

Contract Design Under Uncertainty

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Joint work with Paul Dütting, Yingkai Li, and Inbal Talgam-Cohen

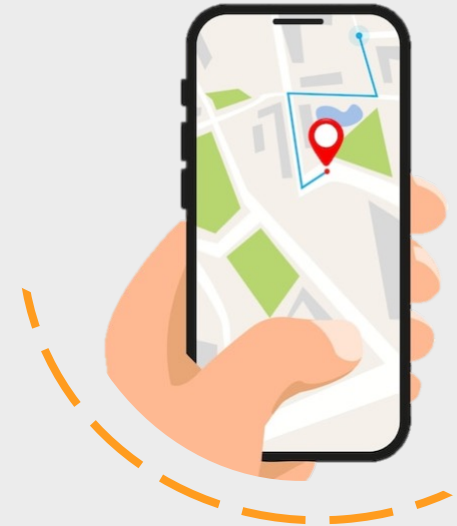
Google - Mountain View (Global HQ)

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Modern Algorithms in Society

- Interactions with **self-interested** individuals
- Challenges beyond **computational** tractability
- Take **incentives** into account



Social Media Marketing

- **Paid Ads** - in the social platform feed
- **Influencer Marketing** - a brand hires popular users



Paid Ads

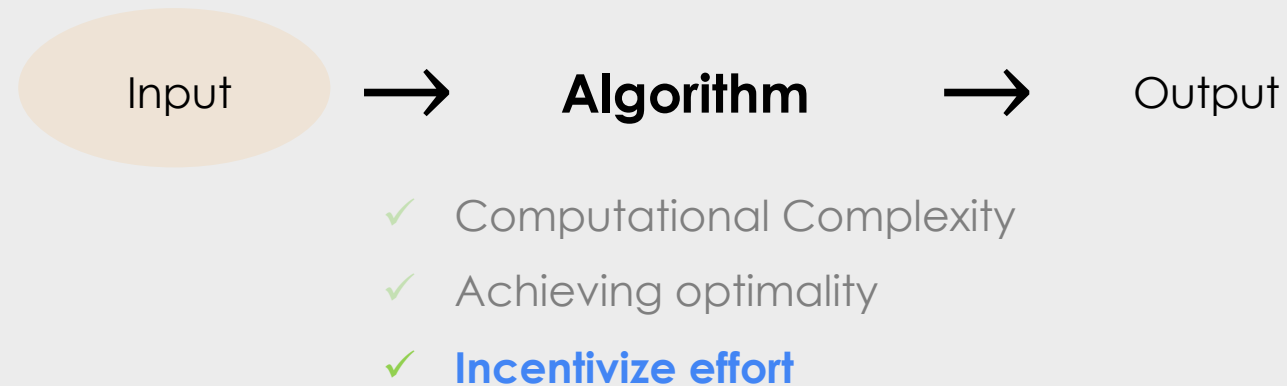
- Algorithmic **auction** - advertisers bid → allocation and payments



- ✓ Computational Complexity
- ✓ Achieving optimality
- ✓ **Incentivize truthfulness**

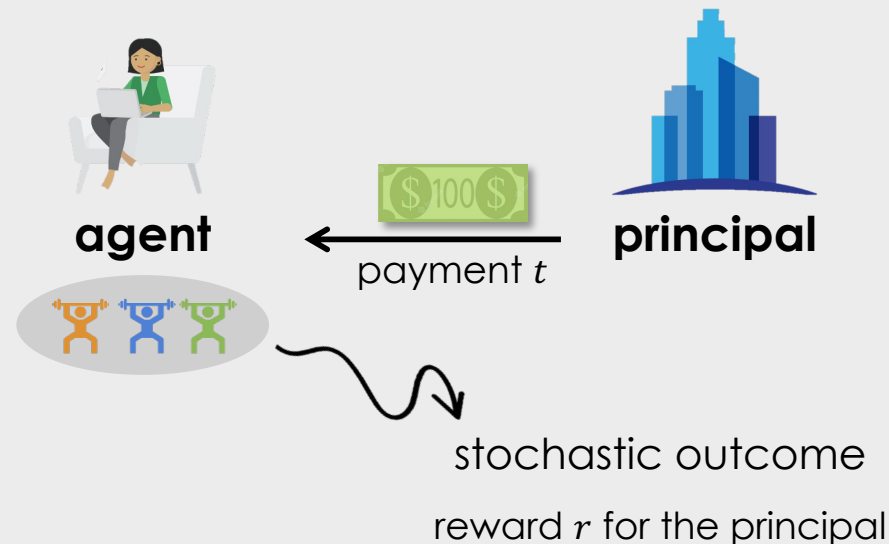
Influencer Marketing

- The algorithm determines a **contract** to incentivize effort



The Principal-Agent Problem [GH83]

- **Moral hazard** - the agent's actions cannot be observed
- **Objective:** a contract maximizing expected **rewards minus payment** $\mathbb{E}[r - t]$



Research Agenda

Generalizations of the classic model:

- **Personalization** for participants from diverse population (with **types**)
- **Multilateral** contracts involving multiple principals or agents
- The need for **simple** contracts

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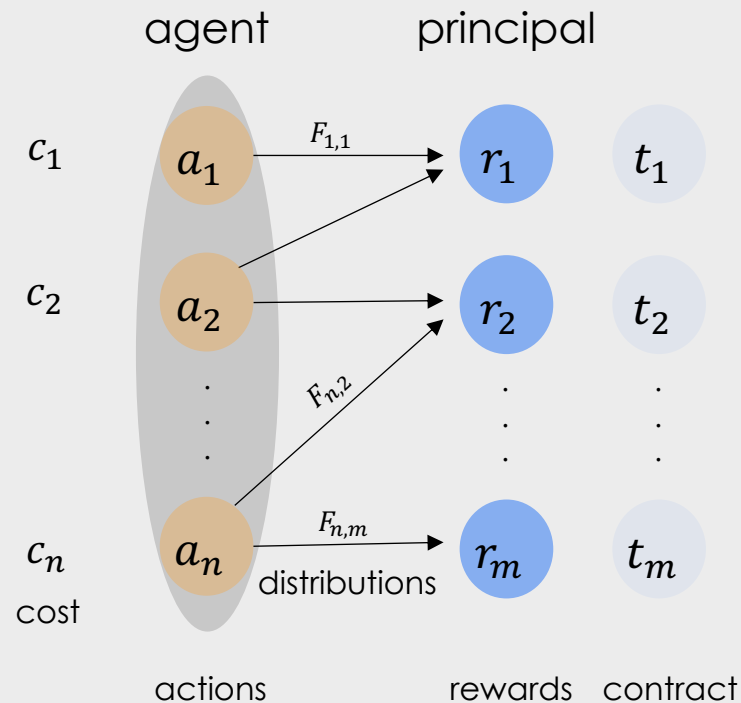
Outline

