▲□▶ ▲□▶ ▲ 三▶ ▲ 三▶ - 三 - のへぐ

Efficient Mechanism Design Bandwidth Allocation in Computer Network

Presenter: Hao MA

Game Theory Course Presentation April 1st, 2014

Summary

▲□▶ ▲□▶ ▲□▶ ▲□▶ = 三 のへで

Efficient Mechanism Design

Efficient Mechanism Design focus on the mechanism that lead to efficient allocation!

Quick-fire Question

Summary



Price of Anarchy?



◆□▶ ◆□▶ ▲□▶ ▲□▶ ■ ののの



Price of Anarchy (PoA): PoA is a measure of the extent to which system efficiency degrades due to selfish behaviour of its agents.

Define *s* as a strategy profile, *S* as the set of all strategy profiles and $E \subseteq S$ is the set of strategies in equilibrium.

For Welfare function W / Cost function C.

$$PoA = \frac{\max_{s \in S} W(s)}{\min_{s \in E} W(s)} = \frac{\max_{s \in E} C(s)}{\min_{s \in S} C(s)}$$

Note: $PoA \ge 1$, and the smaller, the better.

◆□▶ ◆□▶ ▲□▶ ▲□▶ □ のQ@



Price of Anarchy (PoA): PoA is a measure of the extent to which system efficiency degrades due to selfish behaviour of its agents.

Define *s* as a strategy profile, *S* as the set of all strategy profiles and $E \subseteq S$ is the set of strategies in equilibrium.

For Welfare function W / Cost function C.

$$PoA = \frac{\max_{s \in S} W(s)}{\min_{s \in E} W(s)} = \frac{\max_{s \in E} C(s)}{\min_{s \in S} C(s)}$$

Note: $PoA \ge 1$, and the smaller, the better.

▲□▶ ▲□▶ ▲□▶ ▲□▶ ▲□ ● のへぐ

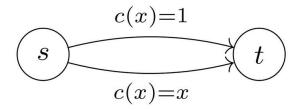


Figure : Pigou's example: selfish routing problem.



Name: Steve

Position: CEO of a big Internet Provider

Personality:

- Cares more about the **best use** of network resources (efficient allocation) than **money** in his pocket (revenue maximization)
- No price discrimination, and charge each user **the same price** for network resource per unit.



Name: Steve

Position: CEO of a big Internet Provider

Personality:

- Cares more about the **best use** of network resources (efficient allocation) than **money** in his pocket (revenue maximization)
- No price discrimination, and charge each user **the same price** for network resource per unit.



Name: Steve

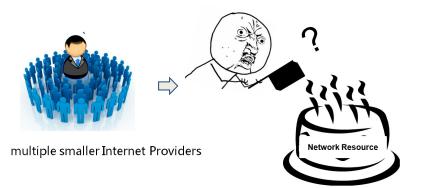
Position: CEO of a big Internet Provider

Personality:

- Cares more about the **best use** of network resources (efficient allocation) than **money** in his pocket (revenue maximization)
- No price discrimination, and charge each user **the same price** for network resource per unit.

◆□▶ ◆□▶ ◆ □▶ ◆ □▶ ─ □ ─ の < @





- A communciation link of capacity C > 0
- R users
- User *r* get capacity *d_r*.
- User *r* receives a utility $U_r(d_r)$
- *U_r(d_r)* is concave, strictly increasing and continuously differentiable with domain *d_r* > 0

Given a complete knowledge and centralized control of the system, the optimization problem becomes

$$\begin{array}{l} \textit{maximize } \sum_{r} U_r(d_r) \qquad \qquad (1) \\ \textit{subject to} : \sum_{r} d_r \leq C; \\ d_r \geq 0, r = 1, ..., R. \end{array}$$

