
PHETS Documentation

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INTRODUCTION

This package offers high-level tools for exploration and visualization of delay coordinate embedding and persistent homology. It is used to investigate the utilization of these together as a signal processing technique.

PHETS encompasses four submodules:

- *signals*
- *phomology*
- *embed*
- *prfstats*

signals holds the `TimeSeries` and `Trajectory` classes, which can be initialized from arrays or text files. Calling the `embed` method of a `TimeSeries` returns a `Trajectory`; calling the `project` method of `Trajectory` returns a `TimeSeries`. `TimeSeries` and `Trajectory` both inherit from `BaseTrajectory`, where all cropping, windowing, and normalization is handled.

phomology holds the `Filtration` class, which is initialized from a `Trajectory` and a dict of filtration parameters. Filtration movies, persistence diagrams, and persistence rank functions are created by calling the respective methods of the `Filtration` class.

embed holds the `embed` function, as well as functions for generating movies. The movies functions take one or more `TimeSeries` and return one or more `Trajectory` objects (created in the process of building the movies).

prfstats holds functions for statistical analysis of PRFs. Generally, they take one or two `Trajectory` objects, create PRFs from the windows of the `Trajectory` objects, do some analysis, and then save plots from the results.

TROUBLESHOOTING

2.1 compiling `find_landmarks.c` on OSX

PHETS requires the OpenMP C library `omp.h`. From what I can tell, OpenMP is not included in clang (the default C compiler on macOS), and may only be installed /configured for recent versions, and not with great ease. For these reasons, we've never tried to run PHETS on clang, and cannot guarantee it will work correctly.

On the other hand, OpenMP works with gcc out of the box, and you may already have a version of gcc installed. If so, determine the version and edit `find_landmarks_c_compile_str` in `config.py` to match. (NOTE: on macOS, gcc is a symlink for clang. This is avoided by including the version number, eg `gcc-5`.)

If you do not have gcc installed, you can do `brew install gcc` and then, as above, tweak `config.py`. You can also tell brew to install a particular version if you would like (anything 5+ should work).

A quick way to test if things are working is to run `python refresh.py` in the PHETS directory. This script will remove a number of temporary files and attempt to compile `find_landmarks.c`

If the compiler is still giving errors (don't mind warnings), try `brew upgrade gcc` or `brew reinstall gcc --without-multilib`

See [here](#) and [here](#) for more information.

REGRESSION TESTS

pytest is used for testing. To run the test suite, type `pytest --tb=short` from the top-level directory. (Running pytest within a subdirectory will only execute the tests for that submodule.)

Each submodule contains a `unit_test` directory. The tests themselves are defined in `unit_tests/test__<submodule>.py`. These are not exactly unit tests – rather, each one calls or initializes a user-facing feature and compares the result to a saved reference. The input data is found in `unit_tests/data` and the references in `<unit_tests/ref>`.

`prfstats/unit_tests/prepare__data.py` should be run regularly. In the case of the `prfstats` module, in order to keep test execution time low, the input data is pre-computed sets of Filtration instances. A small, correct change to PHETS can break Python's ability to load these objects from file, breaking the tests. In this case, run the routines in `prfstats/unit_tests/prepare__data.py`, and the tests should work correctly. Further, while an incorrect change should make at least one or two tests fail, many `prfstats` tests will not fail until the test data is re-generated with `prepare__data.py`. If a couple tests fail, run this file to see what else is broken down the pipeline, and always do so when a major change is completed.

Routines in `unit_tests/prepare__refs.py` should be run *only when you wish to change the behavior of existing functionality*. They should also be run individually (that is, don't change the refs for features that you aren't intentionally modifying).

TO DO

Here are a couple ideas for how PHETS can be improved:

- speed up generation of movies by streaming images to ffmpeg over stdin rather than saving to file, [like so](#)
- rewrite `embed.embed` with numpy for performance – can it be done without a for loop?
- improve `build_filtration.py` readability
- implement switch for memory profiling of filtration construction
- integrate new `utilities.timeit` decorator
- find a faster way than text files to pass data from `find_landmarks` to Python
- streamline passage of params to `find_landmarks`

DOCUMENTATION

This documentation is built with Sphinx. The autodoc extension is used to generate the [library reference](#) from docstrings in the Python code. The text and layout for all other sections (eg this paragraph) is defined in `docs/source/index.rst`.

