

*Statistical Analysis of Networks*

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# Projection & Weighted Graphs

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# Motivating Example *Revisited*

## **Diffusion of Ideas and Technology: The Role of Networks in Influencing the Endorsement and Use of On-Officer Video Cameras**

Journal of Contemporary Criminal Justice

2015, Vol. 31(3) 243–261

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**Jacob T. N. Young<sup>1</sup> and Justin T. Ready<sup>1</sup>**

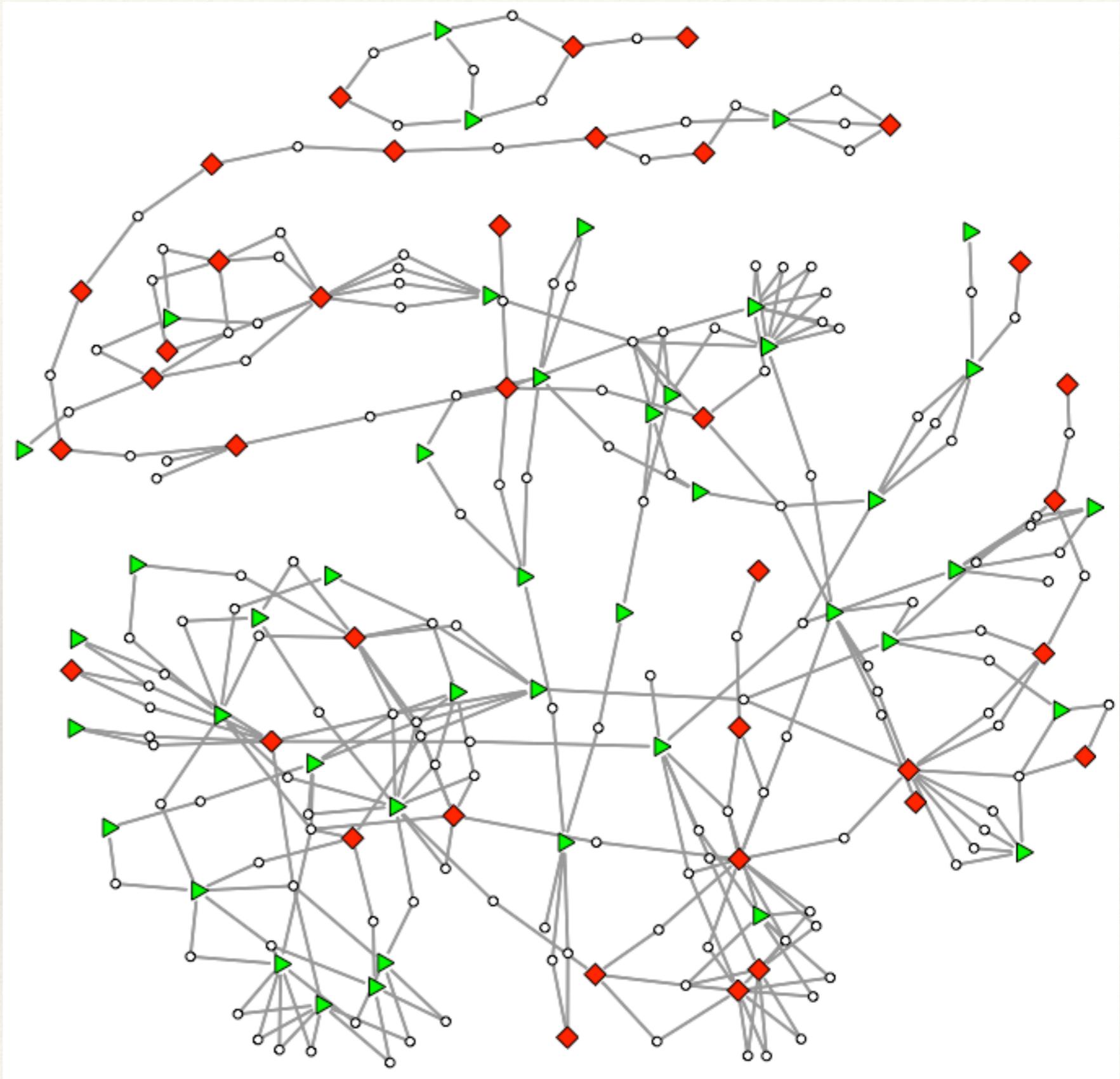
- ❖ Questions:

- ❖ How do police officers “frame” body-worn cameras?
- ❖ Is the meaning officers attribute to cameras created and transmitted in groups?

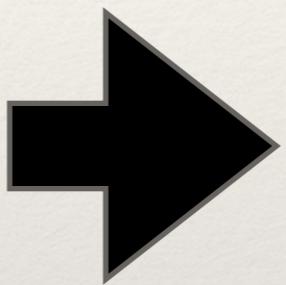
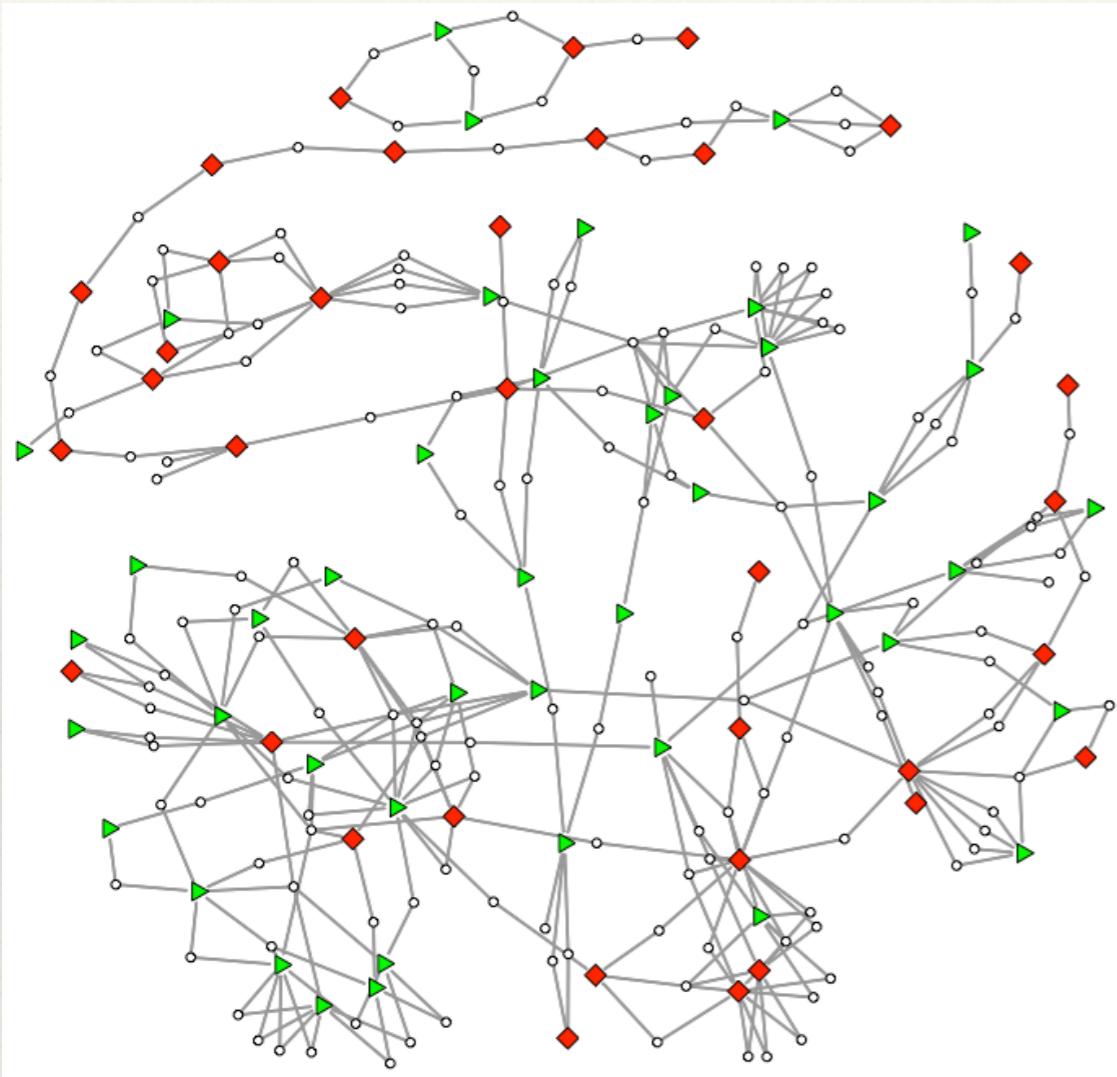
What is the  
concept of  
interest?

How is it  
conceptualized?

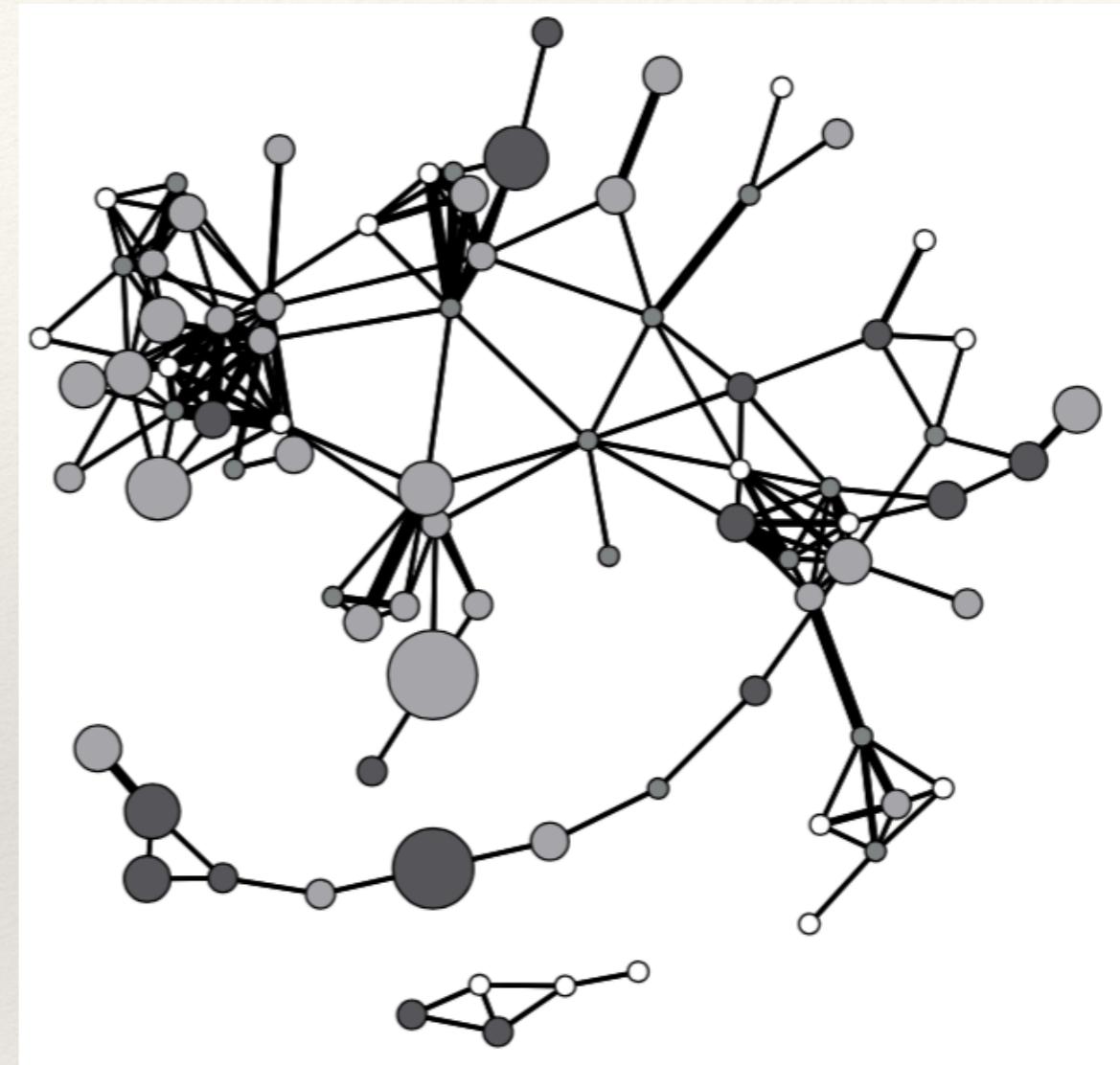
How is it  
operationalized?



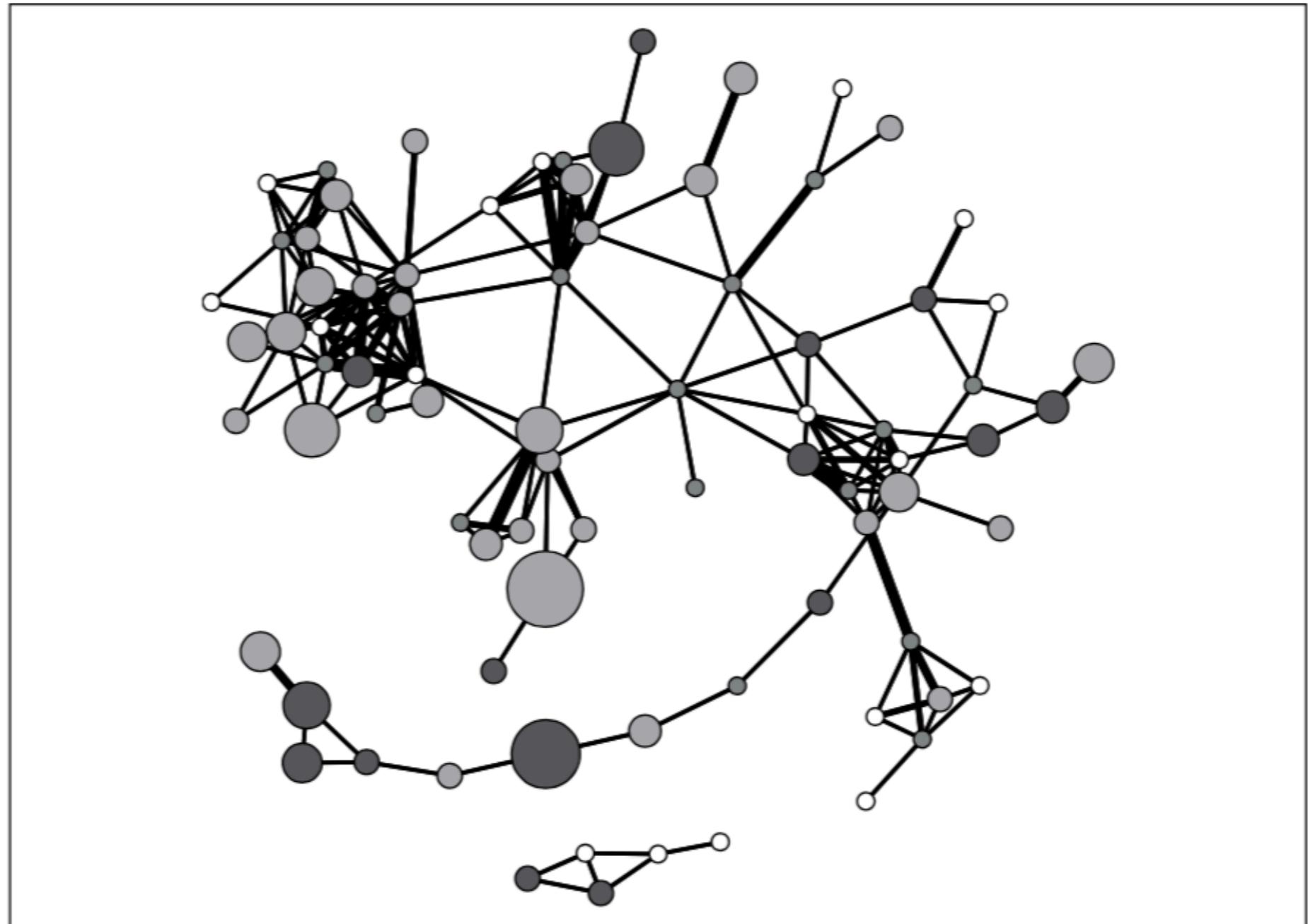
Two-mode network of officers connected by incidents



One-mode network of officers connected to officers



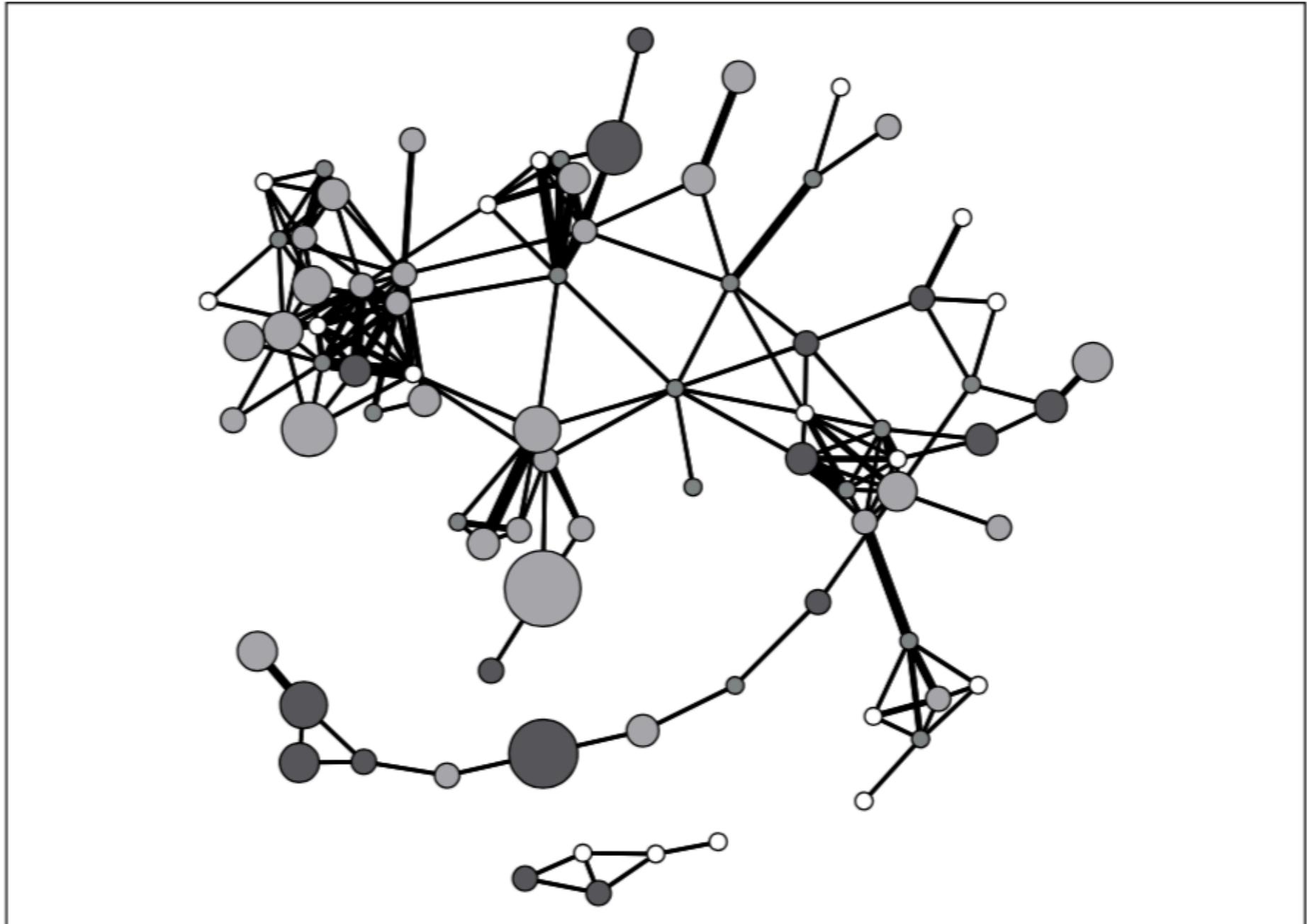
*What do the connections represent in this network?*



**Figure 5.** One-mode network of officers.

Note. Node size is proportional to change in legitimacy: darker = more negative; white = no change. Line size is proportional to number of shared incidents.

