusters/submodels (not used if starting values for model parameters are given with <code>initial_probs</code> and <code>transition_probs</code> ).
transition_probs	A list of matrices of transition probabilities for submodels of each cluster.
initial_probs	A list which contains vectors of initial state probabilities for submodels of each cluster.
formula	Optional formula of class <a href="#">formula</a> for the mixture probabilities. Left side omitted.
data	A data frame containing the variables used in the formula. Ignored if no formula is provided.
coefficients	An optional $k \times l$ matrix of regression coefficients for time-constant covariates for mixture probabilities, where $l$ is the number of clusters and $k$ is the number of covariates. A logit-link is used for mixture probabilities. The first column is set to zero.
cluster_names	A vector of optional names for the clusters.
...	Additional arguments to <code>simulate_transition_probs</code> .

**Value**

Object of class `mhmm` with following elements:

observations	State sequence object or a list of such containing the data.
transition_probs	A matrix of transition probabilities.
emission_probs	A matrix or a list of matrices of emission probabilities.
initial_probs	A vector of initial probabilities.
coefficients	A matrix of parameter coefficients for covariates (covariates in rows, clusters in columns).
X	Covariate values for each subject.
cluster_names	Names for clusters.

state\_names Names for hidden states.  
 symbol\_names Names for observed states.  
 channel\_names Names for channels of sequence data  
 length\_of\_sequences (Maximum) length of sequences.  
 n\_sequences Number of sequences.  
 n\_symbols Number of observed states (in each channel).  
 n\_states Number of hidden states.  
 n\_channels Number of channels.  
 n\_covariates Number of covariates.  
 n\_clusters Number of clusters.

### See Also

[fit\\_model](#) for estimating model parameters; [summary.mhmm](#) for a summary of a mixture model; [separate\\_mhmm](#) for organizing an mhmm object into a list of separate hmm objects; and [plot.mhmm](#) for plotting mixture models.

### Examples

```

# Define sequence data
data("mvad", package = "TraMineR")
mvad_alphabet <- c(
  "employment", "FE", "HE", "joblessness", "school",
  "training"
)
mvad_labels <- c(
  "employment", "further education", "higher education",
  "joblessness", "school", "training"
)
mvad_scores <- c("EM", "FE", "HE", "JL", "SC", "TR")
mvad_seq <- seqdef(mvad, 17:86,
  alphabet = mvad_alphabet, states = mvad_scores,
  labels = mvad_labels, xstep = 6
)

# Initialize the MMM
set.seed(123)
mmm_mvad <- build_mmm(
  observations = mvad_seq,
  n_clusters = 2,
  formula = ~male, data = mvad
)

## Not run:
# Estimate model parameters
mmm_mvad <- fit_model(mmm_mvad)$model

```

```

# Plot model (both clusters in the same plot)
require(igraph)
plot(mmm_mvad,
     interactive = FALSE,
     # Modify legend position and properties
     with.legend = "right", legend.prop = 0.3, cex.legend = 1.2,
     # Define vertex layout
     layout = layout_as_star,
     # Modify edge properties
     edge.label = NA, edge.arrow.size = 0.8, edge.curved = 0.2,
     # Modify vertex label positions (initial probabilities)
     vertex.label.pos = c("left", "right", "right", "left", "left", "right")
)

# Summary of the MMM
summary(mmm_mvad)

## End(Not run)

```

---

cluster_names	<i>Get cluster names from mhmm object</i>
---------------	---

---

### Description

Get cluster names from mhmm object

### Usage

```
cluster_names(object)
```

### Arguments

object            An object of class 'mhmm'.

### Value

A character vector containing the cluster names.

---

cluster_names<-	<i>Set cluster names for mhmm object</i>
-----------------	--

---

### Description

Set cluster names for mhmm object

### Usage

```
cluster_names(object) <- value
```

**Arguments**

object	An object of class 'mhmm'.
value	A character vector containing the new cluster names.

**Value**

The modified object with updated cluster names.

---

colorpalette	<i>Color palettes</i>
--------------	-----------------------

---

**Description**

A list containing ready defined color palettes with distinct colors using iWantHue. By default, seqHMM uses these palettes when assigning colors.

**Format**

A list with 200 color palettes.

**Source**

iWantHue web page <https://medialab.github.io/iwanthue/>

**See Also**

[plot\\_colors](#) for visualization of color palettes. Implementations of iWantHue for R:

- <https://github.com/hoesler/rwantshue>
- <https://github.com/johnbaums/hues>

**Examples**

```
data("colorpalette")
# Color palette with 9 colors
colorpalette[[9]]
# Color palette with 24 colors
colorpalette[[24]]
```

---

estimate_coef	<i>Estimate Regression Coefficients of Mixture Hidden Markov Models</i>
---------------	---

---

**Description**

Function `estimate_coef` estimates the regression coefficients of mixture hidden Markov models and its restricted variants while keeping other parameters fixed.

**Usage**

```
estimate_coef(model, threads = 1)
```

**Arguments**

<code>model</code>	An object of class <code>hmm</code> or <code>mhmm</code> .
<code>threads</code>	Number of threads to use in parallel computing. The default is 1.

---

fit_model	<i>Estimate Parameters of (Mixture) Hidden Markov Models and Their Restricted Variants</i>
-----------	--

---

**Description**

Function `fit_model` estimates the parameters of mixture hidden Markov models and its restricted variants using maximum likelihood. Initial values for estimation are taken from the corresponding components of the model with preservation of original zero probabilities.

**Usage**

```
fit_model(
  model,
  em_step = TRUE,
  global_step = FALSE,
  local_step = FALSE,
  control_em = list(),
  control_global = list(),
  control_local = list(),
  lb,
  ub,
  threads = 1,
  log_space = FALSE,
  constraints = NULL,
  fixed_inits = NULL,
  fixed_emissions = NULL,
  fixed_transitions = NULL,
  ...
)
```



**Arguments**

- model** An object of class `hmm` or `mhmm`.
- em\_step** Logical. Whether or not to use the EM algorithm at the start of the parameter estimation. The default is `TRUE`.
- global\_step** Logical. Whether or not to use global optimization via `nloptr` (possibly after the EM step). The default is `FALSE`.
- local\_step** Logical. Whether or not to use local optimization via `nloptr` (possibly after the EM and/or global steps). The default is `FALSE`.
- control\_em** Optional list of control parameters for the EM algorithm. Possible arguments are
- maxeval** The maximum number of iterations, the default is 1000. Note that iteration counter starts with -1 so with `maxeval=1` you get already two iterations. This is for backward compatibility reasons.
  - print\_level** The level of printing. Possible values are 0 (prints nothing), 1 (prints information at the start and the end of the algorithm), 2 (prints at every iteration), and for mixture models 3 (print also during optimization of coefficients).
  - reltol** Relative tolerance for convergence defined as  $(\log Lik_{new} - \log Lik_{old}) / (\text{abs}(\log Lik_{old}) + 0.1)$ . The default is  $1e-10$ .
  - restart** A list containing options for possible EM restarts with the following components:
    - times** Number of restarts of the EM algorithm using random initial values. The default is 0, i.e. no restarts.
    - transition** Logical. Should the original transition probabilities be varied? The default is `TRUE`.
    - emission** Logical. Should the original emission probabilities be varied? The default is `TRUE`.
    - sd** Standard deviation for `rnorm` used in randomization. The default is 0.25.
    - maxeval** Maximum number of iterations, the default is `control_em$maxeval`
    - print\_level** Level of printing in restarted EM steps. The default is `control_em$print_level`.
    - reltol** Relative tolerance for convergence at restarted EM steps. The default is `control_em$reltol`. If the relative change of the final model of the restart phase is larger than the tolerance for the original EM phase, the final model is re-estimated with the original `reltol` and `maxeval` at the end of the EM step.
    - n\_optimum** Save the log-likelihood values of the `n_optimum` best models (from all estimated models including the the first EM run.). The default is `min(times + 1, 25)`.
    - use\_original** If `TRUE`. Use the initial values of the input model as starting points for the permutations. Otherwise permute the results of the first EM run.
- control\_global** Optional list of additional arguments for `nloptr` argument `opts`. The default values are

	<b>algorithm</b> "NLOPT_GD_MLSL_LDS"
	<b>local_opts</b> list(algorithm = "NLOPT_LD_LBFGS", ftol_rel = 1e-6, xtol_rel = 1e-4)
	<b>maxeval</b> 10000 (maximum number of iterations in global optimization algorithm.)
	<b>maxtime</b> 60 (maximum time for global optimization. Set to 0 for unlimited time.)
control_local	Optional list of additional arguments for <code>nloptr</code> argument opts. The default values are  <b>algorithm</b> "NLOPT_LD_LBFGS" <b>ftol_rel</b> 1e-10 <b>xtol_rel</b> 1e-8 <b>maxeval</b> 10000 (maximum number of iterations)
lb, ub	Lower and upper bounds for parameters in Softmax parameterization. The default interval is $[pmin(-25, 2 * initialvalues), pmax(25, 2 * initialvalues)]$ , except for gamma coefficients, where the scale of covariates is taken into account. Note that it might still be a good idea to scale covariates around unit scale. Bounds are used only in the global optimization step.
threads	Number of threads to use in parallel computing. The default is 1.
log_space	Make computations using log-space instead of scaling for greater numerical stability at a cost of decreased computational performance. The default is FALSE.
constraints	Integer vector defining equality constraints for emission distributions. Not supported for EM algorithm. See details.
fixed_inits	Can be used to fix some of the probabilities to their initial values. Should have same structure as <code>model\$initial_probs</code> , where each element is either TRUE (fixed) or FALSE (to be estimated). Note that zero probabilities are always fixed to 0. Not supported for EM algorithm. See details.
fixed_emissions	Can be used to fix some of the probabilities to their initial values. Should have same structure as <code>model\$emission_probs</code> , where each element is either TRUE (fixed) or FALSE (to be estimated). Note that zero probabilities are always fixed to 0. Not supported for EM algorithm. See details.
fixed_transitions	Can be used to fix some of the probabilities to their initial values. Should have same structure as <code>model\$transition_probs</code> , where each element is either TRUE (fixed) or FALSE (to be estimated). Note that zero probabilities are always fixed to 0. Not supported for EM algorithm. See details.
...	Additional arguments to <code>nloptr</code> .

### Details

The fitting function provides three estimation steps: 1) EM algorithm, 2) global optimization, and 3) local optimization. The user can call for one method or any combination of these steps, but should note that they are performed in the above-mentioned order. The results from a former step are used

as starting values in a latter, except for some of global optimization algorithms (such as MLSL and StoGO) which only use initial values for setting up the boundaries for the optimization.

It is possible to rerun the EM algorithm automatically using random starting values based on the first run of EM. Number of restarts is defined by the `restart` argument in `control_em`. As the EM algorithm is relatively fast, this method might be preferred option compared to the proper global optimization strategy of step 2.

The default global optimization method (triggered via `global_step = TRUE`) is the multilevel single-linkage method (MLSL) with the LDS modification (`NLOPT_GD_MLSL_LDS` as `algorithm` in `control_global`), with L-BFGS as the local optimizer. The MLSL method draws random starting points and performs a local optimization from each. The LDS modification uses low-discrepancy sequences instead of pseudo-random numbers as starting points and should improve the convergence rate. In order to reduce the computation time spent on non-global optima, the convergence tolerance of the local optimizer is set relatively large. At step 3, a local optimization (L-BFGS by default) is run with a lower tolerance to find the optimum with high precision.

There are some theoretical guarantees that the MLSL method used as the default optimizer in step 2 should find all local optima in a finite number of local optimizations. Of course, it might not always succeed in a reasonable time. The EM algorithm can help in finding good boundaries for the search, especially with good starting values, but in some cases it can mislead. A good strategy is to try a couple of different fitting options with different combinations of the methods: e.g. all steps, only global and local steps, and a few evaluations of EM followed by global and local optimization.

By default, the estimation time is limited to 60 seconds in global optimization step, so it is advisable to change the default settings for the proper global optimization.

Any algorithm available in the `nloptr` function can be used for the global and local steps.

Equality constraints for emission distributions can be defined using the argument `constraints`. This should be a vector with length equal to the number of states, with numbers starting from 1 and increasing for each unique row of the emission probability matrix. For example in case of five states with emissions of first and third states being equal, `constraints = c(1, 2, 1, 3, 4)`. Similarly, some of the model parameters can be fixed to their initial values by using arguments `fixed_inits`, `fixed_emissions`, and `fixed_transitions`, where the structure of the arguments should be same as the corresponding model components, so that `TRUE` value means that the parameter should be fixed and `FALSE` otherwise (it is still treated as fixed if it is zero though). For both types of constrains, only numerical optimisation (local or global) is available, and currently the gradients are computed numerically (if needed) in these cases.

In a case where there is no transitions from one state to anywhere (even to itself), the state is defined as absorbing in a way that probability of staying in this state is fixed to 1. See also `'build_mm'` function.

## Value

**logLik** Log-likelihood of the estimated model.

**em\_results** Results after the EM step: log-likelihood (`logLik`), number of iterations (`iterations`), relative change in log-likelihoods between the last two iterations (`change`), and the log-likelihoods of the `n_optimum` best models after the EM step (`best_opt_restart`).

**global\_results** Results after the global step.

**local\_results** Results after the local step.

**call** The matched function call.

## References

Helske S. and Helske J. (2019). Mixture Hidden Markov Models for Sequence Data: The seqHMM Package in R, *Journal of Statistical Software*, 88(3), 1-32. doi:10.18637/jss.v088.i03

## See Also

[build\\_hmm](#), [build\\_mhmm](#), [build\\_mm](#), [build\\_mmm](#), and [build\\_lcm](#) for constructing different types of models; [summary.mhmm](#) for a summary of a MHMM; [separate\\_mhmm](#) for reorganizing a MHMM into a list of separate hidden Markov models; [plot.hmm](#) and [plot.mhmm](#) for plotting model objects; and [ssplot](#) and [mssplot](#) for plotting stacked sequence plots of hmm and mhmm objects.

## Examples

```
# Hidden Markov model for mvad data

data("mvad", package = "TraMineR")

mvad_alphabet <-
  c("employment", "FE", "HE", "joblessness", "school", "training")
mvad_labels <- c(
  "employment", "further education", "higher education",
  "joblessness", "school", "training"
)
mvad_scodes <- c("EM", "FE", "HE", "JL", "SC", "TR")
mvad_seq <- seqdef(mvad, 17:86,
  alphabet = mvad_alphabet,
  states = mvad_scodes, labels = mvad_labels, xtstep = 6
)

attr(mvad_seq, "cpal") <- colorpalette[[6]]

# Starting values for the emission matrix
emiss <- matrix(
  c(
    0.05, 0.05, 0.05, 0.05, 0.75, 0.05, # SC
    0.05, 0.75, 0.05, 0.05, 0.05, 0.05, # FE
    0.05, 0.05, 0.05, 0.4, 0.05, 0.4, # JL, TR
    0.05, 0.05, 0.75, 0.05, 0.05, 0.05, # HE
    0.75, 0.05, 0.05, 0.05, 0.05, 0.05
  ), # EM
  nrow = 5, ncol = 6, byrow = TRUE
)

# Starting values for the transition matrix
trans <- matrix(0.025, 5, 5)
diag(trans) <- 0.9

# Starting values for initial state probabilities
initial_probs <- c(0.2, 0.2, 0.2, 0.2, 0.2)

# Building a hidden Markov model
```

```

init_hmm_mvad <- build_hmm(
  observations = mvad_seq,
  transition_probs = trans, emission_probs = emiss,
  initial_probs = initial_probs
)

## Not run:
set.seed(21)
fit_hmm_mvad <- fit_model(init_hmm_mvad, control_em = list(restart = list(times = 50)))
hmm_mvad <- fit_hmm_mvad$model

## End(Not run)

# save time, load the previously estimated model
data("hmm_mvad")

# Markov model
# Note: build_mm estimates model parameters from observations,
# no need for estimating with fit_model unless there are missing observations

mm_mvad <- build_mm(observations = mvad_seq)

# Comparing likelihoods, MM fits better
logLik(hmm_mvad)
logLik(mm_mvad)

## Not run:
require("igraph") # for layout_in_circle

plot(mm_mvad,
  layout = layout_in_circle, legend.prop = 0.3,
  edge.curved = 0.3, edge.label = NA,
  vertex.label.pos = c(0, 0, pi, pi, pi, 0)
)

#####

#' # Three-state three-channel hidden Markov model
# See ?hmm_biofam for five-state version

data("biofam3c")

# Building sequence objects
marr_seq <- seqdef(biofam3c$marrried,
  start = 15,
  alphabet = c("single", "married", "divorced")
)
child_seq <- seqdef(biofam3c$children,
  start = 15,
  alphabet = c("childless", "children")
)
left_seq <- seqdef(biofam3c$left,

