

Statistical Analysis of Networks

Projection & Weighted Graphs

Motivating Example *Revisited*

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

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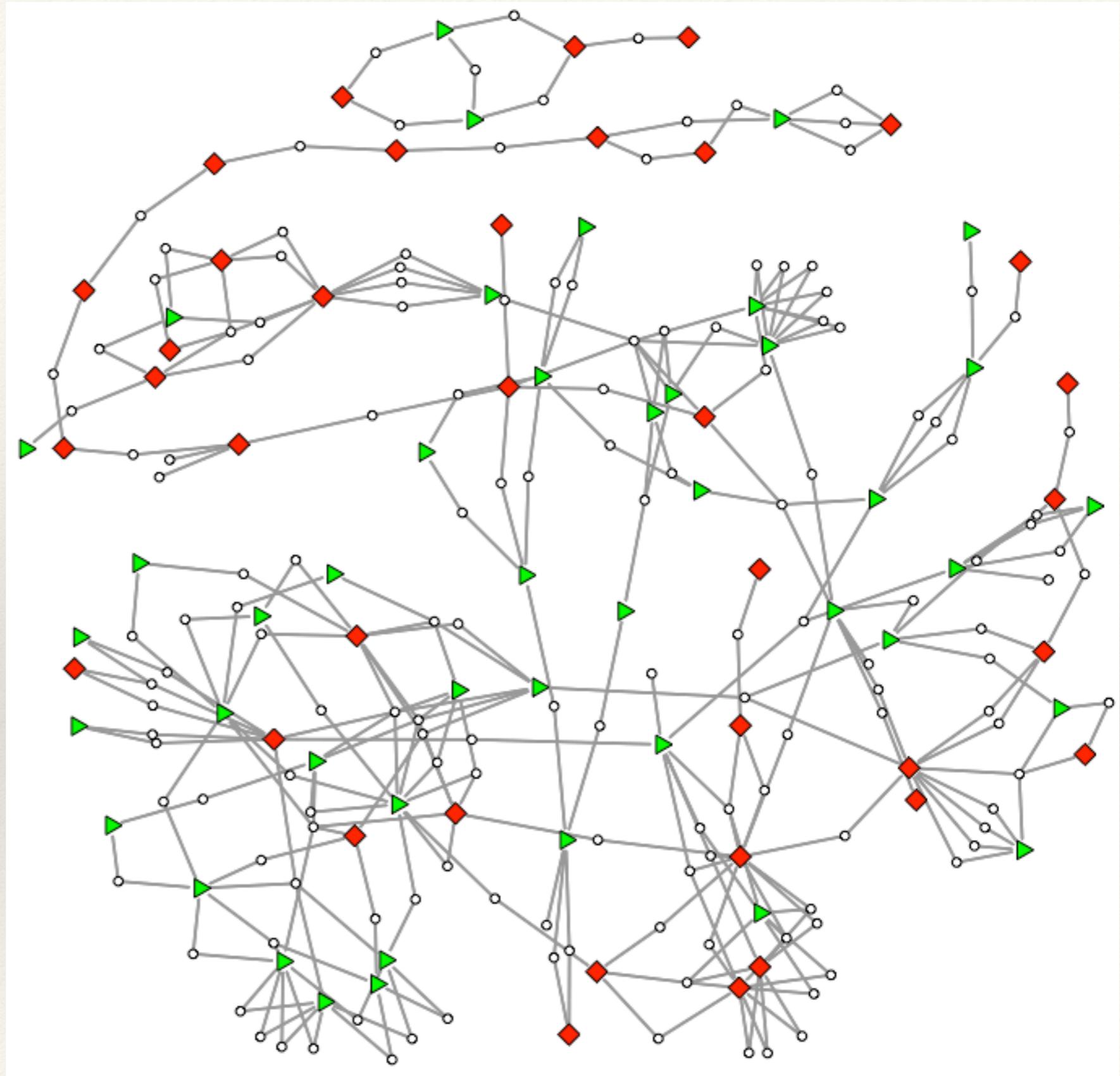
❖ Questions:

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

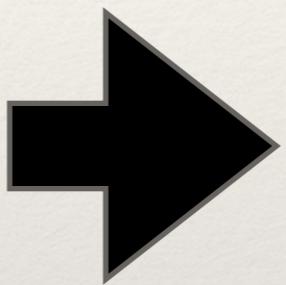
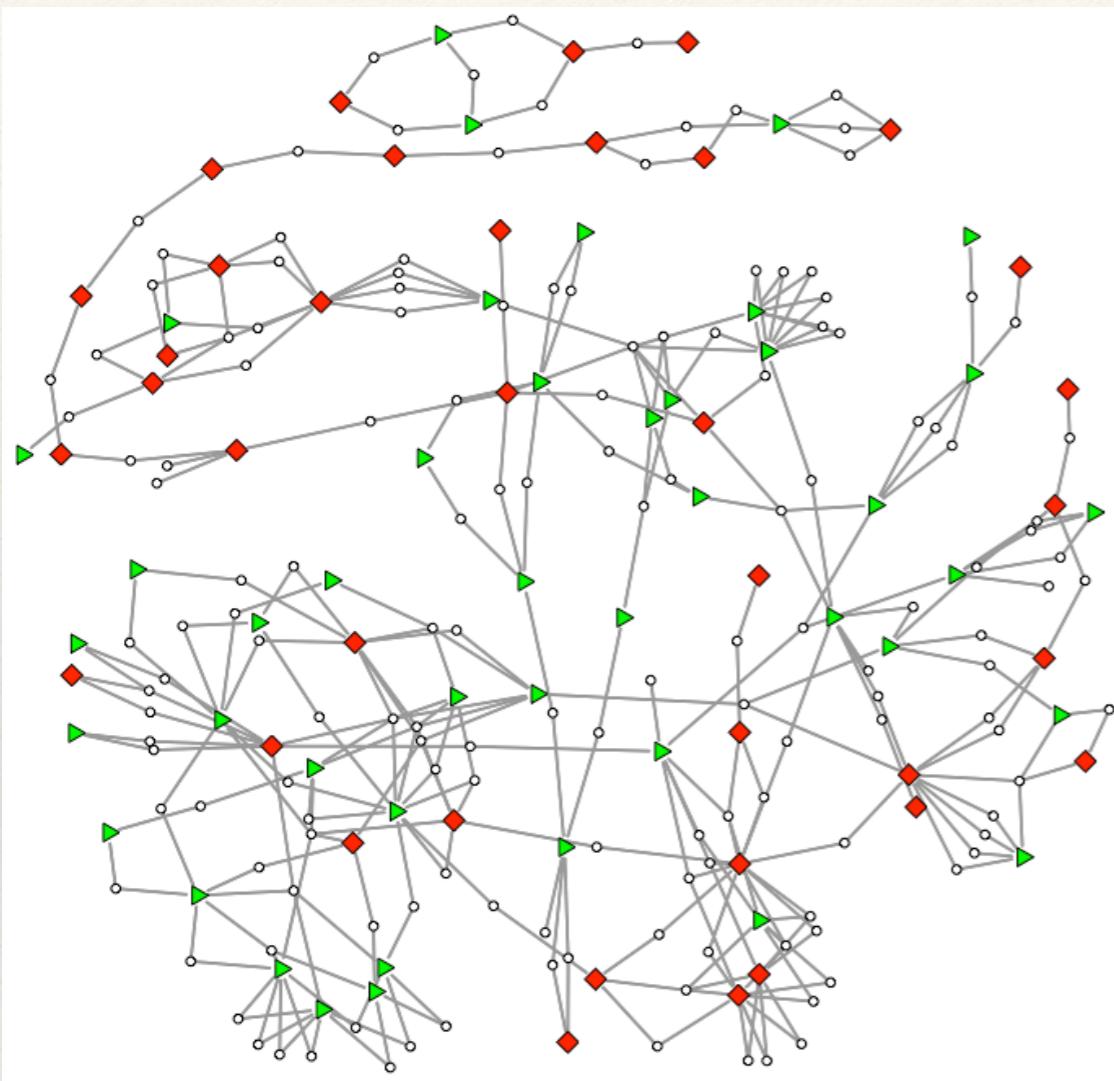
What is the
concept of
interest?

How is it
conceptualized?

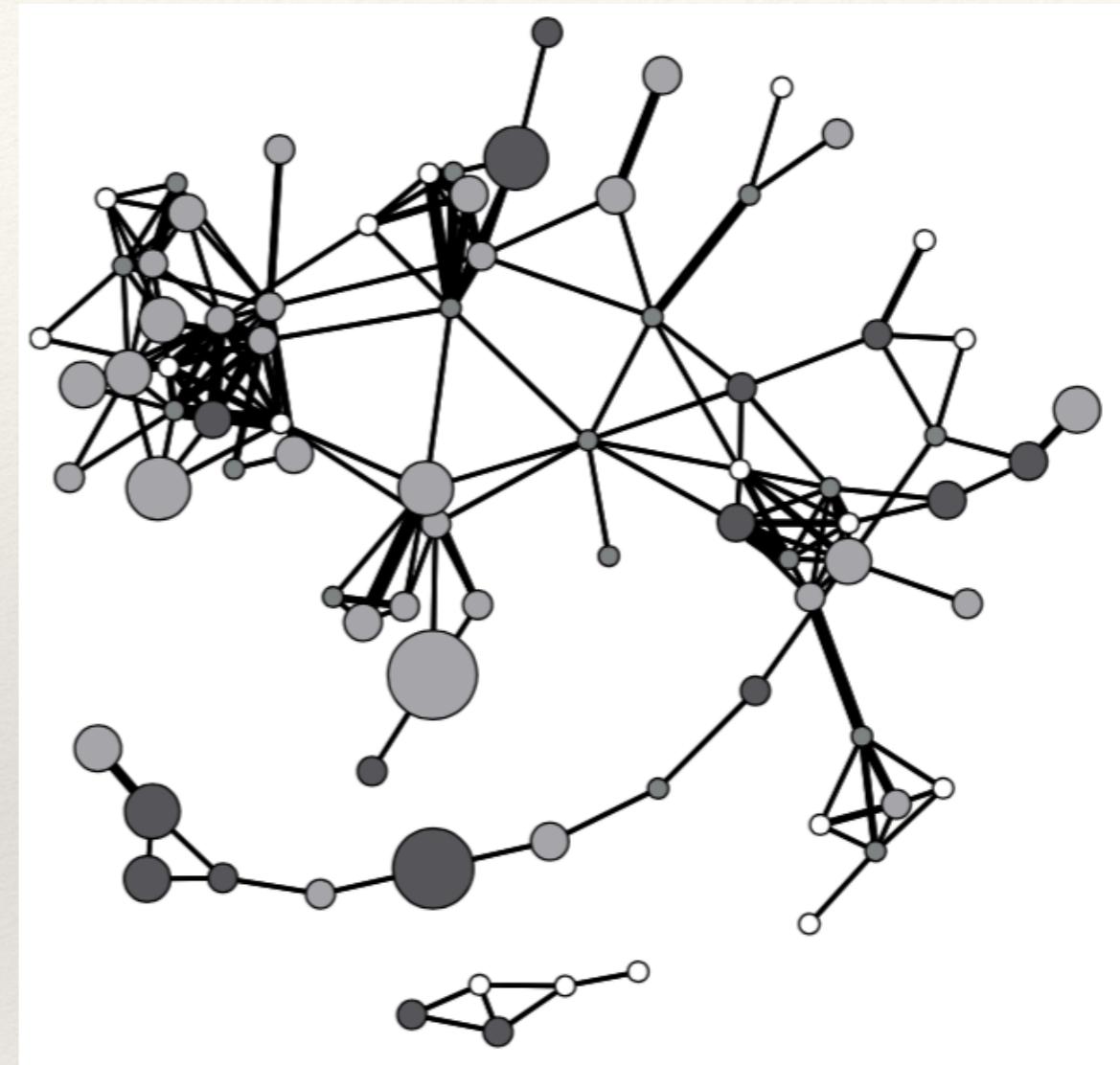
How is it
operationalized?



Two-mode network of officers connected by incidents



One-mode network of officers connected to officers



What do the connections represent in this network?

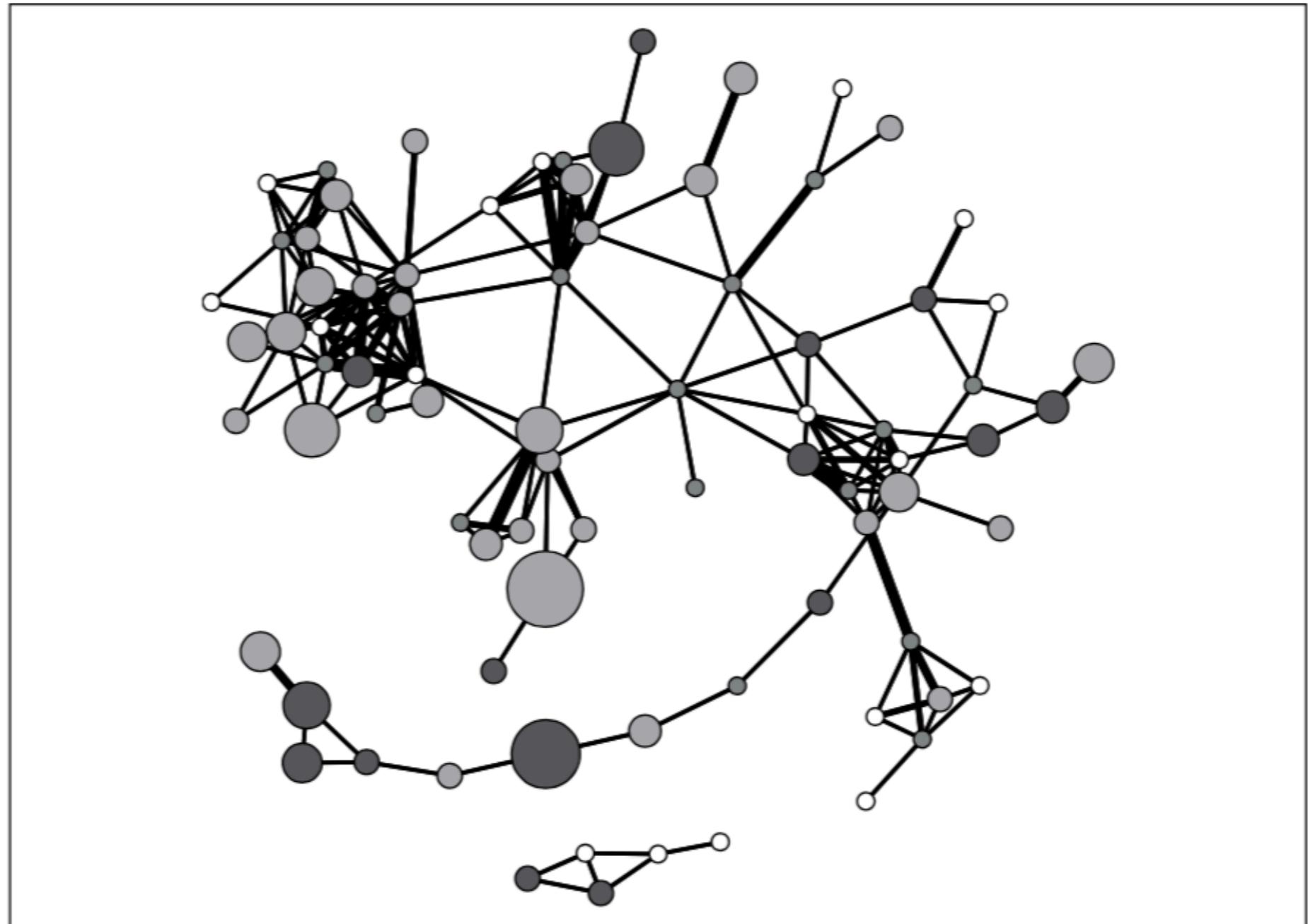


Figure 5. One-mode network of officers.

