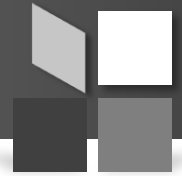


**COMP/MATH 553 Algorithmic
Game Theory**
**Lecture 12: Implementation of the
Reduced Forms and the Structure
of the Optimal Multi-item Auction**

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New Decision Variables



Variables: Interim Allocation rule aka. **“REDUCED FORM”**:

$$\{\pi_i : T_i \longrightarrow [0, 1]^n, p_i : T_i \longrightarrow \mathbb{R}^+\}_{i \in [m]}$$

$$* \pi_{ij}(v_i) : \Pr \left(\begin{array}{c} \mathbf{j} \text{ } \langle \text{bananas} \rangle \text{ } \dashrightarrow \text{ } \langle \text{monkey } i \rangle \text{ } \mathbf{i} \\ \vec{t}_{-i} \sim \mathcal{D}_{-i} \end{array} \mid \langle \text{monkey } i \rangle \text{ } \mathbf{i} \text{ valuation } v_i \right)$$

$$* \hat{p}_i(v_i) : \mathbf{E} \left[\mathbf{price}_i \mid \langle \text{monkey } i \rangle \text{ } \mathbf{i} \text{ valuation } v_i \right]$$



A succinct LP

■ Variables:

- $\pi_{ij}(v_i)$: probability that item j is allocated to bidder i if her reported valuation (*bid*) is v_i **in expectation over every other bidders' valuations (bids)**;
- $p_i(v_i)$: price bidder i pays if her reported valuation (*bid*) is v_i **in expectation over every other bidder's valuations (bids)**

■ Constraints:

- BIC: $\sum_j v_{ij} \cdot \pi_{ij}(v_i) - p_i(v_i) \geq \sum_j v_{ij} \cdot \pi_{ij}(v'_i) - p_i(v'_i)$ for all v_i and v'_i in T_i
- IR: $\sum_j v_{ij} \cdot \pi_{ij}(v_i) - p_i(v_i) \geq 0$ for all v_i in T_i
- Feasibility: exists an auction with this reduced form.

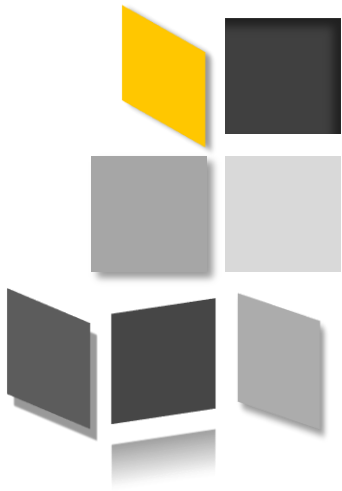
■ Objective:

- Expected revenue: $\sum_i \sum_{v_i \in T_i} \Pr[t_i = v_i] \cdot p_i(v_i)$

Implementation of a Feasible Reduced Form



- ❑ After solving the succinct LP, we find the optimal reduced form π^* and p^* .
- ❑ Can you turn π^* and p^* into an auction whose reduced form is exactly π^* and p^* ?**
- ❑ This is crucial, otherwise being able to solve the LP is meaningless.
- ❑ Will show you a way to implement any feasible reduced form, and it reveals important structure of the revenue-optimal auction!



