

# Recap/Big Picture

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DS 574 LECTURE 10

# Econ→CS

## Input:

Data reported by  
**strategic agents**.



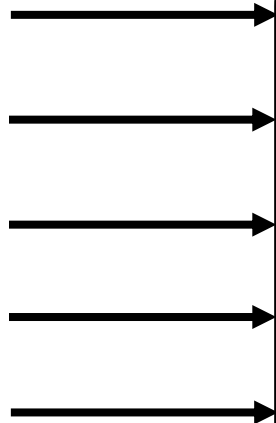
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15



10



**Objective:** Maximize  
buyer's value

**Mechanism**



## Output:

-who gets what  
-who pays (gets  
paid) what



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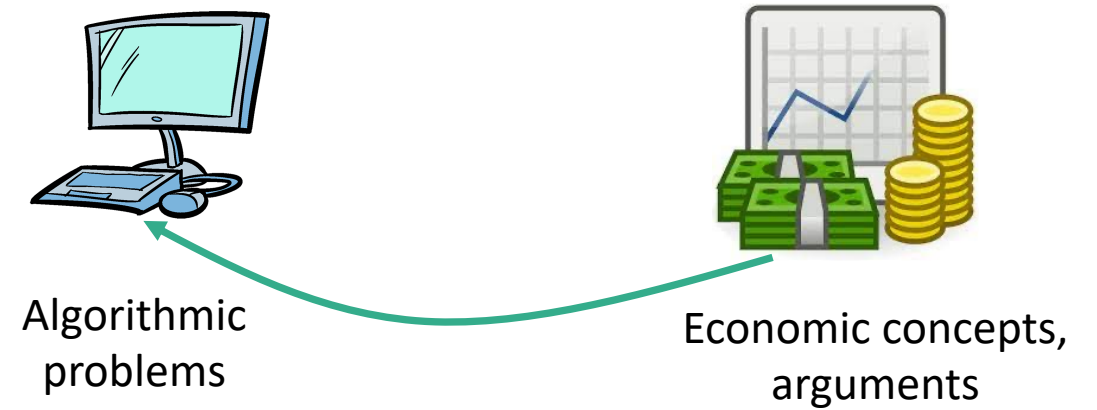
Use **game theory** to reason about  
**incentives** within the **algorithm**  
so that we can **guarantee**  
(approximate) optimality.



Algorithmic  
problems



Economic concepts,  
arguments



# Maximize Social Welfare: 2<sup>nd</sup> Price

**Objective:** Maximize value of the allocation

**Input:** Strategic bids

$b_1$

$b_2$

$b_3$

$b_4$

$b_5$

**Mechanism**

**Output:**

- allocation

highest bidder  $x_1 = 1$

- payment

2<sup>nd</sup> highest bid  $b^2$

**2<sup>nd</sup> Price (Vickrey) Auction is DSIC:**

maxes  $i$ 's utility to have  $b_i = v_i$  independent of all  $b_{-i}$

$b_i > v_i$ : if  $b^2$  is in between,  $i$  wins and overpays

$b_i < v_i$ : if  $b^2$  is in between,  $i$  loses and gets 0 util instead of positive



value  $v_i$

utility  $v_i x_i(\mathbf{b}) - p_i(\mathbf{b})$

bid  $b_i$

$$\Rightarrow b_i = v_i \forall i$$

# Dominant Strategy Incentive Compatibility

More utility for bidding actual value:

$$\underline{v_i x_i(v_i, \mathbf{b}_{-i}) - p_i(v_i, \mathbf{b}_{-i})} \geq v_i x_i(b_i, \mathbf{b}_{-i}) - p_i(b_i, \mathbf{b}_{-i}) \quad \forall i, v_i, b_i, \mathbf{b}_{-i}$$

1) The allocation rule must be **monotone**, or this can't hold.

