

The sensitivity of competitive auctions to prior information

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Motivation

- worst-case (CS) vs. average-case (Econ) analysis
- two sampling-based truthful, randomized competitive auctions
 - ◇ SCS – worst-case 4-competitive
 - ◇ DSOT – worst-case 380-competitive
- empirical evaluation of average revenue
 - ◇ typical distributions
 - ◇ compare to posted-price, omniscient mechanisms
- single-item optimal auction
 - ◇ robustness to noise in the prior distribution

Dual-Price Sampling Optimal Threshold Auction (DSOT)

- Partition bids \mathbf{b} at random into two sets \mathbf{b}' and \mathbf{b}'' .
- Let $p' = \text{opt}(\mathbf{b}')$ and $p'' = \text{opt}(\mathbf{b}'')$ be prices that yield the most revenue in fixed price auctions on \mathbf{b}' and \mathbf{b}'' .
- Use p' as threshold for all bids in \mathbf{b}'' :
 - ◇ all bids in \mathbf{b}'' of value below p' are rejected
 - ◇ all remaining bids win at price p' .
- Use p'' as threshold for all bids in \mathbf{b}' .

Sampling Cost-Sharing Auction (SCS)

— based on CostShare_C .

CostShare_C : Find the largest k such that the highest k bidders' values are at least C/k . Charge each C/k .

$\mathcal{F}(\mathbf{b}) =$ optimal profit of fixed price auction on \mathbf{b} .

- Divide vector \mathbf{b} randomly into \mathbf{b}' and \mathbf{b}'' .

Let $\mathcal{F}' = \mathcal{F}(\mathbf{b}')$ and $\mathcal{F}'' = \mathcal{F}(\mathbf{b}'')$.

- Run $\text{CostShare}_{\mathcal{F}'}$ on \mathbf{b}'' and $\text{CostShare}_{\mathcal{F}''}$ on \mathbf{b}' .

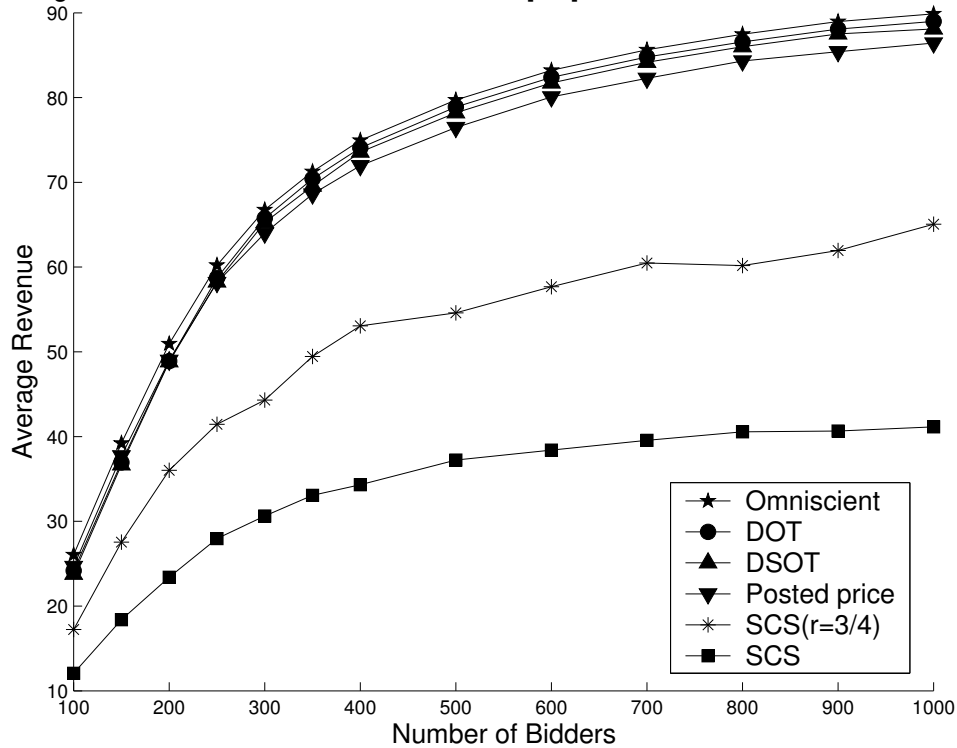
SCS has revenue $\min(\mathcal{F}', \mathcal{F}'')$.