To build this documentation, TeX must be installed, along with the following:

- `texlive-latex-recommended`
- `texlive-fonts-recommended`
- `texlive-latex-extra`
- `latexmk`

I used `sudo apt-get install <package>` for each.

To update the documentation:

```
cd docs
make latexpdf
mv build/latex/PHETS.pdf ../documentation.pdf
```

REFERENCE

6.1 signals

```
class signals.BaseTrajectory(data, crop=(None, None), num_windows=None,  
                             window_length=None, vol_norm=(False, False, False),  
                             time_units='samples', name=None, fname=None)
```

Parameters

- **data** (*str* or *array*) – The filename to load, or array. If a filename, sets *fname*.
- **crop** (*array*, *optional*) – Range of signal to work with. Observes *time_units*. Either or both bounds may be None. format: (start, stop). default: (None, None)
- **num_windows** (*int*, *optional*) – Slice signal into windows evenly spaced windows. default: None
- **window_length** (*int* or *float*, *optional*) – Observes *time_units* if None, *window_length* == *len*(data) / *num_windows* default: None
- **vol_norm** (*arr*, *optional*) – Normalize amplitude by (full, crop, window). default: (False, False, False)
- **time_units** (*str*, *optional*) – 'samples' or 'seconds' Observes *config.SAMPLE_RATE* default: 'samples'
- **name** (*string*, *optional*) – Sets name, a label used for titles for plots. If None and *fname* is not None, name is derived from *fname*. default: None
- **fname** (*string*, *optional*) – If data is not a filename (i.e. is an array), sets *fname*. default: None

```
crop (crop_cmd)
```

Set data to the region of *data_full* specified by *crop_cmd* and *time_units*.

Parameters **crop_cmd** (*array*) – observes *time_units* format: (start, stop)

```
slice (num_windows, window_length=None)
```

Sets windows, an array of evenly spaced windows from data.

Parameters

- **num_windows** (*int*) –
- **window_length** (*int* or *float*, *optional*) – observes 'time_units' if None, *window_length* == *len*(data) / *num_windows* default: None

```
class signals.TimeSeries(data, **kwargs)
```

Bases: *signals.signals.BaseTrajectory*

See [BaseTrajectory](#) for parameter descriptions

embed (*tau*, *m*)

Embed data_full, re-apply crop and slicing.

Parameters

- **tau** (*int* or *float*) – observes time_units
- **m** (*int*) –

Returns

Return type [Trajectory](#)

plot (*filename*)

Plot time series (crop only), save to filename. :param filename: :type filename: str

plot_full (*filename*)

Plot full time series with crop and windows demarcated, save to filename.

Parameters filename (*str*) –

class signals.**Trajectory** (*data*, ***kwargs*)

Bases: signals.signals.BaseTrajectory

See [BaseTrajectory](#) for parameter descriptions

filtrations (*filt_params*, *quiet=True*, *status_str=None*)

Compute filtration for each window of trajectory.

Parameters

- **filt_params** (*dict*) – see Filtration
- **quiet** (*bool*) – terminal output noise

Returns array of Filtration objects

Return type array

plot (*filename*)

Plot time series (crop only), save to filename. :param filename: :type filename: str

plot_full (*filename*)

Plot full trajectory save to filename.

Parameters filename (*str*) –

project (*axis=0*)

Project self.data_full to time series, re-apply crop and slicing.

