

Optimal Project Design

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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 & risk-averse, cash constrained agent
 - *Today:* Agent will be risk-neutral (results generalize to any concave u)
- *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)

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Some Intuition

- *Efficient outcome.* Agent sets $c(F) = 0$ for all F
- Principal responds by offering wage 0 and implementing $F(x) = \mathbb{I}_{\{x=1\}}$
 - *i.e.*, while outcome is efficient, 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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Problem Formulation

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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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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.

- *Informativeness principle*: if the signal is made less informative, then incentivizing the agent becomes more expensive
- Output x acts as a signal about the agent's action
- Binary distribution provides the least informative signal among all distributions with the same mean

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Binary Projects: Proof

- Fix a c & suppose principal offers w^* , implementing F^* (w/ mean μ^*)
- 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_μ is a distribution with support $\{0, 1\}$ and mean μ , 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_μ 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 μ
- 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 π , 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 $1/e$
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- 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

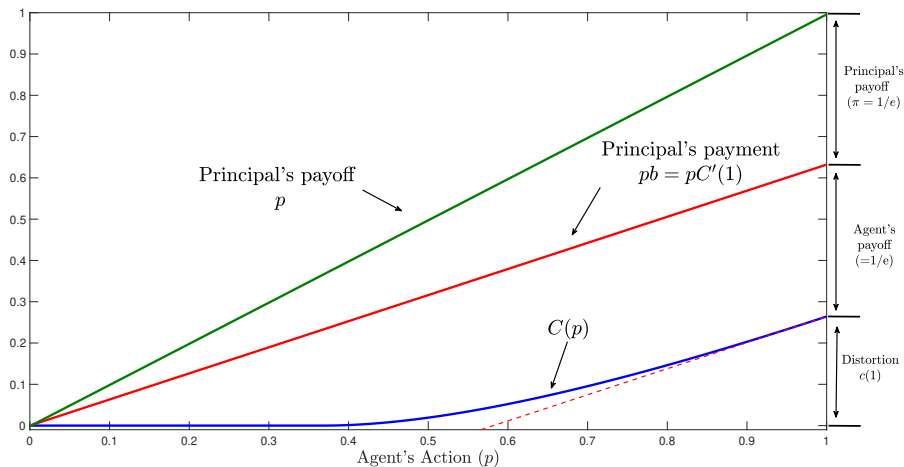
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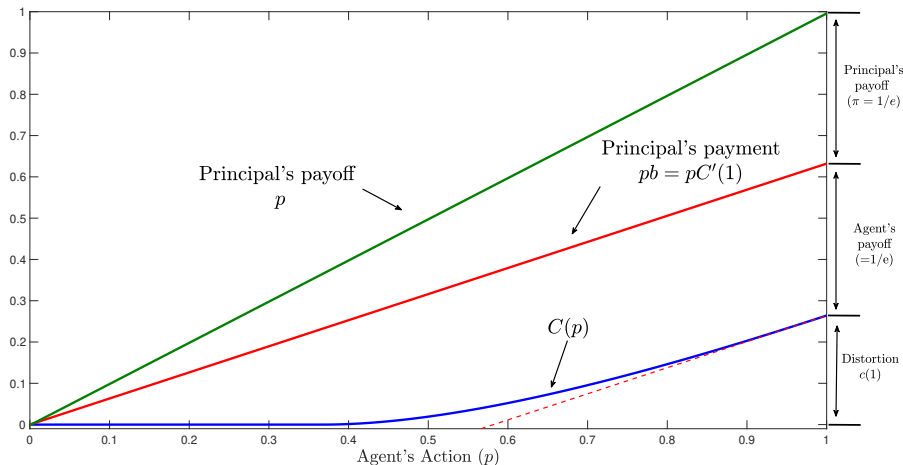
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Graphically



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

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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**:

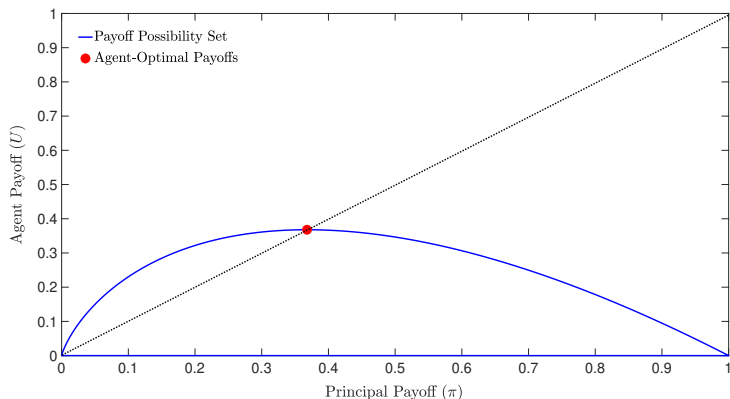
$$\mathcal{P} = \bigcup_{c : \Delta([0,1]) \rightarrow \mathbb{R}_+} V(c).$$

Payoff pairs implementable by an arbitrary binary project

Theorem 3. Payoff Possibility Set

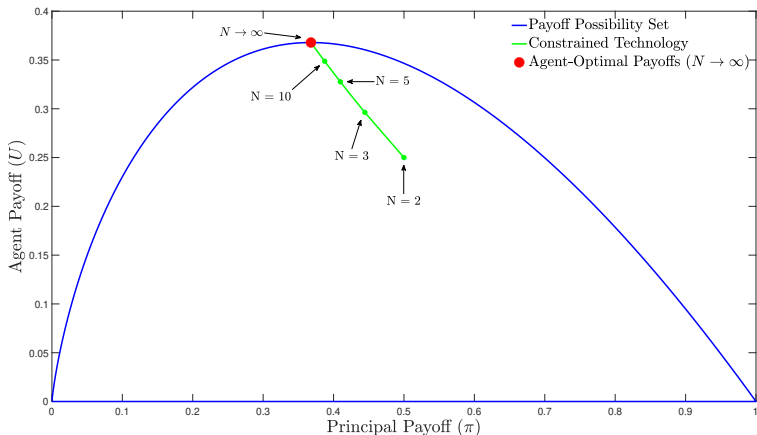
The payoff possibility set is

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



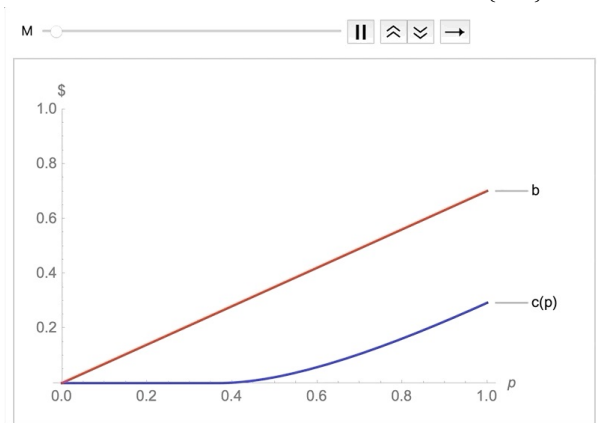
Bounded Project Complexity

- 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

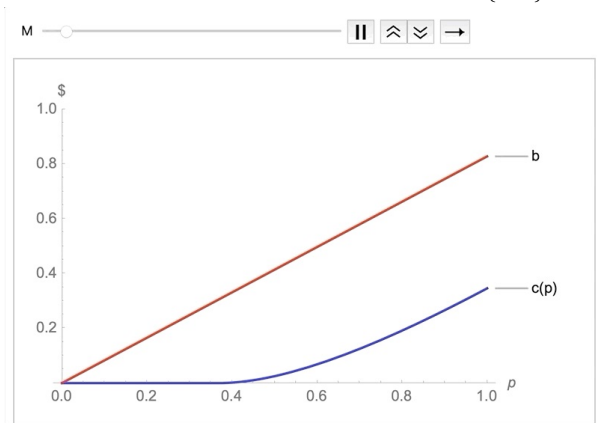
- Suppose agent can choose output distributions with support $[-M, 1]$.
- Suffices to focus on binary projects s.t $F(x) = \mathbb{I}_{\{x=1\}}$ is implemented.



