

Social Network Analysis

Master Sociology 2014-15
University of Amsterdam

Programme for today

Session 1: Monday, Jan 12, REC-H 2.01 (computer lab)

- Methods 1: Introduction to graph theory
- Methods 2: Cohesion

Session 2: Wednesday, Jan 14, REC-H 2.01 (computer lab)

- Methods 3: Cohesion / Centrality
- Methods 4: Centrality

Session 3: Thursday, Jan 15, REC-G S.09

- Theory 2: QAP correlations and regression

Session 4: Friday, Jan 16, 13-15, REC-G S.14

- Test 2

UCInet basics

- We learn the program by doing SNA; mechanics of the program are not the focus of this class
- Quick start guide to UCInet
 - <https://sites.google.com/site/ucinetsoftware/document>
- Working directory
C:/Program Files (x86)/Analytic Technologies/Datafiles
- In this lab, the files we produce will be stored in C:/Temp

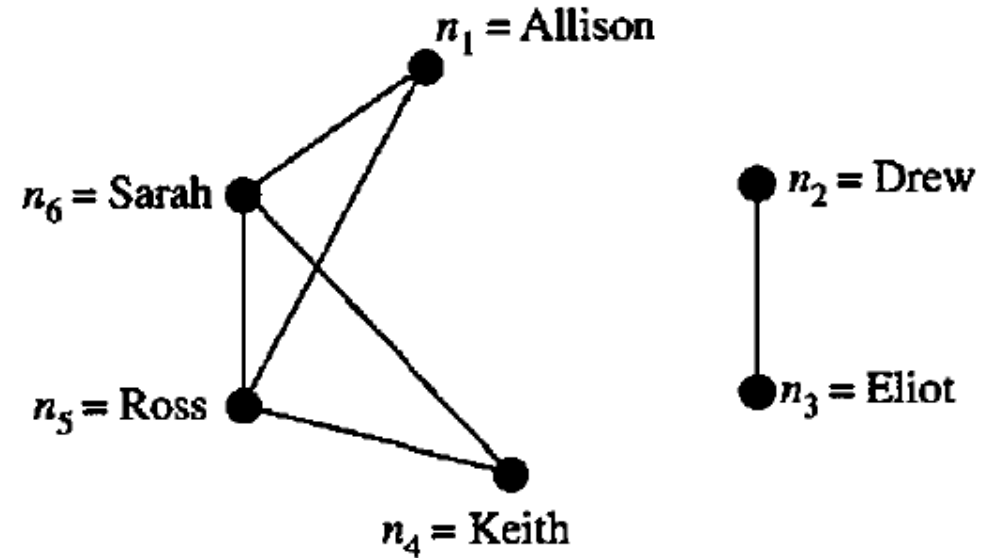
UCInet basics

- UCInet datasets
 - Each dataset consists of two files
 - Extension `##d` are the actual data
 - Extension `##h` contain information about the data
 - Many datasets are already stored in the folder Datafiles
- Import data
 - UCInet spreadsheet (manually or cut and paste)
 - From Excel
 - Data load format (DL)
 - See tutorials on the web

Basic Concepts of Graph Theory

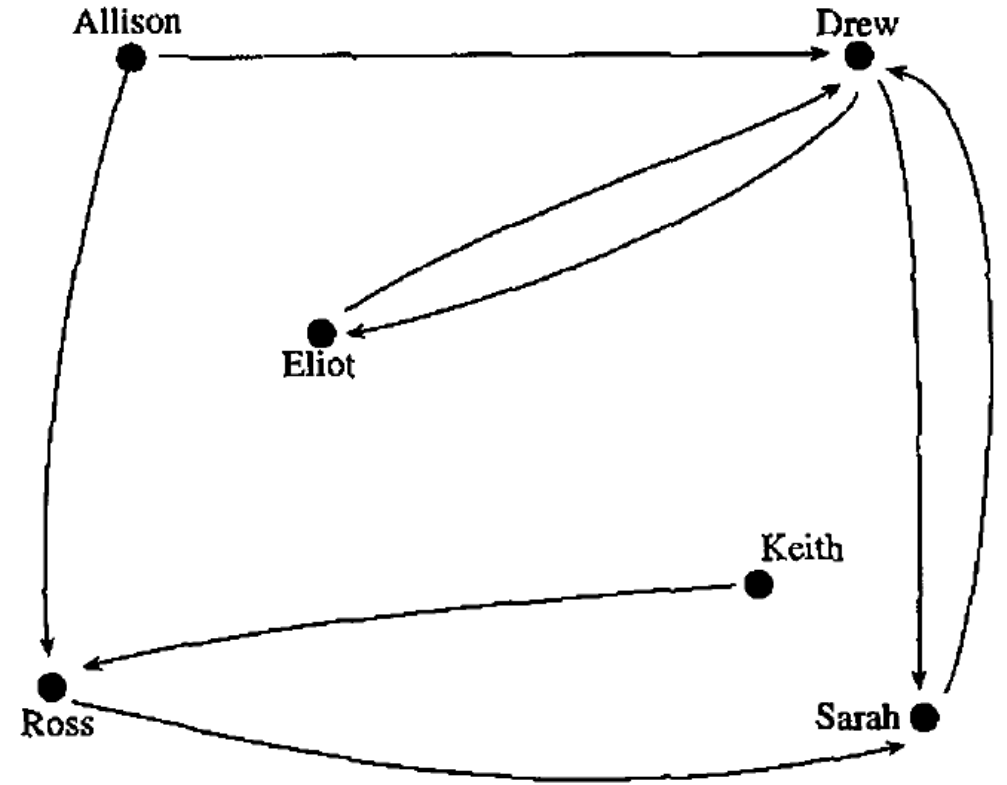
Graph

- Graph $G(V,E)$ consists of
 - Vertices V representing actors
 - Edges E representing ties
- Edge
 - Unordered pair of nodes (u,v)
 - Nodes u and v are adjacent if $(u,v) \in E$
 - E is a subset of all pairs of nodes (those which are tied)



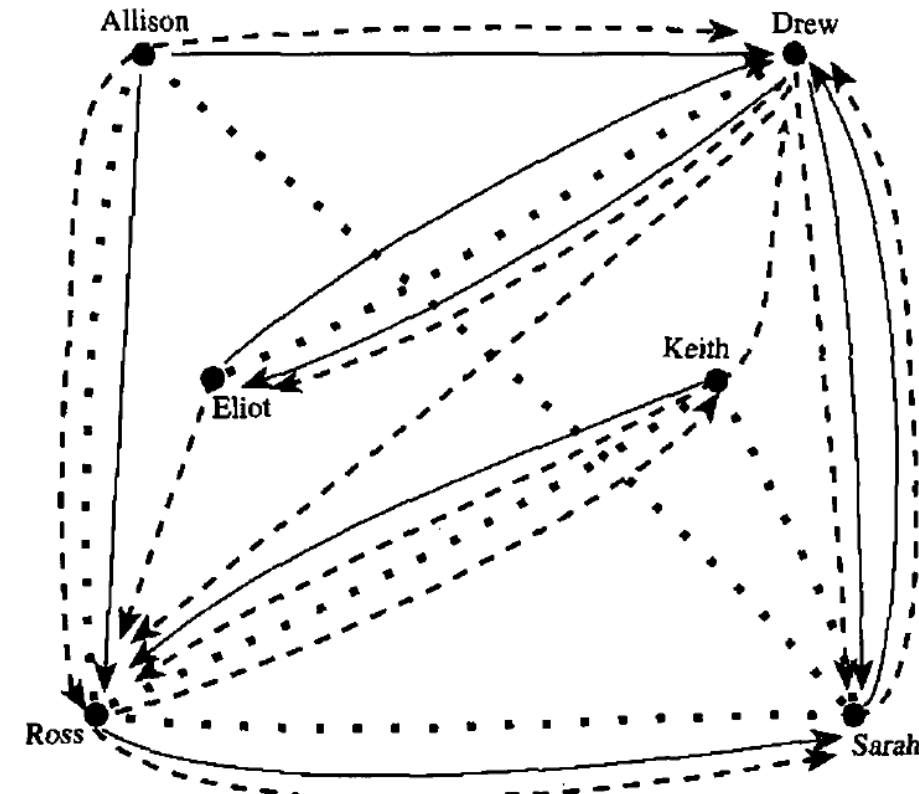
Directed graph

- Digraph $D(V,E)$ consists of
 - Vertices V representing actors
 - Arcs E representing ties
- Arc
 - Ordered pair of nodes (u,v)
 - $(u,v) \in E$ means that u sends arc to v
 - $(u,v) \in E$ does not imply that $(v,u) \in E$



Multiple relations

| Relation 1 Friendship at Beginning | Relation 2 Friendship at End | Relation 3 Lives Near |
|--|------------------------------------|-----------------------------|
| <Allison, Drew> | <Allison, Drew> | (Allison, Ross) |
| <Allison, Ross> | <Allison, Ross> | (Allison, Sarah) |
| <Drew, Sarah> | <Drew, Sarah> | (Drew, Eliot) |
| <Drew, Eliot> | <Drew, Eliot> | (Keith, Ross) |
| <Eliot, Drew> | <Drew, Ross> | (Keith, Sarah) |
| <Keith, Ross> | <Eliot, Ross> | (Ross, Sarah) |
| <Ross, Sarah> | <Keith, Drew> | |
| <Sarah, Drew> | <Keith, Ross> | |
| | <Ross, Keith> | |
| | <Ross, Sarah> | |
| | <Sarah, Drew> | |



