

Equal-Pay Contracts

Tomek Ponitka

Tel Aviv University

Joint work with



Michal Feldman

Tel Aviv University
Microsoft ILDC



Yoav Gal-Tzur

Tel Aviv University



Maya Schlesinger

Tel Aviv University

Contract Design

manager at paper company



Principal



owns a project and receives its **reward**.



revenue from selling paper

team of salespeople



Agents



affect the **expected reward** through their **costly actions**.



thorough customer research

per-sale commission



Contract

incentivizes agents to take **actions** by offering them **shares of reward**.

Model

Known: n agents. Each agent i has set of **individual actions** A_i .
 $f : 2^A \rightarrow \mathbb{R}_{\geq 0}$ maps **taken actions** S to **expected reward** $f(S)$.



a

b

c

A_1



d

e

A_2



f

g

h

i

A_3



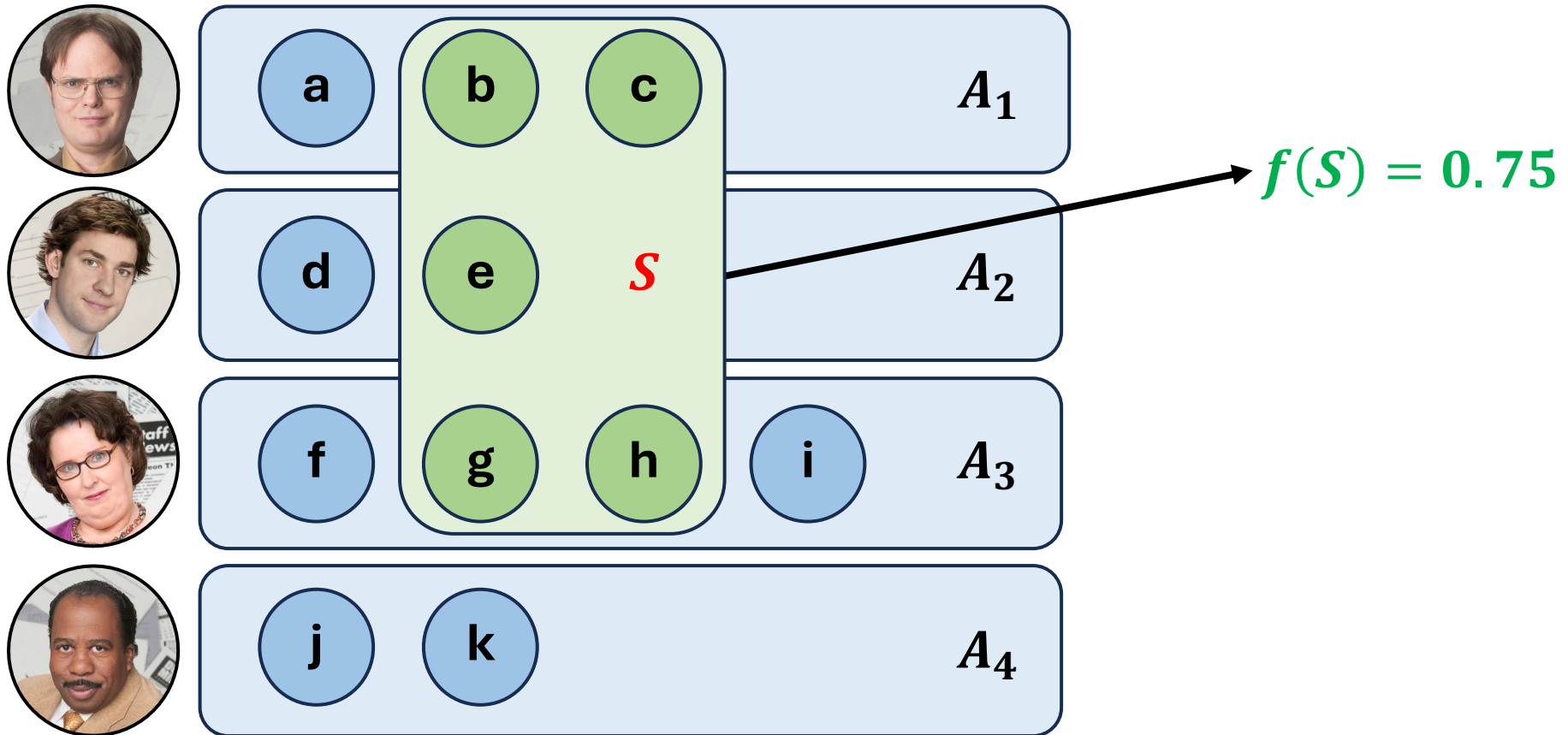
j

k

A_4

Model

Known: n agents. Each agent i has set of **individual actions** A_i .
 $f : 2^A \rightarrow \mathbb{R}_{\geq 0}$ maps **taken actions** S to **expected reward** $f(S)$.



Model

Known: n agents. Each agent i has set of **individual actions** A_i .
 $f : 2^A \rightarrow \mathbb{R}_{\geq 0}$ maps **taken actions** S to **expected reward** $f(S)$.

