

Externalities, Supermodularities, and Equilibrium in Games with Endogenous Network Formation

master thesis

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Introduction

Network with Peer Effect

- Network structure and local interactions play an important role in individual and aggregate behavior
- Externalities of the individual behavior in the network is a key factor for the aggregate behavior
- Especially, we can see the importance of positive externalities
 - R&D network, criminal network, educational network
- Many literatures argue the importance of "peer effects" both theoretically and empirically

Endogeneity of the network

- However, in many works, the network is exogenous and fixed
- When economic agents are faced with some shocks or policy changes, they respond to them and the network will be changed
- Related to "Lucas Critique"

This paper

- This paper argue the endogenous network formation with peer effect
- We consider the model where
 - agents first choose the agents who they connect
 - agents choose the level of effort given the network structure
- We provide
 - the existence of subgame perfect equilibrium, where all agents take pure strategies at each stage
 - the argument about the uniqueness of the equilibrium
 - the discussion for policy implication (key player policy)

- Network with peer effect
 - Ballester et.al.(2006), Calvó-Armengol et.al.(2009), Liu et.al.(2012)
- Endogenous network
 - Acemoglu and Azar(2019), Oberfield(2018), Farboodi(2014)
- Closely related paper
 - Kim et.al.(2017), Hiller, T.(2017)

Model

Model : Setup

- the set of agents : $N = \{1, \dots, n\}$ with $n \geq 2$ and $n < \infty$
- Agents are initially connected in *potential network* g^P
- g^P is represented by adjacency matrix $\mathbf{G}^P = (g_{ij}^P)_{ij}$ where

$$g_{ij}^P = \begin{cases} 1 & \text{(if } i \text{ has a link to } j \text{ in } g^P) \\ 0 & \text{(otherwise)} \end{cases}$$

- Self-loop is not allowed : $g_{ii}^P = 0$ for all $i \in N$
- Agent i 's neighbors in g^P : $N_i(g^P) = \{j \in N | g_{ij}^P = 1\}$

Model : 1st stage

- First, agents simultaneously choose their neighbors from the agents whom they connect in potential network
- This strategy is represented by $\psi_i = (\psi_{i1}, \dots, \psi_{in})$ such that $\psi_{ij} \in \{0, 1\}$ for all $j \in N$, $\psi_{ii} = 0$ for all $i \in N$, and $\psi_{ij} = 0$ for all $j \notin N_i(g^P)$
- Note that $\psi_i \in \Psi_i = \{0, 1\}^n$
- When agent form links, he incurs the link-specific costs $c_{ij} \geq 0$
- "Choosing neighbors" strategy is dependent on g^P and $\mathbf{C} = (c_{ij})_{ij}$, so sometimes we denote $\psi_i(g^P, \mathbf{C})$

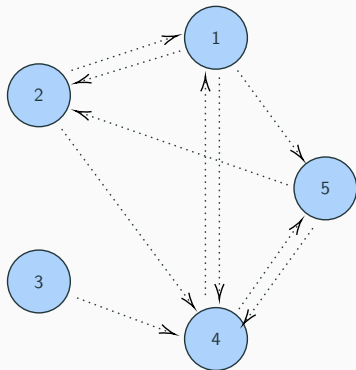
Model : Realized network

- At the end of 1st stage, we can see *realized network* denoted as g
- g is represented by the adjacency matrix \mathbf{G}

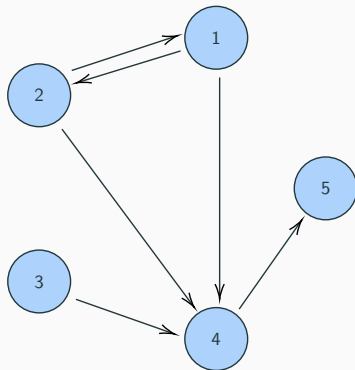
$$g_{ij} = \begin{cases} 1 & (\text{if } \psi_{ij}(g^p, \mathbf{C}) = 1) \\ 0 & (\text{otherwise}) \end{cases}$$

- g depends on $\psi(g^p, \mathbf{C}) = \prod_{i=1}^n \psi_i(g^p, \mathbf{C})$, so we can denote $g(\psi(g^p, \mathbf{C}))$
- We denote $g(\psi)$

