

# Contextual Standard Auctions with Budgets

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## Motivation

GOOGLE AD MANAGER

Rolling out first price auctions to Google Ad Manager partners

- **First-Price Auctions:** Major ad platforms have recently shifted to first-price auctions as the preferred auction format.
- **Contextual Values:** Users have features which determine the value of advertisers.
- **Budgets:** Buyers often have caps on average/total expenditure.

## Model

- **Setup:** Seller (ad platform) auctions one indivisible item among  $n$  buyers (advertisers).
- **Feature-based Values:** Item type represented by feature vector  $\alpha \in A \subset \mathbb{R}^d$ , buyer type represented by weight vector and budget  $(w, B) \in \Theta \subset \mathbb{R}^d$ . Value is given by  $w^T \alpha$ .
- **Bayesian Setting:** Buyer types are drawn from commonly known distribution  $G$ , item types are drawn from commonly known distribution  $F$ , buyer types are private information, item type is revealed before auction

## Symmetric First Price Equilibrium

A strategy  $\beta^* : \Theta \times A \rightarrow \mathbb{R}_{\geq 0}$  such that  $\beta^*(w, B, \alpha)$  is an optimal solution to the following optimization problem a.s.  $(w, B) \sim G$ :

$$\begin{aligned} \max_{b: A \rightarrow \geq 0} \quad & \mathbb{E}_{\alpha, \{\theta_i\}_{i=1}^{n-1}} [(w^T \alpha - b(\alpha)) 1\{b(\alpha) \geq \{\beta^*(\theta_i, \alpha)\}_i\}] \\ \text{s. t.} \quad & \mathbb{E}_{\alpha, \{\theta_i\}_{i=1}^{n-1}} [b(\alpha) 1\{b(\alpha) \geq \{\beta^*(\theta_i, \alpha)\}_i\}] \leq B. \end{aligned}$$

## Pacing Based SFPE

### • IID, No Budgets, No Contexts:

- ▶ Values of all buyers are i.i.d.  $v_i \sim H$
- ▶ Well-known: Symmetric equilibrium exists and is given by

$$\sigma_H(v_i) = E \left[ \max_{j \neq i} v_j \mid \max_{j \neq i} v_j < v \right]$$



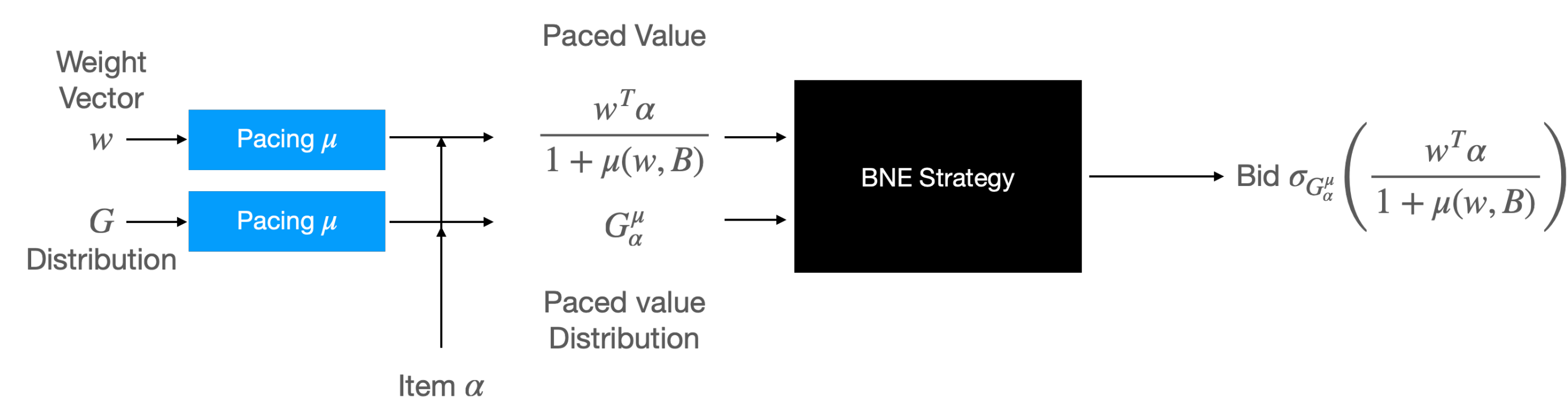
- **Pacing:** Each buyer type  $(w, B)$  has an associated dual variable  $\mu(w, B)$  corresponding to the budget constraint in her optimization problem, her paced value for item  $\alpha$  is given by

$$\frac{w^T \alpha}{1 + \mu(w, B)}$$

- **Pacing Based Strategy:** Each buyer treats her paced value as her true value, assumes that every other buyer does the same, and bids according to  $\sigma_\alpha$

$$\beta(w, B, \alpha) = \sigma_{G_\alpha^\mu} \left( \frac{w^T \alpha}{1 + \mu(w, B)} \right)$$

- **Theorem:** There exists pacing function  $\mu : \Theta \rightarrow \mathbb{R}_{\geq 0}$  such that the pacing based strategy corresponding to  $\mu$  is a SFPE.



