
motifcluster

Release 0.0.2

William George Underwood, Andrew Elliott

May 01, 2020

CONTENTS

1	Introduction	3
2	Installation	5
3	Dependencies	7
4	Documentation	9
5	Tutorial	11
6	Authors	13
7	Links	15
8	Contents	17
9	Index	31
	Python Module Index	33
	Index	35

A Python package for motif-based spectral clustering of weighted directed networks.

INTRODUCTION

The **motifcluster** package provides implementations of motif-based spectral clustering of weighted directed networks in Python. These provide the capability for:

- Building motif adjacency matrices
- Sampling random weighted directed networks
- Spectral embedding with motif adjacency matrices
- Motif-based spectral clustering

The methods are all designed to run quickly on large sparse networks, and are easy to install and use.

INSTALLATION

```
pip install motifcluster
```


DEPENDENCIES

- Networkx
- Numpy
- Scipy
- Scikit-learn

DOCUMENTATION

Documentation for the **motifcluster** package is available on [Read the Docs](#), and also on GitHub in the [doc directory](#).

TUTORIAL

A tutorial for the **motifcluster** package is available on Github in the [tutorial directory](#).

CHAPTER
SIX

AUTHORS

- William George Underwood (maintainer)
- Andrew Elliott

LINKS

- Source code repository on [GitHub](#)
- Package index page on [PyPI](#)
- Documentation on [Read the Docs](#)

CONTENTS

8.1 Clustering methods

Functions for spectral clustering are in *motifcluster.clustering*.

cluster_spectrum (*spectrum*, *num_clusts*)

Get cluster assignments from spectrum using k-means++.

Get a list of cluster assignments from a spectrum, using k-means++ and *num_clusts* clusters.

Parameters

- **spectrum** (*dict*) – A dictionary containing “*vects*”: the matrix of eigenvectors to pass to k-means++.
- **num_clusts** (*int*) – The number of clusters to find.

Returns **cluster_assigns** – A list of integers from 1 to *num_clusts*, representing cluster assignments.

Return type list of int

run_motif_clustering (*adj_mat*, *motif_name*, *motif_type*='struc', *mam_weight_type*='unweighted', *mam_method*='sparse', *num_eigs*=2, *type_lap*='comb', *num_clusts*=2, *restrict*=True, *gr_method*='sparse')

Run motif-based clustering.

Run motif-based clustering on the adjacency matrix of a (weighted directed) network, using a specified motif, motif type, weighting scheme, embedding dimension, number of clusters and Laplacian type. Optionally restrict to the largest connected component before clustering.

Parameters

- **adj_mat** (*matrix*) – Adjacency matrix to be embedded.
- **motif_name** (*str*) – Motif used for the motif adjacency matrix.
- **motif_type** (*str*) – Type of motif adjacency matrix to use. One of “*func*” or “*struc*”.
- **mam_weight_type** (*str*) – Weighting scheme for the motif adjacency matrix. One of “*unweighted*”, “*mean*” or “*product*”.
- **mam_method** (*str*) – The method to use for building the motif adjacency matrix. One of “*sparse*” or “*dense*”.
- **num_eigs** (*int*) – Number of eigenvalues and eigenvectors for the embedding.
- **type_lap** (*str*) – Type of Laplacian for the embedding. One of “*comb*” or “*rw*”.
- **num_clusts** (*int*) – The number of clusters to find.

- **restrict** (*bool*) – Whether or not to restrict the motif adjacency matrix to its largest connected component before embedding.
- **gr_method** (*str*) – Format to use for getting largest component. One of “*sparse*” or “*dense*”.

Returns

- **adj_mat** (*sparse matrix*) – The original adjacency matrix.
- **motif_adj_mat** (*sparse matrix*) – The motif adjacency matrix.
- **comps** (*list*) – The indices of the largest connected component of the motif adjacency matrix (if `restrict=True`).
- **adj_mat_comps** (*matrix*) – The original adjacency matrix restricted to the largest connected component of the motif adjacency matrix (if `restrict=True`).
- **motif_adj_mat_comps** (*matrix*) – The motif adjacency matrix restricted to its largest connected component (if `restrict=True`).
- **vals** (*list*) – A length-`num_eigs` list containing the eigenvalues associated with the Laplace embedding of the (restricted) motif adjacency matrix.
- **vects** (*matrix*) – A matrix containing the eigenvectors associated with the Laplace embedding of the (restricted) motif adjacency matrix.
- **clusts** – A vector containing integers representing the cluster assignment of each vertex in the (restricted) graph.

