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“Neurosis is the inability to tolerate ambiguity” - Sigmund Freud

Welcome to the documentation for freud, a Python package for analyzing particle simulation trajectories of periodic systems. The library contains a diverse array of analysis routines designed for molecular dynamics and Monte Carlo simulation trajectories. Since any scientific investigation is likely to benefit from a range of analyses, freud is designed to work as part of a larger analysis pipeline. In order to maximize its interoperability with other systems, freud works with and returns NumPy arrays.
The recommended method of installing freud is using conda through the conda-forge channel. First download and install miniconda following conda’s instructions. Then, install freud from conda-forge:

```
$ conda install -c conda-forge freud
```

Alternatively, freud can be installed directly from source.

```
$ mkdir build
$ cd build
$ cmake ../
$ make install
```

### 1.1 Examples

Examples are provided as Jupyter notebooks in a separate freud-examples repository. These can be run locally with the `jupyter notebook` command. These examples will also be provided as static notebooks on NBViewer and interactive notebooks on MyBinder.

Visualization of data is done via Bokeh [Bokeh].

### 1.2 Installation

#### 1.2.1 Installing freud

You can either install freud via conda or compile it from source.

**Install via conda**

The code below will install freud from conda-forge.
conda install -c conda-forge freud

Compile from source

The following are **required** for installing freud:

- Python (2.7, 3.5, 3.6)
- NumPy
- Intel Threading Building Blocks (TBB)
- CMake

The following are **optional** for installing freud:

- Cython: The freud repository contains a Cython-generated `_freud.cpp` file that can be used directly. However, Cython is necessary if you wish to recompile this file.

The code that follows creates a build directory inside the freud source directory and builds freud:

```
mkdir build
cd build
cmake ../
# Use `cmake ../ -DENABLE_CYTHON=ON` to rebuild `_freud.cpp`
make install
```

By default, freud installs to the **USER_SITE** directory, which is in `~/.local` on Linux and in `~/Library` on macOS. **USER_SITE** is on the Python search path by default, so there is no need to modify **PYTHONPATH**.

**Note:** freud makes use of submodules. CMake has been configured to automatically initialize and update submodules. However, if this does not work, or you would like to do this yourself, please execute:

```
git submodule update --init
```

### 1.2.2 Unit Tests

The unit tests for freud are included in the repository and are configured to be run using the Python **unittest** library:

```
# Run tests from the tests directory
cd tests
python -m unittest discover .
```

Note that because freud is designed to require installation to run *(i.e. it cannot be run directly out of the build directory)*, importing freud from the root of the repository will fail because it will try and import the package folder. As a result, unit tests must be run from outside the root directory.

### 1.2.3 Documentation

The documentation for freud is hosted online at ReadTheDocs, but you may also build the documentation yourself:
Building the documentation

The following are required for building freud documentation:

- Sphinx

You can install sphinx using conda

```bash
conda install sphinx
```

or from PyPi

```bash
pip install sphinx
```

To build the documentation, run the following commands in the source directory:

```bash
cd doc
make html
# Then open build/html/index.html
```

To build a PDF of the documentation (requires LaTeX and/or PDFLaTeX):

```bash
cd doc
make latexpdf
# Then open build/latex/freud.pdf
```

1.3 Modules

Below is a list of modules in freud. To add your own module, read the development guide.

1.3.1 Bond Module

Overview

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<td>Compute bonds in a 3D system using a ((x, y, z)) coordinate system.</td>
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Details

The bond module allows for the computation of bonds as defined by a map. Depending on the coordinate system desired, either a two or three dimensional array is supplied, with each element containing the bond index mapped to the pair geometry of that element. The user provides a list of indices to track, so that not all bond indices contained in the bond map need to be tracked in computation.
The bond module is designed to take in arrays using the same coordinate systems as the \textit{PMFT Module} in \texttt{freud}.

\textbf{Note:} The coordinate system in which the calculation is performed is not the same as the coordinate system in which particle positions and orientations should be supplied. Only certain coordinate systems are available for certain particle positions and orientations:

- 2D particle coordinates (position: $[x, y, 0]$, orientation: $\theta$):
  - $r, \theta_1, \theta_2$.
  - $x, y$.
  - $x, y, \theta$.
- 3D particle coordinates:
  - $x, y, z$.

\texttt{class \texttt{freud.bond.BondingAnalysis}(num\_particles, num\_bonds)}

Analyze the bond lifetimes and flux present in the system.

\textit{Module author: Eric Harper <harperic@umich.edu>}

\textbf{Parameters}

- \texttt{num\_particles} (\texttt{unsigned int}) – Number of particles over which to calculate bonds.
- \texttt{num\_bonds} (\texttt{unsigned int}) – Number of bonds to track.

\textbf{Variables}

- \texttt{bond\_lifetimes} ($([N\texttt{\_particles}, \texttt{varying}] \texttt{numpy\_ndarray})$) – Bond lifetimes.
- \texttt{overall\_lifetimes} ($([N\texttt{\_particles}, \texttt{varying}] \texttt{numpy\_ndarray})$) – Overall bond lifetimes.
- \texttt{transition\_matrix} ($\texttt{numpy\_ndarray}$) – Transition matrix.
- \texttt{num\_frames} (\texttt{unsigned int}) – Number of frames calculated.
- \texttt{num\_particles} (\texttt{unsigned int}) – Number of tracked particles.
- \texttt{num\_bonds} (\texttt{unsigned int}) – Number of tracked bonds.

\texttt{compute}(\texttt{self}, \texttt{frame\_0}, \texttt{frame\_1})

Calculates the changes in bonding states from one frame to the next.

\textbf{Parameters}

- \texttt{frame\_0} ($([N\texttt{\_particles}, N\texttt{\_bonds}] \texttt{numpy\_ndarray})$) – Current/previous bonding frame (as output from \texttt{BondingR12} modules).
- \texttt{frame\_1} ($([N\texttt{\_particles}, N\texttt{\_bonds}] \texttt{numpy\_ndarray})$) – Next/current bonding frame (as output from \texttt{BondingR12} modules).

\texttt{getBondLifetimes}(\texttt{self})

Return the bond lifetimes.

\textbf{Returns} Lifetime of bonds.

\textbf{Return type} ($[N\texttt{\_particles}, \texttt{varying}] \texttt{numpy\_ndarray}$)

\texttt{getNumBonds}(\texttt{self})

Get number of bonds tracked.
Returns Number of bonds.

Return type unsigned int

**getNumFrames (self)**
Get number of frames calculated.

Returns Number of frames.

Return type unsigned int

**getNumParticles (self)**
Get number of particles being tracked.

Returns Number of particles.

Return type unsigned int

**getOverallLifetimes (self)**
Return the overall lifetimes.

Returns Lifetime of bonds.

Return type \((N_{\text{particles}}, \text{varying})\) numpy.ndarray

**getTransitionMatrix (self)**
Return the transition matrix.

Returns Transition matrix.

Return type numpy.ndarray

**initialize (self, frame_0)**
Calculates the changes in bonding states from one frame to the next.

Parameters frame_0 \(((N_{\text{particles}}, N_{\text{bonds}})\) numpy.ndarray\) – First bonding frame (as output from BondingR12 modules).

**class freud.bond.BondingR12 (r_max, bond_map, bond_list)**
Compute bonds in a 2D system using a \((r, \theta_1, \theta_2)\) coordinate system.

*Module author: Eric Harper <harperic@umich.edu>*

Parameters

- **r_max (float)** – Distance to search for bonds.
- **bond_map (numpy.ndarray)** – 3D array containing the bond index for each \(r, \theta_2, \theta_1\) coordinate.
- **bond_list (numpy.ndarray)** – List containing the bond indices to be tracked, \(\text{bond_list}[i] = \text{bond_index}\).

Variables

- **bonds (numpy.ndarray)** – Particle bonds.
- **box (freud.box.Box)** – Box used in the calculation.
- **list_map (dict)** – The dict used to map bond index to list index.
- **rev_list_map (dict)** – The dict used to map list idx to bond idx.

**compute (self, box, ref_points, ref_orientations, points, orientations, nlist=None)**
Calculates the correlation function and adds to the current histogram.

Parameters
• **box** (*freud.box.Box*) – Simulation box.

• **ref_points** (**(N\_particles, 3) numpy.ndarray**) – Reference points to calculate the bonding.

• **ref\_orientations** (**(N\_particles, 4)**) – Orientations as angles to use in computation.

• **points** (**(N\_particles, 3) numpy.ndarray**) – Points to calculate the bonding.

• **orientations** (**(N\_particles, 4) numpy.ndarray**) – Orientations as angles to use in computation.

• **nlist** (**freud.locality.NeighborList**, optional) – NeighborList to use to find bonds (Default value = None).

**getBonds** (*self*)
- Return the particle bonds.
- **Returns** Particle bonds.
- **Return type** `numpy.ndarray`

**getBox** (*self*)
- Get the box used in the calculation.
- **Returns** `freud.box.Box`.
- **Return type** `freud.box.Box`

**getListMap** (*self*)
- Get the dict used to map bond idx to list idx.
- **Returns** The mapping from bond to particle index.
- **Return type** `dict`

**getRevListMap** (*self*)
- Get the dict used to map list idx to bond idx.
- **Returns** The mapping from particle to bond index.
- **Return type** `dict`

**class** `freud.bond.BondingXY2D` (**x\_max, y\_max, bond\_map, bond\_list**)  
Compute bonds in a 2D system using a \((x, y)\) coordinate system.

*Module author: Eric Harper <harperic@umich.edu>*

**Parameters**

• **x\_max** (**float**) – Maximum \(x\) distance at which to search for bonds.

• **y\_max** (**float**) – Maximum \(y\) distance at which to search for bonds.

• **bond\_map** (**numpy.ndarray**) – 3D array containing the bond index for each \(x, y\) coordinate.

• **bond\_list** (**numpy.ndarray**) – List containing the bond indices to be tracked, `bond_list[i] = bond_index`.

**Variables**

• **bonds** (**numpy.ndarray**) – Particle bonds.

• **box** (**freud.box.Box**) – Box used in the calculation.

• **list\_map** (**dict**) – The dict used to map bond index to list index.
• **rev_list_map** *(dict)* – The dict used to map list idx to bond idx.

**compute** *(self, box, ref_points, ref_orientations, points, orientations, nlist=None)*
Calculates the correlation function and adds to the current histogram.

**Parameters**

- **box** *(freud.box.Box)* – Simulation box.
- **ref_points** *(\(N_{\text{particles}}, 3\)) numpy.ndarray* – Reference points to calculate the bonding.
- **ref_orientations** *(\(N_{\text{particles}}, 4\)) numpy.ndarray* – Orientations as angles to use in computation.
- **points** *(\(N_{\text{particles}}, 3\)) numpy.ndarray* – Points to calculate the bonding.
- **orientations** *(\(N_{\text{particles}}, 4\)) numpy.ndarray* – Orientations as angles to use in computation.
- **nlist** *(freud.locality.NeighborList, optional)* – NeighborList to use to find bonds (Default value = None).

**getBonds** *(self)*
Return the particle bonds.

**Returns** Particle bonds.

**Return type** numpy.ndarray

**getBox** *(self)*
Get the box used in the calculation.

**Returns** freud Box.

**Return type** freud.box.Box

**getListMap** *(self)*
Get the dict used to map list idx to bond idx.

**Returns** The mapping from bond to particle index.

**Return type** dict

**getRevListMap** *(self)*
Get the dict used to map list idx to bond idx.

**Returns** The mapping from particle to bond index.

**Return type** dict

**class** freud.bond.BondingXYT *(x_max, y_max, bond_map, bond_list)*
Compute bonds in a 2D system using a \((x, y, \theta)\) coordinate system.

For each particle in the system determine which other particles are in which bonding sites.

*Module author: Eric Harper* <harperic@umich.edu>

**Parameters**

- **x_max** *(float)* – Maximum \(x\) distance at which to search for bonds.
- **y_max** *(float)* – Maximum \(y\) distance at which to search for bonds.
- **bond_map** *(numpy.ndarray)* – 3D array containing the bond index for each \(x, y\) coordinate.
• **bond_list** (*numpy.ndarray*) – List containing the bond indices to be tracked, 
  `bond_list[i] = bond_index`.

**Variables**

• **bonds** (*numpy.ndarray*) – Particle bonds.

• **box** (*freud.box.Box*) – Box used in the calculation.

• **list_map** (*dict*) – The dict used to map bond index to list index.

• **rev_list_map** (*dict*) – The dict used to map list idx to bond idx.

**compute** *(self, box, ref_points, ref_orientations, points, orientations, nlist=None)*

Calculates the correlation function and adds to the current histogram.

**Parameters**

• **box** (*freud.box.Box*) – Simulation box

• **ref_points** (*(*N* particles, 3) *numpy.ndarray*) – Reference points to calculate the bonding.

• **ref_orientations** (*(*N* particles, 4) *numpy.ndarray*) – Orientations as angles to use in computation.

• **points** (*(*N* particles, 3) *numpy.ndarray*) – Points to calculate the bonding.

• **orientations** (*(*N* particles, 4) *numpy.ndarray*) – Orientations as angles to use in computation.

• **nlist** (*freud.locality.NeighborList*, optional) – NeighborList to use to find bonds (Default value = None).

**getBonds** *(self)*

Return the particle bonds.

**Returns**  
Particle bonds.

**Return type**  
*numpy.ndarray*

**getBox** *(self)*

Get the box used in the calculation.

**Returns**  
freud Box.

**Return type**  
*freud.box.Box*

**getListMap** *(self)*

Get the dict used to map list idx to bond idx.

**Returns**  
The mapping from bond to particle index.

**Return type**  
*dict*

**getRevListMap** *(self)*

Get the dict used to map list idx to bond idx.

**Returns**  
The mapping from particle to bond index.

**Return type**  
*dict*

**class**  
*freud.bond.BondingXYZ (x_max, y_max, z_max, bond_map, bond_list)*

Compute bonds in a 3D system using a *(x, y, z)* coordinate system.

For each particle in the system determine which other particles are in which bonding sites.

*Module author:* Eric Harper <harperic@umich.edu>
Parameters

- \texttt{x\_max (float)} – Maximum \(x\) distance at which to search for bonds.
- \texttt{y\_max (float)} – Maximum \(y\) distance at which to search for bonds.
- \texttt{z\_max (float)} – Maximum \(z\) distance at which to search for bonds.
- \texttt{bond\_map (numpy.ndarray)} – 3D array containing the bond index for each \(x\), \(y\), \(z\) coordinate.
- \texttt{bond\_list (numpy.ndarray)} – List containing the bond indices to be tracked, \(\text{bond\_list}[i] = \text{bond\_index}\).

Variables

- \texttt{bonds (numpy.ndarray)} – Particle bonds.
- \texttt{box (freud.box.Box)} – Box used in the calculation.
- \texttt{list\_map (dict)} – The dict used to map bond index to list index.
- \texttt{rev\_list\_map (dict)} – The dict used to map list idx to bond idx.

\texttt{compute (self, box, ref\_points, ref\_orientations, points, orientations, nlist=None)}

Calculates the correlation function and adds to the current histogram.

Parameters

- \texttt{box (freud.box.Box)} – Simulation box.
- \texttt{ref\_points ((N\_particles, 3) numpy.ndarray)} – Reference points to calculate the bonding.
- \texttt{ref\_orientations ((N\_particles, 4) numpy.ndarray)} – Orientations as angles to use in computation.
- \texttt{points ((N\_particles, 3) numpy.ndarray)} – Points to calculate the bonding.
- \texttt{orientations ((N\_particles, 4) numpy.ndarray)} – Orientations as angles to use in computation.
- \texttt{nlist (freud.locality.NeighborList, optional)} – NeighborList to use to find bonds (Default value = None).

\texttt{getBonds (self)}

Return the particle bonds.

\textbf{Returns} Particle bonds.

\textbf{Return type} \texttt{numpy.ndarray}

\texttt{getBox (self)}

Get the box used in the calculation.

\textbf{Returns} freud Box.

\textbf{Return type} \texttt{freud.box.Box}

\texttt{getListMap (self)}

Get the dict used to map list idx to bond idx.

\textbf{Returns} The mapping from bond to particle index.

\textbf{Return type} \texttt{dict}

\texttt{getRevListMap (self)}

Get the dict used to map list idx to bond idx.
Returns  The mapping from particle to bond index.

Return type  dict

1.3.2 Box Module

Overview

The box module provides the Box class, which defines the geometry of the simulation box. The module natively supports periodicity by providing the fundamental features for wrapping vectors outside the box back into it.

```python
class freud.box.Box(Lx, Ly, Lz, xy, xz, yz, is2D=None)
```

The freud Box class for simulation boxes.

Module author: Richmond Newman <newmanrs@umich.edu>

Module author: Carl Simon Adorf <csadorf@umich.edu>

Module author: Bradley Dice <bdice@bradleydice.com>

Changed in version 0.7.0: Added box periodicity interface

The Box class is defined according to the conventions of the HOOMD-blue simulation software. For more information, please see:


Parameters

- **Lx** *(float)* – Length of side x.
- **Ly** *(float)* – Length of side y.
- **Lz** *(float)* – Length of side z.
- **xy** *(float)* – Tilt of xy plane.
- **xz** *(float)* – Tilt of xz plane.
- **yz** *(float)* – Tilt of yz plane.
- **is2D** *(bool)* – Specify that this box is 2-dimensional, default is 3-dimensional.

Variables

- **xy** *(float)* – The xy tilt factor.
- **xz** *(float)* – The xz tilt factor.
- **yz** *(float)* – The yz tilt factor.
- **L** *(tuple, settable)* – The box lengths
- **Lx** *(tuple, settable)* – The x-dimension length.
- **Ly** *(tuple, settable)* – The y-dimension length.
- **Lz** *(tuple, settable)* – The z-dimension length.
• **Linv** *(tuple)* – The inverse box lengths.
• **volume** *(float)* – The box volume (area in 2D).
• **dimensions** *(int, settable)* – The number of dimensions (2 or 3).
• **periodic** *(list, settable)* – Whether or not the box is periodic.

cube *(type cls, L=None)*

Construct a cubic box with equal lengths.

**Parameters**

- **L** *(float)* – The edge length

from_box *(type cls, box, dimensions=None)*

Initialize a box instance from a box-like object.

**Parameters**

- **box** – A box-like object
- **dimensions** *(int)* – Dimensionality of the box (Default value = None)

**Note:** Objects that can be converted to freud boxes include lists like [Lx, Ly, Lz, xy, xz, yz], dictionaries with keys 'Lx', 'Ly', 'Lz', 'xy', 'xz', 'yz', 'dimensions', namedtuples with properties Lx, Ly, Lz, xy, xz, yz, dimensions, 3x3 matrices (see from_matrix()), or existing freud.box.Box objects.

If any of Lz, xy, xz, yz are not provided, they will be set to 0.

If all values are provided, a triclinic box will be constructed. If only Lx, Ly, Lz are provided, an orthorhombic box will be constructed. If only Lx, Ly are provided, a rectangular (2D) box will be constructed.

If the optional dimensions argument is given, this will be used as the box dimensionality. Otherwise, the box dimensionality will be detected from the dimensions of the provided box. If no dimensions can be detected, the box will be 2D if Lz == 0, and 3D otherwise.

**Returns** The resulting box object.

**Return type** freud.box.Box

from_matrix *(type cls, boxMatrix, dimensions=None)*

Initialize a box instance from a box matrix.

For more information and the source for this code, see: http://hoomd-blue.readthedocs.io/en/stable/box.html

**Parameters**

- **boxMatrix** *(array-like)* – A 3x3 matrix or list of lists
- **dimensions** *(int)* – Number of dimensions (Default value = None)

getCoordinates *(self, f)*

Alias for makeCoordinates()

Deprecated since version 0.8: Use makeCoordinates() instead.

**Parameters**

- **f** *(3 (numpy.ndarray))* – Fractional coordinates (x, y, z) between 0 and 1 within parallelepipedal box.
**getImage** *(self, vec)*

Returns the image corresponding to a wrapped vector.

New in version 0.8.

**Parameters**

vec *(3 numpy.ndarray)* – Coordinates of unwrapped vector.

**Returns**

Image index vector.

**Return type** *(3 numpy.ndarray)*

**getL**(self)

Return the lengths of the box as a tuple *(x, y, z)*.

**Returns**

Dimensions of the box as *(x, y, z)*.

**Return type** *(float, float, float)*

**getLatticeVector** *(self, i)*

Get the lattice vector with index *i*.

**Parameters**

i *(unsigned int)* – Index *(0 ≤ i < d)* of the lattice vector, where *d* is the box dimension *(2 or 3)*.

**Returns**

Lattice vector with index *i*.

**Return type** *list[float, float, float]*

**getLinv**(self)

Return the inverse lengths of the box *(1/Lx, 1/Ly, 1/Lz)*.

**Returns**

Dimensions of the box as *(1/Lx, 1/Ly, 1/Lz)*.

**Return type** *(float, float, float)*

**getLx**(self)

Length of the x-dimension of the box.

**Returns**

This box’s x-dimension length.

**Return type** *float*

**getLy**(self)

Length of the y-dimension of the box.

**Returns**

This box’s y-dimension length.

**Return type** *float*

**getLz**(self)

Length of the z-dimension of the box.

**Returns**

This box’s z-dimension length.

**Return type** *float*

**getPeriodic**(self)

Get the box’s periodicity in each dimension.

**Returns**

Periodic attributes in x, y, z.

**Return type** *list[bool, bool, bool]*

**getPeriodicX**(self)

Get the box periodicity in the x direction.

**Returns**

True if periodic, False if not.
Return type bool

getPeriodicY (self)
Get the box periodicity in the y direction.

Returns True if periodic, False if not.

Return type bool

getPeriodicZ (self)
Get the box periodicity in the z direction.

Returns True if periodic, False if not.

Return type bool

getTiltFactorXY (self)
Return the tilt factor xy.

Returns This box’s xy tilt factor.

Return type float

getTiltFactorXZ (self)
Return the tilt factor xz.

Returns This box’s xz tilt factor.

Return type float

getTiltFactorYZ (self)
Return the tilt factor yz.

Returns This box’s yz tilt factor.

Return type float

getVolume (self)
Return the box volume (area in 2D).

Returns Box volume.

Return type float

is2D (self)
Return if box is 2D (True) or 3D (False).

Returns True if 2D, False if 3D.

Return type bool

makeCoordinates (self, f)
Convert fractional coordinates into real coordinates.

Parameters f ((3) numpy.ndarray) – Fractional coordinates (x, y, z) between 0 and 1 within parallelepipedal box.

Returns Vector of real coordinates (x, y, z).

Return type list[float, float, float]

makeFraction (self, vec)
Convert real coordinates into fractional coordinates.

Parameters vec ((3) numpy.ndarray) – Real coordinates within parallelepipedal box.

