1 Documentation 3
Python Module Index 75
Index 77

Contents
dwave-system is a basic API for easily incorporating the D-Wave system as a sampler in the D-Wave Ocean software stack, directly or through Leap’s cloud-based hybrid solvers. It includes DWaveSampler, a dimod sampler that accepts and passes system parameters such as system identification and authentication down the stack, and LeapHybridSampler, for Leap’s hybrid solvers. It also includes several useful composites—layers of pre- and post-processing—that can be used with DWaveSampler to handle minor-embedding, optimize chain strength, etc.
CHAPTER 1

Documentation

Note: This documentation is for the latest version of dwave-system. Documentation for the version currently installed by dwave-ocean-sdk is here: dwave-system.

1.1 Introduction

dwave-system enables easy incorporation of the D-Wave system as a sampler in either a hybrid quantum-classical solution, using LeapHybridSampler() or dwave-hybrid samplers such as KerberosSampler, or directly using DWaveSampler().

Note: For applications that require detailed control on communication with the remote compute resource (a D-Wave QPU or Leap’s hybrid solvers), see dwave-cloud-client.

D-Wave System Documentation describes D-Wave quantum computers and Leap hybrid solvers, including features, parameters, and properties. It also provides guidance on programming the D-Wave system, including how to formulate problems and configure parameters.

1.1.1 Example

This example solves a small example of a known graph problem, minimum vertex cover. It uses the NetworkX graphic package to create the problem, Ocean’s dwave_networkx to formulate the graph problem as a BQM, and dwave-system’s DWaveSampler() to use a D-Wave system as the sampler. dwave-system’s EmbeddingComposite() handles mapping between the problem graph to the D-Wave system’s numerically indexed qubits, a mapping known as minor-embedding.

```python
>>> import networkx as nx
>>> import dwave_networkx as dnx
(continues on next page)
```
>>> from dwave.system import DWaveSampler, EmbeddingComposite
...
>>> s5 = nx.star_graph(4)  # a star graph where node 0 is hub to four other nodes
>>> sampler = EmbeddingComposite(DWaveSampler())
>>> print(dnx.min_vertex_cover(s5, sampler))
[0]

1.2 Reference Documentation

1.2.1 Samplers

A sampler accepts a binary quadratic model (BQM) and returns variable assignments. Samplers generally try to find minimizing values but can also sample from distributions defined by the BQM.

Contents

- Samplers
  - DWaveSampler
  - DWaveCliqueSampler
  - LeapHybridSampler
  - LeapHybridDQMSampler

DWaveSampler

class DWaveSampler (failover=False, retry_interval=-1, **config)
A class for using the D-Wave system as a sampler.

Uses parameters set in a configuration file, as environment variables, or explicitly as input arguments for selecting and communicating with a D-Wave system. For more information, see D-Wave Cloud Client.

Inherits from dimod.Sampler and dimod.Structured.

Parameters

- failover (bool, optional, default=False) – Switch to a new QPU in the rare event that the currently connected system goes offline. Note that different QPUs may have different hardware graphs and a failover will result in a regenerated nodelist, edgelist, properties and parameters.

- retry_interval (number, optional, default=-1) – The amount of time (in seconds) to wait to poll for a solver in the case that no solver is found. If retry_interval is negative then it will instead propagate the SolverNotFoundError to the user.

- **config – Keyword arguments passed to dwave.cloud.client.Client. from_config()

Note: Prior to version 1.0.0, DWaveSampler used the base client, allowing non-QPU solvers to be selected. To reproduce the old behavior, instantiate DWaveSampler with client='base'.

Chapter 1. Documentation
Examples

This example submits a two-variable Ising problem mapped directly to two adjacent qubits on a D-Wave system. `qubit_a` is the first qubit in the QPU’s indexed list of qubits and `qubit_b` is one of the qubits coupled to it. Other required parameters for communication with the system, such as its URL and an authentication token, are implicitly set in a configuration file or as environment variables, as described in Configuring Access to D-Wave Solvers. Given sufficient reads (here 100), the quantum computer should return the best solution, 1, \(-1\) on `qubit_a` and `qubit_b`, respectively, as its first sample (samples are ordered from lowest energy).

```python
>>> from dwave.system import DWaveSampler
... >>> sampler = DWaveSampler()
... >>> qubit_a = sampler.nodelist[0]
... >>> qubit_b = next(iter(sampler.adjacency[qubit_a]))
... >>> sampleset = sampler.sample_ising({qubit_a: -1, qubit_b: 1},
... ... {},
... ... num_reads=100)
... >>> sampleset.first.sample[qubit_a] == 1 and sampleset.first.sample[qubit_b] == -1
True
```

See Ocean Glossary for explanations of technical terms in descriptions of Ocean tools.

Properties

For parameters and properties of D-Wave systems, see D-Wave System Documentation.

<table>
<thead>
<tr>
<th>DWaveSampler.properties</th>
<th>D-Wave solver properties as returned by a SAPI query.</th>
</tr>
</thead>
<tbody>
<tr>
<td>DWaveSampler.parameters</td>
<td>D-Wave solver parameters in the form of a dict, where keys are keyword parameters accepted by a SAPI query and values are lists of properties in properties for each key.</td>
</tr>
<tr>
<td>DWaveSampler.nodelist</td>
<td>List of active qubits for the D-Wave solver.</td>
</tr>
<tr>
<td>DWaveSampler.edgelist</td>
<td>List of active couplers for the D-Wave solver.</td>
</tr>
<tr>
<td>DWaveSampler.adjacency</td>
<td>Adjacency structure formatted as a dict, where keys are the nodes of the structured sampler and values are sets of all adjacent nodes for each key node.</td>
</tr>
<tr>
<td>DWaveSampler.structure</td>
<td>Structure of the structured sampler formatted as a namedtuple, Structure(nodelist, edgelist, adjacency), where the 3-tuple values are the nodelist, edgelist and adjacency attributes.</td>
</tr>
</tbody>
</table>

dwave.system.samplers.DWaveSampler.properties

DWaveSampler.properties

D-Wave solver properties as returned by a SAPI query.

Solver properties are dependent on the selected D-Wave solver and subject to change; for example, new released features may add properties. D-Wave System Documentation describes the parameters and properties supported on the D-Wave system.
Examples

```python
>>> from dwave.system import DWaveSampler
>>> sampler = DWaveSampler()
>>> sampler.properties  # doctest: +SKIP
{'anneal_offset_ranges': [[-0.2197463755538704, 0.03821687759418928],
[-0.2242514597680286, 0.01718456460967399],
[-0.20860153999435985, 0.05511969218508182],
# Snipped above response for brevity
See Ocean Glossary for explanations of technical terms in descriptions of Ocean tools.

Type  dict
```

```
dwave.system.samplers.DWaveSampler.parameters

DWaveSampler.parameters
D-Wave solver parameters in the form of a dict, where keys are keyword parameters accepted by a SAPI query and values are lists of properties in properties for each key.

Solver parameters are dependent on the selected D-Wave solver and subject to change; for example, new released features may add parameters. D-Wave System Documentation describes the parameters and properties supported on the D-Wave system.

Examples

```python
>>> from dwave.system import DWaveSampler
>>> sampler = DWaveSampler()
>>> sampler.parameters  # doctest: +SKIP
{'anneal_offsets': ['parameters'],
'anneal_schedule': ['parameters'],
'annealing_time': ['parameters'],
'answer_mode': ['parameters'],
'auto_scale': ['parameters'],
# Snipped above response for brevity
See Ocean Glossary for explanations of technical terms in descriptions of Ocean tools.

Type  dict[str, list]
```

```
dwave.system.samplers.DWaveSampler.nodelist

DWaveSampler.nodelist
List of active qubits for the D-Wave solver.

Examples

Node list for one D-Wave 2000Q system (output snipped for brevity).

```python
>>> from dwave.system import DWaveSampler
>>> sampler = DWaveSampler()
>>> sampler.nodelist
[0, 1, 2, ...
```
See Ocean Glossary for explanations of technical terms in descriptions of Ocean tools.

**Type** list

dwave.system.samplers.DWaveSampler.edgelist

**DWaveSampler.edgelist**
List of active couplers for the D-Wave solver.

**Examples**
Coupler list for one D-Wave 2000Q system (output snipped for brevity).

```python
>>> from dwave.system import DWaveSampler
>>> sampler = DWaveSampler()
>>> sampler.edgelist
[(0, 4), (0, 5), (0, 6), (0, 7), ...
```

See Ocean Glossary for explanations of technical terms in descriptions of Ocean tools.

**Type** list

dwave.system.samplers.DWaveSampler.adjacency

**DWaveSampler.adjacency**
Adjacency structure formatted as a dict, where keys are the nodes of the structured sampler and values are sets of all adjacent nodes for each key node.

**Type** dict[variable, set]

dwave.system.samplers.DWaveSampler.structure

**DWaveSampler.structure**
Structure of the structured sampler formatted as a namedtuple, Structure(nodelist, edgelist, adjacency), where the 3-tuple values are the nodelist, edgelist and adjacency attributes.

**Methods**

<table>
<thead>
<tr>
<th>Method</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>DWaveSampler.sample</strong>(bqm, warnings)</td>
<td>Sample from the specified Ising model.</td>
</tr>
<tr>
<td><strong>DWaveSampler.sample_ising</strong>(h, *args, **kwargs)</td>
<td>Sample from an Ising model using the implemented sample method.</td>
</tr>
<tr>
<td><strong>DWaveSampler.sample_qubo</strong>(Q, **parameters)</td>
<td>Sample from a QUBO using the implemented sample method.</td>
</tr>
<tr>
<td><strong>DWaveSampler.validate_anneal_schedule</strong>(...)</td>
<td>Raise an exception if the specified schedule is invalid for the sampler.</td>
</tr>
<tr>
<td><strong>DWaveSampler.to_networkx_graph</strong>(..)</td>
<td>Converts DWaveSampler's structure to a Chimera or Pegasus NetworkX graph.</td>
</tr>
</tbody>
</table>
DWaveSampler.sample(bqm, warnings=None, **kwargs)

Sample from the specified Ising model.

Parameters

- **bqm** (BinaryQuadraticModel) – The binary quadratic model. Must match nodelist and edgelist.
- **warnings** (WarningAction, optional) – Defines what warning action to take, if any. See Warnings. The default behaviour is to ignore warnings.
- ****kwargs – Optional keyword arguments for the sampling method, specified per solver in parameters. D-Wave System Documentation’s solver guide describes the parameters and properties supported on the D-Wave system.

Returns A dimod SampleSet object. In it this sampler also provides timing information in the info field as described in the D-Wave System Documentation’s timing guide.

Return type dimod.SampleSet

Examples

This example submits a two-variable Ising problem mapped directly to two adjacent qubits on a D-Wave system. qubit_a is the first qubit in the QPU’s indexed list of qubits and qubit_b is one of the qubits coupled to it. Given sufficient reads (here 100), the quantum computer should return the best solution, 1, −1 on qubit_a and qubit_b, respectively, as its first sample (samples are ordered from lowest energy).

```python
>>> from dwave.system import DWaveSampler
... ...
>>> sampler = DWaveSampler()
... ...
>>> qubit_a = sampler.nodelist[0]
... ...
>>> qubit_b = next(iter(sampler.adjacency[qubit_a]))
>>> sampleset = sampler.sample_ising({qubit_a: -1, qubit_b: 1},
...                                     {},
...                                     num_reads=100)
>>> sampleset.first.sample[qubit_a] == 1 and sampleset.first.sample[qubit_b] == -1
True
```

See Ocean Glossary for explanations of technical terms in descriptions of Ocean tools.

dwave.system.samplers.DWaveSampler.sample_ising

DWaveSampler.sample_ising(h, *args, **kwargs)

Sample from an Ising model using the implemented sample method.

This method is inherited from the Sampler base class.

Converts the Ising model into a BinaryQuadraticModel and then calls sample().

Parameters

- **h** (dict/list) – Linear biases of the Ising problem. If a dict, should be of the form {v: bias, ...} where is a spin-valued variable and bias is its associated bias. If a list, it is treated as a list of biases where the indices are the variable labels.
• $J_{\text{dict}[(\text{variable, variable}), \text{bias}]}$ – Quadratic biases of the Ising problem.

• **kwargs – See the implemented sampling for additional keyword definitions.

Returns SampleSet

See also:

sample(), sample_qubo()

dwave.system.samplers.DWaveSampler.sample_qubo

DWaveSampler.sample_qubo($Q$, **parameters)

Sample from a QUBO using the implemented sample method.

This method is inherited from the Sampler base class.

Converts the QUBO into a BinaryQuadraticModel and then calls sample().

Parameters

• $Q_{\text{dict}}$ – Coefficients of a quadratic unconstrained binary optimization (QUBO) problem. Should be a dict of the form ${u, v): bias, ...}$ where $u, v$, are binary-valued variables and bias is their associated coefficient.

• **kwargs – See the implemented sampling for additional keyword definitions.

Returns SampleSet

See also:

sample(), sample_ising()

dwave.system.samplers.DWaveSampler.validate_anneal_schedule

DWaveSampler.validate_anneal_schedule(anneal_schedule)

Raise an exception if the specified schedule is invalid for the sampler.

Parameters anneal_schedule (list) – An anneal schedule variation is defined by a series of pairs of floating-point numbers identifying points in the schedule at which to change slope. The first element in the pair is time $t$ in microseconds; the second, normalized persistent current $s$ in the range [0,1]. The resulting schedule is the piecewise-linear curve that connects the provided points.

Raises

• ValueError – If the schedule violates any of the conditions listed below.

• RuntimeError – If the sampler does not accept the anneal_schedule parameter or if it does not have annealing_time_range or max_anneal_schedule_points properties.

As described in D-Wave System Documentation, an anneal schedule must satisfy the following conditions:

• Time $t$ must increase for all points in the schedule.

• For forward annealing, the first point must be (0,0) and the anneal fraction $s$ must increase monotonically.

• For reverse annealing, the anneal fraction $s$ must start and end at $s=1$.

• In the final point, anneal fraction $s$ must equal 1 and time $t$ must not exceed the maximum value in the annealing_time_range property.
• The number of points must be >=2.
• The upper bound is system-dependent; check the `max_anneal_schedule_points` property. For reverse annealing, the maximum number of points allowed is one more than the number given by this property.

Examples

This example sets a quench schedule on a D-Wave system.

```python
>>> from dwave.system import DWaveSampler
>>> sampler = DWaveSampler()
>>> quench_schedule=[[0.0, 0.0], [12.0, 0.6], [12.8, 1.0]]
>>> DWaveSampler().validate_anneal_schedule(quench_schedule)  # doctest: +SKIP
```

**dwave.system.samplers.DWaveSampler.to_networkx_graph**

DWaveSampler`s to_networkx_graph() converts DWaveSampler’s structure to a Chimera or Pegasus NetworkX graph.

**Returns** Either an (m, n, t) Chimera lattice or a Pegasus lattice of size m.

**Return type** networkx.Graph

Examples

This example converts a selected D-Wave system solver to a graph and verifies it has over 2000 nodes.

```python
>>> from dwave.system import DWaveSampler
...
>>> sampler = DWaveSampler()
>>> g = sampler.to_networkx_graph()  # doctest: +SKIP
>>> len(g.nodes) > 2000  # doctest: +SKIP
True
```

**DWaveCliqueSampler**

class DWaveCliqueSampler(**config)

A sampler for solving clique problems on the D-Wave system.

This sampler wraps `find_clique_embedding()` to generate embeddings with even chain length. These embeddings work well for dense binary quadratic models. For sparse models, using `EmbeddingComposite` with `DWaveSampler` is preferred.

**Parameters**

**config** – Keyword arguments, as accepted by `DWaveSampler`

Examples

This example creates a BQM based on a 6-node clique (complete graph), with random ±1 values assigned to nodes, and submits it to a D-Wave system. Parameters for communication with the system, such as its URL and an authentication token, are implicitly set in a configuration file or as environment variables, as described in Configuring Access to D-Wave Solvers.
```python
>>> from dwave.system import DWaveCliqueSampler
>>> import dimod
... >>> bqm = dimod.generators.ran_r(1, 6)
... >>> sampler = DWaveCliqueSampler()  # doctest: +SKIP
>>> sampler.largest_clique_size > 5  # doctest: +SKIP
True
>>> sampleset = sampler.sample(bqm, num_reads=100)  # doctest: +SKIP
```

### Properties

<table>
<thead>
<tr>
<th>Property</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>DWaveCliqueSampler.largest_clique_size</code></td>
<td>The maximum number of variables that can be embedded.</td>
</tr>
<tr>
<td><code>DWaveCliqueSampler.properties</code></td>
<td>A dict containing any additional information about the sampler.</td>
</tr>
<tr>
<td><code>DWaveCliqueSampler.parameters</code></td>
<td>A dict where keys are the keyword parameters accepted by the sampler methods and values are lists of the properties relevant to each parameter.</td>
</tr>
</tbody>
</table>

### dwave.system.samplers.DWaveCliqueSampler.largest_clique_size

`DWaveCliqueSampler.largest_clique_size`

The maximum number of variables that can be embedded.

### dwave.system.samplers.DWaveCliqueSampler.properties

`DWaveCliqueSampler.properties`

A dict containing any additional information about the sampler.

