

Spite vs. Risk: Explaining overbidding^{*}

A theoretical and experimental investigation

Oliver Kirchkamp[†] Wladislaw Mill[‡]

October 22, 2018

In this paper we use an experiment to compare a theory of risk aversion and a theory of spite as an explanation for overbidding in auctions. As a workhorse we use the second-price all-pay and the first-price winner-pay auction. Both risk and spite can be used to rationalize deviations from risk neutral equilibrium bids in auctions.

We exploit that equilibrium predictions in the second-price all-pay auctions for spiteful preferences are different than those for risk averse preferences.

Indeed, we find that spite is a more convincing explanation for bidding behavior in particular for the second-price all-pay auction. Not only can spite rationalize observed bids, also our measure for spite is consistent with observed bids.

Keywords: Auction, Overbidding, Spite, Risk, Experiment

JEL: C91; C72; D44; D91

^{*}We would like to thank the Max Planck Society for financial support through the International Max Planck Research School on Adapting Behavior in a Fundamentally Uncertain World. We are grateful for comments by Nick Netzer, Armin Schmutzler, Maarten Janssen, Juuso Välimäki, Igor Letina, Kremena Valkanova, Ulrich Bergmann, Julia Grünseis, Christoph Engel, Karl Schlag, Simeon Schudy, Nicolas Fugger, seminar audience at the University Zurich, University Maastricht and participants at ESA World Meeting 2016, EEA 2016 and the IMPRS Thesis Workshop 2015. We use R (2017) for the statistical analysis. Data, methods and video instructions can be found at <https://www.kirchkamp.de/research/SpiteVsRisk.html>.

[†]University Jena, School of Economics, Carl-Zeiss-Str. 3, 07743 Jena, oliver@kirchkamp.de, Telefon +49 3641 943240, fax +49 3641 943242.

[‡]University of Mannheim, Department of Economics, L7 3-5, 68131 Mannheim, Germany. Email: mill@uni-mannheim.de, Telefon +49 621 181-1897. This reserach was mainly developed as a member of the IMPRS on Adapting Behavior in a Fundamentally Uncertain World in Jena.

1. Introduction

In this paper we compare spite and risk as possible motives for deviations from risk neutral Bayesian Nash equilibria (RNBNE). We use the second-price all-pay auction as a device to compare risk aversion and spite as motives which determine bids in auctions. We do the same with the first-price winner-pay auction. Using an experiment we find that spite explains bidding behavior in the second-price all-pay auction better than risk while risk seems to be the better predictor in the first-price winner-pay auction.

Auctions are a relevant part of everyday life. Auctions are commonly used as selling mechanisms for example in online auctions (like ebay), government auctions (like spectrum auctions) and at charity events (like silent auctions). In fact we find explicit auction institutions in many markets. Moreover, (all-pay) auctions are also a good model of non-market interaction. For example, fights between animals (Riley, 1980; Smith, 1974)¹, competition between firms (Fudenberg and Tirole, 1986; Ghemawat and Nalebuff, 1985; Oprea et al., 2013), the voluntary provision of public goods (Bilodeau et al., 2004), legal expenditures in litigation environments (Baye et al., 2005), the settlement of strikes, fiscal and political stabilization, the timing of exploratory oil drilling, and many more (see Hörisch and Kirchkamp, 2010, p. 1) are applications of (all-pay-)auctions.

While risk neutral Bayesian Nash equilibria (RNBNE) can be derived for many auction types, empirical and laboratory evidence suggests that auction participants do not always bid according to RNBNE. Many authors find overbidding (or over-dissipation). Often bids are higher than equilibrium bids in all-pay auctions,² in rent-seeking contests,³ and also in winner-pay auctions.⁴ Given the fast use of auctions and the model thereof, it is important to thoroughly understand the reasons and motives for the deviations from RNBNE.

Several authors propose explanations why bids might deviate from RNBNE.⁵

However, there is no agreement in the literature on which explanations are performing best. Several explanations, like risk aversion, joy of winning, anticipated regret etc., have been shown to be of limited usefulness.⁶ The most common explanation is arguably risk aversion. A very recent one, however, is the spite motive.

In this paper we suggest that spite might be, in some situations, a better predictor for overbidding than risk aversion. We study this question using equilibrium analyses and empirical evidence from a conducted experiment.

As a workhorse we use the second-price all-pay auction and first-price winner-pay auction. For both auction types spite leads to an increase in equilibrium bids as long as valuations are not too high. Risk aversion, however, leads to a decrease in bids in the second-price all-pay

¹Smith (1974) uses a war-of-attrition game with common valuations to model fights between animals.

²See Noussair and Silver (2006); Ernst and Thöni (2013); Goeree et al. (2002); Chen et al. (2015); Lugovskyy et al. (2010).

³Potters et al. (1998).

⁴Morgan et al. (2003); Andreoni et al. (2007); Barut et al. (2002).

⁵Filiz-Ozbay and Ozbay (2007, 2010); Cooper and Fang (2008); Andreoni et al. (2007); Cox et al. (1985, 1988); Fibich et al. (2006); Kagel and Levin (1993); Kirchkamp and Reiss (2008); Engelbrecht-Wiggans and Katok (2009); Kirchkamp et al. (2008); Armantier and Treich (2009).

⁶Kagel and Levin (1993); Kirchkamp et al. (2008); Engelbrecht-Wiggans and Katok (2009); Andreoni et al. (2007); Katuscak et al. (2013).

auction and to an increase in bids in the first-price winner-pay auction.

In our experiment, we measure spitefulness, risk, and bids. We find that spite explains bidding behavior better than risk in the second-price all-pay auction. Risk seems to be the better predictor in the first-price winner-pay auction.

We want to make three contributions:

1. We extend the theoretical model of spiteful behavior and risk averse behavior to second-price all-pay auctions.
2. With the help of experiments we investigate bidding behavior in the second-price all-pay auction and in the first-price winner-pay auction.
3. Most importantly, we find that spite is an important determinant of overbidding (relative to RNBNE) in the second-price all-pay auction, but less so in the first-price winner-pay auction.

The remainder of the paper is structured as follows: We briefly summarize the relevant literature in Section 2. In Section 3 we present the model and the theoretical predictions. Section 4 will explain the design of the experiment. In Section 5 we show the results of the experiment. Section 6 concludes.

2. Literature

In this paper we study second-price all-pay auctions and first-price winner-pay auctions with sealed-bids and private information.⁷ We restrict our attention to auctions where the highest bidder wins.⁸ We also assume that the number of bidders is known.⁹

2.1. Literature on overbidding

In many experiments, overbidding (relative to RNBNE) has been observed and explained with the help of a number of motives, ranging from risk aversion, over spite, to anticipated regret. Obviously, we cannot do right by the vast literature on overbidding. Nevertheless, we will present a few selected findings from this literature.

⁷Equilibria for all-pay auctions with common values are provided by Hendricks et al. (1988) and Kovenock et al. (1996). Sacco and Schmutzler (2008) provide mixed strategy equilibria for common value auctions where the prize is influenced by the own behavior. Feess et al. (2008) show a pure equilibrium strategy in case of handicapped players. Klose and Kovenock (2015) show equilibria for the case of externalities which depend on the bidders' identities. Bertoletti (2016) show equilibria for common value all-pay auctions with reserve price. Dechenaux and Mancini (2008) and Baye et al. (2005) model litigation systems with all-pay auctions. The case of affiliated valuations is studied by Krishna and Morgan (1997).

Intermediate situations between the first-price and second-price all-pay auction are studied by Albano (2001).

⁸The survey by Dechenaux et al. (2015) includes rent-seeking games where the ex-post allocation is stochastic and where also bidders who did not submit the highest bid have a chance to win the auction.

⁹Bos (2012) considers the situation where the number of bidders is unknown.

Three particularly important motives to explain overbidding are: risk aversion, anticipated regret and joy of winning. Risk aversion has been suggested by Cox et al. (1985, 1988) as an explanation of overbidding. In the context of all-pay auctions Filiz-Ozbay et al. (2006) study risk averse players to explain overbidding. However, Kagel and Levin (1993), Kirchkamp et al. (2008), and Engelbrecht-Wiggans and Katok (2009) argue that risk aversion might be by itself not enough to explain overbidding. Kagel and Levin (1993) point out that risk aversion does not explain very well bidding behavior in third-price auctions. In equilibrium risk-averse bidders should bid less than the RNBNE. Bidders in their experiment, however, bid more.

Anticipated regret is another motive to explain overbidding in winner-pay auctions. Filiz-Ozbay and Ozbay (2007, 2010) propose that players anticipate their regret after a wrong choice. Katuscak et al. (2013) do not replicate this finding with a large sample and thus argue against anticipated regret.

Joy of winning is another explanation for overbidding, suggested by Cooper and Fang (2008). Andreoni et al. (2007) provide evidence against joy of winning.

A large number of other factors, internal and external to the bidders, have been studied. Among the external factors, Katok and Kwasnica (2008) demonstrate that the speed of the ticking clock in Dutch auctions affects bids. Cox and James (2012) argue that the structure of the presented games have an influence on the bidding behavior of the participants. Kirchkamp et al. (2009) show that outside options also influence bids.

Among the factors internal to bidders Nakajima (2011) explains the deviation of the Dutch auction from the first-price winner-pay auction with the Allais paradox. Güth et al. (2003), Dittrich et al. (2012), and Ockenfels and Selten (2005) relate bidding behavior to learning. Kagel et al. (1987) and Hyndman et al. (2012) show that the provision of information influences overbidding. Anderson et al. (1998) introduce bounded rationality as a reason for overbidding in all-pay auctions. Similarly, Kirchkamp and Reiss (2008) claim that overbidding may result from bidding heuristics. Armantier and Treich (2009) argue that bidders are unable to assess winning probabilities correctly. Chen et al. (2013) provide empirical evidence that the menstrual cycle has an influence on the bidder's decision.

2.2. Literature on spite

In addition to the above-discussed explanations for overbidding, another important motive is spite. Andreoni et al. (2007) suggest that spite may cause overbidding. Bartling and Netzer (2016, p.23) propose that "spiteful preferences are an important determinant of overbidding in the second-price auction". Several recent papers study the impact of spite on equilibrium bids. Morgan et al. (2003) may have been the first to consider spite in the equilibrium for winner-pay auctions. Similarly, Brandt et al. (2007); Sandholm and Tang (2012); Sandholm and Sharma (2010); Mill (2017) study equilibrium bids with spiteful preferences for winner-pay auctions. Nishimura et al. (2011) study spite in common-valuations-auctions and, most recently, Bartling et al. (2017) consider equilibria where bidders could also have spiteful preferences towards the auctioneer.

Not only could we see spite as a convenient explanation of overbidding; several studies show that, indeed, spite is not uncommon in general. Saijo and Nakamura (1995) find spiteful

behavior in Voluntary Contribution Mechanisms.¹⁰ Fehr et al. (2008) use experiments to show that spiteful behavior is rather wide spread in the least developed parts of India. Kimbrough and Reiss (2012) and Bartling et al. (2017) study spite in auctions. Kimbrough and Reiss (2012) observe consistent spiteful behavior in a second-price winner-pay auction. Bartling et al. (2017) show that subjects' bidding behavior is not driven by spite towards the auctioneer.

To the best of our knowledge, no paper studies spite in all-pay auctions. More importantly, no paper has measured spite and combined a theory of spiteful bidding with actually spiteful behavior.

In the next section, we will determine equilibrium bids with spite and risk in the context of the second-price all-pay auction and the first-price winner-pay auction.

3. Model

In the following, we will derive the Bayesian Nash equilibrium for spiteful bidders and for risk averse bidders in the second-price all-pay auction and the first-price winner-pay auction.¹¹

3.1. Second-price all-pay auction

3.1.1. Spite in the second-price all-pay auction

Consider a situation with one prize and two risk neutral bidders, $k \in \{i, j\}$. Bidders have a utility function $u(x)$ and private valuations v_k . Valuations follow a distribution function F with density function f , i.e. $v \sim F(0, \bar{v})$, and $f(x) = dF(x)/dx$. Each bidder k submits a bid b_k following a monotonic bidding function $b_k = \beta_k(v_k)$. Consider the case $b_j \geq b_i$. In the second-price all-pay auction both players pay the second highest bid (b_i). The prize is allocated to the bidder with the highest bid. If $b_i = b_j$, the prize is distributed randomly.

For the candidate equilibrium we assume $\beta_k(0) = 0$.¹² Furthermore, we assume that the first derivative $\beta'_k(x) = d\beta_k(x)/dx$ and the inverse $\beta_k^{-1}(b_k) = v_k$ exist. The payoff of the winning bidder j is $(v_j - b_i)$. The payoff of the losing bidder i is $-b_i$.

In line with the literature on spite in auctions¹³ we assume that a spiteful loser i experiences a disutility $\alpha \cdot (v_j - b_i)$ where α describes the amount of spite. A non-spiteful bidder is characterized by $\alpha = 0$. Here we assume that $\alpha \in [0, 1)$. We do not consider $\alpha < 0$ which could represent sympathy or profit sharing. We also rule out $\alpha > 1$, i.e. that an other bidder's gain is more important than the own loss. For simplicity, we assume that spite only affects

¹⁰Cason et al. (2002) show that this pattern did not prevail in the U.S.

¹¹The risk neutral Bayesian Nash equilibrium for spiteful bidders in the first-price all-pay auction is shown in Appendix A.

¹²We assume a monotonic bidding function. A selfish agent would only win if the opponent has also a valuation of zero and there is no benefit to bidding anything above 0. For a spiteful subject, it might make sense to bid above zero if the bid would be costless (standard second-price winner-pay auction) as this spiteful subject could reduce the payoff of the opponent by this increased bid. However, in the all-pay case, one could never offset the downside of paying for the own bid by making the opponent bid more as long as $\alpha \leq 1$. Hence, zero is the best choice.

¹³See Bartling et al. (2017); Morgan et al. (2003); Brandt et al. (2007); Sandholm and Tang (2012); Sandholm and Sharma (2010); Mill (2017).

the loser of the auction. Furthermore, we assume that α is constant and independent of the valuation.

We call $\Phi_{\text{Spite}}^{\text{II-AP}}(b_i, v_i)$ the payoff of player i :

$$\Phi_{\text{Spite}}^{\text{II-AP}}(b_i, v_i) = \begin{cases} u(v_i - b_j) & \text{if } b_i > b_j \text{ (i wins)} \\ u\left(\frac{v_i - b_i}{2}\right) & \text{if } b_i = b_j \text{ (a tie)} \\ u(-b_i - \alpha(v_j - b_i)) & \text{if } b_i < b_j \text{ (j wins)} \end{cases} \quad (1)$$

We assume that bidder i with valuation v makes a bid b . The opponent, bidder j with valuation v_j , makes a bid $b_j = \beta_j(v_j)$. The expected utility of a spiteful bidder i is given as follows:

$$\mathbb{E}(b, v) = \underbrace{\int_0^{\beta_j^{-1}(b)} u(v - \beta_j(v_j)) f(v_j) dv_j}_{\text{bidder i wins and obtains the prize and pays the loser's bid}} + \underbrace{\int_{\beta_j^{-1}(b)}^{\bar{v}} u(-b - \alpha(v_j - b)) f(v_j) dv_j}_{\text{bidder i loses and pays the own bid and additionally experiences spite}} \quad (2)$$

Rearranging the FOC yields:

$$\beta_j'(\beta_j^{-1}(b)) = \frac{(u(v - b) - u(-b - \alpha(\beta_j^{-1}(b) - b))) f(\beta_j^{-1}(b))}{(1 - \alpha) \int_{\beta_j^{-1}(b)}^{\bar{v}} u(-b - \alpha(v_j - b))' f(v_j) dv_j}$$

For the symmetric equilibrium and risk neutrality we obtain

$$\beta_j'(v) = \frac{v + \alpha(v - b) f(v)}{(1 - \alpha)(1 - F(v))} = \frac{v(1 + \alpha) f(v)}{(1 - \alpha)(1 - F(v))} - \frac{\alpha(b) f(v)}{(1 - \alpha)(1 - F(v))}. \quad (3)$$

Solving the differential Equation (3) with initial value $b(0) = 0$ gives us the symmetric equilibrium bidding function $b_{\text{Spite}}^{\text{II-AP}}$:

$$b_{\text{Spite}}^{\text{II-AP}}(v) = \frac{\alpha + 1}{1 - \alpha} (1 - F(v))^{\frac{\alpha}{1 - \alpha}} \int_0^v s f(s) (1 - F(s))^{\frac{1}{\alpha - 1}} ds = \frac{\alpha + 1}{\alpha} \left(v - \frac{\int_0^v (1 - F(s))^{\frac{\alpha}{\alpha - 1}} ds}{(1 - F(v))^{\frac{\alpha}{\alpha - 1}}} \right) \quad (4)$$

For $\alpha = 0$, Equation (4) becomes the familiar equilibrium bidding function for second-price all-pay auctions without spite:

$$b^{\text{II-AP}} := b_{\alpha=0}^{\text{II-AP}} = \int_0^v s f(s) (1 - F(s))^{-1} ds$$

For uniformly distributed valuations, $F(x) = x$, we have the following equilibrium bid:

$$b_{\text{Spite}}^{\text{II-AP}}(v) = \frac{(\alpha + 1)}{\alpha(2\alpha - 1)} \left((1 - \alpha) \left((1 - v)^{\frac{\alpha}{1 - \alpha}} - 1 \right) + v\alpha \right) \quad (5)$$

From Equation (5) we have $\lim_{\alpha \rightarrow 0} b_{\text{Spite}}^{\text{II-AP}}(v) = -\log(1 - v) - v$ and $\lim_{\alpha \rightarrow 1} b_{\text{Spite}}^{\text{II-AP}}(v) = 2v$. Figure 1 illustrates the case of uniform valuations. The left graph in the Figure shows that bids are monotonically increasing in valuations. To simplify the notation we assume in the following that valuations $v \in [0, 1]$. It is easy to see the following:

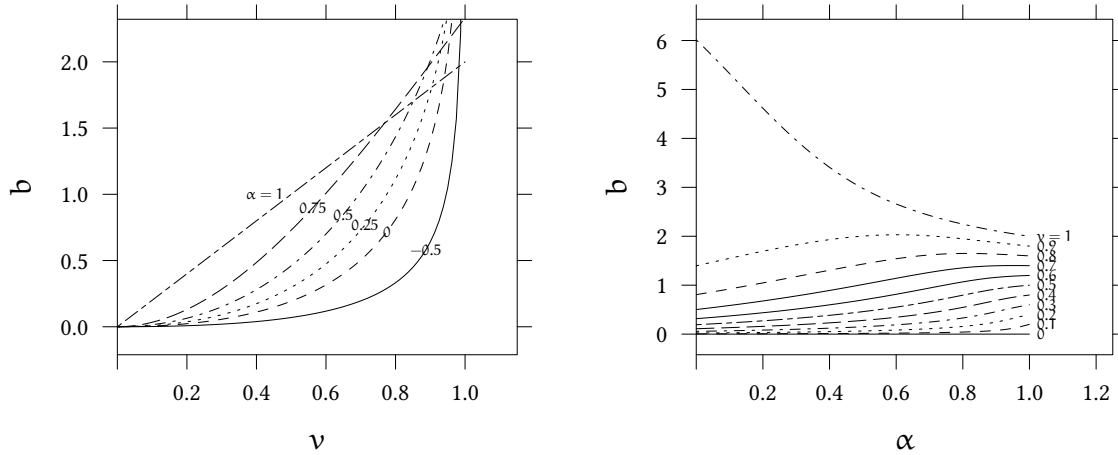


Figure 1: Equilibrium bids in second-price all-pay auctions for spiteful bidders. Equilibrium bids in second-price all-pay auctions for different valuations v and different levels of spite α for uniformly distributed valuations (see Equation (5)).

Proposition 1. *The bidding function in the second-price all-pay auction is increasing in bidder's valuation:*

$$\frac{db_{\text{Spite}}^{\text{II-AP}}}{dv} \geq 0$$

The proof is shown in Appendix B. The right part of Figure 1 shows that bids are increasing in spite if valuations are sufficiently small. For large valuations, equilibrium bids decrease when spite increases. Looking again at Equation (5) we find the following:

Observation 1. *For the case of uniformly distributed valuations in the second-price all-pay auction bids increase in spite for low valuations and they decrease in spite for high valuations.*

Things are different in the first-price all-pay auction. In Proposition 8 in Appendix A we show that in first-price all-pay auctions bids are always increasing in spite.

3.1.2. Risk aversion in the second-price all-pay auction

To compare spite with risk aversion we will derive the equilibrium bidding function for risk averse subjects. We assume that the risk preferences can be described as constant absolute risk aversion (CARA).¹⁴ Again we assume two players $k \in \{i, j\}$ who are competing for an object which each player values with $v_k \in [0, 1]$. Valuation are drawn from a distribution with density function $F(v)$ and distribution function $f(v)$. Both bidders use a bidding function $\beta_k(v_k)$. Both players have the same utility function $u(x) = -r e^{(-x/r)}$. Here we rule out spite, i.e. $\alpha = 0$. As above we assume that bidder i with valuation v makes a bid b . The opponent,

¹⁴We use CARA and not CRRA since in all-pay auctions subjects will experience negative payoffs. Hence, CRRA would imply complex utilities, which is difficult to interpret.