Returns A fractional coordinate vector.
Return type  list[float, float, float]

**set2D** *(self, val)*
Set the dimensionality to 2D (True) or 3D (False).

Parameters**  **val** *(bool)* – 2D=True, 3D=False.

**setL** *(self, L)*
Set all side lengths of box to L.

Parameters  **L** *(float)* – Side length of box.

**setPeriodic** *(self, x, y, z)*
Set the box’s periodicity in each dimension.

Parameters
- **x** *(bool)* – True if periodic in x, False if not.
- **y** *(bool)* – True if periodic in y, False if not.
- **z** *(bool)* – True if periodic in z, False if not.

**setPeriodicX** *(self, val)*
Set the box periodicity in the x direction.

Parameters  **val** *(bool)* – True if periodic, False if not.

**setPeriodicY** *(self, val)*
Set the box periodicity in the y direction.

Parameters  **val** *(bool)* – True if periodic, False if not.

**setPeriodicZ** *(self, val)*
Set the box periodicity in the z direction.

Parameters  **val** *(bool)* – True if periodic, False if not.

**square** *(type cls, L=None)*
Construct a 2-dimensional (square) box with equal lengths.

Parameters  **L** *(float)* – The edge length.

**to_dict** *(self)*
Return box as dictionary.

Returns  Box parameters

Return type  dict

**to_matrix** *(self)*
Returns the box matrix (3x3).

Returns  box matrix

Return type  list of lists, shape 3x3

**to_tuple** *(self)*
Returns the box as named tuple.

Returns  Box parameters

Return type  namedtuple

**unwrap** *(self, vecs, imgs)*
Unwrap a given array of vectors inside the box back into real space, using an array of image indices that determine how many times to unwrap in each dimension.
Parameters

- **vecs** ((3) or (N, 3) numpy.ndarray) – Single vector or array of N vectors. The vectors are modified in place.
- **imgs** ((3) or (N, 3) numpy.ndarray) – Single image index or array of N image indices.

Returns: Vectors unwrapped by the image indices provided.

Return type: (3) or (N, 3) numpy.ndarray

wrap (self, vecs)
Wrap a given array of vectors from real space into the box, using the periodic boundaries.

Note: Since the origin of the box is in the center, wrapping is equivalent to applying the minimum image convention to the input vectors.

Parameters: vecs ((3) or (N, 3) numpy.ndarray) – Single vector or array of N vectors. The vectors are altered in place and returned.

Returns: Vectors wrapped into the box.

Return type: (3) or (N, 3) numpy.ndarray

### 1.3.3 Cluster Module

**Overview**

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<td>Finds clusters in a set of points.</td>
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**Details**

The cluster module aids in finding and computing the properties of clusters of points in a system.

**class** freud.cluster.Cluster (box, rcut)
Finds clusters in a set of points.

Given a set of coordinates and a cutoff, freud.cluster.Cluster will determine all of the clusters of points that are made up of points that are closer than the cutoff. Clusters are 0-indexed. The class contains an index array, the cluster_idx attribute, which can be used to identify which cluster a particle is associated with: cluster_obj.cluster_idx[i] is the cluster index in which particle i is found. By the definition of a cluster, points that are not within the cutoff of another point end up in their own 1-particle cluster.

Identifying micelles is one primary use-case for finding clusters. This operation is somewhat different, though. In a cluster of points, each and every point belongs to one and only one cluster. However, because a string of points belongs to a polymer, that single polymer may be present in more than one cluster. To handle this situation, an optional layer is presented on top of the cluster_idx array. Given a key value per particle (i.e. the polymer id), the computeClusterMembership function will process cluster_idx with the key values in mind and provide a list of keys that are present in each cluster.

*Module author: Joshua Anderson <joaander@umich.edu>*

**Parameters**
Variables

- **box** (*freud.box.Box*) – Box used in the calculation.
- **num_clusters** (*int*) – The number of clusters.
- **num_particles** (*int*) – The number of particles.
- **cluster_idx** (*\(N_{\text{particles}}\) numpy.ndarray*) – The cluster index for each particle.
- **cluster_keys** (*list\(\ (\text{list})\)*) – A list of lists of the keys contained in each cluster.

**computeClusterMembership** (*self, keys*)

Compute the clusters with key membership. Loops over all particles and adds them to a list of sets. Each set contains all the keys that are part of that cluster. Get the computed list with **getClusterKeys()**.

**Parameters**

- **keys** (*\(N_{\text{particles}}\) numpy.ndarray*) – Membership keys, one for each particle.

**computeClusters** (*self, points, nlist=None, box=None*)

Compute the clusters for the given set of points.

**Parameters**

- **points** (*\(N_{\text{particles}}, 3\) np.ndarray*) – Particle coordinates.
- **nlist** (*freud.locality.NeighborList*, optional) – Object to use to find bonds (Default value = None).
- **box** (*freud.box.Box*, optional) – Simulation box (Default value = None).

**getBox** (*self*)

Return the stored freud Box.

**Returns** freud Box.

**Return type** *freud.box.Box*

**getClusterIdx** (*self*)

Returns 1D array of Cluster idx for each particle.

**Returns** 1D array of cluster idx.

**Return type** *\(N_{\text{particles}}\) numpy.ndarray*

**getClusterKeys** (*self*)

Returns the keys contained in each cluster.

**Returns** List of lists of each key contained in clusters.

**Return type** *list*

**getNumClusters** (*self*)

Returns the number of clusters.

**Returns** Number of clusters.

**Return type** *int*
**getNumParticles** *(self)*

Returns the number of particles.

**Returns**  Number of particles.

**Return type**  int

```python
class freud.cluster.ClusterProperties(box)
```

Routines for computing properties of point clusters.

Given a set of points and cluster ids (from `Cluster`, or another source), `ClusterProperties` determines the following properties for each cluster:

- Center of mass
- Gyration tensor

The computed center of mass for each cluster (properly handling periodic boundary conditions) can be accessed with `getClusterCOM()`. This returns a \((N_{clusters}, 3)\) numpy.ndarray.

The \(3 \times 3\) gyration tensor \(G\) can be accessed with `getClusterG()`. This returns a numpy.ndarray, shape=\((N_{clusters} \times 3 \times 3)\). The tensor is symmetric for each cluster.

*Module author: Joshua Anderson <joaander@umich.edu>*

**Parameters**

- **box** *(freud.box.Box)* – Simulation box.

**Variables**

- **box** *(freud.box.Box)* – Box used in the calculation.
- **num_clusters** *(int)* – The number of clusters.
- **cluster_COM** \((N_{clusters}, 3)\) numpy.ndarray – The center of mass of the last computed cluster.
- **cluster_G** \((N_{clusters}, 3, 3)\) numpy.ndarray – The cluster \(G\) tensors computed by the last call to `computeProperties()`.
- **cluster_sizes** \((N_{clusters})\) numpy.ndarray – The cluster sizes computed by the last call to `computeProperties()`.

```python
computeProperties(self, points, cluster_idx, box=None)
```

Compute properties of the point clusters. Loops over all points in the given array and determines the center of mass of the cluster as well as the \(G\) tensor. These can be accessed after the call to `computeProperties()` with `getClusterCOM()` and `getClusterG()`.

**Parameters**

- **points** \((N_{particles}, 3)\) np.ndarray – Positions of the particles making up the clusters.
- **cluster_idx** np.ndarray – List of cluster indexes for each particle.
- **box** *(freud.box.Box, optional)* – Simulation box (Default value = None).

```python
getBox(self)
```

Return the stored `freud.box.Box` object.

**Returns**  freud Box

**Return type**  freud.box.Box

```python
getClusterCOM(self)
```

Returns the center of mass of the last computed cluster.

**Returns**  Cluster center of mass coordinates \((x, y, z)\).
Return type: \((N_{\text{clusters}}, 3)\) \text{numpy.ndarray}

\textbf{getClusterG}(\textit{self})

Returns the cluster \(G\) tensors computed by the last call to \texttt{computeProperties()}.

\textbf{Returns} List of gyration tensors for each cluster.

\textbf{Return type:} \((N_{\text{clusters}}, 3)\) \text{numpy.ndarray}

\textbf{getClusterSizes}(\textit{self})

Returns the cluster sizes computed by the last call to \texttt{computeProperties()}.

\textbf{Returns} Sizes of each cluster.

\textbf{Return type:} \((N_{\text{clusters}})\) \text{numpy.ndarray}

\textbf{getNumClusters}(\textit{self})

Count the number of clusters found in the last call to \texttt{computeProperties()}.

\textbf{Returns} Number of clusters.

\textbf{Return type:} \text{int}

1.3.4 Density Module

Overview

| \texttt{freud.density.FloatCF} | Computes the pairwise correlation function \((p \times q)(r)\) between two sets of points with associated values \(p\) and \(q\). |
| \texttt{freud.density.ComplexCF} | Computes the pairwise correlation function \((p \times q)(r)\) between two sets of points with associated values \(p\) and \(q\). |
| \texttt{freud.density.GaussianDensity} | Computes the density of a system on a grid. |
| \texttt{freud.density.LocalDensity} | Computes the local density around a particle. |
| \texttt{freud.density.RDF} | Computes RDF for supplied data. |

Details

The density module contains various classes relating to the density of the system. These functions allow evaluation of particle distributions with respect to other particles.

Correlation Functions

\textbf{class} \texttt{freud.density.FloatCF}(\textit{rmax, dr})

Computes the pairwise correlation function \((p \times q)(r)\) between two sets of points with associated values \(p\) and \(q\).

Two sets of points and two sets of real values associated with those points are given. Computing the correlation function results in an array of the expected (average) product of all values at a given radial distance.

The values of \(r\) where the correlation function is computed are controlled by the \textit{rmax} and \textit{dr} parameters to the constructor. \textit{rmax} determines the maximum distance at which to compute the correlation function and \textit{dr} is the step size for each bin.
Note: 2D: `freud.density.FloatCF` properly handles 2D boxes. The points must be passed in as \([x, y, 0]\). Failing to set \(z=0\) will lead to undefined behavior.

Self-correlation: It is often the case that we wish to compute the correlation function of a set of points with itself. If given the same arrays for both `points` and `ref_points`, we omit accumulating the self-correlation value in the first bin.

Module author: Matthew Spellings <mspells@umich.edu>

Parameters

- `rmax (float)` – Distance over which to calculate.
- `dr (float)` – Bin size.

Variables

- `RDF ((N_{bins}) \text{numpy.ndarray})` – Expected (average) product of all values at a given radial distance.
- `box (freud.box.Box)` – Box used in the calculation.
- `counts ((N_{bins}) \text{numpy.ndarray})` – The counts of each histogram bin.
- `R ((N_{bins}) \text{numpy.ndarray})` – The values of bin centers.

`accumulate (self, box, ref_points, refValues, points, values, nlist=None)`
Calculates the correlation function and adds to the current histogram.

Parameters

- `box (freud.box.Box)` – Simulation box.
- `ref_points ((N_{particles}, 3) \text{numpy.ndarray})` – Reference points to calculate the local density.
- `refValues ((N_{particles}) \text{numpy.ndarray})` – Values to use in computation.
- `points ((N_{particles}, 3) \text{numpy.ndarray})` – Points to calculate the bonding.
- `values ((N_{particles})` – Values to use in computation.
- `nlist (freud.locality.NeighborList, optional)` – NeighborList to use to find bonds (Default value = None).

`compute (self, box, ref_points, refValues, points, values, nlist=None)`
Calculates the correlation function for the given points. Will overwrite the current histogram.

Parameters

- `box (freud.box.Box)` – Simulation box
- `ref_points ((N_{particles}, 3) \text{numpy.ndarray})` – Reference points to calculate the local density.
- `refValues ((N_{particles}) \text{numpy.ndarray})` – Values to use in computation.
- `points ((N_{particles}, 3) \text{numpy.ndarray})` – Points to calculate the local density.
- `values ((N_{particles})` – Values to use in computation.
- `nlist (freud.locality.NeighborList, optional)` – NeighborList to use to find bonds (Default value = None).
**getBox**(self)
Get the box used in the calculation.

Returns  freud Box.

Return type  freud.box.Box

**getCounts**(self)
Get counts of each histogram bin.

Returns  Counts of each histogram bin.

Return type  \((N_{\text{bins}})\) numpy.ndarray

**getR**(self)
Get bin centers.

Returns  Values of bin centers.

Return type  \((N_{\text{bins}})\) numpy.ndarray

**getRDF**(self)
Returns the radial distribution function.

Returns  Expected (average) product of all values at a given radial distance.

Return type  \((N_{\text{bins}})\) numpy.ndarray

**reduceCorrelationFunction**(self)
Reduces the histogram in the values over N processors to a single histogram. This is called automatically by `freud.density.FloatCF.getRDF()`, `freud.density.FloatCF.getCounts()`.

**resetCorrelationFunction**(self)
Resets the values of the correlation function histogram in memory.

---

**class**  freud.density.ComplexCF(rmax, dr)
Computes the pairwise correlation function \(\langle p \ast q \rangle (r)\) between two sets of points with associated values \(p\) and \(q\).

Two sets of points and two sets of complex values associated with those points are given. Computing the correlation function results in an array of the expected (average) product of all values at a given radial distance.

The values of \(r\) where the correlation function is computed are controlled by the \(r_{\text{max}}\) and \(dr\) parameters to the constructor. \(r_{\text{max}}\) determines the maximum distance at which to compute the correlation function and \(dr\) is the step size for each bin.

**Note:** 2D: `freud.density.ComplexCF` properly handles 2D boxes. The points must be passed in as \([x, y, 0]\). Failing to set \(z=0\) will lead to undefined behavior.

Self-correlation: It is often the case that we wish to compute the correlation function of a set of points with itself. If given the same arrays for both `points` and `ref_points`, we omit accumulating the self-correlation value in the first bin.

**Module author:** Matthew Spellings <mspells@umich.edu>

**Parameters**

- \(r_{\text{max}}\) (`float`) – Distance over which to calculate.
- \(dr\) (`float`) – Bin size.

**Variables**
• RDF ((N\text{bins}) \text{numpy.ndarray}) – Expected (average) product of all values at a given radial distance.
• box (freud.box.Box) – Box used in the calculation.
• counts ((N\text{bins}) \text{numpy.ndarray}) – The counts of each histogram bin.
• R ((N\text{bins}) \text{numpy.ndarray}) – The values of bin centers.

accumulate (self, box, ref_points, refValues, points, values, nlist=None)
Calculates the correlation function and adds to the current histogram.

Parameters

• box (freud.box.Box) – Simulation box.
• ref_points ((N\text{particles}, 3) \text{numpy.ndarray}) – Reference points to calculate the local density.
• refValues ((N\text{particles}) \text{numpy.ndarray}) – Values to use in computation.
• points ((N\text{particles}, 3) \text{numpy.ndarray}) – Points to calculate the bonding.
• values ((N\text{particles}) – Values to use in computation.
• nlist (freud.locality.NeighborList, optional) – NeighborList to use to find bonds (Default value = None).

compute (self, box, ref_points, refValues, points, values, nlist=None)
Calculates the correlation function for the given points. Will overwrite the current histogram.

Parameters

• box (freud.box.Box) – Simulation box.
• ref_points ((N\text{particles}, 3) \text{numpy.ndarray}) – Reference points to calculate the local density.
• refValues ((N\text{particles}) \text{numpy.ndarray}) – Values to use in computation.
• points ((N\text{particles}, 3) \text{numpy.ndarray}) – Points to calculate the bonding.
• values ((N\text{particles}) – Values to use in computation.
• nlist (freud.locality.NeighborList, optional) – NeighborList to use to find bonds (Default value = None).

getBox (self)
Get the box used in the calculations.

Returns freud Box.

Return type freud.box.Box

getCounts (self)
Get the counts of each histogram bin.

Returns Counts of each histogram bin.

Return type (N\text{bins}) \text{numpy.ndarray}

getR (self)
Get The value of bin centers.

Returns Values of bin centers.

Return type (N\text{bins}) \text{numpy.ndarray}
getRDF(self)
Get the RDF.

Returns  Expected (average) product of all values at a given radial distance.
Return type  \((N_{\text{bins}}) \text{ numpy.ndarray}\)

reduceCorrelationFunction(self)
Reduces the histogram in the values over N processors to a single histogram. This is called automatically by \texttt{freud.density.ComplexCF.getRDF()}, \texttt{freud.density.ComplexCF.getCounts()}.

resetCorrelationFunction(self)
Resets the values of the correlation function histogram in memory.

Gaussian Density
class freud.density.GaussianDensity(*args)
Computes the density of a system on a grid.
Replaces particle positions with a Gaussian blur and calculates the contribution from the grid based upon the distance of the grid cell from the center of the Gaussian. The dimensions of the image (grid) are set in the constructor, and can either be set equally for all dimensions or for each dimension independently.

- Constructor Calls:

  - Initialize with all dimensions identical:
    \[
    \texttt{freud.density.GaussianDensity(width, r\_cut, dr)}
    \]

  - Initialize with each dimension specified:
    \[
    \texttt{freud.density.GaussianDensity(width\_x, width\_y, width\_z, r\_cut, dr)}
    \]

Module author: Joshua Anderson <joaander@umich.edu>

Parameters
- \texttt{width} (\texttt{unsigned int}) – Number of pixels to make the image.
- \texttt{width\_x} (\texttt{unsigned int}) – Number of pixels to make the image in x.
- \texttt{width\_y} (\texttt{unsigned int}) – Number of pixels to make the image in y.
- \texttt{width\_z} (\texttt{unsigned int}) – Number of pixels to make the image in z.
- \texttt{r\_cut} (\texttt{float}) – Distance over which to blur.
- \texttt{sigma} (\texttt{float}) – Sigma parameter for Gaussian.

Variables
- \texttt{box} (\texttt{freud.box.Box}) – Box used in the calculation.
- \texttt{gaussian\_density} ((\texttt{w_x, w_y, w_z}) \texttt{numpy.ndarray}) – The image grid with the Gaussian density.
- \texttt{counts} ((\texttt{N\_bins}) \texttt{numpy.ndarray}) – The counts of each histogram bin.
- \texttt{R} ((\texttt{N\_bins}) \texttt{numpy.ndarray}) – The values of bin centers.

calculate (\texttt{self, box, points})
Calculates the Gaussian blur for the specified points. Does not accumulate (will overwrite current image).

Parameters
Local Density

class freud.density.LocalDensity(r_cut, volume, diameter)

Computes the local density around a particle.

The density of the local environment is computed and averaged for a given set of reference points in a sea of data points. Providing the same points calculates them against themselves. Computing the local density results in an array listing the value of the local density around each reference point. Also available is the number of neighbors for each reference point, giving the user the ability to count the number of particles in that region.

The values to compute the local density are set in the constructor. r_cut sets the maximum distance at which to calculate the local density. volume is the volume of a single particle. diameter is the diameter of the circumsphere of an individual particle.

Note: 2D: freud.density.LocalDensity properly handles 2D boxes. The points must be passed in as [x, y, 0]. Failing to set z=0 will lead to undefined behavior.

Parameters

- r_cut (float) – Maximum distance over which to calculate the density.
- volume (float) – Volume of a single particle.
- diameter (float) – Diameter of particle circumsphere.

Variables

- box (freud.box.Box) – Box used in the calculation.
- density ((N_particles) numpy.ndarray) – Density per particle.
- num_neighbors ((N_particles) numpy.ndarray) – Number of neighbors for each particle.

compute (self, box, ref_points, points=None, nlist=None)

Calculated the local density for the specified points. Does not accumulate (will overwrite current data).

Parameters

- box (freud.box.Box) – Simulation box.
freud Documentation, Release 0.9.0

- **ref_points** ((\(N_{\text{particles}}, 3\)) numpy.ndarray) – Reference points to calculate the local density.
- **points** ((\(N_{\text{particles}}, 3\)) numpy.ndarray) – Points to calculate the local density.
- **nlist** ([freud.locality.NeighborList, optional]) – NeighborList to use to find bonds (Default value = None).

**getBox** *(self)*
Get the box used in the calculation.

Returns: freud Box.

Return type: freud.box.Box

**getDensity** *(self)*
Get the density array for each particle.

Returns: Density array for each particle.

Return type: (\(N_{\text{particles}}\)) numpy.ndarray

**getNumNeighbors** *(self)*
Return the number of neighbors for each particle.

Returns: Number of neighbors for each particle.

Return type: (\(N_{\text{particles}}\)) numpy.ndarray

### Radial Distribution Function

class [freud.density.RDF](rmax, dr, rmin=0)
Computes RDF for supplied data.

The RDF \(g(r)\) is computed and averaged for a given set of reference points in a sea of data points. Providing the same points calculates them against themselves. Computing the RDF results in an RDF array listing the value of the RDF at each given \(r\), listed in the \(r\) array.

The values of \(r\) to compute the RDF are set by the values of \(r_{\text{min}}, r_{\text{max}}, \text{dr}\) in the constructor. \(r_{\text{max}}\) sets the maximum distance at which to calculate the \(g(r)\), \(r_{\text{min}}\) sets the minimum distance at which to calculate the \(g(r)\), and \(\text{dr}\) determines the step size for each bin.

*Module author: Eric Harper <harperic@umich.edu>*

**Note:** 2D: [freud.density.RDF](rmax, dr, rmin=0) properly handles 2D boxes. The points must be passed in as \([x, y, 0]\). Failing to set \(z=0\) will lead to undefined behavior.

**Parameters**
- **rmax** *(float)* – Maximum distance to calculate.
- **dr** *(float)* – Distance between histogram bins.
- **rmin** *(float)* – Minimum distance to calculate, default 0.

**Variables**
- **box** ([freud.box.Box]) – Box used in the calculation.
- **RDF** ((\(N_{\text{bins}}\)) numpy.ndarray) – Histogram of RDF values.
- **R** ((\(N_{\text{bins}}, 3\)) numpy.ndarray) – The values of bin centers.
• \texttt{n_r ((N_{bins}, 3) numpy.ndarray)} – Histogram of cumulative RDF values.

Changed in version 0.7.0: Added optional \texttt{rmin} argument.

\textbf{accumulate} (\texttt{self, box, ref_points, points, nlist=None})

Calculates the RDF and adds to the current RDF histogram.

\textbf{Parameters}

• \texttt{box (freud.box.Box)} – Simulation box.

• \texttt{ref_points ((N_{particles}, 3) numpy.ndarray)} – Reference points to calculate the local density.

• \texttt{points ((N_{particles}, 3) numpy.ndarray)} – Points to calculate the bonding.

• \texttt{nlist (freud.locality.NeighborList, optional)} – NeighborList to use to find bonds (Default value = None).

\textbf{compute} (\texttt{self, box, ref_points, points, nlist=None})

Calculates the RDF for the specified points. Will overwrite the current histogram.