```

```

    start = 15,
    alphabet = c("with parents", "left home")
  )

# Define colors
attr(marr_seq, "cpal") <- c("violetred2", "darkgoldenrod2", "darkmagenta")
attr(child_seq, "cpal") <- c("darkseagreen1", "coral3")
attr(left_seq, "cpal") <- c("lightblue", "red3")

# Starting values for emission matrices

emiss_marr <- matrix(NA, nrow = 3, ncol = 3)
emiss_marr[1, ] <- seqstatf(marr_seq[, 1:5])[, 2] + 1
emiss_marr[2, ] <- seqstatf(marr_seq[, 6:10])[, 2] + 1
emiss_marr[3, ] <- seqstatf(marr_seq[, 11:16])[, 2] + 1
emiss_marr <- emiss_marr / rowSums(emiss_marr)

emiss_child <- matrix(NA, nrow = 3, ncol = 2)
emiss_child[1, ] <- seqstatf(child_seq[, 1:5])[, 2] + 1
emiss_child[2, ] <- seqstatf(child_seq[, 6:10])[, 2] + 1
emiss_child[3, ] <- seqstatf(child_seq[, 11:16])[, 2] + 1
emiss_child <- emiss_child / rowSums(emiss_child)

emiss_left <- matrix(NA, nrow = 3, ncol = 2)
emiss_left[1, ] <- seqstatf(left_seq[, 1:5])[, 2] + 1
emiss_left[2, ] <- seqstatf(left_seq[, 6:10])[, 2] + 1
emiss_left[3, ] <- seqstatf(left_seq[, 11:16])[, 2] + 1
emiss_left <- emiss_left / rowSums(emiss_left)

# Starting values for transition matrix
trans <- matrix(c(
  0.9, 0.07, 0.03,
  0, 0.9, 0.1,
  0, 0, 1
), nrow = 3, ncol = 3, byrow = TRUE)

# Starting values for initial state probabilities
inits <- c(0.9, 0.09, 0.01)

# Building hidden Markov model with initial parameter values
init_hmm_bf <- build_hmm(
  observations = list(marr_seq, child_seq, left_seq),
  transition_probs = trans,
  emission_probs = list(emiss_marr, emiss_child, emiss_left),
  initial_probs = inits
)

# Fitting the model with different optimization schemes

# Only EM with default values
hmm_1 <- fit_model(init_hmm_bf)
hmm_1$logLik # -24179.1

```

```

# Only L-BFGS
hmm_2 <- fit_model(init_hmm_bf, em_step = FALSE, local_step = TRUE)
hmm_2$logLik # -22267.75

# Global optimization via MLSTL_LDS with L-BFGS as local optimizer and final polisher
# This can be slow, use parallel computing by adjusting threads argument
# (here threads = 1 for portability issues)
hmm_3 <- fit_model(
  init_hmm_bf,
  em_step = FALSE, global_step = TRUE, local_step = TRUE,
  control_global = list(maxeval = 5000, maxtime = 0), threads = 1
)
hmm_3$logLik # -21675.42

# EM with restarts, much faster than MLSTL
set.seed(123)
hmm_4 <- fit_model(init_hmm_bf, control_em = list(restart = list(times = 5)))
hmm_4$logLik # -21675.4

# Global optimization via StOGO with L-BFGS as final polisher
# This can be slow, use parallel computing by adjusting threads argument
# (here threads = 1 for portability issues)
set.seed(123)
hmm_5 <- fit_model(
  init_hmm_bf,
  em_step = FALSE, global_step = TRUE, local_step = TRUE,
  lb = -50, ub = 50, control_global = list(
    algorithm = "NLOPT_GD_STOGO",
    maxeval = 2500, maxtime = 0
  ), threads = 1
)
hmm_5$logLik # -21675.4

#####

# Mixture HMM

data("biofam3c")

## Building sequence objects
marr_seq <- seqdef(biofam3c$married,
  start = 15,
  alphabet = c("single", "married", "divorced")
)
child_seq <- seqdef(biofam3c$children,
  start = 15,
  alphabet = c("childless", "children")
)
left_seq <- seqdef(biofam3c$left,
  start = 15,
  alphabet = c("with parents", "left home")
)

```

```
## Choosing colors
attr(marr_seq, "cpal") <- c("#AB82FF", "#E6AB02", "#E7298A")
attr(child_seq, "cpal") <- c("#66C2A5", "#FC8D62")
attr(left_seq, "cpal") <- c("#A6CEE3", "#E31A1C")

## Starting values for emission probabilities
# Cluster 1
B1_marr <- matrix(
  c(
    0.8, 0.1, 0.1, # High probability for single
    0.8, 0.1, 0.1,
    0.3, 0.6, 0.1, # High probability for married
    0.3, 0.3, 0.4
  ), # High probability for divorced
  nrow = 4, ncol = 3, byrow = TRUE
)

B1_child <- matrix(
  c(
    0.9, 0.1, # High probability for childless
    0.9, 0.1,
    0.9, 0.1,
    0.9, 0.1
  ),
  nrow = 4, ncol = 2, byrow = TRUE
)

B1_left <- matrix(
  c(
    0.9, 0.1, # High probability for living with parents
    0.1, 0.9, # High probability for having left home
    0.1, 0.9,
    0.1, 0.9
  ),
  nrow = 4, ncol = 2, byrow = TRUE
)

# Cluster 2

B2_marr <- matrix(
  c(
    0.8, 0.1, 0.1, # High probability for single
    0.8, 0.1, 0.1,
    0.1, 0.8, 0.1, # High probability for married
    0.7, 0.2, 0.1
  ),
  nrow = 4, ncol = 3, byrow = TRUE
)

B2_child <- matrix(
  c(
    0.9, 0.1, # High probability for childless
    0.9, 0.1,
```



```
    0.9, 0.1,
    0.1, 0.9
  ),
  nrow = 4, ncol = 2, byrow = TRUE
)

B2_left <- matrix(
  c(
    0.9, 0.1, # High probability for living with parents
    0.1, 0.9,
    0.1, 0.9,
    0.1, 0.9
  ),
  nrow = 4, ncol = 2, byrow = TRUE
)

# Cluster 3
B3_marr <- matrix(
  c(
    0.8, 0.1, 0.1, # High probability for single
    0.8, 0.1, 0.1,
    0.8, 0.1, 0.1,
    0.1, 0.8, 0.1, # High probability for married
    0.3, 0.4, 0.3,
    0.1, 0.1, 0.8
  ), # High probability for divorced
  nrow = 6, ncol = 3, byrow = TRUE
)

B3_child <- matrix(
  c(
    0.9, 0.1, # High probability for childless
    0.9, 0.1,
    0.5, 0.5,
    0.5, 0.5,
    0.5, 0.5,
    0.1, 0.9
  ),
  nrow = 6, ncol = 2, byrow = TRUE
)

B3_left <- matrix(
  c(
    0.9, 0.1, # High probability for living with parents
    0.1, 0.9,
    0.5, 0.5,
    0.5, 0.5,
    0.1, 0.9,
    0.1, 0.9
  ),
  nrow = 6, ncol = 2, byrow = TRUE
)
```

```

# Starting values for transition matrices
A1 <- matrix(
  c(
    0.80, 0.16, 0.03, 0.01,
    0, 0.90, 0.07, 0.03,
    0, 0, 0.90, 0.10,
    0, 0, 0, 1
  ),
  nrow = 4, ncol = 4, byrow = TRUE
)

A2 <- matrix(
  c(
    0.80, 0.10, 0.05, 0.03, 0.01, 0.01,
    0, 0.70, 0.10, 0.10, 0.05, 0.05,
    0, 0, 0.85, 0.01, 0.10, 0.04,
    0, 0, 0, 0.90, 0.05, 0.05,
    0, 0, 0, 0, 0.90, 0.10,
    0, 0, 0, 0, 0, 1
  ),
  nrow = 6, ncol = 6, byrow = TRUE
)

# Starting values for initial state probabilities
initial_probs1 <- c(0.9, 0.07, 0.02, 0.01)
initial_probs2 <- c(0.9, 0.04, 0.03, 0.01, 0.01, 0.01)

# Birth cohort
biofam3c$covariates$cohort <- cut(biofam3c$covariates$birthyr, c(1908, 1935, 1945, 1957))
biofam3c$covariates$cohort <- factor(
  biofam3c$covariates$cohort,
  labels = c("1909-1935", "1936-1945", "1946-1957")
)

# Build mixture HMM
init_mhmm_bf <- build_mhmm(
  observations = list(marr_seq, child_seq, left_seq),
  initial_probs = list(initial_probs1, initial_probs1, initial_probs2),
  transition_probs = list(A1, A1, A2),
  emission_probs = list(
    list(B1_marr, B1_child, B1_left),
    list(B2_marr, B2_child, B2_left),
    list(B3_marr, B3_child, B3_left)
  ),
  formula = ~ sex + cohort, data = biofam3c$covariates,
  channel_names = c("Marriage", "Parenthood", "Residence")
)

# Fitting the model with different settings

# Only EM with default values

```

```

mhmm_1 <- fit_model(init_mhmm_bf)
mhmm_1$logLik # -12713.08

# Only L-BFGS
mhmm_2 <- fit_model(init_mhmm_bf, em_step = FALSE, local_step = TRUE)
mhmm_2$logLik # -12966.51

# Use EM with multiple restarts
set.seed(123)
mhmm_3 <- fit_model(init_mhmm_bf, control_em = list(restart = list(times = 5, transition = FALSE)))
mhmm_3$logLik # -12713.08

## End(Not run)

# Left-to-right HMM with equality constraint:

set.seed(1)

# Transition matrix
# Either stay or move to next state
A <- diag(c(0.9, 0.95, 0.95, 1))
A[1, 2] <- 0.1
A[2, 3] <- 0.05
A[3, 4] <- 0.05

# Emission matrix, rows 1 and 3 equal
B <- rbind(
  c(0.4, 0.2, 0.3, 0.1),
  c(0.1, 0.5, 0.1, 0.3),
  c(0.4, 0.2, 0.3, 0.1),
  c(0, 0.2, 0.4, 0.4)
)

# Start from first state
init <- c(1, 0, 0, 0)

# Simulate sequences
sim <- simulate_hmm(
  n_sequences = 100,
  sequence_length = 20, init, A, B
)

# initial model, use true values as inits for faster estimation here
model <- build_hmm(sim$observations, init = init, trans = A, emiss = B)

# estimate the model subject to constraints:
# First and third row of emission matrix are equal (see details)
fit <- fit_model(model,
  constraints = c(1, 2, 1, 3),
  em_step = FALSE, local_step = TRUE
)
fit$model

```

```
## Fix some emissions:

fixB <- matrix(FALSE, 4, 4)
fixB[2, 1] <- fixB[1, 3] <- TRUE # these are fixed to their initial values
fit <- fit_model(model,
  fixed_emissions = fixB,
  em_step = FALSE, local_step = TRUE
)
fit$model$emission_probs
```

---

forward\_backward

*Forward and Backward Probabilities for Hidden Markov Model*


---

### Description

The `forward_backward` function computes scaled forward and backward probabilities of a hidden Markov model.

### Usage

```
forward_backward(model, forward_only = FALSE, log_space = FALSE, threads = 1)
```

### Arguments

<code>model</code>	Object of class <code>hmm</code> or <code>mhmm</code> .
<code>forward_only</code>	If <code>TRUE</code> , only forward probabilities are computed. The default is <code>FALSE</code> .
<code>log_space</code>	Compute forward and backward probabilities in logarithmic scale instead of scaling. The default is <code>FALSE</code> .
<code>threads</code>	Number of threads used in parallel computing. The default is 1.

### Value

List with components

<code>forward_probs</code>	If <code>log_space = FALSE</code> , scaled forward probabilities, i.e. probability of state given observations up to that time point. If <code>log_space = TRUE</code> , logarithms of non-scaled forward probabilities.
<code>backward_probs</code>	Scaled backward probabilities ( <code>log_space = FALSE</code> ), or logarithms of non-scaled backward probabilities ( <code>log_space = TRUE</code> ).
<code>scaling_factors</code>	Sum of non-scaled forward probabilities at each time point. Only computed if <code>log_space = FALSE</code> .

In case of multiple observations, these are computed independently for each sequence.

## Examples

```
# Load a pre-defined MHMM
data("mhmm_biofam")

# Compute forward and backward probabilities
fb <- forward_backward(mhmm_biofam)

# The most probable hidden state at time t
# given the observations up to time t for the first subject:
apply(fb$forward_probs[, , 1], 2, which.max)
```

---

gridplot

*Plot Multidimensional Sequence Plots in a Grid*

---

## Description

Function gridplot plots multiple ssp objects to a grid.

## Usage

```
gridplot(
  x,
  nrow = NA,
  ncol = NA,
  byrow = FALSE,
  with.legend = "auto",
  legend.pos = "auto",
  legend.pos2 = "center",
  title.legend = "auto",
  ncol.legend = "auto",
  with.missing.legend = "auto",
  row.prop = "auto",
  col.prop = "auto",
  cex.legend = 1
)
```

## Arguments

x	A list of <a href="#">ssp</a> objects.
nrow, ncol	Optional arguments to arrange plots.
byrow	Controls the order of plotting. Defaults to FALSE, i.e. plots are arranged column-wise.
with.legend	Defines if and how the legends for the states are plotted. The default value "auto" (equivalent to TRUE and "many") creates separate legends for each requested plot. Other possibilities are "combined" (all legends combined) and FALSE (no legend).

legend.pos	Defines the positions of the legend boxes relative to the whole plot. Either one of "bottom" (equivalent to "auto") or "right", or a numerical vector of grid cells (by order) to print the legends to (the cells must be in one row/column).
legend.pos2	Defines the positions of the legend boxes relative to the cell(s). One of "bottomright", "bottom", "bottomleft", "left", "topleft", "top" (the default), "topright", "right" and "center".
title.legend	The titles for the legend boxes. The default "auto" takes the titles from the channel labels provided by the first object in x. NA prints no title.
ncol.legend	(A vector of) the number of columns for the legend(s). The default "auto" creates one column for each legend.
with.missing.legend	If set to "auto" (the default), a legend for the missing state is added automatically if one or more of the sequences in data contain missing states. With the value TRUE a legend for the missing state is added in any case; equivalently FALSE omits the legend for the missing state.
row.prop	Sets the proportions of the row heights of the grid. The default value is "auto" for even row heights. Takes a vector of values from 0 to 1, with values summing to 1.
col.prop	Sets the proportion of the column heights of the grid. The default value is "auto" for even column widths. Takes a vector of values from 0 to 1, with values summing to 1.
cex.legend	Expansion factor for setting the size of the font for the labels in the legend. The default value is 1. Values lesser than 1 will reduce the size of the font, values greater than 1 will increase the size.

### See Also

[ssp](#) for defining the plot before using `gridplot`, and [plot.ssp](#) for plotting only one `ssp` object.