Improving SCS

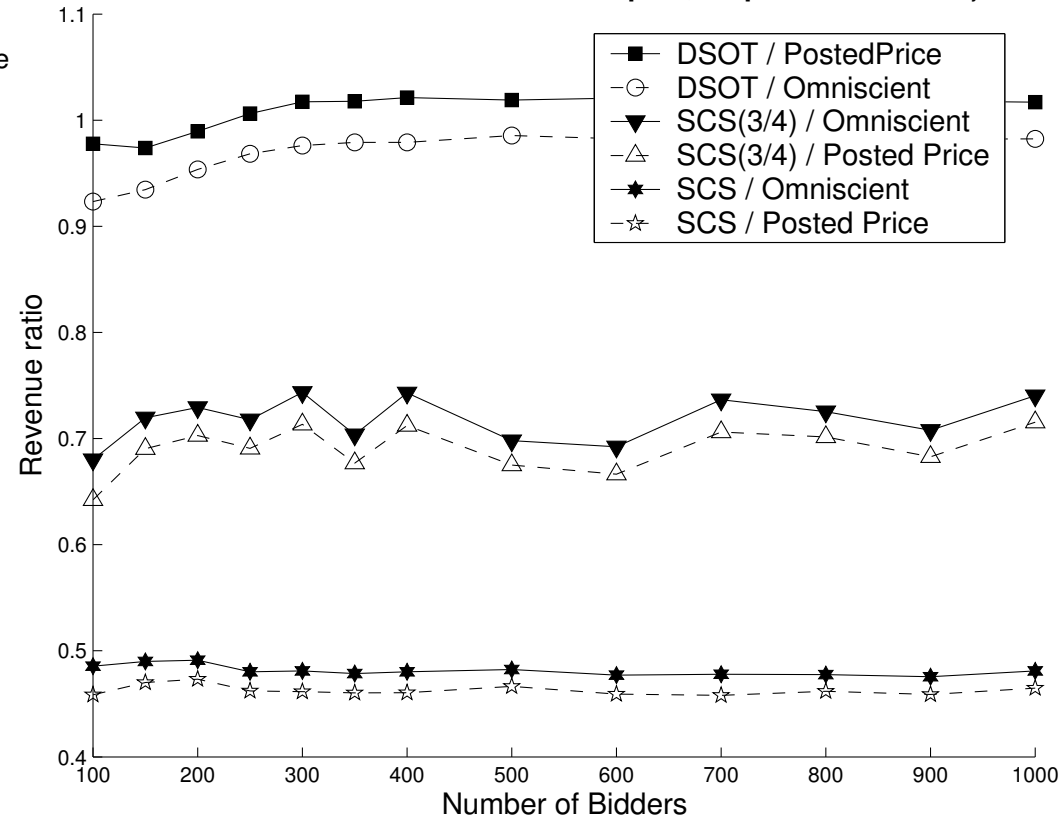
- b' can support cost ≤ 80 ; b'' can support cost ≤ 100
- SCS asks b' to support 100 and b'' to support 80 \Rightarrow Profit **80**
- **Idea:** ask b' to support cost $\frac{3}{4} * 100 = 75$ and b'' to support $\frac{3}{4} * 80 = 60$; \Rightarrow Profit **135**
- **Intuition:** many iid bids $\Rightarrow b'$ and b'' support close costs \Rightarrow can use $\frac{3}{4}$
- We formalize intuition for $U[0;1]$ distribution
 - ◇ may work for continuous well-behaved distributions
 - ⊗ Don't try this at home!
- Empirically, improved $SCS(\frac{3}{4})$ performs significantly better than standard SCS

U[0;1] distribution, bounded supply

Average revenue for i.i.d. bids from a Uniform[0;1] distribution with 100 available objects

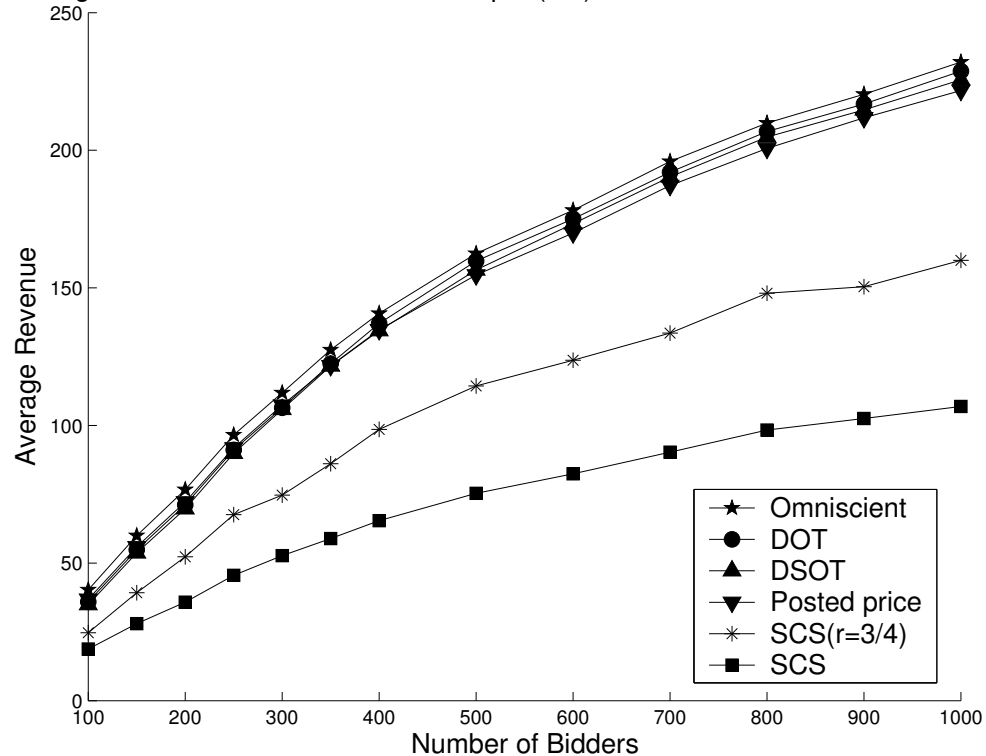


Revenue ratios for noise-free auctions for the Uniform[0.000;1.000] distribution with 100 objects

