

Microeconomics 4

Jan Knoepfle

Spring 2022

Organisation

Instructor: Jan Knoepfle, jan.knoepfle@aalto.fi

Lectures:

- **Mondays and Tuesdays 10-12h**, Economicum seminar room 3-4
- streamed on Zoom for non-Helsinki students

Office hours:

- By appointment (email) either in person or over Zoom
- Feel free to reach out actively whenever we can help!

MyCourses Forum: To help your peers, post questions directly on forum whenever possible

Slides uploaded in advance (incomplete) and after the lectures (completed)

Textbooks:

- **Mailath: Modeling Strategic Behavior.**
- **Borgers: An Introduction to the Theory of Mechanism Design.**
- **Mas-Colell, Whinston, Green: Microeconomic Theory.**
- **Krishna: Auction Theory.**
- **Salanié: The Economics of Contracts.**

Problem Sets

TA: Eero Mäenpää, eero.maenpaa@aalto.fi

Exercises:

- 4 problem sets, posted on MyCourses one week before due date
- Exercise sessions with Eero **Mondays 14-16h** in Economicum seminar room 3-4
Dates: 21.03, 28.03, ~~14.04.~~^{04.04.} 02.05
- Hand in your solutions to problem set on MyCourses before exercise session
- Model solutions uploaded after exercise sessions

Requirements:

- At least 50% of solutions to problem sets
- Pass final exam

Grades based on exam only

Information Economics

Information Economics

Introduction

- Micro 3: framework to analyse interaction in **given game** and predict outcome

- Micro 3: framework to analyse interaction in **given game** and predict outcome
- Micro 4: we want to design the **optimal 'game'** to achieve the 'best' outcome

- Micro 3: framework to analyse interaction in **given game** and predict outcome
- Micro 4: we want to design the **optimal 'game'** to achieve the 'best' outcome
- Asymmetric Information poses main problem

- Micro 3: framework to analyse interaction in **given game** and predict outcome
- Micro 4: we want to design the **optimal 'game'** to achieve the 'best' outcome
- Asymmetric Information poses main problem
- Examples for such 'designed games'
 - Sales procedures
 - Voting mechanisms
 - Employment contracts

Two main classes of design problems:

1) Adverse Selection (hidden information)

Uninformed party cannot see **characteristic** of informed party. **Uninformed** moves first.

Concepts: Screening and Mechanism Design

Two main classes of design problems:

1) Adverse Selection (hidden information)

Uninformed party cannot see **characteristic** of informed party. **Uninformed** moves first.

Concepts: Screening and Mechanism Design

2) Moral Hazard (hidden action)

Uninformed party does not see **action** of informed party. **Uninformed** moves first.

Concepts: Contract Theory

Two main classes of design problems:

1) Adverse Selection (hidden information)

Uninformed party cannot see **characteristic** of informed party. **Uninformed** moves first.

Concepts: Screening and Mechanism Design

2) Moral Hazard (hidden action)

Uninformed party does not see **action** of informed party. **Uninformed** moves first.

Concepts: Contract Theory

Will also discuss a class of 'fixed games' where asymmetric information is crucial:

(3) Sender-Receiver Games

Uninformed party does not see **characteristic** of informed party. **Informed** moves first.

Signaling games, (evidence) disclosure games, ...

Adverse Selection

Adverse Selection

Screening

Screening

Screening:

- Principal and **one agent**
- Agent has private information about his preferences
- Principal design (commits to) a mechanism

Simple example:

- One Seller, one Buyer
- Seller owns a phone, **her own valuation 0**, Buyer has valuation $\theta \in \Theta = [\underline{\theta}, \bar{\theta}]$
- Utilities from sale at price p : $u_S(p) = p$ and $u_B(\theta, p) = \theta - p$
- **Buyer knows θ** (we call θ his private type)
- Seller only knows that θ is drawn from distribution F . Assume that $F'(\theta) = f(\theta) > 0$

What is optimal selling procedure (mechanism) for the seller?

