

VAGUE BY DESIGN: PERFORMANCE EVALUATION AND LEARNING FROM WAGES

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INTRODUCTION

- Performance evaluation is a key aspect of labor contracts and organization design
 - Many ways to evaluate: Shop floor control, consumer scores, product testing, sales,...
 - Digitization provides a growing number of possibilities
- Performance evaluations are an important source of information in the workplace
- Inform the firm about the worker's performance
 - Necessary basis of incentivizing effort via performance pay
 - Classic results show more information is better Holmström '79, Grossman&Hart '83
- Inform the worker about his performance
 - Learn about ability/match with the job
 - Confidence in his capability to succeed and sense of agency

Dual role of performance evaluation: basis of *incentives* and agent *learning*

- How do these two aspects interact?
- How to optimally design performance evaluation when it shapes worker confidence?

Dual role of performance evaluation: basis of *incentives* and agent *learning*

- How do these two aspects interact?
- How to optimally design performance evaluation when it shapes worker confidence?
- Two-period model of moral hazard with uncertain match-specific ability
- Principal designs evaluation of output and contingent wages
 - Fully flexible evaluation: Could observe true contribution to profits
 - Commitment to performance pay
- Learning about the agent's ability based on these evaluations

THIS PAPER: PREVIEW OF RESULTS

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Dual role of performance evaluation: basis of *incentives* and agent *learning*

- Agent learning imposes a cost on the principal
- Binary case: Optimal evaluation is noisy & tough
- General case:
 - lower censorship
 - use information about effort, shroud information about ability

- Literature
- 2x2 Model
 - Transformation to Belief Space
 - Terminal Period: Agent Learning Imposes a Cost on the Principal
 - Initial Period: Optimal Evaluation – Noisy and Thoughtful
- General Case
- Extensions

RELATED LITERATURE

- Design of information
Kolotilin et al. '22, Doval&Skreta forthcoming
and performance pay:
Georgiadis&Szentes '20, Hoffmann et al. '21, Li&Yang '20
- Implicit incentives and information design:
Ely&Szydlowski '20, Hörner&Lambert '21, Smolin '20
- More information can increase the cost of incentives:
Fang&Moscarini '05, Jehiel '14, Nafziger '09, Meyer&Vickers '97

2x2 Model

THE MODEL

- Two time periods $t \in \{1, 2\}$, common discount factor δ .
- Agent
 - risk averse with utility index $u = \sqrt{w}$ (this talk) and reservation utility U
 - observable but nonverifiable effort $e_t \in \{0, 1\}$ at cost $c \cdot e$
 - time-invariant ability $\theta \in \{\theta_H, \theta_L\}$, high with prior probability μ .
- Principal
 - risk neutral
 - implements high effort

- Output is
 - high or low, $y_t \in \{y_L, y_H\}$, high with probability

type \ effort	$e_t = 0$	$e_t = 1$
$\theta = \theta_L$	a	$a + b$
$\theta = \theta_H$	$a + \Delta a$	$a + b + \Delta a + \Delta b$

- Effort is productive: $b \geq 0$
- Ability is productive: $\Delta a \geq 0$
- Complementarities: Δb
Log-Supermodular: $\frac{\Delta b}{b} > \frac{\Delta a}{a}$

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INFORMATION, CONTRACTS AND COMMITMENT

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 - a signal structure $S, p(s|y_t)$, and
 - wages w as a function the signal.

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 - a signal structure $S, p(s|y_t)$, and
 - wages w as a function the signal.
- Agent observes the contract and makes participation and effort decision
- Output is not observed
- Principal and agent observe the signal realization, wages, and effort
- Update beliefs to $\hat{\mu}(s) \in [\underline{\mu}, \bar{\mu}]$

THE CONTRACTING PROBLEM

First Period

$$\Pi_1 = \max_{S, p, w} \mathbb{E}[y|e = 1, \mu] + \int_S p(s|e = 1, \mu) \left(\Pi_2(\hat{\mu}(s)) - w(s) \right) ds \quad (1)$$

$$\text{s.t. } \int_S p(s|e = 1, \mu) u(w(s)) ds - c \geq U \quad (P_1)$$

$$\int_S p(s|e = 1, \mu) u(w(s)) ds - c \geq \int_S p(s|e = 0, \mu) u(w(s)) ds \quad (IC_1)$$

