Package ‘depmix’

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Description Fits (multigroup) mixtures of latent or hidden Markov models on mixed categorical and continuous (timeseries) data. The Rdonlp2 package can optionally be used for optimization of the log-likelihood and is available from R-forge.
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Depmix utility functions

Description

These functions are used internally by depmix functions. They should not be called directly unless you know what you’re doing.

Usage

checkSetRecode(dat,dmm,tdcov,printlevel=1)
recode(dat,xm)
fblo(x,i,bigB)
fbuo(x,i,bigB)
ppar(x,z)
recitt(itemtypes)
pp(x)
np(x)
pa2conr(x)
paridx(nstates,itemtypes,mat,idx1=0,idx2=0,it=0,comp=1,group=1)
fresp(x,pars)
bdiag(x)
cl2st(cluster,dat,dmm)
cl2stob(cluster,dat,dmm)
kmstart(dat,dmm)
poststart(dat,dmm)
tr2stin(sttr)

Arguments

dat,xm See markovdata help.
dmm,itemtypes,nstates See dmm help.
cluster Some clustering of the data, eg from kmeans or posterior estimates that can be used to arrive at starting values for parameters.
sttr transition matrix starting values.
printlevel,tdcov See depmix help.
x,mat,idx1,idx2,it,comp,group
A vector/matrix(name)/indices et cetera.
i,z,bigB,pars More internal stuff.
Details

Function bdiag takes as argument a list of matrices and returns the blockdiagonal matrix formed from these, and the other entries padded with zeroes. This function is from package assist by Chunlei Ke and Yuedong Wang.

Value

Most of these functions are used for their side-effect, ie sending stuff to C-routines, or returning recoded stuff (data, itemtypes) et cetera.

Author(s)

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discrimination Discrimination Learning Data

Description

This data set is from a simple discrimination learning experiment. It consists of 192 binary series of responses of different lengths. This is a subset of the data described by Raijmakers et al. (2001), and it is analyzed much more extensively using latent Markov models and depmix in Schmittmann et al. (2006) and Visser et al. (2006).

Usage

data(discrimination)

Format

An object of class markovdata.

Source


Description

dmm creates an object of class dmm, a dependent mixture model.

lca creates an object of class dmmLlca, a latent class model or an independent mixture model.

Usage

dmm(nstates, itemtypes, modname = NULL, fixed = NULL, stval = NULL, conrows = NULL, conpat = NULL, tdfix = NULL, tdst = NULL, linmat = NULL, snames = NULL, inames = NULL)

## S3 method for class 'dmm'
summary(object, specs=FALSE, precision=3, se=NULL, ...)

lca(nclasses, itemtypes, modname = NULL, fixed = NULL, stval = NULL, conrows = NULL, conpat = NULL, linmat = NULL, snames = NULL, inames = NULL)

Arguments

nstates The number of latent states/classes of the model.
nclasses The number of classes of an lca model, i.e. the number of states in a dmm model. They are now called classes because they do not change over time.
itemtypes A vector of length nitems providing the type of measurement, 1 for gaussian data, 2 for a binary item, n>3 for categorical items with n answer possibilities. Answer categories are assumed to be unordered categorical. Ordinal responses can be implemented using inequality and/or linear constraints.
modname A character string with the name of the model, good when fitting many models. Components of mixture models keep their own names. Names are printed in the summary. Boring default names are provided.
fixed A vector of length the number of parameters of the model indicating whether parameters are fixed (0) or not (>0). This may be identical to conpat (see below).
stval Start values of the parameters. These will be random if not specified. Start values must be specified (for all parameters) if there are fixed parameters.
conrows Argument conrows can be used to specify general constraints between parameters. See details below.
conpat Argument conpat can be used to specify fixed parameters and equality constraints. It cannot be used in conjunction with fixed. See details below.
The first is a logical vector indicating (with 1's) which parameters are dependent on covariates (it should have length npars). tdst provides the starting values for the regression parameters. Using tdcov=TRUE in fitdmm will actually fit the regression parameters. The covariate itself has to be specified in the data as "covariate" (see help on markovdata) and should be scaled to 0-1.