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

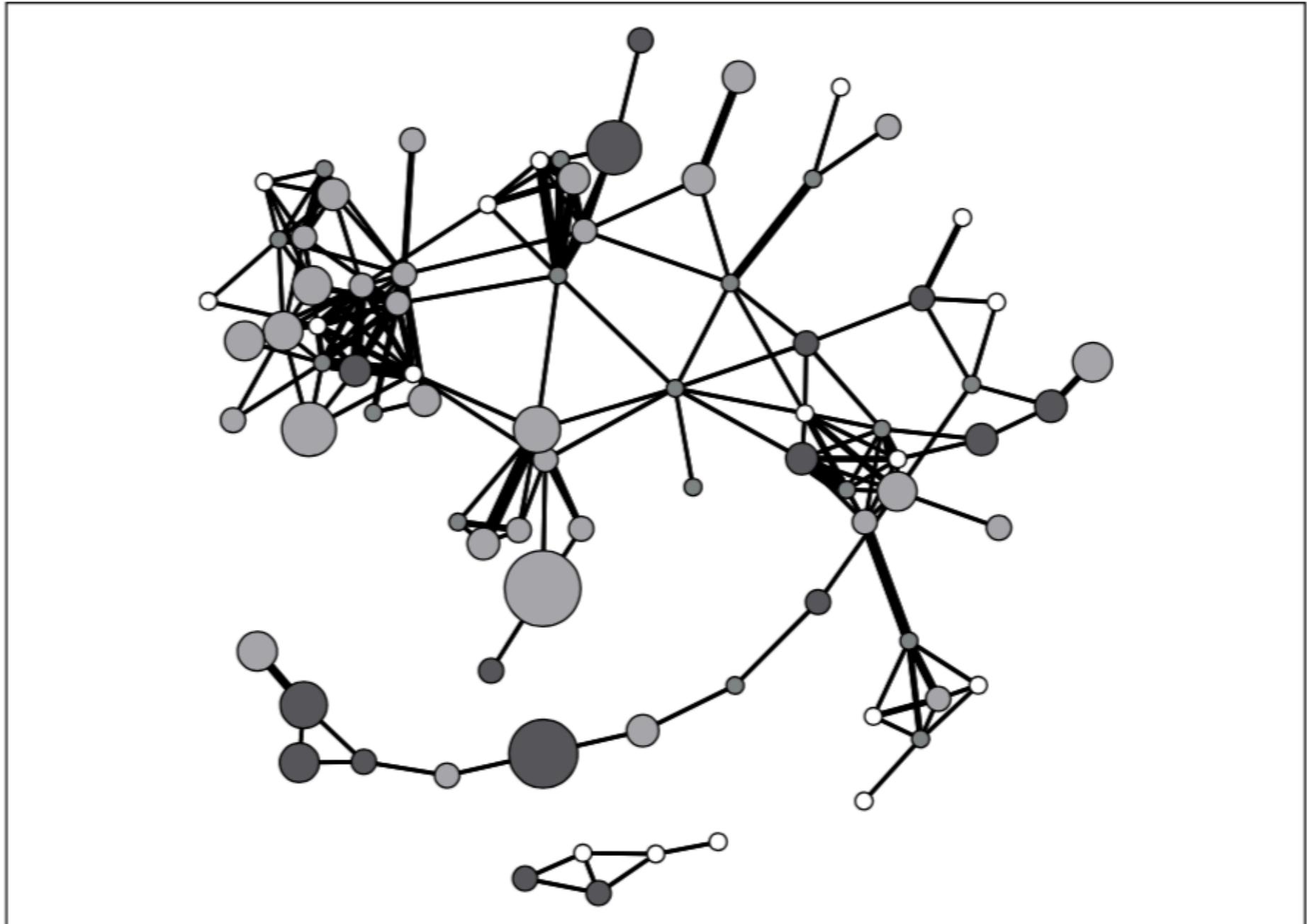


Figure 5. One-mode network of officers.

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

Projection & Weighted Graphs

Learning Goals

- ❖ 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?

Projection

- ❖ 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.

Projection

- ❖ 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:
 - ❖ $A_{ij}^T = A_{ji}$

Projection

- ❖ 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^T**

	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

Projection

- ❖ 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: $5 \times 6 \times 6 \times 5 = 5 \times 5$

Projection

- ❖ 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
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D	0	1	1	1	1
E	0	0	1	0	0
F	0	0	1	0	1

Matrix **A^T**

	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^T**

	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

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

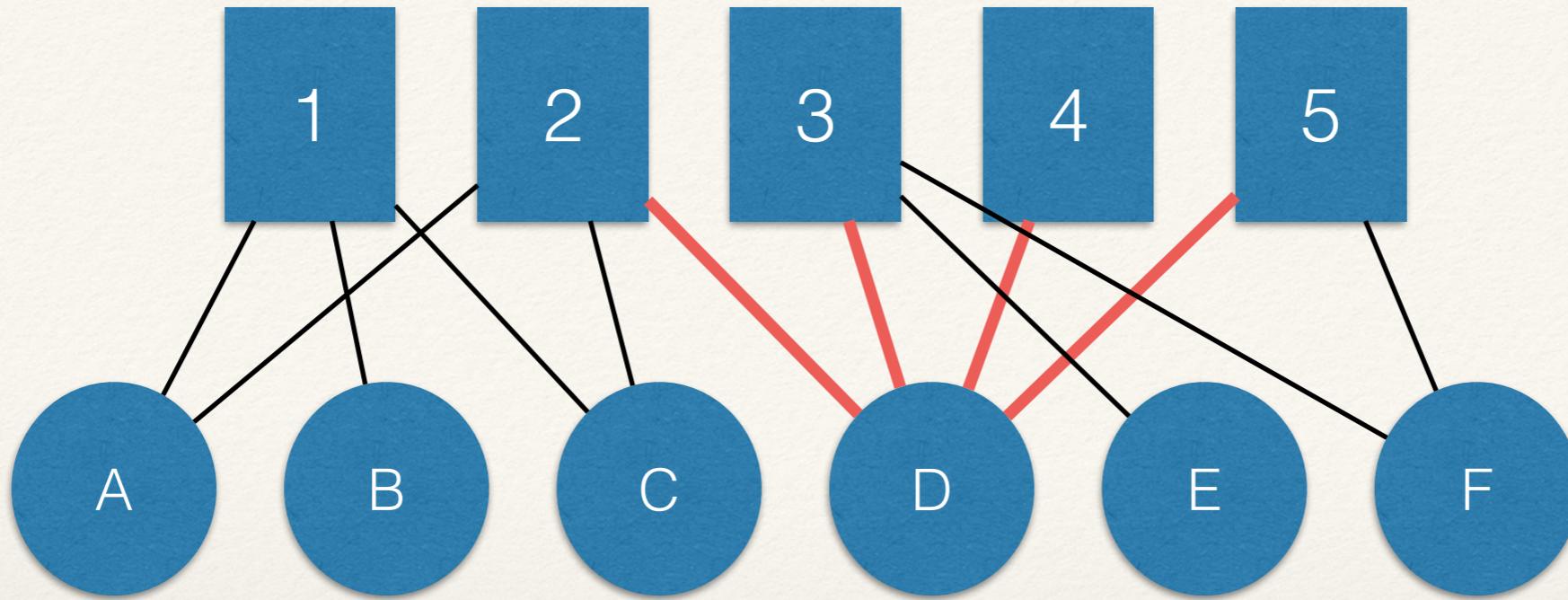
Projection by Multiplication

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

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



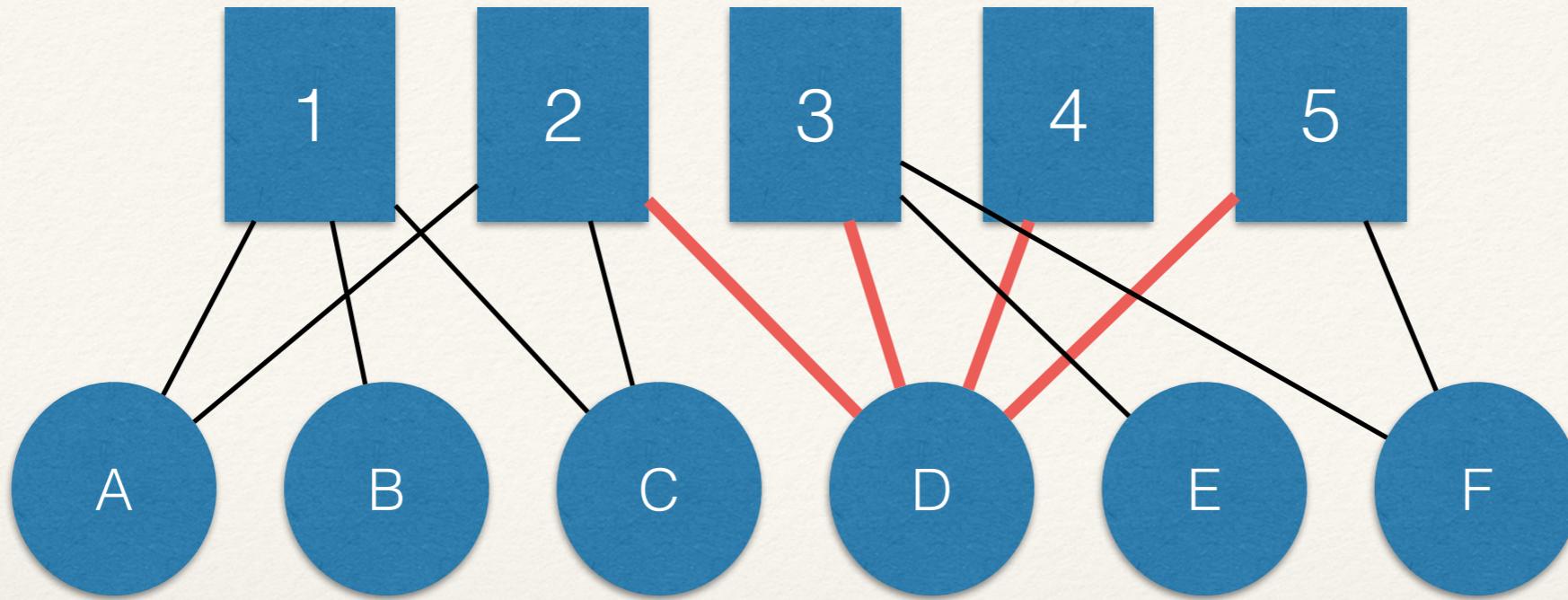
$$\mathbf{A} \times \mathbf{A}^T = \mathbf{P}$$

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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

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

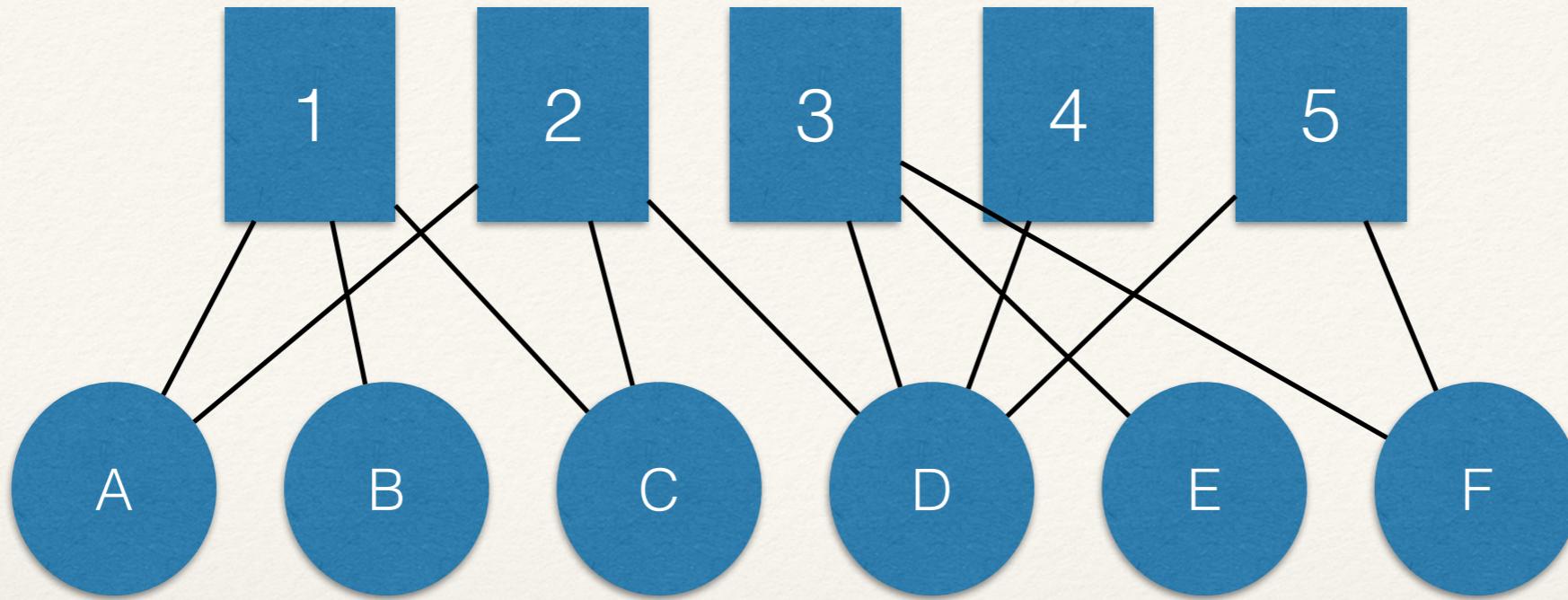


$$\mathbf{A} \times \mathbf{A}^T = \mathbf{P}$$

	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?

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

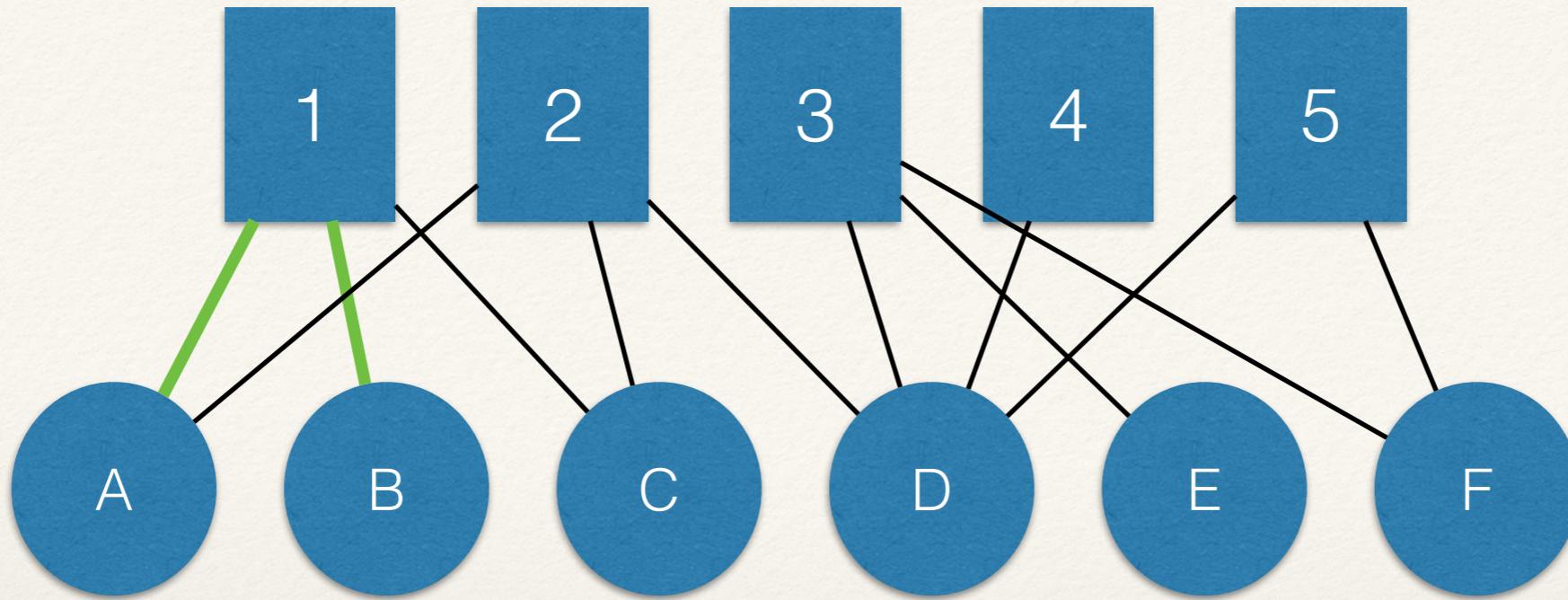


$$\mathbf{A} \times \mathbf{A}^T = \mathbf{P}$$

	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)

