

# NON-MARKET ALLOCATION MECHANISMS

## OPTIMAL DESIGN WITH INVESTMENT INCENTIVES

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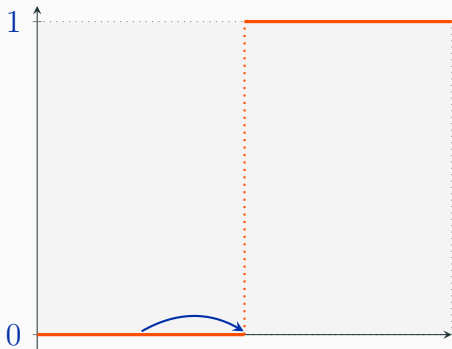
## This paper

How should **allocation/selection mechanisms** be designed, when accounting for agents' **investment incentives**?

# RANDOMIZATION

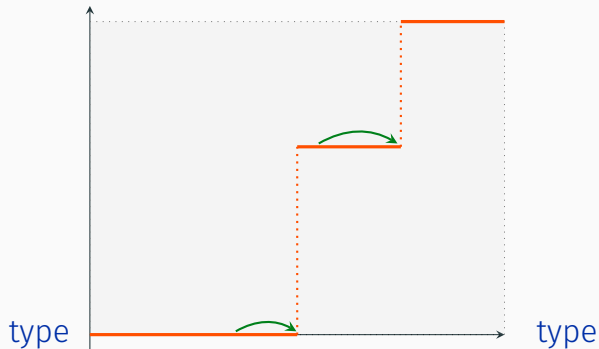
Intuitively, **randomization** might help by spreading investment incentives (but complicated tradeoff).

$\mathbb{P}(\text{select.})$



(a) Pass-fail

$\mathbb{P}(\text{select.})$



(b) Some randomization.

# CONTRIBUTION

## Main theorem

If the investment cost is **quadratic** and the principal is **selecting in the upper tail**, a **pass-fail rule** is **optimal** for the principal.

- A **firmer foundation** for the use of such rules.
- A possible **explanation** for the prevalence of pass-fail rules.
- We also characterize the optimal pass-fail and provide some comparative statics.

## EXTENSIONS

Three extensions:

- (i) **Capacity** constrained principal.
- (ii) **Utilitarian welfare** maximization (weight on the agent).
- (iii) Implementation through **information design**.

### Results

- Pass-fail rules **remain optimal** in (i) and (ii).
  - (i) Capacity constraint **lowers** the optimal cutoff.
  - (ii) Accounting for agents' costs **increases** the optimal cutoff.
- Relaxing the principal's **commitment power** as in (iii) does not reduce her optimal payoff.



# MODEL

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# SETUP

## Agent

- **Natural type:**  $\theta \in \Theta = [\underline{\theta}, \bar{\theta}]$  with  $\underline{\theta} < 0 < \bar{\theta}$ .
- **Distribution:** cdf  $F: \Theta \rightarrow [0, 1]$ .
- **Final type:**  $t \in T = \mathbb{R}$ .
- **Cost:**  $\gamma c(t, \theta)$  where  $\gamma > 0$ .

## Principal

- Can observe  $t$
- Cannot observe  $\theta$  but knows  $F$ .
- **Allocation/Selection:**  $a \in \{0, 1\}$ .

# PAYOFFS

## Agent's payoff

Allocation net of investment cost:

$$v(a, t) = a - \gamma c(t, \theta).$$

## Principal's payoff

Final type conditional on allocation:

$$\pi(a, t) = at$$

## TIMING AND INCENTIVE-COMPATIBILITY

1. The principal **commits** to a **selection rule**  $\sigma: T \rightarrow [0, 1]$  which is publicly revealed.
2. Agents observe  $\theta$  and choose an **investment rule**  $\tau: \Theta \rightarrow T$ .
3. Agents selected with probability  $\sigma(\tau(\theta))$

### Definition (Incentive-compatibility)

- An investment rule  $\tau$  is **incentive-compatible** under selection rule  $\sigma$  if, for all  $\theta \in \Theta$ :

$$\tau(\theta) \in \arg \max_{t \in T} \sigma(t) - \gamma c(t, \theta).$$

- An investment rule  $\tau$  is **implementable** if there exists a selection rule  $\sigma$  under which  $\tau$  is incentive-compatible.

