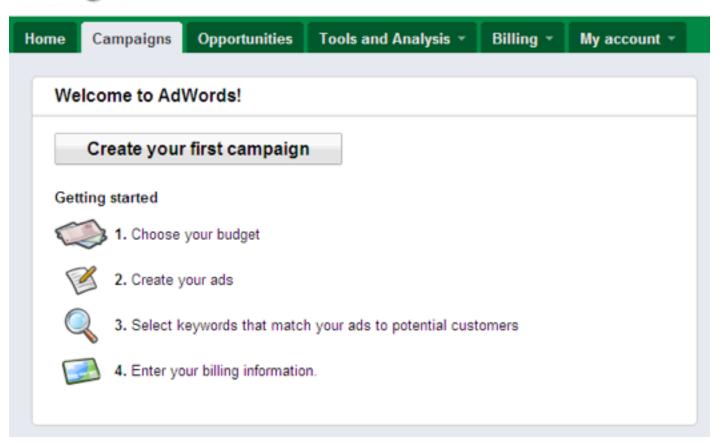
# POLYHEDRAL CLINCHING AUCTIONS BEYOND HARD BUDGET CONSTRAINTS

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Google Research NYC

- **Item values** are an useful abstraction but often intangible.
- Typically, buyers care about the items (impressions) only in **aggregate**.
- Aggregate statistics about an auction result: **budget spent**, **average cpc**, ...

# Google<sup>-</sup>AdWords



### Bidding and budget

Bidding option ? Basic options | Advanced options OIII manually set my bids for clicks AdWords will set my bids to help maximize clicks within my target budget Budget ? S per day Actual daily spend may vary. ?

### Bidding and budget

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Budget 🔃	\$ pe Actual daily spen	r day d may vary. ?			

### Bidding and budget

Bidding option (2)	O Focus on clicks - use maximum CPC bids				
	O Manual bidding for clicks				
	Automatic bidding to try to maximize clicks for your target budget     Focus on conversions (Conversion Optimizer) - use CPA bids     Advanced options				
	Bid type				
	O Max CPA: Most you would want to pay for any particular conversion				
	Target CPA: Average amount you would like to pay for each conversion				
	Set an initial target CPA (1-per-click) bid				
	Start with recommended bid: \$13.20      This is accounted by the second se				
	This is approximately equivalent to a maximum CPC bid of \$1.86				
	O Start with custom bid \$ Target CPA				
	After you click 'Save' you will be able to set individual ad group bids.				
	Your current maximum CPC bids will be saved so you can return to manual bidding later.				
	Save Cancel				

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- Few techniques for budgeted settings.
- [Ausubel], [Dobzinski, Lavi, Nisan]: clinching auctions
- Extended in many directions in previous years:
  - general environments: [Fiat et al], [Colini-Baldeschi et al], [Goel, Mirrokni, PL], [Dobzinski, PL]
  - revenue: [Bhattacharya et al], [Devanur, Ha, Hartline]
  - online settings: [Goel, Mirrokni, PL]

- Two issues with current state of affairs:
  - **Clinching** is all we know how to do
  - Our knowledge is (mostly) limited to hard budget constraints.

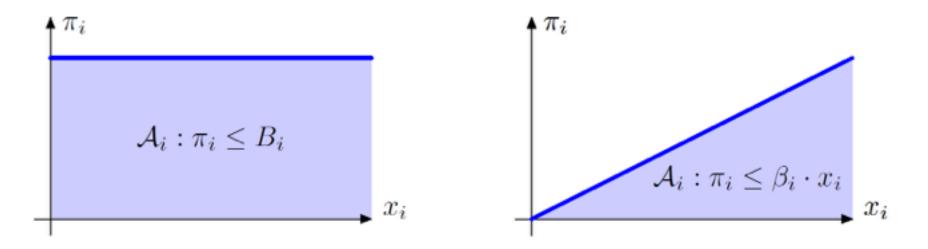
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**Plan:** Address the second issue.

