



# 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  
FLORIAN DÖRFLER

AUTOMATIC  
CONTROL  
LABORATORY 

**ETH** zürich

# OBSERVATIONS

Actors decide with whom they want to interact.



01

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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



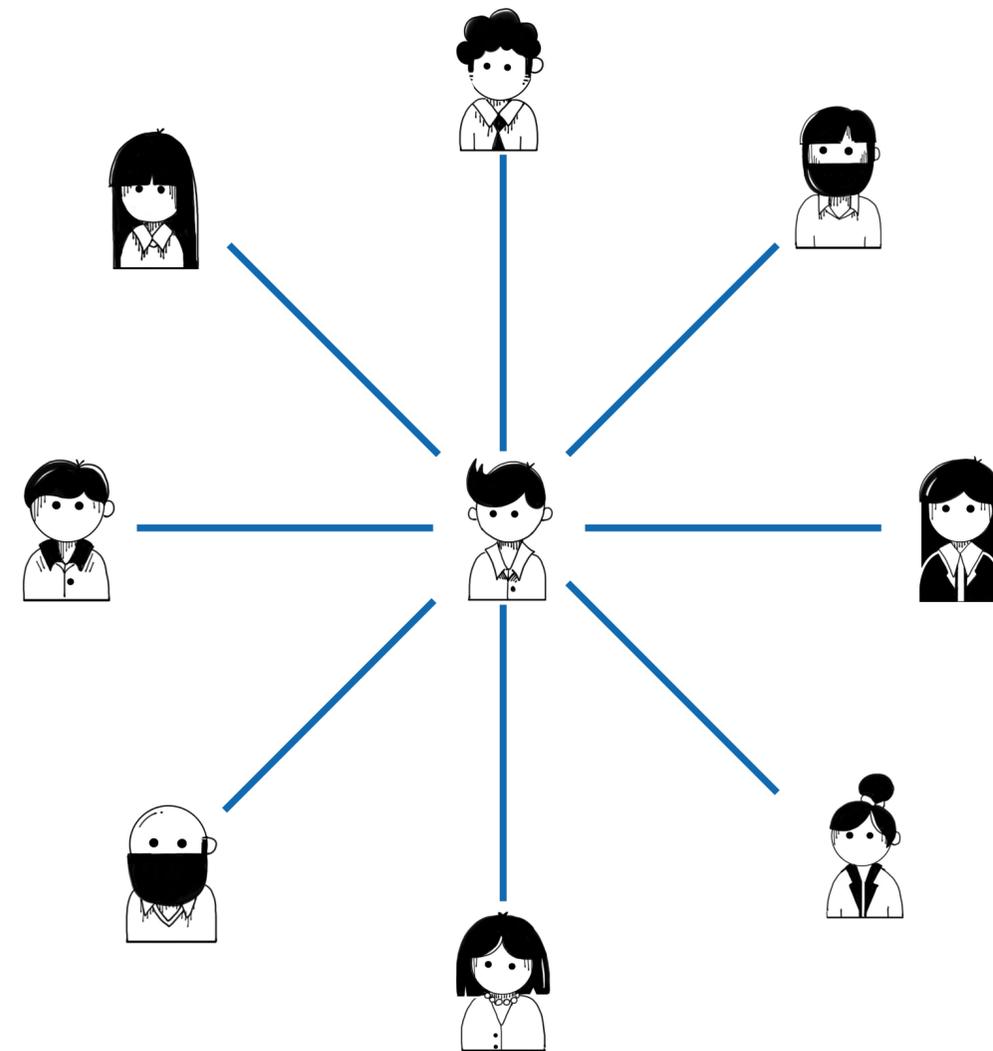
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"It's not what you know, it's who you know" is one of those phrases

# SOCIAL NETWORK POSITIONS' BENEFITS

## Social Influence

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



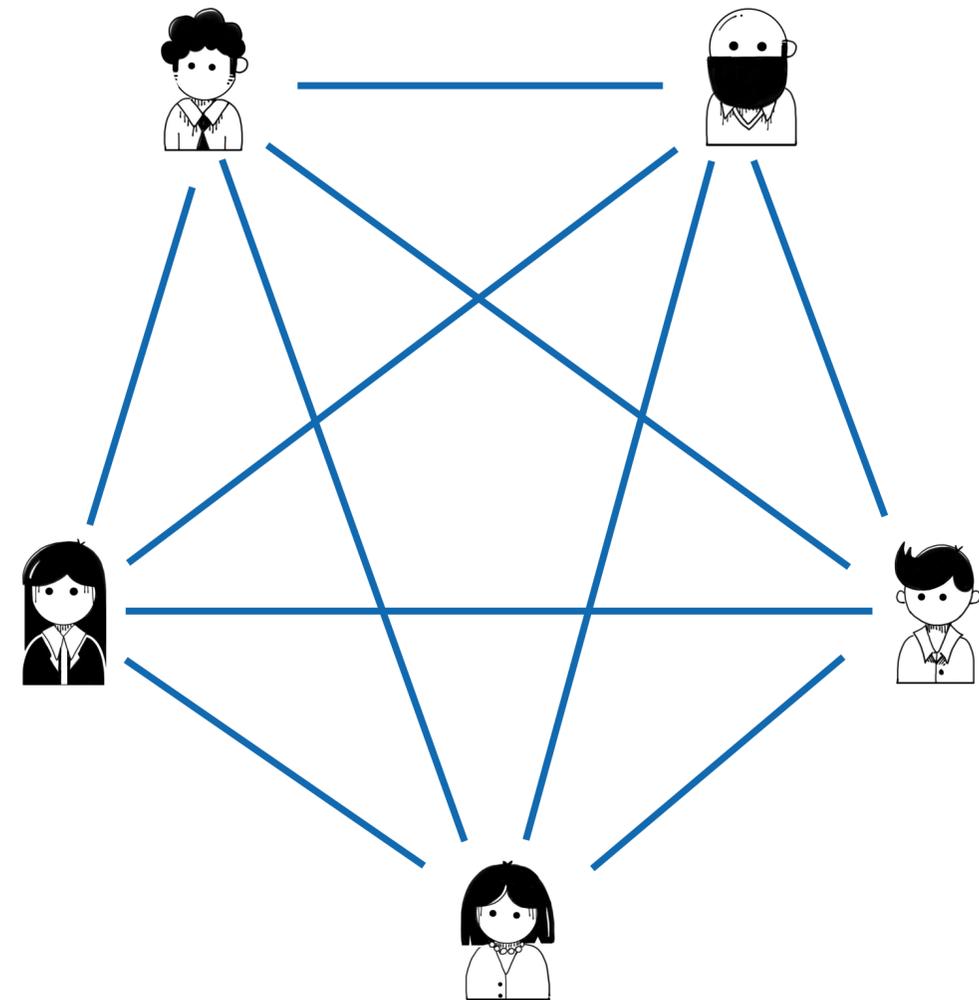
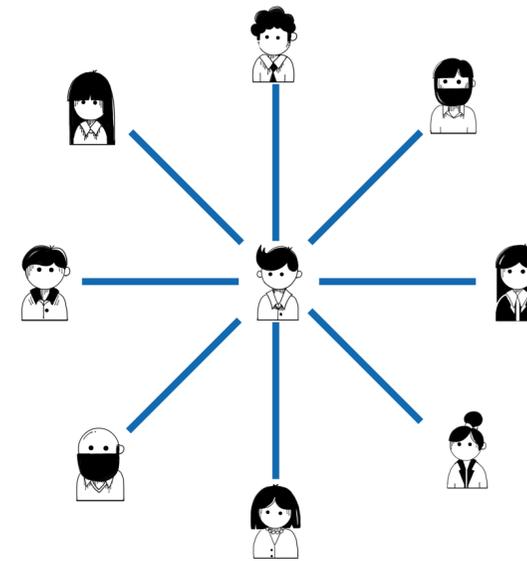
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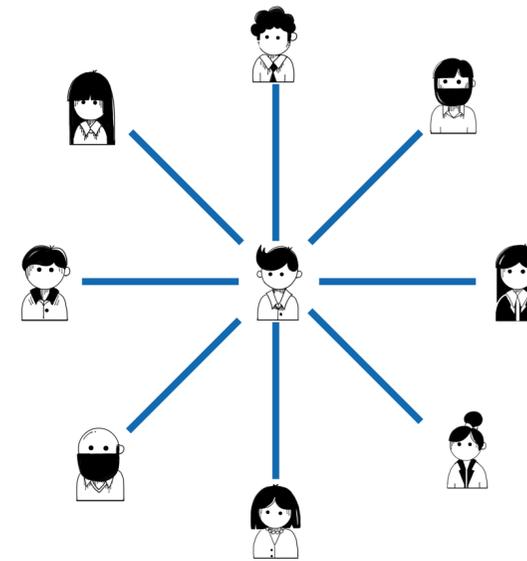
The more our friends' friends are our friends, the safer we feel.