▲□▶ ▲□▶ ▲ 三▶ ▲ 三▶ - 三■ - のへぐ

• A communciation link of capacity C > 0

R users

- User r get capacity d_r .
- User *r* receives a utility $U_r(d_r)$
- *U_r(d_r)* is concave, strictly increasing and continuously differentiable with domain *d_r* > 0

Given a complete knowledge and centralized control of the system, the optimization problem becomes

$$\begin{array}{l} \textit{maximize } \sum_{r} U_r(d_r) \qquad \qquad (1)\\ \textit{subject to}: \sum_{r} d_r \leq C;\\ d_r \geq 0, r=1,...,R. \end{array}$$

- A communciation link of capacity C > 0
- R users
- User *r* get capacity *d_r*.
- User *r* receives a utility $U_r(d_r)$
- *U_r(d_r)* is concave, strictly increasing and continuously differentiable with domain *d_r* > 0

Given a complete knowledge and centralized control of the system, the optimization problem becomes

$$\begin{array}{l} \textit{maximize } \sum_{r} U_r(d_r) \qquad \qquad (1) \\ \textit{subject to} : \sum_{r} d_r \leq C; \\ d_r \geq 0, r = 1, ..., R. \end{array}$$

- A communciation link of capacity C > 0
- R users
- User r get capacity d_r.
- User *r* receives a utility $U_r(d_r)$
- *U_r(d_r)* is concave, strictly increasing and continuously differentiable with domain *d_r* > 0

Given a complete knowledge and centralized control of the system, the optimization problem becomes

$$\begin{array}{l} \textit{maximize } \sum_{r} U_r(d_r) \qquad \qquad (1) \\ \textit{subject to} : \sum_{r} d_r \leq C; \\ d_r \geq 0, r = 1, ..., R. \end{array}$$

▲□▶ ▲□▶ ▲ 三▶ ▲ 三▶ - 三■ - のへぐ

- A communciation link of capacity *C* > 0
- R users
- User *r* get capacity *d_r*.
- User *r* receives a utility $U_r(d_r)$
- *U_r(d_r)* is concave, strictly increasing and continuously differentiable with domain *d_r* > 0

Given a complete knowledge and centralized control of the system, the optimization problem becomes

$$\begin{array}{l} \textit{maximize } \sum_{r} U_r(d_r) \qquad \qquad (1) \\ \textit{subject to} : \sum_{r} d_r \leq C; \\ d_r \geq 0, r = 1, ..., R. \end{array}$$

- A communciation link of capacity C > 0
- R users
- User *r* get capacity *d_r*.
- User *r* receives a utility $U_r(d_r)$
- *U_r(d_r)* is concave, strictly increasing and continuously differentiable with domain *d_r* > 0

Given a complete knowledge and centralized control of the system, the optimization problem becomes

$$\begin{array}{l} \textit{maximize } \sum_{r} U_r(d_r) & (1) \\ \textit{subject to} : \sum_{r} d_r \leq C; \\ d_r \geq 0, r = 1, ..., R. \end{array}$$

- A communciation link of capacity C > 0
- R users
- User *r* get capacity *d_r*.
- User *r* receives a utility $U_r(d_r)$
- *U_r(d_r)* is concave, strictly increasing and continuously differentiable with domain *d_r* > 0

Given a complete knowledge and centralized control of the system, the optimization problem becomes

$$\begin{array}{l} \textit{maximize } \sum_{r} U_{r}(d_{r}) \qquad \qquad (1) \\ \textit{subject to} : \sum_{r} d_{r} \leq C; \\ d_{r} \geq 0, r = 1, ..., R. \end{array}$$

◆□▶ ◆□▶ ◆臣▶ ◆臣▶ 臣 のへぐ

Problem?

Summary

Problem?

Utility functions are not available to the manager.

What should Steve do?



Summary

◆□▶ ◆□▶ ◆ □▶ ◆ □▶ ─ □ ─ の < @

Problem?

Utility functions are not available to the manager.

What should Steve do?