Findings:  
Officers views  
of cameras  
changed based  
on who they  
interacted with  
through the  
network



**Figure 5.** One-mode network of officers.

Note. Node size is proportional to change in legitimacy: darker = more negative; white = no change. Line size is proportional to number of shared incidents.

*Statistical Analysis of Networks*

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# Projection & Weighted Graphs

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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 create **unipartite** graphs from **bipartite** graphs?
  - ❖ What is the difference between **dichotomized** projections and **summation** projections?

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# Projection

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- ❖ The process by which we map the connectivity between modes to a single mode.
  - ❖ Example
    - ❖ Two-mode network is people in groups.
    - ❖ By projecting, we get:
      - ❖ One-mode network of people connected to people *by* groups.
      - ❖ One-mode network of groups connected *by* people.

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# Projection

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- ❖ Breiger (1974)
  - ❖ We can build the adjacency matrix for each projected network through matrix algebra.
  - ❖ Specifically, multiplying an adjacency matrix by its transpose.
  - ❖ The transpose of a matrix simply reverses the columns and rows:
    - ❖  $A_{ij}^T = A_{ji}$

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# Projection

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- ❖ Breiger (1974)
  - ❖ The two-mode,  $N \times M$ , adjacency matrix, when multiplied by its **transpose**, produces either:
    - ❖ An  $M \times M$  matrix (ties among  $M$  nodes via  $N$ ).
    - ❖ An  $N \times N$  matrix (ties among  $N$  nodes via  $M$ ).

# Transposition

Matrix **A**

	1	2	3	4	5
A	1	1	0	0	0
B	1	0	0	0	0
C	1	1	0	0	0
D	0	1	1	1	1
E	0	0	1	0	0
F	0	0	1	0	1

order is 6x5

Matrix **A<sup>T</sup>**

	A	B	C	D	E	F
1	1	1	1	0	0	0
2	1	0	1	1	0	0
3	0	0	0	1	1	1
4	0	0	0	1	0	0
5	0	0	0	1	0	1

order is 5x6

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# Projection

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- ❖ Matrix Multiplication Rules

- ❖ To multiply two matrices, the number of columns in the first matrix must match the number of rows in the second matrix.
- ❖ Example:  $5 \times 6 \times 6 \times 5$  works, but not  $5 \times 6 \times 5 \times 6$
- ❖ The **product** matrix has the number of rows equal to the first matrix and the number of columns equal to the second matrix.
- ❖ Example:  $5 \times 6 \times 6 \times 5 = 5 \times 5$

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# Projection

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- ❖ Product Matrix
  - ❖ The product matrix is the projected graph.
  - ❖ Recall that there are two:
    - ❖  $A \times A^t$  (the “people” matrix  $P$ )
    - ❖ And the  $A^t \times A$  (the “group” matrix  $G$ )
      - ❖ *What does each one mean?*

# Matrix Multiplication

Matrix **A**

	1	2	3	4	5
A	1	1	0	0	0
B	1	0	0	0	0
C	1	1	0	0	0
D	0	1	1	1	1
E	0	0	1	0	0
F	0	0	1	0	1

order is 6x5

# Matrix Multiplication

Matrix **A**

	1	2	3	4	5
A	1	1	0	0	0
B	1	0	0	0	0
C	1	1	0	0	0
D	0	1	1	1	1
E	0	0	1	0	0
F	0	0	1	0	1

Matrix **A<sup>T</sup>**

	A	B	C	D	E	F
1	1	1	1	0	0	0
2	1	0	1	1	0	0
3	0	0	0	1	1	1
4	0	0	0	1	0	0
5	0	0	0	1	0	1

X

order is 6x5

order is 5x6

# Matrix Multiplication

Matrix **A**

	1	2	3	4	5
A	1	1	0	0	0
B	1	0	0	0	0
C	1	1	0	0	0
D	0	1	1	1	1
E	0	0	1	0	0
F	0	0	1	0	1

Matrix **A<sup>T</sup>**

	A	B	C	D	E	F
1	1	1	1	0	0	0
2	1	0	1	1	0	0
3	0	0	0	1	1	1
4	0	0	0	1	0	0
5	0	0	0	1	0	1

X

order is 6x5

order is 5x6

The product matrix is  
6x6

# Projection by Multiplication

$$\mathbf{A} \times \mathbf{A}^T = \mathbf{P}$$

	A	B	C	D	E	F
A	2	1	2	1	0	0
B	1	1	1	0	0	0
C	2	1	2	1	0	0
D	1	0	1	4	1	2
E	0	0	0	1	1	1
F	0	0	0	2	1	2

$$6 \times 5 \times 5 \times 6 = 6 \times 6$$

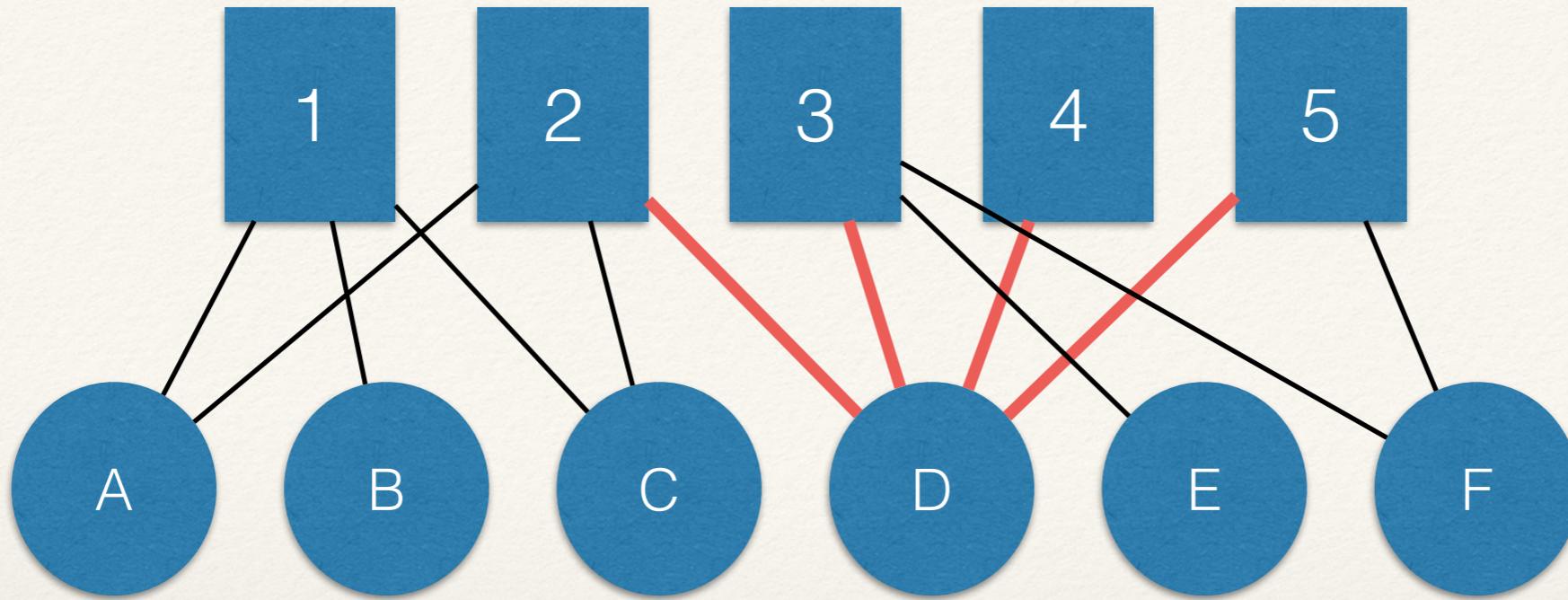
# Projection by Multiplication

We want to know how people are connected by groups (i.e. the rows of our two-mode adjacency matrix)

$$\mathbf{A} \times \mathbf{A}^T = \mathbf{P}$$

	A	B	C	D	E	F
A	2	1	2	1	0	0
B	1	1	1	0	0	0
C	2	1	2	1	0	0
D	1	0	1	4	1	2
E	0	0	0	1	1	1
F	0	0	0	2	1	2

$$6 \times 5 \times 5 \times 6 = 6 \times 6$$



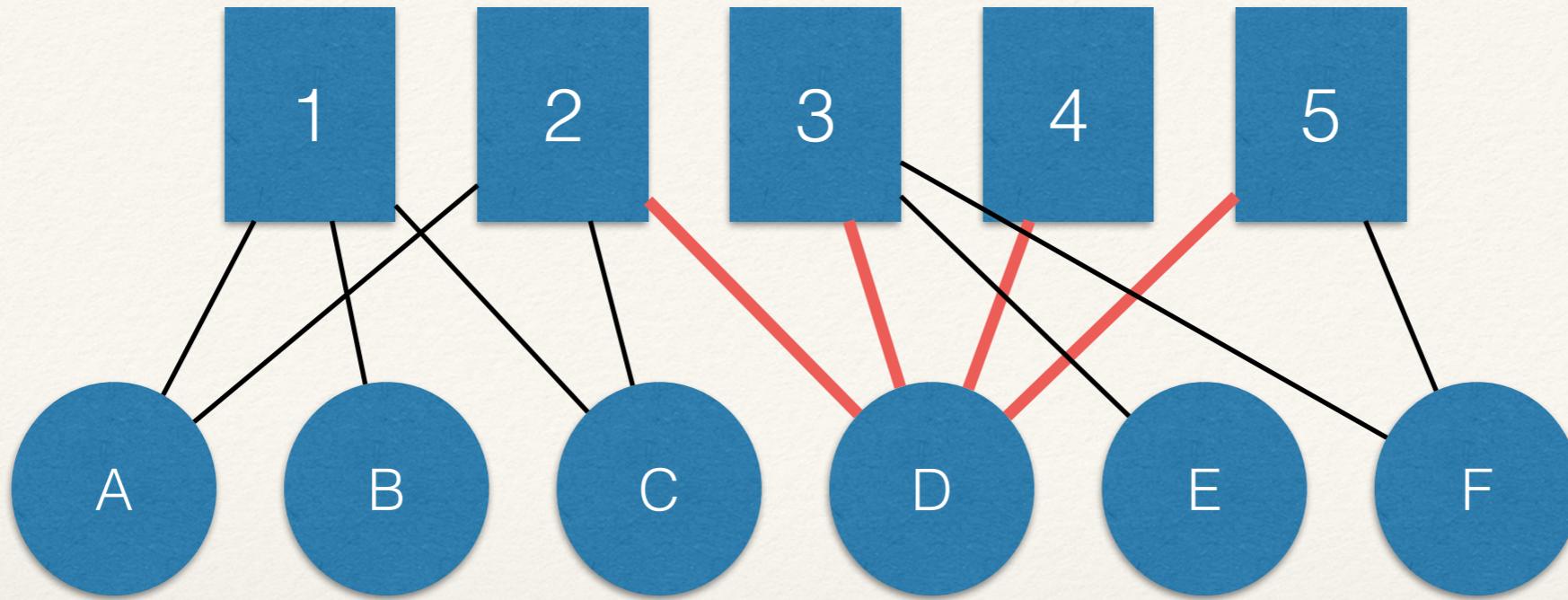
$$\mathbf{A} \times \mathbf{A}^T = \mathbf{P}$$

