

The Optimality of Upgrade Pricing

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These slides at: bit.ly/upgradeprices
Utilities: github.com/indraos/multi-goods-helpers

Introduction

- ▶ Selling a base product and upgrades that cannot be purchased separately is frequently observed and an, arguably, simple mechanism
- ▶ Upgrade Pricing (UP): Sale of inclusion-ordered bundles
This presentation: When is Upgrade Pricing optimal?
- ▶ We will study **robust**—w.r.t. type distribution for fixed support—optimality
- ▶ Several other approaches exist:
 - ▶ Demand Profiles: Wilson 1993
 - ▶ Monge-Kantorovich Duality: Daskalakis, Deckelbaum, and Tzamos 2017, Kash and Frongillo 2016
 - ▶ Lagrangian Duality: Cai, Devanur, and Weinberg 2016, Carroll 2017, Haghpasand and Hartline 2020 [▶ More](#)

When is Grand Bundling Optimal?

Theorem (Haghpanah and Hartline 2020)

Grand bundling is robustly optimal for any distribution iff the type support is a subset of a line through the origin or a -45 -degree line.

- ▶ Mixed bundling often dominates separate pricing McAfee, McMillan, and Whinston 1989a

Hypothesis:

UP is robustly optimal for a larger class of type supports

Model

- ▶ Monopolist sells d goods, zero costs
- ▶ Additive buyer utility $u((q, t); \theta) = \sum_{j=1}^d \theta^j q^j - t$
- ▶ n types $\Theta = \{\theta_1, \theta_2, \dots, \theta_n\} \subseteq \mathbb{R}^d$ ▶ Beyond Finite Support ▶ Beyond Paths
- ▶ Buyer type $\theta \sim F \in \Delta(\Theta)$, probability mass function f
- ▶ By revelation principle, buyer can design direct mechanism $(q_1, t_1), (q_2, t_2), \dots, (q_n, t_n) \in [0, 1]^d \times \mathbb{R}_+$
- ▶ Designer wishes to maximize revenue $\sum_{\theta \in \Theta} f_i t_i$
- ▶ A mechanism is **upgrade pricing** if $\{q_1, q_2, \dots, q_n\}$ can be (totally) ordered in inclusion/component-wise order

Informal Description of of Results

Theorem (Regularity, informal)

If F is “regular” and “weakly monotone”, then UP is optimal.

Theorem (Ironing, informal)

If $\text{supp } F$ has “monotone marginal rate of substitution”, F is “weakly monotone”, and additional technical conditions on F hold, then UP is optimal.

► [More on Necessity](#)

Theorem (Separate Sales and Upgrades, informal)

If types are monotone with respect to component-wise partial order, then upgrade pricing implementability is equivalent to implementation via separate pricing.

Section 1

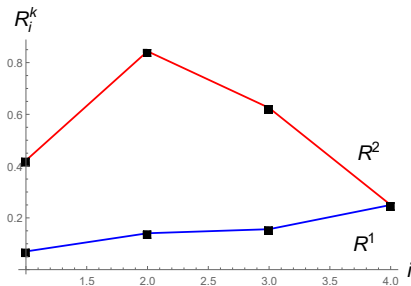
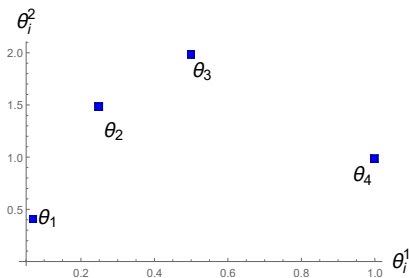
Optimality of Upgrade Pricing with Regular Distributions

Towards Optimality for Regular Distributions

Definition

$$R_i^j = (1 - F_i) \theta_i^j$$

is the **pseudo-revenue** from item j and type θ_i .



