



GAME THEORETICAL INFERENCE OF HUMAN BEHAVIOUR IN SOCIAL NETWORKS

[N. Pagan & F. Dörfler, "Game theoretical inference of human behaviour in social networks". Nature Communications (forthcoming).]

Workshop on "Network Dynamics in the
Social, Economic, and Financial Sciences"

Torino, 07.11.2019

NICOLÒ PAGAN
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AUTOMATIC
CONTROL
LABORATORY **ifa**

ETH zürich

OBSERVATIONS

Actors decide with whom they want to interact.



01

OBSERVATIONS

Actors decide with whom they want to interact.

Network positions provide benefits to the actors.

01

Forbes

3,853 views | Sep 11, 2017, 10:09am

Using Social Networks To Advance Your Career



Adi Gaskell Contributor



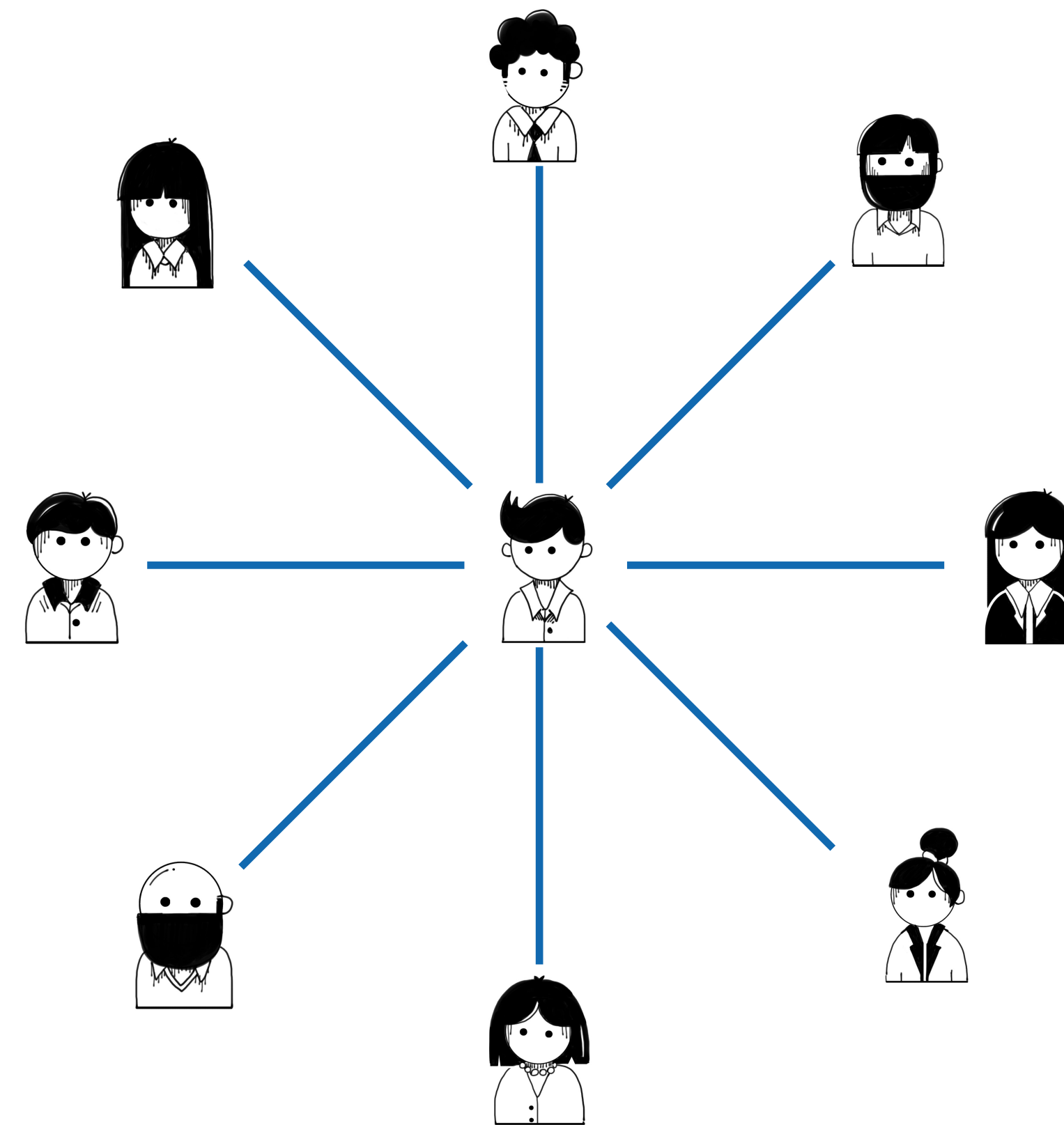
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"It's not what you know, it's who you know" is one of those phrases

SOCIAL NETWORK POSITIONS' BENEFITS

Social Influence

The more people we are connected to, the more we can influence them.



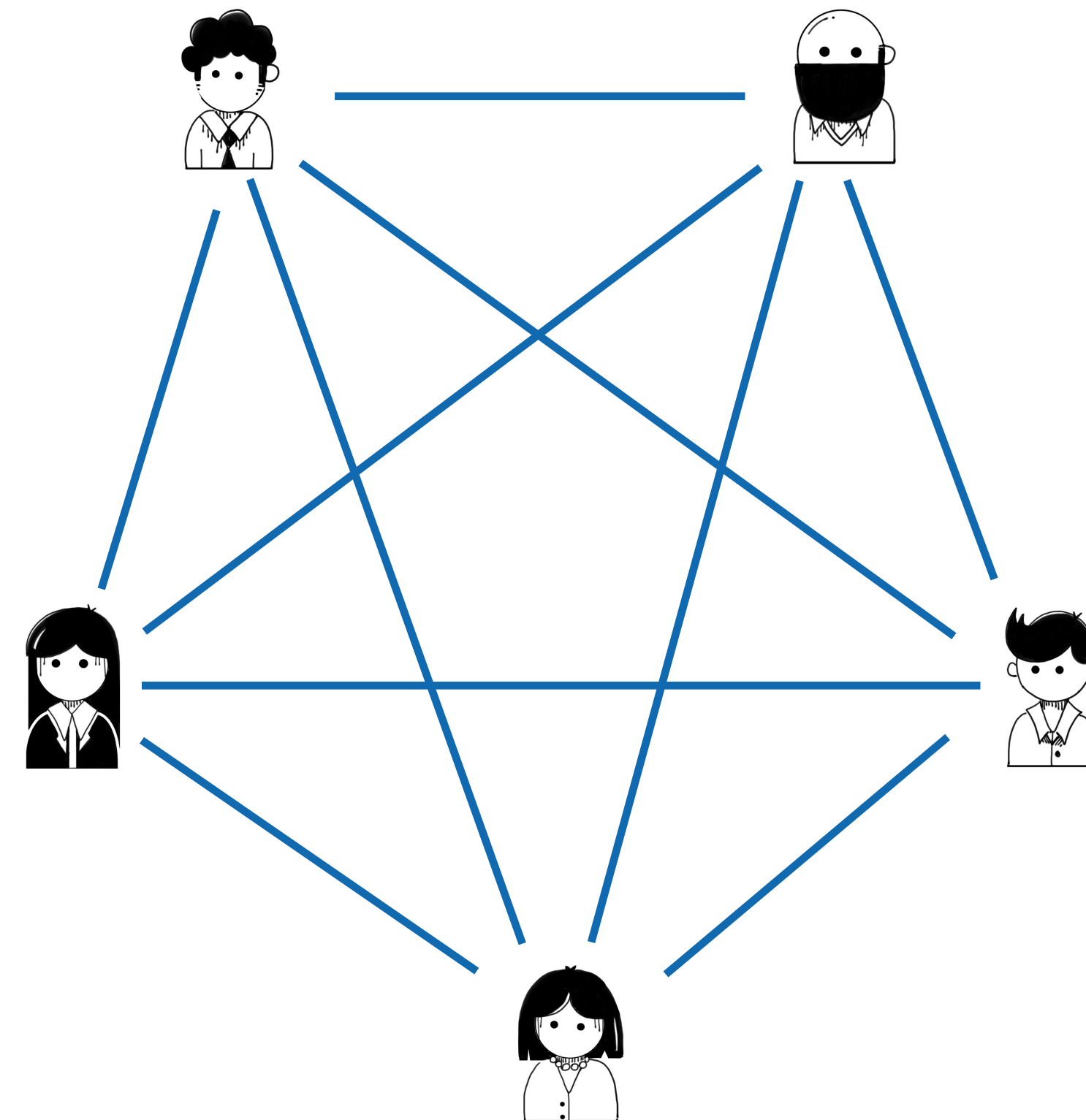
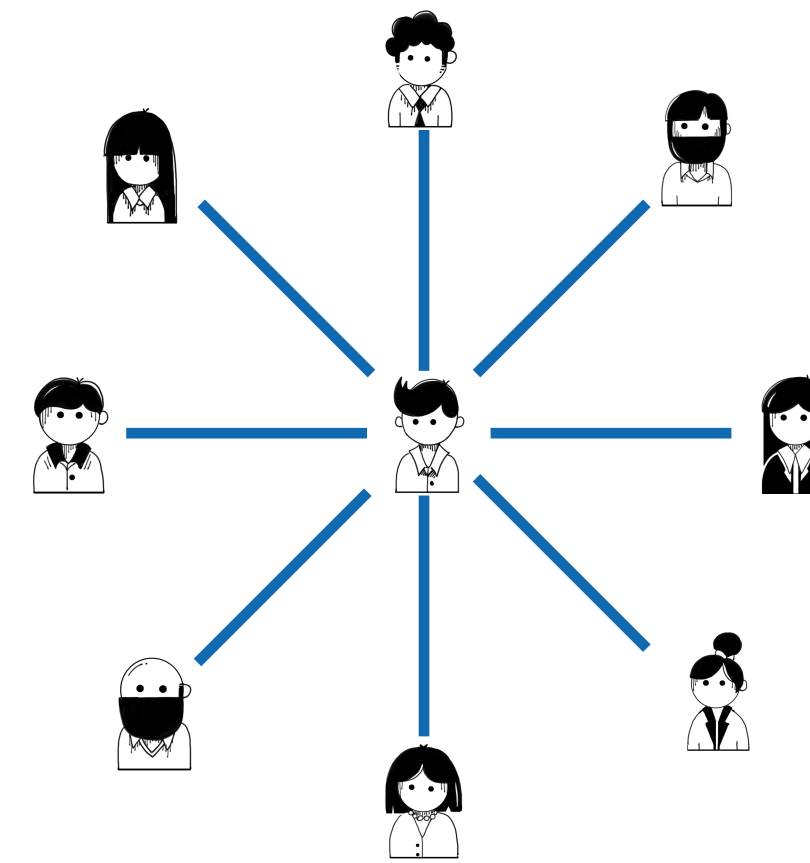
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Social Support

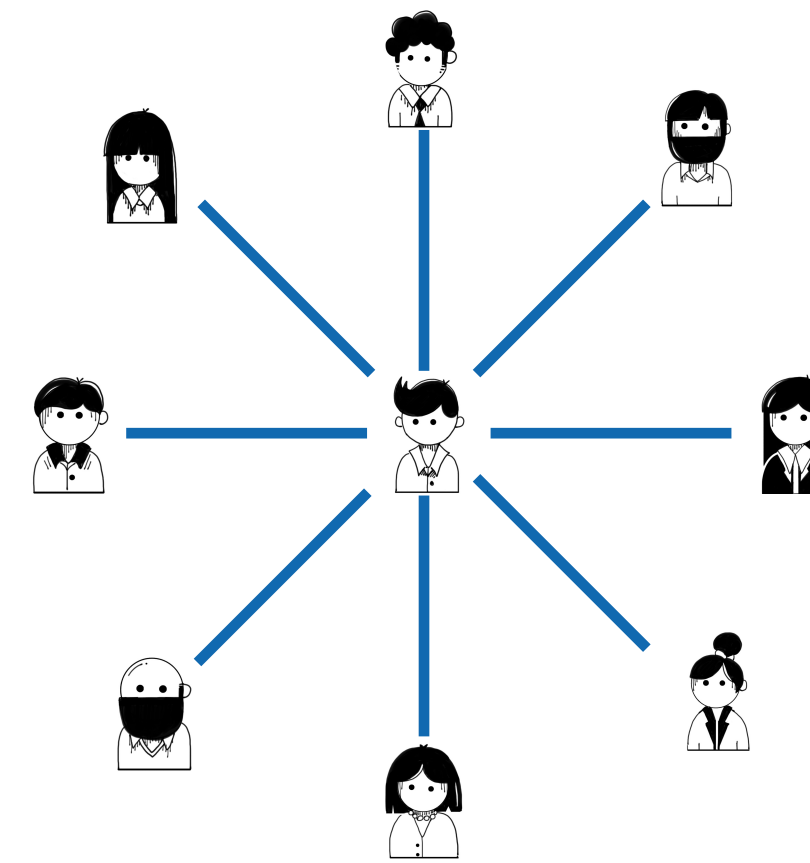
The more our friends' friends are our friends, the safer we feel.



