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Computational economics for self-organizing networks

Reputation for autonomic systems

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Computational Economics

- Game Theory (GT) and Mechanism Design (MD) with an **algorithmic complexity** flavor
- Focus of this presentation:
 - Algorithms and protocols concerned with **self-organizing** communication networks
 - System designers' perspective
- Analytical framework
 - Informal background information
 - Application to real systems

Motivations

- Fundamental assumption made in the past:
 - Entities correctly execute protocols/algorithms, except for faulty or malicious ones
- However, the system designer has no control on individuals' (software agents, humans ...) behavior
 - Agents can/will **deviate from prescribed behavior, if beneficial for them**
- Opportunistic/selfish behavior **prevent** from the realization of system **designers' objectives**

Game Theory: traditional approach

- Framework to **model** systems and participants' behavior
- Definition of a game:
 - Set of players
 - Preference ordering over system outcomes
 - Set of available strategies
- **Rationality** and **Self-interest** : individuals are able to predict and evaluate the outcome of a game, and maximize their gains

GT: traditional approach (2)

- **Utility Function**: represents the (real) system
- Solution concepts: Dominant strategy, Nash/Bayesian
 - Prove of existence is hard, finding the equilibrium point requires many assumptions
- GOAL 1: find the **system operating point**
- GOAL 2: study **strategies**, their stability properties, repeated interactions

Mechanism Design: traditional approach

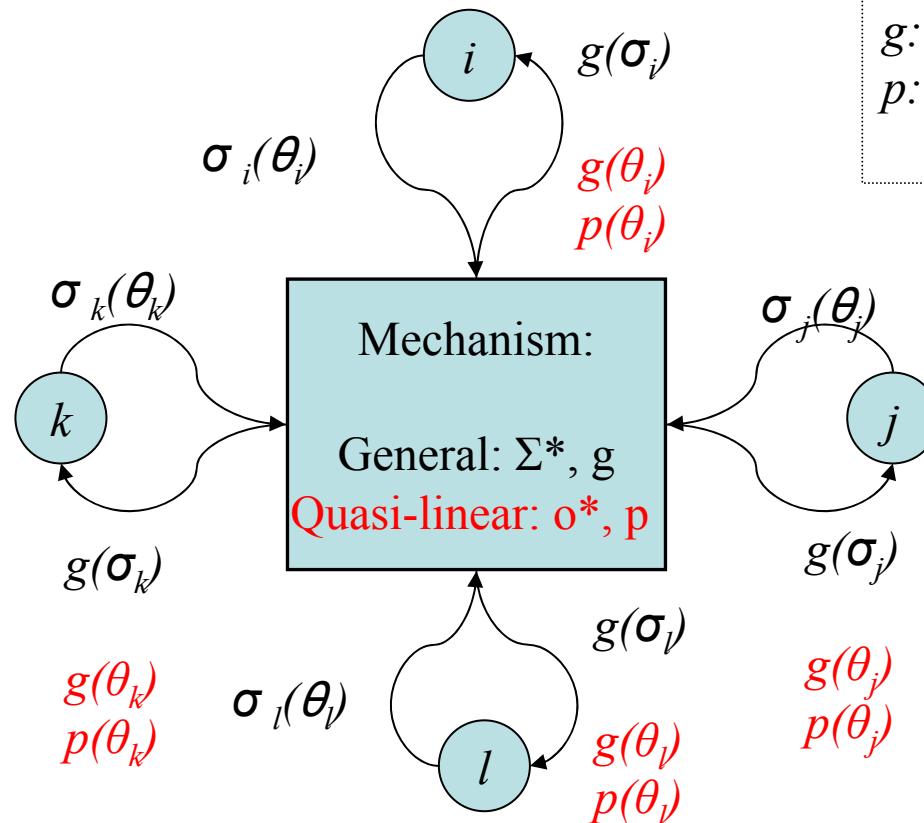
- How to design systems so that individuals' selfish behavior results in the desired system-wide goals
 - System objectives termed **Social Choice Function**:
e.g., efficiency (system welfare), stability, and fairness
- GT models the system and the entities
 - Equilibrium solution concepts borrowed from GT
- Individuals i hold **private information** about their preferences, termed *type* θ_i
 - Selfishness = players **lie** about their type

Traditional MD at a glance

- Quasi-linear utility function: $u_i(\theta_i, o) = v_i(\theta_i, o^*) + p_i$
- **Incentive & Allocation** (that implements the SCF)
 - Allocation example: $g(\theta) = o^* = \operatorname{argmax} (\operatorname{Sum} v_i(\theta, o)) = \operatorname{SCF}(\theta)$
 - Payment example: pay as much as a cheater would gain
- Desirable properties:
 - **Truthfulness / strategy proofness**: truth revelation is dominant
 - **Incentive compatibility**: system outcome = social function

