



Critical properties of Heider balance on multiplex networks

Krishnadas Mohandas, Krzysztof Suchecki, and Janusz A. Holyst



**Group of Physics in Economics
and Social Sciences**

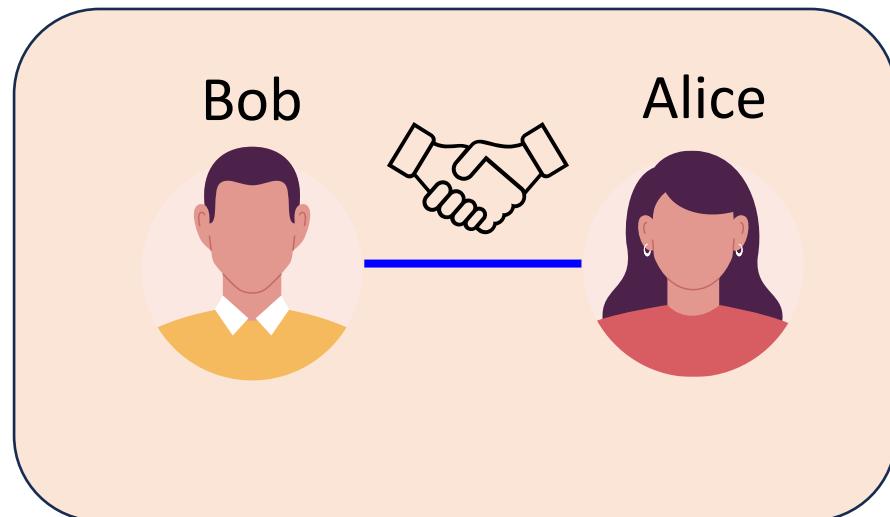


**Faculty
of Physics**

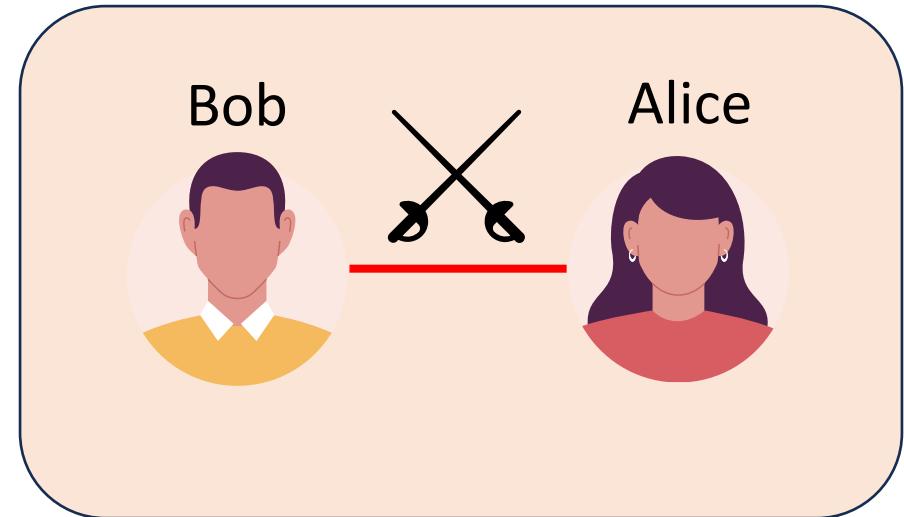
WARSAW UNIVERSITY OF TECHNOLOGY

Introduction

- Relations can be **friendly** or **Hostile**

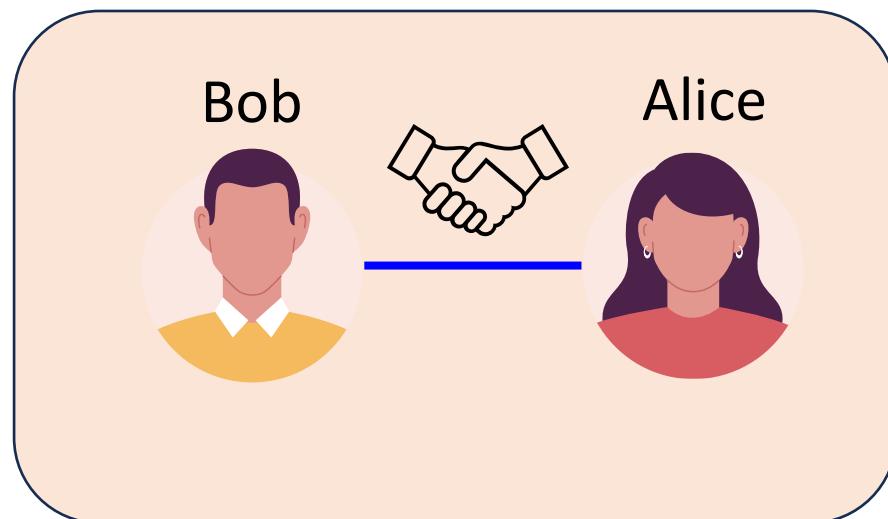


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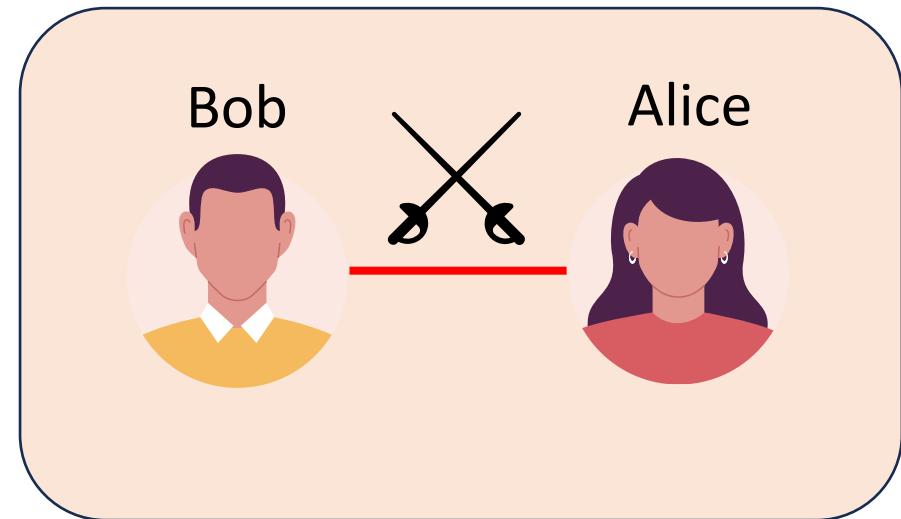


Introduction

- Relations can be **Positive** or **Negative**

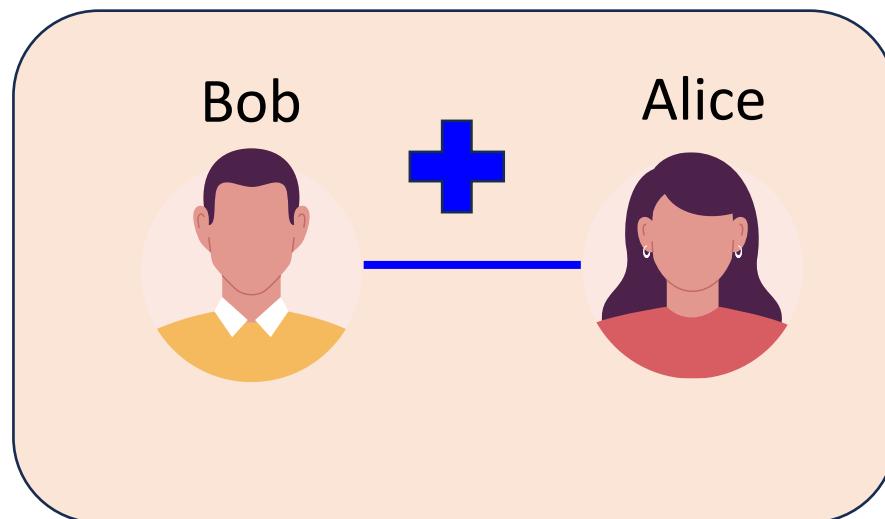


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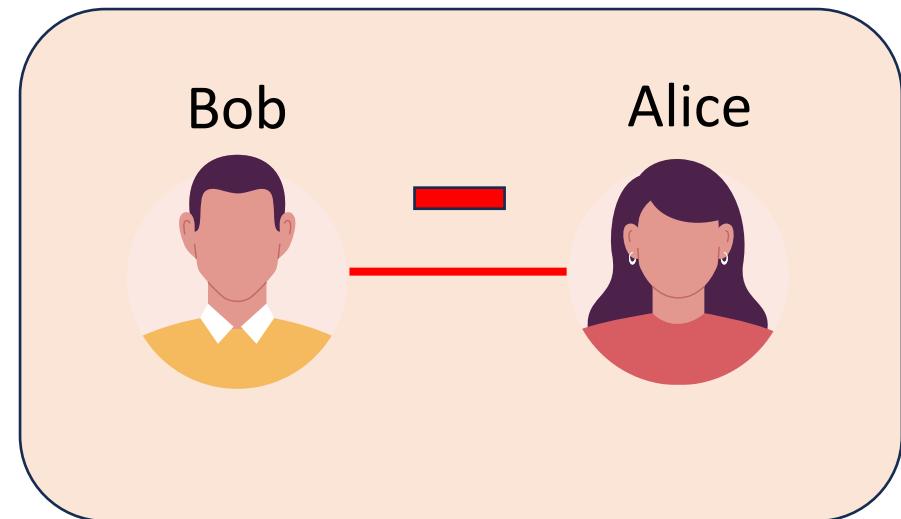


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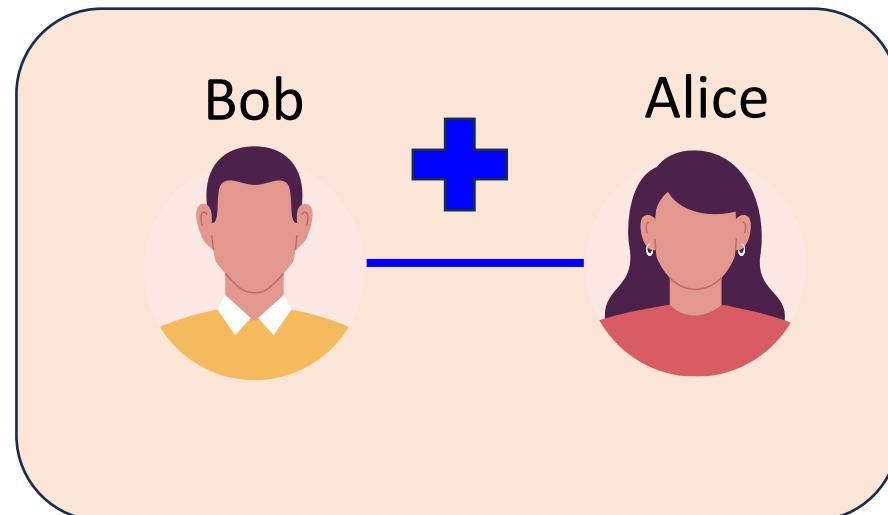
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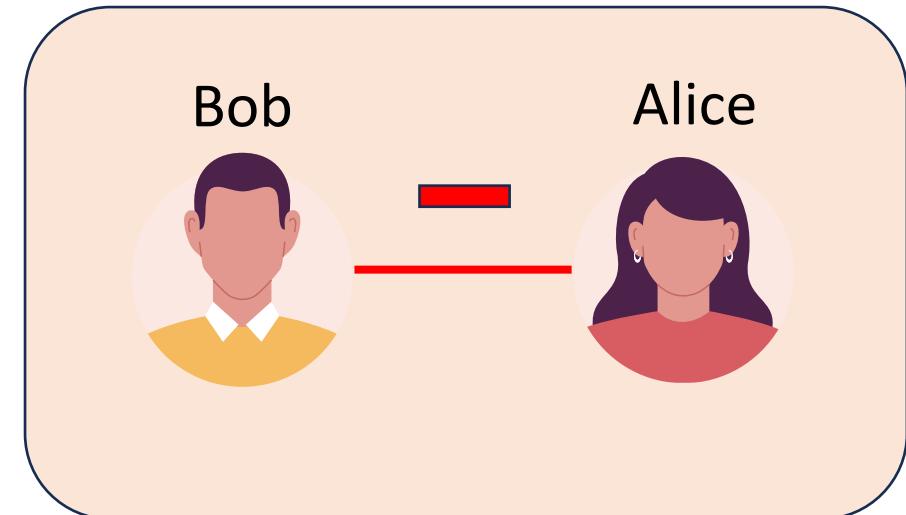
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Signed Networks

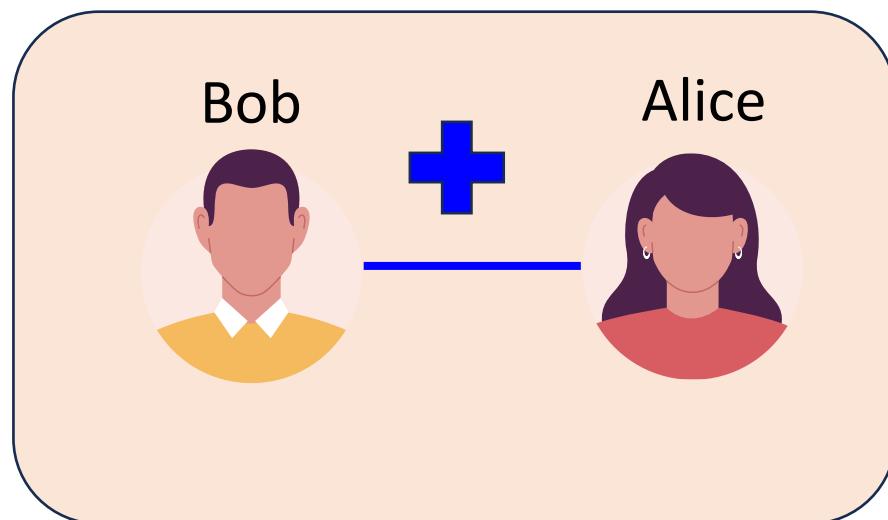


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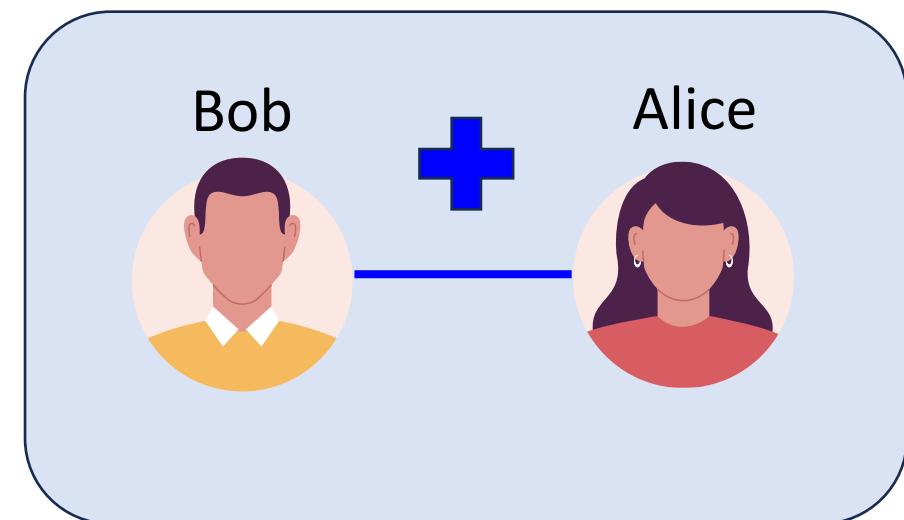


Introduction

- People usually have a consistency in their relationships.
- **Example:** Consistency in behavior at work and outside work.

