ProbabilisticNetwork

Exported

 $\label{eq:QA::Array} $$ \{Float64, 2\}, L::Array\{Int64, 1\}$) Q - a measure of modularity$

This measures Barber's bipartite modularity based on a matrix and a list of module labels.

source: ProbabilisticNetwork/src/./modularity.jl:28

Q(P::ProbabilisticNetwork.Partition) Q – a measure of modularity

This measures Barber's bipartite modularity based on a Partition object, and update the object in the process.

source: ProbabilisticNetwork/src/./modularity.jl:46

 $Qr(A::Array{Float64, 2}, L::Array{Int64, 1})$ Qr – a measure of realized modularity

Measures Poisot's realized modularity, based on a a matrix and a list of module labels. Realized modularity takes values in the [0;1] interval, and is the proportion of interactions established *within* modules.

Note that in some situations, Qr can be *lower* than 0. This reflects a partition in which more links are established between than within modules.

source: ProbabilisticNetwork/src/./modularity.jl:64

 $\mathbf{Qr}(\mathbf{P}::\mathbf{ProbabilisticNetwork.Partition})$ Qr – a measure of realized modularity

Measures Poisot's realized modularity, based on a Partition object.

source: ProbabilisticNetwork/src/./modularity.jl:79

f(p): (p(1-p))
source: ProbabilisticNetwork/src/./proba_utils.jl:63
${\bf centrality_katz(A::Array\{Float64,\ 2\})} {\bf Measures} \ {\bf Katz's} \ {\bf centrality} \ {\bf for} \\ {\bf each} \ {\bf node} \ {\bf in} \ {\bf a} \ {\bf unipartite} \ {\bf network}.$
Keyword arguments
 a (def. 0.1), the weight of each subsequent connection k (def. 5), the maximal path length considered
source: ProbabilisticNetwork/src/./centrality.jl:10
<pre>connectance(A::Array{Float64, 2}) Expected connectance for a proba- bilistic matrix, measured as the number of expected links, divided by the size of the matrix. source: ProbabilisticNetwork/src/./connectance.jl:15</pre>
connectance_var(A::Array{Float64, 2}) Expected variance of the connectance for a probabilistic matrix, measured as the variance of the number of links divided by the squared size of the matrix. source: ProbabilisticNetwork/src/./connectance.jl:23
<pre>degree(A::Array{Float64, 2}) Expected degree source: ProbabilisticNetwork/src/./degree.jl:20</pre>

degree_in(A::Array{Float64, 2}) Expected number of ingoing degrees source: ProbabilisticNetwork/src/./degree.jl:14 degree_out(A::Array{Float64, 2}) Expected number of outgoing degrees source: ProbabilisticNetwork/src/./degree.jl:8 i_esp(p::Float64) Expected value of a single Bernoulli event Simply f(p): p source: ProbabilisticNetwork/src/./proba_utils.jl:37 i_var(p::Float64) Variance of a single Bernoulli event f(p): p(1-p)source: ProbabilisticNetwork/src/./proba utils.jl:50 links(A::Array{Float64, 2}) Expected number of links for a probabilistic matrix**source:** ProbabilisticNetwork/src/./connectance.jl:2 links_var(A::Array{Float64, 2}) Expected variance of the number of links for a probabilistic matrix **source:** ProbabilisticNetwork/src/./connectance.jl:8

m_var(p::Array{Float64, N}) Variance of a series of multiplicative Bernoulli events source: ProbabilisticNetwork/src/./proba utils.jl:72 make_bernoulli(A::Array{Float64, 2}) Generate a random 0/1 matrix from probabilities Returns a matrix B of the same size as A, in which each element B(i,j) is 1 with probability A(i,j). **source:** ProbabilisticNetwork/src/./matrix_utils.jl:22 make_binary(A::Array{Float64, 2}) Returns the adjacency/incidence matrix from a probability matrix Returns a matrix B of the same size as A, in which each element B(i,j) is 1 if A(i,j) is greater than 0. **source:** ProbabilisticNetwork/src/./matrix_utils.jl:73 make_threshold(A::Array{Float64, 2}, k::Float64) Generate a random 0/1 matrix from probabilities Returns a matrix B of the same size as A, in which each element B(i,j) is 1 if A(i,j) is > k. This is probably unwise to use this function since this practice is of questionnable relevance, but it is included for the sake of exhaustivity. k must be in [0;1]. source: ProbabilisticNetwork/src/./matrix_utils.jl:56 make_unipartite(A::Array{Float64, 2}) Transforms a bipartite network into a unipartite network Note that this function returns an asymetric unipartite network.

