



# NetworkX

Analyzing and Manipulating Networks in Python

PyData Hamburg 24/10/2018 - Christoph Martin, [me@devmartin.de](mailto:me@devmartin.de)

**Network:** a pattern of interconnections among a set of things

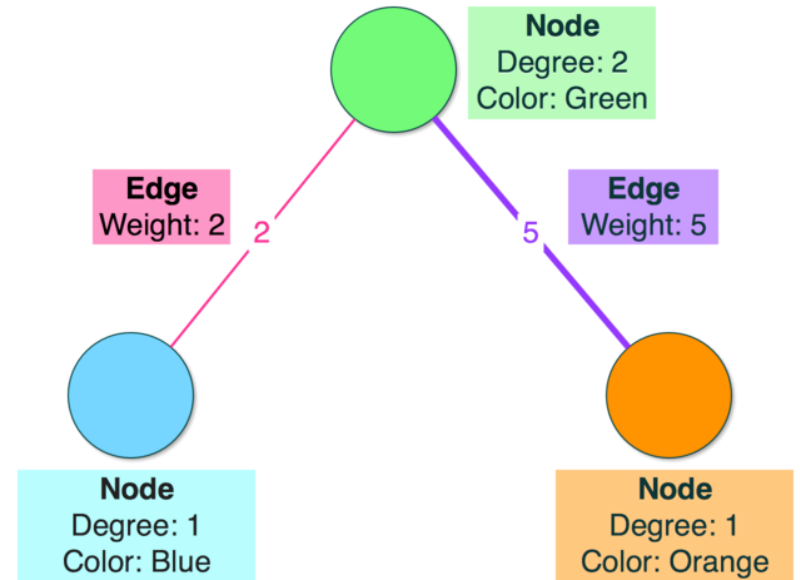
**Node / Vertex:** The entity of analysis which has a relationship. Node is used in the network context, vertex is used in the graph theory context, but both terms are often used interchangeably.

**Link / Edge / Relationship:** The connections between the nodes. Link is used in the network context, edge is used in the graph theory context, and all words are used interchangeably with *relationship*.

**Attributes:** Both nodes and edges can store attributes, which contain additional data about that object.

**Weight:** A common *attribute* of edges, used to indicate *strength* or *value* of a relationship.

**Degree:** Number of edges a node has.



A network is [unweighted/weighted] and [undirected/directed] and may contains parallel edges and / or self loops.



Network	Nodes	Edges	Directionality
Internet	Routers	Internet connections	Undirected
WWW	Webpages	Links	Directed
Power Grid	Power plants, transformers	Cables	Undirected
Mobile-Phone Calls	Subscribers	Calls	Directed
Email	Email addresses	Emails	Directed
Science Collaboration	Scientists	Co-authorships	Undirected
Actor Network	Actors	Co-acting	Undirected
Citation Network	Papers	Citations	Directed
E. Coli Metabolism	Metabolites	Chemical reactions	Directed
Protein Interactions	Proteins	Binding interactions	Undirected

If your data is not 'tabular' but contains relationships / similarities between the observations, then you might want to build a network based on that data..

# Working with networks in Python

Today: NetworkX

NetworkX is a Python package for the creation, manipulation, and study of the structure, dynamics, and functions of complex networks. (<https://networkx.github.io/>)

- Data structures for graphs, digraphs, and multigraphs
- Many standard graph algorithms, network structure and analysis measures
- Nice API
- Open-Source

## Basics

Documentation (good): <https://networkx.github.io/documentation/stable/>  
(<https://networkx.github.io/documentation/stable/>).

```
# installation  
$ pip install networkx
```

Current stable version (2.2) of NetworkX supports Python 2.7, 3.5, 3.6, or 3.7.

