

Robust Contract Design

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Contract/Mechanism Design

- Contract/mechanism design
 - central to many economic problems
 - procurement
 - regulation
 - team design
 - ...

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- Difficulties
 - **asymmetric information**

Contract/Mechanism Design

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 - procurement
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 - ...
- Difficulties
 - **asymmetric information**
 - **uncertainty**
 - agents' higher-order information (Ben's talk)
 - other primitives (e.g., type distribution, demand, technology)

Different Approaches

- Bayesian/SEU
- Worst-case optimality
- Bayesian design with conservative payoff guarantee constraints
- Lexicographic max-min
- Competitive ratio
- Min-max regret

- **Bayesian (subjective expected utility) design**
 - procurement: Baron-Myerson (1982)...
- **Worst-case optimality**
 - ...Garrett (2014)...
- **Bayesian design with conservative payoff guarantees**
 - Mishra-Patil-Pavan (2026)
- **Lexicographic max-min**
 - Kambhampati (2025)
- **Competitive ratio**
 - Bergemann-Heumann-Morris (2025)
- **Min-max regret**
 - Guo-Shmaya (2024)
- **Also related:**
 - **Robust control, model mis-specification**
 - Bergemann-Morris (2005), Bergemann-Schlag (2008), Lopomo-Rigotti-Shannon (2021), Cerreia-Vioglio-Hansen-Maccheroni-Marinacci (2025)
 - **Un-dominated design**
 - Mishra-Patil (2025), Borgers-Li-Wang (2025)
 - **Robust information design**
 - Dworczak-Pavan (2023), Cheng-Klibanoff-Mukerji-Renou (2024)

Plan

- 1 Introduction
- 2 **Procurement**
- 3 Bayesian design
- 4 Worst-case optimality
- 5 Bayesian design with conservative payoff guarantees
- 6 Lexicographic max-min
- 7 Competitive ratio
- 8 Min-max regret
- 9 Conclusions

Procurement Environment

- **Players**

- Designer (buyer)
- Agent (supplier)

- **Contracting variables**

- output $q \in [0, \bar{q}]$ (intensity of trade)
- transfer t

Environment

- **Payoffs**

- Value of q to designer:

$$V(q) = \int_0^q P(s)ds$$

P is inverse demand, $D = P^{-1}$ is direct demand

- Seller's cost θq
- Players' outside option payoff: 0
- Designer payoff:

$$V(q) - t$$

- Seller's payoff:

$$t - \theta q$$

- **Asymmetric information**

- θ : seller's private information

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Bayesian Design

Bayesian approach

- **Designer has model** (V^*, F^*)

- Value function: $V^*(q) = \int_0^q P^*(s) ds$
- Cost technology: θ drawn from abs. cont F^* with $f^*(\theta) > 0$ over $\Theta = [\underline{\theta}, \bar{\theta}]$ and “**virtual cost**”

$$z^*(\theta) := \theta + \frac{F^*(\theta)}{f^*(\theta)}$$

cont. and increasing (**regularity**)

- **(Direct) procurement mechanism** $M = (q, t)$

- quantity schedule $q : \Theta \rightarrow \mathbb{R}_+$
- transfer schedule $t : \Theta \rightarrow \mathbb{R}$
- Rent to type θ :

$$u(\theta) := t(\theta) - \theta q(\theta)$$

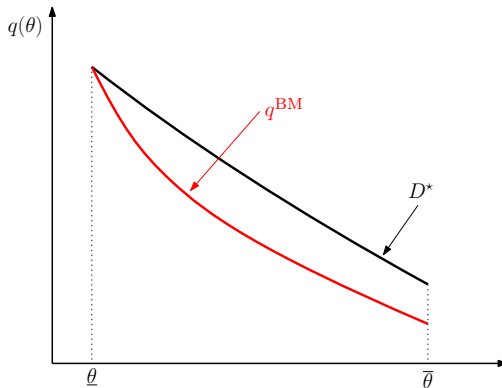
- $\mathcal{M}$: set of IC and IR mechanisms
- Given $M = (q, u) \in \mathcal{M}$, **expected welfare** under (V^*, F^*)

$$W(M; V^*, F^*) := \int [V^*(q(\theta)) - \theta q(\theta) - u(\theta)] F^*(d\theta)$$

SEU/Bayesian Benchmark: Baron Myerson

Proposition

SEU/Bayesian optimum: $q^{BM}(\theta) := D^*(z^*(\theta))$, where $z^*(\theta) := \theta + \frac{F^*(\theta)}{f^*(\theta)}$ is “virtual cost”.



(BM-Proposition)

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Worst Case Optimality

Worst-Case Optimality

- Designer has **admissible set (of models)**: $\mathcal{A} = \mathcal{V} \times \mathcal{F}$

- $\mathcal{V}$: set of possible (gross) value fns
- each $V \in \mathcal{V}$ abs continuous, increasing, concave
 - $\mathcal{P}$: set of corresponding *inverse demand functions*, with

$$V(q) = \int_0^q P(s) ds \quad \forall q$$

- $\mathcal{D}$: set of corresponding *direct demand functions*, $D = P^{-1}$
- Lowest (inverse) demand function: $\underline{P}$
any $q \geq 0$ and $P \in \mathcal{P}$

$$P(q) \geq \underline{P}(q)$$

$\underline{P}$: strictly decreasing, continuous, and s.t.

$$\lim_{q \rightarrow 0^+} \underline{P}(q) > \bar{\theta}$$

- $\mathcal{F} = \text{CDF}(\Theta)$: all cdfs with support contained in $\Theta = [\underline{\theta}, \bar{\theta}]$
(more general $\mathcal{F}$, w. lowest element $\underline{F}$ can be considered)

Environment

- $\mathcal{M}$: set of IC and IR mechanisms
- Designer's welfare under $M \in \mathcal{M}$, given (V, F) :

$$W(M; V, F) := \int [V(q(\theta)) - \theta q(\theta) - u(\theta)] F(d\theta)$$

Definition

Welfare guarantee for $M \in \mathcal{M}$,

$$G(M) := \inf_{(V, F) \in \mathcal{A}} W(M; V, F).$$

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Definition

Worst case optimality

$$\max_{M \in \mathcal{M}} G(M)$$

Worst-case Optimality

- Let

$$q_\ell := \arg \max_q \{ \underline{V}(q) - \bar{\theta}q \} = \underline{D}(\bar{\theta})$$

be efficient output when $V = \underline{V}$ and $\theta = \bar{\theta}$ and

$$G^* := \underline{V}(q_\ell) - \bar{\theta}q_\ell$$

Lemma

For any IC and IR mechanism $M = (q, u) \in \mathcal{M}$

$$G(M) = \inf_{\theta \in \Theta} \left\{ \underline{V}(q(\theta)) - \theta q(\theta) - \int_{\theta}^{\bar{\theta}} q(y) dy - u(\bar{\theta}) \right\}.$$

Furthermore,

$$G(M) \leq G^*$$

Worst-case Optimality

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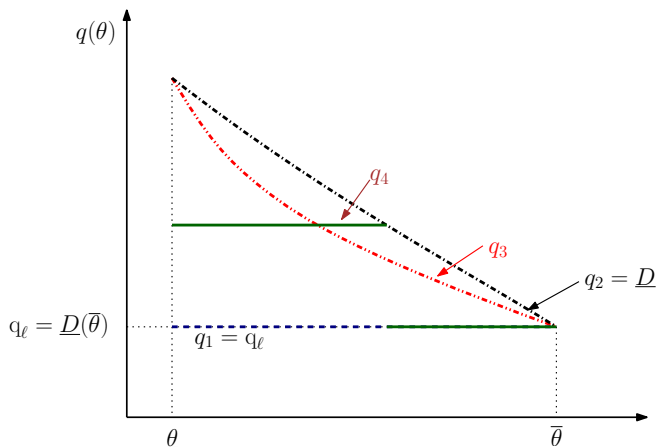
$$G(M) \leq G^*$$

- Worst case always under lowest demand **but not necessarily highest cost**

(Max-guarantee-proof)

Worst-case Optimality

- “Flat” mechanism $M_1 = (q_1, t_1)$ with $q_1(\theta) = q_\ell$ all θ , delivers maximal guarantee G^*
- ...so does $M_2 = (q_2, t_2)$ with $q_2(\theta) = \underline{D}(\theta)$ all θ
- ...but also mechanisms $M_3 = (q_3, t_3)$ and $M_4 = (q_4, t_4)$...
- ...continuum of worst-case optimal mechanisms



Worst-case Optimality: Full Characterization

Lemma

$M = (q, u)$ is worst-case optimal iff

- 1 q non-increasing,
- 2 $u(\theta) = \int_{\theta}^{\bar{\theta}} q(y) dy$ all θ
- 3 for all θ ,

$$\underline{W}(\theta, q) := \underline{V}(q(\theta)) - \theta q(\theta) - \int_{\theta}^{\bar{\theta}} q(y) dy \geq G^*$$

WCO: Majorizations

Lemma

Take any non-increasing q s.t. $q(\bar{\theta}) = q_\ell$. Following are equivalent

$$\underline{W}(\theta, q) := \underline{V}(q(\theta)) - \theta q(\theta) - \int_{\theta}^{\bar{\theta}} q(y) dy \geq G^* \quad \forall \theta \in \Theta$$

$$\int_{\theta}^{\bar{\theta}} q(y) dy \leq \int_{\theta}^{\bar{\theta}} \underline{D}(y) dy - \underbrace{\int_{\theta}^{\underline{P}(q(\theta))} (\underline{D}(y) - q(\theta)) dy}_{DWL(\theta, q(\theta)) \geq 0} \quad \forall \theta \in \Theta$$

$$\int_{\theta}^{\bar{\theta}} q(y) dy \leq \int_{\theta}^{\bar{\theta}} \underline{D}(y) dy \quad \forall \theta \in \Theta \quad \text{and} \quad \underline{W}(\theta, q) \geq G^*$$

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Bayesian Design under Conservative Payoff Guarantees

Bayesian Design with Conservative Payoff Guarantees

- In many settings designer **does have** model (V^*, F^*)
- However, designer needs approval from
 - regulatory authorities
 - upstream supervisor
 - ...