linmat
A complete matrix of linear constraints. This argument is intended for internal use only, it is used by the fit routine to re-create the model with the fitted parameter values. Warning: use of this argument results in complete replacement of the otherwise created matrix A, which contains e.g. sum constraints for transition matrix parameters. If linmat is provided, make sure it is correct, otherwise strange results may occur in fitting models.

snames
Names for the states may be provided in snames. Defaults are State1, State2 etc. These are printed in the summary.

inames
Names for items may be provided in inames. Defaults are Item1, Item2 etc. They are printed in the summary.

dmm
Object of class dmm.

precision
Precision sets the number of digits to be printed in the summary functions.

se
Vector with standard errors, these are passed on from the summary.fit function if and when ses are available.

specs,...
Internal use.

object
An object of class dmm.

Details
The function dmm creates an object of class dmm and sets random initial parameter values if these are not provided. Even though dmm is not a mixture of Markov models, the mixture parameter is included in the parameter vector. This is important when specifying constraints. Parameters are ordered as follows: the first parameter(s) are the mixing proportions of the mixture of Markov and/or latent class models. I.e., when a single latent class model or a single Markov chain is fitted, this mixture proportion has value 1.0 and is it is fixed in estimation. After the mixing proportions, the next parameters in the parameter vector are the transition matrix parameters, the square of nstates in row-major order. That is, first the transition probabilities from state 1 to all the other states are given, then the probabilities from state 2 to all the other states etc. Next are the observation matrix parameters. These are provided consecutively for each state/class. I.e a trichotomous item model with two states has 6 observation parameters; the first three are the probabilities of observing category 1, 2 and 3 respectively in state 1 (which sum to one), and then similarly for state 2. As another example: suppose we have model for one binary item and one gaussian item, in that order, we would have 4 observation parameters for each state, first the probabilities of observing a symbol from category 1 or 2 in state 1, the two parameters, the mean and standard deviation for state 1, and then the same state 2 (see the example in fitdmm with data from rudy). Finally the initial state probabilities are provided, in the order of the states. In the case of a latent class model or a finite mixture model, these parameters are usually denote as the mixture proportions.

Linear constraints can be set using arguments conrows and conpat. conrows must be contain nc by npars values, in row major order, with nc the number of contraints to be specified. conrows is used to define general linear constraints. A row of conrows must contain the partial derivatives of a general linear constraint with respect to each of the parameters. Suppose we want the constraint
x1 \cdot 2 \cdot x2 = 0\), one row of conrows should contain a 1 in position one and -2 in position and zeroes in the remaining positions. In the function \texttt{mixdmm conrows} is understood to specify linear constraints on the mixing proportions only. As a consequence, it is not possible to easily constrain parameters between components of a mixture model.

\texttt{conpat} can be used as a shortcut for both fixed and conrows. It must be a single vector of length npars containing 0's (zeros) for fixed parameters, 1's (ones) for free parameters and higher numbers for possibly equality constrained parameters. E.g. \texttt{conpat=c(1,1,0,2,2,3,3,3)} would indicate that pars 1 and 2 are freely estimated, par 3 is fixed at its startvalue (which must be provided in this case), par 4 and 5 are to estimated equal and pars 6, 7 and 8 are also to be estimated equal.

\textbf{Value}

\texttt{dmm} returns an object of class \texttt{dmm} which has its own summary method. This will print the parameter values, itemtypes, number of (free) parameters, and the number of states. There is no print method. Using print will print all fields of the model which is a list of the following:

- \texttt{modname}: See above.
- \texttt{nstates}: See above.
- \texttt{snames}: See above.
- \texttt{nitem}: The number of items (\texttt{=length(itemtypes)}).
- \texttt{itemtypes}: See above.
- \texttt{inames}: See above.
- \texttt{npars}: The total parameter count of the model.
- \texttt{npars.total}: The total number of parameters of when the covariate parameters are included.
- \texttt{freepars}: The number of freely estimated parameters (it is computed as \texttt{sum(as.logical(fixed))}-\texttt{rank(qr(A))}).
- \texttt{freepars.nod}: The number of freely estimated parameters (it is computed as \texttt{sum(as.logical(fixed))}-\texttt{rank(qr(A))}); this version without the covariate parameters.
- \texttt{pars}: A vector of length npars containing parameter values.
- \texttt{fixed}: \texttt{fixed} is a (logical) vector of length npars specifying which parameters are fixed and which are not.
- \texttt{A}: The matrix \texttt{A} contains the general linear constraints of the model. \texttt{nrow(A)} is the number of linear constraints. \texttt{A} starts with a number of rows for the sum constraints for the transition, observation and initial state parameters, after which the user provided constraints are added.
- \texttt{bu,bl}: \texttt{bu} and \texttt{bl} represent the upper and lower bounds of the parameters and the constraints. These vectors are each of length npars + \texttt{nrow(A)}.
- \texttt{blin,bulin}: The lower and upper bounds of the linear constraints.
- \texttt{td,tdin,tdtr,tdob,tdfit}: Logicals indicating whether there covariates, in which parameters they are, and whether they are estimated or not (the latter is used to decide whether to print those values or not).
- \texttt{st}: Logical indicating whether the model has user specified starting values.
lca returns an object of class dmm, lca, and is otherwise identical to a dmm object. The only difference is that the transition matrix parameters are irrelevant, and consequently they are not printed in the summary function.

Author(s)
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References

See Also
mixdmm on defining mixtures of dmm’s, mgdmm for defining multi group models, and generate for generating data from models.

Examples

```r
# create a 2 state model with one continuous and one binary response
# with start values provided in st
st <- c(1,0.9,0.1,0.2,0.8,2,1,0.7,0.3,5,2,0.2,0.8,0.5,0.5)
mod <- dmm(nsta=2, itemt=c(1,2), stval=st)
summary(mod)

# 2 class latent class model with equal conditional probabilities in each class
stv=c(1,rep(c(0.9,0.1),5),rep(c(0.1,0.9),5),0.5,0.5)
# here the conditional probs of the first item are set equal to those in
# the subsequent items
conpat=c(1,rep(c(2,5),5),rep(c(4,5),5),1,1)
lc=lca(ncl=2, itemtypes=rep(2,5), conpat=conpat, stv=stv)
summary(lc)
```

Description

*fitdmm* fits mixtures of hidden/latent Markov models on arbitrary length time series of mixed categorical and continuous data. This includes latent class models and finite mixture models (for time series of length 1), which are in effect independent mixture models.

*posterior* computes the most likely latent state sequence for a given dataset and model.
Usage

fitdmm(dat, dmm, printlevel = 1, poster = TRUE, tdcov = 0,
   ses = TRUE, method = "optim", vfactor = 15, der = 1, iterlim = 100,
   kmst = !dmm$st, kmrep = 5, postst = FALSE)
loglike(dat, dmm, tdcov = 0, grad = FALSE, hess = FALSE, set
   = TRUE, grInd = 0, sca = 1, printlevel = 1)
posterior(dat, dmm, tdcov = 0, printlevel = 1)
computeSes(dat, dmm)
bootstrap(object, dat, samples = 100, pvalonly = 0, ...)
## S3 method for class 'fit'
summary(object, precision = 3, fd = 1, ...)
oneliner(object, precision = 3)

Arguments

dat An object (or list of objects) of class md, see markovdata. If dat is a list of objects of class md a multigroup model is fitted on these data sets.
dmm An object (or a list of objects) of class dmm, see dmm. If dmm is a list of objects of class dmm, these are taken to components of a mixture of dmm's model and will be coerced to class mixdmm. In any case, the model that is fitted is a multigroup mixture of dmm's with default ngroups = 1 and number of components = 1.
printlevel printlevel controls the output provided by the C-routines that are called to optimize the parameters. The default of 1 provides minimal output: just the initial and final loglikelihood of the model. Setting higher values will provide more output on the progress the iterations.
poster By default posteriors are computed, the result of which can be found in fit$post.
method This is the optimization algorithm that is used. donlp2 from the Rdonlp2 package is the default method. There is optional support for NPSOL.
der Specifies whether whether derivatives are to be used in optimization.
vfactor vfactor controls optimization in optim and nlm. Since in those routines there is no possibility for enforcing constraints, constraints are enforced by adding a penalty term to the loglikelihood. The penalty term is printed at the end of optimization if it is not close enough to zero. This may have several reasons. When parameters are estimated at bounds for example. This can be solved by fixing those parameters on their boundary values. When this is not acceptable the vfactor may be increased such that the penalty is larger and the probability that they actually hold in the fitted model is correspondingly higher.
tdcov Logical, when set to TRUE, given that the model and data have covariates, the corresponding parameters will be estimated.
ses Logical, determines whether standard errors are computed after optimization.
iterlim The iteration limit for npsol, defaults to 100, which may be too low for large models.
grad logical: if TRUE the gradients are returned.
The function `fitdmm` optimizes the parameters of a mixture of dms using a general purpose optimization routine subject to linear constraints on the parameters.

