

# Monotone comparative statics

Federico Echenique

Caltech – SS205a

November 15, 2021

# Comparative statics



# Comparative statics

Consider a firm with production function  $f : \mathbb{R} \rightarrow \mathbb{R}$ , hiring labor at salary  $w$  and selling its product at price  $p$ .

Profit of the firm when it employs  $l$  units of labor is

$$g(w, l) = pf(l) - wl.$$

Labor demand is  $l(w) = \operatorname{argmax}_{l \geq 0} pf(l) - wl$ .

To do comparative statics, suppose we can apply the implicit function theorem.

We need  $f$  to be smooth, and assume an interior solution to the profit maximization problem.

From the first order condition  $pf'(l(w)) - w = 0$  we obtain:  
 $pf''(l(w))l'(w) - 1 = 0$ . Meaning that  $l'(w) = 1/pf''(l)$ . Then it would seem that downward-sloping labor demand hinges on  $f$  being concave.

As we shall see this idea is misleading.

## Proposition

Let  $f : \mathbb{R}^2 \rightarrow \mathbb{R}$  be twice differentiable, with

$$\frac{\partial^2 f(x, t)}{\partial x \partial t} \geq 0,$$

and suppose that  $a, b : \mathbb{R} \rightarrow \mathbb{R}$  are monotone increasing, with  $a(t) < b(t)$  for all  $t$ . Then there is

$$x^*(t) \in \operatorname{argmax}\{f(x, t) : x \in [a(t), b(t)]\}$$

that is monotone increasing.

# Partial orders

A binary relation  $\leq$  on a set  $X$  is:

- ▶ *reflexive* if  $(\forall x \in X)(x \leq x)$ ;
- ▶ *antisymmetric* if  $(\forall x, y \in X)(x \leq y \text{ and } y \leq x \implies x = y)$ ;
- ▶ *transitive* if

$$(\forall x, y, z \in X)(y \leq x \text{ and } z \leq y \implies z \leq x);$$

- ▶ a *partial order* if it is reflexive, antisymmetric and transitive;
- ▶ a *linear order* if  $(\forall x, y \in X)(x \leq y \text{ or } y \leq x)$ .

Obs. linear orders are also called complete.

A pair  $(X, \leq)$ , where  $X$  is a set and  $\leq$  is a partial order on  $X$ , is called a *partially ordered set*, or a PO set.

# Examples

The set  $(\mathbb{R}^n, \leq)$  is a PO set.

Let  $C_b(A)$  be the set all continuous and bounded functions  $f : A \subseteq \mathbb{R} \rightarrow \mathbb{R}$ , and define  $f \leq g$  if  $f(x) \leq g(x)$  for all  $x$ . Then  $(C_b(A), \leq)$  is a PO set.

Let  $\Omega$  be a non-empty set, and denote by  $2^\Omega$  the set of all subsets of  $\Omega$ . Then  $(2^\Omega, \subseteq)$  is a PO set.

The English alphabet ordered lexicographically.

On the set  $\mathbb{Z}_+$  of positive integers, let  $a \leq b$  if  $b$  is divisible by  $a$ . Then  $(\mathbb{Z}_+, \leq)$  is a PO set.

# Examples

Let  $V$  be a real vector space and  $P$  be a pointed convex cone in  $V$  (meaning that if  $x, y \in P$   $\alpha \in (0, 1)$  and  $\lambda > 0$  then  $\lambda x \in P$  and  $\alpha x + (1 - \alpha)y \in P$ ; and that if  $x, -x \in P$  then  $x = 0$ ).

Then  $P$  defines a partial order by  $x \leq y$  iff  $y - x \in P$ .

## Exercise

Verify that indeed this defines a partial order.

