Package ‘fastnet’

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degree.collect

Degrees of nodes

Description
Collect the degrees for all nodes in a network.

Usage
degree.collect(net)

Arguments
net The input network.

Details
Obtain the degrees for all nodes.

Value
A vector.

Author(s)
Xu Dong, Nazrul Shaikh.

Examples
## Not run:
x.deg <- degree.collect(x)
summary(x.deg)
## End(Not run)

degree.dist
Plot of the degree distribution of a network

Description
Plot the degree distribution of a network.

Usage
degree.dist(net, cumulative = TRUE, log = TRUE)
Arguments

net  The input network.
cumulative  A logical index asking whether a cumulative distribution should be returned.
log  A logical index asking whether a logarithm-scaled distribution should be returned.

Details

Plot the degree distribution of a network.

Value

A .gif plot.

Author(s)

Xu Dong

Examples

```r
## Not run:
x <- net.erdos.renyi.gnp(1000, 0.01)

## Plot the standard degree distribution of x.
degree.dist(x, cumulative = FALSE, log = FALSE)

## Plot the degree distribution of x, with a logarithm scale.
degree.dist(x, cumulative = FALSE, log = TRUE)

## Plot the cumulative degree distribution of x.
degree.dist(x, cumulative = TRUE, log = FALSE)

## Plot the cumulative degree distribution of x, with a logarithm scale.
degree.dist(x, cumulative = TRUE, log = TRUE)

## End(Not run)
```

degree.hist  Histogram of the degree distribution of a network

Description

Plot the histogram of all degrees of a network.

Usage

degree.hist(g, breaks = 100)
Arguments

- \texttt{g} The input network.
- \texttt{breaks} A single number giving the number of cells for the histogram.

Details

Plot the histogram of all degrees of a network.

Value

A .gif plot.

Author(s)

Xu Dong

Examples

```r
## Not run:
x <- net.erdos.renyi.gnp(1000, 0.05)
degree.hist(x)
## End(Not run)
```

---

draw.net \hspace{1cm} \textit{Plot of a small network}

Description

Plot a small network.

Usage

\begin{verbatim}
draw.net(net)
\end{verbatim}

Arguments

- \texttt{net} The input network.

Details

Plot a small network.

Value

A .gif plot.
**from.adjacency**

**Author(s)**  
Xu Dong, Nazrul Shaikh.

**Examples**

```r  
## Not run:  
x <- net::ring.lattice(12,4)  
draw.net(x)  
## End(Not run)  
```

---

**Description**

Transform an adjacency matrix to an ego-centric list form used in fastnet.

**Usage**

```r  
from.adjacency(adj.mat)  
```

**Arguments**

- `adj.mat`: The input adjacency matrix

**Value**

A list containing the nodes of the network and their respective neighbors.

**Author(s)**

Xu Dong, Christian Llano.

**Examples**

```r  
adj.mat <- matrix(c(0,1,0,0,1,0,0,0,0,0,1,0,0,0,1,0,0,0,0,1,0,0,0,1,0,0,0,0,1,0,0,0,1,0), nrow = 4, ncol = 4)  
g <- from.adjacency(adj.mat)  
```
from.edgelist

Edgelist to fastnet

Description
Transform an edgelist to an ego-centric list form used in fastnet.

Usage
from.edgelist(edgelist)

Arguments
edgelist  A 2-column data frame, in which the 1st column represents the start nodes, and the 2nd column represents the destination nodes.

Details
Most network data repositories choose to store the data in an edgelist form. This function helps user to load it in fastnet.

Value
A list containing the nodes of the network and their respective neighbors.

Author(s)
Xu Dong

Examples
edgelist <- data.frame(from=c(1, 3, 2, 3, 3), to=c(4, 5, 6, 5, 7))
g <- from.edgelist(edgelist)

from.igraph

Transform an igraph object to a fastnet object

Description
Transform an igraph object to an ego-centric list form used in fastnet.

Usage
from.igraph(net.igraph)
Arguments

net.igraph  The input igraph object.

Value

A list containing the nodes of the network and their respective neighbors.

Author(s)

Xu Dong.