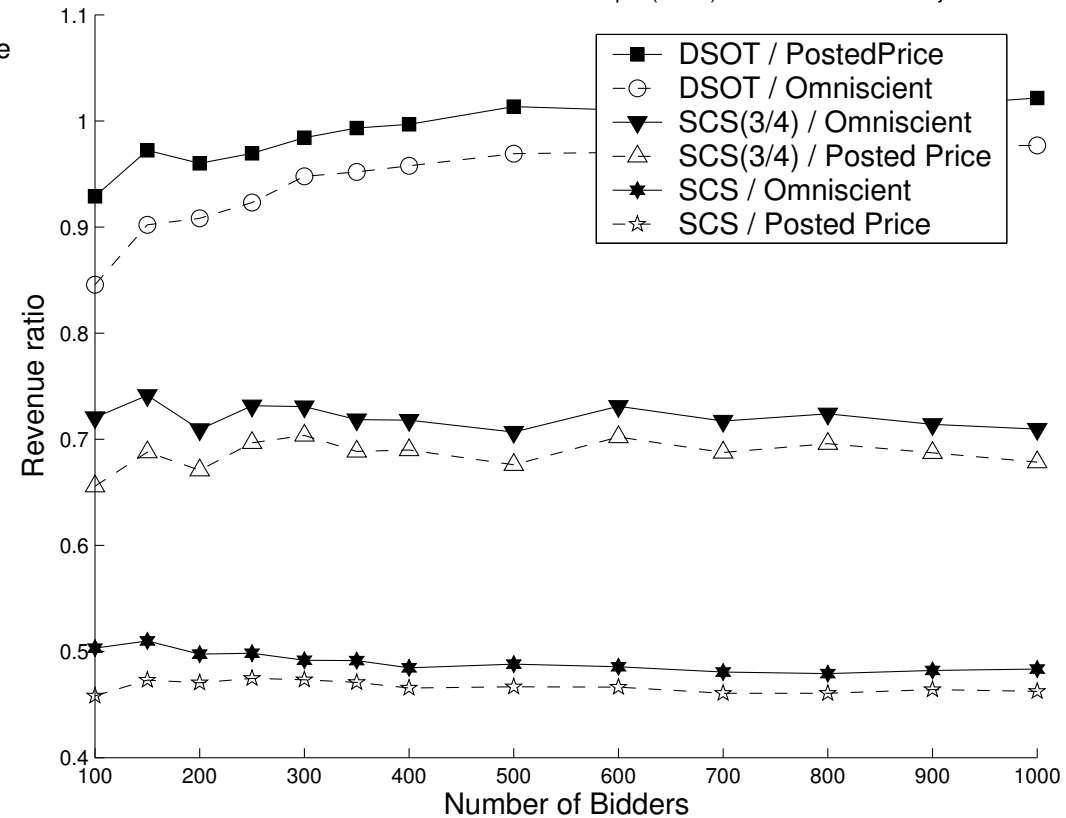


Exponential distribution, bounded supply

Average revenue for i.i.d. bids from a Expon(1.0) distribution with 100 available objects

