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The Price of Uncertainty in Security Games

Presented by Nicolas Christin at the Eight Workshop on the Economics of Information Security (WEIS 2009). University College London, June 2009.

Motivation

- Lack of good metrics to characterize judicious security investments
 - Marketing pitches vs. defensible metrics
 - Assessing penalties for cybercrime
- Economic models help, but usually assume full rationality and perfect information
- In practice:
 - Limited information due to size and complexity of network
 - Failure to discover optimal strategies
 - Failure to implement the chosen strategies

→ *How valuable is information in the context of security decision making?*

→ *How do we even measure that?*

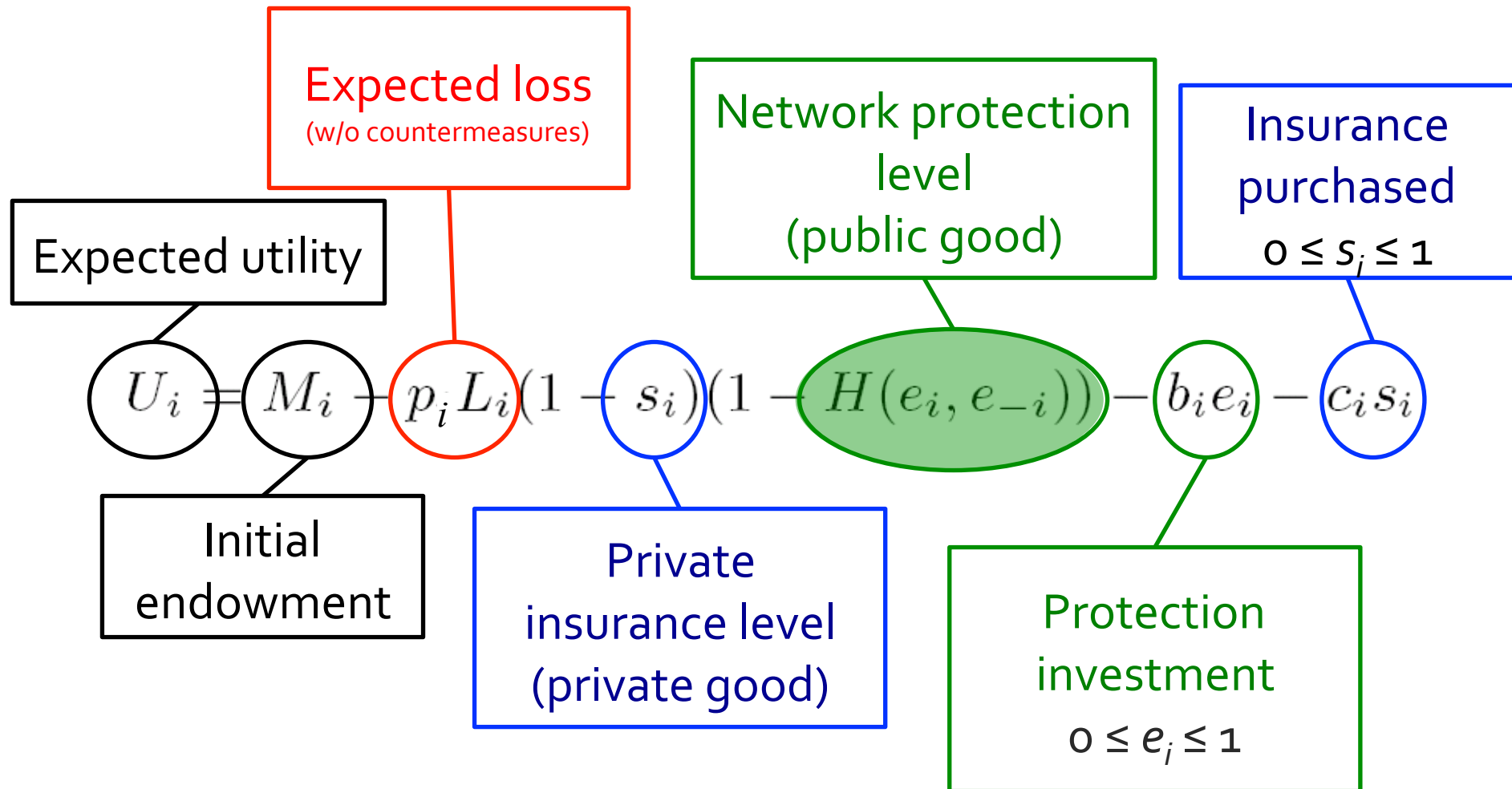
Contribution and Approach

- Propose and evaluate set of metrics to quantify value of information in information security decision-making
- Based on stylized network security games analysis
 - Under different information conditions
 - Under different expertise conditions

Background: Security Models

- Originally proposed in [GCC:WWW'o8, GCC:EC'o8] and presented at last year's WEIS
- Two key components of a security strategy
 - Self-protection (e.g., patching system vulnerabilities)
 - Joint protection level determined by all participants of a network
 - Public good
 - Self-insurance (e.g., having good backups)
 - Individual level of loss reduction
 - Private good

General Utility Model



Different contribution functions

- Weakest-link: $H(e_i, e_{-i}) = \min(e_i, e_{-i})$
 - Example: corporate network protection
 - $U_i = M_i - p_i L_i (1 - s_i) (1 - \min(e_i, e_{-i})) - b_i e_i - c_i s_i$
- Best shot: $H(e_i, e_{-i}) = \max(e_i, e_{-i})$
 - Example: Censorship resilient networks (see: Tor)
