

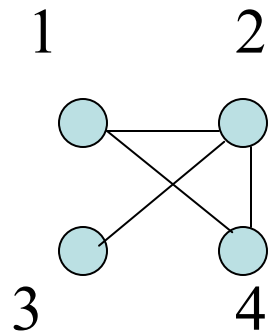
Basic Network Properties and How to Calculate Them

- Data input
- Number of nodes
- Number of edges
- Symmetry
- Degree sequence
- Average nodal degree
- Degree distribution
- Clustering coefficient
- Average hop distance (simplified average distance)
- Connectedness
- Finding and characterizing connected clusters

Matlab Preliminaries

- In Matlab everything is an array
- So
 - $a = [2 \ 5 \ 28 \ 777]$
 - for $i = a$
 - do something
 - end
- Will “do something” for $i =$ the values in array a
- $[a;b;c]$ yields a column containing $a \ b \ c$
- $[a \ b \ c]$ yields a row containing $a \ b \ c$

Matrix Representation of Networks



$$\gg A = [0 \ 1 \ 0 \ 1; 1 \ 0 \ 1 \ 1; 0 \ 1 \ 0 \ 0; 1 \ 1 \ 0 \ 0]$$

A =

$$\begin{array}{cccc} 0 & 1 & 0 & 1 \\ 1 & 0 & 1 & 1 \\ 0 & 1 & 0 & 0 \\ 1 & 1 & 0 & 0 \end{array}$$

»

A is called the adjacency matrix

$A(i, j) = 1$ if there is an arc from i to j

$A(j, i) = 1$ if there is also an arc from j to i

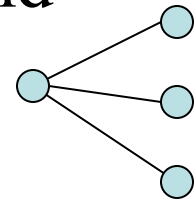
In this case A is symmetric and the graph is undirected

Directed, Undirected, Simple

- Directed graphs have one-way *arcs* or links
- Undirected graphs have two-way *edges* or links
- Mixed graphs have some of each
- Simple graphs have no multiple links between nodes and no self-loops

Basic Facts About Undirected Graphs

- Let n be the number of nodes and m be the number of edges
- Then *average nodal degree* is $\langle k \rangle = 2m/n$
- The *Degree sequence* is a list of the nodes and their respective degrees
- The sum of these degrees is $\sum_{i=1}^n d_i = 2m$
- $D = \text{sum}(A)$ in Matlab
- $\text{sum}(\text{sum}(A)) = 2m$
- Valid degree sequences add to an even number



$$D = [3 \ 1 \ 1 \ 1]$$

More Definitions and Calculations

- Geodesic - shortest path between two nodes
- Average path length = avg of all geodesics
- Graph diameter = longest geodesic

Data Input

- Edge list: each row is an arc, the numbers of the nodes it connects are in columns 1 and 2, with the first node being the source and the second being the destination of the arc
 - N_i, N_j
 - N_k, N_l
 - An edge = 2 arcs, one in each direction, requiring two entries; often only one arc is listed so you have to know if the matrix is intended to be symmetric or not
- Node list: Node #1 in column 1, the node numbers of nodes it links to by outgoing edges in the next columns, however many are necessary
 - N_1, N_i, N_j, N_k
 - N_2, N_l, N_m
- The easy way to generate these is to use Excel, save as “.csv” and load into Matlab using `dlmread('filename.csv');`

Example Node and Edge Lists

◇	A	B	C	D	E
1	1	23			
2	2	8			
3	3	4			
4	4	3	5	7	
5	5	4	6	11	
6	6	5			
7	7	4	9	11	
8	8	2	9	25	26
9	9	7	8	20	
10	10	72			
11	11	5	7	13	20

Part of a nodelist.
It is symmetric.

```
1 2
1 8
1 43
1 49
1 50
1 51
1 52
1 53
1 54
1 55
2 3
2 9
2 44
2 56
3 4
3 10
3 45
3 57
3 58
4 5
4 11
4 46
4 59
4 60
```

Part of an edgelist.
It is intended to be symmetric but each edge appears only once so you have to make the resulting matrix symmetric yourself!

