

# ODSCWest23\_ex1

October 25, 2023

```
[114]: # First example notebook for the ODSC West 2023 Workshop:  
# https://odsc.com/speakers/  
→using-graphs-for-large-feature-engineering-pipelines/
```

```
[188]: import datetime  
  
import pandas as pd  
  
from graphreduce.node import GraphReduceNode  
from graphreduce.graph_reduce import GraphReduce  
from graphreduce.enum import ComputeLayerEnum, PeriodUnit
```

## 1 defining a node

```
[189]: class CustomerNode(GraphReduceNode):  
    def do_annotate(self):  
        self.df[self.colabbr('name_length')] = self.df[self.colabbr('name')].  
→apply(lambda x: len(x))  
  
    def do_filters(self):  
        pass  
  
    def do_normalize(self):  
        pass  
  
    def do_post_join_annotate(self):  
        pass  
  
    def do_reduce(self, reduce_key, *args, **kwargs):  
        pass  
  
    def do_labels(self, reduce_key, *args, **kwargs):  
        pass
```

```
[212]: class OrderNode(GraphReduceNode):  
    def do_annotate(self):
```

```

        pass

    def do_filters(self):
        pass

    def do_normalize(self):
        pass

    def do_post_join_annotate(self):
        pass

    def do_reduce(self, reduce_key):
        return self.prep_for_features().groupby(self.colabbr(reduce_key)).agg(
            **{
                self.colabbr(f'{self.pk}_count') : pd.NamedAgg(column=self.
↪colabbr(self.pk), aggfunc='count'),
                self.colabbr(f'amount_sum') : pd.NamedAgg(column=self.
↪colabbr('amount'), aggfunc='sum')
            }
        ).reset_index()

    def do_labels(self, reduce_key):
        return self.prep_for_labels().groupby(self.colabbr(reduce_key)).agg(
            **{
                self.colabbr(f'{self.pk}_had_order') : pd.NamedAgg(column=self.
↪colabbr(self.pk), aggfunc='count')
            }
        ).reset_index()

```

## 2 Instantiate the node

```

[191]: cust = CustomerNode(
        pk='id',
        prefix='cust',
        fpath='dat/cust.csv',
        fmt='csv',
        compute_layer=ComputeLayerEnum.pandas,
    )

```

2023-10-25 18:25:50 [warning ] no `date\_key` set for <GraphReduceNode: fpath=dat/cust.csv fmt=csv>

```

[192]: cust.do_data()

```

```

[193]: cust.df

```

```
[193]:      cust_id cust_name
      0         1      wes
      1         2     john
```

```
[ ]:
```

```
[ ]:
```

```
[194]: order = OrderNode(
        pk='id',
        prefix='order',
        fpath='dat/orders.csv',
        fmt='csv',
        compute_layer=ComputeLayerEnum.pandas,
    )
```

```
2023-10-25 18:25:52 [warning ] no `date_key` set for <GraphReduceNode:
fpath=dat/orders.csv fmt=csv>
```

```
[195]: order.do_data()
```

```
[196]: order.df
```

```
[196]:      order_id  order_customer_id  order_ts  order_amount
      0         1                  1  2023-05-12          10.0
      1         2                  1  2023-06-01          11.5
      2         3                  2  2023-01-01         100.0
      3         4                  2  2022-08-05         150.0
      4         5                  1  2023-07-01         325.0
      5         6                  2  2023-07-02          23.0
      6         7                  1  2023-07-14        12000.0
```

### 3 Run operations

```
[197]: order.do_annotate()
```

```
[198]: # pre-annotate
      cust.df
```

```
[198]:      cust_id cust_name
      0         1      wes
      1         2     john
```

```
[199]: cust.do_annotate()
```

```
[200]: cust.df
```

```
[200]:
```

	cust_id	cust_name	cust_name_length
0	1	wes	3
1	2	john	4

```
[ ]:
```

## 4 Handling time

```
[201]: len(order.df)
```

```
[201]: 7
```

```
[202]: len(order.prep_for_features())
```

```
[202]: 7
```

```
[213]: # we didn't provide a date key or date information
order = OrderNode(
    pk='id',
    prefix='order',
    fpath='dat/orders.csv',
    fmt='csv',
    compute_layer=ComputeLayerEnum.pandas,
    compute_period_val=365,
    compute_period_unit=PeriodUnit.day,
    label_period_val=30,
    label_period_unit=PeriodUnit.day,
    cut_date=datetime.datetime(2023, 6, 1),
    date_key='ts'
)
```

```
[214]: order.do_data()
```

```
[215]: print(len(order.df))
```

```
7
```

```
[216]: print(len(order.prep_for_features()))
```

```
4
```

```
[217]: order.prep_for_features()
```

```
[217]:
```

	order_id	order_customer_id	order_ts	order_amount
0	1	1	2023-05-12	10.0
1	2	1	2023-06-01	11.5

2	3	2	2023-01-01	100.0
3	4	2	2022-08-05	150.0

```
[ ]:
```

```
[218]: order.prep_for_labels()
```

```
[218]:   order_id  order_customer_id  order_ts  order_amount
      4         5              1  2023-07-01         325.0
```

```
[ ]:
```

## 5 Adding operations to a node.

```
[209]: order.do_reduce('customer_id')
```

```
[209]:   order_customer_id  order_id_count
      0                 1              2
      1                 2              2
```

```
[219]: # let's add a sum of the order amount
      order.do_reduce('customer_id')
```

```
[219]:   order_customer_id  order_id_count  order_amount_sum
      0                 1              2              21.5
      1                 2              2             250.0
```

```
[ ]:
```

```
[ ]:
```

## 6 Constructing a graph.

```
[220]: help(GraphReduce)
```

Help on class GraphReduce in module graphreduce.graph\_reduce:

```
class GraphReduce(networkx.classes.digraph.DiGraph)
|   GraphReduce(name: str = 'graph_reduce', parent_node:
Optional[graphreduce.node.GraphReduceNode] = None, fmt: str = 'parquet',
compute_layer: graphreduce.enum.ComputeLayerEnum = None, cut_date:
datetime.datetime = datetime.datetime(2023, 10, 25, 16, 41, 54, 934704),
compute_period_val: Union[int, float] = 365, compute_period_unit:
graphreduce.enum.PeriodUnit = <PeriodUnit.day: 'day'>, has_labels: bool = False,
label_period_val: Union[int, float, NoneType] = None, label_period_unit:
```

```

Optional[graphreduce.enum.PeriodUnit] = None, spark_sqlctx:
pyspark.sql.context.SQLContext = None, feature_function: Optional[str] = None,
dynamic_propagation: bool = False, type_func_map: Dict[str, List[str]] =
{'int64': ['min', 'max', 'sum'], 'str': ['first'], 'object': ['first'],
'float64': ['min', 'max', 'sum'], 'bool': ['first'], 'datetime64': ['first']},
storage_client: Optional[graphreduce.storage.StorageClient] = None, *args,
**kwargs)
|
|   Method resolution order:
|       GraphReduce
|       networkx.classes.digraph.DiGraph
|       networkx.classes.graph.Graph
|       builtins.object
|
|   Methods defined here:
|
|   __init__(self, name: str = 'graph_reduce', parent_node:
Optional[graphreduce.node.GraphReduceNode] = None, fmt: str = 'parquet',
compute_layer: graphreduce.enum.ComputeLayerEnum = None, cut_date:
datetime.datetime = datetime.datetime(2023, 10, 25, 16, 41, 54, 934704),
compute_period_val: Union[int, float] = 365, compute_period_unit:
graphreduce.enum.PeriodUnit = <PeriodUnit.day: 'day'>, has_labels: bool = False,
label_period_val: Union[int, float, NoneType] = None, label_period_unit:
Optional[graphreduce.enum.PeriodUnit] = None, spark_sqlctx:
pyspark.sql.context.SQLContext = None, feature_function: Optional[str] = None,
dynamic_propagation: bool = False, type_func_map: Dict[str, List[str]] =
{'int64': ['min', 'max', 'sum'], 'str': ['first'], 'object': ['first'],
'float64': ['min', 'max', 'sum'], 'bool': ['first'], 'datetime64': ['first']},
storage_client: Optional[graphreduce.storage.StorageClient] = None, *args,
**kwargs)
|       Constructor for GraphReduce
|
|       Args:
|           name : the name of the graph reduce
|           parent_node : parent-most node in the graph, if doing reductions the
granularity to which to reduce the data
|           fmt : the format of the dataset
|           compute_layer : compute layer to use (e.g., spark)
|           cut_date : the date to cut off history
|           compute_period_val : the amount of time to consider during the
compute job
|           compute_period_unit : the unit for the compute period value (e.g.,
day)
|           has_labels : whether or not the compute job computes labels, when
True `prep_for_labels()` and `compute_labels` will be called
|           label_period_val : amount of time to consider when computing labels
|           label_period_unit : the unit for the label period value (e.g., day)
|           spark_sqlctx : if compute layer is spark this must be passed

```

```

|         feature_function : optional custom feature function
|         dynamic_propagation : optional to dynamically propagate children
data upward, useful for very large compute graphs
|         type_func_match : optional mapping from type to a list of functions
(e.g., {'int' : ['min', 'max', 'sum'], 'str' : ['first']})
|
|     __repr__(self)
|         Return repr(self).
|
|     __str__(self)
|         Returns a short summary of the graph.
|
|         Returns
|         -----
|         info : string
|             Graph information as provided by `nx.info`
|
|         Examples
|         -----
|         >>> G = nx.Graph(name="foo")
|         >>> str(G)
|         "Graph named 'foo' with 0 nodes and 0 edges"
|
|         >>> G = nx.path_graph(3)
|         >>> str(G)
|         'Graph with 3 nodes and 2 edges'
|
|     add_entity_edge(self, parent_node: graphreduce.node.GraphReduceNode,
relation_node: graphreduce.node.GraphReduceNode, parent_key: str, relation_key:
str, relation_type: str = 'parent_child', reduce: bool = True)
|         Add an entity relation
|
|     assign_parent(self, parent_node: graphreduce.node.GraphReduceNode)
|         Assign the parent-most node in the graph
|
|     depth_first_generator(self)
|         Depth-first traversal over the edges
|
|     do_transformations(self)
|         Perform all graph transformations
|         1) hydrate graph
|         2) check for duplicate prefixes
|         3) filter data
|         4) clip anomalies
|         5) annotate data
|         6) depth-first edge traversal to: aggregate / reduce features and labels
|         6a) optional alternative feature_function mapping
|         6b) join back to parent edge

```

```

|         6c) post-join annotations if any
|         7) repeat step 6 on all edges up the hierarchy
|
|     get_children(self, node: graphreduce.node.GraphReduceNode) ->
List[graphreduce.node.GraphReduceNode]
|         Get the children of a given node
|
|     hydrate_graph_attrs(self, attrs=['cut_date', 'compute_period_val',
'compute_period_unit', 'has_labels', 'label_period_val', 'label_period_unit',
'compute_layer', 'feature_function', 'spark_sqlctx', '_storage_client'])
|         Hydrate the nodes in the graph with parent
|         attributes in `attrs`
|
|     hydrate_graph_data(self)
|         Hydrate the nodes in the graph with their data
|
|     join(self, parent_node: graphreduce.node.GraphReduceNode, relation_node:
graphreduce.node.GraphReduceNode, relation_df=None)
|         Join the child or peer nnode to the parent node
|
|         Optionally pass the `child_df` directly
|
|     plot_graph(self, fname: str = 'graph.html')
|         Plot the graph
|
|         Args
|             fname : file name to save the graph to - should be .html
|             notebook : whether or not to render in notebook
|
|     prefix_uniqueness(self)
|         Identify children with duplicate prefixes, if any
|
|     -----
|     Readonly properties defined here:
|
|     parent
|
|     -----
|     Methods inherited from networkx.classes.digraph.DiGraph:
|
|     add_edge(self, u_of_edge, v_of_edge, **attr)
|         Add an edge between u and v.
|
|         The nodes u and v will be automatically added if they are
|         not already in the graph.
|
|         Edge attributes can be specified with keywords or by directly
|         accessing the edge's attribute dictionary. See examples below.

```



## Parameters

-----

`u_of_edge, v_of_edge` : nodes

Nodes can be, for example, strings or numbers.

Nodes must be hashable (and not None) Python objects.

`attr` : keyword arguments, optional

Edge data (or labels or objects) can be assigned using keyword arguments.

## See Also

-----

`add_edges_from` : add a collection of edges

## Notes

-----

Adding an edge that already exists updates the edge data.

Many NetworkX algorithms designed for weighted graphs use an edge attribute (by default `'weight'`) to hold a numerical value.

## Examples

-----

The following all add the edge `e=(1, 2)` to graph `G`:

```
>>> G = nx.Graph() # or DiGraph, MultiGraph, MultiDiGraph, etc
>>> e = (1, 2)
>>> G.add_edge(1, 2) # explicit two-node form
>>> G.add_edge(*e) # single edge as tuple of two nodes
>>> G.add_edges_from([(1, 2)]) # add edges from iterable container
```

Associate data to edges using keywords:

```
>>> G.add_edge(1, 2, weight=3)
>>> G.add_edge(1, 3, weight=7, capacity=15, length=342.7)
```

For non-string attribute keys, use subscript notation.

```
>>> G.add_edge(1, 2)
>>> G[1][2].update({0: 5})
>>> G.edges[1, 2].update({0: 5})
```

```
add_edges_from(self, ebunch_to_add, **attr)
```

Add all the edges in `ebunch_to_add`.

## Parameters

-----

`ebunch_to_add` : container of edges

Each edge given in the container will be added to the graph. The edges must be given as 2-tuples (u, v) or 3-tuples (u, v, d) where d is a dictionary containing edge data.

attr : keyword arguments, optional

Edge data (or labels or objects) can be assigned using keyword arguments.

See Also

-----

add\_edge : add a single edge

add\_weighted\_edges\_from : convenient way to add weighted edges

Notes

-----

Adding the same edge twice has no effect but any edge data will be updated when each duplicate edge is added.

Edge attributes specified in an ebunch take precedence over attributes specified via keyword arguments.

Examples

-----

```
>>> G = nx.Graph() # or DiGraph, MultiGraph, MultiDiGraph, etc
>>> G.add_edges_from([(0, 1), (1, 2)]) # using a list of edge tuples
>>> e = zip(range(0, 3), range(1, 4))
>>> G.add_edges_from(e) # Add the path graph 0-1-2-3
```

Associate data to edges

```
>>> G.add_edges_from([(1, 2), (2, 3)], weight=3)
>>> G.add_edges_from([(3, 4), (1, 4)], label="WN2898")
```

add\_node(self, node\_for\_adding, \*\*attr)

Add a single node `node\_for\_adding` and update node attributes.

Parameters

-----

node\_for\_adding : node

A node can be any hashable Python object except None.

attr : keyword arguments, optional

Set or change node attributes using key=value.

See Also

-----

add\_nodes\_from

Examples

-----

```

>>> G = nx.Graph() # or DiGraph, MultiGraph, MultiDiGraph, etc
>>> G.add_node(1)
>>> G.add_node("Hello")
>>> K3 = nx.Graph([(0, 1), (1, 2), (2, 0)])
>>> G.add_node(K3)
>>> G.number_of_nodes()
3

```

Use keywords set/change node attributes:

```

>>> G.add_node(1, size=10)
>>> G.add_node(3, weight=0.4, UTM=("13S", 382871, 3972649))

```

#### Notes

-----

A hashable object is one that can be used as a key in a Python dictionary. This includes strings, numbers, tuples of strings and numbers, etc.

On many platforms hashable items also include mutables such as NetworkX Graphs, though one should be careful that the hash doesn't change on mutables.

```

add_nodes_from(self, nodes_for_adding, **attr)
    Add multiple nodes.

```

#### Parameters

-----

`nodes_for_adding` : iterable container

A container of nodes (list, dict, set, etc.).

