

# Price Theory Encompassing

MIT 14.126, Fall 2001

1

## Road Map

- ◆ Traditional Doctrine: Convexity-Based
- ◆ Disentangling Three Big Ideas
  - Convexity: Prices & Duality
  - Order: Comparative Statics, Positive Feedbacks, Strategic Complements
  - Value Functions: Differentiability and Characterizations, Incentive Equivalence Theorems

2

## Convexity and Traditional Price Theory

## Convexity at Every Step

- ◆ Global or local convexity conditions imply
  - Existence of prices
  - Comparative statics, using *second-order conditions*
  - Dual representations, which lead to...
    - ◆ Hotelling's lemma
    - ◆ Shephard's lemma
    - ◆ Samuelson-LeChatelier principle
- ◆ "Convexity" is at the core idea on which the whole analysis rests.

# Samuelson-LeChatelier Principle

- ◆ Idea: Long-run demand is “more elastic” than short-run demand.
- ◆ Formally, the statement applies to smooth demand functions for sufficiently small price changes.
- ◆ Let  $p=(p_x, w, r)$  be the current vector of output and input prices and let  $p'$  be the long-run price vector that determined the current choice of a fixed input, say capital.
- ◆ Theorem: If the demand for labor is differentiable at this point, then:

$$\left. \frac{\partial l^L}{\partial w} \right|_p \leq \left. \frac{\partial l^S}{\partial w} \right|_{p, p'=p} \leq 0$$

5

# Varian's Proof

- ◆ Long- and short-run profit functions defined:

$$\pi^L(p) = \max_{k, l} p_x f(k, l) - wl - rk$$

$$\pi^S(p, p') = \max_l p_x f(k^*(p'), l) - wl - rk^*(p')$$

- ◆ Long-run profits are higher:

$$\pi^L(p) \geq \pi^S(p, p') \text{ for all } p, p' \text{ and } \pi^L(p) = \pi^S(p, p)$$

- ◆ So, long-run demand “must be” more elastic:

$$\left. \frac{\partial^2 \pi^L}{\partial w^2} \right|_p \geq \left. \frac{\partial^2 \pi^S}{\partial w^2} \right|_{p, p'=p} \geq 0 \text{ and } \left. \frac{\partial l^L}{\partial w} \right|_p \leq \left. \frac{\partial l^S}{\partial w} \right|_{p, p'=p} \leq 0$$

6

# A Robust “Counterexample”

- ◆ The production set consists of the convex hull of these three points, with free disposal allowed:

Capital	Labor	Output
0	0	0
1	1	1
0	2	1

- ◆ Fix the price of output at 9 and the price of capital at 3, and suppose the wage rises from  $w=2$  to  $w=5$ . Demands are:

$$l^L(2) = 2, l^S(5, 2) = 0, l^L(5) = 1$$

- ◆ *Long run labor demand falls less than short-run labor demand.*
- ◆ Robustness: Tweaking the numbers or “smoothing” the production set does not alter this conclusion.

# An Alternative Doctrine

## Disentangling Three Ideas

# Separating the Elements

## ◆ Convexity

- Proving existence of prices
- Dual representations of convex sets
- Dual representations of optima

## ◆ Order

- Comparative statics
- Positive feedbacks (LeChatelier principle)
- Strategic complements

## ◆ Envelopes

- Useful with dual functions
- Multi-stage optimizations
- Characterizing information rents

9

# Invariance Chart

Conclusions about $\max_{x \in S} f(x, t)$	Transformations of Choice Variable
Supporting ("Lagrangian") prices exist	Linear ("convexity preserving")
Optimal choices increase in parameter	Order-preserving
Long-run optimum change is larger, same direction	Order-preserving
Value function derivative formula: $V'(t) = f_2(x^*(t), t)$	One-to-one

10

# Convexity Alone

# Pure Applications of Convexity

## ◆ Separating Hyperplane Theorem

- Existence of prices
- Existence of probabilities
- Existence of Dual Representations
  - ◆ Example: Bondavera-Shapley Theorem
  - ◆ Example: Linear programming duality

## ◆ "Alleged" Applications of Duality

- Hotelling's lemma
- Shephard's lemma

11

12

## Separating Hyperplane Theorem

◆ Theorem. Let  $S$  be a non-empty, closed convex set in  $\mathbb{R}^N$  and  $x \notin S$ . Then there exists  $p \in \mathbb{R}^N$  such that

$$p \cdot x > \max\{p \cdot y \mid y \in S\}$$

◆ Proof. Let  $y \in S$  be the nearest point in  $S$  to  $x$ . Let

$$p = (x - y) / \|x - y\|$$

- Argue that such a point  $y$  exists.
- Argue that  $p \cdot x > p \cdot y$ .
- Argue that if  $z \in S$  and  $p \cdot z > p \cdot y$ , then for some small positive  $t$ ,  $tz + (1-t)y$  is closer to  $x$  than  $y$  is.

13

## Dual Characterizations

◆ Corollary. If  $S$  is a closed convex set, then  $S$  is the intersection of the closed "half spaces" containing it.

- Defining

$$\pi(p) = \max\{p \cdot x \mid x \in S\}$$

- it must be true that

$$S = \bigcap_{p \in \mathbb{R}^N} \{x \mid p \cdot x \leq \pi(p)\}$$

14

## Convexity and Quantification

◆ The following conditions on a closed set  $S$  in  $\mathbb{R}^N$  are equivalent

- $S$  is convex
- For every  $x$  on the boundary of  $S$ , there is a supporting hyperplane for  $S$  through  $x$ .
- For every concave objective function  $f$  there is some  $\lambda$  such that the maximizers of  $f(x)$  subject to  $x \in S$  are maximizers of  $f(x) + \lambda \cdot x$  subject to  $x \in \mathbb{R}^N$ .

