# Package 'LARkmeans'

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Title K-means Clustering				
Version 1.0				
Oate 2016-11-09 Author Lasse Rintakumpu Maintainer Lasse Rintakumpu <rintakumpu@gmail.com> Oescription Provides a custom class for k-means clustering.</rintakumpu@gmail.com>				
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				VignetteBuilder knitr
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LARkmeans-package				
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LARkmeans-package K-means Clustering				
Description				

## Details

Type Package

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Provides a custom class for k-means clustering.

Index: This package was not yet installed at build time.

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## Author(s)

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## References

Forgy, E. W. (1965) Cluster analysis of multivariate data: efficiency vs interpretability of classifications. Biometrics **21**, 768–769.

Hamerly, G.; Elkan, C. (2002) Alternatives to the k-means algorithm that find better clusterings (PDF). Proceedings of the eleventh international conference on Information and knowledge management (CIKM).

Lloyd, S. P. (1957, 1982) Least squares quantization in PCM. Technical Note, Bell Laboratories. Published in 1982 in IEEE Transactions on Information Theory **28**, 128–137.

distEuclidean

Euclidean Distance

## **Description**

Calculates the Euclidean distance between two matrices with k columns.

## Usage

```
distEuclidean(x, y)
```

## Arguments

```
x a matrix of size n x k.
y a matrix of size m x k.
```

## Value

a distance matrix of size n x m.

## See Also

dist

## **Examples**

```
## The function is currently defined as
function (x, y)
{
    z <- matrix(0, nrow = nrow(x), ncol = nrow(y))
    for (k in 1:nrow(y)) {
        z[, k] <- sqrt(colSums((t(x) - y[k, ])^2))
    }
    return(z)
}</pre>
```

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## **Description**

Performs k-means clustering m times on a data matrix. fitted returns a vector with the class labels of the best run.

## Usage

```
kMeans(X, k, m=10, ind, max.iter=50, ...)
## S3 method for class 'kMeans'
fitted(object)
```

## **Arguments**

Χ	a numeric matrix of data.
k	the desired number of clusters.
m	the number of times to run the clustering algorithm. The default is 10.
ind	a numeric vector of columns indicating the variables used in the clustering.
max.iter	the maximum number of iterations for a single run of the clustering algorithm. The default is 50.
	not used.

## **Details**

The matrix data given by X is clustered by the standard k-means method, also known as Lloyd-Forgy method (1957 & 1965). This method aims at minimizing the within-cluster sum of squares objective and thus assigns the clusters by the smallest Euclidean distance of observation to the cluster center.

The Random Partition method as described by Hamerly and Elkan (2002) is used for computing the initial cluster means.

## Value

kMeans returns an object of class kMeans which has a print, summary, predict, plot and a fitted method. It is a list with the following components:

Cbest the vector of the best group labels.

ObjBest the value of the objective function for the best solution.

CentroidsBest the matrix containing the centroids of the best solution.

m the number of repetitions.k the number of groups.

Xname name of the data set used for the clustering.

Ind the value of input ind.

Y the data used for the clustering.

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Best value of which of the runs was the best.

Call a matrix with n rows and m columns giving the cluster assignments for all m runs.

ObjAll a vector having the objective functions of all runs.

StatusAll a vector having the status from all runs.

#### References

Forgy, E. W. (1965) Cluster analysis of multivariate data: efficiency vs interpretability of classifications. Biometrics **21**, 768–769.

Hamerly, G.; Elkan, C. (2002) Alternatives to the k-means algorithm that find better clusterings (PDF). Proceedings of the eleventh international conference on Information and knowledge management (CIKM).

Lloyd, S. P. (1957, 1982) Least squares quantization in PCM. Technical Note, Bell Laboratories. Published in 1982 in IEEE Transactions on Information Theory **28**, 128–137.

#### See Also

kmeans

## **Examples**

```
# Example using random data from three different populations
# with one variable
set.seed(63555)
exampleData <- matrix(nrow=90, ncol=1)</pre>
exampleData[1:30, 1] <- rnorm(30, mean=3, sd=1)
exampleData[31:60, 1] <- rnorm(30, mean=6, sd=1)
exampleData[61:90, 1] <- rnorm(30, mean=9, sd=1)
kMeansResult <- kMeans(exampleData, k=3)</pre>
kMeansResult
# K-Means clustering for iris
# Number of runs: 10
# Status of best run: converged
fitted(kMeansResult)
summary(kMeansResult)
#K-Means clustering for exampleData
#Clusters to be detected: 3
#Cluster sizes detected: 30 28 32
#Number of runs: 10
#Status of best run: converged
#Criterion value: 3522.284
#Summary of criterion values:
#Min: 3522.284
#Q1: 3546.504
```

plot.kMeans 5

#Mean: 3543.455 #Q3: 3546.504 #Max: 3546.504

plot.kMeans

Plot Method for K-Means Clustering

## **Description**

Plot method for objects of class "kMeans". Calls plot function to produce a scatter plot of the data values and cluster assignments provided by the given kMeans object.

## Usage

```
plot.kMeans(X, col, pch, ...)
```

## **Arguments**

X an object of the class kMeans.
 col a specification for the plotting color.
 pch either an integer specifying a symbol or a single character to be used as the default in plotting points. See points for possible values and their interpretation.
 ... further arguments to be passed to or from methods.

## See Also

par

## **Examples**

```
set.seed(63555)
exampleData <- matrix(nrow=90, ncol=1)
exampleData[1:30, 1] <- rnorm(30, mean=3, sd=1)
exampleData[31:60, 1] <- rnorm(30, mean=6, sd=1)
exampleData[61:90, 1] <- rnorm(30, mean=9, sd=1)
kMeansResult <- kMeans(exampleData, k=3)
plot(kMeansResult)</pre>
```

predict.kMeans

Predict Method for K-Means Clustering

## Description

This function assigns observations in the data matrix newData the most likeliest clusters using the best solution from a kMeans object.

## Usage

```
predict.kMeans(X, newData)
```

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## **Arguments**

X object of class kMeans.

newData a data matrix or data frame having the same columns as the original X when

kMeans was called.

## Value

Returns a vector of cluster assignments for newData based on the kMeans object.

## See Also

**k**Means

## **Examples**

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