Some features may require other packages (e.g.: matplotlib for plotting, numpy/scipy for faster computations)

```
In [1]: import numpy as np
import pandas as pd

import networkx as nx
import matplotlib.pyplot as plt
plt.style.use('seaborn')
%matplotlib inline
```

```
In [4]: g = nx.Graph()
g.add_node('alice'), g.add_node('bob')
g.add_edge('alice', 'bob')
g.add_node(sum), g.add_node(nx.Graph) #hashable objects can be nodes
g.add_edge(sum, nx)
draw_network(g)
```

`α.classes.graph.Graph'>`



`x' from '/home/crs/.virtualenvs/nxpydata/lib/python3.6/site-packages/networkx/___ii`

`<built-in function sum>`

```
In [5]: g.add_nodes_from(['clair', 'dave'])
g.add_edges_from([('bob', 'clair'), ('alice', 'clair'), ('clair', 'dave')])

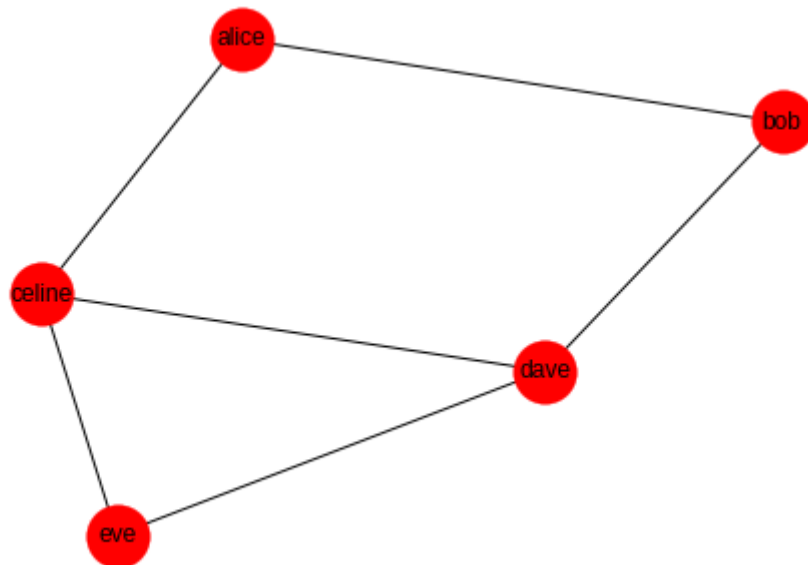
draw_network(g)
```



```
nx.classes.graph.Graph>
nx' from '/home/crs/.virtualenvs/nxpydata/lib/python3.6/site-packages/networkx/_
<built-in function sum>
```



```
In [6]: edges = [['alice', 'bob'], ['alice', 'celine'], ['celine', 'dave'],  
                ['bob', 'dave'], ['dave', 'eve'], ['celine', 'eve']]  
g = nx.from_edgelist(edges)  
draw_network(g)
```



Graph classes in networkx: `nx.Graph`, `nx.DiGraph`, `nx.MultiGraph`, and `nx.MultiDiGraph`

```
In [7]: g = nx.from_edgelist(edges, create_using=nx.DiGraph)
draw_network(g)
print(nx.info(g))
```

Name:

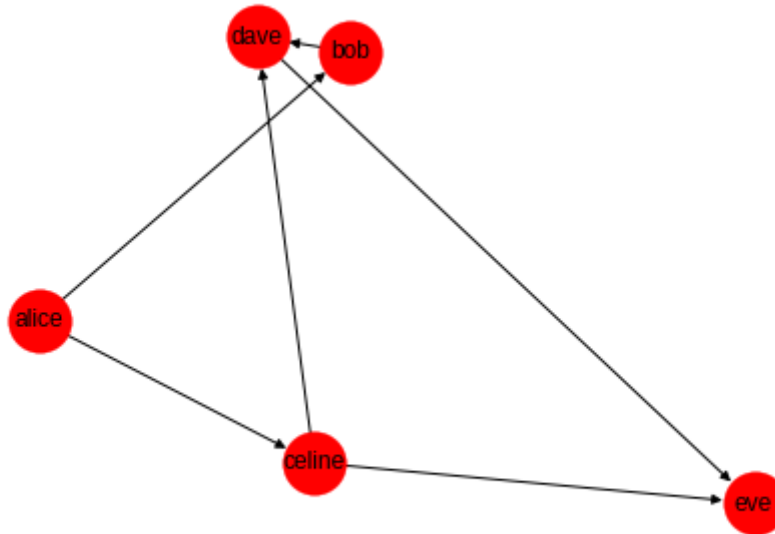
Type: DiGraph

Number of nodes: 5

Number of edges: 6

Average in degree: 1.2000

Average out degree: 1.2000



```
In [8]: g.is_directed()
```

Out[8]: True

NetworkX can read many popular graph formats, e.g.: gml, graphml, edgelist, adjlist, ..

```
In [9]: !zcat house.edgelist.gz | head -n3
```

```
alice bob  
alice celine  
bob dave
```

```
In [10]: g = nx.read_edgelist('house.edgelist.gz')  
print(nx.info(g))
```

```
Name:  
Type: Graph  
Number of nodes: 5  
Number of edges: 6  
Average degree: 2.4000
```

We can also use pandas DataFrames to create networks..

```
In [11]: dfrm = pd.read_csv('house.csv')
dfrm.tail(2)
```

```
Out[11]:
```

	source	target	intensity
4	celine	eve	3
5	dave	eve	1

```
In [12]: g = nx.from_pandas_edgelist(dfrm, source='source', target='target',
                                     edge_attr='intensity')
g.graph['name'] = 'house'
```

```
In [13]: print(nx.info(g))
```

```
Name: house
Type: Graph
Number of nodes: 5
Number of edges: 6
Average degree:  2.4000
```

```
In [14]: g.edges(data=True)
```

```
Out[14]: EdgeDataView([('alice', 'bob', {'intensity': 1}), ('alice', 'celine', {'intensity': 1}), ('bob', 'dave', {'intensity': 2}), ('celine', 'dave', {'intensity': 2}), ('celine', 'eve', {'intensity': 3}), ('dave', 'eve', {'intensity': 1})])
```

```
In [15]: g.edges[('celine', 'dave')]
```

```
Out[15]: {'intensity': 2}
```

```
In [16]: g.nodes(data=True)
```

```
Out[16]: NodeDataView({'alice': {}, 'bob': {}, 'celine': {}, 'dave': {}, 'eve': {}})
```