	A	B	C	D	E	F
A	2	1	2	1	0	0
B	1	1	1	0	0	0
C	2	1	2	1	0	0
D	1	0	1	4	1	2
E	0	0	0	1	1	1
F	0	0	0	2	1	2

The diagonal is the count of ties **the person** has with two-mode vertices

For example, D is in 4 groups

$$6 \times 5 \times 5 \times 6 = 6 \times 6$$

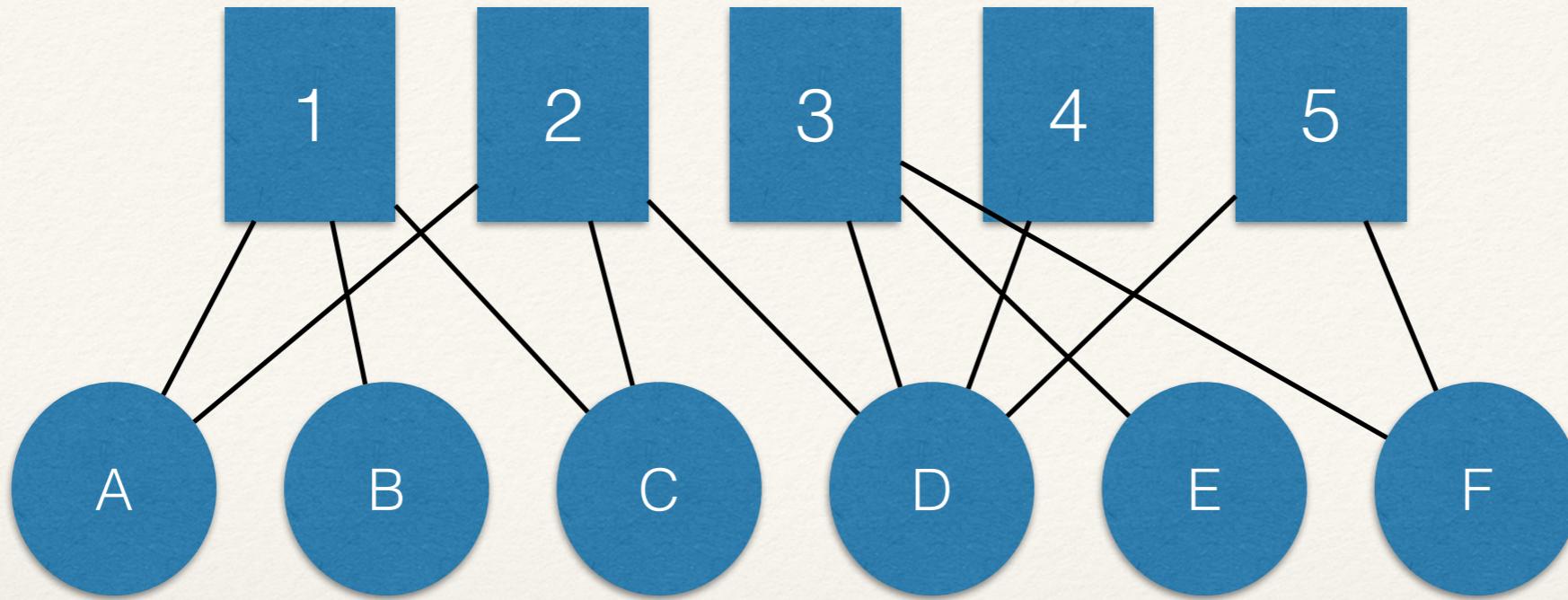


$$\mathbf{A} \times \mathbf{A}^T = \mathbf{P}$$

	A	B	C	D	E	F
A	2	1	2	1	0	0
B	1	1	1	0	0	0
C	2	1	2	1	0	0
D	1	0	1	4	1	2
E	0	0	0	1	1	1
F	0	0	0	2	1	2

*What statistic does the diagonal give us?*

$$6 \times 5 \times 5 \times 6 = 6 \times 6$$

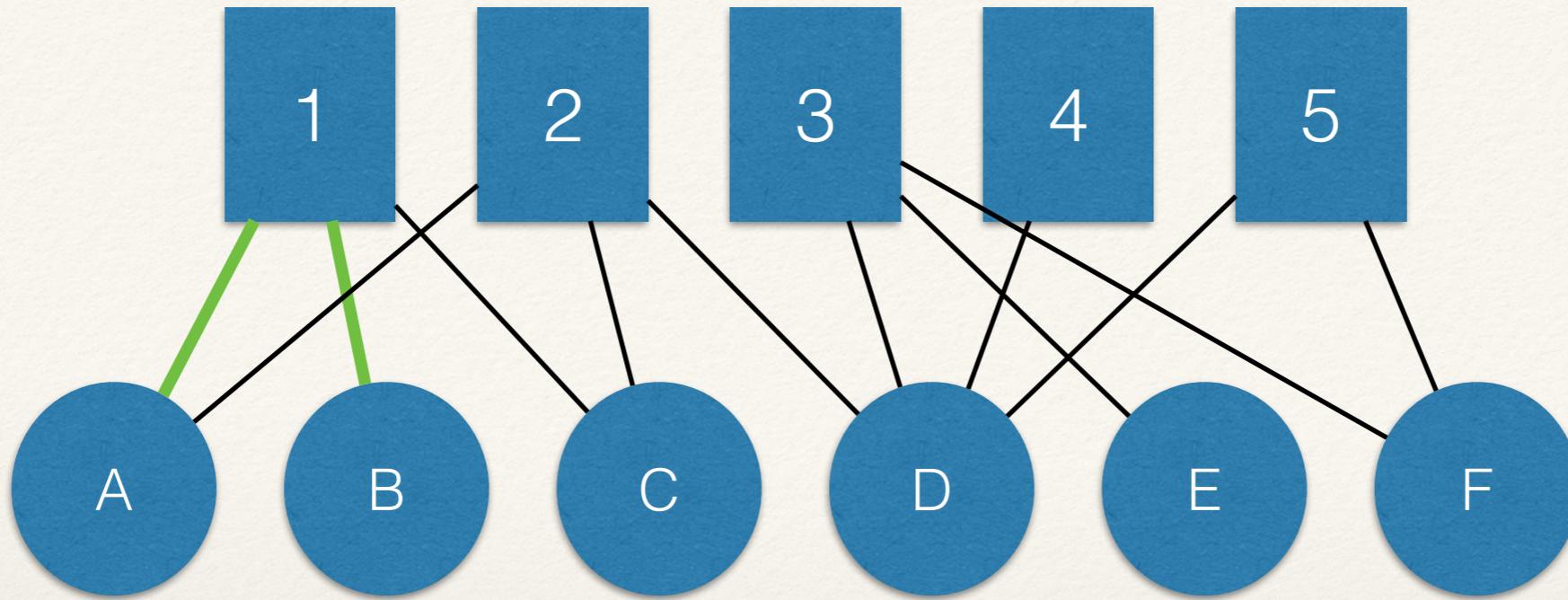


$$\mathbf{A} \times \mathbf{A}^T = \mathbf{P}$$

	A	B	C	D	E	F
A	2	1	2	1	0	0
B	1	1	1	0	0	0
C	2	1	2	1	0	0
D	1	0	1	4	1	2
E	0	0	0	1	1	1
F	0	0	0	2	1	2

Note, that the projection forces the product matrix to be symmetric  
(i.e. undirected graph)

$$6 \times 5 \times 5 \times 6 = 6 \times 6$$



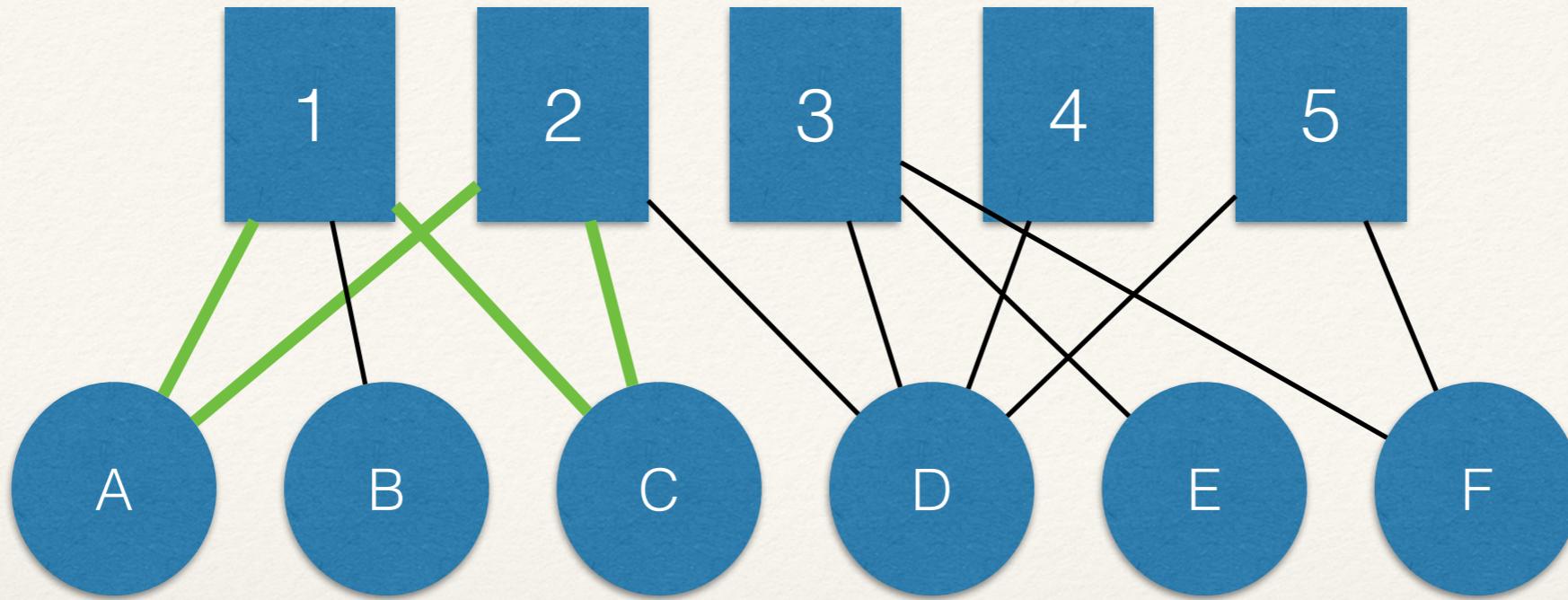
$$\mathbf{A} \times \mathbf{A}^T = \mathbf{P}$$

	A	B	C	D	E	F
A	2	1	2	1	0	0
B	1	1	1	0	0	0
C	2	1	2	1	0	0
D	1	0	1	4	1	2
E	0	0	0	1	1	1
F	0	0	0	2	1	2

The off-diagonal entries indicate the number of ways that vertices in the first mode are connected by vertices in the second mode

A and B are linked through a single vertex, 1

$$6 \times 5 \times 5 \times 6 = 6 \times 6$$



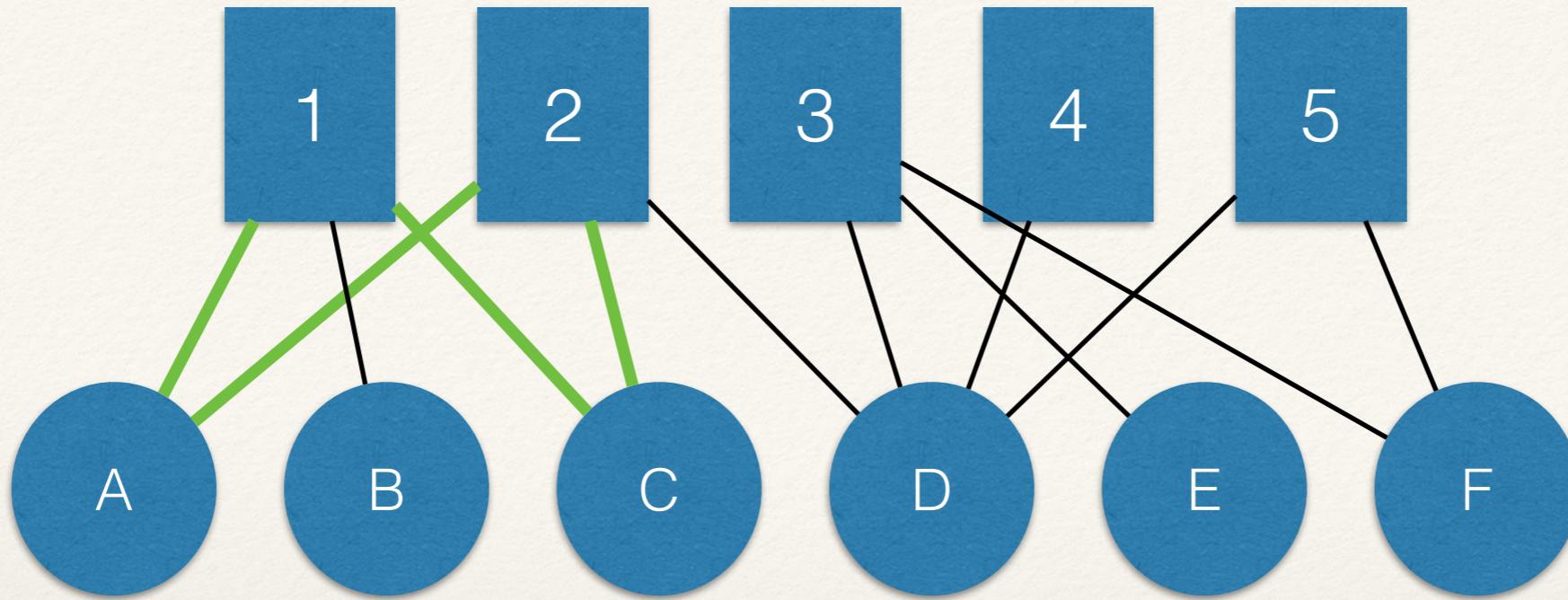
$$\mathbf{A} \times \mathbf{A}^T = \mathbf{P}$$

	A	B	C	D	E	F
A	2	1	2	1	0	0
B	1	1	1	0	0	0
C	2	1	2	1	0	0
D	1	0	1	4	1	2
E	0	0	0	1	1	1
F	0	0	0	2	1	2

A and C are linked through two vertices, 1 and 2

So, if these are groups, A and C are members of 2 of the same groups

$$6 \times 5 \times 5 \times 6 = 6 \times 6$$



$$\mathbf{A} \times \mathbf{A}^T = \mathbf{P}$$

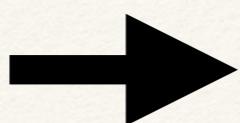
	A	B	C	D	E	F
A	2	1	2	1	0	0
B	1	1	1	0	0	0
C	2	1	2	1	0	0
D	1	0	1	4	1	2
E	0	0	0	1	1	1
F	0	0	0	2	1	2

**NOTE: these are counts  
of shared vertices, not  
edge counts**

$$6 \times 5 \times 5 \times 6 = 6 \times 6$$

	A	B	C	D	E	F
A	2	1	2	1	0	0
B	1	1	1	0	0	0
C	2	1	2	1	0	0
D	1	0	1	4	1	2
E	0	0	0	1	1	1
F	0	0	0	2	1	2

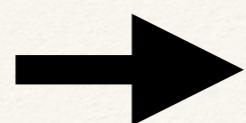
	A	B	C	D	E	F
A	2	1	2	1	0	0
B	1	1	1	0	0	0
C	2	1	2	1	0	0
D	1	0	1	4	1	2
E	0	0	0	1	1	1
F	0	0	0	2	1	2



	A	B	C	D	E	F
A	0	1	1	1	0	0
B	1	0	1	0	0	0
C	1	1	0	1	0	0
D	1	0	1	0	1	1
E	0	0	0	1	0	1
F	0	0	0	1	1	0

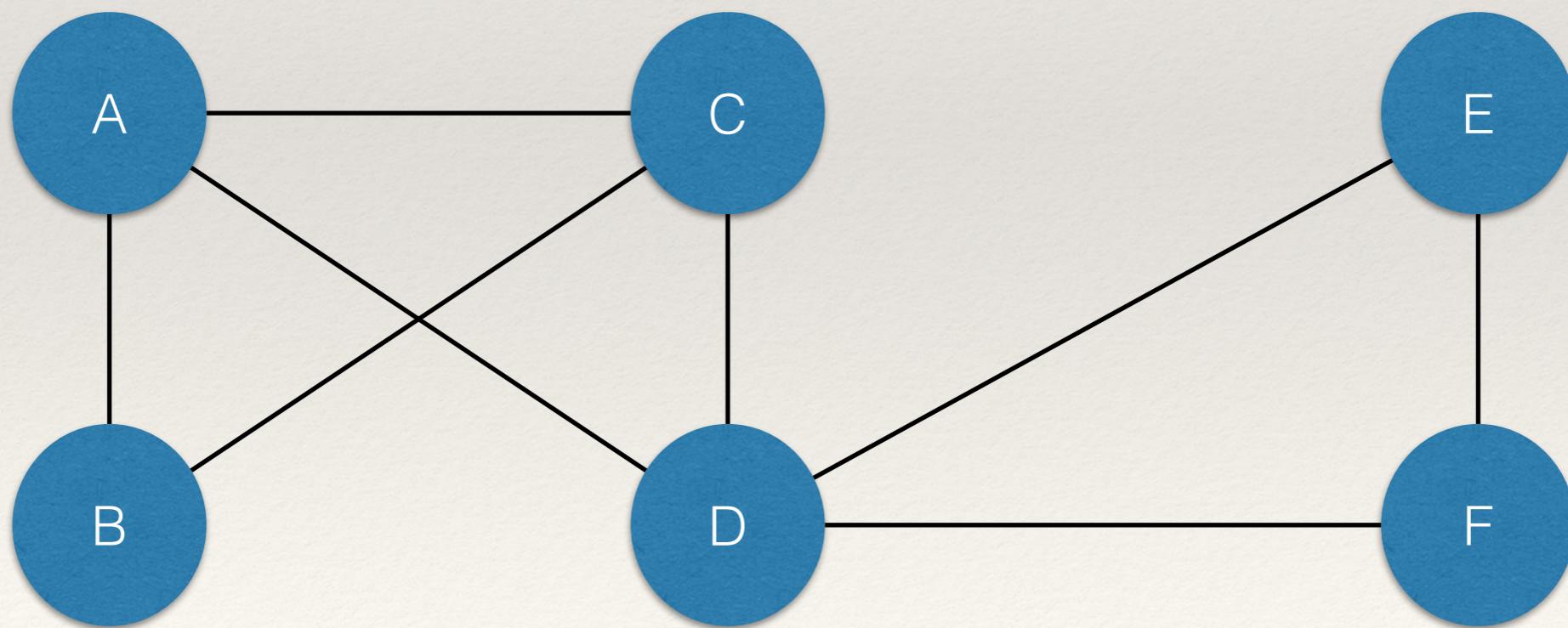
If we treat any tie greater than 0 as binary, *called **dichotomizing***, and recode the diagonal as 0, we get an undirected, one-mode network

	A	B	C	D	E	F
A	2	1	2	1	0	0
B	1	1	1	0	0	0
C	2	1	2	1	0	0
D	1	0	1	4	1	2
E	0	0	0	1	1	1
F	0	0	0	2	1	2



	A	B	C	D	E	F
A	0	1	1	1	0	0
B	1	0	1	0	0	0
C	1	1	0	1	0	0
D	1	0	1	0	1	1
E	0	0	0	1	0	1
F	0	0	0	1	1	0

If we treat any tie greater than 0 as binary, *called **dichotomizing***, and recode the diagonal as 0, we get an undirected, one-mode network