Second Period

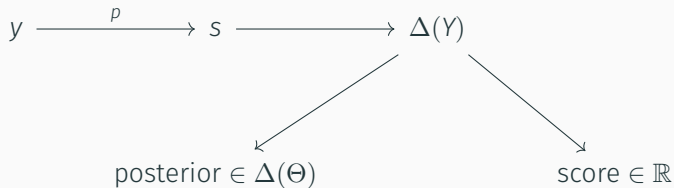
$$\Pi_2(\hat{\mu}) = \max_{S, p, w} \mathbb{E}[y|e = 1, \hat{\mu}] - \int_S p(s|e = 1, \hat{\mu}) w(s) ds \quad (2)$$

$$\text{s.t. } \int_S p(s|e = 1, \hat{\mu}) u(w(s)) ds - c \geq U \quad (P_2)$$

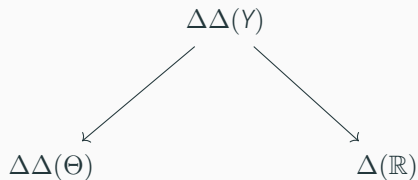
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$$\begin{aligned} \max_{S, p, w} \quad & \mathbb{E}[y|e=1] + \mathbb{E}_p \left(\Pi_2(\hat{\mu}(s)) - w(s) \right) \\ \text{s.t.} \quad & \mathbb{E}_p u(w(s)) - c \geq U & (P_1) \\ & \mathbb{E}_p \frac{p(s|e=1) - p(s|e=0)}{p(s|e=1)} u(w(s)) \geq c & (IC_1) \end{aligned}$$

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 \end{aligned}$$

$$\Delta\Delta(\Theta) \longleftrightarrow \Delta(\mathbb{R})$$

Proposition

The contracting problem can be written as a choice of a distribution over posteriors, m , with mean μ and support on $[\underline{\mu}, \bar{\mu}]$ and a mapping from posteriors to wages.

- Wages deterministic as a function of the posterior: $w(s) \mapsto w(\hat{\mu})$
- Probability over signal realizations on-path: $p(s|e_t = 1, \mu) \mapsto m(\hat{\mu})$
- Score:

$$\begin{aligned} p(s|e_t = 1, \mu) - p(s|e_t = 0, \mu) &\mapsto (b + \Delta b \mu) (p(s|y_H) - p(s|y_L)) \\ &\mapsto (b + \Delta b \mu) \frac{\hat{\mu} - \mu}{(\Delta a + \Delta b) \mu (1 - \mu)} m(\hat{\mu}) \end{aligned}$$

$$\Pi_1 = \max_{m,w} \mathbb{E}[y|e=1, \mu] + \int m(\hat{\mu}) (\delta \Pi_2(\hat{\mu}) - w(\hat{\mu})) d\hat{\mu} \quad (3)$$

$$\text{s.t. } \int u(w(\hat{\mu})) m(\hat{\mu}) d\hat{\mu} - c \geq U \quad (P_1)$$

$$\int (b + \Delta b \mu) \frac{\hat{\mu} - \mu}{(\Delta a + \Delta b) \mu (1 - \mu)} u(w(\hat{\mu})) m(\hat{\mu}) d\hat{\mu} \geq c \quad (IC_1)$$

$$\int \hat{\mu} m(\hat{\mu}) d\hat{\mu} = \mu; \quad \text{supp}(m) \subset [\underline{\mu}, \bar{\mu}] \quad (BP)$$

THE FINAL PERIOD

- Pure incentive problem, no motive to shape learning
- Classic result of Grossman and Hart (1983):

Proposition

The optimal evaluation in the final period is fully informative.

THE IMPACT OF INFORMATION

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 - Increases continuation profit
 2. Agent has more information when choosing effort
 - More expensive to satisfy incentive compatibility
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} scales with Δb :
interaction of
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- Second-period IC:

$$u(w_H) - u(w_L) = \frac{c}{b + \hat{\mu}\Delta b}$$

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- Required bonus inversely proportional to a linear function of beliefs
 - Optimistic agent cheaper to motivate
 - Uncertain agent is cheaper to motivate
 - Given change in belief has a larger effect at low beliefs

Proposition

If the technology is log-supermodular, Π_2 , is strictly concave in the posterior belief. Furthermore, it is more concave at low posteriors, $\Pi_2''' > 0$.