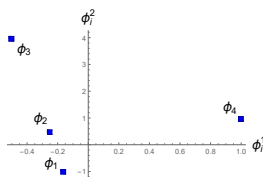
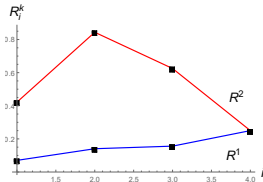
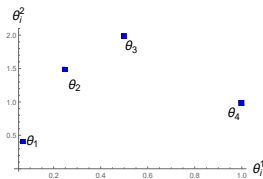
Regularity

Definition

F is **regular** if $i \mapsto R_i^j$ is single-peaked for all goods $j \in [d]$.

► Equivalent view: Slopes cross zero only once

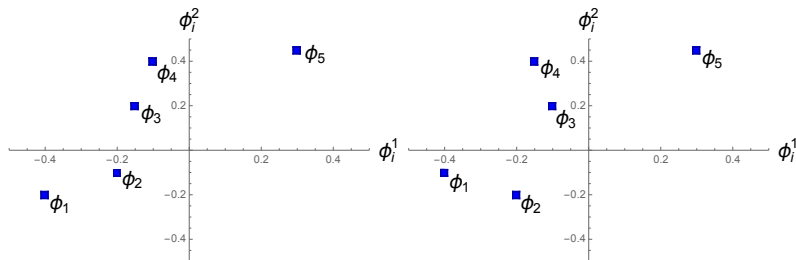
$$\begin{aligned} \frac{R_i^k - R_{i+1}^k}{f_i} &= \frac{\theta_i^k(1 - F_i) - \theta_{i+1}^k(1 - F_{i+1})}{f_i} \\ &= \theta_i^k - \frac{1 - F_{i+1}}{f_i}(\theta_{i+1}^k - \theta_i^k) =: \phi_i^k \end{aligned}$$



Weak Monotonicity

Definition

F is **weakly monotone** if $\theta_i^j \leq \theta_{i'}^j$, for any $i \leq \arg \max_i R_i^j \leq i'$, $j \in [d]$



Regularity Theorem

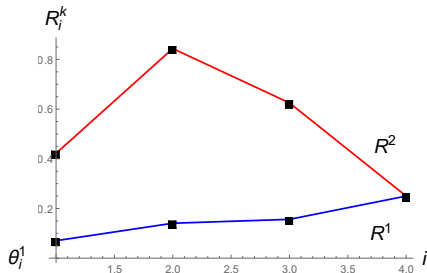
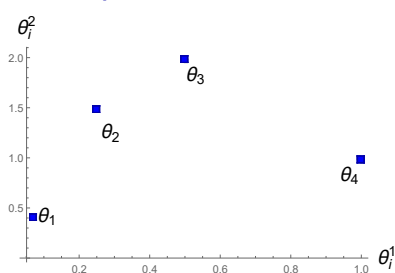
Theorem (Regularity)

If F is regular and weakly monotone, then UP is optimal. In particular, the following is an allocation of an optimal mechanism:

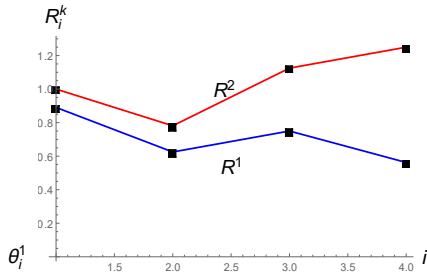
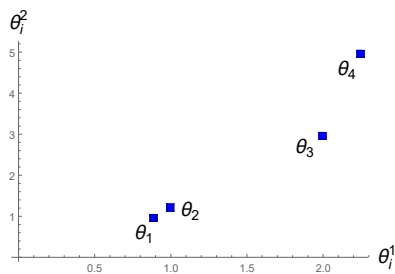
$$q_i^j = \mathbb{1}_{i \geq \arg \max_i R_i^j}. \quad (1)$$

What We Capture, What We Do Not

We Capture



We Do Not Capture



Proof Strategy

$$q_i^j = \mathbb{1}_{i \geq \arg \max_i R_i^j}. \quad (!)$$

Proof Strategy.

