

# Optimal Project Design

George Georgiadis (Northwestern Kellogg)

*with D. Garrett (TSE), A. Smolin (TSE), and B. Szentes (LSE)*

# Motivation

- Rents due to agency problems is key determinant of economic welfare
- Determinants of these frictions are usually part of model description
  - In adverse selection models, distribution of types typically exogenous
  - In moral hazard models, production technology taken as given
- If an agent's payoff depends on agency frictions, then he is likely to take actions to generate these frictions optimally.

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Revisit standard principal-agent model under moral hazard to understand how an agent might gain by designing the production technology optimally.

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# Model

- *Players.* Risk-neutral principal & agent, and latter is cash-constrained
- *Timing.*
  - i. Agent chooses a “project”  $c : \Delta([0, 1]) \rightarrow \mathbb{R}_+$ ; *i.e.*, a map from every output distribution with support on  $[0, 1]$  to a (nonnegative) cost.
  - ii. Principal offers a wage scheme  $w : [0, 1] \rightarrow \mathbb{R}_+$
  - iii. Agent chooses an “action”  $F \in \Delta([0, 1])$
  - iv. Output  $x \sim F$  and payoffs are realized
- *Payoffs.*
  - Agent:  $\mathbb{E}_F[w(x)] - c(F)$
  - Principal:  $\mathbb{E}_F[x - w(x)]$
  - Both players have outside option 0

# Applications

- An entrepreneur (agent) seeks funding from a VC (principal)
- Before contracting, the entrepreneur must develop a business plan, specifying various aspects of his production function
- Conceivable he has at least some flexibility in choosing the biz plan.
- If VC has a lot of bargaining power, the entrepreneur benefits from putting forward a biz plan that exacerbates moral hazard problem.
- *Remark:* Abstract away from constraints in the agent's flexibility.
- More broadly, employees can often influence aspects of production function (e.g., assignment of projects, goals, evaluation metrics, etc), which provides an opportunity to shape their production technology.

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## Some Intuition

- *First Best.*
  - Agent sets  $c(F) = 0$  for all  $F$
  - Principal responds by offering wage 0 and implementing  $F(x) = \mathbb{I}_{\{x=1\}}$
- Outcome is efficient but the agent is left with no rents!
- *Mechanism.* Agent chooses the project to make the moral hazard problem severe, which will enable him to extract rents.



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## Problem Formulation

- *Principal.* Given project  $c$ , she solves:

$$\begin{aligned}
 & \max_{w(\cdot), F} \mathbb{E}_F[x - w(x)] \\
 & \text{s.t. } \mathbb{E}_F[w(x)] - c(F) \geq \mathbb{E}_{\tilde{F}}[w(x)] - c(\tilde{F}) \quad \text{for all } \tilde{F} \\
 & \quad w(x) \geq 0 \text{ for all } x \\
 & \quad F \in \Delta([0, 1])
 \end{aligned}$$

Denote the optimal contract by  $w^c$  and implemented action by  $F^c$ .

- *Agent.* Chooses the optimal project by solving:

$$\begin{aligned}
 & \max c(F^c) - \mathbb{E}_{F^c}[w^c(x)] \\
 & \text{s.t. } c : \Delta([0, 1]) \rightarrow \mathbb{R}_+
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# Main Results

- 1 Optimal project is *coarse*: all feasible actions generate binary output
  - Binary projects effectively restrict the contracting space, making it more expensive for the principal to motivate the agent.
- 2 Action space is *rich*: Optimal (binary) project comprises
  - continuum of zero-cost actions where project succeeds with some prob.
  - a high cost action which guarantees success
  - a spectrum of actions in between.
- 3 *Inefficiency*: Maximal output realized in equilibrium at bloated costs
- 4 *Rents*: The agent extracts all rents
- 5 Characterization of payoff allocations for any production technology

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## A Simple Example

- Suppose the agent is restricted to choosing a project comprising two actions,  $F_L$  and  $F_H$ , with binary output; *i.e.*,  $\text{supp}(F_i) = \{0, 1\}$
- Easy to solve analytically and show that:
  - $F_L$  costs 0 and leads to  $x = 1$  with probability  $1/2$  (otherwise  $x = 0$ )
  - $F_H$  costs  $1/4$  and leads to  $x = 1$  with probability 1
  - Principal sets  $w(0) = 0$  and  $w(1) = 1/2$ , implementing  $F_H$
- *Remarks:*
  - Clearly,  $c(F_L) = 0$ : otherwise, agent can uniformly decrease costs
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## Simple Example Continued

- Can the agent benefit from choosing a 3<sup>rd</sup> action?