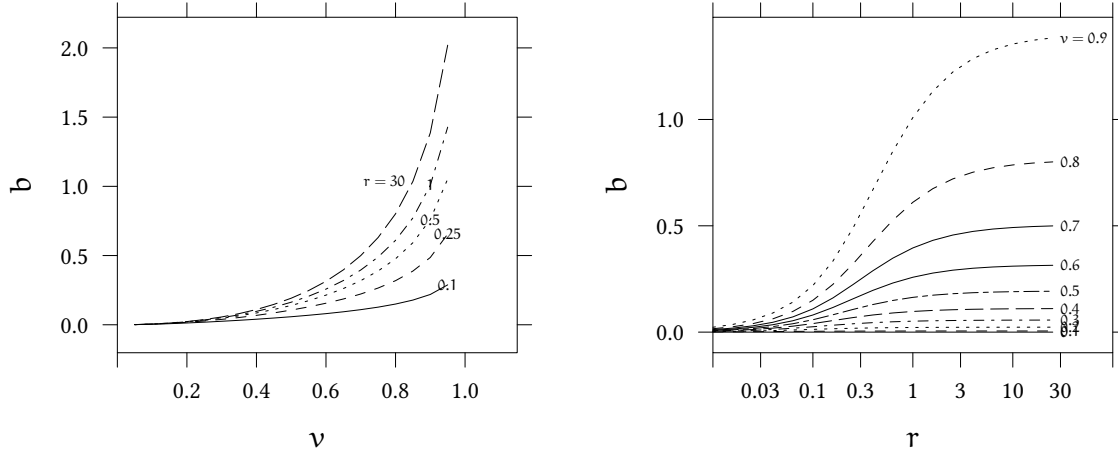


Figure 2: Equilibrium bids in second-price all-pay auctions for risk averse bidders. Equilibrium bids in second-price all-pay auctions for different valuations v and different levels of risk r with uniform distributions of valuations (see Equation (7)). Increasing r indicates decreasing risk aversion (for $r = \infty$ we would have risk neutrality).

bidder j with valuation v_j , makes a bid $b_j = \beta_j(v_j)$. The expected utility of a risk-averse bidder i in the second-price all-pay auction is given by the following equation:

$$\mathbb{E}(b, v) = \underbrace{\int_0^{\beta_j^{-1}(b)} u(v - \beta_j(v_j)) f(v_j) dv_j}_{\text{bidder } i \text{ wins and obtains the prize and pays the loser's bid}} + \underbrace{\int_{\beta_j^{-1}(b)}^{\bar{v}} u(-b) f(v_j) dv_j}_{\text{bidder } i \text{ loses and pays the own bid}} \quad (6)$$

Rearranging the FOC yields:

$$\beta_j'(\beta_j^{-1}(b)) = \frac{(u(v - b) - u(-b)) f(\beta_j^{-1}(b))}{\int_{\beta_j^{-1}(b)}^{\bar{v}} u(-b)' f(v_j) dv_j} = \frac{\left(-e^{-\frac{b-v}{r}} + e^{-\frac{b}{r}}\right) r f(\beta_j^{-1}(b))}{\int_{\beta_j^{-1}(b)}^{\bar{v}} e^{-\frac{b}{r}} f(v_j) dv_j}$$

Assuming symmetry, i.e. $\beta_j^{-1}(b) = v$, we get:

$$\beta_j'(v) = \frac{r \cdot e^{-\frac{b}{r}} (1 - e^{-\frac{v}{r}}) f(v)}{e^{-\frac{b}{r}} (1 - F(v))}$$

Hence the equilibrium bid is as follows:

$$\beta_{\text{Risk}}^{\text{II-AP}}(v) = \int_0^v \frac{r(1 - e^{-\frac{s}{r}}) f(s)}{(1 - F(s))} ds \quad (7)$$

Figure 2 illustrates the case of uniformly distributed valuations. From Equation (7) we can conclude the following:

Proposition 2. *The equilibrium bid of a risk averse bidder is smaller than the bid of a risk neutral bidder:*

$$\beta_{\text{Risk}}^{\text{II-AP}}(v) \leq \beta_{\text{RNBNE}}^{\text{II-AP}}(v)$$

The proof of Proposition 2 is shown in Appendix B.

3.2. First-price winner-pay auction

3.2.1. Spite in the first-price winner-pay auction

Morgan et al. (2003) introduced spite to the first-price and second-price winner-pay auction. The equilibrium bid in the first-price winner-pay auction for the two-player case is given by:

$$b_{\text{Spite}}^{\text{I}}(v) = v - \int_0^v \frac{F(t)^{1+\alpha} dt}{F(v)^{1+\alpha}} \quad (8)$$

Morgan et al. (2003) point out that more spiteful bidders have a steeper equilibrium bidding function in valuations. This can easily be seen for the case of a uniform distribution where (8) implies that $b_{\text{Spite}}^{\text{I}}(v) = (1 + \alpha)/(2 + \alpha)v$. Figure 3 shows equilibrium bids in first-price winner-pay auctions for different valuations v and different levels of spite α with uniform distributions of valuations.

Morgan et al. (2003) also show that spiteful subjects overbid in equilibrium in the first-price winner-pay auction.

Proposition 3 (Morgan et al. (2003)). *A spiteful subject bids more than a risk neutral selfish subject in the first-price winner-pay auction:*

$$\beta_{\text{S}}^{\text{I}}(v) \geq \beta_{\text{RNBNE}}^{\text{I}}(v)$$

3.2.2. Risk in the first-price winner-pay auction

Riley and Samuelson (1981) and Maskin and Riley (1984) show that in first-price winner-pay auctions risk averse bidders bid more than risk neutral bidders (see Riley and Samuelson, 1981, Proposition 4).

Proposition 4 (Riley and Samuelson (1981)). *In the first-price winner-pay auction a risk averse bidder bids more than a risk neutral bidder:*

$$\beta_{\text{RNBNE}}^{\text{I}}(v) \leq \beta_{\text{Risk}}^{\text{I}}(v)$$

Morgan et al. (2003) show that in the first-price winner-pay auction bids for risk averse and spiteful bidders are equivalent in the sense that for each risk-preference ρ^* one can always find a spite factor α^* such that the bidding function for a risk averse bidder is identical to the bidding function of a spiteful, but risk neutral bidder. In particular, Morgan et al. (2003, Proposition 4) show that risk averse bidders with CRRA utility $u(v) = v^\rho$ use the same bidding function as a spiteful bidder with spite parameter $\alpha = 1/\rho - 1$.

Figure 1 shows equilibrium bids of spiteful bidders for different valuations v and different levels of spite α . We could draw the same figure for risk averse bidders with risk aversion $\rho = 1/(1 + \alpha)$.

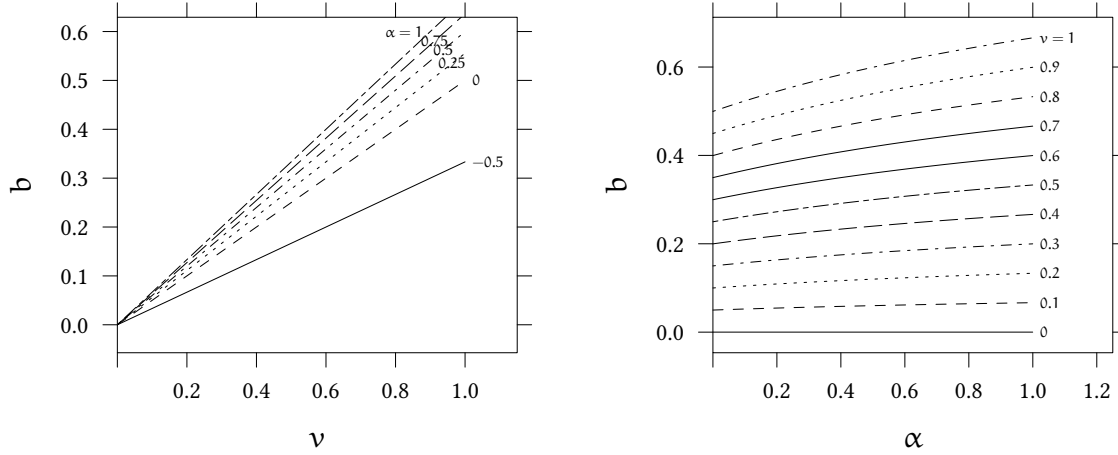


Figure 3: Equilibrium bids in first-price winner-pay auctions for spiteful bidders. Equilibrium bids in first-price winner-pay auctions for different valuations v and different levels of spite α with uniform distributions of valuations (see Equation (8)).

3.3. Revenue in the second-price all-pay auction and the first-price winner-pay auction

For spiteful bidders, we can derive the following proposition.¹⁵

Proposition 5. *For spiteful bidders revenues can be ranked as follows:*

$$\mathbb{E}(m_{\text{Spite}}^{\text{II-AP}}(v)) \geq \mathbb{E}(m_{\text{Spite}}^{\text{I-AP}}(v)) \geq \mathbb{E}(m_{\text{Spite}}^{\text{I}}(v)) \geq \mathbb{E}(m_{\text{selfish}}^{\text{I}}(v)) \quad (9)$$

The proof of Proposition 5 can be found in Appendix C.3.

Proposition 5 states, in particular, that for spiteful bidders revenue is larger in the second-price all-pay auction than in the first-price winner-pay auction.

If bidders are risk averse we can derive the opposite result:

Proposition 6. *For risk averse bidders revenues can be ranked as follows:*

$$\mathbb{E}(m_{\text{Risk}}^{\text{II-AP}}(v)) \leq \mathbb{E}(m_{\text{Risk}}^{\text{I}}(v))$$

The proof is show in Appendix C.4.

4. Design of the experiment and Hypotheses

To investigate the model presented above, we use a laboratory experiment. In the experiment, we first measure preferences for spitefulness and for risk. We will discuss the different

¹⁵In Appendix C we also derive revenue ranking for the first-price all-pay auction and also the second-price winner-pay auction.

measures of these preferences in Section 4.1. In the next step of the experiment participants bid either in a second-price all-pay auction or in a first-price winner-pay auction. We will discuss bidding behavior in Section 4.2. In Section 4.3 we discuss the payment of subject and Section 4.4 depicts the hypotheses for the experiment.

4.1. Preferences for Spitefulness and Risk

To measure preferences for risk we use a Holt and Laury (2002) task. We will discuss this measure in Section 4.1.1. To measure spiteful preferences we did not find a standard task. We use, hence, three different measures. One of the measures we use has been proposed by Marcus et al. (2014). We discuss this measure in Section 4.1.2. Another measure has been proposed by Kimbrough and Reiss (2012). We discussed their measure in Section 4.1.3. We also propose our own measure in Section 4.1.4. Each measure was explained to participants in great detail using video-instructions.¹⁶

4.1.1. Risk according to Holt and Laury (2002)

We measure preferences for risk with the help of a Holt and Laury (2002) task. This measure uses ten paired lottery choices.¹⁷ Each choice compares a risky lottery and a less risky lottery. The ten choices differ in the probabilities of the good outcomes of the lotteries. Participants who choose a large number of the risky options are considered more risk loving. Participants who choose more of the safer options are considered more risk averse.

4.1.2. Spite according to Marcus et al. (2014)

In the questionnaire by Marcus et al. (2014) participants are asked to rate 17 statements. Here are two examples:¹⁸

- If I am checking out at a store and I feel like the person in line behind me is rushing me, then I will sometimes slow down and take extra time to pay.
- I would rather no one get extra credit in a class if it meant that others would receive more credit than me.

Participants were asked to indicate their agreement on a scale between 1 and 5. Higher scores on the scale indicate more spitefulness. The measure of spitefulness with this task is the average agreement with the statements. The distribution of spitefulness with this measure is shown in the left part of Figure 4.

¹⁶Appendix F.2 provides the text of the videos. The videos can be found at <https://www.kirchkamp.de/research/SpiteVsRisk.html>.

¹⁷Lotteries are shown in Table 4 in Appendix D.1. Details of the implementation are illustrated in Appendix F.1, Second Task (B).

¹⁸All statements are shown in Appendix D.2.

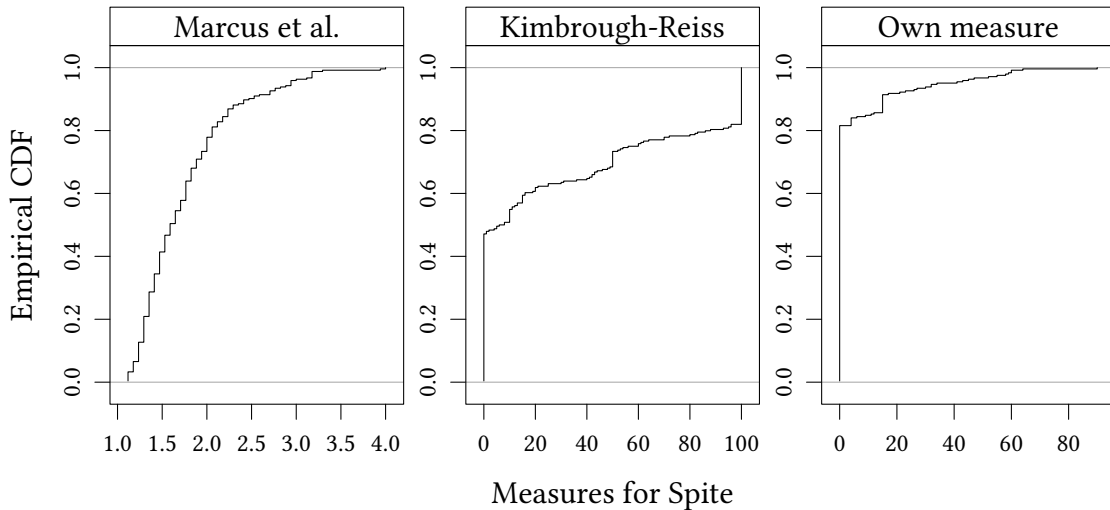


Figure 4: Distribution of Measures for Spite.

4.1.3. Spite according to Kimbrough and Reiss

As a second measure for spitefulness we use a modification of Kimbrough and Reiss (2012) who observe spiteful behavior with the help of a variant of a second-price auction.¹⁹ We first asked participants to supply a bid function for a second price auction with one opponent. Then valuations were generated randomly, bids were determined according to the stated bid functions, and participants were informed about the outcome of each auction, i.e. participants were told who had won the auction and the winner's bid. Separately for the won and lost auction participants could then increase or to keep their own bid by a percentage (between 0 and 100%) of the difference between the winner's and the loser's bid. Hence, bidders could not change the allocation. They could only diminish the winner's payoff. Furthermore, we also elicit the willingness to pay for this bid adaptation.

Participants who had increased their losing bid are considered spiteful. The spite-measure is a continuous measure between 0% (no adjustment) and 100% (if the loser increases the own bid up to the winner's bid). The distribution of spite for this measure is shown in the middle of Figure 4.

4.1.4. Our own measure for spitefulness

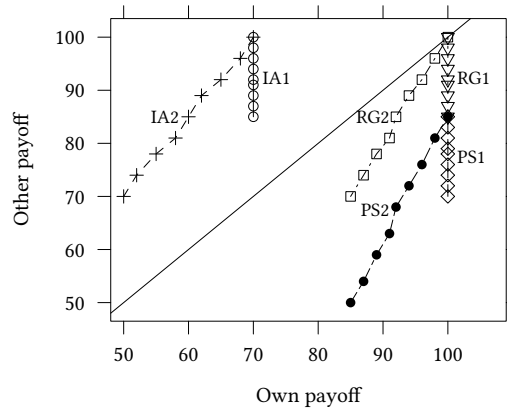
For our own measure of spitefulness we ask participants to decide six times among 9 possible allocations similar to the SVO-Slider measure by Murphy et al. (2011) and Murphy and Ackerman (2014). Figure 5 shows the six sets we use.²⁰ For each set participants had to chose their preferred allocation.

In each of the six sets the allocation with the highest payoff for the other player also

¹⁹See Appendix D.4 for details.

²⁰Details of the allocations are shown in Appendix D.3.

You	70	70	70	70	70	70	70	70	70
IA1	○	○	○	○	○	○	○	○	○
Other	100	98	96	94	92	91	89	87	85
You	70	68	65	62	60	58	55	52	50
IA2	○	○	○	○	○	○	○	○	○
Other	100	96	92	89	85	81	78	74	70
You	100	100	100	100	100	100	100	100	100
RG1	○	○	○	○	○	○	○	○	○
Other	100	98	96	94	92	91	89	87	85
You	100	98	96	94	92	91	89	87	85
RG2	○	○	○	○	○	○	○	○	○
Other	100	96	92	89	85	81	78	74	70
You	100	100	100	100	100	100	100	100	100
PS1	○	○	○	○	○	○	○	○	○
Other	85	83	81	79	78	76	74	72	70
You	100	98	96	94	92	91	89	87	85
PS2	○	○	○	○	○	○	○	○	○
Other	85	81	76	72	68	63	59	54	50



(a) The six allocation sets for our own slider measure as shown on the screen. (b) A graphical representation of the six allocation sets.

Figure 5: Own measure of spitefulness.

For each of the six sets players choose one allocation. For each set we consider the Pareto efficient allocation not spiteful. Less efficient allocations will be considered more spiteful.

maximizes the own payoff. Deviations from this allocation only reduce the payoff of the other player. These deviations never increase the own payoff. A deviation can, hence, be seen as a sign of spitefulness. This deviation is costless in sets IA1, RG1 and PS1. It is costly in IA2, RG2, and PS2.

While one reason for these deviations can be spite, there are other explanations. Deviations in sets IA1 and IA2 can be a sign of “inequality aversion”. Deviations in sets RG1 and RG2 can be a sign of “concerns for relative gain”.

As a measure for spitefulness we take the sum of points by which the payoff of the other player is reduced. Anybody who is not spiteful would leave 570 points to the other player. The lowest possible number of points a spiteful person could leave to the other is 430. This maximally spiteful person would, hence, reduce the payoff of the other by 140 points. Higher values indicate, hence, higher spitefulness.

Notably, based on this measure only 18% of participants were behaving spitefully at all. Only 12% of participants were willing to pay for this behavior. A distribution of the combined spite measure is shown in the right graph in Figure (4).

4.1.5. Other controls

We use the slider measure by Murphy et al. (2011) and Murphy and Ackerman (2014) to control for social value orientation and inequality aversion. We use the questionnaire of Back et al. (2013) to control for rivalry.

4.2. Design of the auction

After measuring preferences for spite,²¹ SVO and risk preferences participants played either the second-price all-pay auction or the first-price winner-pay auction. We explained to participants in great detail (using video-instructions) the rules of the auction.²² Participants played the auction for 15 rounds with stranger matching within matching groups of 6.

We use the strategy method to elicit bid functions. In each round participants were asked to state a bid for valuations of 0, 10, 20, . . . , 90, 100. Figure 6 shows an example of the bidding interface. Bids for intermediate valuations were linearly interpolated. To give more feedback in each round, each pair of bidders played ten auctions, each time for a random pair of valuations. Figure 7 shows an example of the feedback interface. For each of the ten auctions participants learn their own valuation, their own bid, and their opponent's bid. Participants also learn the outcome of the auction and how much they had won or lost.

4.3. Payment

Participants were paid at the end of the experiment for one random task, i.e. either one lottery from the risk-measure or one allocation from the SVO-Slider-Measure or the Spite-Measure or the adaptation of Kimbrough and Reiss (2012) or one of the auctions.²³ For each task we converted ECU (experimental currency unit) to Euros using separate rates to make sure that for the different tasks average payoffs were similar. For the same reason participants received a higher initial endowment in the all-pay auction.

4.4. Hypotheses

4.4.1. Bids in the first-price winner-pay auction:

We use the Bayesian Nash equilibrium as a benchmark. Following Propositions 3 and 4 we should expect spiteful agents and risk averse agents to bid more than non-spiteful risk neutral agents.

Hypothesis 1.1. *Increased spitefulness will lead to higher bids in the first-price winner-pay auction.*

Hypothesis 1.2. *Increased risk aversion will lead to higher bids in the first-price winner-pay auction.*

²¹The implementation of Kimbrough and Reiss (2012) and our all-pay auction were counterbalanced as both parts are auctions and we want to control for order effects here.

²²Appendix F.2 provides the text of the videos. The videos can be found at <https://www.kirchkamp.de/research/SpiteVsRisk.html>.

²³In case of the all-pay auction only one of the 10×15 auctions was paid out.

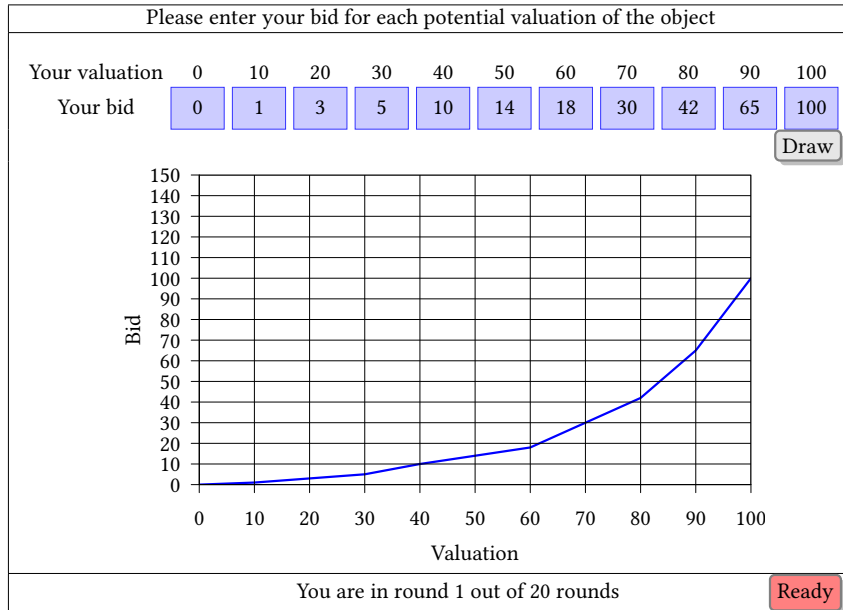


Figure 6: Interface of the bidding stage.

Imputing the bidding function for the possible valuations between 0 and 100. The bidding function is drawn after the input of the respective bids.

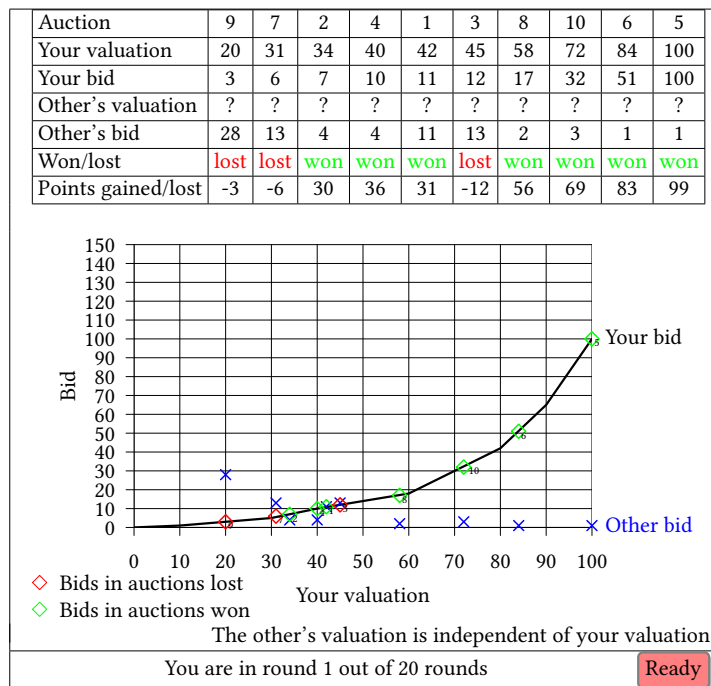


Figure 7: Interface of the feedback stage.

Mapping the 10 random valuations and the respective bids on the bidding function. Additionally subjects could see the opponent's bid, whether they won and the amount they won/lost.

4.4.2. Bids in the second-price all-pay auction:

Following Observation 1 we expect that in the second-price all-pay auction bids increase in spite for low valuations and they decrease in spitefulness for high valuations.²⁴

Hypothesis 2.1. *Bids increase in spite for low valuations and they decrease in spite for high valuations.*

We expect, hence, that bidders with spiteful preferences will bid more than the RNBNE for small valuations. They will bid less than the RNBNE for large valuations. Following Proposition 2 we expect that risk averse agents will underbid compared to risk neutral agents.

