

*Statistical Analysis of Networks*

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# Degree Centrality

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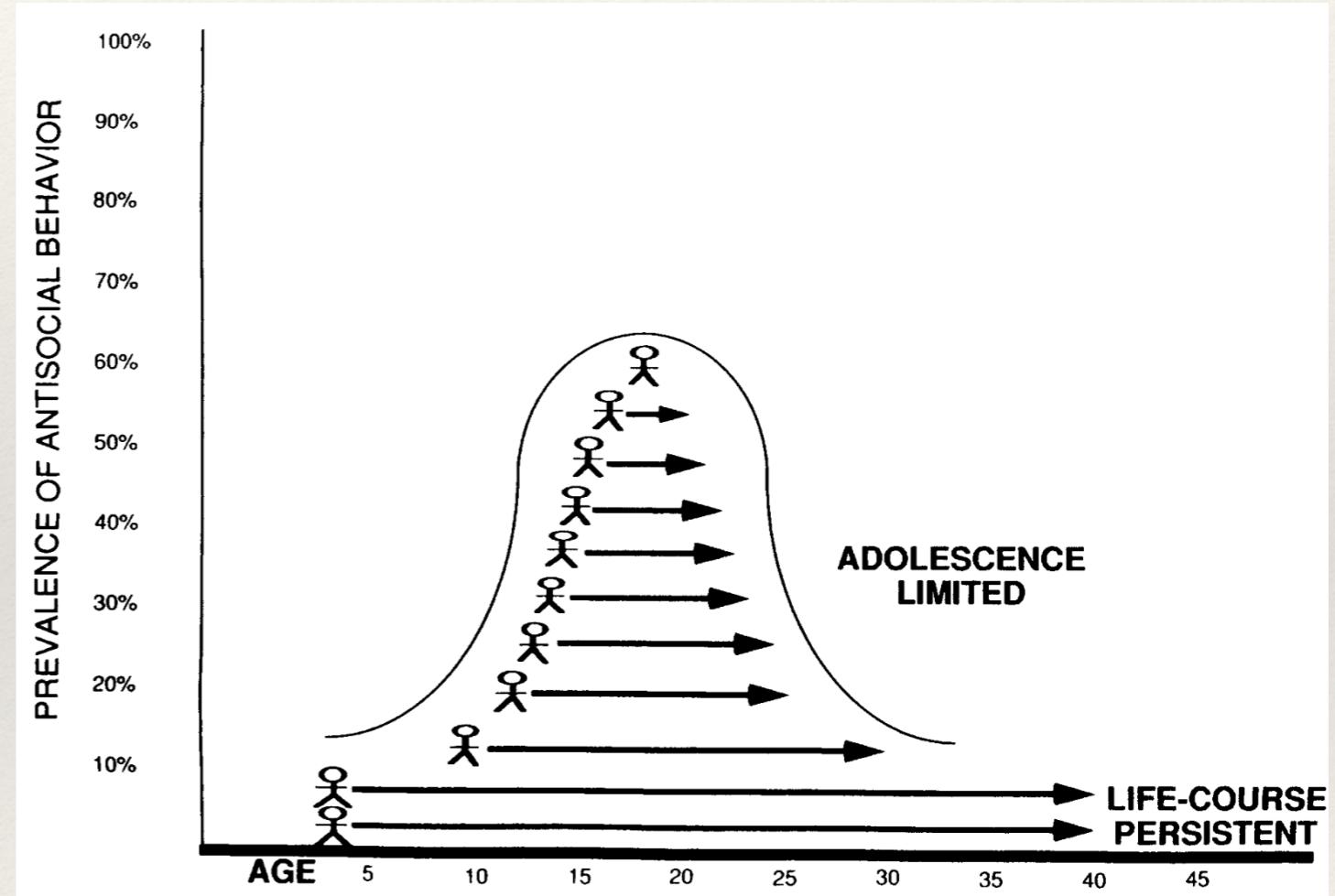
# Motivating Example

- ❖ Question: Why is there a dramatic increase in delinquency during adolescence?



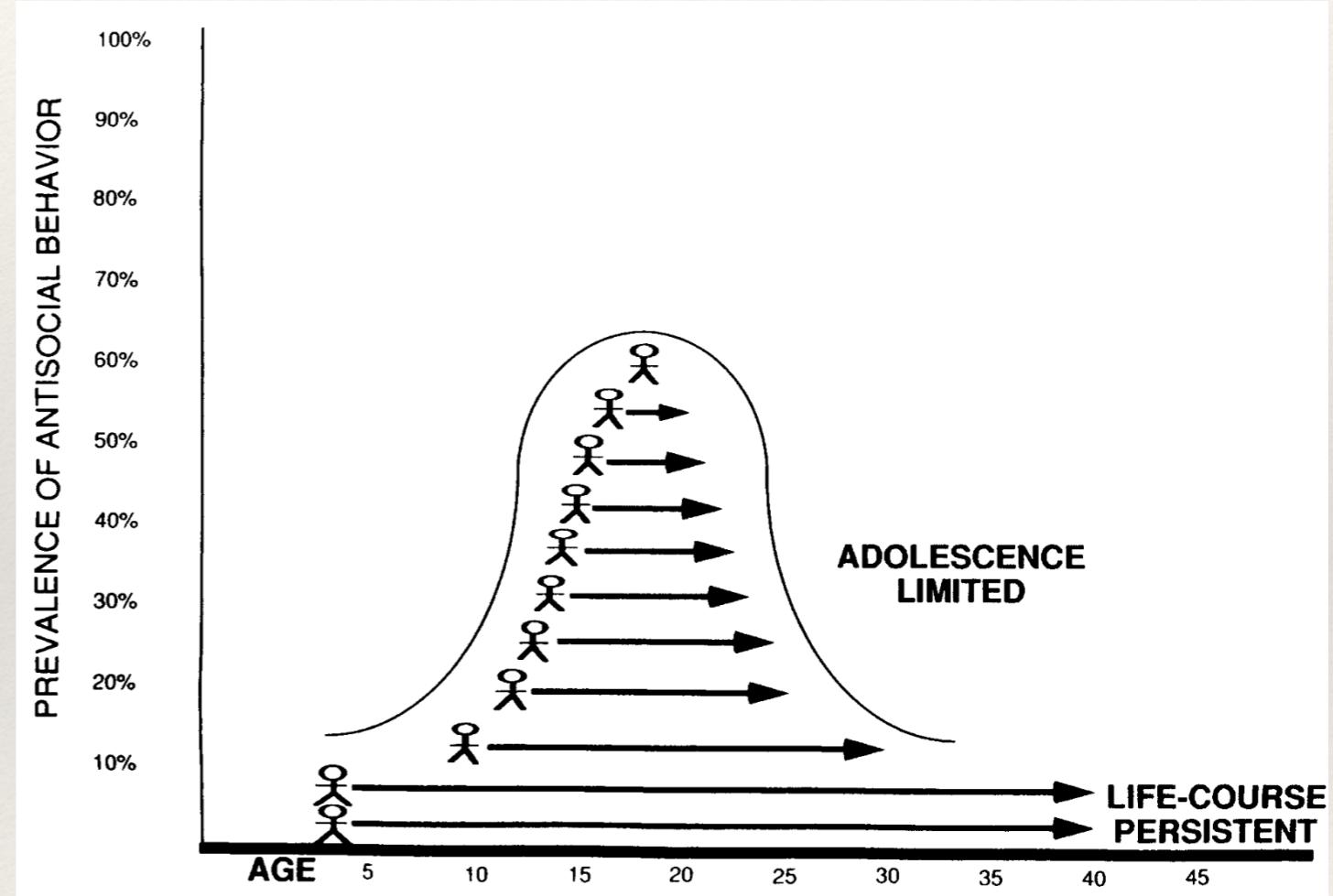
# Motivating Example

- ❖ Question: Why is there a dramatic increase in delinquency during adolescence?
- ❖ Argument: Moffitt's (1993) Dual-taxonomy theory.



# Motivating Example

- ❖ Question: Why is there a dramatic increase in delinquency during adolescence?
- ❖ Argument: Moffitt's (1993) Dual-taxonomy theory.
- ❖ *So what?*



# Motivating Example

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- ❖ Dual-taxonomy theory argues that the causal mechanism generating the dramatic increase in delinquency during adolescence is social **mimicry**
- ❖ As a consequence, life-course persistent individuals should occupy “more influential positions in the peer social structure” (Moffitt 1993: 687) and should be “moving toward central positions, during early adolescence” (Moffitt 1997: 28).
- ❖ This is a hypothesis about the trajectory of centrality for a set of individuals

# Motivating Example

J Youth Adolescence (2014) 43:104–115  
DOI 10.1007/s10964-013-9946-0

EMPIRICAL RESEARCH

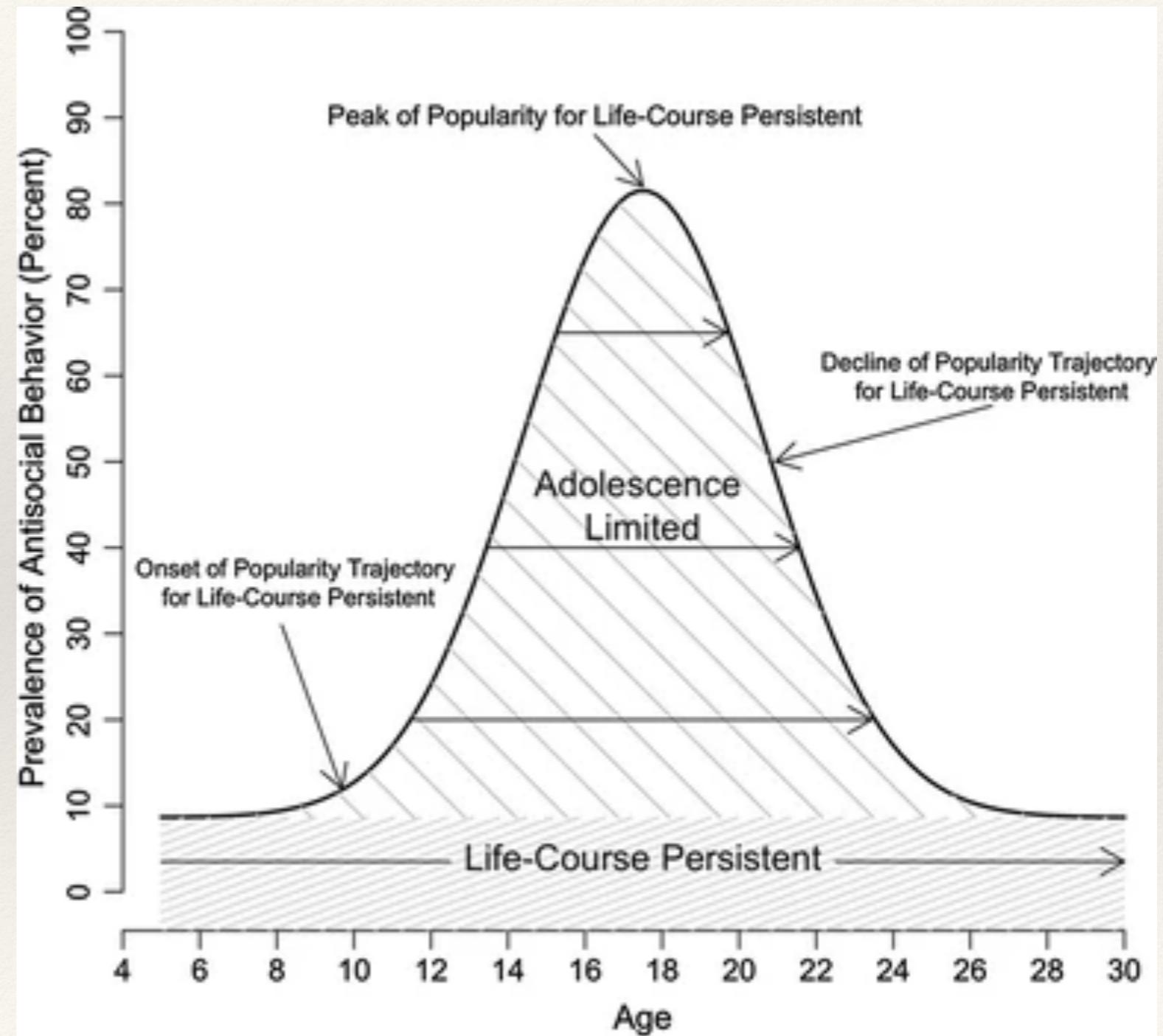
## **“Role Magnets”? An Empirical Investigation of Popularity Trajectories for Life-Course Persistent Individuals During Adolescence**

Jacob T. N. Young

- ❖ <https://link.springer.com/article/10.1007/s10964-013-9946-0>
- ❖ Goal: Examine the developmental trajectory of popularity during adolescence for individuals showing persistent violence into young adulthood.