- Strong complementarity of effort and ability: Agent learning dominates
- Principal has an incentive to conceal information
- Information aversion strongest at the bottom: avoid pessimistic agents

SOLVING THE FULL PROBLEM

- First period: Incentives and learning
 - Incentives: More informative evaluation *decreases* agency cost *this period*
 - Learning: More informative evaluation *increases* agency cost *next period*

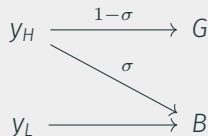
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- First period: Incentives and learning
 - Incentives: More informative evaluation *decreases* agency cost *this period*
 - Learning: More informative evaluation *increases* agency cost *next period*
- Information design problem, with:
 - Endogenous payoffs (wages are designed)
 - Additional constraints (participation and incentive compatibility)
- Binary state does not guarantee binary evaluation (Le Treust&Tomala, 2019)

THE OPTIMAL CONTRACT

Theorem 1

The optimal contract in the first period is (essentially) unique, with a binary ($S = \{G, B\}$) and tough evaluation structure. Let $\sigma \in [0, 1)$ denote the degree of vagueness. The optimal contract consists of



- a good evaluation and associated high wage, only if output was good,
- a bad evaluation and associated low wage: always after output was bad, with prob. σ after output was good.

PROOF OF THEOREM 1: OUTLINE

$$\mathcal{L}(w, m; \underbrace{(\lambda_P, \lambda_{IC})}_{\lambda})$$

Lagrangian of the contracting problem including (P) and (IC)

Concavification of the partially maximized Lagrangian (Georgiadis&Szentes '20)

PROOF OF THEOREM 1: OUTLINE

$$\mathcal{L}(w, m; \lambda)$$

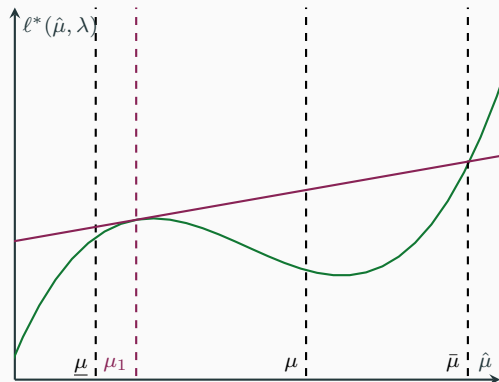
Optimal Wages given m, λ : Standard moral hazard problem $\mapsto w^*(\hat{\mu}; \lambda)$

objective is an expectation given λ : $\mathcal{L}(w^*(\hat{\mu}; \lambda), m; \lambda) = \int \ell^*(\hat{\mu}; \lambda) m(\hat{\mu}) \, d\hat{\mu}$

Information Design given λ : Concavification of $\ell^* \mapsto m^*(\hat{\mu}; \lambda)$

$m^*(\cdot; \lambda)$ binary and tough for all λ

Duality: \mapsto Solution exists and features of m^* hold in the optimal contract



- Unconstrained information design
- Payoff $\ell^*(\hat{\mu}; \lambda)$
- $\frac{\partial^3}{\partial \hat{\mu}^3} \ell^*(\hat{\mu}; \lambda) > 0$
 - Convex $\Rightarrow m$ fully informative
 - Concave-convex: For low μ , learning effect dominates $\Rightarrow m$ tough
- This for given λ , but $\lambda(m)$!

- Optimal evaluation is binary
 - Motive to control learning does not add complexity to the evaluation
 - Data aggregated into a pass-fail signal
- Noisy evaluation can be optimal
 - Preserve agent's uncertainty
- The optimal noise is asymmetric
 - Tough evaluation: Avoid unwarranted praise, embrace unwarranted reprimand
 - “Drill-sergeant mentality” is part of optimal organization design
- Prevent very low posteriors
 - Costly to motivate, change in posterior has a large effect
- Result of joint design of evaluation and wages

WHAT IF EFFORT AND ABILITY ARE NOT STRONG COMPLEMENTS?