15

Order Alone

16

# "Order" Concepts & Results

- ◆ Order-related definitions
- ◆ Optimization problems
  - Comparative statics for separable objectives
  - An improved LeChatelier principle
  - Comparative statics with non-separable "trade-offs"
- ◆ Equilibrium w/ Strategic Complements
  - Dominance and equilibrium
  - Comparative statics
  - Adaptive Learning
  - LeChatelier principle for equilibrium

17

# Two Aspects of Complements

- ◆ Constraints
  - Activities are complementary if doing one enables doing the other...
  - ...or at least doesn't prevent doing the other.
    - ◆ This condition is described by sets that are sublattices.
- ◆ Payoffs
  - Activities are complementary if doing one makes it weakly more profitable to do the other...
    - ◆ This is described by supermodular payoffs.
  - ...or at least doesn't change the other from being profitable to being unprofitable
    - ◆ This is described by payoffs satisfying a single crossing condition.

18

# Definitions: "Lattice"

- ◆ Given a partially ordered set  $(X, \geq)$ , define
  - The "join":  $x \vee y = \inf \{z \in X \mid z \geq x, z \geq y\}$ .
  - The "meet":  $x \wedge y = \sup \{z \in X \mid z \leq x, z \leq y\}$ .

- ◆  $(X, \geq)$  is a "lattice" if

$$(\forall x, y \in X) x \wedge y, x \vee y \in X$$

- ◆ Example:  $X = \mathbf{R}^N$ ,

$$x \geq y \text{ if } x_i \geq y_i, i = 1, \dots, N$$

$$(x \wedge y)_i = \min(x_i, y_i); i = 1, \dots, N$$

$$(x \vee y)_i = \max(x_i, y_i); i = 1, \dots, N$$

19

# Definitions, 2

- ◆  $(X, \geq)$  is a "complete lattice" if for every non-empty subset  $S$ , a greatest lower bound  $\inf(S)$  and a least upper bound  $\sup(S)$  exist in  $X$ .
- ◆ A function  $f: X \rightarrow \mathbf{R}$  is "supermodular" if
$$(\forall x, y \in X) f(x) + f(y) \leq f(x \wedge y) + f(x \vee y)$$
- ◆ A function  $f$  is "submodular" if  $-f$  is supermodular.

20

## Definitions, 3

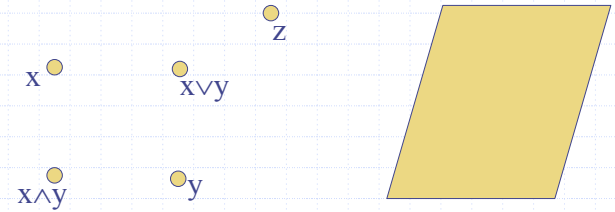
- ◆ Given two subsets  $S, T \subset X$ , " $S$  is *as high as*  $T$ ," written  $S \geq T$ , means

$$\begin{aligned} & [x \in S \text{ and } y \in T] \\ \Rightarrow & [x \vee y \in S \text{ and } x \wedge y \in T] \end{aligned}$$

- ◆ A function  $x^*$  is "*isotone*" (or "*weakly increasing*") if
$$t \geq t' \Rightarrow x^*(t) \geq x^*(t')$$
  - "Nondecreasing" is not used because...
- ◆ A set  $S$  is a "*sublattice*" if  $S \geq S$ .

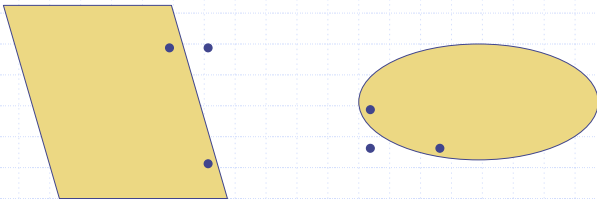
21

## Sublattices of $\mathbf{R}^2$



22

## Not Sublattices



- ◆ Convexity, order and topology are mostly independent concepts. However, in  $\mathbf{R}$ , these concepts coincide
- Topology:  $S = \text{compact set with boundary } \{a, b\}$
- Convexity:  $S = \{\alpha a + (1 - \alpha)b \mid \alpha \in [0, 1]\}$
- Order:  $S = [a, b] = \{x \mid a \leq x \leq b\}$

23

## "Pairwise" Supermodularity

- ◆ Theorem (Topkis). Let  $f: \mathbf{R}^N \rightarrow \mathbf{R}$ . The following are equivalent:
  - $f$  is supermodular
  - For all  $n \neq m$  and  $x_{-nm}$ , the restriction  $f(\cdot, \cdot, x_{-nm}): \mathbf{R}^2 \rightarrow \mathbf{R}$  is supermodular.

24

# Proof of Pairwise Supermodularity

- ◆  $\Rightarrow$  This direction follows from the definition.
- ◆  $\Leftarrow$  Given  $x \neq y$ , suppose for notational simplicity that

$$x_i = \begin{cases} \max(x_i, y_i) & \text{for } i = 1, \dots, n \\ \min(x_i, y_i) & \text{for } i = n + 1, \dots, N \end{cases}$$

- ◆ Then,

$$\begin{aligned} f(x \vee y) - f(y) &= \sum_{i=1}^n (f(x_1, \dots, x_i, y_{i+1}, \dots, y_N) - f(x_1, \dots, x_{i-1}, y_i, \dots, y_N)) \\ &\geq \sum_{i=1}^n [f(x_1, \dots, x_i, y_{i+1}, \dots, y_n, x_{n+1}, \dots, x_N) \\ &\quad - f(x_1, \dots, x_{i-1}, y_i, \dots, y_n, x_{n+1}, \dots, x_N)] \\ &= f(x) - f(x \wedge y) \end{aligned}$$

**QED**

25

# "Pairwise" Sublattices

- ◆ Theorem (Topkis). Let  $S$  be a sublattice of  $\mathbf{R}^N$ . Define

$$S_{ij} = \{x \in \mathbf{R}^N \mid (\exists z \in S) x_i = z_i, x_j = z_j\}$$

$$\text{Then, } S = \bigcap_{i,j} S_{ij}.$$

- ◆ Remark. Thus, a sublattice can be expressed as a collection of constraints on pairs of arguments. In particular, undecomposable constraints like

$$x_1 + x_2 + x_3 \leq 1$$

can never describe in a sublattice.

