NetworKit User Guide

This text is meant as an introduction to using NetworKit for network analysis. We assume that you have read the Readme and successfully built the core library and the Python module.

Startup

Start a Python 3 shell. The standard shell works fine, although IPython may be what you really want for interactive programming and data analysis. These example sessions were created with IPython.

NetworKit can be imported like a standard Python module:

```
or
In[2]: from NetworKit import *
```

In the following, we will assume the latter way for brevity.

Reading Graphs from Disk

In[1]: import NetworKit

There is a large variety of formats for storing graph data in files. For NetworKit, the currently best supported format is the METIS adjacency format. Various example graphs in this format can be found here. We download the file PGP-giantcompo.graph to disk. The needed reader class named METISGraphReader is in the graphio submodule:

```
In[3]: G = graphio.METISGraphReader().read("PGPgiantcompo.graph")
   [BEGIN] reading graph G(n=10680, m=24316) from METIS file: 10% 20.0094%
   30.0187% 40.0281% 50.0375% 60.0468% 70.0562% 80.0655% 90.0749% [DONE]
```

More reader classes can be found in the graphio module. However, there is also a convenient function in the top namespace which tries to guess the input format and select the appropriate reader:

```
In[4]: G = readGraph("PGPgiantcompo.graph")
  [BEGIN] reading graph G(n=10680, m=24316) from METIS file: 10% 20.0094%
  30.0187% 40.0281% 50.0375% 60.0468% 70.0562% 80.0655% 90.0749% [DONE]
```

The Graph Object

The variable G now contains the network represented as an undirected graph.

```
In [10]: G
Out[10]: <_NetworKit.Graph at 0x7f71c9e519c0>
```

For NetworKit, nodes are just integer indexes and edges are tuples of nodes. Galready provides a number of methods:

```
In [11]: G.

G.addEdge G.edges G.hasEdge
G.markAsWeighted G.numberOfEdges G.removeEdge G.toString
G.addNode G.getName G.isMarkedAsWeighted
G.nodes G.numberOfNodes G.removeNode G.weight
```

For example, we could check if nodes 42 and 43 are connected:

```
In [20]: G.hasEdge(42,43)
Out[20]: False
```

To work with the raw list of nodes or edges, call G.nodes() and G.edges() respectively.

Using Algorithms

Connected Components

As a first example of a network analysis kernel, we determine the connected components of the graph by issuing the following commands:

```
In [22]: cc = properties.ConnectedComponents()
In [23]: cc.run(G)
```

The cc object has now performed the calculation and results can be requested through different method calls.

```
In [24]: cc.numberOfComponents()
Out[24]: 1
In [25]: cc.sizeOfComponent(0)
```

Out[25]: 10680

Not surprisingly, PGPgiantcompo has a single connected component containing all nodes.

Community Detection

NetworKit contains several algorithms for parallel community detection, contained in the community submodule. The PLM class implements a parallel heuristic known as the Louvain method. Let's apply it to 'G

```
In [26]: zeta = community.PLM().run(G)
```

Computation will not take long for a relatively small network like PGPgiantcompo. The resulting partition is stored in the variable zeta, which internally stores a mapping from node index to community index. It also provides useful methods:

```
In [27]: zeta.numberOfClusters()
Out[27]: 90
```

We see that 90 communities have been found. Let's examine their sizes:

```
In [30]: zeta.clusterSizes()
Out[30]: [215, 355, 171, 563, 385, 235, 275, 137, 17, 691,
    360, 20, 482, 295, 99, 353, 121, 140, 262, 185, 212, 40, 368,
    82, 85, 78, 356, 16, 364, 772, 271, 31, 64, 77, 313, 333,
    28, 129, 67, 17, 15, 51, 12, 23, 27, 21, 57, 7, 37, 24, 69,
    7, 26, 297, 28, 50, 45, 83, 24, 41, 47, 16, 28, 14, 25, 14,
    8, 11, 68, 33, 26, 37, 6, 16, 19, 17, 76, 29, 27, 13, 8,
    12, 14, 7, 18, 24, 14, 25, 9, 11]
```

