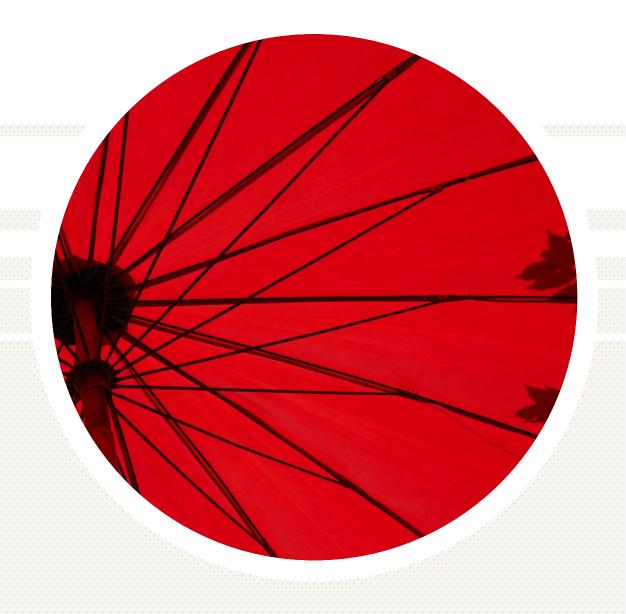




What Network Analysis Can Teach Us About How We Collaborate

Deanna Schneider, Division of Extension





# **Pilot Project**

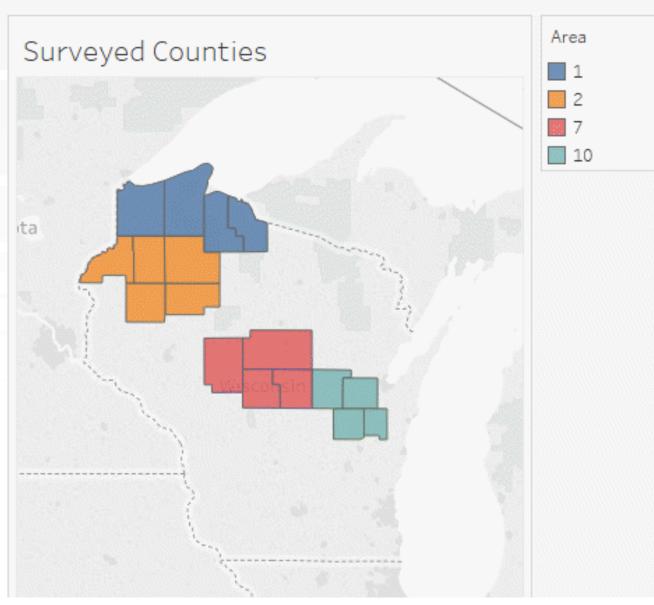
Surveyed 80 employees in 4 areas asking...

### How often do you:

- Communicate
- Collaborate

#### Added:

- Employment Data
- Zoom Data





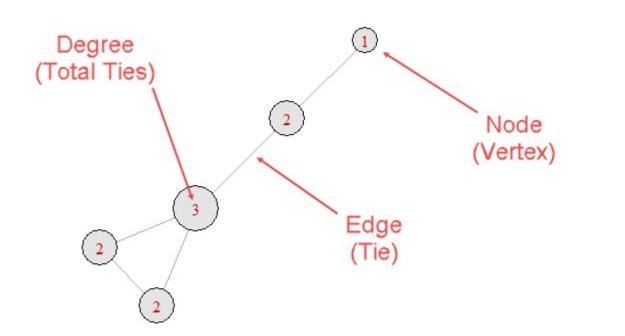
### **Tools and Process**

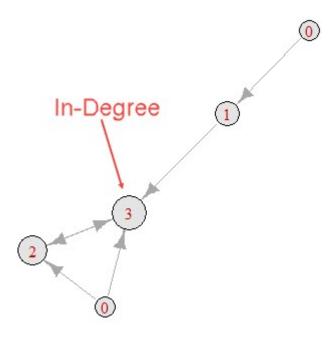
- Collect Data (Qualtrics)
- Clean Data (Excel)
- Descriptive Statistics (R/Kumu)
- Visualizations (R/Kumu)
- Modeling (R)



#### What's a Network?

A group or system of interconnected people or things.

