

# Package: MajMinKmeans (via r-universe)

September 15, 2024

**Type** Package

**Title** k-Means Algorithm with a Majorization-Minimization Method

**Version** 0.1.0

**Maintainer** Sheikhi Ayyub <sheikhy.a@uk.ac.ir>

**Description** A hybrid of the K-means algorithm and a Majorization-Minimization method to introduce a robust clustering. The reference paper is: Julien Mairal, (2015) <doi:10.1137/140957639>. The two most important functions in package 'MajMinKmeans' are cluster\_km() and cluster\_MajKm(). Cluster\_km() clusters data without Majorization-Minimization and cluster\_MajKm() clusters data with Majorization-Minimization method. Both of these functions calculate the sum of squares (SS) of clustering. Another useful function is MajMinOptim(), which helps to find the optimum values of the Majorization-Minimization estimator.

**Imports** MASS

**License** GPL-3

**Encoding** UTF-8

**RoxygenNote** 7.3.1

**NeedsCompilation** no

**Author** Sheikhi Ayyub [aut, cre], Yaghoobi Mohammad Ali [aut]

**Date/Publication** 2024-05-17 09:20:09 UTC

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

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

**RemoteRef** HEAD

**RemoteSha** e33c3afc78b5a4e62879872b3f0dd8440bba0533

## Contents

clusters_km . . . . .	2
clusters_MajKm . . . . .	3

Euclid . . . . .	3
kmeans . . . . .	4
MajMinOptim . . . . .	5
<b>Index</b>	<b>6</b>

---

clusters_km	<i>clustering results of the k-mean algorithm</i>
-------------	---------------------------------------------------

---

### Description

clusters data into two clusters. This function uses the kmeans function to cluster the data and exports the clustering results as well as the sum of square (SS) of clustering using the Euclidian distance.

### Usage

```
clusters_km(x, k = 2)
```

### Arguments

x	matrix of data (dim 1: samples (must be equal to dim 1 of X), dim 2: attributes (must be equal to dim 2 of X))
k	number of clusters ( this version considers 2 clusters )

### Value

sum of square (SS) of clustering

### Examples

```
{
X=rbind(matrix(rnorm(1000*2 ,4,.1),1000,2),matrix(rnorm(1000*2, 3, 0.2),1000,2))
M<- X[sample(nrow(X), 2),]
clusters_km(X,2)
}
```

---

clusters_MajKm	<i>clustering results of the majorized k-mean algorithm</i>
----------------	-------------------------------------------------------------

---

**Description**

clusters data into two clusters with a majorization k-means. This function uses a hybrid of the k-means and the majorization-minimization method to cluster the data and exports the clustering results as well as the sum of square (SS) of clustering.

**Usage**

```
clusters_MajKm(X, k = 2, La)
```

**Arguments**

X	matrix of data (dim 1: samples (must be equal to dim 1 of X), dim 2: attributes (must be equal to dim 2 of X))
k	number of clusters ( this version considers 2 clusters )
La	the tuning parameter

**Value**

sum of square (SS) of clustering and the 'delta' (difference of two successive majorization function).

**Examples**

```
{
X=rbind(matrix(rnorm(1000*2),4,.1),1000,2),matrix(rnorm(1000*2),3,0.2),1000,2))
M <- X[sample(nrow(X), 2),]
clusters_MajKm(X,2, 0.5)
}
```

---

Euclid	<i>Euclidian distance</i>
--------	---------------------------

---

**Description**

Calculates the Euclidian distance between points. This function can be used in the kmeans function to do the clustering procedure using the Euclidian distance.

**Usage**

```
Euclid(x, mu)
```

**Arguments**

`x` matrix of data (dim 1: samples (must be equal to dim 1 of X), dim 2: attributes (must be equal to dim 2 of X))

`mu` initial seleted centroids (randomly or another method).

**Value**

Euclidian distance between two points.

**Examples**

```
{
X=rbind(matrix(rnorm(1000*2 ,4,.1),1000,2),matrix(rnorm(1000*2, 3, 0.2),1000,2))
M <- X[sample(nrow(X), 2),]
Euclid(X,M)
}
```

---

kmeans

*k-means function*


---

**Description**

k-means algorithm in clustering. This function export the clustered results based on one replication of the k-means method

**Usage**

```
kmeans(x, centers, nItter = 4)
```

**Arguments**

`x` matrix of data (dim 1: samples (must be equal to dim 1 of X), dim 2: attributes (must be equal to dim 2 of X))

`centers` initial seleted centroids (randomly or another method)

`nItter` Number of itteration function

**Value**

clustered results based on k-means methods.

**Examples**

```
{
X=rbind(matrix(rnorm(1000*2 ,4,.1),1000,2),matrix(rnorm(1000*2, 3, 0.2),1000,2))
M <- X[sample(nrow(X), 2),]
kmeans(X,M, 4)
}
```

---

MajMinOptim	<i>majorization-minimization optimization</i>
-------------	-----------------------------------------------

---

**Description**

Finding the optimized majorization-minimization centers

**Usage**

```
MajMinOptim(X, Z, M, eps, lambda)
```

**Arguments**

X	matrix of data (dim 1: samples (must be equal to dim 1 of X), dim 2: attributes (must be equal to dim 2 of X))
Z	is a n by k matrix where for all i and j, $z_{i,j}$ is abinary variable that is equal to 1 if the case i is assigned to cluster j and zero otherwise. (dim 1: samples (must be equal to dim 1 of X), dim 2: attributes (must be equal to dim 2 of X))
M	initial seleted centroids (randomly or another method)
eps	a threshold value assumed as 0.0001
lambda	a threshold value assumed as 0.5

**Value**

The optimized majorization-minimization centers.

**Examples**

```
{
X=rbind(matrix(rnorm(1000*2),4,.1),1000,2),matrix(rnorm(1000*2),3,0.2),1000,2))
M <- X[sample(nrow(X), 2),]
distsToCenters <- Euclid(X, M)
clusters <- apply(distsToCenters, 1, which.min)
Z <- matrix(0, nrow = NROW(X), ncol = 1)
for(i in 1:NROW(X))
  if (clusters[[i]] == 1)
    Z[i,]=clusters[[i]]
Z=cbind(Z, 1-Z)
MajMinOptim(X,Z,M ,eps=1e-4, lambda=.5)
}
```

# Index

clusters\_km, [2](#)  
clusters\_MajKm, [3](#)  
Euclid, [3](#)  
kmeans, [4](#)  
MajMinOptim, [5](#)