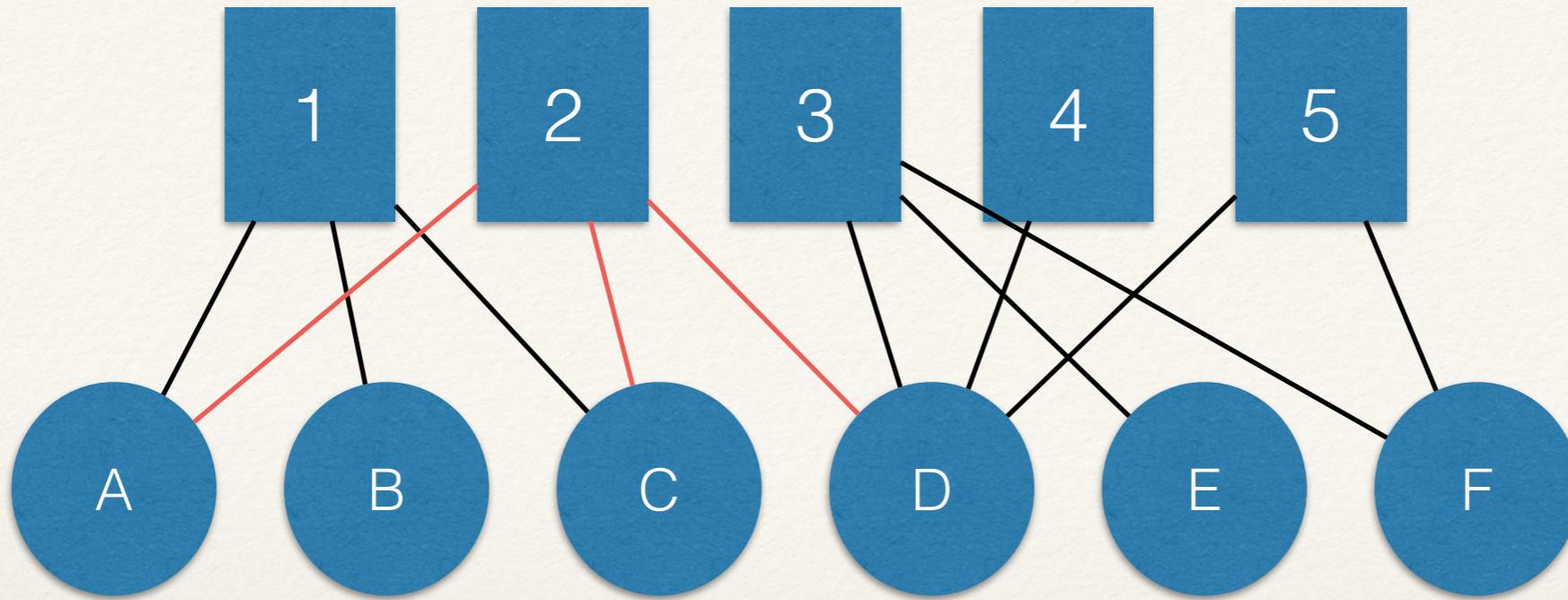
# Projection by Multiplication

We want to know how groups are connected by people  
(i.e. the columns of our two-mode adjacency matrix)

$$\mathbf{A}^T \times \mathbf{A} = \mathbf{G}$$

	1	2	3	4	5
1	3	2	0	0	0
2	2	3	1	1	1
3	0	1	3	1	2
4	0	1	1	1	1
5	0	1	2	1	2

$$5 \times 6 \times 6 \times 5 = 5 \times 5$$



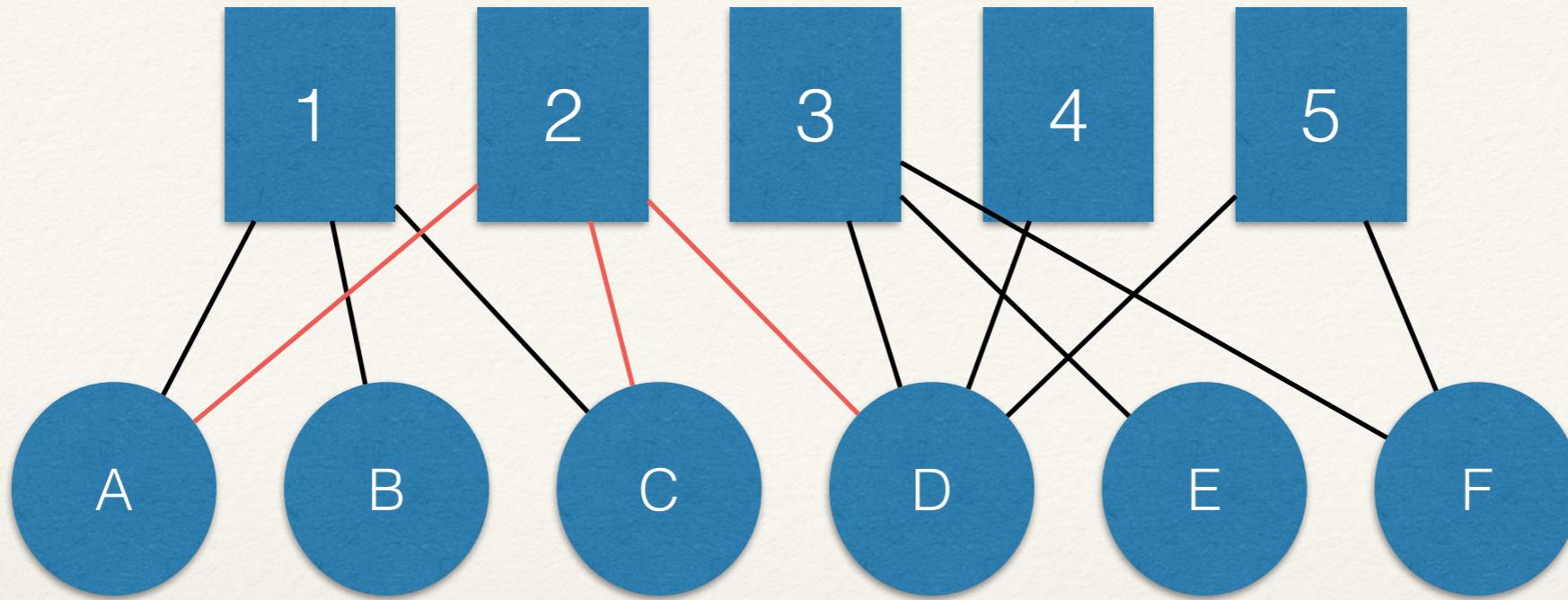
$$\mathbf{A}^T \times \mathbf{A} = \mathbf{G}$$

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1	3	2	0	0	0
2	2	3	1	1	1
3	0	1	3	1	2
4	0	1	1	1	1
5	0	1	2	1	2

The diagonal is the count of ties the **group** has with two-mode vertices

For example, 2 has 3 people

$$5 \times 6 \times 6 \times 5 = 5 \times 5$$

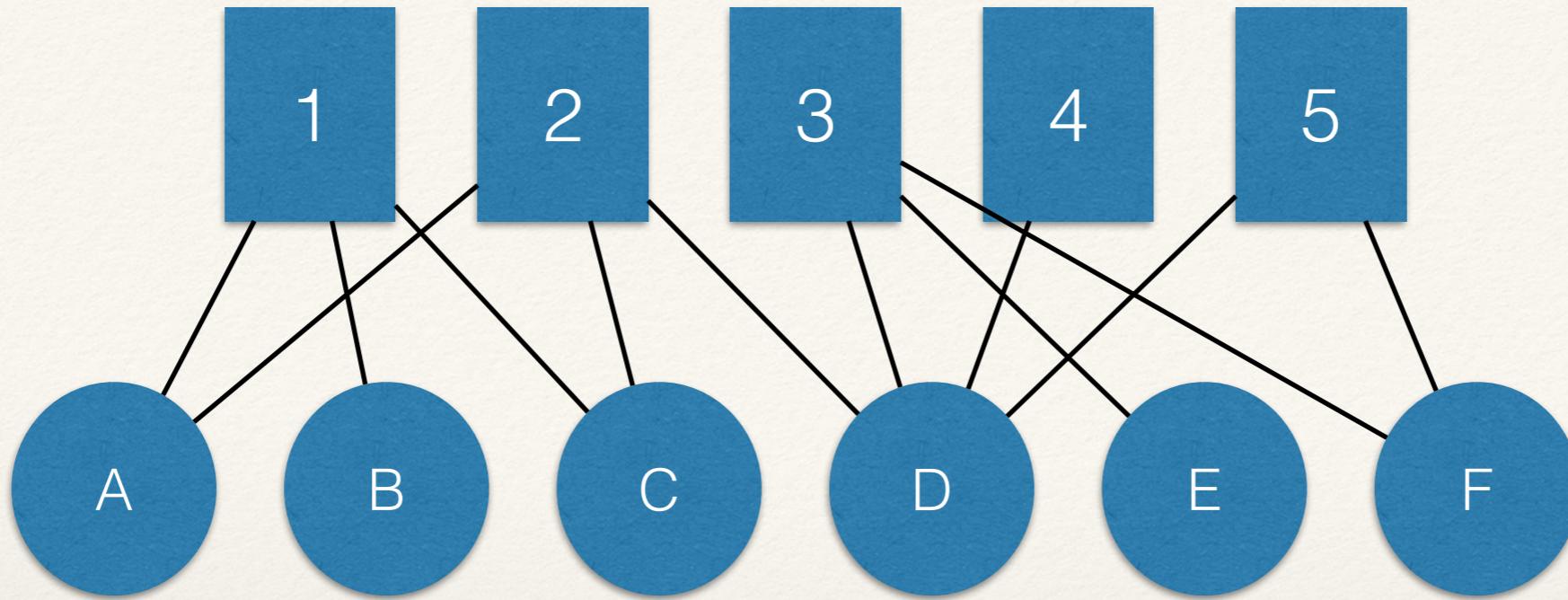


$$\mathbf{A}^T \times \mathbf{A} = \mathbf{G}$$

	1	2	3	4	5
1	3	2	0	0	0
2	2	3	1	1	1
3	0	1	3	1	2
4	0	1	1	1	1
5	0	1	2	1	2

*What statistic does the diagonal give us?*

$$5 \times 6 \times 6 \times 5 = 5 \times 5$$

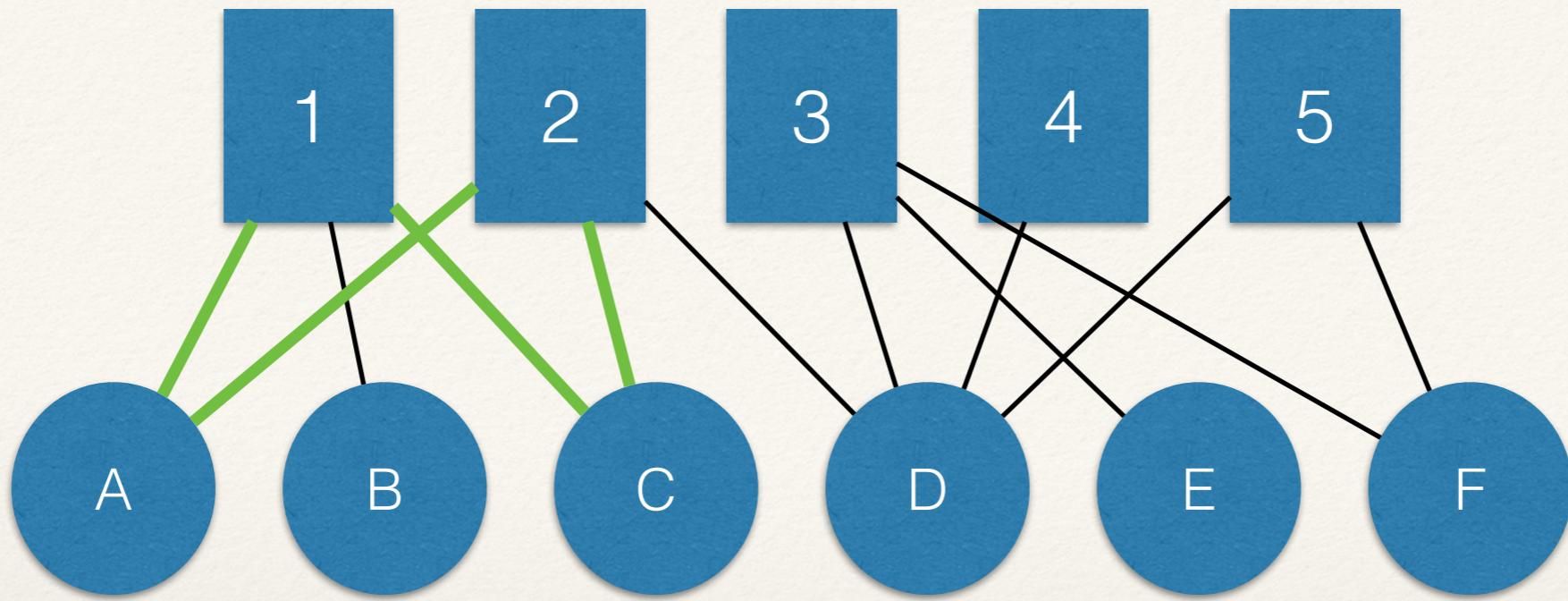


$$\mathbf{A}^T \times \mathbf{A} = \mathbf{G}$$

	1	2	3	4	5
1	3	2	0	0	0
2	2	3	1	1	1
3	0	1	3	1	2
4	0	1	1	1	1
5	0	1	2	1	2

Note, that the projection forces the product matrix to be symmetric  
(i.e. undirected graph)

$$5 \times 6 \times 6 \times 5 = 5 \times 5$$



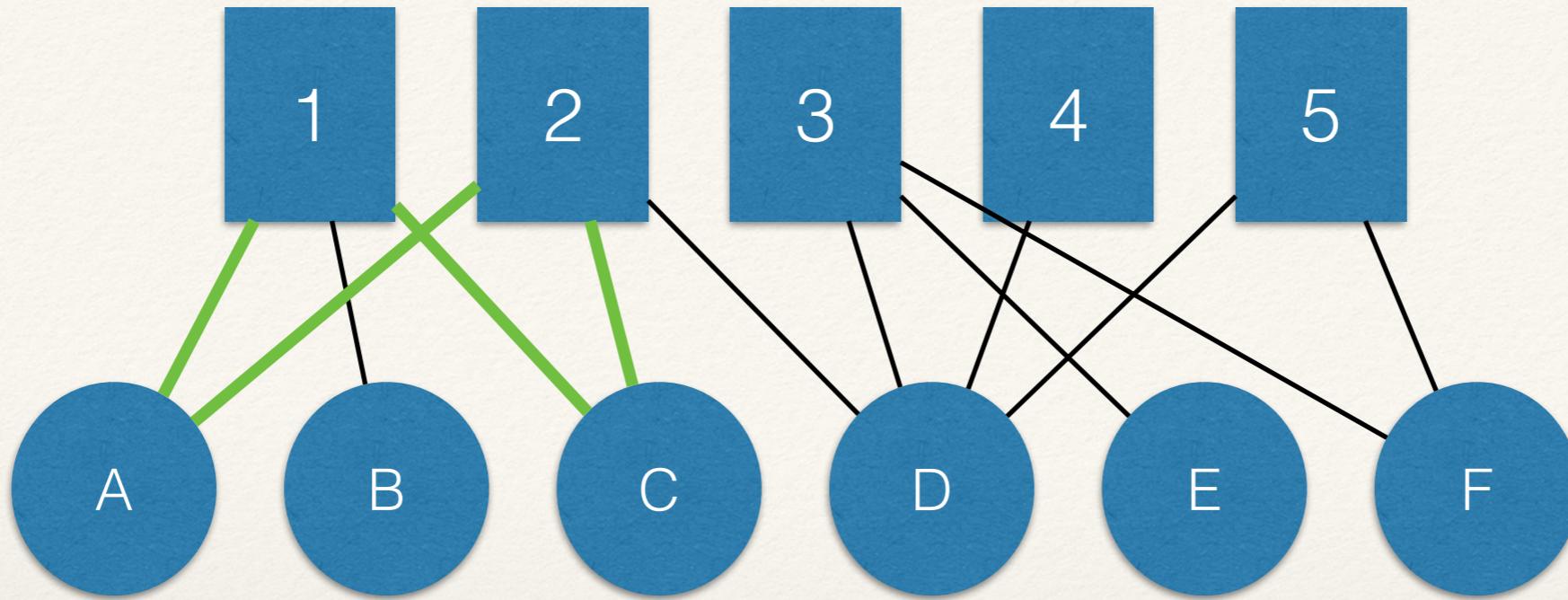
$$\mathbf{A}^T \times \mathbf{A} = \mathbf{G}$$

	1	2	3	4	5
1	3	2	0	0	0
2	2	3	1	1	1
3	0	1	3	1	2
4	0	1	1	1	1
5	0	1	2	1	2

The off-diagonal entries indicate the number of ways that vertices in the second mode are connected by vertices in the first mode

1 and 2 are connected by 2 vertices, A and C

$$5 \times 6 \times 6 \times 5 = 5 \times 5$$



$$\mathbf{A}^T \times \mathbf{A} = \mathbf{G}$$

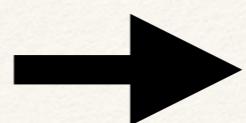
	1	2	3	4	5
1	3	2	0	0	0
2	2	3	1	1	1
3	0	1	3	1	2
4	0	1	1	1	1
5	0	1	2	1	2

**NOTE: these are counts  
of shared vertices, not  
edge counts**

$$5 \times 6 \times 6 \times 5 = 5 \times 5$$

	1	2	3	4	5
1	3	2	0	0	0
2	2	3	1	1	1
3	0	1	3	1	2
4	0	1	1	1	1
5	0	1	2	1	2

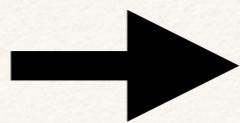
	1	2	3	4	5
1	3	2	0	0	0
2	2	3	1	1	1
3	0	1	3	1	2
4	0	1	1	1	1
5	0	1	2	1	2