Type: `dict`

### dwave.system.samplers.DWaveCliqueSampler.parameters

`DWaveCliqueSampler.parameters`

A dict where keys are the keyword parameters accepted by the sampler methods and values are lists of the properties relevant to each parameter.

Type: `dict`

### Methods

<table>
<thead>
<tr>
<th>Method</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>DWaveCliqueSampler.largest_clique()</code></td>
<td>The clique embedding with the maximum number of source variables.</td>
</tr>
<tr>
<td><code>DWaveCliqueSampler.sample(bqm[, chain_strength])</code></td>
<td>Sample from the specified binary quadratic model.</td>
</tr>
</tbody>
</table>

Continued on next page
**DWaveCliqueSampler.sample** *(bqm, chain_strength=None, **kwargs) – Sample from the specified binary quadratic model.*

**Parameters**

- `bqm` *(BinaryQuadraticModel) – Any binary quadratic model with up to largest_clique_size variables. This BQM is embedded using a clique embedding.*
- `chain_strength` *(float, optional) – The (relative) chain strength to use in the embedding. By default a chain strength of \(1.5\sqrt{N}\) where \(N\) is the size of the largest clique, as returned by largest_clique_size.*
- `**kwargs` *(Optional keyword arguments for the sampling method, specified per solver in parameters. D-Wave System Documentation’s solver guide describes the parameters and properties supported on the D-Wave system. Note that auto_scale is not supported by this sampler, because it scales the problem as part of the embedding process.)*

**DWaveCliqueSampler.sample_ising** *(h, J, **parameters) – Sample from an Ising model using the implemented sample method.*

**Returns** *(SampleSet)*

**See also:**

`sample()`, `sample_qubo()`
DWaveCliqueSampler.sample_qubo(Q, **parameters)

Sample from a QUBO using the implemented sample method.
This method is inherited from the Sampler base class.

Converts the QUBO into a BinaryQuadraticModel and then calls sample().

Parameters

- Q (dict) – Coefficients of a quadratic unconstrained binary optimization (QUBO) problem. Should be a dict of the form {(u, v): bias, ...} where u, v, are binary-valued variables and bias is their associated coefficient.
- **kwargs – See the implemented sampling for additional keyword definitions.

Returns

SampleSet

See also:
sample(), sample_ising()

LeapHybridSampler

class LeapHybridSampler(solver=None, connection_close=True, **config)

A class for using Leap’s cloud-based hybrid solvers.

Uses parameters set in a configuration file, as environment variables, or explicitly as input arguments for selecting and communicating with a hybrid solver. For more information, see D-Wave Cloud Client.

Inherits from dimod.Sampler.

Parameters **config – Keyword arguments passed to dwave.cloud.client.Client.from_config().

Examples

This example builds a random sparse graph and uses a hybrid solver to find a maximum independent set.

```python
>>> import dimod
>>> import networkx as nx
>>> import dwave_networkx as dnx
>>> import numpy as np
>>> from dwave.system import LeapHybridSampler
...
>>> # Create a maximum-independent set problem from a random graph
>>> problem_node_count = 300
>>> G = nx.random_geometric_graph(problem_node_count, radius=0.0005*problem_node_count)
>>> qubo = dnx.algorithms.independent_set.maximum_weighted_independent_set_qubo(G)
>>> bqm = dimod.BQM.from_qubo(qubo)
...
>>> # Find a good solution
>>> sampler = LeapHybridSampler()  # doctest: +SKIP
>>> sampleset = sampler.sample(bqm)  # doctest: +SKIP
```
Properties

<table>
<thead>
<tr>
<th>LeapHybridSampler.properties</th>
<th>Solver properties as returned by a SAPI query.</th>
</tr>
</thead>
<tbody>
<tr>
<td>LeapHybridSampler.parameters</td>
<td>Solver parameters in the form of a dict, where keys are keyword parameters accepted by a SAPI query and values are lists of properties in properties for each key.</td>
</tr>
</tbody>
</table>

**dwave.system.samplers.LeapHybridSampler.properties**

LeapHybridSampler.properties
Solver properties as returned by a SAPI query.
Solver properties are dependent on the selected solver and subject to change.
Type: dict

**dwave.system.samplers.LeapHybridSampler.parameters**

LeapHybridSampler.parameters
Solver parameters in the form of a dict, where keys are keyword parameters accepted by a SAPI query and values are lists of properties in properties for each key.
Solver parameters are dependent on the selected solver and subject to change.
Type: dict[str, list]

Methods

<table>
<thead>
<tr>
<th>LeapHybridSampler.sample(bqm[, time_limit])</th>
<th>Sample from the specified binary quadratic model.</th>
</tr>
</thead>
<tbody>
<tr>
<td>LeapHybridSampler.sample_ising(h, J, ...)</td>
<td>Sample from an Ising model using the implemented sample method.</td>
</tr>
<tr>
<td>LeapHybridSampler.sample_qubo(Q, **parameters)</td>
<td>Sample from a QUBO using the implemented sample method.</td>
</tr>
</tbody>
</table>

**dwave.system.samplers.LeapHybridSampler.sample**

LeapHybridSampler.sample(bqm, time_limit=None, **kwargs)
Sample from the specified binary quadratic model.

Parameters

- **bqm** (dimod.BinaryQuadraticModel) – The binary quadratic model.
- **time_limit** (int) – Maximum run time, in seconds, to allow the solver to work on the problem. Must be at least the minimum required for the number of problem variables, which is calculated and set by default. The minimum time for a hybrid solver is specified as a piecewise-linear curve defined by a set of floating-point pairs, the minimum_time_limit field under properties. The first element in each pair is the number of problem variables; the second is the minimum required time. The minimum time for any particular number of variables is a linear interpolation calculated on two pairs that represent the relevant range for the given number of variables. For example, if LeapHybridSampler().properties[“minimum_time_limit”] returns [[1, 0.1], [100, 10.0], [1000, 20.0]], then
the minimum time for a 50-variable problem is 5 seconds, the linear interpolation of the first
two pairs that represent problems with between 1 to 100 variables.

• **kwargs – Optional keyword arguments for the solver, specified in parameters.

Returns A `dimod SampleSet` object.

Return type `dimod.SampleSet`

Examples

This example builds a random sparse graph and uses a hybrid solver to find a maximum independent set.

```python
>>> import dimod
>>> import networkx as nx
>>> import dwave_networkx as dnx
>>> import numpy as np
...
>>> # Create a maximum-independent set problem from a random graph
>>> problem_node_count = 300
>>> G = nx.random_geometric_graph(problem_node_count, radius=0.0005*problem_node_count)
>>> qubo = dnx.algorithms.independent_set.maximum_weighted_independent_set_qubo(G)
>>> bqm = dimod.BQM.from_qubo(qubo)
...
>>> # Find a good solution
>>> sampler = LeapHybridSampler() # doctest: +SKIP
>>> sampleset = sampler.sample(bqm) # doctest: +SKIP
```

`dwave.system.samplers.LeapHybridSampler.sample_ising`

`LeapHybridSampler.sample_ising(h, J, **parameters)`

Sample from an Ising model using the implemented sample method.

This method is inherited from the `Sampler` base class.

Converts the Ising model into a `BinaryQuadraticModel` and then calls `sample()`.

Parameters

- `h (dict/list)` – Linear biases of the Ising problem. If a dict, should be of the form `{v: bias, ...}` where is a spin-valued variable and `bias` is its associated bias. If a list, it is treated as a list of biases where the indices are the variable labels.

- `J (dict[(variable, variable), bias])` – Quadratic biases of the Ising problem.

- **kwargs – See the implemented sampling for additional keyword definitions.

Returns `SampleSet`

See also:

`sample()`, `sample_qubo()`
LeapHybridSampler.sample_qubo \(Q, \{\text{parameters}\}\)

Sample from a QUBO using the implemented sample method.

This method is inherited from the Sampler base class.

Converts the QUBO into a BinaryQuadraticModel and then calls sample().

Parameters

- \(Q (\text{dict})\) – Coefficients of a quadratic unconstrained binary optimization (QUBO) problem. Should be a dict of the form \{(u, v): bias, \ldots\} where \(u, v\) are binary-valued variables and \(bias\) is their associated coefficient.

- \(**kwargs\) – See the implemented sampling for additional keyword definitions.

Returns SampleSet

See also: sample(), sample_ising()
Methods

LeapHybridDQMSampler.sample_dqm(dqm[, time_limit=None, compressed=False, **kwargs])

Sample from the specified binary quadratic model.

Parameters

- **bqm** (dimod.DiscreteQuadraticModel) – The binary quadratic model.
- **time_limit** (int) – The maximum run time in seconds.
- **kwargs** – Optional keyword arguments for the solver, specified in parameters.

Returns

A sample set.

Return type

dimod.SampleSet

1.2.2 Composites

dimod composites that provide layers of pre- and post-processing (e.g., minor-embedding) when using the D-Wave system.

Contents

- **Composites**
  - **CutOffs**
    - CutOffComposite
    - PolyCutOffComposite
  - **Embedding**
    - AutoEmbeddingComposite
    - EmbeddingComposite
    - FixedEmbeddingComposite
    - LazyFixedEmbeddingComposite
    - TilingComposite
    - VirtualGraphComposite
  - **Reverse Anneal**
    - ReverseBatchStatesComposite
    - ReverseAdvanceComposite
CutOffs

Prunes the binary quadratic model (BQM) submitted to the child sampler by retaining only interactions with values commensurate with the sampler’s precision.

CutOffComposite

class CutOffComposite(child_sampler, cutoff, cutoff_vartype=<Vartype.SPIN: frozenset({1, -1})>, comparison=<built-in function lt>)

Composite to remove interactions below a specified cutoff value.

Prunes the binary quadratic model (BQM) submitted to the child sampler by retaining only interactions with values commensurate with the sampler’s precision as specified by the `cutoff` argument. Also removes variables isolated post- or pre-removal of these interactions from the BQM passed on to the child sampler, setting these variables to values that minimize the original BQM’s energy for the returned samples.

Parameters

- sampler (dimod.Sampler) – A dimod sampler.
- cutoff (number) – Lower bound for absolute value of interactions. Interactions with absolute values lower than cutoff are removed. Isolated variables are also not passed on to the child sampler.
- cutoff_vartype (Vartype/str/set, default='SPIN') – Variable space to execute the removal in. Accepted input values:
  - Vartype.SPIN, 'SPIN', {-1, 1}
  - Vartype.BINARY, 'BINARY', {0, 1}
- comparison (function, optional) – A comparison operator for comparing interaction values to the cutoff value. Defaults to operator.lt().

Examples

This example removes one interaction, 'ac': -0.7, before embedding on a D-Wave system. Note that the lowest-energy sample for the embedded problem is {'a': 1, 'b': -1, 'c': -1} but with a large enough number of samples (here num_reads=1000), the lowest-energy solution to the complete BQM is likely found and its energy recalculated by the composite.

```python
>>> import dimod
>>> sampler = DWaveSampler(solver={'qpu': True})
>>> bqm = dimod.BinaryQuadraticModel({'a': -1, 'b': 1, 'c': 1}, ...
    ...   {'ab': -0.8, 'ac': -0.7, 'bc': -1},
    ...   0,
    ...   dimod.SPIN)
>>> CutOffComposite(AutoEmbeddingComposite(sampler), 0.75).sample(bqm,
...   num_reads=1000).first.energy
-3.5
```

Properties

- **CutOffComposite.child**
  The child sampler.

Continued on next page
Table 9 – continued from previous page

<table>
<thead>
<tr>
<th>Method</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CutOffComposite.children</td>
<td>List of child samplers that that are used by this composite.</td>
</tr>
<tr>
<td>CutOffComposite.properties</td>
<td>A dict containing any additional information about the sampler.</td>
</tr>
<tr>
<td>CutOffComposite.parameters</td>
<td>A dict where keys are the keyword parameters accepted by the sampler methods and values are lists of the properties relevant to each parameter.</td>
</tr>
</tbody>
</table>

**dwave.system.composites.CutOffComposite.child**

CutOffComposite.child

The child sampler. First sampler in Composite.children.

Type: Sampler

**dwave.system.composites.CutOffComposite.children**

CutOffComposite.children

List of child samplers that that are used by this composite.

**dwave.system.composites.CutOffComposite.properties**

CutOffComposite.properties

A dict containing any additional information about the sampler.

**dwave.system.composites.CutOffComposite.parameters**

CutOffComposite.parameters

A dict where keys are the keyword parameters accepted by the sampler methods and values are lists of the properties relevant to each parameter.

**Methods**

<table>
<thead>
<tr>
<th>Method</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>CutOffComposite.sample(bqm, **parameters)</td>
<td>Cut off interactions and sample from the provided binary quadratic model.</td>
</tr>
<tr>
<td>CutOffComposite.sample_ising(h, J, **parameters)</td>
<td>Sample from an Ising model using the implemented sample method.</td>
</tr>
<tr>
<td>CutOffComposite.sample_qubo(Q, **parameters)</td>
<td>Sample from a QUBO using the implemented sample method.</td>
</tr>
</tbody>
</table>

**dwave.system.composites.CutOffComposite.sample**

CutOffComposite.sample(bqm, **parameters)

Cut off interactions and sample from the provided binary quadratic model.

Prunes the binary quadratic model (BQM) submitted to the child sampler by retaining only interactions with value commensurate with the sampler’s precision as specified by the cutoff argument. Also removes variables isolated post- or pre-removal of these interactions from the BQM passed on to the child sampler, setting these
variables to values that minimize the original BQM’s energy for the returned samples.

Parameters

- **bqm** (dimod.BinaryQuadraticModel) – Binary quadratic model to be sampled from.
- **parameters** – Parameters for the sampling method, specified by the child sampler.

Returns dimod.SampleSet

Examples

See the example in CutOffComposite.

dwave.system.composites.CutOffComposite.sample_ising

CutOffComposite.sample_ising(h, J, **parameters)
Sample from an Ising model using the implemented sample method.
This method is inherited from the Sampler base class.
Converts the Ising model into a BinaryQuadraticModel and then calls sample().

Parameters

- **h** (dict/list) – Linear biases of the Ising problem. If a dict, should be of the form {v: bias, ...} where is a spin-valued variable and bias is its associated bias. If a list, it is treated as a list of biases where the indices are the variable labels.
- **J** (dict[(variable, variable), bias]) – Quadratic biases of the Ising problem.
- **kwargs** – See the implemented sampling for additional keyword definitions.

Returns SampleSet

See also:

test(), sample_qubo()
PolyCutOffComposite

Prunes the polynomial submitted to the child sampler by retaining only interactions with values commensurate with the sampler’s precision.

class PolyCutOffComposite(child_sampler, cutoff, cutoff_vartype=<Vartype.SPIN: frozenset({1, -1}), comparison=<built-in function lt>)

Composite to remove polynomial interactions below a specified cutoff value.

Prunes the binary polynomial submitted to the child sampler by retaining only interactions with values commensurate with the sampler’s precision as specified by the cutoff argument. Also removes variables isolated post- or pre-removal of these interactions from the polynomial passed on to the child sampler, setting these variables to values that minimize the original polynomial’s energy for the returned samples.

Parameters

- **sampler** (dimod.PolySampler) – A dimod binary polynomial sampler.
- **cutoff** (number) – Lower bound for absolute value of interactions. Interactions with absolute values lower than cutoff are removed. Isolated variables are also not passed on to the child sampler.
- **cutoff_vartype** (Vartype/str/set, default='SPIN') – Variable space to do the cutoff in. Accepted input values:
  - Vartype.SPIN, 'SPIN', {-1, 1}
  - Vartype.BINARY, 'BINARY', {0, 1}
- **comparison** (function, optional) – A comparison operator for comparing the interaction value to the cutoff value. Defaults to operator.lt().

Examples

This example removes one interaction, 'ac': 0.2, before submitting the polynomial to child sampler ExactSolver.

```python
>>> import dimod
>>> sampler = dimod.HigherOrderComposite(dimod.ExactSolver())
>>> poly = dimod.BinaryPolynomial({'a': 3, 'abc':-4, 'ac': 0.2}, dimod.SPIN)
>>> PolyCutOffComposite(sampler, 1).sample_poly(poly).first.sample['a']
-1
```

Properties

- **PolyCutOffComposite.child**
  - The child sampler.
- **PolyCutOffComposite.children**
  - List of child samplers that are used by this composite.
- **PolyCutOffComposite.properties**
  - A dict containing any additional information about the sampler.
- **PolyCutOffComposite.parameters**
  - A dict where keys are the keyword parameters accepted by the sampler methods and values are lists of the properties relevant to each parameter.
The child sampler. First sampler in Composite.children.

Type Sampler

List of child samplers that that are used by this composite.

A dict containing any additional information about the sampler.

A dict where keys are the keyword parameters accepted by the sampler methods and values are lists of the properties relevant to each parameter.

Cutoff and sample from the provided binary polynomial. Prunes the binary polynomial submitted to the child sampler by retaining only interactions with values commensurate with the sampler’s precision as specified by the cutoff argument. Also removes variables isolated post- or pre-removal of these interactions from the polynomial passed on to the child sampler, setting these variables to values that minimize the original polynomial’s energy for the returned samples.