Step 1. Principal commits to a **contract** $\vec{\alpha} = (\alpha_1, \dots, \alpha_n) \in [0, 1]^n$.
offering agent i an α_i -fraction of the reward.



$$\alpha_1 = 0.1$$



$$\alpha_2 = 0.2$$



$$\alpha_3 = 0.3$$



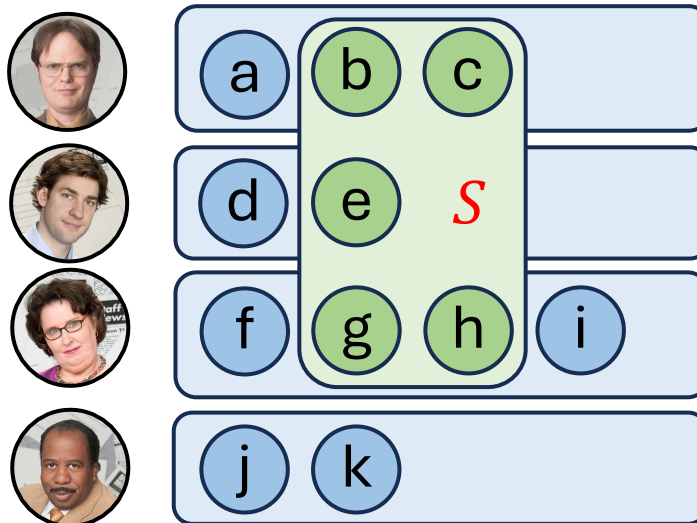
$$\alpha_4 = 0$$

Model

Known: n agents. Each agent i has set of **individual actions** A_i .
 $f : 2^A \rightarrow \mathbb{R}_{\geq 0}$ maps **taken actions** S to **expected reward** $f(S)$.

Step 1. Principal commits to a **contract** $\vec{\alpha} = (\alpha_1, \dots, \alpha_n) \in [0, 1]^n$.
offering agent i an α_i -fraction of the reward.

Step 2. Each agent i strategically chooses a **subset** of actions to take.
Agents form a **pure Nash equilibrium** $S \subseteq A$.



Model

Known: n agents. Each agent i has set of **individual actions** A_i .
 $f : 2^A \rightarrow \mathbb{R}_{\geq 0}$ maps **taken actions** S to **expected reward** $f(S)$.

Step 1. Principal commits to a **contract** $\vec{\alpha} = (\alpha_1, \dots, \alpha_n) \in [0, 1]^n$.
offering agent i an α_i -fraction of the reward.

Step 2. Each agent i strategically chooses a **subset** of actions to take.
Agents form a **pure Nash equilibrium** $S \subseteq A$.



Agent i 's utility:

$$\alpha_i \cdot f(S) - \sum_{a \in S_i} c_a$$

Principal's utility:

$$(1 - \sum_{i=1}^n \alpha_i) \cdot f(S)$$

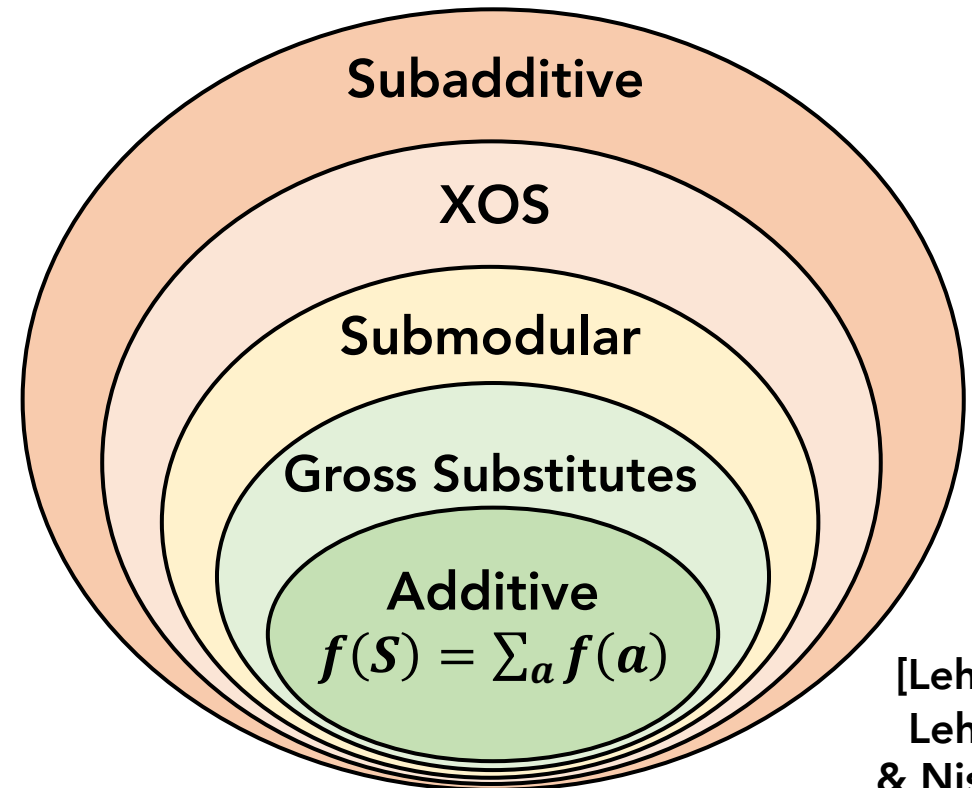
All About f

The function $f : 2^A \rightarrow \mathbb{R}_{\geq 0}$ has exponential representation size.
We access f via either:

Value Queries
Input: $S \subseteq A$
Output: $f(S)$

Demand Queries
Input: $p \in \mathbb{R}^A$
Output: $S \in \operatorname{argmax}_{S \subseteq A} (f(S) - \sum_{a \in S} p_a)$

This **hierarchy** of set functions captures the combinatorial structure of $f : 2^A \rightarrow \mathbb{R}_{\geq 0}$.



[Lehmann,
Lehmann
& Nisan '01]

Optimization Problem

maximize

$$\text{principal's utility} \\ (1 - \sum_{i=1}^n \alpha_i) \cdot f(S)$$

subject to **Nash equilibrium** constraints

Substantial interest in **efficient algorithms** for finding near-optimal contracts under different **structural assumptions** on f .

[Babaioff, Feldman, Nisan, EC 2006 & WINE 2006 & SAGT 2009]

[Dütting, Ezra, Feldman, Kesselheim, STOC 2023 & SODA 2025]

[Deo-Campo Vuong, Dughmi, Patel, Prasad, SODA 2024]

[Ezra, Feldman, Schlesinger, ITCS 2024]

[Alon, Castiglioni, Chen, Ezra, Li, Talgam-Cohen, EC 2025]

[Aharoni, Hoefer, Talgam-Cohen, EC 2025]

[Castiglioni, Chen, Li, 2025 & 2025]

[Feldman, Gal-Tzur, Ponitka, Schlesinger, EC 2025 & ITCS 2026]

[Doron-Arad, Shachnai, Shmerler, Talgam-Cohen, **EC 2026**]

[Dütting, Ezra, Feldman, Kesselheim, **EC 2026**]

[Ding, Li, Sun, 2026]

[Lavi, Shachnai, Talgam-Cohen, 2026]

Pay Gap

maximize

$$\text{principal's utility} \\ (1 - \sum_{i=1}^n \alpha_i) \cdot f(S)$$

subject to **Nash equilibrium** constraints

Problem: Optimal contracts might induce a large **pay gap**.

Discriminatory
Payments

$$\alpha_1 = 0.5$$



$$\alpha_2 = 0.05$$



$$\alpha_3 = 0.005$$



$$\alpha_4 = 0$$



Conflicts with fairness desiderata, institutional rules, legal constraints.

Equal Pay Act of 1963





March 17, 2016

Finance Committee approves bill limiting the pay of executives in financial corporations

Share:



In an unprecedented measure, the Knesset Finance Committee unanimously approved Wednesday evening a bill that would lower the maximum salary ceiling for executives in financial corporations from NIS 3.5 million to NIS 2.5 million a year, or 35 times the salary of the lowest-paid worker, whichever is lower.



Equal-Pay Contracts

A contract $\alpha = (\alpha_1, \dots, \alpha_n)$ is **equal-pay** if $\alpha_i = \alpha_j$ for agents i, j with $\alpha_i, \alpha_j \neq 0$.

Equal
Payments

$$\alpha_1 = 0.2$$



$$\alpha_2 = 0.2$$



$$\alpha_3 = 0.2$$



$$\alpha_4 = 0$$



All **hired** agents
receive the same payment.

In the paper: A contract $\alpha = (\alpha_1, \dots, \alpha_n)$ is **γ -equal-pay** if $\alpha_i \geq \gamma \cdot \alpha_j$ for agents i, j with $\alpha_i, \alpha_j \neq 0$.

Research Direction

Question 1 (Computational Aspects):

Can we efficiently compute nearly-optimal **equal-pay** contracts?

Question 2 (Price of Equality):

How much utility is lost by restricting contracts to be **equal-pay** relative to **unconstrained** contracts?

Our Contributions

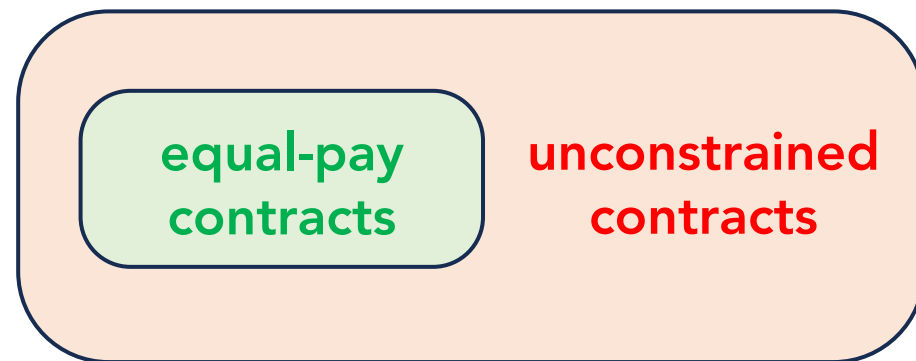
1. We completely characterize the **computational complexity** of **equal-pay** contracts.

