

Profit-Maximizing Envy-free Pricing

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Envy-free pricing

Seller: (m items)

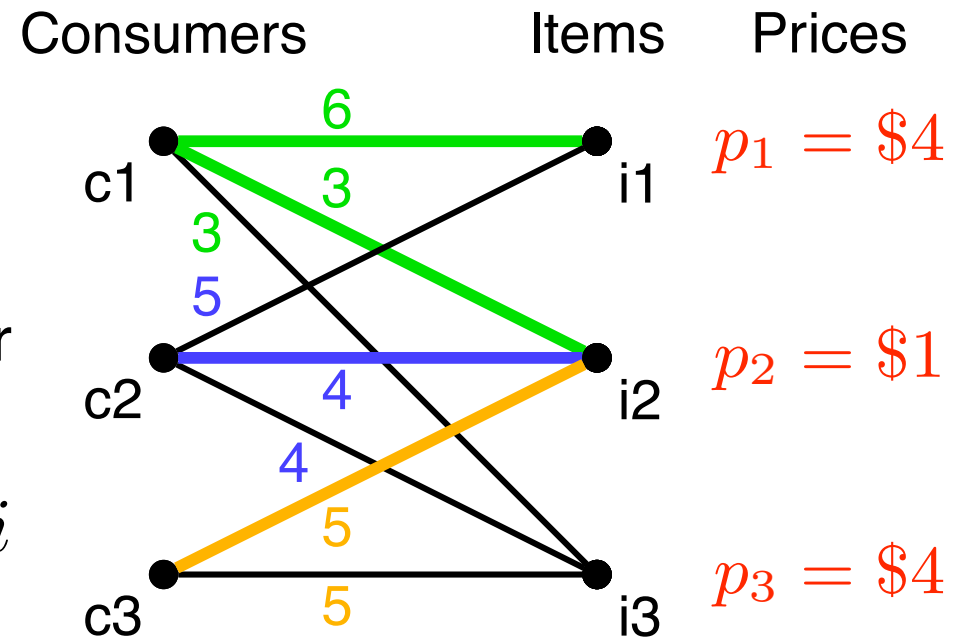
- Sets price p_j for item j .

Consumers: (n consumers)

- Consumer i has *valuation* v_{ij} for item j .
- For item j at price p_j , consumer i has *utility*: $u_{ij} = v_{ij} - p_j$.
- Consumer desired item to maximize utility.

Agreement: (*envy-free pricing*)

- Must have allocation such that all consumers are happy.



For Consumer 3:

$$u_{32} = \$4$$

$$u_{33} = \$1$$

Envy-free pricing

Given: the valuations v_{ij}

Find: prices p_j and allocation

so as: to maximize seller profit $\sum_j \text{allocated } p_j$

s.t.: envy-free constraint:

if customer i is allocated item j ,

then $p_j \leq v_{ij}$

and for every j' , we have $v_{ij'} - p_{j'} \leq v_{ij} - p_j$.

Motivation

- Envy-free pricing has been studied in Economics for 50 years [Walras 1954]
- Envy-free \Rightarrow consumer has no incentive to change allocation; Maximum-profit \Rightarrow seller has no incentive to change prices.
- “Price of truthfulness” in combinatorial auctions: to analyze performance, compare truthful mechanism to envy-free (full information) pricing

Main result

Envy-free pricing has an $O(\log n)$ approximation algorithm

Walrasian Equilibrium

Definition: *Walrasian Equilibrium*, an envy-free pricing with unallocated items at price zero.

Examples:

- for unlimited supply, all items at price zero.
- for limited supply, all items at price zero is not Walrasian Equilib.

Algorithm: *Vickrey-Clarke-Groves (VCG)*:

1. Allocate items via *maximum weighted matching (MM)*.
2. For (i, j) in matching, give item j to consumer i at price:

$$p_j = v_{i,j} - \text{MM}(V) + \text{MM}(V_{-i}).$$

Theorem: [Leonard 83] VCG outputs a Walrasian Equilibrium.

Reserve Prices

Definition: *reserve price*, a lower bound on the sale price of an item.

Reserve prices often used to obtain more revenue from VCG.
(e.g., Bayesian optimal auction [Myerson 81])

Walrasian + Reserve

Definition: *Walrasian Equilibrium with reserve price r* , a envy-free pricing with unallocated items at price r .