# SOCIAL NETWORK POSITIONS' BENEFITS

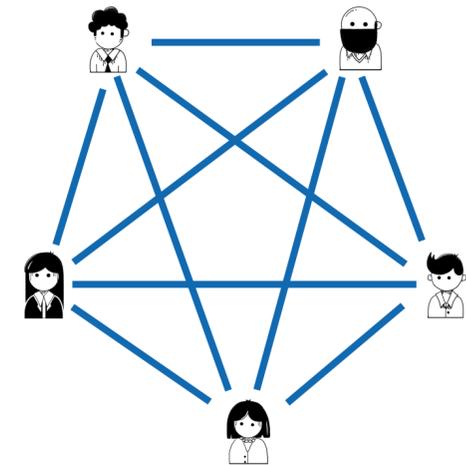
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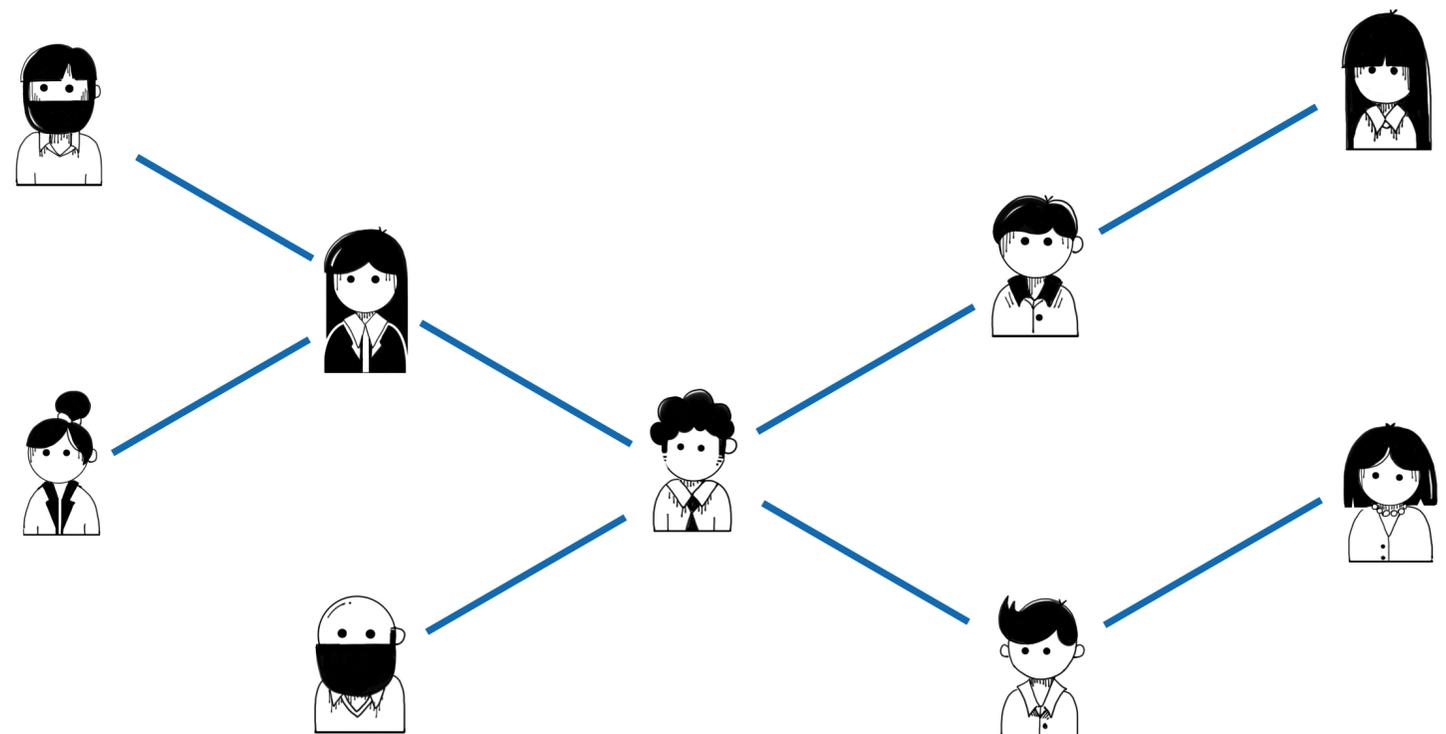
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## Brokerage

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



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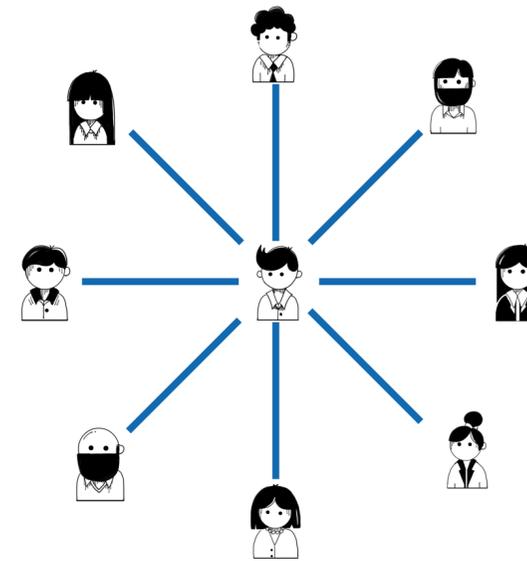
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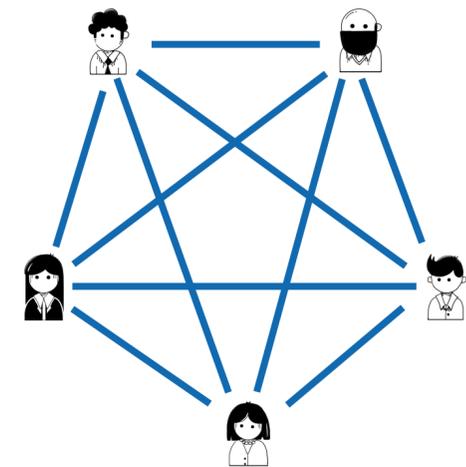
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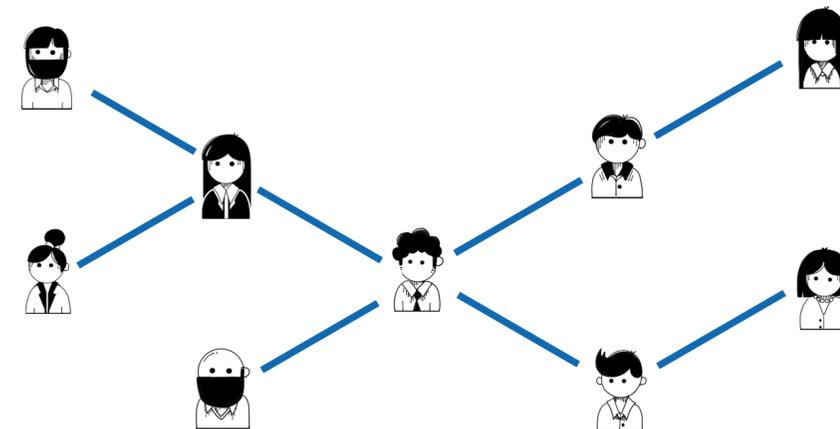
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Degree Centrality



Clustering coefficient



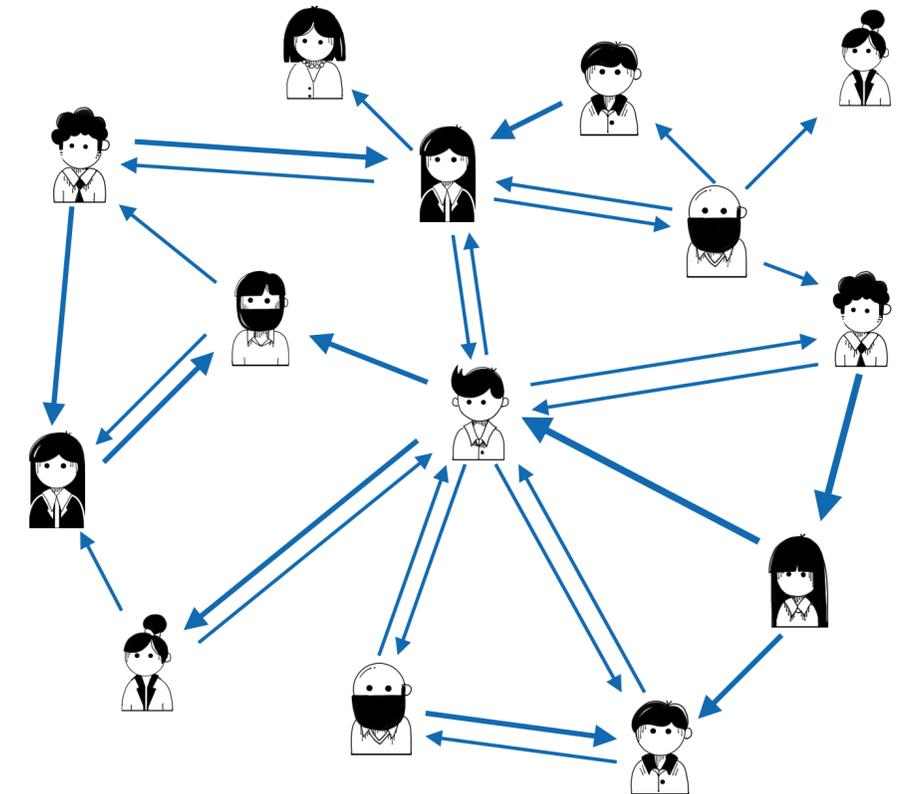
Betweenness Centrality

# 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

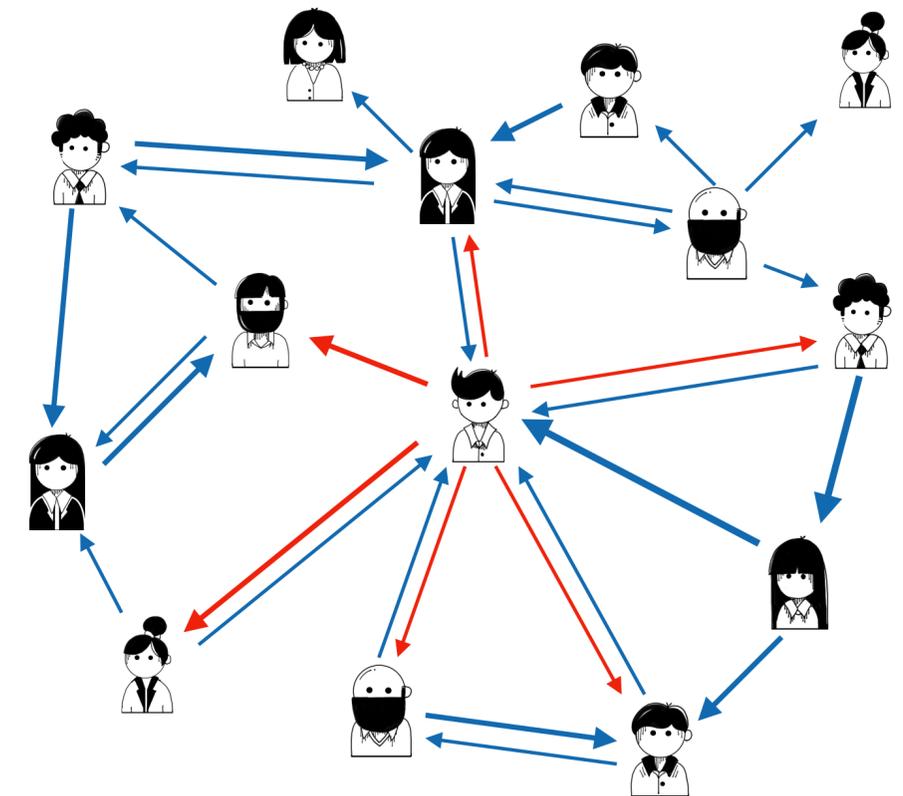
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A typical action of agent  $i$  is:

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



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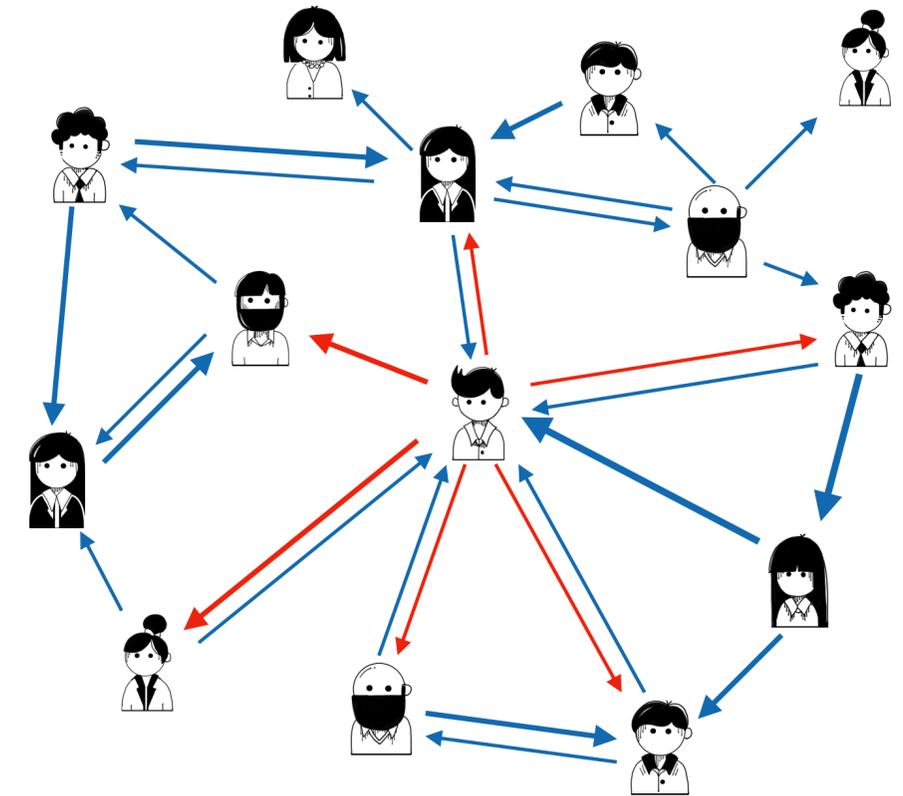
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Every agent  $i$  is endowed with a payoff function  $V_i$  and is looking for

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

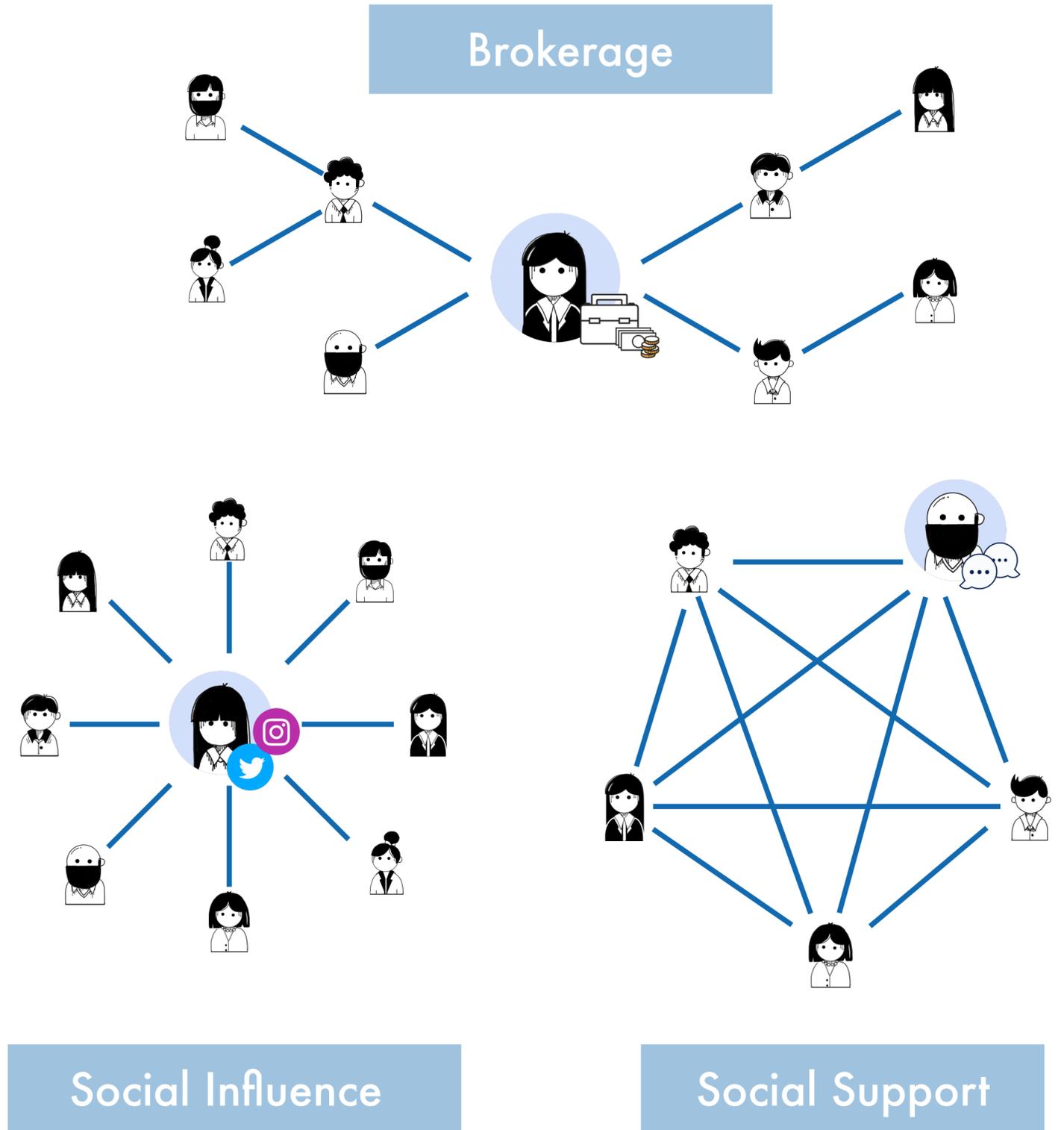


# SOCIAL NETWORK FORMATION MODEL

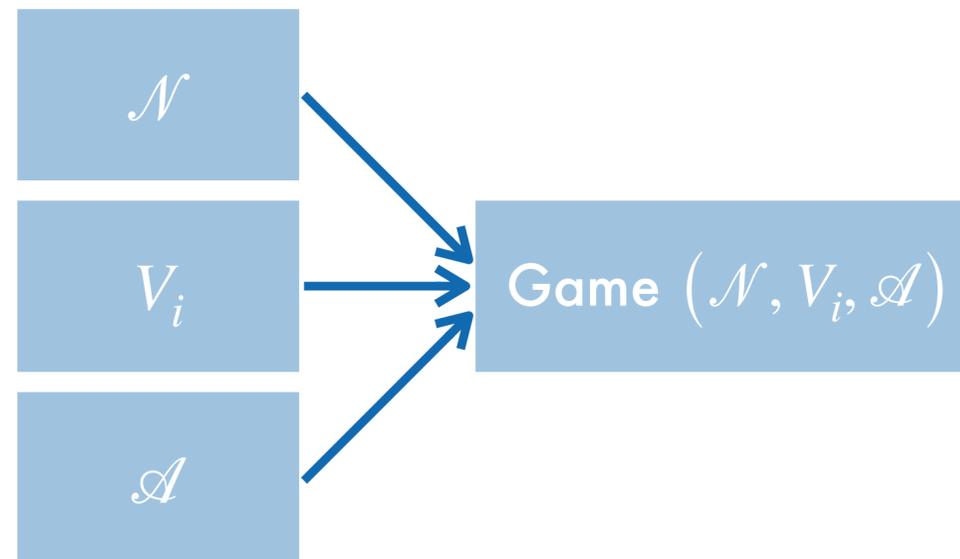
Parametric payoff function:

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

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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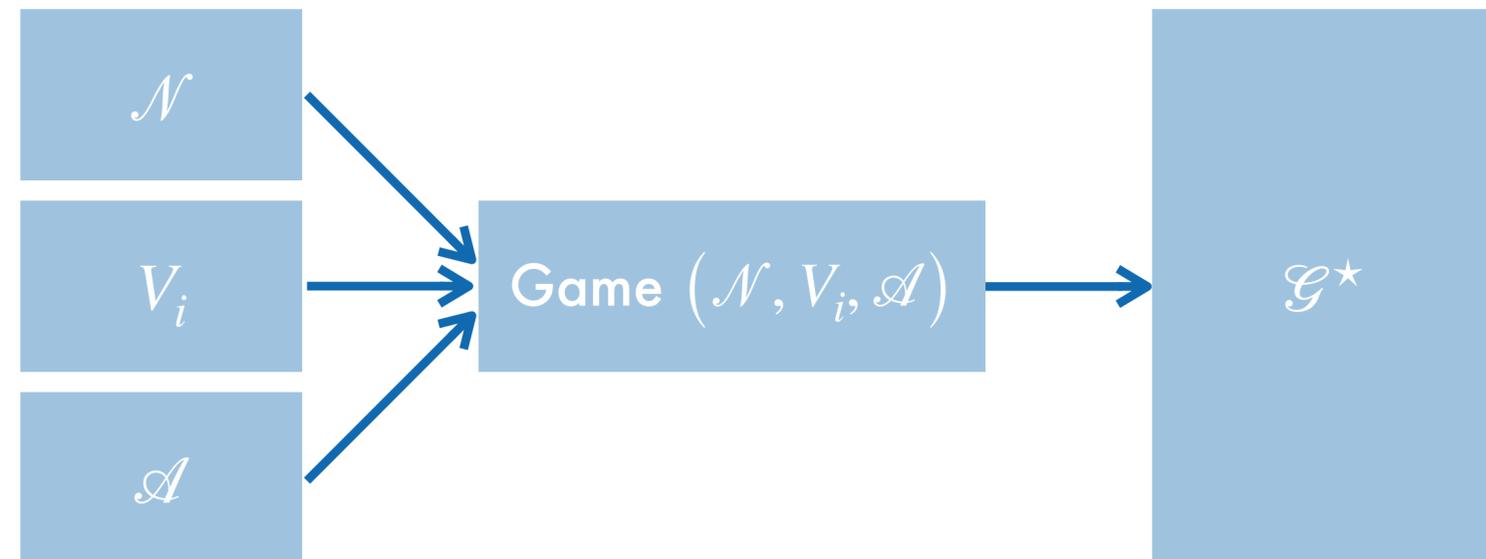