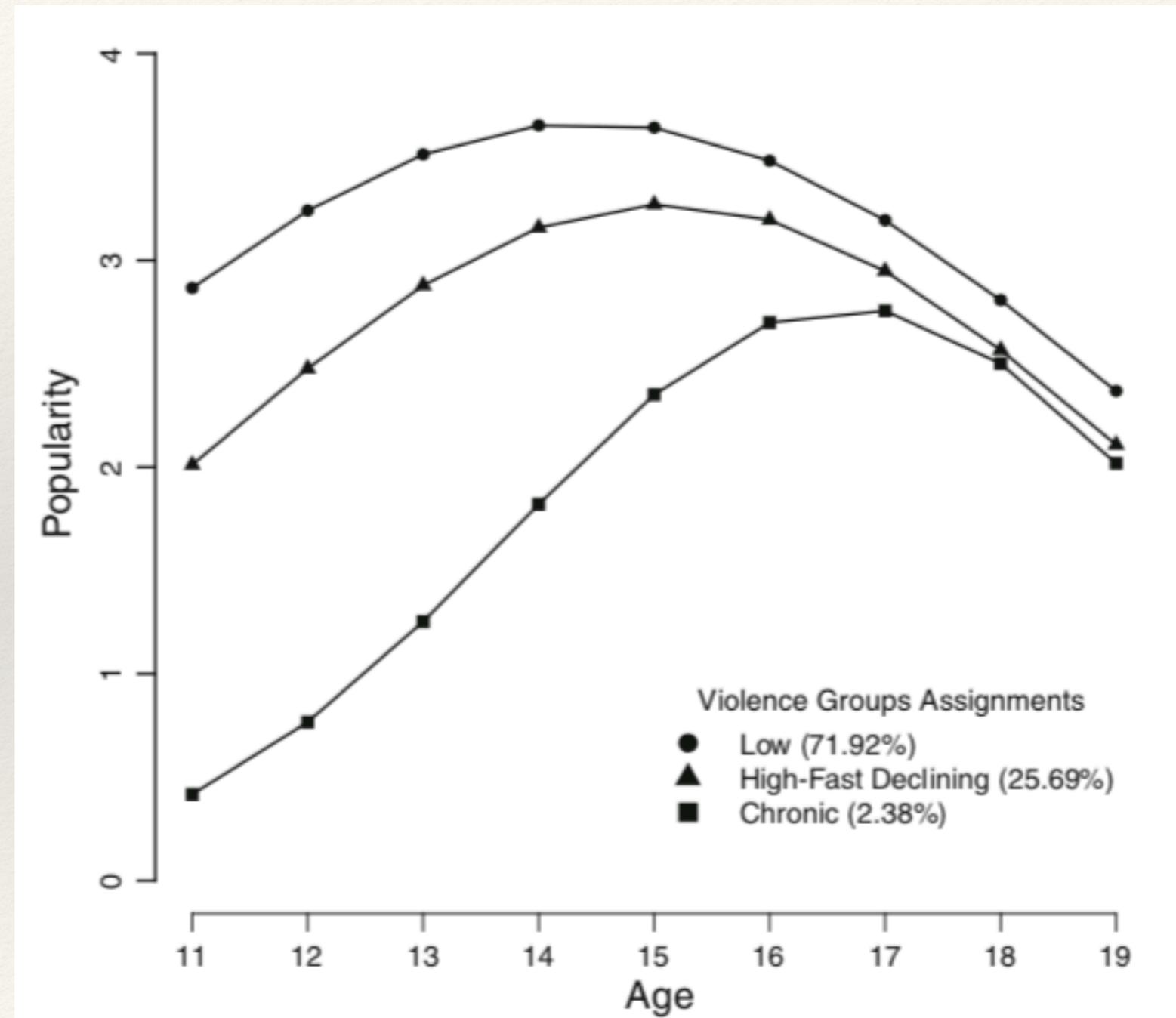
# Motivating Example

- ❖ This is a hypothesis about the trajectory of centrality for a set of individuals



# Motivating Example

- ❖ Findings: Chronically violent individuals showed a more precipitous increase in centrality during adolescence.



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# *Another* Motivating Example

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- ❖ How do you reduce hate speech in online forums?

# *Another Motivating Example*

- ❖ “Deplatforming” is a type of **strategic network disruption**
- ❖ Why should it work?  
According to this intervention,  
what does the network look  
like in online hate  
organizations? (Draw it)



# *Another Motivating Example*

- ❖ “Deplatforming” is a type of **strategic network disruption**
- ❖ Does it work?

**PNAS**

RESEARCH ARTICLE

POLITICAL SCIENCES

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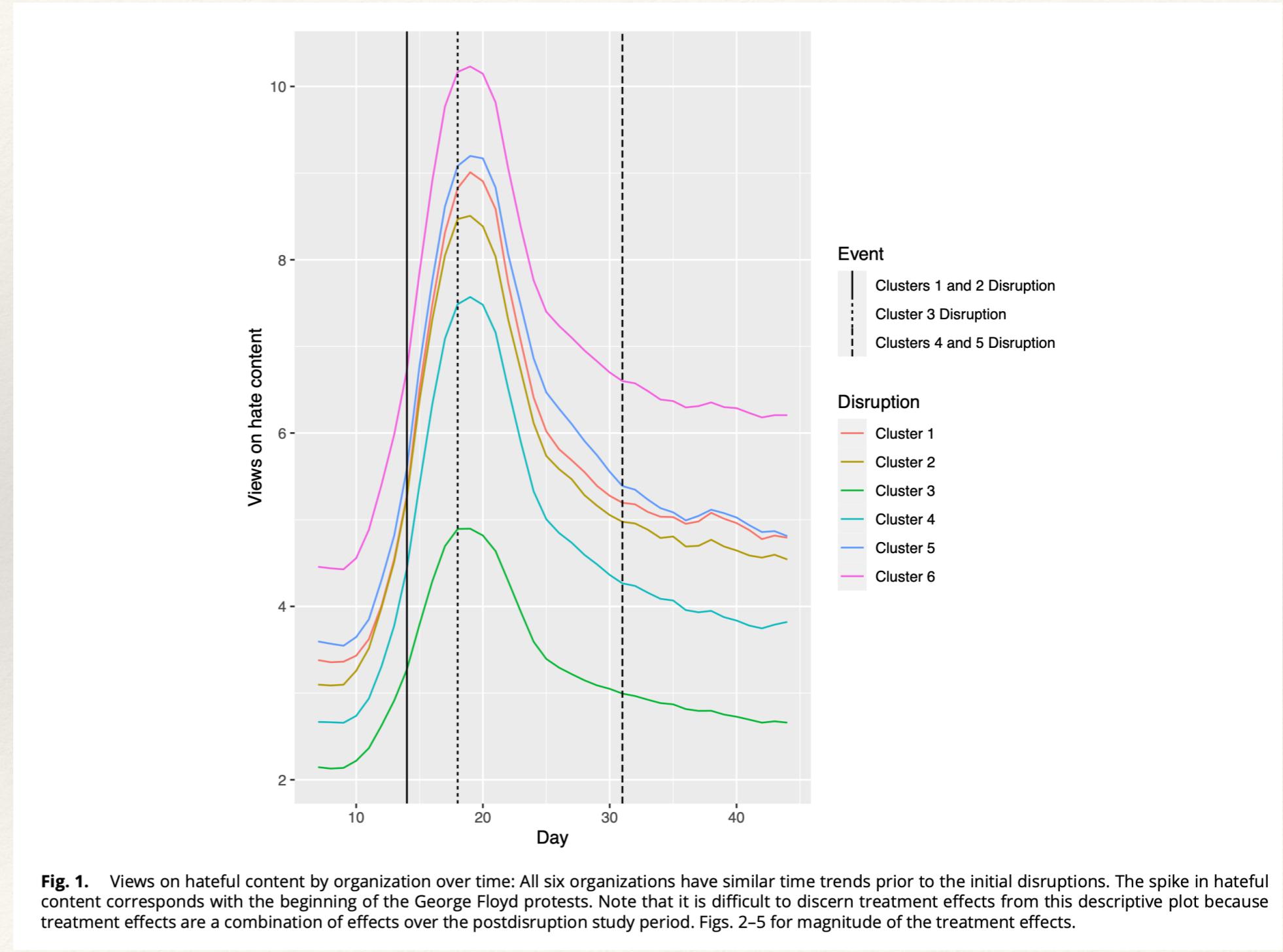
## **Disrupting hate: The effect of deplatforming hate organizations on their online audience**

Daniel Robert Thomas<sup>a,1</sup> and Laila A. Wahedi<sup>a</sup>

Edited by Timothy Wilson, University of Virginia, Charlottesville, VA; received August 17, 2022; accepted January 20, 2023

# *Another Motivating Example*

- ❖ Yes!
- ❖ Hate speech declined after the disruption events.



*Statistical Analysis of Networks*

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# Degree Centrality

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# Learning Goals

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- ❖ At the end of the lecture, you should be able to answer these questions:
  - ❖ How can we conceptualize “centrality”.
  - ❖ How can we operationalization centrality as “degree”.
  - ❖ How do you calculate degree centrality for undirected and directed graphs?
  - ❖ What are the descriptive properties of degree centrality?

When we say a *node* is “central,”  
what do we mean conceptually?

# Concepts and Operationalization

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- ❖ Speaking generally, network *position* should be interesting and important
  - ❖ As a description of the position of a **node/vertex** as a:
    - ❖ dependent variable (e.g. are taller individuals more likely to be trusted?)
    - ❖ independent variable (e.g. are more popular adolescents more likely to succeed in school?)
  - ❖ And, as a description of an entire **network**:
    - ❖ Is this needle-sharing network hierarchical or decentralized?

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# Conceptualization

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- ❖ “Everyone agrees, it seems, that centrality is an important structural attribute of networks. All concede that it is related to a high degree to other important group properties and processes. But there consensus ends.” (Freeman, 1978/1979: 217)
- ❖ The type of measure we use depends on the substantive question of interest.
  - ❖ Various measures of centrality are correlated, but they operationalize different concepts.

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# Conceptualization

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- ❖ Concepts and unit of analysis:
  - ❖ Point centrality (degree, betweenness, closeness)
  - ❖ Graph centrality (compactness)

# Undirected Networks

# Degree Centrality: Undirected Binary Graphs

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- ❖ In an undirected binary graph, *actor degree centrality* measures the extent to which a node connects to all other nodes in a social network.
- ❖ In other words, the number of edges incident with a node.
  - ❖ This is symbolized as:  $d(n_i)$
  - ❖ For an undirected binary graph, the degree  $d(n_i)$  is the row or column sum.

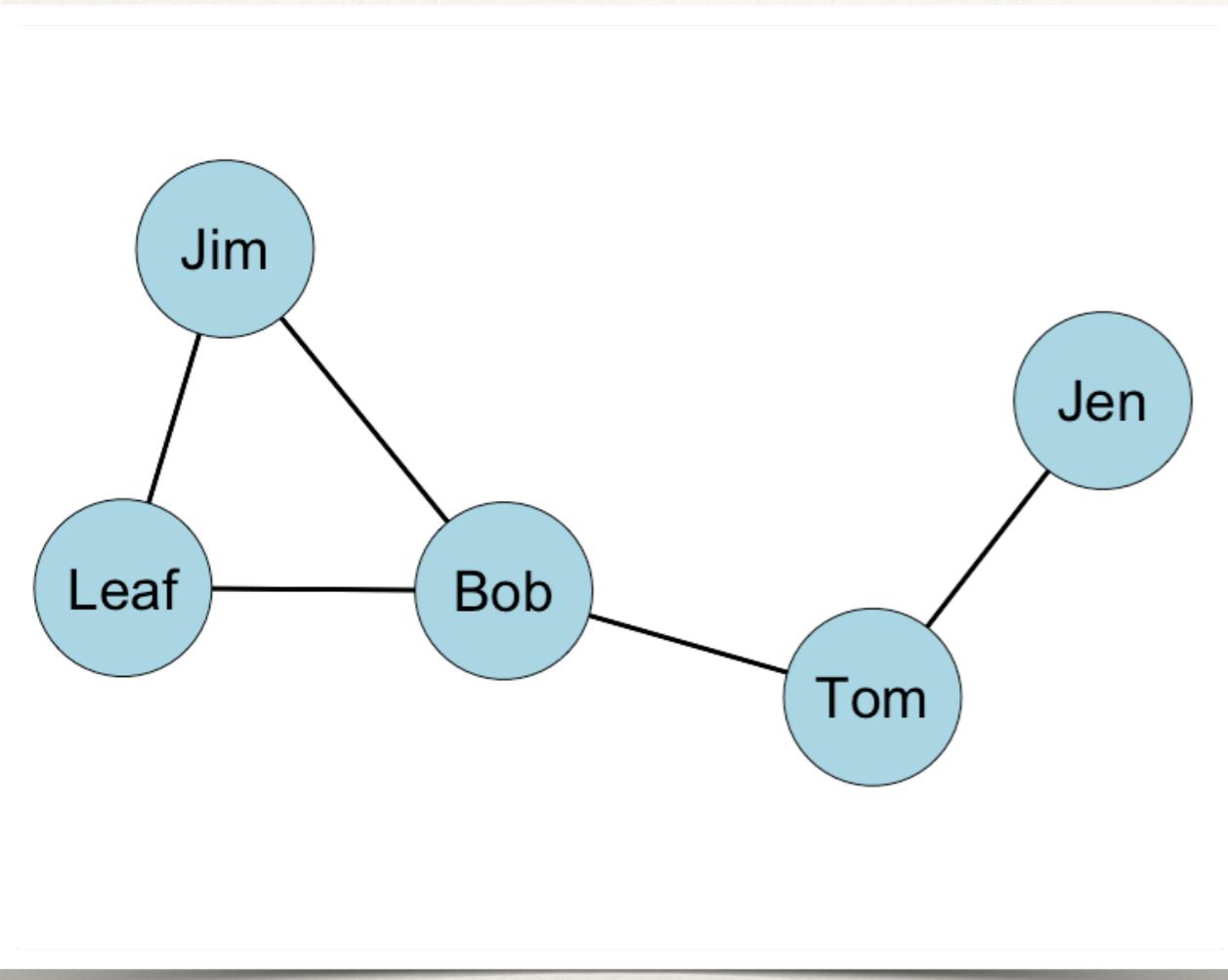
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# Degree Centrality: Undirected Binary Graphs

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$$C_D(n_i) = d(n_i) = \sum_j x_{ij} = \sum_j x_{ji}$$