SOCIAL NETWORK POSITIONS' BENEFITS

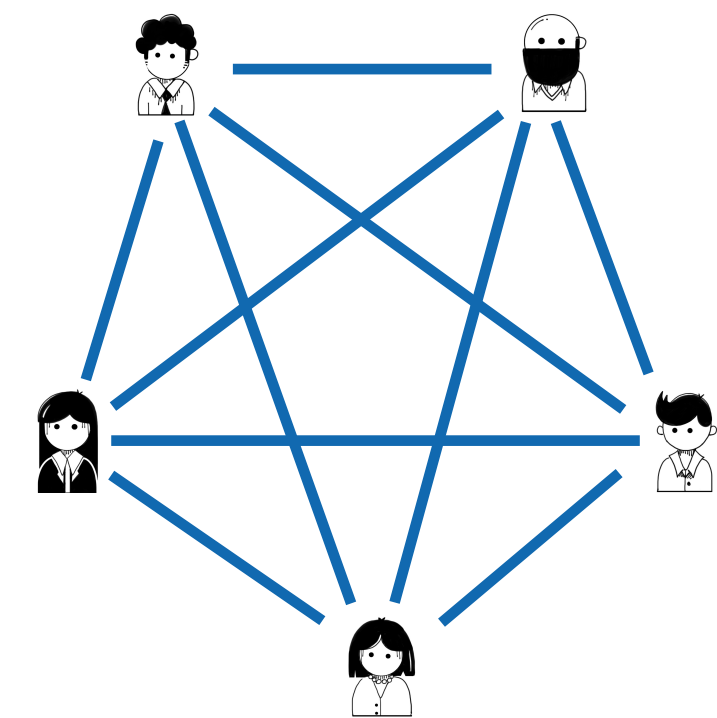
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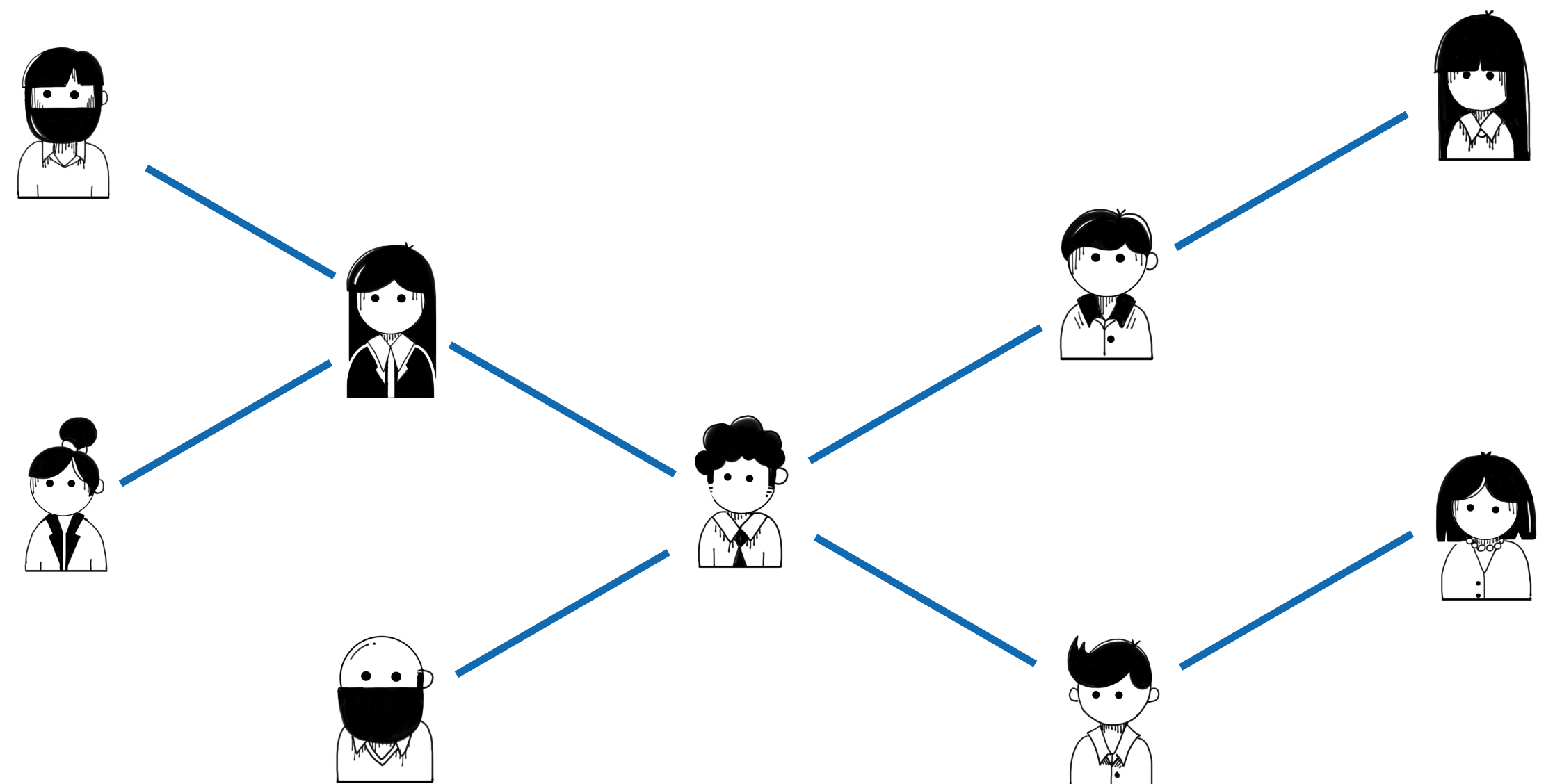
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Brokerage

The more we are on the path between people, the more we can control.



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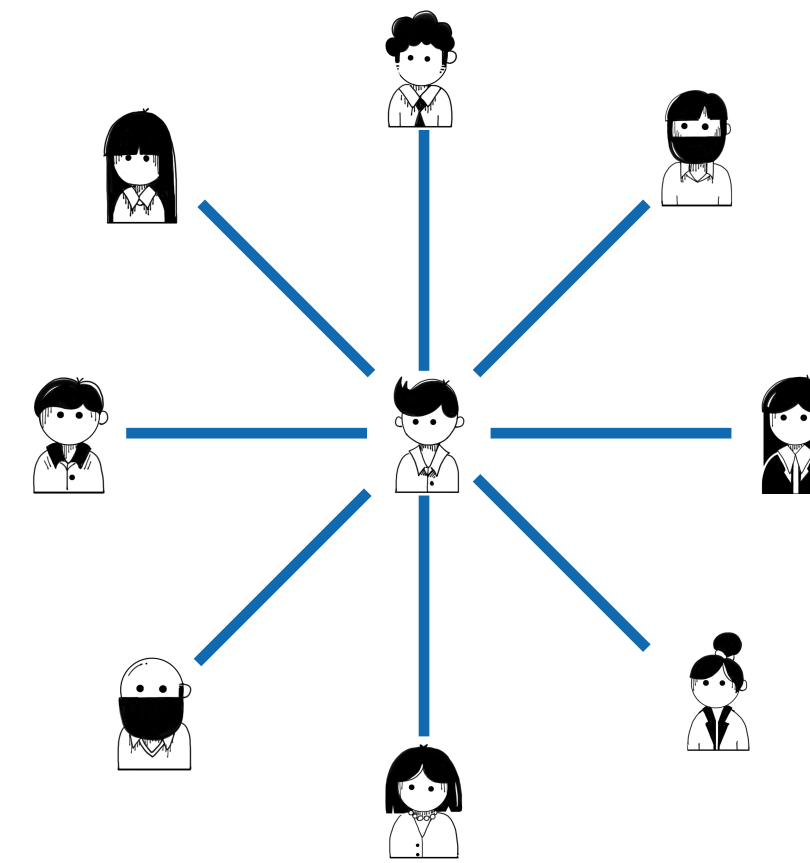
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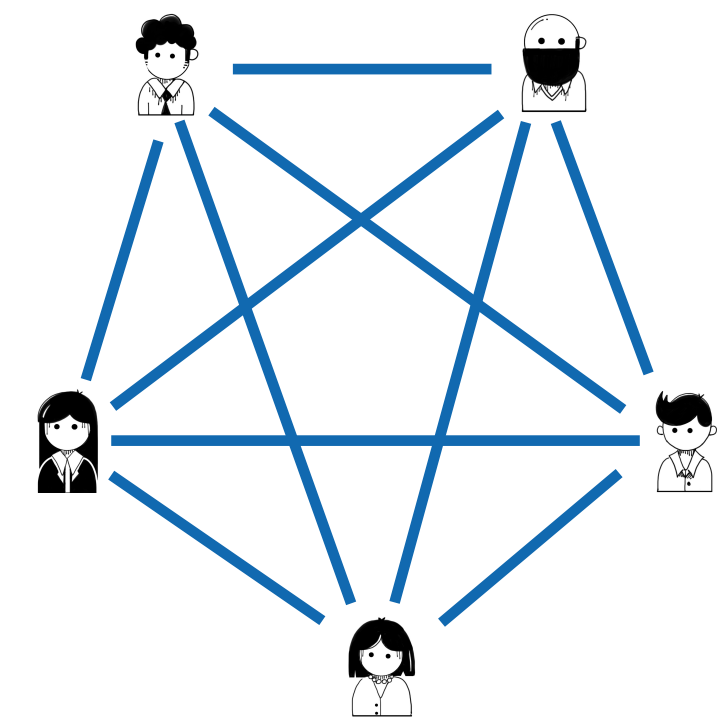
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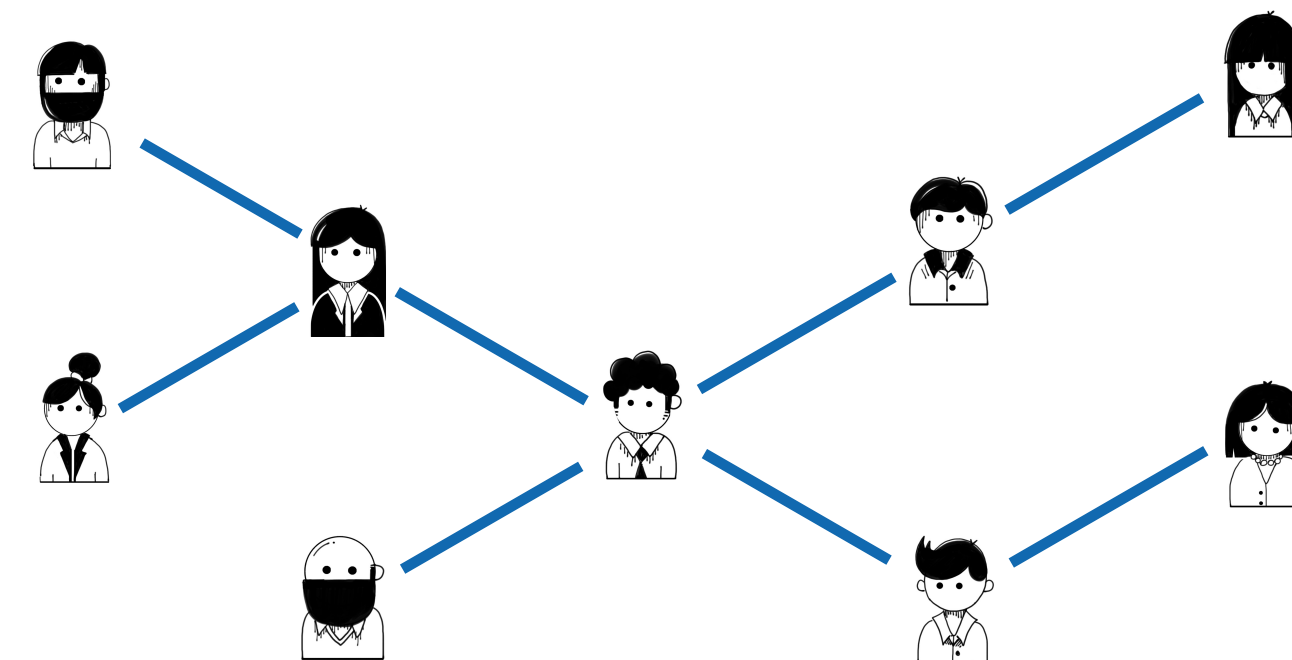
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Degree Centrality



Clustering coefficient



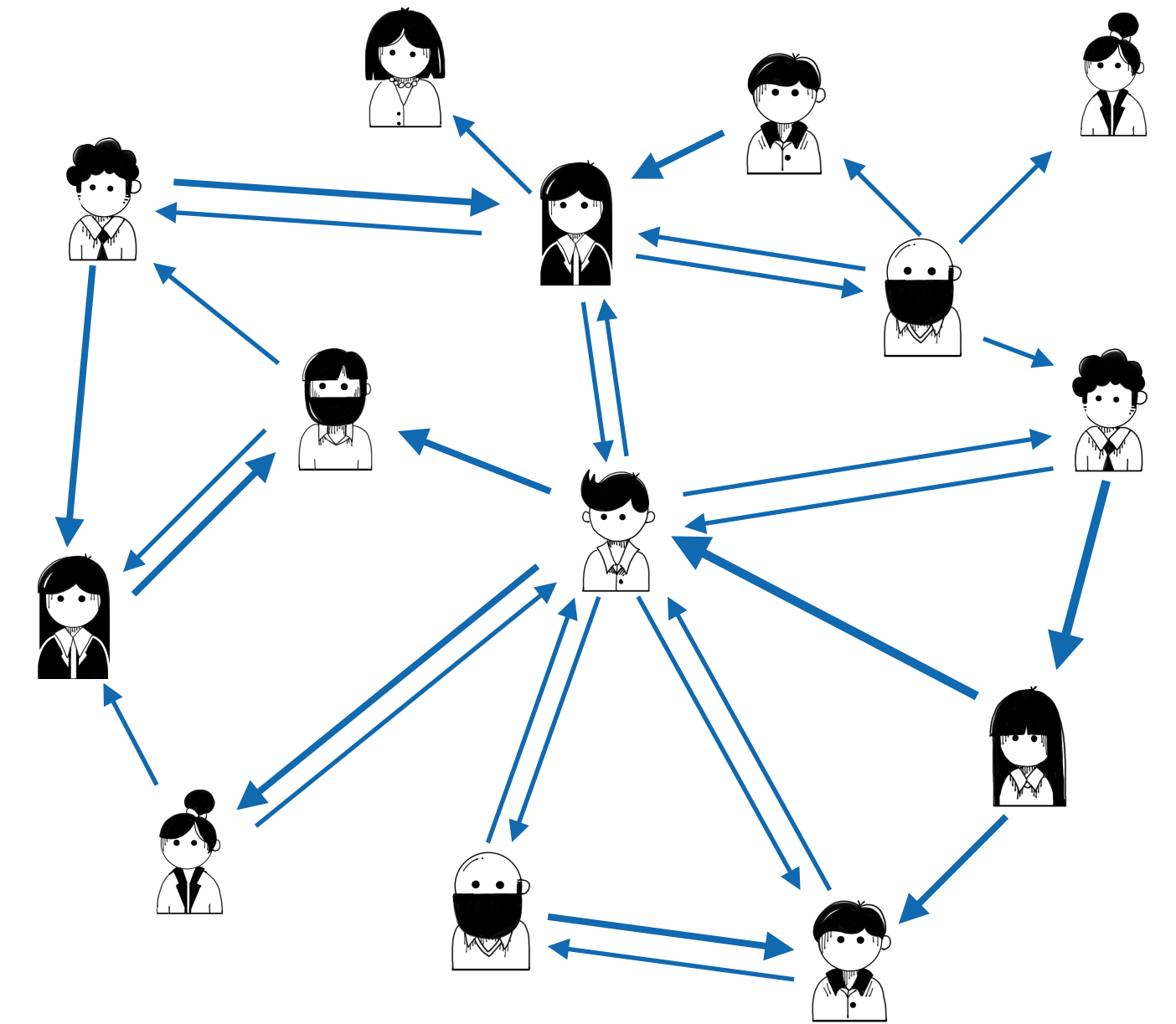
Betweenness Centrality

SOCIAL NETWORK FORMATION MODEL

Directed weighted network \mathcal{G} with $\mathcal{N} = \{1, \dots, N\}$ agents.

$a_{ij} \in [0, 1]$ quantifies the importance of the friendship among i

and j from i 's point of view.