Matlab Data Formats

- Comma delimited, space delimited, tab delimited
 - All can be read using `dlmread` and written using `dlmwrite`
- Lotus 1-2-3 format `wk1` can be exchanged between Excel, Matlab, and UCINET
 - Matlab uses `wk1read` and `wk1write`
 - UCINET uses excel matrix input or small matrices can be pasted into UCINET's matrix
 - Matlab also uses `xlsread` and `xlswrite` if you don't need UCINET compatibility

Data Input and Misc Ops

- `adjbuilde` builds adjacency matrix from edge list
- `adjbuildn` builds adjacency matrix from node list
- `diagnoseMatrix` tests for power law
- Miscellaneous data conversion
 - `adj2str` adjacency matrix to Matlab data structure
 - `adj2pajek` for input to Pajek graph software
 - `adj2inc` adjacency matrix to incidence matrix
 - `str2adj` reverses `adj2str`

More Matlab Preliminaries

- Logical expressions:
- $B=A > 1$ returns a matrix B with entries = 1 everywhere matrix A has an entry > 1
- B is a logical matrix and can't be operated on like numerical matrices can (can't multiply, invert, etc)
- $B=B+0$ converts B to a numerical matrix
- `Unitize(A)` makes all non-zero entries in $B = 1$
 - `function unitize=unitize(A)`
 - `% unitizes a matrix, makes all non zero entries = 1`
 - `unitize=A>0;`
 - `unitize=unitize+0;`

Still More Matlab Preliminaries

- A' = the transpose of A
- `sum(A)` adds up each column and stores the result as a row vector
- `sum(A')` adds up each row and stores the result as a row vector
- `sum(sum(A))` adds up all the entries in A
- `length(x)` counts the number of entries in vector x
- `find(x logical expr)` returns the subscripts of entries in x that satisfy the logical expression using linear indexing
 - Linear indexing gives every entry one subscript
- `length(find(x logical expr))` tells how many entries in x satisfy the logical expression
- `[i,j]= find(x logical expr)` returns the i and j subscripts of entries in matrix x that satisfy the logical expression

More Preliminaries

- $\text{idx}=[a\ b\ c]$
- $B=A(\text{idx},\text{idx})$
- $B(i,j)=A(\text{one of the combinations of } a,b,c \text{ in pairs})$

– $A =$

1	2	3	4	5
6	7	8	9	10
11	12	13	14	15
16	17	18	19	20
21	22	23	24	25

– $\text{idx} = [1\ 2\ 5]$

– $B =$

1	2	5
6	7	10
21	22	25

Number of nodes

- Normally this is just the size of the network measured by the number of rows in the adjacency matrix: $size(A, 1)$ in Matlab
 - **function** numnodes=numnodes(A)
 - %finds number of nodes in A including isolates
 - numnodes=size(A,1);
- But if there are isolated nodes, it's useful to count only the non-isolated ones
- So numnonisolationnodes works differently
 - **function** nodes = numnonisolationnodes(A)
 - % counts non-isolated nodes in a matrix
 - A=unitize(A+A');
 - nodes=min(length(find(sum(A')~=0)),length(find(sum(A)~=0)));

Number of Links or Edges

- If A is symmetric (the network is undirected) then links are called edges and the number of edges is the number of entries in the adjacency matrix/2
- Max = $n(n-1)/2$; min = $n-1$
- If A is asymmetric (the network is directed or mixed) then directed links are called arcs and the number of edges+number of arcs = the same thing as above
- So numedges symmetrizes the matrix before calculating