- Intermediate case: Weak complements/substitutes
 - More information increases the continuation value
 - Fully informative evaluation
- Strong substitutes: $\frac{\Delta b}{b} + \frac{\Delta a}{1-a} < 0$
 - More information decreases the continuation value
 - Noisy and *lenient* evaluation
- Lenient Evaluation
 - Let some failures slip (but punish others harshly)
 - Lack of reprimand not very informative: Avoids complacency

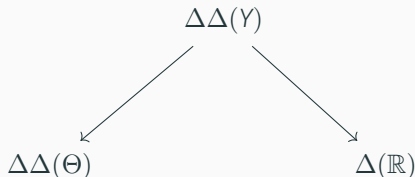
The General Case

POSTERIOR SPACE

Consider now a general $Y \subset \mathbb{R}$ with pdf $f(y|e, \mu)$

$$\begin{aligned} \max_{S, p, w} \quad & \mathbb{E}[y|e = 1] + \mathbb{E}_p \left(\Pi_2(\hat{\mu}(s)) - w(s) \right) \\ \text{s.t.} \quad & \mathbb{E}_p u(w(s)) - c \geq U \end{aligned} \tag{P_1}$$

$$\mathbb{E}_p \frac{p(s|e = 1) - p(s|e = 0)}{p(s|e = 1)} u(w(s)) \geq c \tag{IC_1}$$



- Rewrite the program as a choice of $\Phi \in \Delta\Delta Y$
- Let $\boldsymbol{\mu} \in [0, 1]^{|Y|}$, $\mathbf{x} \in \mathbb{R}^{|Y|}$ denote the vector of posteriors and scores given y

$$\begin{aligned} \max_{w, \Phi} \mathbb{E} \left(\Pi_2(\boldsymbol{\mu} \cdot \phi) - w(\phi) \right) \\ \text{s.t. } \mathbb{E} u(w(\phi)) - c \geq U & \quad (P_1) \\ \mathbb{E} (\mathbf{x} \cdot \phi) u(w(\phi)) \geq c & \quad (IC_1) \\ \mathbb{E} \phi = f(\cdot | 1, \mu_0) \end{aligned}$$

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- **The 1-dimensional case:** The following are equivalent
 - $x \in \text{span}(\mu, \mathbf{1})$
 - there is a bijection between feasible distributions over posteriors and feasible distributions over scores
 - $f = f(\cdot|0, 0) + g(e, \theta)(f(\cdot|1, 1) - f(\cdot|0, 0))$

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- 1-dim: Sufficiently strong complementarities $\Rightarrow \Pi_2''' > 0 \Rightarrow$ lower censorship

Extensions

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- Many periods
 - Not analytically tractable: lack of control over shape of continuation value
 - Numerically: Same structure within period; noisier evaluation early in the relationship

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 - As little information as possible about ability

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 - As much information as possible about effort
 - As little information as possible about ability
- Optimal Performance Evaluation
 - Noisy, even though wages could condition on true y
 - Strong complementarity: avoid very low posterior beliefs (tough/lower-censorship)

OUTLOOK

- Preference across given information sources: conduct, not results!
 - Salary differences between workers: mostly driven by types, so should be concealed
- Affects task design: Harder/easier to keep agents motivated
- Career Concerns: informationally opposite forces
 - information about effort and ability inseparably intertwined
 - here: source of friction; CC: source of incentives

Thank You!

UTILITY FUNCTION

- Sufficient condition on utility function
- $w = u^{-1}$, “cost of utility”

Assumption 1

1. (No incentives at probability zero) $\frac{w(x)}{x} \rightarrow \infty$ as $x \rightarrow \infty$.
2. (Decreasing curvature) $w''' \leq 0$.
3. (Bounded changes in curvature) $\frac{w'''(u_H)}{w''(u_H)} \geq -A$.

Condition

- Satisfied for CRRA $u(x) = \frac{x^{1-\gamma}}{1-\gamma}$
 - if $\gamma \leq 1/2$ and U sufficiently large.
 - Always satisfied for $\gamma = \frac{1}{2}$

STEP 1: OPTIMAL WAGES

- Let $\mathcal{L}(m, w; (\lambda_P, \lambda_{IC}))$ denote the Lagrangian associated to the problem.
- Solving for the optimal wage given λ yields

$$w^*(\hat{\mu}, \lambda) = u'^{-1} \left(\left(\lambda_P + \lambda_{IC} \frac{b + \Delta b \mu}{(\Delta a + \Delta b) \mu (1 - \mu)} (\hat{\mu} - \mu) \right)^{-1} \right)$$

