

# The Persuasion Duality

Piotr Dworczak & Anton Kolotilin  
(Northwestern University) (UNSW Business School)

University of Oxford, October, 2025

# Motivation

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- An explosion of interest in **Bayesian persuasion** following Kamenica and Gentzkow (2011)
- Concavification not always tractable
- A number of papers propose **duality** as a tool to solve information design problems:
  - Kolotilin (2018);
  - Dworczak and Martini (2019);
  - Dizdar and Kováč (2020);
  - Kolotilin, Corrao, and Wolitzky (2025);
  - (Galperti, Levkun, and Perego (2023));
  - ...

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- The optimal dual variable, interpreted as a price function, is a **supergradient of the concave closure of the objective function at the prior belief**.
- Our results unify and generalize existing duality results in persuasion.
- This minicourse introduces methodology and illustrates it in applications.

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Sender chooses a signal/experiment to reveal information about a state to Receiver, who then takes an action.

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- Sender designs a signal:
  - Signal realization space  $S$ ;
  - Conditional probability  $\pi(s|\omega)$  for each  $s \in S$  and  $\omega \in \Omega$ ;

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- Receiver chooses action  $a \in A$  (compact).
- Receiver's utility is  $u(a, \omega)$  and Sender's utility is  $v(a, \omega)$ ;  
both are continuous

## Belief-based approach

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- After observing  $s$ , Receiver uses Bayes' rule to update his belief from the *prior*  $\mu_0 \in \Delta(\Omega)$  to the *posterior*  $\mu \in \Delta(\Omega)$ .

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$$a^*(\mu) \in \arg \max_{a \in A} \mathbb{E}_\mu[u(a, \omega)],$$

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- Sender's *indirect utility* from posterior  $\mu$  is

$$V(\mu) = \mathbb{E}_\mu[v(a^*(\mu), \omega)].$$

- A signal  $\pi$  induces a distribution  $\tau$  over posteriors  $\mu$ , so Sender's expected utility is  $\mathbb{E}_\tau[V(\mu)]$ .

## Splitting lemma

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**Lemma.** There exists a signal  $\pi$  that induces a distribution of posteriors  $\tau \in \Delta(\Delta(\Omega))$  iff  $\mathbb{E}_\tau[\mu] = \mu_0$ .

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*Proof:* The only if part follows from the law of iterated expectations.

# Splitting lemma

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**Lemma.** There exists a signal  $\pi$  that induces a distribution of posteriors  $\tau \in \Delta(\Delta(\Omega))$  iff  $\mathbb{E}_\tau[\mu] = \mu_0$ .

*Proof:* The only if part follows from the law of iterated expectations. The if part is shown by construction. Indeed, in the finite case, define, for all  $\omega \in \Omega$  and all  $\mu$  in  $\text{supp}(\tau)$ ,

$$\begin{aligned}\pi(\mu|\omega) &= \frac{\mu(\omega)\tau(\mu)}{\mu_0(\omega)} \\ \implies \Pr(\omega|\mu) &= \frac{\pi(\mu|\omega)\mu_0(\omega)}{\tau(\mu)} = \mu(\omega).\end{aligned}$$

# Concavification

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Sender's problem is to find a distribution of posteriors  $\tau$  to

$$\text{maximize } \mathbb{E}_\tau[V(\mu)]$$

$$\text{subject to } \mathbb{E}_\tau[\mu] = \mu_0.$$

Smallest concave function that is everywhere greater than  $V$  is called *concave closure* of  $V$  and is denoted by  $\widehat{V}$ .

**Concavification.** The value of Sender's problem is  $\widehat{V}(\mu_0)$ .

## Recap of linear programming

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Fix an  $m \times n$  matrix  $A$ , an  $n$ -vector  $b$ , and an  $m$ -vector  $c$ .

If the *primal problem* is to find an  $m$ -vector  $x \geq 0$  to

$$\begin{aligned} & \text{maximize } xc \\ & \text{subject to } xA = b, \end{aligned}$$

then the *dual problem* is to find an  $n$ -vector  $y$  to

$$\begin{aligned} & \text{minimize } by \\ & \text{subject to } Ay \geq c. \end{aligned}$$

## Recap of linear-programming duality

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Primal: find  $x \geq 0$  to      Dual: find  $y$  to  
maximize  $xc$       minimize  $by$   
subject to  $xA = b$ ,      subject to  $Ay \geq c$ .

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**Weak duality.** If  $x$  and  $y$  are feasible solutions, then

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**Optimality criterion.** If feasible solutions  $x$  and  $y$  satisfy  $xc = by$ , then they are optimal, as, for any feasible solutions  $\tilde{x}$  and  $\tilde{y}$ ,

$$\tilde{x}c \leq by = xc \quad \text{and} \quad b\tilde{y} \geq xc = by.$$

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$$\tilde{xc} \leq b\tilde{y} = xc \quad \text{and} \quad b\tilde{y} \geq xc = by.$$

**Strong duality.** If the primal and dual admit feasible solutions, then both have optimal solutions  $x$  and  $y$ , and they satisfy  $xc = by$ .

# Primal Problem in Dworczak and Kolotilin (2024)

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Find a distribution of posteriors  $\tau \in \Delta(\Delta(\Omega))$  to

$$\begin{aligned} & \text{maximize} \int_{\Delta(\Omega)} V(\mu) d\tau(\mu) \\ & \text{subject to} \int_{\Delta(\Omega)} \mu d\tau(\mu) = \mu_0 \end{aligned} \tag{P}$$

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where

- $(\Omega, \rho)$  is a compact metric space
- $\mu_0 \in \Delta(\Omega)$  is a prior belief
- $V : \Delta(\Omega) \rightarrow \mathbb{R}$  is bounded and u.s.c. in the weak $^*$  topology

# Technical Background

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- Let  $L(\Omega)$  denote the set of Lipschitz functions on  $\Omega$
- **Fact 3.** The space dual to  $(M(\Omega), \|\cdot\|_{KR})$  is  $L(\Omega)$
- Note that  $(\Delta(\Delta(\Omega)), \|\cdot\|_{KR})$  is also a compact metric space

## Dual problem

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The dual problem is to find a *price function*  $p \in L(\Omega)$  to

$$\begin{aligned} & \text{minimize} \int_{\Omega} p(\omega) d\mu_0(\omega) \\ \text{subject to } & \int_{\Omega} p(\omega) d\mu(\omega) \geq V(\mu) \text{ for all } \mu \in \Delta(\Omega). \end{aligned} \tag{D}$$

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The *concave closure* of  $V$  at  $\mu_0$  is the value of the primal problem:

$$\widehat{V}(\mu_0) := \sup_{\tau \in \mathcal{T}(\mu_0)} \int_{\Delta(\Omega)} V(\mu) d\tau(\mu).$$

$\widehat{V}(\mu_0)$  is the supremum of  $z$  such that  $(z, \mu_0)$  belongs to the convex hull of the graph of  $V$  on  $\Delta(\Omega)$ .

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The *concave envelope* of  $V$  at  $\mu_0$  is the value of the dual problem:

$$\overline{V}(\mu_0) := \inf_{p \in \mathcal{P}(V)} \int_{\Omega} p(\omega) d\mu_0(\omega).$$

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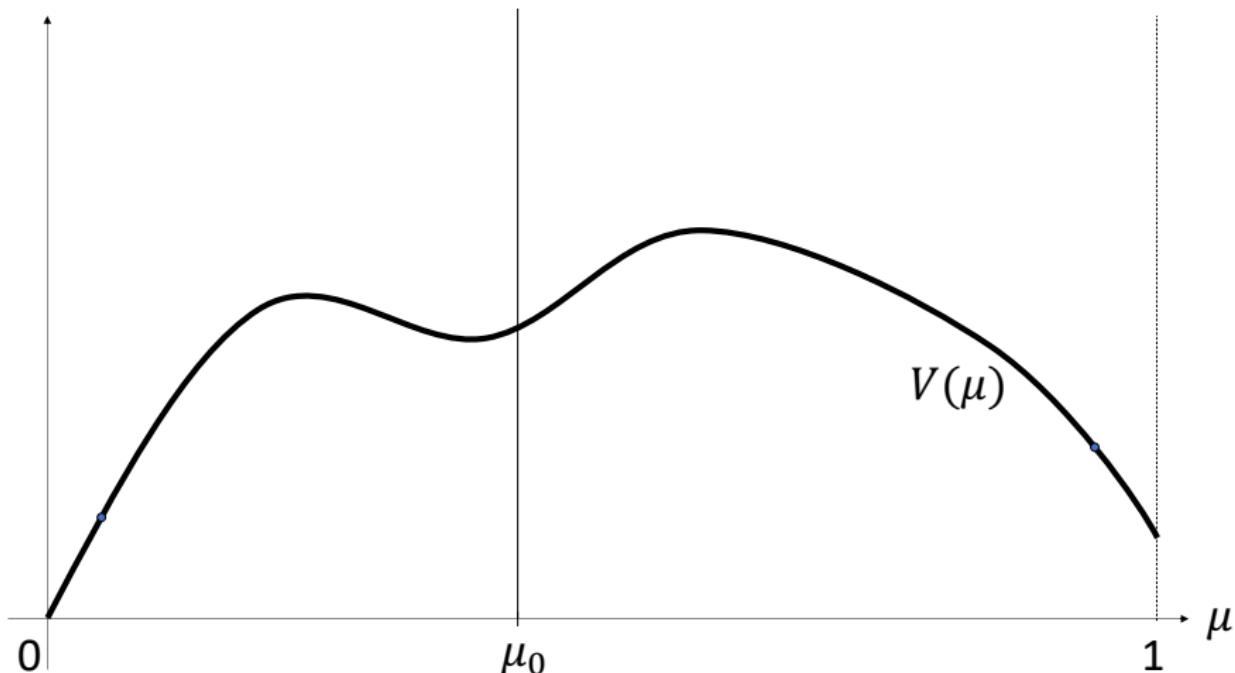
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$\overline{V}(\mu_0)$  is the infimum at  $\mu_0$  of continuous linear functions on  $M(\Omega)$  that bound  $V$  from above on  $\Delta(\Omega)$ , because the space  $L(\Omega)$  is dual to  $(M(\Omega), \|\cdot\|_{KR})$ .

