# CEMMAP Lectures on Contracts & Market Microstructure

2. Sealed Bid Auctions

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How useful are models of competitive equilibrium in assessing conduct and performance?

- The two welfare theorems nicely encapsulate the close connection between exploiting the potential gains from trade and a price vector separating suppliers from demanders.
- But how is price set in competitive equilibrium?
  - We imagine a fictional Walrasian auctioneer
  - estting market clearing prices
  - 9 by maximizing a weighted sum of individual utilities
  - to mimic a *centralized* omniscient social planner.
- Competitive equilibria are not *implementable*:
  - loosely speaking a noncooperative game with a Nash equilibrium that reproduces the competitive allocations does not exist.
- Furthermore the empirical results inferred from structural econometric models of competitive equilibrium, about how well the private economy performs, are ambiguous.

- This is a motivation for exploring the institutional structure of markets, digging more deeply into how markets function.
- An obvious place to start these investigation are auction mechanisms:
  - **1** There is single agent on one side of the market, the auctioneer.
  - In the simplest case only one (indivisible) is traded.
  - On apply standard noncooperative definitions of games
    - players, moves, information sets and payoffs
    - Nash equilibrium refinements (such as perfection)
- (Variations on) these models can show how:
  - the number of traders on the other side of the market determine the equilibrium.
  - endogenous search for a market affects the trading outcome.
  - 9 what factors affect the choice of the trading mechanism.

#### • Strategically equivalent auctions:

- first price sealed bid (FPSB) auctions (highest bidder wins and pays his bid).
- Dutch auctions (auctioneer reduces the price until a bidder accepts the offer)

• Strategically equivalent auctions when there are private values:

- second price sealed bid (SPSB) auctions (highest bidder wins, paying highest losing bid).
- Japanese (button) auctions players exit as the auctioneer raises the price

(last remaining bidder wins at price second last exits).

**Basic framework** 

- We first consider a first price sealed bid (FPSB) auction:
  - for *N* players (predetermined outside the model)
  - with independent private values.
- By FPSB we mean that each player  $n \in \{1, ..., N\}$  simultaneously submits a bid denoted by  $b_n \in \mathbf{R}^+$ , and that the player submitting the highest bid is awarded the (single) object up for auction, and pays what he or she bid.

#### First Price Sealed Bid (FPSB) Auction

Best replies in equilibrium

• Let W(b) denote the probability of winning the auction with bid b. That is:

$$W(b) \equiv \mathsf{Pr}\left\{b_k \leq b ext{ for all } k = 1, \dots, N
ight\}$$

• Then the maximization problem faced by player *n* can be written as:

$$\max_{b}(v_{n}-b)W(b)$$

• The first order condition (FOC) is:

$$(v_n - b_n)W'(b_n) - W(b_n) = 0$$
<sup>(1)</sup>

• The second order condition (SOC) of the optimization problem is:

$$0 > SOC \equiv \frac{\partial}{\partial b} FOC = \frac{\partial}{\partial b} [(v - b)W'(b) - W(b)]$$
$$= (v - b)W''(b) - 2W'(b)$$

• Totally differentiating the FOC with respect to b and v yields:

$$0=W'\left(b_{n}\right)dv_{n}+\left[\left(v_{n}-b_{n}\right)W''\left(b_{n}\right)-2W'\left(b_{n}\right)\right]db_{n}$$

and hence:

$$\frac{db_{n}}{dv_{n}} = \frac{-W'(b_{n})}{(v_{n} - b_{n})W''(b_{n}) - 2W'(b_{n})} > 0$$

because  $W'(b_n) > 0$  and the denominator of the quotient is the SOC.

 We infer that if players are in a pure strategy equilibrium with an interior solution, then b<sub>n</sub> is increasing in v<sub>n</sub>.