- Partially maximized Lagrangian, $\sup_w \mathcal{L}(m, w; (\lambda_P, \lambda_{IC}))$, is posterior separable

$$\begin{aligned} \mathcal{L}(m, w^*(\hat{\mu}, \lambda); (\lambda_P, \lambda_{IC})) = & \int \left\{ P_\mu^1 Y + \delta \Pi_2(\hat{\mu}) - w^*(\hat{\mu}, \lambda) \right. \\ & + \lambda_P (u(w^*(\hat{\mu}, \lambda)) - c - U) \\ & \left. + \lambda_{IC} \left(\frac{b + \Delta b \mu}{(\Delta a + \Delta b) \mu (1 - \mu)} (\hat{\mu} - \mu) u(w(\hat{\mu}, \lambda)) - c \right) \right\} m(\hat{\mu}) d\hat{\mu} \end{aligned}$$

STEP 2: INFORMATION DESIGN

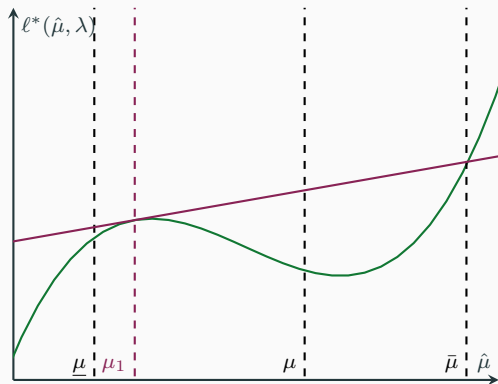
- Unconstrained information design problem with payoff $\ell^*(\hat{\mu}; \lambda)$
- The objective is either convex or concave-convex since

$$\frac{\partial^3}{\partial \hat{\mu}^3} \ell^*(\hat{\mu}; \lambda) = \lambda_{IC} \left(\frac{b + \Delta b \mu}{(\Delta a + \Delta b) \mu (1 - \mu)} \right) \frac{\partial^2}{\partial \hat{\mu}^2} u(w(\hat{\mu}; \lambda)) + \delta \Pi_2'''(\hat{\mu}) > 0$$

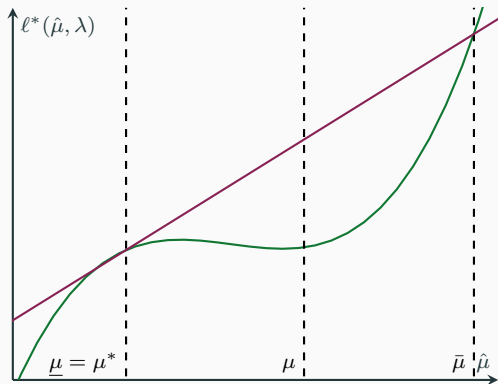
Lemma

For any λ_{IC} , there exists a unique solution to the information design problem. It induces at most two posteriors: the highest feasible posterior $\bar{\mu}$ with probability $m(\bar{\mu}) \in [0, \frac{\mu - \underline{\mu}}{\bar{\mu} - \underline{\mu}}]$ and a low posterior, $\mu^ \in [\underline{\mu}, \mu]$ with $m(\mu^*) \in [\frac{\bar{\mu} - \mu}{\bar{\mu} - \underline{\mu}}, 1]$.*

STEP 2: INFORMATION DESIGN



(a) Interior solution.



(b) Corner solution.

STEP 3: STRONG DUALITY

- We need to show strong duality in the general problem, i.e.

$$\inf_{\lambda \geq 0} \sup_{w, m \text{ s.t. (BP)}} \mathcal{L}(m, w; \lambda) = \sup_{w, m \text{ s.t. (BP)}} \inf_{\lambda \geq 0} \mathcal{L}(m, w; \lambda)$$

- Two steps: [1] Wages

Lemma

The wage setting problem satisfies strong duality, i.e.

$$\sup_w \inf_{\lambda \geq 0} \mathcal{L}(m, w; \lambda) = \inf_{\lambda \geq 0} \sup_w \mathcal{L}(m, w; \lambda).$$

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- Two steps: [2] Information Design

Lemma

The information design problem satisfies strong duality, i.e.