Parameters axis (*int*) –

Returns

Return type [TimeSeries](#)

6.2 phomology

class phomology.**Filtration** (*traj*, *params*, *silent=False*, *save=True*)

Parameters

- **traj** ([Trajectory](#)) – trajectory from which to build the filtration

- **params** (*dict*) – options for landmark selection, witness complex, distance modification, etc. see `build_filtration.build_filtration()`
- **silent** (*bool, optional*) – suppress stdout
- **save** (*bool or str, optional*) – Save the filtration to file for later use. If `save` is a path/filename, save filtration to `save`; `save` should end with `.p`. Otherwise, if `save` is `True`, save filtration to `phomology/filtrations/filt.p`. In this case, filtration may be loaded by calling `load_filtration()` without providing the filename parameter.
default: `True`

intervals ()

Returns birth and death times for holes in the complex filtration

Return type Intervals

movie (*filename, **kwargs*)
build filtration visualization

Parameters

- **filename** (*str*) – Output path/filename. Should end in `'mp4'` or other movie format.
- **color_scheme** (*str, optional*) – `None`, `'highlight new'`, or `('birth time gradient', cycles)` where `cycles` is an `int` default: `None`
- **camera_angle** (*array*) – For 3D mode. (azimuthal, elevation) in degrees.
default: `(70, 45)`
- **alpha** (*float*) – Opacity of simplexes
default: `1`
- **dpi** (*int*) – plot resolution – dots per inch

Returns

Return type `None`

pd ()

Returns persistence diagram

Return type `PD`

plot_complex (*i, filename, **kwargs*)
plot complex at `i`th step of the filtration

Parameters

- **i** (*int*) –
- **filename** (*str*) – Output path/filename. Should end in `'png'` or other supported image format.

Returns

Return type `None`

plot_pd (*filename*)
plot the persistence diagram

Parameters **filename** (*str*) – Output path/filename. Should end in `'png'` or other supported image format.

Returns**Return type** None**plot_prf** (*filename*)

plot the persistence rank function

Parameters **filename** (*str*) – Output path/filename. Should end in ‘.png’ or other supported image format.**Returns****Return type** None**prf** ()**Returns** persistence rank function**Return type** PRF**phomology.load_filtration** (*filename=None*)

load a filtration from file

Parameters **filename** (*str*) – Path/filename. Should end with .p**phomology.build_filtration.build_filtration** (*input_file_name, parameter_set, silent=False*)**Parameters**

- **input_file_name** (*str*) –
- **parameter_set** (*dict*) – Options for filtration and landmark selection. Defaults are set in `config.py`

GENERAL:**num_divisions** Number of (epsilon) steps in filtration. The filtration parameter will be divided up equally in the interval [`min_filtration_param`, `max_filtration_param`].

default: 50

max_filtration_param The maximum value for the filtration parameter. If it is a negative integer, -x, the program will automatically choose the max filtration parameter such that the highest dimensional simplex constructed is of dimension x - 1.

default: -20

min_filtration_param The minimum value for the filtration parameter. Zero is usually fine.

default: 0

start How many lines to skip in the input file before reading data in.

default: 0

worm_length How many witnesses the program will read from the data file. If set to None, the program will read the file to the end. In general, a reasonable cap we have found is 10,000 witnesses and 200 or less landmarks.

default: None

ds_rate The ratio of number of witnesses / number of landmarks.

default: 50

landmark_selector "maxmin" How the landmarks are selected from among the witnesses. Only options are "EST" for equally spaced in time and "maxmin" for a max-min distance algorithm.

default: "maxmin"

WITNESS RELATION:

absolute The standard fuzzy witness relation says that a witness witnesses a simplex if the distance from the witness to each of the landmarks is within epsilon *more* than the distance to the closest landmark. If using the absolute relation, the closest landmark is dropped from the calculation, and the distance from a witness to each of the landmarks must be within epsilon of zero.

default: False

use_cliques If this is set to True, than witnesses are only used to connect edges, and higher simplices (faces, solids, etc.) are inferred from the 1-skeleton graph using the Bron-Kerbosch maximal clique finding algorithm. This can be useful in reducing noise if several of the false holes are triangles.

default: False

simplex_cutoff If not equal to zero, this caps the number of landmarks a witness can witness. Note: this does not effect automatic max_filtration_param selection.