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

Negative Payoffs

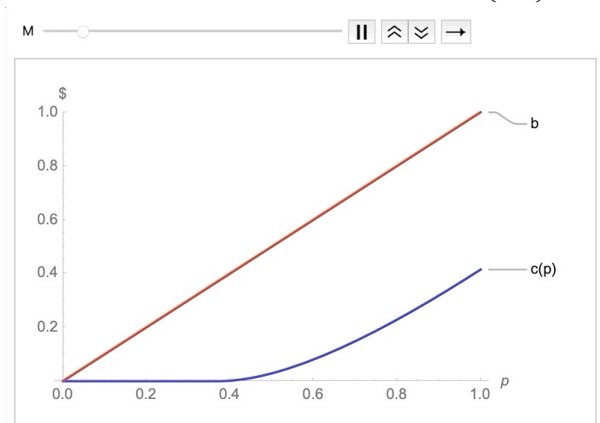
- Suppose agent can choose output distributions with support $[-M, 1]$.
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- As $M \uparrow$, both $C(\cdot)$ and b are shifted upwards.

Negative Payoffs

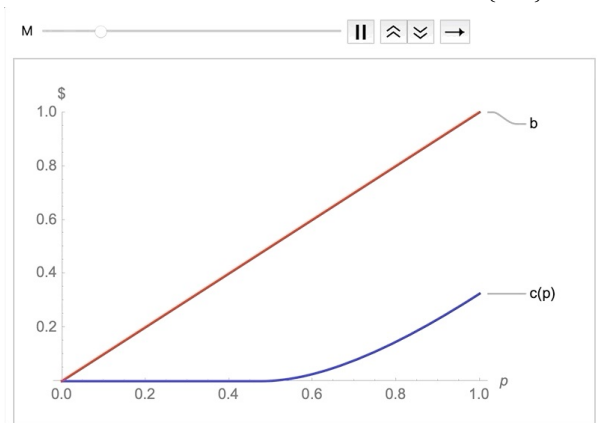
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- For M sufficiently large, $b = 1$, and agent extracts all surplus.

Negative Payoffs

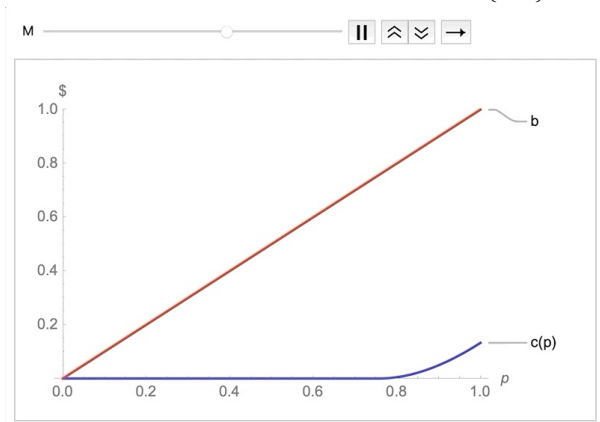
- Suppose agent can choose output distributions with support $[-M, 1]$.
- Suffices to focus on binary projects s.t $F(x) = \mathbb{I}_{\{x=1\}}$ is implemented.



- 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.

Related Literature (Incomplete List)

- *Principal-agent models:*
 - *Informativeness principle:* Holmström (1979), Chaigneau et al. (2019)
 - *Limited liability:* Innes (1990), Poblete and Spulber (2012)
 - *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)