\textbf{Parameters}

• \texttt{box (freud.box.Box)} – Simulation box.

• \texttt{ref_points ((N_{particles}, 3) numpy.ndarray)} – Reference points to calculate the local density.

• \texttt{points ((N_{particles}, 3) numpy.ndarray)} – Points to calculate the bonding.

• \texttt{nlist (freud.locality.NeighborList)} – NeighborList to use to find bonds (Default value = None)

\textbf{getBox} (\texttt{self})

Get the box used in the calculation.

\textbf{Returns} freud Box.

\textbf{Return type} \texttt{freud.box.Box}

\textbf{getNr} (\texttt{self})

Get the histogram of cumulative RDF values.

\textbf{Returns} Histogram of cumulative RDF values.

\textbf{Return type} \texttt{(N_{bins}, 3) numpy.ndarray}

\textbf{getR} (\texttt{self})

Get values of the histogram bin centers.

\textbf{Returns} Values of the histogram bin centers.

\textbf{Return type} \texttt{(N_{bins}, 3) numpy.ndarray}

\textbf{getRDF} (\texttt{self})

Histogram of RDF values.

\textbf{Returns} Histogram of RDF values.

\textbf{Return type} \texttt{(N_{bins}, 3) numpy.ndarray}

\textbf{reduceRDF} (\texttt{self})

Reduces the histogram in the values over N processors to a single histogram. This is called automatically by \texttt{freud.density.RDF.getRDF()}, \texttt{freud.density.RDF.getNr()}.
resetRDF \[\text{(self)}\]

Resets the values of RDF in memory.

### 1.3.5 Environment Module

**Overview**

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<th>Function</th>
<th>Description</th>
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<td>\texttt{freud.environment.BondOrder}</td>
<td>Compute the bond order diagram for the system of particles.</td>
</tr>
<tr>
<td>\texttt{freud.environment.LocalDescriptors}</td>
<td>Compute a set of descriptors (a numerical “fingerprint”) of a particle’s local environment.</td>
</tr>
<tr>
<td>\texttt{freud.environment.MatchEnv}</td>
<td>Clusters particles according to whether their local environments match or not, according to various shape matching metrics.</td>
</tr>
<tr>
<td>\texttt{freud.environment.Pairing2D}</td>
<td>Compute pairs for the system of particles.</td>
</tr>
<tr>
<td>\texttt{freud.environment.AngularSeparation}</td>
<td>Calculates the minimum angles of separation between particles and references.</td>
</tr>
</tbody>
</table>

**Details**

The environment module contains functions which characterize the local environments of particles in the system. These methods use the positions and orientations of particles in the local neighborhood of a given particle to characterize the particle environment.

**class \texttt{freud.environment.BondOrder} \[(r_{\text{max}}, k, n, n_{\text{BinsT}}, n_{\text{BinsP}})\]**

Compute the bond order diagram for the system of particles.

Available modes of calculation:

- If \texttt{mode='bod'} (Bond Order Diagram, default): Create the 2D histogram containing the number of bonds formed through the surface of a unit sphere based on the azimuthal \((\theta)\) and polar \((\phi)\) angles.
- If \texttt{mode='lbod'} (Local Bond Order Diagram): Create the 2D histogram containing the number of bonds formed, rotated into the local orientation of the central particle, through the surface of a unit sphere based on the azimuthal \((\theta)\) and polar \((\phi)\) angles.
- If \texttt{mode='obcd'} (Orientation Bond Correlation Diagram): Create the 2D histogram containing the number of bonds formed, rotated by the rotation that takes the orientation of neighboring particle \(j\) to the orientation of each particle \(i\), through the surface of a unit sphere based on the azimuthal \((\theta)\) and polar \((\phi)\) angles.
- If \texttt{mode='oocd'} (Orientation Orientation Correlation Diagram): Create the 2D histogram containing the directors of neighboring particles (\(\hat{z}\) rotated by their quaternion), rotated into the local orientation of the central particle, through the surface of a unit sphere based on the azimuthal \((\theta)\) and polar \((\phi)\) angles.

*Module author: Erin Teich <erteich@umich.edu>*

**Parameters**

- \texttt{r_{\text{max}}} (float) – Distance over which to calculate.
- \texttt{k} (unsigned int) – Order parameter i. To be removed.
- \texttt{n} (unsigned int) – Number of neighbors to find.
- \texttt{n_{\text{bins T}}} (unsigned int) – Number of \(\theta\) bins.
- `n_bins_p(unsigned int)` – Number of \( \phi \) bins.

**Variables**

- `bond_order((N_\phi, N_\theta) numpy.ndarray)` – Bond order.
- `box(freud.box.Box)` – Box used in the calculation.
- `theta((N_\theta) numpy.ndarray)` – The values of bin centers for \( \theta \).
- `phi((N_\phi) numpy.ndarray)` – The values of bin centers for \( \phi \).
- `n_bins_theta(unsigned int)` – The number of bins in the \( \theta \) dimension.
- `n_bins_phi(unsigned int)` – The number of bins in the \( \phi \) dimension.

**accumulate** *(self, box, ref_points, ref_orientations, points, orientations, str mode='bod', nlist=None)*

Calculates the correlation function and adds to the current histogram.

**Parameters**

- `box(freud.box.Box)` – Simulation box.
- `ref_points((N_{particles}, 3) numpy.ndarray)` – Reference points to calculate bonds.
- `ref_orientations((N_{particles}, 4) numpy.ndarray)` – Reference orientations to calculate bonds.
- `points((N_{particles}, 3) numpy.ndarray)` – Points to calculate the bonding.
- `orientations((N_{particles}, 3) numpy.ndarray)` – Orientations to calculate the bonding.
- `mode(str, optional)` – Mode to calculate bond order. Options are 'bod', 'lbod', 'obcd', or 'oocd' (Default value = 'bod').
- `nlist(freud.locality.NeighborList, optional)` – NeighborList to use to find bonds (Default value = None).

**compute** *(self, box, ref_points, ref_orientations, points, orientations, mode='bod', nlist=None)*

Calculates the bond order histogram. Will overwrite the current histogram.

**Parameters**

- `box(freud.box.Box)` – Simulation box.
- `ref_points((N_{particles}, 3) numpy.ndarray)` – Reference points to calculate bonds.
- `ref_orientations((N_{particles}, 4) numpy.ndarray)` – Reference orientations to calculate bonds.
- `points((N_{particles}, 3) numpy.ndarray)` – Points to calculate the bonding.
- `orientations((N_{particles}, 3) numpy.ndarray)` – Orientations to calculate the bonding.
- `mode(str, optional)` – Mode to calculate bond order. Options are 'bod', 'lbod', 'obcd', or 'oocd' (Default value = 'bod').
- `nlist(freud.locality.NeighborList, optional)` – NeighborList to use to find bonds (Default value = None).

**getBondOrder** *(self)*

Get the bond order.

**Returns** Bond order.

**Return type** \((N_\phi, N_\theta) numpy.ndarray\)
**getBox**(self)
Get the box used in the calculation.

Returns freud Box.

Return type `freud.box.Box`

**getNBinsPhi**(self)
Get the number of bins in the \(\phi\)-dimension of histogram.

Returns \(N_\phi\)

Return type `unsigned int`

**getNBinsTheta**(self)
Get the number of bins in the \(\theta\)-dimension of histogram.

Returns \(N_\theta\).

Return type `unsigned int`

**getPhi**(self)
Get \(\phi\).

Returns Values of bin centers for \(\phi\).

Return type \( (N_\phi) \) numpy.ndarray

**getTheta**(self)
Get \(\theta\).

Returns Values of bin centers for \(\theta\).

Return type \( (N_\theta) \) numpy.ndarray

**reduceBondOrder**(self)
Reduces the histogram in the values over N processors to a single histogram. This is called automatically by `freud.environment.BondOrder.getBondOrder()`.

**resetBondOrder**(self)
Resets the values of the bond order in memory.

**class** `freud.environment.LocalDescriptors(box, nNeigh, lmax, rmax)`
Compute a set of descriptors (a numerical “fingerprint”) of a particle’s local environment.

The resulting spherical harmonic array will be a complex-valued array of shape \((\text{num} \_ \text{bonds}, \text{num} \_ \text{sphs})\). Spherical harmonic calculation can be restricted to some number of nearest neighbors through the `num_neighbors` argument; if a particle has more bonds than this number, the last one or more rows of bond spherical harmonics for each particle will not be set.

*Module author: Matthew Spellings <mspells@umich.edu>*

**Parameters**

- **num_neighbors** (`unsigned int`) – Maximum number of neighbors to compute descriptors for.
- **lmax** (`unsigned int`) – Maximum spherical harmonic \(l\) to consider.
- **rmax** (`float`) – Initial guess of the maximum radius to looks for neighbors.
- **negative_m** (`bool`) – True if we should also calculate \(Y_{lm}\) for negative \(m\).

**Variables**

- **sph** ((\(N_{\text{bonds}}, \text{SphWidth}\)) numpy.ndarray) – A reference to the last computed spherical harmonic array.
• **num_particles** (*unsigned int*) – The number of particles.
• **num_neighbors** (*unsigned int*) – The number of neighbors.
• **l_max** (*unsigned int*) – The maximum spherical harmonic $l$ to calculate for.
• **r_max** (*float*) – The cutoff radius.

**compute**(*self, box, unsigned int num_neighbors, points_ref, points=None, orientations=None, mode='neighborhood', nlist=None*)
Calculates the local descriptors of bonds from a set of source points to a set of destination points.

**Parameters**
• **box** (*freud.box.Box*) – Simulation box.
• **num_neighbors** (*unsigned int*) – Number of neighbors to compute with or to limit to, if the neighbor list is precomputed.
• **points_ref** (*((N_{particles}, 3) numpy.ndarray]*) – Source points to calculate the order parameter.
• **points** (*((N_{particles}, 3) numpy.ndarray, optional]*) – Destination points to calculate the order parameter (Default value = None).
• **orientations** (*((N_{particles}, 4) numpy.ndarray, optional]*) – Orientation of each reference point (Default value = None).
• **mode** (*str, optional*) – Orientation mode to use for environments, either 'neighborhood' to use the orientation of the local neighborhood, 'particle_local' to use the given particle orientations, or 'global' to not rotate environments (Default value = 'neighborhood').
• **nlist** (*freud.locality.NeighborList, optional*) – Neighborlist to use to find bonds or 'precomputed' if using **computeNList()** (Default value = None).

**computeNList**(*self, box, points_ref, points=None*)
Compute the neighbor list for bonds from a set of source points to a set of destination points.

**Parameters**
• **box** (*freud.box.Box*) – Simulation box.
• **points_ref** (*((N_{particles}, 3) numpy.ndarray]*) – Source points to calculate the order parameter.
• **points** (*((N_{particles}, 3) numpy.ndarray, optional]*) – Destination points to calculate the order parameter (Default value = None).

**getLMax**(*self*)
Get the maximum spherical harmonic $l$ to calculate for.

**Returns** $l$.

**Return type** unsigned int

**getNP**(*self*)
Get the number of particles.

**Returns** $N_{particles}$.

**Return type** unsigned int

**getNSphs**(*self*)
Get the number of neighbors.

**Returns** $N_{neighbors}$.
Return type unsigned int

getRMax(self)
    Get the cutoff radius.
    Returns r.
    Return type float

getSph(self)
    Get a reference to the last computed spherical harmonic array.
    Returns Order parameter.
    Return type (N_{bonds}, \text{SphWidth}) \text{numpy.ndarray}

class freud.environment.MatchEnv(box, rmax, k)
    Clusters particles according to whether their local environments match or not, according to various shape matching metrics.

    Module author: Erin Teich <erteich@umich.edu>

    Parameters
    • box (\text{freud.box.Box}) – Simulation box.
    • rmax (float) – Cutoff radius for cell list and clustering algorithm. Values near the first minimum of the RDF are recommended.
    • k (unsigned int) – Number of nearest neighbors taken to define the local environment of any given particle.

    Variables
    • tot_environment ((N_{particles}, N_{neighbors}, 3) \text{numpy.ndarray}) – All environments for all particles.
    • num_particles (unsigned int) – The number of particles.
    • num_clusters (unsigned int) – The number of clusters.

cluster(self, points, threshold, hard_r=False, registration=False, global_search=False, env_nlist=None, nlist=None)
    Determine clusters of particles with matching environments.

    Parameters
    • points ((N_{particles}, 3) \text{numpy.ndarray}) – Destination points to calculate the order parameter.
    • threshold (float) – Maximum magnitude of the vector difference between two vectors, below which they are “matching.”
    • hard_r (bool) – If True, add all particles that fall within the threshold of m_rmaxsq to the environment.
    • registration (bool) – If True, first use brute force registration to orient one set of environment vectors with respect to the other set such that it minimizes the RMSD between the two sets.
    • global_search (bool) – If True, do an exhaustive search wherein the environments of every single pair of particles in the simulation are compared. If False, only compare the environments of neighboring particles.
    • env_nlist (\text{freud.locality.NeighborList}, optional) – Neighborlist to use to find the environment of every particle (Default value = None).
• `nlist` (*freud.locality.NeighborList, optional*) – Neighborlist to use to find neighbors of every particle, to compare environments (Default value = None).

**getClusters**(self)
Get a reference to the particles, indexed into clusters according to their matching local environments.

**Returns** Clusters.

**Return type** \((N_{\text{particles}})\) numpy.ndarray

**getEnvironment**(self, i)
Returns the set of vectors defining the environment indexed by i.

**Parameters**
- `i` (unsigned int) – Environment index.

**Returns** The array of vectors.

**Return type** \((N_{\text{neighbors}}, 3)\) numpy.ndarray

**getNP**(self)
Get the number of particles.

**Returns** \(N_{\text{particles}}\).

**Return type** unsigned int

**getNumClusters**(self)
Get the number of clusters.

**Returns** \(N_{\text{clusters}}\).

**Return type** unsigned int

**getTotEnvironment**(self)
Returns the matrix of all environments for all particles.

**Returns** The array of vectors.

**Return type** \((N_{\text{particles}}, N_{\text{neighbors}}, 3)\) numpy.ndarray

**isSimilar**(self, refPoints1, refPoints2, threshold, registration=False)
Test if the motif provided by refPoints1 is similar to the motif provided by refPoints2.

**Parameters**
- `refPoints1` \((N_{\text{particles}}, 3)\) numpy.ndarray – Vectors that make up motif 1.
- `refPoints2` \((N_{\text{particles}}, 3)\) numpy.ndarray – Vectors that make up motif 2.
- `threshold` (float) – Maximum magnitude of the vector difference between two vectors, below which they are considered “matching.”
- `registration` (bool, optional) – If True, first use brute force registration to orient one set of environment vectors with respect to the other set such that it minimizes the RMSD between the two sets (Default value = False).

**Returns** A doublet that gives the rotated (or not) set of refPoints2, and the mapping between the vectors of refPoints1 and refPoints2 that will make them correspond to each other. Empty if they do not correspond to each other.

**Return type** tuple \(((N_{\text{particles}}, 3)\) numpy.ndarray, map[int, int])

**matchMotif**(self, points, refPoints, threshold, registration=False, nlist=None)
Determine clusters of particles that match the motif provided by refPoints.

**Parameters**
• `points` `numpy.ndarray` – Particle positions.

• `refPoints` `numpy.ndarray` – Vectors that make up the motif against which we are matching.

• `threshold` `float` – Maximum magnitude of the vector difference between two vectors, below which they are considered “matching.”

• `registration` `bool`, optional – If True, first use brute force registration to orient one set of environment vectors with respect to the other set such that it minimizes the RMSD between the two sets (Default value = False).

• `nlist` `freud.locality.NeighborList`, optional – Neighborlist to use to find bonds (Default value = None).

`minRMSDMotif` (`self`, `points`, `refPoints`, `registration=False, nlist=None`)  
Rotate (if registration=True) and permute the environments of all particles to minimize their RMSD with respect to the motif provided by refPoints.

**Parameters**

• `points` `numpy.ndarray` – Particle positions.

• `refPoints` `numpy.ndarray` – Vectors that make up the motif against which we are matching.

• `registration` `bool`, optional – If True, first use brute force registration to orient one set of environment vectors with respect to the other set such that it minimizes the RMSD between the two sets (Default value = False).

• `nlist` `freud.locality.NeighborList`, optional – Neighborlist to use to find bonds (Default value = None).

**Returns** Vector of minimal RMSD values, one value per particle.

**Return type** `numpy.ndarray`  

`minimizeRMSD` (`self`, `refPoints1`, `refPoints2`, `registration=False`)  
Get the somewhat-optimal RMSD between the set of vectors refPoints1 and the set of vectors refPoints2.

**Parameters**

• `refPoints1` `numpy.ndarray` – Vectors that make up motif 1.

• `refPoints2` `numpy.ndarray` – Vectors that make up motif 2.

• `registration` `bool`, optional – If true, first use brute force registration to orient one set of environment vectors with respect to the other set such that it minimizes the RMSD between the two sets (Default value = False).

**Returns** A triplet that gives the associated min_rmsd, rotated (or not) set of refPoints2, and the mapping between the vectors of refPoints1 and refPoints2 that somewhat minimizes the RMSD.

**Return type** `tuple`  

`setBox` (`self`, `box`)  
Reset the simulation box.

**Parameters** `box` `freud.box.Box` – Simulation box.

**Note:** Pairing2D is deprecated and is replaced with `Bond Module`. 
class freud.environment.Pairing2D(rmax, k, compDotTol)
Compute pairs for the system of particles.

Module author: Eric Harper <harperic@umich.edu>
Depreciated since version 0.8.2: Use freud.bond instead.

Parameters
- `rmax` (float) – Distance over which to calculate.
- `k` (unsigned int) – Number of neighbors to search.
- `compDotTol` (float) – Value of the dot product below which a pair is determined.

Variables
- `match` ((N\_particles) numpy.ndarray) – The match.
- `pair` ((N\_particles) numpy.ndarray) – The pair.
- `box` (freud.box.Box) – Box used in the calculation.

```
class freud.environment.AngularSeparation(box, rmax, n)
Calculates the minimum angles of separation between particles and references.
```

Module author: Erin Teich
Module author: Andrew Karas

Parameters

- **rmax** *(float)* – Cutoff radius for cell list and clustering algorithm. Values near the first minimum of the RDF are recommended.
- **n** *(int)* – The number of neighbors.

Variables

- **nlist** *(freud.locality.NeighborList)* – The neighbor list.
- **n_p** *(unsigned int)* – The number of particles used in computing the last set.
- **n_ref** *(unsigned int)* – The number of reference particles used in computing the neighbor angles.
- **n_global** *(unsigned int)* – The number of global orientations to check against.

`computeGlobal` *(self, global_ors, ors, equiv_quats)*

Calculates the minimum angles of separation between `global_ors` and `ors`, checking for underlying symmetry as encoded in `equiv_quats`.

Parameters

- **ors** *((N_particles, 3) numpy.ndarray)* – Orientations to calculate the order parameter.
- **global_ors** *(N_particles, 4) numpy.ndarray)* – Reference orientations to calculate the order parameter.
- **equiv_quats** *(N_particles, 4) numpy.ndarray)* – The set of all equivalent quaternions that takes the particle as it is defined to some global reference orientation. Important: `equiv_quats` must include both \(q\) and \(-q\), for all included quaternions.

`computeNeighbor` *(self, box, ref_ors, ors, ref_points, points, equiv_quats, nlist=None)*

Calculates the minimum angles of separation between `ref_ors` and `ors`, checking for underlying symmetry as encoded in `equiv_quats`.

Parameters

- **box** *(freud.box.Box)* – Simulation box.
- **orientations** *(N_particles, 3) numpy.ndarray)* – Orientations to calculate the order parameter.
- **ref_orientations** *(N_particles, 4) numpy.ndarray)* – Reference orientations to calculate the order parameter.
- **ref_points** *(N_particles, 3) numpy.ndarray)* – Reference points to calculate the order parameter.
- **points** *(N_particles, 3) numpy.ndarray)* – Points to calculate the order parameter.
- **nlist** *(freud.locality.NeighborList, optional)* – NeighborList to use to find bonds (Default value = None).
- **equiv_quats** *(N_particles, 4) numpy.ndarray, optional)* – The set of all equivalent quaternions that takes the particle as it is defined to some global reference orientation. Important: `equiv_quats` must include both \(q\) and \(-q\), for all included quaternions.
- **nlist** – Neighborlist to use to find bonds (Default value = None).

`getGlobalAngles` *(self)*

The global angles in radians
Returns Angles in radians.

Return type \( (N_{\text{particles}}, N_{\text{global}}) \) \( \text{numpy.ndarray} \)

getNGlobal (self)
Get the number of global orientations to check against.

Returns \( N_{\text{global orientations}} \).

Return type unsigned int

getNP (self)
Get the number of particles used in computing the last set.

Returns \( N_{\text{particles}} \).

Return type unsigned int

getNReference (self)
Get the number of reference particles used in computing the neighbor angles.

Returns \( N_{\text{particles}} \).

Return type unsigned int

getNeighborAngles (self)
The neighbor angles in radians.

Returns Angles in radians.

Return type \( (N_{\text{reference}}, N_{\text{neighbors}}) \) \( \text{numpy.ndarray} \)

1.3.6 Index Module

Overview

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<td>freud-style indexer for flat arrays.</td>
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Details

The index module exposes the 1-dimensional indexer utilized in freud at the C++ level. At the C++ level, freud utilizes flat arrays to represent multidimensional arrays. \( N \)-dimensional arrays with \( n_i \) elements in each dimension \( i \) are represented as 1-dimensional arrays with \( \prod_{i=1}^{N} n_i \) elements.

class freud.index.Index2D (*args)
freud-style indexer for flat arrays.

Once constructed, the object provides direct access to the flat index equivalent:

- Constructor Calls:
  
  Initialize with all dimensions identical:

  ```python
  freud.index.Index2D(w)
  ```

  Initialize with each dimension specified:

  ```python
  freud.index.Index2D(w, h)
  ```
Note: freud indexes column-first i.e. Index2D(i, j) will return the 1-dimensional index of the $i^{th}$ column and the $j^{th}$ row. This is the opposite of what occurs in a numpy array, in which array[i, j] returns the element in the $i^{th}$ row and the $j^{th}$ column.

Module author: Joshua Anderson <joaander@umich.edu>

Parameters

- w (unsigned int) – Width of 2D array (number of columns).
- h (unsigned int) – Height of 2D array (number of rows).

Variables num_elements (unsigned int) – Number of elements in the array.

Example:

```python
index = Index2D(10)
i = index(3, 5)
```

__call__(self, i, j)

Parameters

- i (unsigned int) – Column index.
- j (unsigned int) – Row index.

Returns Index in flat (e.g. 1-dimensional) array.

Return type unsigned int

gNumElements (self)

Get the number of elements in the array.

