

## 4. Lattices and Supermodularity

Daisuke Oyama

Mathematical Economics

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# Partially Ordered Sets

## Definition 4.1

A binary relation  $\preceq$  on a set  $X$  is a *partial order* if it satisfies the following:

- ▶ **Transitivity:**  
for all  $x, y, z \in X$ , if  $x \preceq y$  and  $y \preceq z$ , then  $x \preceq z$ .
- ▶ **Reflexivity:**  
for all  $x \in X$ ,  $x \preceq x$ .
- ▶ **Antisymmetry:**  
for all  $x, y \in X$ , if  $x \preceq y$  and  $y \preceq x$ , then  $x = y$ .
  
- ▶ A *partially ordered set* (or *poset*) is a set  $X$  with a partial order  $\preceq$  on  $X$ , denoted  $(X, \preceq)$ .

# Examples

- ▶  $(\mathbb{R}, \leq)$ , where  $\leq$  is the usual order on  $\mathbb{R}$ .

In fact, it is a totally ordered set:

$\leq$  also satisfies *completeness*: for all  $x, y \in \mathbb{R}$ ,  $x \leq y$  or  $y \leq x$ .

- ▶  $(\mathbb{R}^n, \leq)$ , where  $\leq$  is the vector order on  $\mathbb{R}^n$ .

- ▶  $(2^X, \subset)$ , where  $2^X$  is the set of all subsets of a set  $X$ , and  $\subset$  is set inclusion.

## Upper/lower bounds, ...

- ▶ Let  $(X, \preceq)$  be a partially ordered set, and let  $S \subset X$ .
- ▶  $x \in X$  is an *upper bound* of  $S$  if  $y \preceq x$  for all  $y \in S$ .  
 $x \in X$  is a *lower bound* of  $S$  if  $x \preceq y$  for all  $y \in S$ .
- ▶  $x \in X$  is a *greatest* (or *largest*) *element* of  $S$  if  $x \in S$ , and  $x$  is an upper bound of  $S$ .  
 $x \in X$  is a *least* (or *smallest*) *element* of  $S$  if  $x \in S$ , and  $x$  is a lower bound of  $S$ .
- ▶  $x \in X$  is a *maximal element* of  $S$  if  $x \in S$ , and  $x \preceq y$  and  $y \in S \implies y = x$ .  
 $x \in X$  is a *minimal element* of  $S$  if  $x \in S$ , and  $y \preceq x$  and  $y \in S \implies y = x$ .

- ▶ If the set of upper bounds of  $S$  has a least element, then it is called the *least upper bound*, or *supremum*, of  $S$ , and denoted  $\sup_X S$ .

That is,  $x = \sup_X S$  if and only if

1.  $y \preceq x$  for all  $y \in S$ ; and
2. if  $y \preceq z$  for all  $y \in S$ , then  $x \preceq z$ .

- ▶ If the set of lower bounds of  $S$  has a greatest element, then it is called the *greatest lower bound*, or *infimum*, of  $S$ , and denoted  $\inf_X S$ .

That is,  $x = \inf_X S$  if and only if

1.  $x \preceq y$  for all  $y \in S$ ; and
2. if  $z \preceq y$  for all  $y \in S$ , then  $z \preceq x$ .

- ▶  $\sup_X S$  is a greatest (least) element of  $S$  if and only if  $\sup_X S \in S$  ( $\inf_X S \in S$ ).

## (Abstract) Lattices

- ▶ For  $x, y \in X$ , write

$$x \vee_X y = \sup_X \{x, y\}, \quad x \wedge_X y = \inf_X \{x, y\}.$$

(If there is no risk of confusion, we just write  $x \vee y$  and  $x \wedge y$ .)

### Definition 4.2

A partially ordered set  $(X, \preceq)$  is a *lattice* if  $x \vee_X y$  and  $x \wedge_X y$  exist for all  $x, y \in X$ .

## Example

- ▶  $(\mathbb{R}, \leq)$  is a lattice.
  - ▶  $x \vee y = \max\{x, y\}$ ,  $x \wedge y = \min\{x, y\}$
- ▶  $(\mathbb{R}^n, \leq)$  is a lattice.
  - ▶  $x \vee y \in \mathbb{R}^n$ : the vector such that  $(x \vee y)_i = \max\{x_i, y_i\}$
  - ▶  $x \wedge y \in \mathbb{R}^n$ : the vector such that  $(x \wedge y)_i = \min\{x_i, y_i\}$
- ▶  $(2^X, \subset)$  is a lattice.
  - ▶  $S \vee T = S \cup T$ ,  $S \wedge T = S \cap T$
- ▶  $X = \{(0, 0), (1, 0), (0, 1)\} \subset \mathbb{R}^2$   
 $(X, \leq)$  is not a lattice.
  - ▶ The set of upper bounds of  $\{(1, 0), (0, 1)\}$  is empty, so  $(1, 0) \vee_X (0, 1)$  does not exist.

▶  $X = \{(0, 0), (1, 0), (0, 1), (2, 2)\} \subset \mathbb{R}^2$

$(X, \leq)$  is a lattice.

▶  $(1, 0) \vee_X (0, 1) = (2, 2)$

▶ Note that  $(1, 0) \vee_{\mathbb{R}^2} (0, 1) = (1, 1)$ .

$\implies X$  is not a *sublattice* of  $(\mathbb{R}^2, \leq)$  (to be defined later).

▶  $X = \{(0, 0), (1, 0), (0, 1)\} \cup \{(x_1, x_2) \mid x_1 = x_2, x_1 > 2\} \subset \mathbb{R}^2$

$(X, \leq)$  is not a lattice.

▶ The set of upper bounds of  $\{(1, 0), (0, 1)\}$ ,

$\{(x_1, x_2) \mid x_1 = x_2, x_1 > 2\}$ , does not have a least element.



# Sublattices

## Definition 4.3

For a lattice  $(X, \lesssim)$ ,  $K \subset X$  is a *sublattice* of  $(X, \lesssim)$  if  $x \vee_X y \in K$  and  $x \wedge_X y \in K$  for all  $x, y \in K$ .