SOCIAL NETWORK FORMATION MODEL

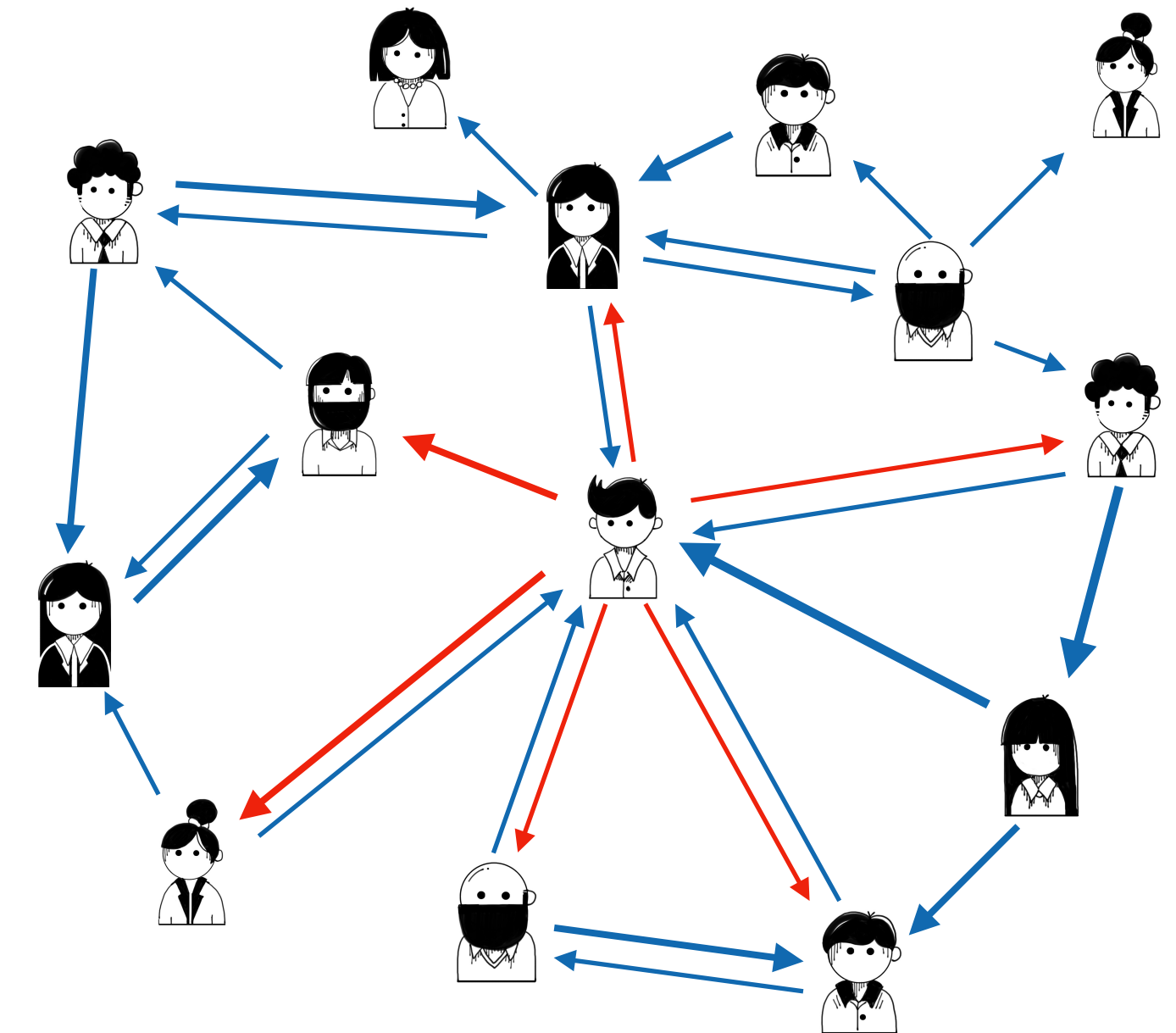
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A typical action of agent i is:

$$a_i = [a_{i1}, \dots, a_{i,i-1}, a_{i,i+1}, \dots, a_{iN}] \in \mathcal{A} = [0, 1]^{N-1},$$



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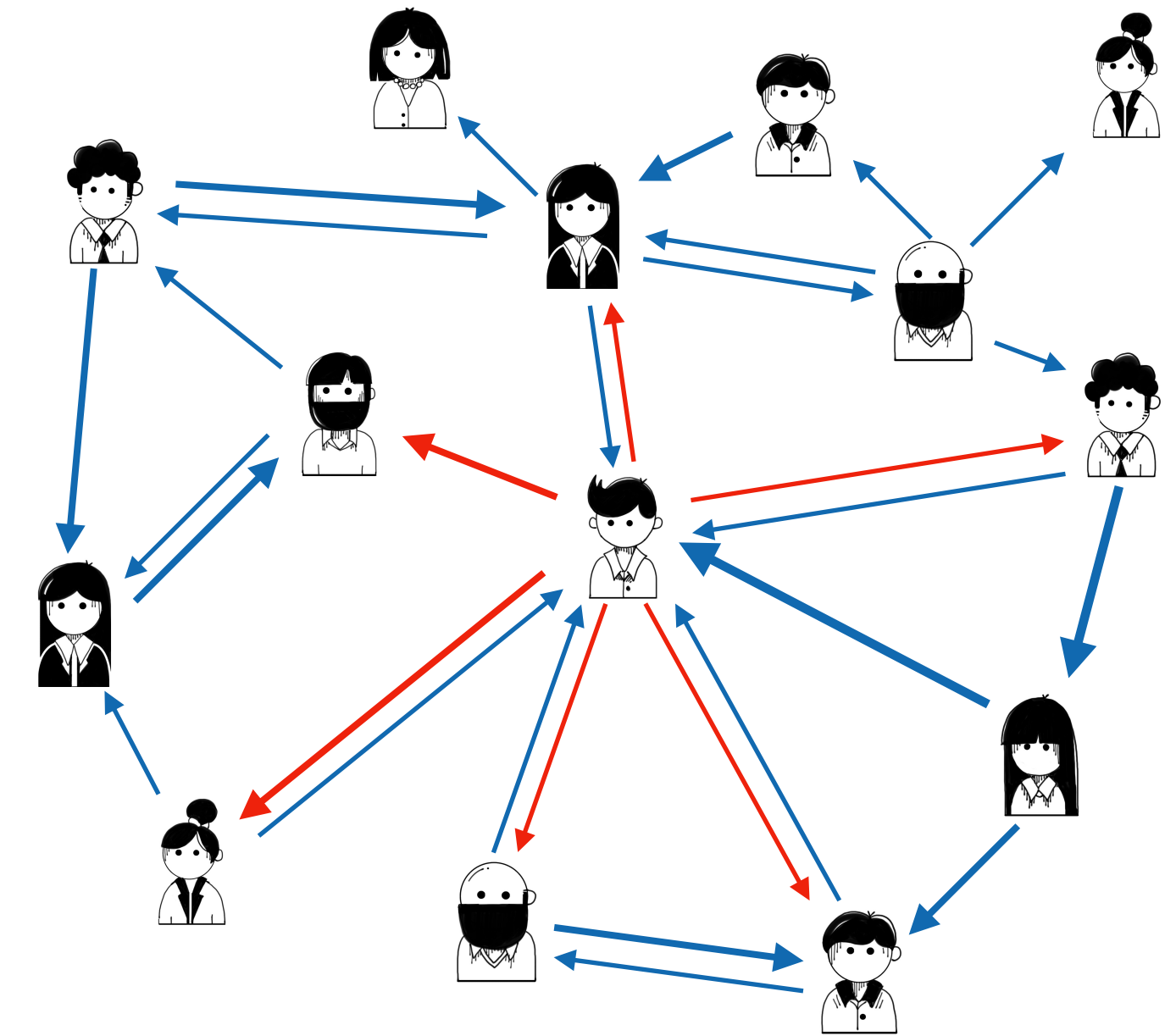
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Every agent i is endowed with a payoff function V_i and is looking for

$$a_i^* \in \arg \max_{a_i \in \mathcal{A}} V_i(a_i, \mathbf{a}_{-i})$$

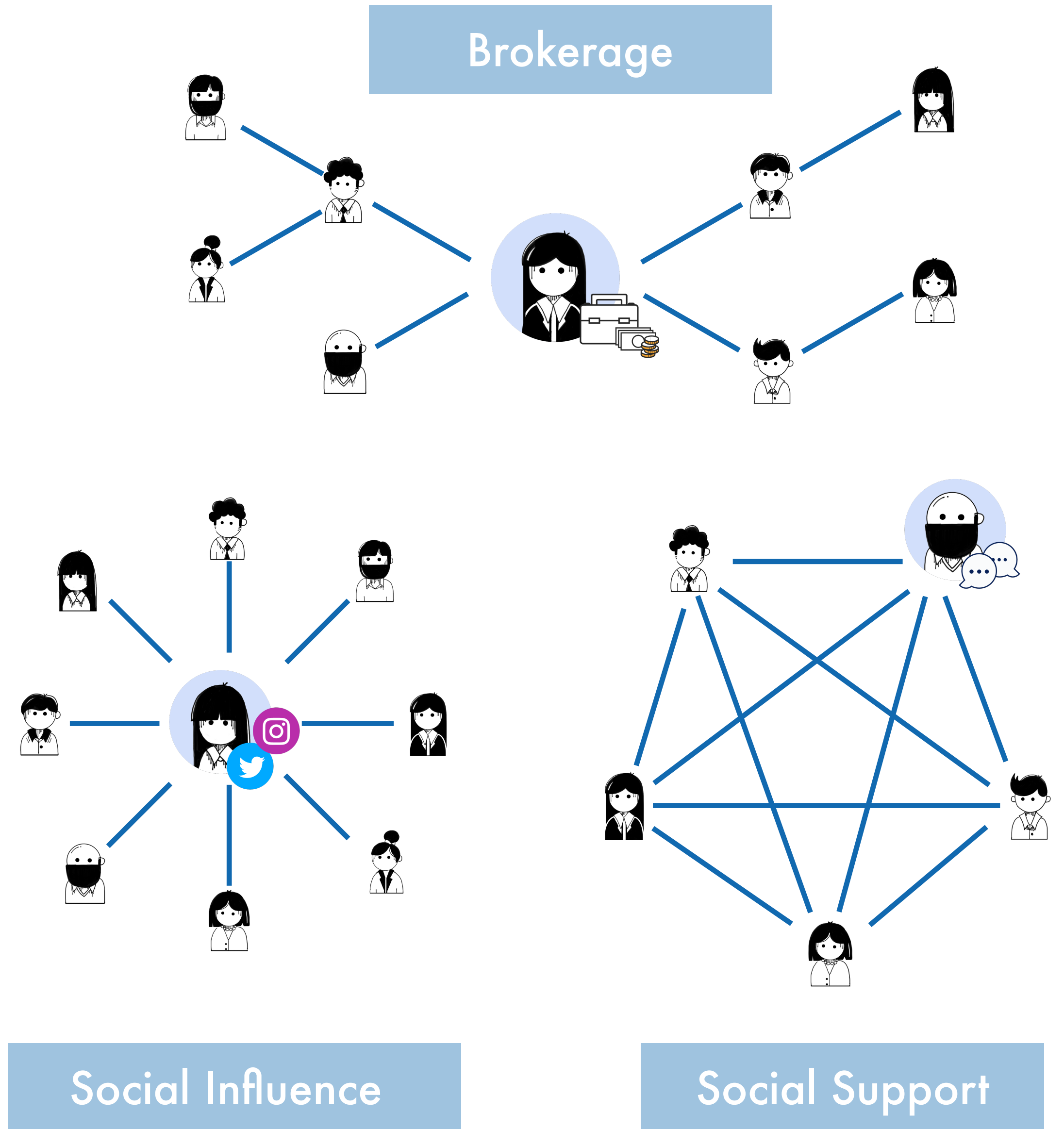


SOCIAL NETWORK FORMATION MODEL

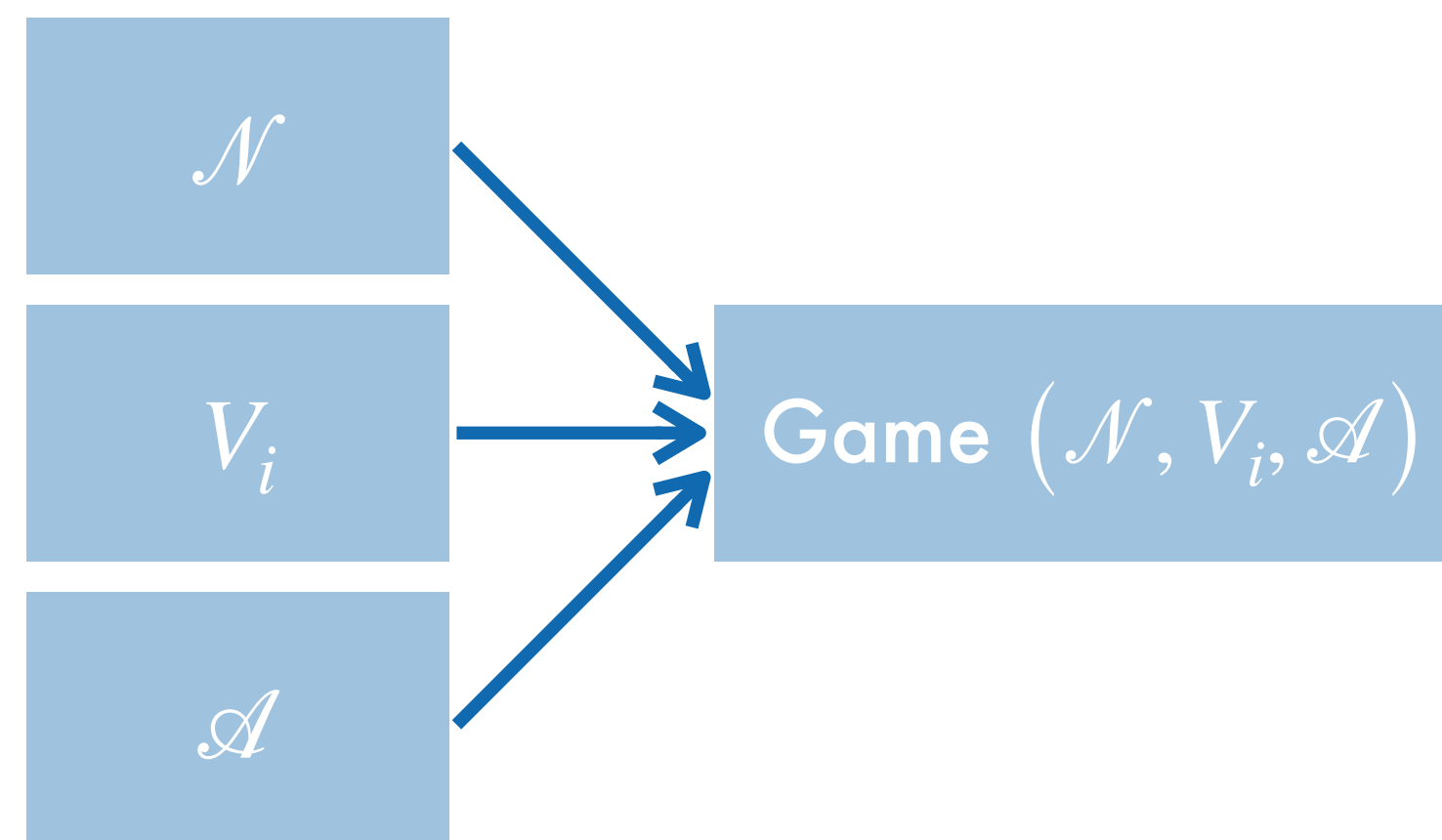
Parametric payoff function:

$$V_i(a_i, \mathbf{a}_{-i}, \theta_i) = \theta_i^T \text{benefit}(a_i, \mathbf{a}_{-i}) - \text{cost}(a_i),$$

where $\theta_i \in \Theta$, are the **individual** parameters relative to the different contributions, e.g., social influence, social support, brokerage.



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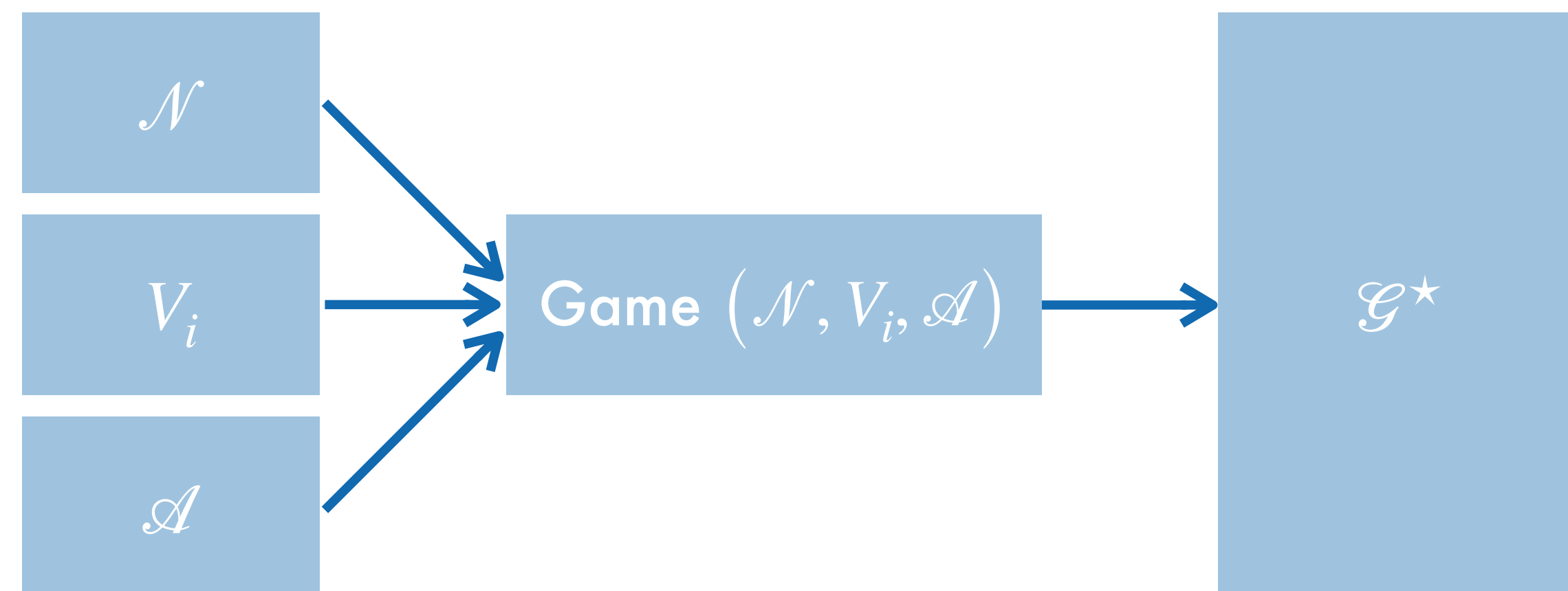


NASH EQUILIBRIUM

Definition.