Later in the class, we'll work with a special case where  $P$  is defined via a collection of linear functions  $\mathcal{F}$ :

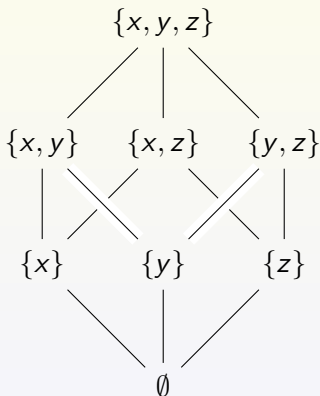
$$P = \{x \in V : f(x) \geq 0 \text{ for all } f \in \mathcal{F}\}$$

## Exercise

Show that if  $\mathcal{F}$  is a collection of linear functions such that  $0 = \cap\{f^{-1}(0) : f \in \mathcal{F}\}$ , then  $P$  as defined is a pointed convex cone.

# Partial orders

Finite PO sets are represented by a *Hasse diagram*. For ex.  $c(2^X, \subseteq)$ , by set inclusion:





Let  $(X, \leq)$  be a PO set and  $A \subseteq X$  be a subset of  $X$ .

Then  $A^u = \{x \in X : \forall z \in A, z \leq x\}$  is the set of *upper bounds* of  $A$ ; and

$A^l = \{x \in X : \forall z \in A, x \leq z\}$  is the set of *lower bounds* of  $A$ .

The *least upper bound*, or *supremum* of  $A$ , if it exists, is a smallest element of  $A^u$ . So the supremum of  $A$ , denoted  $\sup A$ , is  $x \in A^u$  s.t  $x \leq z$  for all  $z \in A^u$ .

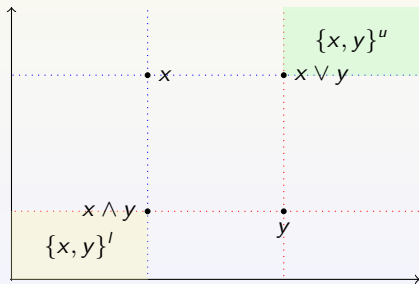
The *greatest lower bound*,  $\inf A$ , is  $x \in A^l$  such that  $z \leq x$  for all  $z \in A^l$ .

# Lattices

For sets with two elements, we use a special notation to denote infimum and supremum.

For  $x, y \in X$ , we let  $x \vee y = \sup\{x, y\}$   $x \wedge y = \inf\{x, y\}$ .

We term  $x \vee y$  the *join* of  $x$  and  $y$ ; and  $x \wedge y$  the *meet* of  $x$  and  $y$ .



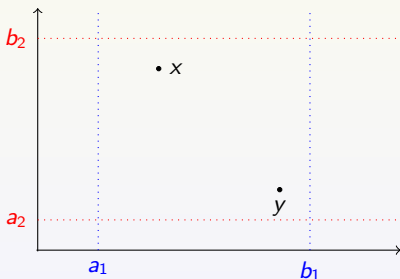
## Definition

A PO set  $(X, \leq)$  is a *lattice* if, for all  $x, y \in X$ ,  $x \wedge y$  and  $x \vee y$  exist in  $X$ .

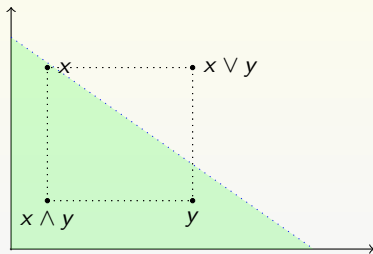
# Lattices

Example:

A *rectangle* in  $\mathbb{R}^n$  is a subset  $\prod_{i=1}^n [a_i, b_i]$  of  $\mathbb{R}^n$ , with  $a_i < b_i$  for all  $i = 1, \dots, n$ . A rectangle is a lattice.



# Not a lattice



A lattice  $(X, \leq)$  is *complete* if for all  $B \subseteq X$ ,  $\inf B$  and  $\sup B$  exists in  $X$ .<sup>1</sup>

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<sup>1</sup>This terminology is potentially confusing. We often use complete order to talk about a linear order.