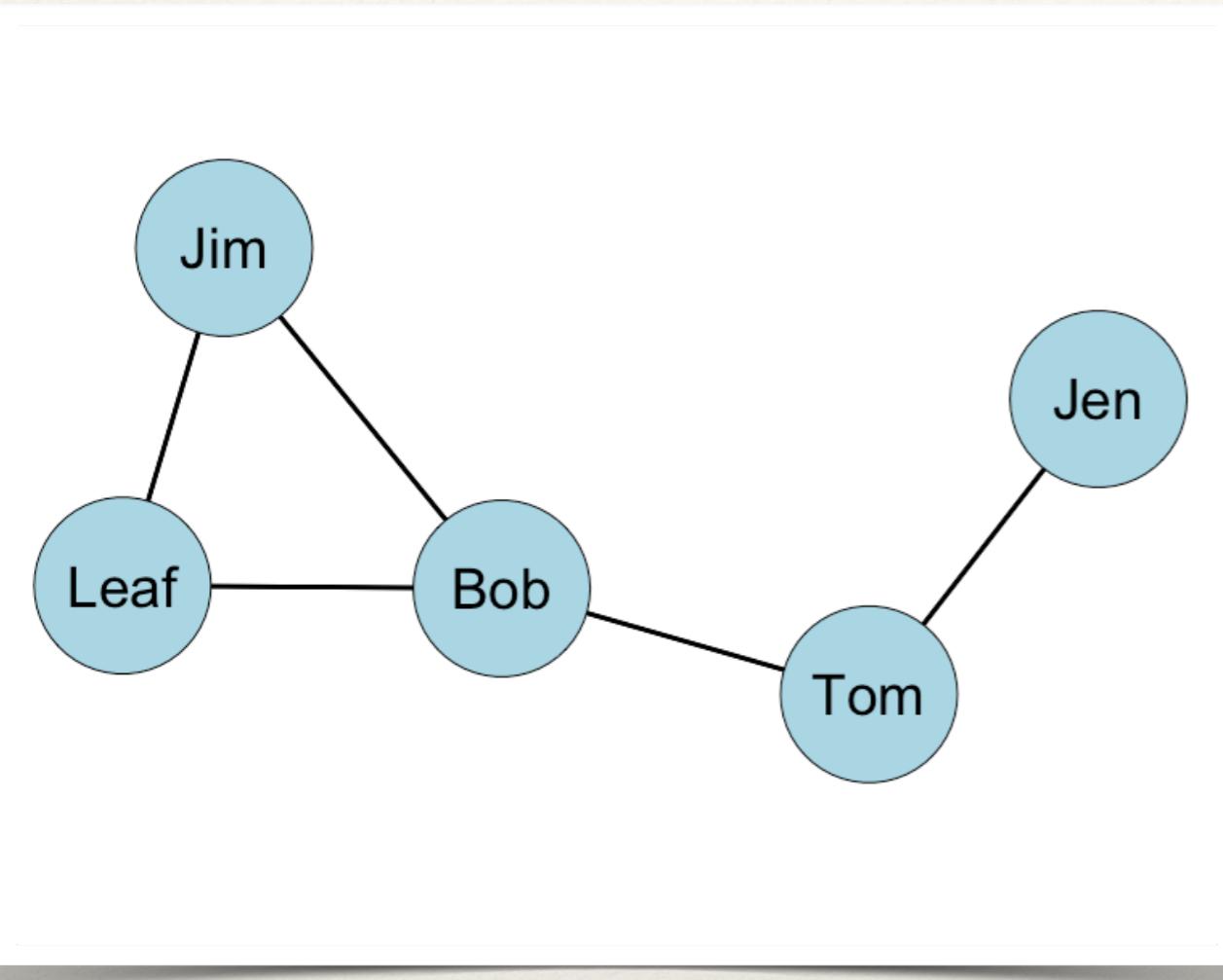
# Example: Undirected, Binary Network



*What is the degree for each node in this graph?*

	Jen	Tom	Bob	Leaf	Jim
Jen	0	1	0	0	0
Tom	1	0	1	0	0
Bob	0	1	0	1	1
Leaf	0	0	1	0	1
Jim	0	0	1	1	0

# Example: Undirected, Binary Network



	Jen	Tom	Bob	Leaf	Jim
Jen	0	1	0	0	0
Tom	1	0	1	0	0
Bob	0	1	0	1	1
Leaf	0	0	1	0	1
Jim	0	0	1	1	0

Note that the column sum and row sum are the same.

# Degree Centrality: Undirected Binary Graphs

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- ❖ Actor degree centrality not only reflects each node's connectivity to other nodes but also depends on the size of the network,  $g$ .
- ❖ Larger networks will have a higher maximum possible degree centrality value.
  - ❖ *Solution?*

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# Standardized Degree Centrality: Undirected Binary Graphs

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- ❖ Standardize!
  - ❖ Take into account the number of nodes and the maximum possible nodes to which  $i$  could be connected.
  - ❖ That is,  $g-1$ .

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# Standardized Degree Centrality: Undirected Binary Graphs

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$$C'_D(n_i) = \frac{d(n_i)}{g - 1} = \frac{\sum_j x_{ij}}{g - 1}$$

## Standardized Degree Centrality: Undirected Binary Graphs

$$C'_D(n_i) = \frac{d(n_i)}{g - 1} = \frac{\sum_j x_{ij}}{g - 1}$$



Why do we subtract 1?

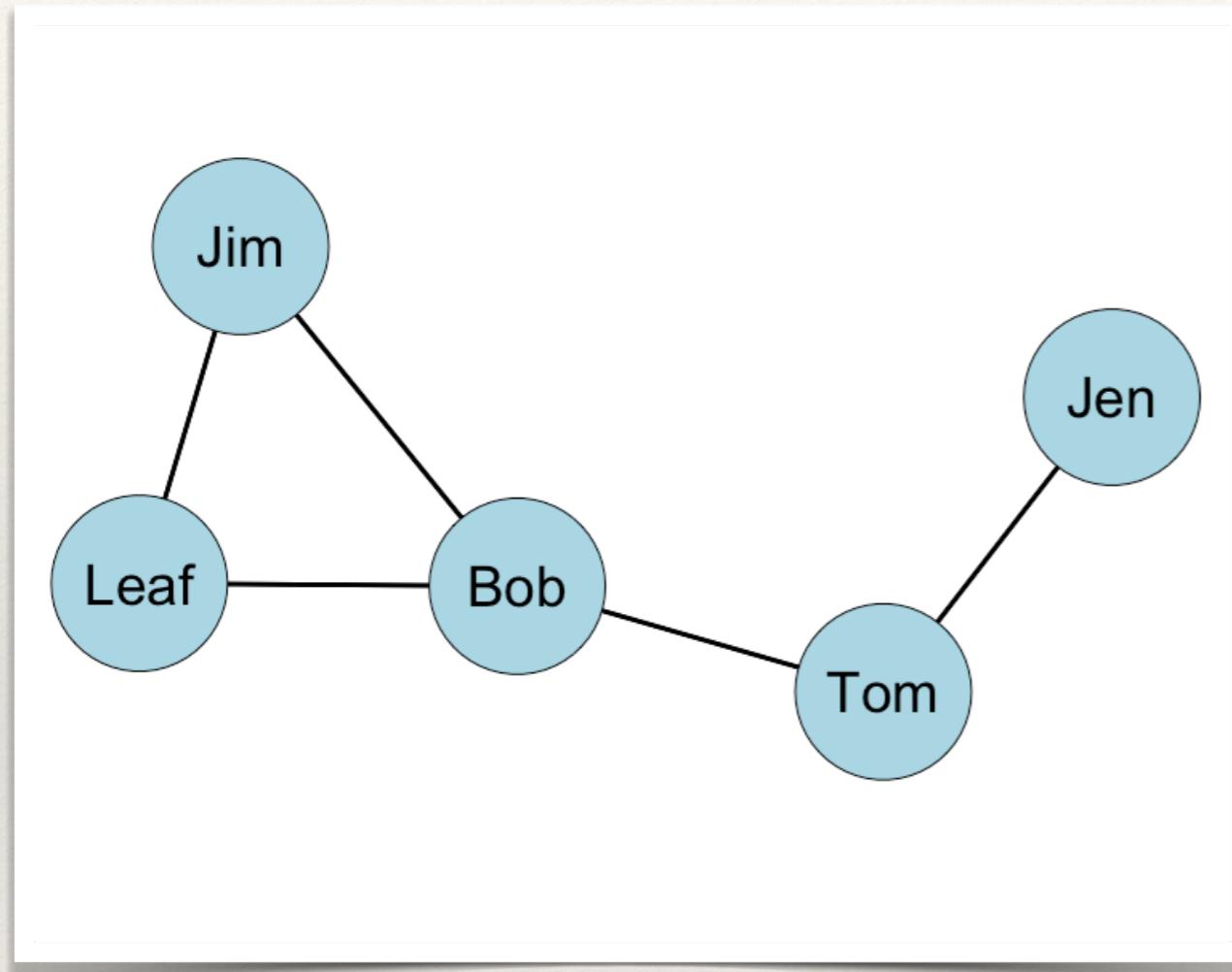
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## Standardized Degree Centrality: Undirected Binary Graphs

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- ❖ This yields the proportion of the network members with ties to actor  $i$ .
  - ❖ This varies between 0 (no connections; isolate) to 1 (ties to every actor).

# Example: Undirected, Binary Network



## Raw Degree Centrality

Jen = 1

Tom = 2

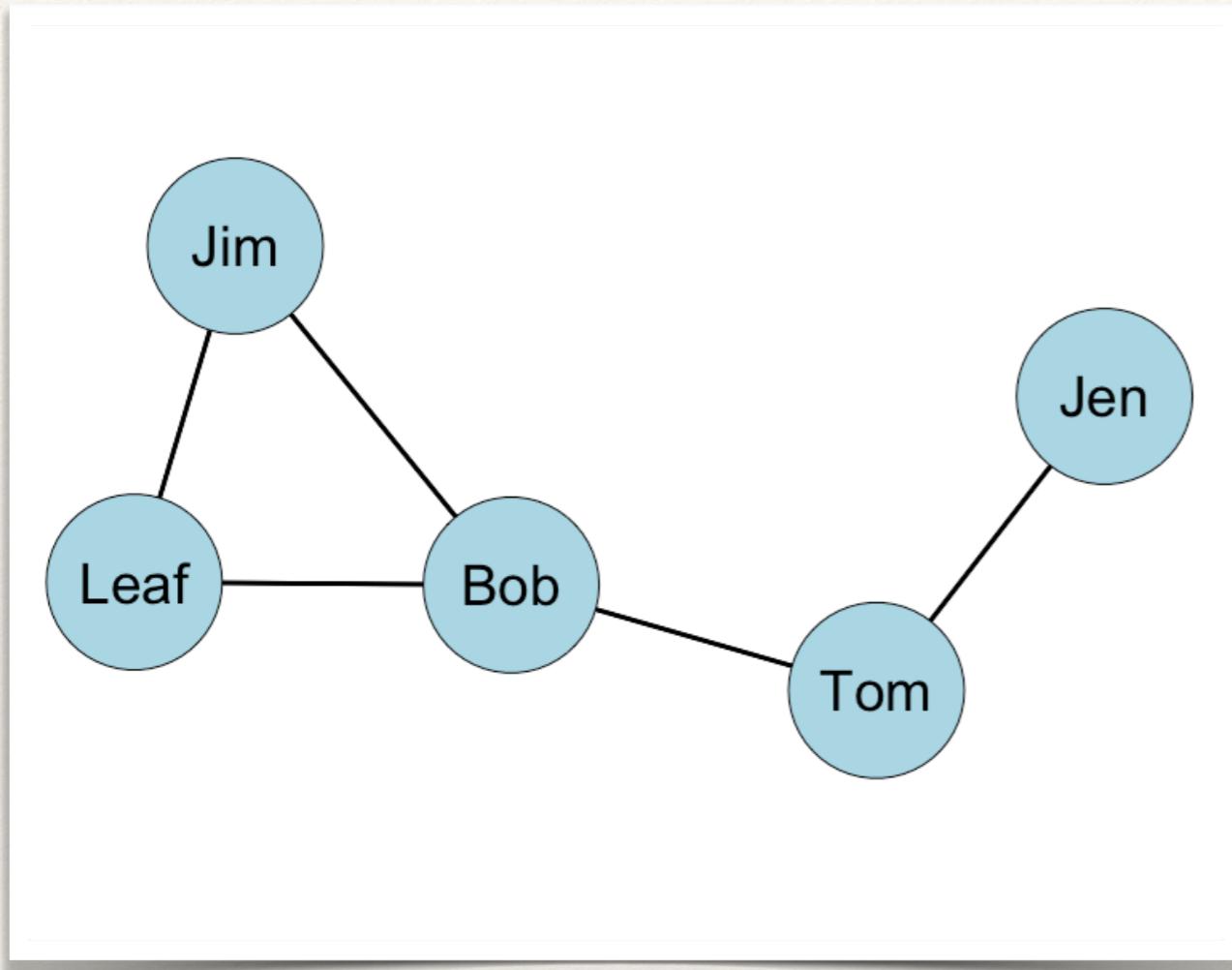
Bob = 3

Leaf = 2

Jim = 2

*What is the standardized degree centrality score for each node?*

# Example: Undirected, Binary Network



Standardized Degree  
Centrality

$$\text{Jen} = 1/4 = 0.25$$

$$\text{Tom} = 2/4 = 0.50$$

$$\text{Bob} = 3/4 = 0.75$$

$$\text{Leaf} = 2/4 = 0.50$$

$$\text{Jim} = 2/4 = 0.50$$

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# Summarizing Degree Centrality

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- ❖ We can examine the summary statistics for degree centrality by inspecting the **mean**.

# Mean Degree (undirected)

Sum up the  
degrees for each  
actor

$$\bar{d} = \frac{\sum_{i=1}^g d(n_i)}{g} = \frac{2L}{g}$$

Divide by number  
of actors

# Mean Degree (undirected)

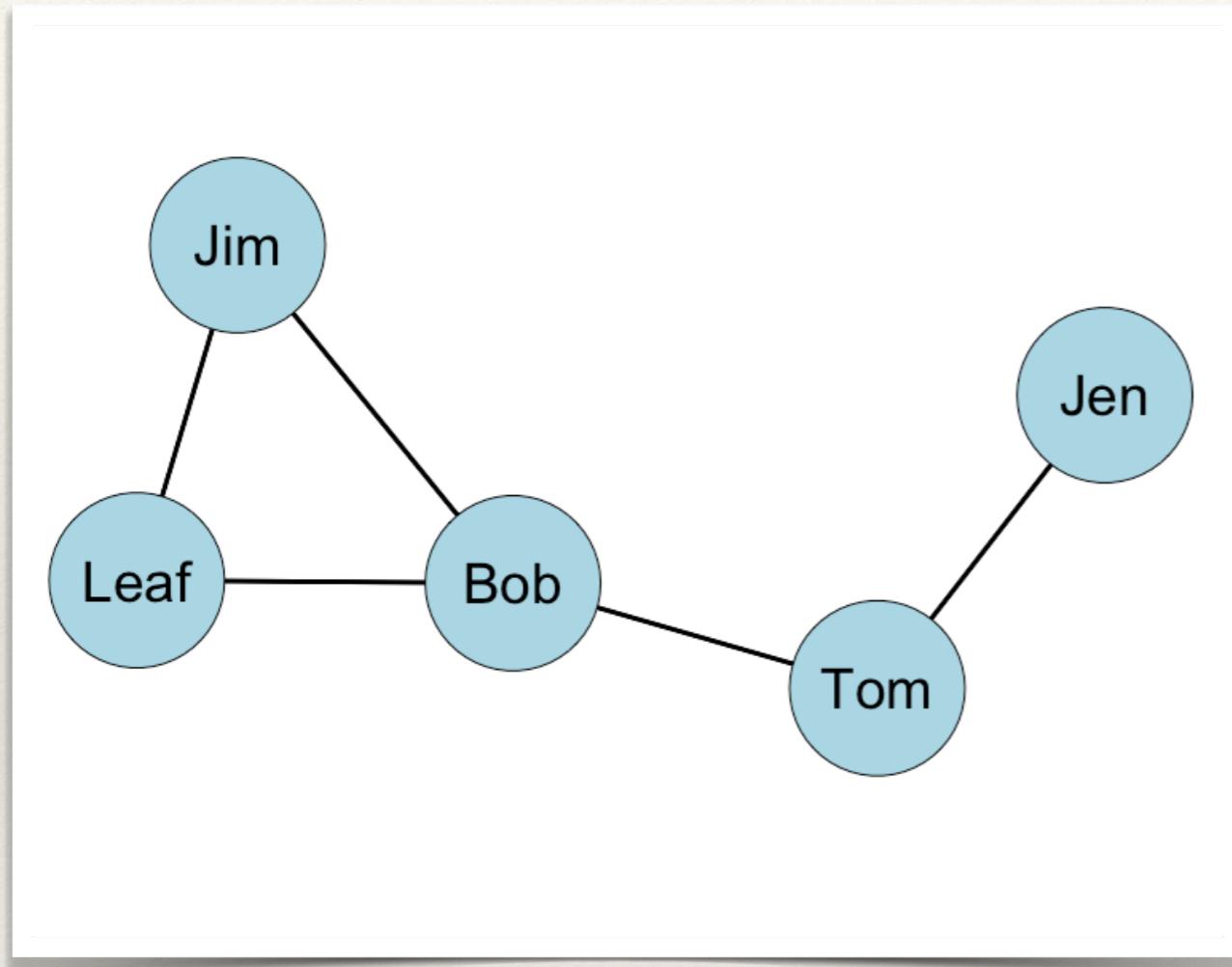
$$\bar{d} = \frac{\sum_{i=1}^g d(n_i)}{g} = \frac{2L}{g}$$