Screening – an example

Let's first consider a special and simple class of mechanisms: **posted-price mechanisms**

- Seller posts a price p and buyer decides whether to buy at this price or not
- Seller's expected profit from price p : $\Pi(p) = \underbrace{(1 - F(p))}_{\mathbb{P}[\text{sale}]} \underbrace{p}_{\text{price}}$ iff $p \leq \theta$
- Seller-optimal posted-price must maximise $\Pi(p)$

What is marginal effect on $\Pi(p)$ if increasing p ?

$$\frac{d\Pi(p)}{dp} = \underbrace{(1 - F(p))}_{\substack{\text{gain from} \\ \text{all types} \\ \text{who buy}}} - \underbrace{f(p) \cdot p}_{\substack{\text{loss from marginal} \\ \text{type } \theta = p}}$$
$$\frac{d\Pi(p)}{dp} = 0 \Leftrightarrow p^* = \frac{1 - F(p^*)}{f(p^*)}$$

Screening – an example

Can the seller do better?

- Seller could bargain multiple rounds, offer lotteries at different prices, . . .

Screening – an example

Can the seller do better?

- Seller could bargain multiple rounds, offer lotteries at different prices, . . .
- Problem: space of possible selling procedures is very large

Screening – an example

Can the seller do better?

- Seller could bargain multiple rounds, offer lotteries at different prices, ...
- Problem: space of possible selling procedures is very large

What are the **fixed components** of our problem?

- Space of outcomes: allocation prob. $q \in [0, 1]$ and transfer $t \in \mathbb{R}$ from seller to buyer
- Preferences: seller: $-t$, buyer: $\theta q + t$ with $\theta \sim F$

Screening – an example

Can the seller do better?

- Seller could bargain multiple rounds, offer lotteries at different prices, ...
- Problem: space of possible selling procedures is very large

What are the **fixed components** of our problem?

- Space of outcomes: allocation prob. $q \in [0, 1]$ and transfer $t \in \mathbb{R}$ from seller to buyer
- Preferences: seller: $-t$, buyer: $\theta q + t$ with $\theta \sim F$

What are all **possible mechanisms**?

1. Seller commits to game:

- space of strategies S
- outcome functions $q: S \rightarrow [0, 1]$ and $t: S \rightarrow \mathbb{R}$

2. Buyer (knows θ and) chooses strategy $s(\theta) \in S$

The set of all possible mechanisms of the form $\Gamma = (S, (q, t))$ is quite large!

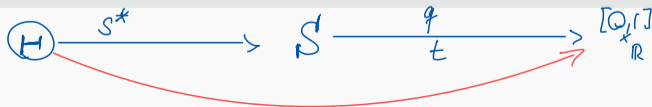
Theorem (Revelation Principle)

Take any mechanism

$$\Gamma = (S, (q(s), t(s))_{s \in S})$$

and optimal agent strategy

$$s_{\Gamma}^*: \Theta \rightarrow S.$$



Theorem (Revelation Principle)

Take any mechanism

$$\Gamma = \left(S, (q(s), t(s))_{s \in S} \right)$$

and optimal agent strategy

$$s_{\Gamma}^* : \Theta \rightarrow S.$$

There is a **direct** mechanism

$$\hat{\Gamma} = \left(\Theta, (\hat{q}(\theta), \hat{t}(\theta))_{\theta \in \Theta} \right)$$

such that **truthtelling**

$$s_{\hat{\Gamma}}^* : \Theta \rightarrow \Theta \text{ with } s_{\hat{\Gamma}}^*(\theta) = \theta$$

is an optimal strategy for the agent and the outcome is the same as in mechanism Γ .