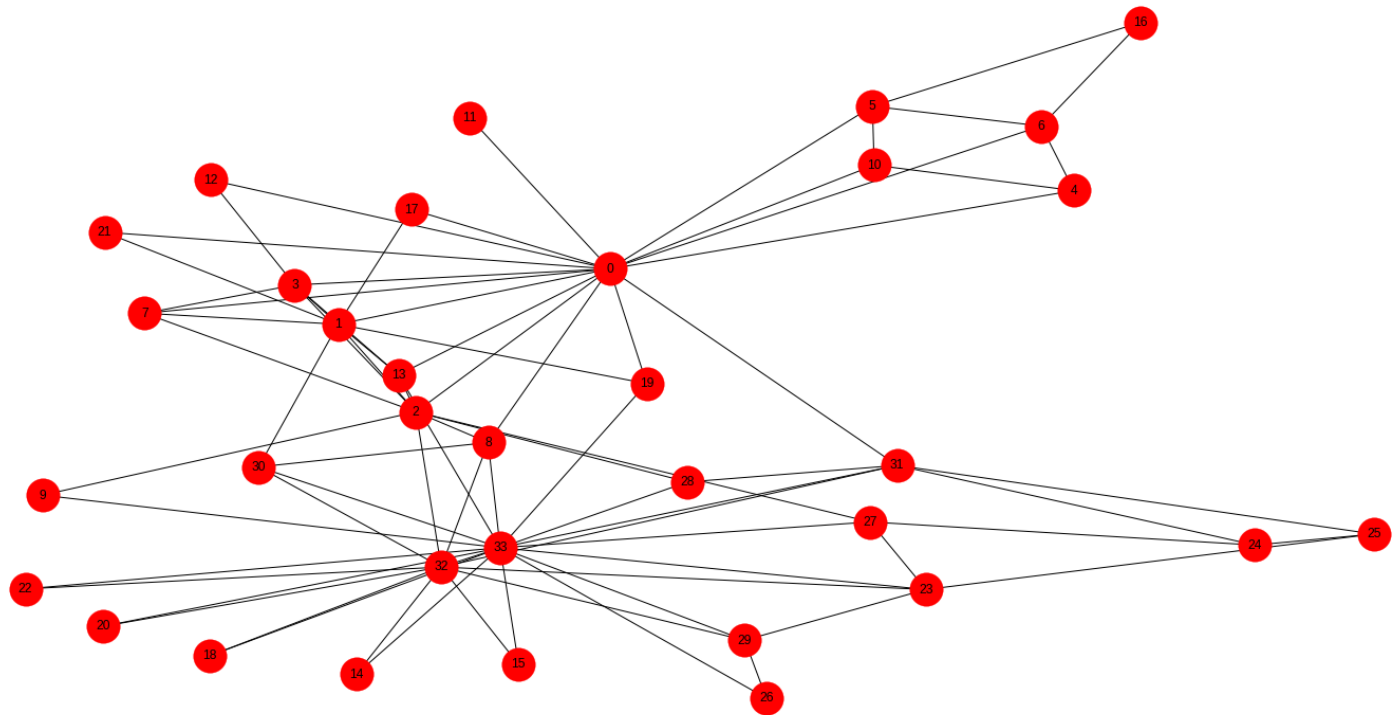


## Some network analysis - Zachary's karate club

Zachary's karate club is a well-known social network. The data was collected from the members of a university karate club by Wayne Zachary in 1977. Each node represents a member of the club, and each edge represents a tie between two members of the club. The network is undirected. An often discussed problem using this dataset is to find the two groups of people into which the karate club split after an argument between two teachers.

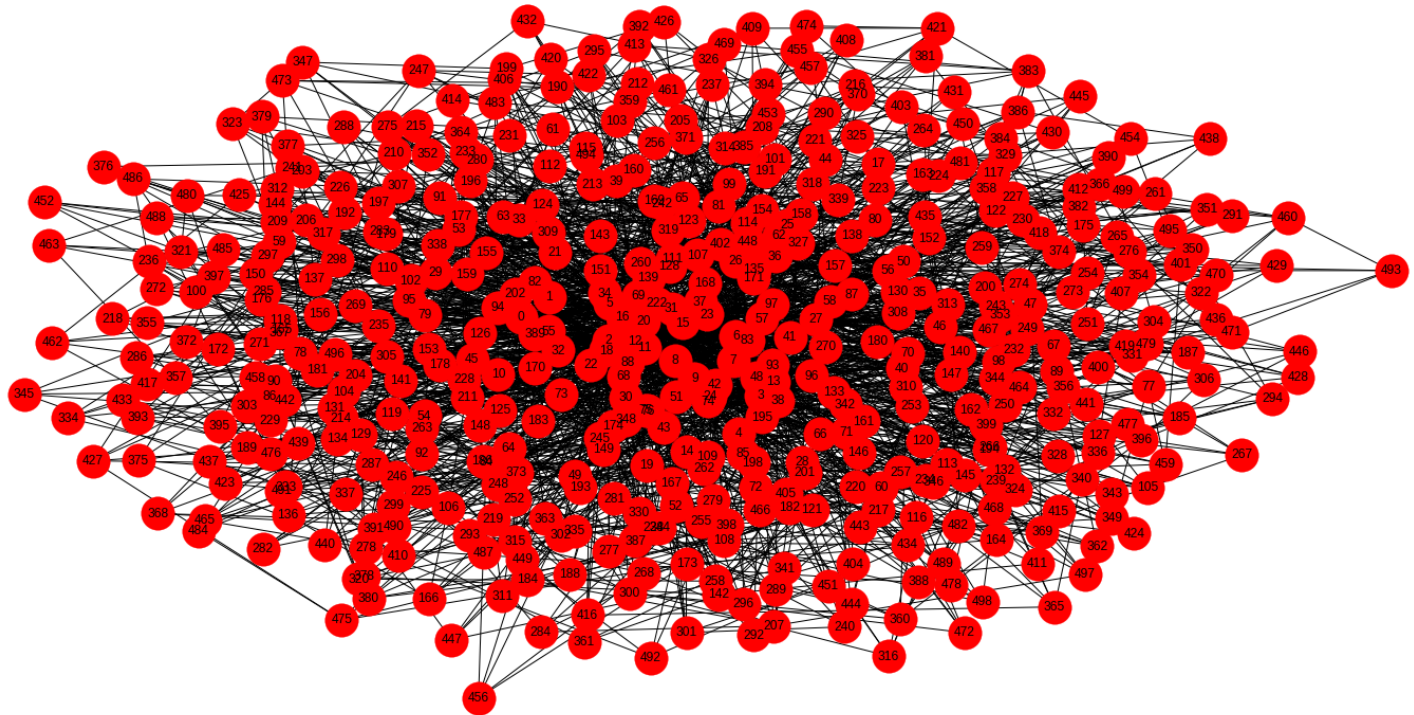
```
In [17]: karate = nx.karate_club_graph()
```

```
In [18]: plt.figure(figsize=(20,10))  
draw_network(karate)
```