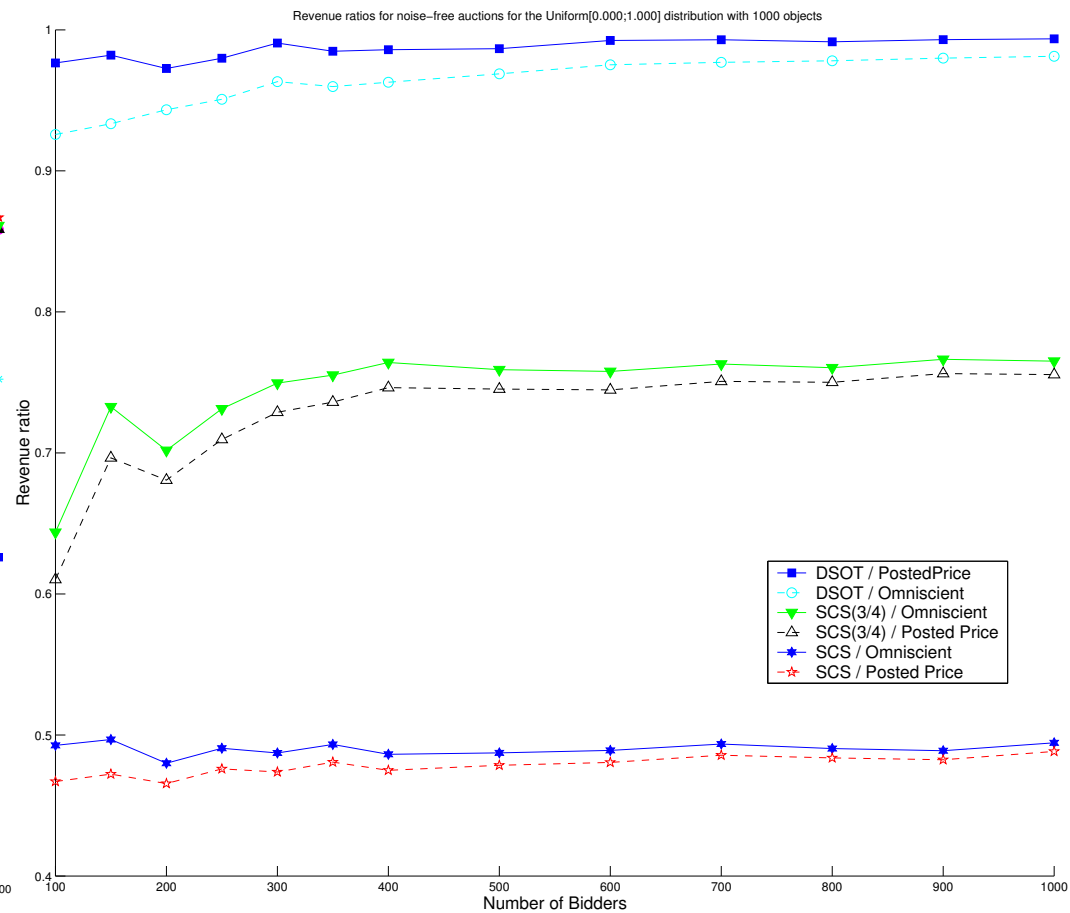
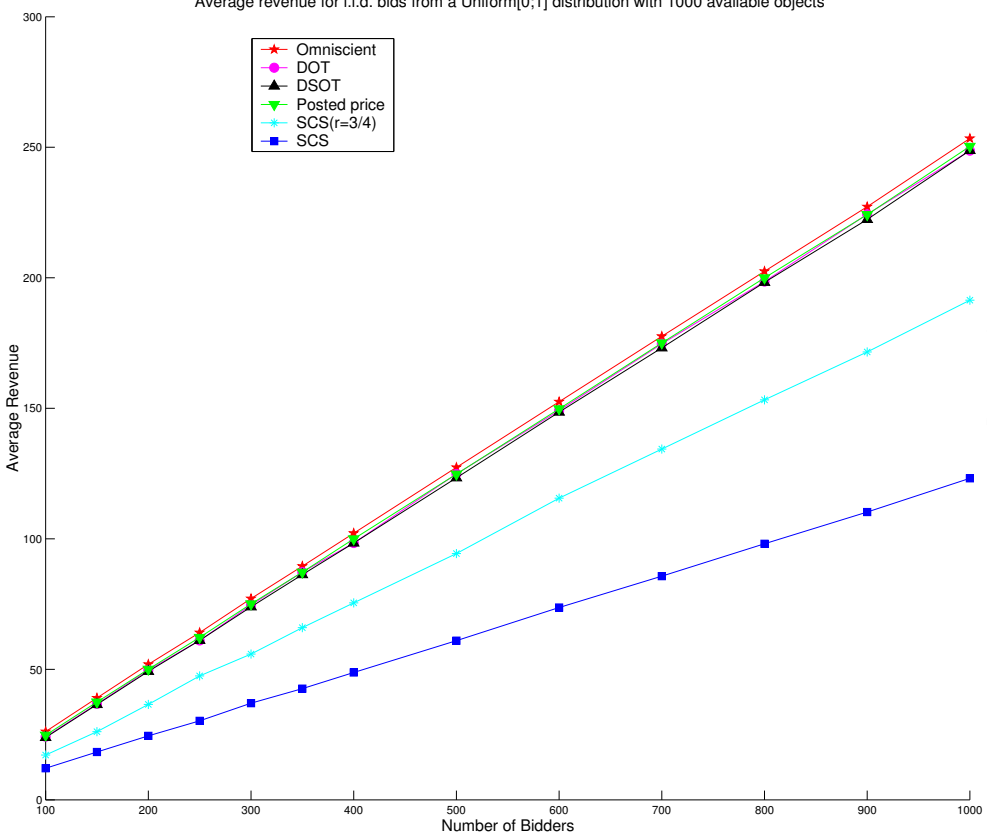


Revenue ratios for noise-free auctions for the Expon(1.000) distribution with 100 objects