- **Single-parameter** model of types [ADT.C. EC'21]
 - Motivated by single-parameter auction design
- **Characterization** of the design space [ADT.C. EC'21]
- Counter-intuitive and **undesirable** properties of **optimal contracts** [ADLT.C. EC'23]
- **Linear contracts** (aka commission-based) are near-optimal [ADLT.C. EC'23]

Recent works on contracts with types. Myerson (1982), Guruganesh et al. (2021), Castiglioni et al. (2021), Gottlieb and Moreira (2022), Casto-Pire et al. (2022).

The Model

- The classic principal-agent model by Grossman-Hart (1983)

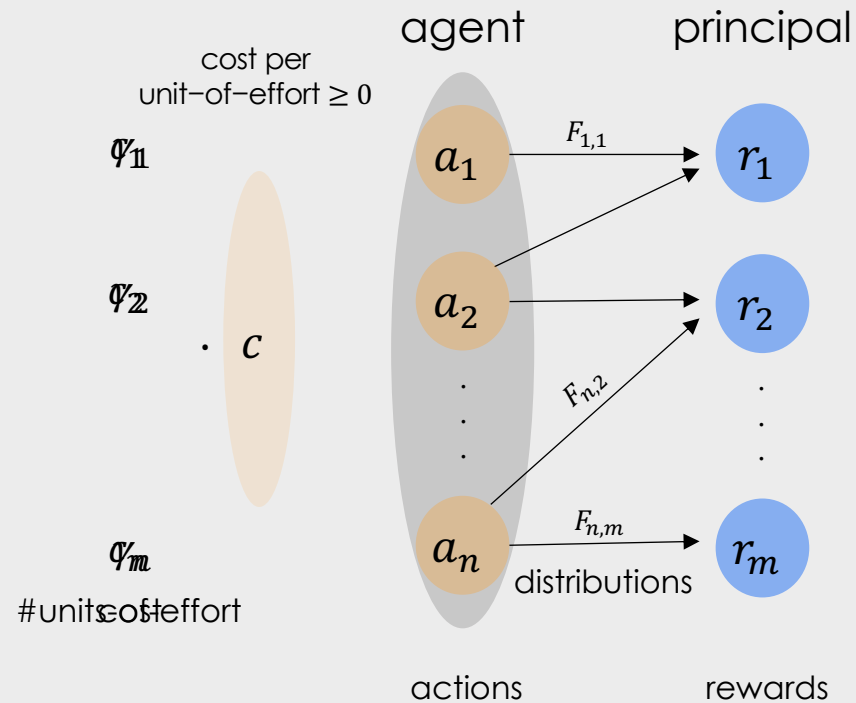


Notation. $T_i = \mathbb{E}_{j \sim F_i}[t_j]$, $R_i = \mathbb{E}_{j \sim F_i}[r_j]$

- Agent's **action i^*** maximizes $T_i - c_i$
- Principal's **revenue** $R_{i^*} - T_{i^*}$

The Model

- Type is drawn from G with density g supported on $\mathcal{C} = [\underline{c}, \bar{c}]$



The Model

- A **contract** (x, t) consists of two mappings:
 - **Payment rule** $t: \mathcal{C} \rightarrow \mathbb{R}^m$ from types to a **paymet** scheme
 - **Allocation rule** $x: \mathcal{C} \rightarrow [n]$ from types to an **action** recommendation

Notation. $T_i^c = \mathbb{E}_{j \sim F_i}[t(c)_j]$

- **Agent** c report \hat{c} and action $i^*(c)$ maximize $T_{i^*(c)}^{\hat{c}} - \gamma i^*(c)c$
- A contract (x, t) is **incentive compatible (IC)** if $c = \hat{c}$ and $x(c) = i^*(c)$
- **Principal's contract** maximizes $\mathbb{E}_{c \sim G}[R_{x(c)} - T_{x(c)}^c]$ s.t. IC
- **Welfare** sum of utilities $\mathbb{E}_{c \sim G}[R_{i^*(c)} - \gamma i^*(c)c]$

Design Space Characterization

Definition. Allocation rule x is **implementable** if exists payment rule t s.t. (x, t) is IC