The network \mathcal{G}^* is a Nash Equilibrium if for all agents i :

$$V_i(a_i, \mathbf{a}_{-i}^* | \theta_i) \leq V_i(a_i^*, \mathbf{a}_{-i}^* | \theta_i), \forall a_i \in \mathcal{A}$$



NASH EQUILIBRIUM

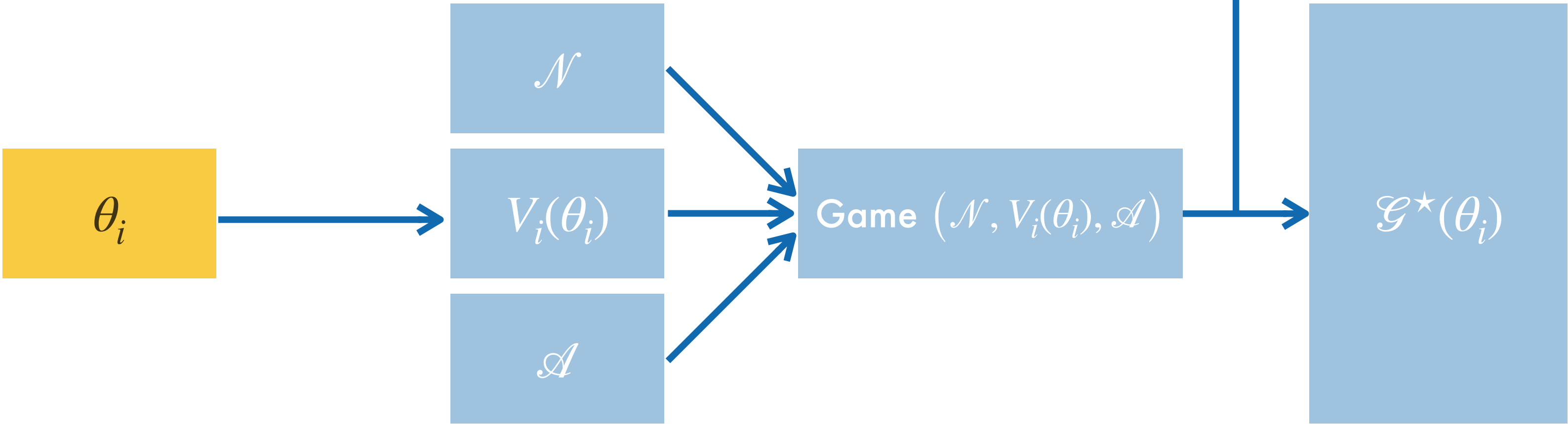
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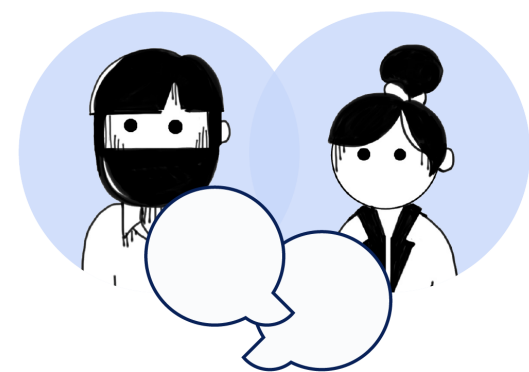
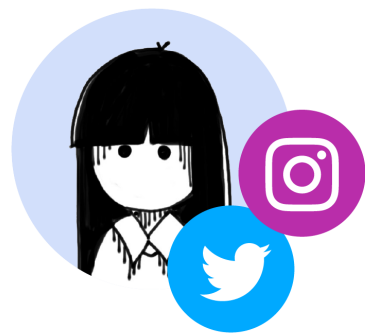
$$V_i(a_i, \mathbf{a}_{-i}^* | \theta_i) \leq V_i(a_i^*, \mathbf{a}_{-i}^* | \theta_i), \forall a_i \in \mathcal{A}$$

\implies

$$\forall i, a_i^* \in \arg \max_{a_i \in \mathcal{A}} V_i(a_i, \mathbf{a}_{-i}^* | \theta_i)$$



INDIVIDUAL BEHAVIOUR θ_i



Question: Given θ_i , which \mathcal{G}^* is in equilibrium?

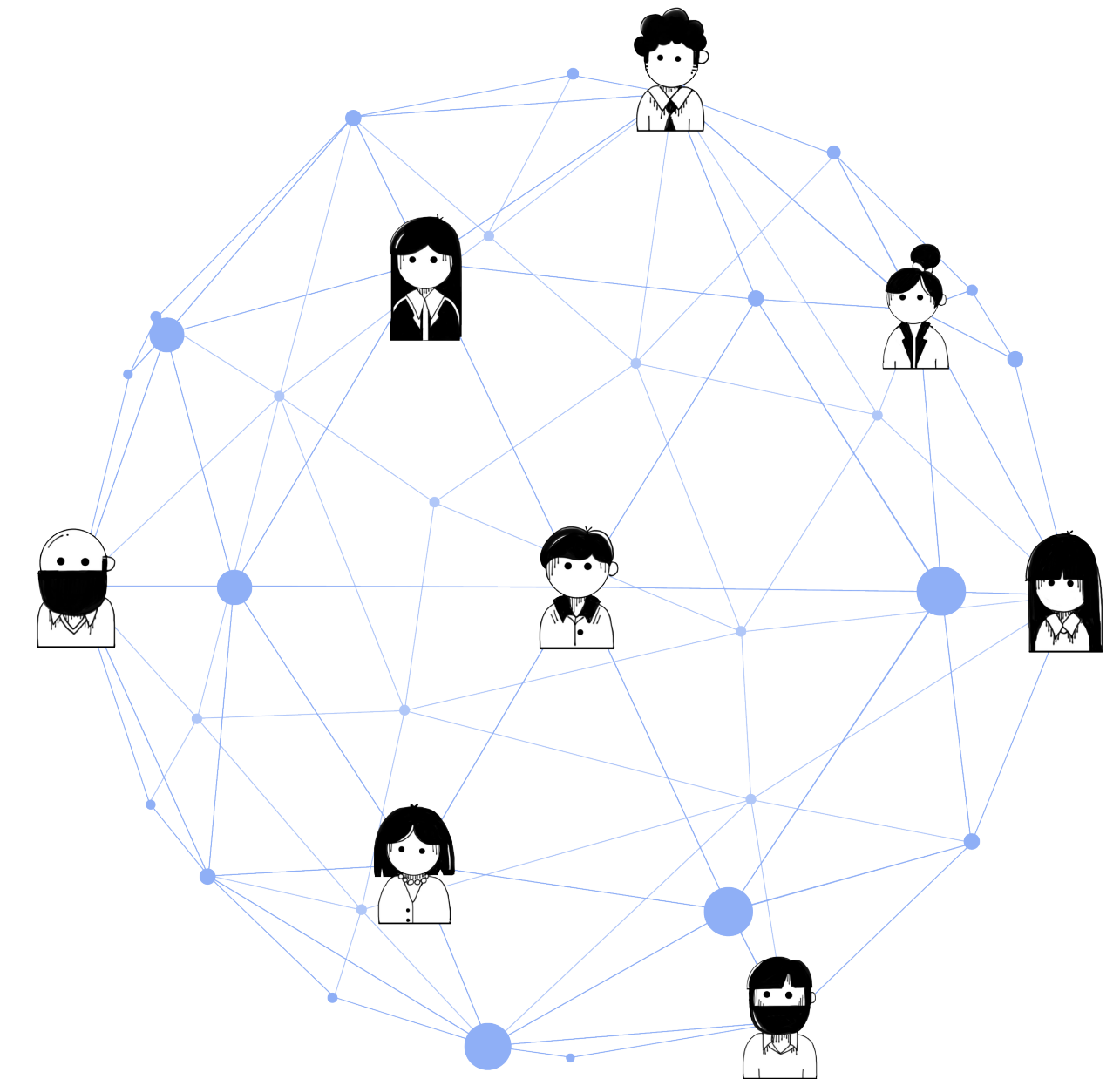
DETERMINE

$$\forall i, a_i^* \in \arg \max_{a_i \in \mathcal{A}} V_i(a_i, \mathbf{a}_{-i}^* | \theta_i)$$

STRATEGIC NETWORK FORMATION MODEL

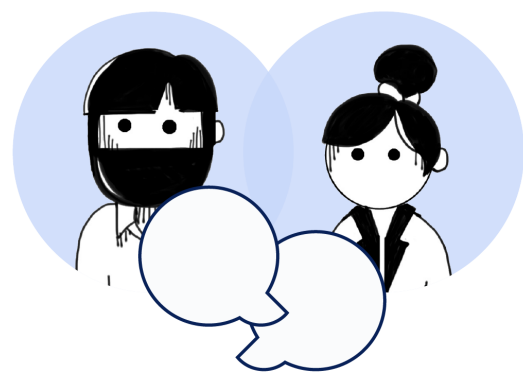
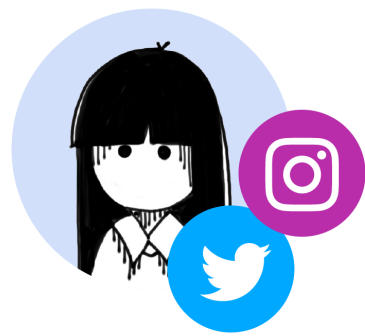


STRATEGIC PLAY

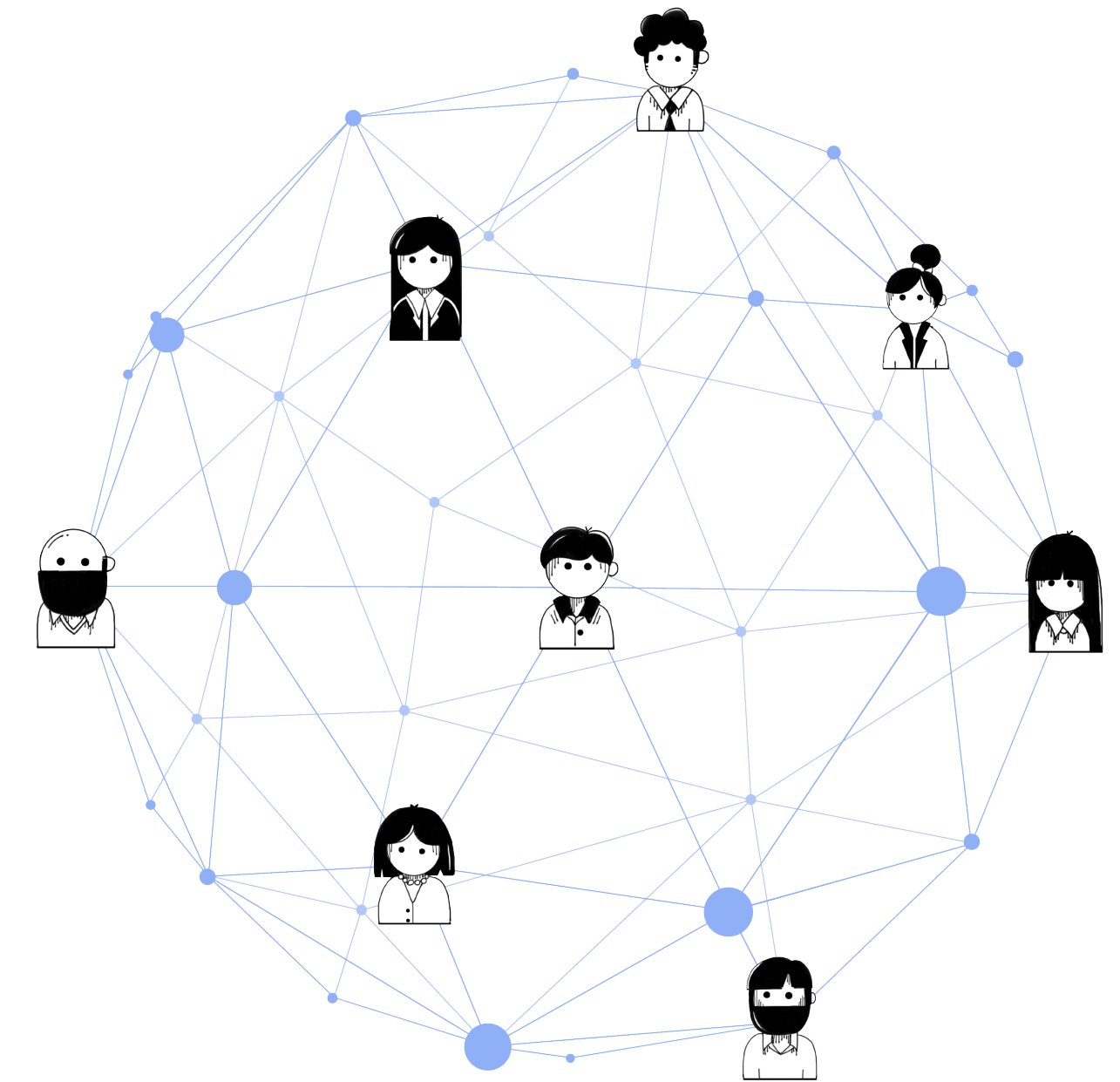


SOCIAL NETWORK STRUCTURE $\mathcal{G}^*(\theta_i)$

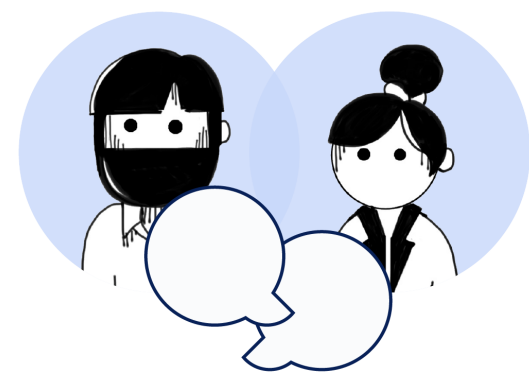
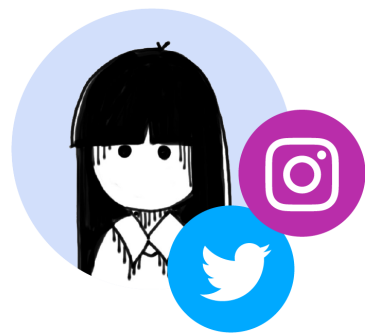
**INDIVIDUAL
BEHAVIOUR θ_i**



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INDIVIDUAL BEHAVIOUR θ_i



Question: Given \mathcal{G}^* , for which θ_i is \mathcal{G}^* in equilibrium?

