

# Notes on Mechanism Design

Kyohei OKUMURA

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- This study notes are mainly based on the lecture note written by Valimaki in 2018.

## 1 Single Agent

- One principal v.s. one agent.
- $a \in A$ : allocation,  $\theta \in \Theta$ : agent's private info.  $\theta \sim F(\theta)$ .  $u^P(a, \theta)$ ,  $u^A(a, \theta)$ .
- We often assume quasi-linear payoff functions:
  - $a := (x, t)$ ,  $u^P(a, \theta) := v^P(x, \theta) + t$ ,  $u^A(a, \theta) := v^A(x, \theta) - t$ .
- A mechanism is a pair  $M := (\Sigma, \phi)$ , where  $\Sigma$  is a message space and  $\phi : \Sigma \rightarrow \Delta(A)$ .
- Agent's strategy:  $\sigma : \Theta \rightarrow \Delta(\Sigma)$ . Principal commits to a mechanism  $M$ .
- Consider a social choice function  $\psi : \Theta \rightarrow A$ . We want to know whether  $\psi$  is implementable (, i.e., achievable in equilibrium,) or not.
- As for implementability, we can discuss it focusing only on direct mechanisms, assuming  $\Sigma := \Theta$ , w.l.o.g. (Revelation principle)

### 1.1 Revenue Equivalence

- In §1.1 and §1.2, we assume that the parameter space is a closed interval  $\Theta := [\underline{\theta}, \bar{\theta}] \subseteq \mathbb{R}$ .

#### 1.1.1 Milgrom and Segal (2002), Envelope Theorem

- $\Theta := [\underline{\theta}, \bar{\theta}]$ .  $f(\cdot, \theta) : X \rightarrow \mathbb{R}$ .  $\{f(\cdot, \theta)\}_{\theta \in \Theta}$ .
- $V(\theta) := \max_{x \in X} f(x, \theta)$ .  $X^*(\theta) := \operatorname{argmax}_{x \in X} f(x, \theta)$

**Def. 1.1** (Selection). *A function  $x^* : \Theta \rightarrow X$  is a selection from  $X^*$  if  $x^*(\theta) \in X^*(\theta)$  for all  $\theta \in \Theta$ .*

**Thm. 1.1** (Milgrom and Segal (2002)). *Assume the following:*

- For any  $x \in X$ ,  $f(x, \cdot) : \Theta \rightarrow \mathbb{R}$  is absolutely continuous on  $\Theta$ .
- For any  $x \in X$ ,  $f(x, \cdot) : \Theta \rightarrow \mathbb{R}$  is differentiable on  $\Theta$ .

*Then, the following holds:*

- $V$  is absolutely continuous.
- For any selection  $x^*$  from  $X^*$ ,  $V(\theta) = V(\underline{\theta}) + \int_{\underline{\theta}}^{\theta} f_{\theta}(x^*(s), s) ds$ .

*Proof.* Note that the absolute continuity of  $f(x, \theta)$  implies that  $f_{\theta}(x, \theta) \in L^1(\Theta)$  for any  $x \in X$ .

(i)  $V$  is **absolutely continuous**. It is sufficient to show that  $V$  is Lipschitz continuous. Fix any  $\theta', \theta$ . Since any integrable function is bounded, for any  $x$  there exists  $L > 0$  s.t.  $|f_\theta(x, \theta)| \leq L$  for almost all  $\theta \in \Theta$ .

$$\begin{aligned} |V(\theta') - V(\theta)| &= \left| \max_{x'} f(x', \theta') - \max_x f(x, \theta) \right| \\ &\leq \max_x |f(x, \theta') - f(x, \theta)| = \max_x \left| \int_{\theta'}^{\theta} f_\theta(x, s) ds \right| \\ &\leq L \cdot |\theta' - \theta| \end{aligned}$$

(ii) Fix any selection  $x^*$  from  $X^*$ . By the result of (i),

$$V(\theta) = V(\underline{\theta}) + \int_{\underline{\theta}}^{\theta} V'(s) ds$$

Fix any selection  $x^*$  and  $\theta', \theta$  such that  $\theta' > \theta$ . By the definition of  $V$  and  $x^*$ ,

$$\begin{aligned} V(\theta) &= f(x^*(\theta), \theta) \geq f(x^*(\theta'), \theta) \\ V(\theta') &= f(x^*(\theta'), \theta') \geq f(x^*(\theta), \theta') \end{aligned}$$