Q. What do implementable allocation rules look like?

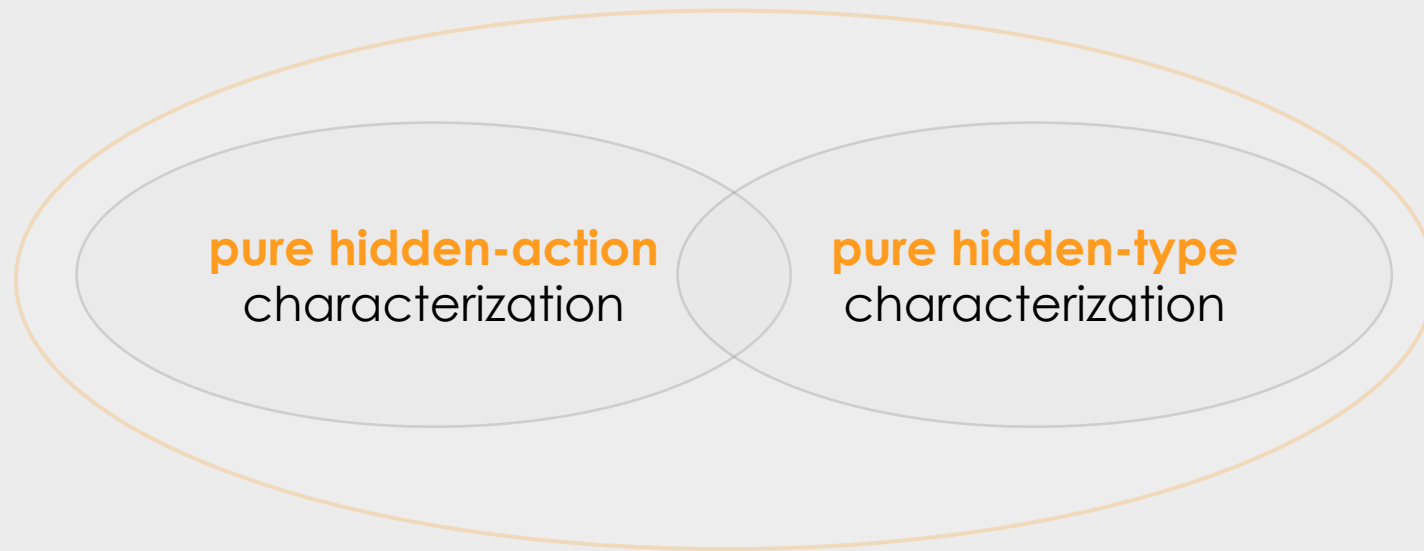
pure hidden-action
characterization

pure hidden-type
characterization

Design Space Characterization

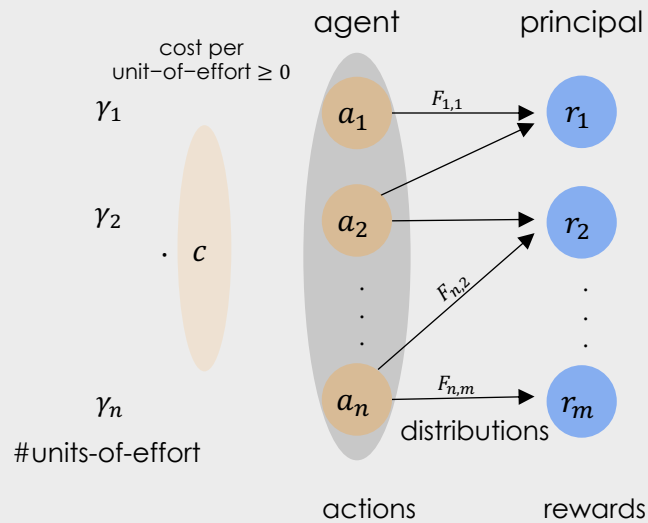
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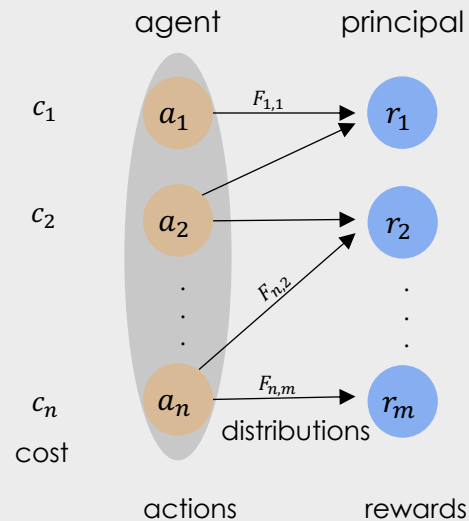
Pure hidden action



Design Space Characterization

Pure hidden action

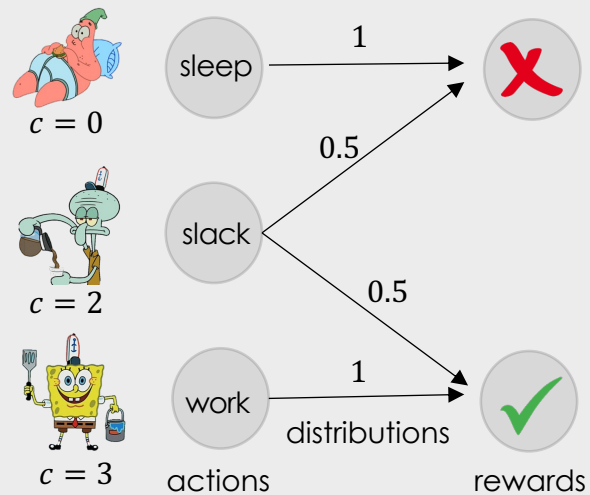
- x is **implementable** \Leftrightarrow exists no **deviation scheme** λ_k s.t. (1) dominant distribution $\sum_k \lambda_k F_k \geq F_x$
(2) strictly lower cost $\sum_k \lambda_k c_k < c_x$



Design Space Characterization

Pure hidden action

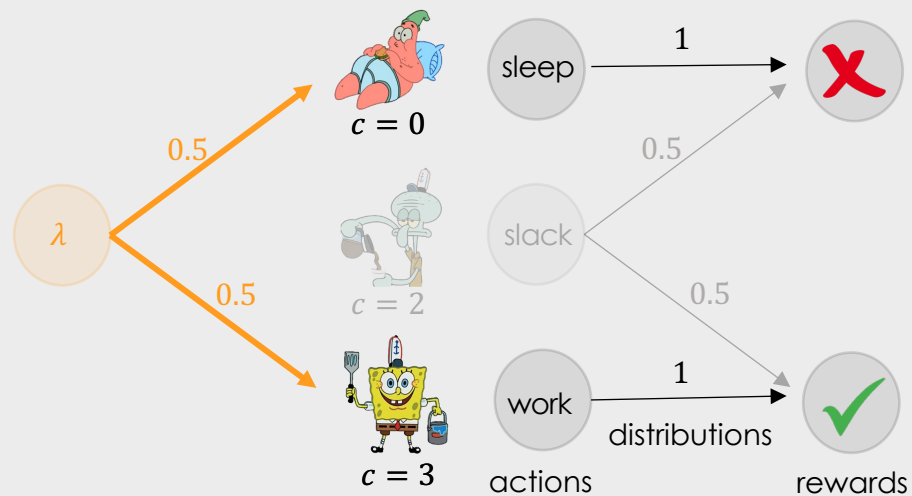
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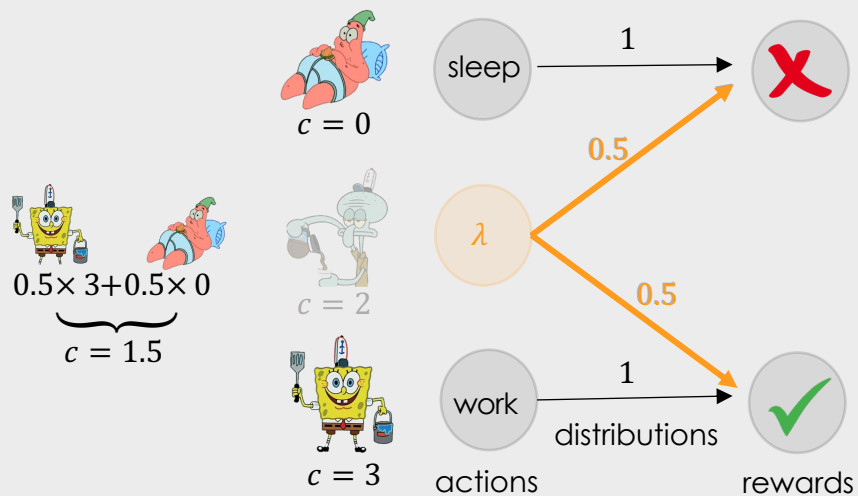