◆□▶ ◆□▶ ▲□▶ ▲□▶ ■ ののの

Suggested Mechanism

Proportional Allocation Mechanism: Each user *r* gives a payment w_r ($w_r \ge 0$) to Steve . Given the vector $\mathbf{w} = (w_1, \cdots, w_r)$, Steve chooses a capacity allocation $\mathbf{d} = (d_1, \cdots, d_r)$. Each user is charges with the same price $\mu > 0$, leading to $d_r = \frac{w_r}{\mu}$.

$$\sum_{r} \frac{\mathbf{w}_{r}}{\mu} = \mathbf{C} \Rightarrow \mu = \frac{\sum_{r} \mathbf{w}_{r}}{\mathbf{C}}$$

Quick-fire Question

Introduction

Problem Formulation

Summary

▲□▶ ▲□▶ ▲□▶ ▲□▶ = 三 のへで

Suggested Mechanism

Direct?

Logic flow of the analysis

 Price-taking Agent Model: Users do not anticipate the effect of their actions on the prices of the link per unit (μ), and they consider the price to be fixed and they select the best declarations w_r given μ.

 \Downarrow relaxation

• Price-Anticipating Agent Model: Users can anticipate the effects of their actions.

Proportional Allocation Mechanism: Price-taking Agent Model

Given a price $\mu > 0$, user *r* try to maximize its payoff function for $w_r \ge 0$:

$$P_r(w_r;\mu) = U_r\left(rac{w_r}{\mu}
ight) - w_r$$
 (Quasilinear in w_r)

A pair (\mathbf{w}, μ) is a *competitive equilibrium* if users maximize their payoff

$$P_r(w_r;\mu) \ge P_r(\hat{w}_r;\mu) \quad \forall \hat{w}_r \ge 0, r$$

[Kelly 2007] shows that when users are **price-takers**, there **exists** a competitive equilibrium, and the resulting allocation **solves** the optimization problem (1)

Proportional Allocation Mechanism: Price-taking Agent Model

Given a price $\mu > 0$, user *r* try to maximize its payoff function for $w_r \ge 0$:

$${{\it P}_r}({\it w_r};\mu) = {\it U_r}\left(rac{{\it w_r}}{\mu}
ight) - {\it w_r} \left({\it Quasilinear} \; {\it in} \; {\it w_r}
ight)$$

A pair (\mathbf{w}, μ) is a *competitive equilibrium* if users maximize their payoff

$$P_r(w_r;\mu) \geq P_r(\hat{w}_r;\mu) \quad \forall \hat{w}_r \geq 0, r$$

[Kelly 2007] shows that when users are **price-takers**, there **exists** a competitive equilibrium, and the resulting allocation **solves** the optimization problem (1)

(日) (日) (日) (日) (日) (日) (日)

Theorem

[KELLY 1997] Assume that for each user r, the utility function U_r is concave, strictly increasing, and continuously differentiable. Then there exists a competitive equilibrium, i.e., a vector $\mathbf{w} = (w_1, \dots, w_r) \ge 0$ and a scalar $\mu > 0$ satisfying

$$P_r(w_r; \mu) \ge P_r(\hat{w}_r; \mu) \quad \forall \hat{w}_r \ge 0, \ r$$
$$\mu = \frac{\sum_r w_r}{C}$$

In this case, the scalar μ is uniquely determined, and the vector $\mathbf{d} = \frac{\mathbf{w}}{\mu}$ is a solution to the optimization problem (1). If the functions U_r are strictly concave, then \mathbf{w} is uniquely determined as well.

▲□▶ ▲□▶ ▲ 三▶ ▲ 三▶ - 三 - のへぐ

Proof

Step 1: Aim: Find the equivalent/optimality condition for the competitive equilibrium.