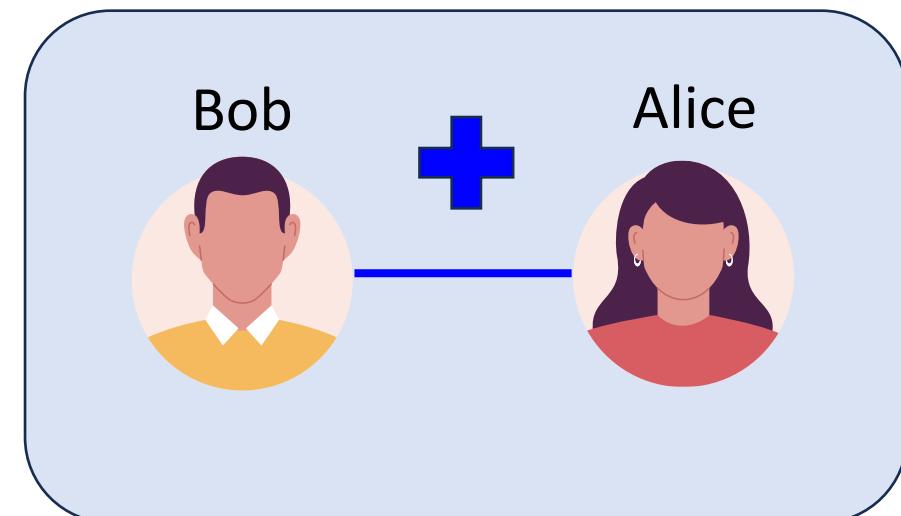
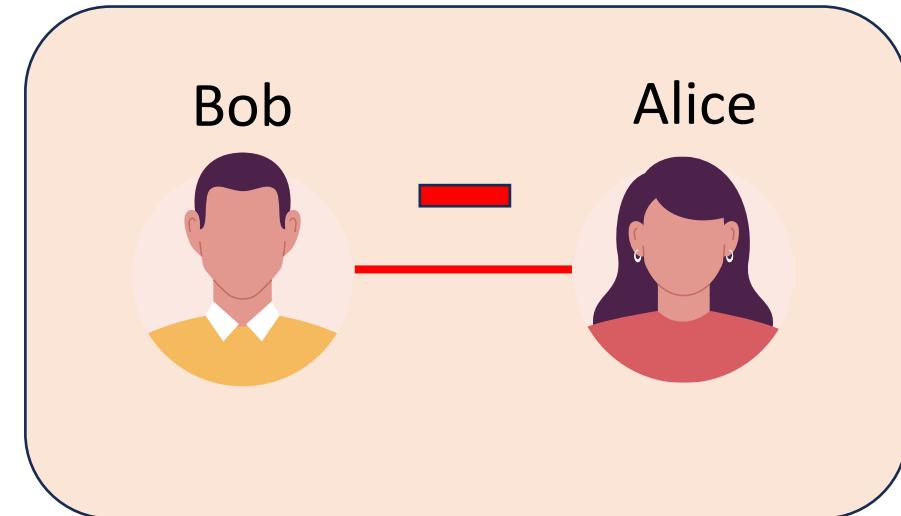


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Introduction

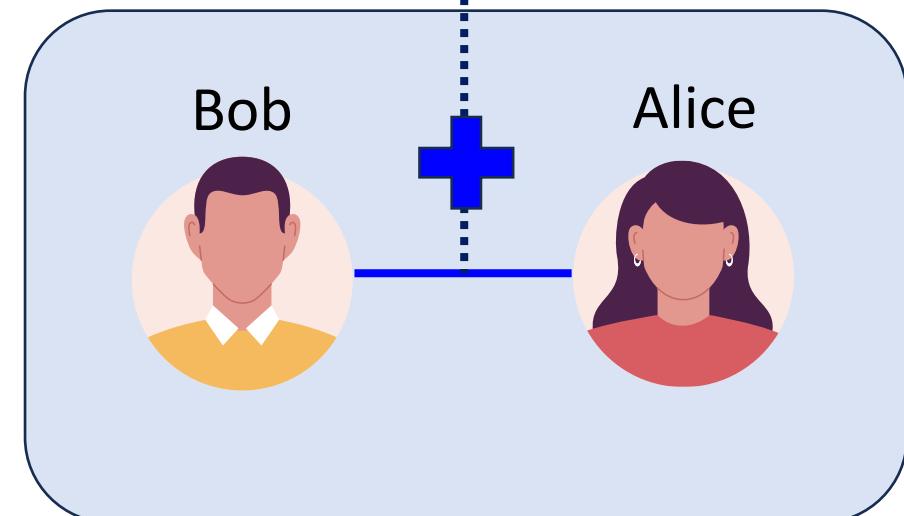
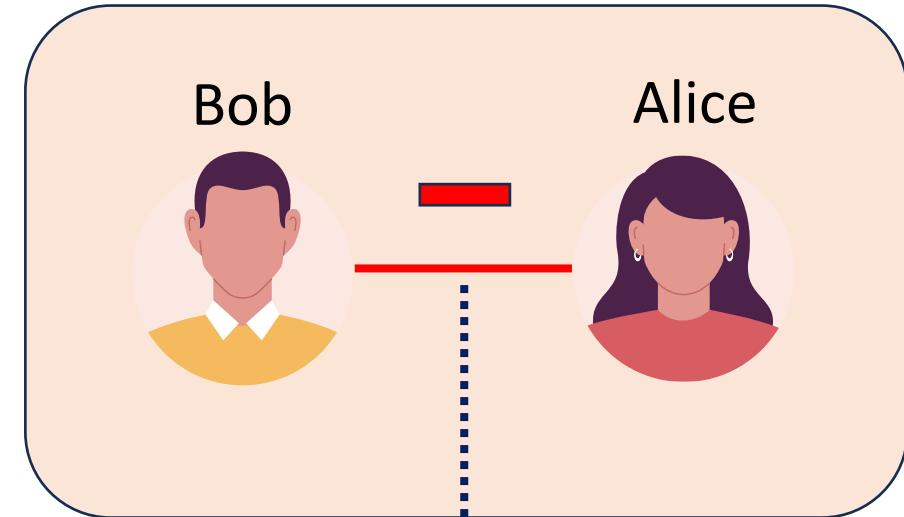
- Sports club and work relationship
- Variations in consistency lead to a cognitive dissonance



Introduction

- Officially friends and privately enemies?
- Variations in consistency lead to a cognitive dissonance

Interactions between different relations linking the same pair of individuals

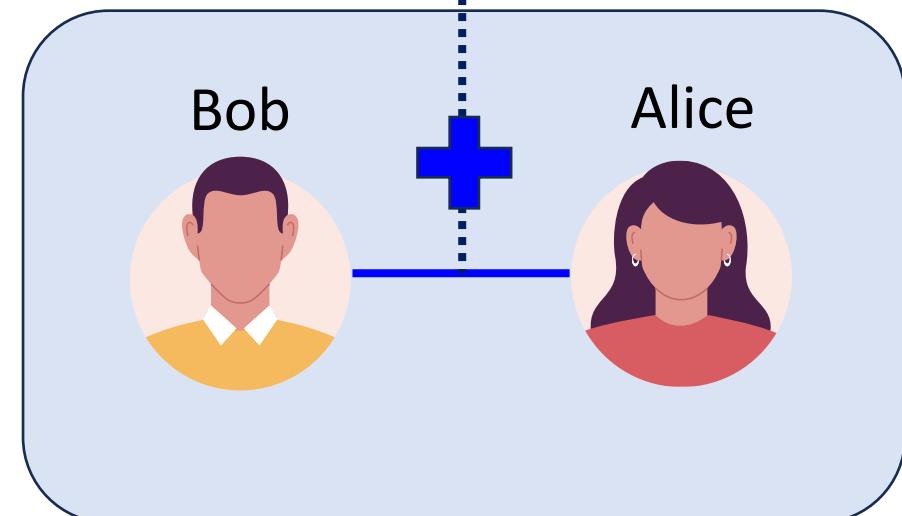
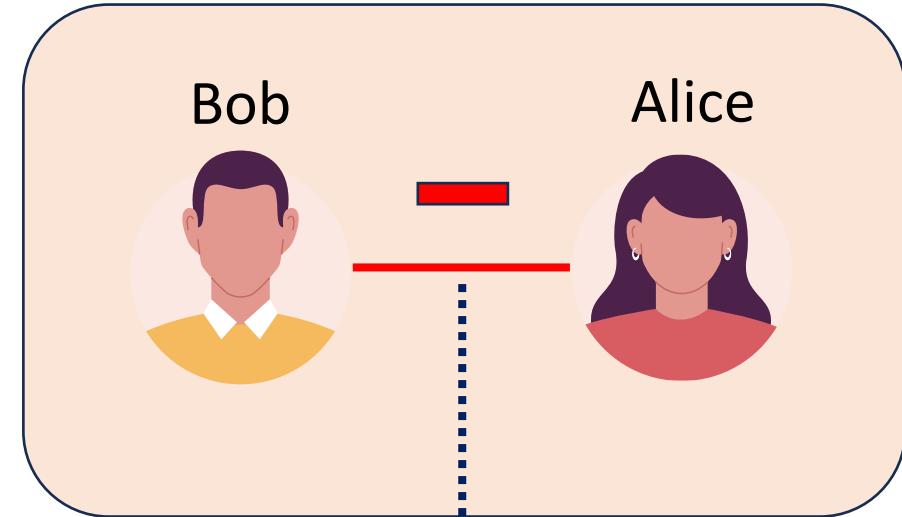


Introduction

- Officially friends and privately enemies?
- Variations in consistency lead to a cognitive dissonance

Interactions between different relations linking the same pair of individuals

Interactions is for the relations and not between the people



Outline

Social science Lens

Studying different relationships to understand their impact on social balance.

What happens to the relationship layers

Two layers in different relations.

Missing connection

Heider Balance to multiplex networks

Simulations

Duplex network with contrasting relationship layers.

HB with missing links

Physics Lens

How multiplex network effect the Heider balance

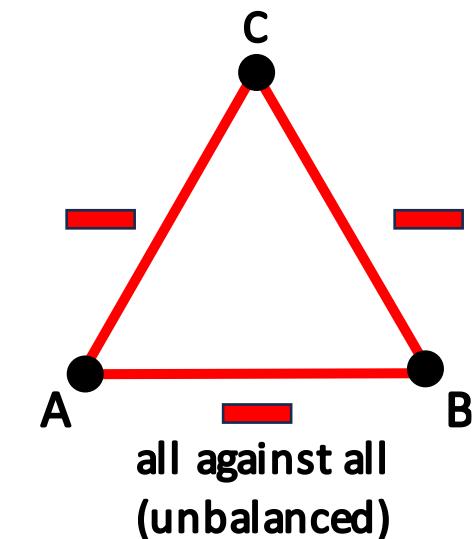
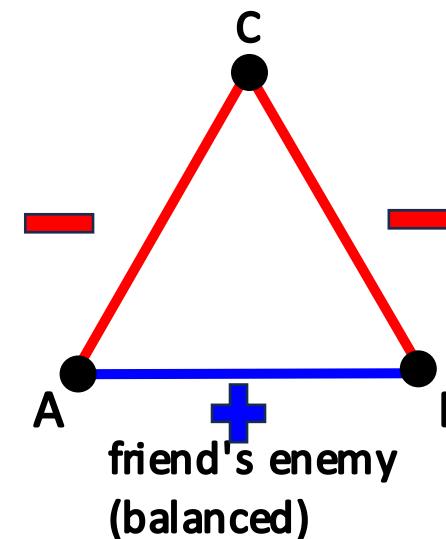
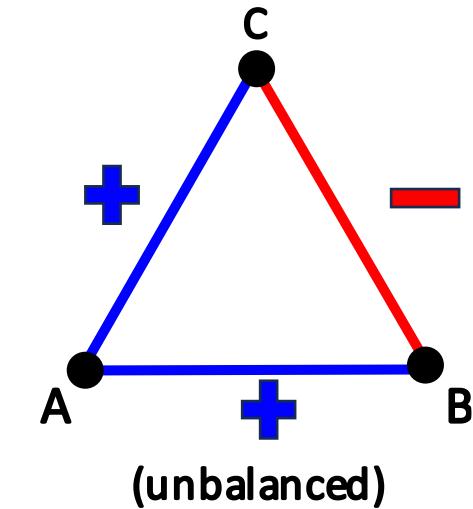
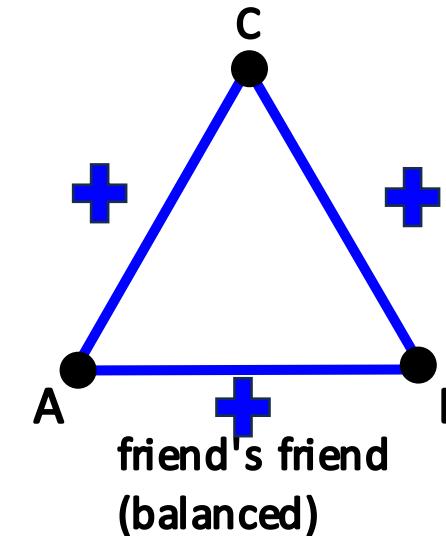
Monte Carlo simulation assesses how changes in one layer affect network balance.