Sociometric notation

Friendship at Beginning of Year

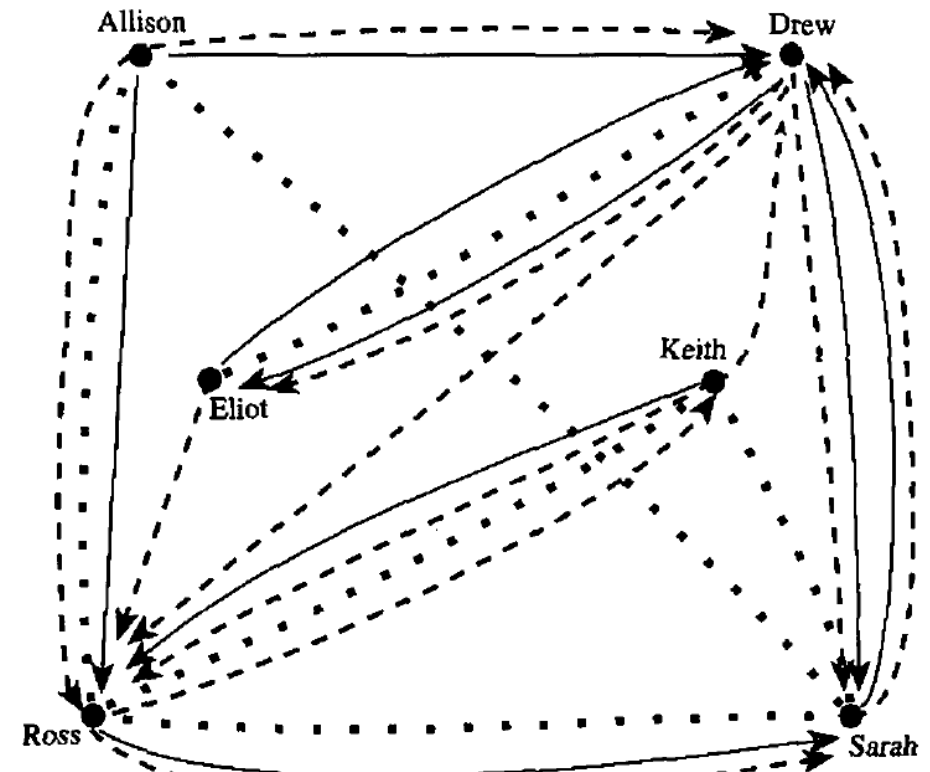
| | Allison | Drew | Eliot | Keith | Ross | Sarah |
|---------|---------|------|-------|-------|------|-------|
| Allison | - | 1 | 0 | 0 | 1 | 0 |
| Drew | 0 | - | 1 | 0 | 0 | 1 |
| Eliot | 0 | 1 | - | 0 | 0 | 0 |
| Keith | 0 | 0 | 0 | - | 1 | 0 |
| Ross | 0 | 0 | 0 | 0 | - | 1 |
| Sarah | 0 | 1 | 0 | 0 | 0 | - |

Friendship at End of Year

| | Allison | Drew | Eliot | Keith | Ross | Sarah |
|---------|---------|------|-------|-------|------|-------|
| Allison | - | 1 | 0 | 0 | 1 | 0 |
| Drew | 0 | - | 1 | 0 | 1 | 1 |
| Eliot | 0 | 0 | - | 0 | 1 | 0 |
| Keith | 0 | 1 | 0 | - | 1 | 0 |
| Ross | 0 | 0 | 0 | 1 | - | 1 |
| Sarah | 0 | 1 | 0 | 0 | 0 | - |

Lives Near

| | Allison | Drew | Eliot | Keith | Ross | Sarah |
|---------|---------|------|-------|-------|------|-------|
| Allison | - | 0 | 0 | 0 | 1 | 1 |
| Drew | 0 | - | 1 | 0 | 0 | 0 |
| Eliot | 0 | 1 | - | 0 | 0 | 0 |
| Keith | 0 | 0 | 0 | - | 1 | 1 |
| Ross | 1 | 0 | 0 | 1 | - | 1 |
| Sarah | 1 | 0 | 0 | 1 | 1 | - |



Valued ties

- We can assign values to ties
 - Strength of affective relationship
 - Frequency of contact
 - Amount of money
 - Probability of passing on information
- These are called valued graphs

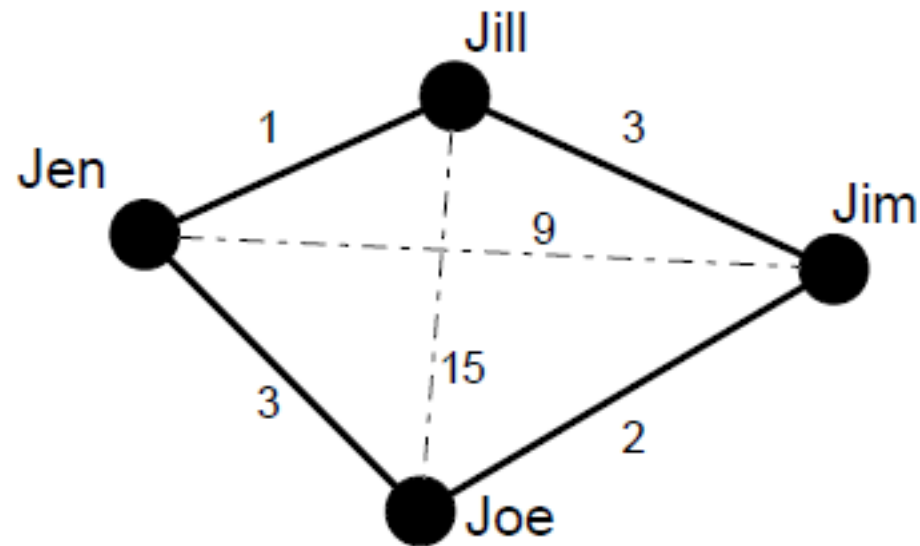
Adjacency matrices for binary and valued ties

Friendship

| | Jim | Jill | Jen | Joe |
|------|-----|------|-----|-----|
| Jim | - | 1 | 0 | 1 |
| Jill | 1 | - | 1 | 0 |
| Jen | 0 | 1 | - | 1 |
| Joe | 1 | 0 | 1 | - |

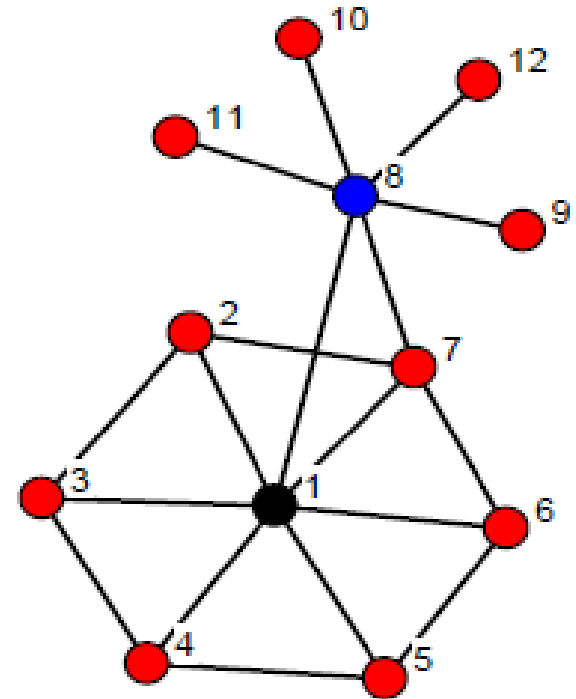
Proximity

| | Jim | Jill | Jen | Joe |
|------|-----|------|-----|-----|
| Jim | - | 3 | 9 | 2 |
| Jill | 3 | - | 1 | 15 |
| Jen | 9 | 1 | - | 3 |
| Joe | 2 | 15 | 3 | - |



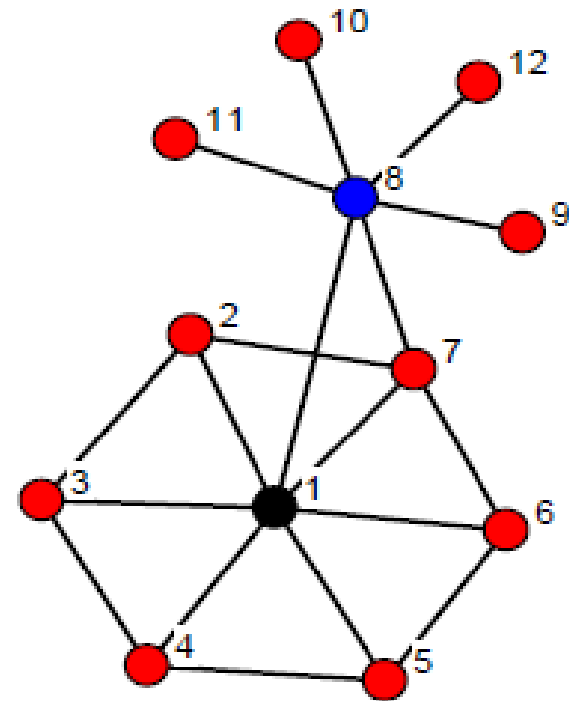
Degree

- The number of edges incident upon a node



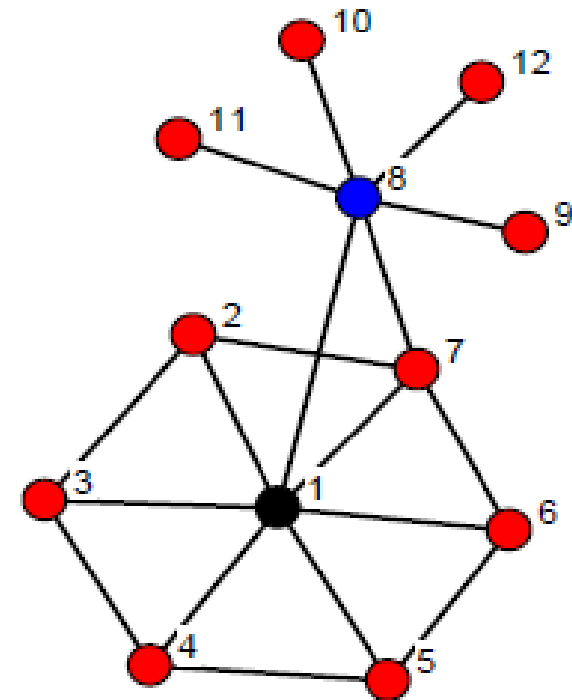
Walks, trails, and paths

- Walk: Unrestricted
 - The most general kind of sequence
 - 1-2-3-1-2-7-1-2
- Trail: Cannot repeat edge
 - Walk in which all edges are distinct
 - 1-2-3-1-7-8-10
- Path: Cannot repeat node
 - Walk in which all nodes and edges are distinct
 - 1-2-3-4-5-6-7-8



