The Strange Role of Information Asymmetry in Auctions — Does More Accurate Value Estimation Benefit a Bidder?

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Abstract

We study the second-price auction in which bidders have asymmetric information regarding the item’s value. Each bidder’s value for the item depends on a private component and a public component. While each bidder observes their own private component, they hold different and asymmetric information about the public component. We characterize the equilibrium of this auction game and study how the asymmetric bidder information affects their equilibrium bidding strategies. We also discover multiple surprisingly counter-intuitive equilibrium phenomena. For instance, a bidder may be better off if she is less informed regarding the public component. Conversely, a bidder may sometimes be worse off if she obtains more accurate estimation about the auctioned item. Our results suggest that efforts devoted by bidders to improve their value estimations, as widely seen in today’s online advertising auctions, may not always be to their benefit.

Introduction

Thanks to convenient access to information and the significant advances of machine learning techniques, we can now make unprecedentedly accurate predictions about various uncertain quantities. Naturally, these predictions have also changed how our decisions are made. One prominent example, among many others, is the significant effort devoted by both Internet companies and online advertisers to predicting an Internet user’s interest in their ads. Such predicted level of interest, typically characterized by the click through rate (Richardson, Dominowska, and Ragno 2007), will then affect the advertiser’s value, and thus the bid, for the user’s impression. More generally, when we make many other purchases (e.g., a house, used car or an artwork), we tend to collect as much data as possible in order to reduce our uncertainty about the item’s value estimation.

An important question, however, is whether such more accurate estimations always benefit us. At a first glance, this answer might seem obvious — how could knowing more ever be harmful? Certainly, if an agent’s payoff, with possible uncertainty, only depends on its own decision and is not affected by any other agent’s actions (i.e., there are no externalities), then more accurate estimations always improve the agent’s decision (Blackwell 1953). However, the answer turns out to be highly non-trivial once an agent’s payoff has externalities, and specifically, is affected by other self-interested agents’ actions in a strategic setup. This raises the following basic and intriguing question:

In real-world strategic settings with uncertainty, does more accurate estimation always benefit the agent?

In this paper, we investigate the above question in the fundamental sealed-bid second price auction (a.k.a., Vickrey auction) which is widely used in the online advertising industry (Edelman, Ostrovsky, and Schwarz 2007). Specifically, we consider a second-price auction with \(n + 1\) bidders \(0, 1, \ldots, n\). The valuation of the auctioned item for each bidder \(i\) is the sum of a private component \(V_i\) and a public component \(C\) that is common to all bidders. While \(V_i\)’s and \(C\) are all drawn from publicly known distributions, bidders have asymmetric information about the public component \(C\); that is, (only) bidder 0 can observe the realized \(C\). As a motivating example, consider an ad auction among insurance companies who all bid for an Internet user searching the keyword of “insurance”. The common component \(C\) here captures the total profit any insurance company could earn from this particular user, which depends on the user’s attributes and typically does not differ much across insurance companies. The private component \(V_i\) describes the specific operational cost of insurance company \(i\). Bidder 0 could be a performance-driven advertiser who delegates her bidding to professional marketing agencies which indeed have much more accurate information about Internet users.

This work aims at understanding how the special bidder 0’s extra information may affect the equilibrium as well as agents’ utilities by comparing two different situations: (1) the asymmetric world as we described above; and (2) the symmetric world in which bidder 0 also cannot observe \(C\).

Summary of results. As a necessary step, we first characterize the equilibrium of the above game under information asymmetry. Specifically, when there are only two bidders, 0 and 1, we provide an explicit characterization of the equilibrium under mild assumptions. As an application of this characterization, we prove that there exists a critical value \(t\)
for the uninformed bidder 1 such that he bids more conserva-
tively than his bid in the symmetric world whenever his value \( v_1 \leq t \), but bids more aggressively when \( v_1 \geq t \). For \( n > 1 \), we derive differential equations that characterize the bidding equilibrium. As an application of this characteriza-
tion, we show that uninformed bidders’ bids decrease as \( n \) increases; that is, more competition leads to more conserva-
tive bidding in the asymmetric world, whereas bidders’ equilibrium bid remain the same in the symmetric world.

We then leverage our equilibrium characterization to dis-
cover multiple quite surprising facts, which illustrate the in-
tricate role of information asymmetry in auctions. We show
that the informed bidder 0 may be worse off in the asym-
metric world — i.e., bidder 0’s information advantage may
be harmful to her. Surprisingly, we also find that an unin-
formed bidder may be better off in the asymmetric world
— i.e., ignorance of her value may increase a bidder’s util-
ty at equilibrium. Finally, we observe that the auctioneer’s revenue may decrease as the uninformed bidders learn ad-

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\[ \text{formed bidder may be better off in the asymmetric world} \]

\[ \text{that the informed bidder} \]

\[ \text{equilibrium bid remain the same in the symmetric world.} \]

\[ \text{We remark that such a counter example was discovered re-} \]

\[ \text{cently in (Syrgkanis, Kempe, and Tardos 2015) as well, but with} \]

\[ \text{a much more involved analysis than ours.} \]

\[ \text{The study of information asymmetry in auc-} \]

\[ \text{tions has a long history, dating back to the seminal work of} \]

\[ \text{Wilson (1967). Most previous works have focused on } \]

\[ \text{first price auction, and like us, have assumed that one bidder is} \]

\[ \text{fully informed whereas all other bidders are not informed} \]

\[ \text{(Wilson 1967; Milgrom and Weber 1982; Engelbrecht-} \]