**YES!**

- In the optimal project:
  - $F_i$  leads to  $x = 1$  w.p  $p_i$ , where  $p_L < p_M < p_H$  &  $c(F_L) < c(F_M) < c(F_H)$
  - Principal implements  $F_H$ , wherein  $x = 1$  with probability 1
- Conditional on implementing  $F_H$ , intermediate action  $F_M$  is useful for the agent because it determines the optimal bonus.
- $F_L$  determines if implementing  $F_H$  is optimal for principal.
  - Absent this action,  $F_M$  would be implementable with bonus =  $c(F_M)$ , which could be preferable for the principal (reducing rents to 0).
- *Actions support each other*, enabling agent to extract rents

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# Plan of Attack

- *Theorem 1:* Show it suffices to restrict attention to binary projects
  - Given an arbitrary project, we construct a new project such that  $c(F) < 1$  iff  $\text{supp}(F) = \{0, 1\}$ , and the agent is (weakly) better off.
- This dramatically reduces the dimensionality of the problem so that:
  - In Stage 1, the agent assigns a cost  $C(p) \geq 0$  to each  $p = \Pr\{x = 1\}$
  - In Stage 2, the principal offers a bonus contract  $w(x) = b\mathbb{I}_{\{x=1\}}$
  - In Stage 3, agent chooses  $p$  at a cost  $C(p)$
- *Theorem 2:* Characterize the optimal project (in closed form)



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# Properties of an Optimal Project

## Theorem 1.

- For any project  $c$ , there exists another project,  $\tilde{c}$ , such that
  - i.  $\tilde{c}(F) < 1$  if and only if  $\text{supp}(F) = \{0, 1\}$  (i.e., output is binary), and
  - ii. the principal optimally implements  $F(x) = \mathbb{I}_{\{x=1\}}$  (i.e.,  $x = 1$  w.p 1),

which gives the agent a (weakly) larger expected payoff.

- The principal optimally rewards those outputs which are indicative of the target action, and punishes those indicative of a deviation.
- Binary projects restrict the contracting space, limiting the principal's *screening* ability, and increasing the expected payment to the agent.

## Binary Projects: Proof

- Fix a  $c$  & suppose principal offers  $w^*$ , implementing  $F^*$  (w/ mean  $\mu^*$ )
- Construct a new project  $\tilde{c}$ : For each  $\mu \in [0, 1]$ , define

$$B_\mu = (1 - \mu) + \mu \mathbb{I}_{\{x=1\}} \quad \text{and}$$

$$\tilde{c}(B_\mu) = \inf \{c(F) : \mathbb{E}_F[x] = \mu\}$$

*i.e.*,  $B_\mu$  is a distribution with support  $\{0, 1\}$  and mean  $\mu$ , and we assign it the cost of the cheapest distribution in  $c$  with same mean.

- Given  $\tilde{c}$ , wolog, the principal offers a *bonus contract*  $w(x) = b \mathbb{I}_{\{x=1\}}$ , or equivalently, a *linear contract*  $w(x) = bx$ .

## Binary Projects: Proof

- Consider the problem of implementing any action at max profit

$$\Pi(F) = \sup_{w(\cdot) \geq 0} \{ \mathbb{E}_F[x - w(x)] : F \text{ is IC} \}, \text{ and}$$

$$\tilde{\Pi}(B_\mu) = \sup_{b \in [0,1]} \{ (1-b)\mu : B_\mu \text{ is IC} \},$$

in the original and the new project,  $c$  and  $\tilde{c}$ , respectively.

- Lemma 1:* For any  $F$  such that  $\mathbb{E}_F[x] = \mu$ ,  $\tilde{\Pi}(B_\mu) \leq \Pi(F)$ .  
i.e., implementing  $B_\mu$  is less profitable than an  $F$  with same mean.