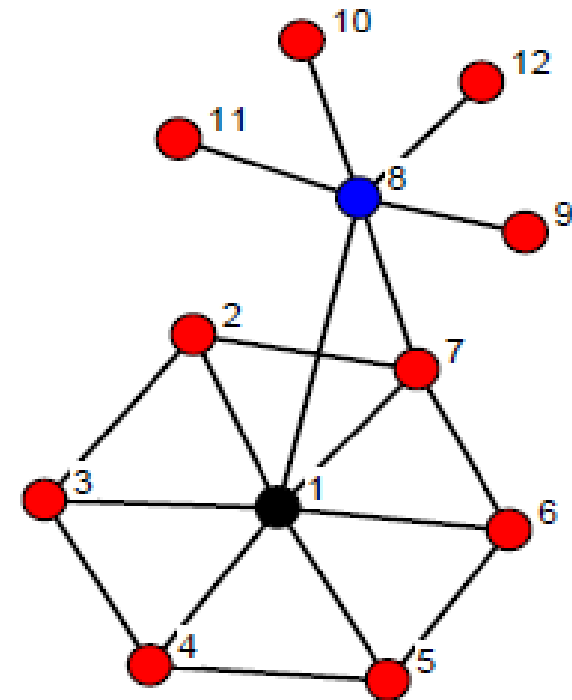
Walks, trails, and paths

- Walk: Unrestricted
 - The most general kind of sequence
 - 1-2-3-1-2-7-1-2
 - Example: ?
- Trail: Cannot repeat edge
 - Walk in which all edges are distinct
 - 1-2-3-1-7-8-10
 - Example: ?
- Path: Cannot repeat node
 - Walk in which all nodes and edges are distinct
 - 1-2-3-4-5-6-7-8
 - Example: ?

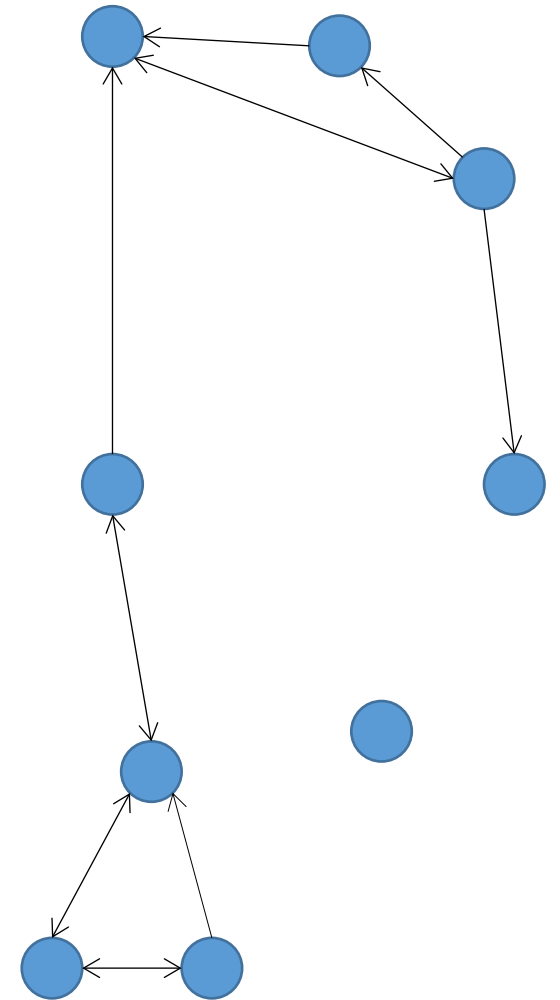


Walks, trails, and paths

- Walk: Unrestricted
 - The most general kind of sequence
 - 1-2-3-1-2-7-1-2
 - Example: Dollar bill
- Trail: Cannot repeat edge
 - Walk in which all edges are distinct
 - 1-2-3-1-7-8-10
 - Example: Gossip
- Path: Cannot repeat node
 - Walk in which all nodes and edges are distinct
 - 1-2-3-4-5-6-7-8
 - Example: Virus

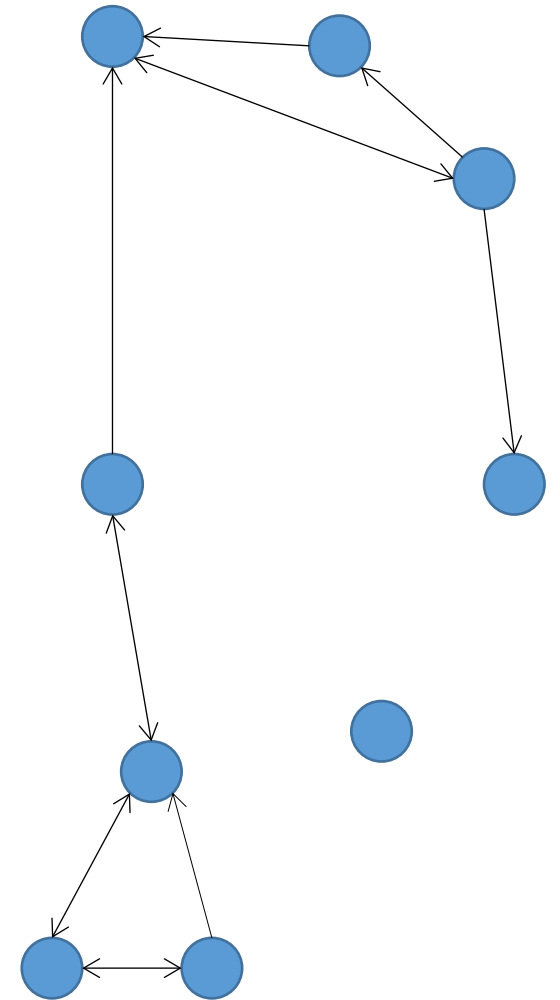


Indegree and Outdegree



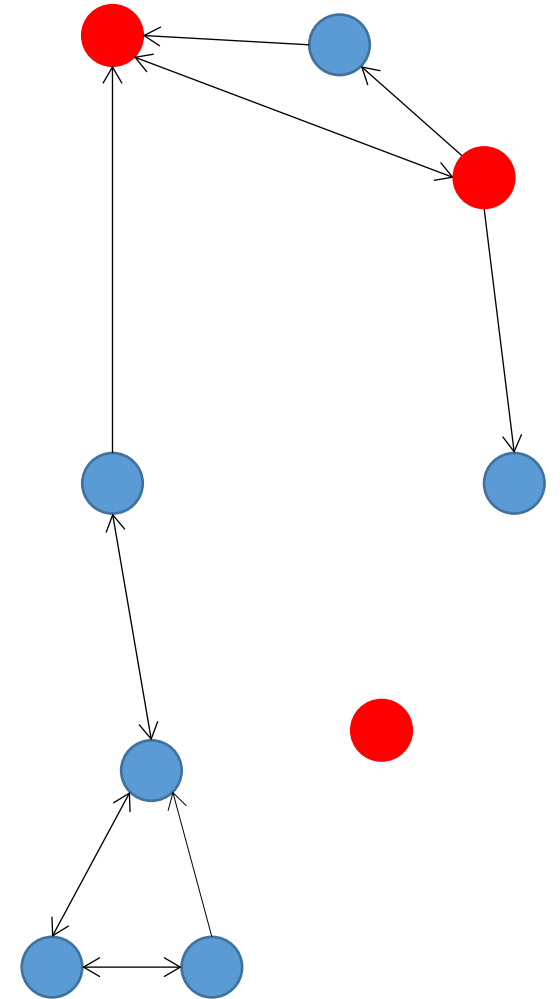
Indegree and Outdegree

- This applies to directed graphs only
- Indegree: Arcs that terminate at node
- Outdegree: Arcs that originate at node
- Average Indegree = Average Outdegree