- ▶ If  $K \subset X$  is a sublattice of  $(X, \lesssim)$ , then  $(K, \lesssim)$  is a lattice, but not vice versa.
- ▶ (According to this definition, Definition 7.1 in the textbook is defining a sublattice of  $\mathbb{R}^n$ .)
- ▶  $X = \{(0, 0), (1, 0), (0, 1), (1, 1)\}$  is a sublattice of  $(\mathbb{R}^2, \leq)$ .
- ▶  $X = \{(0, 0), (1, 0), (0, 1), (2, 2)\}$  is not a sublattice of  $(\mathbb{R}^2, \leq)$ .

# Complete Lattices

## Definition 4.4

A lattice  $(X, \lesssim)$  is *complete* if  $\sup_X S$  and  $\inf_X S$  exist for all  $S \subset X$

(where  $\sup_X \emptyset = \inf_X X$  and  $\inf_X \emptyset = \sup_X X$  by convention).

- ▶ This property is called “compact” in the textbook.  
We follow the “standard” terminology here.
- ▶ Any lattice  $(X, \lesssim)$  with finite  $X$  is a complete lattice, but not always if  $X$  is infinite.
- ▶  $X = [0, 1] \subset \mathbb{R}$   
 $(X, \leq)$  is a complete lattice.
- ▶  $X = [0, 1) \subset \mathbb{R}$   
 $(X, \leq)$  is not a complete lattice.

▶  $X = [0, 1) \cup \{2\} \subset \mathbb{R}$

$(X, \leq)$  is a complete lattice.

▶  $\sup_X [0, 1) = 2$

▶ Note that  $\sup_{\mathbb{R}} [0, 1) = 1$ .

$\implies X$  is not a complete sublattice of  $(\mathbb{R}, \leq)$ .

# Complete sublattices

## Definition 4.5

For a lattice  $(X, \preceq)$ ,  $K \subset X$  is a complete sublattice of  $(X, \preceq)$  if  $\sup_X S$  and  $\inf_X S$  exist in  $K$  for all  $S \subset K$ ,  $S \neq \emptyset$ .

# Complete (Sub-)Lattices in $\mathbb{R}^n$

## Proposition 4.1

1. *If  $K \subset \mathbb{R}^n$ ,  $K \neq \emptyset$ , is compact and  $(K, \leq)$  is a lattice, then  $(K, \leq)$  is a complete lattice.*
2. *For  $X \subset \mathbb{R}^n$ , suppose that  $(X, \leq)$  is a lattice.  
If  $K \subset X$  is compact and a sublattice of  $(X, \leq)$ , then  $K$  is a complete sublattice of  $(X, \leq)$ .*

► Part 1 is a special case of part 2. (Let  $K = X$ .)

## Proof

2.

- ▶ Let  $S \subset K$ ,  $S \neq \emptyset$ .

We want to show that  $\sup_X S$  exists in  $K$ .

(The existence of  $\inf_X S$  in  $K$  can be shown symmetrically.)

- ▶ Let  $U \subset X$  be the set of upper bounds of  $S$  in  $X$ :  
 $U = \{u \in X \mid s \leq u \text{ for all } s \in S\}$ .

For the moment, assume that  $U \neq \emptyset$ . We prove this later.

- ▶ For  $(s, u) \in S \times U$ , write  $[s, u] = \{x \in \mathbb{R}^n \mid s \leq x \leq u\}$ , which is a closed set.
- ▶ We want to show that  $\bigcap_{(s,u) \in S \times U} [s, u] \cap K \neq \emptyset$ .

If  $\underline{u} \in \bigcap_{(s,u) \in S \times U} [s, u] \cap K$ , then

- ▶  $\underline{u} \in K$  ( $\subset X$ );
- ▶  $s \leq \underline{u}$  for all  $s \in S$ :  $\underline{u}$  is an upper bound of  $S$ ;
- ▶  $\underline{u} \leq u$  for all  $u \in U$ .  $\therefore \sup_X S = \underline{u} \in K$ .

- ▶ Take any  $(s^1, u^1), \dots, (s^K, u^K) \in S \times U$ .
- ▶ Since  $X$  is a lattice,  $\bar{s} = \sup_X \{s^1, \dots, s^K\}$  exists in  $X$ .
- ▶ Since  $K$  is a sublattice of  $X$ ,  $\bar{s} \in K$ .
- ▶ For each  $k = 1, \dots, K$ , since  $u^k$  is an upper bound of  $\{s^1, \dots, s^K\}$ , we have  $\bar{s} \leq u^k$ .
- ▶ Therefore  $\bigcap_{k=1}^K [s^k, u^k] \cap K \neq \emptyset$ .
- ▶ By the compactness of  $K$ , this implies that  $\bigcap_{(s,u) \in S \times U} [s, u] \cap K \neq \emptyset$ .
- ▶ Finally, we show that  $U \neq \emptyset$ .
  - ▶ Write  $[s, \infty) = \{x \in \mathbb{R}^n \mid s \leq x\}$ , which is a closed set.
  - ▶ By the compactness of  $K$ , a similar argument as above shows that  $\bigcap_{s \in S} [s, \infty) \cap K \neq \emptyset$ .
  - ▶ Thus  $U = \bigcap_{s \in S} [s, \infty) \cap X \supset \bigcap_{s \in S} [s, \infty) \cap K \neq \emptyset$ .

# Complete (Sub-)Lattices in $\mathbb{R}^n$

## Proposition 4.2

*For a sublattice  $K \subset \mathbb{R}^n$  of  $(\mathbb{R}^n, \leq)$ ,  $K$  is a complete sublattice of  $(\mathbb{R}^n, \leq)$  if and only if it is a compact set.*



## Proof

- ▶ “If” part:

Follows from Proposition 4.1.

- ▶ “Only if” part:

Boundedness:  $K$  is contained in a bounded set  $\{x \in \mathbb{R}^n \mid \inf_{\mathbb{R}^n} K \leq x \leq \sup_{\mathbb{R}^n} K\}$ .

Closedness: If  $\{x^k\} \subset K$  and  $x^k \rightarrow x^*$ , then let  $y^k = \inf_{\mathbb{R}^n} \{x^m\}_{m \geq k} \in K$ , and let  $\bar{y} = \sup_{\mathbb{R}^n} \{y^k\} \in K$ .