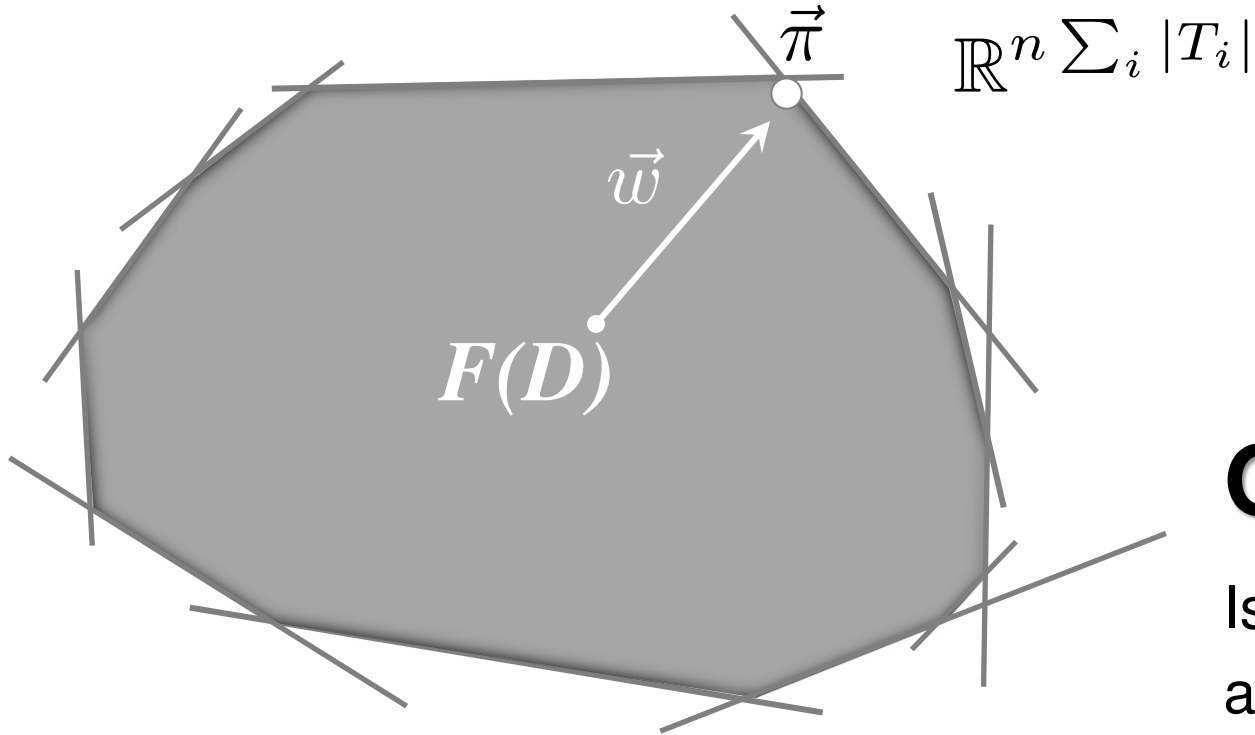
Implementation of a Feasible Reduced Form

Set of *Feasible* Reduced Forms



- Reduced form is collection $\{\pi_i : T_i \longrightarrow [0, 1]^n\}$;
- Can view it as a vector $\vec{\pi} \in \mathbb{R}^{n \sum_i |T_i|}$;
- Let's call set of feasible reduced forms $F(D) \in \mathbb{R}^{n \sum_i |T_i|}$;
- **Claim 1: $F(D)$ is a convex polytope.**
- **Proof: *Easy!***
 - A feasible reduced form $\vec{\pi}$ is implemented by a feasible allocation rule M .
 - M is a distribution over deterministic feasible allocation rules, of which there is a finite number. So: $M = \sum_{\ell=1}^k p_\ell \cdot M_\ell$, where M_ℓ is **deterministic**.
 - Easy to see: $\vec{\pi} = \sum_{\ell=1}^k p_\ell \cdot \vec{\pi}(M_\ell)$
- So, $F(D) = \left(\begin{array}{c} \text{convex hull of reduced forms of} \\ \text{feasible deterministic mechanisms} \end{array} \right)$

Set of *Feasible* Reduced Forms



Q:

Is there a **simple** allocation rule implementing the corners?

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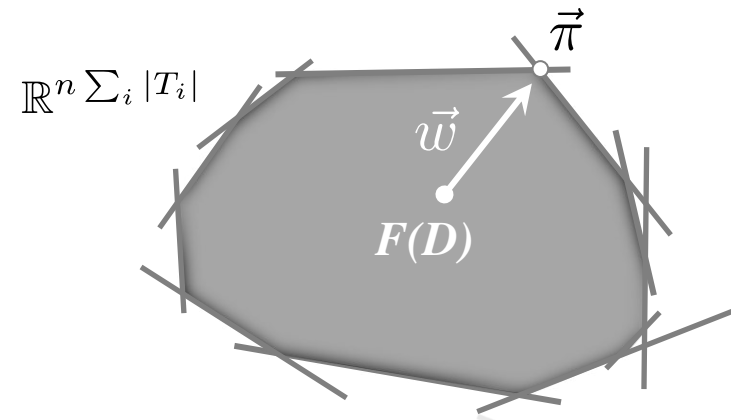


virtual welfare maximizing interim rule when virtual value functions are the f_i 's



expected **virtual** welfare of an allocation rule with interim rule π'

interpretation: **virtual** value derived by bidder i when given item j when his type is A



