Connectivity

V. Batagelj

Connectivity

Condensation

Bow-tie

Other connectivities

Important nodes

Closeness

Betweeness

Hubs and authorities

Clustering

Introduction to Network Analysis using Pajek 4. Structure of networks 2 Connectivity

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Outline

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- Connectivity
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- Other connectivities
- Important nodes
- Closeness
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- Clustering

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e-mail: vladimir.batagelj@fmf.uni-lj.si Current version of slides (April 17, 2018 at 02:59): http://vladowiki.fmf.uni-lj.si/doku.php?id=pajek:ev:pde



Walks

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length |s| of the walk s is the number of links it contains.

s = (j, h, l, g, e, f, h, l, e, c, b, a)|s| = 11

A walk is closed iff its initial and terminal node coincide.

If we don't consider the direction of the links in the walk we get a semiwalk or chain.

trail – walk with all links different path – walk with all nodes different

cycle – closed walk with all internal nodes different

A graph is acyclic if it doesn't contain any cycle.



Shortest paths



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The diameter of a graph equals to the distance between the most distant pair of nodes: $D = \max_{u,v \in \mathcal{V}} d(u, v)$.

Network/Create New Network/Subnetwork with Paths/



Shortest paths





Equivalence relations and Partitions

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A relation R on \mathcal{V} is an equivalence relation iff it is reflexive $\forall v \in \mathcal{V} : vRv$, symmetric $\forall u, v \in \mathcal{V} : uRv \Rightarrow vRu$, and transitive $\forall u, v, z \in \mathcal{V} : uRz \land zRv \Rightarrow uRv$.

Each equivalence relation determines a partition into equivalence classes $[v] = \{u : vRu\}.$

Each partition **C** determines an equivalence relation $uRv \Leftrightarrow \exists C \in \mathbf{C} : u \in C \land v \in C$.

k-neighbors of *v* is the set of nodes on 'distance' *k* from *v*, $N^k(v) = \{u \in v : d(v, u) = k\}.$ The set of all *k*-neighbors, k = 0, 1, ... of *v* is a partition of \mathcal{V} .

k-neighborhood of
$$v$$
, $N^{(k)}(v) = \{u \in v : d(v, u) \leq k\}$.

Network/Create Partition/k-Neighbors



Motorola's neighborhood



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The thickness of edges is a square root of its value.



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Node u is reachable from node v iff there exists a walk with initial node v and terminal node u.

Node v is weakly connected with node u iff there exists a semiwalk with v and u as its end-nodes.

Node v is strongly connected with node u iff they are mutually reachable.

Weak and strong connectivity are equivalence relations. Equivalence classes induce weak/strong components.

Network/Create Partition/Components/



Weak components



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Reordering the nodes of network such that the nodes from the same class of weak partition are put together we get a matrix representation consisting of diagonal blocks – weak components.

Most problems can be solved separately on each component and afterward these solutions combined into final solution.



Special graphs - bipartite, tree

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A graph $\mathcal{G} = (\mathcal{V}, \mathcal{L})$ is bipartite iff its set of nodes \mathcal{V} can be partitioned into two sets \mathcal{V}_1 and \mathcal{V}_2 such that every link from \mathcal{L} has one end-node in \mathcal{V}_1 and the other in \mathcal{V}_2 .

A weakly connected graph \mathcal{G} is a tree iff it doesn't contain loops and semicycles of length at least 3.



Condensation



delete all loops and identify parallel arcs the obtained reduced grap is acyclic. For every acyclic graph an ordering / level function $i: \mathcal{V} \to \mathbb{N}$ exists s.t. $(u, v) \in \mathcal{A} \Rightarrow i(u) < i(v)$.



Condensation – Example

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Network/Create Partition/Components/Strong [1] Operations/Network+Partition/Shrink Network [1][0] Network/Create New Network/Transform/Remove/Loops [yes] Network/Acyclic Network/Depth Partition/Acyclic Partition/Make Permutation Permutation/Inverse Permutation select partition [Strong Components] Operations/Partition+Permutation/Functional Composition Partition*Permut Partition/Make Permutation select [original network] File/Network/Export Matrix to EPS/Using Permutation





Internal structure of strong components



Let d be the largest common divisor of lengths of closed walks in a strong component.

The component is said to be simple, iff d = 1; otherwise it is periodical with a period d.

The set of nodes \mathcal{V} of strongly connected directed graph $\mathcal{G} = (\mathcal{V}, R)$ can be partitioned into d clusters $\mathcal{V}_1, \mathcal{V}_2, \ldots, \mathcal{V}_d$, s.t. for every arc $(u, v) \in R$ holds $u \in \mathcal{V}_i \Rightarrow v \in \mathcal{V}_{(imod d)+1}$.

