



Non Clairvoyant Dynamic Mechanism Design

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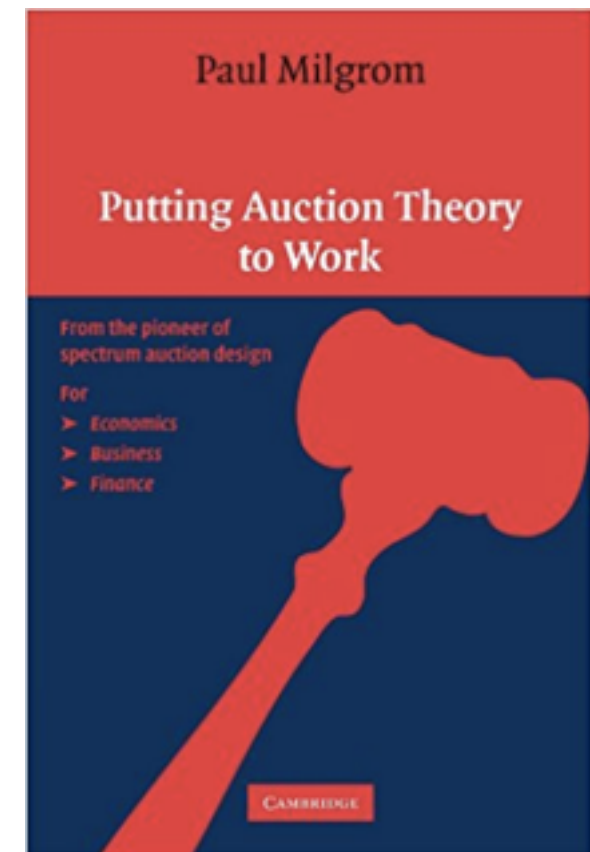
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Next generation of ad auction

- Classic auctions found their way to the web
- Designed for different domains: art, spectrum, ...
- Internet ad auctions are different: repeated and the buyer cares about the aggregate result.
- Why use dynamic auctions ?
 - Can improve both revenue and efficiency over static auctions (no tradeoffs)
 - Can generate arbitrarily more revenue than static auctions.
 - Combines the best of real time auctions and guaranteed contracts.



Towards practical dynamic auctions

- Current state:
 - beautiful mathematical theory [...]
 - polynomial time algorithms [PPPR], [ADH]
 - understanding of competition complexity [LP]
- Barriers to a practical implementation:
 - DP / LP solutions are not scalable
 - relies on accurate forecasts
 - assumes too much of buyer rationality / knowledge

Repeated Auctions Model

- Single buyer model

- For timestep $t = 1 \dots T$
 - item arrives (ad impression)
 - buyer observes his type $v_t \sim F_t$
(sellers $\leftarrow$ public info, buyer $\leftarrow$ public info + private cookies)
 - agent reports value $\hat{v}_t$
 - allocation with probability $x_t(\hat{v}_{1..t})$ and pays $p_t(\hat{v}_{1..t})$
 - buyer gets utility $u_t^{v_t} = v_t x_t(\hat{v}_{1..t}) - p_t(\hat{v}_{1..t})$

- Buyer wants to maximize continuation utilities

$$u_t^{v_t}(\hat{v}_{1..t}; F_{1..T}) + \mathbb{E}_{F_{t+1..T}} \left[\sum_{\tau=t+1}^T u_{\tau}^{v_{\tau}}(\hat{v}_{1..\tau}; F_{1..T}) \right]$$

Design Space

- The auction is represented by allocation and payments:

$$x_t : \Theta^t \times (\Delta\Theta)^T \rightarrow [0, 1] \quad x_t(\hat{v}_{1..t}; F_{1..T})$$

$$p_t : \Theta^t \times (\Delta\Theta)^T \rightarrow \mathbb{R}_+ \quad p_t(\hat{v}_{1..t}; F_{1..T})$$