2. We obtain tight bounds for the **price of equality** of $\Theta(\log n / \log \log n)$.

3. We resolve two open problems for **computational complexity** of **unconstrained** contracts.

Hardness Results

Equal-pay contracts provide a useful lens for studying **computational complexity** of **unconstrained** contracts.

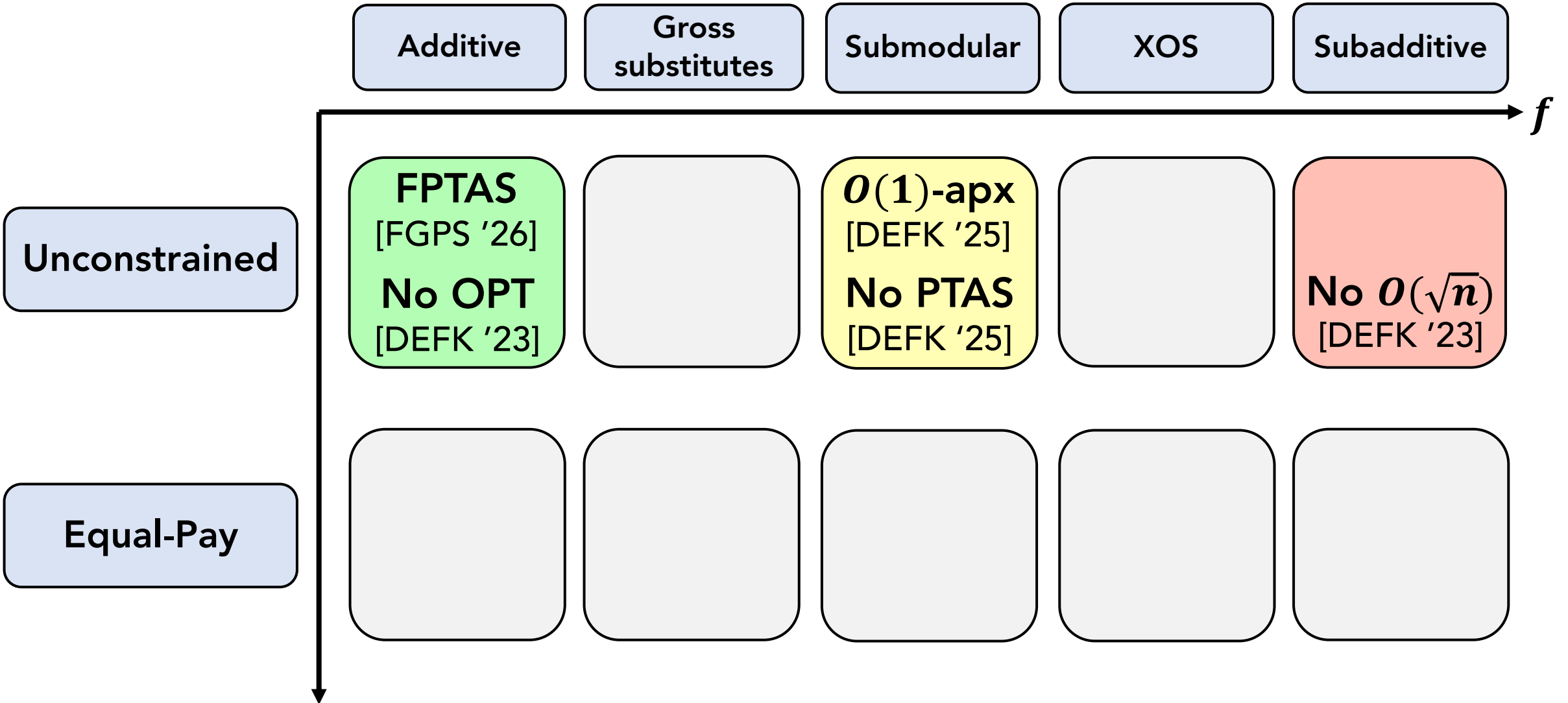


In many hardness results for **unconstrained** contracts, the optimal contract is **equal-pay**.

[DEFK '23] [EFS '24] [DEFK '25]

