

# Cohesion

## Relational and Group

# Relational vs Group

- **Relational or dyadic cohesion** refers to pairwise social closeness
- **Network cohesion** refers to the cohesion of an entire group

# Ways to Approach This

- Many ways to define or theoretically conceive of cohesion
  - cohesion → outcome
  - What is the mechanism that would relate cohesion to the outcome of interest?
  - Define cohesion consistent with this mechanism
- For each way, we can then devise an operational measurement
  - Don't confuse the measure with the construct

# Adjacency & Strength of Tie

- Raw dyadic data
- Positive ties
- Guttman scale of social closeness or obligation
- Valued relations
  - Frequency of interaction
  - Duration of relation
  - Intensity

# Multiplexity

- Multiplexity is often what is meant by “relational embeddedness”
  - As in economic ties being embedded in social ties
- Combination of (the right set of) ties can be seen as yielding greater closeness than just one tie

# Embedded Ties

- A tie  $(u,v)$  is structurally embedded if there exists node  $p$  (possibly several nodes) such that  $(u,p) \in E$  and  $(v,p) \in E$ 
  - I.e, then endpoints  $u$  and  $v$  have “friends” in common

# Simmelian Ties

- Krackhardt's definition:
- A dyad has a simmelian tie if it is reciprocal ties to each other and to third parties
- The value of a simmelian tie is the number of third parties they have in common
  - Ideally, it is the number of cliques they have in common

# Reachability

- If there exists a path from  $u$  to  $v$  of any length, then  $v$  is said to be reachable from  $u$
- The reachability matrix  $R$  in which  $r_{ij} = 1$  if I can reach  $j$  records the relational cohesions in the graph
- Is a weak form of cohesion – minimal in fact
- Can define a weak form of Simmelian ties on the reachability graph



# Geodesic Distance

Adjacency

	a	b	c	d	e	f	g	h	i	j
a	0	1	1	1	0	0	0	0	0	0
b	1	0	1	0	1	0	0	0	0	0
c	1	1	0	1	0	0	0	0	0	0
d	1	0	1	0	1	0	0	0	0	0
e	0	1	0	1	0	1	0	0	0	0
f	0	0	0	0	1	0	1	0	1	0
g	0	0	0	0	0	1	0	1	0	1
h	0	0	0	0	0	0	1	0	1	1
i	0	0	0	0	0	1	0	1	0	1
j	0	0	0	0	0	0	1	1	1	0

Geodesic Distance

	a	b	c	d	e	f	g	h	i	j
a	0	1	1	1	2	3	4	5	4	5
b	1	0	1	2	1	2	3	4	3	4
c	1	1	0	1	2	3	4	5	4	5
d	1	2	1	0	1	2	3	4	3	4
e	2	1	2	1	0	1	2	3	2	3
f	3	2	3	2	1	0	1	2	1	2
g	4	3	4	3	2	1	0	1	2	1
h	5	4	5	4	3	2	1	0	1	1
i	4	3	4	3	2	1	2	1	0	1
j	5	4	5	4	3	2	1	1	1	0

More nuance in the representation of non-connection

# Reciprocal Distance

	a	b	c	d	e	f	g	h	i	j
a	0.00	1.00	1.00	1.00	0.50	0.33	0.25	0.20	0.25	0.20
b	1.00	0.00	1.00	0.50	1.00	0.50	0.33	0.25	0.33	0.25
c	1.00	1.00	0.00	1.00	0.50	0.33	0.25	0.20	0.25	0.20
d	1.00	0.50	1.00	0.00	1.00	0.50	0.33	0.25	0.33	0.25
e	0.50	1.00	0.50	1.00	0.00	1.00	0.50	0.33	0.50	0.33
f	0.33	0.50	0.33	0.50	1.00	0.00	1.00	0.50	1.00	0.50
g	0.25	0.33	0.25	0.33	0.50	1.00	0.00	1.00	0.50	1.00
h	0.20	0.25	0.20	0.25	0.33	0.50	1.00	0.00	1.00	1.00
i	0.25	0.33	0.25	0.33	0.50	1.00	0.50	1.00	0.00	1.00
j	0.20	0.25	0.20	0.25	0.33	0.50	1.00	1.00	1.00	0.00

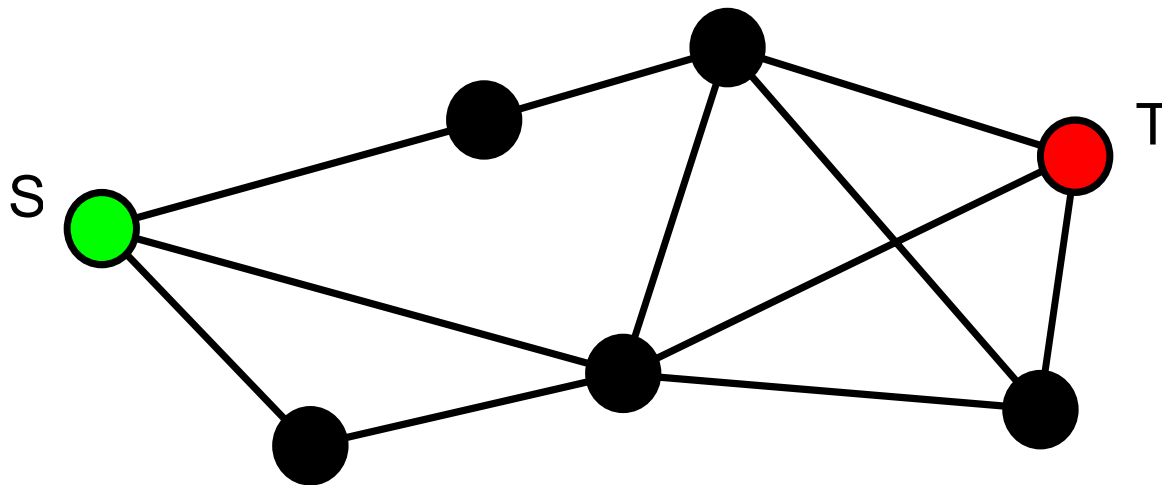
# Number of Walks\*

		1	2	3	4	5	6	7	8	9	10
		a	b	c	d	e	f	g	h	i	j
		---	---	---	---	---	---	---	---	---	---
1	a	194	167	195	167	154	50	30	12	30	12
2	b	167	188	167	188	115	82	22	30	22	30
3	c	195	167	194	167	154	50	30	12	30	12
4	d	167	188	167	188	115	82	22	30	22	30
5	e	154	115	154	115	150	59	82	50	82	50
6	f	50	82	50	82	59	150	115	154	115	154
7	g	30	22	30	22	82	115	188	167	188	167
8	h	12	30	12	30	50	154	167	194	167	195
9	i	30	22	30	22	82	115	188	167	188	167
10	j	12	30	12	30	50	154	167	195	167	194