Or, multiply the  
number of edges  
by 2.

$$\frac{2L}{g}$$

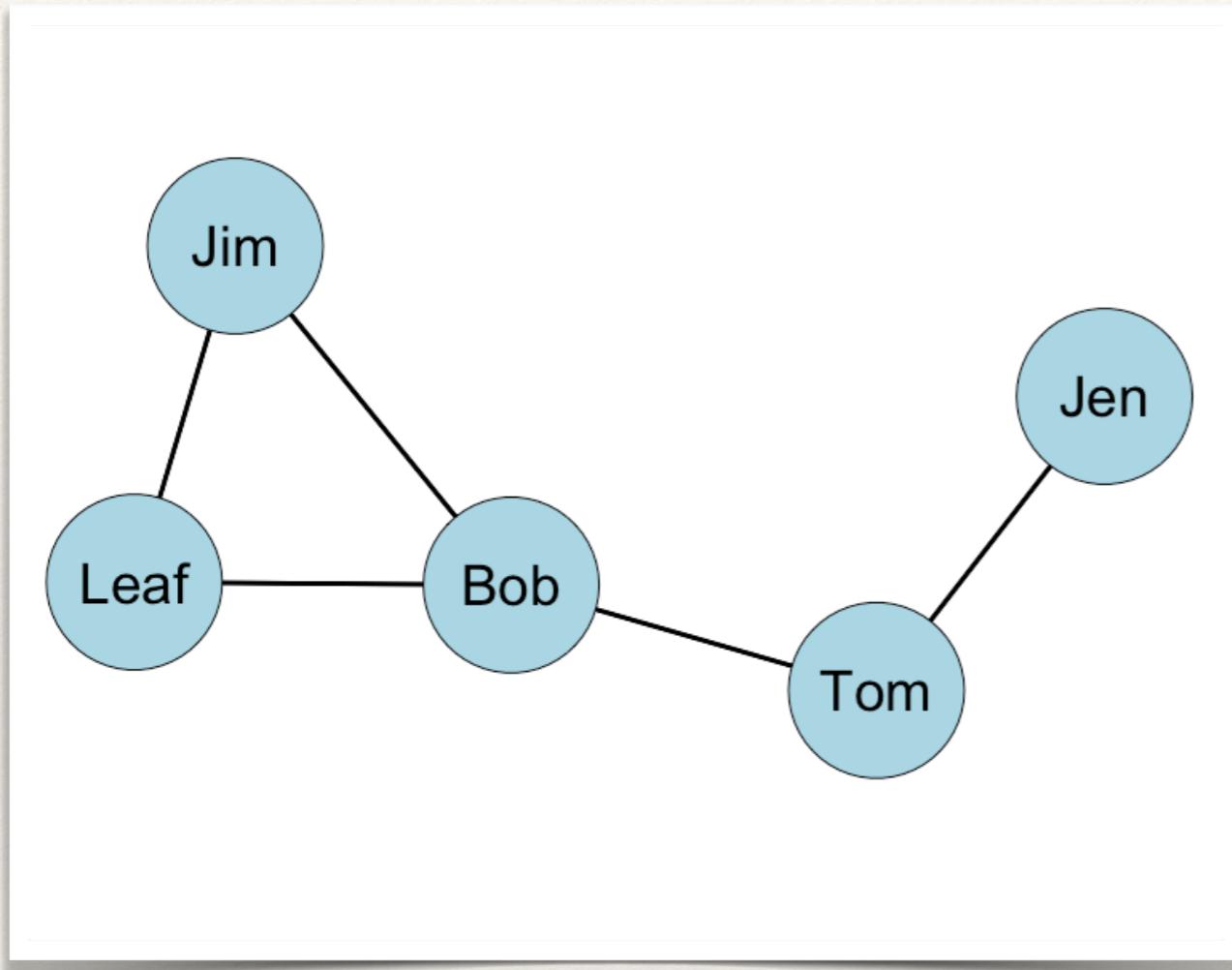
↑  
Divide by number  
of actors

# Example: Undirected, Binary Network



*What is the mean degree for this graph?*

# Example: Undirected, Binary Network



$$\bar{d} = \frac{\sum_{i=1}^g d(n_i)}{g} = \frac{2L}{g} = \frac{2 * 5}{5} = \frac{10}{5} = 2$$

*What is the mean degree for this graph?*

# Summarizing Degree Centrality

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- ❖ The average degree is an important property of a network.
  - ❖ *Why?*
    - ❖ What does a network with a high average degree look like? A low average degree?
      - ❖ **Draw a picture of each...(I will wait)**

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# Summarizing Degree Centrality

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- ❖ Note that degrees form a distribution.
  - ❖ The average degree tells us the central tendency of that distribution.
- ❖ What is another way we can describe a distribution?
  - ❖ (Hint: think back to your stats courses)

# Summarizing Degree Centrality

- ❖ Dispersion!
- ❖ We can describe the dispersion in the centrality scores.
  - ❖ In sna, this is referred to as *centralization*.
- ❖ *Group degree centralization* measures the extent to which the actors in a social network differ from one another in their individual degree centralities.
  - ❖ This should sound familiar, what does the standard deviation tell you?

# Standard Deviation

$$SD_y = \sqrt{\frac{\sum (y_i - \mu)^2}{N}}$$



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# Group Degree Centralization

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$$C_A = \frac{\sum_{i=1}^g [C_A(n^*) - C_A(n_i)]}{\max \sum_{i=1}^g [C_A(n^*) - C_A(n_i)]}$$

# Group Degree Centralization

$$C_A = \frac{\sum_{i=1}^g [C_A(n^*) - C_A(n_i)]}{\max \sum_{i=1}^g [C_A(n^*) - C_A(n_i)]}$$

Largest actor degree centrality scored observed

Degree centrality for actor  $i$

# Group Degree Centralization

$$C_A = \frac{\sum_{i=1}^g [C_A(n^*) - C_A(n_i)]}{\max \sum_{i=1}^g [C_A(n^*) - C_A(n_i)]}$$

Sum of observed differences between the largest actor centrality and all others

Theoretical maximum possible sum of those differences

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# Summarizing Degree Centrality

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- ❖ Note that this is a generic measure (thanks Freeman, 1979!)
  - ❖ We can calculate the denominator as  $(g-1)(g-2)$  (Thanks Wasserman & Faust, 1994!)