## ASSUMPTIONS: COST

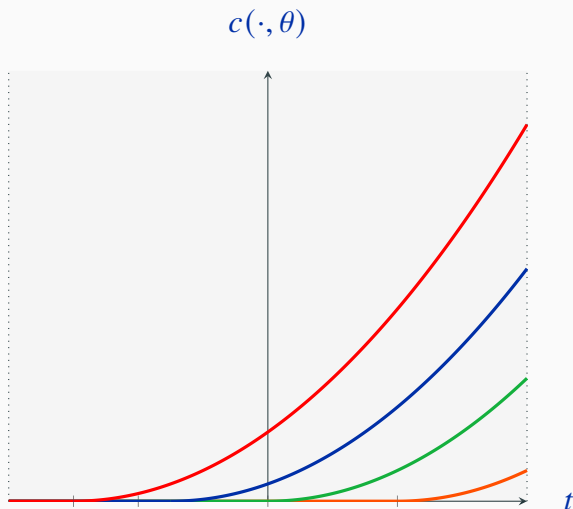
### Assumption (Quadratic cost)

The *cost function*  $c: T \times \Theta \rightarrow \mathbb{R}_+$  is given by:

$$c(t, \theta) = \frac{(t - \theta)^2}{2} \mathbb{1}_{t \geq \theta}.$$

Define  $\theta_0$  by:

$$\gamma c(0, \theta_0) = 1 \Leftrightarrow \theta_0 = -\sqrt{2/\gamma}$$

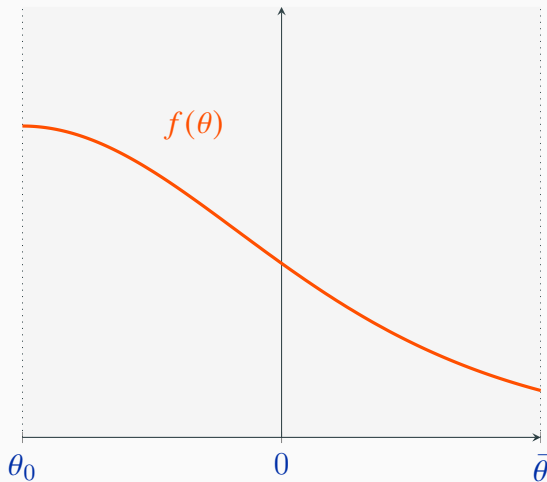


## ASSUMPTIONS: DISTRIBUTION

Assumption (Selection in the upper tail)

The cdf  $F$  admits a *density function*  $f$  which is:

- (i) *strictly positive on  $\Theta$* ;
- (ii) *differentiable*;
- (iii) *decreasing*:  $f'(\theta) \leq 0$  for  $\theta \geq \theta_0$ .



## PRINCIPAL'S PROBLEM

Principal's **ex-ante expected payoff** is:

$$\int_{\underline{\theta}}^{\bar{\theta}} \tau(\theta) \sigma(\tau(\theta)) f(\theta) d\theta$$

Principal's program

$$\begin{aligned} & \underset{\sigma, \tau}{\text{maximize}} && \int_{\underline{\theta}}^{\bar{\theta}} \tau(\theta) \sigma(\tau(\theta)) f(\theta) d\theta \\ & \text{subject to} && \tau(\theta) \in \arg \max_{t \in T} \sigma(t) - \gamma c(t, \theta), \quad \forall \theta \in \Theta \end{aligned}$$

## MAIN RESULT

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## MAIN RESULT

### Definition

A selection rule  $\sigma$  is a  $t^\dagger$ -pass-fail rule if there exists a selection cutoff  $t^\dagger \in T$  such that, for almost every  $t \in T$ :

$$\sigma(t) = \mathbb{1}_{t \geq t^\dagger}.$$

### Theorem

For any  $\gamma > 0$ , there exists a strictly positive selection cutoff  $t_\gamma^*$  such that the  $t_\gamma^*$ -pass-fail rule is optimal.

# OVERVIEW OF THE PROOF

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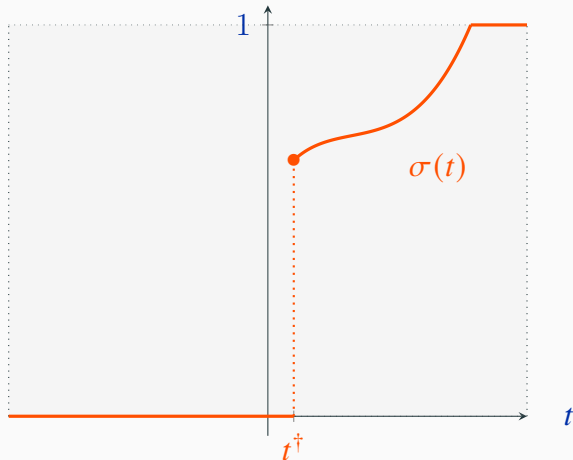
# MONOTONE SELECTION RULES

## Lemma

We can restrict attention w.l.o.g. to selection rules such that:

- (i)  $\sigma(t) = 0$  for all  $t < 0$ , and;
- (ii)  $\sigma$  is non-decreasing;

Let  $t^\dagger = \inf\{t \in T \mid \sigma(t) > 0\}$ .



