NetworkX

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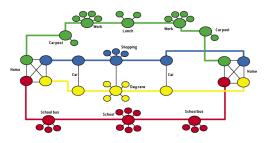
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- 1. Introduction and history
- Getting started
- 3. Working with data
- 4. Writing algorithms
- 5. Live demo
- 6. The future
- 7. Questions

Introduction

Example: social networks and epidemics

Understand epidemic outbreak of diseases through modeling Build social networks from detailed census data Run dynamic models for smallpox, SARS, flu, etc.



Building a social network

Goal: find a good intervention strategy
If Smallpox Strikes Portland...

Social network of one person

by: Chris L. Barrett, Stephen G. Eubank, James P. Smith Scientific American (March 2005)

Goals: Why we started project

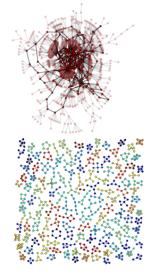
We needed:

- Tool to study the structure and dynamics of social, biological, and infrastructure networks
- Ease-of-use and rapid development in a collaborative, multidisciplinary environment
- Open-source tool base that can easily grow in a multidisciplinary environment with non-expert users and developers
- An easy interface to existing code bases written in C, C++, and FORTRAN
- ► To painlessly slurp in large nonstandard data sets
- No existing API or graph implementation that was suitable
- Inspired by Guido van Rossum's 1998 Python graph representation essay
- First public release in April 2005

Features: NetworkX in one slide

Python language package for exploration and analysis of networks and network algorithms

- Data structures for representing many types of networks, or graphs, (simple graphs, directed graphs, and graphs with parallel edges and self loops)
- Nodes can be any (hashable) Python object
- Edges can contain arbitrary data
- Many network science algorithms (centrality, paths, graph generators)
- Flexibility ideal for representing networks found in many different fields
- Many unit and functional tests
- Online up-to-date documentation
- Open source software (BSD license), Github developer site
- Works with Python 3



Design decisions

NetworkX defines no custom node objects or edge objects

- "node-centric" view of network
- ▶ Nodes: whatever you put in (hashable) with optional node data
- Edges: tuples with optional edge data
- Edge, node data can be anything

NetworkX is all Python

(Other projects use custom compiled code and Python: Boost Graph, igraph, Graphviz)

- Focus on computational network modeling not software tool development
- Move fast to design new algorithms or models

Feature: Simple use, adding nodes

Start Python Import NetworkX using "nx" as a short name

```
>>> import networkx as nx
```

The basic *Graph* class is used to hold the network information. Nodes can be added as follows:

```
>>> G=nx.Graph()
>>> G.add_node(1) # integer
>>> G.add_node('a') # string
>>> print G.nodes()
['a', 1]
```

Feature: nodes can be "anything"

Nodes can be any hashable object such as strings, numbers, files, functions, and more

```
>>> import math
>>> G.add_node(math.cos) # cosine function
>>> fh=open('tmp.txt','w')
>>> G.add_node(fh) # file handle
>>> print G.nodes()
[<built-in function cos>,
<open file 'tmp.txt', mode 'w' at 0x30dc38>]
```

Feature: edges are just pairs of nodes

Edges, or links, between nodes are represented as tuples of nodes. They can be added simply

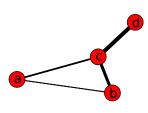
```
>>> G.add_edge(1,'a')
>>> G.add_edge('b',math.cos)
>>> print G.edges()
[('b', <built-in function cos>), ('a', 1)]
```

If the nodes do not already exist they are automatically added to the graph.

Feature: Edge can hold arbitrary data

Any Python object is allowed as edge data (e.g. number, string, image, file, ip address) Edge data assigned and stored in a Python dictionary (default empty).