\*Of length of length 6 or less

# Independent Paths

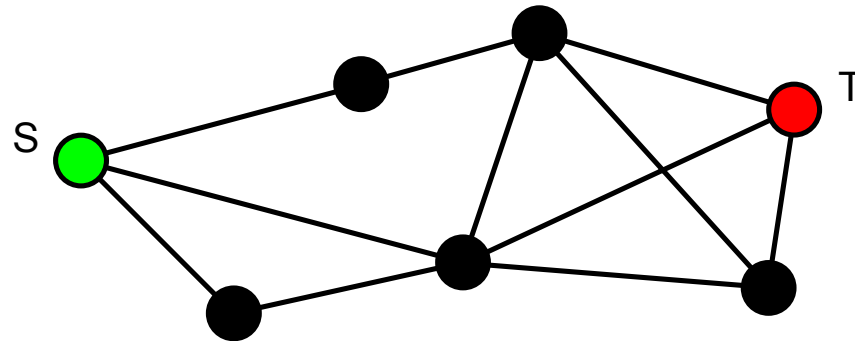
- A set of paths is node-independent if they share no nodes (except beginning and end)
  - They are line-independent if they share no lines



- 2 node-independent paths from S to T
- 3 line-independent paths from S to T

# Connectivity

- Line connectivity  
 $\lambda(s,t)$  is the minimum number of lines that must be removed to disconnect  $s$  from  $t$
- Node connectivity  
 $\kappa(s,t)$  is minimum number of nodes that must be removed to disconnect  $s$  from  $t$

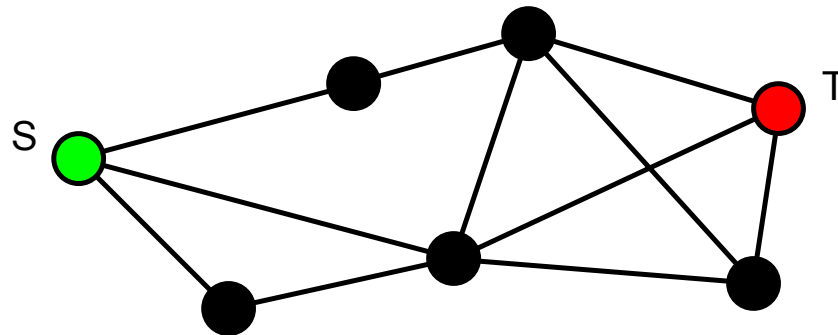


# Menger's Theorem

- Menger proved that the number of line independent paths between  $s$  and  $t$  equals the line connectivity  $\lambda(s,t)$
- And the number of node-independent paths between  $s$  and  $t$  equals the node connectivity  $\kappa(u,v)$

# Maximum Flow

- If ties are pipes with capacity of 1 unit of flow, what is the maximum # of units that can flow from s to t?
- Ford & Fulkerson show this was equal to the number of line-independent paths



# Group Cohesion

- Whole network measures can be
  - Averages of dyadic cohesion
  - Measures not easily reducible to dyadic measures

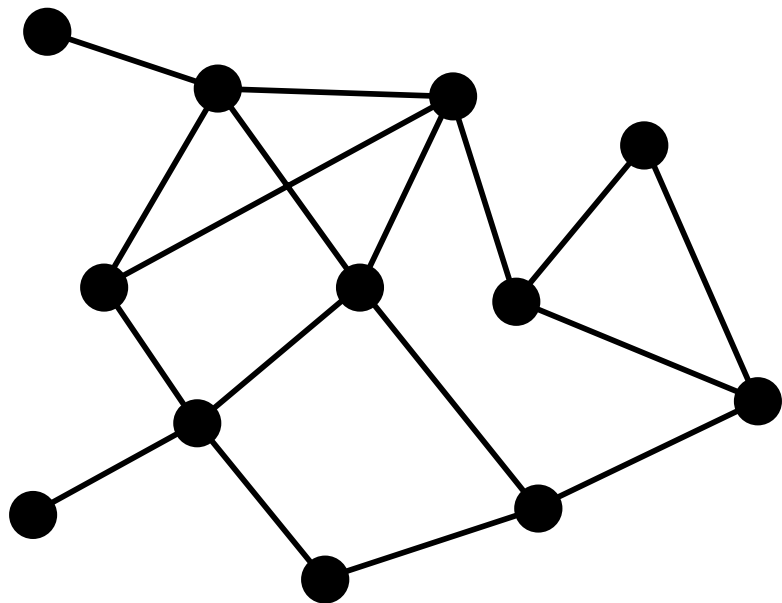


# Measures of Group Cohesion

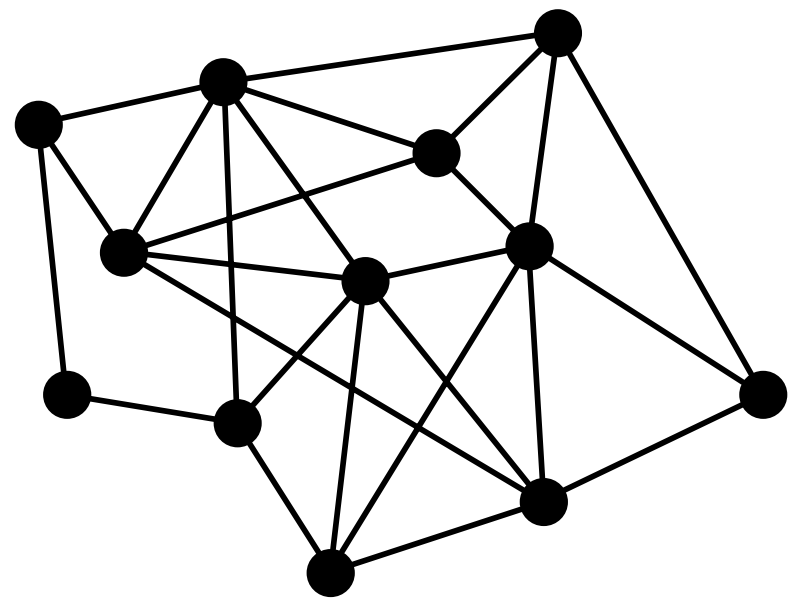
- Density & Average degree
- Average Distance and Diameter
- Number of components
- Fragmentation
- Distance-weighted Fragmentation
- Cliques per node
- Connectivity
- Centralization
- Core/Peripheriness
- Transitivity (clustering coefficient)

# Density

- Number of ties, expressed as percentage of the number of ordered/unordered pairs