OR

A container of (node, attribute dict) tuples.

Node attributes are updated using the attribute dict.

`attr` : keyword arguments, optional (default= no attributes)

Update attributes for all nodes in nodes.

Node attributes specified in nodes as a tuple take precedence over attributes specified via keyword arguments.

#### See Also

-----

`add_node`

#### Examples

-----

```

>>> G = nx.Graph() # or DiGraph, MultiGraph, MultiDiGraph, etc
>>> G.add_nodes_from("Hello")
>>> K3 = nx.Graph([(0, 1), (1, 2), (2, 0)])
>>> G.add_nodes_from(K3)

```

```
>>> sorted(G.nodes(), key=str)
[0, 1, 2, 'H', 'e', 'l', 'o']
```

Use keywords to update specific node attributes for every node.

```
>>> G.add_nodes_from([1, 2], size=10)
>>> G.add_nodes_from([3, 4], weight=0.4)
```

Use (node, attrdict) tuples to update attributes for specific nodes.

```
>>> G.add_nodes_from([(1, dict(size=11)), (2, {"color": "blue"})])
>>> G.nodes[1]["size"]
11
>>> H = nx.Graph()
>>> H.add_nodes_from(G.nodes(data=True))
>>> H.nodes[1]["size"]
11
```

adj = <functools.cached\_property object>

Graph adjacency object holding the neighbors of each node.

This object is a read-only dict-like structure with node keys and neighbor-dict values. The neighbor-dict is keyed by neighbor to the edge-data-dict. So `G.adj[3][2]['color'] = 'blue'` sets the color of the edge `(3, 2)` to `"blue"`.

Iterating over `G.adj` behaves like a dict. Useful idioms include `for nbr, datadict in G.adj[n].items():`.

The neighbor information is also provided by subscripting the graph. So `for nbr, foovalue in G[node].data('foo', default=1):` works.

For directed graphs, `G.adj` holds outgoing (successor) info.

clear(self)

Remove all nodes and edges from the graph.

This also removes the name, and all graph, node, and edge attributes.

Examples

-----

```
>>> G = nx.path_graph(4) # or DiGraph, MultiGraph, MultiDiGraph, etc
>>> G.clear()
>>> list(G.nodes)
[]
>>> list(G.edges)
[]
```

```

| clear_edges(self)
|     Remove all edges from the graph without altering nodes.
|
|     Examples
|     -----
|
|     >>> G = nx.path_graph(4) # or DiGraph, MultiGraph, MultiDiGraph, etc
|     >>> G.clear_edges()
|     >>> list(G.nodes)
|     [0, 1, 2, 3]
|     >>> list(G.edges)
|     []
|
| degree = <functools.cached_property object>
|     A DegreeView for the Graph as G.degree or G.degree().
|
|     The node degree is the number of edges adjacent to the node.
|     The weighted node degree is the sum of the edge weights for
|     edges incident to that node.
|
|     This object provides an iterator for (node, degree) as well as
|     lookup for the degree for a single node.
|
|     Parameters
|     -----
|
|     nbunch : single node, container, or all nodes (default= all nodes)
|         The view will only report edges incident to these nodes.
|
|     weight : string or None, optional (default=None)
|         The name of an edge attribute that holds the numerical value used
|         as a weight. If None, then each edge has weight 1.
|         The degree is the sum of the edge weights adjacent to the node.
|
|     Returns
|     -----
|
|     DiDegreeView or int
|         If multiple nodes are requested (the default), returns a
|         `DiDegreeView`
|         mapping nodes to their degree.
|         If a single node is requested, returns the degree of the node as an
integer.
|
|     See Also
|     -----
|
|     in_degree, out_degree
|
|     Examples
|     -----
|
|     >>> G = nx.DiGraph() # or MultiDiGraph

```

```

| >>> nx.add_path(G, [0, 1, 2, 3])
| >>> G.degree(0) # node 0 with degree 1
| 1
| >>> list(G.degree([0, 1, 2]))
| [(0, 1), (1, 2), (2, 2)]
|
| edges = <functools.cached_property object>
|     An OutEdgeView of the DiGraph as G.edges or G.edges().
|
|     edges(self, nbunch=None, data=False, default=None)
|
|     The OutEdgeView provides set-like operations on the edge-tuples
|     as well as edge attribute lookup. When called, it also provides
|     an EdgeDataView object which allows control of access to edge
|     attributes (but does not provide set-like operations).
|     Hence, `G.edges[u, v]['color']` provides the value of the color
|     attribute for edge `(u, v)` while
|     `for (u, v, c) in G.edges.data('color', default='red'):`
|     iterates through all the edges yielding the color attribute
|     with default `'red'` if no color attribute exists.
|
|     Parameters
|     -----
|     nbunch : single node, container, or all nodes (default= all nodes)
|         The view will only report edges from these nodes.
|     data : string or bool, optional (default=False)
|         The edge attribute returned in 3-tuple (u, v, ddict[data]).
|         If True, return edge attribute dict in 3-tuple (u, v, ddict).
|         If False, return 2-tuple (u, v).
|     default : value, optional (default=None)
|         Value used for edges that don't have the requested attribute.
|         Only relevant if data is not True or False.
|
|     Returns
|     -----
|     edges : OutEdgeView
|         A view of edge attributes, usually it iterates over (u, v)
|         or (u, v, d) tuples of edges, but can also be used for
|         attribute lookup as `edges[u, v]['foo']`.
|
|     See Also
|     -----
|     in_edges, out_edges
|
|     Notes
|     -----
|     Nodes in nbunch that are not in the graph will be (quietly) ignored.
|     For directed graphs this returns the out-edges.

```

## Examples

-----

```
>>> G = nx.DiGraph() # or MultiDiGraph, etc
>>> nx.add_path(G, [0, 1, 2])
>>> G.add_edge(2, 3, weight=5)
>>> [e for e in G.edges]
[(0, 1), (1, 2), (2, 3)]
>>> G.edges.data() # default data is {} (empty dict)
OutEdgeDataView([(0, 1, {}), (1, 2, {}), (2, 3, {'weight': 5})])
>>> G.edges.data("weight", default=1)
OutEdgeDataView([(0, 1, 1), (1, 2, 1), (2, 3, 5)])
>>> G.edges([0, 2]) # only edges originating from these nodes
OutEdgeDataView([(0, 1), (2, 3)])
>>> G.edges(0) # only edges from node 0
OutEdgeDataView([(0, 1)])
```

`has_predecessor(self, u, v)`  
Returns True if node u has predecessor v.

This is true if graph has the edge  $u \leftarrow v$ .

`has_successor(self, u, v)`  
Returns True if node u has successor v.

This is true if graph has the edge  $u \rightarrow v$ .

`in_degree = <functools.cached_property object>`  
An InDegreeView for (node, in\_degree) or in\_degree for single node.

The node in\_degree is the number of edges pointing to the node.  
The weighted node degree is the sum of the edge weights for edges incident to that node.

This object provides an iteration over (node, in\_degree) as well as lookup for the degree for a single node.

## Parameters

-----

`nbunch` : single node, container, or all nodes (default= all nodes)  
The view will only report edges incident to these nodes.

`weight` : string or None, optional (default=None)  
The name of an edge attribute that holds the numerical value used as a weight. If None, then each edge has weight 1.  
The degree is the sum of the edge weights adjacent to the node.

Returns

-----

If a single node is requested

deg : int

In-degree of the node

OR if multiple nodes are requested

nd\_iter : iterator

The iterator returns two-tuples of (node, in-degree).

See Also

-----

degree, out\_degree

Examples

-----

```
>>> G = nx.DiGraph()
```

```
>>> nx.add_path(G, [0, 1, 2, 3])
```

```
>>> G.in_degree(0) # node 0 with degree 0
```

```
0
```

```
>>> list(G.in_degree([0, 1, 2]))
```

```
[(0, 0), (1, 1), (2, 1)]
```

in\_edges = <functools.cached\_property object>

An InEdgeView of the Graph as G.in\_edges or G.in\_edges().

in\_edges(self, nbunch=None, data=False, default=None):

Parameters

-----

nbunch : single node, container, or all nodes (default= all nodes)

The view will only report edges incident to these nodes.

data : string or bool, optional (default=False)

The edge attribute returned in 3-tuple (u, v, ddict[data]).

If True, return edge attribute dict in 3-tuple (u, v, ddict).

If False, return 2-tuple (u, v).

default : value, optional (default=None)

Value used for edges that don't have the requested attribute.

Only relevant if data is not True or False.

Returns

-----

in\_edges : InEdgeView

A view of edge attributes, usually it iterates over (u, v)

or (u, v, d) tuples of edges, but can also be used for

attribute lookup as `edges[u, v]['foo']`.

See Also

-----



```

|     edges
|
| is_directed(self)
|     Returns True if graph is directed, False otherwise.
|
| is_multigraph(self)
|     Returns True if graph is a multigraph, False otherwise.
|
| neighbors = successors(self, n)
|
| out_degree = <functools.cached_property object>
|     An OutDegreeView for (node, out_degree)
|
|     The node out_degree is the number of edges pointing out of the node.
|     The weighted node degree is the sum of the edge weights for
|     edges incident to that node.
|
|     This object provides an iterator over (node, out_degree) as well as
|     lookup for the degree for a single node.
|
|     Parameters
|     -----
|
|     nbunch : single node, container, or all nodes (default= all nodes)
|         The view will only report edges incident to these nodes.
|
|     weight : string or None, optional (default=None)
|         The name of an edge attribute that holds the numerical value used
|         as a weight.  If None, then each edge has weight 1.
|         The degree is the sum of the edge weights adjacent to the node.
|
|     Returns
|     -----
|
|     If a single node is requested
|     deg : int
|         Out-degree of the node
|
|     OR if multiple nodes are requested
|     nd_iter : iterator
|         The iterator returns two-tuples of (node, out-degree).
|
|     See Also
|     -----
|
|     degree, in_degree
|
|     Examples
|     -----
|
|     >>> G = nx.DiGraph()
|     >>> nx.add_path(G, [0, 1, 2, 3])

```

```

|     >>> G.out_degree(0)  # node 0 with degree 1
|     1
|     >>> list(G.out_degree([0, 1, 2]))
|     [(0, 1), (1, 1), (2, 1)]
|
| out_edges = <functools.cached_property object>
|     An OutEdgeView of the DiGraph as G.edges or G.edges().
|
|     edges(self, nbunch=None, data=False, default=None)
|
| The OutEdgeView provides set-like operations on the edge-tuples
| as well as edge attribute lookup. When called, it also provides
| an EdgeDataView object which allows control of access to edge
| attributes (but does not provide set-like operations).
| Hence, `G.edges[u, v]['color']` provides the value of the color
| attribute for edge `(u, v)` while
| `for (u, v, c) in G.edges.data('color', default='red'):`
| iterates through all the edges yielding the color attribute
| with default `'red'` if no color attribute exists.
|
| Parameters
| -----
| nbunch : single node, container, or all nodes (default= all nodes)
|     The view will only report edges from these nodes.
| data : string or bool, optional (default=False)
|     The edge attribute returned in 3-tuple (u, v, ddict[data]).
|     If True, return edge attribute dict in 3-tuple (u, v, ddict).
|     If False, return 2-tuple (u, v).
| default : value, optional (default=None)
|     Value used for edges that don't have the requested attribute.
|     Only relevant if data is not True or False.
|
| Returns
| -----
| edges : OutEdgeView
|     A view of edge attributes, usually it iterates over (u, v)
|     or (u, v, d) tuples of edges, but can also be used for
|     attribute lookup as `edges[u, v]['foo']`.
|
| See Also
| -----
| in_edges, out_edges
|
| Notes
| -----
| Nodes in nbunch that are not in the graph will be (quietly) ignored.
| For directed graphs this returns the out-edges.

```

## Examples

-----

```
>>> G = nx.DiGraph() # or MultiDiGraph, etc
>>> nx.add_path(G, [0, 1, 2])
>>> G.add_edge(2, 3, weight=5)
>>> [e for e in G.edges]
[(0, 1), (1, 2), (2, 3)]
>>> G.edges.data() # default data is {} (empty dict)
OutEdgeDataView([(0, 1, {}), (1, 2, {}), (2, 3, {'weight': 5})])
>>> G.edges.data("weight", default=1)
OutEdgeDataView([(0, 1, 1), (1, 2, 1), (2, 3, 5)])
>>> G.edges([0, 2]) # only edges originating from these nodes
OutEdgeDataView([(0, 1), (2, 3)])
>>> G.edges(0) # only edges from node 0
OutEdgeDataView([(0, 1)])
```

pred = <functools.cached\_property object>

Graph adjacency object holding the predecessors of each node.

This object is a read-only dict-like structure with node keys and neighbor-dict values. The neighbor-dict is keyed by neighbor to the edge-data-dict. So `G.pred[2][3]['color'] = 'blue'` sets the color of the edge `(3, 2)` to `"blue"`.

Iterating over `G.pred` behaves like a dict. Useful idioms include `for nbr, datadict in G.pred[n].items():`. A data-view not provided by dicts also exists: `for nbr, foovalue in G.pred[node].data('foo'):`. A default can be set via a `default` argument to the `data` method.

predecessors(self, n)

Returns an iterator over predecessor nodes of n.

A predecessor of n is a node m such that there exists a directed edge from m to n.

## Parameters

-----

n : node

A node in the graph

## Raises

-----

NetworkXError

If n is not in the graph.

## See Also

-----

successors

```

remove_edge(self, u, v)
    Remove the edge between u and v.

    Parameters
    -----
    u, v : nodes
        Remove the edge between nodes u and v.

    Raises
    -----
    NetworkXError
        If there is not an edge between u and v.

    See Also
    -----
    remove_edges_from : remove a collection of edges

    Examples
    -----
    >>> G = nx.Graph() # or DiGraph, etc
    >>> nx.add_path(G, [0, 1, 2, 3])
    >>> G.remove_edge(0, 1)
    >>> e = (1, 2)
    >>> G.remove_edge(*e) # unpacks e from an edge tuple
    >>> e = (2, 3, {"weight": 7}) # an edge with attribute data
    >>> G.remove_edge(*e[:2]) # select first part of edge tuple

remove_edges_from(self, ebunch)
    Remove all edges specified in ebunch.

    Parameters
    -----
    ebunch: list or container of edge tuples
        Each edge given in the list or container will be removed
        from the graph. The edges can be:

            - 2-tuples (u, v) edge between u and v.
            - 3-tuples (u, v, k) where k is ignored.

    See Also
    -----
    remove_edge : remove a single edge

    Notes
    -----
    Will fail silently if an edge in ebunch is not in the graph.

```

#### Examples

-----

```
>>> G = nx.path_graph(4) # or DiGraph, MultiGraph, MultiDiGraph, etc
>>> ebunch = [(1, 2), (2, 3)]
>>> G.remove_edges_from(ebunch)
```

```
remove_node(self, n)
```

Remove node n.

Removes the node n and all adjacent edges.

Attempting to remove a non-existent node will raise an exception.

#### Parameters

-----

n : node

A node in the graph

#### Raises

-----

NetworkXError

If n is not in the graph.

#### See Also

-----

remove\_nodes\_from

#### Examples

-----

```
>>> G = nx.path_graph(3) # or DiGraph, MultiGraph, MultiDiGraph, etc
>>> list(G.edges)
[(0, 1), (1, 2)]
>>> G.remove_node(1)
>>> list(G.edges)
[]
```

```
remove_nodes_from(self, nodes)
```

Remove multiple nodes.

#### Parameters

-----

nodes : iterable container

A container of nodes (list, dict, set, etc.). If a node in the container is not in the graph it is silently ignored.

#### See Also

-----

remove\_node

## Examples

-----

```
>>> G = nx.path_graph(3) # or DiGraph, MultiGraph, MultiDiGraph, etc
>>> e = list(G.nodes)
>>> e
[0, 1, 2]
>>> G.remove_nodes_from(e)
>>> list(G.nodes)
[]
```

`reverse(self, copy=True)`

Returns the reverse of the graph.