Show that  $x^* = \bar{y}$ :

- ▶ For any  $\varepsilon > 0$ , there exists  $k$  such that  $\bar{y} - \varepsilon \mathbf{1} \leq y^k$ , and hence  $\bar{y} - \varepsilon \mathbf{1} \leq x^m$  for all  $m \geq k$ . Therefore,  $\bar{y} - \varepsilon \mathbf{1} \leq x^*$ . Since  $\varepsilon > 0$  is arbitrary, this implies  $\bar{y} \leq x^*$ .
- ▶ For any  $\varepsilon > 0$ , there exists  $k$  such that  $x^* - \varepsilon \mathbf{1} \leq x^m$  for all  $m \geq k$ , and hence  $x^* - \varepsilon \mathbf{1} \leq y^k$ . Therefore,  $x^* - \varepsilon \mathbf{1} \leq \bar{y}$ . Since  $\varepsilon > 0$  is arbitrary, this implies that  $x^* \leq \bar{y}$ .

## Tarski's Fixed Point Theorem

- ▶ For partially ordered sets  $(X, \preceq_X)$  and  $(Y, \preceq_Y)$ , a function  $f: X \rightarrow Y$  is *non-decreasing* (or *isotone*, or *order-preserving*) if  $f(x) \preceq_Y f(x')$  whenever  $x \preceq_X x'$ .

### Proposition 4.3 (Tarski's Fixed Point Theorem)

Suppose that  $(X, \preceq)$  is a complete lattice, and that  $f: X \rightarrow X$  is a non-decreasing function. Let  $X^* \subset X$  be the set of fixed points of  $f$ .

1.  $\sup\{x \in X \mid x \preceq f(x)\}$  and  $\inf\{x \in X \mid f(x) \preceq x\}$  are the greatest and the least elements of  $X^*$ .  
In particular,  $X^* \neq \emptyset$ .
2.  $(X^*, \preceq)$  is a complete lattice.

- ▶  $X^*$  is not a sublattice in  $X$  in general.

# Proof

1.

▶ Let  $X' = \{x \in X \mid x \preceq f(x)\}$ .

$X' \neq \emptyset$  since  $\inf X \in X'$ .

▶ Denote  $x^* = \sup X' \in X$ .

We show that  $x^* \in X^*$ .

▶ Take any  $x \in X'$ , where  $x \preceq f(x)$  and  $x \preceq x^*$ .

By the monotonicity of  $f$ , we have  $f(x) \preceq f(x^*)$ , so that  $x \preceq f(x^*)$ .

Since this holds for any  $x \in X'$ , we have

$$x^* \preceq f(x^*). \tag{1}$$

- ▶ By the monotonicity of  $f$ , (1) implies that  $f(x^*) \lesssim f(f(x^*))$ . This means that  $f(x^*) \in X'$ .

Therefore, we have

$$f(x^*) \lesssim x^*. \tag{2}$$

- ▶ By (1) and (2), we have  $x^* = f(x^*)$ , i.e.,  $x^* \in X^*$ .
- ▶ For any  $x \in X^*$ , we have  $x \in X'$ , and therefore  $x \lesssim x^*$ .

Thus,  $x^*$  is the greatest element of  $X^*$ .

- ▶ A symmetric argument shows that  $\inf\{x \in X \mid f(x) \lesssim x\}$  is the least element of  $X^*$ .

2.

- ▶ Take any  $S \subset X^*$ .
- ▶ Denote  $\bar{s} = \sup_X S \in X$  and  $Z = \{x \in X \mid \bar{s} \preceq x\}$ .  
 $(Z, \preceq)$  is a complete lattice.
- ▶ We have  $f(Z) \subset Z$ .
  - ▶ Take any  $z \in Z$ .
  - ▶ For any  $x \in S$  ( $\subset X^*$ ), we have  $x = f(x) \preceq f(\bar{s}) \preceq f(z)$ .
  - ▶ This shows that  $\bar{s} = \sup_X S \preceq z$ , i.e.,  $f(z) \in Z$ .
- ▶ Thus, the restriction  $f|_Z$  of  $f$  to  $Z$  is a non-decreasing function from the complete lattice  $Z$  to itself.
- ▶ Let  $Z^*$  ( $\subset X^*$ ) denote the set of fixed points of  $f|_Z$ , which is the set of upper bounds of  $S$  in  $X^*$ .

By part 1,  $Z^*$  has a least element, which is  $\sup_{X^*} S$ .

- ▶ A symmetric argument shows that  $\inf_{X^*} S$  exists in  $X^*$ .
- ▶ Thus,  $(X^*, \lesssim)$  is a complete lattice.

### Proposition 4.4

*Suppose that  $(X, \preceq_X)$  is a complete lattice and  $(Y, \preceq_Y)$  is a lattice, and that  $f: X \times Y \rightarrow X$  is a non-decreasing function. Then the greatest and the least fixed points of  $f(\cdot, y)$  are non-decreasing in  $y$ .*

## Proof

▶ Let  $\bar{x}^*(y) \in X$  denote the greatest fixed point of  $f(\cdot, y)$ .

▶ Let  $y' \lesssim y''$ .

Let  $Z = \{x \in X \mid \bar{x}^*(y') \lesssim x\}$ .

$(Z, \lesssim)$  is a complete lattice.

▶  $f(\cdot, y'')$  maps  $Z$  into itself:

for any  $x \in Z$ , we have

$$\bar{x}^*(y') = f(\bar{x}^*(y'), y') \lesssim f(x, y') \lesssim f(x, y'').$$

▶ By Tarski's Fixed Point Theorem,  $f(\cdot, y'')$  has a fixed point in  $Z$ , and its greatest fixed point in  $Z$  is the greatest fixed point in  $X$ ,

that is,  $\bar{x}^*(y'') \in Z$ , or  $\bar{x}^*(y') \lesssim \bar{x}^*(y'')$ .



