

# *Flexible Moral Hazard Problems*

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# OVERVIEW

Classic moral hazard model:

- Effort is either binary, or belongs to an interval.
- Main result: contracts are motivated by informativeness.
- Consequently, contracts are monotone only under MLRP.

Current paper:

- Allow agent to choose *any* output distribution.
- Contracts pinned down by an output-by-output FOC.
- Monotone costs  $\implies$  monotone contracts.
- In particular: Informativeness plays no role.

# Two Examples

## COMMON SETUP FOR EXAMPLES

A principal (she) contracts with an agent (he).

- Compact set  $X \subset \mathbb{R}$  of possible outputs.
- Principal offers agent a (bounded) contract:  $w : X \rightarrow \mathbb{R}$ .
- Agent can opt out and get  $u_0$ .
- If opts in, agent covertly chooses  $\alpha \in \mathcal{A} \subseteq \Delta(X)$ .
- Effort costs:  $C : \mathcal{A} \rightarrow \mathbb{R}_+$ , continuous, increasing in FOSD.
- Payoffs:

$$\text{Principal: } x - w \quad \text{Agent: } u(w) - C(\alpha).$$

$u$ : strictly increasing, differentiable, unbounded, concave.

# STANDARD BINARY EFFORT MODEL

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So:  $w$  is monotone  $\iff$  MLRP holds.

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Implications:

- Cost minimization is trivial:  $\min w(L)$  s.t. IR.
- IC contracts are monotone:

$$w(H) = u^{-1}(u \circ w(L) + C'(\alpha)) \geq u^{-1}(u \circ w(L)) = w(L).$$

Model



## OUR MODEL

A principal (she) contracts with an agent (he).

- Compact set  $X \subset \mathbb{R}$  of possible outputs.
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- **Limited liability:**  $w(\cdot) \geq 0$ .
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**$u$ :** increasing, continuous, unbounded &  $u(0) = 0$ .

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$$\lim_{\epsilon \downarrow 0} \frac{1}{\epsilon} [C(\alpha + \epsilon(\beta - \alpha)) - C(\alpha)] = \int k_\alpha(x) (\beta - \alpha)(dx)$$

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(if  $X$  is finite:  $C$  smooth  $\iff$  differentiable, which holds a.e.)

# FIRST-ORDER APPROACH

**Lemma.** For a bounded  $v : X \rightarrow \mathbb{R}$ , and  $\alpha \in \mathcal{A}$ ,

$$\alpha \in \arg \max_{\beta \in \mathcal{A}} \left[ \int v(x) \beta(dx) - C(\beta) \right]$$

if and only if

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(the “only if” direction also works if  $C$  is not convex)

# RELATIONSHIP TO STANDARD FOC

Consider the problem:

$$\max_{x \in [0,1]} [xv - c(x)]$$

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Standard way of writing FOC for optimal  $x^* \in (0, 1)$  is

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An equivalent way of writing the above condition is:

$$x^* \in \operatorname{argmax}_{x \in [0,1]} [xv - xc'(x^*)].$$

The lemma generalizes the second formulation.

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## PROOF: NECESSITY

Fix  $\beta \in \mathcal{A}$ , and for every  $\epsilon \in [0, 1]$ , define  $\beta_\epsilon := \alpha + \epsilon(\beta - \alpha)$ .

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$$0 \geq \frac{1}{\epsilon} \left[ \int v(x)(\beta_\epsilon - \alpha)(dx) \right] - \frac{1}{\epsilon} [C(\beta_\epsilon) - C(\alpha)]$$

(because  $\alpha \in \operatorname{argmax}_{\beta \in \mathcal{A}} [\int v(x)\beta(dx) - C(\beta)]$ )

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(by Gateaux differentiability)

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- Agent covertly chooses  $\alpha \in \mathcal{A} = \Delta(X)$ .
- Effort costs  $C : \mathcal{A} \rightarrow \mathbb{R}_+$ : convex, increasing, smooth.
- Limited liability:  $w \geq 0$ .
- Feasible contracts:  $W = \{w : X \rightarrow \mathbb{R}_+ : \text{bounded}\}$ .
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# CHARACTERIZATION OF IC

A contract-distribution pair  $(w, \alpha) \in W \times \mathcal{A}$  is **IC** if and only if

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**Proposition.**  $(w, \alpha)$  is IC if and only if a  $m \in \mathbb{R}$  exists such that

$$w(x) \leq u^{-1}(k_\alpha(x) + m)$$

for all  $x$ , and with equality  $\alpha$ -almost surely.