Use Dijkstra's algorithm to find the shortest path:



```
>>> G=nx.Graph()
>>> G.add_edge('a','b',weight=0.3)
>>> G.add_edge('b','c',weight=0.5)
>>> G.add_edge('a','c',weight=2.0)
>>> G.add_edge('c','d',weight=1.0)
>>> print nx.shortest_path(G,'a','d')
['a', 'c', 'd']
>>> print nx.shortest_path(G,'a','d',weighted=True)
['a', 'b', 'c', 'd']
```

Feature: testing

NetworkX has many tests that can be run by users

```
>>> import networkx
>>> networkx.test(verbositv=2)
Doctest: networkx.utils ... ok
Conversion from digraph to array to digraph. ... ok
Conversion from digraph to matrix to digraph. ... ok
Conversion from graph to array to graph. ... ok
Conversion from graph to matrix to graph, ... ok
Conversion from weighted digraph to array to weighted digraph. ... ok
Conversion from non-square array. ... ok
Conversion from digraph to sparse matrix to digraph. ... ok
Conversion from graph to sparse matrix to graph. ... ok
Conversion from weighted digraph to sparse matrix to weighted digraph. ... ok
Conversion from weighted graph to sparse matrix to weighted graph. ... ok
Conversion from graph to sparse matrix to graph with nodelist. ... ok
Conversion from non-square sparse array. ... ok
Doctest: networkx ... ok
Ran 855 tests in 4.334s
0K
```

Feature: Online, up-to-date documentation





Feature: Python expressivity - a simple algorithm

Python is easy to write and read

Breadth First Search

```
from collections import deque

def breadth_first_search(g, source):
    queue = deque([(None, source)])
    enqueued = set([source])
    while queue:
        parent, n = queue.popleft()
        yield parent, n
        new = set(g[n]) - enqueued
        enqueued |= new
        queue.extend([(n, child) for child in new])
```

Credit: Matteo Dell'Amico

Feature: Compact code - building new generators

Directed Scale-Free Graphs

Béla Bollobás* Christian Borgs† Jennifer Chayes‡ Oliver Riordan§

2 The model

We consider a directed graph which grows by adding single edges at discrete time steps. At each such step a vertex may or may not also be added. For simplicity we allow multiple edges and loops. More precisely, let α , β , γ , δ_{in} and δ_{out} be non-negative real numbers, with $\alpha + \beta + \gamma = 1$. Let G_0 be any fixed initial directed graph, for example a single vertex without edges, and let t_0 be the number of edges of G₀. (Depending on the parameters, we may have to assume $t_0 \ge 1$ for the first few steps of our process to make sense.) We set $G(t_0) = G_0$, so at time t the graph G(t) has exactly t edges, and a random number n(t) of vertices. In what follows, to choose a vertex vof G(t) according to $d_{out} + \delta_{out}$ means to choose v so that $Pr(v = v_i)$ is proportional to $d_{out}(v_i) + \delta_{out}$, i.e., so that $Pr(v = v_i) = (d_{out}(v_i) + \delta_{out})/(t + \delta_{out}n(t))$. To choose v according to $d_{in} + \delta_{in}$ means to choose v so that $Pr(v = v_i) = (d_{in}(v_i) + \delta_{in})/(t + \delta_{in}n(t))$. Here $d_{out}(v_i)$ and $d_{in}(v_i)$ are the out-degree and in-degree of v_i , measured in the graph G(t).

For $t \ge t_0$ we form G(t+1) from G(t) according the the following rules:

- (A) With probability α , add a new vertex v together with an edge from v to an existing vertex w, where w is chosen according to $d_{in} + \delta_{in}$.
- (B) With probability β, add an edge from an existing vertex v to an existing vertex w, where v and w are chosen independently, v according to d_{out} + δ_{out}, and w according to d_{in} + δ_{in}.
- (C) With probability γ, add a new vertex w and an edge from an existing vertex v to w, where v is chosen according to d_{out} + δ_{out}.

Feature: Compact code - building new generators

```
import bisect
 2 import random
 3 from networkx import MultiDiGraph
 4
   def scale free graph(n, alpha=0.41,beta=0.54,delta in=0.2,delta out=0):
       def choose node(G, distribution, delta):
 6
 7
            cumsum = 0.0
 8
           psum = float(sum(distribution.values()))+float(delta)*len(distribution)
 9
            r = random.random()
10
           for i in range(0, len(distribution));
11
               cumsum += (distribution[i]+delta)/psum
12
                if r < cumsum:
13
                    break
14
            return i
15
16
       G = MultiDiGraph()
       G.add edges from([(0,1),(1,2),(2,0)])
       gamma = 1 - alpha - beta
18
19
20
       while len(G)<n:
21
            r = random.random()
22
           if r < alpha:
23
                v = len(G)
24
               w = choose node(G, G.in degree(), delta in)
25
            elif r < alpha+beta:
26
                v = choose node(G, G,out degree(),delta out)
27
               w = choose_node(G, G.in_degree(),delta_in)
28
            else.
29
                v = choose node(G, G.out degree(), delta out)
30
                w = len(G)
31
           G.add edge(v.w)
32
        return G
```