Numedges

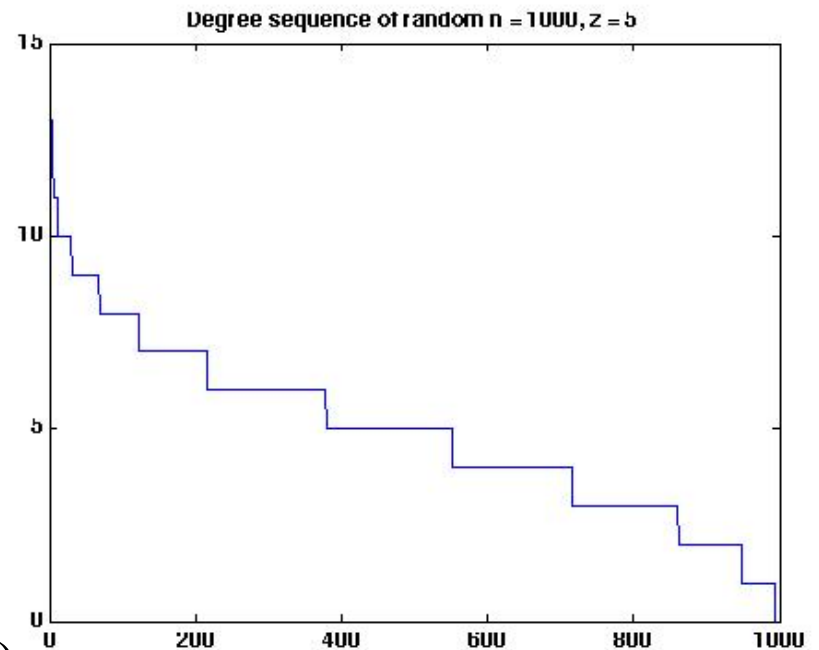
```
function numedges = numedges(A)
%counts edges in matrix A
%works when not all nodes have edges
AT=A+A';
numedges=sum(sum(AT~=0))/2;
```


Symmetry

- If $A = A'$ (transpose) then the network is symmetric (undirected)
- In Matlab just ask *isequal(A,A')*
- The answer = 1 if yes, 0 if no
- To find non-symmetric entries, use
 - $[i,j]=\text{find}(A \sim A')$

Degree Sequence

- `kvec(A)` finds the degrees of the nodes
 - function `kvec = kvec(A)`
 - %returns the k vector of a symmetric matrix
 - `kvec=sum(A);`
- Directed graphs have k_{in} and k_{out}
 - `kin=sum(A)`
 - `kout=sum(A')`
- Routine `pds` plots degree seq vs percent of nodes
- Plot to right made with
 - `plot(kvec(A))`
 - after sorting `kvec(A)` in descending order using
 - `kvsorted=sort(kvec,'descend')`



Average Nodal Degree

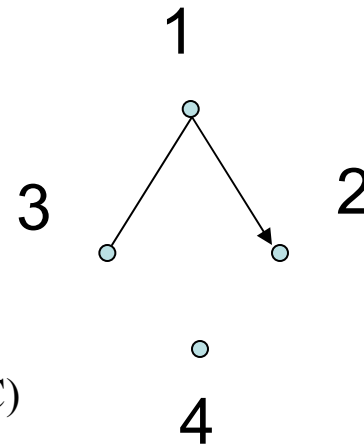
- Variously denoted \bar{k} or $\langle k \rangle$ (“k-hat”)
- Standard khat:
 - `function khat = khat(A)`
 - %function to find average nodal degree of a symmetric matrix
 - %includes isolated nodes, that is, nodes with $k = 0$
 - %to find $\langle k \rangle$ excluding isolates, use `khatnoniso`
 - `numedges2=numedges(A)*2;`
 - `nodes=size(A,1);`
 - `khat=numedges2/nodes;`
- `khatnoniso`:
 - `function khatnoniso = khat(A)`
 - %function to find average nodal degree of a symmetric matrix
 - %does not include isolated nodes, that is, nodes with $k = 0$
 - %to find $\langle k \rangle$ including isolates, use `khat`
 - `numedges2=numedges(A)*2;`
 - `nodes=numnonisnodes(A);`
 - `khatnoniso=numedges2/nodes;`