26

# Proof of Pairwise Sublattices

It is immediate that  $S \subset \bigcap_{i,j} S_{ij}$ . For the reverse,

suppose  $x \in \bigcap_{i,j} S_{ij}$ . Then,  $(\exists z^{ij} \in S) z_i^{ij} = x_i$  and  $z_j^{ij} = x_j$ .

Define  $z^j = \vee_j z^{ij} \in S$ . For all  $j$ ,  $z_j^j \geq x_j, z_i^j = x_i$ .

So,  $z = \wedge_i z^i \in S$  satisfies  $z_i = x_i$  for all  $i$ .

**QED**

27

# Complementarity

- ◆ Complementarity/supermodularity has equivalent characterizations:

- Higher marginal returns

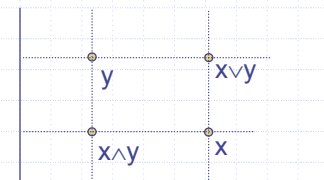
$$f(x \vee y) - f(x) \geq f(y) - f(x \wedge y)$$

- Nonnegative mixed second differences

$$[f(x \vee y) - f(x)] - [f(y) - f(x \wedge y)] \geq 0$$

- For smooth objectives, non-negative mixed second derivatives:

$$\frac{\partial^2 f}{\partial x_i \partial x_j} \geq 0 \text{ for } i \neq j$$



28



# Monotonicity Theorem

- ◆ Theorem (Topkis). Let  $f: X \times \mathbf{R} \rightarrow \mathbf{R}$  be a supermodular function and define

$$x^*(t) \equiv \operatorname{argmax}_{x \in S(t)} f(x, t).$$

If  $t \geq t'$  and  $S(t) \geq S(t')$ , then  $x^*(t) \geq x^*(t')$ .

- ◆ Corollary. Let  $f: X \times \mathbf{R} \rightarrow \mathbf{R}$  be a supermodular function and suppose  $S(t)$  is isotone. Then, for each  $t$ ,  $S(t)$  and  $x^*(t)$  are sublattices.
- ◆ Proof of Corollary. Trivially,  $t \geq t$ , so  $S(t) \geq S(t)$  and  $x^*(t) \geq x^*(t)$ . **QED**

29

# Proof of Monotonicity Theorem

- ◆ Suppose that  $f$  is supermodular and that  $x \in x^*(t), x' \in x^*(t'), t > t'$ .
- ◆ Then,  $(x \wedge x') \in S(t'), (x \vee x') \in S(t)$ .  
So,  $f(x, t) \geq f(x \vee x', t)$  and  $f(x', t') \geq f(x \wedge x', t')$ .
- ◆ If either any of these inequalities are strict then their sum contradicts supermodularity:

$$f(x, t) + f(x', t') > f(x \wedge x', t') + f(x \vee x', t).$$

**QED**

30

# Necessity for Separable Objectives

- ◆ Theorem (Milgrom). Let  $f: \mathbf{R}^N \times \mathbf{R} \rightarrow \mathbf{R}$  be a supermodular function and suppose  $S$  is a sublattice.

$$\text{Let } x_{g,S}^*(t) \equiv \operatorname{argmax}_{x \in S} f(x, t) + \sum_{n=1}^N g_n(x_n).$$

- ◆ Then, the following are equivalent:
  - $f$  is supermodular
  - For all  $g_1, \dots, g_N: \mathfrak{R} \rightarrow \mathfrak{R}$ ,  $x_{g,S}^*(t)$  is isotone.

## Remarks:

- This is a "robust monotonicity" theorem.
- The function  $g(x) \equiv \sum g_n(x_n)$  is "modular":  
 $g(x) + g(y) = g(x \wedge y) + g(x \vee y)$ .

# Proof

- ◆  $\Rightarrow$  Follows from Topkis's theorem.
- ◆  $\Leftarrow$  It suffices to show "pairwise supermodularity." Hence, it is sufficient to show that supermodularity is necessary when  $N=2$ . We treat the case of two choice variables; the treatment of a choice variable and parameter is similar.
- ◆ Let  $x, y \in \mathfrak{R}^2$  be unordered:  $x_1 > y_1, x_2 < y_2$
- ◆ Fix
 
$$g_i(z_i) = \begin{cases} -\infty & \text{if } z_i \notin \{x_i, y_i\} \\ f(x \wedge y) - f(x) & \text{if } z_i = x_i, i = 1 \\ f(x \wedge y) - f(y) & \text{if } z_i = y_i, i = 2 \\ 0 & \text{otherwise} \end{cases}$$
- ◆ If  $f(x) + f(y) > f(x \wedge y) + f(x \vee y)$ , then  $x_g^* = \{x, y, x \wedge y\}$  is not a sublattice, so  $-(x^*(t) \geq x^*(t))$ . **QED**

30



## Application: Production Theory

- ◆ Problem:

$$\max_{k,l} pf(k,l) - L(l,w) - K(k,r)$$

- ◆ Suppose that  $L$  is supermodular in the natural order, for example,  $L(l,w) = wl$ .
  - Then,  $-L$  is supermodular when the order on  $l$  is reversed.
  - $L^*(w)$  is nonincreasing in the natural order.
- ◆ If  $f$  is supermodular, then  $k^*(w)$  is also nonincreasing.
  - That is, capital and labor are "price theory complements."
- ◆ If  $f$  is supermodular with the reverse order, then capital and labor are "price theory substitutes."