Feature: leveraging libraries

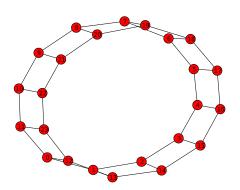
Use well-tested numerical and statistical libraries E.g. convert Graphs to and from NumPy (and SciPy sparse) matrices Example: Find eigenvalue spectrum of the graph Laplacian

```
1 >>> L=nx.laplacian(G)
2 >>> print L # a NumPy matrix
3 [[ 1. -1. 0. 0. 0. 0.]
_{4} [-1. 2. -1. 0. 0. 0.]
[0. -1. 2. -1. 0. 0.]
[0. 0. -1. 2. -1. 0.]
7 \quad [0. \quad 0. \quad 0. \quad -1. \quad 2. \quad -1.]
   [ 0. 0. 0. -1. 1.]]
9 >>> import numpy.linalg
10 >>> print numpy.linalg.eigvals(L)
11 3.7321e+00 3.0000e+00 2.0000e+00
     1.0000e+00 -4.0235e-17 2.6795e-01]
12
```

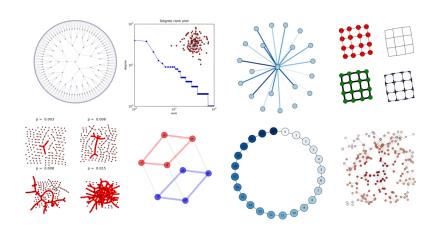
Feature: drawing

Built-in interface to Matplotlib plotting package Node positioning algorithms based on force-directed, spectral, and geometric methods

```
>>> G = nx.circular_ladder_graph(12)
>>> nx.draw(G) # Matplotlib under the hood
```



Drawing with Matplotlib



Getting Started

- Running Python and loading NetworkX
- Creating a Graph, adding nodes and edges
- Finding what is in NetworkX
- Interacting with NetworkX graphs
- Graph generators and operators
- Basic analysis of graphs

Running Python and loading NetworkX

IPython Command line

```
File Edit View Terminal Help
aric@ll:~$ ipython
Python 2.6.4 (r264:75706, Dec 7 2009, 18:43:55)
Type "copyright", "credits" or "license" for more information.
IPvthon 0.10 -- An enhanced Interactive Pvthon.
          -> Introduction and overview of IPvthon's features.
%quickref -> Quick reference.
help -> Python's own help system.
object? -> Details about 'object'. ?object also works, ?? prints more.
In [1]: import networkx as nx
In [2]: help(nx)
In [3]: nx?
In [4]:
```

No GUI http://www.cryptonomicon.com/beginning.html

Command line vs executing file

You can type commands interactively or put them in a file and run them.

```
File Edit View Terminal Help
aric@ll:~$ cat mv program.pv
import networkx as nx
print "imported networkx"
aric@ll:~$ python my program.py
imported networkx
aric@ll:~$ ipvthon
Python 2.6.4 (r264:75706, Dec 7 2009, 18:43:55)
Type "copyright", "credits" or "license" for more information.
IPvthon 0.10 -- An enhanced Interactive Pvthon.
          -> Introduction and overview of IPython's features.
%quickref -> Ouick reference.
help -> Pvthon's own help system.
object? -> Details about 'object', ?object also works, ?? prints more,
In [1]: run my program.py
imported networkx
In [2]: import networkx as nx
In [3]: print "imported networkx"
----> print("imported networkx")
imported networkx
In [4]:
```

Creating a graph

The basic *Graph* object is used to hold the network information. Create an empty graph with no nodes and no edges:

```
import networkx as nx

G = nx.Graph()
```

The graph G can be grown in several ways.