## AGENT'S PSEUDO UTILITY

- Indirect utility and pseudo utility:

$$\begin{aligned} U(\theta) &:= \max_{t \in T} \sigma(t) - \gamma c(t, \theta) \\ &= \gamma \left( \underbrace{\max_{t \in T} \left\{ t\theta + \frac{\sigma(t)}{\gamma} - \frac{t^2}{2} \right\}}_{:=u(\theta)} - \frac{\theta^2}{2} \right) \end{aligned}$$

- Then  $u(\theta)$  is convex and the Envelope theorem implies

$$u'(\theta) = \tau(\theta) \quad \text{a.e.}$$

- Furthermore, we have:

$$\sigma(\tau(\theta)) = \gamma \left( u(\theta) + \frac{u'(\theta)^2}{2} - \theta u'(\theta) \right) \quad \text{a.e.}$$

## VARIATIONAL PROGRAM

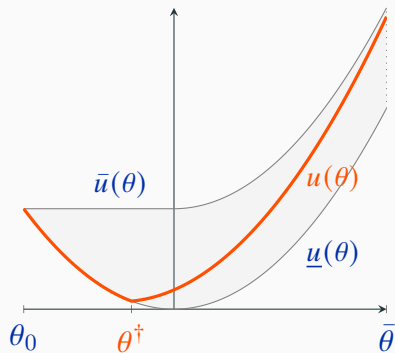
### Principal's program

$$\underset{u \in \mathcal{U}}{\text{maximize}} V(u) := \int_{\theta_0}^{\bar{\theta}} u'(\theta) \left( u(\theta) + \frac{u'(\theta)^2}{2} - \theta u'(\theta) \right) f(\theta) d\theta$$

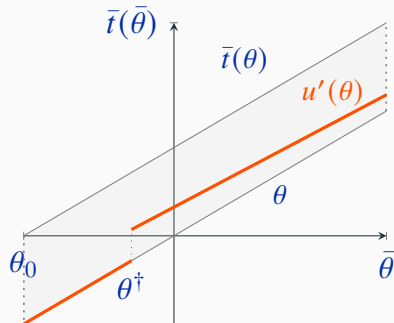
where,  $u \in \mathcal{U}$  (feasible pseudo utility functions) iff:

- (i)  $u$  is a **convex function**.
- (ii)  $\underline{u}(\theta) := \theta^2/2 \leq u(\theta) \leq 1/\gamma + (\theta^2/2)\mathbb{1}_{\theta \geq 0} =: \bar{u}(\theta)$
- (iii)  $\theta \leq u'(\theta) \leq \theta + \sqrt{2/\gamma} =: \bar{t}(\theta)$
- (iv)  $u(\theta) + u'(\theta)^2/2 - \theta u'(\theta) \leq 1/\gamma$

# FEASIBLE PSEUDO UTILITIES: ILLUSTRATION



(a) An admissible pseudo-utility  $u$ .



(b) Its induced investment rule  $u'$ .

## SOLUTIONS AS EXTREME POINTS

Let  $\mathcal{U}(\theta^\dagger) := \left\{ u \in \mathcal{U} \mid \theta^\dagger = \sup\{\theta \in \Theta \mid u(\theta) = \underline{u}(\theta)\} \right\}$  for any  $\theta^\dagger \in [\theta_0, \bar{\theta}]$ .

### Principal's program

$$\underset{\theta^\dagger \in [\theta_0, \bar{\theta}], u \in \mathcal{U}(\theta^\dagger)}{\text{maximize}} \quad V_{\theta^\dagger}(u) = \int_{\theta^\dagger}^{\bar{\theta}} u'(\theta) \left( u(\theta) + \frac{u'(\theta)^2}{2} - \theta u'(\theta) \right) f(\theta) d\theta$$

### Lemma

- For any  $\theta^\dagger \in [\theta_0, \bar{\theta}]$ , the set  $\mathcal{U}(\theta^\dagger)$  is *convex* and *compact*.
- For any  $\theta^\dagger \in [\theta_0, \bar{\theta}]$ , the functional  $V_{\theta^\dagger} : \mathcal{U}(\theta^\dagger) \rightarrow \mathbb{R}$  is *upper-semicontinuous*, and, if  $f$  is *decreasing*, it is also *convex*.
- Therefore  $V_{\theta^\dagger}(u)$  has a maximizer that is an *extreme point* of  $\mathcal{U}(\theta^\dagger)$ .