$$\vec{\pi} \in \arg\max_{\vec{\pi}' \in F(D)} \{ \vec{\pi}' \cdot \vec{w} \}$$

$$\vec{\pi}' \cdot \vec{w} = \sum_i \sum_j \sum_{A \in T_i} \pi'_{ij}(A) w_{ij}(A) \quad \text{----- (1)}$$

$$= \sum_i \sum_j \sum_{A \in T_i} \pi'_{ij}(A) f_{ij}(A) \Pr[t_i = A] \quad \text{--- (2)}$$

$$f_{ij}(A) := \frac{w_{ij}(A)}{\Pr_{\mathcal{D}}[t_i = A]}$$

Is there a **simple** allocation rule implementing a corner?



virtual welfare maximizing interim rule when virtual value functions are the f_i 's



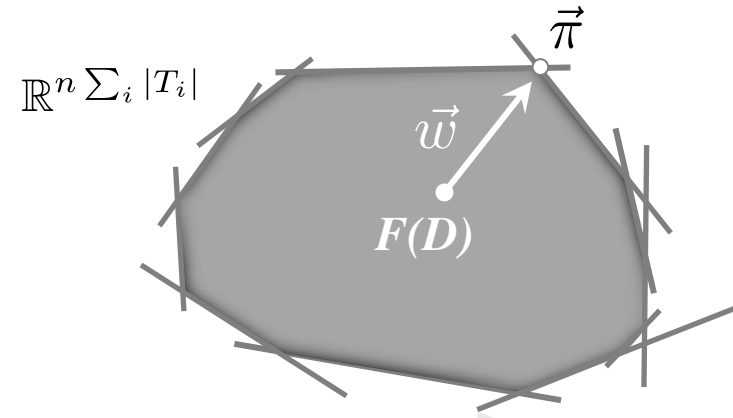
Q: Can you name an algorithm doing this?

A: YES, the VCG allocation rule (w/ virtual value functions $f_i, i=1, \dots, m$)

= : **virtual-VCG**($\{f_i\}$)

interpretation: **virtual** value derived by bidder i when given item j when his type is A

$$f_{ij}(A) := \frac{w_{ij}(A)}{\Pr_{\mathcal{D}}[t_i = A]}$$

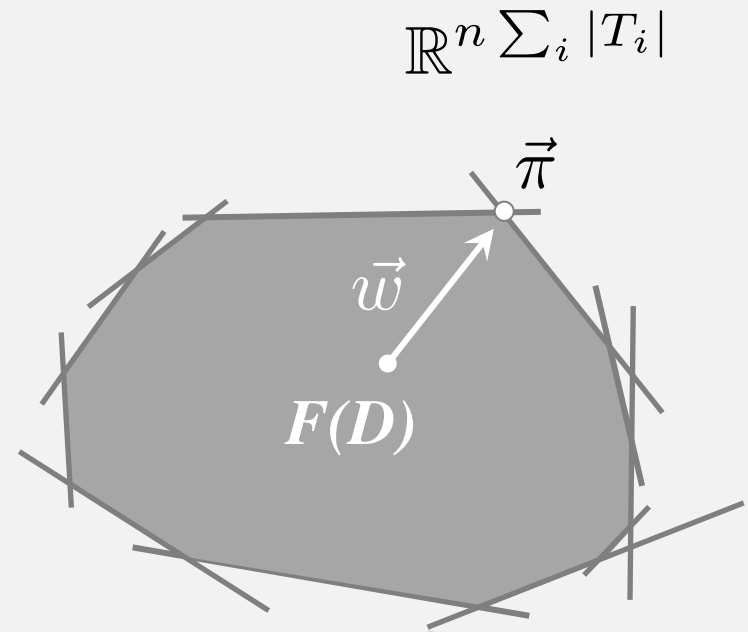


$$\bar{\pi} \in \text{argmax}_{\pi' \in F(D)} \{ \pi' \cdot \bar{w} \}$$



➔ $F(D)$ is a **Convex Polytope** whose corners are **implementable** by **virtual VCG** allocation rules.

How about implementing any point **inside** $F(D)$?