Value

`fitdmm` returns an object of class `fit` which has a summary method that prints the summary of the fitted model, and the following fields:

- `dateLtimeusedLtotMem`: The date that the model was fitted, the time it took to so and the memory usage.
- `loglike`: The loglikelihood of the fitted model.
- `aic`: The AIC of the fitted model.
- `bic`: The BIC of the fitted model.
- `mod`: The fitted model.
- `post`: See function posterior for details.

`loglike` returns a list of the following:

- `logl`: The loglikelihood.
The 

gr, grset: gr contains the gradients. grset is a logical vector giving information as to which gradients are set, currently all gradients are set except the gradients for the mixing proportions.

hs, hsset: hs contains the hessian. hsset is a logical vector giving information as to which elements are computed.

posterior: returns lists of the following:

states: A matrix of dimension 2+sum(nstates) by sum(length(ntimes)) containing in the first column the a posteriori component, in the second column the a posteriori state and in the remaining column the posterior probabilities of all states.

comp: Contains the posterior component number for each independent realization; all ones for a single component model.

compute5es: returns a vector of length npars with the standard errors and a matrix hs with the hessian used to compute them. The routine is not fail safe and can produce errors, i.e., when the (corrected) hessian is singular; a warning is issued when the hessian is close to being singular.

bootstrap: returns an object of class fit with three extra fields, the bootstrapped standard errors, bse, a matrix with goodness-of-fit measures of the bootstrap samples, i.e., logl, AIC and BIC and pbetter, which is the proportion of bootstrap samples that resulted in better fits than the original model.

summary.fit: pretty-prints the outputs.

oneliner: returns a vector of loglike, aic, bic, mod$npars, mod$freepars, date.

Note

fitdmm fits time series of arbitrary length and mixtures of dmm's, where, to the best of my knowledge, other packages are limited due to the different optimization routines that are commonly used for these types of models.

Author(s)

Ingmar Visser <i.visser@uva.nl>, Development of this package was supported by European Commission grant 51652 (NEST) and by a VENI grant from the Dutch Organization for Scientific Research (NWO).

References


See Also

dmm, markovdata
Examples

```r
# COMBINED RT AND CORRECT/INCORRECT SCORES from a 'switching' experiment

data(speed)
mod <- dmm(nsta=2,itemt=c(1,2))  # gaussian and binary items
ll <- loglike(speed,mod)
fit1 <- fitdmm(dat=speed,dmm=mod)
summary(fit1)
ll <- loglike(speed,fit1)

# bootstrap
## Not run:
pst <- posterior(dat=speed,dmm=fit1)
bs <- bootstrap(fit1,speed,samples=50)

## End(Not run) # end not run

# add some constraints using conpat
conpat=rep(1,15)
conpat[1]=0
conpat[14:15]=0
conpat[8:9]=0

# use starting values from the previous model fit, except for the guessing
# parameters which should really be 0.5
stv=c(1,0.96,0.04,0.94,0.916,0.52,0.2,0.5,0.5,0.6,0.39,0.24,0.098,0.9,0.1)
mod=dmm(nstates=2,itemt=c("n",2),stval=stv,conpat=conpat)
fit2 <- fitdmm(dat=speed,dmm=mod)
summary(fit2)

# add covariates to the model to incorporate the fact the accuracy pay off changes per trial
# 2-state model with covariates + other constraints

## Not run:

conpat=rep(1,15)
conpat[1]=0
conpat[8:9]=0
conpat[14:15]=0
conpat[2]=2
conpat[5]=2
stv=c(1,0.9,0.1,0.1,0.9,5.5,0.2,0.5,0.5,0.5,0.6,0.4,0.25,0.9,0.1,0.1)
tdfix=rep(0,15)
tdfix[2:5]=1
stcov=rep(0,15)
stcov[2:5]=c(-0.4,0.4,0.15,-0.15)

mod<-dmm(nstates=2,itemt=c("n",2),stval=stv,conpat=conpat,tdfix=tdfix,tdst=stcov,
modname="twoboth+cov")
```
fit3 <- fitdmm(dat = speed, dmm = mod, tdcov = 1, der = 0, ses = 0, vfa = 80)
summary(fit3)