Hence,

$$\begin{aligned} V(\theta') - V(\theta) &\leq f(x^*(\theta'), \theta') - f(x^*(\theta'), \theta) \\ \frac{V(\theta') - V(\theta)}{\theta' - \theta} &\leq \frac{f(x^*(\theta'), \theta') - f(x^*(\theta'), \theta)}{\theta' - \theta} \end{aligned}$$

Similarly,

$$\begin{aligned} V(\theta) - V(\theta') &\leq f(x^*(\theta'), \theta) - f(x^*(\theta'), \theta') \\ \frac{V(\theta) - V(\theta')}{\theta - \theta'} &\geq \frac{f(x^*(\theta'), \theta) - f(x^*(\theta'), \theta')}{\theta - \theta'} \end{aligned}$$

Note that by assumption  $f(x, \cdot)$  is differentiable at all  $\theta \in \Theta$ . Therefore, if  $V$  is differentiable at  $\theta$ , we have  $V'(\theta) = f_\theta(x^*(\theta), \theta)$ .  $\square$

### 1.1.2 RET

- Focus on the agent's utility:  $u := u^A$ .
- $A := \phi(\Theta)$ .  $V(\theta) := \max_{a \in A} u(a, \theta)$ .  $A^*(\theta) := \arg\max_{a \in A} u(a, \theta)$ .
- Assume that  $u(a, \cdot)$  is absolutely continuous and differentiable on  $\Theta$  for all  $a \in A$ .
- By incentive compatibility,  $\phi(\theta) \in A^*(\theta)$  for all  $\theta \in \Theta$ :  $\phi$  is a selection from  $A^*$ .

**Thm. 1.2** (Revenue Equivalence Theorem).

$$V(\theta) = V(\underline{\theta}) + \int_{\underline{\theta}}^{\theta} u_\theta(\phi(s), s) ds$$

In particular, under quasi-linear utility,

$$V(\theta) = V(\underline{\theta}) + \int_{\underline{\theta}}^{\theta} v_\theta(x(s), s) ds$$

$$t(\theta) = v(x(\theta), \theta) - V(\underline{\theta}) - \int_{\underline{\theta}}^{\theta} v_\theta(x(s), s) ds$$

*Proof.* Milgrom and Segal.

As for quasi-linear cases, since  $u((x, t), \theta) = v(x, \theta) - t$ , we have  $u_\theta = v_\theta$ . Note that  $V(\theta) = v(x(\theta), \theta) - t(\theta)$ .  $\square$

- RET states that under any IC mechanism, except for the constant  $V(\underline{\theta})$ , the transfer from the agent to the principal is uniquely determined once the allocation rule  $x$  is fixed.

## 1.2 Characterization of IC

### 1.2.1 Monotone Comparative Statics

This subsection is based on the lecture slides by John K.-H. Quah:

<http://www.johnquah.com/lecture-slides.html>

- Consider parameterized optimization problems.
- We often want to know how optimizers and optimal values change according to the changes in parameters.
- comparative statics = Sensitivity analysis
- Implicit function theorem: Not only the direction of changes but also the rate of change. Many assumptions are required.
- Monotone comparative statics: Only the direction of changes. Fewer assumptions.
- $\Theta \subseteq \mathbb{R}$ . Two functions  $g : \Theta \rightarrow \mathbb{R}$  and  $f : \Theta \rightarrow \mathbb{R}$ .

**Def. 1.2** (Single Crossing).  $g$  dominates  $f$  by single crossing property (SCP),  $g \succsim_{SC} f$ , if for all  $x'' > x'$ ,

- $f(x'') - f(x') \geq 0 \implies g(x'') - g(x') \geq 0$
- $f(x'') - f(x') > 0 \implies g(x'') - g(x') > 0$

$\{f(\cdot, \theta)\}_{\theta \in \Theta}$  is an SCP family if

$$\forall \theta'' > \theta'; f(\cdot, \theta'') \succsim_{SC} f(\cdot, \theta')$$

**Def. 1.3** (Increasing Differences).  $g$  dominates  $f$  by increasing differences,  $g \succsim_{IN} f$ , if for all  $x'' > x'$ ,

$$g(x'') - g(x') \geq f(x'') - f(x').$$

$\{f(\cdot, \theta)\}_{\theta \in \Theta}$  satisfies increasing differences if

$$\forall \theta'' > \theta'; f(\cdot, \theta'') \succsim_{IN} f(\cdot, \theta')$$

**Def. 1.4** (Strictly Increasing Differences).  $g$  dominates  $f$  by strictly increasing differences,  $g \succsim_{SID} f$ , if for all  $x'' > x'$ ,

$$g(x'') - g(x') > f(x'') - f(x').$$

$\{f(\cdot, \theta)\}_{\theta \in \Theta}$  satisfies strictly increasing differences (SID) if

$$\forall \theta'' > \theta'; f(\cdot, \theta'') \succ_{SID} f(\cdot, \theta')$$

**Rem. 1.1.**  $g \succsim_{IN} f$  implies  $g \succsim_{SC} f$ .

**Rem. 1.2.**  $\{f(\cdot, \theta)\}_{\theta \in \Theta}$  satisfies SID iff  $\{f(x, \cdot)\}_{x \in X}$  satisfies SID.