Examples

## Not run:
library("igraph")
net.igraph <- erdos.renyi.game(100, 0.1)
g <- from.igraph(net.igraph)
## End(Not run)

Description

Import a statnet object.

Usage

from.statnet(net.statnet, ncores = detectCores())

Arguments

net.statnet  The input statnet object.

ncores  The number of cores to be used.

Value

A list containing the nodes of the network and their respective neighbors.

Author(s)

Xu Dong
Examples

## Not run:
library("ergm")
library("doParallel")
data("flo")
nflo <- network(flo, loops = TRUE)
fflo <- from.statnet(nflo)
## End(Not run)

---

**metric.cluster.global**

*Global Clustering Coefficient*

**Description**

Calculate the global clustering coefficient of a graph.

**Usage**

```r
metric.cluster.global(g)
```

**Arguments**

- `g` The input network.

**Details**

The global clustering coefficient measures the ratio of (closed) triples versus the total number of all possible triples in network `g`. `metric.cluster.global()` calculates the global clustering coefficient of `g`.

**Value**

A real constant.

**Author(s)**

Xu Dong, Nazrul Shaikh.

**References**


**Examples**

## Not run:
x <- net.erdos.renyi.gnp(1000, 0.01)
metric.cluster.global(x)
## End(Not run)
metric.cluster.mean  Mean Local Clustering Coefficient

Description

Calculate the average local clustering coefficient of a graph.

Usage

metric.cluster.mean(g)

Arguments

g  The input network.

Details

The local clustering coefficient of a node is the ratio of the triangles connected to the node and the triples centered on the node. metric.cluster.mean() calculates the (estimated) average clustering coefficient for all nodes in graph g with a justified error.

Value

A real constant.

Author(s)

Xu Dong, Nazrul Shaikh.

References


Examples

## Not run:
x <- net.erdos.renyi.gnp(n = 1000, ncores = 3, p = 0.06)
metric.cluster.mean(x)
## End(Not run)
metric.cluster.median  Median Local Clustering Coefficient

Description

Calculate the median local clustering coefficient of a graph.

Usage

metric.cluster.median(g)

Arguments

g  The input network.

Details

The local clustering coefficient of a node is the ratio of the triangles connected to the node and the triples centered on the node. metric.cluster.median() calculates the (estimated) median clustering coefficient for all nodes in graph g with a justified error.

Value

A real constant.

Author(s)

Xu Dong, Nazrul Shaikh.

References


Examples

## Not run:
x <- net.erdos.renyi.gnp(1000, 0.1)
metric.cluster.median(x)
## End(Not run)
**Effective Degree**

**Description**
Calculate the effective degree of a network.

**Usage**
```r
metric.degree.effective(g, effective_rate = 0.9)
```

**Arguments**
- `g`: The input network.
- `effective_rate`: The effective rate (0.9 is set by default).

**Details**
The effective degree

**Value**
A real constant.

**Author(s)**
Xu Dong, Nazrul Shaikh.

**Examples**
```r
## Not run:
metric.degree.effective(x)
## End(Not run)
```

---

**Degree Entropy**

**Description**
Calculate the degree entropy of a graph.

**Usage**
```r
metric.degree.entropy(g)
```
Arguments

  g        The input network.

Details

  Calculates the degree entropy of graph g, i.e.

  \[
  \text{Entropy}(g) = - \sum_{i=1}^{n} i \times \log_2(i)
  \]

Value

  A real constant.

Author(s)

  Xu Dong, Nazrul Shaikh.

References


Examples

  ## Not run:
  x <- net.erdos.renyi.gnp(1000, 0.01)
  metric.degree.entropy(x)
  ## End(Not run)

---

**metric.degree.max**  
**Maximal Degree**

Description

  Calculate the maximal degree of a graph.

Usage

  metric.degree.max(g)

Arguments

  g        The input network.

Details

  The maximal degree.
Efficient Maximal Degree

Description
Calculate the efficient maximal degree of a graph.

Usage
metric.degree.max.efficient(g)

Arguments

  g          The input network.

Details
The efficient maximal degree is the 90

Value
A real constant.

