

How Dull Are Monotonic Mechanisms

Sergiu Hart

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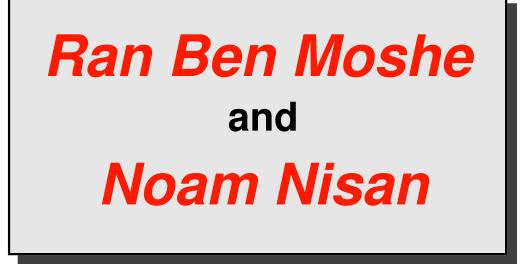
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Joint work with



The Hebrew University of Jerusalem

SERGIU HART C 2022 - p. 3



Ran Ben Moshe, Sergiu Hart, and Noam Nisan "Monotonic Mechanisms for Selling Multiple Goods" (2022)

www.ma.huji.ac.il/hart/abs/mech-monot.html



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• 1 SELLER

ho 1 BUYER

• 1 SELLER

- 1 BUYER
- k GOODS (ITEMS)

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- 1 BUYER
- k goods (items)

OBJECTIVE:

MAXIMIZE the **REVENUE** of the **SELLER**

• 1 SELLER

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- J BUYER
- k goods (items)
 - values of GOODS to BUYER :

 $X=\left(X_{1},X_{2},...,X_{k}
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 - additive valuation (good 1 and good $2 = X_1 + X_2$)
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quasi-linear utilities (i.e., additive in monetary payments)

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SELLER :

no value and no cost for the GOODS

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 $\mathsf{Rev}(X) :=$ optimal revenue from valuation X



ONE GOOD (k = 1):

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Myerson 1981

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\checkmark SELLER posts a PRICE p

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$$\mathsf{Rev}(X) = \max_p p \cdot (1 - F(p))$$

Myerson 1981



$X \sim \begin{cases} 10 & \text{with probability } 1/2 \\ 22 & \text{with probability } 1/2 \end{cases}$

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One Good: Example

$$X \sim \begin{cases} 10 & \text{with probability } 1/2 \\ 22 & \text{with probability } 1/2 \end{cases}$$

$$p = 10 \rightarrow R = 10 \cdot 1 = 10$$

$$p = 22 \rightarrow R = 22 \cdot 1/2 = 11 \quad \longleftarrow$$

 $\mathsf{Rev}(X) = 11$ p = 22

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Multiple Goods $(k \ge 2)$



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No simple useful characterization of solution

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There are valuations X such that *N*-REV(X) = 1 and REV(X) = ∞
For every ε > 0 there are bounded X s.t. *N*-REV(X) < ε ⋅ REV(X)

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Multiple Goods $(k \ge 2)$

• There are valuations X such that \mathcal{N} -REV(X) = 1 and REV $(X) = \infty$ • For every $\varepsilon > 0$ there are bounded X s.t. \mathcal{N} -REV $(X) < \varepsilon \cdot \text{REV}(X)$

Hart and Nisan 2013/2019 (Briest, Chawla, Kleinberg, Weinberg 2010/2015 for $k \ge 3$)

Multiple Goods $(k \ge 2)$

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For example:

selling separately

Multiple Goods (k > 2)

- selling separately
- selling bundled

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- all deterministic mechanisms

Multiple Goods (k > 2)

- selling separately
- selling bundled
- all deterministic mechanisms
- mechanisms with bounded "menus" (at most m choices, for finite m)

Multiple Goods $(k \ge 2)$

Simple mechanisms cannot guarantee any positive fraction of the optimal revenue

Multiple Goods (k > 2)

- No simple useful characterization of solution
- Hard to solve even in simple cases
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"CONCEPTUAL COMPLEXITY"



Monotonicity

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Monotonicity of Revenue

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BUYER's willingness to pay increases



BUYER's willingness to pay increases

\Rightarrow **SELLER's revenue increases**

Monotonicity of Revenue

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correct for one good

Non-Monotonicity of Revenue

BUYER's willingness to pay increases

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- correct for one good
- FALSE for multiple goods !

Non-Monotonicity of Revenue

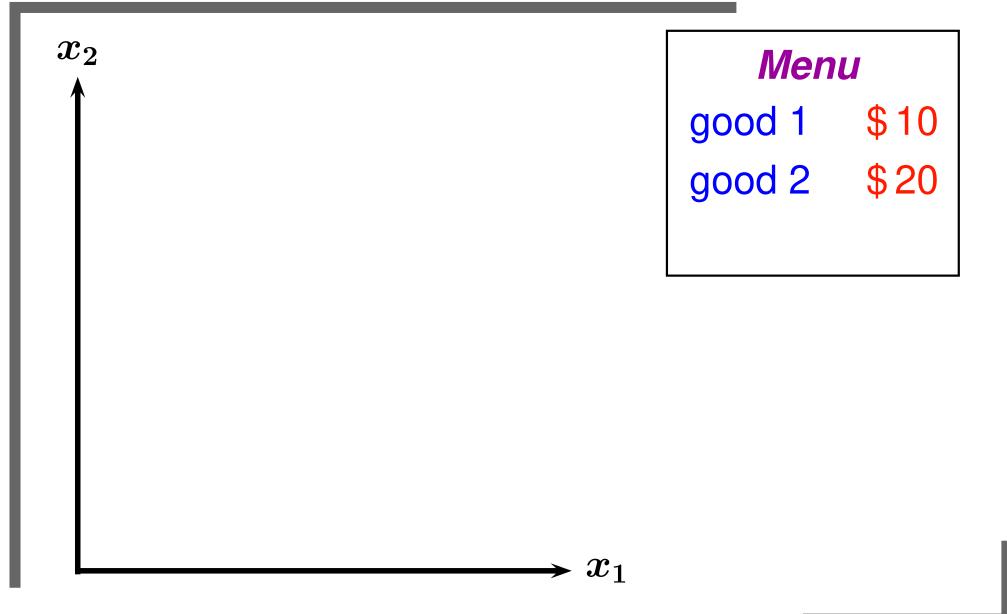
BUYER's willingness to pay increases

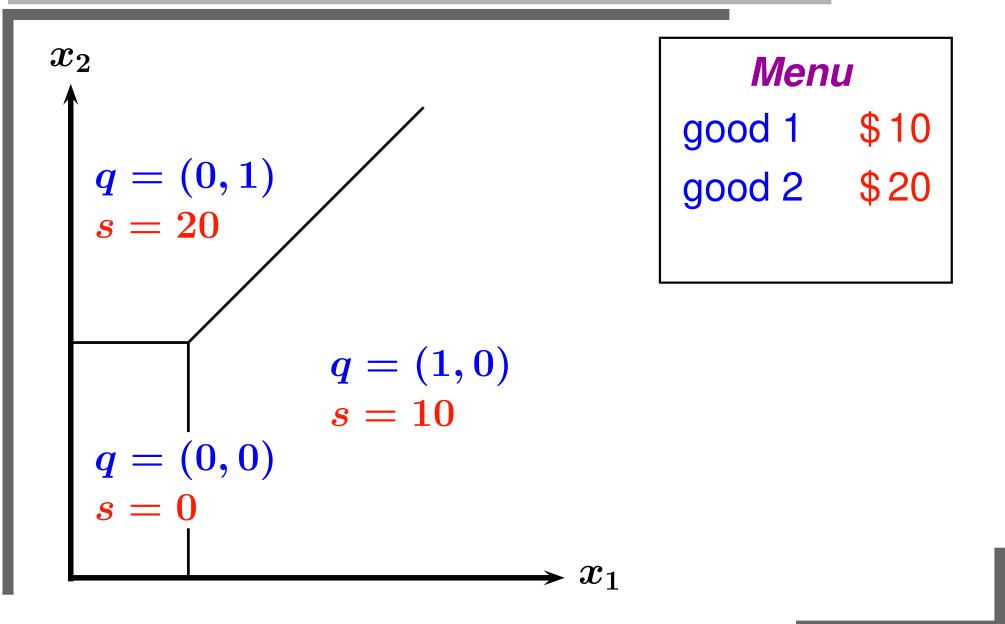
- \Rightarrow **SELLER's revenue increases**
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Hart and Reny 2014

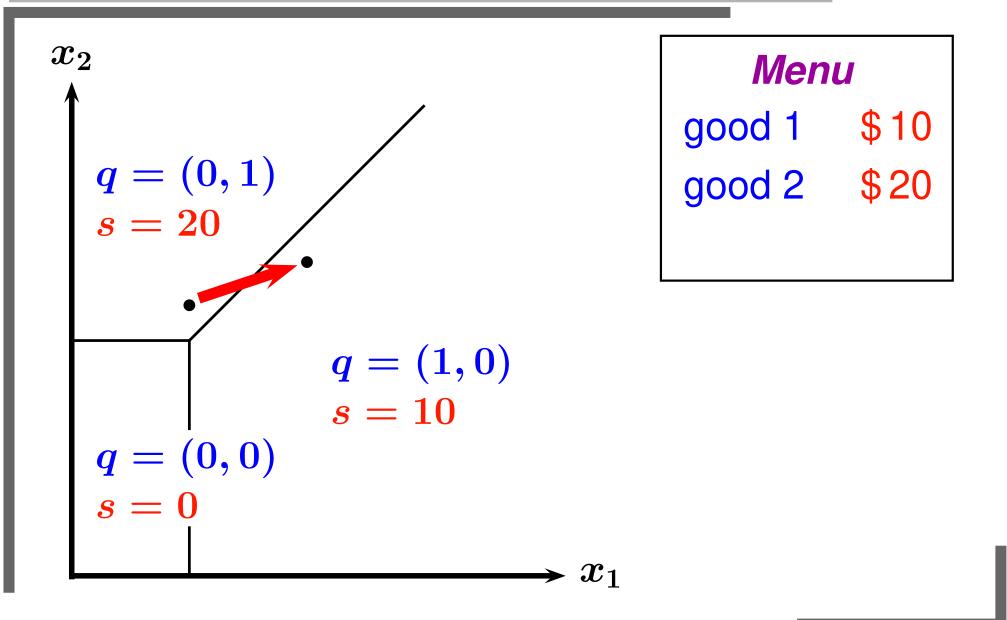
	Menu	
	good 1 good 2	\$10
	good 2	\$20

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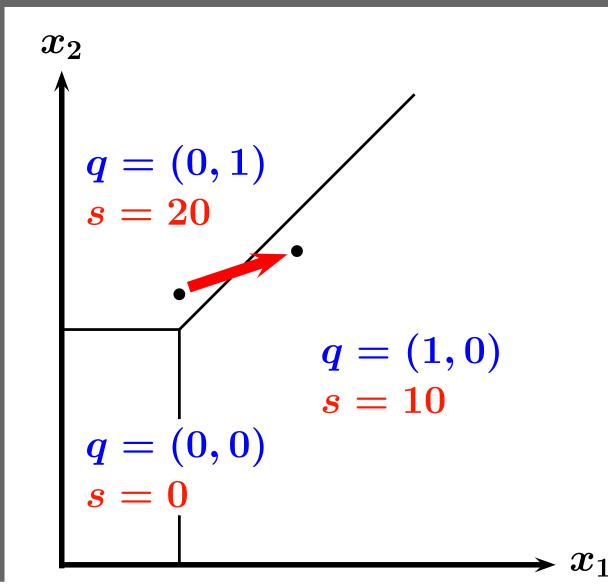