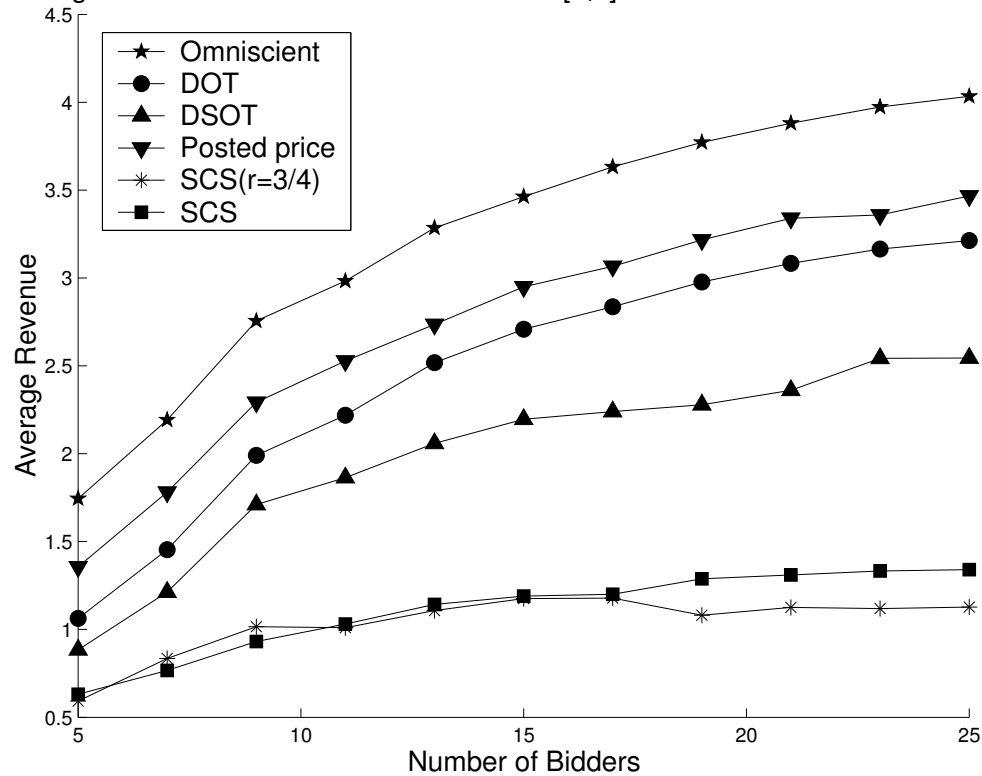
U[0;1] distribution, unbounded supply

Average revenue for i.i.d. bids from a Uniform[0;1] distribution with 1000 available objects

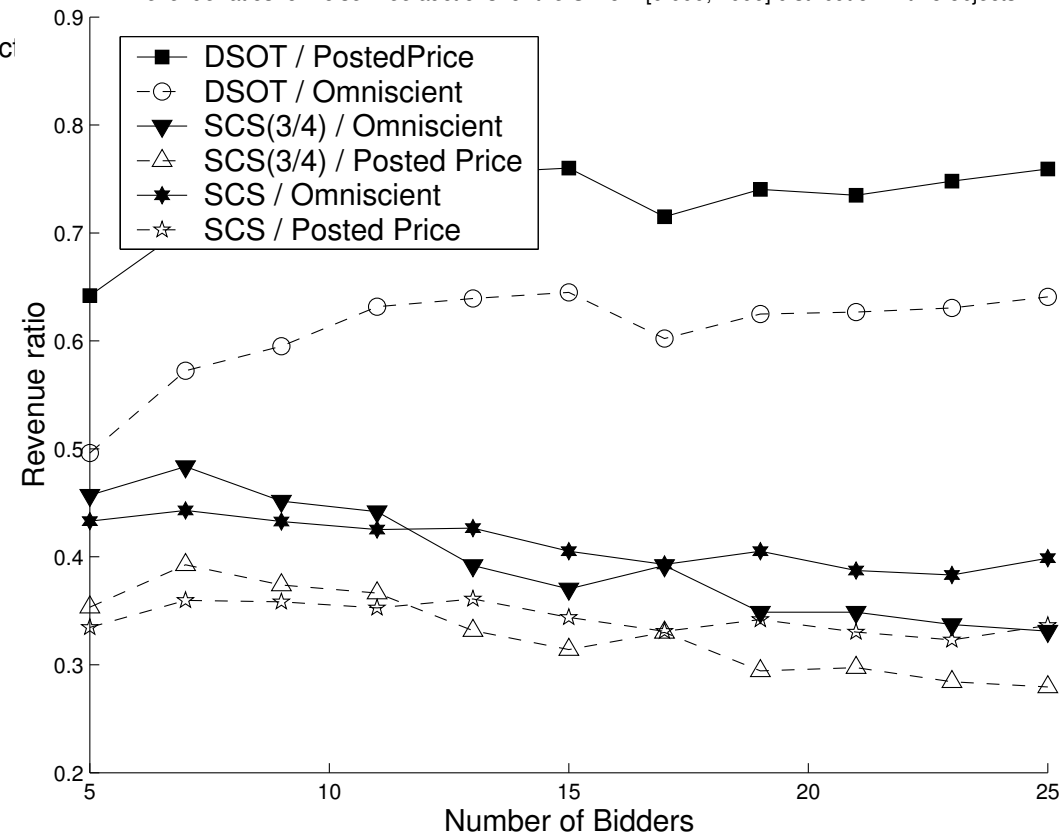


Few items, few bidders, U[0;1] distribution

Average revenue for i.i.d. bids from a Uniform[0;1] distribution with 5 available objects