Hard Budgets: 
$$u_i(x_i, \pi_i) = \begin{cases} v_i(x_i) - \pi_i, & \pi_i \leq B_i \\ -\infty, & \end{cases}$$

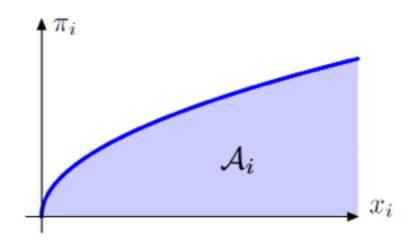
Average budgets: 
$$u_i(x_i, \pi_i) = \begin{cases} v_i(x_i) - \pi_i, & \pi_i \leq \beta_i x_i \\ -\infty, & \text{o.w.} \end{cases}$$

Generic constr: 
$$u_i(x_i, \pi_i) = \begin{cases} v_i(x_i) - \pi_i, & (x_i, \pi_i) \in \mathcal{A}_i \\ -\infty, & \text{o.w.} \end{cases}$$



### Generic admissible set:

•  $(0,0) \in \mathcal{A}_i$ 



• right-down closeness

 $(x_i, \pi) \in \mathcal{A}_i \Rightarrow (x_i + \delta, \pi_i - \delta') \in \mathcal{A}_i, \forall \delta, \delta' \ge 0$ 

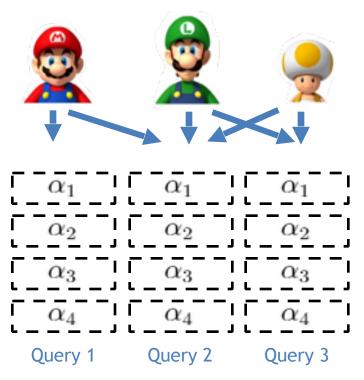
- convexity: distributions over admissible outcomes are admissible
- topological closeness

# Setting

- n agents with (private) value  $v_i$  per item (say clicks) and (public) admissible set  $\mathcal{A}_i$
- allocation constraints  $\mathbf{P} \subseteq \mathbb{R}^n$  (polymatroid) i.e. sponsored search, one-sided-matching, flows, spanning trees, ...

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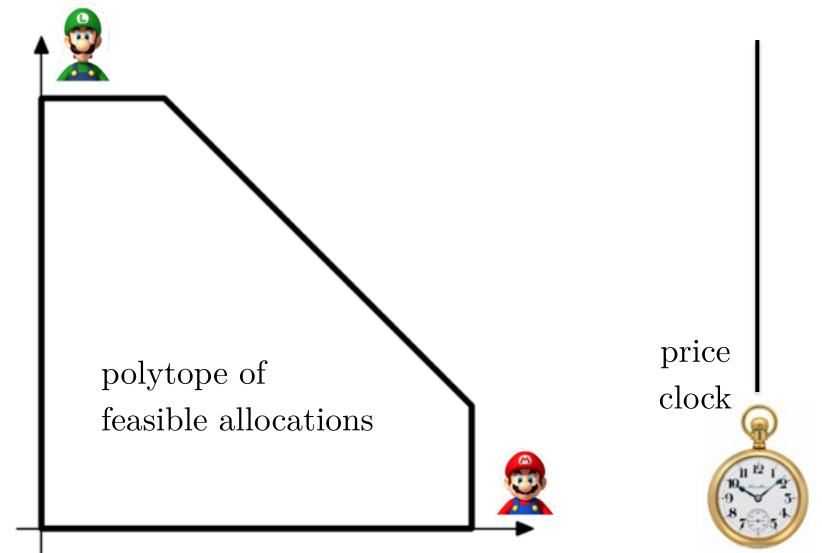