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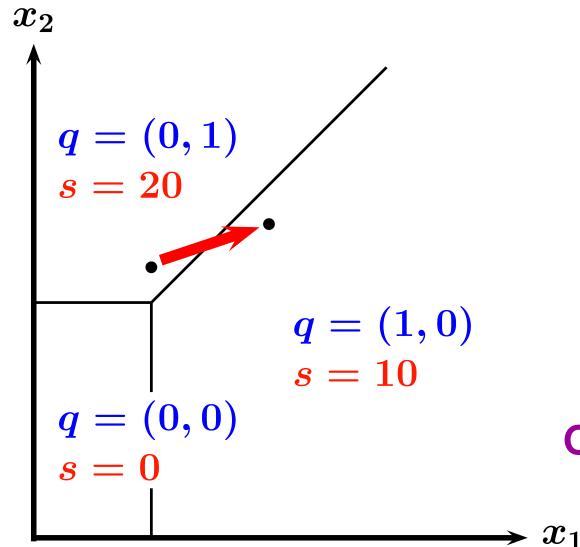
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(10, 23) pays \$20 (20, 27) pays \$10

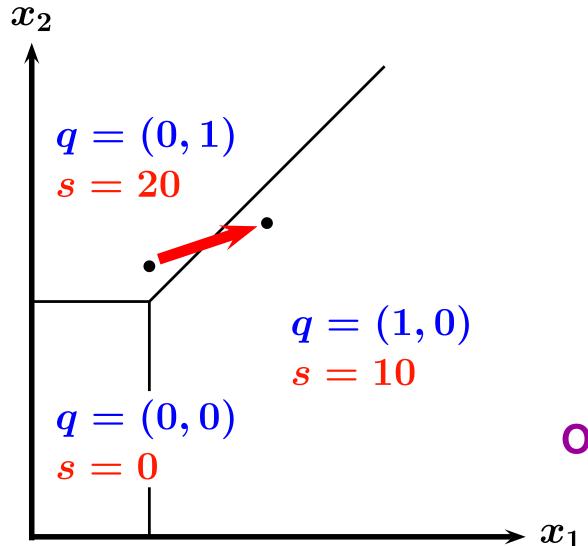
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(10, 23) pays \$20
(20, 27) pays \$10

Optimal for some *X***?**

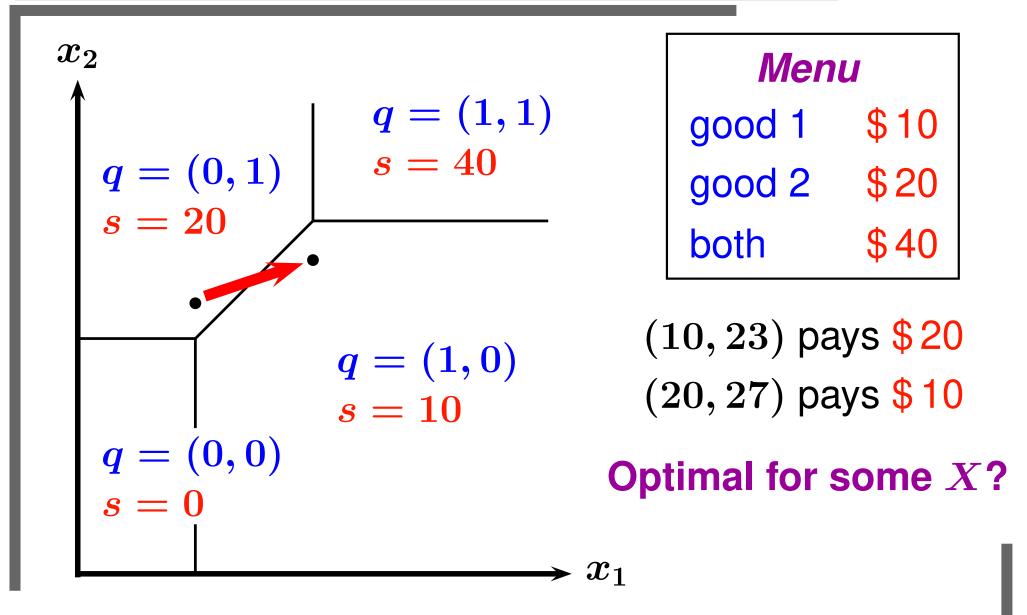


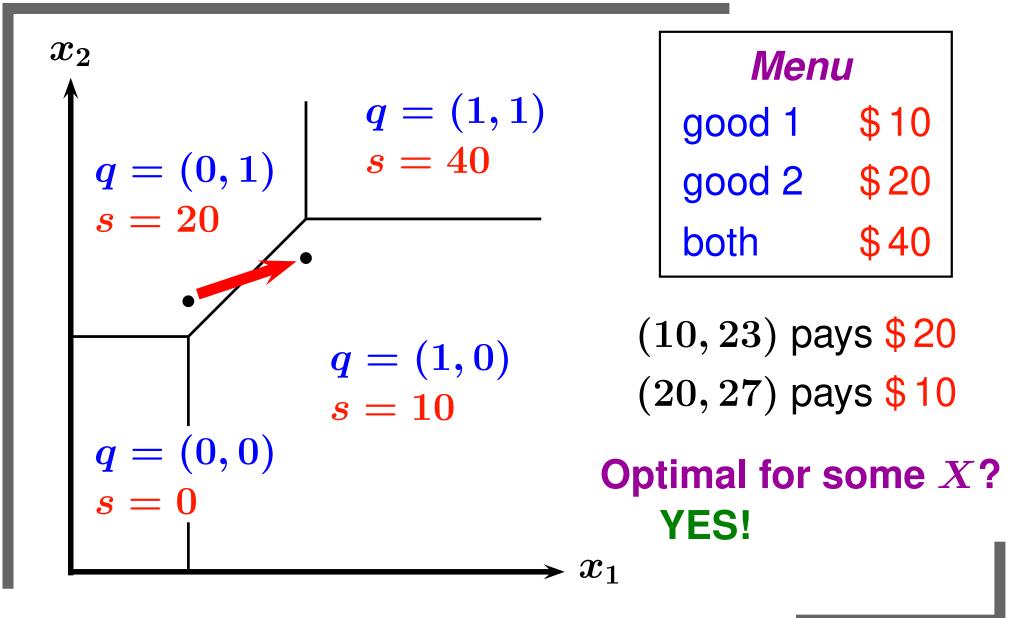


(10, 23) pays \$20 (20, 27) pays \$10

Optimal for some X? No!

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There are simple 2-good valuations X for which the above NON-MONOTONIC mechanism MAXIMIZES REVENUE

Non-Monotonicity

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 - moreover: unique maximizer; robust

Non-Monotonicity

- There are simple 2-good valuations X for which the above NON-MONOTONIC mechanism MAXIMIZES REVENUE
 - moreover: unique maximizer; robust
- There are simple 2-good valuations X, X' such that

 $X' \ge X$ but $\mathsf{Rev}(X') < \mathsf{Rev}(X)$

Conclusion: NON-MONOTONIC mechanisms are <u>needed</u> in order to obtain the maximal revenue

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- Conclusion: NON-MONOTONIC mechanisms are <u>needed</u> in order to obtain the maximal revenue
- Question: How much additional revenue can one gain by using NON-MONOTONIC mechanisms?
 - Answer 1: a non-negligible amount
 - Answer 2: <u>most</u> of the revenue !



The Setup

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(Direct) *mechanism* $\mu = (q, s)$:

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Allocation function

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$$s:\mathbb{R}^k_+ o\mathbb{R}$$

• s(x) = payment from **BUYER** with valuation x to **SELLER**



BUYER payoff function

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$$\mathbf{b}(x) = \mathbf{q}(x) \cdot x - \mathbf{s}(x)$$

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• INDIVIDUAL RATIONALITY (IR) $b(x) \ge 0$ for all x

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• INDIVIDUAL RATIONALITY (IR) $b(x) \geq 0$ for all x

INCENTIVE COMPATIBILITY (IC) $q(y) \cdot x - s(y) \leq b(x) \text{ for all } x, y$



•		

Optimal Revenue

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Payoff of SELLER from mechanism $\mu = (q, s)$

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Optimal REVENUE

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Payoff of SELLER from mechanism $\mu = (q, s)$

$${f R}(\mu;X):=\mathbb{E}[{f s}(X)]$$

Optimal REVENUE

$$\mathsf{Rev}(X) := \sup_{\mu} \mathbb{R}(\mu; X)$$

supremum is taken over all (IR and IC) mechanisms μ

• A mechanism $\mu = (q, s)$ is **MONOTONIC** if its payment function *s* is nondecreasing:

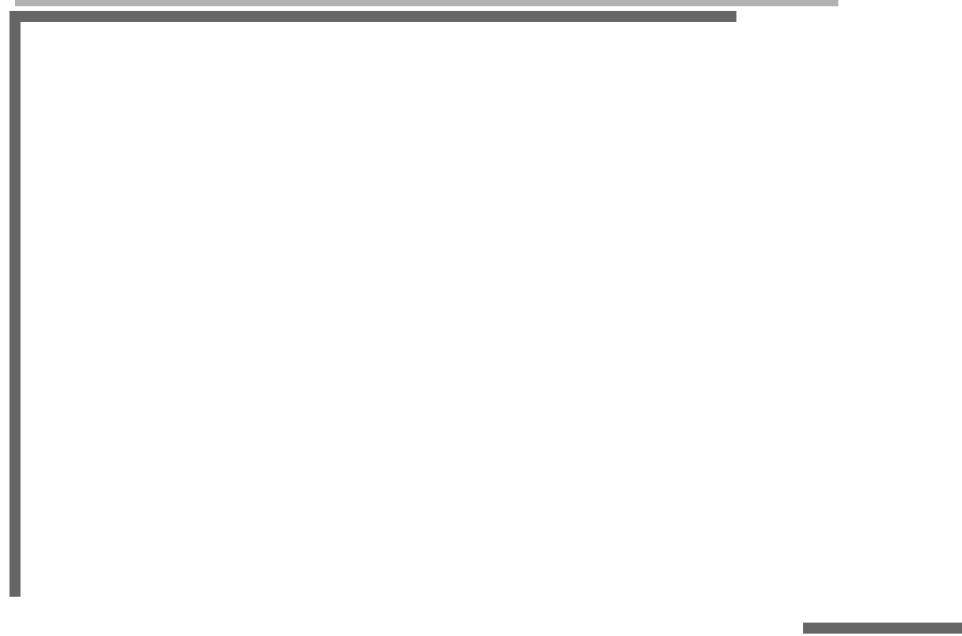
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 $x \geq y$ implies $\mathbf{s}(x) \geq \mathbf{s}(y)$

• MONREV(X) := maximal revenueobtainable by **MONOTONIC** mechanisms

Monotonicity of Revenue



Monotonicity of Revenue

Claim. If $X \ge Y$ (more generally: if X first order stochastically dominates Y) then

$\mathsf{MONRev}(X) \geq \mathsf{MONRev}(Y)$

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Proof. For every monotonic mechanism μ : $X \ge Y \Rightarrow s(X) \ge s(Y)$

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Proof. For every monotonic mechanism μ : $X \ge Y \Rightarrow s(X) \ge s(Y)$ $\Rightarrow \mathbb{E}[s(X)] \ge \mathbb{E}[s(Y)]$ $R(\mu; X) > R(\mu; Y)$

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Proof 2. For every price p $p \cdot \mathbb{P}[X > p] \ge p \cdot \mathbb{P}[Y > p]$



Monotonic Revenue

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Theorem. For every k-good valuation X $MONREV(X) \leq k \cdot BREV(X)$

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Proof. Put $X^{max} := max_{1 \leq i \leq k}X_i$.

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Theorem. For every k-good valuation X $MONREV(X) \leq k \cdot BREV(X)$

Proof. Put $X^{max} := max_{1 \leq i \leq k}X_i$. Then: $\mathsf{MONREV}(X_1,...,X_k)$

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Then: $MONREV(X_1, ..., X_k)$ $\leq MONREV(X^{max}, ..., X^{max})$

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Then: $\mathsf{MONREV}(X_1, ..., X_k)$ $\leq \mathsf{MONREV}(X^{max}, ..., X^{max})$ $\leq \mathsf{REV}(X^{max}, ..., X^{max})$ $= k \cdot \mathsf{REV}(X^{max})$

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Monotonic Revenue

Monotonic Revenue

Corollary. Let $k \geq 2$.

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• There are k-good valuations X such that $\mathsf{MONREV}(X) = 1 \quad \text{and} \quad \mathsf{REV}(X) = \infty$

Corollary. Let $k \geq 2$.

There are k-good valuations X such that
 MONREV(X) = 1 and REV(X) = ∞

 For every ε > 0 there are bounded X s.t.

 $\mathsf{MonRev}(X) < \varepsilon \cdot \mathsf{Rev}(X)$

Corollary. Let k > 2.

Proof.

• There are k-good valuations X such that **MONREV**(X) = 1 and $\text{REV}(X) = \infty$

• For every $\varepsilon > 0$ there are bounded X s.t. $\mathsf{MONRev}(X) < \varepsilon \cdot \mathsf{Rev}(X)$

 $\frac{\mathsf{MONRev}}{\mathsf{Rev}} \leq k \cdot \frac{\mathsf{BRev}}{\mathsf{Rev}}$ Use Hart and Nisan 2013/2019 (Briest et al 2010/2015 for k > 3) for **BREV**

Corollary. Let $k \geq 2$.

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 MONREV(X) < ε ⋅ REV(X)</p>

• There are bounded X such that $\mathsf{MONRev}(X) \leq \frac{k^2}{2^k-1} \cdot \mathsf{DRev}(X)$

Corollary. Let $k \geq 2$.

There are k-good valuations X such that
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 For every ε > 0 there are bounded X s.t.

 $\mathsf{MonRev}(X) < \varepsilon \cdot \mathsf{Rev}(X)$

$$\mathsf{MONRev}(X) \leq rac{k^2}{2^k-1} \cdot \mathsf{DRev}(X)$$

Proof. Use Hart and Nisan 2013/2019

Monotonic vs. Separate



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Monotonic vs. Separate

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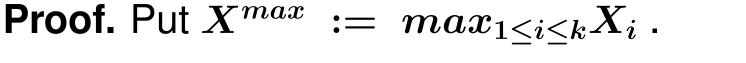
Then: $MonRev(X_1,...,X_k)$ $\leq MonRev(X^{max},...,X^{max})$ $\leq Rev(X^{max},...,X^{max})$ $= k \cdot Rev(X^{max})$

Monotonic vs. Separate

Theorem. For every k-good valuation X**MONREV** $(X) \le k \cdot SREV(X)$

Proof. Put
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Then: $MONREV(X_1, ..., X_k)$ $\leq MONREV(X^{max}, ..., X^{max})$ $\leq REV(X^{max}, ..., X^{max})$ $= k \cdot REV(X^{max})$ $\leq k \cdot (REV(X_1) + ... + REV(X_k))$



Theorem. For every k-good valuation X

 $MONREV(X) \le k \cdot SREV(X)$

Then: $MonRev(X_1,...,X_k) \leq MonRev(X^{max},...,X^{max}) \leq Rev(X^{max},...,X^{max}) = k \cdot Rev(X^{max}) \leq k \cdot (Rev(X_1) + ... + Rev(X_k))$

Monotonic vs. Separate

 $\mathbb{P}[X^{max} > p] \leq \mathbb{P}[X_1 > p] + ... + \mathbb{P}[X_k > p]$

Monotonic vs. Separate

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$$X^{max} \ := \ max_{1 \leq i \leq k} X_i$$
 .

Then: $MONREV(X_1, ..., X_k)$ $\leq MONREV(X^{max}, ..., X^{max})$ $\leq REV(X^{max}, ..., X^{max})$ $= k \cdot REV(X^{max})$ $\leq k \cdot (REV(X_1) + ... + REV(X_k))$

 $p \cdot \mathbb{P}[X^{max} > p] \leq p \cdot (\mathbb{P}[X_1 > p] + ... + \mathbb{P}[X_k > p])$

Theorem. For every k-good valuation X MONREV $(X) \leq k \cdot \min\{BREV(X), SREV(X)\}$

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SERGIU HART © 2022 - p. 34

Theorem. For every k-good valuation X MONREV $(X) \leq k \cdot \min\{\mathsf{BREV}(X), \mathsf{SREV}(X)\}$

- Tight?
 - BREV: Yes

Theorem. For every k-good valuation X MONREV $(X) \leq k \cdot \min\{\mathsf{BREV}(X), \mathsf{SREV}(X)\}$

Jight?

BREV: Yes

There are k i.i.d. goods s.t.

 $\mathsf{SRev}(X) > (k - \varepsilon) \mathsf{BRev}(X)$

(Hart and Nisan 2012/2017)

Theorem. For every k-good valuation X MONREV $(X) \leq k \cdot \min\{\mathsf{BREV}(X), \mathsf{SREV}(X)\}$

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There are k i.i.d. goods s.t. $MONREV(X) \ge SREV(X) > (k - \varepsilon)BREV(X)$ (Hart and Nisan 2012/2017)

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- Jight?
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Jight?

BREV: Yes

SREV: ??

There are k i.i.d. goods s.t.

 $\mathsf{BRev}(X) \geq \Omega(\log k) {\boldsymbol{\cdot}} \mathsf{SRev}(X)$

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 - BREV: Yes
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 $\mathsf{MONRev}(X) \geq \mathsf{BRev}(X) \geq \Omega(\log k) \cdot \mathsf{SRev}(X)$

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Theorem. For every k-good valuation X MONREV $(X) \leq k \cdot \min\{\mathsf{BREV}(X), \mathsf{SREV}(X)\}$

- Tight?
 - BREV: Yes
 - SREV: ?? [between $\Omega(\log k)$ and k]

There are k i.i.d. goods s.t.