Screening – direct IC mechanisms

Thanks to revelation principle, without loss to focus on direct truthful mechanisms

Seller's optimisation problem:

$$\max_{\substack{q: \Theta \rightarrow [0,1] \\ t: \Theta \rightarrow \mathbb{R}}} \int_{\underline{\theta}}^{\bar{\theta}} -t(\theta) f(\theta) d\theta \quad \text{such that}$$

- ① buyer is willing to participate
- ② buyer is willing to reveal true type

Screening – direct IC mechanisms

Thanks to revelation principle, without loss to focus on direct truthful mechanisms

Seller's optimisation problem:

$$\max_{\substack{q:\Theta\rightarrow[0,1] \\ t:\Theta\rightarrow\mathbb{R}}} \int_{\underline{\theta}}^{\bar{\theta}} -t(\theta)f(\theta) d\theta \quad \text{such that}$$

for all $\theta \in \Theta$: $\theta q(\theta) + t(\theta) \geq 0$

for all $\theta, \hat{\theta} \in \Theta$: $\theta q(\theta) + t(\theta) \geq \theta q(\hat{\theta}) + t(\hat{\theta})$

individual rationality \rightarrow (IR_{θ})

incentive compatibility \leftarrow $(IC_{\theta, \hat{\theta}})$

Screening – direct IC mechanisms

Thanks to revelation principle, without loss to focus on direct truthful mechanisms

Seller's optimisation problem:

$$\max_{\substack{q:\Theta\rightarrow[0,1] \\ t:\Theta\rightarrow\mathbb{R}}} \int_{\underline{\theta}}^{\bar{\theta}} -t(\theta)f(\theta) d\theta \quad \text{such that}$$

$$\text{for all } \theta \in \Theta: \quad \theta q(\theta) + t(\theta) \geq 0 \quad (IR_{\theta})$$

$$\text{for all } \theta, \hat{\theta} \in \Theta: \quad \theta q(\theta) + t(\theta) \geq \theta q(\hat{\theta}) + t(\hat{\theta}) \quad (IC_{\theta, \hat{\theta}})$$

...still a lot of constraints

Screening – incentive compatible allocations and transfers

Type θ 's utility from report $\hat{\theta}$ is

$$\underline{V(\theta, \hat{\theta})} = \theta q(\hat{\theta}) + t(\hat{\theta}), \quad \text{with } \underline{V(\theta)} \equiv \underline{V(\theta, \hat{\theta})} \Big|_{\hat{\theta}=\theta}$$

Screening – incentive compatible allocations and transfers

Type θ 's utility from report $\hat{\theta}$ is

$$V(\theta, \hat{\theta}) = \theta q(\hat{\theta}) + t(\hat{\theta}), \quad \text{with } V(\theta) \equiv V(\theta, \hat{\theta}) \Big|_{\hat{\theta}=\theta}$$

The IC constraints imply two important conditions:

1. $q(\cdot)$ must be weakly increasing in θ

Screening – incentive compatible allocations and transfers

Type θ 's utility from report $\hat{\theta}$ is

$$V(\theta, \hat{\theta}) = \theta q(\hat{\theta}) + t(\hat{\theta}), \quad \text{with } V(\theta) \equiv V(\theta, \hat{\theta}) \Big|_{\hat{\theta}=\theta}$$

The IC constraints imply two important conditions:

1. $q(\cdot)$ must be weakly increasing in θ

2. $V'(\theta) = q(\theta)$ and we can integrate so that $V(\theta) - V(\underline{\theta}) = \int_{\underline{\theta}}^{\theta} q(s) ds$

Rough idea:

• Assume (!) that functions $q(\cdot)$ and $t(\cdot)$ are differentiable

• IC $_{\theta, \hat{\theta}}$ says that $V(\theta, \hat{\theta})$ must be max at $\hat{\theta} = \theta$

"FOC": $\frac{\partial}{\partial \hat{\theta}} V(\theta, \hat{\theta}) \Big|_{\hat{\theta}=\theta} = 0 \Leftrightarrow \theta q'(\hat{\theta}) + t'(\hat{\theta}) \Big|_{\hat{\theta}=\theta} = 0$

$$\frac{dV(\theta, \theta)}{d\theta} = q(\theta) + \underbrace{\theta q'(\theta) + t'(\theta)}_{=0}$$

