

Oblivious Rounding and the Integrality Gap

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Setting: A Maximization Problem (V, X)

IN GENERAL

X = set of feasible solutions

V = set of linear objective functions

X, V are sets of non-negative real vectors

EXAMPLE: MAX-CUT IN COMPLETE
WEIGHTED GRAPHS OF n VERTICES

X = set of cuts

V = set of edge weight functions

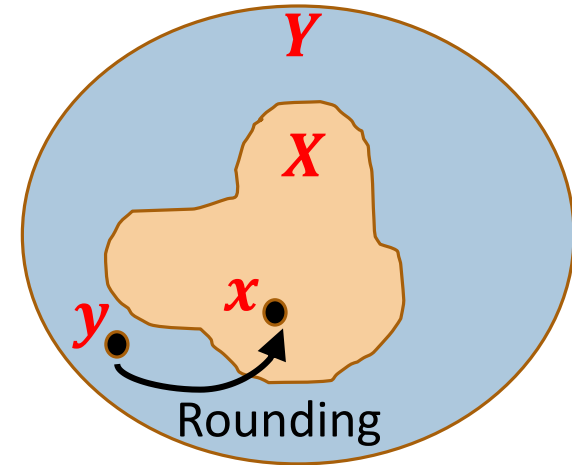
X, V are sets of vectors of dimension $\binom{n}{2}$
(vectors in X are $\{0,1\}$ -vectors)

Given $v \in V$, find $x \in X$ that maximizes $v \cdot x$

Oblivious Rounding

Classic approach to hard maximization problem (V, X) :

1. Relax (V, X) to (V, Y) where $X \subset Y$ and Y is fractional
2. Given $v \in V$ find $y \in Y$, guarantees $v \cdot y$
3. Round y to $x \in X$



The **approximation ratio** of the rounding is $\frac{v \cdot x}{v \cdot y}$ (in the worst case)

If step 3 does not use v , we call the rounding “**oblivious**”*

*Not to be confused with [\[Young’95\]](#)

Examples from the Literature

OBLIVIOUS ROUNDING

Threshold rounding for vertex cover

- [Hochbaum'82]

Randomized rounding for set cover

- [Raghavan-Thompson'87]

Random hyperplane rounding for max-cut

- [Goemans-Williamson'95]

Welfare maximization for **submodular** valuations

- [Feige'09, Feige-Vondrak'10]

NON-OBLIVIOUS ROUNDING

Rounding of SDPs for CSP

- [Raghavendra-Steurer'09]

Facility location

- [Li'13]

Welfare maximization for **gross substitutes** valuations

- [Nisan-Segal'06]

Main Question

For which problems and relaxations can we expect oblivious rounding to give a good approximation ratio?

A question about **information**

- Rounding **not** restricted to be in polynomial time

Answer useful for:

- Algorithm designers
- Mechanism designers (details in a few slides)

Main Result & Application

Main Result

[Informal] The approximation ratio of the best oblivious rounding scheme for a given relaxation = the integrality gap of the problem's closure