### Examples

```
## Not run:
data("biofam3c")

# Creating sequence objects
child_seq <- seqdef(biofam3c$children, start = 15)
marr_seq <- seqdef(biofam3c$married, start = 15)
left_seq <- seqdef(biofam3c$left, start = 15)

## Choosing colors
attr(child_seq, "cpal") <- c("#66C2A5", "#FC8D62")
attr(marr_seq, "cpal") <- c("#AB82FF", "#E6AB02", "#E7298A")
attr(left_seq, "cpal") <- c("#A6CEE3", "#E31A1C")

# Preparing plot for state distribution plots of observations for women
ssp_f <- ssp(
  list(
```

```

    child_seq[biofam3c$covariates$sex == "woman", ],
    marr_seq[biofam3c$covariates$sex == "woman", ],
    left_seq[biofam3c$covariates$sex == "woman", ]
  ),
  type = "d", plots = "obs", title = "Women",
  ylab = c("Children", "Married", "Left home")
)

# Preparing plot for state distribution plots of observations for men
# (Updating the previous plot, only arguments that change values)
ssp_m <- update(ssp_f,
  title = "Men",
  x = list(
    child_seq[biofam3c$covariates$sex == "man", ],
    marr_seq[biofam3c$covariates$sex == "man", ],
    left_seq[biofam3c$covariates$sex == "man", ]
  )
)

# Plotting state distribution plots of observations for women and men in two columns
gridplot(list(ssp_f, ssp_m), ncol = 2, with.legend = FALSE)

# Preparing plots for women's state distributions
ssp_f2 <- ssp(
  list(
    marr_seq[biofam3c$covariates$sex == "woman", ],
    child_seq[biofam3c$covariates$sex == "woman", ],
    left_seq[biofam3c$covariates$sex == "woman", ]
  ),
  type = "d", border = NA, with.legend = FALSE,
  title = "State distributions for women", title.n = FALSE, xtlab = 15:30,
  ylab.pos = c(1, 2, 1), ylab = c("Married", "Children", "Left home")
)

# The same plot with sequences instead of state distributions
ssp_f3 <- update(
  ssp_f2,
  type = "I", sortv = "mds.obs", title = "Sequences for women"
)

# State distributions with men's data
ssp_m2 <- update(
  ssp_f2,
  title = "State distributions for men",
  x = list(
    marr_seq[biofam3c$covariates$sex == "man", ],
    child_seq[biofam3c$covariates$sex == "man", ],
    left_seq[biofam3c$covariates$sex == "man", ]
  )
)

# Men's sequences
ssp_m3 <- update(

```

```

    ssp_m2,
    type = "I", sortv = "mds.obs", title = "Sequences for men"
  )

# Plotting state distributions and index plots of observations
# for women and men in two columns (+ one column for legends)
gridplot(
  list(ssp_f2, ssp_f3, ssp_m2, ssp_m3),
  ncol = 3, byrow = TRUE,
  with.legend = "combined", legend.pos = "right", col.prop = c(0.35, 0.35, 0.3)
)

# The same with different positioning and fixed cells for legends
gridplot(
  list(ssp_f2, ssp_f3, ssp_m2, ssp_m3),
  ncol = 2, nrow = 3, byrow = TRUE,
  # defining the legend positions by the cell numbers
  legend.pos = 3:4
)

## End(Not run)

```

---

 hidden\_paths

*Most Probable Paths of Hidden States*


---

### Description

Function `hidden_paths` computes the most probable path of hidden states of a (mixture) hidden Markov model given the observed sequences.

### Usage

```
hidden_paths(model, respect_void = TRUE)
```

### Arguments

<code>model</code>	A hidden Markov model of class <code>hmm</code> or a mixture HMM of class <code>mhmm</code> .
<code>respect_void</code>	If <code>TRUE</code> (default), states at the time points corresponding to TraMineR's void in the observed sequences are set to void in the hidden state sequences as well.

### Value

The most probable paths of hidden states as an `stslist` object (see [seqdef](#)). The log-probability is included as an attribute `log_prob`.

### See Also

[hmm\\_biofam](#) for information on the model used in the example; and [seqIplot](#), [ssplot](#), or [mssplot](#) for plotting hidden paths.



**Examples**

```
# Load a pre-defined HMM
data("hmm_biofam")

# Compute the most probable hidden state paths given the data and the model
mpp <- hidden_paths(hmm_biofam)

# Plot hidden paths for the first 100 individuals
ssplot(mpp, type = "I", tlim = 1:100)

# Because the model structure is so sparse that the posterior probabilities are
# mostly peaked to single state at each time point, the joint probability of
# observations and most probable paths of hidden states is almost identical to
# log-likelihood:

sum(attr(mpp, "log_prob"))
logLik(hmm_biofam)
```

---

hmm\_biofam

*Hidden Markov model for the biofam data*


---

**Description**

A five-state hidden Markov model (HMM) fitted for the [biofam](#) data.

**Format**

A hidden Markov model of class `hmm`; a left-to-right model with four hidden states.

**Details**

The model is loaded by calling `data(hmm_biofam)`. It was created with the following code:

```
data("biofam3c")

# Building sequence objects
marr_seq <- seqdef(biofam3c$married, start = 15,
  alphabet = c("single", "married", "divorced"))
child_seq <- seqdef(biofam3c$children, start = 15,
  alphabet = c("childless", "children"))
left_seq <- seqdef(biofam3c$left, start = 15,
  alphabet = c("with parents", "left home"))

## Choosing colors
attr(marr_seq, "cpal") <- c("violetred2", "darkgoldenrod2", "darkmagenta")
attr(child_seq, "cpal") <- c("darkseagreen1", "coral3")
attr(left_seq, "cpal") <- c("lightblue", "red3")
```

```
init <- c(0.9, 0.05, 0.02, 0.02, 0.01)

# Starting values for transition matrix
trans <- matrix(
  c(0.8, 0.10, 0.05, 0.03, 0.02,
    0, 0.9, 0.05, 0.03, 0.02,
    0, 0, 0.9, 0.07, 0.03,
    0, 0, 0, 0.9, 0.1,
    0, 0, 0, 0, 1),
  nrow = 5, ncol = 5, byrow = TRUE)

# Starting values for emission matrices
emiss_marr <- matrix(
  c(0.9, 0.05, 0.05, # High probability for single
    0.9, 0.05, 0.05,
    0.05, 0.9, 0.05, # High probability for married
    0.05, 0.9, 0.05,
    0.3, 0.3, 0.4), # mixed group
  nrow = 5, ncol = 3, byrow = TRUE)

emiss_child <- matrix(
  c(0.9, 0.1, # High probability for childless
    0.9, 0.1,
    0.1, 0.9,
    0.1, 0.9,
    0.5, 0.5),
  nrow = 5, ncol = 2, byrow = TRUE)

emiss_left <- matrix(
  c(0.9, 0.1, # High probability for living with parents
    0.1, 0.9,
    0.1, 0.9,
    0.1, 0.9,
    0.5, 0.5),
  nrow = 5, ncol = 2, byrow = TRUE)

initmod <- build_hmm(
  observations = list(marr_seq, child_seq, left_seq),
  initial_probs = init, transition_probs = trans,
  emission_probs = list(emiss_marr, emiss_child,
    emiss_left),
  channel_names = c("Marriage", "Parenthood", "Residence"))

fit_biofam <- fit_model(initmod, em = FALSE, local = TRUE)
hmm_biofam <- fit_biofam$model
```

**See Also**

Examples of building and fitting HMMs in [build\\_hmm](#) and [fit\\_model](#); and [biofam](#) for the original data and [biofam3c](#) for the three-channel version used in this model.

**Examples**

```
# Plotting the model
plot(hmm_biofam)
```

---

hmm\_mvad

*Hidden Markov model for the mvad data*


---

**Description**

A hidden Markov model (MMM) fitted for the [mvad](#) data.

**Format**

A hidden Markov model of class `hmm`; unrestricted model with six hidden states.

**Details**

Model was created with the following code:

```
data("mvad", package = "TraMineR")

mvad_alphabet <-
  c("employment", "FE", "HE", "joblessness", "school", "training")
mvad_labels <- c("employment", "further education", "higher education",
  "joblessness", "school", "training")
mvad_scodes <- c("EM", "FE", "HE", "JL", "SC", "TR")
mvad_seq <- seqdef(mvad, 17:86, alphabet = mvad_alphabet,
  states = mvad_scodes, labels = mvad_labels, xtstep = 6)

attr(mvad_seq, "cpal") <- colorpalette[[6]]

# Starting values for the emission matrix
emiss <- matrix(
  c(0.05, 0.05, 0.05, 0.05, 0.75, 0.05, # SC
    0.05, 0.75, 0.05, 0.05, 0.05, 0.05, # FE
    0.05, 0.05, 0.05, 0.4, 0.05, 0.4, # JL, TR
    0.05, 0.05, 0.75, 0.05, 0.05, 0.05, # HE
    0.75, 0.05, 0.05, 0.05, 0.05, 0.05), # EM
  nrow = 5, ncol = 6, byrow = TRUE)
```

```

# Starting values for the transition matrix
trans <- matrix(0.025, 5, 5)
diag(trans) <- 0.9

# Starting values for initial state probabilities
initial_probs <- c(0.2, 0.2, 0.2, 0.2, 0.2)

# Building a hidden Markov model
init_hmm_mvad <- build_hmm(observations = mvad_seq,
  transition_probs = trans, emission_probs = emiss,
  initial_probs = initial_probs)

set.seed(21)
fit_hmm_mvad <- fit_model(init_hmm_mvad, control_em = list(restart = list(times = 100)))
hmm_mvad <- fit_hmm_mvad$model

```

**See Also**

Examples of building and fitting HMMs in [build\\_hmm](#) and [fit\\_model](#); and [mvad](#) for more information on the data.

**Examples**

```

data("hmm_mvad")

# Plotting the model
plot(hmm_mvad)

```

---

logLik.hmm

*Log-likelihood of the Hidden Markov Model*


---

**Description**

Function `logLik.hmm` computes the log-likelihood value of a hidden Markov model.

**Usage**

```

## S3 method for class 'hmm'
logLik(object, partials = FALSE, threads = 1, log_space = FALSE, ...)

```

**Arguments**

<code>object</code>	A hidden Markov model of class <code>hmm</code> .
<code>partials</code>	Return a vector containing the individual contributions of each sequence to the total log-likelihood. The default is <code>FALSE</code> , which returns the sum of all log-likelihood components.

threads	Number of threads to use in parallel computing. The default is 1.
log_space	Make computations using log-space instead of scaling for greater numerical stability at the cost of decreased computational performance. The default is TRUE.
...	Ignored.

**Value**

Log-likelihood of the hidden Markov model. This is an object of class `logLik` with attributes `nobs` and `df` inherited from the model object.

**See Also**

[build\\_hmm](#) and [fit\\_model](#) for building and fitting Hidden Markov models.

---

logLik.mhmm

---

*Log-likelihood of the Mixture Hidden Markov Model*


---

**Description**

Function `logLik.mhmm` computes the log-likelihood value of a mixture hidden Markov model.

**Usage**

```
## S3 method for class 'mhmm'
logLik(object, partials = FALSE, threads = 1, log_space = FALSE, ...)
```

**Arguments**

object	A mixture hidden Markov model of class <code>mhmm</code> .
partials	Return a vector containing the individual contributions of each sequence to the total log-likelihood. The default is FALSE, which returns the sum of all log-likelihood components.
threads	Number of threads to use in parallel computing. The default is 1.
log_space	Make computations using log-space instead of scaling for greater numerical stability at the cost of decreased computational performance. The default is TRUE.
...	Ignored.

**Value**

Log-likelihood of the mixture hidden Markov model. This is an object of class `logLik` with attributes `nobs` and `df` inherited from the model object.

**See Also**

[build\\_mhmm](#) and [fit\\_model](#) for building and fitting mixture Hidden Markov models.

---

mc_to_sc	<i>Transform a Multichannel Hidden Markov Model into a Single Channel Representation</i>
----------	--

---

### Description

Transforms data and parameters of a multichannel model into a single channel model. Observed states (symbols) are combined and parameters multiplied across channels.

### Usage

```
mc_to_sc(model, combine_missing = TRUE, all_combinations = FALSE, cpal)
```

### Arguments

model	An object of class <code>hmm</code> or <code>mhmm</code> .
combine_missing	Controls whether combined states of observations at time $t$ are coded missing (coded with <code>*</code> in <code>stslists</code> ) if one or more of the channels include missing information at time $t$ . Defaults to <code>TRUE</code> . <code>FALSE</code> keeps missing states as they are, producing more states in data; e.g. <code>single/childless/*</code> where the observation in channel 3 is missing.
all_combinations	Controls whether all possible combinations of observed states are included in the single channel representation or only combinations that are found in the data. Defaults to <code>FALSE</code> , i.e. only actual observations are included.
cpal	The color palette used for the new combined symbols. Optional in a case where the number of symbols is less or equal to 200 (in which case the <code>seqHMM::colorpalette</code> is used).

### Details

Note that in case of no missing observations, the log-likelihood of the original and transformed models are identical but the AIC and BIC can be different as the model attribute `df` is recomputed based on the single channel representation.

### See Also

[build\\_hmm](#) and [fit\\_model](#) for building and fitting Hidden Markov models; and [hmm\\_biofam](#) for information on the model used in the example.

### Examples

```
# Loading a hidden Markov model of the biofam data (hmm object)
data("hmm_biofam")

# Convert the multichannel model to a single-channel model
```

```

sc <- mc_to_sc(hmm_biofam)

# Likelihoods of the single-channel and the multichannel model are the same
# (Might not be true if there are missing observations)
logLik(sc)
logLik(hmm_biofam)

```

---

mc_to_sc_data	<i>Merge Multiple Sequence Objects into One (from Multichannel to Single Channel Data)</i>
---------------	--

---

## Description

Function `mc_to_sc_data` combines observed states of multiple sequence objects into one, time point by time point.

## Usage

```
mc_to_sc_data(data, combine_missing = TRUE, all_combinations = FALSE, cpal)
```

## Arguments

<code>data</code>	A list of state sequence objects ( <code>stslists</code> ) created with the <a href="#">seqdef</a> function.
<code>combine_missing</code>	Controls whether combined states of observations at time <code>t</code> are coded missing (coded with <code>*</code> in <code>stslists</code> ) if one or more of the channels include missing information at time <code>t</code> . Defaults to <code>TRUE</code> . <code>FALSE</code> keeps missing states as they are, producing more states in data; e.g. <code>single/childless/*</code> where the observation in channel 3 is missing.
<code>all_combinations</code>	Controls whether all possible combinations of observed states are included in the single channel representation or only combinations that are found in the data. Defaults to <code>FALSE</code> , i.e. only actual observations are included.
<code>cpal</code>	The color palette used for the new combined symbols. Optional in a case where the number of symbols is less or equal to 200 (in which case the <code>seqHMM:::colorpalette</code> is used).

## See Also

[mc\\_to\\_sc](#) for transforming multichannel `hmm` or `mhmm` objects into single-channel representations; [ssplot](#) for plotting multiple sequence data sets in the same plot; and [seqdef](#) for creating state sequence objects.

**Examples**

```

# Load three-channel sequence data
data("biofam3c")

# Building sequence objects
marr_seq <- seqdef(biofam3c$marrried,
  start = 15,
  alphabet = c("single", "married", "divorced")
)
child_seq <- seqdef(biofam3c$children,
  start = 15,
  alphabet = c("childless", "children")
)
left_seq <- seqdef(biofam3c$left,
  start = 15,
  alphabet = c("with parents", "left home")
)

# Define colors
attr(marr_seq, "cpal") <- c("violetred2", "darkgoldenrod2", "darkmagenta")
attr(child_seq, "cpal") <- c("darkseagreen1", "coral3")
attr(left_seq, "cpal") <- c("lightblue", "red3")

# Converting multichannel data to single-channel data
sc_data <- mc_to_sc_data(list(marr_seq, child_seq, left_seq))

# 10 combined states
alphabet(sc_data)

# Colors for combined states
attr(sc_data, "cpal") <- colorpalette[[14]][1:10]

# Plotting sequences for the first 10 subjects
ssplot(
  list(
    "Marriage" = marr_seq, "Parenthood" = child_seq,
    "Residence" = left_seq, "Combined" = sc_data
  ),
  type = "I",
  tlim = 1:10
)

# Including all combinations (whether or not available in data)
sc_data_all <- mc_to_sc_data(list(marr_seq, child_seq, left_seq),
  all_combinations = TRUE
)

# 12 combined states, 2 with no observations in data
seqstatf(sc_data_all)

```



**Description**

A mixture hidden Markov model (MHMM) fitted for the `biofam` data.

**Format**

A mixture hidden Markov model of class `mhmm`: three clusters with left-to-right models including 4, 4, and 6 hidden states. Two covariates, `sex` and `cohort`, explaining the cluster membership.