DETERMINE

$$\forall i, a_i^* \in \arg \max_{a_i \in \mathcal{A}} V_i(a_i, \mathbf{a}_{-i}^* | \theta_i)$$

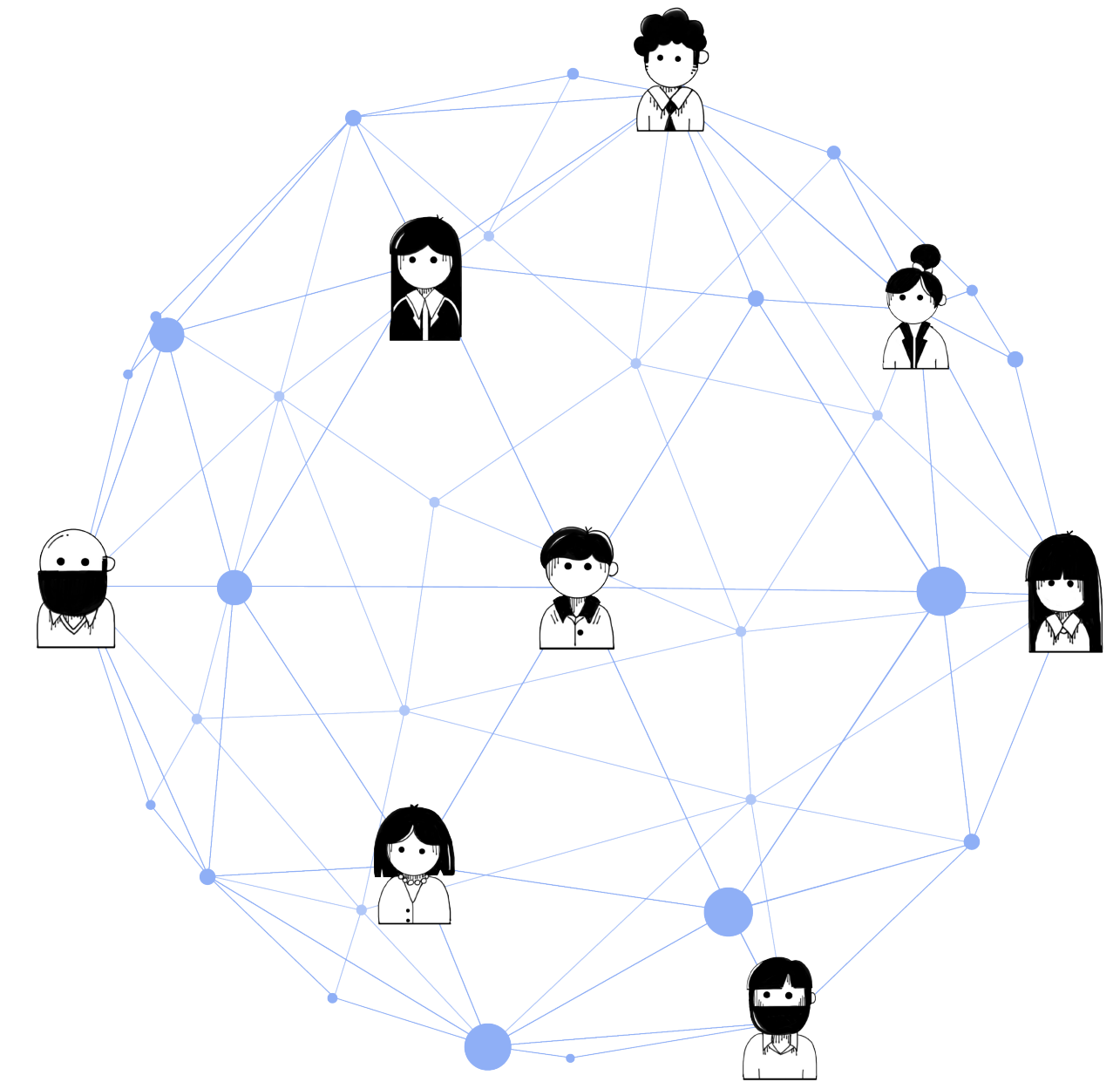
STRATEGIC NETWORK FORMATION MODEL



STRATEGIC PLAY

GAME-THEORETICAL INFERENCE

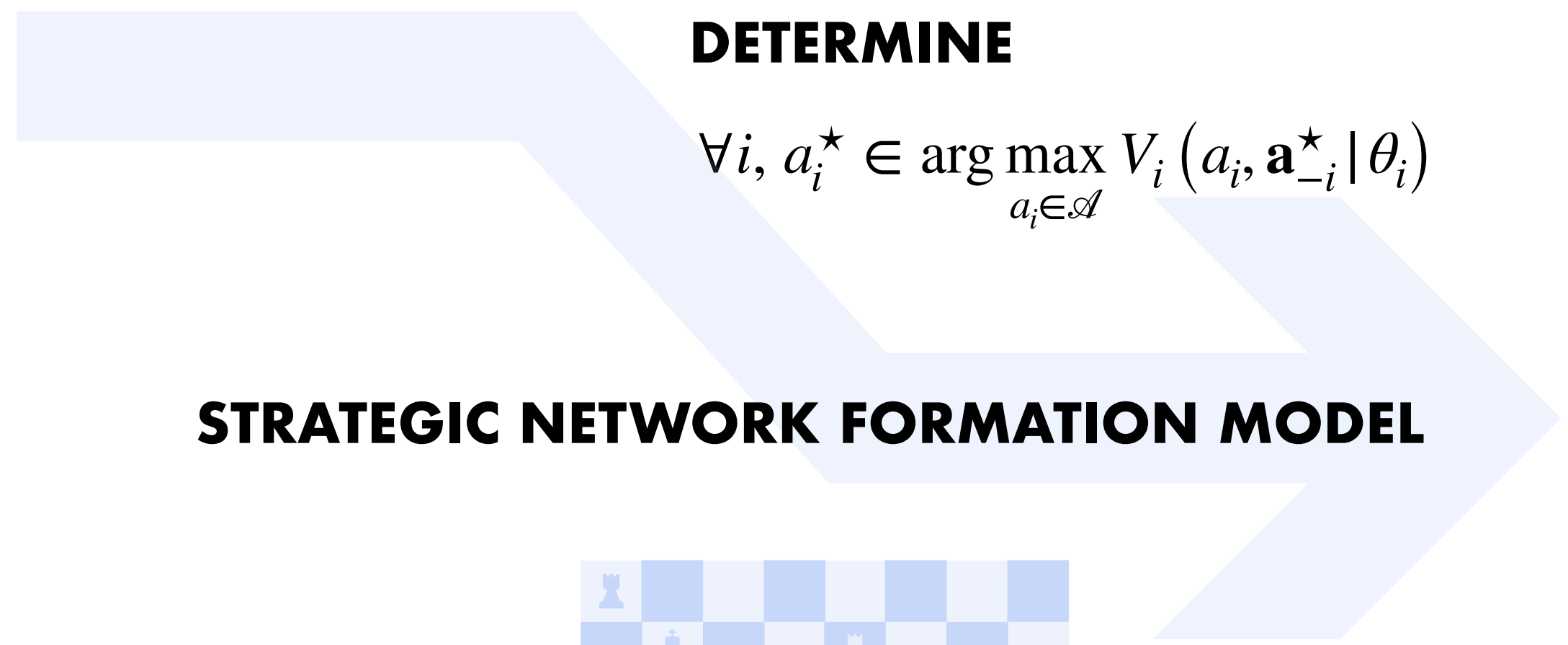
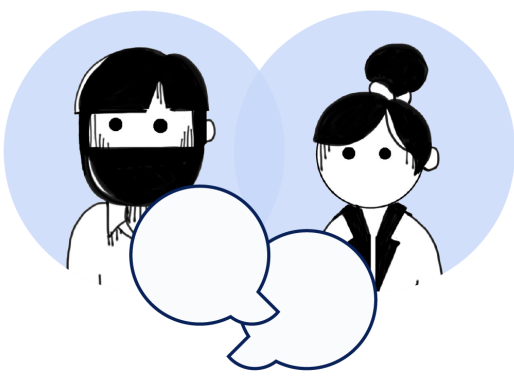
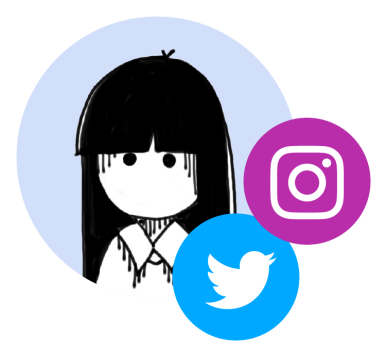
$$\forall i, \theta_i^* \text{ s.t. } V_i(a_i, \mathbf{a}_{-i}^*, \theta_i^*) \leq V_i(a_i^*, \mathbf{a}_{-i}^*, \theta_i^*), \forall a_i \in \mathcal{A}$$



SOCIAL NETWORK STRUCTURE $\mathcal{G}^*(\theta_i)$

Question: Given \mathcal{G}^* , for which θ_i is \mathcal{G}^* in equilibrium?

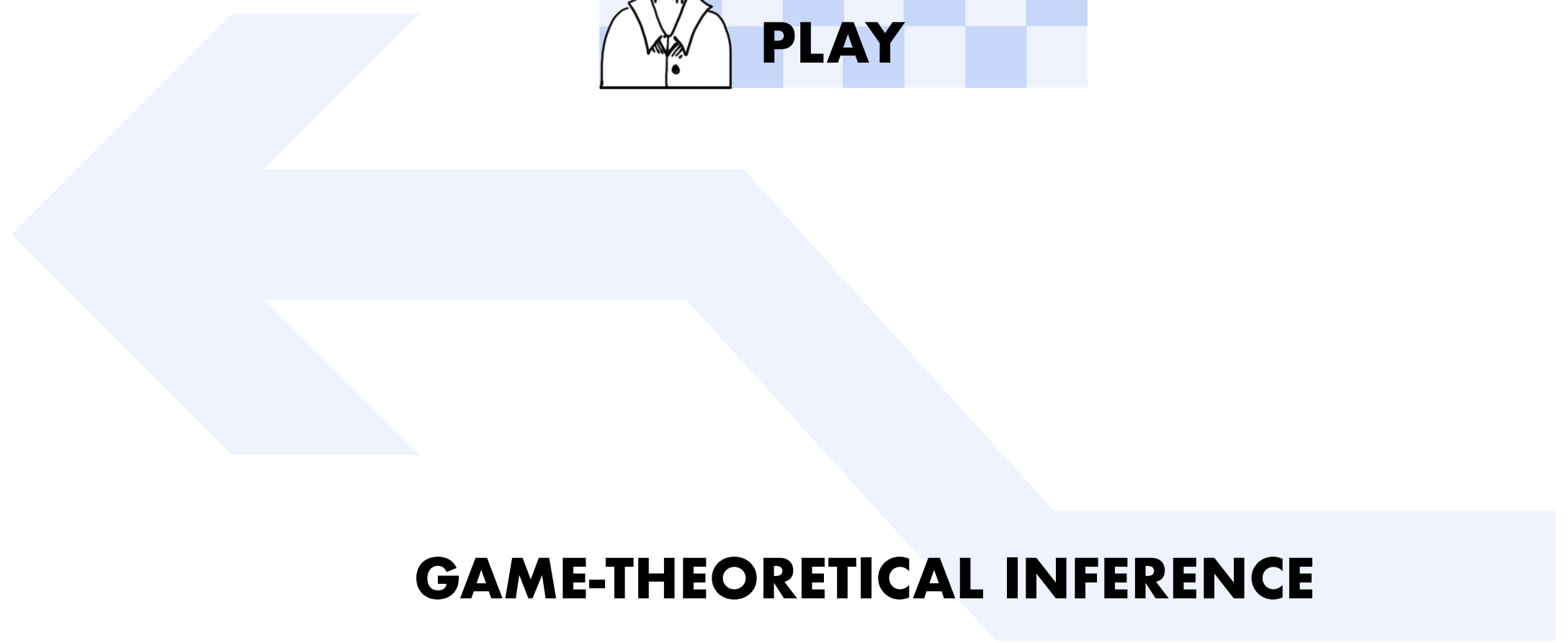
INDIVIDUAL BEHAVIOUR θ_i



DETERMINE

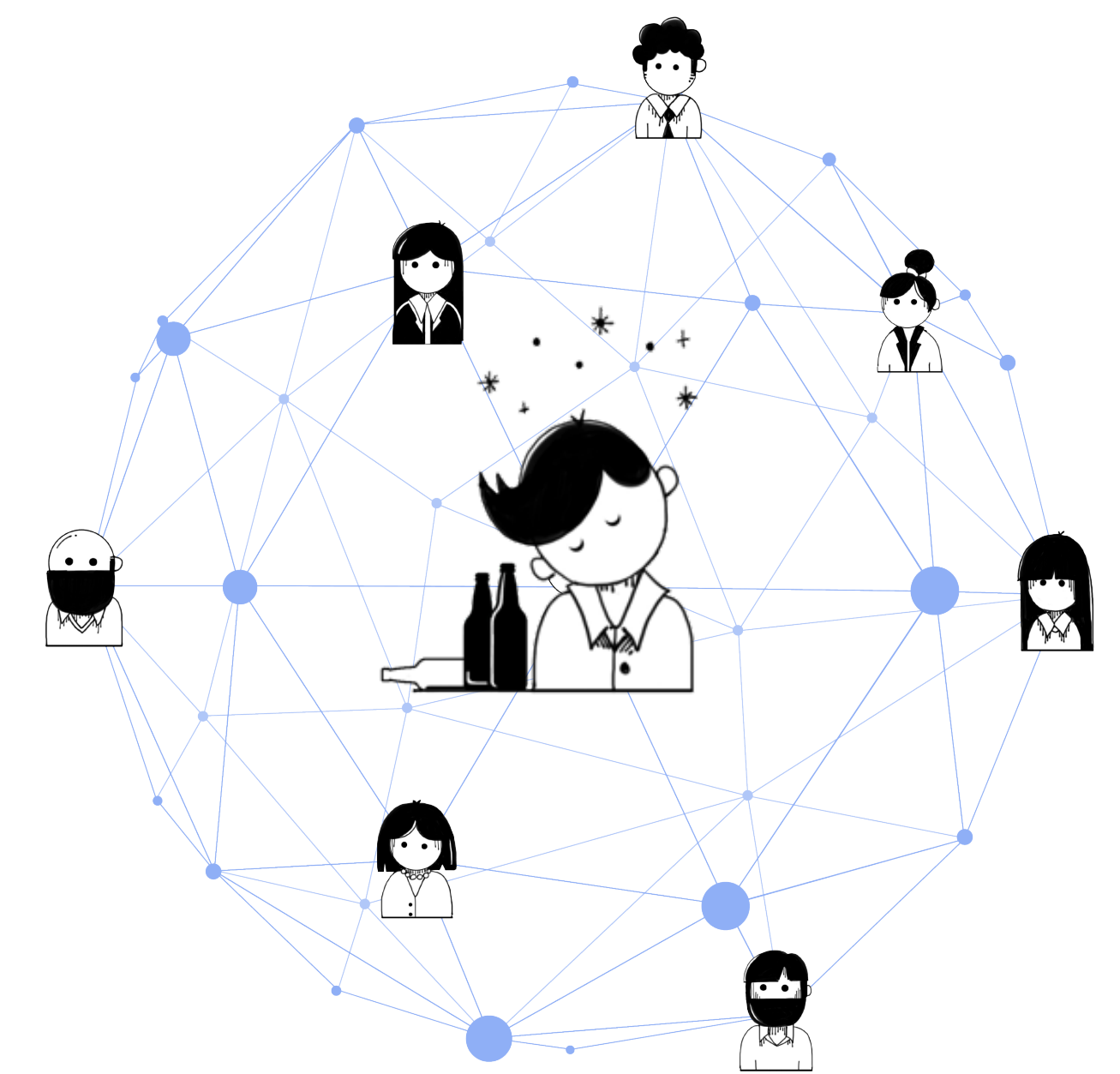
$$\forall i, a_i^* \in \arg \max_{a_i \in \mathcal{A}} V_i(a_i, \mathbf{a}_{-i}^* | \theta_i)$$