default: 0

weak Uses a completely different relation. The filtration parameter k specifies that each witness will witness a simplex of its k-nearest neighbors. If this relation is used, max_filtration_param should be a positive integer, and num_divisions and min_filtration_param will be ignored.

default: False

use_twir Uses a completely different algorithm. TODO: insert your description here. Note: this works best with EST landmark selection. If max-min is used, be sure to set time_order_landmarks to True.

default: False

DISTANCE DISTORTIONS:

d_speed_amplify The factor by which distances are divided if the witness is at a relatively high speed.

default: 1

d_orientation_amplify The factor by which distances are divided if the witness and the landmark are travelling in similar directions.

default: 1

d_stretch The factor by which distances are divided if the vector from the witness to the landmark is in a similar direction (possibly backwards) as the direction in which time is flowing at the witness.

default: 1

d_ray_distance_amplify TODO: change this parameter. Right now, as long as the number is not 1, this will multiply the distance between two points by the distance between the closest points on the parameterized rays.

default: 1

d_use_hamiltonian If this is not zero, this will override all the above distortions. Distance will be computed using not only position coordinates, but also velocity coordinates. Velocity components are scaled by the value of this parameter (before squaring). If the value is negative, then the absolute value of the parameter is used, but the unit velocities are used instead of the actual velocities.

default: 0

use_ne_for_maxmin Whether or not to apply the above distance distortions to the max-min landmark selection (not recommended). Has no effect if landmark selector is EST.

default: False

MISC:

connect_time_1_skeleton If this is set to True, then on the first step of the filtration, each landmark will be adjoined by an edge to the next landmark in time. Note: this works best with EST landmark selection. If max-min is used, be sure to set `time_order_landmarks` to True.

default: False

reentry_filter Attempts to limit high dimensional simplices by requiring that landmarks get far away then come back. This only works if using cliques. Note: this works best with EST landmark selection. If max-min is used, be sure to set `time_order_landmarks` to True.

default: False

dimension_cutoff Simplexes with dimension greater than the dimension cutoff will be separated into their lower dimensional subsets when writing to the output file. This is very handy, as both Perseus and PHAT seem to take exponential time as a function of the dimension of a simplex. The caveat is that all homology greater than or equal to the dimension cutoff will be inaccurate. Thus, if one cares about Betti 2, dimension cutoff should be at least 3.

still valid / in use ??? setting to 0 doesn't affect tests

default: 2

store_top_simplices If there is a dimension cutoff in use, this parameter determines at which point in the process the simplices are decomposed. By setting this to False, smaller simplices will be stored when they are discovered. This makes the output file a bit smaller, but takes a bit longer. The results will be left unchanged.

default: True

- **silent** (*bool*) – Suppress stdout

Returns (simplexes, (landmarks, witnesses), eps)

Return type array

6.3 embed

`embed.embed(data, tau, m)`

Parameters

- **data** (*array*) – one dimensional (time series)

- **tau** (*int*) – delay (samples)
- **m** (*int*) – target dimension

Returns m-dimensional embedding

Return type array

`embed.movies.compare_multi` (*path1*, *path2*, *i_arr*, *out_fname*, *crop*, *time_units*, *m*, *tau*, *framerate=1*)

Embed two sets of files, varying file index, and view side-by-side

Parameters

- **path1** (*str*) – format-ready string, eg
'datasets/time_series/C134C/{:02d}-C134C.txt '
- **path2** (*str*) – format-ready string, eg
'datasets/time_series/C135B/{:02d}-C135B.txt '
- **i_arr** (*array*) – file indices
- **out_fname** (*str*) – path/filename for movie, should probably end with .mp4
- **crop** (*2-tuple of int or float*) – (start, stop). observes *time_units*. (None, None) works.
- **time_units** (*str*) – 'samples' or 'seconds', used for crop and tau
- **m** (*int*) – embedding dimension
- **tau** (*int or float*) – embedding delay, observes *time_units*
- **framerate** (*int, optional*) – movies frames per second
default: 1

Returns [trajs1, trajs2]

Return type 2d array of Trajectory instances

`embed.movies.compare_vary_tau` (*ts1*, *ts2*, *out_fname*, *m*, *tau*, *framerate=1*)

Like vary_tau, but two signals side-by-side.

