

# Package ‘uHMM’

April 22, 2016

**Type** Package

**Title** Construct an Unsupervised Hidden Markov Model

**Version** 1.0

**Date** 2016-03-18

**Author** Emilie POISSON-CAILLAULT [aut], Paul TERNYNCK [aut, cre]

**Maintainer** Paul TERNYNCK <terynck@lisic.univ-littoral.fr>

**Description** Construct a Hidden Markov Model with states learnt by unsupervised classification.

**License** GPL (>= 2)

**RoxygenNote** 5.0.1

**LazyData** TRUE

**Depends** R (>= 3.0.0), stats, grDevices

**Imports** tcltk, tcltk2, tkplot, HMM, clValid, class, cluster,  
FactoMineR, corrplot, chron

**NeedsCompilation** no

**Repository** CRAN

**Date/Publication** 2016-04-22 16:25:47

## R topics documented:

uHMM-package . . . . .	2
computeGap . . . . .	2
cutCalculation . . . . .	3
emissionMatrix . . . . .	4
FastSpectralNJW . . . . .	5
HMMparams . . . . .	6
KmeansAutoElbow . . . . .	7
KpartitionNJW . . . . .	8
MarelCarnot . . . . .	9
selfKNN . . . . .	10
spectralPamClusteringNg . . . . .	11
transitionMatrix . . . . .	12
uHMMinterface . . . . .	13
ZPGaussianSimilarity . . . . .	13

**Index****15**

---

uHMM-package*Construct an unsupervised Hidden Markov Model*

---

**Description**

This package proposes an interface to detect usual or extreme events in a dataset and to characterize their dynamic, by building an unsupervised Hidden Markov Model (use [uHMInterface](#) to launch the interface). Functions can also be used out of the interface to build an uHMM.

**Details**

Package:	uHMM
Version:	1.0
Date:	2016-04-13
Depends:	R (>= 3.0.0), stats, grDevices
Import:	tcltk, tcltk2, tkplot, HMM, clValid, class, cluster, FactoMineR, corrplot, chron
License:	GPL (>=2)
LazyLoad:	yes

**Author(s)**

Emilie Poisson-Caillault and Paul Ternynck

Maintainer: <[emilie.caillault@lisic.univ-littoral.fr](mailto:emilie.caillault@lisic.univ-littoral.fr)>

**Source**

Rousseeuw, Kevin, et al. "Hybrid hidden Markov model for marine environment monitoring." Selected Topics in Applied Earth Observations and Remote Sensing, IEEE Journal of 8.1 (2015): 204-213.

---

computeGap*Compute gap between eigenvalues of a similarity matrix*

---

**Description**

Find the highest gap between eigenvalues of a similarity matrix. The 2 first eigenvalues are considered as equal to each other (the gap between the 2 first eigenvalues is set to 0).

**Usage**

```
computeGap(similarity, Gmax)
```

### Arguments

- similarity a similarity matrix.  
 Gmax the maximum gap value allowed (only the first Gmax eigenvalues will be taken into account).

### Value

The function returns a list containing the following components:

- gap a vector indicating the gap between similarity matrix eigenvalues (the gap between the 2 first eigenvalues is set to 0)  
 Kmax an integer indicating the index of the highest gap (the highest gap is between the Kmax-th and the (Kmax+1)-th eigenvalues)

### Examples

```
x <- rbind(matrix(rnorm(50, mean = 0, sd = 0.3), ncol = 2),
            matrix(rnorm(50, mean = 2, sd = 0.3), ncol = 2),
            matrix(rnorm(50, mean = 4, sd = 0.3), ncol = 2))

similarity<-ZPGaussianSimilarity(x,7)
Gap<-computeGap(similarity,10)
plot(1:length(Gap$gap),Gap$gap,type="h",
main=paste("Gap criteria =",Gap$K),ylab="gap value",xlab="eigenvalues")

x=(runif(1000)*4)-2;y=(runif(1000)*4)-2
keep<-which((x*x2+y*y2<0.5)|(x*x2+y*y2>1.5**2 & x*x2+y*y2<2**2 ))
data<-data.frame(x,y)[keep,]
plot(data)

similarity<-ZPGaussianSimilarity(data,1)
Gap<-computeGap(similarity,10)
plot(1:length(Gap$gap),Gap$gap,type="h",
main=paste("Gap criteria =",Gap$K),ylab="gap value",xlab="eigenvalues")
```

### Description

Compute intra and inter-cluster cuts from the similarity matrix of a dataset.