Average Hop Distance

- Simplified average distance between two randomly selected nodes, assuming all node-node distances = 1
- Found by taking powers of A
- A^2 contains (i,j) entries = 1 where node i links to node j via a 2-hop path; A^3 contains (i,j) entries = 1 where node i links to node j via a 3-hop path, etc.
- If $A^{power}(i,j)$ flips from 0 to 1 for the first time when $power=k$ then there is a path of length k from i to j .
- When all possible paths have been found, the network's diameter $d = k$. `distmat.m` works this way.
- Can probably be generalized to non-unity distances if they obey the triangle inequality
- Otherwise actual shortest paths must be found, computationally intensive

Hop Distance Example Using distmat.m

- `>> AC=[0 1 1 0;0 0 0 0;1 0 0 0;0 0 0 0]`
- `AC =`
 - 0 1 1 0
 - 0 0 0 0
 - 1 0 0 0
 - 0 0 0 0
- `>> [distance_matrix, avg_dist, diameter]=distmat(AC)`
- `distance_matrix =`
 - 0 1 1 0
 - 0 0 0 0
 - 1 2 0 0
 - 0 0 0 0
- `avg_dist =`
 - 1.2500 (5 total length of 4 paths)
- `diameter =`
 - 2



row:col = from:to

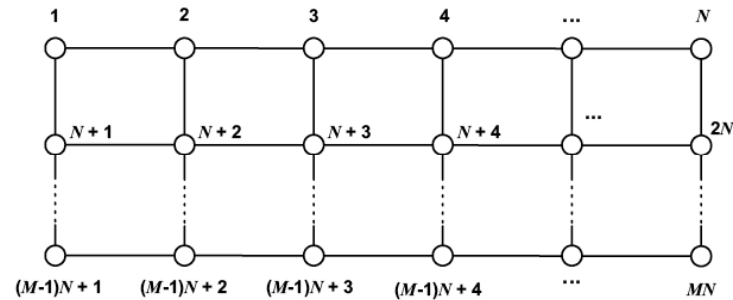
Average Hop Distance for Regular Network Structures

- Grids, stars, stars with concentric circles, etc
- Of interest to analysts of transportation networks and cellphone networks
 - Transport: nodes are transfer stations
 - Cellphones: nodes are transmission towers
- Lots of work done by Leonard Miller, NIST
- Evaluates the average hop distance formula from a few slides back in closed form for several such networks
- http://w3.antd.nist.gov/wctg/netanal/netanal_netmodels.html

CONNECTIVITY PROPERTIES OF MESH AND RING/MESH NETWORKS

L. E. Miller

2 April 2001



the average hop distance for a mesh network is

$$\bar{m} = \frac{N + M}{3}$$

$$Connectivity = \frac{M \times 2(N - 1) + 2(M - 1) \times N}{NM(NM - 1)} = \frac{2[M(2N - 1) - N]}{NM(NM - 1)}$$

$$Connectivity = \frac{actual_links}{max_poss_links} = \frac{m}{n(n-1)/2} = \frac{\langle k \rangle}{n-1} = clust_{random}$$


$$numnodes = MN$$

$$numedges = M(N - 1) + N(M - 1)$$

$$\langle k \rangle = 2 \frac{M(N - 1) + N(M - 1)}{MN} = 4 - \frac{2}{M} - \frac{2}{N}$$