The reverse is a graph with the same nodes and edges but with the directions of the edges reversed.

## Parameters

-----

`copy` : bool optional (default=True)

If True, return a new DiGraph holding the reversed edges.

If False, the reverse graph is created using a view of the original graph.

`succ = <functools.cached_property object>`

Graph adjacency object holding the successors of each node.

This object is a read-only dict-like structure with node keys and neighbor-dict values. The neighbor-dict is keyed by neighbor to the edge-data-dict. So `G.succ[3][2]['color'] = 'blue'` sets the color of the edge `(3, 2)` to `"blue"`.

Iterating over `G.succ` behaves like a dict. Useful idioms include `for nbr, datadict in G.succ[n].items():`. A data-view not provided by dicts also exists: `for nbr, foovalue in G.succ[node].data('foo'):` and a default can be set via a `default` argument to the `data` method.

The neighbor information is also provided by subscripting the graph. So `for nbr, foovalue in G[node].data('foo', default=1):` works.

For directed graphs, `G.adj` is identical to `G.succ`.

`successors(self, n)`

Returns an iterator over successor nodes of `n`.

A successor of `n` is a node `m` such that there exists a directed edge from `n` to `m`.

## Parameters

-----

n : node

A node in the graph

Raises

-----

NetworkXError

If n is not in the graph.

See Also

-----

predecessors

Notes

-----

neighbors() and successors() are the same.

to\_undirected(self, reciprocal=False, as\_view=False)

Returns an undirected representation of the digraph.

Parameters

-----

reciprocal : bool (optional)

If True only keep edges that appear in both directions  
in the original digraph.

as\_view : bool (optional, default=False)

If True return an undirected view of the original directed graph.

Returns

-----

G : Graph

An undirected graph with the same name and nodes and  
with edge (u, v, data) if either (u, v, data) or (v, u, data)  
is in the digraph. If both edges exist in digraph and  
their edge data is different, only one edge is created  
with an arbitrary choice of which edge data to use.  
You must check and correct for this manually if desired.

See Also

-----

Graph, copy, add\_edge, add\_edges\_from

Notes

-----

If edges in both directions (u, v) and (v, u) exist in the  
graph, attributes for the new undirected edge will be a combination of  
the attributes of the directed edges. The edge data is updated  
in the (arbitrary) order that the edges are encountered. For

more customized control of the edge attributes use `add_edge()`.

This returns a "deepcopy" of the edge, node, and graph attributes which attempts to completely copy all of the data and references.

This is in contrast to the similar `G=DiGraph(D)` which returns a shallow copy of the data.

See the Python copy module for more information on shallow and deep copies, <https://docs.python.org/3/library/copy.html>.

Warning: If you have subclassed `DiGraph` to use dict-like objects in the data structure, those changes do not transfer to the Graph created by this method.

#### Examples

-----

```
>>> G = nx.path_graph(2) # or MultiGraph, etc
>>> H = G.to_directed()
>>> list(H.edges)
[(0, 1), (1, 0)]
>>> G2 = H.to_undirected()
>>> list(G2.edges)
[(0, 1)]
```

-----  
Methods inherited from `networkx.classes.graph.Graph`:

`__contains__(self, n)`

Returns True if `n` is a node, False otherwise. Use: '`n in G`'.

#### Examples

-----

```
>>> G = nx.path_graph(4) # or DiGraph, MultiGraph, MultiDiGraph, etc
>>> 1 in G
True
```

`__getitem__(self, n)`

Returns a dict of neighbors of node `n`. Use: '`G[n]`'.

#### Parameters

-----

`n` : node

A node in the graph.

#### Returns

-----



```

adj_dict : dictionary
    The adjacency dictionary for nodes connected to n.

Notes
-----
G[n] is the same as G.adj[n] and similar to G.neighbors(n)
(which is an iterator over G.adj[n])

Examples
-----
>>> G = nx.path_graph(4) # or DiGraph, MultiGraph, MultiDiGraph, etc
>>> G[0]
AtlasView({1: {}})

__iter__(self)
    Iterate over the nodes. Use: 'for n in G'.

Returns
-----
niter : iterator
    An iterator over all nodes in the graph.

Examples
-----
>>> G = nx.path_graph(4) # or DiGraph, MultiGraph, MultiDiGraph, etc
>>> [n for n in G]
[0, 1, 2, 3]
>>> list(G)
[0, 1, 2, 3]

__len__(self)
    Returns the number of nodes in the graph. Use: 'len(G)'.

Returns
-----
nnodes : int
    The number of nodes in the graph.

See Also
-----
number_of_nodes: identical method
order: identical method

Examples
-----
>>> G = nx.path_graph(4) # or DiGraph, MultiGraph, MultiDiGraph, etc
>>> len(G)
4

```

```
add_weighted_edges_from(self, ebunch_to_add, weight='weight', **attr)
    Add weighted edges in `ebunch_to_add` with specified weight attr
```

#### Parameters

-----

ebunch\_to\_add : container of edges

Each edge given in the list or container will be added to the graph. The edges must be given as 3-tuples (u, v, w) where w is a number.

weight : string, optional (default= 'weight')

The attribute name for the edge weights to be added.

attr : keyword arguments, optional (default= no attributes)

Edge attributes to add/update for all edges.

#### See Also

-----

add\_edge : add a single edge

add\_edges\_from : add multiple edges

#### Notes

-----

Adding the same edge twice for Graph/DiGraph simply updates the edge data. For MultiGraph/MultiDiGraph, duplicate edges are stored.

#### Examples

-----

```
>>> G = nx.Graph() # or DiGraph, MultiGraph, MultiDiGraph, etc
```

```
>>> G.add_weighted_edges_from([(0, 1, 3.0), (1, 2, 7.5)])
```

```
adjacency(self)
```

Returns an iterator over (node, adjacency dict) tuples for all nodes.

For directed graphs, only outgoing neighbors/adjacencies are included.

#### Returns

-----

adj\_iter : iterator

An iterator over (node, adjacency dictionary) for all nodes in the graph.

#### Examples

-----

```
>>> G = nx.path_graph(4) # or DiGraph, MultiGraph, MultiDiGraph, etc
```

```
>>> [(n, nbrdict) for n, nbrdict in G.adjacency()]
```

```
[(0, {1: {}}), (1, {0: {}, 2: {}}), (2, {1: {}, 3: {}}), (3, {2: {}})]
```

```
copy(self, as_view=False)
```

Returns a copy of the graph.

The copy method by default returns an independent shallow copy of the graph and attributes. That is, if an attribute is a container, that container is shared by the original and the copy. Use Python's ``copy.deepcopy`` for new containers.

If ``as_view`` is True then a view is returned instead of a copy.

Notes

-----

All copies reproduce the graph structure, but data attributes may be handled in different ways. There are four types of copies of a graph that people might want.

Deepcopy -- A "deepcopy" copies the graph structure as well as all data attributes and any objects they might contain. The entire graph object is new so that changes in the copy do not affect the original object. (see Python's `copy.deepcopy`)

Data Reference (Shallow) -- For a shallow copy the graph structure is copied but the edge, node and graph attribute dicts are references to those in the original graph. This saves time and memory but could cause confusion if you change an attribute in one graph and it changes the attribute in the other. NetworkX does not provide this level of shallow copy.

Independent Shallow -- This copy creates new independent attribute dicts and then does a shallow copy of the attributes. That is, any attributes that are containers are shared between the new graph and the original. This is exactly what ``dict.copy()`` provides. You can obtain this style copy using:

```
>>> G = nx.path_graph(5)
>>> H = G.copy()
>>> H = G.copy(as_view=False)
>>> H = nx.Graph(G)
>>> H = G.__class__(G)
```

Fresh Data -- For fresh data, the graph structure is copied while new empty data attribute dicts are created. The resulting graph is independent of the original and it has no edge, node or graph attributes. Fresh copies are not enabled. Instead use:

```
>>> H = G.__class__()
>>> H.add_nodes_from(G)
>>> H.add_edges_from(G.edges)
```

View -- Inspired by dict-views, graph-views act like read-only versions of the original graph, providing a copy of the original structure without requiring any memory for copying the information.

See the Python copy module for more information on shallow and deep copies, <https://docs.python.org/3/library/copy.html>.

#### Parameters

-----

`as_view` : bool, optional (default=False)

If True, the returned graph-view provides a read-only view of the original graph without actually copying any data.

#### Returns

-----

`G` : Graph

A copy of the graph.

#### See Also

-----

`to_directed`: return a directed copy of the graph.

#### Examples

-----

```
>>> G = nx.path_graph(4) # or DiGraph, MultiGraph, MultiDiGraph, etc
```

```
>>> H = G.copy()
```

`edge_subgraph(self, edges)`

Returns the subgraph induced by the specified edges.

The induced subgraph contains each edge in ``edges`` and each node incident to any one of those edges.

#### Parameters

-----

`edges` : iterable

An iterable of edges in this graph.

#### Returns

-----

`G` : Graph

An edge-induced subgraph of this graph with the same edge attributes.

#### Notes

-----

The graph, edge, and node attributes in the returned subgraph

view are references to the corresponding attributes in the original graph. The view is read-only.

To create a full graph version of the subgraph with its own copy of the edge or node attributes, use::

```
G.edge_subgraph(edges).copy()
```

Examples

-----

```
>>> G = nx.path_graph(5)
>>> H = G.edge_subgraph([(0, 1), (3, 4)])
>>> list(H.nodes)
[0, 1, 3, 4]
>>> list(H.edges)
[(0, 1), (3, 4)]
```

```
get_edge_data(self, u, v, default=None)
```

Returns the attribute dictionary associated with edge (u, v).

This is identical to `G[u][v]` except the default is returned instead of an exception if the edge doesn't exist.

Parameters

-----

u, v : nodes

default: any Python object (default=None)

Value to return if the edge (u, v) is not found.

Returns

-----

edge\_dict : dictionary

The edge attribute dictionary.

Examples

-----

```
>>> G = nx.path_graph(4) # or DiGraph, MultiGraph, MultiDiGraph, etc
>>> G[0][1]
{}
```

Warning: Assigning to `G[u][v]` is not permitted.

But it is safe to assign attributes `G[u][v]['foo']`

```
>>> G[0][1]["weight"] = 7
>>> G[0][1]["weight"]
7
>>> G[1][0]["weight"]
7
```

```

|
| >>> G = nx.path_graph(4) # or DiGraph, MultiGraph, MultiDiGraph, etc
| >>> G.get_edge_data(0, 1) # default edge data is {}
| {}
| >>> e = (0, 1)
| >>> G.get_edge_data(*e) # tuple form
| {}
| >>> G.get_edge_data("a", "b", default=0) # edge not in graph, return 0
| 0
|
| has_edge(self, u, v)
|     Returns True if the edge (u, v) is in the graph.
|
|     This is the same as `v in G[u]` without KeyError exceptions.
|
|     Parameters
|     -----
|     u, v : nodes
|         Nodes can be, for example, strings or numbers.
|         Nodes must be hashable (and not None) Python objects.
|
|     Returns
|     -----
|     edge_ind : bool
|         True if edge is in the graph, False otherwise.
|
|     Examples
|     -----
|     >>> G = nx.path_graph(4) # or DiGraph, MultiGraph, MultiDiGraph, etc
|     >>> G.has_edge(0, 1) # using two nodes
|     True
|     >>> e = (0, 1)
|     >>> G.has_edge(*e) # e is a 2-tuple (u, v)
|     True
|     >>> e = (0, 1, {"weight": 7})
|     >>> G.has_edge(*e[:2]) # e is a 3-tuple (u, v, data_dictionary)
|     True
|
|     The following syntax are equivalent:
|
|     >>> G.has_edge(0, 1)
|     True
|     >>> 1 in G[0] # though this gives KeyError if 0 not in G
|     True
|
| has_node(self, n)
|     Returns True if the graph contains the node n.
|

```

Identical to ``n in G``

Parameters

-----

`n` : node

Examples

-----

```
>>> G = nx.path_graph(3) # or DiGraph, MultiGraph, MultiDiGraph, etc
```

```
>>> G.has_node(0)
```

```
True
```

It is more readable and simpler to use

```
>>> 0 in G
```

```
True
```

`nbunch_iter(self, nbunch=None)`

Returns an iterator over nodes contained in `nbunch` that are also in the graph.

The nodes in `nbunch` are checked for membership in the graph and if not are silently ignored.

Parameters

-----

`nbunch` : single node, container, or all nodes (default= all nodes)

The view will only report edges incident to these nodes.

Returns

-----

`niter` : iterator

An iterator over nodes in `nbunch` that are also in the graph.

If `nbunch` is `None`, iterate over all nodes in the graph.

Raises

-----

`NetworkXError`

If `nbunch` is not a node or sequence of nodes.

If a node in `nbunch` is not hashable.

See Also

-----

`Graph.__iter__`

Notes

-----

When `nbunch` is an iterator, the returned iterator yields values

directly from nbunch, becoming exhausted when nbunch is exhausted.

To test whether nbunch is a single node, one can use  
"if nbunch in self:", even after processing with this routine.

If nbunch is not a node or a (possibly empty) sequence/iterator or None, a :exc:`NetworkXError` is raised. Also, if any object in nbunch is not hashable, a :exc:`NetworkXError` is raised.

nodes = <functools.cached\_property object>

A NodeView of the Graph as G.nodes or G.nodes().

Can be used as `G.nodes` for data lookup and for set-like operations. Can also be used as `G.nodes(data='color', default=None)` to return a NodeDataView which reports specific node data but no set operations. It presents a dict-like interface as well with `G.nodes.items()` iterating over `(node, nodedata)` 2-tuples and `G.nodes[3]['foo']` providing the value of the `foo` attribute for node `3`. In addition, a view `G.nodes.data('foo')` provides a dict-like interface to the `foo` attribute of each node. `G.nodes.data('foo', default=1)` provides a default for nodes that do not have attribute `foo`.

#### Parameters

-----

data : string or bool, optional (default=False)

The node attribute returned in 2-tuple (n, ddict[data]).

If True, return entire node attribute dict as (n, ddict).

If False, return just the nodes n.

default : value, optional (default=None)

Value used for nodes that don't have the requested attribute.

Only relevant if data is not True or False.

#### Returns

-----

##### NodeView

Allows set-like operations over the nodes as well as node attribute dict lookup and calling to get a NodeDataView.

A NodeDataView iterates over `(n, data)` and has no set operations.

A NodeView iterates over `n` and includes set operations.

When called, if data is False, an iterator over nodes.

Otherwise an iterator of 2-tuples (node, attribute value)

where the attribute is specified in `data`.

If data is True then the attribute becomes the

entire data dictionary.

#### Notes



-----

If your node data is not needed, it is simpler and equivalent to use the expression `for n in G`, or `list(G)`.