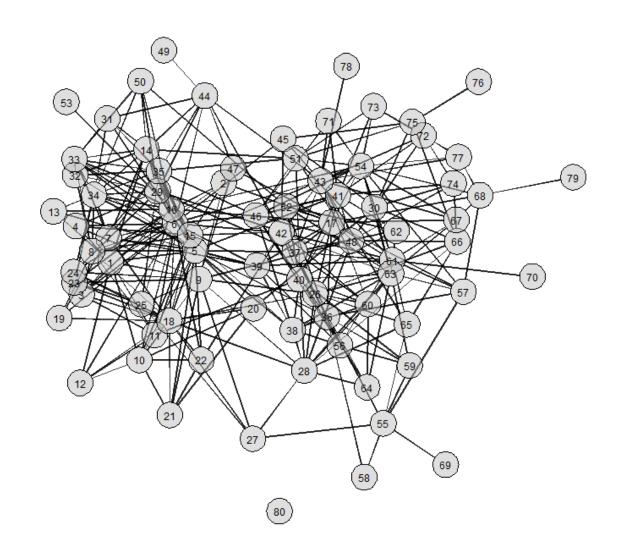




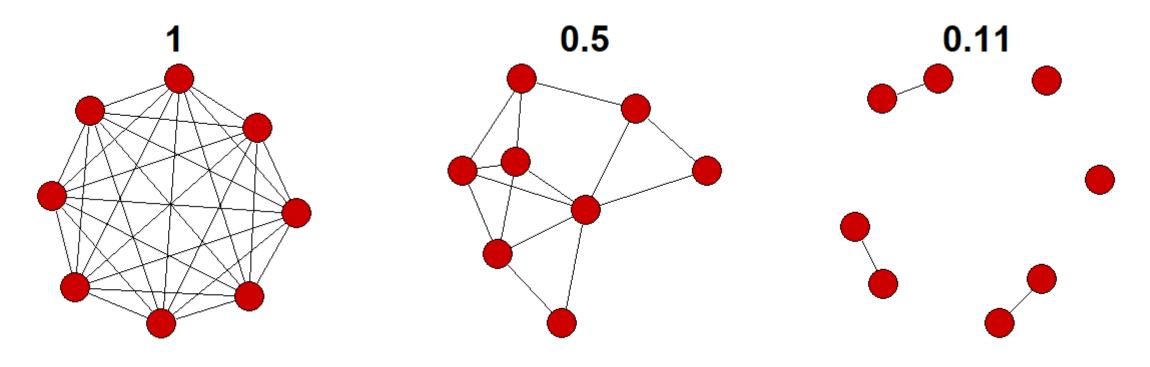
### **Basic Network Statistics**



- 80 nodes
- 339 edges
- 2 components
- 1 isolate

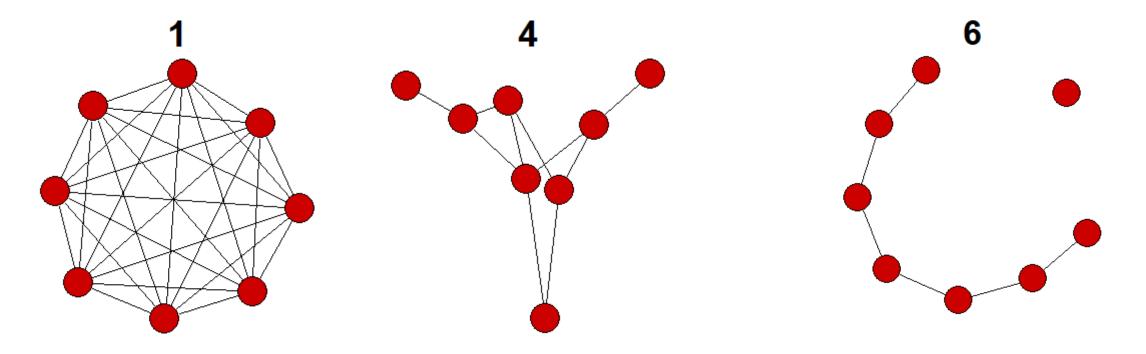


## **Basic Network Statistics: Density**



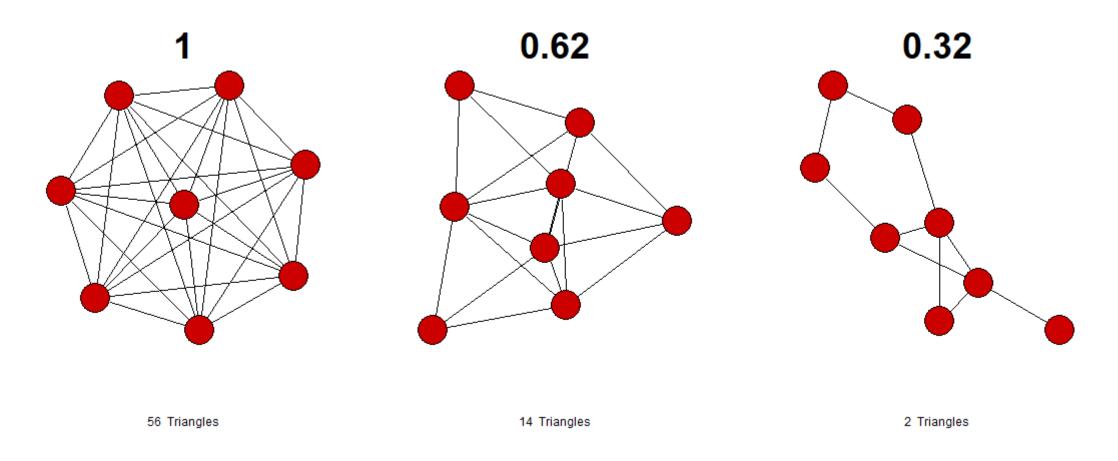
Density = the proportion of observed edges to possible edges

