



Microeconomic Theory

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All models are wrong, but some are useful.

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Chapter 1 Correspondence: $\Psi : X \rightarrow 2^Y$ (@ Lec 07 of ECON 204)

Definition 1.1 (Correspondence)

A **correspondence** $\Psi : X \rightarrow 2^Y$ from X to Y is a function from X to 2^Y , that is, $\Psi(x) \subseteq Y$ for every $x \in X$. (2^Y is the set of all subsets of Y)



Example 1.1 Let $u : \mathbb{R}_+^n \rightarrow \mathbb{R}$ be a continuous utility function, $y > 0$ and $p \in \mathbb{R}_{++}^n$, that is, $p_i > 0$ for each i .

Define $\Psi : \mathbb{R}_{++}^n \times \mathbb{R}_{++} \rightarrow 2^{\mathbb{R}_+^n}$ by

$$\Psi(p, y) = \operatorname{argmax} u(x)$$

$$\text{s.t. } x \geq 0$$

$$p \cdot x \leq y$$

Ψ is the demand correspondence associated with the utility function u ; typically $\Psi(p, y)$ is multi-valued.

1.1 Continuity of Correspondences

1.1.1 Upper/Lower Hemicontinuous

Let $X \subseteq \mathbb{E}^n$, $Y \subseteq \mathbb{E}^m$, and $\Psi : X \rightarrow 2^Y$.

Definition 1.2 (Upper Hemicontinuous)

Ψ is **upper hemicontinuous** (uhc) at $x_0 \in X$ if, for every open set V with $\Psi(x_0) \subseteq V$, there is an open set U with $x_0 \in U$ s.t.

$$x \in U \Rightarrow \Psi(x) \subseteq V$$



Upper hemicontinuity reflects the requirement that Ψ doesn't "jump down/implode in the limit" at x_0 . (A set to "jump down" at the limit x_0 : It should mean the set suddenly gets smaller – it "implodes in the limit" – that is, there is a sequence $x_n \rightarrow x_0$ and points $y_n \in \Psi(x_n)$ that are far from every point of $\Psi(x_0)$ as $n \rightarrow \infty$.)

Definition 1.3 (Lower Hemicontinuous)

Ψ is **lower hemicontinuous** (lhc) at $x_0 \in X$ if, for every open set V with $\Psi(x_0) \cap V \neq \emptyset$, there is an open set U with $x_0 \in U$ s.t.

$$x \in U \Rightarrow \Psi(x) \cap V \neq \emptyset$$



Lower hemicontinuity reflects the requirement that Ψ doesn't "jump up/explode in the limit" at x_0 . (A set to "jump up" at the limit x_0 : It should mean that the set suddenly gets bigger – it "explodes in the limit" – that is,

there is a sequence $x_n \rightarrow x_0$ and a point $y_0 \in \Psi(x_0)$ that is far from every point of $\Psi(x_n)$ as $n \rightarrow \infty$.)

Definition 1.4 (Continuous Correspondence)

Ψ is **continuous** at $x_0 \in X$ if it is both **uhc** and **lhc** at x_0 .



Proposition 1.1

Ψ is upper hemicontinuous (respectively lower hemicontinuous, continuous) if it is uhc (respectively lhc, continuous) at every $x \in X$.

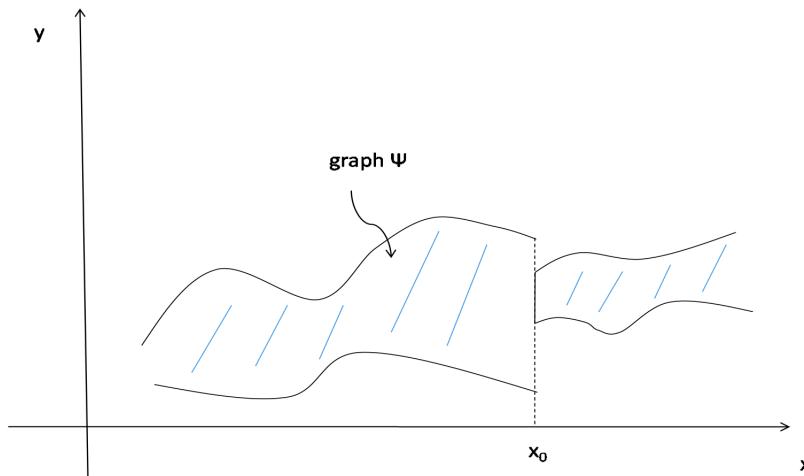


Figure 1.1: The correspondence Ψ ‘implodes in the limit’ at x_0 . Ψ is not upper hemicontinuous at x_0 .

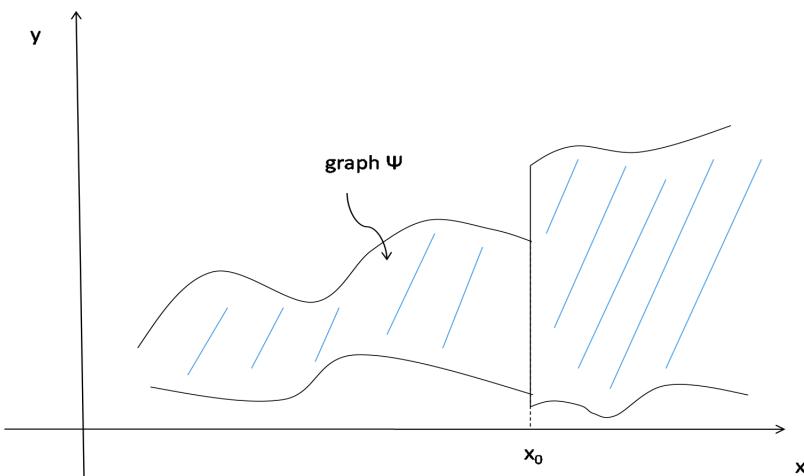


Figure 1.2: The correspondence Ψ ‘explodes in the limit’ at x_0 . Ψ is not lower hemicontinuous at x_0 .

1.1.2 Theorem: $\Psi(x) = \{f(x)\}$ is uhc $\Leftrightarrow f$ is continuous

Theorem 1.1 ($\Psi(x) = \{f(x)\}$ is uhc $\Leftrightarrow f$ is continuous)

Let $X \subseteq \mathbb{E}^n$, $Y \subseteq \mathbb{E}^m$ and $f : X \rightarrow Y$. Let $\Psi : X \rightarrow 2^Y$ be defined by $\Psi(x) = \{f(x)\}$ for all $x \in X$.

Then Ψ is uhc if and only if f is continuous.



1.1.3 Berge's Maximum Theorem: the set of maximizers is uhc with non-empty compact values

Theorem 1.2 (Berge's Maximum Theorem)

Let $X \subseteq \mathbb{R}^n$ and $Y \subseteq \mathbb{R}^m$. Consider the function $f : X \times Y \rightarrow \mathbb{R}$ and the correspondence $\Gamma : Y \rightarrow 2^X$.

Define $v(y) = \max_{x \in \Gamma(y)} f(x, y)$ and the set of maximizers

$$\Omega(y) = \operatorname{argmax}_{x \in \Gamma(y)} f(x, y) = \{x : f(x, y) = v(y)\}$$

Suppose f and Γ are continuous, and that Γ has non-empty compact values. Then, v is continuous and Ω is uhc with non-empty compact values.



1.2 Graph of Correspondence

An alternative notion of continuity looks instead at properties of the graph of the correspondence.

Definition 1.5 (Graph of Correspondence)

The **graph** of a correspondence $\Psi : X \rightarrow 2^Y$ is the set

$$\operatorname{graph}\Psi = \{(x, y) \in X \times Y : y \in \Psi(x)\}$$



1.2.1 Closed Graph

By the definition of continuous function $f : \mathbb{R}^n \rightarrow \mathbb{R}$, each convergent sequence $\{(x_n, y_n)\}$ in graph f converges to a point (x, y) in graph f , that is, graph f is closed.

Definition 1.6 (Closed Graph)

Let $X \subseteq \mathbb{E}^n$, $Y \subseteq \mathbb{E}^m$. A correspondence $\Psi : X \rightarrow 2^Y$ has closed graph if its graph is a closed subset of $X \times Y$, that is, if for any sequences $\{x_n\} \subseteq X$ and $\{y_n\} \subseteq Y$ such that $x_n \rightarrow x \in X$, $y_n \rightarrow y \in Y$ and $y_n \in \Psi(x_n)$ for each n , then $y \in \Psi(x)$.



Example 1.2 Consider the correspondence $\Psi(x) = \begin{cases} \{\frac{1}{x}\}, & \text{if } x \in (0, 1] \\ \{0\}, & \text{if } x = 0 \end{cases}$ ("implode in the limit")

Let $V = (-0.1, 0.1)$. Then $\Psi(0) = \{0\} \subseteq V$, but no matter how close x is to 0, $\Psi(x) = \{\frac{1}{x}\} \not\subseteq V$, so Ψ is not

uhc at 0. However, note that Ψ has closed graph.

1.3 Closed-valued, Compact-valued, and Convex-valued Correspondences

Definition 1.7 (Closed-valued, Compact-valued, and Convex-valued Correspondences)

Given a correspondence $\Psi : X \rightarrow 2^Y$,

1. Ψ is **closed-valued** if $\Psi(x)$ is a closed subset of Y for all x ;
2. Ψ is **compact-valued** if $\Psi(x)$ is compact for all x .
3. Ψ is **convex-valued** if $\Psi(x)$ is convex for all x .



1.3.1 Closed-valued, uhc and Closed Graph

For closed-valued correspondences these concepts can be more tightly connected. A closed-valued and upper hemicontinuous correspondence must have closed graph. For a closed-valued correspondence with a compact range, upper hemicontinuity is equivalent to closed graph.

Theorem 1.3 (uhc and Closed Graph)

Let $X \subseteq \mathbb{E}^n$, $Y \subseteq \mathbb{E}^m$, and $\Psi : X \rightarrow 2^Y$.

1. Ψ is **closed-valued** and **uhc** $\Rightarrow \Psi$ has **closed graph**.
2. Ψ is **closed-valued** and **uhc** $\Leftarrow \Psi$ has **closed graph**. (If Y is **compact**)



Theorem 1.4

Let $X \subseteq \mathbb{E}^n$, $Y \subseteq \mathbb{E}^m$, and $\Psi : X \rightarrow 2^Y$. If Ψ has **closed graph** and there is an **open set** W with $x_0 \in W$ and a **compact set** Z such that $x \in W \cap X \Rightarrow \Psi(x) \subseteq Z$, then Ψ is **uhc** at x_0 .



1.3.2 Theorem: compact-valued, uhs correspondence of compact set is compact

Theorem 1.5

Let X be a compact set and $\Psi : X \rightarrow 2^X$ be a non-empty, compact-valued upper-hemicontinuous correspondence. If $C \subseteq X$ is compact, then $\Psi(C)$ is compact.



Proof 1.1

Given the compact-valued Ψ , we can have an open cover of $\Psi(C)$, $\{U_\lambda : \lambda \in \Lambda\}$. So $\forall x \in C$, there exists $U_{l(x)}$, $l(x) \in \Lambda$ such that $U_{l(x)}$ is an open cover of $\Psi(x)$.

Consider a $c \in C$. Since Ψ is uhs and $\Psi(c) \subseteq U_{l(c)}$, there exists open set V_c s.t. $c \in V_c$ and $\Psi(x) \subseteq U_{l(c)}$, $\forall x \in V_c \cap C$.

$\{V_c : c \in C\}$ is an open cover of C . Because C is compact, there is a finite subcover $\{V_{c_i} : i = 1, \dots, m\}, m \in \mathbb{N}$, where $\{c_i : i = 1, \dots, m\} \subseteq C$.

Because $\Psi(x) \subseteq U_{l(c_i)}, \forall x \in V_{c_i} \cap C$ and $\{V_{c_i} : i = 1, \dots, m\}, m \in \mathbb{N}$ is a open cover for C , we can infer $\{U_{l(c_i)} : i = 1, \dots, m\}$ is a finite subcover of $\{U_{l(c)} : c \in C\}$ for $\Psi(C)$. Hence, $\Psi(C)$ is compact.

1.4 Fixed Points for Correspondences (@ Lec 13 of ECON 204)

1.4.1 Definition

Definition 1.8 (Fixed Points for Correspondences)

Let X be nonempty and $\psi : X \rightarrow 2^X$ be a correspondence. A point $x^* \in X$ is a fixed point of ψ if $x^* \in \psi(x^*)$.



Note We only need x^* to be in $\psi(x^*)$, not $\{x^*\} = \psi(x^*)$. That is, ψ need not be single-valued at x^* . So x^* can be a fixed point of ψ but there may be other elements of $\psi(x^*)$ different from x^* .

1.4.2 Kakutani's Fixed Point Theorem: uhs, compact, convex values correspondence has a fixed point over compact convex set

Theorem 1.6 (Kakutani's Fixed Point Theorem)

Let $X \subseteq \mathbb{R}^n$ be a non-empty, **compact**, **convex** set and $\psi : X \rightarrow 2^X$ be an **upper hemi-continuous** correspondence with non-empty and **convex** values. Then ψ has a fixed point in X .



1.4.3 Theorem: \exists compact set $C = \bigcap_{i=0}^{\infty} \Psi^i(X)$ s.t. $\Psi(C) = C$

Theorem 1.7

Let (X, d) be a compact metric space and let $\Psi(x) : X \rightarrow 2^X$ be a upper-hemicontinuous, compact-valued correspondence, such that $\Psi(x)$ is non-empty for every $x \in X$. There exists a compact non-empty subset $C \subseteq X$, such that $\Psi(C) \equiv \bigcup_{x \in C} \Psi(x) = C$.



Proof 1.2

Let's construct a sequence $\{C_n\}$ such that $C_0 = X$, $C_1 = \Psi(C_0), \dots, C_n = \Psi(C_{n-1}), \dots$ We claim that $C = \bigcap_{i=0}^{\infty} C_i$ is a non-empty compact set and satisfies $\Psi(C) = C$.

1. Because we can infer $\Psi(X_1) \subseteq \Psi(X_2)$ if $X_1 \subseteq X_2$, $X = C_0 \supseteq C_1 \Rightarrow C_1 = \Psi(C_0) \supseteq C_2 = \Psi(C_1), \dots$, so $C_0 \supseteq C_1 \supseteq \dots \supseteq C_n \supseteq \dots$. Hence, C is not empty.

2. Because X is compact, by the theorem 1.5, we can infer C_n is compact for all n . Then, C_n is closed for all n , so C is closed. Because C is a closed set of compact set X , C is compact.
3. $C \subseteq C_n, \forall n \Rightarrow \Psi(C) \subseteq \Psi(C_n), \forall n \Rightarrow \Psi(C) \subseteq C$
4. Assume $C \subseteq \Psi(C)$ doesn't hold, that is $\exists y \in C$ s.t. $y \notin \Psi(C)$. Because $y \in C$ and $C_0 \supseteq C_1 \supseteq \dots \supseteq C_n \supseteq \dots$, there exists $k \in C_n$ for all n s.t. $y \in \Psi(k)$. $k \in \cap_{i=1}^{\infty} C_i = C$, so $\Psi(k) \subseteq \Psi(C)$, which contradicts to $y \notin \Psi(C)$. Hence, $C \subseteq \Psi(C)$.

All in all the claim " $C = \cap_{i=0}^{\infty} C_i$ is a non-empty compact set and satisfies $\Psi(C) = C$ " is proved.

Chapter 2 Preference and Utility Function

Based on

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2.1 Preferences

2.1.1 Preference Relation

Definition 2.1 (Weak, Strict, Indifference)

\succeq referred to as the **weak preference relation**: " x is at least as good as y ". (ordinal);

"**No better than**": $y \preceq x$ if and only if $x \succeq y$.

"**Strict preference**": $x \succ y$ if and only if $x \succeq y$ and not $y \succeq x$.

"**Indifference**": $x \sim y$ if and only if $x \succeq y$ and $y \succeq x$.



2.1.2 Basic Assumptions

2.1.3 Rational Preference

Definition 2.2 (Rational Relation = Preference)

A binary relation \succeq on X is a **preference relation** if it is a weak order, i.e., **complete** and **transitive**.

Rationality: \succeq is **rational** if and only if it is **complete** and **transitive**.

- \succeq is **complete** iff $\forall x, y \in X, x \succeq y$ or $y \succeq x$.
- \succeq is **transitive** iff $\forall x, y, z \in X$, if $x \succeq y$ and $y \succeq z$, then $x \succeq z$.



The completeness means

- Any two bundles can be compared
- Indifference is allowed

The transitivity

- like transitivity of the real numbers
- extends pairwise preferences to longer chains in the logical way.

2.2 Utility Function

2.2.1 Utility Function \Leftrightarrow Rational Preference

Definition 2.3 (Utility Function)

We can say a function $u : X \rightarrow \mathbb{R}$ represents \succeq if $\forall x, y \in X$,

$$x \succeq y \Leftrightarrow u(x) \geq u(y)$$



Proposition 2.1 (Rational $\succeq \Rightarrow \exists u(\cdot)$)

If \exists a function $u : X \rightarrow \mathbb{R}$ represents \succeq , then \succeq is rational (i.e., completeness and transitivity)



Note The reverse may not true.

2.2.2 Convex Preference

Definition 2.4 (Convex \succeq)

\succeq is **convex** if for every $x \in X$ the $\{y \in X : y \succeq x\}$ is convex, i.e., $y_1 \succeq x$ and $y_2 \succeq x \Rightarrow \alpha y_1 + (1 - \alpha)y_2 \succeq x$ for all $\alpha \in [0, 1]$.



Convex relations imply *averages are preferred to extremes*.

Definition 2.5 (Strictly Convex)

\succeq is **strictly convex** iff $\forall x, y, z \in X$, if $x \succeq z$ and $y \succeq z$, then $\alpha x + (1 - \alpha)y \succ z$ for all $\alpha \in (0, 1)$



2.2.3 Convex Preference \Leftrightarrow Quasiconcave Utility Function

Definition 2.6 (Quasi-Concave Function)

A function u is **quasi-concave** if and only if for all $t \in \mathbb{R}$, $\{x \in X : u(x) \geq t\}$ is convex.

$$\forall x, y \in X, t \in \mathbb{R}, 0 \leq a \leq 1 : u(x) \geq t, u(y) \geq t \Rightarrow u(ax + (1 - a)y) \geq t$$



Proposition 2.2 (Concave Function \Rightarrow Quasi-Concave Function)

Any function that is concave is also quasi-concave.



Proposition 2.3 (Convex $\succeq \Leftrightarrow$ quasi-concave $u(\cdot)$)

\succeq is convex, $\Leftrightarrow \exists$ a quasi-concave $u(\cdot)$ that represents \succeq .



2.3 Preferences over Nearby Bundles

2.3.1 Monotone Preference

Definition 2.7 (Monotone \succeq)

\succeq is **monotone** if $x, y \in X$ with $x \geq y \Rightarrow x \succeq y$ (and $x > y \Rightarrow x \succ y$).



Proposition 2.4 (Monotone $\succeq \Rightarrow$ monotone $u(\cdot)$)

If \succeq is monotone, then \exists a monotone $u(\cdot)$ that represents \succeq .



Note Complete, transitive, and monotone are three assumptions that made by all theories (either EU or non-EU).

2.3.2 Strongly monotone

Definition 2.8 (Strongly Monotone \succeq)

\succeq is **strongly monotone** if and only if for any $x = (x_1, \dots, x_n), y = (y_1, \dots, y_n) \in X$, if $\forall i : x_i \geq y_i$ and $\exists j$ such that $x_j > y_j$, then $x \succ y$.



(When we compare elements that have more than one dimension, strongly monotone holds if at least one relation is not equal.)

$$A = (1, 1), B = (2, 1), C = (1, 2), D = (2, 2)$$

Strongly monotone can infer that $D \succ B \succ A, D \succ C \succ A$.

2.3.3 Local Non-Satiation

Even weaker assumptions will ensure that the consumer's choice exhausts their budget.

Definition 2.9 (Local Nonsatiation)

For any bundle x , there is a nearby bundle y in the consumption set such that y is preferred to x . That is, for all $x \in X$ and every $\varepsilon > 0$,

$$\exists y \in |x - y| < \varepsilon, \text{ s.t. } y \succ x$$



We have

Strong Monotonicity \Rightarrow Monotonicity \Rightarrow Local Nonsatiation

2.4 Common Assumptions of Preference

\succeq	u
monotone	\implies nondecreasing
strongly monotone	\implies strictly increasing
continuous	\implies continuous (Debreu's Theorem)
convex	\implies quasi-concave (but not concave)
strictly convex	\implies strictly concave (and strictly quasi-concave)
homothetic (and continuous)	\implies continuous and homogeneous
(so-called) quasi-linear	\implies quasi-linear
(so-called) differentiable	\implies differentiable
separable	\implies separable (form)
strongly separable	\implies additively separable (form)

Figure 2.1: Properties of Preference and Utility Function

2.4.1 Independence of Preference

The 'standard' model of decisions under risk is based on von Neumann and Morgenstern Expected Utility (EU), which requires the independence assumption.

Definition 2.10 (Independence of Preference)

Independence: For any $x, y, z \in X$ and $0 < \alpha < 1$, if $x \succeq y$ then $\alpha x + (1 - \alpha)z \succeq \alpha y + (1 - \alpha)z$.



2.4.2 Continuous Preference

Definition 2.11 (Continuous \succeq)

\succeq is **continuous** on X if and only if for any sequence $\{x^n, y^n\}_{n=1}^{\infty}$ with $x^n \succeq y^n$ and we note $x = \lim_{n \rightarrow \infty} x^n$, $y = \lim_{n \rightarrow \infty} y^n$, we have $x \succeq y$ (i.e., the graph $\{(x, y) \mid x \succeq y \subseteq X \times X\}$ is closed).



Proposition 2.5 (Debreu's Theorem, Continuous $\succeq \Rightarrow$ continuous $u(\cdot)$)

If \succeq is continuous (on X , a convex subset of \mathbb{R}^k), then \exists a continuous $u(\cdot)$ that represents \succeq .



Example 2.1 (Lexicographic preferences (not continuous)) Under Lexicographic preference \succ , $x \succ y$ if and only if

- $x_1 > y_1$, or
- $x_1 = y_1$, and $x_2 > y_2$, or
- $x_1 = y_1$ and $x_2 = y_2$ and $x_3 > y_3$, or
- etc.

Under Lexicographic preferences, there is no indifference.

We can find the Lexicographic preference violates continuity: $(1 + \frac{1}{n}, 1) \succ (1, 2)$ and $\lim (1 + \frac{1}{n}, 1) = (1, 1) \prec (1, 2)$.

Example 2.2 (Utility Representation for Lexicographic Preferences) Consider the lexicographic preference \succeq over the restricted domain $X = (\mathbb{Q} \cap [0, 1]) \times [0, 1]$. Enumerate the rationals in $[0, 1]$ as $\mathbb{Q} \cap [0, 1] = \{q^1, q^2, q^3, \dots\}$ where $q^i \neq q^j$ if $i \neq j$. The utility representation of this preference is

$$u(x_1, x_2) = \sum_{q^i < q^j} \frac{1}{2^i} + \frac{1}{2^j} x_2, \text{ where } q^j = x_1$$

2.4.3 Homothetic Preference

Definition 2.12 (Homotheticity)

\succeq are homothetic if $x \succeq y \Rightarrow \alpha x \succeq \alpha y$ for all $\alpha > 0$.



Proposition 2.6 (Homothetic preference \Leftrightarrow homogeneous $u(\cdot)$)

A continuous \succeq is homothetic $\Leftrightarrow \exists$ a continuous homogeneous $u(\cdot)$ that represents \succeq such that $u(\alpha x) = \alpha u(x)$ for all $x > 0$.



2.4.4 Quasi-linearity

Definition 2.13 (Quasi-Linearity)

\succeq on X is **quasi-linear** on x_1 if

$$x \succeq y \Rightarrow (x + \epsilon e_1) \succeq (y + \epsilon e_1)$$

where $e_1 = (1, 0, \dots, 0)$ and $\epsilon > 0$.



Theorem 2.1 (Quasi-Linearity $\Leftrightarrow u(x) = x_1 + v(x_{-1})$)

A continuous \succeq on $(-\infty, \infty) \times \mathbb{R}_+^{K-1}$ is quasi-linear in $x_1 \Leftrightarrow \exists$ a $u(\cdot)$ that represents \succeq such that

$$u(x) = x_1 + v(x_{-1})$$

where $v(\cdot)$ satisfies $(v(x_{-1}), 0, \dots, 0) \sim (0, x_{-1})$.



2.4.5 Separability

Definition 2.14 (Separability)

\succeq satisfies **separability** if for any x_i

$$(x_i, x_{-i}) \succeq (x'_i, x_{-i}) \Leftrightarrow (x_i, x'_{-i}) \succeq (x'_i, x'_{-i})$$



Theorem 2.2 (Separability \Rightarrow Additive $u(\cdot)$)

\succeq with **separability** admits additive u -representation

$$u(x) = v_1(x_1) + \cdots + v_K(x_K)$$



Note Strong assumption, usually ignored in practice.

2.4.6 Differentiable Preference

Consider a vector of values $v(x) \in \mathbb{R}_+^K$ for the K commodities and a feasible direction $x + \varepsilon d \in X$ from x for small enough $\varepsilon > 0$.

d is considered improvement if and only if

$$d \cdot v(x) > 0$$

Given $v(x) : X \rightarrow \mathbb{R}_+^K$, let

$$D_v(x) = \{d : d \cdot v(x) > 0\}$$

be the set of directions that are improvements relative to x .

$d \in \mathbb{R}^k$ is an improvement direction at x if there is $\lambda^* > 0$ such that λd is an improvement

$$x + \lambda d \succ x$$

for any $\lambda \leq \lambda^*$. Let $D_\succeq(x)$ be the set of all improvement directions at x .

Any improvement is an improvement direction if

- \succeq are strictly convex.
- \succeq are convex, strongly monotonic, and continuous.

Definition 2.15 (Differentiable Preference)

\succeq is **differentiable** if there exists a function $v(x) : X \rightarrow \mathbb{R}_+^K$ such that

$$D_\succeq(x) = D_v(x), \forall x \in X$$



Example 2.3 \succeq represented by

- (1). $\alpha x_1 + \beta x_2$ for $\alpha, \beta > 0$ are differentiable: $v(x) = (\alpha, \beta)$.

(2). $\min\{\alpha x_1, \beta x_2\}$ are differentiable where $\alpha x_1 \neq \beta x_2$: $v(x) = \begin{cases} (1, 0) & \text{if } \alpha x_1 < \beta x_2 \\ (0, 1) & \text{otherwise} \end{cases}$

Proposition 2.7 (Sufficient condition for differentiable \succeq)

Any (monotonic and convex) \succeq can be represented by a (strongly monotonic and quasi-concave) and differentiable u is differentiable.



Chapter 3 Choice Theory

3.1 Choice

Let $\mathcal{B} = 2^X$ (all subsets of X) and $B \in \mathcal{B}$ be the all potential alternatives that can be chosen.

The choice of an agent can be represented by $C(B) \subseteq B, \forall B \in \mathcal{B}$.

Definition 3.1 (Choice Correspondence (More than one choice))

A choice correspondence C assigns a non-empty subset for every non-empty set A

$$\emptyset \neq C(A) \subseteq A$$



Definition 3.2 (Induced Choice Rule)

Given a binary relation \succeq , the **induced choice rule** C_{\succeq} is defined by $C(A) = C_{\succeq}(A) = \{x \in A : x \succeq y, \forall y \in A\}, \forall A \subseteq X$.

A choice function c can be **rationalizable** if there is a preference relation \succeq on X such that $c = c_{\succeq}$.



Definition 3.3 (Revealed Preference)

Given a choice rule \succeq , its **revealed preference relation** \succeq_C is defined by $x \succeq_C y$ if there exists some A such that $x, y \in A$ and $x \in C(A)$.



Proposition 3.1

If C is rationalized by \succeq , then $\succeq = \succeq_C$.



Definition 3.4 (Rubinstein's Condition α)

A choice function c satisfies **condition α** if for any two problems A, B , if $A \subseteq B$ and $c(B) \in A$, then $c(A) = c(B)$.



3.1.1 Choice Function

Definition 3.5 (Choice Function)

A **choice function** c such that $c(A) \in A$ which specifies a unique element for each nonempty subset $A \subseteq X$ (no indifferent preferences).



Proposition 3.2 (Rubinstein's Condition $\alpha \Rightarrow$ Rationalizable Choice Function c)

(1). Let c be a choice function defined on a domain containing at least all subsets of X of size of at most 3. If c satisfies condition α , then there is a preference \succeq on X such that $c = c_{\succeq}$.

(2). Let c be a choice function with a domain D satisfying that if $A, B \in D$, then $A \cup B \in D$. If c satisfies condition α , then there is a preference relation \succeq on X such that $c = c_\succeq$.



3.1.2 Choice Correspondence

Definition 3.6 (Sen's α or Independence of Irrelevant Alternatives)

If $a \in A \subseteq B$, then $a \in C(B) \Rightarrow a \in C(A)$.



Definition 3.7 (Sen's β)

If $a, b \in A \subseteq B$, then $a, b \in C(A)$ and $b \in C(B) \Rightarrow a \in C(B)$.



α and β are equivalent to WARP.

Definition 3.8 (Weak Axiom of Revealed Preference (WARP))

Given a choice structure (C, \mathcal{B}) satisfies **WARP**. If $\exists B \in \mathcal{B}$ with $x, y \in B$, such that $x \in C(B)$. Then, $\forall B' \in \mathcal{B}$ with $x, y \in B'$, $y \in C(B') \Rightarrow x \in C(B')$.

Or we can say,

$$x, y \in B \cap B', x \in C(B), \text{ and } y \in C(B') \Rightarrow x \in C(B')$$



Proposition 3.3 (Rational \Rightarrow WARP)

Given \succeq is rational, then $(C_\succeq^*, \mathcal{B})$ satisfies WARP.

$(C_\succeq^* \text{ is the choice rule that picks the maximal alternatives by } \succeq)$



Proposition 3.4 (Sen's Condition $\alpha, \beta \Rightarrow$ Rationalizable Choice Correspondence C)

Let C be a choice correspondence defined on a domain containing at least all subsets of X of size of at most 3. If C satisfies condition α and β , then there is a preference \succeq on X such that $C = C_\succeq$.



3.2 Revealed Preference

Given choice data (p^t, x^t) , we say u -function rationalizes the observed behavior (p^t, x^t) if for all $t = 1, \dots, T$, $p^t x^t \geq p^t x \Rightarrow u(x^t) \geq u(x)$, that is, $u(\cdot)$ achieves its maximum value on the budget set at the chosen bundles. If “locally non-satiated” u -function, $p^t x^t > p^t x \Rightarrow u(x^t) > u(x)$.

Definition 3.9 (Revealed Preferred)

We say x^t is

- $x^t R^D x$: directly revealed preferred to x , if $p^t x^t \geq p^t x$; (x is available under p^t)
- $x^t P^D x$: strictly directly revealed preferred to x , if $p^t x^t > p^t x$;

- $x^t Rx$: indirectly revealed preferred to x , if \exists a sequence $\{x_k\}_{k=1}^K$ with $x_1 = x^t$ and $x_K = x$ such that $x_k R^D x_{k+1}$ for all $k = 1, \dots, K-1$, i.e., $p^t x^t = p^t x_1 \geq p^t x_2 \geq \dots \geq p^t x_K = p^t x$.

Definition 3.10 (Generalized Axiom of Revealed Preference (GARP))

Consider two observations (p^t, x^t) and (p^s, x^s) , GARP is satisfied if

$$x^t Rx^s \Rightarrow \text{not } x^s P^D x^t$$

$$\text{i.e., } x^t Rx^s \Rightarrow p^s x^t \geq p^s x^s$$

GARP is a generalization of various other revealed preference tests

Definition 3.11

Weak Axiom of Revealed Preference (WARP):

$$x^t R^D x^s, x^t \neq x^s \Rightarrow \text{not } x^s P^D x^t$$

$$\text{i.e., } p^t x^t \geq p^t x^s, x^t \neq x^s \Rightarrow p^s x^t \geq p^s x^s$$

Strong Axiom of Revealed Preference (SARP):

$$x^t Rx^s, x^t \neq x^s \Rightarrow \text{not } x^s Rx^t$$

Theorem 3.1 (Afriat's Theorem)

The following conditions are equivalent:

1. The data satisfies GARP;
2. There exists a non-satiated u -function that rationalizes the data;
3. There exists a concave, monotonic, continuous, non-satiated u -function that rationalizes the data.
4. There exist positive numbers (u^t, λ^t) for $t = 1, \dots, T$ that satisfy the so-called Afriat inequalities:

$$u^s \leq u^t + \lambda^t p^t(x^s - x^t), \forall t, s$$

3.3 Choice under Uncertainty

We want to model an uncertain prospect corresponding forms of function u .

The literature contains (basically) three sets of answers to these questions, differing in whether uncertainty is objective or subjective.

- o Objective uncertainty: von Neumann-Morgenstern (vNM).
- o Subjective uncertainty: Savage.
- o Horse lottery-roulette wheel theory: Anscombe and Aumann (A-A)

3.3.1 von Neumann-Morgenstern (vNM)

The set of prizes is defined by X and the set of probability measures (or distributions) over X is denoted by P .

A compound lottery: If $p, q \in P$ and $\alpha \in [0, 1]$, then there is an element $\alpha p + (1 - \alpha)q \in P$ which is defined by taking the convex combinations of the probabilities of each prize separately, or

$$(\alpha p + (1 - \alpha)q)(x) = \alpha p(x) + (1 - \alpha)q(x)$$

$(\alpha p + (1 - \alpha)q)$ represents a compound lottery.

Definition 3.12 (Three Axioms)

Three Axioms

(A1) \succ is a preference relation (asymmetric and negatively transitive);

(A2) For all $p, q, r \in P$ and $\alpha \in [0, 1]$, $p \succ q \Rightarrow \alpha p + (1 - \alpha)r \succ \alpha q + (1 - \alpha)r$.

(A3) For all $p, q, r \in P$ such that $p \succ q \succ r$, $\exists \alpha, \beta \in (0, 1)$ such that

$$\alpha p + (1 - \alpha)r \succ q \succ \beta p + (1 - \beta)r$$



Theorem 3.2 (vNM)

\succ on P satisfies axioms (A1)-(A3) if and only if there exists a function $u : X \rightarrow \mathbb{R}$ such that

$$p \succ q \Leftrightarrow \sum_x p(x)u(x) > \sum_x q(x)u(x) \quad (*)$$

Moreover, u is unique up to a positive affine transformation: there is another u' represents \succ in the sense of (*) if and only if there exists $c > 0$ and d such that

$$u'(\cdot) = cu(\cdot) + d$$



Remark

- o If u represents \succ then so will $v(\cdot) = f(u(\cdot))$ for any **strictly increasing** f .
- o $k(p) = \sum_x p(x)u(x)$ gives an ordinal representation of \succ .

Lemma 3.1 (Four Lemmas obtained by the three axioms)

If \succ satisfies (A1) to (A3), then

(L1). If $p \succ q$ and $0 \leq \alpha < \beta \leq 1$, then

$$\beta p + (1 - \beta)q \succ \alpha p + (1 - \alpha)q$$

(L2). If $p \succeq q \succeq r$ and $p \succ r \Rightarrow$ there exists a unique $\alpha^* \in [0, 1]$ such that

$$q \sim \alpha^* p + (1 - \alpha^*)r$$

(L3). If $p \sim q$ and $\alpha \in [0, 1] \Rightarrow$ for all $r \in P$,

$$\alpha p + (1 - \alpha)r \succ \alpha q + (1 - \alpha)r$$

(L4). For any $x \in X$, let δ_x be the probability distribution degenerate at x , that is $\delta_x(x') =$

$$\begin{cases} 1, & \text{if } x' = x \\ 0, & \text{if } x' \neq x \end{cases}$$

For all $p \in P$, we have $x_1, x_2 \in X$ such that

$$\delta_{x_1} \succeq p \succeq \delta_{x_2}$$



3.3.2 Savage (1954)

Consider the situation that what the decision maker chooses depends critically on his/her subjectively assesses as the odds of the outcomes.

The basics of the Savage formulation:

- o a set of X of prizes/consequences;
- o a set S of the nature (states of the world).

Each $s \in S$ is a compilation of all characteristics/factors about which the DM is uncertain and which are relevant to the consequences that will result from her/his choice. The set S is an exhaustive list of mutually exclusive states — only one $s \in S$ will be the realized state.

We denote the choice space by H , as the set of all functions from S to X ($H = X^S$).

Savage seeks to find a subjective taste (the utility function) $u(\cdot)$ and a subjective belief (the probability measure) π such that

$$h \succ h' \Leftrightarrow \sum_{s \in S} \pi(s)u(h(s)) > \sum_{s \in S} \pi(s)u(h'(s))$$

Note that, it contains an assumption that $u(\cdot)$ is a function about x which doesn't depend on the state of the world when it receives x .