Algorithm: *Vickrey-Clarke-Groves with reserve price r (VCG_r)*:

1. Construct V' : add two dummy consumers for each item with valuation r .
(breaking ties in favor of real consumers)
2. Run VCG on V' and output prices.

Lemma: The VCG_r prices are a Walrasian Equilibrium with reserve price r .

Proof: If price for unsold item j is less than r ,

- in $VCG(V')$, one dummy consumer for item j would envy, thus
- prices of $VCG(V')$ would not be Walrasian for V' .

Log n Approximation

Lemma: If MM sells k items at price $\geq p$, then $\text{VCG}_p \geq kp/2$.

Proof:

- Consider $(i, j) \in \text{MM}$ at price above p .
- Either i or j is matched in VCG_p .
(otherwise we could add (i, j) to VCG_p)
- Therefore, number of matched vertices in VCG_p is $\geq k$.
- Number of matched items in VCG_p is $\geq k/2$.

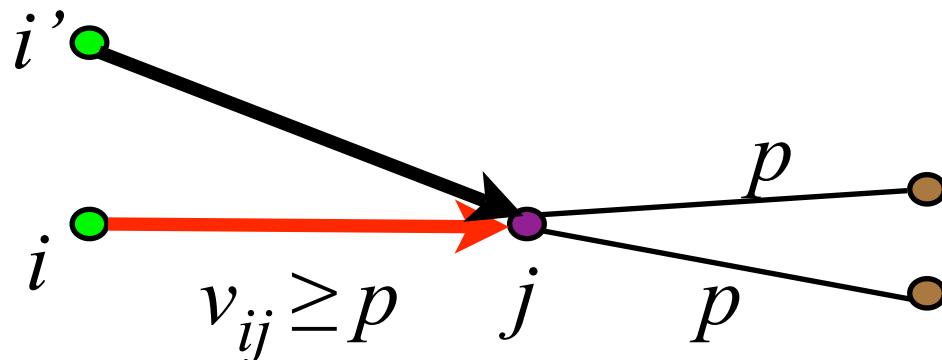
Algorithm: Limited Supply Logarithmic Approximation:

1. Run $\text{MM}(V)$ to compute prices p_1, \dots, p_m .
2. Output VCG_{p_i} with highest profit.

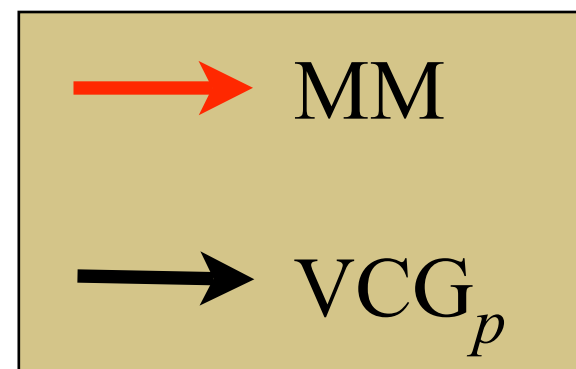
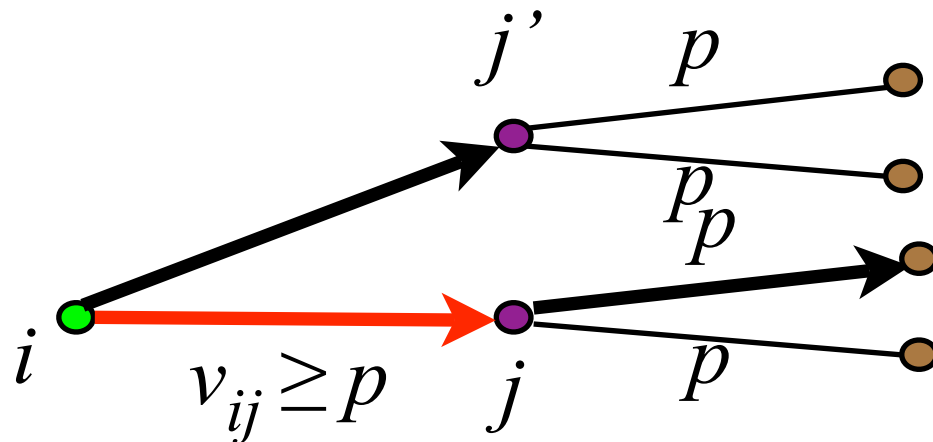
Analysis: Profit $\geq \max_i \frac{ip_i}{2} \geq \frac{\sum_i p_i}{2 \ln n} = \frac{\text{MM}}{2 \ln n} \geq \frac{\text{OPT}}{2 \ln n}$

Proof of lemma

if j is sold to a real consumer



if j is not sold to a real consumer



Open

- **Open:** Is there a matching lower bound?
- **Known:** APX-hard, by reduction from Vertex Cover in bounded degree graphs

Open Problem #1: Pricing over Time

A special case of Envy-free pricing.