Author(s)
Xu Dong, Nazrul Shaikh.

Examples
## Not run: x <- net.erdos.renyi.gnp(1000, 0.01)
metric.degree.max.efficient(x)
## End(Not run)
**metric.degree.mean**  
*Mean Degree*

**Description**  
Calculate the mean degree of a graph.

**Usage**  
metric.degree.mean(g)

**Arguments**  
g  
The input network.

**Details**  
The mean degree is the average value of the degrees of all nodes in graph g.

**Value**  
A real constant.

**Author(s)**  
Xu Dong, Nazrul Shaikh.

**Examples**  
```r  
## Not run:  
x <- net.erdos.renyi.gnp(1000, 0.01)  
metric.degree.mean(x)  
## End(Not run)
```

---

**metric.degree.median**  
*Median Degree*

**Description**  
Calculate the median degree of a graph.

**Usage**  
metric.degree.median(g)
Arguments

\( g \) The input network.

Details

The median degree is the median value of the degrees of all nodes in graph \( g \).

Value

A real constant.

Author(s)

Xu Dong, Nazrul Shaikh.

Examples

```r
## Not run:
x <- net.erdos.renyi.gnp(1000, 0.01)
metric.degree.median(x)
## End(Not run)
```

---

**metric.degree.min**  
**Minimal Degree**

Description

Calculate the minimal degree of a network.

Usage

```r
metric.degree.min(g)
```

Arguments

\( g \) The input network.

Details

The minimal degree.

Value

A real constant.

Author(s)

Xu Dong, Nazrul Shaikh.
Examples

```r
## Not run:
metric.degree.min(x)
## End(Not run)
```

### Standard Deviation of Degree Distribution

**Description**

Calculate the standard deviation of all degrees of a network.

**Usage**

```r
metric.degree.sd(g)
```

**Arguments**

- `g`: The input network.

**Details**

The standard deviation of all degrees of a network.

**Value**

A real constant.

**Author(s)**

Xu Dong, Nazrul Shaikh.

**Examples**

```r
## Not run:
metric.degree.sd(x)
## End(Not run)
```
**Description**

Calculate the average path length of a graph.

**Usage**

```r
metric.distance.apl(Network, probability = 0.95, error = 0.03,
                     Cores = detectCores(), full.apl = FALSE)
```

**Arguments**

- **Network**
  - The input network.
- **probability**
  - The confidence level probability.
- **error**
  - The sampling error.
- **Cores**
  - Number of cores to use in the computations. By default uses `parallel` function `detectCores()`.
- **full.apl**
  - It will calculate the sampling version by default. If it is set to true, the population APL will be calculated and the rest of the parameters will be ignored.

**Details**

The average path length (APL) is the average shortest path lengths of all pairs of nodes in graph `Network`. `metric.distance.apl` calculates the population APL and estimated APL of graph `g` with a sampling error set by the user.

The calculation uses a parallel load balancing approach, distributing jobs equally among the cores defined by the user.

**Value**

A real value.

**Author(s)**

Luis Castro, Nazrul Shaikh.

**References**


**Examples**

```r
## Not run:
## Default function
x <- net.erdos.renyi.gnp(1000, 0.01)
m.NUMBER = diameter(x)
## Population APL
m.NUMBER = diameter(x, full = TRUE)
## Sampling at 99% level with an error of 10% using 5 cores
m.NUMBER = diameter(Network = x, probability = 0.99, error = 0.1, Cores = 5)
```

**Description**

Calculate the diameter of a graph.

**Usage**

```r
m.NUMBER = diameter(Network, probability = 0.95, error = 0.03, 
Cores = detectCores(), full = TRUE)
```

**Arguments**

- `Network`: The input network.
- `probability`: The confidence level probability
- `error`: The sampling error
- `Cores`: Number of cores to use in the computations. By default uses `parallel` function `detectCores()`.
- `full`: It will calculate the popular full version by default. If it is set to FALSE, the estimated diameter will be calculated.

**Details**

The diameter is the largest shortest path lengths of all pairs of nodes in graph `Network`.

**Value**

A real value.
Author(s)

Luis Castro, Nazrul Shaikh.