 $\mathsf{MONRev}(X) \geq \mathsf{BRev}(X) \geq \Omega(\log k) \cdot \mathsf{SRev}(X)$

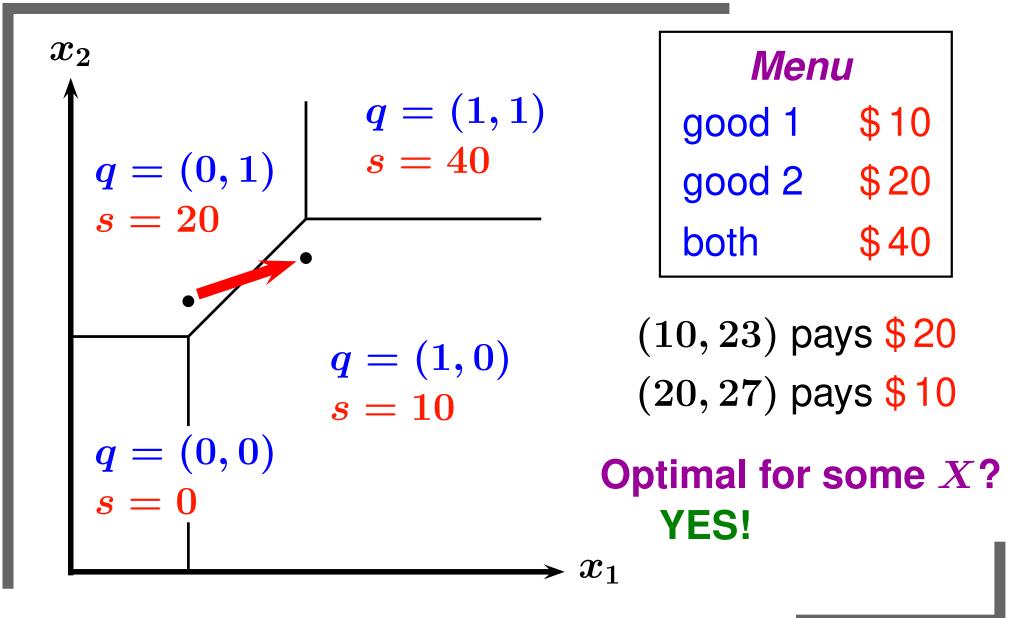
(Hart and Nisan 2012/2017)

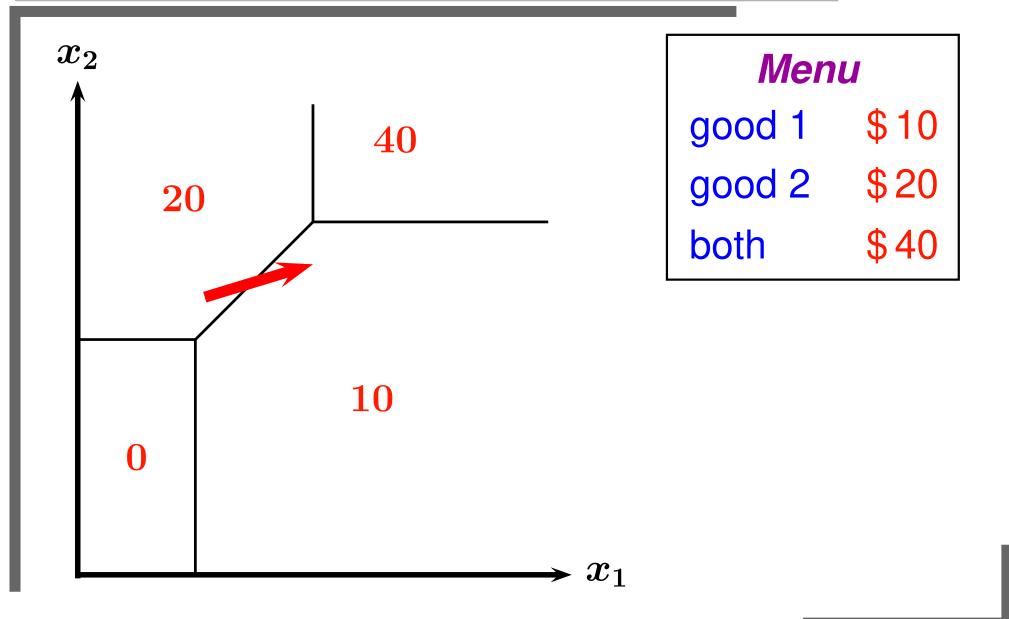
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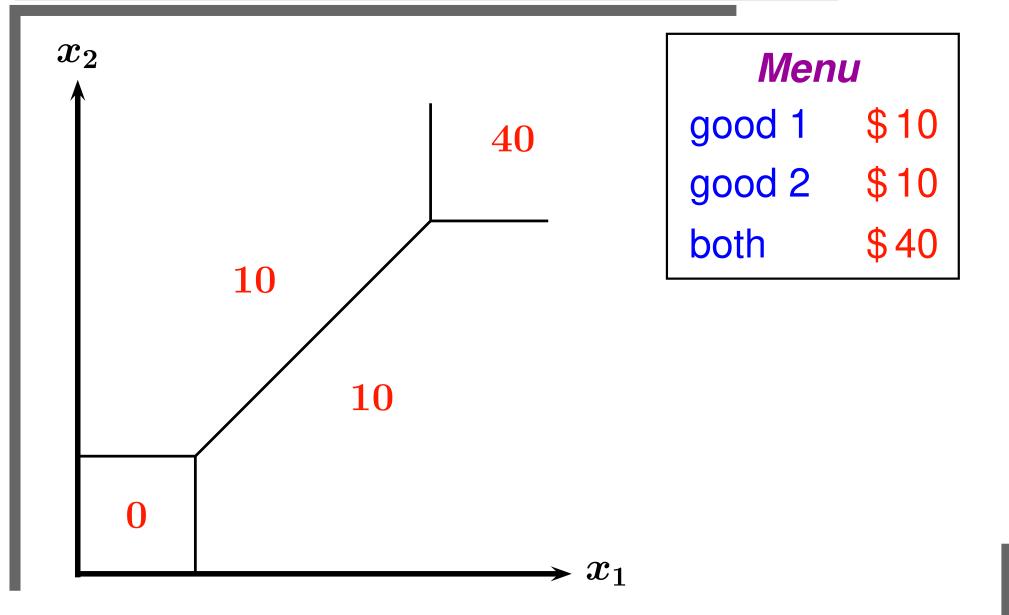
- Jight?
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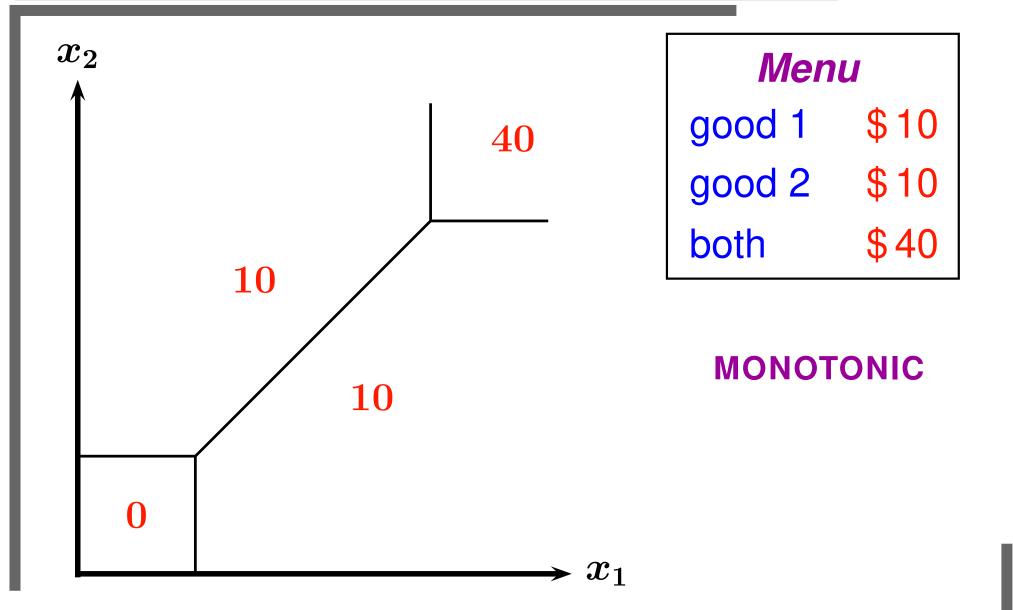


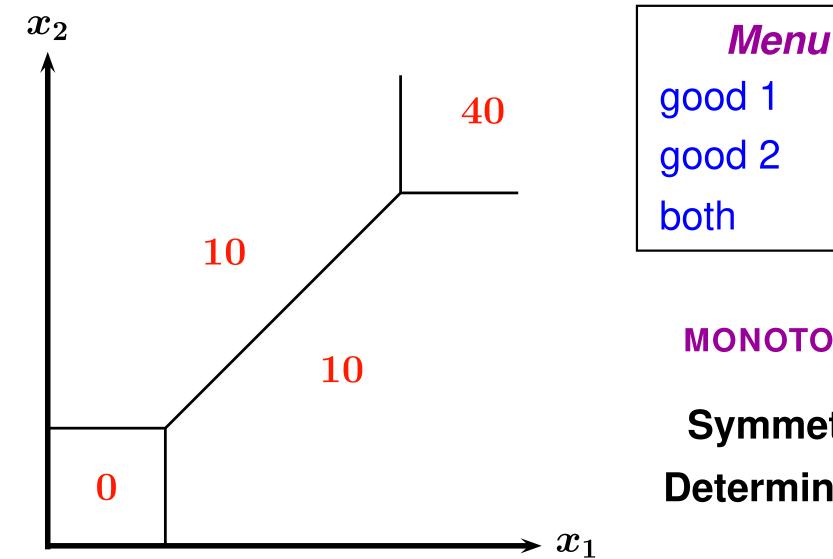
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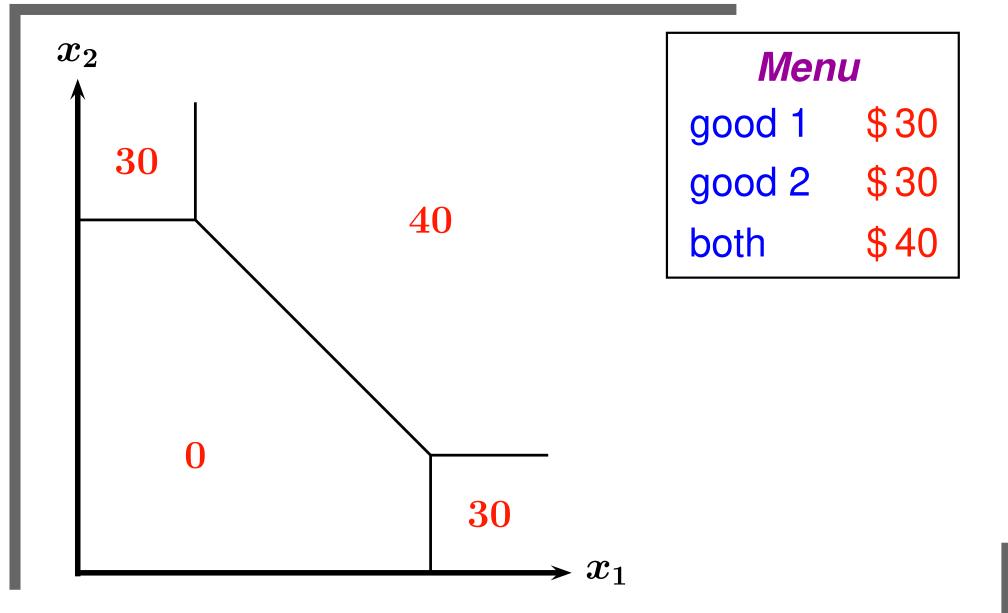