Adjacency Matrix for Grids

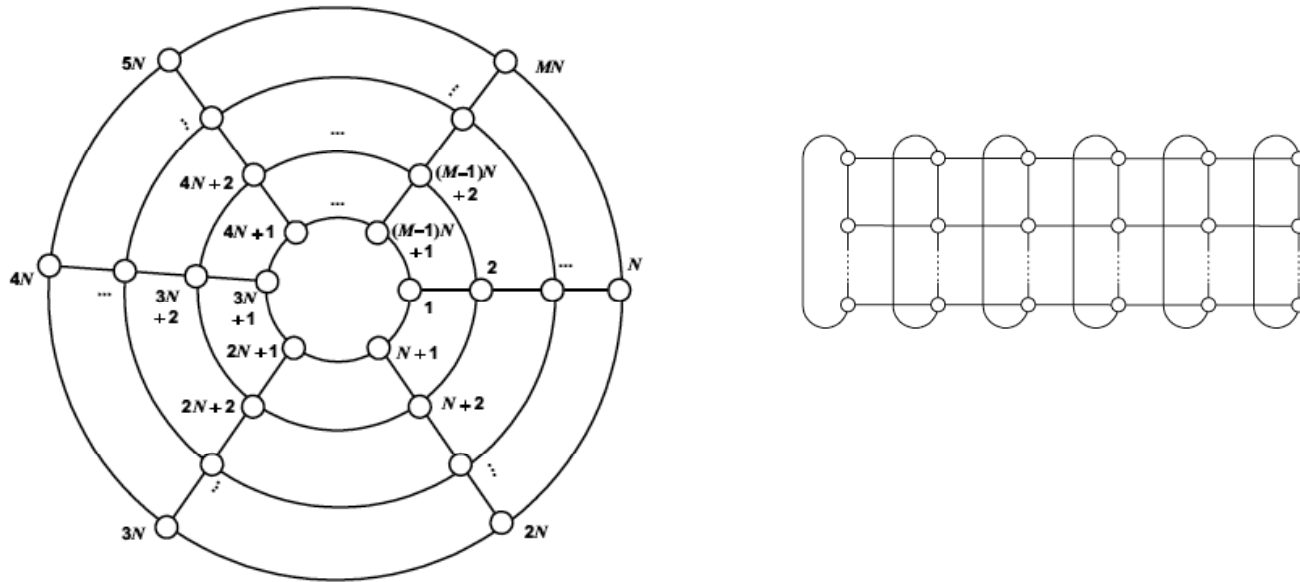
The adjacency (one-hop connectivity) matrix for such a network, in which a 1 entry at (i, j) indicates a connection from node i to node j and a 0 entry at (i, j) indicates no connection from node i to node j , is an $NM \times NM$ matrix with the form given by

Should read $A_{NM \times MN}$ 

$$A_{M \times N} = \begin{bmatrix} A_N & I_N & 0 & \cdots & 0 \\ I_N & A_N & I_N & \cdots & 0 \\ 0 & I_N & A_N & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & A_N \end{bmatrix} \quad (1)$$

where I_N is the $N \times N$ identity matrix and A_N is the $N \times N$ adjacency matrix for a single row of N nodes connected in tandem. The structure of A_N is easily determined to be a matrix of 0s, except for $N - 1$ 1s on the first upper diagonal and $N - 1$ 1s on the first lower diagonal, for example,

$$A_6 = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 0 & 1 & 0 \end{bmatrix} \quad (2)$$



the average hop distance for a ring-mesh network is

$$\bar{m} = \frac{1}{12(NM - 1)} \times \begin{cases} 4M(N^2 - 1) + 3N(M^2 - 1), & M \text{ odd} \\ 4M(N^2 - 1) + 3NM^2, & M \text{ even} \end{cases}$$

$$\text{Connectivity} = \frac{M \times 2(N - 1) + 2M \times N}{NM(NM - 1)} = \frac{2(2N - 1)}{N(NM - 1)}$$

$$\text{numnodes} = MN$$

$$\text{numedges} = M(N - 1) + N(M - 1) + N$$

$$\langle k \rangle = 2 \frac{M(N - 1) + N(M - 1) + N}{MN} = 2 \frac{2MN - M}{MN} = 4 - \frac{2}{N}$$