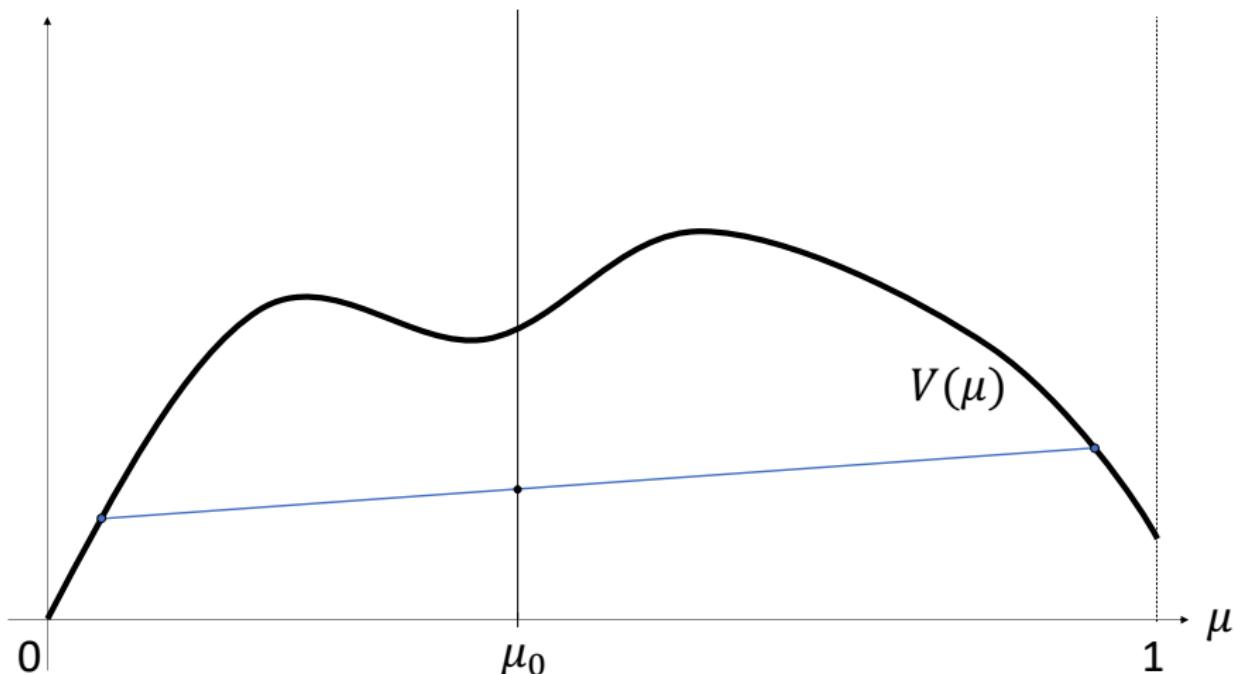
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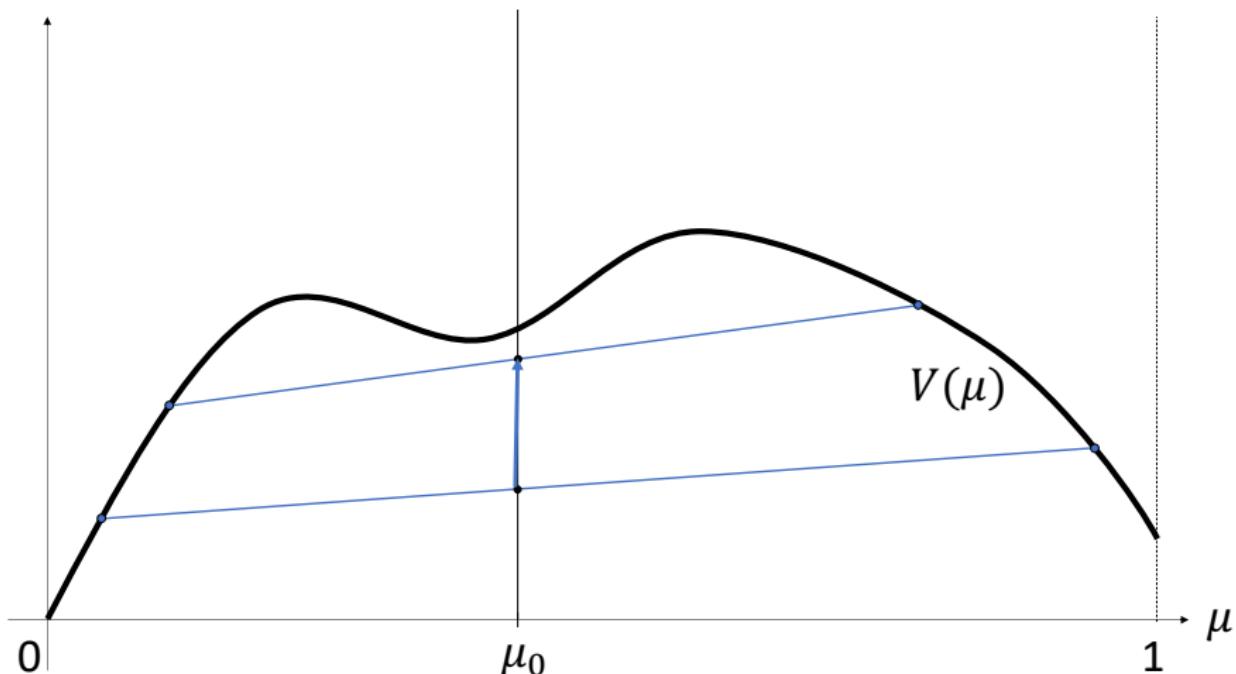
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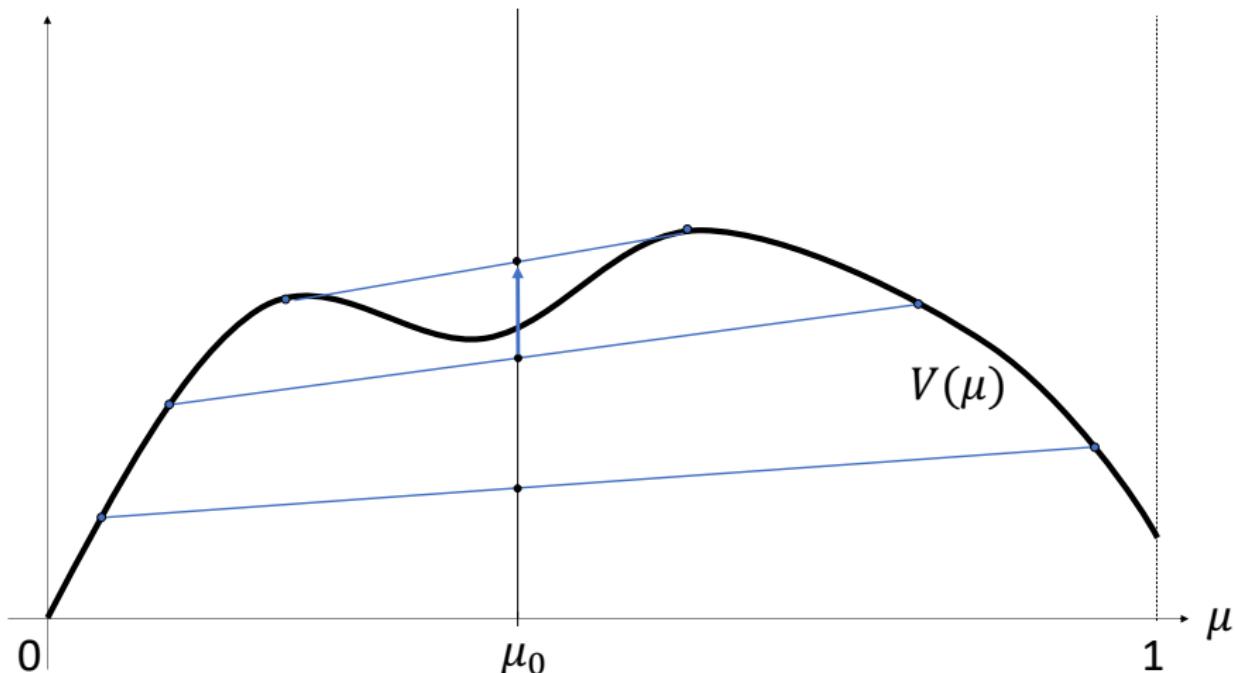
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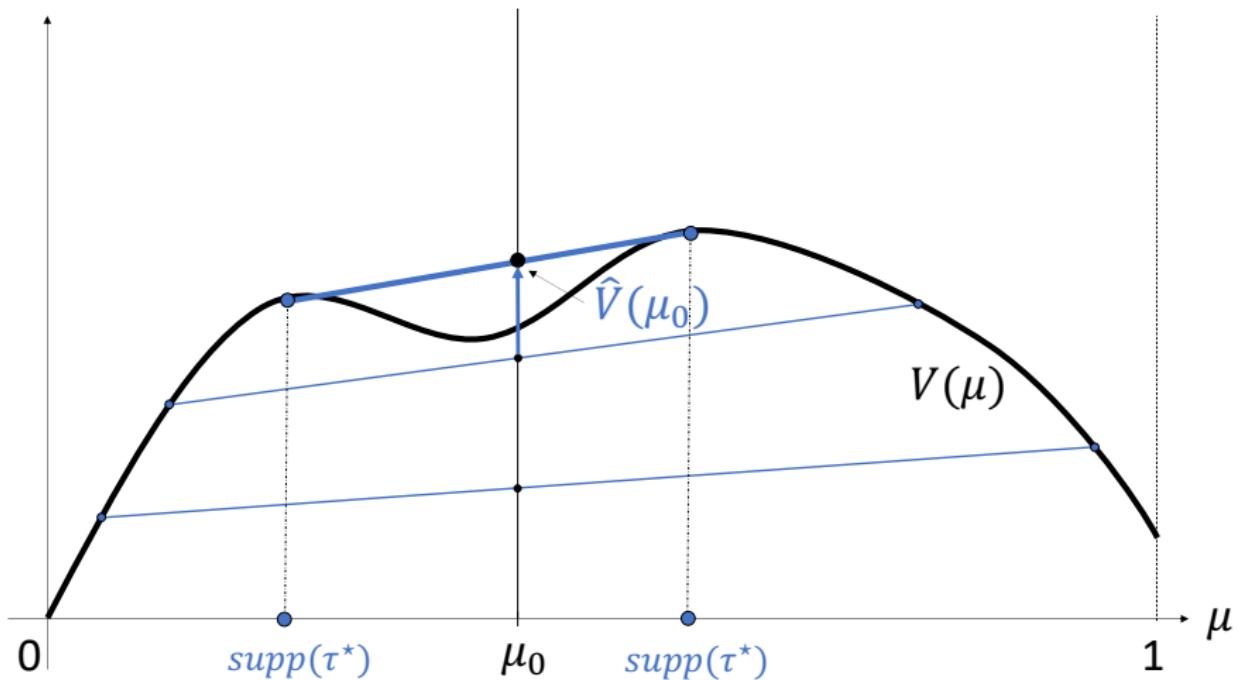
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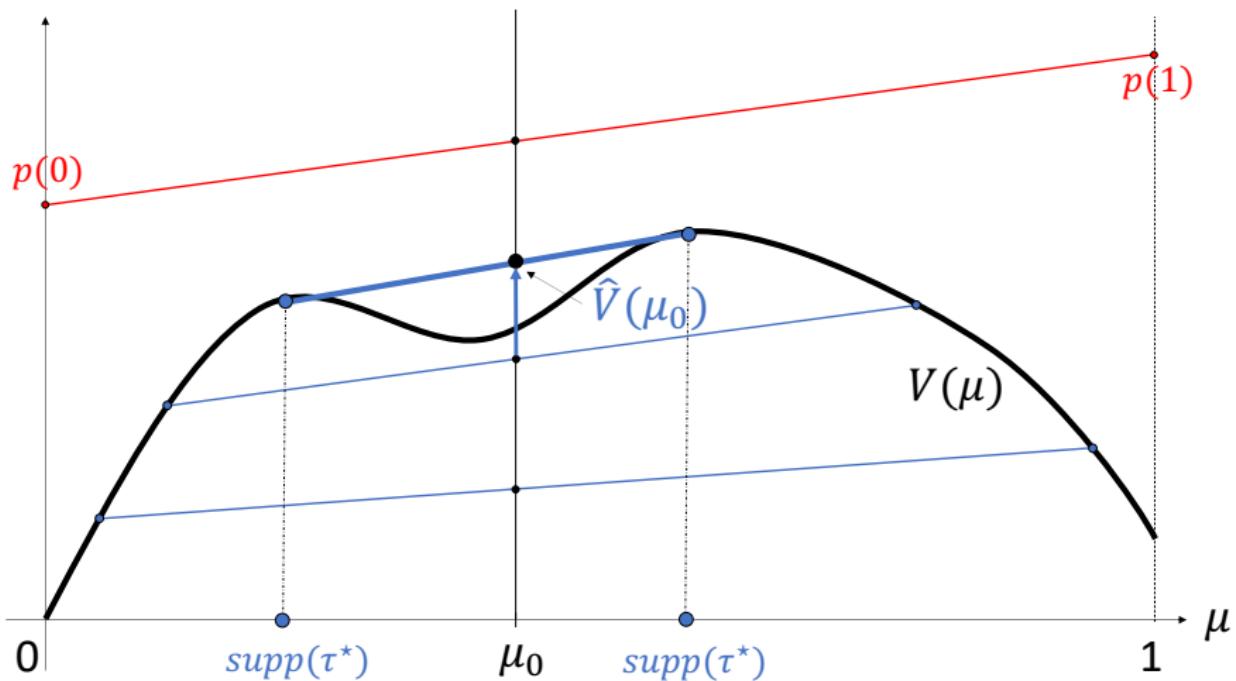
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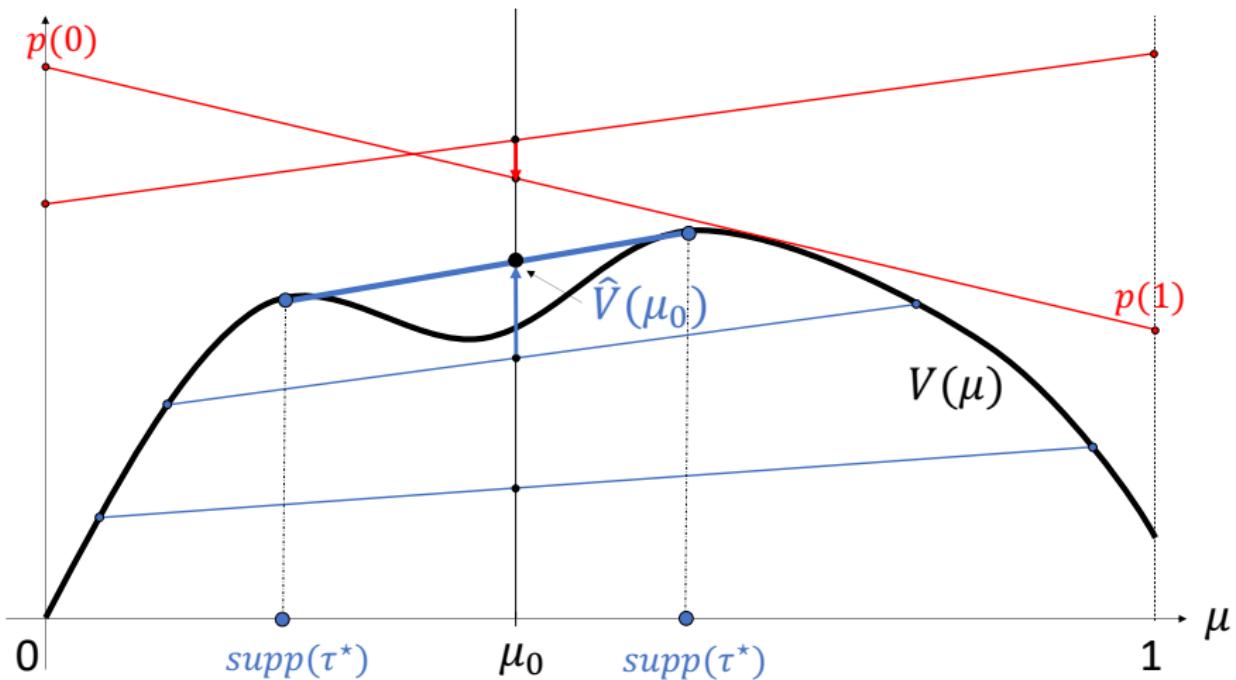
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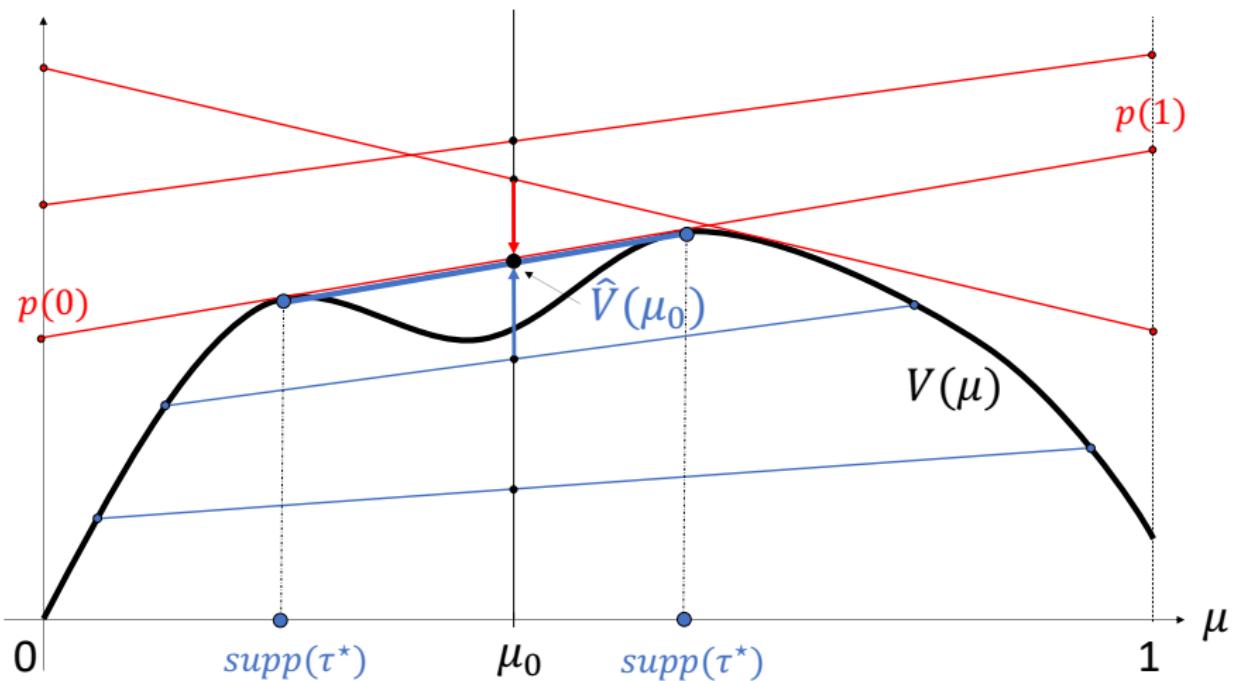
