Mechanism Design with Ambiguity: IMSI Tutorial (Module 2 of Decision Making and Uncertianty)

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- Principal may simply ask for the information, but agents may not report it truthfully unless the principal gives them incentives to do so by monetary payments or some other instrument that she controls.
- But provision of incentives involve costly trade-offs... hence the question designing game forms that provide incentives to elicit private information optimally, e.g., maximizing expected revenue in an auction.

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- We start by explaining the basic idea the design optimal auction in the regular case and then discuss how the design has to be adjusted to take into account ambiguity averse agents.

Optimal Auction Problem in the "Regular Case"

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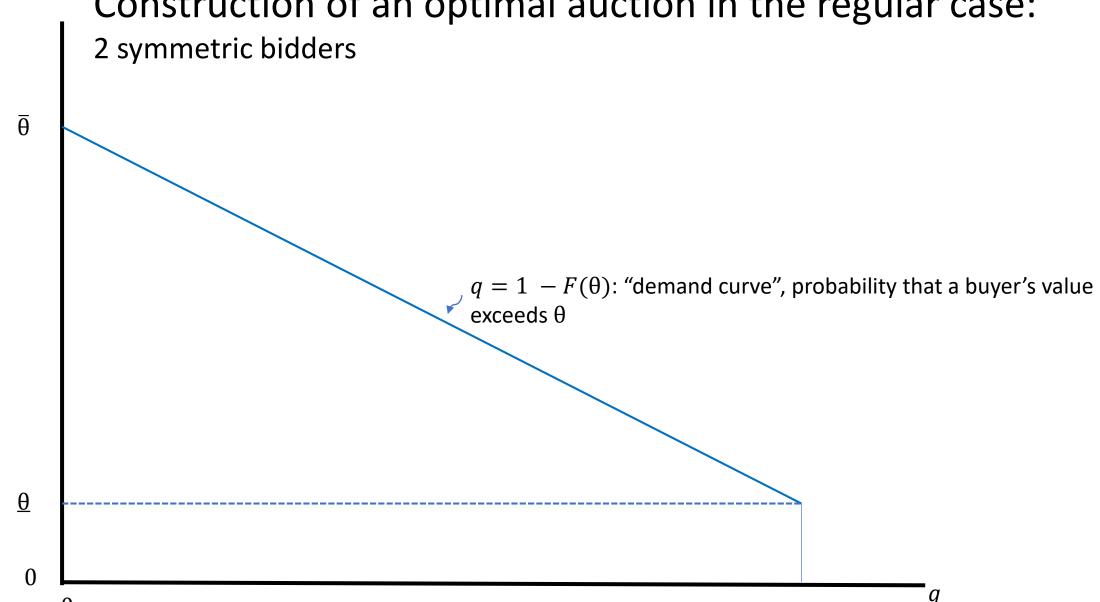
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- The seller has a value known to be zero.
- What sales mechanism will maximize the seller's expected profits?

■ For each bidder, graph the inverse of his cumulative distribution function, with value θ on the Y or "price axis" and the probability that the buyer's value exceeds a certain value, $1 - F(\theta) \equiv q$, on the X or "quantity axis."

Construction of an optimal auction in the regular case:



- For each bidder, graph the inverse of his cumulative distribution function, with value θ on the Y or "price axis" and the probability that the buyer's value exceeds a certain value, $1 - F(\theta) \equiv q$, on the X or "quantity axis."
- For each bidder, we then have something that looks like a demand curve, with the bidder's maximum possible value being the price at a quantity of zero and the bidder's minimum possible value being the price at a quantity of one.