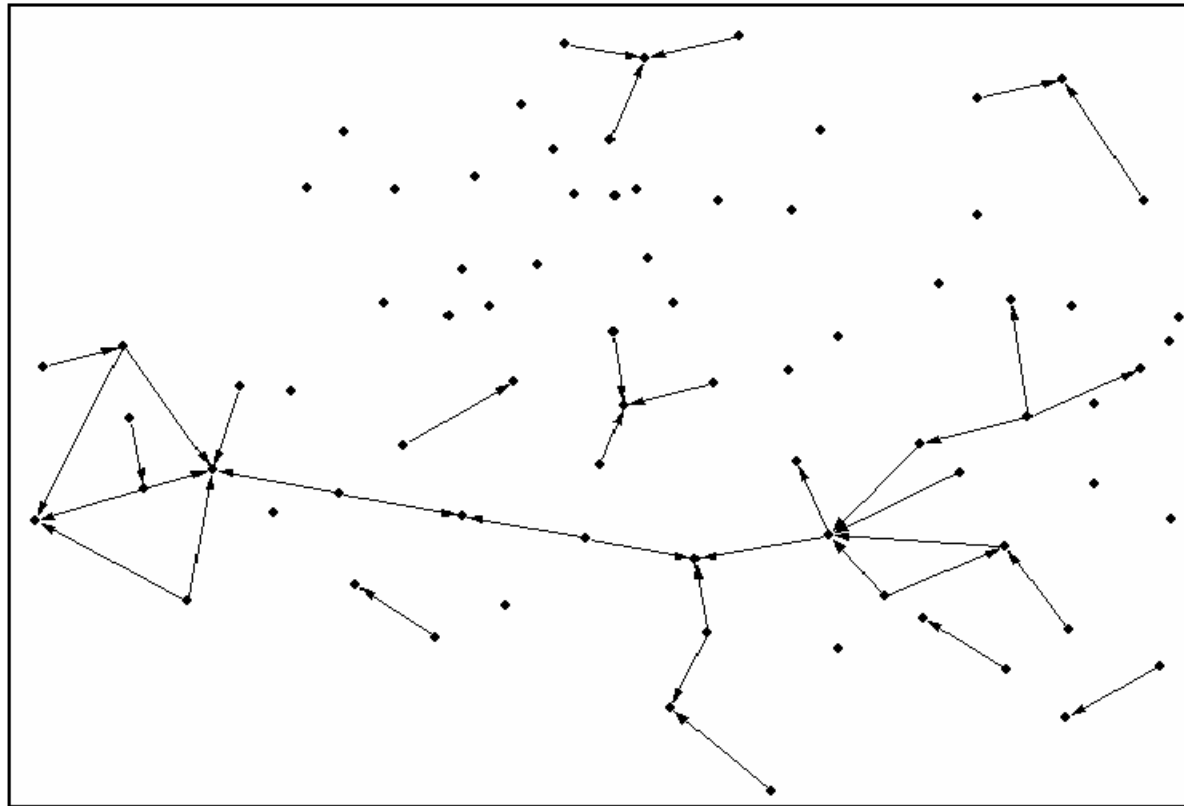


Low Density (25%)  
Avg. Dist. = 2.27



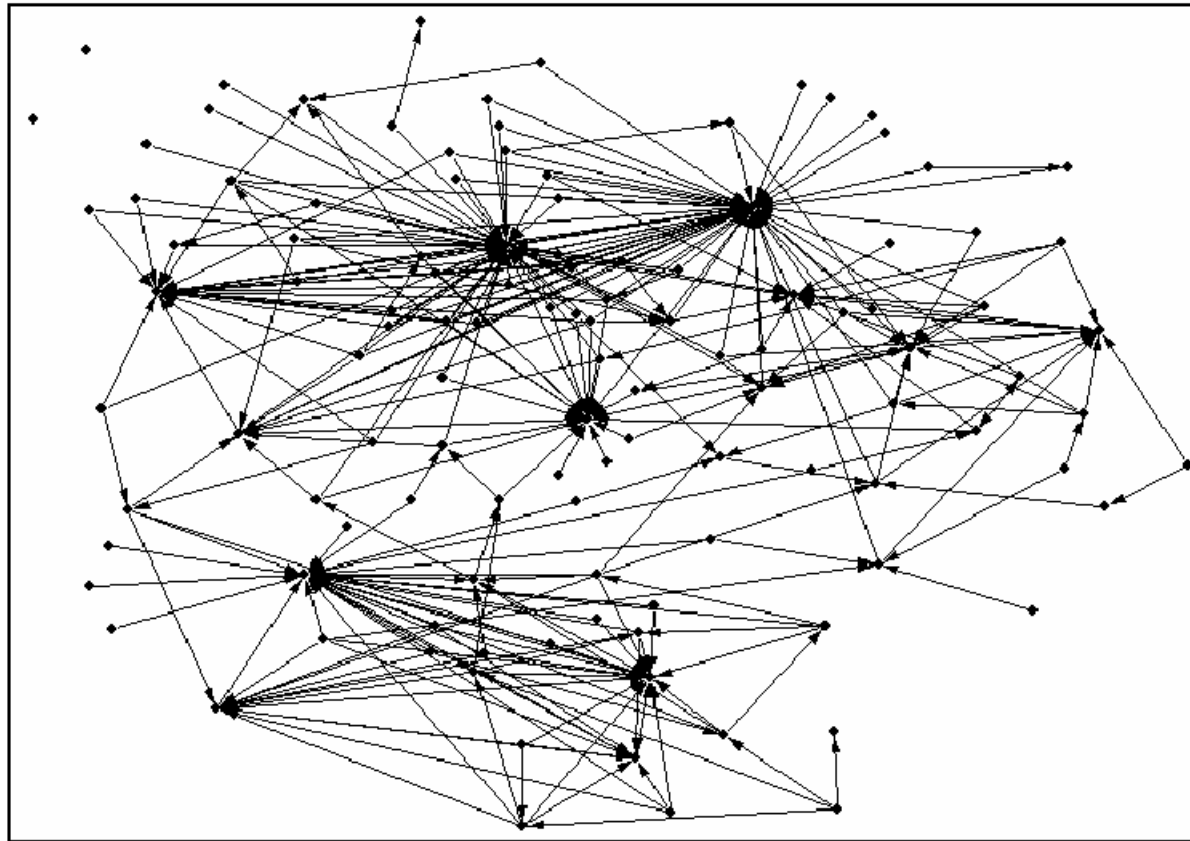
High Density (39%)  
Avg. Dist. = 1.76

# Help With the Rice Harvest



Village 1

# Help With the Rice Harvest

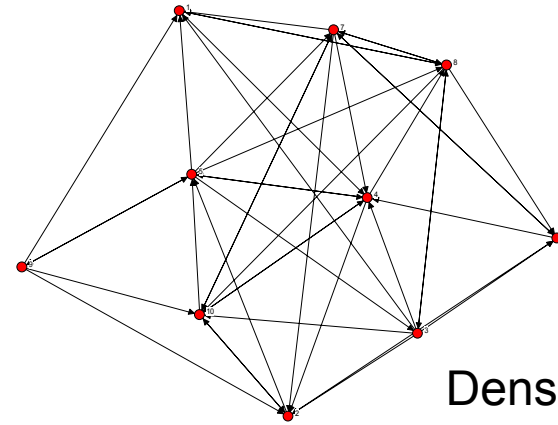


Which  
village  
is more  
likely to  
survive?

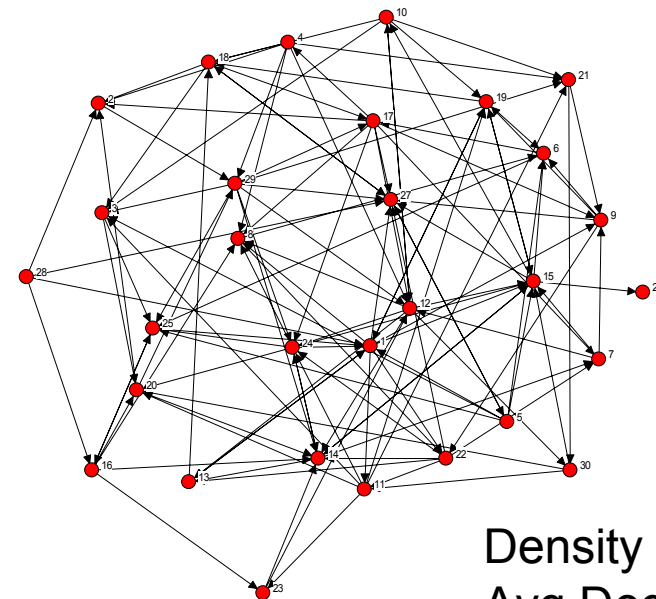
Village 2

# Average Degree

- Average number of links per person
- Is same as  $\text{density} \times (n-1)$ , where  $n$  is size of network
  - Density is just normalized avg degree
  - divide by max possible
- Often more intuitive than density



