

# Robust Delegation

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September 8, 2025

# Introduction

Organizations face two sources of uncertainty (Cyert and March (1963)).

External uncertainty: fluctuations in the competitive environment, operational constraints, or technological developments

Internal uncertainty: heterogeneity in the preferences, skills, and beliefs of the individuals charged with making those decisions.

Delegation literature has placed considerable attention on understanding how external uncertainty shapes internal decision-making processes.

However, internal uncertainty may be equally consequential

# Introduction

Study the role of internal uncertainty in the design of delegation rules

Delegation where principal has limited information on agent's preference

Doesn't know how agent trades off among sub-optimal options

Uncertainty may give more or less discretion to agent

Uncertainty makes delegation simple: no holes, convex, even full delegation

Uncertainty in how agents exploit/manipulate rules for private interest

Sophisticated rules leave more space for manipulation

# Model

# Model

Principal delegates agent to take an action  $a \in A$ , a compact subset of  $\mathbb{R}^n$

Agent privately know state  $\theta \in \Theta$ , a compact subset of  $\mathbb{R}^m$

Principal does not know state  $\theta$ , only holds a belief  $F(\theta)$  with density.

Principal's continuous utility function  $v(a, \theta)$ : quasi-concave in  $a$

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Principal's continuous utility function  $v(a, \theta)$ : quasi-concave in  $a$

Principal does not perfectly know agent's utility function  $u(a, \theta)$

# Model

Principal delegates a compact delegation set  $D$  to agent.

In state  $\theta$ , agent with utility  $u$  chooses an action

$$\tilde{a}(\theta; u, D) \in \operatorname{argmax}_{a \in D} u(a, \theta)$$

Principal's expected payoff from delegation set  $D$  with a given  $u$  is

$$\mathbb{E}_F[v(\tilde{a}(\theta; u, D), \theta)].$$

Principal is uncertain about agent's utility function  $u$

# Model: Uncertainty in Utility

Principal considers every  $u \in \mathcal{U}$  possible.

Principal knows agent's preferred action is in  $a^*(\theta)$ , continuous in  $\theta$ ,

$$a^*(\theta) = \arg \max_{a \in A} u(a, \theta)$$

Perfect identification, can generalize to partial identification  $A^*(\theta)$

$$\mathcal{U} \equiv \{u \mid \text{continuous in } (a, \theta), \text{ strictly quasiconcave in } a, \\ a^*(\theta) = \arg \max_{a \in A} u(a, \theta)\}$$

Our results also hold for other functional forms  $\mathcal{U}_{gq} \subset \mathcal{U}_{qc} \subset \mathcal{U}_{ss}$



# Model: Max-min Design

Principal considers every  $u \in \mathcal{U}$  possible.

She is ambiguity-averse and evaluates the performance of a set  $D$  by its *worst-case expected payoff*

There are two possible max-min design framework

$$\sup_D \inf_{u \in \mathcal{U}} \mathbb{E}_F[v(\tilde{a}(\theta; u, D), \theta)]$$
$$\sup_D \mathbb{E}_F[\inf_{u \in \mathcal{U}} v(\tilde{a}(\theta; u, D), \theta)].$$

We prove they are equivalent: focus on the second one from now

# Model Discussion

Asymmetric treatment on two sources of uncertainty: state + preferences

Preferences are higher dimensional objective  $\mathbb{R}^{A \times \Theta}$  than states  $\Theta$

Sampling preferences requires revealed-preference designs—demanding repeated observations of the decision problem

Screening over high-dimensional preferences is not that tractable

Better connect with classical delegation literature

# Literature Review

Bayesian Delegation in Uni-dimension:

Holmström (1984), Melumad and Shibano (1991), Alonso and Matouschek (2008), Amador and Bagwell (2013) etc.

Bayesian Delegation in Multi-dimension:

Alonso et al. (2014), Gan et al. (2023), Frankel (2016), Kleiner (2022)

Robust Design:

Robust in  $A$ ,  $F(\theta)$ , agents' belief about  $\theta$ , agents' equilibrium play

Robust Delegation in  $v(\theta, a)$ : Frankel (2014)

# Analysis

# Uncertainty in Behaviors

Uncertainty in preference  $\rightarrow$  Uncertainty in how agent manipulate rules

For a  $D$ , define the  $D$ -Admissible Set correspondence,  $A_D(\theta)$ , the set of decisions selected by agents with preferences in  $\mathcal{U}$  when the state is  $\theta$ , i.e.,

$$A_D(\theta) \equiv \{a \in D : \exists u \in \mathcal{U}, \forall a' \in D, u(a, \theta) \geq u_A(a', \theta)\}.$$

Principal doesn't know how agent makes tradeoffs when constrained.