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



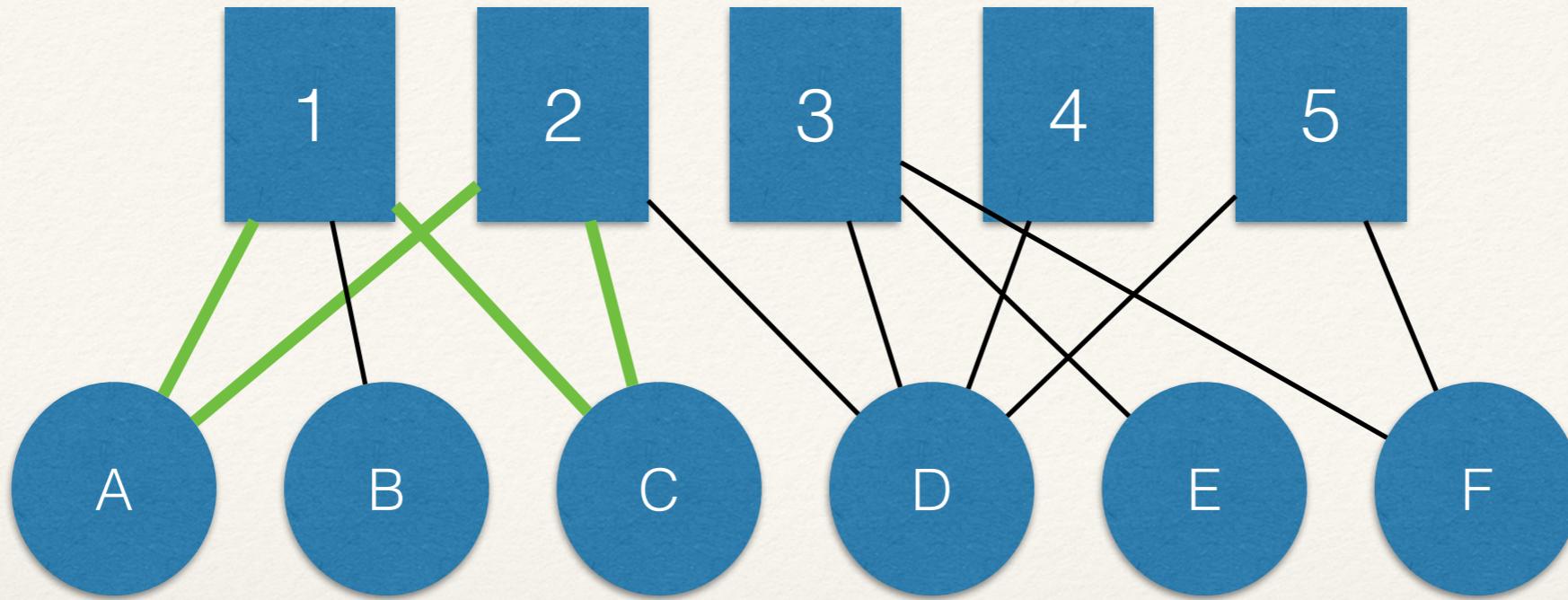
$$\mathbf{A} \times \mathbf{A}^T = \mathbf{P}$$

	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

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



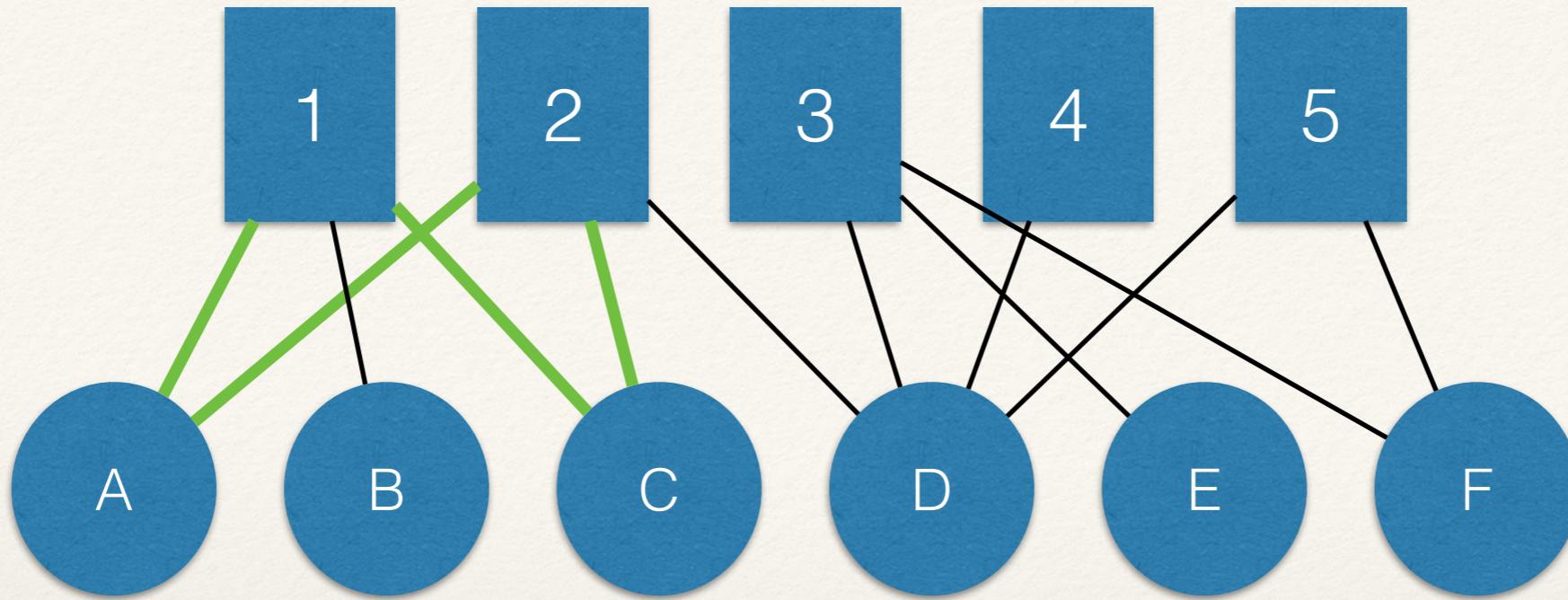
$$\mathbf{A} \times \mathbf{A}^T = \mathbf{P}$$

	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

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



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

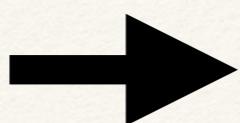
	A	B	C	D	E	F
A	2	1	2	1	0	0
B	1	1	1	0	0	0
C	2	1	2	1	0	0
D	1	0	1	4	1	2
E	0	0	0	1	1	1
F	0	0	0	2	1	2

**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

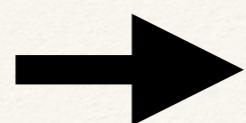
	A	B	C	D	E	F
A	2	1	2	1	0	0
B	1	1	1	0	0	0
C	2	1	2	1	0	0
D	1	0	1	4	1	2
E	0	0	0	1	1	1
F	0	0	0	2	1	2



	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

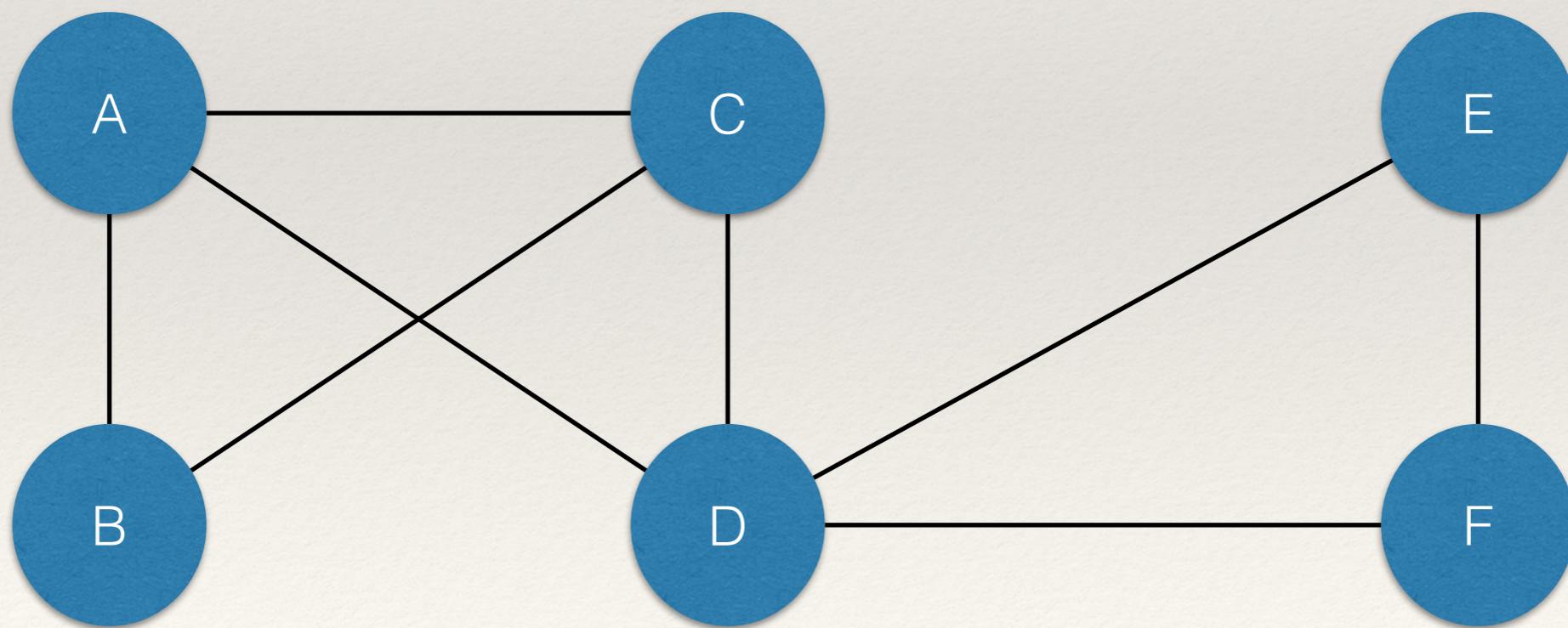
If we treat any tie greater than 0 as binary, *called **dichotomizing***, and recode the diagonal as 0, we get an undirected, one-mode network

	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

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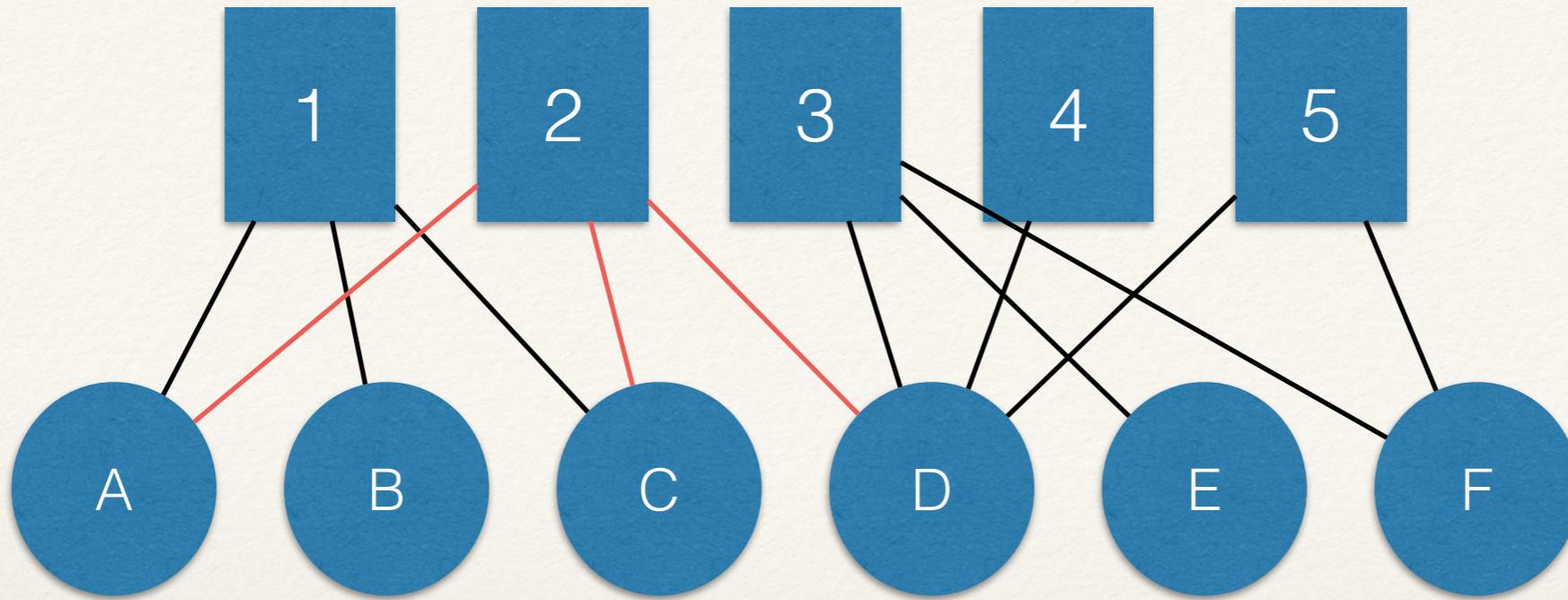
Projection by Multiplication

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

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



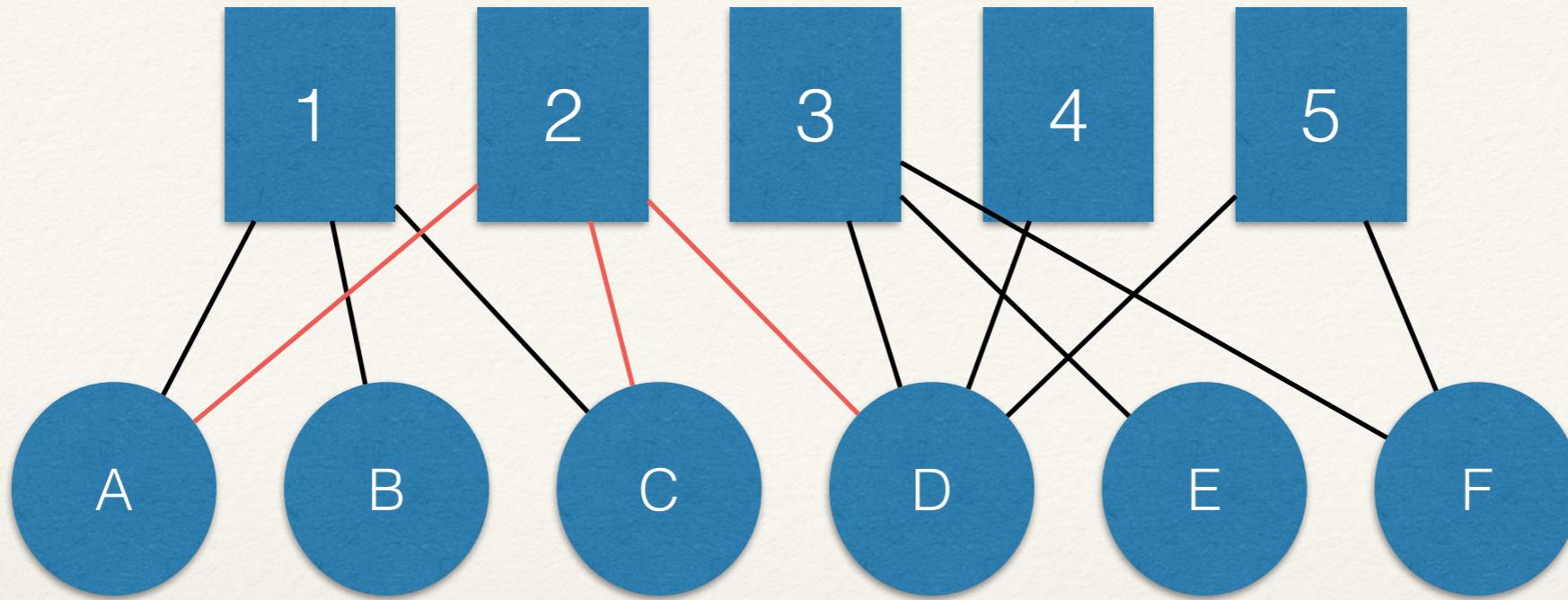
$$\mathbf{A}^T \times \mathbf{A} = \mathbf{G}$$