 - $U_i = M_i - p_i L_i (1 - s_i) (1 - \max(e_i, e_{-i})) - b_i e_i - c_i s_i$
- Total effort: $H(e_i, e_{-i}) = \frac{1}{N} \sum_i e_i$
 - Example: Peer-to-peer (swarming) transfers (see: BitTorrent)
 - $U_i = M_i - p_i L_i (1 - s_i) (1 - \frac{1}{N} \sum_k e_k) - b_i e_i - c_i s_i$

Uncertainty

- Expected losses may differ among players.
- Expected losses for other players may be unknown.
 - We assume that all expected losses are UID (uniformly and independently distributed) in $[0, L]$.
- Some players may not take into account the expected losses of others.

Information Conditions

- Complete Information
 - You know all players' expected losses, including your own. E.g., (weakest link):
 - $$U_i = M - p_i L \left(1 - \min_{j=1}^N e_j\right) (1 - s_i) - b e_i - c s_i$$
- Incomplete Information
 - You know you own expected loss but not others'. You know the distribution. E.g.,
 - $$U_i = M - p_i L \left(1 - E\left(\min_{j=1}^N e_j\right)\right) (1 - s_i) - b e_i - c s_i$$

A Mixed Economy

- *One expert player* acts strategically based on all available information.
- *All other players* choose levels of protection and insurance based on a straightforward cost-benefit analysis, ignoring behavior of others.

- perceived utility:

$$U_i = M - p_j L(1 - e_j)(1 - s_j) - be_j - cs_j$$

- actual utility:

$$U_i = M - p_j L \left(1 - \min_{k=1}^N e_k\right) (1 - s_j) - be_j - cs_j$$

Methodology

- For each information condition: complete and incomplete
 - Compute an expected utility for the expert player
 - Expert player's strategy: best-response to the behavior of the naive players.
- We take an additional expected value over all attack probabilities
 - Leave the final "expected utility" as a function of parameters known under incomplete information.