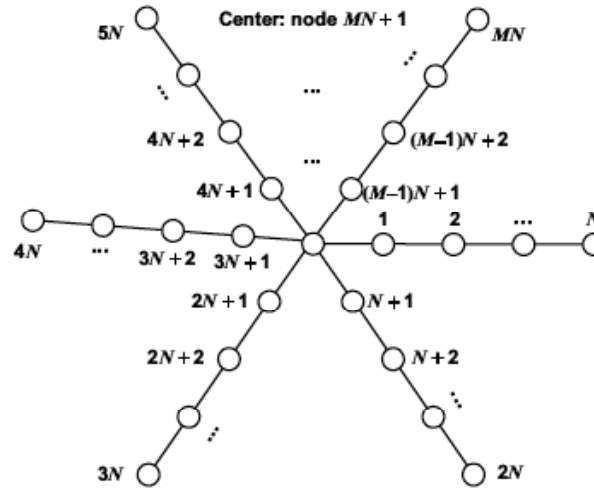
CONNECTIVITY PROPERTIES OF STAR AND STAR/MESH NETWORKS

L. E. Miller

2 April 2001

1. STAR NETWORK

A star network connected by bidirectional links can be modeled as M "rays" of N nodes plus a center node, as illustrated in the following figure:



$$numnodes = MN + 1$$

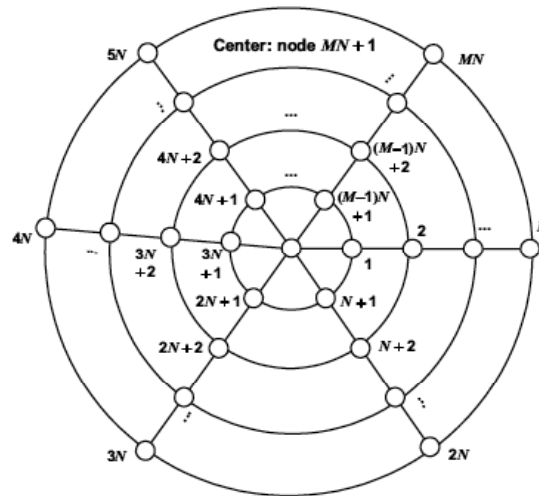
$$numedges = MN$$

$$\langle k \rangle = 2 \frac{MN}{MN + 1}$$

the average hop distance for a star network is

$$\bar{m} = \frac{(N + 1)(2MN + M - N + 1)}{3(MN + 1)}$$

$$Connectivity = \frac{M \times 2(N - 1) + 2M \times 1}{NM(NM + 1)} = \frac{2}{NM + 1}$$



the average hop distance for a star-mesh network is

$$\bar{m} = \frac{MN^2 + 6MN - M - 9N + 3}{3(MN + 1)}$$

$$\text{Connectivity} = \frac{M \times 2(N - 1) + 2M \times N + 2M \times 1}{NM(NM + 1)} = \frac{4}{NM + 1}$$

$$\text{numnodes} = MN + 1$$

$$\text{numedges} = \text{staredges} + \text{ringedges} = MN + MN = 2MN$$

$$\langle k \rangle = 4 \frac{MN}{MN + 1}$$

Clustering Coefficient

- Asks if two neighbors of a node are neighbors
- This can be answered for every node and then the average taken as the network's overall C
- Normalize by $k(k-1)/2$ = the max number of triangles for a node whose degree = k
- Calculated by finding A^3 and looking at the diagonal, whose entries count the number of paths of length 3 that return to the start node
- $A^n(i,j)$ = number of paths from i to j of length n
- Paths of length 3 that return to their start are triangles
- If the network is undirected then there will be 2 for each triangle since it counts both directions

Clustering Coeff Code

```
function [clust,cc] = clust(A)
% the clustering coefficient modified from code by Ed Schneiderman, Johns Hopkins U
A=sortbyk(A); % sorts A so first node has biggest k, etc
% we will calculate the average clustering coefficient ONLY for those
% vertices where the number of neighbors is >1. We can calculate it
% for all vertices as well, by defining that if the vertex has zero
% neighbors, its clustering coeff is zero, and if it has one neighbor,
% its clustering coeff is one.
s=size(A,1);
T=A*A*A;
triangles=0.5*diag(T);
% a vector of s elements, each element being the # of triangles around that vertex
% note that we will use the *undirected* network here

k=sum(A,2); % a vector with the number of edges for each vertex

c_avg=0;
cc=zeros(s,1);
N_k_morethanone=0;

for i=1:s
if k(i)>1
    cc(i)=2*triangles(i)/(k(i)*(k(i)-1)); % each node's clustering coeff
c_avg=c_avg+cc(i); N_k_morethanone=N_k_morethanone+1;
end
end
c_avg=c_avg/N_k_morethanone;
clust=c_avg;
```