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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

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

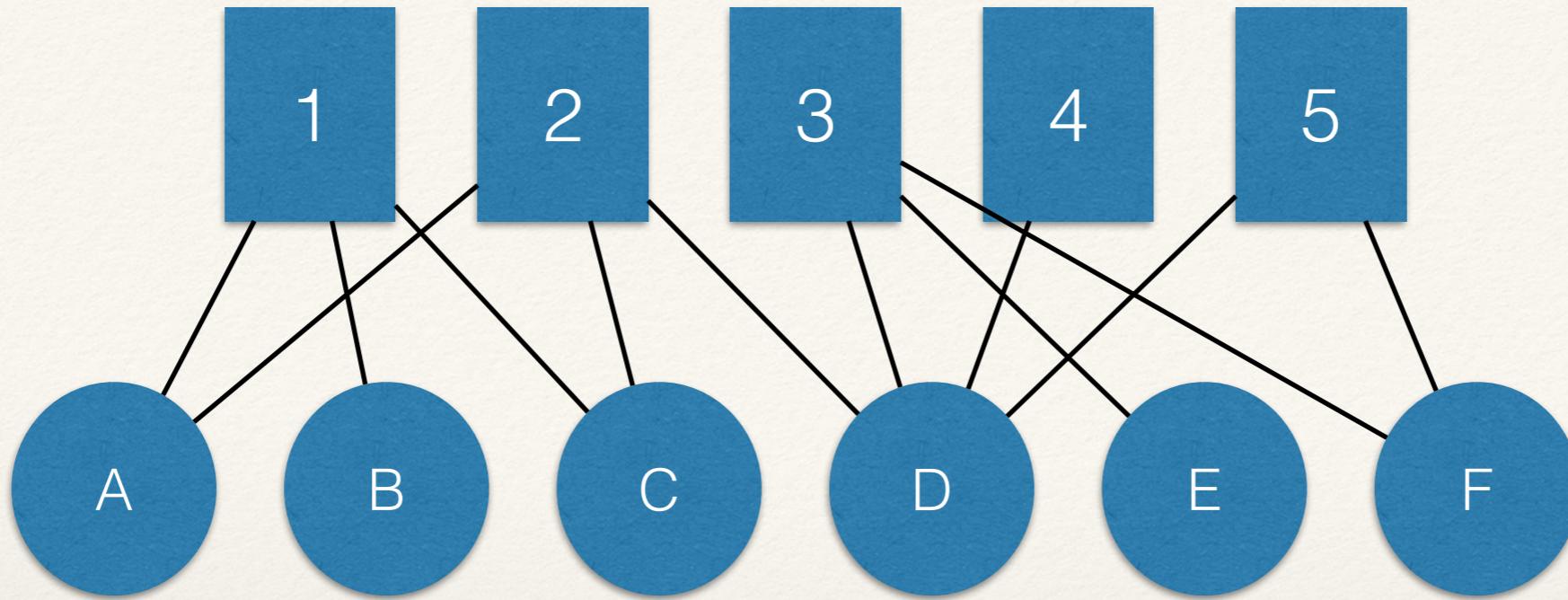


$$\mathbf{A}^T \times \mathbf{A} = \mathbf{G}$$

	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?

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

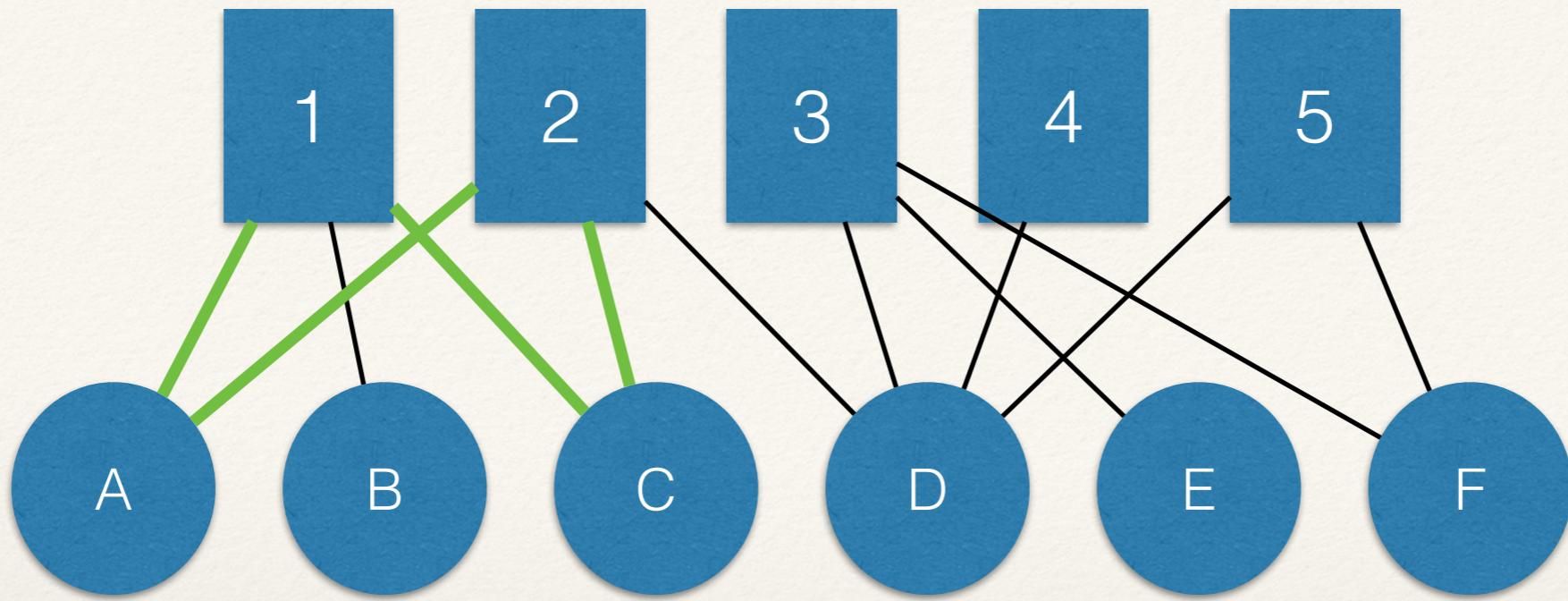


$$\mathbf{A}^T \times \mathbf{A} = \mathbf{G}$$

	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)

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



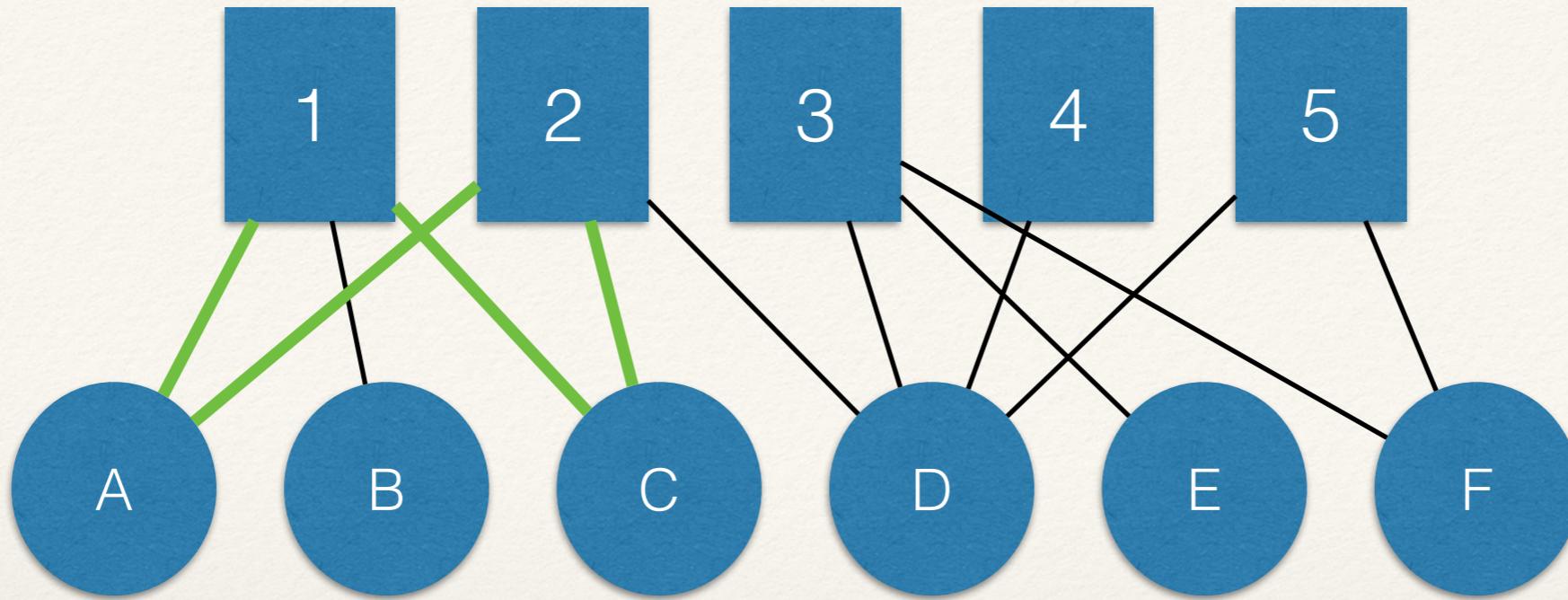
$$\mathbf{A}^T \times \mathbf{A} = \mathbf{G}$$

	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

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



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

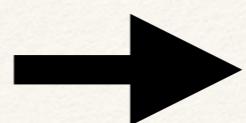
	1	2	3	4	5
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4	0	1	1	1	1
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**NOTE: these are counts
of shared vertices, not
edge counts**

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

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

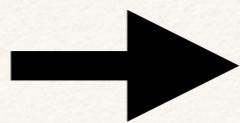
	1	2	3	4	5
1	3	2	0	0	0
2	2	3	1	1	1
3	0	1	3	1	2
4	0	1	1	1	1
5	0	1	2	1	2



	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

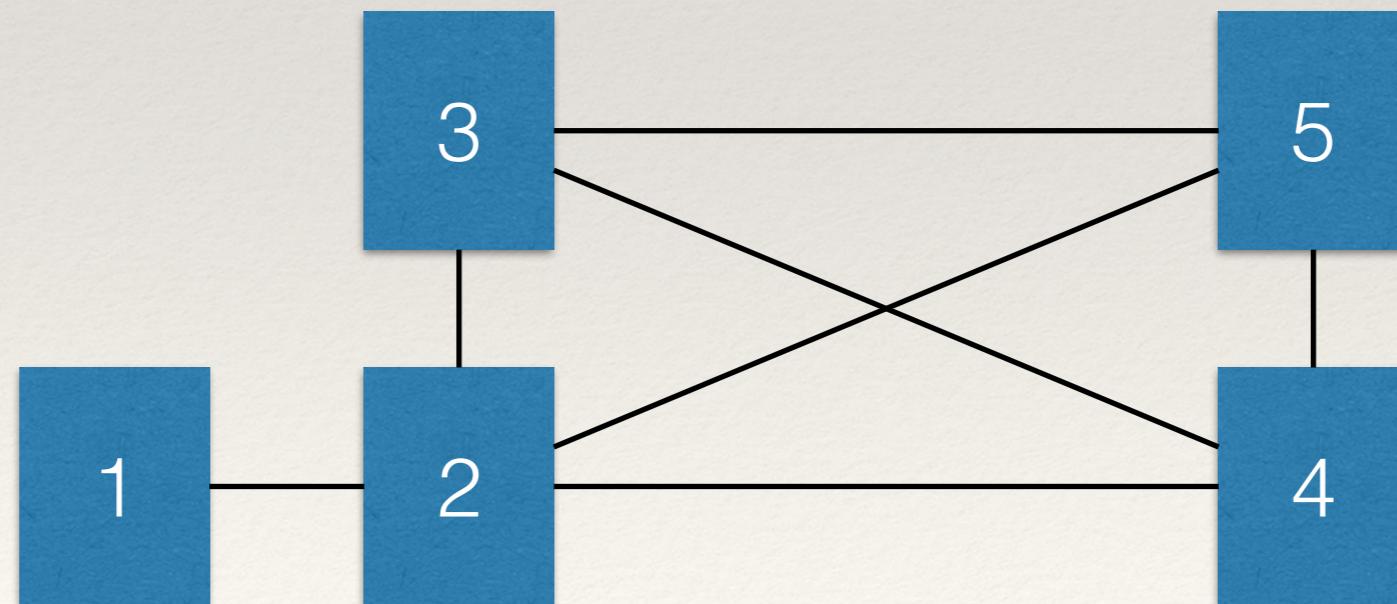
If we treat any tie greater than 0 as binary, *called **dichotomizing***, and recode the diagonal as 0, we get an undirected, one-mode network

	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

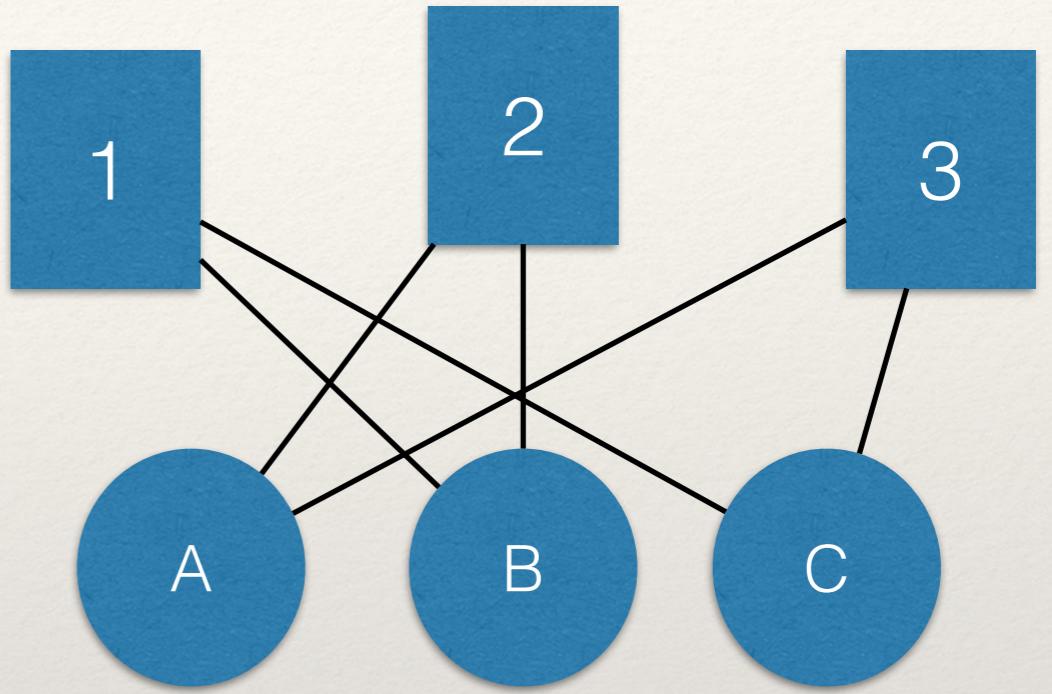
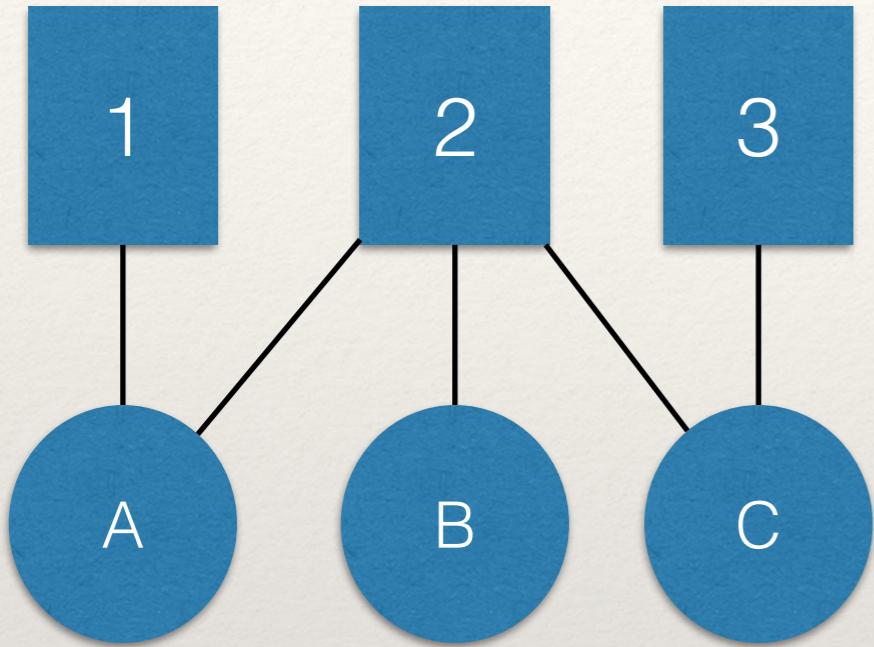
If we treat any tie greater than 0 as binary, *called **dichotomizing***, and recode the diagonal as 0, we get an undirected, one-mode network