Returns Number of elements in the array.

Return type unsigned int

class freud.Index3D (*args)
freud-style indexer for flat arrays.

Once constructed, the object provides direct access to the flat index equivalent:

- Constructor Calls:

  Initialize with all dimensions identical:

  ```python
  freud.Index3D(w)
  ```

  Initialize with each dimension specified:

  ```python
  freud.Index3D(w, h, d)
  ```

Note: freud indexes column-first i.e. Index3D(i, j, k) will return the 1-dimensional index of the $i^{th}$ column, $j^{th}$ row, and the $k^{th}$ frame. This is the opposite of what occurs in a numpy array, in which array[i, j, k] returns the element in the $i^{th}$ frame, $j^{th}$ row, and the $k^{th}$ column.

Module author: Joshua Anderson <joaander@umich.edu>

Parameters
- \( w(\text{unsigned int}) \) – Width of 2D array (number of columns).
- \( h(\text{unsigned int}) \) – Height of 2D array (number of rows).
- \( d(\text{unsigned int}) \) – Depth of 2D array (number of frames).

**Variables**

- \( \text{num\_elements}(\text{unsigned int}) \) – Number of elements in the array.

Example:

```python
index = Index3D(10)
i = index(3, 5, 4)
```

```python
__call__(self, i, j, k)
```

**Parameters**

- \( i(\text{unsigned int}) \) – Column index.
- \( j(\text{unsigned int}) \) – Row index.
- \( k(\text{unsigned int}) \) – Frame index.

**Returns**

Index in flat (e.g. 1-dimensional) array.

**Return type**

unsigned int

```python
getNumElements(self)
```

Get the number of elements in the array.

**Returns**

Number of elements in the array.

**Return type**

unsigned int

### 1.3.7 Interface Module

**Overview**

`freud.interface.InterfaceMeasure`

Measures the interface between two sets of points.

**Details**

The interface module contains functions to measure the interface between sets of points.

```python
class freud.interface.InterfaceMeasure(box, r_cut)
```

Measures the interface between two sets of points.

**Module author:** Matthew Spellings `<mspells@umich.edu>`

**Parameters**

- \( \text{box}(\text{freud.box.Box}) \) – Simulation box.
- \( \text{r\_cut}(\text{float}) \) – Distance to search for particle neighbors.

```python
compute(self, ref\_points, points, nlist=None)
```

Compute and return the number of particles at the interface between the two given sets of points.

**Parameters**

- \( \text{ref\_points}((N_{points}, 3) \text{numpy.ndarray}) \) – One set of particle positions.
• **points** \((N_{\text{particles}}, 3) \text{ numpy.ndarray}\) – Other set of particle positions.

• **nlist** (freud.locality.NeighborList, optional) – Neighborlist to use to find bonds (Default value = None).

### 1.3.8 KSpace Module

**Overview**

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**Details**

Modules for calculating quantities in reciprocal space, including Fourier transforms of shapes and diffraction pattern generation.

**Structure Factor**

**class** freud.kspace.SFactor3DPoints \((box, g)\)

Compute the full 3D structure factor of a given set of points.

Given a set of points \(\vec{r}_i\), SFactor3DPoints computes the static structure factor \(S(\vec{q}) = C_0 \left| \sum_{m=1}^N e^{i \vec{q} \cdot \vec{r}_m} \right|^2\).

In this expression, \(C_0\) is a scaling constant chosen so that \(S(0) = 1\), and \(N\) is the number of particles.

\(S\) is evaluated on a grid of \(q\)-values \(\vec{q} = \frac{2\pi}{L_x} h \hat{i} + \frac{2\pi}{L_y} k \hat{j} + \frac{2\pi}{L_z} l \hat{k}\) for integer \(h, k, l : [-g, g]\) and \(L_x, L_y, L_z\) are the box lengths in each direction.

After calling `compute()`, access the \(q\) values with `getQ()`, the static structure factor values with `getS()`, and (if needed) the un-squared complex version of \(S\) with `getSComplex()`. All values are stored in 3D numpy.ndarray structures. They are indexed by \(a, b, c\) where \(a = h + g, b = k + g, c = l + g\).
Note: Due to the way that numpy arrays are indexed, access the returned S array as \( S[c, b, a]\) to get the value at \( q = (qx[a], qy[b], qz[c])\).

Parameters

- **box** *(freud.box.Box)* – The simulation box.
- **g** *(int)* – The number of grid points for \( q \) in each direction is \( 2g+1 \).

**compute** *(points)*
Compute the static structure factor of a given set of points.

After calling `compute()`, you can access the results with `getS()`, `getSComplex()`, and the grid with `getQ()`.

**Parameters**

- **points** *(\(N\_\text{particles}, 3\) numpy.ndarray)* – Points used to compute the static structure factor.

**getQ()**
Get the \( q \) values at each point.

The structure factor \( S[c, b, a] \) is evaluated at the vector \( q = (qx[a], qy[b], qz[c]) \).

**Returns**

- \( (qx, qy, qz) \).

**Return type**
tuple

**getS()**
Get the computed static structure factor.

**Returns**
The computed static structure factor as a copy.

**Return type**
\( (X, Y) \) numpy.ndarray

**getSComplex()**
Get the computed complex structure factor (if you need the phase information).

**Returns**
The computed static structure factor, as a copy, without taking the magnitude squared.

**Return type**
\( (X, Y) \) numpy.ndarray

class freud.kspace.AnalyzeSFactor3D(S)
Analyze the peaks in a 3D structure factor.

Given a structure factor \( S(q) \) computed by classes such as `SFactor3DPoints`, `AnalyzeSFactor3D` performs a variety of analysis tasks.

- Identifies peaks.
- Provides a list of peaks and the vector \( \vec{q} \) positions at which they occur.
- Provides a list of peaks grouped by \( q^2 \)
- Provides a full list of \( S(|q|) \) values vs \( q^2 \) suitable for plotting the 1D analog of the structure factor.
- Scans through the full 3D peaks and reconstructs the Bravais lattice.

**Note:** All of these operations work in an indexed integer \( q \)-space \( h, k, l \). Any peak position values returned must be multiplied by \( 2\pi/L \) to get to real \( q \) values in simulation units.

**Parameters**

- **S** *(numpy.ndarray)* – Static structure factor to analyze.
**getPeakDegeneracy** (*cut*)

Get a dictionary of peaks indexed by \( q^2 \).

Parameters

- **cut** (*numpy.ndarray*) – All \( S(q) \) values greater than cut will be counted as peaks.

Returns

A dictionary with keys \( q^2 \) and a list of peaks for the corresponding values.

Return type *dict*

**getPeakList** (*cut*)

Get a list of peaks in the structure factor.

Parameters

- **cut** (*float*) – All \( S(q) \) values greater than cut will be counted as peaks.

Returns

peaks, \( q \) as lists.

Return type *list*

**getSvsQ()**

Get a list of all \( S(|q|) \) values vs \( q^2 \).

Returns

S, qsquared.

Return type *numpy.ndarray*

class **freud.kspace.SingleCell3D**(*k*, *ndiv*, *dK*, *boxMatrix*)

SingleCell3D objects manage data structures necessary to call the Fourier Transform functions that evaluate FTs for given form factors at a list of \( K \) points. SingleCell3D provides an interface to helper functions to calculate \( K \) points for a desired grid from the reciprocal lattice vectors calculated from an input boxMatrix. State is maintained as *set_* and *update_* functions invalidate internal data structures and as fresh data is restored with *update_* function calls. This should facilitate management with a higher-level UI such as a GUI with an event queue.

I’m not sure what sort of error checking would be most useful, so I’m mostly allowing ValueErrors and such exceptions to just occur and then propagate up through the calling functions to be dealt with by the user.

Parameters

- **ndiv** (*int*) – The resolution of the diffraction image grid.
- **k** (*float*) – The angular wave number of the plane wave probe (Currently unused).
- **dK** (*float*) – The k-space unit associated with the diffraction image grid spacing.
- **boxMatrix** ((*N* *particles*, 3) *numpy.ndarray*) – The unit cell lattice vectors as columns in a 3x3 matrix.
- **scale** (*float*) – nm per unit length (default 1.0).

Note:

- The *set_* functions take a single parameter and cause other internal data structures to become invalid.
- The *update_* and calculate functions restore the validity of these structures using internal data.
- The functions are separate to make it easier to avoid unnecessary computation such as when changing multiple parameters before seeking output or when wrapping the code with an interface with an event queue.

**add_ptype** (*name*)

Create internal data structures for new particle type by name.

Particle type is inactive when added because parameters must be set before FT can be performed.
Parameters

**name** *(str)* – particle name

**calculate** (*args, **kwargs)*

Calculate FT. The details and arguments will vary depending on the form factor chosen for the particles.

For any particle type-dependent parameters passed as keyword arguments, the parameter must be passed as a list of length \( \max(p_{\text{type}}) + 1 \) with indices corresponding to the particle types defined. In other words, type-dependent parameters are optional (depending on the set of form factors being calculated), but if included must be defined for all particle types.

**get_form_factors** ()

Get form factor names and indices.

**Returns** List of factor names and indices.

**Return type** list

**get_ptypes** ()

Get ordered list of particle names.

**Returns** List of particle names.

**Return type** list

**remove_ptype** *(name)*

Remove internal data structures associated with ptype name.

**Parameters**

- *name* *(str)* – Particle type to remove.

**set_active** *(name)*

Set particle type active.

**Parameters**

- *name* *(str)* – Particle name.

**set_box** *(boxMatrix)*

Set box matrix.

**Parameters**

- *boxMatrix* *(numpy.ndarray)* – Unit cell box matrix.

**set_dK** *(dK)*

Set grid spacing in diffraction image.

**Parameters**

- *dK* *(float)* – Difference in \( K \) vector between two adjacent diffraction image grid points.

**set_form_factor** *(name, ff)*

Set scattering form factor.

**Parameters**

- *name* *(str)* – Particle type name.
- *ff* *(str)* – Scattering form factor named in **get_form_factors**().

**set_inactive** *(name)*

Set particle type inactive.

**Parameters**

- *name* *(str)* – Particle name.

**set_k** *(k)*

Set angular wave number of plane wave probe.

---

**Note:** This shouldn’t usually be necessary, since particle types may be set inactive or have any of their properties updated through **set_** methods.
Parameters $k (\text{float}) - |k_0|$.

`set_ndiv(ndiv)`
Set number of grid divisions in diffraction image.

Parameters `ndiv` (int) – Define diffraction image as `ndiv x ndiv` grid.

`set_param(particle, param, value)`
Set named parameter for named particle.

Parameters

- `particle` (str) – Particle name.
- `param` (str) – Parameter name.
- `value` (float) – Parameter value.

`set_rq(name, position, orientation)`
Set positions and orientations for a particle type.

To best maintain valid state in the event of changing numbers of particles, position and orientation are updated in a single method.

Parameters

- `name` (str) – Particle type name.
- `position` ((N,3) numpy.ndarray) – Array of particle positions.
- `orientation` ((N,4) numpy.ndarray) – Array of particle quaternions.

`set_scale(scale)`
Set scale factor. Store global value and set for each particle type.

Parameters `scale` (float) – nm per unit for input file coordinates.

`update_K_constraint()`
Recalculate constraint used to select $K$ values.

The constraint used is a slab of epsilon thickness in a plane perpendicular to the $k_0$ propagation, intended to provide easy emulation of TEM or relatively high-energy scattering.

`update_Kpoints()`
Update $K$ points at which to evaluate FT.

**Note:** If the diffraction image dimensions change relative to the reciprocal lattice, the $K$ points need to be recalculated.

`update_bases()`
Update the direct and reciprocal space lattice vectors.

**Note:** If scale or boxMatrix is updated, the lattice vectors in direct and reciprocal space need to be recalculated.

**class freud.kspace.FTfactory**
Factory to return an FT object of the requested type.

`addFT(name, constructor, args=None)`
Add an FT class to the factory.

Parameters
• **name** (*str*) – Identifying string to be returned by `getFTlist()`.
• **constructor** (*str*) – Class / function name to be used to create new FT objects.
• **args** (*list*) – Set default argument object to be used to construct FT objects.

`getFTlist()`  
Get an ordered list of named FT types.

    Returns  List of FT names.
    Return type  list

`getFTObject (i, args=None)`  
Get a new instance of an FT type from list returned by `getFTlist()`.

    Parameters
    • **i** (*int*) – Index into list returned by `getFTlist()`.
    • **args** – Argument object used to initialize FT, overriding default set at `addFT()`.

`class freud.kspace.FTbase`  
Base class for FT calculation classes.

    getFT ()
    Return Fourier Transform.

    Returns  Fourier Transform.
    Return type  numpy.ndarray

    get_density (density)
    Get density.

    Returns  Density.
    Return type  numpy.complex64

    get_parambyname (name)
    Get named parameter for object.

    Parameters
    • **name** (*str*) – Parameter name. Must exist in list returned by `get_params()`.

    Returns  Parameter value.
    Return type  float

    get_params ()
    Get the parameter names accessible with `set_parambyname()`.

    Returns  Parameter names.
    Return type  list

    get_scale ()
    Get scale.

    Returns  Scale.
    Return type  float

    set_K (K)
    Set $K$ points to be evaluated.

    Parameters
    • **K** (numpy.ndarray) – List of $K$ vectors at which to evaluate FT.
**set_density**(density)
Set density.

Parameters **density**(numpy.complex64) – Density.

**set_parambyname**(name, value)
Set named parameter for object.

Parameters

- **name**(str) – Parameter name. Must exist in list returned by get_params().
- **value**(float) – Parameter value to set.

**set_rq**(r, q)
Set r, q values.

Parameters

- **r**(float) – r.
- **q**(float) – q.

**set_scale**(scale)
Set scale.

Parameters **scale**(float) – Scale.

---

**class freud.kspace.FTdelta**
Fourier transform a list of delta functions.

**compute**(args, **kwargs)
Compute FT.
Calculate \( S = \sum_{\alpha} e^{-iK \cdot r_{\alpha}} \).

**set_K**(K)
Set K points to be evaluated.

Parameters **K**(numpy.ndarray) – List of K vectors at which to evaluate FT.

**set_density**(density)
Set density.

Parameters **density**(numpy.complex64) – density

**set_rq**(r, q)
Set r, q values.

Parameters

- **r**(float) – r.
- **q**(float) – q.

**set_scale**(scale)
Set scale.

Parameters **scale**(float) – Scale.

---

Note: For a scale factor, \( \lambda \), affecting the scattering density \( \rho(r), S_{\lambda}(k) = \lambda^3 * S(\lambda * k) \)

**class freud.kspace.FTsphere**
Fourier transform for sphere.
Calculate \( S = \sum_{\alpha} e^{-iKr_{\alpha}} \).

**get_radius()**
Get radius parameter.

If appropriate, return value should be scaled by `get_parambyname('scale')` for interpretation.

**Returns** Unscaled radius.

**Return type** float

**set_radius(radius)**
Set radius parameter.

**Parameters**
- **radius** (float) – Sphere radius will be stored as given, but scaled by scale parameter when used by methods.

---

**class freud.kspace.FTpolyhedron**
Fourier Transform for polyhedra.

**compute(** *args, **kwargs)**
Compute FT.

Calculate \( S = \sum_{\alpha} e^{-iKr_{\alpha}} \).

**get_radius()**
Get radius parameter.

If appropriate, return value should be scaled by `get_parambyname('scale')` for interpretation.

**Returns** Unscaled radius.

**Return type** float

**set_K(K)**
Set \( K \) points to be evaluated.

**Parameters**
- **K** (numpy.ndarray) – List of \( K \) vectors at which to evaluate FT.

**set_density(density)**
Set density.

**Parameters**
- **density** (numpy.complex64) – Density.

**set_params** (verts, facets, norms, d, areas, volume)
Construct list of facet offsets.

**Parameters**
- **verts** ((\(N_{\text{particles}}\), 3) numpy.ndarray) – Vertex coordinates.
- **facets** ((\(N_{\text{facets}}\), 3) numpy.ndarray) – Facet vertex indices.
- **norms** ((\(N_{\text{facets}}\), 3) numpy.ndarray) – Facet normals.
- **d** ((\(N_{\text{facets}} - 1\)) numpy.ndarray) – Facet distances.
- **area** ((\(N_{\text{facets}} - 1\)) numpy.ndarray) – Facet areas.
- **volume** (float) – Polyhedron volume.

**set_radius(radius)**
Set radius of in-sphere.

**Parameters**
- **radius** (float) – Radius of inscribed sphere without scale applied.

**set_rq(r, q)**
Set \( r, q \) values.
Parameters

- \( r (\text{float}) = r \).
- \( q (\text{float}) = q \).

class freud.kspace.FTconvexPolyhedron
Fourier Transform for convex polyhedra.

Parameters

- \( \text{hull} (\left( N_{\text{verts}}, 3 \right) \text{numpy.array}) = \text{Convex hull object.} \)

Spoly2D \( (i, k) \)
Calculate Fourier transform of polygon.

Parameters

- \( i (\text{float}) = \text{Face index into self.hull simplex list.} \)
- \( k (\text{numpy.array}) = \text{Angular wave vector at which to calculate } S(i). \)

Spoly3D \( (k) \)
Calculate Fourier transform of polyhedron.

Parameters

- \( k (\text{int}) = \text{Angular wave vector at which to calculate } S(i). \)

compute_py (*args, **kwargs)
Compute FT.

Calculate \( P = F \ast S: \)

- \( S = \sum_{\alpha} e^{-iK \cdot r_{\alpha}}. \)
- \( F \) is the analytical form factor for a polyhedron, computed with \( \text{Spoly3D}(). \)

get_radius()
Get radius parameter.

If appropriate, return value should be scaled by get_parambyname('scale') for interpretation.

Returns

Unscaled radius.

Return type

float

set_radius (radius)
Set radius of in-sphere.

Parameters

- \( \text{radius} (\text{float}) = \text{Radius of inscribed sphere without scale applied.} \)

Diffraction Patterns

class freud.kspace.DeltaSpot
Base class for drawing diffraction spots on a 2D grid.

Based on the dimensions of a grid, determines which grid points need to be modified to represent a diffraction spot and generates the values in that subgrid. Spot is a single pixel at the closest grid point.

Parameters

- \( \text{shape} \) – Number of grid points in each dimension.
- \( \text{extent} \) – Range of x,y values associated with grid points.

get_gridPoints()
Get indices of sub-grid.

Based on the type of spot and its center, return the grid mask of points containing the spot.
**makeSpot** (*cval*)
Generate intensity value(s) at sub-grid points.

**Parameters**
- **cval** (*numpy.complex64*) – Complex valued amplitude used to generate spot intensity.

**set_xy** (*x, y*)
Set *x, y* values of spot center.

**Parameters**
- **x** (*float*) – *x* value of spot center.
- **y** (*float*) – *y* value of spot center.

**class** **freud.kspace.GaussianSpot**
Draw diffraction spot as a Gaussian blur.

Grid points filled according to Gaussian at spot center.

**Parameters**
- **shape** – Number of grid points in each dimension.
- **extent** – Range of *x, y* values associated with grid points.

**makeSpot** (*cval*)
Generate intensity value(s) at sub-grid points.

**Parameters**
- **cval** (*numpy.complex64*) – Complex valued amplitude used to generate spot intensity.

**set_sigma** (*sigma*)
Define Gaussian.

**Parameters**
- **sigma** (*float*) – Width of the Gaussian spot.

**set_xy** (*x, y*)
Set *x, y* values of spot center.

**Parameters**
- **x** (*float*) – *x* value of spot center.
- **y** (*float*) – *y* value of spot center.

**Utilities**

**class** **freud.kspace.Constraint**
Constraint base class.

Base class for constraints on vectors to define the API. All constraints should have a ‘radius’ defining a bounding sphere and a ‘satisfies’ method to determine whether an input vector satisfies the constraint.

**satisfies** (*v*)
Constraint test.

**Parameters**
- **v** – Vector to test against constraint.

**class** **freud.kspace.AlignedBoxConstraint**
Axis-aligned Box constraint.

Tetragonal box aligned with the coordinate system. Consider using a small z dimension to serve as a plane plus or minus some epsilon. Set R < L for a cylinder.
satisfies(v)
  Constraint test.

Parameters  
v – Vector to test against constraint.

freud.kspace.constrainedLatticePoints()
Generate a list of points satisfying a constraint.

Parameters

•  v1 (numpy.ndarray) – Lattice vector 1 along which to test points.
•  v2 (numpy.ndarray) – Lattice vector 2 along which to test points.
•  v3 (numpy.ndarray) – Lattice vector 3 along which to test points.
•  constraint (Constraint) – Constraint object to test lattice points against.

freud.kspace.reciprocalLattice3D()
Calculate reciprocal lattice vectors.

3D reciprocal lattice vectors with magnitude equal to angular wave number.

Parameters

•  a1 (numpy.ndarray) – Real space lattice vector 1.
•  a2 (numpy.ndarray) – Real space lattice vector 2.
•  a3 (numpy.ndarray) – Real space lattice vector 3.

Returns  Reciprocal space vectors.

Return type  list

Note:
For unit test, \( \text{dot}(g[i], a[j]) = 2 \pi \text{ * diracDelta}(i, j) \): list of reciprocal lattice vectors

freud.kspace.meshgrid2(*arrs)
Computes an n-dimensional meshgrid.

source: http://stackoverflow.com/questions/1827489/numpy-meshgrid-in-3d

Parameters  
arrs (list) – Arrays to meshgrid.

Returns  Tuple of arrays.

Return type  tuple

1.3.9 Locality Module

Overview

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<th>Class representing a certain number of “bonds” between particles.</th>
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The locality module contains data structures to efficiently locate points based on their proximity to other points.

**class freud.locality.NeighborList**

Class representing a certain number of “bonds” between particles. Computation methods will iterate over these bonds when searching for neighboring particles.

NeighborList objects are constructed for two sets of position arrays A (alternatively reference points; of length \(n_A\)) and B (alternatively target points; of length \(n_B\)) and hold a set of \((i, j) : i < n_A, j < n_B\) index pairs corresponding to near-neighbor points in A and B, respectively.

For efficiency, all bonds for a particular reference particle \(i\) are contiguous and bonds are stored in order based on reference particle index \(i\). The first bond index corresponding to a given particle can be found in \(\log(n_{bonds})\) time using `find_first_index()`.

*Module author: Matthew Spellings* <mspells@umich.edu>

New in version 0.6.4.

**Note:** Typically, there is no need to instantiate this class directly. In most cases, users should manipulate `freud.locality.NeighborList` objects received from a neighbor search algorithm, such as `freud.locality.LinkCell`, `freud.locality.NearestNeighbors`, or `freud.voronoi.Voronoi`.