Hypothesis 2.2. *Increased risk aversion leads to lower bids.*

4.4.3. Revenue

Following Proposition 5 we expect the following:

Hypothesis 3.1. *Revenue in the second-price all-pay auction is higher than in the first-price winner-pay auction, if bidders are spiteful.*

Following Proposition 6 we expect the following:

Hypothesis 3.2. *Revenue in the second-price all-pay auction is lower than in the first-price winner-pay auction, if bidders are risk averse.*

Hypotheses 3.1 and 3.2 make different predictions. If spite is a motive driving bidding behavior, then we should find support for Hypothesis 3.1. If, however, risk aversion is driving the behavior, we expect to find the opposite (Hypothesis 3.2).

As the first-price winner-pay auction and the second-price all-pay auction are behaviorally very different auction formats we will compare the revenue in the two auction formats in Appendix E.5.

5. Results

We conducted the experiments in June 2015 (the second-price all-pay auction) and in May 2017 (the first-price winner-pay auction) at the laboratory of the school of economics of the University of Jena (Germany). We recruited 244 participants in 14 sessions using the online recruiting platform ORSEE (Greiner, 2015). We implemented the experiment using z-Tree (Fischbacher, 2007). Instructions were presented as 25-minute-videos followed by test questions for the auction and for the spite-measure based on Kimbrough and Reiss (2012). The entire experiment lasted for about 100 minutes. Participants earned on average 15.83€ (≈ 9.5 € an hour), which was at that time slightly above the minimum wage. We had 41% male and 59% female participants with a median age of 24. Participants were on average in their third year of studying and about 14% were students of business or economics.²⁵

²⁴In Appendix A we discuss the first-price all-pay auction. There we predict that spiteful bidders will bid more, in particular when their valuation is high.

²⁵In the second-price all-pay auction 138 subjects were recruited in 8 sessions with 45% male participants with a median age of 24 with 12% of subjects being students of business or economics.

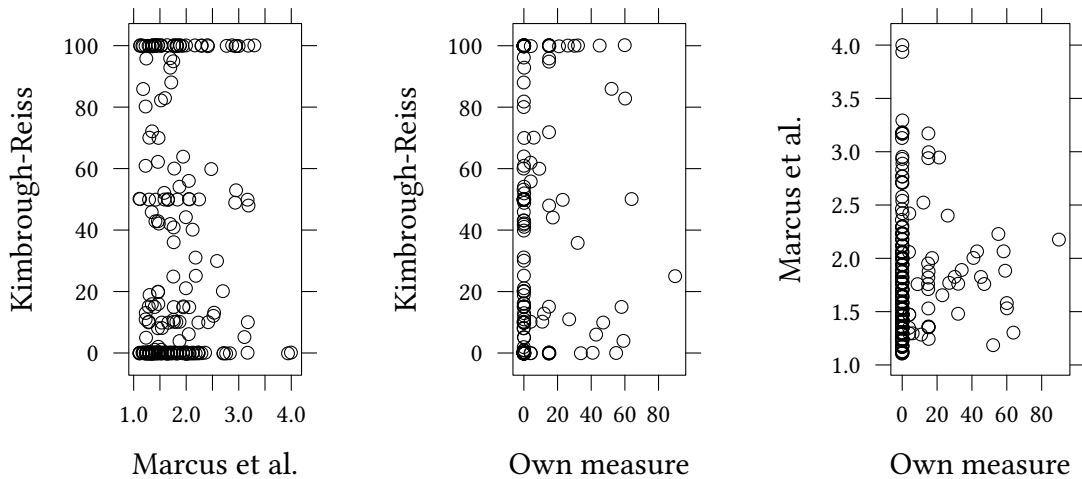


Figure 8: Joint Distribution of Measures for Spite.

5.1. Measures of Spite

Figure 8 shows the joint distribution of the three measures for spite. There is no evident correlation. For the three instruments we find a Cronbach α of 0.118 (CI=[0.0263,0.215]). The two behavioral measures are correlated significantly ($r = 0.137$, $p = 0.033$). The questionnaire is not significantly correlated with the two behavioral measures ($r = 0.079$, $p \geq 0.05$; $r = 0.061$, $p \geq 0.05$). Apparently, the three instruments seem to measure different aspects of spiteful preferences.

Having said that, we find substantial consistency within each scale. For the questionnaire (Marcus et al., 2014) we find a Cronbach α of 0.863 (CI=[0.829,0.9]). For our own measure we find a Cronbach α of 0.707 (CI=[0.635,0.789]).

Neither the questionnaire nor our own measure seems to be strictly one-dimensional. For the questionnaire, we find that the first element of a principal component analysis explains 33.2% of the variance, (CI=[27.7,37.6]). For our own measure, we find that the first element of a principal component analysis explains 76.6% of the variance, (CI=[65.8,85.9]).

As there is, in general, no way to disentangle which of the three spite-measures is “really” measuring spite we will look at the combined (normalized) measures.²⁶

5.2. Aggregated Bids

In this section we will present an overview of bidding behavior based on aggregated bids. In Section 5.3 we will continue with a more detailed model to explain individual bids.

In the first-price winner-pay auction 106 subjects were recruited in 6 sessions with 36% male participants with a median age of 25 with 16% of subjects being students of business or economics.

²⁶In Appendix E.4 we provide the main regressions with the individual measures. The results are very similar.

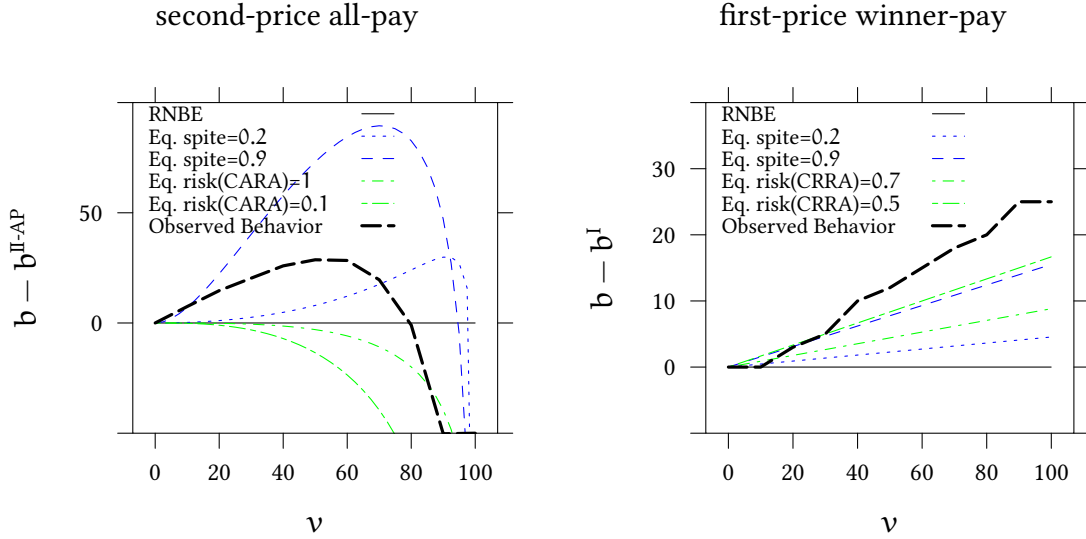


Figure 9: Median overbidding: Theory and observations.

The left graph shows median overbidding ($b - b^{\text{II-AP}}$) in the second-price all-pay auction. As a reference we include theoretical overbidding for spiteful ($\alpha > 0$) and for risk averse (CARA, $r < \infty$) agents. The right graph shows median overbidding ($b - b^{\text{I}}$) in the first-price winner-pay auction. As a reference we include theoretical overbidding for spiteful ($\alpha > 0$) and for risk averse (CRRA, $\rho < 1$) agents.

Figure 9 shows overbidding, i.e. the difference between average bids minus RNBNE bids in the two auction formats.

The right part of Figure 9 shows behavior in the first-price winner-pay auction. For low valuations overbidding is approximately zero. For larger valuations overbidding increases. Both spiteful preferences and risk aversion are in line with this behavior.

The left part of Figure 9 shows behavior in the second-price all-pay auction. For this format, spiteful preferences and risk aversion make quite different predictions. Risk aversion would predict underbidding for all valuations. Spiteful preferences would predict overbidding for intermediate valuations and underbidding only for very large valuations. Observed bids seem to follow the pattern predicted by spiteful preferences, and not the one predicted by risk aversion. We find overbidding up to a rather high valuation and underbidding afterwards.

While the figure suggests that spite might be a more convincing explanation than risk for most valuations, risk aversion is still in line with the observed underbidding for high valuations. Could it be that risk explains bids better at least for large valuations? Could a model of risk averse agents perhaps perform so well for large valuations that this extra performance compensates the comparatively worse performance of risk aversion for small valuations?

To answer this question formally, we estimate the following model:

$$\text{Bid}_{i,t,j,v} = \beta_{\text{II-AP}}^{\text{T}} + \zeta_{i,j} + \eta_j + \epsilon_{i,j,k,l} \quad (10)$$

where $\text{Bid}_{i,t,j,v}$ is the bid of subject i in group j in period t for valuation v . $\zeta_{i,j}$ is a random effect for bidder i in group j , η_j is a random effect for group j , and $\epsilon_{i,j,k,l}$ is the residual. $\beta_{\text{II-AP}}^{\text{T}}$

is the theoretical bidding function for the second-price all-pay auction according to either Equation (5) or (7). T indicates the type of the model: spiteful preferences (Equation (5)) or risk aversion (Equation (7)).

We fit the parameters (either α (spite parameter) or r (risk parameter)) of the theoretical bidding function by maximizing the log-likelihood of the model.²⁷ We find the model with spiteful agents significantly better ($\chi^2(df=0)= 50.316$, $p \leq 0.001$) than the model without spite. The model with risk averse agents, however, is not significantly better ($\chi^2(df=0)= 0$) than the one without risk aversion.

For the second-price all-pay auction we conclude the following:

Result 1.1. *Behavior in the second-price all-pay auction is significantly better described by a theory of spite than a theory of risk aversion.*

For the first-price winner-pay auction we have seen in Section 3.2.2 that equilibrium bids for a risk aversion bidder with risk aversion ρ are equivalent to equilibrium bids for a spiteful bidder with spite parameter $\alpha = 1/\rho - 1$. Hence, for the first-price winner-pay auction both theories describe behavior equally well.

Result 1.2. *Spite and risk perform equally well in describing behavior in the first-price winner-pay auction.*

5.2.1. The impact of spite and risk on bids

Let us next check whether the elicited preferences for spite and risk contribute to an explanation of observed bids.

First-price winner-pay auction Figure 10 is an extension of the right part of Figure 9. Similar to Figure 9 also Figure 10 shows median overbidding, i.e. bids minus RNBNE bids. Different from Figure 9, Figure 10 is based on a median split to divide participants into more and less spiteful ones on the left and into more or less risk averse ones on the right. The figure also includes equilibrium predictions for different levels of spite on the left and for risk aversion on the right.

For the first-price winner-pay auction Figure 10 does not suggest a substantial influence of spite or risk aversion on bids.

Second-price all-pay auction Figure 11 shows overbidding for the second-price all-pay auction for participants with different preferences for spite or risk. As in Figure 10 we use a median split for spite in the left part of Figure 11 and a median split for risk in the right part of Figure 11. We also include equilibrium overbidding for different levels of spite in the left part and for different levels of risk preferences in the right part.

²⁷We use a limited-memory modification of the Broyden-Fletcher-Goldfarb-Shanno quasi-Newton method to find the maximum while restricting all α to be in $(0, 1)$ and r to be in $(0, 10^{10})$. The best performing α is 0.664 (i.e. substantially spiteful) while the best performing r is 10^{10} (i.e. close to risk neutrality).

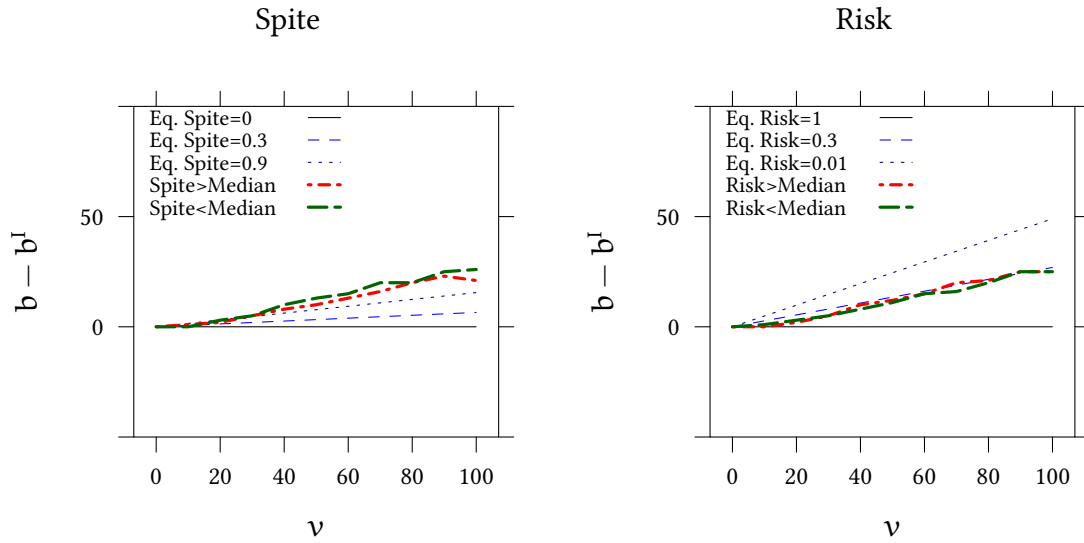


Figure 10: Spite and risk in the first-price winner-pay auction.

The left graph shows theoretical overbidding for spiteful-agents as well as median overbidding for above and below median spiteful experiment-participants. The right graph shows theoretical overbidding for risk averse-agents as well as median overbidding for above and below median risk averse experiment-participants. The theoretical predictions for risk averse subjects are derived from Morgan et al. (2003) based on CRRA-risk-preferences (risk of one denotes risk neutrality and decreasing numbers indicate increasing risk aversion).

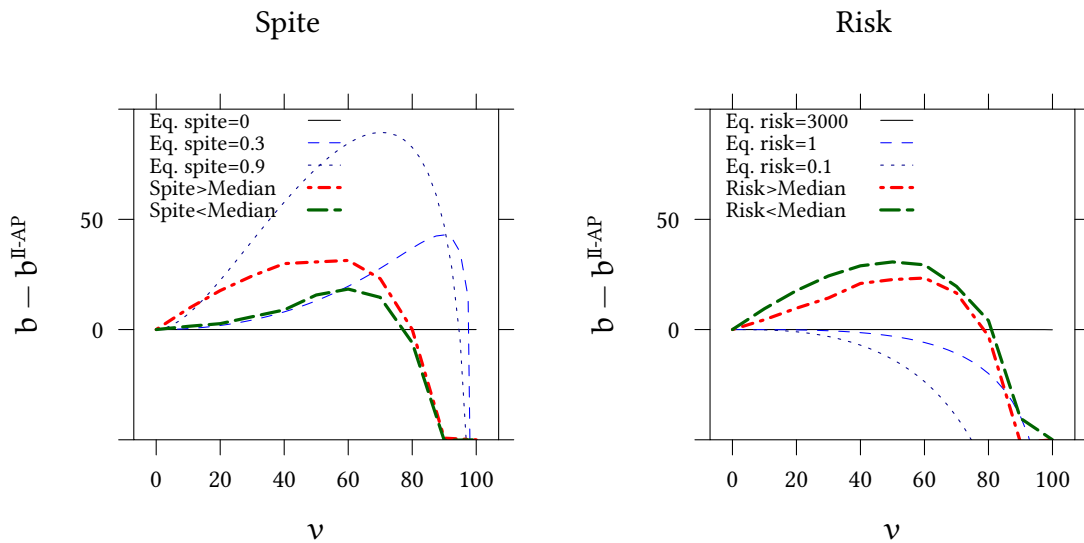


Figure 11: Median overbidding in the second-price all-pay auction.

The left graph shows theoretical overbidding for spiteful-agents as well as median overbidding for above and below median spiteful experiment-participants. The right graph shows theoretical overbidding for risk averse-agents (CARA) as well as median overbidding for above and below median risk averse experiment-participants. Risk of infinity denotes risk neutrality and decreasing numbers indicate increasing risk aversion.

	Second-price all-pay auction	First-price winner-pay auction
Spite	3.70* (1.67)	0.05 (0.40)
Risk	-7.39* (3.05)	0.09 (0.79)
Constant	11.48*** (3.05)	11.61*** (0.99)
Observations	138	106
Log Likelihood	-682.60	-370.37
Akaike Inf. Crit.	1,375.20	750.75
Bayesian Inf. Crit.	1,389.83	764.06
Notes:	p :*** < .001** < .01* < .05+ < .10	

Table 1: Mixed effects model of the average overbidding as a function of spite and risk. The table shows estimation results of overbidding in both auction types.

As predicted by theory, the difference between bids of high-spite subjects and low-spite subjects increases up to a high valuation and decreases fast afterwards. The difference between bids of high risk aversion subjects and low risk aversion subjects is negatively increasing, as predicted by theory.

A formal comparison What we have seen in Figures 10 and 11 can also be confirmed more formally. We use a mixed effects model (Equation (11)) to explain average overbidding (per participant) for the two auction formats.

$$\overline{\text{Bid}_{i,j} - b^I} = \beta_0 + \beta_{\text{Spite}} \text{Spite}_{i,j} + \beta_{\text{Risk}} \text{Risk}_{i,j} + \eta_j + \epsilon_{i,j} \quad (11)$$

We call $\overline{\text{Bid}_{i,j} - b^I}$ the average overbidding of participant i in group j over all valuations and all rounds that participant played. $\text{Spite}_{i,j}$ is the sum of the three spite measures for participant i in group j . $\text{Risk}_{i,j}$ is the risk aversion for this person, and η_j is the group specific random effect. Table 1 shows estimation results.

As we have seen in Figure 11 we confirm for the second-price all-pay auction that spite leads to overbidding. Risk aversion leads to underbidding. Both observations are in line with theory (Equations (5) and (7)).

As we have seen in Figure 10, in the first-price winner-pay auction neither risk nor spite contribute significantly to overbidding.

Summary of aggregate results To summarise this section, we have found for the first-price winner-pay auction that bids can be rationalised equally well with risk aversion or spiteful preferences. However, our measures of these preferences do not seem to explain actual bids.

Things are different for the second-price all-pay auction. Here, our measure for spite and our measure for risk preferences explains actual bids in line with the equilibrium prediction. More spiteful bidders bid more, as they should. More risk averse bidders bid less, again as they should. Most importantly, however, we find that most of the deviation of bids from RNBNE bids seems to be due to spiteful preference, not due to risk aversion.

5.3. Individual bids

Let us next turn to individual bids. We will start with the first-price winner-pay auction in Section 5.3.1 and turn to the second-price all-pay auction in Section 5.3.2.

5.3.1. First-price winner-pay auction

In Section 5.2.1 we did not find a substantial effect of preferences for risk and spite on aggregate bids for the first-price winner-pay auction. To present a more detailed image we will use individual bids in the current section. We will employ a linear mixed effects model with several controls.²⁸ Equation (12) specifies the different econometric models:

$$\begin{aligned}
 \text{Bid}_{i,t,j,v} - b^l &= \beta_0 + \beta_1 \text{Period} + \beta_2 v + \zeta_{i,j} + \eta_j + \epsilon_{i,j,k,l} + C_M & (12) \\
 C_1 &= 0 \\
 C_2 &= \beta_3 \text{Spite}_i + \beta_4 \text{Spite}_i \times v \\
 C_3 &= C_2 + \beta_5 \mathbb{1}_{\text{Gender}=\text{♀}} + \beta_6 \text{Risk}_i + \beta_7 \text{rivalry}_i + \beta_8 \text{SVO}_i + \beta_9 \text{IA}_i \\
 C_4 &= \beta_{10} \text{Risk}_i + \beta_{11} \text{Risk}_i \times v \\
 C_5 &= C_4 + \beta_{12} \mathbb{1}_{\text{Gender}=\text{♀}} + \beta_{13} \text{Spite}_i + \beta_{14} \text{rivalry}_i + \beta_{15} \text{SVO}_i + \beta_{16} \text{IA}_i
 \end{aligned}$$

where $\zeta_{i,j}$ is a random effect for bidder i in group j , η_j is a random effect for group j , and $\epsilon_{i,j,k,l}$ is the residual. C_1 is a base specification. Specification C_2 and C_3 control for spite. C_4 and C_5 control for risk.

Table 2 shows estimation results for Equation (12).

Model C_1 assumes a simple linear relationship between valuation v and bids. This can be rationalised with a theory of risk averse bidders but also with a theory of spiteful bidders.

Result 2.1. *Overbidding in the first-price winner-pay auction is consistent with the theory of spiteful-agents and also with theory on risk averse agents.*

Models C_2 and C_3 allow us to investigate Hypothesis 1.1. We find that, contrary to the theoretical prediction, more spite leads to a less steep bidding function (the interaction of $\text{Spite} \times v$ is negative and significant).

Result 2.2. *Contrary to the theoretical prediction, more spite leads to a less steep bidding slope in the first-price winner-pay auction (the interaction of $\text{Spite} \times v$ is negative and significant).*

With models C_4 and C_5 we will investigate Hypothesis 1.2. We find that risk aversion leads, in line with theory, to steeper bids (the interaction of $\text{Risk} \times v$ is positive and significant).

Result 2.3. *In line with theory, more risk aversion leads to a steeper bidding slope in the first-price winner-pay auction.*

²⁸Actually, for the first-price winner-pay auction there is not a big difference between estimating bids and overbidding. Since overbidding in the first-price winner-pay auction is just the bidding behavior minus half of the valuation, estimating bids would give us exactly the same coefficients, but the coefficient of valuations on bidding behavior, which would be increased by 1/2.

	C ₁	C ₂	C ₃	C ₄	C ₅
Period	−0.09*** (0.02)	−0.09*** (0.02)	−0.09*** (0.02)	−0.09*** (0.02)	−0.09*** (0.02)
v	0.21*** (0.003)	0.21*** (0.003)	0.21*** (0.003)	0.21*** (0.003)	0.21*** (0.003)
Spite		0.53 (0.40)	0.33 (0.51)		−0.15 (0.50)
Spite × v		−0.01*** (0.002)	−0.01*** (0.002)		
Risk			−0.06 (0.79)	−1.46 ⁺ (0.80)	−1.59* (0.80)
Risk × v				0.03*** (0.003)	0.03*** (0.003)
Male			−4.98** (1.65)		−4.98** (1.65)
Rivalry			0.86 (0.86)		0.86 (0.86)
SVO			0.04 (0.06)		0.04 (0.06)
IA			0.46 (0.63)		0.46 (0.63)
Constant	1.96 ⁺ (1.01)	1.96 ⁺ (1.02)	2.92 ⁺ (1.60)	1.96 ⁺ (1.02)	2.92 ⁺ (1.60)
Observations	17,490	17,490	17,490	17,490	17,490
Log Likelihood	−69,248.54	−69,233.82	−69,227.71	−69,200.08	−69,194.64
Akaike Inf. Crit.	138,509.10	138,483.60	138,481.40	138,416.10	138,415.30
Bayesian Inf. Crit.	138,555.70	138,545.80	138,582.40	138,478.30	138,516.30

Notes:

p :*** < .001** < .01* < .05⁺ < .10

Table 2: Estimation results for Equation (12) (overbidding for the first-price winner-pay auction).