Let  $(X, \leq)$  be a lattice and  $A \subseteq X$ . We say that  $A$  is a *sublattice* of  $(X, \leq)$  if, for all  $x, y \in A$ ,  $x \wedge y, x \vee y \in A$ .

Let  $A$  be a sublattice of  $(X, \leq)$ . We say that  $A$  is *subcomplete* if, for all  $B \subseteq A$ ,  $\inf B, \sup B \in A$ .

# Strong set order

Let  $(X, \leq)$  be a lattice. For  $A, B \subseteq X$ , say that  $A$  is *smaller than  $B$  in the strong set order* (or *induced set order*; and denoted by  $A \sqsubseteq B$ ) if

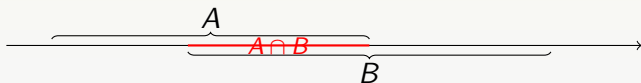
$$x \in A \text{ and } y \in B \implies x \wedge y \in A \text{ and } x \vee y \in B.$$



# Strong set order

Example with  $X = \mathbb{R}$ .

$A \sqsubseteq B$ :



## Lemma

$\sqsubseteq$  is antisymmetric and transitive on  $2^X \setminus \{\emptyset\}$ .

Obs.  $x \wedge y \leq x$  so  $(x \wedge y) \vee x = x$

And  $x \leq x \vee y$  so  $(x \vee y) \wedge x = x$

First anti-symmetry. Let  $A \sqsubseteq B$  and  $B \sqsubseteq A$ . Let  $x \in A$  and  $y \in B$ . Then  $x \wedge y \in A$  and  $x \vee y \in B$ , as  $A \sqsubseteq B$ . Then  $x = (x \vee y) \wedge x \in B$  and  $y = y \vee (x \wedge y) \in A$ , as  $B \sqsubseteq A$ . Thus  $A = B$ .

Now to show transitivity: Let  $A \sqsubseteq B$  and  $B \sqsubseteq C$ . Let  $x \in A$  and  $y \in C$ . Choose  $z \in B$ . Note:

$$x \vee y = x \vee ((y \wedge z \vee y)) = (x \vee (y \wedge z)) \vee y.$$

Now,  $y \wedge z \in B$ , as  $B \sqsubseteq C$ . Then  $x \vee (y \wedge z) \in B$ , as  $A \sqsubseteq B$ . So  $x \vee y \in C$ , as  $B \sqsubseteq C$ .

Similarly,

$$x \wedge y = (x \wedge (x \vee z)) \wedge y = x \wedge ((x \vee z) \wedge y).$$

Now,  $(x \vee z) \in B$ , so  $(x \vee z) \wedge y \in B$ . Then  $x \wedge y \in A$ .

Observe that  $A \sqsubseteq B$  iff  $A$  is a sublattice of  $(X, \leq)$ .

If we denote by  $L(X)$  the set of all nonempty sublattices of  $(X, \leq)$ , we obtain:

## Theorem

$(L(X), \sqsubseteq)$  is a PO set.

# Increasing differences

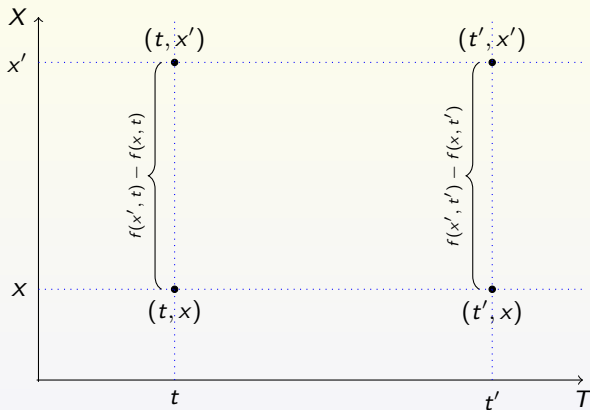
Let  $(X, \leq)$  be a lattice and  $(T, \leq')$  be a PO set.