is not implementable

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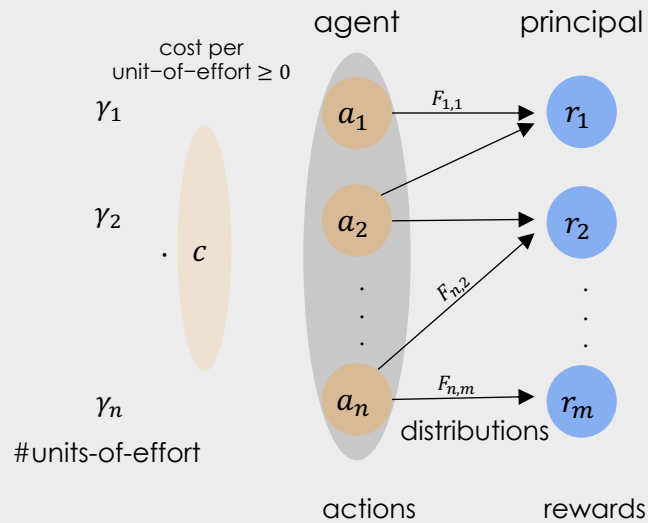


is not implementable

- LP duality approach

Design Space Characterization

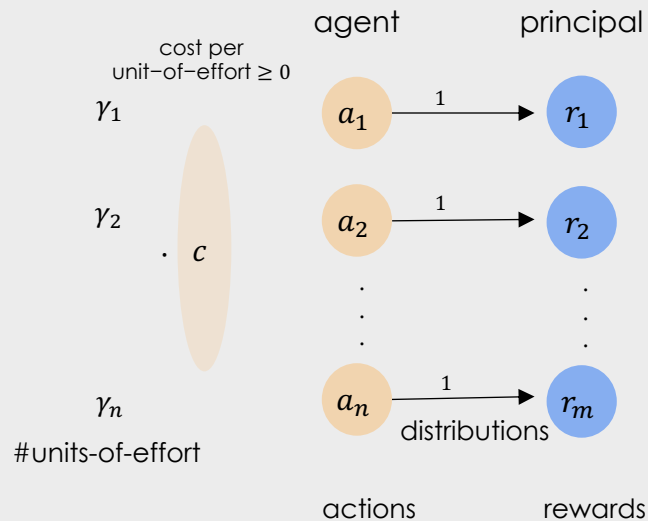
Pure hidden type



Design Space Characterization

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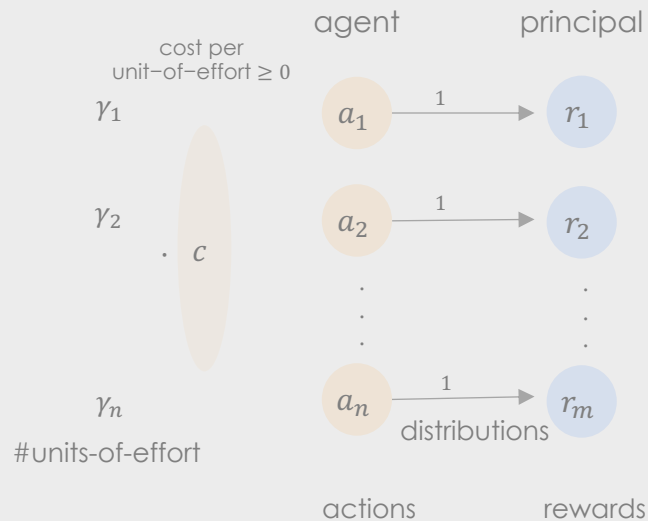
- x is **implementable** $\Leftrightarrow x$ is **monotone** [Myerson 1981]



Design Space Characterization

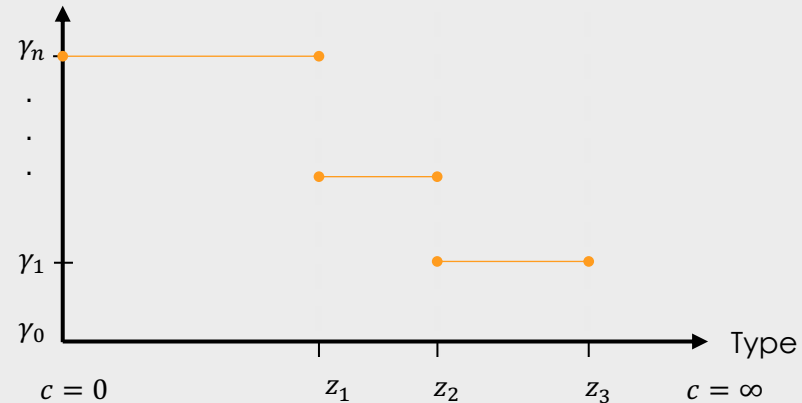
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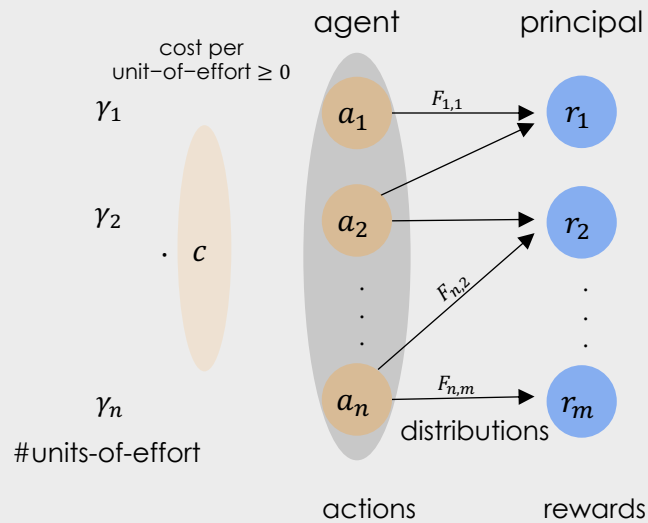
Monotone allocation rule $x: \mathcal{C} \rightarrow [n]$