The table shows estimation results for the different controls C₁, C₂, C₃, C₄, and C₅. Spite is the sum of the three spite measures. IA is the sum of the inequality aversion score obtained from the slider measure and the score obtained from inequality allocation of our own spite measure.

Adding extra controls (in C₃ and C₅) for gender, rivalry, social value orientation, and inequality aversion does not change the coefficients for spite and risk. Gender is a highly significant factor of overbidding. Furthermore, overbidding decreases over time, which could be seen a sign of learning.

5.3.2. Second-price all-pay auction

In Section 5.2.1 we found for the second-price all-pay auction a substantial effect of preferences for risk and spite on aggregate bids. In the current section we will use individual bids to present a more detailed picture. Similar to Section 5.3.1 we will use a mixed effects model.²⁹ Since overbidding for the first-price winner-pay auction is (in equilibrium) linear in valuations, we used a specification linear in valuations in Section 5.3.1. For the second-price all-pay auction matters are different. Here, overbidding is non-linear in valuations. Hence, we follow a non-linear approach in the current section. Specifically, we use a generalized additive model (GAM) where overbidding is modeled as a smooth function of the valuation.³⁰

A second non-linearity that we have to account for is that in equilibrium of the second-

²⁹We are mainly interested in overbidding-behavior. Nevertheless, we also estimate bidding behavior in Appendix E.1.

³⁰We used the default thin plate regression spline. Note that also cubic regression spline, cyclic cubic regression splines and P-splines (a specific version of B-Splines) result in qualitatively the same outcome, as can be seen in Appendix E. We also estimate the same regression by using piece wise linear splines – again results are robust to this specification.

price all-pay auction spite leads to a non-linear increase in bids.³¹ Also risk aversion has a non-linear effect on bids.³² For increasing spite, we expect increasing overbidding up to a certain level, and underbidding for high valuations. For increasing risk aversion we expect increasing underbidding which becomes stronger for high valuations. To simplify the interpretation of our results, we use a piece-wise linear function with a constant slope for valuations below 50 and a constant slope for valuations above 50.³³ We compare five different models:

$$\begin{aligned}
\text{Bid}_{i,t,j,v} - b^{\text{I-AP}} &= \beta_0 + \beta_1 \text{Period} + \zeta_{i,j} + \eta_j + \epsilon_{i,j,k,l} + C'_M & (13) \\
C'_1 &= s(v) \\
C'_2 &= C'_1 + \beta_2 \text{Spite}_i + \beta_3 \text{Spite}_i \cdot v_{[0,50]}(v) + \beta_4 \text{Spite}_i \cdot v_{[50,100]}(v) \\
C'_3 &= C'_2 + \beta_5 \text{IA}_i + \beta_6 \mathbb{1}_{\text{Gender}=\text{♀}} + \beta_7 \text{Risk}_i + \beta_8 \text{rivalry}_i + \beta_9 \text{SVO}_i \\
C'_4 &= C'_1 + \beta_{10} \text{Risk}_i + \beta_{11} \text{Risk}_i \cdot v_{[0,50]}(v) + \beta_{12} \text{Risk}_i \cdot v_{[50,100]}(v) \\
C'_5 &= C'_2 + \beta_{13} \text{IA}_i + \beta_{14} \mathbb{1}_{\text{Gender}=\text{♀}} + \beta_{15} \text{Spite}_i + \beta_{16} \text{rivalry}_i + \beta_{17} \text{SVO}_i
\end{aligned}$$

where $\zeta_{i,j}$ is a random effect for bidder i in group j , η_j is a random effect for group j , and $\epsilon_{i,j,k,l}$ is the residual. $s(v)$ is the thin plate regression spline over the valuation. $v_{[0,50]}(v)$ and $v_{[50,100]}(v)$ are defined as follows:³⁴

$$v_{[0,50]}(v) = \begin{cases} v & \text{if } v \leq 50 \\ 0 & \text{otherwise} \end{cases} \quad (14)$$

$$v_{[50,100]}(v) = \begin{cases} v & \text{if } v > 50 \\ 0 & \text{otherwise} \end{cases} \quad (15)$$

Estimation results are shown in Table 3. Figure 12 shows estimation results for the fitted spline.

Figure 12 shows the spline $s(v)$ from Equation (13). In line with Hypothesis 2.1 we see that, for all specifications, overbidding first increases up to a certain point and then, for high valuations, turns into underbidding.

Result 3.1. *In line with spiteful preferences, bidders bid more than the RNBNE for small valuations and, respectively, less for large valuations.*

Table 3 provides estimation results for Equation (13). Hypothesis 2.1 can be assessed with the help of models C'_2 and C'_3 . Indeed, with increasing spite overbidding increases more strongly for $v < 50$ and then more slowly for $v > 50$.

Result 3.2. *Bids increase in spite for low valuations and they increase less for high valuations.*

³¹See Equation (5), Figure 1 for bids and Figure 11 for overbidding.

³²See Equation (7), Figure 2 for bids and Figure 11 for overbidding.

³³Technically: We use a B-spline of degree 1 with one knot at 50. In Appendix E.3 we alternatively model the non-linearity by using squared valuations. The results are robust to those specifications.

³⁴Results are robust to using a cut-off different from 50.

	C'_1	C'_2	C'_3	C'_4	C'_5
Period	-0.40*** (0.05)	-0.40*** (0.05)	-0.40*** (0.05)	-0.40*** (0.05)	-0.40*** (0.05)
Spite		2.57 (1.70)	3.34 ⁺ (1.97)		4.10* (1.95)
Spite $\times v_{[0,50]}$		1.49** (0.47)	1.49** (0.47)		
Spite $\times v_{[50,100]}$		0.31 (0.43)	0.31 (0.43)		
Risk			-6.02* (2.91)	-4.36 (3.10)	-3.62 (2.96)
Risk $\times v_{[0,50]}$				-3.48*** (0.86)	-3.48*** (0.86)
Risk $\times v_{[50,100]}$				-3.02*** (0.78)	-3.02*** (0.78)
Male			-19.05** (6.11)		-19.05** (6.11)
Rivalry			-0.70 (3.09)		-0.70 (3.09)
SVO			0.41 ⁺ (0.24)		0.41 ⁺ (0.24)
IA			-1.84 (2.51)		-1.84 (2.51)
Constant	14.92*** (3.15)	14.89*** (3.11)	14.83* (6.48)	14.87*** (3.10)	14.83* (6.48)
Observations	23760	23760	23760	23760	23760
Log Likelihood	-120506.69	-120499.39	-120490.12	-120493.49	-120484.68
Akaike Inf. Crit	241027.38	241018.78	241010.24	241006.97	240999.36
Bayesian Inf. Crit.	241083.91	241099.54	241131.38	241087.73	241120.5

Notes: p :*** < .001** < .01* < .05⁺ < .10;

Table 3: Estimation results for Equation (13) (overbidding in the second-price all-pay auction).

The table shows estimation results for the different controls C'_1 , C'_2 , C'_3 , C'_4 , and C'_5 . Thin plate regression splines are used for $s(v)$. Spite is the sum of the three spite measures. IA is the sum of the inequality aversion score obtained from the slider measure and the score obtained from inequality allocation of our own spite measure. Standard errors are shown in parenthesis.

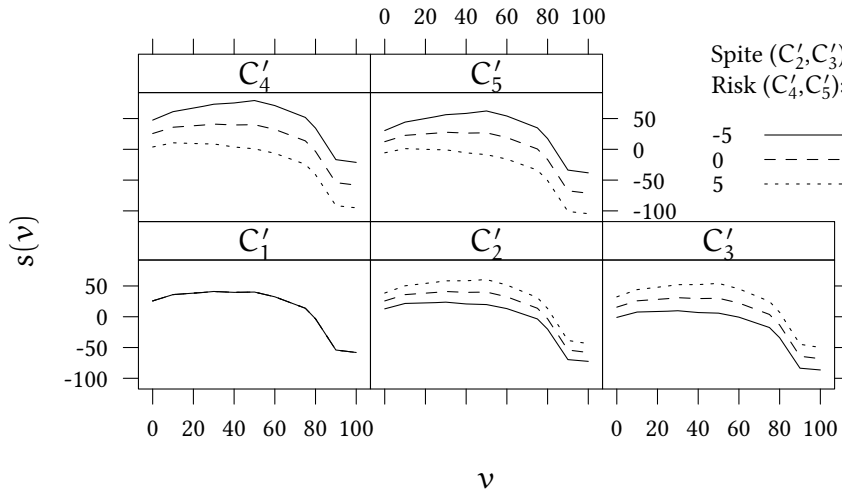


Figure 12: Estimation results for the spline from Equation (13) (overbidding). The Figure show splines for different controls C'_1 , C'_2 , C'_3 , C'_4 , and C'_5 and for different (normalized) levels of spite (in Controls C'_2 , C'_3) and different (normalized) levels of risk (in Controls C'_4 , C'_5).

Hypothesis 2.2 can be assessed with the help of models C'_4 and C'_5 : Indeed, risk aversion leads to underbidding which is increasing in the valuation.

Result 3.3. *Increased risk aversion leads to lower bids.*

Adding extra controls (in C_3 and C_5) for gender, rivalry, social value orientation, and inequality aversion does not change the coefficients for spite and risk. As with our estimation for the first-price winner-pay auction in Equation (12), also for the second-price all-pay auction gender is a highly significant factor of overbidding and subjects decrease overbidding over time.

All in all, estimation results for Equation (13) suggest that the theory of spiteful bidding performs rather well. More spitefulness leads to more overbidding in particular for smaller valuations. We also find support for the theory of risk averse bidders, i.e. that more risk aversion leads to more underbidding. However, as we have already seen for aggregate results in Section 5.2, overall behavior is more in line with spiteful bidding than with risk aversion – we find mostly overbidding. Furthermore, overbidding is increasing on average in particular for intermediate valuations.

6. Discussion and Conclusion

In this paper we want to contribute – with the help of theory and experiments – to a better understanding of bidding behavior in auctions. We propose that spite could be a relevant factor to explain bids. As a workhorse we use the second-price all-pay auction and the first-price winner-pay auction.

We show that, in equilibrium, spite and risk should have an influence on bids in these two auction formats. For the second-price all-pay auction spite leads to overbidding and risk-aversion leads to underbidding. For the first-price winner-pay auction spite has the same effect as risk-aversion: both lead to overbidding.

For the participants in our experiment we use three different measures of spiteful preferences: a questionnaire, an allocation task and an experimental design similar to Kimbrough and Reiss (2012). We use a Holt and Laury (2002) task to measure risk-preferences.

Our measures of spiteful preferences and risk-aversion predict bidding behavior, in particular in the second-price all-pay auction, quite well. As predicted by theory, more spiteful bidders make higher bids. More risk-averse bidders make lower bids. Most importantly, in the second-price all-pay auction the overall effect of spite seems to dominate the effect of risk.

Bidding behavior in the first-price winner-pay auction is also well explained by risk preferences. Spite performs less well as a predictor in the first-price winner-pay auction. Contrary to the equilibrium prediction, more spitefulness seems to reduce the slope of the bidding function in the first-price winner-pay auction. In line with the equilibrium prediction more risk-aversion leads to an increase in the slope of the bidding function.

To summarise, spite seems to be a very appealing explanation for bidding behavior in the second-price all-pay auction. However, we cannot generalize this findings to the first-price winner-pay auction. One might speculate that spite might be the better predictor in a more

aggressive context (like the second-price all-pay auction) but perhaps not good as a predictor for some more conventional auctions.

We have also seen that the concept of spite itself seems to be hard to grasp. The correlation of the three measures for spite we are using is positive. However, the correlation is not too large. Also our approach, to simply sum up the normalized values of each measure, is pragmatic.

Further possible extensions of our work could focus on the model of spite. In this paper we have assumed that only the loser of an auction is spiteful. Furthermore, we have treated spite only as a constant, independent of valuation and bid and identical for all members of the population. Further theoretical work might relax these assumptions.

Despite these limitations we, nevertheless, conclude that spite is a relevant and important motive in auctions. In particular, our results seem to suggest that the spite motive could be as relevant and important as risk aversion, in particular competitive situations.

References

- Albano, G. L. (2001). A class of all-pay auctions with affiliated information. *Louvain Economic Review*, 67(1):31–38.
- Anderson, S., Goeree, J. K., and Holt, C. A. (1998). Rent seeking with bounded rationality: An analysis of the all-pay auction. *Journal of Political Economy*, 106(4):828–853.
- Andreoni, J., Che, Y.-K., and Kim, J. (2007). Asymmetric information about rivals' types in standard auctions: An experiment. *Games and Economic Behavior*, 59(2):240–259.
- Armantier, O. and Treich, N. (2009). Subjective probabilities in games: an application to the overbidding puzzle. *International Economic Review*, 50(4):1079–1102.
- Back, M. D., Küfner, A. C. P., Dufner, M., Gerlach, T. M., Rauthmann, J. F., and Denissen, J. J. A. (2013). Narcissistic admiration and rivalry: Disentangling the bright and dark sides of narcissism. *Journal of Personality and Social Psychology*, 105(10):1013–1037.
- Bartling, B., Gesche, T., and Netzer, N. (2017). Does the absence of human sellers bias bidding behavior in auction experiments? *Journal of the Economic Science Association*, 3(1):44–61.
- Bartling, B. and Netzer, N. (2016). An externality-robust auction: Theory and experimental evidence. *Games and Economic Behavior*, 97:186–204.
- Barut, Y., Kovenock, D., and Noussair, C. (2002). A comparison of multiple-unit all-pay and winner-pay auctions under incomplete information. *International Economic Review*, 43(3):675–708.
- Baye, M. R., Kovenock, D., and de Vries, C. G. (2005). Comparative analysis of litigation systems: An auction-theoretic approach. *The Economic Journal*, 115(505):583–601.

- Bertoletti, P. (2016). Reserve prices in all-pay auctions with complete information. *Research in Economics*, 70(3):446–453.
- Bilodeau, M., Childs, J., and Mestelman, S. (2004). Volunteering a public service: an experimental investigation. *Journal of Public Economics*, 88(12):2839–2855.
- Bos, O. (2012). Wars of attrition and all-pay auctions with stochastic competition. *Journal of Mathematical Economics*, 48(2):83–91.
- Brandt, F., Tuomas, S., and Yoav, S. (2007). Spiteful bidding in sealed-bid auctions. In *Proceedings of the Twentieth International Joint Conference on Artificial Intelligence (IJCAI)*.
- Cason, T. N., Saijo, T., and Yamato, T. (2002). Voluntary participation and spite in public good provision experiments: An international comparison. *Experimental Economics*, 5(2):133–153.
- Chen, Y., Katuscak, P., and Ozdenoren, E. (2013). Why can't a woman bid more like a man? *Games and Economic Behavior*, 77(1):181–213.
- Chen, Z. C., Ong, D., and Sheremeta, R. M. (2015). The gender difference in the value of winning. *Economics Letters*, 137:226–229.
- Cooper, D. J. and Fang, H. (2008). Understanding overbidding in second price auctions: An experimental study. *The Economic Journal*, 118(532):1572–1595.
- Cox, J., Smith, V., and Walker, J. (1988). Theory and individual behavior of first-price auctions. *Journal of Risk and Uncertainty*, 1(1):61–99.
- Cox, J. C. and James, D. (2012). Clocks and trees: Isomorphic dutch auctions and centipede games. *Econometrica*, 80(2):883–903.
- Cox, J. C., Smith, V. L., and Walker, J. M. (1985). Experimental development of sealed-bid auction theory: Calibrating controls for risk aversion. *American Economic Review*, 75(2):160–65.
- Dechenaux, E., Kovenock, D., and Sheremeta, R. (2015). A survey of experimental research on contests, all-pay auctions and tournaments. *Experimental Economics*, pages 609–669.
- Dechenaux, E. and Mancini, M. (2008). Auction-theoretic approach to modeling legal systems: An experimental analysis. *Applied Economics Research Bulletin*, 2:142–177.
- Dittrich, D. A. V., Güth, W., Kocher, M. G., and Pezanis-Christou, P. (2012). Loss aversion and learning to bid. *Economica*, 79(314):226–257.
- Engelbrecht-Wiggans, R. and Katok, E. (2009). A direct test of risk aversion and regret in first price sealed-bid auctions. *Decision Analysis*, 6(2):75–86.
- Ernst, C. and Thöni, C. (2013). Bimodal Bidding in Experimental All-Pay Auctions. *Games*, 4(4):608–623.

- Feess, E., Muehlheusser, G., and Walzl, M. (2008). Unfair contests. *Journal of Economics*, 93(3):267–291.
- Fehr, E., Hoff, K., and Kshetramade, M. (2008). Spite and development. *American Economic Review*, 98(2):494–99.
- Fibich, G., Gavious, A., and Sela, A. (2006). All-pay auctions with risk-averse players. *International Journal of Game Theory*, 34(4):583–599.
- Filiz-Ozbay, E. and Ozbay, E. Y. (2007). Auctions with anticipated regret: Theory and experiment. *American Economic Review*, 97(4):1407–1418.
- Filiz-Ozbay, E. and Ozbay, E. Y. (2010). Anticipated loser regret in third price auctions. *Economics Letters*, 107(2):217–219.
- Fischbacher, U. (2007). z-tree: Zurich toolbox for ready-made economic experiments. *Experimental Economics*, 10(2):171–178.
- Fudenberg, D. and Tirole, J. (1986). A theory of exit in duopoly. *Econometrica*, 54(4):943–60.
- Ghemawat, P. and Nalebuff, B. (1985). Exit. *RAND Journal of Economics*, 16(2):184–194.
- Goeree, J. K., Holt, C. A., and Palfrey, T. R. (2002). Quantal response equilibrium and overbidding in private-value auctions. *Journal of Economic Theory*, 104(1):247–272.
- Greiner, B. (2015). Subject pool recruitment procedures: organizing experiments with orsee. *Journal of the Economic Science Association*, 1(1):114–125.
- Güth, W., Ivanova-Stenzel, R., Königstein, M., and Strobel, M. (2003). Learning to bid - an experimental study of bid function adjustments in auctions and fair division games. *The Economic Journal*, 113(487):477–494.
- Hendricks, K., Weiss, A. M., and Wilson, C. A. (1988). The war of attrition in continuous time with complete information. *International Economic Review*, 29(4):663–80.
- Holt, C. A. and Laury, S. K. (2002). Risk aversion and incentive effects. *American Economic Review*, 92(5):1644–1655.
- Hyndman, K., Ozbay, E. Y., and Sujarittanonta, P. (2012). Rent seeking with regretful agents: Theory and experiment. *Journal of Economic Behavior & Organization*, 84(3):866–878.
- Hörisch, H. and Kirchkamp, O. (2010). Less fighting than expected. *Public Choice*, 144(1):347–367.
- Kagel, J. H., Harstad, R. M., and Levin, D. (1987). Information Impact and Allocation Rules in Auctions with Affiliated Private Values: A Laboratory Study. *Econometrica*, 55(6):1275–1304.

- Kagel, J. H. and Levin, D. (1993). Independent private value auctions: Bidder behaviour in first-, second- and third-price auctions with varying numbers of bidders. *Economic Journal*, 103(419):868–79.
- Katok, E. and Kwasnica, A. (2008). Time is money: The effect of clock speed on seller’s revenue in dutch auctions. *Experimental Economics*, 11(4):344–357.
- Katuscak, P., Michelucci, F., and Zajicek, M. (2013). Does Anticipated Regret Really Matter? Revisiting the Role of Feedback in Auction Bidding. Cerge-ei working papers, The Center for Economic Research and Graduate Education - Economics Institute, Prague.
- Kimbrough, E. O. and Reiss, J. P. (2012). Measuring the distribution of spitefulness. *PLOS ONE*, 7(8):1–8.
- Kirchkamp, O., Poen, E. E., and Reiss, J. P. (2009). Outside options: Another reason to choose the first-price auction. *European Economic Review*, 53(2):153–169.
- Kirchkamp, O. and Reiss, J. P. (2008). Heterogeneous bids in auctions with rational and markdown bidders - theory and experiment. Jena Economic Research Papers 2008-066, Friedrich-Schiller-University Jena, Max-Planck-Institute of Economics.
- Kirchkamp, O., Reiss, J. P., and Sadrieh, A. (2008). A pure variation of risk in private-value auctions. Research Memorandum 050, Maastricht University, Maastricht Research School of Economics of Technology and Organization (METEOR).
- Klose, B. and Kovenock, D. (2015). The all-pay auction with complete information and identity-dependent externalities. *Economic Theory*, 59(1):1–19.
- Kovenock, D., Baye, M. R., and de Vries, C. G. (1996). The all-pay auction with complete information. *Economic Theory*, 8(2):291–305.
- Krishna, V. and Morgan, J. (1997). An analysis of the war of attrition and the all-pay auction. *Journal of Economic Theory*, 72(2):343–362.
- Lugovskyy, V., Puzzello, D., and Tucker, S. (2010). An experimental investigation of overdispersion in the all pay auction. *European Economic Review*, 54(8):974–997.
- Marcus, D. K., Zeigler-Hill, V., Mercer, S. H., and Norris, A. L. (2014). The psychology of spite and the measurement of spitefulness. *Psychological Assessment*, 26(2):563–574.
- Maskin, E. and Riley, J. (1984). Optimal auctions with risk averse buyers. *Econometrica*, 52(6):1473–1518.
- Mill, W. (2017). The spite motive in third price auctions. *Economics Letters*, 161:71–73.
- Morgan, J., Steiglitz, K., and Reis, G. (2003). The Spite Motive and Equilibrium Behavior in Auctions. *Contributions in Economic Analysis & Policy*, 2(1).

