

Network Analysis with Python and NetworkX Cheat Sheet by RJ Murray (murenei) via cheatography.com/58736/cs/15946/

## Basic graph manipulation

import networkx as nx
G=nx.G raph()
G=nx.M ult iGr aph()
G.add_ edg es_fro $m([(0,1),(0,2),(1,3),(2$,
4)]
nx.dra w_n etw orkx(G)
G.add_ nod e(' A', rol e=' man ager')
G.add_ edg e(' A', 'B' ,re lation = 'friend')

- edge

| Create a graph | nodes of bipartite graph |
| :---: | :---: |
| allowing <br> parallel bipart ite.pr oje cte d_g raph (B, X) edges | Bipartite projected |
| Create <br> graph <br> from <br> edges | graph - <br> nodes <br> with <br> bipartite <br> friends in |
| Draw the | common |
| P=bipa rti te.w ei ght ed_ pro jec ted _gr <br> Add a $\operatorname{aph}(B, X)$ <br> node | projected <br> graph <br> with |
| Add an edge | weights <br> (number |
| Set attribute of a node | of friends in common) |

G.node ['A'], G.edge [('A', 'B')]



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Page 1 of 5 .

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| Network Connectivity: Connected Components |  |
| :---: | :---: |
| nx.is_ con nec ted (G) | Is there a path between every pair of nodes? |
| nx.num ber _co nne cte d_c omp on e nts(G) | \# separate components |
| nx. nod e_c onn ect ed_com pon en $t(G, N)$ | Which connected component does $N$ belong to? |
| nx.is_ str ong ly_ con nec ted(G) | Is the network connected directionally? |
| nx.is_ wea kly _co nne cted(G) | Is the directed network connected if assumed undirected? |


| Common Graphs |
| :--- |
| G=nx.k ara te_ clu b_g raph Karate club graph (social <br> () <br> network)  <br> G=nx.p ath _gr aph (n) Path graph with n nodes <br> G=rand om_ reg ula r_g rap - Random d-regular graph on <br> $h(d, n)$ n-nodes |

See NetworkX Graph Generators reference for more.
Also see "An Atlas of Graphs" by Read and Wilson (1998).

Influence Measures and Network Centralization

| Influence Measures and Network Centralization (cont) |  |
| :---: | :---: |
| nx.edg e_b etw een nes s_c ent ral ity(G) | E |
| nx.edg e_b etw een nes s_c ent ral ity _su b G , \{s ubset\}) | $\text { se } \mathrm{t}$ |
| Normalization: Divide by number of pairs of nodes. |  |
| PageRank and Hubs \& Authorities Algorithms |  |
| nx.pag era nk(G, alpha=0.8) | Scaled <br> PageRank <br> of $G$ with <br> dampeninc <br> parameter |
| h, a=nx.hi ts (G) | HITS <br> algorithm - <br> outputs 2 <br> dictio- <br> naries <br> (hubs, <br> authorities', |
| h,a=nx.hi ts( G,m ax_ ite $r=1$ 0, n orm ali zed =True) | Constrained HITS and normalized by sum at each stage |

Centrality measures make different assumptions about what it means to be a "central" node. Thus, they produce different rankings.

| Network Evolution-Real-world Applications |  |
| :--- | :--- |
| G.degree (), G.in_d egree (), G.out_deg ree () | Distributior <br> of node |
| Preferential Attachment Model | degrees <br> Results in <br> power law |
|  | -> many <br> nodes with |
|  | low <br> degrees; |
|  | few with |
| high |  |
| degrees |  |


| dc=nx.d eg ree _ce ntr ali ty (G) | ```Degree G=bara bas i_a lbe rt_gra ph(n,m) centrality for network``` | Preferentic <br> Attachmen <br> Model with |
| :---: | :---: | :---: |
| dc [node] | Degree centrality for a node | $n$ nodes and each new node |
| nx.in_ deg ree _ce ntr ali ty (G), nx.out _de g re e_c ent ral ity(G) | DC for directed networks | attaching <br> to $m$ <br> existing <br> nodes |
| cc=nx.c lo sen ess _ce ntr ali ty ( G, n orm al | Closeness |  |
| i zed =True) | centrality Small World model (norma- <br> lised) for <br> the <br> network | High <br> average <br> degree <br> (global <br> clustering) |
| cc [node] | Closeness <br> centrality <br> for an <br> individual | and low <br> average <br> shortest <br> path |
|  | $\mathrm{G}=$ watt s_s tro gat z_g rap $\mathrm{h}(\mathrm{n}, \mathrm{k}, \mathrm{p})$ | Small |
| bC=nx.b et wee nne ss_ cen tra lity (G) |  | World <br> network of <br> $n$ nodes, connected to its $k$ nearest neighbours, with chance $p \mathrm{c}$ rewiring |
| ..., normal ize d=T rue ,...) | Normalized <br> betwee- <br> nness <br> centrality |  |
| ..., endpoi nts =False, ...) | BC excluding endpoints |  |
| ..., K=10,...) | BC approx- ${ }^{G}=$ conn ect ed_ wat ts_ str oga tz_ gra ph( imated $n, k, p, t)$ <br> using <br> random <br> sample of <br> K nodes | $t=\max$ <br> iterations <br> to try to <br> ensure <br> connected <br> graph |
| ```nx.bet wee nne ss_ cen tra lit y_s ubs et( G, { sub set})``` | BC G=newm an_ wat ts_ str oga tz_ gra ph( $n, k$ calculated ${ }^{\prime}$ p) on subset | $p=$ probab <br> ility of <br> adding (no <br> rewiring) |
|  | Link Prediction measures | How likely are 2 nodes to connect, given an existing network |

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Page 2 of 5.

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| Network Evolution - Real-world Applications (cont |  |
| :---: | :---: |
| nx.com mon _ne igh bor s(G, n1, n2) | Calc common neighbors of nodes n1, n2 |
| nx.jac car d_c oef fic ient (G) | Normalised common neighbors measure |
| nx.res our ce_ all oca tio n_i ndex <br> (G) | Calc RAI of all nodes not already connected by an edge |
| nx.ada mic_ad ar_ ind ex(G) | As per RAI but with log of degree of common neighbor |
| nx.pre fer ent ial _at tac hme nt(G ) | Product of two nodes' degrees |
| Community Common Neighbors | Common neighbors but with bonus if they belong in same 'community' |
| nx.cn_ sou nda raj an_ hop cro ft(n 1, n2) | CCN score for $n 1$, n2 |
| G.node ['A '][ 'co mmu nit y']=1 | Add community attribute to node |
| nx.ra_ ind ex_ sou nda raj an_ hop cro ft(G) | Community <br> Resource <br> Allocation score |

These scores give only an indication of whether 2 nodes are likely to connect.
To make a link prediction, you would use these scores as features in a classification ML model.


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Page 3 of 5 .

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