\[ \text{Wiggins, Milgrom, and Weber 1983). However, our focus in} \]

\[ \text{this work is a different yet similarly basic auction format,} \]

\[ \text{the second price auction.} \]

\[ \text{The seminal work of Milgrom and Weber (1982) proved} \]

\[ \text{the linkage principle (as mentioned above) for commonly} \]

\[ \text{seen auction formats with symmetric bidders and positively} \]

\[ \text{correlated (a.k.a., associated) bidder signals. Hausch (1987)} \]

\[ \text{and another interesting recent work (Syrgkanis, Kempe,} \]

\[ \text{and Tardos 2015) analyze the equilibrium of both first and sec-
ond price auctions but with non-symmetric bidders. Both} \]

\[ \text{works consider only two bidders and demonstrate the fail-} \]

\[ \text{ure of the linkage principle in their models. Most of these} \]

\[ \text{previous works assume common value. However, our model} \]

\[ \text{can have many bidders and uses a more general bidder value} \]

\[ \text{model (which includes the common value model as a special} \]

\[ \text{case with } V_i \equiv 0 \text{ for each bidder } i \text{). On the other hand, the} \]

\[ \text{information structure in our model is simpler by assuming} \]

\[ \text{one bidder fully observes } C \text{ while all others observe no in-} \]

\[ \text{formation. Therefore, our results are not quite comparable.} \]

\[ \text{Moreover, these previous works are all primarily focused on the} \]

\[ \text{auctioneer’s revenue, whereas we mainly study bidders’ bid-} \]

\[ \text{ding behavior and utility changes at equilibrium. To our} \]

\[ \text{knowledge, this perspective has not been investigated much} \]

\[ \text{before.} \]

\[ \text{Our work is also related to recent literature on design-} \]

\[ \text{ing signaling schemes to influence bidders’ bidding strategy} \]

\[ \text{(Emek et al. 2012; Badanidiyuru, Bhawalkar, and Xu 2018;} \]

\[ \text{Li and Das 2019; Bergemann et al. 2022). However, these} \]

\[ \text{works also focuses on the auctioneer’s revenue. Our work} \]

\[ \text{can be viewed as understanding the equilibrium under one} \]

\[ \text{particular signaling scheme, i.e., with full information re-} \]

\[ \text{vealed to exactly one bidder.} \]

\[ \text{Preliminaries} \]

\[ \text{Basic Setup.} \]

\[ \text{We consider a single-item second-price auc-
ction with } n+1 \text{ bidders, denoted by set } [n] = \{0, 1, \ldots, n\}, \]

\[ \text{and correlated bidder values. Bidder } i \text{ has value } V_i + C \text{ for} \]

\[ \text{the item where } V_i \text{ is the private value that depends only on } i, \]

\[ \text{and } C \text{ is the common value that is the same for ev-
ery bidder. For any } i \in [n], V_i \text{ is a random variable drawn from a con-
tinuous distribution with strictly positive probability density function (PDF) } f_i(v) \text{ supported on } [l_i, u_i] \text{ for} \]

\[ \text{some } 0 \leq l_i < u_i. \text{ Let } F_i(v) \text{ be the corresponding cumu-
lative distribution function (CDF). The common value } C \text{ is also a random variable drawn from a continuous di-
tribution with strictly positive PDF } \]

\[ \text{g(c) supported on } [l_c, u_c] \text{ for} \]

\[ \text{0 \leq l_c < u_c; G(c) is the CDF. By convention, we use capital} \]

\[ \text{letters (e.g., } V_i \text{) to denote random variables and small letters} \]

\[ \text{(e.g., } v_i \text{) denote the realization. As a standard assumption,} \]

\[ \text{V_i's and } C \text{ are assumed to all be independent.} \]

\[ \text{In the second-price auction, each bidder } i \in [n] \text{ submits a} \]

\[ \text{bid } b_i. \text{ The highest bid wins, receives the auctioned item, and} \]

\[ \text{pays the second highest bid. Let } i^* \text{ be the winning bidder and} \]

\[ \text{b_{i^*} be the second highest bid. The payoff of } i^* \text{ is } v_{i^*} + c - b_{i^*} \text{ and the payoffs of all other bidders are } 0. \]

\[ \text{Information Structure.} \]

\[ \text{All distributions are common knowledge, i.e., known to all bidders. Each bidder } i \text{ can observe the} \]

\[ \text{realization of his own private value } v_i. \text{ However, } \]

\[ \text{bidder 0 is the only special bidder who additionally observes the} \]

\[ \text{realized common value of } c. \text{ So the game exhibits infor-
modation asymmetry — bidder 0 has more accurate estimation of } c \text{. As widely known in the literature (Milgrom and Wei-
ber 1982), under information asymmetry, truthful bidding ceases to be an equilibrium in second-price auctions. Nev-
ethertheless, the following lemma shows that truthful bidding} \]

\[ \text{remains a dominant strategy for bidder 0 (but not for other} \]

\[ \text{bidders; see examples later). Its proof follows the same} \]

\[ \text{argument as the truthfulness proof in a standard VCG mecha-
nism, and thus is omitted.} \]

\[ \text{Lemma 1. Truthful bidding is a dominant strategy for bid-
der 0.} \]

\[ \text{Consequently, we shall assume throughout the paper that} \]

\[ \text{bidder 0 always bids truthfully. Our equilibrium charac-
terization is therefore focused on finding the bidding strategy of the} \]

\[ \text{uninformed bidders.} \]

\[ \text{Additional discussion on our model.} \]

\[ \text{When studying information asymmetry in auctions, it is common to consider correlated bidder values since when} \]