- Murphy, R. O. and Ackerman, K. A. (2014). Social value orientation: Theoretical and measurement issues in the study of social preferences. *Personality and Social Psychology Review*, 18(1):13–41.
- Murphy, R. O., Ackerman, K. A., and Handgraaf, M. J. J. (2011). Measuring social value orientation. *Judgment and Decision Making*, 6(8):771–781.
- Nakajima, D. (2011). First-price auctions, dutch auctions, and buy-it-now-prices with allais paradox bidders. *Theoretical Economics*, 6(3):473–498.
- Nishimura, N., Cason, T. N., Saijo, T., and Ikeda, Y. (2011). Spite and reciprocity in auctions. *Games*, 2(3):365–411.
- Noussair, C. and Silver, J. (2006). Behavior in all-pay auctions with incomplete information. *Games and Economic Behavior*, 55(1):189–206.
- Ockenfels, A. and Selten, R. (2005). Impulse balance equilibrium and feedback in first price auctions. *Games and Economic Behavior*, 51(1):155–170.
- Oprea, R., Wilson, B. J., and Zillante, A. (2013). War of attrition: evidence from a laboratory experiment on market exit. *Economic Inquiry*, 51(4):2018–2027.
- Potters, J., de Vries, C. G., and van Winden, F. (1998). An experimental examination of rational rent-seeking. *European Journal of Political Economy*, 14(4):783–800.
- R Core Team (2017). *R: A Language and Environment for Statistical Computing*. R Foundation for Statistical Computing, Vienna, Austria.
- Riley, J. and Samuelson, W. F. (1981). Optimal auctions. *American Economic Review*, 71(3):381–92.
- Riley, J. G. (1980). Strong evolutionary equilibrium and the war of attrition. *Journal of Theoretical Biology*, 82(3):383–400.
- Sacco, D. and Schmutzler, A. (2008). All-pay auctions with negative prize externalities: Theory and experimental evidence. SOI - Working Papers 0806, Socioeconomic Institute - University of Zurich.
- Saijo, T. and Nakamura, H. (1995). The "spite" dilemma in voluntary contribution mechanism experiments. *Journal of Conflict Resolution*, 39(3):535–560.
- Sandholm, T. and Sharma, A. (2010). Asymmetric spite in auctions. In *Association for the Advancement of Artificial Intelligence*.
- Sandholm, T. and Tang, P. (2012). Optimal auctions for spiteful bidders. In *Association for the Advancement of Artificial Intelligence*.
- Smith, J. M. (1974). The theory of games and the evolution of animal conflicts. *Journal of Theoretical Biology*, 47(1):209–221.

Appendix – for online publication only

A. Bids in the first-price all-pay auction

Two bidders i and j have private valuations for a prize with values v_k , $k \in \{i, j\}$. Both submit a bid b_k following a monotonic bidding function $\beta_k(v_k)$ with first derivative $\beta'_k(x) \equiv \frac{d\beta(x)}{dx}$ and inverse $\beta_k^{-1}(b_k) = v_k$. Valuations are distributed according to F , i.e. $v \sim F(0, \bar{v})$, and $f(x) = \frac{dF(x)}{dx}$. The prize is allocated to the player with the highest bid. If $v_i = v_j$, the prize is distributed randomly. We assume that both bidders have the same utility function $u(x)$.

Following a standard argument in equilibrium $\beta_k(0) = 0$.³⁵ u is the utility function and we assume bidders to be risk neutral.

To integrate spite into the model we assume that the losing participant gains an additional disutility of α -times the payoff of the winning player. Assume without loss of generality that $b_j > b_i$, i.e. bidder j is the winner. Then $\alpha \cdot (v_j - \beta_j(v_j))$ reflects bidder i 's spite.³⁶ The absence of spite is equivalent $\alpha = 0$. We assume that $\alpha \in [0, 1]$, i.e. we do not consider $\alpha < 0$ which could represent sympathy or profit sharing.

The payoff of player i is as follows:

$$\Phi_{\text{Spite}}^{\text{I-AP}}(\beta_i, v_i) = \begin{cases} u(v_i - \beta_i(v_i)) & \text{if } \beta_i > \beta_j \text{ (i wins)} \\ u\left(\frac{v_i - \beta_i(v_i)}{2}\right) & \text{if } \beta_i = \beta_j \text{ (a tie)} \\ u(-\beta_i(v_i) - \alpha(v_j - \beta_j(v_j))) & \text{if } \beta_i < \beta_j \text{ (j wins)} \end{cases} \quad (16)$$

We follow the standard approach and assume that bidder i with valuation v_i makes a bid b . The expected utility of this bidder is given as follows:

$$\mathbb{E}(b, v) = \underbrace{\int_0^{\beta_j^{-1}(b)} u(v - b) f(v_j) dv_j}_{\text{bidder i wins and obtains the prize}} + \underbrace{\int_{\beta_j^{-1}(b)}^{\bar{v}} u(-b - \alpha(v_j - \beta_j(v_j))) f(v_j) dv_j}_{\text{bidder i loses and pays the own bid and additionally experiences spite}} \quad (17)$$

Rearranging the FOC leads to:

$$\beta'_j(\beta_j^{-1}(b)) = \frac{(u(v - b) - u(-b - \alpha(\beta_j^{-1}(b) - \beta_j(\beta_j^{-1}(b))))f(\beta_j^{-1}(b)))}{\int_0^{\beta_j^{-1}(b)} u(v - b)'f(v_j) dv_j + \int_{\beta_j^{-1}(b)}^{\bar{v}} u(-b - \alpha(v_j - \beta_j(v_j)))'f(v_j) dv_j}$$

Assuming a symmetric equilibrium bidding function, we obtain the following condition:

$$\beta'_j(v) = \frac{((v - b) - (-b - \alpha(v - b)))f(v)}{1} = (1 + \alpha)v f(v) - \alpha f(v)b \quad (18)$$

³⁵We assume a monotonic bidding function. A selfish agent would only win if the opponent has also a valuation of zero. As one cannot influence the payoff of the opponent there is no benefit in bidding above zero as one has to pay for this bid. Hence, zero is the best choice.

³⁶Note that spite is similar but not equivalent to the negative aspect of inequality aversion. The latter would consider relative gain $\alpha \cdot (v_j - \beta_j(v_j) - \beta_i(v_i))$.

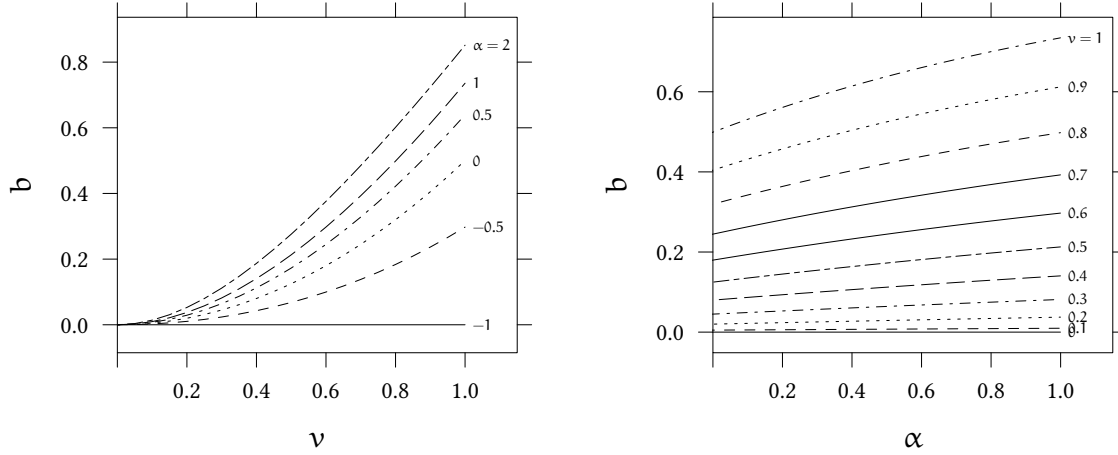


Figure 13: Equilibrium bids in first-price all-pay auctions.

Equilibrium bids in first-price all-pay auctions for different valuations v and different levels of spite α with uniform distributions of valuations (see Equation (21)).

Solving the ODE (18) with the initial value $b(0) = 0$ we obtain the symmetric equilibrium bidding function $b_{\text{Spite}}^{\text{I-AP}}$:

$$b_{\text{Spite}}^{\text{I-AP}}(v) = e^{-\alpha F(v)} \int_0^v (\alpha + 1) s f(s) e^{\alpha F(s)} ds = \frac{\alpha + 1}{\alpha} \left(v - \frac{1}{e^{\alpha F(v)}} \int_0^v e^{\alpha F(s)} ds \right) \quad (19)$$

For $\alpha = 0$, Equation (19) becomes the familiar equilibrium bidding function for all-pay auctions without spite:

$$b^{\text{I-AP}} := b_{\alpha=0}^{\text{I-AP}} = \int_0^v s f(s) ds \quad (20)$$

For uniformly distributed valuations, $F(x) = x$, we have

$$b_{\text{Spite}}^{\text{I-AP}}(v) = \frac{\alpha + 1}{\alpha} \left(v + \frac{e^{-\alpha v} - 1}{\alpha} \right). \quad (21)$$

The left part of Figure 13 illustrates in particular the following:

Proposition 7. *The bidding function in the first-price all-pay auction is increasing in the bidder's valuation:*

$$\frac{db_{\text{Spite}}^{\text{I-AP}}}{dv} \geq 0$$

Proof of Proposition 7:

$$\begin{aligned} \frac{db_{\text{Spite}}^{\text{I-AP}}}{dv} &= \frac{1}{e^{\alpha F(v)}} \left(-\alpha f(v) \int_0^v (\alpha + 1) s f(s) e^{\alpha F(s)} ds + (\alpha + 1) v f(v) e^{\alpha F(v)} \right) \\ &= \frac{(\alpha + 1) f(v)}{e^{\alpha F(v)}} \int_0^v e^{\alpha F(s)} ds \geq 0 \end{aligned}$$

■

The right part of Figure 13 illustrates the following:

Proposition 8. *Bids in the first-price all-pay auction are increasing in spite:*

$$\frac{db_{\text{Spite}}^{\text{I-AP}}}{d\alpha} \geq 0$$

Proof of Proposition 8: We want to show the following:

$$\begin{aligned} \frac{db_{\text{Spite}}^{\text{I-AP}}}{d\alpha} &= \frac{1}{e^{\alpha F(v)}} \left(\int_0^v sf(s) e^{\alpha F(s)} ((\alpha + 1)(F(s) - F(v)) + 1) ds \right) \\ &= \int_0^v sf(s) \underbrace{e^{\alpha(F(s)-F(v))} ((\alpha + 1)(F(s) - F(v)) + 1)}_{:=Q(\alpha, F(v), F(s))} ds \\ &\geq 0 \end{aligned}$$

Let us rewrite $F(v) = w$ and $F(s) = z$ and therefore $F(z)^{-1} = s$ and $F(w)^{-1} = v$ are the case. Hence, $Q(\alpha, w, z) := e^{\alpha(z-w)} ((\alpha + 1)(z - w) + 1)$.

When we now consider the derivative of $Q(\alpha, w, z)$ in α we see that:

$$\begin{aligned} \frac{d(Q(\alpha, w, z))}{d\alpha} &= (-z + w) e^{\alpha(z-w)} (-\alpha z + \alpha w - z + w - 2) \\ &= \underbrace{(z - w)}_{\leq 0} e^{\alpha(z-w)} \underbrace{\left(\underbrace{(\alpha + 1)}_{\leq 2} \underbrace{(z - w)}_{\geq -1} + 2 \right)}_{\geq 0} \\ &= \leq 0 \end{aligned}$$

Also considering the derivative of $Q(\alpha, w, z)$ in w we see that:

$$\begin{aligned} \frac{d(Q(\alpha, w, z))}{dw} &= \left(-1 + (-z + w) \alpha^2 + (-2 - z + w) \alpha \right) e^{\alpha(z-w)} \\ &= -e^{\alpha(z-w)} \left(1 + \alpha \underbrace{\left(\underbrace{(\alpha + 1)}_{\leq 2} \underbrace{(z - w)}_{\geq -1} + 2 \right)}_{\geq 0} \right) \\ &= \leq 0 \end{aligned}$$

Our goal is to show that the following equation holds (which would be Proposition 8)

$$\int_0^{F(w)^{-1}} F(z)^{-1} Q(\alpha, w, z) dz \geq 0$$

Using the derivatives of $Q(\alpha, w, z)$ in w and in α we get:

$$\begin{aligned} \int_0^{F(w)^{-1}} F(z)^{-1} Q(\alpha, w, z) dz &\geq \int_0^{F(w)^{-1}} F(z)^{-1} Q(1, w, z) dz \geq \int_0^{F(1)^{-1}} F(z)^{-1} Q(1, 1, z) dz \\ \int_0^{F(1)^{-1}} F(z)^{-1} Q(1, 1, z) dz &= \int_0^{F(1)^{-1}} F(z)^{-1} e^{z-1} ((2)(z-1) + 1) dz \\ &= \int_0^{F(1)^{-1}} F(z)^{-1} \underbrace{e^{z-1} ((2z-1))}_{=Q(1,1,z)} dz \end{aligned}$$

Considering this we see that this is positive even for the lowest possible values.

$$\int_0^1 Q(1, 1, z) dz = 3 \cdot e^{-1} - 1 \approx 0.1 \geq 0$$

Now we can also show that the function $Q(\alpha, w, z)$ is increasing in z

$$\begin{aligned} \frac{d(Q(\alpha, w, z))}{dz} &= - \left(-1 + (-z + w) \alpha^2 + (-2 - z + w) \alpha \right) e^{\alpha(z-w)} \\ &= e^{\alpha(z-w)} \left(1 + \alpha \underbrace{\left(\underbrace{(\alpha + 1)}_{\leq 2} \underbrace{(z - w) + 2}_{\geq -1} \right)}_{\geq 0} \right) \\ &= \geq 0 \end{aligned}$$

Now as we know that $F(z)^{-1}$ is an increasing function. Moreover, we we know that $\frac{d(Q(\alpha, w, z))}{dz} \geq 0$ and also $\int_0^1 Q(\alpha, w, z) dz \geq \int_0^1 Q(1, 1, z) dz \geq 0$. Thus, we conclude the following:

$$\int_0^{F(w)^{-1}} F(z)^{-1} Q(\alpha, w, z) dz \geq \int_0^{F(1)^{-1}} F(z)^{-1} e^{z-1} ((2)(z-1) + 1) dz \geq 0 \quad \blacksquare$$

In particular bids in the first-price all-pay auction with spite ($\alpha > 0$) are larger than or equal to bids without spite ($\alpha = 0$), i.e. $b_{\text{Spite}}^{\text{I-AP}} - b^{\text{I-AP}} \geq 0$.

Proposition 9. *The difference between equilibrium bids with spite and without spite is increasing in bidder's valuation (for uniform distribution):*

$$\frac{d(b_{\text{Spite}}^{\text{I-AP}} - b^{\text{I-AP}})}{dv} \geq 0$$

Proof of Proposition 9: To prove that the deviation is increasing in valuation in the first-price

winner-pay auction with uniform distribution we can use the result of (7)

$$\begin{aligned}
\frac{d(b_{\text{Spite}}^{\text{I-AP}} - b_{\text{selfish}}^{\text{I-AP}})}{dv} &= \frac{1}{e^{\alpha F(v)}} \left(-\alpha f(v) \int_0^v (\alpha + 1) s f(s) e^{\alpha F(s)} ds + (\alpha + 1) v f(v) e^{\alpha F(v)} \right) - v f(v) \\
&= f(v) \left(\frac{(\alpha + 1)}{e^{\alpha F(v)}} \int_0^v \underbrace{e^{\alpha F(s)}}_{\geq 1} ds - v \right) \\
&= \alpha f(v) \left(\frac{(\alpha + 1)}{e^{\alpha F(v)}} \left(\frac{v e^{\alpha F(v)}}{\alpha} - \int_0^v s f(s) e^{\alpha F(s)} ds \right) - \frac{v}{\alpha} \right) \\
&= \alpha f(v) \left(v - \frac{(\alpha + 1)}{e^{\alpha F(v)}} \int_0^v s f(s) e^{\alpha F(s)} ds \right) \\
&= \frac{\alpha f(v)}{e^{\alpha F(v)}} \left(\underbrace{v e^{\alpha F(v)} - (\alpha + 1) \int_0^v s f(s) e^{\alpha F(s)} ds}_M \right)
\end{aligned}$$

To see that this is indeed positive we show that the derivative of M is positive:

$$\begin{aligned}
\frac{d(M)}{dv} &= e^{\alpha F(v)} + v \alpha f(v) e^{\alpha F(v)} - (1 + \alpha) v f(v) e^{\alpha F(v)} \\
\leftrightarrow &= e^{\alpha F(v)} (1 - v \cdot f(v))
\end{aligned}$$

Obviously $\frac{d(M)}{dv}$ is positive for $f(v) = 1$. We can also easily see that $M(0) = 0$. Therefore, the deviation is increasing in valuation in the first-price winner-pay auction for $\alpha \geq 0$ for uniform distribution. \blacksquare

Proposition 9 is actually quite intuitive. Consider two bidders competing for an object in the first-price winner-pay auction. For a low valuation, the probability of winning is low, too. In this situation, bidders will suffer from spite almost always. As a result, the overall value of the auction is small and bids will be small, too. If, on the other hand, valuations are high bidders will want to avoid the disutility from losing by increasing their bids. As a result, we see that when spite increases also bids increase.

B. Bids in the second-price all-pay auction

Proof of Proposition 1:

$$\begin{aligned}
\frac{db_{\text{Spite}}^{\text{II-AP}}}{dv} &= \underbrace{\frac{1 + \alpha}{1 - \alpha} (1 - F(v))^{\frac{2\alpha - 1}{1 - \alpha}} f(v)}_{q \geq 0} \left(v (1 - F(v))^{\frac{\alpha}{\alpha - 1}} - \frac{\alpha}{1 - \alpha} \int_0^v s f(s) (1 - F(s))^{\frac{1}{\alpha - 1}} ds \right) \\
&= q \left(v (1 - F(v))^{\frac{\alpha}{\alpha - 1}} - v (1 - F(v))^{\frac{\alpha}{\alpha - 1}} + \int_0^v (1 - F(s))^{\frac{\alpha}{\alpha - 1}} ds \right) \\
&\geq 0
\end{aligned}$$

\blacksquare

Proof of Proposition 2:

$$\beta_{\text{Risk}}^{\text{II-AP}}(0) = 0$$

$$\beta_{\text{RNBNE}}^{\text{II-AP}}(0) = 0$$

Let us proof by contradiction. We know that risk averse and risk neutral players start at the same point. We assume for now that risk averse players have a higher slope compared to risk neutral subjects:

$$\beta_{\text{Risk}}^{\text{II-AP}'}(v) \geq \beta_{\text{RNBNE}}^{\text{II-AP}'}(v)$$

$$\frac{r(1 - e^{-\frac{v}{r}})f(v)}{(1 - F(v))} \geq \frac{vf(v)}{(1 - F(v))}$$

$$r(1 - e^{-\frac{v}{r}}) \geq v$$

$$(1 - e^{-\frac{v}{r}}) \geq \frac{v}{r} \quad \text{here we use } r \geq 0$$

$$(e^{\frac{v}{r}} - 1) \geq e^{\frac{v}{r}} \frac{v}{r}$$

$$(e^m - 1) \geq^{**} e^m m \quad \text{we substitute } \frac{v}{r} = m$$

$$\underbrace{e^m(1 - m) - 1}_{L(m)} \geq 0$$

We can show that $L(m)$ is decreasing ($\frac{\partial L(m)}{\partial m} = -me^m$) in m and as $L(0) = 0$ we obtain a contradiction as $e^m(1 - m) - 1 \leq 0 \forall m \in \mathbb{R}_+$. ■

C. Revenue

We have seen that the introduction of spite could explain overbidding in all-pay auctions. In addition, it would be interesting to see some results on revenue ranking in case of spiteful bidders.

In this paper we are only looking at two players and therefore the revenue of the seller is just two times the ex-ante expected payment, which is the bid multiplied by the probability to pay the bid:

$$\text{Expected payment : } m(v) = \text{Bid}(v) \cdot \text{Prob}(\text{Paying Bid})$$

$$\text{Ex-ante expected payment : } \mathbb{E}[m(v)] := \text{expected revenue for one player}$$

$$= \frac{1}{2} \cdot \text{expected revenue for the seller}$$

Now we can look at the revenues of the all-pay auctions with spite:

C.1. Revenue in the first-price all-pay auction

Proposition 10. *The revenue in the first-price all-pay auctions with two players is given by:*

$$R_{\text{Spite}}^{\text{I-AP}} = 2\mathbb{E}(m_{\text{Spite}}^{\text{I-AP}}) = 2 \int_0^1 \frac{(e^{-\alpha F(s)} - e^{-\alpha})}{\alpha} (\alpha + 1) s f(s) e^{\alpha F(s)} ds \quad (22)$$

The revenue in the second-price all-pay auctions with two players is given by:

$$R_{\text{Spite}}^{\text{II-AP}} = 2\mathbb{E}(m_{\text{Spite}}^{\text{II-AP}}) = 2 \int_0^1 2 \frac{\alpha + 1}{2 - \alpha} s f(s) (1 - F(s)) ds \quad (23)$$

Proof of Proposition 10:

$$\begin{aligned} R_{\text{Spite}}^{\text{I-AP}} = 2\mathbb{E}(m_{\text{Spite}}^{\text{I-AP}}) &= 2 \int_0^1 b_{\text{Spite}}^{\text{I-AP}} f(v) dv \\ &= 2 \int_0^1 f(v) e^{-\alpha F(v)} \int_0^v (\alpha + 1) s f(s) e^{\alpha F(s)} ds dv \\ &= 2 \int_0^1 \int_s^1 f(v) e^{-\alpha F(v)} dv (\alpha + 1) s f(s) e^{\alpha F(s)} ds \\ (\text{for } \alpha \neq 0 :) &= 2 \int_0^1 \frac{e^{-\alpha F(v)}}{-\alpha} \Big|_s^1 (\alpha + 1) s f(s) e^{\alpha F(s)} ds \\ &= 2 \int_0^1 \frac{(e^{-\alpha F(s)} - e^{-\alpha})}{\alpha} (\alpha + 1) s f(s) e^{\alpha F(s)} ds \end{aligned}$$

$$\begin{aligned} R_{\text{Spite}}^{\text{II-AP}} = 2\mathbb{E}(m_{\text{Spite}}^{\text{II-AP}}) &= 2 \int_0^1 b_{\text{Spite}}^{\text{II-AP}} (2(1 - F(v))) f(v) dv \\ &= 2 \int_0^1 \frac{\alpha + 1}{1 - \alpha} \int_0^v s f(s) (1 - F(s))^{\frac{1}{\alpha-1}} ds (1 - F(v))^{\frac{\alpha}{1-\alpha}} (2(1 - F(v))) f(v) dv \\ &= 2 \int_0^1 \int_s^1 (1 - F(v))^{\frac{1}{1-\alpha}} 2f(v) dv \frac{\alpha + 1}{1 - \alpha} s f(s) (1 - F(s))^{\frac{1}{\alpha-1}} ds \\ &= 2 \int_0^1 (1 - F(v))^{\frac{2-\alpha}{1-\alpha}} 2 \frac{1 - \alpha}{\alpha - 2} \Big|_s^1 \frac{\alpha + 1}{1 - \alpha} s f(s) (1 - F(s))^{\frac{1}{\alpha-1}} ds \\ &= 2 \int_0^1 \left((1 - F(s))^{\frac{2-\alpha}{1-\alpha}} 2 \frac{1 - \alpha}{2 - \alpha} \right) \frac{\alpha + 1}{1 - \alpha} s f(s) (1 - F(s))^{\frac{1}{\alpha-1}} ds \\ &= 2 \int_0^1 2 \frac{\alpha + 1}{2 - \alpha} s f(s) (1 - F(s)) ds \end{aligned}$$

■

C.2. The impact of spite on revenue

Proposition 11. *Spite increases revenue in case of the first-price and second-price all-pay auction.*

$$\mathbb{E}(m_{\text{Spite}}^{\text{I-AP}}) \geq \mathbb{E}(m_{\text{selfish}}) \quad (24)$$

$$\mathbb{E}(m_{\text{Spite}}^{\text{II-AP}}) \geq \mathbb{E}(m_{\text{selfish}}) \quad (25)$$

