

Simple item auction price of anarchy  
n players  $v_1 \dots v_n$  value for item

SW giving item to bidder i at price p

$$SW = \underbrace{(v_i - p)}_{\text{value for winner}} + \underbrace{p}_{\text{value for auctioneer}} + \underbrace{0 + \dots + 0}_{\text{value for others}} = v_i$$

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Way to prove PoA:

$$u_i(s_i^*, s_{-i})$$

$s_i^*$  special strategy

$s$  any other strategy vector

Example  
bid  $v_i/2$

Claim  $s$  any bid-vector

$$u_i\left(\frac{v_i}{2}, s_{-i}\right) \geq \frac{1}{2} v_i - \text{price}(s)$$

$$\text{price}(s) = \max_j s_j$$

Proof: • OK if  $\frac{v_i}{2}$  wins

$$\Rightarrow u_i\left(\frac{v_i}{2}, s_{-i}\right) = v_i - \frac{v_i}{2}$$

• if  $\frac{v_i}{2}$  loses

$$\text{price}(s) > \frac{v_i}{2}$$

Also true  $u_i\left(\frac{v_i}{2}, s_{-i}\right) \geq 0$

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Price of anarchy proof

$$\sum_i u_i\left(\frac{v_i}{2}, s_{-i}\right) \geq \frac{1}{2} \max_j v_j - \text{price}(s)$$

reasons use top for approx  $v_j$   
second one for the rest

Assume  $s$  Nash (pure strategy)

$$\underbrace{\sum_i u_i(s)}_{\text{red}} \geq \sum_i u_i\left(\frac{v_i}{2}, s_{-i}\right) \geq \underbrace{\frac{1}{2} \max_j v_j}_{\text{blue}} - \underbrace{\text{price}(s)}_{\text{red}}$$

$$\text{SW}(s) \geq \frac{1}{2} \max_j v_j = \frac{1}{2} \text{OPT}$$

# General framework for PoA in auctions

$$\sum_i u_i(s_i^*, s_{-i}) \geq 2 \text{OPT} - \text{Rev}(s)$$

↑  
Opt social  
welfare

↑  
revenue  
for auctioneer  
at  $s$

$$\Rightarrow \text{SW}(s) \geq \lambda \cdot \text{OPT}$$

all Nash equilibria  $s$

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Next mixed Nash:  $\sigma$  Nash  
same here: use expectations

$$\begin{aligned} \sum_{i \in \text{SNR}} E(u_i(s)) &\geq \sum_{i \in \text{SNR}} E(u_i(\frac{v_i}{2}, s_{-i})) \\ &\geq \cancel{E}(\lambda \cdot \text{OPT}) - E(\text{price}(\lambda)) \end{aligned}$$

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& no-regret learning outcome

$$\sum_t u_i(s^t) \geq \sum_t u_i(\frac{v_i}{2}, s_{-i}^t) - \text{Reg}$$

discretize options  $\varepsilon v_i$  rounding

$$\frac{1}{\varepsilon} \text{ options to bid} \quad \text{Reg} = \sqrt{2 \ln \frac{1}{\varepsilon} T}$$

$$\sum_t \sum_i u_i(s_t^*) \geq \sum_t \sum_i u_i\left(\frac{v_j}{2}, s_{-i}^*\right) - n \text{Reg}$$

$$\geq \sum_t \frac{1}{2} \max_j v_j - \sum_t \text{price}(s_t^*) - n \text{Reg}$$

$$\sum_t \text{sw}(s_t^*) \geq \frac{1}{2} T \max_j v_j - n \text{Reg}$$

All pay! need

$$\sum_i u_i(s_i^*, s_{-i}) \geq \lambda \max_j v_j - \text{Rev}(s)$$

$j$  not max value  $s_j^* = 0$

$$u_i(0, s_{-i}) \geq 0$$

$j$  is max  $s_j^*$  units  $[0, v_j]$

$$E(u_{-j}(s_j^*, s_{-j}))$$

$$= \frac{1}{v_j} \int_0^{v_j} u_{-j}(x, s_{-j}) dx$$

$$= v_j \left( -\frac{v_j}{v_j} - p \right) - \frac{v_j^2}{2} = \frac{1}{2} v_j^2 - p$$

exp payment

$$p = \max_{i \in J} s_i$$

prob of winning  $\frac{v_i - p}{v_i}$

Result:

$$E\left(\sum u_i(s_i^*, s_{-i})\right) \geq \frac{1}{2} \max_j v_j - \text{Rev}(s)$$

$$\Rightarrow SW(s) \geq \frac{1}{2} \text{OPT}$$