\[ \text{values are independent, one bidder’s information would not affect any other bidders’ valuations. In this literature, the} \]

\[ \text{valuation model with a com-
mon and a private component is prevalent (Milgrom and We-
ber 1982; Abraham et al. 2013; Li and Das 2019). Moreover,} \]

\[ \text{it also strictly generalizes the classic common-value auction} \]

\[ \text{(Emek et al. 2012; Badanidiyuru, Bhawalkar, and Xu 2018;} \]

\[ \text{Li and Das 2019; Bergemann et al. 2022). However, these} \]

\[ \text{works also focuses on the auctioneer’s revenue. Our work} \]

\[ \text{can be viewed as understanding the equilibrium under one} \]

\[ \text{particular signaling scheme, i.e., with full information re-} \]

\[ \text{vealed to exactly one bidder.} \]
model (when $V_i = 0, \forall i$) originating from the seminal work of Wilson (1966; 1967). Most related to our specific model is that of (Arnosti, Beck, and Milgrom 2016), motivated by online advertising, in which $C$ captures the attributes of the auctioned user impression that are valuable to all advertisers (i.e., bidders), $V_i$ captures bidder $i$’s private value per impression, and they are independent. In another motivating domain of art auction, $C$ captures the resale value of the auctioned artwork and $V_i$ captures bidder $i$’s personal preferences about the artwork (Goetzmann and Spiegel 1995).

**Theme and structure of the remainder.** A main theme of this paper is to study whether bidder $0$’s more accurate value estimation would necessarily benefit herself and harm other bidders. To do so, we first characterize the equilibrium of the game in Sections and for the cases of two bidders and many bidders, respectively. Section presents our findings and insights by analyzing the equilibrium.

**Equilibrium Analysis: Two-Bidder Cases**

To understand bidder payoffs at equilibrium, we must be able to characterize the equilibrium in the first place. In this section, we focus on the setting with two bidders, i.e., bidder 0 and bidder 1, and provide an explicit equilibrium characterization under reasonable distribution assumptions.

Since bidder 0 always bids her true value, we only need to characterize bidder 1’s equilibrium bidding function, denoted as $b^*(v_1)$, which maps her realized private value $v_1$ to the optimal bid $b^*(v_1)$. Particularly, given any private value $v_1 \in [l_1, u_1]$, the equilibrium bid of bidder 1 is the $b^*$ that maximizes her expected utility, as a function of $v_1$ and $b$, expressed as follows:

$$U(b; v_1) = \mathbb{E}_{v_0 \sim U_{v_0}} [v_1 + C - (V_0 + C) | b \geq V_0 + C] \cdot \Pr(b \geq V_0 + C)$$

$$= \int_{v_0 + c \leq b} (v_1 - v_0) f_0(v_0) g(c) dv_0 dc$$

$$= \int_0^b \int_0^{b-v_0} (v_1 - v_0) f_0(v_0) g(c) dc dv_0$$

$$= \int_0^b (v_1 - v_0) f_0(v_0) G(b - v_0) dv_0$$

(1)

Since $V_0$ has support $[l_0, u_0]$. If $v_1 \geq u_0$, for any $v_0$ we have $(v_1 - v_0) f_0(v_0) G(b - v_0) \geq 0$. Therefore, $U(b; v_1)$ will be maximized by any $b \geq u_0 + u_c$. That is, bidder 1 simply bids high enough to secure a winning position. On the other hand, if $v_1 \leq l_0$, bidder 1 will bid as low as possible as any winning will result in negative utility. Therefore, in the rest of this section we focus on the non-trivial case $v_1 \in (l_0, u_0)$.

By definition, the equilibrium bidding function $b^*(v_1) = \arg\max_{b \in \mathbb{R}} U(b; v_1)$ for any $v_1$. Unfortunately, we show that the optimal bid $b^*$ generally does not admit tractable structure and may have multiple maxima because the function $U(b; v_1)$ as a function of $b$ is non-convex in general for a fixed $v_1$. This is illustrated in the following example.

**Example 1.** Consider the problem with $V_0 \sim U_0 \left( \begin{array}{c} 0 \\ 3 \\ 7 \end{array} \right)$ where $U(I)$ denotes the uniform distribution over set $I$. The common component $C \sim \frac{1}{2} (\text{Beta}(2, 7) + \text{Beta}(7, 2))$, i.e., the uniform mixture of Beta($2, 7$) and Beta($7, 2$). Simple calculation shows that $g(c) = 28c(1-c)^6 + c^6(1-c)$ for all $c \in [0, 1]$. In this case, the explicit expression of bidder 1’s expected utility $U(b; v_1)$, as a function of $b$, is already rather complicated and thus omitted here. The following figure plots $U(b; v_1)$ for $v_1 = 3/8$. As we can see, $U(b; v_1)$ is neither convex nor concave, and has multiple local maxima.

![Figure 1: Plots of function $U(b; v_1)$ with $v_1 = 3/8$.](image_url)

Example 1 reveals some intricacies in characterizing bidder 1’s equilibrium bidding strategy due to multiple optimal bids. Generally, the optimal bidding function does not admit an explicit characterization. Our main result in this section is to identify a fairly general condition, under which bidder 1’s optimal bidding strategy can be explicitly characterized and also easily computed.

**Theorem 1.** If $g(c)$ is log-concave,\(^2\) then $b^*(v_1)$ is bidder 1’s equilibrium (i.e., optimal) bid for $v_1$ if and only if $b^*(v_1)$ satisfies equation $v_1 = \mathbb{E}(V_0 | V_0 + C = b^*(v_1))$. Moreover, $\mathbb{E}(V_0 | V_0 + C = b)$ is non-decreasing in $b$, therefore $b^*(v_1)$ can be efficiently computed for each $v_1$ via binary search.