# EXTREME POINTS: NECESSARY CONDITIONS

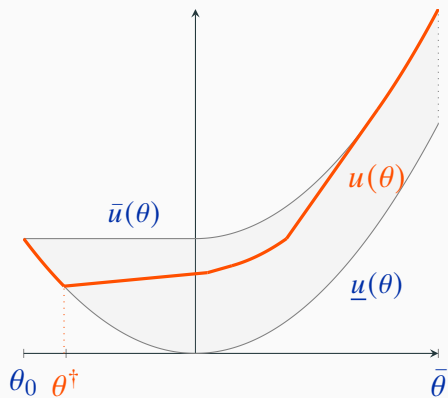
## Proposition

If  $u$  is an extreme point of  $\mathcal{U}(\theta^\dagger)$ , it must be a sequence of *affine* and *quadratic* pieces. Moreover,  $u$  cannot be quadratic below zero.

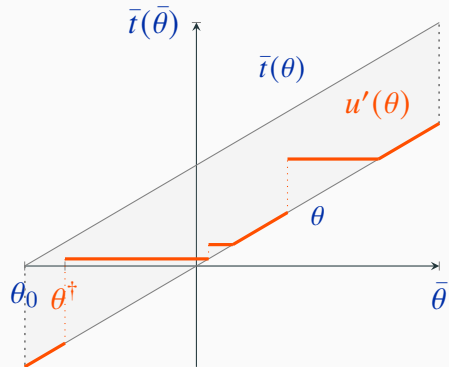
- Affine:  $u(\theta) = a\theta + b$  for some  $a, b$ .
- Quadratic:  $u(\theta) = \frac{\theta^2}{2} + c$  for some  $c \in [0, \gamma/2]$ .
- Let  $\mathcal{C}(\theta^\dagger)$  be the set of such functions = our set of *candidates*.



# EXTREME POINTS: EXAMPLE

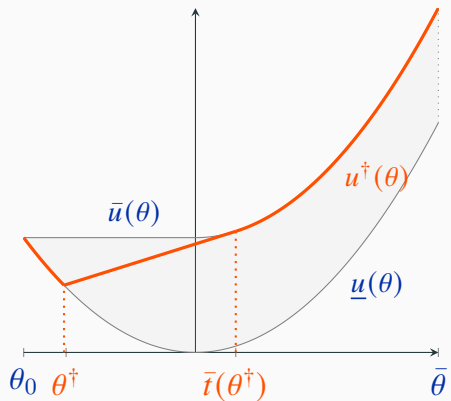


(a) An extreme point  $u$ .

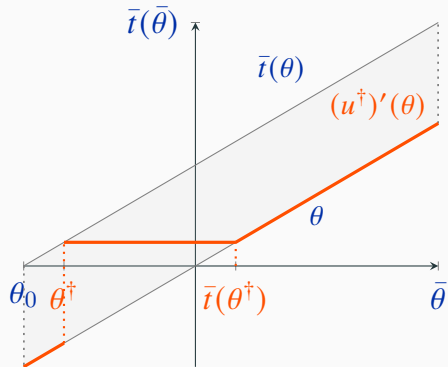


(b) Its induced investment rule  $u'$ .

## OUR CANDIDATE: PASS-FAIL



(a) Pseudo-utility under pass-fail.



(b) Investment under pass-fail.

## PASS-FAIL OPTIMALITY

- Let  $DV_{\theta^\dagger}(u)(h)$  be the Gâteaux derivative of  $V_{\theta^\dagger}$  at  $u$  in direction  $h$ .
- By convexity of  $V_{\theta^\dagger}$ , for any  $u \in \mathcal{C}(\theta^\dagger)$ :

$$V_{\theta^\dagger}(u^\dagger) - V_{\theta^\dagger}(u) \geq DV_{\theta^\dagger}(u)(u^\dagger - u).$$

- We prove that, for any  $u \in \mathcal{C}(\theta^\dagger)$ :

$$DV_{\theta^\dagger}(u)(u^\dagger - u) = \int_{\theta^\dagger}^{\bar{\theta}} \underbrace{(\alpha(\theta) f(\theta) + \beta(\theta) f'(\theta))}_{\geq 0} \underbrace{(u^\dagger(\theta) - u(\theta))}_{\geq 0} d\theta \geq 0.$$

□