	1	2	3	4	5
1	0	1	0	0	0
2	1	0	1	1	1
3	0	1	0	1	1
4	0	1	1	0	1
5	0	1	1	1	0

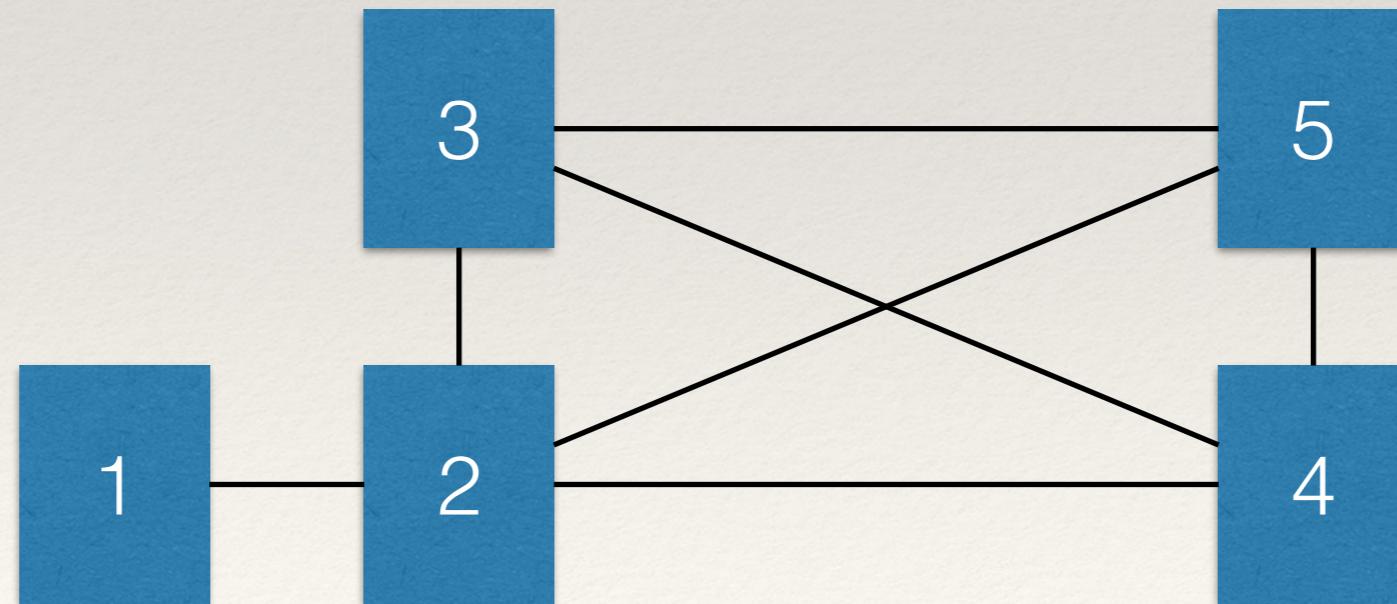
If we treat any tie greater than 0 as binary, *called **dichotomizing***, and recode the diagonal as 0, we get an undirected, one-mode network

	1	2	3	4	5
1	3	2	0	0	0
2	2	3	1	1	1
3	0	1	3	1	2
4	0	1	1	1	1
5	0	1	2	1	2



	1	2	3	4	5
1	0	1	0	0	0
2	1	0	1	1	1
3	0	1	0	1	1
4	0	1	1	0	1
5	0	1	1	1	0

If we treat any tie greater than 0 as binary, *called **dichotomizing***, and recode the diagonal as 0, we get an undirected, one-mode network



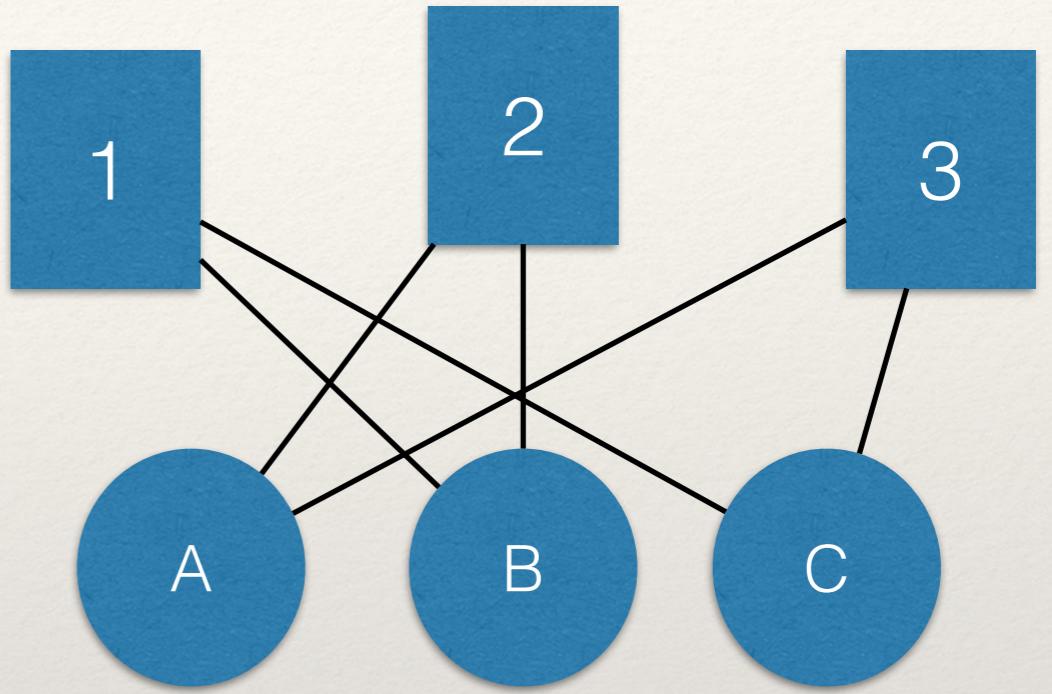
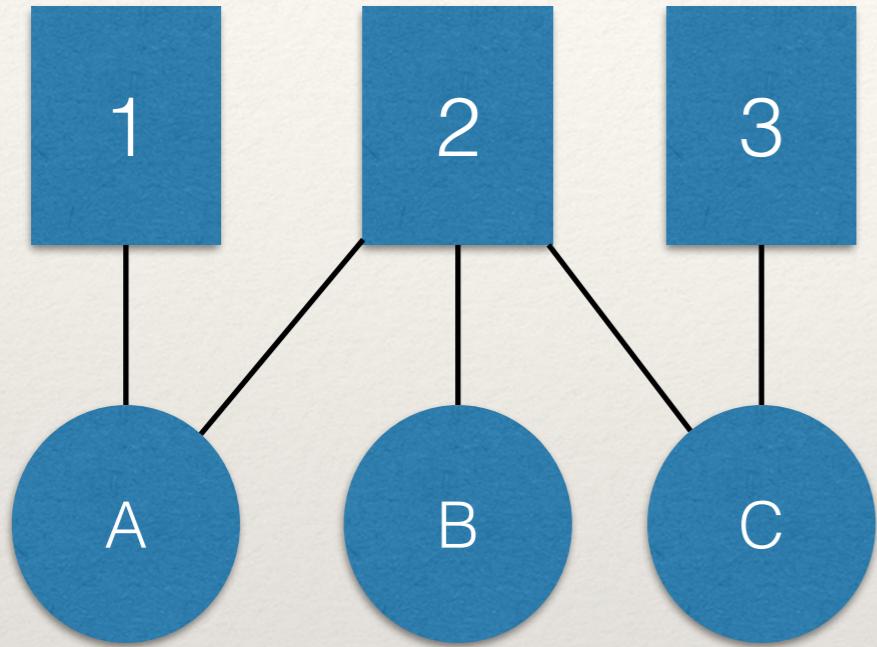
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# Projection

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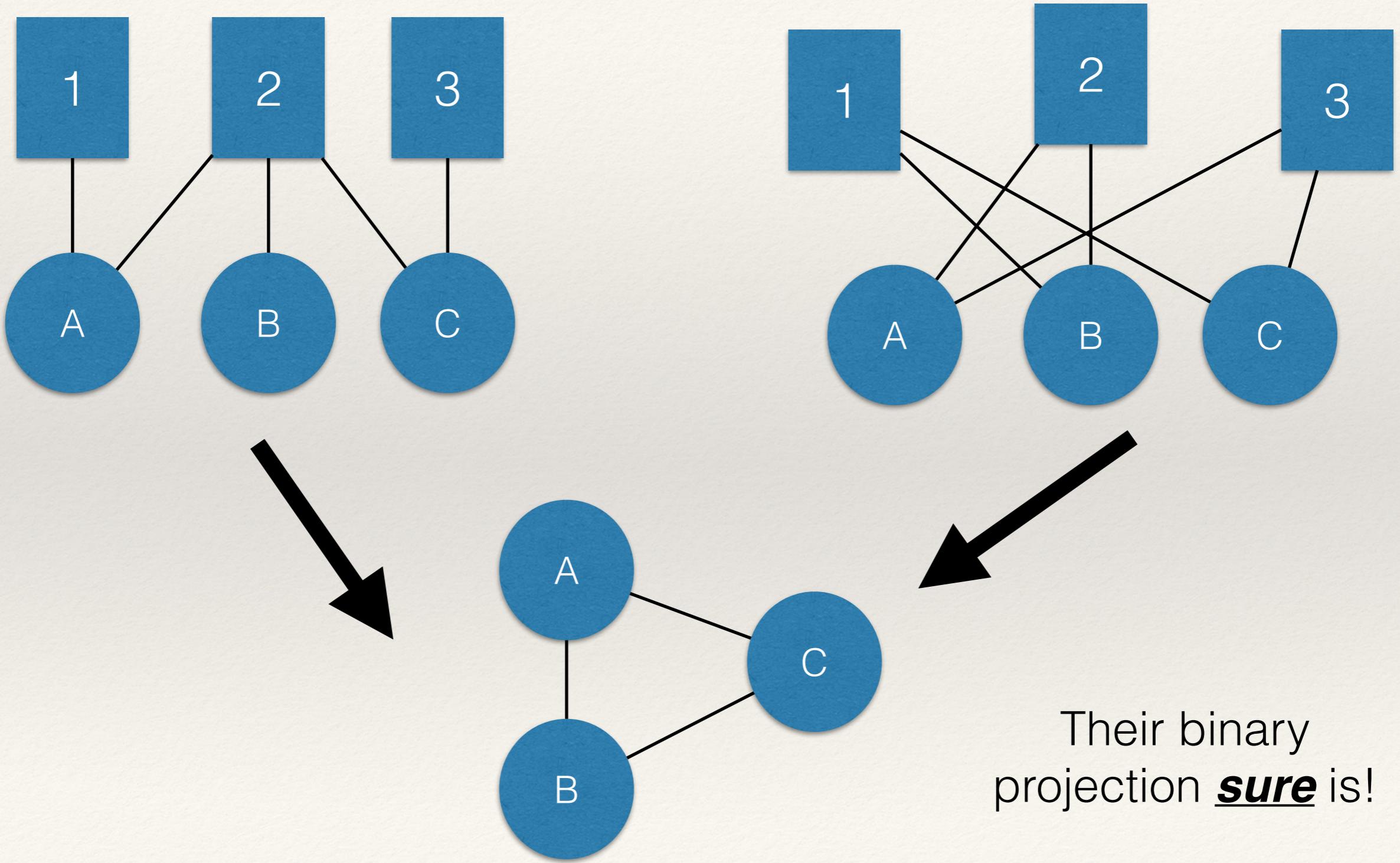
- ❖ To project, or not to project?
  - ❖ As noted by many scholars, there is data loss when we project and binarize the data.
  - ❖ Sometimes, this can be misleading.

# Projection and Data Loss

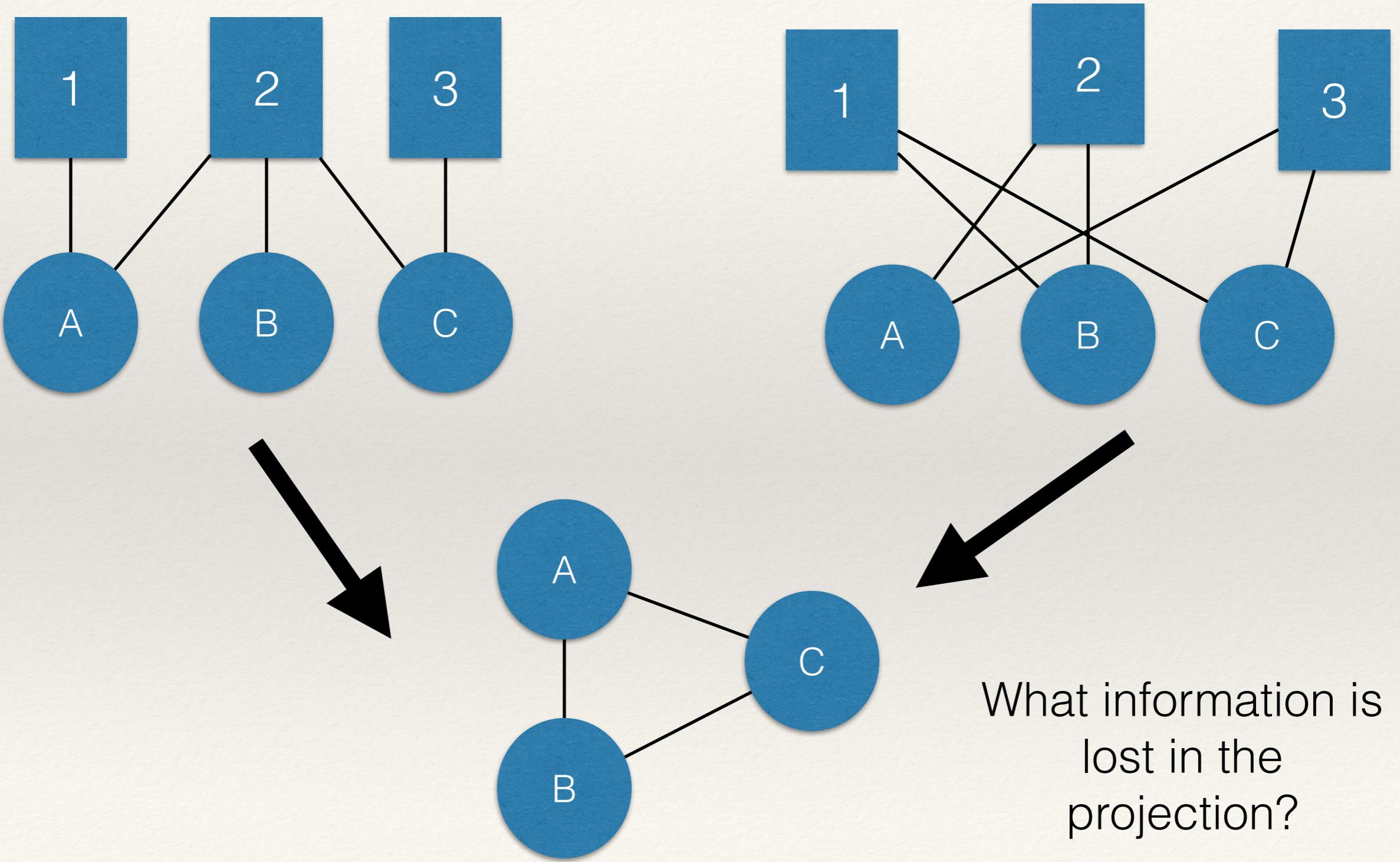


Are these bipartite graphs the same?

# Projection and Data Loss



# Projection and Data Loss



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# Projection

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- ❖ So what do we do?
  - ❖ When you can, “keep it real” by keeping it two-mode.
  - ❖ If you must project, minimize data loss by weighting edges.

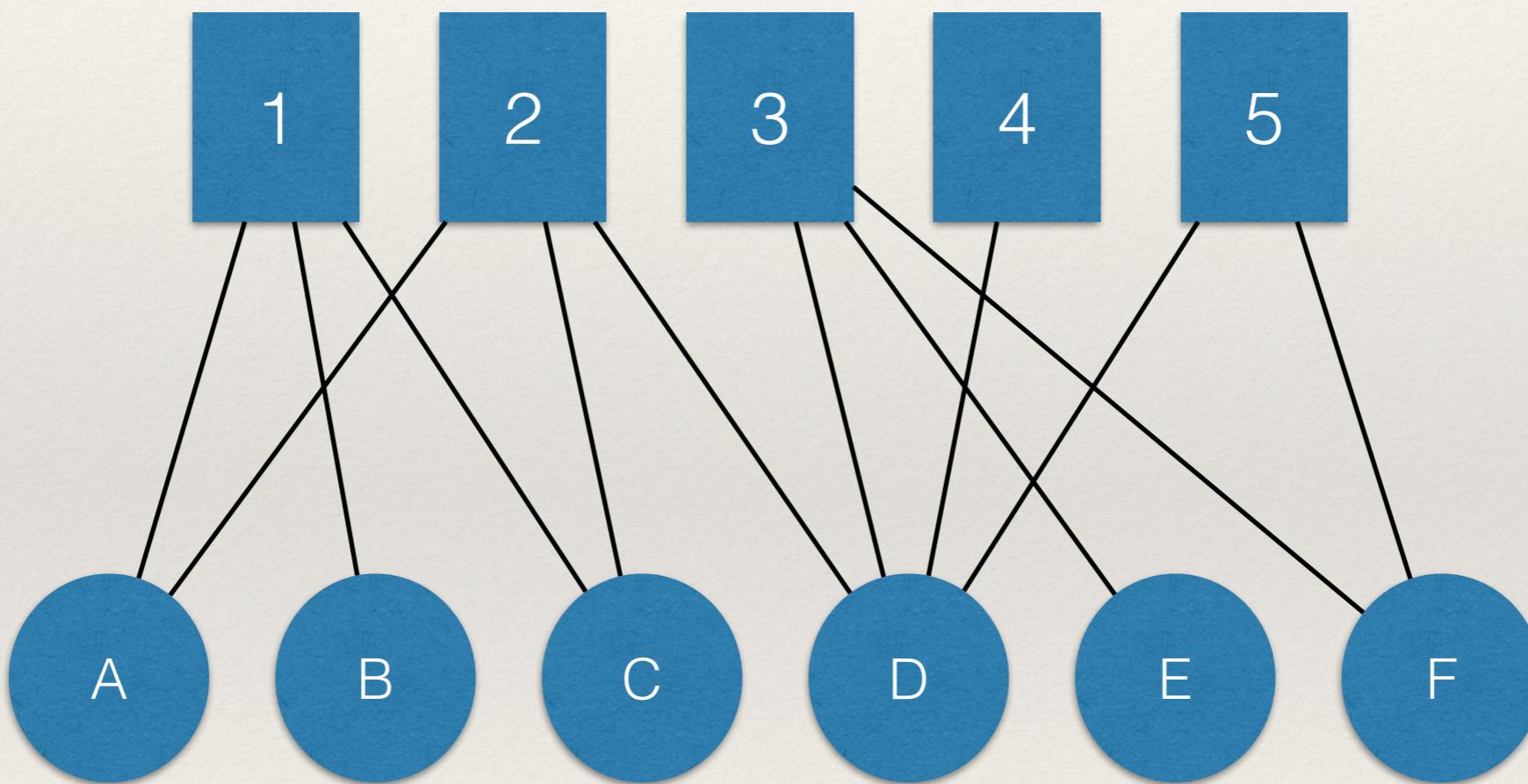
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# Weighted Edges

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- ❖ We can use the information from the bipartite graph to weight the edges in the network.
- ❖ These weights can be used in a plot and / or in the analysis.
- ❖ The most common method is to sum the ties between two actors (i.e. *summation method*).

