

# Package: ordinalClust (via r-universe)

May 21, 2026

**Type** Package

**Title** Ordinal Data Clustering, Co-Clustering and Classification

**Version** 1.3.5.1

**Date** 2026-04-17

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**Description** Ordinal data classification, clustering and co-clustering using model-based approach with the BOS (Binary Ordinal Search) distribution for ordinal data (Christophe Biernacki and Julien Jacques (2016) <[doi:10.1007/s11222-015-9585-2](https://doi.org/10.1007/s11222-015-9585-2)>).

**License** GPL (>= 2)

**Imports** Rcpp (>= 0.12.11), methods

**LinkingTo** Rcpp, RcppProgress, RcppArmadillo, BH

**Suggests** knitr, rmarkdown, caret, ggplot2

**VignetteBuilder** knitr

**LazyData** true

**Depends** R (>= 3.3)

**NeedsCompilation** yes

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**Repository** <https://cran.r-universe.dev>

**Date/Publication** 2026-04-21 22:03:59 UTC

**RemoteUrl** <https://github.com/cran/ordinalClust>

**RemoteRef** HEAD

**RemoteSha** 989959cc6b5a33c3bfeb6fb073cd35b5acff4b62

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bosclassif	<i>Function to perform a classification</i>
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### Description

This function performs a classification algorithm on a dataset with ordinal features, and a label variable that belongs to  $(1,2,\dots,kr)$ . The classification function provides two classification models. The first model, (chosen by the argument  $kc=0$ ), is a multivariate BOS model with the assumption that, conditional on the class of the observations, the features are independent. The second model is a parsimonious version of the first model. Parsimony is introduced by grouping the features into clusters (as in co-clustering) and assuming that the features of a cluster have a common distribution.

### Usage

```
bosclassif(x, y, idx_list=c(1), kr, kc=0, init, nbSEM, nbSEMBurn,
           nbindmini, m=0, percentRandomB=0)
```

### Arguments

x	Matrix made of ordinal data of dimension $N \times J_{tot}$ . The features with same numbers of levels must be placed side by side. The missing values should be coded as NA.
y	Vector of length N. It should represent the classes corresponding to each row of x. Must be labeled with numbers $(1,2,\dots,kr)$ .
idx_list	Vector of length D. This argument is useful when variables have different numbers of levels. Element d should indicate where the variables with number of levels m[d] begin in matrix x.
kr	Number of row classes.
kc	Vector of length D. The d-th element indicates the number of column clusters. Set to 0 to choose a classical multivariate BOS model.
m	Vector of length D. The d-th element defines the number of levels of the ordinal data.
nbSEM	Number of SEM-Gibbs iterations realized to estimate parameters.
nbSEMBurn	Number of SEM-Gibbs burn-in iterations for estimating parameters. This parameter must be inferior to nbSEM.

nbindmini	Minimum number of cells belonging to a block.
init	String that indicates the kind of initialisation. Must be one of the following strings: "kmeans", "random" or "randomBurnin".
percentRandomB	Vector of length 1. Indicates the percentage of resampling when init is equal to "randomBurnin".

### Value

Return an object. The slots are:

@zr	Vector of length N with resulting row partitions.
@zc	List of length D. The d-th item is a vector of length J[d] representing the column partitions for the group of variables d.
@J	Vector of length D. The d-th item represents the number of columns for d-th group of variables.
@W	List of length D. Item d is a matrix of dimension J*kc[d] such that W[j,h]=1 if j belongs to cluster h.
@V	Matrix of dimension N*kr such that V[i,g]=1 if i belongs to cluster g.
@icl	ICL value for co-clustering.
@kr	Number of row classes.
@name	Name of the result.
@number_distrib	Number of groups of variables.
@pi	Vector of length kr. Row mixing proportions.
@rho	List of length D. The d-th item represents the column mixing proportion for the d-th group of variables.
@dlist	List of length d. The d-th item represents the indexes of group of variables d.
@kc	Vector of length D. The d-th element represents the number of clusters column H for the d-th group of variables.
@m	Vector of length D. The d-th element represents the number of levels of the d-th group of variables.
@nbSEM	Number of SEM-Gibbs algorithm iteration.
@params	List of length D. The d-th item represents the blocks parameters for a group of variables d.
@xhat	List of length D. The d-th item represents the dataset of the d-th group of variables, with missing values completed.