Bayesian Design with Conservative Payoff Guarantees

- In many settings designer **does have** model (V^*, F^*)
- However, designer needs approval from
 - regulatory authorities
 - upstream supervisor
 - ...
- **Approver**
 - same objective as designer
 - but **does not trust designer's confidence in model**
 - limited expertise
 - legal requirements/impediments
 - value-at-risk obligations
 - ...
 - asks for satisfactory **welfare guarantees**

Alternative Interpretation (of robustness constraint)

- Designer does not fully trust her own model
 - first prepares for worst (in case model is mis-specified)
 - then uses model to select among worst-case-optimal mechanisms
- Two interpretations equivalent

Bayesian Design with Conservative Payoff Guarantees

- Supervisor approves only mechanisms with highest welfare guarantee

Definition

Short-list (of approved mechanisms)

$$\mathcal{M}^{\text{SL}} \equiv \arg \max_{M \in \mathcal{M}} G(M)$$

Bayesian Design with Conservative Payoff Guarantees

- Supervisor approves only mechanisms with highest welfare guarantee

Definition

Short-list (of approved mechanisms)

$$\mathcal{M}^{\text{SL}} \equiv \arg \max_{M \in \mathcal{M}} G(M)$$

- Mechanisms with largest guarantee
- More permissive lists: $\mathcal{M}_{\gamma}^{\text{SL}} \equiv \{M \in \mathcal{M} : G(M) \geq \gamma G^*\}$, $\gamma \in [0, 1]$

Designer's Problem

- Choose mechanism that maximizes expected welfare **under conjectured model** (V^*, F^*) over SL of worst-case-optimal mechanisms

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Definition

Mechanism M **robustly optimal** if

$$M \in \arg \max_{M' \in \mathcal{M}^{\text{SL}}} W(M'; V^*, F^*)$$

Robust Optimality

- Robustly optimal schedule q^{OPT} solves

$$\max_q \int_{\underline{\theta}}^{\bar{\theta}} \left[V^*(q(\theta)) - z^*(\theta)q(\theta) \right] F^*(d\theta)$$

s.t.

q non-increasing

$$\underline{W}(\theta, q) := \underline{V}(q(\theta)) - \theta q(\theta) - \int_{\theta}^{\bar{\theta}} q(y) dy \geq G^* \quad \forall \theta \in \Theta$$

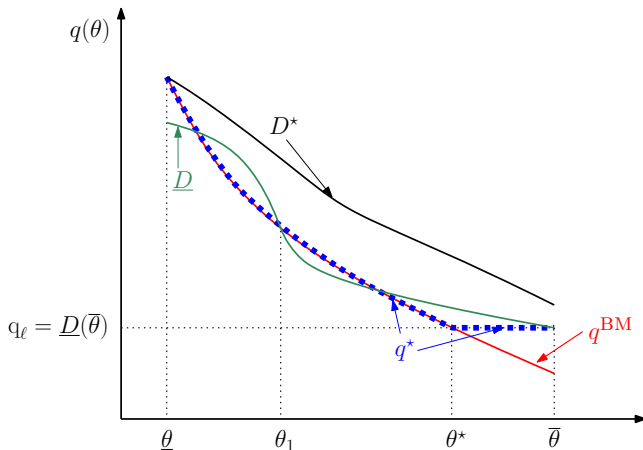
where $z^*(\theta) := \theta + \frac{F^*(\theta)}{f^*(\theta)}$ is **virtual cost** under conjectured model (V^*, F^*) .

Baron-Myerson-with-Quantity-Floor

Definition

BM-with-quantity-floor:

$$q^*(\theta) := \max\{q^{\text{BM}}(\theta), q_\ell\}$$



Robust Optimality of BM-with-Quantity-Floor

Proposition

BM-with-quantity-floor $M^ = (q^*, u^*)$ robustly optimal iff $\underline{W}(\underline{\theta}, q^*) \geq G^*$ and*

$$\int_{\underline{\theta}}^{\bar{\theta}} q^*(y) dy \leq \int_{\underline{\theta}}^{\bar{\theta}} \underline{D}(y) dy \quad \forall \theta \in \Theta$$

Furthermore, when $M^ = (q^*, u^*)$ robustly optimal, any $M^{\text{OPT}} = (q^{\text{OPT}}, u^{\text{OPT}})$ s.t.*

$$q^{\text{OPT}}(\theta) = q^*(\theta) \quad \forall \theta > \underline{\theta}$$

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Corollary

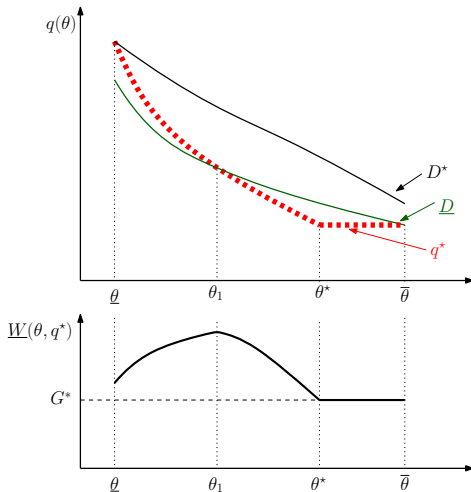
When uncertainty only over type distribution (more generally, $D^ = \underline{D}$)*

- *BM-with-quantity-floor robustly optimal*
- *efficiency at both bottom and top*

(BM-floor: Proof)

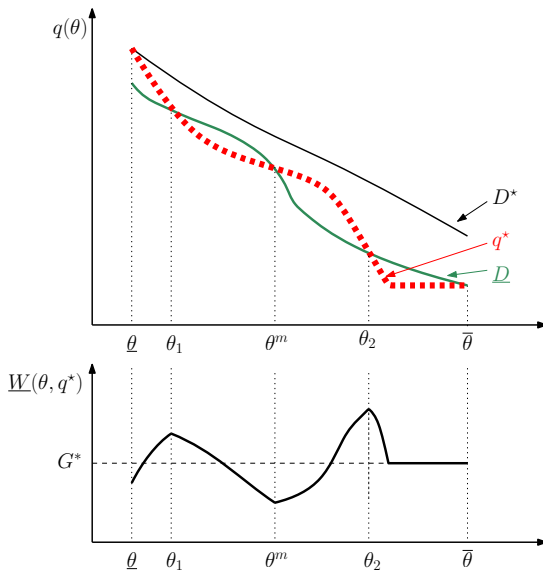
Robust Optimality of BM-with-Quantity-Floor

Suppose $D^* \neq \underline{D}$, q^{BM} single-crosses $\underline{D}$ from above, and $\underline{W}(\theta, q^*) \geq G^*$. Then BM-with-quantity-floor robustly optimal

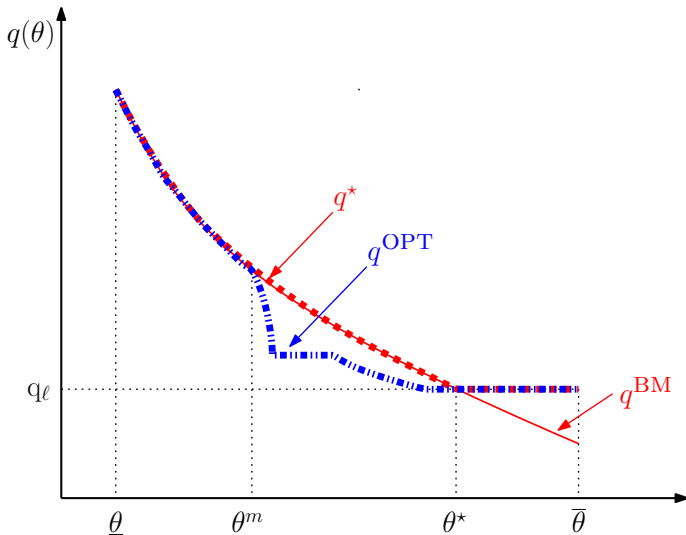


(Monotonicity Lemma)

Non-robustness of BM-with-quantity-floor



General Case



Robust Optimality: General Case

- **Reduced procurement for intermediate costs**

- avoids over-procurement and excessive rents when $D < D^*$

- **Increased procurement for high costs**

- avoids inefficiencies motivated by rent extraction (for lower types)

- **Low costs**

- welfare higher than G^* even when $D < D^*$
- no adjustments necessary

(Proposition: General Case)

(More General Cost Uncertainty)

Downstream Markets and Price Regulation

- Suppose supplier sells to **downstream mkt**
- Idea: use **price regulation** for **demand discovery**
- (Direct) price mechanism $\tilde{M} = (p, t)$
 - **price schedule** $p : \Theta \rightarrow \mathbb{R}_+$
 - **transfer schedule** $t : \Theta \times \mathcal{D} \rightarrow \mathbb{R}$
 - conditioning t on realized demand D
 - irrelevant under qty regulation
 - **protects seller from uncertainty under price regulation**

(EPIC and EPIR)

Price Regulation: Welfare Guarantee

Lemma

For any price mechanism $\tilde{M} \in \tilde{\mathcal{M}}$,

$$G(\tilde{M}) \leq G^*.$$

There exists price mechanism $\underline{\tilde{M}} \in \tilde{\mathcal{M}}$ s.t.

$$G(\underline{\tilde{M}}) = G^*.$$

Any $\tilde{M} \in \tilde{\mathcal{M}}^{\text{SL}}$ (i.e., for which $G(\tilde{M}) = G^*$) s.t.

$$p(\bar{\theta}) = \bar{\theta}$$

and $\tilde{u}(\bar{\theta}, \underline{D}) = 0$.