**Thm. 1.3** (Milgrom and Shannon (1994)).  $X \subseteq \mathbb{R}$ .  $f, g : X \rightarrow \mathbb{R}$ .

$$[\forall Y \subseteq X; \operatorname{argmax}_{x \in Y} g(x) \geq \operatorname{argmax}_{x \in Y} f(x)] \iff g \succsim_{SC} f$$

Note that, for  $Y, Z \subseteq \mathbb{R}$ ,

$$Y \geq Z \stackrel{\Delta}{\iff} [y \in Y, z \in Z \implies y \vee x \in Y, y \wedge z \in Z.]$$

*Proof.* .

$\Rightarrow$ ) We show contrapositive. Suppose that  $g \not\prec_{SC} f$ . There exist  $x'', x'$  such that  $x'' > x'$  and at least one of the following holds:

$$f(x'') \geq f(x'), g(x'') < g(x') \quad (1)$$

or

$$f(x'') > f(x'), g(x'') \leq g(x') \quad (2)$$

Let  $Y := \{x', x''\}$ ,  $G_Y := \operatorname{argmax}_{x \in Y} g(x)$  and  $F_Y := \operatorname{argmax}_{x \in Y} f(x)$ . In case of (1),  $x' \vee x'' \notin G_Y$ . In case of (2),  $x' \wedge x'' \notin F_Y$ .

$\Leftarrow$ ) Fix any  $Y \subseteq X$  and  $x'', x' \in Y$  such that  $x' \in G_Y$  and  $x'' \in F_Y$ . We need to show that  $x' \vee x'' \in G_Y$  and  $x' \wedge x'' \in F_Y$ . First, since  $x'' \in F_Y$ , we have  $f(x'') \geq f(x')$ . By assumption,  $g(x'') \geq g(x')$ . Since  $x' \in G_Y$ , we have  $x'' \in G_Y$  and  $x' \vee x'' \in G_Y$ .

Next, we show  $f(x'') = f(x')$ . Note that this implies that  $x' \wedge x'' \in F_Y$ . Suppose toward contradiction that  $f(x'') > f(x')$ . Then, since  $g \succ_{SC} f$ , we have  $g(x'') > g(x')$ . This contradicts  $x' \in G_Y$ .  $\square$

### 1.2.2 Characterization of IC

**Ass. 1.** *Quasi-linear utility.*  $v(x, \theta)$  is absolutely continuous and differentiable on  $\Theta$  for all  $x$ .

**Lem. 1.1.** *Let  $V(\theta) := v(x(\theta), \theta) - t(\theta)$ . If a mechanism  $(x, t)$  is IC, then*

$$V(\theta) = V(\underline{\theta}) + \int_{\underline{\theta}}^{\theta} v_{\theta}(x(s), s) ds \quad (\text{LIC})$$

*Proof.* RET.  $\square$

**Lem. 1.2.** *If a mechanism  $(x, t)$  is IC and  $\{v(\cdot, \theta)\}_{\theta \in \Theta}$  satisfies SID, then*

$$x(\theta) \text{ is non-decreasing in } \theta. \quad (\text{M})$$

*Proof.* Fix  $\theta'', \theta'$  such that  $\theta'' > \theta'$ . Since  $\{v(\cdot, \theta)\}_{\theta \in \Theta}$  satisfies SID,  $v(\cdot, \theta'') \succsim_{SID} v(\cdot, \theta')$ . Suppose toward contradiction that  $x(\theta'') < x(\theta')$ . Since  $v(\cdot, \theta'') \succsim_{SID} v(\cdot, \theta')$ ,

$$v(x(\theta'), \theta'') - v(x(\theta''), \theta'') > v(x(\theta'), \theta') - v(x(\theta''), \theta') \geq 0$$

This violates IC. A contradiction.  $\square$

- The lemmas above shows that, assuming  $\{v(\cdot, \theta)\}_{\theta \in \Theta}$  satisfies SID, IC of  $(x, t)$  implies (LIC) and (M).
- We can show that the converse also holds.