### Variables

- **index_i** (`np.ndarray`) – The reference point indices from the last set of points this object was evaluated with. This array is read-only to prevent breakage of `find_first_index()`.
- **index_j** (`np.ndarray`) – The reference point indices from the last set of points this object was evaluated with. This array is read-only to prevent breakage of `find_first_index()`.
- **weights** (`(n_{bonds}) np.ndarray`) – The per-bond weights from the last set of points this object was evaluated with.
- **segments** (`(N_{ref,points}) np.ndarray`) – A segment array, which is an array of length \(N_{ref}\) indicating the first bond index for each reference particle from the last set of points this object was evaluated with.
- **neighbor_counts** (`(N_{ref,points}) np.ndarray`) – A neighbor count array, which is an array of length \(N_{ref}\) indicating the number of neighbors for each reference particle from the last set of points this object was evaluated with.

**Example:**

```python
# Assume we have position as Nx3 array
lc = LinkCell(box, 1.5).compute(box, positions)
nlist = lc.nlist

# Get all vectors from central particles to their neighbors
```

(continues on next page)
rijs = positions[nlist.index_j] - positions[nlist.index_i]
box.wrap(rijs)

**copy** *(self, other=None)*

Create a copy. If other is given, copy its contents into this object. Otherwise, return a copy of this object.

**Parameters**

- **other** *(freud.locality.NeighborList, optional)* – A Neighborlist to copy into this object (Default value = None).

**filter** *(self, filt)*

Removes bonds that satisfy a boolean criterion.

**Parameters**

- **filt** *(np.ndarray)* – Boolean-like array of bonds to keep (True means the bond will not be removed).

**Note:** This method modifies this object in-place.

**Example:**

```python
# Keep only the bonds between particles of type A and type B
nlist.filter(types[nlist.index_i] != types[nlist.index_j])
```

**filter_r** *(self, box, ref_points, points, float rmax, float rmin=0)*

Removes bonds that are outside of a given radius range.

**Parameters**

- **box** *(freud.box.Box)* – Simulation box.
- **ref_points** *(N_particles, 3) numpy.ndarray)* – Reference points to use for filtering.
- **points** *(N_particles, 3) numpy.ndarray)* – Target points to use for filtering.
- **rmax** *(float)* – Maximum bond distance in the resulting neighbor list.
- **rmin** *(float, optional)* – Minimum bond distance in the resulting neighbor list (Default value = 0).

**find_first_index** *(self, unsigned int i)*

Returns the lowest bond index corresponding to a reference particle with an index ≥ i.

**Parameters**

- **i** *(unsigned int)* – The particle index.

**from_arrays** *(type cls, Nref, Ntarget, index_i, index_j, weights=None)*

Create a NeighborList from a set of bond information arrays.

**Parameters**

- **Nref** *(int)* – Number of reference points (corresponding to index_i).
- **Ntarget** *(int)* – Number of target points (corresponding to index_j).
- **index_i** *(np.array)* – Array of integers corresponding to indices in the set of reference points.
- **index_j** *(np.array)* – Array of integers corresponding to indices in the set of target points.
- **weights** *(np.array, optional)* – Array of per-bond weights (if None is given, use a value of 1 for each weight) (Default value = None).
class freud.locality.IteratorLinkCell
Iterates over the particles in a cell.

Module author: Joshua Anderson <joaander@umich.edu>

Example:

```python
# Grab particles in cell 0
for j in linkcell.itercell(0):
    print(positions[j])
```

next(self)
Implements iterator interface

class freud.locality.LinkCell(box, cell_width)
Supports efficiently finding all points in a set within a certain distance from a given point.

Module author: Joshua Anderson <joaander@umich.edu>

Parameters

- **box** (*freud.box.Box*) – Simulation box.
- **cell_width** (*float*) – Maximum distance to find particles within.

Variables

- **box** (*freud.box.Box*) – Simulation box.
- **num_cells** (*unsigned int*) – The number of cells in the box.
- **nlist** (*freud.locality.NeighborList*) – The neighbor list stored by this object, generated by `compute()`.
- **index_i** (*np.ndarray*) – The reference point indices from the last set of points this object was evaluated with. This array is read-only to prevent breakage of `find_first_index()`.
- **index_j** (*np.ndarray*) – The reference point indices from the last set of points this object was evaluated with. This array is read-only to prevent breakage of `find_first_index()`.
- **weights** (*([N_{bonds}] np.ndarray*) – The per-bond weights from the last set of points this object was evaluated with.

Note: 2D: `freud.locality.LinkCell` properly handles 2D boxes. The points must be passed in as `[x, y, 0]`. Failing to set `z=0` will lead to undefined behavior.

Example:

```python
# Assume positions are an Nx3 array
lc = LinkCell(box, 1.5)
lc.computeCellList(box, positions)
for i in range(positions.shape[0]):
    # Cell containing particle i
    cell = lc.getCell(positions[0])
    # List of cell's neighboring cells
    cellNeighbors = lc.getCellNeighbors(cell)
    # Iterate over neighboring cells (including our own)
    for neighborCell in cellNeighbors:
        # Iterate over particles in each neighboring cell
```

(continues on next page)
for neighbor in lc.itercell(neighborCell):
    pass  # Do something with neighbor index

# Using NeighborList API
dens = density.LocalDensity(1.5, 1, 1)
dens.compute(box, positions, nlist=lc.nlist)

compute (self, box, ref_points, points=None, exclude_i=None)
Update the data structure for the given set of points and compute a NeighborList.

Parameters
- **box** (freud.box.Box) – Simulation box.
- **ref_points** ((N_particles, 3) numpy.ndarray) – Reference point coordinates.
- **points** ((N_particles, 3) numpy.ndarray, optional) – Point coordinates (Default value = None).
- **exclude_i** (bool, optional) – True if pairs of points with identical indices should be excluded; if None, is set to True if points is None or the same object as ref_points (Default value = None).

computeCellList (self, box, ref_points, points=None, exclude_i=None)
Update the data structure for the given set of points and compute a NeighborList.

Parameters
- **box** (freud.box.Box) – Simulation box.
- **ref_points** ((N_particles, 3) numpy.ndarray) – Reference point coordinates.
- **points** ((N_particles, 3) numpy.ndarray, optional) – Point coordinates (Default value = None).
- **exclude_i** (bool, optional) – True if pairs of points with identical indices should be excluded; if None, is set to True if points is None or the same object as ref_points (Default value = None).

getBox (self)
Get the freud Box.

Returns freud Box.

Return type freud.box.Box

ggetCell (self, point)
Returns the index of the cell containing the given point.

Parameters **point** ((3) numpy.ndarray) – Point coordinates (x, y, z).

Returns Cell index.

Return type unsigned int

ggetCellNeighbors (self, cell)
Returns the neighboring cell indices of the given cell.

Parameters **cell** (unsigned int) – Cell index.

Returns Array of cell neighbors.

Return type (N_neighbors) numpy.ndarray
**getNumCells** *(self)*  
Get the number of cells in this box.

- **Returns**: The number of cells in this box.  
- **Return type**: unsigned int

**itercell** *(self, unsigned int cell)*  
Return an iterator over all particles in the given cell.

- **Parameters**  
  - **cell** *(unsigned int)* – Cell index.  
- **Returns**: Iterator to particle indices in specified cell.
- **Return type**: iter

**class** **freud.locality.NearestNeighbors** *(rmax, n_neigh, scale=1.1, strict_cut=False)*  
Supports efficiently finding the $N$ nearest neighbors of each point in a set for some fixed integer $N$.

- **strict_cut** == True: $r_{\text{max}}$ will be strictly obeyed, and any particle which has fewer than $N$ neighbors will have values of UINT_MAX assigned.
- **strict_cut** == False (default): $r_{\text{max}}$ will be expanded to find the requested number of neighbors. If $r_{\text{max}}$ increases to the point that a cell list cannot be constructed, a warning will be raised and the neighbors already found will be returned.

*Module author: Eric Harper <harperic@umich.edu>*

**Parameters**

- **rmax** *(float)* – Initial guess of a distance to search within to find $N$ neighbors.
- **n_neigh** *(unsigned int)* – Number of neighbors to find for each point.
- **scale** *(float)* – Multiplier by which to automatically increase $r_{\text{max}}$ value if the requested number of neighbors is not found. Only utilized if **strict_cut** is False. Scale must be greater than 1.
- **strict_cut** *(bool)* – Whether to use a strict $r_{\text{max}}$ or allow for automatic expansion, default is False.

**Variables**

- **UINTMAX** *(unsigned int)* – Value of C++ UINTMAX used to pad the arrays.
- **box** *(freud.box.Box)* – Simulation box.
- **num_neighbors** *(unsigned int)* – The number of neighbors this object will find.
- **n_ref** *(unsigned int)* – The number of particles this object found neighbors of.
- **r_max** *(float)* – Current nearest neighbors search radius guess.
- **r_sq_list** *(([N_particles, N_neighbors] numpy.ndarray)* – The Rsq values list.
- **nlist** *(freud.locality.NeighborList)* – The neighbor list stored by this object, generated by **compute**.  

**Example:**

```python
nn = NearestNeighbors(2, 6)
nn.compute(box, positions, positions)
```

(continues on next page)
hexatic = order.HexOrderParameter(2)
hexatic.compute(box, positions, nlist=nn.nlist)

**compute** *(self, box, ref_points, points=None, exclude_ii=None)*
Update the data structure for the given set of points.

**Parameters**
- **box** *(freud.box.Box)* – Simulation box.
- **ref_points** *((N\text{particles}, 3) \text{numpy.ndarray})* – Reference point coordinates.
- **points** *((N\text{particles}, 3) \text{numpy.ndarray}, optional)* – Point coordinates (Default value = None).
- **exclude_ii** *(bool, optional)* – True if pairs of points with identical indices should be excluded; if None, is set to True if points is None or the same object as ref_points (Default value = None).

**getBox** *(self)*
Get the freud Box.

**Returns** freud Box.

**Return type** freud.box.Box

**getNRef** *(self)*
Get the number of particles this object found neighbors of.

**Returns** The number of particles this object found neighbors of.

**Return type** unsigned int

**getNeighborList** *(self)*
Return the entire neighbor list.

**Returns** Neighbor List.

**Return type** *(N\text{particles}, N\text{neighbors}) \text{numpy.ndarray}*

**getNeighbors** *(self, unsigned int i)*
Return the \(N\) nearest neighbors of the reference point with index \(i\).

**Parameters** **i** *(unsigned int)* – Index of the reference point whose neighbors will be returned.

**getNumNeighbors** *(self)*
The number of neighbors this object will find.

**Returns** The number of neighbors this object will find.

**Return type** unsigned int

**getRMax** *(self)*
Return the current neighbor search distance guess.

**Returns** Nearest neighbors search radius.

**Return type** float

**getRsq** *(self, unsigned int i)*
Return the squared distances to the \(N\) nearest neighbors of the reference point with index \(i\).

**Parameters** **i** *(unsigned int)* – Index of the reference point of which to fetch the neighboring point distances.
Returns Squared distances to the \( N \) nearest neighbors.

Return type \( (N_{\text{particles}}) \text{numpy.ndarray} \)

getRsqList \((self)\)
Return the entire Rsq values list.

Returns Rsq list.

Return type \( (N_{\text{particles}}, N_{\text{neighbors}}) \text{numpy.ndarray} \)

gUINTMAX \((self)\)
Returns Value of C++ UINTMAX used to pad the arrays.

Return type unsigned int

gWrappedVectors \((self)\)
Return the wrapped vectors for computed neighbors. Array padded with -1 for empty neighbors.

Returns Wrapped vectors.

Return type \( (N_{\text{particles}}) \text{numpy.ndarray} \)

setCutMode \((self, \text{strict\_cut})\)
Set mode to handle rmax by Nearest Neighbors.

- \( \text{strict\_cut} == \text{True} \): rmax will be strictly obeyed, and any particle which has fewer than \( N \) neighbors will have values of UINT_MAX assigned.
- \( \text{strict\_cut} == \text{False} \): rmax will be expanded to find the requested number of neighbors. If rmax increases to the point that a cell list cannot be constructed, a warning will be raised and the neighbors already found will be returned.

Parameters \text{strict\_cut} \((\text{bool})\) – Whether to use a strict rmax or allow for automatic expansion.

setRMax \((self, \text{float rmax})\)
Update the neighbor search distance guess.

Parameters \text{rmax} \((\text{float})\) – Nearest neighbors search radius.

### 1.3.10 Order Module

**Overview**

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Continued on next page
The order module contains functions which compute order parameters for the whole system or individual particles. Order parameters take bond order data and interpret it in some way to quantify the degree of order in a system using a scalar value. This is often done through computing spherical harmonics of the bond order diagram, which are the spherical analogue of Fourier Transforms.

```python
class freud.order.CubaticOrderParameter(t_initial, t_final, scale, n_replicates, seed)
```

Compute the cubatic order parameter \([HajiAkbari2015]\) for a system of particles using simulated annealing instead of Newton-Raphson root finding.

*Module author: Eric Harper <harperic@umich.edu>*

**Parameters**

- `t_initial (float)` – Starting temperature.
- `t_final (float)` – Final temperature.
- `scale (float)` – Scaling factor to reduce temperature.
- `n_replicates (unsigned int)` – Number of replicate simulated annealing runs.
- `seed (unsigned int)` – Random seed to use in calculations. If None, system time is used.

**compute (self, orientations)**

Calculates the per-particle and global order parameter.

**Parameters** orientations \((N_{\text{particles}}, 4)\) \(\text{numpy.ndarray}\) – Orientations as angles to use in computation.

**get_cubatic_order_parameter (self)**

Get cubatic order parameter.

**Returns** Cubatic order parameter.

**Return type** float
get_cubatic_tensor(self)
Get cubatic tensor.

Returns  Rank 4 tensor corresponding to cubatic tensor.
Return type  (3, 3, 3, 3) numpy.ndarray

generate_r4_tensor(self)
Get R4 Tensor.

Returns  Rank 4 tensor corresponding to each individual particle orientation.
Return type  (3, 3, 3, 3) numpy.ndarray

generate_global_tensor(self)
Get global tensor.

Returns  Rank 4 tensor corresponding to global orientation.
Return type  (3, 3, 3, 3) numpy.ndarray

generate_orientation(self)
Get orientations.

Returns  Orientation of global orientation.
Return type  (4) numpy.ndarray

generate_particle_op(self)
Get per-particle order parameter.

Returns  Cubatic order parameter.
Return type  np.ndarray

generate_particle_tensor(self)
Get per-particle cubatic tensor.

Returns  Rank 5 tensor corresponding to each individual particle orientation.
Return type  (N\text{particles}, 3, 3, 3, 3) numpy.ndarray

generate_scale(self)
Get scale.

Returns  Value of scale.
Return type  float

generate_t_final(self)
Get final temperature.

Returns  Value of final temperature.
Return type  float

generate_t_initial(self)
Get initial temperature.

Returns  Value of initial temperature.
Return type  float

class freud.order.NematicOrderParameter(u)
Compute the nematic order parameter for a system of particles.

Module author: Jens Glaser <jsglaser@umich.edu>
New in version 0.7.0.

**Parameters**

- **u** 
  ((3) numpy.ndarray) – The nematic director of a single particle in the reference state (without any rotation applied).

**compute**(self, orientations)
Calculates the per-particle and global order parameter.

**Parameters**

- **orientations** 
  ((\(N_{\text{particles}}\), 4) numpy.ndarray) – Orientations to calculate the order parameter.

**get_director**(self)

The director (eigenvector corresponding to the order parameter).

**Returns**
The average nematic director.

**Return type**
(3) numpy.ndarray

**get_nematic_order_parameter**(self)

The nematic order parameter.

**Returns**
Nematic order parameter.

**Return type**
float

**get_nematic_tensor**(self)

The nematic Q tensor.

**Returns**
3x3 matrix corresponding to the average particle orientation.

**Return type**
(3,3) numpy.ndarray

**get_particle_tensor**(self)

The full per-particle tensor of orientation information.

**Returns**
3x3 matrix corresponding to each individual particle orientation.

**Return type**
(\(N_{\text{particles}}\),3,3) numpy.ndarray

**class** freud.order.HexOrderParameter(rmax, k, n)

Calculates the \(k\)-atic order parameter for each particle in the system.

The \(k\)-atic order parameter for a particle \(i\) and its \(n\) neighbors \(j\) is given by:

\[
\psi_k(i) = \frac{1}{n} \sum_j^n e^{k \phi_{ij}}
\]

The parameter \(k\) governs the symmetry of the order parameter while the parameter \(n\) governs the number of neighbors of particle \(i\) to average over. \(\phi_{ij}\) is the angle between the vector \(r_{ij}\) and \((1,0)\).

**Note:** 2D: freud.cluster.Cluster properly handles 2D boxes. The points must be passed in as \([x, y, 0]\). Failing to set \(z=0\) will lead to undefined behavior.

**Module author:** Eric Harper <harperic@umich.edu>

**Parameters**

- **rmax**
  (float) – +/- r distance to search for neighbors.

- **k**
  (unsigned int) – Symmetry of order parameter (\(k = 6\) is hexatic).

- **n**
  (unsigned int) – Number of neighbors (\(n = k\) if \(n\) not specified).

**Variables**

- **psi** 
  ((\(N_{\text{particles}}\)) numpy.ndarray) – Order parameter.
• **box** (*freud.box.Box*) – Box used in the calculation.

• **num_particles** (*unsigned int*) – Number of particles.

• **k** (*unsigned int*) – Symmetry of the order parameter.

**compute** (*self, box, points, nlist=None*)
Calculates the correlation function and adds to the current histogram.

**Parameters**

• **box** (*freud.box.Box*) – Simulation box.

• **points** (*\(N_{\text{particles}}, 3\) numpy.ndarray*) – Points to calculate the order parameter.

• **nlist** (*freud.locality.NeighborList*) – Neighborlist to use to find bonds.

**getBox** (*self*)
Get the box used in the calculation.

**Returns** freud Box.

**Return type** *freud.box.Box*

**getK** (*self*)
Get the symmetry of the order parameter.

**Returns** \(k\).

**Return type** *unsigned int*

**getNP** (*self*)
Get the number of particles.

**Returns** \(N_{\text{particles}}\).

**Return type** *unsigned int*

**getPsi** (*self*)
Get the order parameter.

**Returns** Order parameter.

**Return type** (*\(N_{\text{particles}}\) numpy.ndarray*)

---

class *freud.order.TransOrderParameter* (*rmax, k, n*)
Compute the translational order parameter for each particle.

**Module author:** Michael Engel <engelmm@umich.edu>

**Parameters**

• **rmax** (*float*) – +/- \(r\) distance to search for neighbors.

• **k** (*float*) – Symmetry of order parameter (\(k = 6\) is hexatic).

• **n** (*unsigned int*) – Number of neighbors (\(n = k\) if \(n\) not specified).

**Variables**

• **d_r** (*\(N_{\text{particles}}\) numpy.ndarray*) – Reference to the last computed translational order array.

• **box** (*freud.box.Box*) – Box used in the calculation.

• **num_particles** (*unsigned int*) – Number of particles.

**compute** (*self, box, points, nlist=None*)
Calculates the local descriptors.
Parameters

- **box** (*freud.box.Box*) – Simulation box.
- **points** ([$N_{\text{particles}}$, 3] numpy.ndarray) – Points to calculate the order parameter.
- **nlist** (*freud.locality.NeighborList*) – Neighborlist to use to find bonds.

### getBox (self)

Get the box used in the calculation.

**Returns**  freud.Box.

**Return type**  freud.box.Box

### getDr (self)

Get a reference to the last computed spherical harmonic array.

**Returns**  Order parameter.

**Return type**  ($N_{\text{particles}}$) numpy.ndarray

### getNP (self)

Get the number of particles.

**Returns**  $N_{\text{particles}}$.

**Return type**  unsigned int

---

**class**  *freud.order.LocalQl (box, rmax, l, rmin)*


Implements the local rotationally invariant $Q_l$ order parameter described by Steinhardt. For a particle $i$, we calculate the average $Q_l$ by summing the spherical harmonics between particle $i$ and its neighbors $j$ in a local region:

$$Q_{lm}(i) = \frac{1}{N_b} \sum_{j=1}^{N_b} Y_{lm}(\theta(\hat{r}_{ij}), \phi(\hat{r}_{ij})).$$

This is then combined in a rotationally invariant fashion to remove local orientational order as follows:

$$Q_l(i) = \sqrt{\frac{4\pi}{2l+1} \sum_{m=-l}^{l} |Q_{lm}|^2}.$$

Added first/second shell combined average $Q_l$ order parameter for a set of points:

- Variation of the Steinhardt $Q_l$ order parameter
- For a particle $i$, we calculate the average $Q_l$ by summing the spherical harmonics between particle $i$ and its neighbors $j$ and the neighbors $k$ of neighbor $j$ in a local region.

**Module author:** Xiyu Du `<xiyudu@umich.edu>`

**Module author:** Vyas Ramasubramani `<vramasub@umich.edu>`

---

Parameters

- **box** (*freud.box.Box*) – Simulation box.
- **rmax** (float) – Cutoff radius for the local order parameter. Values near the first minimum of the RDF are recommended.
- **l** (unsigned int) – Spherical harmonic quantum number $l$. Must be a positive number.
- **rmin** (float) – Can look at only the second shell or some arbitrary RDF region.
• **box** (*freud.box.Box*) – Box used in the calculation.

• **num_particles** (*unsigned int*) – Number of particles.

• **Ql** (*[N_{particles}] numpy.ndarray*) – The last computed $Q_l$ for each particle (filled with NaN for particles with no neighbors).

• **ave_Ql** (*[N_{particles}] numpy.ndarray*) – The last computed $\bar{Q}_l$ for each particle (filled with NaN for particles with no neighbors).

• **norm_Ql** (*[N_{particles}] numpy.ndarray*) – The last computed $Q_l$ for each particle normalized by the value over all particles (filled with NaN for particles with no neighbors).

• **ave_norm_Ql** (*[N_{particles}] numpy.ndarray*) – The last computed $\bar{Q}_l$ for each particle normalized by the value over all particles (filled with NaN for particles with no neighbors).

### compute

```python
compute(self, points, nlist=None)
```
Compute the local rotationally invariant $Q_l$ order parameter.

**Parameters**

- **points** (*[N_{particles}, 3] numpy.ndarray*) – Points to calculate the order parameter.
- **nlist** (*freud.locality.NeighborList*, optional) – Neighborlist to use to find bonds (Default value = None).

### computeAve

```python
computeAve(self, points, nlist=None)
```
Compute the local rotationally invariant $Q_l$ order parameter.

**Parameters**

- **points** (*[N_{particles}, 3] numpy.ndarray*) – Points to calculate the order parameter.
- **nlist** (*freud.locality.NeighborList*, optional) – Neighborlist to use to find bonds (Default value = None).

### computeAveNorm

```python
computeAveNorm(self, points, nlist=None)
```
Compute the local rotationally invariant $Q_l$ order parameter.