$$\sup_{m \text{ s.t. (BP)}} \inf_{\lambda \geq 0} \int \ell^*(\hat{\mu}; \lambda) m(\hat{\mu}) \, d\hat{\mu} = \inf_{\lambda \geq 0} \sup_{m \text{ s.t. (BP)}} \int \ell^*(\hat{\mu}; \lambda) m(\hat{\mu}) \, d\hat{\mu}.$$

A SIMPLIFIED PROBLEM

- Define a simplified problem, using binary and tough evaluation

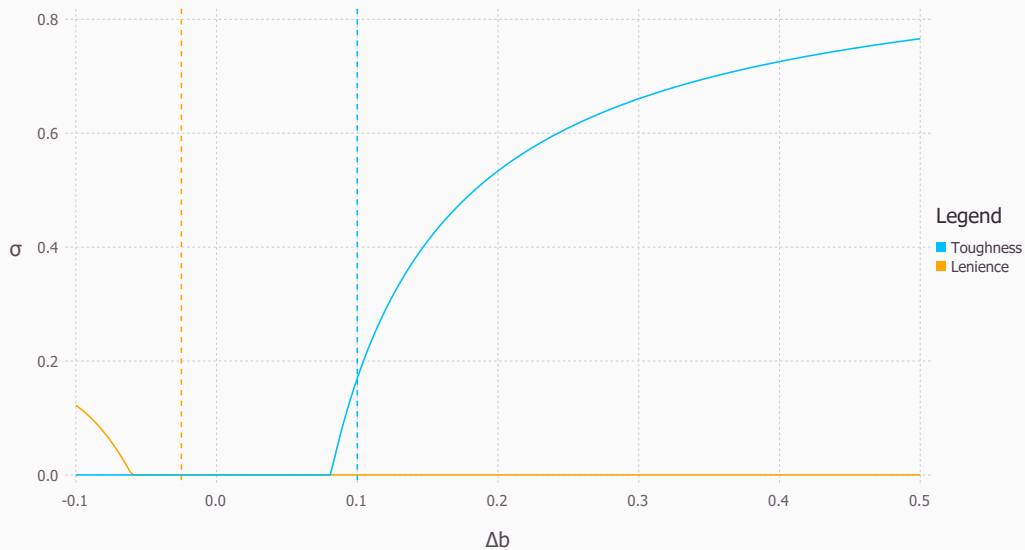
$$\max_{m_1, m_2, \mu_1, w_1, w_2} \mathbb{E}[y_1 | e = 1, \mu] + m_1(\Pi_2(\mu_1) - w_1) + m_2(\Pi_2(\bar{\mu}) - w_2) \quad (4)$$

$$\text{s.t. } m_1 u(w_1) + m_2 u(w_2) - c \geq U \quad (\text{P})$$

$$\frac{b + \Delta b \mu}{(\Delta a + \Delta b) \mu (1 - \mu)} \sum_i m_i (\mu_i - \mu) u(w_i) \geq c \quad (\text{IC})$$

$$m_1 \mu_1 + m_2 \bar{\mu} = \mu; \quad m_1 + m_2 = 1; \quad \mu_1 \geq \underline{\mu} \quad (\text{BP})$$

COMPLEMENTS AND SUBSTITUTES

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Assumption (1*)

1. (No incentives at probability zero) $\frac{w(x)}{x} \rightarrow \infty$ as $x \rightarrow \infty$.
2. (Increasing curvature) $w''' \geq 0$.
3. (Bounded changes in curvature)

$$\frac{3(b + \mu\Delta b)\Delta b}{c(a\Delta b - b\Delta a)} \geq \frac{w'''(u_L)}{w''(u_L)},$$

where $u_L = U - \frac{a + \mu\Delta a}{b + \mu\Delta b}c$.