**Lem. 1.3.** *Assume that  $\{v(\cdot, \theta)\}_{\theta \in \Theta}$  satisfies (SID). If the conditions (LIC) and (M) hold, then  $(x, t)$  is IC.*

*Proof.* Fix any  $\theta, \theta'$ . We need to show that  $v(x(\theta), \theta) - t(\theta) \geq v(x(\theta'), \theta) - t(\theta')$ . Note that, by (LIC), we have

$$t(\theta) = v(x(\theta), \theta) - V(\underline{\theta}) - \int_{\underline{\theta}}^{\theta} v_{\theta}(x(s), s) ds$$

Then,

$$\begin{aligned} & [v(x(\theta), \theta) - t(\theta)] - [v(x(\theta'), \theta) - t(\theta')] \\ &= [v(x(\theta), \theta) - t(\theta)] - [v(x(\theta'), \theta) + v(x(\theta'), \theta') - v(x(\theta'), \theta') - t(\theta')] \\ &= \int_{\underline{\theta}}^{\theta} v_{\theta}(x(s), s) ds - \int_{\underline{\theta}}^{\theta'} v_{\theta}(x(s), s) ds - [v(x(\theta'), \theta) - v(x(\theta'), \theta')] \\ &= \int_{\theta'}^{\theta} v_{\theta}(x(s), s) ds - \int_{\theta'}^{\theta} v_{\theta}(x(s), \theta') ds = \int_{\theta'}^{\theta} \underbrace{[v_{\theta}(x(s), s) - v_{\theta}(x(s), \theta')]}_{\geq 0 \because ID} ds \geq 0 \end{aligned}$$

$\square$

**Thm. 1.4** (Characterization of IC). *Assume that  $\{v(\cdot, \theta)\}_{\theta}$  satisfies SID. Then,*

$$(x, t) \text{ is IC} \iff x \text{ is non-decreasing, and } t \text{ is calculated by (LIC)}$$

### 1.3 General Case: Rochet's Theorem and Cyclical Monotonicity

- Consider quasi-linear utility cases.
- Characterize IC mechanisms.

**Def. 1.5** (weak monotonicity). *An allocation rule  $x : \Theta \rightarrow A$  is weakly monotone if*

$$\forall \theta, \theta'; [v(x(\theta), \theta') - v(x(\theta), \theta)] + [v(x(\theta'), \theta) - v(x(\theta'), \theta')] \leq 0$$

**Prop. 1.1.** *If  $(x, t)$  is IC, then  $x$  is weakly monotone.*

**Def. 1.6** (cyclical monotonicity).

$$S := \{(\theta^1, \dots, \theta^{k+1}) \mid \forall i \in [k+1]; \theta^i \in \Theta, \theta^1 = \theta^{k+1}, k \in \mathbb{Z}^+\}$$

*An allocation rule  $x$  is cyclically monotone if, for any  $(\theta^1, \dots, \theta^{k+1}) \in S$ ,*

$$\sum_{i=1}^k [v(x^i, \theta^{i+1}) - v(x^i, \theta^i)] \leq 0, \text{ where } x^i := x(\theta^i) \quad (\text{CM})$$

**Thm. 1.5** (Rochet (1987)).

$$\exists t; (x, t) : \text{IC} \iff x \text{ is cyclically monotone.}$$

*Proof.* .

$\Rightarrow$ ) Easy.

$\Leftarrow$ ) Fix  $\theta_0 \in \Theta$ .

$$S(\theta) := \{(\theta^1, \dots, \theta^{k+1}) \mid \forall i \in [k+1]; \theta^i \in \Theta, \theta^1 = \theta_0, \theta^{k+1} = \theta, k \in \mathbb{Z}^+\}$$

$$V(\theta) := \sup_{(\theta^1, \dots, \theta^{k+1}) \in S(\theta)} \sum_{i=1}^k [v(x^i, \theta^{i+1}) - v(x^i, \theta^i)]$$

**(i)**  $[V(\theta_0) = 0.]$  By CM,  $V(\theta_0) \leq 0$ . Considering the case where  $k := 1$ , we see that  $(\theta_0, \theta_0) \in S(\theta_0)$  satisfies  $[v(x^1, \theta^2) - v(x^1, \theta^1)] = 0$ . Therefore,  $V(\theta_0) = 0$ .