- Maximal guarantee same as with quantity mechanisms
- Price cap at $\bar{\theta}$

Robust Optimality: Baron-Myerson-with-price-cap

- Designer's price regulation problem

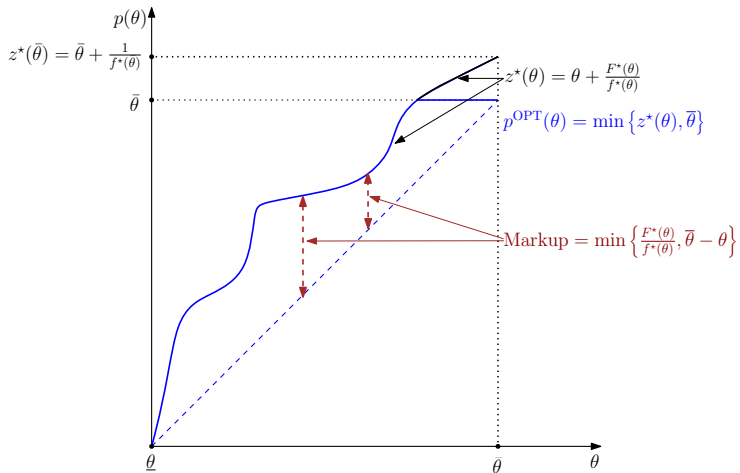
$$\max_{\tilde{M} \in \tilde{M}^{SL}} W(\tilde{M}; D^*, F^*)$$

Proposition

In any robustly optimal price mechanism $\tilde{M}^{OPT} = (p^{OPT}, \tilde{u}^{OPT})$

$$p^{OPT}(\theta) = \min\{z^*(\theta), \bar{\theta}\} \quad \forall \theta \in \Theta$$

Robust Optimality: Baron-Myerson-with-price-cap



(Proof Sketch)

Quantity vs Price Regulation?

- When is qty regulation superior to price regulation?
 - Weitzman 78: depends
 - Bayesian design: price wins
 - Worst-case optimality: tie
 - Robust optimality?
- M^{OPT} : robustly optimal **quantity** mechanism
- $\tilde{M}^{\text{OPT}}$: robustly optimal **price** mechanism

Definition

Price regulation dominates if

$$\widetilde{W}(\tilde{M}^{\text{OPT}}; D^*, F^*) \geq W(M^{\text{OPT}}; V^*, F^*)$$

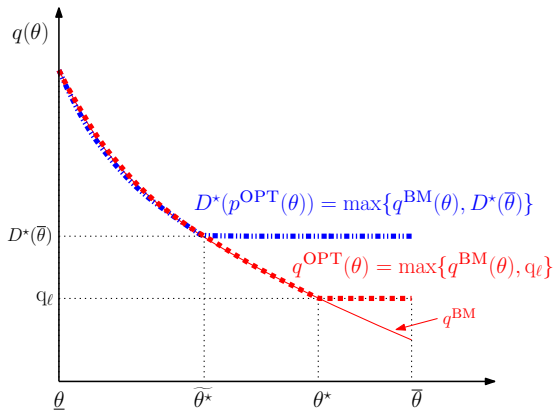
Quantity regulation dominates if

$$\widetilde{W}(\tilde{M}^{\text{OPT}}; D^*, F^*) \leq W(M^{\text{OPT}}; V^*, F^*)$$

Quantity Regulation Dominates

Proposition

If BM-with-quantity-floor is robustly optimal, then quantity regulation dominates (strictly if $D^(\bar{\theta}) > \underline{D}(\bar{\theta})$).*

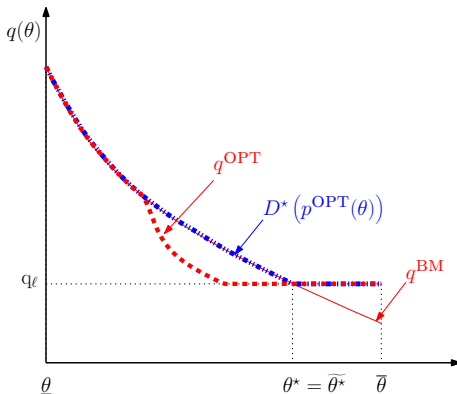


(Reasoning)

Price Regulation Dominates

Proposition

If BM-with-quantity-floor **not** robustly optimal and $D^*(\bar{\theta}) = \underline{D}(\bar{\theta})$, price regulation strictly dominates.



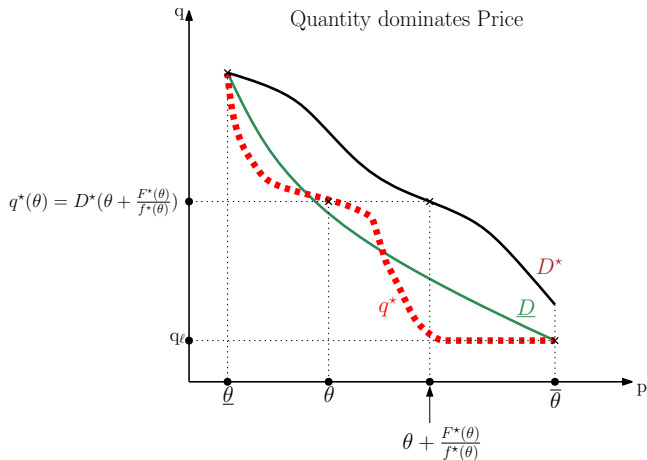
(Reasoning)

Quantity vs Price Regulation: Primitives

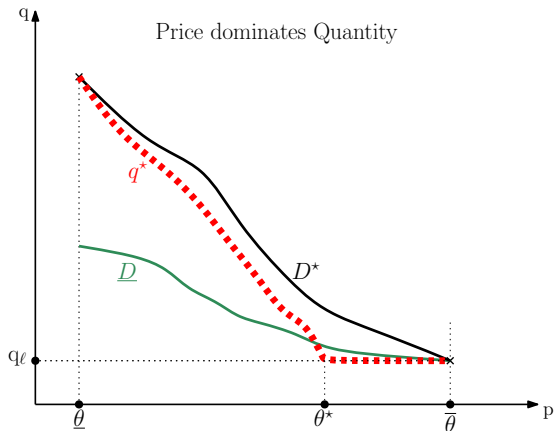
- Qty regulation dominates when
 - markups $F^*(p)/f^*(p)$ large relative to $D^* - \underline{D}$
- Price regulation dominates when
 - markups $F^*(p)/f^*(p)$ small relative to $D^* - \underline{D}$

(Proposition: Qty vs Price Regulation)

Quantity Regulation Dominates



Price Regulation Dominates



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Lexicographic Robustness

Lexicographic Robustness: Kambhampati (2026)

- Max-min optimality: many (continuum of) mechanisms
- (Endogenous) lexicographic beliefs refine beyond worst case
- Properly selected belief systems **eliminate rent-extraction vs efficiency tradeoff**

Lexicographic Probability Systems

- Suppose V known to designer
- Expected payoff of M under F : $w(M; F) := W(M; V, F)$
- Uncertainty: $\mathcal{F} \subset \text{CDF}(\Theta)$
 - each $F \in \mathcal{F}$: finite support $\Theta^\# := (\theta_1, \theta_2, \dots, \theta_N)$, with $f(\theta) := \Pr(\theta)$
- **Lexicographic probability system (LPS)**: ordered list of beliefs

$$F = (F_1, \dots, F_K)$$

- **Lexicographic order**: Let $a, b \in \mathbb{R}^K$. Say $a \geq_L b$ if

$$b_k > a_k \text{ for some } k \in \{1, \dots, K\} \implies a_j > b_j \text{ for some } j < k$$

- Type θ **F-infinitely more likely** than θ' ($\theta >_F \theta'$) if

$$\min\{k : f_k(\theta) > 0\} < \min\{k : f_k(\theta') > 0\}$$

- $\theta \geq_F \theta'$ if not true that $\theta' >_F \theta$
- Mechanisms compared by expected payoff vector

$$w_F(M) \equiv (w(M; F_1), \dots, w(M; F_K))$$

- $M \in \mathcal{M}$ **F-optimal** iff $w_F(M) \geq_L w_F(M')$ all $M' \in \mathcal{M}$

- **Definition:** F **adversarial** wrt $M = (q, t)$ if

$$f_1(\theta) > 0 \Rightarrow V(q(\theta)) - t(\theta) \leq V(q(\theta')) - t(\theta'), \forall \theta' \in \Theta^\#$$

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Definitions

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- **Definition:** F has **full support** if $\forall \theta \in \Theta^\#, \exists k \in \{1, \dots, K\}$ s.t. $f_k(\theta) > 0$.