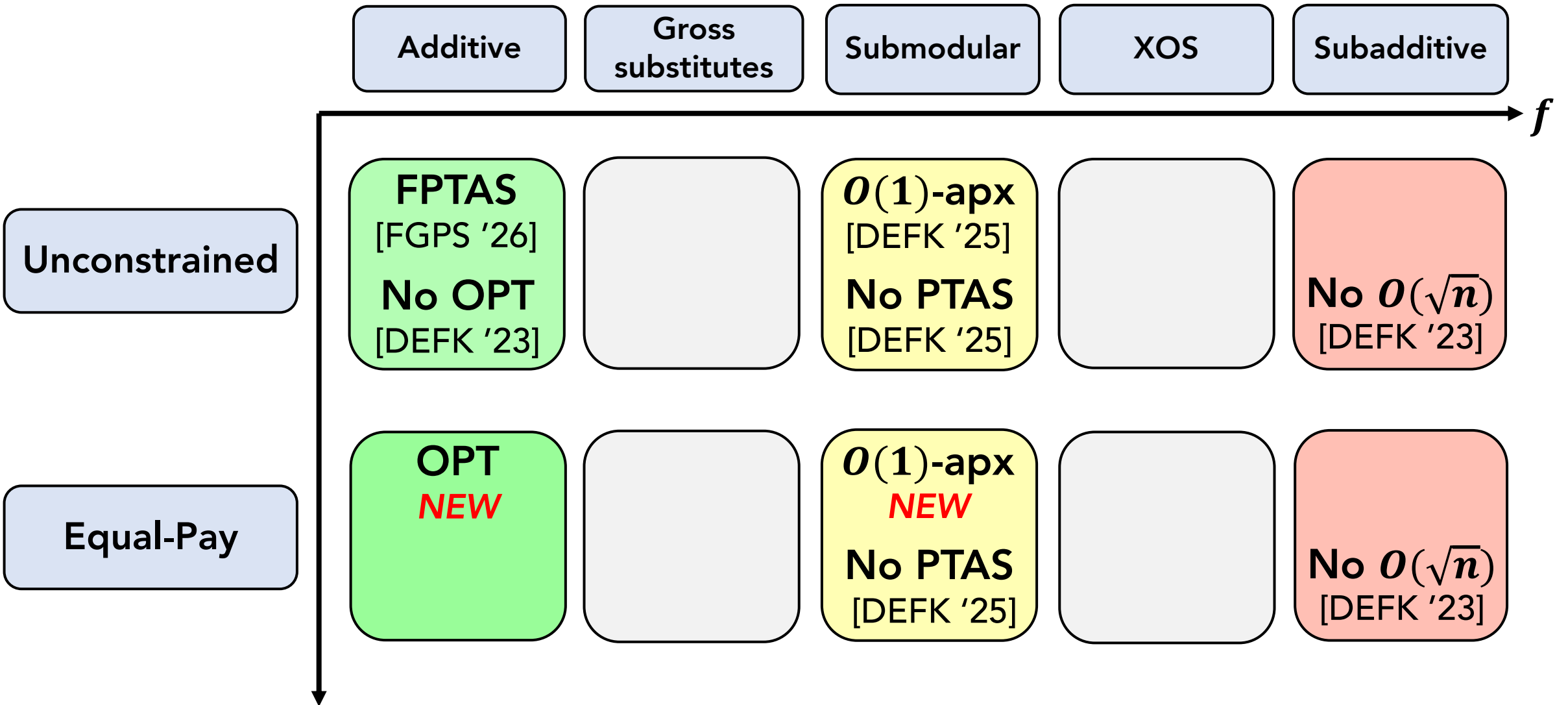
Our analysis of **equal-pay** contracts provides two new hardness results for **unconstrained** contracts.