Projection

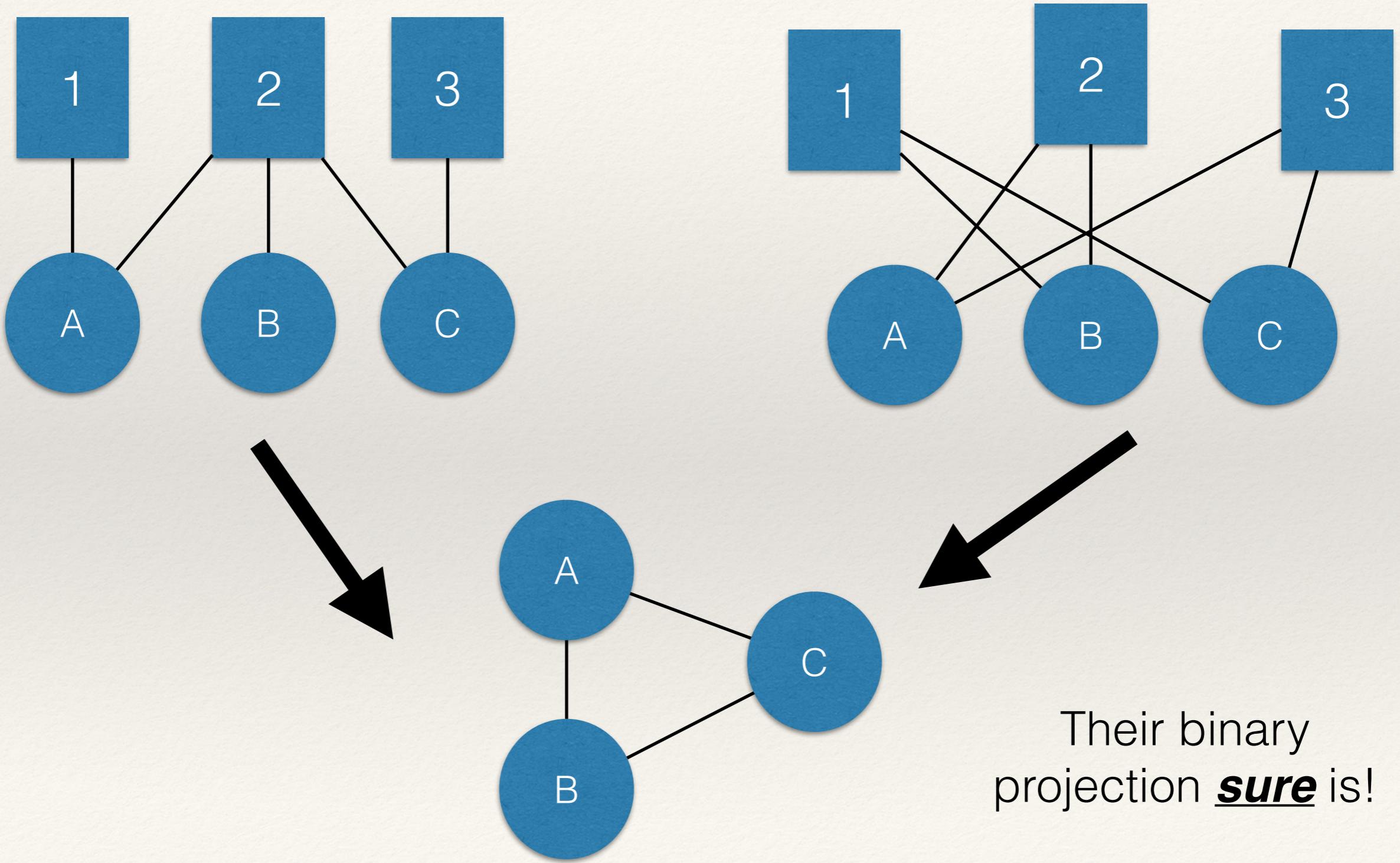
- ❖ To project, or not to project?
 - ❖ As noted by many scholars, there is data loss when we project and binarize the data.
 - ❖ Sometimes, this can be misleading.

Projection and Data Loss

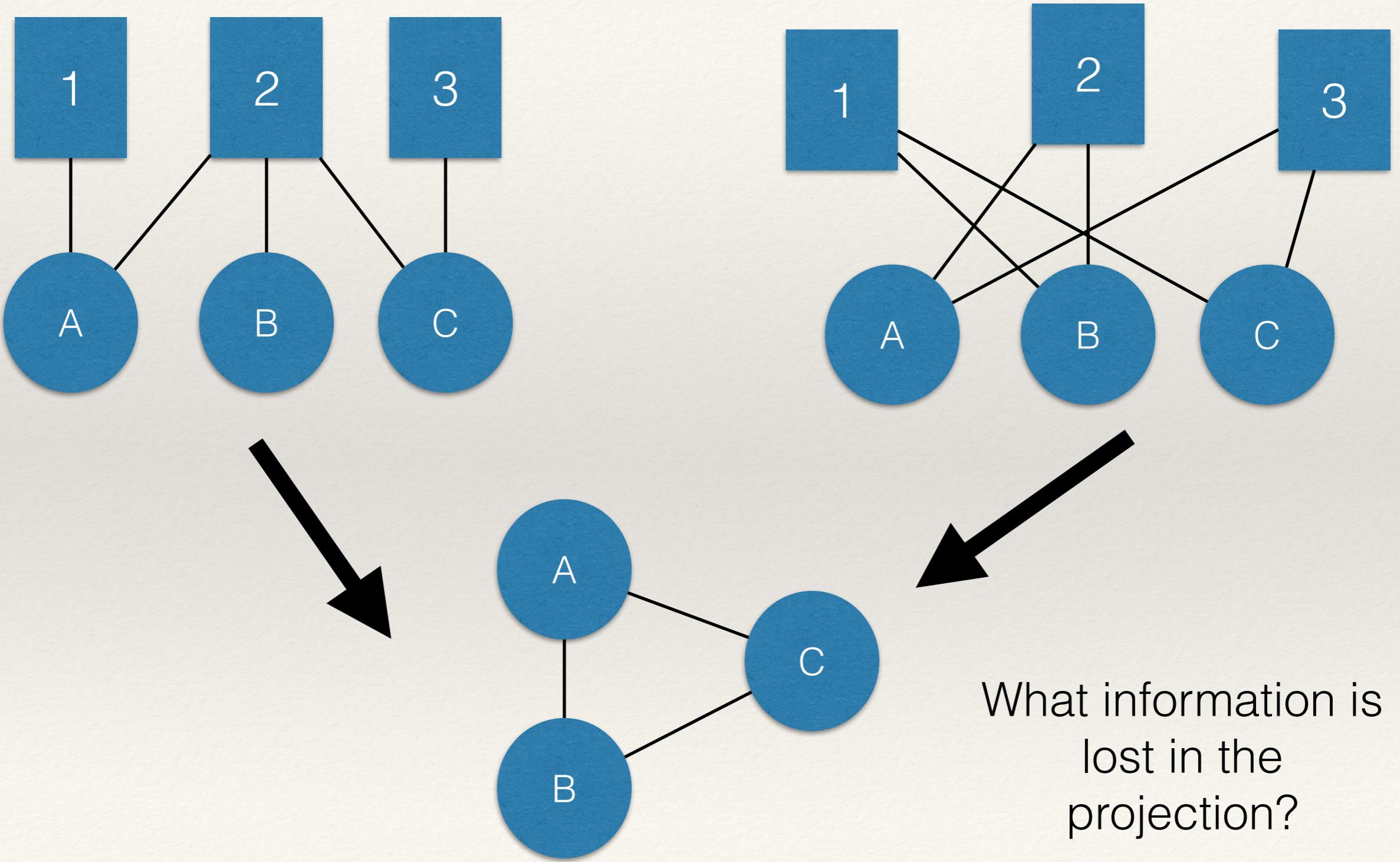


Are these bipartite graphs the same?

Projection and Data Loss



Projection and Data Loss



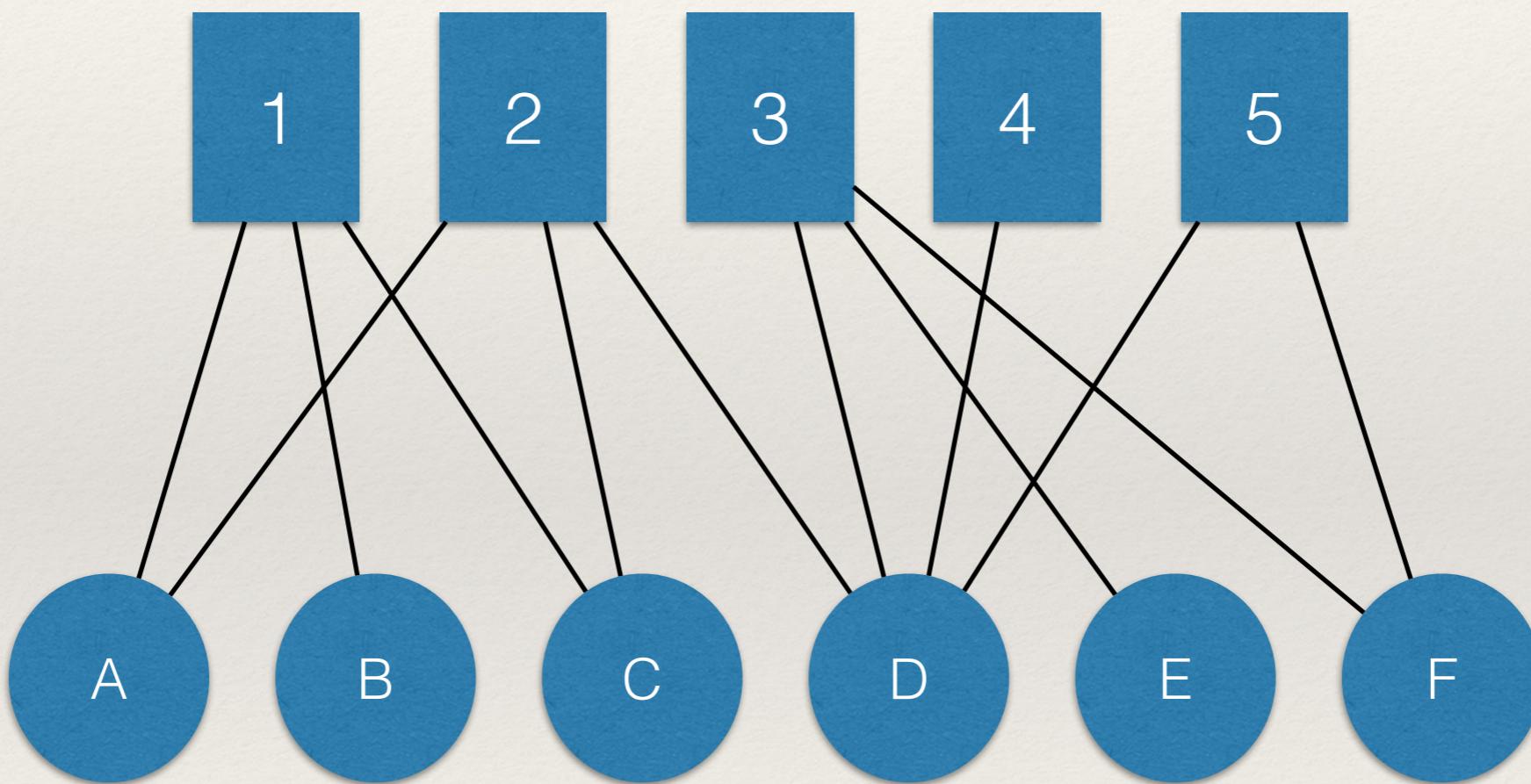
Projection

- ❖ 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.

Weighted Edges

- ❖ 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

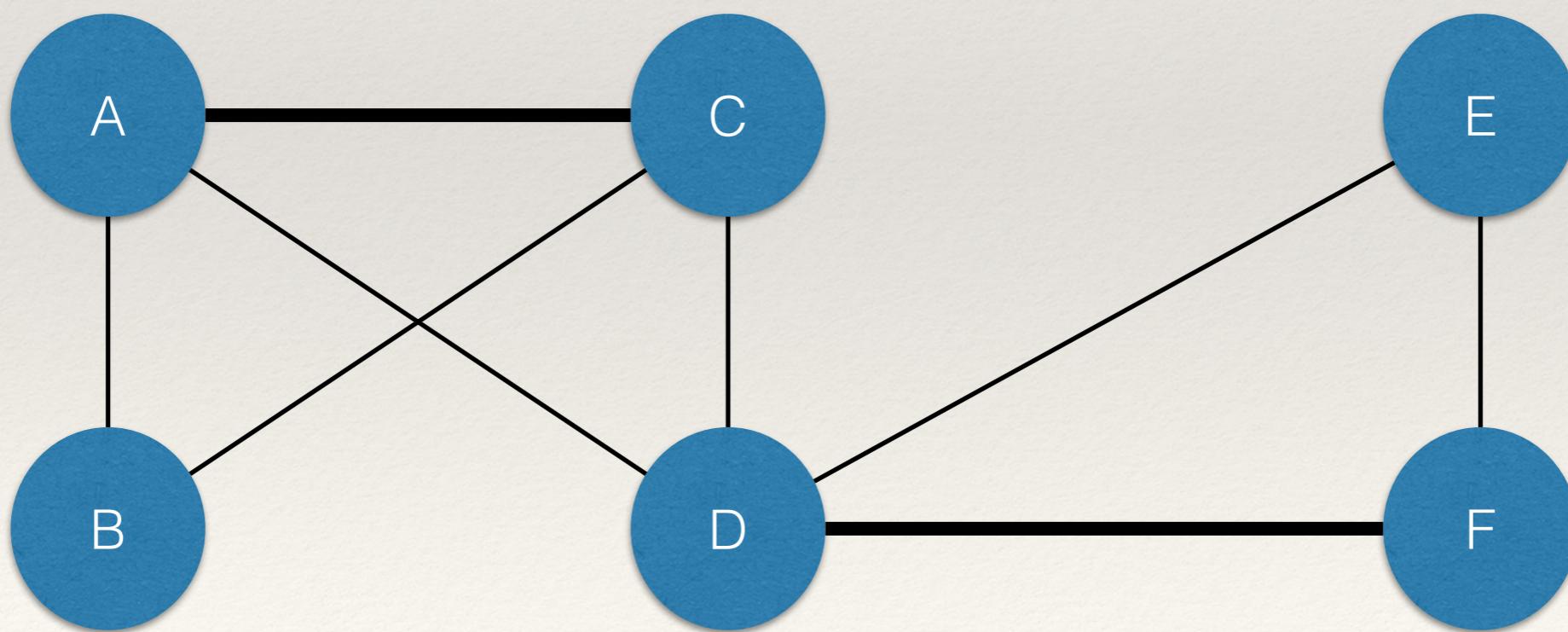
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$$\mathbf{A} \times \mathbf{A}^T = \mathbf{P}$$