**(ii)**  $[V(\theta) < \infty \text{ for all } \theta \in \Theta.]$  Fix any  $(\theta^1, \dots, \theta^{k+1}) \in S(\theta)$ .

$$\begin{aligned} 0 = V(\theta_0) &\geq \sum_{i=1}^k [v(x^i, \theta^{i+1}) - v(x^i, \theta^i)] + [v(x^{k+1}, \theta_0) - v(x^{k+1}, \theta^{k+1})] \\ &= \sum_{i=1}^k [v(x^i, \theta^{i+1}) - v(x^i, \theta^i)] + [v(x(\theta), \theta_0) - v(x(\theta), \theta)] \\ &\therefore \sum_{i=1}^k [v(x^i, \theta^{i+1}) - v(x^i, \theta^i)] \leq v(x(\theta), \theta) - v(x(\theta), \theta_0) \\ &\therefore V(\theta) \leq v(x(\theta), \theta) - v(x(\theta), \theta_0) \end{aligned}$$

**(iii) [Construct the transfer rule]** Fix any  $\theta, \theta'$ . By the same argument as in (ii), we can show that

$$V(\theta) \geq V(\theta') + v(x(\theta'), \theta) - v(x(\theta'), \theta')$$

Define  $t(\theta) := v(x(\theta), \theta) - V(\theta)$ . With this  $t$ , a mechanism  $(x, t)$  satisfies IC:

$$v(x(\theta), \theta) - t(\theta) - (v(x(\theta'), \theta) - t(\theta')) = V(\theta) - V(\theta') - v(x(\theta'), \theta) + v(x(\theta'), \theta') \geq 0$$

□

## 1.4 Optimizing over Incentive Compatible Mechanisms

**Ass. 2** (Assumptions for IC characterization). In §1.4, we assume that (1) utility function is quasi linear, (2)  $v(x, \theta)$  is absolutely continuous and differentiable on  $\Theta$  for all  $x$ , and (3)  $\{v(\cdot, \theta)\}_\theta$  has SID.

**Ass. 3** (Private Values).  $v^P(x, \theta) \equiv v^P(x)$

**Ass. 4** (Absolutely continuous distribution). The distribution function  $F$  is absolutely continuous, i.e., there exists  $f : \Theta \rightarrow \mathbb{R}_+$  s.t.  $F(x) := \int_{\underline{\theta}}^x f(s)ds$

- Optimal Mechanism = Revenue Maximizing Mechanism
- By Thm. 1.4,  $(x, t)$  is IC iff  $x$  is nondecreasing and  $t$  is calculated by Envelope theorem.

$$\begin{aligned} [\text{Expected Revenue}] &= \mathbb{E}_\theta \left[ t(\theta) + v^P(x(\theta)) \right] \\ &= \mathbb{E}_\theta \left[ v(x(\theta), \theta) - V(\underline{\theta}) - \int_{\underline{\theta}}^{\theta} v_\theta(x(s), s)ds + v^P(x(\theta)) \right] \\ &= \mathbb{E}_\theta \left[ S(x(\theta), \theta) - V(\underline{\theta}) - \int_{\underline{\theta}}^{\theta} v_\theta(x(s), s)ds \right] \end{aligned}$$

$$\begin{aligned} \mathbb{E}_\theta \left[ \int_{\underline{\theta}}^{\theta} v_\theta(x(s), s)ds \right] &= \int_{\underline{\theta}}^{\bar{\theta}} \int_{\underline{\theta}}^{\theta} v_\theta(x(s), s)ds dF(\theta) \\ &= \int_{\underline{\theta}}^{\bar{\theta}} \int_s^{\bar{\theta}} v_\theta(x(s), s)dF(\theta)ds \\ &= \int_{\underline{\theta}}^{\bar{\theta}} (1 - F(s))v_\theta(x(s), s)ds \\ &= \int_{\underline{\theta}}^{\bar{\theta}} (1 - F(\theta))v_\theta(x(\theta), \theta)d\theta \\ &= \int_{\underline{\theta}}^{\bar{\theta}} v_\theta(x(\theta), \theta) \frac{1 - F(\theta)}{f(\theta)} dF(\theta) \\ &= \mathbb{E}_\theta \left[ v_\theta(x(\theta), \theta) \frac{1 - F(\theta)}{f(\theta)} \right] \end{aligned}$$

$$\therefore [\text{Expected Revenue}] = \mathbb{E}_\theta \left[ S(x(\theta), \theta) - V(\underline{\theta}) - v_\theta(x(\theta), \theta) \frac{1 - F(\theta)}{f(\theta)} \right]$$