Definitions

- **Definition:** M **robust** if $\exists F$ adversarial wrt M s.t. M is F -optimal.

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Definitions

- **Definition:** M **robust** if $\exists F$ adversarial wrt M s.t. M is F -optimal.
- **Definition:** M is **perfectly robust** if $\exists F$ adversarial wrt M with full support s.t. M is F -optimal.
- **Definition:** M is **properly robust** if $\exists F$ strongly adversarial wrt M with full support s.t. M is F -optimal.

Definitions

- **Definition:** $M = (q^{FB}, t^{FB})$ is **FB-efficient** (with minimal transfers) if,

① $q^{FB}(\theta) = \arg \max_{q \in [0, \bar{q}]} \{V(q) - \theta q\}$ all $\theta \in \Theta^\#$

- ② No rent at the bottom:

$$t^{FB}(\theta_N) = \theta_N q^{FB}(\theta_N)$$

- ③ Upward-adjacent IC constraints bind, i.e., all $j = 1, \dots, N - 1$

$$t^{FB}(\theta_j) - \theta_j q^{FB}(\theta_j) = t^{FB}(\theta_{j+1}) - \theta_j q^{FB}(\theta_{j+1})$$

Theorem

M robust iff

- ① *FB-efficiency at “bottom,” i.e., with highest-cost type: $q(\theta_N) = q^{FB}(\theta_N)$*
- ② *$t(\theta_N) = t^{FB}(\theta_N)$*
- ③ *lowest welfare attained at “bottom”:*

$$V(q(\theta)) - t(\theta) \geq V(q(\theta_N)) - t(\theta_N) \quad \forall \theta \in \Theta^\#$$

Theorem

M robust iff

- ① *FB-efficiency at “bottom,” i.e., with highest-cost type: $q(\theta_N) = q^{FB}(\theta_N)$*
- ② *$t(\theta_N) = t^{FB}(\theta_N)$*
- ③ *lowest welfare attained at “bottom”:*

$$V(q(\theta)) - t(\theta) \geq V(q(\theta_N)) - t(\theta_N) \quad \forall \theta \in \Theta^\#$$

- Hence, robustness implies WCO (saddle point)

(Robustness and WCO: Proof)

Perfect Robustness and Undomination

Definition: $M = (q, t)$ **undominated** if there is no IC-IR $M' = (q', t')$ s.t.

$$V(q'(\theta)) - t'(\theta) \geq V(q(\theta)) - t(\theta) \quad \forall \theta \in \Theta^\#$$

with inequality strict for some $\theta \in \Theta^\#$.

Theorem

M **perfectly robust** iff (a) *robust* and (b) *undominated*.

(Perfect Robustness and Undomination: Proof)

Proper Robustness and Efficiency

Theorem

M **properly robust** *iff* *FB-efficient*.

- Hence, in any properly robust mechanism:
 - **efficient procurement from all types**
 - **minimal transfer compatible with IC-IR of all types**

(Proper Robustness and Efficiency: Proof)

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Competitive Ratio

Competitive Ratio: Bergemann-Heumann-Morris (2025)

- Buyer knows V (equivalently, D) but has no prior over seller's cost.
- Objective: guarantee maximal fraction of efficient social surplus.
- Main result: **constant utility-share mechanism optimal.**

Environment

- Same as above, but with constant-elasticity gross value fn

$$V^*(q) = \frac{q^\sigma}{\sigma}, \quad \sigma \in (0, 1)$$

- Seller privately knows increasing, convex cost fn C (e.g., $C(q) = \theta q$)
- Designer commits to DRM (t, q) – equivalently, tariff $T : [0, \bar{q}] \rightarrow \mathbb{R}$
- Given (C, T) , seller chooses $q \in [0, \bar{q}]$ to max $T(q) - C(q)$
- Let

$$Q(C, T) \in \arg \max_{q \in [0, \bar{q}]} \{T(q) - C(q)\}.$$

- Designer's ex-post payoff: $U(C, T) := V(Q(C, T)) - T(Q(C, T))$.

Competitive Ratio

- Complete-information efficient surplus:

$$W(C) := \max_{q \in [0, \bar{q}]} \{V(q) - C(q)\}$$

- Relative performance of T at cost C :

$$R(C, T) := \frac{U(C, T)}{W(C)}.$$

Definition

Competitive ratio is

$$\sup_{T \in \mathcal{T}} \inf_{C \in \mathcal{C}} R(C, T).$$

Constant-Share Mechanism

Definition

Constant-share mechanism pays

$$T_z(q) = z V(q), \quad z \in (0, 1).$$

- Buyer keeps $(1 - z)V(q)$.

Linear Costs: Exact Competitive Ratio

- Suppose admissible set $\mathcal{C}$ s.t., for any $C \in \mathcal{C}$, there exists $\theta \in \Theta$ s.t. for all $q \in [0, \bar{q}]$,

$$C(q) = \theta q$$

- Then,

$$Q(C, T_z) = \left(\frac{z}{\theta}\right)^{1/(1-\sigma)}$$

and

$$R(C, T_z) = \frac{U(C, T_z)}{W(C)} = \frac{1-z}{1-\sigma} \frac{1}{z^{\sigma/(1-\sigma)}},$$

which is independent of θ .

Theorem 1

Optimal constant-share mechanism is $z^ = \sigma$, and*

$$\sup_T \inf_{C \in \mathcal{C}} \frac{U(C, T)}{W(C)} = \sigma^{\frac{\sigma}{1-\sigma}}.$$

Linear Costs: Exact Competitive Ratio

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Theorem 1

Optimal constant-share mechanism is $z^ = \sigma$, and*

$$\sup_T \inf_{C \in \mathcal{C}} \frac{U(C, T)}{W(C)} = \sigma^{\frac{\sigma}{1-\sigma}}.$$

- No mechanism, linear or not, improves over optimal constant-share one
- Guarantee falls from 1 to $1/e$ as demand becomes more elastic
- Optimal constant-share mech procures $\sigma^{1/(1-\sigma)}$ of efficient output $Q^{FB}(C) = \theta^{\frac{1}{\sigma-1}}$ in each state: **distortions for all types θ**

Why Constant-Share Optimal?

- Non-constant transfer rule favors some output scales over others.
- Nature selects linear cost whose optimal output $Q(C, T)$ yields low scale.
- Constant sharing equalizes relative performance across scales.
- Constant-share mechanism Bayes-optimal for cdf $F(\theta) = (\theta/\bar{\theta})^\alpha$ on $[0, \bar{\theta}]$, with $\alpha \rightarrow \sigma/(1 - \sigma)$

General Convex Costs

- With arbitrary increasing convex costs C , constant-share mechanism T_z with $z = \sigma$ no longer exactly optimal.
- Let $z^*(\sigma)$ maximize worst-case ratio, i.e.,

$$z^*(\sigma) \in \arg \max_{z \in [0,1]} \inf_{C \in \mathcal{C}} \frac{U(C, T_z)}{W(C)}$$

- Let

$$B(\sigma) := \frac{1 - z^*(\sigma)}{(1 - \sigma)(z^*(\sigma))^{\sigma/(\sigma-1)} + \sigma z^*(\sigma)} = \inf_{C \in \mathcal{C}} \frac{U(C, T_{z^*(\sigma)})}{W(C)}.$$

Theorem

For every mechanism T ,

$$\inf_{C \in \mathcal{C}} \frac{U(C, T)}{W(C)} \leq B(\sigma),$$

with equality iff $T = T_{z^*(\sigma)}$.

