

# Applied economic networks - VI

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# Today's class

- ▶ Network games - continuation
- ▶ Amine's presentation: *Peer effects and social networks in education*

## A more interesting game

- ▶ Each player  $i \in N = \{1, 2, \dots, n\}$  simultaneously selects an action  $a_i \geq 0$ .
- ▶ Payoff depend on everyone's action  $u_i(a_1, a_2, \dots, a_n)$ .

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Best response

$$BR_i(\mathbf{a}_i) = \alpha \sum_j w_{ij} a_j + b_i$$

where  $\alpha > 0$ .

What if  $w_{ij} = \frac{1}{d_i}$  for all  $j$  such that  $g_{ij} = 1$ ?

## Why this best response?

- ▶ Can be shown that it corresponds to the linear-in-means model (peer effects)
- ▶ In a dynamic setting this is the DeGroot updating rule (naive learning – in two lessons)
- ▶ In general has deep interpretation
- ▶ It is linear..

# What payoff?

The best response described before can be derived from two different payoff functions:



$$u_i(\mathbf{a}) = -\frac{a_i^2}{2} + \alpha \sum_j w_{ij} a_i a_j + b_i a_i$$

- ▶ By setting  $b_i = (1 - \alpha)y_i$ :

$$u_i(\mathbf{a}) = -\alpha \sum_j (a_i - a_j)^2 - (1 - \alpha)(a_i - y_i)^2$$

## Solving the game

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- ▶  $(\mathbf{I} - \alpha \mathbf{W}) \mathbf{a} = \mathbf{b}$
- ▶ We need to invert the matrix  $(\mathbf{I} - \alpha \mathbf{W})$ .

## Solving the game - contd

- ▶ All rows of  $\mathbf{W}$  sum up to 1 (row stochastic matrix).
- ▶  $\alpha \in (0, 1)$ .

### Result

The game has a unique Nash equilibrium given by:

$$\mathbf{a}^* = (\mathbf{I} - \alpha\mathbf{W})^{-1}\mathbf{b}$$

Moreover,  $(\mathbf{I} - \alpha\mathbf{W})$  is a non-negative matrix.

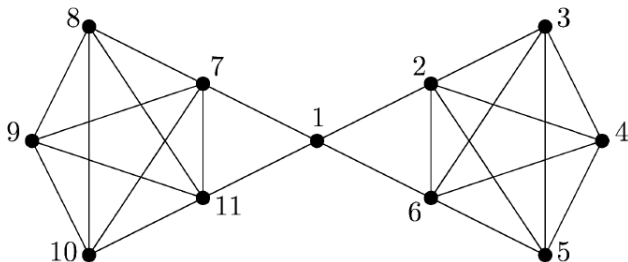
## Further result

Nash equilibrium and network structure

The actions in equilibrium are directly proportional to the Bonacich centrality.

- ▶ This model has first been studied by *Ballester, Calvo-armengol and Zenou, ECMA, 2006.*

## The key player

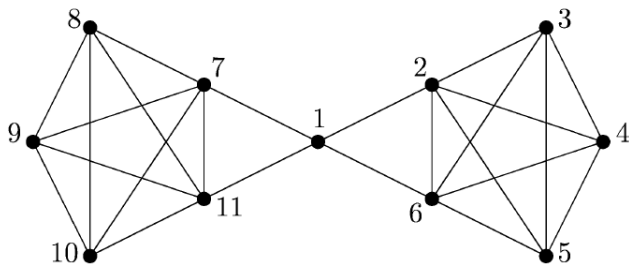


Who is the key player?

- ▶ Player 1 connects the network
- ▶ Removing player 2 we disrupt the highest number of links

Who has the highest impact on direct and indirect effects?

## The key player



Who is the key player? Result from BCZ 2006

- ▶ If  $\alpha$  is low enough player 2 is the key player
- ▶ If  $\alpha$  is large enough, player 1 is the key player  $\implies$  Indirect effects matter more

## Standard notation:

- ▶  $n$  agents
- ▶ Agents choose an action  $x_i \geq 0$ ,  $\mathbf{x}_{-i}$  denotes the vector of actions other than  $i$ .
- ▶  $g_{ij} = g_{ji} = 1$  indicates a link between  $i$  and  $j$ . Undirected network.
- ▶  $\delta > 0$  measure how much  $i$  and  $j$  affect each other's payoff.



# Payoff

Focus on linear best reply games.

A public good game

$$U_i(x_i, \mathbf{x}_{-i}; \delta, \mathbf{G}) = b_i(x_i + \delta \sum_j g_{ij} x_j) - \kappa_i x_i$$

where  $\kappa_i$  is the marginal cost.

