

Robust Delegation

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Introduction

Delegating decision rights to agents is common in organizations

Address a mismatch between authority and information:

The agent is privately informed about the state.

Examples

Fiscal policy, budget allocation

Monopoly regulation, pricing regulation

Healthcare Treatment

Corporate management, investment opportunity

WTO, assessing tariff response to dumping

Introduction

Agent's preference may not be perfectly aligned with the principal

Constraint what the agent can choose may be beneficial

The literature on delegation: what is the optimal delegation set?

Classical Problem in the literature

In one dimension, when is (simple) interval delegation optimal?

Optimal delegation can be sophisticated (digging holes)

Delegation sets in reality:

Often simple delegation rule, sometimes 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

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

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

Model: Max-min Design I

Principal delegates a compact delegation set D to agent.

In state θ , 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: Max-min Design II

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*

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

The max-min design problem is to find delegation set D that solves

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

Model: Uncertainty in Utility I

What is the structure of \mathcal{U} ?

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

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

To fix idea: perfect identification: $A^*(\theta) = a^*(\theta)$

$\mathcal{U} \equiv \{u \mid \text{continuous and strictly quasiconcave in } a,$

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

Across different θ , no restrictions: pick preference for a in each θ

Model Discussion

Asymmetric treatment on two sources of uncertainty: state + preferences

Think about a long-last institution with agents coming and leaving

Preferences are higher dimensional objective $\mathbb{R}^{A \times \Theta}$, hard to measure
(revealed preference) max-min a standard tool for large uncertainty

Better connect with classical delegation literature

Technically hard to do Bayesian design on generic preferences set

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 θ , agents' equilibrium play

Robust Delegation in θ : Li and Hu (2024)

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

Analysis

Two Max-min Formation

The max-min design problem is to find delegation set D that solves

$$\max_D \inf_{u \in \mathcal{U}} \int_{\Theta} v(\tilde{a}(\theta; u, D), \theta) dF(\theta)$$

With \mathcal{U} in our framework, we can prove it is equivalent to

$$\max_D \int_{\Theta} \inf_{u \in \mathcal{U}} [v(\tilde{a}(\theta; u, D), \theta)] dF(\theta)$$

Ex-post non trivial proof on measurability

Use the second one for analysis starting from now

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 θ , i.e.,

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

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

Principal's worst payoff in state θ 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 θ 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

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

Lemma (Existence)

The robust delegation problem (1) has a solution.

What are properties of optimal robust delegation?

Visible Set

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

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

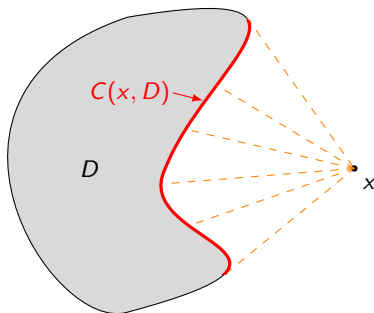
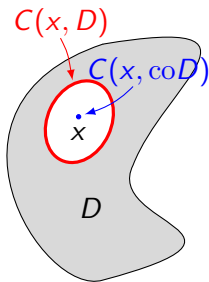


Figure: Illustration of visible set

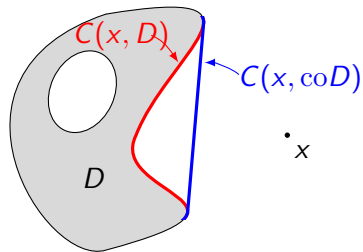
The Geometry of Visible Sets

Lemma

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



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



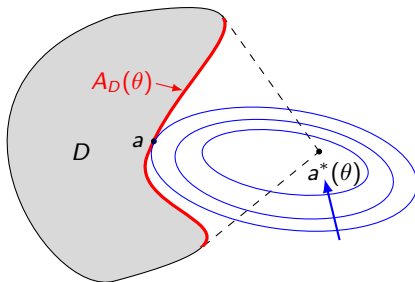
(b) $x \notin \text{co}D$

The Geometry of Admissible Sets

Lemma

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

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



The Geometry of Admissible Sets

Combining two previous lemmas we get

Lemma

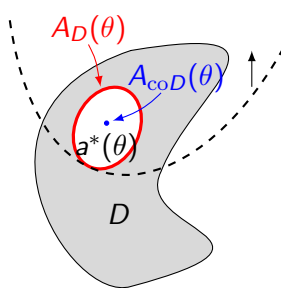
$$\text{cl}(A_{\text{co}D}(\theta)) \subseteq \text{co}[\text{cl}(A_D(\theta))]$$