Parameters

- poly (dimod.BinaryPolynomial) – Binary polynomial to be sampled from.
- **parameters – Parameters for the sampling method, specified by the child sampler.

Returns dimod.SampleSet
Examples

See the example in \texttt{PolyCutOffComposite}.

\pkg{dwave.system.composites.PolyCutOffComposite}.\texttt{sample_hising}

\begin{verbatim}
PolyCutOffComposite.\texttt{sample_hising}(h, J, **kwargs)

Sample from a higher-order Ising model.

Convert the given higher-order Ising model to a \texttt{BinaryPolynomial} and call \texttt{sample_poly()}.

Parameters

- \texttt{h (dict)} – Variable biases of the Ising problem as a dict of the form \{v: bias, \ldots\}, where \texttt{v} is a variable in the polynomial and \texttt{bias} its associated coefficient.
- \texttt{J (dict)} – Interaction biases of the Ising problem as a dict of the form \{(u, v, \ldots): bias\}, where \texttt{u, v} are spin-valued variables in the polynomial and \texttt{bias} their associated coefficient.
- **\texttt{kwargs} – See \texttt{sample_poly()} for additional keyword definitions.

Returns SampleSet

See also:

\texttt{sample_poly()}, \texttt{sample_hubo()}
\end{verbatim}

\pkg{dwave.system.composites.PolyCutOffComposite}.\texttt{sample_hubo}

\begin{verbatim}
PolyCutOffComposite.\texttt{sample_hubo}(H, **kwargs)

Sample from a higher-order unconstrained binary optimization problem.

Convert the given higher-order unconstrained binary optimization problem to a \texttt{BinaryPolynomial} and then call \texttt{sample_poly()}.

Parameters

- \texttt{H (dict)} – Coefficients of the HUBO as a dict of the form \{(u, v, \ldots): bias, \ldots\}, where \texttt{u, v} are binary-valued variables in the polynomial and \texttt{bias} their associated coefficient.
- **\texttt{kwargs} – See \texttt{sample_poly()} for additional keyword definitions.

Returns SampleSet

See also:

\texttt{sample_poly()}, \texttt{sample_hising()}
\end{verbatim}

Embedding

Minor-embed a problem BQM into a D-Wave system.

Embedding composites for various types of problems and application. For example:

- \texttt{EmbeddingComposite} for a problem with arbitrary structure that likely requires heuristic embedding.
- \texttt{AutoEmbeddingComposite} can save unnecessary embedding for problems that might have a structure similar to the child sampler.
- \texttt{LazyFixedEmbeddingComposite} can benefit applications that resubmit a BQM with changes in some values.
AutoEmbeddingComposite

class AutoEmbeddingComposite(child_sampler, **kwargs)

Maps problems to a structured sampler, embedding if needed.

This composite first tries to submit the binary quadratic model directly to the child sampler and only embeds if a \texttt{dimod.exceptions.BinaryQuadraticModelStructureError} is raised.

Parameters

- \texttt{child_sampler} (\texttt{dimod.Sampler}) – Structured dimod sampler, such as a \texttt{DWaveSampler}().
- \texttt{find_embedding} (\texttt{function}, \texttt{optional}) – A function \texttt{find_embedding(S, T, **kwargs)} where \texttt{S} and \texttt{T} are edgelists. The function can accept additional keyword arguments. Defaults to \texttt{minorminer.find_embedding}().
- \texttt{kwargs} – See the \texttt{EmbeddingComposite} class for additional keyword arguments.

Properties

<table>
<thead>
<tr>
<th>AutoEmbeddingComposite.child</th>
<th>The child sampler.</th>
</tr>
</thead>
<tbody>
<tr>
<td>AutoEmbeddingComposite.parameters</td>
<td>None</td>
</tr>
<tr>
<td>AutoEmbeddingComposite.properties</td>
<td>None</td>
</tr>
</tbody>
</table>

\texttt{dwave.system.composites.AutoEmbeddingComposite.child}

\texttt{AutoEmbeddingComposite.child}

The child sampler. First sampler in \texttt{Composite.children}.

Type \texttt{Sampler}

\texttt{dwave.system.composites.AutoEmbeddingComposite.parameters}

\texttt{AutoEmbeddingComposite.parameters = None}

\texttt{dwave.system.composites.AutoEmbeddingComposite.properties}

\texttt{AutoEmbeddingComposite.properties = None}

Methods

<table>
<thead>
<tr>
<th>AutoEmbeddingComposite.sample(bqm, **parameters)</th>
<th>Sample from the provided binary quadratic model.</th>
</tr>
</thead>
<tbody>
<tr>
<td>AutoEmbeddingComposite.sample_ising(h, J,...)</td>
<td>Sample from an Ising model using the implemented sample method.</td>
</tr>
<tr>
<td>AutoEmbeddingComposite.sample_qubo(Q,...)</td>
<td>Sample from a QUBO using the implemented sample method.</td>
</tr>
</tbody>
</table>
AutoEmbeddingComposite.sample

Sample from the provided binary quadratic model.

**Parameters**

- **bqm** (*dimod.BinaryQuadraticModel*) – Binary quadratic model to be sampled from.
- **chain_strength** (*float/mapping/callable, optional*) – Magnitude of the quadratic bias (in SPIN-space) applied between variables to create chains. The energy penalty of chain breaks is 2 * \(chain\_strength\). If a mapping is passed, a chain-specific strength is applied. If a callable is passed, it will be called on \(chain\_strength(bqm, embedding)\) and should return a float or mapping, to be interpreted as above. By default, \(chain\_strength\) is calculated with \(uniform\_torque\_compensation()\).
- **chain_break_method** (*function/list, optional*) – Method or methods used to resolve chain breaks. If multiple methods are given, the results are concatenated and a new field called “chain_break_method” specifying the index of the method is appended to the sample set. See \(unembed\_sampleset()\) and \(dwave.embedding.chain\_breaks\).
- **chain_break_fraction** (*bool, optional, default=True*) – Add a \(chain\_break\_fraction\) field to the unembedded response with the fraction of chains broken before unembedding.
- **embedding_parameters** (*dict, optional*) – If provided, parameters are passed to the embedding method as keyword arguments. Overrides any \(embedding\_parameters\) passed to the constructor.
- **return_embedding** (*bool, optional*) – If True, the embedding, chain strength, chain break method and embedding parameters are added to \(dimod\._SampleSet.info\) of the returned sample set. The default behaviour is defined by \(return\_embedding\_default\), which itself defaults to False.
- **warnings** (*WarningAction, optional*) – Defines what warning action to take, if any. See \(warnings\). The default behaviour is defined by \(warnings\_default\), which itself defaults to \(IGNORE\).
- ****parameters – Parameters for the sampling method, specified by the child sampler.

**Returns** *dimod.SampleSet*

**Examples**

See the example in EmbeddingComposite.

dwave.system.composites.AutoEmbeddingComposite.sample_ising

AutoEmbeddingComposite.sample_ising(*h, J, **parameters*)

Sample from an Ising model using the implemented sample method.

This method is inherited from the Sampler base class.

Converts the Ising model into a *BinaryQuadraticModel* and then calls \(sample()\).

**Parameters**
• \( h(\text{dict/list}) \) – Linear biases of the Ising problem. If a dict, should be of the form \( \{v: \text{bias}, \ldots\} \) where is a spin-valued variable and \( \text{bias} \) is its associated bias. If a list, it is treated as a list of biases where the indices are the variable labels.

• \( J(\text{dict}[(\text{variable, variable}), \text{bias}]) \) – Quadratic biases of the Ising problem.

• **kwargs – See the implemented sampling for additional keyword definitions.

Returns SampleSet

See also:

\( \text{sample()}, \text{sample_qubo()} \)

dwave.system.composites.AutoEmbeddingComposite.sample_qubo

AutoEmbeddingComposite.sample_qubo(\( Q, **\text{parameters} \))

Sample from a QUBO using the implemented sample method.

This method is inherited from the Sampler base class.

Converts the QUBO into a BinaryQuadraticModel and then calls \( \text{sample}() \).

Parameters

• \( Q(\text{dict}) \) – Coefficients of a quadratic unconstrained binary optimization (QUBO) problem. Should be a dict of the form \( \{(u, v): \text{bias}, \ldots\} \) where \( u, v \) are binary-valued variables and \( \text{bias} \) is their associated coefficient.

• **kwargs – See the implemented sampling for additional keyword definitions.

Returns SampleSet

See also:

\( \text{sample()}, \text{sample_ising()} \)

EmbeddingComposite

class EmbeddingComposite(\( \text{child_sampler}, \text{find_embedding=}\text{function find_embedding}, \text{embedding_parameters=None, scale_aware=False, child_structure_search=}\text{function child_structure_dfs}\))

Maps problems to a structured sampler.

Automatically minor-embeds a problem into a structured sampler such as a D-Wave system. A new minor-embedding is calculated each time one of its sampling methods is called.

Parameters

• \( \text{child_sampler} \) (dimod.Sampler) – A dimod sampler, such as a DWaveSampler, that accepts only binary quadratic models of a particular structure.

• \( \text{find_embedding} \) (function, optional) – A function \( \text{find_embedding}(S, T, **\text{kwargs}) \) where \( S \) and \( T \) are edgelists. The function can accept additional keyword arguments. Defaults to \text{minorminer.find_embedding}().

• \( \text{embedding_parameters} \) (dict, optional) – If provided, parameters are passed to the embedding method as keyword arguments.

• \( \text{scale_aware} \) (bool, optional, default=False) – Pass chain interactions to child samplers that accept an ignored_interactions parameter.
• **child_structure_search**  *(function, optional) – A function*  
  `child_structure_search(sampler)`  
  *that accepts a sampler and returns the dimod. Structured.structure.*  
  *Defaults to dimod.child_structure_dfs().*

### Examples

```
>>> from dwave.system import DWaveSampler, EmbeddingComposite
...  
>>> sampler = EmbeddingComposite(DWaveSampler())
>>> h = {'a': -1., 'b': 2}
>>> J = {('a', 'b'): 1.5}
>>> sampleset = sampler.sample_ising(h, J, num_reads=100)
>>> sampleset.first.energy
-4.5
```

### Properties

<table>
<thead>
<tr>
<th>Property Name</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>EmbeddingComposite.child</td>
<td>The child sampler.</td>
</tr>
<tr>
<td>EmbeddingComposite.parameters</td>
<td>Parameters in the form of a dict.</td>
</tr>
<tr>
<td>EmbeddingComposite.properties</td>
<td>Properties in the form of a dict.</td>
</tr>
<tr>
<td>EmbeddingComposite.return_embedding_default</td>
<td>Defines the default behaviour for sample()’s return_embedding kwarg.</td>
</tr>
<tr>
<td>EmbeddingComposite.warnings_default</td>
<td>Defines the default behavior for sample()’s warnings kwarg.</td>
</tr>
</tbody>
</table>

**dwave.system.composites.EmbeddingComposite.child**

*EmbeddingComposite.child*

The child sampler. First sampler in Composite.children.

*Type*  
Sampler

**dwave.system.composites.EmbeddingComposite.parameters**

*EmbeddingComposite.parameters = None*

Parameters in the form of a dict.

For an instantiated composed sampler, keys are the keyword parameters accepted by the child sampler and parameters added by the composite.

*Type*  
dict[str, list]

**dwave.system.composites.EmbeddingComposite.properties**

*EmbeddingComposite.properties = None*

Properties in the form of a dict.

Contains the properties of the child sampler.

*Type*  
dict
dwave.system.composites.EmbeddingComposite.return_embedding_default

EmbeddingComposite.return_embedding_default = False

Defines the default behaviour for sample()'s return_embedding kwarg.

dwave.system.composites.EmbeddingComposite.warnings_default

EmbeddingComposite.warnings_default = 'ignore'

Defines the default behaviour for sample()'s warnings kwarg.

Methods

<table>
<thead>
<tr>
<th>Method</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>EmbeddingComposite.sample(bqm[,...])</td>
<td>Sample from the provided binary quadratic model.</td>
</tr>
<tr>
<td>EmbeddingComposite.sample_ising(h, J, ...)</td>
<td>Sample from an Ising model using the implemented sample method.</td>
</tr>
<tr>
<td>EmbeddingComposite.sample_qubo(Q, **parameters)</td>
<td>Sample from a QUBO using the implemented sample method.</td>
</tr>
</tbody>
</table>

**dwave.system.composites.EmbeddingComposite.sample**

EmbeddingComposite.sample(bqm, chain_strength=None, chain_break_method=None, chain_break_fraction=True, embedding_parameters=None, return_embedding=None, warnings=None, **parameters)

Sample from the provided binary quadratic model.

**Parameters**

- **bqm** (dimod.BinaryQuadraticModel) – Binary quadratic model to be sampled from.

- **chain_strength** (float/mapping/callable, optional) – Magnitude of the quadratic bias (in SPIN-space) applied between variables to create chains. The energy penalty of chain breaks is 2 * chain_strength. If a mapping is passed, a chain-specific strength is applied. If a callable is passed, it will be called on chain_strength(bqm, embedding) and should return a float or mapping, to be interpreted as above. By default, chain_strength is calculated with uniform_torque_compensation().

- **chain_break_method** (function/list, optional) – Method or methods used to resolve chain breaks. If multiple methods are given, the results are concatenated and a new field called “chain_break_method” specifying the index of the method is appended to the sample set. See unembed_sampleset() and dwave.embedding.chain_breaks.

- **chain_break_fraction** (bool, optional, default=True) – Add a chain_break_fraction field to the unembedded response with the fraction of chains broken before unembedding.

- **embedding_parameters** (dict, optional) – If provided, parameters are passed to the embedding method as keyword arguments. Overrides any embedding_parameters passed to the constructor.

- **return_embedding** (bool, optional) – If True, the embedding, chain strength, chain break method and embedding parameters are added to dimod.SampleSet.info of the returned sample set. The default behaviour is defined by return_embedding_default, which itself defaults to False.
• **warnings** (*WarningAction*, optional) – Defines what warning action to take, if any. See `warnings`. The default behaviour is defined by `warnings_default`, which itself defaults to `IGNORE`.

• **parameters** – Parameters for the sampling method, specified by the child sampler.

Returns `dimod.SampleSet`

**Examples**

See the example in `EmbeddingComposite`.

**dwave.system.composites.EmbeddingComposite.sample_ising**

`EmbeddingComposite.sample_ising(h, J, **parameters)`

Sample from an Ising model using the implemented sample method.

This method is inherited from the `Sampler` base class.

Converts the Ising model into a `BinaryQuadraticModel` and then calls `sample()`.

**Parameters**

- `h` (dict/list) – Linear biases of the Ising problem. If a dict, should be of the form `{v: bias, ...}` where `v` is a spin-valued variable and `bias` is its associated bias. If a list, it is treated as a list of biases where the indices are the variable labels.

- `J` (dict[(variable, variable), bias]) – Quadratic biases of the Ising problem.

- **kwargs** – See the implemented sampling for additional keyword definitions.

**Returns** `SampleSet`

See also:

`sample()`, `sample_qubo()`

**dwave.system.composites.EmbeddingComposite.sample_qubo**

`EmbeddingComposite.sample_qubo(Q, **parameters)`

Sample from a QUBO using the implemented sample method.

This method is inherited from the `Sampler` base class.

Converts the QUBO into a `BinaryQuadraticModel` and then calls `sample()`.

**Parameters**

- `Q` (dict) – Coefficients of a quadratic unconstrained binary optimization (QUBO) problem. Should be a dict of the form `{(u, v): bias, ...}` where `u, v` are binary-valued variables and `bias` is their associated coefficient.

- **kwargs** – See the implemented sampling for additional keyword definitions.

**Returns** `SampleSet`

See also:

`sample()`, `sample_ising()`
FixedEmbeddingComposite

class FixedEmbeddingComposite(child_sampler, embedding=None, source Adjacency=None, **kwargs)

Maps problems to a structured sampler with the specified minor-embedding.

Parameters

- **child_sampler** ([dimod.Sampler]) – Structured dimod sampler such as a D-Wave system.
- **embedding** ([dict[hashable, iterable], optional]) – Mapping from a source graph to the specified sampler’s graph (the target graph).
- **source adjacency** ([dict[hashable, iterable]]) – Deprecated. Dictionary to describe source graph as {node: [node neighbours]}.
- **kwargs** – See the EmbeddingComposite class for additional keyword arguments. Note that find_embedding and embedding_parameters keyword arguments are ignored.