#### Examples

There are two simple ways of getting a list of all nodes in the graph:

```
>>> G = nx.path_graph(3)
>>> list(G.nodes)
[0, 1, 2]
>>> list(G)
[0, 1, 2]
```

To get the node data along with the nodes:

```
>>> G.add_node(1, time="5pm")
>>> G.nodes[0]["foo"] = "bar"
>>> list(G.nodes(data=True))
[(0, {'foo': 'bar'}), (1, {'time': '5pm'}), (2, {})]
>>> list(G.nodes.data())
[(0, {'foo': 'bar'}), (1, {'time': '5pm'}), (2, {})]

>>> list(G.nodes(data="foo"))
[(0, 'bar'), (1, None), (2, None)]
>>> list(G.nodes.data("foo"))
[(0, 'bar'), (1, None), (2, None)]

>>> list(G.nodes(data="time"))
[(0, None), (1, '5pm'), (2, None)]
>>> list(G.nodes.data("time"))
[(0, None), (1, '5pm'), (2, None)]

>>> list(G.nodes(data="time", default="Not Available"))
[(0, 'Not Available'), (1, '5pm'), (2, 'Not Available')]
>>> list(G.nodes.data("time", default="Not Available"))
[(0, 'Not Available'), (1, '5pm'), (2, 'Not Available')]
```

If some of your nodes have an attribute and the rest are assumed to have a default attribute value you can create a dictionary from node/attribute pairs using the `default` keyword argument to guarantee the value is never `None`:

```
>>> G = nx.Graph()
>>> G.add_node(0)
>>> G.add_node(1, weight=2)
>>> G.add_node(2, weight=3)
>>> dict(G.nodes(data="weight", default=1))
```

```

    {0: 1, 1: 2, 2: 3}

number_of_edges(self, u=None, v=None)
    Returns the number of edges between two nodes.

Parameters
-----
u, v : nodes, optional (default=all edges)
    If u and v are specified, return the number of edges between
    u and v. Otherwise return the total number of all edges.

Returns
-----
nedges : int
    The number of edges in the graph. If nodes `u` and `v` are
    specified return the number of edges between those nodes. If
    the graph is directed, this only returns the number of edges
    from `u` to `v`.

See Also
-----
size

Examples
-----
For undirected graphs, this method counts the total number of
edges in the graph:

>>> G = nx.path_graph(4)
>>> G.number_of_edges()
3

If you specify two nodes, this counts the total number of edges
joining the two nodes:

>>> G.number_of_edges(0, 1)
1

For directed graphs, this method can count the total number of
directed edges from `u` to `v`:

>>> G = nx.DiGraph()
>>> G.add_edge(0, 1)
>>> G.add_edge(1, 0)
>>> G.number_of_edges(0, 1)
1

number_of_nodes(self)

```

```

| Returns the number of nodes in the graph.
|
| Returns
| -----
| nnodes : int
|     The number of nodes in the graph.
|
| See Also
| -----
| order: identical method
| __len__: identical method
|
| Examples
| -----
| >>> G = nx.path_graph(3) # or DiGraph, MultiGraph, MultiDiGraph, etc
| >>> G.number_of_nodes()
| 3
|
| order(self)
|     Returns the number of nodes in the graph.
|
| Returns
| -----
| nnodes : int
|     The number of nodes in the graph.
|
| See Also
| -----
| number_of_nodes: identical method
| __len__: identical method
|
| Examples
| -----
| >>> G = nx.path_graph(3) # or DiGraph, MultiGraph, MultiDiGraph, etc
| >>> G.order()
| 3
|
| size(self, weight=None)
|     Returns the number of edges or total of all edge weights.
|
| Parameters
| -----
| weight : string or None, optional (default=None)
|     The edge attribute that holds the numerical value used
|     as a weight. If None, then each edge has weight 1.
|
| Returns
| -----

```

size : numeric

The number of edges or  
(if weight keyword is provided) the total weight sum.

If weight is None, returns an int. Otherwise a float  
(or more general numeric if the weights are more general).

See Also

-----

number\_of\_edges

Examples

-----

```
>>> G = nx.path_graph(4) # or DiGraph, MultiGraph, MultiDiGraph, etc
>>> G.size()
3
```

```
>>> G = nx.Graph() # or DiGraph, MultiGraph, MultiDiGraph, etc
>>> G.add_edge("a", "b", weight=2)
>>> G.add_edge("b", "c", weight=4)
>>> G.size()
2
>>> G.size(weight="weight")
6.0
```

subgraph(self, nodes)

Returns a SubGraph view of the subgraph induced on `nodes`.

The induced subgraph of the graph contains the nodes in `nodes`  
and the edges between those nodes.

Parameters

-----

nodes : list, iterable

A container of nodes which will be iterated through once.

Returns

-----

G : SubGraph View

A subgraph view of the graph. The graph structure cannot be  
changed but node/edge attributes can and are shared with the  
original graph.

Notes

-----

The graph, edge and node attributes are shared with the original graph.  
Changes to the graph structure is ruled out by the view, but changes  
to attributes are reflected in the original graph.

To create a subgraph with its own copy of the edge/node attributes use:  
`G.subgraph(nodes).copy()`

For an inplace reduction of a graph to a subgraph you can remove nodes:  
`G.remove_nodes_from([n for n in G if n not in set(nodes)])`

Subgraph views are sometimes NOT what you want. In most cases where you want to do more than simply look at the induced edges, it makes more sense to just create the subgraph as its own graph with code like:

```
::  
  
    # Create a subgraph SG based on a (possibly multigraph) G  
    SG = G.__class__()  
    SG.add_nodes_from((n, G.nodes[n]) for n in largest_wcc)  
    if SG.is_multigraph():  
        SG.add_edges_from((n, nbr, key, d)  
                           for n, nbrs in G.adj.items() if n in largest_wcc  
                           for nbr, keydict in nbrs.items() if nbr in largest_wcc  
                           for key, d in keydict.items()))  
    else:  
        SG.add_edges_from((n, nbr, d)  
                           for n, nbrs in G.adj.items() if n in largest_wcc  
                           for nbr, d in nbrs.items() if nbr in largest_wcc)  
    SG.graph.update(G.graph)
```

#### Examples

```
-----  
>>> G = nx.path_graph(4) # or DiGraph, MultiGraph, MultiDiGraph, etc  
>>> H = G.subgraph([0, 1, 2])  
>>> list(H.edges)  
[(0, 1), (1, 2)]
```

`to_directed(self, as_view=False)`  
Returns a directed representation of the graph.

#### Returns

-----

G : DiGraph

A directed graph with the same name, same nodes, and with each edge (u, v, data) replaced by two directed edges (u, v, data) and (v, u, data).

#### Notes

-----

This returns a "deepcopy" of the edge, node, and graph attributes which attempts to completely copy

all of the data and references.

This is in contrast to the similar `D=DiGraph(G)` which returns a shallow copy of the data.

See the Python copy module for more information on shallow and deep copies, <https://docs.python.org/3/library/copy.html>.

Warning: If you have subclassed Graph to use dict-like objects in the data structure, those changes do not transfer to the DiGraph created by this method.

#### Examples

-----

```
>>> G = nx.Graph() # or MultiGraph, etc
>>> G.add_edge(0, 1)
>>> H = G.to_directed()
>>> list(H.edges)
[(0, 1), (1, 0)]
```

If already directed, return a (deep) copy

```
>>> G = nx.DiGraph() # or MultiDiGraph, etc
>>> G.add_edge(0, 1)
>>> H = G.to_directed()
>>> list(H.edges)
[(0, 1)]
```

`to_directed_class(self)`

Returns the class to use for empty directed copies.

If you subclass the base classes, use this to designate what directed class to use for ``to_directed()`` copies.

`to_undirected_class(self)`

Returns the class to use for empty undirected copies.

If you subclass the base classes, use this to designate what directed class to use for ``to_directed()`` copies.

`update(self, edges=None, nodes=None)`

Update the graph using nodes/edges/graphs as input.

Like `dict.update`, this method takes a graph as input, adding the graph's nodes and edges to this graph. It can also take two inputs: edges and nodes. Finally it can take either edges or nodes. To specify only nodes the keyword ``nodes`` must be used.

The collections of edges and nodes are treated similarly to the `add_edges_from/add_nodes_from` methods. When iterated, they should yield 2-tuples (u, v) or 3-tuples (u, v, datadict).

#### Parameters

`edges` : Graph object, collection of edges, or None

The first parameter can be a graph or some edges. If it has attributes ``nodes`` and ``edges``, then it is taken to be a Graph-like object and those attributes are used as collections of nodes and edges to be added to the graph.

If the first parameter does not have those attributes, it is treated as a collection of edges and added to the graph.

If the first argument is None, no edges are added.

`nodes` : collection of nodes, or None

The second parameter is treated as a collection of nodes to be added to the graph unless it is None.

If ``edges`` is None and ``nodes`` is None an exception is raised.

If the first parameter is a Graph, then ``nodes`` is ignored.

#### Examples

```
>>> G = nx.path_graph(5)
>>> G.update(nx.complete_graph(range(4, 10)))
>>> from itertools import combinations
>>> edges = (
...     (u, v, {"power": u * v})
...     for u, v in combinations(range(10, 20), 2)
...     if u * v < 225
... )
>>> nodes = [1000] # for singleton, use a container
>>> G.update(edges, nodes)
```

#### Notes

If you want to update the graph using an adjacency structure it is straightforward to obtain the edges/nodes from adjacency. The following examples provide common cases, your adjacency may be slightly different and require tweaks of these examples::

```
>>> # dict-of-set/list/tuple
>>> adj = {1: {2, 3}, 2: {1, 3}, 3: {1, 2}}
>>> e = [(u, v) for u, nbrs in adj.items() for v in nbrs]
>>> G.update(edges=e, nodes=adj)

>>> DG = nx.DiGraph()
>>> # dict-of-dict-of-attribute
>>> adj = {1: {2: 1.3, 3: 0.7}, 2: {1: 1.4}, 3: {1: 0.7}}
```

```

>>> e = [
...     (u, v, {"weight": d})
...     for u, nbrs in adj.items()
...     for v, d in nbrs.items()
... ]
>>> DG.update(edges=e, nodes=adj)

>>> # dict-of-dict-of-dict
>>> adj = {1: {2: {"weight": 1.3}, 3: {"color": 0.7, "weight": 1.2}}}
>>> e = [
...     (u, v, {"weight": d})
...     for u, nbrs in adj.items()
...     for v, d in nbrs.items()
... ]
>>> DG.update(edges=e, nodes=adj)

>>> # predecessor adjacency (dict-of-set)
>>> pred = {1: {2, 3}, 2: {3}, 3: {3}}
>>> e = [(v, u) for u, nbrs in pred.items() for v in nbrs]

>>> # MultiGraph dict-of-dict-of-dict-of-attribute
>>> MDG = nx.MultiDiGraph()
>>> adj = {
...     1: {2: {0: {"weight": 1.3}, 1: {"weight": 1.2}}},
...     3: {2: {0: {"weight": 0.7}}},
... }
>>> e = [
...     (u, v, ekey, d)
...     for u, nbrs in adj.items()
...     for v, keydict in nbrs.items()
...     for ekey, d in keydict.items()
... ]
>>> MDG.update(edges=e)

```

See Also

-----

add\_edges\_from: add multiple edges to a graph

add\_nodes\_from: add multiple nodes to a graph

---

Data descriptors inherited from `networkx.classes.graph.Graph`:

`__dict__`

dictionary for instance variables (if defined)

`__weakref__`

list of weak references to the object (if defined)



```

| name
|     String identifier of the graph.
|
|     This graph attribute appears in the attribute dict G.graph
|     keyed by the string `"name"`. as well as an attribute (technically
|     a property) `G.name`. This is entirely user controlled.
|
| -----
| Data and other attributes inherited from networkx.classes.graph.Graph:
|
| adjlist_inner_dict_factory = <class 'dict'>
|     dict() -> new empty dictionary
|     dict(mapping) -> new dictionary initialized from a mapping object's
|         (key, value) pairs
|     dict(iterable) -> new dictionary initialized as if via:
|         d = {}
|         for k, v in iterable:
|             d[k] = v
|     dict(**kwargs) -> new dictionary initialized with the name=value pairs
|         in the keyword argument list.  For example:  dict(one=1, two=2)
|
| adjlist_outer_dict_factory = <class 'dict'>
|     dict() -> new empty dictionary
|     dict(mapping) -> new dictionary initialized from a mapping object's
|         (key, value) pairs
|     dict(iterable) -> new dictionary initialized as if via:
|         d = {}
|         for k, v in iterable:
|             d[k] = v
|     dict(**kwargs) -> new dictionary initialized with the name=value pairs
|         in the keyword argument list.  For example:  dict(one=1, two=2)
|
| edge_attr_dict_factory = <class 'dict'>
|     dict() -> new empty dictionary
|     dict(mapping) -> new dictionary initialized from a mapping object's
|         (key, value) pairs
|     dict(iterable) -> new dictionary initialized as if via:
|         d = {}
|         for k, v in iterable:
|             d[k] = v
|     dict(**kwargs) -> new dictionary initialized with the name=value pairs
|         in the keyword argument list.  For example:  dict(one=1, two=2)
|
| graph_attr_dict_factory = <class 'dict'>
|     dict() -> new empty dictionary

```

```

| dict(mapping) -> new dictionary initialized from a mapping object's
| (key, value) pairs
| dict(iterable) -> new dictionary initialized as if via:
|     d = {}
|     for k, v in iterable:
|         d[k] = v
| dict(**kwargs) -> new dictionary initialized with the name=value pairs
| in the keyword argument list. For example: dict(one=1, two=2)
|
|
| node_attr_dict_factory = <class 'dict'>
| dict() -> new empty dictionary
| dict(mapping) -> new dictionary initialized from a mapping object's
| (key, value) pairs
| dict(iterable) -> new dictionary initialized as if via:
|     d = {}
|     for k, v in iterable:
|         d[k] = v
| dict(**kwargs) -> new dictionary initialized with the name=value pairs
| in the keyword argument list. For example: dict(one=1, two=2)
|
|
| node_dict_factory = <class 'dict'>
| dict() -> new empty dictionary
| dict(mapping) -> new dictionary initialized from a mapping object's
| (key, value) pairs
| dict(iterable) -> new dictionary initialized as if via:
|     d = {}
|     for k, v in iterable:
|         d[k] = v
| dict(**kwargs) -> new dictionary initialized with the name=value pairs
| in the keyword argument list. For example: dict(one=1, two=2)

```

```

[282]: cust = CustomerNode(
|     pk='id',
|     prefix='cust',
|     fpath='dat/cust.csv',
|     fmt='csv',
|     compute_layer=ComputeLayerEnum.pandas,
| )
|
| order = OrderNode(
|     pk='id',
|     prefix='order',
|     fpath='dat/orders.csv',
|     fmt='csv',

```

```
compute_layer=ComputeLayerEnum.pandas,  
date_key='ts'  
)
```

```
2023-10-25 18:36:07 [warning ] no `date_key` set for <GraphReduceNode:  
fpath=dat/cust.csv fmt=csv>
```

```
[283]: gr = GraphReduce(  
        name='odsc_first_graph',  
        parent_node=cust,  
        fmt='csv',  
        compute_layer=ComputeLayerEnum.pandas,  
        compute_period_val=365,  
        compute_period_unit=PeriodUnit.day,  
        label_period_val=30,  
        label_period_unit=PeriodUnit.day,  
        cut_date=datetime.datetime(2023, 6, 12),  
        has_labels=True  
)
```

```
[284]: # show how attribute push down works  
# gr.hydrate_graph_attrs
```

```
[285]: gr.add_node(cust)  
gr.add_node(order)
```

```
[286]: gr.hydrate_graph_attrs()
```

```
2023-10-25 18:36:12 [info      ] hydrating attributes for CustomerNode  
2023-10-25 18:36:12 [info      ] hydrating attributes for OrderNode
```

```
[287]: order.compute_period_val
```

```
[287]: 365
```

```
[288]: order.cut_date
```

```
[288]: datetime.datetime(2023, 6, 12, 0, 0)
```

```
[289]: # add an edge
```

```
[290]: help(gr.add_entity_edge)
```