Network/Create Partition/

Components/Strong-Periodic



Bonhoure's periodical graph. Pajek data



Bow-tie structure of the Web graph

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Kumar &: The Web as a graph

Let S be the largest strong component in network \mathcal{N} ; \mathcal{W} the weak component containing S; \mathcal{I} the set of nodes from which S can be reached; \mathcal{O} the set of nodes reachable from S; \mathcal{T} (tubes) set of nodes (not in S) on paths from \mathcal{I} to \mathcal{O} ; $\mathcal{R} = \mathcal{W} \setminus (\mathcal{I} \cup S \cup \mathcal{O} \cup \mathcal{T})$ (tendrils); and $\mathcal{D} = \mathcal{V} \setminus \mathcal{W}$. The partition

$$\{\mathcal{I}, \mathcal{S}, \mathcal{O}, \mathcal{T}, \mathcal{R}, \mathcal{D}\}$$

is called the **bow-tie** partition of \mathcal{V} .

Note: chains can exist in the set \mathcal{R} .

Network/Create Partition/Bow-Tie

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Biconnectivity

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Nodes u and v are biconnected iff they are connected (in both directions) by two independent (no common internal node) paths. Biconnectivity determines a partition of the set of links.

A node is an articulation node iff its deletion increases the number of weak components in a graph.

A link is a **bridge** iff its deletion increases the number of weak components in a graph.

Network/Create New Network/with Bi-Connected Components.

The notion of biconnectivity can be generalized do k-connectivity. No very efficient algorithm for k > 3 exists.



k-connectivity

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- Node connectivity κ of graph \mathcal{G} is equal to the smallest number of nodes that, if deleted, induce a disconnected graph or the trivial graph K_1 .
 - Link connectivity λ of graph \mathcal{G} is equal to the smallest number of links that, if deleted, induce a disconnected graph or the trivial graph K_1 .

```
Whitney's inequality: \kappa(\mathcal{G}) \leq \lambda(\mathcal{G}) \leq \delta(\mathcal{G}) .
```

```
Graph \mathcal{G} is (node) k-connected, if \kappa(\mathcal{G}) \ge k and is link k-connected, if \lambda(\mathcal{G}) \ge k.
```

Whitney / Menger theorem: Graph G is node/link k-connected iff every pair of nodes can be connected with k node/link internally disjoint (semi)walks.



Triangular and short cycle connectivities

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In an undirected graph we call a triangle a subgraph isomorphic to K_3 .

A sequence $(T_1, T_2, ..., T_s)$ of triangles of \mathcal{G} (node) triangularly connects vertices $u, v \in \mathcal{V}$ iff $u \in T_1$ and $v \in T_s$ or $u \in T_s$ and $v \in T_1$ and $\mathcal{V}(T_{i-1}) \cap \mathcal{V}(T_i) \neq \emptyset$, i = 2, ...s. It edge triangularly connects vertices $u, v \in \mathcal{V}$ iff a stronger version of the second condition holds $\mathcal{E}(T_{i-1}) \cap \mathcal{E}(T_i) \neq \emptyset$, i = 2, ...s.



Node triangular connectivity is an equivalence on \mathcal{V} ; and edge triangular connectivity is an equivalence on \mathcal{E} . See the paper.



Triangular connectivity in directed graphs

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If a graph \mathcal{G} is mixed we replace edges with pairs of opposite arcs. In the following let $\mathcal{G} = (\mathcal{V}, \mathcal{A})$ be a simple directed graph without loops. For a selected arc $(u, v) \in \mathcal{A}$ there are only two different types of directed triangles: **cycl**ic and **transitive**.



For each type we get the corresponding triangular network \mathcal{N}_{cyc} and \mathcal{N}_{tra} by determining the corresponding weight w_{cyc} or w_{tra} to its arcs, counting the number of cyclic/transitive triangles that contain the arc. We remove arcs with weight zero. The notion of triangular connectivity can be extended to the notion of short (semi) cycle connectivity.

Network/Create New Network/with Ring Counts/3-Rings/Directed



Edge-cut at level 16 of triangular network of Erdős collaboration graph

Connectivity





Arc-cut at level 11 of transitive triangular network of ODLIS dictionary





Important vertices in network

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To identify important / interesting elements (nodes, links) in a network we often try to express our intuition about important / interesting element using an appropriate measure (index, weight) following the scheme

larger is the measure value of an element, more important / interesting is this element

Too often, in analysis of networks, researchers uncritically pick some measure from the literature. For formal approach see Roberts.

It seems that the most important distinction between different node indices is based on the view/decision whether the network is considered directed or undirected.



