# MS&E 233 Game Theory, Data Science and Al Lecture 11

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(by courtesy) Computer Science and Electrical Engineering

Institute for Computational and Mathematical Engineering

#### **Computational Game Theory for Complex Games**

- Basics of game theory and zero-sum games (T)
- Basics of online learning theory (T)
- Solving zero-sum games via online learning (T)
- HW1: implement simple algorithms to solve zero-sum games
- Applications to ML and AI (T+A)
- HW2: implement boosting as solving a zero-sum game
- Basics of extensive-form games
- Solving extensive-form games via online learning (T)
- HW3: implement agents to solve very simple variants of poker
- General games, equilibria and online learning (T)
- Online learning in general games

(3)

 HW4: implement no-regret algorithms that converge to correlated equilibria in general games

#### **Data Science for Auctions and Mechanisms**

- Basics and applications of auction theory (T+A)
- Basic Auctions and Learning to bid in auctions (T)
- HW5: implement bandit algorithms to bid in ad auctions

- Optimal auctions and mechanisms (T)
- Simple vs optimal mechanisms (T)
  - HW6: calculate equilibria in simple auctions, implement simple and optimal auctions, analyze revenue empirically
  - Optimizing mechanisms from samples (T)
  - Online optimization of auctions and mechanisms (T)
- HW7: implement procedures to learn approximately optimal auctions from historical samples and in an online manner

#### **Further Topics**

- Econometrics in games and auctions (T+A)
- A/B testing in markets (T+A)
- HW8: implement procedure to estimate values from bids in an auction, empirically analyze inaccuracy of A/B tests in markets

#### **Guest Lectures**

- Mechanism Design for LLMs, Renato Paes Leme, Google Research
- Auto-bidding in Sponsored Search Auctions, Kshipra Bhawalkar, Google Research

# Sum: Vickrey-Clarke-Groves (VCG)

A universal welfare maximizing auction/mechanism!

For any mechanism design setting, it guarantees that:

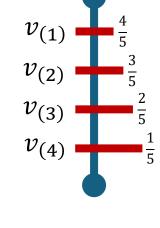
- 1. It is dominant strategy truthful
- 2. It always chooses the welfare maximizing outcome/allocation
- 3. All bidders have non-negative utility
- 4. All payments are non-negative

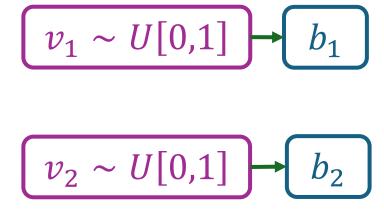
For special case of single-item auction = Second-Price Auction

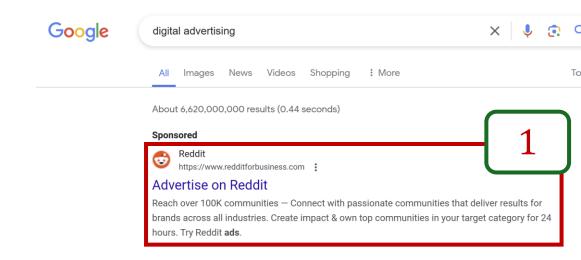
# What if we want to maximize revenue?

# Let's go back to basics: Single-Item Auction

- How much revenue does the second-price auction achieve?  $\operatorname{Rev} = E\big[v_{(2)}\big] = E\big[\min(v_1,v_2)\big] = 1/3$
- Can we do better?





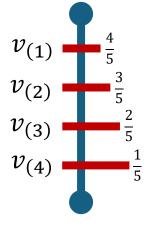


# Let's go back to basics: Single-Item Auction

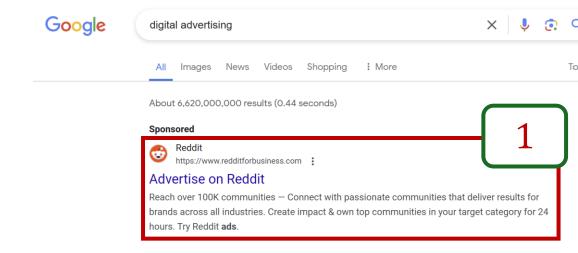
What if we only had one bidder?

$$Rev = E[v_{(2)}] = 0$$

Can we do better?





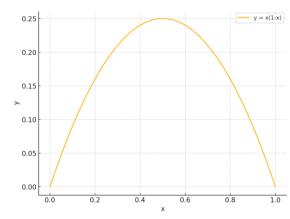


What if we post a reserve price?

# Let's go back to basics: Single-Item Auction

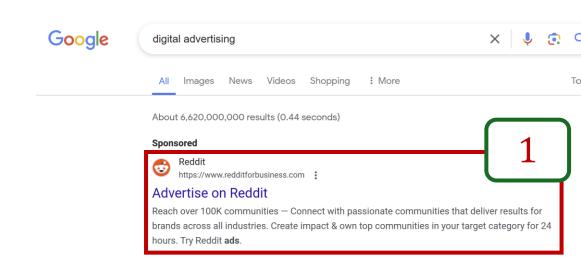
• Auctioneer: "If you bid less than r, I will not accept your bid and not show any ad on the page! If you win you must pay r."

$$Rev(r) = E[r \ 1(v \ge r)] = r \ (1 - r) \Rightarrow Rev(1/2) = 1/4$$



- Is the auction truthful?
- Is the auction efficient?

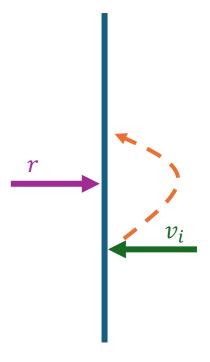




### Truthfulness of Mechanism

Suppose I bid my value. Would I want to deviate?