Model : 1st stage



potential network



realized network

Figure 1: Difference between potential network and realized network

Model : 2nd stage

- Given realized network, each agent $i = 1, \dots, n$ simultaneously exerts an effort $x_i \geq 0$
- Denote $\mathbf{x} = (x_1, \dots, x_n)$
- Payoff function is

$$u_i(\mathbf{x}, \psi, \mathbf{C}, \phi) = v_i(\mathbf{x}, g(\psi), \phi) - \sum_{j=1}^n g_{ij}(\psi) c_{ij}$$

where

$$v_i(\mathbf{x}, g(\psi), \phi) = \alpha_i x_i - \frac{1}{2} x_i^2 + \phi \sum_{j=1}^n g_{ij}(\psi) x_i x_j$$

- $\phi > 0$ and cross term represent the peer effect
- Optimal level of effort depends on the realized network

Equilibrium

- **Definition** : Given g^P and \mathbf{C} , the network g^* is equilibrium network if $g^* = g(\psi^*)$ where ψ^* is the strategies in the pure-strategy subgame perfect equilibrium.
- We focus on pure strategy equilibrium to consider the non-stochastic network

2nd stage equilibrium

- **Assumption** : $\phi\rho(\mathbf{G}^p) < 1$ where $\rho(\cdot)$ is spectral radius
- **Proposition** : Under Assumption, for any realized network g , the subgame has a unique Nash equilibrium \mathbf{x}^* , which is interior and given by

$$\mathbf{x}^* = (\mathbf{I} - \phi\mathbf{G})^{-1}\alpha$$

- We can calculate as

$$v_i^*(\mathbf{x}^*(g(\psi)), g(\psi), \phi) = \frac{1}{2}x_i^*(g(\psi))^2$$

Best response dynamics : Step 0

Let $g^{(0)}$ be the initial realized network where $g^{(0)} = g^p$, that is, choosing strategy is $\psi_i^{(0)} \in \Psi_i$ such that $\psi_{ij}^{(0)} = 1$ for all $j \in N_i(g^p)$. Compute each players' optimal effort and payoffs. Denote the set of agents who do not take best response as $NB^{(0)}$:

$$\begin{aligned} NB^{(0)} = \{ & i \in N \mid \exists \tilde{\psi}_i \subset N_i(g^p) \\ & \text{s.t. } u_i(\mathbf{x}^*(g(\tilde{\psi}_1, \psi_{-i}^{(0)}), \tilde{\psi}_i, \psi_{-i}^{(0)}, \mathbf{C}, \phi) > u_i(\mathbf{x}^*(g(\psi^{(0)})), \psi^{(0)}, \mathbf{C}, \phi) \\ & \text{and } \tilde{\psi}_i \neq \psi_i^{(0)} \} \end{aligned}$$

Go into Step 1.

Best response dynamics : Step $k \geq 1$

Check whether $NB^{(k-1)}$ is empty or not.

If $NB^{(k-1)} = \emptyset$, define $g^* = g(\psi^{(k-1)})$ and terminate the algorithm.

Otherwise, choose a agent $i \in NB^{(k-1)}$ randomly. i changes her strategy from $\psi_i^{(k-1)}$ to $\psi_i^{(k)}$ such that

$$u_i(\mathbf{x}^*(g(\psi_i^{(k)}, \psi_{-i}^{(k-1)})), \psi_i^{(k)}, \psi_{-i}^{(k-1)}, \mathbf{C}, \phi) >$$

$$u_i(\mathbf{x}^*(g(\psi^{(k-1)})), \psi^{(k-1)}, \mathbf{C}, \phi), \text{ and for any } j (\neq i), \psi_j^{(k)} = \psi_j^{(k-1)}.$$

Then, new network $g(\psi^{(k)})$ is realized. Compute each players' payoffs and define $NB^{(k)}$:

$$NB^{(k)} = \{i \in N | \exists \tilde{\psi}_i \subset N_i(g^P)$$

$$\text{s.t. } u_i(\mathbf{x}^*(g(\tilde{\psi}_1, \psi_{-i}^{(k)}), \tilde{\psi}_i, \psi_{-i}^{(k)}, \mathbf{C}, \phi) > u_i(\mathbf{x}^*(g(\psi^{(k)})), \psi^{(k)}, \mathbf{C}, \phi)$$

$$\text{and } \tilde{\psi}_i \neq \psi_i^{(k)}\}$$

Proceed to Step $k + 1$.

