Chapter 4: Graphs

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DRAFT

Graph theory

See Alberto Montresor theory here: http://disi.unitn.it/~montreso/sp/slides/06-grafi.pdf (http://disi.unitn.it/~montreso/sp/slides/06-grafi.pdf)

See <u>Graphs on the book (https://interactivepython.org/runestone/static/pythonds/Graphs/toctree.html)</u>

In particular, see:

Vocabulary and definitions
 (https://interactivepython.org/runestone/static/pythonds/Graphs/VocabularyandDefinitions.html)

To keep it short, a graph is a set of vertices linked by edges.

Directed graphs

First off, download the Python skeleton (graphs.py) to modify. Solutions are in a separate file (graphs_solutions.py).

In this worksheet we are going to use so called Directed Graphs (DiGraph for brevity), that is graphs that have *directed* edges: each edge can be pictured as an arrow linking source node *a* to target node *b*. With such an arrow, you can go from *a* to *b* but you cannot go from *b* to *a* unless there is another edge in the reverse direction.

- A DiGraph for us can also have no edges or no verteces at all.
- A vertex for us can be anything, a string like 'abc', the number 3, etc
- In our model, edges simply link vertices and have no weights
- The DiGraph is represented as an adjacency list, mapping each vertex to the verteces it is linked to.

QUESTION: DiGraph model is thus good for dense or sparse graphs?

Serious graphs

In this worksheet we follow the Do-It-Yourself methodology and create graph classes from scratch for didactical purposes. Of course, in Python world you have alread nice libraries entirely devoted to graphs like networkx. (https://networkx.github.io/). If you have huge graphs you might consider big data tools like Spark GraphX (http://spark.apache.org/graphx/) which is programmable in Python.

1) Building graphs

IMPORTANT: All the functions until <u>1.8 has_edge() excluded</u> are already provided and you don't need to implement them!

1.1) Building basics

Let's look at the constructor <u>__init__</u> and add_vertex. They are already provided and you don't need to implement it:

```
class DiGraph:
             init (self):
        def
             \overline{\#} The \overline{	ext{cl}}ass just holds the dictionary _edges: as keys it has the vert
   eces, and
             # to each vertex associates a list with the verteces it is linked to.
             self. edges = {}
        def add vertex(self, vertex):
             """ Adds vertex to the DiGraph. A vertex can be any object.
                 If the vertex already exist, does nothing.
             if vertex not in self. edges:
                 self._edges[vertex] = []
You will see that inside it just initializes edges. So the only way to create a DiGraph is with a call like
In [4]:
g = DiGraph()
DiGraph provides an str method to have a nice printout:
In [5]:
print g
DiGraph()
You can add then vertices to the graph like so:
In [6]:
g.add vertex('a')
g.add_vertex('b')
g.add vertex('c')
```

In [7]:

```
print g
```

a: [] b: [] c: []

Adding a vertex twice does nothing:

```
In [8]:
g.add vertex('a')
print g
a: []
b: []
c: []
Once you added the verteces, you can start adding directed edges among them with the method add edge:
   def add edge(self, vertex1, vertex2):
            """ Adds an edge to the graph, from vertex1 to vertex2
                If verteces don't exist, raises an Exception.
                If there is already such an edge, exits silently.
            if not vertex1 in self. edges:
                raise Exception("Couldn't find source vertex:" + str(vertex1))
            if not vertex2 in self. edges:
                raise Exception("Couldn't find target vertex:" + str(vertex2))
            if not vertex2 in self. edges[vertex1]:
                self. edges[vertex1].append(vertex2)
In [9]:
g.add edge('a', 'c')
print g
a: ['c']
b: []
c: []
In [10]:
g.add_edge('a', 'b')
print g
a: ['c', 'b']
b: []
c: []
Adding an edge twice makes no difference:
In [11]:
g.add_edge('a', 'b')
print g
a: ['c', 'b']
b: []
c: []
```

Notice a DiGraph can have self-loops too (also called *caps*):

```
In [12]:
```

```
g.add_edge('b', 'b')
print g

a: ['c', 'b']
b: ['b']
c: []
```

1.2) dig()

dig() is a shortcut to build graphs, it is already provided and you don't need to implement it. **USE IT ONLY WHEN TESTING**, **NOT IN THE DiGraph CLASS CODE** !!!!

With no parameter prints the empty graph:

```
In [13]:
```

```
print dig()
```

DiGraph()

To build more complex graphs, provide pairs source vertex / target verteces list like in the following examples:

```
In [14]:
```

```
print dig('a',['b','c'])
a: ['b', 'c']
b: []
c: []
```

In [15]:

```
b: ['b']
c: ['a']
```

1.3) Equality

Graphs for us are equal irrespectively of the order in which elements in adjacency lists are specified. So for example these two graphs will be considered equal:

```
In [16]:
```

```
dig('a', ['c', 'b']) == dig('a', ['b', 'c'])
```

Out[16]:

True

1.4) Basic querying

There are some provided methods to query the DiGraph: adj, verteces, is empty

1.5) adj

To obtain the edges, you can use the method adj(self, vertex). It is already provided and you don't need to implement it:

```
def adj(self, vertex):
    """ Returns the verteces adjacent to vertex.

    NOTE: verteces are returned in a NEW list.
    Modifying the list will have NO effect on the graph!

if not vertex in self._edges:
    raise Exception("Couldn't find a vertex " + str(vertex))

return self. edges[vertex][:]
```

In [17]:

['b', 'c']

Let's check we actually get back a new list (so modifying the old one won't change the graph):

```
In [18]:
```

```
lst.append('d')
print lst

['b', 'c', 'd']

In [19]:
print g.adj('a')
```

```
['c', 'b']
```

NOTE: This technique of giving back copies is also called *defensive copying*: it prevents users from modifying the internal data structures of a class instance in an uncontrolled manner. For example, if we allowed them direct access to the internal verteces list, they could add duplicate edges, which we don't allow in our model. If instead we only allow users to add edges by calling add_edge, we are sure the constraints for our model will always remain satisfied.

1.6) is_empty()

We can check if a DiGraph is empty. It is already provided and you don't need to implement it:

```
def is_empty(self):
    """ A DiGraph for us is empty if it has no verteces and no edges """
    return len(self. edges) == 0
```

```
In [20]:
print dig().is_empty()

True
In [21]:
print dig('a',[]).is empty()
```

False

1.7) verteces()

To obtain the verteces, you can use the function verteces. (NOTE for Italians: method is called verteces, with two es!!!). It is already provided and you don't need to implement it:

In [22]:

Notice it returns a *set*, as verteces are stored as keys in a dictionary, so they are not supposed to be in any particular order. When you print the whole graph you see them vertically ordered though, for clarity purposes:

In [23]:

```
print g
a: ['c', 'b']
b: ['c']
c: []
```

Verteces in the edges list are instead stored and displayed in the order in which they were inserted.

1.8) has_edge

Enough for talking! Implement this method in DiGraph:

1.9) full graph

```
Implement this function outside the class definition. It is not a method of DiGraph!
```

```
def full graph(verteces):
```

""" Returns a DiGraph which is a full graph with provided verteces list.

raise Exception("TODO IMPLEMENT ME!")

1.10) dag

Implement this function outside the class definition. It is not a method of DiGraph!

```
def dag(verteces):
```

"" Returns a DiGraph which is DAG (Directed Acyclic Graph) made out of provided verteces list

Provided list is intended to be in topological order.
NOTE: a DAG is ACYCLIC, so caps (self-loops) are not allowed !!

raise Exception("TODO IMPLEMENT ME!")

1.11) list_graph

Implement this function outside the class definition. It is not a method of DiGraph!

```
def list graph(n):
```

""" Return a graph of n verteces displaced like a monodirectional list: 1 -> 2 -> 3 -> ... -> n

Each vertex is a number i, 1 <= i <= n and has only one edge connecting it

to the following one in the sequence If n = 0, return the empty graph. if n < 0, raises an Exception.

raise Exception("TODO IMPLEMENT ME!")

1.12) star_graph

Implement this function outside the class definition. It is not a method of DiGraph!

```
def star graph(n):
```

""" Returns graph which is a star with n nodes

First node is the center of the star and it is labeled with 1. This node is linked

to all the others. For example, for n=4 you would have a graph like t his:

If n = 0, the empty graph is returned If n < 0, raises an Exception

raise Exception("TODO IMPLEMENT ME!")

2) Manipulate graphs

You will now implement some methods to manipulate graphs.

```
2.1) remove_vertex
```

```
def remove vertex(self, vertex):
            """ Removes the provided vertex and returns it
                If the vertex is not found, raises an Exception.
            raise Exception("TODO IMPLEMENT ME!")
2.2) reverse
   def reverse(self):
            """ Reverses the direction of all the edges """
            raise Exception("TODO IMPLEMENT ME!")
2.3) has self loops
   def has self loops(self):
            """ Returns True if the graph has any self loop (a.k.a. cap), False o
   therwise """
            for source in self._edges:
                if source in self. edges[source]:
                    return True
            return False
2.4) remove_self_loops
   def remove_self_loops(self):
    """ Removes all of the self-loops edges (a.k.a. caps)
```

NOTE: Removes just the edges, not the verteces!

raise Exception("TODO IMPLEMENT ME!")

3) Query graphs

You can query graphs the "Do it yourself" way with Depth First Search (DFS) or Breadth First Search (BFS).