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D	1	0	1	4	1	2
E	0	0	0	1	1	1
F	0	0	0	2	1	2

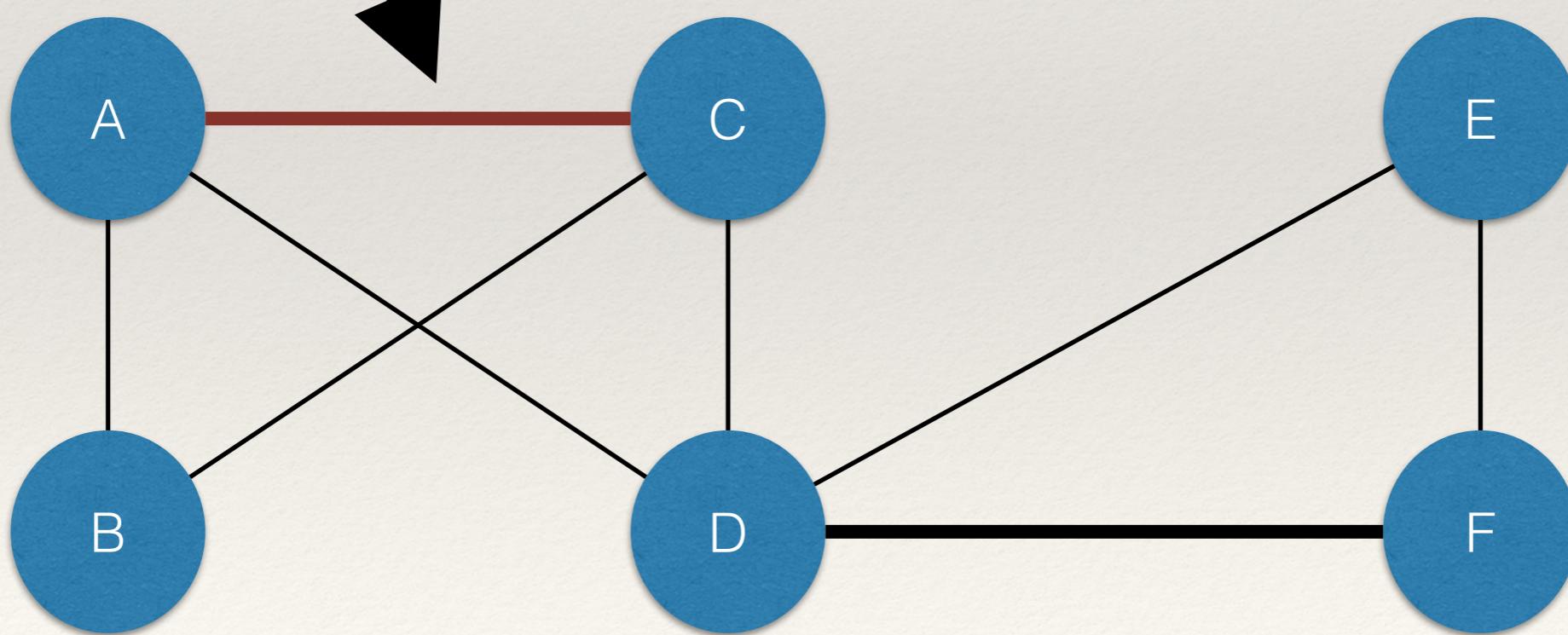
The off-diagonal entries are the tie weights



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

	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

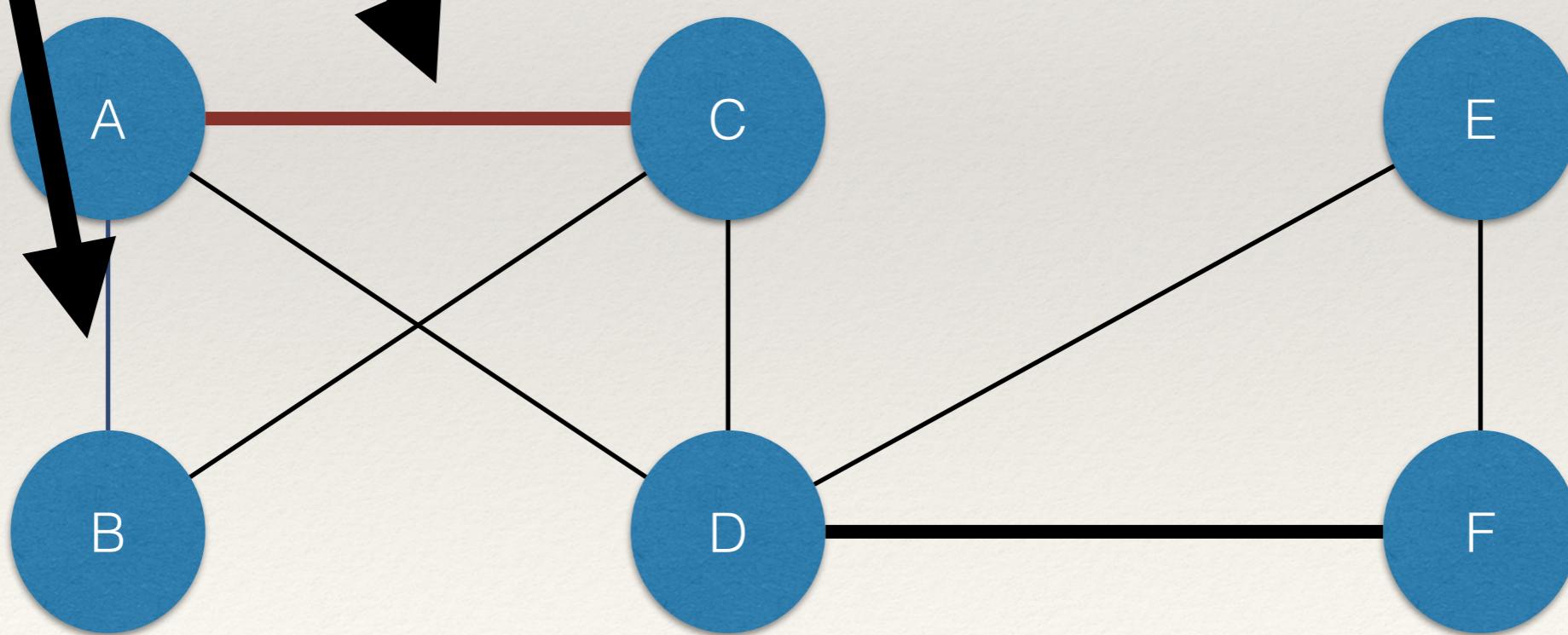
The weight corresponds to the number of nodes in the second mode that the two actors share in the bipartite graph.



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

	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

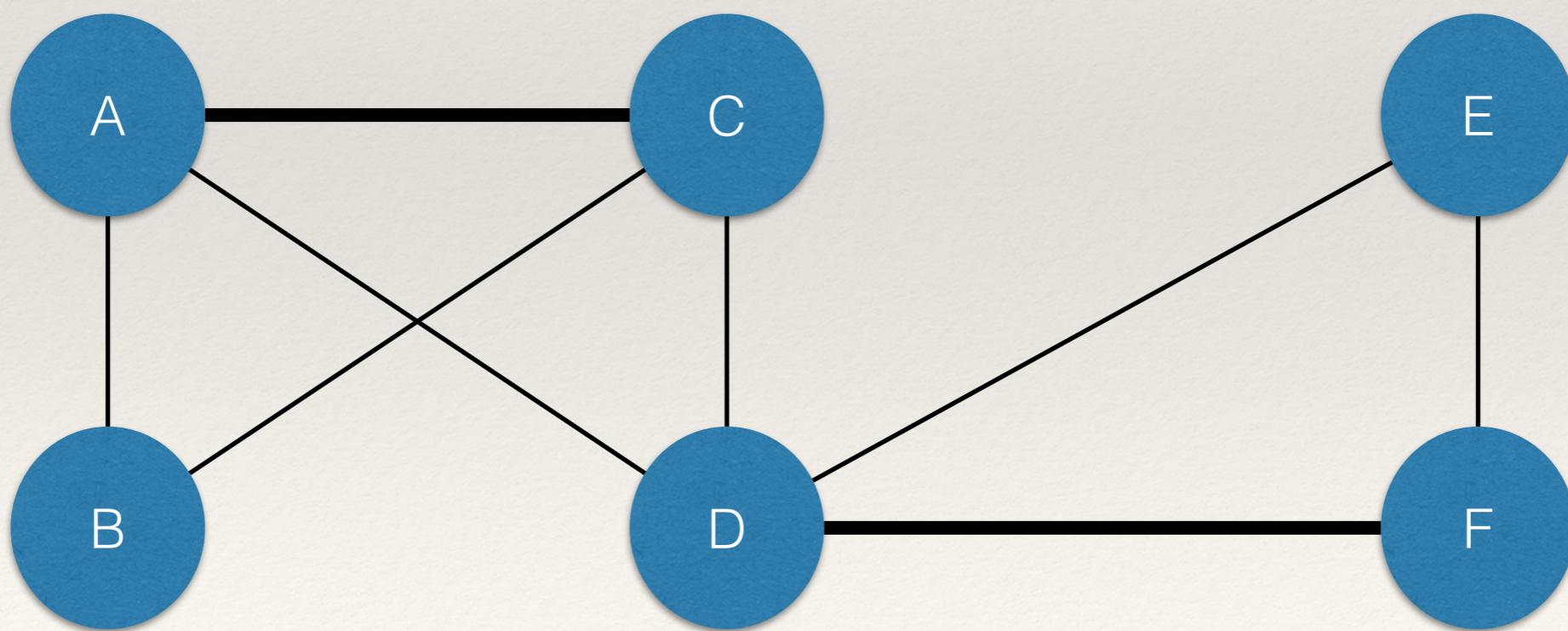
The weight corresponds to the number of nodes in the second mode that the two actors share in the bipartite graph.