3.4 Social Choice

Notations:

1. We consider finite set of alternatives X and finite set of agents I .
2. We use \mathcal{B} to denotes the set of all preference relations.
3. We use $\mathcal{R} \subseteq \mathcal{B}$ to denotes the set of all rational preference relations.
4. We use $\succeq \in \mathcal{R}$ to represents individual rational preference relation.

3.4.1 Social Welfare Function and Properties

Definition 3.13 (Social Welfare Function (SWF))

A **social welfare function** (SWF) is a mapping

$$f : \mathcal{A} \subseteq \mathcal{R}^I \rightarrow \mathcal{B}$$

$\succeq = f(\succeq_1, \dots, \succeq_I)$ is interpreted as the **social preference relation**. It doesn't need to be rational (i.e., complete and transitive).



Definition 3.14 (SWF's Properties)

A social welfare function $f : \mathcal{A} \rightarrow \mathcal{B}$

- o has **unrestricted domain** (UD) if $\mathcal{A} = \mathcal{R}^n$;
- o is **transitive** (T) if $f(\succeq_1, \dots, \succeq_I)$ is transitive for all $(\succeq_1, \dots, \succeq_I) \in \mathcal{A}$;
- o is **nondictatorial** (ND) if there is no agent $i \in I$ such that $\forall \{x, y\} \subseteq X x \succeq_i y \Rightarrow x \succeq y$. (That is there is no distinguished voter who can choose the winner).
- o is **weakly Pareto** (PA) if, $\forall \{x, y\} \subseteq X$ and any preference profile $(\succeq_1, \dots, \succeq_I) \in \mathcal{A}$, we have $x \succeq_i y, \forall i \in I \Rightarrow x \succeq y$.
- o is **independent of irrelevant alternatives** (IIA) if, $\forall \{x, y\} \subseteq X$, and any \succeq and \succeq' with $\succeq_i|_{x,y} = \succeq'_i|_{x,y}, \forall i \in I$, if $x \succeq y$ then $x \succeq' y$.



3.4.2 Arrow's Theorem

Theorem 3.3 (Arrow's impossibility theorem)

Suppose $|X| \geq 3$, $\mathcal{A} = \mathcal{R}^I$ (UD). Then if a SWF f satisfies T, PA, and IIA, then it fails to be ND.



Proof 3.1

Yu, N. N. (2012). A one-shot proof of Arrow's impossibility theorem. *Economic Theory*, 523-525.

Chapter 4 Demand Theory

4.1 Utility Maximization Problem (UMP)

Budget set is given by $B = \{x \in X \subseteq \mathbb{R}_+^K : p \cdot x \leq w\}$, where w is the DM's wealth and p is the vector of prices. Without losing generality, we can assume $w = 1$.

The DM's problem is finding the \succeq -optimal bundle $x \in B(p)$. With the corresponding utility function $u(x)$, we can consider a consumer's problem

$$\begin{aligned} & \max_{x \in X} u(x) \\ & \text{s.t. } p \cdot x \leq w \end{aligned} \tag{UMP}$$

The set \succeq -optimal bundle is represented by $x(p, w)$.

4.1.1 Marshallian Demand: Existence and Properties

Proposition 4.1 (Continuous Preference \Rightarrow Solution $x(p, w)$ Existence)

If $\succeq (u(\cdot))$ is continuous, then all such problems have a solution $x(p, w)$.



Proof 4.1

By the Weierstrass Extreme Value Theorem.

Proposition 4.2 (Convex Preference \Rightarrow Convex $x(p, w)$)

If \succeq is convex ($u(\cdot)$ is quasi-concave), then $x(p, w)$ is convex.



Proof 4.2

Suppose $x, x' \in X$. The optimal utility $u^* = u(x) = u(x')$. For any $\alpha \in [0, 1]$, let $x'' = \alpha x + (1 - \alpha)x'$.

Because \succeq is convex, we have $u(\cdot)$ is quasi-concave, that is $u(x'') \geq u^*$. x'' is also feasible. So, $x'' \in x(p, w)$.

Proposition 4.3 (Strictly Convex Preference \Rightarrow Singleton $x(p, w)$)

If \succeq is strictly convex ($u(\cdot)$ is strictly quasi-concave), then $x(p, w)$ is (at most) a singleton.



Proposition 4.4 (Differentiable Preference \Rightarrow Marginal Utility equals to Price)

If \succeq is differentiable, $x^* \in x(p, w)$, and the vector of marginal values at x^* (as defined above) is denoted by $v(x^*) = (v_1(x^*), \dots, v_K(x^*))$, where $v_k(x^*)$ is usually taken by $\frac{\partial u}{\partial x_k}(x^*)$ in "classic" problem. Then,

we have

$$\frac{v_k(x^*)}{v_j(x^*)} = \frac{p_k}{p_j} \text{ for any } x_k^*, x_j^* > 0$$

and for any k with $x_k^* > 0$ (consumed commodity)

$$\frac{v_k(x^*)}{p_k} \geq \frac{v_j(x^*)}{p_j} \text{ for any } j \neq k \quad (*)$$



Corollary 4.1 (Sufficient Conditions for Optimality)

If \succeq is strongly monotonic, convex, continuous, and differentiable and if $p \cdot x^* = w$ and $(*)$ is satisfied then $x^* \in x(p, w)$



Definition 4.1 (Rationalize)

\succeq **fully rationalize** the demand function x if for any (p, w) , the bundle $x(p, w)$ is the unique \succeq -maximal bundle within B .

A monotonic \succeq **rationalize** the demand function x if for any (p, w) , the bundle $x(p, w)$ is a \succeq -maximal bundle within B .



The unique solution is called Marshallian (Uncompensated) Demand.

Proposition 4.5 (Properties of Marshallian Demand)

- (i). **Walras' Law:** If \succeq is local nonsatiation, $\forall x^* \in x(p, w) : p \cdot x^* = w$.
- (ii). **Homogeneity of degree zero in (p, w) :** $x(\alpha p, \alpha w) \equiv x(p, w)$, $\forall \alpha > 0$.
- (iii). **Continuous in prices and in wealth if the \succeq is continuous.**



Proposition 4.6 (Weak Axiom of Revealed Preference of Marshallian Demand)

If demand is single valued then WARP(3.8) is equivalent to

$$p \cdot y' \leq w \text{ and } y \neq y' \Rightarrow p' \cdot y > w$$

where $y \equiv x(p, w)$ and $y' \equiv x(p', w')$. (y' is feasible under (p, w) but $y = x(p, w)$, which means y is better and it can't be feasible under (p', w') .)



4.1.2 Lagrangian Approach: $\frac{\partial u(x^*)}{\partial x_i} = \lambda^* p_i$ and $\lambda^* \left(x_i(p, w) + \sum_{j=1}^K p_j \frac{\partial x_j}{\partial p_i} \right) = 0$

The Lagrangian of the problem is

$$L(x, \lambda) = u(x) - \lambda(p \cdot x - w)$$

By the KKT necessary conditions, we have

$$\frac{\partial L}{\partial x_i}(x^*, \lambda^*) = \frac{\partial u(x^*)}{\partial x_i} - \lambda^* p_i = 0, \forall i = 1, \dots, K$$

$$\lambda^* \geq 0 \text{ and } \lambda^*(p \cdot x^* - w) = 0$$

Based on that, we have

Lemma 4.1

- (i). $\frac{\partial u(x^*)}{\partial x_i} = \lambda^* p_i;$
- (ii). $\lambda^* \left(x(p, w) + p \cdot \frac{\partial x(p, w)}{\partial p} \right) = 0 \text{ i.e., } \lambda^* \left(x_i(p, w) + \sum_{j=1}^K p_j \frac{\partial x_j}{\partial p_i} \right) = 0.$



4.1.3 Envelope Theorem $\Rightarrow \lambda^* = \frac{\partial u(x(p, w))}{\partial w}$

Theorem 4.1 (Envelope Theorem)

Consider the constrained maximization problem,

$$\max_{x \in \mathbb{R}^n} f(x; \theta)$$

$$\text{s.t. } g(x; \theta) \leq 0$$

where $x \in \mathbb{R}^n$ is the choice variable and $\theta \in \mathbb{R}^m$ is some parameter. Let f, g be continuously differentiable real-valued functions.

- Let the value function of the problem be $V(\theta) \triangleq f(x^*(\theta), \theta).$
- The Lagrangian for this problem is

$$L(x, \lambda; \theta) = f(x; \theta) - \lambda g(x; \theta)$$

- Let x^* and λ^* denote the optimized values of the variables.

(By KKT necessary conditions, we have $\frac{\partial f}{\partial x}(x^*; \theta) = \lambda^* \frac{\partial g}{\partial x}(x^*; \theta)$ and $\lambda^* g(x^*; \theta) = 0$)

Then the following is true for any $\bar{\theta} \in \mathbb{R}^m$

$$\frac{\partial V}{\partial \theta_i}(\bar{\theta}) = \frac{\partial L}{\partial \theta_i}(x^*, \lambda^*; \bar{\theta}) = \frac{\partial f}{\partial \theta_i}(x^*; \bar{\theta}) - \lambda^* \frac{\partial g}{\partial \theta_i}(x^*; \bar{\theta})$$



Proof 4.3

The proof of the envelope theorem is a straightforward calculation.

Firstly, by KKT necessary conditions, we have $\frac{\partial f}{\partial x}(x^*; \bar{\theta}) = \lambda^* \frac{\partial g}{\partial x}(x^*; \bar{\theta})$ and $\lambda^* g(x^*; \bar{\theta}) = 0 \Rightarrow$

$\lambda^* \left[\frac{\partial g}{\partial x}(x^*; \bar{\theta}) \frac{\partial x^*(\bar{\theta})}{\partial \theta_i} + \frac{\partial g}{\partial \theta_i}(x^*; \bar{\theta}) \right] = 0$. Then we have

$$\begin{aligned}\frac{\partial V}{\partial \theta_i}(\bar{\theta}) &= \frac{\partial f}{\partial \theta_i}(x^*; \bar{\theta}) + \frac{\partial f}{\partial x}(x^*; \bar{\theta}) \frac{\partial x^*(\bar{\theta})}{\partial \theta_i} \\ &\quad \left(\text{by } \frac{\partial f}{\partial x}(x^*; \bar{\theta}) = \lambda^* \frac{\partial g}{\partial x}(x^*; \bar{\theta}) \right) \\ &= \frac{\partial f}{\partial \theta_i}(x^*; \bar{\theta}) + \lambda^* \frac{\partial g}{\partial x}(x^*; \bar{\theta}) \frac{\partial x^*(\bar{\theta})}{\partial \theta_i} \\ &\quad \left(\text{by } \lambda^* \left[\frac{\partial g}{\partial x}(x^*; \bar{\theta}) \frac{\partial x^*(\bar{\theta})}{\partial \theta_i} + \frac{\partial g}{\partial \theta_i}(x^*; \bar{\theta}) \right] = 0 \right) \\ &= \frac{\partial f}{\partial \theta_i}(x^*; \bar{\theta}) - \lambda^* \frac{\partial g}{\partial \theta_i}(x^*; \bar{\theta})\end{aligned}$$

Corollary 4.2

$$\lambda^* = \frac{\partial u(x(p, w))}{\partial w}.$$



Proof 4.4

By the envelope theorem, we have $\frac{\partial u(x(p, w))}{\partial w} = \frac{\partial L}{\partial w}|_{x^*, \lambda^*} = \lambda^*$.

4.1.4 Indirect Utility Function $v(p, w) \equiv u(x(p, w))$

Proposition 4.7 (Properties of Indirect Utility Function)

1. $v(p, w)$ is homogeneous of degree zero in (p, w) ;
2. $v(p, w)$ is strictly increasing in w and non-increasing in p_i ;
3. $v(p, w)$ is quasi-convex, that is the set $\{p : v(p, w) \leq u\}$ is convex for all $u \in \mathbb{R}$.
4. $\lambda^* = \frac{\partial v(p, w)}{\partial w}$ (Corollary 4.2).



4.1.5 Roy's Identity $x_i^* = -\frac{\frac{\partial v}{\partial p_i}}{\frac{\partial v}{\partial w}}$: recover $x(p, w)$ from $v(p, w)$

Proposition 4.8 (Roy's Identity)

$$x_i^*(p, w) = -\frac{\frac{\partial v(p, w)}{\partial p_i}}{\frac{\partial v(p, w)}{\partial w}}.$$



Proof 4.5

By the definition,

$$v(p, w) \equiv u(x(p, w))$$

$$\begin{aligned} \frac{\partial v}{\partial p_i} &\equiv \sum_{j=1}^K \frac{\partial u}{\partial x_j} \frac{\partial x_j}{\partial p_i} \\ &= \sum_{j=1}^K \lambda^* p_j \frac{\partial x_j}{\partial p_i} && \text{by } \frac{\partial u(x^*)}{\partial x_i} = \lambda^* p_i \\ &= -\lambda^* x_i^* && \text{by } \lambda^* \left(x_i(p, w) + \sum_{j=1}^K p_j \frac{\partial x_j}{\partial p_i} \right) = 0 \\ x_i^* &= -\frac{\frac{\partial v}{\partial p_i}}{\frac{\partial v}{\partial w}} && \text{by } \lambda^* = \frac{\partial v(p, w)}{\partial w} \end{aligned}$$

4.2 Expenditure Minimization Problem (EMP)

Consider the duality

$$\begin{aligned} \min_{x \in X} \quad & p \cdot x \\ \text{s.t.} \quad & u(x) \geq u \end{aligned} \tag{EMP}$$

The optimal solutions are represented by $h(p, u)$. With uniqueness, we call it *Hicksian (compensated) demand*.

4.2.1 Hicksian Demand $h(p, u)$: Properties

Proposition 4.9 (Properties of Hicksian Demand)

(i). $h(p, u)$ is homogeneous of degree zero in p :

$$h(tp, u) = h(p, u), \forall t \in \mathbb{R}_+$$

(ii). $u(x)$ is strictly quasi-concave $\Rightarrow h(p, u)$ is unique;

(iii). For $u > u(0)$ and $u(\cdot)$ is locally non-satiated, constraint is active: for all $x^* \in h(p, u)$,

$$u(x^*) = u$$

Lemma 4.2 ($\sum_{j=1}^K \frac{\partial u}{\partial x_j} \frac{\partial h_j}{\partial p_i} = 0$)

If $u(x)$ is strictly quasi-concave, the Hicksian demand satisfies $\sum_{j=1}^K \frac{\partial u}{\partial x_j} \frac{\partial h_j}{\partial p_i} = 0$

Proof 4.6

$$u(h(p, u)) \equiv u \Rightarrow \sum_{j=1}^K \frac{\partial u}{\partial x_j} \frac{\partial h_j}{\partial p_i} = 0.$$

4.2.2 Expenditure Function $e(p, u) \equiv p \cdot h(p, u)$

Given the Hicksian demand $h(p, u)$, we can define the expenditure function as $e(p, u) \equiv p \cdot h(p, u)$.

Proposition 4.10 (Properties of Expenditure Function)

(i). $e(p, u)$ is homogeneous of degree 1 in p :

$$e(tp, u) = tp \cdot h(tp, u) = tp \cdot h(p, u) = te(p, u)$$

(ii). $e(p, u)$ is strictly increasing in u , non-decreasing in p_i ;

(iii). $e(p, u)$ is **concave** in p ;

(iv). $e(p, u)$ is continuous in p for all $p >> 0$;

(v). For all $x^* \in h(p, u)$, $x^* \in x(p, e(p, u))$;

(vi). For $w > 0$, $e(p, v(p, w)) \equiv w$;

(vii). $e(p, u)$'s derivative property:

$$\frac{\partial e(p, u)}{\partial p_i} \equiv h_i(p, u)$$



Proof 4.7 (Proof for concavity)

Suppose the price of good 1 increases from p_1^0 to p_1^1 : $p^0 \rightarrow p^1$. Set $p^a = ap^0 + (1 - a)p^1$, $0 \leq a \leq 1$.

So, $p^0 \leq p^a \leq p^1$ and

$$\begin{aligned} e(p^a, u) &= p^a \cdot h(p^a, u) \\ &= (ap^0 + (1 - a)p^1) \cdot h(p^a, u) \\ &= a[p^0 \cdot h(p^a, u)] + (1 - a)[p^1 \cdot h(p^a, u)] \\ h(p^a, u) \text{ is feasible in both EMP, but not optimal solutions} \\ &\geq a[p^0 \cdot h(p^0, u)] + (1 - a)[p^1 \cdot h(p^1, u)] \\ &= ae(p^0, u) + (1 - a)e(p^1, u) \end{aligned}$$

Proof 4.8 (Proof for Derivative)1. Direct proof:

$$\begin{aligned}
 e(p, u) &\equiv p \cdot h(p, u) \\
 \frac{\partial e}{\partial p_i} &\equiv \sum_{j=1}^K p_j \frac{\partial h_j}{\partial p_i} + h_i \\
 &\equiv \sum_{j=1}^K \frac{1}{\lambda^*} \frac{\partial u(x^*)}{\partial x_j} \frac{\partial h_j}{\partial p_i} + h_i \quad \text{by } \frac{\partial u(x^*)}{\partial x_i} = \lambda^* p_i \\
 &= h_i \quad \text{by } \sum_{j=1}^K \frac{\partial u}{\partial x_j} \frac{\partial h_j}{\partial p_i} = 0
 \end{aligned}$$

2. Envelope Theorem Proof:

$$\begin{aligned}
 L(x, \lambda; (p, u)) &= p \cdot x - \lambda(u(x) - u) \\
 \frac{\partial e(p, u)}{\partial p_i} &= \left. \frac{\partial L(x, \lambda; (p, u))}{\partial p_i} \right|_{x^* = h(p, u)} = x_i|_{x^* = h(p, u)} = h_i(p, u)
 \end{aligned}$$

4.2.3 Law of Compensated Demand: $\frac{\partial h_i}{\partial p_i} \leq 0$ **Corollary 4.3 (Law of Compensated Demand)**

Hicksian demand is downward sloping in its own price,

$$\frac{\partial h_i}{\partial p_i} \leq 0$$

**Proof 4.9**

By the concavity of $e(p, u)$ (4.10), we can conclude $\nabla^2 e(p, u) \preceq 0$ (negative semi-definite). Then, we know its diagonal elements are non-positive $\frac{\partial e^2}{\partial^2 p_i} = \frac{\partial h_i}{\partial p_i} \leq 0$.

4.2.4 Shifts in Hicksian Demand: $\frac{\partial h_i}{\partial u} \equiv \frac{\partial x_i}{\partial w} \frac{\partial e}{\partial u}$, same direction as $\frac{\partial x_i}{\partial w}$

How does Hicksian demand curve shift when utility changes?

$$\begin{aligned}
 h_i(p, u) &\equiv x_i(p, e(p, u)) \\
 \frac{\partial h_i}{\partial u} &\equiv \frac{\partial x_i}{\partial w} \frac{\partial e}{\partial u}
 \end{aligned}$$

We know $\frac{\partial e}{\partial u} > 0$, so the direction of Hicksian demand shift is the same as $\frac{\partial x_i}{\partial w}$.

- Normal good: increasing utility shifts h_i to the right.
- Inferior good: increasing utility shifts h_i to the left.

4.3 UMP and EMP

4.3.1 Slutsky Equation: substitution effect and income effect

Slutsky: how change of p_j (price in good j) affects x_i (the demand of product i).

Proposition 4.11 (Slutsky Equation)

$$\frac{\partial x_i(p, w)}{\partial p_j} = \underbrace{\frac{\partial h_i(p, u)}{\partial p_j}}_{\text{substitution effect}} - \underbrace{\frac{\partial x_i(p, w)}{\partial w} x_j(p, w)}_{\text{income effect}}$$



Proof 4.10

$$\begin{aligned} h_i(p, u) &\equiv x_i(p, e(p, u)) \\ \frac{\partial h_i}{\partial p_j} &\equiv \frac{\partial x_i}{\partial p_j} + \frac{\partial x_i}{\partial w} \frac{\partial e}{\partial p_j} \\ &\equiv \frac{\partial x_i}{\partial p_j} + \frac{\partial x_i}{\partial w} h_j(p, u) \\ &\equiv \frac{\partial x_i}{\partial p_j} + \frac{\partial x_i}{\partial w} x_j(p, e(p, u)) \end{aligned}$$

- **Substitution effect:** $\frac{\partial h_i}{\partial p_j}$, the change of relative prices change with constant utility will change the x_i .
- **Income (Wealth) effect:** $-\frac{\partial x_i}{\partial w} x_j(p, w)$, the change of price can be seen as change of wealth, which will also impact the x_i .

4.3.2 Relationship Between UMP and EMP

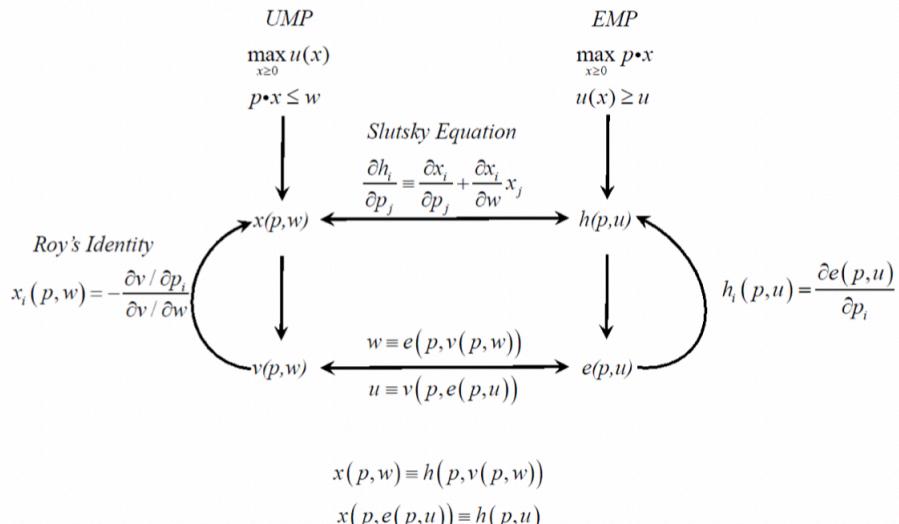


Figure 4.1: Relationship Between UMP and EMP

Chapter 5 General Equilibrium

5.1 Exchange Economy

1. There are L perfectly divisible commodities indexed by $l = 1, \dots, L$ over \mathbb{R}^L .
2. There are m agents, indexed by $i = 1, \dots, m$. $N = \{1, \dots, m\}$.
3. Each agent has a preference relation \succeq_i on \mathbb{R}_+^L represented by a utility function $u_i : \mathbb{R}_+^L \rightarrow \mathbb{R}$.
4. Each agent has a vector of initial endowments $w_i \in \mathbb{R}_+^L$.
5. The aggregate endowment is $w = \sum_{i=1}^m w_i$.

Example 5.1 (Endowment Box Economy) The endowment box economy has 2 goods ($L = 2$) and 2 agents ($m = 2$). The commodity space is \mathbb{R}^2 .

Each agent's consumption set is $\mathbb{R}_+^2 = \{x \in \mathbb{R}^2 : x \geq 0\}$.

Each agent $i = a, b$ has preference relation \succ_i over \mathbb{R}_+^2 represented by a utility function $u_i : \mathbb{R}_+^2 \rightarrow \mathbb{R}$.

Each agent has a vector of initial endowments $w_i = (w_{i1}, w_{i2}) \in \mathbb{R}^2$.

Definition 5.1 (Allocation)

An **allocation** in an exchange economy is an assignment of goods to agents $x = (x_1, \dots, x_m) \in \mathbb{R}_+^{L \times m}$ such that $\sum_{i=1}^m x_i = w$.



5.1.1 Pareto Optimal/Efficient

Definition 5.2 (Pareto Optimal)

An allocation x is **Pareto optimal/efficient** if there doesn't exist an allocation y s.t. $u_i(y_i) \geq u_i(x_i)$ ($y_i \succeq_i x_i$) for each i and $u_j(y_j) > u_j(x_j)$ ($y_j \succ_j x_j$) for some j .



Consider the following social planner's problem:

- o Fix an agent j and $\{\bar{u}_i\}_{i \neq j}$,

$$\begin{aligned}
 & \max_{(x_1, \dots, x_n) \in \mathbb{R}_+^{L \times m}} u_j(x_j) \\
 & \text{s.t. } u_i(x_i) \geq \bar{u}_i, i \neq j \\
 & \quad \sum_{i=1}^m x_i = w \\
 & \quad x_i \geq 0, \forall i
 \end{aligned} \tag{P}$$

Proposition 5.1 (P.O. \Leftrightarrow Solutions of Problem (P))

Suppose each agent's utility function is continuous and strongly monotone. Then, an allocation x^* in an exchange economy is Pareto-Optimal iff it is a solution of Problem (P) for some choice of $\{\bar{u}_i\}_{i \neq j}$.

Proof 5.1

1. " \Leftarrow ": Suppose x^* is a solution to Problem (P) for $\{\bar{u}_i\}_{i \neq j}$. Suppose by the way of contradiction that x^* is not Pareto-Optimal. Then there is another allocation \hat{x} such that

- (i). Either: $u_j(\hat{x}_j) > u_j(x_j^*)$ and $u_i(\hat{x}_i) \geq u_i(x_i^*)$ for all $i \neq j$.
- (ii). Or: $u_j(\hat{x}_j) \geq u_j(x_j^*)$, $u_k(\hat{x}_k) \geq u_k(x_k^*)$ for some $k \neq j$, and $u_i(\hat{x}_i) \geq u_i(x_i^*)$ for all $i \neq j, k$.

Suppose (i) holds: Since \hat{x} is an allocation, $\sum_{i=1}^m \hat{x}_i = m$ and $\hat{x}_i \geq 0, \forall i$. By assumption and x^* is solution of Problem (P), $u_i(\hat{x}_i) \geq u_i(x_i^*) \geq \bar{u}_i$ for all $i \neq j$. So, \hat{x} satisfies the constraints of Problem (P). Because $u_j(\hat{x}_j) > u_j(x_j^*)$, x^* is not the solution to Problem (P). Contradiction!

Suppose (ii) holds: Prove by constructing another allocation \tilde{x} as follows: By continuity, $\exists \epsilon > 0$ sufficiently small s.t. $u_k((1 - \epsilon)\hat{x}_k) \geq u_k(x_k^*)$. Set $\tilde{x}_k = (1 - \epsilon)\hat{x}_k$, $\tilde{x}_j = \hat{x}_j + \epsilon\hat{x}_k$, and $\tilde{x}_i = \hat{x}_i$ for all $i \neq j, k$. Then, $\sum_{i=1}^m \tilde{x}_i = \sum_{i=1}^m \hat{x}_i = w$, $u_i(\tilde{x}_i) \geq u_i(x_i^*) \geq \bar{u}_i$ for all $i \neq j$ and $u_j(\tilde{x}_j) > u_j(x_j^*) \geq \bar{u}_j$ (by strong monotonicity). Hence, x^* is not the solution to Problem (P). Contradiction!

2. " \Rightarrow ": Suppose x^* is Pareto-Optimal. Set $\bar{u}_i = u_i(x^*)$ for all $i \neq j$.

Claim 5.1

x^* solves Problem (P) given $\{\bar{u}_i\}_{i \neq j}$.

Firstly, x^* is feasible for Problem (P). Then, suppose by the way of contradiction that there is another allocation x' such that $\sum_{i=1}^m x'_i = w$, $u_i(x'_i) \geq \bar{u}_i = u_i(x_i^*)$ for all $i \neq j$, and $u_j(x'_j) > u_j(x_j^*)$. Hence, x^* is not Pareto-Optimal, which is a contradiction!

Proposition 5.2

From the first-order condition (FOC) of Problem (P), a necessary condition for interior Pareto-Optimal allocations when each u_i is also differentiable is

- o $Du_j(x_j^*) = \lambda_i Du_i(x_i^*)$ for some $\lambda_i > 0$ and $\forall i \neq j$ where $x_i^* >> 0, \forall i$.

5.1.2 Individually Rational, Block, Core

Are all Pareto-Optimal allocations equally likely are reasonable?

How the initial endowment allocation affects the Pareto-Optimal allocation?

One agent should block any proposed trades leading to allocations that generate lower utility.

Definition 5.3 (Individually Rational)

A bundle x_i is **individually rational** (IR) for agent i if $x_i \succeq_i w_i$.

An allocation $x = (x_1, \dots, x_m)$ is **individually rational** (IR) if $x_i \succeq_i w_i$ for all $i = 1, \dots, m$.



Let $N := \{1, \dots, m\}$ be the set of agents. A **coalition** is a nonempty subset $S \subseteq N$.

Example 5.2 With two agents $\{a, b\}$, there are 3 possible coalitions: $\{a\}, \{b\}, \{a, b\}$.

Definition 5.4 (Block)

A coalition S can **block** an allocation $x = (x_1, \dots, x_m)$ if there exists bundles $y_i \in \mathbb{R}_+^L$ for all $i \in S$ s.t.

1. $\sum_{i \in S} y_i = \sum_{i \in S} w_i$
2. $y_i \succeq_i x_i$ for all $i \in S$;
3. $y_j \succ_j x_j$ for some $j \in S$.



Definition 5.5 (Core)

The **core** is the set of allocations that cannot be blocked by any coalition.



Note (Core and P.O.)

- o Every allocation in the core is Pareto-Optimal (directly by definition).
- o Not every Pareto-Optimal allocation is in the core.
- o For the two agent case, the core is the set of individually rational Pareto-Optimal allocations.

5.1.3 Competitive Equilibrium

Assumption Suppose there are markets for all available goods and all agents are price-takers in these markets.

Given a vector of price $p \in \mathbb{R}^L$, agent i chooses x_i^* to solve the following problem:

$$\max_{x_i \in \mathbb{R}_+^L} u_i(x_i)$$

$$s.t. p \cdot x_i \leq p \cdot w_i$$

Definition 5.6 (Competitive Equilibrium)

Given endowment $w = (w_i)_{i \in N}$. A **competitive (Walrasian) equilibrium** in an exchange economy is a pair $p^* \in \mathbb{R}^L$ (price vector over L goods) and an allocation $x^* = (x_i^*)_{i \in N}$ such that:

- (i). $x_i^* \in \text{argmax } u_i(x)$ s.t. $p^* \cdot x_i \leq p^* \cdot w_i, \forall i \in N$.
- (ii). $\sum_{i \in N} x_i^* = w$.

We call $x^* = (x_i^*)_{i \in N}$ the competitive equilibrium (Walrasian) allocation

and p^* the competitive equilibrium (Walrasian) price vector.



Demand notations:

Definition 5.7 (Excess Demand)

Let $x_i(p) := x_i(p, p \cdot w_i)$ denote agent i 's **demand** given the price vector $p \in \mathbb{R}^L$ and income $p \cdot w_i$.

Agent i 's **individual excess demand at p** is $x_i(p) - w_i$.

The **aggregate excess demand** at p is $\sum_{i \in N} x_i(p) - w$.



Note (Excess Demand and Competitive Equilibrium)

- p^* is a competitive equilibrium price vector if and only if $0 \in \sum_{i \in N} x_i(p) - w$
- (x^*, p^*) is a competitive equilibrium if and only if x^* satisfies $x_i^* \in x_i(p^*), \forall i$ and $\sum_{i \in N} x_i^* - w = 0$.

5.1.4 First Welfare Theorem: CE \Rightarrow P.O.

Given non-satiated preference, every CE is P.O. (P.E.).

Theorem 5.1 (First-order (fundamental) Welfare Theorem: CE \Rightarrow P.O.)

If each agent's preference relation is locally non-satiated, then every competitive equilibrium allocation is Pareto optimal (Pareto efficient).



Proof 5.2

Let $x^* = (x_1^*, \dots, x_n^*)$ be a CE allocation with corresponding CE price vector p^* .

Suppose by way of contradiction that x^* is not P.O. allocation. Then, there is another allocation $\hat{x} = (\hat{x}_1, \dots, \hat{x}_n)$ such that $\hat{x}_j \succ_j x_j^*$ for some $j \in N$ and $\hat{x}_i \succeq_i x_i^*$ for all $i \neq j$.

By the definition of CE, \hat{x}_j should not be affordable for j , i.e., $p^* \hat{x}_j > p^* w_j$. By the definitions of non-satiated and CE, $p^* \hat{x}_i \geq p^* w_i$ for all $i \neq j$. (If $p^* \hat{x}_i < p^* w_i$, $\exists \tilde{x}_i$ s.t. $p^* \tilde{x}_i \leq p^* w_i$ and $\tilde{x}_i \succ_i \hat{x}_i \succeq_i x_i^*$, which contradicts to the definition of CE.)

Add up all inequalities, we get

$$p^* \cdot \left(\sum_{i=1}^n \hat{x}_i \right) > p^* \cdot \left(\sum_{i=1}^n w_i \right) = p^* \cdot w$$

which contradicts to the definition of a feasible allocation that $\sum_i^n \hat{x}_i = w$. Hence, x^* is P.O.



Note This requires **only local non-satiation** of preferences. In particular, does not require convexity of preferences.

5.1.5 CE \Rightarrow IR; CE \subseteq P.O. \cap Core



Note [CE \Rightarrow IR] At any prices p , an agent can always afford their initial endowment, so by revealed preference, every CE allocation is individually rational.

Corollary 5.1 (CE Allocation is in Core)

If each agent's preference relation is locally non-satiated, then every CE allocation is in the core.

**Proof 5.3**

In exercise.

Note Not every P.O. allocation is in the core. But every CE allocation is a P.O. allocation in the core.

$$\text{CE} \subseteq \text{P.O.} \cap \text{Core}$$

5.1.6 Equilibrium with Transfers

What scope does planner have for redistribution using only decentralized market mechanism?

Not every P.O. allocation is “equitable.” To implement a more ”equitable” allocation. Some possible mechanisms:

- will need taxes or transfers (should be budget-balancing, i.e., no money leaves the economy).
- taxes/transfers should be lump-sum.

Definition 5.8 (“Supportable” as a Price Equilibrium with Transfers)

An allocation x^* is **supportable** as a **price equilibrium with transfers** if there exists a price vector $p^* \in \mathbb{R}_+^L$ and lump-sum budget-balancing transfers $\{T_i : i = 1, \dots, m\}$ so that $\sum_{i=1}^m T_i = 0$, such that $\forall i$:

$$x_i^* \in \arg \max_{x \in \mathbb{R}_+^L \text{ s.t. } p^* \cdot x_i \leq p^* \cdot w_i + T_i} u_i(x_i)$$



5.1.7 Second Welfare Theorem: sufficient condition for P.O. be supported as a price equilibrium with transfers

Remark Is every P.O. allocation supportable as a price equilibrium with transfers? **No.** (e.g. a non-convex indifference curve (preference relation): for a P.O. allocation, there exists an allocation such that gives a bundle with lower cost but equal utility for an agent.)

Formally, a sufficient condition can be given:

Theorem 5.2 (Second Welfare Theorem)

If each consumers' preference relation is convex, continuous, and strongly monotone, then every interior P.O. allocation in an exchange economy can be supported as a price equilibrium with transfers.



To give the proof of the theorem, we need firstly give some definitions and results.

Definition 5.9 (Supported by a price; Supported)

An allocation $\bar{x} = (\bar{x}_1, \dots, \bar{x}_m)$ in an exchange economy is **supported by a non-zero price vector** $p \in \mathbb{R}^L$ if

$$\forall i : x_i \succeq_i \bar{x}_i \Rightarrow p \cdot x_i \geq p \cdot \bar{x}_i$$

If an allocation $\bar{x} = (\bar{x}_1, \dots, \bar{x}_m)$ is **supported**, then a common price p supports each agent's "better-than set" at \bar{x}_i ($\{x_i \in \mathbb{R}_+^L : x_i \succeq_i \bar{x}_i\}$): $\forall x_i \in \{x_i \in \mathbb{R}_+^L : x_i \succeq_i \bar{x}_i\} : p \cdot x_i \geq p \cdot \bar{x}_i$.



Note An allocation is supported as a price equilibrium with transfers \Leftrightarrow the allocation that is supported (i.e., all agents' bundles are supported by a common price).

Recall:

Theorem 5.3 (Separating Hyperplane Theorem)

Let $A, B \subseteq \mathbb{R}^n$ be non-empty disjoint, convex sets. Then $\exists p \in \mathbb{R}^n, p \neq 0$, s.t.

$$p \cdot a \leq p \cdot b, \forall a \in A, \forall b \in B$$

**Proof 5.4 (Second Welfare Theorem 5.2)**

Let $x^* = (x_1^*, \dots, x_m^*)$ be an interior P.O. allocation, so $x_i^* >> 0, \forall i$. Let

$$P_i := \{x_i \in \mathbb{R}_+^L : u_i(x_i) > u_i(x_i^*)\}, \forall i$$

Properties about P_i :

- (1). By strong monotonicity, $P_i \neq \emptyset$ (interior allocation) for all i .
- (2). By convexity, P_i is convex for all i .