**Details**

The model was created with the following code:

```
data("biofam3c")

## Building sequence objects
marr_seq <- seqdef(biofam3c$married, start = 15,
  alphabet = c("single", "married", "divorced"))
child_seq <- seqdef(biofam3c$children, start = 15,
  alphabet = c("childless", "children"))
left_seq <- seqdef(biofam3c$left, start = 15,
  alphabet = c("with parents", "left home"))

## Choosing colors
attr(marr_seq, "cpal") <- c("#AB82FF", "#E6AB02", "#E7298A")
attr(child_seq, "cpal") <- c("#66C2A5", "#FC8D62")
attr(left_seq, "cpal") <- c("#A6CEE3", "#E31A1C")

## Starting values for emission probabilities
# Cluster 1
B1_marr <- matrix(
  c(0.8, 0.1, 0.1, # High probability for single
    0.8, 0.1, 0.1,
    0.3, 0.6, 0.1, # High probability for married
    0.3, 0.3, 0.4), # High probability for divorced
  nrow = 4, ncol = 3, byrow = TRUE)

B1_child <- matrix(
  c(0.9, 0.1, # High probability for childless
    0.9, 0.1,
    0.9, 0.1,
    0.9, 0.1),
  nrow = 4, ncol = 2, byrow = TRUE)
```

```
B1_left <- matrix(
  c(0.9, 0.1, # High probability for living with parents
    0.1, 0.9, # High probability for having left home
    0.1, 0.9,
    0.1, 0.9),
  nrow = 4, ncol = 2, byrow = TRUE)

# Cluster 2

B2_marr <- matrix(
  c(0.8, 0.1, 0.1, # High probability for single
    0.8, 0.1, 0.1,
    0.1, 0.8, 0.1, # High probability for married
    0.7, 0.2, 0.1),
  nrow = 4, ncol = 3, byrow = TRUE)

B2_child <- matrix(
  c(0.9, 0.1, # High probability for childless
    0.9, 0.1,
    0.9, 0.1,
    0.1, 0.9),
  nrow = 4, ncol = 2, byrow = TRUE)

B2_left <- matrix(
  c(0.9, 0.1, # High probability for living with parents
    0.1, 0.9,
    0.1, 0.9,
    0.1, 0.9),
  nrow = 4, ncol = 2, byrow = TRUE)

# Cluster 3

B3_marr <- matrix(
  c(0.8, 0.1, 0.1, # High probability for single
    0.8, 0.1, 0.1,
    0.8, 0.1, 0.1,
    0.1, 0.8, 0.1, # High probability for married
    0.3, 0.4, 0.3,
    0.1, 0.1, 0.8), # High probability for divorced
  nrow = 6, ncol = 3, byrow = TRUE)

B3_child <- matrix(
  c(0.9, 0.1, # High probability for childless
    0.9, 0.1,
    0.5, 0.5,
    0.5, 0.5,
    0.5, 0.5,
    0.1, 0.9),
  nrow = 6, ncol = 2, byrow = TRUE)
```

```

B3_left <- matrix(
  c(0.9, 0.1, # High probability for living with parents
    0.1, 0.9,
    0.5, 0.5,
    0.5, 0.5,
    0.1, 0.9,
    0.1, 0.9),
  nrow = 6, ncol = 2, byrow = TRUE)

# Starting values for transition matrices
A1 <- matrix(
  c(0.80, 0.16, 0.03, 0.01,
    0, 0.90, 0.07, 0.03,
    0, 0, 0.90, 0.10,
    0, 0, 0, 1),
  nrow = 4, ncol = 4, byrow = TRUE)

A2 <- matrix(
  c(0.80, 0.10, 0.05, 0.03, 0.01, 0.01,
    0, 0.70, 0.10, 0.10, 0.05, 0.05,
    0, 0, 0.85, 0.01, 0.10, 0.04,
    0, 0, 0, 0.90, 0.05, 0.05,
    0, 0, 0, 0, 0.90, 0.10,
    0, 0, 0, 0, 0, 1),
  nrow = 6, ncol = 6, byrow = TRUE)

# Starting values for initial state probabilities
initial_probs1 <- c(0.9, 0.07, 0.02, 0.01)
initial_probs2 <- c(0.9, 0.04, 0.03, 0.01, 0.01, 0.01)

# Birth cohort
biofam3c$covariates$cohort <- factor(cut(biofam3c$covariates$birthyr,
  c(1908, 1935, 1945, 1957)), labels = c("1909-1935", "1936-1945", "1946-1957"))

# Build mixture HMM
init_mhmm_bf <- build_mhmm(
  observations = list(marr_seq, child_seq, left_seq),
  initial_probs = list(initial_probs1, initial_probs1, initial_probs2),
  transition_probs = list(A1, A1, A2),
  emission_probs = list(list(B1_marr, B1_child, B1_left),
    list(B2_marr, B2_child, B2_left),
    list(B3_marr, B3_child, B3_left)),
  formula = ~sex + cohort, data = biofam3c$covariates,
  channel_names = c("Marriage", "Parenthood", "Residence"))

# Fitting the model

```

```
mhmm_biofam <- fit_model(init_mhmm_bf)$model
```

### See Also

Examples of building and fitting MHMMs in [build\\_mhmm](#) and [fit\\_model](#); and [biofam](#) for the original data and [biofam3c](#) for the three-channel version used in this model.

### Examples

```
data("mhmm_biofam")

# use conditional_se = FALSE for more accurate standard errors
# (these are considerably slower to compute)
summary(mhmm_biofam$model)

if (interactive()) {
  # Plotting the model for each cluster (change with Enter)
  plot(mhmm_biofam)
}
```

---

mhmm\_mvad

*Mixture hidden Markov model for the mvad data*

---

### Description

A mixture hidden Markov model (MHMM) fitted for the [mvad](#) data.

### Format

A mixture hidden Markov model of class mhmm: two clusters including 3 and 4 hidden states. No covariates.

### Details

The model is loaded by calling `data(mhmm_mvad)`. It was created with the following code:

```
data("mvad", package = "TraMineR")

mvad_alphabet <-
  c("employment", "FE", "HE", "joblessness", "school", "training")
mvad_labels <- c("employment", "further education", "higher education",
  "joblessness", "school", "training")
mvad_scodes <- c("EM", "FE", "HE", "JL", "SC", "TR")
mvad_seq <- seqdef(mvad, 17:86, alphabet = mvad_alphabet,
  states = mvad_scodes, labels = mvad_labels, xtstep = 6)

attr(mvad_seq, "cpal") <- colorpalette[[6]]
```

```

# Starting values for the emission matrices
emiss_1 <- matrix(
  c(0.01, 0.01, 0.01, 0.01, 0.01, 0.95,
    0.95, 0.01, 0.01, 0.01, 0.01, 0.01,
    0.01, 0.01, 0.01, 0.95, 0.01, 0.01),
  nrow = 3, ncol = 6, byrow = TRUE)

emiss_2 <- matrix(
  c(0.01, 0.01, 0.01, 0.06, 0.90, 0.01,
    0.01, 0.95, 0.01, 0.01, 0.01, 0.01,
    0.01, 0.01, 0.95, 0.01, 0.01, 0.01,
    0.95, 0.01, 0.01, 0.01, 0.01, 0.01),
  nrow = 4, ncol = 6, byrow = TRUE)

# Starting values for the transition matrix

trans_1 <- matrix(
  c(0.95, 0.03, 0.02,
    0.01, 0.98, 0.01,
    0.01, 0.01, 0.98),
  nrow = 3, ncol = 3, byrow = TRUE)

trans_2 <- matrix(
  c(0.97, 0.01, 0.01, 0.01,
    0.01, 0.97, 0.01, 0.01,
    0.01, 0.01, 0.97, 0.01,
    0.01, 0.01, 0.01, 0.97),
  nrow = 4, ncol = 4, byrow = TRUE)

# Starting values for initial state probabilities
initial_probs_1 <- c(0.5, 0.25, 0.25)
initial_probs_2 <- c(0.4, 0.4, 0.1, 0.1)

# Building a hidden Markov model with starting values
init_mhmm_mvad <- build_mhmm(observations = mvad_seq,
  transition_probs = list(trans_1, trans_2),
  emission_probs = list(emiss_1, emiss_2),
  initial_probs = list(initial_probs_1, initial_probs_2))

# Fit the model
set.seed(123)
mhmm_mvad <- fit_model(init_mhmm_mvad, control_em = list(restart = list(times = 25)))$model

```

### See Also

Examples of building and fitting MHMMs in [build\\_mhmm](#) and [fit\\_model](#); and [mvad](#) for more information on the data.

## Examples

```
data("mhmm_mvad")

summary(mhmm_mvad)

if (interactive()) {
  # Plotting the model for each cluster (change with Enter)
  plot(mhmm_mvad)
}
```

---

mssplot

*Interactive Stacked Plots of Multichannel Sequences and/or Most Probable Paths for Mixture Hidden Markov Models*

---

## Description

Function `mssplot` plots stacked sequence plots of observation sequences and/or most probable hidden state paths for each model of the `mhmm` object (model chosen according to the most probable path).

## Usage

```
mssplot(
  x,
  ask = FALSE,
  which.plots = NULL,
  hidden.paths = NULL,
  plots = "obs",
  type = "d",
  tlim = 0,
  sortv = NULL,
  sort.channel = 1,
  dist.method = "OM",
  with.missing = FALSE,
  missing.color = NULL,
  title = NA,
  title.n = TRUE,
  cex.title = 1,
  title.pos = 1,
  with.legend = "auto",
  ncol.legend = "auto",
  with.missing.legend = "auto",
  legend.prop = 0.3,
  cex.legend = 1,
  hidden.states.colors = "auto",
  hidden.states.labels = "auto",
```

```

    xaxis = TRUE,
    xlab = NA,
    xtlab = NULL,
    xlab.pos = 1,
    ylab = "auto",
    hidden.states.title = "Hidden states",
    yaxis = FALSE,
    ylab.pos = "auto",
    cex.lab = 1,
    cex.axis = 1,
    respect_void = TRUE,
    ...
)

```

### Arguments

<code>x</code>	Mixture hidden Markov model object of class <code>mhmm</code> .
<code>ask</code>	If TRUE and <code>which.plots</code> is NULL, <code>plot.mhmm</code> operates in interactive mode, via <a href="#">menu</a> . Defaults to FALSE.
<code>which.plots</code>	The number(s) of the requested model(s) as an integer vector. The default NULL produces all plots.
<code>hidden.paths</code>	Output from the <a href="#">hidden_paths</a> function. The default value NULL computes hidden paths automatically, if needed.
<code>plots</code>	What to plot. One of "obs" for observations (the default), "hidden.paths" for most probable paths of hidden states, or "both" for observations and hidden paths together.
<code>type</code>	The type of the plot. Available types are "I" for index plots and "d" for state distribution plots (the default). See <a href="#">seqplot</a> for details.
<code>tlim</code>	Indexes of the subjects to be plotted (the default is 0, i.e. all subjects are plotted). For example, <code>tlim = 1:10</code> plots the first ten subjects in data.
<code>sortv</code>	A sorting variable or a sort method (one of "from.start", "from.end", "mds.obs", or "mds.hidden") for <code>type = "I"</code> . The value "mds.hidden" is only available when <code>which = "both"</code> and <code>which = "hidden.paths"</code> . Options "mds.obs" and "mds.hidden" automatically arrange the sequences according to the scores of multidimensional scaling (using <a href="#">cmdscale</a> ) for the observed data or hidden states paths. MDS scores are computed from distances/dissimilarities using a metric defined in argument <code>dist.method</code> . See <a href="#">plot.stslist</a> for more details on "from.start" and "from.end".
<code>sort.channel</code>	The number of the channel according to which the "from.start" or "from.end" sorting is done. Sorting according to hidden states is called with value 0. The default value is 1 (the first channel).
<code>dist.method</code>	The metric to be used for computing the distances of the sequences if multidimensional scaling is used for sorting. One of "OM" (optimal matching, the default), "LCP" (longest common prefix), "RLCP" (reversed LCP, i.e. longest common suffix), "LCS" (longest common subsequence), "HAM" (Hamming distance), and "DHD" (dynamic Hamming distance). Transition rates are used

	for defining substitution costs if needed. See <a href="#">seqdef</a> for more information on the metrics.
<code>with.missing</code>	Controls whether missing states are included in state distribution plots ( <code>type = "d"</code> ). The default is <code>FALSE</code> .
<code>missing.color</code>	Alternative color for representing missing values in the sequences. By default, this color is taken from the <code>missing.color</code> attribute of the sequence object.
<code>title</code>	A vector of main titles for the graphics. The default is <code>NA</code> : if <code>title.n = TRUE</code> , the name of the cluster and the number of subjects is plotted. <code>FALSE</code> prints no titles, even when <code>title.n = TRUE</code> .
<code>title.n</code>	Controls whether the number of subjects is printed in the main titles of the plots. The default is <code>TRUE</code> : <code>n</code> is plotted if <code>title</code> is anything but <code>FALSE</code> .
<code>cex.title</code>	Expansion factor for setting the size of the font for the main titles. The default value is 1. Values lesser than 1 will reduce the size of the font, values greater than 1 will increase the size.
<code>title.pos</code>	Controls the position of the main titles of the plots. The default value is 1. Values greater than 1 will place the title higher.
<code>with.legend</code>	Defines if and where the legend for the states is plotted. The default value "auto" (equivalent to <code>TRUE</code> and "right") creates separate legends for each requested plot and positions them on the right-hand side of the plot. Other possible values are "bottom", "right.combined", and "bottom.combined", of which the last two create a combined legend in the selected position. <code>FALSE</code> prints no legend.
<code>ncol.legend</code>	(A vector of) the number of columns for the legend(s). The default "auto" creates one column for each legend.
<code>with.missing.legend</code>	If set to "auto" (the default), a legend for the missing state is added automatically if one or more of the sequences in the data/channel contains missing states and <code>type = "I"</code> . If <code>type = "d"</code> missing states are omitted from the legends unless <code>with.missing = TRUE</code> . With the value <code>TRUE</code> a legend for the missing state is added in any case; equivalently <code>FALSE</code> omits the legend for the missing state.
<code>legend.prop</code>	Sets the proportion of the graphic area used for plotting the legend when <code>with.legend</code> is not <code>FALSE</code> . The default value is 0.3. Takes values from 0 to 1.
<code>cex.legend</code>	Expansion factor for setting the size of the font for the labels in the legend. The default value is 1. Values lesser than 1 will reduce the size of the font, values greater than 1 will increase the size.
<code>hidden.states.colors</code>	A vector of colors assigned to hidden states. The default value "auto" uses the colors assigned to the <code>stlist</code> object (created with <a href="#">seqdef</a> ) if <code>hidden.paths</code> is given; otherwise colors from <a href="#">colorpalette</a> are automatically used.
<code>hidden.states.labels</code>	Labels for the hidden states. The default value "auto" uses the names provided in <code>x\$state_names</code> if <code>x</code> is an <code>hmm</code> object; otherwise the number of the hidden state.
<code>xaxis</code>	Controls whether an x-axis is plotted below the plot at the bottom. The default value is <code>TRUE</code> .



<code>xlab</code>	An optional label for the x-axis. If set to NA, no label is drawn.
<code>xtlab</code>	Optional labels for the x-axis tick labels. If unspecified, the column names of the <code>seqdata</code> sequence object are used (see <a href="#">seqdef</a> ).
<code>xlab.pos</code>	Controls the position of the x-axis label. The default value is 1. Values greater than 1 will place the label further away from the plot.
<code>ylab</code>	Labels for the channels shown as labels for y-axes. A vector of names for each channel (observations). The default value "auto" uses the names provided in <code>x\$channel_names</code> if <code>x</code> is an <code>hmm</code> object; otherwise the names of the list in <code>x</code> if given, or the number of the channel if names are not given. FALSE prints no labels.
<code>hidden.states.title</code>	Optional label for the hidden state plot (in the y-axis). The default is "Hidden states".
<code>yaxis</code>	Controls whether or not to plot the y-axis. The default is FALSE.
<code>ylab.pos</code>	Controls the position of the y axis labels (labels for channels and/or hidden states). Either "auto" or a numerical vector indicating how far away from the plots the titles are positioned. The default value "auto" positions all titles on line 1. Shorter vectors are recycled.
<code>cex.lab</code>	Expansion factor for setting the size of the font for the axis labels. The default value is 1. Values lesser than 1 will reduce the size of the font, values greater than 1 will increase the size.
<code>cex.axis</code>	Expansion factor for setting the size of the font for the x-axis tick labels. The default value is 1. Values lesser than 1 will reduce the size of the font, values greater than 1 will increase the size.
<code>respect_void</code>	If TRUE (default), states at the time points corresponding to TraMineR's void in the observed sequences are set to void in the hidden state sequences as well.
<code>...</code>	Other arguments to be passed on to <a href="#">seqplot</a> .

### See Also

[build\\_mhmm](#) and [fit\\_model](#) for building and fitting mixture hidden Markov models, [hidden\\_paths](#) for computing the most probable paths (Viterbi paths) of hidden states, [plot.mhmm](#) for plotting `mhmm` objects as directed graphs, and [colorpalette](#) for default colors.

### Examples

```
# Loading mixture hidden Markov model (mhmm object)
# of the biofam data
data("mhmm_biofam")

# Plotting the first cluster only
mssplot(mhmm_biofam, which.plots = 1)

if (interactive()) {
  # Interactive plot
  mssplot(mhmm_biofam)
}
```

---

`plot.hmm`*Plot hidden Markov models*

---

**Description**

Function `plot.hmm` plots a directed graph with pie charts of emission probabilities as vertices/nodes.