Examples

```
>>> adj_mat = np.array(range(1, 10)).reshape((3, 3))
>>> run_motif_clustering(adj_mat, "M1")
```

8.2 Adjacency and indicator matrices

Functions for building adjacency and indicator matrices are in *motifcluster.indicators*.

build_G (*adj_mat*)

Build sparse adjacency matrix.

Build the sparse adjacency matrix *G* from a graph adjacency matrix.

Parameters **adj_mat** (*matrix*) – The original adjacency matrix.

Returns **G** – The adjacency matrix in sparse form.

Return type sparse matrix

build_Gd (*adj_mat*)

Build double-edge adjacency matrix.

Build the sparse double-edge adjacency matrix *Gd* from a graph adjacency matrix.

Parameters **adj_mat** (*matrix*) – The original adjacency matrix.

Returns **Gd** – A double-edge adjacency matrix in sparse form.

Return type sparse matrix

_build_Gp (*adj_mat*)

Build product matrix.

Build the sparse product matrix *Gp* from a graph adjacency matrix.

Parameters *adj_mat* (*matrix*) – The original adjacency matrix.

Returns **Gp** – A product matrix in sparse form.

Return type sparse matrix

_build_Gs (*adj_mat*)

Build single-edge indicator matrix.

Build the sparse single-edge adjacency matrix *Gs* from a graph adjacency matrix.

Parameters *adj_mat* (*matrix*) – The original adjacency matrix.

Returns **Gs** – A single-edge adjacency matrix in sparse form.

Return type sparse matrix

_build_Id (*adj_mat*)

Build identity matrix.

Build the sparse identity matrix *Id* from a graph adjacency matrix.

Parameters *adj_mat* (*matrix*) – The original adjacency matrix.

Returns **Id** – An identity matrix in sparse form.

Return type sparse matrix

_build_J (*adj_mat*)

Build directed indicator matrix.

Build the sparse directed indicator matrix *J* from a graph adjacency matrix.

Parameters *adj_mat* (*matrix*) – The original adjacency matrix.

Returns **J** – A directed indicator matrix in sparse form.

Return type sparse matrix

_build_J0 (*adj_mat*)

Build missing-edge indicator matrix.

Build the missing-edge indicator matrix *J0* from a graph adjacency matrix.

Parameters *adj_mat* (*matrix*) – The original adjacency matrix.

Returns **J0** – A missing-edge indicator matrix.

Return type sparse matrix

_build_Jd (*adj_mat*)

Build double-edge indicator matrix.

Build the sparse double-edge indicator matrix *Jd* from a graph adjacency matrix.

Parameters *adj_mat* (*matrix*) – The original adjacency matrix.

Returns **Jd** – A double-edge indicator matrix in sparse form.

Return type sparse matrix

`_build_Je` (*adj_mat*)

Build edge-and-diagonal matrix.

Build the sparse edge-and-diagonal matrix *Ie* from a graph adjacency matrix.

Parameters **adj_mat** (*matrix*) – The original adjacency matrix.

Returns **Ie** – An edge-and-diagonal matrix in sparse form.

Return type sparse matrix

`_build_Jn` (*adj_mat*)

Build vertex-distinct indicator matrix.

Build the vertex-distinct indicator matrix *Jn* from a graph adjacency matrix.

Parameters **adj_mat** (*matrix*) – The original adjacency matrix.

Returns **Jn** – A vertex-distinct indicator matrix.

Return type sparse matrix

`_build_Js` (*adj_mat*)

Build single-edge indicator matrix.

Build the sparse single-edge indicator matrix *Js* from a graph adjacency matrix.

Parameters **adj_mat** (*matrix*) – The original adjacency matrix.

Returns **Js** – A single-edge indicator matrix in sparse form.

Return type sparse matrix

8.3 Motif adjacency matrices

Functions for building motif adjacency matrices are in *motifcluster.motifadjacency*.

build_motif_adjacency_matrix (*adj_mat*, *motif_name*, *motif_type*='struc',
mam_weight_type='unweighted', *mam_method*='sparse')

Build a motif adjacency matrix.

Build a motif adjacency matrix from an adjacency matrix. Entry (*i*, *j*) of a motif adjacency matrix is the sum of the weights of all motifs containing both nodes *i* and *j*.