Traditional MD: a sketch

- σ, θ
strategies
- i *individuals*
- *Mechanism*
= center



σ, θ : reports by agents

g : outcome rule
 p : payment rule

Traditional MD: goals

- Guide the system operating point to an optimal position, from the designer's perspective
- Formulate and study incentives or payment schemes to stimulate participants to behave as prescribed by the original protocol/algorithm, assuming selfish behavior

From traditional MD to Algorithmic MD to Distributed MD (1)

- Limitations of traditional MD
 - Computationally **unbounded** agents
 - **Centralized** approach (the mechanism, the center)
 - Cost-free, error-free communication with the center
- Limitations that derive from GT
 - Solution concepts often too “demanding”
 - **No repeated** interactions

From traditional MD to Algorithmic MD to Distributed MD (2)

Algorithmic MD (AMD)

- Considers computational complexity of MD
 - Agent level: valuation and strategic complexity
 - Infrastructure level: outcome computation and communication complexity
- Approaches:
 - Polynomial time solutions (Nisan *et. al.*)
 - **Protocol compatibility** (Feigenbaum, Shenker)

From traditional MD to Algorithmic MD to Distributed MD (3)

Distributed (Algorithmic) MD

- Eliminates the need for a computational center
 - Agents compute the outcome rule and **payments**
- Theory is still in its infancy, many open problems:
 - Communication security
 - Privacy issues
 - Self-sustaining systems issues: **budget balance is not always guaranteed**

Recent advances

- **Automatic Mechanism Design**
 - Consider complexity of mechanism definition
 - Deals with the limited (a couple!!) number of available mechanism/incentive schemes and their limitations
 - Quasi-linear utilities
 - No payments
 - Different social choice functions (SCF, defined by the system designer)
- **Interaction with Security and Cryptography**
 - Cryptography has emerged as a fundamental tool, especially in distributed settings
 - Reputation/trust has emerged as a useful metric to prioritize interaction
- Towards **autonomic systems**: the designers' role is still preponderant in all analytical models, because of the SCF
 - Let the system participants dynamically define the SCF

Application: Mobile ad hoc networks

- Prescribed behavior: forward packets for others
- Problem: forwarding cost energy, which is scarce
- Nodes deviate from prescribed behavior

- Game theoretical models: **equilibrium = no cooperation**
 - Prisoner's Dilemma
 - Cyclic dependent networks

- Solution: incentive schemes or (?) reputation
- **Cooperation strategies**: TFT, G-TFT, CORE, ...
- *Evolutionary game theory*: blends game theory and genetic programming

Application: Mobile ad hoc networks (Continued)

- Application of mechanism design: Ad hoc-VCG
 - Node **declare its cost** for forwarding a packet
 - **Node can lie** to be on the shortest path or to be excluded from it
- Payment for an intermediate node on the shortest path from a source S to a destination D = declared cost for forwarding a packet plus a premium
- Premium = the difference of the total cost of the shortest path from S to D without v_i and the shortest path from S to D with v_i (and its declared cost)
- The payment is defined in such a way that node v_i will get the same amount independent of what it declares as its forwarding cost.
- Budget balance problem
- Centralized solution`

Application: Content Distribution

- Prescribed behavior: share upload capacity
- Problem: limited resource, **free-riding**

- Game theoretical models
 - Marriage theorem
 - Bipartite matching
 - Network formation games

- Practical instance: BitTorrent, **peer selection strategy**

Conclusion

- Powerful analytical frameworks
- From theory to practice is often an open problem
 - Many impossibility results!
- Question: what is the difference between reputation in MANET and peer selection in P2P?
 - Hint: “**shadow of the future**”

MD Example: VCG example

- Second-best sealed bid auction
 - Quasi-linear utility
 - Valuation function over (single) item allocation
- Players **reveal their type** to the center (*i.e.* mechanism)
 - But, don't need to know other players' values/types
- Center **computes system allocation** to maximize SCF
 - Center allocates item to highest bid
 - Winner pays the **second highest price**
- Truthful revelation is dominant
- Not budget-balanced

Generalized VCG, formal definition

- Center collects $\theta = (\theta_1, \dots, \theta_n)$ from agents
- $g(\theta)$: select o^* to maximize $\sum_i v_i(\theta, o)$
- $p(\theta)$: agent i pays $\sum_{j \neq i} (v_j(\theta_j, o^{-i}) - v_j(\theta_j, o^*))$
where
 o^{-i} solves $\max_a \sum_{j \neq i} (v_j(\theta_j, o))$

- Theorem: the VCG mechanism is truthful and implements the SCF