Elasticity Governs Guarantee

- Optimal share $z^*(\sigma)$ increases with σ : elastic demand requires stronger supply incentives.
- Numerically, $z^*(\sigma)$ stays close to σ :

$$|z^*(\sigma) - \sigma| \leq 0.12.$$

- Guarantee declines with σ and is close to linear:

$$|B(\sigma) - (1 - \sigma)| \leq 0.16.$$

- Unlike linear-cost case, $\lim_{\sigma \rightarrow 1} B(\sigma) = 0$: flexible value functions permit adversarial Nature to grasp most surplus

Plan

- 1 Introduction
- 2 Procurement
- 3 Bayesian design
- 4 Worst-case optimality
- 5 Bayesian design with conservative payoff guarantees
- 6 Lexicographic max-min
- 7 Competitive ratio
- 8 **Min-max regret**
- 9 Conclusions

Min-max Regret

Min-Max Regret: Guo-Shmaya (2025)

- Monopolist sells quantity $q \in [0, \bar{q}]$, with $\bar{q} = 1$, in downstream market w. inverse demand P s.t. $P(0) = \bar{v}$.
- Monopolist observes
 - inverse demand P (decreasing and upper semicontinuous)
 - cost fn C (increasing, lower semicontinuous, with $C(0) = 0$)
- Designer (regulator) faces uncertainty over (P, C)
- Designer's policy $\rho(q, p)$ specifies monopolist's net revenue when transacting at (q, p)
- Taxation Principle: any outcome of any IC-IR DRM $M = (q(P, C), t(P, C))$ implementable with tariff ρ
- Revenue ρ funded with consumer's money
- Given (C, P) and ρ , monopolist chooses (q, p) , with $p \leq P(q)$ to maximize

$$\rho(q, p) - C(q).$$

Objective and Regret

- Regulator's payoff

$$RS_{\alpha} = CS + \alpha \Pi, \quad \alpha \in [0, 1].$$

- $\alpha = 0$: consumers' surplus
- $\alpha = 1$: total surplus

Objective and Regret

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- $\alpha = 1$: total surplus

Definition 2

Given (P, C) , ex-post regret under ρ

$$\text{Regret} = CIP(P, C) - RS_{\alpha}(\rho, P, C; q^*(\rho, P, C), p^*(\rho, P, C)),$$

where $CIP(P, C)$ is designer's complete-information payoff and (q^*, p^*) is firm's best response to ρ given (P, C) .

Objective and Regret

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where $CIP(P, C)$ is designer's complete-information payoff and (q^*, p^*) is firm's best response to ρ given (P, C) .

Optimal policy ρ^* minimizes ex-post regret over all (P, C) in admissible set $\mathcal{A} = \mathcal{P} \times \mathcal{C}$ and all firm's best responses (q^*, p^*)

Three Sources of Regret

- ① **Rents to firm:** high revenue transfers surplus from consumers to firm.
- ② **Underproduction:** weak marginal incentives leave valuable consumers unserved.
- ③ **Overproduction:** subsidies may induce output whose cost exceeds consumer value.

Three Sources of Regret

- 1 **Rents to firm:** high revenue transfers surplus from consumers to firm.
- 2 **Underproduction:** weak marginal incentives leave valuable consumers unserved.
- 3 **Overproduction:** subsidies may induce output whose cost exceeds consumer value.

Designer problem

No single instrument controls all three margins. Optimal policy combines

- average-revenue cap
- piece-rate subsidy
- cap on total subsidy.

Worst-Case Regret

- Let $f_\alpha \equiv 1/(2 - \alpha)$ and

$$r_\alpha = \max_{q \in [0,1], p \in [0, f_\alpha \bar{v}]} \min \{q(1 - f_\alpha)\bar{v} - qp \log q; q(f_\alpha \bar{v} - p)\}.$$

Theorem

Every policy has worst-case regret greater or equal to r_α , and bound is attainable.

Worst-Case Regret

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Theorem

Every policy has worst-case regret greater or equal to r_α , and bound is attainable.

As α rises, redistribution matters less and regulator tolerates higher firm revenue to improve output incentives.

Polar Case: Consumer Surplus

- When $\alpha = 0$ (i.e., designer maximizes CS), $f_0 = 1/2$ and optimal policy is

$$\rho(q, p) = \min \left\{ \frac{q\bar{v}}{2}, qp \right\}.$$

- Equivalently, **price cap** at $\bar{v}/2$.
- Cap limits firm's extraction of consumers' surplus.
- No subsidy needed because firm's profit receives zero welfare weight.

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- Cap limits firm's extraction of consumers' surplus.
- No subsidy needed because firm's profit receives zero welfare weight.

Implication: **min-max-regret foundation for widely used regulatory instrument.**

Polar Case: Total Surplus

- When $\alpha = 1$ (i.e., designer maximizes TS), $f_1 = 1$ and optimal policy is

$$\rho(q, p) = \min\{q\bar{v}, qp + r_1\}.$$

- Equivalently, **piece-rate subsidy** $(\bar{v} - p)$ with **total subsidy cap** at r_1 .

Polar Case: Total Surplus

- When $\alpha = 1$ (i.e., designer maximizes TS), $f_1 = 1$ and optimal policy is

$$\rho(q, p) = \min\{q\bar{v}, qp + r_1\}.$$

- Equivalently, **piece-rate subsidy** $(\bar{v} - p)$ with **total subsidy cap** at r_1 .

Trade-off

Subsidy mitigates underproduction at low quantities; subsidy cap prevents large losses from overproduction.

Optimal Policy for Any Welfare Weight

- Let

$$s_\alpha := (\sup q(f_\alpha \bar{v} - p) : q(1 - f_\alpha)\bar{v} - qp \log q > r_\alpha, (q, p) \in [0, 1] \times [0, f_\alpha \bar{v}])^+$$

be smallest subsidy cap consistent with regret r_α .

Theorem

For every $s \in [s_\alpha, r_\alpha]$,

$$\rho(q, p) = \min\{qf_\alpha \bar{v}, qp + s\}$$

achieves worst-case regret r_α .

Optimal Policy for Any Welfare Weight

- Let

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Theorem

For every $s \in [s_\alpha, r_\alpha]$,

$$\rho(q, p) = \min\{qf_\alpha \bar{v}, qp + s\}$$

achieves worst-case regret r_α .

Optimal policy uses same instruments as in polar cases $\alpha = 0, 1$, i.e.,

- **piece-rate subsidy** ($f_\alpha \bar{v} - p$) at low prices (prevents underproduction)
- **cap on average revenue** at $f_\alpha \bar{v}$ (prevents excessive CS extraction)
- **cap on total subsidy** at s (prevents overproduction)

Additional Knowledge

- Suppose all $P \in \mathcal{P}$ s.t. $\underline{P} \leq P \leq \overline{P}$.
- Let

$$\underline{V}(q) = \int_0^q \underline{P}(z) dz \quad \text{and} \quad \overline{V}(q) = \int_0^q \overline{P}(z) dz$$

and note that (p, q) reveals that CS is at least

$$\Theta(q, p) = \int_0^q \max\{\underline{P}(z), p\} dz.$$

Theorem

Optimal policy is

$$\rho(q, p) = \min\{f_\alpha \overline{V}(q), \Theta(q, p) + s - d(q)\}$$

where

$$d(q) := \max_{0 \leq z \leq q} \{\underline{V}(z) - f_\alpha \overline{V}(z)\}$$

(with s appropriately chosen).

- Policy mitigates rents while avoiding too much (too little) production

Additional Knowledge

- With no uncertainty over demand (i.e., $\underline{P} = \overline{P} = P$)

$$\rho(q, p) = f_{\alpha} \overline{V}(q)$$

i.e., revenue is constant fraction f_{α} of gross consumer value (as with competitive ratio).

Conclusions

Conclusions

- Robust contracting
 - central to many design problems
- Traditional approach to robustness: worst-case optimality
 - multiple (continuum of) WCO mechanisms, many dominated
- Other model-free approaches deliver sharper predictions
 - competitive ratio
 - min-max regret
 - lexicographic min-max
- **Bayesian design subject to guarantee constraints**
 - blends ideas/techniques from Bayesian design with robustness analysis
 - designer uses model/conjecture to select over mechanisms yielding satisfactory guarantee

Conclusions

- A lot remains to be done
 - dynamics and experimentation (Suraj's lecture)
 - multi-dimensional screening (Rahul's lecture)
 - many agents and information robustness (Ben's lecture)
 - robust control, micro-foundations, and macro-finance (Lars' lecture)
 - value at risk and macro (Cosmin's lecture)
 - competing designers (largely unexplored)
 - ...

THANK YOU

Baron Myerson: Proof

Lemma

Mechanism $M = (q, u)$ IC and IR iff

- ① q non-increasing
- ② for all $\theta \in \Theta$,

$$u(\theta) = u(\bar{\theta}) + \int_{\theta}^{\bar{\theta}} q(z) dz$$

- ③ $u(\bar{\theta}) \geq 0$

Baron Myerson: Proof

- Expected payoff of any IC and IR $M \in \mathcal{M}$ under model (V^*, F^*) :

$$\int_{\underline{\theta}}^{\bar{\theta}} \left[V^*(q(\theta)) - z^*(\theta)q(\theta) \right] F^*(d\theta) - u(\bar{\theta})$$

- $q^{BM}(\theta)$ maximizes virtual surplus $V^*(q) - z^*(\theta)q$, all θ

- Regularity:

$\Rightarrow q^{BM}$ decreasing

$\Rightarrow q^{BM}$ (along with $u^{BM}(\bar{\theta}) = 0$ and $u(\theta) = \int_{\underline{\theta}}^{\bar{\theta}} q(z) dz$) **SEU-optimal**

Max-guarantee-proof

For any IC and IR mechanism $M = (q, u)$

$$\begin{aligned} W(M; V, F) &:= \int \{V(q(\theta)) - \theta q(\theta) - u(\theta)\} F(d\theta) \\ &\geq \int \{\underline{V}(q(\theta)) - \theta q(\theta) - u(\theta)\} F(d\theta) \\ &\geq \inf_{\theta \in \Theta} [\underline{V}(q(\theta)) - \theta q(\theta) - u(\theta)] \end{aligned}$$

Hence,

$$G(M) \geq \inf_{\theta \in \Theta} [\underline{V}(q(\theta)) - \theta q(\theta) - u(\theta)]$$