## Proposition 4.5

*Let  $X \subset \mathbb{R}^n$  be a compact set, and suppose that  $(X, \leq)$  has a least element  $\underline{x}$ . Suppose that  $f: X \rightarrow X$  is non-decreasing and continuous.*

*Then the sequence  $x^k = f(x^{k-1})$  with  $x^0 = \underline{x}$  converges to the least fixed point.*

## Proof

- ▶ By the monotonicity of  $f$ , we have

$$x^0 = \underline{x} \leq f(x^0) = x^1,$$

$$x^1 = f(x^0) \leq f(x^1) = x^2,$$

$$x^2 = f(x^1) \leq f(x^2) = x^3,$$

⋮

- ▶ By the boundedness of  $X$ ,  $x^k$  converges to some  $\underline{x}^*$ , and by the closedness of  $X$ ,  $\underline{x}^* \in X$ .
- ▶ By the continuity of  $f$ ,  $\underline{x}^* = f(\underline{x}^*)$ , i.e.,  $\underline{x}^*$  is a fixed point of  $f$ .

- ▶ Let  $\hat{x} \in X$  be a fixed point of  $f$ .
- ▶ By the monotonicity of  $f$ , we have

$$x^0 = \underline{x} \leq \hat{x},$$

$$x^1 = f(x^0) \leq f(\hat{x}) = \hat{x},$$

$$x^2 = f(x^1) \leq f(\hat{x}) = \hat{x},$$

$\vdots$

- ▶ Therefore,  $\underline{x}^* \leq \hat{x}$ .
- ▶ This shows that  $\underline{x}^*$  is the least fixed point.

## Application: Games with Monotone Best Responses

- ▶  $\mathcal{I} = \{1, \dots, I\}$ : Set of players
- ▶  $S_i$ : Set of strategies of player  $i \in \mathcal{I}$   
Partially ordered by  $\preceq_i$
- ▶ Assumption: For all  $i \in \mathcal{I}$ ,  $(S_i, \preceq_i)$  is a complete lattice.
- ▶  $\preceq$ : Product partial order on  $S = \prod_{i \in \mathcal{I}} S_i$   
 $(s_i)_{i \in \mathcal{I}} \preceq (s'_i)_{i \in \mathcal{I}}$  if and only if  $s_i \preceq_i s'_i$  for all  $i \in \mathcal{I}$
- ▶  $\preceq_{-i}$ : Product partial order on  $S_{-i} = \prod_{j \neq i} S_j$   
 $(s_j)_{j \neq i} \preceq_{-i} (s'_j)_{j \neq i}$  if and only if  $s_j \preceq_j s'_j$  for all  $j \neq i$
- ▶  $u_i: S \rightarrow \mathbb{R}$ : Payoff function of player  $i \in \mathcal{I}$
- ▶ Denote this game by  $G$ .

- ▶ Best response correspondence (in pure strategies) of player  $i$ :

$$b_i(s_{-i}) = \{s_i \in S_i \mid u_i(s_i, s_{-i}) \geq u_i(s'_i, s_{-i}) \text{ for all } s_i \in S_i\}$$

- ▶ Assumption:  $G$  has monotone best responses,

i.e., for all  $i \in \mathcal{I}$ ,

- ▶ for all  $s_{-i} \in S_{-i}$ ,  $b_i(s_{-i})$  has a greatest element  $\bar{b}_i(s_{-i})$  and a least element  $\underline{b}_i(s_{-i})$ , and
  - ▶  $\bar{b}_i(s_{-i})$  and  $\underline{b}_i(s_{-i})$  are non-decreasing in  $s_{-i}$ .
- ▶ (We will later discuss what conditions on the primitives of the game guarantee this assumption to hold.)

## Examples

Coordination game:

	$L_2$	$R_2$
$L_1$	4, 4	0, 2
$R_1$	2, 0	3, 3

- ▶  $b_i(L_j) = L_i, b_i(R_j) = R_i$
- ▶ With orders  $L_i \prec_i R_i$ , the best responses are non-decreasing.

Battle of the sexes:

	$L_2$	$R_2$
$L_1$	0, 0	2, 1
$R_1$	1, 2	0, 0

- ▶  $b_i(L_j) = R_i, b_i(R_j) = L_i$
- ▶ With orders  $L_1 \prec_1 R_1$  and  $R_2 \prec_2 L_2$ , the best responses are non-decreasing.

Matching pennies:

	$L_2$	$R_2$
$L_1$	1, -1	-1, 1
$R_1$	-1, 1	1, -1

- ▶  $b_1(L_2) = L_1$ ,  $b_1(R_2) = R_1$   
 $b_2(L_1) = R_2$ ,  $b_2(R_1) = L_2$
- ▶ With any orders, the best responses cannot be non-decreasing simultaneously for both players.



# Existence of Pure-Strategy Nash Equilibria

## Proposition 4.6

*Suppose that the game  $G$  has monotone best responses.*

*Then  $G$  has a pure-strategy Nash equilibrium.*

*In particular, there are a greatest and a least pure-strategy Nash equilibria.*

## Proof

- ▶ The function  $\bar{b}: S \rightarrow S$  defined by  $\bar{b}(s) = (\bar{b}_1(s_{-1}), \dots, \bar{b}_n(s_{-i}))$  is a non-decreasing function from the complete lattice  $S$  to itself.
- ▶ By Tarski's Fixed Point Theorem, a greatest fixed point of  $\bar{b}$  exists, which is the greatest pure-strategy Nash equilibrium.

# Supermodular Functions

## Definition 4.6

For a lattice  $(X, \preceq)$ , a function  $f: X \rightarrow \mathbb{R}$  is said to be *supermodular* if

$$f(x) + f(x') \leq f(x \vee x') + f(x \wedge x')$$

for all  $x, x' \in X$ .

- ▶  $f$  is said to be *strictly supermodular* if  $f(x) + f(x') < f(x \vee x') + f(x \wedge x')$  whenever neither  $x \preceq x'$  nor  $x' \preceq x$ .
- ▶  $f$  is said to be (strictly) *submodular* if  $-f$  is (strictly) supermodular.

## Example

- ▶ Let  $X = \{x \in \mathbb{R}^2 \mid \underline{x} \leq x \leq \bar{x}\}$  for some  $\underline{x}, \bar{x} \in \mathbb{R}^2$ .

Suppose that  $f: X \rightarrow \mathbb{R}$  is supermodular.

- ▶ Consider  $(x'_1, x''_2)$  and  $(x''_1, x'_2)$  with  $x'_1 \leq x''_1$  and  $x'_2 \leq x''_2$ .