# split the data into three time series
data(speed)

r1 = markovdata(dat = speed[1:168,], item = itemtypes(speed))
r2 = markovdata(dat = speed[169:302,], item = itemtypes(speed))
r3 = markovdata(dat = speed[303:439,], item = itemtypes(speed))

# define 2-state model with constraints
conpat = rep(1, 15)
conpat[1] = 0
conpat[8:9] = 0
conpat[14:15] = 0
stv = c(1, 0.9, 0.1, 0.1, 0.9, 0.5, 0.2, 0.5, 0.5, 0.4, 0.25, 0.9, 0.1, 0.1)
mod = dmm(nstates = 2, item = c("n", 2), stval = stv, conpat = conpat)

# define 3-group model with equal transition parameters, and no
# equalities between the observer parameters
mgr <- mgdmm(dmm = mod, ng = 3, trans = TRUE, observer = FALSE)

fitmg <- fitdmm(dat = list(r1, r2, r3), dmm = mgr)
summary(fitmg)

## End(Not run) # end not run

## LEARNING DATA AND MODELS (with absorbing states)

## Not run:
data(discrimination)

# all or none model with error prob in the learned state
fixed = c(0, 0, 0, 1, 1, 1, 1, 0, 0, 0, 0)
stv = c(1, 1, 0.0, 0.03, 0.97, 0.1, 0.9, 0.5, 0.5, 0, 1)
allor <- dmm(nstates = 2, itemtypes = 2, fixed = fixed, stval = stv, modname = "All-or-none")

# Concept identification model: learning only after an error
st = c(1, 1, 0, 0, 0, 0.5, 0.5, 0.25, 0.25, 0.05, 0.95, 0, 1, 1, 0, 0.25, 0.375, 0.375)
# fix some parameters
fx = rep(0, 19)
fx[8:12] = 1
fx[17:19] = 1

# add a couple of constraints
conr1 <- rep(0, 19)
conr1[9] = 1
conr1[10] = -1
conr2 <- rep(0, 19)
conr2[18] = 1
conr2[19] = -1
conr3 <- rep(0, 19)
generate


conr3[8]=1
conr3[17]=-2
conr=c(conr1,conr2,conr3)
cim <- dmm(nstates=3,itemtypes=2, fixed=fx, conrows=conr, stval=st, modname="CIM")

# define a mixture of the above models ...
mix <- mixdmm(dmm=list(allor,cim),modname="MixAllCim")

# ... and fit it on the combined data discrimination
fitmix <- fitdmm(discrimination,mix)
summary(fitmix)

## End(Not run) # end not run

---

**generate**

*Generate data from a dependent mixture model*

**Description**

`generate` generates a dataset according to a given `dmm`.

**Usage**

```r
generate(ntimes,dmm,nreal=1)
```

**Arguments**

- `ntimes`: The number of repeated measurements, i.e., the length of the time series (this may be a vector containing the lengths of independent realizations).
- `dmm`: Object of class `dmm` or `mixdmm`.
- `nreal`: The number of independent realizations that is to generated. Each of them will have the dimension of `ntimes`; all this does is replace `ntimes` by `rep(ntimes,nreal)`.

**Details**

`generate` generates a date set of the specified dimensions `ntimes` and `nreal` using the parameter values in `dmm`, which should be an object of class `dmm` or `mixdmm`. `generate` does not handle multi group models, which can be run separately.

This function is used in the bootstrap'ing routine to compute standard errors based on parametric bootstraps.
Markovdata creates an object of class `markovdata`, to be used by `fitdmm`.