A function  $f : X \times T \rightarrow \mathbb{R}$  has *increasing differences* if, for all  $x, x' \in X$  and  $t, t' \in T$  with  $x < x'$  and  $t < t'$

$$f(x', t) - f(x, t) \leq f(x', t') - f(x, t').$$

(Put differently, if the function  $t \mapsto f(x', t) - f(x, t)$  is monotone increasing, for each  $x, x' \in X$  with  $x < x'$ .)

# Increasing differences





A function  $f : X \times T \rightarrow \mathbb{R}$  has *strictly increasing differences* if, for all  $x, x' \in X$  and  $t, t' \in T$  with  $x < x'$  and  $t < t'$

$$f(x', t) - f(x, t) < f(x', t') - f(x, t').$$

We say that the function has increasing differences “in  $(x, t)$ .”

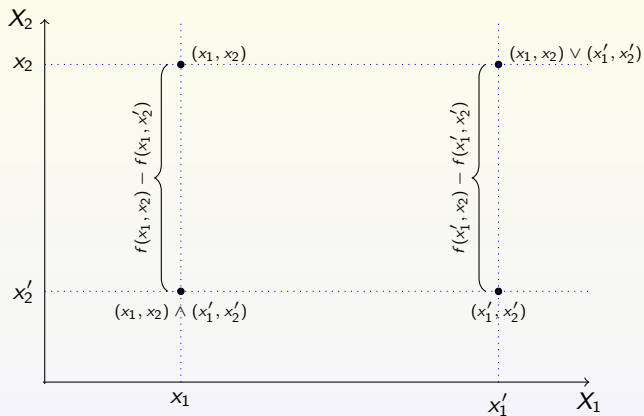
Obs. that a function has increasing differences in  $(x, t)$  iff it has increasing differences in  $(t, x)$ .

## Definition

A function  $f : X \rightarrow \mathbb{R}$  is *supermodular* if, for all  $x, x' \in X$ ,

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

# Supermodularity



Supermodularity requires that

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

which in the figure means that

$$f(x_1, x_2) - f(x_1, x'_2) \leq f(x'_1, x_2) - f(x'_1, x'_2).$$

So in this case, supermodularity is the same as increasing differences in each dimension.

Supermodularity is often viewed in economics as a notion of complementarity because of the connection to increasing differences. In the figure, think of  $f$  as production from two inputs, or the utility of consumption from two goods.

Supermodularity says that any “marginal” increase along the  $X_2$  dimension is aided by increases in the  $X_1$  dimension. The two inputs, or the two consumption goods, are therefore complements.

Two elements  $x, x' \in X$  are *unordered* if  $x \not\leq x'$  and  $x' \not\leq x$ .

A function  $f : X \rightarrow \mathbb{R}$  is *strictly supermodular* if, for all unordered  $x, x' \in X$ ,

$$f(x) + f(x') < f(x \vee x') + f(x \wedge x').$$

## Remark

If  $f$  and  $g$  are supermodular functions, and  $\lambda > 0$ , then  $f + g$  and  $\lambda f$  are supermodular functions.

There is, however, a supermodular  $f$  and strictly increasing  $h : \mathbb{R} \rightarrow \mathbb{R}$  such that  $h \circ f$  is not supermodular. *Supermodularity is a cardinal property.*

# Quasi-supermodularity

Let  $(X, \leq)$  and  $(T, \leq)$  be PO sets.

A function  $f : X \times T \rightarrow \mathbb{R}$  satisfies the *single-crossing property* in  $(x, t)$  ( $x \in X$  and  $t \in T$ ) if, for all  $x, x' \in X$  and  $t, t' \in T$ .

$$f(x, t) \leq f(x', t) \implies f(x, t') \leq f(x', t')$$

$$f(x, t) < f(x', t) \implies f(x, t') < f(x', t')$$

when  $x \leq x'$  and  $t \leq t'$ .

Moreover, if

$$f(x, t) \leq f(x', t) \implies f(x, t') < f(x', t')$$

for  $t < t'$  then we say that  $f$  satisfies the *strict single crossing property*.