Because $\underline{V} \in \mathcal{V}$ and, for each θ , Dirac distribution at θ is in $\mathcal{F}$

$$G(M) \leq \inf_{\theta \in \Theta} [\underline{V}(q(\theta)) - \theta q(\theta) - u(\theta)]$$

Hence,

$$G(M) = \inf_{\theta \in \Theta} [\underline{V}(q(\theta)) - \theta q(\theta) - u(\theta)]$$

Because $u(\bar{\theta}) \geq 0$ and

$$\underline{V}(q(\bar{\theta})) - \bar{\theta} q(\bar{\theta}) \leq \underline{V}(q_\ell) - \bar{\theta} q_\ell := G^*$$

$$G(M) \leq G^*$$

BM-with-floor: Proof

- Full program:

$$\max_q \int_{\underline{\theta}}^{\bar{\theta}} \left[V^*(q(\theta)) - z^*(\theta)q(\theta) \right] F^*(d\theta)$$

q non-increasing

$$\underline{W}(\theta; q) := \underline{V}(q(\theta)) - \theta q(\theta) - \int_{\underline{\theta}}^{\bar{\theta}} q(y) dy \geq G^* \quad \forall \theta \in \Theta$$

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- Relaxed program:

$$\max_q \int_{\underline{\theta}}^{\bar{\theta}} \left[V^*(q(\theta)) - z^*(\theta)q(\theta) \right] F^*(d\theta)$$

$$q(\theta) \geq q_\ell \quad \forall \theta \in \Theta$$

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$q(\theta) \geq q_\ell \quad \forall \theta \in \Theta$

- BM-with-quantity-floor schedule q^* solves relaxed program
- Majorization lemma: conditions in Prop $\Rightarrow \underline{W}(\theta; q^*) \geq G^*$ all θ

Monotonicity Lemma

- Recall

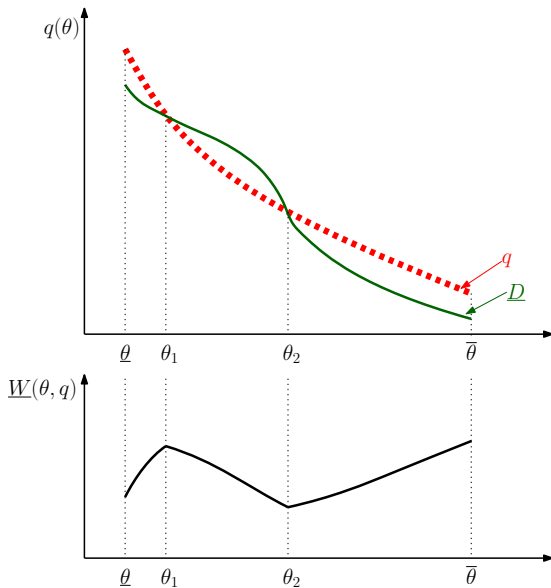
$$\underline{W}(\theta, q) := \underline{V}(q(\theta)) - \theta q(\theta) - \int_{\theta}^{\bar{\theta}} q(y) dy$$

Lemma (Monotonicity)

For any non-increasing q and interval $I \subseteq \Theta$,

- 1 If $0 < q(\theta) \leq \underline{D}(\theta)$ for all $\theta \in I$, then $\underline{W}(\cdot, q)$ non-increasing over I
(decreasing if q decreasing and $q(\theta) < \underline{D}(\theta)$ all $\theta \in I$)
- 2 If $q(\theta) > \underline{D}(\theta)$ for all $\theta \in I$, then $\underline{W}(\cdot, q)$ non-decreasing over I
(increasing if q decreasing)

Monotonicity Lemma: Illustration



Proposition: General Case

Let

$$\theta^m := \max\{\arg \min_{\theta \in \Theta} \underline{W}(\theta, q^*)\}$$

$$\theta^* := \begin{cases} (q^{\text{BM}})^{-1}(q_\ell) & \text{if } q^{\text{BM}}(\bar{\theta}) < q_\ell \\ \bar{\theta} & \text{otherwise} \end{cases}$$

Proposition

When BM-with-quantity-floor **not** robustly optimal,

- (a) $q^{\text{OPT}}(\theta) = q^{\text{BM}}(\theta)$ for all $\theta \in (\underline{\theta}, \theta^m)$
- (b) $q^{\text{OPT}}(\theta) \leq q^{\text{BM}}(\theta)$ for all $\theta \in (\theta^m, \theta^*)$
- (c) $q^{\text{OPT}}(\theta) = q_\ell$ for all $\theta \in [\theta^*, \bar{\theta}]$

Furthermore, if $\theta^* < \bar{\theta}$, then $q^{\text{OPT}}(\theta) < q^{\text{BM}}(\theta)$ for all $\theta \in (\theta^m, \theta^*)$. If, instead, $\theta^* = \bar{\theta}$, there exists $\theta^{**} \in [\theta^m, \theta^*]$ such that $q^{\text{OPT}}(\theta) < q^{\text{BM}}(\theta)$ for all $\theta \in (\theta^m, \theta^{**})$ and $q^{\text{OPT}}(\theta) = \underline{D}(\theta)$ for all $\theta \in [\theta^{**}, \bar{\theta}]$.

More General Cost Uncertainty

- Suppose there exist $\underline{F} \in \text{CDF}(\Theta)$ s.t.

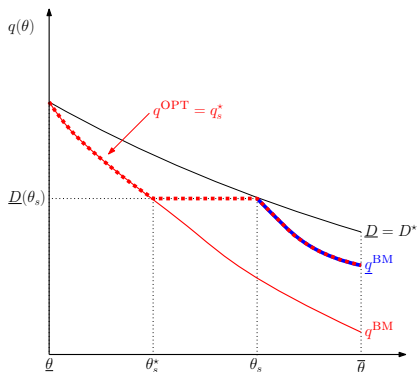
$$\mathcal{F} = \{F \in \text{CDF}(\Theta) : \underline{F}(\theta) \leq F(\theta), \forall \theta\}$$

with $F^* \in \mathcal{F}$

- So far, $\underline{F}(\theta) = \mathbb{I}(\theta \geq \bar{\theta})$
- Generalization also accommodates for unbounded Θ

More General Cost Uncertainty

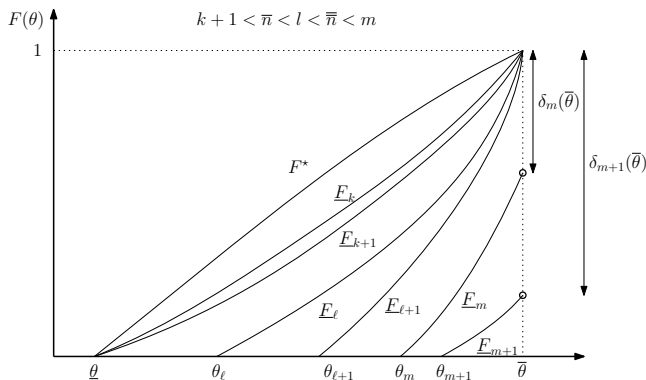
- Suppose $\underline{F}(\theta) = 0$ for all $\theta \leq \theta_s$, with $\theta_s \in (\underline{\theta}, \bar{\theta})$, and regular over $[\theta_s, \bar{\theta})$
- For simplicity, $D^* = \underline{D}$



- Optimal mechanism bridges two BM schedules
 - Low costs: BM under $(F^*, \underline{V})$
 - High costs: BM under $(\underline{F}, \underline{V})$
 - No distortion at top **and middle** (θ_s)

Variations in Cost Uncertainty

- Suppose $\bar{\theta}$ finite
- (Monotone) sequence $(\underline{F}_n)$ converging to $\mathbb{I}(\theta \geq \bar{\theta})$



Variations in Cost Uncertainty

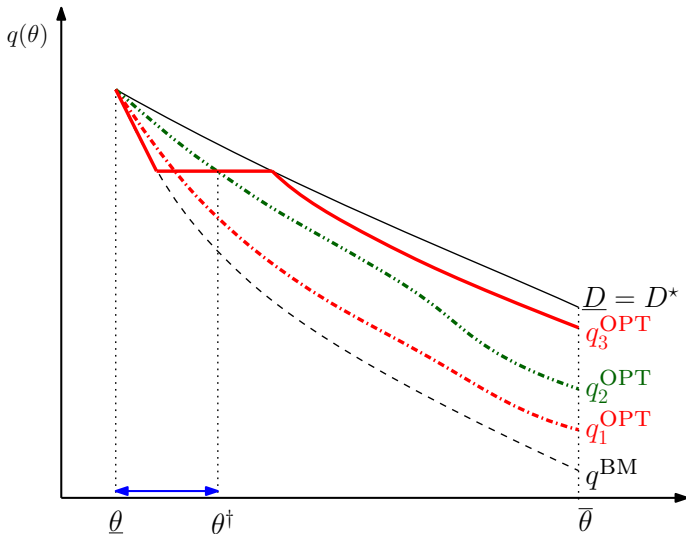
Proposition

Suppose $D^ = \underline{D}$. There exist monotone sequences $(\underline{F}_n)$ s.t., for every $\theta \in (\underline{\theta}, \bar{\theta})$, there exists $n(\theta) \in \mathbb{N}$ s.t.*

$$q_{n+1}^{OPT}(\theta) \begin{cases} \geq \\ \leq \end{cases} q_n^{OPT}(\theta) \text{ for } \begin{cases} n \leq n(\theta) - 1 \\ n > n(\theta) \end{cases}$$