- ▶ Observe that (!) is upgrade pricing ✓
- ▶ Write down a dual to the monopolist's problem
- ▶ Propose a dual certificate of optimality for (!)



Duality

- ▶ Introduce dual variables λ_{ij} , $i \in [n], j \in \{0\} \cup [n]$;
- ▶ λ_{ij} corresponds to $\text{IC}(i \rightarrow j)$, λ_{i0} corresponds to $\text{IR}(i)$
- ▶ Define **virtual value** $\phi_i^\lambda = \theta_i - \sum_{j=1}^n \lambda_{ji}(\theta_j - \theta_i) \in \mathbb{R}^d$

Lemma (Duality, compare Cai, Devanur, and Weinberg 2016)

A mechanism $(q_i, t_i)_{i \in \{0\} \cup [n]}$ maximizes revenue if and only if there are multipliers $\lambda_{ji} \geq 0$, $j \in [n]$, $i \in \{0\} \cup [n]$ such that

Virtual Welfare Maximization $(q_i)_{i \in [n]}$ optimizes

$$\max_{(q_i)_{i \in [n]} \in [0,1]^n} \sum_{i=1}^n f_i \langle q_i \cdot \phi_i^\lambda \rangle$$

Feasibility of Flow $f_i = \sum_{j=0}^n \lambda_{ij} - \sum_{j=1}^n \lambda_{ji}$ for all $i \in [n]$

Compl. Slackness $\lambda_{ji}(\langle q_j, \theta_j \rangle - t_j - \langle q_i, \theta_j \rangle - t_i) = 0$ for
 $j \in [n], i \in \{0\} \cup [n]$

Implementability There are transfers t s.th. (q, t) is implementable

Virtual Values for Regular Distributions

- ▶ Virtual values depend on dual variables λ_{ij}
- ▶ $\lambda_{ji} = \mathbb{1}_{j=i+1}(1 - F_i)$ gives virtual values

$$\phi_i^\lambda := \theta_i - \frac{1 - F_{i-1}}{f_i}(\theta_{i+1} - \theta_i) = \phi_i.$$

Optimality for Regular Distributions.

- ▶ Check that $\lambda_{ji} = \mathbb{1}_{j=i+1}(1 - F_i)$ is a dual certificate

Virtual Welfare Maximization ✓

Feasibility of Flow ✓

Complementary Slackness ✓

Implementability ✓

▶ More formal

▶ Duality Lemma



Section 2

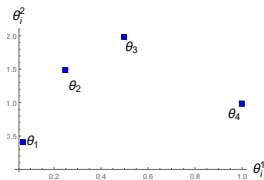
Optimality with Ironing

Definition

We say that a type space Θ has **monotone marginal rates of substitution** if for any $i, j \in [n]$, $l, k \in [d]$

$$i \leq j \text{ and } k \leq l \implies \frac{\theta_i^k}{\theta_i^l} \leq \frac{\theta_j^k}{\theta_j^l}$$

► Relation to Ratio Monotonicity



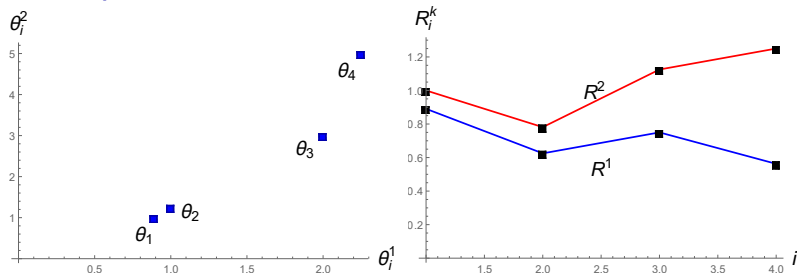
Theorem (With Ironing)

If F has monotone marginal rates of substitution and has weakly monotone types, and is **Mostly Regular**, then UP is optimal with allocation