## Remark

If  $f : X \times T \rightarrow \mathbb{R}$  has increasing differences in  $(x, t)$  then it satisfies the single crossing property.

Let  $(X, \leq)$  be a lattice. A function  $f : X \rightarrow \mathbb{R}$  is *quasi-supermodular* if, for all  $x, x' \in X$

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

$$f(x \wedge x') < f(x') \implies f(x) < f(x \vee x')$$

## Remark

If  $f : X \rightarrow \mathbb{R}$  is supermodular then it is quasi-supermodular.

## Theorem

Let  $(X_i, \leq_i)$  be a lattice for  $i = 1, \dots, n$ . If  $f : \times_{i=1}^n X_i \rightarrow \mathbb{R}$  has increasing differences in  $(x_i, x_j)$ , for  $i \neq j$ , and  $x_i \mapsto f(x_i, x_{-i})$ , for all  $i$  is supermodular, then  $f$  is supermodular.

## Corollary

Let  $X \subseteq \mathbb{R}^n$  be a lattice under the usual order on  $\mathbb{R}^n$ . If  $f : X \rightarrow \mathbb{R}$  has increasing differences in any two variables then it is supermodular.

## Proposition

Let  $X_i \subseteq \mathbb{R}$  be an open set, and  $X = \times_{i=1}^n X_i \subseteq \mathbb{R}^n$ . Let  $f : X \rightarrow \mathbb{R}$  be twice continuously differentiable. If

$$\frac{\partial^2 f(x)}{\partial x_i \partial x_j} \geq 0$$

for  $i \neq j$  then  $f$  has increasing differences in  $(x_i, x_j)$ ; and if this inequality holds for all  $i \neq j$  then  $f$  is supermodular.

# Supermodularity: Examples

- ▶  $f(x_1, x_2) = x_1 x_2$
- ▶  $f(x_1, \dots, x_n) = K \prod x_i^{a_i}$ ,  $K > 0$  and  $a_i > 0$
- ▶  $\log[D_1(p_1, p_2)(p_1 - c)]$ , when

$$p_2 \mapsto \frac{\partial \log D_1(p_1, p_2)}{\partial p_1}$$

is monotone increasing. When  $D_1$  is a demand function (for “differentiated products”) this means that demand elasticity

$$\epsilon = - \frac{\partial \log D_1(p_1, p_2)}{\partial \log p_1}$$

is decreasing in  $p_2$ .

# Supermodularity: Examples

- ▶  $f(x_1, x_2) = g(x_1 - x_2)$ , when  $g$  is a concave function.  
In particular, in the previous example,  $\log[D_1(p_1, p_2)(p_1 - c)]$  is supermodular in  $(p_1, c)$  as the log is concave.

# Monotonicity theorem

Let  $(X, \leq)$  be a lattice,  $(T, \leq')$  a PO set, and  $f : X \times T \rightarrow \mathbb{R}$  a function. Denote by

$$M(t, S) = \operatorname{argmax}\{f(x, t) : x \in S\}$$

the set of maximizers of  $f$  over the set  $S \subseteq X$  for fixed  $t \in T$ .

## Theorem (Milgrom and Shannon's Monotonicity Theorem)

$M(t, S)$  is monotone increasing iff

- ▶  $x \mapsto f(x, t)$  is quasi-supermodular;
- ▶  $f$  satisfies the single crossing property in  $(x, t)$



# Monotonicity theorem: Proof

( $\Leftarrow$ ) Let  $t \leq t'$  and  $S \sqsubseteq S'$ .

Let  $x \in M(t, S)$  and  $x' \in M(t', S')$ . Since  $S \sqsubseteq S'$ ,  $x \wedge x' \in S$  and  $x \vee x' \in S'$ . Now,  $x \wedge x' \in S$  and  $x \in M(t, S)$  means that  $f(x \wedge x', t) \leq f(x, t)$ .

By SCP,  $f(x \wedge x', t') \leq f(x, t')$ .

QSM implies that  $f(x', t') \leq f(x \vee x', t')$ .