A duplex network in different states

Erdős Rényi graph

Heider Balance theory

- Friend of my friend is my friend
- Enemy of my enemy is my friend
- Friend of my enemy is my enemy
- Enemy of my friend is my enemy



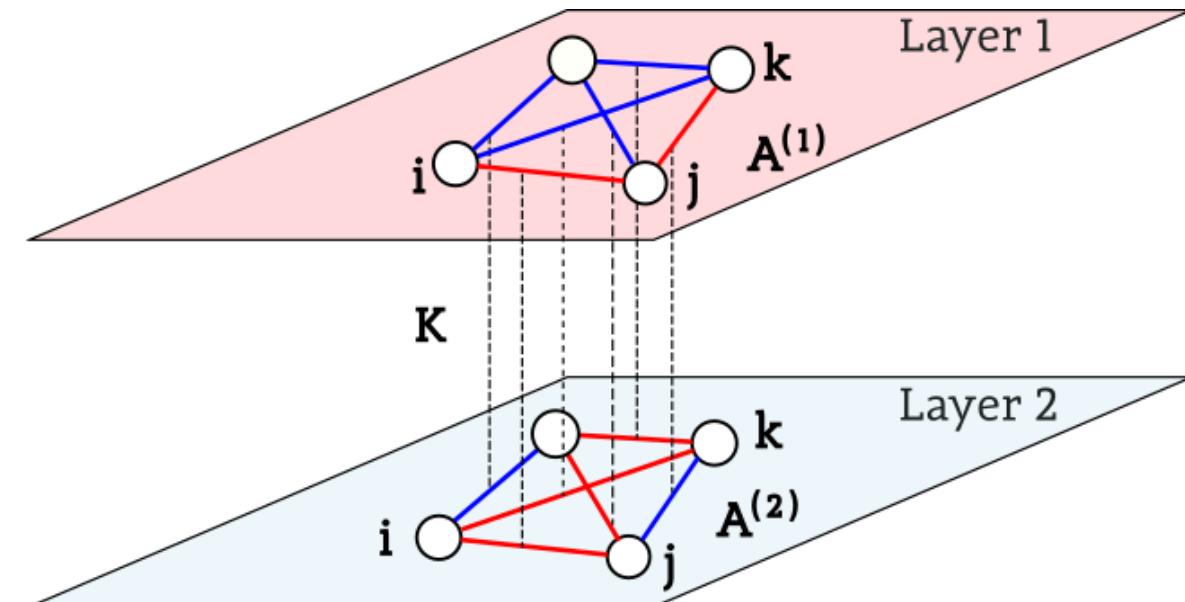
Model



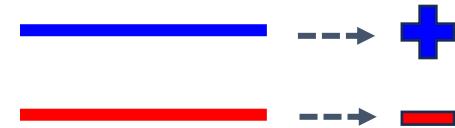
- Assumptions:**
- $x_{ij}^{(\alpha)} = \pm 1$
 - Each layer corresponds to a different type of relationship
 - Interlayer interactions exist between different relations for the same pair of agents

Hamiltonian (Utility Potential):

$$H = - \sum_{\alpha=1}^L A^{(\alpha)} \sum_{i>j>k} x_{ij}^{(\alpha)} x_{jk}^{(\alpha)} x_{ki}^{(\alpha)} - K \sum_{\alpha>\beta} \sum_{i>j} x_{ij}^{(\alpha)} x_{ij}^{(\beta)}$$



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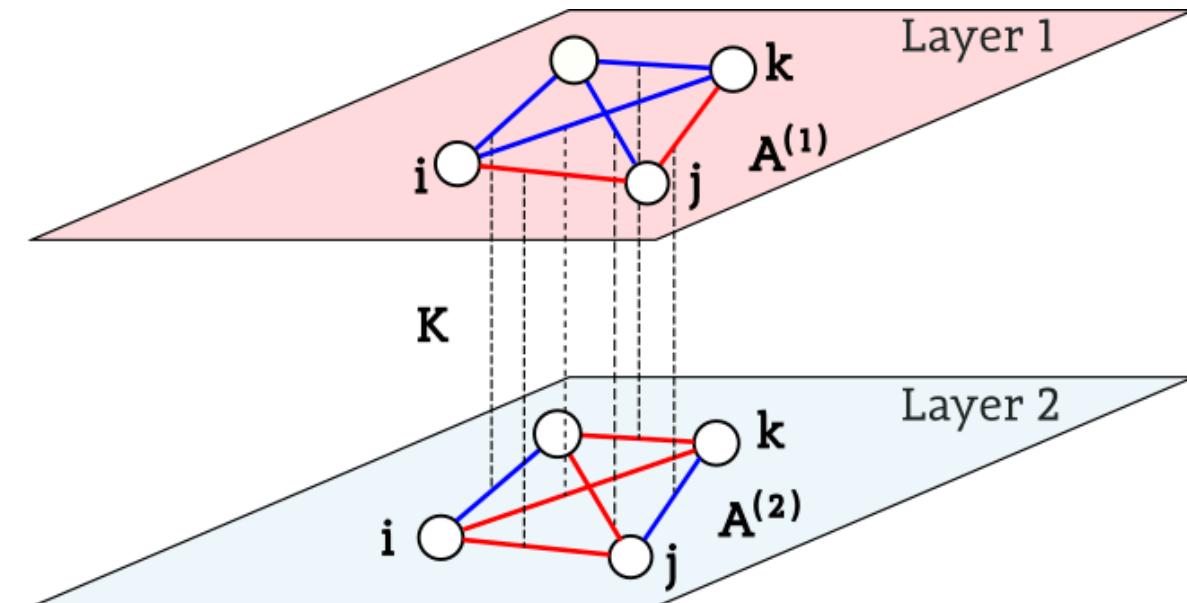


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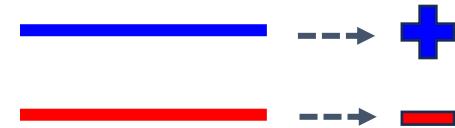
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Heider interactions
(in the layer)



Model



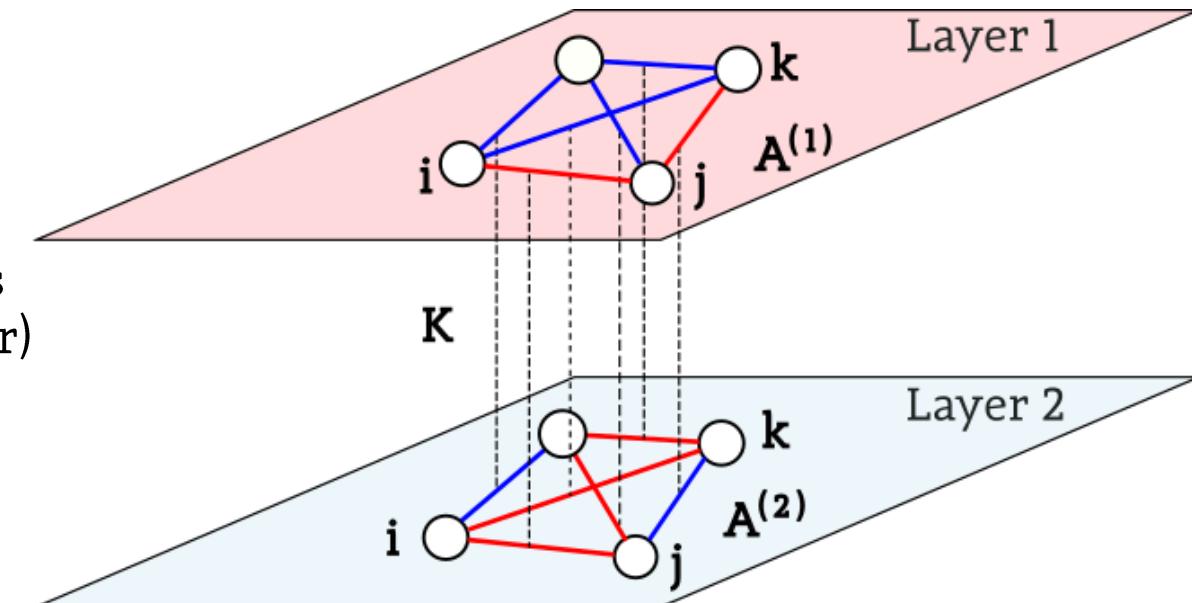
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(in the layer)

Ising Interactions
(between the layer)



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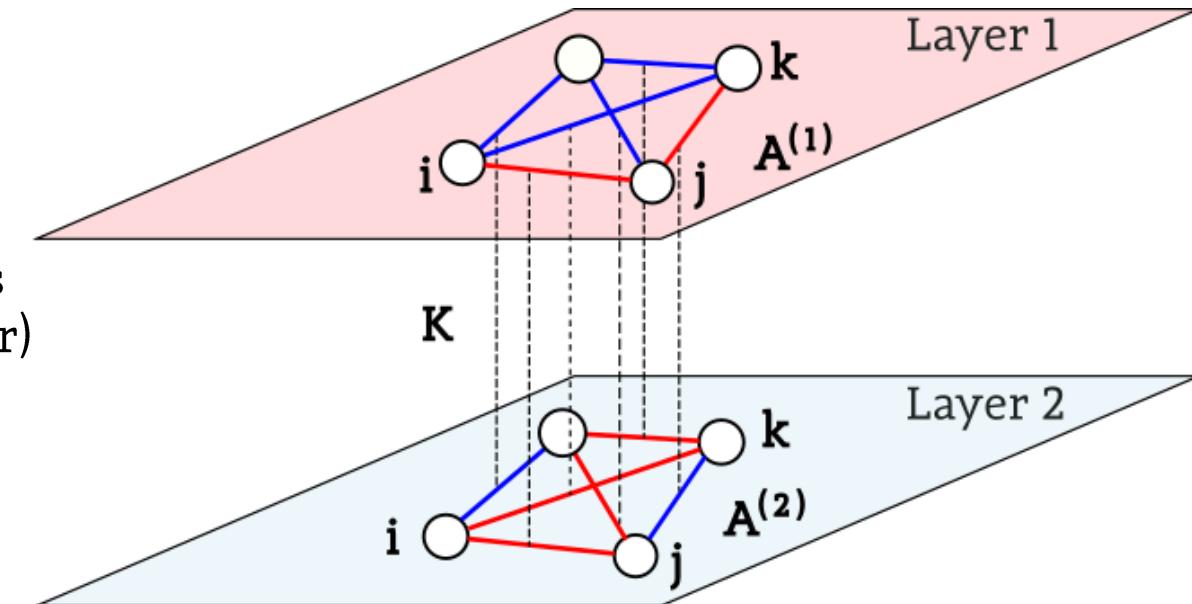
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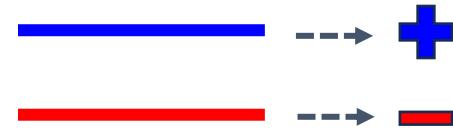
Heider interactions
(in the layer)

Ising Interactions
(between the layer)

$\alpha = 1, 2 \dots, L$ indicates the layer
 A = Heider interaction strength
 K = Ising Interaction strength