Examples

To embed a triangular problem (a problem with a three-node complete graph, or clique) in the Chimera topology, you need to chain two qubits. This example maps triangular problems to a composed sampler (based on the unstructured ExactSolver) with a Chimera unit-cell structure.

```python
>>> import dimod
>>> import dwave_networkx as dnx
>>> from dwave.system import FixedEmbeddingComposite
... >>> c1 = dnx.chimera_graph(1)
>>> embedding = {'a': [0, 4], 'b': [1], 'c': [5]}
... >>> structured_sampler = dimod.StructureComposite(dimod.ExactSolver(),
... c1.nodes, c1.edges)
... >>> sampler = FixedEmbeddingComposite(structured_sampler, embedding)
... >>> sampler.edgelist
[('a', 'b'), ('a', 'c'), ('b', 'c')]
```

Properties

- `FixedEmbeddingComposite.properties`
- `FixedEmbeddingComposite.parameters`
- `FixedEmbeddingComposite.children`
- **FixedEmbeddingComposite.child** The child sampler.
- `FixedEmbeddingComposite.nodelist` Nodes available to the composed sampler.
- `FixedEmbeddingComposite.edgelist` Edges available to the composed sampler.
- `FixedEmbeddingComposite.adjacency` Adjacency structure for the composed sampler.
- `FixedEmbeddingComposite.structure` Structure of the structured sampler formatted as a namedtuple, Structure(nodelist, edgelist, adjacency), where the 3-tuple values are the nodelist, edgelist and adjacency attributes.
FixedEmbeddingComposite.properties = None

FixedEmbeddingComposite.parameters = None

FixedEmbeddingComposite.children = None

FixedEmbeddingComposite.child
The child sampler. First sampler in Composite.children.
    Type Sampler

FixedEmbeddingComposite.nodelist
    Nodes available to the composed sampler.
    Type list

FixedEmbeddingComposite.edgelist
    Edges available to the composed sampler.
    Type list

FixedEmbeddingComposite.adjacency
    Adjacency structure for the composed sampler.
    Type dict[variable, set]

FixedEmbeddingComposite.structure
    Structure of the structured sampler formatted as a namedtuple, Structure(nodelist, edgelist, adjacency), where the 3-tuple values are the nodelist, edgelist and adjacency attributes.
## Methods

<table>
<thead>
<tr>
<th>Method</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>FixedEmbeddingComposite.sample(bqm, **parameters)</code></td>
<td>Sample the binary quadratic model.</td>
</tr>
<tr>
<td><code>FixedEmbeddingComposite.sample_ising(h, J, ...)</code></td>
<td>Sample from an Ising model using the implemented sample method.</td>
</tr>
<tr>
<td><code>FixedEmbeddingComposite.sample_qubo(Q, ...)</code></td>
<td>Sample from a QUBO using the implemented sample method.</td>
</tr>
</tbody>
</table>

### `dwave.system.composites.FixedEmbeddingComposite.sample`

```python
FixedEmbeddingComposite.sample(bqm, **parameters)
```
Sample the binary quadratic model.

On the first call of a sampling method, finds a minor-embedding for the given binary quadratic model (BQM). All subsequent calls to its sampling methods reuse this embedding.

**Parameters**

- `bqm (dimod.BinaryQuadraticModel)` – Binary quadratic model to be sampled from.
- `chain_strength (float/mapping/callable, optional)` – Magnitude of the quadratic bias (in SPIN-space) applied between variables to form a chain, with the energy penalty of chain breaks set to 2 * `chain_strength`. If a mapping is passed, a chain-specific strength is applied. If a callable is passed, it will be called on `chain_strength(bqm, embedding)` and should return a float or mapping, to be interpreted as above. By default, `chain_strength` is calculated with `uniform_torque_compensation()`.
- `chain_break_method (function, optional)` – Method used to resolve chain breaks during sample unembedding. See `unembed_sampleset()`.
- `chain_break_fraction (bool, optional, default=True)` – Add a ‘chain_break_fraction’ field to the unembedded response with the fraction of chains broken before unembedding.
- `embedding_parameters (dict, optional)` – If provided, parameters are passed to the embedding method as keyword arguments. Overrides any `embedding_parameters` passed to the constructor. Only used on the first call.
- `**parameters` – Parameters for the sampling method, specified by the child sampler.

**Returns** `dimod.SampleSet`

### `dwave.system.composites.FixedEmbeddingComposite.sample_ising`

```python
FixedEmbeddingComposite.sample_ising(h, J, **parameters)
```
Sample from an Ising model using the implemented sample method.

This method is inherited from the `Sampler` base class.

Converts the Ising model into a `BinaryQuadraticModel` and then calls `sample()`.

**Parameters**

- `h (dict/list)` – Linear biases of the Ising problem. If a dict, should be of the form `{v: bias, ...}` where is a spin-valued variable and `bias` is its associated bias. If a list, it is treated as a list of biases where the indices are the variable labels.
• $J(\text{dict}[(\text{variable}, \text{variable}), \text{bias}])$ – Quadratic biases of the Ising problem.

• **kwargs – See the implemented sampling for additional keyword definitions.

Returns SampleSet

See also:
sample(), sample_qubo()

dwave.system.composites.FixedEmbeddingComposite.sample_qubo

FixedEmbeddingComposite.sample_qubo($Q$, **parameters)

Sample from a QUBO using the implemented sample method.

This method is inherited from the Sampler base class.

Converts the QUBO into a BinaryQuadraticModel and then calls sample().

Parameters

• $Q(\text{dict})$ – Coefficients of a quadratic unconstrained binary optimization (QUBO) problem. Should be a dict of the form $\{u, v\}: \text{bias}, \ldots$ where $u, v,$ are binary-valued variables and $\text{bias}$ is their associated coefficient.

• **kwargs – See the implemented sampling for additional keyword definitions.

Returns SampleSet

See also:
sample(), sample_ising()

LazyFixedEmbeddingComposite

class LazyFixedEmbeddingComposite(child_sampler, find_embedding=<function minorminer.find_embedding>, embedding_parameters=None, scale_aware=False, child_structure_search=<function child_structure_dfs>)

Maps problems to the structure of its first given problem.

This composite reuses the minor-embedding found for its first given problem without recalculating a new minor-embedding for subsequent calls of its sampling methods.

Parameters

• child_sampler (dimod.Sampler) – Structured dimod sampler.

• find_embedding (function, default=func:minorminer.find_embedding) – A function find_embedding($S, T$, **kwargs) where $S$ and $T$ are edgelists. The function can accept additional keyword arguments. The function is used to find the embedding for the first problem solved.

• embedding_parameters (dict, optional) – If provided, parameters are passed to the embedding method as keyword arguments.
Examples

```python
>>> from dwave.system import LazyFixedEmbeddingComposite, DWaveSampler
...
>>> sampler = LazyFixedEmbeddingComposite(DWaveSampler())
>>> sampler.nodelist is None  # no structure prior to first sampling
True
>>> __ = sampler.sample_ising({}, {('a', 'b'): -1})
>>> sampler.nodelist  # has structure based on given problem
['a', 'b']
```

Properties

- **LazyFixedEmbeddingComposite.parameters**
  - Nodes available to the composed sampler.
- **LazyFixedEmbeddingComposite.properties**
  - Edges available to the composed sampler.
- **LazyFixedEmbeddingComposite.nodelist**
  - Adjacency structure for the composed sampler.
- **LazyFixedEmbeddingComposite.structure**
  - Structure of the structured sampler formatted as a namedtuple, `Structure(nodelist, edgelist, adjacency)`, where the 3-tuple values are the `nodelist`, `edgelist` and `adjacency` attributes.
LazyFixedEmbeddingComposite.\texttt{adjacency}  
Adacency structure for the composed sampler.  
\textbf{Type} \ \texttt{dict[variable, set]}

LazyFixedEmbeddingComposite.\texttt{structure}  
Structure of the structured sampler formatted as a \texttt{namedtuple}, \texttt{Structure(nodelist, edgelist, adjacency)}, where the 3-tuple values are the \texttt{nodelist}, \texttt{edgelist} and \texttt{adjacency} attributes.

\textbf{Methods}

| LazyFixedEmbeddingComposite.\texttt{sample}(bqm, ...) | Sample the binary quadratic model. |
| LazyFixedEmbeddingComposite.\texttt{sample_ising}(h, ...) | Sample from an Ising model using the implemented sample method. |
| LazyFixedEmbeddingComposite.\texttt{sample_qubo}(Q, ...) | Sample from a QUBO using the implemented sample method. |

\textbf{Parameters}

- \texttt{bqm} (\texttt{dimod.BinaryQuadraticModel}) – Binary quadratic model to be sampled from.
- \texttt{chain_strength} (\texttt{float/mapping/callable, optional}) – Magnitude of the quadratic bias (in SPIN-space) applied between variables to form a chain, with the energy penalty of chain breaks set to 2 * \texttt{chain_strength}. If a mapping is passed, a chain-specific strength is applied. If a callable is passed, it will be called on \texttt{chain_strength(bqm, embedding)} and should return a float or mapping, to be interpreted as above. By default, \texttt{chain_strength} is calculated with \texttt{uniform_torque_compensation()}.  
- \texttt{chain_break_method} (\texttt{function, optional}) – Method used to resolve chain breaks during sample unembedding. See \texttt{unembed_sampleset()}.  
- \texttt{chain_break_fraction} (\texttt{bool, optional, default=True}) – Add a ‘chain_break_fraction’ field to the unembedded response with the fraction of chains broken before unembedding.  
- \texttt{embedding_parameters} (\texttt{dict, optional}) – If provided, parameters are passed to the embedding method as keyword arguments. Overrides any \texttt{embedding_parameters} passed to the constructor. Only used on the first call.  
- **\texttt{parameters} – Parameters for the sampling method, specified by the child sampler.

\textbf{Returns} \texttt{dimod.SampleSet}
LazyFixedEmbeddingComposite.sample_ising(h, J, **parameters)
Sample from an Ising model using the implemented sample method.

This method is inherited from the Sampler base class.

Converts the Ising model into a BinaryQuadraticModel and then calls sample().

Parameters
- **h (dict/list)** – Linear biases of the Ising problem. If a dict, should be of the form `{v: bias, ...}` where is a spin-valued variable and bias is its associated bias. If a list, it is treated as a list of biases where the indices are the variable labels.
- **J (dict[(variable, variable), bias])** – Quadratic biases of the Ising problem.
- ****kwargs – See the implemented sampling for additional keyword definitions.

Returns SampleSet

See also:
sample(), sample_qubo()

dlave.system.composites.LazyFixedEmbeddingComposite.sample_qubo

LazyFixedEmbeddingComposite.sample_qubo(Q, **parameters)
Sample from a QUBO using the implemented sample method.

This method is inherited from the Sampler base class.

Converts the QUBO into a BinaryQuadraticModel and then calls sample().

Parameters
- **Q (dict)** – Coefficients of a quadratic unconstrained binary optimization (QUBO) problem. Should be a dict of the form `{(u, v): bias, ...}` where u, v, are binary-valued variables and bias is their associated coefficient.
- ****kwargs – See the implemented sampling for additional keyword definitions.

Returns SampleSet

See also:
sample(), sample_ising()

dlave.system.composites.TilingComposite

class TilingComposite(sampler, sub_m, sub_n, t=4)
Composite to tile a small problem across a Chimera-structured sampler.

Enables parallel sampling for small problems (problems that are minor-embeddable in a small part of a D-Wave solver’s Chimera graph).

Notation CN refers to a Chimera graph consisting of an NxN grid of unit cells, where each unit cell is a bipartite graph with shores of size t. The D-Wave 2000Q QPU supports a C16 Chimera graph: its 2048 qubits are logically mapped into a 16x16 matrix of unit cell of 8 qubits (t=4).
A problem that can be minor-embedded in a single unit cell, for example, can therefore be tiled across the unit cells of a D-Wave 2000Q as 16x16 duplicates. This enables sampling 256 solutions in a single call.

**Parameters**

- **sampler** (*dimod.Sampler*) – Structured dimod sampler such as a *DWaveSampler*.
- **sub_m** (*int*) – Number of rows of Chimera unit cells for minor-embedding the problem once.
- **sub_n** (*int*) – Number of columns of Chimera unit cells for minor-embedding the problem once.
- **t** (*int, optional, default=4*) – Size of the shore within each Chimera unit cell.

**Examples**

This example submits a two-variable QUBO problem representing a logical NOT gate to a D-Wave system. The QUBO—two nodes with biases of -1 that are coupled with strength 2—needs only any two coupled qubits and so is easily minor-embedded in a single unit cell. Composite TilingComposite tiles it multiple times for parallel solution: the two nodes should typically have opposite values.

```python
>>> from dwave.system import DWaveSampler, EmbeddingComposite
>>> from dwave.system import TilingComposite...
>>> qpu_2000q = DWaveSampler(solver={'topology__type': 'chimera'})
>>> sampler = EmbeddingComposite(TilingComposite(qpu_2000q, 1, 1, 4))
>>> Q = {(1, 1): -1, (1, 2): 2, (2, 1): 0, (2, 2): -1}
>>> sampleset = sampler.sample_qubo(Q)
>>> len(sampleset)> 1
True
```

See Ocean Glossary for explanations of technical terms in descriptions of Ocean tools.

**Properties**

- **TilingComposite.properties** Properties in the form of a dict.
- **TilingComposite.parameters** Parameters in the form of a dict.
- **TilingComposite.children** The single wrapped structured sampler.
- **TilingComposite.child** The child sampler.
- **TilingComposite.nodelist** List of active qubits for the structured solver.
- **TilingComposite.edgelist** List of active couplers for the D-Wave solver.
- **TilingComposite.adjacency** Adjacency structure formatted as a dict, where keys are the nodes of the structured sampler and values are sets of all adjacent nodes for each key node.
- **TilingComposite.structure** Structure of the structured sampler formatted as a namedtuple, *Structure*(nodelist, edgelist, adjacency), where the 3-tuple values are the *nodelist*, edgelist and adjacency attributes.
**TilingComposite**

- **properties**: None
  - Properties in the form of a dict.
  - Type: dict

- **parameters**: None
  - Parameters in the form of a dict.
  - Type: dict[str, list]

- **children**: None
  - The single wrapped structured sampler.
  - Type: list

- **child**: None
  - The child sampler. First sampler in Composite.children.
  - Type: Sampler

- **nodelist**: None
  - List of active qubits for the structured solver.
  - Type: list

- **edgelist**: None
  - List of active couplers for the D-Wave solver.
  - Type: list

- **adjacency**: None
  - Adjacency structure formatted as a dict, where keys are the nodes of the structured sampler and values are sets of all adjacent nodes for each key node.
  - Type: dict[variable, set]
TilingComposite.structure

Structure of the structured sampler formatted as a namedtuple, `Structure(nodelist, edgelist, adjacency)`, where the 3-tuple values are the `nodelist`, `edgelist` and `adjacency` attributes.

Methods

<table>
<thead>
<tr>
<th>Method</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>TilingComposite.sample(bqm, **kwargs)</code></td>
<td>Sample from the specified binary quadratic model.</td>
</tr>
<tr>
<td><code>TilingComposite.sample_ising(h, J, **parameters)</code></td>
<td>Sample from an Ising model using the implemented sample method.</td>
</tr>
<tr>
<td><code>TilingComposite.sample_qubo(Q, **parameters)</code></td>
<td>Sample from a QUBO using the implemented sample method.</td>
</tr>
</tbody>
</table>

TilingComposite.sample

Sample from the specified binary quadratic model.

Parameters

- `bqm` ([dimod.BinaryQuadraticModel]) – Binary quadratic model to be sampled from.
- `**kwargs` – Optional keyword arguments for the sampling method, specified per solver.

Returns `dimod.SampleSet`

Examples

This example submits a simple Ising problem of just two variables on a D-Wave system. Because the problem fits in a single Chimera unit cell, it is tiled across the solver’s entire Chimera graph, resulting in multiple samples (the exact number depends on the working Chimera graph of the D-Wave system).

```python
>>> from dwave.system import DWaveSampler, EmbeddingComposite
>>> from dwave.system import TilingComposite
... qpu_2000q = DWaveSampler(solver={'topology__type': 'chimera'})
>>> sampler = EmbeddingComposite(TilingComposite(qpu_2000q, 1, 1, 4))
>>> response = sampler.sample_ising({}, {('a', 'b'): 1})
>>> len(response)                      # doctest: +SKIP
246
```

See Ocean Glossary for explanations of technical terms in descriptions of Ocean tools.

dwave.system.composites.TilingComposite.sample_qubo

Sample from a QUBO using the implemented sample method.

This method is inherited from the `Sampler` base class.

Converting the Ising model into a `BinaryQuadraticModel` and then calls `sample()`.

Parameters
• **kwargs – See the implemented sampling for additional keyword definitions.

Returns SampleSet

See also:

sample(), sample_qubo()

dwave.system.composites.TilingComposite.sample_qubo

TilingComposite.sample_qubo(Q, **parameters)

Sample from a QUBO using the implemented sample method.

This method is inherited from the Sampler base class.

Converts the QUBO into a BinaryQuadraticModel and then calls sample().

Parameters

• Q(dict) – Coefficients of a quadratic unconstrained binary optimization (QUBO) problem. Should be a dict of the form \((u, v)\colon bias, \ldots\) where \(u, v\) are binary-valued variables and bias is their associated coefficient.

• **kwargs – See the implemented sampling for additional keyword definitions.

Returns SampleSet

See also:

sample(), sample_ising()

VirtualGraphComposite

class VirtualGraphComposite(sampler, embedding, chain_strength=None, flux_biases=None, flux_bias_num_reads=1000, flux_bias_max_age=3600)

Composite to use the D-Wave virtual graph feature for minor-embedding.

Calibrates qubits in chains to compensate for the effects of biases and enables easy creation, optimization, use, and reuse of an embedding for a given working graph.

Inherits from dimod.ComposedSampler and dimod.Structured.