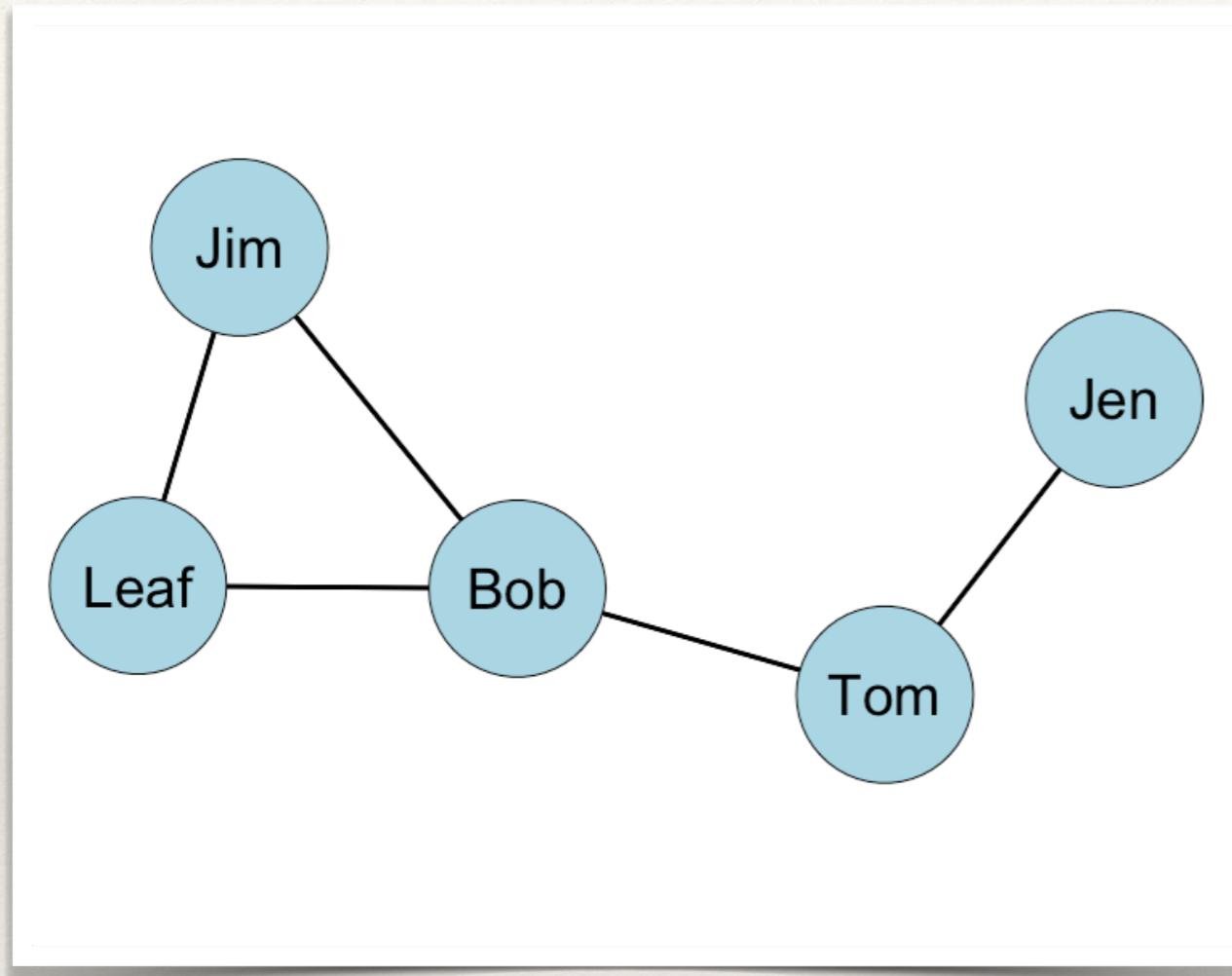
# Index of Group Degree Centralization

$$C_D = \frac{\sum_{i=1}^g [C_D(n^*) - C_D(n_i)]}{[(g-1)(g-2)]}$$

Sum of observed differences between the largest actor centrality and all others

The maximum possible sum of differences

# Example: Undirected, Binary Network



## Raw Degree Centrality

Jen = 1

Tom = 2

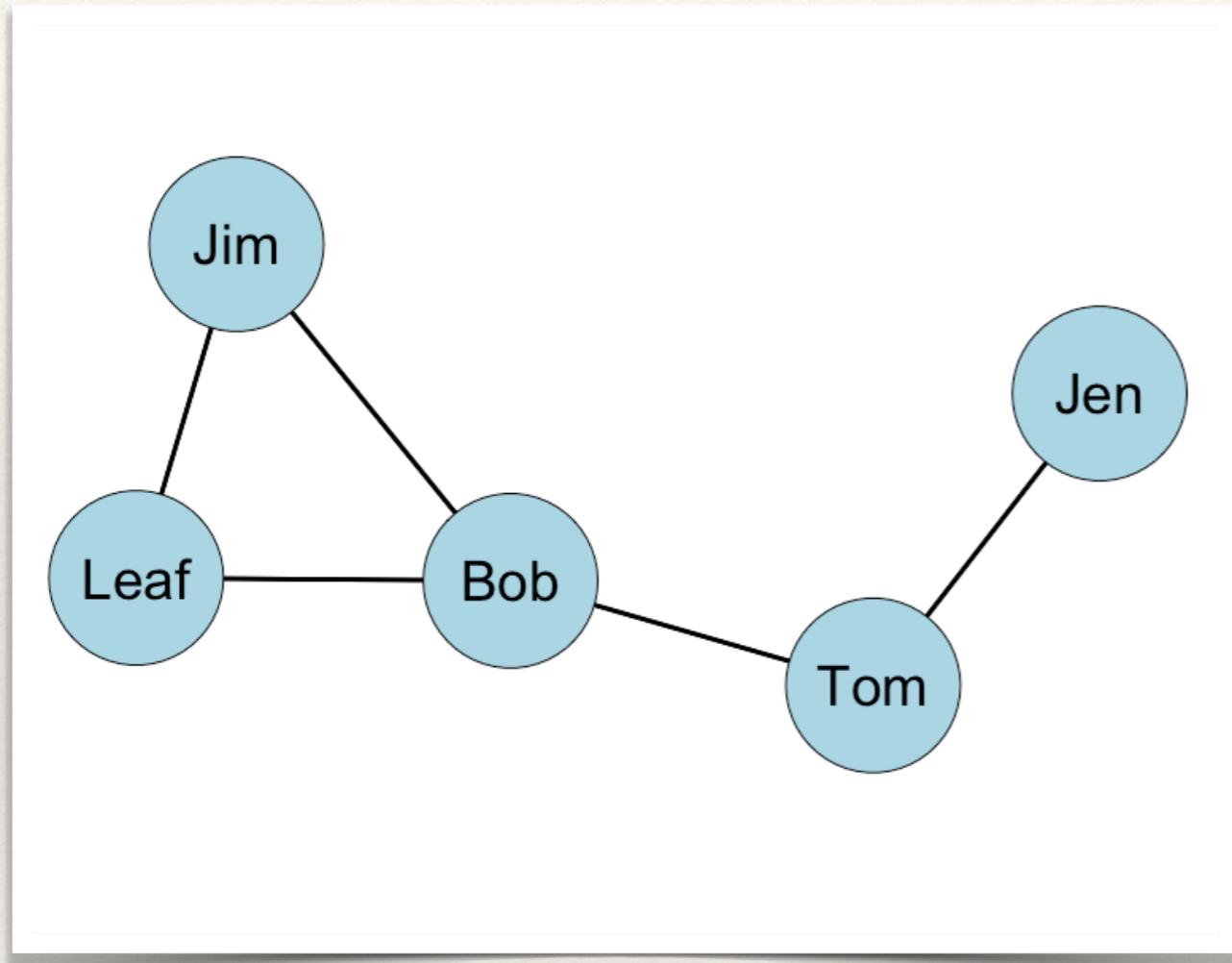
Bob = 3

Leaf = 2

Jim = 2

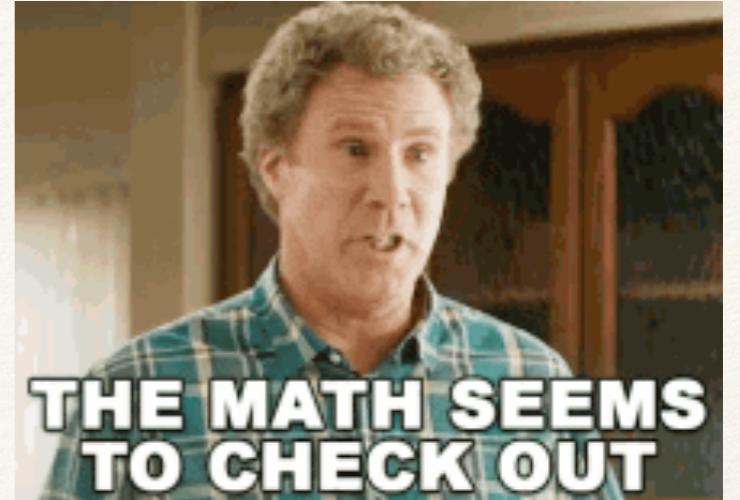
*What is the index of degree centralization for this graph?*

# Example: Undirected, Binary Network



0.4167

$$C_D = \frac{\sum_{i=1}^g [C_D(n^*) - C_D(n_i)]}{[(g-1)(g-2)]}$$



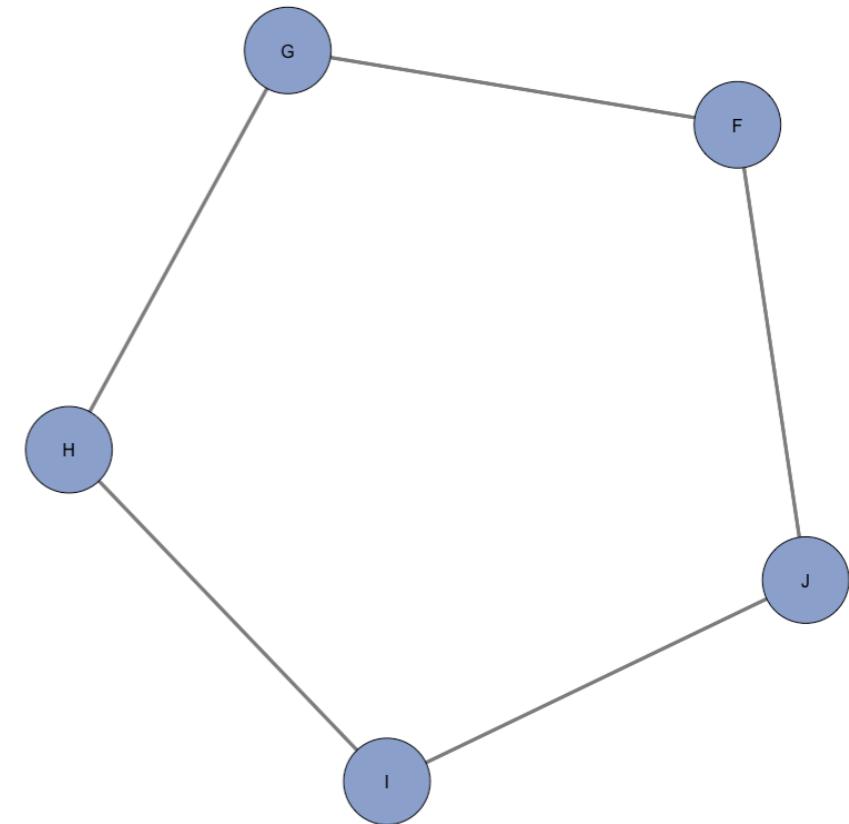
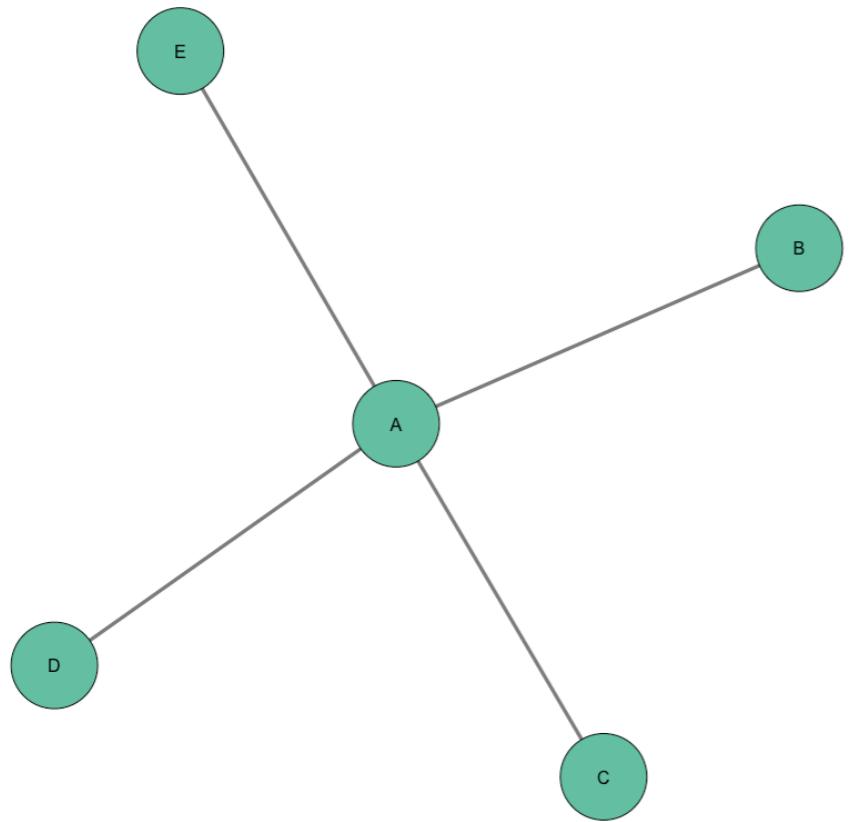
$$= \frac{(3-1) + (3-2) + (3-3) + (3-2) + (3-2)}{(5-1)(5-2)}$$

$$= \frac{2+1+0+1+1}{4*3} = \frac{5}{12} = 0.4167$$

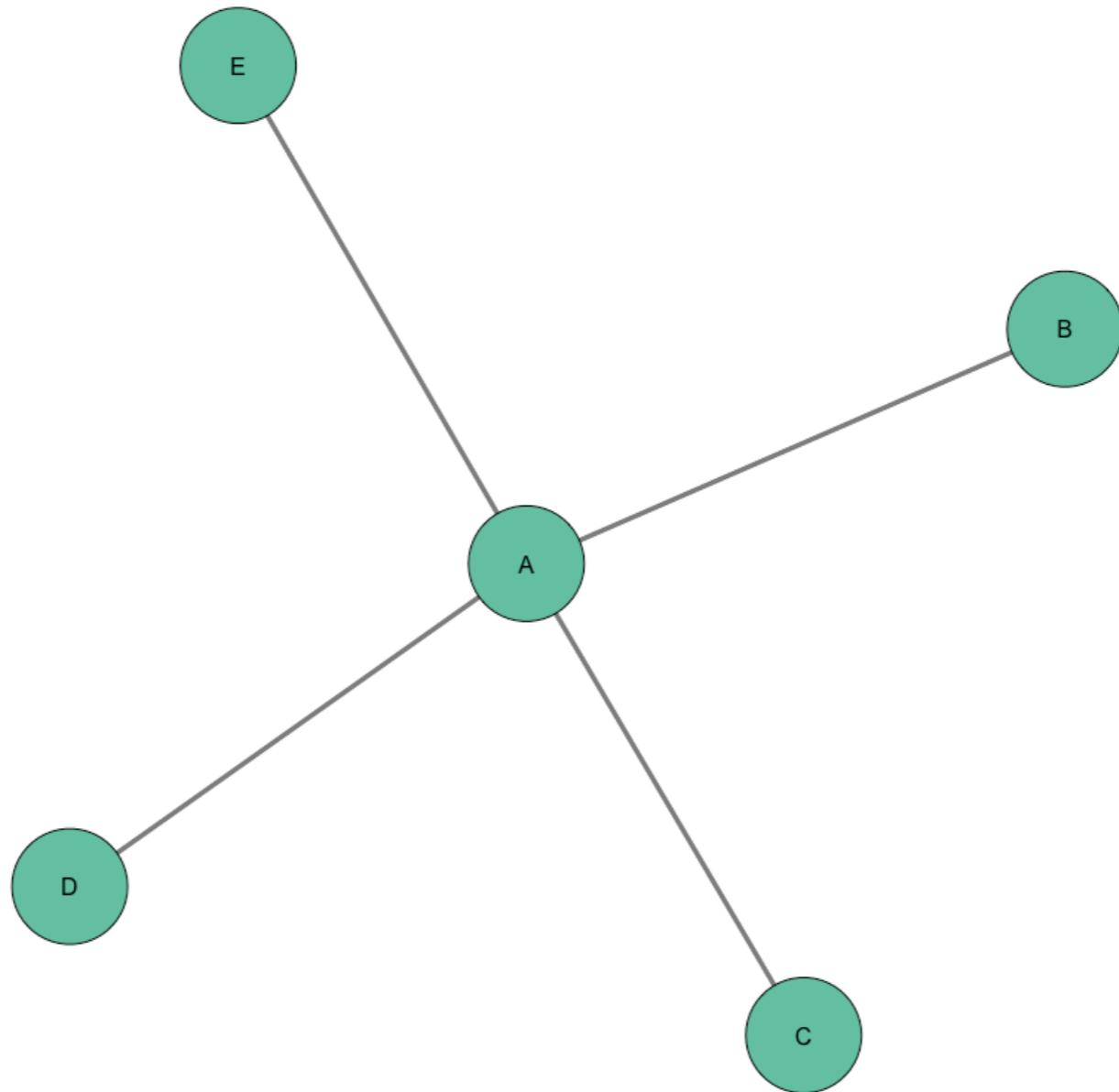
# Summarizing Degree Centrality

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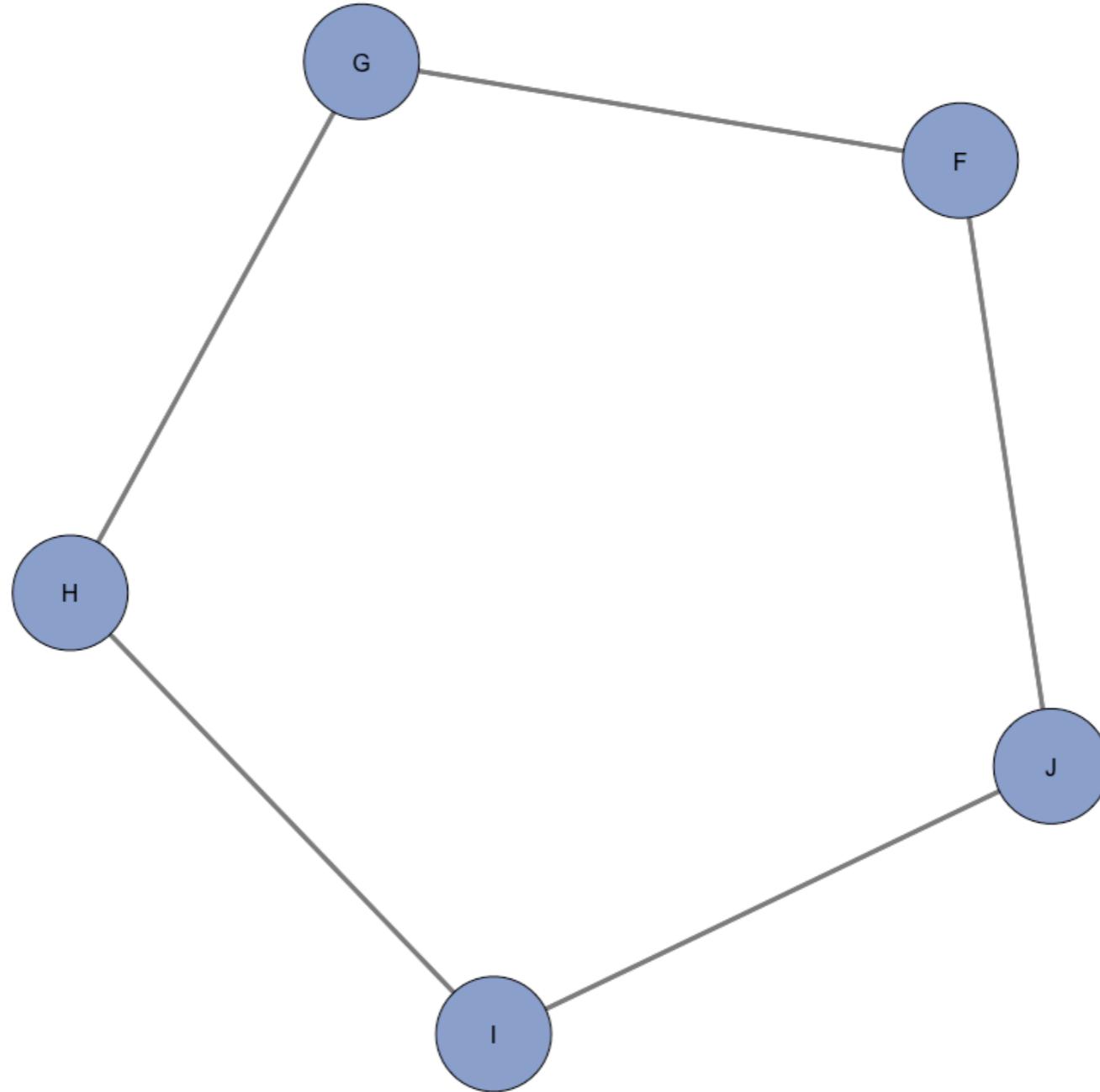
- ❖ When degree centrality is evenly dispersed, the numerator will be zero, and the quotient will be close to 0.
- ❖ When there is considerable inequality in the actor degrees, the quotient will be closer to 1.
  - ❖ Thus, closer to 1 indicates that the graph is hierarchically structured.