Model



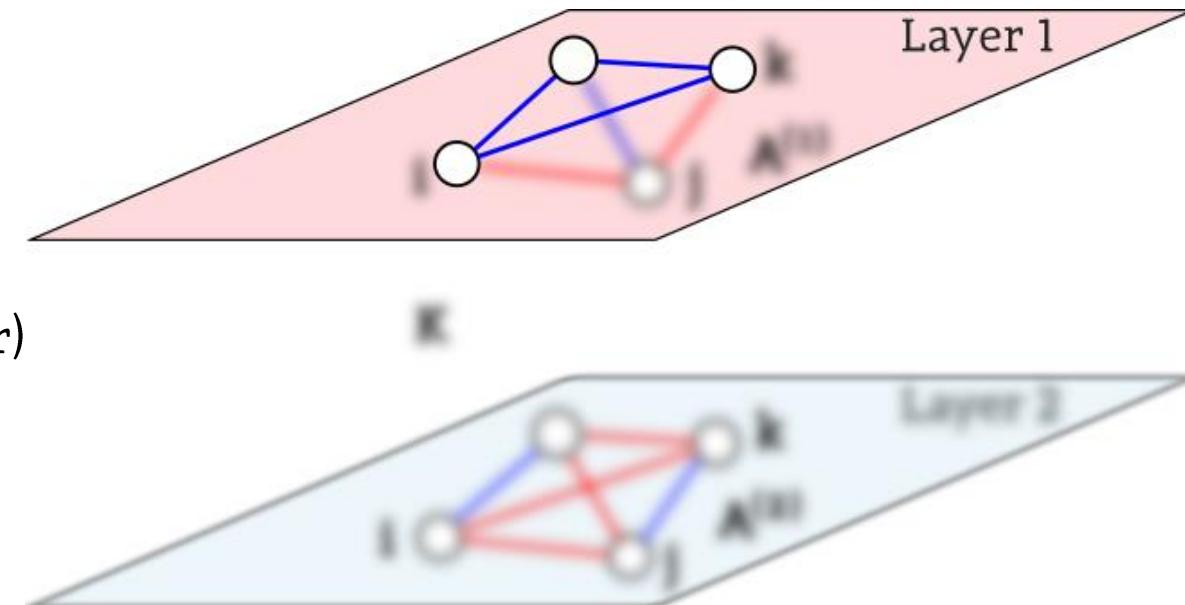
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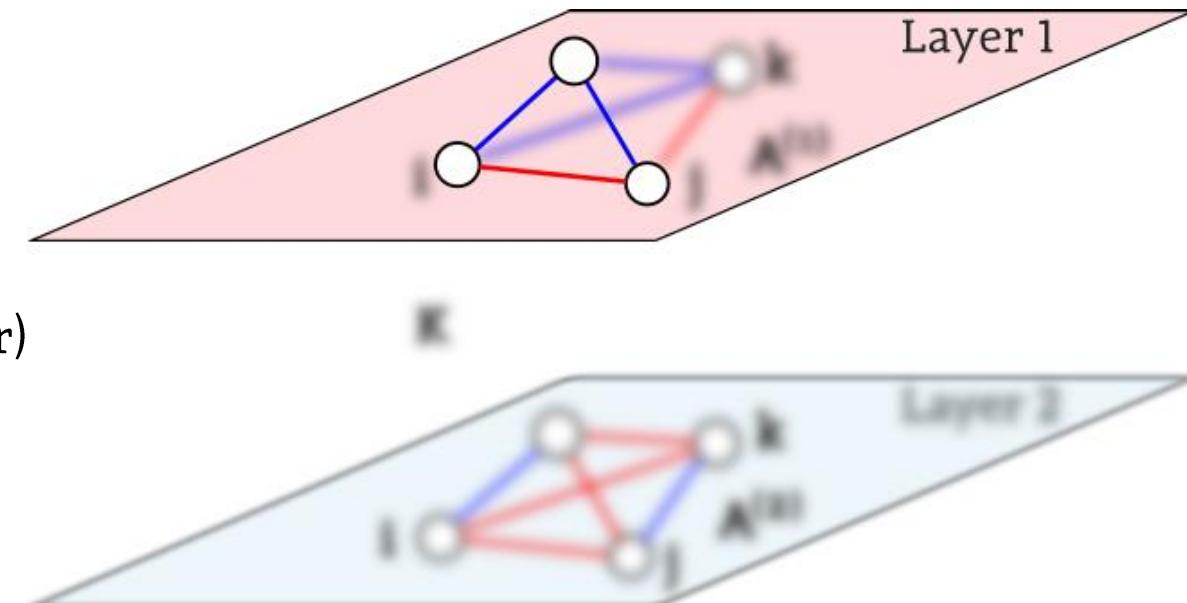
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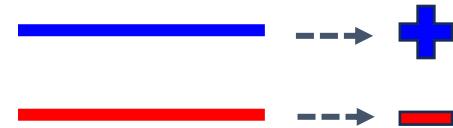
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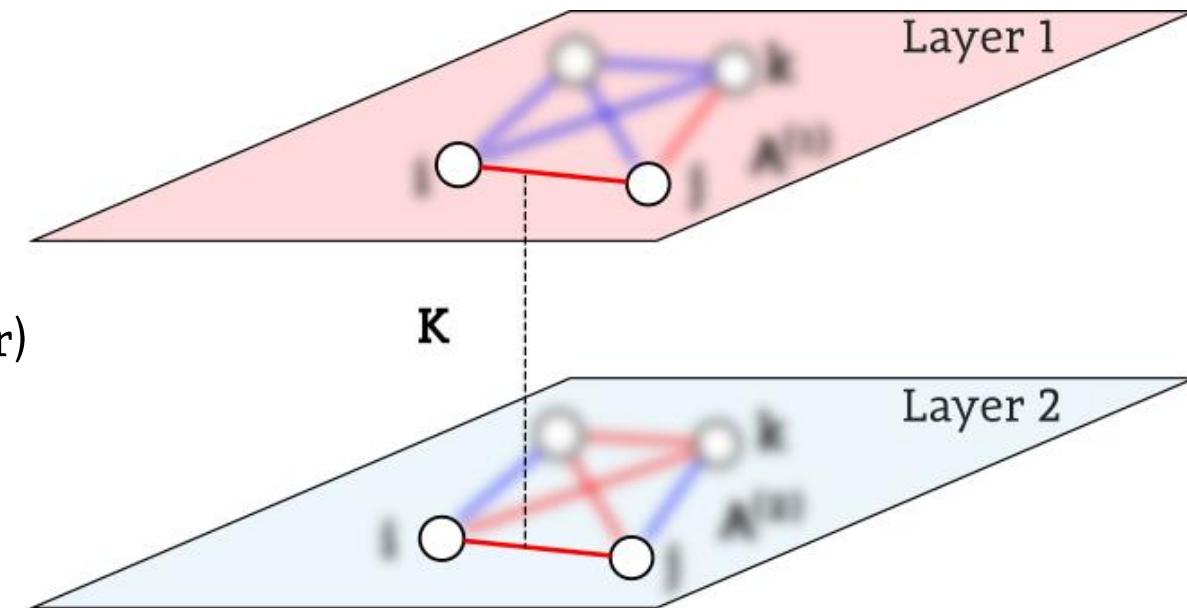
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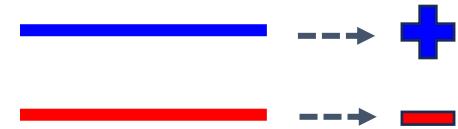
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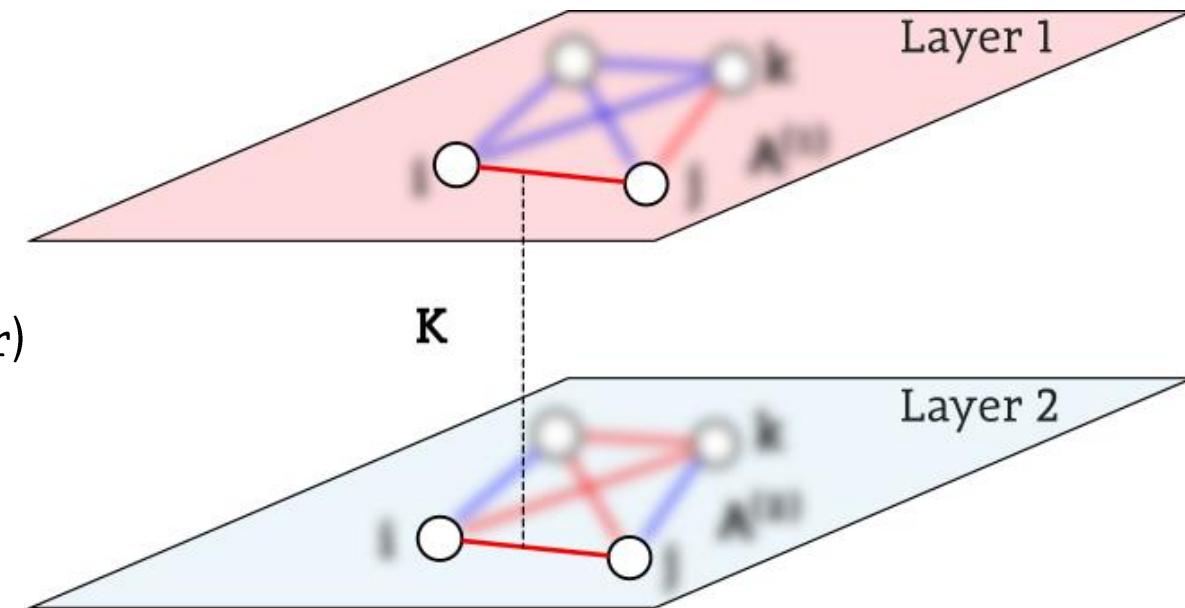
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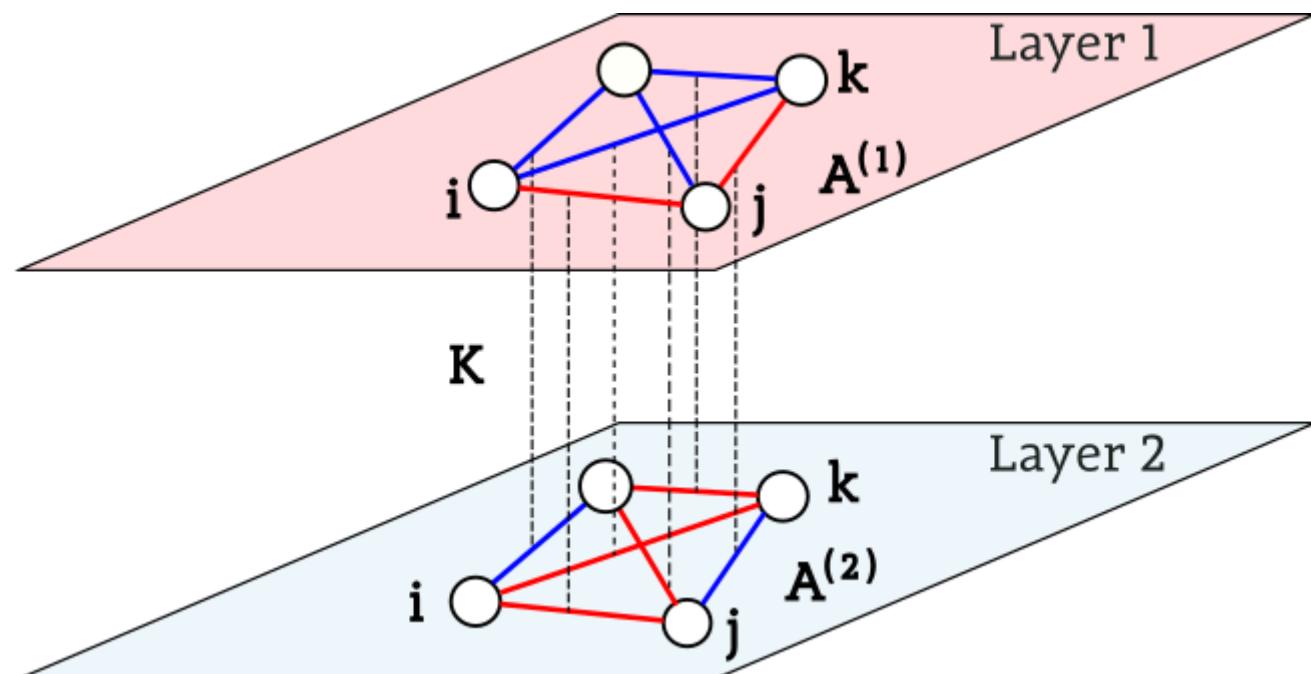
Complete graph !



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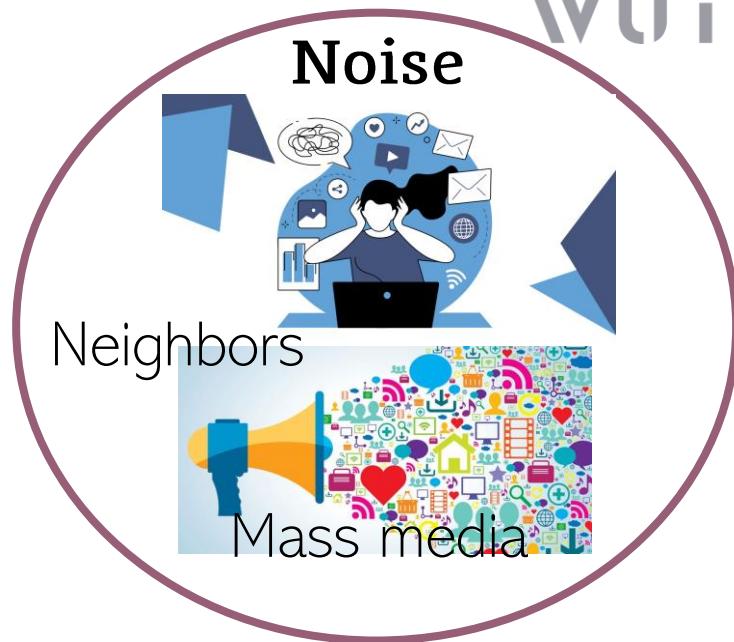
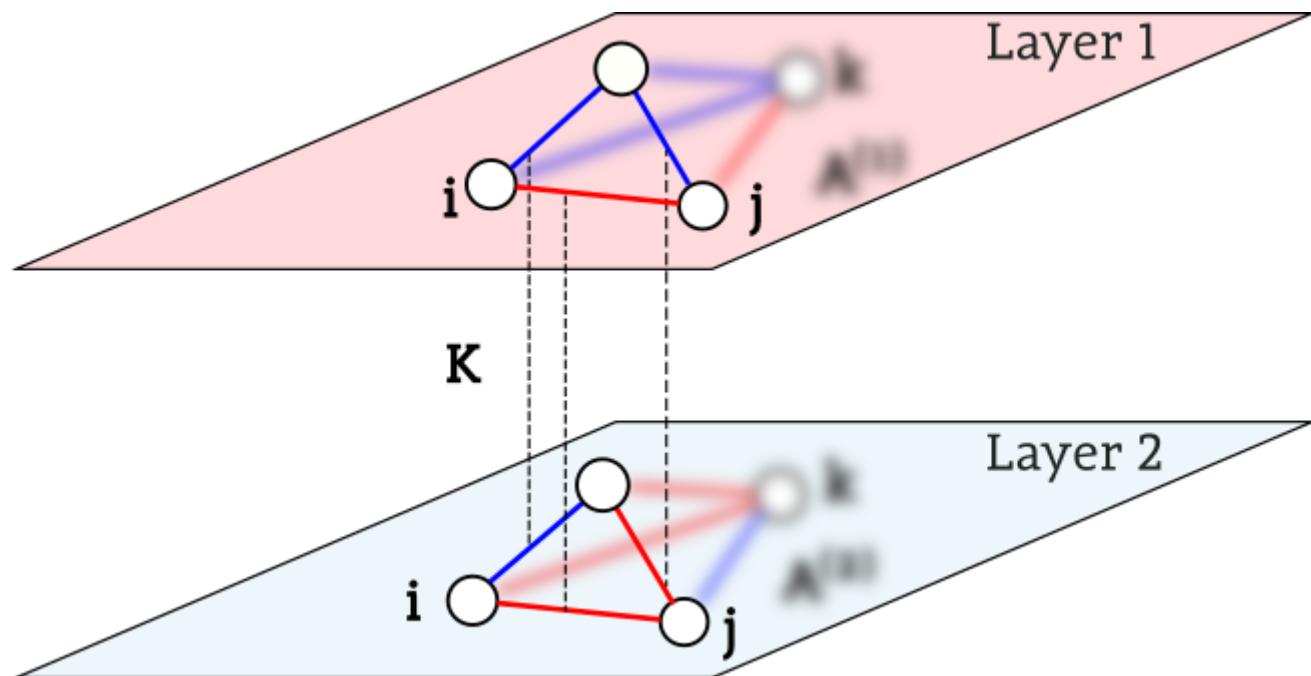
How the connection evolve

What happens to the links when the system evolves?