... Important nodes in network

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This gives us two main types of indices:

- networks containing directed links (we replace edges by pairs of opposite arcs): measures of importance; with two subgroups: measures of influence, based on out-going arcs; and measures of support, based on incoming arcs;
- measures of centrality, based on all links.

For undirected networks all three types of measures coincide. If we change the direction of all arcs (replace the relation with its inverse relation) the measure of influence becomes a measure of support, and vice versa.



... Important nodes in network

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The real meaning of measure of importance depends on the relation described by a network. For example the most 'important' person for the relation ' $_{--}$ doesn't like to work with $_{--}$ ' is in fact the least popular person.

Removal of an important node/link from a network produces a substantial change in structure/functioning of the network.



Normalization

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Let $p: \mathcal{V} \to \mathbb{R}$ be an index in network $\mathcal{N} = (\mathcal{V}, \mathcal{L}, p)$. If we want to compare indices p over different networks we have to make them comparable. Usually we try to achieve this by normalization of p. Let $\mathcal{N} \in \mathbf{N}(\mathcal{V})$, where $\mathbf{N}(\mathcal{V})$ is a selected family of networks over the same set of nodes V,

$$p_{max} = \max_{\mathcal{N} \in \mathbf{N}(\mathcal{V})} \max_{v \in \mathcal{V}} p_{\mathcal{N}}(v) \quad \text{and} \quad p_{min} = \min_{\mathcal{N} \in \mathbf{N}(\mathcal{V})} \min_{v \in \mathcal{V}} p_{\mathcal{N}}(v)$$

then we define the normalized index as

$$p'(v) = rac{p(v) - p_{min}}{p_{max} - p_{min}} \in [0, 1]$$

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Degrees

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The simplest index are the degrees of nodes. Since for simple networks deg_{min} = 0 and deg_{max} = n - 1, the corresponding normalized indices are centrality $\deg'(v) = \frac{\deg(v)}{n-1}$ and similary support $indeg'(v) = \frac{indeg(v)}{n}$ influence $outdeg'(v) = \frac{outdeg(v)}{n}$ Instead of degrees in original network we can consider also the degrees with respect to the reachability relation (transitive closure). Network/Create Partition/Degree Network/Create Vector/Centrality/Degree Network/Create Vector/Centrality/Proximity Prestige



Closeness

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Most indices are based on the distance d(u, v) between vertices in a network $\mathcal{N} = (\mathcal{V}, \mathcal{L})$. Two such indices are radius $r(v) = \max_{u \in \mathcal{V}} d(v, u)$ total closeness $S(v) = \sum_{u \in \mathcal{V}} d(v, u)$ These two indices are measures of influence – to get measures of support we have to replace in definitions d(u, v) with d(v, u). If the network is not strongly connected r_{max} and S_{max} are equal to ∞ . Sabidussi (1966) introduced a related measure 1/S(v); or in its normalized form

closeness $cl(v) = \frac{n-1}{\sum_{u \in \mathcal{V}} d(v, u)}$

 $D = \max_{u,v \in \mathcal{V}} d(v, u)$ is called the diameter of network.

Network/Create Vector/Centrality/Closeness Network/Create New Network/Subnetwork with Paths/Info on Diameter



Betweeness

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Important are also the nodes that can control the information flow in the network. If we assume that this flow uses only the shortest paths (geodesics) we get a measure of betweeness (Anthonisse 1971, Freeman 1977)

$$b(v) = \frac{1}{(n-1)(n-2)} \sum_{\substack{u,t \in \mathcal{V}: g_u, t > 0 \\ u \neq v, t \neq v, u \neq t}} \frac{g_{u,t}(v)}{g_{u,t}}$$

where $g_{u,t}$ is the number of geodesics from u to t; and $g_{u,t}(v)$ is the number of those among them that pass through node v.

For computation of geodesic matrix see Brandes.

Network/Create Vector/Centrality/Betweenness



Padgett's Florentine families



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		close	between
1.	Acciaiuoli	0.368421	0.000000
2.	Albizzi	0.482759	0.212454
з.	Barbadori	0.437500	0.093407
4.	Bischeri	0.400000	0.104396
5.	Castellani	0.388889	0.054945
6.	Ginori	0.333333	0.000000
7.	Guadagni	0.466667	0.254579
8.	Lamberteschi	0.325581	0.000000
9.	Medici	0.560000	0.521978
10.	Pazzi	0.285714	0.000000
11.	Peruzzi	0.368421	0.021978
12.	Ridolfi	0.500000	0.113553
13.	Salviati	0.388889	0.142857
14.	Strozzi	0.437500	0.102564
15.	Tornabuoni	0.482759	0.091575



Hubs and authorities

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To each node v of a network $\mathcal{N} = (\mathcal{V}, \mathcal{L})$ we assign two values: quality of its content (authority) x_v and quality of its references (hub) y_v .