**Usage**

```
## S3 method for class 'hmm'
plot(
  x,
  layout = "horizontal",
  pie = TRUE,
  vertex.size = 40,
  vertex.label = "initial.probs",
  vertex.label.dist = "auto",
  vertex.label.pos = "bottom",
  vertex.label.family = "sans",
  loops = FALSE,
  edge.curved = TRUE,
  edge.label = "auto",
  edge.width = "auto",
  cex.edge.width = 1,
  edge.arrow.size = 1.5,
  edge.label.family = "sans",
  label.signif = 2,
  label.scientific = FALSE,
  label.max.length = 6,
  trim = 1e-15,
  combine.slices = 0.05,
  combined.slice.color = "white",
  combined.slice.label = "others",
  with.legend = "bottom",
  ltext = NULL,
  legend.prop = 0.5,
  cex.legend = 1,
  ncol.legend = "auto",
  cpal = "auto",
  cpal.legend = "auto",
  legend.order = TRUE,
  main = NULL,
  withlegend,
  ...
)
```

**Arguments**

<code>x</code>	A hidden Markov model object of class <code>hmm</code> created with <code>build_hmm</code> (or <code>build_mm</code> ). Multichannel <code>hmm</code> objects are automatically transformed into single-channel objects. See function <code>mc_to_sc</code> for more information on the transformation.
<code>layout</code>	specifies the layout of vertices (nodes). Accepts a numerical matrix, a <code>layout_</code> function (without quotation marks), or either of the predefined options "horizontal" (the default) and "vertical". Options "horizontal" and "vertical" position vertices at the same horizontal or vertical line. A two-column numerical matrix can be used to give x and y coordinates of the vertices. The <code>layout_</code> functions available in the <code>igraph</code> package offer other automatic layouts for graphs.
<code>pie</code>	Are vertices plotted as pie charts of emission probabilities? Defaults to TRUE.
<code>vertex.size</code>	Size of vertices, given as a scalar or numerical vector. The default value is 40.
<code>vertex.label</code>	Labels for vertices. Possible options include "initial.probs", "names", NA, and a character or numerical vector. The default "initial.probs" prints the initial probabilities of the model and "names" prints the names of the hidden states as labels. NA prints no labels.
<code>vertex.label.dist</code>	Distance of the label of the vertex from its center. The default value "auto" places the label outside the vertex.
<code>vertex.label.pos</code>	Positions of vertex labels, relative to the center of the vertex. A scalar or numerical vector giving position(s) as radians or one of "bottom" ( $\pi/2$ as radians), "top" ( $-\pi/2$ ), "left" ( $\pi$ ), or "right" ( $0$ ).
<code>vertex.label.family</code> , <code>edge.label.family</code>	Font family to be used for vertex/edge labels. See argument <code>family</code> in <code>par</code> for more information.
<code>loops</code>	Defines whether transitions back to same states are plotted.
<code>edge.curved</code>	Defines whether to plot curved edges (arcs, arrows) between vertices. A logical or numerical vector or scalar. Numerical values specify curvatures of edges. The default value TRUE gives curvature of 0.5 to all edges. See <code>igraph.plotting</code> for more information.
<code>edge.label</code>	Labels for edges. Possible options include "auto", NA, and a character or numerical vector. The default "auto" prints transition probabilities as edge labels. NA prints no labels.
<code>edge.width</code>	Width(s) for edges. The default "auto" determines widths according to transition probabilities between hidden states. Other possibilities are a scalar or a numerical vector of widths.
<code>cex.edge.width</code>	An expansion factor for edge widths. Defaults to 1.
<code>edge.arrow.size</code>	Size of the arrow in edges (constant). Defaults to 1.5.
<code>label.signif</code>	Rounds labels of model parameters to specified number of significant digits, 2 by default. Ignored for user-given labels.

<code>label.scientific</code>	Defines if scientific notation should be used to describe small numbers. Defaults to FALSE, e.g. 0.0001 instead of 1e-04. Ignored for user-given labels.
<code>label.max.length</code>	Maximum number of digits in labels of model parameters. Ignored for user-given labels.
<code>trim</code>	Scalar between 0 and 1 giving the highest probability of transitions that are plotted as edges, defaults to 1e-15.
<code>combine.slices</code>	Scalar between 0 and 1 giving the highest probability of emission probabilities that are combined into one state. The default value is 0.05.
<code>combined.slice.color</code>	Color of the combined slice that includes the smallest emission probabilities (only if argument "combine.slices" is greater than 0). The default color is white.
<code>combined.slice.label</code>	The label for combined states (when argument "combine.slices" is greater than 0) to appear in the legend.
<code>with.legend</code>	Defines if and where the legend of state colors is plotted. Possible values include "bottom" (the default), "top", "left", and "right". FALSE omits the legend.
<code>ltext</code>	Optional description of (combined) observed states to appear in the legend. A vector of character strings. See <a href="#">seqplot</a> for more information.
<code>legend.prop</code>	Proportion used for plotting the legend. A scalar between 0 and 1, defaults to 0.5.
<code>cex.legend</code>	Expansion factor for setting the size of the font for labels in the legend. The default value is 1. Values lesser than 1 will reduce the size of the font, values greater than 1 will increase the size.
<code>ncol.legend</code>	The number of columns for the legend. The default value "auto" sets the number of columns automatically.
<code>cpal</code>	Optional color palette for (combinations of) observed states. The default value "auto" uses automatic color palette. Otherwise a vector of length $\times n_{\text{symbols}}$ is given, i.e. the argument requires a color specified for all (combinations of) observed states even if they are not plotted (if the probability is less than <code>combine.slices</code> ).
<code>cpal.legend</code>	Optional color palette for the legend, only considered when <code>legend.order</code> is FALSE. Should match <code>ltext</code> .
<code>legend.order</code>	Whether to use the default order in the legend, i.e., order by appearance (first by hidden state, then by emission probability). TRUE by default.
<code>main</code>	Main title for the plot. Omitted by default.
<code>withlegend</code>	Deprecated. Use <code>with.legend</code> instead.
<code>...</code>	Other parameters passed on to <a href="#">plot.igraph</a> such as <code>vertex.color</code> , <code>vertex.label.cex</code> , or <code>edge.lty</code> .

### See Also

[build\\_hmm](#) and [fit\\_model](#) for building and fitting Hidden Markov models, [mc\\_to\\_sc](#) for transforming multistate hmm objects into single-channel objects, [hmm\\_biofam](#) and [hmm\\_mvad](#) for information on the models used in the examples, and [plot.igraph](#) for the general plotting function of directed graphs.

**Examples**

```

# Multichannel data, left-to-right model

# Loading a HMM of the biofam data
data("hmm_biofam")

# Plotting hmm object
plot(hmm_biofam)

# Plotting HMM with
plot(hmm_biofam,
     # varying curvature of edges
     edge.curved = c(0, -0.7, 0.6, 0.7, 0, -0.7, 0),
     # legend with two columns and less space
     ncol.legend = 2, legend.prop = 0.4,
     # new label for combined slice
     combined.slice.label = "States with probability < 0.05"
)

# Plotting HMM with given coordinates
plot(hmm_biofam,
     # layout given in 2x5 matrix
     # x coordinates in the first column
     # y coordinates in the second column
     layout = matrix(c(
       1, 3, 3, 5, 3,
       0, 0, 1, 0, -1
     ), ncol = 2),
     # larger vertices
     vertex.size = 50,
     # straight edges
     edge.curved = FALSE,
     # thinner edges and arrows
     cex.edge.width = 0.5, edge.arrow.size = 1,
     # varying positions for vertex labels (initial probabilities)
     vertex.label.pos = c(pi, pi / 2, -pi / 2, 0, pi / 2),
     # different legend properties
     with.legend = "top", legend.prop = 0.3, cex.legend = 1.1,
     # Fix axes to the right scale
     xlim = c(0.5, 5.5), ylim = c(-1.5, 1.5), rescale = FALSE,
     # all states (not combining states with small probabilities)
     combine.slices = 0,
     # legend with two columns
     ncol.legend = 2
)

# Plotting HMM with own color palette
plot(hmm_biofam,
     cpal = 1:10,
     # States with emission probability less than 0.2 removed
     combine.slices = 0.2,
     # legend with two columns

```

```

    ncol.legend = 2
  )

# Plotting HMM without pie graph and with a layout function
require("igraph")
# Setting the seed for a random layout
set.seed(1234)
plot(hmm_biofam,
     # Without pie graph
     pie = FALSE,
     # Using an automatic layout function from igraph
     layout = layout_nicely,
     vertex.size = 30,
     # Straight edges and probabilities of moving to the same state
     edge.curved = FALSE, loops = TRUE,
     # Labels with three significant digits
     label.signif = 3,
     # Fixed edge width
     edge.width = 1,
     # Remove edges with probability less than 0.01
     trim = 0.01,
     # Hidden state names as vertex labels
     vertex.label = "names",
     # Labels inside vertices
     vertex.label.dist = 0,
     # Fix x-axis (more space on the right-hand side)
     xlim = c(-1, 1.3)
)

# Single-channel data, unrestricted model

# Loading a hidden Markov model of the mvad data (hmm object)
data("hmm_mvad")

# Plotting the HMM
plot(hmm_mvad)

# Checking the order of observed states (needed for the next call)
require(TraMineR)
alphabet(hmm_mvad$observations)

# Plotting the HMM with own legend (note: observation "none" nonexistent in the observations)
plot(hmm_mvad,
     # Override the default order in the legend
     legend.order = FALSE,
     # Colours in the pies (ordered by the alphabet of observations)
     cpal = c("purple", "pink", "brown", "lightblue", "orange", "green"),
     # Colours in the legend (matching to ltext)
     cpal.legend = c("orange", "pink", "brown", "green", "lightblue", "purple", "gray"),
     # Labels in the legend (matching to cpal.legend)
     ltext = c("school", "further educ", "higher educ", "training", "jobless", "employed", "none")
)

```

```

require("igraph")
plot(hmm_mvad,
  # Layout in circle (layout function from igraph)
  layout = layout_in_circle,
  # Less curved edges with smaller arrows, no labels
  edge.curved = 0.2, edge.arrow.size = 0.9, edge.label = NA,
  # Positioning vertex labels (initial probabilities)
  vertex.label.pos = c("right", "right", "left", "left", "right"),
  # Less space for the legend
  legend.prop = 0.3
)

```

---

plot.mhmm

---

*Interactive Plotting for Mixed Hidden Markov Model (mhmm)*


---

### Description

Function `plot.mhmm` plots a directed graph of the parameters of each model with pie charts of emission probabilities as vertices/nodes.

### Usage

```

## S3 method for class 'mhmm'
plot(
  x,
  interactive = TRUE,
  ask = FALSE,
  which.plots = NULL,
  nrow = NA,
  ncol = NA,
  byrow = FALSE,
  row.prop = "auto",
  col.prop = "auto",
  layout = "horizontal",
  pie = TRUE,
  vertex.size = 40,
  vertex.label = "initial.probs",
  vertex.label.dist = "auto",
  vertex.label.pos = "bottom",
  vertex.label.family = "sans",
  loops = FALSE,
  edge.curved = TRUE,
  edge.label = "auto",
  edge.width = "auto",
  cex.edge.width = 1,
  edge.arrow.size = 1.5,
  edge.label.family = "sans",

```

```

label.signif = 2,
label.scientific = FALSE,
label.max.length = 6,
trim = 1e-15,
combine.slices = 0.05,
combined.slice.color = "white",
combined.slice.label = "others",
with.legend = "bottom",
ltext = NULL,
legend.prop = 0.5,
cex.legend = 1,
ncol.legend = "auto",
cpal = "auto",
main = "auto",
withlegend,
...
)

```

### Arguments

x	A hidden Markov model object of class mhmm created with <a href="#">build_mhmm</a> (or <a href="#">build_mmm</a> or <a href="#">build_lcm</a> ). Multichannel mhmm objects are automatically transformed into single-channel objects. See function <a href="#">mc_to_sc</a> for more information on the transformation.
interactive	Whether to plot each cluster in succession or in a grid. Defaults to TRUE, i.e. clusters are plotted one after another.
ask	If TRUE and which.plots is NULL, plot.mhmm operates in interactive mode, via <a href="#">menu</a> . Defaults to FALSE. Ignored if interactive = FALSE.
which.plots	The number(s) of the requested cluster(s) as an integer vector. The default NULL produces all plots.
nrow, ncol	Optional arguments to arrange plots in a grid. Ignored if interactive = TRUE.
byrow	Controls the order of plotting in a grid. Defaults to FALSE, i.e. plots are arranged column-wise. Ignored if interactive = TRUE.
row.prop	Sets the proportions of the row heights of the grid. The default value is "auto" for even row heights. Takes a vector of values from 0 to 1, with values summing to 1. Ignored if interactive = TRUE.
col.prop	Sets the proportion of the column heights of the grid. The default value is "auto" for even column widths. Takes a vector of values from 0 to 1, with values summing to 1. Ignored if interactive = TRUE.
layout	specifies the layout of vertices (nodes). Accepts a numerical matrix, a <a href="#">layout_</a> function (without quotation marks), or either of the predefined options "horizontal" (the default) and "vertical". Options "horizontal" and "vertical" position vertices at the same horizontal or vertical line. A two-column numerical matrix can be used to give x and y coordinates of the vertices. The <a href="#">layout_</a> functions available in the <a href="#">igraph</a> package offer other automatic layouts for graphs.



pie	Are vertices plotted as pie charts of emission probabilities? Defaults to TRUE.
vertex.size	Size of vertices, given as a scalar or numerical vector. The default value is 40.
vertex.label	Labels for vertices. Possible options include "initial.probs", "names", NA, and a character or numerical vector. The default "initial.probs" prints the initial probabilities of the model and "names" prints the names of the hidden states as labels. NA prints no labels.
vertex.label.dist	Distance of the label of the vertex from its center. The default value "auto" places the label outside the vertex.
vertex.label.pos	Positions of vertex labels, relative to the center of the vertex. A scalar or numerical vector giving position(s) as radians or one of "bottom" ( $\pi/2$ as radians), "top" ( $-\pi/2$ ), "left" ( $\pi$ ), or "right" ( $0$ ).
vertex.label.family, edge.label.family	Font family to be used for vertex/edge labels. See argument family in <a href="#">par</a> for more information.
loops	Defines whether transitions back to same states are plotted.
edge.curved	Defines whether to plot curved edges (arcs, arrows) between vertices. A logical or numerical vector or scalar. Numerical values specify curvatures of edges. The default value TRUE gives curvature of 0.5 to all edges. See <a href="#">igraph.plotting</a> for more information.
edge.label	Labels for edges. Possible options include "auto", NA, and a character or numerical vector. The default "auto" prints transition probabilities as edge labels. NA prints no labels.
edge.width	Width(s) for edges. The default "auto" determines widths according to transition probabilities between hidden states. Other possibilities are a scalar or a numerical vector of widths.
cex.edge.width	An expansion factor for edge widths. Defaults to 1.
edge.arrow.size	Size of the arrow in edges (constant). Defaults to 1.5.
label.signif	Rounds labels of model parameters to specified number of significant digits, 2 by default. Ignored for user-given labels.
label.scientific	Defines if scientific notation should be used to describe small numbers. Defaults to FALSE, e.g. 0.0001 instead of $1e-04$ . Ignored for user-given labels.
label.max.length	Maximum number of digits in labels of model parameters. Ignored for user-given labels.
trim	Scalar between 0 and 1 giving the highest probability of transitions that are plotted as edges, defaults to $1e-15$ .
combine.slices	Scalar between 0 and 1 giving the highest probability of emission probabilities that are combined into one state. The default value is 0.05.
combined.slice.color	Color of the combined slice that includes the smallest emission probabilities (only if argument "combine.slices" is greater than 0). The default color is white.

<code>combined.slice.label</code>	The label for combined states (when argument <code>"combine.slices"</code> is greater than 0) to appear in the legend.
<code>with.legend</code>	Defines if and where the legend of state colors is plotted. Possible values include <code>"bottom"</code> (the default), <code>"top"</code> , <code>"left"</code> , and <code>"right"</code> . <code>FALSE</code> omits the legend.
<code>ltext</code>	Optional description of (combined) observed states to appear in the legend. A vector of character strings. See <a href="#">seqplot</a> for more information.
<code>legend.prop</code>	Proportion used for plotting the legend. A scalar between 0 and 1, defaults to 0.5.
<code>cex.legend</code>	Expansion factor for setting the size of the font for labels in the legend. The default value is 1. Values lesser than 1 will reduce the size of the font, values greater than 1 will increase the size.
<code>ncol.legend</code>	The number of columns for the legend. The default value <code>"auto"</code> sets the number of columns automatically.
<code>cpal</code>	Optional color palette for (combinations of) observed states. The default value <code>"auto"</code> uses automatic color palette. Otherwise a vector of length <code>x\$n_symbols</code> is given, i.e. the argument requires a color specified for all (combinations of) observed states even if they are not plotted (if the probability is less than <code>combine.slices</code> ).
<code>main</code>	Optional main titles for plots. The default <code>"auto"</code> uses <code>cluster_names</code> as titles, <code>NULL</code> prints no titles.
<code>withlegend</code>	Deprecated. Use <code>with.legend</code> instead.
<code>...</code>	Other parameters passed on to <a href="#">plot.igraph</a> such as <code>vertex.color</code> , <code>vertex.label.cex</code> , or <code>edge.lty</code> .