Moreover, for all θ , there exists $j, k \in \mathbb{N}$ with $j < k$, s.t. $q_k^{OPT}(\theta) < q_j^{OPT}(\theta)$

Variations in Cost Uncertainty



Price Mechanisms: EPIC and EPIR

Definition

Price mechanism $\tilde{M} = (p, t)$ **EPIC** and **EPIR** if, for all $\theta, \theta' \in \Theta$ and $D \in \mathcal{D}$,

$$t(\theta, D) - \theta D(p(\theta)) \geq t(\theta', D) - \theta D(p(\theta'))$$

$$\tilde{u}(\theta, D) := t(\theta, D) - \theta D(p(\theta)) \geq 0$$

- $\tilde{\mathcal{M}}$: set of EPIC and EPIR price mechanisms $\tilde{M} := (p, \tilde{u})$

Price Mechanisms: EPIC and EPIR

Lemma

$\tilde{M} = (p, t)$ is EPIC and EPIR iff p is non-decreasing, and for every $\theta \in \Theta$ and $D \in \mathcal{D}$,

$$\tilde{u}(\theta, D) = \tilde{u}(\bar{\theta}, D) + \int_{\theta}^{\bar{\theta}} D(p(y)) dy,$$

with $\tilde{u}(\bar{\theta}, D) \geq 0$.

Robust Optimality of Baron-Myerson-with-price-cap

- Welfare under $\tilde{M} = (p, \tilde{u}) \in \tilde{\mathcal{M}}$ given (D, F) :

$$\tilde{W}(\tilde{M}; D, F) := \int \{V(D(p(\theta))) - \theta D(p(\theta)) - \tilde{u}(\theta, D)\} F(d\theta)$$

Robust Optimality of Baron-Myerson-with-price-cap

$$\max_{p, \{\bar{u}(\bar{\theta}, D) \geq 0\}} \int_{\underline{\theta}}^{\bar{\theta}} \left[V^*(D^*(p(\theta))) - z^*(\theta) D^*(p(\theta)) \right] F^*(d\theta) - \bar{u}(\bar{\theta}, D^*)$$

s.t.

p non-decreasing

$$\int_0^{D(p(\theta))} D^{-1}(y) dy - \theta D(p(\theta)) - \int_{\theta}^{\bar{\theta}} D(p(y)) dy - \bar{u}(\bar{\theta}, D) \geq G^* \quad \forall \theta \in \Theta \quad \forall D \in \mathcal{D}$$

$$p(\bar{\theta}) = \bar{\theta}$$

Robust Optimality of Baron-Myerson-with-price-cap

- Relaxed program:

$$\max_{p, \{\tilde{u}(\bar{\theta}, D) \geq 0\}} \int_{\bar{\theta}}^{\bar{\theta}} \left[V^*(D^*(p(\theta))) - z^*(\theta) D^*(p(\theta)) \right] F^*(d\theta) - \tilde{u}(\bar{\theta}, D^*)$$

s.t.

p non-decreasing

$$\int_0^{D(\bar{\theta})} D^{-1}(y) dy - \bar{\theta} D(\bar{\theta}) - \tilde{u}(\bar{\theta}, D) \geq G^* \quad \forall D \in \mathcal{D}$$

$$p(\bar{\theta}) = \bar{\theta}$$

- $p^{\text{OPT}}(\theta) = \min\{z^*(\theta), \bar{\theta}\}$, along with $\{\tilde{u}^{\text{OPT}}(\bar{\theta}, D)\}$ chosen to satisfy remaining robustness constraints (blue inequalities), solves relaxed program
- Because $p^{\text{OPT}}(\theta) \geq \theta$ all θ , remaining robustness constraint satisfied

Quantity Regulation Dominates

- Robustness under price mechanism $\Rightarrow$ **price cap** at $\bar{\theta}$
- When $D^*(\bar{\theta}) > \underline{D}(\bar{\theta})$
 $\Rightarrow$ **over-procurement (vis-a-vis quantity mechanism) for high costs**
- When BM-with-quantity-floor mechanism robustly optimal
 $\Rightarrow$ **quantity regulation dominates**

Price Regulation Dominates

- When BM-with-quantity-floor mechanism **not** robustly optimal

⇒ **downward qty adjustments** for intermediate costs (under D^* vis-a-vis q^{BM})

- adjustments necessary to avoid over-procurement when $D < D^*$
 - under price regulation, demand adjusts itself
 - adjustment do not occur when $D = D^*$
- If price cap at $\bar{\theta}$ does not induce over-procurement vis-a-vis q_ℓ

⇒ **price regulation strictly dominates**

Qty vs Price: Primitives

Let $\underline{\delta} \equiv \inf\{D^*(\theta) - \underline{D}(\theta) : \theta \in \Theta\}$, $\bar{\delta} \equiv \sup\{D^*(\theta) - \underline{D}(\theta) : \theta \in \Theta\}$,
 $\underline{f} \equiv \inf\{f^*(\theta) : \theta \in \Theta\}$ and $\bar{f} \equiv \sup\{f^*(\theta) : \theta \in \Theta\}$.

Proposition

Suppose D^* is bi-Lipschitz continuous over $[0, P^*(0)]$ with constants $0 < k \leq K < \infty$.

- Quantity regulation dominates if, for all $\theta \in \Theta$,

$$\frac{F^*(\theta)}{f^*(\theta)} \geq \frac{D^*(\theta) - \underline{D}(\theta)}{k},$$

(strictly if $D^*(\bar{\theta}) - \underline{D}(\bar{\theta}) > 0$)

- When, instead, for all $\theta \in \Theta$,

$$\frac{F^*(\theta)}{f^*(\theta)} \leq \frac{D^*(\theta) - \underline{D}(\theta)}{K \left(1 + \frac{\bar{\delta}}{\underline{\delta}} \sqrt{\frac{2\bar{f}}{\underline{f}}}\right)},$$

and $\underline{\delta} > 0$, price regulation strictly dominates.

Robustness and WCO: Proof

- **Sufficiency** ($\Leftarrow$): Take $F = (\delta_N)$, where δ_N is Dirac's delta at θ_N . Then, F is adversarial wrt M . Furthermore, M is F -optimal. Hence M robust.

Robustness and WCO: Proof

- **Necessity** ($\Rightarrow$): Suppose $V(q(\theta)) - t(\theta) < V(q^{FB}(\theta_N)) - t^{FB}(\theta_N)$ for some $\theta \in \Theta^\#$. Then, under any F adversarial wrt M , first-order payoff

$$w(M; F_1) < V(q^{FB}(\theta_N)) - t^{FB}(\theta_N).$$

Consider “flat-mechanism”

$$\tilde{M} = (\tilde{q}, \tilde{t}) \text{ s.t. } \tilde{q}(\theta) = q^{FB}(\theta_N) \text{ and } \tilde{t}(\theta) = t^{FB}(\theta_N) \quad \forall \theta \in \Theta^\#$$

$\tilde{M}$ is IC and IR and $w_F(\tilde{M}) >_L w_F(M)$. Hence M not robust.

- Furthermore, if $(q(\theta_N), t(\theta_N)) \neq (q^{FB}(\theta_N), t^{FB}(\theta_N))$, Nature can bring designer's payoff strictly below $V(q^{FB}(\theta_N)) - t^{FB}(\theta_N)$, and hence M not WCO (and therefore not robust).

Perfect Robustness and Undomination: Proof

- **Necessity** ($\Rightarrow$): Perfect robustness means there exists full-support F adversarial wrt M s.t. M is F -optimal.
 - This means M robust.
 - Suppose M dominated (say by M'). Then $w_F(M') >_L w_F(M)$ for any F of full support. Hence, M not perfectly robust.

Perfect Robustness and Undomination: Proof

- **Sufficiency** ($\Leftarrow$): Suppose M robust and undominated. Then

$$\begin{cases} V(q(\theta)) - t(\theta) \geq V(q(\theta_N)) - t(\theta_N) \quad \forall \theta \in \Theta^\# \\ (q(\theta_N), t(\theta_N)) = (q^{FB}(\theta_N), t^{FB}(\theta_N)) \end{cases}$$

- This means that Dirac δ_N at θ_N is adversarial wrt M and, for any IC-IR $M' \in \mathcal{M}$,

$$V(q'(\theta_N)) - t'(\theta_N) \leq V(q^{FB}(\theta_N)) - t^{FB}(\theta_N)$$

(Hence, M' delivers weakly less than M against δ_N)

- Because M undominated, $\exists$ full-support $\tilde{F} \in \Delta(\Theta^\#)$ s.t., for any IC-IR $M' \in \mathcal{M}$,

$$\sum_{\theta \in \Theta} f(\theta) \cdot (V(q(\theta)) - t(\theta)) \geq \sum_{\theta \in \Theta} f(\theta) \cdot (V(q'(\theta)) - t'(\theta))$$

(Proof requires some work based on linear programming).