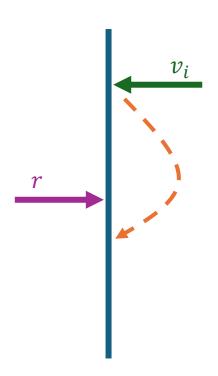
- Case 1. My value is below reserve price
- Only way to change anything is bid above
- But then I get negative utility as I pay more than value



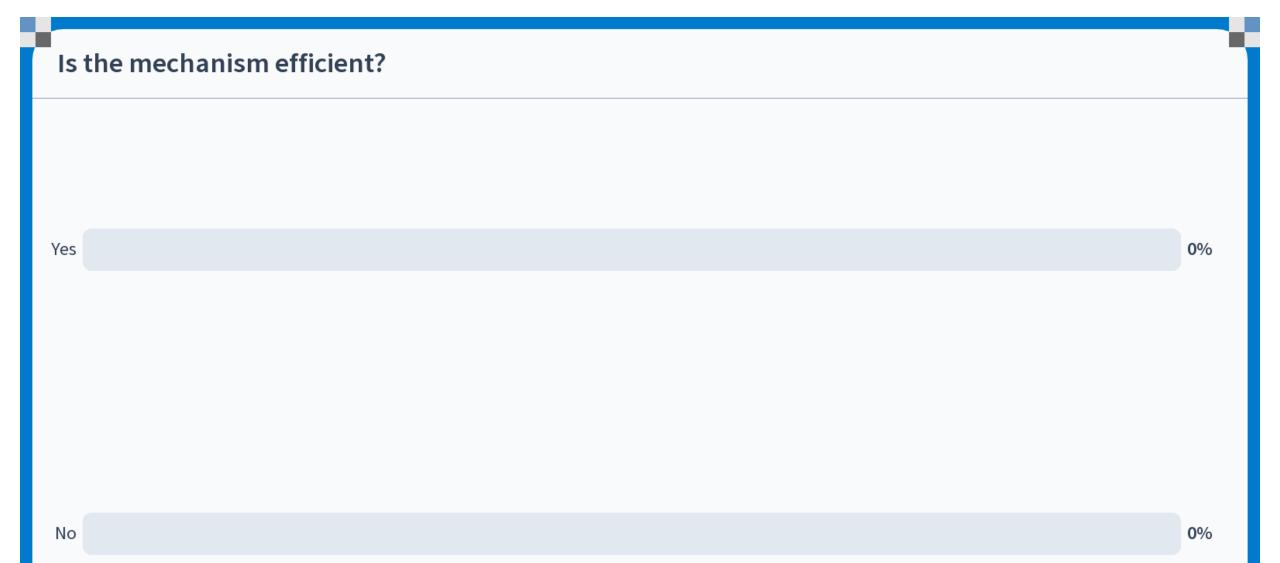
### Truthfulness of Mechanism

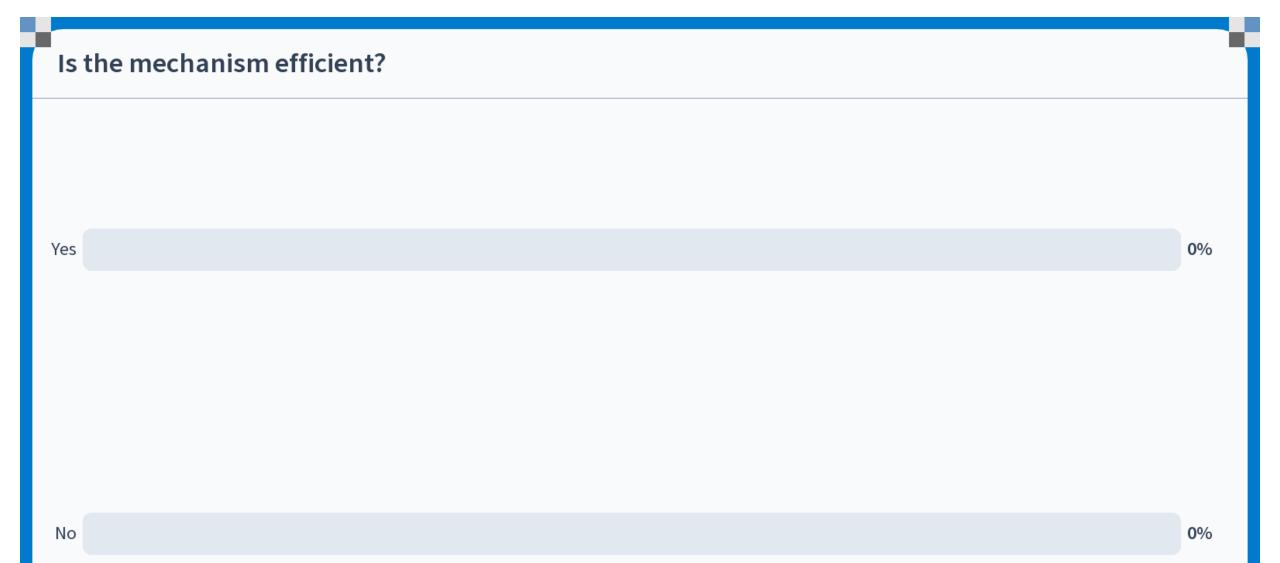
Suppose I bid my value. Would I want to deviate?

- Case 2. My value is above reserve price
- I get non-negative utility
- Only way to change anything is bid below
- But then I get zero utility as I lose



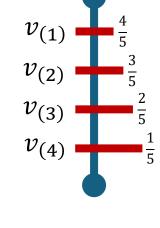


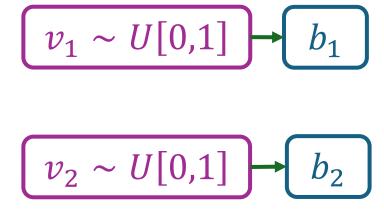


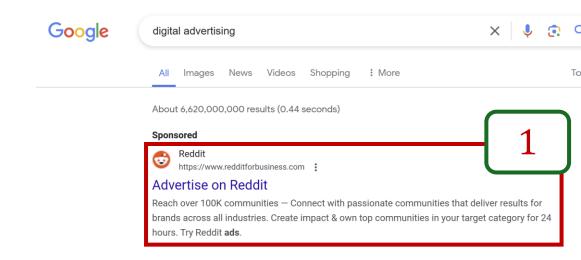


# Let's go back to basics: Single-Item Auction

- How much revenue does the second-price auction achieve?  $\operatorname{Rev} = E\big[v_{(2)}\big] = E\big[\min(v_1,v_2)\big] = 1/3$
- Can we do better?





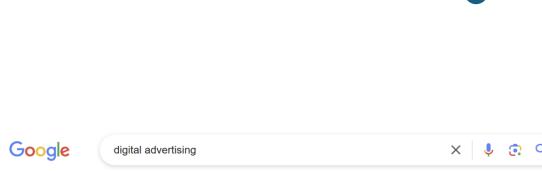


# Let's go back to basics: Single-Item Auction

• Auctioneer: "If you bid less than r, I will not accept your bid! If you win you must pay  $\max(\mathbf{b}_2, r)$ ."

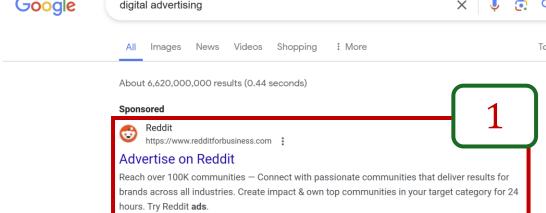
$$Rev(1/2) = E[\max(v_{(2)}, r) 1(v_{(1)} \ge r)] = 5/12$$

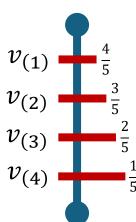
Can we do better?



$$v_1 \sim U[0,1] \longrightarrow b_1$$

$$v_2 \sim U[0,1] \longrightarrow b_2$$





# How do we optimize over all possible mechanisms!

# Single-Parameter Settings

- ullet Each bidder has some value  $v_i$  for being allocated
- Bidders submit a reported value  $b_i$  (without loss of generality)
- Mechanism decides on an allocation  $x \in X \subseteq \{0,1\}^n$
- Mechanism fixes a probabilistic allocation rule:

$$x(b) \in \Delta(X)$$

- First question. Given an allocation rule, when can we find a payment rule p so that the overall mechanism is truthful?
- If we can find such a payment, we will say that x is implementable

### Some Shorthand Notation

- Let's fix bidder i and what other bidders bid  $b_{-i}$
- For simplicity of notation, we drop index i and  $b_{-i}$

What properties does the function

$$x(v) \equiv x_i(v, b_{-i})$$

need to satisfy, so that x is implementable?

Can we find a truthful payment function

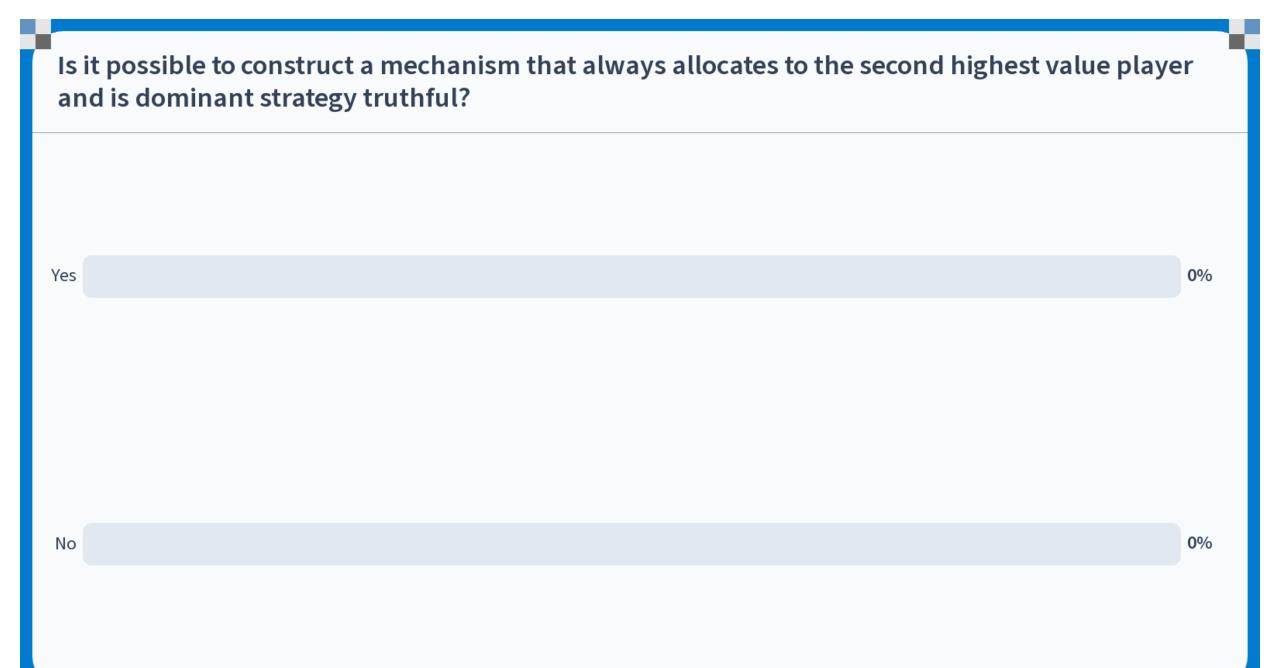
$$p(v) \equiv p(v, b_{-i})$$

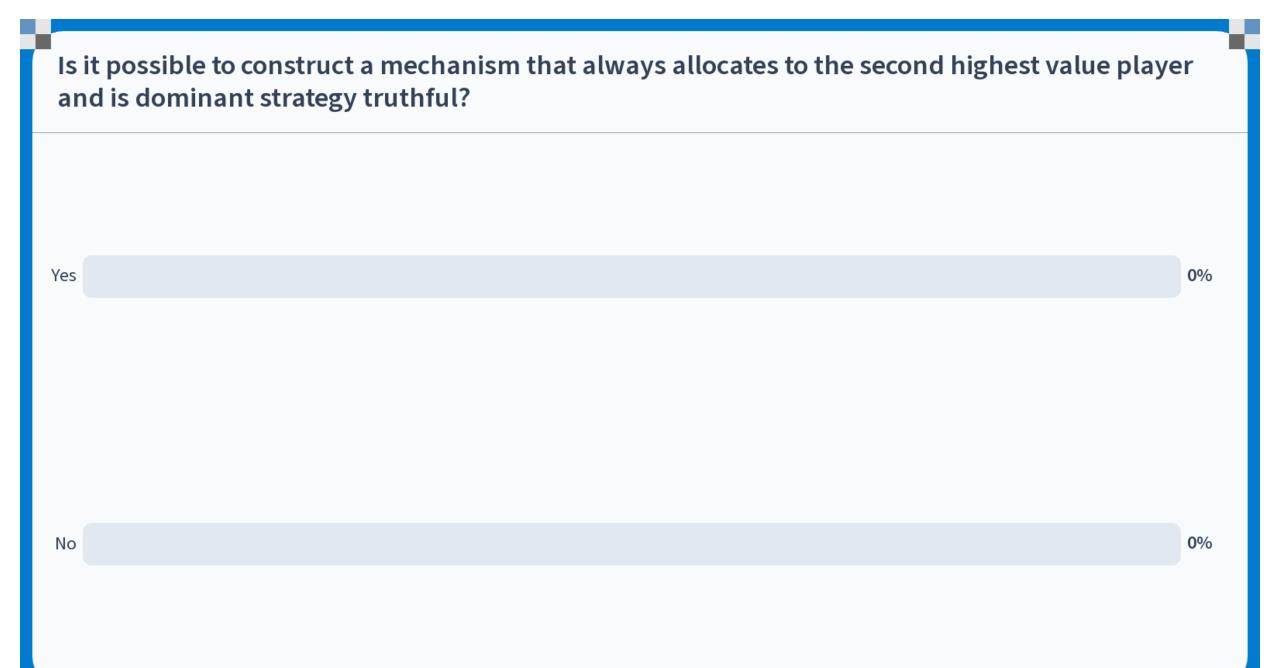
Is it possible to construct a mechanism that always allocates to the second highest value player and is dominant strategy truthful?

Is it possible to construct a mechanism that always allocates to the second highest value player and is dominant strategy truthful?

Yes

No





# Suppose it is possible

Suppose that we both bid truthfully

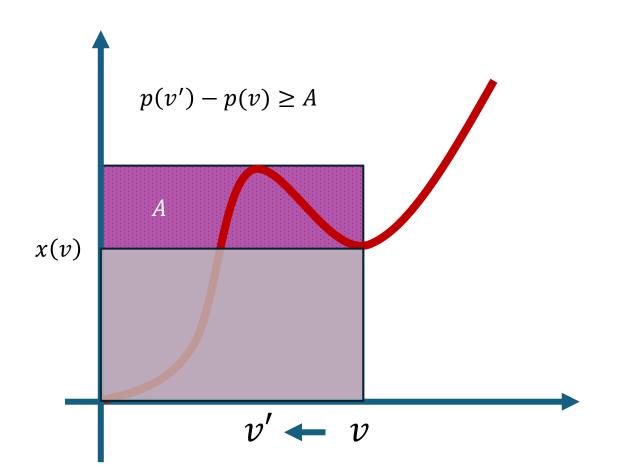
Suppose that I am the highest value bidder

• No matter what the payment rule is, I can always reduce my bid to the second highest bid minus  $\epsilon$ 

 By doing so, I am paying at most the second highest bid and I am winning deterministically

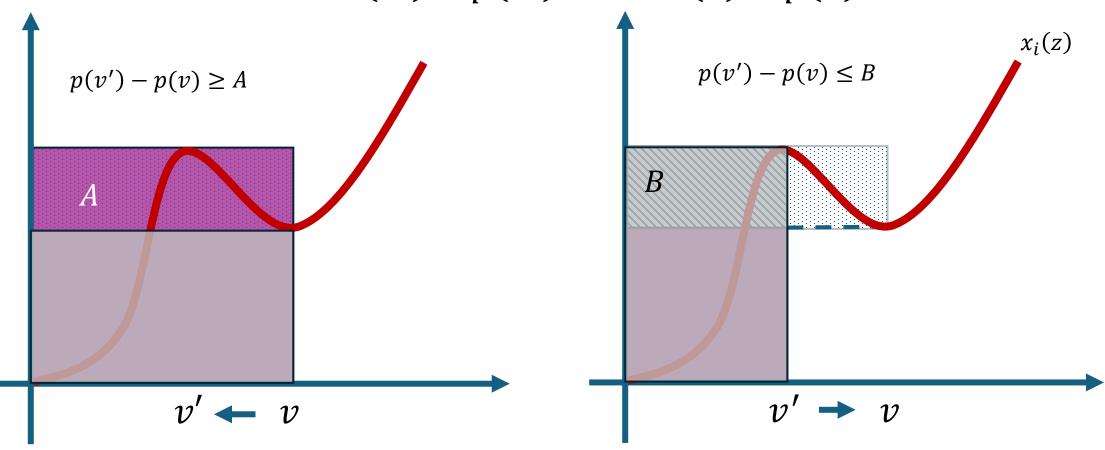
## Implementable Rules are Monotone

$$v \cdot x(v) - p(v) \ge v \cdot x(v') - p(v')$$

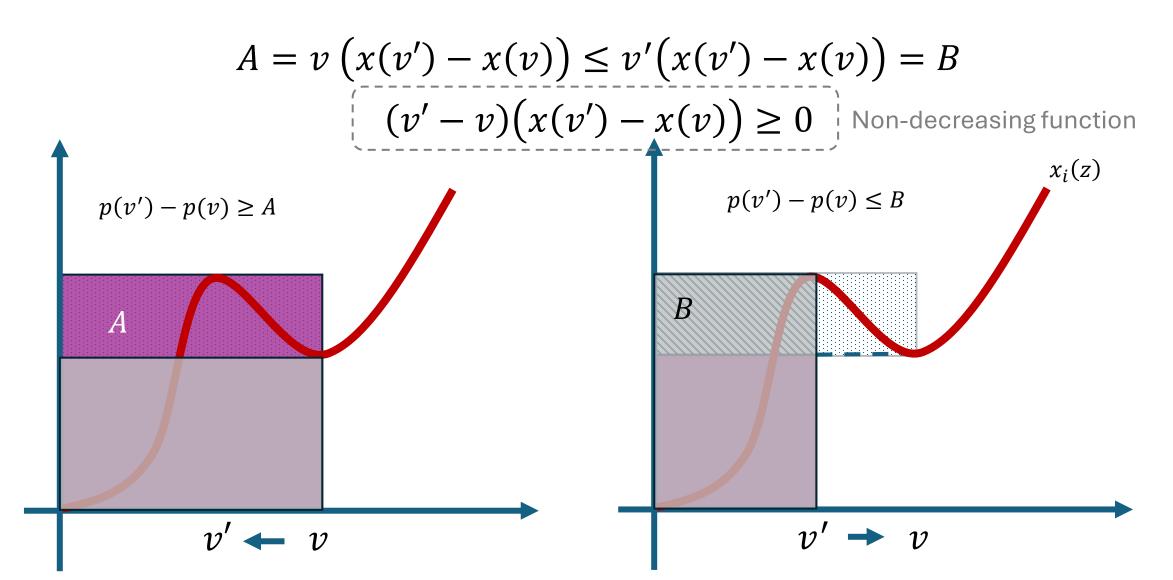


## Implementable Rules are Monotone

$$v \cdot x(v) - p(v) \ge v \cdot x(v') - p(v')$$
$$v' \cdot x(v') - p(v') \ge v' \cdot x(v) - p(v)$$



### Implementable Rules are Monotone



Any implementable allocation rule must be monotone!

"If not allocated with value v, I should not be allocated if I report a lower value!"

# Uniqueness of Payment Rule

• I should not want to deviate locally up or down infinitesimally

$$u(v) \ge v \cdot x(v + \epsilon) - p(v + \epsilon) = u(v + \epsilon) - \epsilon \cdot x(v + \epsilon)$$
  
$$u(v) \ge v \cdot x(v - \epsilon) - p(v - \epsilon) = u(v - \epsilon) + \epsilon \cdot x(v - \epsilon)$$

• Dividing over by  $\epsilon$ , restricts the rate of change of utility

$$\frac{u(v+\epsilon) - u(v)}{\varepsilon} \le x(v+\epsilon)$$

$$\frac{u(v) - \epsilon}{\varepsilon} \ge x(v-\epsilon)$$

• If u was differentiable, then taking the limit of the above as  $\epsilon \to 0$ 

$$x(v) \le u'(v) \le x(v) \Rightarrow u'(v) = x(v) \Rightarrow u(v) - u(0) = \int_0^v x(z) dz$$

Under any truthful payment rule 
$$u(v) = u(0) + \int_{0}^{v} x(z) dz$$

# Discontinuity of Allocation Rule

 Even though allocation rule can be discontinuous, because it is monotone, it is Riemann integrable

$$\int_0^v x(z) dz = \lim_{\epsilon \to 0} \sum_k x(z + \epsilon) \cdot \epsilon \ge \lim_{\epsilon \to 0} \sum_k u(z + \epsilon) - u(z) = u(v) - u(0)$$

$$\int_0^v x(z) dz = \lim_{\epsilon \to 0} \sum_k x(z - \epsilon) \cdot \epsilon \le \lim_{\epsilon \to 0} \sum_k u(z) - u(z - \epsilon) = u(v) - u(0)$$

Under any truthful payment rule 
$$u(v) = u(0) + \int_{0}^{v} x(z) dz$$

# What does that imply about payments

Since utility is value minus payment

$$v x(v) - p(v) = -p(0) + \int_0^v x(z) dz$$

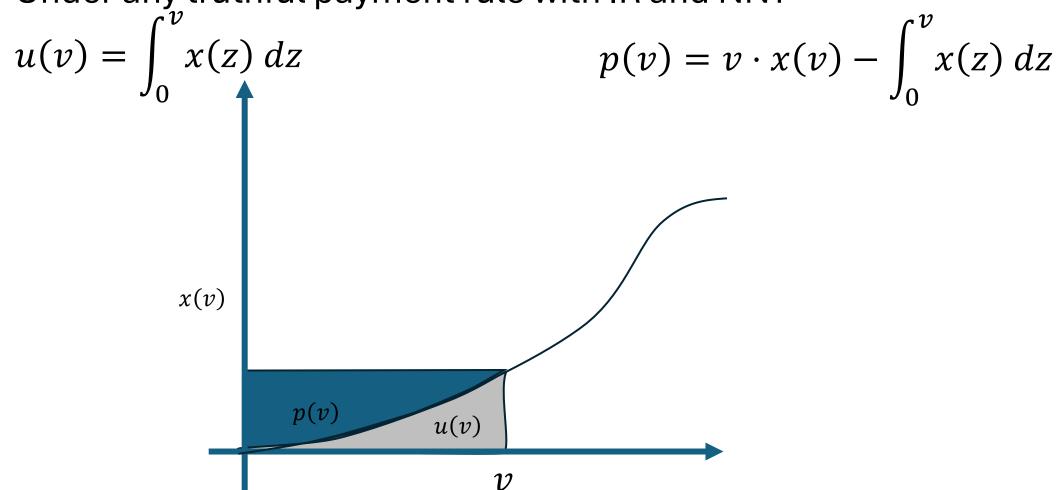
- Non-Negative-Transfers (NNT). We never have negative payments  $p(0) \geq 0$
- Individually Rational (IR). We never give bidders negative utility  $p(0) \leq 0$
- Thus, payment at 0 should be zero!

# Under any truthful payment rule that satisfies NNT and IR

$$p(v) = v \cdot x(v) - \int_0^v x(z) dz$$

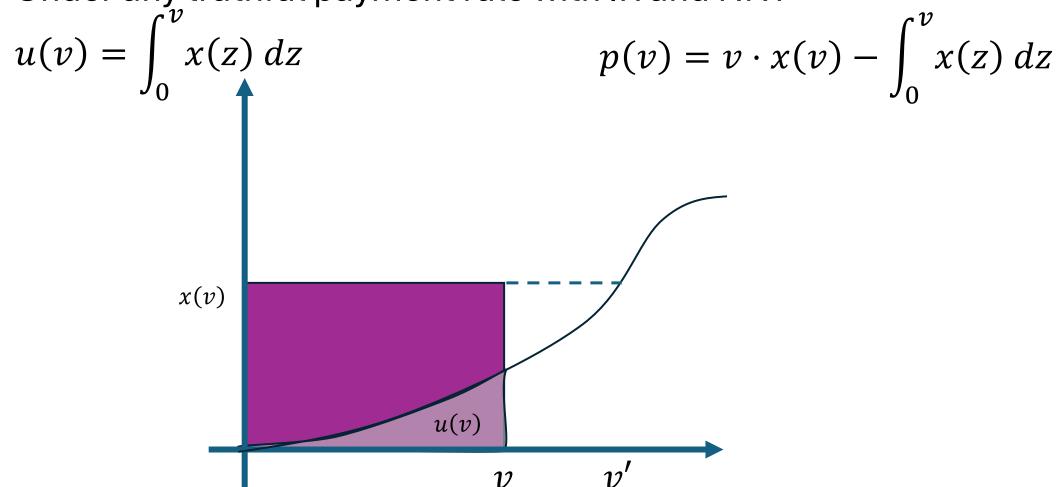
# Visualizing Utility

Under any truthful payment rule with IR and NNT



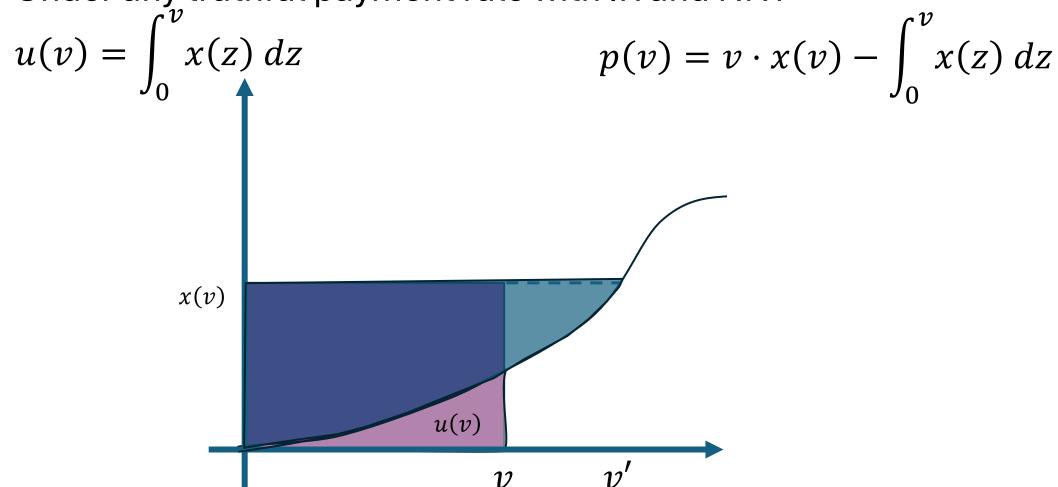
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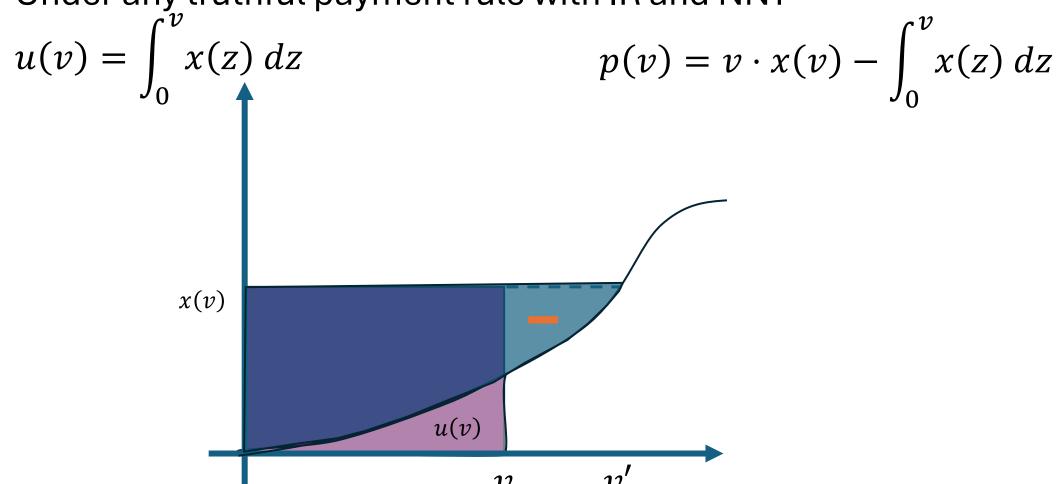
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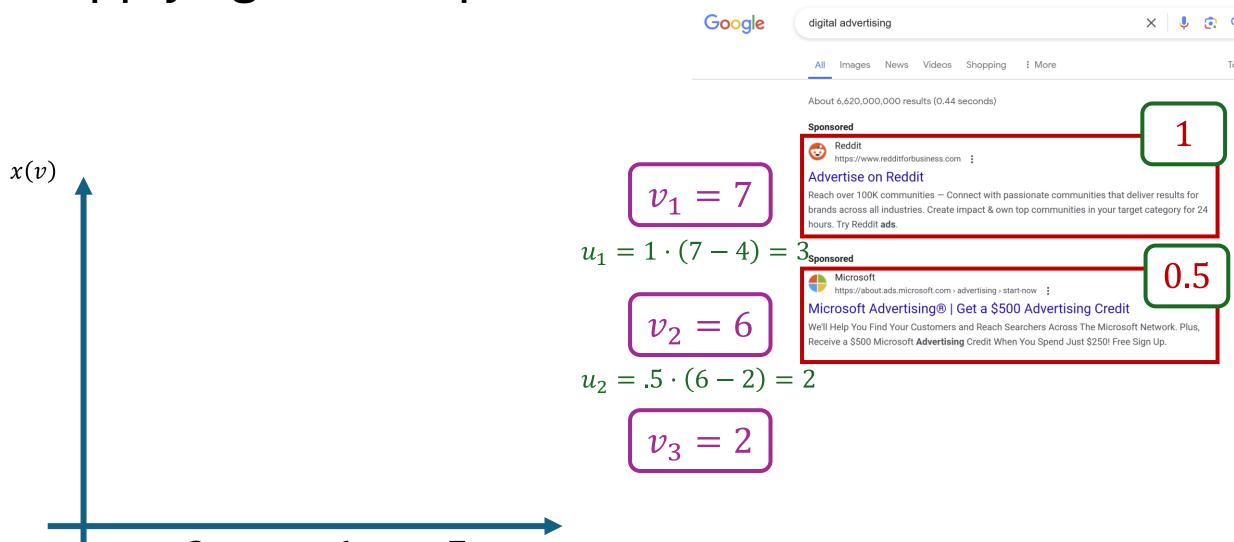
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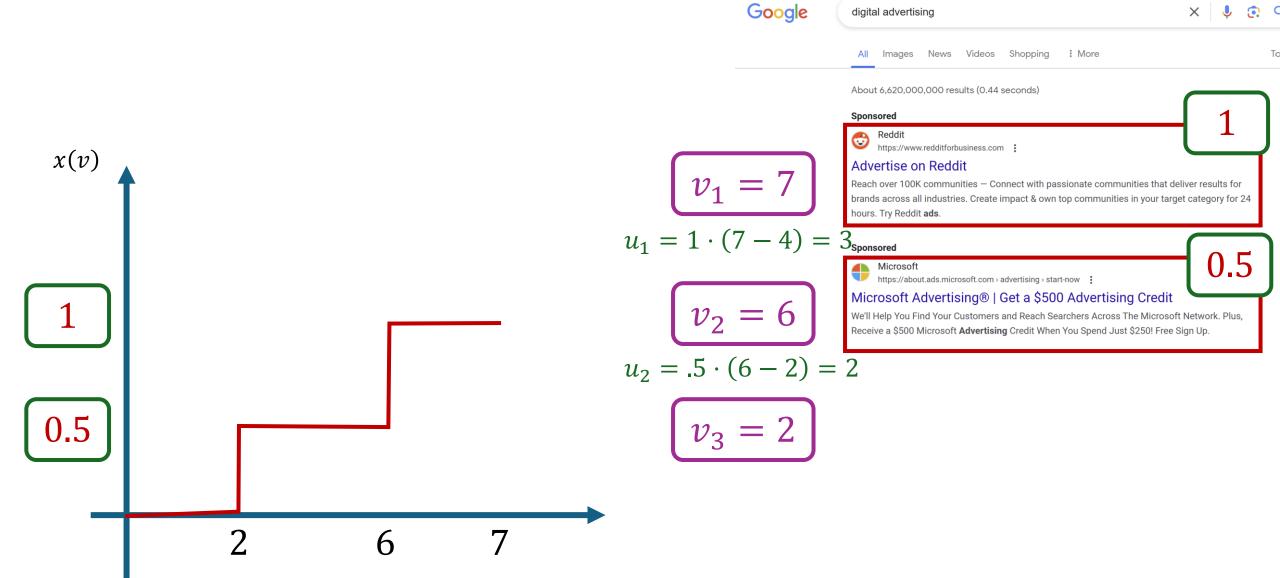


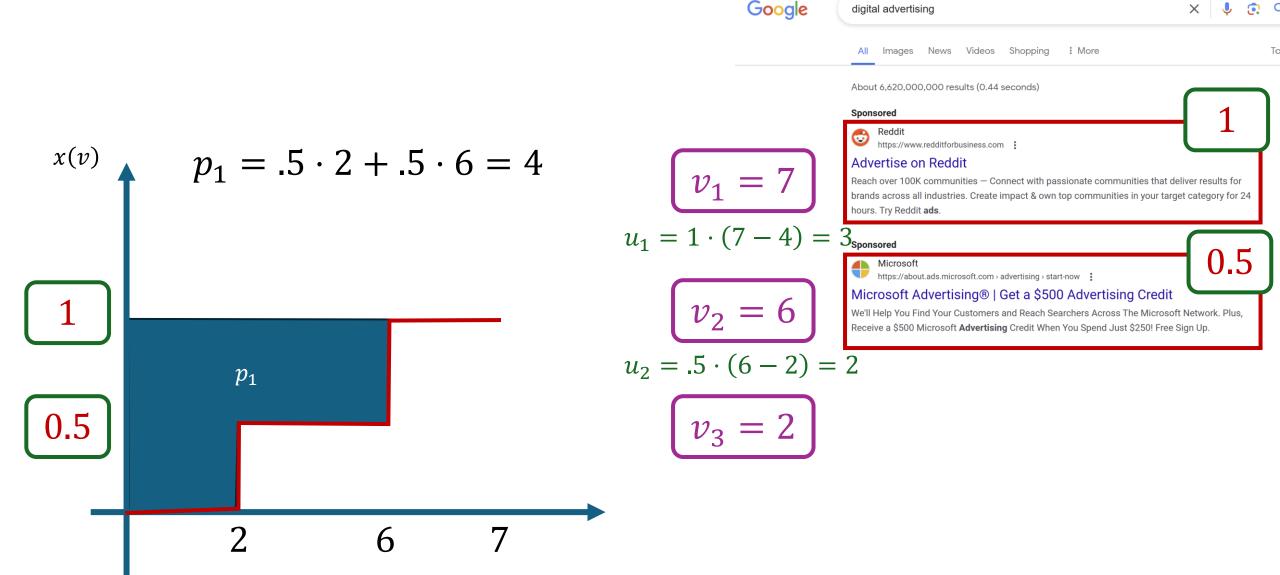
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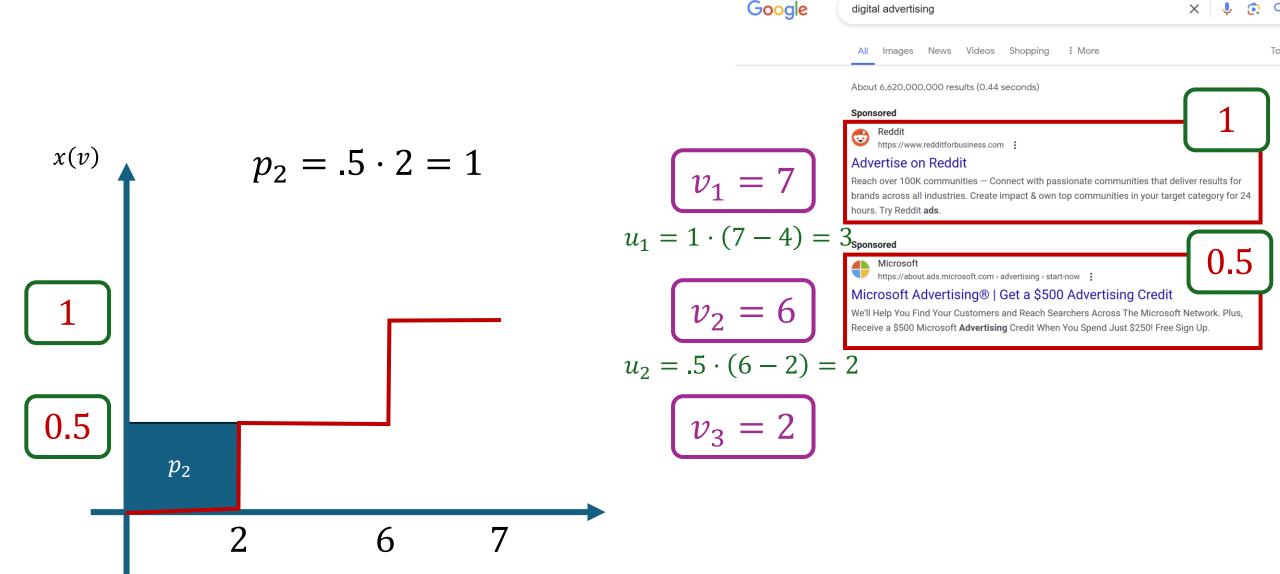
Under any truthful payment rule with IR and NNT

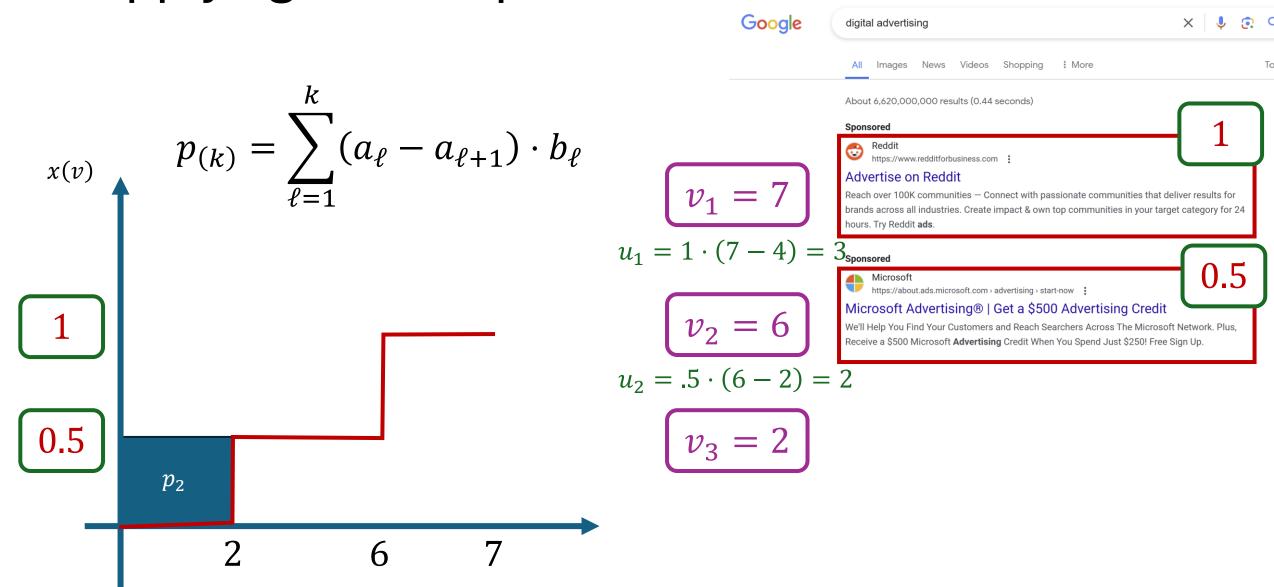












# Given an allocation rule, the payment is uniquely determined!

# Optimizing over allocation rules

# Myerson's Theorem

- Let x, p be any DSIC mechanism
- Suppose each value  $v_i \sim F_i$  independently and let  $v=(v_1,\dots,v_n)$   $E[p_i(v)]=E[v_i\phi_i(v_i)]$

where  $\phi_i(v_i)$  is bidder i's "virtual value".

• Letting  $F_i$  the CDF and  $f_i$  the density:

$$\phi_i(v_i) = v_i - \frac{1 - F_i(v_i)}{f_i(v_i)}$$

• Assuming  $\phi_i(v_i)$  is monotone non-decreasing, then the optimal DSIC mechanism is the mechanism that allocates to the highest virtual value bidder (or none if highest virtual value is negative)

# Back to Uniform Example

- If  $v_i \sim U[0,1]$  then F(v) = v and f(v) = 1
- Virtual value simplifies to

$$\phi_i(v_i) = v_i - (1 - v_i) = 2v_i - 1$$

 We should allocate to the highest virtual value player, as long as the highest virtual value is non-negative

$$v_i \ge 1/2$$

- Since all virtual value functions are the same, allocating to the highest virtual value is the same as allocating to the highest value
- Simply: Second Price with a reserve price of 1/2!

# Myerson's Theorem

• Consider the revenue contribution of a single bidder i and drop other bids and index from notation

$$E[p(v)] = E\left[v x(v) - \int_0^v x(z)dz\right] = E\left[v \hat{x}(v) - \int_0^v \hat{x}(z)dz\right]$$

- Allocation  $\hat{x}(z)$  is the expected allocation over other bidder values  $\hat{x}(z) = E_{v_{-i}}[x(z,v_{-i})]$
- We can do an exchange of the integrals:

$$E\left[\int_0^v \hat{x}(z) dz\right] = \int_{v=0}^\infty \int_{z=0}^v \hat{x}(z) dz f(v) dv = \int_{z=0}^\infty \hat{x}(z) \int_{v=z}^\infty f(v) dv dz$$
$$= \int_{z=0}^\infty \hat{x}(z) \left(1 - F(z)\right) dz = E\left[\hat{x}(v) \frac{1 - F(v)}{f(v)}\right]$$

### **Appendix:** Deriving the Optimal Reserve

• Bidders are symmetric. Revenue is twice the revenue we collect from each bidder

$$\begin{aligned} \operatorname{Rev}_{1}(r) &= E[\max(v_{2}, r) \ 1(v_{1} \geq \max(v_{2}, r))] \\ &= E[v_{2} \mid v_{2} \in [r, v_{1}]] \Pr(v_{2} \in [r, v_{1}] | v_{1} \geq r) \Pr(v_{1} \geq r) + r \Pr(v_{2} \leq r) \Pr(v_{1} \geq r) \\ &= \int_{r}^{1} \frac{v + r}{2} (v - r) dv + r^{2} (1 - r) \\ &= \int_{r}^{1} \frac{v^{2} - r^{2}}{2} dv + r^{2} (1 - r) \\ &= \left(\frac{1 - r^{3}}{6} - \frac{r^{2}}{2} (1 - r) + r^{2} (1 - r)\right) \\ &= \frac{1 - r^{3}}{6} + \frac{r^{2} (1 - r)}{2} = \frac{1 - r^{3} + 3r^{2} - 3r^{3}}{6} = \frac{1 + 3r^{2} - 4r^{3}}{6} \end{aligned}$$

The first order condition

$$(\text{Rev}_1(r))' = r(1 - 2r) = 0 \Rightarrow r = 1/2$$