## References

Helske S. and Helske J. (2019). Mixture Hidden Markov Models for Sequence Data: The seqHMM Package in R, *Journal of Statistical Software*, 88(3), 1-32. doi:10.18637/jss.v088.i03

## See Also

[build\\_mhmm](#) and [fit\\_model](#) for building and fitting mixture hidden Markov models; [plot.igraph](#) for plotting directed graphs; and [mhmm\\_biofam](#) and [mhmm\\_mvad](#) for the models used in examples.

## Examples

```
# Loading mixture hidden Markov model (mhmm object)
# of the biofam data
data("mhmm_biofam")

# Plotting only the first cluster
plot(mhmm_biofam, which.plots = 1)

if (interactive()) {
  # Plotting each cluster (change with Enter)
  plot(mhmm_biofam)

  # Choosing the cluster (one at a time)
```

```

plot(mhmm_biofam, ask = TRUE)

# Loading MHMM of the mvad data
data("mhmm_mvad")

# Plotting models in the same graph (in a grid)
# Note: the plotting window must be high enough!
set.seed(123)
plot(mhmm_mvad,
     interactive = FALSE,
     # automatic layout, legend on the right-hand side
     layout = layout_nicely, with.legend = "right",
     # Smaller and less curved edges
     edge.curved = 0.2, cex.edge.width = 0.5, edge.arrow.size = 0.7,
     vertex.label.pos = -4 * pi / 5, vertex.label.dist = 5
    )
}

```

plot.ssp

*Stack Multichannel Sequence Plots and/or Most Probable Paths Plots  
from Hidden Markov Models*

## Description

Function `plot.ssp` plots stacked sequence plots from `ssp` objects defined with [ssp](#).

## Usage

```
## S3 method for class 'ssp'
plot(x, ...)
```

## Arguments

<code>x</code>	An <code>ssp</code> object.
<code>...</code>	Ignored.

## References

Helske S. and Helske J. (2019). Mixture Hidden Markov Models for Sequence Data: The seqHMM Package in R, Journal of Statistical Software, 88(3), 1-32. doi:10.18637/jss.v088.i03

## See Also

[ssp](#) for more examples and information on defining the plot before using `plot.ssp`; [ssplot](#) for straight plotting of `ssp` objects; and [gridplot](#) for plotting multiple `ssp` objects.

**Examples**

```

data("biofam3c")

## Building sequence objects
child_seq <- seqdef(biofam3c$children, start = 15)
marr_seq <- seqdef(biofam3c$married, start = 15)
left_seq <- seqdef(biofam3c$left, start = 15)

## Choosing colors
attr(child_seq, "cpal") <- c("#66C2A5", "#FC8D62")
attr(marr_seq, "cpal") <- c("#AB82FF", "#E6AB02", "#E7298A")
attr(left_seq, "cpal") <- c("#A6CEE3", "#E31A1C")

# Plotting state distribution plots of observations
ssp1 <- ssp(list(child_seq, marr_seq, left_seq))
plot(ssp1)

```

---

plot\_colors

*Plot Colorpalettes*


---

**Description**

Function `plot_colors` plots colors and their labels for easy visualization of a colorpalette.

**Usage**

```
plot_colors(x, labels = NULL)
```

**Arguments**

`x` A vector of colors.  
`labels` A vector of labels for colors. If omitted, given color names are used.

**See Also**

See e.g. the [colorpalette](#) data and RColorBrewer package for ready-made color palettes.

**Examples**

```

plot_colors(colorpalette[[5]], labels = c("one", "two", "three", "four", "five"))

plot_colors(colorpalette[[10]])

plot_colors(1:7)

plot_colors(c("yellow", "orange", "red", "purple", "blue", "green"))

plot_colors(rainbow(15))

```

---

posterior_probs	<i>Posterior Probabilities for (Mixture) Hidden Markov Models</i>
-----------------	---

---

**Description**

Function `posterior_probs` computes the posterior probabilities of hidden states of a (mixture) hidden Markov model.

**Usage**

```
posterior_probs(model, log_space = FALSE)
```

**Arguments**

<code>model</code>	A (mixture) hidden Markov model of class <code>hmm</code> or <code>mhmm</code> .
<code>log_space</code>	Compute posterior probabilities in logarithmic scale. The default is <code>FALSE</code> .

**Value**

Posterior probabilities. In case of multiple observations, these are computed independently for each sequence.

**Examples**

```
# Load a pre-defined MHMM
data("mhmm_biofam")

# Compute posterior probabilities
pb <- posterior_probs(mhmm_biofam)

# Locally most probable states for the first subject:
pb[, , 1]
```

---

<code>print.hmm</code>	<i>Print Method for a Hidden Markov Model</i>
------------------------	---

---

**Description**

Prints the parameters of a (mixture) hidden Markov model.

**Usage**

```
## S3 method for class 'hmm'
print(x, digits = 3, ...)

## S3 method for class 'mhmm'
print(x, digits = 3, ...)

## S3 method for class 'summary.mhmm'
print(x, digits = 3, ...)
```

**Arguments**

`x` Hidden Markov model of class `hmm` or `mhmm`.  
`digits` Minimum number of significant digits to print.  
`...` Further arguments to `print.default`.

**See Also**

[build\\_hmm](#) and [fit\\_model](#) for building and fitting hidden Markov models.

---

<code>separate_mhmm</code>	<i>Reorganize a mixture hidden Markov model to a list of separate hidden Markov models (covariates ignored)</i>
----------------------------	---

---

**Description**

The `separate_mhmm` function reorganizes the parameters of a `mhmm` object into a list where each list component is an object of class `hmm` consisting of the parameters of the corresponding cluster.

**Usage**

```
separate_mhmm(model)
```

**Arguments**

`model` Mixture hidden Markov model of class `mhmm`.

**Value**

List with components of class `hmm`.

**See Also**

[build\\_mhmm](#) and [fit\\_model](#) for building and fitting MHMMs; and [mhmm\\_biofam](#) for more information on the model used in examples.

**Examples**

```
# Loading mixture hidden Markov model (mhmm object)
# of the biofam data
data("mhmm_biofam")

# Separate models for clusters
sep_hmm <- separate_mhmm(mhmm_biofam)

# Plotting the model for the first cluster
plot(sep_hmm[[1]])
```

seqdef

*Imported Functions from TraMineR***Description**

Imported functions for convenience. For details, see the corresponding help pages of [seqstatf](#), [alphabet](#) and [seqdef](#).

seqHMM

*The seqHMM package***Description**

The seqHMM package is designed for fitting hidden (or latent) Markov models (HMMs) and mixture hidden Markov models (MHMMs) for social sequence data and other categorical time series. The package supports models for one or multiple subjects with one or multiple interdependent sequences (channels). External covariates can be added to explain cluster membership in mixture models. The package provides functions for evaluating and comparing models, as well as functions for easy plotting of multichannel sequences and hidden Markov models. Common restricted versions of (M)HMMs are also supported, namely Markov models, mixture Markov models, and latent class models.

**Details**

Maximum likelihood estimation via the EM algorithm and direct numerical maximization with analytical gradients is supported. All main algorithms are written in C++. Parallel computation is implemented via OpenMP.

**References**

Helske S. and Helske J. (2019). Mixture Hidden Markov Models for Sequence Data: The seqHMM Package in R, *Journal of Statistical Software*, 88(3), 1-32. doi:10.18637/jss.v088.i03

---

seqHMM-deprecated      *Deprecated function(s) in the seqHMM package*

---

### Description

These functions are provided for compatibility with older version of the seqHMM package. They will be eventually completely removed.

### Usage

```
fit_hmm(  
  model,  
  em_step = TRUE,  
  global_step = FALSE,  
  local_step = FALSE,  
  control_em = list(),  
  control_global = list(),  
  control_local = list(),  
  lb,  
  ub,  
  threads = 1,  
  log_space = FALSE,  
  ...  
)
```

```
fit_mhmm(  
  model,  
  em_step = TRUE,  
  global_step = FALSE,  
  local_step = FALSE,  
  control_em = list(),  
  control_global = list(),  
  control_local = list(),  
  lb,  
  ub,  
  threads = 1,  
  log_space = FALSE,  
  ...  
)
```

```
trim_hmm(  
  model,  
  maxit = 0,  
  return_loglik = FALSE,  
  zerotol = 1e-08,  
  verbose = TRUE,  
  ...  
)
```



)

**Arguments**

- model** An object of class `hmm` or `mhmm`.
- em\_step** Logical. Whether or not to use the EM algorithm at the start of the parameter estimation. The default is `TRUE`.
- global\_step** Logical. Whether or not to use global optimization via `nloptr` (possibly after the EM step). The default is `FALSE`.
- local\_step** Logical. Whether or not to use local optimization via `nloptr` (possibly after the EM and/or global steps). The default is `FALSE`.
- control\_em** Optional list of control parameters for the EM algorithm. Possible arguments are
- maxeval** The maximum number of iterations, the default is 1000. Note that iteration counter starts with -1 so with `maxeval=1` you get already two iterations. This is for backward compatibility reasons.
  - print\_level** The level of printing. Possible values are 0 (prints nothing), 1 (prints information at the start and the end of the algorithm), 2 (prints at every iteration), and for mixture models 3 (print also during optimization of coefficients).
  - reltol** Relative tolerance for convergence defined as  $(\log Lik_{new} - \log Lik_{old}) / (abs(\log Lik_{old}) + 0.1)$ . The default is `1e-10`.
  - restart** A list containing options for possible EM restarts with the following components:
    - times** Number of restarts of the EM algorithm using random initial values. The default is 0, i.e. no restarts.
    - transition** Logical. Should the original transition probabilities be varied? The default is `TRUE`.
    - emission** Logical. Should the original emission probabilities be varied? The default is `TRUE`.
    - sd** Standard deviation for `rnorm` used in randomization. The default is `0.25`.
    - maxeval** Maximum number of iterations, the default is `control_em$maxeval`
    - print\_level** Level of printing in restarted EM steps. The default is `control_em$print_level`.
    - reltol** Relative tolerance for convergence at restarted EM steps. The default is `control_em$reltol`. If the relative change of the final model of the restart phase is larger than the tolerance for the original EM phase, the final model is re-estimated with the original `reltol` and `maxeval` at the end of the EM step.
    - n\_optimum** Save the log-likelihood values of the `n_optimum` best models (from all estimated models including the the first EM run.). The default is `min(times + 1, 25)`.
    - use\_original** If `TRUE`. Use the initial values of the input model as starting points for the permutations. Otherwise permute the results of the first EM run.

control_global	Optional list of additional arguments for <code>nloptr</code> argument opts. The default values are <b>algorithm</b> "NLOPT_GD_MLSL_LDS" <b>local_opts</b> list(algorithm = "NLOPT_LD_LBFGS", ftol_rel = 1e-6, xtol_rel = 1e-4) <b>maxeval</b> 10000 (maximum number of iterations in global optimization algorithm.) <b>maxtime</b> 60 (maximum time for global optimization. Set to 0 for unlimited time.)
control_local	Optional list of additional arguments for <code>nloptr</code> argument opts. The default values are <b>algorithm</b> "NLOPT_LD_LBFGS" <b>ftol_rel</b> 1e-10 <b>xtol_rel</b> 1e-8 <b>maxeval</b> 10000 (maximum number of iterations)
lb, ub	Lower and upper bounds for parameters in Softmax parameterization. The default interval is $[pmin(-25, 2 * initialvalues), pmax(25, 2 * initialvalues)]$ , except for gamma coefficients, where the scale of covariates is taken into account. Note that it might still be a good idea to scale covariates around unit scale. Bounds are used only in the global optimization step.
threads	Number of threads to use in parallel computing. The default is 1.
log_space	Make computations using log-space instead of scaling for greater numerical stability at a cost of decreased computational performance. The default is FALSE.
...	Additional arguments to <code>nloptr</code> .
maxit	Number of iterations. After zeroing small values, the model is refitted, and this is repeated until there is nothing to trim or <code>maxit</code> iterations are done.
return_loglik	Return the log-likelihood of the trimmed model together with the model object. The default is FALSE.
zerotol	Values smaller than this are trimmed to zero.
verbose	Print results of trimming. The default is TRUE.

---

simulate\_hmm

*Simulate hidden Markov models*


---

### Description

Simulate sequences of observed and hidden states given parameters of a hidden Markov model.

**Usage**

```
simulate_hmm(  
  n_sequences,  
  initial_probs,  
  transition_probs,  
  emission_probs,  
  sequence_length  
)
```

**Arguments**

`n_sequences`      Number of simulations.

`initial_probs`    A vector of initial state probabilities.

`transition_probs`  
                    A matrix of transition probabilities.

`emission_probs`   A matrix of emission probabilities or a list of such objects (one for each channel).

`sequence_length`   Length for simulated sequences.

**Value**

A list of state sequence objects of class `stslst`.

**See Also**

[build\\_hmm](#) and [fit\\_model](#) for building and fitting hidden Markov models; [ssplot](#) for plotting multiple sequence data sets; [seqdef](#) for more information on state sequence objects; and [simulate\\_mhmm](#) for simulating mixture hidden Markov models.

**Examples**

```
# Parameters for the HMM  
emission_probs <- matrix(c(0.5, 0.2, 0.5, 0.8), 2, 2)  
transition_probs <- matrix(c(5 / 6, 1 / 6, 1 / 6, 5 / 6), 2, 2)  
initial_probs <- c(1, 0)  
  
# Setting the seed for simulation  
set.seed(1)  
  
# Simulating sequences  
sim <- simulate_hmm(  
  n_sequences = 10, initial_probs = initial_probs,  
  transition_probs = transition_probs,  
  emission_probs = emission_probs,  
  sequence_length = 20  
)  
  
ssplot(sim, sortv = "mds.obs", type = "I")
```

---

 simulate\_initial\_probs

*Simulate Parameters of Hidden Markov Models*


---

### Description

These are helper functions for quick construction of initial values for various model building functions. Mostly useful for global optimization algorithms which do not depend on initial values.

### Usage

```
simulate_initial_probs(n_states, n_clusters = 1)

simulate_transition_probs(
  n_states,
  n_clusters = 1,
  left_right = FALSE,
  diag_c = 0
)

simulate_emission_probs(n_states, n_symbols, n_clusters = 1)
```

### Arguments

n_states	Number of states in each cluster.
n_clusters	Number of clusters.
left_right	Constrain the transition probabilities to upper triangular. Default is FALSE.
diag_c	A constant value to be added to diagonal of transition matrices before scaling.
n_symbols	Number of distinct symbols in each channel.

### See Also

[build\\_hmm](#), [build\\_mhmm](#), [build\\_mm](#), [build\\_mmm](#), and [build\\_lcm](#) for constructing different types of models.

---

 simulate\_mhmm

*Simulate Mixture Hidden Markov Models*


---

### Description

Simulate sequences of observed and hidden states given the parameters of a mixture hidden Markov model.

**Usage**

```
simulate_mhmm(
  n_sequences,
  initial_probs,
  transition_probs,
  emission_probs,
  sequence_length,
  formula,
  data,
  coefficients
)
```

**Arguments**

<code>n_sequences</code>	The number of simulations.
<code>initial_probs</code>	A list containing vectors of initial state probabilities for the submodel of each cluster.
<code>transition_probs</code>	A list of matrices of transition probabilities for the submodel of each cluster.
<code>emission_probs</code>	A list which contains matrices of emission probabilities or a list of such objects (one for each channel) for the submodel of each cluster. Note that the matrices must have dimensions $s \times m$ where $s$ is the number of hidden states and $m$ is the number of unique symbols (observed states) in the data.
<code>sequence_length</code>	The length of the simulated sequences.
<code>formula</code>	Covariates as an object of class <code>formula</code> , left side omitted.
<code>data</code>	An optional data frame, a list or an environment containing the variables in the model. If not found in data, the variables are taken from <code>environment(formula)</code> .
<code>coefficients</code>	An optional $k \times l$ matrix of regression coefficients for time-constant covariates for mixture probabilities, where $l$ is the number of clusters and $k$ is the number of covariates. A logit-link is used for mixture probabilities. The first column is set to zero.

**Value**

A list of state sequence objects of class `stslst`.

**See Also**

[build\\_mhmm](#) and [fit\\_model](#) for building and fitting mixture hidden Markov models; [ssplot](#) for plotting multiple sequence data sets; [seqdef](#) for more information on state sequence objects; and [simulate\\_hmm](#) for simulating hidden Markov models.