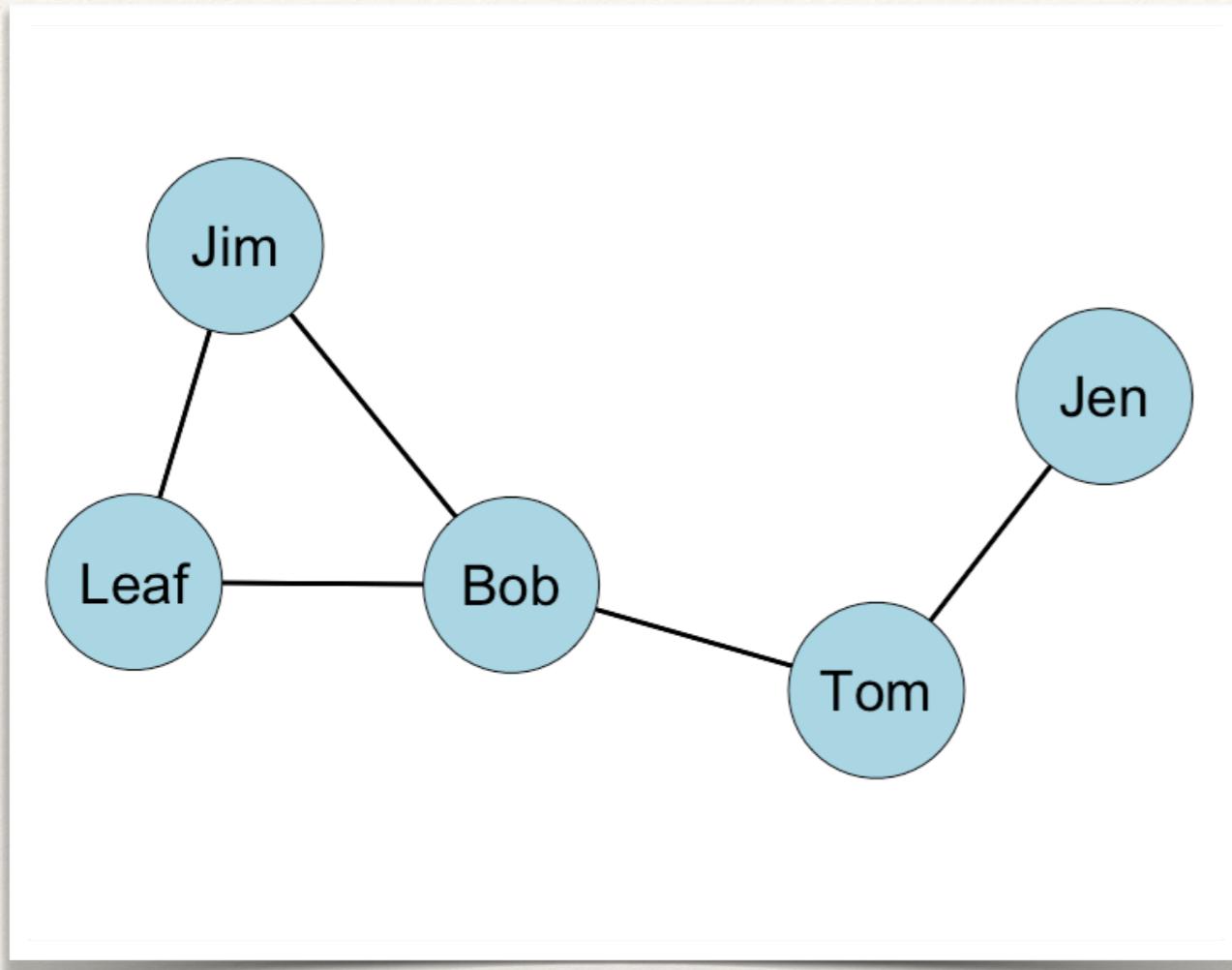
$$C_D = \frac{\sum_{i=1}^g [C_D(n^*) - C_D(n_i)]}{[(g-1)(g-2)]} = \frac{(4-4) + (4-1) + (4-1) + (4-1) + (4-1)}{(5-1)(5-2)} = \frac{0 + 3 + 3 + 3 + 3}{4 * 3} = \frac{12}{12} = 1.0$$



$$C_D = \frac{\sum_{i=1}^g [C_D(n^*) - C_D(n_i)]}{[(g-1)(g-2)]} = \frac{(2-2) + (2-2) + (2-2) + (2-2) + (2-2)}{(5-1)(5-2)} = \frac{0+0+0+0+0}{4*3} = \frac{0}{12} = 0.0$$



# Example: Undirected, Binary Network



*How should we interpret  
this value?*

0.4167

# Directed Networks

# Degree Centrality: Directed Binary Graphs

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- ❖ In a directed binary graph, *actor degree centrality* can be broken down into indegree and outdegree centrality.
  - ❖ **Indegree**,  $C_I(n_i)$ , measures the number of ties that  $i$  receives.
    - ❖ For the sociomatrix  $X_{ij}$ , the indegree for  $i$  is the column sum.
  - ❖ **Outdegree**,  $C_O(n_i)$ , measures the number of ties that  $i$  sends.
    - ❖ For the sociomatrix  $X_{ij}$ , the outdegree for  $i$  is the row sum.

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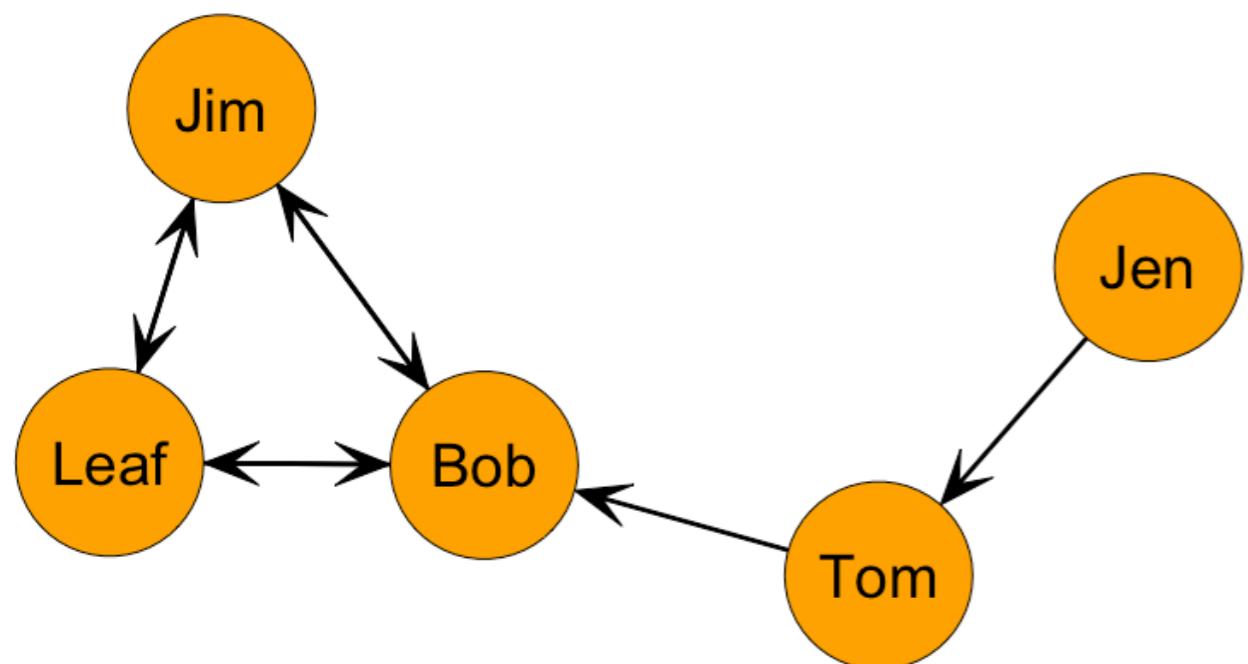
# Degree Centrality: Directed Binary Graphs

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$$C_I(n_i) = \sum_j x_{ji}$$

$$C_O(n_i) = \sum_j x_{ij}$$

# Example: Directed, Binary Network



	Jen	Tom	Bob	Leaf	Jim
Jen	0	1	0	0	0
Tom	0	0	1	0	0
Bob	0	0	0	1	1
Leaf	0	0	1	0	1
Jim	0	0	1	1	0

*What is the indegree and outdegree  
for each node in the graph?*

# Example: Directed, Binary Network

*NOTE: These both sum to the same value*

## Raw Indegree Centrality

Jen = 0

Tom = 1

Bob = 3

Leaf = 2

Jim = 2

**TOTAL: 8**

## Raw Outdegree Centrality

Jen = 1

Tom = 1

Bob = 2

Leaf = 2

Jim = 2

**TOTAL: 8**



# Degree Centrality: Directed Binary Graphs

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- ❖ Recall that actor degree centrality not only reflects each node's connectivity to other nodes but also depends on the size of the network,  $g$ .
- ❖ Larger networks will have a higher maximum possible degree centrality value.
  - ❖ We can standardize, or normalize, the same way by dividing by  $g-1$ .

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# Standardized Degree Centrality: Directed Binary Graphs

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$$C'_I(n_i) = \frac{C_I(n_i)}{g - 1}$$

$$C'_O(n_i) = \frac{C_O(n_i)}{g - 1}$$

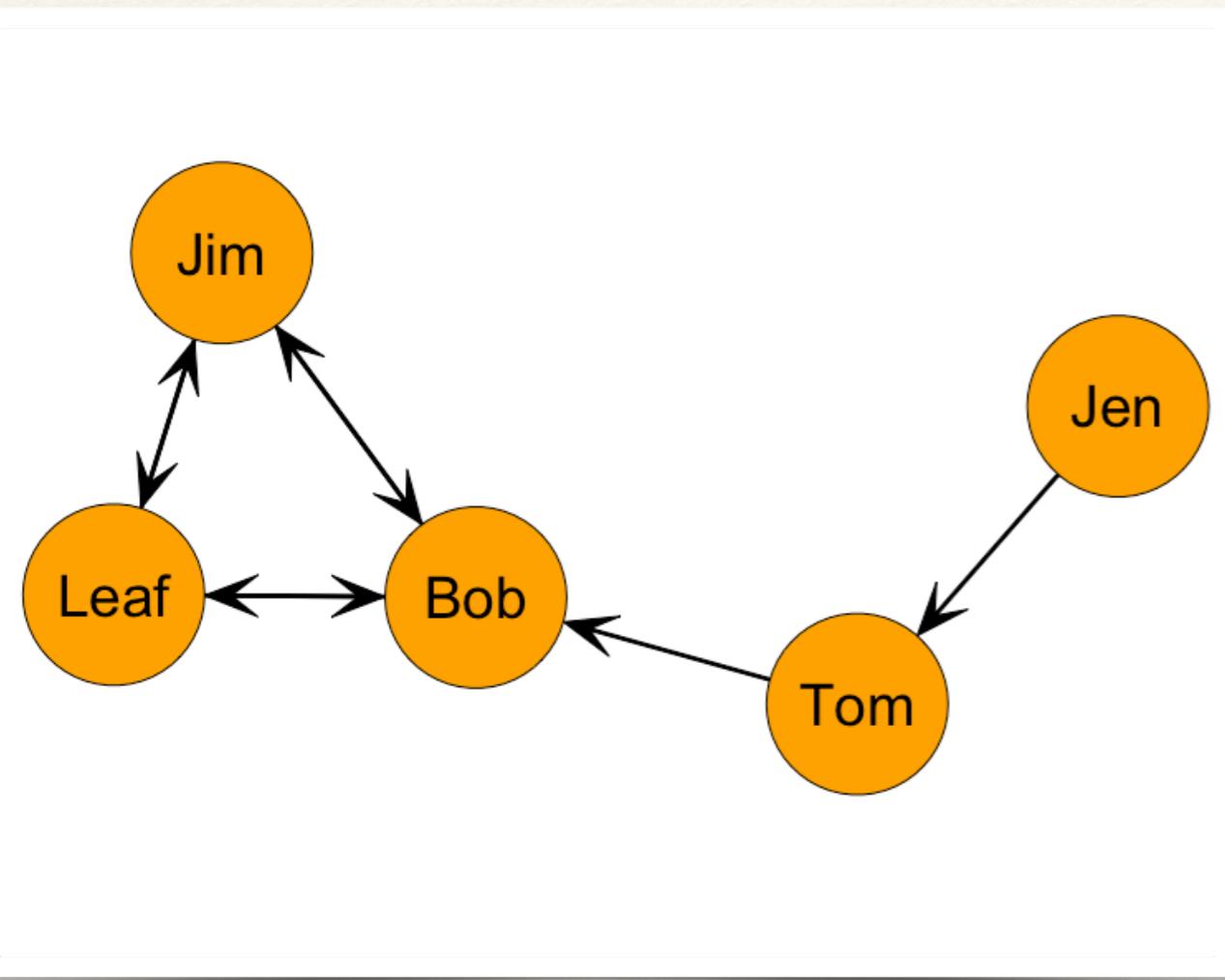
# Standardized Degree Centrality: Directed Binary Graphs