**Parameters**

- **points** (*[N_{particles}, 3] numpy.ndarray*) – Points to calculate the order parameter.
- **nlist** (*freud.locality.NeighborList*, optional) – Neighborlist to use to find bonds (Default value = None).

### computeNorm

```python
computeNorm(self, points, nlist=None)
```
Compute the local rotationally invariant $Q_l$ order parameter.

**Parameters**

- **points** (*[N_{particles}, 3] numpy.ndarray*) – Points to calculate the order parameter.
- **nlist** (*freud.locality.NeighborList*, optional) – Neighborlist to use to find bonds (Default value = None).

### getAveQl

```python
getAveQl(self)
```
Get a reference to the last computed $Q_l$ for each particle. Returns NaN instead of $Q_l$ for particles with no neighbors.

**Returns** Order parameter.

**Return type** (*[N_{particles}] numpy.ndarray*)

### getBox

```python
getBox(self)
```
Get the box used in the calculation.
Returns `freud.Box`.

**Return type** `freud.box.Box`

### getNP(self)

Get the number of particles.

**Returns** `N_{\text{particles}}`

**Return type** `unsigned int`

### getQl(self)

Get a reference to the last computed $Q_l$ for each particle. Returns NaN instead of $Q_l$ for particles with no neighbors.

**Returns** Order parameter.

**Return type** `(N_{\text{particles}})\text{ numpy.ndarray}`

### getQlAveNorm(self)

Get a reference to the last computed $Q_l$ for each particle. Returns NaN instead of $Q_l$ for particles with no neighbors.

**Returns** Order parameter.

**Return type** `(N_{\text{particles}})\text{ numpy.ndarray}`

### getQlNorm(self)

Get a reference to the last computed $Q_l$ for each particle. Returns NaN instead of $Q_l$ for particles with no neighbors.

**Returns** Order parameter.

**Return type** `(N_{\text{particles}})\text{ numpy.ndarray}`

### setBox(self, box)

Reset the simulation box.

**Parameters**

- `box (freud.box.Box)` – Simulation box.

---

**class** `freud.order.LocalQlNear(box, rmax, l, kn)`


Implements the local rotationally invariant $Q_l$ order parameter described by Steinhardt. For a particle $i$, we calculate the average $Q_l$ by summing the spherical harmonics between particle $i$ and its neighbors $j$ in a local region:

$$Q_{lm}(i) = \frac{1}{N_i} \sum_{j=1}^{N_i} Y_{lm}(\theta(\vec{r}_{ij}), \phi(\vec{r}_{ij}))$$

This is then combined in a rotationally invariant fashion to remove local orientational order as follows:

$$Q_l(i) = \sqrt{\frac{4\pi}{2l+1} \sum_{m=-l}^{l} |Q_{lm}|^2}$$

Added first/second shell combined average $Q_l$ order parameter for a set of points:

- Variation of the Steinhardt $Q_l$ Order parameter.
- For a particle $i$, we calculate the average $Q_l$ by summing the spherical harmonics between particle $i$ and its neighbors $j$ and the neighbors $k$ of neighbor $j$ in a local region.

**Module author:** Xiyu Du `<xiyudu@umich.edu>`

**Module author:** Vyas Ramasubramani `<vramasub@umich.edu>`
Parameters

- **box** (*freud.box.Box*) – Simulation box.
- **rmax** (*float*) – Cutoff radius for the local order parameter. Values near the first minimum of the RDF are recommended.
- **l** (*unsigned int*) – Spherical harmonic quantum number $l$. Must be a positive number.
- **kn** (*unsigned int*) – Number of nearest neighbors. must be a positive integer.

Variables

- **box** (*freud.box.Box*) – Box used in the calculation.
- **num_particles** (*unsigned int*) – Number of particles.
- **Ql** (*numpy.ndarray*) – The last computed $Q_l$ for each particle (filled with NaN for particles with no neighbors).
- **ave_Ql** (*numpy.ndarray*) – The last computed $\bar{Q}_l$ for each particle (filled with NaN for particles with no neighbors).
- **norm_Ql** (*numpy.ndarray*) – The last computed $Q_l$ for each particle normalized by the value over all particles (filled with NaN for particles with no neighbors).
- **ave_norm_Ql** (*numpy.ndarray*) – The last computed $\bar{Q}_l$ for each particle normalized by the value over all particles (filled with NaN for particles with no neighbors).

```python
class LocalWl
```

Implements the local rotationally invariant $W_l$ order parameter described by Steinhardt that can aid in distinguishing between FCC, HCP, and BCC.

Added first/second shell combined average $W_l$ order parameter for a set of points:

```python
class LocalWl
```
• Variation of the Steinhardt $W_l$ order parameter.
• For a particle $i$, we calculate the average $W_l$ by summing the spherical harmonics between particle $i$ and its neighbors $j$ and the neighbors $k$ of neighbor $j$ in a local region.

Module author: Xiyu Du <xiyudu@umich.edu>
Module author: Vyas Ramasubramani <vramasub@umich.edu>

Parameters

• **box** (*freud.box.Box*) – Simulation box.
• **rmax** (*float*) – Cutoff radius for the local order parameter. Values near the first minimum of the RDF are recommended.
• **l** (*unsigned int*) – Spherical harmonic quantum number $l$. Must be a positive number
• **rmin** (*float*) – Lower bound for computing the local order parameter. Allows looking at, for instance, only the second shell, or some other arbitrary RDF region.

Variables

• **box** (*freud.box.Box*) – Box used in the calculation.
• **num_particles** (*unsigned int*) – Number of particles.
• **Ql** (*$N_{\text{particles}}$ numpy.ndarray*) – The last computed $Q_l$ for each particle (filled with NaN for particles with no neighbors).
• **ave_Ql** (*$N_{\text{particles}}$ numpy.ndarray*) – The last computed $\bar{Q}_l$ for each particle (filled with NaN for particles with no neighbors).
• **norm_Ql** (*$N_{\text{particles}}$ numpy.ndarray*) – The last computed $Q_l$ for each particle normalized by the value over all particles (filled with NaN for particles with no neighbors).
• **ave_norm_Ql** (*$N_{\text{particles}}$ numpy.ndarray*) – The last computed $\bar{Q}_l$ for each particle normalized by the value over all particles (filled with NaN for particles with no neighbors).
• **Wl** (*$N_{\text{particles}}$ numpy.ndarray*) – The last computed $W_l$ for each particle (filled with NaN for particles with no neighbors).
• **ave_Wl** (*$N_{\text{particles}}$ numpy.ndarray*) – The last computed $\bar{W}_l$ for each particle (filled with NaN for particles with no neighbors).
• **norm_Wl** (*$N_{\text{particles}}$ numpy.ndarray*) – The last computed $W_l$ for each particle normalized by the value over all particles (filled with NaN for particles with no neighbors).
• **ave_norm_Wl** (*$N_{\text{particles}}$ numpy.ndarray*) – The last computed $\bar{W}_l$ for each particle normalized by the value over all particles (filled with NaN for particles with no neighbors).

**getAveWl**(self)
Get a reference to the last computed $W_l$ for each particle. Returns NaN instead of $W_l$ for particles with no neighbors.

Returns Order parameter.
Return type (*$N_{\text{particles}}$* numpy.ndarray)

**getWl**(self)
Get a reference to the last computed $W_l$ for each particle. Returns NaN instead of $W_l$ for particles with no neighbors.

Returns Order parameter.
Return type (*$N_{\text{particles}}$* numpy.ndarray)
getWlAveNorm (self)
Get a reference to the last computed $W_l$ for each particle. Returns NaN instead of $W_l$ for particles with no neighbors.

**Returns**  Order parameter.

**Return type**  $(N_{\text{particles}})$ numpy.ndarray

getWlNorm (self)
Get a reference to the last computed $W_l$ for each particle. Returns NaN instead of $W_l$ for particles with no neighbors.

**Returns**  Order parameter.

**Return type**  $(N_{\text{particles}})$ numpy.ndarray

class freud.order.LocalWlNear (box, rmax, l, kn)

Implements the local rotationally invariant $W_l$ order parameter described by Steinhardt that can aid in distinguishing between FCC, HCP, and BCC.

Added first/second shell combined average $W_l$ order parameter for a set of points:

- Variation of the Steinhardt $W_l$ order parameter.
- For a particle $i$, we calculate the average $W_l$ by summing the spherical harmonics between particle $i$ and its neighbors $j$ and the neighbors $k$ of neighbor $j$ in a local region.

*Module author: Xiyu Du <siyudu@umich.edu>*

*Module author: Vyas Ramasubramani <vramasub@umich.edu>*

**Parameters**

- **box** (*freud.box.Box*) – Simulation box.
- **rmax** (*float*) – Cutoff radius for the local order parameter. Values near the first minimum of the RDF are recommended.
- **l** (*unsigned int*) – Spherical harmonic quantum number $l$. Must be a positive number
- **kn** (*unsigned int*) – Number of nearest neighbors. Must be a positive number.

**Variables**

- **box** (*freud.box.Box*) – Box used in the calculation.
- **num_particles** (*unsigned int*) – Number of particles.
- **Ql** (*$(N_{\text{particles}})$ numpy.ndarray*) – The last computed $Q_l$ for each particle (filled with NaN for particles with no neighbors).
- **ave_Ql** (*$(N_{\text{particles}})$ numpy.ndarray*) – The last computed $\bar{Q}_l$ for each particle (filled with NaN for particles with no neighbors).
- **norm_Ql** (*$(N_{\text{particles}})$ numpy.ndarray*) – The last computed $Q_l$ for each particle normalized by the value over all particles (filled with NaN for particles with no neighbors).
- **ave_norm_Ql** (*$(N_{\text{particles}})$ numpy.ndarray*) – The last computed $\bar{Q}_l$ for each particle normalized by the value over all particles (filled with NaN for particles with no neighbors).
- **Wl** (*$(N_{\text{particles}})$ numpy.ndarray*) – The last computed $W_l$ for each particle (filled with NaN for particles with no neighbors).
• **ave_Wl** (**(N\_particles\) numpy.ndarray**) – The last computed \(\bar{W}_l\) for each particle (filled with NaN for particles with no neighbors).

• **norm_Wl** (**(N\_particles\) numpy.ndarray**) – The last computed \(W_l\) for each particle normalized by the value over all particles (filled with NaN for particles with no neighbors).

• **ave\_norm_Wl** (**(N\_particles\) numpy.ndarray**) – The last computed \(\bar{W}_l\) for each particle normalized by the value over all particles (filled with NaN for particles with no neighbors).

**computeAve** (**self, points, nlist=None**)

Compute the local rotationally invariant \(Q_l\) order parameter.

**Parameters**

• **points** (**(N\_particles\, 3) numpy.ndarray**) – Points to calculate the order parameter.

• **nlist** (**freud.locality.NeighborList**, optional) – Neighborlist to use to find bonds (Default value = None).

**computeAveNorm** (**self, points, nlist=None**)

Compute the local rotationally invariant \(Q_l\) order parameter.

**Parameters**

• **points** (**(N\_particles\, 3) numpy.ndarray**) – Points to calculate the order parameter.

• **nlist** (**freud.locality.NeighborList**, optional) – Neighborlist to use to find bonds (Default value = None).

**computeNorm** (**self, points, nlist=None**)

Compute the local rotationally invariant \(Q_l\) order parameter.

**Parameters**

• **points** (**(N\_particles\, 3) numpy.ndarray**) – Points to calculate the order parameter.

• **nlist** (**freud.locality.NeighborList**, optional) – Neighborlist to use to find bonds (Default value = None).

class **freud.order.SolLiq** (**box, rmax, Qthreshold, Sthreshold, l**)

**SolLiq**(box, rmax, Qthreshold, Sthreshold, l)

Computes dot products of \(Q_{lm}\) between particles and uses these for clustering.

*Module author: Richmond Newman <newmanrs@umich.edu>*

**Parameters**

• **box** (**freud.box.Box**) – Simulation box.

• **rmax** (**float**) – Cutoff radius for the local order parameter. Values near first minimum of the RDF are recommended.

• **Qthreshold** (**float**) – Value of dot product threshold when evaluating \(Q_{lm}(i)Q_{lm}(j)\) to determine if a neighbor pair is a solid-like bond. (For \(l = 6\), 0.7 generally good for FCC or BCC structures).

• **Sthreshold** (**unsigned int**) – Minimum required number of adjacent solid-link bonds for a particle to be considered solid-like for clustering. (For \(l = 6\), 6-8 is generally good for FCC or BCC structures).

• **l** (**unsigned int**) – Choose spherical harmonic \(Q_l\). Must be positive and even.

**Variables**

• **box** (**freud.box.Box**) – Box used in the calculation.
• **largest_cluster_size** (*unsigned int*) – The largest cluster size. Must call a compute method first.

• **cluster_sizes** (*unsigned int*) – The sizes of all clusters.

• **largest_cluster_size** – The largest cluster size. Must call a compute method first.

• **Ql_mi** (*\(N_{\text{particles}}\) numpy.ndarray*) – The last computed \(Q_{\text{lim}}\) for each particle.

• **clusters** (*\(N_{\text{particles}}\) numpy.ndarray*) – The last computed set of solid-like cluster indices for each particle.

• **num_connections** (*\(N_{\text{particles}}\) numpy.ndarray*) – The number of connections per particle.

• **Ql_dot_ij** (*\(N_{\text{particles}}\) numpy.ndarray*) – Reference to the qldot_ij values.

• **num_particles** (*unsigned int*) – Number of particles.

**compute** *(self, points, nlist=None)*

Compute the local rotationally invariant \(Q_l\) order parameter.

**Parameters**

• **points** (*\(N_{\text{particles}}, 3\) numpy.ndarray*) – Points to calculate the order parameter.

• **nlist** (*freud.locality.NeighborList*, optional) – Neighborlist to use to find bonds (Default value = None).

**computeSolLiqNoNorm** *(self, points, nlist=None)*

Compute the local rotationally invariant \(Q_l\) order parameter.

**Parameters**

• **points** (*\(N_{\text{particles}}, 3\) numpy.ndarray*) – Points to calculate the order parameter.

• **nlist** (*freud.locality.NeighborList*, optional) – Neighborlist to use to find bonds (Default value = None).

**computeSolLiqVariant** *(self, points, nlist=None)*

Compute the local rotationally invariant \(Q_l\) order parameter.

**Parameters**

• **points** (*\(N_{\text{particles}}, 3\) numpy.ndarray*) – Points to calculate the order parameter.

• **nlist** (*freud.locality.NeighborList*, optional) – Neighborlist to use to find bonds (Default value = None).

**getBox** *(self)*

Get the box used in the calculation.

**Returns** fred Box.

**Return type** *fred.box.Box*

**getClusterSizes** *(self)*

Return the sizes of all clusters.

**Returns** The cluster sizes.

**Return type** (*\(N_{\text{clusters}}\) numpy.ndarray*)

**getClusters** *(self)*

Get a reference to the last computed set of solid-like cluster indices for each particle.

**Returns** Clusters.
Return type \((N_{\text{particles}})\) numpy.ndarray

getLargestClusterSize (self)  
Returns the largest cluster size. Must call a compute method first.
  
Returns Largest cluster size.
  
Return type unsigned int

getNP (self)  
Get the number of particles.
  
Returns \(N_p\).
  
Return type unsigned int

getNumberOfConnections (self)  
Get a reference to the number of connections per particle.
  
Returns Clusters.
  
Return type \((N_{\text{particles}})\) numpy.ndarray

getQldot_\_ij (self)  
Get a reference to the qldot_\_ij values.
  
Returns The qldot values.
  
Return type \((N_{\text{clusters}})\) numpy.ndarray

getQlmi (self)  
Get a reference to the last computed \(Q_{lm}\) for each particle.
  
Returns Order parameter.
  
Return type \((N_{\text{particles}})\) numpy.ndarray

setBox (self, box)  
Reset the simulation box.
  
Parameters box (freud.box.Box) – Simulation box.

setClusteringRadius (self, rcutCluster)  
Reset the clustering radius.
  
Parameters rcutCluster (float) – Radius for the cluster finding.

class freud.order.SolLiqNear (box, rmax, Qthreshold, Sthreshold, l)  
SolLiqNear(box, rmax, Qthreshold, Sthreshold, l, kn=12)
Computes dot products of \(Q_{lm}\) between particles and uses these for clustering.

Module author: Richmond Newman <newmanrs@umich.edu>

Parameters
  
  • box (freud.box.Box) – Simulation box.
  
  • rmax (float) – Cutoff radius for the local order parameter. Values near the first minimum of the RDF are recommended.

  • Qthreshold (float) – Value of dot product threshold when evaluating \(Q_{lm}(i)Q_{lm}(j)\) to determine if a neighbor pair is a solid-like bond. (For \(l = 6\), 0.7 generally good for FCC or BCC structures).
• **Sthreshold** (*unsigned int*) – Minimum required number of adjacent solid-link bonds for a particle to be considered solid-like for clustering. (For \( l = 6 \), 6-8 is generally good for FCC or BCC structures).

• **l** (*unsigned int*) – Choose spherical harmonic \( Q_l \). Must be positive and even.

• **kn** (*unsigned int*) – Number of nearest neighbors. Must be a positive number.

**Variables**

• **box** (*freud.box.Box*) – Box used in the calculation.

• **largest_cluster_size** (*unsigned int*) – The largest cluster size. Must call a compute method first.

• **cluster_sizes** (*unsigned int*) – The sizes of all clusters.

• **largest_cluster_size** – The largest cluster size. Must call a compute method first.

• **Ql_mi** (\( (N_{\text{particles}}) \text{numpy.ndarray} \)) – The last computed \( Q_{lm} \) for each particle.

• **clusters** (\( (N_{\text{particles}}) \text{numpy.ndarray} \)) – The last computed set of solid-like cluster indices for each particle.

• **num_connections** (\( (N_{\text{particles}}) \text{numpy.ndarray} \)) – The number of connections per particle.

• **Ql_dot_ij** (\( (N_{\text{particles}}) \text{numpy.ndarray} \)) – Reference to the qldot_ij values.

• **num_particles** (*unsigned int*) – Number of particles.

**compute**(*self, points, nlist=None*)

Compute the local rotationally invariant \( Q_l \) order parameter.

**Parameters**

• **points** (\( (N_{\text{particles}}, 3) \text{numpy.ndarray} \)) – Points to calculate the order parameter.

• **nlist** (*freud.locality.NeighborList*) – Neighborlist to use to find bonds.

**computeSolLiqNoNorm**(*self, points, nlist=None*)

Compute the local rotationally invariant \( Q_l \) order parameter.

**Parameters**

• **points** (\( (N_{\text{particles}}, 3) \text{numpy.ndarray} \)) – Points to calculate the order parameter.

• **nlist** (*freud.locality.NeighborList*) – Neighborlist to use to find bonds.

**computeSolLiqVariant**(*self, points, nlist=None*)

Compute the local rotationally invariant \( Q_l \) order parameter.

**Parameters**

• **points** (\( (N_{\text{particles}}, 3) \text{numpy.ndarray} \)) – Points to calculate the order parameter.

• **nlist** (*freud.locality.NeighborList*) – Neighborlist to use to find bonds.

**Deprecated Classes**

The below functions have all either been deprecated or moved to the *Environment Module* module.
Bond Order

class freud.order.BondOrder(rmax, k, n, nBinsT, nBinsP)

Note: This class is only retained for backwards compatibility. Please use freud.environment.BondOrder instead.

Deprecated since version 0.8.2: Use freud.environment.BondOrder instead.

Local Descriptors

class freud.order.LocalDescriptors(box, nNeigh, lmax, rmax)

Note: This class is only retained for backwards compatibility. Please use freud.environment.LocalDescriptors instead.

Deprecated since version 0.8.2: Use freud.environment.LocalDescriptors instead.

Environment Matching

class freud.order.MatchEnv(box, rmax, k)

Note: This class is only retained for backwards compatibility. Please use freud.environment.MatchEnv instead.

Deprecated since version 0.8.2: Use freud.environment.MatchEnv instead.

Pairing

Note: Pairing2D is deprecated and is replaced with Bond Module.

class freud.order.Pairing2D(rmax, k, compDotTol)

Note: This class is only retained for backwards compatibility. Please use freud.bond instead.

Deprecated since version 0.8.2: Use freud.bond instead.
Angular Separation

class freud.order.AngularSeparation(box, rmax, n)

**Note:** This class is only retained for backwards compatibility. Please use `freud.environment.AngularSeparation` instead.

Deprecated since version 0.8.2: Use `freud.environment.AngularSeparation` instead.

1.3.11 Parallel Module

Overview

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<th>Context manager for managing the number of threads to use.</th>
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<td>freud.parallel.setNumThreads</td>
<td>Set the number of threads for parallel computation.</td>
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Details

The `freud.parallel` module controls the parallelization behavior of freud, determining how many threads the TBB-enabled parts of freud will use. By default, freud tries to use all available threads for parallelization unless directed otherwise, with one exception.

`parallel.setNumThreads(nthreads=None)`
Set the number of threads for parallel computation.

*Module author: Joshua Anderson <joaander@umich.edu>*

**Parameters**
- `nthreads` *(int, optional)* – Number of threads to use. If None (default), use all threads available.

class freud.parallel.NumThreads(N=None)
Context manager for managing the number of threads to use.

*Module author: Joshua Anderson <joaander@umich.edu>*

**Parameters**
- `N` *(int)* – Number of threads to use in this context. Defaults to None, which will use all available threads.

1.3.12 PMFT Module

Overview

<table>
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<td>Computes the PMFT [vanAndersKlotsa2014] [vanAndersAhmed2014] for systems described by coordinates (x, y, \theta) listed in the (x, y, \text{and } t) arrays.</td>
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<table>
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<tr>
<th>Function</th>
<th>Computes the PMFT in coordinates $x, y$ listed in the $x$ and $y$ arrays.</th>
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<tr>
<td><code>freud.pmft.PMFTXY2D</code></td>
<td>[vanAndersKlotsa2014]</td>
</tr>
<tr>
<td><code>freud.pmft.PMFTXYZ</code></td>
<td>[vanAndersKlotsa2014]</td>
</tr>
</tbody>
</table>

Details

The PMFT Module allows for the calculation of the Potential of Mean Force and Torque (PMFT) in a number of different coordinate systems. The PMFT is defined as the negative algorithm of positional correlation function (PCF). A given set of reference points is given around which the PCF is computed and averaged in a sea of data points. The resulting values are accumulated in a PCF array listing the value of the PCF at a discrete set of points. The specific points are determined by the particular coordinate system used to represent the system.