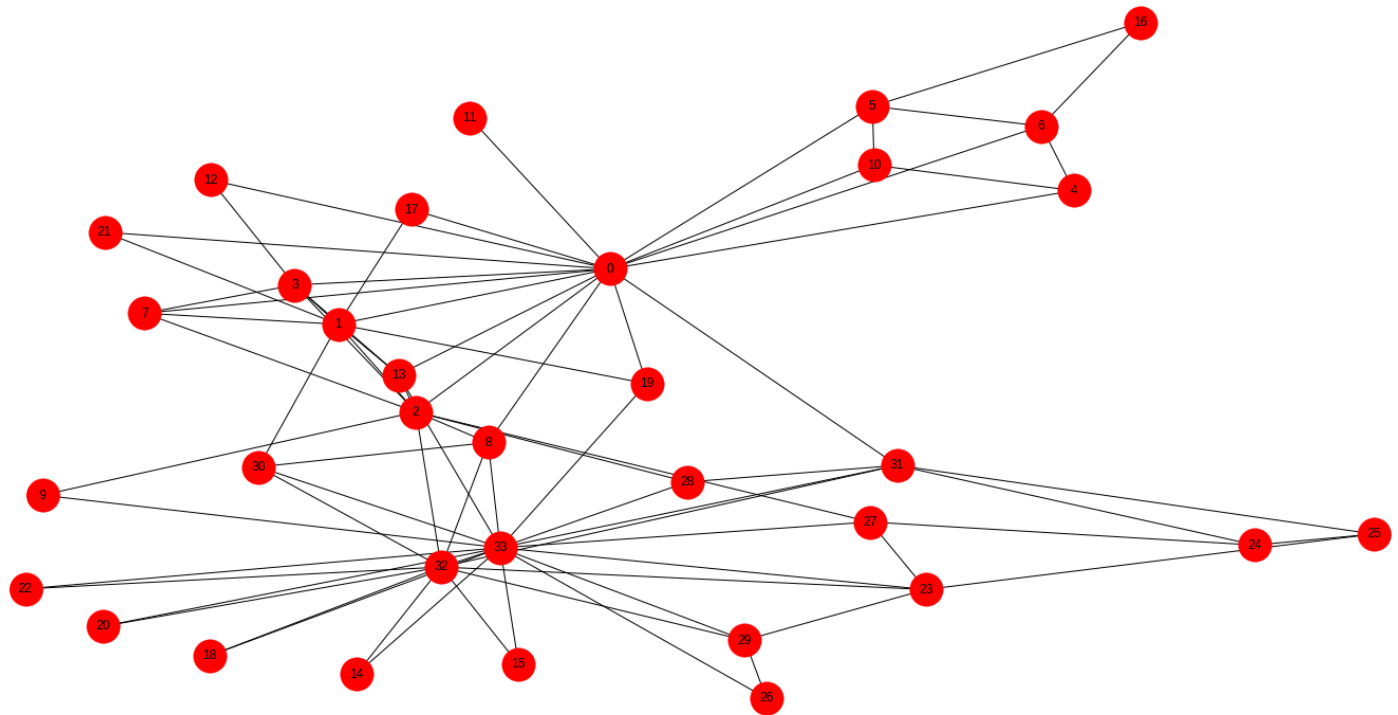




```
In [19]: plt.figure(figsize=(20,10))  
draw_network(nx.barabasi_albert_graph(500,5))
```



```
In [20]: plt.figure(figsize=(20,10))  
draw_network(karate)
```



```
In [21]: print(nx.info(karate))
```

```
Name: Zachary's Karate Club  
Type: Graph  
Number of nodes: 34  
Number of edges: 78  
Average degree: 4.5882
```

```
In [22]: karate.number_of_nodes(), karate.number_of_edges()
```

```
Out[22]: (34, 78)
```