### Usage

```
cutCalculation(similarity, label, K)
```

**Arguments**

- `similarity` a similarity matrix.  
`label` vector of cluster sequencing.  
`K` number of clusters. (= nbCluster CALCULE DANS LA FONCTION ???)

**Details**

intra cluster cut :

$$Cut(g_k, g_l) = \sum_{i=1, x(i) \in g_k}^{N_p} \sum_{j=1, x(j) \in g_l}^{N_p} w(x(i), x(j))$$

**Value**

The function returns a list containing:

- `mncut` the inter-cluster cut, i.e. K-sum(`ratioCutVol`).  
`ratioCutVol` vector of intra-cluster cuts, one component per cluster.

**Examples**

```
x<-rbind(matrix(runif(100),ncol=2),matrix(runif(100)+2,ncol=2),matrix(runif(20)*3,ncol=2))
similarity<-ZPGaussianSimilarity(x,7)%%t(ZPGaussianSimilarity(x,7))
km<-kmeans(similarity,2)
label<-km$cluster
plot(x,col=km$cluster)
cutCalculation(similarity,label,length(unique(label)))
```

<code>emissionMatrix</code>	<i>Emission matrix estimation</i>
-----------------------------	-----------------------------------

**Description**

This function estimates the emission matrix of a Hidden Markov Model from vectors of state and symbol sequencing.

**Usage**

```
emissionMatrix(states, symbols)
```

**Arguments**

- `states` a numeric vector of state sequencing.  
`symbols` a numeric vector of symbol sequencing.

**Value**

Estimated emission matrix.

**See Also**[HMMparams](#)**Examples**

```
states<-c(1,1,3,2,1,2,1,3)
symbols<-c(4,1,3,1,4,4,4,2)
B<-emissionMatrix(states,symbols)
B
```

FastSpectralNJW

*Jordan Fast Spectral Algorithm***Description**

Perform the Jordan spectral algorithm for large databases. Data are sampled, using K-means with Elbow criteria, before being classified.

**Usage**

```
FastSpectralNJW(data, nK = NULL, Kech = 2000, StopCriteriaElbow = 0.97,
neighbours = 7, method = "", nb.iter = 10, uHMMinterface = FALSE,
console = NULL, tm = NULL)
```

**Arguments**

<code>data</code>	numeric matrix or dataframe.
<code>nK</code>	number of clusters desired. If <code>NULL</code> , optimal number of clusters will be computed using gap criteria.
<code>Kech</code>	maximum number of representative points in sampled data.
<code>StopCriteriaElbow</code>	maximum (minimum ?) de variance expliquees des points representatifs souhaite.
<code>neighbours</code>	number of neighbours considered for the computation of local scale parameters.
<code>method</code>	string specifying the spectral classification method desired, either "PAM" (for spectral kmedoids) or "" (for "spectral kmeans").
<code>nb.iter</code>	number of iterations.
<code>uHMMinterface</code>	logical indicating whether the function is used via the <code>uHMMinterface</code> .
<code>console</code>	frame of the <code>uHMM</code> interface in which messages should be displayed (only if <code>uHMMinterface=TRUE</code> ).
<code>tm</code>	a one row dataframe containing text to display in the <code>uHMMinterface</code> (only if <code>uHMMinterface=TRUE</code> ).