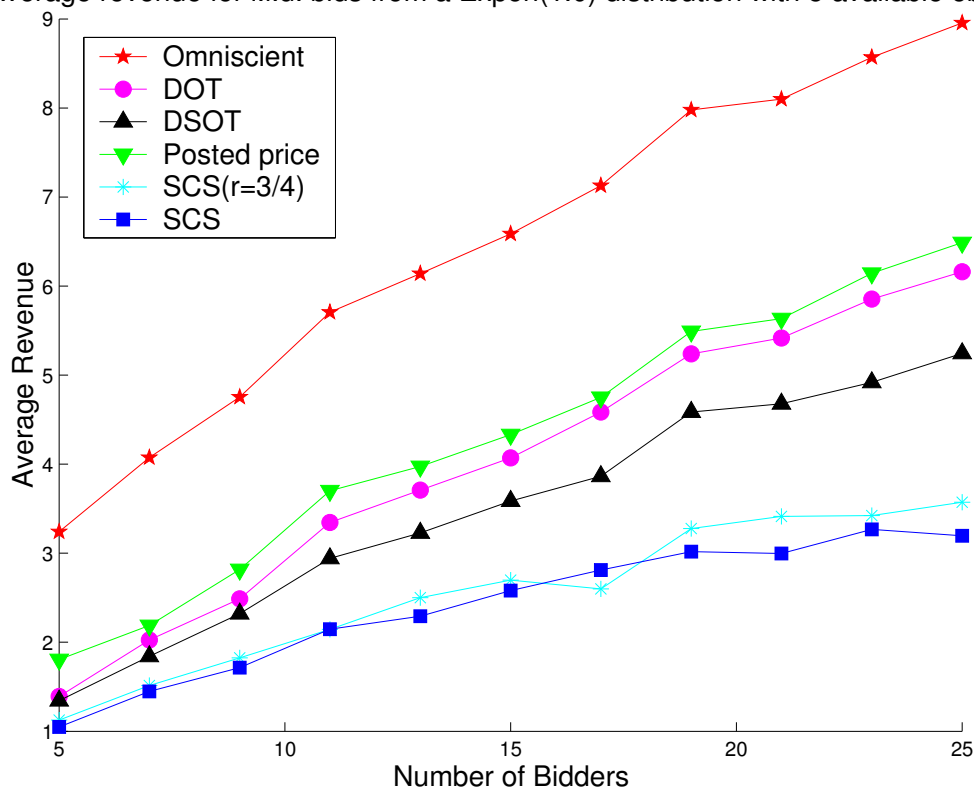


Revenue ratios for noise-free auctions for the Uniform[0.000;1.000] distribution with 5 objects

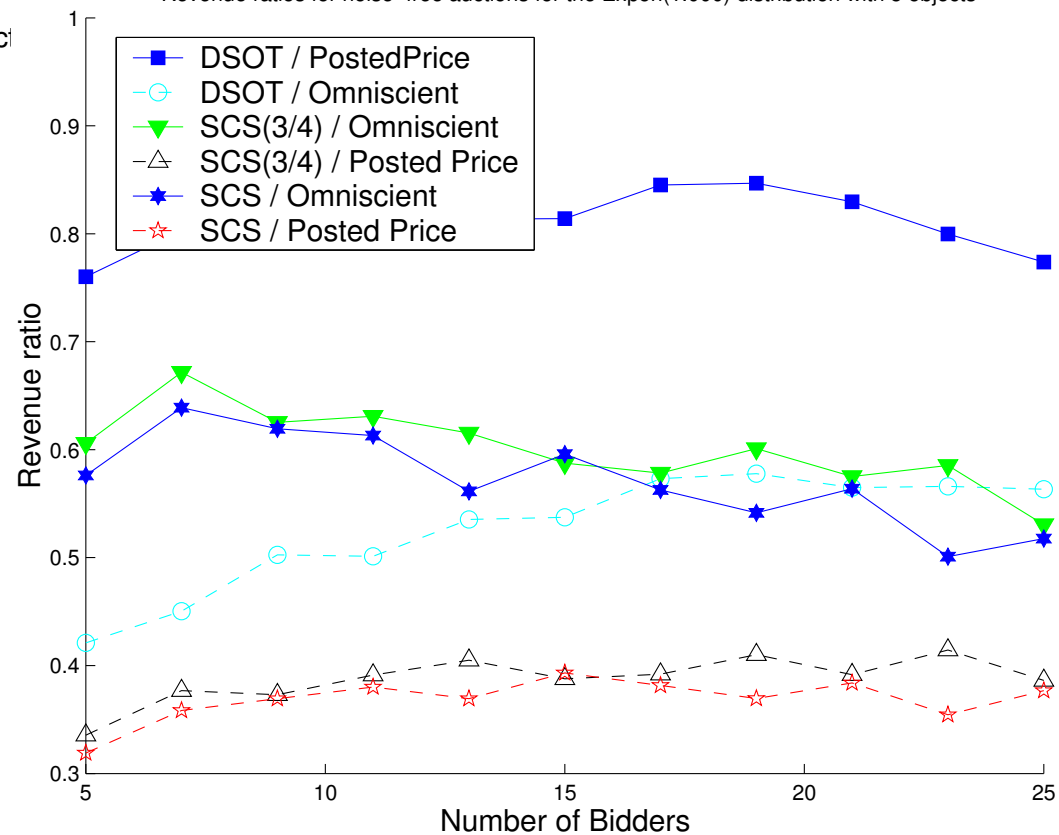


Few items, few bidders, Exponential distribution

Average revenue for i.i.d. bids from a Expon(1.0) distribution with 5 available objects