The density of a graph is defined as a ratio of the number of edges to the number of possible edges in a graph.

```
In [23]: nx.density(karate), 78 / (34*33/2)
```

```
Out[23]: (0.13903743315508021, 0.13903743315508021)
```

The diameter of a graph is the length of the longest shortest path in a graph.

```
In [24]: nx.diameter(karate)
```

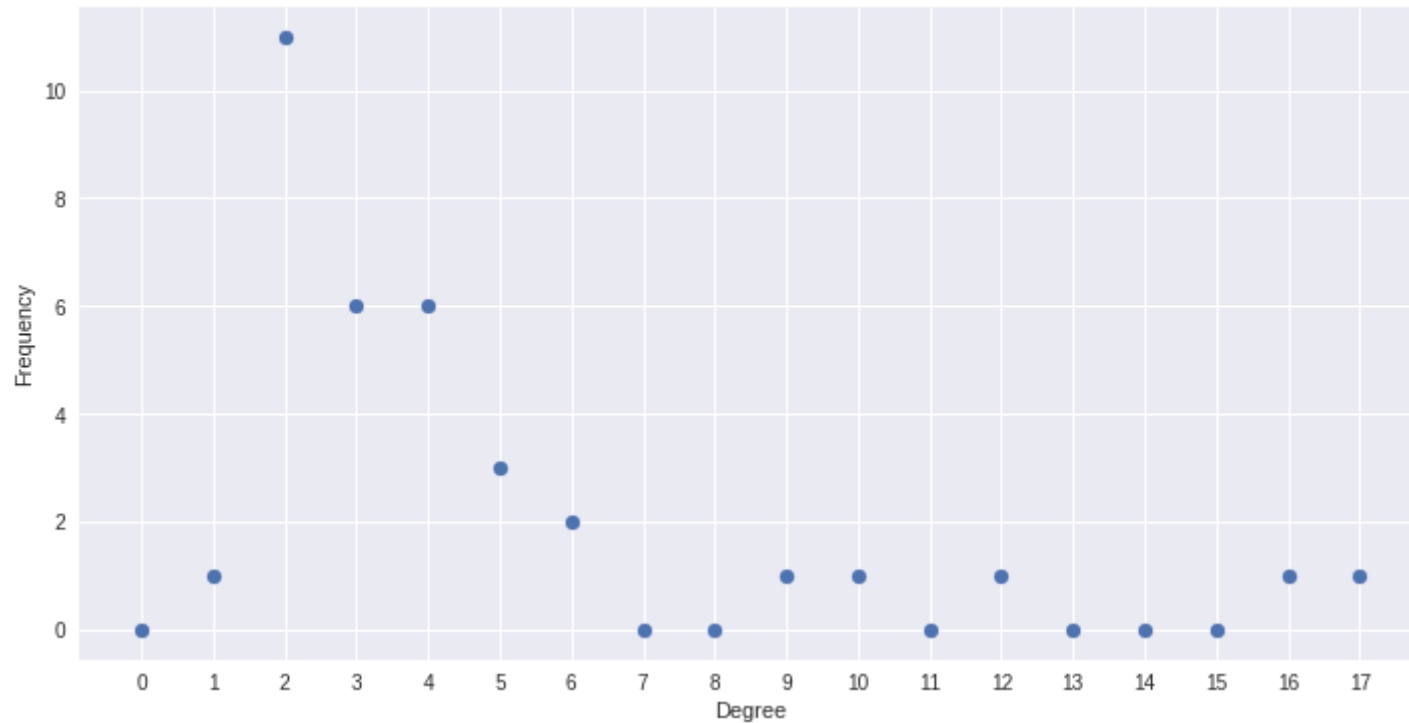
```
Out[24]: 5
```

The degree of a node in a network is the number of connections it has to other nodes.

```
In [25]: nx.degree(karate, nbunch=[32, 33, 1])
```

```
Out[25]: DegreeView({32: 12, 33: 17, 1: 9})
```

```
In [26]: deg_hist = nx.degree_histogram(karate)
plt.figure(figsize=(12,6))
plt.scatter(np.arange(len(deg_hist)), deg_hist)
plt.xlabel('Degree'), plt.ylabel('Frequency')
_ = plt.xticks(np.arange(len(deg_hist)))
```



Centrality measures map a real number to all nodes in a graph. They don't use attributes, they only rely on the graph structure. The centrality values are usually used to measure the 'importance' of individual nodes in a graph.

## Examples

```
nx.degree_centrality,  
nx.closeness_centrality,  
nx.betweenness_centrality,  
nx.pagerank
```

.. and many other available in the `nx.algorithms.centrality` module.

Usage: First argument is the graph. Return value is a dict: key are the nodes, values are the centrality values for the corresponding node.

# Visualizing Networks

```
In [ ]: nx.draw(g, with_labels=True)
```

Networkx offers a variety of layout algorithms, e.g.:

```
nx.layout.circular_layout  
nx.layout.random_layout  
nx.layout.fruchterman_reingold_layout # often useful
```

```
In [ ]: pos = nx.fruchterman_reingold_layout(g, seed=42)
```

```
In [ ]: nx.draw(g, with_labels=True, pos=pos) # fix node position
```



```
In [ ]: edgelist_high = [x for x in g.edges(data=True) if x[-1]['intensity'] > 1]