- The motif is specified by name and the type of motif instance can be one of:
 - Functional: motifs should appear as subgraphs.
 - Structural: motifs should appear as induced subgraphs.
- The weighting scheme can be one of:
 - Unweighted: the weight of any motif instance is one.
 - Mean: the weight of any motif instance is the mean of its edge weights.
 - Product: the weight of any motif instance is the product of its edge weights.

Parameters

- **adj_mat** (*matrix*) – Adjacency matrix from which to build the motif adjacency matrix.
- **motif_name** (*str*) – Motif used for the motif adjacency matrix.
- **motif_type** (*str*) – Type of motif adjacency matrix to build. One of “func” or “struc”.

- **mam_weight_type** (*str*) – The weighting scheme to use. One of “*unweighted*”, “*mean*” or “*product*”.
- **mam_method** (*str*) – Which formulation to use. One of “*dense*” or “*sparse*”. The sparse formulation avoids generating large dense matrices so tends to be faster for large sparse graphs.

Returns A motif adjacency matrix.

Return type sparse matrix

Examples

```
>>> adj_mat = np.array(range(1, 10)).reshape((3, 3))
>>> build_motif_adjacency_matrix(adj_mat, "M1", "func", "mean")
```

mam_M1 (*adj_mat*, *motif_type*, *mam_weight_type*)

Perform the motif adjacency matrix calculations for motif M1.

Parameters

- **adj_mat** (*matrix*) – Adjacency matrix from which to build the motif adjacency matrix.
- **motif_type** (*str*) – Type of motif adjacency matrix to build.
- **mam_weight_type** (*str*) – The weighting scheme to use. One of “*unweighted*”, “*mean*” or “*product*”.

Returns A motif adjacency matrix.

Return type sparse matrix

mam_M10 (*adj_mat*, *motif_type*, *mam_weight_type*, *mam_method*)

Perform the motif adjacency matrix calculations for motif M10.

Parameters

- **adj_mat** (*matrix*) – Adjacency matrix from which to build the motif adjacency matrix.
- **motif_type** (*str*) – Type of motif adjacency matrix to build.
- **mam_weight_type** (*str*) – The weighting scheme to use. One of “*unweighted*”, “*mean*” or “*product*”.
- **mam_method** (*str*) – Which formulation to use. One of “*dense*” or “*sparse*”.

Returns A motif adjacency matrix.

Return type sparse matrix

mam_M11 (*adj_mat*, *motif_type*, *mam_weight_type*, *mam_method*)

Perform the motif adjacency matrix calculations for motif M11.

Parameters

- **adj_mat** (*matrix*) – Adjacency matrix from which to build the motif adjacency matrix.
- **motif_type** (*str*) – Type of motif adjacency matrix to build.
- **mam_weight_type** (*str*) – The weighting scheme to use. One of “*unweighted*”, “*mean*” or “*product*”.
- **mam_method** (*str*) – Which formulation to use. One of “*dense*” or “*sparse*”.

Returns A motif adjacency matrix.

Return type sparse matrix

mam_M12 (*adj_mat*, *motif_type*, *mam_weight_type*, *mam_method*)

Perform the motif adjacency matrix calculations for motif M12.

Parameters

- **adj_mat** (*matrix*) – Adjacency matrix from which to build the motif adjacency matrix.
- **motif_type** (*str*) – Type of motif adjacency matrix to build.
- **mam_weight_type** (*str*) – The weighting scheme to use. One of “*unweighted*”, “*mean*” or “*product*”.
- **mam_method** (*str*) – Which formulation to use. One of “*dense*” or “*sparse*”.

Returns A motif adjacency matrix.

Return type sparse matrix

mam_M13 (*adj_mat*, *motif_type*, *mam_weight_type*, *mam_method*)

Perform the motif adjacency matrix calculations for motif M13.

Parameters

- **adj_mat** (*matrix*) – Adjacency matrix from which to build the motif adjacency matrix.
- **motif_type** (*str*) – Type of motif adjacency matrix to build.
- **mam_weight_type** (*str*) – The weighting scheme to use. One of “*unweighted*”, “*mean*” or “*product*”.
- **mam_method** (*str*) – Which formulation to use. One of “*dense*” or “*sparse*”.

Returns A motif adjacency matrix.

Return type sparse matrix

mam_M2 (*adj_mat*, *motif_type*, *mam_weight_type*)

Perform the motif adjacency matrix calculations for motif M2.