- Suppose the principal were restricted to linear contracts in  $c$ . Then:

$$\Pi_{lin}(F) = \tilde{\Pi}(B_\mu) \quad \text{for all } F \text{ with mean } \mu.$$

- Absent this restriction, her profit is weakly larger; i.e.,  $\Pi(F) \geq \Pi_{lin}(F)$ .

# Binary Projects: Proof

- Define  $B^* = B_{\mu^*}$  and  $b^* = \mathbb{E}_{F^*}[w^*(x)]/\mu^* < 1$
- If  $w(x) = b^* \mathbb{I}_{\{x=1\}}$  implements  $B^*$ , then:
  - 1 It makes the same expected payment to the agent as  $w^*$ .
  - 2 It generates profit equal to  $\Pi(F^*)$  for the principal.
- If  $b^*$  does *not* implement  $B^*$ , adjust cost  $\tilde{c}(B^*) = \inf_{\mu} \{b^* \mu - c(B_{\mu})\}$
- *Lemma 2*: Principal cannot implement  $B^*$  with any  $b < b^*$ .
  - Suppose  $B^*$  can be implemented by some  $b < b^*$
  - If  $\tilde{c}(B^*)$  was adjusted, this contradicts the above definition of  $\tilde{c}(B^*)$ .
  - If  $\tilde{c}(B^*)$  was not, then the premise contradicts Lemma 1.

## Binary Projects: Proof

- By assumption,  $F^*$  is optimal in  $c$ ; *i.e.*,  $\Pi(F^*) \geq \Pi(F)$  for all  $F$
- By Lemma 1,  $\tilde{\Pi}(B_\mu) \leq \Pi(F)$  for any  $F$  with mean  $\mu$
- By construction,  $\tilde{\Pi}(B^*) = \Pi(F^*)$ , and therefore,

$$\tilde{\Pi}(B^*) \geq \tilde{\Pi}(B_\mu) \quad \text{for all } \mu$$

*i.e.*, the principal optimally implements  $B^*$  in  $\tilde{c}$ .

- Also by construction, agent is weakly better off relative to  $\{c, w^*\}$ .
- If  $\mu^* = 1$ , then the proof is complete.

## Binary Projects: Proof

- Suppose  $\mu^* < 1$ . Since  $b^*$  implements  $B^*$ , the following IC is satisfied

$$b^* \mu^* - \tilde{c}(B^*) \geq b^* \mu - \tilde{c}(B_\mu) \quad \text{for all } \mu.$$

- *Observation:* This constraint is slack for all  $\mu > \mu^*$ .
  - If not,  $b^*$  implements  $B_{\mu'}$  for some  $\mu' > \mu^*$  giving principal bigger profit
- Therefore, wolog, we can adjust  $\tilde{c}(B_\mu) = \infty$  for all  $\mu > \mu^*$ .
- Multiply bonus  $b^*$ , costs and success prob.  $\Pr\{x = 1\}$  by  $1/\mu^* > 1$ .
  - Payoffs are scaled up and IC constraints are unchanged.
- **Summary:** New project comprises only actions with support  $\{0, 1\}$ , principal optimally implements  $x = 1$  w.p. 1, and agent is better off.

# Implication

- By Theorem 1, it suffices to restrict attention to:
  - Actions such that

$$x = \begin{cases} 1 & \text{with probability } p \\ 0 & \text{with probability } 1 - p \end{cases}$$

- Cost function  $C(p) \geq 0$  such that principal optimally implements  $p = 1$
  - Bonus contracts  $w(x) = b\mathbb{I}_{\{x=1\}}$  for some  $b \geq 0$  to be chosen.
- We will solve the problem using backward induction



## Heuristic Characterization – Stage 2

- Fix a cost function  $C(\cdot)$ . Then the principal solves

$$\begin{aligned} \max \quad & p(1 - b) \\ \text{s.t.} \quad & pb - C(p) \geq \tilde{p}b - C(\tilde{p}) \quad \text{for all } \tilde{p} \in [0, 1] \\ & p \in [0, 1] \quad \text{and} \quad b \geq 0 \end{aligned}$$

- Guess that  $C$  is twice differentiable and convex. Then we can replace the agent's IC constraint with its first-order condition:

$$b = C'(p)$$

and rewrite the principal's problem as

$$\pi := \max_p p [1 - C'(p)]$$

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# Heuristic Characterization – Stage 1

- The agent solves

$$\begin{aligned} \max_{C(\cdot) \geq 0} \quad & p^* b - C(p^*) \\ \text{s.t.} \quad & p^* [1 - C'(p^*)] \geq p [1 - C'(p)] \quad \text{for all } p \quad (\text{IC}_P) \end{aligned}$$

where  $p^* = 1$  by Theorem 1, and  $b = C'(p^*)$  from the agent's FOC.