implementable

Myerson's  
Lemma

2) DSIC payments are completely determined by the allocation rule:

$$p_i(v_i, \mathbf{b}_{-i}) = \int_0^{v_i} z x'_i(z, \mathbf{b}_{-i}) dz = \underline{v_i x_i(v_i, \mathbf{b}_{-i})} - \int_0^{v_i} x_i(z, \mathbf{b}_{-i}) dz$$

Payment  
Identity

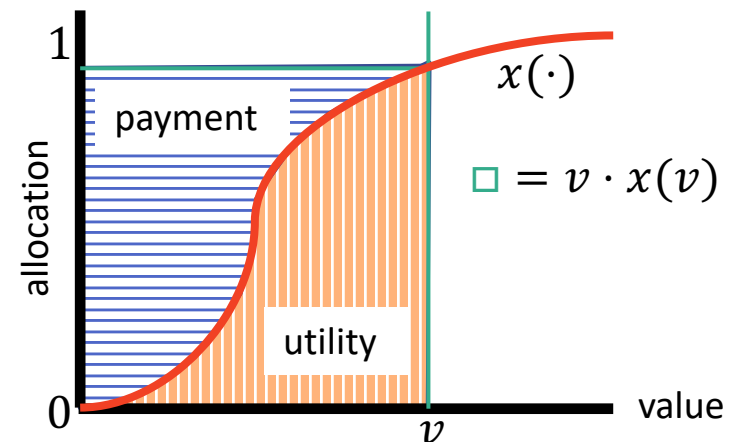


value  $v_i$

utility  $v_i x_i(\mathbf{b}) - p_i(\mathbf{b})$

bid  $b_i$

$$\Rightarrow b_i = v_i \quad \forall i$$



# Maximize Social Welfare: 1<sup>st</sup> Price

**Objective:** Maximize value of the allocation

**Input:** Strategic bids

$b_1$

$b_2$

$b_3$

$b_4$

$b_5$

**Mechanism**

**Output:**

- allocation
- payment

highest bidder  
own bid

$x_1 = 1$   
 $b^1$

**1<sup>st</sup> Price Auction is not DSIC:**

$b_i = v_i$  means utility is 0,  
better have  $b_i < v_i$



value  $v_i$

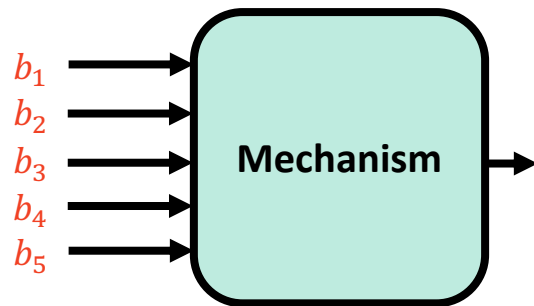
utility  $v_i x_i(\mathbf{b}) - p_i(\mathbf{b})$

bid  $b_i$

# The Bayesian Setting: Stages

Each bidder  $i$ 's value  $v_i$  is drawn from a distribution with CDF  $F_i$  and pdf  $f_i$

- $F_1, \dots, F_n$  are common knowledge to all bidders and the auctioneer
- $F_i(x) = \Pr[v_i \leq x]$
- $f_i(x) = \frac{d}{dx} F_i(x)$



**ex ante:** no values are known. mechanism announced.

**interim:**  $i$  knows  $v_i$ , Bayesian updates given this  
bidders submit bids

**ex post:** outcome announced. know  $v_1, \dots, v_n$



value  $v_i$

utility  $v_i x_i(\mathbf{b}) - p_i(\mathbf{b})$

bid  $b_i$

**needed:**

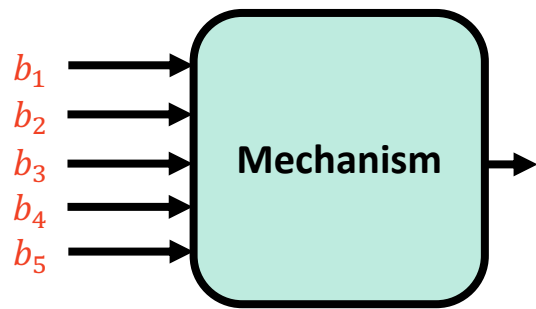
- for bidders to reason about other bidders' behavior (BNE)
- for auctioneer to reason about objective in expectation