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- (ii) Cheapest contract implement  $\alpha$  is  $w_{m_\alpha^*,\alpha}$  for

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- (iii) Every  $\alpha$  can be implemented with a monotone contract (since  $C$  is FOSD monotone,  $k_\alpha$  is increasing).

# Profit Maximization

# THE PRINCIPAL'S PROBLEM

Let  $w_\alpha := w_{m_\alpha^*, \alpha}$  be the cost minimizing wage implementing  $\alpha$ .

The principal's problem is:

$$\max_{\alpha \in \mathcal{A}} \left[ \int x \alpha(dx) - \int w_\alpha(x) \alpha(dx) \right].$$



## ADDITIONAL ASSUMPTIONS

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**Nice Agent's Payoffs.**  $u$  is continuously differentiable, and  $u' > 0$ .

**Continuous Derivative.** The mapping  $\alpha \mapsto k_\alpha$  is weak\*-supnorm continuous.

**2nd Order Differentiability.** Every  $\alpha$  admits a continuous function  $h_\alpha : X \times X \rightarrow \mathbb{R}$  such that for

$$\lim_{\epsilon \rightarrow 0} \frac{1}{\epsilon} [k_{\alpha + \epsilon(\beta - \alpha)}(\cdot) - k_\alpha(\cdot)] = \int h(\cdot, y)(\beta - \alpha)(dy),$$

where convergence is in the supnorm.

(for finite  $X$ : equivalent to twice differentiability).

# PRINCIPAL FIRST ORDER CONDITION

Define the function:

$$\chi_{\alpha}(x) = \int \frac{h_{\alpha}(x, y)}{u' \circ w_{\alpha}(y)} \alpha(dy).$$

**Theorem.**

A profit maximizing  $\alpha^*$  exists. Moreover,  $\alpha^*$  must solve

$$\max_{\alpha \in \mathcal{A}} \int [x - w_{\alpha^*}(x) - \chi_{\alpha^*}(x)] \alpha(dx).$$

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For every  $\alpha$ , let

$$\pi_\alpha(x) := x - w_\alpha(x) - \chi_\alpha(x).$$

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- (ii) If  $\pi_\alpha$  is strictly *quasiconvex*  $\forall \alpha$ , then  $\text{supp } \alpha^* \subseteq \{L, H\}$ .



# NICE PROFIT MAXIMIZING DISTRIBUTIONS

For every  $\alpha$ , let

$$\pi_\alpha(x) := x - w_\alpha(x) - \chi_\alpha(x).$$

**Corollary.** Suppose  $X = [L, H]$  and  $\alpha^*$  maximizes profits. Then,

- (i) If  $\pi_\alpha$  is strictly *quasiconcave*  $\forall \alpha$ , then  $|\text{supp } \alpha^*| = 1$ .
- (ii) If  $\pi_\alpha$  is strictly *quasiconvex*  $\forall \alpha$ , then  $\text{supp } \alpha^* \subseteq \{L, H\}$ .
- (iii) If  $w_\alpha + \chi_\alpha$  is a non-affine & analytic  $\forall \alpha$ ,  $\alpha^*$  is discrete.

# FLEXIBLE MORAL HAZARD PROBLEMS

We showed that in flexible moral hazard problems:

- Incentive compatibility pins down contract.
- Cost minimization is trivial.
- Every distribution can be implemented.
- Wages are monotone without loss.

# FLEXIBLE MORAL HAZARD PROBLEMS

We showed that in flexible moral hazard problems:

- Incentive compatibility pins down contract.
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- Every distribution can be implemented.
- Wages are monotone without loss.

Also obtained results about principal optimality.

- First order approach is valid.
- Optimality of single, binary, and discrete distributions.

## RELATED LITERATURE

- Flexible models with specific functional forms:
  - CARA utility, monetary effort costs, finite  $X$ : Holmstrom and Milgrom (1987).
  - Mean-measurable costs: Diamond (1998), Barron, Georgiadis, and Swinkels (2020).
  - $f$ -Divergence costs, finite  $X$ : Hebert (2018), Bonham (2021), Mattsson and Weibull (2022), Bonham and Riggs-Cragun (2023).
- Flexible Monitoring: Georgiadis and Szentes (2020), Mahzoon, Shourideh, and Zetlin-Joines (2022), Wong (2023).
- Robust contracting: Carroll (2015), Antic (2022), Antic and Georgiadis (2022), Carroll and Walton (2022).

Thanks!