NetworkX includes many graph generator functions and facilities to read and write graphs in many formats.

```
1 # One node at a time
2 >>> G.add node(1) # "method" of G
3
4 # A list of nodes
5 >>> G.add nodes from([2,3])
6
7 # A container of nodes
\approx >>> H = nx.path graph(10)
9 >>> G.add nodes from(H) # G now contains the nodes of H
10
11 # In contrast, you could use the graph H as a node in G.
12 >>> G.add node(H) # G now contains Graph H as a node
```

Nodes can be any hashable object such as strings, numbers, files, functions, and more.

G can also be grown by adding edges.

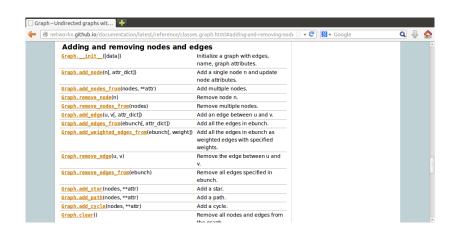
```
1 # Single edge
2 >>> G.add edge(1,2)
_3 >>> e = (2,3)
4 >>> G.add edge(*e) # unpack edge tuple*
5
6 # List of edges
7
8 >>> G.add edges from([(1,2),(1,3)])
10 # Container of edges
11 >>> G.add edges from(H.edges())
```

If the nodes do not already exist they are automatically added to the graph. You can demolish the graph similarly with

```
G.remove_node, G.remove_nodes_from,
G.remove_edge, G.remove_edges_from.
```

- How do I find out the names of the methods like add_edge?
- ▶ How do I see what is in my graph?

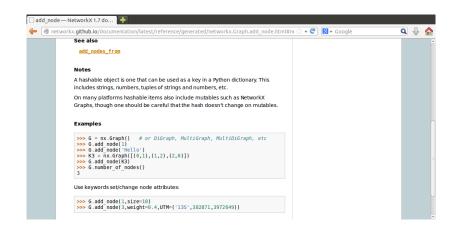
What's in NetworkX?



What's in NetworkX?



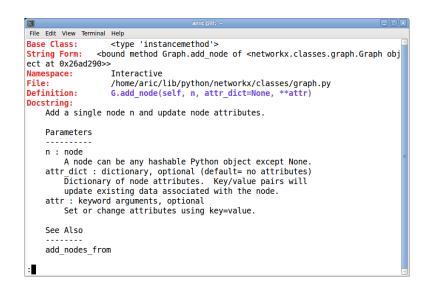
What's in Networkx?



What's in Networkx?

```
aric@ll: ~
                                                                           File Edit View Terminal Help
aric@ll:~$ ipython
Python 2.6.4 (r264:75706, Dec 7 2009, 18:43:55)
Type "copyright", "credits" or "license" for more information.
IPython 0.10 -- An enhanced Interactive Python.
          -> Introduction and overview of IPvthon's features.
%quickref -> Quick reference.
help -> Python's own help system.
object? -> Details about 'object'. ?object also works, ?? prints more.
In [1]: import networkx as nx
In [2]: G=nx.Graph()
In [3]: G.add
                          G.add nodes from
G.add cycle
G.add edge
                         G.add path
G.add edges from
                          G.add star
G.add node
                          G.add weighted edges from
In [3]: G.add node?
In [4]:
```

What's in Networkx?



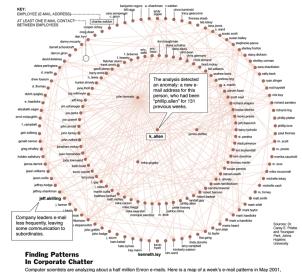
```
File Edit View Terminal Help
Base Class:
                  <type 'instancemethod'>
String Form:
               <bound method Graph.add node of <networkx.classes.graph.Graph obj</pre>
ect at 0x26ad290>>
Namespace:
                  Interactive
File:
                 /home/aric/lib/python/networkx/classes/graph.py
Definition:
                  G.add node(self, n, attr dict=None, **attr)
Docstring:
   Add a single node n and update node attributes.
    Parameters
   n : node
        A node can be any hashable Python object except None.
   attr dict : dictionary, optional (default= no attributes)
        Dictionary of node attributes. Key/value pairs will
        update existing data associated with the node.
    attr: keyword arguments, optional
        Set or change attributes using kev=value.
    See Also
    add nodes from
:
```

Demo

Adding attributes to graphs, nodes, and edges

(Almost) any Python object is allowed as graph, node, and edge data.