## First Price Sealed Bid (FPSB) Auction

Bayesian Nash Equilibrium with monotone bidding

- From now on we assume that players are in a (pure strategy) Bayesian equilibrium with bids that are monotone increasing in valuations.
- That is we consider Bayesian Nash Equilibrium (BNE) in which bidders follow a strategy β : V → B ≡ [0,∞) where β (v) is increasing in v.
- Then  $\beta(v)$  has an inverse, which we denote by  $\alpha : \mathbf{B} \to \mathbf{V}$  such that  $\alpha [\beta(v)] = v$  for all v.
- Letting G(b) denote the distribution of bids, it follows that:

$$W(b) \equiv \Pr \left\{ b_k \leq b_n \text{ for all } k = 1, \dots, N \right\} = G(b_n)^{N-1}$$

• From the monotonicity property of the BNE:

$$G(b) = F(\alpha(b))$$

## Identifying IPV from FPSB auctions

From the probability of winning

- Assume our data set consists of all the bids recorded in *I* auctions in which the same equilibrium is played.
- Let  $b_n^i$  for  $n \in \{1, ..., N\}$  and  $i \in \{1, ..., I\}$  denote the bid by player n in the  $i^{th}$  auction.
- Suppose the probability of winning the  $i^{th}$  auction,  $W_i(b)$ , and hence its derivative  $W'_i(b)$ , are identified.
- Rewriting the FOC (1) then proves  $v_n^i$  is identified:

$$v_n^i = b_n^i + \frac{W_i\left(b_n^i\right)}{W_i'\left(b_n^i\right)}$$
<sup>(2)</sup>

- For example assume v<sub>n</sub> ∈ V is iid (IPV) drawn from F(v), then W(b), v<sup>i</sup><sub>n</sub> and F(v) are identified.
- This logic extends to models where  $v_n$  is drawn from  $F(v | z_n)$  where:
  - each bidder knows  $z_n$ , some background variables.
  - bidders know, or do not know, the z values of their rivals.
  - The analyst observes  $z_n$  and bidders' information sets.

### Identifying IPV from FPSB auctions

From the bid distribution (Guerre, Perrigne and Voung, 2003)

- Alternatively note that the probability distribution of bids and its density,  $G_i(b)$  and  $G'_i(b)$ , are identified.
- Focusing on the IPV case (purely for notational ease) the probability *n* wins with *b<sub>n</sub>* is:

$$W(b_n) = G(b_n)^{N-1}$$

implying

$$W'(b_n) = (N-1) G(b_n)^{N-2} G'(b_n)$$

• We rewrite the FOC, Equation (1) as:

$$v_{n}^{i} = b_{n}^{i} + \frac{W(b_{n}^{i})}{W'(b_{n}^{i})} = b_{n}^{i} + \frac{G(b_{n}^{i})}{(N-1)G'(b_{n}^{i})}$$
(3)

• This shows  $v_n^i$  and hence F(v) can also be directly identified off the bidding distribution G(b).

#### Identifying the IPV Distribution from FPSB auctions The distribution of winning bids

- Now suppose our data set consists of only the winning bid recorded in *I* auctions in which the same equilibrium is played.
- Let  $b^i$  for  $i \in \{1, ..., I\}$  denote the winning bid in the  $i^{th}$  auction.
- Thus the distribution of winning bids, denoted by  $H(b^i)$ , is identified.
- Since the winning bid is defined as the highest one, H(b) is just the probability that all the bids are less than b, implying:

$$H\left(b
ight)=\mathsf{Pr}\left\{b_{n}^{i}\leq b ext{ for all } n=1,\ldots,N
ight\}=\mathcal{G}(b)^{N}$$

• Consequently:

$$G(b) = H(b)^{\frac{1}{N}}$$
(4)

and

$$G'(b) = \frac{1}{N} H(b)^{\frac{1}{N}-1} H'(b)$$
(5)

• This shows the bidding distribution is identified from the data generating process of the winner's bid.