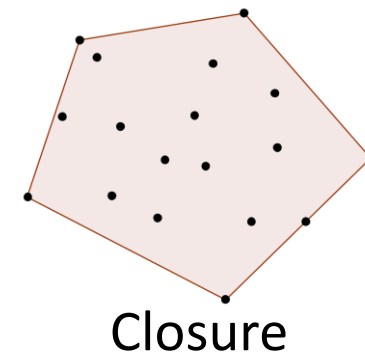
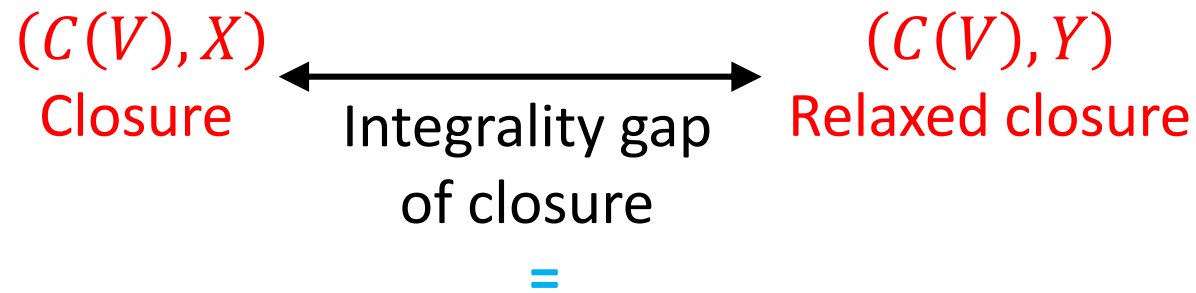
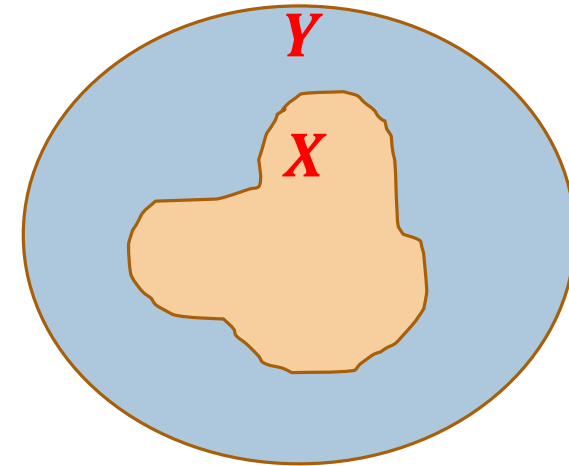
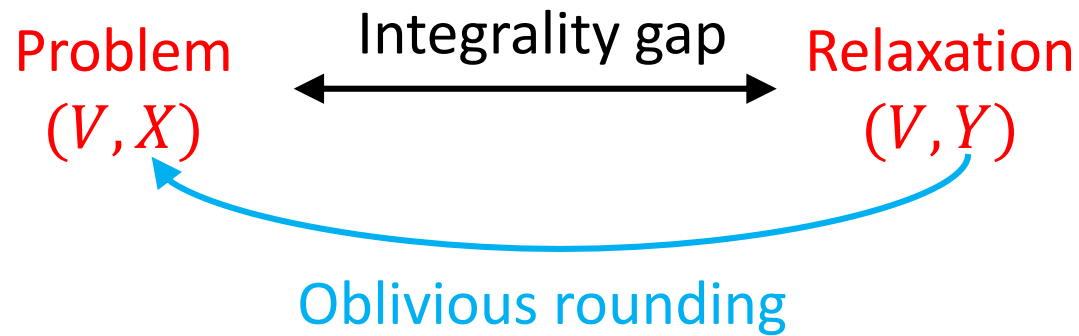
The closure of problem (V, X) is $(C(V), X)$

- where $C(V)$ is the convex closure of V



Corollary: If a problem is closed ($V = C(V)$) then oblivious rounding can achieve the integrality gap

Main Result Illustration



Oblivious rounding approximation ratio

An Application: Welfare Maximization

Informally: Allocate m indivisible items among buyers to maximize total value

- Each buyer i has a valuation function $v_i: 2^{[m]} \rightarrow \mathbb{R}_{\geq 0}$
- Valuations belong to classes (e.g., additive, submodular, ...)

More formally:

- $v \in V$ = the buyers' valuations, from class V
- $x \in X$ = an allocation
- $y \in Y$ = an allocation as if the items were divisible

(all sets of vectors of dimension 2^m)

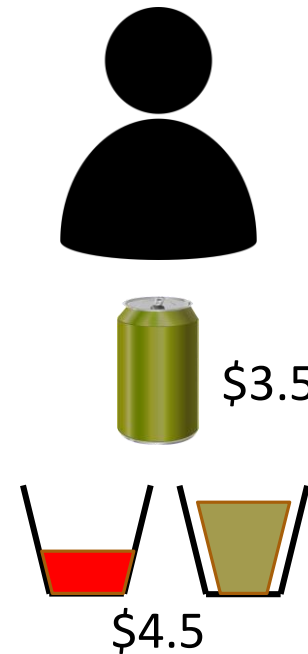
Main thm: Approximation ratio of oblivious rounding = integrality gap of problem's closure