# The Bayesian Setting: Incentive Compatibility

Each bidder  $i$ 's value  $v_i$  is drawn from a known distribution  $F_i$

**BIC:** 
$$\mathbb{E}_{v_{-i}}[v_i x_i(v_i, v_{-i}) - p_i(v_i, v_{-i})] \geq \mathbb{E}_{v_{-i}}[v_i x_i(b_i, v_{-i}) - p_i(b_i, v_{-i})] \quad \forall i, v_i, b_i$$

NOT  $\forall \mathbf{b}_{-i}$  but in  $\mathbb{E}_{v_{-i}}$ !



**interim:**  $i$  knows  $v_i$ , Bayesian updates given this bidders submit bids

$$v_i \hat{x}_i(v_i) - \hat{p}_i(v_i) \geq v_i \hat{x}_i(b_i) - \hat{p}_i(b_i) \quad \forall i, v_i, b_i$$

$$\hat{x}_i(b_i) = \mathbb{E}_{v_{-i}}[x_i(b_i, v_{-i})]$$

$$\hat{p}_i(b_i) = \mathbb{E}_{v_{-i}}[p_i(b_i, v_{-i})]$$



value  $v_i$

utility  $v_i x_i(\mathbf{b}) - p_i(\mathbf{b})$

bid  $b_i$

**ex post:** outcome announced. know  $v_1, \dots, v_n$

$$x_i(b_i, \mathbf{b}_{-i})$$

$$p_i(b_i, \mathbf{b}_{-i})$$

**DSIC:** 
$$v_i x_i(v_i, \mathbf{b}_{-i}) - p_i(v_i, \mathbf{b}_{-i}) \geq v_i x_i(b_i, \mathbf{b}_{-i}) - p_i(b_i, \mathbf{b}_{-i}) \quad \forall i, v_i, b_i, \mathbf{b}_{-i}$$

# Nash Equilibrium vs. Incentive-Compatibility

A mechanism is [concept] Incentive-Compatible if in the mechanism, truthful reporting is a [concept] Nash Equilibrium. (i.e. [concept] \in Dominant Strategy, Bayes-Nash, Ex Post\*)

\*sincere bidding may be required instead of truthful

**BNE:** Best-response strategies  $\sigma$  form a Bayes-Nash Equilibrium (BNE) in  $(x, p)$  when

$$\mathbb{E}_{v_{-i}}[v_i x_i(\sigma_i(v_i), \sigma_{-i}(v_{-i})) - p_i(\sigma_i(v_i), \sigma_{-i}(v_{-i}))] \geq \mathbb{E}_{v_{-i}}[v_i x_i(b_i, \sigma_{-i}(v_{-i})) - p_i(b_i, \sigma_{-i}(v_{-i}))] \quad \forall i, v_i, b_i$$