#### Identifying the IPV Distribution from FPSB auctions Identification when only the winning bid is observed

• Substituting Equations (4) and (5) back into Equation (3) gives:

$$v^{i} = b^{i} + \frac{G(b^{i})}{(N-1) G'(b^{i})} = b^{i} + \frac{NH(b)}{(N-1) H'(b)}$$

- This identifies the winning valuations, and hence their distribution, denoted by  $F_W(v)$ .
- But the distribution of the winning valuations is a one to one mapping of the distribution of all the valuations:

$$F_W(v) = \Pr\left\{v_n \leq v \text{ for all } n = 1, \dots, N\right\} = F(v)^N$$

• Therefore F(v) is identified off the winning bids alone using the equation:

$$F(v) = F_W(v)^{\frac{1}{N}}$$

### Second Price Sealed Bid (SPSB) Auctions

A weakly dominant strategy

- Now suppose as before:
  - each bidder knows her own valuation;
  - makes sealed bid (that is bids simultaneously).
- But instead of a FPSB auction, consider a SPSB auction, where the highest bidder wins the auction but only pays the second highest bid.
- Now it is a weakly dominant strategy for (each) *n* to bid her expected valuation, *v<sub>n</sub>*.
- Intuitively, compared with bidding  $v_n$ :
  - bidding more implies winning some auctions that yield negative expected value, but leaves unchanged the expected value of any other auction that would be won;
  - bidding less implies losing some auctions that yield positive expected value, but leaves unchanged the expected value of any other auction that she would win.

# Second Price Sealed Bid (SPSB) Auctions

#### A picture proof



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- Now assume F(v) is the common distribution of valuations as before.
- Note first the obvious point that because players bid their valuations in SPSB auctions with private valuations, F(v) is trivially identified if all the bids are observed.
- Now suppose only the winning price is observed.
- Then the probability distribution of the second highest valuation, which we now denote by  $F_{N-1,N}(v)$ , is identified.

# Identifying the IPV Distribution from SPSB Auctions

Identification when only the winning bid is observed

- More generally, let  $F_{i,N}(v)$  denote the distribution of the  $i^{th}$  order statistic.
- One can show (Arnold, Balakrishnan and Nagaraja,1992 for example):

$$F_{i,N}(v) = \frac{N!}{(N-i)!(i-1)!} \int_0^{F(v)} t^{i-1} \left(1-t\right)^{N-i} dt \qquad (6)$$

- We can interpret (6) as a mapping from distribution functions in valuations into distribution functions of ordered valuations.
- To prove F(v) is identified, fix  $v_0$  and note:
  - $F_{i,N}(v_0)$  is identified (and a consistent estimator is obtained from the sample distribution).
  - **②** The right side of (6) is strictly increasing in candidate values of  $F(v_0)$  since  $t^{i-1} (1-t)^{N-i} > 0$  for all  $t \in [0, 1]$ .
  - **③** Therefore the LHS and RHS of (6) cross only once at  $v_0$ .
  - Hence the mapping defining  $F_{i,N}(v)$  from F(v) is invertible.

Revisiting best replies in SPSB auctions with private values

- Suppose there are N risk neutral bidders. Bidder n:
  - has valuation  $v_n$ , the utility gain from winning the auction.
  - receives signal  $x_n \equiv v_n + \epsilon_n$ , where  $E[\epsilon_n | x_n] = 0$ .
- The literature focuses on perfect Bayesian equilibria in weakly undominated pure strategies (Athey and Haile, 2006).
- Let  $b_n \equiv \beta_n(x_n, N)$  denote the equilibrium strategy of bidder *n*.
- In a second price auction with private values, it is a weakly dominant strategy for (each) *n* to bid his expected valuation, setting:

$$\beta_n(x_n, N) = x_n \equiv E[v_n | x_n]$$

• Note the same logic applies to *n* individually if  $v_n = x_n$ , regardless of the correlation structure between valuations and the other bidders' information.

#### Bidder Information

Revisiting best replies in FPSB auctions with private values

 Denote the bid distribution function for the maximum equilibrium bid of the n<sup>th</sup> bidder's rivals, which is independent of x<sub>n</sub>, as:

$$G(b_m | n, N) = \Pr\left[\max_{n' \in N \setminus n} \{b_{n'}\} \le b_m | n, N
ight]$$

• Then *b<sub>n</sub>* solves:

$$b_n = \arg\max_b \int_{-\infty}^b (x_n - b) G'(b_m | n, N) db_m$$

• The first order condition is:

$$x_n = b_n + \frac{G(b_n | n, N)}{G'(b_n | n, N)}$$

- This FOC reduces to (2) when v<sub>n</sub> = x<sub>n</sub> and the valuations of the bidders are *iid*.
- In both cases both  $W(b_n)$  and  $G(b_n | x_n, N)$  represent the probability of *n* winning the auction with bid  $b_n$ .

