



A Decision-Analytic Approach for P2P Cooperation Policy Setting

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Outline

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Our Motivation & Goal

- Overall performance of P2P systems depends on resource contributions of individual peers.
- Rational peers decide on their cooperation policies according to their **individual** utilities.
- **Inherent conflict** among individual utilities of the rational peers results in
 - free-riding
 - unfair contribution
 - low participation
- Our goal is dealing with the inherent individual utility conflicts to improve **overall performance** of the system.

Our Approach

- We employ **decision-theory** to model cooperation policy setting of participating peers:
 - Each peer chooses its strategy according to **observable** strategies of the other peers.
 - Through a swarm-based iterative learning process:
 - Rational peers set their cooperation policies so as to maximize their own utility.
 - Their decisions are coordinated in a distributed manner to improve the social welfare of the system.
- The game-theoretic analysis lacks an **explicit** and **tractable** handling of the individual strategy dynamics present in the interactions among individual peers



SYSTEM MODEL

Individual-based Lagrangian Swarm Model

- Interacting participants of a P2P system exhibit general properties of an individual based Lagrangian swarm model:
 - composed of many individual peers;
 - the interactions are based on local information exchange;
 - emergence;
 - self-organization.
- We made two modifications to adopt this model in the context of a P2P system:
 - Distributed local objectives (utility functions) are defined for individual peers.
 - The interaction of particles is represented as a non-cooperative game.

Definitions

- We assume that N peers $p_i ; i:1, \dots, N$ participate in the system
- Policy (d_i)
 - a peer's policy is its level of cooperation (a numerical assessment of the peer's contributed resources to the system)
- Strategy (s_i)
 - the strategy of a peer reflects its decision on the change in its cooperation level (policy)
- Utility (U_i)
 - A peer's utility is determined by its strategy choices and depends on several parameters - discussed as follows.

Utility Function

- Cost and Benefit

- the total cost for participating in the system with cooperation level of d_i will be $c_i d_i$
- the benefit of cooperation of p_j to p_i is represented by $b_{ij} d_j$; where b_{ij} is measured (e.g.) as the inverse of latency

- Incentives for high contribution

- it is modeled by a monotonically increasing function of the cooperation policy of a peer p_j , denoted by bc_j

- Utility:

$$U_i = bc_i \cdot \sum_{j \in N} b_{ij} \cdot d_j - c_i d_i \quad ; b_{ii} \equiv 0$$



DECISION-ANALYTIC APPROACH

Overall

- Observable strategies of other peers are monitored by each peer in a sequence of iterations.
- Based on this empirical evidence, each peer can decide rationally on a strategy in every iteration.
- This chain of decisions are made based on a method inspired by particle swarm optimization (PSO).
- Through this chain of decisions each participating peer concludes its final cooperation policy with respect to the other peers' behavior.

More Formally

- To maximize its expected utility U_i , each peer p_i sets its final cooperation policy through an iterative decision making process:
 - p_i monitors the strategies of the other peers in its neighborhood N_i locally and evaluates their strategies.
 - It chooses its strategy s_i^{next} in the next iteration with respect to the evaluation result and to its own experience:

$$s_i^{next} = s_i^{current} + r_1 c_1 (d_p - d_i^{current}) + r_2 c_2 (d_n - d_i^{current})$$

- d_p is the best previous policy of p_i and d_n denotes the best policy of the other peers in N_i .
- Then the cooperation policy d_i of peer p_i is revised as follows:

$$d_i^{next} = d_i^{current} + s_i^{next}$$



ANALYSIS - EVALUATION

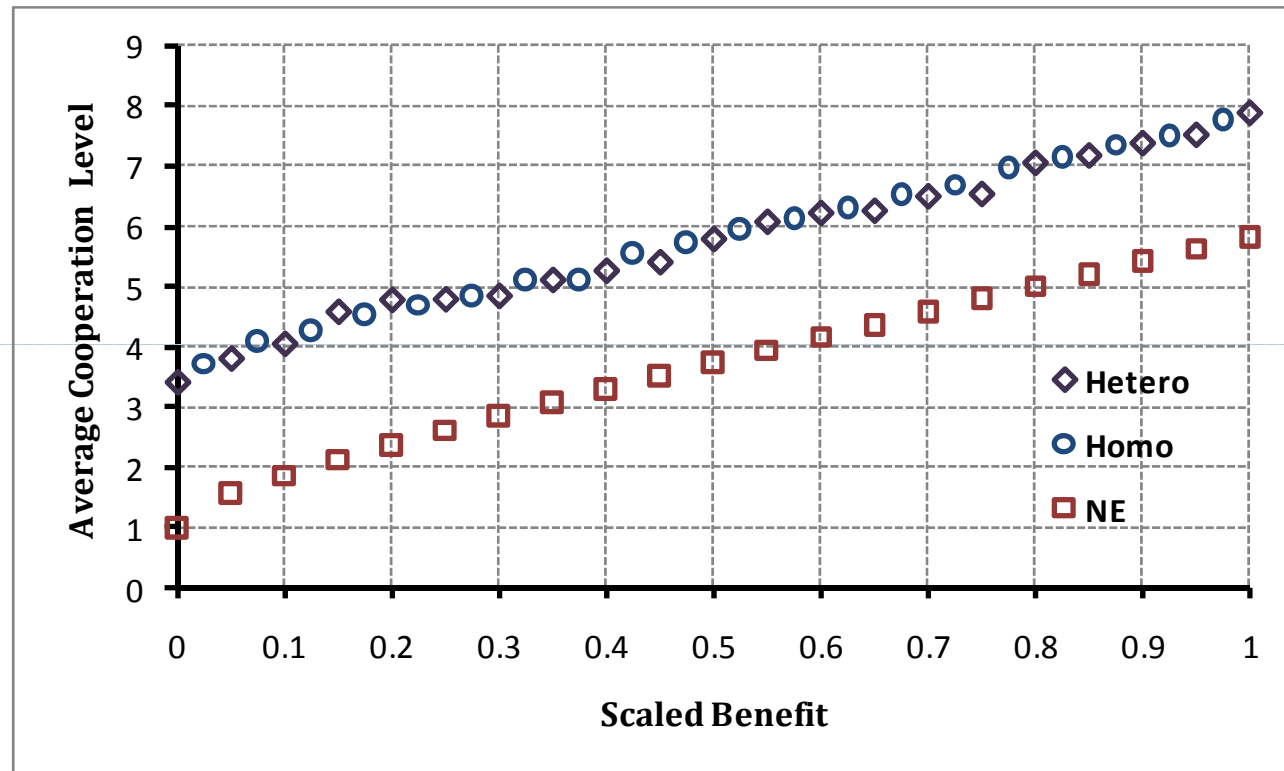
NE Analysis

- We employ Nash equilibrium analysis to investigate the predicted strategies for the participating peers by the decision-analytic approach.
- According to [Buragohain et al. P2PComputing03] for a similar quantitative model of the system in a homogeneous setting (for all $p_i, b_{ij} = b, c_i = c$), the NE is given by:

$$d^* = (b(N-1)/2c - 1) \pm ((b(N-1)/2c - 1)^2 - 1)^{1/2}$$

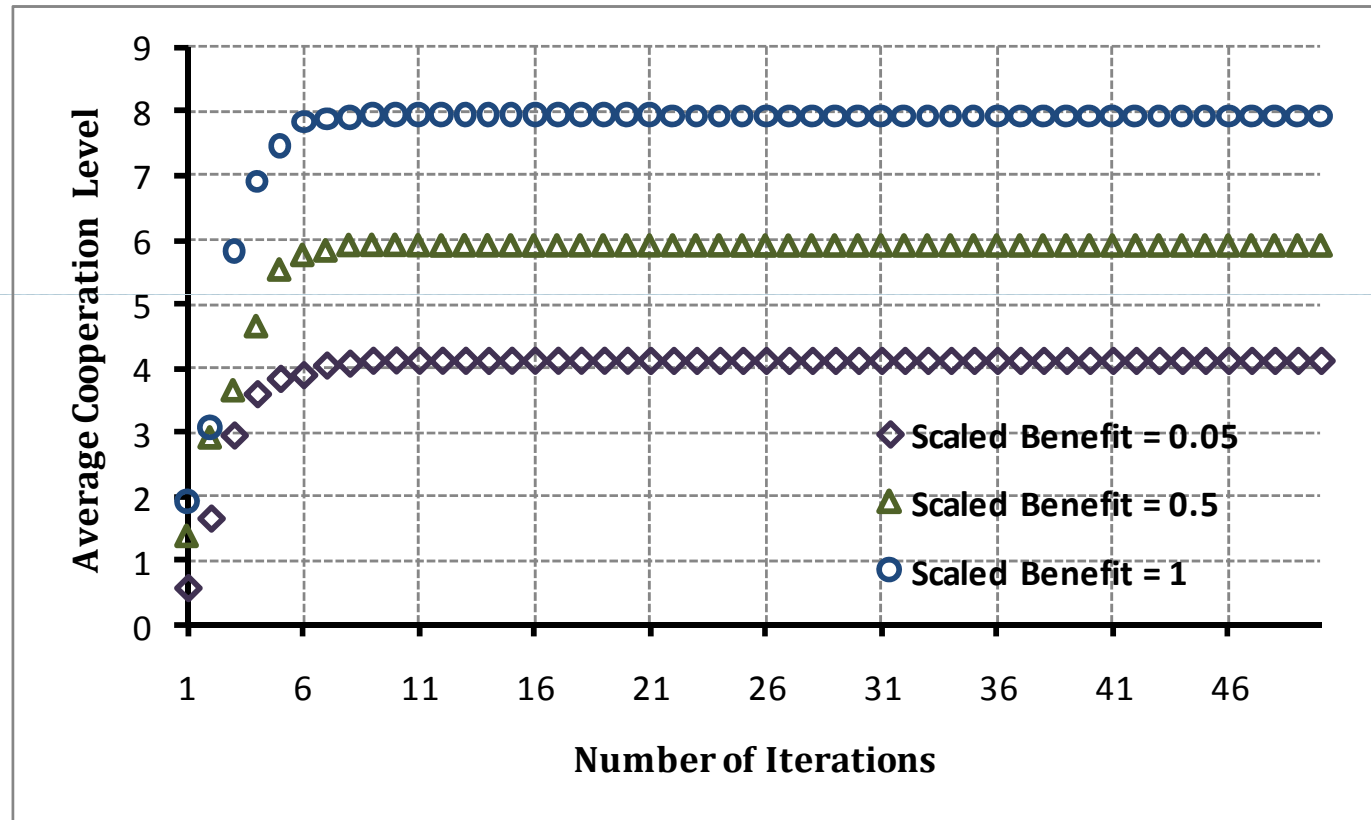
- As we numerically show:
 - The expected NE of the game is not the Pareto-optimal one.
 - The outcome derived from the proposed decision-analytic approach would make all players better-off.

The comparison of the average cooperation level



- Tendency toward Pareto efficiency
- Better outcome than NE
- Both homogeneous and heterogeneous settings evolve similarly

Convergence to a set of Pareto efficient strategy



- Fast convergence regardless of the target cooperation level



CONCLUSION

Conclusion – Future Work

- We propose a decision-analytic approach based on the modified swarm model, to set and coordinate rational decisions of the individual peers on their cooperation policies in a distributed manner.
- The resulting cooperation policies constitute the final set of decisions that maximize rational peers' utility in-line with the **social welfare** of the system.
 - Incentive-compatible for peers to follow
- Our approach quickly approximates a **Pareto-optimal** operating point of the system.
- In our future work, we will investigate information exchange mechanisms that involve incentives for neighbor truthfulness or own observation and verification.

**THANK YOU FOR YOUR
ATTENTION.
MORE QUESTIONS TO:**

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