Parameters

• sampler (DWaveSampler) – A dimod dimod.Sampler. Typically a DWaveSampler or derived composite sampler; other samplers may not work or make sense with this composite layer.

• embedding(dict[hashable, iterable]) – Mapping from a source graph to the specified sampler’s graph (the target graph).

• chain_strength (float, optional, default=None) – Desired chain coupling strength. This is the magnitude of couplings between qubits in a chain. If None, uses the maximum available as returned by a SAPI query to the D-Wave solver.
• **flux_biases** *(list/False/None, optional, default=None)* – Per-qubit flux bias offsets in the form of a list of lists, where each sublist is of length 2 and specifies a variable and the flux bias offset associated with that variable. Qubits in a chain with strong negative J values experience a J-induced bias; this parameter compensates by recalibrating to remove that bias. If False, no flux bias is applied or calculated. If None, flux biases are pulled from the database or calculated empirically.

• **flux_bias_num_reads** *(int, optional, default=1000)* – Number of samples to collect per flux bias value to calculate calibration information.

• **flux_bias_max_age** *(int, optional, default=3600)* – Maximum age (in seconds) allowed for a previously calculated flux bias offset to be considered valid.

---

**Attention:** D-Wave’s virtual graphs feature can require many seconds of D-Wave system time to calibrate qubits to compensate for the effects of biases. If your account has limited D-Wave system access, consider using `FixedEmbeddingComposite` instead.

### Examples

This example uses `VirtualGraphComposite` to instantiate a composed sampler that submits a QUBO problem to a D-Wave solver. The problem represents a logical AND gate using penalty function $P = xy - 2(xy + y)z + 3z$, where variables $x$ and $y$ are the gate’s inputs and $z$ the output. This simple three-variable problem is manually minor-embedded to a single Chimera unit cell: variables $x$ and $y$ are represented by qubits 1 and 5, respectively, and $z$ by a two-qubit chain consisting of qubits 0 and 4. The chain strength is set to the maximum allowed found from querying the solver’s extended J range. In this example, the ten returned samples all represent valid states of the AND gate.

```python
>>> from dwave.system import DWaveSampler, VirtualGraphComposite
>>> embedding = {'x': {1}, 'y': {5}, 'z': {0, 4}}
>>> qpu_2000q = DWaveSampler(solver={'topology__type': 'chimera'})
>>> qpu_2000q.properties['extended_j_range']
[-2.0, 1.0]
>>> sampler = VirtualGraphComposite(qpu_2000q, embedding, chain_strength=2)  # doctest: +SKIP
>>> Q = {('x', 'y'): -2, ('x', 'z'): -2, ('y', 'z'): 3}
>>> sampleset = sampler.sample_qubo(Q, num_reads=10)  # doctest: +SKIP
>>> print(sampleset)
['BINARY', 4 rows, 10 samples, 3 variables]
```

See Ocean Glossary for explanations of technical terms in descriptions of Ocean tools.

### Properties

- `VirtualGraphComposite.properties`
- `VirtualGraphComposite.parameters`
- `VirtualGraphComposite.children`

Continued on next page
## dwave.system.composites.VirtualGraphComposite

<table>
<thead>
<tr>
<th>Property</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>child</code></td>
<td>The child sampler.</td>
</tr>
<tr>
<td><code>nodelist</code></td>
<td>Nodes available to the composed sampler.</td>
</tr>
<tr>
<td><code>edgelist</code></td>
<td>Edges available to the composed sampler.</td>
</tr>
<tr>
<td><code>adjacency</code></td>
<td>Adjacency structure for the composed sampler.</td>
</tr>
<tr>
<td><code>structure</code></td>
<td>Structure of the structured sampler formatted as a namedtuple, Structure(nodelist, edgelist, adjacency), where the 3-tuple values are the nodelist, edgelist and adjacency attributes.</td>
</tr>
</tbody>
</table>

### dwave.system.composites.VirtualGraphComposite.properties

VirtualGraphComposite.properties = None

### dwave.system.composites.VirtualGraphComposite.parameters

VirtualGraphComposite.parameters = None

### dwave.system.composites.VirtualGraphComposite.children

VirtualGraphComposite.children = None

### dwave.system.composites.VirtualGraphComposite.child

VirtualGraphComposite.child

The child sampler. First sampler in Composite.children.

Type: Sampler

### dwave.system.composites.VirtualGraphComposite.nodelist

VirtualGraphComposite.nodelist

Nodes available to the composed sampler.

Type: list

### dwave.system.composites.VirtualGraphComposite.edgelist

VirtualGraphComposite.edgelist

Edges available to the composed sampler.

Type: list

### dwave.system.composites.VirtualGraphComposite.adjacency

VirtualGraphComposite.adjacency

Adjacency structure for the composed sampler.

Type: dict[variable, set]
VirtualGraphComposite.structure

Structure of the structured sampler formatted as a namedtuple, Structure(nodelist, edgelist, adjacency), where the 3-tuple values are the nodelist, edgelist and adjacency attributes.

Methods

<table>
<thead>
<tr>
<th>Method</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>VirtualGraphComposite.sample(bqm,...)</td>
<td>Sample from the given Ising model.</td>
</tr>
<tr>
<td>VirtualGraphComposite.sample_ising(h, J,...)</td>
<td>Sample from an Ising model using the implemented sample method.</td>
</tr>
<tr>
<td>VirtualGraphComposite.sample_qubo(Q,...)</td>
<td>Sample from a QUBO using the implemented sample method.</td>
</tr>
</tbody>
</table>

dwave.system.composites.VirtualGraphComposite.sample

Sample from the given Ising model.

Parameters

- **h** (list/dict) – Linear biases of the Ising model. If a list, the list’s indices are used as variable labels.
- **J** (dict of (int, int) – float): Quadratic biases of the Ising model.
- **apply_flux_bias_offsets** (bool, optional) – If True, use the calculated flux_bias offsets (if available).
- ****kwargs – Optional keyword arguments for the sampling method, specified per solver.

Examples

This example uses VirtualGraphComposite to instantiate a composed sampler that submits an Ising problem to a D-Wave solver. The problem represents a logical NOT gate using penalty function $P = xy$, where variable $x$ is the gate’s input and $y$ the output. This simple two-variable problem is manually minor-embedded to a single Chimera unit cell: each variable is represented by a chain of half the cell’s qubits, $x$ as qubits 0, 1, 4, 5, and $y$ as qubits 2, 3, 6, 7. The chain strength is set to half the maximum allowed found from querying the solver’s extended J range. In this example, the ten returned samples all represent valid states of the NOT gate.

```python
>>> from dwave.system import DWaveSampler, VirtualGraphComposite
>>> embedding = {'x': {0, 4, 1, 5}, 'y': {2, 6, 3, 7}}
>>> qpu_2000q = DWaveSampler(solver={'topology__type': 'chimera'})
>>> qpu_2000q.properties['extended_j_range']
[-2.0, 1.0]
>>> sampler = VirtualGraphComposite(qpu_2000q, embedding, chain_strength=1) # doctest: +SKIP
>>> h = {}
>>> J = {('x', 'y'): 1}
>>> sampleset = sampler.sample_ising(h, J, num_reads=10) # doctest: +SKIP
>>> print(sampleset) # doctest: +SKIP
x y energy num_oc. chain_
0 -1 +1 -1.0 6 0.0
|continues on next page|
Reverse Anneal

Composites that do batch operations for reverse annealing based on sets of initial states or anneal schedules.
ReverseBatchStatesComposite

class ReverseBatchStatesComposite(child_sampler)
   Composite that reverse anneals from multiple initial samples. Each submission is independent from one another.

   Parameters sampler(dimod.Sampler) – A dimod sampler.

Examples

This example runs 100 reverse anneals each from two initial states on a problem constructed by setting random ±1 values on a clique (complete graph) of 15 nodes, minor-embedded on a D-Wave system using the DWaveCliqueSampler sampler.

   >>> import dimod
   >>> from dwave.system import DWaveCliqueSampler, ReverseBatchStatesComposite
   ...  
   >>> sampler = DWaveCliqueSampler()  # doctest: +SKIP
   >>> sampler_reverse = ReverseBatchStatesComposite(sampler)  # doctest: +SKIP
   >>> schedule = [[0.0, 1.0], [10.0, 0.5], [20, 1.0]]
   ...  
   >>> bqm = dimod.generators.ran_r(1, 15)
   >>> init_samples = [{i: -1 for i in range(15)}, {i: 1 for i in range(15)}]
   >>> sampleset = sampler_reverse.sample(bqm,
   ...               anneal_schedule=schedule,
   ...               initial_states=init_samples,
   ...               num_reads=100,
   ...               reinitialize_state=True)  # doctest: +SKIP

Properties

<table>
<thead>
<tr>
<th>ReverseBatchStatesComposite.child</th>
<th>The child sampler.</th>
</tr>
</thead>
<tbody>
<tr>
<td>ReverseBatchStatesComposite.children</td>
<td>List of child samplers that are used by this composite.</td>
</tr>
<tr>
<td>ReverseBatchStatesComposite.properties</td>
<td>A dict containing any additional information about the sampler.</td>
</tr>
<tr>
<td>ReverseBatchStatesComposite.parameters</td>
<td>A dict where keys are the keyword parameters accepted by the sampler methods and values are lists of the properties relevent to each parameter.</td>
</tr>
</tbody>
</table>

1.2. Reference Documentation
ReverseBatchStatesComposite.properties

A dict containing any additional information about the sampler.

Type dict

ReverseBatchStatesComposite.parameters

A dict where keys are the keyword parameters accepted by the sampler methods and values are lists of the properties relevant to each parameter.

Type dict

Methods

<table>
<thead>
<tr>
<th>Method</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>ReverseBatchStatesComposite.sample(bqm, ...)</td>
<td>Sample the binary quadratic model using reverse annealing from multiple initial states.</td>
</tr>
<tr>
<td>ReverseBatchStatesComposite.sample_ising(h, ...)</td>
<td>Sample from an Ising model using the implemented sample method.</td>
</tr>
<tr>
<td>ReverseBatchStatesComposite.sample_qubo(Q, ...)</td>
<td>Sample from a QUBO using the implemented sample method.</td>
</tr>
</tbody>
</table>

ReverseBatchStatesComposite.sample(bqm, **parameters)

Sample the binary quadratic model using reverse annealing from multiple initial states.

Parameters

- **bqm** *(dimod.BinaryQuadraticModel)* – Binary quadratic model to be sampled from.
- **parameters** – Parameters for the sampling method, specified by the child sampler.

Returns *dimod.SampleSet* that has initial_state field.

Examples

This example runs 100 reverse anneals each from two initial states on a problem constructed by setting random ±1 values on a clique (complete graph) of 15 nodes, minor-embedded on a D-Wave system using the DWaveCliqueSampler sampler.

```python
>>> import dimod
>>> from dwave.system import DWaveCliqueSampler, ReverseBatchStatesComposite...
>>> sampler = DWaveCliqueSampler()    # doctest: +SKIP
>>> sampler_reverse = ReverseBatchStatesComposite(sampler)    # doctest: +SKIP
>>> schedule = [[0.0, 1.0], [10.0, 0.5], [20, 1.0]]
...    # doctest: +SKIP
>>> bqm = dimod.generators.ran_r(1, 15)
>>> init_samples = [{i: -1 for i in range(15)}, {i: 1 for i in range(15)}]
>>> sampleset = sampler_reverse.sample(bqm,
```
ReverseBatchStatesComposite.sample_ising(h, J, **parameters)

Sample from an Ising model using the implemented sample method.

This method is inherited from the Sampler base class.

Converts the Ising model into a BinaryQuadraticModel and then calls sample().

Parameters

- **h (dict/list)** – Linear biases of the Ising problem. If a dict, should be of the form \{v: bias, ...\} where is a spin-valued variable and bias is its associated bias. If a list, it is treated as a list of biases where the indices are the variable labels.

- **J (dict[(variable, variable), bias])** – Quadratic biases of the Ising problem.

- ****kwargs – See the implemented sampling for additional keyword definitions.

Returns SampleSet

See also: sample(), sample_qubo()

dwave.system.composites.ReverseBatchStatesComposite.sample_qubo

ReverseBatchStatesComposite.sample_qubo(Q, **parameters)

Sample from a QUBO using the implemented sample method.

This method is inherited from the Sampler base class.

Converts the QUBO into a BinaryQuadraticModel and then calls sample().

Parameters

- **Q (dict)** – Coefficients of a quadratic unconstrained binary optimization (QUBO) problem. Should be a dict of the form \{(u, v): bias, ...\} where u, v, are binary-valued variables and bias is their associated coefficient.

- ****kwargs – See the implemented sampling for additional keyword definitions.

Returns SampleSet

See also: sample(), sample_ising()

ReverseAdvanceComposite

class ReverseAdvanceComposite(child_sampler)

Composite that reverse anneals an initial sample through a sequence of anneal schedules.
If you do not specify an initial sample, a random sample is used for the first submission. By default, each subsequent submission selects the most-found lowest-energy sample as its initial state. If you set reinitialize_state to False, which makes each submission behave like a random walk, the subsequent submission selects the last returned sample as its initial state.

**Parameters**

- **sampler** (*dimod.Sampler*) — A dimod sampler.

**Examples**

This example runs 100 reverse anneals each for three schedules on a problem constructed by setting random ±1 values on a clique (complete graph) of 15 nodes, minor-embedded on a D-Wave system using the *DWaveCliqueSampler* sampler.