How the connection evolve

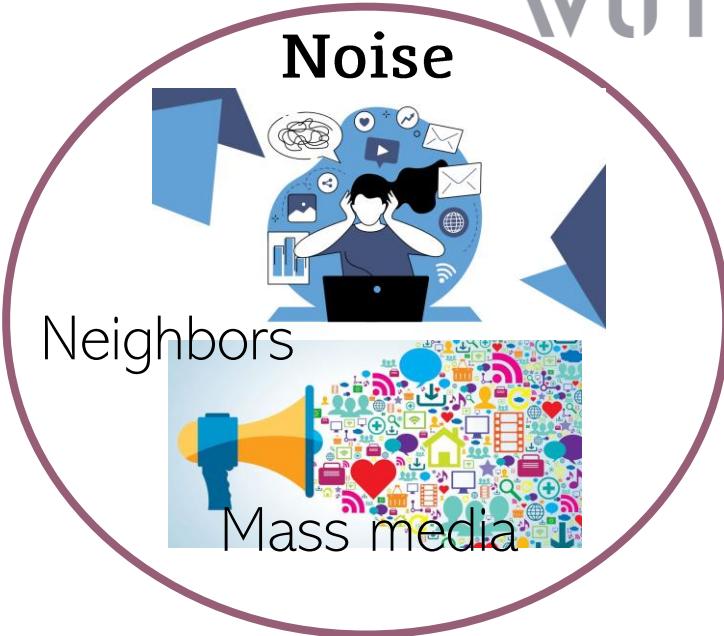
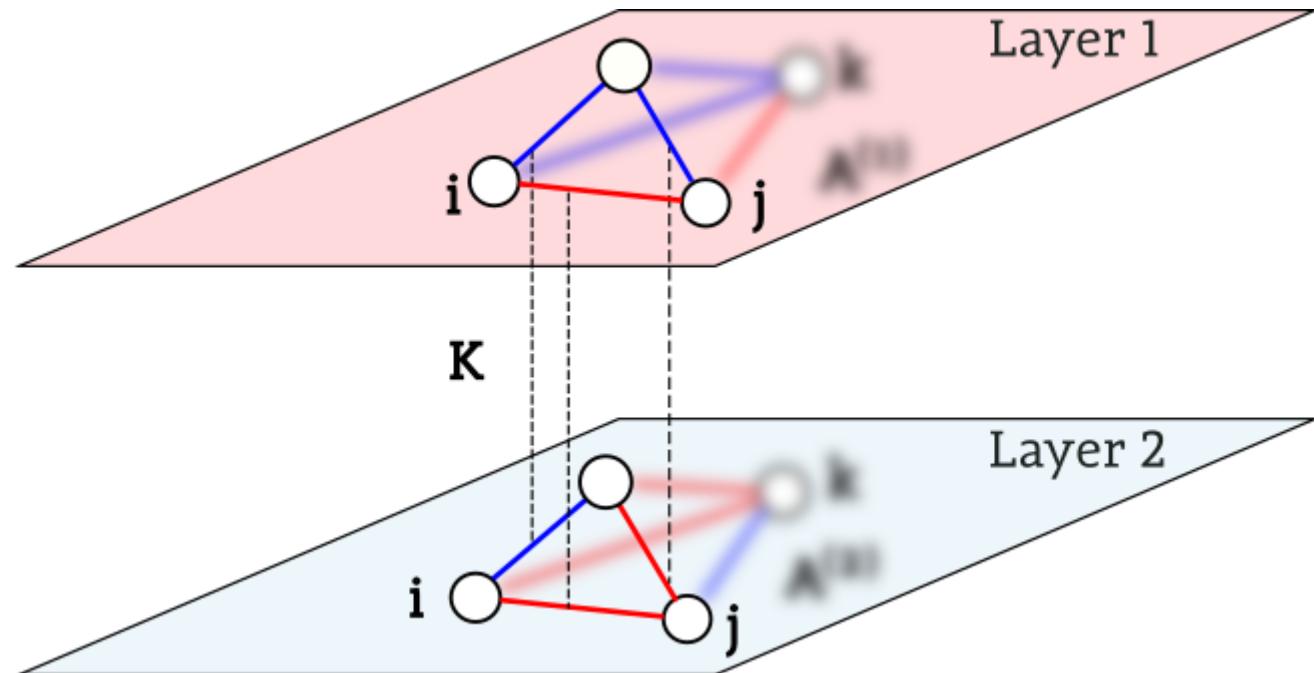
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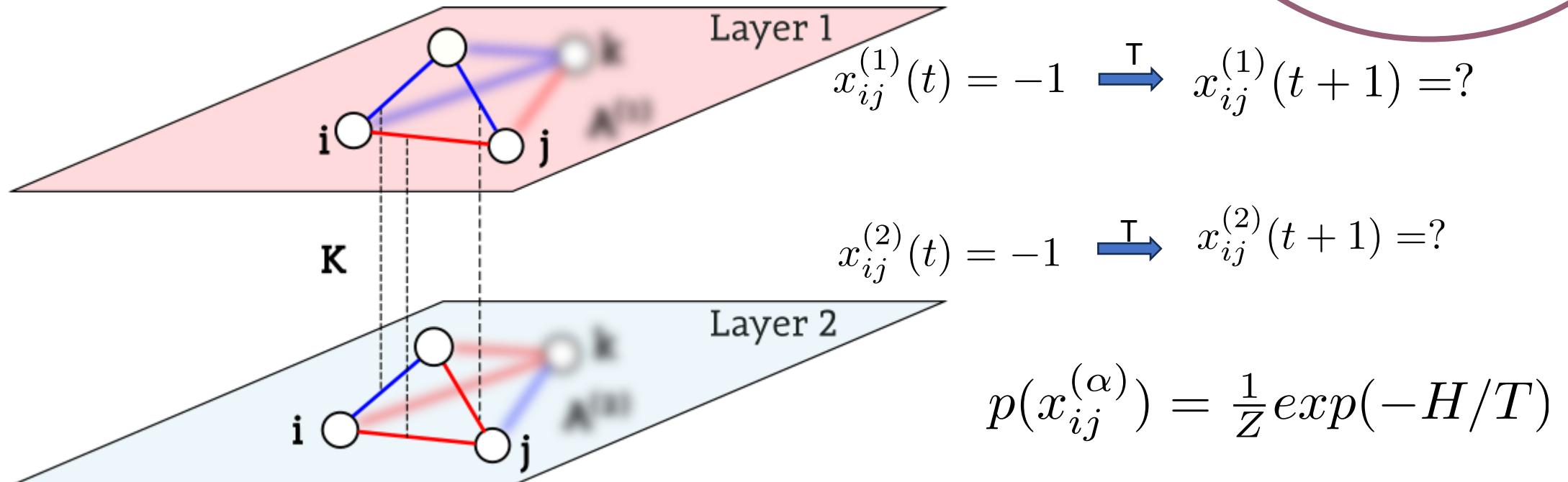
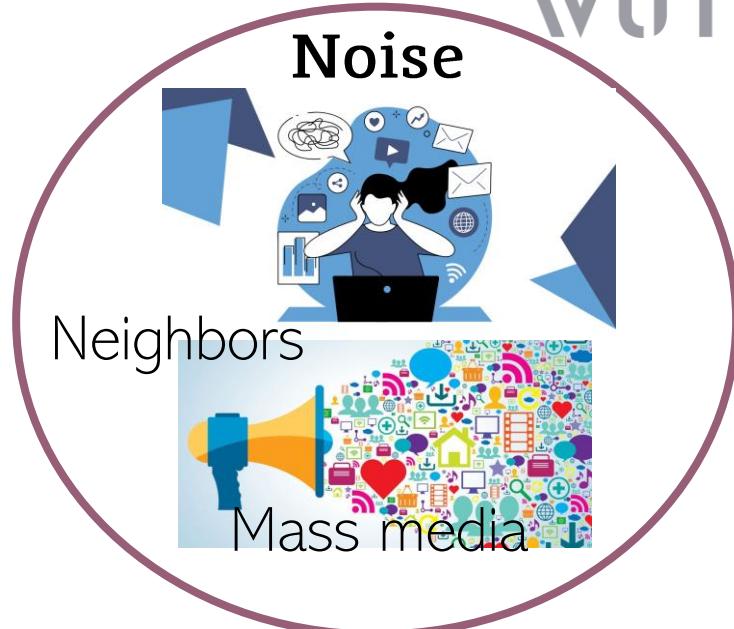
Temperature



How the connection evolve

What happens to the links when the system evolves?

Temperature



Analytical Approach:

MEAN FIELD APPROXIMATION

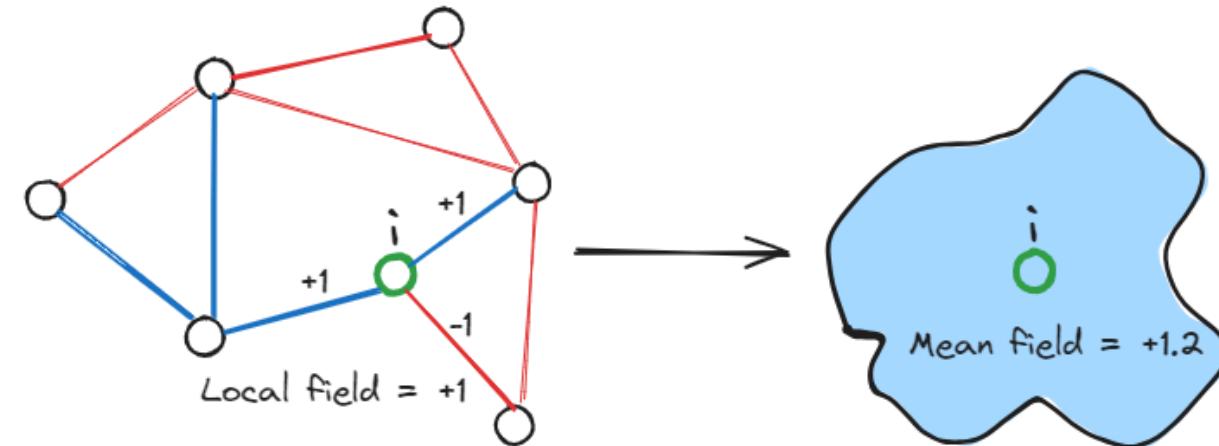
Microstate

Mesostate

Analytical Approach:

- Replacing individual interactions with a single effective interaction representing the average effect of the system.

MEAN FIELD APPROXIMATION



MF Approximation :Microstate

For a duplex network

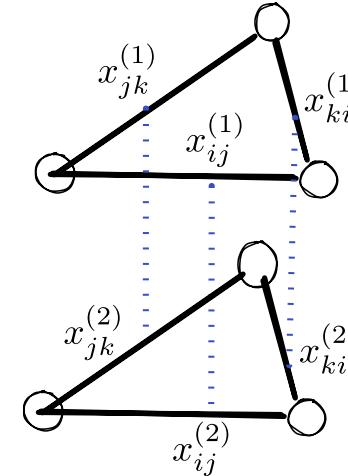
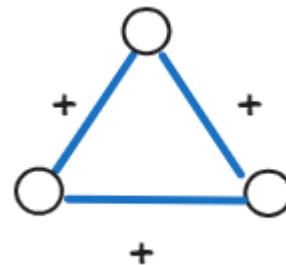
Elementary Subsystems: Pair of coupled links $\vec{x}_{ij} = [x_{ij}^{(1)}, x_{ij}^{(2)}]$

Interactions: These pairs experience interlayer interactions proportional to $[\langle x^{(1)} \rangle^2, \langle x^{(2)} \rangle^2]$

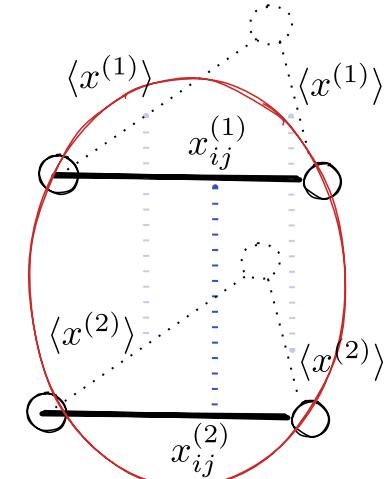
The probabilities of states of this pair are described by the **canonical ensemble** $P(\vec{x}_{ij}) = \frac{\exp(-E(\vec{x}_{ij})/T)}{\sum_{\vec{x}_{mn}} \exp(-E(\vec{x}_{mn})/T)}$

$$E(\vec{x}_{ij}) = -A(N-2)x_{ij}^{(1)}(x^{(1)})^2 - A(N-2)x_{ij}^{(2)}(x^{(2)})^2 - Kx_{ij}^{(1)}x_{ij}^{(2)}$$

Average polarization $\langle \vec{x}_{ij} \rangle = \sum_{\vec{x}_{ij}} P(\vec{x}_{ij})\vec{x}_{ij}$



\approx



$$P\left(\begin{array}{c} \text{---} \\ | \\ \text{---} \end{array}\right) = \frac{\exp\left(-\frac{1}{T}H\left(\begin{array}{c} \text{---} \\ | \\ \text{---} \end{array}\right)\right)}{\exp\left(-\frac{1}{T}H\left(\begin{array}{c} \text{---} \\ | \\ \text{---} \end{array}\right)\right) + \exp\left(-\frac{1}{T}H\left(\begin{array}{c} \text{---} \\ | \\ \text{---} \end{array}\right)\right) + \exp\left(-\frac{1}{T}H\left(\begin{array}{c} \text{---} \\ | \\ \text{---} \end{array}\right)\right) + \exp\left(-\frac{1}{T}H\left(\begin{array}{c} \text{---} \\ | \\ \text{---} \end{array}\right)\right)}$$

Diagram illustrating the states and their probabilities:

- States are represented by horizontal lines connecting two circles (sites).
- Two states are shown on the left:
 - Top state: Two sites connected by a blue line.
 - Bottom state: Two sites connected by a red line.
- An orange arrow points from the bottom state to the top state, labeled $x_{ij}^{(1)}$.
- A second row of states is shown below:
 - Top state: Two sites connected by a blue line.
 - Bottom state: Two sites connected by a red line.
- A third row of states is shown below:
 - Top state: Two sites connected by a red line.
 - Bottom state: Two sites connected by a blue line.
- A fourth row of states is shown below:
 - Top state: Two sites connected by a red line.
 - Bottom state: Two sites connected by a red line.