Price of Uncertainty

- Goal: measure how much uncertainty costs an expert player
 - Quantify a payoff differential between full information condition and limited information condition
 - Payoff depend on 5 parameters: initial endowment M , cost of protection b and cost of insurance c , number of players N , and magnitude of losses L
 - Need to reduce the number of parameters through the definition of the metric
- Three possible metrics
 - Difference metric
 - Payoff-ratio metric
 - Cost-ratio metric

Payoff Difference Metric

$$\max_{b,c \in [0,L]} [\text{Expected Payoff Complete}(b, c, L, L, N) - \text{Expected Payoff Incomplete}(b, c, L, L, N)]$$

- Worst-case difference in payoff between complete and incomplete information
 - Maximum taken over all possible prices for protection and insurance
- An insignificant price of uncertainty yields an output of zero
- The metric's output increases w/ the significance of the price of uncertainty

Payoff Ratio Metric

$$\max_{b,c \in [0,L]} \left[\frac{\text{Expected Payoff Complete}(b, c, L, L, N)}{\text{Expected Payoff Incomplete}(b, c, L, L, N)} \right]$$

- Somewhat analogous to “price of anarchy”
 - payoff-ratio of a game’s socially optimal equilibrium to its worst case Nash equilibrium
- Currency independent
- An insignificant price of uncertainty yields an output of one
- The metric’s output increases w/ the significance of the price of uncertainty

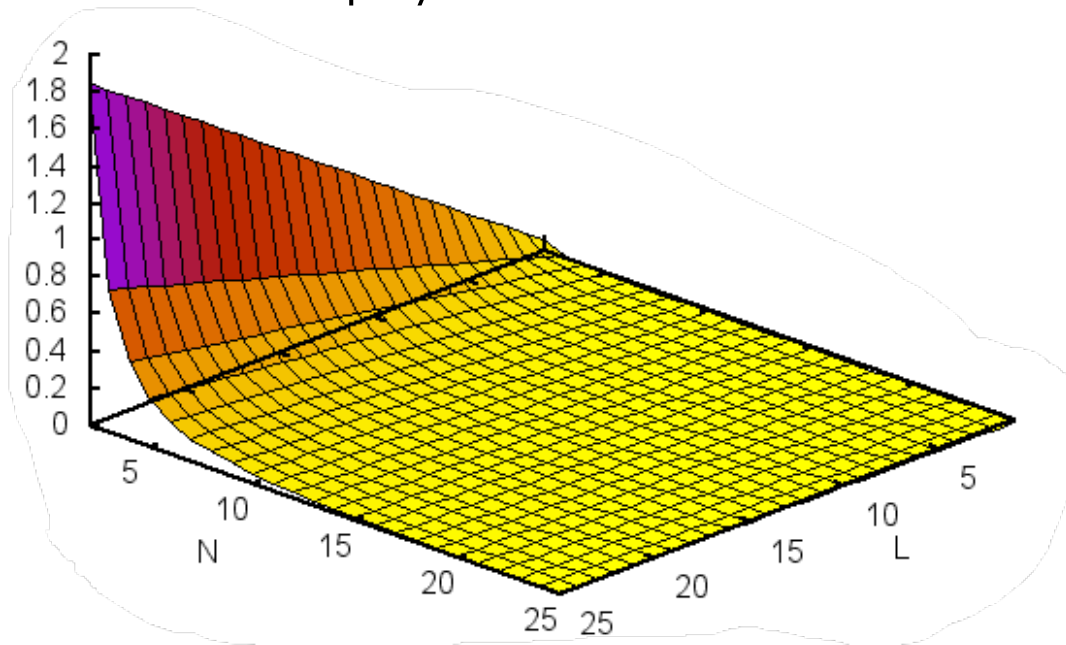
Cost Ratio Metric

$$\min_{b,c \in [0,L]} \left[\frac{\text{Expected Payoff Complete}(b, c, L, 0, N)}{\text{Expected Payoff Incomplete}(b, c, L, 0, N)} \right]$$

- Similar to the payoff-ratio metric, but with a different canonical choice of zero for the initial endowment M
 - Simpler algebraic analysis due to an abundance of term cancellations
- An insignificant price of uncertainty yields an output of one
- The metric's output *decreases to zero* w/ the significance of the price of uncertainty