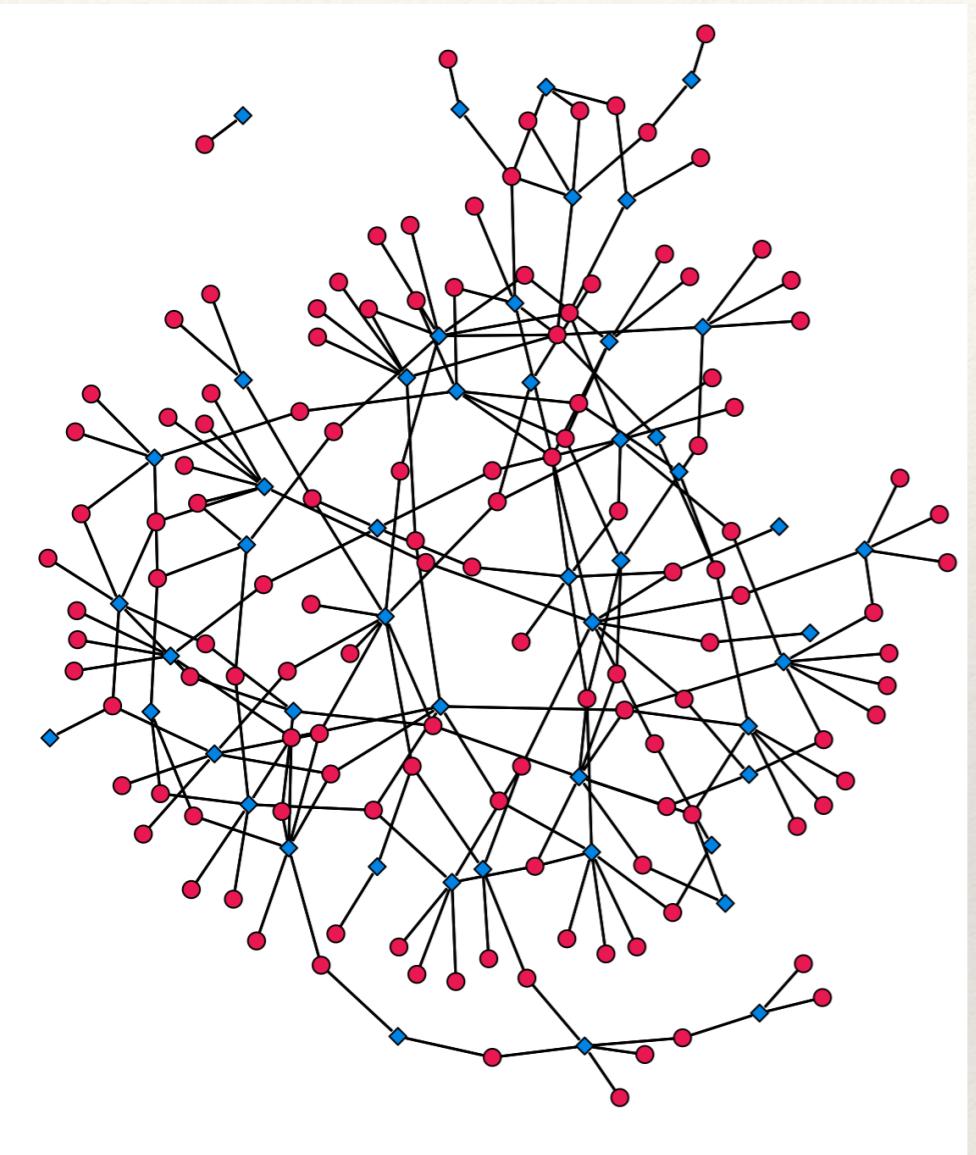


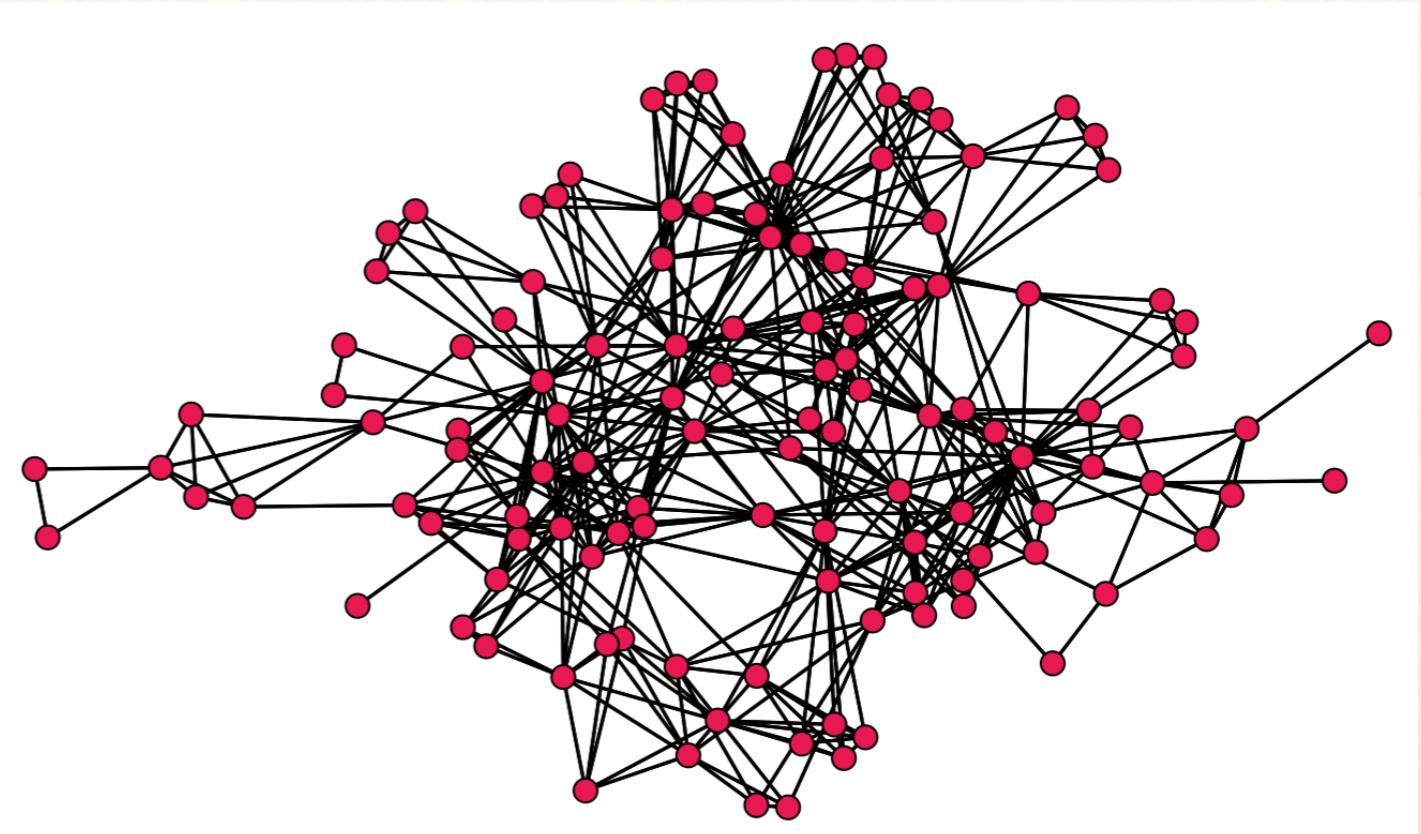
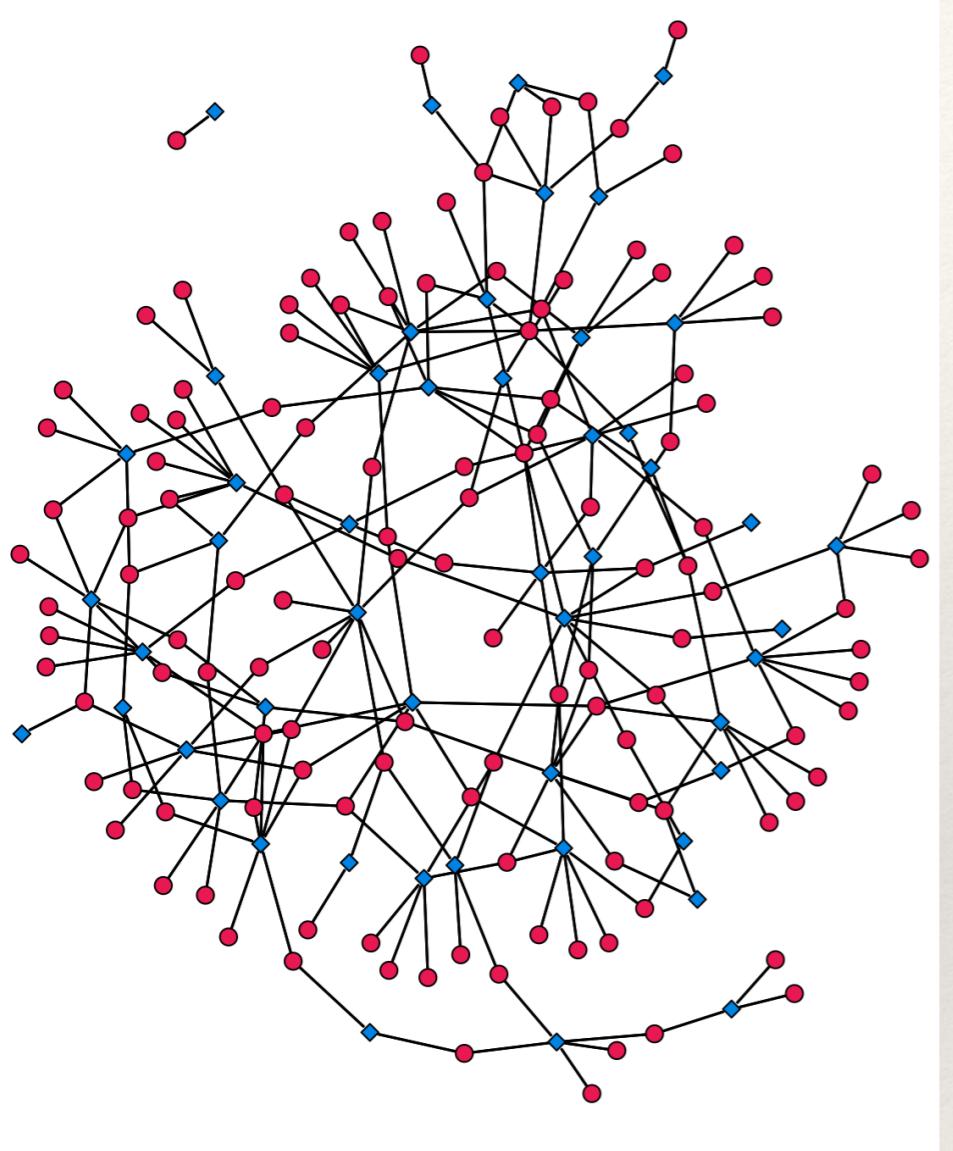
$$\mathbf{A} \times \mathbf{A}^T = \mathbf{P}$$

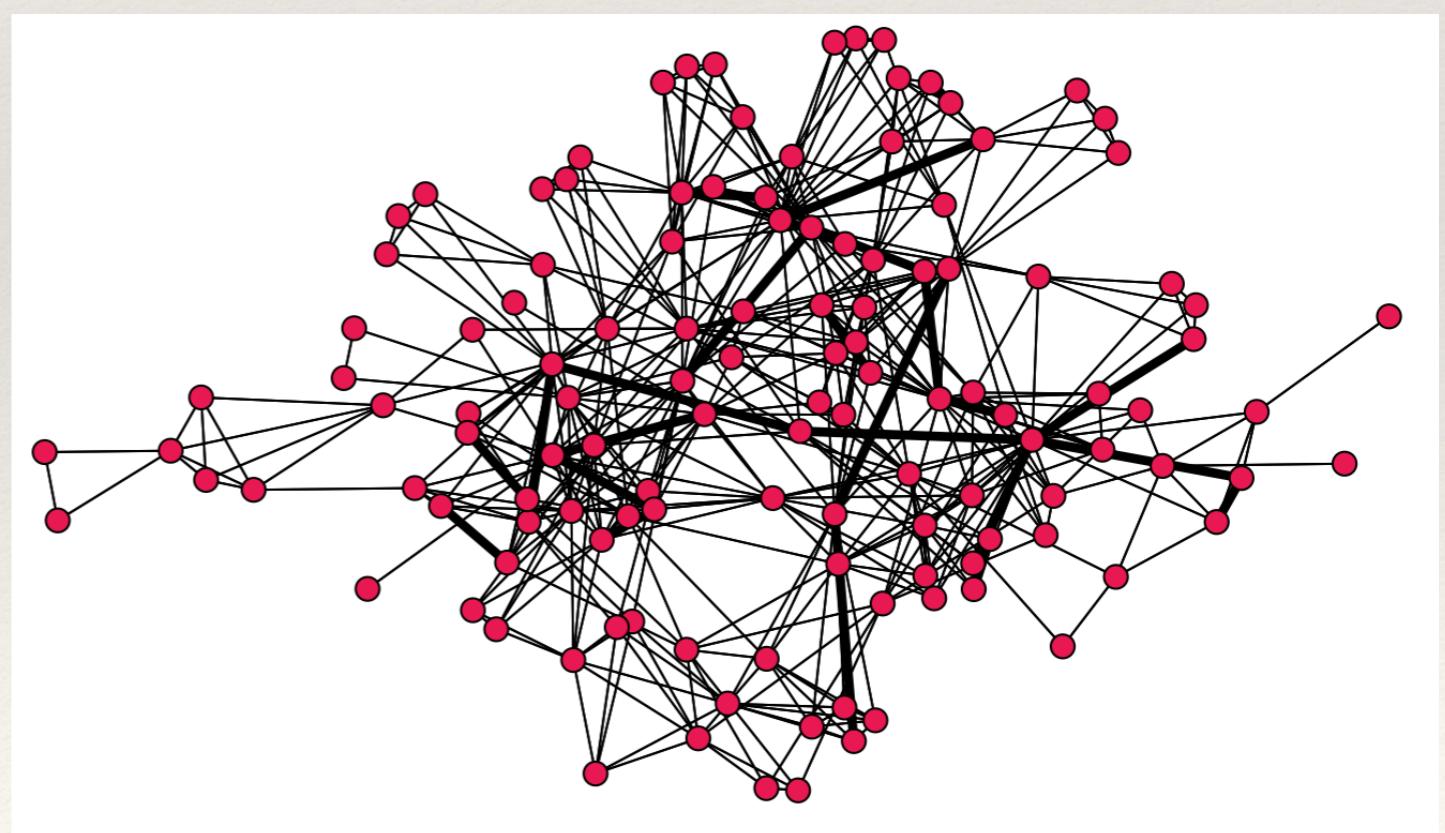
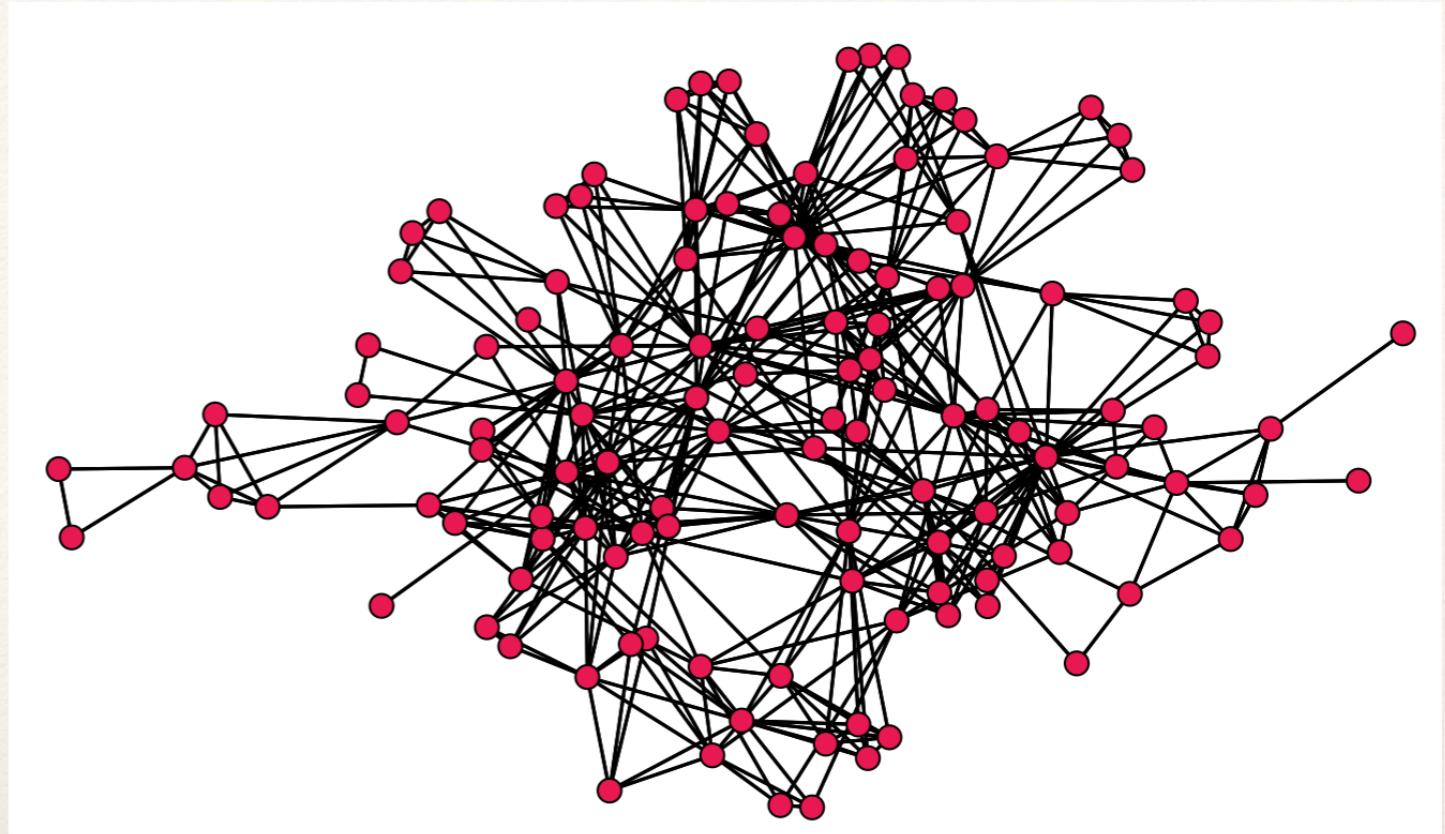
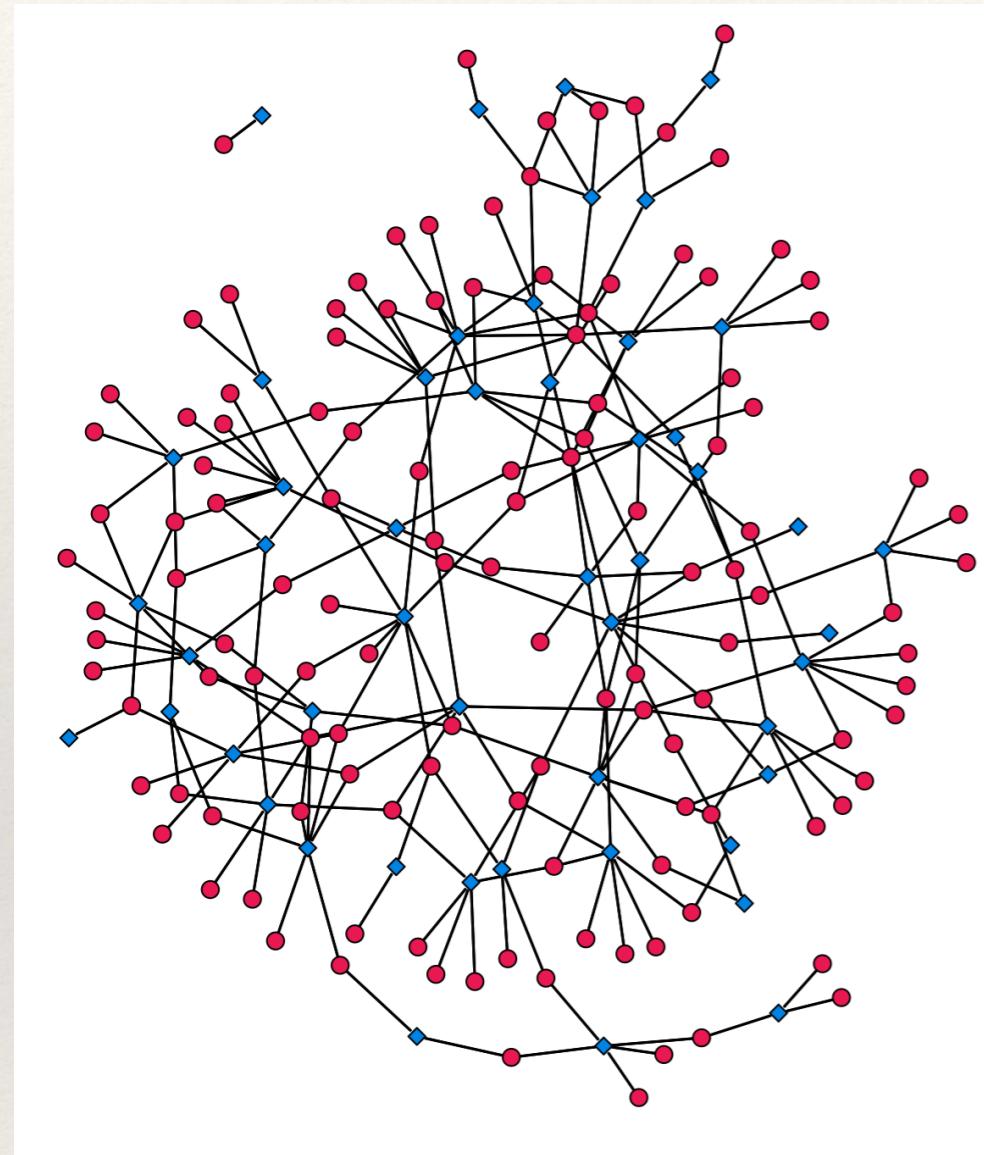
	A	B	C	D	E	F
A	2	1	2	1	0	0
B	1	1	1	0	0	0
C	2	1	2	1	0	0
D	1	0	1	4	1	2
E	0	0	0	1	1	1
F	0	0	0	2	1	2

