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

master thesis

Yuya Furusawa December 7, 2019

U-Tokyo, GSE

Introduction

Network with Peer Effect

- Network structure and local interactions play an important role in individual and aggregate nehavior
- Externalities of the individual behavior in the network is a key factor for the aggeregate bahavior
- 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

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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)

Related Literature

- 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 $n \ge 2$ and $n < \infty$
- Agents are initially connected in potential network g^p
- ullet g^p is represented by adjacency matrix $oldsymbol{G}^p = \left(g^p_{ij}
 ight)_{ii}$ 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 $i \in N$
- ullet Agent i's neighbors in g^p : $N_i(g^p) = \{j \in N | g^p_{ij} = 1\}$

Model: 1st stage

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

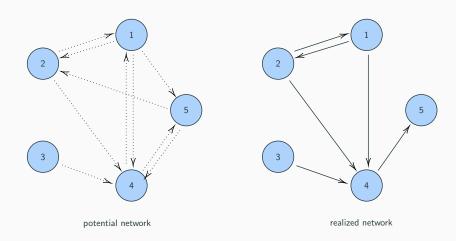
Model: Realized network

- At the end of 1st stage, we can see realized network denoted as g
- ullet g is represented by tha adjacency matrix $oldsymbol{G}$

$$g_{ij} = egin{cases} 1 & ext{(if } \psi_{ij}(g^p, m{C}) = 1) \\ 0 & ext{(otherwise)} \end{cases}$$

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

Model: 1st stage



 $\textbf{Figure 1:} \ \, \mathsf{Difference} \ \, \mathsf{between} \ \, \mathsf{potential} \ \, \mathsf{network} \ \, \mathsf{and} \ \, \mathsf{realized} \ \, \mathsf{network}$

Model: 2nd stage

- Given realized network, each agent $i=1,\cdots,n$ simultaneously excerts 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

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

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

Equilibrium

Equilibrium

- **Definition**: Given g^p and 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** : $\phi \rho(\mathbf{G}^p)$ where $\rho(\cdot)$ is spectral radius
- **Proposition**: Under Assumption, for any realized network g, the subgame has a unique Nash equilibrium 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, chooosing 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 $NB^{(0)}$:

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

Go into Step 1.

Best response dynamics : Step $k \ge 1$

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

$$\begin{split} \textit{NB}^{(k)} &= \{i \in \textit{N} | \exists \tilde{\psi}_i \subset \textit{N}_i(\textit{g}^\textit{p}) \\ \text{s.t. } \textit{u}_i(\textit{\textbf{x}}^*(\textit{g}(\tilde{\psi}_1, \psi_{-i}^{(k)}), \tilde{\psi}_i, \psi_{-i}^{(k)}, \textit{\textbf{C}}, \phi) > \textit{u}_i(\textit{\textbf{x}}^*(\textit{g}(\psi^{(k)})), \psi^{(k)}, \textit{\textbf{C}}, \phi) \\ \text{and } \tilde{\psi}_i \neq \psi_i^{(k)} \} \end{split}$$

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 excert more effort by the strategic complementarity in 2nd stage. Agent *i*'s increased effort makes agents who have a link to him excert 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

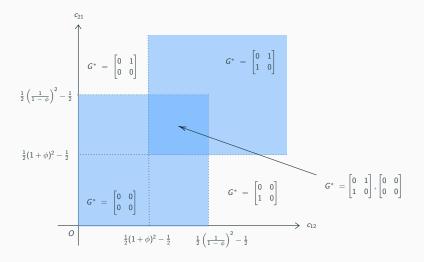


Figure 2: Equilibrium network region

Uniqueness of the equilibrium

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

Comparative Statics

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

$$g(\psi^*(g^p,\hat{\boldsymbol{C}}))\supseteq g(\psi^*(g^p,\boldsymbol{C}))$$

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

Phase transition

ullet Example : Suppose n=5, lpha=(1,1,1,1,1), and $\phi=1/5$

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

Phase transition

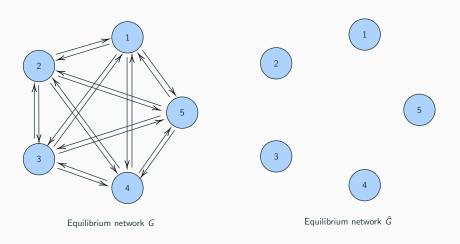


Figure 3: Equilibrium networks

Finding the Key Player

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,

$$i \in \arg\max_{i \in \mathcal{N}} \{x^*(g(\psi(g^{p^{-i}}, \boldsymbol{C}))) - x^*(g(\psi(g^{p^{-i}}, \boldsymbol{C}^{-i})))\}$$

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 strcture to realized network

Endogenous and Exogenous

ullet Example : Suppose n=5, lpha=(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 with highest effort	agent 2
key player in endogenous network	agent 1
key player in exogenous network	agent 3

Endogenous and Exogenous

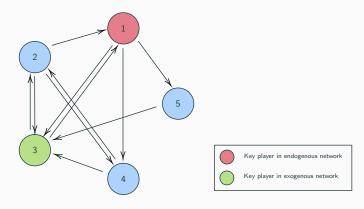


Figure 4: Equilibrium network G^* and key players

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

Future work

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