Best Shot, Payoff Difference

Best-shot: Payoff difference as a function of number of players N and losses L

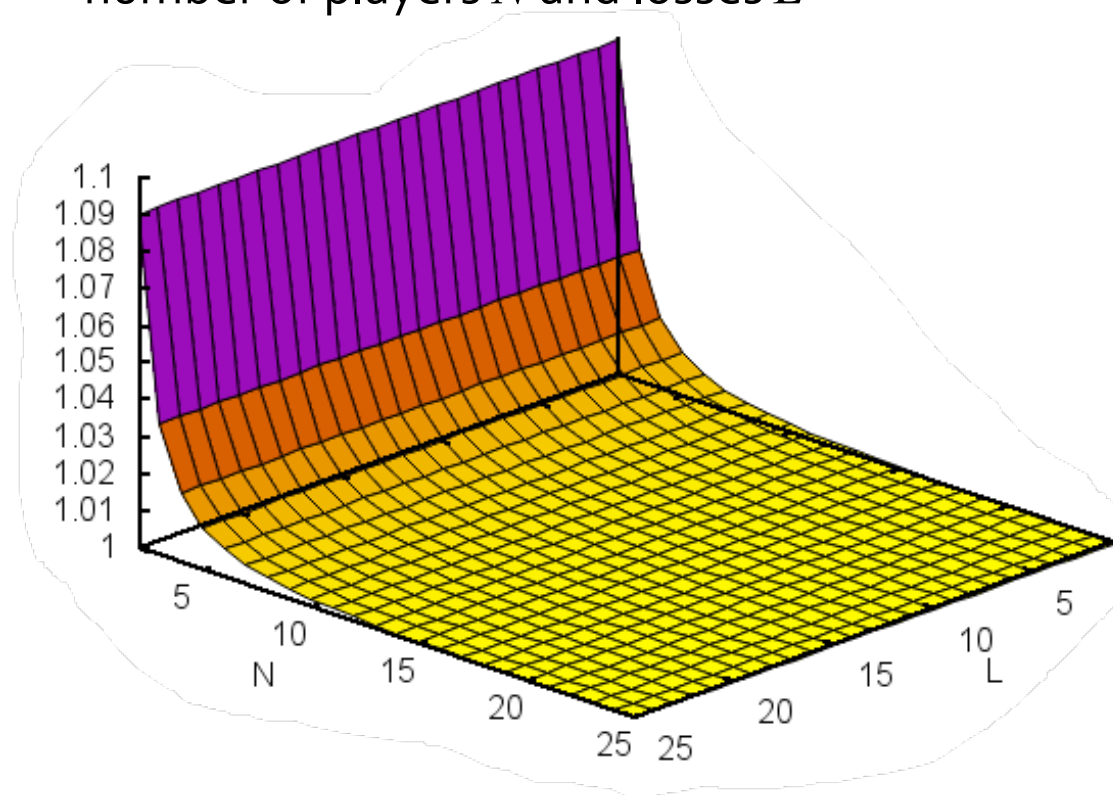


(note: the paper also contains plots for the maximizing (cost of protection, cost of insurance) pairs)

- Payoff difference increases with the potential losses
- Payoff difference decreases when the number of players increases
 - Unless losses are in $L \approx O(N^2)$