- Constraints:

- Dynamic Incentive Compatibility (DIC)

$$v_t \in \operatorname{argmax}_{\hat{v}_t} u_t^{v_t}(\hat{v}_t \dots) + \mathbb{E}_{F_{t+1..T}} [\sum_{\tau=t+1}^T u_{\tau}^{v_{\tau}}(\hat{v}_t \dots)]$$

- Ex-post Individual Rationality (ep-IR) $\sum_t u_t \geq 0$

- Objective function: $\text{REV}^*(F_{1..T}) = \max \mathbb{E}_{F_{1..T}} [\sum_t p_t(v_{1..t})]$

Cassandra's curse

- Optimal mechanism requires seller to know all distributions in advance (to solve the DP).
- The definition of DIC require buyer and seller to agree on distributions $F_{t+1}, F_{t+2}, \dots, F_T$.
- Can we get mechanism that doesn't require common knowledge about the future ?



-  DIC:

$$v_t \in \operatorname{argmax}_{\hat{v}_t} u_t^{v_t}(\hat{v}_t \dots) + \mathbb{E}_{F_{t+1} \dots F_T} [\sum_{\tau=t+1}^T u_{\tau}^{v_{\tau}}(\hat{v}_t \dots)]$$

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- Can we get mechanism that doesn't require common knowledge about the future ?
- Super-DIC: for any beliefs $\tilde{F}_{t+1..T}$
$$v_t \in \operatorname{argmax}_{\hat{v}_t} u_t^{v_t}(\hat{v}_t \dots) + \mathbb{E}_{\tilde{F}_{t+1..T}} [\sum_{\tau=t+1}^T u_{\tau}^{v_{\tau}}(\hat{v}_t \dots)]$$
- Theorem (Cassandra's curse): Under super-DIC the revenue optimal mechanism is the optimal static auction.



Non-Clairvoyance

- **Non-Clairvoyance:** mechanism is measurable with respect to
i.e. $v_{1..t}, F_{1..t}$ $x_t(v_{1..t}; F_{1..t}), p_t(v_{1..t}; F_{1..t})$

- Entangled design: consider two items sequences:

$$\left[\begin{array}{c} \text{🍏} \\ F_a \end{array} \quad \begin{array}{c} \text{🍇} \\ F_g \end{array} \right] \quad \left[\begin{array}{c} \text{🍏} \\ F_a \end{array} \quad \begin{array}{c} \text{🍊} \\ F_o \end{array} \right]$$

the non-clairvoyant mechanism needs to use the same rule for item 1. The clairvoyant can use different rules depending on what comes next.

- **DIC for Non-Clairvoyant:** buyers don't need to know the future to check DIC. The only requirement is that distribution F_t will be common knowledge in step t .

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Non Clairvoyant Revenue Approx

- **Benchmark:** the optimal dynamic mechanism that knows all the distributions $\text{REV}^*(F_{1..T})$.
- A NonClairvoyant auction is an α -approximation if for all distributions $F_{1..T}$ we have

$$\text{REV}(F_{1..T}) \geq \alpha \text{REV}^*(F_{1..T})$$

Non Clairvoyant Revenue Approx

Theorem: Every non-clairvoyant policy is at most a $1/2$ -approximation to the optimal clairvoyant revenue.

Theorem: For multiple buyers there is a non-clairvoyant policy that is at least $1/5$ -approx to the opt clairvoyant.

Theorem: Can be improved to $1/2$ for two periods and for $1/3$ for one buyer and multiple periods.

Technique: Bank Account Mechanisms

Theorem: Every non-clairvoyant policy is “isomorphic” to a bank account mechanism.