The proof of Theorem 1 is rather involved and can be found in the online appendix. Key to our proof is to rule out all bad situations similar to Example 1, for which a characterization of the optimal bid is challenging. To overcome this issue, we leverage the log-concavity assumption to prove the following crucial property of random variables, which may be of independent interest.

**Lemma 2.** If $g(c)$ is log-concave, then $\mathbb{E}(V_0 | V_0 + C = b)$ is non-decreasing in $b$.

This property allows us to show that bidder 1’s utility function must increase first and then decrease. This allows us to characterize the optimal bid as the one with derivative equaling 0, leading to the condition in Theorem 1.

We remark that the log-concavity assumption in Theorem 1 is widely adopted for modeling value distributions (Bagnoli and Bergstrom 2006). In fact, many commonly used distributions — e.g., normal, exponential, logistic, Laplace and uniform distributions over a convex set — are all log-concave. It is easy to verify that the density function $g(c)$ in Example 1 is not log-concave. The following example is an application of Theorem 1.

\(^2\)A function $g$ is log-concave if $\log(g)$ is a concave function.
Example 2. Consider the case where $V_0$ is drawn from the log-concave distribution $\mathcal{U}([0, \alpha])$ for some $\alpha \in (0, 1]$ and $C \sim \mathcal{U}([0, 1])$. Applying Theorem 1, we have

$$b^*(v_1) = \begin{cases} 2v_1 & v_1 \in [0, \alpha/2); \\ any \ bid \ in \ [\alpha, 1] & v_1 = \alpha/2; \\ 2v_1 + 1 - \alpha & v_1 \in (\alpha/2, \alpha]. \end{cases}$$

Example 2 already reveals interesting insights regarding bidding behaviors under information asymmetry. When $v_1 < \frac{1}{4} \alpha$, bidder 1 bids $2v_1$, which is less than $v_1 + E(C)$, despite knowing that the common value $C$ has high expected value $E(C) = 0.5$. On the other hand, when $v_1 > \frac{3}{4} \alpha$, bidder 1 dramatically increases her bid to $2v_1 + 1 - \alpha > 1$, which is much larger than both her private component $v_1$ and the expectation of the common value. That is, bidder 1 bids conservatively when her private value is small but bids aggressively otherwise. It turns out that this intuition can be made precise, as shown in the following proposition.

**Proposition 1.** If $f_0(v_0)$ and $g(c)$ are both log-concave, then there exists a critical value $t \in [l_1, u_1]$, such that $b^*(v_1) \leq v_1 + E(C)$ for any $v_1 < t$ and $b^*(v_1) \geq v_1 + E(C)$ for any $v_1 > t$.

Notice that if bidder 0 cannot observe the realized common value $c$ either (so both bidders have equal information regarding $C$), then bidder 1’s equilibrium bid will be $v_1 + E(C)$. Proposition 1 shows that the critical value $t$ is precisely the value which determines whether bidder 1 will bid more conservatively or aggressively, compared to the case where both bidders do not observe $C$.

**Proof of Proposition 1.** We start with a corollary of Lemma 2. In particular, by exchanging the role of $V_0$ and $C$, we have the following conclusion.

**Corollary 1.** If $f_0(v)$ is log-concave, then $E(C|V_0 + C = b) = \frac{n decreasing in b.**

By Theorem 1, we know that the optimal bid $b^*$ for $v_1$ satisfies $v_1 = E(V_0|V_0 + C = b^*) = b^* - E(C|V_0 + C = b^*)$. So $b^* = v_1 + E(C|V_0 + C = b^*)$. Therefore,

$$b^* - v_1 + E(C) = E(C|V_0 + C = b^* - E(C).$$

Observe that $\lim_{v_1 \to 0^+} b^*(l_0 + \epsilon) = l_0 + l_c$, i.e., if bidder 1’s value equals the minimum possible bidder 0 value $l_0$, the $l_0 = E(V_0|V_0 + C = b^*(l_0))$ condition implies that $b^*(l_0) = l_0 + l_c$. Thus $\lim_{v_1 \to 0^+} E(C|V_0 + C = b^*(l_0 + \epsilon)) = l_c < E(C).$ Similarly, $\lim_{v_1 \to 0^+} E(C|V_0 + C = b^*(u_0 - \epsilon)) = u_c > E(C).$ By Corollary 1, if we increase $v_1$ from $l_0$ to $u_0$, the sign of $b^*(v_1) - (v_1 + E(C))$ will change from being negative to being positive. There must exist a critical point $t$ such that $b^*(t) = t + E(C)$, while $b^*(v_1) \leq v_1 + E(C)$ for all $v_1 < t$ and $b^*(v_1) \geq v_1 + E(C)$ for all $v_1 > t$.

**Equilibria Analysis: Many-Bidder Cases**

In this section, we consider the case with $n \geq 2$ uninformed bidders. Like most previous works (Milgrom and Weber 1982; Engelbrecht-Wiggans, Milgrom, and Weber 1983;Arnold, Beck, and Milgrom 2016), we focus on symmetric uninformed bidders — i.e., $V_i$’s are independent and identically distributed (i.i.d.). They are drawn from the same continuous distribution $f(v)$ supported on the entire interval $[l, u]$ for all bidder $i = 1, 2, ..., n$. However, we do allow $f_i \neq f$ to be different. Note that if bidder $i$ ($i > 0$) has a value $v$ less than $l_0$, he would always prefer not to win the auction and thus bid 0. Therefore, we shall w.l.o.g. assume $l \geq l_0$. Also similar to these previous work, we focus on symmetric equilibria among uninformed bidders. By “symmetric equilibrium” we mean the same bidding strategy $b^*(v)$ for any uninformed bidder $i > 0$.