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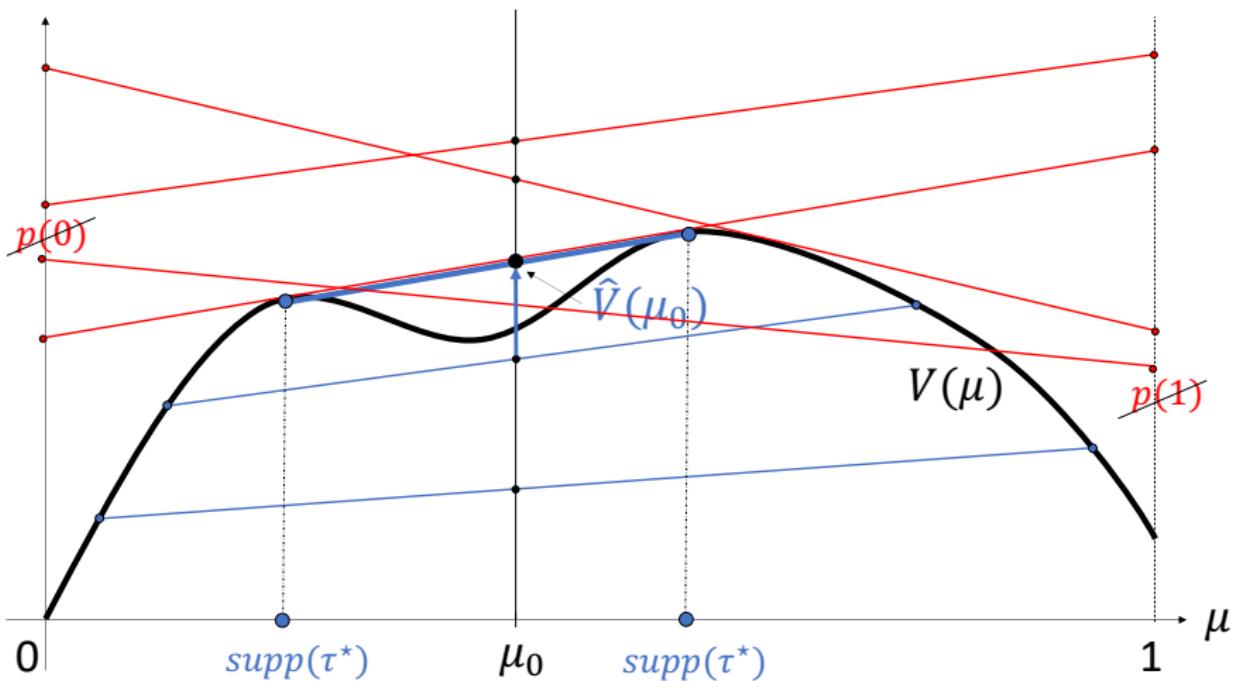
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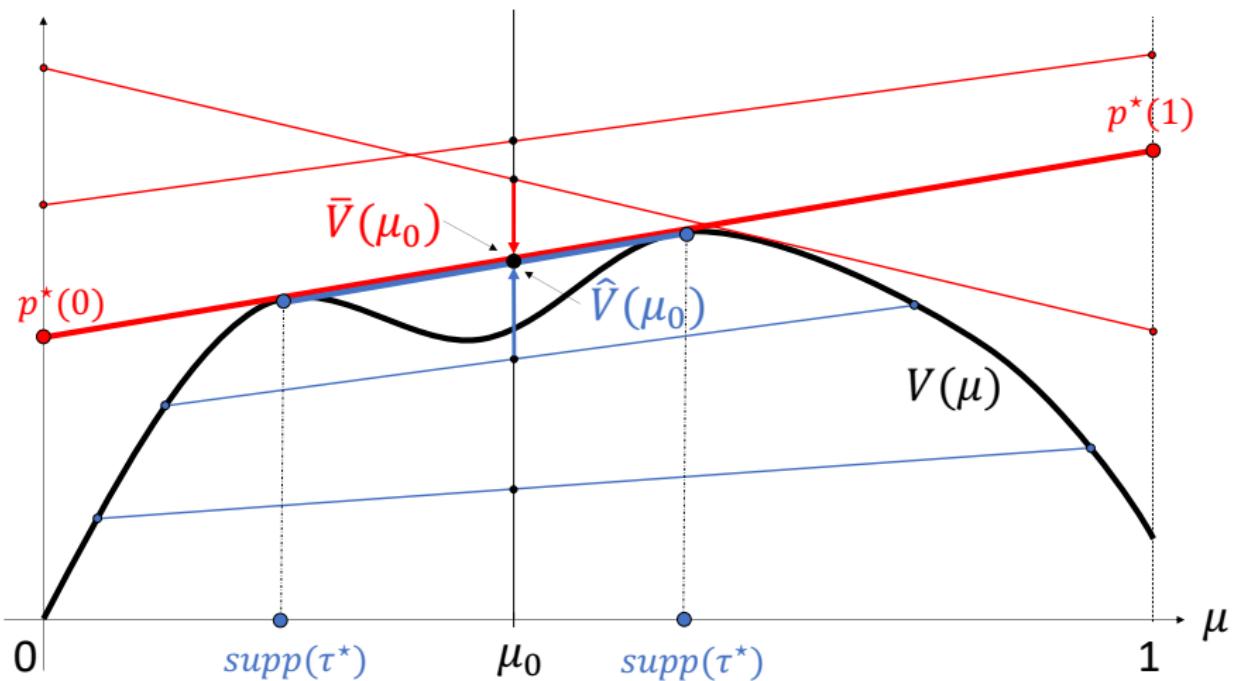
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- Process  $\mu \in \Delta(\Omega)$  operated at unit level consumes measure  $\mu$  of resources and generates income  $V(\mu)$ .
- Production plan  $\tau \in \Delta(\Delta(\Omega))$  describes the level at which each process  $\mu$  is operated.
- The primal problem is to find a production plan  $\tau \in \Delta(\Delta(\Omega))$  that exhausts endowment  $\mu_0$  and maximizes total income.