■ From the demand curve for each bidder draw a "marginal revenue" curve, calculated the way we always calculate marginal revenue curves from demand curves: Multiply "quantity," $q = 1 - F(\theta)$, times "price," $\theta = F^{-1}(1 - q)$, and take the derivative with respect to "quantity":

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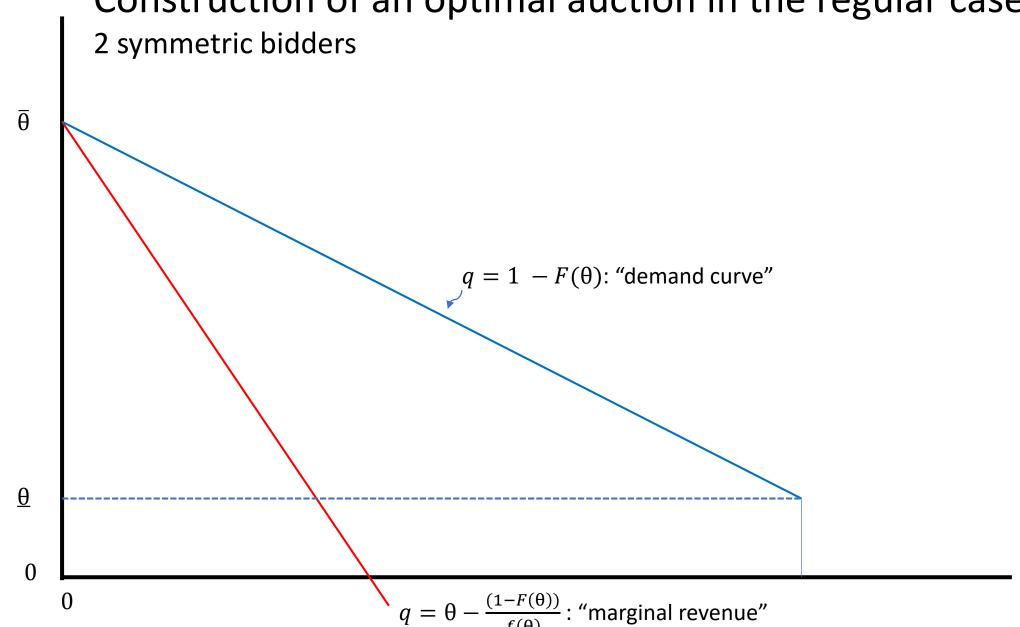
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■ Define $MR(\theta) = -\infty$ for all $\theta < \underline{\theta}$. Assume, $MR(\theta)$ is monotonically increasing in θ .

Conduct the following second-price auction with reserve price:

■ Have each bidder announce his value. The reserve price is set at r, s.t. MR(r) = 0.

Optimal /tuction

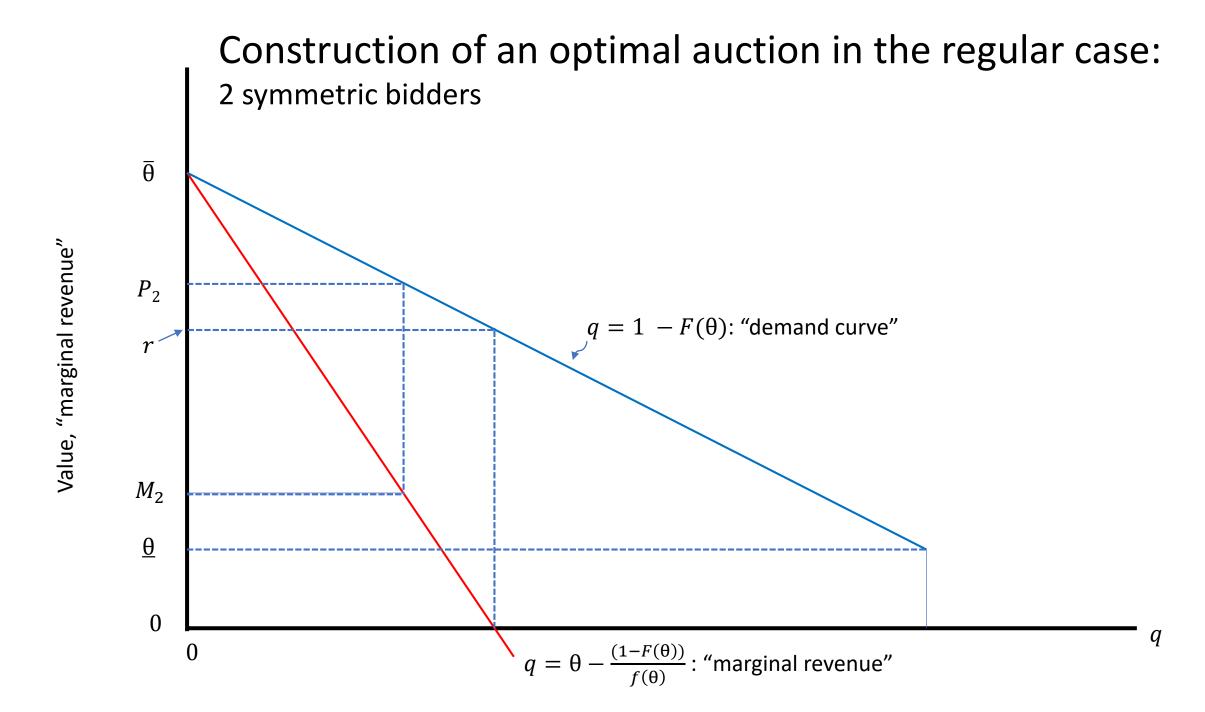
Regular case contd.