# NASH EQUILIBRIUM

*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

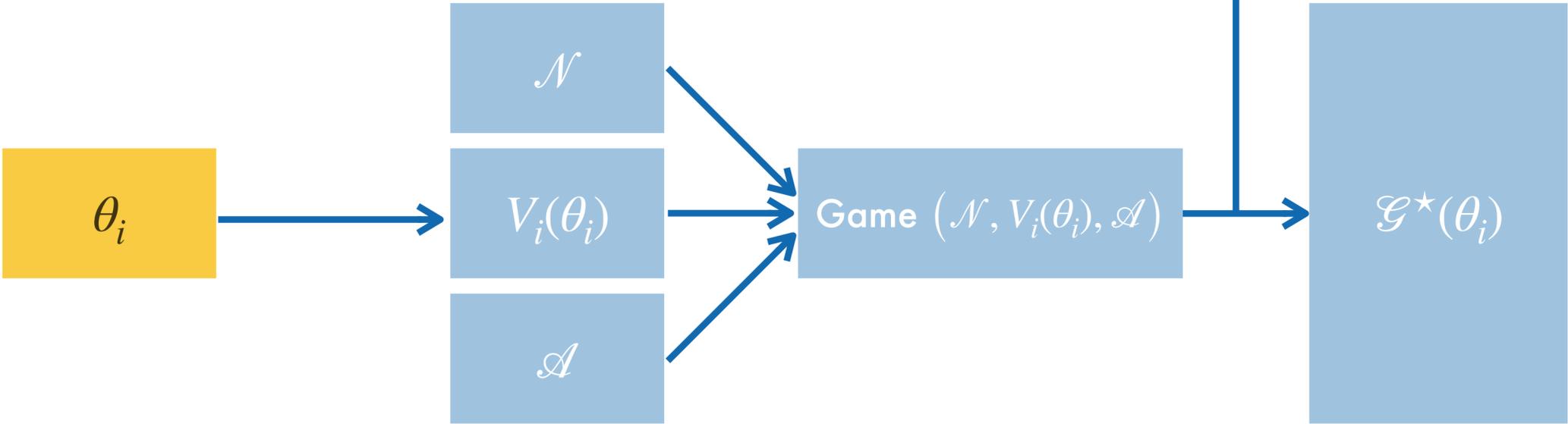
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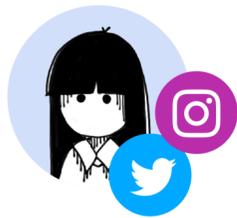
$$V_i(a_i, \mathbf{a}_{-i}^* | \theta_i) \leq V_i(a_i^*, \mathbf{a}_{-i}^* | \theta_i), \forall a_i \in \mathcal{A}$$

$\Rightarrow$

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



# INDIVIDUAL BEHAVIOUR $\theta_i$



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

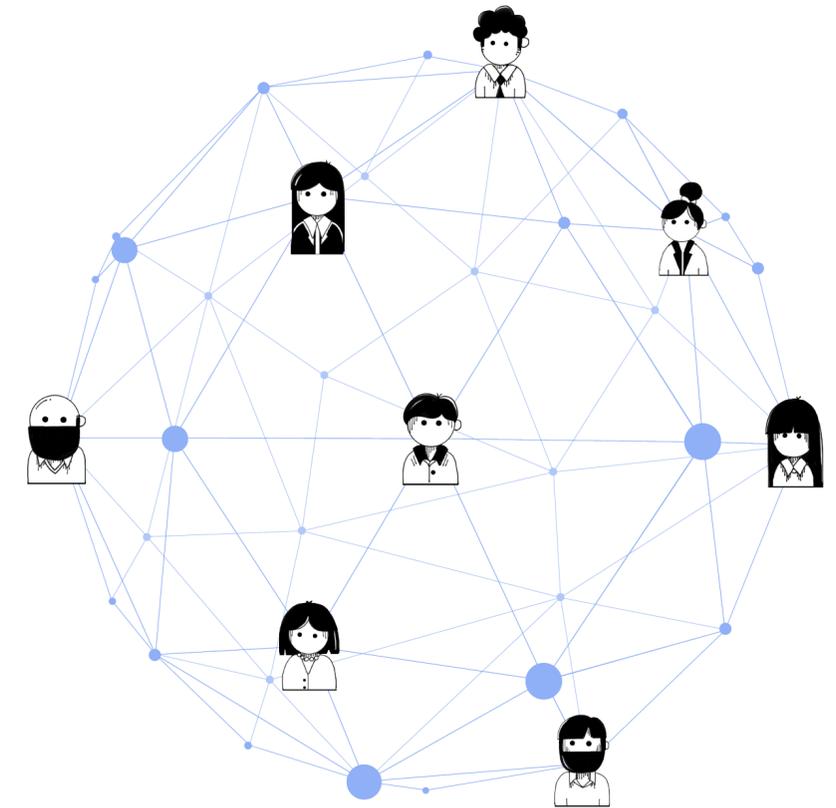
## DETERMINE

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

## STRATEGIC NETWORK FORMATION MODEL

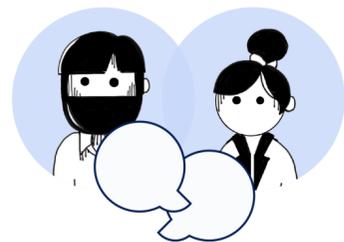
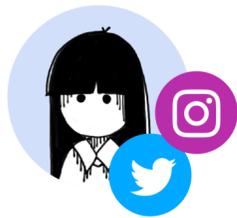


STRATEGIC PLAY

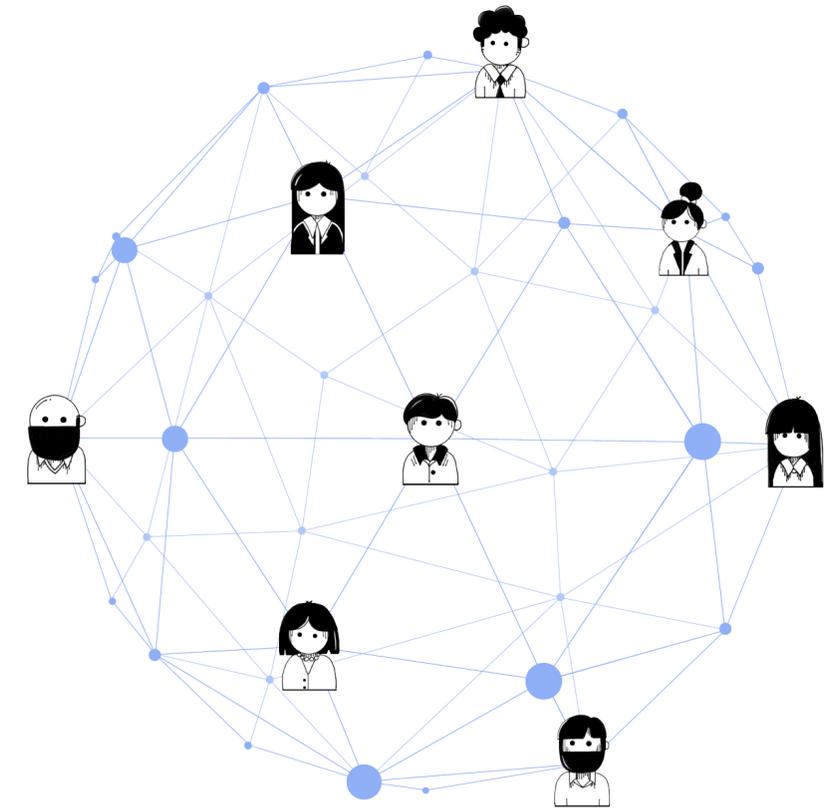


SOCIAL NETWORK STRUCTURE  $\mathcal{G}^*(\theta_i)$

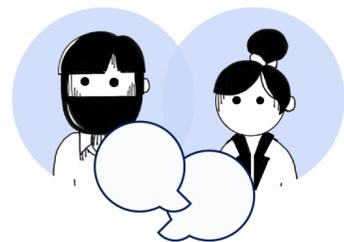
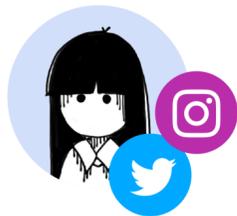
**INDIVIDUAL  
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Question: Given  $\mathcal{G}^*$ , for which  $\theta_i$  is  $\mathcal{G}^*$  in equilibrium?



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

## DETERMINE

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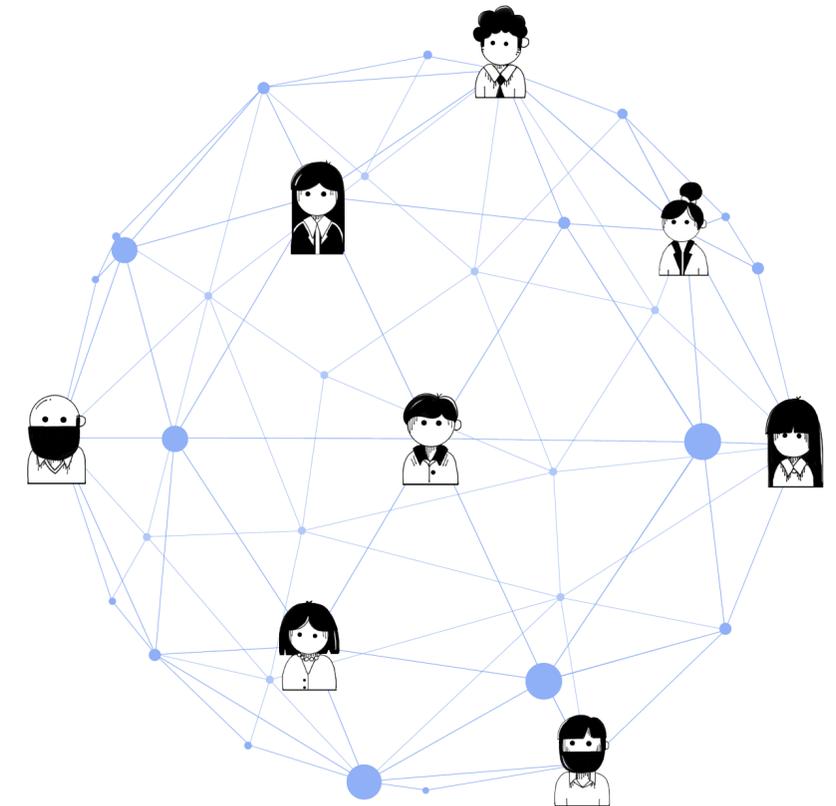
## STRATEGIC NETWORK FORMATION MODEL