Let

$$\begin{aligned} P &:= P_1 + \cdots + P_m \\ &= \left\{ z \in \mathbb{R}_+^L : z = \sum_{i=1}^m x_i \text{ for some } (x_1, \dots, x_m) \in \mathbb{R}_+^{L \times m} \text{ s.t. } u_i(x_i) > u_i(x_i^*), \forall i \right\} \end{aligned}$$

Properties about P :

- (1). By construction, $P \neq \emptyset$.
- (2). P is convex: because the sum of convex sets is convex.
- (3). $w \notin P$: this follows from the Pareto optimality of $x^* = (x_1^*, \dots, x_m^*)$.

As $\{w\}$ is convex and $\{w\} \cap P = \emptyset$, by the Separating Hyperplane Theorem 5.3, $\exists p \in \mathbb{R}^L, p \neq 0$ s.t.

$p \cdot z \geq p \cdot w$ for all $z \in P$.

- o Fix j and suppose $u_j(x_j) > u_j(x_j^*)$ for some $x_j \in \mathbb{R}_+^L$. By continuity, $\exists \epsilon \in (0, 1)$ sufficient small s.t. $u_j((1 - \epsilon)x_j) > u_j(x_j^*)$. Let $y_j := (1 - \epsilon)x_j$. For $i \neq j$: set $y := x_i^* + \frac{\epsilon}{m-1}x_j$. By

strong monotonicity, $u_i(y_i) > u_i(x_i^*)$, $\forall i \neq j$. So, $\sum_{i=1}^m y_i \in P$ by the definition of P .

Then,

$$\begin{aligned} p \cdot \left(\sum_{i=1}^m y_i \right) &\geq p \cdot w \\ \text{By } \sum_{i=1}^m x_i^* = w, \quad p \cdot \left(\sum_{i=1}^m y_i - \sum_{i=1}^m x_i^* \right) &\geq 0 \\ p \cdot (x_j - x_j^*) &\geq 0 \\ p \cdot x_j &\geq p \cdot x_j^* \end{aligned}$$

That is, with p , $u_j(x_j) > u_j(x_j^*) \Rightarrow p \cdot x_j \geq p \cdot x_j^*$.

By strong monotonicity, $u_j(x_j^* + (0, 0, \dots, 0, 1, 0, \dots, 0)) > u_j(x_j^*)$, hence, $p \cdot (x_j^* + (0, 0, \dots, 0, 1, 0, \dots, 0)) \geq p \cdot x_j^* \Rightarrow p \cdot (0, 0, \dots, 0, 1, 0, \dots, 0) \geq 0$. That is, $p_i \geq 0, \forall i$. By definition, $p \neq 0$, $p > 0$.

By assumption $x_j^* >> 0$, $p \cdot x_j^* > 0$. Now suppose $\exists x_j \in \mathbb{R}_+^L$ s.t. $u_j(x_j) > u_j(x_j^*)$ and $p \cdot x_j = p \cdot x_j^*$. By continuity, $\exists \delta \in (0, 1)$ s.t. $u_j(\delta x_j) > u_j(x_j^*)$. By what we show above, $u_j(x_j) > u_j(x_j^*) \Rightarrow p \cdot x_j \geq p \cdot x_j^*$. We have $p \cdot x_j > \delta p \cdot x_j = p \cdot (\delta x_j) \geq p \cdot x_j^* > 0$. There is a contradiction. Hence, we prove that

$$u_j(x_j) > u_j(x_j^*) \Rightarrow p \cdot x_j > p \cdot x_j^*$$

- o Let the transfers be $T_i := p \cdot x_i^* - p \cdot w_i, \forall i$ such that $\sum_i T_i = p \cdot (\sum_i x_i^* - \sum_i w_i) = 0$.

All in all,

$$x_i^* \in \arg \max_{x \in \mathbb{R}_+^L \text{ s.t. } p^* \cdot x_i \leq p^* \cdot w_i + T_i} u_i(x_i)$$

x^* is a price equilibrium with transfers $\{T_i\}_i$ and the price vector p .

5.1.8 Second Welfare Theorem: P.O. with Endowments Used \Rightarrow CE

Theorem 5.4 (Second Welfare Theorem (corollary))

Suppose that interior x^* is Pareto efficient and consumers receive endowment worth $p \cdot w^i = p \cdot x^{i*}$ for all $i = 1, \dots, m$. Then, if a competitive equilibrium exists for such w , then x^* is a competitive equilibrium allocation.



Proof 5.5

By the Second Welfare Theorem 5.2, interior P.O. allocation x^* can be supported by transfers $\{T_i\}_{i=1}^m$. Then, $p \cdot x_i \leq p \cdot w_i + T_i$. Because $p \cdot w^i = p \cdot x^{i*}$, $T_i = 0, \forall i$. So, x^* is exactly a competitive equilibrium allocation.

5.1.9 Walras' Law in Competitive Equilibrium

Recall that “ p is a competitive equilibrium price vector” $\Leftrightarrow “0 \in \sum_{i=1}^m x_i(p) - w.”$

 **Note** Only relative prices matter, as the Marshallian demand has homogeneity of degree zero: $x(\lambda p) = x(p)$.

Hence, if p^* is a competitive equilibrium price vector, so is λp^* , $\forall \lambda$, which correspond to the same competitive equilibrium.

Remark Are markets independent? No.

If \succeq_i is locally non-satiated for all i , then $\forall i: p \cdot x_i(p) = p \cdot w_i, \forall p$. Adding over agents: $p \cdot \sum_{i=1}^m x_i(p) = p \cdot w, \forall p$.

This is **Walras' Law** in aggregate level: $p \cdot [\sum_{i=1}^m x_i(p) - w] = 0, \forall p$.

Remark If Walras' Law holds and there exists $p^* >> 0$ such that all markets but one clear, then the p^* must clear the last market too.

Let $Z(p) = \sum_{i=1}^m x_i(p) - w$. By Walras' Law, $p \cdot Z(p) = 0, \forall p$. Suppose that exists $p^* >> 0$ such that $Z_l(p^*) = 0, l = 1, \dots, L-1$. Then,

$$0 = p^* \cdot Z(p^*) = \sum_{l=1}^L p_l^* \cdot Z_l(p^*) = p_L^* Z_L(p^*)$$

$$p_L^* > 0 \Rightarrow Z_L(p^*) = 0$$

5.2 Private Ownership Production Economy

1. There are L perfectly divisible goods. The commodity space is \mathbb{R}^L .
2. There are m consumers. Each consumer $i = 1, \dots, m$ has a preference relation \succeq_i represented by a utility function $u_i : \mathbb{R}_+^L \rightarrow \mathbb{R}$, an initial endowment $w_i \in \mathbb{R}_+^L$, and owns shares $\{\theta_{ij} : j = 1, \dots, J\}$ in the J firms, where $\theta_{ij} \geq 0, \forall i, j$ and $\sum_{i=1}^m \theta_{ij} = 1, \forall j$.
3. There are J firms. Each firm $j = 1, \dots, J$ has a production set $Y_j \subseteq \mathbb{R}^L$ that is nonempty, (representing the constraints of production).

 **Note** Standard sign convention regarding net output vectors: y represents net output;

$y_k \leq 0 \Rightarrow$ good k is a net input in y ;

$y_k \geq 0 \Rightarrow$ good k is a net output in y .

4. The set of allocation is

$$\mathcal{A} := \left\{ (x, y) = (\underbrace{x_1, \dots, x_m}_{\text{consumption}}, \underbrace{y_1, \dots, y_J}_{\text{production}}) \in \mathbb{R}^{L \times m} \times \mathbb{R}^{L \times J} : \sum_{i=1}^m x_i = \sum_{j=1}^J y_j + \sum_{i=1}^m w_i, y_j \in Y_j, \forall j \right\} \quad (\text{A})$$

Given $p \in \mathbb{R}^L$, firm j 's problem is to choose production plan y_j^* s.t. $y_j^* \in y_j(p) = \arg \max p \cdot y_j$

$$y_j^* \in y_j(p) = \arg \max_{y_j \in Y_j} p \cdot y_j \quad (\text{ystar})$$

Given $p \in \mathbb{R}^L$ and production plans in $\{y_j(p), j = 1, \dots, J\}$, consumer i 's problem is to choose x_i^* s.t.

$$\begin{aligned} x_i^* \in x_i(p) &= \arg \max_{x_i \in \mathbb{R}_+^L} u_i(x_i) \\ \text{s.t. } p \cdot x_i &\leq p \cdot w_i + \sum_{j=1}^J \theta_{ij} p \cdot y_j(p) \end{aligned} \quad (\text{xstar})$$

5.2.1 Competitive Equilibrium

Definition 5.10 (Competitive Equilibrium)

An allocation (x^*, y^*) and a price vector $p^* \in \mathbb{R}^L$ are a *competitive equilibrium* in a private ownership production economy if

(i). $x_i^* \in x_i(p^*)$ (given by (xstar)) for all agent i . That is,

$$\begin{aligned} x_i^* \in x_i(p^*) &= \arg \max_{x_i \in \mathbb{R}_+^L} u_i(x_i) \\ \text{s.t. } p^* \cdot x_i &\leq p^* \cdot w_i + \sum_{j=1}^J \theta_{ij} p^* \cdot y_j^*(p^*) \end{aligned}$$

(ii). $y_j^* \in y_j(p^*)$ (given by (ystar)) for all firm j . That is,

$$y_j^* \in y_j(p^*) = \arg \max_{y_j \in Y_j} p^* \cdot y_j$$

(iii). Market clearing: $(x^*, y^*) \in \mathcal{A}$ (given by (A)). That is

$$\sum_{i=1}^m x_i^* = \sum_{j=1}^J y_j^* + \sum_{i=1}^m w_i$$



Example 5.3 (Representative Agent Model) There is a single consumer ($m = 1$) and a single firm ($J = 1$).

For example, suppose there are $L = 2$ goods: time (label/leisure) x_l and consumption x_c .

Suppose the firm's production set is defined by a production function $f : \mathbb{R}_+ \rightarrow \mathbb{R}$, so

$$Y := \{(-y_l, y_c) \in \mathbb{R}^2 : y_l \geq 0, y_c \leq f(y_l)\}$$

The set of feasible consumption bundles is

$$\hat{Y} := \left(\underbrace{Y + \{\omega\}}_{\{y+w: y \in Y\}} \right) \cap \mathbb{R}_+^L$$

5.2.2 Pareto Optimal

Definition 5.11 (Pareto Optimal)

An allocation (x^*, y^*) in a private ownership production economy is **Pareto optimal** if there is no other allocation (x, y) s.t. $x_i \succeq_i x_i^*, \forall i$ and $x_h \succ_h x_h^*$ for some h .



5.2.3 First-Welfare Theorem (production)

Theorem 5.5 (First-Welfare Theorem)

If each consumer's preference relation is locally non-satiated, then every competitive equilibrium allocation in a private ownership production economy is Pareto optimal.



Proof 5.6

Let (x^*, y^*) be a competitive equilibrium allocation with corresponding equilibrium price vector p^* .

Suppose by the way of contradiction that (x, y) is not Pareto optimal. That is, \exists an allocation (x, y) s.t. $x_i \succeq_i x_i^*, \forall i$ and $x_h \succ_h x_h^*$ for some h . Then, by the (xstar),

$$p^* \cdot x_h > p^* \cdot w_h + \sum_{j=1}^J \theta_{hj} p^* \cdot y_j^*$$

and by local non-satiation,

$$p^* \cdot x_i \geq p^* \cdot w_i + \sum_{j=1}^J \theta_{ij} p^* \cdot y_j^*$$

Adding together

$$\begin{aligned} \sum_{i=1}^m p^* \cdot x_i &> \sum_{i=1}^m p^* \cdot w_i + \sum_{j=1}^J p^* \cdot y_j^* \\ \Rightarrow \sum_{i=1}^m p^* \cdot x_i - \sum_{i=1}^m p^* \cdot w_i &= p^* \cdot \left[\sum_{i=1}^m x_i - \sum_{i=1}^m w_i \right] > \sum_{j=1}^J p^* \cdot y_j^* \end{aligned}$$

As $\sum_{i=1}^m x_i = \sum_{j=1}^J y_j + \sum_{i=1}^m w_i$, we have $\sum_{i=1}^m x_i - \sum_{i=1}^m w_i = \sum_{j=1}^J y_j$,

$$\sum_{j=1}^J p^* \cdot y_j = p^* \cdot \left[\sum_{i=1}^m x_i - \sum_{i=1}^m w_i \right] > \sum_{j=1}^J p^* \cdot y_j^*$$

There is a contradiction, since this implies there is a firm j and $y_j \in Y_j$ s.t. $p^* \cdot y_j > p^* \cdot y_j^*$, which contradicts to the assumption that y_j^* maximizes profits for firm j at p^* ((ystar)).

5.2.4 Equilibrium with Transfers

Definition 5.12 (“Supportable” as a Price Equilibrium with Transfers)

An allocation (x^*, y^*) in a private ownership production economy is **supportable** as a **price equilibrium with transfers** if there exists a price vector $p^* \in \mathbb{R}^L$ and lump-sum budget-balancing transfers $\{T_i : i = 1, \dots, m\}$ so that $\sum_{i=1}^m T_i = 0$, such that:

1. $\forall i$,

$$\begin{aligned} x_i^* &\in \arg \max_{x \in \mathbb{R}_+^L} u_i(x_i) \\ \text{s.t. } p^* \cdot x_i &\leq p^* \cdot w_i + \sum_{j=1}^J \theta_{ij} p^* \cdot y_j^* + T_i \end{aligned}$$

2. $\forall j$,

$$y_j^* \in \arg \max_{y_j \in Y_j} p^* \cdot y_j$$

3. Feasibility:

$$\sum_{i=1}^m x_i^* = \sum_{j=1}^J y_j^* + \sum_{i=1}^m w_i$$



5.2.5 Second Welfare Theorem (production)

Production economy is a more general form of exchange economy. It is the same that not every P.O. allocation can be supported as a price signal with transfers.

Theorem 5.6 (Second Welfare Theorem (production))

If each consumer's preference relation is continuous, strongly monotone, and convex, and each firm's production set is convex, then every interior P.O. allocation in a private ownership production economy can be supported as a price equilibrium with transfers.



Proof 5.7

Let (x^*, y^*) be an interior P.O. allocation, so $x_i^* >> 0, \forall i$. The same as exchange economy, for each agent i , let $P_i := \{x_i \in \mathbb{R}_+^L : u_i(x_i) > u_i(x_i^*)\}$ and let

$$P := P_1 + \cdots + P_m$$

$$= \left\{ z \in \mathbb{R}_+^L : z = \sum_{i=1}^m x_i \text{ for some } (x_1, \dots, x_m) \in \mathbb{R}_+^{L \times m} \text{ s.t. } u_i(x_i) > u_i(x_i^*), \forall i \right\}$$

Then P is non-empty and convex.

In production side, let

$$Y := \sum_{j=1}^J Y_j = \left\{ y \in \mathbb{R}^L : y = \sum_{j=1}^J y_j \text{ for some } (y_1, \dots, y_J) \text{ s.t. } y_j \in Y_j, \forall j \right\}$$

Then $Y + \{w\}$ is non-empty and convex.

Claim 5.2

$P \cap (Y + \{w\}) = \emptyset$. This follows from the assumption that (x^*, y^*) is P.O. (There is no allocation gives higher utilities while satisfies constraints).



By the Separating Hyperplane Theorem 5.3, $\exists p \in \mathbb{R}^L, p \neq 0, \text{s.t.}$

$$p \cdot z \geq p \cdot (y + w), \forall z \in P \text{ and } y \in Y \quad (\text{p:SHT})$$

For each i , set

$$T_i := p \cdot x_i^* - p \cdot w_i - \sum_{j=1}^J \theta_{ij} p \cdot y_j^*$$

Then,

$$p \cdot x_i^* = p \cdot w_i + \sum_{j=1}^J \theta_{ij} p \cdot y_j^* + T_i$$

and

$$\begin{aligned} \sum_{i=1}^m T_i &= \sum_{i=1}^m \left(p \cdot x_i^* - p \cdot w_i - \sum_{j=1}^J \theta_{ij} p \cdot y_j^* \right) \\ &= p \cdot \left(\sum_{i=1}^m x_i^* - \sum_{i=1}^m w_i - \sum_{j=1}^J y_j^* \right) \\ &= p \cdot 0 = 0 \end{aligned}$$

So, $\{T_i : i = 1, \dots, m\}$ are budget balancing.

Claim 5.3

$$p \cdot z \geq p \cdot \left(\sum_{i=1}^m x_i^* \right) = p \cdot \left(\sum_{j=1}^J y_j^* + \sum_{i=1}^m w_i \right) \geq p \cdot (y + w), \forall z \in P \text{ and } y \in Y$$



To prove this, first note the feasibility, $\sum_{j=1}^J y_j^* + \sum_{i=1}^m w_i \in Y + \{w\}$ and $\sum_{i=1}^m x_i^* = \sum_{j=1}^J y_j^* + \sum_{i=1}^m w_i$. By the p:SHT,

$$p \cdot z \geq p \cdot \left(\sum_{j=1}^J y_j^* + \sum_{i=1}^m w_i \right) = p \cdot \left(\sum_{i=1}^m x_i^* \right), \forall z \in P$$

Now using the strong monotonicity, for each i , we can choose a sequence $\{x_i^n\} \subseteq P_i$ s.t. $x_i^n \rightarrow x_i^*$ (e.g.

$x_i^n = (1 + \frac{1}{n}) x_i^*$). Let $z^n = \sum_{i=1}^m x_i^n$ for all n . Then, $z^n \in P$ for all n and $z^n \rightarrow \sum_{i=1}^m x_i^*$.

Let $y \in Y$ be arbitrary. By the **p:SHT**,

$$\begin{aligned} p \cdot z^n &\geq p \cdot (y + w), \forall n \\ \Rightarrow \lim_{n \rightarrow \infty} p \cdot z^n &= p \cdot \left(\sum_{i=1}^m x_i^* \right) \geq p \cdot (y + w) \\ \Rightarrow p \cdot \left(\sum_{j=1}^J y_j^* + \sum_{i=1}^m w_i \right) &= p \cdot \left(\sum_{i=1}^m x_i^* \right) \geq p \cdot (y + w) \end{aligned}$$

That is, claim 5.3 is proved.

Claim 5.4

$$\forall j: p \cdot y_j^* \geq p \cdot y_j, \forall y_j \in Y_j$$



To show this, we fix k and $y_k \in Y_k$, such that $y_k + \sum_{j \neq k} y_j^* \in Y$. By claim 5.3,

$$\begin{aligned} p \cdot \left(\sum_{j=1}^J y_j^* + w \right) &\geq p \cdot \left(y_k + \sum_{j \neq k} y_j^* + w \right) \\ \Rightarrow p \cdot y_k^* &\geq p \cdot y_k \end{aligned}$$

Hence, claim 5.4 is proved.

Claim 5.5

$$\forall i: u_i(x_i) > u_i(x_i^*) \Rightarrow p \cdot x_i > p \cdot x_i^*.$$



Note that in the proof for the SWT in exchange economy, it is sufficient to show $\forall i: u_i(x_i) > u_i(x_i^*) \Rightarrow p \cdot x_i \geq p \cdot x_i^*$. Fix h and let $x_h \in P_h$. So, $u_h(x_h) > u_h(x_h^*)$. By the continuity and strong monotonicity of preference, we have $x_h + \sum_{i \neq h} x_i^* \in P$ (we can increase each x_i^* a little and reduce x_h). Hence, by 5.3,

$$\begin{aligned} p \cdot (x_h + \sum_{i \neq h} x_i^*) &\geq p \cdot \left(\sum_{i=1}^m x_i^* \right) \\ \Rightarrow p \cdot x_h &\geq p \cdot x_h^* \end{aligned}$$

Hence, claim 5.5 is proved.

All in all, SWT is proved.

5.3 Existence of Competitive Equilibrium

5.3.1 Excess Demand in Exchange Economies

Assumption Suppose

- each consumer's preference relation is continuous, strongly monotone, and strictly convex, and

- $\sum_i w_i > 0$.

Based on this assumption 5.3.1, we have

- Each agent's demand function $x_i : \mathbb{R}_{++}^L \rightarrow \mathbb{R}_+^L$ is well-defined, continuous, homogeneous of degree 0, and satisfies Walras' Law (for individual).
- Excess demand function $Z : \mathbb{R}_{++} \rightarrow \mathbb{R}^L$ given by

$$Z(p) = \sum_{i=1}^m x_i(p) - \sum_{i=1}^m w_i$$

is

Definition 5.13 (Condition (1) to (4))

Given the Assumption 5.3.1, following conditions are satisfied:

- (1). Continuous;
- (2). Homogeneous of degree 0;
- (3). Satisfies Walras' Law: $p \cdot Z(p) = 0, \forall p$;
- (4). Bounded below: $\exists s > 0$ s.t. $Z_l(p) \geq -s, \forall p, \forall l = 1, \dots, L$.



5.3.2 Excess Demand in Production Economies

We use the same assumption 5.3.1 for consumers' preferences, and we add a assumption on the production side.

Assumption Suppose each firm's production set Y_j is

- closed,
- strictly convex ($y, y' \in Y_j \Rightarrow \alpha y + (1 - \alpha)y' \in \text{int}Y_j, \forall \alpha \in (0, 1)$),
- bounded above ($\exists \bar{y}_j \in \mathbb{R}^L$ s.t. $y \leq \bar{y}_j, \forall y \in Y_j$), and
- $0 \in Y_j$.

Based on this assumption 5.3.2, we have

- Each firm's supply function $y_j : \mathbb{R}_{++}^L \rightarrow \mathbb{R}^L$ is well-defined, continuous, and homogeneous of degree 0.

Based on assumption 5.3.1 and 5.3.2, we have

- Excess demand function $Z : \mathbb{R}_{++} \rightarrow \mathbb{R}^L$ given by

$$Z(p) = \sum_{i=1}^m x_i(p) - \sum_{j=1}^J y_j(p) - \sum_{i=1}^m w_i$$

is

Definition 5.14 (Condition (1) to (4))

Given the Assumption 5.3.2, following conditions are satisfied:

- (1). Continuous;
- (2). Homogeneous of degree 0;

- (3). Satisfies Walras' Law: $p \cdot Z(p) = 0, \forall p;$
- (4). Bounded below: $\exists s > 0$ s.t. $Z_l(p) \geq -s, \forall p, \forall l = 1, \dots, L.$



- If $Z(p^*) = 0$, then p^* is a competitive equilibrium price vector, with corresponding equilibrium allocation $(x_1(p^*), \dots, x_m(p^*), y_1(p^*), \dots, y_J(p^*))$.

5.3.3 Boundary Condition

Since Z is homogeneous of degree 0, we can normalize prices, set

$$\Delta := \left\{ p \in \mathbb{R}_+^L : \sum_{l=1}^L p_l = 1 \right\}$$

We give other notations:

$$\partial\Delta := \{p \in \Delta : p_l = 0 \text{ for some } l\}$$

$$\text{int}\Delta := \{p \in \Delta : p_l > 0, \forall l\} = \Delta \cap \mathbb{R}_{++}^L$$

Consider an exchange economy. Let $p \in \partial\Delta$. Let $w_i \gg 0$. If \succeq_i is strongly monotone on \mathbb{R}_{++}^L , then demand of agent i is undefined at p_i (infinity for the zero price good).

So, we add a condition for excess demand Z :

Definition 5.15 (Condition (5))

- (5). If $p^n \in \text{int}\Delta, \forall n$ and $p^n \rightarrow p$, where $p \in \partial\Delta$, then $\max_l\{Z_l(p^n)\} \rightarrow +\infty$.



Note

- Condition (5) holds in an exchange economy with assumption 5.3.1;
- Condition (5) holds in a production economy with assumption 5.3.1 and 5.3.2;
- Condition (5) is not true in general, and the condition (5) does not imply $p_l^n \rightarrow 0 \Rightarrow Z_l(p^n) \rightarrow +\infty$ (relative prices matter!)
- By Walras' Law and lower bound on Z , then the converse holds:

$$Z_l(p^n) \rightarrow +\infty \Rightarrow p_l^n \rightarrow 0$$

So, if condition (1) to (5) hold and $p^n \rightarrow p$ where $p^n \gg 0$ and $p_l > 0$, then $\{Z_l(p^n)\}$ is bounded.

5.3.4 Existence of Competitive Equilibrium

Theorem 5.7 (Condition (1) to (5) $\Rightarrow \exists$ a competitive equilibrium)

Let $Z : \mathbb{R}_{++}^L \rightarrow \mathbb{R}^L$ be a function s.t. condition (1) to (5) are satisfied, that is

- (1). Continuous;
- (2). Homogeneous of degree 0;
- (3). Satisfies Walras' Law: $p \cdot Z(p) = 0, \forall p$;
- (4). Bounded below: $\exists s > 0$ s.t. $Z_l(p) \geq -s, \forall p, \forall l = 1, \dots, L$.
- (5). If $p^n \in \text{int}\Delta, \forall n$ and $p^n \rightarrow p$, where $p \in \partial\Delta$, then $\max_l \{Z_l(p^n)\} \rightarrow +\infty$.

 **Note** (Fulfilled when Assumption 5.3.1 and 5.3.2 are satisfied).

Then, $\exists \bar{p} \in \mathbb{R}_{++}^L$ s.t. $Z(\bar{p}) = 0$. 

Proof 5.8

First, restrict attention to $p \in \Delta = \left\{ p \in \mathbb{R}_+^L : \sum_{l=1}^L p_l = 1 \right\}$.

- o For $p \in \text{int}\Delta$, define a subset of good $\Lambda(p) \subseteq \{1, \dots, L\}$ by

$$\Lambda(p) := \left\{ l \in \{1, \dots, L\} : Z_l(p) = \max_k Z_k(p) \right\}$$

- o For $p \in \Delta \setminus \text{int}\Delta = \partial\Delta$, let

$$\Lambda(p) := \{l \in \{1, \dots, L\} : p_l = 0\}$$

Then, note $\Lambda(p) \neq \emptyset, \forall p \in \Delta$.

Define the correspondence $\varphi : \Delta \rightarrow 2^\Delta$ that maps a price vector p to a set of prices by

$$\varphi(p) := \{q \in \Delta : q_l = 0, \forall l \notin \Lambda(p)\}$$

As $\Lambda(p) \neq \emptyset, \forall p \in \Delta$, we have $\varphi(p) \neq \emptyset$ for all p and

$$\varphi(p) := \begin{cases} \{q \in \Delta : q \in \arg \max_{\tilde{q} \in \Delta} \tilde{q} \cdot Z(p)\} & \text{if } p \in \text{int}\Delta \\ \{q \in \Delta : q \cdot p = 0\} & \text{if } p \in \partial\Delta \end{cases}$$

 **Note**

1. $\varphi(p) \subseteq \partial\Delta \Leftrightarrow \Lambda(p) \neq \{1, \dots, L\}$
2. $p \in \partial\Delta \Rightarrow p$ is not a fixed point of φ .
3. p is a fixed point of $\varphi \Leftrightarrow p \in \text{int}\Delta$ and $\Lambda(p) = \{1, \dots, L\} \Leftrightarrow p \in \text{int}\Delta$ and $\exists m \in \mathbb{R}$ s.t. $Z_l(p) = m, \forall l = \{1, \dots, L\}$.
4. p is a fixed point of $\varphi \Leftrightarrow p \in \text{int}\Delta$ and $Z(p) = 0$. (By Walras' Law: $0 = p \cdot Z(p) = m \sum_l p_l = m$.)
5. $Z(p) \neq 0 \Rightarrow \Lambda(p) \neq \{1, \dots, L\} \Rightarrow \varphi(p) \subseteq \partial\Delta$.

Now it suffices to show φ has a fixed point. Note that $\forall p \in \Delta$, $\varphi(p)$ is non-empty, convex, and compact, and Δ is non-empty, convex, and compact.

Claim 5.6

φ has closed graph.



Let $p^n \rightarrow p \in \Delta$ and $q^n \rightarrow q \in \Delta$ where $(p^n, q^n) \in \text{graph } \varphi \forall n$. We want to show $(p, q) \in \text{graph } \varphi$ (i.e., $q \in \varphi(p)$):

- o Case 1: Suppose $p \in \text{int}\Delta$. Since $p >> 0$, assume without losing generality, $p^n >> 0 \forall n$. Suppose $l \notin \Lambda(p)$, we must show $q_l = 0$. Since $p >> 0$, $l \notin \Lambda(p) \Rightarrow Z_l(p) < \max_k Z_k(p)$. Since Z is continuous, $\exists N$, such that $\forall n \geq N$, $Z_l(p^n) < \max_k Z_k(p^n) \Rightarrow l \notin \Lambda(p^n), \forall n \geq N \Rightarrow q_l^n = 0, \forall n \geq N \Rightarrow q_l = \lim_{n \rightarrow \infty} q_l^n = 0$. So, $q \in \varphi(p)$.
- o Case 2: Suppose $p \in \partial\Delta$. Without losing generality, we write $p = (0, \dots, 0, p_{r+1}, \dots, p_L)$, where $p_l > 0$ for all $l = r+1, \dots, L$. So, $\Lambda(p) = \{1, \dots, r\}$ and $\varphi(p) = \{\tilde{q} \in \Delta : \tilde{q}_l = 0, l = r+1, \dots, L\}$.
 - Case 2A: Suppose $\{p^n\}$ has a subsequence in $\text{int}\Delta$. Without losing generality, let $\{p^n\}$ denote this subsequence. Since $p^n \rightarrow p \in \partial\Delta$, $\max_k Z_k(p^n) \rightarrow +\infty$. Also, by Walras' Law and lower bound in Z , $\{Z_l(p^n)\}$ is bounded for $l = r+1, \dots, L$. Since $p^n \in \text{int}\Delta, \forall n, \exists N_2$ s.t. $\forall n \geq N_2, \Lambda(p^n) \subseteq \{1, \dots, r\}$. Since $q^n \in \varphi(p^n), \forall n$ and $q_l^n = 0, \forall l = r+1, \dots, L, \forall n \geq N_2$, we have $q_l = \lim_n q_l^n = 0$. Hence, $q \in \varphi(p)$.
 - Case 2B: No subsequence of $\{p^n\}$ lies in $\text{int}\Delta$. Without losing generality, take $\{p^n\} \subseteq \partial\Delta$. Now, because $p_l > 0$ for $l = r+1, \dots, L$, $\exists N_3$ s.t. $\forall n \geq N_3, p_l^n > 0$ for $l = r+1, \dots, L$. Then, $\Lambda(p^n) \subseteq \{1, \dots, r\}, \forall n \geq N_3$. By the same argument above (in Case 2A), we have $q^n \in \varphi(p^n), \forall n \Rightarrow q_l^n = 0, \forall l = r+1, \dots, L, \forall n \geq N_3 \Rightarrow q_l = \lim_{n \rightarrow \infty} q_l^n = 0, \forall l = r+1, \dots, L$. Hence, $q \in \varphi(p)$.

All in all, φ has closed graph. By Kakutani's Fixed Point Theorem, φ has a fixed point. By above augment, $\bar{p} \in \text{int}\Delta$ and $Z(\bar{p}) = 0$.

Corollary 5.2

If an exchange economy satisfies assumption 5.3.1, then it has a competitive equilibrium. If a private ownership production economy satisfies assumption 5.3.1 and 5.3.2, then it also has a competitive equilibrium.



5.4 Uniqueness of Equilibrium

When is the equilibrium unique?

One condition:

Definition 5.16 (Strong Weak Axiom)

The function $Z : \mathbb{R}_{++}^L \rightarrow \mathbb{R}^L$ satisfies the strong weak axiom if for any $\bar{p} \in \mathbb{R}_{++}^L$ s.t. $Z(\bar{p}) = 0$ and any $p \in \mathbb{R}_{++}^L$ s.t. $p \neq \alpha\bar{p}, \forall \alpha > 0$,

$$\bar{p} \cdot Z(p) > 0$$



Theorem 5.8 (Uniqueness of Equilibrium)

If $Z : \mathbb{R}_{++}^L \rightarrow \mathbb{R}^L$ satisfies condition (1)-(5) and strong weak axiom, then there is a unique $p^* \in \text{int}\Delta$ s.t. $Z(p^*) = 0$.



Proof 5.9

Since Z satisfies condition (1)-(5), $\exists p^* \in \text{int}\Delta$ s.t. $Z(p^*) = 0$. By the strong weak axiom, if $p \in \text{int}\Delta$ and $p \neq p^*$, then $p^* \cdot Z(p) > 0 \Rightarrow Z(p) \neq 0$. So, there is a unique $p^* \in \text{int}\Delta$ s.t. $Z(p^*) = 0$.

Example 5.4

1. In an exchange economy with a representative consumer with strictly quasi-concave, strongly monotone, and C^1 (first-order continuously differentiable) utility function and $\omega >> 0$, the excess demand function satisfies the strong weak axiom.
2. If Z satisfies gross substitutes (for each l , Z_l is increasing in $p_k, \forall k \neq l$), then Z satisfies the strong weak axiom.

5.5 Market Demand and Observable Implications

Given an outcome, can we say it is obtained from an economy?

Restrict to exchange economies for simplicity.

Theorem 5.9 (Sonnenschein-Mantel-Debreu (SMD) Theorem)

Let $Z : \mathbb{R}_{++}^L \rightarrow \mathbb{R}^L$ be a function that is continuous and satisfies Walras' Law ($p \cdot Z(p) = 0, \forall p$). Then $\forall \epsilon > 0$ there is an exchange economy with L consumers having continuous, strictly convex, strongly monotone preferences, and endowments $\{w_i : i = 1, \dots, L\} \subseteq \mathbb{R}_+^L$, s.t., the excess demand function for this economy is equivalent to Z on $\Delta^\epsilon = \{p \in \Delta : p_l \geq \epsilon, \forall l\}$.



Theorem 5.10 (Mas-Colell Theorem)

Let $E \subseteq \text{int}\Delta$ be compact. Then there exists an exchange economy with L consumers for which E is the set of competitive equilibrium prices.



Example 5.5 Suppose $L = 2$, $Z : \mathbb{R}_{++}^2 \rightarrow \mathbb{R}^2$ satisfies condition (1)-(5). Choose good 2 as numéraire, and set $p = (p_1, 1)$. Then,

1. If p_1 is close to 0, then $Z_1(p) > 0$ by the boundary condition (5) and lower bound condition (4).
2. If p_1 is large, then $Z_2(p) > 0$ by the boundary condition (5) and lower bound condition (4). So, $Z_1(p) = -\frac{1}{p_1}Z_2(p) < 0$ by Walras' Law (condition (3)).
3. $\exists p_1^* \text{ s.t. } Z_1(p^*) = 0$ by continuous Z (condition (1)), and p^* is an equilibrium price vector by Walras' Law (condition (3)).

5.6 Comparative Statics and Local Uniqueness

Restrict to exchange economies for simplicity.

Let $\vec{w} = (w_1, \dots, w_m) \in \mathbb{R}_+^{L \times m}$ denote profile of initial endowments.

Let $\mathcal{E}(\vec{w})$ denote the economy with fixed preference relations $\{\succeq_i : i = 1, \dots, m\}$ and endowment profile \vec{w} .

Let $Z : \mathbb{R}_{++}^L \times \mathbb{R}_{++}^{L \times m} \rightarrow \mathbb{R}^L$ denote excess demand as a function of (p, \vec{w}) , so

$$Z(p, \vec{w}) := \sum_{i=1}^m x_i(p, p \cdot w_i) - \sum_{i=1}^m w_i$$

Let

$$Z_{-L}(p, \vec{w}) := (Z_1(p, \vec{w}), \dots, Z_{L-1}(p, \vec{w}))$$

Normalize $p_L = 1$. Given \vec{w} , equilibrium in $\mathcal{E}(\vec{w})$ corresponds to p s.t. $Z_{-L}(p, \vec{w}) = 0$.

Assume $Z(\cdot, \vec{w})$ is C^1 and satisfies conditions (1)-(5), $\forall \vec{w}$.

Definition 5.17 (Regular Equilibrium)

Given \vec{w} , an equilibrium price vector p is a **regular equilibrium** if $D_p Z_{-L}(p, \vec{w})$ is non-singular (has full rank $L - 1$).


Definition 5.18 (Regular/Critical Economy)

If every equilibrium in the economy $\mathcal{E}(\vec{w})$ is regular, then $\mathcal{E}(\vec{w})$ is a **regular economy**. An economy that is not regular is a **critical economy**.



Proposition 5.3

1. Regular equilibria are locally unique.

$(\exists \text{ open set } V \text{ with } p^* \in V \text{ s.t. if } p \in V, \text{ then } Z_{-L}(p, \vec{w}) = 0 \Leftrightarrow p = p^*)$

2. A regular economy has finitely many equilibria.

3. In a regular economy, local equilibrium comparative statics are determinate.

4. $\mathcal{E}(\vec{w})$ is a regular economy if 0 is a regular value of $Z(\cdot, \vec{w})$. ($D_p Z(0, \vec{w})$ is non-singular).