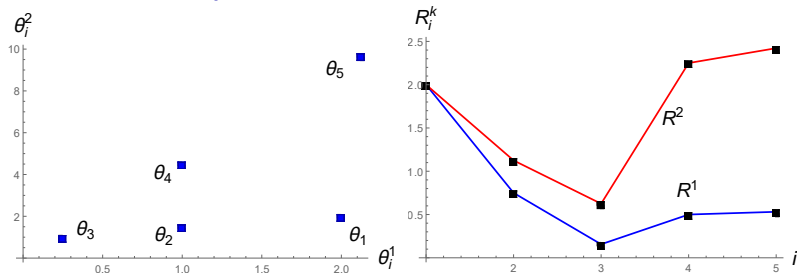
$$q^j(\theta_i) = 1_{i \geq \arg \max_i R_i^j}. \quad (2)$$

What We Capture, What We Do Not

We Capture



We Do Not Capture



Entangled Virtual Values

- ▶ Wanted: λ_{ij} ; we think of it as **ironing virtual values**
- ▶ Challenge: Virtual values for different goods are entangled

Lemma (Ordered Slopes)

If F has monotone marginal rate of substitution and λ is downward, then for $1 \leq k \leq l \leq d$,

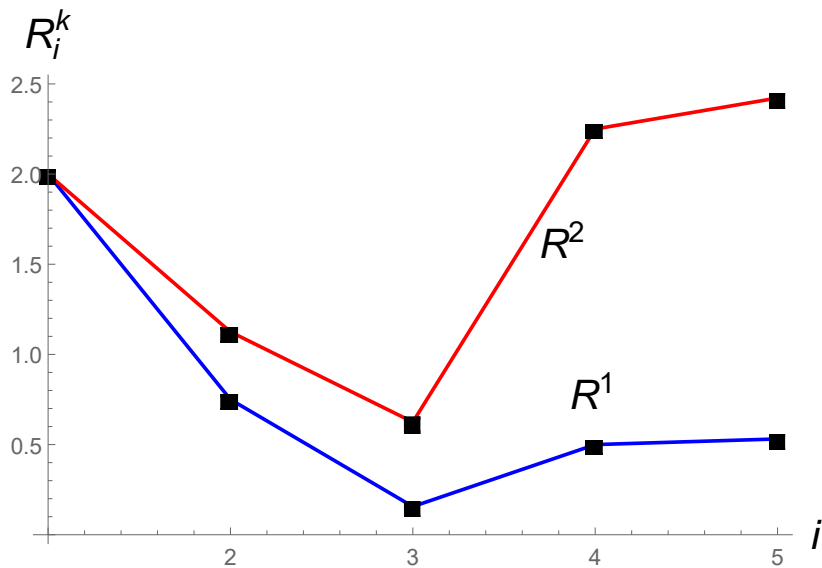
$$\frac{\phi_i^k}{\theta_i^k} \leq \frac{\phi_i^l}{\theta_i^l}.$$

Proof.

$$\begin{aligned} \frac{\phi_i^{\lambda,k}}{\theta_i^k} &= \frac{\theta_i^k - \sum_{j=1}^n \lambda_{ji}(\theta_j^k - \theta_i^k)}{\theta_i^k} = 1 + \sum_{j=i}^n \lambda_{ji} - \sum_{j=i}^n \lambda_{ji} \frac{\theta_j^k}{\theta_i^k} \\ &\leq 1 + \sum_{j=i}^n \lambda_{ji} - \sum_{j=i}^n \lambda_{ji} \frac{\theta_j^l}{\theta_i^l} = \frac{\theta_i^l - \sum_{j=1}^n \lambda_{ji}(\theta_j^l - \theta_i^l)}{\theta_i^l} = \frac{\phi_i^{\lambda,l}}{\theta_i^l}. \end{aligned}$$