Parameters

- **adj_mat** (*matrix*) – Adjacency matrix from which to build the motif adjacency matrix.
- **motif_type** (*str*) – Type of motif adjacency matrix to build.
- **mam_weight_type** (*str*) – The weighting scheme to use. One of “*unweighted*”, “*mean*” or “*product*”.

Returns A motif adjacency matrix.

Return type sparse matrix

mam_M3 (*adj_mat*, *motif_type*, *mam_weight_type*)

Perform the motif adjacency matrix calculations for motif M3.

Parameters

- **adj_mat** (*matrix*) – Adjacency matrix from which to build the motif adjacency matrix.
- **motif_type** (*str*) – Type of motif adjacency matrix to build.
- **mam_weight_type** (*str*) – The weighting scheme to use. One of “*unweighted*”, “*mean*” or “*product*”.

Returns A motif adjacency matrix.

Return type sparse matrix

mam_M4 (*adj_mat, mam_weight_type*)

Perform the motif adjacency matrix calculations for motif M4.

Parameters

- **adj_mat** (*matrix*) – Adjacency matrix from which to build the motif adjacency matrix.
- **mam_weight_type** (*str*) – The weighting scheme to use. One of “*unweighted*”, “*mean*” or “*product*”.

Returns A motif adjacency matrix.

Return type sparse matrix

mam_M5 (*adj_mat, motif_type, mam_weight_type*)

Perform the motif adjacency matrix calculations for motif M5.

Parameters

- **adj_mat** (*matrix*) – Adjacency matrix from which to build the motif adjacency matrix.
- **motif_type** (*str*) – Type of motif adjacency matrix to build.
- **mam_weight_type** (*str*) – The weighting scheme to use. One of “*unweighted*”, “*mean*” or “*product*”.

Returns A motif adjacency matrix.

Return type sparse matrix

mam_M6 (*adj_mat, motif_type, mam_weight_type*)

Perform the motif adjacency matrix calculations for motif M6.

Parameters

- **adj_mat** (*matrix*) – Adjacency matrix from which to build the motif adjacency matrix.
- **motif_type** (*str*) – Type of motif adjacency matrix to build.
- **mam_weight_type** (*str*) – The weighting scheme to use. One of “*unweighted*”, “*mean*” or “*product*”.

Returns A motif adjacency matrix.

Return type sparse matrix

mam_M7 (*adj_mat, motif_type, mam_weight_type*)

Perform the motif adjacency matrix calculations for motif M7.

Parameters

- **adj_mat** (*matrix*) – Adjacency matrix from which to build the motif adjacency matrix.
- **motif_type** (*str*) – Type of motif adjacency matrix to build.
- **mam_weight_type** (*str*) – The weighting scheme to use. One of “*unweighted*”, “*mean*” or “*product*”.

Returns A motif adjacency matrix.

Return type sparse matrix

mam_M8 (*adj_mat, motif_type, mam_weight_type, mam_method*)

Perform the motif adjacency matrix calculations for motif M8.

Parameters

- **adj_mat** (*matrix*) – Adjacency matrix from which to build the motif adjacency matrix.

- **motif_type** (*str*) – Type of motif adjacency matrix to build.
- **mam_weight_type** (*str*) – The weighting scheme to use. One of “*unweighted*”, “*mean*” or “*product*”.
- **mam_method** (*str*) – Which formulation to use. One of “*dense*” or “*sparse*”.

Returns A motif adjacency matrix.

Return type sparse matrix

mam_M9 (*adj_mat*, *motif_type*, *mam_weight_type*, *mam_method*)

Perform the motif adjacency matrix calculations for motif M9.

Parameters

- **adj_mat** (*matrix*) – Adjacency matrix from which to build the motif adjacency matrix.
- **motif_type** (*str*) – Type of motif adjacency matrix to build.
- **mam_weight_type** (*str*) – The weighting scheme to use. One of “*unweighted*”, “*mean*” or “*product*”.
- **mam_method** (*str*) – Which formulation to use. One of “*dense*” or “*sparse*”.

Returns A motif adjacency matrix.

Return type sparse matrix

mam_Mcoll (*adj_mat*, *motif_type*, *mam_weight_type*, *mam_method*)

Perform the motif adjacency matrix calculations for motif Mcoll.