Theorem 5.11 (C^1 Demand Functions \Rightarrow Regular Economy)

Suppose each agent $i = 1, \dots, m$ has a C^1 demand function: $x_i : \mathbb{R}_{++}^L \times \mathbb{R}_{++}^L \rightarrow \mathbb{R}_+^L$. Then almost all economy are regular. That is,

$$\left\{ \vec{w} \in \mathbb{R}_{++}^{L \times m} : \mathcal{E}(\vec{w}) \text{ is a critical economy} \right\}$$

has Lebesgue measure zero in $\mathbb{R}^{L \times m}$.


Proof 5.10

To prove DZ_{-L} has full rank, it is sufficient to show the sub-matrix has rank $L - 1$.

Denote the derivative matrix of $DZ_{-L}(p, \vec{w})$ by A . The (j, k) item of the matrix is

$$\frac{\partial x_{jk}}{\partial p_k} + \frac{\partial x_{jk}}{\partial w} \omega_k$$

Compute the derivatives of Z_{-L} with respect to the initial endowment of consumer 1 (ω_1). Denote the derivative matrix by A . The (j, k) item of the matrix is

$$A_{jk} = \begin{cases} \frac{\partial x_{1j}}{\partial w}(p, p \cdot \omega_1)p_k - 1, & k = j \\ \frac{\partial x_{1j}}{\partial w}(p, p \cdot \omega_1)p_k, & k \neq j \end{cases}, j = 1, \dots, L-1, k = 1, \dots, L$$

Minus $k = 1, \dots, L-1$ column by $\frac{p_k}{p_L}$ times L^{th} column. We can get

$$\begin{bmatrix} & & A_{1L} \\ -I_{L-1 \times L-1} & & \vdots \\ & & A_{L-1,L} \end{bmatrix}$$

which has rank $L - 1$. Hence, A has rank $L - 1$.

Hence, 0 is a regular value of Z_{-L} . So, the result follows from the Transversality Theorem.

5.7 General Equilibrium with Uncertainty

Set-up:

1. There are L (physical) goods.
2. There are two time periods, $t = 0, t = 1$.

3. At date $t = 0$, the state of nature is determinate.
4. At date $t = 1$, there are S ($s \in \{1, \dots, S\}$) possible states of nature, and all uncertainty resolves at $t = 1$.
5. Hence, the commodity space is $\mathbb{R}^{L \times (S+1)}$.
6. Each consumer has preference relation \succeq_i over $\mathbb{R}_+^{L \times (S+1)}$ represented by utility function u_i and initial endowment under different states $w_i = (w_{i0}, w_{i1}, \dots, w_{iS}) \in \mathbb{R}_+^{L \times (S+1)}$, where w_{i0} is the endowment at $t = 0$ and w_{is} is the endowment at $t = 1$ with state $s \in \{1, \dots, S\}$.

5.7.1 Basic Settings: (Complete) Contingent Commodities, Arrow-Debreu Equilibrium

Definition 5.19 (Contingent Commodity)

A unit of the **contingent commodity** (or **Arrow security**) l_s is a claim to receive a unit of good l if and only if state s occurs at date $t = 1$



Suppose at date $t = 0$, there are markets for “date $t = 0$ consumption” and “a complete set of Arrow securities”.

Given price vector $p \in \mathbb{R}^{L \times (S+1)}$, agent i ’s budget set is

$$B_i(p) = \left\{ x \in \mathbb{R}_+^{L \times (S+1)} : p \cdot x \leq p \cdot w_i \right\}$$

Definition 5.20 (Arrow-Debreu Equilibrium)

A competitive equilibrium in this model is an **Arrow-Debreu equilibrium**.



5.7.2 General: Asset Markets and Radner Equilibrium

There is a market of spots. Assets are used for the trading across different stages.



Note Simplify: assume all assets payoff in units of good 1.

An asset is defined by its return vector $r \in \mathbb{R}^S$.

Example 5.6

1. Arrow securities $l_{\bar{s}}$ (for good 1): $r_s = \begin{cases} 1, & s = \bar{s} \\ 0, & s \neq \bar{s} \end{cases}$
2. Riskless bond: $r_s = 1, \forall s$.
3. Another asset: $r_s = \begin{cases} 1, & s \text{ is even} \\ -1, & s \text{ is odd} \end{cases}$. Hence, $r = (-1, 1, \dots, (-1)^{|S|})$.

Suppose there are K assets traded at date $t = 0$, indexed by $k = 1, \dots, K$. Summarize their payoffs in an $S \times K$

matrix.

$$R := \begin{bmatrix} r_{11} & \cdots & r_{K1} \\ \vdots & \vdots & \vdots \\ r_{1S} & \cdots & r_{KS} \end{bmatrix}$$

where r_{ks} is the asset k 's return in state s .

Assume assets are in zero total supply. Let $z_i \in \mathbb{R}^k$ denote agent i 's portfolio. So,

$$z_{ik} = \# \text{ units of asset } k \text{ bought/sold by agent } i$$

Let $q = (q_1, \dots, q_K) \in \mathbb{R}^K$ denote the vector of asset prices.

Given an asset payoff structure R , the payoff from the portfolio z_i is $Rz_i \in \mathbb{R}^S$.

Let $p_s \in \mathbb{R}^L$ denote price vector expected at date 0 to hold in the spot market at date $t = 1$ if state s occurs.

Definition 5.21 (Radner Equilibrium)

A consumption allocation (x_1^*, \dots, x_m^*) , portfolio profile (z_1^*, \dots, z_m^*) , spot price vectors $(p_0^*, p_1^*, \dots, p_S^*)$, and asset price vector q^* are a **Radner equilibrium** if

- For every agent i : (x_i^*, z_i^*) solves

$$\max_{(x_i, z_i) \in \mathbb{R}_+^{L \times (S+1)} \times \mathbb{R}^K} u_i(x_i) \\ \text{s.t. } \left. \begin{array}{l} p_0^* \cdot x_{i0} + q^* \cdot z_i \leq p_0^* \cdot w_{i0} \\ p_s^* \cdot x_{is} \leq p_s^* \cdot w_{is} + p_{1s}^* (Rz_i)_s, \forall s \end{array} \right\} \triangleq B_i(p^*, q^*, R)$$

(reminds that we assume all assets are payoff in good 1).

- $\sum_{i=1}^m z_i^* = 0$, $\sum_{i=1}^m x_i^* = \sum_{i=1}^m w_i$.



Note Normalize: $p_{1s} = 1, \forall s = 1, \dots, S$.

Then consumer i 's budget set is

$$B_i(p, q, R) := \{x_i \in \mathbb{R}_+^{L \times (S+1)} : \exists z \in \mathbb{R}^K \text{ s.t. } p_0 \cdot x_{i0} + q \cdot z \leq p_0 \cdot w_{i0} \text{ and } p_s \cdot (x_{is} - w_{is}) \leq (Rz)_s, \forall s\}$$

Definition 5.22 (Complete Asset Structure R)

The asset structure with return matrix R is **complete** if $\text{rank } R = S$. And the asset structure is **incomplete** if $\text{rank } R < S$.



Theorem 5.12 (Arrow-Debreu Equilibrium \Leftrightarrow Radner Equilibrium)

Suppose preferences are strongly monotone and the asset structure is complete.

- If (x^*, p^*) is an Arrow-Debreu equilibrium, then there exists a portfolio profile z^* and asset price vector q^* such that (x^*, p^*, q^*, z^*) is a Radner equilibrium.
- If (x^*, p^*, q^*, z^*) is a Radner equilibrium, then there exists a vector $\mu \in \mathbb{R}_{++}^S$ (i.e., common beliefs)

such that $(x^*, (p_0^*, \mu_1 p_1^*, \dots, \mu_S p_S^*))$ is an Arrow-Debreu equilibrium.



Definition 5.23 (Arbitrage-free)

For asset structure with return matrix R , the asset price vector $q \in \mathbb{R}^K$ is **arbitrage-free** if $\nexists z \in \mathbb{R}^K$ s.t.

$q \cdot z \leq 0$ and $Rz \geq 0$ with strict inequality for one.



Proposition 5.4 (Strongly Monotone \Rightarrow Arbitrage-free)

If preferences are strongly monotone, then in any Radner equilibrium, asset prices must be arbitrage-free.



We can also infer the components of assets from arbitrage-free prices.

Theorem 5.13 (Reconstruct Assets from Arbitrage-free Prices)

Consider an asset structure with return matrix R . If

- $q \in \mathbb{R}^K$ is arbitrage-free, or
- $q \in \mathbb{R}^K$ is a Radner equilibrium asset prices with $r_k \geq 0, r_k \neq 0, \forall k$.

then there exists $\mu \in \mathbb{R}_{++}^S$ such that $q = \mu^T R$, that is, s.t.

$$q_k = \sum_s \mu_s r_{ks}, \forall k$$



Chapter 6 Game Theory

Based on

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- Mas-Colell, Whinston, and Green, Microeconomic Theory, Oxford University Press (1995).
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6.1 Basic Game Theory

6.1.1 Action and Domination Theorem

Let A be the finite set of possible actions and Ω be the finite set of possible states. A function can map the action and state to a value, $u(a, \omega)$. It can be represented by $\vec{u}(a) = \{u(a, \omega)\}_{\omega \in \Omega}$. It is common in game theory to assume the utility function is given or known.

A **mixed action** is a probability distribution over A , $\sigma \in \Delta(A)$.

A **belief** of the agent is a probability distribution over Ω , $\mu \in \Delta(\Omega)$.

Definition 6.1 (Optimal and Justifiable Mixed Action)

A mixed action $\sigma \in \Delta(A)$ is **optimal** given $\mu \in \Delta(\Omega)$ if

$$\mathbb{E}_\mu u(\sigma, \tilde{\omega}) \geq \mathbb{E}_\mu u(\sigma', \tilde{\omega}), \forall \sigma' \in \Delta(A)$$

A mixed action $\sigma \in \Delta(A)$ is **justifiable** if it is optimal for some belief $\mu \in \Delta(\Omega)$.



Definition 6.2 (Dominant and Dominated Action)

A mixed action $\sigma \in \Delta(A)$ is **dominant** if

$$u(\sigma, \omega) > u(\sigma', \omega), \forall \omega \in \Omega, \sigma' \in \Delta(A), \sigma \neq \sigma'$$

A mixed action $\sigma \in \Delta(A)$ is **dominated** if

$$u(\sigma, \omega) < u(\sigma', \omega), \forall \omega \in \Omega, \text{ and for some } \sigma' \in \Delta(A)$$

In this case we say σ' dominates σ .



Theorem 6.1 (Domination Theorem: Justifiable = Not Dominated)

A mixed action is justifiable if and only if it is not dominated.

**Proof 6.1**

\Rightarrow is easily proved by the definition. We focus on proving \Leftarrow :

Let $\mathcal{U} = \{\vec{u}(\sigma) : \sigma \in \Delta(A)\}$ and σ^* be an undominated mixed action. Then, we have $\mathcal{U} \cap (\{\vec{u}(\sigma^*)\} + \mathbb{R}_{++}^\Omega) = \emptyset$. Because \mathcal{U} and $\{\vec{u}(\sigma^*)\} + \mathbb{R}_{++}^\Omega$ are disjoint, convex, and nonempty, we can use the Separating Hyperplane Theorem 5.3: $\exists p \in \mathbb{R}^\Omega, p \neq 0$ such that $p \cdot a \leq p \cdot b, \forall a \in \mathcal{U}, b \in (\{\vec{u}(\sigma^*)\} + \mathbb{R}_{++}^\Omega)$.

Claim 6.1

$$p \cdot \vec{u}(\sigma) \leq p \cdot \vec{u}(\sigma^*), \forall \sigma \in \Delta(A).$$

**Proof 6.2**

For any positive number m , $\vec{u}(\sigma^*) + (\frac{1}{m}, \dots, \frac{1}{m}) \in \{\vec{u}(\sigma')\} + \mathbb{R}_{++}^\Omega$. So, for any $\sigma \in \Delta(A)$, $p \cdot \vec{u}(\sigma) \leq p \cdot (\vec{u}(\sigma^*) + (\frac{1}{m}, \dots, \frac{1}{m}))$. By taking limit, $p \cdot \vec{u}(\sigma^*) = \lim_{m \rightarrow \infty} p \cdot (\vec{u}(\sigma^*) + (\frac{1}{m}, \dots, \frac{1}{m})) \geq p \cdot \vec{u}(\sigma)$.

Claim 6.2

$$p > 0.$$

**Proof 6.3**

Prove by the contradiction. Suppose $p_\omega < 0$ for some $\omega \in \Omega$. Let $z = (\epsilon, \dots, \epsilon) + M\mathbb{1}_\omega$, $M > 0, \epsilon > 0$. So, $\vec{u}(\sigma^*) + z \in (\{\vec{u}(\sigma^*)\} + \mathbb{R}_{++}^\Omega)$. We have $p \cdot \vec{u}(\sigma^*) \leq p \cdot (\vec{u}(\sigma^*) + z)$ by the result of SHT. There is a contradiction since $p_\omega < 0$. So, we have $p \geq 0$. Because $p \neq 0$, $p > 0$ is proved.

Finally, we normalize p to $\mu = \frac{1}{\sum_\omega p_\omega} p$. Then, σ^* is optimal for the belief μ , which means σ^* is justifiable.

6.1.2 Extensive Game

Definition 6.3 (History)

The sequences of actions are called **histories**. $h' = (\underbrace{a_1, \dots, a_n}_{h:\text{prefix of } h'}, a_{n+1}, \dots) \in H$. We call h' is the **continuation** of h . h is a **terminal** of H if there is no continuation of h in H . ($\emptyset \in H$.)

**Definition 6.4 (Extensive form Perfect Information Game)**

An extensive form game with perfect information is defined as $G = \{N, A, H, Z, P, O, o, \succ_{n \in N}\}$, where N is the set of players, A is the set of actions, H is the set of all histories, Z is the set of all histories that are terminals, $P : H/Z \rightarrow N$ is a mapping from a non-terminal histories to a player (who moves after a

non-terminal history), O is the set of outcomes, and o is a function from Z to O .

A PIG is finite horizon if there is a bound on the length of its histories.



We denote $A(h)$ as the actions available to player $P(h)$ after a history h .

Let $H_i = \{h \in H/Z : i = P(h)\}$ be the set of histories that player i moves after.

Definition 6.5 (Strategy)

A **strategy** is defined as a function $s_i : H_i \rightarrow A$ for which $s_i(h) \in A(h), \forall h \in H_i$. Let S_i be the set of all strategies available to the player i . A **strategy profile** is a collection of strategy $s = (s_i)_{i \in N}$.



Definition 6.6 (Subgame)

A **subgame** of a PIG $G = \{N, A, H, Z, P, O, o, \succ_{n \in N}\}$ is a game (a PIG) that starts after a given finite history $h \in H$. Formally, the subgame $G(h)$ associated with $h = (h_1, \dots, h_n) \in H$ is $G(h) = \{N, A, H_h, Z, P_h, O, o_h, \succ_{n \in N}\}$, where

$$H_h = \{(a_1, a_2, \dots) : (h_1, \dots, h_n, a_1, a_2, \dots) \in H\}$$

$$o_h(h') = o(hh'), P_h(h') = P(hh')$$

A strategy s of G defines a strategy s_h of $G(h)$ by $s_h(h') = s(hh')$.



Definition 6.7 (Subgame Perfect Equilibrium (SPNE))

A **subgame perfect equilibrium (SPNE)** of G is a strategy profile s^* such that for every subgame $G(h)$ it holds that $h' \mapsto s_i^*(hh')$ is an optimal strategy in $G(h)$, given beliefs that the rest of the players behave according to s_{-i}^* (or its restriction to $G(h)$).



Definition 6.8 (Profitable Deviation)

Let s be a strategy profile. We say that s'_i is a **profitable deviation** from s for player i at history h if s'_i is a strategy for G such that

$$o_h(s'_i, s_{-i}) \succ_i o_h(s)$$



Note that a strategy profile has no profitable deviations iff it's a SPNE.

Theorem 6.2 (The one-deviation principle)

Let $G = (N, A, H, O, o, P, \{\preceq_i\}_{i \in N})$ be a finite horizon, extensive form game with perfect information.

Let s be a strategy profile that is not a subgame perfect equilibrium. There exists some history h and a profitable deviation \bar{s}_i for player $i = P(h)$ in $G(h)$ such that $\bar{s}_i(k) = s_i(k)$ for all $k \neq h$.



- Let $G = (N, A, H, O, o, P, \{\preceq_i\}_{i \in N})$ be a PIG.
- $A(\emptyset)$ is the set of allowed initial actions for player $i = P(\emptyset)$. For each $b \in A(\emptyset)$, let $s^{G(b)}$ be some strategy

profile for the subgame $G(b)$.

- Given some $a \in A(\emptyset)$, we denote by s^a the strategy profile for G in which player $i = P(\emptyset)$ chooses the initial action a , and for each action $b \in A(\emptyset)$ the subgame $G(b)$ is played according to $s^{G(b)}$.
- So $s_i^a(\emptyset) = a$ and for every player $j, b \in A(\emptyset)$ and $bh \in H \setminus Z$, $s_j^a(bh) = s_j^{G(b)}(h)$.

Lemma 6.1 (Backward Induction)

Let $G = (N, A, H, Z, O, o, P, \{\preceq_i\}_{i \in N})$ be a finite PIG. Assume that for each $b \in A(\emptyset)$ the subgame $G(b)$ has a subgame perfect equilibrium $s^{G(b)}$. Let $i = P(\emptyset)$ and let a be the \succ_i -maximizer over $A(\emptyset)$ of $o_a(s^{G(a)})$. Then s^a is a subgame perfect equilibrium of G .



6.1.3 Strategic Form Game

Definition 6.9 (Normal Form Game)

A game in **normal form** is denoted by

$$G = \left(\underbrace{N}_{\text{players}}, \underbrace{\{S_i\}_{i \in N}}_{\text{Strategy Set}}, \underbrace{\{u_i(\cdot)\}_{i \in N}}_{\text{VNM utility}} \right)$$

$u_i : \prod_{i \in I} S_i \rightarrow \mathbb{R}$ is the utility function that maps all players' strategies to a player's utilities.

A finite game is a normal-form game in which the set of players N is a finite set, and the set of strategy profiles S is finite.



Definition 6.10 (Mixed/Pure Strategy)

A mixed strategy for player i is a probability distribution $\sigma_i \in \Delta(S_i)$.

Elements of S_i are called pure strategies.



Definition 6.11 (Dominant/Dominated Strategy)

A strategy $\sigma_i \in \Delta(S_i)$ is a **dominant strategy** for i in G , if we have $u_i(\sigma_i, \sigma_{-i}) > u_i(\sigma'_i, \sigma_{-i}), \forall \sigma'_i \neq \sigma_i, \sigma_{-i} \in \times_{j \neq i} \Delta(S_j)$.

A strategy $\sigma_i \in \Delta(S_i)$ is a **dominated strategy** for i in G , if $\exists \sigma'_i \neq \sigma_i, u_i(\sigma_i, \sigma_{-i}) < u_i(\sigma'_i, \sigma_{-i}), \forall \sigma_{-i} \in \times_{j \neq i} \Delta(S_j)$.

A strategy $\sigma_i \in \Delta(S_i)$ is a **weakly dominated strategy** for i in G , if $\exists \sigma'_i \neq \sigma_i, u_i(\sigma_i, \sigma_{-i}) \leq u_i(\sigma'_i, \sigma_{-i}), \forall \sigma_{-i} \in \times_{j \neq i} \Delta(S_j)$ and there is a $\sigma_{-i} \in \times_{j \neq i} \Delta(S_j), u_i(\sigma_i, \sigma_{-i}) < u_i(\sigma'_i, \sigma_{-i})$



Lemma 6.2

1. A dominant strategy is always pure.

2. A strategy σ'_i dominates σ_i iff $u_i(\sigma'_i, s_{-i}) > u_i(\sigma_i, s_{-i})$, for all pure strategy profiles $s_{-i} \in S_{-i}$.



Definition 6.12 (Belief, Best Response)

A **belief** for player i is a probability distribution $\mu \in \Delta(S_{-i})$.

A strategy $\sigma_i \in \Delta(S_i)$ is the **best response** to beliefs μ if it solves the problem of $\max_{\sigma_i \in \Delta(S_i)} u_i(\sigma_i, s_{-i})$.

Denote the set of all best responses to μ by $\beta_i(\mu)$.

**Lemma 6.3 (Mixed Strategy is BR iff its Pure Strategies are Indifferent)**

A mixed strategy σ_i is in $\beta_i(\mu)$ iff every pure strategy in the support of σ_i is in $\beta_i(\mu)$. In particular, every strategy in the support of σ_i yields the same payoff to i .

**Theorem 6.3 (Domination Theorem rephrased)**

In a finite game, a strategy is dominated iff there is no belief to which it is a best response.

**Definition 6.13 (Algorithm: Iterated Elimination of Dominated Strategies (IEDS))**

Let $(N, (S_i), (u_i))$ be a finite game; $N = [n]$.

- We define (inductively) n sequences of sets of mixed strategies.
- Let $D_i^0 = \Delta(S_i)$.
- Given $D_1^{k-1}, \dots, D_n^{k-1}$, let

$$D_i^k = \left\{ \sigma_i : \nexists \bar{\sigma}_i : u_i(\sigma_i, \sigma_{-i}) < u_i(\bar{\sigma}_i, \sigma_{-i}) \forall \sigma_{-i} \in \times_{j \neq i} D_j^{k-1} \right\}.$$

- Note that $\{D_i^k\}$ is a decreasing sequence of sets.
- Let $D_i = \cap_{k=0}^{\infty} D_i^k$.
- The set $D = \times_{i=1}^n D_i$ be the set of strategies that survive the iterated elimination of dominated strategies.

A game is called **dominance-solvable** if D is a singleton.

**Definition 6.14 (Rationalizable Strategies)**

- $R_i^0 = \Delta(S_i)$.
- Given $R_1^{k-1}, \dots, R_n^{k-1}$, Let

$$Z_i^k = \left\{ s_i \in S_i : \sigma_i(s_i) > 0 \text{ for some } \sigma_i \in R_i^{k-1} \right\}$$

$$R_i^k = \left\{ \sigma_i \in \Delta(S_i) : \exists \mu \in \Delta(\times_{j \neq i} Z_j^k) \text{ s.t. } \sigma_i \in \beta_i(\mu) \right\}$$

Note: $\{R_i^k\}_{k=0}^{\infty}$ is a decreasing sequence of sets.

Let $R_i = \cap_{k=0}^{\infty} R_i^k$.

The **rationalizable strategies** are the elements of $R = \times_{i=1}^n R_i$.



Lemma 6.4

In a finite game, R is always non-empty and contains a pure strategy profile.

**Proposition 6.1**

$\sigma_i \in \Delta(S_i)$ is **rationalizable** iff there are sets $Z_1, \dots, Z_n, Z_j \subseteq S_j$ such that

1. $\sigma_i \in \beta_i(\mu_i)$ for some $\mu_i \in \Delta(\times_{h \neq i} Z_h)$.
2. for every $s_j \in Z_j$ there is $\mu_j \in \Delta(\times_{h \neq j} Z_h)$ such that $s_j \in \beta_j(\mu_j)$.

**Corollary 6.1 (Rationalizable = IEDS)**

Rationalizable strategies are exactly the strategies survive the iterated elimination of dominated strategies,

$$R = D$$

**6.1.4 Nash Equilibrium and Existence****Definition 6.15 (Nash Equilibrium)**

A strategy profile $\Sigma = (\sigma_1, \dots, \sigma_I)$ is a **Nash** equilibrium of the game G if for every $i \in I$, we have:

$u_i(\sigma_i^*, \sigma_{-i}^*) \geq u_i(\sigma'_i, \sigma_{-i}^*), \forall \sigma'_i \in \Delta(S_i)$ (no profitable deviation). In other words,

1. σ_i is the best response to beliefs $\mu_i \in \Delta(S_{-i})$
2. $\mu_i = \sigma_{-i}$ (correct beliefs).



1. In rationalizable strategies, beliefs can be incorrect.
2. In a Nash equilibrium, beliefs are correct. Any strategy in a Nash equilibrium is rationalizable.

Definition 6.16 (Best Response Correspondence)

In a Nash equilibrium the player i 's best response correspondence $\beta_i : \Delta(S_{-i}) \rightarrow 2^{\Delta(S_i)}$ is defined as $\beta_i(\sigma_{-i}) = \arg \max_{\sigma_i \in \Delta(S_i)} u_i(\sigma_i, \sigma_{-i})$. Let $\beta(\sigma) = \times_{i \in I} \beta_i(\sigma_{-i})$. Then σ is a Nash equilibrium iff $\beta(\sigma) = \sigma$. β is called the **best response correspondence** of the game.

**Theorem 6.4 (Existence of Nash Equilibrium)**

A Nash equilibrium exists in a finite game Γ , if for all $i \in I$,

- (i). S_i is non-empty, convex, compact, subset of \mathbb{R}^m (i.e., for some finite dimensions of real numbers).
- (ii). $u_i(s_i, \dots, s_I)$ is continuous in (s_i, \dots, s_I) and quasi-concave in any s_i .

**Proof 6.4**

We prove a lemma for the best response correspondence $\beta_i(s_{-i}) = \arg \max_{s_i \in S_i} u_i(s_i, s_{-i})$ firstly.

Lemma 6.5

Suppose $\{S_i\}_{i \in I}$ are non-empty. Suppose that S_i is compact and convex and u_i is continuous in (s_i, \dots, s_I) and quasi-concave in any s_i , then best response correspondence $\beta_i(s_{-i})$ is non-empty, convex-valued and uhc.

**Proof 6.5**

This lemma is proved by Berge's Maximum Theorem (Theorem 1.2).

Consider the best response correspondence of the game β with $\beta(s_i, \dots, s_I) = \{\beta_1(s_{-1}), \dots, \beta_I(s_{-I})\}$.

As we proved β is non-empty, convex-valued and uhc from S to S where S is non-empty, compact, and convex. By the Kakutani's Fixed Point Theorem (Theorem 1.6), we have β has a fixed point $s \in S$, which should be the Nash equilibrium.

6.1.5 Bayesian Game

Definition 6.17 (Bayesian Game)

A **Bayesian game** is defined by

$$\Gamma = (I, \Omega, \{A_i\}_{i \in I}, \{u_i(\cdot)\}_{i \in I}, \{\Theta_i\}_{i \in I}, \{F_i\}_{i \in I})$$

where Ω is the state space, $u_i : A \times \Omega$ is i 's payoff function, and $F_i \in \Delta(\Omega \times \Theta_i)$ is the (prior) distribution of the player i 's type.

**Definition 6.18 (Normal-form Bayesian game)**

Assume a finite game. The **normal-form game** can be represented by

$$(I, (S_i, U_i)_{i \in I})$$

defined by letting S_i be the set of strategies based on types $s_i : \Theta_i \rightarrow A_i$ and

$$U_i(s) = \sum_{\omega \in \Omega} \sum_{(\theta_i)_{i \in I} \in \Theta} p(\omega, \theta_1, \dots, \theta_I) u_i(s_1(\theta_1), \dots, s_I(\theta_I), \omega)$$

for all $s \in S$.

A **Bayesian Nash equilibrium** (BNE) of a Bayesian game is a strategy profile (s_1, \dots, s_n) that is a Nash equilibrium of the derived normal-form game.

**Definition 6.19 (Best Response, Interim Payoff)**

s_i is a BR to s_{-i} iff for all θ_i , $s_i(\theta_i)$ maximizes the **interim payoff** of player i . The interim payoff is

defined by the expected payoff given the type θ_i of player i by playing action a_i .

$$\mathbb{E}_{\omega \in \Omega, \tilde{\theta}_{-i} \in \Theta_{-i}} [u_i(a_i, s_{-i}(\tilde{\theta}_{-i}), \omega) | \theta_i]$$



6.1.6 Zero-sum Game

Let $G = (\{1, 2\}, (S_1, S_2), (u_1, u_2))$. Suppose there is a constant c so that $u_1(s) + u_2(s) = c, \forall s \in S$. Then G is equivalent to a zero-sum game.

Definition 6.20 (Saddle Point)

Let X, Y be sets and $f : X \times Y \rightarrow \mathbb{R}$ a real function. $(x^*, y^*) \in X \times Y$ is a **saddle point** of f if $x^* \in \operatorname{argmax}_{x \in X} f(x, y^*)$ and $y^* \in \operatorname{argmin}_{y \in Y} f(x^*, y)$. That is,

$$f(x, y^*) \leq f(x^*, y^*) \leq f(x^*, y), \forall x \in X, y \in Y$$



Consider a zero-sum game of two players. The strategy of the player 1 is max-min strategy, which is given by

$$\max_{\sigma_1} \min_{\sigma_2} u(\sigma_1, \sigma_2)$$

and the strategy of the player 2 is min-max strategy, which is given by

$$\min_{\sigma_2} \max_{\sigma_1} u(\sigma_1, \sigma_2)$$

Proposition 6.2 (min-max ≥ max-min)

Min-max strategy is always better than max-min strategy. That is,

$$\max_{\sigma_1} \min_{\sigma_2} u(\sigma_1, \sigma_2) \leq \min_{\sigma_2} \max_{\sigma_1} u(\sigma_1, \sigma_2)$$



Proof 6.6

As $u(\sigma'_1, \sigma_2) \leq \max_{\sigma_1} u(\sigma_1, \sigma_2)$ for all σ'_1 , we have

$$\begin{aligned} \min_{\sigma_2} u(\sigma'_1, \sigma_2) &\leq \min_{\sigma_2} \max_{\sigma_1} u(\sigma_1, \sigma_2), \forall \sigma'_1 \\ \Rightarrow \max_{\sigma_1} \min_{\sigma_2} u(\sigma_1, \sigma_2) &\leq \min_{\sigma_2} \max_{\sigma_1} u(\sigma_1, \sigma_2) \end{aligned}$$

We shall prove that these are in fact equal in the zero-sum game.

Definition 6.21 (Value of a zero-sum game)

A **value** for a zero-sum game G is a number $v \in \mathbb{R}$ for which there exists a strategy profile $(\bar{\sigma}_1, \bar{\sigma}_2)$ such

that

$$u(\bar{\sigma}_1, \sigma_2) \geq v \quad \text{for all } \sigma_2$$

$$u(\sigma_1, \bar{\sigma}_2) \leq v \quad \text{for all } \sigma_1$$



Note: $v = u(\bar{\sigma}_1, \bar{\sigma}_2)$. The value is unique (if it exists), and represents a guaranteed payoff for the players.
 (Uniqueness: Suppose there are two values, v and $v' > v$, achieved by profiles σ and σ' . Then when 2 plays σ_2 we have that $u(\sigma'_1, \sigma_2) \geq v'$ because σ'_1 guarantees v' . And when 1 plays σ'_1 we have $v \geq u(\sigma'_1, \sigma_2)$ because σ_2 guarantees v . So $v' > v$ leads to a contradiction.)

Proposition 6.3

The following statements are equivalent

1. $(\bar{\sigma}_1, \bar{\sigma}_2)$ is a Nash equilibrium.
2. $v = u(\bar{\sigma}_1, \bar{\sigma}_2)$.



Corollary 6.2

If (σ_1^*, σ_2^*) and $(\bar{\sigma}_1, \bar{\sigma}_2)$ are Nash equilibria of G , then so are the profiles $(\bar{\sigma}_1, \sigma_2^*)$ and $(\sigma_1^*, \bar{\sigma}_2)$.



Theorem 6.5 (Minimax Theorem)

Let G be a zero-sum game. There is a strategy profile (σ_1^*, σ_2^*) s.t.

$$\max_{\sigma_1} \min_{\sigma_2} u(\sigma_1, \sigma_2) = v = \min_{\sigma_2} \max_{\sigma_1} u(\sigma_1, \sigma_2)$$

where $v = u(\sigma_1^*, \sigma_2^*)$ is the value of the game and (σ_1^*, σ_2^*) is a Nash equilibrium.



Proof 6.7

Let $n_i = |S_i|$. Let $\vec{u}(\sigma_2) := \{u(s_1, \sigma_2) : s_1 \in S_1\}$. Let $\mathbb{C} = \{\vec{u}(\sigma_2) : \sigma_2 \in \Delta(S_2)\}$. We can find \mathbb{C} is convex and compact.

Let $m(x) := \max\{x_i : i = 1, \dots, n_1\}$. Then, the player 2's min max payoff is given by $v := \inf\{m(x) : x \in \mathbb{C}\}$. By compactness, exists strategy σ_2^* such that $\vec{u}(\sigma_2^*) = (v, v, \dots, v)$. That is, $u(s_1, \sigma_2^*) = v, \forall s_1 \in S_1$.

Let $\mathbb{A} := \{z \in \mathbb{R}^{n_1} : z << (v, v, \dots, v)\} = (v, v, \dots, v) - \mathbb{R}_{++}^{n_1}$. As \mathbb{C} and \mathbb{A} are disjoint. By SHT, we can find a $p \neq 0$ s.t. $p \cdot \mathbb{A} \leq p \cdot \mathbb{C}$. $p > 0$ since \mathbb{A} can have arbitrary small elements in any dimension.

Then, we normalize p to be in $\Delta(S_1)$, denote it by σ_1^* .

By limitation, $v = \sigma_1^* \cdot (v, v, \dots, v) = \lim_{\epsilon \rightarrow 0^+} \sigma_1^* \cdot (v - \epsilon, v - \epsilon, \dots, v - \epsilon) \leq \sigma_1^* \cdot \vec{u}(\sigma_2)$, $\forall \sigma_2 \in \Delta(S_2)$.

Hence, $u(s_1, \sigma_2^*) \leq m(\vec{u}(\sigma_2^*)) = v = u(\sigma_1^*, \sigma_2^*) \leq u(\sigma_1^*, \sigma_2)$

6.1.7 Correlated equilibrium

Suppose there is a mediator that give advices to each player based on a distribution $p \in \Delta(S)$. A player doesn't know other players' advices but knows the distribution p .

Definition 6.22 (Correlated Equilibrium)

A **correlated equilibrium** of G is any probability distribution $p \in \Delta(S)$ such that, for all i and $s_i, s'_i \in S_i$, the player i can't get a higher expected profit than by following the advice,

$$\sum_{s_{-i} \in S_{-i}} p(s_i, s_{-i}) u_i(s_i, s_{-i}) \geq \sum_{s_{-i} \in S_{-i}} p(s_i, s_{-i}) u_i(s'_i, s_{-i})$$

where LHS is the expected profit of player i when he receives an advice s_i from the mediator.



Let G be a finite game,

Proposition 6.4

If we identify a Nash equilibrium σ of G with a probability distribution on $\Delta(S)$, then any Nash equilibrium of G is also a correlated equilibrium.



Proposition 6.5

The set of correlated equilibria is a non-empty, convex, compact subset of $\Delta(S)$.



For $p \in \Delta(S)$, the **marginal distribution** on S_i is given by

$$p_i(s_i) = \sum_{s_{-i} \in S_{-i}} p(s_i, s_{-i})$$

Proposition 6.6

A correlated equilibrium that is the independent mixture of its marginal distributions is a Nash equilibrium.



Proposition 6.7 (Correlated Equilibrium Strategy \Rightarrow Rationalizable)

Let $G = (S_i, u_i)_{i=1}^n$ be a finite normal-form game, and ρ a correlated equilibrium of G . Suppose that the profile (s_i, s_{-i}) receives strictly positive probability in ρ : $\rho(s_i, s_{-i}) > 0$. s_i is rationalizable.



Proof 6.8

Let Z_i be the set of pure strategies of player i that receive strictly positive probability in ρ_i . For each

$s_i \in Z_i$ we have that $\sum_{s_{-i} \in S_{-i}} \rho(s_i, s_{-i}) \geq \sum_{s_{-i} \in S_{-i}} \rho(s'_i, s_{-i}), \forall s'_i \in S_i$, so for any $s'_i \in S_i$:

$$\begin{aligned}\mathbb{E}_{\mu_i} u_i(s_i, s_{-i}) &= \sum_{s_{-i} \in S_{-i}} \frac{\rho(s_i, s_{-i})}{\sum_{\tilde{s}_{-i} \in S_{-i}} \rho(s_i, \tilde{s}_{-i})} u_i(s_i, s_{-i}) \\ &\geq \sum_{s_{-i} \in S_{-i}} \frac{\rho(s_i, s_{-i})}{\sum_{\tilde{s}_{-i} \in S_{-i}} \rho(s_i, \tilde{s}_{-i})} u_i(s'_i, s_{-i}) = \mathbb{E}_{\mu_i} u_i(s'_i, s_{-i})\end{aligned}$$

where $\mu_i \in \Delta(S_{-i})$ are the beliefs over S_{-i} obtained by ρ conditioning on s_i . This means that s_i is a best response to beliefs μ_i . Since i and $s_i \in Z_i$ are arbitrary, we are done.