Mean field solution

Self consistency equations:

$$x^{(1)} = f_1(x^{(1)}, x^{(2)})$$

$$x^{(2)} = f_2(x^{(1)}, x^{(2)})$$

$$f_\alpha(x^{(1)}, x^{(2)}) = \frac{e^{2d} \sinh(a[(x^{(1)})^2 + (x^{(2)})^2]) + \sinh(a(-1)^\alpha [(x^{(2)})^2 - (x^{(1)})^2])}{e^{2d} \cosh(a[(x^{(1)})^2 + (x^{(2)})^2]) + \cosh(a[(x^{(1)})^2 - (x^{(2)})^2])}$$

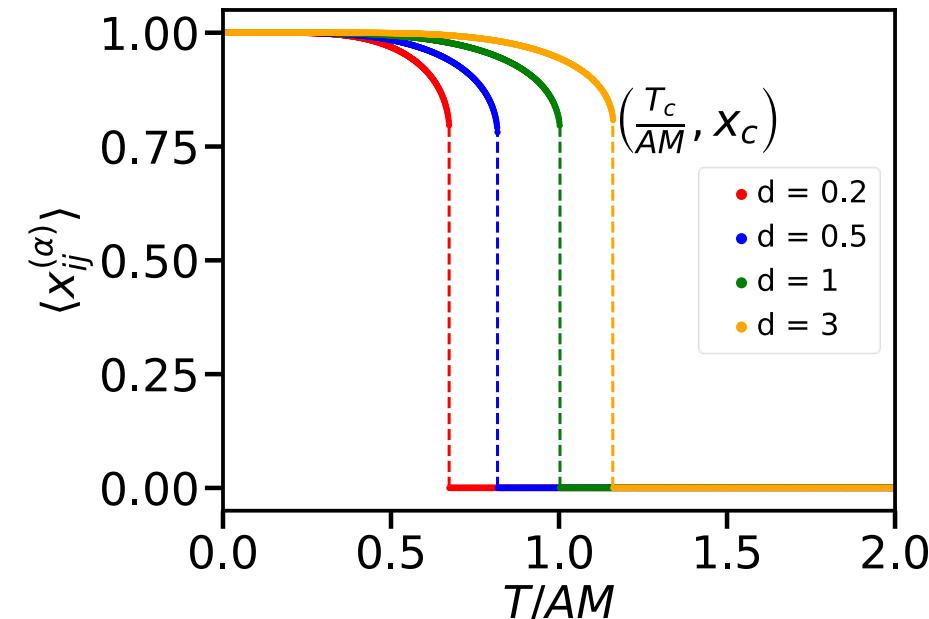
$$a = \frac{AM}{T}$$

$$d = \frac{K}{AM}$$

Re-scaled intralayer and interlayer interaction strength

Observations:

- Temperature and Polarization:** Increasing temperature T leads to a continuous decrease up to x_c .
- Critical Point:** At T_c , the mean polarization abruptly jumps to zero, indicating a first-order transition.
- Unidirectional Transition:** This first-order transition always goes from a polarized to an unpolarized state. Reversed transition does not occur;
- Dependence on Coupling Strength K:** When increasing the coupling strength (K), the critical temperature (T_c) also increases,



Finding critical temperature

Dynamical equations:

$$x^{(1)}(t+1) = f_1(x^{(1)}(t), x^{(2)}(t))$$

$$x^{(2)}(t+1) = f_2(x^{(1)}(t), x^{(2)}(t))$$

Jacobian to find largest Eigenvalue

$$x^{(1)} = x^{(2)} \equiv x$$

$$\lambda_+ = 1$$

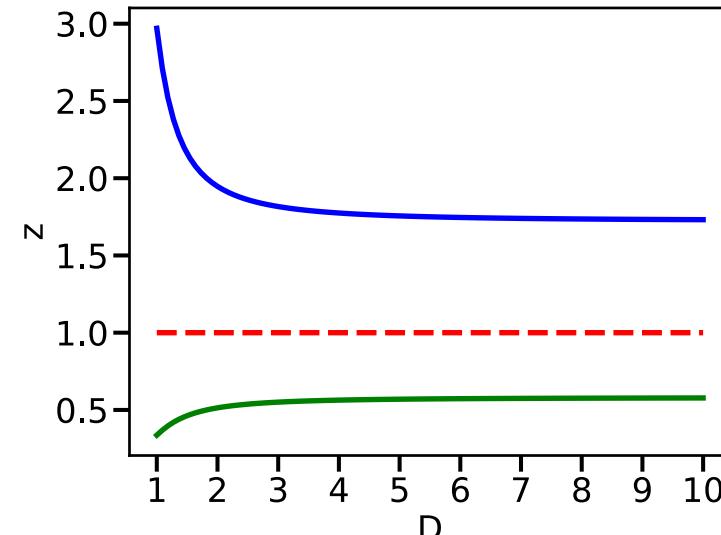
$$z = e^{ax^2}$$

$$D = e^d$$

Calculate the largest value of z for a fixed value of D

$$8 \ln z = \frac{(z^4 - 1)(z^4 D^2 + 2z^2 + D^2)}{z^2(z^4 + 2z^2 D^2 + 1)}$$

$$\frac{T_c}{AM} = \frac{x_c^2}{\ln z} = \left(\frac{1}{\ln z} \right) \left(\frac{D^2(z^4 - 1)}{D^2(z^4 + 1) + 2z^2} \right)^2$$



MF Approximation

Microstate

- Set of links between i and j .
- L number of self consistency equation
- Lower order (duplex, triplex network)

MF Approximation

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Mesostate

- Number of positive links L^+ among all L links in the set.
- One MF equation.
- Higher order multiplex network

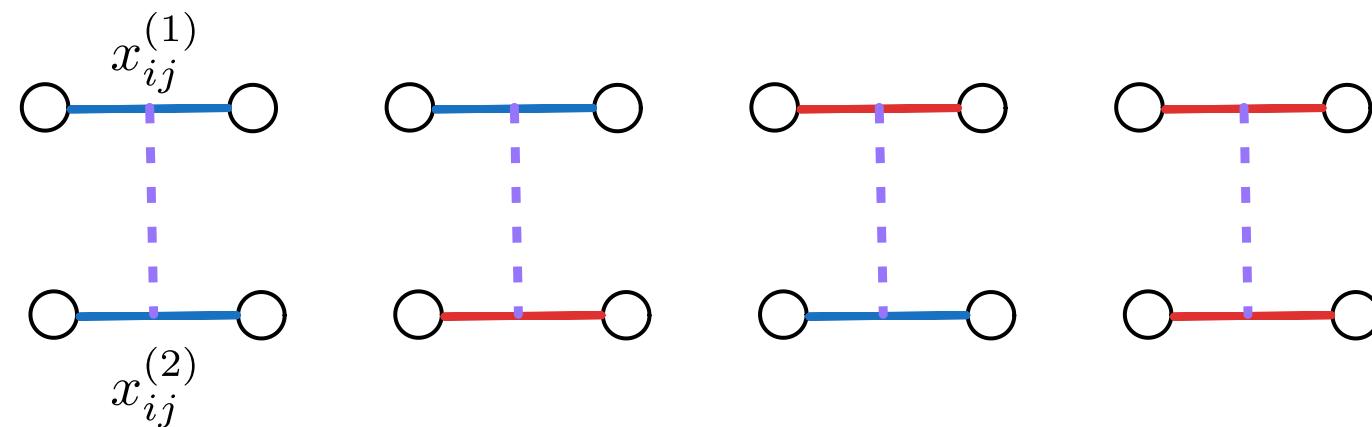
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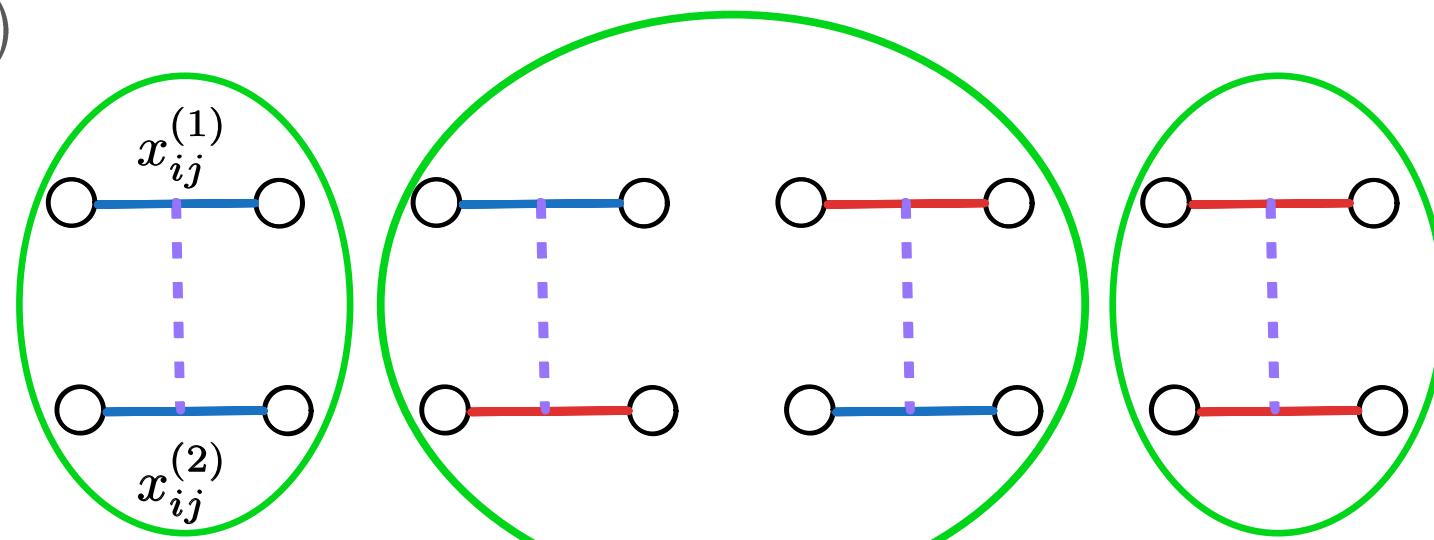
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MF Approximation : mesostate

Mean link polarization

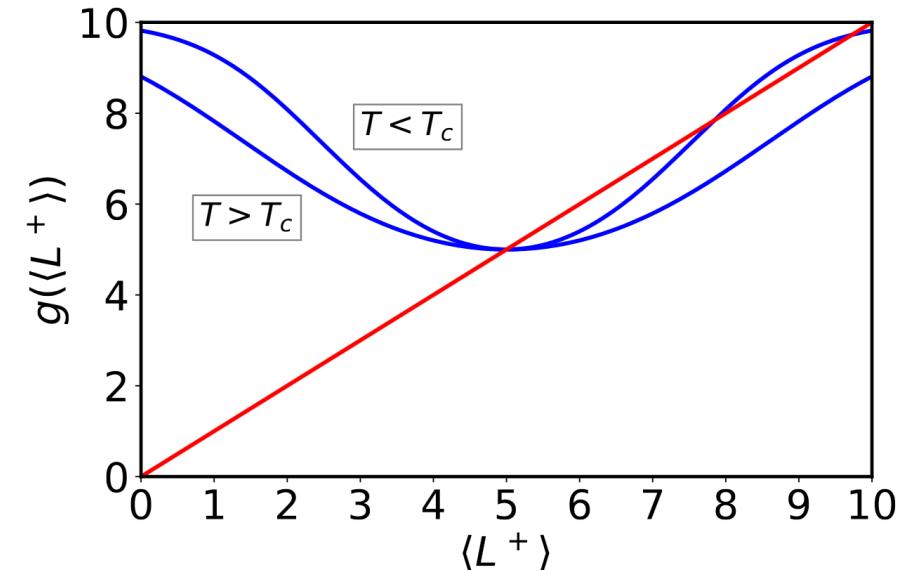
$$\langle x \rangle = \left\langle \frac{L^+ - L^-}{L} \right\rangle = \frac{2\langle L^+ \rangle}{L} - 1$$

$$\mathcal{M} = \binom{L}{L^+}$$

Self-consistent equation

$$\langle L^+ \rangle = g(\langle L^+ \rangle) = \frac{\sum_{L^+=0}^L L^+ \binom{L}{L^+} e^{-E(L^+)/T}}{\sum_{L^+=0}^L \binom{L}{L^+} e^{-E(L^+)/T}}$$