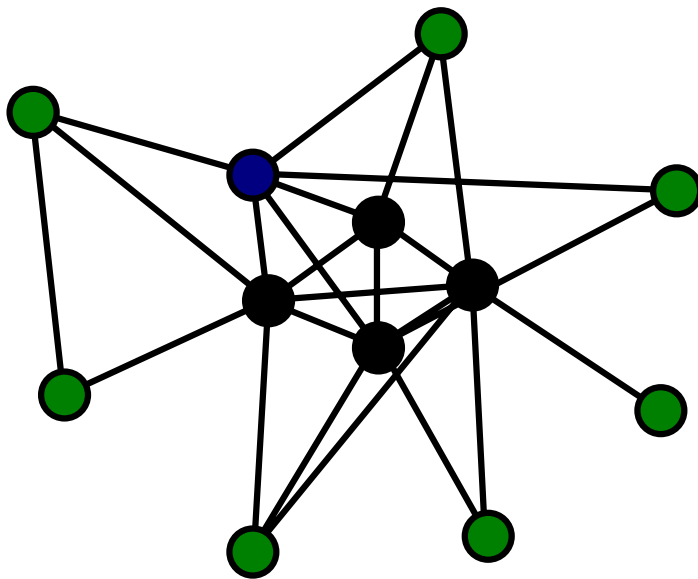
Density 0.47  
Avg Deg 4



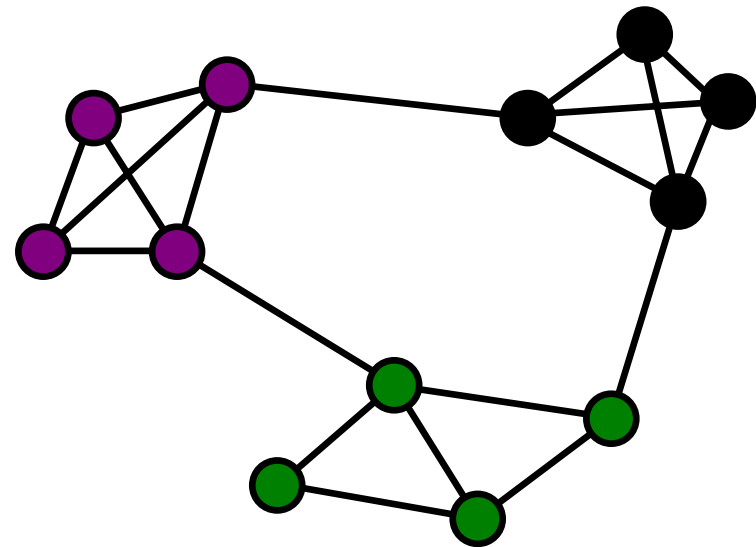
Density 0.14  
Avg Deg 4

# Average Distance

- Average geodesic distance between all pairs of nodes



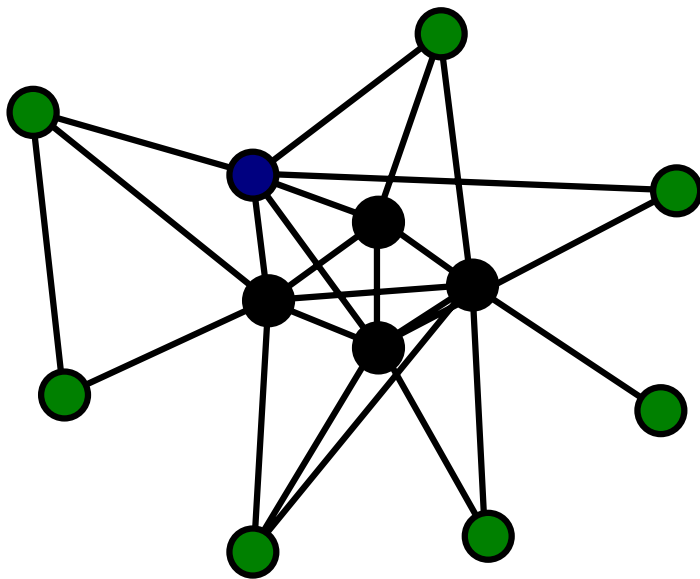
avg. dist. = 1.9



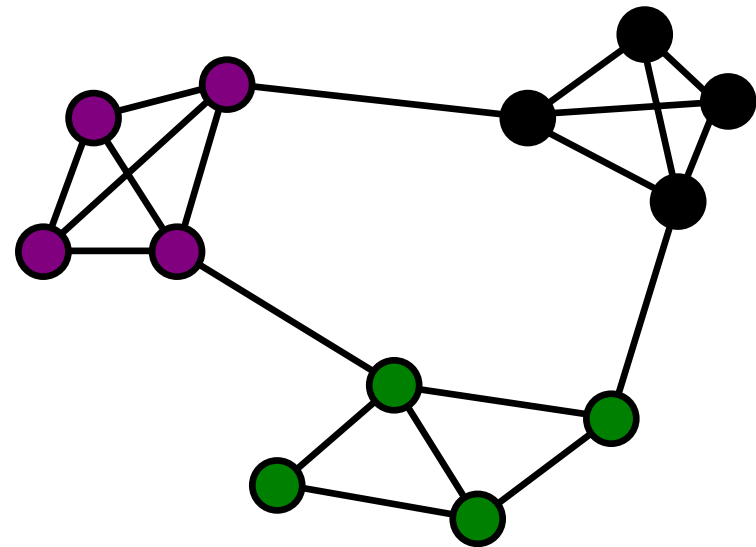
avg. dist. = 2.4

# Diameter

- Maximum distance



Diameter = 3



Diameter = 3

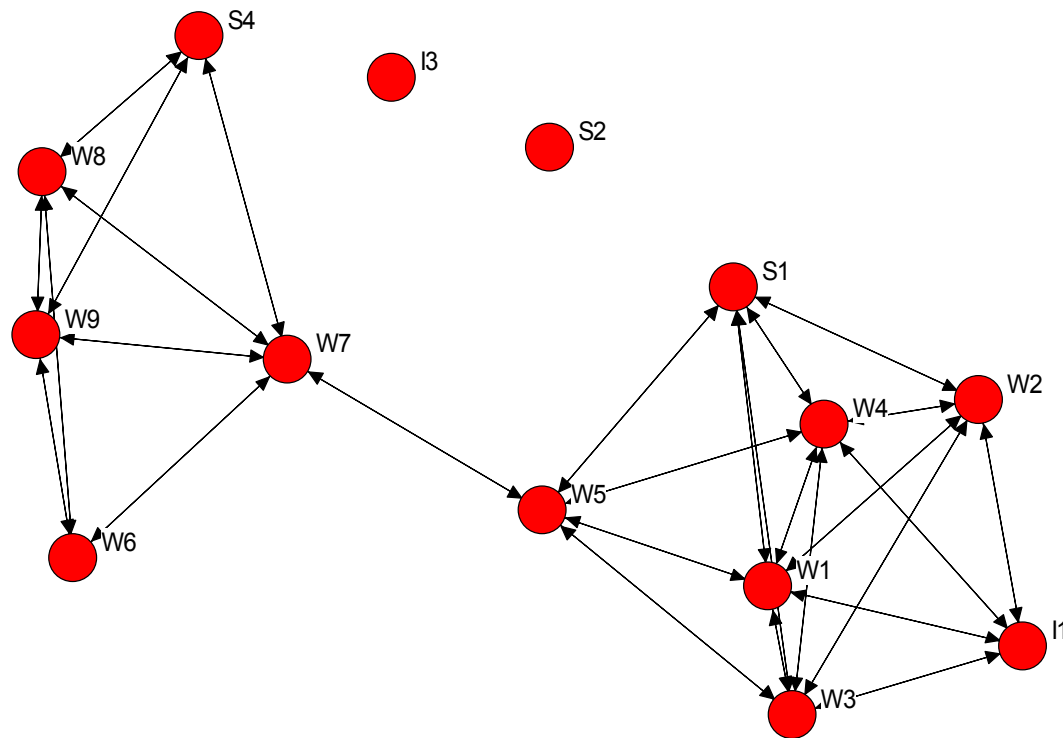
# Fragmentation Measures

- Component ratio
- F measure of fragmentation
- Breadth (Distance-weighted fragmentation) B