**Details**

Algorithme de Jordan pour un grand jeu de donnees : echantillonage puis spectral

**Value**

The function returns a list containing:

<code>sim</code>	similarity matrix of representative points, multiplied by its transpose ( <a href="#">ZPGaussianSimilarity</a> ).
<code>label</code>	vector of cluster sequencing.
<code>gap</code>	number of clusters.
<code>labelElbow</code>	vector of prototype sequencing.
<code>vpK</code>	matrix containing, in columns, the K first normalised eigen vectors of the data similarity matrix.
<code>valp</code>	vector containing the K first eigen values of the data similarity matrix.
<code>echantillons</code>	matrix of prototypes coordinates.
<code>label.echantillons</code>	vector containing the cluster of each prototype.
<code>numSymbole</code>	vector containing the nearest prototype of each data item.

**See Also**

[KmeansAutoElbow](#) [ZPGaussianSimilarity](#) [knn](#) [silhouette](#) [dunn](#) [connectivity](#) [dist](#)

**Examples**

```
x=(runif(1000)*4)-2;y=(runif(1000)*4)-2
keep<-which((x**2+y**2<0.5)|(x**2+y**2>1.5**2 & x**2+y**2<2**2 ))
data<-data.frame(x,y)[keep,]

cl<-FastSpectralNJW(data,2)
plot(data,col=cl$label)
```

**Description**

This function is used by the [uHMMinterface](#) to estimate parameters of a Hidden Markov Model.

**Usage**

```
HMMparams(stateSeq, symbolSeq)
```

**Arguments**

<code>stateSeq</code>	a numeric vector of state sequencing.
<code>symbolSeq</code>	a numeric vector of symbol sequencing.

**Value**

HMMparams returns a list containing :

trans	The transition matrix.
emis	The emission matrix.
startProb	The vector of initial probability distribution (initial states are supposed equiprobable).

**See Also**

[transitionMatrix](#) [emissionMatrix](#)

KmeansAutoElbow

*KmeansAutoElbow function*

**Description**

KmeansAutoElbow performs k-means clustering on a dataframe with selection of optimal number of clusters using elbow criteria.

**Usage**

```
KmeansAutoElbow(features, Kmax, StopCriteria = 0.99, graph = FALSE)
```

**Arguments**

features	dataframe or matrix of raw data.
Kmax	maximum number of clusters allowed.
StopCriteria	elbow method cumulative explained variance > criteria to stop K-search. (???)
graph	boolean, if TRUE figures are plotted.

**Details**

KmeansAutoElbow returns partition and K number of groups according to kmeans clustering and Elbow method

**Value**

The function returns a list containing the following components:

K	number of clusters in data according to explained variance and kmeans algorithm.
res.kmeans	an object of class "kmeans" (see <a href="#">kmeans</a> ) containing classification results.

**See Also**

[kmeans](#)

## Examples

```
x <- rbind(matrix(rnorm(300, mean = 0, sd = 0.3), ncol = 2),
            matrix(rnorm(100, mean = 2, sd = 0.3), ncol = 2),
            matrix(rnorm(100, mean = 4, sd = 0.3), ncol = 2))
colnames(x) <- c("x", "y")
km<-KmeansAutoElbow(x, round(dim(x)/25,0)[1],StopCriteria=0.99,graph=TRUE)
plot(x,col=km$res.kmeans$cluster)
points(km$res.kmeans$centers, col = 1:km$K, pch = 16)
```

**KpartitionNJW**

*KpartitionNJW function*

## Description

Perform spectral classification on the similarity matrix of a dataset (Ng et al. (2001) algorithm), using kmeans algorithm on data projected in the space of its K first eigen vectors.

## Usage

```
KpartitionNJW(similarity, K)
```

## Arguments

similarity	matrix of similarity.
K	number of clusters.

## Value

The function returns a list containing:

label	vector of cluster sequencing.
centres	matrix of cluster centers in the space of the K first normalised eigen vectors.
vecteursPropresProjK	matrix containing, in columns, the K first normalised eigen vectors of the similarity matrix.
valeursPropresK	vector containing the K first eigen values of the similarity matrix.
vecteursPropres	matrix containing, in columns, eigen vectors of the similarity matrix.
valeursPropres	vector containing eigen values of the similarity matrix.
inertieZ	vector of within-cluster sum of squares, one component per cluster.