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- The dual problem is to find feasible prices that minimize the total cost of buying up all the resources.

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1. All resources are used in production:

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$$\int p(\omega) d\mu(\omega) = V(\mu), \quad \text{for all } \mu \in \text{supp}(\tau).$$

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2. Operating processes make zero profits:

$$\int p(\omega) d\mu(\omega) = V(\mu), \quad \text{for all } \mu \in \text{supp}(\tau).$$

3. No entrant can make strictly positive profits:

$$\int p(\omega) d\mu(\omega) \geq V(\mu), \quad \text{for all } \mu \in \Delta(\Omega).$$

## Interpretation as a production problem

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**Weak duality is the first welfare theorem.** If a pair  $(\tau, p)$  is a competitive equilibrium, then  $\tau$  solves Producer's problem, and  $p$  solves Dealer's problem.

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This is because, for any feasible  $\tau$  and  $p$ , we have

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**Strong duality is the second welfare theorem.** Producer's and Dealer's problems admit optimal solutions  $\tau$  and  $p$ . Any such solutions constitute a competitive equilibrium.

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- Assume that  $\text{supp}(\mu_0) = \Omega$ .
- Weak duality yields sufficient optimality conditions.
- Strong duality yields necessary optimality conditions.
- Specifically,  $\tau \in \mathcal{T}(\mu_0)$  is optimal iff there exists  $p \in L(\Omega)$ :

$$\int_{\Omega} p(\omega) d\mu(\omega) \geq V(\mu) \text{ for all } \mu \in \Delta(\Omega),$$

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- **Primal and dual attainment** additionally require existence of solutions to the primal and dual problems, respectively.
- We use the term **strong duality** when there is no duality gap and both primal and dual attainment hold.

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**Optimality criterion.** If  $\tau \in \mathcal{T}(\mu_0)$  and  $p \in \mathcal{P}(V)$  satisfy  $\int V d\tau = \int p d\mu_0$ , then  $\tau$  attains  $\widehat{V}(\mu_0)$  and  $p$  attains  $\overline{V}(\mu_0)$ .

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$$\int V d\tilde{\tau} \leq \int p d\mu_0 = \int V d\tau \quad \text{and} \quad \int \tilde{p} d\mu_0 \geq \int V d\tau = \int p d\mu_0.$$

## Primal attainment

---

**Primal attainment.** The primal problem has an optimal solution.

*Proof:* Follows from the Weierstrass theorem, because we maximize bounded upper semi-continuous function  $\tau \rightarrow \int V d\tau$  on compact set  $\mathcal{T}(\mu_0)$ .  $\mathcal{T}(\mu_0)$  is compact, because it is a closed subset (by continuity of  $\tau \rightarrow \int \mu d\tau$ ) of a compact set  $\Delta(\Delta(\Omega))$ .

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- Let  $\varphi : E \rightarrow \mathbb{R} \cup \{+\infty\}$ , and let  $\varphi^* : E^* \rightarrow \mathbb{R} \cup \{+\infty\}$  be its Legendre transform.
- Fenchel-Moreau Theorem: If  $\varphi : E \rightarrow \mathbb{R} \cup \{+\infty\}$  is convex and lower semi-continuous, then  $\varphi^{**} = \varphi$ .

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*Proof outline continued:*

- Define the function  $\varphi$  on  $M(\Omega)$  as

$$\varphi(\eta) = \begin{cases} -\sup_{\tau \in \mathcal{T}(\eta)} \int_{\Delta(\Omega)} V(\mu) d\tau(\mu), & \eta \in \Delta(\Omega), \\ +\infty, & \eta \notin \Delta(\Omega). \end{cases}$$

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- By the F-M theorem, we get  $\varphi = \varphi^{**}$  on  $\Delta(\Omega)$ .

## Dual attainment

---

$\widehat{V}$  is *superdifferentiable* at  $\mu_0$  if there is a continuous linear function  $H$  on  $M(\Omega)$  (*supporting hyperplane* of  $\widehat{V}$  at  $\mu_0$ ), represented as  $H(\mu) = \int p d\mu$  with  $p \in L(\Omega)$  (*supergradient* of  $\widehat{V}$  at  $\mu_0$ ), such that

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When  $\Omega$  is infinite,

- concavity does not imply continuity on the interior of the domain,
- the set  $\Delta(\Omega)$  has an empty (relative) interior.

