

# Package ‘leiden’

July 23, 2019

**Type** Package

**Title** R Implementation of Leiden Clustering Algorithm

**Version** 0.3.1

**Date** 2019-07-23

**Description** Implements the 'Python leidenalg' module to be called in R.  
Enables clustering using the leiden algorithm for partition a graph into communities.  
See the 'Python' repository for more details: <<https://github.com/vtraag/leidenalg>>  
Traag et al (2018) From Louvain to Leiden: guaranteeing well-connected communities. <arXiv:1810.08473>.

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**URL** <https://github.com/TomKellyGenetics/leiden>

**Imports** methods, reticulate, Matrix, igraph

**Encoding** UTF-8

**LazyData** true

**RoxygenNote** 6.1.1

**Suggests** data.table, tibble, devtools, covr, testthat, spelling,  
knitr, rmarkdown, RColorBrewer

**Language** en-US

**VignetteBuilder** knitr

**NeedsCompilation** no

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**Repository** CRAN

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leiden

*Run Leiden clustering algorithm*


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### Description

Implements the Leiden clustering algorithm in R using reticulate to run the Python version. Requires the python "leidenalg" and "igraph" modules to be installed. Returns a vector of partition indices.

### Usage

```
leiden(object, partition_type = c("RBConfigurationVertexPartition",
  "ModularityVertexPartition", "RBERVertexPartition", "CPMVertexPartition",
  "MutableVertexPartition", "SignificanceVertexPartition",
  "SurpriseVertexPartition"), initial_membership = NULL,
  weights = NULL, node_sizes = NULL, resolution_parameter = 1,
  seed = NULL, n_iterations = 2L)
```

### Arguments

<code>object</code>	An adjacency matrix compatible with <code>igraph</code> object or an input graph as an <code>igraph</code> object (e.g., shared nearest neighbours).
<code>partition_type</code>	Type of partition to use. Defaults to <code>RBConfigurationVertexPartition</code> . Options include: <code>ModularityVertexPartition</code> , <code>RBERVertexPartition</code> , <code>CPMVertexPartition</code> , <code>MutableVertexPartition</code> , <code>SignificanceVertexPartition</code> , <code>SurpriseVertexPartition</code> (see the Leiden python module documentation for more details)
<code>initial_membership, weights, node_sizes</code>	Parameters to pass to the Python <code>leidenalg</code> function (defaults <code>initial_membership=None</code> , <code>weights=None</code> ).
<code>resolution_parameter</code>	A parameter controlling the coarseness of the clusters
<code>seed</code>	Seed for the random number generator. By default uses a random seed if nothing is specified.
<code>n_iterations</code>	Number of iterations to run the Leiden algorithm. By default, 2 iterations are run. If the number of iterations is negative, the Leiden algorithm is run until an iteration in which there was no improvement.
<code>...</code>	Arguments to be passed to methods

### Value

A partition of clusters as a vector of integers

**Examples**

```

#check if python is available
modules <- reticulate::py_module_available("leidenalg") && reticulate::py_module_available("
if(modules){
#generate example data
adjacency_matrix <- rbind(cbind(matrix(round(rbinom(4000, 1, 0.8))), 20, 20),
                           matrix(round(rbinom(4000, 1, 0.3))), 20, 20),
                           matrix(round(rbinom(400, 1, 0.1))), 20, 20)),
                        cbind(matrix(round(rbinom(400, 1, 0.3))), 20, 20),
                           matrix(round(rbinom(400, 1, 0.8))), 20, 20),
                           matrix(round(rbinom(4000, 1, 0.2))), 20, 20)),
                        cbind(matrix(round(rbinom(400, 1, 0.3))), 20, 20),
                           matrix(round(rbinom(4000, 1, 0.1))), 20, 20),
                           matrix(round(rbinom(4000, 1, 0.9))), 20, 20))

rownames(adjacency_matrix) <- 1:60
colnames(adjacency_matrix) <- 1:60
#generate partitions
partition <- leiden(adjacency_matrix)
table(partition)

#generate partitions at a lower resolution
partition <- leiden(adjacency_matrix, resolution_parameter = 0.5)
table(partition)

#generate example weights
weights <- sample(1:10, sum(adjacency_matrix!=0), replace=TRUE)
partition <- leiden(adjacency_matrix, weights = weights)
table(partition)

#generate example weighted matrix
adjacency_matrix[adjacency_matrix == 1] <- weights
partition <- leiden(adjacency_matrix)
table(partition)

# generate (unweighted) igraph object in R
library("igraph")
adjacency_matrix[adjacency_matrix > 1] <- 1
snn_graph <- graph_from_adjacency_matrix(adjacency_matrix)
partition <- leiden(snn_graph)
table(partition)

# generate (weighted) igraph object in R
library("igraph")
adjacency_matrix[adjacency_matrix >= 1] <- weights
snn_graph <- graph_from_adjacency_matrix(adjacency_matrix, weighted = TRUE)
partition <- leiden(snn_graph)
table(partition)

# pass weights to python leidenalg
adjacency_matrix[adjacency_matrix >= 1 ] <- 1
snn_graph <- graph_from_adjacency_matrix(adjacency_matrix, weighted = NULL)

```

```
weights <- sample(1:10, sum(adjacency_matrix!=0), replace=TRUE)
partition <- leiden(snn_graph, weights = weights)
table(partition)

# run only if python is available (for testing)
}
```