Welfare Max: The Relaxation

Problem: Indivisible items



Relaxation: Divisible items



Welfare = 5.5

Welfare = \$7

Integrality gap

Welfare Max: The Relaxation $X \rightarrow Y$

The relaxation used in \approx all welfare approximation algorithms: [Configuration LP](#)

Problem:

$$\max_{i,S} \sum x_{i,S} v_{i,S}$$

s.t.

$$\sum_S x_{i,S} \leq 1 \text{ for every buyer } i$$

$$\sum_{i,S:j \in S} x_{i,S} \leq 1 \text{ for every item } j$$

$$x_{i,S} \in \{0,1\}$$

Relaxation:

$$\max_{i,S} \sum y_{i,S} v_{i,S}$$

s.t.

$$\sum_S y_{i,S} \leq 1 \text{ for every buyer } i$$

$$\sum_{i,S:j \in S} y_{i,S} \leq 1 \text{ for every item } j$$

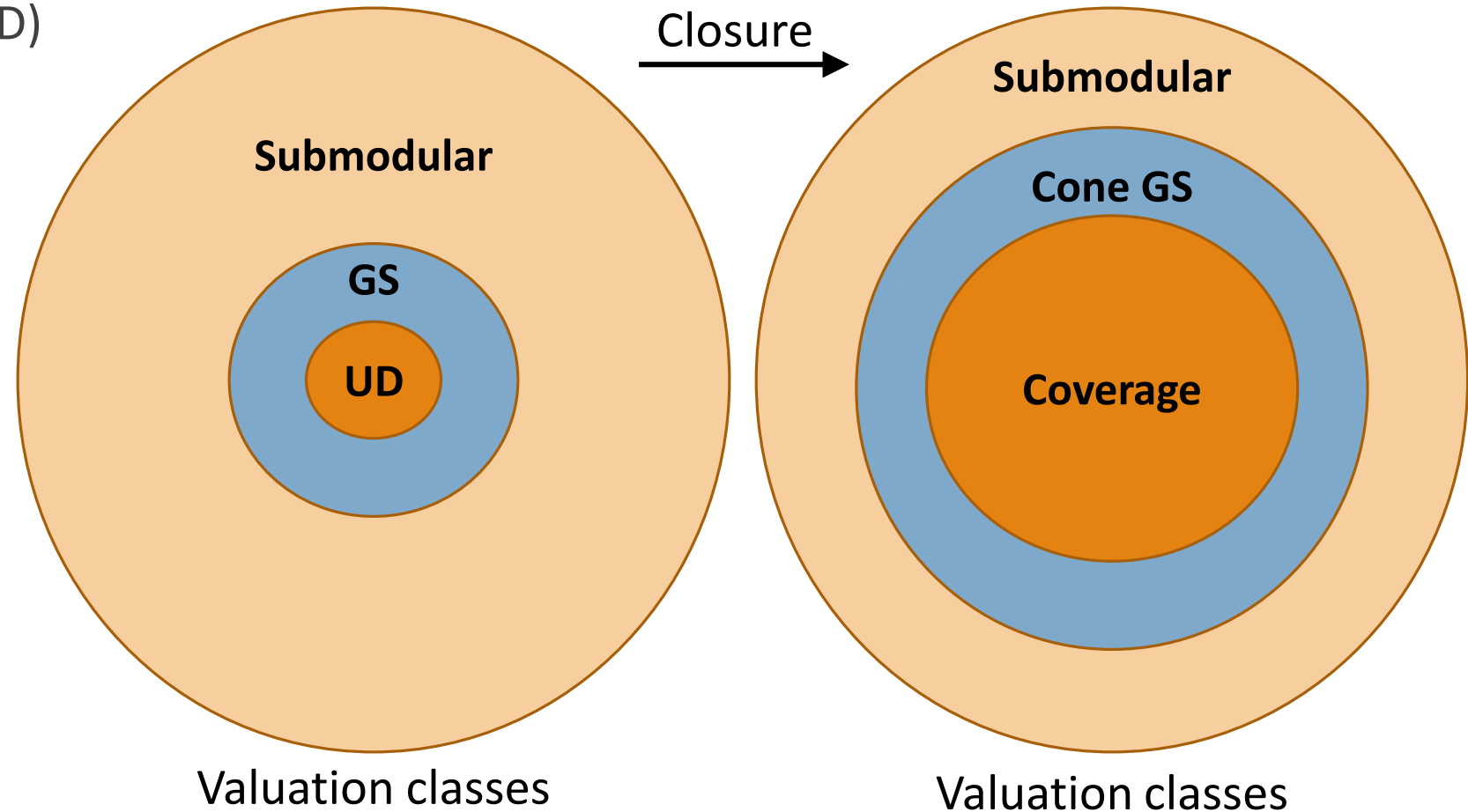
$$y_{i,S} \geq 0$$

Main thm: Approximation ratio of oblivious rounding = integrality gap of problem's closure

Welfare Max: The Closure $V \rightarrow C(V)$

Let V be unit-demand (UD) valuations:

- $v_i(S) = \max_{\text{item } j \in S} \{v(j)\}$
- Subclass of gross substitutes (GS)
- So integrality gap of configuration LP = 1



Main Result Applied to Welfare Max.

Theorem: For welfare maximization with **unit-demand** valuations, **oblivious** rounding of solutions to the configuration LP achieves ≤ 0.782 approximation (**0.833** for 2 buyers)

- Despite the integrality gap of **1** for **unit-demand/GS**

Conclusion: For welfare maximization, “ignorance is not always bliss”

- Need to know the valuations to round the fractional allocation

Proof: The integrality gap of the configuration LP for **coverage** valuations is no better than **0.782** [*cf.* Feige-Vondrak'10], and coverage is the closure of unit-demand

Implications for Mechanism Design

Advantages of oblivious rounding for welfare maximization with **strategic** buyers:

1. Incentive compatibility
 - [Duetting-Kesselheim-Tardos'15]: Can embed into a mechanism that approximately maximizes welfare **in equilibrium**
2. Fairness
 - Treats buyers equally, approximation ratio holds **per buyer**
3. Communication