6.1.8 Quantal Response Equilibrium

Imagine choosing $a \in A$.

Definition 6.23 (Quantal Response (softmax))

A **quantal response** is a function $\gamma : \mathbb{R}^A \rightarrow \Delta(A)$ mapping from a vector of utility values v to a probability distribution over actions, which satisfies that

- $\gamma(v) \gg 0$ for all v (interior);
- γ is continuous,
- γ is monotonic ($v_h < v_j \rightarrow \gamma_h(v) < \gamma_j(v)$)
- γ is responsive $\left(v_j < v'_j \rightarrow \gamma_j(v) < \gamma_j(v'_j, v_{-j})\right)$



Interpret $\gamma(v)$ as the probability of choosing each of $a \in A$ alternatives when v is the vector of utility values of the alternatives in A .

Interpretation: mistakes or random utility.

One common quantal response function is the logistic function:

$$\gamma_j(v) = \frac{e^{\lambda v_j}}{\sum_{h \in I} e^{\lambda v_h}}$$

for $\lambda > 0$, where λ captures how close the quantal response is to choosing according to the largest values of v .

Fix a normal-form game $G = \{N, \{S_i, i \in N\}, \{u_i, i \in N\}\}$. Let $\gamma_i : \mathbb{R}^{S_i} \rightarrow \Delta(S_i)$ be a quantal response for player i .

Definition 6.24 (Quantal Response Equilibrium)

A **quantal response equilibrium** of G is a strategy profile σ^* such that

$$\sigma_i^* = \gamma_i(\vec{u}_i(\sigma_{-i}^*))$$

where $\vec{u}_i(\sigma_{-i}^*) = (u_i(s_i, \sigma_{-i}^*))_{s_i \in S_i}$.



Proposition 6.8

Every finite normal-form game, with any profile of quantal responses, has a quantal response equilibrium.



Observation: λ measures the distance to Nash.

Proposition 6.9

Let $\{\lambda^k\}$ be a sequence such that $\lim_{k \rightarrow \infty} \lambda^k = \infty$ and $\sigma^(\lambda^k)$ be a QRE when the logistic quantal responses take parameter value λ^k . If $\{\sigma^*(\lambda^k)\}$ is a convergent sequence, it converges to a Nash equilibrium.*



6.2 Knowledge and Common Knowledge

6.2.1 Knowledge and Information

1. Let Ω be a (finite) set of possible states of the world. Information is provided by a subset of Ω . The smaller a subset is, the more information it provides.
2. Subsets $E \subseteq \Omega$ are **events**.
- 3.

Definition 6.25 (Information Function)

We define the function $P : \Omega \rightarrow 2^\Omega$ an **information function**. $P(\omega)$ is the set of states that the agent considers possible when the actual state is ω .



When the state is ω the decision-maker knows only that the state is in the set $P(\omega)$. Means that, if $\omega' \in P(\omega)$, then when the state is ω , information doesn't allow one to distinguish between ω and ω' .

- 4.

Definition 6.26 (Knowledge)

Knowledge is modeled through a function $K : 2^\Omega \rightarrow 2^\Omega$. $K(E)$ (which we write as KE) is the set of states at which the agent knows that the event E has occurred. That is, given an information function $P : \Omega \rightarrow 2^\Omega$, the **knowledge** $K : 2^\Omega \rightarrow 2^\Omega$ is defined as

$$KE = \{\omega \in \Omega : P(\omega) \subseteq E\}$$

We can say **KE** is the set of states that “the agent knows E .”



5. Given a state ω , $\{E \subseteq \Omega : KE \ni \omega\} = \{E \subseteq \Omega : P(\omega) \subseteq E\}$ is the set of all events that the agent “knows” (all events that the agent believes are possible). The most accurate information provided by it is $\cap\{E \subseteq \Omega : KE \ni \omega\}$. Hence,

$$P(\omega) = \cap\{E \subseteq \Omega : KE \ni \omega\}$$

These two equations provide the back and fourth relationship between knowledge and information. However, we don't give any restrictions for the settings of the knowledge or the information function, so they can be any forms.

6.2.2 Partitional Information Function

Definition 6.27 (P1&P2 Conditions for Information Function)

Usually, we assume following two conditions of a information function:

- P1. $\omega \in P(\omega)$ for every $\omega \in \Omega$. (Reality will not be excluded from agent's information structure.)
- P2. If $\omega' \in P(\omega)$ then $P(\omega') = P(\omega)$.



Definition 6.28 (S5 Conditions for Knowledge)

There are 5 axioms of knowledge that can restrict the form of knowledge.

1. $K\Omega = \Omega$

Alice knows that some state of the world has occurred.

2. $KA \cap KB = K(A \cap B)$

Alice knows A and knows B iff she knows A and B .

3. $KA \subseteq A$ (Axiom of knowledge)

If Alice knows A , then A has indeed occurred (some state in A is true).

4. $KKA = KA$ (Axiom of positive introspection)

If Alice knows A then she knows that she knows A .

5. $(KA)^c = K((KA)^c)$ (Axiom of negative introspection)

If Alice doesn't know A then she knows that she doesn't know A .

They are not independent.



Definition 6.29 (Partitional)

Information function P (and associated knowledge operator K) is **partitional** if $\{P(\omega) : \omega \in \Omega\}$ constitute a partition of Ω .



When P is partitional we abuse notation and denote the partition by P as well.

Theorem 6.6 (S5 \Leftrightarrow P1&P2 \Leftrightarrow Partitional)

For a P (and associated K), following conditions are equivalent:

- o P/K is partitional;
- o P satisfies P1 and P2.
- o K satisfies S5.



Example 6.1 (Partitional) $\Omega = \{\omega_1, \omega_2, \omega_3, \omega_4\}$.

1. Information Structure: $P(\omega_1) = P(\omega_2) = \{\omega_1, \omega_2\}$, $P(\omega_3) = \{\omega_3\}$, and $P(\omega_4) = \{\omega_4\}$.

2. Knowledge Function: $K(\{\omega_3, \omega_4\}) = \{\omega_3, \omega_4\}$, $K(\{\omega_1, \omega_3\}) = \{\omega_3\}$.

Example 6.2 (Non-Partitional) $\Omega = \{\text{bark, don't bark}\}$. $P(\omega) = \begin{cases} \{\text{bark}\} & \text{if } \omega = \text{bark} \\ \{\text{bark, don't bark}\} & \text{if } \omega = \text{don't bark} \end{cases}$

Then, $K(\{\text{bark}\}) = \{\text{bark}\}$ and $K(\{\text{don't bark}\}) = \emptyset$. Axiom 5 is violated.

6.2.3 Self-evident and Algebra

Definition 6.30 (Self-evident)

An event $A \in 2^\Omega$ is **self-evident** if

$$KA = A$$

That is, the agent “knows A” if and only if A happens.



Proposition 6.10 (Self-evident \Leftrightarrow Unions of Elements of Partition)

“A is self-evident” if and only if it is the union (1 or more) of elements of the partition P .



Proof 6.9

A is self-evident iff $A = KA = \{\omega : P(\omega) \subseteq A\}$.

Definition 6.31 (Algebra)

A collection of events Σ is an **algebra** if it satisfies:

1. $\Omega \in \Sigma$;
2. If $A \in \Sigma$ then $A^c \in \Sigma$;
3. If $A, B \in \Sigma$ then $A \cup B \in \Sigma$.



Let Σ be the **collection of self-evident events**.

Corollary 6.3 (Set of Self-Evident Events is an Algebra)

Then Σ is an algebra (in fact, $\Sigma = \Sigma_P$, the algebra generated by the partition P).



Example 6.3 $\Omega = \{\omega_1, \omega_2, \omega_3\}$

1. $P_1 = \{\{\omega_1, \omega_2\}, \{\omega_3\}\}$, $\Sigma_1 = \{\emptyset, \{\omega_1, \omega_2\}, \{\omega_3\}, \Omega\}$.

2. $P_2 = \{\{\omega_1\}, \{\omega_2\}, \{\omega_3\}\}$, $\Sigma_2 = \{\emptyset, \{\omega_1\}, \{\omega_2\}, \{\omega_3\}, \{\omega_1, \omega_2\}, \{\omega_1, \omega_3\}, \{\omega_2, \omega_3\}, \Omega\}$.

3. $P_3 = \{\{\omega_1\}, \{\omega_2, \omega_3\}\}$, $\Sigma_3 = \{\emptyset, \{\omega_1\}, \{\omega_2, \omega_3\}, \Omega\}$.

Corollary 6.4 ($KA \in \Sigma$)

For any A (despite whether it is in Σ) and K is partitional, we have

$$KA = \cup\{S \in \Sigma : S \subseteq A\} \in \Sigma$$

So KA is always self-evident. And, since any algebra is closed under unions, it follows that KA is the largest element of Σ that is contained in A .

**Proof 6.10**

We prove by two directions:

(1). $\omega \in KA \Rightarrow \omega \in \cup\{S \in \Sigma : S \subseteq A\}$: Given $\omega \in KA$, we have $P(\omega) \subseteq A$. Since K is partitional, $P(\omega) \in \Sigma$. Therefore, $P(\omega) \in \{S \in \Sigma : S \subseteq A\}$. Hence, $\omega \in \cup\{S \in \Sigma : S \subseteq A\}$.

(2). $\omega \in \cup\{S \in \Sigma : S \subseteq A\} \Rightarrow \omega \in KA$: Given $\omega \in \cup\{S \in \Sigma : S \subseteq A\}$, there exists $S \in \Sigma$ such that $S \subseteq A$ and $\omega \in S$. Since Σ is the collection of self-evident events, we have $KS = S$. Hence, $\omega \in KS \subseteq KA$.

Then, we can recover $P(\omega)$ from Σ by

$$P(\omega) = \cap\{S \in \Sigma : \omega \in S\} \in \Sigma$$

All in all, a knowledge space can be defined by P , K , or Σ .

6.2.4 Common Knowledge

Consider a finite set of N agents, each with a (partitional) knowledge function $K_i : 2^\Omega \rightarrow 2^\Omega, i \in N$.

Definition 6.32 (Refinement and Coarsening of Sub-algebras)

Let Σ, Π be two sub-algebras of some algebra. We say that Σ is a **refinement** of Σ and Π is a **coarsening** of Σ if $\Pi \subseteq \Sigma$.



Example 6.4 Consider two sub-algebras of $\{\emptyset, \{\omega_1\}, \{\omega_2\}, \{\omega_3\}, \{\omega_1, \omega_2\}, \{\omega_1, \omega_3\}, \{\omega_2, \omega_3\}, \Omega\}$,

$$\Sigma = \{\emptyset, \{\omega_1\}, \{\omega_2\}, \{\omega_3\}, \{\omega_1, \omega_2\}, \{\omega_1, \omega_3\}, \{\omega_2, \omega_3\}, \Omega\} \text{ and } \Pi = \{\emptyset, \Omega\}$$

where $\Pi \subseteq \Sigma$. The Π provides a less inaccurate information structure than Σ .

Definition 6.33 (Meet and Join of Sub-algebras)

The **meet** of two algebras $\Sigma_1, \Sigma_2 \subseteq \Sigma$ is the finest sub-algebra of Σ that is a coarsening of each Σ_i .

Their **join** is the coarsest sub-algebra of Σ that is a refinement of each Σ_i .



Definition 6.34 (Common Knowledge)

An event A is said to be **common knowledge** at $\omega \in \Omega$ if for any sequence $i_1, i_2, \dots, i_k \in N$ it holds that

$$\omega \in K_{i_1} K_{i_2} \cdots K_{i_k} A$$



Let $\Sigma_C = \cap_i \Sigma_i$ be the meet of the player's algebras.

Proposition 6.11

The following are equivalent:

1. $C \in \Sigma_C$.
2. $K_i C = C, \forall i \in N$.

**Definition 6.35 (K_C)**

Recall that $K_i A = \cup\{S \in \Sigma_i : S \subseteq A\}$. Analogously, we define a knowledge operator from Σ_C :

$$K_C A = \cup\{S \in \Sigma_C : S \subseteq A\}$$



Since $K_C A \in \Sigma_i$ for any i , we have $K_i K_C A = K_C A$. Hence, $K_C A$ is common knowledge at $\omega \in K_C A$. As Σ_C is also the set of $\{K_C A : A \in 2^\Omega\}$, we can conclude that

Proposition 6.12 ($E \in \Sigma_C \Leftrightarrow E$ is common knowledge at $\omega \in E$)

For any $E \in \Sigma_C$, E is common knowledge at $\omega \in E$. Conversely, if E is common knowledge at any $\omega \in E$, then $E \in \Sigma_C$ ($E = K_i E$ for any i).



Example 6.5 Consider the following game, which is a simplification of the popular board game “Clue.” There are two decks of cards, with four cards each. Deck 1 has numbered cards, with numbers 1,2,3 and 4. Deck 2 has cards with the suites: \heartsuit , \diamondsuit , \spadesuit and \clubsuit . One card from each deck is set aside; turned face down and not seen by any of the players. Of the remaining cards, Player 1 gets all the odd-numbered cards from Deck 1 while Player 2 gets the even cards. From the remainder of Deck 2, Player 1 gets the red cards (\heartsuit and \diamondsuit) while 2 gets the black cards (\spadesuit and \clubsuit). For example, if the cards set aside are 4 and \diamondsuit , then P1 gets 1, 3 and \heartsuit , while P2 gets 2, \clubsuit and \spadesuit . Players don't see how many cards the other player received.

A state of the world is a pair of cards that was set aside (and the point of the game is to find out what these are).

Show that the event

$$E = \{(3, \clubsuit), (3, \spadesuit), (1, \clubsuit), (1, \spadesuit)\}$$

is common knowledge when the cards set aside are the number 3 and \clubsuit .

Answer The state is a pair (n, s) , with n being a number and s a suit. At any state in E , n is odd and s is a black suit. So for any such state the players' information functions take values: $P_1(n, s) = \{(n, \clubsuit), (n, \spadesuit)\}$

and $P_2(n, s) = \{(1, s), (3, s)\}$.

Thus, for any $(n, s) \in E$, $P_i(n, s) \subseteq E$. This means that E is the union of elements of the partition P_i , and thus E is self-evident for each player. So $E \in \Sigma_1 \cap \Sigma_2$ and it is common knowledge at any $(n, s) \in E$.

6.2.5 Application: agreeing to disagree

Definition 6.36 (Belief Space)

Suppose

1. a finite set N of players,
2. a state-space Ω ,
3. each agent endowed with a partitional knowledge K_i ; resulting in an algebra Σ_i on Ω ,
4. each agent endowed with a prior belief $\mu_i \in \Delta(\Omega)$

The tuple $(N, \Omega, \{\mu_i\}_{i \in N}, \{\Sigma_i\}_{i \in N})$ is a **belief space**.



Fix $\mu \in \Delta(\Omega)$, the expectation of a random variable $X : \Omega \rightarrow \mathbb{R}$ according to μ is denoted by $\mathbb{E}_\mu X$.

Suppose that $\mu(B) \in (0, 1)$, a conditional probability can be written as $\mu(A|B) = \frac{\mu(A \cap B)}{\mu(B)}$, $\forall A \subseteq \Omega$. Conditional expectation $\mathbb{E}[X|B] = \mathbb{E}_{\mu(\cdot|B)} X$.

Given an algebra Σ_i (which represent a knowledge space), we define the expectation of X conditioned on player i 's information at ω is

$$\mathbb{E}[X|\Sigma_i](\omega) = \mathbb{E}[X|P_i(\omega)] = \frac{\sum_{\omega' \in P_i(\omega)} \mu(\omega') X(\omega')}{\mu(P_i(\omega))}$$

(where we assume $\mu(P_i(\omega)) > 0$.)

An event that $\mathbb{E}[X|\Sigma_i] = q$ is the event that, given i 's information (Σ_i), her conditional expectation for the random variable X is q : $\{\omega \in \Omega : \mathbb{E}[X|\Sigma_i](\omega) = q\}$.

Proposition 6.13

The Law of Iterated (Total) Expectations: let S be an element of an algebra Π , and let $X : \Omega \rightarrow \mathbb{R}$ be a r.v. Then,

$$\mathbb{E}[X|S] = \mathbb{E}[\mathbb{E}[X|\Pi]|S]$$



Consider a belief space $(N, \Omega, \{\mu_i\}_{i \in N}, \{\Sigma_i\}_{i \in N})$ and a **common prior** $\mu = \mu_i, \forall i \in N$.

Theorem 6.7 (Aumann's Agreement Theorem, 1976)

Let $X : \Omega \rightarrow \mathbb{R}$ be a random variable. Suppose that, at ω_0 , it is common knowledge of posteriors that $\mathbb{E}[X|\Sigma_i] = q_i$ for $i = 1, \dots, n$ and some $(q_1, q_2, \dots, q_n) \in \mathbb{R}^n$. Then $q_1 = q_2 = \dots = q_n$.



Proof 6.11

For $i \in N$, let A_i be the event that $\mathbb{E}[X|\Sigma_i] = q_i$. By the common knowledge hypothesis there is a $C \in \Sigma_C = \cap_i \Sigma_i$ such that $\omega_0 \in C \subset \cap_i A_i$. Hence, $\mathbb{E}[X|\Sigma_i](\omega) = q_i$ for all $\omega \in C$. Thus, for all i ,

$$\mathbb{E}[X|C] = \mathbb{E}[\mathbb{E}[X|\Sigma_i]|C] = \mathbb{E}[q_i|C] = q_i, \forall i \in N$$

Corollary 6.5

If two players have common priors over a finite space, and it is common knowledge that their posteriors for some event S are q_1 and q_2 , then $q_1 = q_2$.



6.3 Extensive-Form Games

In this section, we want to study extensive-form games with incomplete information.

We have discussed the extensive form game with perfect information (PIG), which is denoted by $G = \{N, A, H, Z, P, O, o, \succ_r\}$. P is a function mapping from H/Z (non-terminal history) to N (one of the players). Now we shall allow two or more players to make simultaneous moves, and the P is a function from H/Z to $2^N/\{\emptyset\}$, the power set of the set of players, so that after history $h \in H$ the set of players who play simultaneously is $P(h)$.

Without losing generality, we can represent an extensive-form game as $\Gamma = (N, A, H, P, \{u_i\}_{i \in N})$.

For each player $i \in N$, let

$$H_i = \{h \in H/Z : i \in P(h)\}$$

be the set of histories at which i moves.

A **strategy** for player i is a function $s_i : H_i \rightarrow A$ for which

$$s_i(h) \in A_i(h), \forall h \in H_i$$

Let S_i be the set of all strategies available for player i . A strategy profile is a collection of all players' strategies.

Thus, we obtain a normal-form game $(N, (S_i, u_i)_{i \in N})$ defined from Γ .

6.3.1 Mixed and Behavioral Strategies

In this example, a mixed strategy of the P1 can be represented by a probability distribution over $\{OU, OD, IU, ID\}$.

To replicate mixed strategy, we can use **behavioral strategy** that an agent's (probabilistic) choice at each node is independent of his/her choices at other nodes.

Example 6.6

1. Mixed Strategy: OU with probability $\frac{1}{2}$ and ID with probability $\frac{1}{2}$.
2. Behavioral Strategy: play O and I with prob $\frac{1}{2}$ each at $h = \emptyset$; and then D with prob 1 at history $h = I$.

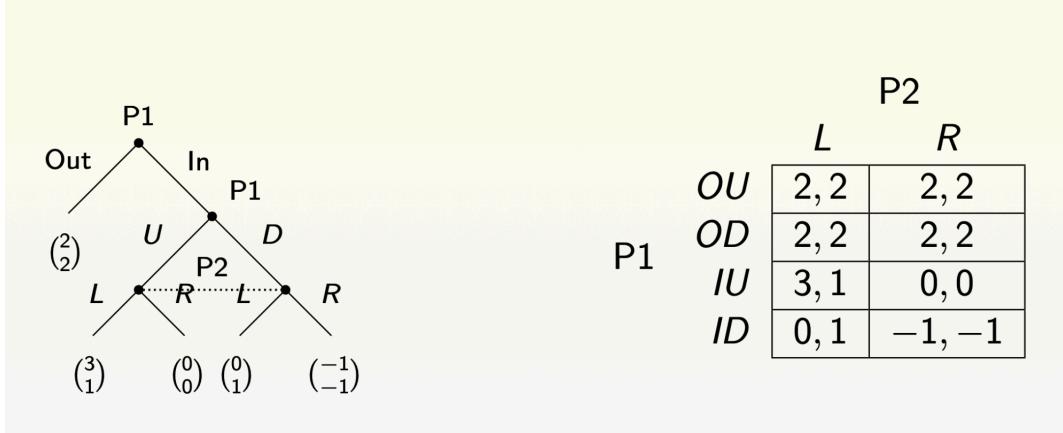


Figure 6.1: Example

Definition 6.37 (Behavioral Strategy)

A **behavioral strategy** for player i is a function that maps histories $h \in H_i$ into probability distributions over $A_i(h)$,

$$\sigma_i(h) \in \Delta(A_i(h))$$



A profile of behavioral strategies, $\sigma = \{\sigma_i : i \in N\}$, includes a probability distribution over Z . So does a profile of mixed strategies. Then $u_i(\sigma)$ is the expected payoff.

Theorem 6.8 (Outcome Equivalent)

Behavioral and mixed strategies are “outcome equivalent” in games of perfect recall.

**6.3.2 Subgame Perfect Equilibrium****Definition 6.38**

A strategy profile σ^* is a **Nash equilibrium** if, for all i ,

1. σ_i^* is a best response to beliefs μ_i ,
2. μ_i coincides with σ_{-i}^* .



When s is a pure-strategy Nash eq. $h(s)$ is the equilibrium path of s .

Definition 6.39

A subgame of an extensive-form game $G = \{N, A, H, Z, P, O, \succ_{n \in N}\}$ is a game that starts after a given finite history $h \in H$.

Formally, the subgame $\Gamma(h)$ associated with $h = (h_1, \dots, h_n) \in H$ is $\Gamma(h) = (N, A, H_h, P_h, \{u_{i,h}\}_{i \in N})$, where $H_h = \{(a_1, a_2, \dots) : (h_1, \dots, h_n, a_1, a_2, \dots) \in H\}$. $P_h(h') = P(hh')$ for all non-terminal $h' \in H_h$, and $u_{i,h} = u_i(hh')$ for any terminal $h' \in H$.



A behavioral strategy σ_i of i in Γ defines a behavioral strategy $\sigma_i(\cdot|h)$ of $\Gamma(h)$ by $\sigma_i(h'|h) = \sigma_i(hh')$.

Definition 6.40 (Subgame Perfect Equilibrium)

A **subgame perfect equilibrium**[SPNE/SPE] of Γ is a strategy profile σ^* such that for every subgame $\Gamma(h)$ it holds that σ^* (more precisely, its restriction to H_h) is a Nash equilibrium of $\Gamma(h)$.



Over time $t = 1, \dots, T$, the set of actions is A^T .

Lemma 6.6

σ^* is a SPNE iff its restriction $\sigma^*(\cdot|h)$ is SPNE of $\Gamma(h)$, for all subgames $\Gamma(h)$.

**6.3.3 Infinite Horizon Games****Definition 6.41 (Finite Horizon)**

An extensive-form game has a **finite horizon** if there is a bound on the length of any history in Z .



In these games, SPNE can be found through “generalized backwards induction.”

Definition 6.42 (Continuous At Infinity Utility Function)

A utility function $v : A^T \rightarrow \mathbb{R}$ is **continuous at infinity** if

$$\lim_{t \rightarrow \infty} \sup\{|v(\hat{h}) - v(h)| : h, \hat{h} \in A^T\} = 0$$

That is, eventually, a negligible effect on utility.

**Definition 6.43 (Discounted Utility)**

We focus on additively separable models with a sub-utility $u : A \rightarrow \mathbb{R}$ that is the same for each t . We model a **discounted utility** by assuming a discount factor $\delta \in [0, 1]$:

$$v(a_1, \dots) = (1 - \delta) \sum_{t=1}^{\infty} \delta^{t-1} u(a_t)$$



Can think of δ as the probability that the game will go on for another period, and $1 - \delta$ the prob that the game will end.

“Ending” events are independent: A geometric distribution.

The probability that the game ends at time t is $(1 - \delta)^{t-1}$ and $\sum(1 - \delta)\delta^t u(a_t)$ becomes expected utility.

Proposition 6.14

$v(a_1, \dots) = (1 - \delta) \sum_{t=1}^{\infty} \delta^{t-1} u(a_t)$ is continuous at infinity.

**6.3.4 Repeated Games**

Let $G_0 = (N, \{A_i : i \in N\}, \{u_i : i \in N\})$ be a normal-form game. We let $A = \prod_{i \in N} A_i$. We only consider games in which A is compact and each u_i is continuous.

Let T be the number of periods (or repetitions) of the repeated game. T can be either finite or infinite. G_0 is the stage game.

Definition 6.44 (Repeated Game)

A T -repeated, δ -discounted, game of G_0 is an extensive form game $\Gamma = (N, A, H, P, \{u_i\}_{i \in N}, \delta)$, where

1. $P(h) = N$ for all $h \in H/Z$.
2. The set of histories H has terminal histories $Z = A^T$; recall that Z uniquely determines H as the set of all prefixes of Z .
3. $v_i : Z \rightarrow \mathbb{R}$ is a payoff function of the form

$$v((a_t)_{t=1}^T) = \begin{cases} (1 - \delta) \sum_{t=1}^{\infty} \delta^{t-1} u(a_t), & \delta \in (0, 1) \\ \sum_{t=1}^T u_i(a_t), & \delta = 1 \end{cases}$$



Given a strategy profile s , we obtain a terminal history $z \in Z$, and (abusing notation) write $v_i(s)$ for $v_i(z)$.

We call $(v_1(s), \dots, v_n(s))$ the **payoff profile** associated with s .

Definition 6.45 ((Profitable) One-Stage Deviation)

A **one-stage deviation** from σ by i is a strategy σ'_i that coincides with σ_i everywhere except at a single history.

A **profitable one-stage deviation** from σ by i is a strategy σ'_i that coincides with σ_i at a single history h , and for which $v_i(\sigma'_i, \sigma_{-i}|h) > v_i(\sigma_i, \sigma_{-i}|h)$.



Proposition 6.15 (SPNE \Leftrightarrow No Profitable One-Stage Deviation)

Let Γ be a discounted, infinitely repeated game. Then a strategy-profile is a SPNE of Γ iff no player has a profitable one-stage deviation.



Corollary 6.6 (NE \Rightarrow SPNE)

If s is a pure-strategy Nash equilibrium of G_0 , then the strategy profile in which each player chooses s_i after any history, is a SPNE of infinitely-repeated discounted G_0 .



Proposition 6.16 (Nash Reversion Folk Theorem)

Let σ be a Nash equilibrium of the finite stage game G_0 . Let v be a feasible payoff profile s.t. $u_i(\sigma) < v_i$ for all players i . Then there is $\underline{\delta}$ s.t. if $\delta \geq \underline{\delta}$ then there exists a SPNE of infinitely repeated G_0 with discount factor δ in which players' equilibrium payoffs is the profile v .



No Nash equilibrium that has higher payoffs for all players than Nash equilibrium can be obtained with relatively high discount factor in an infinite game.

6.4 Signaling Game

6.4.1 Canonical Game

Definition 6.46 (Canonical Game)

1. There are two players: **S** (sender) and **R** (receiver).
2. **S** holds more information than **R**: the value of some random variable t with support \mathcal{T} . (We say that t is the **type** of **S**)
3. Prior belief of **R** concerning t are given by a probability distribution ρ over \mathcal{T} (common knowledge)
4. **S** sends a **signal** $s \in \mathcal{S}$ to **R** drawn from a signal set \mathcal{S} .
5. **R** receives this signal, and then takes an **action** $a \in \mathcal{A}$ drawn from a set \mathcal{A} (which could depend on the signal s that is sent).
6. **S**'s payoff is given by a function $u : \mathcal{T} \times \mathcal{S} \times \mathcal{A} \rightarrow \mathbb{R}$ and **R**'s payoff is given by a function $v : \mathcal{T} \times \mathcal{S} \times \mathcal{A} \rightarrow \mathbb{R}$.



6.4.2 Nash Equilibrium

Definition 6.47 (Strategy)

A **behavior strategy** for **S** is given by a function $\sigma : \mathcal{T} \times \mathcal{S} \rightarrow [0, 1]$ such that $\sum_s \sigma(t, s)$ for each t .

A **behavior strategy** for **R** is given by a function $\alpha : \mathcal{S} \times \mathcal{A} \rightarrow [0, 1]$ such that $\sum_a \alpha(s, a)$ for each s .



Definition 6.48 (Nash Equilibrium)

Behavior strategies α and σ form a **Nash equilibrium** if and only if

1. For all $t \in \mathcal{T}$,

$$\sigma(t, s) > 0 \text{ implies } \sum_a \alpha(s, a)u(t, s, a) = \max_{s' \in \mathcal{S}} (\sum_a \alpha(s', a)u(t, s', a))$$

2. For each $s \in \mathcal{S}$ such that $\sum_t \sigma(t, s)\rho(t) > 0$,

$$\alpha(s, a) > 0 \text{ implies } \sum_t \mu(t; s)v(t, s, a) = \max_{a'} \sum_t \mu(t; s)v(t, s, a')$$

where $\mu(t; s)$ is the **R**'s posterior belief about t given s , $\mu(t; s) = \frac{\sigma(t, s)\rho(t)}{\sum_{t'} \sigma(t', s)\rho(t')}$ if $\sum_t \sigma(t, s)\rho(t) > 0$ and $\mu(t; s) = 0$ otherwise.



Definition 6.49 (Separating & Pooling Equilibrium)

An equilibrium (σ, α) is called a **separating** equilibrium if each type t sends different signals; i.e., the set \mathcal{S} can be partitioned into (disjoint) sets $\{\mathcal{S}_t; t \in \mathcal{T}\}$ such that $\sigma(t, \mathcal{S}_t) = 1$. An equilibrium (σ, α) is

called a **pooling** equilibrium if there is a single signal s^* that is sent by all types; i.e., $\sigma(t, s^*) = 1$ for all $t \in \mathcal{T}$.



6.4.3 Single-crossing

6.4.3.1 Situation over real line

Consider the situation that $\mathcal{T}, \mathcal{S}, \mathcal{A} \subseteq \mathbb{R}$ and \geq is the usual "greater than or equal to" relationship.

1. We let $\Delta\mathcal{A}$ denote the set of probability distributions on \mathcal{A} .
2. For each $s \in \mathcal{S}$ and $\mathcal{T}' \subseteq \mathcal{T}$, we let $\Delta\mathcal{A}(s, \mathcal{T}')$ be the set of mixed strategies that are the best responses by \mathbf{R} to $s \in \mathcal{S}$ for some probability distribution with support \mathcal{T}' .
3. For $\alpha \in \Delta\mathcal{A}$, we write $u(t, s, \alpha) \triangleq \sum_{a \in \mathcal{A}} u(t, s, a)\alpha(a)$.

Definition 6.50 (Single-crossing)

The data of the game are said to satisfy the **single-crossing property** if the following holds: If $t \in \mathcal{T}$, $(s, \alpha) \in \mathcal{S} \times \Delta\mathcal{A}$ and $(s', \alpha') \in \mathcal{S} \times \Delta\mathcal{A}$ are such that $\alpha \in \Delta\mathcal{A}(s, \mathcal{T})$, $\alpha' \in \Delta\mathcal{A}(s', \mathcal{T})$, $s > s'$ and $u(t, s, \alpha) \geq u(t, s', \alpha')$, then for all $t' \in \mathcal{T}$ such that $t' > t$, $u(t', s, \alpha) \geq u(t', s', \alpha')$.



6.5 Adverse Selection

Consider a labor market that has many identical firms. In competitive equilibrium, firms' profits are 0. Firms are price (wage) takers, risk-neutral, and CRS. There are continuum of workers with productivity levels $\theta \in [\underline{\theta}, \bar{\theta}]$ (Assume workers work if it is indifferent for them between employment and non-employment).

1. $\theta \sim F$, $F(\cdot)$ is a c.d.f. over $[\underline{\theta}, \bar{\theta}]$.
2. N is the total mass of workers.
3. Type θ worker has a reservation utility $r(\theta)$.
 - o Suppose the competitive equilibrium wages are $\theta = w^*(\theta)$.
 - o An allocation is denoted by $I : [\underline{\theta}, \bar{\theta}] \rightarrow \{0, 1\}$, where $I(\theta) = 0$ denotes θ is unemployed and $I(\theta) = 1$ denotes θ is employed.
 - o Aggregate welfare = sum of utilities of all participants

$$= N \int_{\underline{\theta}}^{\bar{\theta}} [I(\theta) \times \theta + [1 - I(\theta)]r(\theta)] dF(\theta)$$

Then we have the optimal allocation satisfies

$$I^*(\theta) = \begin{cases} 1, & \theta > r(\theta) \\ 0, & \theta = r(\theta) \\ 0, & \theta < r(\theta) \end{cases}$$

In the asymmetric information case,

Definition 6.51

w is CE wage if $w = \mathbb{E}[\theta|r(\theta) \leq w]$.



6.5.1 Adverse Selection

Assumption

- (A1). r is strictly increasing in θ .
- (A2). $F(\cdot)$ has a strictly positive density, $F(\theta) > 0, \forall \theta \in [\underline{\theta}, \bar{\theta}]$.
- (A3). $r(\theta) \leq \theta$ (outside option is worse than productivity, i.e., full employment is optimal).

Lemma 6.7

Under A1-A3, $\Phi(w) := \mathbb{E}[\theta|r(\theta) \leq w]$ is well-defined, continuous, and non-decreasing.



Hence, there exists underemployment, which makes 1st welfare theorem fails. There may exist multiple CEs, where the one with the highest wage Pareto dominates others.

Example 6.7 Suppose $\theta \in [0, 2]$, $F(\theta) = \frac{\theta}{2}$, $f(\theta) = \frac{1}{2}$, $r(\theta) = \alpha\theta, \alpha \in (0, 1)$.

$$\mathbb{E}[\theta|r(\theta) \leq w] = \mathbb{E}\left[\theta|\theta \leq \frac{w}{\alpha}\right] = \begin{cases} 1, & w \geq 2\alpha \\ \frac{1}{F(\frac{w}{\alpha})} \int_0^{\frac{w}{\alpha}} \theta f(\theta) d\theta = \frac{w}{2\alpha}, & w \leq 2\alpha \end{cases}$$

CEs are given by $\mathbb{E}[\theta|r(\theta) \leq w] = w$. $w^* = 0$ is always CE and $w^* = 1$ is CE if $\alpha \leq \frac{1}{2}$.

6.5.2 Game Theoretical Approach

1. Suppose there are two firms setting wages simultaneously.
2. Workers observe the wages in stage 1 and make an employment decision.

Let W^* be the set of CE wages and $w^* := \max W^*$.

Lemma 6.8

$\forall w' \in (w^*, \bar{\theta}]: \mathbb{E}[\theta|r(\theta) \leq w'] < w'$.



Proof 6.12

Suppose by the contradiction that $\exists w' \in (w^*, \bar{\theta}]$ s.t. $\mathbb{E}[\theta|r(\theta) \leq w'] \geq w'$. Since $\mathbb{E}[\theta|r(\theta) \leq \bar{\theta}] < \bar{\theta}$, there must exist a $w'' \in [w', \bar{\theta})$ s.t. $\mathbb{E}[\theta|r(\theta) \leq w''] = w''$ by intermediate value theorem, which contradicts to the definition of w^* .

Proposition 6.17

- (i). If $w^* > r(\underline{\theta})$ and $\exists \epsilon > 0$ s.t. $\mathbb{E}[\theta|r(\theta) \leq w'] > w'$, $\forall w' \in (w^* - \epsilon, w^*)$. Then, there is a unique SPE where both firms set wage = w^* .
- (ii). If $w^* = r(\underline{\theta})$ (complete market shutdown at w^*), there are multiple SPE that all give the same outcome as complete market shutdown where both firms set wage = w^* .

**Proof 6.13****Lemma 6.9**

In all SPE, firms make zero profits.

**Proof 6.14**

Suppose not, i.e., at least one firm makes strictly positive profits. Then, the total profits of firms 1&2,

$$\Pi = M(\bar{w}) [\mathbb{E}[\theta|r(\theta) \leq \bar{w}] - \bar{w}] > 0$$

where \bar{w} is the max wage set by the two firms and $M(\bar{w})$ is the mass of workers willing to work at \bar{w} . At least one firm, i , makes profit $\leq \frac{\Pi}{2}$. Then, i 's profits from setting $\bar{w} + \delta$, with $\delta \rightarrow 0^+$, is higher:

$$\begin{aligned} & M(\bar{w} + \delta) [\mathbb{E}[\theta|r(\theta) \leq \bar{w} + \delta] - \bar{w} - \delta] \\ & \geq M(\bar{w}) [\mathbb{E}[\theta|r(\theta) \leq \bar{w} + \delta] - \bar{w} - \delta] \rightarrow \Pi \text{ as } \delta \rightarrow 0 \end{aligned}$$

Hence, the i has incentive to deviate.