Before characterizing the equilibrium, we first show a useful property of any symmetric equilibrium. That is, the equilibrium bid strictly increases in the private value $v$. Its proof is technical and deferred to the online appendix.

**Lemma 3. (Strict Bid Monotonicity)** Any symmetric equilibrium bid $b^*(v)$ is strictly increasing in the private value $v$. In particular, for any $l_0 < v < u_0$, $b^*(v) < b^*(v')$.

Lemma 3 implies some structural properties of the equilibrium bidding function $b^*(v)$. First, $b^*(v)$ is almost everywhere differentiable, since it is a bounded strictly increasing function. Moreover, the distribution of the bid $b^*(V)$, denoted as $P(b)$, is atomless since $b^*(v)$ is strictly increasing. Next, we characterize the symmetric equilibrium by providing a first-order characterization of the bid distribution $P(b)$. We can easily recover $b^*(v)$ from $P(b)$ since $F(v) = P(b^*(v))$, i.e., $b^*(v) = P^{-1}(F(v))$.

**Theorem 2.** Suppose a symmetric equilibrium exists, then it is the unique symmetric equilibrium. Moreover, the bid distribution $P(b)$ of this symmetric equilibrium is characterized by the following first-order differential equation:

$$P' = \frac{P \int_0^b (F^{-1}(P) - x) f_0(x) g(b - x) dx}{(n - 1) \int_0^b (F^{-1}(P) - x) F_0(x) g(b - x) dx},$$

for any $b \in [\min^*, u_c + u]$ with boundary condition $P(\min^*) = 0$, where $\min^*$ is the maximum $b$ satisfying $b = l + E(C|V_0 + C \leq b)$. A proof sketch of Theorem 2 goes as follows. We start by characterizing the equilibrium utility of any uninformed bidder, as a function of the functional $P(b)$. By definition of equilibrium, we know that the bid $b$ must be the optimal bid for the bidder value $v = F^{-1}(P(b))$ for any $b$, implying that the derivative of the revenue, viewed as a function of $b$ for the corresponding given $v$, must equal 0. This gives rise to a differential equation as stated in Equation (3). The formal proof is technical, and is deferred to the online appendix, in which we also provide a numerical algorithm for computing an approximation of the distribution $P(b)$ based on the finite element method.

In the next section, we shall derive a closed form solution to the above differential equation for uniformly random distributions. Before concluding this section, we briefly mention how Equation (3) is related to our equilibrium characterization for two bidder cases ($n = 1$) as in Theorem 1.

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5Recall that bidder 0 has more accurate estimation of $C$ and always bids truthfully due to Lemma 1.
Since the value \( v \) and the corresponding equilibrium bid \( b \) satisfies \( v = F^{-1}(P(b)) \), when \( n = 1 \), we must have the numerator of the RHS of Equation (3) equal to 0, in order to yield bounded \( P' \). With \( v = F^{-1}(P(b)) \), this exactly implies \( v = \int_{0}^{b} f_{0}(x)g(b-x)\,dx = 0 \), or equivalently,

\[
v = \int_{0}^{b} x f_{0}(x)g(b-x)\,dx = E(V_{0}|V_{0} + C = b)
\]

The above condition is precisely the characterization of Theorem 1.

**Strange Roles of Information Asymmetry**

Armed with the equilibrium characterization in Section and , we are now ready to study how information asymmetry affects bidders’ utilities in a second price auction. We shall compare the following two different situations:

- **Asymmetric World**: only bidder 0 observes realized \( c \) (our current model).
- **Symmetric World**: none of the bidders observe the realized \( c \).

**The Harm of Overtly Possessing Information**

At a first glance, one may think that bidder 0, as the only informed bidder, would always get higher utility in the asymmetric world. Our first surprising finding is that this is not the case. In fact, more accurate value estimation may be harmful for all possible bidder value realizations.

**Proposition 2.** Suppose all bidders know that they are in the asymmetric world. There are instances in which bidder 0 in this asymmetric world has strictly less utility than her utility in the symmetric world for any possible bidder 0 type. That is, overtly possessing information can be harmful to the informed bidder 0.

**Proof.** Our proof constructs such an instance with two bidders, and then utilize our equilibrium characterization in Theorem 1 and Proposition 1. Our main insight is that if the uninformed bidder’s valuation is close to 0 with high probability, then the uninformed bidder will tend to bid aggressively under information asymmetry. Specifically, the critical value in Proposition 1 would be very small. So that with almost probability 1, bidder 1 will over bid compared to the case when bidder 0 is not informed and thus causes utility decrease for bidder 0. Next, we provide a formal construction.

Consider the case where the valuation density of bidder 0 is \( \frac{\lambda c^{\lambda-1}}{\Gamma(\lambda)} \) for \( v_{0} \in [0, 1] \), i.e., a truncated exponential distribution. We will mostly think of \( \lambda \) as large enough. The density function for bonus value \( c \) and bidder 1’s valuation is the same — the uniform distribution on \([0, 1]\). Notice that these two distributions are both log-concave.