References


Examples

```r
## Not run:
## Default function
x <- net.erdos.renyi.gnp(1000,0.01)
metric.distance.diameter(x)
## Population APL
m = metric.distance.diameter(x, full=TRUE)
## Sampling at 99% level with an error of 10% using 5 cores
metric.distance.diameter(Network = x, probability=0.99, error=0.1, Cores=5)
## End(Not run)
```

---

**metric.distance.effdia**

*Effective Diameter*

**Description**

Calculate the effective diameter of a graph.

**Usage**

```r
metric.distance.effdia(Network, probability = 0.95, error = 0.03,
                         effective_rate = 0.9, Cores = detectCores(), full = TRUE)
```

**Arguments**

- **Network**: The input network.
- **probability**: The confidence level probability.
- **error**: The sampling error.
- **effective_rate**: The effective rate (by default it is set to be 0.9).
- **Cores**: Number of cores to use in the computations. By default uses `parallel` function `detectCores()`.
- **full**: It will calculate the popular full version by default. If it is set to FALSE, the estimated diameter will be calculated.
**metric.distance.meanecc**

**Details**

The diameter is the largest shortest path lengths of all pairs of nodes in graph `Network`. `metric.distance.diameter` calculates the (estimated) diameter of graph `Network` with a justified error.

**Value**

A real value.

**Author(s)**

Luis Castro, Nazrul Shaikh.

**References**


**Examples**

```r
## Not run:
##Default function
x <- net.erdos.renyi.gnp(1000, 0.01)
metric.distance.effdia(x)
##Population APL
metric.distance.effdia(x, full=TRUE)
##Sampling at 99% level with an error of 10% using 5 cores
metric.distance.effdia(Network = x, probability=0.99, error=0.1, Cores=5)

## End(Not run)
```

**metric.distance.meanecc**

*Mean Eccentricity*

**Description**

Calculate the mean eccentricity of a graph.

**Usage**

```r
metric.distance.meanecc(g, p)
```

**Arguments**

- `g` - The input network.
- `p` - The sampling probability.
Details
The mean eccentricities of all nodes in graph \( g \). Calculates the (estimated) mean eccentricity of graph \( g \) with a justified error.

Value
A real constant.

Author(s)
Xu Dong, Nazrul Shaikh.

References

Examples
```r
## Not run:
x <- net.erdos.renyi.gnp(1000, 0.01)
metric.distance.meanecc(x, 0.01)
## End(Not run)
```

---

**Median Eccentricity**

**Description**
Calculate the (estimated) median eccentricity of a graph.

**Usage**
```
metric.distance.medianecc(g, p)
```

**Arguments**
- \( g \) The input network.
- \( p \) The sampling probability.

**Details**
Is the median eccentricities of all nodes in graph \( g \). \texttt{metric.distance.medianecc} calculates the (estimated) median eccentricity of graph \( g \) with a justified error.
Value

A real constant.

Author(s)

Xu Dong, Nazrul Shaikh.

References


Examples

```r
# Not run:
x <- net.erdos.renyi.gnp(1000, 0.01)
metric.distance.medianecc(x, 0.01)
# End(Not run)
```

---

**Description**

Calculate the median path length (MPL) of a network.

**Usage**

```r
metric.distance.mpl(Network, probability = 0.95, error = 0.03,
Cores = detectCores(), full = FALSE)
```

**Arguments**

- `Network` - The input network.
- `probability` - The confidence level probability
- `error` - The sampling error
- `Cores` - Number of cores to use in the computations. By default `detectCores()` from `parallel`
- `full` - It calculates the sampling version by default. If it is set to true, the population MPL will be calculated and the rest of the parameters will be ignored.