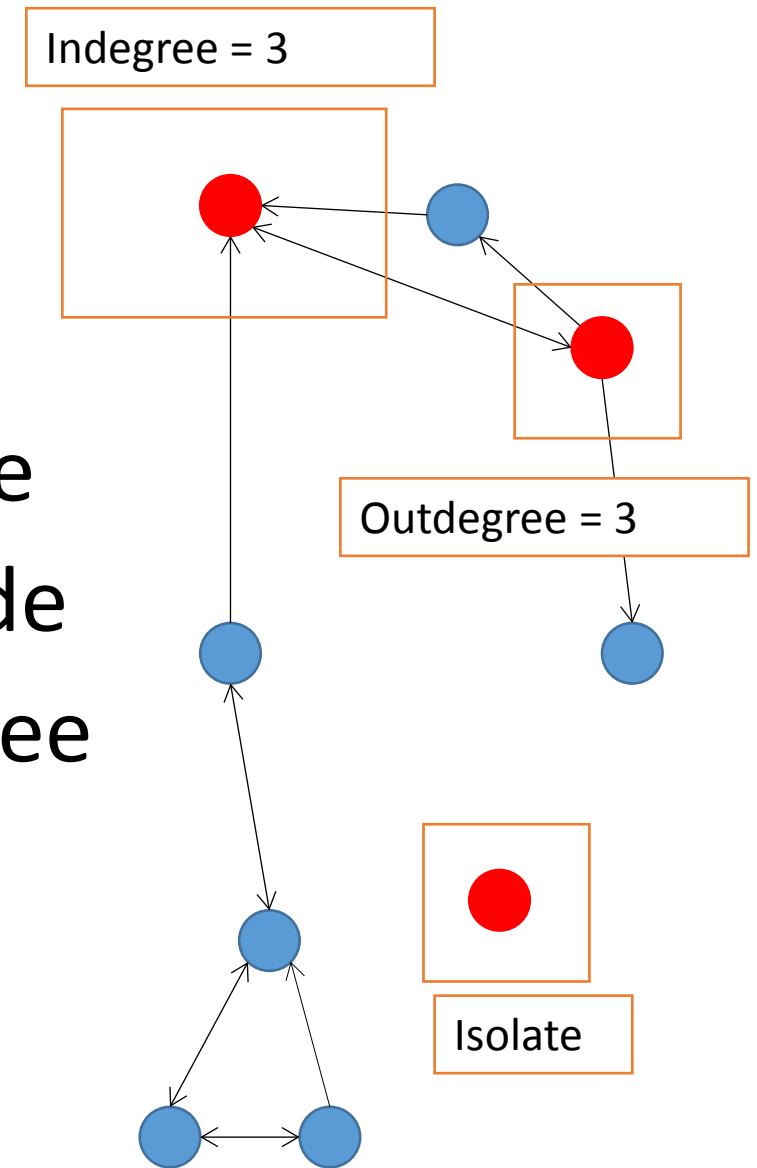
Indegree and Outdegree

- This applies to directed graphs only
- Indegree: Arcs that terminate at node
- Outdegree: Arcs that originate at node
- Average Indegree = Average Outdegree



Indegree and Outdegree

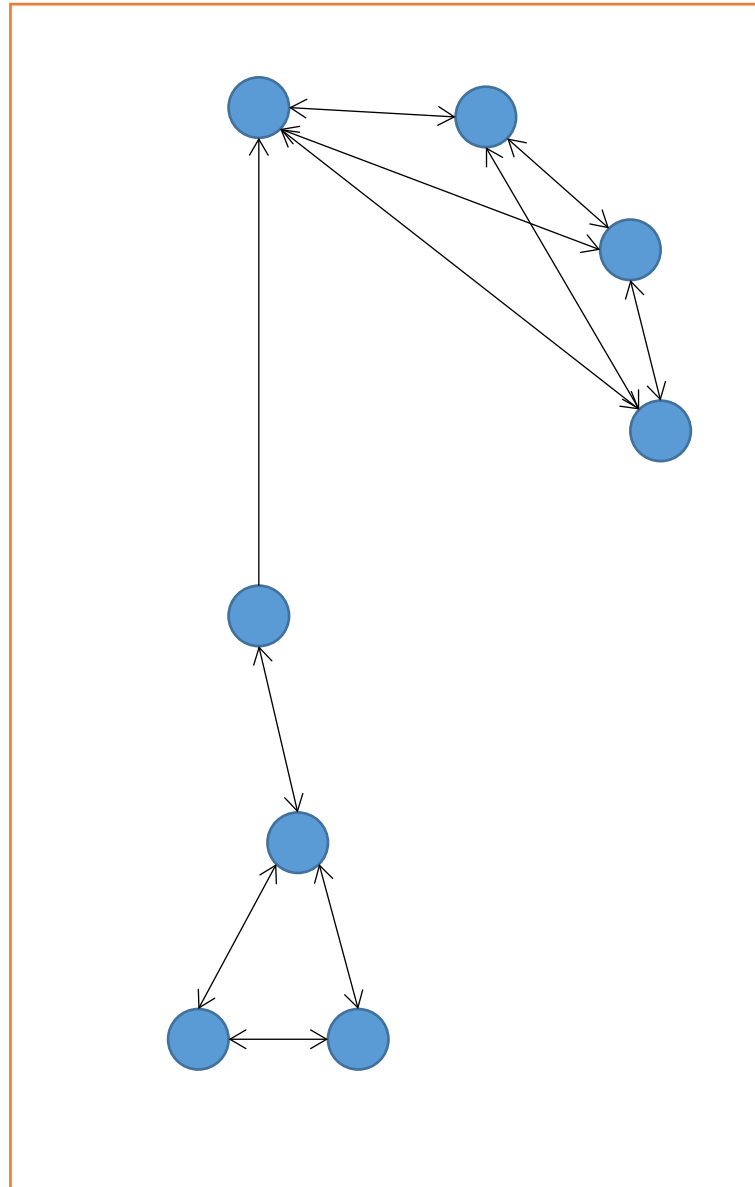
- This applies to directed graphs only
- Indegree: Arcs that terminate at node
- Outdegree: Arcs that originate at node
- Average Indegree = Average Outdegree



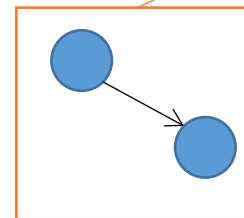
Component

Component

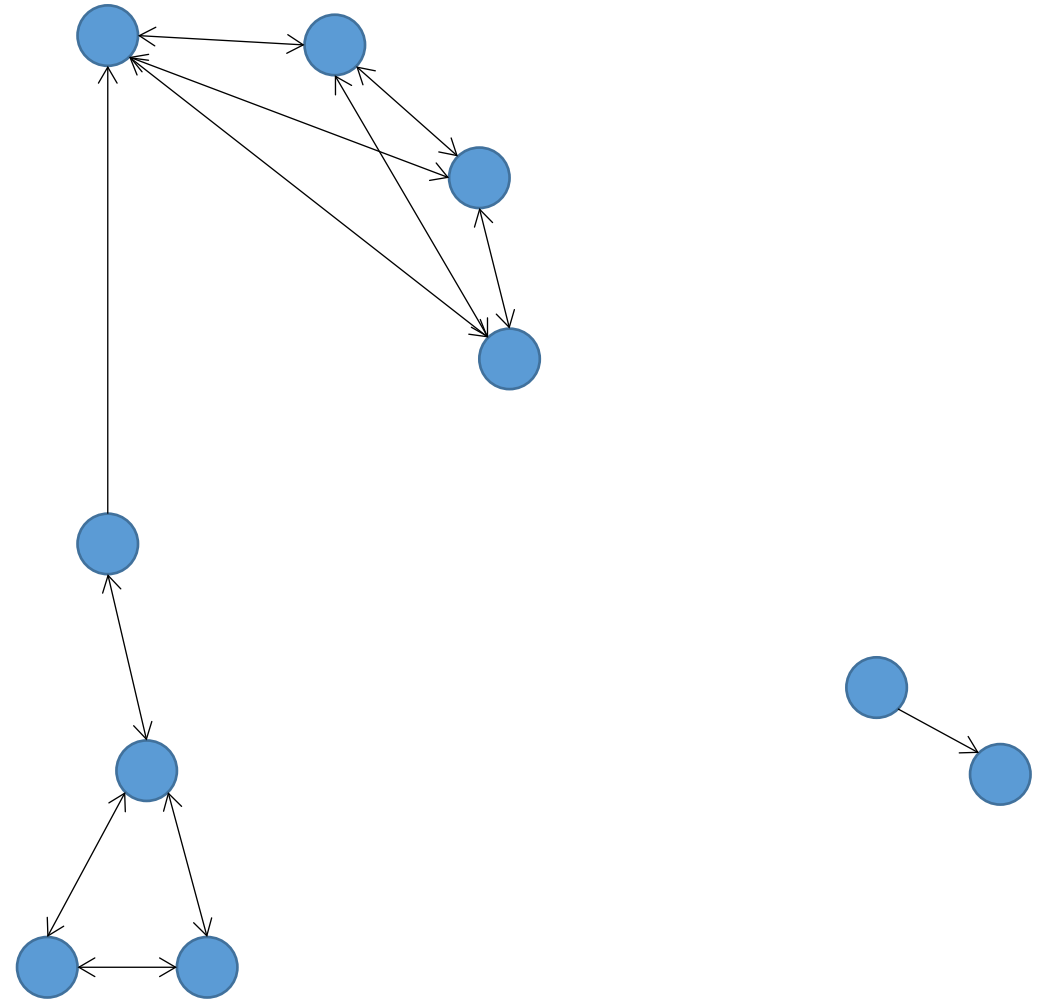
- Maximal set of nodes in which every node can reach every other node (no matter how long the path)
- If something flows through a certain type of line, it cannot flow between components
- Connected graph: One component
- Weak component: Every node reaches the other, ignoring direction
- Beware: Components do not define networks