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- If more than one bidder has a +ve MR, the highest bidder gets the item at a price $P = MR^{-1}(M_2)$, where M_2 is the MR of the 2nd highest bid: winner pays the lowest value he could have had and still have had the maximal marginal revenue.

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- The other buyer pays nothing.



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 - while any lie that does not affect the outcome also does not affect the amount paid by the liar and so does not change the liar's utility.

The case with ambiguity averse bidders

Bose, Ozdenoren and Pape (2006)

2 risk-neutral potential buyers, as before, except now they have maxmin expected utility preferences:

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- The set ∆_B represents each bidder's belief about the other bidder's valuation.
- lacksquare The seller is ambiguity neutral, i.e., $\triangle_{\mathcal{S}} = \{ F \}$.
- We assume that the seller's distribution F is a focal point, and bidders allow for an ε -order amount of noise around this focal distribution.

• We consider a direct mechanism where bidders simultaneously report their types. The mechanism stipulates a probability $x_i(\widetilde{\theta}, \theta')$ for assigning the item and a transfer rule $t_i(\widetilde{\theta}, \theta')$ as a function of reported types.

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- We assume that type θ of bidder i chooses a report $\widetilde{\theta}$ to maximize

$$\min_{G \in \triangle_{B}} \int \left(x_{i}(\widetilde{\theta}, \theta')\theta - t_{i}(\widetilde{\theta}, \theta') \right) dG \left(\theta' \right)$$

Seller finds a mechanism $\left\{(x_i,t_i)_{i=1,2}\right\}$ that solves,

$$\max_{\left\{(x_{i},t_{i})_{i=1,2}\right\}}\int\left(t_{1}(\theta,\theta')+t_{2}(\theta,\theta')\right)dF(\theta)dF\left(\theta'\right)\text{ s.t.}$$

$$(IC)\quad:\quad\min_{G\in\triangle_{B}}\int\left(x_{i}(\theta,\theta')\theta-t_{i}(\theta,\theta')\right)dG\left(\theta'\right)$$

$$\geq\quad\min_{G\in\triangle_{B}}\int\left(x_{i}(\widetilde{\theta},\theta')\theta-t_{i}(\widetilde{\theta},\theta')\right)dG\left(\theta'\right)$$

$$(IR)\quad:\quad\min_{G\in\triangle_{B}}\int\left(x_{i}(\theta,\theta')\theta-t_{i}(\theta,\theta')\right)dG\left(\theta'\right)\geq0$$

where $(x_1(\theta, \theta') + x_2(\theta', \theta)) \le 1$, for all $\theta, \theta' \in [\underline{\theta}, \overline{\theta}]$.