STRATEGIC PLAY

## GAME-THEORETICAL INFERENCE

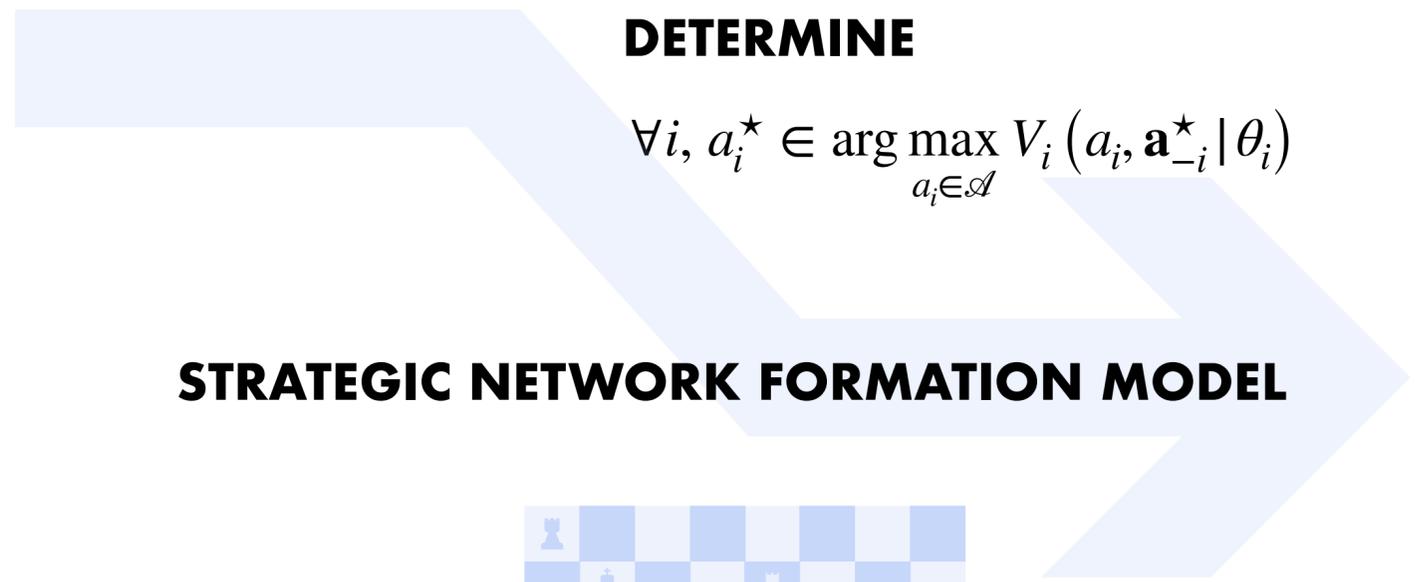
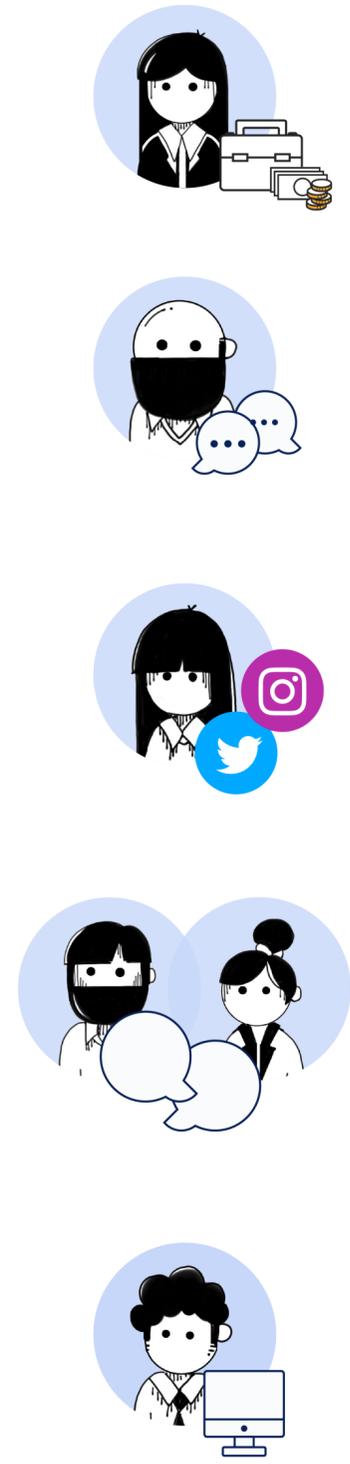
$$\forall i, \theta_i^* \text{ s.t. } V_i(a_i, \mathbf{a}_{-i}^*, \theta_i^*) \leq V_i(a_i^*, \mathbf{a}_{-i}^*, \theta_i^*), \forall a_i \in \mathcal{A}$$



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

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

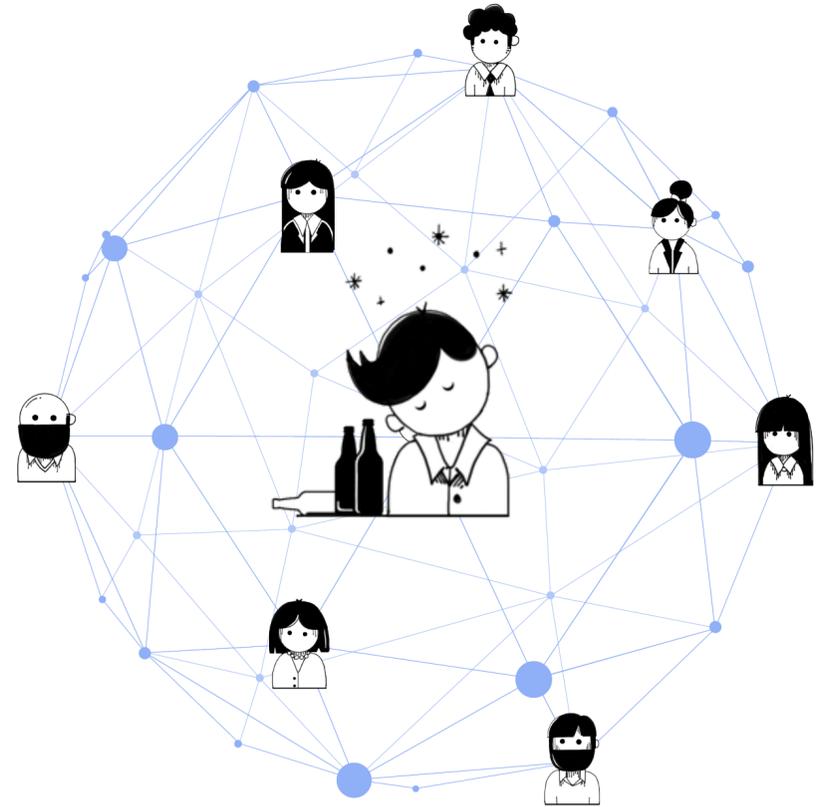
### INDIVIDUAL BEHAVIOUR $\theta_i$



### STRATEGIC NETWORK FORMATION MODEL



$\theta_i^*$  providing the **most rational** explanation to NE



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

# HOMOGENEOUS RATIONAL AGENTS

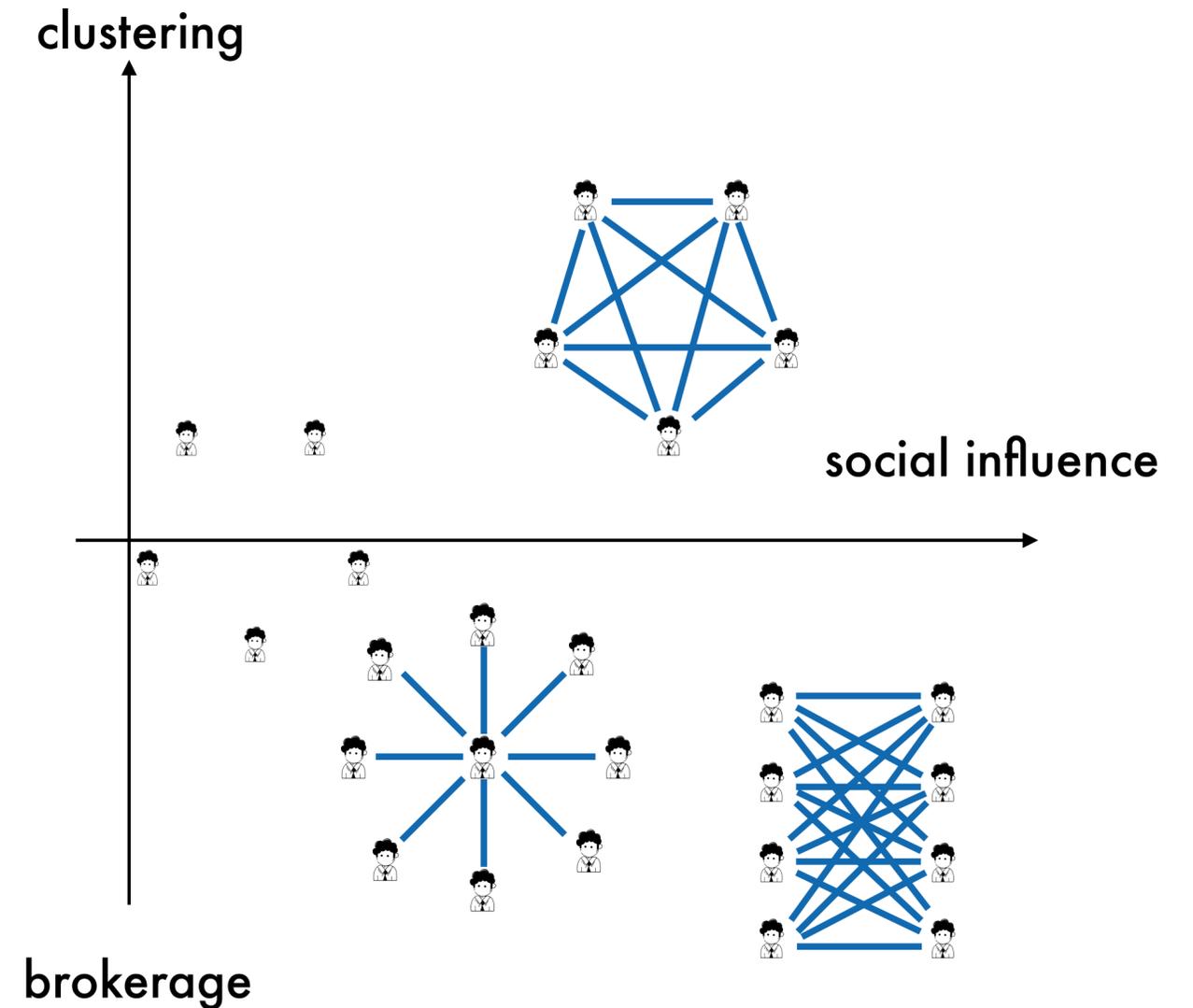
## Assumption.