Help on method add\_entity\_edge in module graphreduce.graph\_reduce:

```
add_entity_edge(parent_node: graphreduce.node.GraphReduceNode, relation_node:  
graphreduce.node.GraphReduceNode, parent_key: str, relation_key: str,  
relation_type: str = 'parent_child', reduce: bool = True) method of
```

graphreduce.graph\_reduce.GraphReduce instance  
Add an entity relation

```
[291]: gr.add_entity_edge(  
    parent_node=cust,  
    relation_node=order,  
    parent_key='id',  
    relation_key='customer_id',  
    reduce=True  
)
```

```
[292]: gr.do_transformations()
```

```
2023-10-25 18:36:15 [info    ] hydrating graph attributes  
2023-10-25 18:36:15 [info    ] hydrating attributes for CustomerNode  
2023-10-25 18:36:15 [info    ] hydrating attributes for OrderNode  
2023-10-25 18:36:15 [info    ] hydrating graph data  
2023-10-25 18:36:15 [info    ] checking for prefix uniqueness  
2023-10-25 18:36:15 [info    ] running filters, normalize, and annotations for  
<GraphReduceNode: fpath=dat/cust.csv fmt=csv>  
2023-10-25 18:36:15 [info    ] running filters, normalize, and annotations for  
<GraphReduceNode: fpath=dat/orders.csv fmt=csv>  
2023-10-25 18:36:15 [info    ] depth-first traversal through the graph from  
source: <GraphReduceNode: fpath=dat/cust.csv fmt=csv>  
2023-10-25 18:36:15 [info    ] reducing relation <GraphReduceNode:  
fpath=dat/orders.csv fmt=csv>  
2023-10-25 18:36:15 [info    ] joining <GraphReduceNode: fpath=dat/orders.csv  
fmt=csv> to <GraphReduceNode: fpath=dat/cust.csv fmt=csv>  
2023-10-25 18:36:15 [info    ] computed labels for <GraphReduceNode:  
fpath=dat/orders.csv fmt=csv>
```

```
[293]: gr.parent_node.df
```

```
[293]:
```

	cust_id	cust_name	cust_name_length	order_customer_id	order_id_count	\
0	1	wes	3	1	2	
1	2	john	4	2	2	

  

	order_amount_sum	order_customer_id_dupe	order_id_had_order
0	21.5	1	1
1	250.0	2	1

```
[ ]:
```

## 7 Constructing a graph without reducing relations.

```
[268]: cust = CustomerNode(
        pk='id',
        prefix='cust',
        fpath='dat/cust.csv',
        fmt='csv',
        compute_layer=ComputeLayerEnum.pandas,
    )

    order = OrderNode(
        pk='id',
        prefix='order',
        fpath='dat/orders.csv',
        fmt='csv',
        compute_layer=ComputeLayerEnum.pandas,
        date_key='ts'
    )

    gr = GraphReduce(
        name='odsc_first_graph',
        parent_node=cust,
        fmt='csv',
        compute_layer=ComputeLayerEnum.pandas,
        compute_period_val=365,
        compute_period_unit=PeriodUnit.day,
        label_period_val=30,
        label_period_unit=PeriodUnit.day,
        cut_date=datetime.datetime(2023, 6, 12)
    )

    gr.add_node(cust)
    gr.add_node(order)

    gr.add_entity_edge(
        parent_node=cust,
        relation_node=order,
        parent_key='id',
        relation_key='customer_id',
        reduce=False
    )
```

```
2023-10-25 18:35:17 [warning ] no `date_key` set for <GraphReduceNode:
fpath=dat/cust.csv fmt=csv>
```

```
[269]: gr.do_transformations()
```

```
2023-10-25 18:35:24 [info      ] hydrating graph attributes
2023-10-25 18:35:24 [info      ] hydrating attributes for CustomerNode
```

```

2023-10-25 18:35:24 [info      ] hydrating attributes for OrderNode
2023-10-25 18:35:24 [info      ] hydrating graph data
2023-10-25 18:35:24 [info      ] checking for prefix uniqueness
2023-10-25 18:35:24 [info      ] running filters, normalize, and annotations for
<GraphReduceNode: fpath=dat/cust.csv fmt=csv>
2023-10-25 18:35:24 [info      ] running filters, normalize, and annotations for
<GraphReduceNode: fpath=dat/orders.csv fmt=csv>
2023-10-25 18:35:24 [info      ] depth-first traversal through the graph from
source: <GraphReduceNode: fpath=dat/cust.csv fmt=csv>
2023-10-25 18:35:24 [info      ] doing nothing with relation node
<GraphReduceNode: fpath=dat/orders.csv fmt=csv>
2023-10-25 18:35:24 [info      ] joining <GraphReduceNode: fpath=dat/orders.csv
fmt=csv> to <GraphReduceNode: fpath=dat/cust.csv fmt=csv>

```

```
[270]: gr.parent_node.df
```

```

[270]:
  cust_id cust_name  cust_name_length  order_id  order_customer_id \
0         1      wes                  3         1                  1
1         1      wes                  3         2                  1
2         1      wes                  3         5                  1
3         1      wes                  3         7                  1
4         2     john                  4         3                  2
5         2     john                  4         4                  2
6         2     john                  4         6                  2

  order_ts  order_amount
0  2023-05-12         10.0
1  2023-06-01         11.5
2  2023-07-01        325.0
3  2023-07-14       12000.0
4  2023-01-01         100.0
5  2022-08-05         150.0
6  2023-07-02         23.0

```

## 8 Constructing a graph and automating feature generation.

```
[294]: help(GraphReduce)
```

Help on class GraphReduce in module graphreduce.graph\_reduce:

```

class GraphReduce(networkx.classes.digraph.DiGraph)
|   GraphReduce(name: str = 'graph_reduce', parent_node:
Optional[graphreduce.node.GraphReduceNode] = None, fmt: str = 'parquet',
compute_layer: graphreduce.enum.ComputeLayerEnum = None, cut_date:
datetime.datetime = datetime.datetime(2023, 10, 25, 16, 41, 54, 934704),
compute_period_val: Union[int, float] = 365, compute_period_unit:
graphreduce.enum.PeriodUnit = <PeriodUnit.day: 'day'>, has_labels: bool = False,

```

```

label_period_val: Union[int, float, NoneType] = None, label_period_unit:
Optional[graphreduce.enum.PeriodUnit] = None, spark_sqlctx:
pyspark.sql.context.SQLContext = None, feature_function: Optional[str] = None,
dynamic_propagation: bool = False, type_func_map: Dict[str, List[str]] =
{'int64': ['min', 'max', 'sum'], 'str': ['first'], 'object': ['first'],
'float64': ['min', 'max', 'sum'], 'bool': ['first'], 'datetime64': ['first']},
storage_client: Optional[graphreduce.storage.StorageClient] = None, *args,
**kwargs)
|
|   Method resolution order:
|       GraphReduce
|       networkx.classes.digraph.DiGraph
|       networkx.classes.graph.Graph
|       builtins.object
|
|   Methods defined here:
|
|   __init__(self, name: str = 'graph_reduce', parent_node:
Optional[graphreduce.node.GraphReduceNode] = None, fmt: str = 'parquet',
compute_layer: graphreduce.enum.ComputeLayerEnum = None, cut_date:
datetime.datetime = datetime.datetime(2023, 10, 25, 16, 41, 54, 934704),
compute_period_val: Union[int, float] = 365, compute_period_unit:
graphreduce.enum.PeriodUnit = <PeriodUnit.day: 'day'>, has_labels: bool = False,
label_period_val: Union[int, float, NoneType] = None, label_period_unit:
Optional[graphreduce.enum.PeriodUnit] = None, spark_sqlctx:
pyspark.sql.context.SQLContext = None, feature_function: Optional[str] = None,
dynamic_propagation: bool = False, type_func_map: Dict[str, List[str]] =
{'int64': ['min', 'max', 'sum'], 'str': ['first'], 'object': ['first'],
'float64': ['min', 'max', 'sum'], 'bool': ['first'], 'datetime64': ['first']},
storage_client: Optional[graphreduce.storage.StorageClient] = None, *args,
**kwargs)
|       Constructor for GraphReduce
|
|       Args:
|           name : the name of the graph reduce
|           parent_node : parent-most node in the graph, if doing reductions the
granularity to which to reduce the data
|           fmt : the format of the dataset
|           compute_layer : compute layer to use (e.g., spark)
|           cut_date : the date to cut off history
|           compute_period_val : the amount of time to consider during the
compute job
|           compute_period_unit : the unit for the compute period value (e.g.,
day)
|           has_labels : whether or not the compute job computes labels, when
True `prep_for_labels()` and `compute_labels` will be called
|           label_period_val : amount of time to consider when computing labels
|           label_period_unit : the unit for the label period value (e.g., day)

```

```

|         spark_sqlctx : if compute layer is spark this must be passed
|         feature_function : optional custom feature function
|         dynamic_propagation : optional to dynamically propagate children
data upward, useful for very large compute graphs
|         type_func_match : optional mapping from type to a list of functions
(e.g., {'int' : ['min', 'max', 'sum'], 'str' : ['first']})
|
|     __repr__(self)
|         Return repr(self).
|
|     __str__(self)
|         Returns a short summary of the graph.
|
|         Returns
|         -----
|         info : string
|             Graph information as provided by `nx.info`
|
|         Examples
|         -----
|         >>> G = nx.Graph(name="foo")
|         >>> str(G)
|         "Graph named 'foo' with 0 nodes and 0 edges"
|
|         >>> G = nx.path_graph(3)
|         >>> str(G)
|         'Graph with 3 nodes and 2 edges'
|
|     add_entity_edge(self, parent_node: graphreduce.node.GraphReduceNode,
relation_node: graphreduce.node.GraphReduceNode, parent_key: str, relation_key:
str, relation_type: str = 'parent_child', reduce: bool = True)
|         Add an entity relation
|
|     assign_parent(self, parent_node: graphreduce.node.GraphReduceNode)
|         Assign the parent-most node in the graph
|
|     depth_first_generator(self)
|         Depth-first traversal over the edges
|
|     do_transformations(self)
|         Perform all graph transformations
|         1) hydrate graph
|         2) check for duplicate prefixes
|         3) filter data
|         4) clip anomalies
|         5) annotate data
|         6) depth-first edge traversal to: aggregate / reduce features and labels
|         6a) optional alternative feature_function mapping

```



```

|         6b) join back to parent edge
|         6c) post-join annotations if any
|         7) repeat step 6 on all edges up the hierarchy
|
|     get_children(self, node: graphreduce.node.GraphReduceNode) ->
List[graphreduce.node.GraphReduceNode]
|         Get the children of a given node
|
|     hydrate_graph_attrs(self, attrs=['cut_date', 'compute_period_val',
'compute_period_unit', 'has_labels', 'label_period_val', 'label_period_unit',
'compute_layer', 'feature_function', 'spark_sqlctx', '_storage_client'])
|         Hydrate the nodes in the graph with parent
|         attributes in `attrs`
|
|     hydrate_graph_data(self)
|         Hydrate the nodes in the graph with their data
|
|     join(self, parent_node: graphreduce.node.GraphReduceNode, relation_node:
graphreduce.node.GraphReduceNode, relation_df=None)
|         Join the child or peer nnode to the parent node
|
|         Optionally pass the `child_df` directly
|
|     plot_graph(self, fname: str = 'graph.html')
|         Plot the graph
|
|     Args
|         fname : file name to save the graph to - should be .html
|         notebook : whether or not to render in notebook
|
|     prefix_uniqueness(self)
|         Identify children with duplicate prefixes, if any
|
|     -----
|     Readonly properties defined here:
|
|     parent
|
|     -----
|     Methods inherited from networkx.classes.digraph.DiGraph:
|
|     add_edge(self, u_of_edge, v_of_edge, **attr)
|         Add an edge between u and v.
|
|         The nodes u and v will be automatically added if they are
|         not already in the graph.
|
|         Edge attributes can be specified with keywords or by directly

```

accessing the edge's attribute dictionary. See examples below.

#### Parameters

`u_of_edge, v_of_edge` : nodes

Nodes can be, for example, strings or numbers.

Nodes must be hashable (and not None) Python objects.

`attr` : keyword arguments, optional

Edge data (or labels or objects) can be assigned using keyword arguments.

#### See Also

`add_edges_from` : add a collection of edges

#### Notes

Adding an edge that already exists updates the edge data.

Many NetworkX algorithms designed for weighted graphs use an edge attribute (by default `'weight'`) to hold a numerical value.

#### Examples

The following all add the edge `e=(1, 2)` to graph `G`:

```
>>> G = nx.Graph() # or DiGraph, MultiGraph, MultiDiGraph, etc
>>> e = (1, 2)
>>> G.add_edge(1, 2) # explicit two-node form
>>> G.add_edge(*e) # single edge as tuple of two nodes
>>> G.add_edges_from([(1, 2)]) # add edges from iterable container
```

Associate data to edges using keywords:

```
>>> G.add_edge(1, 2, weight=3)
>>> G.add_edge(1, 3, weight=7, capacity=15, length=342.7)
```

For non-string attribute keys, use subscript notation.

```
>>> G.add_edge(1, 2)
>>> G[1][2].update({0: 5})
>>> G.edges[1, 2].update({0: 5})
```

`add_edges_from(self, ebunch_to_add, **attr)`

Add all the edges in `ebunch_to_add`.

#### Parameters

ebunch\_to\_add : container of edges  
 Each edge given in the container will be added to the graph. The edges must be given as 2-tuples (u, v) or 3-tuples (u, v, d) where d is a dictionary containing edge data.  
 attr : keyword arguments, optional  
 Edge data (or labels or objects) can be assigned using keyword arguments.

See Also

-----

add\_edge : add a single edge  
 add\_weighted\_edges\_from : convenient way to add weighted edges

Notes

-----

Adding the same edge twice has no effect but any edge data will be updated when each duplicate edge is added.

Edge attributes specified in an ebunch take precedence over attributes specified via keyword arguments.

Examples

-----

```
>>> G = nx.Graph() # or DiGraph, MultiGraph, MultiDiGraph, etc
>>> G.add_edges_from([(0, 1), (1, 2)]) # using a list of edge tuples
>>> e = zip(range(0, 3), range(1, 4))
>>> G.add_edges_from(e) # Add the path graph 0-1-2-3
```

Associate data to edges

```
>>> G.add_edges_from([(1, 2), (2, 3)], weight=3)
>>> G.add_edges_from([(3, 4), (1, 4)], label="WN2898")
```

```
add_node(self, node_for_adding, **attr)
    Add a single node `node_for_adding` and update node attributes.
```

Parameters

-----

node\_for\_adding : node  
 A node can be any hashable Python object except None.  
 attr : keyword arguments, optional  
 Set or change node attributes using key=value.

See Also

-----

add\_nodes\_from

Examples

```

-----
>>> G = nx.Graph() # or DiGraph, MultiGraph, MultiDiGraph, etc
>>> G.add_node(1)
>>> G.add_node("Hello")
>>> K3 = nx.Graph([(0, 1), (1, 2), (2, 0)])
>>> G.add_node(K3)
>>> G.number_of_nodes()
3

```

Use keywords set/change node attributes:

```

>>> G.add_node(1, size=10)
>>> G.add_node(3, weight=0.4, UTM=("13S", 382871, 3972649))

```

#### Notes

-----  
A hashable object is one that can be used as a key in a Python dictionary. This includes strings, numbers, tuples of strings and numbers, etc.

On many platforms hashable items also include mutables such as NetworkX Graphs, though one should be careful that the hash doesn't change on mutables.

`add_nodes_from(self, nodes_for_adding, **attr)`  
Add multiple nodes.

#### Parameters

-----  
`nodes_for_adding` : iterable container  
A container of nodes (list, dict, set, etc.).  
OR  
A container of (node, attribute dict) tuples.  
Node attributes are updated using the attribute dict.  
`attr` : keyword arguments, optional (default= no attributes)  
Update attributes for all nodes in nodes.  
Node attributes specified in nodes as a tuple take precedence over attributes specified via keyword arguments.