Proof of Proposition 11: We first compare the revenues of spiteful bidders in first-price all-pay auctions and standard-selfish revenue

$$\begin{aligned} R_{\text{Spite}}^{\text{I-AP}} - R^{\text{selfish}} &= \\ 2\mathbb{E}(m_{\text{Spite}}^{\text{I-AP}} - m_{\text{selfish}}) &= 2 \frac{\alpha + 1}{\alpha} \int_0^1 s f(s) (1 - e^{-\alpha} e^{\alpha F(s)}) ds - 2 \int_0^1 s f(s) (1 - F(s)) ds \\ &= 2 \int_0^1 s f(s) \left(\frac{\alpha + 1}{\alpha} - \frac{\alpha + 1}{\alpha} e^{\alpha(F(s)-1)} - 1 + F(s) \right) ds \\ &= 2 \int_0^1 s f(s) \left(\frac{1}{\alpha} - \frac{\alpha + 1}{\alpha} e^{\alpha(F(s)-1)} + F(s) \right) ds \end{aligned}$$

It is quite straightforward that this difference is positive as we know that $b_{\text{Spite}}^{\text{I-AP}} \geq b^{\text{I-AP}}$. To see this:

$$\begin{aligned} b_{\text{Spite}}^{\text{I-AP}} &\geq b^{\text{I-AP}} \\ \rightarrow f(v) \cdot b_{\text{Spite}}^{\text{I-AP}} &\geq f(v) \cdot b^{\text{I-AP}} \\ \rightarrow \int_0^1 f(v) \cdot b_{\text{Spite}}^{\text{I-AP}} dv &\geq \int_0^1 f(v) \cdot b^{\text{I-AP}} dv \\ \rightarrow \mathbb{E}(m_{\text{Spite}}^{\text{I-AP}}) &\geq \mathbb{E}(m_{\text{selfish}}) \\ \rightarrow 2\mathbb{E}(m_{\text{Spite}}^{\text{I-AP}}) &\geq 2\mathbb{E}(m_{\text{selfish}}) \\ R_{\text{Spite}}^{\text{I-AP}} &\geq R^{\text{selfish}} \end{aligned}$$

We now compare the revenues of spiteful bidders in second-price all-pay auctions and standard-selfish revenue

$$\begin{aligned} \mathbb{E}(m_{\text{Spite}}^{\text{II-AP}} - m_{\text{selfish}}) &= 2 \int_0^1 2 \frac{\alpha + 1}{2 - \alpha} s f(s) (1 - F(s)) ds - 2 \int_0^1 s f(s) (1 - F(s)) ds \\ &= 2 \int_0^1 s f(s) (1 - F(s)) \left(2 \frac{\alpha + 1}{2 - \alpha} - 1 \right) ds \end{aligned}$$

As we are interested just whether the difference is positive we check only: $2 \frac{\alpha + 1}{2 - \alpha} - 1 \geq 0 \leftrightarrow 2\alpha + 2 \geq 2 - \alpha \leftrightarrow 3\alpha \geq 0$. Obviously, $\mathbb{E}(m_{\text{Spite}}^{\text{II-AP}} - m_{\text{selfish}})$ is positive for all $\alpha \geq 0$.

$$\Rightarrow R_{\text{Spite}}^{\text{II-AP}} \geq R^{\text{selfish}} \quad \forall \alpha \geq 0 \quad \blacksquare$$

C.3. Revenue with spite under the different auction formats

We have seen that the revenues are higher if bidders are experiencing spite relative to the selfish case. In the following we want to rank revenues for the different auction formats (first-price and second-price winner-pay and all-pay auctions, respectively) for spiteful bidders.

Proof of Proposition 5: We will prove the different parts of Proposition 5 one after the other. In Lemma 1 we will show that $\mathbb{E}(m_{\text{Spite}}^{\text{II-AP}}(v)) \geq \mathbb{E}(m_{\text{Spite}}^{\text{I-AP}}(v))$. In Lemma 2 we show that $\mathbb{E}(m_{\text{Spite}}^{\text{I-AP}}(v)) \geq \mathbb{E}(m_{\text{Spite}}^{\text{I}}(v))$. Finally we will show in Lemma 3 that $\mathbb{E}(m_{\text{Spite}}^{\text{I}}(v)) \geq \mathbb{E}(m_{\text{selfish}}(v))$ which completes the proof of Proposition 5.

Let us first look at the revenues of the all-pay auctions.

Lemma 1. *The expected revenue of the second-price all-pay auction in case of spiteful bidders is higher than the expected revenue of the first-price all-pay auction with spiteful bidders.*

$$\mathbb{E}(m_{\text{Spite}}^{\text{I-AP}}) \leq \mathbb{E}(m_{\text{Spite}}^{\text{II-AP}}) \quad (26)$$

Proof of Lemma 1:

$$\begin{aligned} \mathbb{E}(m_{\text{Spite}}^{\text{I-AP}} - m_{\text{Spite}}^{\text{II-AP}}) &= \int_0^1 \frac{(e^{-\alpha F(s)} - e^{-\alpha})}{\alpha} (\alpha + 1) s f(s) e^{\alpha F(s)} ds - \int_0^1 2 \frac{\alpha + 1}{2 - \alpha} s f(s) (1 - F(s)) ds \\ &= (\alpha + 1) \int_0^1 s f(s) \left(\frac{1}{\alpha} - \frac{1 e^{\alpha F(s)}}{\alpha e^\alpha} - \frac{2}{2 - \alpha} (1 - F(s)) \right) ds \\ &= (\alpha + 1) \int_0^1 s f(s) \left(\frac{(2)e^\alpha - 3\alpha e^\alpha - 2e^{\alpha F(s)} + e^{\alpha F(s)} \alpha + 2\alpha F(s) e^\alpha}{(2 - \alpha)(\alpha) e^\alpha} \right) ds \\ &= (\alpha + 1) \int_0^1 s f(s) \left(\frac{(\alpha - 2)e^{\alpha F(s)} + 2(1 + (F(s) - \frac{3}{2})\alpha) e^\alpha}{(2 - \alpha)(\alpha) e^\alpha} \right) ds \\ &= (\alpha + 1) \int_0^1 s f(s) \underbrace{\left(\frac{(\alpha - 2)e^{\alpha(F(s)-1)} + 2(1 + (F(s) - \frac{3}{2})\alpha)}{(2 - \alpha)(\alpha)} \right)}_{: = n(F(s), \alpha)} ds \end{aligned}$$

If we could show that $n(F(s), \alpha) \leq 0$ it would be obvious that the difference is negative. Therefore, we show that this is negative by contradiction.

$$\begin{aligned} n(F(s), \alpha) &> 0 \rightarrow \frac{(\alpha - 2)e^{\alpha(F(s)-1)} + 2(1 + (F(s) - \frac{3}{2})\alpha)}{(2 - \alpha)(\alpha)} > 0 \\ \Leftrightarrow \frac{(\alpha - 2)}{2} &> -e^{-\alpha(F(s)-1)} (1 + (F(s) - \frac{3}{2})\alpha) \\ \rightarrow \frac{(\alpha - 2)}{2} e^{-1 + \frac{1}{2}\alpha} &> -e^{(1 + (F(s) - \frac{3}{2})\alpha)} (1 + (F(s) - \frac{3}{2})\alpha) \\ \Leftrightarrow \mathcal{W}\left(\frac{(\alpha - 2)}{2} e^{-1 + \frac{1}{2}\alpha}\right) &> -(1 + (F(s) - \frac{3}{2})\alpha) \\ \Leftrightarrow F(s) &> \frac{3}{2} - \frac{1}{\alpha} - \frac{\mathcal{W}\left(\frac{(\alpha - 2)}{2} e^{\frac{(\alpha - 2)}{2}}\right)}{\alpha} \\ \Leftrightarrow F(s) &> \frac{3}{2} - \frac{1}{\alpha} - \frac{(\alpha - 2)}{2\alpha} = 1 \end{aligned}$$

Whereas $\mathcal{W}(\cdot)$ represents the Lambert W-function. ■

Next we are going to look at the expected revenues of the winner-pay vs. all-pay auction:

Lemma 2. *Expected revenue with the first-price all-pay auction is higher than in the first-price winner-pay auction if bidders are spiteful.*

$$\mathbb{E}(m_{\text{Spite}}^{\text{I-AP}}) \geq \mathbb{E}(m_{\text{Spite}}^{\text{I}}) \quad (27)$$

Expected revenue with the second-price all-pay auction is higher than in the first-price winner-pay auction if bidders are spiteful.

$$\mathbb{E}(m_{\text{Spite}}^{\text{II-AP}}) \geq \mathbb{E}(m_{\text{Spite}}^{\text{I}}) \quad (28)$$

Proof of Lemma 2:

$$\begin{aligned} R_{\text{Spite}}^{\text{I}} - R_{\text{Spite}}^{\text{I-AP}} &= 2\mathbb{E}(m_{\text{Spite}}^{\text{I}} - m_{\text{Spite}}^{\text{I-AP}}) \\ &= 2\frac{1+\alpha}{1-\alpha} \int_0^1 (s f(s)(F(s)^\alpha - F(s)) \\ &\quad - 2 \int_0^1 \frac{(e^{-\alpha F(s)} - e^{-\alpha})}{\alpha} (\alpha+1) s f(s) e^{\alpha F(s)} ds \\ &= 2(1+\alpha) \int_0^1 s f(s) \underbrace{\left(\frac{F(s)^\alpha - F(s)}{1-\alpha} - \frac{1 - e^{\alpha(F(s)-1)}}{\alpha} \right)}_{:=g(F(s),\alpha)} ds \end{aligned}$$

Let us now set $g(F(s), \alpha) := g(k, \alpha)$:

$$\begin{aligned} \frac{dg(k, \alpha)}{dk} &= \frac{\frac{\alpha k^\alpha}{k} - 1}{1-\alpha} + e^{\alpha(k-1)} \\ \frac{\frac{dg(k, \alpha)}{dk}}{dk} &= \alpha \left(e^{\alpha(k-1)} - k^{\alpha-2} \right) \end{aligned}$$

Now we want to prove that the second derivative of $g(k, \alpha)$ is negative. Hence, we assume the opposite.

$$\begin{aligned} \alpha \left(e^{\alpha(k-1)} - k^{\alpha-2} \right) &> 0 \\ \rightarrow e^{\alpha(k-1)} k^{2-\alpha} &> 1 \\ \rightarrow e^{\alpha(k-1)} k > e^{\alpha(k-1)} k^{2-\alpha} &> 1 \\ \rightarrow e^{\alpha(k)} k \alpha &> e^\alpha \alpha \\ \rightarrow k \alpha &> \mathcal{W}(e^\alpha \alpha) = \alpha \\ \rightarrow k &> 1 \end{aligned}$$

Whereas $\mathcal{W}(\cdot)$ represents the Lambert W-function.

Thus, we know that $\max(k) = 1$ and hence $k \not> 1$ and hence we know that $\frac{dg(k, \alpha)}{dk} \leq 0$.

$$\begin{aligned} \left(\frac{dg(k, \alpha)}{dk} \right) \leq 0 &\rightarrow \min\left(\frac{dg(k, \alpha)}{dk}\right) = \frac{dg(k, \alpha)}{dk} \Big|_{k=1} = 0 \\ \Rightarrow \frac{dg(k, \alpha)}{dk} \geq 0 &\rightarrow \max(g(k, \alpha)) = g(k, \alpha) \Big|_{k=1} = 0 \\ \Rightarrow g(k, \alpha) \leq 0 \end{aligned}$$

Therefore :

$$2(1 + \alpha) \int_0^1 s f(s) \underbrace{\left(\frac{F(s)^\alpha - F(s)}{1 - \alpha} - \frac{1 - e^{\alpha(F(s)-1)}}{\alpha} \right)}_{=g(s,\alpha) \leq 0} ds \leq 0$$

$$\Rightarrow \mathbb{E}(m_{\text{Spite}}^I - m_{\text{Spite}}^{\text{I-AP}}) \leq 0 \Rightarrow R_{\text{Spite}}^{\text{I-AP}} \geq R_{\text{Spite}}^I$$

The second part of the lemma is just straightforward, as we know due to Lemma 1 that the revenue of the second-price all-pay auction is higher than the revenue of the first-price all-pay auction. Thus, we can easily conclude that

$$\mathbb{E}(m_{\text{Spite}}^{\text{II-AP}}) \geq \mathbb{E}(m_{\text{Spite}}^I) \Rightarrow R_{\text{Spite}}^{\text{II-AP}} \geq R_{\text{Spite}}^I \quad \blacksquare$$

Lemma 3. *Spite increases revenue in case of the first-price winner-pay auction.*

$$\mathbb{E}(m_{\text{Spite}}^{\text{I-WP}}) \geq \mathbb{E}(m_{\text{selfish}}) \quad (29)$$

Proof of Lemma 3: This Lemma is shown in Morgan et al. (2003) as part of Proposition 4.4 and is proved in Appendix A.4.

Summarizing our results we have seen that in theory spiteful bidders would bid more than selfish bidders. Now, we have seen that this behavior would lead to more revenue for the seller. Also, we have seen that the revenue equivalence principle is not applicable with spiteful bidders, and the result is that all-pay auctions are yielding higher revenue than the first-price winner-pay auction. Moreover, we have seen that the second-price all-pay auctions yield even higher revenue than the first-price all-pay auction.

C.4. Revenue with risk aversion

Proof of Proposition 6: The proof follows easily from the following two lemmas:

Lemma 4. *A seller's revenue in the second-price all-pay auction is lower if bidders are risk averse.*

Lemma 5. *A seller's revenue in the first-price winner-pay auction is higher if bidders are risk averse.*

The first Lemma is obvious as we already have shown that risk averse subjects are underbidding in the second-price all-pay auction and hence, a bidder has a smaller expected payment and hence, a seller has a smaller expected revenue.

The second Lemma is also straightforward as a seller has a higher expected revenue if bidders overbid (see Riley and Samuelson, 1981, Proposition 4). \blacksquare

D. Measuring preferences for risk and spitefulness

D.1. Risk preferences

The lotteries for the Holt and Laury (2002) task are shown in Table 4. Details of the implementation are illustrated in Appendix F.1, Second Task (B).

Lottery A	Lottery B
In 1 out of 10 cases you will earn 1800 points and in 9 out of 10 cases you will earn 1440 points	In 1 out of 10 cases you will earn 3465 points and in 9 out of 10 cases you will earn 90 points
In 2 out of 10 cases you will earn 1800 points and in 8 out of 10 cases you will earn 1440 points	In 2 out of 10 cases you will earn 3465 points and in 8 out of 10 cases you will earn 90 points
In 3 out of 10 cases you will earn 1800 points and in 7 out of 10 cases you will earn 1440 points	In 3 out of 10 cases you will earn 3465 points and in 7 out of 10 cases you will earn 90 points
In 4 out of 10 cases you will earn 1800 points and in 6 out of 10 cases you will earn 1440 points	In 4 out of 10 cases you will earn 3465 points and in 6 out of 10 cases you will earn 90 points
In 5 out of 10 cases you will earn 1800 points and in 5 out of 10 cases you will earn 1440 points	In 5 out of 10 cases you will earn 3465 points and in 5 out of 10 cases you will earn 90 points
In 6 out of 10 cases you will earn 1800 points and in 4 out of 10 cases you will earn 1440 points	In 6 out of 10 cases you will earn 3465 points and in 4 out of 10 cases you will earn 90 points
In 7 out of 10 cases you will earn 1800 points and in 3 out of 10 cases you will earn 1440 points	In 7 out of 10 cases you will earn 3465 points and in 3 out of 10 cases you will earn 90 points
In 8 out of 10 cases you will earn 1800 points and in 2 out of 10 cases you will earn 1440 points	In 8 out of 10 cases you will earn 3465 points and in 2 out of 10 cases you will earn 90 points
In 9 out of 10 cases you will earn 1800 points and in 1 out of 10 cases you will earn 1440 points	In 9 out of 10 cases you will earn 3465 points and in 1 out of 10 cases you will earn 90 points
In 10 out of 10 cases you will earn 1800 points and in 0 out of 10 cases you will earn 1440 points	In 10 out of 10 cases you will earn 3465 points and in 0 out of 10 cases you will earn 90 points

Table 4: Choices in the Holt and Laury (2002) task.

D.2. Spitefulness – Marcus et al. (2014)

The measure of Marcus et al. (2014) is based on a rating of 17 statements. Participants are asked to indicate their agreement on a scale between 1 and 5. Higher scores on the scale indicate more spitefulness. Marcus et al. (2014) propose to use the average of the stated agreements as a measure for spitefulness.

- I would be willing to take a punch if it meant that someone I did not like would receive two punches.
- I would be willing to pay more for some goods and services if other people I did not like had to pay even more.
- If I was one of the last students in a classroom taking an exam and I noticed that the instructor looked impatient, I would be sure to take my time finishing the exam just to irritate him or her.
- If my neighbor complained about the appearance of my front yard, I would be tempted to make it look worse just to annoy him or her.
- It might be worth risking my reputation in order to spread gossip about someone I did not like.
- If I am going to my car in a crowded parking lot and it appears that another driver wants my parking space, then I will make sure to take my time pulling out of the parking space.
- I hope that elected officials are successful in their efforts to improve my community even if I opposed their election. (reverse scored)
- If my neighbor complained that I was playing my music too loud, then I might turn up the music even louder just to irritate him or her, even if meant I could get fined.
- I would be happy receiving extra credit in a class even if other students received more points than me. (reverse scored)
- Part of me enjoys seeing the people I do not like fail even if their failure hurts me in some way.
- If I am checking out at a store and I feel like the person in line behind me is rushing me, then I will sometimes slow down and take extra time to pay.
- It is sometimes worth a little suffering on my part to see others receive the punishment they deserve.
- I would take on extra work at my job if it meant that one of my co-workers who I did not like would also have to do extra work.
- If I had the opportunity, then I would gladly pay a small sum of money to see a classmate who I do not like fail his or her final exam.
- There have been times when I was willing to suffer some small harm so that I could punish someone else who deserved it.
- I would rather no one get extra credit in a class if it meant that others would receive more credit than me.
- If I opposed the election of an official, then I would be glad to see him or her fail even if their failure hurt my community.

D.3. Spitefulness – Own Measure

Our own spite measure is assessing spite similar to the social value orientation task of Murphy et al. (2011) and Murphy and Ackerman (2014). In their slider task participants are presented with 6 (or 15, if inequality aversion is also measured) sets of allocations. Each set contains 9 allocations. Each allocation determines the own payoff and the payoff of the other participant. Participants have to choose a preferred allocation for each set.

Similarly, our spite measure uses six sets of allocations. As in Murphy et al. (2011) and Murphy and Ackerman (2014), each set contains 9 allocations. An overview of the six sets is shown in Figure 5.

The leftmost allocation is always the non spiteful allocation and the rightmost allocation is always the maximally spiteful allocation. In the experiment each set was shown on a separate screen. Two sets were presented in reverse order.

Each of the six tasks is supposed to measure one feature of spite. The sets IA1 and IA2 are measuring spite when it is behaviorally in line with inequality aversion. A decision maker with positive concerns for social efficiency would choose the allocation with the highest payoff for the other player since this choice also maximizes the own payoff. A spiteful person but also an inequality averse person would choose possibly a different allocation. In IA1 being spiteful has no cost. Decision makers get 70 ECU for sure and can basically reduce the payoff of the opponent. In IA2 spitefulness has a cost. In both IA1 and IA2 it may be that the motivation of the decision maker of not maximizing the payoff of the other player could be to either harm the other (spite) or to decrease the overall inequality.

RG1 and RG2 are measuring spite when spite is behaviorally in line with relative gain. Again, a decision maker with positive concerns for social efficiency would choose the allocation with the highest payoff for the other player since this choice also maximizes the own payoff. In RG1 a person who deviates from this choice is considered spiteful as this person decreases the payoff of the opponent. However, this behavior would also be in line with the behavior of an agent who wants to have relatively better payoff compared to the opponent (which is often considered spite). RG2 is a variant of RG1 where the spiteful choice is costly.

In PS1 and PS2 the efficient outcome implies already a positive relative standing of the decision maker who can only decrease the payoff of the other player. We take the last two sets as extreme spite. PS2 is a variant of PS1 where the spiteful choice is costly.

The allocation of the overall spite in this measure can be seen in Figure 4 (on the right).

The decisions of the individual set can be seen in Table 5.

D.4. Spitefulness – Kimbrough and Reiss (2012)

In the original paper by Kimbrough and Reiss (2012) participants were matched into groups of three and played 16 rounds of a second-price winner-pay auction. Participants would bid for an object for which they had an individual induced value $v \sim U[500, 1000]$. After the auction participants did a real effort task. Thereafter, participants learned whether they had won or lost the auction. In a next (and crucial) step, participants could increase their bid from the earlier auction. They also had a possibility to buy the object they were competing for at a random price $p \sim U[300, 500]$ if they lost.

Submeasures	No Spite in %	Spite in %	Average Spite
IA	84.00	16.00	3.17
IA-WP	91.00	9.00	1.36
RG	97.00	3.00	0.19
RG-WP	95.00	5.00	0.77
PS	96.00	4.00	0.42
PS-WP	96.00	4.00	0.31
Σ	82.00	18.00	4.87

Table 5: The allocation of choices considered (non-)spiteful in the six allocational tasks of our own spite measure.

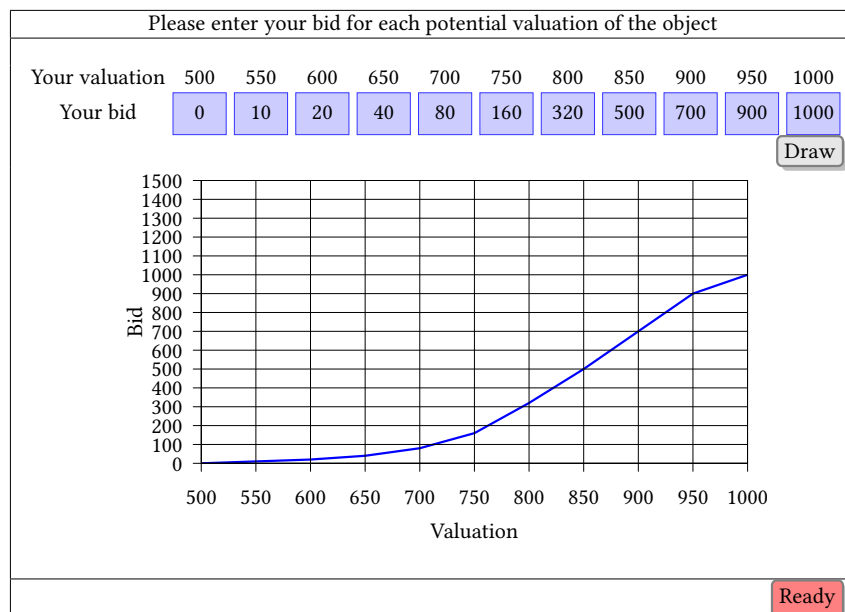


Figure 14: Interface of the Kimbrough and Reiss (2012)-spite measure. Imputing the bidding function for the possible valuations between 500 and 1000. The bidding function is drawn after the input of the respective bids.