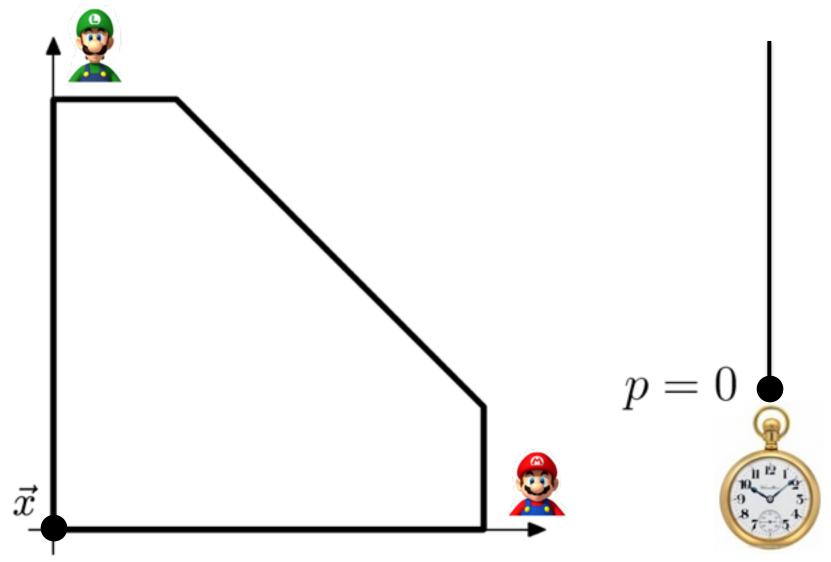
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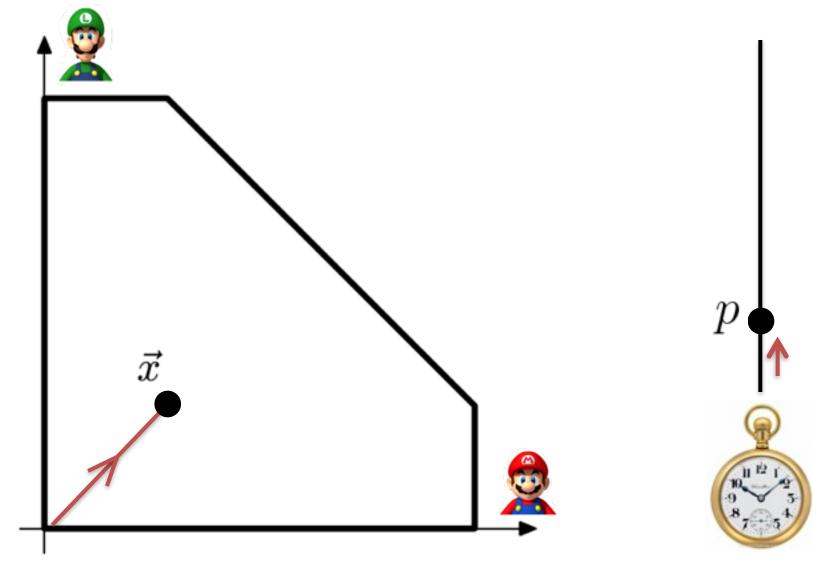
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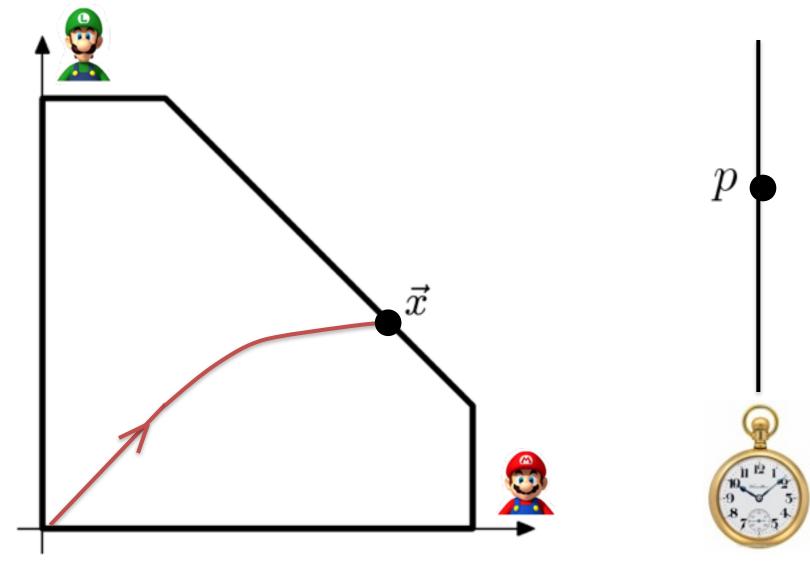
# Goal