### Author(s)

Margot Seloisse, Julien Jacques, Christophe Biernacki.

**Examples**

```
# loading the real dataset
data("dataqol.classif")

set.seed(5)

# loading the ordinal data
M <- as.matrix(dataqol.classif[,2:29])

# creating the classes values
y <- as.vector(dataqol.classif$death)

# sampling datasets for training and to predict
nb.sample <- ceiling(nrow(M)*2/3)
sample.train <- sample(1:nrow(M), nb.sample, replace=FALSE)

M.train <- M[sample.train,]
M.validation <- M[-sample.train,]
nb.missing.validation <- length(which(M.validation==0))
m <- c(4)
M.validation[which(M.validation==0)] <- sample(1:m, nb.missing.validation,replace=TRUE)

y.train <- y[sample.train]
y.validation <- y[-sample.train]

# configuration for SEM algorithm
nbSEM=50
nbSEMBurn=40
nbindmini=1
init="kmeans"

# number of classes to predict
kr <- 2
# different kc to test with cross-validation
kcol <- 1

res <- bosclassif(x=M.train,y=y.train,kr=kr,kc=kcol,m=m,
                 nbSEM=nbSEM,nbSEMBurn=nbSEMBurn,
                 nbindmini=nbindmini,init=init)

predictions <- predict(res, M.validation)
```

---

bosclust	<i>Function to perform a clustering</i>
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---

### Description

This function performs a clustering algorithm on ordinal data by using the multiple latent block model (see references for further details). It allows the user to define  $D$  groups of variables that have different numbers of levels. The BOS distribution is used, and the parameters inference is obtained using the SEM-Gibbs algorithm.

### Usage

```
bosclust(x, idx_list=c(1), kr, init, nbSEM, nbSEMBurn,
         nbindmini, m=0, percentRandomB=0)
```

### Arguments

<code>x</code>	Matrix made of ordinal data of dimension $N \times J_{tot}$ . The features with the same numbers of levels must be placed side by side. The missing values should be coded as NA.
<code>idx_list</code>	Vector of length $D$ . This argument is useful when variables have different numbers of levels. Element $d$ should indicate where the variables with number of levels $m[d]$ begin in matrix $x$ .
<code>kr</code>	Number of row clusters.
<code>m</code>	Vector of length $D$ . The $d$ -th element defines the number of levels of the ordinal data.
<code>nbSEM</code>	Number of SEM-Gibbs iterations realized to estimate the parameters.
<code>nbSEMBurn</code>	Number of SEM-Gibbs burn-in iterations for estimating parameters. This parameter must be inferior to <code>nbSEM</code> .
<code>nbindmini</code>	Minimum number of cells belonging to a block.
<code>init</code>	String that indicates the kind of initialisation. Must be one of the following words : "kmeans", "random" or "randomBurnin".
<code>percentRandomB</code>	Vector of length 1. Indicates the percentage of resampling when <code>init</code> is equal to "randomBurnin".

### Value

<code>@V</code>	Matrix of dimension $N \times kr$ such that $V[i,g]=1$ if $i$ belongs to cluster $g$ .
<code>@zr</code>	Vector of length $N$ with resulting row partitions.
<code>@pi</code>	Vector of length $kr$ . This corresponds to the row mixing proportions.
<code>@m</code>	Vector of length $D$ . The $d$ -th element represents the number of levels of the $d$ -th group of variables.
<code>@icl</code>	ICL value for clustering.

@name	Name of the result.
@params	List of length D. The d-th item stores the resulting position and precision parameters mu and pi.
@paramschain	List of length nbSEMBurn. The parameters of the blocks are stored for each iteration of the SEM-Gibbs algorithm.
@xhat	List of length D. The d-th item represents the dataset of the d-th group of variables, with missing values completed.
@zrchain	Matrix of dimension nbSEM*N. Row i represents the row cluster partitions at iteration i.
@pichain	List of length nbSEM. Item i is a vector of length kr that contains the row mixing proportions at iteration i.