### Usage

```r
markovdata(dat, itemtypes, nitems = length(itemtypes), ntimes = length(as.matrix(dat))/nitems, replicates = rep(1, length(ntimes)), inames = NULL, dname = NULL, xm = NA)
```

```r
# S3 method for class 'md'
summary(object, ...)
# S3 method for class 'md'
plot(x, nitems = 1:(min(5, dim(x)[2])),
     nind = 1:(min(5, length(attributes(x)$ntimes))),...)
```
# S3 method for class 'md'
print(x, ...)

dname(object)
names(object)
itemtypes(object)
replicates(object)

ncov(object)
inames(object)
nitems(object)
ind(object)

Arguments

dat An R object to be coerced to markovdata, a data frame or matrix.

itemtypes A vector providing the types of measurement with possible values 'continuous', 'categorical', and 'covariate'. This is mainly only used to rearrange the data when there are covariates in such a way that the covariate is in the last column. Only one covariate is supported in estimation of models.

ntimes The number of repeated measurements, ie the length of the time series (this may be a vector containing the lengths of independent realizations). It defaults the number of rows of the data frame or data matrix.

replicates Using this argument case weights can be provided. This is particularly usefull in eg latent class analysis with categorical variables when there usually are huge numbers of replicates, ie identical response patterns. depmix computes the raw data log likelihood for each case separately. Thus, when there are many replicates of a case a lot of computation time is saved by specifying case weights instead of providing the full data set.

inames The names of items. These default to the column names of matrices or dataframes.

dname The name of the dataset, used in summary, print and plot functions.

xm xm is the missing data code. It can be any value but zero. Missing data are recoded into NA.

object,x An object of class md.

Further arguments passed on to plot and summary.

nitems,nind In the plot function, these arguments control which data are to be plotted, ie nitems indicates a range of items, and nind a range of realizations, respectively.

Details

The function markovdata coerces a given data frame or matrix to be an object of class md such that it can be used in fitdmm. The md object has its own summary, print and plot methods.

The functions dname, itemtypes, ntimes, and replicates retrieve the respective attributes with these names; similarly ncov, nitems, inames, and ind retrieve the number of covariates, the number of items (the number of columns of the data), the column names and the number of independent realizations respectively.
Value

An md-object is a matrix of dimensions \(\text{sum(ntimes)} \times \text{nitems}\), containing the measured variables and covariates rearranged such that the covariate appears in the last column. The column names are inames and the matrix has three further attributes:

- **dname**: The name of the data set.
- **itemtypes**: See above.
- **ntimes**: See above. This will be a vector computed as ntimes=rep(ntimes,nreal).
- **replicates**: The number of replications of each case, used as weights in computing the log likelihood.

Author(s)

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See Also

dmm, depmix

Examples

```r
x &lt;= rnorm(100,10,2)
y &lt;= ifelse(runif(100) &lt; 0.5,0,1)
z &lt;= matrix(c(x,y),100,2)
md &lt;= markovdata(z,itemtypes=c("cont","cat"))
summary(md)

data(speed)
summary(speed)
plot(speed,nind=2)

# split the data into three data sets
# (to perform multi group analysis)
r1 &lt;= markovdata(dat=speed[1:168],item=itemtypes(speed))
r2 &lt;= markovdata(dat=speed[169:302],item=itemtypes(speed))
r3 &lt;= markovdata(dat=speed[303:439],item=itemtypes(speed))
summary(r2)
```

**mgdmm**

*Multi group model specification*

Description

mgdmm creates an object of class mgd, a multi-group model, from a given model of either class dmm or class mixdmm or lists of these.
mgdmm

Usage

mgdmm(dmm, ng=1, modname=NULL, trans=FALSE, obser=FALSE, init=FALSE, conpat=NULL)
## S3 method for class 'mgd'
summary(object, specs=FALSE, precision=3, se=NULL, ...)