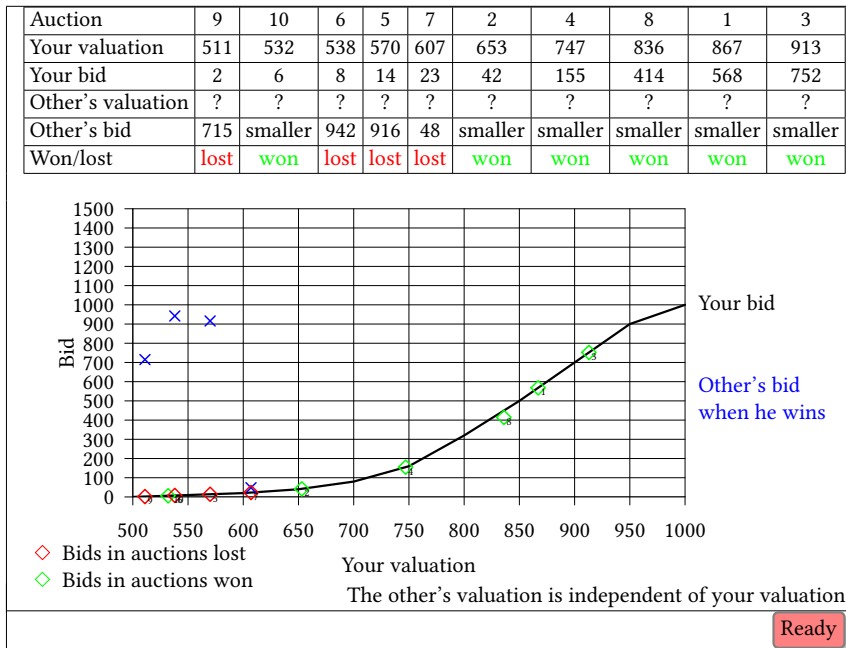


Figure 15: Interface of the feedback of each auction.

Mapping the 10 random valuations and the respective bids on the bidding function. Additionally subjects could see the opponent's bid (if the opponent won) and whether they won or lost.

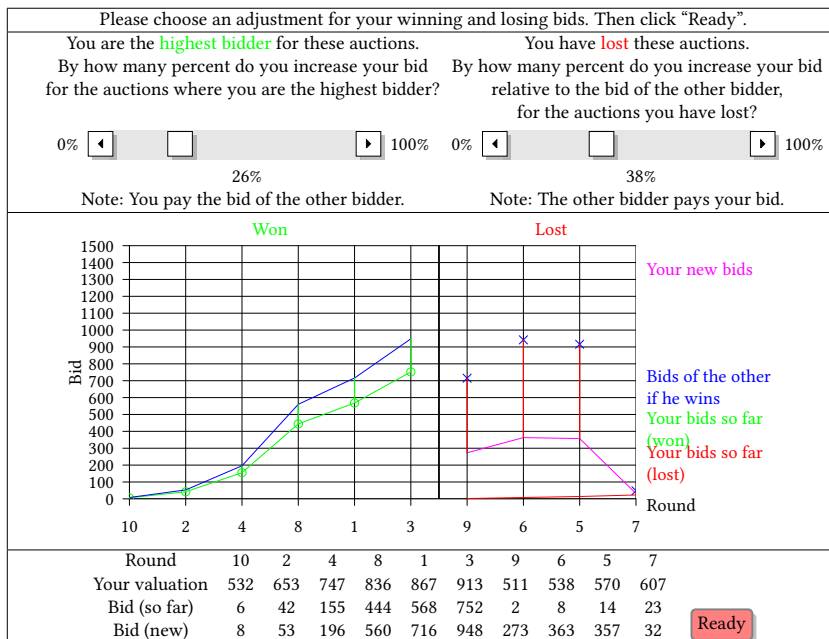


Figure 16: Interface of the bid adaptation.

To reduce the demand effect participants were allowed to increase their losing but also the winning bid. Auctions were ordered so that participants made decisions for auctions they had won in the left part of the screen and for auctions they had lost in the right part.

We change some aspects of Kimbrough and Reiss (2012)'s design. We excluded the outside option. We also excluded the real effort task. We also use the strategy method to elicit one bid function for all auctions. Furthermore, we measure the willingness to pay for the adaptation of the bid. All in all, our measure consists of the following four stages:

Stage 1 Participants ($I_{\{1,2\}}$) submit an initial bid function ($B_{I_{\{1,2\}}}(v)$). Here we use the strategy method (see Figure 14). We present participants with the possible valuations between 500 and 1000 in steps of 50. They were asked to indicate their bid for each valuation .

Stage 2 10 random valuations with $v \sim U[500, 1000]$ were drawn for each participant. For each participant we use their bid function to determine the bid for each valuation. Each valuation and bid of each pair represents one auction. Participants were then informed about the highest bid and the winner in each of the 10 auctions (see Figure 15).

Stage 3 Participants were asked separately for the auctions they had lost and for the auctions they had won by how much they wanted to increase their bids. They could increase their bids by any percentage between 0 and 100% of the difference between winning and losing bid. Hence, the outcome of the auction could not be affected by the final bids. In any case, the initial bids still determined who had won which auction. The final bids only determined how much the winner needed to pay. The interface is shown in Figure 16.³⁷

Stage 4 We essentially use a second-price winner-pay auction to elicit the individual willingness to pay for the adjustment from Stage 3. Participants were randomly matched into pairs with a new partner. They were asked to state how much they were willing to pay for their final bid to be implemented. For each pair the final bids of the person who stated a higher willingness to pay were implemented. That participants had to pay the willingness to pay of their partner from stage 4. Since we use a second-price winner-pay auction it is a dominant strategy for participants to reveal the true willingness to pay for the adjustment of bids. Here, we do not use this data as this stage is arguable rather complicated for subjects to grasp.

E. Further regressions

E.1. Estimating bidding behavior in the second-price all-pay auction

In the main part of the paper we estimated the overbidding behavior for the second-price all-pay auction. In this subsection we will estimate the bidding behavior directly. To estimate the bidding behavior we will use a mixed-effects model, as spite, risk, social value orientation (SVO) etc. are fixed effects but the individuals and the matching-group are random effects. In

³⁷An indicator, that there may be a demand effect existing (or participants did not fully understand this part) is that 67% of the participants increased their bid also in the winning case. It may also be, that participants wanted to ensure that they won or they experienced joy of winning—but in any case behavior can be driven by these motivations only if the task is not completely understood.

line with the overbidding, we expect increased spite will lead to higher bids for intermediate valuations. We compare five different models which differ only in the controls C_1, \dots, C_5 .

$$\begin{aligned} \text{Bid}_{i,t,j,v} &= \beta_0 + \beta_1 \text{Period} + \beta_2 v_{[0,50]} + \beta_3 v_{[50,100]} + \zeta_{i,j} + \eta_j + \epsilon_{i,j,k,l} + C_M & (30) \\ C_1 &= 0 \\ C_2 &= \beta_4 \text{Spite}_i + \beta_5 \text{Spite}_i \times v_{[0,50]} + \beta_6 \text{Spite}_i \times v_{[50,100]} \\ C_3 &= C_2 + \beta_7 \mathbb{1}_{\text{Gender}=\text{♀}} + \beta_8 \text{Risk}_i + \beta_9 \text{rivalry}_i + \beta_{10} \text{SVO}_i + \beta_{11} \text{IA}_i \\ C_4 &= \beta_{12} \text{Risk}_i + \beta_{13} \text{Risk}_i \times v_{[0,50]} + \beta_{14} \text{Risk}_i \times v_{[50,100]} \\ C_5 &= C_4 + \beta_{15} \mathbb{1}_{\text{Gender}=\text{♀}} + \beta_{16} \text{Spite}_i + \beta_{17} \text{rivalry}_i + \beta_{18} \text{SVO}_i + \beta_{19} \text{IA}_i \end{aligned}$$

where $\zeta_{i,j}$ is a random effect for bidder i in group j , η_j is a random effect for group j , and $\epsilon_{i,j,k,l}$ is the residual. $v_{[0,50]}(v)$ and $v_{[50,100]}(v)$ are defined in Equation (14) and (15) above.

	C_1	C_2	C_3	C_4	C_5
Period	−0.40*** (0.05)	−0.40*** (0.05)	−0.40*** (0.05)	−0.40*** (0.05)	−0.40*** (0.05)
$v_{[0,50]}$	30.63*** (0.86)	30.63*** (0.86)	30.63*** (0.86)	30.63*** (0.86)	30.63*** (0.86)
$v_{[50,100]}$	64.43*** (0.78)	64.43*** (0.78)	64.43*** (0.78)	64.43*** (0.78)	64.43*** (0.78)
Spite		2.57 (1.71)	3.34 ⁺ (2.02)		4.10* (2.00)
Spite $\times v_{[0,50]}$		1.49** (0.47)	1.49** (0.47)		
Spite $\times v_{[50,100]}$		0.31 (0.43)	0.31 (0.43)		
Risk			−6.02* (2.99)	−4.36 (3.12)	−3.62 (3.03)
Risk $\times v_{[0,50]}$				−3.48*** (0.86)	−3.48*** (0.86)
Risk $\times v_{[50,100]}$				−3.02*** (0.78)	−3.02*** (0.78)
Male			−19.05** (6.27)		−19.05** (6.27)
Rivalry			−0.70 (3.17)		−0.70 (3.17)
SVO			0.41 ⁺ (0.25)		0.41 ⁺ (0.25)
IA			−1.84 (2.58)		−1.84 (2.58)
Constant	28.16*** (3.21)	28.13*** (3.17)	28.07*** (6.67)	28.11*** (3.16)	28.07*** (6.67)
Observations	23,760	23,760	23,760	23,760	23,760
Log Likelihood	−120,475.50	−120,466.70	−120,449.40	−120,458.90	−120,442.80
Akaike Inf. Crit.	240,965.00	240,953.30	240,928.90	240,937.90	240,915.60
Bayesian Inf. Crit.	241,021.50	241,034.10	241,050.00	241,018.60	241,036.70

Notes:

p : *** < .001 ** < .01 * < .05⁺ < .10

Table 6: Estimation of Equation (30).

Spite is the sum of the three (normalized) spite measures. IA is the sum of the (normalized) inequality aversion score obtained from the slider measure and the (normalized) score obtained from inequality allocation of our own spite measure.

Estimation results are shown in Table 6. It can be seen that spite has a significant positive effect on the bidding behavior for intermediate valuations. This is in line with theory: with increasing spite one would find more overbidding for low valuations. For high valuations ($v \in [50, 100]$) spite has no significant effect.

Concerning risk, we can see that increasing risk aversion also leads to lower bids for intermediate valuations and even stronger decrease in bids for high valuations. This is also in line with theory on risk averse bidding.

Obviously valuations also have a significant and positive influence on bids. Furthermore, female bidders bid more than men. The decrease in bidding over the rounds could be interpreted as a learning effect of overbidding.

Result 4.1. *Spite has a significant positive effect on bids for intermediate valuations.*

Result 4.2. *Risk has a significant negative effect on bids.*

E.2. Estimation results for Equation (13) with alternative spline implementations

Table 8, 9 and 10 show the estimation results for Equation (13) with B-Splines, cyclic cubic splines and P-splines. Table 7 shows also the estimation results for Equation (13), however, instead of using a spline we use piece-wise linear splines. It is evident that the results are robust to these alternative implementations.

	C'_1	C'_2	C'_3	C'_4	C'_5
Period	-0.40*** (0.05)	-0.40*** (0.05)	-0.40*** (0.05)	-0.40*** (0.05)	-0.40*** (0.05)
Valuation _[0,25]	13.59*** (1.14)	13.59*** (1.14)	13.59*** (1.14)	13.59*** (1.14)	13.59*** (1.14)
Valuation _[25,50]	14.24*** (0.95)	14.24*** (0.95)	14.24*** (0.95)	14.24*** (0.95)	14.24*** (0.95)
Valuation _[50,75]	-16.50*** (1.02)	-16.50*** (1.02)	-16.50*** (1.02)	-16.50*** (1.02)	-16.50*** (1.02)
Valuation _[75,100]	-95.69*** (1.03)	-95.69*** (1.03)	-95.69*** (1.03)	-95.69*** (1.03)	-95.69*** (1.03)
Spite	3.33* (1.69)	2.57 (1.71)	3.34 ⁺ (2.02)		4.10* (2.00)
Spite \times $v_{[0,50]}$		1.49** (0.48)	1.49** (0.48)		
Spite \times $v_{[50,100]}$		0.31 (0.43)	0.31 (0.43)		
Risk			-6.02* (2.99)	-4.36 (3.12)	-3.62 (3.03)
Risk \times $v_{[0,50]}$				-3.48*** (0.88)	-3.48*** (0.88)
Risk \times $v_{[50,100]}$				-3.02*** (0.79)	-3.02*** (0.79)
Male			-19.05** (6.27)		-19.05** (6.27)
Rivalry			-0.70 (3.17)		-0.70 (3.17)
SVO			0.41 ⁺ (0.25)		0.41 ⁺ (0.25)
IA			-1.84 (2.58)		-1.84 (2.58)
Constant	27.82*** (3.21)	27.82*** (3.21)	27.75*** (6.68)	27.80*** (3.20)	27.75*** (6.68)
Observations	23,760	23,760	23,760	23,760	23,760
Log Likelihood	-120,829.50	-120,824.10	-120,806.90	-120,816.60	-120,800.40
Akaike Inf. Crit.	241,679.00	241,672.30	241,647.90	241,657.20	241,634.90
Bayesian Inf. Crit.	241,759.70	241,769.20	241,785.10	241,754.10	241,772.20

Notes:

p :*** < .001** < .01* < .05⁺ < .10

Table 7: Estimation results for Equation (13) (overbidding) with piece wise linear splines. The table shows estimation results for the different controls C'_1 , C'_2 , C'_3 , C'_4 , and C'_5 . Splines have knots at valuations 25, 50, and 75. Spite is the sum of the three spite measures. IA is the sum of the inequality aversion score obtained from the slider measure and the score obtained from inequality allocation of our own spite measure.

	C'_1	C'_2	C'_3	C'_4	C'_5
Period	-0.40*** (0.05)	-0.40*** (0.05)	-0.40*** (0.05)	-0.40*** (0.05)	-0.40*** (0.05)
Valuation	31.14*** (2.25)	31.14*** (2.25)	31.14*** (2.25)	31.14*** (2.25)	31.14*** (2.25)
Valuation ²	35.25*** (1.66)	35.25*** (1.66)	35.25*** (1.66)	35.25*** (1.66)	35.25*** (1.66)
Valuation ³	-93.63*** (1.10)	-93.63*** (1.10)	-93.63*** (1.10)	-93.63*** (1.10)	-93.63*** (1.10)
Spite		2.57 (1.71)	3.34 ⁺ (2.02)		4.10* (2.00)
Spite $\times v_{[0,50]}$		1.49** (0.48)	1.49** (0.48)		
Spite $\times v_{[50,100]}$		0.31 (0.43)	0.31 (0.43)		
Risk			-6.02* (2.99)	-4.36 (3.12)	-3.62 (3.03)
Risk $\times v_{[0,50]}$				-3.48*** (0.88)	-3.48*** (0.88)
Risk $\times v_{[50,100]}$				-3.02*** (0.79)	-3.02*** (0.79)
Male			-19.05** (6.27)		-19.05** (6.27)
Rivalry			-0.70 (3.17)		-0.70 (3.17)
SVO			0.41 ⁺ (0.25)		0.41 ⁺ (0.25)
IA			-1.84 (2.58)		-1.84 (2.58)
Constant	25.73*** (3.24)	25.70*** (3.21)	25.64*** (6.69)	25.68*** (3.20)	25.64*** (6.69)
Observations	23,760	23,760	23,760	23,760	23,760
Log Likelihood	-120,946.20	-120,937.50	-120,920.30	-120,930.00	-120,913.80
Akaike Inf. Crit.	241,908.30	241,897.00	241,872.50	241,882.00	241,859.60
Bayesian Inf. Crit.	241,972.90	241,985.80	242,001.70	241,970.80	241,988.90

Notes:

p :*** < .001** < .01* < .05⁺ < .10

Table 8: Estimation results for Equation (13) (overbidding) with B-Splines (degree three). The table shows estimation results for the different controls C'_1 , C'_2 , C'_3 , C'_4 , and C'_5 . B-Splines with degree three are used. Spite is the sum of the three spite measures. IA is the sum of the inequality aversion score obtained from the slider measure and the score obtained from inequality allocation of our own spite measure.

	C'_1	C'_2	C'_3	C'_4	C'_5
Period	-0.40*** (0.06)	-0.40*** (0.06)	-0.40*** (0.06)	-0.40*** (0.06)	-0.40*** (0.06)
Spite		2.57 (1.71)	3.34 ⁺ (1.97)		4.10* (1.95)
Spite $\times v_{[0,50]}$		1.49** (0.52)	1.49** (0.52)		
Spite $\times v_{[50,100]}$		0.31 (0.47)	0.31 (0.47)		
Risk			-6.02* (2.91)	-4.36 (3.11)	-3.62 (2.97)
Risk $\times v_{[0,50]}$				-3.48*** (0.95)	-3.48*** (0.95)
Risk $\times v_{[50,100]}$				-3.02*** (0.86)	-3.02*** (0.86)
Male			-19.05** (6.11)		-19.05** (6.11)
Rivalry			-0.70 (3.09)		-0.70 (3.09)
SVO			0.41 ⁺ (0.24)		0.41 ⁺ (0.24)
IA			-1.84 (2.51)		-1.84 (2.51)
Constant	14.92*** (3.16)	14.89*** (3.12)	14.83* (6.48)	14.87*** (3.11)	14.83* (6.48)
Observations	23760	23760	23760	23760	23760
Log Likelihood	-122872.72	-122866.39	-122857.12	-122861.47	-122852.67
Akaike Inf. Crit	245757.43	245750.77	245742.24	245740.95	245733.33
Bayesian Inf. Crit.	245805.89	245823.46	245855.3	245813.63	245846.39

Notes: p :*** < .001** < .01* < .05⁺ < .10;

Table 9: Estimation results for Equation(13) (overbidding) with cyclic cubic regression splines.

The table shows estimation results for the different controls C'_1 , C'_2 , C'_3 , C'_4 , and C'_5 . cyclic cubic regression splines are used. Spite is the sum of the three spite measures. IA is the sum of the inequality aversion score obtained from the slider measure and the score obtained from inequality allocation of our own spite measure.

	C'_1	C'_2	C'_3	C'_4	C'_5
Period	-0.40*** (0.05)	-0.40*** (0.05)	-0.40*** (0.05)	-0.40*** (0.05)	-0.40*** (0.05)
Spite		2.57 (1.70)	3.34 ⁺ (1.97)		4.10* (1.95)
Spite $\times v_{[0,50]}$		1.49** (0.47)	1.49** (0.47)		
Spite $\times v_{[50,100]}$		0.31 (0.42)	0.31 (0.43)		
Risk			-6.02* (2.91)	-4.36 (3.10)	-3.62 (2.96)
Risk $\times v_{[0,50]}$				-3.48*** (0.86)	-3.48*** (0.86)
Risk $\times v_{[50,100]}$				-3.02*** (0.78)	-3.02*** (0.78)
Male			-19.05** (6.11)		-19.05** (6.11)
Rivalry			-0.70 (3.09)		-0.70 (3.09)
SVO			0.41 ⁺ (0.24)		0.41 ⁺ (0.24)
IA			-1.84 (2.51)		-1.84 (2.51)
Constant	14.92*** (3.15)	14.89*** (3.11)	14.83* (6.48)	14.87*** (3.10)	14.83* (6.48)
Observations	23760	23760	23760	23760	23760
Log Likelihood	-120506.61	-120499.31	-120490.04	-120493.4	-120484.6
Akaike Inf. Crit	241027.22	241018.61	241010.08	241006.8	240999.19
Bayesian Inf. Crit.	241083.75	241099.37	241131.21	241087.56	241120.33

Notes:

p :*** < .001** < .01* < .05⁺ < .10;

Table 10: Estimation results for Equation (13) (overbidding) with P splines.

The table shows estimation results for the different controls C'_1 , C'_2 , C'_3 , C'_4 , and C'_5 . P-splines are used. Spite is the sum of the three spite measures. IA is the sum of the inequality aversion score obtained from the slider measure and the score obtained from inequality allocation of our own spite measure.

E.3. Polynomial approach (instead of linear spline)

Table 11 show the estimation results for Equation (13) using the following econometric model:

$$\begin{aligned}
\text{Bid}_{i,t,j,v} - b^{\text{II-AP}} &= \beta_0 + \beta_1 \text{Period} + \zeta_{i,j} + \eta_j + \epsilon_{i,j,k,l} + C'_M \\
C'_1 &= s(v) \\
C'_2 &= C'_1 + \beta_3 \text{Spite}_i + \beta_4 \text{Spite}_i \cdot v + \beta_5 \text{Spite}_i \cdot v^2 \\
C'_3 &= C'_2 + \beta_6 \text{IA}_i + \beta_7 \mathbb{1}_{\text{Gender}=\text{♀}} + \beta_8 \text{Risk}_i + \beta_9 \text{rivalry}_i + \beta_{10} \text{SVO}_i \\
C'_4 &= C'_1 + \beta_{11} \text{Risk}_i + \beta_{12} \text{Risk}_i \cdot v + \beta_{13} \text{Risk}_i \cdot v^2 \\
C'_5 &= C'_2 + \beta_{14} \text{IA}_i + \beta_{15} \mathbb{1}_{\text{Gender}=\text{♀}} + \beta_{16} \text{Spite}_i + \beta_{17} \text{rivalry}_i + \beta_{18} \text{SVO}_i \quad (31)
\end{aligned}$$

where $\zeta_{i,j}$ is a random effect for bidder i in group j , η_j is a random effect for group j , and $\epsilon_{i,j,k,l}$ is the residual. $s(v)$ is the thin plate regression spline over the valuation.