Components

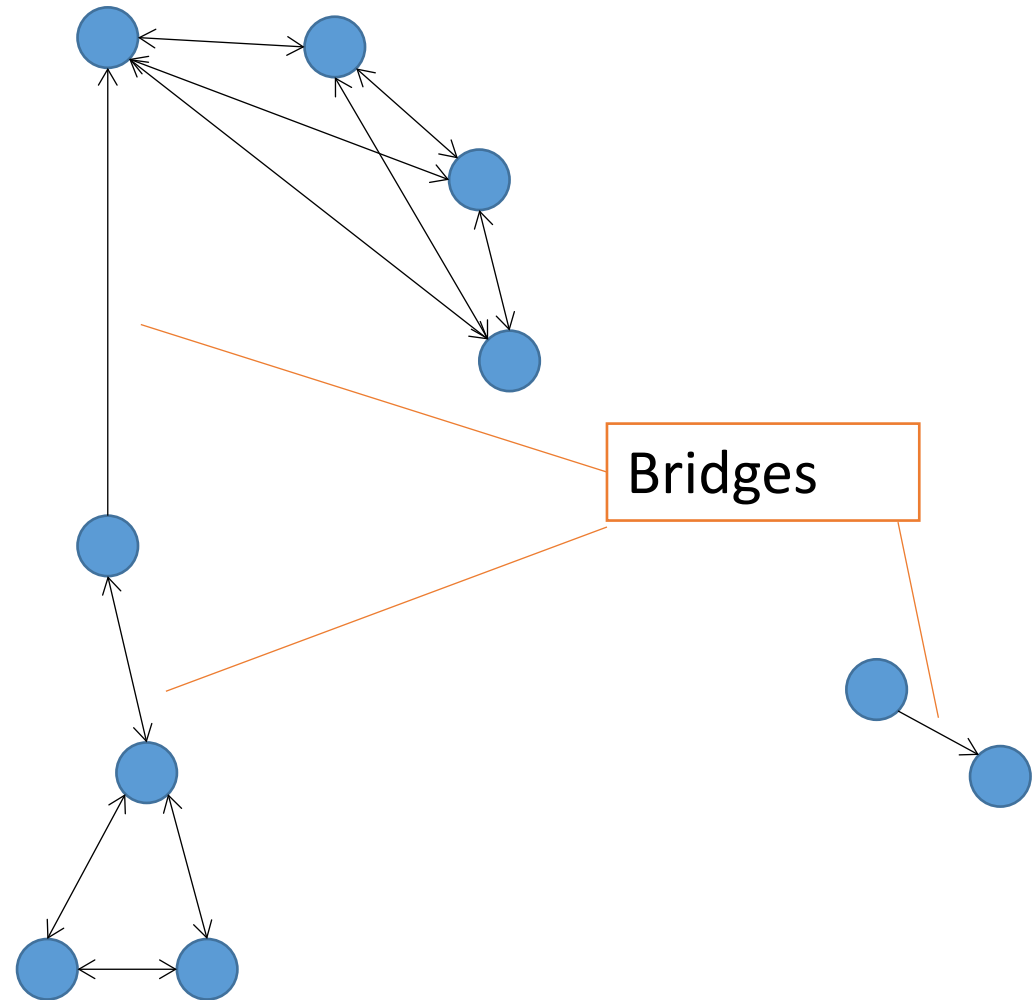


Bridge



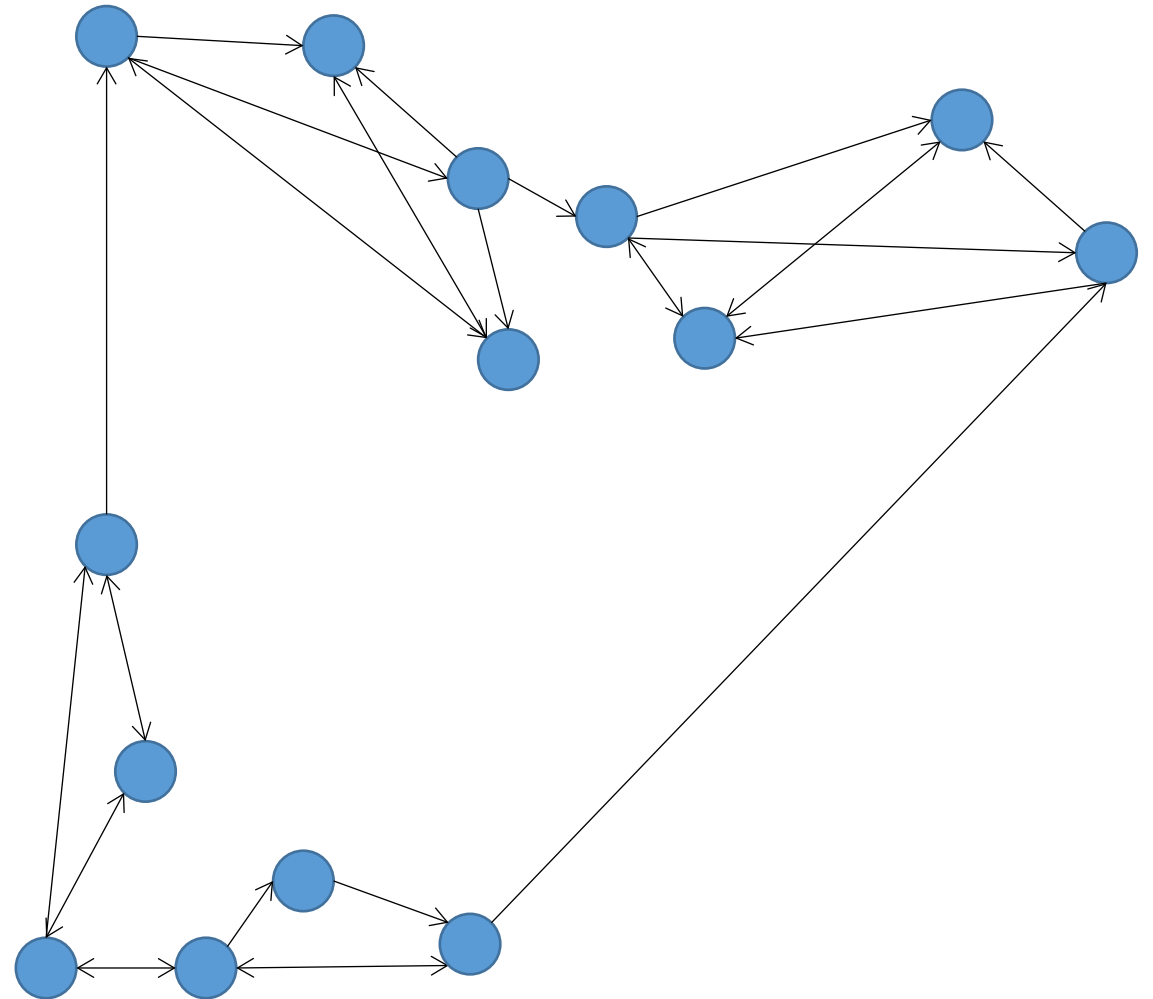
Bridge

- A tie that would increase the number of components if removed



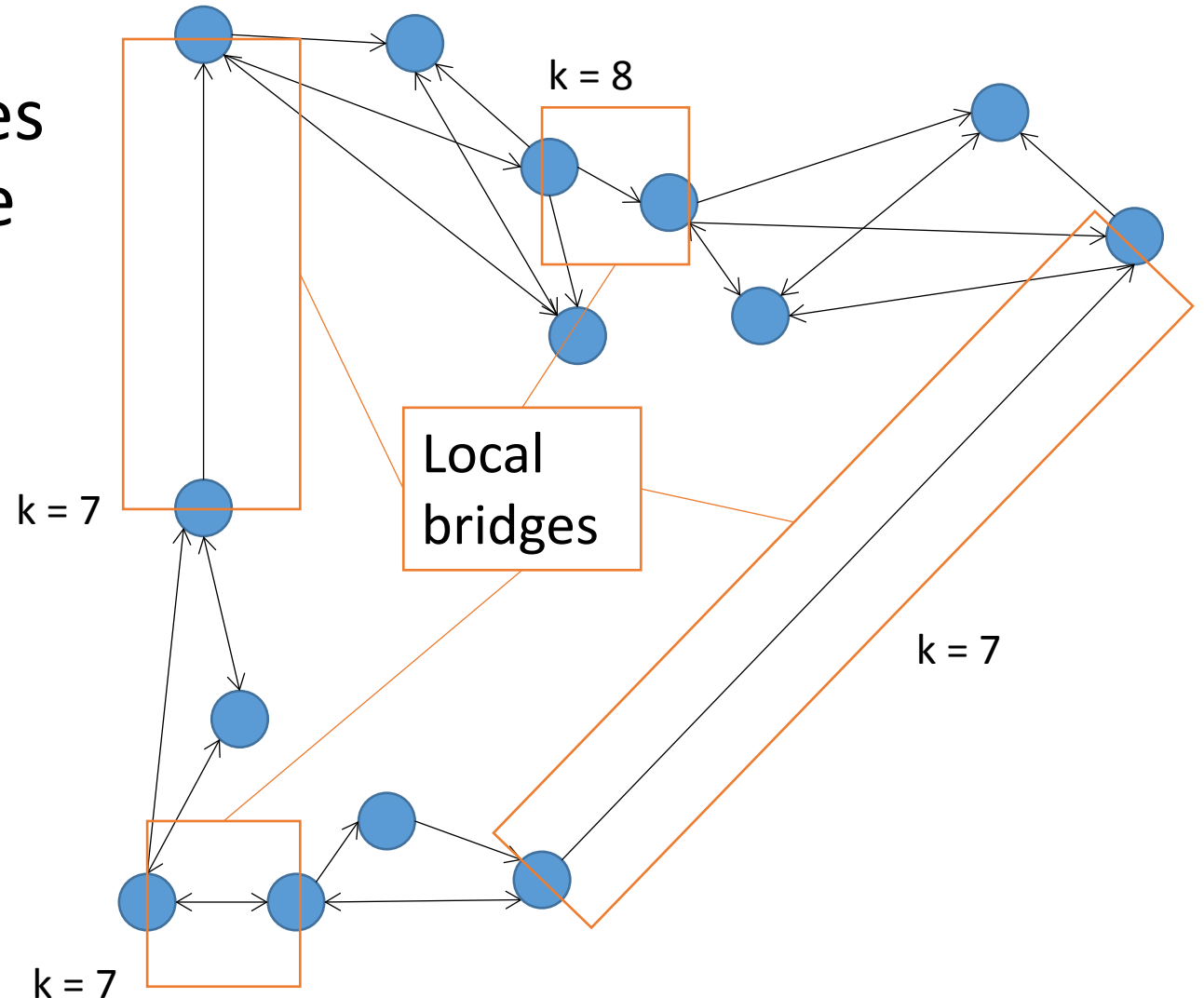
Local bridge of degree k

- A tie that connects nodes that would otherwise be at least k steps apart



Local bridge of degree k

- A tie that connects nodes that would otherwise be at least k steps apart



Cohesion

What is cohesion?

- Generally: How strongly connected a network is
- Why this that relevant? What happens in cohesive clusters?

What is cohesion?

- Generally: How strongly connected a network is
- Why this that relevant? What happens in cohesive clusters?
- Cohesive clusters
 - Enforce behavioral norms
 - Share and make sense of information
 - Develop an identity, ingroup-outgroup
 - Can coordinate for collective action
 - Share a common fate

Why care about cohesion?

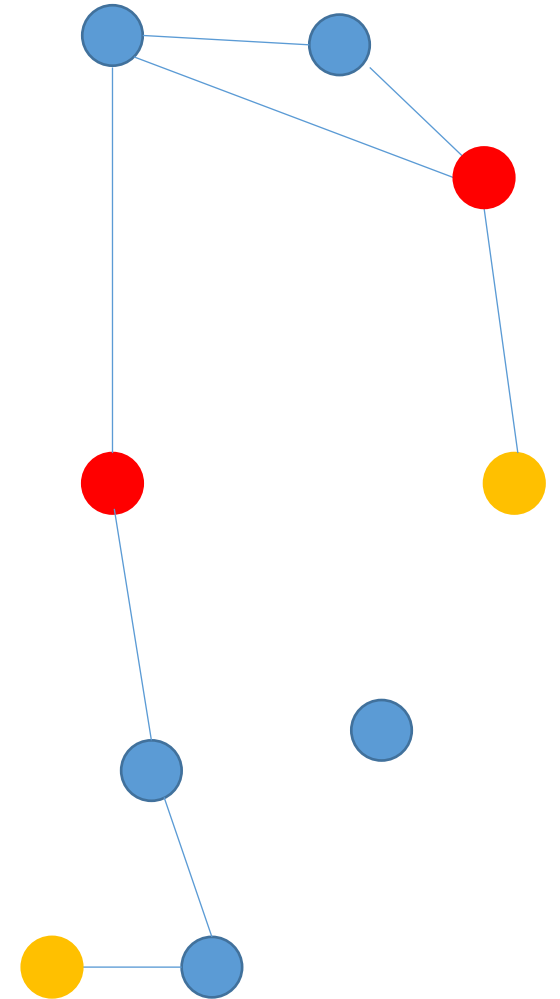