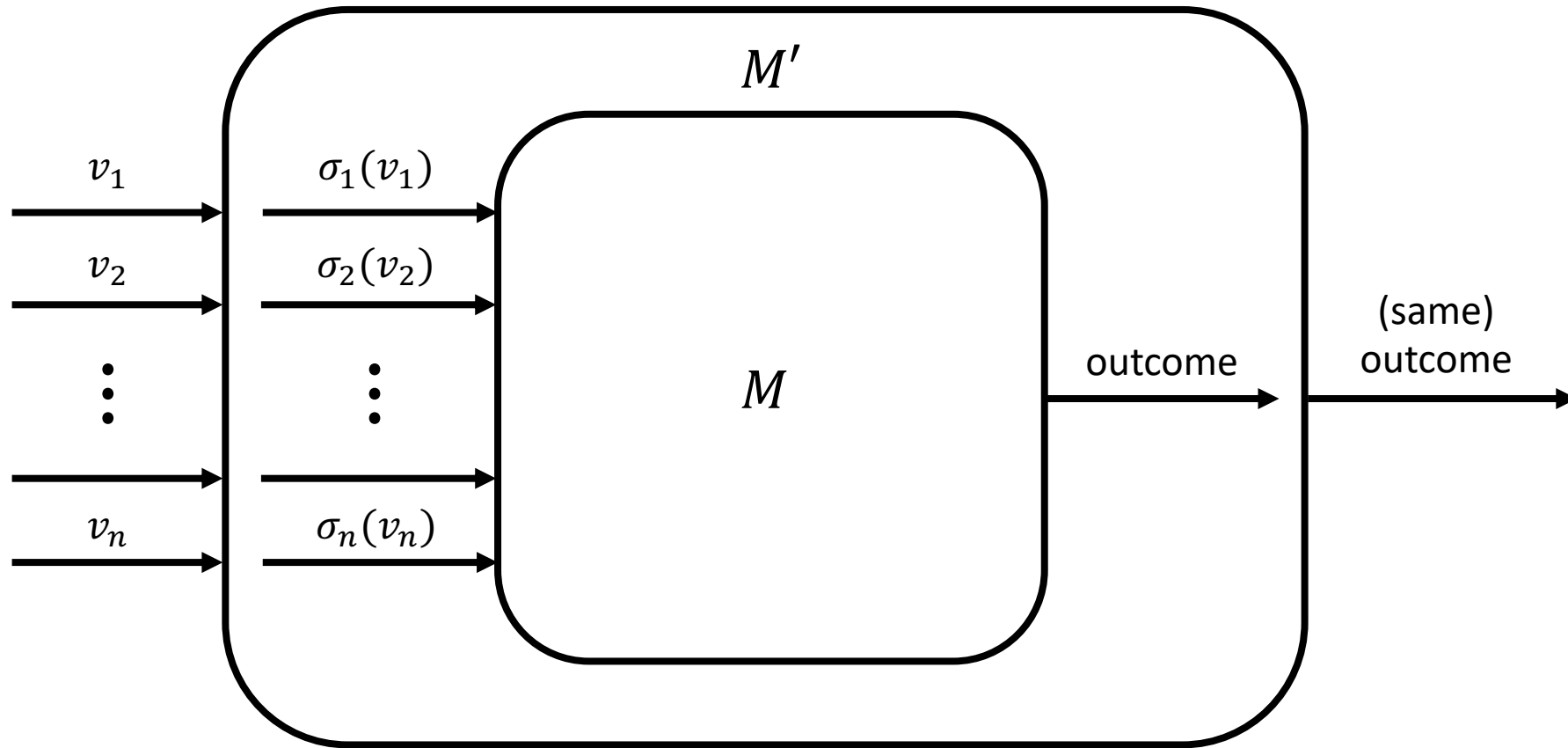
**BIC:** A mechanism  $(x, p)$  is Bayesian Incentive-Compatible (BIC) when

$$\mathbb{E}_{v_{-i}}[v_i x_i(v_i, v_{-i}) - p_i(v_i, v_{-i})] \geq \mathbb{E}_{v_{-i}}[v_i x_i(b_i, v_{-i}) - p_i(b_i, v_{-i})] \quad \forall i, v_i, b_i$$



# Revelation Principle + Revenue Equivalence

Revelation Principle: It is without loss to focus on [DS/B/EP]IC mechanisms.



Revenue Equivalence: Mechs w/ the same outcome have the same  $\mathbb{E}[\text{Rev}]$ .

# Maximizing Revenue

How can we max revenue? Can't just charge  $v_i$  – not IC. Still need the payment identity.