### Author(s)

Margot Seloisse, Julien Jacques, Christophe Biernacki.

### Examples

```
library(ordinalClust)
data("dataqol")
set.seed(5)

# loading the ordinal data
M <- as.matrix(dataqol[,2:29])

m = 4

krow = 4

nbSEM=50
nbSEMBurn=40
nbindmini=2
init = "random"

object <- bosclust(x=M,kr=krow, m=m, nbSEM=nbSEM,
  nbSEMBurn=nbSEMBurn, nbindmini=nbindmini, init=init)
```

---

boscoclust

*Function to perform a co-clustering*

---

### Description

This function runs a co-clustering algorithm on ordinal data by using the latent block model (see references for further details). A BOS distribution is used, and the parameters inference is obtained using the SEM-Gibbs algorithm.

**Usage**

```
boscoclust(x=matrix(0,nrow=1,ncol=1), idx_list=c(1), kr, kc, init, nbSEM, nbSEMBurn,
           nbRepeat=1, nbindmini, m=0, percentRandomB=0)
```

**Arguments**

x	Matrix made of ordinal data of dimension $N \times J_{tot}$ . The features with the same numbers of levels must be placed side by side. The missing values should be coded as NA.
idx_list	Vector of length D. This argument is useful when variables have different numbers of levels. Element d should indicate where the variables with number of levels m[d] begin in matrix x.
kr	Number of row classes.
kc	Vector of length D. The d-th element indicates the number of column clusters.
m	Vector of length D. The d-th element defines the number of levels of the ordinal data.
nbSEM	Number of SEM-Gibbs iterations realized to estimate parameters.
nbSEMBurn	Number of SEM-Gibbs burn-in iterations for estimating parameters. This parameter must be inferior to nbSEM.
nbRepeat	Number of times sampling on rows and columns will be done for each SEM-Gibbs iteration.
nbindmini	Minimum number of cells belonging to a block.
init	String that indicates the kind of initialisation. Must be one of the following words : "kmeans", "random" or "randomBurnin".
percentRandomB	Vector of length 2. Indicates the percentage of resampling when init is equal to "randomBurnin".

**Value**

@V	Matrix of dimension $N \times kr$ such that $V[i,g]=1$ if i belongs to cluster g.
@icl	ICL value for co-clustering.
@name	Name of the result.
@paramschain	List of length nbSEMBurn. The parameters of the blocks are stored for each iteration of the SEM-Gibbs algorithm.
@pichain	List of length nbSEM. Item i is a vector of length kr that contains the row mixing proportions at iteration i.
@rhochain	List of length nbSEM. Item i is a list of length D whose d-th element contains the column mixing proportions of the group of variables d, for iteration i.
@zc	List of length D. The d-th item is a vector of length J[d] representing the column partitions for the group of variables d.
@zr	Vector of length N with resulting row partitions.
@W	List of length D. Item d is a matrix of dimension $J \times kc[d]$ such that $W[j,h]=1$ if j belongs to cluster h.

@m	Vector of length D. The d-th element represents the number of levels of d-th group of variables.
@params	List of length D. The d-th item represents the blocks parameters for a group of variables d.
@pi	Vector of length kr. This corresponds to the row mixing proportions.
@rho	List of length D. The d-th item represents the column mixing proportion for the d-th group of variables.
@xhat	List of length D. The d-th item represents the dataset of the d-th group of variables, with missing values completed.
@zrchain	Matrix of dimension nbSEM*N. Row i represents the row cluster partitions at iteration i.
@zcchain	List of length D. Item d is a matrix of dimension nbSEM*J[d]. Row i represents the column cluster partitions at iteration i.

### Author(s)

Margot Selosse, Julien Jacques, Christophe Biernacki.