33

## Application: Pricing Decisions

- ◆ A monopolist facing demand  $D(p,t)$  produces at unit cost  $c$ .

$$\begin{aligned} p^*(t) &= \operatorname{argmax}_p (p - c)D(p,t) \\ &= \operatorname{argmax}_{p > c} \log(p - c) + \log(D(p,t)) \end{aligned}$$

- ◆  $p^*(c,t)$  is always isotone in  $c$ . It is also isotone in  $t$  if  $\log(D(p,t))$  is supermodular in  $(p,t)$ , which is the same as being supermodular in  $(\log(p),t)$ , which means that increases in  $t$  make demand less elastic:

$$\frac{\partial \log D(p,t)}{\partial \log(p)} \text{ nondecreasing in } t$$

34

## Application: Auction Theory

- ◆ A firm's value of winning an item at price  $p$  is  $U(p,t)$ , where  $t$  is the firm's type. (Losing is normalized to zero.) A bid of  $p$  wins with probability  $F(p)$ .
- ◆ Question: Can we conclude that  $p(t)$  is nondecreasing, without knowing  $F$ ?

$$\begin{aligned} p_F^*(t) &= \operatorname{argmax}_p U(p,t)F(p) \\ &= \operatorname{argmax}_p \log(U(p,t)) + \log(F(p)) \end{aligned}$$

- ◆ Answer: Yes, if and only if  $\log(U(p,t))$  is supermodular.

35

## Long v Short-Run Demand

- ◆ Notation. Let  $I^S(w,w')$  be the short-run demand for labor when the current wage is  $w$  and the wage determining fixed inputs is  $w'$ .
- ◆ Setting  $w = w'$  in  $I^S$  gives the long run demands.
- ◆ Samuelson-LeChatelier principle:

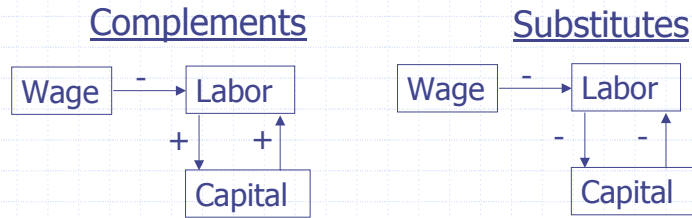
$$0 \geq I_1(w,w) \geq \frac{d}{dw} I(w,w).$$

- which can be restated revealingly as:

$$0 \geq I_2(w,w).$$

36

# Milgrom-Roberts Analysis



## Remarks:

- This analysis involves no assumptions about convexity, divisibility, etc.
- For smooth demands, symmetry of the substitution matrix implies that, locally, one of the two cases above applies.

37

# Improved LeChatelier Principle

◆ Let  $H(x, y, t)$  be supermodular and  $S$  a sublattice.

◆ Let  $(x^*(t), y^*(t)) = \max_{(x, y) \in S} \arg \max H(x, y, t)$

◆ Let  $x^*(t, t') = \max_{x \in \{x' | (x', y^*(t')) \in S\}} \arg \max H(x, y^*(t'), t)$

◆ Theorem (Milgrom & Roberts).  $x^*$  is isotone in both arguments. In particular, if  $t > t'$ , then

$$x^*(t) = x^*(t, t) \geq x^*(t, t') \geq x^*(t', t') = x^*(t')$$

38

# Proof

◆ By the Topkis Monotonicity Theorem,  $y^*(t) \geq y^*(t')$

◆ Applying the same theorem again, for all  $t, t'' > t'$

$$x^*(t', t') = \max_{x \in \{x' | (x', y^*(t')) \in S\}} \arg \max H(x, y^*(t'), t')$$

$$\leq x^*(t, t') = \max_{x \in \{x' | (x', y^*(t')) \in S\}} \arg \max H(x, y^*(t'), t)$$

$$\leq x^*(t, t'') = \max_{x \in \{x' | (x', y^*(t'')) \in S\}} \arg \max H(x, y^*(t''), t)$$

◆ Setting  $t=t''$  completes the proof.

# Long v Short-Run Demand

◆ Theorem. Let  $w > w'$ . Suppose capital and labor are **complements**, i.e.,  $f(k, l)$  is supermodular in the natural order. If demand is single-valued at  $w$  and  $w'$ , then

$$l^S(w, w) \leq l^S(w, w') \leq l^S(w', w')$$

◆ Theorem. Let  $w > w'$ . Suppose capital and labor are **substitutes**, i.e.,  $f(k, l)$  is supermodular when capital is given its reverse order. If demand is single-valued at  $w$  and  $w'$ , then

$$l^S(w, w) \leq l^S(w, w') \leq l^S(w', w')$$

39

# Non-separable Objectives

◆ Consider an optimization problem featuring “trade-offs” among effects.