**Details**

The median path length (MPL) is the median shortest path lengths of all pairs of nodes in `Network`. `metric.distance.mpl(g)` calculates the population MPL OR estimated MPL of network g with a sampling error set by the user. The calculation uses a parallel load balancing approach, distributing jobs equally among the cores defined by the user.
**metric.eigen.mean**

**Value**
A real integer

**Author(s)**
Luis Castro, Nazrul Shaikh.

**References**


**Examples**
```r
## Not run:
#Default function
x <- net.erdos.renyi.gnp(1000,0.01)
metric.distance.mpl(x)
##Population MPL
metric.distance.mpl(x, full=TRUE)
##Sampling at 99% level with an error of 10% using 5 cores
metric.distance.mpl(Network = x, probability=0.99, error=0.1, Cores=5)

## End(Not run)
```

**Description**
Calculate the mean eigenvalue centrality of a graph.

**Usage**
```
metric.eigen.mean(g)
```

**Arguments**
- `g`: The input network.

**Details**
`metric.eigen.mean` calculates the mean eigenvalue centrality score of graph `g`. 

---

**metric.eigen.mean**  
*Mean Eigenvalue Centrality*

---
**Value**

A real constant.

**Author(s)**

Xu Dong, Nazrul Shaikh.

**References**


**Examples**

```r
## Not run:
x <- net.erdos.renyi.gnp(1000, 0.01)
metric.eigen.mean(x)
## End(Not run)
```

---

**metric.eigen.median**  
*Median Eigenvalue Centrality*

**Description**

Calculate the median eigenvalue centrality of a graph.

**Usage**

```r
metric.eigen.median(g)
```

**Arguments**

- `g`  
The input network.

**Details**

`metric.eigen.median` calculates the median eigenvalue centrality score of graph `g`.

**Value**

A real constant.

**Author(s)**

Xu Dong, Nazrul Shaikh.
References


Examples

```r
# Not run:
x <- net.erdos.renyi.gnp(1000, 0.01)
metric.eigen.median(x)
# End(Not run)
```

<table>
<thead>
<tr>
<th>metric.eigen.value</th>
<th>Eigenvalue Score</th>
</tr>
</thead>
</table>

Description

Calculate the eigenvalue centrality score of a graph.

Usage

`metric.eigen.value(g)`

Arguments

g The input network.

Details

`metric.eigen.value` calculates the eigenvalue centrality score of graph `g`.

Value

A real constant.

Author(s)

Xu Dong, Nazrul Shaikh.

References


Examples

```r
## Not run:
x <- net.erdos.renyi.gnp(1000, 0.01)
metric.eigen.value(x)
## End(Not run)
```

Description

Calculate the density of a graph.

Usage

```r
metric.graph.density(g)
```

Arguments

- `g`: The input network.

Details

Computes the ratio of the number of edges and the number of possible edges.

Value

A real constant.

Author(s)

Xu Dong, Nazrul Shaikh.

Examples

```r
## Not run:
x <- net.erdos.renyi.gnp(1000, 0.01)
metric.graph.density(x)
## End(Not run)
```
neighbors

Neighbors of an agent in a network

Description

Presents all neighbors of a given node.

Usage

neighbors(netL nodeid)

Arguments

net

The input network.

nodeid

The ID of the input node.

Details

Neighbors of a node are nodes that directly connect to this node.

Value

A vector.

Author(s)

Xu Dong

Examples

## Not run:
x <- net.ring.lattice(12, 4)
neighbors(x, 2)
## End(Not run)

net.barabasi.albert

Barabasi-Albert Scale-free Graph

Description

Simulate a scale-free network using a preferential attachment mechanism (Barabasi and Albert, 1999)

Usage

net.barabasi.albert(n, m, ncores = detectCores(), d = FALSE)
Arguments

- **n**: Number of nodes of the network.
- **m**: Number of nodes to which a new node connects at each iteration.
- **ncores**: Number of cores, by default `detectCores()` from `parallel`.
- **d**: A logical value determining whether the generated network is a directed or undirected (default) network.

Details

Starting with $m$ nodes, the preferential attachment mechanism adds one node and $m$ edges in each step. The edges will be placed with one end on the newly-added node and the other end on the existing nodes, according to probabilities that associate with their current degrees.

Value

A list containing the nodes of the network and their respective neighbors.

Author(s)

Luis Castro, Xu Dong, Nazrul Shaikh.

References


Examples

```r
## Not run:
x <- net.barabasi.albert(1000, 20) # using default ncores
## End(Not run)
```

Description

Simulate a (connected) caveman network of $m$ cliques of size $k$.