Parameters

- **ts1** (*TimeSeries*) –
- **ts2** (*TimeSeries*) –
- **out_fname** (*str*) – path/filename for movie, should probably end with .mp4
- **m** (*int*) – embedding dimension
- **tau** (*int or float*) – embedding delay, observes *ts.time_units*
- **framerate** (*int, optional*) – movie frames per second
default: 1

Returns [trajs1, trajs2]

Return type 2d array of Trajectory instances

`embed.movies.slide_window` (*ts*, *out_fname*, *m*, *tau*, *framerate=1*)

Movie depicting embedding of each window of *ts*.

Parameters

- **ts** (*TimeSeries*) – pre-windowed (e.g., initialized with *num_windows* or call *slice* method before passing to this function).
- **out_fname** (*str*) – path/filename for movie, should probably end with *.mp4*
- **m** (*int*) – embedding dimension
- **tau** (*int or float*) – embedding delay, observes *ts.time_units*
- **framerate** (*int, optional*) – movie frames per second
default: 1

Returns**Return type** *Trajectory*`embed.movies.vary_tau(ts, out_fname, m, tau, framerate=1)`Movie depicting embedding of *ts* over a range of *tau*.**Parameters**

- **ts** (*TimeSeries*) –
- **out_fname** (*str*) – path/filename for movie, should probably end with *.mp4*
- **m** (*int*) – embedding dimension
- **tau** (*int or float*) – observes *ts.time_units*
- **framerate** (*int, optional*) – movie frames per second
default: 1

Returns array of *Trajectory* instances**Return type** 1-d array

6.4 prfstats

`prfstats.plot_dists_to_ref(path, out_filename, filt_params, i_ref, i_arr, weight_func=None,
load_filts=False, save_filts=True, samples=False, quiet=True)`

Take a set of trajectories as text files with one specified as the reference, compute a prf for each, plot the L2 distance from each prf to the reference prf.

Parameters

- **path** (*str*) – format-ready string. example: to use a set of files named *traj00.txt*, *traj01.txt*, *traj02.txt*, etc, located in *path/to/files*, use *'path/to/files/traj{:02d}.txt'*
- **out_filename** (*str*) – path/filename for plot output
- **filt_params** (*dict*) – see *phomology.build_filtration.build_filtration()*
- **i_ref** (*int*) – the reference prf is generated from *path.format(i_ref)*
- **i_arr** (*array of ints*) – prfs are generated from *[path.format(i) for i in i_arr]*
- **weight_func** (*lambda, optional*) – weight function applied to all prfs before computing distances
default: *None*

- **load_filts** (*bool or str, optional*) – If True, loads filtration set generated by previous run. If a filename, loads filtration set from specified file.

default: False

- **save_filts** (*bool or str, optional*) – if True, saves the filtration set such that `load_filts=True` may be used in subsequent runs.

if a filename, saves filtration set to specified file.

default: True

- **samples** (*dict or int, optional*) – If of the form `{'interval': i, 'filt_step': j}`, plots persistence diagram, persistence rank function, and *j*-th step of filtration for every *i*-th input file. If of the form `{'interval': i}` or simply *i*, plots persistence diagram, persistence rank function, and full filtration movie for every *i*-th input file.

These plots are saved to `output/prfstats/samples/`

default: False

- **quiet** (*bool, optional*) – less terminal output

default: True

Returns distances to reference prf

Return type 1-d array

```
prfstats.plot_dists_to_means(traj1, traj2, out_filename, filt_params, weight_func=None,
                             samples=False, load_filts=False, save_filts=True, quiet=True)
```

Take two pre-windowed Trajectories, generate prf for all windows, find mean prf for each Trajectory, plot distances from prfs to mean prf.