- Using that  $C'(1) = 1 - \pi$ , we can rewrite this maximization program as

$$\begin{aligned} \max \quad & 1 - \pi - \int_0^1 C'(q) dq \\ \text{s.t.} \quad & C'(p) \geq 1 - \frac{\pi}{p} \quad \text{for all } p < 1 \\ & C(\cdot) \geq 0 \quad \text{and} \quad \pi \in [0, 1] \end{aligned}$$

## Heuristic Characterization – Stage 1 (Continued)

- *Step 1:* For (any) fixed  $\pi$ , we solve

$$\begin{aligned} \max_{C(\cdot) \geq 0} \quad & 1 - \pi - \int_0^1 C'(p) dp \\ \text{s.t.} \quad & C'(p) \geq 1 - \frac{\pi}{p} \quad \text{for all } p < 1 \end{aligned}$$

- Objective decreases in  $C'(p)$  and constraint imposes lower bound. So

$$C'(p) = \left[ 1 - \frac{\pi}{p} \right]^+$$

- *Step 2:* Plugging  $C'(\cdot)$  into the agent's objective, we solve

$$\max_{\pi \in [0,1]} \{-\pi \ln \pi\} = \frac{1}{e} \quad \text{and} \quad \pi^* = \frac{1}{e};$$

*i.e.*, the principal's, as well as the agent's payoff is  $1/e$ .

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# Characterization

## Theorem 2. Optimal Project

- There exists an optimal project in which the agent chooses

$$C'(p) = \begin{cases} 0 & \text{if } p \leq 1/e \\ 1 - \frac{1}{pe} & \text{if } p > 1/e \end{cases}$$

- The principal offers bonus contract with  $b = 1 - 1/e$
  - Each player obtains payoff equal to  $1/e$
- 
- The agent chooses a convex cost function s.t any  $p \leq 1/e$  is costless, while larger  $p$ 's are progressively more expensive and the principal is indifferent across any bonus contract with  $b \in [0, 1 - 1/e]$ .
  - Principal's profit  $\pi = 1/e$ , and agent captures all rents for  $p > 1/e$ .

# Characterization

## Theorem 2. Optimal Project

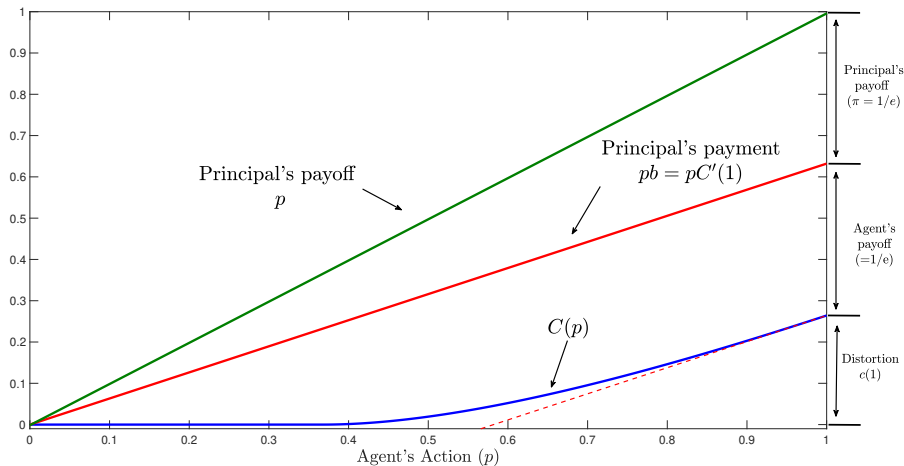
- There exists an optimal project in which the agent chooses

$$C'(p) = \begin{cases} 0 & \text{if } p \leq 1/e \\ 1 - \frac{1}{pe} & \text{if } p > 1/e \end{cases}$$

- The principal offers bonus contract with  $b = 1 - 1/e$
  - Each player obtains payoff equal to  $1/e$
- 
- The agent chooses a convex cost function s.t any  $p \leq 1/e$  is costless, while larger  $p$ 's are progressively more expensive and the principal is indifferent across any bonus contract with  $b \in [0, 1 - 1/e]$ .
  - Principal's profit  $\pi = 1/e$ , and agent captures all rents for  $p > 1/e$ .