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Following Gale (1967), we say that  $\widehat{V}$  has *bounded steepness* at  $\mu_0$  if there exists a constant  $L$  such that

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Intuitively, bounded steepness says that the marginal increase in the value of the persuasion problem is bounded above for a small perturbation of the prior.

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**Dual Attainment.** The following statements are equivalent:

1. The dual problem has an optimal solution.
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*Takeaway:* Duality holds without any extra assumptions in finite state spaces, but additional regularity conditions are needed otherwise.

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- Because we represent the weak<sup>\*</sup> topology using the norm  $\|\cdot\|_{KR}$ , we can use the result of Gale (1967).

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- Remains to prove that existence of solution to the dual problem is equivalent to superdifferentiability of  $\hat{V}$  at the prior  $\mu_0$ .

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$\widehat{V}$  is superdifferentiable at  $\mu_0 \implies$  dual attainment:

- Since  $\widehat{V}$  is superdifferentiable at  $\mu_0$ , there exists a continuous linear function  $H$  on  $M(\Omega)$  such that

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- Main takeaway: The optimal price function  $p$  is a supergradient of the concave closure  $\widehat{V}$  at the prior  $\mu_0$ .

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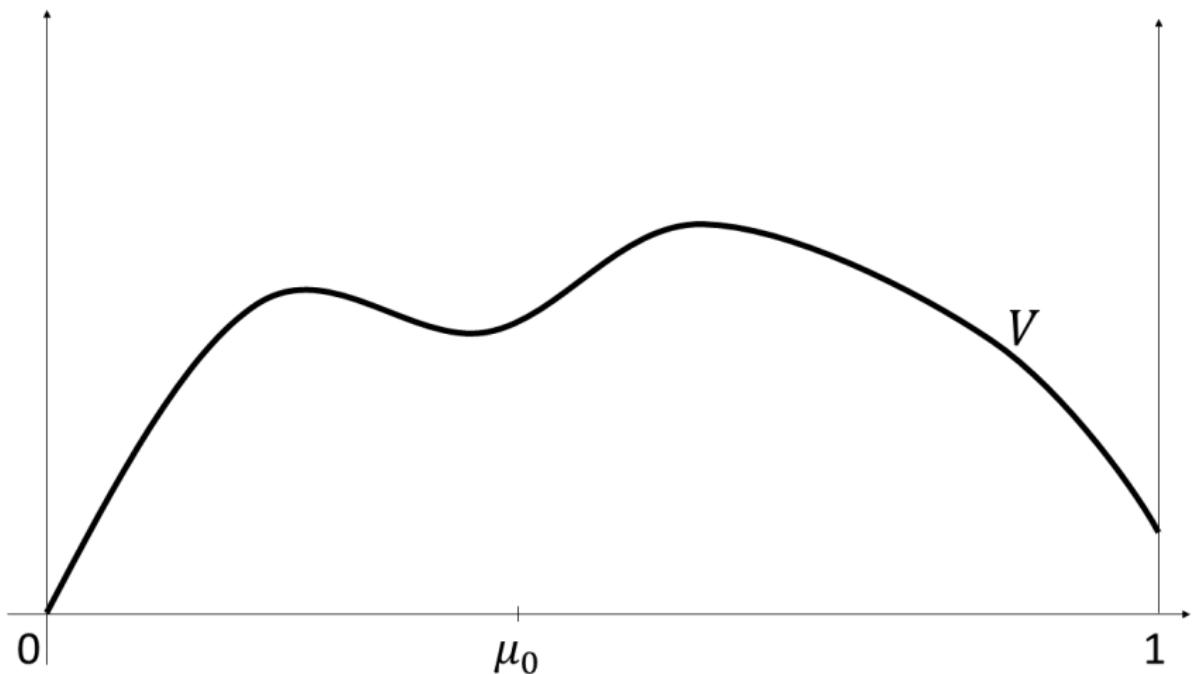
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- Thus,  $p$  is a supergradient of  $\hat{V}$  at  $\mu_0$ .