Principal's worst payoff in state  $\theta$  when agent selects from  $D$  is

$$\underline{V}_D(\theta) \equiv \inf_{a \in A_D(\theta)} v(a, \theta).$$

# Rewrite Design Problem

Principal's worst payoff in state  $\theta$  when agent selects from  $D$  is

$$\underline{V}_D(\theta) \equiv \inf_{a \in A_D(\theta)} v(a, \theta).$$

Principal's optimal robust delegation problem can be written as

$$\max_{\text{compact } D \subset A} \int_{\Theta} \underline{V}_D(\theta) dF. \quad (1)$$

To study the property of  $\underline{V}_D$ , we need geometric property of  $A_D(\theta)$

# Visible Set

Define  $C(x, D)$  the set of visible points on  $D$  from point  $x$

$$C(x, D) \equiv \{a \in D : [a, x] \cap D = a\}$$

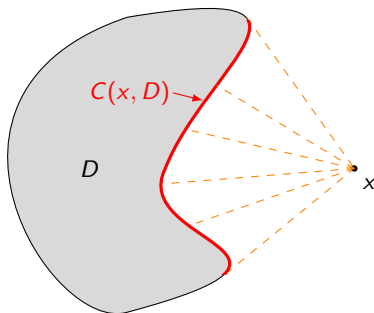


Figure: Illustration of visible set

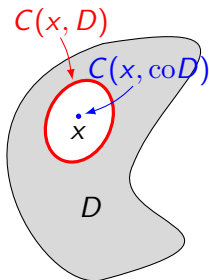
# The Geometry of Visible Sets

## Lemma

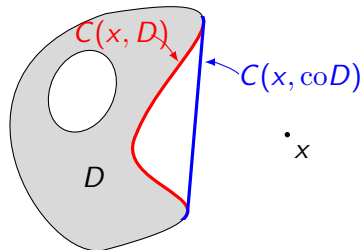
$C(x, D)$  is lower hemi-continuous in  $D$ .

## Lemma

$C(x, \text{co}D) \subseteq \text{co}C(x, D)$ .



(a)  $x \in \text{co}D \setminus D$



(b)  $x \notin \text{co}D$

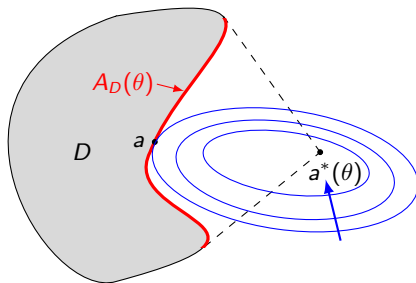


# Admissible Sets = Visible Sets

## Lemma

For any non-empty and compact  $D \subset A$  and  $\theta \in \Theta$  we have

$$\text{cl}[A_D(\theta)] = \text{cl}[C(a^*(\theta), D)]$$



# Existence of Optimal Solution

First, we use the equivalence to prove existence

## Lemma

*The max-min optimal delegation set exists.*

$$\max_{\text{compact } D \subset A} \int_{\Theta} \underline{v}_D(\theta) dF = \max_{\text{compact } D \subset A} \int_{\Theta} \min_{a \in \text{cl}[C(a^*(\theta), D)]} v(a, \theta) dF.$$

Endow  $D$  with Hausdorff metric

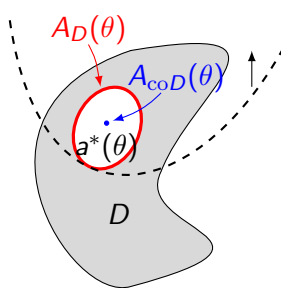
Prove  $\text{cl}[C(a, D)]$  is lower hemi-continuous in  $D$

Prove  $\underline{v}_D(\theta)$  is upper semi-continuous in  $D$

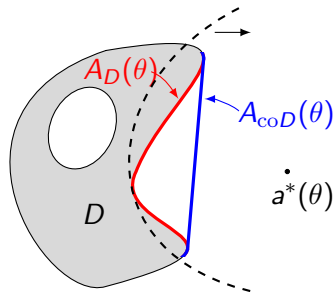
# The Optimality of Convex Set

## Proposition

*Any delegation set  $D$  is weakly out-performed by its convex hull  $\text{co}D$ .*



(a)  $a^*(\theta) \in \text{co}D \setminus D$



(b)  $a^*(\theta) \notin \text{co}D$

# The Optimality of Convex Set

For the reverse direction, denote  $A^* = \{a^*(\Theta)\}$

## Proposition

*Suppose  $A^*$  is convex,  $a^*(\theta)$  admits positive density in  $A^*$ , and the principal's payoff function  $v(\cdot, \theta)$  is strictly quasi-concave for each  $\theta$ . Then, any optimal delegation set is convex.*