- Keeps a state variable b_t (balance) for each buyer
- Chooses a per-period IC mechanism based on balance

$$x_t(v_t, b_t), p_t(v_t, b_t)$$

with the balance-independence property

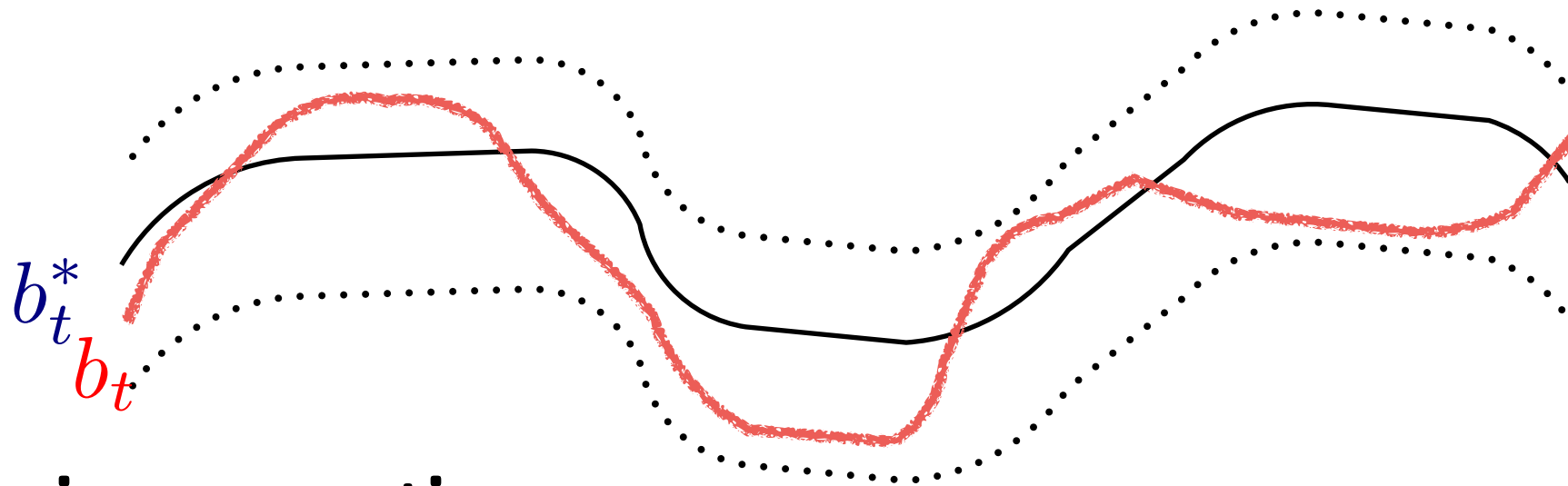
$$\mathbb{E}[v_t x_t(v_t, b_t) - p_t(v_t, b_t)] = \text{const} \geq 0$$

- Updates balance:

$$0 \leq b_{t+1} \leq b_t + [v_t x_t - p_t]$$

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Other nice properties:

- framework to design and prove lower bounds on dynamic mechanisms
- computationally efficient (multi-buyer, multi-item)
- no pre-processing required (LP or DP)

1/3-approximation policy

Keep a variable b called balance initialized to zero.

For every period t , receive an item with distribution F_t

Sell 1/3 of the item with each of the following auctions:

- Myerson's auction for F_t
- Give the item for free and increment balance $b = b + v_t$
- For $f = \min(b, \mathbb{E}_{F_t}[v_t])$
charge f before the buyer can see the item
post a price of r such that $\mathbb{E}(v_t - r)^+ = f$
decrement balance $b = b - f$

Balance independence property: $\mathbb{E}[\text{utility}]$ is balance independent.

Extension to Multiple buyers

Single buyers (1/3 approx)

1/3 item: Myerson

1/3 item: Give for free

1/3 item: Dynamic posted price

Multiple buyers (1/5 approx)

1/3 item: Myerson

2/3 item: Second price auction

2/3 item: Dynamic money
burning auction [HR]

Thanks

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<https://ssrn.com/abstract=2873701>