- This means M perfectly robust (it suffices to take LPS $F = (\delta_N, \tilde{F})$)

Proper Robustness and Efficiency: Proof

- **Sufficiency:** Let $F = (\delta_N, \dots, \delta_1)$ and note that F is strongly adversarial wrt M^{FB} (because $V(q^{FB}(\theta)) - t^{FB}(\theta)$ is decreasing in θ) and has full support. It thus suffices to show that M^{FB} is F -optimal.

Proof by induction. No IC-IR mechanism yields higher profits against δ_N .

Furthermore, any mechanism $M \in \mathcal{M}$ that yields same payoff as M^{FB} against δ_N is s.t.

- $(q(\theta_N), t(\theta_N)) = (q^{FB}(\theta_N), t^{FB}(\theta_N))$
- leaves rent $u(\theta_{N-1})$ to θ_{N-1} no smaller than $t^{FB}(\theta_{N-1}) - \theta_{N-1}q^{FB}(\theta_{N-1})$;
This because

$$(q^{FB}(\theta_{N-1}), t^{FB}(\theta_{N-1})) = \arg \max_{q, t} \{V(q) - \theta_{N-1}q - u(\theta_{N-1})\}$$

$$\text{s.t. } u(\theta_{N-1}) \geq (\theta_N - \theta_{N-1})q^{FB}(\theta_N)$$

- Inductively, $w_F(M^{FB}) \geq_L w_F(M)$ for all IC-IR $M \in \mathcal{M}$.
- This means M^{FB} properly robust.

Proper Robustness and Efficiency: Proof

- **Sufficiency (intuition):** Under M^{FB} , each type supplies efficiently and receives minimal transfer among all mechanisms implementing efficient trade. Hence, any IC-IR mechanism delivering higher payoff against some type $\theta_k < \theta_N$ must deliver smaller rent to θ_k . But this means lower payoff against some less-efficient type $\theta_l > \theta_k$. Because these less-efficient types are infinitely more likely under $F = (\delta_N, \dots, \delta_1)$, M^{FB} is F -optimal against $F = (\delta_N, \dots, \delta_1)$.

Proper Robustness and Efficiency: Proof

Necessity follows from following 3 results:

- ① $M = (q, t) \in \mathcal{M}$ properly robust $\implies V(q(\theta)) - t(\theta)$ non-increasing.
- ② $M = (q, t) \in \mathcal{M}$ properly robust $\implies M$ robust.
- ③ $M = (q, t) \in \mathcal{M}$ properly robust and $(q(\theta_i), t(\theta_i)) = (q^{FB}(\theta_i), t^{FB}(\theta_i))$ for all $i = N - k, \dots, N \implies$

$$(q(\theta_{N-k-1}), t(\theta_{N-k-1})) = (q^{FB}(\theta_{N-k-1}), t^{FB}(\theta_{N-k-1}))$$

Proper Robustness and Efficiency: Proof

Result 1: $M = (q, t) \in \mathcal{M}$ properly robust $\implies V(q(\theta)) - t(\theta)$ non-increasing.

Proof of Result 1: Suppose contrary, i.e., $\exists j$ s.t.

$$V(q(\theta_{j-1})) - t(\theta_{j-1}) < V(q(\theta_j)) - t(\theta_j) \quad (*)$$

Let $J := \sup\{j : (*) \text{ holds}\}$. Consider $M^+ = (q^+, t^+)$ s.t.

$$\begin{cases} (q^+(\theta_j), t^+(\theta_j)) = (q(\theta_j), t(\theta_j)) & \text{for all } j = 1, \dots, J \\ (q^+(\theta_j), t^+(\theta_j)) = (q(\theta_j), t(\theta_j)) & \text{for all } j = J + 1, \dots, N \end{cases}$$

Clearly, $M^+ \in \mathcal{M}$: because $(q(\theta_J), t(\theta_J)) \succeq_J (q(\theta_j), t(\theta_j))$ for all $j \geq J$,
 $(q(\theta_J), t(\theta_J)) \succeq_m (q(\theta_j), t(\theta_j))$ for all $m \leq J$ all $j \geq J$, as $t - \theta q$ has SCP.

Proper Robustness and Efficiency: Proof

Proof of Result 1 (continued): $w_F(M^+) >_L w_F(M)$ for any full-support F strongly adversarial wrt M . To see this, let $k^* \equiv \min\{k : f_k(\theta_J) > 0\}$.

- For all $\theta \geq \theta_J$, $V(q^+(\theta)) - t^+(\theta) = V(q(\theta)) - t(\theta)$.
- For any $\theta < \theta_J$, if $f_k(\theta) > 0$ for some $k \leq k^*$, then $\theta \geq_F \theta_J$ and hence

$$V(q(\theta)) - t(\theta) \leq V(q(\theta_J)) - t(\theta_J)$$

because F strongly adversarial wrt M .

- Hence, for any $k = 1, \dots, k^*$,

$$\sum_{\theta \in \Theta} f_k(\theta) \cdot (V(q(\theta)) - t(\theta)) \leq \sum_{\theta \in \Theta} f_k(\theta) \cdot (V(q^+(\theta)) - t^+(\theta))$$

- Furthermore,

$$\sum_{\theta \in \Theta} f_{k^*}(\theta) \cdot (V(q(\theta)) - t(\theta)) < \sum_{\theta \in \Theta} f_{k^*}(\theta) \cdot (V(q^+(\theta)) - t^+(\theta))$$

- Hence $w_F(M^+) >_L w_F(M)$. Therefore, M not properly robust.

Proper Robustness implies FB Efficiency: Proof

Result 2: $M = (q, t) \in \mathcal{M}$ properly robust $\implies M$ robust.

Proof of Result 2: Because M properly robust, $\exists$ full-support F strongly adversarial wrt M s.t. M F -optimal. That F strongly adversarial implies F adversarial. This means M robust (and hence WCO). Q.E.D.

- In turn, this implies $(q(\theta_N), t(\theta_N)) = (q^{FB}(\theta_N), t^{FB}(\theta_N))$

Proper Robustness and Efficiency: Proof

Result 3: $M = (q, t) \in \mathcal{M}$ properly robust and $(q(\theta_i), t(\theta_i)) = (q^{FB}(\theta_i), t^{FB}(\theta_i))$ for all $i = r, \dots, N \implies$

$$(q(\theta_{r-1}), t(\theta_{r-1})) = (q^{FB}(\theta_{r-1}), t^{FB}(\theta_{r-1}))$$

Proof of Result 3: Suppose not. Then $w_F(M^{FB}) >_L w_F(M)$ for any full-support F strongly adversarial wrt M . To see this, first observe

$$V(q^{FB}(\theta_j)) - t^{FB}(\theta_j) > V(q(\theta_{r-1})) - t(\theta_{r-1}) \quad \forall j < r \quad (**)$$

In particular, for $j = r - 1$,

$$V(q^{FB}(\theta_{r-1})) - t^{FB}(\theta_{r-1}) > V(q(\theta_{r-1})) - t(\theta_{r-1})$$

This follows from

$$(q^{FB}(\theta_{r-1}), t^{FB}(\theta_{r-1})) = \arg \max_{q, t} \{V(q) - \theta_{r-1}q - \Pi(q, t; \theta_{r-1})\}$$

$$\begin{aligned} \text{s.t. } \Pi(q, t; \theta_{r-1}) &\equiv t - \theta_{r-1}q \\ &\geq t^{FB}(\theta_r) - \theta_{r-1}q^{FB}(\theta_r) \end{aligned}$$

Because $V(q^{FB}(\theta)) - t^{FB}(\theta)$ non-increasing, $(**)$ follows.

Proper Robustness and Efficiency: Proof

Proof of Result 3 (continued):

Let $J = r - 1$ if $V(q(\theta_{r-2})) - t(\theta_{r-2}) > V(q(\theta_{r-1})) - t(\theta_{r-1})$. Else,

$$J \equiv \min\{j : V(q(\theta_j)) - t(\theta_j) = V(q(\theta_{r-1})) - t(\theta_{r-1})\}$$

Because M properly robust, $V(q(\theta)) - t(\theta)$ non-increasing. Hence, for any $\theta < \theta_J$ and any $j = J, \dots, N$, $\theta_j >_{\mathbb{F}} \theta$ under any full-support $\mathbb{F}$ strongly adversarial wrt M . But then, $w_{\mathbb{F}}(M^{FB}) >_L w_{\mathbb{F}}(M)$. In fact, let $\hat{k} \equiv \min\{k : f_k(\theta_{r-1}) > 0\}$.

- For all $k < \hat{k}$, previous property and fact that $\mathbb{F}$ strongly adversarial wrt M ,

$$\sum_{\theta \in \Theta} f_k(\theta) \cdot (V(q(\theta)) - t(\theta)) \leq \sum_{\theta \in \Theta} f_k(\theta) \cdot (V(q^{FB}(\theta)) - t^{FB}(\theta))$$

- Furthermore, for $k = \hat{k}$,

$$\sum_{\theta \in \Theta} f_{\hat{k}}(\theta) \cdot (V(q(\theta)) - t(\theta)) < \sum_{\theta \in \Theta} f_{\hat{k}}(\theta) \cdot (V(q^{FB}(\theta)) - t^{FB}(\theta))$$

- Hence, induction step applies. Q.E.D.
- Results 1-3 imply that if M **properly robust** then $M = M^{FB}$ (i.e., FB-efficient)