- Why do some communities, organizations, families perform certain functions better than others?
- How does a community's, family's, organization's internal structure affect its functions?
- Examples:
 - Charlestown resisting the urban renewal plan
 - Failure to establish a work council
 - Performance of R & D teams

Investigating cohesion: A family of measures

- Distance
- Density
- Reciprocity
- Transitivity
- (Non-)Fragmentation
- ...

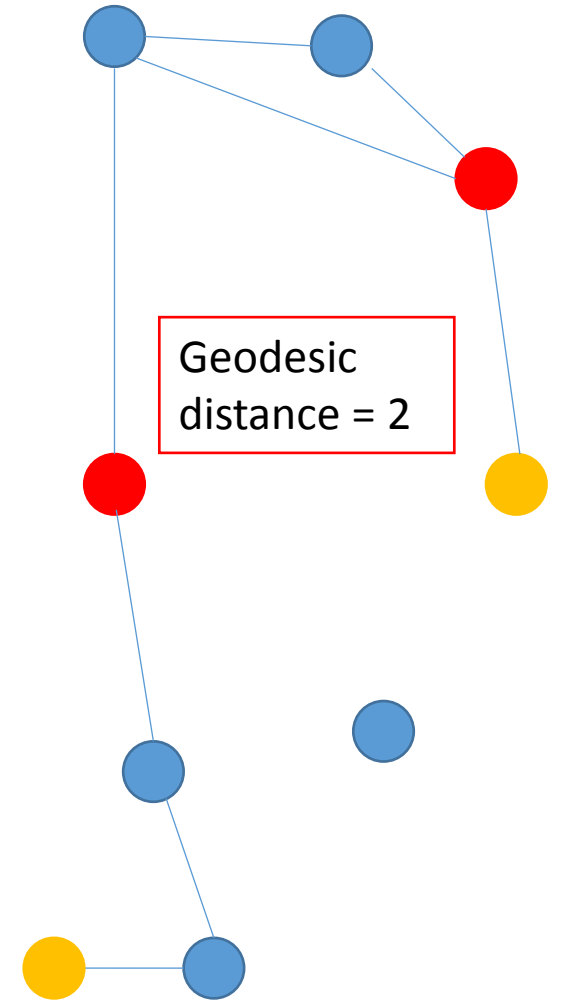
Distance

- The shortest path between two nodes is the geodesic distance
- The longest shortest path in a network is the diameter
- Remember: Six degrees of separation
→ Geodesic distance is a measure of this



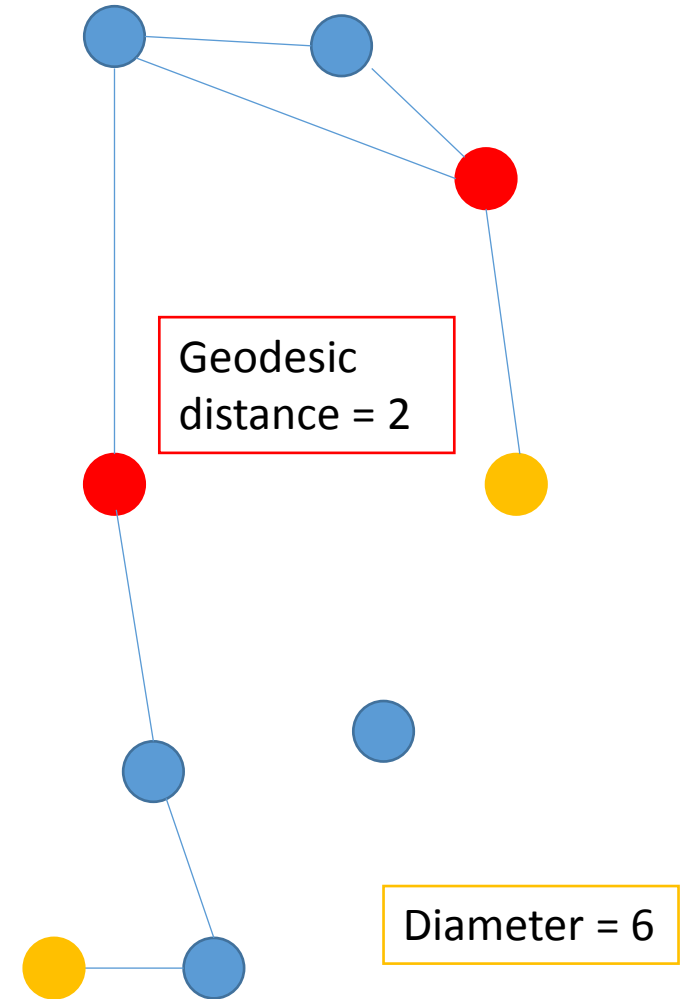
Distance

- The shortest path between two nodes is the geodesic distance
- The longest shortest path in a network is the diameter
- Remember: Six degrees of separation
→ Geodesic distance is a measure of this