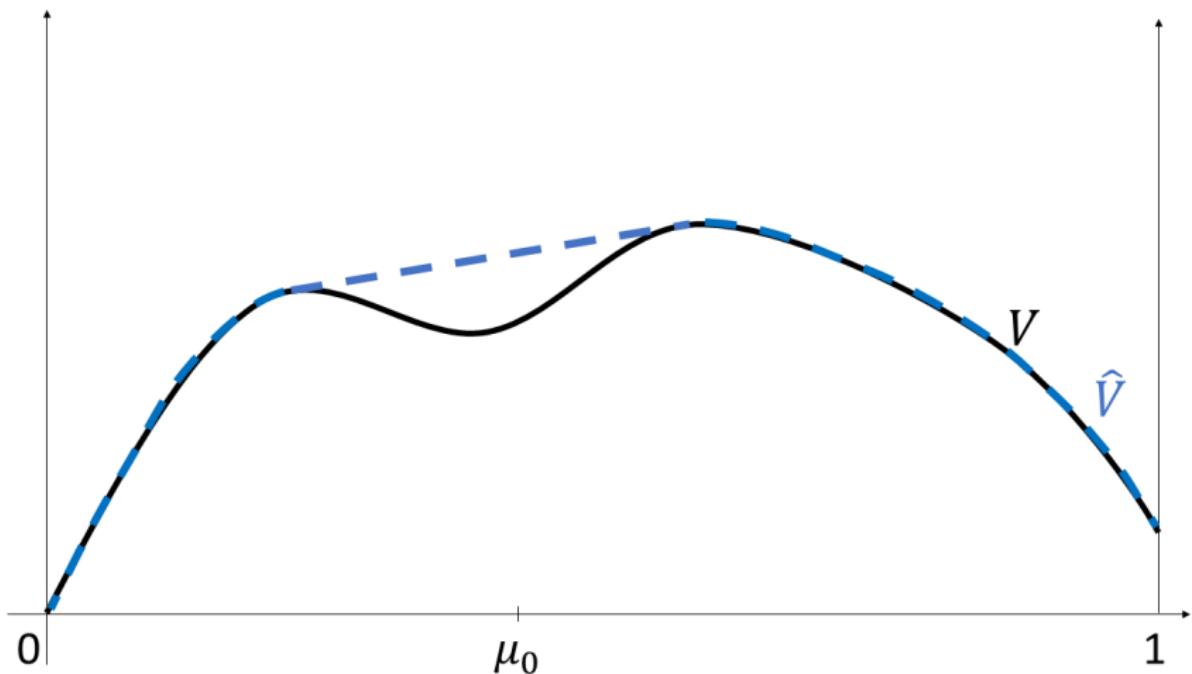
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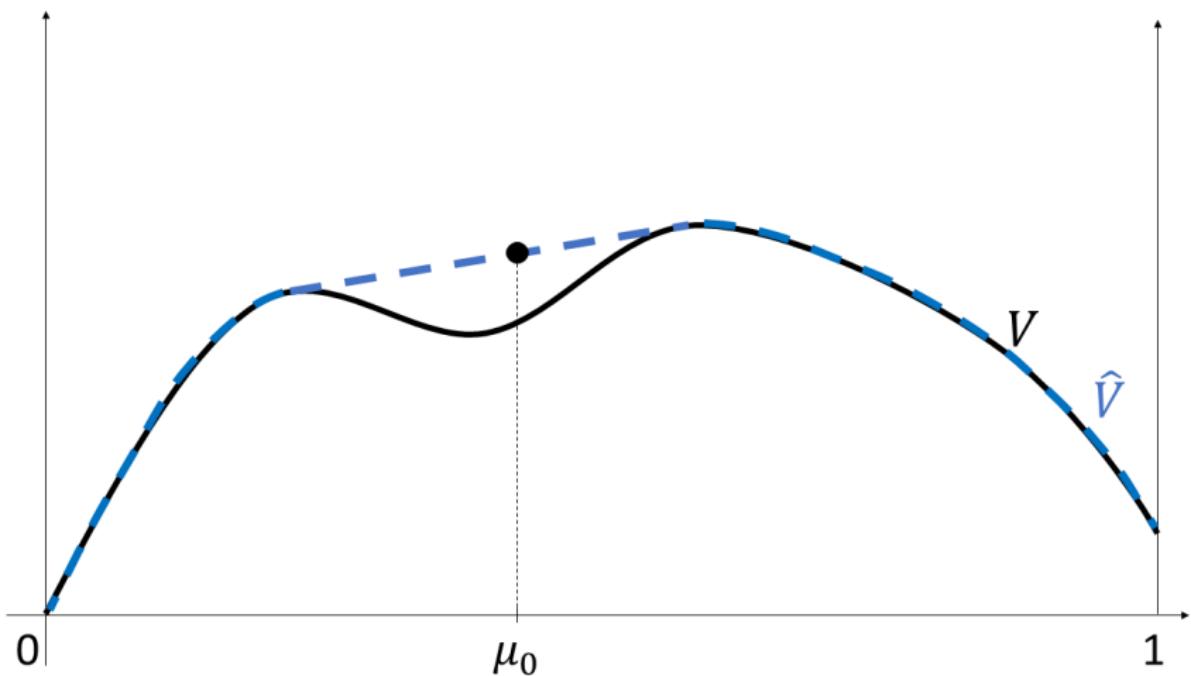
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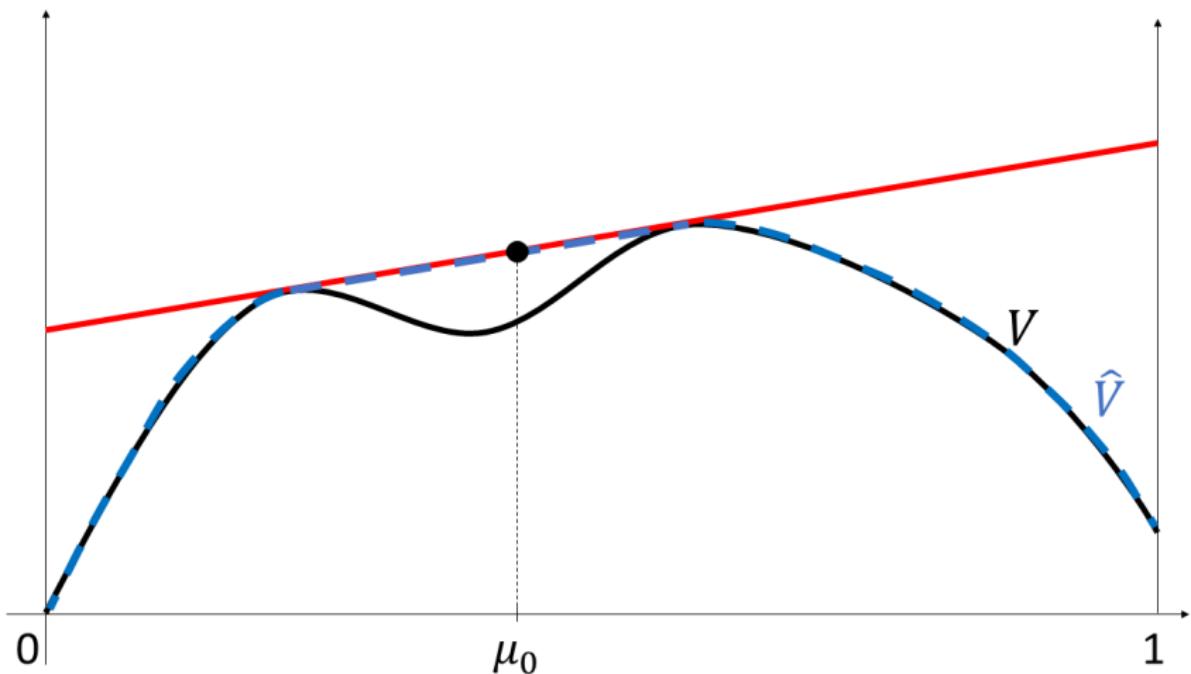
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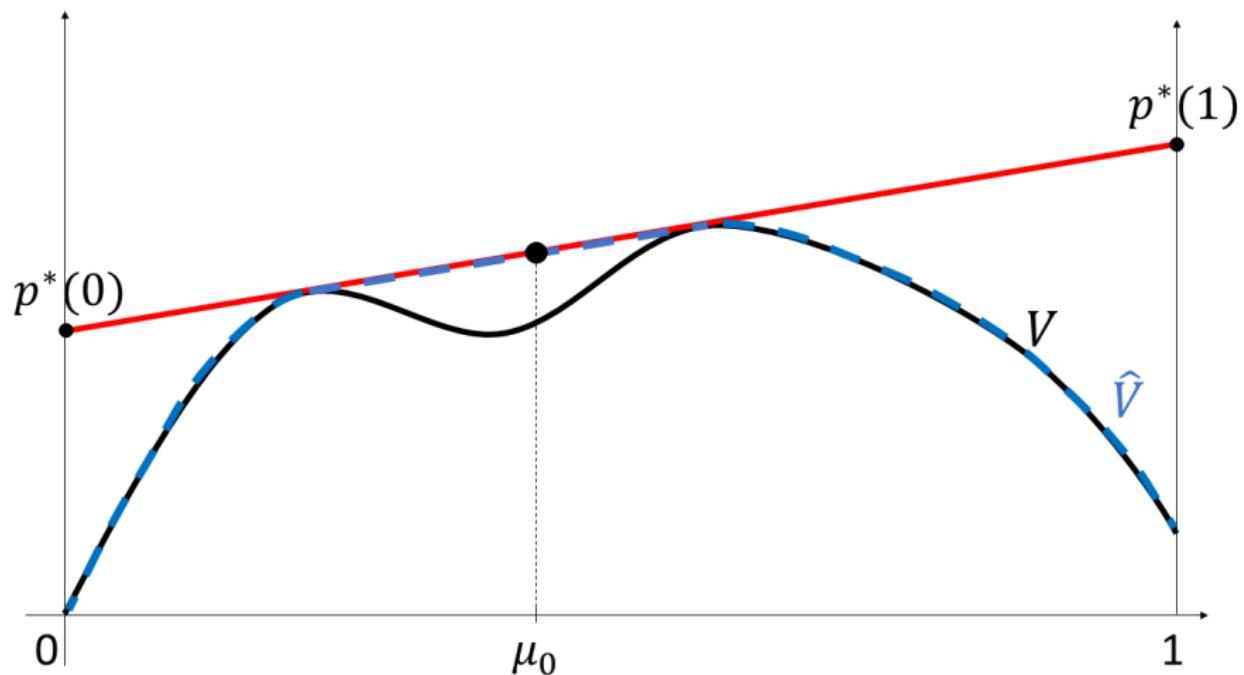
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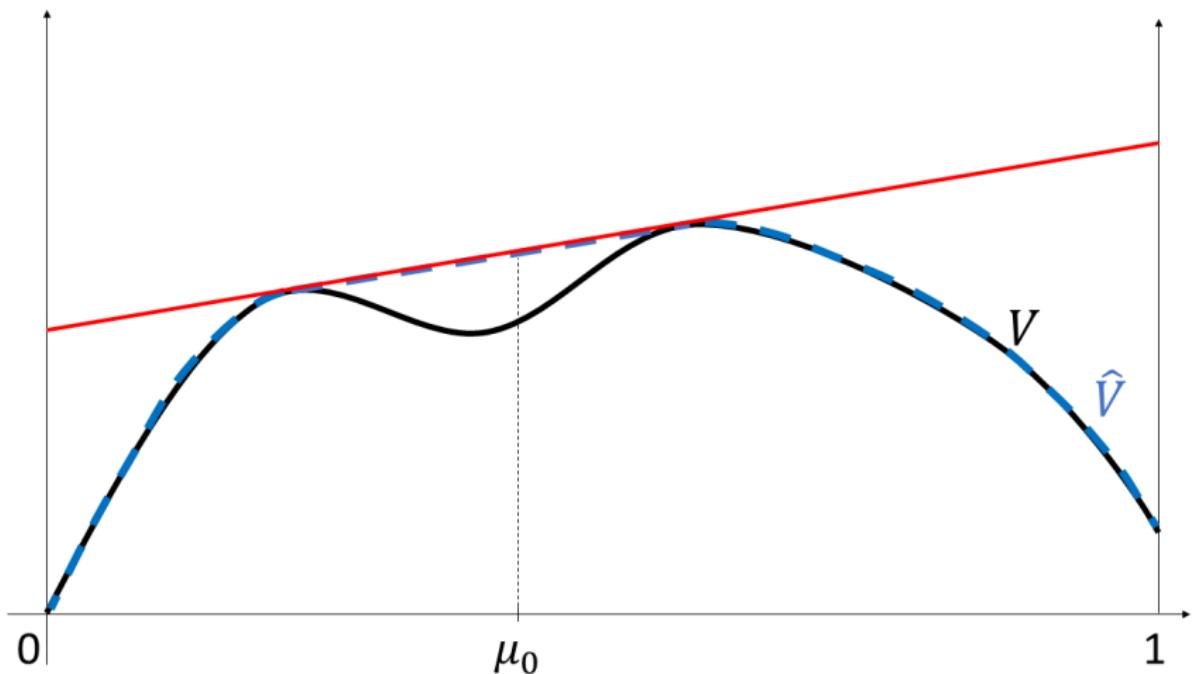
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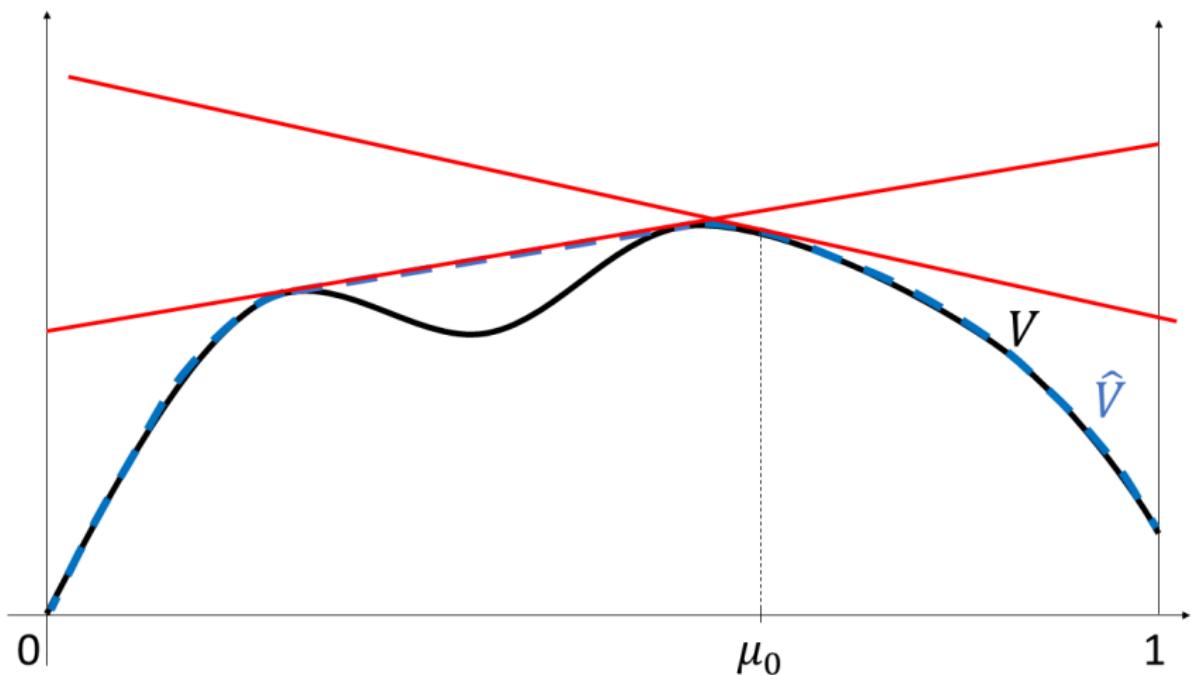
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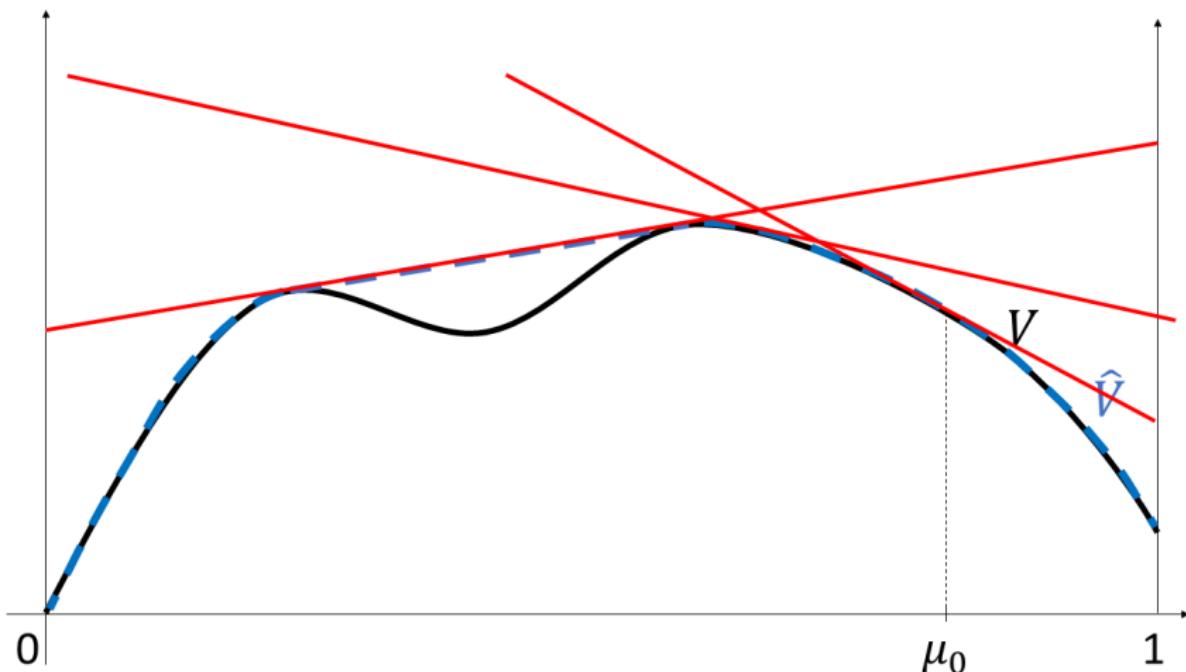
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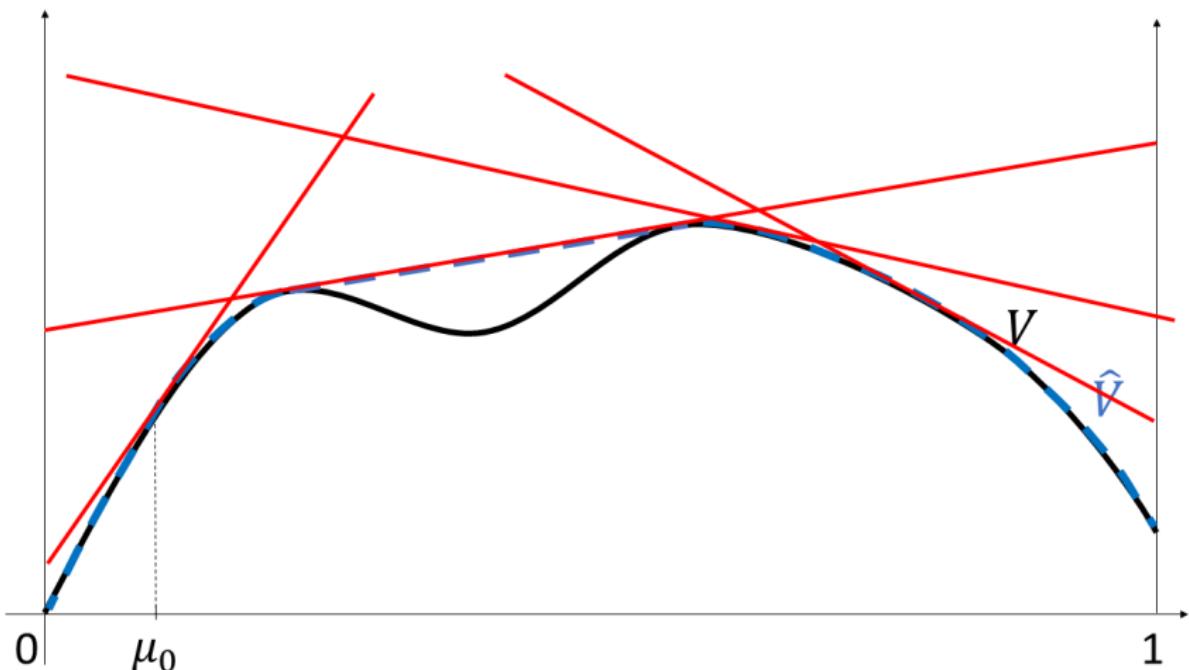
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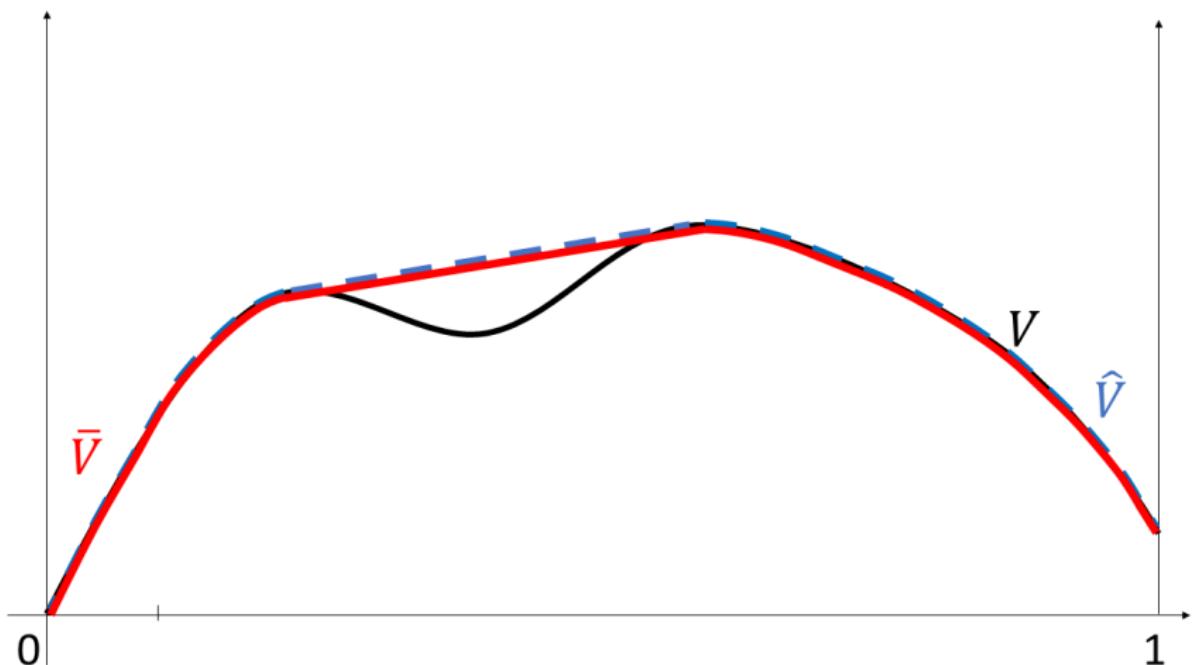
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## Lipschitz Preservation