# Projection



	A	B	C	D	E	F
A	2	1	2	1	0	0
B	1	1	1	0	0	0
C	2	1	2	1	0	0
D	1	0	1	4	1	2
E	0	0	0	1	1	1
F	0	0	0	2	1	2



	A	B	C	D	E	F
A	0	1	1	1	0	0
B	1	0	1	0	0	0
C	1	1	0	1	0	0
D	1	0	1	0	1	1
E	0	0	0	1	0	1
F	0	0	0	1	1	0

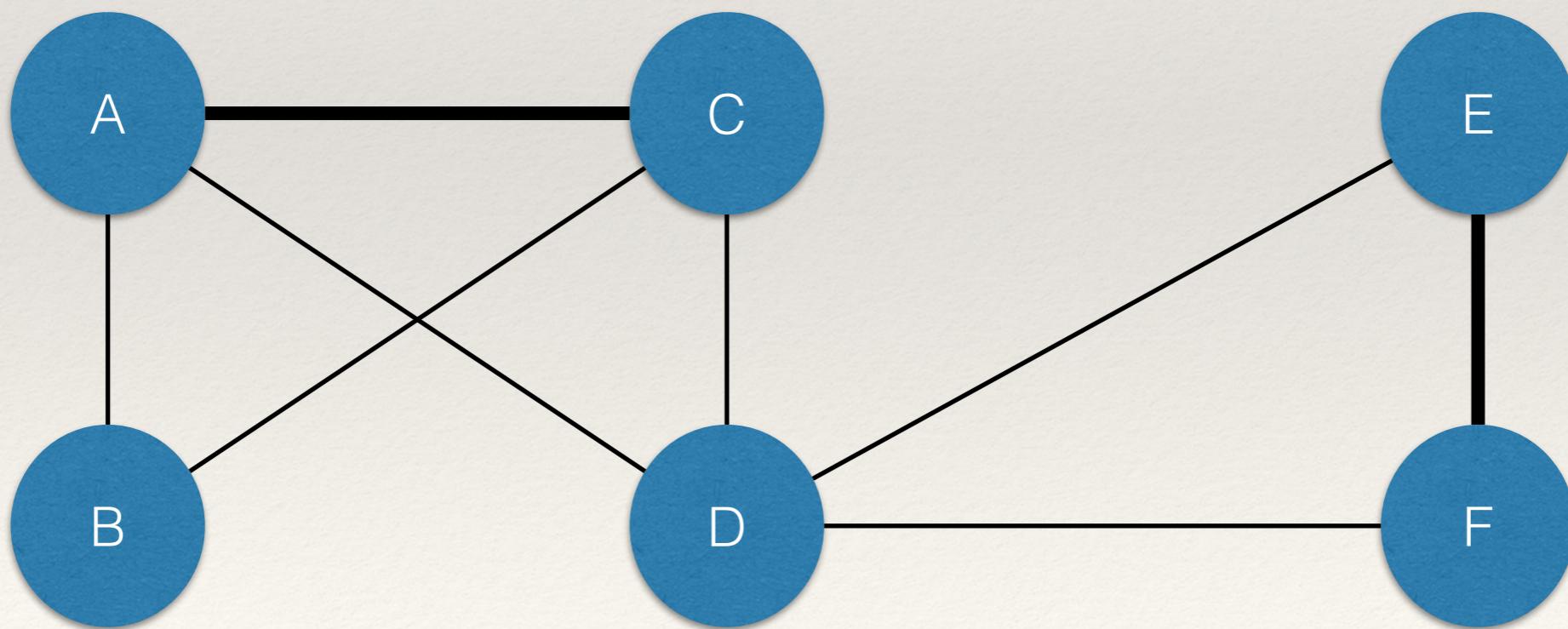
If we treat any tie greater than 0 as binary, called **dichotomizing**, and recode the diagonal as 0, we get an undirected, one-mode network

	A	B	C	D	E	F
A	2	1	2	1	0	0
B	1	1	1	0	0	0
C	2	1	2	1	0	0
D	1	0	1	4	1	2
E	0	0	0	1	1	1
F	0	0	0	2	1	2

$$\mathbf{A} \times \mathbf{A}^T = \mathbf{P}$$

	A	B	C	D	E	F
A	2	1	2	1	0	0
B	1	1	1	0	0	0
C	2	1	2	1	0	0
D	1	0	1	4	1	2
E	0	0	0	1	1	1
F	0	0	0	2	1	2

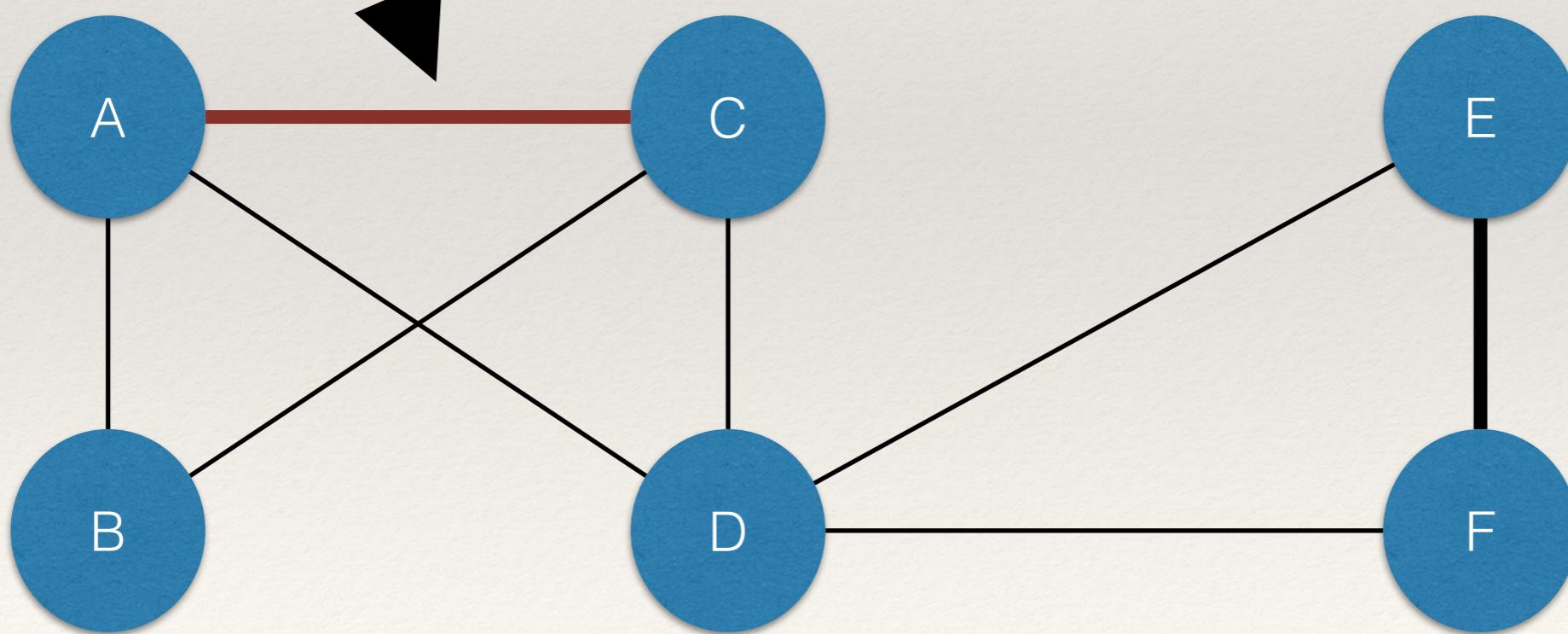
The off-diagonal entries are the tie weights



$$\mathbf{A} \times \mathbf{A}^T = \mathbf{P}$$

	A	B	C	D	E	F
A	2	1	2	1	0	0
B	1	1		0	0	0
C	2	1	2	1	0	0
D	1	0	1	4	1	2
E	0	0	0	1	1	1
F	0	0	0	2	1	2

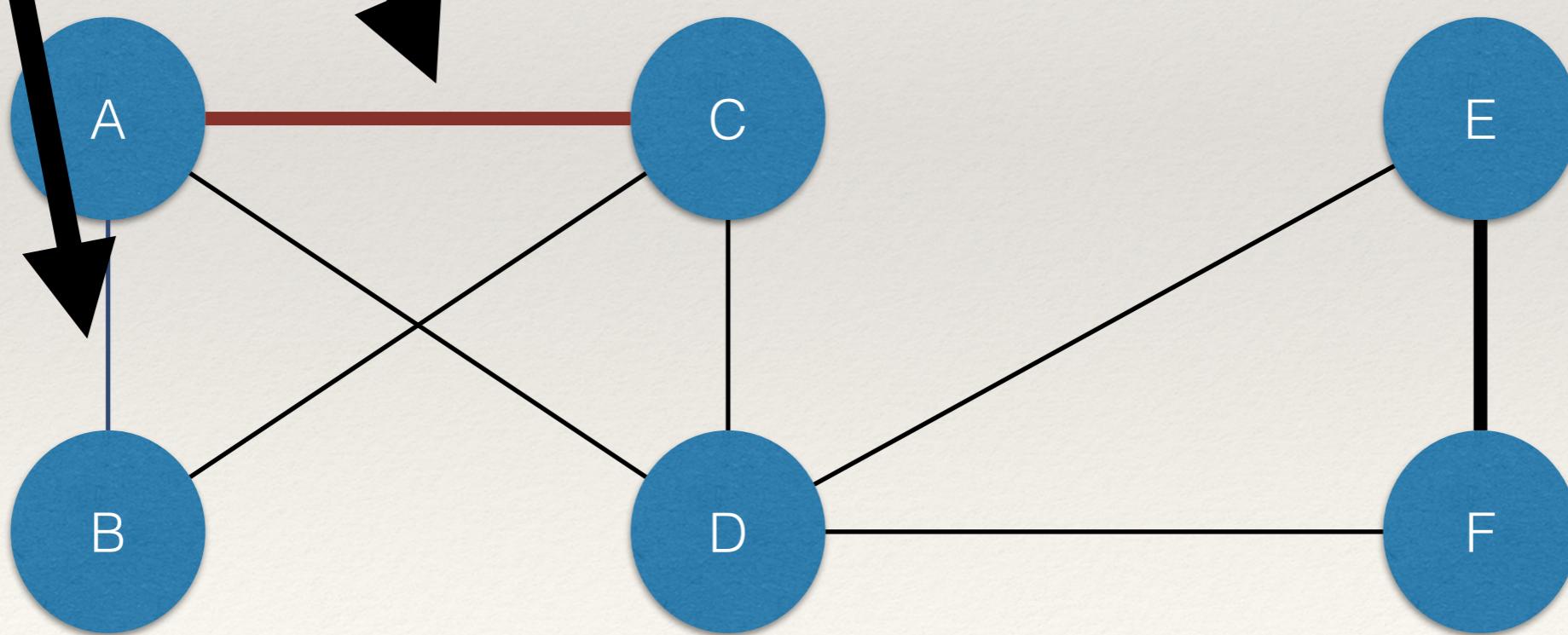
The weight corresponds to the number of nodes in the second mode that the two actors share in the bipartite graph.



$$\mathbf{A} \times \mathbf{A}^T = \mathbf{P}$$

	A	B	C	D	E	F
A	2	1	2	1	0	0
B	1	1		0	0	0
C	2	1	2	1	0	0
D	1	0	1	4	1	2
E	0	0	0	1	1	1
F	0	0	0	2	1	2

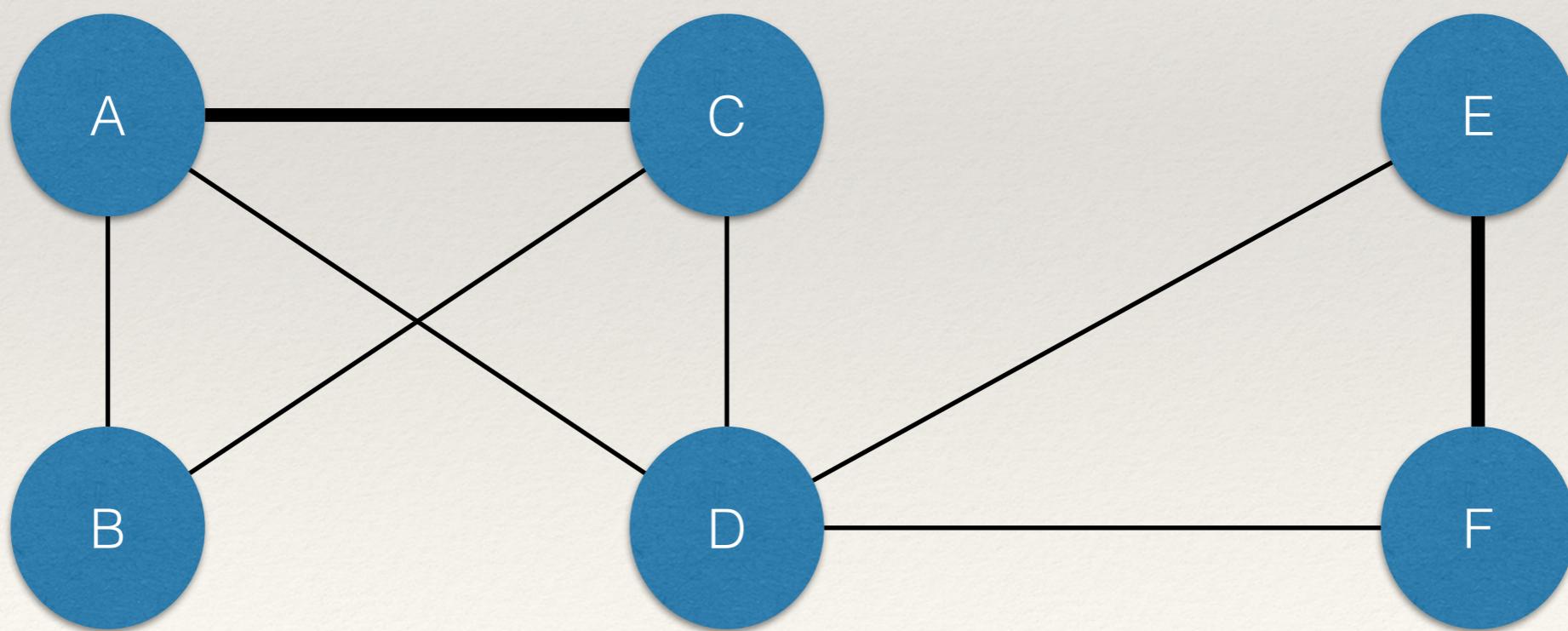
The weight corresponds to the number of nodes in the second mode that the two actors share in the bipartite graph.