The Computational Landscape*



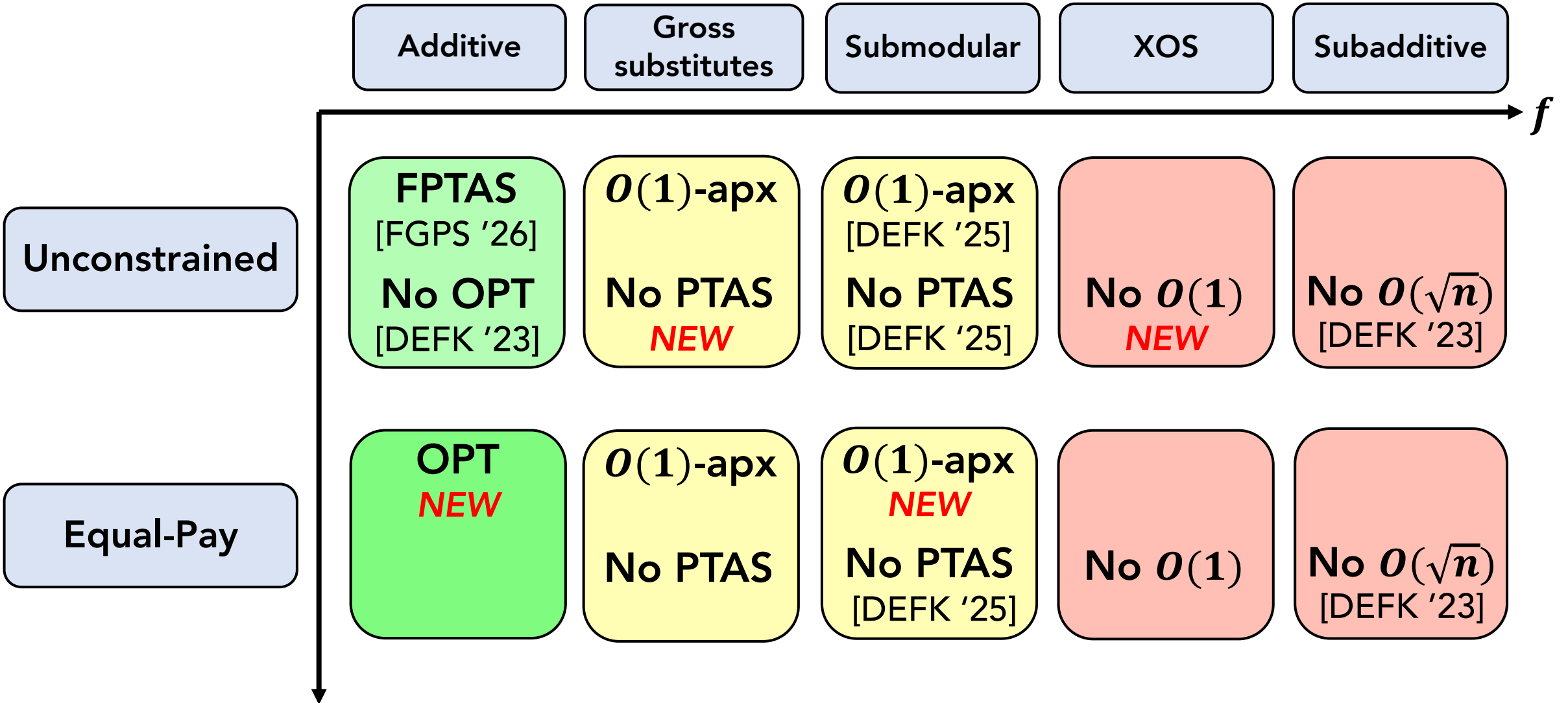
*with demand queries

The Computational Landscape*



*with demand queries

The Computational Landscape*



*with demand queries

Our Techniques

Positive Result for Submodular:
Approximate demand queries with **agent cardinality constraints**

Negative Result for XOS:
Hide a special set

Negative Result for Gross Substitutes:
Encoding a **coverage structure** over the agents

Our Techniques

Theorem: For submodular f , an **$O(1)$ -approximation** to the optimal **equal-pay** contract can be computed in **poly time** with demand queries.

Demand Queries

Input: $p \in \mathbb{R}^A$

Output: $S \in \operatorname{argmax}_{S \subseteq A} (f(S) - \sum_{a \in S} p_a)$

Demand queries are **agent-agnostic**:
They ignore **ownership of actions**.

Payments are generally lower when
fewer agents control the actions.

Lemma: Demand queries with **agent cardinality constraints** can be **approximately** simulated by polynomially many value queries.

We run the **distorted greedy** algorithm [HFWK '19] recursively on two levels: at the **agent level**, and, for each agent, at the **action level**.

Our Techniques

Information-Theoretic Hardness: For XOS f , **no** algorithm making polynomially many value or demand queries achieves an **$O(1)$ -approx.**

- (a) The **hidden set** is required to find a **good contract**.
- (b) Since demand queries are **agent-agnostic**, identifying the hidden set requires **exponentially many** such queries.

hidden set of agents T

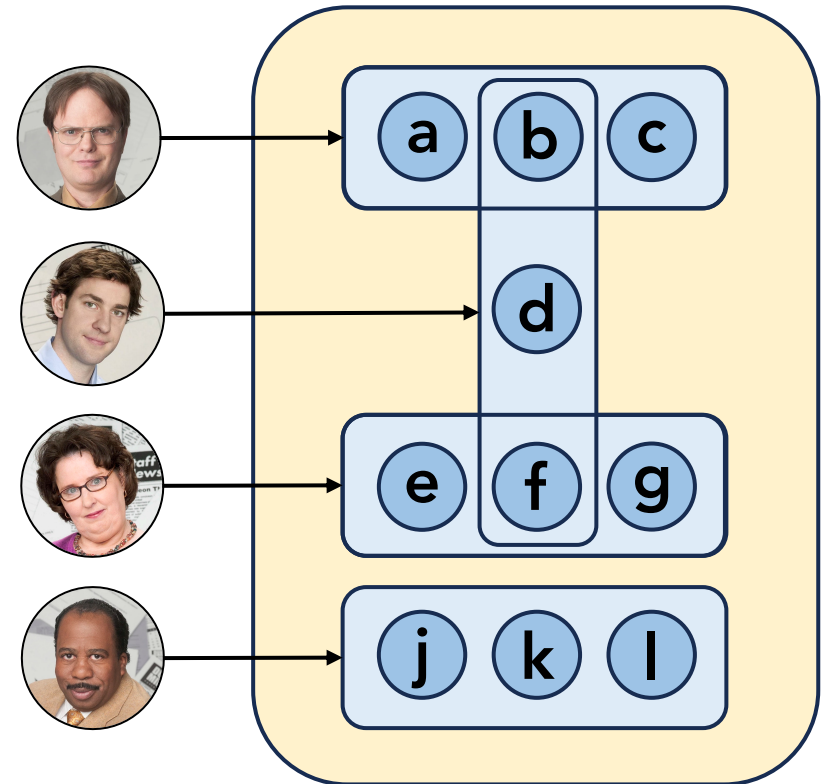


Our Techniques

Computational Hardness: For gross substitutes f , there is **no PTAS** unless $P = NP$.