Expected Revenue  $= \mathbb{E}_v \left[ \sum_i p_i(\mathbf{v}) \right] = \mathbb{E}_v \left[ \sum_i x_i(\mathbf{v}) \varphi_i(v_i) \right] =$  Expected Virtual Welfare

Only DSIC if  $\varphi_i(v_i)$  is monotone

plug in the payment identity

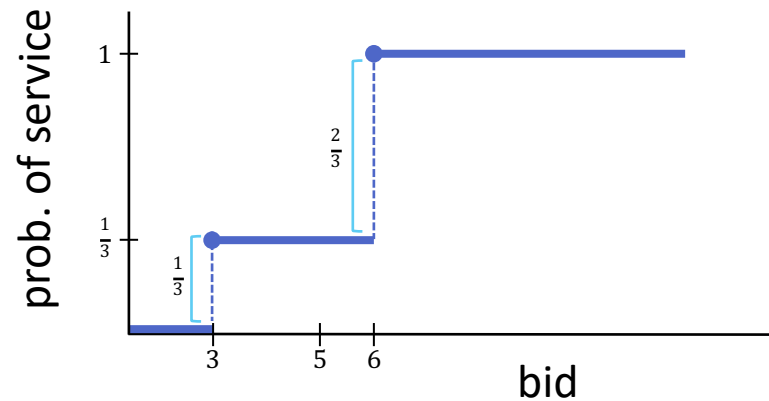
To max rev, choose  $x$  to maximize this

For virtual  
value functions

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

# How else can we express revenue?

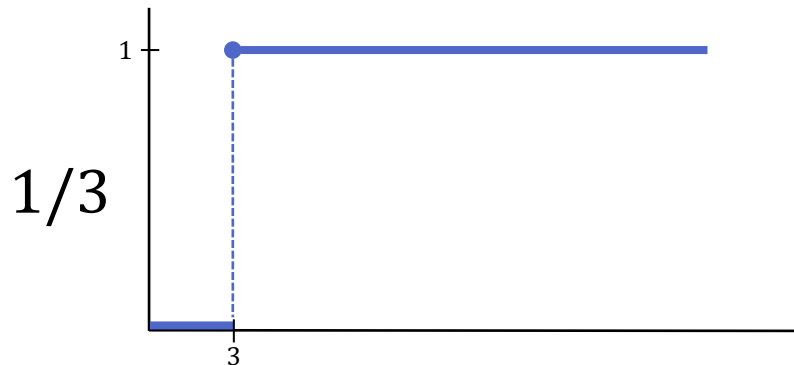
Any allocation rule can be expressed as a distribution of prices.



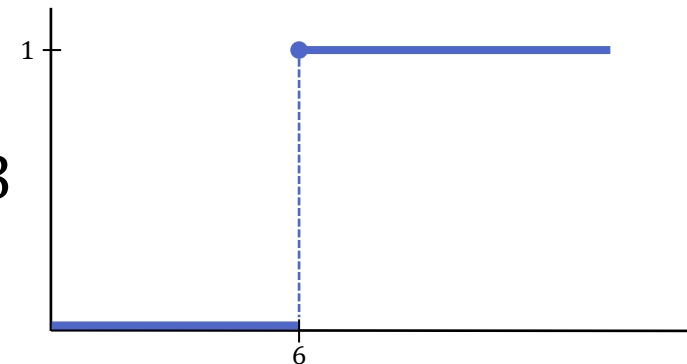
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Menu