$$\begin{aligned}\text{cl}(A_{\text{co}D}(\theta)) &= \text{cl}[\cup_{a \in A^*(\theta)} C(a, \text{co}D)] \\ &\subseteq \text{cl}[\cup_{a \in A^*(\theta)} \text{co}C(a, D)] \\ &\subseteq \text{cl}[\text{co}[\cup_{a \in A^*(\theta)} C(a, D)]] \\ &= \text{co}[\text{cl}(A_D(\theta))]\end{aligned}$$

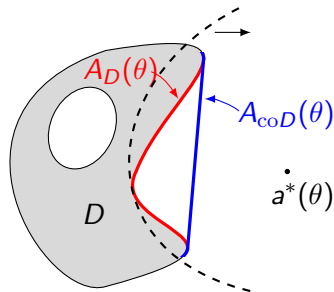
The Optimality of Convex Set

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 any optimal set to be convex, suppose $a^*(\theta) = A^*(\theta)$, and denote $A^* = \{a^*(\Theta)\}$

Proposition

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

Counterexample if A^* is not convex: aligned agent

Properties of Optimal Convex Set I

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

Proposition

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

Properties of Optimal Convex Set II

There has to be sufficient discretion for delegation to be valuable

Lemma

Suppose that $a^(\theta) = A^*(\theta)$, $a^*(\theta)$ 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 $\text{aff}(D)$

Examples

After establishing some general properties, more insights from examples

One dimension: Interval Delegation (skip)

Disk Example: Full delegation or No delegation

Cube Example: Full delegation and High dimensionality

Simplest Delegation: Disk Example

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

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

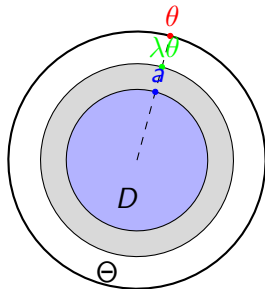
Agent wants to match the state: $a^*(\theta) = \theta$

Principal's most preferred action at state θ 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 $g(z) = -z^2$

Disk: Known Preference

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



P and A are aligned in “direction”, misaligned in “distance”

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

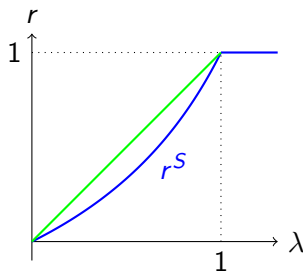


Figure: $F \sim U(B(0, 1))$: $r^S = \frac{\lambda}{2-\lambda}$

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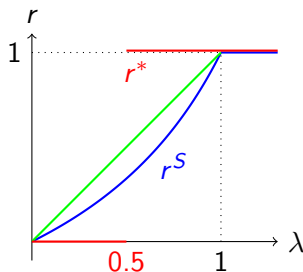
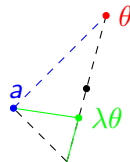
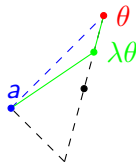
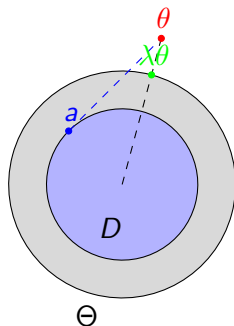


Figure: $F \sim U(B(0, 1))$: $r^S = \frac{\lambda}{2-\lambda}$

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 $r \in [0, 1]$.*

P and A are superficially aligned in “direction”, misaligned in “distance”

Constraining only “distance” is not robustly optimal

as it triggers unpredicted deviation in “direction”

Full Delegation: Cube Example

The state space Θ 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

Extensions I

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

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

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