STRATEGIC NETWORK FORMATION MODEL



GAME-THEORETICAL INFERENCE

θ_i^* providing the **most rational** explanation to NE



SOCIAL NETWORK STRUCTURE $\mathcal{G}^*(\theta_i)$

HOMOGENEOUS RATIONAL AGENTS

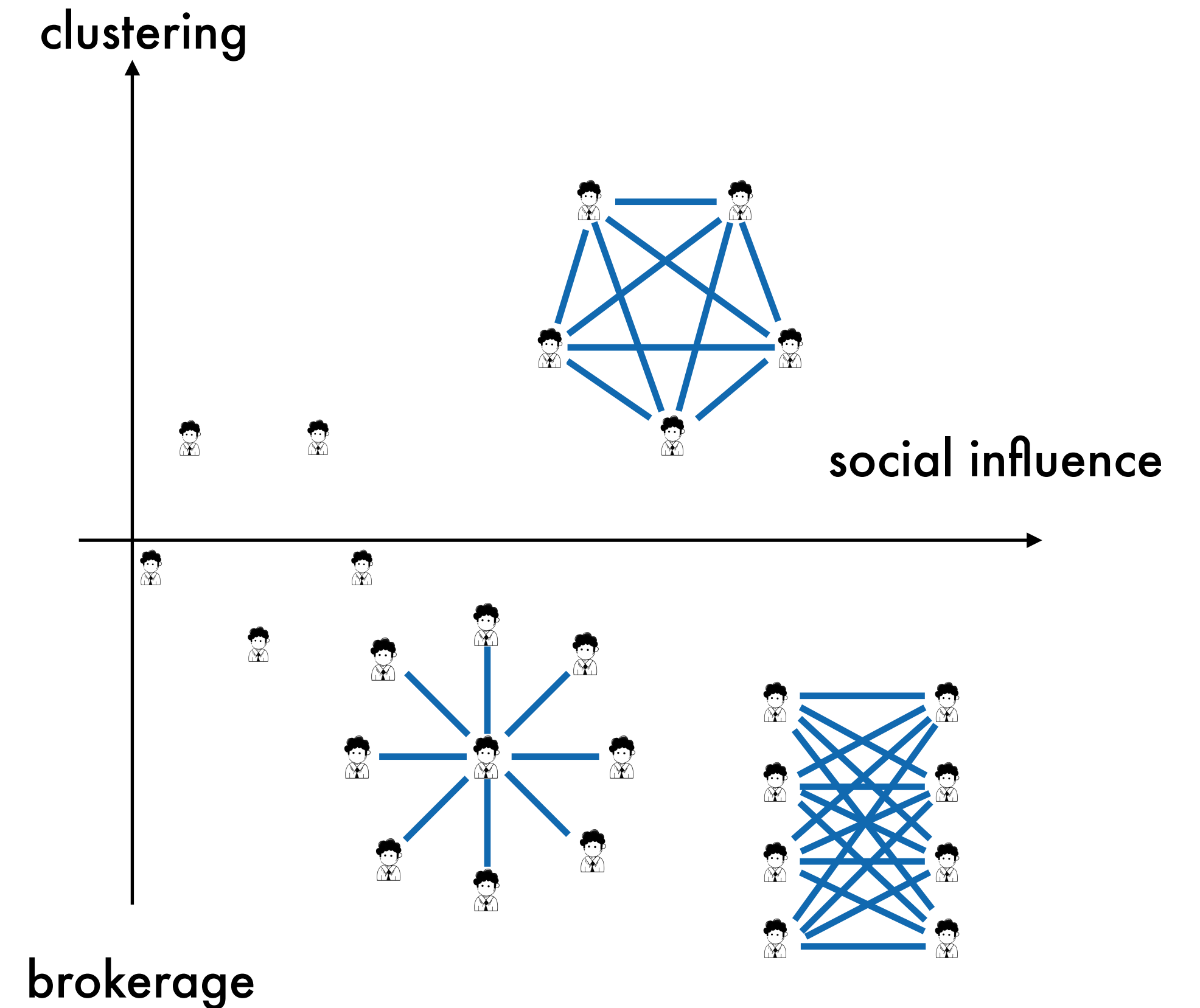
Assumption.

Individual preferences $\theta_i = \theta$, for all agents i , and fully rational behaviour.

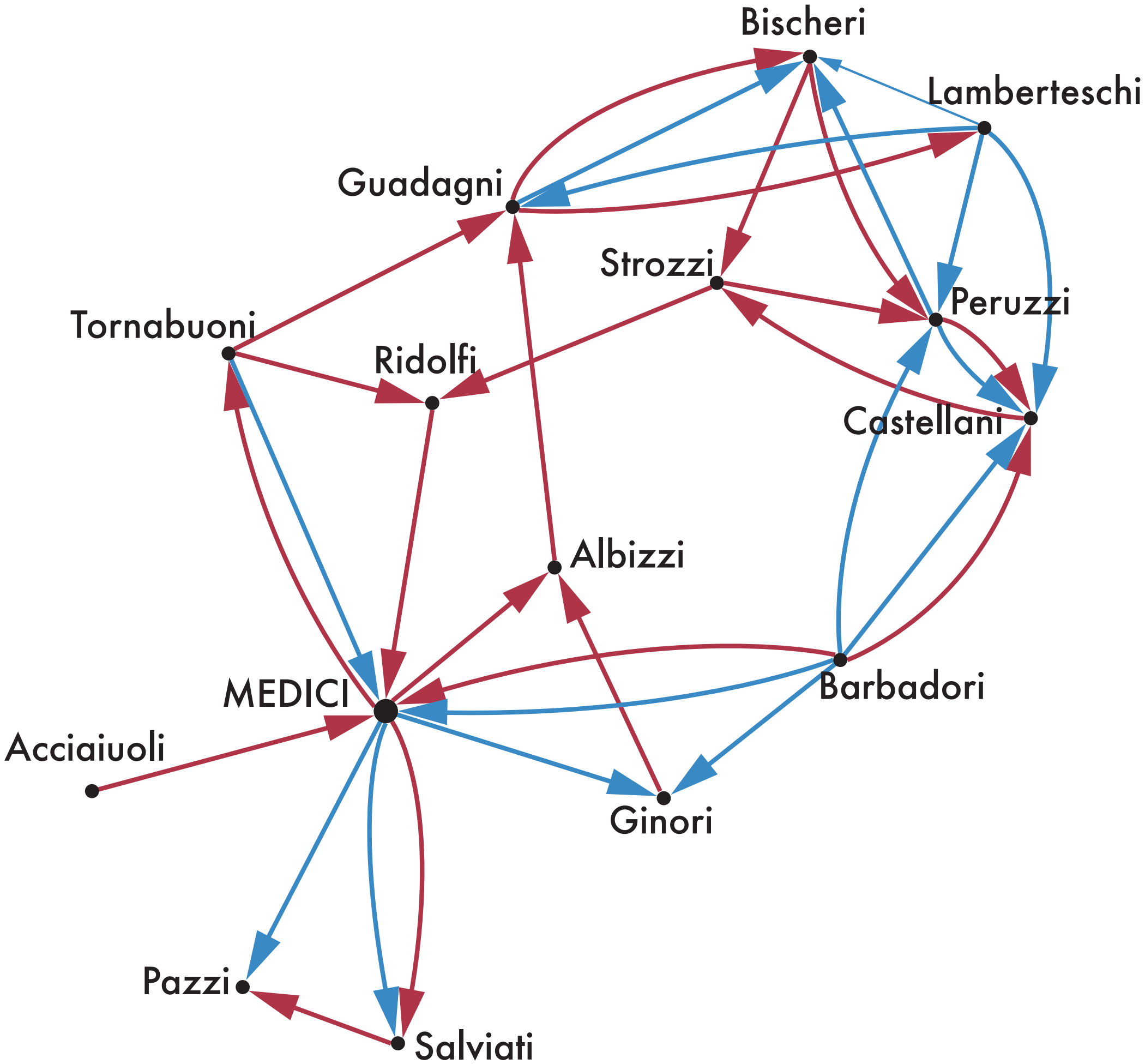
For specific **network motifs**, analytical parametric necessary conditions can be derived through **Variational Inequality**. Sufficiency can also be established.

Confirm known results in Strategic Network Formation literature

IFF conditions, parameter space analysis, NE and Pairwise Nash Equilibrium results.



REAL WORLD NETWORKS?



INVERSE OPTIMIZATION PROBLEM

Error function.

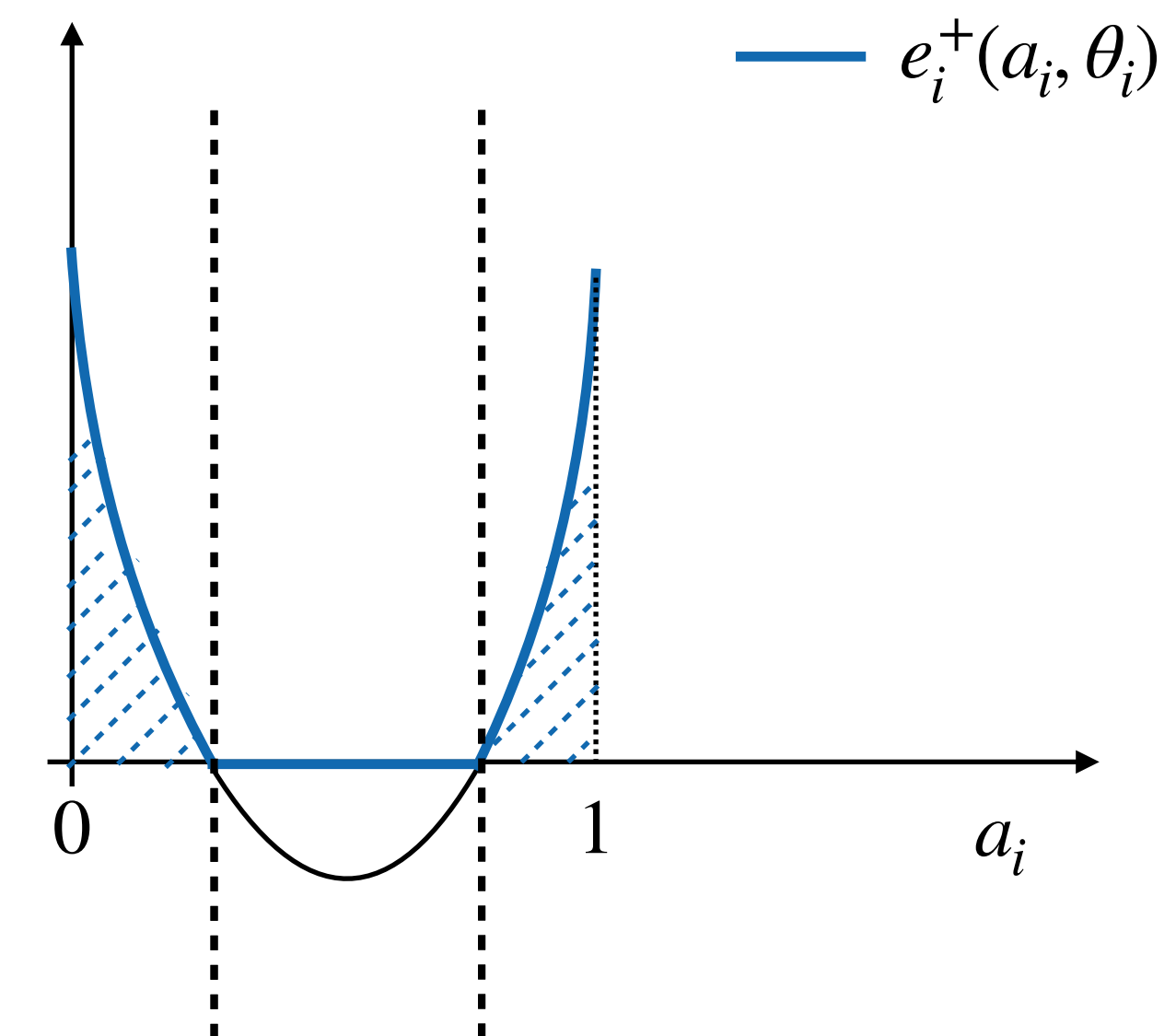
$$e_i(a_i, \theta_i) := V_i(a_i, \theta_i | \mathbf{a}_{-i}^*) - V_i(a_i^*, \theta_i | \mathbf{a}_{-i}^*)$$

$e_i^+(a_i, \theta_i) := \max\{0, e_i(a_i, \theta_i)\} > 0$ corresponds to a violation of the Nash equilibrium condition



Distance function.

$$d_i(\theta_i) := \left(\int_{\mathcal{A}} e_i^+(a_i, \theta_i)^2 da_i \right)^{1/2} = \|e_i^+(a_i, \theta_i)\|_{L_2(\mathcal{A})}$$



$$e_i(a_i, \theta_i) < 0$$

No violations: can be neglected

INVERSE OPTIMIZATION PROBLEM

Problem [Minimum NE-Distance Problem].

Given a network \mathcal{G}^* of N agents, for all agents i find the vectors of preferences θ_i^* such that

$$\theta_i^* \in \arg \min_{\theta_i \in \Theta} d_i^2(\theta_i)$$

Theorem [Convexity of the objective function].

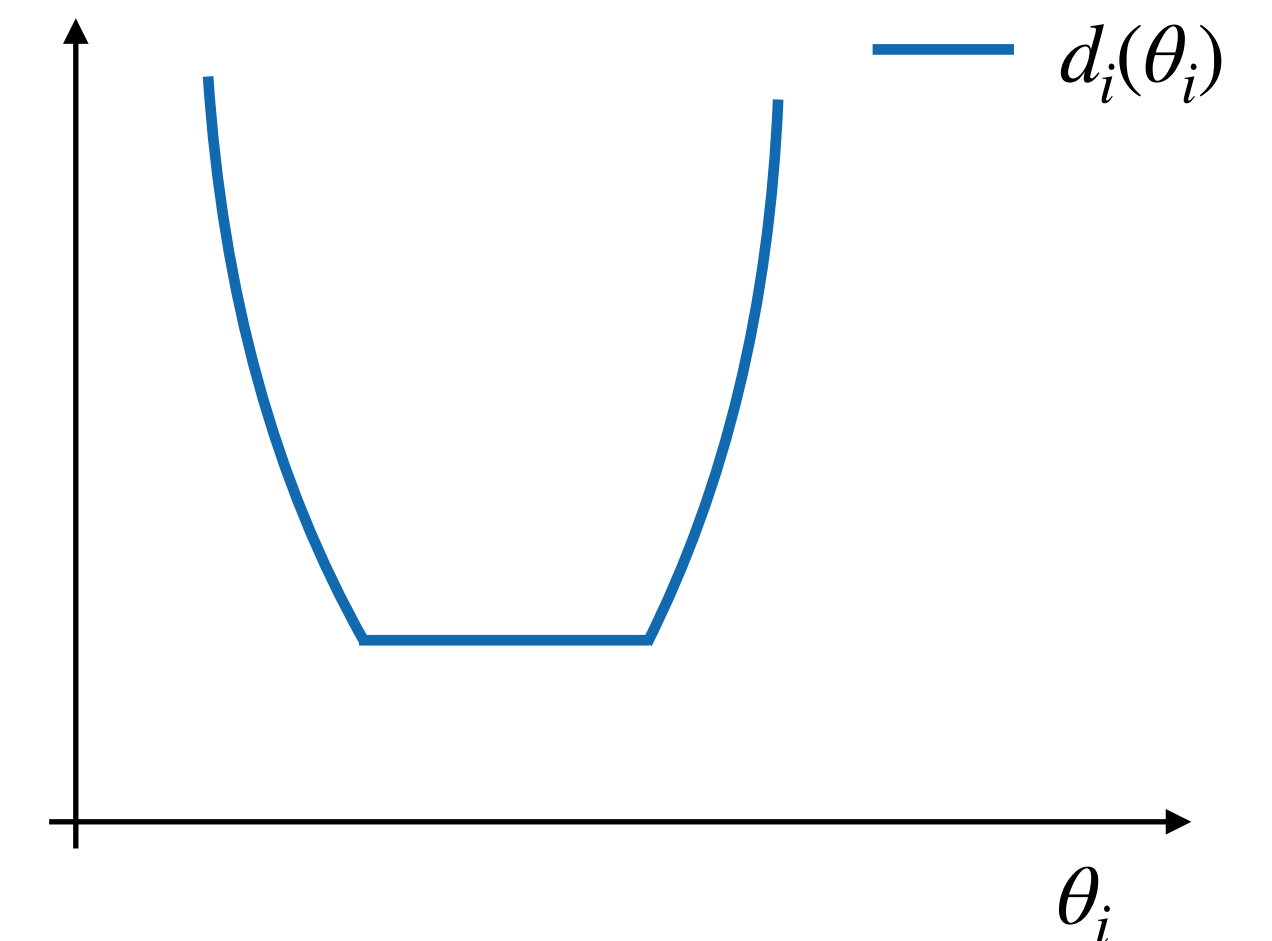
Let $e_i(a_i, \theta_i) : \mathbb{R}^n \times \mathbb{R}^p \rightarrow \mathbb{R}$ be a continuous function of $a_i \in \mathbb{R}^n$ and $\theta_i \in \mathbb{R}^p$, and linear in θ_i , and let \mathcal{A} be a compact subset of \mathbb{R}^n . Consider the squared distance function:

$$d_i^2(\theta_i) := \int_{\mathcal{A}} \left(\max \{0, e_i(a_i, \theta_i)\} \right)^2 dx = \|e_i^+(a_i, \theta_i)\|_{L_2(\mathcal{A})}^2$$

Then $d_i^2(\theta_i)$ is continuously differentiable, and its gradient reads as

$$\nabla_{\theta} d_i^2(\theta) = \int_{\mathcal{A}} 2 \nabla_{\theta_i} (e_i(a_i, \theta_i)) \max \{0, e_i(a_i, \theta_i)\} da_i.$$

Moreover, d_i^2 is convex.