- The principal solves the following revenue maximization problem:

$$\max_{x(\cdot), V(\underline{\theta})} \mathbb{E}_\theta \left[ S(x(\theta), \theta) - V(\underline{\theta}) - v_\theta(x(\theta), \theta) \frac{1 - F(\theta)}{f(\theta)} \right] \text{ s.t. } x(\cdot) : \text{increasing.}$$

- It is optimal to set  $V(\underline{\theta}) := 0$ , assuming that the outside option value is zero. Then, the problem can be reduced to

$$\max_{x(\cdot)} \mathbb{E}_\theta \left[ \underbrace{S(x(\theta), \theta) - v_\theta(x(\theta), \theta) \frac{1 - F(\theta)}{f(\theta)}}_{(\star)} \right] \text{ s.t. } x(\cdot) : \text{nondecreasing.}$$

- If  $v^P(\theta) \equiv 0$ ,  $(\star) = v(x(\theta), \theta) - v_\theta(x(\theta), \theta) \frac{1 - F(\theta)}{f(\theta)} =:$  [the virtual valuation of the bidder.]

## Finding a Solution

- Just ignoring the monotonicity of  $x$  and solve the relaxed problem. Fix  $\theta \in \Theta$ , and solve maximization problem for each  $\theta$ .
- [Is the argument below valid in case  $x$  is not  $\mathcal{C}^1$  on  $\Theta$ ?]

**Ass. 5.** Assume that  $v$  is linear in  $\theta$ .

**Ass. 6.** Assume the interior solution. (?)

$$\begin{aligned}
 & \max_x S(x, \theta) - v_\theta(x, \theta) \frac{1 - F(\theta)}{f(\theta)} \\
 & S_x(x, \theta) - v_{\theta x}(x, \theta) \frac{1 - F(\theta)}{f(\theta)} = 0 \quad (\text{FOC})
 \end{aligned}$$

$$\begin{aligned}
 & \underbrace{\left( \underbrace{S_{x\theta}(x, \theta)}_{=v_{x\theta}(x, \theta) \text{ (}\because PV\text{)}} - v_{\theta x}(x, \theta) \frac{d}{d\theta} \frac{1 - F(\theta)}{f(\theta)} - \underbrace{v_{\theta x\theta}(x, \theta)}_{=0} \frac{1 - F(\theta)}{f(\theta)} \right)}_{=v_{x\theta}(x, \theta) \left( 1 - \frac{d}{d\theta} \frac{1 - F(\theta)}{f(\theta)} \right)} d\theta \\
 & + \left( \underbrace{S_{xx}(x, \theta) - v_{\theta xx}(x, \theta) \frac{1 - F(\theta)}{f(\theta)}}_{\leq 0 \text{ (}\because SOC\text{)}} \right) dx = 0 \\
 & \iff \left( v_{x\theta}(x, \theta) - v_{x\theta}(x, \theta) \frac{d}{d\theta} \frac{1 - F(\theta)}{f(\theta)} \right) d\theta + (\dots) dx = 0 \\
 & \therefore \operatorname{sgn} \left( \frac{dx}{d\theta} \right) = \operatorname{sgn} \left( v_{x\theta}(x, \theta) - v_{x\theta}(x, \theta) \frac{d}{d\theta} \frac{1 - F(\theta)}{f(\theta)} \right)
 \end{aligned}$$

- Note that since  $v$  has SID,  $v_{x\theta} \geq 0$ .
- The following condition (MHR: monotone hazard rate condition) is sufficient in order for  $x$  to be nondecreasing:

$$\frac{d}{d\theta} \frac{1 - F(\theta)}{f(\theta)} \leq 0 \iff \frac{d}{d\theta} \frac{f(\theta)}{1 - F(\theta)} \geq 0$$

## 2 Many Agents

### 2.1 Setup

**Def. 2.1** (Solution Concepts).  $(M, \phi)$ : mechanism

1.  $\sigma = (\sigma_1, \dots, \sigma_N)$  is a dominant strategy equilibrium (DSE) if

$$\forall i \forall \theta_i \forall m_i \forall m_{-i} \forall \theta_{-i}; \quad u_i(\phi(\sigma_i(\theta_i), m_{-i}; \theta_i, \theta_{-i})) \geq u_i(\phi(m_i, m_{-i}; \theta_i, \theta_{-i}))$$

2.  $\sigma$  is ex-post equilibrium if

$$\forall i \forall \theta_i \forall m_i \forall \theta_{-i}; \quad u_i(\phi(\sigma_i(\theta_i), \sigma_{-i}(\theta_{-i}); \theta_i, \theta_{-i})) \geq u_i(\phi(m_i, \sigma_{-i}(\theta_{-i}); \theta_i, \theta_{-i}))$$

In words, as long as the other players are truthful, their type is not important for a players' incentives for truthful reporting.

