Network Economics: two examples

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#### (A) Diffusion in Social Networks

#### (B) Economics of Information Security

#### (1) Diffusion Model

inspired from game theory and statistical physics.

## (2) Results

from a mathematical analysis.

#### (3) Adding Clustering

joint work with Emilie Coupechoux



## Crossing the Chasm (Moore 1991)

Malcolm Gladwell

#### (1) Diffusion Model

#### (2) Results

#### (3) Adding Clustering

#### (1) Coordination game...







• Both receive payoff q.

Both receive payoff
1-q>q.



• Both receive nothing.

#### (1)...on a network.

Everybody start with
Field
Total payoff = sum of the payoffs with each

neighbor.

• A seed of nodes switches to tak

(Blume 95, Morris 00)

#### (1) Threshold Model

- State of agent i is represented by
- $X_{i} = \begin{cases} 0 & \text{if } & \text{icq} \\ 1 & \text{if } & \text{take} \end{cases}$ • Switch from from icq to take if:

$$\sum_{j \sim i} X_j \ge qd_i$$

#### (1) Model for the network?

*p* == 0.04

p == 0.05





#### Statistical physics: bootstrap percolation.

#### (1) Model for the network?



#### (1) Random Graphs

- Random graphs with given degree sequence introduced by (Molloy and Reed, 95).
- Examples:
  - Erdös-Réyni graphs,  $G(n,\lambda/n)$ .
  - Graphs with power law degree distribution.
- We are interested in large population asymptotics.
- Average degree is  $\lambda$ .
- No clustering: C=0.

#### (1) Diffusion Model q = relative threshold $\lambda = average degree$

#### (2) Results

#### (3) Adding Clustering

(1) Diffusion Model q = relative threshold $\lambda = average degree$ 

(2) Results

(3) Adding Clustering

#### (2) Contagion (Morris 00)

- Does there exist a finite groupe of players such that their action under best response dynamics spreads contagiously everywhere?
- Contagion threshold:  $q_c$  = largest q for which contagious dynamics are possible.



#### (2)Another example: d-regular trees



 $q_c = \frac{1}{d}$ 



Seed = one node, λ=3 and q=0.24 (source: the Technoverse blog)



Seed = one node, λ=3 and 1/q>4 (source: the Technoverse blog)

#### (2) Some experiments



Seed = one node, λ=3 and q=0.24 (or 1/q>4) (source: the Technoverse blog)

#### (2) Contagion threshold



#### (2) A new Phase Transition



#### (2) Pivotal players

 Giant component of players requiring only one neighbor to switch: deg <1/q.</li>



#### (2) q above contagion threshold

- New parameter: size of the seed as a fraction of the total population  $0 < \alpha < 1$ .
- Monotone dynamic  $\rightarrow$  only one final state.



#### (2)Minimal size of the seed, q>1/4



#### (2) q>1/4, low connectivity



#### (2) q>1/4, high connectivity



Connectivity inhibits the global cascade, but once it occurs, it facilitates its diffusion.

#### (2) Equilibria for q<q<sub>c</sub>

- Trivial equilibria: all 0 / all 1
- Initial seed applies best-response, hence can switches back. If the dynamic converges, it is an equilibrium.
- Robustness of all 0 equilibrium?
- Initial seed = 2 pivotal neighbors

-> pivotal equilibrium

#### (2) Strength of Equilibria for $q < q_c$

10 Mean number of trials to 6 switch 4 from all 0 2. to pivotal equilibrium 7 2 3 6 1 5 λ

Mean degree

In Contrast with (Montanari, Saberi 10) Their results for q≈1/2

#### (2) Coexistence for $q < q_c$ Size giant 0.4component 0.3 0.2 Connected **Players 0** Players 1 0.1 0 1.3 1.2 1.0 1.1 λ Coexistence

#### (1) Diffusion Model

#### (2) Results

#### (3) Adding Clustering joint work with Emilie Coupechoux

# (3) Simple model with tunable clustering

• Clustering coefficient:

# $C = \frac{3 \text{ number of triangles}}{\text{number of connected triples}}$

• Adding cliques:



#### (3) Contagion threshold with clustering



#### (3) Side effect of clustering!

0.9 0.8 Fraction of 0.7 0.6 pivotal players 0.5 and size of 0.4 0.3 the cascade 0.2 0.1 0 8 2 3 5 6 7 9 Mean Degree Pivotal players in the graph with no clustering Cascade size in the graph with no clustering Pivotal players in the graph with positive clustering Cascade size in the graph with positive clustering

#### Conclusion for (A)

- Simple tractable model:
  - Threshold rule
  - Random network : heterogeneity of population
  - Tunable degree/clustering
- 1 notion: Pivotal Players and 2 regimes:
  - Low connectivity: tipping point / clustering hurts
  - High connectivity: chasm / clustering helps activation
- Open problems:
  - Size of optimal seed? Dynamics of the diffusion? More than 2 states?

#### (A) Diffusion in Social Networks

#### (B) Economics of Information Security

#### (1) Network Security Games

#### (2) Complete Information

(3) Incomplete Information

#### (1) Network Security as a Public Good



#### (1) Local Best Response

# $X_i = \left\{ \begin{array}{ll} 1 & \text{if Secure} \\ 0 & \text{if Not protected} \end{array} \right.$

Weakest link:

$$X_i = \min_{j \sim i} X_j$$

• Best shot:  $X_i = \min_{j \sim i} (1 - X_j)$ 

#### (1) Network Security Games

#### (2) Complete Information

(3) Incomplete Information

#### (2) Complete Information

Weakest link

 $X_i = \min_{j \sim i} X_j$ 

Only trivial equilibria: all 0/ all 1.

• Best shot  
$$X_i = \min_{j \sim i} (1 - X_j)$$











#### (2) Slight extensions WL

- Weakest link:  $X_i = \min_{j \sim i} X_j = \mathbf{1}(\sum_{j \sim i} X_j \ge d_i)$
- Weakest link with parameter K:

$$X_i = \mathbf{1}(\sum_{j \sim i} X_j \ge d_i - K)$$

• Change of variables:

$$Y_i = 1 - X_i = \mathbb{1}(\sum_{j \sim i} Y_j \ge K + 1)$$

Bootstrap percolation!

• Richer structure of equilibria.

#### (2) Slight extensions BS

- Best shot:
- $X_i = \min_{j \sim i} (1 X_j) = \mathbf{1}(\sum_{j \sim i} X_j \le 0)$ 
  - Best shot with parameter K:

$$X_i = \mathbf{1}(\sum_{j \sim i} X_j \le K)$$

 Equilibria: Maximal Independent Set of order K+1. Bramoullé Kranton (07)

#### (2) Best Shot with parameter K=1



#### (1) Network Security Games

#### (2) Complete Information

(3) Incomplete Information

#### (3) Incomplete Information

- Bayesian game. Galeotti et al. (10)
- Type = degree d
- Neighbors' actions are i.i.d. Bern(γ)
- Network externality function:

 $h(\gamma, d) = P(|oss|X_i = 0) - P(|oss|X_i = 1)$ 

corresponds to the price, an agent with degree d is ready to invest in security.

#### (3) Weakest link

• Network externality function:

$$h(\gamma, d) = E[\min_{i=1}^{d} \text{Bern}_{i}(\gamma)] - 0$$
$$= \gamma^{d}$$

- This function being increasing, incentives are aligned in the population.
- This gives a Coordination problem for the game. Lelarge (12)

#### (3) Best shot

• Network externality function:

$$h(\gamma, d) = 1 - E[\max_{i=1}^{d} \operatorname{Bern}_{i}(\gamma)]$$
$$= (1 - \gamma)^{d}$$

- This function being decreasing, incentives are not aligned in the population.
- This gives a Free rider problem for the game. Lelarge (12)

#### Conclusion (B)

- Information structure of the game is crucial: network externality function when incomplete information.
- Technology is not enough! There is a need to design economic incentives to ensure the deployment of security technologies.
- Open problems:
  - Dynamics? mean field games? more quantitative results? Local and global interactions?

#### Thank you!