$$C'_I(n_i) = \frac{C_I(n_i)}{g - 1}$$

*Why do we subtract 1?*

$$C'_O(n_i) = \frac{C_O(n_i)}{g - 1}$$

# Example: Directed, Binary Network

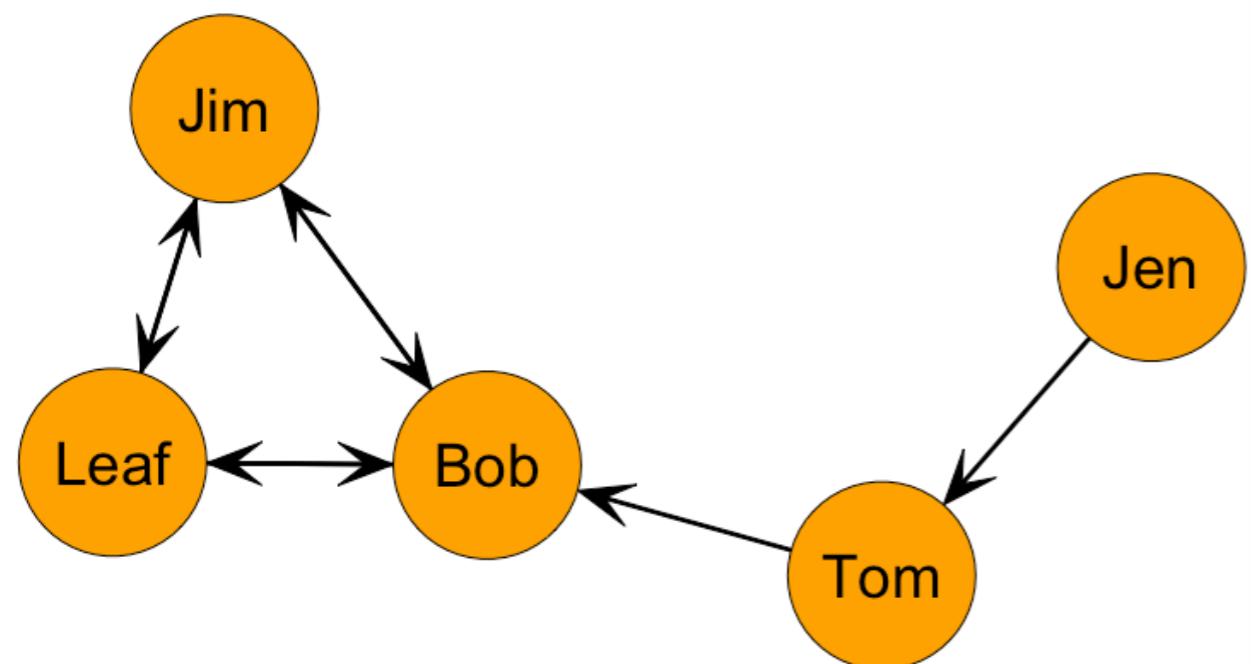


## Raw Indegree Centrality

Jen = 0  
Tom = 1  
Bob = 3  
Leaf = 2  
Jim = 2

*What is the standardized indegree  
and outdegree centrality score for  
each node?*

# Example: Directed, Binary Network



## Standardized Indegree Centrality

$$\text{Jen} = 0/4 = 0$$

$$\text{Tom} = 1/4 = 0.25$$

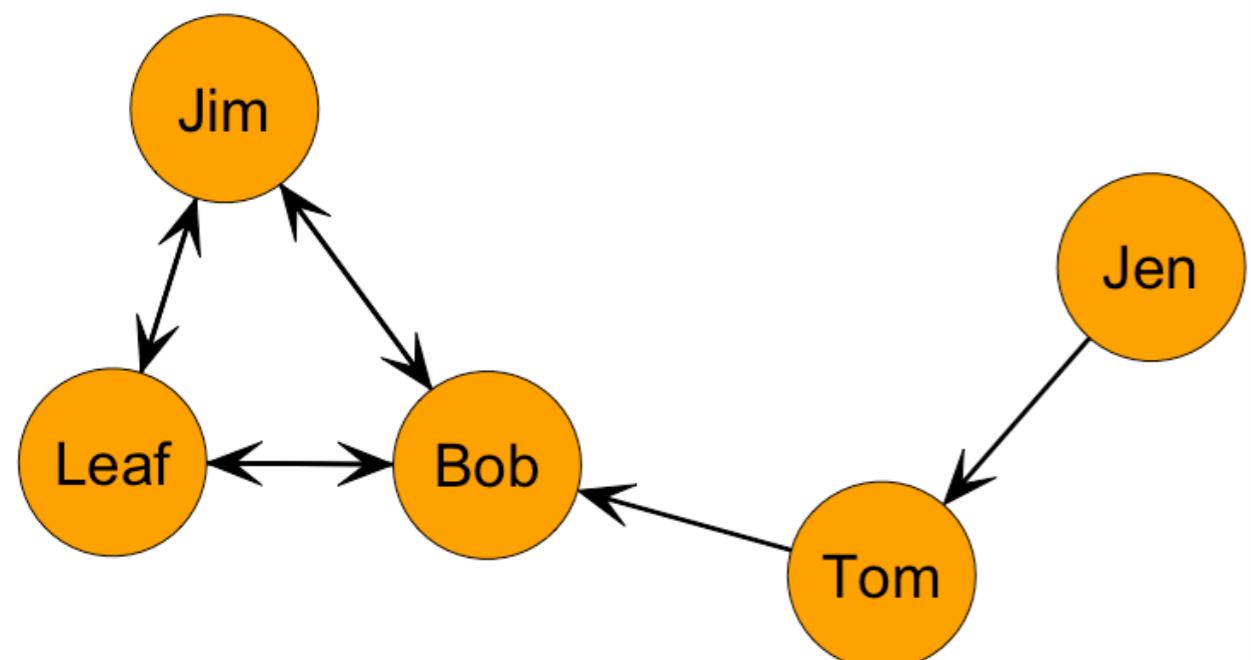
$$\text{Bob} = 3/4 = 0.75$$

$$\text{Leaf} = 2/4 = 0.50$$

$$\text{Jim} = 2/4 = 0.50$$

*What is the standardized indegree  
and outdegree centrality score for  
each node?*

# Example: Directed, Binary Network



*What is the standardized indegree  
and outdegree centrality score for  
each node?*

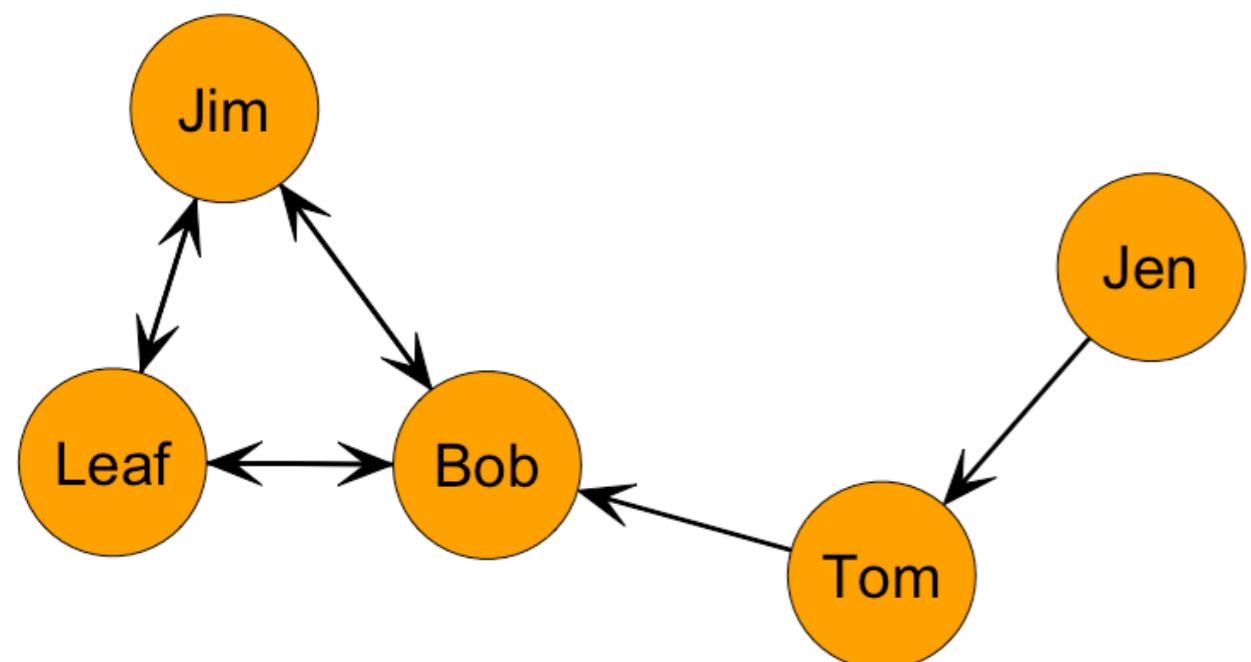
## Raw Indegree Centrality

Jen = 0  
Tom = 1  
Bob = 3  
Leaf = 2  
Jim = 2

## Raw Outdegree Centrality

Jen = 1  
Tom = 1  
Bob = 2  
Leaf = 2  
Jim = 2

# Example: Directed, Binary Network



*What is the standardized indegree and outdegree centrality score for each node?*

## Standardized Indegree Centrality

$$\text{Jen} = 0/4 = 0$$

$$\text{Tom} = 1/4 = 0.25$$

$$\text{Bob} = 3/4 = 0.75$$

$$\text{Leaf} = 2/4 = 0.50$$

$$\text{Jim} = 2/4 = 0.50$$

## Standardized Outdegree Centrality

$$\text{Jen} = 1/4 = 0.25$$

$$\text{Tom} = 1/4 = 0.25$$

$$\text{Bob} = 2/4 = 0.50$$

$$\text{Leaf} = 2/4 = 0.50$$

$$\text{Jim} = 2/4 = 0.50$$

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# Summarizing Degree Centrality

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- ❖ As before, we can examine the summary statistics for degree centrality by inspecting the **mean**.

# Mean Degree (directed)

$$\bar{d} = \frac{\sum_{i=1}^g C_I(n_i)}{g} = \frac{\sum_{i=1}^g C_O(n_i)}{g} = \frac{L}{g}$$

Or, number of edges

Divide by number of actors

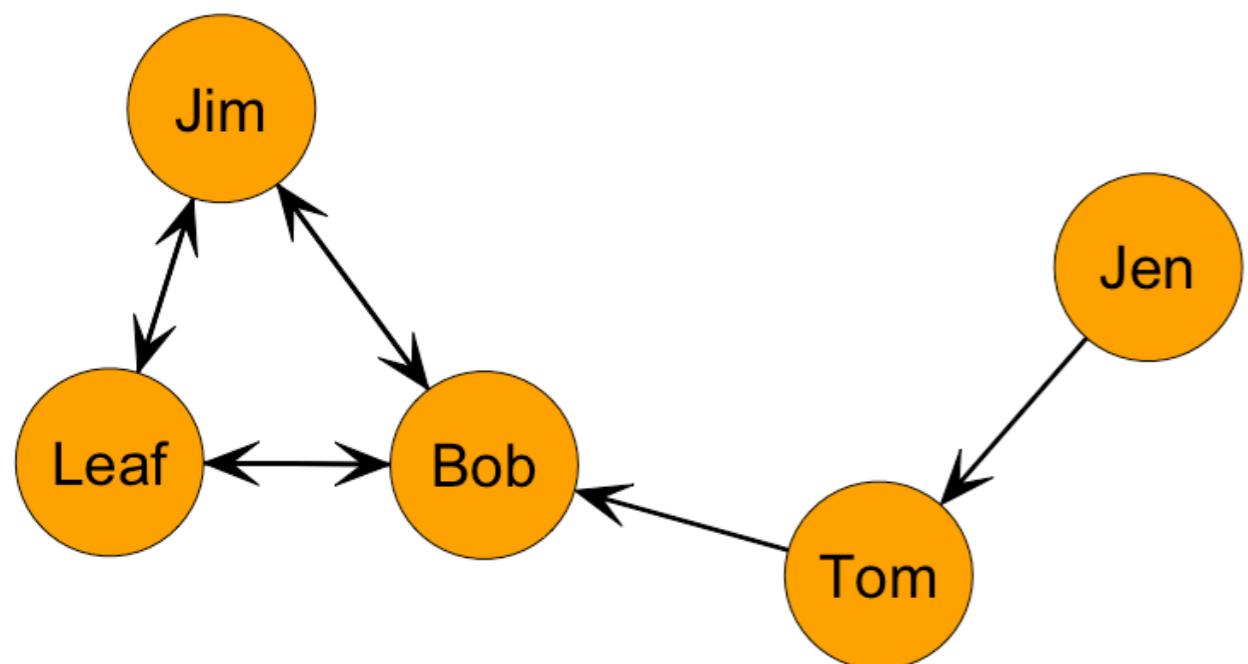
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# Summarizing Degree Centrality

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- ❖ The mean indegree is equal to the mean outdegree.
  - ❖ *Why?*

# Example: Directed, Binary Network



$$\bar{d} = \frac{C_I(n_i)}{g} = \frac{C_O(n_i)}{g} = \frac{L}{g} = \frac{8}{5} = 1.6$$

*What is the mean indegree/  
outdegree for this graph?*

# Summarizing Degree Centrality

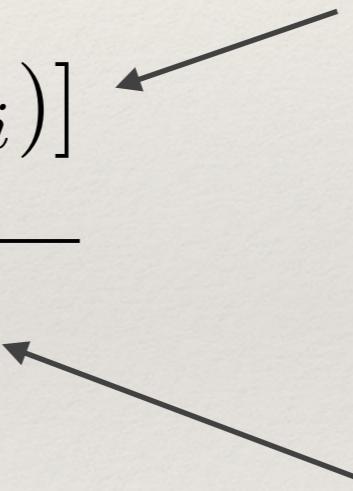
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- ❖ We can also calculate how centralized the graph itself is.
  - ❖ *Group degree centralization* measures the extent to which the actors in a social network differ from one another in their individual degree centralities.
  - ❖ The difference here is that the denominator is  $(g-1)^2$  or  $(g-1)(g-1)$ .
    - ❖ Note that the numerator may differ though for **indegree** and **outdegree**.