Usage

```r
net.caveman(m, k, ncores = detectCores())
```

Arguments

- **m**: Number of cliques (or caves) in the network.
- **k**: Number of nodes per clique.
- **ncores**: Number of cores, by default `detectCores()` from `parallel`.
Details

The (connected) caveman network is formed by connecting a set of isolated $k$ - cliques (or "caves"), neighbor by neighbor and head to toe, using one edge that removed from each clique such that all $m$ cliques form a single circle (Watts 1999). The total number of nodes, i.e. $n$, in this network is given by $k + m$.

Value

A list containing the nodes of the network and their respective neighbors.

Author(s)

Xu Dong, Nazrul Shaikh.

References


Examples

```r
# Not run:
x <- net.caveman(50, 20) # using ncores by default
# End(Not run)
```

---

net.complete  Complete Network

Description

Simulate a complete (or full) network.

Usage

```r
net.complete(n, ncores = detectCores())
```

Arguments

- **n**: Number of nodes of the network.
- **ncores**: Number of cores, by default detectCores() from parallel.

Details

The $n$ nodes in the network are fully connected.

Note that the input $n$ should not exceed 10000, for the sake of memory overflow.
Value

A list containing the nodes of the network and their respective neighbors.

Author(s)

Xu Dong, Nazrul Shaikh.

Examples

```r
## Not run:
x <- net.complete(1000) # using ncores by default
## End(Not run)
```

Description

Simulate a random network with \( n \) nodes and \( m \) edges, according to Erdos and Renyi (1959).

Usage

```r
net.erdos.renyi.gnm(n, m, ncores = detectCores(), d = TRUE)
```

Arguments

- \( n \): Number of nodes of the network.
- \( m \): Number of edges of the network.
- \( ncores \): Number of cores, by default `detectCores()` from parallel.
- \( d \): A logical value determining whether is a network directed (default) or undirected.

Details

In this (simplest) random network, \( m \) edges are formed at random among \( n \) nodes. When \( d = \text{TRUE} \) is a directed network.

Value

A list containing the nodes of the network and their respective neighbors.

Author(s)

Xu Dong, Nazrul Shaikh.
References


Examples

```r
## Not run:
x <- net.erdos.renyi.gnp(1000, 100)
## End(Not run)
```

net.erdos.renyi.gnp  Directed / Undirected Erdos-Renyi G(n, p) network

Description

Simulate a random network with \( n \) nodes and a link connecting probability of \( p \), according to Edos and Renyi (1959).

Usage

```r
net.erdos.renyi.gnp(n, p, ncores = detectCores(), d = TRUE)
```

Arguments

- `n`: Number of nodes of the network.
- `p`: Connecting probability.
- `ncores`: Number of cores, by default `detectCores()` from `parallel`.
- `d`: A logical value determining whether is a network directed (default) or indiected.

Details

In this (simplest) random network, each edge is formed at random with a constant probability. When \( d = \) TRUE is a directed network.

Value

A list containing the nodes of the network and their respective neighbors.

Author(s)

Luis Castro, Xu Dong, Nazrul Shaikh.

References

Examples

```r
## Not run:
x <- net.erdos.renyi.gnp(1000, 0.01)
## End(Not run)
```

---

### Holme-Kim Network

**Description**

Simulate a scale-free network with relatively high clustering, comparing to B-A networks (Holme and Kim, 1999).

**Usage**

```r
net.holme.kim(n, m, pt)
```

**Arguments**

- `n`: Number of nodes of the network.
- `m`: Number of nodes to which a new node connects at each iteration.
- `pt`: Triad formation probability after each preferential attachment mechanism.

**Details**

The Holme-Kim network model is a simple extension of B-A model. It adds an additional step, called "Triad formation", with the probability `pt` that compensates the low clustering in B-A networks.

**Value**

A list containing the nodes of the network and their respective neighbors.

**Author(s)**

Xu Dong, Nazrul Shaikh

**References**


**Examples**

```r
## Not run:
x <- net.holme.kim(1000, 20, 0.1)
## End(Not run)
```
Description

Simulate a random network with a power-law degree distribution that has an exponential cutoff, according to Newman et al. (2001).