Since  $x' \in M(t', S')$  and  $x \vee x' \in S'$  we obtain that  $x \vee x' \in M(t', S')$ .

Now:  $x', x \vee x' \in M(t', S')$  imply that  $f(x', t') = f(x \vee x', t')$ . So we must have, by single crossing and supermodularity, that  $f(x \wedge x', t) = f(x, t)$ . Thus  $x \wedge x' \in M(t, S)$ .

# Monotonicity theorem: Proof

( $\implies$ ) Conversely, suppose that  $M(t, S)$  is monotone increasing. Let  $x, x' \in X$  and  $t, t' \in T$ .

First consider  $S = \{x, x \wedge x'\}$  and  $S' = \{x', x \vee x'\}$ .

Note that  $S \sqsubseteq S'$ .

Suppose that  $f(x \wedge x', t) \leq f(x, t)$ ; so  $x \in M(t, S)$ .

Then  $M(t, S) \sqsubseteq M(t, S')$  implies that we must have  $f(x', t) \leq f(x \vee x', t)$  (as  $f(x', t) > f(x \vee x', t)$  would imply  $x' \in M(t, S')$  and  $x \vee x' \notin M(t, S')$ ). Similarly,  $f(x \wedge x', t) < f(x, t)$  implies  $f(x', t) < f(x \vee x', t)$ .

# Monotonicity theorem: Proof

Let  $x < x'$  and  $t < t'$ . Consider  $S = \{x, x'\}$ . If  $f(x, t) \leq f(x', t)$ , then  $x' \in M(t, S) \subseteq M(t', S)$  implies that  $f(x, t') \leq f(x', t')$ . Similarly,  $f(x, t) < f(x', t)$  implies that  $f(x, t') < f(x', t')$ .

# Monotonicity theorem

The following result follows directly from the more general Milgrom and Shannon result.

## Corollary (Topkis's monotonicity theorem)

Let  $x \mapsto f(x, t)$  be supermodular and satisfy increasing differences in  $(x, t)$ . Then  $M(t, S)$  is monotone increasing.

## Application: labor demand

Let  $g(l, w) = pf(l) - wl$ , where  $p, w, l$  are all real variables:  $p$  is price,  $w$  is wage and  $l$  is labor.

$f : \mathbb{R} \rightarrow \mathbb{R}$  is a production function. Note that  $\pi$  satisfies the single-crossing property in  $(-w, l)$ .

So labor demand is monotone decreasing. This result holds without any assumptions on  $f$ ; it is purely a consequence of the interaction between wages and labor: at higher wages, any given increase in labor use gives a higher cost increase.

## Application: labor demand

Suppose you want to generalize the result to more than one factor.

Let  $g(z, w) = pf(z) - w \cdot z$ , where  $z = (z_1, z_2)$  is a vector of production factors and  $w = (w_1, w_2)$  is a vector of factor prices.

As before,  $p$  is price and  $f$  a production function.

Now factor demand is decreasing when  $f$  is supermodular. In this case we do need some assumptions on production, but it is a natural assumption because we want the increase in one factor go “move” the demand for other factors in the same direction.

## Application: differentiated products

Let  $\pi_i(p_i, p_{-i}, c_i) = (p_i - c_i)D_i(p_i, p_{-i})$  be the profit function of a firm  $i$ .

$D_i$  is the demand facing firm  $i$  when prices are  $p = (p_1, \dots, p_n) = (p_i, p_{-i})$  and  $c_i$  is the (constant) marginal cost.

Suppose that demand elasticity

$$\epsilon = -\frac{\partial \log D_1(p_1, p_2)}{\partial \log p_1}$$

is decreasing in  $p_2$ .

## Application: differentiated products

By a similar calculation to what we already did, we see that  $g_i(p_i, p_{-i}, c_i)$  has increasing differences in  $(p_i, c_i)$  and in  $(p_i, p_{-i})$ .

So the optimal price will be increasing in  $c_i$  and also in competitors' prices.