# Index of Group Degree Centralization

$$C_D = \frac{\sum_{i=1}^g [C_D(n^*) - C_D(n_i)]}{[(g - 1)^2]}$$

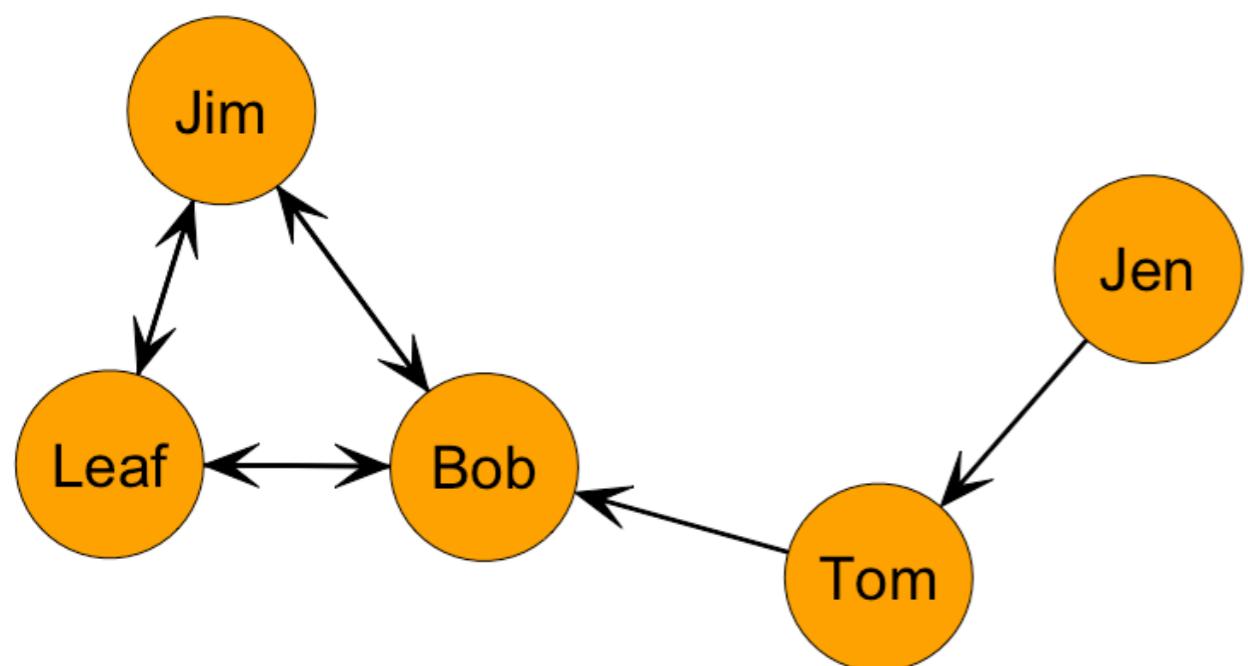
Sum of observed differences between the largest actor centrality and all others



The maximum possible sum of differences

*Note the difference (see W&F p. 199)*

# Example: Directed, Binary Network



## Raw Indegree Centrality

Jen = 0

Tom = 1

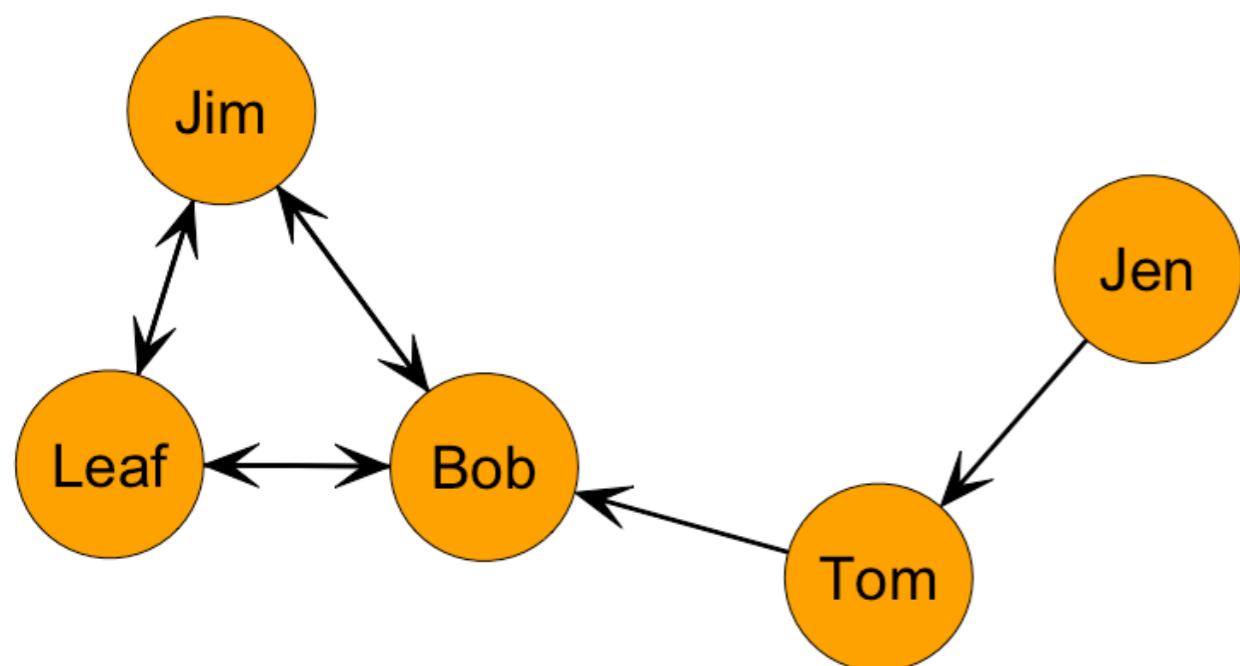
Bob = 3

Leaf = 2

Jim = 2

*What is the index of indegree centralization for this graph?*

# Example: Directed, Binary Network



## Raw Indegree Centrality

Jen = 0

Tom = 1

Bob = 3

Leaf = 2

Jim = 2

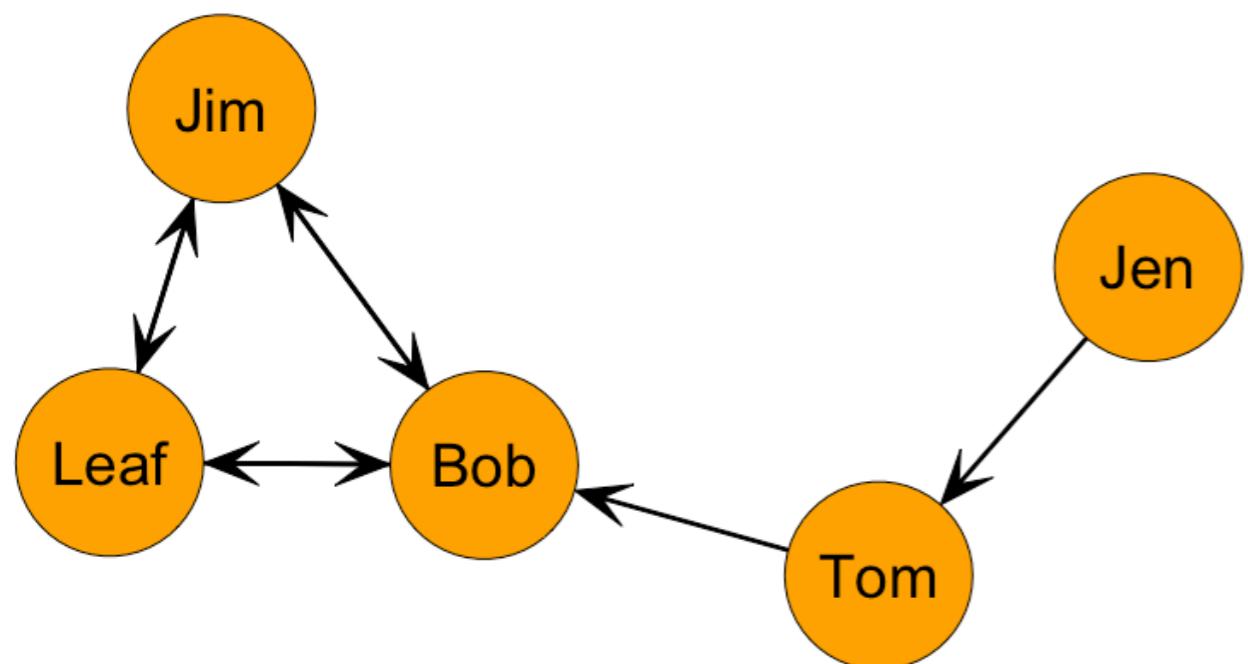
*What is the index of indegree centralization for this graph?*

0.4375

$$C_I = \frac{\sum_{i=1}^g [C_I(n^*) - C_I(n_i)]}{[(g-1)(g-1)]} =$$

$$= \frac{(3-0) + (3-1) + (3-3) + (3-2) + (3-2)}{(5-1)(5-1)} = \frac{3+2+0+1+1}{4*4} = \frac{7}{16} = 0.4375$$

# Example: Directed, Binary Network



## Raw Outdegree Centrality

Jen = 1

Tom = 1

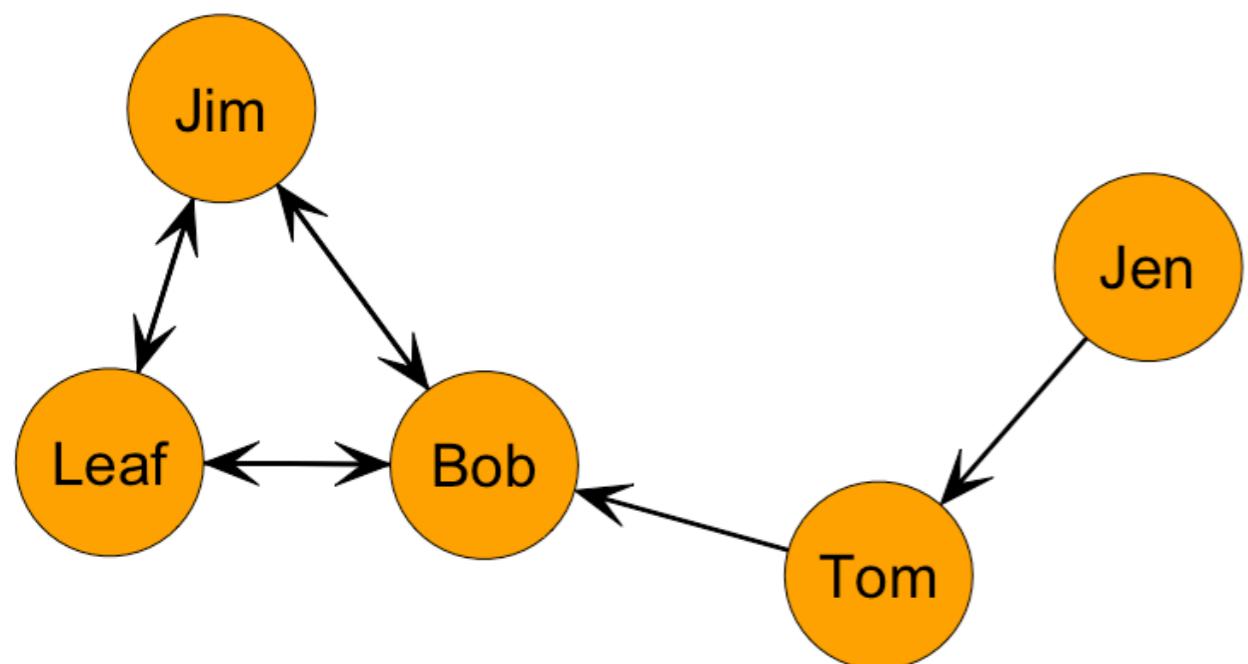
Bob = 2

Leaf = 2

Jim = 2

*What is the index of outdegree centralization for this graph?*

# Example: Directed, Binary Network



## Raw Outdegree Centrality

Jen = 1

Tom = 1

Bob = 2

Leaf = 2

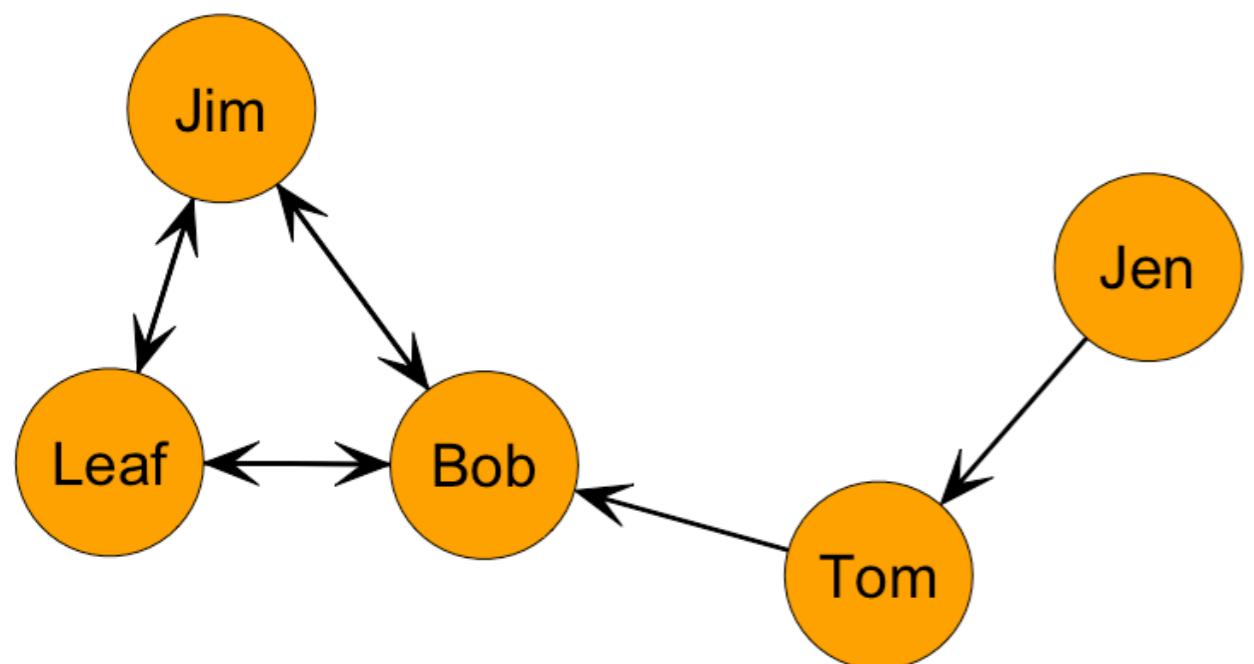
Jim = 2

*What is the index of outdegree centralization for this graph?*

0.125

$$C_O = \frac{\sum_{i=1}^g [C_O(n^*) - C_O(n_i)]}{[(g-1)(g-1)]}$$
$$= \frac{(2-1) + (2-1) + (2-2) + (2-2) + (2-2)}{(5-1)(5-1)} = \frac{1+1+0+0+0}{4*4} = \frac{2}{16} = 0.125$$

# Example: Directed, Binary Network



$$C_I = 0.4375$$

$$C_O = 0.125$$

*What do the differences in the centralization scores tell us about the graph?*

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# Learning Goals

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- ❖ At the end of the lecture, you should be able to answer these questions:
  - ❖ How can we conceptualize “centrality”.
  - ❖ How can we operationalization centrality as “degree”.
  - ❖ How do you calculate degree centrality for undirected and directed graphs?
  - ❖ What are the descriptive properties of degree centrality?

Questions?