Parameters

- **traj1** (*Trajectory*) – Must be pre-windowed. This is accomplished by initializing with the `num_windows` parameter, calling the `slice` method, or using `TimeSeries.embed()`.
- **traj2** (*Trajectory*) – see *traj1*
- **out_filename** (*str*) – Path/filename for plot output. Should probably end with `.png`.
- **filt_params** (*dict*) – See `phomology.build_filtration.build_filtration()`
- **weight_func** (*lambda, optional*) – Weight function applied to all prfs before doing statistics

default: None

- **load_filts** (*bool or 2-tuple of str, optional*) – If True, loads filtration sets generated by previous run. If a tuple of filenames, loads filtration sets from specified files.

default: False

- **save_filts** (*bool or 2-tuple of str, optional*) – If True, saves the filtration sets such that `load_filts=True` may be used in subsequent runs. If a tuple of filenames, saves filtration sets to specified files.

default: True

- **samples** (*dict or int, optional*) – If of the form `{'interval': i, 'filt_step': j}`, plots persistence diagram, persistence rank function, and jrd step of filtration for every *i*th input file. If of the form `{'interval': i}` or simply *i*, plots persistence diagram, persistence rank function, and full filtration movie for every *i*th input file.

These plots are saved to `output/prfstats/samples/`

default: False

- **quiet** (*bool, optional*) – less terminal output

default: True

Returns [`dists_1_vs_1`, `dists_2_vs_1`, `dists_1_vs_2`, `dists_2_vs_2`]

Return type 2d array

`prfstats.plot_clusters(traj1, traj2, out_filename, filt_params, weight_func=None, samples=False, load_filts=False, save_filts=True, quiet=True)`

Just like `plot_dists_to_means()` except distances are plotted in a more succinct manner.

`prfstats.plot_rocs(traj1, traj2, out_filename, filt_params, k, vary_param=None, weight_func=None, load_filts=False, save_filts=True, samples=False, quiet=True)`

Take two pre-windowed Trajectories, generate prf for all windows, take half of prfs from each set of windows ('every other') to train two "L2Classifiers"s, use other half of prfs to test, report results as ROC curves.

Parameters

- **traj1** (*Trajectory*) – Must be pre-windowed. This is accomplished by initializing with the `num_windows` parameter, calling the `slice` method, or using `TimeSeries.embed()`.
- **traj2** (*Trajectory*) – see `traj1`
- **out_filename** (*str*) – Path/filename for plot output. Should probably end with `.png`.
- **filt_params** (*dict*) – See `phomology.build_filtration.build_filtration()`
- **k** (*1-d array of float*) – parameterizes true and false positive rates, determines thresholds
- **weight_func** (*lambda, optional*) – Weight function applied to all prfs before doing statistics
default: None
- **vary_param** (*2-tuple, optional*) – (*p*, (*val1*, *val2*, *val3*, ...)) where *p* is a string – either `'weight_func'` or any filtration parameter (see `build_filtration()`) default: None
- **load_filts** (*bool or 2-tuple of str, optional*) – If True, loads filtration sets generated by previous run. If a tuple of filenames, loads filtration sets from specified files.
default: False
- **save_filts** (*bool or 2-tuple of str, optional*) – If True, saves the filtration sets such that `load_filts=True` may be used in subsequent runs. If a tuple of filenames, saves filtration sets to specified files.
default: True

- **samples** (*dict or int, optional*) – If of the form `{'interval': i, 'filt_step': j}`, plots persistence diagram, persistence rank function, and jrd step of filtration for every *i*th input file. If of the form `{'interval': i}` or simply *i*, plots persistence diagram, persistence rank function, and full filtration movie for every *i*th input file.

These plots are saved to `output/prfstats/samples/`

default: False

- **quiet** (*bool, optional*) – less terminal output

default: True

Returns roc data for both classifiers

Return type array

```
prfstats.plot_variance(traj, out_filename, filt_params, vary_param_1, vary_param_2=None,
                      legend_labels_1=None, legend_labels_2=None, weight_func=None,
                      samples=False, quiet=True, annot_hm=False, load_filts=False,
                      save_filts=True, heatmaps=True)
```

Take pre-windowed Trajectory, compute prf for all windows for a range of one or two `'vary_param's`, compute various statistics on prfs per `vary_params`, plot results.