Screening – change order of integration

Inserting $-t(\theta) = \theta q(\theta) - V(\theta) = \theta q(\theta) - \int_{\underline{\theta}}^{\theta} q(s) ds - V(\underline{\theta})$ into max. problem gives

$$\max_{q: \Theta \rightarrow [0,1], V(\underline{\theta})} \int_{\underline{\theta}}^{\bar{\theta}} \left[\theta q(\theta) - \int_{\underline{\theta}}^{\theta} q(s) ds - V(\underline{\theta}) \right] f(\theta) d\theta \quad \text{s.t. } V(\underline{\theta}) \geq 0, \text{ and } q(\cdot) \text{ increasing}$$

$$\int_{\underline{\theta}}^{\bar{\theta}} \int_{\underline{\theta}}^{\theta} q(s) ds f(\theta) d\theta$$

change order
of integration
 $\underline{\theta} \leq s \leq \theta \leq \bar{\theta}$

$$\int_{\underline{\theta}}^{\bar{\theta}} \int_{s}^{\bar{\theta}} f(\theta) d\theta \cdot q(s) ds = \int_{\underline{\theta}}^{\bar{\theta}} (1 - F(s)) q(s) ds$$

$$\stackrel{s \rightarrow \theta}{=} \int_{\underline{\theta}}^{\bar{\theta}} \frac{(1 - F(\theta))}{f(\theta)} q(\theta) f(\theta) d\theta$$

Screening – virtual value

Changing the order of integration gives

$$\max_{q: \Theta \rightarrow [0,1], V(\underline{\theta})} \int_{\underline{\theta}}^{\bar{\theta}} \left[\theta - \frac{1 - F(\theta)}{f(\theta)} \right] q(\theta) f(\theta) d\theta - V(\underline{\theta}) \quad s.t. \quad V(\underline{\theta}) \geq 0, \quad q(\cdot) \text{ increasing}$$

- We can choose optimal $q(\cdot)$ pointwise (if result satisfies monotonicity constraint)
- $J(\theta) \equiv \theta - \frac{1 - F(\theta)}{f(\theta)}$ is called the **virtual valuation** of type θ
- We say that distribution F is **regular** if $J(\theta)$ is increasing

Screening – virtual value

Changing the order of integration gives

$$\max_{q: \Theta \rightarrow [0,1], V(\underline{\theta})} \int_{\underline{\theta}}^{\bar{\theta}} \left[\theta - \frac{1 - F(\theta)}{f(\theta)} \right] q(\theta) f(\theta) d\theta - V(\underline{\theta}) \quad s.t. \quad V(\underline{\theta}) \geq 0, \quad q(\cdot) \text{ increasing}$$

- We can choose optimal $q(\cdot)$ pointwise (if result satisfies monotonicity constraint)
- $J(\theta) \equiv \theta - \frac{1 - F(\theta)}{f(\theta)}$ is called the **virtual valuation** of type θ
- We say that distribution F is **regular** if $J(\theta)$ is increasing
- If F regular, optimal allocation is $q^*(\theta) = \begin{cases} 0 & \text{if } \theta < \theta^* \\ 1 & \text{if } \theta \geq \theta^*, \end{cases}$ with $\theta^* = \frac{1 - F(\theta^*)}{f(\theta^*)}$
- What if J is not monotone? \rightarrow exercise

Screening example – recap

We have used many tools today that we will develop in more detail further on:

- Revelation Principle
- Characterising IC in terms of allocation rule only
- Virtual valuations

These have allowed us to derive several results:

- We can solve for optimal mechanisms
- Utilities and transfers are pinned down almost entirely by IC allocation rule
- **Posted price mechanisms are optimal!**