Parameters

- **adj_mat** (*matrix*) – Adjacency matrix from which to build the motif adjacency matrix.
- **motif_type** (*str*) – Type of motif adjacency matrix to build.
- **mam_weight_type** (*str*) – The weighting scheme to use. One of “*unweighted*”, “*mean*” or “*product*”.
- **mam_method** (*str*) – Which formulation to use. One of “*dense*” or “*sparse*”.

Returns A motif adjacency matrix.

Return type sparse matrix

mam_Md (*adj_mat*, *mam_weight_type*)

Perform the motif adjacency matrix calculations for motif Md.

Parameters

- **adj_mat** (*matrix*) – Adjacency matrix from which to build the motif adjacency matrix.
- **mam_weight_type** (*str*) – The weighting scheme to use. One of “*unweighted*”, “*mean*” or “*product*”.

Returns A motif adjacency matrix.

Return type sparse matrix

mam_Mexpa (*adj_mat*, *motif_type*, *mam_weight_type*, *mam_method*)

Perform the motif adjacency matrix calculations for motif Mexpa.

Parameters

- **adj_mat** (*matrix*) – Adjacency matrix from which to build the motif adjacency matrix.
- **motif_type** (*str*) – Type of motif adjacency matrix to build.

- **mam_weight_type** (*str*) – The weighting scheme to use. One of “*unweighted*”, “*mean*” or “*product*”.
- **mam_method** (*str*) – Which formulation to use. One of “*dense*” or “*sparse*”.

Returns A motif adjacency matrix.

Return type sparse matrix

mam_Ms (*adj_mat*, *motif_type*, *mam_weight_type*)

Perform the motif adjacency matrix calculations for motif Ms.

Parameters

- **adj_mat** (*matrix*) – Adjacency matrix from which to build the motif adjacency matrix.
- **motif_type** (*str*) – Type of motif adjacency matrix to build.
- **mam_weight_type** (*str*) – The weighting scheme to use. One of “*unweighted*”, “*mean*” or “*product*”.

Returns A motif adjacency matrix.

Return type sparse matrix

8.4 Network sampling

Functions for random sampling of weighted directed networks are in *motifcluster.sampling*.

demonstration_graph ()

Generate a small graph for demonstrations.

Generate the sparse and dense adjacency matrices of a small weighted directed graph, for demonstrating methods and running tests.

Returns

- **adj_mat_dense** (*matrix*) – the adjacency matrix in dense form.
- **adj_mat_sparse** (*sparse matrix*) – the adjacency matrix in sparse form.

sample_bsbm (*source_block_sizes*, *dest_block_sizes*, *bipartite_connection_matrix*, *bipartite_weight_matrix=None*, *sample_weight_type='unweighted'*)

Sample a bipartite stochastic block model (BSBM).

Sample the (weighted) adjacency matrix of a (weighted) bipartite stochastic block model (BSBM) with specified parameters.

Parameters

- **source_block_sizes** (*list of int*) – A list containing the size of each block of source vertices.
- **dest_block_sizes** (*list of int*) – A list containing the size of each block of destination vertices.
- **bipartite_connection_matrix** (*matrix*) – A matrix containing the source block to destination block connection probabilities.
- **sample_weight_type** (*str*) – The type of weighting scheme. One of “*unweighted*”, “*constant*” or “*poisson*”.
- **weight_matrix** (*matrix*) – A matrix containing the source block to destination block weight parameters. Unused for *sample_weight_type* = “*constant*”. Defaults to *None*.

Returns **adj_mat** – A randomly sampled (weighted) adjacency matrix of a BSBM.

Return type sparse matrix

Examples

```
>>> source_block_sizes = [10, 10]
>>> dest_block_sizes = [10, 10, 10]
>>> bipartite_connection_matrix = np.array([0.8, 0.5, 0.1, 0.1, 0.5, 0.8]).
↳reshape((2, 3))
>>> bipartite_weight_matrix = np.array([20, 10, 2, 2, 10, 20]).reshape((2, 3))
>>> sample_bsbm(block_sizes, bipartite_connection_matrix,
... bipartite_weight_matrix, "poisson")
```

sample_dsbm (*block_sizes*, *connection_matrix*, *weight_matrix*=None, *sample_weight_type*='unweighted')

Sample a directed stochastic block model (DSBM).

Sample the (weighted) adjacency matrix of a (weighted) directed stochastic block model (DSBM) with specified parameters.