### **Basic Network Statistics: Diameter**



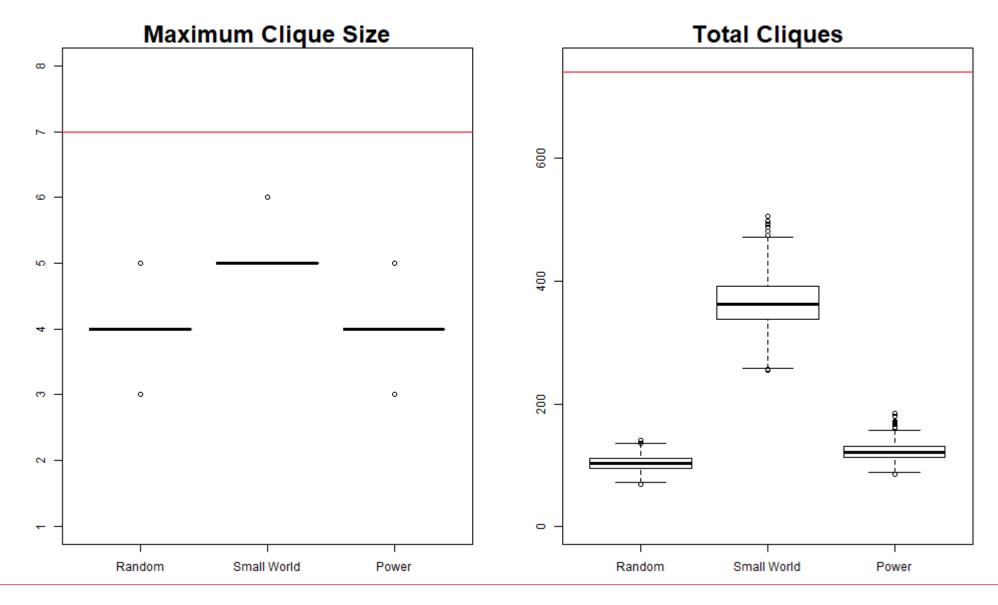
Diameter = the shortest longest path

### **Basic Network Statistics: Transitivity**

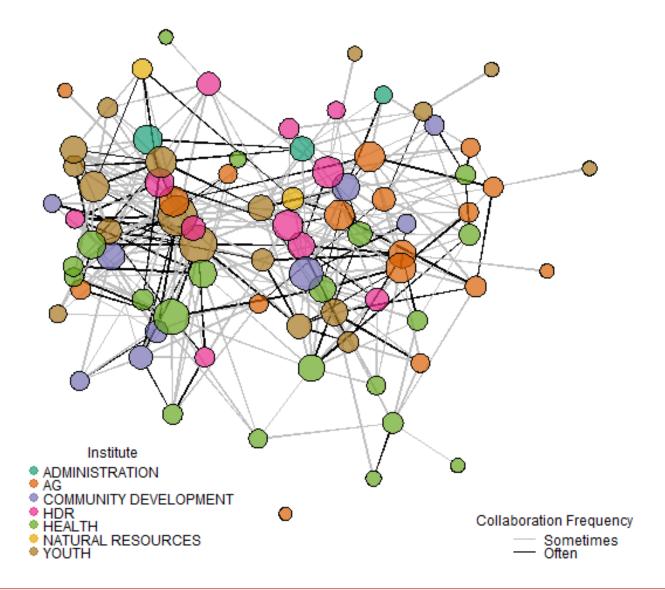


Transitivity = the proportion of closed triangles to possible triangles

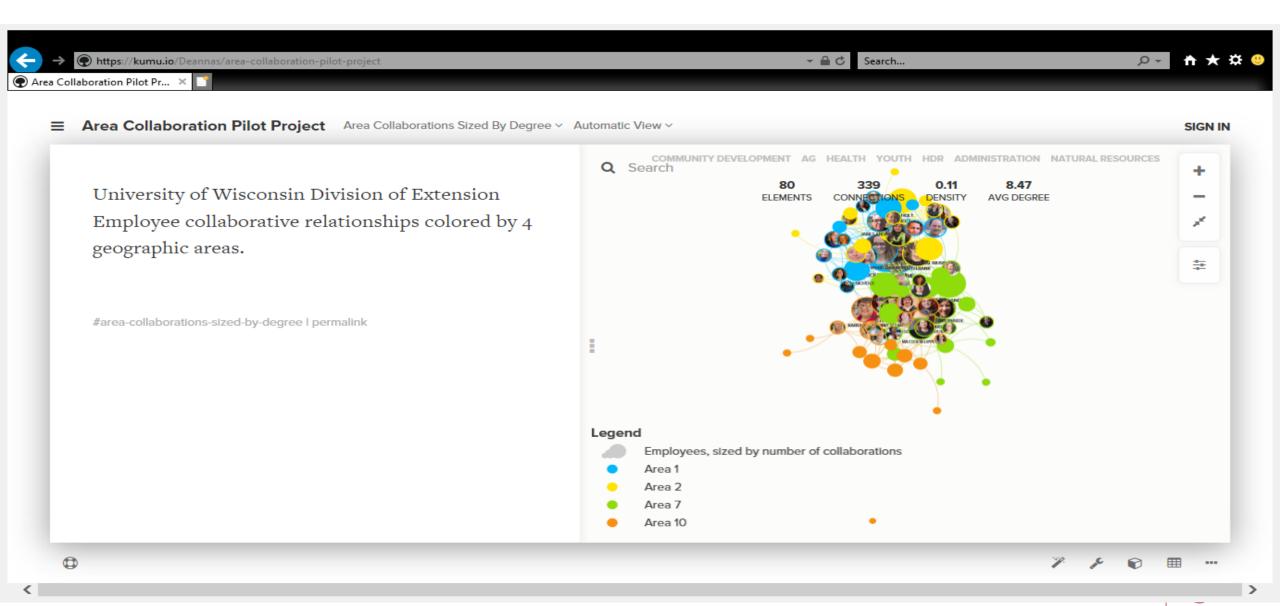
# **Clique Comparison**



# Visualizing Our Network – Institute Affiliation

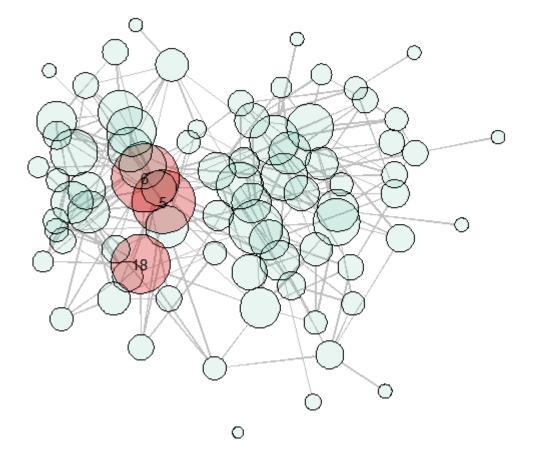


## https://kumu.io

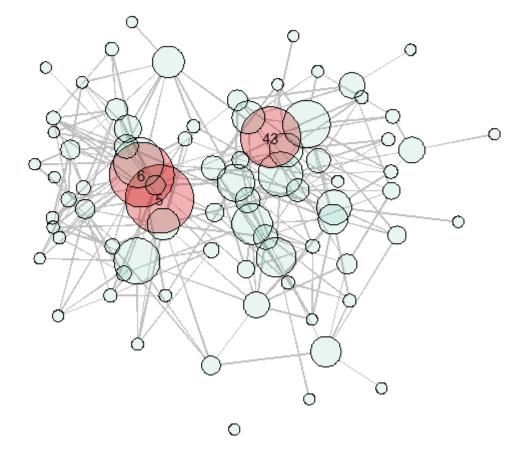


### **Actor Prominence**

#### Degree



#### **Betweenness**



# **Inspecting Our Prominent Actors**

ID	Area	Location	Department	Institute	Sex	Years in Job
5	1	IRON	YFCD	YOUTH	M	6.69
6	1	ASHLAND	YFCD	YOUTH	M	13.32
18	1	ASHLAND	YFCD	HEALTH	F	0.1
43	10	OUTAGAMIE	YFCD	HDR	F	28.53

## **Modeling with ERGM**

- Exponential Random Graph Model
- Defines probabilities of ties between 2 nodes
- Based on node, edge, and network structure characteristics.
- Interpreted like logistic regression models

### Some cautions

- Correlation, not causation
- We can only model the data we have