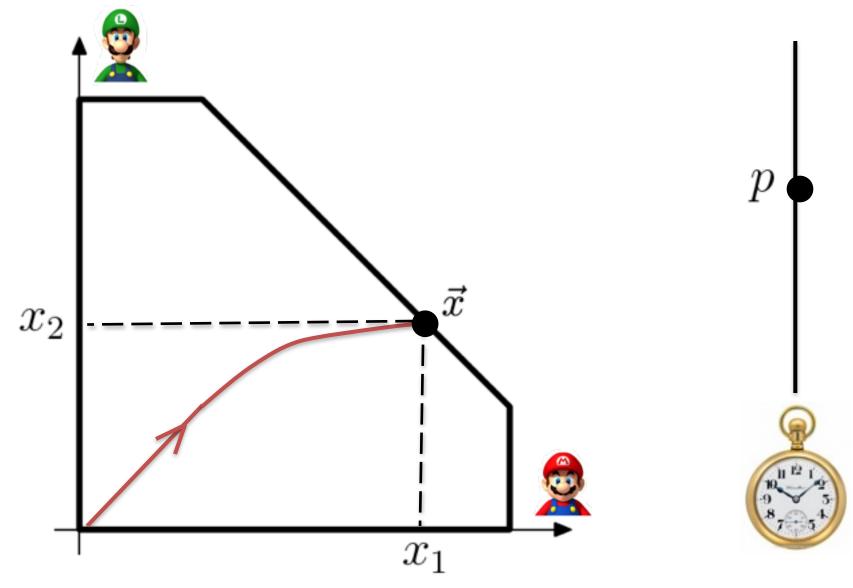
- truthful auction
- admissible outcomes
- Pareto efficient: no alternative outcome where each agent and the auctioneer weakly improve and at least one strictly improves.





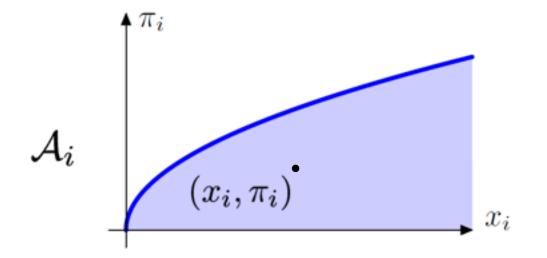






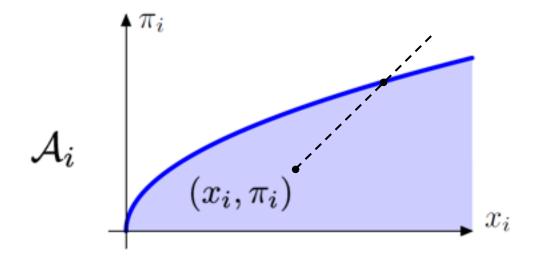
We initialize  $(x_i = 0, \pi_i = 0)$  and set prices and update for all prices p

For each price p we compute the demands of each agent



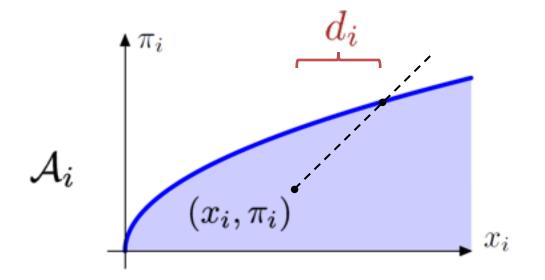
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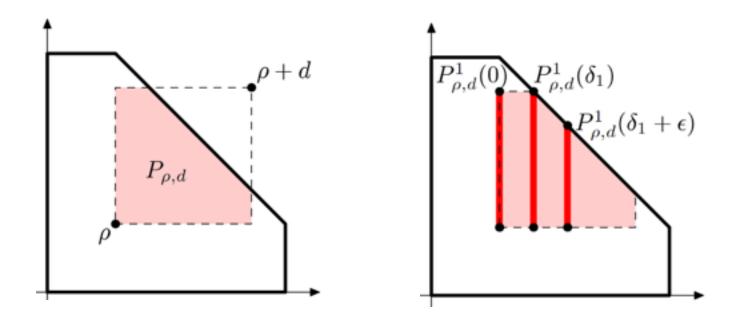
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Clinching: find for each agent maximum amount  $\delta_i$  that one can allocate to him without making the allocations of the other players infeasible.

$$x_i = x_i + \delta_i \qquad \pi_i = \pi_i + p\delta_i$$

various new techniques needed to prove Pareto-optimality for generic  $\mathcal{A}_i$  :