## References

Ng Andrew, Y., M. I. Jordan, and Y. Weiss. "On spectral clustering: analysis and an algorithm [C]." Advances in Neural Information Processing Systems (2001).

## Examples

```
#####
x <- rbind(matrix(rnorm(100, mean = 0, sd = 0.3), ncol = 2),
            matrix(rnorm(100, mean = 2, sd = 0.3), ncol = 2),
            matrix(rnorm(100, mean = 4, sd = 0.3), ncol = 2))

similarity<-ZPGaussianSimilarity(x,7)
similarity=similarity%*%t(similarity)
sp<-KpartitionNJW(similarity,3)
plot(x,col=sp$label)

#####
x <- rbind(data.frame(x=1:100+(runif(100)-0.5)*2,y=runif(100)/5),
            data.frame(x=1:100+(runif(100)-0.5)*2,y=runif(100)/5+1),
            data.frame(x=1:100+(runif(100)-0.5)*2,y=runif(100)/5+2))

similarity<-ZPGaussianSimilarity(x,7)
similarity=similarity%*%t(similarity)
sp<-KpartitionNJW(similarity,3)
plot(x,col=sp$label)

#####
x=(runif(1000)*4)-2;y=(runif(1000)*4)-2
keep<-which((x**2+y**2<0.5)|(x**2+y**2>1.5**2 & x**2+y**2<2**2 ))
data<-data.frame(x,y)[keep,]

similarity=ZPGaussianSimilarity(data, 7)
similarity=similarity%*%t(similarity)
sp<-KpartitionNJW(similarity,2)

plot(data,col=sp$label)
```

## Description

The MarelCarnot data set gives the measurements of 14 physico-chemical and biological parameters performed by the Marel-Carnot station (Boulogne-sur-Mer, France), at high frequency resolution.

## Usage

`MarelCarnot`

## Format

A data frame with 131487 rows and 16 columns.

## Details

Dates	date of measurement	(YYYY:MM:DD)
Hours	time of measurement	(HH:MM:SS)
C_NI1	nitrate concentration	(in $\mu\text{mol/L}$ )
C_PO1	phosphate concentration	(in $\mu\text{mol/L}$ )
C_O21	corrected dissolved oxygen	(in mg/L)
C_SI1	silicate concentration	(in $\mu\text{mol/L}$ )
CSAL1	salinity	(in PSU)
CSAT1	oxygen saturation	(in %)
ETCO1	air temperature	(in degrees Celsius)
E_LU1	P.A.R	(in $\mu\text{mol of photons/s/m}^2$ )
E_O21	uncorrected dissolved oxygen	(in mg/L)
E_PH1	pH	
E_TU1	turbidity	(in NTU)
ECHL1	fluorescence	(in FFU)
E_TA	water temperature	(in degrees Celsius)
XMAHH	water level	(in m)

## Source

Lefebvre Alain (2015). MAREL Carnot data and metadata from Coriolis Data Centre. SEANOE.  
<http://doi.org/10.17882/39754>

*selfKNN*

*Self KNN*

## Description

This function performs the k-Nearest Neighbour algorithm without class estimation, but only computation of distances and neighbours.

## Usage

```
selfKNN(train, K = 1)
```

## Arguments

- |       |                                  |
|-------|----------------------------------|
| train | numeric matrix or data frame.    |
| K     | number of neighbours considered. |

**Value**

The function returns a list with the following components:

- |     |  |
|-----|--|
| D   | matrix of squared root of the distances between observations and their nearest neighbours. |
| idx | Index of K nearest neighbours of each observation.   |

**Examples**

```
x<-matrix(runif(10),ncol=2)
plot(x,pch=c("1","2","3","4","5"))
selfKNN(x,K=4)
```

**spectralPamClusteringNg**

*spectralPamClusteringNg function*

**Description**

Perform spectral classification on the similarity matrix of a dataset, using pam algorithm (a more robust version of K-means) on projected data.