INVERSE OPTIMIZATION PROBLEM - SOLUTION

First-order optimality condition

$$0 = \nabla_{\theta_i}(d_i^2(\theta_i)) = 2 \int_{\mathcal{A}} \nabla_{\theta_i}(e_i(a_i, \theta_i)) \max \{0, e_i(a_i, \theta_i)\} da_i.$$

max operator within $(N - 1)$ - dimensional integral

INVERSE OPTIMIZATION PROBLEM - SOLUTION

Search for an approximate solution. Consider a finite set of possible actions (samples)

$$\left\{ a_i^j \right\}_{j=1}^{n_i} \subset \mathcal{A}$$

Let $e_i^j(\theta_i) = e_i(a_i^j, \theta_i)$ and $e_i^{j,+}(\theta_i) = e_i^+(a_i^j, \theta_i)$ be the corresponding error and positive error at the samples.

Approximate the distance function as

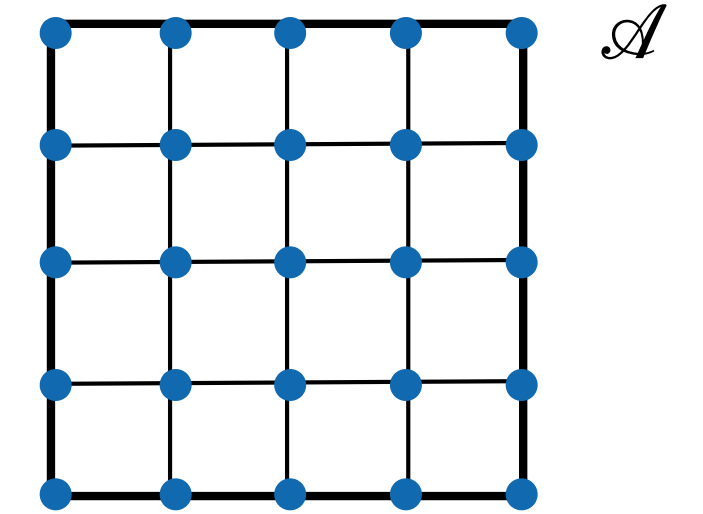
$$\tilde{d}_i(\theta_i) := \left(\sum_{j=1}^{n_i} \left(e_i^{j,+}(\theta_i) \right)^2 \right)^{1/2} = \|\mathbf{e}_i^+\|_2$$

Problem [Discrete Minimum NE-Distance Problem].

Given a network \mathcal{G}^* of N agents, for all agents i find the vectors of preferences $\hat{\theta}_i$ such that

$$\hat{\theta}_i \in \arg \min_{\theta_i \in \Theta} \tilde{d}_i^2(\theta_i)$$

Same property of the original problem (Convexity)



INVERSE OPTIMIZATION PROBLEM - SOLUTION

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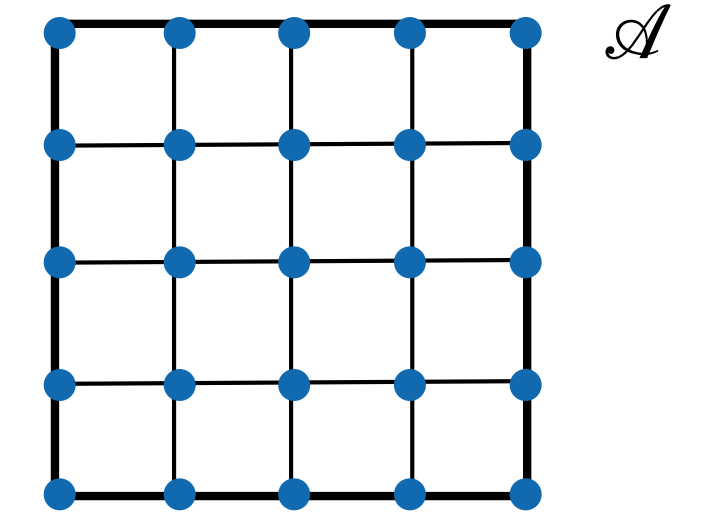
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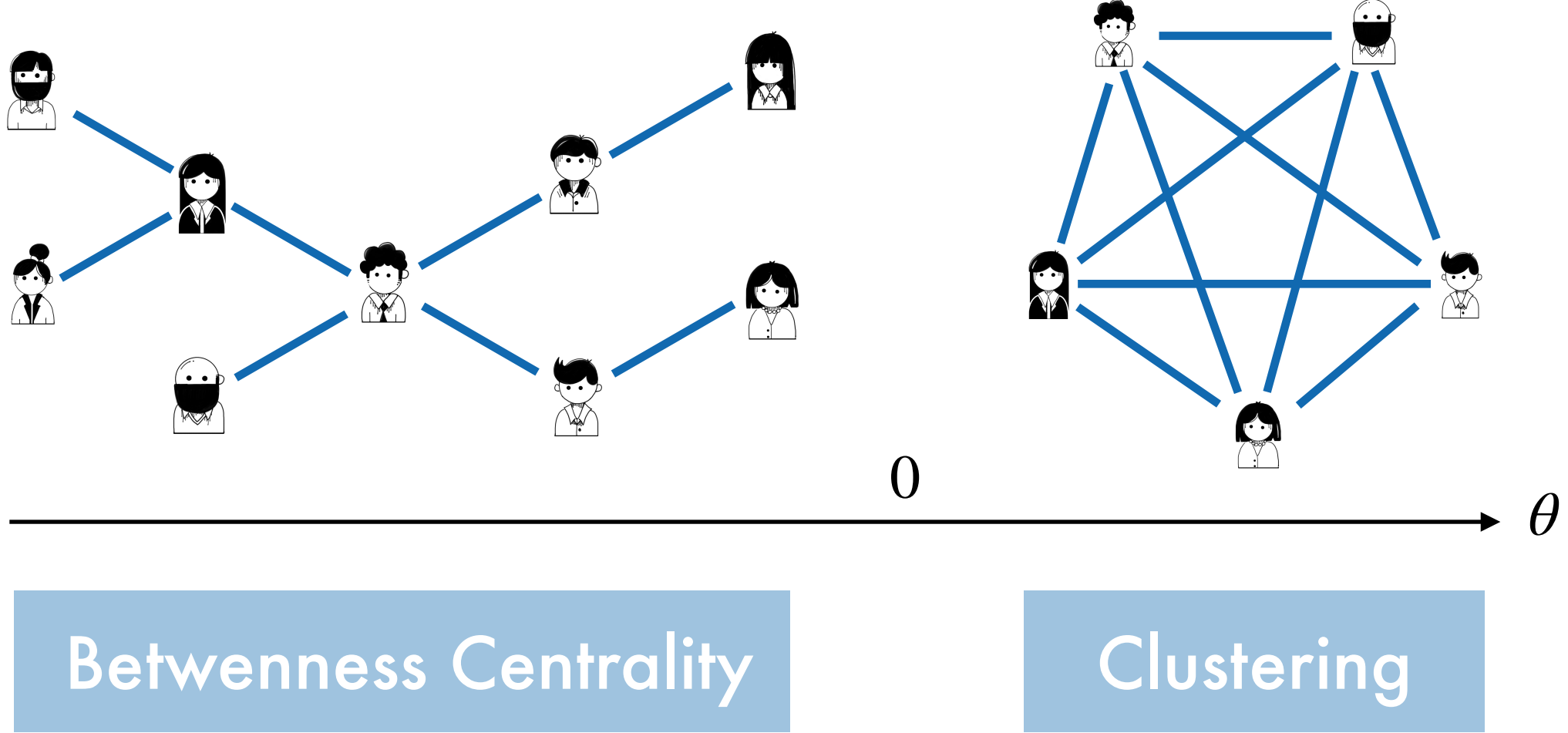
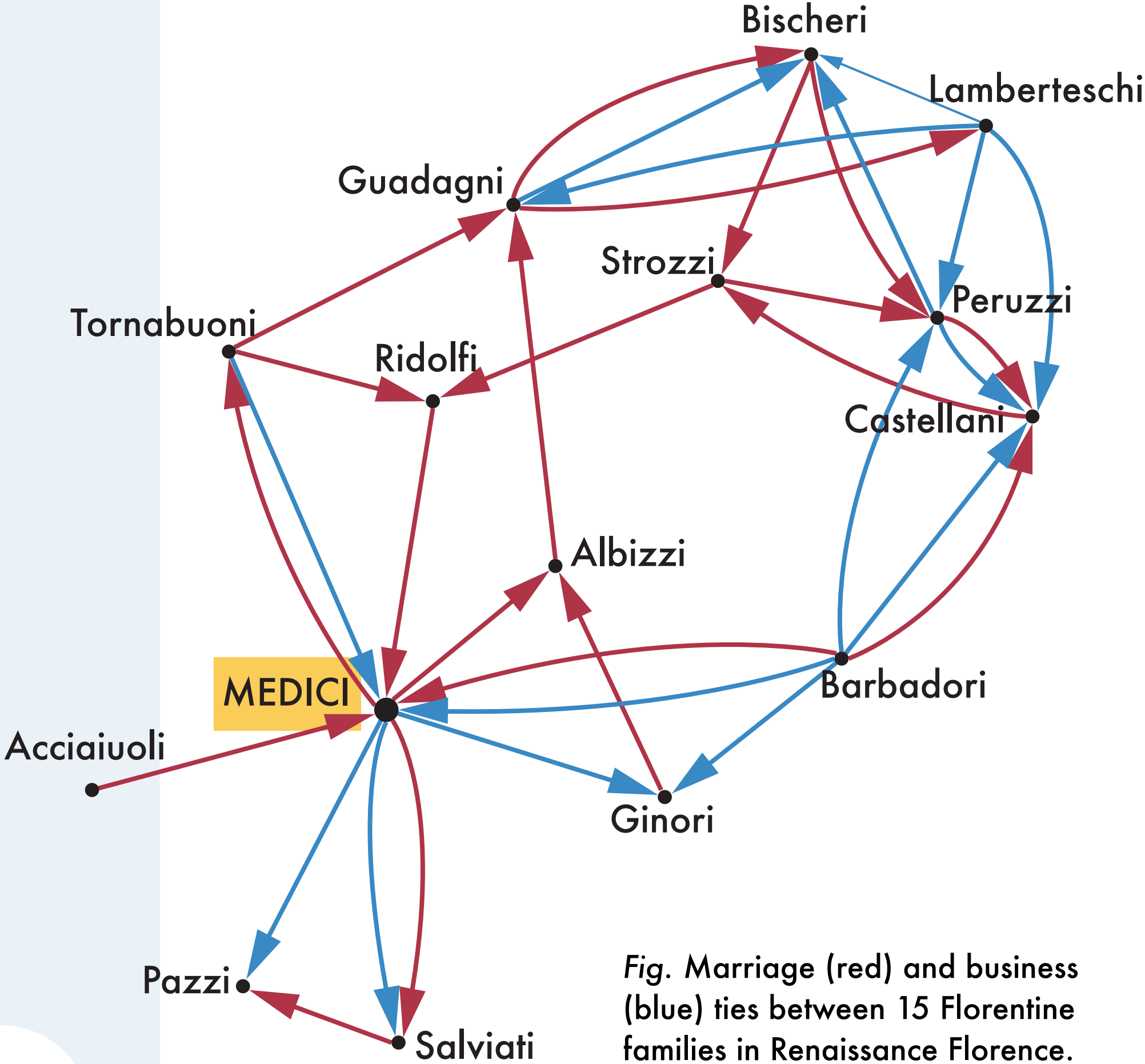
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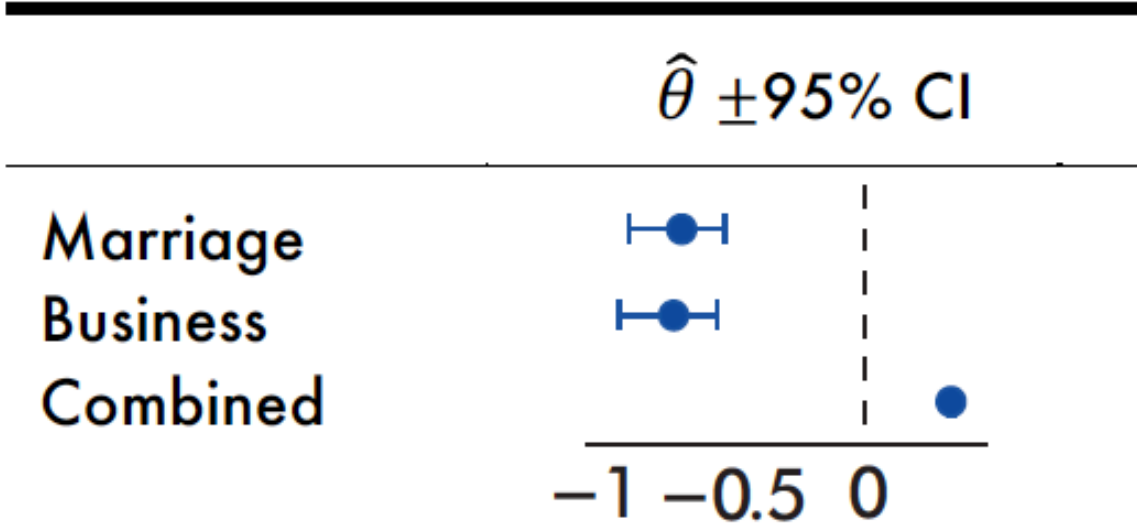


Similar to Generalized Least Square Regression

RENAISSANCE FLORENCE NETWORK



Behaviour estimation of the Medici family



Padgett, J. F., & Ansell, C. K. (1993). Robust Action and the Rise of the Medici, 1400-1434. *American journal of sociology*, 98(6), 1259-1319.