A good authority is selected by good hubs; and good hub points to good authorities (see Kleinberg).

$$x_v = \sum_{u:(u,v)\in\mathcal{L}} y_u$$
 and $y_v = \sum_{u:(v,u)\in\mathcal{L}} x_u$

Let **W** be a matrix of network \mathcal{N} and **x** and **y** authority and hub vectors. Then we can write these two relations as $\mathbf{x} = \mathbf{W}^T \mathbf{y}$ and $\mathbf{y} = \mathbf{W} \mathbf{x}$.

We start with $\mathbf{y} = [1, 1, ..., 1]$ and then compute new vectors \mathbf{x} and \mathbf{y} . After each step we normalize both vectors. We repeat this until they stabilize.

We can show that this procedure converges. The limit vector \mathbf{x}^* is the principal eigen vector of matrix $\mathbf{W}^T \mathbf{W}$; and \mathbf{y}^* of matrix $\mathbf{W} \mathbf{W}^T$.



... Hubs and authorities

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Similar procedures are used in search engines on the web to evaluate the importance of web pages. PageRank, PageRank / Google, HITS / AltaVista, SALSA, theory.

Network/Create New Network/Subnetwork with Paths/Info on Diameter Network/Create Vector/Centrality/Closeness Network/Create Vector/Centrality/Betweeness Network/Create Vector/Centrality/Hubs-Authorities Network/Create Vector/Centrality/Clustering Coefficients

Examples: Krebs, Krempl. World Cup 1998 in Paris, 22 national teams. A player from first country is playing in the second country.

There exist other measures based on eigen-values and eigen-vectors such as Katz, Bonachich and Brandes. See also Borgatti.



... Hubs and authorities: football

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Hubs and authorities

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Exporters (hubs)



Importers (authorities)



Clustering

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Let $\mathcal{G} = (\mathcal{V}, \mathcal{E})$ be simple undirected graph. Clustering in node v is usually measured as a quotient between the number of links in subgraph $\mathcal{G}^1(v) = \mathcal{G}(N^1(v))$ induced by the neighbors of node v and the number of links in the complete graph on these nodes:

$$C(v) = \begin{cases} \frac{2|\mathcal{L}(\mathcal{G}^{1}(v))|}{\deg(v)(\deg(v)-1)} & \deg(v) > 1\\ 0 & \text{otherwise} \end{cases}$$

We can consider also the size of node neighborhood by the following correction $% \left({{{\rm{cons}}} \right) = 0} \right)$

$$C_1(v) = rac{\deg(v)}{\Delta}C(v)$$

where Δ is the maximum degree in graph \mathcal{G} . This measure attains its largest value in nodes that belong to an isolated clique of size Δ . Network/Create Vector/Clustering Coefficients/CC2 paper

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User defined indices

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Xingqin Qi et al. defined in their paper Terrorist Networks, Network Energy and Node Removal a new measure of centrality based on Laplacian energy – Laplacian centrality

$$L(v) = \deg(v)(\deg(v) + 1) + 2\sum_{u \in N(v)} \deg(u)$$

select the network Network/Create Vector/Centrality/Degree/All Operations/Network+Vector/Neighbours/Sum/All [False] Vector/Transform/Multiply by [2] select the degree vector as First select the degree vector as Second Vectors/Multiply (First*Second) Vectors/Add (First+Second) select the 2*sum on neighbors as Second Vectors/Add (First+Second) dispose auxiliary vectors File/Vector/Change Label [Laplace All centrality]

macro



Network centralization measures

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Extremal approach: Let $p : \mathcal{V} \to \mathbb{R}$ be an index in network $\mathcal{N} = (\mathcal{V}, \mathcal{L})$. We introduce the quantities

$$p^{\star} = \max_{v \in \mathcal{V}} p(v)$$

$$D = \sum_{v \in \mathcal{V}} (p^{\star} - p(v))$$

$$D^{\star} = \max\{D(\mathcal{N}) : \mathcal{N} \text{ is a network on } \mathcal{V}\}$$

Then we can define centralization with respect to p

$$C_p(\mathcal{N}) = \frac{D(\mathcal{N})}{D^*}$$

Usually the most centralized graph is the star S_n and the least centralized is the complete graph K_n .



... Network centralization measures

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Variational approach: The other approach is based on the variance. First we compute the average node centrality

$$\overline{p} = \frac{1}{n} \sum_{v \in \mathcal{V}} p(v)$$

and then define

$$V_p(\mathcal{N}) = \frac{1}{n} \sum_{v \in \mathcal{V}} (p(v) - \overline{p})^2$$

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