**Usage**

```
spectralPamClusteringNg(similarity, K)
```

**Arguments**

- |            |                      |
|------------|----------------------|
| similarity | matrix of similarity |
| K          | number of clusters   |

**Value**

The function returns a list containing:

- |                      |   |
|----------------------|---|
| label                | vector of cluster sequencing.   |
| centres              | matrix of cluster medoids (similar in concept to means, but medoids are members of the dataset) in the space of the K first normalised eigen vectors. |
| id.med               | integer vector of indices giving the medoid observation numbers.  |
| vecteursPropresProjK | matrix containing, in columns, the K first normalised eigen vectors of the similarity matrix.   |
| valeursPropresK      | vector containing the K first eigen values of the similarity matrix.  |
| vecteursPropres      | matrix containing, in columns, eigen vectors of the similarity matrix.  |

`valeursPropres` vector containing eigen values of the similarity matrix.  
`cluster.info` matrix, each row gives numerical information for one cluster. These are the cardinality of the cluster (number of observations), the maximal and average dissimilarity between the observations in the cluster and the cluster's medoid, the diameter of the cluster (maximal dissimilarity between two observations of the cluster), and the separation of the cluster (minimal dissimilarity between an observation of the cluster and an observation of another cluster).

## References

Ng Andrew, Y., M. I. Jordan, and Y. Weiss. "On spectral clustering: analysis and an algorithm [C]." Advances in Neural Information Processing Systems (2001).

## See Also

[pam](#)

`transitionMatrix`      *Transition matrix estimation*

## Description

This function estimates the transition matrix of a (Hidden) Markov Model from a vector of state sequencing.

## Usage

`transitionMatrix(states)`

## Arguments

`states`      a numeric vector of state sequencing.

## Value

Estimated transition matrix.

## See Also

[HMMparams](#)

## Examples

```
states<-c(1,1,3,2,1,2,1,3)
A<-transitionMatrix(states)
A
```

---

**uHMMinterface***Graphical Interface to Build an uHMM*

---

**Description**

A user-friendly interface to detect usual or extreme events in a dataset and to characterize their dynamic, by building an unsupervised Hidden Markov Model.

**Usage**

```
uHMMinterface(uHMMenv = NULL)
```

**Arguments**

**uHMMenv** an environment in which data and results will be stored. If NULL, a local environment will be created.

**Value**

Results are saved in the directory chosen by the user.

**References**

Rousseeuw, Kevin, et al. "Hybrid hidden Markov model for marine environment monitoring." Selected Topics in Applied Earth Observations and Remote Sensing, IEEE Journal of 8.1 (2015): 204-213.

---

**ZPGaussianSimilarity** *Similarity matrix with local scale parameter*

---

**Description**

Compute and return the similarity matrix of a data frame using gaussian kernel with a local scale parameter for each data point, rather than a unique scale parameter.

**Usage**

```
ZPGaussianSimilarity(data, K)
```

**Arguments**

**data** a matrix or numeric data frame.  
**K** number of neighbours considered to compute scale parameters.

**Value**

The matrix of similarity.

**References**

Zelnik-Manor, Lihi, and Pietro Perona. "Self-tuning spectral clustering." Advances in neural information processing systems. 2004.

**Examples**

```
x <- rbind(matrix(rnorm(50, mean = 0, sd = 0.3), ncol = 2))
similarity<-ZPGaussianSimilarity(x,7)
```

# Index

computeGap, 2  
connectivity, 6  
cutCalculation, 3  
  
dist, 6  
dunn, 6  
  
emissionMatrix, 4, 7  
  
FastSpectralNJW, 5  
  
HMMparams, 5, 6, 12  
  
kmeans, 7  
KmeansAutoElbow, 6, 7  
knn, 6  
KpartitionNJW, 8  
  
MarelCarnot, 9  
  
pam, 12  
  
selfKNN, 10  
silhouette, 6  
spectralPamClusteringNg, 11  
  
transitionMatrix, 7, 12  
  
uHMM-package, 2  
uHMMinerface, 2, 6, 13  
  
ZPGaussianSimilarity, 6, 13