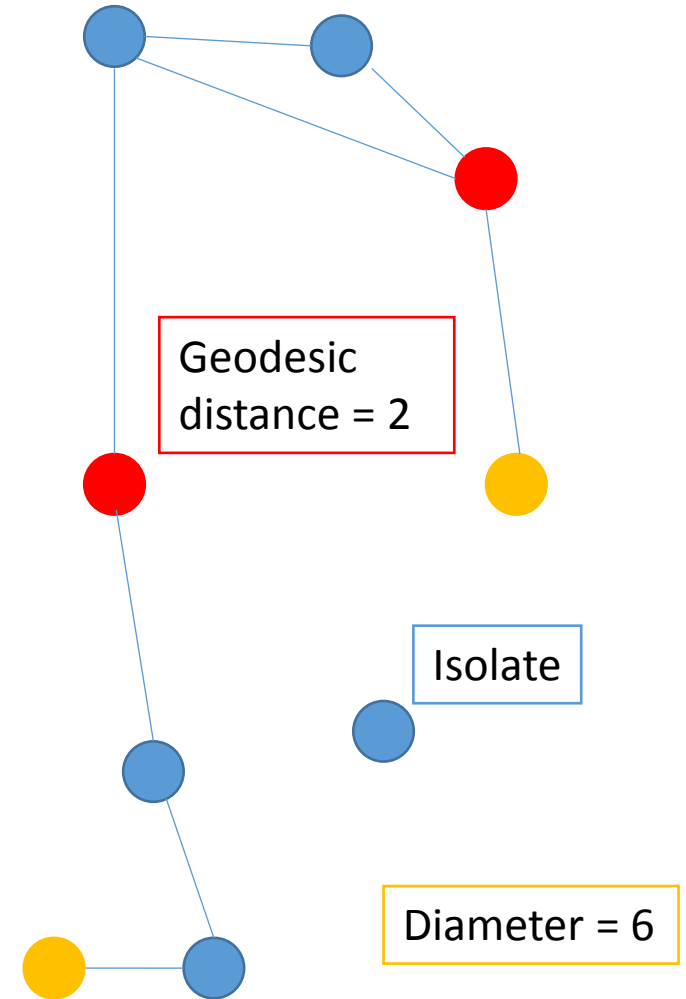
Distance

- The shortest path between two nodes is the geodesic distance
- The longest shortest path in a network is the diameter
- Remember: Six degrees of separation
→ Geodesic distance is a measure of this



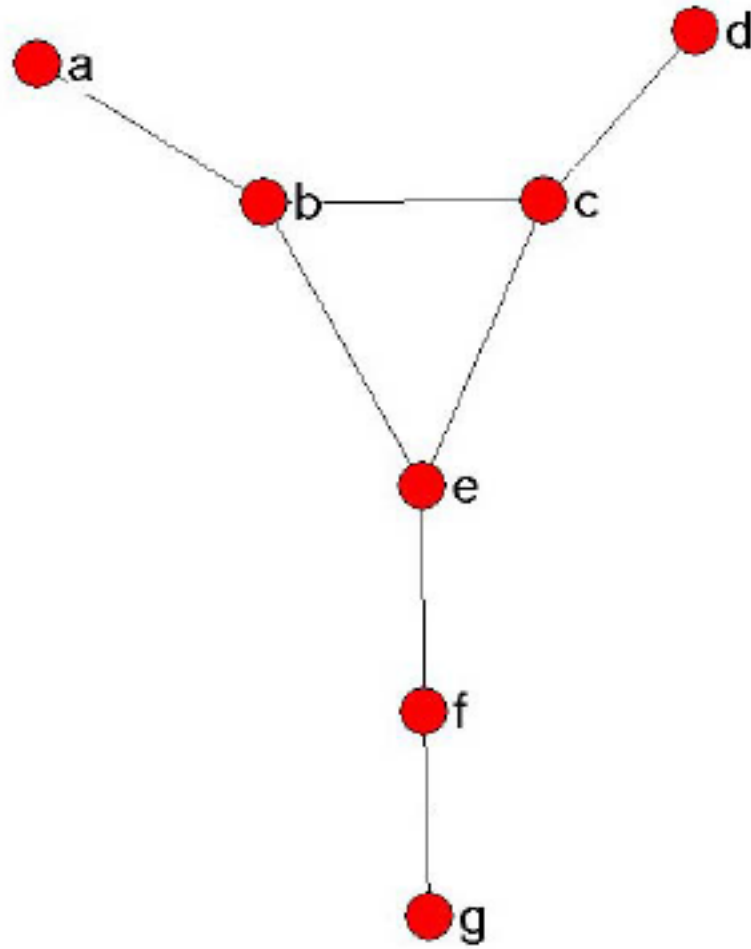
Distance

- The shortest path between two nodes is the geodesic distance
- The longest shortest path in a network is the diameter
- Remember: Six degrees of separation
→ Geodesic distance is a measure of this



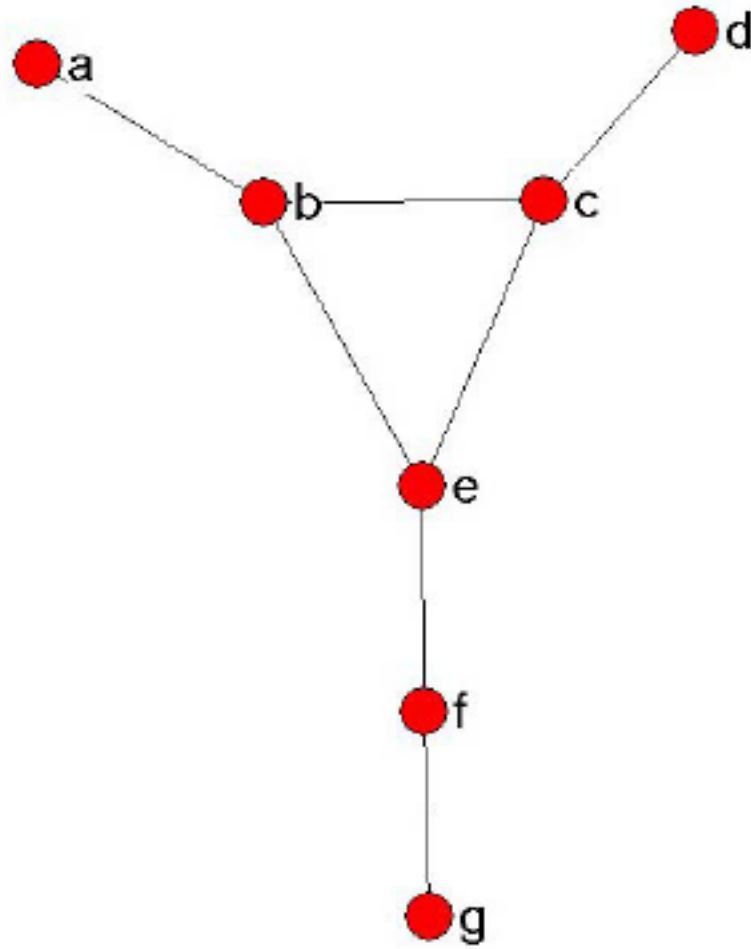
Geodesic distance matrix

| | a | b | c | d | e | f | g |
|---|---|---|---|---|---|---|---|
| a | | | | | | | |
| b | | | | | | | |
| c | | | | | | | |
| d | | | | | | | |
| e | | | | | | | |
| f | | | | | | | |
| g | | | | | | | |



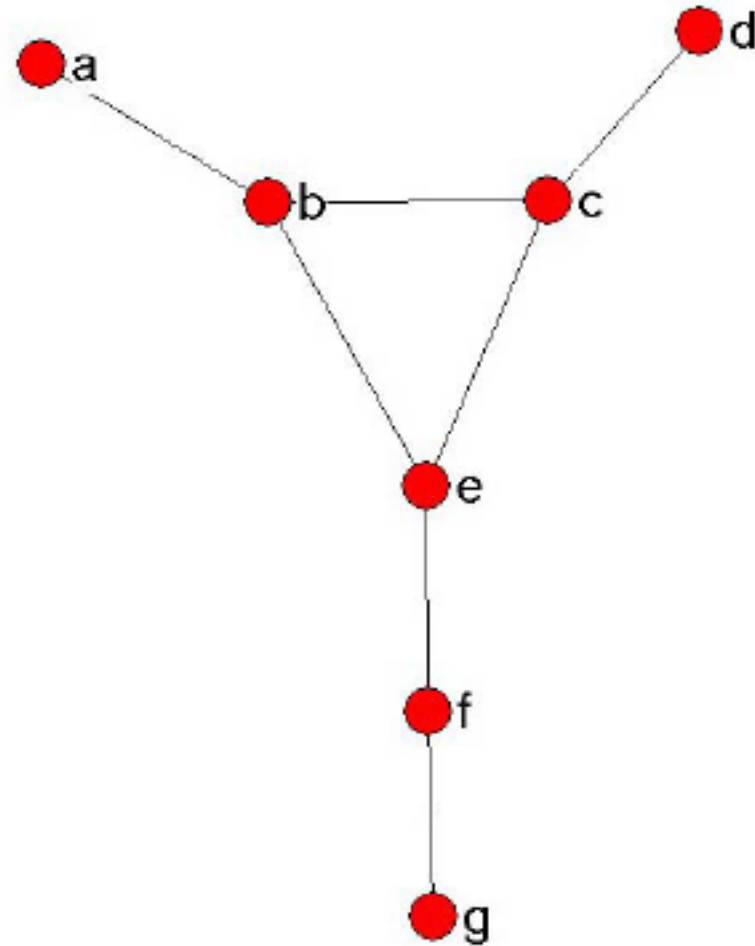
Geodesic distance matrix

| | a | b | c | d | e | f | g |
|---|---|---|---|---|---|---|---|
| a | 0 | 1 | 2 | 3 | 2 | 3 | 4 |
| b | | | | | | | |
| c | | | | | | | |
| d | | | | | | | |
| e | | | | | | | |
| f | | | | | | | |
| g | | | | | | | |