Given $\mu > 0$, **w** satisfy

$$P_r(w_r;\mu) \geq P_r(\hat{w}_r;\mu) \quad \forall \hat{w}_r \geq 0, r$$

if and only if

$$\frac{dP_r(w_r;\mu)}{dw_r} = 0 \quad \text{if } w_r > 0$$
$$\frac{dP_r(0;\mu)}{dw_r} \le 0 \quad \text{if } w_r = 0$$

 $(P_r \text{ is also concave})$ namely

Proof

Step 2: Aim: There exists a **d** that satisfies constraints of similar form . **What We know**: at least one optimal solution to the optimization problem exists (Why?) Lagrangian:

$$\mathcal{L}(\mathbf{d},\mu) = \sum_{r} U_{r}(d_{r}) - \mu \left(\sum_{r} d_{r} - C\right)$$

Slater constraint qualification $\surd \Rightarrow$ existence of $\mu ~\surd$ so the optimal ${\bf d}$ will satisfy

$$egin{aligned} U_r^{'}\left(d_r
ight) &= \mu & ext{if } d_r > 0 \ U_r^{'}(0) &\leq \mu & ext{if } d_r = 0 \ \sum_r d_r &= C. \end{aligned}$$

There exists a pair (d, μ) that satisfy the constraints above, and μ is unique and $\mu > 0$. Quick-fire Question

Proof

- Step 3: If the pair (d,μ) satifies constraint in Step 2, let w = μd. and (w,μ) will satisfy the constraint in Step 1 (i.e. competitive equilibrium)
- Step 4: If w and μ > 0 satisfy constraint in step 1 (i.e. competitive equilibrium), let d = ^w/_μ, and (d,μ) will satisfies constraints in Step 2.
- Step 5: Complete the proof.

Proportional Allocation Mechanism: Price-Anticipating Agent Model

Now the agents know that they can affect the price!

It is possible to show that a Nash equilibrium exists and that is unique.

Theorem

[Johari 2004] Let $R \ge 2$, let d^{CE} be an allocation profile achievable in competitive equilibrium and let d^{NE} be the unique allocation profile achievable in Nash equilibrium. Then any profile of valuation functions U_r for which $\forall r, U_r(0) \ge 0$ satisfies

$$\sum_{r} U_r(d_r^{NE}) \geq \frac{3}{4} \sum_{i} U_r(d_r^{CE}).$$

Quick-fire Question

・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・
 ・

Proportional Allocation Mechanism: Price-Anticipating Agent Model

In other words, the price of anarchy is $\frac{4}{3}$. Even in the worst case, the strategic behaviour by agents will only cause a small reduction in social welfare.

◆□▶ ◆□▶ ▲□▶ ▲□▶ □ のQ@

Proportional Allocation Mechanism: Price-Anticipating Model

Something Else:

- Not bad!
- It achieves minimal price of anarchy, as compared to a broad family of mechanisms in which
 - agents' declarations are a single scalar;
 - the mechanism charges all users the same rate.
- When mechanism is allowed to charge users at different prices, a VCG-like mechanism can be used to achieve full effciency.

▲□▶ ▲□▶ ▲ 三▶ ▲ 三▶ - 三■ - のへぐ

Summary

• In a game where users of a congested single resource anticipate the effect of their actions on the price of the resource, the aggregate utility received by the users is at least 3/4 of the maximum possible aggregate utility.

References

Johari 2004	Johari, Ramesh, and John N. Tsitsiklis. "Efficiency loss in a network resource allocation game." Mathematics of Operations Research 29.3 (2004): 407-435.
Kelly 1997	Kelly, Frank. "Charging and rate control for elastic traffic." European transactions on Telecommunications 8.1 (1997): 33-37.
Shoham 2009	Shoham, Yoav, and Kevin Leyton-Brown. Multiagent systems: Algorithmic, game-theoretic, and logical foundations. Cambridge University Press, 2009.
Vijay 1998	Krishna, Vijay, and Motty Perry. "Efficient mechanism design." Penn State University, mimeo (1998).

Summary