Lemma 6.10

In all SPE, firm i sets $w_i \leq w^*$, $i \in \{1, 2\}$.

**Proof 6.15**

Directly given by Lemma 6.8 and Lemma 6.9.

- (i): In SPE, no firm i sets $w_i < w^*$: suppose $w_i < w^*$ and let $j \neq i$, take any w'_j s.t. $w'_j \in (w_i, w^*)$ and $w'_j > w^* - \epsilon$. Then, j gets profit: $M(w'_j) [\mathbb{E}[\theta|r(\theta) \leq w'_j] - w'_j] > 0$ (by Case (i)'s conditions).
- (ii): By Lemma 6.10, both firms set $w_i \leq w^* = r(\underline{\theta})$. Check that $\{(w_1, w_2) : w_1, w_2 \leq w^*\}$ is SPE

wage profiles.

6.5.3 Planner Intervention

Planner can't observe the true type θ .

The planner's tools:

1. Take over the firms.
2. w_e , employment wage.
3. w_u , unemployment wage.

s.t. budget balanced.

Definition 6.52 (Constrained Efficient)

A CE w is **constrained efficient** if it cannot be Pareto improved upon by an intervention by the planner. ♣

Proposition 6.18 ($w^* := \max W^*$ is constrained efficient)

Let W^* be the set of CE wages. $w^* := \max W^*$ is constrained efficient. ♠

Proof 6.16

Note that both firms are making zero profits by the Lemma 6.9. Any CE wage $w \neq w^*$ can be Pareto improved by $\{w_e = w^*, w_u = 0\}$. Then, we prove w^* can't be Pareto improved.

1. Case 1: if w^* gives full-employment in CE, then w^* is Pareto efficient.
2. Case 1: suppose w^* doesn't give full-employment in CE.

Consider taking an intervention $w_e \& w_u$. Then, $\{\theta : r(\theta) + w_u \leq w_e\} = [\underline{\theta}, \hat{\theta}]$ for some $\hat{\theta} \in [\underline{\theta}, \bar{\theta}]$ such that

$$r(\hat{\theta}) + w_u = w_e \quad (6.1)$$

The budget balanced gives

$$w_e F(\hat{\theta}) + w_u (1 - F(\hat{\theta})) = \int_{\underline{\theta}}^{\hat{\theta}} \theta dF(\theta) \quad (6.2)$$

Plug (6.1) into (6.2):

$$\begin{cases} w_u(\hat{\theta}) = \int_{\underline{\theta}}^{\hat{\theta}} \theta dF(\theta) - r(\hat{\theta})F(\hat{\theta}) = F(\hat{\theta}) (\mathbb{E}[\theta | \theta \leq \hat{\theta}] - r(\hat{\theta})) \\ w_e(\hat{\theta}) = \int_{\underline{\theta}}^{\hat{\theta}} \theta dF(\theta) + r(\hat{\theta})(1 - F(\hat{\theta})) \end{cases}$$

Let θ^* be s.t. $r(\theta^*) = w^*$. Because w^* is a CE price, $\mathbb{E}[\theta | \theta \leq \theta^*] = r(\theta^*) = w^*$. So, CE with w^* can be implemented by $w_u(\theta^*) = 0$ and $w_e(\theta^*) = w^*$.

- (a). If $\hat{\theta} < \theta^*$. $\underline{\theta}$ is worse off under the intervention since $w_e(\hat{\theta}) < w^*$.

(b). If $\hat{\theta} > \theta^*$. $\bar{\theta}$ is worse off under the intervention since $w_u(\hat{\theta}) = F(\hat{\theta}) (\mathbb{E}[\theta | \theta \leq \hat{\theta}] - r(\hat{\theta})) < 0$ by the Lemma 6.8

6.5.4 Signaling

Suppose the worker $\theta \in [\underline{\theta}, \bar{\theta}]$ can properly and costlessly reveal his type to the firms. Then,

Lemma 6.11

All workers reveal their types.



Spence's Job Market Signaling Model One worker has productivity $\theta \in \{\theta_L, \theta_H\}$ with $P(\theta_H) = \lambda$. The worker signal through his education with cost $e > 0$. The education doesn't change his productivity. The payoff of the worker is the wage minus the cost:

$$u(w, e | \theta) = w - c(e, \theta)$$

where $c(0, \theta) = 0$, $c_e(e, \theta) := \frac{\partial c(e, \theta)}{\partial e} > 0$, $c_\theta(e, \theta) := \frac{\partial c(e, \theta)}{\partial \theta} < 0$, and $c_{e\theta}(e, \theta) := \frac{\partial^2 c(e, \theta)}{\partial e \partial \theta} < 0$ (Single-Crossing Property, the difference $c(e, \theta_L) - c(e, \theta_H)$ is increasing in e (i.e., $c_e(e, \theta_L) - c_e(e, \theta_H) > 0$), which means if $c(e, \theta_L)$ and $c(e, \theta_H)$ intersect as functions of e , they only intersect at one time.)

1. Stage 0: Nature chooses the $\theta \in \{\theta_L, \theta_H\}$ with $P(\theta_H) = \lambda$.
2. Stage 1: The worker learns θ and chooses $e(\theta) \geq 0$.
3. Stage 2: Firms observe $e(\theta)$. Then, they simultaneously make wage offers w_1 and w_2 .
4. Stage 3: The worker observes w_1, w_2 and makes employment decision.

Let $r(\theta_L)$ and $r(\theta_H) = 0$. Let $\mu(e) \in [0, 1]$ be the probability that in the beginning of stage 2, firms think that the worker is θ_H type with probability $\mu(e)$ when observing e . The corresponding expected productivity (the highest wage) that the firm can pay is

$$w(e) = \mu(e)\theta_H + (1 - \mu(e))\theta_L$$

In stage 2, both firm will set $w(e)$ (complete competition).

Definition 6.53 (Perfect Bayesian Equilibrium)

A PBE is a strategy profile $(e^*(\theta_L), e^*(\theta_H), w_1^* : \mathbb{R}_+ \rightarrow \mathbb{R}, w_2^* : \mathbb{R}_+ \rightarrow \mathbb{R})$, and a belief $\mu^* : \mathbb{R} \rightarrow [0, 1]$ such that

1. $\forall \theta \in \{\theta_L, \theta_H\}$, the worker strategy optimal given firm strategies.
2. The belief $\mu^*(e)$ is derived from $\lambda, e^*(\theta_L), e^*(\theta_H)$ via Bayes' rule whenever possibly (on the equilibrium path). Outside the equilibrium path the belief $\mu^*(e)$ is arbitrarily.
3. Firms offer wages that form a NE of the stage 2 game, where their belief $\mu^*(e)$ about their workers'



type. (sequential rationality).

We simplify the game by backward induction:

1. Stage 3: The worker chooses the highest wage off if it is ≥ 0 .
2. Stage 2: After observing $e(\theta)$, firms chooses the wage as the expected productivity in NE,

$$w^*(e) = \mu^*(e)\theta_H + (1 - \mu^*(e))\theta_L$$

because it is a Bertrand competition.

Separating Equilibrium In separating equilibrium, $e^*(\theta_L) \neq e^*(\theta_H)$.

Lemma 6.12

In any separating PBE, $w^*(e^*(\theta)) = \theta, \forall \theta \in \{\theta_L, \theta_H\}$.



Proof 6.17

By Bayes' rule, after firm observe $e^*(\theta_L)$, $\mu^*(e^*(\theta_L)) = 0$. Then, $w^*(e^*(\theta_L)) = \theta_L$. ($e^*(\theta_H)$ is similar.)



Lemma 6.13

In separating PBE, low type always chooses zero education, $\theta^*(\theta_L) = 0$.



Proof 6.18

If not, the low type worker always has profitable deviation, $\theta^*(\theta_L) = 0$.

Lemma 6.14

Define \underline{e} and \bar{e} such that

1. $\theta_L = \theta_H - c(\underline{e}, \theta_L)$ (the lowest effort can prevent the low type from mimicking high type) and
2. $\theta_L = \theta_H - c(\bar{e}, \theta_H)$ (the highest effort can prevent the high type from pooling with low type).

Then, in all separating PBES, $e \in [\underline{e}, \bar{e}]$.

Conversely, $\forall \hat{e} \in [\underline{e}, \bar{e}]$, there is a separating PBE where $e^*(\theta_H) = \hat{e}$.



These different PBES are Pareto ranked. High type prefers the PBE with a lower e (the best is the one with $e^*(\theta_H) = \underline{e}$.)

Pooling PBE $e^*(\theta) = e^*, \theta \in \{\theta_L, \theta_H\}, \mu^*(e^*) = \lambda$, and $w^*(e^*) = \mathbb{E}[\theta]$.

Lemma 6.15

Define e' by $\theta_L = \mathbb{E}[\theta] - c(e', \theta_L)$ (the highest effort can prevent the low type from choosing $e = 0$ and get $w = \theta_L$.)

Then, for all pooling PBE, $e^*(\theta_L) = e^*(\theta_H) = e^* \in [0, e']$. Conversely, for all $\hat{e} \in [0, e']$, there is a

pooling PBE with $e^* = \hat{e}$.



6.5.5 Cho-Kreps Intuitive Criterion

Definition 6.54 (Equilibrium Dominated Message)

A message is **equilibrium dominated** for a type if the type must do strictly worse by sending the message than it does in equilibrium (i.e., payoff in eq. is strictly better than maximum payoff from deviating).



Definition 6.55 (Cho-Kreps Intuitive Criterion)

If an information set is off the eq. path and a message is eq. dominated for a type, then beliefs should assign zero probability to the message coming from that type (if possible).



Fix a PBE $e^*(\theta), \theta \in \{\theta_L, \theta_H\}, \mu^*(\cdot)$ (We know $w_1^*(e) = w_2^*(e) = \mu(e)\theta_H + (1 - \mu(e))\theta_L$). Let $u^*(\theta), \theta \in \{\theta_L, \theta_H\}$ be the PBE utility of the type θ worker.

The criterion requires the (off-path) belief $\mu^*(e) := P(\tilde{\theta} = \theta_H|e) = 1 - P(\tilde{\theta} = \theta_L|e)$ satisfies

$$P(\tilde{\theta} = \theta|e) = 0, \forall e, \theta$$

such that

1. $u^*(\theta) > \max_{w \in [\underline{\theta}, \bar{\theta}]} [w - c(e, \theta)]$
2. $\exists \theta' \text{ s.t. } u^*(\theta') \leq \max_{w \in [\underline{\theta}, \bar{\theta}]} [w - c(e, \theta')]$ (make sure the sum of beliefs given e is nonzero.)

In this application, the only PBE that survives Intuitive Criterion is the best separating PBE, $e^*(\theta_H) = \underline{e}$ (the lowest effort).

6.5.6 Screening Model

Workers can undertake a contractible/observable task level $t \geq 0$. The utility of a worker is defined by $u(w, t, \theta) := w - c(t, \theta)$, where $c(\cdot, \cdot)$ satisfies the same assumption as in signaling model 6.5.4.

The Game follows

1. Stage 1: Two firms simultaneously determine sets of contracts, (w, t) .
2. Stage 2: The worker observes all offer contracts and makes employment decision. (If indifference, choose lower task contract, favor employment over unemployment. If contracts of firms are indifferent, choose each with probability 1/2.)

The null contract is $(w, t) = (0, 0)$. Assume WLOG at stage 1, each firm appears a non-empty set of contracts.

Perfect Information

Proposition 6.19 (Perfect Information)

If firms can observe the worker types, then in SPE firms make zero profit and type θ_i worker signs $(w_i^*, t_i^*) = (\theta_i, 0)$.



Proof 6.19

Claim 6.3

Firms make zero profits from this contract.



Proof 6.20

Suppose not,

- $w_i^* > \theta_i \Rightarrow$ negative profits, firms benefit from offering null contract.
- $w_i^* < \theta_i \Rightarrow$ Let Π be the total profits of the firms. Then one of the firms makes profit $\leq \frac{\Pi}{2}$.

Then, this firm can benefit from offering $(w_i^* + \Delta, t_i^*)$, where $\Delta \rightarrow 0^+$.

Then, we prove the firms must choose $(w_i^*, t_i^*) = (\theta_i, 0)$. Suppose by the way of contradiction that $t_i^* > 0$.

Then, one firm can profitably deviate by offering $(w_i^*, 0)$.

Asymmetric Information

Lemma 6.16

In any SPE, firms obtain zero profits,



Proof 6.21

Firms must make profits ≥ 0 . Suppose by the way of contradiction that the total profit $\Pi > 0$. Let (w_L, t_L) be the contract signed by θ_L and (w_H, t_H) be the contract signed by θ_H . One firm can profitably deviate by offering $(w_L + \Delta, t_L)$ and $(w_H + \Delta, t_H)$, where $\Delta \in (0, \Pi)$.

Lemma 6.17

There is no pooling SPE.



Proof 6.22

Suppose for a contradiction, \exists an SPE where both worker types sign $(w_p = \mathbb{E}[\theta], t_p)$. Suppose one firm offers (w_p, t_p) , then another firm can only employ high type workers by offering (\tilde{w}, \tilde{t}) , where $\tilde{w} - c(\tilde{t}, \theta_H) > \mathbb{E}[\theta] - c(t_p, \theta_H)$, $\tilde{w} - c(\tilde{t}, \theta_L) < \mathbb{E}[\theta] - c(t_p, \theta_L)$, and $\tilde{w} < \theta_H$. (The existence is given by $\frac{\partial^2 c(t, \theta)}{\partial t \partial \theta} < 0$.)

Lemma 6.18

Let (w_L, t_L) be the contract signed by θ_L and (w_H, t_H) be the contract signed by θ_H in separating SPE.

Then, $w_L = \theta_L$ and $w_H = \theta_H$.

**Proof 6.23**

Suppose $w_i > \theta_i, i \in \{L, H\}$, firms benefit from not offering this contract. So, $w_L \leq \theta_L$ and $w_H \leq \theta_H$.

1. $w_L = \theta_L$: Suppose $w_L < \theta_L$. Either firm can profitably deviate by setting (w'_L, t_L) such that $w_L < w'_L < \theta_L$. This offer can win all low-type workers and get a positive profit from hiring them. If $w'_L - c(t_L, \theta_H) \geq w_H - c(t_H, \theta_H)$, the offer can also hire high-type workers, which also give positive profit for the firm. Hence, there is a contradiction.
2. $w_H = \theta_H$: Suppose $w_H < \theta_H$, firms get positive profits, which contradicts to the Lemma 6.16.

Lemma 6.19

θ_L signs the contract $(\theta_L, 0)$ in SPE.

**Proof 6.24**

Suppose $t_L > 0$. One firm can profitably deviate by offering $(\theta_L - \Delta, 0)$.

Proposition 6.20

In any (pure strategy) SPE, θ_L signs $(w_L, t_L) = (\theta_L, 0)$ and θ_H signs $(w_H, t_H) = (\theta_H, t_H)$, where t_H solves

$$\theta_H - c(t_H, \theta_L) = \theta_L$$



If $\lambda := P(\theta_H)$ is high, the pure SPE may not exist (exist (\tilde{w}, \tilde{t}) can attract both types and make positive profit).

Cross subsidizing deviation by a firm (prices one product above its market value to fund another product), (\tilde{w}, \tilde{t}) (signed by low type) and $(\tilde{\tilde{w}}, \tilde{\tilde{t}})$ (signed by high type), is a profitable deviation if λ is large enough.

Chapter 7 Mechanism Design

7.1 Mechanism Design

Design incentives for agents to reveal their types or achieve particular society outcomes.

Given

1. the set of agents I with utility function $u_i(x; \theta_i), i \in I$,
2. alternatives (outcomes for the society) X ,
3. types (of agents) $\Theta = (\Theta_1, \dots, \Theta_I)$ with prior probability ϕ over Θ ,
4. and a **social choice function** (SCF) $f : \Theta \rightarrow X$.

Definition 7.1 (Mechanism $\Gamma = (S, g)$)

A **mechanism** is represented as

$$\Gamma = (S, g)$$

where $S \triangleq (S_1, \dots, S_I)(S_1, \dots, S_I)$ represents the set of strategies, S_i represents the strategy set of agent i , and $g : S \triangleq (S_1, \dots, S_I) \rightarrow X$ is the outcome function that determines the social outcome. 

A **Bayesian game induced by Γ** is $(I, S, \Theta, \phi, \tilde{u})$, where the payoffs functions are

$$\tilde{u}_i(s; \theta_i) = u_i(g(s); \theta_i)$$

for all $i \in I$, $s \in S$, and $\theta_i \in \Theta_i$.

7.1.1 Implement in Dominant Strategies

Definition 7.2 (Γ Implements f)

A mechanism Γ (indirectly) **implements** a social choice function (SCF) f if there exists an “equilibrium”

$s^*(\cdot) = (s_1^*(\cdot), \dots, s_I^*(\cdot))$ of the Bayesian game induced by Γ such that

$$g(s_1^*(\theta_1), \dots, s_I^*(\theta_I)) = f(\theta_1, \dots, \theta_I)$$

for all $(\theta_1, \dots, \theta_I) \in \Theta$. Here the “equilibrium” is a dominant strategy equilibrium or BNE. 

That is, the equilibrium in a game induced by Γ gives the same outcome as the outcome of f given by revealing agents’ true types.

Definition 7.3 (Direct Mechanism)

A mechanism is **direct mechanism** if $S_i = \Theta_i$ for all $i \in I$ and $g(\theta) = f(\theta)$ for all $\theta = (\theta_1, \dots, \theta_I) \in \Theta$.

So, a direct mechanism can be represented by $\Gamma = (\Theta, f(\cdot))$. 

A strategy is weakly dominant if for all $\theta_i \in \Theta_i$ and all $s_{-i}(\cdot) \in S_{-i}$, we have $u_i(g(s_i(\theta_i), s_{-i}), \theta_i) \geq u_i(g(s'_i, s_{-i}), \theta_i)$ for all $s'_i \neq s_i$.

Definition 7.4 (Dominant Strategy Equilibrium)

Strategy profile $s^* = (s_1^*(\cdot), \dots, s_I^*(\cdot))$ is a **dominant strategy (D-S) equilibrium** of $\Gamma = (S, g(\cdot))$ if for all $i \in I$ and $\theta_i \in \Theta$, we have, for all $s'_i \in S_i$ and $s_{-i} \in S_{-i}$:

$$u_i(g(s_i^*(\theta_i), s_{-i}), \theta_i) \geq u_i(g(s'_i, s_{-i}), \theta_i)$$

equivalently, in the Bayesian game induced by Γ ,

$$\tilde{u}_i(s_i^*(\theta_i), s_{-i}, \theta_i) \geq \tilde{u}_i(s'_i, s_{-i}, \theta_i) $$

Definition 7.5 (Implement in dominant strategies)

Γ **implements** f in **dominant strategies** if \exists a dominant strategy (D-S) equilibrium s^* of Γ such that $g(s^*(\theta)) = f(\theta)$. 

7.1.2 Dominant-Strategy-Incentive-Compatible (DSIC)/Strategy-Proof**Definition 7.6 (Strategy-Proof, DSIC)**

f is **strategy-proof** (also called dominant-strategy-incentive-compatible, **DSIC**) if

$$s_i^*(\theta_i) = \theta_i, \quad \forall \theta_i \in \Theta_i, i \in I$$

is a dominant strategy (D-S) equilibrium of the direct mechanism $\Gamma = (\Theta, f(\cdot))$. 

Theorem 7.1 (Revelation Principle)

If \exists a mechanism $\Gamma = (S, g(\cdot))$ that implements f in dominant strategies. Then f is strategy-proof (DSIC). 

Proof 7.1

Given Γ implements f in dominant strategies, there is a D-S equilibrium $s^* = (s_1^*(\cdot), \dots, s_I^*(\cdot))$ such that $g(s^*(\theta)) = f(\theta)$.

By the definition of D-S equilibrium,

$$u_i(g(s_i^*(\theta_i), s_{-i}), \theta_i) \geq u_i(g(s'_i, s_{-i}), \theta_i)$$

By substituting $g(s^*(\theta)) = f(\theta)$, we have

$$u_i(f(\theta_i, \theta_{-i}), \theta_i) \geq u_i(f(\theta'_i, \theta_{-i}), \theta_i), \quad \forall \theta'_i \in \Theta_i$$

which gives that f is DSIC.

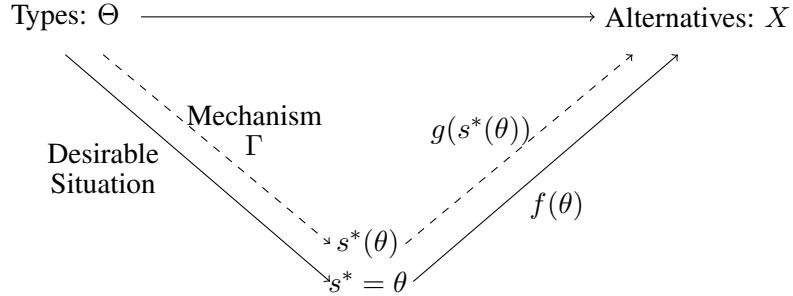


Figure 7.1: How Mechanism Design works

7.1.3 Bayesian-Incentive-Compatible (BIC)

Definition 7.7 (BIC)

f is Bayesian-incentive-compatible (B.I.C.) if

$$s_i^*(\theta_i) = \theta_i, \quad \forall \theta_i \in \Theta_i, i \in I$$

is a BNE of the Bayesian game induced by the direct mechanism $\Gamma = (\Theta, f(\cdot))$.



Theorem 7.2 (Revelation Principle (BIC))

If \exists a mechanism $\Gamma = (S, g(\cdot))$ that implements f in BNE. Then f is BIC.



BIC is a weaker condition than DSIC.

7.1.4 Negative Results: dictatorial SCF f

Theorem 7.3 (Gibbard-Satterthwaite Theorem)

Suppose that $|X| \geq 3$ and a social choice function f is surjective (i.e., $\forall x \in X \exists(\theta_1, \dots, \theta_I) \in \Theta$ s.t. $f(\theta_1, \dots, \theta_I) = x$). Then, f is strategy-proof (DSIC) $\Leftrightarrow f$ is dictatorial (3.14, i.e., $\exists i^* \in \{1, \dots, I\}$ such that $f(\theta) \in \operatorname{argmax}_{x \in X} u_{i^*}(x; \theta_{i^*})$ for all $\theta \in \Theta$).



Note: By the revelation principle, under the conditions of the Theorem, there is no mechanism that implements a non-dictatorial SCF f in dominant strategies.

7.2 Quasi-linear Model

Consider $x = (k, \underbrace{t_1, \dots, t_I}_{t}) \in X = K \times \mathbb{R}^I$, in our example, K represents a set of choices for projects and \mathbb{R}^I represents the set of transfers for all agents.

Each agent has a quasi-linear function that represents her utility:

$$u_i(k, t, \theta_i) = v_i(k, \theta_i) + t_i$$

where $v : K \times \Theta \rightarrow \mathbb{R}$ represents the utility without transfers.

Let $p(\cdot) = (k(\cdot), t(\cdot))$ represents the “project-choice rule” $k : \Theta \rightarrow K$ and the “transfer rule” $t : \Theta \rightarrow \mathbb{R}^I$.

Definition 7.8 (ex-post efficient)

$k(\cdot) : \Theta \rightarrow K$ is **ex-post efficient** if $\nexists (\theta \in \Theta, k' \in K, t = (t_1, \dots, t_I) \in \mathbb{R}^I)$ such that

$$(1). \sum_{i=1}^I t_i = 0$$

$$(2). v_i(k', \theta_i) + t_i > v_i(k(\theta), \theta_i), \forall i \in I$$

i.e., we can't get a higher total social welfare. (Because of the transfers, a higher social welfare can make everyone better off.)



Proposition 7.1 (ex-post efficient \Leftrightarrow maximizing the sum of utilities)

\forall project-choice rule $k(\cdot)$, $k(\cdot)$ is ex-post efficient if and only if $k(\cdot)$ maximizes the sum of utilities, i.e.,

$\forall \theta \in \Theta$ and $\forall k' \in K$,

$$\sum_{i=1}^I v_i(k(\theta), \theta_i) \geq \sum_{i=1}^I v_i(k', \theta_i)$$



Proof 7.2

“ \Leftarrow ”: Suppose by the way of contradiction that there exists (θ, k', t) such that $\sum_{i=1}^I t_i = 0$ and $v_i(k', \theta_i) + t_i > v_i(k(\theta), \theta_i), \forall i \in I$. Sum together, there is a contradiction.

“ \Rightarrow ”: Suppose by the way of contradiction that there exists (θ, k') , $\sum_{i=1}^I v_i(k(\theta), \theta_i) < \sum_{i=1}^I v_i(k', \theta_i)$.

Then, we can define a t such that satisfies $\sum_{i=1}^I t_i = 0$ and $v_i(k', \theta_i) + t_i > v_i(k(\theta), \theta_i), \forall i \in I$.

Let $\Delta = \sum_{i=1}^I v_i(k', \theta_i) - \sum_{i=1}^I v_i(k(\theta), \theta_i)$, then $t_i = v_i(k(\theta), \theta_i) - v_i(k', \theta_i) + \frac{\Delta}{I}, \forall i \in I$ is the transfer-choice we want.

7.2.1 Vickrey-Clarke-Groves Mechanism

Proposition 7.2 (VCG Mechanism)

Suppose $k^*(\cdot)$ is ex-post efficient project choice rule. For each $i \in \{1, \dots, I\}$, let $h_i : \Theta_{-i} \rightarrow \mathbb{R}$ be an arbitrary function.

Define the transfer rule $t(\cdot)$ as follows

$$t_i(\theta_i, \theta_{-i}) = \sum_{j \neq i} v_j(k^*(\theta_i, \theta_{-i}), \theta_j) + h_i(\theta_{-i})$$

Then the SCF $f(\cdot) = (k^*(\cdot), t(\cdot))$ is DSIC. ♠

Proof 7.3

Take any $i, \theta \in \Theta$ and let $\hat{\theta}_i \in \Theta_i$. Reporting truthfully gives higher profits than misreporting $\hat{\theta}_i$:

$$\begin{aligned} v_i(k^*(\theta_i, \theta_{-i}), \theta_i) + t_i(\theta_i, \theta_{-i}) &= v_i(k^*(\theta_i, \theta_{-i}), \theta_i) + \sum_{j \neq i} v_j(k^*(\theta), \theta_j) + h_i(\theta_{-i}) \\ &= \sum_{j=1}^I v_j(k^*(\theta), \theta_j) + h_i(\theta_{-i}) \\ &\geq \sum_{j=1}^I v_j(k^*(\hat{\theta}_i, \theta_{-i}), \theta_j) + h_i(\theta_{-i}) \\ &= v_i(k^*(\hat{\theta}_i, \theta_{-i}), \theta_i) + t_i(\hat{\theta}_i, \theta_{-i}) \end{aligned}$$

Hence, VCG mechanism with SCF $f(\cdot) = (k^*(\cdot), t(\cdot))$ is DSIC.

Definition 7.9 (Pivotal VCG Mechanism (Special Case))

Let $h_i(\theta_{-i}) = \max_{k \in K} \sum_{j \neq i} v_j(k, \theta_j)$.

$$t_i(\theta_i, \theta_{-i}) = \sum_{j \neq i} v_j(k^*(\theta_i, \theta_{-i}), \theta_j) - \max_{k \in K} \sum_{j \neq i} v_j(k, \theta_j) \leq 0$$

1. i is **pivotal** if $k^*(\theta)$ doesn't maximize $\max_{k \in K} \sum_{j \neq i} v_j(k, \theta_j)$.
2. i is **not pivotal** if $k^*(\theta)$ maximizes $\max_{k \in K} \sum_{j \neq i} v_j(k, \theta_j)$.

💡 Note i is **not pivotal** $\Rightarrow t_i(\theta) = 0$. ♣

Example 7.1 Suppose $k \in \{0, 1\}, \theta \in \Theta \subset \mathbb{R}^I, v_i(k, \theta_i) = k\theta_i$. Since $\sum_{i=1}^I v_i(k, \theta_i) = k \sum_{i=1}^I \theta_i$, $k^*(\cdot)$ is ex-post efficient: $k^*(\theta) = 1 \Leftrightarrow \sum_{i=1}^I \theta_i \geq 0$. The pivotal VCG transfers:

$$t_i(\theta) = \begin{cases} \sum_{j \neq i} \theta_j - 0 & \text{if } \sum_{j=1}^I \theta_j \geq 0 > \sum_{j \neq i} \theta_j \\ 0 - \sum_{j \neq i} \theta_j & \text{if } \sum_{j=1}^I \theta_j < 0 \leq \sum_{j \neq i} \theta_j \\ 0 & \text{otherwise} \end{cases}$$

Example 7.2 (Second Price Auction) One indivisible object to be allocated to one of $1, \dots, I$. Social decision is deciding who gets the object, $K = \{1, \dots, I\}$, $k = i$ means “ i receives the object”. $\Theta_i \subseteq \mathbb{R}_+, \theta_i \in \Theta_i$

denotes i 's valuation for the object $v_i(k, \theta_i) = \begin{cases} \theta_i, & \text{if } k = i \\ 0, & \text{if } k \neq i \end{cases}$. Ex-post efficient $k^*(\cdot)$ allocates the object to the individual with the highest valuation. The pivotal VCG transfers:

$$t_i(\cdot) = \begin{cases} 0 - \theta^{(2)}(\text{the second highest}), & \text{if } k^*(\theta) = i \text{ (} i \text{ is pivotal)} \\ 0 = \theta^{(1)} - \theta^{(1)}, & \text{if } k^*(\theta) \neq i \end{cases}$$

Example 7.3 (Uniform-Price Auction) m -identical indivisible objects ($m < I$). Each agent can consume 0 or 1 object. $K = \{M \subset \{1, \dots, I\} \mid |M| = m\}$, where $k = M$ is the set of agents who receive an object.

$v_i(k, \theta_i) = \begin{cases} \theta_i, & i \in k \\ 0, & i \notin k \end{cases}$. The ex-post efficient $k^*(\cdot)$ allocates the objects to top m -valuation agents. The pivotal VCG transfers:

$$t_i(\theta) = \begin{cases} \left(\sum_{j=1}^m \theta_{(j)} - \theta_i\right) - \left(\sum_{j=1}^{m+1} \theta_{(j)} - \theta_i\right) = -\theta^{(m+1)} & , i \in k^*(\theta) \\ 0 & , i \notin k^*(\theta) \end{cases}$$

Example 7.4 (Package Auction) 2 identical indivisible objects to be allocated $I = 3$ agents. Each agent can consume 0, 1, or 2 units. $K = \{(1, 1, 0), (1, 0, 1), (0, 1, 1), (2, 0, 0), (0, 2, 0), (0, 0, 2)\}$. $\Theta_i = \{\theta_i = (v_1, v_2) \in \mathbb{R}_+^2 \mid v_2 \geq v_1 \geq 0\}$. Consider an example,

	θ_1	θ_2	θ_3
v_1	3	4	1
v_2	4	5	6

The ex-post efficient $k^*(\theta) = (1, 1, 0)$. Then,

$$t_1(\theta) = 4 - 6 = -2, t_2(\theta) = 3 - 6 = -3, t_3(\theta) = 7 - 7 = 0$$

7.2.2 Uniqueness of VCG Mechanism

Assumption K is a compact subset of a topological space which all singletons are closed (metric spaces $K \subset \mathbb{R}^n$, K compact, or any finite K .)

Let V_{usc} be the set of upper hemicontinuous functions $v : K \rightarrow \mathbb{R}$. (v is upper hemicontinuous if $\forall \alpha \in \mathbb{R} : \{k \in K \mid v(k) \geq \alpha\}$ is closed.)

Facts: A upper hemicontinuous function attains the maximum over a compact set. Sum of upper hemicontinuous functions is upper hemicontinuous.

Proposition 7.3 (Green & Laffont 1979)

Suppose that $\forall i : \{v_i(\cdot, \theta_i) : K \rightarrow \mathbb{R} \mid \theta_i \in \Theta_i\} = V_{usc}$. Then, any ex-post efficient and DSIC direct mechanism is a VCG mechanism.



Proof 7.4

Take any $f(\cdot) = (k^*(\cdot), t_i(\cdot))$ such that it is ex-post efficient and DSIC. We prove it is VCG mechanism by showing there is a h_i satisfies the definition of VCG mechanism.

Define $\forall i, h_i : \Theta \rightarrow \mathbb{R}$ such that

$$h_i(\theta) = - \sum_{j \neq i} v_j(k^*(\theta_i, \theta_{-i}), \theta_j) + t_i(\theta_i, \theta_{-i})$$

We want to show $h_i(\theta)$ is independent of θ_i and is actually $h_i(\theta_{-i})$.

That is, $\forall \theta_i, \hat{\theta}_i, \theta_{-i}$, we want to show $h_i(\theta_i, \theta_{-i}) = h_i(\hat{\theta}_i, \theta_{-i})$.

Lemma 7.1

If $k^*(\theta_i, \theta_{-i}) = k^*(\hat{\theta}_i, \theta_{-i})$, then $h_i(\theta_i, \theta_{-i}) = h_i(\hat{\theta}_i, \theta_{-i})$. ♥

Proof 7.5

$k^*(\theta_i, \theta_{-i}) = k^*(\hat{\theta}_i, \theta_{-i})$ requires

$$v_i(k^*(\theta_i, \theta_{-i}), \theta_i) + t_i(\theta_i, \theta_{-i}) \geq v_i(k^*(\hat{\theta}_i, \theta_{-i}), \theta_i) + t_i(\hat{\theta}_i, \theta_{-i})$$

$$v_i(k^*(\hat{\theta}_i, \theta_{-i}), \hat{\theta}_i) + t_i(\hat{\theta}_i, \theta_{-i}) \geq v_i(k^*(\theta_i, \theta_{-i}), \hat{\theta}_i) + t_i(\theta_i, \theta_{-i})$$

Since $v_i(k^*(\hat{\theta}_i, \theta_{-i}), \hat{\theta}_i) = v_i(k^*(\theta_i, \theta_{-i}), \hat{\theta}_i)$, we have $t_i(\hat{\theta}_i, \theta_{-i}) = t_i(\theta_i, \theta_{-i})$. Hence, $h_i(\theta_i, \theta_{-i}) = h_i(\hat{\theta}_i, \theta_{-i})$.

1. Case 1: “ $k^*(\theta_i, \theta_{-i}) = k^*(\hat{\theta}_i, \theta_{-i})$ ”, $h_i(\theta_i, \theta_{-i}) = h_i(\hat{\theta}_i, \theta_{-i})$ is given by Lemma 7.1.

2. Case 2: “ $k^*(\theta_i, \theta_{-i}) \neq k^*(\hat{\theta}_i, \theta_{-i})$ ”

Suppose by the way of contradiction $h_i(\theta_i, \theta_{-i}) \neq h_i(\hat{\theta}_i, \theta_{-i})$, WLOG, we consider $h_i(\theta_i, \theta_{-i}) > h_i(\hat{\theta}_i, \theta_{-i})$. There is an $\epsilon > 0$ s.t. $h_i(\theta_i, \theta_{-i}) > h_i(\hat{\theta}_i, \theta_{-i}) + \epsilon$.

Define $v : K \rightarrow \mathbb{R}$ such that

$$v(k) = \begin{cases} - \sum_{j \neq i} v_j(k^*(\theta_i, \theta_{-i}), \theta_j), & \text{if } k = k^*(\theta_i, \theta_{-i}) \\ - \sum_{j \neq i} v_j(k^*(\hat{\theta}_i, \theta_{-i}), \theta_j) + \epsilon, & \text{if } k = k^*(\hat{\theta}_i, \theta_{-i}) \\ -C, & \text{otherwise} \end{cases}$$

where $C > \max_{k \in K} \sum_{j \neq i} v_j(k^*(k, \theta_{-i}), \theta_j)$.

Hence, v is upper hemicontinuous, $v \in V_{usc}$.

By the assumption that $\forall i : \{v_i(\cdot, \theta_i) : K \rightarrow \mathbb{R} \mid \theta_i \in \Theta_i\} = V_{usc}$, we know $\exists \theta_i^\epsilon \in \Theta_i$ s.t.

$$v_i(\cdot, \theta_i^\epsilon) = v(\cdot).$$

- Because $k^*(\cdot)$ is ex-post efficient,

$$\begin{aligned} v_i(k^*(\theta_i^\epsilon, \theta_{-i}), \theta_i^\epsilon) + \sum_{j \neq i} v_j(k^*(\theta_i^\epsilon, \theta_{-i}), \theta_j) &\geq v_i(k^*(\hat{\theta}_i, \theta_{-i}), \theta_i^\epsilon) + \sum_{j \neq i} v_j(k^*(\hat{\theta}_i, \theta_{-i}), \theta_j) \\ &= v(k^*(\hat{\theta}_i, \theta_{-i})) + \sum_{j \neq i} v_j(k^*(\hat{\theta}_i, \theta_{-i}), \theta_j) = \epsilon \end{aligned}$$

By the definition of $v(\cdot)$, we have

$$v_i(k^*(\theta_i^\epsilon, \theta_{-i}), \theta_i^\epsilon) + \sum_{j \neq i} v_j(k^*(\theta_i^\epsilon, \theta_{-i}), \theta_j) = v(k^*(\theta_i^\epsilon, \theta_{-i})) + \sum_{j \neq i} v_j(k^*(\theta_i^\epsilon, \theta_{-i}), \theta_j) \leq \epsilon$$

Hence, we can conclude $k^*(\theta_i^\epsilon, \theta_{-i}) = k^*(\hat{\theta}_i, \theta_{-i})$. Then, by the Lemma 7.1, $h_i(\theta_i^\epsilon, \theta_{-i}) = h_i(\hat{\theta}_i, \theta_{-i})$.