By the supermodularity of  $f$ , we have

$$\begin{aligned} & f(x'_1, x''_2) + f(x''_1, x'_2) \\ & \leq f(x'_1 \vee x''_1, x''_2 \vee x'_2) + f(x'_1 \wedge x''_1, x''_2 \wedge x'_2) \\ & = f(x''_1, x''_2) + f(x'_1, x'_2), \end{aligned}$$

or  $f(x''_1, x'_2) - f(x'_1, x'_2) \leq f(x''_1, x''_2) - f(x'_1, x''_2)$ ,

that is,  $f$  satisfies *increasing differences* in  $(x_1, x_2)$ .

- ▶ Conversely, if  $f$  satisfies increasing differences in  $(x_1, x_2)$ , then it is supermodular.

## Example: Submodular Functions on $\mathbb{R}^2$

- ▶ Let  $X = \{x \in \mathbb{R}^2 \mid \underline{x} \leq x \leq \bar{x}\}$  for some  $\underline{x}, \bar{x} \in \mathbb{R}^2$ .

Suppose that  $f: X \rightarrow \mathbb{R}$  is submodular  
(i.e.,  $-f$  is supermodular).

- ▶ Define the partial order  $\leq^*$  on  $\mathbb{R}^2$  by  
 $(x'_1, x'_2) \leq^* (x''_1, x''_2) \iff x'_1 \leq x''_1, x'_2 \geq x''_2$ .
- ▶ Then  $f$  is supermodular with respect to  $\leq^*$ :

If  $(x'_1, x'_2) \leq^* (x''_1, x''_2)$ , then

$$\begin{aligned} f(x''_1, x'_2) - f(x'_1, x'_2) &= -[(-f(x''_1, x'_2)) - (-f(x'_1, x'_2))] \\ &\leq -[(-f(x''_1, x''_2)) - (-f(x'_1, x''_2))] \\ &= f(x''_1, x''_2) - f(x'_1, x''_2). \end{aligned}$$

- ▶ This “trick” does not work with more than two variables.

### Proposition 4.7

Let  $X = \{x \in \mathbb{R}^n \mid \underline{x} \leq x \leq \bar{x}\}$  for some  $\underline{x}, \bar{x} \in \mathbb{R}^n$ ,  $\underline{x} \ll \bar{x}$ , and suppose that  $f: X \rightarrow \mathbb{R}$  is twice continuously differentiable on  $\text{int } X$  and continuous on  $X$ .

Then  $f$  is supermodular if and only if for all  $i, j = 1, \dots, n$ ,  $i \neq j$

$$\frac{\partial^2 f}{\partial x_i \partial x_j}(x) \geq 0 \text{ for all } x \in \text{int } X.$$

► Example:

$f(x_1, x_2) = x_1^{\alpha_1} x_2^{\alpha_2}$ ,  $\alpha_1, \alpha_2 \geq 0$ , is supermodular on  $\mathbb{R}_+^2$ .

# Optimization

## Proposition 4.8

Let  $(X, \preceq)$  be a lattice.

- ▶ If  $f: X \rightarrow \mathbb{R}$  is supermodular, then  $\arg \max_{x \in X} f(x)$  is a sublattice of  $X$ .
- ▶ If  $f$  is strictly supermodular, then  $\arg \max_{x \in X} f(x)$  is a chain, i.e., for any  $x, x' \in \arg \max_{x \in X} f(x)$ ,  $x \preceq x'$  or  $x' \preceq x$ .

## Proof

1.

- ▶ Suppose that  $x, x' \in \arg \max_{x \in X} f(x)$ .
- ▶ By supermodularity, we have

$$0 \leq f(x) - f(x \wedge x') \leq f(x \vee x') - f(x') \leq 0,$$

which must hold with equality.

- ▶ Thus,  $x \vee x', x \wedge x' \in \arg \max_{x \in X} f(x)$ .

2.

- ▶ If  $x, x' \in \arg \max_{x \in X} f(x)$ , then we have  $f(x \vee x') + f(x \wedge x') \leq f(x) + f(x')$ .
- ▶ If neither  $x \preceq x'$  nor  $x' \preceq x$ , then this contradicts the strict supermodularity.

## Proposition 4.9

Let  $X$  and  $Y$  be lattices, and suppose that  $f: X \times Y \rightarrow \mathbb{R}$  is supermodular.

Assume that  $v(y) = \sup_{x \in X} f(x, y)$  is finite for all  $y \in Y$ .

Then  $v$  is supermodular.

### Proof

- ▶ Let  $y, y' \in Y$ .
- ▶ For any  $x, x' \in X$ , we have

$$\begin{aligned} &v(y \vee_Y y') + v(y \wedge_Y y') \\ &\geq f(x \vee_X x', y \vee_Y y') + f(x \wedge_X x', y \wedge_Y y') \\ &\geq f(x, y) + f(x', y'). \end{aligned}$$

- ▶ Since this holds for all  $x, x' \in X$ , it follows that  $v(y \vee_Y y') + v(y \wedge_Y y') \geq v(y) + v(y')$ .



# Monotone Comparative Statics

## Definition 4.7

For partially ordered sets  $(X, \preceq_X)$  and  $(Y, \preceq_Y)$ , a function  $f: X \times Y \rightarrow \mathbb{R}$  satisfies *increasing differences* in  $(x, y)$  if

$$f(x'', y') - f(x', y') \leq f(x'', y'') - f(x', y'')$$

whenever  $x' \preceq_X x''$  and  $y' \preceq_Y y''$ .

## Proposition 4.10

Suppose that

- ▶  $X \subset \mathbb{R}^n$ ,  $X \neq \emptyset$ , is compact and a lattice (with respect to  $\leq$ );
- ▶  $(Y, \preceq_Y)$  is a partially ordered set; and
- ▶  $f: X \times Y \rightarrow \mathbb{R}$  satisfies the following:
  - ▶  $f(\cdot, y)$  is continuous for each  $y \in Y$ ;
  - ▶  $f(\cdot, y)$  is supermodular for each  $y \in Y$ ; and
  - ▶  $f(x, y)$  satisfies increasing differences in  $(x, y)$ .