Usage

net.random.plc(n, cutoff, exponent)

Arguments

n The number of the nodes in the network.
cutoff Exponential cutoff of the degree distribution of the network.
exponent Exponent of the degree distribution of the network.

Details

The generated random network has a power-law degree distribution with an exponential degree cutoff.

Value

A list containing the nodes of the network and their respective neighbors.

Author(s)

Xu Dong, Nazrul Shaikh

References


Examples

```r
## Not run:
x <- net.random.plc(1000, 10, 2)
## End(Not run)
```
Description
Simulate a rewired caveman network of m cliques of size k, and with a link rewiring probability p.

Usage
net.rewired.caveman(nc, m, p)

Arguments
nc Number of cliques (or caves) in the network.

m Number of nodes per clique.

p Link rewiring probability.

Details
The rewired caveman network is built on the corresponding regular caveman network with m cliques of size k. Then the links in this caveman network are rewired with probability p.

Value
A list containing the nodes of the network and their respective neighbors.

Author(s)
Xu Dong, Nazrul Shaikh

References

Examples
## Not run:
x <- net.rewired.caveman(50, 20, 0.0005)
## End(Not run)
**Description**

Simulate a network with a $k$-regular ring lattice structure.

**Usage**

```r
net.ring.lattice(n, k)
```

**Arguments**

- `n`: Number of nodes in the network.
- `k`: Number of edges per node.

**Details**

The $n$ nodes are placed on a circle and each node is connected to the nearest $k$ neighbors.

**Value**

A list containing the nodes of the network and their respective neighbors.

**Author(s)**

Xu Dong, Nazrul Shaikh

**References**


**Examples**

```r
## Not run:
x <- net.ring.lattice(1000, 10)
## End(Not run)
```
net.watts.strogatz

Watts-Strogatz Small-world Network

Description

Simulate a small-world network according to the model of Watts and Strogatz (1998).

Usage

net.watts.strogatz(n, k, re)

Arguments

n  The number of the nodes in the network (or lattice).
k  Number of edges per node.
re Rewiring probability.

Details

The formation of Watts-Strogatz network starts with a ring lattice with \( n \) nodes and \( k \) edges per node, then each edge is rewired at random with probability \( re \).

Value

A list containing the nodes of the network and their respective neighbors.

Author(s)

Xu Dong, Nazrul Shaikh

References


Examples

```r
## Not run:
x <- net.watts.strogatz(1000, 10, 0.05)
## End(Not run)
```
preview.deg

Preview of the degree distribution of a network

Description
Present the first 10 degrees of a network.

Usage
preview.deg(g)

Arguments
g The input network.

Details
Present the first 10 degrees of a network.

Value
A vector.

Author(s)
Xu Dong, Nazrul Shaikh.

Examples
## Not run:
x <- net.ring.lattice(12,4)
preview.deg(x)
## End(Not run)

preview.net

Preview of a network

Description
Present the first 10 ego-centric lists of a network.

Usage
preview.net(net)
Arguments

net The input network.

Details

the connection condition of the first 10 nodes in a network.

Value

A list.

Author(s)

Xu Dong, Nazrul Shaikh.

Examples

## Not run:
```r
x <- net.ring.lattice(12,4)
preview.net(x)
## End(Not run)
```

Description

Coerce a fastnet object to edgelist.

Usage

to.edgelist(network, ncores)

Arguments

network A fastnet object.
ncores The number of cores to be used.

Value

A 2-column list with each row representing an edge, from source to destination

Author(s)

Xu Dong
Examples

## Not run:
g <- net.erdos.renyi.gnp(100, 0.1)
el <- to.edgelist(g)
## End(Not run)

Description

Coerce a fastnet object to an igraph object

Usage

to.igraph(g)

Arguments

- **g**: A fastnet object

Value

An igraph object

Author(s)

Xu Dong

Description

Coerce a fastnet object to a tidygraph object.

Usage

to.tidygraph(g)

Arguments

- **g**: A fastnet object.

Value

A tidygraph object
Author(s)

Xu Dong
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