Two Ways to Define Clustering Coeff

- Newman Equation (3)
- Newman Equation (5)
- These are called `clustEq3` and `clustEq5`

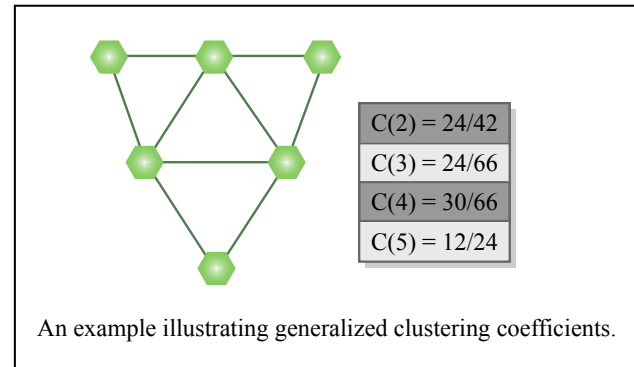
Generalized Clustering Coefficient

- Asks for loops of any length, not just 3
- This is a lot harder to do
- Paper by Huang discusses this

Link Prediction Based on Graph Topology: The Predictive Value of the Generalized Clustering Coefficient

Zan Huang
Department of Supply Chain and Information Systems
Pennsylvania State University
419 Business Building
University Park, PA, 16802
zanhuang@psu.edu

LinkKDD'06, August 20, 2006, Philadelphia, Pennsylvania, USA.
Copyright 2006 ACM 1-59593-446-6/06/0008...\$5.00



An example illustrating generalized clustering coefficients.

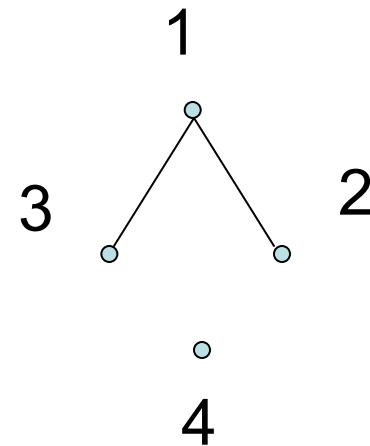
Image by MIT OpenCourseWare.

Connectedness

```
function yn = isconnected(g)
% isconnected(g) -- determine if g is a connected graph
% written by Ed Schneiderman, Johns Hopkins U, comments added by Whitney
% Gives the right answer only if g is undirected
% Works by asking if node 1 can be reached from every node
% If we set x(j)=1 before starting the while loop, it asks if node j can be
% reached from every node
% By testing each node this way, we can see if a directed graph is
% connected. This is how isconnectedasym works.
% modified to work when not all nodes have edges 9-14-06
% g = g+0.;
n = length(g);
x = zeros(n,1);
x(1)=1;% by definition, node 1 can be reached from node 1
while(1)
    y = x; % remembers previous x
    x = g*x + x;% g*x(j) = 1 when node j has been reached from node 1.
                % Adding in x just remembers previously reached nodes.
    x = x>0;% unitize x
    if (x==y) ;% no new nodes were reached
        break
    end
end
if (sum(x)<numnodes(g) )
    yn = 0;% we did not reach every node
else
    yn = 1;% we reached every node, so it's connected
end
```


Example of Connectedness Calculation

$$\begin{array}{r}
 \text{AC1} = \\
 \begin{array}{cccc}
 0 & 1 & 1 & 0 \\
 1 & 0 & 0 & 0 \\
 1 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0
 \end{array}
 \end{array}
 \quad
 \begin{array}{r}
 \text{length}(\text{AC1}) \\
 = \\
 4
 \end{array}$$



$$\begin{array}{r}
 x = \\
 \begin{array}{c}
 1 \\
 0 \\
 0 \\
 0
 \end{array}
 \end{array}$$