Reduction from a promise variant of **maximum coverage** [EFS '24]

Matroid structure over **actions**
Coverage structure over **agents**



Research Direction

Question 1 (Computational Aspects):

Can we efficiently compute nearly-optimal **equal-pay** contracts?

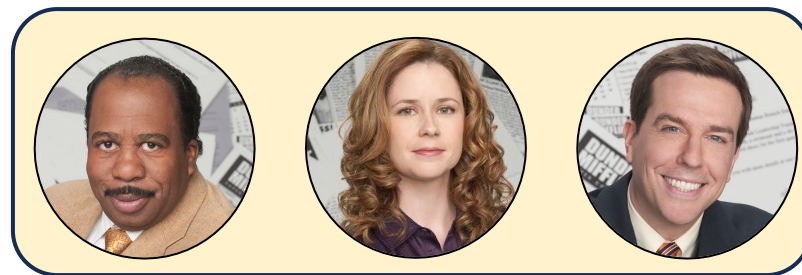
Question 2 (Price of Equality):

How much utility is lost by restricting contracts to be **equal-pay** relative to **unconstrained** contracts?

Price of Equality

Theorem: For XOS f , the optimal **equal-pay** contract gives an $\Theta(\log n / \log \log n)$ -approx. to the optimal **unconstrained** contract.

We sort the agents by decreasing α_i^* and apply a bucketing argument.



Large Agent

$$\alpha_i^* > 1/2$$

1 bucket

Middle Agents

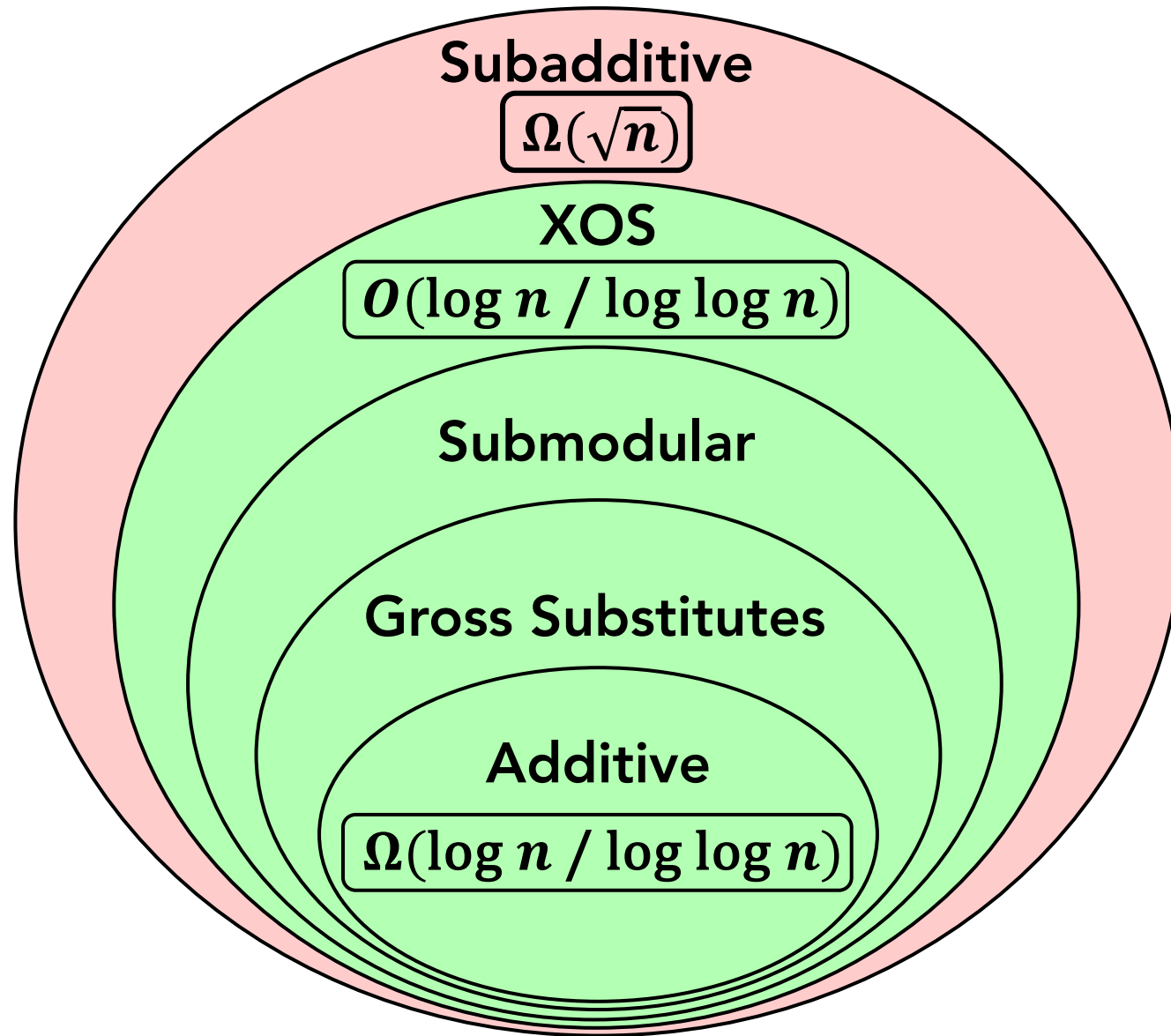
$\Theta(\log n / \log \log n)$ buckets

Small Agents

$$\alpha_i^* < 1/(2n)$$

1 bucket

Price of Equality



Other Works on Fair Contracts

Fairness is an **emerging frontier** in algorithmic contract design.