# Component Ratio

- No. of components divided by number of nodes



Component ratio =  $3/14 = 0.21$

# F Measure of Fragmentation

- Proportion of pairs of nodes that are unreachable from each other

$$F = 1 - \frac{2 \sum_{i>j} r_{ij}}{n(n-1)}$$

$r_{ij} = 1$  if node  $i$  can reach node  $j$  by a path of any length  
 $r_{ij} = 0$  otherwise

- If all nodes reachable from all others (i.e., one component), then  $F = 0$
- If graph is all isolates, then  $F = 1$

# Computation Formula for F Measure

- No ties across components, and all reachable within components, hence can express in terms of size of components

$$F = 1 - \frac{\sum_k s_k (s_k - 1)}{n(n - 1)}$$

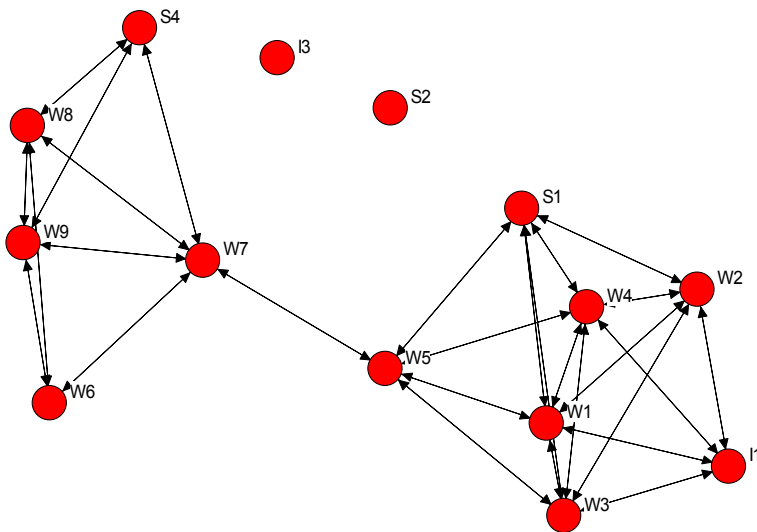
$S_k$  = size of  $k$ th component

# Computational Example

## Games Data

Comp	Size	Sk(Sk-1)
1	1	0
2	1	0
3	12	132
<hr/>		
	14	132

$$\underline{0.2747} = 14/(132*131) = F$$



# Heterogeneity/Concentration

- Sum of squared proportion of nodes falling in each component, where  $s_k$  gives size of  $k$ th component:

$$H = 1 - \sum_k \left( \frac{s_k}{n} \right)^2$$

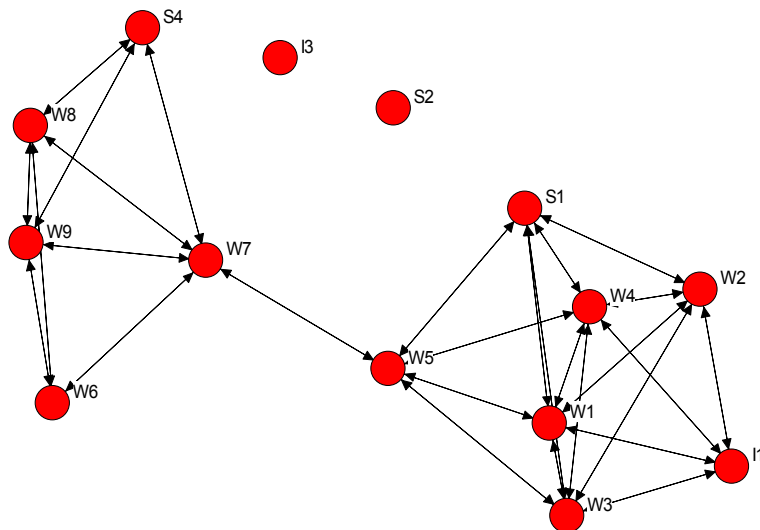
- Maximum value is  $1-1/n$
- Can be normalized by dividing by  $1-1/n$ . If we do, we obtain the F measure

$$F = 1 - \frac{\sum_k s_k (s_k - 1)}{n(n - 1)}$$

# Heterogeneity Example

Games Data

Comp	Size	Prop	Prop <sup>2</sup>
1	1	0.0714	0.0051
2	1	0.0714	0.0051
3	12	0.8571	0.7347
<hr/>			
	14	1.0000	0.7449



Heterogeneity = 0.255

# Breadth

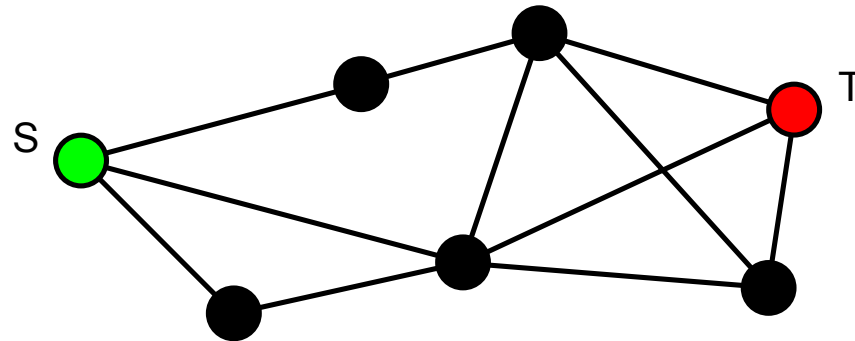
- Distance-Weighted Fragmentation
- Use average of the reciprocal of distance
  - letting  $1/\infty = 0$

$$B = 1 - \frac{\sum_{i,j} \frac{1}{d_{ij}}}{n(n-1)}$$

- Bounds
  - lower bound of 0 when every pair is adjacent to every other (entire network is a clique)
  - upper bound of 1 when graph is all isolates