Then

1. for each  $y \in Y$ ,  $\arg \max_{x \in X} f(x, y)$  is a nonempty, complete sublattice of  $X$ ; and
2. its greatest and least elements are non-decreasing in  $y$ .

## Proof

- ▶ Write  $X^*(y) = \arg \max_{x \in X} f(x, y)$ .
- ▶ By the compactness of  $X$  and the continuity of  $f(\cdot, y)$ ,  $X^*(y)$  is nonempty and compact.
- ▶ By the compactness, the lattice  $X$  is a complete lattice by Proposition 4.1.
- ▶ Thus, together with the supermodularity of  $f(\cdot, y)$ ,  $X^*(y)$  is a sublattice of  $X$  by Proposition 4.8.
- ▶ Again by Proposition 4.1,  $X^*(y)$  is a complete sublattice.
- ▶ Denote the greatest and the least elements of  $X^*(y)$  by  $\bar{x}^*(y)$  and  $\underline{x}^*(y)$ , respectively.

- ▶ Suppose that  $y' \preceq_Y y''$ .

If  $x' \in X^*(y')$  and  $x'' \in X^*(y'')$ , then

$$\begin{aligned}
 0 &\leq f(x', y') - f(x' \wedge_X x'', y') && \text{(by } x' \in X^*(y')) \\
 &\leq f(x' \vee_X x'', y') - f(x'', y') && \text{(by supermodularity)} \\
 &\leq f(x' \vee_X x'', y'') - f(x'', y'') && \text{(by increasing differences)} \\
 &\leq 0 && \text{(by } x'' \in X^*(y''))
 \end{aligned}$$

which must hold with equality.

- ▶ Thus,  $x' \wedge_X x'' \in X^*(y')$  and  $x' \vee_X x'' \in X^*(y'')$ .
- ▶ In particular, we must have
  - ▶  $\underline{x}^*(y') \preceq_X \underline{x}^*(y') \wedge_X \underline{x}^*(y'')$ , so that  $\underline{x}^*(y') \preceq_X \underline{x}^*(y'')$ ; and
  - ▶  $\bar{x}^*(y') \vee_X \bar{x}^*(y'') \preceq_X \bar{x}^*(y'')$ , so that  $\bar{x}^*(y') \preceq_X \bar{x}^*(y'')$ .

# Supermodular Games

- ▶  $\mathcal{I} = \{1, \dots, I\}$ : Set of players
- ▶  $S_i \subset \mathbb{R}^{n_i}$ : Set of strategies of player  $i \in \mathcal{I}$   
Partially ordered by  $\leq$  on  $\mathbb{R}^{n_i}$ 
  - ▶  $S_i \subset \mathbb{R}^{n_i}$ : compact
- ▶  $u_i: S \rightarrow \mathbb{R}$ : Payoff function of player  $i \in \mathcal{I}$  ( $S = \prod_{j \in \mathcal{I}} S_j$ )
  - ▶  $u_i(s_i, s_{-i})$ : continuous in  $s_i$  for each  $s_{-i}$  and continuous in  $s_{-i}$  for each  $s_i$
- ▶ Denote this game by  $G$ .
- ▶  $G$  is called a *supermodular game* if for each  $i \in \mathcal{I}$ ,
  - ▶  $S_i$  is a complete lattice;
  - ▶  $u_i(s_i, s_{-i})$  is supermodular in  $s_i$  for each  $s_{-i}$ ; and
  - ▶  $u_i(s_i, s_{-i})$  satisfies increasing differences in  $(s_i, s_{-i})$ .

# Smooth Supermodular Games

- ▶ The game is supermodular if the following are satisfied:

For each  $i \in \mathcal{I}$ :

- ▶  $S_i = \{s_i \in \mathbb{R}^{n_i} \mid \underline{s}_i \leq s_i \leq \bar{s}_i\}$  for some  $\underline{s}_i \ll \bar{s}_i$ ;
- ▶  $u_i$  is twice continuously differentiable on  $\text{int } S$ , and continuous on  $S$ ;
- ▶ for all  $s \in \text{int } S$ ,  $\frac{\partial^2 u_i}{\partial s_{ik} \partial s_{i\ell}}(s) \geq 0$  for all  $k, \ell = 1, \dots, n_i$ ,  $k \neq \ell$ .
- ▶ for all  $s \in \text{int } S$ ,  $\frac{\partial^2 u_i}{\partial s_{ik} \partial s_{jm}}(s) \geq 0$  for all  $j \neq i$ , and for all  $k = 1, \dots, n_i$  and all  $m = 1, \dots, n_j$ .

## Proposition 4.11

Suppose that the game  $G$  is a supermodular game.

1.  $G$  has monotone best responses,  
i.e., the greatest and the least best responses  $\bar{b}_i(s_{-i})$  and  $\underline{b}_i(s_{-i})$  are well defined and non-decreasing in  $s_{-i}$ .
2.  $G$  has a greatest and a least pure-strategy Nash equilibria  $\bar{s}^*$  and  $\underline{s}^*$ .
3. Let  $\bar{s} = (\bar{s}_i)_{i \in \mathcal{I}}$  and  $\underline{s} = (\underline{s}_i)_{i \in \mathcal{I}}$  be the greatest and the least strategy profiles. Then the sequences  $\bar{s}^k = \bar{b}(\bar{s}^{k-1})$ ,  $\bar{s}^0 = \bar{s}$  and  $\underline{s}^k = \underline{b}(\underline{s}^{k-1})$ ,  $\underline{s}^0 = \underline{s}$  converge to  $\bar{s}^*$  and  $\underline{s}^*$ , respectively (where  $\bar{b}(s) = (\bar{b}_i(s_{-i}))_{i \in \mathcal{I}}$  and  $\underline{b}(s) = (\underline{b}_i(s_{-i}))_{i \in \mathcal{I}}$ ).

## Proof

▶ Part 1: By Proposition 4.10.

▶ Part 2: By Proposition 4.6.

▶ Part 3:

$\{\underline{s}^k\}$  is increasing and bounded above, and thus converges to some  $s^* \in S$ .

▶ For any  $s_i \in S_i$ ,  $u_i(\underline{s}_i^k, \underline{s}_{-i}^{k-1}) \geq u_i(s_i, \underline{s}_{-i}^{k-1})$  for all  $k$ .