 - Does not access exponential-sized valuations

Motivates understanding the possibilities/limitations of oblivious rounding

Sketch of Main Proof

Approx. Ratio and Integrality Gap at y

Fix a fractional solution $y \in Y$

INTEGRALITY GAP OF CLOSURE AT y

$$\inf_{v \in C(V)} \max_{x \in X} \frac{v \cdot x}{v \cdot y}$$

First choose worst-case v from the closure

Then find the best integral solution x

APPROXIMATION RATIO OF BEST OBLIVIOUS ROUNDING AT y

$$\max_{x \in C(X)} \inf_{v \in V} \frac{v \cdot x}{v \cdot y}$$

First choose best randomized rounding

Then find the worst-case v for this rounding

This is where the obliviousness comes in

Applying the Minimax Theorem

Fix a fractional solution $y \in Y$

MINIMIZING "OBJECTIVE" PLAYER

$$\inf_{v \in C(V)} \max_{x \in X} \frac{v \cdot x}{v \cdot y}$$

Choose minimizing mixed strategy

MAXIMIZING "ROUNDING" PLAYER

$$\max_{x \in C(X)} \inf_{v \in V} \frac{v \cdot x}{v \cdot y}$$

Choose maximizing mixed strategy

Summary

Many commonly-used rounding schemes are **oblivious**

- Do not use the objective to round a fractional solution

We study when oblivious rounding suffices for **good approximation**

- Focus on **information**

Approximation ratio equals the **integrality gap** of a related problem – the **closure**

Application to **welfare maximization**

- Oblivious rounding does not suffice for **unit-demand**, **gross substitutes**
- Suffices for **submodular**

Another tool for the **toolbox** of algorithm and mechanism designers

Directions for Future Research

1. Use the understanding of the potential and limitations of oblivious rounding as a **guide in designing rounding** schemes
 - for problems for which tight approximation ratios not yet known
 - e.g., when best known approximation is oblivious but the problem is **not** closed
2. Possibilities/limitations of **polynomial-time** oblivious rounding
3. **Other properties** of combinatorial problems predicting the success/failure of rounding techniques