By Proposition 1, we know that there exists a critical value \( \hat{v} \), such that the uninformed bidder bids strictly more than \( v_{0} + E(C) = v_{0} + \frac{1}{\sqrt{2}} \). We now show that in the constructed example, \( \hat{v} < \frac{1}{2} \). In fact, as \( \lambda \to \infty \), bidder 1 will always bid \( v_{1} + 1 \).

Specifically, based on Theorem 1, we can derive the relation between value \( v_{1} \) and its corresponding bid \( b_{1} \) for bidder 1, as follows:

\[
v_{1} = \frac{1}{\lambda} - \frac{b_{1}}{e^{\lambda b_{1}} - 1}, \forall b_{1} \in [0, 1] \tag{4}
\]

\[
v_{1} = \frac{1}{\lambda} + \frac{(b_{1} - 1)e^{\lambda(2-b_{1})} - 1}{e^{\lambda(2-b_{1})} - 1}, \forall b_{1} \in [1, 2] \tag{5}
\]

By Equation (4) we know that, for any \( b_{1} \leq 1 \), we must have \( v_{1} < \frac{1}{2} \). Therefore, for any \( v_{1} \geq \frac{1}{2} \), its bid \( b_{1} \) must be decided by Equation (5). As \( \lambda \to \infty \), the right hand side of Equation (5) tends to \( b_{1} - 1 \). This induces that bidder 1 tends to bid \( v_{1} + 1 \) for all \( v_{1} \geq \frac{1}{2} \) as \( \lambda \to \infty \). In addition, the probability that \( v_{1} \leq \frac{1}{2} \) also tends to 0, since \( v_{1} \) is drawn from a uniform distribution over \([0, 1]\). Therefore, as \( \lambda \to \infty \), bidder 1 almost surely bids \( v_{1} + 1 \) and the utility of bidder 0 tends to

\[
\int_{v_{0} + c \geq v_{1} + 1} (v_{0} + c - v_{1} - 1)f(v_{1})g(c)\,dv_{1}\,dc = \frac{1}{6}v_{0}^{3}
\]

Notice that if bidder 0 is not informed, both bidders will “truthfully” bid \( v_{1} + \frac{1}{2} \), in which case bidder 0, with value \( v_{0} \), gets expected utility \( \int_{v_{0} \geq v_{1}} (v_{0} - v_{1})f(v_{1})\,dv_{1} = \frac{1}{2}v_{0}^{2} \), which is greater than \( \frac{1}{6}v_{0}^{3} \) for any \( v_{0} \in [0, 1] \).

Intuitively, the proof of Proposition 2 illustrates that when a “weak” bidder (with generally smaller private value) is publicly known to be informed, a “strong” bidder (with generally larger private value) will tend to bid more aggressively, and this may cause utility loss to the weak bidder. Notably, it is important to assume all bidders know that they are in the asymmetric world, i.e., bidder 0 overty possesses information. A simple observation reveals that if bidder 0 is covertly informed,\(^4\) her utility will always increase. This is because when she is covertly informed, bidding the true value \( v_{0} + c \) dominates the strategy of bidding \( v_{0} + E(C) \), under which she will already get utility that is equal to her utility when she is uninformed. This suggests that any bidder may prefer covertly acquiring information, especially for weak bidders.

**Possible Blessing of Less Information**

Next, we continue to explore uninformed bidders’ utilities. A natural conjecture would be that uninformed bidders may always prefer the symmetric world, i.e., perhaps the asymmetric world always leads to utility decrease for the disadvantaged uninformed bidders.

Unsurprisingly, this can indeed happen. For instance, consider previous two-bidder Example 2 with \( \alpha = 1 \), i.e., \( f_{0}(x) \) and \( g(c) \) are both uniform distribution over \([0, 1]\). We have shown that in the asymmetric world, bidder 1’s bidding function is \( b(v_{1}) = 2v_{1} \) for all \( v_{1} \in [0, 1] \). Basic algebraic cal-

\(^4\)More precisely, all bidders thought that they are in the symmetric world but bidder 0 covertly gets information to make it actually the asymmetric world.
cularion yields that in the asymmetric world, bidder 1’s expected utility is as follows:

\[
\text{ExpU}_i = \begin{cases} \frac{2}{3}v_1^3 & v_1 \in [0, 0.5]; \\ -\frac{2}{3}v_1^3 + 2v_1^2 - v_1 + \frac{1}{6} & v_1 \in [0.5, 1]. \end{cases}
\]

On the other hand, in the symmetric world, it is straightforward to verify that bidder 1’s equilibrium bid is \(v_1 + 0.5\), resulting in expected utility \(\frac{5}{36}\), which turns out to be larger than the above utility in the asymmetric world for any \(v_1\).

What is surprising, however, is that the above “disadvantaged” situation for uninformed bidders is not always harmful. There are situations where the uninformed bidders can do better in the asymmetric world.

**Proposition 3.** There are instances in which an uninformed bidder of certain type derives strictly more utility in the asymmetric world than her utility in the corresponding symmetric world.

**Proof.** We consider the setting with \(n + 1\) bidders and all distributions \((f_i, g)\) are the uniform distribution over \([0, 1]\). Due to valuation symmetry among uninformed bidders, let \(i(>0)\) denote a generic bidder of our interest and \(v\) denote \(i\)’s private value.

We start by analyzing the simpler case, i.e., the symmetric world. In this case, bidder 0 does not observe \(c\); it is easy to see that the bidder \(i\) bids \(v + \frac{1}{2}\) at the unique dominant-strategy equilibrium. Therefore, the expected utility for bidder \(i\) with basic value \(v\) is \(f_0(v) = \int v \, dx = \frac{1}{n+1}v^{n+1}\).