# Properties of Optimal Convex Set I

There has to be sufficient discretion for delegation to be valuable

## Lemma

*Suppose that  $a^*(\theta)$  admits positive density in  $A^*$ , and let the delegation set  $D$  be low-dimensional in the sense that  $\dim(\text{aff}(D)) < \dim(\text{aff}(A^*))$ . Then  $D$  is worse than no delegation.*

All of  $D$  is visible to any  $a^*$  outside  $\text{aff}(D)$

# Properties of Optimal Convex Set II

A set is strictly convex if all boundary points are extreme points

## Proposition

*Suppose  $A^* \subset \mathbb{R}^2$  is strictly convex,  $a^*(\theta)$  admits positive density in  $A^*$ , and the principal's payoff function  $v(\cdot, \theta)$  is strictly concave for each  $\theta$ . Then, any optimal delegation set is strictly convex.*

With large uncertainty about how agent makes trade-offs along different dimensions, one linear quota is not optimal

# Examples

After establishing some general properties, more insights from examples

Disk Example: Simplest Delegation, Discretion can be more or less

Cube Example: No Micro-management

One dimension: Interval Delegation, Robustness to more knowledge

# Simplest Delegation: Disk Example

The state space  $\Theta$  is a unit disk  $B(0,1)$  in  $\mathbb{R}^2$

$\theta$  is distributed radially symmetrically with full support, meaning its density  $f$  satisfies:  $f(\theta) = f(\theta') > 0$  if  $|\theta| = |\theta'|$

Agent wants to match the state:  $a^*(\theta) = \theta$

Principal's most preferred action at state  $\theta$  is  $\lambda\theta$  for some  $\lambda \in [0,1]$

$v(a, \theta)$  is given by  $g(|a - \lambda\theta|)$ , where  $g : \mathbb{R}_+ \rightarrow \mathbb{R}$  is a concave, strictly decreasing loss function. For example, when  $g(z) = -z^2$



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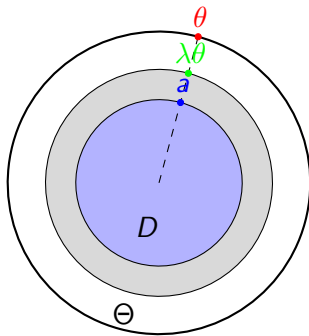
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# Disk: Known Preference

Suppose the agent's preference is known and is given by  $h(|a - \lambda\theta|)$



When  $\lambda < 1$ , Optimal delegation  $D$  is a disk with radius  $r^S(\lambda)$

# How Optimal Radius Changes

- 1) Delegation is tight with known preference  $r^S(\lambda) \leq \lambda$
- 2) Optimal radius  $r^S(\lambda)$  is strictly increasing

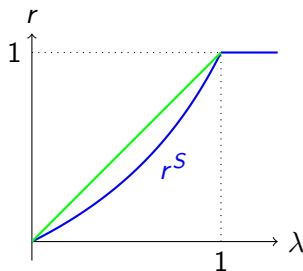


Figure:  $F \sim U(B(0,1))$ :  $r^S = \frac{\lambda}{2-\lambda}$

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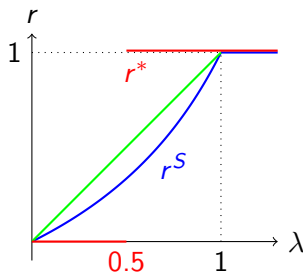
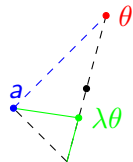
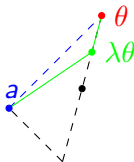
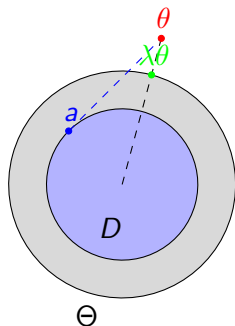


Figure:  $F \sim U(B(0, 1))$ :  $r^S = \frac{\lambda}{2-\lambda}$

# Disk: Unknown Preference

What is the worst action?



# Disk: Robust Design

The max-min solution is bang-bang:

## Proposition

*if  $\lambda < \frac{1}{2}$ ,  $D^* = \{0\}$  is uniquely optimal;*

*if  $\lambda > \frac{1}{2}$ , full delegation  $D^* = A^*$  is uniquely optimal;*

*if  $\lambda = \frac{1}{2}$ ,  $D^*$  is optimal if and only if  $D^* = B(0, r)$  for some  $r \in [0, 1]$ .*

With internal uncertainty, agent may get more or less discretion

Sophisticated constrained delegation is suboptimal

# Full Delegation: Cube Example

The state space  $\Theta$  is the cube  $[0, 1]^{n_1+n_2}$ , with density  $f$

Every state happens with a minimum possibility:  $\exists \gamma \in (0, 1), f(\theta) \geq \gamma \forall \theta$

Agent wants to match the state:  $a^*(\theta) = \theta$ .