Existence of the equilibrium network

- **Theorem** : Best response dynamics algorithm terminates in finite steps, and converged network represents the equilibrium network
- Given 2nd stage Nash equilibrium, 1st stage game becomes *supermodular game*, so pure strategy Nash equilibrium
- By supermodularity, BR dynamics returns the largest equilibrium
- **Intuition** : When agent i forms more links, he exerts more effort by the strategic complementarity in 2nd stage. Agent i 's increased effort makes agents who have a link to him exert more effort, so all agents' level of effort weakly increases. Increasing level of efforts makes the agents to connect more agents.

Uniqueness of the equilibrium

- Equilibrium network may not be unique

- **Example** : Consider $n = 2$, $\alpha = (1, 1)$, and $\mathbf{G}^p = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$.

Then $\begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$ and $\begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$ can be both equilibrium network for some c_{12} and c_{21} .

Uniqueness of the equilibrium

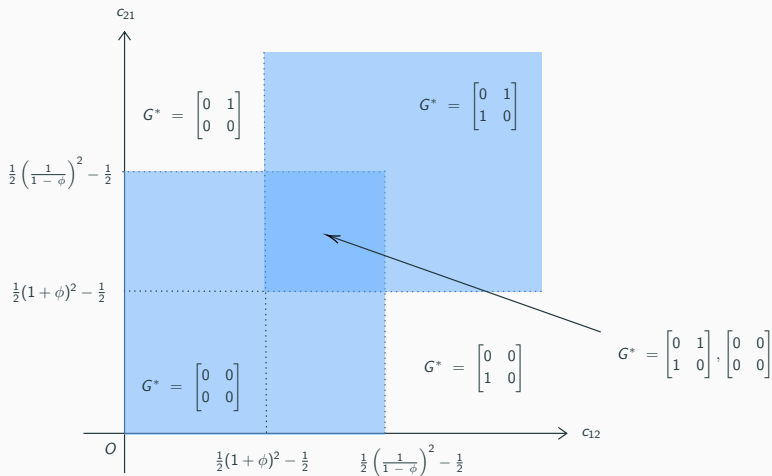


Figure 2: Equilibrium network region

- **Theorem** : The equilibrium network derived by BR dynamics is unique

- **Proposition** : Given the potential network g^P . Consider the cost $\hat{\mathbf{C}}$ and \mathbf{C} with $\hat{\mathbf{C}} \leq \mathbf{C}$. Then,

$$g(\psi^*(g^P, \hat{\mathbf{C}})) \supseteq g(\psi^*(g^P, \mathbf{C}))$$

- When the link formation costs increase, the equilibrium network becomes denser

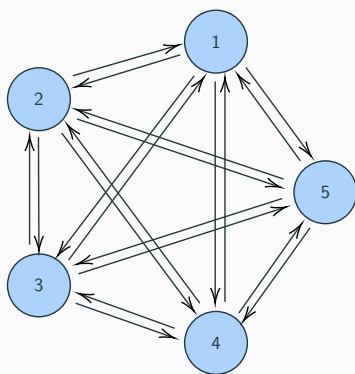
Phase transition

- **Example** : Suppose $n = 5$, $\alpha = (1, 1, 1, 1, 1)$, and $\phi = 1/5$

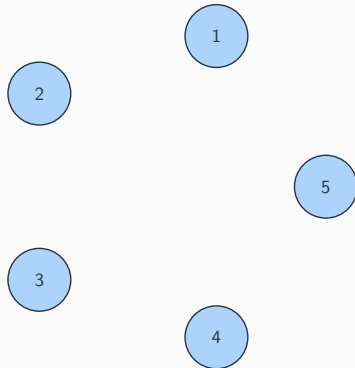
- $\mathbf{C} = \begin{bmatrix} 0 & 3 & 3 & 3 & 3 \\ 3 & 0 & 3 & 3 & 3 \\ 3 & 3 & 0 & 3 & 3 \\ 3 & 3 & 3 & 0 & 3 \\ 3 & 3 & 3 & 3 & 0 \end{bmatrix} \Rightarrow \mathbf{G}^* = \begin{bmatrix} 0 & 1 & 1 & 1 & 1 \\ 1 & 0 & 1 & 1 & 1 \\ 1 & 1 & 0 & 1 & 1 \\ 1 & 1 & 1 & 0 & 1 \\ 1 & 1 & 1 & 1 & 0 \end{bmatrix}$