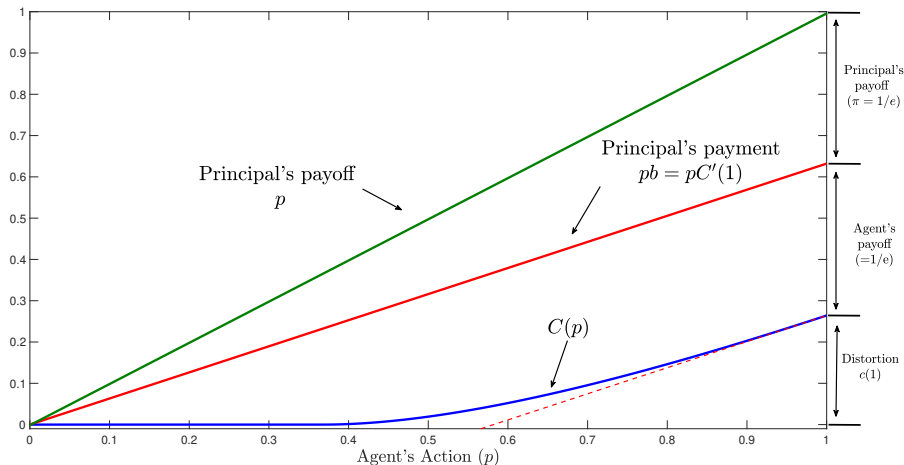


## Graphically



- To capture rents, agent commits to rent seeking activity costing  $C(p)$ .

## Graphically



- To capture rents, agent commits to rent seeking activity costing  $C(p)$ .

## Payoff pairs implementable by an arbitrary binary project

- Insofar, we have assumed the agent can choose *any* cost function

$$c : \Delta([0, 1]) \rightarrow \mathbb{R}_+$$

- Suppose the agent is constrained and must choose among a subset of these cost functions.
- **Q:** Can we make any predictions regarding surplus allocation?
- Let  $V(c) = \{\pi^*, U^*\}$  be the set of equilibrium payoffs for given  $c$ , and define the **payoff possibility set**:

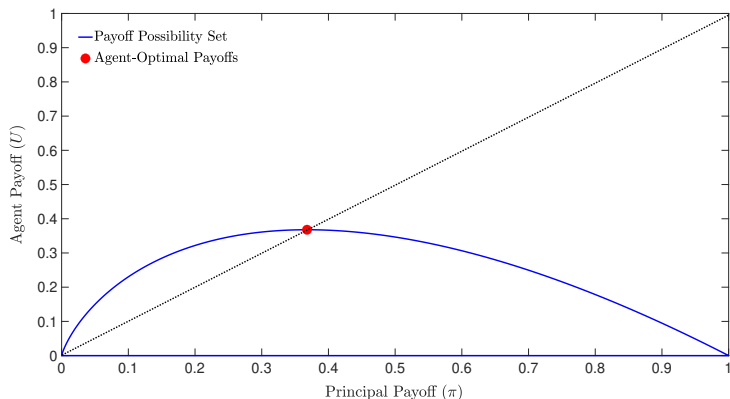
$$\mathcal{P} = \bigcup_{c : \Delta([0,1]) \rightarrow \mathbb{R}_+} V(c).$$