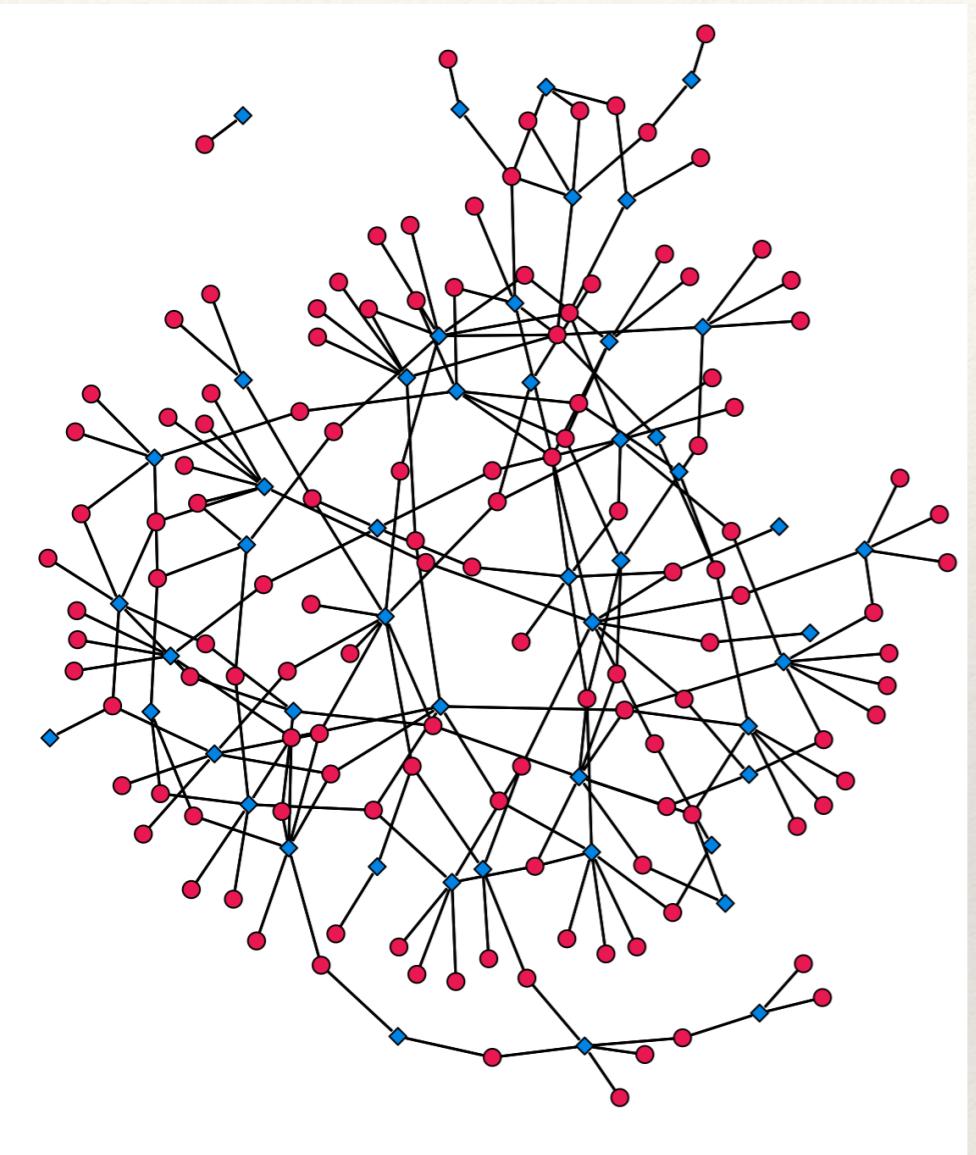


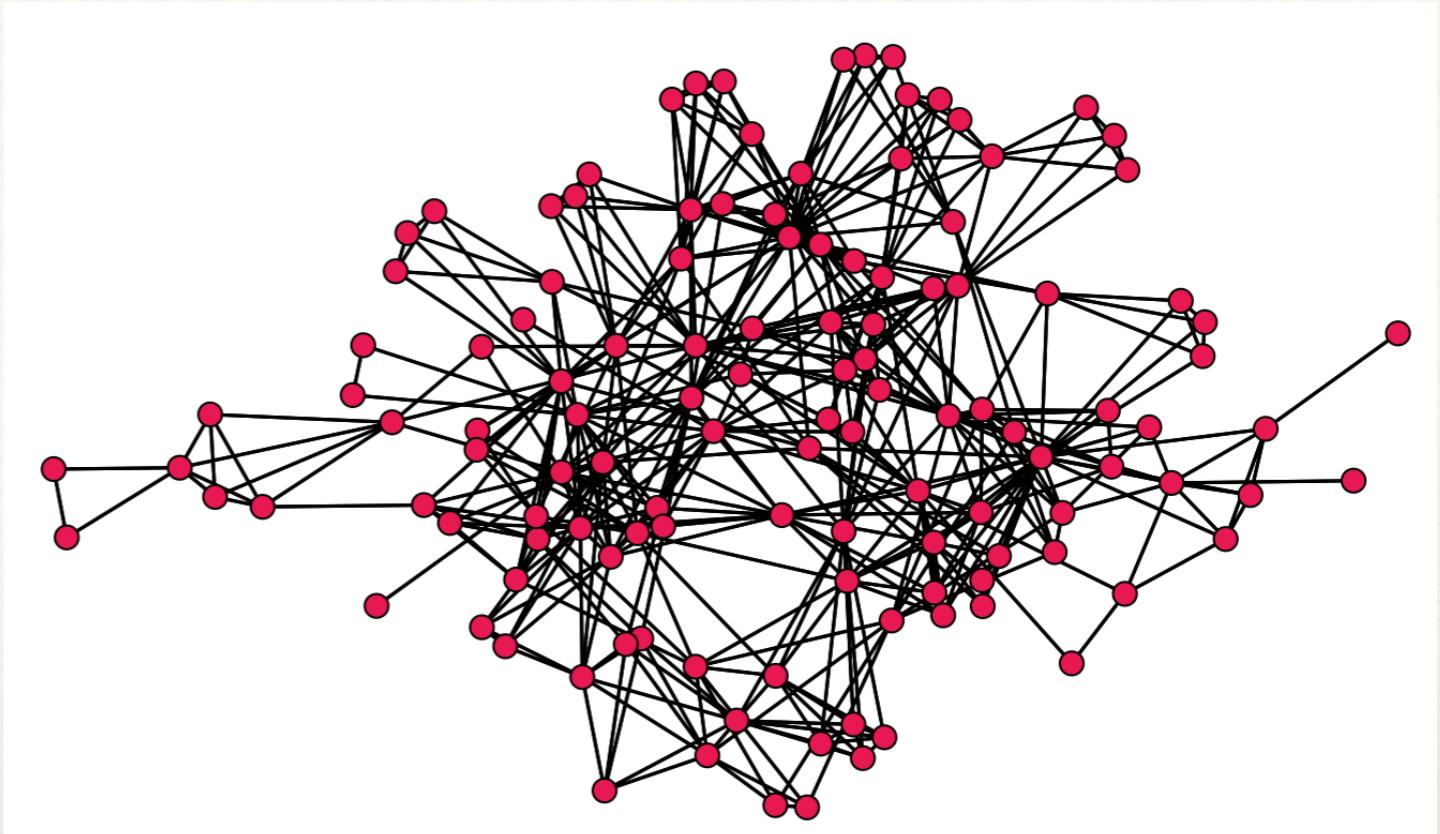
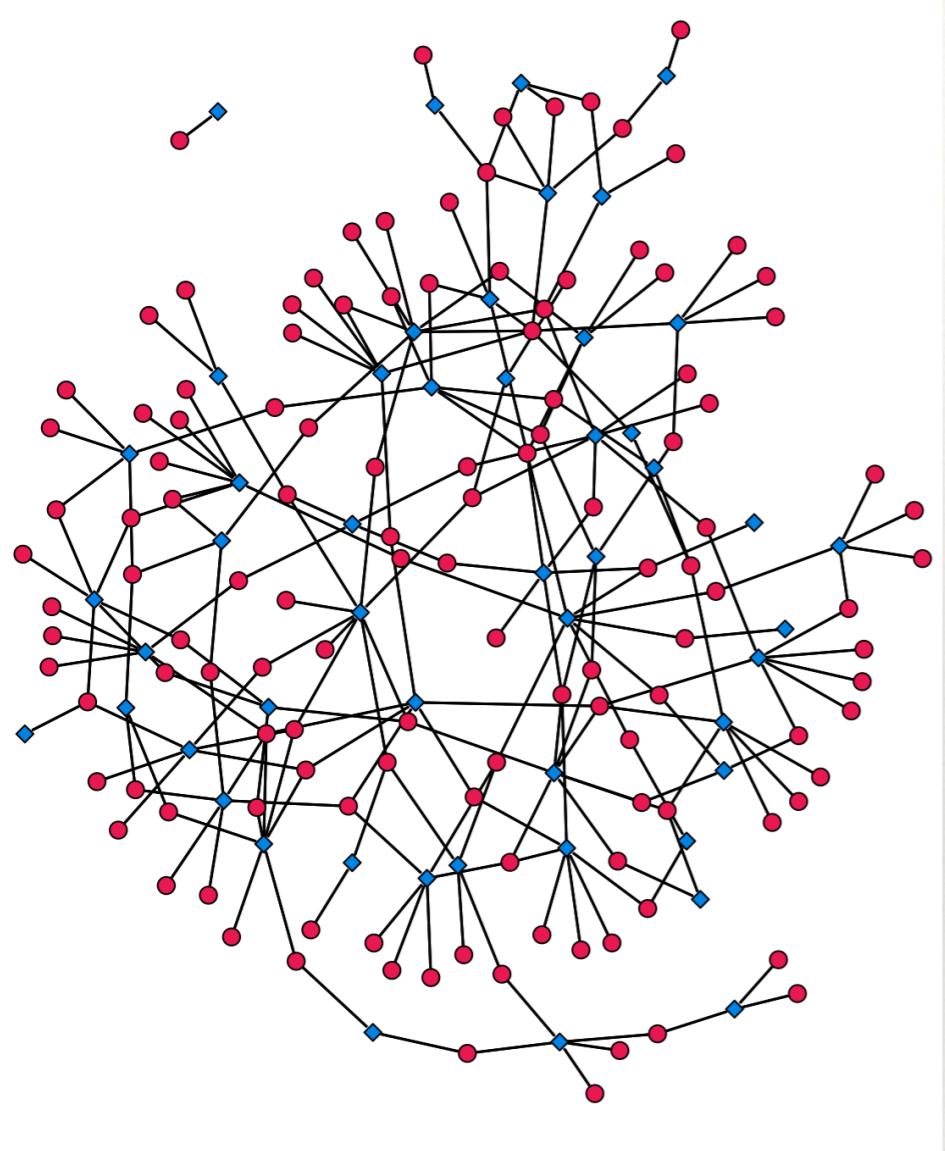
$$\mathbf{A} \times \mathbf{A}^T = \mathbf{P}$$

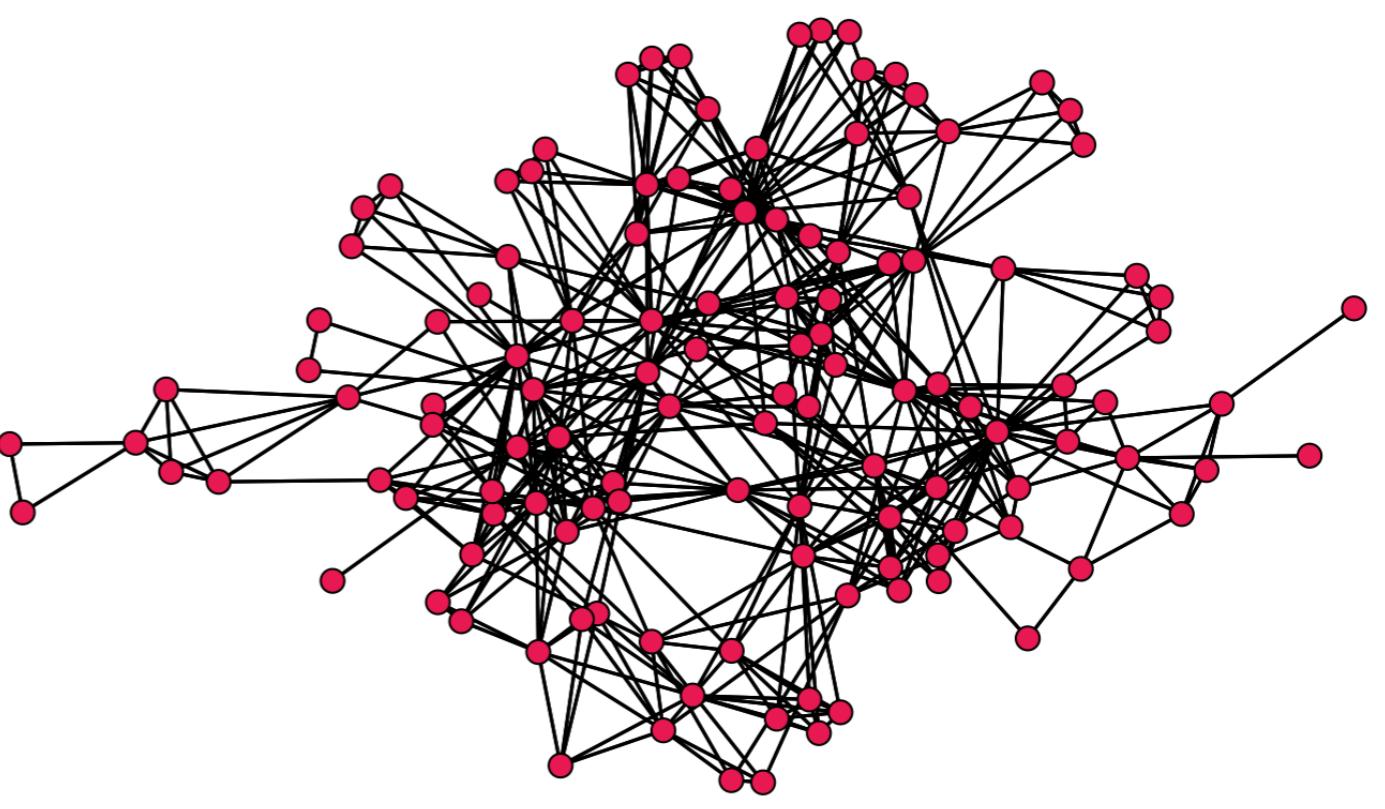
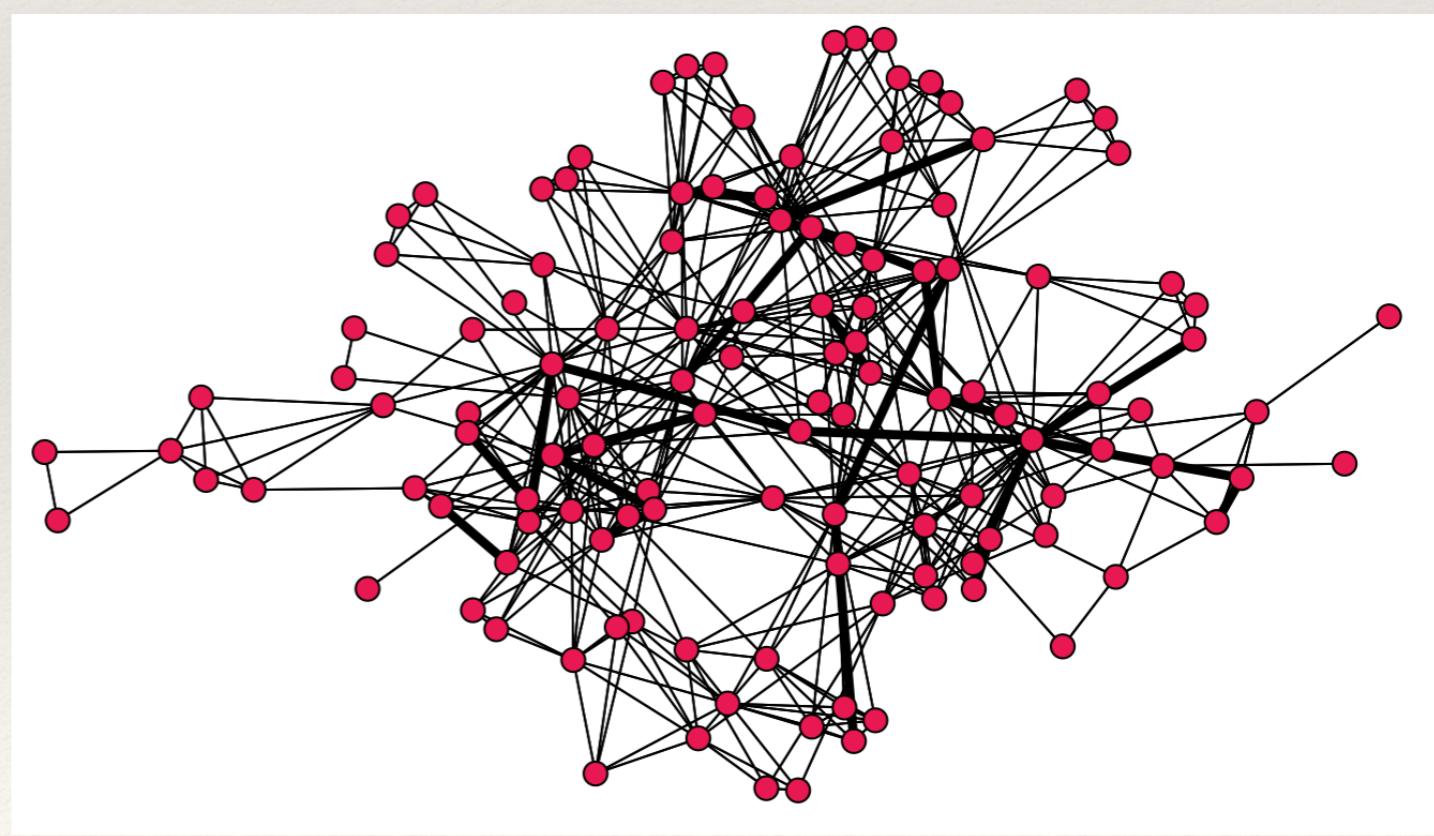
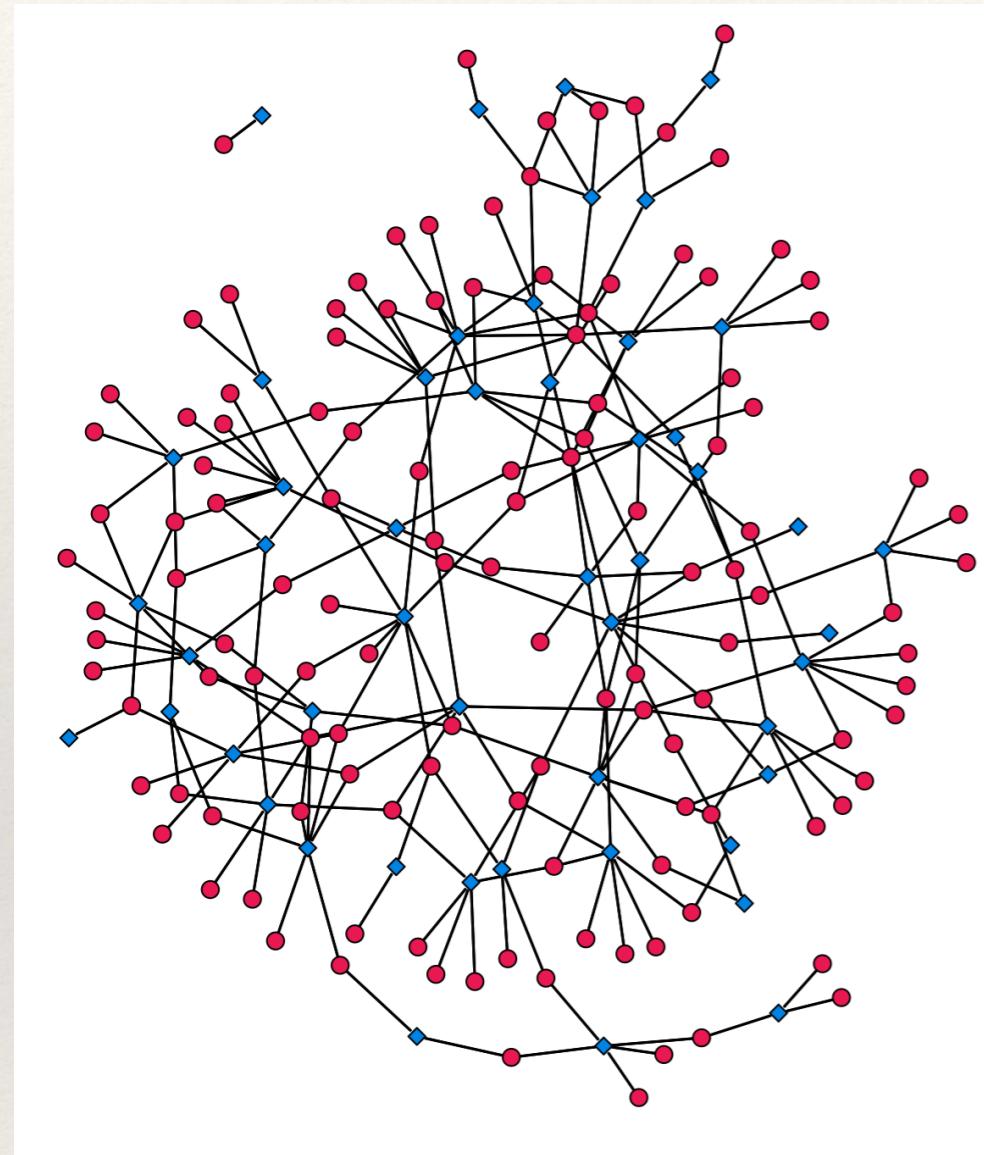
	A	B	C	D	E	F
A	2	1	2	1	0	0
B	1	1	1	0	0	0
C	2	1	2	1	0	0
D	1	0	1	4	1	2
E	0	0	0	1	1	1
F	0	0	0	2	1	2