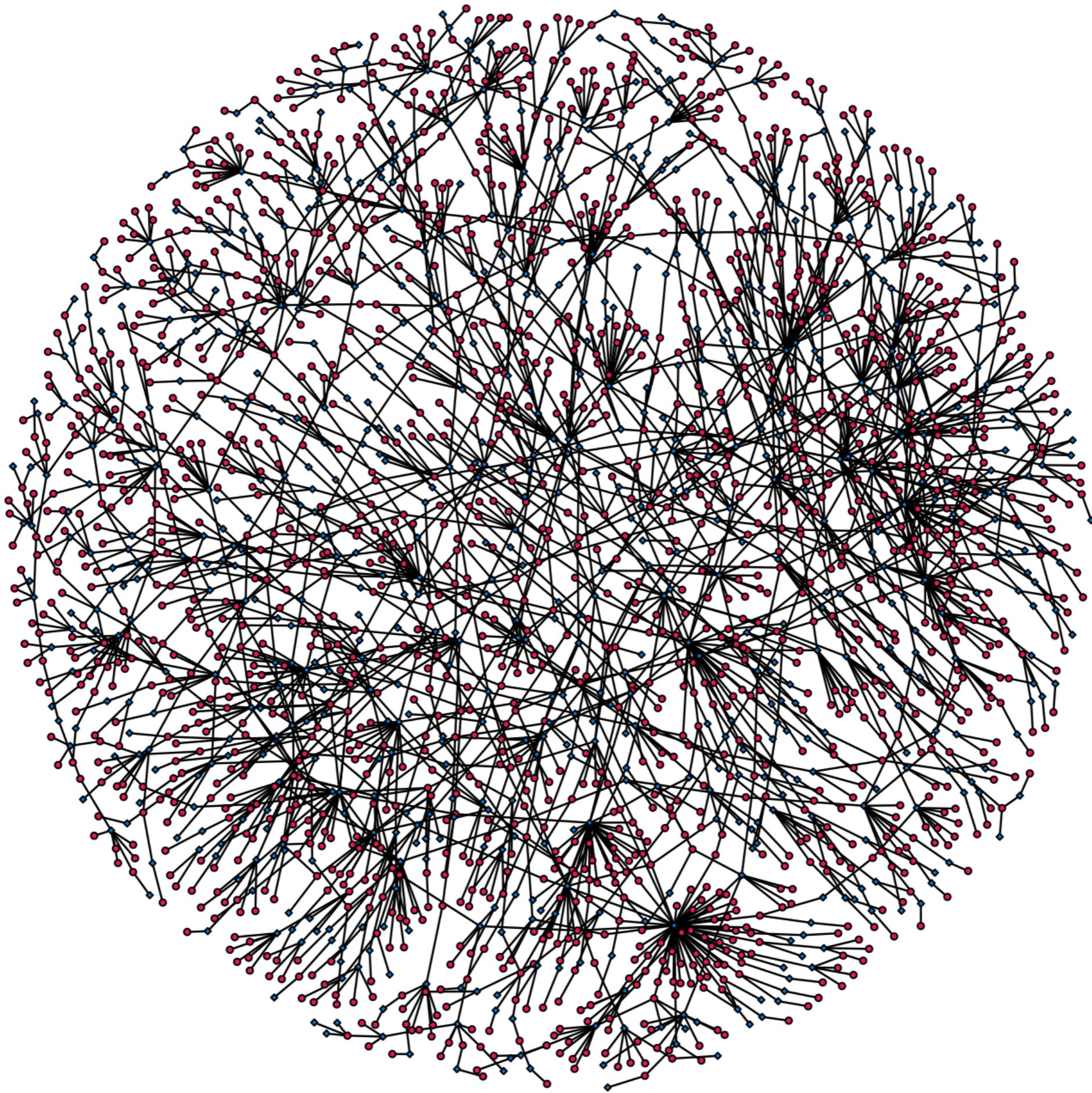
These weights are returned as the product matrix.

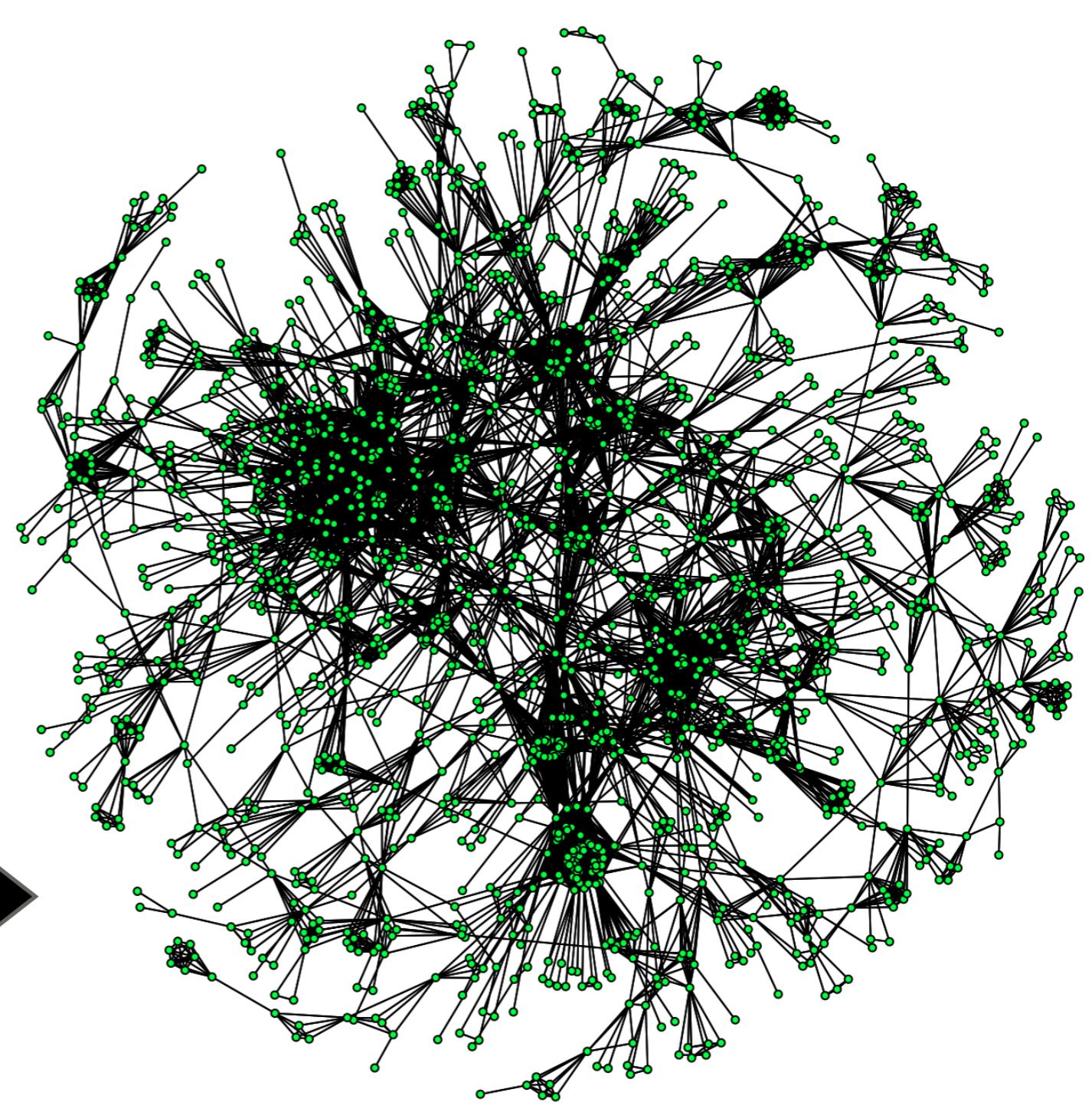
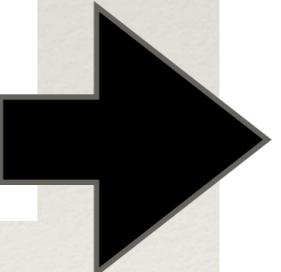
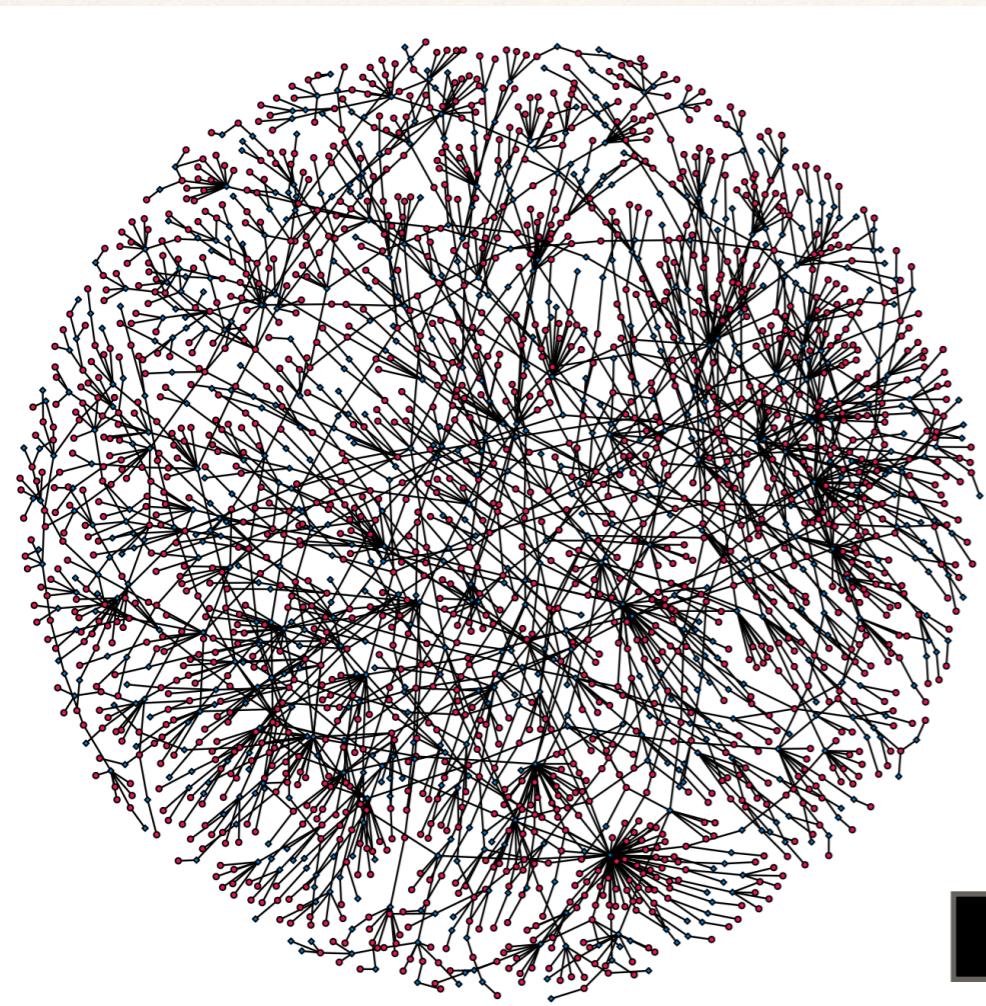


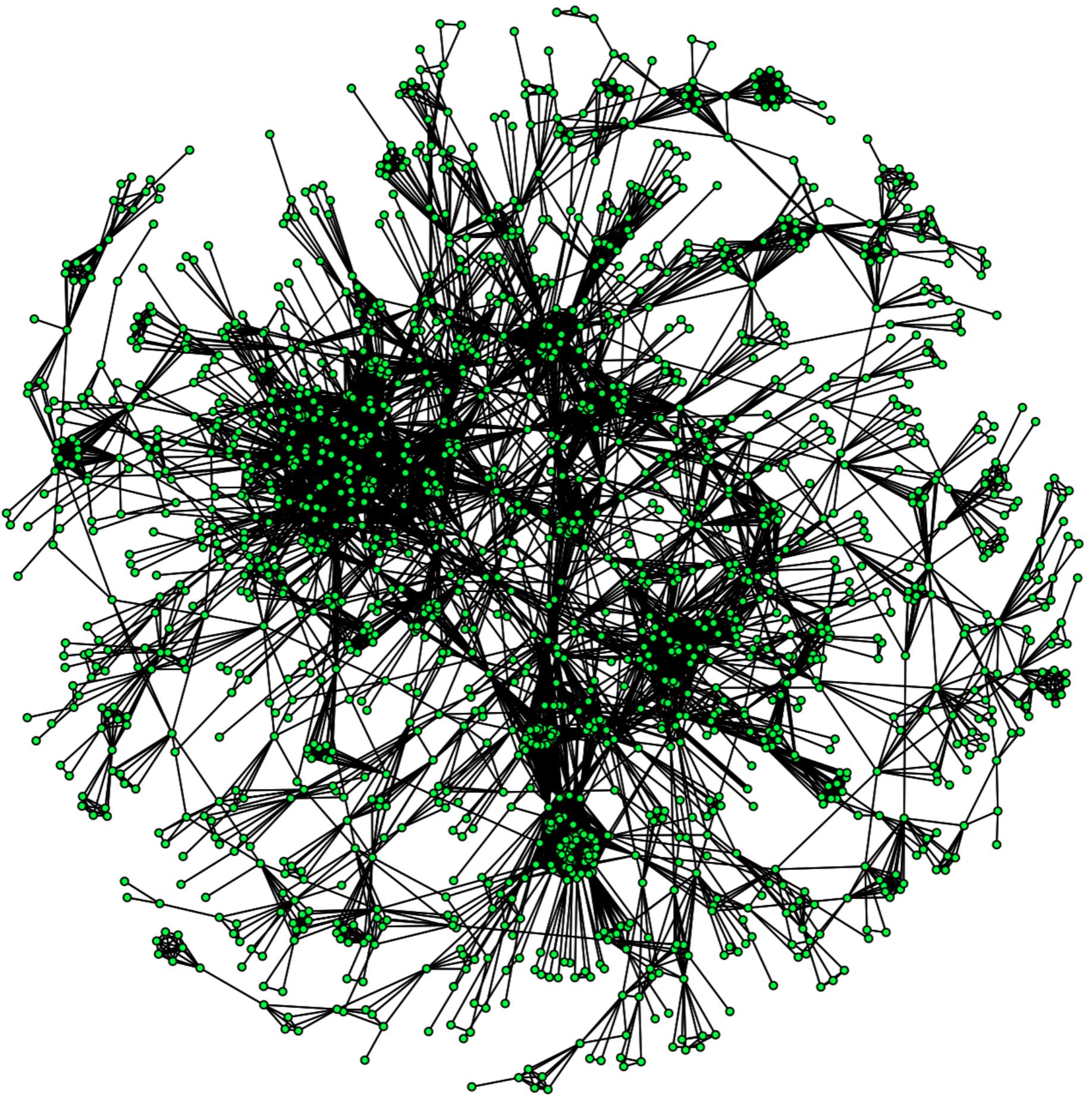












Learning Goals

- ❖ 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?