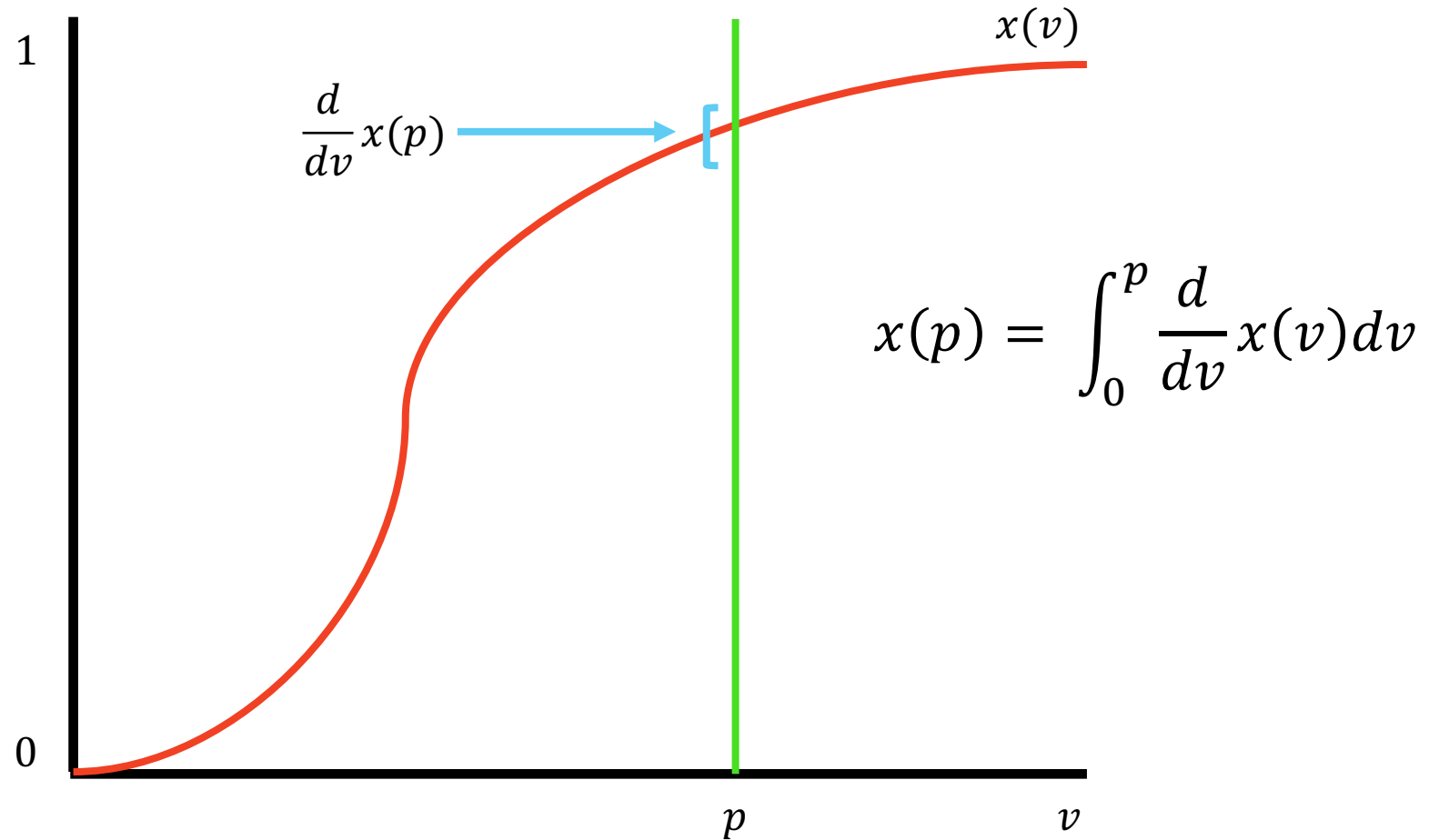
$$\begin{aligned} & (1, \frac{1}{3} \$3 + \frac{2}{3} \$6) \\ & (\frac{1}{3}, \frac{1}{3} \$3) \\ & (0, 0) \end{aligned}$$



+ 2/3



# Any allocation is a distribution over prices



# What is our revenue for a price $p$ ?

Single-bidder revenue curve  $R(p) = p \cdot \Pr_v[v \geq p] = p \cdot [1 - F(p)]$

Moving to quantile space:

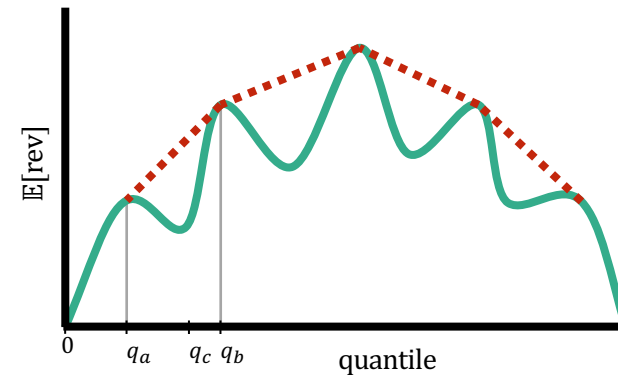
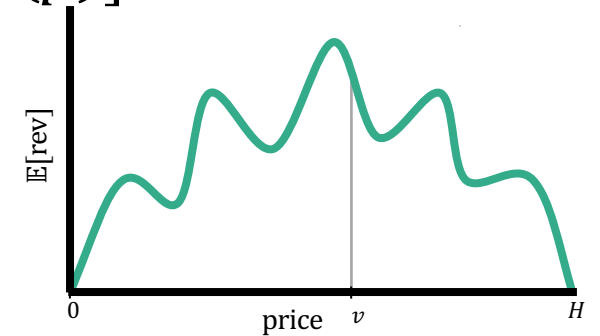
$$q = 1 - F(v) \quad v(q) = F^{-1}(1 - q) \quad q \sim U[0,1]$$

Single-bidder revenue curve in quantile space

$$P(q) = v(q) \cdot q$$

Happily,  $\frac{d}{dq} P(q) = \varphi(v(q))$

We define  $\frac{d}{dq} \bar{P}(q) = \bar{\varphi}(v(q))$  where is  $\bar{P}(\cdot)$  the concave closure of  $P(\cdot)$ .



# Maximizing Revenue

For virtual value functions

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

Only DSIC if  $\varphi_i(v_i)$  is monotone

Expected Revenue

$$= \mathbb{E}_v \left[ \sum_i p_i(\mathbf{v}) \right] = \mathbb{E}_v \left[ \sum_i x_i(\mathbf{v}) \varphi_i(v_i) \right] =$$

Expected Virtual Welfare

True by payment identity OR

$$\frac{d}{dq} P(q) = \varphi(v(q))$$

To max rev, choose  $x$  to maximize this

$$= \mathbb{E}_v \left[ \sum_i x_i(\mathbf{v}) \bar{\varphi}_i(v_i) \right]$$

with  $x = 0$  when  $\bar{\varphi}_i \neq \varphi_i$

# Multiparameter Social Welfare: VCG is DSIC

$$x := \operatorname{argmax} \sum_j v_j(x_j(b_i, \mathbf{b}_{-i}))$$

More utility for bidding actual value:

$$\underline{v_i(x_i(v_i, \mathbf{b}_{-i})) - p_i(v_i, \mathbf{b}_{-i})} \geq v_i(x_i(b_i, \mathbf{b}_{-i})) - p_i(b_i, \mathbf{b}_{-i}) \quad \forall i, v_i, b_i, \mathbf{b}_{-i}$$

$i$  wants to max wrt  $(v_i, \mathbf{b}_{-i})$

$$p_i(b_i, \mathbf{b}_{-i}) = \sum_{j \neq i} v_j(x_j(0, \mathbf{b}_{-i})) - \underline{\sum_{j \neq i} v_j(x_j(b_i, \mathbf{b}_{-i}))}$$



value  $v_i$

utility  $v_i x_i(\mathbf{b}) - p_i(\mathbf{b})$

bid  $b_i$

max w/o  $i$ ,  
unrelated to  $i$ 's bid

curr welf w/o  $i$ ,  
 $x$  is defined to max  
wrt  $\mathbf{b}$