# Payoff pairs implementable by an arbitrary binary project

## Theorem 3. Payoff Possibility Set

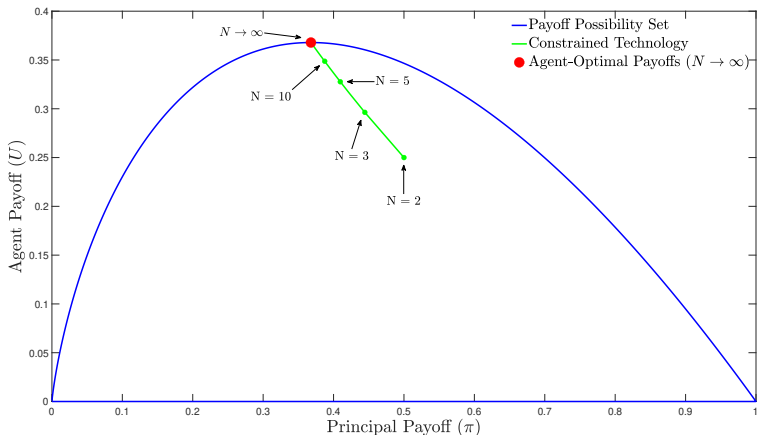
The payoff possibility set is

$$\mathcal{P} = \text{co}(\{\pi, -\pi \log \pi\} : \pi \in [0, 1]).$$



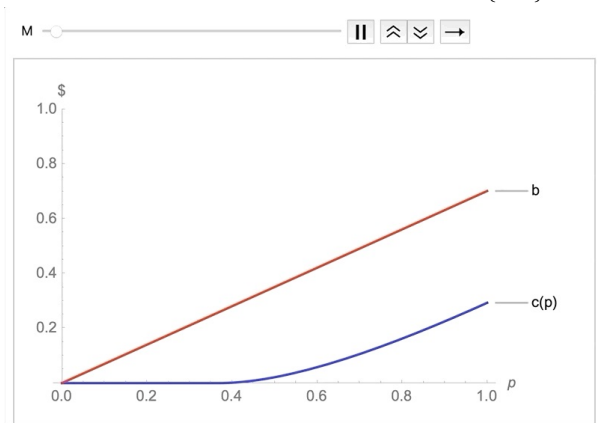
# Bounded Project Complexity

- Suppose the agent can choose a project with at most  $N$  actions.
- By Theorem 1, wolog, he chooses  $p_i \in [0, 1]$  and  $C(p_i) \geq 0$  for each  $i$



## Negative Payoffs

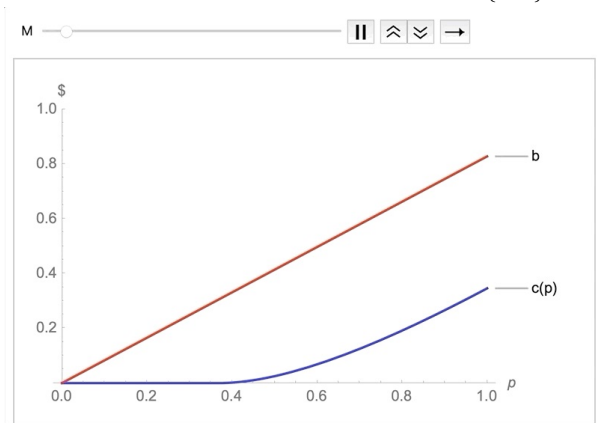
- Suppose agent can choose output distributions with support  $[-M, 1]$ .
- Suffices to focus on binary projects s.t  $F(x) = \mathbb{I}_{\{x=1\}}$  is implemented.



- When  $M = 0$ ,  $C(\cdot)$  and  $b$  are given in Theorem 2.

## Negative Payoffs

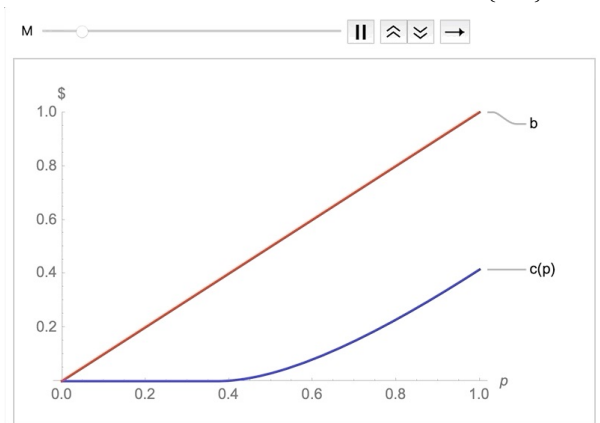
- Suppose agent can choose output distributions with support  $[-M, 1]$ .
- Suffices to focus on binary projects s.t  $F(x) = \mathbb{I}_{\{x=1\}}$  is implemented.