Geodesic distance matrix

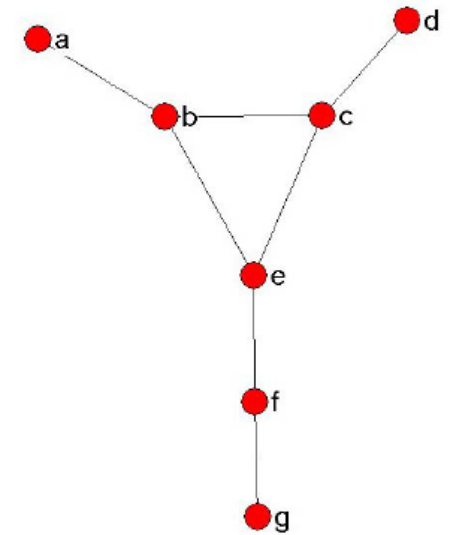
| | a | b | c | d | e | f | g |
|---|---|---|---|---|---|---|---|
| a | 0 | 1 | 2 | 3 | 2 | 3 | 4 |
| b | 1 | 0 | 1 | 2 | 1 | 2 | 3 |
| c | 2 | 1 | 0 | 1 | 1 | 2 | 3 |
| d | 3 | 2 | 1 | 0 | 2 | 3 | 4 |
| e | 2 | 1 | 1 | 2 | 0 | 1 | 2 |
| f | 3 | 2 | 2 | 3 | 1 | 0 | 1 |
| g | 4 | 3 | 3 | 4 | 2 | 1 | 0 |



Geodesic versus adjacency matrix

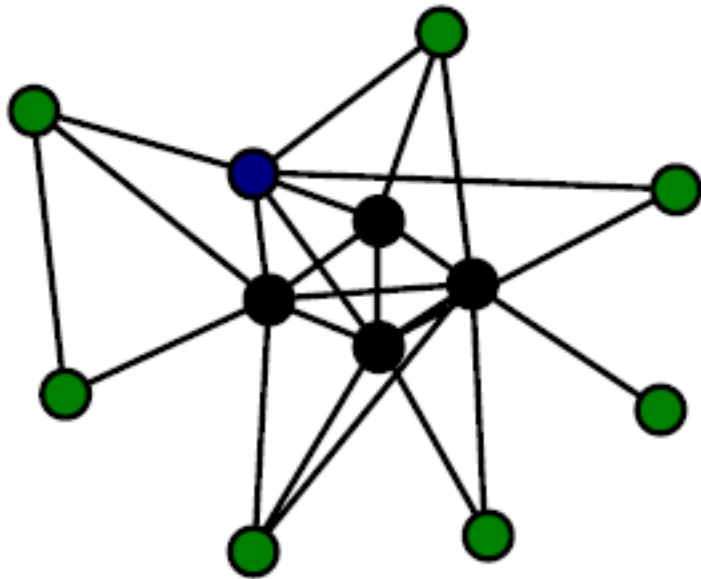
| | a | b | c | d | e | f | g |
|---|---|---|---|---|---|---|---|
| a | 0 | 1 | 2 | 3 | 2 | 3 | 4 |
| b | 1 | 0 | 1 | 2 | 1 | 2 | 3 |
| c | 2 | 1 | 0 | 1 | 1 | 2 | 3 |
| d | 3 | 2 | 1 | 0 | 2 | 3 | 4 |
| e | 2 | 1 | 1 | 2 | 0 | 1 | 2 |
| f | 3 | 2 | 2 | 3 | 1 | 0 | 1 |
| g | 4 | 3 | 3 | 4 | 2 | 1 | 0 |

| | a | b | c | d | e | f | g |
|---|---|---|---|---|---|---|---|
| a | 0 | 1 | 0 | 0 | 0 | 0 | 0 |
| b | 1 | 0 | 1 | 0 | 1 | 0 | 0 |
| c | 0 | 1 | 0 | 1 | 1 | 0 | 0 |
| d | 0 | 0 | 1 | 0 | 0 | 0 | 0 |
| e | 0 | 1 | 1 | 0 | 0 | 1 | 0 |
| f | 0 | 0 | 0 | 0 | 1 | 0 | 1 |
| g | 0 | 0 | 0 | 0 | 0 | 1 | 0 |

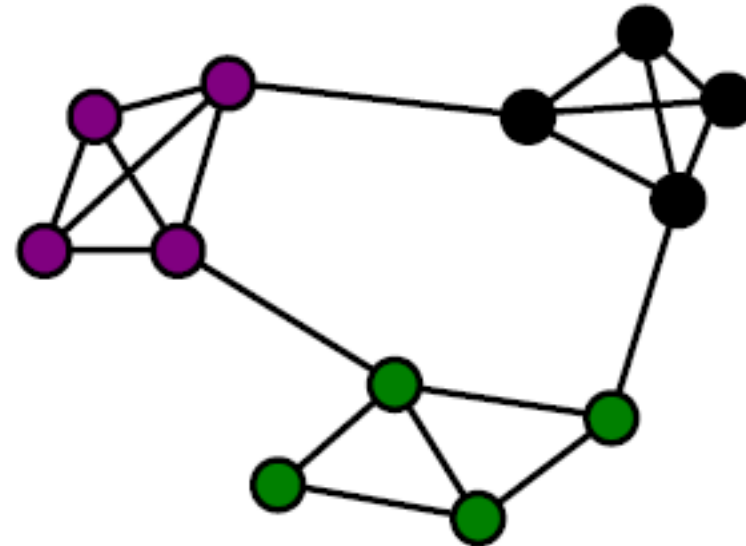


Average distance

- Average geodesic distance between all pairs of nodes



Core/Periphery
c/p fit = 0.97, avg. dist. = 1.9



Clique structure
c/p fit = 0.33, avg. dist. = 2.4

Distance in UCInet

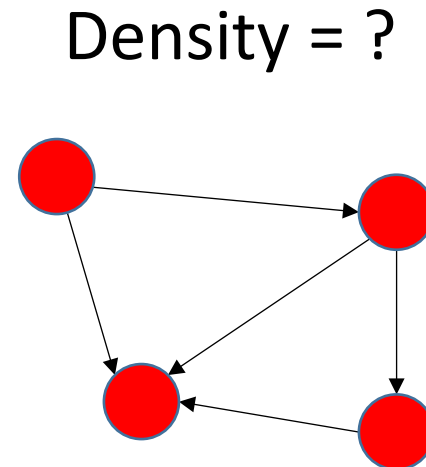
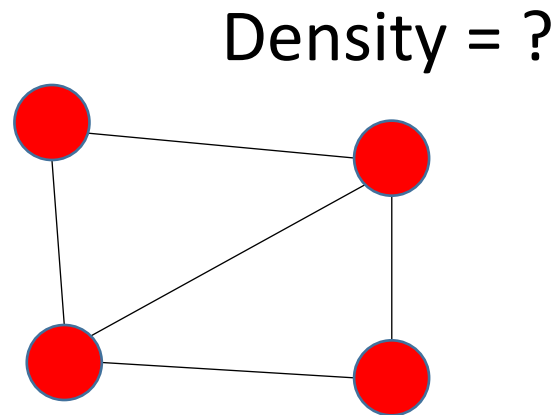
Density

Density

- A measure for the connectedness of a network
- The sum of ties divided by the sum of possible ties
- Undirected graph: Number of possible ties = $n*(n-1) / 2$
 - The relationship A to B is the same as the relationship B to A
- Digraph: twice as many possible edges: $n*(n-1)$
 - The relationship A to B is not the same as the relationship B to A

Density

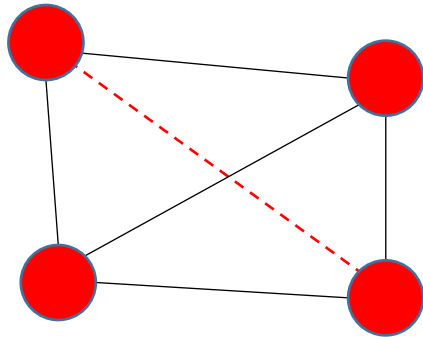
- A measure for the connectedness of a network
- The sum of ties divided by the sum of possible ties
- Undirected graph: Number of possible ties = $n*(n-1) / 2$
 - The relationship A to B is the same as the relationship B to A
- Digraph: twice as many possible edges: $n*(n-1)$
 - The relationship A to B is not the same as the relationship B to A



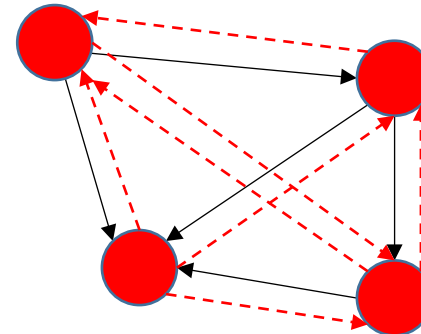
Density

- A measure for the connectedness of a network
- The sum of ties divided by the sum of possible ties
- Undirected graph: Number of possible ties = $n*(n-1) / 2$
 - The relationship A to B is the same as the relationship B to A
- Digraph: twice as many possible edges: $n*(n-1)$
 - The relationship A to B is not the same as the relationship B to A

Density = $5/6 = 0.83$



Density = $5/12 = 0.42$



Density in reflexive and non-reflexive ties

| | Reflexive | Non-Reflexive |
|------------|-----------------------|--------------------------|
| Undirected | $= \frac{T}{n^2 / 2}$ | $= \frac{T}{n(n-1) / 2}$ |
| Directed | $= \frac{T}{n^2}$ | $= \frac{T}{n(n-1)}$ |

Density

- What does it mean socially if a network is dense?

Density

- What does it mean socially if a network is dense?
 - High level of social capital
 - High degree of social control
 - High speed of information flow

Density in UCInet