Parameters

- **block_sizes** (*list of int*) – A list containing the size of each block of vertices.
- **connection_matrix** (*matrix*) – A matrix containing the block-to-block connection probabilities.
- **sample_weight_type** (*str*) – The type of weighting scheme. One of “unweighted”, “constant” or “poisson”.
- **weight_matrix** (*matrix*) – A matrix containing the block-to-block weight parameters. Unused for *sample_weight_type* = “constant”. Defaults to None.

Returns **adj_mat** – A randomly sampled (weighted) adjacency matrix of a DSBM.

Return type sparse matrix

Examples

```
>>> block_sizes = [10, 10]
>>> connection_matrix = np.array([0.8, 0.1, 0.1, 0.8]).reshape((2, 2))
>>> weight_matrix = np.array([10, 3, 3, 10]).reshape((2, 2))
>>> sample_dsbm(block_sizes, connection_matrix, weight_matrix, "poisson")
```

8.5 Spectral methods

Functions relating to spectral methods are in *motifcluster.spectral*.

_get_first_eigs (*some_mat*, *num_eigs*)

Compute first few eigenvalues and eigenvectors of a matrix.

Compute the first few eigenvalues (by magnitude) and associated eigenvectors of a matrix.

Parameters

- **some_mat** (*matrix*) – Matrix for which eigenvalues and eigenvectors are to be calculated.
- **num_eigs** (*int*) – Number of eigenvalues and eigenvectors to calculate.

Returns

- **vals** (*list*) – A length-*num_eigs* list of the first few eigenvalues.
- **vects** (*matrix*) – A *some_mat.shape[0]* by *num_eigs* matrix of the associated eigenvectors.

build_laplacian (*adj_mat*, *type_lap*='rw')

Build a Laplacian matrix.

Build a Laplacian matrix (combinatorial Laplacian or random-walk Laplacian) from a symmetric (weighted) graph adjacency matrix.

Parameters

- **adj_mat** (*matrix*) – Symmetric adjacency matrix from which to build the Laplacian.
- **type_lap** (*str*) – Type of Laplacian to build. One of “*comb*” (combinatorial) or “*rw*” (random-walk).

Returns The specified Laplacian matrix.

Return type sparse matrix

Examples

```
>>> adj_mat = np.array(range(1, 10)).reshape((3, 3))
>>> build_laplacian(adj_mat, "rw")
```

run_laplace_embedding (*adj_mat*, *num_eigs*, *type_lap*='rw')

Run Laplace embedding.

Run Laplace embedding on a symmetric (weighted) adjacency matrix with a specified number of eigenvalues and eigenvectors.

Parameters

- **adj_mat** (*matrix*) – Symmetric adjacency matrix to be embedded.
- **num_eigs** (*int*) – Number of eigenvalues and eigenvectors for the embedding.
- **type_lap** (*str*) – Type of Laplacian for the embedding. One of “*comb*” (combinatorial) or “*rw*” (random-walk).

Returns

- **vals** (*list*) – The length-*num_eigs* list of the first few eigenvalues of the Laplacian.
- **vects** (*matrix*) – An *adj_mat.shape[0]* by *num_eigs* matrix of the associated eigenvectors.

Examples

```
>>> adj_mat = np.array(range(1, 10)).reshape((3, 3))
>>> run_laplace_embedding(adj_mat, 2, "rw")
```

run_motif_embedding (*adj_mat*, *motif_name*, *motif_type*='struc', *mam_weight_type*='unweighted',
mam_method='sparse', *num_eigs*=2, *type_lap*='rw', *restrict*=True,
gr_method='sparse')

Run motif embedding.

Calculate a motif adjacency matrix for a given motif and motif type, optionally restrict it to its largest connected component, and then run Laplace embedding with specified Laplacian type and number of eigenvalues and eigenvectors.

Parameters

- **adj_mat** (*matrix*) – Adjacency matrix to be embedded.
- **motif_name** (*str*) – Motif used for the motif adjacency matrix.
- **motif_type** (*str*) – Type of motif adjacency matrix to use. One of “*func*” or “*struc*”.
- **mam_weight_type** (*str*) – Weighting scheme for the motif adjacency matrix. One of “*unweighted*”, “*mean*” or “*product*”.
- **mam_method** (*str*) – The method to use for building the motif adjacency matrix. One of “*sparse*” or “*dense*”.
- **num_eigs** (*int*) – Number of eigenvalues and eigenvectors for the embedding.
- **type_lap** (*str*) – Type of Laplacian for the embedding. One of “*comb*” or “*rw*”.
- **restrict** (*bool*) – Whether or not to restrict the motif adjacency matrix to its largest connected component before embedding.
- **gr_method** (*str*) – Format to use for getting largest component. One of “*sparse*” or “*dense*”.