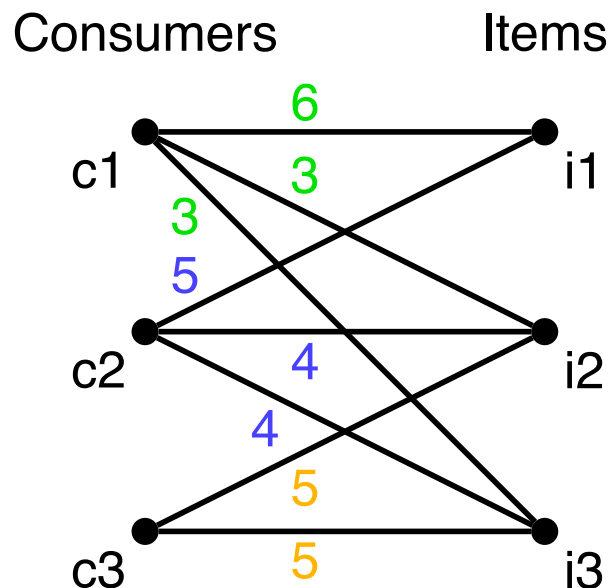
- Customer: “I want to buy a Boston-Bologna ticket between June 16 and June 18 and pay at most \$600.”
- supply c_t of seats available at date t ; customer = $[s_i, t_i]$, valuation v_i
- **Open problem:** design an algorithm that is better than the $\log n$ approximation (the case $c_t = 1 \forall t$ would already be interesting)
- **Not known:** not known to be NP-hard
- **Known:** Solvable by dynamic programming if unlimited supply $c_t \geq n \forall t$

Unlimited Supply

Definition: *unlimited supply* special case: the number of copies of each item is $> n$.

(E.g., pricing in-flight movies.)

Definition: Item j and j' are *identical* iff $v_{ij} = v_{ij'}$ for all i .



Unlimited supply implications:

- Identical items sold at same price \Leftrightarrow envy-free.
- Given prices, consumers pick favorite item.

Open problem #2: Unlimited supply envy-free pricing

- **Open problem:** Is there an $O(1)$ -approximation algorithm if all items are in unlimited supply?
- **Known hardness:** APX-hard
- **Known algorithm:** $O(\log n)$ -approximation algorithm by our main result or by the best-uniform-price algorithm
- **Failed attempt:** Linear programming relaxation + randomized rounding. Variables $x(i, j, p) = 1$ if customer i buys item j at price p . And $y(j, p) = 1$ if item j is offered at price p . Difficulty: our models all seemed to have unbounded integrality gap!

Open problem #3: Unlimited Supply Tollbooth problem

- Variation: customers want to buy bundles instead of single items.
- **The problem:** Items are edges forming a tree, bundles are paths, customer i wants to buy a specific path (or nothing at all) and pay at most v_i to buy all the edges on the path. Unlimited supply: edges have infinite capacity.
- **Known hardness:** APX-hard
- **Known algorithm:** $O(\log n)$ approximation
- **Known special case:** solvable by dynamic programming if all paths end at same node
- **Open problem:** Is there an $O(1)$ -approximation algorithm?

Open problem #4: Unlimited Supply Highway problem

- Customers want to buy bundles
- **The problem:** items are edges forming a path, bundles are subpaths, customer i wants to buy $[s_i, t_i]$ and pay at most an integer sum v_i to buy all the edges in the interval. Edges have infinite capacity.
- **No known hardness.**
- **Known algorithm:** $O(\log n)$ approximation
- **Known special case:** solvable by dynamic programming if all bundles have bounded length $t_i - s_i = O(1)$
- **Open problem:** Is there an $O(1)$ -approximation algorithm, or maybe even an exact algorithm?