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

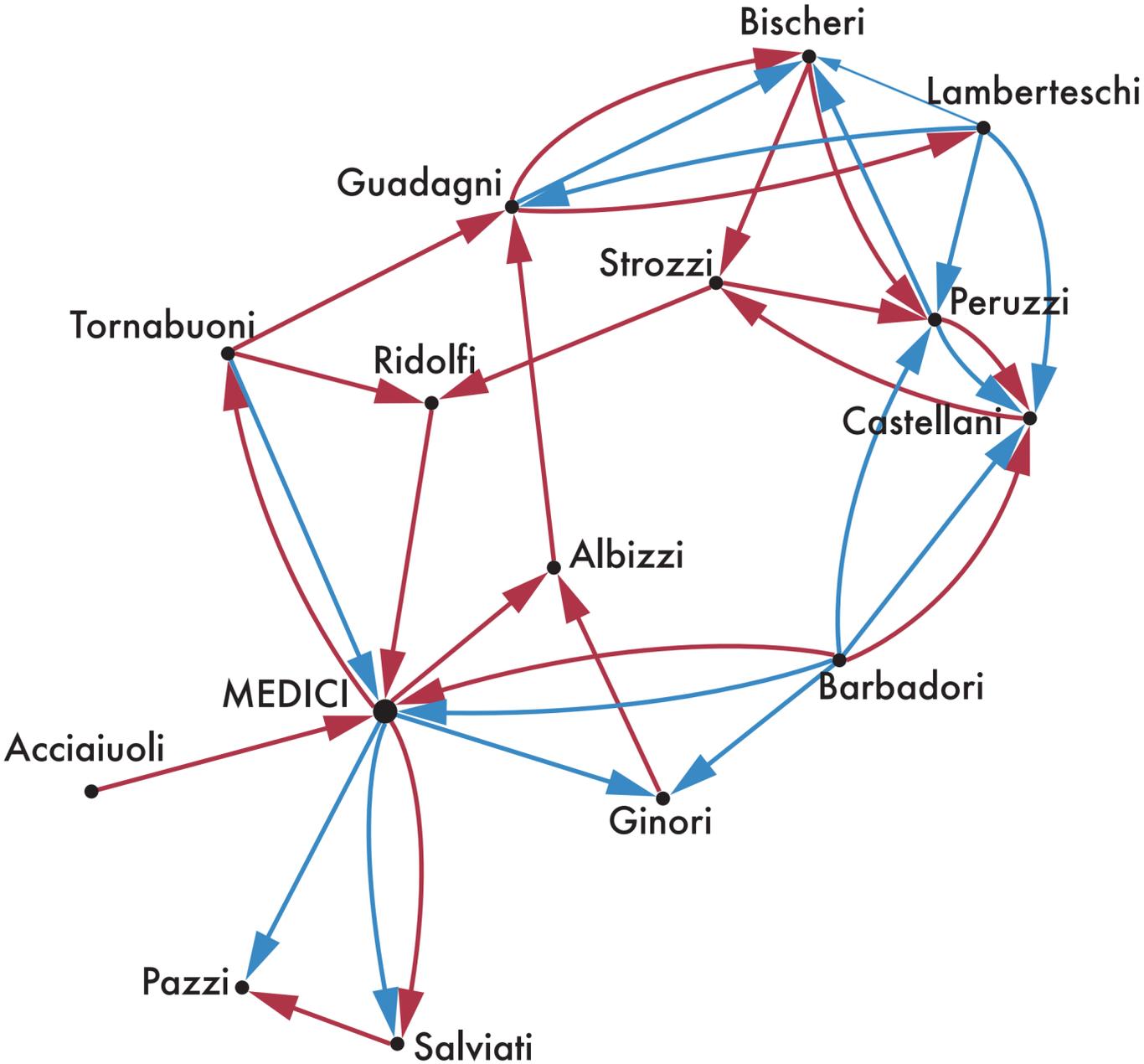
For specific **network motifs**, analytical parametric necessary conditions can be derived through **Variational Inequality**. Sufficiency can also be established.

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.**

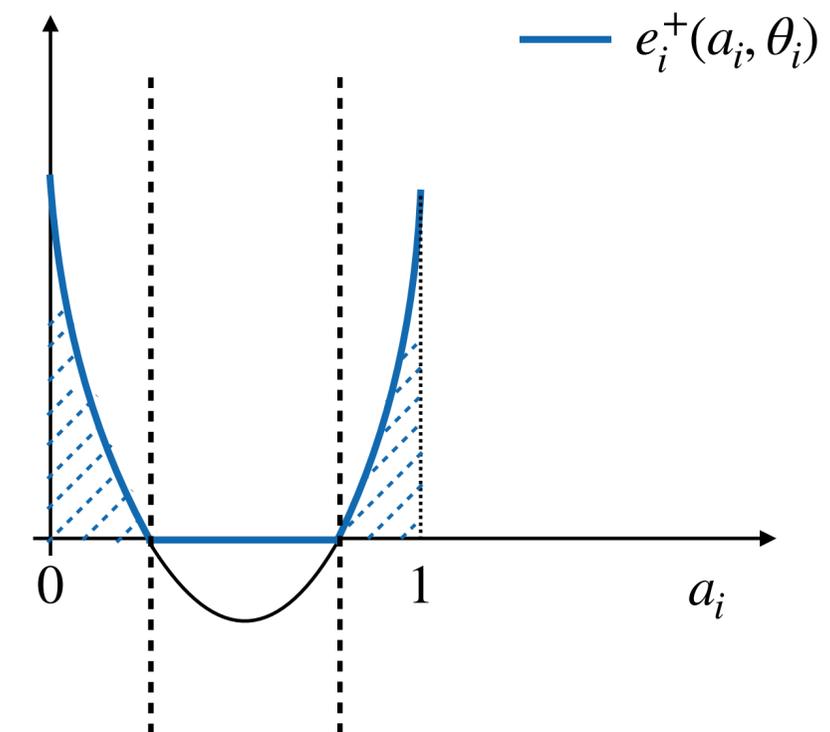
$$e_i(a_i, \theta_i) := V_i(a_i, \theta_i | \mathbf{a}_{-i}^*) - V_i(a_i^*, \theta_i | \mathbf{a}_{-i}^*)$$

$e_i^+(a_i, \theta_i) := \max\{0, e_i(a_i, \theta_i)\} > 0$  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  $\theta_i^*$  such that

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

*Theorem [Convexity of the objective function].*

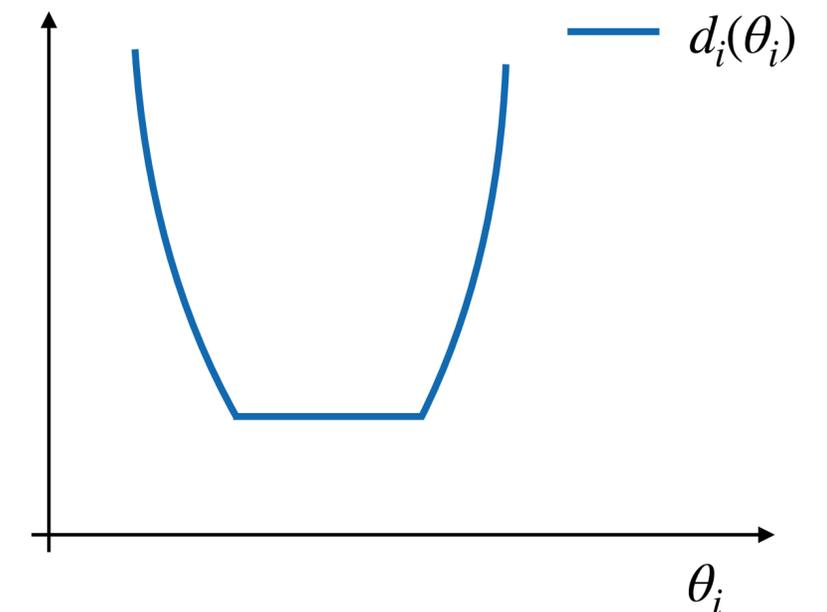
Let  $e_i(a_i, \theta_i) : \mathbb{R}^n \times \mathbb{R}^p \rightarrow \mathbb{R}$  be a continuous function of  $a_i \in \mathbb{R}^n$  and  $\theta_i \in \mathbb{R}^p$ , and linear in  $\theta_i$ , and let  $\mathcal{A}$  be a compact subset of  $\mathbb{R}^n$ . Consider the squared distance function:

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

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

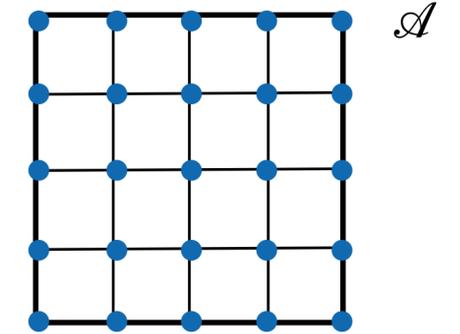
$$\tilde{d}_i(\theta_i) := \left( \sum_{j=1}^{n_i} \left( e_i^{j,+}(\theta_i) \right)^2 \right)^{1/2} = \|\mathbf{e}_i^+\|_2$$