- Networks are fluid. Our model is static.

### The Final Fitted Model

TERM	ESTIMATE	P-VALUE
edges	-6.0198056	< 1e-04 ***
No. of Zoom Minutes	0.0015947	0.00868 **
No. of Zoom Participants	-0.0698162	0.00994 **
Faculty > Academic Staff	0.6368449	< 1e-04 ***
Limited > Academic Staff	0.8557039	< 1e-04 ***
Matching Employee Class	0.3055619	0.01302 *
Matching Area 1	1.6459646	< 1e-04 ***
Matching Area 10	1.3329971	< 1e-04 ***
Matching Area 2	1.6425401	< 1e-04 ***
Matching Area 7	1.1366766	< 1e-04 ***
Matching County	1.9273653	< 1e-04 ***
Matching Department	0.2871169	0.02873 *
Matching Institute	1.7435603	< 1e-04 ***
Amount of Communication	0.3389310	0.00150 **
No. Of Shared Collaborators	0.5946841	< 1e-04 ***

### **Odds Ratios**

TERM	LOWER	ODDS RATIO	UPPER
Matching County	4.79	6.87	9.85
Matching Institute	4.36	5.72	7.50
Matching Area 1	3.86	5.19	6.97
Matching Area 2	3.81	5.17	4.05
Matching Area 10	2.83	3.79	5.09
Matching Area 7	2.40	3.12	4.05
Limited > Academic Staff	1.6	2.5	3.41
Faculty > Academic Staff	1.63	1.90	2.20
No. of Shared Collaborators	1.64	1.81	2.00
Freq. of Communication	1.14	1.40	1.73
Matching Employee Class	1.07	1.36	1.73
Matching Department	1.03	1.33	1.72
No. of Zoom Minutes	1.000	1.002	1.003
No. of Zoom Participants	0.88	0.93	0.98
Edges	0.0002	0.0024	0.0031

## Will they collaborate?

#### **Health & Well-Being Academic Staff**

• 1000 minutes in Zoom meetings with 20 participants







### **Community Development Academic Staff**

• 100 minutes in Zoom meetings with 2 participants

## Will they collaborate?

#### **Health & Well-Being Academic Staff**

• 1000 minutes in Zoom meetings with 20 participants







#### **Health & Well-Being Academic Staff**

100 minutes in Zoom meetings with 2 participants

## Will they collaborate?

#### **Health & Well-Being Academic Staff**

• 1000 minutes in Zoom meetings with 20 participants





Likelihood of collaborating 79.6%

#### **Health & Well-Being Academic Staff**

100 minutes in Zoom meetings with 2 participants

# **Final Thoughts**

## Network analysis is:

- Relatively easy to do
- Interesting and understandable

### helps us:

- Understand how people are connected
- See who is left out
- Predict relationships in our network