Returns

- **adj_mat** (*sparse matrix*) – The original adjacency matrix.
- **motif_adj_mat** (*sparse matrix*) – The motif adjacency matrix.
- **comps** (*list*) – The indices of the largest connected component of the motif adjacency matrix (if restrict=True).
- **adj_mat_comps** (*matrix*) – The original adjacency matrix restricted to the largest connected component of the motif adjacency matrix (if restrict=True).
- **motif_adj_mat_comps** (*matrix*) – The motif adjacency matrix restricted to its largest connected component (if restrict=True).
- **vals** (*list*) – A length-*num_eigs* list containing the eigenvalues associated with the Laplace embedding of the (restricted) motif adjacency matrix.
- **vects** – A matrix containing the eigenvectors associated with the Laplace embedding of the (restricted) motif adjacency matrix.

Examples

```
adj_mat = np.array(range(1, 10)).reshape((3, 3)) run_motif_embedding(adj_mat, “M1”)
```

8.6 Utility functions

Assorted utility functions for the motifcluster module are in *motifcluster.utils*.

`_a_b_one` (*a_mat*, *b_mat*)

Compute a right-multiplication with the ones matrix.

Compute $a * (b @ \text{one_mat})$ where a , b , ones_mat are square matrices of the same size, and ones_mat contains all entries equal to one. The product $*$ is an entry-wise (Hadamard) product, while $@$ represents matrix multiplication. This method is more efficient than the naive approach when a or b are sparse.

Parameters **a**, **b** (*matrix*) – Square matrices of the same size.

Returns The sparse square matrix $a * (b @ one_mat)$.

Return type sparse matrix

`_a_one_b` (*a_mat*, *b_mat*)

Compute a left-multiplication with the ones matrix.

Compute $a * (one_mat @ b)$ where a , b , $ones_mat$ are square matrices of the same size, and $ones_mat$ contains all entries equal to one. The product $*$ is an entry-wise (Hadamard) product, while $@$ represents matrix multiplication. This method is more efficient than the naive approach when a or b are sparse.

Parameters **a, b** (*matrix*) – Square matrices of the same size.

Returns The sparse square matrix $a * (one_mat @ b)$.

Return type sparse matrix

`_drop0_killdiag` (*some_mat*)

Set diagonal entries to zero and sparsify.

Set the diagonal entries of a matrix to zero and convert it to sparse form.

Parameters **some_mat** (*matrix*) – A square matrix.

Returns **sparse_mat** – A sparse-form copy of *some_mat* with its diagonal entries set to zero.

Return type sparse matrix

`_random_sparse_matrix` (*m*, *n*, *p*, *sample_weight_type*='constant', *w*=1)

Build a random sparse matrix.

Build a sparse matrix of size $m * n$ with non-zero probability p . Edge weights can be unweighted, constant-weighted or Poisson-weighted.

Parameters

- **m, n** (*int*) – Dimension of matrix to build is (m , n).
- **p** (*float*) – Probability that each entry is non-zero (before weighting).
- **sample_weight_type** (*str*) – Type of weighting scheme.
- **w** (*float*) – Weight parameter.

Returns A random sparse matrix.

Return type sparse matrix

`get_largest_component` (*adj_mat*, *gr_method*)

Get largest connected component.

Get the indices of the vertices in the largest connected component of a graph from its adjacency matrix.

Parameters

- **adj_mat** (*matrix*) – An adjacency matrix of a graph.
- **gr_method** (*str*) – Format to use before building the graph. One of “sparse” or “dense”.

Returns **verts_to_keep** – A list of indices corresponding to the vertices in the largest connected component.

Return type list

Examples

```
>>> adj_mat = np.array([0, 1, 0, 0, 0, 0, 0, 0, 0]).reshape((3, 3))
>>> get_largest_component(adj_mat)
```

get_motif_names()

Get common motif names.

Get the names of some common motifs as strings.

Returns **motif_names** – A list of names (strings) of common motifs.