More intricate is the equilibrium analysis in the asymmetric world in which bidder 0 observes \(c\). We will use the equilibrium characterization in Theorem 2, and first pin down the boundary condition. Since \(l = 0\), it is easy to see that the only \(b\) that satisfies \(b = E(C|C + V_0 \leq b)\) is \(b = 0\). Therefore, the bid for the smallest value of bidder \(i\) is 0.

Next, we analyze bidder \(i\)’s bidding function \(b(v)\). Since \(V \sim U[0, 1]\), \(b\) has a corresponding distribution, with randomness inherited from \(V\), with CDF \(P(b)\). Equation (3) is a differential equation that characterizes \(P(b)\). We now solve for \(P(b)\) for the special case that \(f_i, g\) are all uniform distribution over \([0, 1]\). Note that \(P(b) = F_i(v) = v\), we thus have

\[
P_i(b) = \frac{F_0^{-1}(P(b))}{P(b)} = \frac{F_0^{-1}(F_i(v))}{P(b)} = \frac{F_0^{-1}(v)}{P(b)} = v = P(b)
\]

Plugging the above equality into Equation (3), and utilizing the fact that \(f_0, g\) are both uniform distribution over \([0, 1]\), we have

\[
P_i(b) = -\frac{P(b)[P(b)b - b^2/2]}{(n-1)[P(b)s(b) - t(b)]}
\]

with boundary condition \(P(0) = 0\), where function \(s(b) = \int_0^b F_0(x)g(b - x)dx\) is

\[
s(b) = \begin{cases} \frac{1}{2}b^2 & b \in [0, 1]; \\ -\frac{1}{2}b^2 + 2b - 1 & b > 1 \end{cases}
\]

and function \(t(b) = \int_0^b xF_0(x)g(b - x)dx\) is

\[
t(b) = \begin{cases} \frac{1}{b^3}b^3 & b \in [0, 1]; \\ \frac{1}{6}(-2b^3 + 9b^2 - 6b + 1) & b > 1 \end{cases}
\]

It turns out that when \(b \leq 1\), differential inequality has a closed form unique solution \(P(b) = \frac{2n+1}{3n+3}b\), which can be verified simply by plugging it into the function. When \(b > 1\), it appears intractable to have a closed form solution due to much more intricate format of function \(s(b), t(b)\). However, our argument next will only need to focus on \(b \leq 1\) or equivalently, \(v \leq \frac{2n+1}{3n+3}\) for the generic bidder \(i\).

The above derivation shows that at the symmetric equilibrium the uninformed bidder \(i\)’s bidding strategy is \(\frac{2n+1}{3n+3}v\) for private value \(v\) satisfying \(v \leq \frac{2n+1}{3n+3}\) (i.e., the bid is at most 1). We are now ready to compute bidder \(i\)’s utility at such a private value \(v\). For convenience, let \(\alpha = \frac{3n+1}{2n+1}\), so \(b(v) = \alpha v\). We then have

\[
U(v) = \int_{v \geq u, \alpha v \geq v_0 + c} (v + c - \max\{v_0 + c, \alpha u\}) \times (n - 1)u^{-1} du dv_0 dc
\]

\[
= \int_{v \geq u \geq \frac{1}{c}(v_0 + c)} (v + c - \alpha u)(n - 1)u^{-1} du dv_0 dc
\]

\[
= \int_{v \geq u \geq \frac{1}{c}(v_0 + c)} (v - v_0)(n - 1)u^{-1} du dv_0 dc
\]

\[
= \int_{v \geq u \geq \frac{1}{c}(v_0 + c)} (v + c - \alpha u)(n - 1)u^{-2} dc dv_0
\]

\[
= \int_{v \geq u \geq \frac{1}{c}(v_0 + c)} (v - v_0)\left(\frac{v_0 + c}{\alpha}\right)^{-1} dc dv_0
\]

\[
= \int_{v \geq u \geq v_0} (v - v_0)\left(\frac{v_0 + c}{\alpha}\right)^{-1} \left(\frac{v_0 + c}{\alpha}\right)^{-1} v_0^{-1} dv_0
\]

\[
= \int_{v \geq u \geq v_0} \left[\frac{\alpha^2(n - 1)}{2(n + 1)} - \frac{\alpha^3(n - 1)}{3(n + 2)}\right] v_0^{n+1}
\]

\[
+ \frac{\alpha v_0^{n+1}}{n} \left[\alpha - \frac{\alpha^2}{2} - \frac{\alpha}{n + 1} + \frac{\alpha^2}{n + 2}\right]
\]

We now compare \(U(v)\) with the bidder’s utility \(\frac{1}{n+1}v^{n+1}\) in the symmetric world at \(v = \frac{1}{2}\). Considering sufficiently large \(n\) and omitting \(o(1)\) terms, we have

\[
U(v) - \frac{1}{n+1}v^{n+1}
\]

\[
= \frac{\alpha^2(n - 1)}{2(n + 1)} - \frac{\alpha^3(n - 1)}{3(n + 2)} + \frac{\alpha^2 - \alpha^3/2}{n} - O\left(\frac{1}{n^2}\right) - \frac{1}{n + 1}
\]

\[
= \alpha^2\left[\frac{1}{2n} + \frac{\alpha^2 - \alpha^3/2}{n}\right] - O\left(\frac{1}{n^2}\right) - \frac{1}{n + 1}
\]

When \(n\) is large, \(\alpha \approx \frac{3}{7}\), so the above quantity tends to \(\frac{10}{16n} > 0\), omitting \(O\left(\frac{1}{n^2}\right)\) term. Therefore, \(U(v) > \)
In other words, an uninformed bidder with value $v = 1/\alpha$ gets strictly larger utility in the asymmetric world than her utility in the symmetric world.