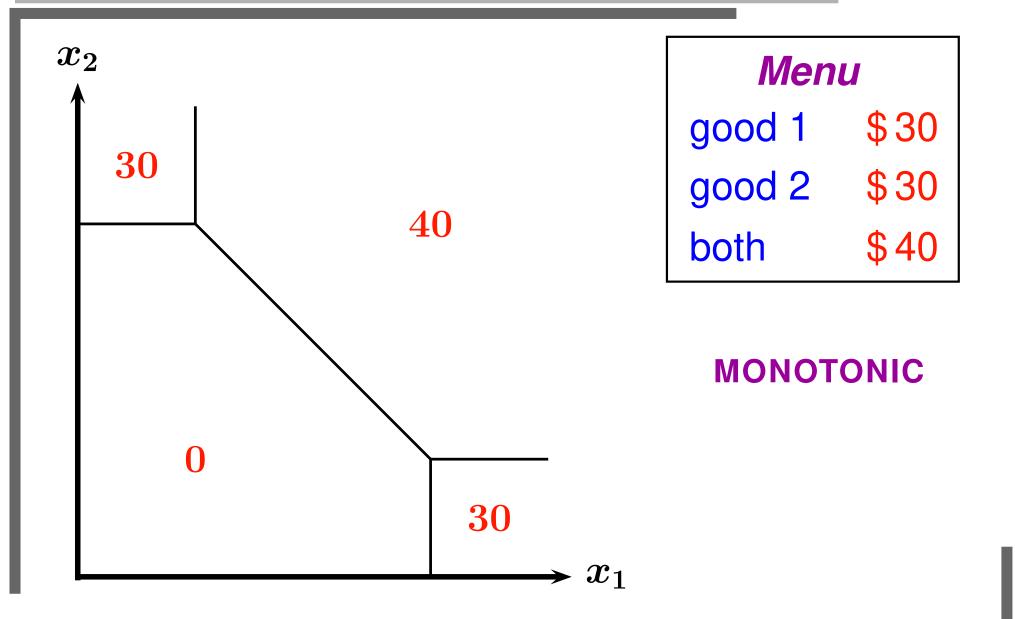
\$10 good 2 \$10 \$40

MONOTONIC

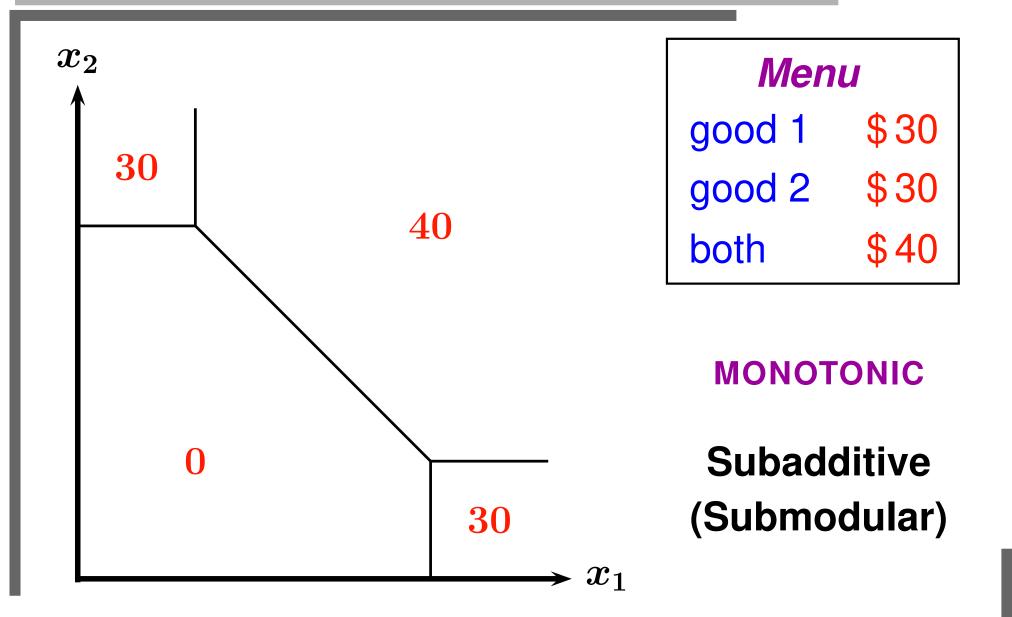
Symmetric Deterministic

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SERGIU HART C 2022 - p. 39



Hart and Reny 2014

SERGIU HART ⓒ 2022 - p. 40

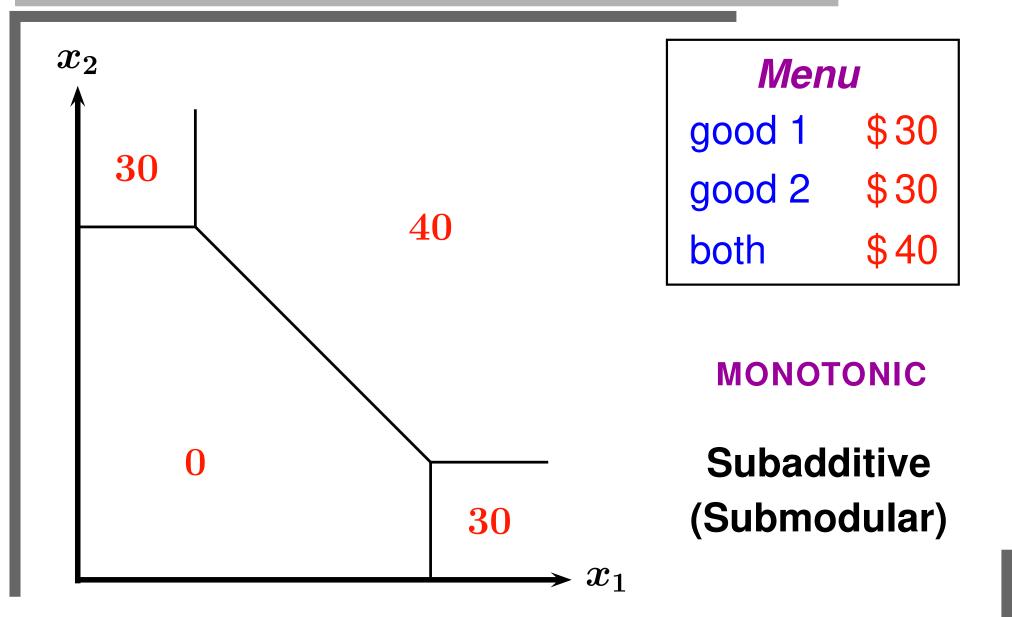
Symmetric deterministic mechanisms

Hart and Reny 2014

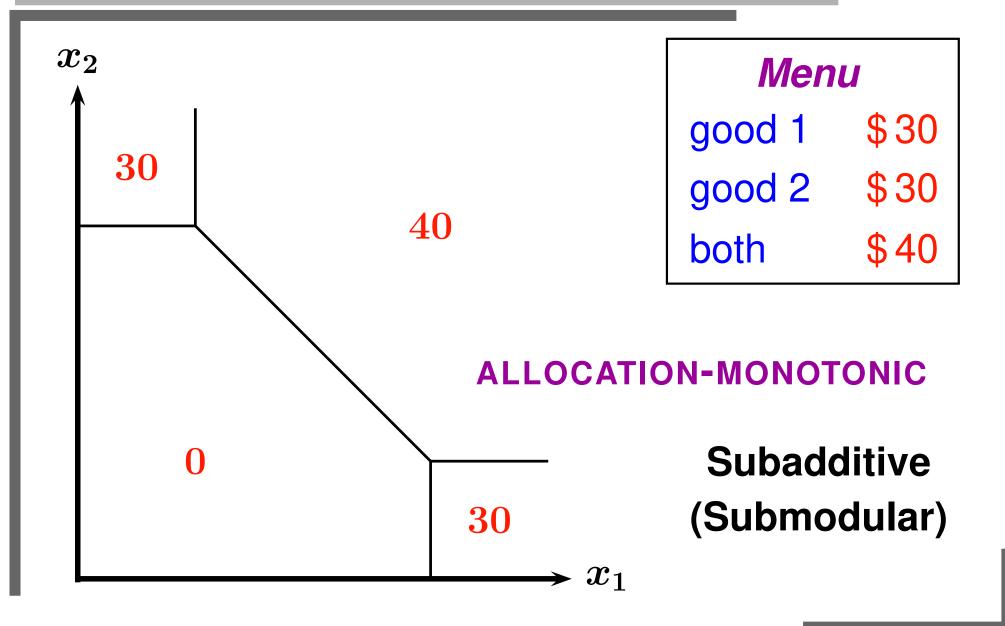
SERGIU HART C 2022 - p. 40

- Symmetric deterministic mechanisms
- Submodular mechanisms

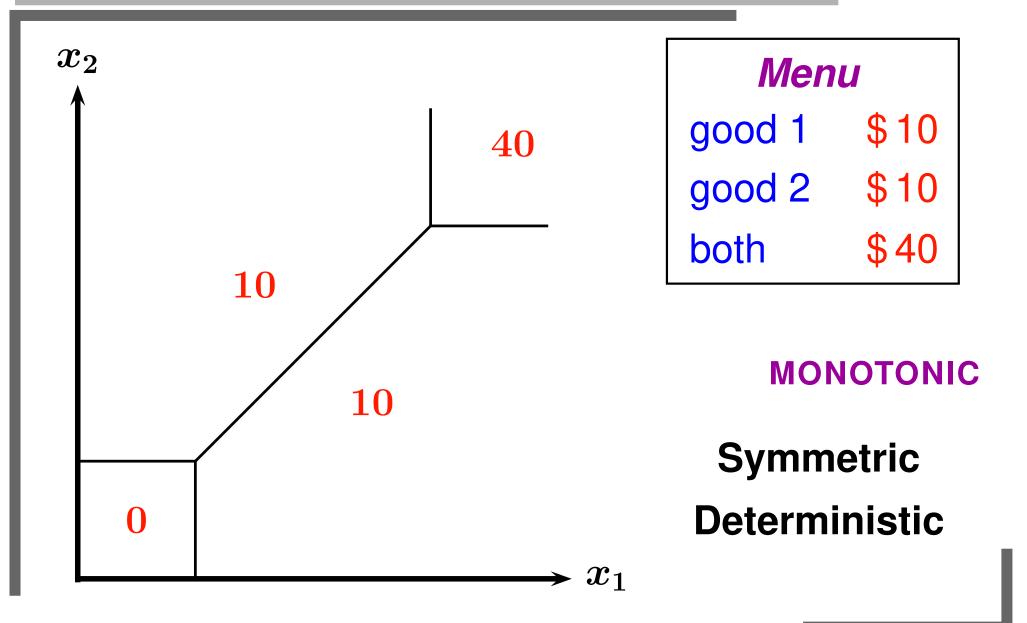
Hart and Reny 2014



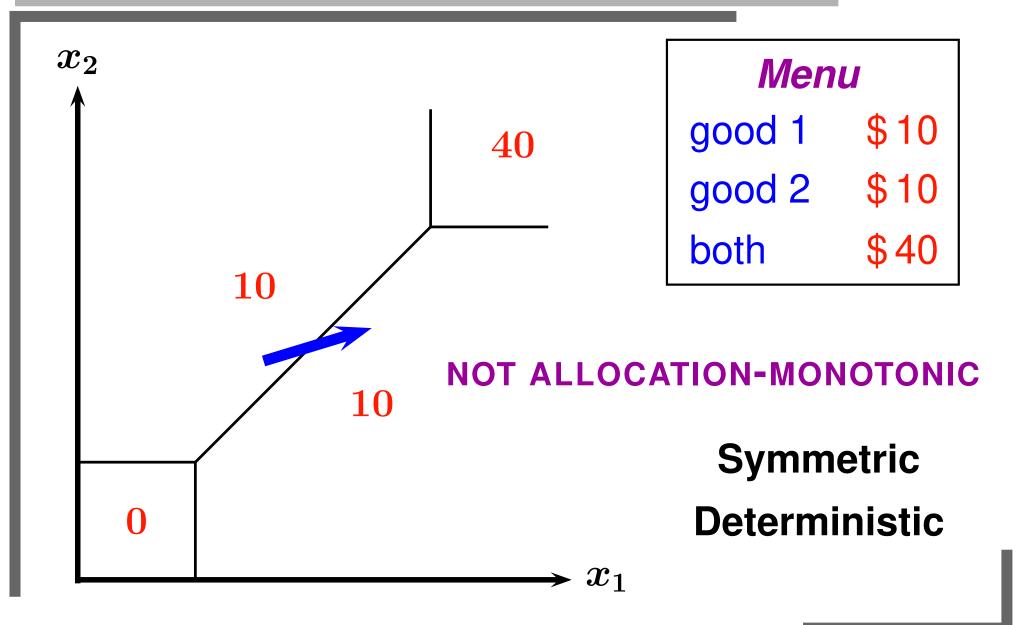
Allocation-Monotonic Mechanism



Monotonic



NOT Allocation-Monotonic





Allocation-Monotonic Mechanisms

SERGIU HART © 2022 - p. 43

• A mechanism $\mu = (q, s)$ is **MONOTONIC** if its payment function *s* is nondecreasing:

 $x \geq y$ implies $\mathbf{s}(x) \geq \mathbf{s}(y)$

Allocation-Monotonic Mechanisms

• A mechanism $\mu = (q, s)$ is **MONOTONIC** if its payment function *s* is nondecreasing:

$$x \geq y$$
 implies ${\color{black} s}(x) \geq {\color{black} s}(y)$

• A mechanism $\mu = (q, s)$ is **ALLOCATION MONOTONIC** if its allocation function q is nondecreasing:

$$x \geq y$$
 implies $\mathbf{q}(x) \geq \mathbf{q}(y)$

Allocation-Monotonic Mechanisms

• A mechanism $\mu = (q, s)$ is **MONOTONIC** if its payment function *s* is nondecreasing:

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• A mechanism $\mu = (q, s)$ is **ALLOCATION MONOTONIC** if its allocation function q is nondecreasing:

$$x \geq y$$
 implies $q(x) \geq q(y)$

• allocation monotonicity \Rightarrow monotonicity

Allocation-Monotonic Mechanisms

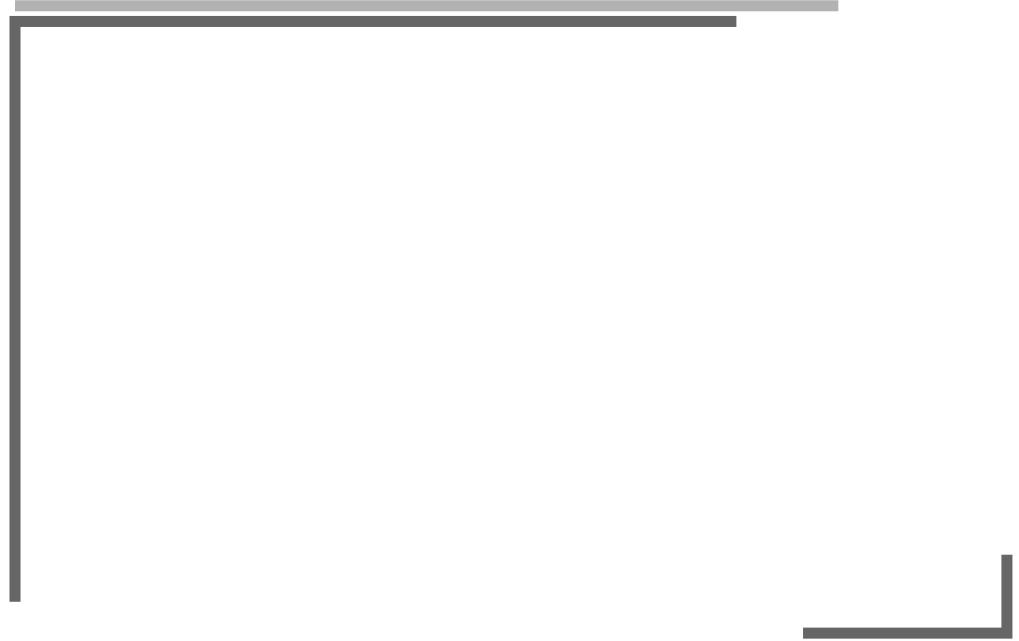
• A mechanism $\mu = (q, s)$ is **MONOTONIC** if its payment function *s* is nondecreasing:

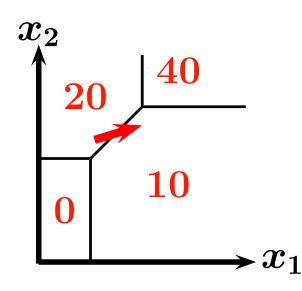
$$x \geq y$$
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• A mechanism $\mu = (q, s)$ is ALLOCATION MONOTONIC if its allocation function q is nondecreasing:

$$x \ge y$$
 implies $q(x) \ge q(y)$

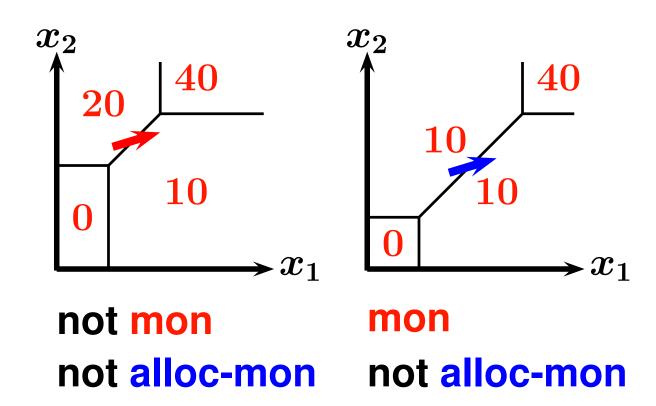
 ■ allocation monotonicity ⇒ monotonicity (by IC)

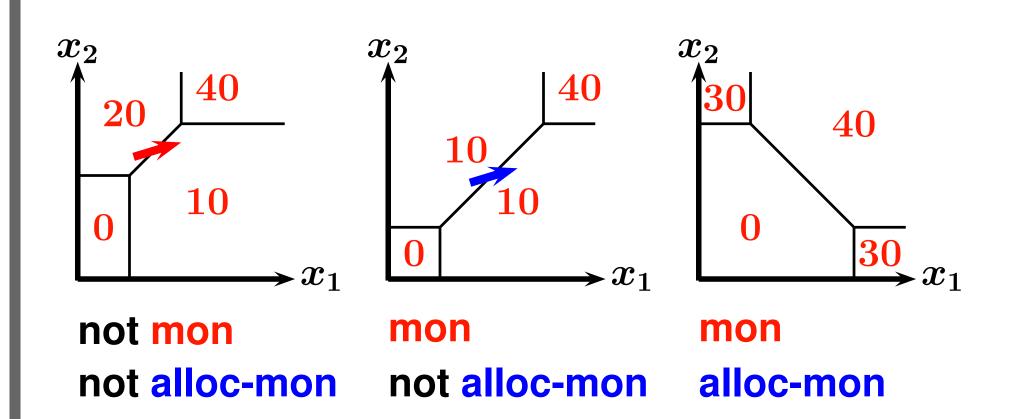




not mon not alloc-mon

SERGIU HART (C) 2022 - p. 45





Deterministic: Pricing

Deterministic: Pricing

Let $\mu = (q, s)$ be a **deterministic** mechanism for k goods. Put $K := \{1, ..., k\}$.