- $\hat{\mathbf{C}} = \begin{bmatrix} 0 & 3 + \epsilon & 3 & 3 & 3 \\ 3 & 0 & 3 & 3 & 3 \\ 3 & 3 & 0 & 3 & 3 \\ 3 & 3 & 3 & 0 & 3 \\ 3 & 3 & 3 & 3 & 0 \end{bmatrix} \Rightarrow \hat{\mathbf{G}}^* = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$

Phase transition



Equilibrium network G



Equilibrium network \hat{G}

Figure 3: Equilibrium networks

Finding the Key Player

- Key player is the agent who has the largest impact on the aggregate behavior of the network
- **Definition** : Agent i is the key player in exogenous network g if, given network g ,

$$i \in \arg \max_{i \in N} \{x^*(g) - x^*(g^{-i})\}$$

where $x^*(g) = \sum_{i=1}^n x_i^*(g)$ and g^{-i} is the network where agent i is removed from the network g

Key player in endogenous network

- **Definition** : Agent i is the key player in endogenous network if, given potential network g^P ,

$$i \in \arg \max_{i \in N} \{x^*(g(\psi(g^{P-i}, \mathbf{C}))) - x^*(g(\psi(g^{P-i}, \mathbf{C}^{-i})))\}$$

where g^{P-i} is the network where agent i is removed from the network g^P

- However, it is difficult to identify key player due to the complexity of the mapping from cost structure to realized network

Difference bet. endogenous and exogenous key player

- **Example** : Suppose $n = 5$, $\alpha = (1, 1, 1, 1, 1)$ and $\phi = 1/5$

- $\mathbf{C} = \begin{bmatrix} 0 & 3.6 & 0.2 & 0.2 & 0.2 \\ 0.3 & 0 & 0.2 & 0.5 & 5.5 \\ 0.2 & 0.2 & 0 & 4.5 & 4.3 \\ 4.1 & 0.2 & 0.4 & 0 & 6.5 \\ 3.2 & 4.1 & 0.3 & 1.0 & 0 \end{bmatrix} \Rightarrow \mathbf{G}^* = \begin{bmatrix} 0 & 0 & 1 & 1 & 1 \\ 1 & 0 & 1 & 1 & 0 \\ 1 & 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \end{bmatrix}$

- Then,

agent 1	$x_1^* = 1.99541284$	key player in endogenous network
agent 2	$x_2^* = 2.12155963$	agent with highest effort
agent 3	$x_3^* = 1.82339450$	key pplayer in exogenous network
agent 4	$x_4^* = 1.78899083$	
agent 5	$x_5^* = 1.36467890$	

Difference bet. endogenous and exogenous key player

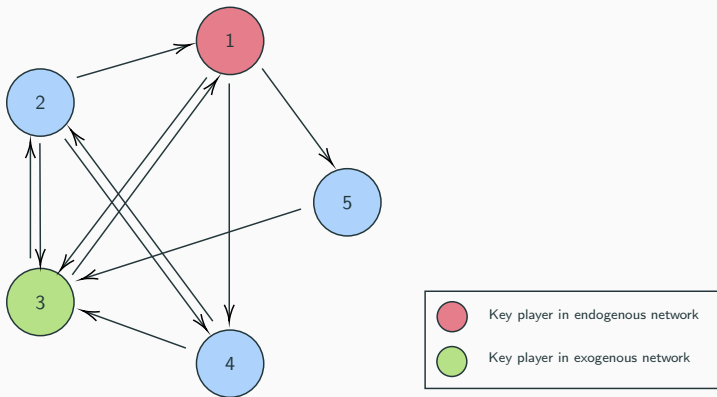


Figure 4: Equilibrium network G^* and key players

Conclusion

Conclusion

- We consider the endogenous network formation with peer effect
- In the model, link formation costs play an important role in determining the network structure and individual and aggregate behaviors
- Due to the supermodularity, we can show the existence of equilibrium network
- We can provide
 - a kind of uniqueness of the equilibrium
 - comparative statics result
 - discussion about key player and policy implication

- link formation with bilateral agreement
- weighted network
- other form of utility function