Return type list

INDEX

- genindex

PYTHON MODULE INDEX

m

- `motifcluster.clustering`, [17](#)
- `motifcluster.indicators`, [18](#)
- `motifcluster.motifadjacency`, [20](#)
- `motifcluster.sampling`, [25](#)
- `motifcluster.spectral`, [26](#)
- `motifcluster.utils`, [28](#)

Symbols

[_a_b_one\(\)](#) (in module *motifcluster.utils*), 28
[_a_one_b\(\)](#) (in module *motifcluster.utils*), 29
[_build_G\(\)](#) (in module *motifcluster.indicators*), 18
[_build_Gd\(\)](#) (in module *motifcluster.indicators*), 18
[_build_Gp\(\)](#) (in module *motifcluster.indicators*), 18
[_build_Gs\(\)](#) (in module *motifcluster.indicators*), 19
[_build_Id\(\)](#) (in module *motifcluster.indicators*), 19
[_build_J\(\)](#) (in module *motifcluster.indicators*), 19
[_build_J0\(\)](#) (in module *motifcluster.indicators*), 19
[_build_Jd\(\)](#) (in module *motifcluster.indicators*), 19
[_build_Je\(\)](#) (in module *motifcluster.indicators*), 19
[_build_Jn\(\)](#) (in module *motifcluster.indicators*), 20
[_build_Js\(\)](#) (in module *motifcluster.indicators*), 20
[_drop0_killdiag\(\)](#) (in module *motifcluster.utils*), 29
[_get_first_eigs\(\)](#) (in module *motifcluster.spectral*), 26
[_random_sparse_matrix\(\)](#) (in module *motifcluster.utils*), 29

B

[build_laplacian\(\)](#) (in module *motifcluster.spectral*), 27
[build_motif_adjacency_matrix\(\)](#) (in module *motifcluster.motifadjacency*), 20

C

[cluster_spectrum\(\)](#) (in module *motifcluster.clustering*), 17

D

[demonstration_graph\(\)](#) (in module *motifcluster.sampling*), 25

G

[get_largest_component\(\)](#) (in module *motifcluster.utils*), 29
[get_motif_names\(\)](#) (in module *motifcluster.utils*), 30

M

[mam_M1\(\)](#) (in module *motifcluster.motifadjacency*), 21
[mam_M10\(\)](#) (in module *motifcluster.motifadjacency*), 21
[mam_M11\(\)](#) (in module *motifcluster.motifadjacency*), 21
[mam_M12\(\)](#) (in module *motifcluster.motifadjacency*), 22
[mam_M13\(\)](#) (in module *motifcluster.motifadjacency*), 22
[mam_M2\(\)](#) (in module *motifcluster.motifadjacency*), 22
[mam_M3\(\)](#) (in module *motifcluster.motifadjacency*), 22
[mam_M4\(\)](#) (in module *motifcluster.motifadjacency*), 22
[mam_M5\(\)](#) (in module *motifcluster.motifadjacency*), 23
[mam_M6\(\)](#) (in module *motifcluster.motifadjacency*), 23
[mam_M7\(\)](#) (in module *motifcluster.motifadjacency*), 23
[mam_M8\(\)](#) (in module *motifcluster.motifadjacency*), 23
[mam_M9\(\)](#) (in module *motifcluster.motifadjacency*), 24
[mam_Mcoll\(\)](#) (in module *motifcluster.motifadjacency*), 24
[mam_Md\(\)](#) (in module *motifcluster.motifadjacency*), 24
[mam_Mexpa\(\)](#) (in module *motifcluster.motifadjacency*), 24
[mam_Ms\(\)](#) (in module *motifcluster.motifadjacency*), 25
 module
 motifcluster.clustering, 17
 motifcluster.indicators, 18
 motifcluster.motifadjacency, 20
 motifcluster.sampling, 25
 motifcluster.spectral, 26
 motifcluster.utils, 28
motifcluster.clustering
 module, 17
motifcluster.indicators
 module, 18
motifcluster.motifadjacency
 module, 20
motifcluster.sampling
 module, 25
motifcluster.spectral
 module, 26
motifcluster.utils
 module, 28

R

[run_laplace_embedding\(\)](#) (in module *motifcluster*), 25

ter.spectral), [27](#)
run_motif_clustering() (*in module motifcluster.clustering*), [17](#)
run_motif_embedding() (*in module motifcluster.spectral*), [27](#)

S

sample_bsbm() (*in module motifcluster.sampling*), [25](#)
sample_dsbm() (*in module motifcluster.sampling*), [26](#)