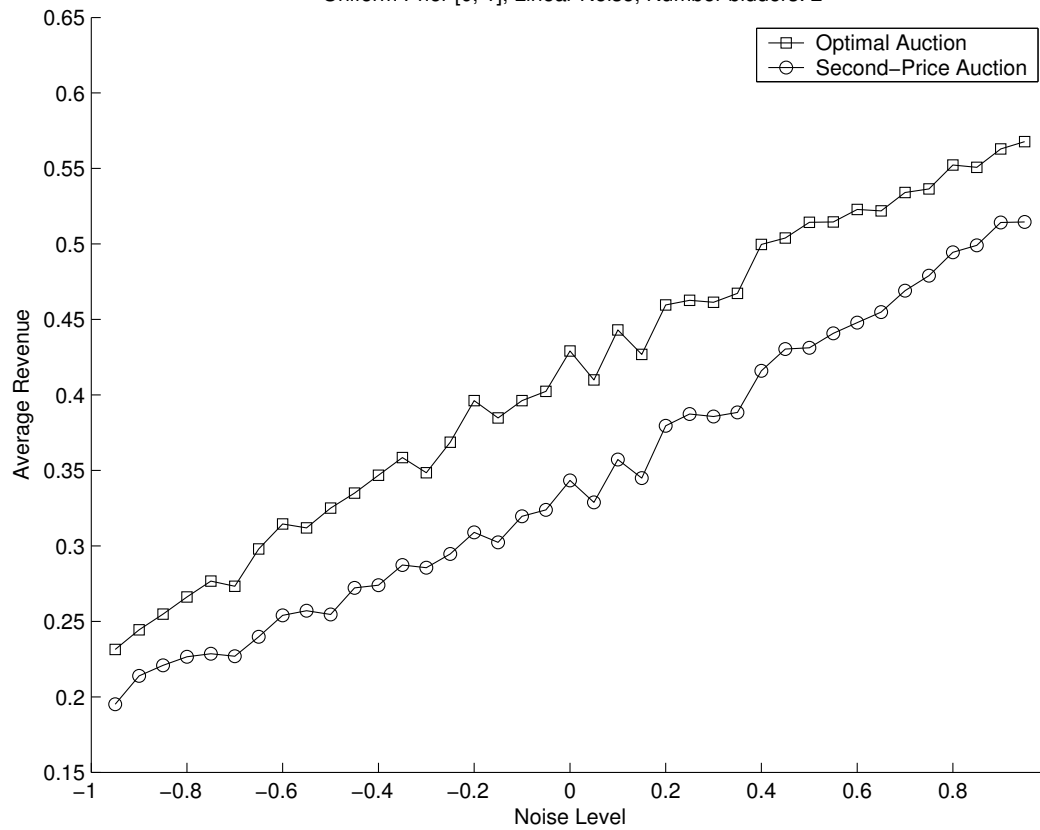


Revenue ratios for noise-free auctions for the Expon(1.000) distribution with 5 objects