- number
- string
- image
- IP address
- email address



when a new name suddenly appeared. Scientists found that this week's pattern differed greatly from others, suggesting different conversations were taking place that might interest investigators. Next step: word analysis of these messages.

Graph attributes

```
1 >>> import networkx as nx
2 # Assign graph attributes when creating a new graph
3
4 >>> G = nx.Graph(day="Friday")
5 >>> G.graph
6 {'day': 'Friday'} # Python dictionary
7
8 # Or you can modify attributes later
10 >>> G.graph['day']='Monday'
11 >>> G. graph
12 {'day': 'Monday'}
```

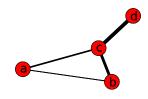
```
1
2 # Add node attributes using add node(), add nodes from() or G.node
3 >>> G.add node(1, time='5pm')
4 >>> G.node[1]['time']
5 '5pm'
6 >>> G.node[1] # Python dictionray
7 {'time': '5pm'}
8
9 >>> G.add nodes from([3], time='2pm') # multiple nodes
10 >>> G.node[1]['room'] = 714 # add new attribute
11
12 >>> G.nodes(data=True)
13 [(1, {'room': 714, 'time': '5pm'}), (3, {'time': '2pm'})]
```

Edge attributes

```
1 # Add edge attributes using add edge(), add edges from(),
2 # subscript notation, or G.edge.
_3 >>> G.add edge(1, 2, weight=4.0)
4 >>> G[1][2]['weight'] = 4.0 # edge already added
5 >>> G.edge[1][2]['weight'] = 4.0 # edge already added
6
7 >>> G[1][2]['weight']
8 4.0
9 >>> G[1][2]
10 {'weight': 4.0}
11
_{12} >>> G. add edges from([(3,4),(4,5)], color='red')
13 >>> G.add edges from([(1,2,{'color':'blue'}), (2,3,{'weight':8})])
14
15 >>> G.edges()
_{16} [(1, 2), (2, 3), (3, 4), (4, 5)]
17 >>> G.edges(data=True)
<sub>18</sub> [(1, 2, {'color': 'blue', 'weight': 4.0}), (2, 3, {'weight': 8}), (3,
```

Weighted graph example

The special attribute 'weight' holds values used by algorithms requiring weighted edges.



Use Dijkstra's algorithm to find the shortest path:

```
1 >>> G=nx.Graph()
2 >>> G.add_edge('a','b',weight=0.3)
3 >>> G.add_edge('b','c',weight=0.5)
4 >>> G.add_edge('a','c',weight=2.0)
5 >>> G.add_edge('a','d',weight=1.0)
6 >>> print nx.shortest_path(G,'a','d')
7 ['a', 'c', 'd']
8 >>> print nx.shortest_path(G,'a','d',weighted=True)
9 ['a', 'b', 'c', 'd']
```

More ways to build graphs: operators and generators

Applying classic graph operations

```
subgraph(G, nbunch) - induce subgraph of G on nodes in nbunch \\ union(G1, G2) - graph union \\ disjoint\_union(G1, G2) - graph union assuming all nodes are different \\ cartesian\_product(G1, G2) - return Cartesian product graph \\ compose(G1,G2) - combine graphs identifying nodes common to both \\ complement(G) - graph complement \\ create\_empty\_copy(G) - return an empty copy of the same graph class \\ convert\_to\_undirected(G) - return an undirected representation of G \\ convert\_to\_directed(G) - return a directed representation of G \\
```

Call a graph generator

```
1 # small graphs
petersen = nx.petersen graph()
3 tutte = nx.tutte graph()
4 maze = nx.sedgewick maze graph()
5 tet = nx.tetrahedral graph()
6
1 # classic graphs
_{8} K 5 = nx.complete graph(5)
_{9} K 3 5 = nx.complete bipartite graph(3,5)
10 barbell = nx.barbell graph(10,10)
lollipop = nx.lollipop graph(10,20)
12
13 # random graphs
_{14} er = nx.erdos renyi graph(100,0.15)
us = nx.watts strogatz graph(30,3,0.1)
_{16} ba = nx.barabasi albert graph(100,5)
red = nx.random lobster(100,0.9,0.9)
```