3.  $\sigma$  is Bayes-Nash equilibrium (BNE) in pure strategies if

$$\forall i \forall \theta_i \forall m_i; \quad \mathbb{E}_{\theta_{-i}|\theta_i}[u_i(\phi(\sigma_i(\theta_i), \sigma_{-i}(\theta_{-i}); \theta_i, \theta_{-i}))] \geq \mathbb{E}_{\theta_{-i}|\theta_i}[u_i(\phi(m_i, \sigma_{-i}(\theta_{-i}); \theta_i, \theta_{-i}))]$$

**Rem. 2.1.** A DSE is ex-post eqm. An ex-post eqm is BNE.

**Rem. 2.2.** The revelation principle holds for all three equilibrium concepts.

### 2.2 Notions of Incentive Compatibility

- DSIC mechanisms, ex-post IC mechanisms, BNIC mechanisms.

### 2.3 Individual Rationality

**Ass. 7.** The value of an outside option is constant on type and normalized to zero.

**Def. 2.2.** Consider an IC mechanism  $(x, t)$ .

1.  $(x, t)$  is ex-post individually rational (IR) if

$$\forall i \forall \theta; \quad v_i(x(\theta), \theta) - t_i(\theta) \geq 0$$

2.  $(x, t)$  is interim IR if

$$\forall i \forall \theta; \quad \mathbb{E}_{\theta_{-i}|\theta_i}[v_i(x(\theta), \theta) - t_i(\theta)] \geq 0$$

3.  $(x, t)$  is ex-ante IR if

$$\forall i \forall \theta; \quad \mathbb{E}_{\theta}[v_i(x(\theta), \theta) - t_i(\theta)] \geq 0$$

**Rem. 2.3.** An ex-post IR mechanism is interim IR and ex-ante IR.

**Rem. 2.4.** For ex-post IR, it is sometimes useful to impose the following restriction: consider the case where other players are not truthful:

$$\forall i \forall m_{-i} \forall \theta; \quad v_i(x(\theta_i, m_{-i}), \theta) - t_i(\theta_i, m_{-i}) \geq 0$$



## 2.4 Bayesian and Dominant Strategy Mechanisms

??

**Def. 2.3** (Standard Independent Private Value Model). *In the standard IPV model, we assume*

1. For all  $i$ ,  $\Theta_i := [\underline{\theta}_i, \bar{\theta}_i]$
2. quasi-linear utility
3. IPV
4. *The assumptions for the envelope theorem hold. (??)*<sup>1</sup>

**Ass. 8.** For all  $i$ ,  $\Theta_i := [\underline{\theta}_i, \bar{\theta}_i]$ ,  $v_i(x_i, \theta_i)$  satisfies SID, and is sufficiently smooth w.r.t.  $\theta_i$ . IPV.

- Define  $V_i$  as follows:

$$V_i(\theta_i) := \max_{m_i} \{ \mathbb{E}_{\theta_{-i}} [v_i(x(m_i, \theta_{-i}), \theta_i) - t_i(m_i, \theta_{-i})] \} = \max_{m_i} \{ \tilde{v}_i(x(m_i), \theta_i) - \tilde{t}_i(m_i) \}$$

- $\tilde{v}_i(x(m_i), \theta_i) := \mathbb{E}_{\theta_{-i}} [v_i(x(m_i, \theta_{-i}), \theta_i)]$ ,  $\tilde{t}_i(m_i) := \mathbb{E}_{\theta_{-i}} [t_i(m_i, \theta_{-i})]$
- Then, we can apply the same argument as in one agent case.

**Rem. 2.5.** Assume that  $v_i(x, \theta_i)$  satisfies SID. Then,

- $(x, t)$ : DSIC implies  $x(m_i, \theta_{-i})$  is non-decreasing in  $m_i$  for all  $\theta_{-i} \in \Theta_i$ .
- $(x, t)$ : BIC implies  $\mathbb{E}_{\theta_{-i}} [x(m_i, \theta_{-i})]$  is non-decreasing in  $m_i$ .