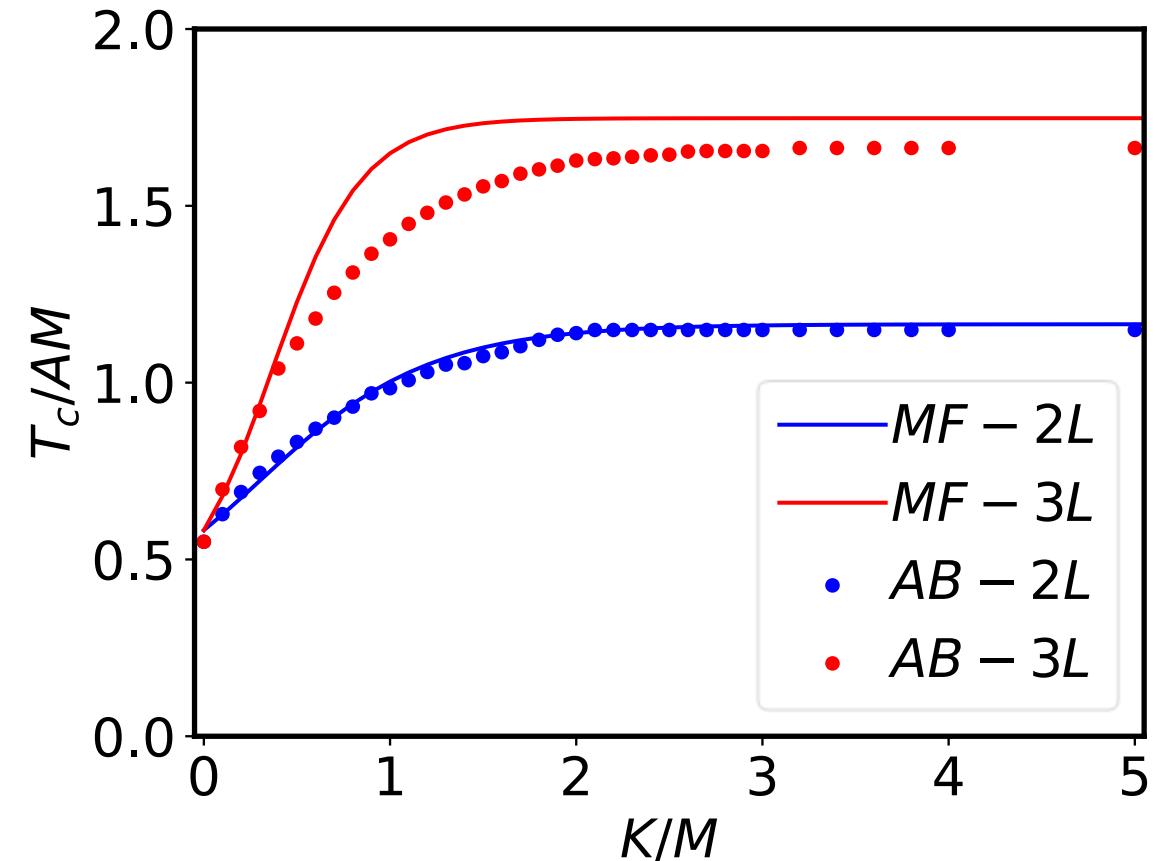
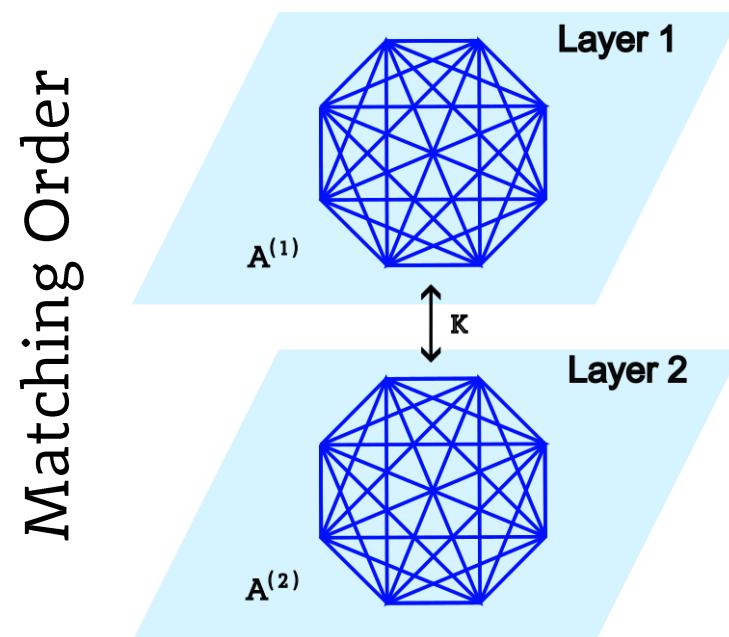
$\langle L_+ \rangle = L$ is a paradise state, and $\langle L_+ \rangle = L/2$ is a disordered state



Saturation of critical temperature

For large coupling strength K-

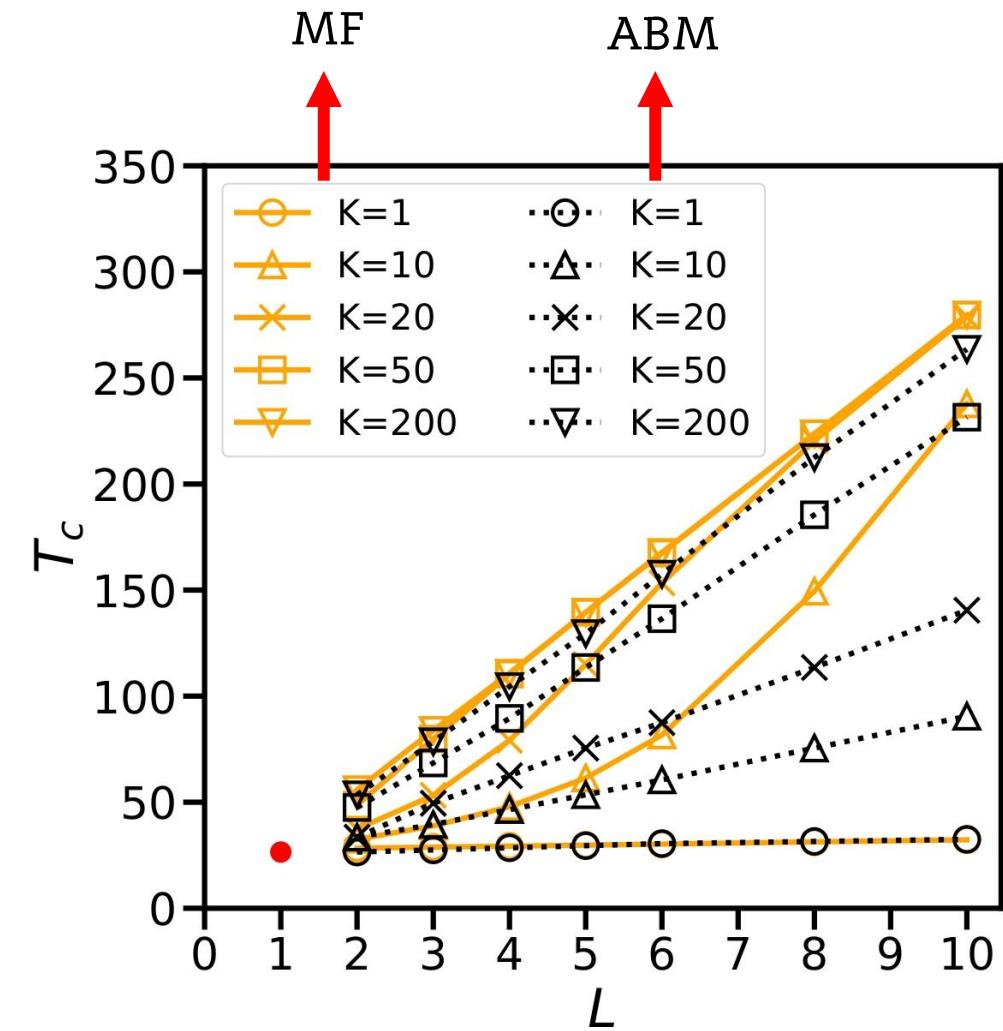
$$T_c|_{\alpha=L, K \rightarrow +\infty} \approx L \cdot T_c|_{\alpha=1}$$



Impact of number of Layers on the Critical temperature

The critical temperature T_c increases with the number of layers L when coupling K is positive

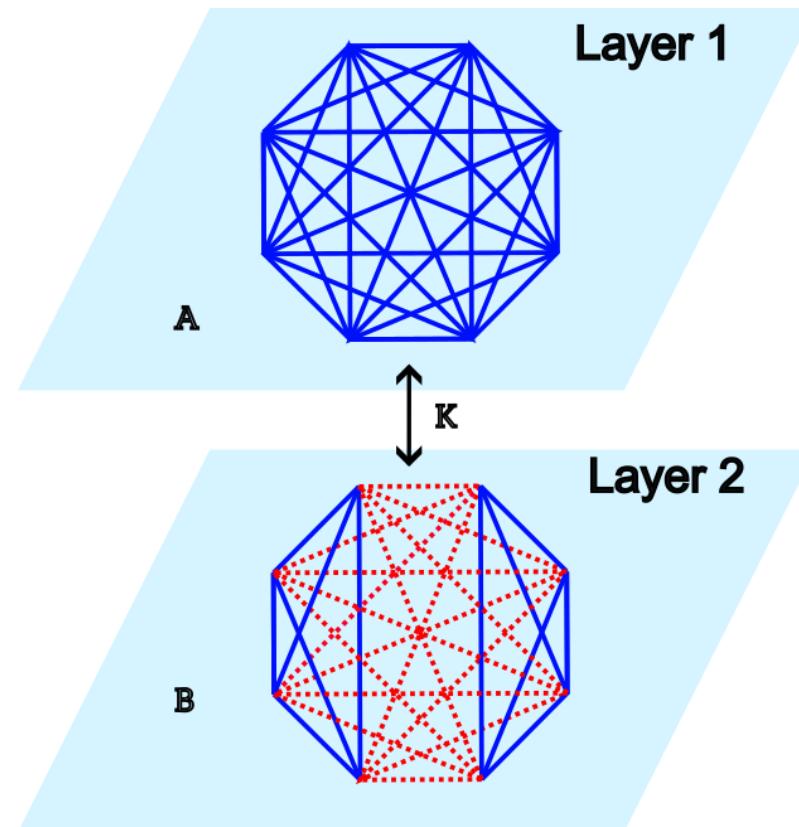
The mean-field approach, while qualitatively correct, is not capable of accurately predicting the behavior of a system with numerous weakly interacting layers.



Critical behavior of the system with layers in different states

What happens when two layers start differently?

Mismatched Order



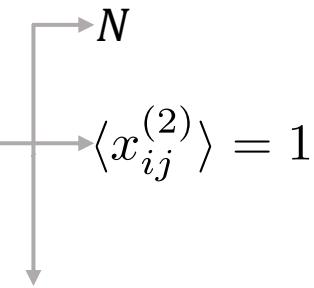
Layer 1

A

Layer 2

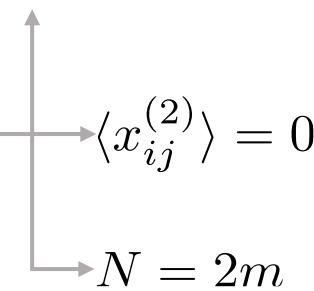
B

Paradise



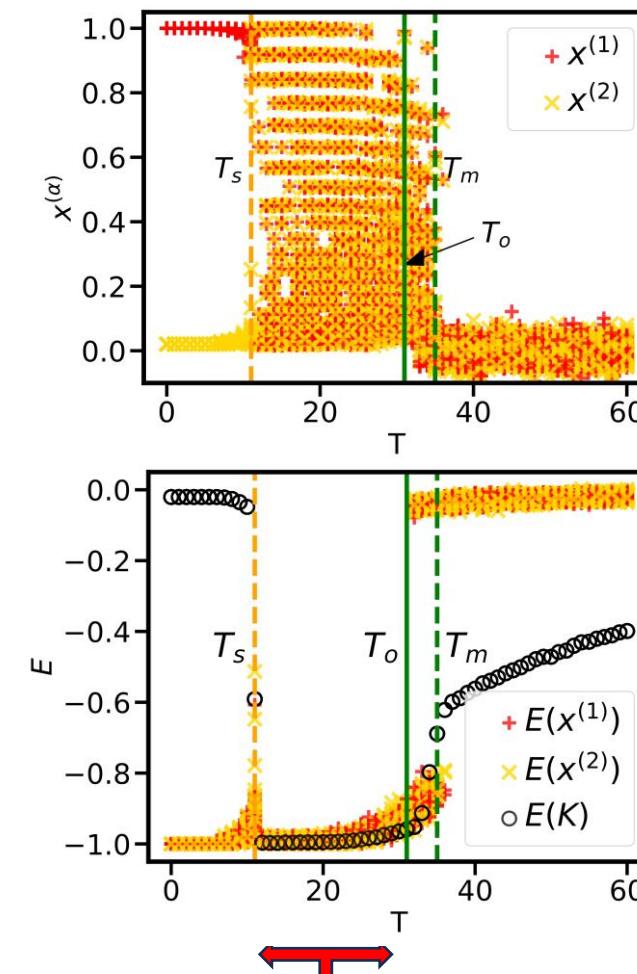
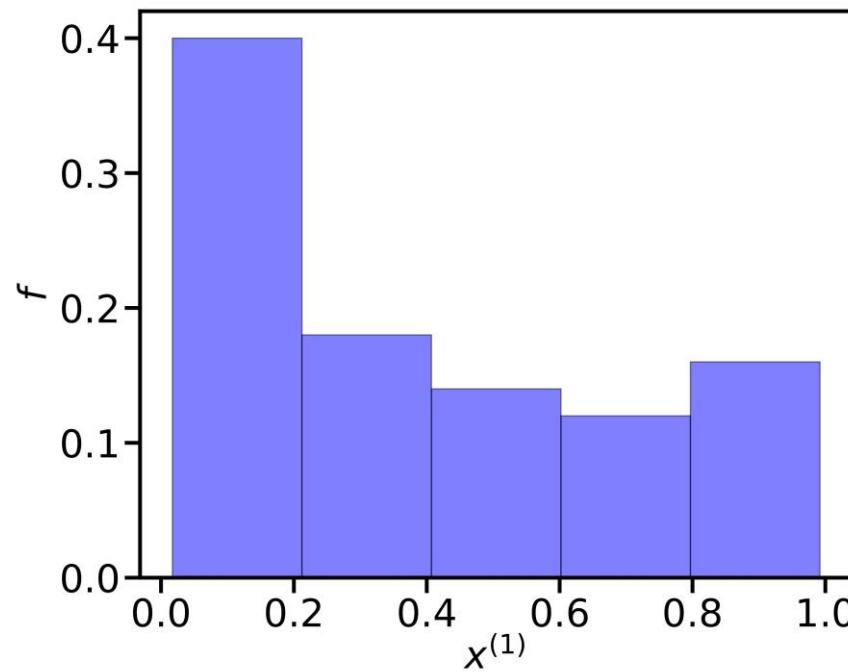
- Heider (Intralayer) Energy in ground state
- Ising (Interlayer) Energy in Excited state

2-Clique



Critical behavior of the system with layers in different states

- Different layers may exist in different states
- Possibilities include one layer ordered and the other disordered, or variations in order within both layers.