- **Novel Proof Technique:** We use a fixed point theorem in the infinite-dimensional space of dual multipliers  $\mu$  by exploiting topological properties of multi-variable functions of bounded variation.

## Standard Auctions and Revenue Equivalence

- **Theorem:** For anonymous standard auctions other than first-price auctions: the BNE strategy (**black box**) changes, but the same equilibrium pacing function (**blue box**) remains the same, leading to revenue equivalence. **This is surprising because revenue equivalence does not hold for strict budget constraints.**

## Price of Anarchy of Liquid Welfare

- **Liquid Welfare:** Maximum revenue that can be extracted from the buyers by an omniscient seller with complete knowledge of their values.
- **Theorem:** Value-pacing-based equilibria of all standard auctions have a Price of Anarchy  $\geq 1/2$  for liquid welfare.

## Numerical Simulations

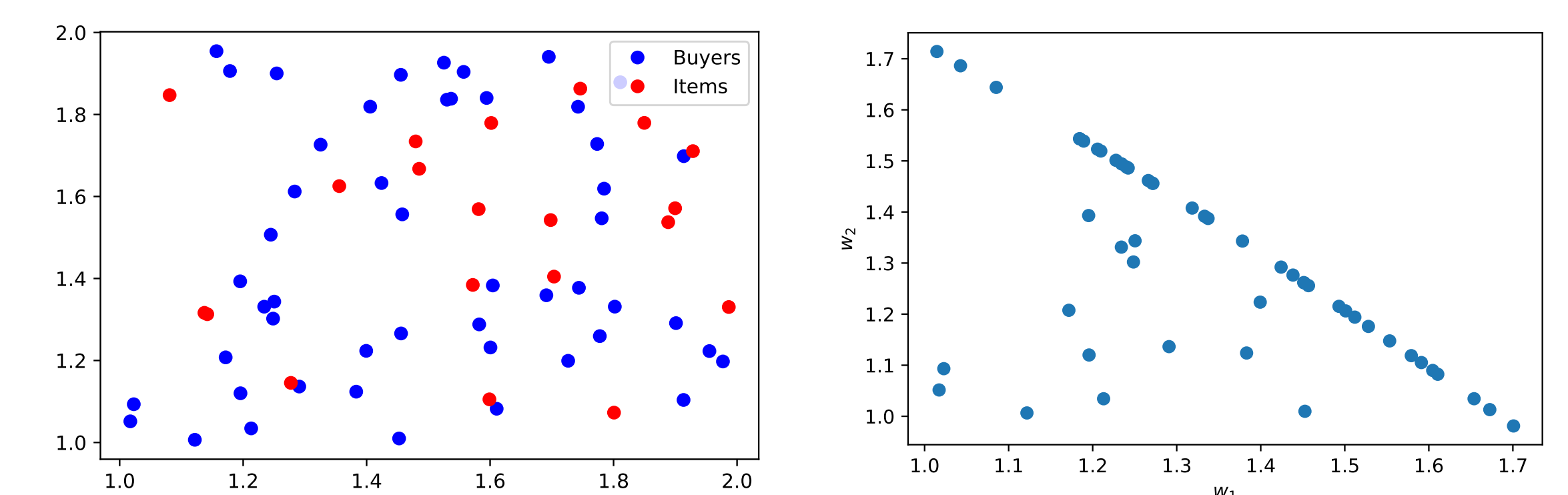


Figure 1: The first plot depicts the set of possible buyer and item types. 3 buyers participate in the auction, each sampling her weight vector uniformly from the blue points, and the item is sampled uniformly from the red points. All buyers have budget  $B = 2$ . The second plot shows the paced weight vectors of the buyers in the SFPE, which was computed through best-response dynamics in the dual space. **In our structural results, we explain why weight vectors get paced down to the same level, as seen in the above plots.**