# Connectivity

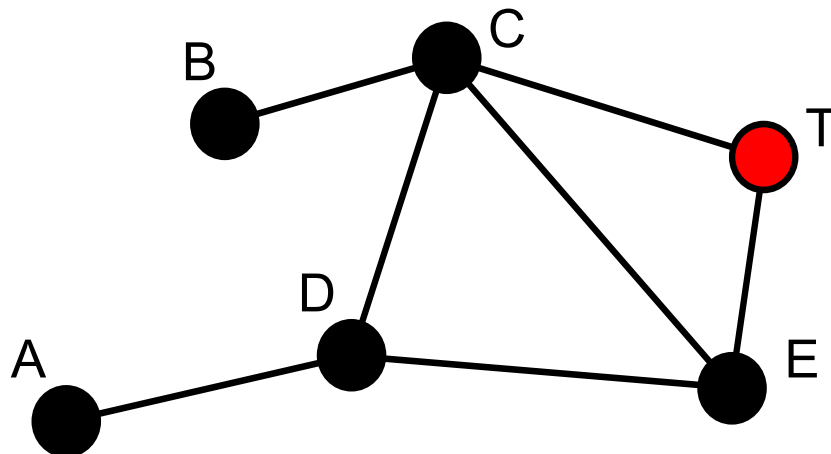
- Line connectivity  $\lambda$  is the minimum number of lines that must be removed to disconnect network
- Node connectivity  $\kappa$  is minimum number of nodes that must be removed to disconnect network





# Transitivity

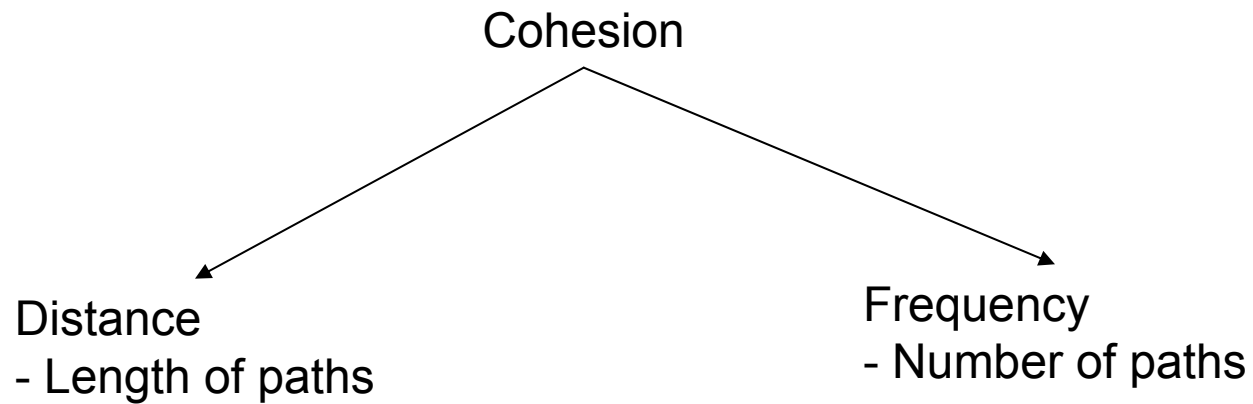
- Proportion of triples with 3 ties as a proportion of triples with 2 or more ties
  - Aka the clustering coefficient



$$cc = 12/26 = 46.15\%$$

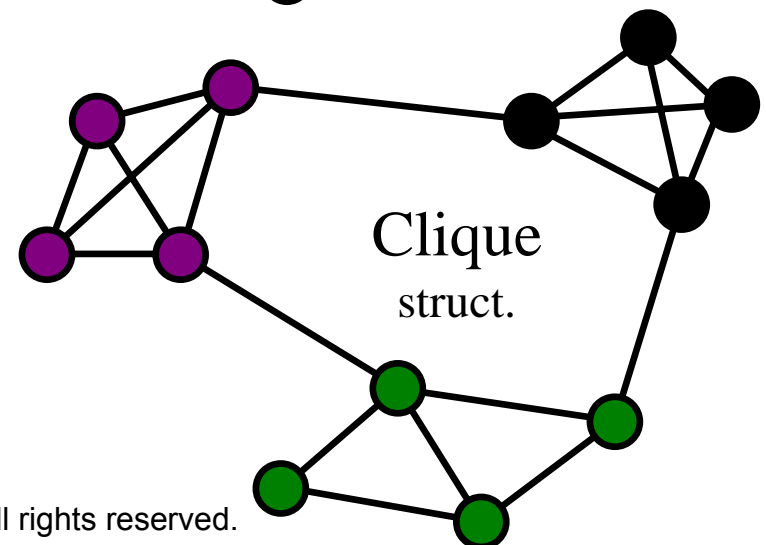
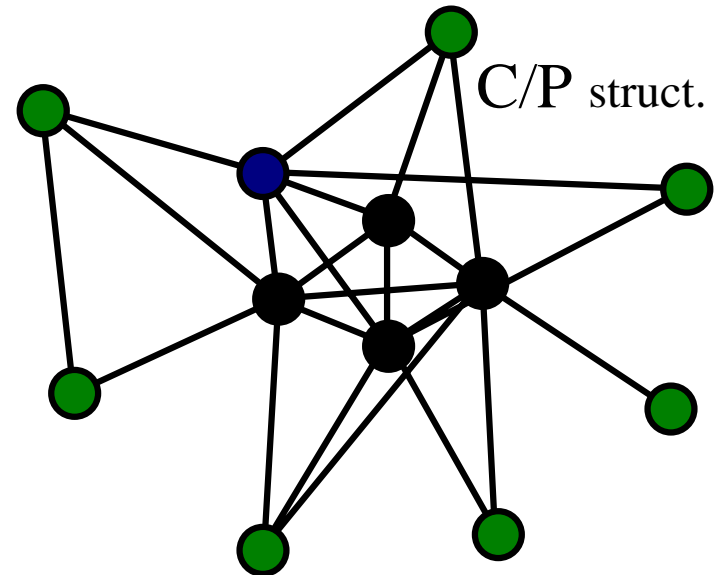
$\{C, T, E\}$  is a transitive triple, but  $\{B, C, D\}$  is not.  $\{A, D, T\}$  is not counted at all.

# Classifying Cohesion



# Core/Periphery Structures

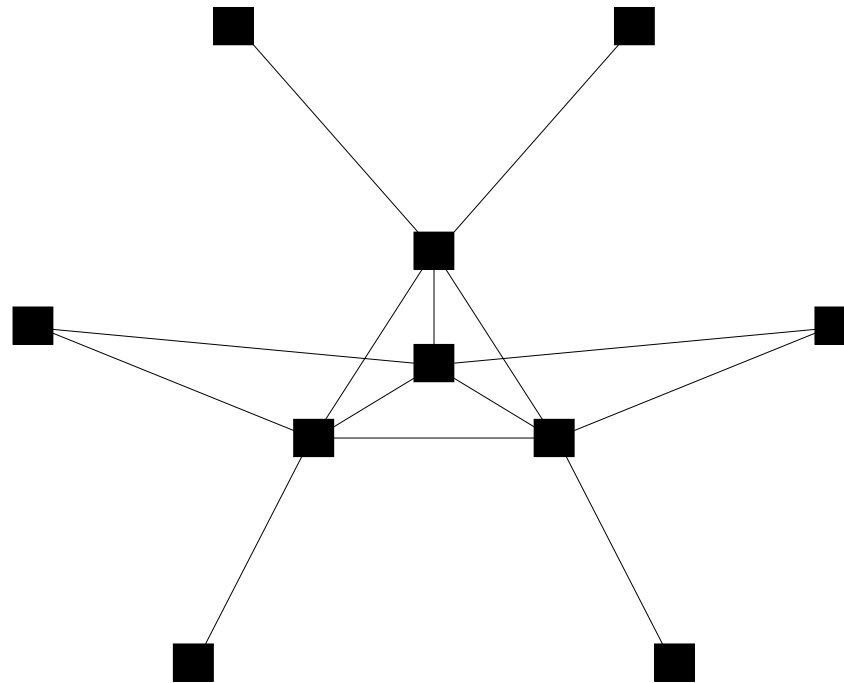
- Does the network consist of a single group (a core) together with hangers-on (a periphery), or
- are there multiple sub-groups, each with their own peripheries?