```python
>>> import dimod
>>> from dwave.system import DWaveCliqueSampler, ReverseAdvanceComposite
...
>>> sampler = DWaveCliqueSampler()  # doctest: +SKIP
>>> sampler_reverse = ReverseAdvanceComposite(sampler)  # doctest: +SKIP
>>> schedule = [[[0.0, 1.0], [t, 0.5], [20, 1.0]] for t in (5, 10, 15)]
...
>>> bqm = dimod.generators.ran_r(1, 15)
>>> init_samples = {i: -1 for i in range(15)}
>>> sampleset = sampler_reverse.sample(bqm,
...     anneal_schedules=schedule,
...     initial_state=init_samples,
...     num_reads=100,
...     reinitialize_state=True)  # doctest: +SKIP
```

**Properties**

- **ReverseAdvanceComposite.child** — The child sampler.
- **ReverseAdvanceComposite.children** — List of child samplers that that are used by this composite.
- **ReverseAdvanceComposite.properties** — A dict containing any additional information about the sampler.
- **ReverseAdvanceComposite.parameters** — A dict where keys are the keyword parameters accepted by the sampler methods and values are lists of the properties relevant to each parameter.
**Type** list[Sampler]

**dwave.system.composites.ReverseAdvanceComposite.properties**

ReverseAdvanceComposite.properties

A dict containing any additional information about the sampler.

**Type** dict

**dwave.system.composites.ReverseAdvanceComposite.parameters**

ReverseAdvanceComposite.parameters

A dict where keys are the keyword parameters accepted by the sampler methods and values are lists of the properties relevant to each parameter.

**Type** dict

**Methods**

<table>
<thead>
<tr>
<th>Method</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>ReverseAdvanceComposite.sample(bqm[, ...])</td>
<td>Sample the binary quadratic model using reverse annealing along a given set of anneal schedules.</td>
</tr>
<tr>
<td>ReverseAdvanceComposite.sample_ising(h, J, ... )</td>
<td>Sample from an Ising model using the implemented sample method.</td>
</tr>
<tr>
<td>ReverseAdvanceComposite.sample_qubo(Q, ... )</td>
<td>Sample from a QUBO using the implemented sample method.</td>
</tr>
</tbody>
</table>

**dwave.system.composites.ReverseAdvanceComposite.sample**

ReverseAdvanceComposite.sample (bqm, anneal_schedules=None, **parameters)

Sample the binary quadratic model using reverse annealing along a given set of anneal schedules.

**Parameters**

- **bqm** (dimod.BinaryQuadraticModel) – Binary quadratic model to be sampled from.
- **anneal_schedules** (list of lists) – Anneal schedules in order of submission. Each schedule is formatted as a list of [time, s] pairs
- **initial_state** (dict, optional) – the state to reverse anneal from. If not provided, it will be randomly generated
- **parameters** – Parameters for the sampling method, specified by the child sampler.

**Returns** dimod.SampleSet that has initial_state and schedule_index fields.

**Examples**

This example runs 100 reverse anneals each for three schedules on a problem constructed by setting random \( \pm 1 \) values on a clique (complete graph) of 15 nodes, minor-embedded on a D-Wave system using the DWaveCliqueSampler sampler.
>>> import dimod
>>> from dwave.system import DWaveCliqueSampler, ReverseAdvanceComposite
... >>> sampler = DWaveCliqueSampler() # doctest: +SKIP
>>> sampler_reverse = ReverseAdvanceComposite(sampler) # doctest: +SKIP
... >>> schedule = [[(0.0, 1.0), [t, 0.5], [20, 1.0]] for t in (5, 10, 15)]
... >>> bqm = dimod.generators.ran_r(1, 15)
>>> init_samples = {i: -1 for i in range(15)}
>>> sampleset = sampler_reverse.sample(bqm,
... anneal_schedules=schedule,
... initial_state=init_samples,
... num_reads=100,
... reinitialize_state=True) # doctest: +SKIP

**dWave.system.composites.ReverseAdvanceComposite.sample_ising**

ReverseAdvanceComposite.sample_ising(h, J, **parameters)
Sample from an Ising model using the implemented sample method.

This method is inherited from the Sampler base class.

Converts the Ising model into a BinaryQuadraticModel and then calls sample().

Parameters

- **h (dict/list)** - Linear biases of the Ising problem. If a dict, should be of the form \{v: bias, ...\} where is a spin-valued variable and bias is its associated bias. If a list, it is treated as a list of biases where the indices are the variable labels.

- **J (dict[(variable, variable), bias])** - Quadratic biases of the Ising problem.

- ****kwargs** - See the implemented sampling for additional keyword definitions.

Returns SampleSet

See also:
sample(), sample_qubo()

**dWave.system.composites.ReverseAdvanceComposite.sample_qubo**

ReverseAdvanceComposite.sample_qubo(Q, **parameters)
Sample from a QUBO using the implemented sample method.

This method is inherited from the Sampler base class.

Converts the QUBO into a BinaryQuadraticModel and then calls sample().

Parameters

- **Q (dict)** - Coefficients of a quadratic unconstrained binary optimization (QUBO) problem. Should be a dict of the form \{(u, v): bias, ...\} where u, v, are binary-valued variables and bias is their associated coefficient.

- ****kwargs** - See the implemented sampling for additional keyword definitions.

Returns SampleSet
See also:

\( \text{sample()}, \text{sample\_ising()} \)

### 1.2.3 Embedding

Provides functions that map binary quadratic models and samples between a source graph and a target graph. For an introduction to minor-embedding, see Minor-Embedding.

#### Generators

Tools for finding embeddings.

#### Generic

**minorminer** is a heuristic tool for minor embedding: given a minor and target graph, it tries to find a mapping that embeds the minor into the target.

\[
\text{minorminer.find\_embedding}(S, T, **params) \quad \text{Heuristically attempt to find a minor-embedding of a graph representing an Ising/QUBO into a target graph.}
\]

**minorminer.find\_embedding**

\[
\text{find\_embedding}(S, T, **params) \quad \text{Heuristically attempt to find a minor-embedding of a graph representing an Ising/QUBO into a target graph.}
\]

**Args:**

- \( S \): an iterable of label pairs representing the edges in the source graph, or a \( \rightarrow \) NetworkX Graph
- \( T \): an iterable of label pairs representing the edges in the target graph, or a \( \rightarrow \) NetworkX Graph
- **params** (optional): see below

**Returns:**

When \( \text{return\_overlap} = \text{False} \) (the default), returns a dict that maps labels in \( S \) to lists of labels in \( T \).

- If the heuristic fails to find an embedding, an empty dictionary is returned

When \( \text{return\_overlap} = \text{True} \), returns a tuple consisting of a dict that maps labels in \( S \) to lists of labels in \( T \) and a bool indicating whether or not a valid embedding was found

When interrupted by Ctrl-C, returns the best embedding found so far

Note that failure to return an embedding does not prove that no embedding exists

**Optional parameters:**
max_no_improvement: Maximum number of failed iterations to improve the current solution, where each iteration attempts to find an embedding for each variable of S such that it is adjacent to all its neighbours. Integer >= 0 (default = 10)

random_seed: Seed for the random number generator that find_embedding uses. Integer >=0 (default is randomly set)

timeout: Algorithm gives up after timeout seconds. Number >= 0 (default is approximately 1000 seconds, stored as a double)

max_beta: Qubits are assigned weight according to a formula \( (\beta^n) \) where \( n \) is the number of chains containing that qubit. This value should never be less than or equal to 1. (default is effectively infinite, stored as a double)

tries: Number of restart attempts before the algorithm stops. On D-WAVE 2000Q, a typical restart takes between 1 and 60 seconds. Integer >= 0 (default = 10)

inner_rounds: the algorithm takes at most this many iterations between restart attempts; restart attempts are typically terminated due to max_no_improvement. Integer >= 0 (default = effectively infinite)

chainlength_patience: Maximum number of failed iterations to improve chainlengths in the current solution, where each iteration attempts to find an embedding for each variable of S such that it is adjacent to all its neighbours. Integer >= 0 (default = 10)

max_fill: Restricts the number of chains that can simultaneously incorporate the same qubit during the search. Integer >= 0, values above 63 are treated as 63 (default = effectively infinite)

threads: Maximum number of threads to use. Note that the parallelization is only advantageous where the expected degree of variables is significantly greater than the number of threads. Integer >= 1 (default = 1)

return_overlap: This function returns an embedding whether or not qubits are used by multiple variables. Set this value to 1 to capture both return values to determine whether or not the returned embedding is valid. Logical 0/1 integer (default = 0)

skip_initialization: Skip the initialization pass. Note that this only works if the chains passed in through initial_chains and fixed_chains are semi-valid. A semi-valid embedding is a collection of chains such that every adjacent pair of variables \((u,v)\) has a coupler \((p,q)\) in the hardware graph where \(p\) is in chain\((u)\) and \(q\) is in chain\((v)\). This can be used on a valid embedding to immediately skip to the chainlength improvement phase. Another good source of semi-valid embeddings is the output of this function with the return_overlap parameter enabled. Logical 0/1 integer (default = 0)

verbose: Level of output verbosity. Integer < 4 (default = 0). When set to 0, the output is quiet until the final result. When set to 1, output looks like this:
initialized
max qubit fill 3; num maxfull qubits=3
embedding trial 1
max qubit fill 2; num maxfull qubits=21
embedding trial 2
embedding trial 3
embedding trial 4
embedding trial 5
embedding found.
max chain length 4; num max chains=1
reducing chain lengths
max chain length 3; num max chains=5

When set to 2, outputs the information for lower levels and also
reports progress on minor statistics (when searching for an
embedding, this is when the number of maxfull qubits decreases;
when improving, this is when the number of max chains decreases)
When set to 3, report before each before each pass. Look here when
tweaking `tries`, `inner_rounds`, and `chainlength_patience`
When set to 4, report additional debugging information. By default,
this package is built without this functionality. In the c++
headers, this is controlled by the CPPDEBUG flag

Detailed explanation of the output information:
max qubit fill: largest number of variables represented in a qubit
num maxfull: the number of qubits that has max overfill
max chain length: largest number of qubits representing a single variable
num max chains: the number of variables that has max chain size

interactive: If `logging` is None or False, the verbose output will be printed
to stdout/stderr as appropriate, and keyboard interrupts will stop the
embedding
process and the current state will be returned to the user. Otherwise, output
will be directed to the logger `logging.getLogger(minorminer.__name__)` and
keyboard interrupts will be propagated back to the user. Errors will use
`logger.error()`, verbosity levels 1 through 3 will use `logger.info()` and
level
4 will use `logger.debug()`. bool, default False

initial_chains: Initial chains inserted into an embedding before
fixed_chains are placed, which occurs before the initialization
pass. These can be used to restart the algorithm in a similar state
to a previous embedding; for example, to improve chainlength of a
valid embedding or to reduce overlap in a semi-valid embedding (see
skip_initialization) previously returned by the algorithm. Missing
or empty entries are ignored. A dictionary, where initial_chains[i]
is a list of qubit labels.

fixed_chains: Fixed chains inserted into an embedding before the
initialization pass. As the algorithm proceeds, these chains are not
allowed to change, and the qubits used by these chains are not used by
other chains. Missing or empty entries are ignored. A dictionary, where
fixed_chains[i] is a list of qubit labels.

restrict_chains: Throughout the algorithm, we maintain the condition
that chain[i] is a subset of restrict_chains[i] for each i, except
those with missing or empty entries. A dictionary, where
restrict_chains[i] is a list of qubit labels.
suspend_chains: This is a metafeature that is only implemented in the Python interface. suspend_chains[i] is an iterable of iterables; for example
suspend_chains[i] = [blob_1, blob_2],
with each blob_j an iterable of target node labels.
this enforces the following:
  for each suspended variable i,
    for each blob_j in the suspension of i,
      at least one qubit from blob_j will be contained in the
      chain for i

we accomplish this through the following problem transformation
  for each iterable blob_j in suspend_chains[i],
    * add an auxiliary node Zij to both source and target graphs
    * set fixed_chains[Zij] = [Zij]
    * add the edge (i,Zij) to the source graph
    * add the edges (q,Zij) to the target graph for each q in blob_j

Chimera

Minor-embedding in Chimera-structured target graphs.

<table>
<thead>
<tr>
<th>Function</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>chimera.find_clique_embedding(k, m[, n, t, target_edges=None])</td>
<td>Find an embedding for a clique in a Chimera graph.</td>
</tr>
<tr>
<td>chimera.find_biclique_embedding(a, b, m[, n, t])</td>
<td>Find an embedding for a biclique in a Chimera graph.</td>
</tr>
<tr>
<td>chimera.find_grid_embedding(dim, m[, n, t])</td>
<td>Find an embedding for a grid in a Chimera graph.</td>
</tr>
</tbody>
</table>

**find_clique_embedding** (k, m=None, n=None, t=None, target_edges=None)

Find an embedding for a clique in a Chimera graph.

Given the node labels or size of a clique (fully connected graph) and size or edges of the target Chimera graph, attempts to find an embedding.

**Parameters**

- k (int/iterable) – Clique to embed. If k is an integer, generates an embedding for a clique of size k labelled [0,k-1]. If k is an iterable of nodes, generates an embedding for a clique of size len(k) labelled for the given nodes.
- m (int) – Number of rows in the Chimera lattice.
- n (int, optional, default=m) – Number of columns in the Chimera lattice.
- t (int, optional, default 4) – Size of the shore within each Chimera tile.
- target_edges (iterable[edge]) – A list of edges in the target Chimera graph. Nodes are labelled as returned by chimera_graph().

**Returns** An embedding mapping a clique to the Chimera lattice.

**Return type** dict
Examples

The first example finds an embedding for a $K_4$ complete graph in a single Chimera unit cell. The second for an alphanumerically labeled $K_3$ graph in 4 unit cells.

```python
>>> from dwave.embedding.chimera import find_clique_embedding
... >>> embedding = find_clique_embedding(4, 1, 1)
>>> embedding
# doctest: +SKIP
{0: [4, 0], 1: [5, 1], 2: [6, 2], 3: [7, 3]}
```

```python
>>> from dwave.embedding.chimera import find_clique_embedding
... >>> embedding = find_clique_embedding(['a', 'b', 'c'], m=2, n=2, t=4)
>>> embedding
# doctest: +SKIP
{'a': [20, 16], 'b': [21, 17], 'c': [22, 18]}
```

dwave.embedding.chimera.find_biclique_embedding

find_biclique_embedding\((a, b, m, n=None, t=None, target_edges=None)\)

Find an embedding for a biclique in a Chimera graph.

Given a biclique (a bipartite graph where every vertex in a set in connected to all vertices in the other set) and a target Chimera graph size or edges, attempts to find an embedding.

Parameters

- **a** (int/iterable) – Left shore of the biclique to embed. If a is an integer, generates an embedding for a biclique with the left shore of size a labelled [0,a-1]. If a is an iterable of nodes, generates an embedding for a biclique with the left shore of size len(a) labelled for the given nodes.
- **b** (int/iterable) – Right shore of the biclique to embed. If b is an integer, generates an embedding for a biclique with the right shore of size b labelled [0,b-1]. If b is an iterable of nodes, generates an embedding for a biclique with the right shore of size len(b) labelled for the given nodes.
- **m** (int) – Number of rows in the Chimera lattice.
- **n** (int, optional, default=m) – Number of columns in the Chimera lattice.
- **t** (int, optional, default 4) – Size of the shore within each Chimera tile.
- **target_edges** (iterable[edge]) – A list of edges in the target Chimera graph. Nodes are labelled as returned by chimera_graph()?

Returns

A 2-tuple containing:

- dict: An embedding mapping the left shore of the biclique to the Chimera lattice.
- dict: An embedding mapping the right shore of the biclique to the Chimera lattice.

Return type tuple

Examples

This example finds an embedding for an alphanumerically labeled biclique in a single Chimera unit cell.
>>> from dwave.embedding.chimera import find_biclique_embedding
...
>>> left, right = find_biclique_embedding(['a', 'b', 'c'], ['d', 'e'], 1, 1)
>>> print(left, right)  # doctest: +SKIP
{'a': [4], 'b': [5], 'c': [6]} {'d': [0], 'e': [1]}

dwave.embedding.chimera.find_grid_embedding

find_grid_embedding(dim, m=None, n=None, t=4)
Find an embedding for a grid in a Chimera graph.

Given grid dimensions and a target Chimera graph size, attempts to find an embedding.

Parameters

- dim (iterable[int]) – Sizes of each grid dimension. Length can be between 1 and 3.
- m (int) – Number of rows in the Chimera lattice.
- n (int, optional, default=m) – Number of columns in the Chimera lattice.
- t (int, optional, default 4) – Size of the shore within each Chimera tile.

Returns
An embedding mapping a grid to the Chimera lattice.

Return type  dict

Examples

This example finds an embedding for a 2x3 grid in a 12x12 lattice of Chimera unit cells.

```python
>>> from dwave.embedding.chimera import find_grid_embedding
...
>>> embedding = find_grid_embedding([2, 3], m=12, n=12, t=4)
>>> embedding
# doctest: +SKIP
{(0, 0): [0, 4],
 (0, 1): [8, 12],
 (0, 2): [16, 20],
 (1, 0): [24, 28],
 (1, 1): [36, 40],
 (1, 2): [48, 52]}
```

Pegasus

Minor-embedding in Pegasus-structured target graphs.

pegasus.find_clique_embedding(k[, m, ...])  
Find an embedding for a clique in a Pegasus graph.

dwave.embedding.pegasus.find_clique_embedding

find_clique_embedding(k, m=None, target_graph=None)
Find an embedding for a clique in a Pegasus graph.

Given a clique (fully connected graph) and target Pegasus graph, attempts to find an embedding by transforming
the Pegasus graph into a $K_{2,2}$ Chimera graph and then applying a Chimera clique-finding algorithm. Results are converted back to Pegasus coordinates.

**Parameters**

- $k$ (int/iterable/networkx.Graph) – A complete graph to embed, formatted as a number of nodes, node labels, or a NetworkX graph.
- $m$ (int) – Number of tiles in a row of a square Pegasus graph. Required to generate an $m$-by-$m$ Pegasus graph when $target_graph$ is None.
- $target_graph$ (networkx.Graph) – A Pegasus graph. Required when $m$ is None.

**Returns** An embedding as a dict, where keys represent the clique’s nodes and values, formatted as lists, represent chains of pegasus coordinates.

**Return type** dict

**Examples**

This example finds an embedding for a $K_3$ complete graph in a 2-by-2 Pegasus graph.

```python
>>> from dwave.embedding.pegasus import find_clique_embedding
...
>>> print(find_clique_embedding(3, 2))  # doctest: +SKIP
{0: [10, 34], 1: [35, 11], 2: [32, 12]}
```

**Utilities**

- embed_bqm(source_bqm[, embedding, ...]) Embed a binary quadratic model onto a target graph.
- embed_ising(source_h, source_J, embedding, ...) Embed an Ising problem onto a target graph.
- embed_qubo(source_Q, embedding, target_adjacency) Embed a QUBO onto a target graph.
- unembed_sampleset(target_sampleset, ...) Unembed a sample set.

**embed_bqm**

Embed a binary quadratic model onto a target graph.

**Parameters**

- $source_bqm$ (BinaryQuadraticModel) – Binary quadratic model to embed.
- $embedding$ (dict/EmbeddedStructure) – Mapping from source graph to target graph as a dict of form $\{s: \{t, \ldots\}, \ldots\}$, where $s$ is a source-model variable and $t$ is a target-model variable. Alternately, an EmbeddedStructure object produced by, for example, EmbeddedStructure(target_adjacency.edges(), embedding). If embedding is a dict, target_adjacency must be provided.
- $target_adjacency$ (dict/networkx.Graph, optional) – Adjacency of the target graph as a dict of form $\{t: Nt, \ldots\}$, where $t$ is a variable in the target graph and $Nt$ is its set of neighbours. This should be omitted if and only if embedding is an EmbeddedStructure object.
- $chain_strength$ (float/mapping/callable, optional) – Magnitude of the
quadratic bias (in SPIN-space) applied between variables to form a chain, with the energy penalty of chain breaks set to 2 * chain_strength. If a mapping is passed, a chain-specific strength is applied. If a callable is passed, it will be called on chain_strength(source_bqm, embedding) and should return a float or mapping, to be interpreted as above. By default, chain_strength is calculated with uniform_torque_compensation().

- **smear_vartype** (Vartype, optional, default=None) – Determines whether the linear bias of embedded variables is smeared (the specified value is evenly divided as biases of a chain in the target graph) in SPIN or BINARY space. Defaults to the Vartype of source_bqm.

**Returns** Target binary quadratic model.

**Return type** BinaryQuadraticModel

**Examples**

This example embeds a triangular binary quadratic model representing a $K_3$ clique into a square target graph by mapping variable $c$ in the source to nodes 2 and 3 in the target.