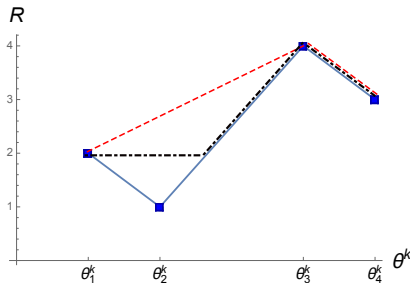


Ordered Slopes



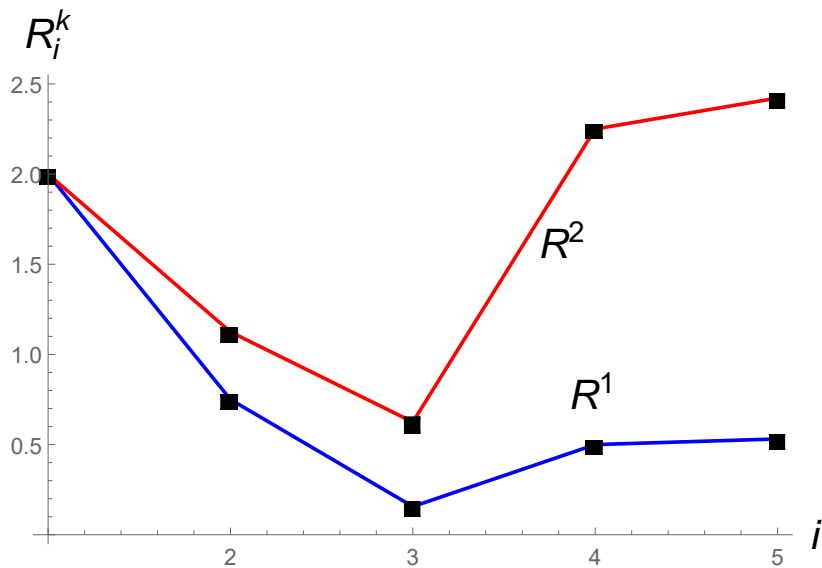
Idea of the Ironing

- ▶ We know that as long as λ is downward that the signs of virtual values are ordered
- ▶ Idea: Iron intervals to zero virtual value: dimension-wise **quasi-concave closure**

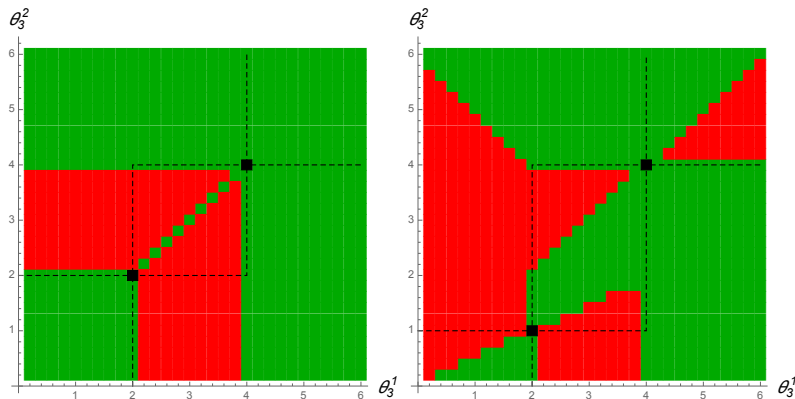


- ▶ Inside of ironing intervals no big problem
- ▶ Main care needed at the boundary of the interval.

Ordered Slopes



The Power of Upgrade Pricing Beyond Grand Bundling



► Upgrade Pricing is more powerful than Grand Bundling.

► For a Fixed Distribution

Section 3

Upgrade Pricing and Separate Pricing

Upgrade Pricing and Separate Pricing

- ▶ UP generalizes Grand Bundling when types are MRS, but not on a line through the origin
- ▶ As we show: Upgrade Pricing and Separate Pricing are equally powerful for a larger class of type supports
- ▶ Types are **monotone** if $\theta_i^j \leq \theta_{i+1}^j$ for any $i \in [n-1], j \in [d]$

Definition

A mechanism is **separate pricing** if it has a representation

$$q_i^k = \begin{cases} 1 & \theta_i^k \geq p^k \\ 0 & \text{else,} \end{cases} \quad t_i = \sum_{k=1}^d p_k 1_{q_i^k=1}.$$

for some $p^k \in \mathbb{R}_+, k \in [d]$.