Best Shot, Payoff Ratio

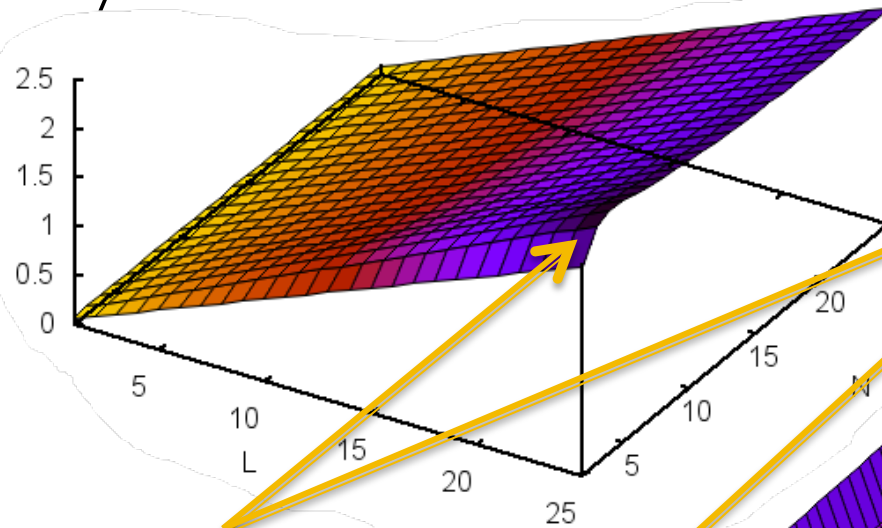
Best-shot: Payoff ratio as a function of number of players N and losses L



- Payoff ratio independent of L
- Payoff ratio decreases when the number of players increases
- Fairly insignificant overall!
 - 10% at most
- Not shown here: cost ratio metric *always* equal to zero! (significant?!)

Weakest-Link Game

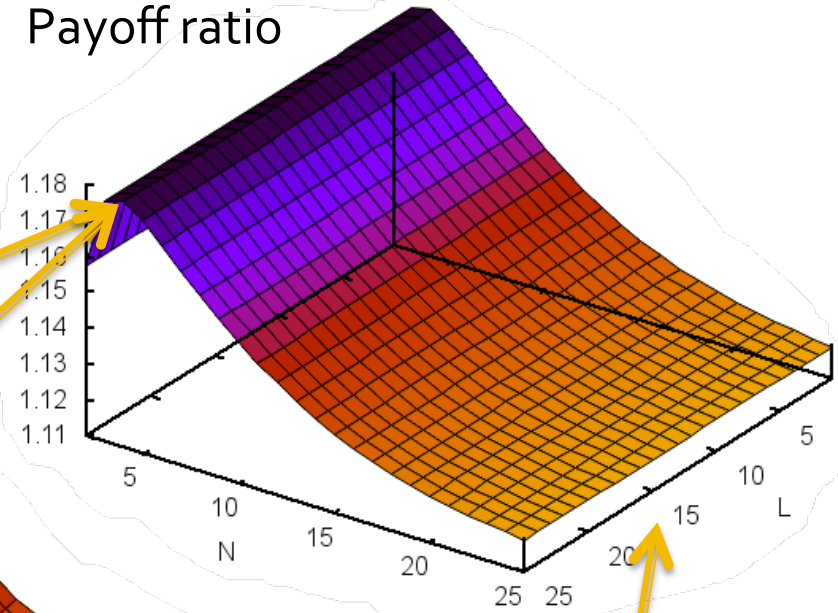
Payoff difference



Highest value for 4 players

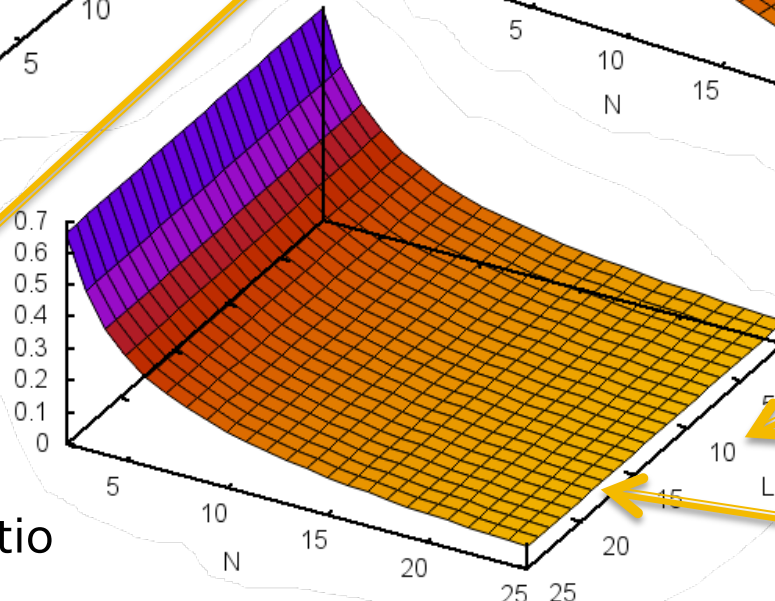
Slightly more significant (18%), but not catastrophic