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Checking bounded steepness of  $\hat{V}$  at  $\mu_0$  can be quite difficult.

## Lipschitz Preservation

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Checking bounded steepness of  $\hat{V}$  at  $\mu_0$  can be quite difficult.

**Lipschitz Preservation.** If  $V$  is Lipschitz on  $\Delta(\Omega)$ , then so is  $\hat{V}$ . Thus,  $\hat{V}$  has bounded steepness at each  $\mu_0 \in \Delta(\Omega)$ , ensuring that the dual problem has an optimal solution.

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## Complementary slackness

---

**Complementary slackness.** Let  $V$  be Lipschitz. Distribution  $\tau \in \mathcal{T}(\mu_0)$  is an optimal solution iff there exists  $p \in \mathcal{P}(V)$  such that

$$V(\mu) = \int_{\Omega} p(\omega) d\mu(\omega), \quad \text{for all } \mu \in \text{supp}(\tau). \quad (\text{C})$$

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*Proof:* The if part follows from weak duality: If (C) holds, then

$$\int_{\Delta(\Omega)} V(\mu) d\tau(\mu) = \int_{\Delta(\Omega)} \int_{\Omega} p(\omega) d\mu(\omega) d\tau(\mu) = \int_{\Omega} p(\omega) d\mu_0(\omega).$$

The only if part follows from no duality gap and dual attainment: If  $\tau \in \mathcal{T}(\mu_0)$  is optimal, then there exists an optimal  $p \in \mathcal{P}(V)$  such that

$$\int_{\Delta(\Omega)} \left( \int_{\Omega} p(\omega) d\mu(\omega) - V(\mu) \right) d\tau(\mu) = 0.$$

# Persuasion and Matching

Anton Kolotilin & Roberto Corrao & Alexander Wolitzky

## Non-linear persuasion

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In *non-linear persuasion* of Kolotilin, Corrao, and Wolitzky (2023), states and actions are one-dimensional but the utilities are not linear in the state.

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- Similar to classical matching or optimal transport, except supply on one side of the market (marginal over actions) is endogenous and determined by receiver's obedience condition.
- Useful for characterizing key properties of optimal signals, or even a unique optimal signal.
- For example, each optimal signal is pairwise under a non-singularity condition on utilities.

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- Strictly concave Receiver's utility ( $u_{aa}(a, \omega) < 0$ ) with interior optimal action  $a^*(\mu)$  given by  $\int_{\Omega} u_a(a^*(\mu), \omega) d\mu(\omega) = 0$  for all  $\mu$ .