#### See Also

-----  
`add_node`

#### Examples

-----  
>>> G = nx.Graph() # or DiGraph, MultiGraph, MultiDiGraph, etc  
>>> G.add\_nodes\_from("Hello")  
>>> K3 = nx.Graph([(0, 1), (1, 2), (2, 0)])

```

| >>> G.add_nodes_from(K3)
| >>> sorted(G.nodes(), key=str)
| [0, 1, 2, 'H', 'e', 'l', 'o']
|
| Use keywords to update specific node attributes for every node.
|
| >>> G.add_nodes_from([1, 2], size=10)
| >>> G.add_nodes_from([3, 4], weight=0.4)
|
| Use (node, attrdict) tuples to update attributes for specific nodes.
|
| >>> G.add_nodes_from([(1, dict(size=11)), (2, {"color": "blue"})])
| >>> G.nodes[1]["size"]
| 11
| >>> H = nx.Graph()
| >>> H.add_nodes_from(G.nodes(data=True))
| >>> H.nodes[1]["size"]
| 11
|
| adj = <functools.cached_property object>
| Graph adjacency object holding the neighbors of each node.
|
| This object is a read-only dict-like structure with node keys
| and neighbor-dict values. The neighbor-dict is keyed by neighbor
| to the edge-data-dict. So `G.adj[3][2]['color'] = 'blue'` sets
| the color of the edge `(3, 2)` to `"blue"`.
|
| Iterating over G.adj behaves like a dict. Useful idioms include
| `for nbr, datadict in G.adj[n].items():`.
|
| The neighbor information is also provided by subscripting the graph.
| So `for nbr, foovalue in G[node].data('foo', default=1):` works.
|
| For directed graphs, `G.adj` holds outgoing (successor) info.
|
| clear(self)
| Remove all nodes and edges from the graph.
|
| This also removes the name, and all graph, node, and edge attributes.
|
| Examples
| -----
| >>> G = nx.path_graph(4) # or DiGraph, MultiGraph, MultiDiGraph, etc
| >>> G.clear()
| >>> list(G.nodes)
| []
| >>> list(G.edges)
| []

```

```

|
| clear_edges(self)
|     Remove all edges from the graph without altering nodes.
|
|     Examples
|     -----
|     >>> G = nx.path_graph(4) # or DiGraph, MultiGraph, MultiDiGraph, etc
|     >>> G.clear_edges()
|     >>> list(G.nodes)
|     [0, 1, 2, 3]
|     >>> list(G.edges)
|     []
|
| degree = <functools.cached_property object>
|     A DegreeView for the Graph as G.degree or G.degree().
|
|     The node degree is the number of edges adjacent to the node.
|     The weighted node degree is the sum of the edge weights for
|     edges incident to that node.
|
|     This object provides an iterator for (node, degree) as well as
|     lookup for the degree for a single node.
|
|     Parameters
|     -----
|     nbunch : single node, container, or all nodes (default= all nodes)
|         The view will only report edges incident to these nodes.
|
|     weight : string or None, optional (default=None)
|         The name of an edge attribute that holds the numerical value used
|         as a weight. If None, then each edge has weight 1.
|         The degree is the sum of the edge weights adjacent to the node.
|
|     Returns
|     -----
|     DiDegreeView or int
|         If multiple nodes are requested (the default), returns a
|         `DiDegreeView`
|         mapping nodes to their degree.
|         If a single node is requested, returns the degree of the node as an
integer.
|
|     See Also
|     -----
|     in_degree, out_degree
|
|     Examples
|     -----

```

```

| >>> G = nx.DiGraph() # or MultiDiGraph
| >>> nx.add_path(G, [0, 1, 2, 3])
| >>> G.degree(0) # node 0 with degree 1
| 1
| >>> list(G.degree([0, 1, 2]))
| [(0, 1), (1, 2), (2, 2)]
|
| edges = <functools.cached_property object>
| An OutEdgeView of the DiGraph as G.edges or G.edges().
|
| edges(self, nbunch=None, data=False, default=None)
|
| The OutEdgeView provides set-like operations on the edge-tuples
| as well as edge attribute lookup. When called, it also provides
| an EdgeDataView object which allows control of access to edge
| attributes (but does not provide set-like operations).
| Hence, `G.edges[u, v]['color']` provides the value of the color
| attribute for edge `(u, v)` while
| `for (u, v, c) in G.edges.data('color', default='red'):`
| iterates through all the edges yielding the color attribute
| with default `red` if no color attribute exists.
|
| Parameters
| -----
| nbunch : single node, container, or all nodes (default= all nodes)
|         The view will only report edges from these nodes.
| data : string or bool, optional (default=False)
|       The edge attribute returned in 3-tuple (u, v, ddict[data]).
|       If True, return edge attribute dict in 3-tuple (u, v, ddict).
|       If False, return 2-tuple (u, v).
| default : value, optional (default=None)
|         Value used for edges that don't have the requested attribute.
|         Only relevant if data is not True or False.
|
| Returns
| -----
| edges : OutEdgeView
|       A view of edge attributes, usually it iterates over (u, v)
|       or (u, v, d) tuples of edges, but can also be used for
|       attribute lookup as `edges[u, v]['foo']`.
|
| See Also
| -----
| in_edges, out_edges
|
| Notes
| -----
| Nodes in nbunch that are not in the graph will be (quietly) ignored.

```

For directed graphs this returns the out-edges.

#### Examples

-----

```
>>> G = nx.DiGraph() # or MultiDiGraph, etc
>>> nx.add_path(G, [0, 1, 2])
>>> G.add_edge(2, 3, weight=5)
>>> [e for e in G.edges]
[(0, 1), (1, 2), (2, 3)]
>>> G.edges.data() # default data is {} (empty dict)
OutEdgeDataView([(0, 1, {}), (1, 2, {}), (2, 3, {'weight': 5})])
>>> G.edges.data("weight", default=1)
OutEdgeDataView([(0, 1, 1), (1, 2, 1), (2, 3, 5)])
>>> G.edges([0, 2]) # only edges originating from these nodes
OutEdgeDataView([(0, 1), (2, 3)])
>>> G.edges(0) # only edges from node 0
OutEdgeDataView([(0, 1)])
```

`has_predecessor(self, u, v)`  
Returns True if node u has predecessor v.

This is true if graph has the edge  $u \leftarrow v$ .

`has_successor(self, u, v)`  
Returns True if node u has successor v.

This is true if graph has the edge  $u \rightarrow v$ .

`in_degree = <functools.cached_property object>`  
An InDegreeView for (node, in\_degree) or in\_degree for single node.

The node in\_degree is the number of edges pointing to the node.  
The weighted node degree is the sum of the edge weights for edges incident to that node.

This object provides an iteration over (node, in\_degree) as well as lookup for the degree for a single node.

#### Parameters

-----

`nbunch` : single node, container, or all nodes (default= all nodes)  
The view will only report edges incident to these nodes.

`weight` : string or None, optional (default=None)  
The name of an edge attribute that holds the numerical value used as a weight. If None, then each edge has weight 1.  
The degree is the sum of the edge weights adjacent to the node.



Returns

-----

If a single node is requested

deg : int

In-degree of the node

OR if multiple nodes are requested

nd\_iter : iterator

The iterator returns two-tuples of (node, in-degree).

See Also

-----

degree, out\_degree

Examples

-----

```
>>> G = nx.DiGraph()
>>> nx.add_path(G, [0, 1, 2, 3])
>>> G.in_degree(0) # node 0 with degree 0
0
>>> list(G.in_degree([0, 1, 2]))
[(0, 0), (1, 1), (2, 1)]
```

in\_edges = <functools.cached\_property object>

An InEdgeView of the Graph as G.in\_edges or G.in\_edges().

in\_edges(self, nbunch=None, data=False, default=None):

Parameters

-----

nbunch : single node, container, or all nodes (default= all nodes)

The view will only report edges incident to these nodes.

data : string or bool, optional (default=False)

The edge attribute returned in 3-tuple (u, v, ddict[data]).

If True, return edge attribute dict in 3-tuple (u, v, ddict).

If False, return 2-tuple (u, v).

default : value, optional (default=None)

Value used for edges that don't have the requested attribute.

Only relevant if data is not True or False.

Returns

-----

in\_edges : InEdgeView

A view of edge attributes, usually it iterates over (u, v)

or (u, v, d) tuples of edges, but can also be used for

attribute lookup as `edges[u, v]['foo']`.

See Also

```

|         -----
|         edges
|
| is_directed(self)
|     Returns True if graph is directed, False otherwise.
|
| is_multigraph(self)
|     Returns True if graph is a multigraph, False otherwise.
|
| neighbors = successors(self, n)
|
| out_degree = <functools.cached_property object>
|     An OutDegreeView for (node, out_degree)
|
|     The node out_degree is the number of edges pointing out of the node.
|     The weighted node degree is the sum of the edge weights for
|     edges incident to that node.
|
|     This object provides an iterator over (node, out_degree) as well as
|     lookup for the degree for a single node.
|
|     Parameters
|     -----
|
|     nbunch : single node, container, or all nodes (default= all nodes)
|         The view will only report edges incident to these nodes.
|
|     weight : string or None, optional (default=None)
|         The name of an edge attribute that holds the numerical value used
|         as a weight.  If None, then each edge has weight 1.
|         The degree is the sum of the edge weights adjacent to the node.
|
|     Returns
|     -----
|
|     If a single node is requested
|     deg : int
|         Out-degree of the node
|
|     OR if multiple nodes are requested
|     nd_iter : iterator
|         The iterator returns two-tuples of (node, out-degree).
|
|     See Also
|     -----
|
|     degree, in_degree
|
|     Examples
|     -----
|
|     >>> G = nx.DiGraph()

```

```

| >>> nx.add_path(G, [0, 1, 2, 3])
| >>> G.out_degree(0) # node 0 with degree 1
| 1
| >>> list(G.out_degree([0, 1, 2]))
| [(0, 1), (1, 1), (2, 1)]
|
| out_edges = <functools.cached_property object>
| An OutEdgeView of the DiGraph as G.edges or G.edges().
|
| edges(self, nbunch=None, data=False, default=None)
|
| The OutEdgeView provides set-like operations on the edge-tuples
| as well as edge attribute lookup. When called, it also provides
| an EdgeDataView object which allows control of access to edge
| attributes (but does not provide set-like operations).
| Hence, `G.edges[u, v]['color']` provides the value of the color
| attribute for edge `(u, v)` while
| `for (u, v, c) in G.edges.data('color', default='red'):`
| iterates through all the edges yielding the color attribute
| with default `'red'` if no color attribute exists.
|
| Parameters
| -----
| nbunch : single node, container, or all nodes (default= all nodes)
|         The view will only report edges from these nodes.
| data : string or bool, optional (default=False)
|       The edge attribute returned in 3-tuple (u, v, ddict[data]).
|       If True, return edge attribute dict in 3-tuple (u, v, ddict).
|       If False, return 2-tuple (u, v).
| default : value, optional (default=None)
|         Value used for edges that don't have the requested attribute.
|         Only relevant if data is not True or False.
|
| Returns
| -----
| edges : OutEdgeView
|       A view of edge attributes, usually it iterates over (u, v)
|       or (u, v, d) tuples of edges, but can also be used for
|       attribute lookup as `edges[u, v]['foo']`.
|
| See Also
| -----
| in_edges, out_edges
|
| Notes
| -----
| Nodes in nbunch that are not in the graph will be (quietly) ignored.
| For directed graphs this returns the out-edges.

```

## Examples

-----

```
>>> G = nx.DiGraph() # or MultiDiGraph, etc
>>> nx.add_path(G, [0, 1, 2])
>>> G.add_edge(2, 3, weight=5)
>>> [e for e in G.edges]
[(0, 1), (1, 2), (2, 3)]
>>> G.edges.data() # default data is {} (empty dict)
OutEdgeDataView([(0, 1, {}), (1, 2, {}), (2, 3, {'weight': 5})])
>>> G.edges.data("weight", default=1)
OutEdgeDataView([(0, 1, 1), (1, 2, 1), (2, 3, 5)])
>>> G.edges([0, 2]) # only edges originating from these nodes
OutEdgeDataView([(0, 1), (2, 3)])
>>> G.edges(0) # only edges from node 0
OutEdgeDataView([(0, 1)])
```

pred = <functools.cached\_property object>

Graph adjacency object holding the predecessors of each node.

This object is a read-only dict-like structure with node keys and neighbor-dict values. The neighbor-dict is keyed by neighbor to the edge-data-dict. So `G.pred[2][3]['color'] = 'blue'` sets the color of the edge `(3, 2)` to `"blue"`.

Iterating over `G.pred` behaves like a dict. Useful idioms include `for nbr, datadict in G.pred[n].items():`. A data-view not provided by dicts also exists: `for nbr, foovalue in G.pred[node].data('foo'):`. A default can be set via a `default` argument to the `data` method.

predecessors(self, n)

Returns an iterator over predecessor nodes of n.

A predecessor of n is a node m such that there exists a directed edge from m to n.

## Parameters

-----

n : node

A node in the graph

## Raises

-----

NetworkXError

If n is not in the graph.

## See Also

-----

```

|     successors
|
| remove_edge(self, u, v)
|     Remove the edge between u and v.
|
|     Parameters
|     -----
|     u, v : nodes
|         Remove the edge between nodes u and v.
|
|     Raises
|     -----
|     NetworkXError
|         If there is not an edge between u and v.
|
|     See Also
|     -----
|     remove_edges_from : remove a collection of edges
|
|     Examples
|     -----
|     >>> G = nx.Graph() # or DiGraph, etc
|     >>> nx.add_path(G, [0, 1, 2, 3])
|     >>> G.remove_edge(0, 1)
|     >>> e = (1, 2)
|     >>> G.remove_edge(*e) # unpacks e from an edge tuple
|     >>> e = (2, 3, {"weight": 7}) # an edge with attribute data
|     >>> G.remove_edge(*e[:2]) # select first part of edge tuple
|
| remove_edges_from(self, ebunch)
|     Remove all edges specified in ebunch.
|
|     Parameters
|     -----
|     ebunch: list or container of edge tuples
|         Each edge given in the list or container will be removed
|         from the graph. The edges can be:
|
|             - 2-tuples (u, v) edge between u and v.
|             - 3-tuples (u, v, k) where k is ignored.
|
|     See Also
|     -----
|     remove_edge : remove a single edge
|
|     Notes
|     -----
|     Will fail silently if an edge in ebunch is not in the graph.

```

### Examples

-----

```
>>> G = nx.path_graph(4) # or DiGraph, MultiGraph, MultiDiGraph, etc
>>> ebunch = [(1, 2), (2, 3)]
>>> G.remove_edges_from(ebunch)
```

```
remove_node(self, n)
```

Remove node n.

Removes the node n and all adjacent edges.

Attempting to remove a non-existent node will raise an exception.

### Parameters

-----

n : node

A node in the graph

### Raises

-----

NetworkXError

If n is not in the graph.

### See Also

-----

remove\_nodes\_from

### Examples

-----

```
>>> G = nx.path_graph(3) # or DiGraph, MultiGraph, MultiDiGraph, etc
>>> list(G.edges)
[(0, 1), (1, 2)]
>>> G.remove_node(1)
>>> list(G.edges)
[]
```

```
remove_nodes_from(self, nodes)
```

Remove multiple nodes.

### Parameters

-----

nodes : iterable container

A container of nodes (list, dict, set, etc.). If a node in the container is not in the graph it is silently ignored.

### See Also

-----

remove\_node

## Examples

```
-----
>>> G = nx.path_graph(3) # or DiGraph, MultiGraph, MultiDiGraph, etc
>>> e = list(G.nodes)
>>> e
[0, 1, 2]
>>> G.remove_nodes_from(e)
>>> list(G.nodes)
[]
```

`reverse(self, copy=True)`

Returns the reverse of the graph.

The reverse is a graph with the same nodes and edges but with the directions of the edges reversed.

### Parameters

`copy` : bool optional (default=True)

If True, return a new DiGraph holding the reversed edges.

If False, the reverse graph is created using a view of the original graph.

`succ = <functools.cached_property object>`

Graph adjacency object holding the successors of each node.

This object is a read-only dict-like structure with node keys and neighbor-dict values. The neighbor-dict is keyed by neighbor to the edge-data-dict. So `G.succ[3][2]['color'] = 'blue'` sets the color of the edge `(3, 2)` to `"blue"`.

Iterating over `G.succ` behaves like a dict. Useful idioms include `for nbr, datadict in G.succ[n].items():`. A data-view not provided by dicts also exists: `for nbr, foovalue in G.succ[node].data('foo'):` and a default can be set via a `default` argument to the `data` method.