**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)



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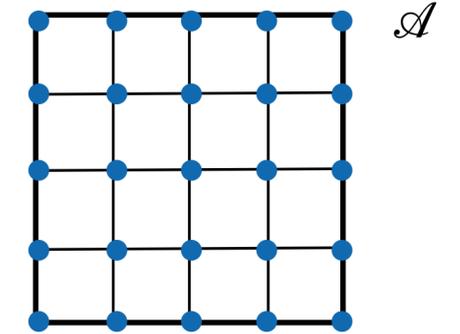
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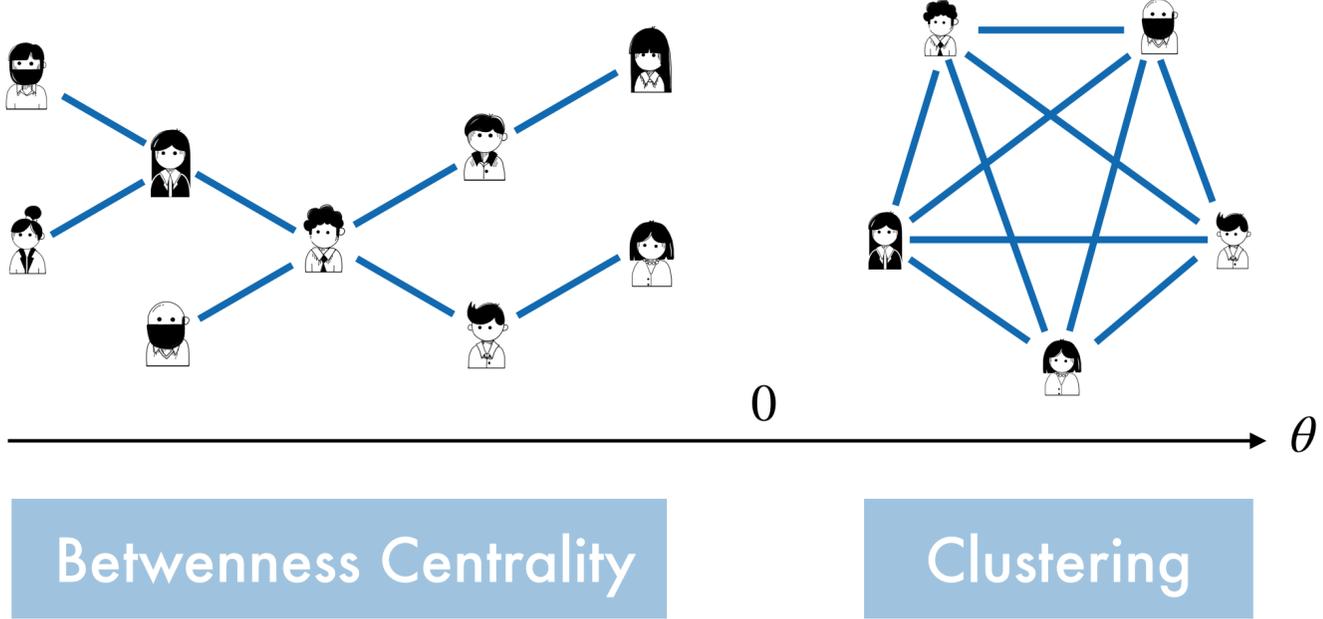
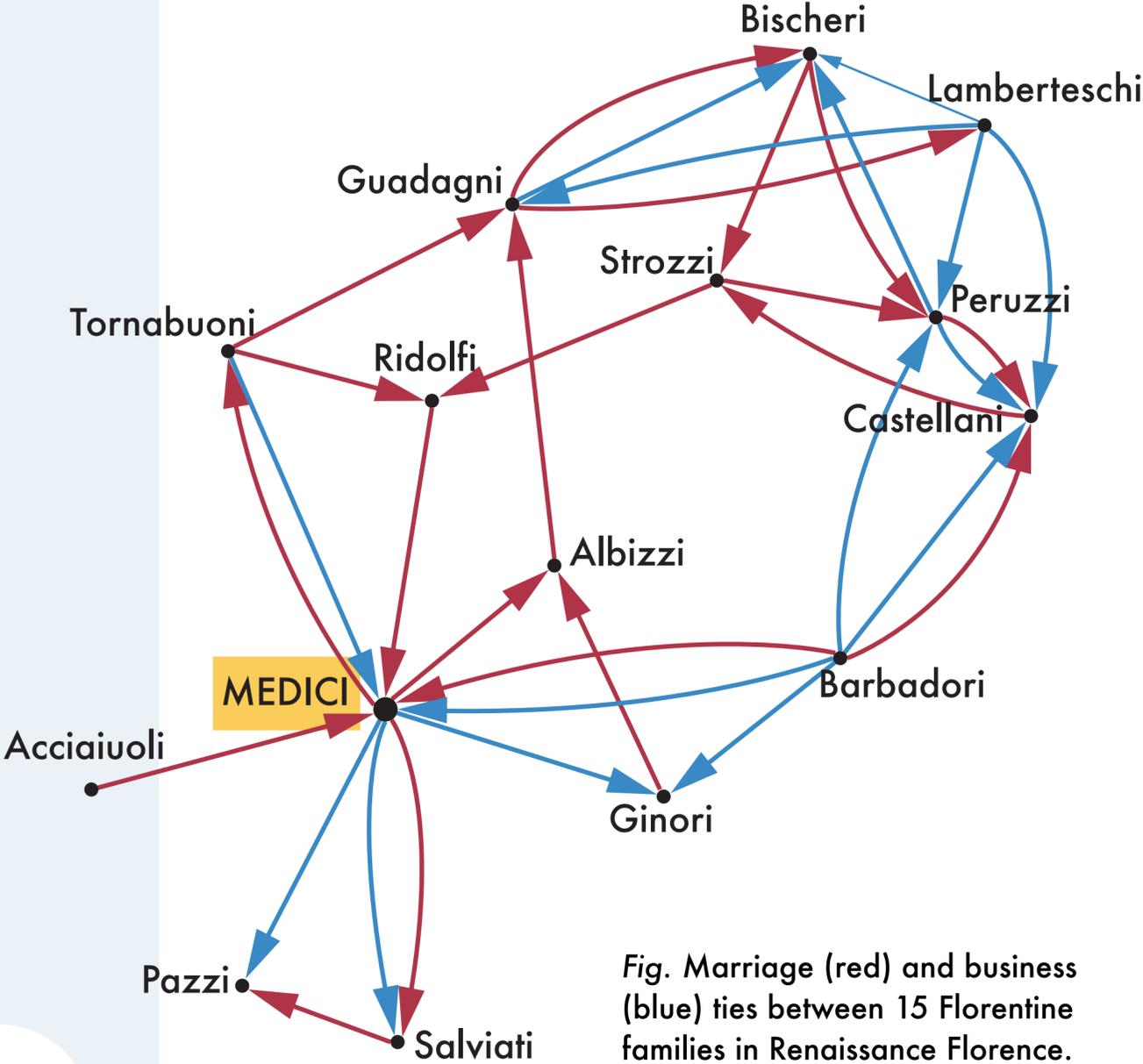
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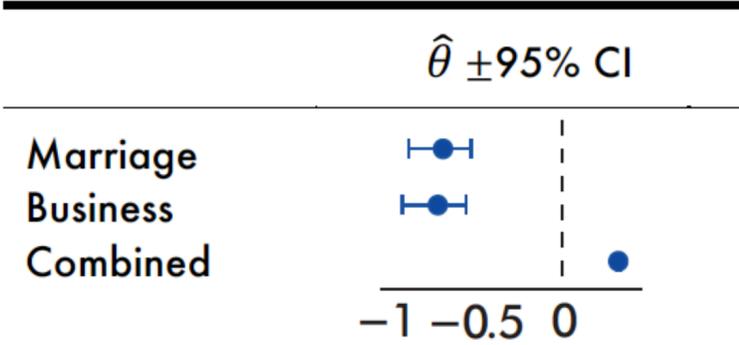