Units-of-Effort



Design Space Characterization

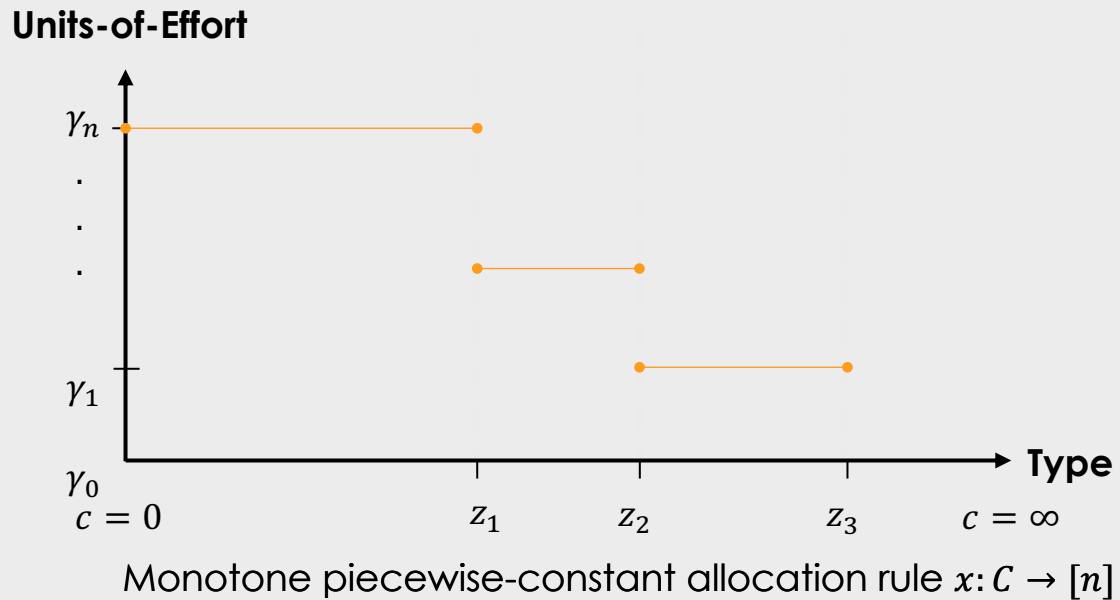
Hidden type and hidden action



Design Space Characterization

Hidden type and hidden action

Proposition [ADT EC'21]. If x is implementable, it is **monotone**



Design Space Characterization

Theorem [ADT EC'21]. x **implementable** \Leftrightarrow exists no **deviation scheme** $\lambda_{(z,k)}$ s.t.

(1) dominant sum of distributions $\sum_{z,k} \lambda_{(z,k)} F_k \geq \sum_z F_{x(z)}$

(2) strictly lower joint cost $\sum_{z,k} \lambda_{(z,k)} \gamma_k^z < \sum_z \gamma_{x(z)}^z$

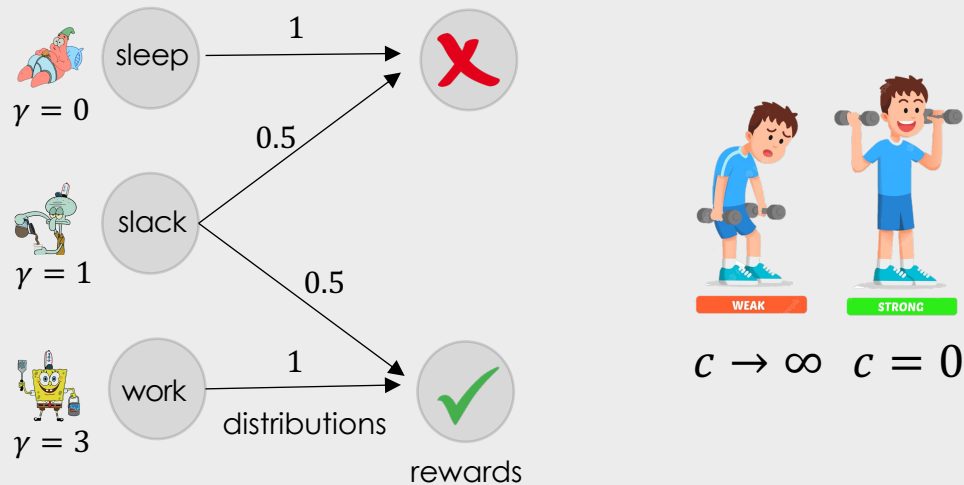
Corollary [ADT EC'21]. Optimal contract is **polytime computable** for const #actions

Hardness for constant #actions in the multi-parameter model [Guruganesh-Schneider-Wang'21]

Example

Theorem [ADT EC'21]. x **implementable** \Leftrightarrow exists no **deviation scheme** $\lambda_{(z,k)}$ s.t.