Parameters

- **traj** (*Trajectory*) – Must be pre-windowed. This is accomplished by initializing with the `num_windows` parameter, calling the `slice` method, or using `TimeSeries.embed()`.
- **out_filename** (*str*) – Path/filename for plot output. Should probably end with `.png`.
- **filt_params** (*dict*) – See `phomology.build_filtration.build_filtration()`
- **vary_param_1** (*tuple*) – (*p*, (*val1*, *val2*, *val3*, ...)) where *p* is a string – either `'weight_func'` or any filtration parameter (see `build_filtration()`)
- **vary_param_2** (*tuple, optional*) – see `vary_param_1` default: None
- **legend_labels_1** (*tuple of str, optional*) – Labels for use when a `vary_param_1` is `'weight_func'` or to otherwise override.
(`'axis'`, (`'tick 1'`, `'tick2'`, `'tick3'`))
- **legend_labels_2** (*tuple of str, optional*) – Labels for use when a `vary_param_2` is `'weight_func'`, or to otherwise override.
(`'legend 1'`, `'legend 2'`, `'legend 3'`)
- **weight_func** (*lambda, optional*) – Weight function applied to all prfs before doing statistics
default: None
- **samples** (*dict or int, optional*) – If of the form `{'interval': i, 'filt_step': j}`, plots persistence diagram, persistence rank function, and jrd step of filtration for every *i*th input file. If of the form `{'interval': i}` or simply *i*, plots persistence diagram, persistence rank function, and full filtration movie for every *i*th input file.

These plots are saved to `output/prfstats/samples/`

default: False

- **quiet** (*bool, optional*) – less terminal output
default: True
- **annot_hm** (*bool, optional*) – Annotate heatmaps – may be broken
default: True
- **load_filts** (*bool or 2-tuple of str, optional*) – If True, loads filtration sets generated by previous run. If a tuple of filenames, loads filtration sets from specified files.
default: False
- **save_filts** (*bool or 2-tuple of str, optional*) – If True, saves the filtration sets such that `load_filts=True` may be used in subsequent runs. If a tuple of filenames, saves filtration sets to specified files.
default: True
- **heatmaps** (*bool, optional*) – If True, plot pointwise stats to `output/prfstats/heatmaps`.
default: True

Returns scaler statistics

Return type array

```
prfstats.pairwise_mean_dists (traj, filt_params, vary_param_1, vary_param_2,
                             weight_func=None, samples=False, quiet=True, load_filts=False,
                             save_filts=True)
```

Parameters

- **traj** (*Trajectory*) – Must be pre-windowed. This is accomplished by initializing with the `num_windows` parameter, calling the `slice` method, or using `TimeSeries.embed()`.
- **filt_params** (*dict*) – See `phomology.build_filtration.build_filtration()`
- **vary_param_1** (*tuple*) – (`p`, (`val1`, `val2`, `val3`, ...)) where `p` is a string – either `'weight_func'` or any filtration parameter (see `build_filtration()`)
- **vary_param_2** (*tuple, optional*) – see `vary_param_1` default: None
- **weight_func** (*lambda, optional*) – Weight function applied to all prfs before doing statistics
default: None
- **samples** (*dict or int, optional*) – If of the form `{'interval': i, 'filt_step': j}`, plots persistence diagram, persistence rank function, and jrd step of filtration for every `i`th input file. If of the form `{'interval': i}` or simply `i`, plots persistence diagram, persistence rank function, and full filtration movie for every `i`th input file.
These plots are saved to `output/prfstats/samples/`
default: False
- **load_filts** (*bool or 2-tuple of str, optional*) – If True, loads filtration sets generated by previous run. If a tuple of filenames, loads filtration sets from specified files.
default: False

- **save_filts** (*bool or 2-tuple of str, optional*) – If True, saves the filtration sets such that `load_filts=True` may be used in subsequent runs. If a tuple of filenames, saves filtration sets to specified files.

default: True

- **quiet** (*bool, optional*) – less terminal output

default: True

Returns distances

Return type 2-d array

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