

# Package ‘leidenAlg’

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**Type** Package

**Title** Implements the Leiden Algorithm via an R Interface

**Version** 1.1.3

## Description

An R interface to the Leiden algorithm, an iterative community detection algorithm on networks. The algorithm is designed to converge to a partition in which all subsets of all communities are locally optimally assigned, yielding communities guaranteed to be connected. The implementation proves to be fast, scales well, and can be run on graphs of millions of nodes (as long as they can fit in memory). The original implementation was constructed as a python interface “leidenalg” found here: <<https://github.com/vtraag/leidenalg>>. The algorithm was originally described in Traag, V.A., Waltman, L. & van Eck, N.J. “From Louvain to Leiden: guaranteeing well-connected communities”. Sci Rep 9, 5233 (2019) <[doi:10.1038/s41598-019-41695-z](https://doi.org/10.1038/s41598-019-41695-z)>.

**License** GPL-3

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**Encoding** UTF-8

**LazyData** true

**Depends** R (>= 3.5.0), Matrix

**Imports** graphics, grDevices, igraph, methods, parallel, Rcpp (>= 1.0.5), sccore, stats

**Suggests** pbapply, testthat (>= 3.1.0)

**LinkingTo** Rcpp, RcppArmadillo, RcppEigen

**SystemRequirements** GNU make (optional), libxml2 (optional), glpk (>= 4.57, optional)

**RoxygenNote** 7.2.3

**URL** <https://github.com/kharchenkolab/leidenAlg>

**BugReports** <https://github.com/kharchenkolab/leidenAlg/issues>

**NeedsCompilation** yes

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---

as.dendrogram.fakeCommunities

*Returns pre-calculated dendrogram*

---

### Description

Returns pre-calculated dendrogram

### Usage

```
## S3 method for class 'fakeCommunities'
as.dendrogram(object, ...)
```

### Arguments

object	fakeCommunities object
...	further parameters for generic

### Value

dendrogram

**Examples**

```
rLeidenComm = suppressWarnings(rleiden.community(exampleGraph, n.cores=1))
as.dendrogram.fakeCommunities(rLeidenComm)
```

---

exampleGraph

*Conos graph*


---

**Description**

Conos graph

**Usage**

```
exampleGraph
```

**Format**

An object of class igraph of length 100.

---

find\_partition

*Finds the optimal partition using the Leiden algorithm*


---

**Description**

Finds the optimal partition using the Leiden algorithm

**Usage**

```
find_partition(graph, edge_weights, resolution = 1, niter = 2)
```

**Arguments**

graph	The igraph graph to define the partition on
edge_weights	Vector of edge weights. In weighted graphs, a real number is assigned to each (directed or undirected) edge. For an unweighted graph, this is set to 1. Refer to igraph, weighted graphs.
resolution	Numeric scalar, resolution parameter controlling communities detected (default=1.0) Higher resolutions lead to more communities, while lower resolutions lead to fewer communities.
niter	Number of iterations that the algorithm should be run for (default=2)

**Value**

A vector of membership values

**Examples**

```
library(igraph)
library(leidenAlg)

g <- make_star(10)
E(g)$weight <- seq(ecount(g))
find_partition(g, E(g)$weight)
```

---

find\_partition\_rcpp     *Refer to the R function find\_partition() For notes of the graph object, refer to <https://igraph.org/c/doc/igraph-Basic.html>*

---

**Description**

Refer to the R function find\_partition() For notes of the graph object, refer to <https://igraph.org/c/doc/igraph-Basic.html>

**Usage**

```
find_partition_rcpp(
  edgelist,
  edgelist_length,
  num_vertices,
  direction,
  edge_weights,
  resolution = 1,
  niter = 2L
)
```

**Arguments**

edgelist	The graph edge list
edgelist_length	integer The length of the graph edge list
num_vertices	integer The number of vertices in the graph
direction	boolean Whether the graph is directed or undirected
edge_weights	Vector of edge weights. In weighted graphs, a real number is assigned to each (directed or undirected) edge. For an unweighted graph, this is set to 1. Refer to igraph, weighted graphs.
resolution	Numeric scalar, resolution parameter controlling communities detected (default=1.0) Higher resolutions lead to more communities, while lower resolutions lead to fewer communities.
niter	Number of iterations that the algorithm should be run for (default=2)

**Value**

A vector of membership values

**Examples**

```
library(igraph)
edgelist <- as.vector(t(igraph::as_edgelist(exampleGraph, names=FALSE))) - 1
edgelist_length <- length(edgelist)
num_vertices <- length(igraph::V(exampleGraph)) - 1
direction <- igraph::is_weighted(exampleGraph)
find_partition_rcpp(edgelist, edgelist_length, num_vertices, direction, E(exampleGraph)$weight)
```

---

find\_partition\_with\_rep

*Finds the optimal partition using the Leiden algorithm with replicate starts*

---

**Description**

This function performs Leiden algorithm `nrep` times and returns the result from the run with the maximum quality.

Since Leiden algorithm has stochastic process, repeating stochastically may improve the result. However, users should be aware of whether there is indeed a community structure with exploration, rather than blindly trusting the returned result that comes with the highest quality value.

The random number generator (RNG) is not re-seeded at each new start of community detection, in order to keep the independence of each replicate. To get reproducible result, users can run `set.seed()` before calling these functions.

`find_partition` only performs the community detection once and the reproducibility can also be ensured with `set.seed()`.