Synchronization between the layers!

Critical behavior of the system with layers in different states

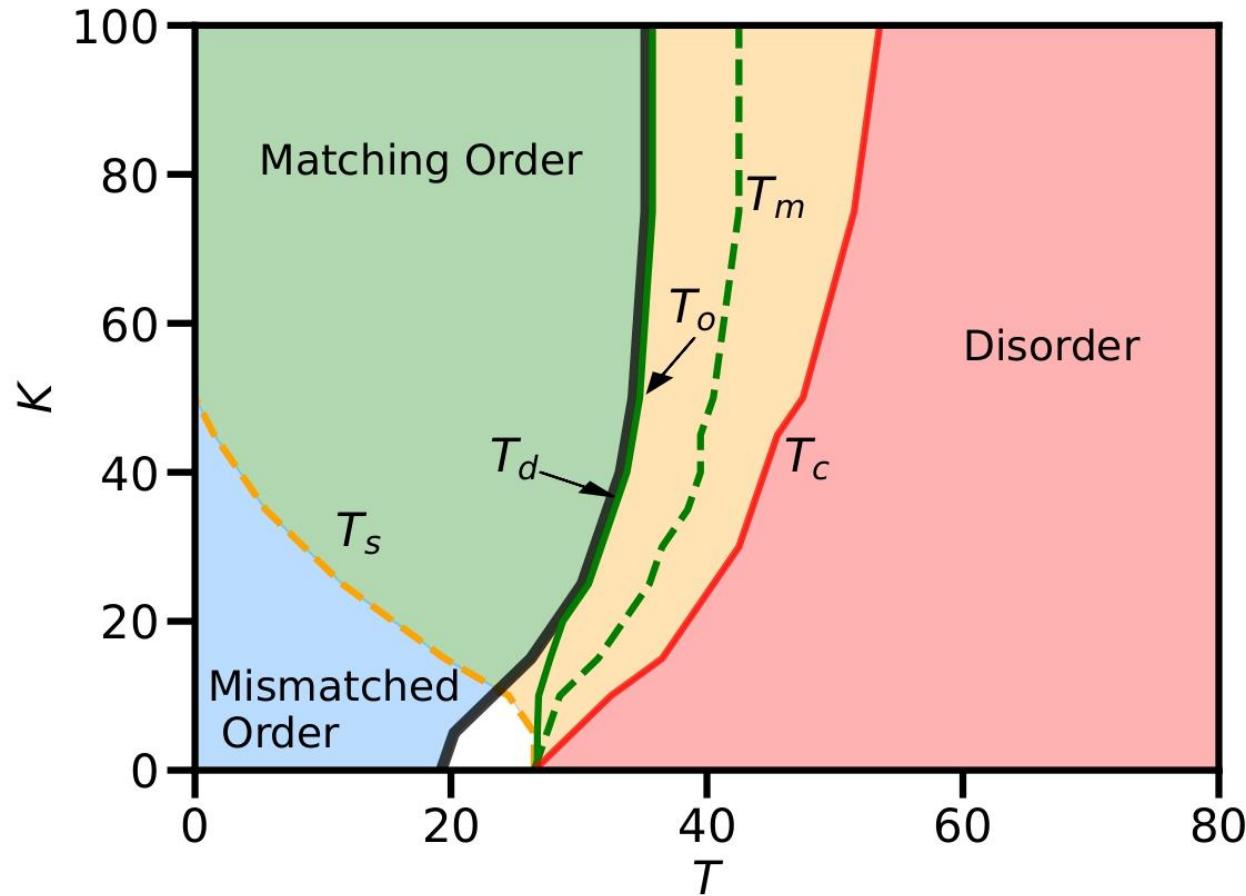
Temperature T_s :

Critical temperature where different states of layers synchronize.

T_m : The limit of retaining order after synchronization

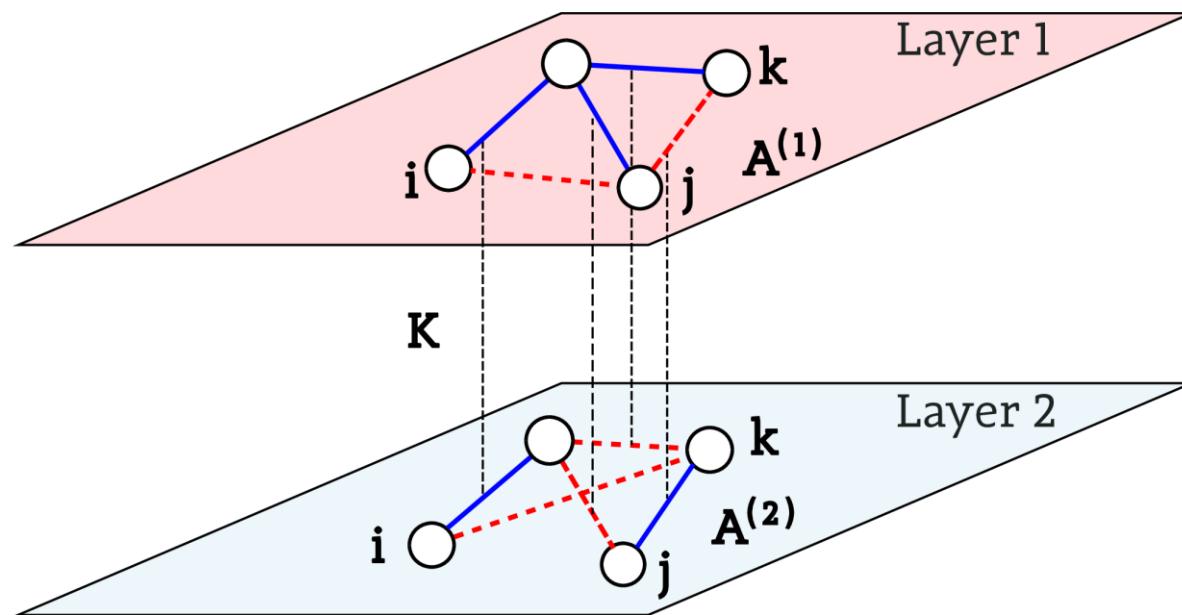
T_o : The limit of spontaneous order

T_c : Critical temperature for the existence of ordered states



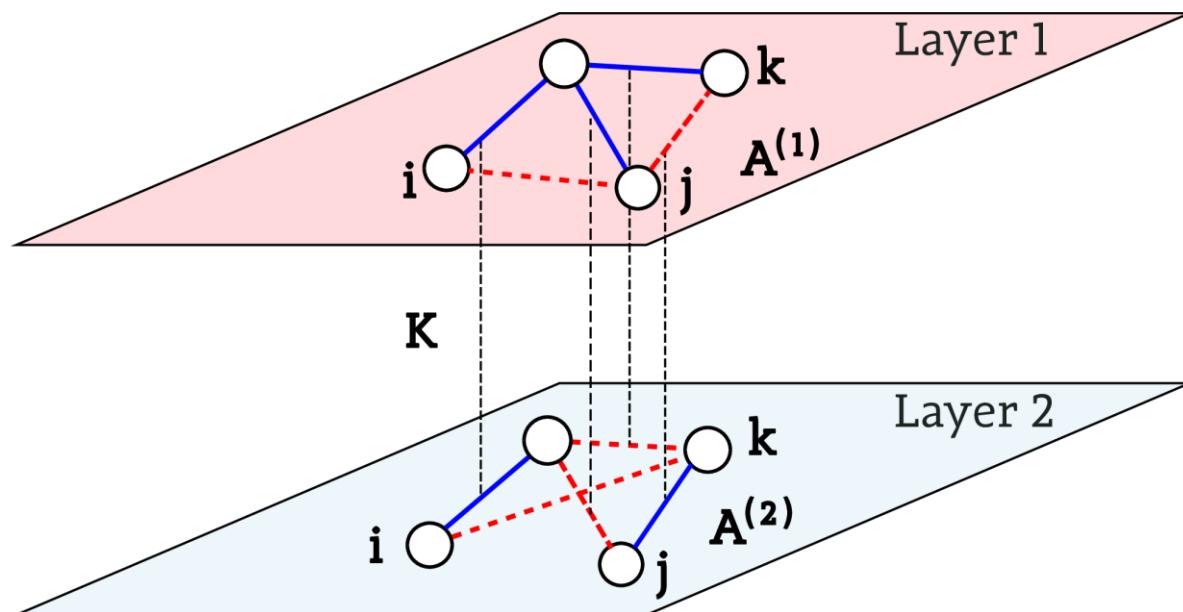
Not everyone knows everyone !

Missing links $\rightarrow x_{ij} = 0$ (not a true link state)



Not everyone knows everyone !

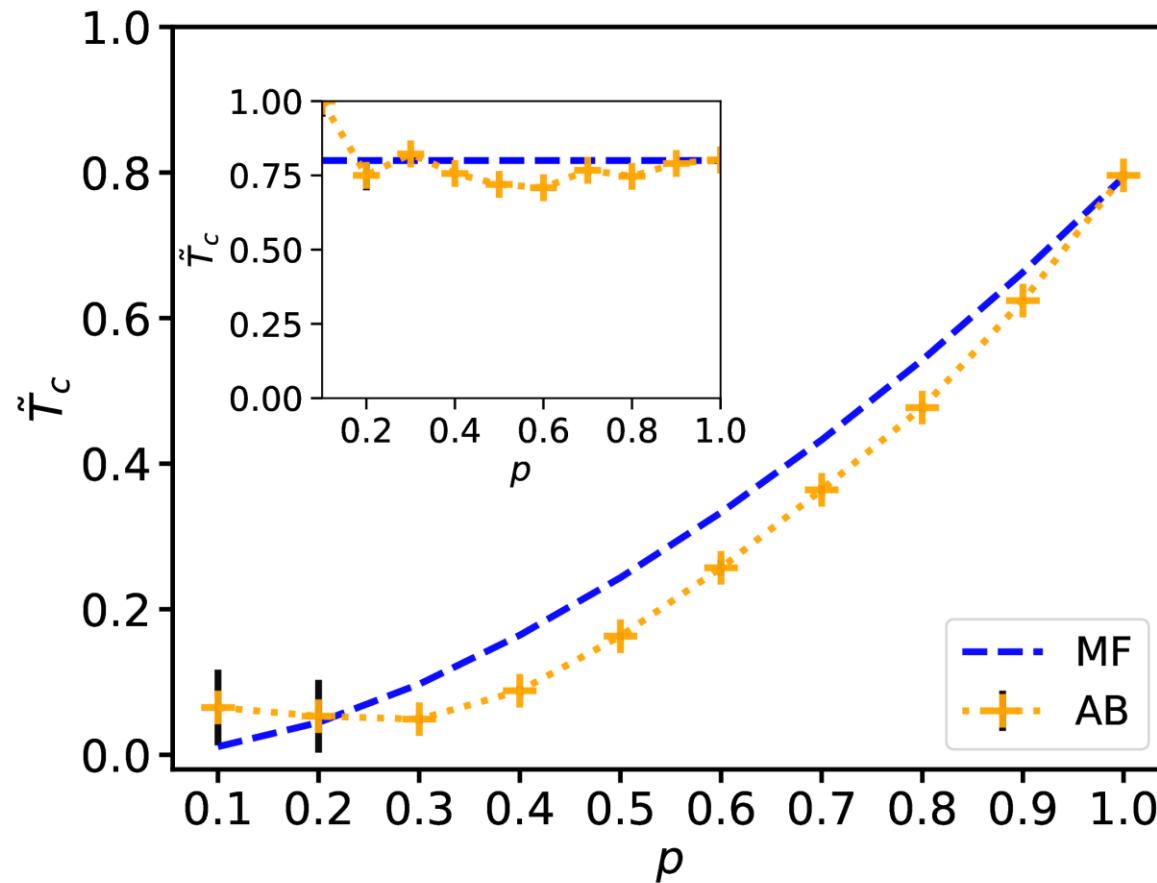
Missing links $\rightarrow x_{ij} = 0$ (not a true link state)



Erdős-Rényi graph !

Heider interaction strength scales with p^2 , while interlayer interaction strength scales with p .

Critical temperature



$$T_c = A \cancel{p}^2 M \left(\frac{1}{\ln z} \right) \left(\frac{D^2(z^4 - 1)}{D^2(z^4 + 1) + 2z^2} \right)^2$$

But T_c not increases like p^2 !

Important results

Social science Lens

Transition from a state of stability to unrest

Tendency of relationships in one layer to align with those in the other.

Influence of balanced or imbalanced relationships becomes increasingly pronounced.

Physics Lens

Temperature rise causes a multiplex network's shift from paradise to disorder, surpassing single-layer critical temperatures.

Synchronization of interlayer relations

Complete graph

Transition

Duplex network with contrasting relationship layers

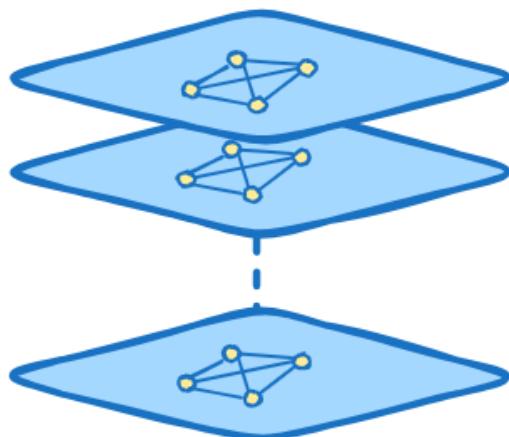
Random graph

Interaction

Heider interaction scales like p^2
Interlayer interaction scales like p .

Main Message

Building several layers of social interactions make the social structures more stable



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ALPHORN PROJECT

- [1] Mohandas, K., Suchecki, K. and Hołyst, J.A., 2024. Physical Review E, 109(4), p.044306.
- [2] P. J. Górska, K. Kułakowski, P. Gawroński, and J. A. Hołyst, Sci. Rep. 7, 16047 (2017).

Thank you