Deterministic: Pricing

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• The PRICE of a set of goods $A \subseteq K$: p(A) := s(x) for x with q(x) = A

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• $p: 2^K \rightarrow [0, \infty]$ is the (canonical) PRICING FUNCTION of μ (nondecreasing function)

• The function p is SUBMODULAR if for all A, B: $p(A) + p(B) \ge p(A \cup B) + p(A \cap B)$

• The function p is SUBMODULAR if for all A, B: $p(A) + p(B) \ge p(A \cup B) + p(A \cap B)$ \Leftrightarrow for all $i, j \notin A$: $p(A \cup \{i\}) - p(A) \ge p(A \cup \{i, j\}) - p(A \cup \{j\})$

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- The mechanism μ is **SUBMODULAR** if its (canonical) pricing function is submodular

Deterministic: Allocation Monotonic

Deterministic: Allocation Monotonic

Let μ be a tie-favorable **deterministic** mechanism

Theorem. μ is allocation monotonic $\Leftrightarrow \mu$ is submodular

 $\{AMON\} = \{SUBMOD\}$

SERGIU HART (C) 2022 - p. 48

General: Allocation Monotonicity



General: Allocation Monotonicity

Let μ be a tie-favorable general (probabilistic) mechanism

Theorem.

- μ is submodular
- $\Rightarrow \mu$ is allocation monotonic
- $\Rightarrow \mu$ is separably subadditive

$\{\texttt{SUBMOD}\} \subset \{\texttt{AMON}\} \subset \{\texttt{SEP SUBADD}\}$





General: Pricing

Let $\mu = (q, s)$ be a mechanism for k goods. • The PRICE of an allocation $g \in [0, 1]^k$: p(g) := s(x) for x with q(x) = g

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 p(g) := s(x) for x with q(x) = g
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 - If g is never allocated, put $p(g) := \sup_x (g \cdot x b(x))$
- $p: [0,1]^k \rightarrow [0,\infty]$ is the (canonical) **PRICING FUNCTION** of μ (nondecreasing, convex, closed function)
 - The convex functions b and p are Fenchel conjugates

Submodular Pricing

Submodular Pricing

The function p is **SUBMODULAR** if

• for all
$$g, h$$
 in $[0, 1]^k$:
 $p(g) + p(h) \ge p(g \lor h) + p(g \land h)$

Submodular Pricing

The function p is **SUBMODULAR** if

- for all g, h in $[0, 1]^k$: $p(g) + p(h) \ge p(g \lor h) + p(g \land h)$
- \Leftrightarrow for all g and orthogonal $d_1, d_2 \ge 0$: $p(g+d_2) - p(g) \ge p(g+d_1+d_2) - p(g+d_1)$

The function p is **SUBMODULAR** if

- for all g, h in $[0, 1]^k$: $p(g) + p(h) \ge p(g \lor h) + p(g \land h)$
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Marginal price of good *i* decreases as allocation of good $j \neq i$ increases The function p is **SUBMODULAR** if

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Marginal price of good *i* decreases as allocation of good $j \neq i$ increases

• If p is differentiable: $rac{\partial^2 p}{\partial g_i \partial g_j} \leq 0$ for all i
eq j



The function p is **SEPARABLY SUBADDITIVE** if

for all orthogonal g, h in $[0, 1]^k$:
 $p(g+h) \leq p(g) + p(h)$

The function p is **SEPARABLY SUBADDITIVE** if

- for all orthogonal g, h in $[0, 1]^k$: $p(g+h) \leq p(g) + p(h)$
- \Leftrightarrow for all g : $p(g) \leq p(g_1, 0, ..., 0) + ... + p(0, ..., 0, g_k)$

The function p is **SEPARABLY SUBADDITIVE** if

- for all orthogonal g, h in $[0, 1]^k$: $p(g+h) \leq p(g) + p(h)$
- \Leftrightarrow for all g : $p(g) \leq p(g_1, 0, ..., 0) + ... + p(0, ..., 0, g_k)$
 - Weaker than subadditivity
 (inequality required for all g, h)

The function p is **SEPARABLY SUBADDITIVE** if

• for all orthogonal g, h in $[0, 1]^k$: $p(g+h) \leq p(g) + p(h)$

$$\Leftrightarrow$$
 for all g : $p(g) \leq p(g_1, 0, ..., 0) + ... + p(0, ..., 0, g_k)$

- Weaker than subadditivity
 (inequality required for all g, h)
- Weaker than submodularity (by p(0) = 0)

Sub... Mechanisms

SERGIU HART © 2022 – p. 53

Sub... Mechanisms

• A mechanism μ is **SUBMODULAR** if its (canonical) pricing function is submodular

Sub... Mechanisms

- A mechanism μ is **SUBMODULAR** if its (canonical) pricing function is submodular
- A mechanism μ is **SEPARABLY SUBADDITIVE** if its (canonical) pricing function is separably subadditive

Allocation Monotonicity



Allocation Monotonicity

Let μ be a tie-favorable **deterministic** mechanism

Theorem. μ is allocation monotonic $\Leftrightarrow \mu$ is submodular

 $\{AMON\} = \{SUBMOD\}$

SERGIU HART (C) 2022 - p. 54







Deterministic mechanisms

 μ allocation monotonic rightarrow [*] b supermodular rightarrow [FC] p submodular $(\mu$ submodular)

SERGIU HART (C) 2022 - p. 55



[FC] = p and b are Fenchel Conjugates

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[*] μ allocation monotonic $\Leftrightarrow b$ supermodular



- [*] μ allocation monotonic $\Leftrightarrow b$ supermodular
 - **Proof.** Assume that **b** is differentiable, then $q = \nabla b$

[*] μ allocation monotonic $\Leftrightarrow b$ supermodular

• **Proof.** Assume that *b* is differentiable, then $q = \nabla b$ (because $b(x) = \max_y(q(y) \cdot x - s(y))$ $= q(x) \cdot x - s(x)$).

[*] μ allocation monotonic $\Leftrightarrow b$ supermodular

• **Proof.** Assume that **b** is differentiable, then $q = \nabla b$ (because $b(x) = \max_y(q(y) \cdot x - s(y))$ $= q(x) \cdot x - s(x)$).

Then:

[*] μ allocation monotonic $\Leftrightarrow b$ supermodular

• **Proof.** Assume that **b** is differentiable, then $q = \nabla b$ (because $b(x) = \max_y(q(y) \cdot x - s(y))$ $= q(x) \cdot x - s(x)$).

Then: **q** nondecreasing

[*] μ allocation monotonic $\Leftrightarrow b$ supermodular

• **Proof.** Assume that *b* is differentiable, then $q = \nabla b$ (because $b(x) = \max_y(q(y) \cdot x - s(y))$ $= q(x) \cdot x - s(x)$).

Then: q nondecreasing $\Leftrightarrow \nabla q \ge 0$

[*] μ allocation monotonic $\Leftrightarrow b$ supermodular

• **Proof.** Assume that *b* is differentiable, then $q = \nabla b$ (because $b(x) = \max_y(q(y) \cdot x - s(y))$ $= q(x) \cdot x - s(x)$).

Then: \boldsymbol{q} nondecreasing $\Leftrightarrow \ \boldsymbol{\nabla} \boldsymbol{q} \ge 0$ $\Leftrightarrow \ \boldsymbol{\nabla}^2 \boldsymbol{b} \ge 0$

[*] μ allocation monotonic $\Leftrightarrow b$ supermodular

• **Proof.** Assume that **b** is differentiable, then $q = \nabla b$ (because $b(x) = \max_y(q(y) \cdot x - s(y))$ $= q(x) \cdot x - s(x)$).

Then: q nondecreasing $\Leftrightarrow \nabla q \ge 0$ $\Leftrightarrow \nabla^2 b \ge 0$ $\Leftrightarrow b \ge 0$

 \Leftrightarrow **b** supermodular

[*] μ allocation monotonic $\Leftrightarrow b$ supermodular

• **Proof.** Assume that **b** is differentiable, then $q = \nabla b$ (because $b(x) = \max_y(q(y) \cdot x - s(y))$ $= q(x) \cdot x - s(x)$).

Then: q nondecreasing

$$\Leftrightarrow \nabla \boldsymbol{q} \ge 0 \\ \Leftrightarrow \nabla^2 \boldsymbol{b} \ge 0$$

 \Leftrightarrow **b** supermodular

Without differentiability: "tie-favorable"



```
\mu allocation monotonic

rightarrow [*]

b supermodular

rightarrow [FC]

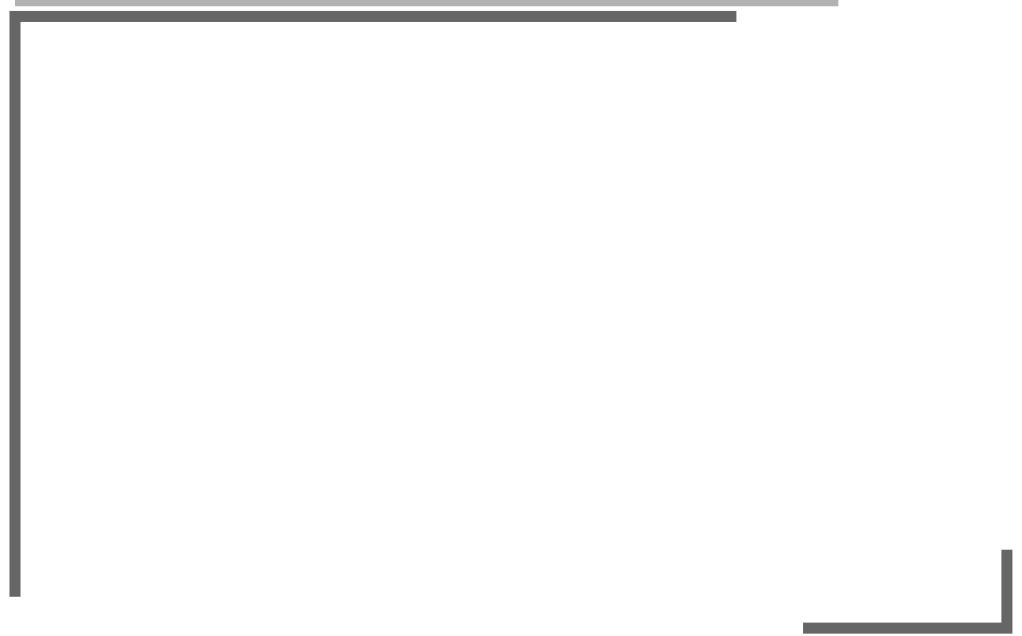
p submodular

(\mu submodular)
```

[FC] = p and b are Fenchel Conjugates

SERGIU HART C 2022 - p. 57

General: Allocation Monotonicity



General: Allocation Monotonicity

Let μ be a tie-favorable general (probabilistic) mechanism

Theorem.