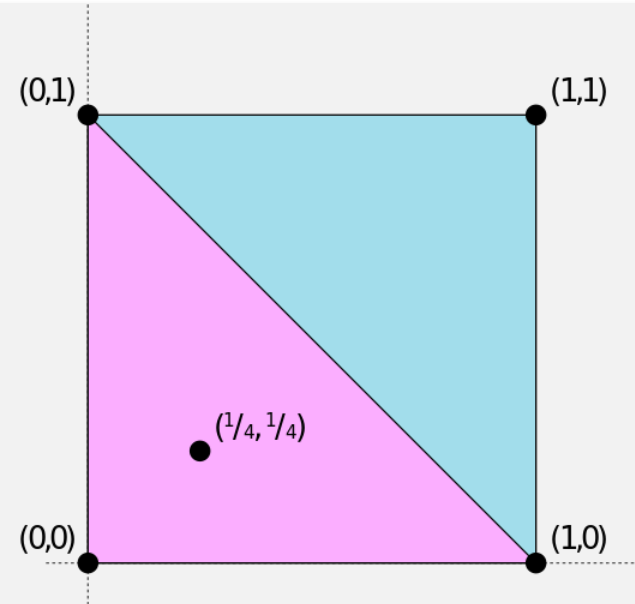
Carathéodory's theorem

If some point x is in the convex hull of P then

$$x = \sum_{p_i \in P} q_i \cdot p_i$$

$$\text{s.t. } \sum_i q_i = 1 \text{ and } q_i \geq 0 \forall i$$

Carathéodory's Theorem: If a point x of \mathbb{R}^d lies in the convex hull of a set P , there is a subset P' of P consisting of $d + 1$ or fewer points such that x lies in the convex hull of P' .



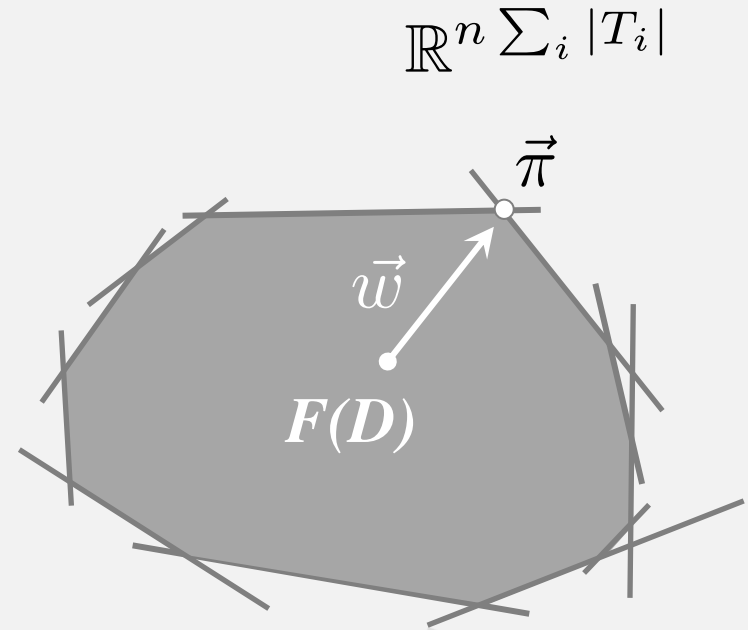
For example:

$$x = \frac{1}{4}(0,1) + \frac{1}{4}(1,0) + \frac{1}{2}(0,0)$$

Characterization Theorem [C.-Daskalakis-Weinberg]



- ➔ Any point inside $F(D)$ is a convex combination (distribution) over the corners.
- ➔ The interim allocation rule of any feasible mechanism can be implemented as a **distribution over virtual VCG allocation rules.**





Structure of the Optimal Auction



Theorem [C.-Daskalaks-Weinberg]: Optimal multi-item auction has the following structure:

1. Bidders submit valuations (t_1, \dots, t_m) to auctioneer.
2. Auctioneer samples virtual transformations f_1, \dots, f_m
3. Auctioneer computes virtual types $t'_i = f_i(t_i)$
4. Virtual welfare maximizing allocation is chosen.

Namely, each item is given to bidder with highest virtual value for that item (if positive)

5. Prices are charged to ensure truthfulness.



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❖ Exact same structure as Myerson!

- in Myerson's theorem: virtual function = deterministic
- here, *randomized* (and they must be)

Interesting Open Problems



- ❑ Another difference: in Myerson's theorem: virtual function is given explicitly, in our result, the transformation is computed by an LP. Is there any structure of our transformation?

- ❑ In single-dimensional settings, the optimal auction is DSIC. In multi-dimensional settings, this is unlikely to be true. What is the gap between the optimal BIC solution and the optimal DSIC solution?