- $x$  is the real-valued choice variable
- $B(x)$  is the “benefits production function”
- Optimal choice is

$$x_B^*(t) = \operatorname{argmax}_{x \in X} \pi(x, B(x), t)$$

41

# Robust Monotonicity Theorem

◆ Define:  $x_B^*(t) = \operatorname{argmax}_{x \in X} \pi(x, B(x), t)$

◆ Theorem. Suppose  $\pi$  is continuously differentiable and  $\pi_2$  is nowhere 0. Then:

$$\left[ (\forall x, y) \frac{\pi_1(x, y, t)}{\pi_2(x, y, t)} \text{ is increasing in } t \right]$$

$$\Rightarrow \left[ \text{For all } B, x_B^*(t) \text{ is isotone} \right]$$

$$\Rightarrow \left[ (\forall x, y) \frac{\pi_1(x, y, t)}{\pi_2(x, y, t)} \text{ is nondecreasing in } t \right]$$

42

# Application: Savings Decisions

◆ By saving  $x$ , one can consume  $F(x)$  in period 2.

$$V(w) = \max_{0 \leq x \leq w} U(w - x, F(x))$$

$$x_F^*(w) = \max \operatorname{argmax}_{0 \leq x \leq w} U(w - x, F(x))$$

◆ Define:  $\pi(x, y, t) = U(t - x, y)$

◆ Analysis. If  $MRS_{xy}$  increases with  $x$ , then optimal savings are isotone in wealth:

$$\left[ \frac{U_1(x, y)}{U_2(x, y)} \text{ increasing in } x \right] \Rightarrow x_F^*(w) \text{ isotone}$$

◆ This is the same condition as found in price theory, when  $F$  is restricted to be linear. Here,  $F$  is unrestricted.

- Also applies to Koopmans consumption-savings model.

43

# Introduction to Supermodular Games

44

## Formulation

- ◆ N players (infinite is okay)
- ◆ Strategy sets  $X_n$  are complete sublattices
  - $\underline{x}_n = \min X_n, \bar{x}_n = \max X_n$
- ◆ Payoff functions  $U_n(x)$  are
  - Continuous
  - "Supermodular with isotone differences"
    - $(\forall n)(\forall x_n, x'_n \in X_n)(\forall x_{-n} \geq x'_{-n} \in X_{-n})$
    - $U_n(x) + U_n(x') \leq U_n(x \wedge x') + U_n(x \vee x')$

45

## Bertrand Oligopoly Models

- ◆ Linear/supermodular Oligopoly:
  - Demand:  $Q_n(x) = A - ax_n + \sum_{j \neq n} b_j x_j$
  - Profit:  $U_n(x) = (x_n - c_n)Q_n(x)$
  - $\frac{\partial U_n}{\partial x_m} = b_m(x_n - c_n)$  which is increasing in  $x_n$
- ◆ Log-supermodular Oligopoly:
  - $\log U_n(x) = \log(x_n - c_n) + \log Q_n(x)$
  - $\frac{\partial^2 U_n}{\partial x_m \partial x_n} \geq 0 \Leftrightarrow \frac{\partial^2 \log Q_n(x)}{\partial \log x_n \partial \log x_m} \geq 0$

46

## Linear Cournot Duopoly

- ◆ Inverse Demand:  $P(x) = A - x_1 - x_2$
- $U_n(x) = x_n P(x) - C_n(x_n)$
- $\frac{\partial U_n}{\partial x_m} = -x_n$

- ◆ Linear Cournot duopoly (but not more general oligopoly) is supermodular if one player's strategy set is given the reverse of its usual order.

47

## Analysis of Supermodular Games

- ◆ Extremal Best Reply Functions
  - $B_n(x) = \max_{x'_n \in X_n} \left( \operatorname{argmax}_{x'_n \in X_n} U_n(x'_n, x_{-n}) \right)$
  - $b_n(x) = \min_{x'_n \in X_n} \left( \operatorname{argmax}_{x'_n \in X_n} U_n(x'_n, x_{-n}) \right)$
  - By Topkis's Theorem, these are isotone functions.
- ◆ Lemma:
  - $\neg[x_n \geq b_n(x)] \Rightarrow [x_n \text{ is strictly dominated by } b_n(x) \vee x_n]$
- ◆ Proof. If  $\neg[x_n \geq b_n(x)]$ , then
  - $U_n(x_n \vee b_n(x), x_{-n}) - U_n(x_n, x_{-n}) \geq U_n(b_n(x), x_{-n}) - U_n(x_n \wedge b_n(x), x_{-n}) > 0$

48

## Rationalizability & Equilibrium

- ◆ Theorem (Milgrom & Roberts): The smallest rationalizable strategies for the players are given by

$$\underline{z} = \lim_{k \rightarrow \infty} b^k(\underline{x})$$

Similarly the largest rationalizable strategies for the players are given by

$$\bar{z} = \lim_{k \rightarrow \infty} B^k(\bar{x})$$

Both are Nash equilibrium profiles.

49

## Proof

- ◆ Notice that  $b^k(\underline{x})$  is an isotone, bounded sequence, so its limit  $\underline{z}$  exists.
- ◆ By continuity of payoffs, its limit is a fixed point of  $b$ , and hence a Nash equilibrium.
- ◆ Any strategy less than  $\underline{z}_n$  is less than some  $b^k_n(\underline{x})$  and hence is deleted during iterated deletion of dominated strategies.
- ◆ **QED**

50

## Comparative Statics

- ◆ Theorem. (Milgrom & Roberts) Consider a family of supermodular games with payoffs parameterized by  $t$ . Suppose that for all  $n$ ,  $x_{-n}$ ,  $U_n(x_n, x_{-n}; t)$  is supermodular in  $(x_n, t)$ . Then

$\bar{z}(t), \underline{z}(t)$  are isotone.

- ◆ Proof. By Topkis's theorem,  $b_t^k(x)$  is isotone in  $t$ . Hence, if  $t > t'$ ,

$$b_t^k(x) \geq b_{t'}^k(x)$$

$$\underline{z}(t) = \lim_{k \rightarrow \infty} b_t^k(x) \geq \lim_{k \rightarrow \infty} b_{t'}^k(x) \geq \underline{z}(t')$$

and similarly for  $\bar{z}$ . **QED**

51

## Adaptive Learning

- ◆ Player  $n$ 's behavior is called "consistent with adaptive learning" if for every date  $t$  there is some date  $t'$  after which  $n$  does not play a strategy that is strictly dominated in the game in which others are restricted to play only strategies they have played since date  $t$ .
- ◆ Theorem (Milgrom & Roberts). In a finite strategy game, if every player's behavior is consistent with adaptive learning, then all eventually play only rationalizable strategies.