# Kinds of CP/Models

- Partitions vs. subgraphs
  - just as in cohesive subgroups
- Discrete vs. continuous
  - classes, or
  - coreness

# A Core/Periphery Structure



# Blocked/Permuted Adjacency Matrix

		C O R E				P E R I P H E R Y					
C O R E		-	1	1	1	1	0	0	1	0	0
		1	-	1	1	0	1	1	0	0	0
		1	1	-	1	0	0	0	1	1	0
		1	1	1	-	1	0	0	0	0	1
P E R I P H E R Y		1	0	0	1	-	0	0	0	0	0
		0	1	0	0	0	-	0	0	0	0
		0	1	0	0	0	0	-	0	0	0
		1	0	1	0	0	0	0	-	0	0
		0	0	1	0	0	0	0	0	-	0
		0	0	0	1	0	0	0	0	0	-

- Core-core is 1-block
- Core-periphery are (imperfect) 1-blocks
- Periphery-periphery is 0-block

# Idealized Blockmodel

	CORE				PERIPHERY					
CORE	-	1	1	1	1	1	1	1	1	1
	1	-	1	1	1	1	1	1	1	1
	1	1	-	1	1	1	1	1	1	1
	1	1	1	-	1	1	1	1	1	1
PERIPHERY	1	1	1	1	-	0	0	0	0	0
	1	1	1	1	0	-	0	0	0	0
	1	1	1	1	0	0	-	0	0	0
	1	1	1	1	0	0	0	-	0	0
	1	1	1	1	0	0	0	0	-	0
	1	1	1	1	0	0	0	0	0	-

$c_i$  = class (core or periphery) that node  $i$  is assigned to

$$\delta_{ij} = \left\{ \begin{array}{l} 1 \text{ if } c_i = \text{CORE} \text{ or } c_j = \text{CORE} \\ 0 \text{ otherwise} \end{array} \right\}$$

# Partitioning a Data Matrix

- Given a graphmatrix, we can randomly assign nodes to either core or periphery
- Search for partition that resembles the ideal



# Assessing Fit to Data

$a_{ij}$  = cell in data matrix

$c_i$  = class (core or periphery) that node  $i$  is assigned to

$$\delta_{ij} = \left\{ \begin{array}{l} 1 \text{ if } c_i = \text{CORE or } c_j = \text{CORE} \\ 0 \text{ otherwise} \end{array} \right\}$$

$$\rho = \sum_{i,j} a_{ij} \delta_{ij}$$

- A Pearson correlation coefficient  $r(A,D)$  is

# Alternative Images

	Core					Periphery				
<i>C</i>	-	1	1	1	1	0	0	0	0	0
<i>o</i>	1	-	1	1	1	0	0	0	0	0
<i>r</i>	1	1	-	1	1	0	0	0	0	0
<i>e</i>	1	1	1	-	1	0	0	0	0	0
	1	1	1	1	-	0	0	0	0	0
<i>P</i>	0	0	0	0	0	-	0	0	0	0
<i>e</i>	0	0	0	0	0	0	-	0	0	0
<i>r</i>	0	0	0	0	0	0	0	-	0	0
<i>i</i>	0	0	0	0	0	0	0	0	-	0
	0	0	0	0	0	0	0	0	0	-

# Alternative Images

	Core					Periphery				
C	-	1	1	1	1	-	-	-	-	-
o	1	-	1	1	1	-	-	-	-	-
r	1	1	-	1	1	-	-	-	-	-
e	1	1	1	-	1	-	-	-	-	-
	1	1	1	1	-	-	-	-	-	-
P	-	-	-	-	-	-	0	0	0	0
e	-	-	-	-	-	0	-	0	0	0
r	-	-	-	-	-	0	0	-	0	0
I	-	-	-	-	-	0	0	0	-	0
	-	-	-	-	-	0	0	0	0	-

# Continuous Model

- $X_{ij} \sim C_i C_j$ 
  - Strength or probability of tie between node  $i$  and node  $j$  is function of product of coreness of each
  - Central players are connected to each other
  - Peripheral players are connected only to core