By continuity, letting  $k \rightarrow \infty$  we have

$$u_i(s_i^*, s_{-i}^*) \geq u_i(s_i, s_{-i}^*). \quad (*)$$

Thus,  $s^*$  is a Nash equilibrium.

▶ For any Nash equilibrium  $\hat{s}$ ,  $\underline{s}^0 \leq \hat{s}$ ,  $\underline{s}^1 = \underline{b}(\underline{s}^0) \leq \underline{b}(\hat{s}) \leq \hat{s}$ ,  $\dots$ , and hence  $s^* \leq \hat{s}$ , i.e.,  $s^*$  is the least Nash equilibrium.



► Proof of (\*):

Suppose that  $f(x, y)$  is continuous in  $x$  for each  $y$  and in  $y$  for each  $x$  and satisfies increasing differences in  $(x, y)$ .

If  $\{(x^k, y^k)\}$  is non-decreasing and converges to  $(x^*, y^*)$ , then  $\lim_{k \rightarrow \infty} f(x^k, y^k) = f(x^*, y^*)$ .

► 
$$f(x^*, y^*) - f(x^k, y^k) = f(x^*, y^*) - f(x^*, y^k) + f(x^*, y^k) - f(x^k, y^k),$$
 where by increasing differences,

$$f(x^*, y^0) - f(x^k, y^0) \leq f(x^*, y^k) - f(x^k, y^k) \leq f(x^*, y^*) - f(x^k, y^*).$$

► Therefore,

$$\begin{aligned} & [f(x^*, y^*) - f(x^*, y^k)] + [f(x^*, y^0) - f(x^k, y^0)] \\ & \leq f(x^*, y^*) - f(x^k, y^k) \\ & \leq [f(x^*, y^*) - f(x^*, y^k)] + [f(x^*, y^*) - f(x^k, y^*)], \end{aligned}$$

where the left and the right hand sides go to 0 as  $k \rightarrow \infty$  by continuity in  $x$  (for  $y = y^0, y^*$ ) and in  $y$  (for  $x = x^*$ ).

## Example: Bertrand Game with Differentiated Products

- ▶ Firms:  $\mathcal{I} = \{1, \dots, I\}$
- ▶ Strategy space of  $i$ :  $S_i = [0, \bar{p}_i]$  (prices)
- ▶  $d_i(p_i, p_{-i})$ : Demand for  $i$ 's product
  - ▶  $\frac{\partial d_i}{\partial p_i} < 0$
  - ▶  $\frac{\partial d_i}{\partial p_j} > 0, j \neq i$  (substitutability)
- ▶  $C_i$ : Total cost
  - ▶  $C_i'' \geq 0$
- ▶ Payoff function of  $i$ :

$$u_i(p_i, p_{-i}) = p_i d_i(p_i, p_{-i}) - C_i(d_i(p_i, p_{-i}))$$

- ▶ Cross derivatives:

$$\frac{\partial^2 u_i}{\partial p_i \partial p_j} = (p_i - C'_i) \frac{\partial^2 d_i}{\partial p_i \partial p_j} + \left(1 - C''_i \frac{\partial d_i}{\partial p_i}\right) \frac{\partial u_i}{\partial p_j}$$

Second term  $> 0$

- ▶ With linear demand  $d_i(p_i, p_{-i}) = a_i - b_i p_i + g_{ij} \sum_{j \neq i} p_j$ ,  $b_i, g_{ij} > 0$  (Problem 7.7), we have  $\frac{\partial^2 d_i}{\partial p_i \partial p_j} = 0$  and therefore  $\frac{\partial^2 u_i}{\partial p_i \partial p_j} > 0$ , so that the game is supermodular.

## Example: Cournot Game with Two Firms

- ▶ Firms:  $\mathcal{I} = \{1, 2\}$
- ▶ Strategy space of 1:  $S_1 = [0, \bar{x}_1]$  (quantities)  
Strategy space of 2:  $S_2 = [-\bar{x}_2, 0]$  (negative of quantities)
- ▶  $P(Q)$ : Inverse demand
  - ▶  $Q = x_1 + (-x_2)$ : total supply
  - ▶  $P' < 0$
- ▶  $C_i$ : Total cost
- ▶ Payoff functions:

$$u_1(x_1, x_2) = P(x_1 - x_2)x_1 - C_1(x_1)$$

$$u_2(x_1, x_2) = P(x_1 - x_2)(-x_2) - C_2(-x_2)$$

- ▶ Cross derivatives:

$$\frac{\partial^2 u_1}{\partial x_1 \partial x_2} = -P''(x_1 - x_2)x_1 - P'(x_1 - x_2)$$

$$\frac{\partial^2 u_2}{\partial x_2 \partial x_1} = P''(x_1 - x_2)x_2 - P'(x_1 - x_2)$$

$$-P' > 0$$

- ▶ With linear inverse demand  $P(Q) = 1 - Q$  (Problem 7.8), we have  $P'' = 0$  and therefore  $\frac{\partial^2 u_i}{\partial x_i \partial x_{-i}} > 0$ , so that the game is supermodular.

## Application: Stable Matchings

- ▶  $M$ : Set of men
- ▶  $W$ : Set of women
- ▶ Assume  $|M| = |W|$ .
- ▶ Each  $m \in M$  has a strict preference ordering  $>^m$  over  $W \cup \{m\}$ ;  
each  $w \in W$  has a strict preference ordering  $>^w$  over  $M \cup \{w\}$ .  
 $x >^i y \cdots i$  ranks  $x$  above  $y$ .
- ▶ Assume
  - ▶ for all  $m \in M$ ,  $w >^m m$  for all  $w \in W$ ; and
  - ▶ for all  $w \in W$ ,  $m >^w w$  for all  $m \in M$ .
- ▶ Write  $x \geq^i y$  for “not  $y >^i x$ ” ( $\iff$  “ $x >^i y$  or  $x = y$ ”).