52

# Equilibrium LeChatelier Principle

## ◆ Formulation

- Consider a parameterized family of supermodular games with payoffs parameterized by  $t$ . Suppose that for all  $n$ ,  $x_{-n}$ ,  $U_n(x_n, x_{-n}; t)$  is supermodular in  $(x_n, t)$ .
- Fixing player 1's strategy at  $z_1(t')$  induces a supermodular game among the remaining players. Let  $y(t, t')$  be the smallest Nash equilibrium in the induced game, with  $y_1(t, t') = z_1(t')$ .

## ◆ Theorem.

- If  $t > t'$ , then  $z(t) \geq y(t, t') \geq z(t')$ .
- If  $t < t'$ , then  $z(t) \leq y(t, t') \leq z(t')$ .

...and a similar conclusion applies to the maximum equilibrium.

53

# Proof

- ◆ Observe (exercise) that

$$z(t) = \underline{y}(t, t), z(t') = \underline{y}(t', t').$$

- ◆ Suppose  $t > t'$ .

- By the comparative statics theorem,  $z$  is isotone, so:

$$z(t) \geq z(t').$$

- Hence, by the comparative statics theorem applied again,  $y$  is isotone, so:

$$\underline{y}(t, t) \geq \underline{y}(t, t') \geq \underline{y}(t', t').$$

**QED**

54

# Envelope Functions Alone

Based on "Envelope Theorems for Arbitrary Choice Sets" by Paul Milgrom & Ilya Segal

# What are "Envelope Theorems"?

- ◆ Envelope theorems deal with the properties of the value function:  $V(t) \equiv \max_{x \in X} f(x, t)$
- ◆ Answer questions about...
  - when  $V$  is differentiable, directionally differentiable, Lipschitz, or **absolutely continuous**
  - when  $V$  satisfies the "envelope formula"  
 $V'(t) = f_t(x, t)$  for  $x \in x^*(t)$
- ◆ "Traditional" envelope theorems assume that set  $X$  is convex and the objective  $f(\cdot, t)$  is concave and differentiable.

55

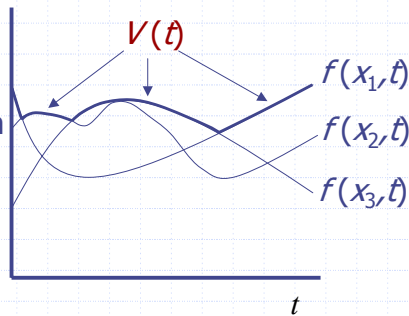
56



# Intuitive Argument

◆ When  $X = \{x_1, x_2, x_3\} \dots$

- $V$  is left- and right-differentiable everywhere
- if  $f_t(x, t)$  is constant on  $x \in X^*(t)$ , then  $V$  is differentiable at  $t$
- envelope formulas apply for
  - ◆  $V'(t) = f_t(x^*(t), t)$
  - ◆  $V'(t+)$  and  $V'(t-)$



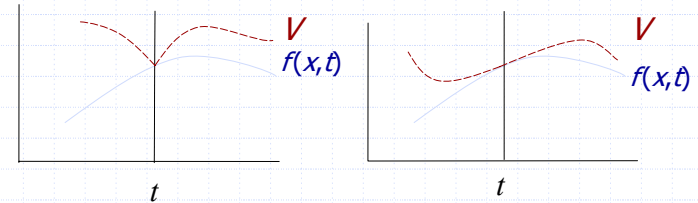
57

# Envelope Derivative Formula

◆ **Theorem 1.** Take  $t \in [0, 1]$  and  $x \in X^*(t)$ , and suppose that  $f_t(x, t)$  exists.

- If  $t < 1$  and  $V'(t+)$  exists, then  $V'(t+) \geq f_t(x, t)$ .
- If  $t > 0$  and  $V'(t-)$  exists, then  $V'(t-) \leq f_t(x, t)$ .
- If  $t \in (0, 1)$  and  $V'(t)$  exists, then  $V'(t) = f_t(x, t)$ .

◆ **Proof:**



58

# Absolute Continuity

◆ **Theorem 2(A).** Suppose that

- $f(x, \cdot)$  is differentiable (or just absolutely continuous) for all  $x \in X$  with derivative (or density)  $f_t$ .
- there exists an integrable function  $b(t)$  such that  $|f_t(x, \cdot)| \leq b(t)$  for all  $x \in X$  and almost all  $t \in [0, 1]$ .

Then  $V$  is absolutely continuous with density satisfying  $|V'(t)| \leq b(t)$ .

# Proof of Theorem 2(A)

◆ Define

$$B(t) = \int_0^t b(s) ds$$

◆ Then for  $t'' > t'$ :

$$\begin{aligned} |V(t'') - V(t')| &\leq \sup_{x \in X} |f(x, t'') - f(x, t')| \\ &= \sup_{x \in X} \left| \int_{t'}^{t''} f_t(x, t) dt \right| \leq \int_{t'}^{t''} \sup_{x \in X} |f_t(x, t)| dt \\ &\leq \int_{t'}^{t''} b(t) dt = |B(t'') - B(t')| \end{aligned}$$

◆ It suffices to prove the theorem for intervals, because open intervals are a basis for the open sets. **QED**

59

60

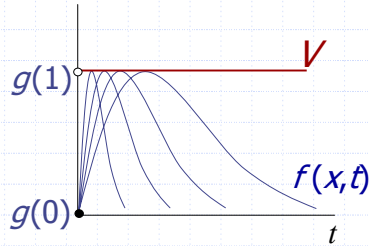


## Why do we need $b(\cdot)$ ?

◆ Let  $X=(0,1]$  and  $f(x,t)=g(t/x)$ , where  $g$  is smooth and single-peaked with unique maximum at 1.