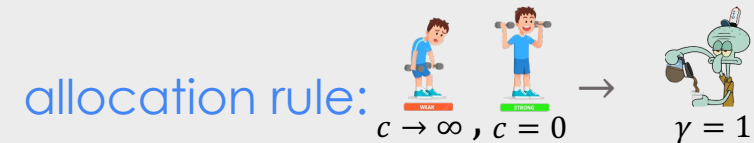
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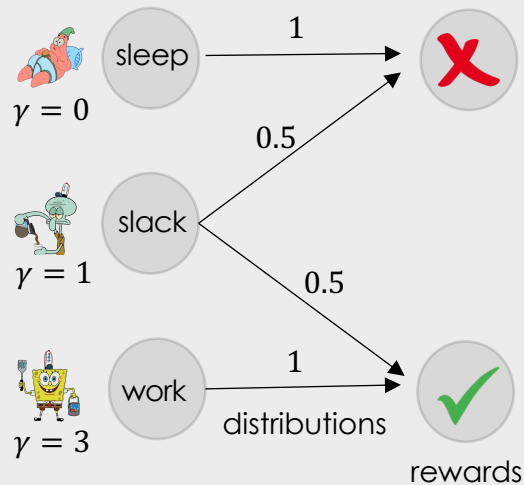
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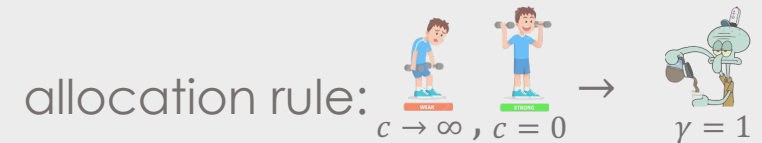
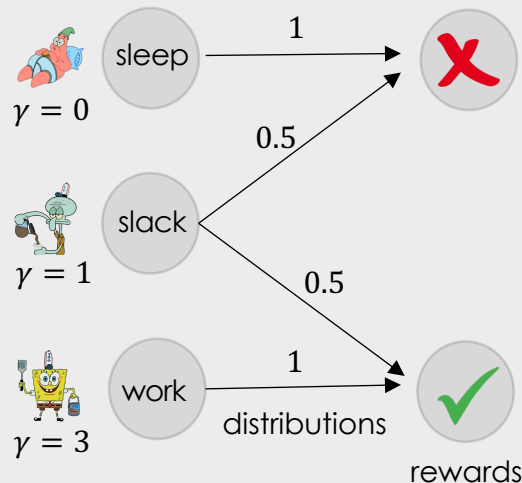
- (1) distributions sum (1,1)
- (2) joint cost ∞



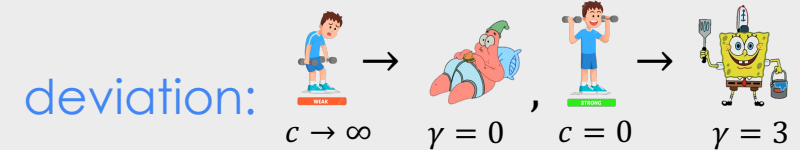
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- (1) distributions sum (1,1)
- (2) joint cost 0

Optimal Contracts and Their Issues

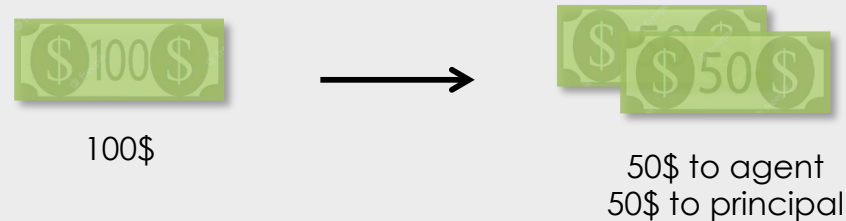
- Informational requirements, extensive analysis ,etc.
- Unintuitive, e.g., non-monotonicity in rewards [DRT EC'19]

Theorem [ADLT EC'23]. In the single dimensional typed model

- Large **menu-size complexity**
- Revenue **non-monotonicity** w.r.t type distribution

Simple Contracts

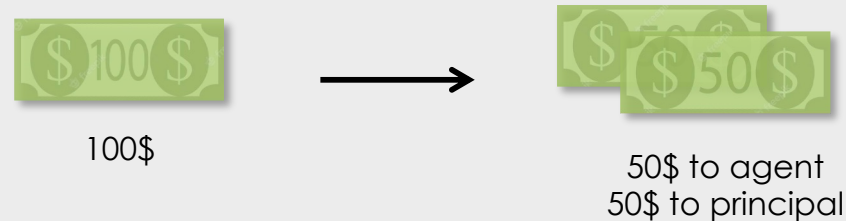
- In a **linear contract**, the principal offers a fixed share $\alpha \in [0,1]$ of the rewards



“It is probably **the great robustness of linear rules** based on aggregates **that accounts for their popularity**. That point is not made as effectively as we would like by our model; we suspect that it cannot be made effectively in any traditional Bayesian model.” [Milgrom and Holmstrom 1987]

Simple Contracts

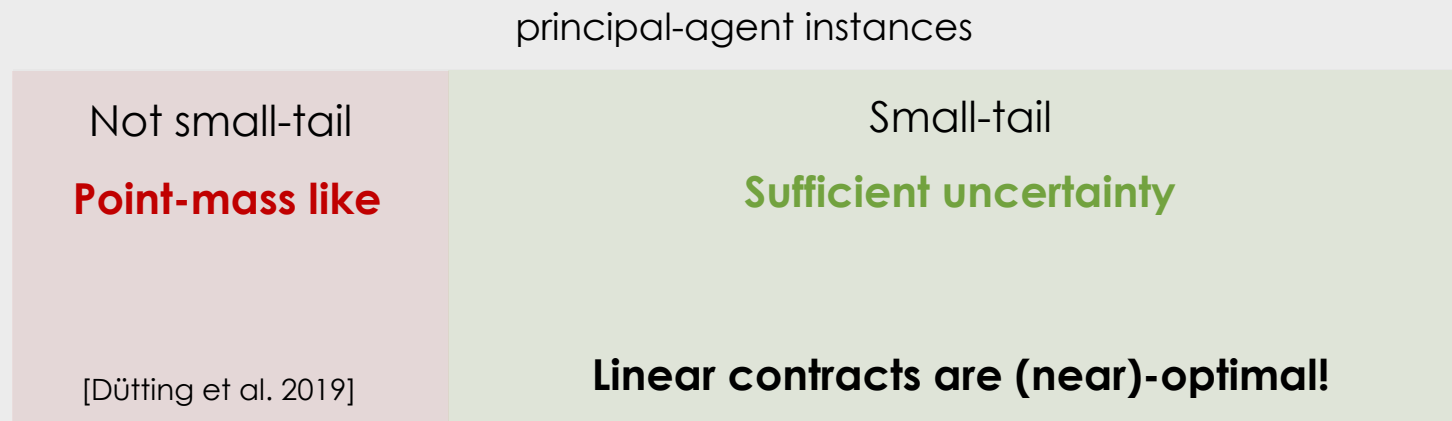
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- **Robustness of linear contracts.** Carroll (2015), Dütting et al. (2019), Yu and Kong (2020), Dai and Toikka (2022), Walton and Carroll (2022)
- **Approximation of linear contracts.** Dütting et al. (2019), Castiglioni et al. (2021), Guruganesh et al. (2021)