Principal's most preferred action  $a_P(\theta)$  satisfies:

$$a_{iP}(\theta) = \theta_i = a_i^*(\theta), \quad \forall i = n_1 + 1, n_1 + 2, \dots, n_1 + n_2$$

Principal's utility  $v(a, \theta) = -\|a - a_P(\theta)\|_q^q$

# Full Delegation: Cube Example

$n_1$  dimension: conflicts of interests       $n_2$  dimension: superficially aligned

## Proposition

*Full delegation is uniquely robust optimal if  $n_2$  is sufficiently larger than  $n_1$ :*

$$\frac{n_2}{n_1} > \frac{(1+q)2^q}{\gamma},$$

Imposing constraints on  $n_1$  dimension triggers distortion on  $n_2$  dimension

Principal refrains from micromanaging agent through restrictions on misaligned decisions when aligned decisions becomes more complex



# The Optimality of Intervals

## Proposition

*Suppose  $\Theta = [\underline{\theta}, \bar{\theta}] \subset \mathbb{R}$ , and  $v(\cdot, \theta)$  is strictly quasiconcave, then any max min optimal delegation set is an interval.*

Finding the optimal interval: optimization over two numbers, simple FOC  
Optimal solution coincides with optimal intervals as Holmstrom (1984)

Justification for interval delegation independent of  $u, F$

# The Optimality of Intervals

If  $\mathcal{U} = \{u\}$ , interval delegation may not be optimal

If  $\mathcal{U} = \mathcal{U}_{qc}$ , interval delegation is always optimal

What if the principal information of  $u$  is in between?

## Proposition

*For any  $\mathcal{U}_1 \subseteq \mathcal{U}_2 \subseteq \mathcal{U}_{qc}$ . If interval delegation is optimal under  $\mathcal{U}_1$ , then interval delegation is also optimal under  $\mathcal{U}_2$ .*

Simple delegation is more desirable if there is more uncertainty.

# Interval Delegation: Supermodularity

If principal knows agent's utility  $u(a, \theta)$  is supermodular (denoted as  $\mathcal{U}_s$ ).

## Proposition

*If the principal's utility  $u(a, \theta)$  is super-modular in  $(a, \theta)$  and concave in  $a$ , and if the uncertainty set is  $\mathcal{U}_s$ , then interval delegation is max min optimal.*

# Extensions I: Money

Money is useless, even if we allow monetary incentives:

$$u(a, \theta, t) = u(a, \theta) + t, \quad u(a, \theta) \in \mathcal{U}$$

If monetary incentive is provided: in the worst case, agent just cares about money and maximizes transfer, which is worse than no delegation

## Extensions II: Partial Identification

We assumed that principal perfectly identifies agent's favorite action  $a^*(\theta)$  in the absence of any constraints.

Now: principal only has partial identification: she only knows agent's favorite action is within  $A^*(\theta)$  in the absence of any constraints.

The characterization of admissible sets  $A_D(\theta)$ , and consequently the optimality of the convex delegation set, generalizes

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The characterization of admissible sets  $A_D(\theta)$ , and consequently the optimality of the convex delegation set, generalizes straightforwardly

## Extensions II: Partial Identification

### Lemma

*Under partial identification, for any non-empty compact  $D \subset A$  and  $\theta \in \Theta$ ,*

$$\text{cl}(A_D(\theta)) = \text{cl}(\cup_{a^*(\theta) \in A^*(\theta)} C(a^*(\theta), D)).$$

*Consequently,*

$$\text{cl}[A_{\text{co}D}(\theta)] \subset \text{co}[\text{cl}(A_D(\theta))].$$

### Proposition

*Under partial identification, for any delegation set  $D$ ,  $\underline{V}_D \leq \underline{V}_{\text{co}D}$  and hence  $\int_{\Theta} \underline{V}_D(\theta) dF \leq \int_{\Theta} \underline{V}_{\text{co}D}(\theta) dF$ . Consequently, there is always a convex max min optimal delegation set.*

# Conclusion

This paper: the principal has limited information on agent's preference

Doesn't know how agent trades off among sub-optimal options

The optimal delegation set is simple: no holes, convex, even full delegation

Our insight: delegation rules are simple because

Uncertainty in how agents exploit/manipulate rules for private interest

Sophisticated rules leave more space for manipulation

**Thank you!**



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