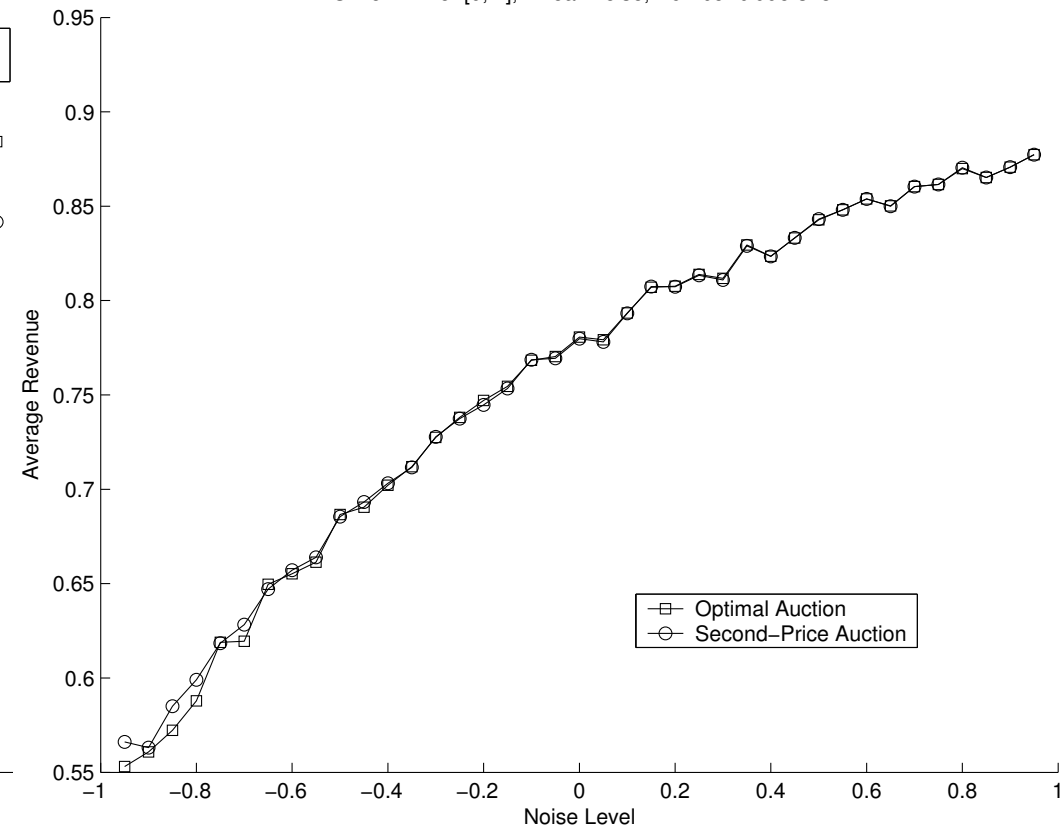


Linear Noise

Uniform Prior [0, 1], Linear Noise, Number bidders: 2

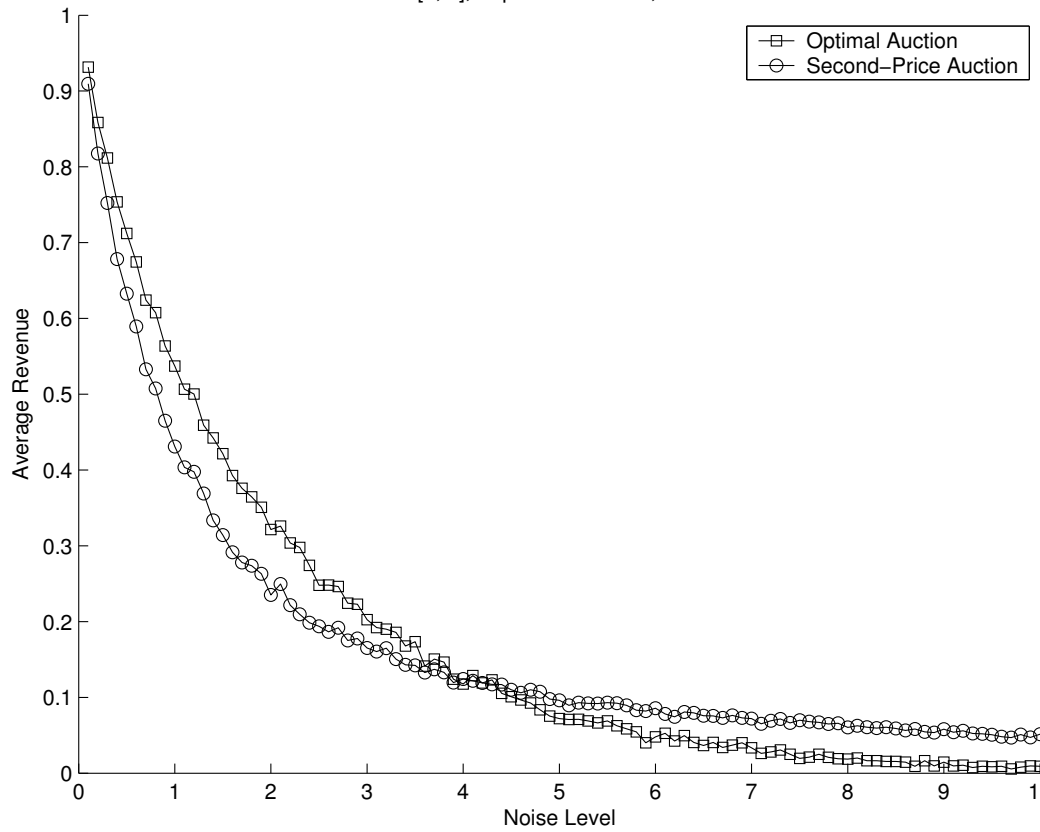


Uniform Prior [0, 1], Linear Noise, Number bidders: 8

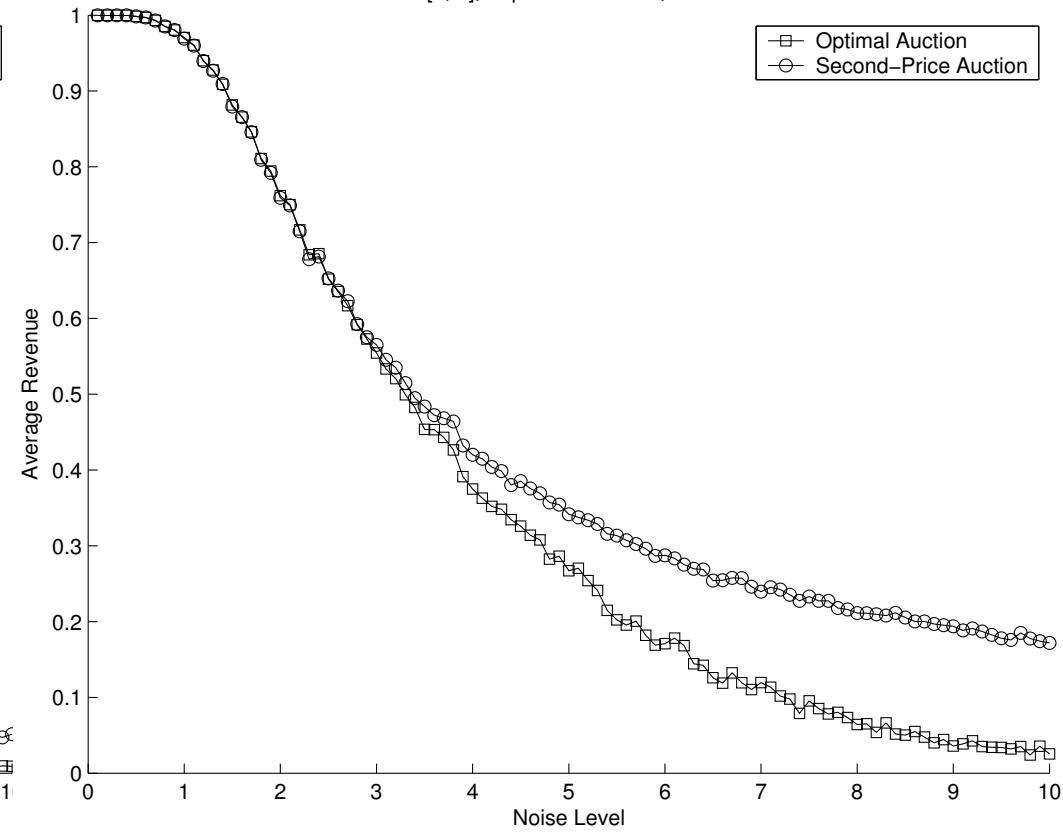


Exponential Noise

Uniform Prior [0, 1], Exponential Noise, Number bidders: 2



Uniform Prior [0, 1], Exponential Noise, Number bidders: 8



Future Work

- mechanism with good worst-case performance and much better average-case performance
- worst-case optimal auction
- low revenue variance with high probability
- average-case theoretical analysis