- ▶ A *matching* is a function  $\mu: M \cup W \rightarrow M \cup W$  such that
  - ▶  $\mu(m) \in W \cup \{m\}$  for all  $m \in M$ ;
  - ▶  $\mu(w) \in M \cup \{w\}$  for all  $w \in W$ ; and
  - ▶  $\mu(m) = w$  if and only if  $\mu(w) = m$ .
- ▶ A pair  $(m, w) \in M \times W$  is a *blocking pair* for matching  $\mu$  if  $w \succ^m \mu(m)$  and  $m \succ^w \mu(w)$ .
- ▶ Matching  $\mu$  is *stable* if there is no blocking pair for  $\mu$ .  
(By assumption, individual rationality is satisfied.)

► Example:

$M$				$W$			
$m_1$ :	$w_2$	$w_1$	$w_3$	$w_1$ :	$m_1$	$m_3$	$m_2$
$m_2$ :	$w_1$	$w_3$	$w_2$	$w_2$ :	$m_3$	$m_1$	$m_2$
$m_3$ :	$w_1$	$w_2$	$w_3$	$w_3$ :	$m_1$	$m_3$	$m_2$

- $\{(m_1, w_1), (m_2, w_2), (m_3, w_3)\}$  is not stable.  
∴  $(m_1, w_2)$  is a blocking pair.
- $\{(m_1, w_1), (m_2, w_3), (m_3, w_2)\}$  is stable.



## Proposition 4.12

*There exists a stable matching.*

*Moreover, there exist*

- ▶ *a stable matching that is most preferred by all  $m \in M$  and least preferred by all  $w \in W$  among all stable matchings; and*
- ▶ *a stable matching that is most preferred by all  $w \in W$  and least preferred by all  $m \in M$  among all stable matchings.*
- ▶ First (formulated and) proved by Gale and Shapley (1962) via the “deferred acceptance algorithm”.
- ▶ We prove by Tarski’s Fixed Point Theorem.

- ▶ A *semi-matching* (or *pre-matching*) is a function  $\mu: M \cup W \rightarrow M \cup W$  such that
  - ▶  $\mu(m) \in W \cup \{m\}$  for all  $m \in M$ ; and
  - ▶  $\mu(w) \in M \cup \{w\}$  for all  $w \in W$ .
- ▶  $X_0$ : Set of all semi-matchings  
 $X \subset X_0$ : Set of all matchings
- ▶ Define the function  $f: X_0 \rightarrow X_0$  by the following:  
 for  $\mu \in X_0$ ,

$$f(\mu)(m) = \arg \max_{>^m} \{w \in W \mid m \geq^w \mu(w)\} \cup \{m\},$$

$$f(\mu)(w) = \arg \max_{>^w} \{m \in M \mid w \geq^m \mu(m)\} \cup \{w\}.$$

- ▶  $X^*$ : Set of fixed points of  $f$

## Proposition 4.13

*Any fixed point of  $f$  is a matching, i.e.,  $X^* \subset X$ .*

### Proof

- ▶ Suppose that  $f(\mu) = \mu$ .
- ▶ Suppose that  $\mu(m) = w$ .
- ▶ Then  $f(\mu)(m) = w$ , which implies  $m \geq^w \mu(w)$ .
- ▶ Also  $f(\mu)(w) \geq^w m$ , and therefore  $\mu(w) \geq^w m$ .
- ▶ Hence  $\mu(w) = m$ .
- ▶ A symmetric argument shows that  $\mu(w) = m \implies \mu(m) = w$ .

## Proposition 4.14

$\mu$  is a stable matching if and only if it is a fixed point of  $f$ , i.e.,  $\mu \in X^*$ .

### Proof

- ▶ Suppose that  $\mu \in X^*$  ( $\subset X$ ).

If  $w >^m \mu(m)$ , then  $f(\mu)(w) \geq^w m$ , hence  $\mu(w) \geq^w m$ .

Hence there is no blocking pair.

- ▶ Suppose that  $\mu \in X \setminus X^*$ .

Suppose that there exists  $m \in M$  such that  $w = \mu(m) \neq w' = f(\mu)(m)$ .

Then  $m >^{w'} \mu(w')$  and  $w' >^m \mu(m)$ .

Hence  $(m, w')$  is a blocking pair.

## Proposition 4.15

$f$  has a fixed point, i.e.,  $X^* \neq \emptyset$ .

### Proof

- ▶ Define the partial order  $\succsim$  on  $X_0$  as follows:  
 $\mu \succsim \nu$  if and only if
  - ▶  $\mu(m) \geq^m \nu(m)$  for all  $m \in M$ , and
  - ▶  $\nu(w) \geq^w \mu(w)$  for all  $w \in W$ .
- ▶ Then  $(X_0, \succsim)$  is a complete lattice.

- ▶  $f: X_0 \rightarrow X_0$  is non-decreasing:

Suppose that  $\mu \lesssim \nu$ .

- ▶ By  $\nu(w) \geq^w \mu(w)$ , we have  $m \geq^w \nu(w) \implies m \geq^w \mu(w)$ .

Therefore,  $f(\mu)(m) \geq^m f(\nu)(m)$ .

- ▶ By  $\mu(m) \geq^m \nu(m)$ , we have  $w \geq^m \mu(m) \implies w \geq^m \nu(m)$ .

Therefore,  $f(\nu)(w) \geq^w f(\mu)(w)$ .

- ▶ Thus, by Tarski's Fixed Point Theorem,  $X^* \neq \emptyset$ .

In particular,  $X^*$  has a greatest element (best for  $M$  and worst for  $W$ ) and a least element (worst for  $M$  and best for  $W$ ).

## Problem 3, Homework 4

1. By Tarski's Fixed Point Theorem,  $f$  has a greatest fixed point  $\bar{x}^*$ .
2.
  - ▶ Let  $X' = \{x \in X \mid \bar{x}^* \leq x\}$ .
  - ▶ For  $x \in X'$ , we have  $g(x) \geq f(x) \geq f(\bar{x}^*) = \bar{x}^*$ , so that  $g(x) \in X'$ .
  - ▶ Thus,  $g$  maps the compact convex set  $X'$  to  $X'$ .
  - ▶ By Brouwer's Fixed Point Theorem,  $g$  has a fixed point  $x^{**}$  in  $X'$ .
  - ▶ For any fixed point  $x^*$  of  $f$ , we have  $x^* \leq \bar{x}^* \leq x^{**}$ .