Payoff ratio



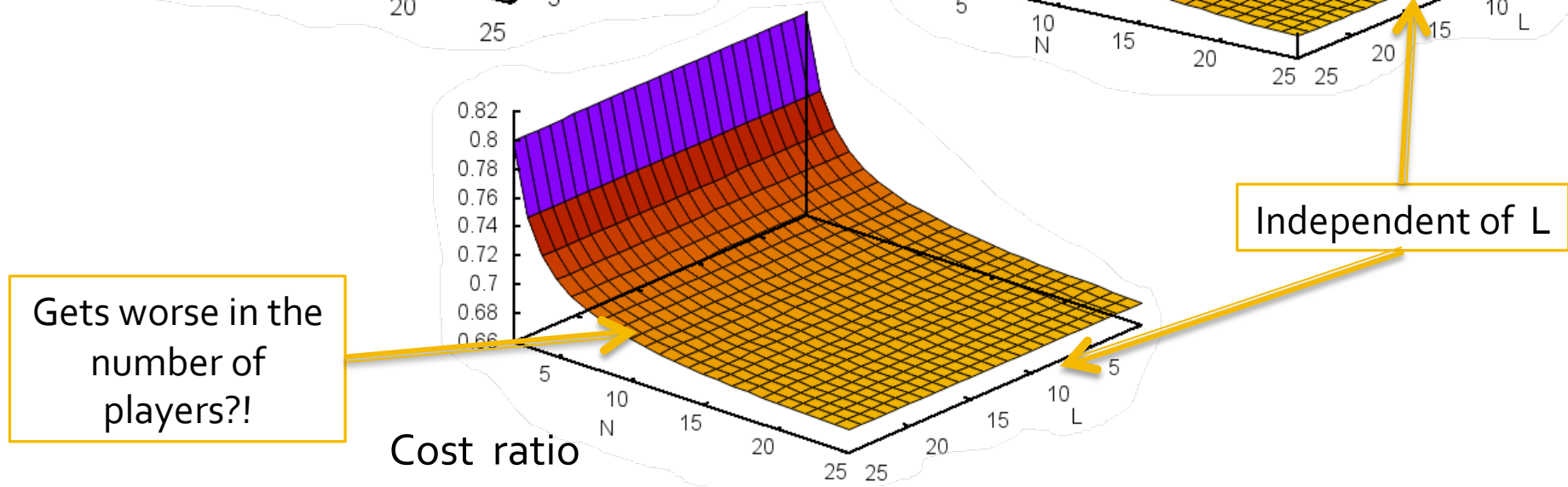
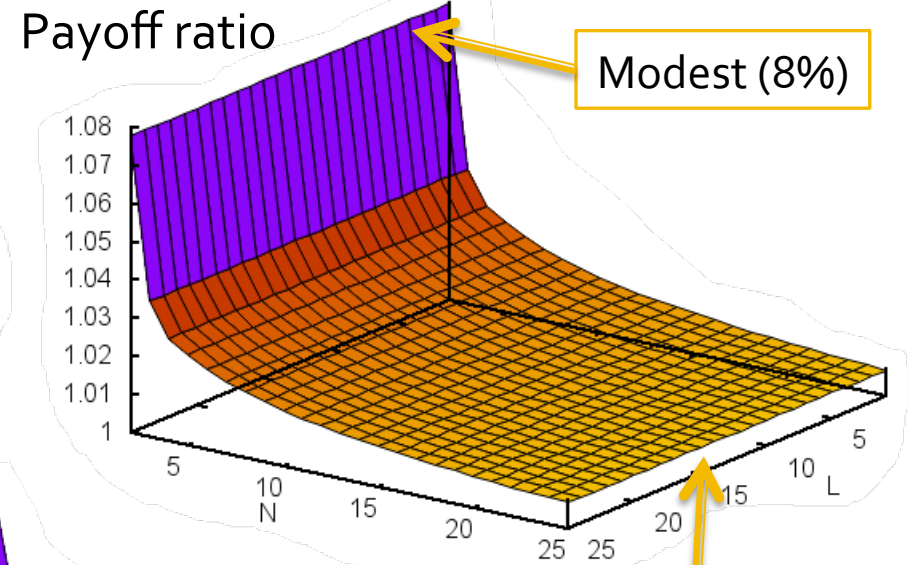
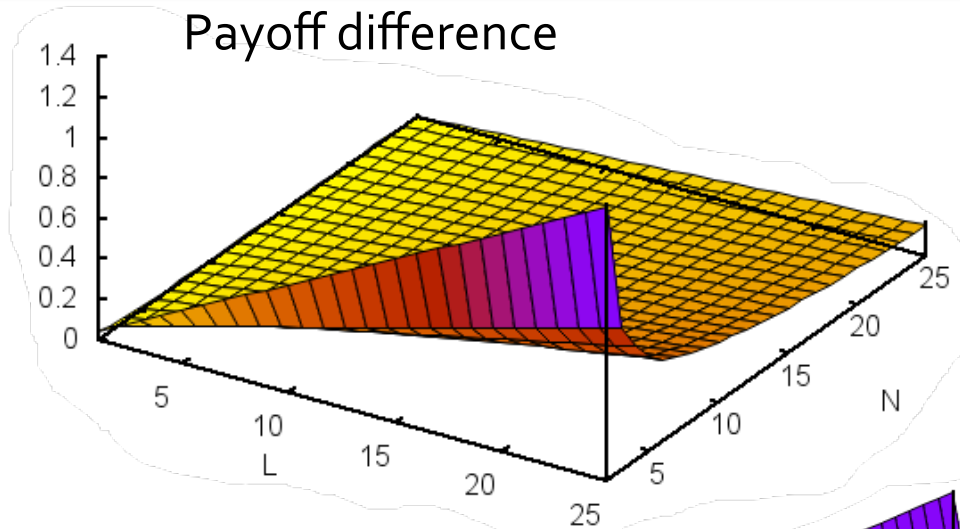
Independent of L