- Because $f(\cdot) = (k^*(\cdot), t_i(\cdot))$ is DSIC, the agent with θ_i^ϵ gets the highest profit from truthfully reporting

$$\begin{aligned} v_i(k^*(\theta_i^\epsilon, \theta_{-i}), \theta_i^\epsilon) + t_i(\theta_i^\epsilon, \theta_{-i}) &\geq v_i(k^*(\theta_i, \theta_{-i}), \theta_i^\epsilon) + t_i(\theta_i, \theta_{-i}) \\ \Leftrightarrow -\sum_{j \neq i} v_j(k^*(\theta_i^\epsilon, \theta_{-i}), \theta_j) + \epsilon + t_i(\theta_i^\epsilon, \theta_{-i}) &\geq -\sum_{j \neq i} v_j(k^*(\theta_i, \theta_{-i}), \theta_j) + t_i(\theta_i, \theta_{-i}) \\ \Leftrightarrow h_i(\theta_i^\epsilon, \theta_{-i}) + \epsilon &\geq h_i(\theta_i, \theta_{-i}) \\ \Leftrightarrow h_i(\hat{\theta}_i, \theta_{-i}) + \epsilon &\geq h_i(\theta_i, \theta_{-i}) \end{aligned}$$

There is a contradiction.

7.2.3 Budget Balancedness of VCG Mechanism

Definition 7.10 (Budget-Balanced VCG Mechanism)

A VCG mechanism is **budget-balanced** if $\sum_{i=1}^I t_i(\theta) = 0$.



Based on the Proposition 7.3, we can show the following corollary.

Corollary 7.1

Suppose $I \geq 2$, $|K| \geq 2$, and $\forall i : \{v_i(\cdot, \theta_i) : K \rightarrow \mathbb{R} \mid \theta_i \in \Theta_i\} = V_{usc}$. Then, there does not exist a budget-balanced VCG mechanism.



Example 7.5 $K = \{0, 1\}$, $\Theta_i = [-1, 1]$, $v_i(k, \theta_i) = k\theta_i$. Take a VCG mechanism $k^*(\cdot)$ ex-post efficient, $h_1 : \Theta_2 \rightarrow \mathbb{R}$, $h_2 : \Theta_1 \rightarrow \mathbb{R}$.

$$\begin{aligned} t_1(\theta) + t_2(\theta) &= v_2(k^*(\theta), \theta_2) + h_1(\theta_2) + v_1(k^*(\theta), \theta_1) + h_2(\theta_1) \\ &= \max\{0, \theta_1 + \theta_2\} + h_1(\theta_2) + h_2(\theta_1) \end{aligned}$$

Suppose by the contradiction that it is a budget-balanced VCG mechanism.

$$\max\{0, \theta_1 + \theta_2\} + h_1(\theta_2) + h_2(\theta_1) = 0$$

We have

$$h_2(1) - h_2(0) = \max\{0, \theta_1\} - \max\{0, \theta_1 + 1\}$$

The LHS is constant and the RHS is a function of θ_1 , which gives a contradiction.

7.2.4 Expected-externality Mechanism (BIC)

Definition 7.11 (EE Mechanism)

$(k^*(\cdot), t(\cdot))$ is an Expected-externality (EE/AGV) mechanism if $k^*(\cdot)$ is ex-post efficient and there are functions $h_i : \Theta_i : \mathbb{R}$ for all i s.t.

$$t_i(\theta) = \mathbb{E}_{\tilde{\theta}_{-i}} \left[\underbrace{\sum_{j \neq i} v_j(k^*(\theta_i, \tilde{\theta}_{-i}), \theta_j)}_{\triangleq \xi_i(\theta_i)} \right] + h_i(\theta_{-i}) \quad (7.1)$$

where $\xi_i(\theta_i) \triangleq \mathbb{E}_{\tilde{\theta}_{-i}} \left[\sum_{j \neq i} v_j(k^*(\theta_i, \tilde{\theta}_{-i})) \right]$ is the expected externality i imposes on others, from i 's interim perspective when her type is θ_i .



Proposition 7.4

EE mechanisms are BIC.



Proof 7.6

Take any i and $\theta_i, \hat{\theta}_i \in \Theta_i$. i 's expected payoff from truthfully reporting is

$$\begin{aligned} & \mathbb{E}_{\tilde{\theta}_{-i}} \left[v_i(k^*(\theta_i, \tilde{\theta}_{-i}), \theta_i) + t_i(\theta_i, \tilde{\theta}_{-i}) \right] \\ (\text{substitute (7.1)}) &= \mathbb{E}_{\tilde{\theta}_{-i}} \left[\sum_{j=1}^I v_j(k^*(\theta_i, \tilde{\theta}_{-i}), \theta_j) \right] + \mathbb{E}_{\tilde{\theta}_{-i}} [h_i(\tilde{\theta}_{-i})] \\ (\text{k^* is ex-post efficient}) &\geq \mathbb{E}_{\tilde{\theta}_{-i}} \left[\sum_{j=1}^I v_j(k^*(\hat{\theta}_i, \tilde{\theta}_{-i}), \theta_j) \right] + \mathbb{E}_{\tilde{\theta}_{-i}} [h_i(\tilde{\theta}_{-i})] \\ &= \mathbb{E}_{\tilde{\theta}_{-i}} \left[\sum_{j=1}^I v_j(k^*(\hat{\theta}_i, \tilde{\theta}_{-i}), \theta_j) \right] + \mathbb{E}_{\tilde{\theta}_{-i}} [t_i(\hat{\theta}_i, \tilde{\theta}_{-i}) - \xi_i(\hat{\theta}_i)] \\ &= \mathbb{E}_{\tilde{\theta}_{-i}} \left[v_i(k^*(\hat{\theta}_i, \tilde{\theta}_{-i}), \theta_i) + t_i(\hat{\theta}_i, \tilde{\theta}_{-i}) \right] \end{aligned}$$

7.2.5 Budget-Balanced EE Mechanism

Budget balancedness requires

$$0 = \sum_{i=1}^I t_i(\theta) = \sum_{i=1}^I [\xi_i(\theta_i) + h_i(\theta_{-i})] \Leftrightarrow \sum_{i=1}^I h_i(\theta_{-i}) = - \sum_{i=1}^I \xi_i(\theta_i)$$

Suppose $h_i(\theta_{-i})$ is in the form of $h_i(\theta_{-i}) = c \sum_{j \neq i} \xi_j(\theta_j)$. Then,

$$\sum_{i=1}^I h_i(\theta_{-i}) = c(I-1) \sum_{i=1}^I \xi_i(\theta_i) \Rightarrow c = -\frac{1}{I-1}$$

Proposition 7.5

The EE mechanism where $h_i(\theta_{-i}) = -\frac{1}{I-1} \sum_{j \neq i} \xi_j(\theta_j)$ is budget-balanced,

$$t_i(\theta) = \xi_i(\theta_i) - \frac{1}{I-1} \sum_{j \neq i} \xi_j(\theta_j)$$



Corollary 7.2

\exists a BIC, ex-post efficient and budget-balanced direct mechanism.



Example 7.6 Project choice with $K = \{0, 1\}$, $\theta_i \sim U[-1, 1]$, $v_j(k, \theta_j) = k\theta_j$. Let $k^*(\cdot)$ be:

$$k^*(\theta) = \begin{cases} 1, & \text{if } \theta_1 + \theta_2 \geq 0 \\ 0, & \text{if } \theta_1 + \theta_2 < 0 \end{cases}$$

Then,

$$\begin{aligned} \xi_i(\theta_i) &\triangleq \mathbb{E}_{\tilde{\theta}_{-i}} \left[v_{-i}(k^*(\theta_i, \tilde{\theta}_{-i})) \right] \\ &= \int_{-1}^{-\theta_i} 0 \times \frac{1}{2} d\tilde{\theta}_{-i} + \int_{-\theta_i}^1 \tilde{\theta}_{-i} \times \frac{1}{2} d\tilde{\theta}_{-i} = \frac{1}{4}(1 - \theta_i^2) \end{aligned}$$

Hence, the budget-balanced EE mechanism is given by

$$t_i(\theta_i) = \xi_i(\theta_i) - \xi_j(\theta_j) = \frac{1}{4}(\theta_j^2 - \theta_i^2)$$

7.2.6 Linear Utility Model

Suppose types are real numbers $\Theta_i = [\underline{\theta}_i, \bar{\theta}_i]$ and $v_i(k, t, \theta_i) = \theta_i \cdot v_i(k) + t_i$, where $v_i : K \rightarrow \mathbb{R}_i$.

Given a direct mechanism $(\Theta, k(\cdot), t(\cdot))$, define interim expected values of $v_i(\cdot)$ and $t_i(\cdot)$:

$$\bar{v}_i(\theta_i) = \mathbb{E}_{\theta_{-i}} [v_i(\theta_i, \theta_{-i})]$$

$$\bar{t}_i(\theta_i) = \mathbb{E}_{\theta_{-i}} [t_i(\theta_i, \theta_{-i})]$$

The expected utility of agent i when all agents report truthfully,

$$\begin{aligned} U_i(\theta_i) &= \mathbb{E}_{\theta_{-i}} [\theta_i v_i(\theta_i, \theta_{-i}) + t_i(\theta_i, \theta_{-i})] \\ &= \theta_i \bar{v}_i(\theta_i) + \bar{t}_i(\theta_i) \end{aligned}$$

Proposition 7.6

A direct mechanism $(\Theta, k(\cdot), t(\cdot))$ is BIC iff $\forall i \in \{1, \dots, I\}$

- (1). $\bar{v}_i(\theta_i)$ is non-decreasing in θ_i .
- (2). $\forall \theta_i \in \Theta_i, U_i(\theta_i) = U_i(\underline{\theta}_i) + \int_{\underline{\theta}_i}^{\theta_i} \bar{v}_i(s) ds$

**Proof 7.7**

\Rightarrow : Given the direct mechanism is BIC. Take any $i, \theta_i, \hat{\theta}_i \in \Theta$, agents with $\theta_i, \hat{\theta}_i$ both report truthfully

$$\begin{aligned}\theta_i \bar{v}_i(\theta_i) + \bar{t}_i(\theta_i) &\geq \theta_i \bar{v}_i(\hat{\theta}_i) + \bar{t}_i(\hat{\theta}_i) \\ \hat{\theta}_i \bar{v}_i(\hat{\theta}_i) + \bar{t}_i(\hat{\theta}_i) &\geq \hat{\theta}_i \bar{v}_i(\theta_i) + \bar{t}_i(\theta_i) \\ \Rightarrow [\theta_i - \hat{\theta}_i][\bar{v}_i(\theta_i) - \bar{v}_i(\hat{\theta}_i)] &\geq 0\end{aligned}$$

Hence, $\bar{v}_i(\theta_i)$ is non-decreasing.

$U_i(\theta_i) = \max_{\hat{\theta}_i \in \Theta_i} [\theta_i \bar{v}_i(\hat{\theta}_i) + t_i(\hat{\theta}_i)]$ by BIC is maximized at $\hat{\theta}_i = \theta_i$.

By Envelope Theorem,

$$\begin{aligned}U_i(\theta_i) &= U_i(\underline{\theta}_i) + \int_{\underline{\theta}_i}^{\theta_i} U'_i(s) ds \\ &= U_i(\underline{\theta}_i) + \int_{\underline{\theta}_i}^{\theta_i} \bar{v}_i(s) ds\end{aligned}$$

\Leftarrow : Take any $i, \theta_i, \hat{\theta}_i \in \Theta$. i 's expected interim payoff from reporting $\hat{\theta}_i$ instead of θ_i is

$$\begin{aligned}&\underbrace{\theta_i \bar{v}_i(\theta_i) + \bar{t}_i(\theta_i)}_{U_i(\theta_i)} - [\theta_i \bar{v}_i(\hat{\theta}_i) + \bar{t}_i(\hat{\theta}_i)] \\ &= U_i(\theta_i) - [U_i(\hat{\theta}_i) + (\theta_i - \hat{\theta}_i)\bar{v}_i(\hat{\theta}_i)] \\ &= U_i(\underline{\theta}_i) + \int_{\underline{\theta}_i}^{\theta_i} \bar{v}_i(s) ds - [U_i(\underline{\theta}_i) + \int_{\underline{\theta}_i}^{\hat{\theta}_i} \bar{v}_i(s) ds] - (\theta_i - \hat{\theta}_i)\bar{v}_i(\hat{\theta}_i) \\ &= \int_{\theta_i}^{\hat{\theta}_i} (\bar{v}_i(\hat{\theta}_i) - \bar{v}_i(s)) ds \geq 0\end{aligned}$$

So the direct mechanism is BIC.

7.3 Auction

Based on

- o Klempner, P. (1998). Auctions with almost common values: The Wallet Game' and its applications.
European Economic Review, 42(3-5), 757-769.

7.3.1 Examples: Auctions with Common-value

- (1). Financial assets;

- (2). Oilfields;
- (3). A takeover target has a common value if the bidders are financial acquirers (e.g. LBO firms) who will follow similar management strategies if successful;
- (4). The Personal Communications Spectrum (PCS) licenses sold by the U.S. Government in the 1995 "Air-waves Auction".

7.3.2 First/Second Price Sealed-bid Auction

- A seller sells a indivisible object.
- There are $N = \{1, \dots, n\}$ bidders, $i \in N$.
- Each bidder has a valuation for the object, $X_i \sim F$, $x_i \in [\underline{x}, \bar{x}]$. p.d.f. $f(\cdot)$ is strictly positive and continuous.
- Strategy of i : $b_i : [\underline{x}, \bar{x}] \rightarrow \mathbb{R}$, a bid function.

Assumption 1. Independence; 2. Symmetry; 3. Private Values; 4. Risk-neutrality.

Let $X = (X_1, \dots, X_n)$. The k^{th} -order statistic, X^k , is the k^{th} the highest value in X_1, \dots, X_n .

Definition 7.12 (Second Price Auction)

Highest bidder wins and pays the second-highest bid. (If more than one bidders bid the highest value, they win with equal probability.)

It can be written as the form of Bayesian game: a bidder i 's utility function is

$$u_i(b_1, \dots, b_n; x_i) = \begin{cases} \frac{1}{|\{j \in N : b_j = b_i\}|} x_i - b^2, & b_i = b^1 \\ 0, & b_i \neq b^1 \end{cases}$$

where b^k is the k -th highest bid.



Theorem 7.4 (Second Price Auction: Bid Truthfully)

In the second-price sealed-bid auction, it is a (weakly) dominant strategy to bid your valuation, i.e.,

$\forall i \in N, \forall x_i \in [\underline{x}, \bar{x}], b_i(x_i) = x_i$.

That is, $\forall i, \forall b'_i \in \mathbb{R}$,

$$u_i(x_i, b_{-i}; x_i) \geq u_i(b'_i, b_{-i}; x_i), \forall b_{-i} \in \mathbb{R}^{n-1}$$

(Moreover, if $\exists b'_i \neq x_i$, then $\exists b_{-i} \in \mathbb{R}^{n-1}$ such that $u_i(x_i, b_{-i}; x_i) > u_i(b'_i, b_{-i}; x_i)$.)



Proof 7.8

Player i has value x_i and treats b_{-i}^1 as a random variable. The payoff conditional on winning is

$$x_i - b_{-i}^1$$

By bidding $b_i = x_i$, i ensures that i wins if $b_i = x_i > b_{-i}^1 \Leftrightarrow x_i - b_{-i}^1 > 0$ and i loses if $b_i = x_i < b_{-i}^1 \Leftrightarrow x_i - b_{-i}^1 < 0$.

Definition 7.13 (First Price Auction)

Highest bidder wins and pays her bid. (If more than one bidder bid the highest value, they win with equal probability.)

It can be written as the form of Bayesian game: a bidder i 's utility function is

$$u_i(b_1, \dots, b_n; x_i) = \begin{cases} \frac{1}{|\{j \in N : b_j = b_i\}|} x_i - b_i, & b_i = b^1 \\ 0, & b_i \neq b^1 \end{cases}$$

where b^k is the k -th highest bid.



Bayesian Nash Equilibrium Analysis of First Price Auction

Conjecture that \exists a BNE with the following properties:

1. Symmetry: $b_1(\cdot) = b_2(\cdot) = \dots = b_n(\cdot) := b(\cdot)$.
2. $b(\cdot)$ is differentiable.
3. $b'(\cdot) > 0$.

Take any bidder i with valuation x_i . Assume i knows $b(\cdot)$ and knows that the other bidder use the same $b(\cdot)$.

Take any $b_i \in \mathbb{R}$ ($b_i := b(x_i)$). (Not that, by the continuity of X_i , it is impossible to tie in this case.)

Then, i 's expected payoff from bidding b_i is

$$P(b(X_j) \leq b_i, \forall j \neq i)(x_i - b_i) = F^{n-1}(b^{-1}(b_i))(x_i - b_i)$$

The necessary F.O.C. gives that optimal b_i satisfies

$$(n-1)f(b^{-1}(b_i)) \frac{1}{b'(b^{-1}(b_i))} F^{n-2}(b^{-1}(b_i))(x_i - b_i) - F^{n-1}(b^{-1}(b_i)) = 0$$

Since $b(\cdot)$ is a symmetric BNE, the optimal b_i must be $b(x_i)$, then $b^{-1}(b_i) = x_i$.

$$(n-1)f(x_i) \frac{1}{b'(x_i)} F^{n-2}(x_i)(x_i - b(x_i)) - F^{n-1}(x_i) = 0$$

Hence,

$$\underbrace{(n-1)f(x_i)F^{n-2}(x_i)x_i + F^{n-1}(x_i)}_{\frac{\partial F^{n-1}(x_i)x_i}{\partial x_i}} - F^{n-1}(x_i) = \underbrace{(n-1)f(x_i)F^{n-2}(x_i)b(x_i) + b'(x_i)F^{n-1}(x_i)}_{\frac{\partial F^{n-1}(x_i)b(x_i)}{\partial x_i}}$$

Taking integral at both sides in $[\underline{x}, x]$,

$$F^{n-1}(x)x - \int_{\underline{x}}^x F^{n-1}(t)dt = F^{n-1}(x)b(x)$$

That is,

$$b(x) = x - \frac{1}{F^{n-1}(x)} \int_{\underline{x}}^x F^{n-1}(t) dt \quad (7.2)$$

Note $b(\cdot)$ is differentiable and $b'(\cdot) > 0$. We can extend $b(\cdot)$ to $[\underline{x}, \bar{x}]$ by setting $b(\underline{x}) = \lim_{x \rightarrow \underline{x}} b(x) = \underline{x}$.

Proposition 7.7 (Symmetric BNE of First Price Auction)

$b(x) = x - \frac{1}{F^{n-1}(x)} \int_{\underline{x}}^x F^{n-1}(t) dt$ is a symmetric BNE of First Price Auction.



Proof 7.9

Any bid higher than $b(\bar{x})$ is suboptimal, and any bid lower than $b(\underline{x})$ is indifferent to the $b(\underline{x})$.

We prove that, for a bidder with x_i , she prefers to bid $b(x_i)$ than $b(y)$, $\forall y$.

Bidding $b(y)$ gives expected payoff

$$\begin{aligned} F^{n-1}(y)(x_i - b(y)) &= F^{n-1}(y)(y - b(y)) + F^{n-1}(y)(x_i - y) \\ (\text{by (7.2)}) &= \int_{\underline{x}}^y F^{n-1}(t) dt + F^{n-1}(y)(x_i - y) \\ &= \int_{\underline{x}}^{x_i} F^{n-1}(t) dt - \underbrace{\int_y^{x_i} [F^{n-1}(t) - F^{n-1}(y)] dt}_{\geq 0, \text{ minimized at } y=x_i} \end{aligned}$$

Theorem 7.5 (Lebrun, 1999)

Consider the bid function $b(\cdot)$ in (7.2), the First Price Auction has essentially unique. (Bidders with types $x > \underline{x}$ bid $b(x)$ and bid for type $x = \underline{x}$ is not pinned down any further than the support $b_i(\underline{x})$ must lie in $(-\infty, b(\underline{x})]$.)



7.4 Revenue Equivalence Theorem

Consider the Optimal Auctions in an Independent Private Values Setting. There is one object and N bidders.

1. Bidders are risk-neutral;
2. Bidders have private valuations;
3. each bidder i 's valuation independently drawn from a strictly increasing c.d.f. $F_i(v)$ (with p.d.f. $f_i(v)$, $v \in \mathcal{X}_i$) that is continuous and bounded below;
4. Seller knows each F_i (use F and f to represent all distributions) and have no value for the object.

Definition 7.14 (General Auction)

A general auction mechanism: bidders have values x and **strategies** $\beta : \mathcal{X} \triangleq \prod_i^N \mathcal{X}_i \rightarrow \mathcal{B}$ generate message (bids) based on their values, then there is an **allocation rule** based on bids $\pi : \mathcal{B} \rightarrow \Delta N$ generates a distribution over all bidders and a **payment rule** $\mu : \mathcal{B} \rightarrow \mathbb{R}^N$ generates payment for all

bidders.



Definition 7.15 (Direct Mechanism)

Consider a situation that bidders follow *revelation principle* that provide their true values. Then the outcome can directly base on the true values.

Then a **direct mechanism** can be represented as (Q, T) , where $Q : \mathcal{X} \rightarrow \Delta N$ is the allocation rule and $T : \mathcal{X} \rightarrow \mathbb{R}^N$ is the payment rule, such that

$$Q(x) = \pi(\beta(x)), \quad T(x) = \mu(\beta(x))$$



Proposition 7.8 (Revelation Principle)

Take any equilibrium of any auction mechanism (\mathcal{B}, π, μ) . There is a distinct direct mechanism (Q, T) that produces the same outcome.



Consider a direct mechanism (Q, T) , an agent i reports v_i while others report their values.

$$\textbf{Expected allocation: } q_i(z_i) = \int_{\mathcal{X}_{-i}} Q_i(z_i, x_{-i}) dF_{-i}(x_{-i})$$

$$\textbf{Expected payment: } t_i(z_i) = \int_{\mathcal{X}_{-i}} T_i(z_i, x_{-i}) dF_{-i}(x_{-i})$$

where Q_i, T_i are i^{th} item of Q, T .

The bidder wants to maximize

$$q_i(z_i)x_i - t_i(z_i)$$

Define the maximum value is

$$u_i(x_i) = \max_{z_i \in \mathcal{X}_i} \{q_i(z_i)x_i - t_i(z_i)\}$$

Assumption The condition for direct mechanism being incentive competitive (IC) is:

$$u_i(x_i) \equiv q_i(x_i)x_i - t_i(x_i) \geq q_i(z_i)x_i - t_i(z_i), \forall x_i, z_i \in \mathcal{X}_i \quad (\text{Ass 1})$$

Firstly, we can compute, for any $z_i \in \mathcal{X}_i$

$$\begin{aligned} & q_i(x_i)z_i - t_i(x_i) \\ &= q_i(x_i)x_i - t_i(x_i) + q_i(x_i)(z_i - x_i) \\ &= u_i(x_i) + q_i(x_i)(z_i - x_i) \end{aligned}$$

Based on the assumption Ass 1, we have

$$u_i(z_i) \geq q_i(x_i)z_i - t_i(x_i) = u_i(x_i) + q_i(x_i)(z_i - x_i)$$

which shows that $u_i(\cdot)$ is convex.

If u_i is differentiable, $u'_i(x_i) = q_i(x_i)$. Then, we can write the **Envelope theorem/condition**:

$$u_i(x_i) = u_i(0) + \int_0^{x_i} q_i(y_i) dy_i$$

which only depends on the allocation rule.

Theorem 7.6 (Revenue Equivalence Theorem)

*If the direct mechanism (Q, T) is incentive competitive (IC), then for all i, x , the **expected payment** is*

$$t_i(x_i) = \underbrace{t_i(0)}_{e.g.=0} + q_i(x_i)x_i - \int_0^{x_i} q_i(y_i) dy_i$$



Proof 7.10

$$\begin{aligned} u_i(x_i) &= q_i(x_i)x_i - t_i(x_i) = u_i(0) + \int_0^{x_i} q_i(y_i) dy_i \\ &\Rightarrow t_i(x_i) = q_i(x_i)x_i - u_i(0) - \int_0^{x_i} q_i(y_i) dy_i \end{aligned}$$

Set $u_i(0) = -t_i(0)$, that is, if i 's value is zero, he pays zero.

Corollary 7.3 (Standard Revenue Equivalence Theorem)

Suppose that values are i.i.d. and bidders are risk-neutral.

Consider any auction and its symmetric and increasing equilibrium, in which the expected payment of bidders have 0 value is 0. Then the expected revenue to the seller is the same.



Proof 7.11

If equilibrium, is symmetric and increasing, then object is always allocated to the bidder with the highest value. Set $t_i(0) = 0$.

Standard Revenue Equivalence Theorem is based on symmetric, independent, and private (uncorrelated) values.

7.5 Optimal Auctions

Goal: Find the **optimal auction** that maximizes the seller's expected revenue subject to individual rationality (IR) and Bayesian incentive compatibility for the buyers.

Definition 7.16 (Virtual Valuation)

Bidder i 's **virtual valuation** is $\phi_i(v_i) = v_i - \frac{1-F_i(v_i)}{f'(v_i)}$



Definition 7.17 (Bidder-Specific Reserve Price)

Bidder i 's bidder-specific reserve price r_i^* is the value for which $\phi_i(r_i^*) = 0$.



Theorem 7.7 (Myerson (1981))

The optimal (single-good) auction in terms of a direct mechanism: The good is sold to the agent $i = \arg \max_i \phi_i(\hat{v}_i)$, as long as $v_i \geq r_i^$. If the good is sold, the winning agent i is charged the smallest valuation that he could have declared while still remaining the winner:*

$$\inf\{v_i^* : \phi_i(v_i^*) \geq 0 \text{ and } \forall j \neq i, \phi_i(v_i^*) \geq \phi_j(\hat{v}_j)\}$$



Chapter 8 Market Design

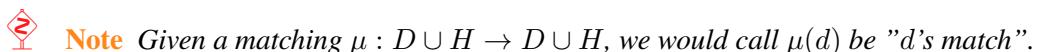
Based on

- Two-Sided Matching: A Study in Game-Theoretic Modeling and Analysis, Roth, Alvin E.& Sotomayor, Matilda, 1990.
- Fleiner, T. (2003). A fixed-point approach to stable matchings and some applications. *Mathematics of Operations research*, 28(1), 103-126.
- Hatfield, J. W., & Kominers, S. D. (2017). Contract design and stability in many-to-many matching. *Games and Economic Behavior*, 101, 78-97.

8.1 Matching One-to-One

Suppose there are doctors (D) and hospitals (H). For a doctor d , define a relation \succeq_d over $H \cup \{d\}$; for a hospital h , define a relation \succeq_h over $D \cup \{h\}$. A matching market is defined by

$$(D, H, \{\succeq_i\}_{i \in D \cup H})$$

 **Note** Given a matching $\mu : D \cup H \rightarrow D \cup H$, we would call $\mu(d)$ be "d's match".

Definition 8.1 (Involution)

A matching $\mu : D \cup H \rightarrow D \cup H$ is **involution** such that

$$\mu(d) \neq d \Rightarrow \mu(\mu(d)) = d, \forall d \in D$$

and

$$\mu(h) \neq h \Rightarrow \mu(\mu(h)) = h, \forall h \in H$$



Definition 8.2 (Stable)

A matching $\mu : D \cup H \rightarrow D \cup H$ is **stable** if it is

- Individually Rational: $\nexists i$ for whom $i > \mu(i)$.
- (Pairwise) Unblocked: $\nexists (d, h)$ such that $d \succ_h \mu(h)$ and $h \succ_d \mu(d)$.



Theorem 8.1 (Gale-Shapley, 1962)

For any matching market, a stable matching μ exists.



Proof 8.1

We prove it by an algorithm:

Definition 8.3 (Deferred Acceptance Algorithm (DA))

At each round, every doctor applies for his most preferred hospital among those haven't rejected him. Each hospital chooses its most preferred doctors among its applicants and the one on the previous waitlist, and then rejects others.



Observation: DA terminates μ . We want to prove

1. μ is IR (obviously);
2. μ is unblocked.

Suppose there is a block (d, h) such that $d \succ_h \mu(h)$ and $h \succ_d \mu(d)$. That is impossible, because the $d \neq \mu(h)$, the d must be rejected by h , which means $h \preceq_d \mu(d)$.



Note We call " h is **achievable** for d " if $\mu(d) = h$ for some stable matching μ .

8.1.1 Matching Markets: One-to-One

Definition 8.4 (D -Optimal Matching)

A matching $\mu : D \cup H \rightarrow D \cup H$ is **D -optimal**, denoted by μ^D , if for any stable μ' we have that $\mu^D \succeq_D \mu'$ (the best stable matching for all doctors).

**Theorem 8.2 (Deferred Acceptance Algorithm \Rightarrow D -Optimal Matching)**

Deferred Acceptance Algorithm (with D proposing) terminates in μ^D .

**Proof 8.2**

...Theorem 2.12 (Gale and Shapley)

Theorem 8.3 (Lone-Wolf Theorem)

The set of matched agent is identical in every stable μ .

**Proof 8.3**

$|\mu^D(H)| \geq |\mu(H)| \geq |\mu^H(H)|$; by symmetry, $|\mu^H(D)| \geq |\mu(D)| \geq |\mu^D(D)|$. Because $|\mu^D(H)| = |\mu^D(D)|$ and $|\mu^H(H)| = |\mu^H(D)|$ by one-to-one, so everything is equal.

8.1.2 Joint and Meet

Definition 8.5 (Joint and Meet)

1. **Join** $\mu \vee_D \mu'$ assign the more preferred match to every d and the less preferred match to every h , that is,

$$\mu \vee_D \mu'(d) = \begin{cases} \mu(d), & \text{if } \mu(d) \geq_d \mu'(d) \\ \mu'(d), & \text{otherwise} \end{cases}, \forall d \in D$$

$$\mu \vee_D \mu'(h) = \begin{cases} \mu(h), & \text{if } \mu(h) \leq_h \mu'(h) \\ \mu'(h), & \text{otherwise} \end{cases}, \forall h \in H$$

2. **Meet** $\mu \wedge_D \mu'$ assign the less preferred match to every d and the more preferred match to every h , that is,

$$\mu \wedge_D \mu'(d) = \begin{cases} \mu(d), & \text{if } \mu(d) \leq_d \mu'(d) \\ \mu'(d), & \text{otherwise} \end{cases}, \forall d \in D$$

$$\mu \wedge_D \mu'(h) = \begin{cases} \mu(h), & \text{if } \mu(h) \geq_h \mu'(h) \\ \mu'(h), & \text{otherwise} \end{cases}, \forall h \in H$$



Theorem 8.4 (Join and Meet of Stable Matchings are Stable)

If μ and μ' are stable, then $\mu \vee_D \mu'$ and $\mu \wedge_D \mu'$ are stable.



8.1.3 Strategic Incentives

- Type = preference list.
- SCF: $f : \Theta \rightarrow \mathcal{M}$, where \mathcal{M} is a set of stable matchings;
- Is f strategy-proof?
- Does there exist a stable and strategy-proof (direct) mechanism?

Definition 8.6

We say a mechanism φ is strategy-proof (SP) if $\varphi(\succ_i, \succ_{-i}) \geq \varphi(\succ'_i, \succ_{-i})$ for all $i \in I$ and \succ'_i and \succ_{-i} .



Theorem 8.5 (Impossibility theorem (Roth))

There is no stable and strategy-proof (SP) mechanism.



The mechanism that yields the D-optimal stable matching (in terms of the stated preferences) makes it a dominant strategy for each doctor to state his true preferences. (Similarly, the mechanism that yields the H-optimal stable matching makes it a dominant strategy for every hospital to state its true preferences.)

Theorem 8.6 (Dubins and Freedman; Roth)

The doctor(D)-optimal stable mechanism is strategy-proof for doctors.

**Proof 8.4**

8.2 Matching Many-to-Many

Contracts are denoted by $x \in X$, $x_D \in D$, $x_H \in H$. $F \triangleq D \cup H$.

Consider a set of contracts $Y \subseteq X$,

- Y_D = doctors listed in Y ;
- Y_d = the contract in Y that list the doctor d ;
- \succ_d over set of contracts that name the doctor d ;
- The set of contracts $f \in F$ chooses from Y : $C_f(Y) = \max_{\succ_f} \{Z \subseteq X : Z \subseteq Y_f\} \subseteq Y_f$;
- The set of contracts doctors choose from Y : $C_D(Y) = \cup_{d \in D} C_d(Y)$.
- The set of contracts doctors reject from Y : $R_D(Y) = Y \setminus C_D(Y)$.

The outcome of matching is $Y \subseteq X$.

Definition 8.7 (Stable Contracts)

$A \subseteq X$ is **stable** if

- Individually Rational (IR): for all $f \in F$: $C_f(A) = A_f$;
- Unblocked: \nexists non-empty $Z \subseteq X$ such that $Z \cap A = \emptyset$ and for all $f \in F$, $Z_f \subseteq C_f(A \cup Z)$.



Example 8.1 Preferences over doctor d : $\{x, y\} > \{x\} > \emptyset > \{y\}$; Preferences over hospital h : $\{y\} > \{x, y\} > \{x\} > \emptyset$.

$$\{x\} \Rightarrow \{x, y\} \Rightarrow \{y\} \Rightarrow \emptyset \Rightarrow \{x\}.$$

Definition 8.8 (Substitutability Condition)

Preference of f satisfies the **substitutability condition** if for all $Y \subseteq X$ and $x, z \in X \setminus Y$:

$$z \notin C_f(Y \cup \{z\}) \Rightarrow z \notin C_f(Y \cup \{z\} \cup \{x\})$$

$$(Y' \subseteq Y \subseteq X \Rightarrow R_f(Y') \subseteq R_f(X)), \text{ where } R \text{ is the rejection choice.}$$



If z is rejected given a set, then it should also be rejected given a larger set.

Theorem 8.7

If contracts are substitutes, then $Y \subseteq X$ is stable if and only if pairwise stable.



Proof 8.5

Prove \Leftarrow : (If not pairwise stable \Rightarrow not stable)

Suppose that Z is a block. So, $Z \subseteq C_f(A \cup Z)$ for all f listed in Z .

We can pick a $z \in Z$ such that $z \in C_f(A \cup Z)$. By the substitutability condition, $z \in C_f(A \cup \{z\})$. So, it is stable.

Theorem 8.8

If contracts are substitutes, then a stable outcome exists.

**Definition 8.9 (Lattice)**

On a **lattice**, $L = (X, <, \wedge, \vee)$ (or we just use $L = (X, <)$), $<$ is a partial order on X in such a way that any two elements x and y of X have a unique greatest lower bound (glb) $x \wedge y$ (meet) and a unique lowest upper bound (lub) $x \vee y$ (join).

**Definition 8.10 (Complete Lattice)**

A lattice $L = (X, <)$ is **complete** if there are both a meet (i.e. a greatest lower bound) and a join (i.e. a least upper bound) for any subset $Y \subseteq X$.

These generalized meet and join operations on Y are denoted by $\wedge Y$ and $\vee Y$.

**Definition 8.11 (Monotone Function over Lattice)**

A function from one lattice to another lattice $f : (X, <) \rightarrow (X', <')$ is **monotone** if $x \leq y \Rightarrow f(x) \leq' f(y)$ for any $x, y \in X$.

**Theorem 8.9 (Tarski 1955)**

Let $L = (X, <)$ be a complete lattice and $f : L \rightarrow L$ be monotone (\leq) function on L . Then, the set $\{x \in L : f(x) = x\}$ of fixed points is a non-empty, complete lattice with order \leq .

**Proof 8.6**

Fleiner, T. (2003). A fixed-point approach to stable matchings and some applications. Mathematics of Operations research, 28(1), 103-126.

If some contracts are not substitute, there are no stable outcomes exist.

8.3 Matching Many-to-One

Settings

- Doctors, D ; Hospitals, H ; Contracts $X = D \times H \times \text{terms}$;
- Hospitals preference \succ_h over 2^X ;

- Doctors preference \succ_d over X (compare one contract with another one contract, not compare over sets of contracts);
- Outcome is $Y \subseteq X$ s.t. $|Y_d| \leq 1$ for all $d \in D$ (a doctor signs at most one contract).