Arguments

modname A character string with the name of the model, good when fitting many models. Components of mixture models keep their own names. Names are printed in the summary. Boring default names are provided.

dmm Object (or list of objects) of class dmm; see details below.

ng Number of groups for a multigroup model.

trans,obser,init Logical arguments specify whether transition parameters, observation parameters and initial state parameters should be estimated equal across groups.

conpat Can be used to specify general linear constraints. See dmm for details.

precision Precision sets the number of digits to be printed in the summary functions.

se Vector with standard errors, these are passed on from the summary.fit function if and when ses are available.

specs,... Internal use.

object An object of class mgd.

Details

The function mgdmm can be used to define an mgd-model or multi group dmm. Its default behavior is to create ng copies of the dmm argument, thereby providing identical starting values for each group’s model. If the dmm argument is a list of models of length ng, the starting values of those models will be used instead. This may save quite some cpu time when fitting large models by providing the parameter values of separately fitted models as starting values. Currently, depmix does not automatically generate starting values for multi group models.

Value

mgdmm returns an object of class mgd which contains all the fields of an object of class dmm and the following extra:

ng ng is the number of groups in the multigroup model.

mixmod mixmod is a list of length ng of mixdmm models for each group.

itemtypes See above.

npars,freepars,pars, fixed,A,b1, bu

The same as above but now for the combined model, here npars equals the sum of npars of the component models plus the mixing proportions.
mixdmm

Description
mixdmm creates an object of class mixdmm, i.e., a mixture of dmm's, given a list of component models of class dmm.

Usage

mixdmm(dmm, modname=NULL, mixprop=NULL, conrows=NULL)

## S3 method for class 'mixdmm'
summary(object, specs=FALSE, precision=3, se=NULL, ...)

Arguments

dmm A list of dmm objects to form the mixture.
modname A character string with the name of the model, good when fitting many models. Components of mixture models keep their own names. Names are printed in the summary. Boring default names are provided.
conrows Argument conrows can be used to specify general constraints between parameters.
mixprop Argument mixprop can be used to set the initial values of the mixing proportions of a mixture of dmm's.
mixdmm

precision  Precision sets the number of digits to be printed in the summary functions.
object      An object of class mixdmm.
specs,...   Internal use. Not functioning currently.
se          Vector with standard errors, these are passed on from the summary.fit function if and when ses are available.

Details

The function mixdmm can be used to define a mixture of dmm's by providing a list of such objects as argument to this function. See the dmm helpfile on how to use the conrows argument. Note that it has to be of length npars, ie including all parameters of the model and not just the mixing proportions.

Value

mixdmm returns an object of class mixdmm which has the same fields as a dmm object. In addition it has the following fields:

nrcomp       The number of components of the mixture model.
mod          A list of the component models, that is a list of objects of class dmm.

Author(s)

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See Also

dmm on defining single component models, and mgdmm on defining multi group models. See generate for generating data.

Examples

# define component 1
# all or none model with error prob in the learned state
fixed = c(0,0,0,1,1,1,1,0,0,0,0)
stv = c(1,1,0,0.07,0.93,0.9,0.1,0.5,0.5,0.1)
allor <- dmm(nstates=2,itemtypes=2,fixed=fixed,stval=stv,modname="All-or-none")

# define component 2
# Concept identification model: learning only after an error
st=c(1,1,0,0,0.5,0.5,0.25,0.25,0.25,0.25,0.25,0.25,0.25,0.25,0.25,0.375,0.375)
# fix some parameters
fx=rep(0,19)
fx[8:12]=1
fx[17:19]=1
# add a couple of constraints
conr1 <- rep(0,19)
conr1[9]=1
conr1[10]=-1
conr2 <- rep(0,19)
Description

This data set is a bivariate series of reaction times and accuracy scores of a single subject switching between slow and accurate responding and fast guessing on a lexical decision task. The slow and accurate responding, and the fast guessing can be modelled using two states, with a switching regime between them. The dataset further contains a third variable called Pacc, representing the relative pay-off for accurate responding, which is on a scale of zero to one. The value of Pacc was varied during the experiment to induce the switching. This data set is a subset of data from experiment 2 in Van der Maas et al, 2005.

Usage

data(speed)

Format

An object of class markovdata.

Source

Han L. J. Van der Maas, Conor V. Dolan and Peter C. M. Molenaar (2005), Phase Transitions in the Trade-Off between Speed and Accuracy in Choice Reaction Time Tasks. Manuscript in revision.
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