Similar to Generalized Least Square Regression

# 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

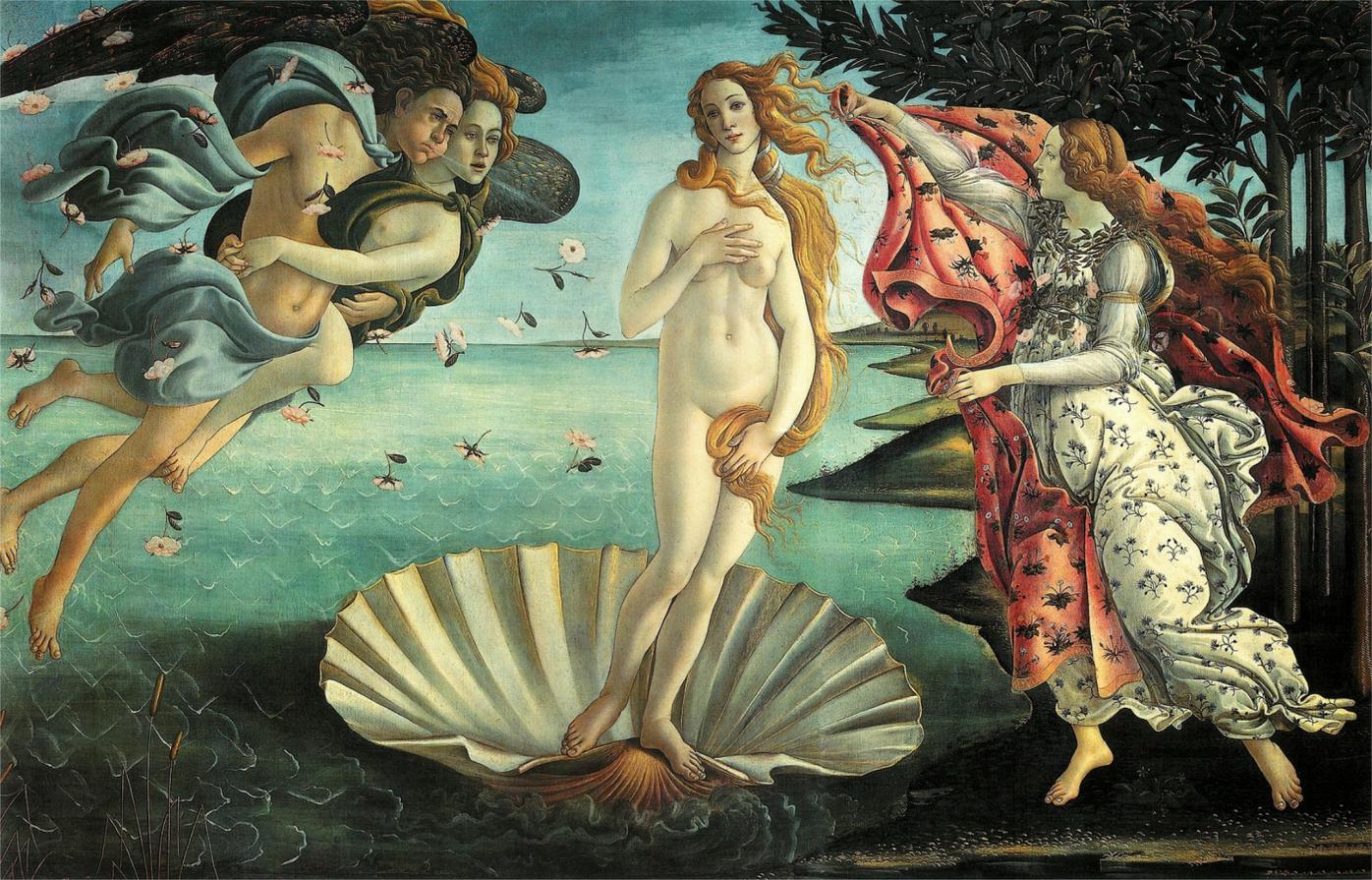
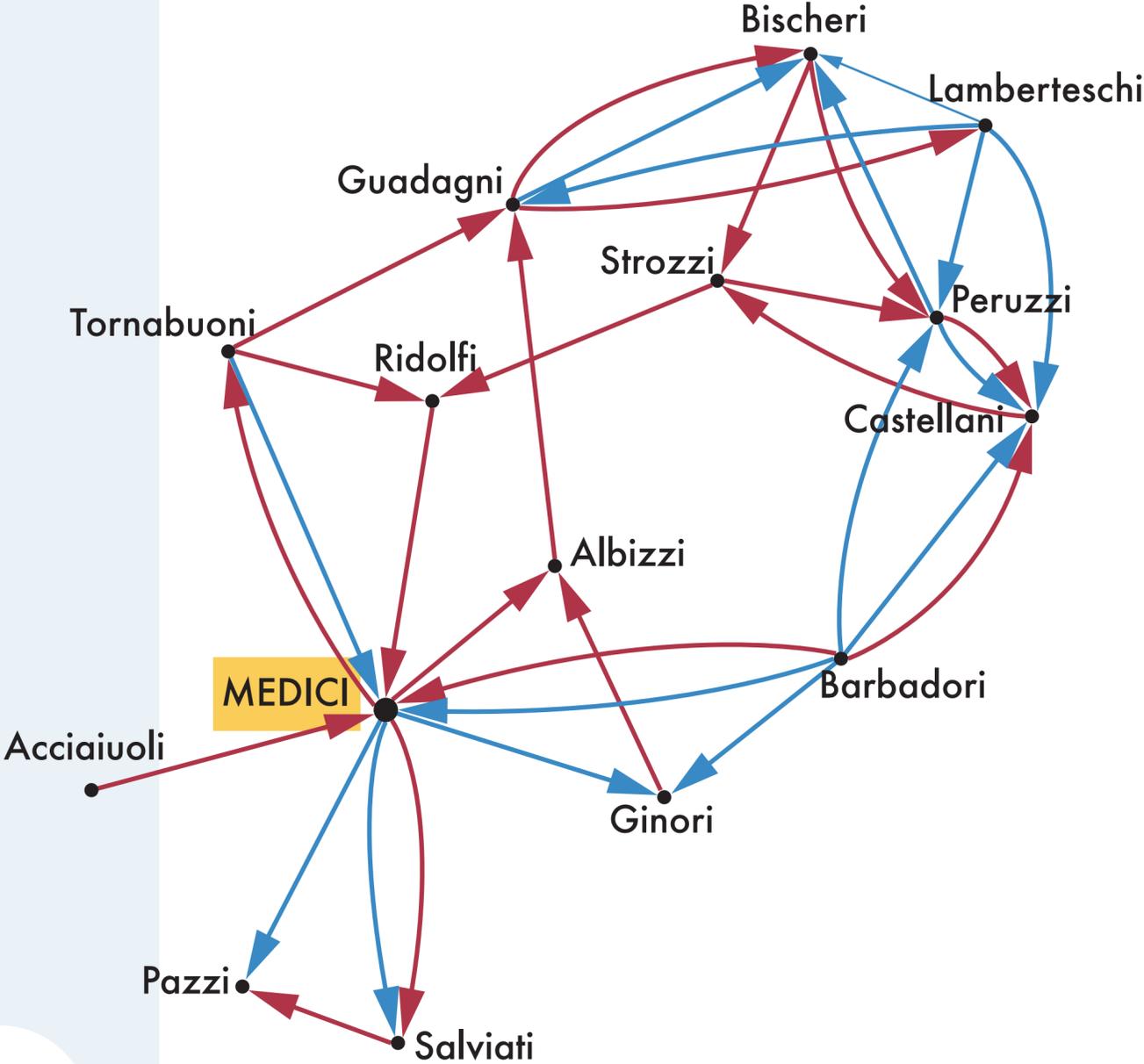
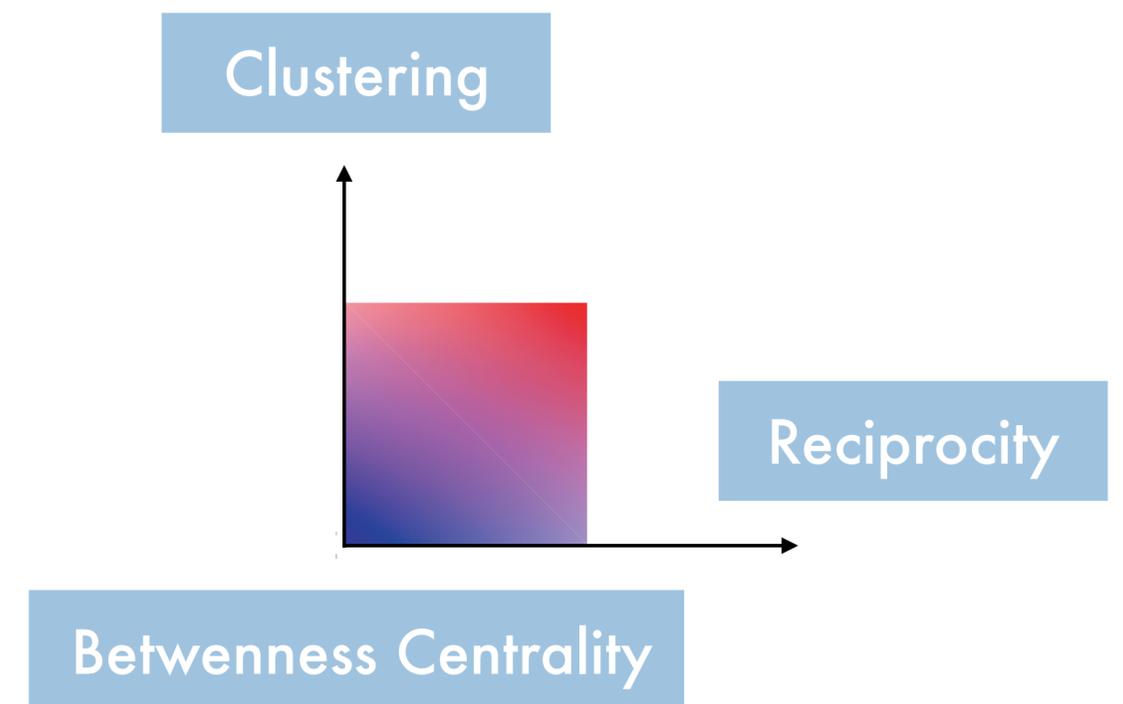
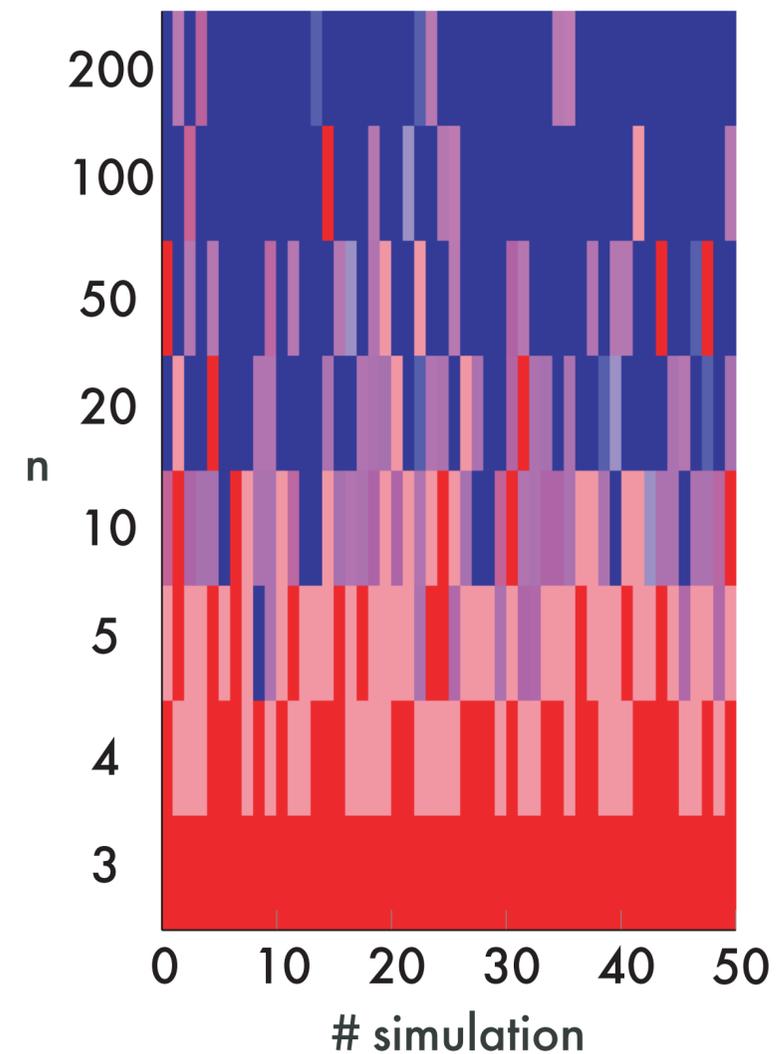


Fig. "La nascita di Venere", 1482-1485, Sandro Botticelli. Galleria degli Uffizi, Firenze.

# 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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