The neighbor information is also provided by subscripting the graph. So `for nbr, foovalue in G[node].data('foo', default=1):` works.

For directed graphs, `G.adj` is identical to `G.succ`.

`successors(self, n)`

Returns an iterator over successor nodes of `n`.

A successor of `n` is a node `m` such that there exists a directed edge from `n` to `m`.

#### Parameters

-----

n : node

A node in the graph

#### Raises

-----

NetworkXError

If n is not in the graph.

#### See Also

-----

predecessors

#### Notes

-----

neighbors() and successors() are the same.

to\_undirected(self, reciprocal=False, as\_view=False)

Returns an undirected representation of the digraph.

#### Parameters

-----

reciprocal : bool (optional)

If True only keep edges that appear in both directions  
in the original digraph.

as\_view : bool (optional, default=False)

If True return an undirected view of the original directed graph.

#### Returns

-----

G : Graph

An undirected graph with the same name and nodes and  
with edge (u, v, data) if either (u, v, data) or (v, u, data)  
is in the digraph. If both edges exist in digraph and  
their edge data is different, only one edge is created  
with an arbitrary choice of which edge data to use.  
You must check and correct for this manually if desired.

#### See Also

-----

Graph, copy, add\_edge, add\_edges\_from

#### Notes

-----

If edges in both directions (u, v) and (v, u) exist in the  
graph, attributes for the new undirected edge will be a combination of  
the attributes of the directed edges. The edge data is updated



in the (arbitrary) order that the edges are encountered. For more customized control of the edge attributes use `add_edge()`.

This returns a "deepcopy" of the edge, node, and graph attributes which attempts to completely copy all of the data and references.

This is in contrast to the similar `G=DiGraph(D)` which returns a shallow copy of the data.

See the Python copy module for more information on shallow and deep copies, <https://docs.python.org/3/library/copy.html>.

Warning: If you have subclassed `DiGraph` to use dict-like objects in the data structure, those changes do not transfer to the Graph created by this method.

#### Examples

```
-----
>>> G = nx.path_graph(2) # or MultiGraph, etc
>>> H = G.to_directed()
>>> list(H.edges)
[(0, 1), (1, 0)]
>>> G2 = H.to_undirected()
>>> list(G2.edges)
[(0, 1)]
```

---

Methods inherited from `networkx.classes.graph.Graph`:

`__contains__(self, n)`  
Returns True if `n` is a node, False otherwise. Use: '`n in G`'.

#### Examples

```
-----
>>> G = nx.path_graph(4) # or DiGraph, MultiGraph, MultiDiGraph, etc
>>> 1 in G
True
```

`__getitem__(self, n)`  
Returns a dict of neighbors of node `n`. Use: '`G[n]`'.

#### Parameters

`n` : node  
A node in the graph.

Returns

```

|         -----
|         adj_dict : dictionary
|             The adjacency dictionary for nodes connected to n.
|
|         Notes
|         -----
|         G[n] is the same as G.adj[n] and similar to G.neighbors(n)
|         (which is an iterator over G.adj[n])
|
|         Examples
|         -----
|         >>> G = nx.path_graph(4) # or DiGraph, MultiGraph, MultiDiGraph, etc
|         >>> G[0]
|         AtlasView({1: {}})
|
|     __iter__(self)
|         Iterate over the nodes. Use: 'for n in G'.
|
|     Returns
|     -----
|     niter : iterator
|         An iterator over all nodes in the graph.
|
|     Examples
|     -----
|     >>> G = nx.path_graph(4) # or DiGraph, MultiGraph, MultiDiGraph, etc
|     >>> [n for n in G]
|     [0, 1, 2, 3]
|     >>> list(G)
|     [0, 1, 2, 3]
|
|     __len__(self)
|         Returns the number of nodes in the graph. Use: 'len(G)'.
|
|     Returns
|     -----
|     nnodes : int
|         The number of nodes in the graph.
|
|     See Also
|     -----
|     number_of_nodes: identical method
|     order: identical method
|
|     Examples
|     -----
|     >>> G = nx.path_graph(4) # or DiGraph, MultiGraph, MultiDiGraph, etc
|     >>> len(G)

```

4

`add_weighted_edges_from(self, ebunch_to_add, weight='weight', **attr)`  
Add weighted edges in ``ebunch_to_add`` with specified weight attr

Parameters

`ebunch_to_add` : container of edges

Each edge given in the list or container will be added to the graph. The edges must be given as 3-tuples (u, v, w) where w is a number.

`weight` : string, optional (default= 'weight')

The attribute name for the edge weights to be added.

`attr` : keyword arguments, optional (default= no attributes)

Edge attributes to add/update for all edges.

See Also

`add_edge` : add a single edge

`add_edges_from` : add multiple edges

Notes

Adding the same edge twice for Graph/DiGraph simply updates the edge data. For MultiGraph/MultiDiGraph, duplicate edges are stored.

Examples

```
>>> G = nx.Graph() # or DiGraph, MultiGraph, MultiDiGraph, etc
>>> G.add_weighted_edges_from([(0, 1, 3.0), (1, 2, 7.5)])
```

`adjacency(self)`

Returns an iterator over (node, adjacency dict) tuples for all nodes.

For directed graphs, only outgoing neighbors/adjacencies are included.

Returns

`adj_iter` : iterator

An iterator over (node, adjacency dictionary) for all nodes in the graph.

Examples

```
>>> G = nx.path_graph(4) # or DiGraph, MultiGraph, MultiDiGraph, etc
>>> [(n, nbrdict) for n, nbrdict in G.adjacency()]
[(0, {1: {}}), (1, {0: {}, 2: {}}), (2, {1: {}, 3: {}}), (3, {2: {}})]
```

```
copy(self, as_view=False)
    Returns a copy of the graph.
```

The copy method by default returns an independent shallow copy of the graph and attributes. That is, if an attribute is a container, that container is shared by the original and the copy. Use Python's `copy.deepcopy` for new containers.

If `as_view` is True then a view is returned instead of a copy.

#### Notes

-----

All copies reproduce the graph structure, but data attributes may be handled in different ways. There are four types of copies of a graph that people might want.

**Deepcopy** -- A "deepcopy" copies the graph structure as well as all data attributes and any objects they might contain. The entire graph object is new so that changes in the copy do not affect the original object. (see Python's `copy.deepcopy`)

**Data Reference (Shallow)** -- For a shallow copy the graph structure is copied but the edge, node and graph attribute dicts are references to those in the original graph. This saves time and memory but could cause confusion if you change an attribute in one graph and it changes the attribute in the other. NetworkX does not provide this level of shallow copy.

**Independent Shallow** -- This copy creates new independent attribute dicts and then does a shallow copy of the attributes. That is, any attributes that are containers are shared between the new graph and the original. This is exactly what `dict.copy()` provides. You can obtain this style copy using:

```
>>> G = nx.path_graph(5)
>>> H = G.copy()
>>> H = G.copy(as_view=False)
>>> H = nx.Graph(G)
>>> H = G.__class__(G)
```

**Fresh Data** -- For fresh data, the graph structure is copied while new empty data attribute dicts are created. The resulting graph is independent of the original and it has no edge, node or graph attributes. Fresh copies are not enabled. Instead use:

```
>>> H = G.__class__()
>>> H.add_nodes_from(G)
```

```

|         >>> H.add_edges_from(G.edges)
|
| View -- Inspired by dict-views, graph-views act like read-only
| versions of the original graph, providing a copy of the original
| structure without requiring any memory for copying the information.
|
| See the Python copy module for more information on shallow
| and deep copies, https://docs.python.org/3/library/copy.html.
|
| Parameters
| -----
| as_view : bool, optional (default=False)
|     If True, the returned graph-view provides a read-only view
|     of the original graph without actually copying any data.
|
| Returns
| -----
| G : Graph
|     A copy of the graph.
|
| See Also
| -----
| to_directed: return a directed copy of the graph.
|
| Examples
| -----
| >>> G = nx.path_graph(4) # or DiGraph, MultiGraph, MultiDiGraph, etc
| >>> H = G.copy()
|
| edge_subgraph(self, edges)
|     Returns the subgraph induced by the specified edges.
|
| The induced subgraph contains each edge in `edges` and each
| node incident to any one of those edges.
|
| Parameters
| -----
| edges : iterable
|     An iterable of edges in this graph.
|
| Returns
| -----
| G : Graph
|     An edge-induced subgraph of this graph with the same edge
|     attributes.
|
| Notes
| -----

```

The graph, edge, and node attributes in the returned subgraph view are references to the corresponding attributes in the original graph. The view is read-only.

To create a full graph version of the subgraph with its own copy of the edge or node attributes, use::

```
G.edge_subgraph(edges).copy()
```

Examples

```
-----
>>> G = nx.path_graph(5)
>>> H = G.edge_subgraph([(0, 1), (3, 4)])
>>> list(H.nodes)
[0, 1, 3, 4]
>>> list(H.edges)
[(0, 1), (3, 4)]
```

```
get_edge_data(self, u, v, default=None)
```

Returns the attribute dictionary associated with edge (u, v).

This is identical to `G[u][v]` except the default is returned instead of an exception if the edge doesn't exist.

Parameters

```
-----
u, v : nodes
default: any Python object (default=None)
        Value to return if the edge (u, v) is not found.
```

Returns

```
-----
edge_dict : dictionary
        The edge attribute dictionary.
```

Examples

```
-----
>>> G = nx.path_graph(4) # or DiGraph, MultiGraph, MultiDiGraph, etc
>>> G[0][1]
{}
```

Warning: Assigning to `G[u][v]` is not permitted.  
But it is safe to assign attributes `G[u][v]['foo']`

```
>>> G[0][1]["weight"] = 7
>>> G[0][1]["weight"]
7
>>> G[1][0]["weight"]
```

7

```
>>> G = nx.path_graph(4) # or DiGraph, MultiGraph, MultiDiGraph, etc
>>> G.get_edge_data(0, 1) # default edge data is {}
{}
>>> e = (0, 1)
>>> G.get_edge_data(*e) # tuple form
{}
>>> G.get_edge_data("a", "b", default=0) # edge not in graph, return 0
0
```

`has_edge(self, u, v)`

Returns True if the edge (u, v) is in the graph.

This is the same as ``v in G[u]`` without `KeyError` exceptions.

Parameters

-----

u, v : nodes

Nodes can be, for example, strings or numbers.

Nodes must be hashable (and not None) Python objects.

Returns

-----

edge\_ind : bool

True if edge is in the graph, False otherwise.

Examples

-----

```
>>> G = nx.path_graph(4) # or DiGraph, MultiGraph, MultiDiGraph, etc
>>> G.has_edge(0, 1) # using two nodes
True
>>> e = (0, 1)
>>> G.has_edge(*e) # e is a 2-tuple (u, v)
True
>>> e = (0, 1, {"weight": 7})
>>> G.has_edge(*e[:2]) # e is a 3-tuple (u, v, data_dictionary)
True
```

The following syntax are equivalent:

```
>>> G.has_edge(0, 1)
True
>>> 1 in G[0] # though this gives KeyError if 0 not in G
True
```

`has_node(self, n)`

Returns True if the graph contains the node n.

Identical to ``n in G``

Parameters

-----

`n` : node

Examples

-----

```
>>> G = nx.path_graph(3) # or DiGraph, MultiGraph, MultiDiGraph, etc
```

```
>>> G.has_node(0)
```

```
True
```

It is more readable and simpler to use

```
>>> 0 in G
```

```
True
```

`nbunch_iter(self, nbunch=None)`

Returns an iterator over nodes contained in `nbunch` that are also in the graph.

The nodes in `nbunch` are checked for membership in the graph and if not are silently ignored.

Parameters

-----

`nbunch` : single node, container, or all nodes (default= all nodes)

The view will only report edges incident to these nodes.

Returns

-----

`niter` : iterator

An iterator over nodes in `nbunch` that are also in the graph.

If `nbunch` is `None`, iterate over all nodes in the graph.

Raises

-----

`NetworkXError`

If `nbunch` is not a node or sequence of nodes.

If a node in `nbunch` is not hashable.

See Also

-----

`Graph.__iter__`

Notes

-----



When nbunch is an iterator, the returned iterator yields values directly from nbunch, becoming exhausted when nbunch is exhausted.

To test whether nbunch is a single node, one can use "if nbunch in self:", even after processing with this routine.

If nbunch is not a node or a (possibly empty) sequence/iterator or None, a :exc:`NetworkXError` is raised. Also, if any object in nbunch is not hashable, a :exc:`NetworkXError` is raised.

nodes = <functools.cached\_property object>

A NodeView of the Graph as G.nodes or G.nodes().

Can be used as `G.nodes` for data lookup and for set-like operations. Can also be used as `G.nodes(data='color', default=None)` to return a NodeDataView which reports specific node data but no set operations. It presents a dict-like interface as well with `G.nodes.items()` iterating over `(node, nodedata)` 2-tuples and `G.nodes[3]['foo']` providing the value of the `foo` attribute for node `3`. In addition, a view `G.nodes.data('foo')` provides a dict-like interface to the `foo` attribute of each node. `G.nodes.data('foo', default=1)` provides a default for nodes that do not have attribute `foo`.

#### Parameters

-----

data : string or bool, optional (default=False)

The node attribute returned in 2-tuple (n, ddict[data]).

If True, return entire node attribute dict as (n, ddict).

If False, return just the nodes n.

default : value, optional (default=None)

Value used for nodes that don't have the requested attribute.

Only relevant if data is not True or False.

#### Returns

-----

##### NodeView

Allows set-like operations over the nodes as well as node attribute dict lookup and calling to get a NodeDataView.

A NodeDataView iterates over `(n, data)` and has no set operations.

A NodeView iterates over `n` and includes set operations.

When called, if data is False, an iterator over nodes.

Otherwise an iterator of 2-tuples (node, attribute value)

where the attribute is specified in `data`.

If data is True then the attribute becomes the entire data dictionary.

## Notes

-----

If your node data is not needed, it is simpler and equivalent to use the expression `for n in G`, or `list(G)`.