It becomes clear that the communities are quite unevenly sized. We can also ask for the community of a specific node:

```
In [14]: zeta.clusterOf(42)
Out[14]: 10
In [15]: zeta.getMembers(10)
```

output omitted

To get an overview of the community detection result, use the function communities.inspectCommunities. This displays a table with information about community sizes as well as the modularity value, a measure of the goodness of the partition and the presence of modular structure in the network.

```
In [5]: community.inspectCommunities(zeta, G)
```

```
# communities 82
min community size 6
max community size 888
avg. community size 130.244
imbalance 6.77863
modularity 0.87673
```

To target very large graphs (in the order of billions of edges), we recommend the PLP parallel label propagation method, which is faster and scales very well with the number of processors. Be aware however that the two methods can arrive at quite different results:

In [7]: community.inspectCommunities(community.PLP().run(G), G)

# communities	979
min community size	2
max community size	411
avg. community size	10.9091
imbalance	37.3636
modularity	0.799206

For convenience the function community.detectCommunities is also provided, which accepts a graph and options like algorithm type etc.

Compatibility with NetworkX

NetworkX is a Python package with a large collection of network analysis methods and graph algorithms. NetworKit provides functions to convert graph objects and thereby connects the two modules.

```
In [6]: import networkx
```

The function nk2nx converts a NetworKit.Graph to a networkx.Graph:

```
In [4]: nxG = nk2nx(G)
In [5]: nxG
Out[5]: <networkx.classes.graph.Graph at 0x7fb1a453f390>
```

Now we can also use some of the numerous NetworkX functions, e.g.

```
In [13]: networkx.degree_assortativity_coefficient(nxG)
Out[13]: 0.23821137170818882
```

The function nx2nk handles the other direction. For example, we can use graph generators not implemented by NetworKit:

```
In [3]: wsG = nx2nk(networkx.generators.watts_strogatz_graph(1000, 2, 0.1))
```

This opens up a wide range of possibilities which are not yet or will never be implemented within NetworKit. Note however that NetworkX is written mostly in pure Python, its data structures are more memory-intensive and its algorithms do not target very large graphs. You are likely to reach limits of your machine for graphs with millions of edges, while NetworKit aims for good performance for three more orders of magnitude.

Overview of Network Properties

Let us now inspect more basic properties of the graph. The functions for this are mostly located in the properties module (though module structure is subject to change). For example, this function gives us a tuple containing the number of nodes and the number of edges:

```
In [7]: properties.nm(G)
Out[7]: (10680, 24316)
```

To see the most important features of a network at a glance, try the function properties.showProperties. It prints a tabular overview (and can also display histograms on the terminal). For example, this is what it tells us about PGPgiantcompo:

In [11]: properties.showProperties(G)

Network Properties				
=======================================				
Basic Properties				
nodes (n)	10680			
edges (m)	24316			
min. degree	1			
max. degree	205			
avg. degree	4.55356			
isolated nodes	0			
self-loops	0			

density	0.000426		
Path Structure			
connected components size of largest comp diameter avg. eccentricity			
Miscellaneous			
degree assortativity	13814		
Community Structure			
avg. local clustering coefficient PLP community detection			0.265945
PLM community detection		communities modularity	
12.1 Community decode	01011	communities modularity	

Some properties have not been calculated because they would take too much time. For some features, it is currently still necessary to convert the graph to NetworkX. The showProperties function is work in progress and will be continually improved.

Settings

Verbosity

The C++ kernel of Networkit provides basic logging functionality and can output text to the terminal. The amount of detail is controlled by the current log level, which is one of TRACE, DEBUG, INFO, WARN, and ERROR. By default, the log level is ERROR. You can set the log level at runtime and do this from the Python shell: