

# A Biological Auction

## Valparaiso University Talk

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- A strange auction

# Outline

- A strange auction
- The biological connection

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- The biological connection
- A strange auction repeated

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- Best response to a known opponent

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- The biological connection
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- Best response to a known opponent
- **Biologically optimal strategy**

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- Best response to a known opponent
- Biologically optimal strategy
- **Concluding remarks**

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- **Biological interpretation.**

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- Keep track of the strategy you use and its effectiveness.
- Play now!
- What were the most effective strategies?

# Strange Auction Model I

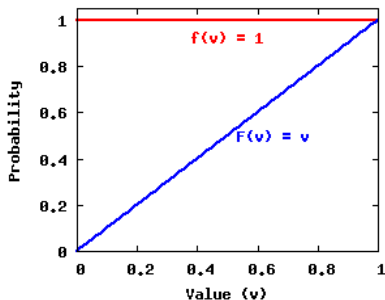
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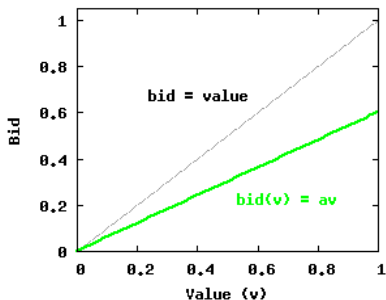
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- I want to choose  $b \geq 0$  to maximize  $\pi(b)$ .

# Payoff Maximization (General Case)

- Maximize the following at  $b = b^*$ :

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- Assume  $\beta$  is strictly increasing and  $F$  is the cdf of  $f$ .

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- Assume  $\beta$  is differentiable.

$$\pi'(b) = \frac{(v - \beta(\beta^{-1}(b)))f(\beta^{-1}(b))}{\beta'(\beta^{-1}(b))} - (1 - F(\beta^{-1}(b))) + \frac{bf(\beta^{-1}(b))}{\beta'(\beta^{-1}(b))}$$

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- Simplify.**

$$\pi'(b) = v f(\beta^{-1}(b)) / \beta'(\beta^{-1}(b)) - (1 - F(\beta^{-1}(b)))$$

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$$\pi'(b) = vf(\beta^{-1}(b))/\beta'(\beta^{-1}(b)) - (1 - F(\beta^{-1}(b)))$$

- First order necessary condition  $\pi'(b^*) = 0$ .

$$0 = vf(\beta^{-1}(b^*))/\beta'(\beta^{-1}(b^*)) - (1 - F(\beta^{-1}(b^*)))$$

# Payoff Maximization (Special Case)

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$$\pi'(b) = v/a - 1 + b/a$$

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- The correct maximum is a trigger strategy.

$$b^* = \begin{cases} 0, & \text{if } v \leq a/2 \\ a, & \text{if } v \geq a/2 \end{cases}$$

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$$\pi(b) = \int_{\beta(u) < b} (v - \beta(u)) f(u) du - b \int_{\beta(u) \geq b} f(u) du$$

- Assume  $\beta(v)$  is the player's payoff maximizing bid, that is,

$$\pi(\beta(v)) \geq \pi(b)$$

for all  $b \geq 0$ .

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- This function is differentiable and increasing from  $\beta(0) = 0$ .

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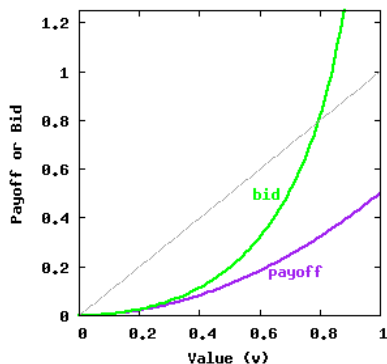
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- $\pi_{\max}(v) = \frac{1}{2}v^2$ .



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- which is positive if  $b < \beta(v)$
- and negative if  $b > \beta(v)$ .

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- The payoff to a player who values the prize at  $v$  and bids  $b$

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- The more you value the prize, the higher your expected payoff.

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- Recall the optimal bidding strategy.

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- Interchange integrals ( $0 \leq u \leq v < \infty$ ).

$$\int_0^\infty \beta(v)f(v) dv = \int_0^\infty \frac{uf(u)}{1 - F(u)} \int_u^\infty f(v) dv du$$

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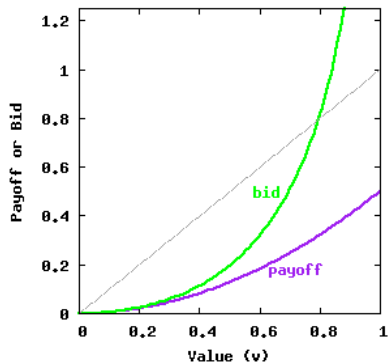
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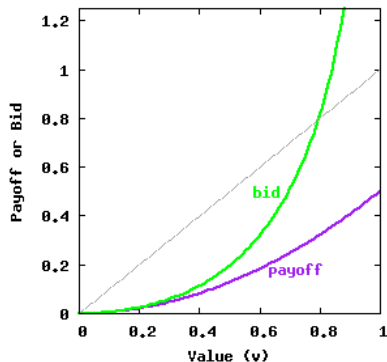
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- The average bid equals the average value.
- For some prize values  $v$ , the bid  $\beta(v)$  is greater than the value!**

# Concluding Remarks

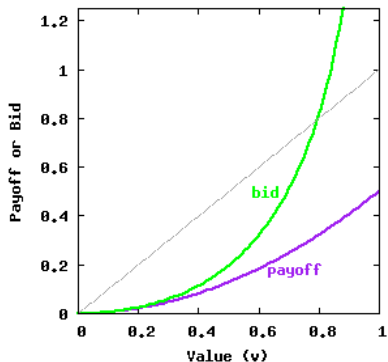


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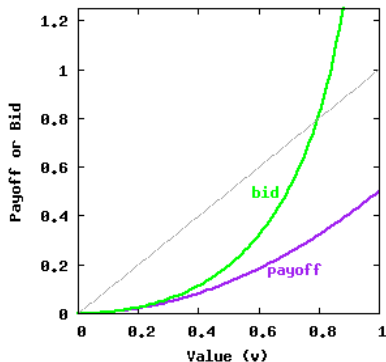
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