

# Package: kkmeans (via r-universe)

May 21, 2026

**Title** Fast Implementations of Kernel K-Means

**Version** 0.1.3

**Description** Implementations several algorithms for kernel k-means. The default 'OTQT' algorithm is a fast alternative to standard implementations of kernel k-means, particularly in cases with many clusters. For a small number of clusters, the implemented 'MacQueen' method typically performs the fastest. For more details and performance evaluations, see Berlinski and Maitra (2025) <[doi:10.1002/sam.70032](https://doi.org/10.1002/sam.70032)>.

**License** GPL-3

**Encoding** UTF-8

**LazyData** true

**Roxygen** list(markdown = TRUE)

**RoxygenNote** 7.3.2

**LinkingTo** Rcpp

**Imports** Rcpp

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

**Date/Publication** 2026-05-20 23:22:13 UTC

**RemoteUrl** <https://github.com/jdberlinski/kkmeans>

**RemoteRef** HEAD

**RemoteSha** 81d7508d3b9c8718f74ed3ee66606780da2cf8d1

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cluster_new	<i>Classify new data based on kkmeans result</i>
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**Description**

Given training data, test data, and kkmeans-results, get which partitions a new set of data belong to

**Usage**

```
cluster_new(test_data, train_data, train_result)
```

**Arguments**

test_data	the new data to be classified
train_data	the data to make classifications from
train_result	kkmeans result containing the result from classifying train_data

**Value**

A vector of class labels for test\_data corresponding to the clusters present in train\_result.

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get_kernel_matrix	<i>Get the kernel matrix for a dataset</i>
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**Description**

Given a dataset, kernel function, and tuning parameter, will return the n x n kernel matrix

**Usage**

```
get_kernel_matrix(data, kern = "g", param1 = 1, param2 = 1)
```

**Arguments**

data	data vector
kern	the kernel to use, one of ('gaussian', 'poly', 'sigmoid', 'laplacian'), can use first letter
param1	first parameter to pass to kernel function.
param2	second parameter to pass to kernel function.

**Value**

An n x n matrix for data given by the specified kernel. The value in position (i, j) corresponds to the kernel function evaluated at data[i, ] and data[j, ].

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get_mknn_dist	<i>Get the average distance to each points k-nearest neighbor</i>
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**Description**

Given a dataset and a value k, will return the value of the average distance from each point to it's k-nearest neighbor

**Usage**

```
get_mknn_dist(data, k = FALSE)
```

**Arguments**

data	the data vector
k	which neighbor to average over

**Value**

The average distance from each point to it's k-th nearest neighbor.

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jump_stat	<i>Function to get jump statistic for varying values of k</i>
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**Description**

Obtains the jump statistic for a particular kernel for the specified number of clusters

**Usage**

```
jump_stat(data, kern = "g", param = 1, k_max, eta, iter_max = 1000L)
```

**Arguments**

data	Numeric data to cluster. This will be converted to a matrix using <code>as.matrix</code> .
kern	The kernel to use.
param	The parameter value to pass to the kernel
k_max	The maximum number of clusters to consider
eta	Power for the jump statistic
iter_max	Maximum number of iterations to use in <code>kkmeans</code> call

**Value**

Sum of squares and value of jump statistic for 1, ..., K chosen clusters

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 kkmeans

*An Efficient Kernel K-Means Algorithm*


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### Description

Performs kernel k-means with the specified kernel using an optimal-transfer quick-transfer algorithm.

### Usage

```
kkmeans(
  data,
  k,
  kern = "g",
  param = 1,
  param2 = 1,
  nstart = 10,
  iter_max = 1000L,
  estimate = FALSE,
  nn = 0,
  init_centers = sample(1:k, size = nrow(data), replace = TRUE),
  method = c("otqt", "macqueen", "lloyd", "ot"),
  trueest = FALSE,
  kmat = NULL,
  random_centers = TRUE
)
```

### Arguments

data	Numeric data to cluster. This will be converted to a matrix using <code>as.matrix</code> .
k	Number of clusters.
kern	Kernel to use, one of ('gaussian', 'poly', 'sigmoid', 'laplacian').
param	value of parameter to pass to kernel function.(eg sigma in gaussian kernel). The Gaussian kernel is $K(x, y) = \exp(- \ x - y\ ^2 / (2*param))$ , and the polynomial kernel is $K(x, y) = (x'y + a) ^ param$
param2	value of second parameter parameter to pass to the kernel function, which correspond to the offset for the sigmoid and polynomial kernels.
nstart	Number of times to run the algorithm. The run with the lowest total within cluster SSE (in feature space) will be returned
iter_max	The maximum number of iterations to allow.
estimate	If using the Gaussian kernel, specifying <code>estimate = "mknn"</code> will use an nn-nearest neighbor method for estimating param.
nn	How many neighbors to consider for mknn estimation.
init_centers	The initial values for cluster membership. If <code>nstart</code> is greater than 1, any start beyond the first iteration will use randomized centers.

method	Which method to use for kernel k-means iteration. One of ("otqt", "macqueen", "lloyd"). "otqt" is a method using optimal-transfer and quick-transfer heuristics similar to the Hartigan and Wong algorithm for k-means clustering.
trueest	Whether or not the within-cluster sum of squares should be recomputed in R after clustering is finished
kmat	kernel matrix, if using a custom kernel
random_centers	if TRUE, then assign k observations as initial clusters, assigning the remaining observations to the closest cluster. Otherwise, assign all observations to clusters at random.

### Value

A list containing the following useful information

**cluster** The final cluster membership.

**centers** A  $k \times p$  matrix, the rows of which contain the centers of the clusters in  $R^n$  (not to be confused with the clusters in feature space)

**wss** The within-cluster sum of squares for each cluster in feature space.

**param** The parameter value used.

### Examples

```
data <- as.matrix(iris[, 1:4])

# cluster using linear kernel (normal k-means)
result <- kkmeans(data, k = 3, kern = "poly", param = 1)

# cluster using gaussian kernel
# estimating the parameter with 3-nearest neighbors
result <- kkmeans(data, k = 3, kern = "g", estimate = "mknn", nn = 3)
```

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matr

*Estimate the bandwidth parameter for a gaussian kernel using MATR*

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### Description

given data and number of clusters  $K$ , choose a bandwidth from a grid according to the MATR algorithm

### Usage

```
matr(dat, k, grid, tol = 0.01)
```

**Arguments**

<code>dat</code>	data vector
<code>k</code>	number of clusters
<code>grid</code>	parameter grid to search on
<code>tol</code>	tolerance for choosing the "best" sigma values. Reducing the tolerance will give larger values of the bandwidth parameter

**Value**

A list with two elements. The first, `l`, contains the trace value for each of the values in `grid`. The second, `best`, contains the best value.

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