## 2.5 Efficient Mechanisms

**Ass. 9.** Quasi-linear utilities, private values.

**Def. 2.4** (Private Values). *The model has private values if*

$$\forall i \in \{0, 1, \dots, N\} \forall \theta_{-i} \theta'_{-i} \in \Theta_{-i}; \quad v_i(x, \theta_i, \theta_{-i}) = v_i(x, \theta_i, \theta'_{-i}) \equiv v_i(x, \theta_i)$$

- Consider the case where  $v_0 \equiv 0$ ,  $t_0 = \sum_{i=1}^N t_i$ .

$$x^*(\theta) \in \operatorname{argmax}_{x \in X} \sum_{i=1}^N v_i(x, \theta) = \operatorname{argmax}_{x \in X} \sum_{i=1}^N v_i(x, \theta_i)$$

**Def. 2.5** (VCG mechanism). *Fix any function  $\tau_i : M_{-i} \rightarrow \mathbb{R}$  for all  $i$ . Define the transfer rule:*

$$t_i^*(m) := - \sum_{j \neq i} v_j(x^*(m), m_j) + \tau_i(m_{-i})$$

*The mechanism  $(x^*, t^*)$  is called VCG mechanism.*

**Rem. 2.6.** VDG mechanism is a DSIC mechanism. ( $\because$  By definition of  $x^*$ .)

**Rem. 2.7** (Green & Laffont(1979)). *If there are no restrictions on the domain of preferences for the players, then VDG mechanisms are the only mechanisms that make truthful revelation a dominant strategy for the efficient allocation rule  $x^*$ .*

**Def. 2.6** (Pivot mechanism (the externality mechanism)). *Pivot mechanism is a VCG mechanism that set*

$$\tau_i(m_i) := \max_{x \in X} \sum_{j \neq i} v_j(x^*(m), m_j)$$

---

<sup>1</sup>Let  $V_i(\theta_i) := \max_{m_i}$

- $W_S(\theta) := \max_{x \in X} \sum_{i \in S} v_i(x, \theta_i)$
- In a pivot mechanism, player  $i$  pays her externality and gets her marginal contribution to the social welfare:

$$t_i(\theta) = W_{[N] \setminus \{i\}}(\theta) - \sum_{j \neq i} v_j(x^*(\theta), \theta_j)$$

$$v_i(x^*(\theta), \theta_i) - t_i(\theta) = W_{[N]}(\theta) - W_{[N] \setminus \{i\}}(\theta)$$

- Correlation in the agents' types makes no difference in the above argument: It works for all private values settings with quasi-linear payoffs.

### 2.5.1 Budget Balance

**Def. 2.7** (Budget Balance).  $(x, t) : \text{mechanism}$ .

1.  $(x, t) : \text{mechanism}$  is ex-post budget balanced if  $\forall \theta; \sum_{i=1}^N t_i(\theta) = 0$ .
2.  $(x, t) : \text{mechanism}$  is ex-ante budget balanced if  $\mathbb{E}_\theta[\sum_{i=1}^N t_i(\theta)] = 0$ .

**Ass. 10.** independent type, i.e.,  $\forall i; \theta_i \perp \theta_{-i}$ .

**Def. 2.8** (AGV mechanism (expected externality mechanism)). A mechanism  $(x^*, t)$ , with the following transfer rule  $t$ , is called AGV mechanism:

$$t_i(m) := - \underbrace{\mathbb{E}_{\theta_{-i}} \left[ \sum_{j \neq i} v_j(x^*(m_i, \theta_{-i}), \theta_j) \right]}_{=: \mathcal{E}_i(m_i)} + \tau_i(m_{-i})$$

**Rem. 2.8.** In AGV mechanism, truthful report is a BNIC. (NB: No need to use independence here.)

**Rem. 2.9.** AGV mechanism with the following  $(\tau_i)_i$  is ex-post BB.

$$\tau_i(m_{-i}) = \frac{\sum_{j \neq i} \mathcal{E}_j(m_j)}{N-1}$$

- ex-post BB does not hold for correlated private values.  $(\mathcal{E}_i(m_i))$  should not depend on  $\theta_i$ .
- It is possible to design variants with similarities to AGV mechanism that work also with correlated values settings.

$$\hat{t}_i(m) := - \sum_{j \neq i} v_j(x^*(m), m_j) + \mathbb{E}_{\theta_i} \left[ \sum_{j=1}^N v_j(x^*(\theta_i, m_{-i}), m_j) \mid \theta_{-i} = m_{-i} \right]$$

- $(x^*, \hat{t})$  is a VCG mechanism, and it is ex-ante BB and ex-ante IC.
- What can be said about incentives to participate interim?

### 2.5.2 BB in BNE

- What more can be done with BNE as the solution concept?
- As long as we assume IPV, all BNE implementing the efficient decision rule have the same expected payoffs.
- The expected payment etc. can be calculated using VCG.

### 2.5.3 IC, BB and Participation