Envy-Free Contracts

[Castiglioni, Chen, Li, 2025]

Nearly-Equal-Pay Contracts

[Ding, Li, Sun, 2026]

Maximizing Social Welfare

[Aharoni, Hoefer, Talgam-Cohen, 2025]

[Feldman, Gal-Tzur, Ponitka, Schlesinger, 2025]

Price of Non-Discrimination in Public Combinatorial Contracts [Feng, Ma, Xiao, 2024]

Fair Contracts [Castiglioni, Chen, Li, 2025]

Envy-Free Contracts with Subsidies [Castiglioni, Chen, Li, 2026]

Anonymous Contracts [Brustle, Dütting, Leonardi, Ponitka, Russo, 2026]

Envy-Free Contracts

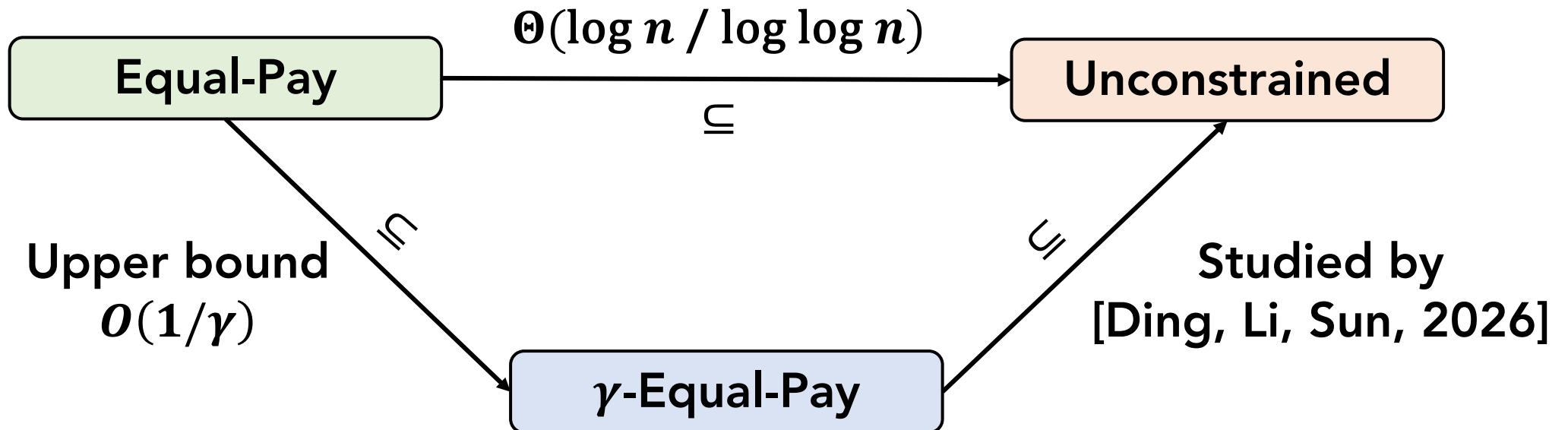
[Castiglioni, Chen, Li, 2025] define **envy-free** contracts.



Open Problem:
Bounding the gap for combinatorial actions and/or XOS f .

Nearly-Equal-Pay Contracts

A contract $\alpha = (\alpha_1, \dots, \alpha_n)$ is **γ -equal-pay** if $\alpha_i \geq \gamma \cdot \alpha_j$ for agents i, j with $\alpha_i, \alpha_j \neq 0$.



Open Problem:
Better bounds for the **equal-pay** vs **γ -equal-pay** gap.

BEST (BEyond STandard) Objectives

For **gross substitutes** f , our computational and price-of-equality results for maximizing **principal's utility** extend to **all BEST objectives**.

Principal's Utility (Second Best): $(1 - \sum_{i=1}^n \alpha_i) \cdot f(S)$

BEST Objectives [FGPS '25]

Expected Reward: $f(S)$

Social Welfare (First Best): $f(S) - c(S)$

Separation: Price of equality for **submodular** f and maximizing **reward** is ∞ .

Open Problem: Price of equality for **submodular** f and maximizing **welfare**.

Takeaways

Equal-Pay Contracts:

(a) motivated by **fairness**.

(b) useful lens for understanding **unconstrained** contracts.

The **price of equality** is $\Theta(\log n / \log \log n)$.

Fairness is an **emerging frontier** in algorithmic contract design.

Thank you!