```python
>>> import networkx as nx
...
>>> target = nx.cycle_graph(4)
>>> # Binary quadratic model for a triangular source graph
>>> h = {'a': 0, 'b': 0, 'c': 0}
>>> J = {('a', 'b'): 1, ('b', 'c'): 1, ('a', 'c'): 1}
>>> bqm = dimod.BinaryQuadraticModel.from_ising(h, J)
>>> # Variable c is a chain
>>> embedding = {'a': {0}, 'b': {1}, 'c': {2, 3}}
>>> # Embed and show the chain strength
>>> target_bqm = dwave.embedding.embed_bqm(bqm, embedding, target)
>>> print(target_bqm.quadratic)
# doctest: +SKIP
{(0, 1): 1.0, (0, 3): 1.0, (1, 2): 1.0, (2, 3): -1.9996979771955565}
```

See also:

embed_ising(), embed_qubo()

dwave.embedding.embed_ising

**embed_ising** (source_h, source_J, embedding, target_adjacency, chain_strength=None)

Embed an Ising problem onto a target graph.

**Parameters**

- **source_h** (dict[variable, bias]/list[bias]) – Linear biases of the Ising problem. If a list, the list’s indices are used as variable labels.
- **source_J** (dict[(variable, variable), bias]) – Quadratic biases of the Ising problem.
- **embedding** (dict) – Mapping from source graph to target graph as a dict of form {s: {t, ... }, ... }, where s is a source-model variable and t is a target-model variable.
- **target_adjacency** (dict/networkx.Graph) – Adjacency of the target graph as a dict of form {t: Nt, ... }, where t is a target-graph variable and Nt is its set of neighbours.
• `chain_strength(float/mapping/callable, optional)` — Magnitude of the quadratic bias (in SPIN-space) applied between variables to form a chain, with the energy penalty of chain breaks set to $2 \times \text{chain_strength}$. If a mapping is passed, a chain-specific strength is applied. If a callable is passed, it will be called on `chain_strength(source_bqm, embedding)` and should return a float or mapping, to be interpreted as above. By default, `chain_strength` is calculated with `uniform_torque_compensation()`.

Returns

A 2-tuple:

- dict[variable, bias]: Linear biases of the target Ising problem.
- dict[(variable, variable), bias]: Quadratic biases of the target Ising problem.

Return type: tuple

Examples

This example embeds a triangular Ising problem representing a $K_3$ clique into a square target graph by mapping variable $c$ in the source to nodes 2 and 3 in the target.

```python
>>> import networkx as nx
... >>> target = nx.cycle_graph(4)
>>> # Ising problem biases
... >>> h = {'a': 0, 'b': 0, 'c': 0}
>>> >>> J = {('a', 'b'): 1, ('b', 'c'): 1, ('a', 'c'): 1}
>>> >>> # Variable c is a chain
... >>> embedding = {'a': {0}, 'b': {1}, 'c': {2, 3}}
>>> >>> # Embed and show the resulting biases
... >>> th, tJ = dwave.embedding.embed_ising(h, J, embedding, target)
>>> >>> th # doctest: +SKIP
{0: 0.0, 1: 0.0, 2: 0.0, 3: 0.0}
>>> >>> tJ # doctest: +SKIP
{(0, 1): 1.0, (0, 3): 1.0, (1, 2): 1.0, (2, 3): -1.0}
```

See also:

`embed_bqm()`, `embed_qubo()`

dwave.embedding.embed_qubo

`embed_qubo(source_Q, embedding, target_adjacency, chain_strength=None)`

Embed a QUBO onto a target graph.

Parameters

- `source_Q(dict[(variable, variable), bias])` — Coefficients of a quadratic unconstrained binary optimization (QUBO) model.
- `embedding(dict)` — Mapping from source graph to target graph as a dict of form `{s: {t, ...}, ...}`, where $s$ is a source-model variable and $t$ is a target-model variable.
- `target_adjacency(dict/networkx.Graph)` — Adjacency of the target graph as a dict of form `{t: Nt, ...}`, where $t$ is a target-graph variable and $Nt$ is its set of neighbours.
- `chain_strength(float/mapping/callable, optional)` — Magnitude of the quadratic bias (in SPIN-space) applied between variables to form a chain, with the energy
penalty of chain breaks set to 2 * chain_strength. If a mapping is passed, a chain-specific strength is applied. If a callable is passed, it will be called on chain_strength(source_bqm, embedding) and should return a float or mapping, to be interpreted as above. By default, chain_strength is calculated with uniform_torque_compensation().

Returns  
Quadratic biases of the target QUBO.

Return type  
dict[(variable, variable), bias]

Examples

This example embeds a triangular QUBO representing a $K_3$ clique into a square target graph by mapping variable $c$ in the source to nodes 2 and 3 in the target.

```python
>>> import networkx as nx
... >>> target = nx.cycle_graph(4)
... >>> # QUBO
... >>> Q = {('a', 'b'): 1, ('b', 'c'): 1, ('a', 'c'): 1}
... >>> # Variable c is a chain
... >>> embedding = { 'a': {0}, 'b': {1}, 'c': {2, 3} }
... >>> # Embed and show the resulting biases
... >>> tQ = dwave.embedding.embed_qubo(Q, embedding, target)
... >>> tQ  # doctest: +SKIP
{ (0, 1): 1.0,  
(0, 3): 1.0,  
(1, 2): 1.0,  
(2, 3): -4.0,  
(0, 0): 0.0,  
(1, 1): 0.0,  
(2, 2): 2.0,  
(3, 3): 2.0 }
```

See also:

embed_bqm(), embed_ising()
and a new field called “chain_break_method” specifying the index of the method is appended to the sample set. Defaults to `majority_vote()`. See `dwave.embedding.chain_breaks`.

- **chain_break_fraction** *(bool, optional, default=False)*—Add a `chain_break_fraction` field to the unembedded `dimod.SampleSet` with the fraction of chains broken before unembedding.

- **return_embedding** *(bool, optional, default=False)*—If True, the embedding is added to `dimod.SampleSet.info` of the returned sample set. Note that if an `embedding` key already exists in the sample set then it is overwritten.

**Returns** Sample set in the source BQM.

**Return type** `SampleSet`

**Examples**

This example unembeds from a square target graph samples of a triangular source BQM.

```python
>>> # Triangular binary quadratic model and an embedding
>>> J = {('a', 'b'): -1, ('b', 'c'): -1, ('a', 'c'): -1}
>>> bqm = dimod.BinaryQuadraticModel.from_ising({}, J)
>>> embedding = {'a': [0, 1], 'b': [2], 'c': [3]}
>>> # Samples from the embedded binary quadratic model
>>> samples = {0: -1, 1: -1, 2: -1, 3: -1}, # [0, 1] is unbroken
...   {0: -1, 1: +1, 2: +1, 3: +1} # [0, 1] is broken
>>> energies = [-3, 1]
>>> embedded = dimod.SampleSet.from_samples(samples, dimod.SPIN, energies)
>>> # Unembed
>>> samples = dwave.embedding.unembed_sampleset(embedded, embedding, bqm)
```

**Diagnostics**

<table>
<thead>
<tr>
<th>Function</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>chain_break_frequency</code></td>
<td>Determine the frequency of chain breaks in the given samples.</td>
</tr>
<tr>
<td><code>diagnose_embedding</code></td>
<td>Diagnose a minor embedding.</td>
</tr>
<tr>
<td><code>is_valid_embedding</code></td>
<td>A simple (bool) diagnostic for minor embeddings.</td>
</tr>
<tr>
<td><code>verify_embedding</code></td>
<td>A simple (exception-raising) diagnostic for minor embeddings.</td>
</tr>
</tbody>
</table>

**`dwave.embedding.chain_break_frequency`**

**chain_break_frequency** *(samples_like, embedding)*

Determine the frequency of chain breaks in the given samples.

**Parameters**

- **samples_like** *(samples_like/dimod.SampleSet)*—A collection of raw samples. 'samples_like' is an extension of NumPy's array_like. See `dimod.as_samples()`.

- **embedding** *(dict)*—Mapping from source graph to target graph as a dict of form `{s: [t,`
where \( s \) is a source-model variable and \( t \) is a target-model variable.

**Returns** Frequency of chain breaks as a dict in the form \( \{ s: f, \ldots \} \), where \( s \) is a variable in the source graph and float \( f \) the fraction of broken chains.

**Return type** dict

**Examples**

This example embeds a single source node, ‘a’, as a chain of two target nodes (0, 1) and uses `chain_break_frequency()` to show that out of two synthetic samples, one ([-1, +1]) represents a broken chain.

```python
>>> import numpy as np
>>> samples = np.array([[-1, +1], [+1, +1]])
>>> embedding = {'a': {0, 1}}
>>> print(dwave.embedding.chain_break_frequency(samples, embedding)['a'])
0.5
```

dwave.embedding.diagnose_embedding

diagnose_embedding \( (emb, source, target) \)

Diagnose a minor embedding.

Produces a generator that lists all issues with the embedding. User-friendly variants of this function are `is_valid_embedding()`, which returns a bool, and `verify_embedding()`, which raises the first observed error.

**Parameters**

- **emb** (dict) – A mapping of source nodes to arrays of target nodes as a dict of form \( \{ s: [t, \ldots], \ldots \} \), where \( s \) is a source-graph variable and \( t \) is a target-graph variable.
- **source** (list/networkx.Graph) – Graph to be embedded as a NetworkX graph or a list of edges.
- **target** (list/networkx.Graph) – Graph being embedded into as a NetworkX graph or a list of edges.

**Yields** Errors yielded in the form `ExceptionClass, arg1, arg2,\ldots`, where the arguments following the class are used to construct the exception object, which are subclasses of `EmbeddingError`.

- **MissingChainError**, snode: a source node label that does not occur as a key of `emb`, or for which `emb[snod]` is empty.
- **ChainOverlapError**, tnode, snode0, snode1: a target node which occurs in both `emb[snod0]` and `emb[snod1]`.
- **DisconnectedChainError**, snode: a source node label whose chain is not a connected subgraph of `target`.
- **InvalidNodeError**, tnode, snode: a source node label and putative target node label that is not a node of `target`.
- **MissingEdgeError**, snode0, snode1: a pair of source node labels defining an edge that is not present between their chains.
Examples

This example diagnoses an invalid embedding from a triangular source graph to a square target graph. A valid embedding, such as \( \text{emb} = \{0: \{1\}, 1: \{0\}, 2: \{2, 3\}\} \), yields no errors.

```python
>>> from dwave.embedding import diagnose_embedding
>>> import networkx as nx

>>> source = nx.complete_graph(3)
>>> target = nx.cycle_graph(4)

>>> embedding = {0: [2], 1: [1, 'a'], 2: [2, 3]}

>>> diagnosis = diagnose_embedding(embedding, source, target)
>>> for problem in diagnosis:
...     print(problem)
... # doctest: +SKIP
...     # for testing purpose only

(<class 'dwave.embedding.exceptions.InvalidNodeError'>, 1, 'a')
(<class 'dwave.embedding.exceptions.ChainOverlapError'>, 2, 2, 0)
```

**dwave.embedding.is_valid_embedding**

**is_valid_embedding**(\(\text{emb}\), \(\text{source}\), \(\text{target}\))

A simple (bool) diagnostic for minor embeddings.

See **diagnose_embedding()** for a more detailed diagnostic and more information.

**Parameters**

- \(\text{emb} (\text{dict})\) – A mapping of source nodes to arrays of target nodes as a dict of form \(\{s: [t, \ldots], \ldots\}\), where \(s\) is a source-graph variable and \(t\) is a target-graph variable.
- \(\text{source} (\text{graph or edgelist})\) – Graph to be embedded.
- \(\text{target} (\text{graph or edgelist})\) – Graph being embedded into.

**Returns** True if \(\text{emb}\) is valid.

**Return type** bool

**dwave.embedding.verify_embedding**

**verify_embedding**(\(\text{emb}\), \(\text{source}\), \(\text{target}\), \(\text{ignore_errors}=()\))

A simple (exception-raising) diagnostic for minor embeddings.

See **diagnose_embedding()** for a more detailed diagnostic and more information.

**Parameters**

- \(\text{emb} (\text{dict})\) – A mapping of source nodes to arrays of target nodes as a dict of form \(\{s: [t, \ldots], \ldots\}\), where \(s\) is a source-graph variable and \(t\) is a target-graph variable.
- \(\text{source} (\text{graph or edgelist})\) – Graph to be embedded.
- \(\text{target} (\text{graph or edgelist})\) – Graph being embedded into.

**Raises** EmbeddingError – A catch-all class for the following errors:

- MissingChainError: A key is missing from \(\text{emb}\) or the associated chain is empty.
- ChainOverlapError: Two chains contain the same target node.
- DisconnectedChainError: A chain is disconnected.
- InvalidNodeError: A chain contains a node label not found in \(\text{target}\).
MissingEdgeError: A source edge is not represented by any target edges.

**Returns** True if no exception is raised.

**Return type** bool

## Chain Strength

Utility functions for calculating chain strength.

### Examples

This example uses `uniform_torque_compensation()`, given a prefactor of 2, to calculate a chain strength that `EmbeddingComposite` then uses.

```python
>>> from functools import partial
>>> from dwave.system import EmbeddingComposite, DWaveSampler
>>> from dwave.embedding.chain_strength import uniform_torque_compensation
... >>> Q = {(0,0): 1, (1,1): 1, (2,3): 2, (1,2): -2, (0,3): -2}
>>> sampler = EmbeddingComposite(DWaveSampler())
>>> # partial() can be used when the BQM or embedding is not accessible
>>> chain_strength = partial(uniform_torque_compensation, prefactor=2)
>>> sampleset = sampler.sample_qubo(Q, chain_strength=chain_strength, return_embedding=True)
>>> sampleset.info['embedding_context']['chain_strength']
1.224744871391589
```

### chain_strength.uniform_torque_compensation

Chain strength that attempts to compensate for torque that would break the chain.

#### chain_strength.scaled(bqm[, embedding, prefactor=1.414])

Chain strength that is scaled to the problem bias range.

### dwave.embedding.chain_strength.uniform_torque_compensation

`uniform_torque_compensation(bqm, embedding=None, prefactor=1.414)`

Chain strength that attempts to compensate for torque that would break the chain.

The RMS of the problem’s quadratic biases is used for calculation.

**Parameters**

- `bqm` (BinaryQuadraticModel) – A binary quadratic model.
- `embedding` (dict/EmbeddedStructure, default=None) – Included to satisfy the `chain_strength` callable specifications for `embed_bqm`.
- `prefactor` (float, optional, default=1.414) – Prefactor used for scaling. For non-pathological problems, the recommended range of prefactors to try is [0.5, 2].

**Returns** The chain strength, or 1 if chain strength is not applicable.

**Return type** float
**dwave.embedding.chain_strength.scaled**

**scaled(bqm, embedding=None, prefactor=1.0)**

Chain strength that is scaled to the problem bias range.

**Parameters**

- **bqm** (*BinaryQuadraticModel*) – A binary quadratic model.
- **embedding** (*dict/EmbeddedStructure, default=None*) – Included to satisfy the `chain_strength` callable specifications for `embed_bqm`.
- **prefactor** (*float, optional, default=1.0*) – Prefactor used for scaling.

**Returns**

The chain strength, or 1 if chain strength is not applicable.

**Return type** float

---

**Chain-Break Resolution**

Unembedding samples with broken chains.

**Generators**

<table>
<thead>
<tr>
<th>chain_breaks.discard(samples, chains)</th>
<th>Discard broken chains.</th>
</tr>
</thead>
<tbody>
<tr>
<td>chain_breaks.majority_vote(samples, chains)</td>
<td>Unembed samples using the most common value for broken chains.</td>
</tr>
<tr>
<td>chain_breaks.weighted_random(samples, chains)</td>
<td>Unembed samples using weighed random choice for broken chains.</td>
</tr>
</tbody>
</table>

---

**dwave.embedding.chain_breaks.discard**

**discard(samples, chains)**

Discard broken chains.

**Parameters**

- **samples** (*samples_like*) – A collection of samples. *samples_like* is an extension of NumPy’s `array_like`. See `dimod.as_samples()`.

- **chains** (*list[array_like]*) – List of chains, where each chain is an array_like collection of the variables in the same order as their representation in the given samples.

**Returns**

A 2-tuple containing:

- numpy.ndarray: Unembedded samples as an array of dtype 'int8'. Broken chains are discarded.
- numpy.ndarray: Indices of rows with unbroken chains.

**Return type** tuple
Examples

This example unembeds two samples that chains nodes 0 and 1 to represent a single source node. The first sample has an unbroken chain, the second a broken chain.

```python
>>> import dimod
>>> import numpy as np
...
>>> chains = [(0, 1), (2,)]
>>> samples = np.array([[1, 1, 0], [1, 0, 0]], dtype=np.int8)
>>> unembedded, idx = dwave.embedding.discard(samples, chains)
>>> unembedded
array([[1, 0]], dtype=int8)
>>> print(idx)
[0]
```

dwave.embedding.chain_breaks.majority_vote

**majority_vote** *(samples, chains)*

Unembed samples using the most common value for broken chains.

**Parameters**

- **samples** *(samples_like)* – A collection of samples. *samples_like* is an extension of NumPy’s array_like. See *dimod.as_samples()*.
- **chains** *(list[array_like]*) – List of chains, where each chain is an array_like collection of the variables in the same order as their representation in the given samples.

**Returns**

A 2-tuple containing:

- numpy.ndarray: Unembedded samples as an nS-by-nC array of dtype ‘int8’, where nC is the number of chains and nS the number of samples. Broken chains are resolved by setting the sample value to that of most the chain’s elements or, for chains without a majority, an arbitrary value.
- numpy.ndarray: Indicies of the samples. Equivalent to np.arange(nS) because all samples are kept and none added.

**Return type**  tuple

**Examples**

This example unembeds samples from a target graph that chains nodes 0 and 1 to represent one source node and nodes 2, 3, and 4 to represent another. Both samples have one broken chain, with different majority values.