- $V(0)=g(0)$ ,  $V(t)=g(1)$ :  $V$  is discontinuous at 0.
- This example has no integrable bound  $b(t)$ :

$$\sup_{x \in (0, \infty)} |f_t(x, t)| = \sup_{x \in (0, \infty)} \left| \frac{1}{t} \left( \frac{t}{x} g' \left( \frac{t}{x} \right) \right) \right| = \frac{1}{t} \sup_{x \in (0, \infty)} |x g'(x)|$$



61

## Envelope Integral Formula

◆ Theorem 2(B). Suppose that, in addition to the assumptions of 2(A), the set of optimizers  $x^*(t)$  is non-empty for all  $t$ . Then for any selection  $x(t) \in x^*(t)$ ,

$$V(s) = V(0) + \int_0^s f_t(x(t), t) dt.$$

62

## Equi-differentiability

◆ Definition. A family of functions  $\{f(x, \cdot)\}_{x \in X}$  is "equi-differentiable" at  $t \in (0, 1)$  if

$$\lim_{t' \rightarrow t} \sup_x \left| \frac{f(x, t') - f(x, t)}{t' - t} - f_t(x, t) \right| = 0$$

◆ If  $X$  is finite, this is the same as simple differentiability.

## Directional Differentiability

◆ Theorem 3. If

- (i)  $\{f(x, \cdot)\}_{x \in X}$  is equi-differentiable at  $t_0$ ,
- (ii)  $x^*(t)$  is non-empty for all  $t$ , and
- (iii)  $\sup_x |f_t(x, t_0)| < \infty$ ,

then for any selection  $x(t) \in x^*(t)$ ,  $V$  is left- and right-differentiable at  $t_0 \in (0, 1)$  and the derivatives satisfy

$$V'(t_0+) = \lim_{t \rightarrow t_0+} f_t(x(t), t_0)$$

$$V'(t_0-) = \lim_{t \rightarrow t_0-} f_t(x(t), t_0)$$

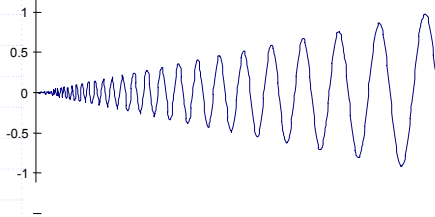
63

64

# Role of "Equi-differentiability"

◆ Simple differentiability (rather than equi-differentiability) is not enough for  $V$  to have left- and right-derivatives:

- Let  $g(t) = t \sin \log(t)$ ,  $f(x, t) = g(t)$  if  $t > \exp(-\pi/2 - 2\pi x)$ ,  $f(x, t) = -t$  otherwise.
- Then,  $V(t) = g(t)$



65

# Continuous Problems

◆ Theorem 4. Suppose  $X$  is a non-empty compact space,  $f$  is upper semi-continuous on  $X$  and  $f_t$  is continuous in  $(x, t)$ . Then,

- $V$  is directionally differentiable

$$V'(t+) = \max_{x \in X^*(t)} f_t(x, t) \text{ for } t \in [0, 1]$$

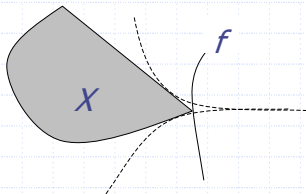
$$V'(t-) = \min_{x \in X^*(t)} f_t(x, t) \text{ for } t \in (0, 1]$$

- In particular,  $V'(t+) \geq V'(t-)$ .
- $V$  is differentiable at  $t$  if any of the following hold:
  - ◆  $V$  is concave (because  $V'(t+) \leq V'(t-)$ )
  - ◆  $t$  is a maximum of  $V(\cdot)$  (because  $V'(t+) \leq V'(t-)$ )
  - ◆  $X^*(t)$  is a singleton (because  $V'(t+) = V'(t-)$ )

66

# Contrast to a "Traditional" Approach

◆ In some approaches, the differentiability of  $x^*$  is used in the argument. However,  $V$  can be differentiable even when  $x^*$  is not. This often happens, for example, in strictly convex problems:



# Applications

67

68

## Hotelling's Lemma

◆ Define:

$$\pi(p) = \max_{x \in X} p \cdot x$$

$$x^*(p) = \operatorname{argmax}_{x \in X} p \cdot x$$

◆ Theorem. Suppose  $X$  is compact. Then,  $\pi'(p)$  exists if and only if  $x^*(p)$  is a singleton, and in that case  $\pi'(p) = x^*(p)$ .

69

## Shephard's Lemma

◆ Define:

$$C(y, p) = \min_{x \in X, x_1 = y} -p_{-1} \cdot x_{-1}$$

$$x^*(p) = \operatorname{arg} \min_{x \in X, x_1 = y} -p_{-1} \cdot x_{-1}$$

◆ Remark: The variable  $x_1$  represents "output" and the other variables represent inputs, measured as negative numbers.

◆ Theorem. Suppose  $X$  is compact. Then,  $\partial C / \partial p$  exists if and only if  $x^*(p)$  is a singleton, and in that case  $\partial C / \partial p = x^*(p)$ .

70

## Multi-Stage Maximization

◆ Stage 1: choose investment  $t \geq 0$ .

◆ Stage 2: choose action vector  $x \in X \neq \emptyset$

◆ Assume:

- $f(x, t)$  is equidifferentiable in  $t$  and  $t^* > 0$
- $f(x, t)$  is u.s.c. in  $x$  and  $X$  is compact

◆ Conclusion: the value function  $V(t)$  is differentiable at  $t^*$  and  $V'(t^*) = 0$ .

- Proof: Apply theorem 4.

## Mechanism Design

◆  $Y$  = set of outcomes

◆ Agent's type is  $t$ , utility is  $f(x, t)$ .