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**Lipschitz property.** Indirect utility  $V(\mu) = \int_{\Omega} v(a^*(\mu), \omega) d\mu(\omega)$  is Lipschitz in  $\mu$ , so general duality applies.

## General duality

---

The primal problem is to find  $\tau \in \Delta(\Delta(\Omega))$  to

$$\begin{aligned} & \text{maximize} \int_{\Delta(\Omega)} V(\mu) d\tau(\mu) \\ & \text{subject to} \int_{\Delta(\Omega)} \mu d\tau(\mu) = \mu_0, \end{aligned} \tag{P}$$

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**General duality theorem.** If  $V$  is Lipschitz in the Kantorovich Rubinstein metric, then *weak duality* and *strong duality* hold.

## Action-based primal problem

---

Action-based primal is to find a *joint distribution*  $\pi \in \Delta(A \times \Omega)$  to

$$\text{maximize} \int_{A \times \Omega} v(a, \omega) d\pi(a, \omega)$$

$$\text{subject to } \int_{A \times \tilde{\Omega}} d\pi(a, \omega) = \int_{\tilde{\Omega}} d\mu_0(\omega) \text{ for all } \tilde{\Omega} \subset \Omega, \quad (P')$$

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- The first constraint is the action-based version of Bayes plausibility, which says that the marginal of  $\pi$  on  $\Omega$  equals  $\mu_0$ .
- The second constraint is the obedience constraint, which says that the expected marginal utility equals zero at the recommended action given the belief it induces.

## Action-based dual problem

---

Action-based dual is to find  $p \in L(\Omega)$  and  $q \in B(A)$  to

$$\text{minimize } \int_{\Omega} p(\omega) d\mu_0(\omega) \quad (D')$$

subject to  $p(\omega) \geq v(a, \omega) + q(a)u_a(a, \omega)$  for all  $(a, \omega) \in A \times \Omega$

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*Interpretation:* Price of state  $\omega$  is no less than Sender's value from inducing any action  $a$  at this state, where this value is the sum of

- Sender's utility,  $v(a, \omega)$ , and
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First-order approach is key for dual constraint taking such a simple form.

## Action-based duality

---

**Bridging belief-based and action-based duality.** A price function  $p \in L(\Omega)$  is feasible (optimal) for (D) iff there exists  $q \in B(A)$  such that  $(p, q)$  is feasible (optimal) for (D').

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*Proof outline:* If  $(p, q)$  is feasible for (D'), then, for all  $\mu \in \Delta(\Omega)$ ,

$$\begin{aligned}\int_{\Omega} p(\omega) d\mu(\omega) &\geq \int_{\Omega} (v(a^*(\mu), \omega) + q(a^*(\mu)) u_a(a^*(\mu), \omega)) d\mu(\omega) \\ &= \int_{\Omega} v(a^*(\mu), \omega) d\mu(\omega) = V(\mu),\end{aligned}$$

so  $p$  is feasible for (D).

## Action-based duality

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*Proof outline continued:* If  $p$  is feasible for (D), then, for all  $\omega_1, \omega_2 \in \Omega$  and  $a \in A$  such that  $u_a(a, \omega_1) < 0 < u_a(a, \omega_2)$ , feasibility for

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$$\implies \inf_{\omega_2: u_a(a, \omega_2) > 0} \frac{p(\omega_2) - v(a, \omega_2)}{u_a(a, \omega_2)} \geq \sup_{\omega_1: u_a(a, \omega_1) < 0} \frac{v(a, \omega_1) - p(\omega_1)}{-u_a(a, \omega_1)}$$

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Thus, we can squeeze in  $q(a)$  between the LHS and RHS, so that  $p(\omega) \geq v(a, \omega) + q(a)u_a(a, \omega)$ , yielding feasibility for (D').

## First-order condition for the action-based dual

---

Given  $q \in B(A)$ , it is optimal for (D') to set

$$p(\omega) = \sup_{a \in A} \{v(a, \omega) + q(a)u_a(a, \omega)\}, \text{ for all } \omega \in \Omega.$$

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Strong duality implies that for all  $\mu$  in the support of an optimal  $\tau$  and for all  $\omega$  in the support of  $\mu$ , the first-order condition holds:

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and thus,

$$q(a^*(\mu)) = \frac{\mathbb{E}_\mu[v_a(a^*(\mu), \omega)]}{-\mathbb{E}_\mu[u_{aa}(a^*(\mu), \omega)]}.$$

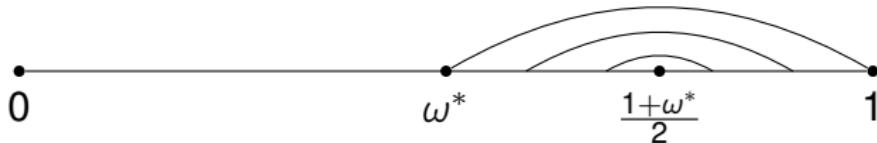
# Pairwise Signals

---

- A signal  $\tau$  is pairwise if each induced posterior  $\mu \in \text{supp}(\tau)$  has at most binary support ( $|\text{supp}(\mu)| \leq 2$ ):

$$\mu = (1 - \lambda)\delta_\omega + \lambda\delta_{\omega'}, \quad \text{for } \omega \leq \omega' \text{ and } \lambda \in [0, 1].$$

- For example, for uniform  $\mu_0$  and any  $\omega^*$ , the signal that reveals  $\omega$  if  $\omega < \omega^*$ , and reveals that the state is either  $\omega$  or  $1 + \omega^* - \omega$  with equal probability if  $\omega \in [\omega^*, \frac{1+\omega^*}{2}]$ , is pairwise.



- Full disclosure ( $\text{supp}(\tau_F) = \{\delta_\omega : \omega \in \Omega\}$ ) is pairwise.
- No disclosure ( $\text{supp}(\tau_N) = \{\mu_0\}$ ) is not pairwise.

# Optimality of pairwise signals

---

Every optimal signal is pairwise if for all  $a$  and  $\omega_1 < \omega_2 < \omega_3$ , we have

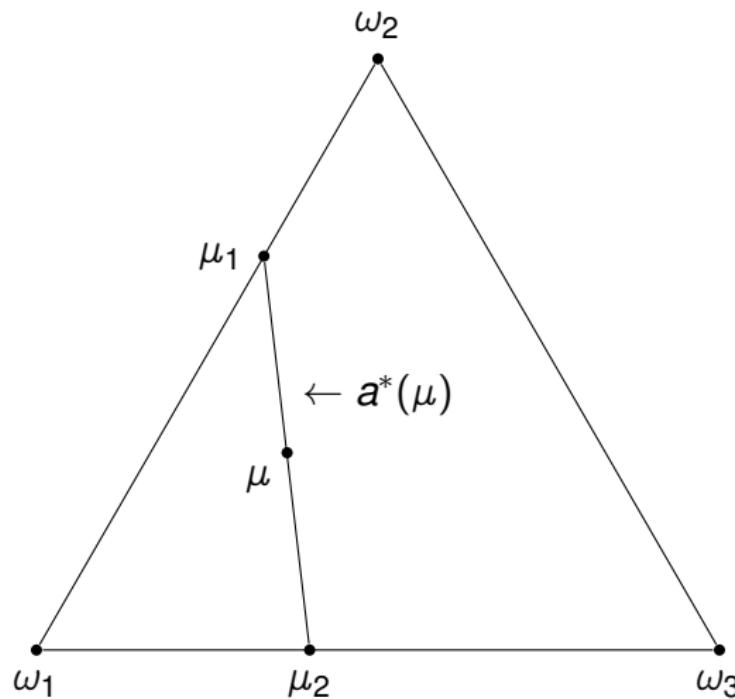
$$\det \begin{pmatrix} v_a(a, \omega_1) & v_a(a, \omega_2) & v_a(a, \omega_3) \\ u_a(a, \omega_1) & u_a(a, \omega_2) & u_a(a, \omega_3) \\ u_{aa}(a, \theta_1) & u_{aa}(a, \omega_2) & u_{aa}(a, \omega_3) \end{pmatrix} \neq 0. \quad (N)$$

- Follows from the strong-duality FOC, as there do not exist an action  $a$ , three states  $\omega_1 < \omega_2 < \omega_3$ , and a vector  $(q(a), q'(a))$ :

$$v_a(a, \omega_i) + q(a)u_{aa}(a, \omega_i) + q'(a)u_a(a, \omega_i), \text{ for } i = 1, 2, 3.$$

- (N) holds, for example, if  $u(a, \omega) = -(a - \omega)^2$  and  $v(a, \omega) = aw(\omega)$  with strictly convex or concave  $w$ .
- (N) fails in the linear case where  $U_a$  and  $V_a$  are linear in  $\theta$  (so pooling intervals of states can be optimal in the linear case).

## Intuition



## Linear persuasion

---

We now assume that  $V(\mu) = v(\mathbb{E}_\mu[\omega])$  for all  $\mu \in \Delta([0, 1])$ .

This is a special case of non-linear persuasion with  $u(a, \omega) = -(a - \omega)^2$  and state-independent  $v$ .

Only the distribution  $\eta$  of posterior means matters, where  $\eta$  is the  $A$ -marginal distribution of  $\pi$ :  $\eta(\tilde{A}) = \pi(\tilde{A}, \Omega)$  for all  $\tilde{A} \subset A$ .

## Primal and Dual

---

The primal problem simplifies to finding  $\eta \in \Delta([0, 1])$  to

$$\begin{aligned} & \text{maximize} \int v(a) d\eta(a) \\ & \text{subject to } \eta \in MPC(\mu_0). \end{aligned} \tag{P''}$$

The dual problem simplifies to finding  $p \in L(\Omega)$  to

$$\begin{aligned} & \text{minimize} \int p(\omega) d\mu_0(\omega) \\ & \text{subject to } p \text{ is convex and } p \geq v. \end{aligned} \tag{D''}$$