PIECE OF ART

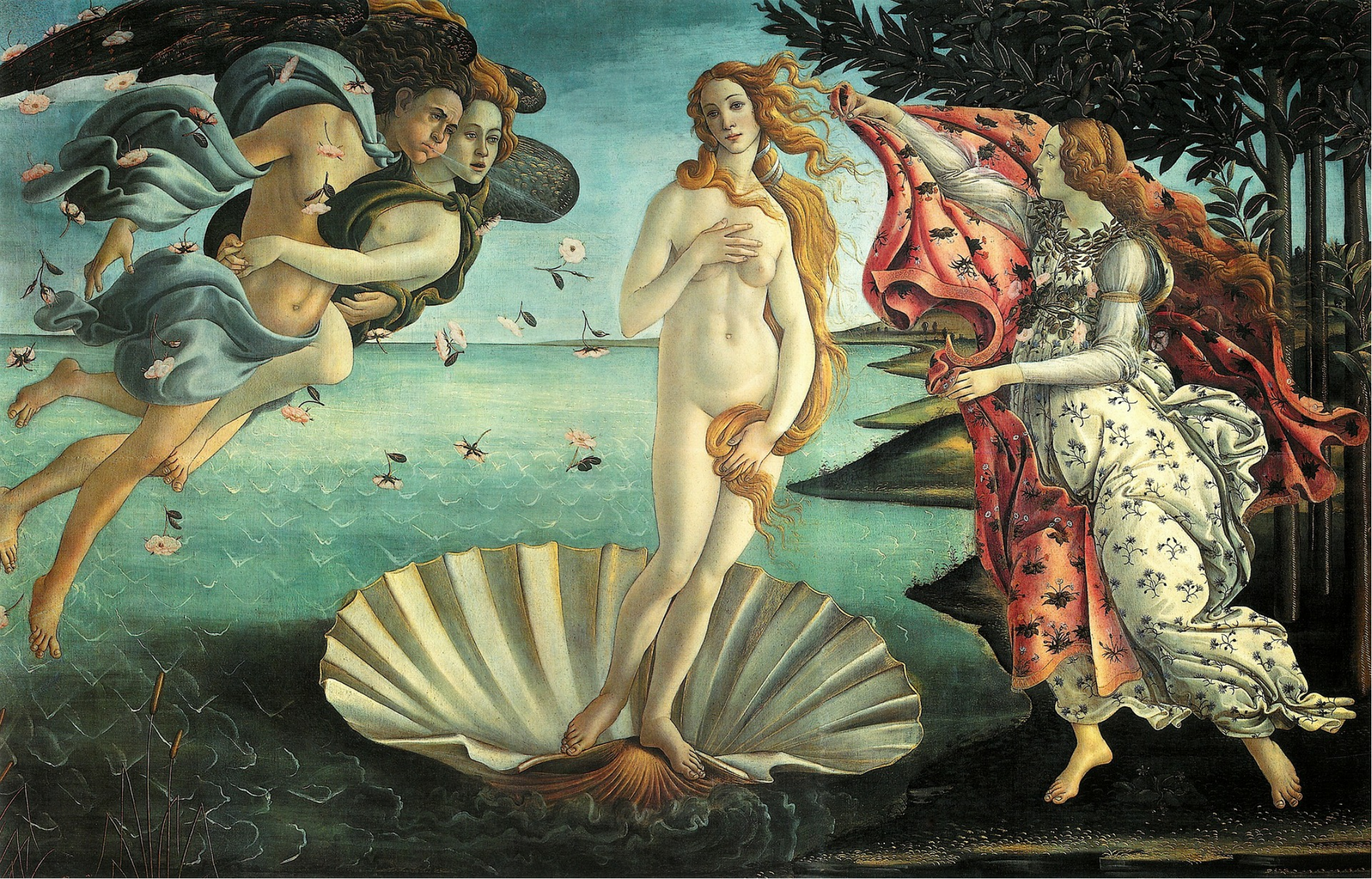
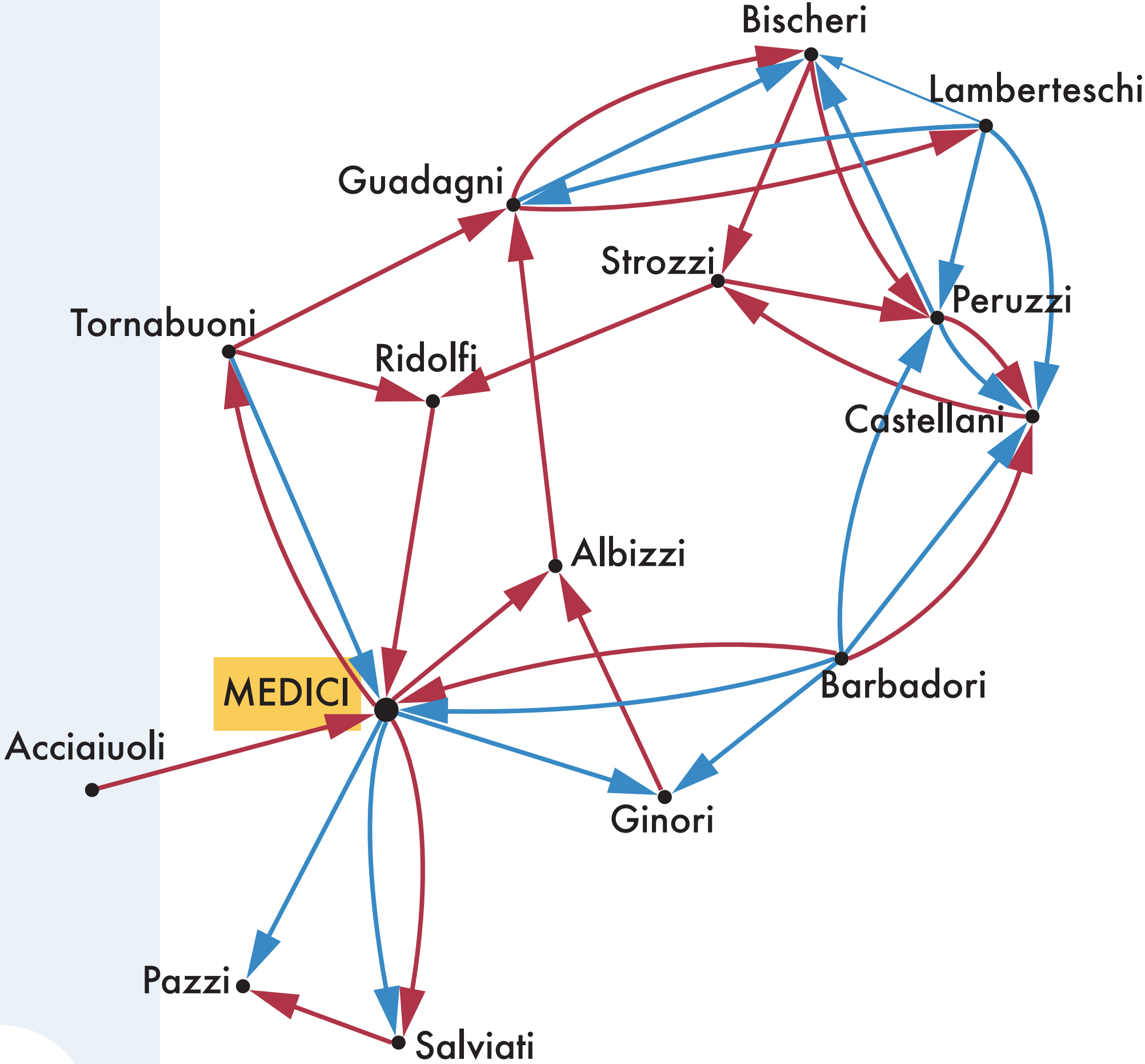
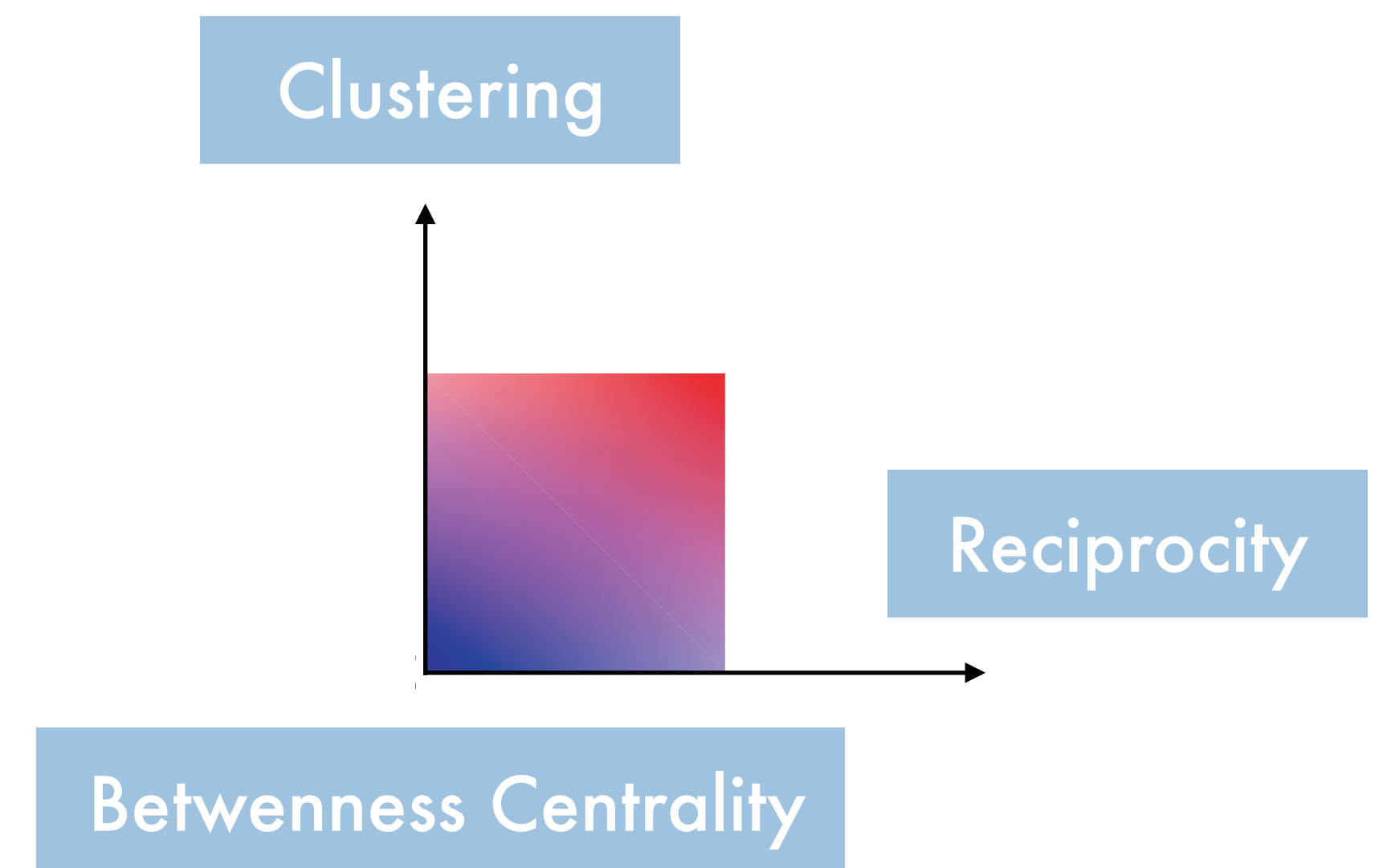
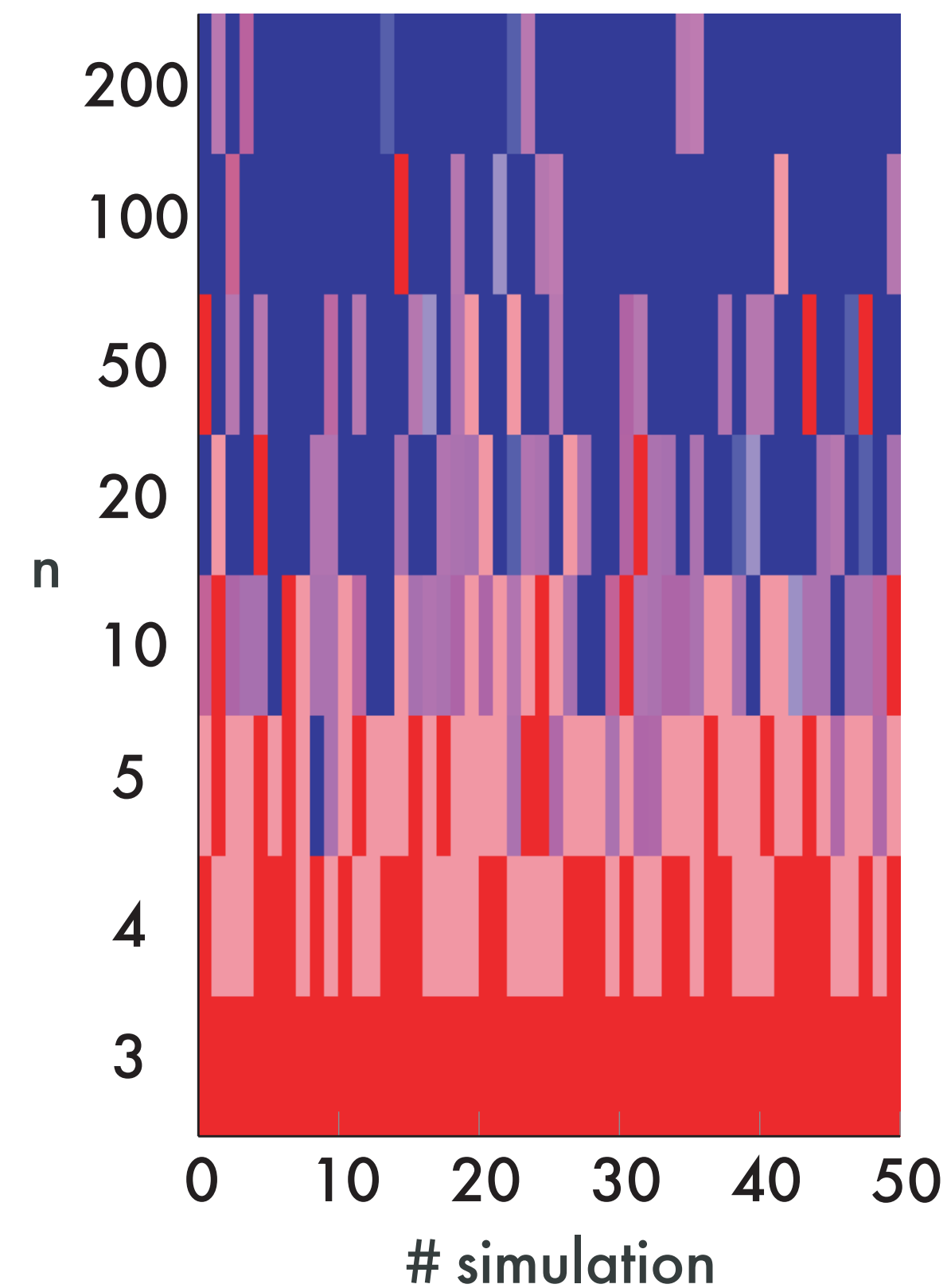


Fig. "La nascita di Venere", 1482-1485, Sandro Botticelli. Galleria degli Uffizi, Firenze.

PREFERENTIAL ATTACHMENT MODEL

Nodes are introduced sequentially.

Each newborn **receives 2 incoming links** from existing nodes (randomly selected, proportionally to the outdegree), and **creates 2 outgoing ties** to existing nodes (randomly selected, proportionally to the indegree).





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