Near-Optimality of Linear Contracts

- $\theta(n)$ separation for **point-mass** distributions [DRT EC'19]
 - Boundary case
- **Approximately optimal** with **sufficient uncertainty**
 - The small-tail assumption



Near-Optimality of Linear Contracts

Theorem [ADLT EC'23]. Revenue benchmark:

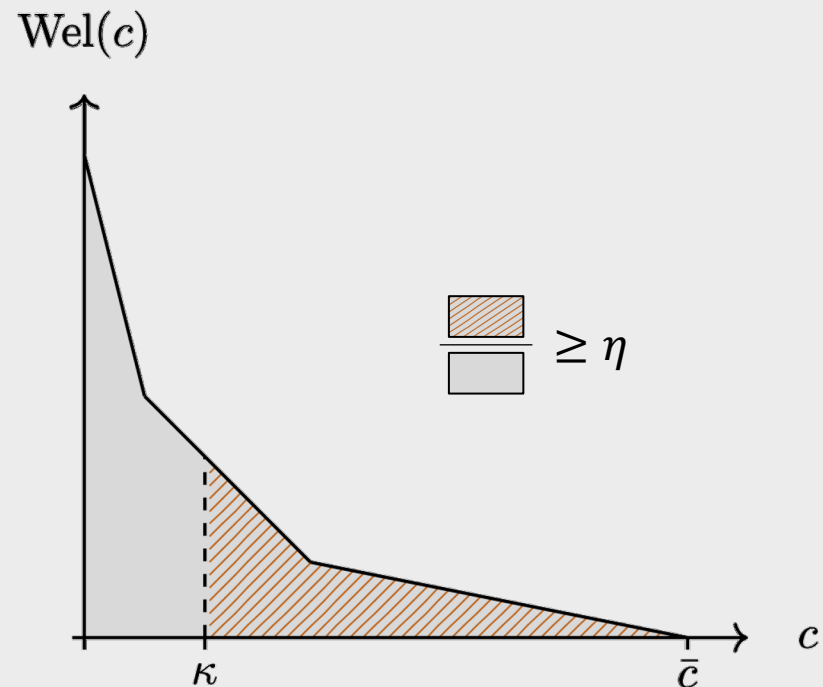
- **3-approximation** for **normal** $\mathcal{N}(\mu, \sigma^2)$ truncated at $c = 0$ with $\sigma \geq 5\eta/2\sqrt{2}$
- **2-approximation** for **uniform** $U[0, \bar{c}]$
 - **Optimal** when $i^*(r, \bar{c}) = 0$
- **2-approximation** for **decreasing densities** (e.g., exponential)

- Constant approximation w.r.t optimal welfare benchmark [ADLT EC'23]

The Small-Tail Assumption

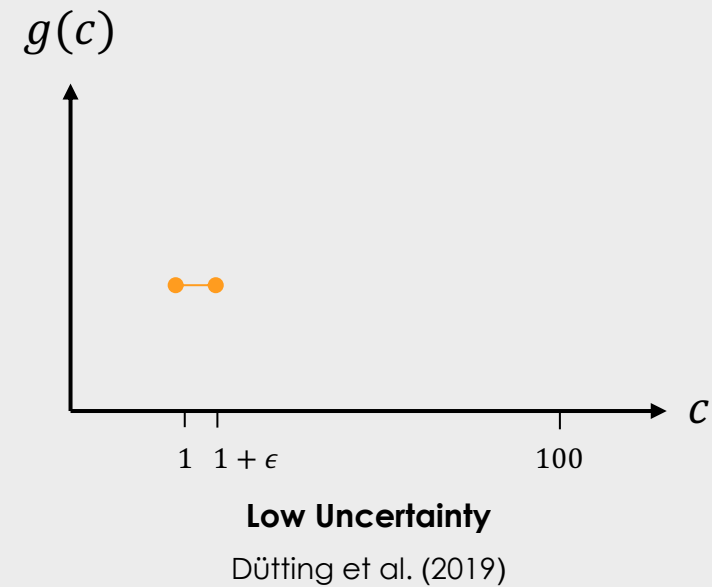
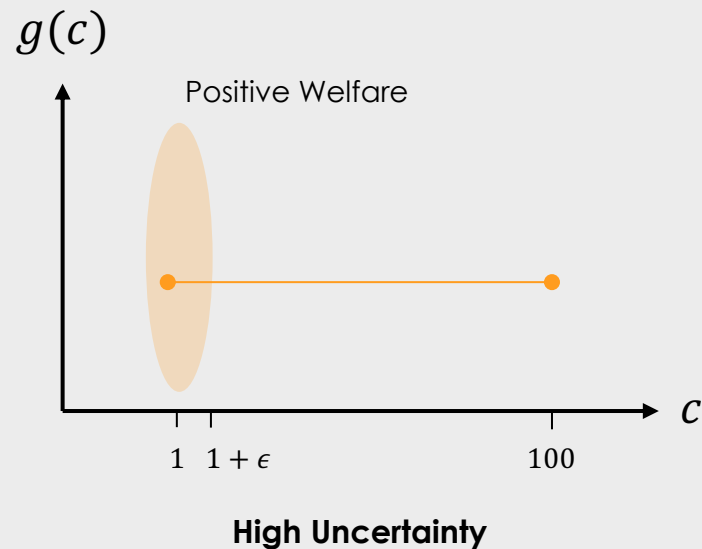
Definition [ADLT EC'23]. Let $\kappa \in [\underline{c}, \bar{c}]$, $\eta \in [0,1]$.