PRIVATE INFORMATION OF THE PRINCIPAL

- Principal chooses
 - Evaluation structure: observed by agent, basis of performance pay and learning
 - Private evaluation: not observed by agent, basis of learning only for principal
- Joint distribution over posteriors: $m_P(\mu_P, \hat{\mu})$
 - Agent observes $m(\hat{\mu}) = \int m_P(\mu_P, \hat{\mu}) d\mu_P$
- Dynamic game with incomplete information
- Agent updates belief based on
 - First-period evaluation
 - Second-period contract offer

PRIVATE INFORMATION OF THE PRINCIPAL

- Unique PBE with passive beliefs: outcome equivalent to optimal contract without private information acquisition
 - Passive beliefs: no updating based on contract offer
 - Principal preferred*
- Private information either revealed or not useful
 - If private information isn't used to adjust second period contract: irrelevant
 - Information used to adjust contract offer: revealed to agent
 - Better to also use it as a basis of performance pay

*Among equilibria that satisfy no-holdup: No rent for the agent in the second period on path.

PRIVATE INFORMATION OF THE PRINCIPAL

- Unique PBE with passive beliefs: outcome equivalent to optimal contract without private information acquisition
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 - If private information isn't used to adjust second period contract: irrelevant
 - Information used to adjust contract offer: revealed to agent
 - Better to also use it as a basis of performance pay
- Remains an equilibrium when principal *has to* acquire private information
- Unique[†] when private information acquisition strategy observed

*Among equilibria that satisfy no-holdup: No rent for the agent in the second period on path.

[†]Under no-holdup and no-signaling-what-you-don't-know.

UNOBSERVABLE EFFORT

- Suppose effort is not observed by the principal
- After a deviation to low effort, signal s
 - Principal has posterior

$$\hat{\mu}(s) = \mu \frac{p(s|y_L) + (a + b + \Delta a + \Delta b) [p(s|y_H) - p(s|y_L)]}{p(s|y_L) + (a + b + (\Delta a + \Delta b)\mu) [p(s|y_H) - p(s|y_L)]}$$

- Agent interprets signal differently:

$$\mu \frac{p(s|y_L) + (a + \Delta a) [p(s|y_H) - p(s|y_L)]}{p(s|y_L) + (a + \Delta a\mu) [p(s|y_H) - p(s|y_L)]}$$

- Agent has private information about the posterior

UNOBSERVABLE EFFORT

- Incentive compatibility in the second period:
 - Slack if agent more optimistic
 - Violated if agent more pessimistic
- “Belief-manipulation motive”
- Double deviations optimal
- First-period IC dynamic: Kink in the principal’s objective at prior μ

$$\int \left\{ \frac{(b + \mu \Delta b)}{\mu(1 - \mu)\Delta b} (\hat{\mu} - \mu) u(w(\hat{\mu})) - \left[1 - \frac{(b + \mu \Delta b)}{\mu(1 - \mu)\Delta b} (\hat{\mu} - \mu) \right] \max\{0, c\Delta b \frac{\mu - \hat{\mu}}{b + \hat{\mu}\Delta b}\} \right\} m(\hat{\mu}) d\hat{\mu} \geq c$$

UNOBSERVABLE EFFORT

- Under $u = \sqrt{\cdot}$ and $\Delta a = 0$: At most *three* evaluation outcomes
 - Neutral signal: Not informative about effort and ability[‡]
 - Conditional on informative evaluation: binary and tough
- Intuition: Avoid outcomes that allow generation of private information

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[‡]In simulations: Never used.

UNOBSERVABLE EFFORT

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 - Neutral signal: Not informative about effort and ability[‡]
 - Conditional on informative evaluation: binary and tough
- Intuition: Avoid outcomes that allow generation of private information
- More complicated with long-run contracting:
 - Principal can induce a learning motive by providing excessive bonuses in $t = 2$
 - Joint design of information and wages in *both periods*

◀ back

[‡]In simulations: Never used.

LONG RUN COMMITMENT: CONTINUATION VALUE

- Principal commits to contract: (S, p, w, v)
 - a signal structure $S, p(s|y)$, realization conditional on contemporaneous output
 - wages w , and
 - continuation value v as a function the signal.