- μ is submodular
- $\Rightarrow \mu$ is allocation monotonic
- $\Rightarrow \mu$ is separably subadditive

$\{\texttt{SUBMOD}\} \subset \{\texttt{AMON}\} \subset \{\texttt{SEP SUBADD}\}$





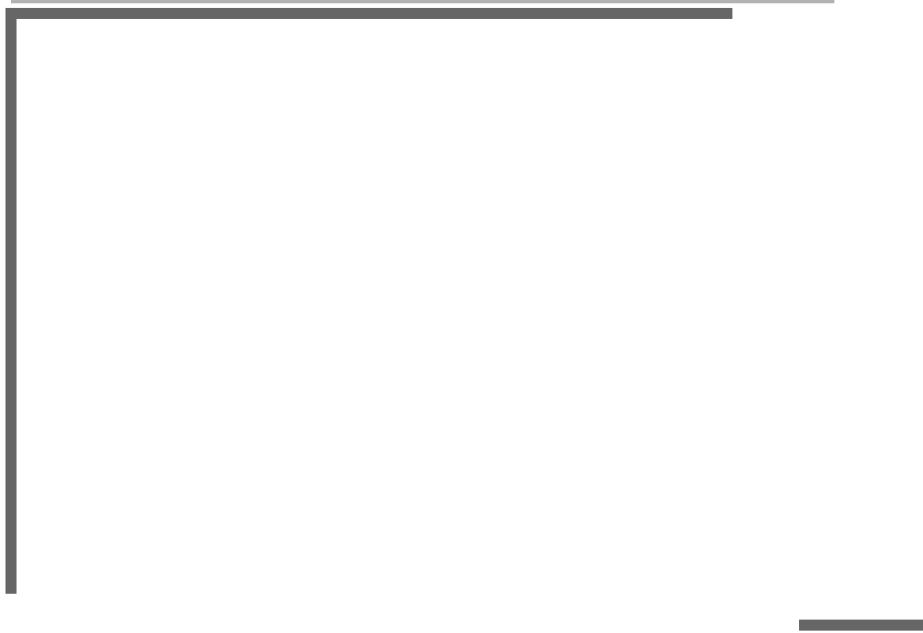


General mechanisms

 $\begin{array}{ll} \mu \text{ allocation monotonic} \\ & \updownarrow [*] \\ \textbf{b} \text{ supermodular } \Rightarrow \textbf{b} \text{ separably superadditive} \\ & \uparrow [\mathsf{FC}] & & \updownarrow [\mathsf{FC}] \\ \textbf{p} \text{ submodular } \Rightarrow \textbf{p} \text{ separably subadditive} \\ (\mu \text{ submodular}) & & (\mu \text{ separably subadditive}) \end{array}$

[FC] = p and b are Fenchel Conjugates









b and **p** are Fenchel conjugates

• **b** and **p** are Fenchel conjugates $\Rightarrow \nabla^2 \mathbf{b} = (\nabla^2 \mathbf{p})^{-1}$

• **b** and **p** are Fenchel conjugates $\Rightarrow \nabla^2 b = (\nabla^2 p)^{-1}$

Therefore:

• **b** and **p** are Fenchel conjugates $\Rightarrow \nabla^2 \mathbf{b} = (\nabla^2 \mathbf{p})^{-1}$

Therefore: *p* submodular

- **b** and **p** are Fenchel conjugates $\Rightarrow \nabla^2 b = (\nabla^2 p)^{-1}$
- Therefore:
 p submodular
 \Leftrightarrow Off-diagonal entries of $\nabla^2 p$ are ≤ 0

- **b** and **p** are Fenchel conjugates $\Rightarrow \nabla^2 \mathbf{b} = (\nabla^2 \mathbf{p})^{-1}$
- Therefore:
 - *p* submodular
 - \Leftrightarrow Off-diagonal entries of $\nabla^2 p$ are ≤ 0
 - \Rightarrow Off-diagonal entries of $(\nabla^2 p)^{-1}$ are ≥ 0

- **b** and **p** are Fenchel conjugates $\Rightarrow \nabla^2 b = (\nabla^2 p)^{-1}$
- Therefore:
 - **p** submodular
 - \Leftrightarrow Off-diagonal entries of $\nabla^2 p$ are ≤ 0
 - \Rightarrow Off-diagonal entries of $(\nabla^2 p)^{-1}$ are ≥ 0
 - \Leftrightarrow **b** supermodular

- **b** and **p** are Fenchel conjugates $\Rightarrow \nabla^2 \mathbf{b} = (\nabla^2 \mathbf{p})^{-1}$
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 - **p** submodular
 - \Leftrightarrow Off-diagonal entries of $\nabla^2 p$ are ≤ 0
 - \Rightarrow Off-diagonal entries of $(\nabla^2 p)^{-1}$ are ≥ 0
 - \Leftrightarrow **b** supermodular

• \notin : already for QUADRATIC mechanisms $q(x) = Ax, \ s(x) = b(x) = \frac{1}{2}x^T\!Ax, \ p(g) = \frac{1}{2}g^T\!A^{-1}g$



General mechanisms

 $\begin{array}{ll} \mu \text{ allocation monotonic} \\ & \updownarrow [*] \\ \textbf{b} \text{ supermodular } \Rightarrow \textbf{b} \text{ separably superadditive} \\ & \uparrow [\mathsf{FC}] & & \updownarrow [\mathsf{FC}] \\ \textbf{p} \text{ submodular } \Rightarrow \textbf{p} \text{ separably subadditive} \\ (\mu \text{ submodular}) & & (\mu \text{ separably subadditive}) \end{array}$

[FC] = p and b are Fenchel Conjugates

Allocation-Monotonic Revenue

Allocation-Monotonic Revenue

Theorem. For every k-good valuation X $AMONREV(X) \leq 2\ln(2k) \cdot SREV(X)$









🥒 Put

 $p'(g) := p(g_1, 0, ..., 0) + ... + p(0, ..., 0, g_k)$



Put

$$p'(g) := p(g_1, 0, ..., 0) + ... + p(0, ..., 0, g_k)$$

• Then: p' is separable



🥒 Put

$$p'(g) := p(g_1, 0, ..., 0) + ... + p(0, ..., 0, g_k)$$

• Then: p' is separable, and

$$rac{1}{k}p'\leq p\leq p'$$

(p nondecreasing and separably subadditive)



🥒 Put

$$p'(g) := p(g_1, 0, ..., 0) + ... + p(0, ..., 0, g_k)$$

• Then: p' is separable, and

$$rac{1}{k}p'\leq p\leq p'$$

(p nondecreasing and separably subadditive)

Apply a result of Chawla, Teng, and Tzamos







Theorem (*Chawla, Teng, and Tzamos 2019*) Let \mathcal{P}' be a cone of nondecreasing and closed k-good pricing functions. Assume that there are constants $0 < c_1 < c_2 < \infty$ such that for every $p \in \mathcal{P}$ there is $p' \in \mathcal{P}'$ satisfying

$$c_1p'(g)\leq p(g)\leq c_2p'(g)$$

for every g; then

$$\mathcal{P} extsf{-}\mathsf{Rev}(X) \leq 2\ln\left(2rac{c_2}{c_1}
ight)\cdot\mathcal{P}' extsf{-}\mathsf{Rev}(X)$$

for every k-good valuation X.



Proof. Let *p* be the canonical pricing function of an allocation monotonic mechanism.

🥒 Put

$$p'(g) := p(g_1, 0, ..., 0) + ... + p(0, ..., 0, g_k)$$

• Then: p' is separable, and

$$rac{1}{k}p'\leq p\leq p'$$

(p nondecreasing and separably subadditive)

Apply a result of Chawla, Teng, and Tzamos

Allocation-Monotonic Revenue

Theorem. For every k-good valuation X $AMONREV(X) \leq 2\ln(2k) \cdot SREV(X)$





SERGIU HART (C) 2022 - p. 67

Symmetric Deterministic Revenue



Symmetric Deterministic Revenue

Theorem. For every k-good valuation X SYMDREV $(X) \leq O(\log^2 k) \cdot SREV(X)$







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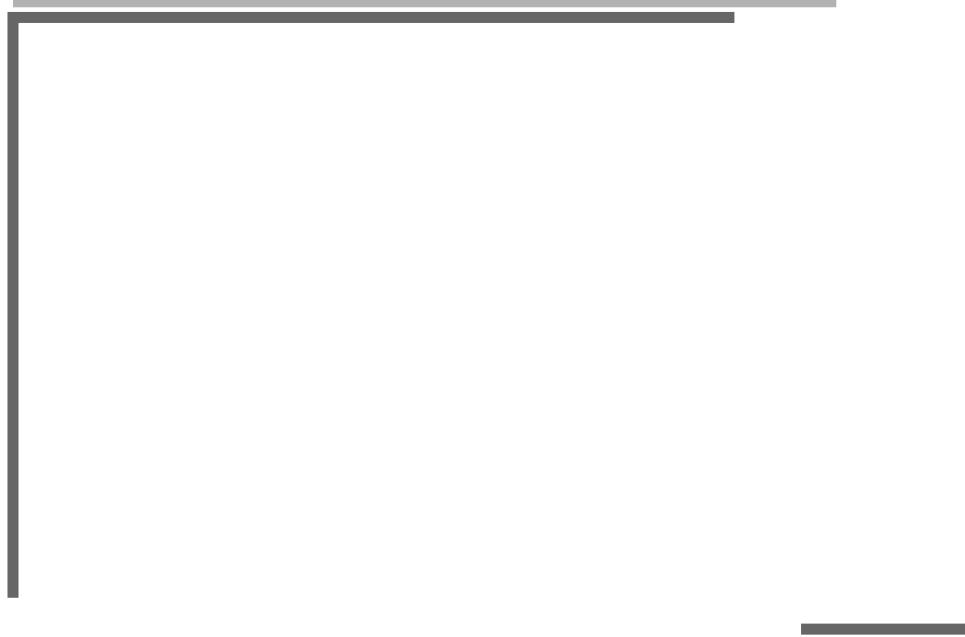


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- Apply the result of Chawla, Teng, and Tzamos



Summary

SERGIU HART © 2022 - p. 70



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