Our next result shows that the above phenomenon may be due to the intense competition among many bidders, under which: (1) each bidder derives little utility in the symmetric world; whereas (2) more conservative bidding among uninformed bidders in the asymmetric world helps them to increase utility when their private values are high. It turns out that with two bidders, the uninformed bidder will always be worse off in the asymmetric world.

**Proposition 4.** When there are two bidders (one informed and one uninformed), the utility of the uninformed bidder in the asymmetric world is always less than her utility in the symmetric world, for any private value $v_1$.

**Proof.** We have

\[
U(b) = \mathbb{E}(v_1 + C - [V_0 + C]|V_0 + C \leq b) \times P(V_0 + C \leq b)
\]

\[
= \int_0^b (v_1 - v_0)f(v_0)G(b - v_0)dv_0
\]

\[
= \int_0^{v_1} (v_1 - v_0)f(v_0)G(b - v_0)dv_0 + \int_{v_1}^{b} (v_1 - v_0)f(v_0)G(b - v_0)dv_0
\]

\[
\leq \int_0^{v_1} (v_1 - v_0)f(v_0)dv_0 + 0
\]

The last term $\int_0^{v_1} (v_1 - v_0)f(v_0)dv_0$ is precisely bidder 1’s utility in the symmetric world.

**Welfare, Revenue, and Failure of the Linkage Principle**

Finally, we make a few simple observations about the effect of information asymmetry on revenue and welfare.

**Observation 1.** The auction’s welfare in the asymmetric world is always weakly less than that in the symmetric world.

**Proof.** This is because if bidder 0 was not informed as in the symmetric world, the second price auction is socially optimal — it always allocates the item to the bidder with higher private values, which is also the bidder with the highest total value since the common component is shared among bidders.

**Example 3** (Winner’s curse and revenue collapse). Consider the degenerate case where $V_i \equiv 0$ and bidders have only the common value $C$, drawn from the uniform distribution on $[0,1]$. In the asymmetric world, bidder 0 always bids his true value $c$ whereas every other bidder’s equilibrium bid is always 0. This is due to the winner’s curse — had any bidder ($i > 0$) won, she must have bid higher than bidder 0’s bid which is the true value $c$ for everyone. The revenue in this asymmetric world is 0, while the revenue in the symmetric world is $\mathbb{E}(C) = 0.5$.

In auction theory, a seminal result of (Milgrom and Weber 1982) is the well-known **linkage principle**. That is, when bidders are symmetric and values are associated, revealing private information of the auctioneer (if she has any) always increases revenue in most commonly seen auctions, including the second-price auction. In other words, transparency always favors revenue.

Recent result by (Syrgkanis, Kempe, and Tardos 2015) shows that this principle fails in setting with two asymmetrically informed bidders, however their proof is involved due to intricate equilibrium analysis under asymmetric information. Next, we use a much simpler example in our setting to give another piece of evidence for the failure of the linkage principle, also with two asymmetrically informed bidders. The key intuition is that the uninformed bidder tends to bid aggressively, which leads to higher revenue for the auctioneer, compared to the situation in which bidder 1 knows exactly the realized $c$. Therefore, the auctioneer may prefers (some) less informed bidders.

**Example 2 Cont’d.** (Failure of the Linkage Principle) Recall the Example 2 setup with two bidders. Now consider $f_1$ as a point-distribution with all point mass at $v_1 = 0.8\alpha$. Bidder 1’s equilibrium bid is (deterministically) $b_1 = 2v_1 + 1 - \alpha = 1 + 0.6\alpha$. Given full information, bidder 0’s dominant strategy is to bid true value $b_0 = v_0 + c$. Note that $\Pr(V_0 + C \geq 1 + 0.6\alpha) = \Pr(1 - C \leq V_0 - 0.6\alpha) = \int_{0.6\alpha}^{\infty} (v_0 - 0.6\alpha) dv_0 = 0.08\alpha$. Therefore,

\[
\Pr(1 - C \leq V_0 - 0.6\alpha) = 0.08\alpha
\]

\[
\Pr(V_0 + C \geq 1 + 0.6\alpha) = \Pr(V_0 + C \geq 1 + 0.6\alpha) = \int_0^{0.6\alpha} (v_0 - 0.6\alpha) dv_0 = 0.08\alpha
\]

Now suppose the auctioneer has private information and can observe realized $c$. If the auctioneer reveals $c$ to bidder 1, then both bidders will bid truthfully, leading to revenue

\[
\mathbb{E}_{v_0,c} \min\{V_0 + C, 0.8\alpha + C\} = \mathbb{E}_{v_0} \min\{V_0, 0.8\alpha\} + \mathbb{E}[C],
\]

which equals 0.5 + 0.48\alpha after basic calculation. Therefore, for any $\alpha \leq 1$, the revenue under no information is larger than that under full information, leading to the failure of the linkage principle.

**Conclusion**

In this paper, we study how information asymmetry affects the equilibrium of a second price auction. The key message revealed from our results is that in strategic interactions as the second-price auction, competition among bidders makes the role of asymmetric information intricate. For instance, less information could be beneficial whereas more information could be harmful. Our work thus leads to new insights about bidders’ behaviors, and shed light on the situations under which bidders may or may not want to obtain fine-grained information about the auctioned item. This leaves many open directions for future research, e.g., understanding the role of information asymmetry in other auction models with different bidder value models and more generally, in other strategic games.
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