LONG RUN COMMITMENT: CONTINUATION VALUE

- Principal commits to contract: (S, p, w, v)
 - a signal structure $S, p(s|y)$, realization conditional on contemporaneous output
 - wages w , and
 - **continuation value v as a function the signal.**
- Assume $u(x) = 2\sqrt{x}$
 - Theorem 1 goes through, delaying *payments* does not affect the mechanism
 - Optimal evaluation: binary and weakly tough

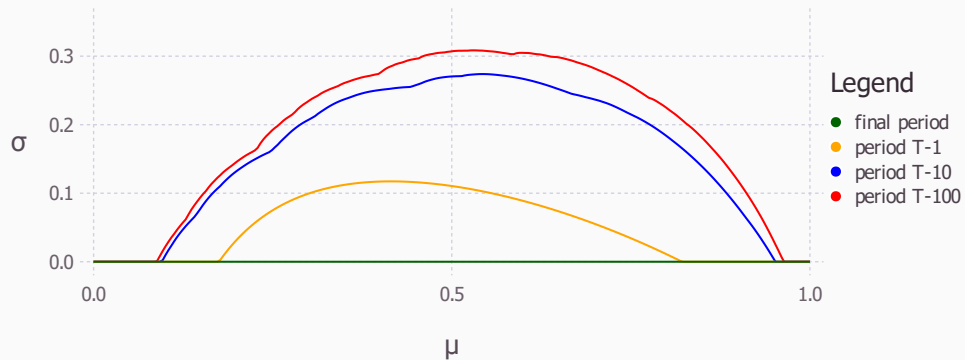
LONG RUN COMMITMENT: FULL COMMITMENT

- Principal commits to long-run contract: $(S_1 \times S_2, p, w)$
 - a signal space $S_1 \times S_2$, p progressively measurable wrt y_t ,
 - and wages w , progressively measurable wrt s_t .
 - Difficult:
 - Agent acquires private info after shirking (effort unobservable to the contract), and
 - the principal can commit to excess bonuses in $t = 2$ (to induce a learning motive).
- ⇒ Characterizing the optimum requires joint design in both periods.

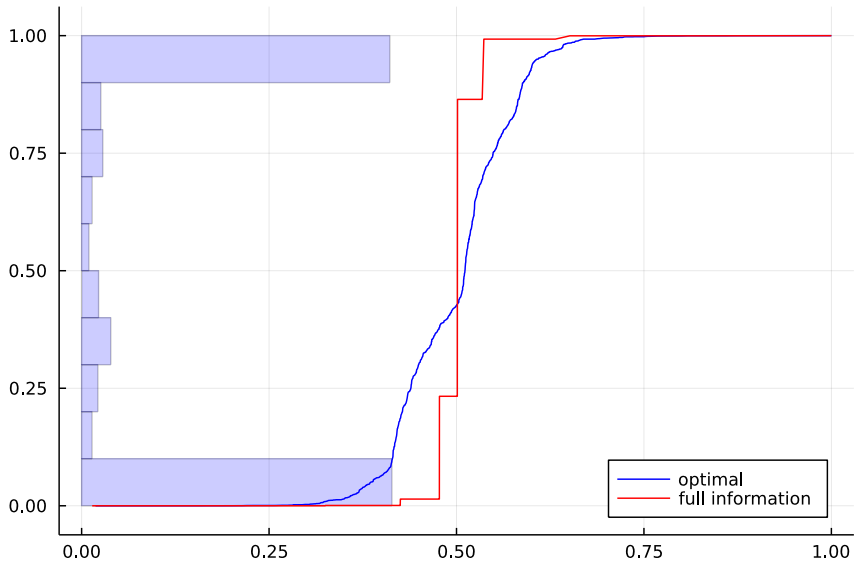
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 - the principal can commit to excess bonuses in $t = 2$ (to induce a learning motive).
- ⇒ Characterizing the optimum requires joint design in both periods.
- Optimum is not tractable. Effect is still in place:
 - Consider optimal contract without fully informative evaluation
 - Bonus for high output in period 1 optimally split between both periods
 - Principal can *postpone* information, but it is *costly*

MANY PERIODS



MANY PERIODS

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Assumption (Bounded changes in curvature)

$$\frac{w'''(u_H)}{w''(u_H)} \geq -\frac{3(b + \mu\Delta b)\Delta b}{c((1-a)\Delta b + b\Delta a)},$$

where $u_H = U + \frac{1-a-\mu\Delta a}{b+\mu\Delta b}c$.

UNDERCONFIDENCE: WHEN IS LEARNING COSTLY?

Proposition

The effect of learning on the principal's continuation value is ambiguous.

- *There exists a threshold \bar{U} such that the continuation value is increasing in information if $U \geq \bar{U}$, and*
- *there exists a threshold $\bar{b} > 0$ such that it is decreasing if $b < \bar{b}$.*