        nx.draw(g, edgelist=edgelist_high, pos=pos, with_labels=True)
```

```
In [ ]: el = nx.get_edge_attributes(g, 'intensity')
```

```
In [ ]: h_pos = nx.fruchterman_reingold_layout(g, seed=42)
nx.draw(g, pos=h_pos, with_labels=True, width=3, style='dashed', node_size=1000,
        node_shape='s')
_ = nx.draw_networkx_edge_labels(g, edge_labels=el, pos=h_pos, font_size=20)
```

```
In [ ]: club_colors = ['red' if v['club'] == "Mr. Hi" else 'lightblue' for k,v in karate
               .nodes(data=True)]
```

```
In [ ]: pos_kt = nx.layout.fruchterman_reingold_layout(karate, seed=1984)
pos_kt[33] -= np.array([0, 0.05])
plt.figure(figsize=(12,6))
dg = karate.degree()
dg = np.array([dg[n] for n in karate.nodes])
nx.draw(karate, node_color=club_colors, pos=pos_kt, with_labels=True, node_size=
dg*100+100)
```

## Alternatives (usually faster)

- graph-tool <https://graph-tool.skewed.de/> (<https://graph-tool.skewed.de/>).
- igraph <http://igraph.org/python/> (<http://igraph.org/python/>).
- NetworkKit <https://networkit.itk.itk.edu/> (<https://networkit.itk.itk.edu/>).

## More information

- Network Science by Albert-László Barabási (<http://networksciencebook.com> (<http://networksciencebook.com>))
- Networks: An Introduction by Mark Newman
- NetworkX Documentation (<https://networkx.github.io/documentation/stable> (<https://networkx.github.io/documentation/stable>))

## Slides

- <https://github.com/crsqg/NetworkXpydata> (<https://github.com/crsqg/NetworkXpydata>).

## Contact

# heading 1 ## heading 2 ### heading 3

<https://github.com/rtidatascience/connected-nx-tutorial>  
[\(https://github.com/rtidatascience/connected-nx-tutorial\)](https://github.com/rtidatascience/connected-nx-tutorial).

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