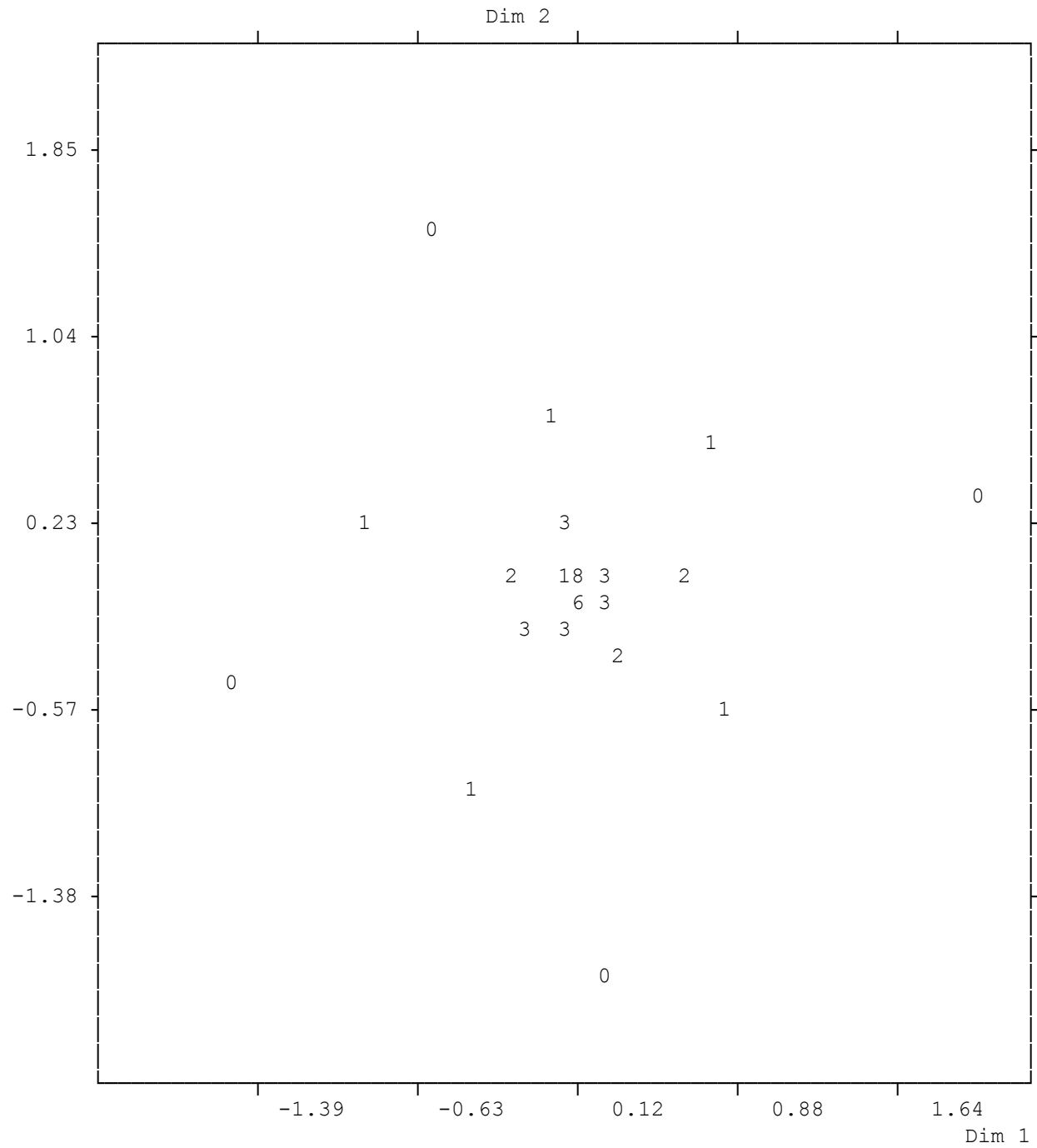
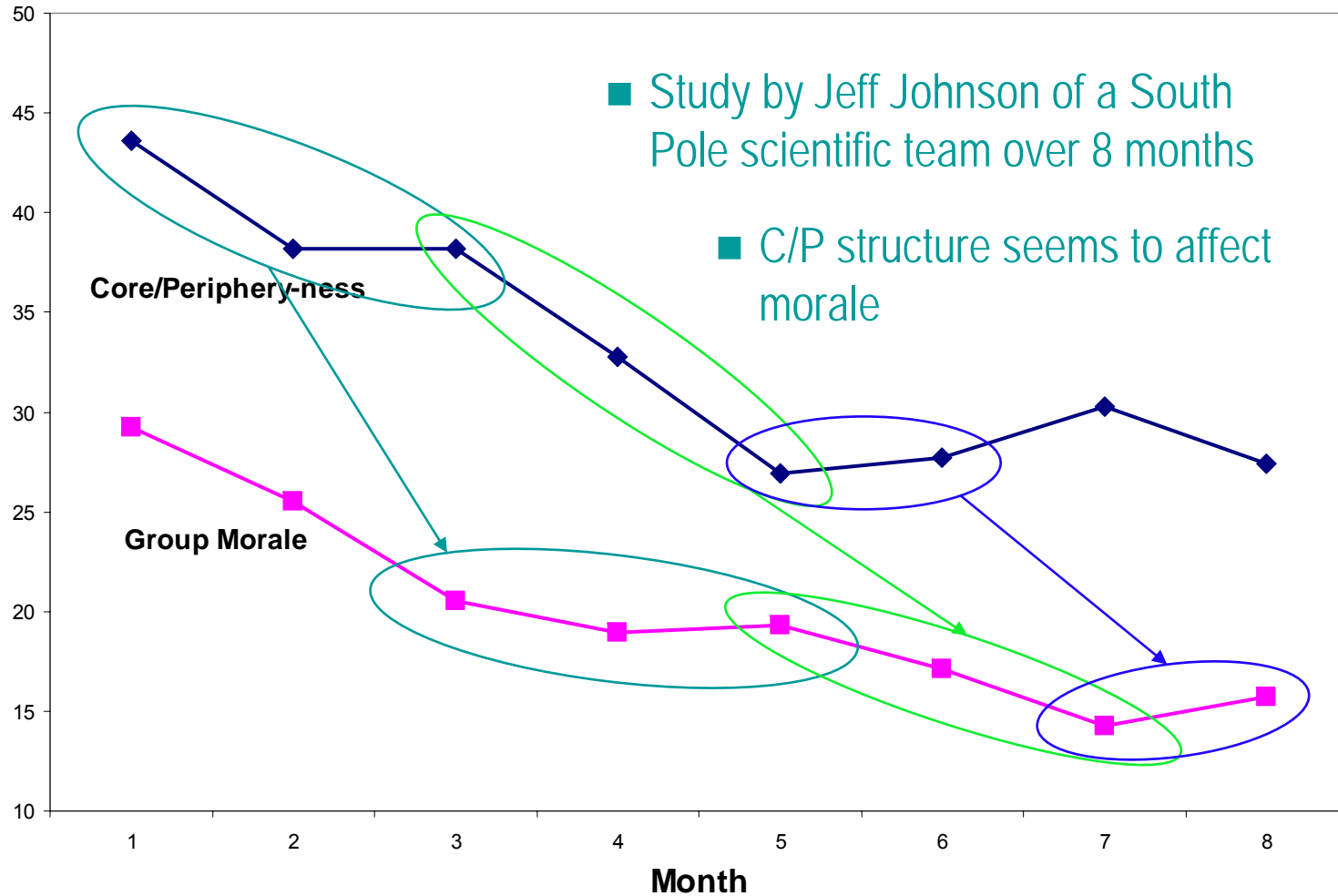


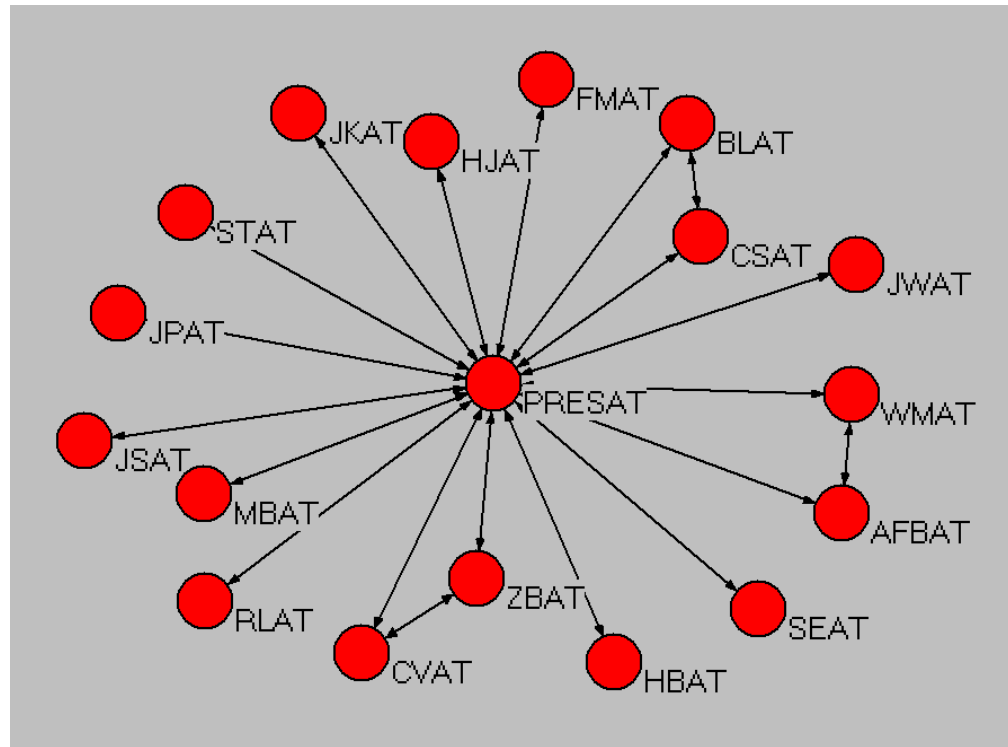
Figure 4: MDS of  $\log_{10}(\text{perio})$

# CP Structures & Morale



# Centralization

- Degree to which network revolves around a single node



*Carter admin.  
Year 1*