Basic analysis of graphs

```
1 >>> G=nx.Graph()
_{2} >>> G.add edges from([(1,2),(1,3)])
3 >>> G.add node("spam")
4
5 # Structure of G can be analyzed using various
6 # graph—theoretic functions
7 >>> nx.connected components(G)
8 [[1, 2, 3], ['spam']]
10 # Functions that return node properties return
11 # dictionaries keyed by node label.
12 >>> nx.degree(G)
13 {1: 2, 2: 1, 3: 1, 'spam': 0}
14
15 >>> sorted(nx.degree(G).values())
16 [0, 1, 1, 2]
17
18 >>> nx.clustering(G)
19 {1: 0.0, 2: 0.0, 3: 0.0, 'spam': 0.0}
```

Working with data

Read a graph stored in a file using common graph formats.

edge lists

adjacency lists

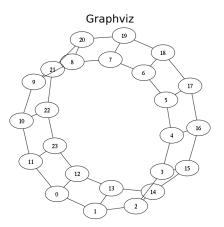
GML

GraphML

Pajek

LEDA

Drawing with other programs



Output to: dot, GML, LEDA, edge list, adjacency list, YAML, sparsegraph6, GraphML

Writing Algorithms

- Examples of some simple algorithms
- Writing a new algorithm

Feature: Python expressivity - a simple algorithm

Python is easy to write and read

Breadth First Search 1 from collections import deque 2 3 def breadth first search(g, source): queue = deque([(None, source)]) enqueued = set([source]) 5 while queue: 6 parent, n = queue.popleft() 7 yield parent, n 8 new = set(g[n]) - enqueued9 enqueued |= new 10 queue.extend([(n, child) for child in new]) 11

Credit: Matteo Dell'Amico

Feature: Python expressivity - a simple algorithm

Python is easy to write and read

Erdős-Rényi Random graph

```
from itertools import combinations
from random import random

def gnp_random_graph(n, p):
    G=empty_graph(n) # NetworkX
    edges=combinations(range(n),2)

for e in edges:
    if random() < p:
        G.add_edge(*e) # NetworkX

return G</pre>
```

Feature: Python expressivity - a simple algorithm

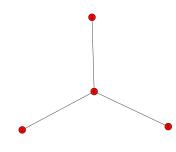
Erdős-Rényi Random graph

```
1 def fast gnp random graph(n, p):
      G = empty graph(n)
2
      G.name="fast_gnp_random_graph(%s,%s)"%(n,p)
3
4
      if not seed is None:
5
           random.seed(seed)
6
7
       if p \le 0 or p \ge 1:
8
           return nx.gnp random graph(n,p)
9
10
      v = 1 # Nodes in graph are from 0,n-1
11
      w = -1
12
      lp = math.log(1.0 - p)
13
      while v < n:
14
           Ir = math.log(1.0 - random.random())
15
          w = w + 1 + int(Ir/Ip)
16
          while w >= v and v < n:
17
               w = w - v
18
               v = v + 1
19
           if v < n:
20
               G.add edge(v, w)
21
      return G
22
```

Degree centrality

For a graph with *n* nodes

$$C_D(v) = \frac{deg(v)}{n-1}$$



```
1 >>> G=nx.star graph(3)
2 >>> print G.edges()
_{3} [(0, 1), (0, 2), (0, 3)]
4 >>> print G.degree(0)
5 3
6 >>> print len(G) # # of nodes
7 4
8 >>> print G.degree(0)/3
a 0
10 >>> print G.degree(0)/3.0
11 1
12 >>> for v in G:
        print v, G.degree(v)/3.0
14 0 1.0
15 1 0.3333333333333
16 2 0.3333333333333
17 3 0.3333333333333
```

```
1 import networkx as nx
2
₃ def_degree centrality(G):
4
5 ____n_=_len(G)_-_1.0_#_forces_floating_point_for_n
6 ____for_v_in_G:
7 print_v,G.degree(v)/n
8
9 ____return
10
11 G_=_nx.star_graph(3)
12 degree_centrality(G)
13 #_0_1.0
14 # 1 0.3333333333333
15 #, 2, 0.3333333333333
16 # 3 0.3333333333333
```