DEF A full insurance mechanism is one where the (ex post) payoff of a given type of a bidder does not vary with the report of the competing bidder. That is, $\{(x_i, t_i)_{i=1,2}\}$ is a full insurance mechanism if, for almost all θ , $(x_i(\theta, \theta')\theta - t_i(\theta, \theta'))$ is constant as a function of θ' .

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- E.g. seller can make strict gains by switching to a full insurance auction from the optimal auction in the regular case.

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- The seller recognizes that he & bidders have different beliefs and will offer "side bets" using transfers.
- The seller can always adjust the transfers of type θ so that, under truth telling, θ gets the same minimum expected utility as he gets in the original mechanism in every state, and thus is fully insured against ambiguity in the new mechanism.

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- No other type wants to imitate θ in the new mechanism: the original mechanism is incentive compatible and imitation in the new mechanism is even worse given that the difference in transfers has weakly positive expected value under any distribution in \triangle_B .
- The seller's distribution is not in the minimizing set for the original mechanism, which means the additional transfers (to the seller) must have strictly positive expected value under the seller's distribution. Thus the seller is strictly better off.

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 \blacksquare First price auction with reserve price r

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- All bidders (regardless of having won or lost) who have bid above the reservation price receives a gift from the seller. For a bidder who bids, say, b (b > r), the amount of the gift is $S(b) = (1 - \varepsilon) \int_{a}^{b} F(y) dy$.

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Mechanism design with ambiguous communication

- Even when a social choice function is not implementable given (possibly ambiguous) prior beliefs, the posterior beliefs resulting from the communication may be such that it becomes implementable.
- Main result shows, in a nutshell, for a social choice function to be implementable, it must be incentive compatible with respect to some set of (possibly ambiguous) beliefs.
 - Furthermore, the messages sent and received at the communication stage must be such that the required posterior belief sets are indeed generated.

A simple example (from Bose and Renou)

There are two players, labeled 1 and 2, two (payoff-relevant) types θ and θ' for each player, and two alternatives x and y. Types are private information.

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- Assume that players have multiple-prior preferences (Gilboa and Schmeidler (1989)) and apply prior-by-prior updating (full Bayesian updating).
- Let P_i be the set of priors of player i about player j's types.

	u_i :	
	θ	θ'
Χ	0	1
У	1	0

Γ.	
θ	θ'
X	У
У	X

	u_i :		
	θ	θ'	
X	0	1	' θ
y	1	0	θ'

 $\begin{array}{c|cc}
\theta & \theta' \\
\hline
x & y \\
y & x
\end{array}$

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the payoff from lying.

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■ If $P_i = \Pi_i \equiv \{(0,1), (1,0)\}$ then f is implementable.

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- Define $\lambda((\omega_1, \omega_2) \mid (\theta_1, \theta_2)) = \lambda_1(\omega_1 \mid \theta_2) \lambda_2(\omega_2 \mid \theta_1)$.
- $\lambda_i(\omega \mid \theta) = 1$, and $\lambda_i(\omega' \mid \theta') = 1$, fully defines the first probability system λ .

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- Alternatively, if the probability system is λ' , player i's posterior belief is (0,1) if he receives the message ω , and (1,0) otherwise.
- Thus, regardless of the message received, player i's set of posteriors is $\Pi_i \equiv \{(0,1), (1,0)\}$, as required.

- Similarly, the second probability system λ' is fully specified by $\lambda_i \left(\omega \mid \theta'\right) = 1$ and $\lambda_i \left(\omega' \mid \theta\right) = 1$
- Clearly, if the probability system is λ , player *i*'s posterior belief is (1,0) if he receives the message ω , and (0,1) otherwise.
- Alternatively, if the probability system is λ' , player i's posterior belief is (0,1) if he receives the message ω , and (1,0) otherwise.
- Thus, regardless of the message received, player i's set of posteriors is $\Pi_i \equiv \{(0,1), (1,0)\}$, as required.
- Caveat: the mechanism exploits the fact that full Bayesian updating is dynamically inconsistent: effectively assumes that agents cannot commit to strategies, so it is not an implementation in (ex ante) Nash equilibrium.

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