```python
>>> import dimod
>>> import numpy as np
...
>>> chains = [(0, 1), (2, 3, 4)]
>>> samples = np.array([[1, 1, 0, 0, 1], [1, 1, 1, 0, 1]], dtype=np.int8)
>>> unembedded, idx = dwave.embedding.majority_vote(samples, chains)
>>> print(unembedded)
[[1 0]
 [1 1]]
```
>>> print(idx)
[0 1]

dwave.embedding.chain_breaks.weighted_random

**weighted_random** *(samples, chains)*
Unembed samples using weighed random choice for broken chains.

**Parameters**

- **samples** *(samples_like)* – A collection of samples. *samples_like* is an extension of NumPy's array_like. See `dimod.as_samples()`.
- **chains** *(list[array_like]*) – List of chains, where each chain is an array_like collection of the variables in the same order as their Representation in the given samples.

**Returns**
A 2-tuple containing:

- `numpy.ndarray`: Unembedded samples as an nS-by-nC array of dtype ‘int8’, where nC is the number of chains and nS the number of samples. Broken chains are resolved by setting the sample value to a random value weighted by frequency of the value in the chain.
- `numpy.ndarray`: Indices of the samples. Equivalent to `np.arange(nS)` because all samples are kept and no samples are added.

**Return type** *tuple*

**Examples**

This example unembeds samples from a target graph that chains nodes 0 and 1 to represent one source node and nodes 2, 3, and 4 to represent another. The sample has broken chains for both source nodes.

```python
>>> import dimod
>>> import numpy as np
... >>> chains = [(0, 1), (2, 3, 4)]
>>> samples = np.array([[1, 0, 1, 0, 1]], dtype=np.int8)
>>> unembedded, idx = dwave.embedding.weighted_random(samples, chains)  # doctest: +SKIP
... array([[1, 1]], dtype=int8)
>>> idx  # doctest: +SKIP
array([0, 1])
```

**Callable Objects**

| `chain_breaks.MinimizeEnergy(bqm, embedding)` | Unembed samples by minimizing local energy for broken chains. |
class MinimizeEnergy(bqm, embedding)

Unembed samples by minimizing local energy for broken chains.

Parameters

- **bqm** (*BinaryQuadraticModel*) – Binary quadratic model associated with the source graph.
- **embedding** (*dict*) – Mapping from source graph to target graph as a dict of form `{s: [t, ...], ...}`, where s is a source-model variable and t is a target-model variable.

Examples

This example embeds from a triangular graph to a square graph, chaining target-nodes 2 and 3 to represent source-node c, and unembeds minimizing the energy for the samples. The first two sample have unbroken chains, the second two have broken chains.

```python
>>> import dimod
>>> import numpy as np...
>>> h = {'a': 0, 'b': 0, 'c': 0}
>>> J = {('a', 'b'): 1, ('b', 'c'): 1, ('a', 'c'): 1}
>>> bqm = dimod.BinaryQuadraticModel.from_ising(h, J)
>>> embedding = {'a': [0], 'b': [1], 'c': [2, 3]}
>>> cbm = dwave.embedding.MinimizeEnergy(bqm, embedding)
>>> samples = np.array([[+1, -1, +1, +1], ...
... [-1, -1, -1, -1], ...
... [-1, -1, +1, -1], ...
... [+1, +1, -1, +1]], dtype=np.int8)
>>> chains = [embedding['a'], embedding['b'], embedding['c']]
>>> unembedded, idx = cbm(samples, chains)
>>> unembedded
array([[ 1, -1, 1],
       [-1, -1, -1],
       [-1, -1, 1],
       [ 1, 1, -1]], dtype=int8)
>>> idx
array([0, 1, 2, 3])
```

__init__(bqm, embedding)

Initialize self. See help(type(self)) for accurate signature.

Methods

__init__(bqm, embedding) Initialize self.

Exceptions

exceptions.EmbeddingError Base class for all embedding exceptions.

exceptions.MissingChainError(snode) Raised if a node in the source graph has no associated chain.
Table 38 – continued from previous page

exceptions.ChainOverlapError(tnode, snode0, . . .)  Raised if two source nodes have an overlapping chain.
exceptions.DisconnectedChainError(snode)  Raised if a chain is not connected in the target graph.
exceptions.InvalidNodeError(snode, tnode)  Raised if a chain contains a node not in the target graph.
exceptions.MissingEdgeError(snode0, snode1)  Raised when two source nodes sharing an edge to not have a corresponding edge between their chains.

`dwave.embedding.exceptions.EmbeddingError`

**exception EmbeddingError**

Base class for all embedding exceptions.

`dwave.embedding.exceptions.MissingChainError`

**exception MissingChainError** (snode)

Raised if a node in the source graph has no associated chain.

**Parameters**

- **snode** – The source node with no associated chain.

`dwave.embedding.exceptions.ChainOverlapError`

**exception ChainOverlapError** (tnode, snode0, snode1)

Raised if two source nodes have an overlapping chain.

**Parameters**

- **tnode** – Location where the chains overlap.
- **snode0** – First source node with overlapping chain.
- **snode1** – Second source node with overlapping chain.

`dwave.embedding.exceptions.DisconnectedChainError`

**exception DisconnectedChainError** (snode)

Raised if a chain is not connected in the target graph.

**Parameters**

- **snode** – The source node associated with the broken chain.

`dwave.embedding.exceptions.InvalidNodeError`

**exception InvalidNodeError** (snode, tnode)

Raised if a chain contains a node not in the target graph.

**Parameters**

- **snode** – The source node associated with the chain.
- **tnode** – The node in the chain not in the target graph.
dwave.embedding.exceptions.MissingEdgeError

**exception MissingEdgeError** *(sn0, sn1)*

Raised when two source nodes sharing an edge to not have a corresponding edge between their chains.

**Parameters**

- **sn0** – First source node.
- **sn1** – Second source node.

**Classes**

**class EmbeddedStructure** *(target_edges, embedding)*

Processes an embedding and a target graph to collect target edges into those within individual chains, and those that connect chains. This is used elsewhere to embed binary quadratic models into the target graph.

**Parameters**

- **target_edges** *(iterable[edge])* – An iterable of edges in the target graph. Each edge should be an iterable of 2 hashable objects.
- **embedding** *(dict)* – Mapping from source graph to target graph as a dict of form `{s: {t, ...}, ...}`, where `s` is a source-model variable and `t` is a target-model variable.

This class is a dict, and acts as an immutable duplicate of embedding.

### 1.2.4 Utilities

Utility functions.

---

`common_working_graph(graph0, graph1)`

Creates a graph using the common nodes and edges of two given graphs.

**Parameters**

- **graph0** – *(dict[dict]/Graph)* A NetworkX graph or a dictionary of dictionaries adjacency representation.
- **graph1** – *(dict[dict]/Graph)* A NetworkX graph or a dictionary of dictionaries adjacency representation.

**Returns** A graph with the nodes and edges common to both input graphs.

**Return type** Graph
Examples

This example creates a graph that represents a quarter (4 by 4 Chimera tiles) of a particular D-Wave system’s working graph.

```python
>>> import dwave_networkx as dnx
>>> from dwave.system import DWaveSampler, common_working_graph
... >>> sampler = DWaveSampler()
>>> C4 = dnx.chimera_graph(4)  # a 4x4 lattice of Chimera tiles
>>> c4_working_graph = common_working_graph(C4, sampler.adjacency)
```

1.2.5 Warnings

- **class WarningAction**
  An enumeration.
- **class ChainBreakWarning**
- **class ChainLengthWarning**
- **class TooFewSamplesWarning**
- **class ChainStrengthWarning**
  Base category for warnings about the embedding chain strength.
- **class EnergyScaleWarning**
  Base category for warnings about the relative bias strengths.
- **class WarningHandler**(action=\texttt{None})

1.3 Installation

Installation from PyPI:

```bash
pip install dwave-system
```

Installation from PyPI with drivers:

**Note:** Prior to v0.3.0, running `pip install dwave-system` installed a driver dependency called `dwave-drivers` (previously also called `dwave-system-tuning`). This dependency has a restricted license and has been made optional as of v0.3.0, but is highly recommended. To view the license details:

```python
from dwave.drivers import __license__
print(__license__)
```

To install with optional dependencies:

```bash
pip install dwave-system[drivers] --extra-index-url https://pypi.dwavesys.com/simple
```

Installation from source:

```bash
pip install -r requirements.txt
python setup.py install
```
Note that installing from source installs `dwave-drivers`. To uninstall the proprietary components:

```
pip uninstall dwave-drivers
```

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d

dwave.embedding.chain_breaks, 65
dwave.embedding.chain_strength, 64
dwave.system.composites.embedding, 23
dwave.system.utilities, 70
Symbols

__init__() (MinimizeEnergy method), 68

A

adjacency (DWaveSampler attribute), 7
adjacency (FixedEmbeddingComposite attribute), 31
adjacency (LazyFixedEmbeddingComposite attribute), 35
adjacency (TilingComposite attribute), 38
adjacency (VirtualGraphComposite attribute), 42
AutoEmbeddingComposite (class in dwave.system.composites), 24

C

chain_break_frequency() (in module dwave.embedding), 61
ChainBreakWarning (class in dwave.system.warnings), 71
ChainLengthWarning (class in dwave.system.warnings), 71
ChainOverlapError, 69
ChainStrengthWarning (class in dwave.system.warnings), 71
child (AutoEmbeddingComposite attribute), 24
child (CutOffComposite attribute), 19
child (EmbeddingComposite attribute), 27
child (FixedEmbeddingComposite attribute), 31
child (PolyCutOffComposite attribute), 22
child (ReverseAdvanceComposite attribute), 48
child (ReverseBatchStatesComposite attribute), 45
child (TilingComposite attribute), 38
child (VirtualGraphComposite attribute), 42
children (CutOffComposite attribute), 19
children (FixedEmbeddingComposite attribute), 31
children (PolyCutOffComposite attribute), 22
children (ReverseAdvanceComposite attribute), 48
children (ReverseBatchStatesComposite attribute), 45
children (TilingComposite attribute), 38
children (VirtualGraphComposite attribute), 42
common_working_graph() (in module dwave.system.utilities), 70
CutOffComposite (class in dwave.system.composites), 18

D

diagnose_embedding() (in module dwave.embedding), 62
discard() (in module dwave.embedding.chain_breaks), 65
DisconnectedChainError, 69
dwave.embedding.chain_breaks (module), 65
dwave.embedding.chain_strength (module), 64
dwave.system.composites.embedding (module), 23
dwave.system.utilities (module), 70
DWaveCliqueSampler (class in dwave.system.samplers), 10
DWaveSampler (class in dwave.system.samplers), 4

E

edgelist (DWaveSampler attribute), 7
edgelist (FixedEmbeddingComposite attribute), 31
edgelist (LazyFixedEmbeddingComposite attribute), 34
edgelist (TilingComposite attribute), 38
edgelist (VirtualGraphComposite attribute), 42
embed_bqm() (in module dwave.embedding), 57
embed_ising() (in module dwave.embedding), 58
embed_qubo() (in module dwave.embedding), 59
EmbeddedStructure (class in dwave.embedding), 70
EmbeddingComposite (class in dwave.system.composites), 26
EmbeddingError, 69
EnergyScaleWarning (class in dwave.system.warnings), 71

F

find_biclique_embedding() (in module dwave.system.composites.embedding), 70
find_clique_embedding() (in module dwave.embedding.chimera), 54
find_clique_embedding() (in module dwave.embedding.pegasus), 56
find_embedding() (in module minorminer), 51
find_grid_embedding() (in module dwave.embedding.chimera), 56
FixedEmbeddingComposite (class in dwave.system.composites), 30
InvalidNodeError, 69
is_valid_embedding() (in module dwave.system.composites), 63

largest_clique() (DWaveCliqueSampler method), 12
largest_clique_size (DWaveCliqueSampler attribute), 11
LazyFixedEmbeddingComposite (class in dwave.system.composites), 33
LeapHybridDQMSampler (class in dwave.system.samplers), 16
LeapHybridSampler (class in dwave.system.samplers), 13

majority_vote() (in module dwave.embedding.chain_breaks), 66
MinimizeEnergy (class in dwave.embedding.chain_breaks), 68
MissingChainError, 69
MissingEdgeError, 70

nodelist (DWaveSampler attribute), 6
nodelist (FixedEmbeddingComposite attribute), 31
nodelist (LazyFixedEmbeddingComposite attribute), 34
nodelist (TilingComposite attribute), 38
nodelist (VirtualGraphComposite attribute), 42

parameters (AutoEmbeddingComposite attribute), 24
parameters (CutOffComposite attribute), 19
parameters (DWaveCliqueSampler attribute), 11
parameters (DWaveSampler attribute), 6
parameters (EmbeddingComposite attribute), 27
parameters (FixedEmbeddingComposite attribute), 31
parameters (LazyFixedEmbeddingComposite attribute), 34
parameters (LeapHybridDQMSampler attribute), 16
parameters (LeapHybridSampler attribute), 14
parameters (PolyCutOffComposite attribute), 22
parameters (ReverseAdvanceComposite attribute), 49
parameters (ReverseBatchStatesComposite attribute), 46
parameters (TilingComposite attribute), 38
parameters (VirtualGraphComposite attribute), 42

return_embedding_default (EmbeddingComposite attribute), 28
ReverseAdvanceComposite (class in dwave.system.composites), 47
ReverseBatchStatesComposite (class in dwave.system.composites), 45

sample() (AutoEmbeddingComposite method), 25
sample() (CutOffComposite method), 19
sample() (DWaveCliqueSampler method), 12
sample() (DWaveSampler method), 8
sample() (EmbeddingComposite method), 28
sample() (FixedEmbeddingComposite method), 32
sample() (LazyFixedEmbeddingComposite method), 35
sample() (LeapHybridSampler method), 14
sample() (ReverseAdvanceComposite method), 49
sample() (ReverseBatchStatesComposite method), 46
sample() (TilingComposite method), 39
sample() (VirtualGraphComposite method), 43
sample_dqm() (LeapHybridDQMSampler method), 17
sample_hising() (PolyCutOffComposite method), 23
sample_hubo() (PolyCutOffComposite method), 23
sample_ising() (AutoEmbeddingComposite method), 25
sample_ising() (CutOffComposite method), 20
sample_ising() (DWaveCliqueSampler method), 12
sample_ising() (DWaveSampler method), 8
sample_ising() (EmbeddingComposite method), 29
sample_ising() (FixedEmbeddingComposite method), 32
sample_ising() (LazyFixedEmbeddingComposite method), 36
sample_ising() (LeapHybridSampler method), 15
sample_ising() (ReverseAdvanceComposite method), 50
sample_ising() (ReverseBatchStatesComposite method), 47
sample_ising() (TilingComposite method), 39
sample_ising() (VirtualGraphComposite method), 44
sample_poly() (PolyCutOffComposite method), 22
sample_qubo() (AutoEmbeddingComposite method), 26
sample_qubo() (CutOffComposite method), 20
sample_qubo() (DWaveCliqueSampler method), 13
sample_qubo() (DWaveSampler method), 9
sample_qubo() (EmbeddingComposite method), 29
sample_qubo() (FixedEmbeddingComposite method), 33
sample_qubo() (LazyFixedEmbeddingComposite method), 36
sample_qubo() (LeapHybridSampler method), 16
sample_qubo() (ReverseAdvanceComposite method), 50
sample_qubo() (ReverseBatchStatesComposite method), 47
sample_qubo() (TilingComposite method), 40
sample_qubo() (VirtualGraphComposite method), 44
scaled() (in module dwave.embedding.chain_strength), 65
structure (DWaveSampler attribute), 7
structure (FixedEmbeddingComposite attribute), 31
structure (LazyFixedEmbeddingComposite attribute), 35
structure (TilingComposite attribute), 39
structure (VirtualGraphComposite attribute), 43

T
TilingComposite (class in dwave.system.composites), 36
to_networkx_graph() (DWaveSampler method), 10
TooFewSamplesWarning (class in dwave.system.warnings), 71

U
unembed_sampleset() (in module dwave.embedding), 60
uniform_torque_compensation() (in module dwave.embedding.chain_strength), 64

V
validate_anneal_schedule() (DWaveSampler method), 9
verify_embedding() (in module dwave.embedding), 63
VirtualGraphComposite (class in dwave.system.composites), 40

W
WarningAction (class in dwave.system.warnings), 71
WarningHandler (class in dwave.system.warnings), 71
warnings_default (EmbeddingComposite attribute), 28
weighted_random() (in module dwave.embedding.chain_breaks), 67