Degree centrality 2

```
1 import networkx as nx
2
3 def degree_centrality(G):
4
      centrality = {} # empty dictionary
5
      n = len(G) - 1.
6
    for v in G:
7
          centrality[v] = G.degree(v)/n
      return centrality
9
10
_{11} G = nx.star graph(3)
12 dc = degree centrality(G)
13 for v in dc:
      print v,dc[v]
15 # 0 1.0
16 # 1 0.3333333333333
17 # 2 0.3333333333333
18 # 3 0.3333333333333
19 print dc
20 # {0: 1.0, 1: 0.33333, 2: 0.33333, 3: 0.33333}
```

```
1 def degree centrality(G):
      centrality = {} # empty dictionary
2
      n = len(G)-1.0 \# forces floating point for n
3
    for v in G:
         centrality[v] = G.degree(v)/n
5
      return centrality
6
7
8 if name ==' main ':
      import networkx as nx
9
      G = nx.star graph(3)
10
  for v,c in degree centrality(G).items():
11
          print v,c
12
13 # 0 1.0
14 # 1 0.3333333333333
15 # 2 0.3333333333333
16 # 3 0.3333333333333
```

```
1 def degree centrality(G):
      """Compute degree centrality for nodes.
2
3
      The degree centrality for a node is the fraction of all other
4
      nodes it is connected to.
5
6
      >>> import networkx as nx
7
      >>> G = nx.star graph(3)
8
      >>> print degree centrality(G)[0]
q
      1.0
10
11
      centrality = {} # empty dictionary
12
      n = len(G)-1.0 \# forces floating point for n
13
      for v in G:
14
          centrality[v] = G.degree(v)/n
15
      return centrality
16
```

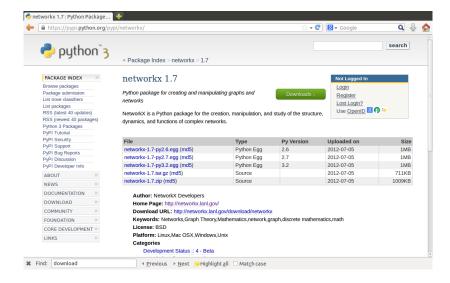
Degree centrality in NetworkX

```
def degree centrality(G):
 2
 3
 4
       The degree centrality for a node v is the fraction of nodes it
 5
 6
 7
 8
 9
       G : graph
10
         A Networkx graph
11
12
       Returns
13
14
15
          Dictionary of nodes with degree centrality as the value.
16
17
       See Also
18
19
20
21
       Notes
22
23
       The degree centrality values are normalized by dividing by the maximum
24
       possible degree in a simple graph n-1 where n is the number of nodes in G.
25
26
       For multigraphs or graphs with self loops the maximum degree might
27
       be higher than n-1 and values of degree centrality greater than 1
       are possible.
28
29
30
        s = 1.0/(len(G)-1.0)
31
       return dict((n,d*s) for n,d in G,degree iter())
```

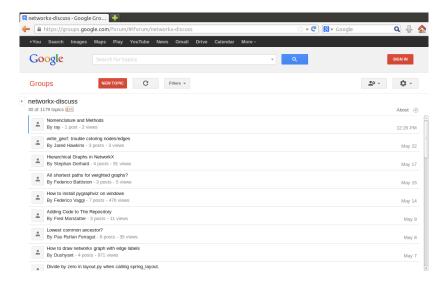
Demo

Future

approx 100K downloads of networkx-1.7



networkx-discuss Google group 1130 members

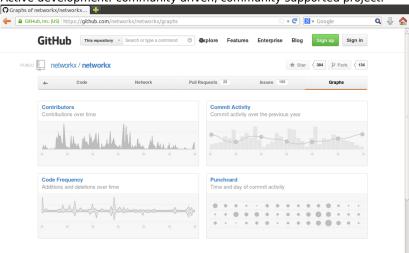


Developer site

http://networkx.github.io/

networkx-1.8 release soon

Active development: community driven, community supported project.



We hope you will contribute (after class is fine).

New features

- More algorithms (contribute)
- Better interaction with graph drawing tools
- Inline integration with IPython notebook
- Community finding algorithms

Questions?