**Usage**

```
find_partition_with_rep(
  graph,
  edge_weights,
  resolution = 1,
  niter = 2,
  nrep = 10
)
```

**Arguments**

<code>graph</code>	The igraph graph to define the partition on
<code>edge_weights</code>	Vector of edge weights. In weighted graphs, a real number is assigned to each (directed or undirected) edge. For an unweighted graph, this is set to 1. Refer to igraph, weighted graphs.

resolution	Numeric scalar, resolution parameter controlling communities detected (default=1.0) Higher resolutions lead to more communities, while lower resolutions lead to fewer communities.
niter	Number of iterations that the algorithm should be run for (default=2)
nrep	Number of replicate starts with random number being updated. (default=10) The result with the best quality will be returned.

**Value**

A vector of membership values

**Examples**

```
library(igraph)

# To run 10 replicates and get the partitioning with the highest quality
membership <- find_partition_with_rep(exampleGraph, E(exampleGraph)$weight, nrep = 10)

# To get reproducible result for every function call, do `set.seed()` right before calling
set.seed(233)
res1 <- find_partition_with_rep(exampleGraph, E(exampleGraph)$weight, resolution = 2)
# Here, no seed was set...
res2 <- find_partition_with_rep(exampleGraph, E(exampleGraph)$weight, resolution = 2)
set.seed(233)
res3 <- find_partition_with_rep(exampleGraph, E(exampleGraph)$weight, resolution = 2)
identical(res1, res2) # FALSE (usually), as no seed as set
identical(res1, res3) # TRUE (always), as set.seed() was used directly before the function call
```

---

`find_partition_with_rep_rcpp`

*Finds the optimal partition using the Leiden algorithm*

---

**Description**

Finds the optimal partition using the Leiden algorithm

**Usage**

```
find_partition_with_rep_rcpp(
  edgelist,
  edgelist_length,
  num_vertices,
  direction,
  edge_weights,
  resolution = 1,
  niter = 2L,
  nrep = 1L
)
```

**Arguments**

edgelist	The graph edge list
edgelist_length	integer The length of the graph edge list
num_vertices	integer The number of vertices in the graph
direction	boolean Whether the graph is directed or undirected
edge_weights	Vector of edge weights. In weighted graphs, a real number is assigned to each (directed or undirected) edge. For an unweighted graph, this is set to 1. Refer to igraph, weighted graphs.
resolution	Numeric scalar, resolution parameter controlling communities detected (default=1.0) Higher resolutions lead to more communities, while lower resolutions lead to fewer communities.
niter	Number of iterations that the algorithm should be run for (default=2)
nrep	Number of replicate starts with random number being updated. (default=10) The result with the best quality will be returned.

**Details**

For notes of the graph object, refer to <https://igraph.org/c/doc/igraph-Basic.html>

**Examples**

```
library(igraph)

edgelist <- as.vector(t(igraph::as_edgelist(exampleGraph, names=FALSE))) - 1
edgelist_len <- length(edgelist) ## The length of the graph edge list
n_vertices <- length(igraph::V(exampleGraph)) - 1 ## The number of vertices in the graph
direct <- igraph::is_weighted(exampleGraph) ## Whether the graph is directed or undirected
edge_weights <- E(exampleGraph)$weight
find_partition_with_rep_rcpp(edgelist, edgelist_len, n_vertices, direct, edge_weights, nrep = 10)
```

---

leiden.community      *Leiden algorithm community detection Detects communities using Leiden algorithm (implementation copied from <https://github.com/vtraag/leidenalg>)*

---

**Description**

Leiden algorithm community detection Detects communities using Leiden algorithm (implementation copied from <https://github.com/vtraag/leidenalg>)

**Usage**

```
leiden.community(graph, resolution = 1, n.iterations = 2)
```

**Arguments**

graph            graph on which communities should be detected  
resolution      resolution parameter (default=1.0) - higher numbers lead to more communities  
n.iterations    number of iterations that the algorithm should be run for (default=2)

**Value**

a fakeCommunities object that returns membership and dendrogram

**Examples**

```
leiden.community(exampleGraph)
```

---

```
membership.fakeCommunities
```

*Returns pre-calculated membership factor*

---

**Description**

Returns pre-calculated membership factor

**Usage**

```
## S3 method for class 'fakeCommunities'  
membership(communities, ...)
```

**Arguments**

communities    fakeCommunities object  
...            further parameters for generic

**Value**

membership factor

**Examples**

```
leidenComm = leiden.community(exampleGraph)  
membership.fakeCommunities(leidenComm)
```



---

rleiden.community	<i>Recursive leiden communities Constructs an n-step recursive clustering, using leiden.community</i>
-------------------	---

---

**Description**

Recursive leiden communities Constructs an n-step recursive clustering, using leiden.community

**Usage**

```
rleiden.community(
  graph,
  max.depth = 2,
  n.cores = parallel::detectCores(logical = FALSE),
  min.community.size = 10,
  verbose = FALSE,
  resolution = 1,
  cur.depth = 1,
  hierarchical = TRUE,
  ...
)
```

**Arguments**

graph	graph
max.depth	Recursive depth (default=2)
n.cores	integer Number of cores to use (default = parallel::detectCores(logical=FALSE)). If logical=FALSE, uses the number of physical CPUs/cores. If logical=TRUE, uses the logical number of CPUS/cores. See parallel::detectCores()
min.community.size	integer Minimal community size parameter for the walktrap communities—Communities smaller than that will be merged (default=10)
verbose	boolean Whether to output progress messages (default=FALSE)
resolution	resolution parameter passed to leiden.community (either a single value, or a value equivalent to max.depth) (default=1)
cur.depth	integer Current depth of clustering (default=1)
hierarchical	boolean If TRUE, calculate hierarchy on the multilevel clusters (default=TRUE)
...	passed to leiden.community

**Value**

a fakeCommunities object that returns membership and dendrogram

**Examples**

```
rleiden.community(exampleGraph, n.cores=1)
```

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