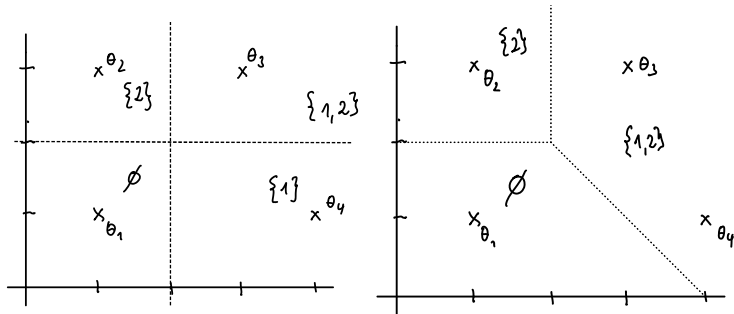
Proof

Theorem (Upgrade Pricing Redundancy)

If Θ is monotone, then the outcome of any upgrade pricing mechanism can be implemented via separate pricing, and vice versa. Otherwise, neither implication needs to hold.

Proof.

- ▶ UP \rightarrow SP: Sell products at the price of upgrades
- ▶ SP \rightarrow UP: Sell upgrades at price of products



Section 4

Epilogue

Conclusion

- ▶ Showed that Upgrade Pricing is more powerful than Grand Bundling with a robust optimality lense
- ▶ Proposed a multi-dimensional ironing targeting the dimension-wise quasi-concave closure, which allows to certify optimality of mechanisms

Future Work:

- ▶ Ironing result without technical conditions
- ▶ Formulations for
 - ▶ continuous
 - ▶ stochastically ordereddistributions
- ▶ Extension to partial bundling in bundles

More

◀ Back

- ▶ Add-On Pricing Ellison 2005
- ▶ Upgrade pricing with vertical heterogeneity Johnson and Myatt 2003
- ▶ Mixed Bundling dominates separate pricing McAfee, McMillan, and Whinston 1989b

Beyond Finite Support

◀ Back

As a generalization of Madarász and Prat 2017, we get the following meta-theorem:

Theorem

Assume that we can show optimality for a class of distributions F with finite support that is contained in a compact interval that satisfies some property P . If the set of finitely supported distributions with property P is dense (with respect to the Wasserstein metric) in the class of continuous distributions with this property, then the property holds also for continuous distributions.

Beyond Paths

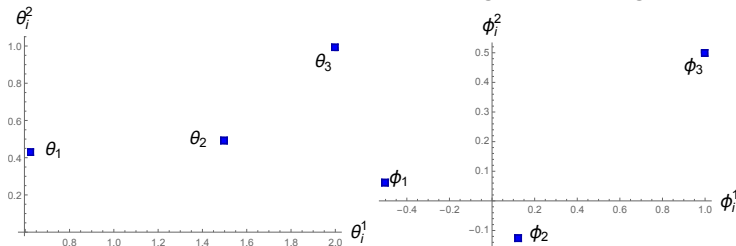
◀ Back

- ▶ Haghpanah and Hartline 2020 and unpublished work from Skrzypacz and Yang use a decomposition introduced in Strassen 1965
- ▶ It relates mixtures on paths to stochastic orders
- ▶ This allows to lift total orders on types to stochastic orders
- ▶ Our technical conditions do not work well with Strassen-type theorems