◆  $M$  = message space.  $h: M \rightarrow Y$  is outcome function.

◆  $X = h(M)$  is set of "accessible outcomes."

◆ Assume that each type has an optimal choice

$$x(t) \in \operatorname{argmax}_{x \in X} f(x, t)$$

70

# Analysis

- ◆ **Corollary 1.** Suppose that the agent's utility function  $f(x, t)$  is differentiable and absolutely continuous in  $t$  for all  $x \in Y$ , and that  $\sup_{x \in Y} f(x, t)$  is integrable on  $[0, 1]$ . Then the agent's equilibrium utility  $V$  in any mechanism implementing a given choice rule  $x$  must satisfy the following integral condition.

$$V(t) = V(0) + \int_0^t f_t(x(s), s) ds.$$

- This had previously been shown only with (sometimes "weak") additional conditions.

73

# Mechanism Design Applications

- ◆ Models in which payoffs are  $v \cdot p - \pi$ , so
 
$$U(v) = U(0) + \int_0^1 v \cdot p^*(sv) ds.$$

## ◆ Theorems

- Green-Laffont Theorem
  - ◆ Uniqueness of Dominant Strategy Mechanisms
- Holmstrom-Williams Theorem
  - ◆ Bayesian Revenue Equivalence
- Myerson-Satterthwaite Theorem
  - ◆ Necessity of Bargaining Inefficiency
- Jehiel-Moldovanu Theorem
  - ◆ Impossibility of Efficiency with Value Interdependencies

74

# Green-Laffont Theorem

- ◆ "Uniqueness of dominant strategy implementation."

- ◆ **Theorem** (Holmstrom's variation). Suppose that

- $M$  is a direct mechanism to implement the efficient outcome in dominant strategies
- the type space is smoothly path-connected.

- ◆ Then,

- the payment function for player  $j$  in mechanism  $M$  is equal to the payment function of the Vickrey-Clarke-Groves pivot mechanism plus some function  $g_j(v_{-j})$  (which depends only on the other player's types).

# Green-Laffont Theorem

- ◆ Given any value vector  $v$ , let  $\{v_j(t) | t \in [0, 1]\}$  be a smooth path connecting some fixed value  $\underline{v}_j$  to  $v_j = v_j(1)$ . By the Envelope Theorem applied to the path parameter  $t$ ,

$$\begin{aligned} U_j(v_j(t), v_{-j}) &= p_j(v_j(t), v_{-j}) \cdot v_j - X_j(v_j(t), v_{-j}) \\ &= U_j(\underline{v}_j, v_{-j}) + \int_0^t p_j(v_j(s), v_{-j}) \cdot v_j'(s) ds \end{aligned}$$

$$\therefore X_j(v_j(1), v_{-j}) = f_j(v_{-j}) + p_j(v_j(1), v_{-j}) \cdot v_j - \int_0^1 p_j(v_j(s), v_{-j}) \cdot v_j'(s) ds$$

$$\text{where } f(v_{-j}) = -U_j(\underline{v}_j, v_{-j})$$

- ◆ So,  $X_j$  is fully determined by the functions  $p$  and  $f_j$ .

75

76

# Holmstrom-Williams' Theorem

- ◆ **Theorem:** Any mechanism that Bayes-Nash implements efficient outcomes on a smoothly path-connected type space entails the same *expected* payments as the Vickrey mechanism, plus some bidder-specific constant.
- ◆ **Proof.** Let  $\{v_j(s), s \in [0,1]\}$  be a path from some fixed value vector to any other value vector. By the Envelope Theorem,

$$U_j(v_j(t)) = p_j(v_j(t)) \cdot v_j(t) - X_j(v_j(t))$$

$$= U_j(v_j(0)) + \int_0^t p_j(v_j(s)) \cdot v_j'(s) ds$$

- ◆ Hence,  $X_j(v)$  is uniquely determined by  $U_j(0)$ . It is equal to  $U_j(0)$  plus the expected payment in the Vickrey mechanism.

77

# Two-Person Bargaining

- ◆ **Assume**
  - there is a buyer with value  $v$  distributed on  $[0,1]$
  - there is a seller with cost  $c$  distributed on  $[0,1]$
- ◆ **The Vickrey-Clarke-Groves mechanism**
  - has each party report its value
  - entails  $p^*(v,c)=1$  if  $v>c$  and  $p^*(v,c)=0$  otherwise
  - payments are
    - ◆ if  $p^*(v,c)=0$ , no payments
    - ◆ if  $p^*(v,c)=1$ , buyer pays  $c$  and the seller receives  $v$

78

# Myerson-Satterthwaite Theorem

- ◆ Expected profits are:
  - $U_B(v) = E[(v-c)1_{\{v>c\}} | v]$ , so  $E[U_B(v)] = E[(v-c)1_{\{v>c\}}]$
  - $U_S(c) = E[(v-c)1_{\{v>c\}} | s]$ , so  $E[U_S(c)] = E[(v-c)1_{\{v>c\}}]$
  - each bidder expects to receive the **entire social surplus**.
- ◆ Apply Holmstrom-Williams theorem:
- ◆ **Theorem (Myerson-Satterthwaite).** There is no mechanism and Bayesian Nash equilibrium such that the mechanism implements for all  $v,c$  with  $v>c$  and
  - $U_B(0) = U_S(1) = 0$  ("voluntary participation by worst type")
  - $E[U_B(v)] + E[U_S(c)] \leq E[(v-c)1_{\{v>c\}}]$  ("balanced expected budget")

79

# Subtleties

- ◆ Consider a model in which:

$$\Pr\{v > 1\} = \Pr\{c < 1\} = 1$$

- ◆ Q: Why doesn't simply trading at price  $p=1$  violate the theorem in this model?
- ◆ A: Because it prescribes trade even when  $c>v$ !

80