source: ProbabilisticNetwork/src/./matrix_utils.jl:7

nestedness(A::Array{Float64, 2}) Nestedness of a matrix, using the Bastolla et al. (XXXX) measure

Returns three values:

- nestedness of the entire matrix
- nestedness of the columns
- nestedness of the rows

source: ProbabilisticNetwork/src/./nestedness.jl:36

nestedness_axis(A::Array{Float64, 2}) Nestedness of a single axis
(called internally by nestedness)

source: ProbabilisticNetwork/src/./nestedness.jl:5

nodiag(A::Array{Float64, 2}) Sets the diagonal to 0

Returns a copy of the matrix A, with the diagonal set to 0. Will fail if the matrix is not square.

source: ProbabilisticNetwork/src/./matrix_utils.jl:38

 $null1(A::Array{Float64, 2})$ Given a matrix A, null1(A) returns a matrix with the same dimensions, where every interaction happens with a probability equal to the connectance of A.

source: ProbabilisticNetwork/src/./nullmodels.jl:7

null2(A::Array{Float64, 2}) Given a matrix A, null2(A) returns a matrix with the same dimensions, where every interaction happens with a probability equal to the degree of each species.

source: ProbabilisticNetwork/src/./nullmodels.jl:46

null3in(A::Array{Float64, 2}) Given a matrix A, null3in(A) returns a matrix with the same dimensions, where every interaction happens with a probability equal to the in-degree (number of predecessors) of each species, divided by the total number of possible predecessors.

source: ProbabilisticNetwork/src/./nullmodels.jl:34

null3out(A::Array{Float64, 2}) Given a matrix A, null3out(A) returns a matrix with the same dimensions, where every interaction happens with a probability equal to the out-degree (number of successors) of each species, divided by the total number of possible successors.

source: ProbabilisticNetwork/src/./nullmodels.jl:20

nullmodel(A::Array{Float64, 2}) This function is a wrapper to generate replicated binary matrices from a template probability matrix A.

If you use julia on more than one CPU, i.e. if you started it with julia $\neg p$ k where k is more than 1, this function will distribute each trial to one worker. Which means that it's fast.

Note that you will get a warning if there are less networks created than have been requested. Not also that this function generates networks, but do not check that their distribution is matching what you expect. Simulated annealing routines will be part of a later release.

Keyword arguments

- n (def. 1000), number of replicates to generate
- max (def. 10000), number of trials to make

source: ProbabilisticNetwork/src/./nullmodels.jl:78

propagate_labels(A::Array{Float64, 2}, kmax::Int64, smax::Int64) A very experimental label propagation method for probabilistic networks

This function is a take on the usual LP method for community detection. Instead of updating labels by taking the most frequent in the neighbors, this algorithm takes each interaction, and transfers the label across it with a probability equal

to the probability of the interaction. It is therefore not generalizable for non-probabilistic networks.

The other pitfall is that there is a need to do a *large* number of iterations to get to a good partition. This method is also untested.

source: ProbabilisticNetwork/src/./modularity.jl:98

ProbabilisticNetwork.Partition Type to store a community partition

This type has three elements:

ullet A, the probability matrix

- L, the array of (integers) module labels
- \bullet Q, if needed, the modularity value

source: ProbabilisticNetwork/src/./modularity.jl:13

Internal

 $\mathbf{@checkprob}(\mathbf{p})$ Quite crude way of checking that a number is a probability The two steps are

- 1. The number should be of the Float64 type if not, will yield a TypeError
- 2. The number should belong to [0,1] if not, will throw a DomainError

signature: checkprob(p)

source: ProbabilisticNetwork/src/./proba_utils.jl:15