Pareto optimality  $\equiv$  no trade

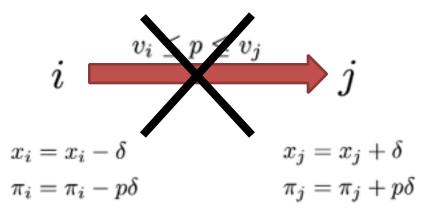
 $v_i < v_j$ 

$$i$$
  $v_i \le p \le v_j$   $j$   
 $x_i = x_i - \delta$   $x_j = x_j + \delta$   
 $\pi_i = \pi_i - p\delta$   $\pi_i = \pi_i + p\delta$ 

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- 1) New outcome not admissible for j
- 2) Violates feasibility constraints

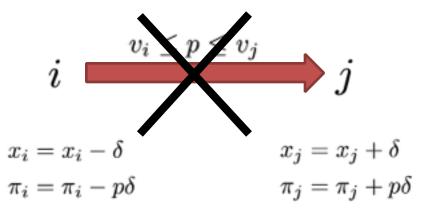


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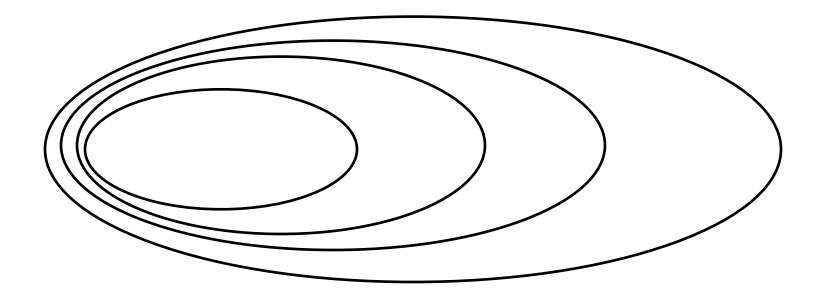
Pareto optimality  $\equiv$  no trade

Hard budgets: no trade at one price p means not trade at any price

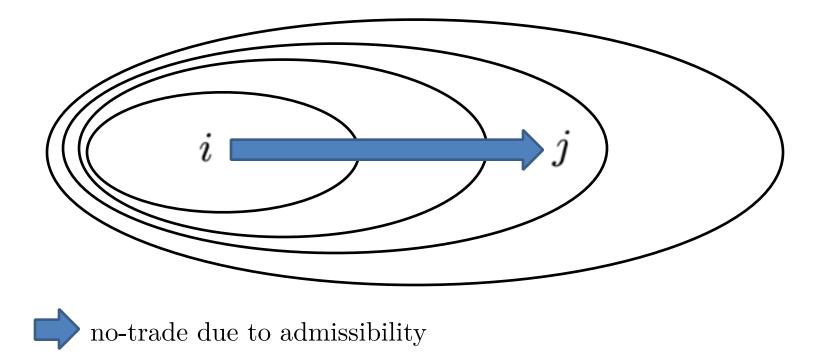
not true anymore...



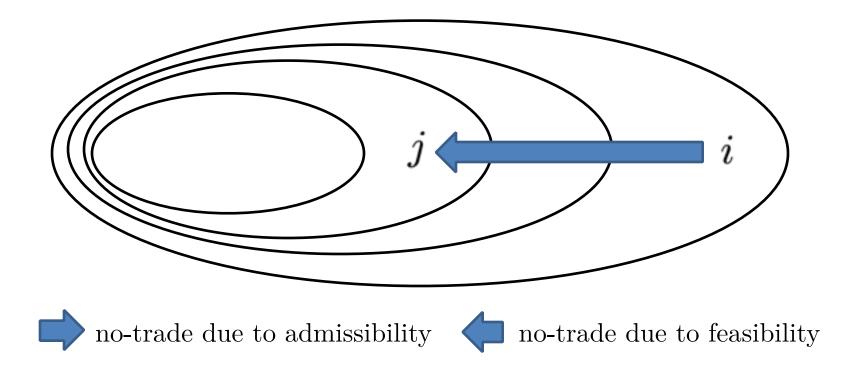
Structure of tight sets lemma: sets of agents



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**Future directions** 

How much further can clinching take us in nonquasilinear settings ?

Average budgets in online settings.

Heuristics in practice inspired by this auction.

Can we go **beyond clinching**?

# Thanks !