
Near Rationality and Competitive Equilibria in Networked Systems

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Background

- Game: Model of interactions between
 - participants (*players*) in a network, who
 - (based on available info) select *strategies*, that
 - yield individual *payoffs* (utilities)
- Nash equilibrium: set of strategies from which no utility-maximizing user has any incentive to deviate

$$\forall \sigma'_i \in \Sigma_i, u_i(\sigma'_i, \sigma^*_{-i}) - u_i(\sigma_i^*, \sigma^*_{-i}) \leq 0$$

Traditional assumptions

- Each player has perfect knowledge of structure of the game
 - Strategies available to all other users
 - Payoffs associated with each strategy
- Each player is perfectly rational
- Players perfectly execute their strategies

*Very stringent assumptions...
... especially in the context of large networks!*

Thesis statement

Slightly expanding the (Nash) solution concept allows to:

- 1) evaluate the robustness of a model to small perturbations*
- 2) help reconcile empirical data with predictions*

For instance, by considering:

$$\underline{\exists \varepsilon > 0} : \forall \sigma'_i \in \Sigma_i, u_i(\sigma'_i, \sigma_{-i}^*) - u_i(\sigma_i^*, \sigma_{-i}^*) \leq \underline{\varepsilon}$$

(sometimes called ε equilibrium)

Case studies

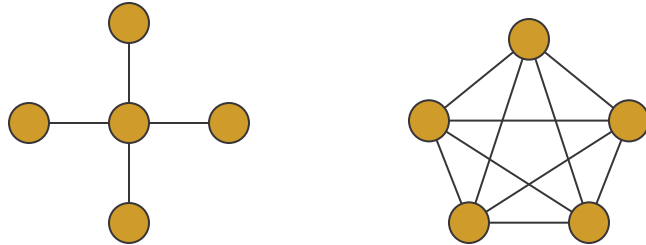
- Three case studies to illustrate our point
 - Network formation
 - Response to a security threat
 - TCP congestion control
- Methodology
 - Describe the game
 - Compare Nash and ϵ -equilibrium outcomes
 - Discuss findings

Case study 1: Network formation

- **Players**
 - Network nodes interested in creating peering connections to other nodes (e.g., ad-hoc network)
- **Utility of a player**
 - Parameterized function of
 - its distance to other nodes (hop count),
 - its connectivity (out degree), and
 - its path loading (number of routes passing through the node)
- **Strategies available to a player**
 - Choose the set of connections maintained with other nodes

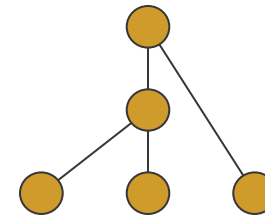
Case study 1: Network formation

■ Nash outcome



Nodes either organize in a star-shaped network, or in a fully connected network (classical result)

■ ε -equilibrium outcome



For a range of parameters (depending on ε), *any* topology forms an ε -equilibrium

■ Findings

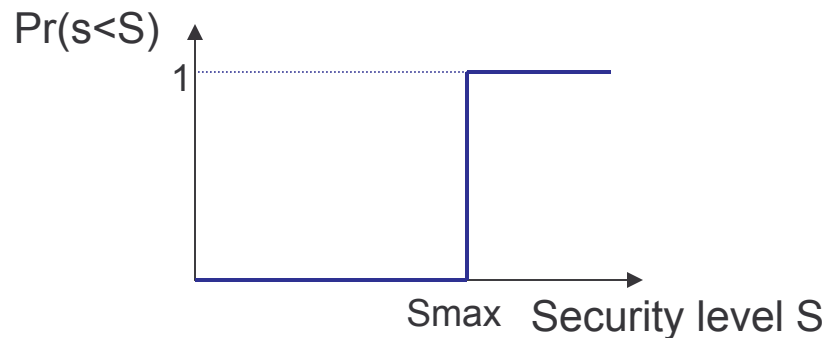
- Even with a small uncertainty ε , the outcome of the model may be significantly different
- Model needs to be refined for complex networks (e.g., the Internet)

Case study 2: Response to security threat

- Simplified model of first stage of a (manual or) semi-automated DDoS attack
- Players
 - Network users, all subject to a security threat of unknown severity
 - The least protected user(s) is (are) compromised
- Utility of a player
 - If not compromised, utility is characterized by a cost function $C(s)$ increasing with their security level s
 - If compromised, utility is characterized by a large penalty $P \gg C(s)$ for any feasible s
- Strategies available to a player
 - Select security level s

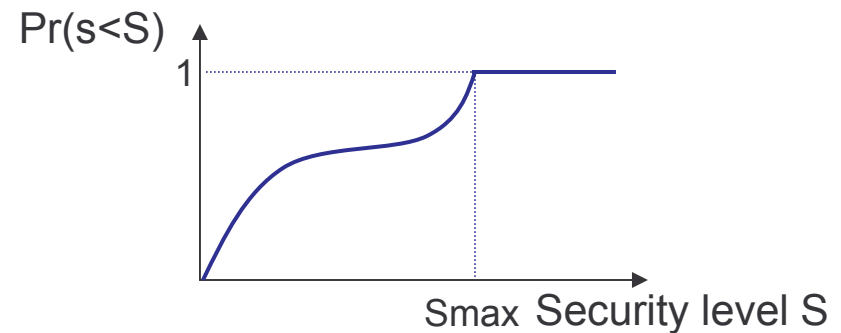
Case study 2: Response to security threat

■ Nash outcome



All users choose the highest security level available

■ ϵ -equilibrium outcome



Dispersion of security levels between no security and highest security

■ Findings

- Empirical data doesn't corroborate predicted Nash outcome
- Simple refinement captures that users are not *perfectly* rational but instead "near rational"

Case study 3: TCP congestion control

- Model initially proposed by [Akella et al. '02]
- Players
 - TCP (SACK) flows sharing bottleneck link(s) in the network
- Utility of a player
 - Throughput (successful transmission rate) obtained by the flow
- Strategies available to a player
 - Change the TCP congestion control parameters
 - additive increase/multiplicative decrease coefficients

Case study 3: TCP congestion control

■ Nash outcome

Congestion control is
completely disabled
(no matter what)

■ ϵ -equilibrium outcome

No change to the default congestion control parameters for large number of flows
(assuming changes to congestion control parameters are bounded over finite time intervals)

■ Findings

- Empirical data doesn't corroborate predicted Nash outcome [Akella et al. '02]
 - ...otherwise the Internet would collapse due to congestion!
- Combining near rationality and model refinement enables us to match observed behavior
- Tweaking TCP stack not worth the hassle

Summary

- Model designers should test (or even allow) for slight deviations from Nash equilibrium (near rationality)
 - Can help assess the robustness of a model
 - Network formation
 - Response to security threat
 - Can help determine if a model is too simplistic
 - Network formation
 - Can help reconcile empirical data with predicted outcome
 - TCP congestion control
 - Response to security threat

Open questions and research agenda

- This is not a plea for using ε -equilibria!
 - Only a convenient (i.e., very simple) tool for our argument
 - Other types of equilibrium (e.g., QRE) potentially useful
 - Model design methodology
 - Accuracy vs. complexity of a model
 - Refine the game definition or the equilibrium concept
 - Mechanism design methodology
 - Impact of near rationality on mechanism performance
 - Application: Security model
 - Refine our model (distinguish between different threats)
 - Gather empirical data (user behavior)
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Questions?