## Examples

-----

There are two simple ways of getting a list of all nodes in the graph:

```
>>> G = nx.path_graph(3)
>>> list(G.nodes)
[0, 1, 2]
>>> list(G)
[0, 1, 2]
```

To get the node data along with the nodes:

```
>>> G.add_node(1, time="5pm")
>>> G.nodes[0]["foo"] = "bar"
>>> list(G.nodes(data=True))
[(0, {'foo': 'bar'}), (1, {'time': '5pm'}), (2, {})]
>>> list(G.nodes.data())
[(0, {'foo': 'bar'}), (1, {'time': '5pm'}), (2, {})]

>>> list(G.nodes(data="foo"))
[(0, 'bar'), (1, None), (2, None)]
>>> list(G.nodes.data("foo"))
[(0, 'bar'), (1, None), (2, None)]

>>> list(G.nodes(data="time"))
[(0, None), (1, '5pm'), (2, None)]
>>> list(G.nodes.data("time"))
[(0, None), (1, '5pm'), (2, None)]

>>> list(G.nodes(data="time", default="Not Available"))
[(0, 'Not Available'), (1, '5pm'), (2, 'Not Available')]
>>> list(G.nodes.data("time", default="Not Available"))
[(0, 'Not Available'), (1, '5pm'), (2, 'Not Available')]
```

If some of your nodes have an attribute and the rest are assumed to have a default attribute value you can create a dictionary from node/attribute pairs using the `default` keyword argument to guarantee the value is never `None`:

```
>>> G = nx.Graph()
>>> G.add_node(0)
>>> G.add_node(1, weight=2)
>>> G.add_node(2, weight=3)
```

```

>>> dict(G.nodes(data="weight", default=1))
{0: 1, 1: 2, 2: 3}

number_of_edges(self, u=None, v=None)
    Returns the number of edges between two nodes.

Parameters
-----
u, v : nodes, optional (default=all edges)
    If u and v are specified, return the number of edges between
    u and v. Otherwise return the total number of all edges.

Returns
-----
nedges : int
    The number of edges in the graph. If nodes `u` and `v` are
    specified return the number of edges between those nodes. If
    the graph is directed, this only returns the number of edges
    from `u` to `v`.

See Also
-----
size

Examples
-----
For undirected graphs, this method counts the total number of
edges in the graph:

>>> G = nx.path_graph(4)
>>> G.number_of_edges()
3

If you specify two nodes, this counts the total number of edges
joining the two nodes:

>>> G.number_of_edges(0, 1)
1

For directed graphs, this method can count the total number of
directed edges from `u` to `v`:

>>> G = nx.DiGraph()
>>> G.add_edge(0, 1)
>>> G.add_edge(1, 0)
>>> G.number_of_edges(0, 1)
1

```

```

| number_of_nodes(self)
|     Returns the number of nodes in the graph.
|
|     Returns
|     -----
|     nnodes : int
|         The number of nodes in the graph.
|
|     See Also
|     -----
|     order: identical method
|     __len__: identical method
|
|     Examples
|     -----
|     >>> G = nx.path_graph(3) # or DiGraph, MultiGraph, MultiDiGraph, etc
|     >>> G.number_of_nodes()
|     3
|
| order(self)
|     Returns the number of nodes in the graph.
|
|     Returns
|     -----
|     nnodes : int
|         The number of nodes in the graph.
|
|     See Also
|     -----
|     number_of_nodes: identical method
|     __len__: identical method
|
|     Examples
|     -----
|     >>> G = nx.path_graph(3) # or DiGraph, MultiGraph, MultiDiGraph, etc
|     >>> G.order()
|     3
|
| size(self, weight=None)
|     Returns the number of edges or total of all edge weights.
|
|     Parameters
|     -----
|     weight : string or None, optional (default=None)
|         The edge attribute that holds the numerical value used
|         as a weight. If None, then each edge has weight 1.
|
|     Returns

```

-----

size : numeric

The number of edges or  
(if weight keyword is provided) the total weight sum.

If weight is None, returns an int. Otherwise a float  
(or more general numeric if the weights are more general).

See Also

-----

number\_of\_edges

Examples

-----

```
>>> G = nx.path_graph(4) # or DiGraph, MultiGraph, MultiDiGraph, etc
```

```
>>> G.size()
```

```
3
```

```
>>> G = nx.Graph() # or DiGraph, MultiGraph, MultiDiGraph, etc
```

```
>>> G.add_edge("a", "b", weight=2)
```

```
>>> G.add_edge("b", "c", weight=4)
```

```
>>> G.size()
```

```
2
```

```
>>> G.size(weight="weight")
```

```
6.0
```

subgraph(self, nodes)

Returns a SubGraph view of the subgraph induced on `nodes`.

The induced subgraph of the graph contains the nodes in `nodes`  
and the edges between those nodes.

Parameters

-----

nodes : list, iterable

A container of nodes which will be iterated through once.

Returns

-----

G : SubGraph View

A subgraph view of the graph. The graph structure cannot be  
changed but node/edge attributes can and are shared with the  
original graph.

Notes

-----

The graph, edge and node attributes are shared with the original graph.  
Changes to the graph structure is ruled out by the view, but changes

to attributes are reflected in the original graph.

To create a subgraph with its own copy of the edge/node attributes use:  
`G.subgraph(nodes).copy()`

For an inplace reduction of a graph to a subgraph you can remove nodes:  
`G.remove_nodes_from([n for n in G if n not in set(nodes)])`

Subgraph views are sometimes NOT what you want. In most cases where you want to do more than simply look at the induced edges, it makes more sense to just create the subgraph as its own graph with code like:

::

```
# Create a subgraph SG based on a (possibly multigraph) G
SG = G.__class__()
SG.add_nodes_from((n, G.nodes[n]) for n in largest_wcc)
if SG.is_multigraph():
    SG.add_edges_from((n, nbr, key, d)
                      for n, nbrs in G.adj.items() if n in largest_wcc
                      for nbr, keydict in nbrs.items() if nbr in largest_wcc
                      for key, d in keydict.items())
else:
    SG.add_edges_from((n, nbr, d)
                      for n, nbrs in G.adj.items() if n in largest_wcc
                      for nbr, d in nbrs.items() if nbr in largest_wcc)
SG.graph.update(G.graph)
```

Examples

-----

```
>>> G = nx.path_graph(4) # or DiGraph, MultiGraph, MultiDiGraph, etc
>>> H = G.subgraph([0, 1, 2])
>>> list(H.edges)
[(0, 1), (1, 2)]
```

`to_directed(self, as_view=False)`

Returns a directed representation of the graph.

Returns

-----

G : DiGraph

A directed graph with the same name, same nodes, and with each edge (u, v, data) replaced by two directed edges (u, v, data) and (v, u, data).

Notes

-----

This returns a "deepcopy" of the edge, node, and

graph attributes which attempts to completely copy all of the data and references.

This is in contrast to the similar `D=DiGraph(G)` which returns a shallow copy of the data.

See the Python copy module for more information on shallow and deep copies, <https://docs.python.org/3/library/copy.html>.

Warning: If you have subclassed Graph to use dict-like objects in the data structure, those changes do not transfer to the DiGraph created by this method.

#### Examples

```
>>> G = nx.Graph() # or MultiGraph, etc
>>> G.add_edge(0, 1)
>>> H = G.to_directed()
>>> list(H.edges)
[(0, 1), (1, 0)]
```

If already directed, return a (deep) copy

```
>>> G = nx.DiGraph() # or MultiDiGraph, etc
>>> G.add_edge(0, 1)
>>> H = G.to_directed()
>>> list(H.edges)
[(0, 1)]
```

`to_directed_class(self)`

Returns the class to use for empty directed copies.

If you subclass the base classes, use this to designate what directed class to use for ``to_directed()`` copies.

`to_undirected_class(self)`

Returns the class to use for empty undirected copies.

If you subclass the base classes, use this to designate what directed class to use for ``to_directed()`` copies.

`update(self, edges=None, nodes=None)`

Update the graph using nodes/edges/graphs as input.

Like `dict.update`, this method takes a graph as input, adding the graph's nodes and edges to this graph. It can also take two inputs: edges and nodes. Finally it can take either edges or nodes.

To specify only nodes the keyword ``nodes`` must be used.

The collections of edges and nodes are treated similarly to the `add_edges_from/add_nodes_from` methods. When iterated, they should yield 2-tuples `(u, v)` or 3-tuples `(u, v, datadict)`.

#### Parameters

`edges` : Graph object, collection of edges, or None

The first parameter can be a graph or some edges. If it has attributes ``nodes`` and ``edges``, then it is taken to be a Graph-like object and those attributes are used as collections of nodes and edges to be added to the graph.

If the first parameter does not have those attributes, it is treated as a collection of edges and added to the graph.

If the first argument is None, no edges are added.

`nodes` : collection of nodes, or None

The second parameter is treated as a collection of nodes to be added to the graph unless it is None.

If ``edges`` is None and ``nodes`` is None an exception is raised.

If the first parameter is a Graph, then ``nodes`` is ignored.

#### Examples

```
>>> G = nx.path_graph(5)
>>> G.update(nx.complete_graph(range(4, 10)))
>>> from itertools import combinations
>>> edges = (
...     (u, v, {"power": u * v})
...     for u, v in combinations(range(10, 20), 2)
...     if u * v < 225
... )
>>> nodes = [1000] # for singleton, use a container
>>> G.update(edges, nodes)
```

#### Notes

It you want to update the graph using an adjacency structure it is straightforward to obtain the edges/nodes from adjacency. The following examples provide common cases, your adjacency may be slightly different and require tweaks of these examples::

```
>>> # dict-of-set/list/tuple
>>> adj = {1: {2, 3}, 2: {1, 3}, 3: {1, 2}}
>>> e = [(u, v) for u, nbrs in adj.items() for v in nbrs]
>>> G.update(edges=e, nodes=adj)

>>> DG = nx.DiGraph()
>>> # dict-of-dict-of-attribute
```



```

>>> adj = {1: {2: 1.3, 3: 0.7}, 2: {1: 1.4}, 3: {1: 0.7}}
>>> e = [
...     (u, v, {"weight": d})
...     for u, nbrs in adj.items()
...     for v, d in nbrs.items()
... ]
>>> DG.update(edges=e, nodes=adj)

>>> # dict-of-dict-of-dict
>>> adj = {1: {2: {"weight": 1.3}, 3: {"color": 0.7, "weight": 1.2}}}
>>> e = [
...     (u, v, {"weight": d})
...     for u, nbrs in adj.items()
...     for v, d in nbrs.items()
... ]
>>> DG.update(edges=e, nodes=adj)

>>> # predecessor adjacency (dict-of-set)
>>> pred = {1: {2, 3}, 2: {3}, 3: {3}}
>>> e = [(v, u) for u, nbrs in pred.items() for v in nbrs]

>>> # MultiGraph dict-of-dict-of-dict-of-attribute
>>> MDG = nx.MultiDiGraph()
>>> adj = {
...     1: {2: {0: {"weight": 1.3}, 1: {"weight": 1.2}}},
...     3: {2: {0: {"weight": 0.7}}},
... }
>>> e = [
...     (u, v, ekey, d)
...     for u, nbrs in adj.items()
...     for v, keydict in nbrs.items()
...     for ekey, d in keydict.items()
... ]
>>> MDG.update(edges=e)

```

See Also

-----

add\_edges\_from: add multiple edges to a graph  
add\_nodes\_from: add multiple nodes to a graph

---

Data descriptors inherited from `networkx.classes.graph.Graph`:

`__dict__`

dictionary for instance variables (if defined)

`__weakref__`

list of weak references to the object (if defined)

```

name
    String identifier of the graph.

    This graph attribute appears in the attribute dict G.graph
    keyed by the string `"name"`. as well as an attribute (technically
    a property) `G.name`. This is entirely user controlled.

-----
Data and other attributes inherited from networkx.classes.graph.Graph:

adjlist_inner_dict_factory = <class 'dict'>
    dict() -> new empty dictionary
    dict(mapping) -> new dictionary initialized from a mapping object's
        (key, value) pairs
    dict(iterable) -> new dictionary initialized as if via:
        d = {}
        for k, v in iterable:
            d[k] = v
    dict(**kwargs) -> new dictionary initialized with the name=value pairs
        in the keyword argument list.  For example:  dict(one=1, two=2)

adjlist_outer_dict_factory = <class 'dict'>
    dict() -> new empty dictionary
    dict(mapping) -> new dictionary initialized from a mapping object's
        (key, value) pairs
    dict(iterable) -> new dictionary initialized as if via:
        d = {}
        for k, v in iterable:
            d[k] = v
    dict(**kwargs) -> new dictionary initialized with the name=value pairs
        in the keyword argument list.  For example:  dict(one=1, two=2)

edge_attr_dict_factory = <class 'dict'>
    dict() -> new empty dictionary
    dict(mapping) -> new dictionary initialized from a mapping object's
        (key, value) pairs
    dict(iterable) -> new dictionary initialized as if via:
        d = {}
        for k, v in iterable:
            d[k] = v
    dict(**kwargs) -> new dictionary initialized with the name=value pairs
        in the keyword argument list.  For example:  dict(one=1, two=2)

graph_attr_dict_factory = <class 'dict'>

```

```

| dict() -> new empty dictionary
| dict(mapping) -> new dictionary initialized from a mapping object's
|   (key, value) pairs
| dict(iterable) -> new dictionary initialized as if via:
|   d = {}
|   for k, v in iterable:
|       d[k] = v
| dict(**kwargs) -> new dictionary initialized with the name=value pairs
|   in the keyword argument list.  For example:  dict(one=1, two=2)
|
| node_attr_dict_factory = <class 'dict'>
| dict() -> new empty dictionary
| dict(mapping) -> new dictionary initialized from a mapping object's
|   (key, value) pairs
| dict(iterable) -> new dictionary initialized as if via:
|   d = {}
|   for k, v in iterable:
|       d[k] = v
| dict(**kwargs) -> new dictionary initialized with the name=value pairs
|   in the keyword argument list.  For example:  dict(one=1, two=2)
|
| node_dict_factory = <class 'dict'>
| dict() -> new empty dictionary
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|   in the keyword argument list.  For example:  dict(one=1, two=2)

```

```

[301]: cust = CustomerNode(
|     pk='id',
|     prefix='cust',
|     fpath='dat/cust.csv',
|     fmt='csv',
|     compute_layer=ComputeLayerEnum.pandas,
| )
|
| order = OrderNode(
|     pk='id',
|     prefix='order',
|     fpath='dat/orders.csv',

```

```

    fmt='csv',
    compute_layer=ComputeLayerEnum.pandas,
    date_key='ts'
)
gr = GraphReduce(
    name='odsc_first_graph',
    parent_node=cust,
    fmt='csv',
    compute_layer=ComputeLayerEnum.pandas,
    compute_period_val=365,
    compute_period_unit=PeriodUnit.day,
    label_period_val=30,
    label_period_unit=PeriodUnit.day,
    cut_date=datetime.datetime(2023, 6, 12),
    dynamic_propagation=True,
    has_labels=True
)

gr.add_node(cust)
gr.add_node(order)

gr.add_entity_edge(
    parent_node=cust,
    relation_node=order,
    parent_key='id',
    relation_key='customer_id',
    reduce=True
)

```

```

2023-10-25 18:37:39 [warning ] no `date_key` set for <GraphReduceNode:
fpath=dat/cust.csv fmt=csv>

```

```
[302]: gr.do_transformations()
```

```

2023-10-25 18:37:40 [info    ] hydrating graph attributes
2023-10-25 18:37:40 [info    ] hydrating attributes for CustomerNode
2023-10-25 18:37:40 [info    ] hydrating attributes for OrderNode
2023-10-25 18:37:40 [info    ] hydrating graph data
2023-10-25 18:37:40 [info    ] checking for prefix uniqueness
2023-10-25 18:37:40 [info    ] running filters, normalize, and annotations for
<GraphReduceNode: fpath=dat/cust.csv fmt=csv>
2023-10-25 18:37:40 [info    ] running filters, normalize, and annotations for
<GraphReduceNode: fpath=dat/orders.csv fmt=csv>
2023-10-25 18:37:40 [info    ] depth-first traversal through the graph from
source: <GraphReduceNode: fpath=dat/cust.csv fmt=csv>
2023-10-25 18:37:40 [info    ] reducing relation <GraphReduceNode:
fpath=dat/orders.csv fmt=csv>
2023-10-25 18:37:40 [info    ] doing dynamic propagation on node

```

```

<GraphReduceNode: fpath=dat/orders.csv fmt=csv>
2023-10-25 18:37:40 [info      ] joining <GraphReduceNode: fpath=dat/orders.csv
fmt=csv> to <GraphReduceNode: fpath=dat/cust.csv fmt=csv>
2023-10-25 18:37:40 [info      ] computed labels for <GraphReduceNode:
fpath=dat/orders.csv fmt=csv>

```

```
[303]: gr.parent_node.df
```

```

[303]:   cust_id cust_name  cust_name_length  order_customer_id  order_id_count  \
0         1      wes                    3                   1             2
1         2     john                    4                   2             2

   order_amount_sum  order_id_min  order_id_max  order_id_sum  \
0              21.5             1             2             3
1             250.0             3             4             7

   order_customer_id_min  order_customer_id_max  order_customer_id_sum  \
0                      1                      1                      2
1                      2                      2                      4

   order_ts_first  order_amount_min  order_amount_max  order_amount_sum_dupe  \
0    2023-05-12             10.0             11.5             21.5
1    2023-01-01             100.0             150.0             250.0

   order_customer_id_dupe  order_id_had_order
0                      1                   1
1                      2                   1

```

```
[ ]:
```