Start at node 1

$\text{AC1} \cdot x$	$x = \text{AC1} \cdot x + x$	$x = \text{AC1} \cdot x + x$	$\text{unitize}(x)$
=	=	=	=
$\begin{array}{c} 0 \\ 1 \\ 1 \\ 0 \end{array}$	$\begin{array}{c} 1 \\ 1 \\ 1 \\ 0 \end{array}$	$\begin{array}{c} 3 \\ 2 \\ 2 \\ 0 \end{array}$	$\begin{array}{c} 1 \\ 1 \\ 1 \\ 0 \end{array}$

So we're done.
 Length(x) < 4 so it's disconnected

We reach nodes 2 and 3 right away and get no further

Network Components

- If a network is disconnected, it consists of isolated clusters and isolated nodes
- Calling sequence on next page for code written by Mo-Han Hsieh finds all these clusters

componentCount calling sequence only

```
% [componentCount] is used to generate the component partition of a matrix.  
% Its input is the adjacency matrix, A.  
% Its outputs are partition, componentList, mainNum, and singletonNum.  
% /partition/ tells which component each node is in.  
% /componentList/ is a list of components and the number of nodes in each  
% component. Its format is: [component ID, number of nodes in it].  
% /mainNum/ is the number of components which have at least two members, and  
% /singletonNum/ is the number of singletons.  
  
function [partition,componentList,mainNum,singletonNum]=componentCount(A)
```

componentCount Example

AC =

0	1	1	0
0	0	0	0
1	0	0	0
0	0	0	0

[aa,bb,cc,dd]=componentCount(AC)

aa =

1
1
1
2

Nodes 1 - 3 are in comp 1
Node 4 is in comp 2

bb =

1 3
2 1

Comp 1 has 3 nodes
Comp 2 has 1 node

cc =

1

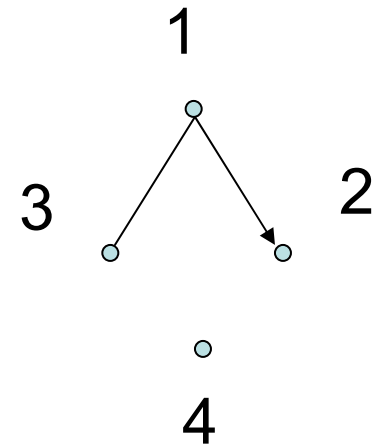
comps with >1 node = 1

dd =

1

singletons = 1

size(bb,1) = number of components = 2



List of Some Useful Routines

- `Adjbuilde`, `adjbuildn` for making adjacency matrices from edge lists and node lists
- `Khat` for finding average nodal degree
- `Numnodes` for counting nodes
- `Numedges` for counting edges
- `Is connected` to see if the network is whole or separated into isolated network components
- `Distmat` for finding shortest paths if node pairs are separated by one hop (don't use `ave_path_length` or `shortest_path`)
- `Randmatrix` for building a Poisson random network (don't use `random_graph` unless you use the default that gives a Poisson network)
- `ClustEq3` or `clustEq5` for finding clustering coefficient
- `Componentcount` for finding separate components

More (later)

- Centrality
 - Node centrality
 - Edge centrality
 - Betweenness centrality
- Calculations
- Degree correlation
 - Joint degree distribution
 - K-nearest neighbors
 - Pearson degree correlation
- Rich club metric
- Degree-preserving rewiring
- Generating a graph that has a specified degree sequence
- Finding Pearson degree correlation
- Finding communities

MIT OpenCourseWare
<http://ocw.mit.edu>

ESD.342 Network Representations of Complex Engineering Systems
Spring 2010

For information about citing these materials or our Terms of Use, visit: <http://ocw.mit.edu/terms>.