Best replies

$$f_i(\mathbf{x}_{-i}, \delta, \mathbf{G}) = \bar{x}_i - \delta \sum_j g_{ij} x_j \quad \text{if} \quad \delta \sum_j g_{ij} x_j < \bar{x}_i$$

and  $f_i = 0$  otherwise.

$\bar{x}_i$  is the value equating marginal cost ( $d$ ) and marginal benefit ( $a$ ), assumed to be equal among players.

## Payoff - II

- ▶ We can obtain the same best reply function by considering the following payoff:

$$\tilde{U}_i(x_i, \mathbf{x}_{-i}; \delta, \mathbf{G}) = \bar{x}_i x_i - \frac{1}{2} x_i^2 - \delta \sum_j g_{ij} x_i x_j$$

- ▶ More models can be considered (e.g. Cournot model with substitutes)

# Nash equilibria

Two definitions:

- ▶  $\mathbf{G}_A$  denotes the links connecting active agents.
- ▶  $\mathbf{G}_{N-A,A}$  denotes the links connecting inactive agents to active agents.

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## *Result*

Actions  $\mathbf{x}$  with active agents  $A$  is a Nash equilibrium if and only if

$$(i) (\mathbf{I} + \delta \mathbf{G}_A) \mathbf{x}_A = \mathbf{1}$$

$$(ii) \delta \mathbf{G}_{N-A,A} \mathbf{x}_A \geq \mathbf{1}$$

This generalizes previous results, for which  $\delta$  was either high (*Bramouille Kranton, 2007*) or low (*Ballester, Calvo-Armengol, Zenou, 2006*).

# Examples

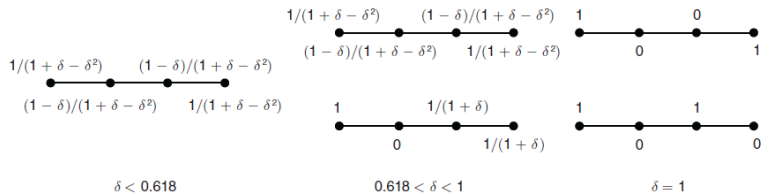


FIGURE 1. EQUILIBRIA IN A LINE NETWORK

# Examples

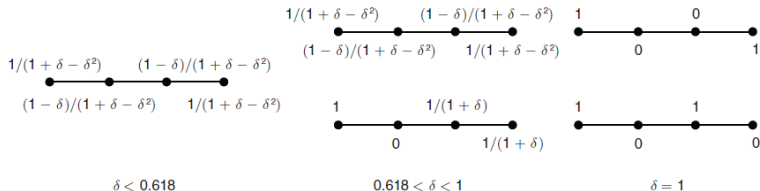


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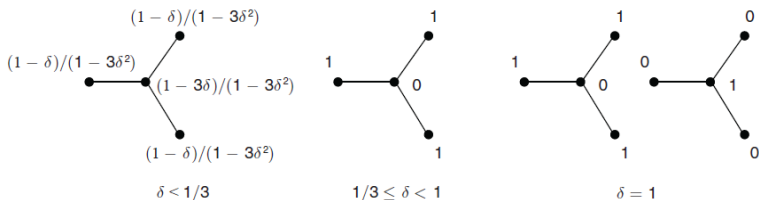


FIGURE 2. EQUILIBRIA IN A STAR NETWORK

## Overview - Potential function

For further results, we need to define conditions for getting all Nash equilibria.

*Potential games (Monderer, Shapley, 1996)*

Consider a game with payoffs  $V_i(x_i, \mathbf{x}_{-i})$ . A function  $\varphi(x_i, \mathbf{x}_{-i})$  is a *potential function* if and only if, for all  $x_i, x'_i$  and all  $\mathbf{x}_{-i}$ :

$$\varphi(x_i, \mathbf{x}_{-i}) - \varphi(x'_i, \mathbf{x}_{-i}) = V_i(x_i, \mathbf{x}_{-i}) - V_i(x'_i, \mathbf{x}_{-i}), \forall i$$

It exists if and only if

$$\partial^2 V_i(\mathbf{x}) / \partial x_i \partial x_j = \partial^2 V_j(\mathbf{x}) / \partial x_j \partial x_i$$

The potential function mimics individual best responses, and therefore allows for simplified analysis. We can find equilibrium actions, given all vectors of other agents' actions.