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#### **Bidder Information**

Best replies in FPSB auctions with common values

• Similar to the private values case, define:

$$\widetilde{G}\left(b_{m}\left|x_{n},\,n,\,N
ight)=\mathsf{Pr}\left[\max_{n^{\prime}\in\mathcal{N}\setminus n}\left\{b_{n^{\prime}}
ight\}\leq b_{m}\left|x_{n},\,n,\,N
ight]
ight]$$

• Suppose the equilibrium is symmetric, and

- denote by  $\beta(x_{n'}, N)$  denote of the bid function of n'.
- suppose  $\beta(x_{n'}, N)$  is strictly increasing in  $x_{n'}$ .
- denote by  $\alpha(b_{n'}, N)$  the inverse of  $\beta(x_{n'}, N)$  in  $x_{n'}$ .
- Then the optimal bid  $b_n$  solves:

$$b_{n} = rgmax_{b} \int_{-\infty}^{b} \left[ v_{n}\left(x_{n}, \alpha\left(b_{m}, N
ight), N
ight) - b 
ight] \widetilde{G}'\left(b_{m}\left|x_{n}, n, N
ight) db_{m}$$

where 
$$v_n(x_n, x_m, N) = E\left[v_n \middle| x_n \text{ and } \max_{n' \in N \setminus n} \{b_{n'}\} = \beta(x_m, N)\right]$$

• Superficially, this FOC resembles the private value auction FOC:

$$v_n(x_n, x_n, N) = b_n + \widetilde{G}(b_n | x_n, n, N) / \widetilde{G}'(b_n | x_n, n, N)$$

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### Identification in FPSB Auctions with Private Values

When all the bids are observed

• Assume  $x_n = v_n$ . From the first order condition:

$$x_n = b_n + \frac{G(b_n | n, N)}{G'(b_n | n, N)}$$

• Recall from its definition that  $G_{m_n}(b_n | n, N)$  is the probability that n wins the auction with  $b_n$ :

$${{G_m}\left( {{b_n}\left| {n,N} 
ight.} 
ight)} = \Pr \left[ {\mathop {\max }\limits_{{n' \in N \setminus n}} {\left\{ {{b_{n'}}} 
ight\}} \le {b_n}\left| {n,N} 
ight.} 
ight]$$

- Thus if all the bids are observed then  $G_m(b_n | x_n, N)$  is identified.
- Hence  $v_n$  is identified (for all bidders in each sampled auction).
- Therefore the probability distribution of  $(v_1, \ldots, v_N)$  in this specialization is identified for any correlation structure.

### Identification Fails in Common Value FPSB Auctions

When all the bids are observed

• Recall that we defined:

$$v_{n}(x_{n}, x_{m}, N) = E\left[v_{n} \left| x_{n} \text{ and } \max_{n' \in N \setminus n} \left\{ b_{n'} \right\} = \beta_{n}(x_{m}, N) 
ight]$$

and derived:

$$v_{n}(x_{n}, x_{n}, N) = b_{n} + \frac{\widetilde{G}(b_{n}|x_{n}, n, N)}{\widetilde{G}'(b_{n}|x_{n}, n, N)}$$

- The basic problem is that the function  $v_n(x_n, x_m, N)$  is only identified at the point  $x_n = x_m$ , not any  $x_m$ .
- Note that every common value model is observationally equivalent to a private value model found by setting  $v_n = v_n (x_n, x'_n, N)$ .
- Thus two common value models with possibly different  $v_n(x_n, x_m, N)$  but the same  $v_n(x_n, x_n, N)$  are (also) observationally equivalent.