- As  $M \uparrow$ , both  $C(\cdot)$  and  $b$  are shifted upwards.

## Negative Payoffs

- Suppose agent can choose output distributions with support  $[-M, 1]$ .
- Suffices to focus on binary projects s.t  $F(x) = \mathbb{I}_{\{x=1\}}$  is implemented.

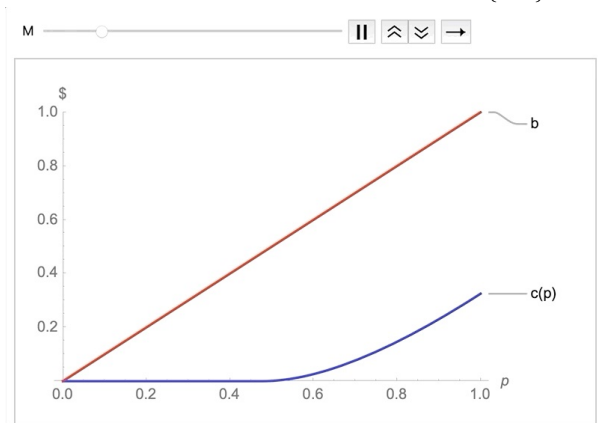


- For  $M$  sufficiently large,  $b = 1$ , and agent extracts all surplus.



## Negative Payoffs

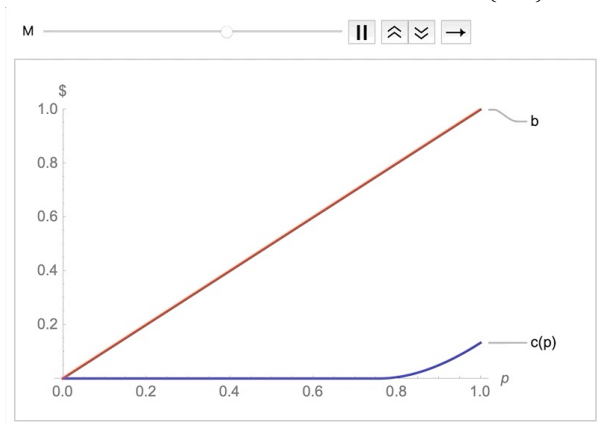
- Suppose agent can choose output distributions with support  $[-M, 1]$ .
- Suffices to focus on binary projects s.t  $F(x) = \mathbb{I}_{\{x=1\}}$  is implemented.



- As  $M \uparrow$  further,  $C(\cdot)$  is shifted downwards, decreasing distortion.

## Negative Payoffs

- Suppose agent can choose output distributions with support  $[-M, 1]$ .
- Suffices to focus on binary projects s.t  $F(x) = \mathbb{I}_{\{x=1\}}$  is implemented.



- As  $M \rightarrow \infty$ ,  $b = 1$  and  $C(\cdot) \rightarrow 0$  leading to efficiency.

# Risk-averse Agent

- Theorem 1 holds if the agent is not *too risk-averse*.

## Corollary 1. Risk-averse Agent

- Let  $u_k(\cdot)$  be a sequence of functions satisfying  $u_k'' < 0 < u_k'$  for each  $k$ , and  $\lim_{k \rightarrow \infty} u_k(\omega) = \omega$  uniformly.
- There exists a  $K$  such that a binary project optimal whenever  $k \geq K$ .
- Theorem 2 — the characterization of the optimal binary project is straightforward for any concave utility function.

## Related Literature (Incomplete List)

- *Principal-agent models:*
  - Mirrlees (1976), Holmström (1979), Innes (1990)
  - *Gaming / multitasking:* Carroll (2015), Barron et al. (2020)
  - *Endogenous monitoring technology:* Georgiadis and Szentes (2020)
- *Sequential mechanism design:*
  - Krähmer and Kovác (2016)
  - Bhaskar et al. (2019)
  - Condorelli and Szentes (2020)

# Discussion

- We consider an agency model of moral hazard in which production technology is endogenous and chosen by the agent.
- The agent optimally designs a project with binary output such that the principal is indifferent between  $b^*$  and any smaller bonus, enabling him to extract all rents.
- *Potential implication.* Promoting more flexibility for workers to design their job as an alternative to regulation (e.g., minimum wages)