## Results - Potential function

Take the game with quadratic payoffs  $\tilde{U}_i$ . It has a *potential function* since  $\partial^2 \tilde{U}_i(\mathbf{x})/\partial x_i \partial x_j = \partial^2 \tilde{U}(\mathbf{x})/\partial x_j \partial x_i = -\delta g_{ij}$ , and it is

$$\varphi(\mathbf{x}; \delta, \mathbf{G}) = \mathbf{x}^T \mathbf{1} - \frac{1}{2} \mathbf{x}^T (\mathbf{I} + \delta \mathbf{G}) \mathbf{x}$$

Consider maximizing the *potential function*. We have then the following result:

### LEMMA 1

The set of Nash equilibria for a given  $\mathbf{G}$  and  $\delta$  corresponds to the set of maxima and saddle points of the potential function

$\varphi(\mathbf{x}; \delta, \mathbf{G})$  on  $\mathbb{R}_+^n$ .



## Results - Uniqueness and lowest eigenvalue

Given *LEMMA 1* a sufficient condition for uniqueness is concavity of the *potential function*. The matrix of second derivatives is given by:

$$\nabla^2 \varphi = -(\mathbf{I} + \delta \mathbf{G})$$

Therefore  $\mathbf{I} + \delta \mathbf{G}$  must be positive definite, which is equivalent to  $1 + \delta \lambda_{\min}(\mathbf{G}) > 0$ .

Result follows straightforward:

### *PROPOSITION 2*

If  $|\lambda_{\min}(\mathbf{G})| < 1/\delta$ , there is a unique Nash equilibrium.

### *Intuition*

If cumulative effect of the ups and downs is high enough (lowest eigenvalue reflects substitutability), multiple equilibria emerge. the larger  $\lambda_{\min}$ , the greater the amplification of actions.

## Stability in an example

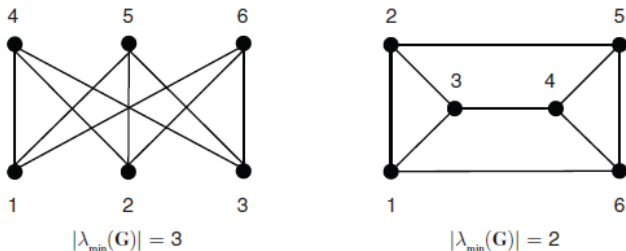


FIGURE 4. LOWEST EIGENVALUE AND STABLE EQUILIBRIA

Imagine a “perturbation”  $\varepsilon$  in agent’s 1 best reply. What happens?

## Results - Stable equilibria

To study stability Nash tâtonnement.

Lowest eigenvalue again is the key of stability.

### *Stable equilibria and Lowest eigenvalue*

An equilibrium  $\mathbf{x}$  with active agents  $A$  and inactive agents is stable if and only if  $|\lambda_{\min}(\mathbf{G}_A)| < 1/\delta$ .

When  $|\lambda_{\min}(\mathbf{G})| < 1/\delta$ , the unique equilibrium is stable.

When  $|\lambda_{\min}(\mathbf{G})| > 1/\delta$ , among the multiple equilibria, all stable equilibria involve inactive agents.

### *Intuition*

We check the perturbation  $\varepsilon$  of an equilibrium that yields the largest change in best replies (*maximal perturbation*). It derives from the eigenvectors associated to the lowest eigenvalue (because Str. Sub.). If the maximal perturbation does not lead the system away, nothing else will.

## Results - Lowest eigenvalue and network structure

Take the characterization of the lowest eigenvalue:

$$\lambda_{\min}(\mathbf{G}) = \min \varepsilon^T \mathbf{G} \varepsilon$$

Now divide agents in two sets (bipartite graph)  $R$  and  $S$  such that  $R = \{i | \varepsilon_i \geq 0\}$  and  $S = \{i | \varepsilon_i < 0\}$ . Then

$$\lambda_{\min}(\mathbf{G}) = \sum_{i,j \in R} \varepsilon_i \varepsilon_j g_{ij} + \sum_{i,j \in S} \varepsilon_i \varepsilon_j g_{ij} + 2 \sum_{i \in R, j \in S} \varepsilon_i \varepsilon_j g_{ij}$$

*Intuition (PROP. 7)*

First two terms captures the link within, and they are always positive. The third term captures between links, and when is larger in magnitude lowers the value of  $\lambda_{\min}(\mathbf{G})$ . This means that actions rebound more in opposite directions (**substitutability**), and it happens when more links between groups are present.

# What's next?

Current research has some open questions:

- ▶ Integrate behavior with network formation
- ▶ Take models of network formation to the data: structural modeling of peer effects
- ▶ Study the impact of homophily, clustering and other network characteristics on behavior

Thank you for your attention