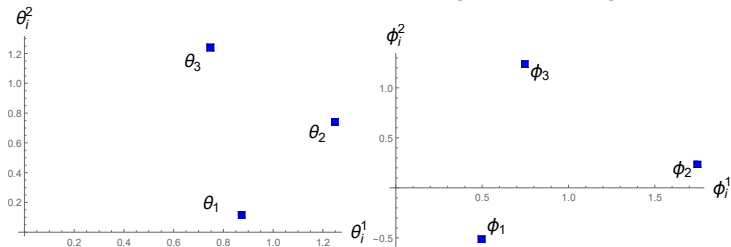
More on Necessity

◀ Back

► Monotonicity without MRS: Strict Upgrade Pricing



► MRS without Monotonicity: No Upgrade Pricing



Lagrangian and Feasibility

◀ Back

$$\begin{aligned}\mathcal{L} &= \sum_{i=1}^n f_i t_i + \sum_{j=1}^n \sum_{i=0}^n \lambda_{ji} (\langle \mathbf{q}_j, \theta_j \rangle - t_j - \langle \mathbf{q}_i, \theta_j \rangle - t_i) \\&= \sum_{i=1}^n t_i \left(f_i - \sum_{j=0}^n \lambda_{ij} + \sum_{j=1}^n \lambda_{ji} \right) + \sum_{j=1}^n \sum_{i=0}^n \lambda_{ji} \langle \mathbf{q}_j, \theta_j \rangle - \lambda_{ji} \langle \mathbf{q}_i, \theta_j \rangle \\&= \sum_{j=1}^n \sum_{i=0}^n \lambda_{ji} \langle \mathbf{q}_j, \theta_j \rangle - \sum_{j=1}^n \sum_{i=0}^n \lambda_{ji} \langle \mathbf{q}_i, \theta_j \rangle \\&= \sum_{j=1}^n \left(\left(\sum_{i=1}^n \lambda_{ij} - \sum_{i=0}^n \lambda_{ji} \right) \langle \mathbf{q}_j, \theta_j \rangle - \sum_{i=0}^n \lambda_{ji} (\langle \mathbf{q}_i, \theta_j \rangle - \langle \mathbf{q}_j, \theta_j \rangle) \right) \\&= \sum_{j=1}^n \left(f_j \langle \mathbf{q}_j, \theta_j \rangle - \sum_{i=0}^n \lambda_{ji} (\langle \mathbf{q}_i, \theta_j \rangle - \langle \mathbf{q}_j, \theta_j \rangle) \right) = \sum_{j=1}^n f_j \langle \mathbf{q}_j, \phi_j \rangle.\end{aligned}$$

Relation to Grand Bundling Optimality

◀ Back

- ▶ In Haghpanah and Hartline 2020, the same theorem is presented in a single-dimensional version, by considering $\phi_i^{\lambda'} = \langle \phi_i^\lambda, \mathbb{1} \rangle$
- ▶ Our analysis that treats dimensions separately, does not allow for this simplification

More formal

◀ Back

- ▶ Implementability is direct from weak monotonicity
- ▶ Feasibility of flow is by definition.
- ▶ Complementary slackness follows as only local downward IC constraints have non-zero dual variables
- ▶ Virtual Welfare Maximization: By single-peakedness and the fact that virtual values are derivatives of pseudo-revenues (from the right), we get that an allocation allocating items right of the maximum of the revenue curve maximizes virtual welfare.

Duality Lemma

◀ Back

Lemma (Duality, compare Cai, Devanur, and Weinberg 2016)

A mechanism $(q_i, t_i)_{i \in \{0\} \cup [n]}$ maximizes revenue if and only if there are multipliers $\lambda_{ji} \geq 0, j \in [n], i \in \{0\} \cup [n]$ such that

Virtual Welfare Maximization $(q_i)_{i \in [n]}$ optimizes

$$\max_{(q_i)_{i \in [n]} \in [0,1]^n} \sum_{i=1}^n f_i \langle q_i \cdot \phi_i^\lambda \rangle$$

Feasibility of Flow $f_i = \sum_{j=0}^n \lambda_{ij} - \sum_{j=1}^n \lambda_{ji}$ for all $i \in [n]$

Compl. Slackness $\lambda_{ji} (\langle q_j, \theta_j \rangle - t_j - \langle q_i, \theta_j \rangle - t_i) = 0$ for
 $j \in [n], i \in \{0\} \cup [n]$