**Note:** The coordinate system in which the calculation is performed is not the same as the coordinate system in which particle positions and orientations should be supplied. Only certain coordinate systems are available for certain particle positions and orientations:

- **2D particle coordinates (position: $[x, y, 0]$, orientation: $\theta$):**
  - $r, \theta_1, \theta_2$.
  - $x, y$.
  - $x, y, \theta$.
- **3D particle coordinates:**
  - $x, y, z$.

**class** `freud.pmft.PMFTR12(r_max, n_r, n_t1, n_t2)`

Computes the PMFT [vanAndersKlotsa2014] [vanAndersAhmed2014] in a 2D system described by $r, \theta_1, \theta_2$.

**Note:** 2D: `freud.pmft.PMFTR12` is only defined for 2D systems. The points must be passed in as $[x, y, 0]$. Failing to set $z=0$ will lead to undefined behavior.

**Module author: Eric Harper <harperic@umich.edu>**

**Module author: Vyas Ramasubramani <vramasub@umich.edu>**

**Parameters**

- `r_max` (*float*) – Maximum distance at which to compute the PMFT.
- `n_r` (*unsigned int*) – Number of bins in $r$.
- `n_t1` (*unsigned int*) – Number of bins in $t_1$.
- `n_t2` (*unsigned int*) – Number of bins in $t_2$.

**Variables**

- `box` (*freud.box.Box*) – Box used in the calculation.
- `bin_counts` (*$[N_r, N_{\theta_2}, N_{\theta_1}]$*) – Bin counts.
freud Documentation, Release 0.9.0

- **PCF** \( (N_r, N_\theta_2, N_\theta_1) \) – The positional correlation function.
- **PMFT** \( (N_r, N_\theta_2, N_\theta_1) \) – The potential of mean force and torque.
- **r_cut** (float) – The cutoff used in the cell list.
- **R** \( (N_r) \text{ numpy.ndarray} \) – The array of r-values for the PCF histogram.
- **T1** \( (N_\theta_1) \text{ numpy.ndarray} \) – The array of T1-values for the PCF histogram.
- **T2** \( (N_\theta_2) \text{ numpy.ndarray} \) – The array of T2-values for the PCF histogram.
- **inverse_jacobian** \( (N_r, N_\theta_2, N_\theta_1) \) – The inverse Jacobian used in the PMFT.
- **n_bins_r** (unsigned int) – The number of bins in the r-dimension of histogram.
- **n_bins_T1** (unsigned int) – The number of bins in the T1-dimension of histogram.
- **n_bins_T2** (unsigned int) – The number of bins in the T2-dimension of histogram.

**accumulate** (self, box, ref_points, ref_orientations, points, orientations, nlist=None)

Calculates the positional correlation function and adds to the current histogram.

**Parameters**

- **box** (freud.box.Box) – Simulation box.
- **ref_points** \( (N_{\text{particles}}, 3) \text{ numpy.ndarray} \) – Reference points to calculate the local density.
- **ref_orientations** \( (N_{\text{particles}}, 4) \text{ numpy.ndarray} \) – Angles of reference points to use in the calculation.
- **points** \( (N_{\text{particles}}, 3) \text{ numpy.ndarray} \) – Points to calculate the local density.
- **orientations** \( (N_{\text{particles}}, 4) \text{ numpy.ndarray} \) – Angles of particles to use in computation.
- **nlist** (freud.locality.NeighborList, optional) – NeighborList to use to find bonds (Default value = None).

**compute** (self, box, ref_points, ref_orientations, points, orientations, nlist=None)

Calculates the positional correlation function for the given points. Will overwrite the current histogram.

**Parameters**

- **box** (freud.box.Box) – Simulation box.
- **ref_points** \( (N_{\text{particles}}, 3) \text{ numpy.ndarray} \) – Reference points to calculate the local density.
- **ref_orientations** \( (N_{\text{particles}}, 4) \text{ numpy.ndarray} \) – Reference orientations as angles to use in computation.
- **points** \( (N_{\text{particles}}, 3) \text{ numpy.ndarray} \) – Points to calculate the local density.
- **orientations** \( (N_{\text{particles}}, 4) \text{ numpy.ndarray} \) – Orientations as angles to use in computation.
- **nlist** (freud.locality.NeighborList, optional) – NeighborList to use to find bonds (Default value = None).

**getBinCounts** (self)

Get the raw bin counts.

**Returns** Bin Counts.

**Return type** \( (N_r, N_\theta_2, N_\theta_1) \text{ numpy.ndarray} \)
```python
getBox(self)
    Get the box used in the calculation.
    Returns freud Box.
    Return type freud.box.Box

getInverseJacobian(self)
    Get the inverse Jacobian used in the PMFT.
    Returns Inverse Jacobian.
    Return type (N_r, N_\theta_2, N_\theta_1) numpy.ndarray

getNBinsR(self)
    Get the number of bins in the r-dimension of histogram.
    Returns N_r.
    Return type unsigned int

getNBinsT1(self)
    Get the number of bins in the T1-dimension of histogram.
    Returns N_\theta_1.
    Return type unsigned int

getNBinsT2(self)
    Get the number of bins in the T2-dimension of histogram.
    Returns N_\theta_2.
    Return type unsigned int

getPCF(self)
    Get the positional correlation function.
    Returns PCF.
    Return type (N_r, N_\theta_2, N_\theta_1) numpy.ndarray

getPMFT(self)
    Get the potential of mean force and torque.
    Returns PMFT.
    Return type (matches PCF) numpy.ndarray

ger(self)
    Get the array of r-values for the PCF histogram.
    Returns Bin centers of r-dimension of histogram.
    Return type (N_r) numpy.ndarray

getRCut(self)
    Get the r_cut value used in the cell list.
    Returns r_cut.
    Return type float

getT1(self)
    Get the array of T1-values for the PCF histogram.
    Returns Bin centers of T1-dimension of histogram.
```
Return type \((N_\theta)\) numpy.ndarray

getT2 (self)
Get the array of T2-values for the PCF histogram.

Returns: Bin centers of T2-dimension of histogram.
Return type \((N_\theta)\) numpy.ndarray

reducePCF (self)
Reduces the histogram in the values over N processors to a single histogram. This is called automatically by freud.pmft.PMFT.PCF().

resetPCF (self)
Resets the values of the PCF histograms in memory.

class freud.pmft.PMFTXYT (x_max, y_max, n_x, n_y, n_t)
Computes the PMFT [vanAndersKlotsa2014] [vanAndersAhmed2014] for systems described by coordinates \(x, y, \theta\) listed in the x, y, and t arrays.
The values of \(x, y, t\) to compute the PCF at are controlled by \(x_{\text{max}}, y_{\text{max}}\) and \(n_{\text{bins}}_{x, y, t}\) parameters to the constructor. The \(x_{\text{max}}\) and \(y_{\text{max}}\) parameters determine the minimum/maximum \(x, y\) values \((\min(\theta) = 0, \max(\theta) = 2\pi)\) at which to compute the PCF and \(n_{\text{bins}}_{x, y, t}\) is the number of bins in \(x, y, t\).

Note: 2D: freud.pmft.PMFTXYT is only defined for 2D systems. The points must be passed in as \([x, y, 0]\). Failing to set \(z=0\) will lead to undefined behavior.

Module author: Eric Harper <harperic@umich.edu>
Module author: Vyas Ramasubramani <vramasub@umich.edu>

Parameters

• \(x_{\text{max}}\) (float) – Maximum x distance at which to compute the PMFT.
• \(y_{\text{max}}\) (float) – Maximum y distance at which to compute the PMFT.
• \(n_x\) (unsigned int) – Number of bins in x.
• \(n_y\) (unsigned int) – Number of bins in y.
• \(n_t\) (unsigned int) – Number of bins in t.

Variables

• \(box\) (freud.box.Box) – Box used in the calculation.
• \(bin\_counts\) \((N_\theta, N_y, N_x)\) numpy.ndarray – Bin counts.
• \(PCF\) \((N_\theta, N_y, N_x)\) numpy.ndarray – The positional correlation function.
• \(PMFT\) \((N_\theta, N_y, N_x)\) numpy.ndarray – The potential of mean force and torque.
• \(r\_cut\) (float) – The cutoff used in the cell list.
• \(X\) \((N_x)\) numpy.ndarray – The array of x-values for the PCF histogram.
• \(Y\) \((N_y)\) numpy.ndarray – The array of y-values for the PCF histogram.
• \(T\) \((N_\theta)\) numpy.ndarray – The array of T-values for the PCF histogram.
• \(jacobian\) (float) – The Jacobian used in the PMFT.
• \(n\_bins\_x\) (unsigned int) – The number of bins in the x-dimension of histogram.
• `n_bins_y` *(unsigned int)* – The number of bins in the y-dimension of histogram.
• `n_bins_T` *(unsigned int)* – The number of bins in the T-dimension of histogram.

**accumulate** *(self, box, ref_points, ref_orientations, points, orientations, nlist=None)*
Calculates the positional correlation function and adds to the current histogram.

**Parameters**

• `box` *(freud.box.Box)* – Simulation box.

• `ref_points` *(((N\_particles, 3) numpy.ndarray)* – Reference points to calculate the local density.

• `ref_orientations` *(((N\_particles, 4) numpy.ndarray)* – Reference orientations as angles to use in computation.

• `points` *(((N\_particles, 3) numpy.ndarray)* – Points to calculate the local density.

• `orientations` *(((N\_particles, 4) numpy.ndarray)* – orientations as angles to use in computation.

• `nlist` *(freud.locality.NeighborList, optional)* – NeighborList to use to find bonds (Default value = None).

**compute** *(self, box, ref_points, ref_orientations, points, orientations, nlist=None)*
Calculates the positional correlation function for the given points. Will overwrite the current histogram.

**Parameters**

• `box` *(freud.box.Box)* – Simulation box.

• `ref_points` *(((N\_particles, 3) numpy.ndarray)* – Reference points to calculate the local density.

• `ref_orientations` *(((N\_particles, 4) numpy.ndarray)* – Reference orientations as angles to use in computation.

• `points` *(((N\_particles, 3) numpy.ndarray)* – Points to calculate the local density.

• `orientations` *(((N\_particles, 4) numpy.ndarray)* – orientations as angles to use in computation.

• `nlist` *(freud.locality.NeighborList, optional)* – NeighborList to use to find bonds (Default value = None).

**getBinCounts** *(self)*
Get the raw bin counts.

**Returns** Bin Counts.

**Return type** *(N\_theta, N\_y, N\_x) numpy.ndarray*

**getBox** *(self)*
Get the box used in the calculation.

**Returns** freud Box.

**Return type** freud.box.Box

**getJacobian** *(self)*
Get the Jacobian used in the PMFT.

**Returns** Jacobian.

**Return type** float
getNBinsT (self)
Get the number of bins in the t-dimension of histogram.

Returns $N_\theta$.
Return type unsigned int

getNBinsX (self)
Get the number of bins in the x-dimension of histogram.

Returns $N_x$.
Return type unsigned int

getNBinsY (self)
Get the number of bins in the y-dimension of histogram.

Returns $N_y$.
Return type unsigned int

getPCF (self)
Get the positional correlation function.

Returns PCF.
Return type $(N_\theta, N_y, N_x)$ numpy.ndarray

getPMFT (self)
Get the potential of mean force and torque.

Returns PMFT.
Return type (matches PCF) numpy.ndarray

getRCut (self)
Get the $r_{cut}$ value used in the cell list.

Returns $r_{cut}$.
Return type float

gET (self)
Get the array of t-values for the PCF histogram.

Returns Bin centers of t-dimension of histogram.
Return type $(N_\theta)$ numpy.ndarray

gETX (self)
Get the array of x-values for the PCF histogram.

Returns Bin centers of x-dimension of histogram.
Return type $(N_x)$ numpy.ndarray

gETY (self)
Get the array of y-values for the PCF histogram.

Returns Bin centers of y-dimension of histogram.
Return type $(N_y)$ numpy.ndarray

reducePCF (self)
Reduces the histogram in the values over N processors to a single histogram. This is called automatically
by freud.pmft.PMFT.PCF().

1.3. Modules
**resetPCF** *(self)*

Resets the values of the PCF histograms in memory.

**class freud.pmft.PMFTXY2D**(x_max, y_max, n_x, n_y)

Computes the PMFT \cite{vanAndersKlotsa2014} \cite{vanAndersAhmed2014} in coordinates \(x, y\) listed in the \(x\) and \(y\) arrays.

The values of \(x\) and \(y\) to compute the PCF at are controlled by \(x_{\text{max}}, y_{\text{max}}, n_x,\) and \(n_y\) parameters to the constructor. The \(x_{\text{max}}\) and \(y_{\text{max}}\) parameters determine the minimum/maximum distance at which to compute the PCF and \(n_x\) and \(n_y\) are the number of bins in \(x\) and \(y\).

**Note:** 2D: **freud.pmft.PMFTXY2D** is only defined for 2D systems. The points must be passed in as \([x, y, 0]\). Failing to set \(z=0\) will lead to undefined behavior.

**Module author:** Eric Harper <harperic@umich.edu>

**Module author:** Vyas Ramasubramani <vramasub@umich.edu>

**Parameters**

- \(x_{\text{max}}\) *(float)* – Maximum \(x\) distance at which to compute the PMFT.
- \(y_{\text{max}}\) *(float)* – Maximum \(y\) distance at which to compute the PMFT.
- \(n_x\) *(unsigned int)* – Number of bins in \(x\).
- \(n_y\) *(unsigned int)* – Number of bins in \(y\).

**Variables**

- **box** *(freud.box.Box)* – Box used in the calculation.
- **bin_counts** *(\(N_y, N_x\)* numpy.ndarray) – Bin counts.
- **PCF** *(\(N_y, N_x\)* numpy.ndarray) – The positional correlation function.
- **PMFT** *(\(N_y, N_x\)* numpy.ndarray) – The potential of mean force and torque.
- **r_cut** *(float)* – The cutoff used in the cell list.
- **X** *(\(N_x\)* numpy.ndarray) – The array of \(x\)-values for the PCF histogram.
- **Y** *(\(N_y\)* numpy.ndarray) – The array of \(y\)-values for the PCF histogram.
- **jacobian** *(float)* – The Jacobian used in the PMFT.
- **n_bins_x** *(unsigned int)* – The number of bins in the \(x\)-dimension of histogram.
- **n_bins_y** *(unsigned int)* – The number of bins in the \(y\)-dimension of histogram.

**accumulate** *(self, box, ref_points, ref_orientations, points, orientations, nlist=None)*

Calculates the positional correlation function and adds to the current histogram.

**Parameters**

- **box** *(freud.box.Box)* – Simulation box.
- **ref_points** *(\(N_{\text{particles}}, 3\)* numpy.ndarray) – Reference points to calculate the local density.
- **ref_orientations** *(\(N_{\text{particles}}, 4\)* numpy.ndarray) – Angles of reference points to use in the calculation.
- **points** *(\(N_{\text{particles}}, 3\)* numpy.ndarray) – Points to calculate the local density.
compute(\texttt{self, box, ref\_points, ref\_orientations, points, orientations, nlist=None})
Calculates the positional correlation function for the given points. Will overwrite the current histogram.

Parameters

- \texttt{box (freud.box.Box)} – Simulation box.
- \texttt{ref\_points (\(N_{\text{particles}}, 3\) \texttt{numpy.ndarray})} – Reference points to calculate the local density.
- \texttt{ref\_orientations (\(N_{\text{particles}}, 4\) \texttt{numpy.ndarray})} – Angles of reference points to use in the calculation.
- \texttt{points (\(N_{\text{particles}}, 3\) \texttt{numpy.ndarray})} – Points to calculate the local density.
- \texttt{orientations (\(N_{\text{particles}}, 4\) \texttt{numpy.ndarray})} – Angles of particles to use in the calculation.
- \texttt{nlist (freud.locality.NeighborList, optional)} – NeighborList to use to find bonds (Default value = None).

getBinCounts(\texttt{self})
Get the raw bin counts (non-normalized).

Returns Bin Counts.
Return type \((N_y, N_x) \texttt{numpy.ndarray}\)

getBox(\texttt{self})
Get the box used in the calculation.

Returns freud Box.
Return type \texttt{freud.box.Box}

getJacobian(\texttt{self})
Get the Jacobian.

Returns Jacobian.
Return type \texttt{float}

getNBinsX(\texttt{self})
Get the number of bins in the x-dimension of histogram.

Returns \(N_x\).
Return type unsigned int

getNBinsY(\texttt{self})
Get the number of bins in the y-dimension of histogram.

Returns \(N_y\).
Return type unsigned int

getPCF(\texttt{self})
Get the positional correlation function.

Returns PCF.
Return type \((N_y, N_x)\) \texttt{numpy.ndarray}

\texttt{getPMFT}(self)
Get the potential of mean force and torque.

Returns PMFT.

Return type (matches PCF) \texttt{numpy.ndarray}

\texttt{getRCut}(self)
Get the \(r_{\text{cut}}\) value used in the cell list.

Returns \(r_{\text{cut}}\).

Return type float

\texttt{getX}(self)
Get the array of x-values for the PCF histogram.

Returns Bin centers of x-dimension of histogram.

Return type \((N_x)\) \texttt{numpy.ndarray}

\texttt{getY}(self)
Get the array of y-values for the PCF histogram.

Returns Bin centers of y-dimension of histogram.

Return type \((N_y)\) \texttt{numpy.ndarray}

\texttt{reducePCF}(self)
Reduces the histogram in the values over \(N\) processors to a single histogram. This is called automatically by \texttt{freud.pmft.PMFT.PCF()}.

\texttt{resetPCF}(self)
Resets the values of the PCF histograms in memory.

\texttt{freud.pmft.PMFTXYZ}(x\_max, y\_max, z\_max, n\_x, n\_y, n\_z)
Computes the PMFT \cite{vanAndersKlotsa2014} \cite{vanAndersAhmed2014} in coordinates \(x, y, z\), listed in the \(x, y, z\) arrays.

The values of \(x, y, z\) to compute the PMF at are controlled by \(x_{\text{max}}, y_{\text{max}}, z_{\text{max}}, n_x, n_y, n_z\) parameters to the constructor. The \(x_{\text{max}}, y_{\text{max}}\), and \(z_{\text{max}}\) parameters determine the minimum/maximum distance at which to compute the PCF and \(n_x, n_y, n_z\) are the number of bins in \(x, y, z\).

Note: 3D: \texttt{freud.pmft.PMFTXYZ} is only defined for 3D systems. The points must be passed in as \([x, y, z]\).

\textit{Module author: Eric Harper <harperic@umich.edu>}

\textit{Module author: Vyas Ramasubramani <vramasub@umich.edu>}

\textbf{Parameters}

- \texttt{x\_max}(float) – Maximum \(x\) distance at which to compute the PMFT.
- \texttt{y\_max}(float) – Maximum \(y\) distance at which to compute the PMFT.
- \texttt{z\_max}(float) – Maximum \(z\) distance at which to compute the PMFT.
- \texttt{n\_x}(unsigned int) – Number of bins in \(x\).
- \texttt{n\_y}(unsigned int) – Number of bins in \(y\).
- \texttt{n\_z}(unsigned int) – Number of bins in \(z\).
• **shiftvec** (*list*) – Vector pointing from [0,0,0] to the center of the PMFT.

**Variables**

- **box** (*freud.box.Box*) – Box used in the calculation.
- **bin_counts** (*((N_z, N_y, N_x) numpy.ndarray*) – Bin counts.
- **PCF** (*((N_z, N_y, N_x) numpy.ndarray*) – The positional correlation function.
- **PMFT** (*((N_z, N_y, N_x) numpy.ndarray*) – The potential of mean force and torque.
- **r_cut** (*float*) – The cutoff used in the cell list.
- **X** (*((N_z) numpy.ndarray*) – The array of x-values for the PCF histogram.
- **Y** (*((N_y) numpy.ndarray*) – The array of y-values for the PCF histogram.
- **Z** (*((N_z) numpy.ndarray*) – The array of z-values for the PCF histogram.
- **jacobian** (*float*) – The Jacobian used in the PMFT.
- **n_bins_x** (*unsigned int*) – The number of bins in the x-dimension of histogram.
- **n_bins_y** (*unsigned int*) – The number of bins in the y-dimension of histogram.
- **n_bins_z** (*unsigned int*) – The number of bins in the z-dimension of histogram.

**accumulate** (*self, box, ref_points, ref_orientations, points, orientations, face_orientations=None, nlist=None*)

Calculates the positional correlation function and adds to the current histogram.

**Parameters**

- **box** (*freud.box.Box*) – Simulation box.
- **ref_points** (*((N_particles, 3) numpy.ndarray*) – Reference points to calculate the local density.
- **ref_orientations** (*((N_particles, 4) numpy.ndarray*) – Angles of reference points to use in the calculation.
- **points** (*((N_particles, 3) numpy.ndarray*) – Points to calculate the local density.
- **orientations** (*((N_particles, 4) numpy.ndarray*) – Angles of particles to use in the calculation.
- **face_orientations** (*((N_particles, 4) numpy.ndarray, optional*) – Orientations of particle faces to account for particle symmetry. If not supplied by user, unit quaternions will be supplied. If a 2D array of shape (*N_f, 4*) or a 3D array of shape (*1, N_f, 4*) is supplied, the supplied quaternions will be broadcast for all particles. (Default value = None).
- **nlist** (*freud.locality.NeighborList, optional*) – NeighborList to use to find bonds (Default value = None).

**compute** (*self, box, ref_points, ref_orientations, points, orientations, face orientations=None, nlist=None*)

Calculates the positional correlation function for the given points. Will overwrite the current histogram.

**Parameters**

- **box** (*freud.box.Box*) – Simulation box.
- **ref_points** (*((N_particles, 3) numpy.ndarray*) – Reference points to calculate the local density.
• `ref_orientations` \((N_{\text{particles}}, 4)\text{ numpy.ndarray}\) – Angles of reference points to use in the calculation.
• `points` \((N_{\text{particles}}, 3)\text{ numpy.ndarray}\) – Points to calculate the local density.
• `orientations` \((N_{\text{particles}}, 4)\text{ numpy.ndarray}\) – Angles of particles to use in the calculation.
• `face_orientations` \((N_{\text{particles}}, 4)\text{ numpy.ndarray}, \text{optional}\) – Orientations of particle faces to account for particle symmetry. If not supplied by user, unit quaternions will be supplied. If a 2D array of shape \((N_f, 4)\) or a 3D array of shape \((1, N_f, 4)\) is supplied, the supplied quaternions will be broadcast for all particles. (Default value = None).
• `nlist` \((\text{freud.locality.NeighborList}, \text{optional})\) – NeighborList to use to find bonds (Default value = None).

`getBinCounts` \((\text{self})\)
Get the raw bin counts.

Returns Bin Counts.

Return type \((N_z, N_y, N_x)\text{ numpy.ndarray}\)

`getBox` \((\text{self})\)
Get the box used in the calculation.

Returns freud Box.