**Examples**

```
emission_probs_1 <- matrix(c(0.75, 0.05, 0.25, 0.95), 2, 2)
emission_probs_2 <- matrix(c(0.1, 0.8, 0.9, 0.2), 2, 2)
```

```

colnames(emission_probs_1) <- colnames(emission_probs_2) <-
  c("heads", "tails")

transition_probs_1 <- matrix(c(9, 0.1, 1, 9.9) / 10, 2, 2)
transition_probs_2 <- matrix(c(35, 1, 1, 35) / 36, 2, 2)
rownames(emission_probs_1) <- rownames(transition_probs_1) <-
  colnames(transition_probs_1) <- c("coin 1", "coin 2")
rownames(emission_probs_2) <- rownames(transition_probs_2) <-
  colnames(transition_probs_2) <- c("coin 3", "coin 4")

initial_probs_1 <- c(1, 0)
initial_probs_2 <- c(1, 0)

n <- 30
set.seed(123)
covariate_1 <- runif(n)
covariate_2 <- sample(c("A", "B"),
  size = n, replace = TRUE,
  prob = c(0.3, 0.7)
)
dataf <- data.frame(covariate_1, covariate_2)

coefs <- cbind(cluster_1 = c(0, 0, 0), cluster_2 = c(-1.5, 3, -0.7))
rownames(coefs) <- c("(Intercept)", "covariate_1", "covariate_2B")

sim <- simulate_mhmm(
  n = n, initial_probs = list(initial_probs_1, initial_probs_2),
  transition_probs = list(transition_probs_1, transition_probs_2),
  emission_probs = list(emission_probs_1, emission_probs_2),
  sequence_length = 20, formula = ~ covariate_1 + covariate_2,
  data = dataf, coefficients = coefs
)

ssplot(sim$observations,
  hidden.paths = sim$states, plots = "both",
  sortv = "from.start", sort.channel = 0, type = "I"
)

hmm <- build_mhmm(sim$observations,
  initial_probs = list(initial_probs_1, initial_probs_2),
  transition_probs = list(transition_probs_1, transition_probs_2),
  emission_probs = list(emission_probs_1, emission_probs_2),
  formula = ~ covariate_1 + covariate_2,
  data = dataf
)

fit <- fit_model(hmm)
fit$model

paths <- hidden_paths(fit$model)

ssplot(list(estimated = paths, true = sim$states),
  sortv = "from.start",

```

```

    sort.channel = 2, ylab = c("estimated paths", "true (simulated)"),
    type = "I"
)

```

---

 ssp

*Define Arguments for Plotting Multichannel Sequences and/or Most Probable Paths from Hidden Markov Models*

---

## Description

Function `ssp` defines the arguments for plotting with `plot.ssp` or `gridplot`.

## Usage

```

ssp(
  x,
  hidden.paths = NULL,
  plots = "obs",
  type = "d",
  tlim = 0,
  sortv = NULL,
  sort.channel = 1,
  dist.method = "OM",
  with.missing = FALSE,
  missing.color = NULL,
  title = NA,
  title.n = TRUE,
  cex.title = 1,
  title.pos = 1,
  with.legend = "auto",
  ncol.legend = "auto",
  with.missing.legend = "auto",
  legend.prop = 0.3,
  cex.legend = 1,
  hidden.states.colors = "auto",
  hidden.states.labels = "auto",
  xaxis = TRUE,
  xlab = NA,
  xtlab = NULL,
  xlab.pos = 1,
  ylab = "auto",
  hidden.states.title = "Hidden states",
  yaxis = FALSE,
  ylab.pos = "auto",
  cex.lab = 1,
  cex.axis = 1,

```

```

    withlegend,
    respect_void = TRUE,
    ...
)

```

## Arguments

<code>x</code>	Either a hidden Markov model object of class <code>hmm</code> or a state sequence object of class <code>stslist</code> (created with the <code>seqdef</code> function) or a list of state sequence objects.
<code>hidden.paths</code>	Output from <code>hidden_paths</code> function. Optional, if <code>x</code> is a <code>hmm</code> object or if <code>type = "obs"</code> .
<code>plots</code>	What to plot. One of <code>"obs"</code> for observations (the default), <code>"hidden.paths"</code> for most probable paths of hidden states, or <code>"both"</code> for observations and hidden paths together.
<code>type</code>	The type of the plot. Available types are <code>"I"</code> for sequence index plots and <code>"d"</code> for state distribution plots (the default). See <code>seqplot</code> for details.
<code>tlim</code>	Indexes of the subjects to be plotted (the default is 0, i.e. all subjects are plotted). For example, <code>tlim = 1:10</code> plots the first ten subjects in data.
<code>sortv</code>	A sorting variable or a sort method (one of <code>"from.start"</code> , <code>"from.end"</code> , <code>"mds.obs"</code> , or <code>"mds.hidden"</code> ) for <code>type = "I"</code> . The value <code>"mds.hidden"</code> is only available when hidden paths are available. Options <code>"mds.obs"</code> and <code>"mds.hidden"</code> automatically arrange the sequences according to the scores of multidimensional scaling (using <code>cmdscale</code> ) for the observed data or hidden states paths. MDS scores are computed from distances/dissimilarities using a metric defined in argument <code>dist.method</code> . See <code>plot.stslist</code> for more details on <code>"from.start"</code> and <code>"from.end"</code> .
<code>sort.channel</code>	The number of the channel according to which the <code>"from.start"</code> or <code>"from.end"</code> sorting is done. Sorting according to hidden states is called with value 0. The default value is 1 (the first channel).
<code>dist.method</code>	The metric to be used for computing the distances of the sequences if multi-dimensional scaling is used for sorting. One of <code>"OM"</code> (optimal matching, the default), <code>"LCP"</code> (longest common prefix), <code>"RLCP"</code> (reversed LCP, i.e. longest common suffix), <code>"LCS"</code> (longest common subsequence), <code>"HAM"</code> (Hamming distance), and <code>"DHD"</code> (dynamic Hamming distance). Transition rates are used for defining substitution costs if needed. See <code>seqdef</code> for more information on the metrics.
<code>with.missing</code>	Controls whether missing states are included in state distribution plots ( <code>type = "d"</code> ). The default is <code>FALSE</code> .
<code>missing.color</code>	Alternative color for representing missing values in the sequences. By default, this color is taken from the <code>missing.color</code> attribute of the sequence object.
<code>title</code>	Main title for the graphic. The default is <code>NA</code> : if <code>title.n = TRUE</code> , only the number of subjects is plotted. <code>FALSE</code> prints no title, even when <code>title.n = TRUE</code> .
<code>title.n</code>	Controls whether the number of subjects (in the first channel) is printed in the title of the plot. The default is <code>TRUE</code> : <code>n</code> is plotted if <code>title</code> is anything but <code>FALSE</code> .



<code>cex.title</code>	Expansion factor for setting the size of the font for the title. The default value is 1. Values lesser than 1 will reduce the size of the font, values greater than 1 will increase the size.
<code>title.pos</code>	Controls the position of the main title of the plot. The default value is 1. Values greater than 1 will place the title higher.
<code>with.legend</code>	Defines if and where the legend for the states is plotted. The default value "auto" (equivalent to TRUE and "right") creates separate legends for each requested plot and positions them on the right-hand side of the plot. Other possible values are "bottom", "right.combined", and "bottom.combined", of which the last two create a combined legend in the selected position. FALSE prints no legend.
<code>ncol.legend</code>	(A vector of) the number of columns for the legend(s). The default "auto" determines number of columns depending on the position of the legend.
<code>with.missing.legend</code>	If set to "auto" (the default), a legend for the missing state is added automatically if one or more of the sequences in the data/channel contains missing states and <code>type = "I"</code> . If <code>type = "d"</code> missing states are omitted from the legends unless <code>with.missing = TRUE</code> . With the value TRUE a legend for the missing state is added in any case; equivalently FALSE omits the legend for the missing state.
<code>legend.prop</code>	Sets the proportion of the graphic area used for plotting the legend when <code>with.legend</code> is not FALSE. The default value is 0.3. Takes values from 0 to 1.
<code>cex.legend</code>	Expansion factor for setting the size of the font for the labels in the legend. The default value is 1. Values lesser than 1 will reduce the size of the font, values greater than 1 will increase the size.
<code>hidden.states.colors</code>	A vector of colors assigned to hidden states. The default value "auto" uses the colors assigned to the <code>stslst</code> object (created with <code>seqdef</code> ) if <code>hidden.paths</code> is given; otherwise colors from <code>colorpalette</code> are automatically used.
<code>hidden.states.labels</code>	Labels for the hidden states. The default value "auto" uses the names provided in <code>x\$state_names</code> if <code>x</code> is an <code>hmm</code> object; otherwise the number of the hidden state.
<code>xaxis</code>	Controls whether an x-axis is plotted below the plot at the bottom. The default value is TRUE.
<code>xlab</code>	An optional label for the x-axis. If set to NA, no label is drawn.
<code>xtlab</code>	Optional labels for the x-axis tick labels. If unspecified, the column names of the <code>seqdata</code> sequence object are used (see <code>seqdef</code> ).
<code>xlab.pos</code>	Controls the position of the x-axis label. The default value is 1. Values greater than 1 will place the label further away from the plot.
<code>ylab</code>	Labels for the channels shown as labels for y-axes. A vector of names for each channel (observations). The default value "auto" uses the names provided in <code>x\$channel_names</code> if <code>x</code> is an <code>hmm</code> object; otherwise the names of the list in <code>x</code> if given, or the number of the channel if names are not given. FALSE prints no labels.

<code>hidden.states.title</code>	Optional label for the hidden state plot (in the y-axis). The default is "Hidden states".
<code>yaxis</code>	Controls whether or not to plot the y-axis. The default is FALSE.
<code>ylab.pos</code>	Controls the position of the y axis labels (labels for channels and/or hidden states). Either "auto" or a numerical vector indicating how far away from the plots the titles are positioned. The default value "auto" positions all titles on line 1. Shorter vectors are recycled.
<code>cex.lab</code>	Expansion factor for setting the size of the font for the axis labels. The default value is 1. Values lesser than 1 will reduce the size of the font, values greater than 1 will increase the size.
<code>cex.axis</code>	Expansion factor for setting the size of the font for the x-axis tick labels. The default value is 1. Values lesser than 1 will reduce the size of the font, values greater than 1 will increase the size.
<code>withlegend</code>	Deprecated. Use <code>with.legend</code> instead.
<code>respect_void</code>	If TRUE (default), states at the time points corresponding to TraMineR's void in the observed sequences are set to void in the hidden state sequences as well.
<code>...</code>	Other arguments to be passed on to <a href="#">seqplot</a> .

### Value

Object of class `ssp`.

### See Also

[plot.ssp](#) for plotting objects created with the `ssp` function; [gridplot](#) for plotting multiple `ssp` objects; [build\\_hmm](#) and [fit\\_model](#) for building and fitting hidden Markov models; [hidden\\_paths](#) for computing the most probable paths of hidden states; and [biofam3c](#) and [hmm\\_biofam](#) for information on the data and model used in the example.

### Examples

```
data("biofam3c")

## Building sequence objects
child_seq <- seqdef(biofam3c$children, start = 15)
marr_seq <- seqdef(biofam3c$married, start = 15)
left_seq <- seqdef(biofam3c$left, start = 15)

## Choosing colors
attr(child_seq, "cpal") <- c("#66C2A5", "#FC8D62")
attr(marr_seq, "cpal") <- c("#AB82FF", "#E6AB02", "#E7298A")
attr(left_seq, "cpal") <- c("#A6CEE3", "#E31A1C")

# Defining the plot for state distribution plots of observations
ssp1 <- ssp(list(
  "Parenthood" = child_seq, "Marriage" = marr_seq,
  "Residence" = left_seq
```

```
))
# Plotting ssp1
plot(ssp1)

## Not run:
# Defining the plot for sequence index plots of observations
ssp2 <- ssp(
  list(child_seq, marr_seq, left_seq),
  type = "I", plots = "obs",
  # Sorting subjects according to the beginning of the 2nd channel (marr_seq)
  sortv = "from.start", sort.channel = 2,
  # Controlling the size, positions, and names for channel labels
  ylab.pos = c(1, 2, 1), cex.lab = 1, ylab = c("Children", "Married", "Residence"),
  # Plotting without legend
  with.legend = FALSE
)
plot(ssp2)

# Plotting hidden Markov models

# Loading data
data("hmm_biofam")

# Plotting observations and most probable hidden states paths
ssp3 <- ssp(
  hmm_biofam,
  type = "I", plots = "both",
  # Sorting according to multidimensional scaling of hidden states paths
  sortv = "mds.hidden",
  # Controlling title
  title = "Biofam", cex.title = 1.5,
  # Labels for x axis and tick marks
  xtlab = 15:30, xlab = "Age"
)
plot(ssp3)

# Computing the most probable paths of hidden states
hid <- hidden_paths(hmm_biofam)
# Giving names for hidden states
library(TraMineR)
alphabet(hid) <- paste("Hidden state", 1:5)

# Plotting observations and hidden state paths
ssp4 <- ssp(
  hmm_biofam,
  type = "I", plots = "hidden.paths",
  # Sequence object of most probable paths
  hidden.paths = hid,
  # Sorting according to the end of hidden state paths
  sortv = "from.end", sort.channel = 0,
  # Controlling legend position, type, and proportion
  with.legend = "bottom.combined", legend.prop = 0.15,
  # Plotting without title and y label
```

```
    title = FALSE, ylab = FALSE
  )
plot(ssp4)

## End(Not run)
```

---

ssplot	<i>Stacked Plots of Multichannel Sequences and/or Most Probable Paths from Hidden Markov Models</i>
--------	---

---

## Description

Function `ssplot` plots stacked sequence plots of sequence object created with the `seqdef` function or observations and/or most probable paths of `hmm` objects.

## Usage

```
ssplot(
  x,
  hidden.paths = NULL,
  plots = "obs",
  type = "d",
  tlim = 0,
  sortv = NULL,
  sort.channel = 1,
  dist.method = "OM",
  with.missing = FALSE,
  missing.color = NULL,
  title = NA,
  title.n = TRUE,
  cex.title = 1,
  title.pos = 1,
  with.legend = "auto",
  ncol.legend = "auto",
  with.missing.legend = "auto",
  legend.prop = 0.3,
  cex.legend = 1,
  hidden.states.colors = "auto",
  hidden.states.labels = "auto",
  xaxis = TRUE,
  xlab = NA,
  xtlab = NULL,
  xlab.pos = 1,
  ylab = "auto",
  hidden.states.title = "Hidden states",
  yaxis = FALSE,
  ylab.pos = "auto",
```

```

    cex.lab = 1,
    cex.axis = 1,
    respect_void = TRUE,
    ...
)

```

### Arguments

<code>x</code>	Either a hidden Markov model object of class <code>hmm</code> or a state sequence object of class <code>stslist</code> (created with the <a href="#">seqdef</a> ) function) or a list of state sequence objects.
<code>hidden.paths</code>	Output from <a href="#">hidden_paths</a> function. Optional, if <code>x</code> is a <code>hmm</code> object or if <code>type = "obs"</code> .
<code>plots</code>	What to plot. One of <code>"obs"</code> for observations (the default), <code>"hidden.paths"</code> for most probable paths of hidden states, or <code>"both"</code> for observations and hidden paths together.
<code>type</code>	The type of the plot. Available types are <code>"I"</code> for sequence index plots and <code>"d"</code> for state distribution plots (the default). See <a href="#">seqplot</a> for details.
<code>tlim</code>	Indexes of the subjects to be plotted (the default is 0, i.e. all subjects are plotted). For example, <code>tlim = 1:10</code> plots the first ten subjects in data.
<code>sortv</code>	A sorting variable or a sort method (one of <code>"from.start"</code> , <code>"from.end"</code> , <code>"mds.obs"</code> , or <code>"mds.hidden"</code> ) for <code>type = "I"</code> . The value <code>"mds.hidden"</code> is only available when hidden paths are available. Options <code>"mds.obs"</code> and <code>"mds.hidden"</code> automatically arrange the sequences according to the scores of multidimensional scaling (using <a href="#">cmdscale</a> ) for the observed data or hidden states paths. MDS scores are computed from distances/dissimilarities using a metric defined in argument <code>dist.method</code> . See <a href="#">plot.stslist</a> for more details on <code>"from.start"</code> and <code>"from.end"</code> .
<code>sort.channel</code>	The number of the channel according to which the <code>"from.start"</code> or <code>"from.end"</code> sorting is done. Sorting according to hidden states is called with value 0. The default value is 1 (the first channel).
<code>dist.method</code>	The metric to be used for computing the distances of the sequences if multi-dimensional scaling is used for sorting. One of <code>"OM"</code> (optimal matching, the default), <code>"LCP"</code> (longest common prefix), <code>"RLCP"</code> (reversed LCP, i.e. longest common suffix), <code>"LCS"</code> (longest common subsequence), <code>"HAM"</code> (Hamming distance), and <code>"DHD"</code> (dynamic Hamming distance). Transition rates are used for defining substitution costs if needed. See <a href="#">seqdef</a> for more information on the metrics.
<code>with.missing</code>	Controls whether missing states are included in state distribution plots ( <code>type = "d"</code> ). The default is <code>FALSE</code> .
<code>missing.color</code>	Alternative color for representing missing values in the sequences. By default, this color is taken from the <code>missing.color</code> attribute of the sequence object.
<code>title</code>	Main title for the graphic. The default is <code>NA</code> : if <code>title.n = TRUE</code> , only the number of subjects is plotted. <code>FALSE</code> prints no title, even when <code>title.n = TRUE</code> .
<code>title.n</code>	Controls whether the number of subjects (in the first channel) is printed in the title of the plot. The default is <code>TRUE</code> : <code>n</code> is plotted if <code>title</code> is anything but <code>FALSE</code> .