Cost ratio



Goes to zero?!

Total Effort Game



Finding: Cost Ratio Is Harmful

- Cost ratio metric always inappropriate in all three scenarios
 - Computing ratios of *very* small quantities
 - A penny divided by a dime yields a 0.1... (remember, going to zero is worse)
 - ... but is not characteristic of large costs!
 - The fact we are dealing with very small quantities is more important
- Behavioral research has shown robust evidence for consumers' preferences for benefits that are presented as large ratios in comparison to small ratios
 - Useful for marketing snake oil, but not for much else

Finding: Uncertainty vs. Expertise

- All metrics show that uncertainty does not significantly penalizes an expert player
- The **more players** in a network, the **less uncertainty** matters
- **Naïve** strategies have a significantly more disastrous impact on payoffs
 - Not shown today
 - Please see paper and related, companion technical report CMU-CyLab-2009-04

Questions?

The Price of Uncertainty in Security Games

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Related papers:

<http://www.andrew.cmu.edu/user/nicolasc/papers-topic.html>

1. Security and Insurance Management in Networks with Heterogeneous Agents [ACM EC'08]
2. Secure or Insure? A Game-Theoretic Analysis of Information Security Games. [WWW'08]
3. Predicted and Observed User Behavior in the Weakest-Link Security Game. [USENIX UPSEC'08]