Return type `freud.box.Box`

`getJacobian` \((\text{self})\)
Get the Jacobian.

Returns Jacobian.

Return type `float`

`getNBinsX` \((\text{self})\)
Get the number of bins in the x-dimension of histogram.

Returns \(N_x\).

Return type unsigned int

`getNBinsY` \((\text{self})\)
Get the number of bins in the y-dimension of histogram.

Returns \(N_y\).

Return type unsigned int

`getNBinsZ` \((\text{self})\)
Get the number of bins in the z-dimension of histogram.

Returns \(N_z\).

Return type unsigned int

`getPCF` \((\text{self})\)
Get the positional correlation function.

Returns PCF.

Return type \((N_z, N_y, N_x)\text{ numpy.ndarray}\)
getPMFT (self)
Get the potential of mean force and torque.

Returns PMFT.
Return type \((N_z, N_y, N_x)\) numpy.ndarray

getRCut (self)
Get the \(r_{\text{cut}}\) value used in the cell list.

Returns \(r_{\text{cut}}\).
Return type float

getX (self)
Get the array of x-values for the PCF histogram.

Returns Bin centers of x-dimension of histogram.
Return type \((N_x)\) numpy.ndarray

getY (self)
Get the array of y-values for the PCF histogram.

Returns Bin centers of y-dimension of histogram.
Return type \((N_y)\) numpy.ndarray

getZ (self)
Get the array of z-values for the PCF histogram.

Returns Bin centers of z-dimension of histogram.
Return type \((N_z)\) numpy.ndarray

reducePCF (self)
Reduces the histogram in the values over N processors to a single histogram. This is called automatically by freud.pmft.PMFTXYZ.PCF().

resetPCF (self)
Resets the values of the PCF histograms in memory.

### 1.3.13 Voronoi Module

Overview

<table>
<thead>
<tr>
<th>freud.voronoi.Voronoi</th>
<th>Compute the Voronoi tessellation of a 2D or 3D system using qhull.</th>
</tr>
</thead>
</table>

Details

The voronoi module contains tools to characterize Voronoi cells of a system.

class freud.voronoi.Voronoi (box, buff)
Compute the Voronoi tessellation of a 2D or 3D system using qhull. This uses scipy.spatial.Voronoi, accounting for periodic boundary conditions.

Module author: Benjamin Schultz <baschult@umich.edu>
Module author: Yina Geng <yinageng@umich.edu>

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Since qhull does not support periodic boundary conditions natively, we expand the box to include a portion of the particles’ periodic images. The buffer width is given by the parameter `buff`. The computation of Voronoi tessellations and neighbors is only guaranteed to be correct if `buff >= L/2` where \( L \) is the longest side of the simulation box. For dense systems with particles filling the entire simulation volume, a smaller value for `buff` is acceptable.

**Parameters**

- **box** (`freud.box.Box`) – Simulation box.
- **buff** (`float`) – Buffer width.

**compute**

Compute Voronoi diagram.

**Parameters**

- **positions** (`(N_{\text{particles}}, 3)` `numpy.ndarray`) – Points to calculate Voronoi diagram for.
- **box** (`freud.box.Box`) – Simulation box (Default value = None).
- **buff** (`float`) – Buffer distance within which to look for images (Default value = None).

**computeNeighbors**

Compute the neighbors of each particle based on the Voronoi tessellation. One can include neighbors from multiple Voronoi shells by specifying `numShells` in `getNeighbors()`. An example of computing neighbors from the first two Voronoi shells for a 2D mesh is shown below.

Retrieve the results with `getNeighbors()`.

**Example:**

```python
from freud import box, voronoi
import numpy as np

vor = voronoi.Voronoi(box.Box(5, 5, is2D=True))
pos = np.array([[0, 0, 0], [0, 1, 0], [0, 2, 0],
                [1, 0, 0], [1, 1, 0], [1, 2, 0],
                [2, 0, 0], [2, 1, 0], [2, 2, 0]], dtype=np.float32)
first_shell = vor.computeNeighbors(pos).getNeighbors(1)
second_shell = vor.computeNeighbors(pos).getNeighbors(2)
print('First shell:', first_shell)
print('Second shell:', second_shell)
```

**Note:** Input positions must be a 3D array. For 2D, set the z value to 0.

**Parameters**

- **positions** (`(N_{\text{particles}}, 3)` `numpy.ndarray`) – Points to calculate Voronoi diagram for.
- **box** (`freud.box.Box`) – Simulation box (Default value = None).
- **buff** (`float`) – Buffer distance within which to look for images (Default value = None).
- **exclude_ii** (`bool`, optional) – True if pairs of points with identical indices should be excluded (Default value = True).
**computeVolumes**
Computes volumes (areas in 2D) of Voronoi cells.
New in version 0.8.
Must call `freud.voronoi.Voronoi.compute()` before this method. Retrieve the results with `freud.voronoi.Voronoi.getVolumes()`.

**getBuffer**
Returns the buffer width.

**Returns** Buffer width.

**Return type** float

**getNeighborList**
Returns a neighbor list object.
In the neighbor list, each neighbor pair has a weight value.
In 2D systems, the bond weight is the “ridge length” of the Voronoi boundary line between the neighboring particles.
In 3D systems, the bond weight is the “ridge area” of the Voronoi boundary polygon between the neighboring particles.

**Returns** Neighbor list.

**Return type** NeighborList

**getNeighbors**
Get `numShells` of neighbors for each particle
Must call `computeNeighbors()` before this method.

**Parameters** `numShells` (int) – Number of neighbor shells.

**getVolumes**
Returns an array of volumes (areas in 2D) corresponding to Voronoi cells.
New in version 0.8.
Must call `freud.voronoi.Voronoi.computeVolumes()` before this method.
If the buffer width is too small, then some polytopes may not be closed (they may have a boundary at infinity), and these polytopes’ volumes/areas are excluded from the list.
The length of the list returned by this method should be the same as the array of positions used in the `freud.voronoi.Voronoi.compute()` method, if all the polytopes are closed. Otherwise try using a larger buffer width.

**Returns** Voronoi polytope volumes/areas.

**Return type** `((N_{cells})\times n_{polytopes})` numpy.ndarray

**getVoronoiPolytopes**
Returns a list of polytope vertices corresponding to Voronoi cells.
If the buffer width is too small, then some polytopes may not be closed (they may have a boundary at infinity), and these polytopes’ vertices are excluded from the list.
The length of the list returned by this method should be the same as the array of positions used in the `freud.voronoi.Voronoi.compute()` method, if all the polytopes are closed. Otherwise try using a larger buffer width.

**Returns** List of numpy.ndarray containing Voronoi polytope vertices.
Return type: list

**setBox**
Reset the simulation box.

**Parameters**
- box (*freud.box.Box*) – Simulation box.

**setBufferWidth**
Reset the buffer width.

**Parameters**
- buff (*float*) – Buffer width.

## 1.4 Development Guide

Contributions to freud are highly encouraged. The pages below offer information about freud’s design goals and how to contribute new modules.

### 1.4.1 Design Principles

#### Vision

The freud library is designed to be a powerful and flexible library for the analysis of simulation output. To support a variety of analysis routines, freud places few restrictions on its components. The primary requirement for an analysis routine in freud is that it should be substantially computationally intensive so as to require coding up in C++: **all freud code should be composed of fast C++ routines operating on systems of particles in periodic boxes.** To remain easy-to-use, all C++ modules should be wrapped in Python code so they can be easily accessed from Python scripts or through a Python interpreter.

In order to achieve this goal, freud takes the following viewpoints:

- In order to remain as agnostic to inputs as possible, freud makes no attempt to interface directly with simulation software. Instead, freud works directly with NumPy [http://www.numpy.org/](http://www.numpy.org/) arrays to retain maximum flexibility.
- For ease of maintenance, freud uses Git for version control; Bitbucket for code hosting and issue tracking; and the PEP 8 standard for code, stressing explicitly written code which is easy to read.
- To ensure correctness, freud employs unit testing using the Python unittest framework. In addition, freud utilizes CircleCI for continuous integration to ensure that all of its code works correctly and that any changes or new features do not break existing functionality.

#### Language choices

The freud library is written in two languages: Python and C++. C++ allows for powerful, fast code execution while Python allows for easy, flexible use. Intel Threading Building Blocks parallelism provides further power to C++ code. The C++ code is wrapped with Cython, allowing for user interaction in Python. NumPy provides the basic data structures in freud, which are commonly used in other Python plotting libraries and packages.

#### Unit Tests

All modules should include a set of unit tests which test the correct behavior of the module. These tests should be simple and short, testing a single function each, and completing as quickly as possible (ideally < 10 sec, but times up to a minute are acceptable if justified).
Make Execution Explicit

While it is tempting to make your code do things “automatically”, such as have a calculate method find all `_calc` methods in a class, call them, and add their returns to a dictionary to return to the user, it is preferred in freud to execute code explicitly. This helps avoid issues with debugging and undocumented behavior:

```python
# this is bad
class SomeFreudClass(object):
    def __init__(self, **kwargs):
        for key in kwargs.keys:
            setattr(self, key, kwargs[key])

# this is good
class SomeOtherFreudClass(object):
    def __init__(self, x=None, y=None):
        self.x = x
        self.y = y
```

Code Duplication

When possible, code should not be duplicated. However, being explicit is more important. In freud this translates to many of the inner loops of functions being very similar:

```c
// somewhere deep in function_a
for (int i = 0; i < n; i++)
{
    vec3[3] pos_i = position[i];
    for (int j = 0; j < n; j++)
    {
        pos_j = position[j];
        // more calls here
    }
}

// somewhere deep in function_b
for (int i = 0; i < n; i++)
{
    vec3[3] pos_i = position[i];
    for (int j = 0; j < n; j++)
    {
        pos_j = position[j];
        // more calls here
    }
}
```

While it might be possible to figure out a way to create a base C++ class all such classes inherit from, run through positions, call a calculation, and return, this would be rather complicated. Additionally, any changes to the internals of the code, and may result in performance penalties, difficulty in debugging, etc. As before, being explicit is better.

However, if you have a class which has a number of methods, each of which requires the calling of a function, this function should be written as its own method (instead of being copy-pasted into each method) as is typical in object-oriented programming.
Python vs. Cython vs. C++

The freud library is meant to leverage the power of C++ code imbued with parallel processing power from TBB with the ease of writing Python code. The bulk of your calculations should take place in C++, as shown in the snippet below:

```python
# this is bad
def badHeavyLiftingInPython(positions):
    # check that positions are fine
    for i, pos_i in enumerate(positions):
        for j, pos_j in enumerate(positions):
            if i != j:
                r_ij = pos_j - pos_i
                # ...
                computed_array[i] += some_val
    return computed_array

# this is good
def goodHeavyLiftingInCPlusPlus(positions):
    # check that positions are fine
    cplusplus_heavy_function(computed_array, positions, len(pos))
    return computed_array
```

In the C++ code, implement the heavy lifting function called above from Python:

```c++
void cplusplus_heavy_function(float* computed_array,
                              float* positions,
                              int n)
{
    for (int i = 0; i < n; i++)
    {
        for (int j = 0; j < n; j++)
        {
            if (i != j)
            {
                r_ij = pos_j - pos_i;
                // ...
                computed_array[i] += some_val;
            }
        }
    }
}
```

Some functions may be necessary to write at the Python level due to a Python library not having an equivalent C++ library, complexity of coding, etc. In this case, the code should be written in Cython and a reasonable attempt to optimize the code should be made.

### 1.4.2 Source Code Conventions

The guidelines below should be followed for any new code added to freud. This guide is separated into three sections, one for guidelines common to Python and C++, one for Python alone, and one for C++.
Both

Naming Conventions

The following conventions should apply to Python, Cython, and C++ code.

- Variable names use `lower_case_with_underscores`
- Function and method names use `lowerCaseWithNoUnderscores`
- Class names use `CapWords`

Python example:

```python
class FreudClass(object):
    def __init__(self):
        pass

def calcSomething(self, position_i, orientation_i, position_j, orientation_j):
    r_ij = position_j - position_i
    theta_ij = calcOrientationThing(orientation_i, orientation_j)

def calcOrientationThing(self, orientation_i, orientation_j):
    ...
```

C++ example:

```cpp
class FreudCPPClass
{
    FreudCPPClass()
    {
    }
    computeSomeValue(int variable_a, float variable_b)
    {
        // do some things in here
    }
};
```

Indentation

- Spaces, not tabs, must be used for indentation
- 4 spaces are required per level of indentation and continuation lines
- There should be no whitespace at the end of lines in the file.
- Documentation comments and items broken over multiple lines should be aligned with spaces

```cpp
class SomeClass
{
    private:
        int m_some_member;   // Documentation for some_member
        int m_some_other_member;  // Documentation for some_other_member
    }

template<class BlahBlah> void some_long_func(BlahBlah with_a_really_long_argument_→list,
                                           int b,
                                           int c);
```
Formatting Long Lines

All code lines should be hand-wrapped so that they are no more than 79 characters long. Simply break any excessively long line of code at any natural breaking point to continue on the next line.

```cpp
cout << "This is a really long message, with "
    << message.length() << "Characters in it."
    << message << endl;
```

Try to maintain some element of symmetry in the way the line is broken. For example, the above long message is preferred over the below:

```cpp
cout << "This is a really long message, with " << message.length() << "Characters in it:"
    << message << endl;
```

There are special rules for function definitions and/or calls:

- If the function definition (or call) cleanly fits within the character limit, leave it all on one line

```cpp
int some_function(int arg1, int arg2)
```

- (Option 1) If the function definition (or call) goes over the limit, you may be able to fix it by simply putting the template definition on the previous line:

```cpp
// go from
template<class Foo, class Bar>
int some_really_long_function_name(int with_really_long, Foo argument, Bar lists)
// to
int some_really_long_function_name(int with_really_long, Foo argument, Bar lists)
```

- (Option 2) If the function doesn’t have a template specifier, or splitting at that point isn’t enough, split out each argument onto a separate line and align them.

```cpp
// Instead of this...
int someReallyLongFunctionName(int with_really_long_arguments, int or, int maybe, ...
    float there, char are, int just, float a, int lot, char of, int them)
// ...use this.
int someReallyLongFunctionName(int with_really_long_arguments,
    int or,
    int maybe,
    float there,
    char are,
    int just,
    float a,
    int lot,
    char of,
    int them)
```

Python

Code in freud should follow PEP 8, as well as the following guidelines. Anything listed here takes precedence over PEP 8, but try to deviate as little as possible from PEP 8. When in doubt, follow these guidelines over PEP 8.
During continuous integration (CI), all Python and Cython code in freud is tested with flake8 to ensure PEP 8 compliance. It is strongly recommended to set up a pre-commit hook to ensure code is compliant before pushing to the repository:

```
flake8 --install-hook git
git config --bool flake8.strict true
```

**Source**

- All code should be contained in Cython files
- Python .py files are reserved for module level docstrings and minor miscellaneous tasks for, e.g., backwards compatibility.
- Semicolons should not be used to mark the end of lines in Python.

**Documentation Comments**

- Documentation is generated using sphinx.
- The documentation should be written according to the Google Python Style Guide.
- A few specific notes:
  - The shapes of NumPy arrays should be documented as part of the type in the following manner: `points ((N, 4)) (:py:class:np.ndarray)): The points...`
  - Constructors should be documented at the class level.
  - Class attributes (including properties) should be documented as class attributes within the class-level docstring.
  - Optional arguments should be documented as such within the type after the actual type, and the default value should be included within the description, e.g., `r_max (float, optional): ... If None (the default), number is inferred...`
  - Properties that are settable should be documented the same way as optional arguments: `Lx (float, settable): Length in x.`
- All docstrings should be contained within the Cython files except module docstrings, which belong in the Python code.
- If you copy an existing file as a template, make sure to modify the comments to reflect the new file.
- Good documentation comments are best demonstrated with an in-code example. Liberal addition of examples is encouraged.

**CPP**

**Indentation**

- C++ code should follow Whitesmith’s style. An extended set of examples follows:

```cpp
class SomeClass
{
  public:
```
SomeClass();
    int SomeMethod(int a);
private:
    int m_some_member;
};

// indent function bodies
int SomeClass::SomeMethod(int a)
{
    // indent loop bodies
    while (condition)
    {
        b = a + 1;
        c = b - 2;
    }

    // indent switch bodies and the statements inside each case
    switch (b)
    {
    case 0:
        c = 1;
        break;
    case 1:
        c = 2;
        break;
    default:
        c = 3;
        break;
    }

    // indent the bodies of if statements
    if (something)
    {
        c = 5;
        b = 10;
    }
    else if (something_else)
    {
        c = 10;
        b = 5;
    }
    else
    {
        c = 20;
        b = 6;
    }

    // omitting the braces is fine if there is only one statement in a body (for loops, if, etc.)
    for (int i = 0; i < 10; i++)
    {
        c = c + 1;
    }

    return c;
    // the nice thing about this style is that every brace lines up perfectly with its mate
}
• TBB sections should use lambdas, not templates

```cpp
void someC++Function(float some_var,
    float other_var)
{
    // code before parallel section
    parallel_for(blocked_range<size_t>({0,n}),
        [=] (const blocked_range<size_t>& r)
        {
            // do stuff
        });
}
```

**Documentation Comments**

• Documentation should be written in doxygen.

### 1.4.3 How to Add New Code

This document details the process of adding new code into freud.

**Does my code belong in freud?**

The freud library is not meant to simply wrap or augment external Python libraries. A good rule of thumb is if the code I plan to write does not require C++, it does not belong in freud. There are, of course, exceptions.

**Create a new branch**

You should branch your code from master into a new branch. Do not add new code directly into the master branch.

**Add a New Module**

If the code you are adding is in a new module, not an existing module, you must do the following:

• Edit `cpp/CMakeLists.txt`
  - Add `$CMAKE_CURRENT_SOURCE_DIR/moduleName` to `include_directories`.
  - Add `moduleName/SubModule.cc` and `moduleName/SubModule.h` to the `FREUD_SOURCES` in `set`.

• Create `cpp/moduleName` folder

• Edit `freud/__init__.py`
  - Add `from . import moduleName` so that your module is imported by default.

• Edit `freud/_freud.pyx`
  - Add `include "moduleName.pxi"`. This must be done to have freud include your Python-level code.

• Create `freud/moduleName.pxi` file
  - This will house the python-level code.
  - If you have a `.pxd` file exposing C++ classes, make sure to import that:
cimport freud._moduleName as moduleName

- Create freud/moduleName.py file
  - Make sure there is an import for each C++ class in your module:

```python
from ._freud import MyC++Class
```

- Create freud/_moduleName.pxd
  - This file will expose the C++ classes in your module to python.
- Add line to doc/source/modules.rst
  - Make sure your new module is referenced in the documentation.
- Create doc/source/moduleName.rst

**Add to an Existing Module**

To add a new class to an existing module, do the following:

- Create cpp/moduleName/SubModule.h and cpp/moduleName/SubModule.cc
  - New classes should be grouped into paired .h, .cc files. There may be a few instances where new classes could be added to an existing .h, .cc pairing.
- Edit freud/moduleName.py file
  - Add a line for each C++ class in your module:

```python
from ._freud import MyC++Class
```

- Expose C++ class in freud/_moduleName.pxd
- Create Python interface in freud/moduleName.pxi

You must include sphinx-style documentation and unit tests.

- Add extra documentation to doc/source/moduleName.rst
- Add unit tests to freud/tests

**1.5 References and Citations**

**1.6 License**

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1.7 Credits

1.7.1 freud Developers

The following people contributed to the development of freud.

Eric Harper, University of Michigan - Former lead developer
  • TBB parallelism
  • PMFT module
  • NearestNeighbors
  • RDF
  • Bonding module
  • Cubatic order parameter
  • Hexatic order parameter
  • Pairing2D

Joshua A. Anderson, University of Michigan - Creator
  • Initial design and implementation
  • IteratorLinkCell
  • LinkCell
  • Various density modules
  • freud.parallel
  • Indexing modules
• cluster.pxi

Matthew Spellings - Former lead developer
• Added generic neighbor list
• Enabled neighbor list usage across freud modules
• Correlation functions
• LocalDescriptors class
• interface.pxi

Erin Teich
• Wrote environment matching module
• BondOrder (with Julia Dshemuchadse)
• Angular separation (with Andrew Karas)
• Contributed to LocalQl development

13. Eric Irrgang
• Authored kspace CPP code

Chrisy Du
• Authored all Steinhardt order parameters

Antonio Osorio

Vyas Ramasubramani - Lead developer
• Ensured pep8 compliance
• Added CircleCI continuous integration support
• Rewrote docs
• Fixed nematic order parameter
• Add properties for accessing class members
• Various minor bug fixes
• Refactored PMFT code
• Refactored Steinhardt order parameter code

Bradley Dice - Lead developer
• Cleaned up various docstrings
• HexOrderParameter bug fixes
• Cleaned up testing code
• Bumpversion support
• Reduced all compile warnings
• Added Python interface for box periodicity
• Added Voronoi support for neighbor lists across periodic boundaries
• Added Voronoi weights for 3D
• Added Voronoi cell volume computation
Richmond Newman
  • Developed the freud box
  • Solid liquid order parameter
Carl Simon Adorf
  • Developed the python box module
Jens Glaser
  • Wrote kspace.pxi front-end
  • Nematic order parameter
Benjamin Schultz
  • Wrote Voronoi module
Bryan VanSaders
Ryan Marson
Tom Grubb
Yina Geng
  • Co-wrote Voronoi neighbor list module
  • Add properties for accessing class members
Carolyn Phillips
  • Initial design and implementation
  • Package name
Ben Swerdlow
James Antonaglia
Mayank Agrawal
  • Co-wrote Voronoi neighbor list module
William Zygmunt
Greg van Anders
James Proctor
Rose Cersonsky
Wenbo Shen
Andrew Karas
  • Angular separation
Paul Dodd
Tim Moore
  • Added optional rmin argument to density.RDF
Michael Engel
  • Translational order parameter
Alex Dutton
BiMap class for MatchEnv

1.7.2 Source code

Eigen (http://eigen.tuxfamily.org/) is included as a git submodule in freud. Eigen is made available under the Mozilla Public License v.2.0 (http://mozilla.org/MPL/2.0/). Its linear algebra routines are used for various tasks including the computation of eigenvalues and eigenvectors.

fsph (https://bitbucket.org/glotzer/fsph) is included as a git submodule in freud. fsph is made available under the MIT license. It is used for the calculation of spherical harmonics, which are then used in the calculation of various order parameters, under the following license:

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Support and Contribution

Please visit our repository on Bitbucket for the library source code. Any issues or bugs may be reported at our issue tracker, while questions and discussion can be directed to our forum. All contributions to freud are welcomed via pull requests! Please see the development guide for more information on requirements for new code.
CHAPTER 3

Indices and tables

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- search


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