### Examples

```
library(ordinalClust)

# loading the real dataset
data("dataqol")
set.seed(5)

# loading the ordinal data
M <- as.matrix(dataqol[,2:29])

# defining different number of categories:
m=4

# defining number of row and column clusters
krow = 4
kcol = 4

# configuration for the inference
nbSEM=50
nbSEMBurn=40
nbindmini=2
init = "randomBurnin"
percentRandomB=c(20,20)

# Co-clustering execution
```

```
object <- boscoclust(x=M,kr=krow,kc=kcol,m=m,nbSEM=nbSEM,  
  nbSEMBurn=nbSEMBurn, nbindmini=nbindmini, init=init, percentRandomB=percentRandomB)
```

---

dataqol

*Questionnaire Responses Of Patients Affected By Breast Cancer*

---

### Description

This dataset contains the responses of 121 patients to 30 questions about their quality of life.

### Usage

```
dataqol
```

### Format

A dataframe with 121 lines and 31 columns. A line represents a patient and a column contains information about the patients.

**Id** patient Id

**q1-q28** responses to 28 questions with the number of levels equal to 4

**q29-q30** responses to 22 questions with the number of levels equal to 7

### Source

The table was determined based on data associated with the package available on: <https://cran.r-project.org/package=QoLR>

---

dataqol.classif

*Questionnaire Responses Of Patients Affected By Breast Cancer*

---

### Description

This dataset contains the responses of 40 patients to 30 questions about their quality of life. Furthermore, a variable indicates if the patient survived the disease.

### Usage

```
dataqol.classif
```

**Format**

A dataframe with 40 lines and 32 columns. A line represents a patient and a column contains information about the patients.

**Id** patient Id

**q1-q28** responses to 28 questions with the number of levels equal to 4

**q29-q30** responses to 22 questions with the number of levels equal to 7

**death** if the patient survived (1) or not (2)

**Source**

The table was determined based on data associated with the package available at: <https://cran.r-project.org/package=QoLR>

---

Msimulated	<i>Matrix of simulated ordinal data</i>
------------	---

---

**Description**

This is a toy dataset for running simple examples.

**Usage**

Msimulated

**Format**

An ordinal data matrix with 60 lines and 50 columns. The number of levels is equal to 3. Four blocks are simulated with  $(\mu, \pi)$  parameters equal to (3,0.5), (2,0.7), (1,0.8) and (2,0.6).

---

pejSim	<i>pejSim</i>
--------	---------------

---

**Description**

This function computes the probability for a level  $e_j$  to be sampled from a BOS distribution of parameters  $(\mu, \pi)$ , with the number of levels equal to  $m$ . It can be used to generate data with a BOS distribution.

**Usage**

pejSim( $e_j, m, \mu, p$ )

**Arguments**

ej	Levels to be sampled
m	Number of levels.
mu	mu parameter for BOS distribution.
pi	pi parameter for BOS distribution.

**Value**

Returns the probability of ej to be sampled from a BOS distribution of parameters (mu,pi), with the number of levels equal to m.

**Author(s)**

Margot Seloisse, Julien Jacques, Christophe Biernacki.

**Examples**

```
library(ordinalClust)
data("dataqol")
set.seed(5)

m=7
nr=10000
mu=5
pi=0.5

probaBOS=rep(0,m)
for (im in 1:m) probaBOS[im]=pejSim(im,m,mu,pi)
M <- sample(1:m,nr,prob = probaBOS, replace=TRUE)
```

---

plot

*~~ Method for the function plot in the **ordinalClust** package ~~*

---

**Description**

Plots the result of a classification, clustering or co-clustering performed using the following functions: bosclassif,bosclust,boscoclust.

**Methods**

```
signature(object = "ResultClassifOrdinal")
signature(object = "ResultClustOrdinal")
signature(object = "ResultCoclustOrdinal")
```

---

predict                      *~~ Method for the function predict in the **ordinalClust** package ~~*

---

### **Description**

Method for the function predict in the **ordinalClust** package.

### **Methods**

signature(object = "ResultClassifOrdinal") Use this method with the result of the function bosclassif, and a new sample to predict the classes.

---

summary                      *~~ Methods for the Function summary in the **ordinalClust** package*  
*~~*

---

### **Description**

Prints the result of a classification, clustering or co-clustering performed using the following functions: bosclassif,bosclust,boscoclust.

### **Methods**

signature(object = "ResultClassifOrdinal")  
signature(object = "ResultClustOrdinal")  
signature(object = "ResultCoclustOrdinal")

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