<code>cex.title</code>	Expansion factor for setting the size of the font for the title. The default value is 1. Values lesser than 1 will reduce the size of the font, values greater than 1 will increase the size.
<code>title.pos</code>	Controls the position of the main title of the plot. The default value is 1. Values greater than 1 will place the title higher.
<code>with.legend</code>	Defines if and where the legend for the states is plotted. The default value "auto" (equivalent to TRUE and "right") creates separate legends for each requested plot and positiones them on the right-hand side of the plot. Other possible values are "bottom", "right.combined", and "bottom.combined", of which the last two create a combined legend in the selected position. FALSE prints no legend.
<code>ncol.legend</code>	(A vector of) the number of columns for the legend(s). The default "auto" determines number of columns depending on the position of the legend.
<code>with.missing.legend</code>	If set to "auto" (the default), a legend for the missing state is added automatically if one or more of the sequences in the data/channel contains missing states and <code>type = "I"</code> . If <code>type = "d"</code> missing states are omitted from the legends unless <code>with.missing = TRUE</code> . With the value TRUE a legend for the missing state is added in any case; equivalently FALSE omits the legend for the missing state.
<code>legend.prop</code>	Sets the proportion of the graphic area used for plotting the legend when <code>with.legend</code> is not FALSE. The default value is 0.3. Takes values from 0 to 1.
<code>cex.legend</code>	Expansion factor for setting the size of the font for the labels in the legend. The default value is 1. Values lesser than 1 will reduce the size of the font, values greater than 1 will increase the size.
<code>hidden.states.colors</code>	A vector of colors assigned to hidden states. The default value "auto" uses the colors assigned to the <code>stslst</code> object (created with <code>seqdef</code> ) if <code>hidden.paths</code> is given; otherwise colors from <code>colorpalette</code> are automatically used.
<code>hidden.states.labels</code>	Labels for the hidden states. The default value "auto" uses the names provided in <code>x\$state_names</code> if <code>x</code> is an <code>hmm</code> object; otherwise the number of the hidden state.
<code>xaxis</code>	Controls whether an x-axis is plotted below the plot at the bottom. The default value is TRUE.
<code>xlab</code>	An optional label for the x-axis. If set to NA, no label is drawn.
<code>xtlab</code>	Optional labels for the x-axis tick labels. If unspecified, the column names of the <code>seqdata</code> sequence object are used (see <code>seqdef</code> ).
<code>xlab.pos</code>	Controls the position of the x-axis label. The default value is 1. Values greater than 1 will place the label further away from the plot.
<code>ylab</code>	Labels for the channels shown as labels for y-axes. A vector of names for each channel (observations). The default value "auto" uses the names provided in <code>x\$channel_names</code> if <code>x</code> is an <code>hmm</code> object; otherwise the names of the list in <code>x</code> if given, or the number of the channel if names are not given. FALSE prints no labels.

<code>hidden.states.title</code>	Optional label for the hidden state plot (in the y-axis). The default is "Hidden states".
<code>yaxis</code>	Controls whether or not to plot the y-axis. The default is FALSE.
<code>ylab.pos</code>	Controls the position of the y axis labels (labels for channels and/or hidden states). Either "auto" or a numerical vector indicating how far away from the plots the titles are positioned. The default value "auto" positions all titles on line 1. Shorter vectors are recycled.
<code>cex.lab</code>	Expansion factor for setting the size of the font for the axis labels. The default value is 1. Values lesser than 1 will reduce the size of the font, values greater than 1 will increase the size.
<code>cex.axis</code>	Expansion factor for setting the size of the font for the x-axis tick labels. The default value is 1. Values lesser than 1 will reduce the size of the font, values greater than 1 will increase the size.
<code>respect_void</code>	If TRUE (default), states at the time points corresponding to TraMineR's void in the observed sequences are set to void in the hidden state sequences as well.
<code>...</code>	Other arguments to be passed on to <a href="#">seqplot</a> .

### See Also

[ssp](#) for creating ssp objects and [plot.ssp](#) and [gridplot](#) for plotting these; [build\\_hmm](#) and [fit\\_model](#) for building and fitting hidden Markov models; [hidden\\_paths](#) for computing the most probable paths of hidden states; and [biofam3c\\_hmm\\_biofam](#) for information on the data and model used in the example.

### Examples

```
data("biofam3c")

# Creating sequence objects
child_seq <- seqdef(biofam3c$children, start = 15)
marr_seq <- seqdef(biofam3c$married, start = 15)
left_seq <- seqdef(biofam3c$left, start = 15)

## Choosing colors
attr(child_seq, "cpal") <- c("#66C2A5", "#FC8D62")
attr(marr_seq, "cpal") <- c("#AB82FF", "#E6AB02", "#E7298A")
attr(left_seq, "cpal") <- c("#A6CEE3", "#E31A1C")

# Plotting state distribution plots of observations
ssplot(list(
  "Children" = child_seq, "Marriage" = marr_seq,
  "Residence" = left_seq
))

## Not run:
# Plotting sequence index plots of observations
ssplot(
```

```

list(child_seq, marr_seq, left_seq),
type = "I",
# Sorting subjects according to the beginning of the 2nd channel (marr_seq)
sortv = "from.start", sort.channel = 2,
# Controlling the size, positions, and names for channel labels
ylab.pos = c(1, 2, 1), cex.lab = 1, ylab = c("Children", "Married", "Residence"),
# Plotting without legend
with.legend = FALSE
)

# Plotting hidden Markov models

# Loading a ready-made HMM for the biofam data
data("hmm_biofam")

# Plotting observations and hidden states paths
ssplot(
  hmm_biofam,
  type = "I", plots = "both",
  # Sorting according to multidimensional scaling of hidden states paths
  sortv = "mds.hidden",
  ylab = c("Children", "Married", "Left home"),
  # Controlling title
  title = "Biofam", cex.title = 1.5,
  # Labels for x axis and tick marks
  xtlab = 15:30, xlab = "Age"
)

# Computing the most probable paths of hidden states
hidden.paths <- hidden_paths(hmm_biofam)
hidden.paths_seq <- seqdef(hidden.paths, labels = paste("Hidden state", 1:5))

# Plotting observations and hidden state paths
ssplot(
  hmm_biofam,
  type = "I", plots = "hidden.paths",
  # Sequence object of most probable paths
  hidden.paths = hidden.paths_seq,
  # Sorting according to the end of hidden state paths
  sortv = "from.end", sort.channel = 0,
  # Controlling legend position, type, and proportion
  with.legend = "bottom", legend.prop = 0.15,
  # Plotting without title and y label
  title = FALSE, ylab = FALSE
)

## End(Not run)

```



**Description**

Get state names from hmm or mhmm object

**Usage**

```
state_names(object)
```

**Arguments**

object            An object of class 'hmm' or 'mhmm'.

**Value**

A character vector containing the state names, or a list of such vectors in 'mhmm' case.

---

state\_names<-            *Set state names for hmm or mhmm object*

---

**Description**

Set state names for hmm or mhmm object

**Usage**

```
state_names(object) <- value
```

**Arguments**

object            An object of class 'hmm' or 'mhmm'.

value            A character vector containing the new state names, or a list of such vectors in 'mhmm' case.

**Value**

The modified object with updated state names.

summary.mhmm

*Summary method for mixture hidden Markov models***Description**

Function `summary.mhmm` gives a summary of a mixture hidden Markov model.

**Usage**

```
## S3 method for class 'mhmm'
summary(
  object,
  parameters = FALSE,
  conditional_se = TRUE,
  log_space = FALSE,
  ...
)
```

**Arguments**

<code>object</code>	Mixture hidden Markov model of class <code>mhmm</code> .
<code>parameters</code>	Whether or not to return transition, emission, and initial probabilities. <code>FALSE</code> by default.
<code>conditional_se</code>	Return conditional standard errors of coefficients. See <a href="#">vcov.mhmm</a> for details. <code>TRUE</code> by default.
<code>log_space</code>	Make computations using log-space instead of scaling for greater numerical stability at cost of decreased computational performance. Default is <code>FALSE</code> .
<code>...</code>	Further arguments to <a href="#">vcov.mhmm</a> .

**Details**

The `summary.mhmm` function computes features from a mixture hidden Markov model and stores them as a list. A `print` method prints summaries of these: log-likelihood and BIC, coefficients and standard errors of covariates, means of prior cluster probabilities, and information on most probable clusters.

**Value**

**transition\_probs** Transition probabilities. Only returned if `parameters = TRUE`.  
**emission\_probs** Emission probabilities. Only returned if `parameters = TRUE`.  
**initial\_probs** Initial state probabilities. Only returned if `parameters = TRUE`.  
**logLik** Log-likelihood.  
**BIC** Bayesian information criterion.  
**most\_probable\_cluster** The most probable cluster according to posterior probabilities.

**coefficients** Coefficients of covariates.  
**vcov** Variance-covariance matrix of coefficients.  
**prior\_cluster\_probabilities** Prior cluster probabilities (mixing proportions) given the covariates.  
**posterior\_cluster\_probabilities** Posterior cluster membership probabilities.  
**classification\_table** Cluster probabilities (columns) by the most probable cluster (rows).

### See Also

[build\\_mhmm](#) and [fit\\_model](#) for building and fitting mixture hidden Markov models; and [mhmm\\_biofam](#) for information on the model used in examples.

### Examples

```
# Loading mixture hidden Markov model (mhmm object)
# of the biofam data
data("mhmm_biofam")

# Model summary
summary(mhmm_biofam)
```

---

trim\_model

*Trim Small Probabilities of Hidden Markov Model*


---

### Description

Function `trim_model` tries to set small insignificant probabilities to zero without decreasing the likelihood.

### Usage

```
trim_model(
  model,
  maxit = 0,
  return_loglik = FALSE,
  zerotol = 1e-08,
  verbose = TRUE,
  ...
)
```

### Arguments

<code>model</code>	Model of class <code>hmm</code> or <code>mhmm</code> for which trimming is performed.
<code>maxit</code>	Number of iterations. After zeroing small values, the model is refitted, and this is repeated until there is nothing to trim or <code>maxit</code> iterations are done.
<code>return_loglik</code>	Return the log-likelihood of the trimmed model together with the model object. The default is <code>FALSE</code> .

zerotol	Values smaller than this are trimmed to zero.
verbose	Print results of trimming. The default is TRUE.
...	Further parameters passed on to <a href="#">fit_model</a> .

**See Also**

[build\\_hmm](#) and [fit\\_model](#) for building and fitting hidden Markov models; and [hmm\\_biofam](#) for information on the model used in the example.

**Examples**

```
data("hmm_biofam")

# Testing if changing parameter values smaller than 1e-03 to zero
# leads to improved log-likelihood.
hmm_trim <- trim_model(hmm_biofam, zerotol = 1e-03, maxit = 10)
```

vcov.mhmm

---

*Variance-Covariance Matrix for Coefficients of Covariates of Mixture  
Hidden Markov Model*

---

**Description**

Returns the asymptotic covariances matrix of maximum likelihood estimates of the coefficients corresponding to the explanatory variables of the model.

**Usage**

```
## S3 method for class 'mhmm'
vcov(object, conditional = TRUE, threads = 1, log_space = FALSE, ...)
```

**Arguments**

object	Object of class mhmm.
conditional	If TRUE (default), the standard errors are computed conditional on other model parameters. See details.
threads	Number of threads to use in parallel computing. Default is 1.
log_space	Make computations using log-space instead of scaling for greater numerical stability at cost of decreased computational performance. Default is FALSE.
...	Additional arguments to function <code>jacobian</code> of <code>numDeriv</code> package.

**Details**

The conditional standard errors are computed using analytical formulas by assuming that the coefficient estimates are not correlated with other model parameter estimates (or that the other parameters are assumed to be fixed). This often underestimates the true standard errors, but is substantially faster approach for preliminary analysis. The non-conditional standard errors are based on the numerical approximation of the full Hessian of the coefficients and the model parameters corresponding to nonzero probabilities. Computing the non-conditional standard errors can be slow for large models as the Jacobian of analytical gradients is computed using finite difference approximation.

**Value**

Matrix containing the variance-covariance matrix of coefficients.

# Index

## \* datasets

- biofam3c, 3
  - colorpalette, 23
  - hmm\_biofam, 41
  - hmm\_mvad, 43
  - mhmm\_biofam, 49
  - mhmm\_mvad, 52
- alphabet, 71
- alphabet (seqdef), 71
- biofam, 3, 4, 41, 43, 49, 52
- biofam3c, 3, 43, 52, 82, 87
- build\_hmm, 4, 28, 43–46, 59, 60, 70, 75, 76, 82, 87, 92
- build\_lcm, 8, 28, 64, 76
- build\_mhmm, 12, 28, 45, 52, 53, 57, 64, 66, 70, 76, 77, 91
- build\_mm, 18, 28, 59, 76
- build\_mmm, 19, 28, 64, 76
- cluster\_names, 22
- cluster\_names<-, 22
- cmdscale, 55, 80, 85
- colorpalette, 23, 56, 57, 68, 81, 86
- estimate\_coef, 24
- fit\_hmm (seqHMM-deprecated), 72
- fit\_mhmm (seqHMM-deprecated), 72
- fit\_model, 6, 10, 14, 21, 24, 43–46, 52, 53, 57, 60, 66, 70, 75, 77, 82, 87, 91, 92
- formula, 9, 13, 20, 77
- forward\_backward, 36
- gridplot, 37, 67, 79, 82, 87
- hidden\_paths, 40, 55, 57, 80, 82, 85, 87
- hmm\_biofam, 40, 41, 46, 60, 82, 87, 92
- hmm\_mvad, 43, 60
- igraph.plotting, 59, 65
- layout\_, 59, 64
- logLik.hmm, 44
- logLik.mhmm, 45
- mc\_to\_sc, 46, 47, 59, 60, 64
- mc\_to\_sc\_data, 47
- menu, 55, 64
- mhmm\_biofam, 49, 66, 70, 91
- mhmm\_mvad, 52, 66
- mssplot, 28, 40, 54
- mvad, 43, 44, 52, 53
- nloptr, 25, 26, 73, 74
- par, 59, 65
- plot.hmm, 6, 19, 28, 58
- plot.igraph, 60, 66
- plot.mhmm, 10, 14, 21, 28, 57, 63
- plot.ssp, 38, 67, 79, 82, 87
- plot.stslist, 55, 80, 85
- plot\_colors, 23, 68
- posterior\_probs, 69
- print.hmm, 69
- print.mhmm (print.hmm), 69
- print.summary.mhmm (print.hmm), 69
- separate\_mhmm, 10, 14, 21, 28, 70
- seqdef, 5, 9, 12, 18, 20, 40, 47, 56, 57, 71, 71, 75, 77, 80, 81, 84–86
- seqHMM, 71
- seqHMM-deprecated, 72
- seqIplot, 40
- seqplot, 55, 57, 60, 66, 80, 82, 85, 87
- seqstatf, 71
- seqstatf (seqdef), 71
- simulate\_emission\_probs (simulate\_initial\_probs), 76
- simulate\_hmm, 74, 77
- simulate\_initial\_probs, 76

`simulate_mhmm`, 75, 76  
`simulate_transition_probs`  
    (`simulate_initial_probs`), 76  
`ssp`, 37, 38, 67, 79, 87  
`ssplot`, 28, 40, 47, 67, 75, 77, 84  
`state_names`, 88  
`state_names<=`, 89  
`summary.mhmm`, 10, 14, 21, 28, 90  
  
`trim_hmm` (`seqHMM`-deprecated), 72  
`trim_model`, 91  
  
`vcov.mhmm`, 90, 92