An instance is **(κ, η) -small-tail** if $\text{Wel}_{[\kappa, \bar{c}]} \geq \eta \text{Wel}_{[\underline{c}, \bar{c}]}$



The Small-Tail Assumption

Depends on the entire principal-agent setting



Universal Approximation Guarantee

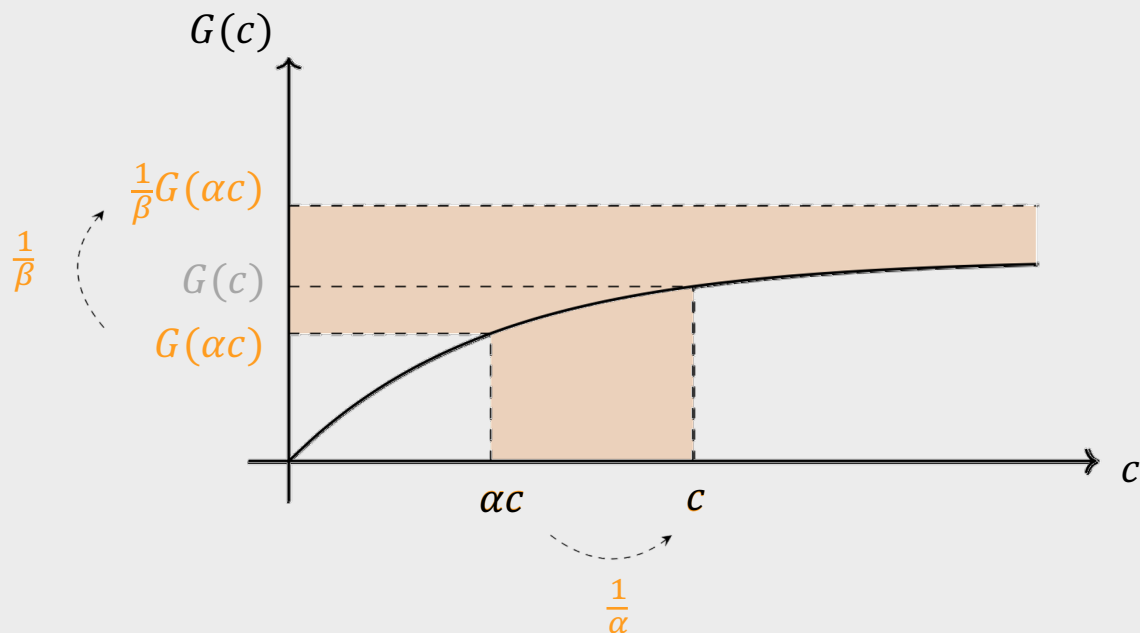
Theorem [ADLT EC'23]. Let $q \in (0,1)$ and $G(c_q) = q$. If for $\alpha, \eta \in (0,1)$ the settings is $(\frac{c_q}{\alpha}, \eta)$ -small-tail then linear contract α is at least $(1 - \alpha)\eta q$ of the **optimal welfare**

Slowly-Increasing Distributions

- Applies to any CDF and captures its **rate of increase**
- Parametric **approximation** of linear contracts
- Any distribution is slowly-increasing for **some parameters**

Slowly-Increasing Distributions

Definition [ADLT EC'23]. Let $\alpha, \beta \in (0,1)$, and $\kappa \in [\underline{c}, \bar{c}]$. A distribution G is (α, β, κ) -**slowly-increasing** if $G(c) \leq \frac{1}{\beta} G(\alpha c) \forall \kappa \leq c$



Approximation for Slowly Increasing

Theorem [ADLT EC'23]. Let $\alpha, \beta, \eta \in (0,1)$, and $\kappa \in [\frac{c}{\alpha}, \bar{c}]$.

Under (α, β, κ) -slowly-increasing and (κ, η) -small-tail
linear contract α is $(1 - \alpha)\beta\eta$ of the optimal welfare

Proof Idea for Slowly Increasing

Step 1. Revenue of linear contract α is at least $1 - \alpha$ of its welfare

$$\begin{array}{ccc} \text{revenue of linear contract} & & \text{welfare of linear contract} \\ \mathbb{E}_{c \sim G}[(1 - \alpha)R_{x(c)}] & \geq (1 - \alpha) \times & \mathbb{E}_{c \sim G}[R_{x(c)} - \gamma_{x(c)}c] \end{array}$$

$$\text{linear contract revenue} \geq (1 - \alpha) \times \text{linear contract welfare} \geq (1 - \alpha) \times \beta \times \text{optimal welfare}$$

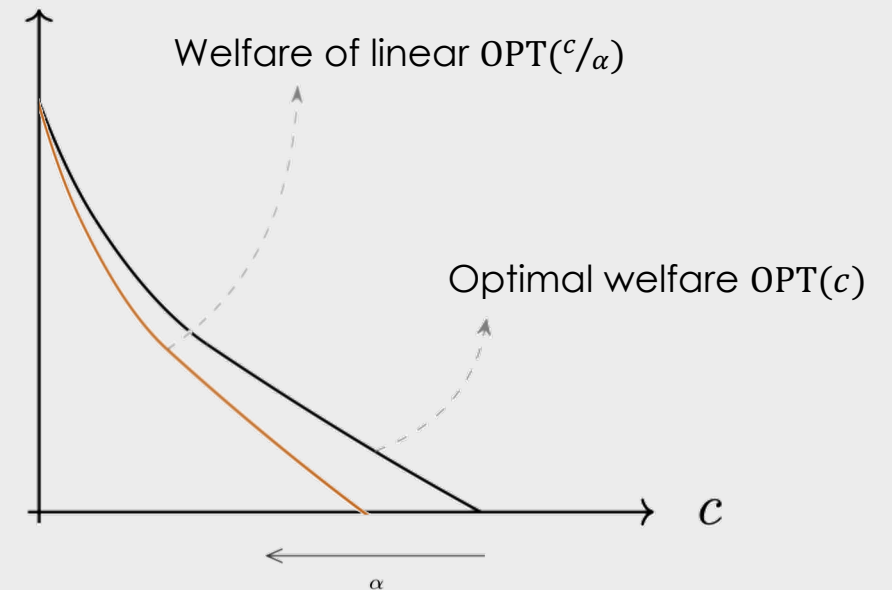
Proof Idea for Slowly Increasing

Step 2. welfare of linear contract α is at least β of the optimal welfare

- When maximizing welfare, the agent maximizes $R_i - \gamma_i c$
- For linear contract α :

The agent maximizes $\alpha R_i - \gamma_i c$ which is $R_i - \gamma_i \frac{c}{\alpha}$

$$\text{linear contract revenue} \geq (1 - \alpha) \times \text{linear contract welfare} \geq (1 - \alpha) \times \beta \times \text{optimal welfare}$$



Summary and Future Directions

- **Single-parameter** model of types
- **Characterization** of the design space
- Counter-intuitive and **undesirable** properties of **optimal contracts**
- **Linear contracts** are near-optimal

Future directions:

- Other **forms of simple contracts** that are near-optimal
- Contracts that involve **multiple agents**
- **Applications** of this theory

Thank You!

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