3.1) Visit and VertexLog

If you noticed, in the skeleton there are two extra classes Visit and VertexLog. Also, in DiGraph the functions dfs and bfs are already provided. The idea here is that both dfs and bfs will traverse the graph and report the intermediate results of the visit inside instances of Visit and VertexLog. At the end of the traversal, they will give back one instance of Visit. Maybe when you do exercises on paper it is convenient to write for example the discovery times inside the nodes of your graphs, but when programming writing intermediate results directly in the verteces of the input graph may cause troubles to the users of your methods. So it is better to store such visit logs in separate data structures: basically, Visit contains a map that associates to each vertex its VertexLog:

```
class Visit:
    """ The visit of a DiGraph visit sequence.

"""

def __init__(self):
    """ Creates a Visit """

self. logs = {}
```

In VertexLog you can put the intermediate info like i.e. discovery_time, or parents of the node if you are interested in building a tree.

```
class VertexLog:
```

```
""" Represents the visit log a single DiGraph vertex
This class is very simple and doesn't even have getters methods.
You can just access fields by using the dot:
    print vertex_log.discovery_time
and set them directly:
    vertex_log.finish_time = 5

If you want, an instances you can set your own fields:
    vertex_log.my_own_field = "whatever"

def __init__(self, vertex):
    self.vertex = vertex
    self.discovery_time = -1
    self.finish_time = -1
    self.parent = None
```

Let's make a simple example:

```
In [24]:
```

```
Visit:
[ {    'discovery_time': 1, 'finish_time': 6, 'parent': None, 'vertex': 'a'},
    {     'discovery_time': 2, 'finish_time': 5, 'parent': 'a', 'vertex': 'b'},
    {     'discovery_time': 3, 'finish_time': 4, 'parent': 'b', 'vertex': 'c'}]
```

Notice we started from 'a', so by default unreachable nodes like d and e were not displayed. Let's try a bfs:

In [25]:

```
print g.bfs('a')
```

Predictably, results are different, you can see it by the parent fields. Note how the finish_time here is always -1 because it is less meaningful to calculate it for a 'bfs'.

You can extract the logs from the Visit object by calling logs():

In [26]:

```
pp(g.dfs('a').logs())
```

By default, they are sorted ascending by discovery time. To see them in descending order, use descendant=False:

In [27]:

```
pp(g.dfs('a').logs(descendant=True))
```

```
[ { 'discovery_time': 3, 'finish_time': 4, 'parent': 'b', 'vertex':
  'c'},
        { 'discovery_time': 2, 'finish_time': 5, 'parent': 'a', 'vertex':
  'b'},
        { 'discovery_time': 1, 'finish_time': 6, 'parent': None, 'vertex':
  'a'}]
```

To see the last timestamp, use last time:

```
In [28]:
```

```
print g.dfs('a').last_time()
```

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3.2) distances()

Try to implement this method of DiGraph:

```
def distances(self, source):
```

Returns a dictionary where the keys are verteces, and each vertex ${\it v}$ is associated

to the *minimal* distance in number of edges required to go from the source

vertex to vertex v. If node is unreachable, the distance will be -1

Source has distance zero from itself Verteces immediately connected to source have distance one.

if source is not a vertex, raises an Exception

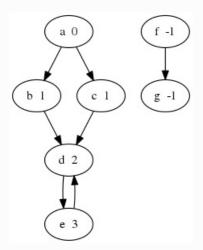
HINT: to implement this, copy and edit either dfs or bfs. Question: w hich one ? $\fill """$

If you look at the following graph, you can see an example of the distances to associate to each vertex, supposing that the source is a. Note that a iself is at distance zero from itself and also that unreachable nodes like f and g will be at distance -1:

In [29]:

```
draw(algolab.show distances(), show='ipynb')
```

Out[29]:



distances ('a') called on this graph would return a map like this:

```
{
    'a':0,
    'b':1,
    'c':1,
    'd':2,
    'e':3,
    'f':-1,
    'g':-1,
}
```

3.2) Play with dfs and bfs

Create small graphs (like linked lists a->b->c, triangles, mini-full graphs, trees - you can also use the functions you defined to create graphs like full_graph, dag, list_graph, star_graph) and try to predict the visit sequence (verteces order, with discovery and finish times) you would have running a dfs or bfs. Then write tests that assert you actually get those sequences when running provided dfs and bfs

3.3) Blow up you computer

Try to call the already implemented function gen_graphs with small numbers for n, like 1, 2, 3, 4 Just with 2 we get back a lot of graphs:

```
def gen_graphs(n):
    """ Returns a list with all the possible 2^(n^2) graphs of size n
    Verteces will be identified with numbers from 1 to n
```

In [30]:

```
print gen graphs(2)
```

```
1: []
2: []
1: []
2: [2]
1: []
2: [1]
1: []
2: [1, 2]
í: [2]
2: []
1: [2]
2: [2]
1: [2]
2: [1]
1: [2]
2: [1, 2]
í: [1]
2: []
1: [1]
2: [2]
1: [1]
2: [1]
í: [1]
2: [1, 2]
1: [1, 2]
2: []
1: [1, 2]
2: [2]
1: [1, 2]
2: [1]
1: [1, 2]
2: [1, 2]
```

]

QUESTION: What happens if you call gen_graphs (10) ? How many graphs do you get back ?

4) Do cool stuff with theory

- find connected components
- determine if a graph is acyclic
- find node distances

In [32]:
<pre>from graphs_solution import * algolab.run(VisitTest)</pre>
•••
Ran 3 tests in 0.003s
0K
T 1001
In [33]:
algolab.run(DiGraphTest)
algolab.run(DiGraphTest)
algolab.run(DiGraphTest)

_ .