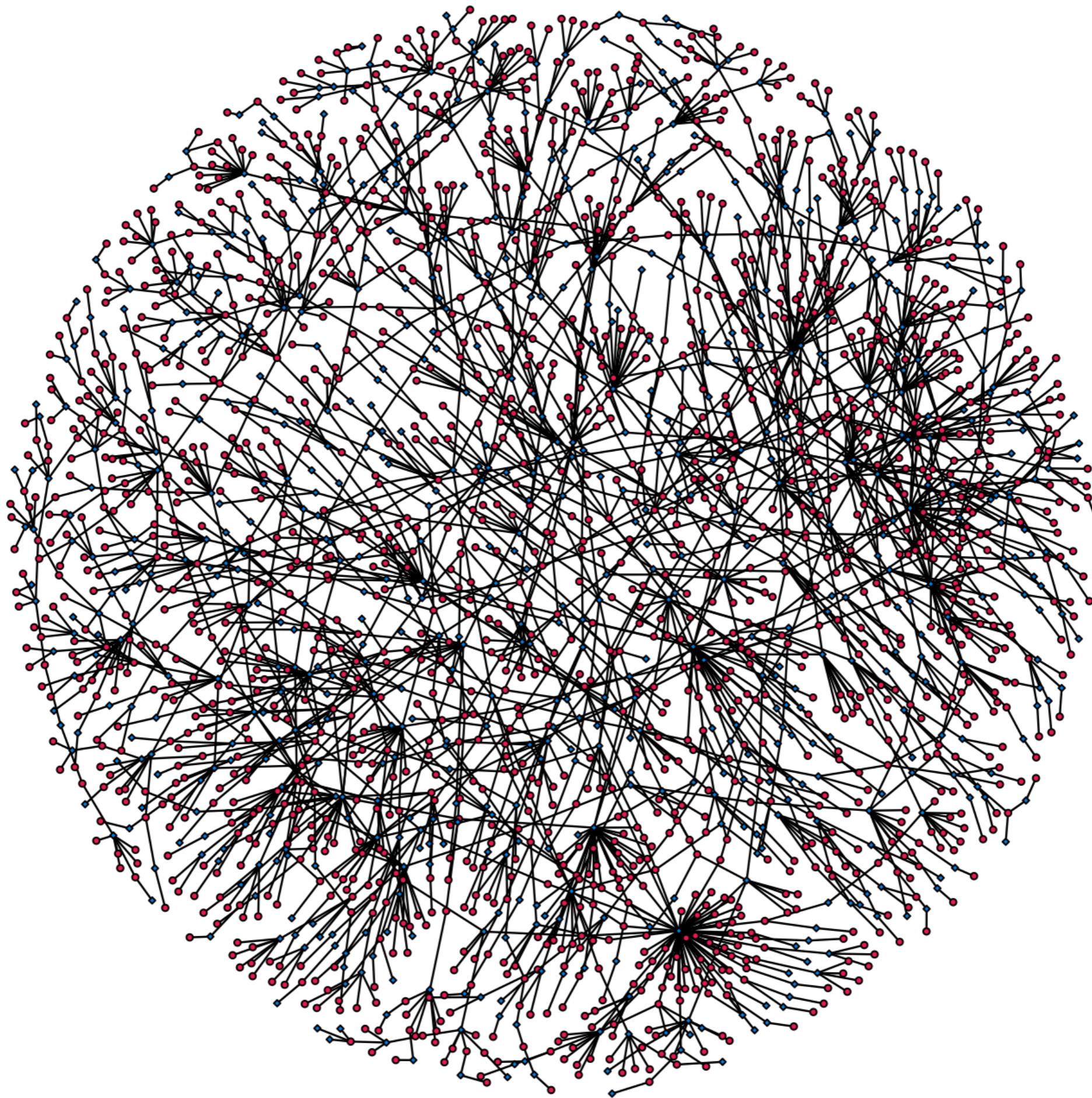
These weights are returned as the product matrix.

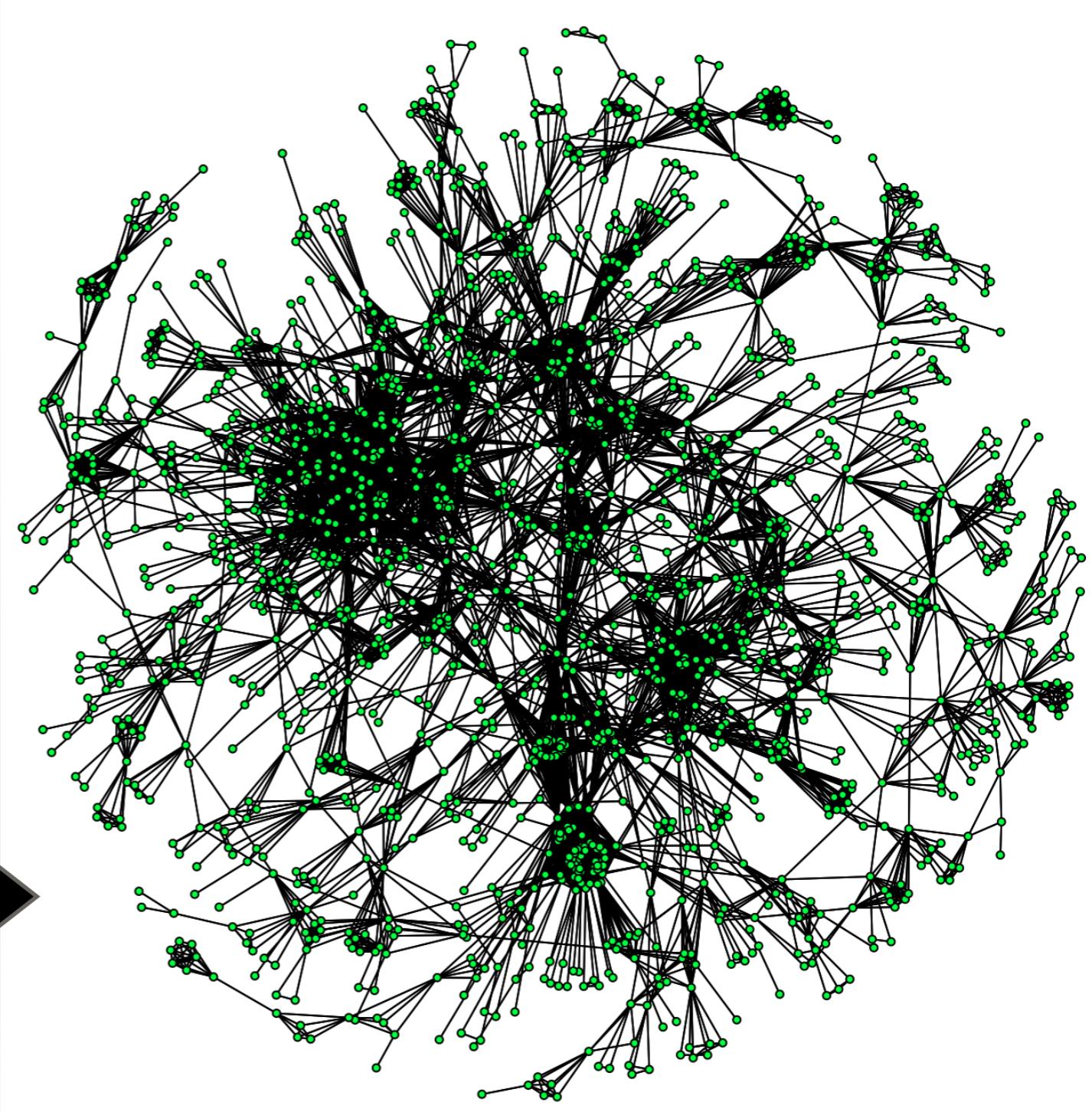
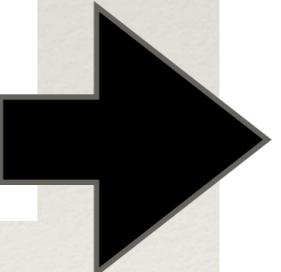
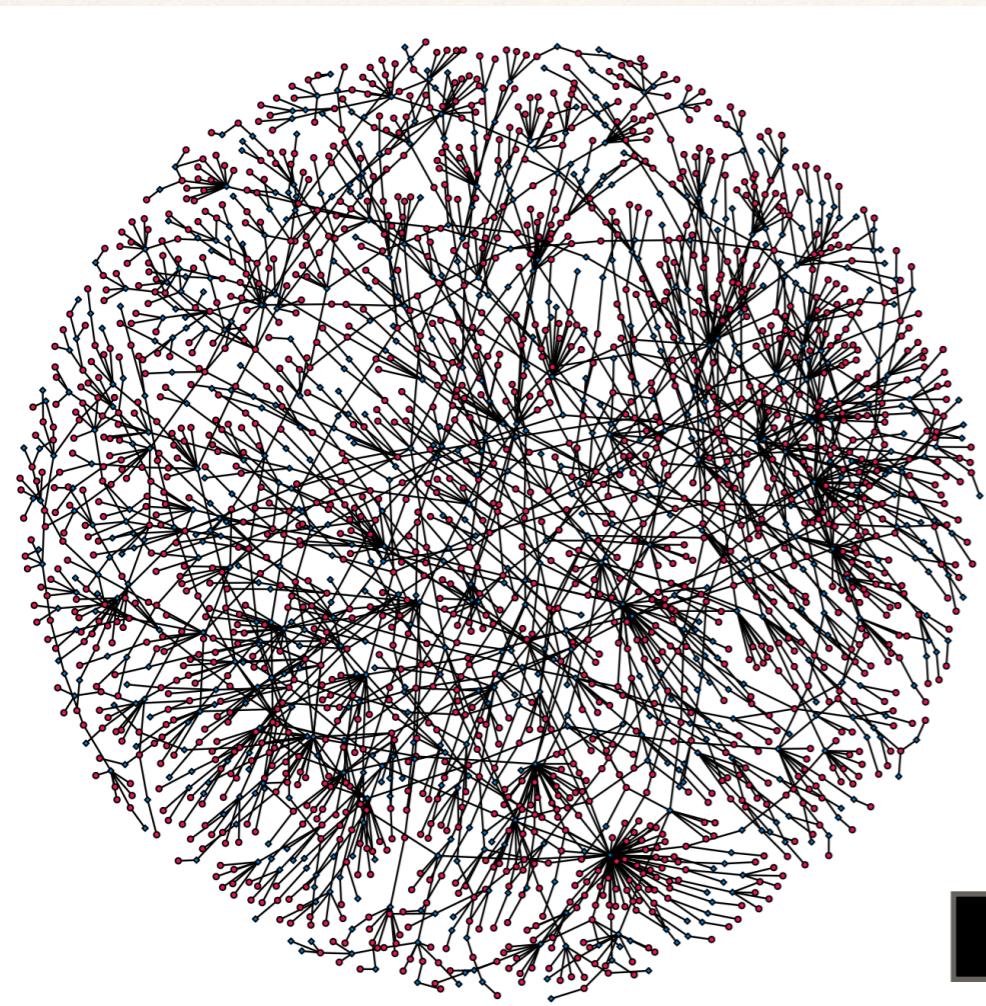


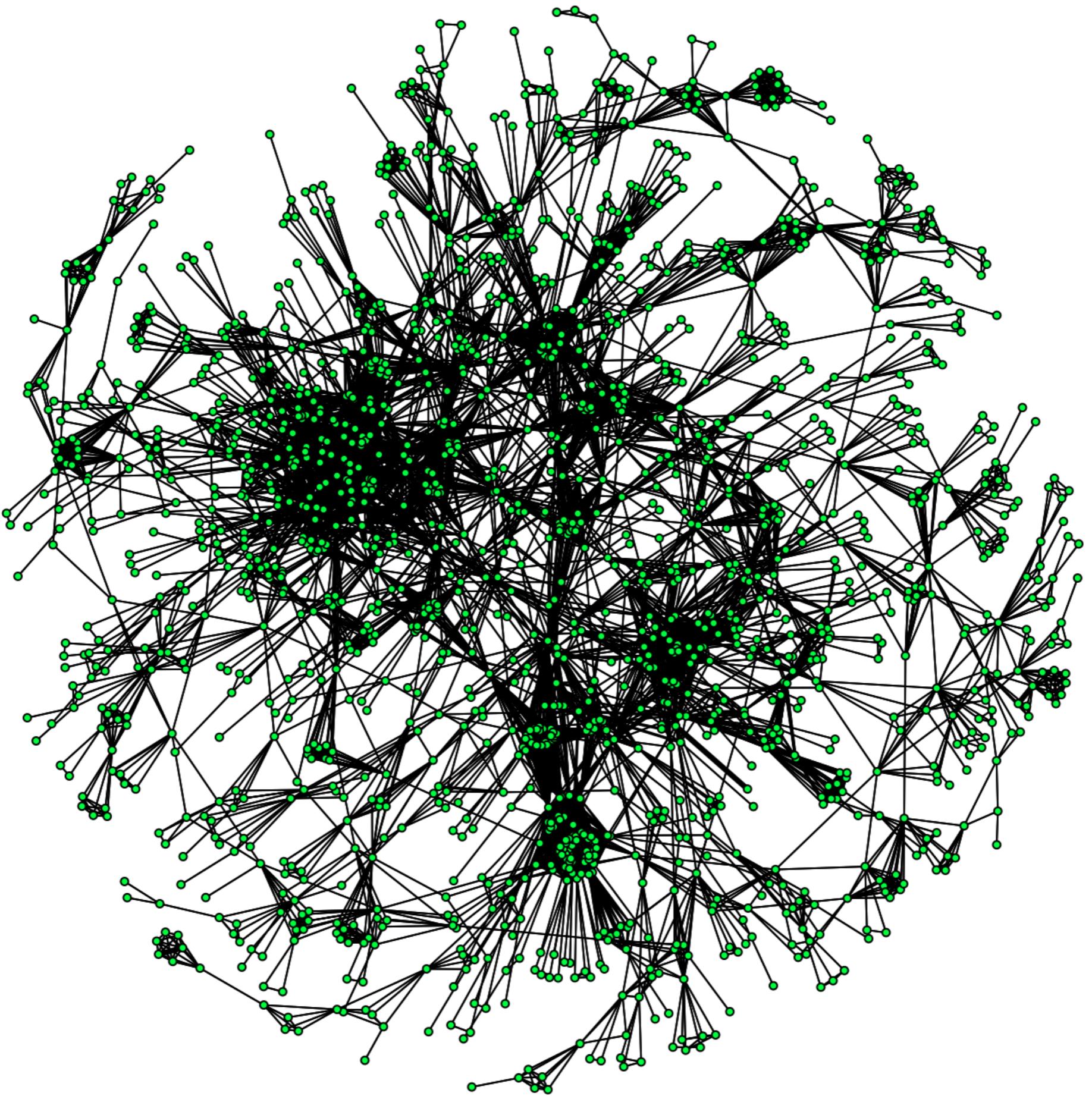












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# Learning Goals

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- ❖ At the end of the lecture, you should be able to answer these questions:
  - ❖ How can we create **unipartite** graphs from **bipartite** graphs?
  - ❖ What is the difference between **dichotomized** projections and **summation** projections?

Questions?