What restriction do we need to have a stable matching? Not as strong as substitute.

Corollary 8.1

Doctor-proposing DA algorithm produces a doctor-optimal stable matching.



Example 8.2 The preferences of agents are

- $d_1 : h_1 \succ h_2; d_2 : h_1 \succ h_2; d_3 : h_2 \succ h_1;$
- $h_1 : d_3 \succ d_1, d_2 \succ d_1 \succ d_2; h_2 : d_1 \succ d_2 \succ d_3.$

There are two stable outcomes

- $(d_1, h_2), (d_3, h_1);$
- $(d_1, h_1), (d_2, h_1), (d_3, h_2).$

Remark Lone-Wolf Theorem doesn't hold.

Assume the d_2 's true preference is $h_2 \succ h_1$ and he reveals it, there is only one stable matching: (d_1, h_2) , (d_3, h_1) . So, the d_2 may benefit from lying.

Remark Strategy-proof doesn't hold.

Definition 8.12 (Law of Aggregate Demand/ Cardinality Monotonicity (CM))

For $h, Y \subseteq Y' \subseteq X \Rightarrow |C_h(Y)| \leq |C_h(Y')|$



Theorem 8.10

Under substitutes and CM, doctor-proposing DA is strategy-proof and LWT holds.



Theorem 8.11 (Rural Hospital Theorem)

Under substitutes / CM, hospitals have same numbers of contracts in every stable outcome.



Cadets-branch matching

Can be found in:

- Jagadeesan, R. (2019). Cadet-branch matching in a Kelso-Crawford economy. *American Economic Journal: Microeconomics*, 11(3), 191-224.

Remark Contracts are not substitutes.

Definition 8.13 (Unilateral Substitute)

Contracts are **unilateral substitutes** if for all $z, x \in X$ and $Y \subseteq X$ such that $z_D \notin Y_D$ if $z \notin C_h(Y \cup \{z\}) \Rightarrow z \notin C_h(Y \cup \{z\} \cup \{x\})$



Remark Preferences of branches satisfying unilateral substitute.

Remark The outcome of doctor-proposing DA algorithm is doctor-optimal and stable.

8.4 Networks

Based on

- Fleiner, T., Jankó, Z., Tamura, A., & Teytelboym, A. (2015). Trading networks with bilateral contracts. arXiv preprint arXiv:1510.01210.
- Fleiner, T., Jankó, Z., Schlotter, I., & Teytelboym, A. (2023). Complexity of stability in trading networks. *International Journal of Game Theory*, 1-20.

Considering a trading network represented by a directed graph, where nodes are firms F and edges X are contracts (income arrow can be understood as buying products and outcome arrow can be understood as selling products).

The choice function of $f \in F$ is represented by C^f , the choice of f over $Y_f \subseteq X_f$ is $C^f(Y_f) \subseteq Y_f$, where X_f is the set of contracts involving f .

The choice sets of buyer side (B) and seller side (S) are defined as

$$C_B^f(Y|Z) \triangleq C^f(Y_f^B \cup Z_f^S) \cap X_f^B$$

$$C_S^f(Z|Y) \triangleq C^f(Z_f^S \cup Y_f^B) \cap X_f^S$$

where Y is the contracts from buyer side and Z is the contracts from seller side.

Definition 8.14 (Irrelevance of Rejected Contracts)

Irrelevance of Rejected Contracts (IRC): $C(A) \subseteq B \subseteq A \Rightarrow C(A) = C(B)$



Definition 8.15 (Fully Substitute)

C^f is **fully substitute** if for $Y' \subseteq Y \subseteq X$ and $Z' \subseteq Z \subseteq X$,

$$R_B^f(Y'|Z) \subseteq R_B^f(Y|Z)$$

$$R_S^f(Z'|Y) \subseteq R_S^f(Z|Y)$$

and

$$R_B^f(Y|Z) \subseteq R_B^f(Y|Z')$$

$$R_S^f(Z|Y) \subseteq R_S^f(Z|Y')$$



Define partial order, $(Y, Z) \geq (Y', Z')$ if $Y \subseteq Y'$ and $Z \supseteq Z'$.

Definition 8.16 (Stable Outcome, Hatfield and Kominers (2012))

An outcome $A \subseteq X$ is stable if it is

1. Individual Rational: $\forall f \in F, C^f(A_f) = A_f$;
2. Unblocked: there is no non-empty set $Z \subseteq X$ s.t. $Z \cap A = \emptyset$ and $\forall f \in F(Z), Z_f \subseteq C^f(A \cup Z)$, where $F(Z)$ is the set of the firms are lined to Z .

**Definition 8.17 (Trail)**

Trail is the set of distinct edges $T = (X^1, X^2, \dots, X^M)$ such that the buyer side (the firm who is the buyer in the edge) of X^i is exactly the seller side (the firm who is the seller in the edge) of X^{i+1} , which is denoted by $b(X^i) = s(X^{i+1})$, $i = 1, \dots, M - 1$.

**Definition 8.18 (Trail-stable Outcome)**

An outcome $A \subseteq X$ is **trail-stable** if its is

1. Individual Rational;
2. There is no locally blocking trail $T = (X^1, X^2, \dots, X^M)$ such that

$$\begin{aligned} X^1 &\in C^{S(X^1)}(A \cup X^1); \\ \{X^i, X^{i+1}\} &\in C^{b(X^i)}(A \cup X^i \cup X^{i+1}); \\ X^M &\in C^{b(X^M)}(A \cup X^M). \end{aligned}$$

**Theorem 8.12 (Fleiner et al. 2016)**

If C^f is fully substitute and IRC for all $f \in F$, then a trail-stable outcome exists.

**Proof 8.7**

$Y \subseteq X$ and $Z \subseteq X$,

$$\Phi(Y, Z) = (X \setminus R_S(Z|Y), X \setminus R_B(Y|Z))$$

where $R_B(Y|Z) = \cup_{f \in F} R_B^f(Y|Z)$.

Claim 8.1

If (Y, Z) is a fixed point of Φ , then $A = Y \cap Z$ is trail-stable outcome.

**Lemma 8.1**

C^f is fully substitute and IRC, and (Y, Z) such that $Y \cap Z = A$, $C_S(Z|Y) = A$, $C_B(Y|Z) = A$.

Then, for a contract $x \in X \setminus A$ and $A \subseteq A' \subseteq X$ if $C_S^{S(x)}(A \cup x|A')$ then $x \in C_S^{S(x)}(Z \cup x|A')$.



Φ will be monotone for the partial order \geq . As $(Y, Z) \geq (Y', Z')$, then $\Phi(Y, Z) \geq \Phi(Y', Z')$. Using Tarski fixed-point theorem, there is a (Y, Z) fixed point.

Read Fleiner, T., Jankó, Z., Tamura, A., & Teytelboym, A. (2015). Trading networks with bilateral contracts. arXiv preprint arXiv:1510.01210.

Proposition 8.1

A is trail-stable $\Rightarrow \exists (Y, Z)$ such that $Y \cap Z = A$ and (Y, Z) is a fixed point of Φ .



8.5 Corporate Game Theory

There is a set of players $N = \{1, \dots, n\}$. The subset of players $S \subseteq N$ is called coalition.

There is a value function about coalition $v : 2^N \rightarrow \mathbb{R}$, which assumes $v(N) \geq \max_{S \subseteq N} v(S)$.

Definition 8.19 (Cooperative Game)

A cooperative game is described by the pair $\langle N, v \rangle$.



Definition 8.20 (Transferable Utility)

Utility is transferable if one player can losslessly transfer part of its utility to another player.



Assume a TU (transferable utility) Economy. Consider a payoffs vector for all players, $x \in \mathbb{R}^n$. The efficiency requires $\sum_{i \in N} x_i = v(N)$. Individual Rational (IR) requires $x_i \geq v(\{i\})$.

8.5.1 Core of Corporate Game and Farkas' lemma

Definition 8.21 (Core)

The **core** is the set of feasible allocations where no coalition of agents can benefit by breaking away from the grand coalition.

$$C(v, N) = \left\{ x \in \mathbb{R}^n : \sum_{i \in N} x_i = v(N), \sum_{i \in S} x_i \geq v(S), \forall S \subseteq N \right\}$$



Theorem 8.13 (Bondareva-Shapley Theorem)

?? The core of $\langle N, v \rangle$ is non-empty ($C(v, N) \neq \emptyset$) if and only if for every function $\alpha : 2^N \setminus \{\emptyset\} \rightarrow [0, 1]$ where $\forall i \in N : \sum_{S: i \in S} \alpha(S) = 1$, the following condition holds:

$$\sum_{S \in 2^N \setminus \{\emptyset\}} \alpha(S) v(S) \leq v(N)$$



Consider $B(N)$ be the solutions to:
$$\begin{cases} \sum_{S: i \in S} y_S = 1, & \forall i \in N \\ y_S \geq 0, & \forall S \subseteq N \end{cases}$$

Lemma 8.2 (Farkas' lemma)

Let $A \in \mathbb{R}^{m \times n}$ and $b \in \mathbb{R}^m$. Then, **exactly one** of the following statement is true

- (1). There exists $x \in \mathbb{R}^n$ such that $Ax = b$ and $x \geq 0$
- (2). There exists $y \in \mathbb{R}^n$ such that $A^T y \geq 0$ and $b^T y < 0$.

**Lemma 8.3****Proof 8.8****Lemma 8.4 ((Alternative) Farkas' lemma)**

Let A be $m \times n$ matrix, $b \in \mathbb{R}^m$ and $F = \{x \in \mathbb{R}^n : Ax \geq b, x \geq 0\}$. Then, either $Cx = d$ or $\exists z$

such that for $y_S \geq 0$, $C^T z - A^T y_S = 0$ and such that $d^T z - b^T y_S < 0$, but not both.



By using this lemma, we can conclude $\begin{cases} v(N)z - \sum_S v(S)y_S < 0 \\ z - \sum y_S = 0 \\ y_S \geq 0 \end{cases}$ must hold at the same time, (let $z = 1$, the last two lines are $B(N)$).

Hence, $\forall y_S \in B(N)$, we have $v(N) \geq \sum_S v(S)y_S$.

8.5.2 Doubly stochastic matrix and Birkhoff-von Neumann Theorem

Consider a matching game between sellers and buyers: $v(\{i, j\}) = v_{ij}$, $v(\{i\}) = 0$ for buyer i and $v(\{j\}) \geq 0$ for seller j .

Core:

$$\begin{aligned} \max_{\alpha} \quad & \sum_i \sum_j v_{ij} \alpha_{ij} \\ \text{s.t.} \quad & \sum_i \alpha_{ij} = 1, \forall j \\ & \sum_j \alpha_{ij} = 1, \forall i \\ & \alpha_{ij} \geq 0 \end{aligned}$$

Definition 8.22 (Doubly Stochastic Matrix)

A **doubly stochastic matrix** is a square matrix $X = (x_{ij})$ of non-negative real numbers, each of whose rows and columns sums to 1.

The class of $n \times n$ doubly stochastic matrices is a convex polytope (convex set in euclidean space) known as the **Birkhoff polytope**.



Theorem 8.14 (Birkhoff-von Neumann Theorem)

A matrix is doubly stochastic if and only if it is a convex combination of permutation matrices.



By this theorem, we can set efficient "integer" assignment.

Can the efficient allocation be competitive equilibrium (CE)?

Theorem 8.15

The core of assignment game is non-empty.

**Proof 8.9**

The duality of core can be written as

$$\begin{aligned} \min \quad & \sum_j u_j^S + \sum_i u_i^B \\ \text{s.t.} \quad & u_j^S + u_i^B \geq v_{ij}, \forall i, j \end{aligned}$$

By strong duality, the minimum value should be equal to $V(N)$.

Hence, $\sum_{j \in T} u_j^S + \sum_{i \in T} u_i^B \geq V(T)$ for a subset $T \subseteq N$. That is, the core is non-empty.

Corollary 8.2

For an assignment game, outcome is in the core if and only if the outcome is CE outcome.



8.6 Constrained Demand Theory

8.6.1 Substitutes Valuation

There are buyers $i \in N$ and goods $j \in J$ with quantities $S \in \mathbb{Z}^J$ sold by a seller.

A buyer's utility is $v(x) - p \cdot x$, where $v(0) = 0$, $p \in \mathbb{R}^J$, and $x \in \{0, 1\}^J$. The buyer's demand is represented by $D(p) \arg\max_x \{v(x) - p \cdot x\}$.

The competitive equilibrium $(p^*, (x^{*i})_{i \in N})$ here are

1. $x^{*i} \in D^i(p^*)$ for every $i \in N$ and
2. $\sum_i x^{*i} \leq S_i$, where the equality holds for $p_i > 0$.

Definition 8.23 (Substitutes Valuation)

A valuation v_i is a **substitutes valuation** if $\forall p : p' = p + \lambda e^j$ ($\lambda > 0$), where $D^i(p) = \{x\}$ and $D^i(p') = \{x'\}$, we have that $x'_k \geq x_k$ for all $k \neq j$. (The increase of product j 's price increases other product's demand).

**Theorem 8.16 (Substitutes Valuation \Rightarrow Competitive Equilibrium Exists)**

If agents have substitutes valuations, then a competitive equilibrium exists.



Theorem 8.17

If there exists an agent without substitutes valuation, then we can construct unit-demand preferences for other agents such that no competitive equilibrium exists.

**8.6.2 Income Effect**

There are buyers $i \in N$ and goods $j \in J$. The endowments (money and goods) of agents are denoted by $w = (w_0, w_I)$.

Outcome: The indivisible (bought) goods is represented by $x_I \in \{0, 1\}^J$ and the (left) divisible money is represented by $x_0 \in (\underline{m}, \infty)$.

$$w_0 = x_0 + p_I \cdot x_I$$

must hold, where p_I is the vector of prices of goods.

Utility Function: An agent's utility function is defined by $u^i : (\underline{m}, \infty) \times \{0, 1\}^J \rightarrow (-\infty, +\infty)$ with assumptions of strictly increasing in x_0 , $\lim_{x_0 \rightarrow \underline{m}} u^i(x_0, x_I) = -\infty$, and $\lim_{x_0 \rightarrow \infty} u^i(x_0, x_I) = +\infty$.

Example 8.3 Examples of feasible utility functions:

1. $u^i(x) = v(x) - p \cdot x$ with $\underline{m} = -\infty$;
2. $u^i(x_0, x_I) = \log(x_0) - \log(-V_Q^i(x_I))$ with $V_Q^i : \{0, 1\}^J \rightarrow (-\infty, 0)$.

Demand:

- o $D_{\text{Marshallian}}^i(p, w) = \{x^* : x^* \in \arg \max_x u^i(x) \text{ s.t. } p \cdot x \leq p \cdot w\}$
- o $D_{\text{Hicksian}}^i(p, u) = \{x^* : x^* \in \arg \min_x p \cdot x \text{ s.t. } u^i(x) \geq u\}$ which is the dual of $D_{\text{Marshallian}}^i$.

Definition 8.24 (Competitive Equilibrium)

Given $(w^i)_{i \in I}$ s.t. $\sum_{i \in N} w_I^i = y_I$. A **competitive equilibrium** is a price vector $p_I^* \in \mathbb{R}^J$ and $x_I^{i*} \in D_{\text{Marshallian}}(p_I^*, w^i)$ for each $i \in N$ such that $\sum_{i \in N} x_I^{i*} = y_I$.



Based on the idea of duality, we can analyze problem based on the dual demand, Hicksian demand.

Definition 8.25 (Hicksian Valuation)

Hicksian valuation is defined by -1 times "the money that can lead to the utility u with goods x_I ":

$$V_{\text{Hicksian}}^i(x_I, u) = -(u^i(\cdot, x_I))^{-1}(u)$$

**Proposition 8.2 (Using Hicksian Valuation to Represent Hicksian Demand)**

$$D_{\text{Hicksian}}^i(p_I, u) = \arg \max_{x_I} \{v_{\text{Hicksian}}^i(x_I, u) - p_I \cdot x_I\}$$

**Proof 8.10**

$$D_{\text{Hicksian}}^i(p_I, u) = \arg \min_{x_I} \{(u^i(\cdot, x_I))^{-1}(u) + p_I \cdot x_I\} = \arg \max_{x_I} \{V_{\text{Hicksian}}^i(x_I, u) - p_I \cdot x_I\}$$

Definition 8.26 (Hicksian Economy)

Hicksian economy: for a profile $(u^i)_{i \in N}$ is a transferable utility (TU) economy in which each agent's "valuation" is a Hicksian valuation V_{Hicksian}^i .



Hicksian Economy works in finding Competitive Equilibrium

Theorem 8.18 (Equilibrium Existence Duality(EED))

Competitive Equilibrium exists for all feasible endowment profiles if and only if Competitive Equilibrium exists in the Hicksian economies for all profiles of utility levels.



Marshallian	Hicksian
Housing Market	Assignment Game
Utility is not Quasi-linear	Utility is Quasi-linear
Unit Demand	Unit Demand
Existence in Housing Market	Existence in Assignment Game
×	Lattice structure and Convexity of structure of CE prices
Net-substitutes	\Rightarrow Substitutes

Like the Theorem 8.16, we want the Hicksian valuations be "substitutes".

Definition 8.27 (Net-Substitutes)

A agent's utility u^i is net-substitutes if $\forall u, D_H^i(p; u) = \{x\}$ and $D_H^i(p'_j, p_{-j}; u) = \{x'\}, p'_j > p_j \Rightarrow x'_k \geq x_k$ for all $k \neq j$.

**Theorem 8.19**

Net-Substitutes Valuation \Rightarrow competitive equilibrium exists.

**Proof 8.11**

Net-substitutes \Rightarrow substitutes holds in Hicksian economy. Hence, CE exists. By 8.18, CE exists in original economy.

Definition 8.28 (Gross-Substitutes)

A agent's utility u^i is gross-substitutes if $\forall w, D_M^i(p; w) = \{x\}$ and $D_M^i(p'_j, p_{-j}; w) = \{x'\}, p'_j > p_j \Rightarrow x'_k \geq x_k$ for all $k \neq j$.



Example 8.4 In quasi housing market, we consider an example, of holding a house which price increases, the demand of another bad house doesn't change under Hicksian demand, which makes net-substitutes hold. But, the Marshallian demand decreases, which makes gross-substitutes don't hold.

Example 8.5 Net, but not gross:

Suppose there is a firm f thinking about workers s_1, s_2 . f values worker at \$5 each, and the hiring budget is \$6;

- $p_1 = 2, p_2 = 4$;
- $p_1 = 3, p_2 = 4$

Obviously, the gross-substitutes (Marshallian Demand) leads to hiring both under $p_1 = 2, p_2 = 4$ and only hiring s_1 under $p_1 = 3, p_2 = 4$.

Let's consider the net-substitutes (Hicksian Demand): As the utility given under $p_1 = 2, p_2 = 4$ is \$10. We can find hiring two workers is still the optimal strategy.

Example 8.6 Net, but no auction:

Suppose there are two identical firms f_1, f_2 and workers s_1, s'_1, s_2 . The value of workers is \$5 each, but a firm want at most one of s_1, s'_1 and has hiring budget \$6. A worker has reservation wage of \$1.

Equilibrium: \$1 for worker s_1, s'_1 and \$5 for s_2 ; One firm hires one of s_1, s'_1 and the other hires s_2 .

8.7 Object Allocation

Exchange: $i \in N$ agent; Agents have strict preference \succ_i over objects. (We use \succ denote $\{\succ_i\}_{i \in N}$).

Two settings:

1. Exchange: an agent shows up with exactly one object.
2. Allocation: One planner owns N objects; agents have \emptyset .

A **mechanism** $\Phi(\succ)$ gives a outcome μ . We want the final outcome μ be

1. Individual Rationality (IR): for all $i \in N$, $\mu_i \succeq i$ (Exchange) and $\mu_i \succeq \emptyset$ (Allocation).
2. Pareto Efficient (PE): $\nexists \mu'$ such that $\mu'_i \succeq \mu_i$ for all $i \in N$, strict for at least one.
3. Strategy-Proof (SP): Φ induces a game. We want that, in this game, truth-telling is a weakly dominant strategy for all agent $i \in N$.

8.7.1 Allocation

(Random) Serial Dictatorship: Randomly order the agents, ask one by one, and allocate a remaining object. \Rightarrow it satisfies IR, PE, SP, but unfair(?).

8.7.2 Exchange

Definition 8.29 (Core)

The **core** is the set of all allocations μ such that there is no $S \subseteq N$ and μ' for which:

- o for $i \in S$, $\mu'_i = j$ for some $j \in S$
- o $\mu'_i \succeq \mu_i$ for all $i \in S$, at least one strict.

Core: IR+PE.



Theorem 8.20 (Core is a Singleton)

There is a unique element in the core.



Proof 8.12

Run the algorithm: Top Trading Cycles (TTC).

Definition 8.30 (Top Trading Cycles (TTC))

Agent = node.

1. Step 1: every agent point at her favorite object/agent.
 - (1A): Find cycles.
 - (1B): Allocate object to agent who is pointing at it in cycle.
 - (1C): Remove the cycle.
2. Step 2: every (remaining) agent point at her favorite object/agent.
 - (2A): Find cycles.
 - (2B): Allocate object to agent who is pointing at it in cycle.
 - (2C): Remove the cycle.
3. Repeat ...



Proposition 8.3

TTC produces an allocation that satisfies IR, PE, SP.



Theorem 8.21 (TTC \Leftrightarrow IR, PE, SP (Ma, 1999))

There is at most 1 IR, PE, SP mechanism (TTC).



Proof 8.13

Definition 8.31

The **size** of a preference profile \succ is the total number of objects agents find acceptable in \succ :

$$S(\succ) = \sum_{i \in N} \#\text{acceptable objects in } \succ_i$$



Consider two Φ and Ψ that disagree for some \succ , the \succ is defined to be bad.

We define the minimal bad profile as a bad profile of minimal size. Consider the two outcomes given by these mechanisms:

$\Phi(\succ)$	same	$A(\Phi)$
$\Psi(\succ)$	same	$A(\Psi)$

the sum of different parts are $A \triangleq A(\Phi) + A(\Psi)$.

Lemma 8.5

If Φ and Ψ are SP, and \succ is a minimal bad profile, then each agent in A has exactly two acceptable objects.



Proof 8.14

Suppose there exists $i \in A$ such that she has > 2 acceptable objects.

Without loss of generality, we consider $\Phi_i(\succ) \succ_i \Psi_i(\succ)$.

Remove all objects from his preference list except $\Phi_i(\succ)$ and endowment of i (call it $\{i\}$). The new preference profile is denoted by \succ'_i .

Since Φ is SP, $\Phi_i(\succ') = \Phi_i(\succ)$; since Ψ is SP, $\Psi_i(\succ') \prec_i \Phi_i(\succ)$.

So, we have \succ' is a bad profile and $S(\succ') < S(\succ)$, a contradiction.

8.8 School Choice

Model:

1. There is a set of schools S ; a school is denoted by $s \in S$; Quota for each s is q_s ;
2. I is the set of all students; A student is denoted by $i \in I$; Student i has preference \succ_i .
3. School places = objects.
4. Each school has a priority order over students π_s .
5. Matching $\mu : I \rightarrow S$ such that $\forall s \in S : \#\mu^{-1}(s) \leq q_s$.
6. Matching violates priority if $\exists s \in S$ such that

(i). $s \succ_i \mu(i)$ and

(ii). either “Wastefulness: $\#\mu^{-1}(s) < q_s$ ” or “Justified Envy: $i \pi_{s,j} \text{ for some } j \in \mu^{-1}(s)$ ”

≈ existence of a blocking pair.

A matching is **stable** if there are no priority violations.

(As we don't consider the preference of j in (ii), it is not true stable \Rightarrow (Pareto) efficient.)

Example 8.7 Boston (Immediate Acceptance)

- (1). Step 1: students apply for favorite schools; school accepts applicants up to capacity and reject rest permanently.
- (2). Step k: students apply for favorite schools among those with capacity and hasn't already rejected them; schools accept applicants up to capacity q_s and reject rest permanently.

Proposition 8.4

DA gives a matching that satisfies stability and SP (not PE).



Run TTC:

Definition 8.32 (Top Trading Cycles (TTC))

Schools and Students (agents) = nodes.

1. Step 1: every agent point at her favorite object/agent.
 - (1A): Find cycles.
 - (1B): Allocate object (school) to agent (student) who is pointing at it in cycle. (Usually based on the students' preference.)
 - (1C): Remove the cycle.
2. Step 2: every (remaining) agent point at her favorite object/agent.
 - (2A): Find cycles.
 - (2B): Allocate object to agent who is pointing at it in cycle.
 - (2C): Remove the cycle.
3. Repeat ...

**Proposition 8.5**

TTC produces an allocation that satisfies PE and SP (not stable).



Hence, we need to make a trade-off between priority violation and efficiency.

Theorem 8.22 (Keslen)

For all $S, \{q_s\}_{s \in S}$, there exists $I, \succ_i, \{\pi_s\}_{s \in S}$ s.t. in the SOSM, every student gets either their last choice or second-last choice.

**Theorem 8.23 (Abdulkadiroğlu, Pathak, Roth, AER)**

There is no (PE+)SP mechanism that Pareto-dominates SOSM.

**Theorem 8.24**

There is no PE+SP mechanism that selects a PE+stable matching whenever it exists.



Definition 8.33 (Kesten/Tang+Yu Algorithm)

Suppose the number of student is not larger than the total capacity $\#I \leq \sum_s q_s$.

- (i). Step 0: Run DA, set SOSM μ_0 . Find under-demanded schools = a school that doesn't reject any students.

Assign $\mu^{-1}(s)$ permanently. Call these schools/students “settled”. Remove all settled schools and students.

- (ii). Step k: Rerun DA on everyone unsettled.

**Definition 8.34 (Priority-Neutral(PN), Reny 2022)**

μ is **priority-neutral(PN)** iff \exists no matching u that can make any student whose priority is violated at μ better off unless u violates the priority of some student and make them worse off.

We call μ is **priority-efficient** if it is PN and PE.

**Theorem 8.25 (Reny 2022)**

1. \exists a unique Priority-efficient matching;
2. Priority efficient \Leftrightarrow SO priority neutral matching;
3. It can be found by the CUTE Algorithm;
4. μ is priority efficient \Leftrightarrow no matching u can make any student better off unless u unless u violates the priority of some student and make them worse off.



8.9 School Choice with Reserves

Consider a school choice model, students can be divided into majority (M) and minority (m), $I = I^M \cup I^m$.

Quotas of schools are represented by $q_s = (q, q^M)$, $s \in S$, where q^M is the quota for majority.

Definition 8.35 (Stability)

A matching is stable if, for all $s \in S$ such that $s \succ_i \mu(i)$,

1. Either: “No Wastefulness: $|\mu^{-1}(s)| = q_s$ ” and “No Justified Envy: $i' \pi_s i$ for all $i' \in \mu^{-1}(s)$ ”
2. Or: $i \in I^M$, “ $|\mu^{-1}(s) \cap I^M| = q_s^M$ ” and “ $i' \pi_s i$ for all $i' \in \mu^{-1}(s) \cap I^M$ ”,

**Definition 8.36 (Stronger Quota)**

A \tilde{q}_s has **stronger quota** than q_s if $\tilde{q}_s = q_s$ but $q_s^M \geq \tilde{q}_s^M$.

**Definition 8.37 (Good Mechanism)**

Mechanism Φ is **good**, if whenever a \tilde{q}_s has stronger quotas than its setting, it doesn't make all minority students worse off.



Theorem 8.26 (Kojima 2012)

There is no stable good mechanism.

**8.9.1 Minority Reserves (slot-specific priority)**

Suppose r_s^m is reserved for minority only. That is $q_s = q_s^M + r_s^m$.

Definition 8.38 (Minority Reserves)

School has minority reserve r_s^m whenever # of admitted minority students is less than r_s^m , then any minority students is admitted ahead of majority students.

**Definition 8.39 (No Blocking Pair)**

No blocking pair if $s \succ_i \mu(i)$, then $|\mu(s)| = q_s$ and,

1. Either: $i \in I^m$ and “ $i' \pi_s i$ for all $i' \in \mu^{-1}(s)$ ”
2. Or: $i \in I^M$, “ $|\mu^{-1}(s) \cap I^m| > r_s^m$ ” and “ $i' \pi_s i$ for all $i' \in \mu^{-1}(s)$ ”
3. Or: $i \in I^M$, “ $|\mu^{-1}(s) \cap I^m| \leq r_s^m$ ” and “ $i' \pi_s i$ for all $i' \in \mu^{-1}(s) \cap I^M$ ”

**Theorem 8.27 (Smart Reserves)**

Suppose μ is a stable matching without affirmative action. Let r_s^m be such that

$$r_s^m \geq |\mu^{-1}(s) \cap I^m|, \forall s \in S$$

Then, either μ is stable w.r.t. r^m or \exists stable matching under r^m that Pareto-dominates μ for I^m .

**8.10 Random Assignment**

Suppose there are agents $i \in I$ and objects $j \in J$, where $|I| = |J|$. Agents have preferences \succ_i over objects, and objects have priorities \triangleright_j over agents.

An allocation is represented by a matrix that each row and each column has sum to 1 probability.

There are two mechanism can be used:

- (i). RSD (Random: draw a priority order \triangleright uniformly.)
- (ii). TTC with uniform random endowment.

Theorem 8.28

These two mechanisms are equivalent (bijection).



RSD is not Pareto-efficiently.

Proposition 8.6

For a row of an allocation matrix $(\tilde{\mu})$ for agent i , $\tilde{\mu}_i \succ_i \tilde{\mu}'_i$

- if and only if $\tilde{\mu}_i \succ_{FOSD} \tilde{\mu}'_i$ (first-order stochastic dominance).
- if and only if $\mathbb{E}U(\tilde{\mu}_i) \geq \mathbb{E}U(\tilde{\mu}'_i)$ under expected utility.

**Definition 8.40**

$\tilde{\mu}$ is **ordinally efficient (sd-efficient)** if there is no $\tilde{\mu}'$ which is \succ_{FOSD} by all agents. (*ex-ante efficient* with respect to cardinal utility)

$\tilde{\mu}$ is **ex-post efficient** if those are only Pareto efficient outcome in the support.

**Definition 8.41**

$\tilde{\mu}$ is **ordinally envy-free** if $\tilde{\mu}_i \succ_{FOSD} \tilde{\mu}_j, \forall i, j$.



RSD is not envy-free.

There exists ordinally efficient and envy-free mechanism.

Definition 8.42 (Probabilistic Serial Algorithm)

Based on the preference of agents:

1. Give each agent his most preferred object with the same proportion such that the sum of each object is at most 1.
2. Repeat by using remaining objects.

Example 8.8 Preference: A: $Obj1 \succ Obj3 \succ Obj2$; B: $Obj1 \succ Obj2 \succ Obj3$; C: $Obj2 \succ Obj3 \succ Obj1$

$$t = \frac{1}{2} \quad A: \frac{1}{2}Obj1; B: \frac{1}{2}Obj1; C: \frac{1}{2}Obj2.$$

$$t = \frac{3}{4} \quad A: \frac{1}{2}Obj1 + \frac{1}{4}Obj3; B: \frac{1}{2}Obj1 + \frac{1}{4}Obj2; C: \frac{3}{4}Obj2.$$

$$t = 1 \quad A: \frac{1}{2}Obj1 + \frac{1}{2}Obj3; B: \frac{1}{2}Obj1 + \frac{1}{4}Obj2 + \frac{1}{4}Obj3; C: \frac{3}{4}Obj2 + \frac{1}{4}Obj3.$$

**Theorem 8.29 (Welfare Theorem)**

Probabilistic Serial Algorithm gives ordinally efficient and envy-free outcome.

**Definition 8.43 (Equal Treatment of Equals (ETE))**

Equal Treatment of Equals: if same preference $\succ_i \Rightarrow$ the same bundle $\tilde{\mu}_i$.

**Proposition 8.7**

For $n = 3$, RSD is ordinally efficient, ETE, Strategy-Proof. (These three properties are incompatible when $n > 3$).



8.11 Random Assignment in School Choice

Example 8.9

- Preference of Agents: $A : s_2 \succ s_3 \succ s_1$; $B : s_2 \succ s_3 \succ s_1$; $C : s_1 \succ s_2 \succ s_3$.
- Priority of Schools: $s_1 : A \succ B \succ C$, $s_2 : C \succ (A, B)$, $s_3 : C \succ B \succ A$.

There are two stable outcomes: $\mu : A - s_2, B - s_3, C - s_1$; $\mu' : A - s_3, B - s_2, C - s_1$.

It can't be strategy proof. In μ , B can lie: $s_2 \succ s_1 \succ s_3$, to make the outcome become μ' . In μ' , A can lie: $s_2 \succ s_1 \succ s_3$, to make the outcome become μ .

Definition 8.44 (Stable Improvement Cycle (S.I.C.))

Each student points at schools they prefer and where he doesn't have a lower priority among those students who prefer students to their assignment.



Theorem 8.30

If a stable matching is not in the student-optimal stable set, then \exists a S.I.C.



Example 8.10

- Preference of Agents: $A : s_2 \succ s_1 \succ s_3$; $B : s_3 \succ s_2 \succ s_1$; $C : s_2 \succ s_3 \succ s_1$.
- Priority of Schools: $s_1 : A \succ (B, C)$, $s_2 : B \succ (A, C)$, $s_3 : C \succ (A, B)$.

DA: $A : s_1, B : s_2, C : s_3$. Another allocation: $A : s_1, B : s_3, C : s_2$.

Consider DA, A wants s_2 ; C also wants s_2 , which has the same priority as A , so A can point at s_2 . B points at s_3 . C can also point at s_2 . So, there is a S.I.C.

8.12 Pseduomarket (FF)

Consider an example that agent A_1 wants a, b for 0.9, A_2 wants a, c for 0.9, A_3 wants b, c for 2. Suppose the budget for each agent is 1.

Reminds that utility is only meaningful for the agent itself. Here, as the budget is the same, the demand of each agent is the same.

8.12.1 Problem of Implementability

An equilibrium (but can't be implemented): A_1 gets $\{\frac{1}{2} : \emptyset; \frac{1}{2} : a + b\}$; A_2 gets $\{\frac{1}{2} : \emptyset; \frac{1}{2} : a + c\}$; A_3 gets $\{\frac{1}{2} : \emptyset; \frac{1}{2} : b + c\}$.

Transfer Utility Economy	Pseduomarket
Allocation $x_j \in X_j, j = 1, \dots, J$	Lottery $\tilde{x}_j \in \mathcal{L}(X_j)$
Price $p \in \mathbb{R}^I$	Budget b_j and Price $p \in \mathbb{R}^I$
$u_j(x) = v_j(x) - p \cdot x$	$V_j(\tilde{x}_j) = \sum_x v_j(x) \mathbb{P}(\tilde{x}_j = x)$
Demand $D_j(p) = \arg \max_x u_j(x)$	$\tilde{D}_j(p) = \arg \max_{\tilde{x}: p \cdot \tilde{x} \leq b_j} V_j(\tilde{x})$
CE: $(p^*, x^*) : x_j^* \in D_j(p^*), \sum_j x_j^* \leq S$ (equality holds for no zero p^*)	RE: $(p^*, \tilde{x}^*) : \tilde{x}_j^* \in \tilde{D}_j(p^*), \sum_j \tilde{x}_j^* \leq S$
S is supply, which equals to $\sum_i \omega_i$ if the economy with endowments.	

We want an allocation being implementable than an allocation (a set of lotteries over agents) $\{w_1, \dots, w_J\} = \mathcal{W} \in \mathcal{L}(\prod_j X_j)$ (feasible bundles for each agent).

Define $\bar{w}_j = \mathbb{E}[w_j]$ and $\bar{\mathcal{W}} = \mathbb{E}[\mathcal{W}]$

Definition 8.45 (Implementable)

A random equilibrium (p^*, \tilde{x}^*) is **implementable** if there exists \mathcal{W} over feasible allocations such that $w_j \in D_j(p^*)$ and $\tilde{x}_j^* = \bar{w}_j, \forall j = 1, \dots, J$.



can be implemented by a distribution of allocations. (BvN)

Proposition 8.8

Random equilibrium always exists.



Definition 8.46 (Rich)

A set of valuations $\mathcal{V}^j = \{v_j(x) : x \in X_j\}$ is **rich** if whenever $v_j(x) \in \mathcal{V}^j$ then $v_j(x) + a \cdot x \in \mathcal{V}^j$ for all $a \in \mathbb{R}^I$. That is $\exists x' \text{ such that } v_j(x') = v_j(x) + a \cdot x$.



Complement may induce unimplementable problem.

Suppose value functions live in V and are rich.

Theorem 8.31

CE exists for all valuations in $V \Leftrightarrow$ RE is implementable for all budgets profiles and all valuations in V .