Implementability There are transfers t s.th. (q, t) is implementable

Relation to Ratio Monotonicity

◀ Back

- ▶ Haghpahan and Hartline 2020 consider a concept called **ratio monotonicity**
- ▶ Ratio monotonicity, as a robust concept, boils down to
$$\frac{\sum_{k=1}^d \theta_i^k}{\theta_i^l} \leq \frac{\sum_{k=1}^d \theta_{i+1}^k}{\theta_{i+1}^l} \text{ for } i \in [n-1] \text{ and } l \in [d]$$
- ▶ An equivalent formulation of monotone MRS is
$$\frac{\sum_{k=1}^l \theta_i^k}{\theta_i^l} \leq \frac{\sum_{k=1}^l \theta_{i+1}^k}{\theta_{i+1}^l} \text{ for } i \in [n-1] \text{ and } l \in [d]$$
- ▶ (Non-trivial) calculations show that ratio-monotonicity's robust property implies that Θ is subset of a line through the origin or a -45° line

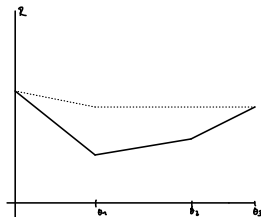
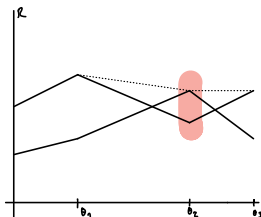
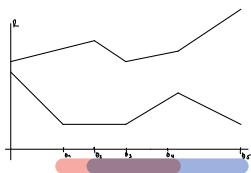
Mostly Regular

◀ Back

Denote by $\overline{R^k}_i$ the quasi-concave closure of $i \mapsto R^k_i$.

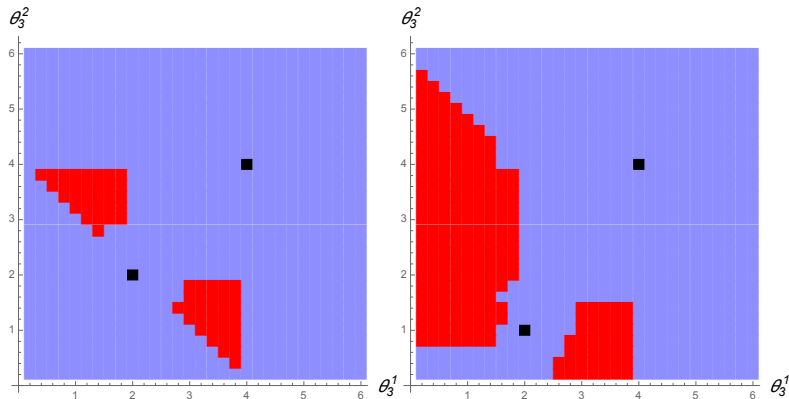
We call a type distribution F *mostly regular* if for some $i^k \in \arg \max_{i \in [n]} R^k_i$ and any i such that $i^k < i \leq i^{k+1}$

1. If $R^k_i \neq \overline{R^k}_i$, then either $R^k_{i-1} \neq \overline{R^k}_{i-1}$ or $R^{l'}_{i-1} = \overline{R^{l'}}_{i-1}$ for $l' \in \{k-1, k+1\}$ (no overlap)
2. $R^l_{i^k} = \overline{R^l}_{i^k}$ for $l \in \{k-1, k+1\}$ (no ironing on maxima)
3. If $i^k \leq i < j \leq i^{k+1} \in [n]$ and $\overline{R^k}_r \neq R^k_r$ for any $i \leq r \leq j$, then $\theta^{k+1}_i \leq \theta^{k+1}_r$ (not too shuffled)



For a Fixed Distribution

◀ Back



- ▶ For fixed distribution (uniform) even more type supports support UP (in purple)
- ▶ We study thus far only robustness, but Upgrade Pricing is even more powerful for fixed distributions