	C'_1	C'_2	C'_3	C'_4	C'_5
Period	-0.40*** (0.05)	-0.40*** (0.05)	-0.40*** (0.05)	-0.40*** (0.05)	-0.40*** (0.05)
Spite		2.53 (1.70)	3.30 ⁺ (1.97)		4.10* (1.95)
Spite $\times v$		0.05** (0.02)	0.05** (0.02)		
Spite $\times v^2$		0.001** (0.001)	0.001** (0.001)		
Risk			-6.02* (2.91)	-4.15 (3.11)	-3.40 (2.97)
Risk $\times v$				-0.10*** (0.03)	-0.10*** (0.03)
Risk $\times v^2$				0.001** (0.001)	0.001** (0.001)
Male			-19.05** (6.11)		-19.05** (6.11)
Rivalry			-0.70 (3.09)		-0.70 (3.09)
SVO			0.41 ⁺ (0.24)		0.41 ⁺ (0.24)
IA			-1.84 (2.51)		-1.84 (2.51)
Constant	14.92*** (3.15)	14.89*** (3.11)	14.83* (6.48)	14.87*** (3.10)	14.83* (6.48)
Observations	23760	23760	23760	23760	23760
Log Likelihood	-120506.69	-120500.4	-120491.13	-120493.22	-120484.42
Akaike Inf. Crit	241027.38	241020.8	241012.26	241006.44	240998.83
Bayesian Inf. Crit.	241083.91	241101.56	241133.4	241087.2	241119.97

Notes:

p :*** < .001** < .01* < .05⁺ < .10;

Table 11: Estimation results for Equation (31)

The table shows estimation results for the different controls C'_1 , C'_2 , C'_3 , C'_4 , and C'_5 . Thin plate regression splines are used. Spite is the sum of the three spite measures. IA is the sum of the inequality aversion score obtained from the slider measure and the score obtained from inequality allocation of our own spite measure.

E.4. Estimating Equation (13) with the individual spite measures

Table 12, 13 and 14 show the estimation results for Equation (13) using the three spite measures separately. The estimations are mainly in line with the results of the normalized combined spite-measure.

	C'_1	C'_2	C'_3	C'_4	C'_5
Period	-0.40*** (0.05)	-0.40*** (0.05)	-0.40*** (0.05)	-0.40*** (0.05)	-0.40*** (0.05)
Spite		2.53 (3.13)	1.75 (3.00)		2.93 (2.96)
Spite $\times v_{[0,50]}$		2.06* (0.86)	2.06* (0.86)		
Spite $\times v_{[50,100]}$		0.91 (0.78)	0.91 (0.78)		
Risk			-5.57 ⁺ (2.94)	-4.36 (3.10)	-3.17 (2.98)
Risk $\times v_{[0,50]}$				-3.48*** (0.86)	-3.48*** (0.86)
Risk $\times v_{[50,100]}$				-3.02*** (0.78)	-3.02*** (0.78)
Male			-18.31** (6.20)		-18.31** (6.20)
Rivalry			1.26 (2.95)		1.26 (2.95)
SVO			0.44 ⁺ (0.25)		0.44 ⁺ (0.25)
IA			0.71 (2.17)		0.71 (2.17)
Constant	14.92*** (3.15)	14.86*** (3.14)	13.94* (6.55)	14.87*** (3.10)	13.94* (6.55)
Observations	23760	23760	23760	23760	23760
Log Likelihood	-120506.69	-120503.12	-120494.32	-120493.49	-120486.38
Akaike Inf. Crit	241027.38	241026.24	241018.64	241006.97	241002.76
Bayesian Inf. Crit.	241083.91	241106.99	241139.77	241087.73	241123.9

Notes:

p :*** < .001** < .01* < .05⁺ < .10;

Table 12: Estimation results for Equation (13) (overbidding) (Kimbrough-Reiss)

The table shows estimation results for the different controls $C'_1, C'_2, C'_3, C'_4,$ and C'_5 . Thin plate regression splines are used. Spite is the Kimbrough-Reiss spite measure. IA is the sum of the inequality aversion score obtained from the slider measure and the score obtained from inequality allocation of our own spite measure.

	C'_1	C'_2	C'_3	C'_4	C'_5
Period	-0.40*** (0.05)	-0.40*** (0.05)	-0.40*** (0.05)	-0.40*** (0.05)	-0.40*** (0.05)
Spite		3.59 (3.14)	2.21 (4.20)		3.71 (4.17)
Spite $\times v_{[0,50]}$		2.42** (0.86)	2.42** (0.86)		
Spite $\times v_{[50,100]}$		1.45 ⁺ (0.78)	1.45 ⁺ (0.78)		
Risk			-5.27 ⁺ (2.93)	-4.36 (3.10)	-2.87 (2.98)
Risk $\times v_{[0,50]}$				-3.48*** (0.86)	-3.48*** (0.86)
Risk $\times v_{[50,100]}$				-3.02*** (0.78)	-3.02*** (0.78)
Male			-18.50** (6.20)		-18.50** (6.20)
Rivalry			1.72 (2.95)		1.72 (2.95)
SVO			0.42 ⁺ (0.25)		0.42 ⁺ (0.25)
IA			-0.83 (2.96)		-0.83 (2.96)
Constant	14.92*** (3.15)	14.96*** (3.12)	14.53* (6.58)	14.87*** (3.10)	14.53* (6.58)
Observations	23760	23760	23760	23760	23760
Log Likelihood	-120506.69	-120501.18	-120493.1	-120493.49	-120486.47
Akaike Inf. Crit	241027.38	241022.37	241016.19	241006.97	241002.95
Bayesian Inf. Crit.	241083.91	241103.12	241137.33	241087.73	241124.08

Notes:

p :*** < .001** < .01* < .05⁺ < .10;

Table 13: Estimation results for Equation (13) (overbidding) (Own measure)

The table shows estimation results for the different controls $C'_1, C'_2, C'_3, C'_4,$ and C'_5 . Thin plate regression splines are used. Spite is the own spite measure. IA is the sum of the inequality aversion score obtained from the slider measure and the score obtained from inequality allocation of our own spite measure.

	C'_1	C'_2	C'_3	C'_4	C'_5
Period	-0.40*** (0.05)	-0.40*** (0.05)	-0.40*** (0.05)	-0.40*** (0.05)	-0.40*** (0.05)
Spite		2.49 (3.13)	6.29 ⁺ (3.50)		6.17 ⁺ (3.46)
Spite $\times v_{[0,50]}$		0.54 (0.86)	0.54 (0.86)		
Spite $\times v_{[50,100]}$		-1.32 ⁺ (0.78)	-1.32 ⁺ (0.78)		
Risk			-6.04* (2.93)	-4.36 (3.10)	-3.63 (2.98)
Risk $\times v_{[0,50]}$				-3.48*** (0.86)	-3.48*** (0.86)
Risk $\times v_{[50,100]}$				-3.02*** (0.78)	-3.02*** (0.78)
Male			-20.58*** (6.22)		-20.58*** (6.22)
Rivalry			-1.61 (3.40)		-1.61 (3.40)
SVO			0.46 ⁺ (0.24)		0.46 ⁺ (0.24)
IA			0.34 (2.16)		0.34 (2.16)
Constant	14.92*** (3.15)	14.91*** (3.15)	14.48* (6.50)	14.87*** (3.10)	14.48* (6.50)
Observations	23760	23760	23760	23760	23760
Log Likelihood	-120506.69	-120503.75	-120493.44	-120493.49	-120485.29
Akaike Inf. Crit	241027.38	241027.5	241016.87	241006.97	241000.59
Bayesian Inf. Crit.	241083.91	241108.25	241138.01	241087.73	241121.73

Notes:

p :*** < .001** < .01* < .05⁺ < .10;

Table 14: Estimation results for Equation (13) (overbidding) (Spite-Score)

The table shows estimation results for the different controls C'_1 , C'_2 , C'_3 , C'_4 , and C'_5 . Thin plate regression splines are used. Spite is the score from the spite questionnaire (Marcus et al., 2014). IA is the sum of the inequality aversion score obtained from the slider measure and the score obtained from inequality allocation of our own spite measure.

E.5. Revenue

To compare the revenue for the seller of both auction types we approximate for every subject the expected revenue extracted by the seller, given the subjects' bidding function and given the behavior of the other bidders. For that purpose we draw for every subject 100 random valuations from a uniform distribution – hence, every subject participates in 100 potential auctions. For each subject we use the bids this subject would make given her bidding function. We match every subject's bids with the bids of every other subject (within an auction type) and estimate the average payment to the seller, for all random valuations. We use this procedure for every round, as subject's bids change throughout the game.³⁸

To obtain the standard errors we bootstrap the revenue means of every subject. Figure 17 shows the bootstrapped estimates for both auction types over rounds. It can be seen that the revenue seems to be higher in the second-price all-pay auction during the first rounds. However, while the revenue of the first-price winner-pay auction decreases only very slowly, the revenue in the second-price all-pay auction decreases more quickly over time. These results (C_3^R) are supported by a mixed-effects regression as reported in Table 15. The initial revenue-surplus of the second-price all-pay auction reduces over time and even swaps for late rounds, which explains why we do not find a significant difference between the average revenues of these two auction types (C_1^R).

³⁸Essentially, this procedure does account for a subject specific effect. To simplify matters we assume here that the group specific effect is negligibly small.

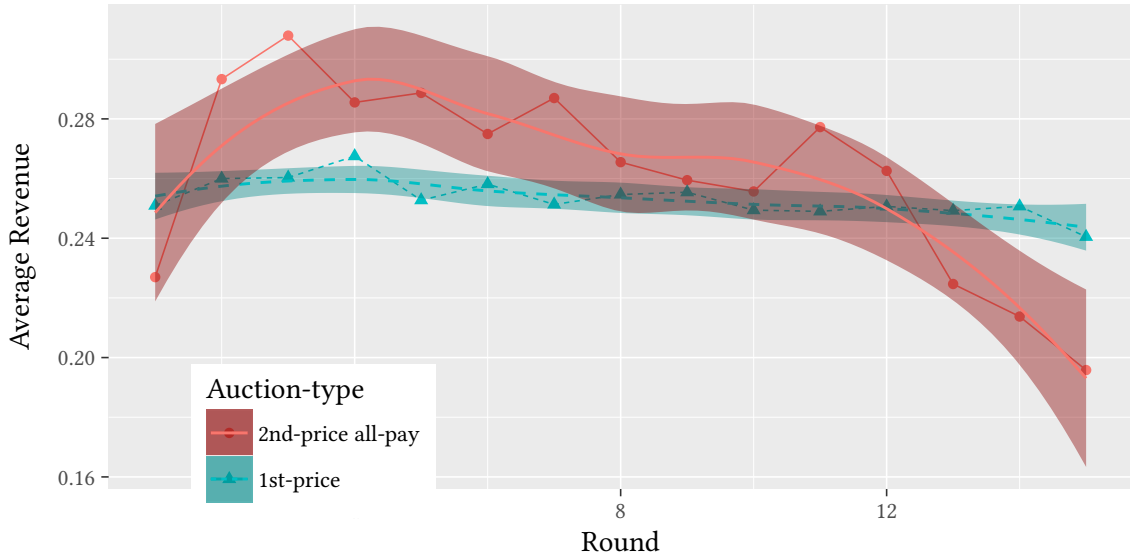


Figure 17: Mean revenues in both auction-types with standard error bands.

Hence, we can neither support Hypothesis 3.1 nor 3.2. It seem like revenue is higher in the second-price all-pay auction in the beginning, which would provide support for the hypothesis that spite plays a more important role than risk aversion. As this effect disappears, however, we cannot say whether risk or spite dominate the behavior on average.

Result 5.1. *Initially the second-price all-pay auction provides higher revenue, which is in line with theory of spiteful behavior. This effect however, disappears over time.*

	C_1^R	C_2^R	C_3^R
1st-price auction	-0.01 (0.02)	-0.01 (0.02)	-0.04* (0.02)
Period		-0.003*** (0.0004)	-0.005*** (0.001)
1st-price auction \times Period			0.004*** (0.001)
Constant	0.26*** (0.01)	0.29*** (0.01)	0.30*** (0.01)
Observations	3,660	3,660	3,660
Log Likelihood	2,631.28	2,653.02	2,657.03
Akaike Inf. Crit.	-5,252.57	-5,294.03	-5,300.06
Bayesian Inf. Crit.	-5,221.54	-5,256.80	-5,256.62

Notes: $p : *** < .001 ** < .01 * < .05^+ < .10$

Table 15: Estimating revenue for both auction-types.

F. Instructions

The experiment was conducted in German. All participants obtained the following handout (translated into English). Participants also saw video instructions, which are available upon request. The video instruction put into writing and translated into English can be found in Appendix F.2.

F.1. Handout

Payoff

- 3.50€ for your participation
- 2.50€ for answering the questionnaire
- Payoff from one Task (either A, or B, or C, or D)

First Task (A)

- Every participant will be assigned another participant
- You will make 21 decisions
- One Task will be randomly paid out
- 1 Point = 6 Euro-cents

Example:

Period: 1 of 1			
For each of the following distributions please indicate the one you prefer the most			
your pay- off	30 35 40 45 50 55 60 65 70	your pay- off	50
	<input type="radio"/> <input type="radio"/> <input type="radio"/> <input type="radio"/> <input checked="" type="radio"/> <input type="radio"/> <input type="radio"/> <input type="radio"/> <input type="radio"/>		
other's pay- off	80 70 60 50 40 30 20 10 0	other's pay- off	40
<input type="button" value="OK"/>			

In this example you obtain 50 points + the points from the decision of another person.

Second Task (B)

In this task you have to decide 10 times between two lotteries A and B. Only one of those 10 decisions will be paid out.

Example:

Lottery A	Lottery B	Your choice
In 1 out of 10 cases you will earn 1800 points and in 9 out of 10 cases you will earn 1440 points	In 1 out of 10 cases you will earn 3465 points and in 9 out of 10 cases you will earn 90 points	Lottery A <input type="radio"/> <input type="radio"/> Lottery B <input type="button" value="OK"/>

In this example you would get the following payoff in one out of 10 cases:

1800 points in case you choose Lottery A and 3465 in case you choose Lottery B.

And you would get the following payoff in 9 out of 10 cases:

1440 points in case you choose Lottery A and 90 in case you choose Lottery B.

- 1 Point = .5 Euro-cents

Task D

Task

- You play 10 auctions with another participant
- If your bid is higher than the bid of the other participant you win the auction. Otherwise, you lose the auction.
- For that purpose 10 random valuations between 500 and 1000 will be drawn for you and your fellow player each.
- Valuation: the amount you obtain in case you win the auction
- Decision: How much do you bid for each of the possible valuations
- In case you win you obtain your valuations as payoff and you have to pay the bid of the loser
- In case you lose you don't get any payoff and you don't have to pay anything.

Procedure:

1 Part: Decision

For all possible valuations between 500 and 1000 you indicate your bid.

2 Part: Result

In this part, you can see your bids and the bids of the other player if he won the auction. You also see which random valuations have been drawn for you and which auctions you won.

3 Part: Adaptation

You can increase your bids. However, you cannot change the outcome of the auction. E.g. if you have lost an auction then it will still stay this way.

2 Part: Implementation

To determine whether your adaptation will be implemented you have to bid with another player for whether the adaptation will be implemented or not.

If you bid more than this other player your adaptation will be implemented and you have to pay the bid of this new player for the adaptation.

If you bid less, you don't pay anything, however, your adaptation will also not be implemented.

Payoff

- 1 point = 0.01 €
- If task D is determined as payoff-relevant only one of the 10 auctions will be paid out
- You additionally get a 5€ payment if this task is paid out

- **+ Payoff =**

If you win the auction: Valuation - Bid of the losers (old or new) - bid for the implementation of the adaptation (in case the adaptation will be implemented for you)

If you lose the auction: - bid for the implementation of the adaptation (in case the adaptation will be implemented for you)

Task C

Task

- You play 15 rounds.
- You play every round 10 auctions with a new participant
- If your bid is higher than the bid of the other participant you win the auction. Otherwise, you lose the auction.
- For that purpose 10 random valuations between 0 and 100 will be drawn for you and your fellow player each. (E.g. both of you will have different valuations)

- Valuation: the amount you obtain in case you win the auction
- Decision: How much you bid for each of the possible valuations
- In case you win you obtain your valuations as payoff and [[First price auction instructions (FSP): you have to pay your own bid]][[second-price all-pay auction instructions (SNPAP): you have to pay the bid of the loser]]
- In case you lose you don't get any payoff and [[FSP: you don't have to pay anything]][[SNPAP: you have to pay your own bid]].

Procedure:

1 Part: Decision

For all possible valuations between 0 and 100 you indicate your bid.

The maximal possible bid is 150 points.

2 Part: Result

In this part, you can see your bids and the bids of the other player. You also see which random valuations have been drawn for you and which of the 10 auctions you won.

Payoff

- 1 point= 0.10 €
- If task C is determined as payoff-relevant only one of the 15 rounds will be paid out.
- If task C is determined as payoff-relevant only one of the 10 auctions will be paid out.
- You additionally get a 7€ payment if this task is paid out.

• **+ Payoff=**

If you win the auction: Valuation [[FSP: -your bid]] [[SNPAP: - Bid of the losers]]

If you lose the auction: [[FSP: 0]] [[SNPAP: - your bid]]

F.2. Text of the Video-instructions

Subjects were instructed with videos. The following shows the written form of the videos translated into English. The German version is available upon request from the authors. The videos can be obtained here: <https://www.kirchkamp.de/research/SpiteVsRisk.html>

Text to the video: General instructions

Welcome to this economic experiment. Today's experiment consists of four sub-experiments. Let us call them, for simplicity, A, B, C, and D. Additional to these tasks you will answer a questionnaire at the end. Let us come to the reimbursement of today's experiment. You will get 3.50€ for the participation in this experiment. You will get additional 2.50€ for answering the questionnaire. And you will get the payment from one of the tasks. Either from Task A, or Task B, or Task C, or Task D. Prior to each task, you will see an instructive video.

Text to the video: SVO (Murphy et al., 2011)

Let us now come to the first sub-experiment. In this sub-experiment every participant will be randomly assigned to another participant. For example, participant A will be assigned participant B, and participant B will be assigned participant C and so every participant will be assigned a different participant. Accordingly, the decision of participant A will be influential for the payoff of participant B and the decision of participant B will have an influence on the payoff of participant C and so forth. You will make 21 decisions over distributions. Only one decision will be randomly picked for payoff in case this sub-experiment is chosen for payoff. Here you see an example for one such decision. The decision consists of choosing one of the distributions. This distribution influences your payoff and the payoff of your fellow participant, who was randomly assigned to you. Let us assume you choose the distribution marked by the red circle. Then you will see your payoff on the top right side. On the lower top side, you can see how much the participant assigned to you will get as payoff. In this example, you earn 50 points. The participant assigned to you gets 40 points in this example. Let us assume this decision will be randomly drawn to be payoff-relevant at the end of the experiment. Let us further assume that you, as player A, choose the decision marked by the red circle. Then you would earn 50 points. Let us further assume that the player, to whom you were randomly assigned, let us call him player Z, chooses the same decision. Then you would get 40 points from this player. In this sub-experiment, every point is worth 6 cents. In the just mentioned example, you would earn 50 points for your decision plus 40 points for the decision of the player who influences your payoff. All together you would earn 90 points, which is worth 5.40€. If this task is chosen for payoff you will earn, in addition to the 3.50€ for participating in the experiment and the 2.50€ for answering the questionnaire, the payoff of one randomly drawn distribution. Please do not forget to click “done” at the end of a decision. If you have any further questions please press the red button on your keyboard and we will come to you. Otherwise, we wish you good luck.

Text to the video: Risk

Let us now come to the second task. Here you have to decide 10 times between Lottery A and Lottery B. Only one of the 10 decisions will be randomly implemented. Here you can see how the interface will later look like for you. In this column, you have to make your decision. Here you can choose between Lottery A and Lottery B. Only one of the 10 decisions will be randomly implemented for you and will influence your payoff. Hence, the first decision could be drawn. Or the fourth. Or the tenth. Which decision will be payoff-relevant for you will be determined randomly by the computer and will be announced to you at the end. Let us take a closer look at one such decision. Let us look for example at the first row. Here you see Lottery A and Lottery B. You now have to decide between Lottery A and Lottery B. In this example you would earn in one out of ten cases the following payoff: 1800 points if you have chosen Lottery A and 3465 points if you have chosen Lottery B. And in nine out of ten cases you would earn the following payoff: 1440 points if you have chosen Lottery A and 90 points if you have chosen Lottery B. In this sub-experiment, every point is worth .50 cents.

If this sub-experiment is drawn for payoff only one lottery will be randomly chosen and the lottery will be played according to your choice. If this task is chosen for payoff you will earn 3.50€ for participating in the experiment and the 2.50€ for answering the questionnaire plus the payoff from this sub-experiment. If you have any further questions please press the red button on your keyboard and we will come to you. Otherwise, we wish you good luck.

Text to the video: Auction

Let us now come to task C. Please note: At the end of this video you will answer 3 control questions to check whether you have understood this task. This task consists of 15 rounds. Each round you will play 10 auctions with a new player. If this sub-experiment is chosen for payoff only one of the auctions will be randomly paid out. In this sub-experiment every point is worth 20 cents. Every auction consists of the following parts: In every auction, two players take part who bid for a prize. In this example player A and player B. Both players value the prize randomly differently. Hence, player A values the prize with valuation A and player B values the prize with valuation B. E.g. valuation corresponds to how worth the prize is to one player. Both submit a bid according to their valuation. Let us assume that the bid of player A is higher than the bid of player B. In this case player A wins the auction and his payoff is: The valuation of player A minus the own bid minus the bid of the loser- in this case player B. Player B loses the auction, e.g. he is not getting any payoff however he still has to pay his bid. Let us now come to the decision in this task. In every round, you play 10 auctions with one randomly assigned player. You will decide for all possible valuations how much you want to bid. Out of all possible valuations, 10 valuations will be drawn randomly by the computer and you will bid according to your decision. To repeat: The payoff of one auction is calculated as the following: If you win the auction you gain your valuation minus your own bid minus the bid of the loser, in this case your co-player. Let us consider the following example: let us assume your valuation is 60 points. And the bid of your co-player for his, to you unknown, valuation is 40. If you have bid for example 50 points, then you win the auction, as you bid more than your co-player. And you obtain the following payoff: Your valuation minus your own bid minus the bid of the loser. Hence, 60 points, because this corresponds to your valuation, minus 50 points, hence, your own bid. Which results in 10 points which equates to 2€. Hence, 60 points, because this corresponds to your valuation, minus 50 points, hence, your own bid. Which results in 10 points which equates to 2€. If you have bid for example 30 points, then you lose the auction, as you bid less than your co-player, who bid 40 points. Hence, you obtain a payoff of 0 points. Hence you pay the bid of the loser. In this case, you would pay 30 points, which equates to 3€. In case both bid the same one player will be randomly announced the winner and the other the loser. Your interface will look like the following. The red circle shows here your possible valuations. In the red marked area, you have to indicate how much you would bid if your valuation would be 0, 10, 20 etc. The maximal possible bid is 150 points. On the button, you see in which of the 15 rounds you are currently in. If you click on "draw" you can see how much you would bid if your randomly drawn valuation is a number between 0 and 10 or between 10 and 20 or 20 and 30 and so on. Every number between 0 and

100 can be randomly picked by the computer to be your valuation. At the bottom, you see the possible valuations and on the left you see your bids according to your function. Let us assume your random valuation is 75. Then you would bid according to your input 40 points. If you are happy with your bidding function please click "done". Here you see the results of every of the 10 auctions in the first round. Here you can see your random valuations for each of the auctions. The red circle shows here how you bid according to your input. And here you see the bid of your co-player. In the red marked area you can see whether you won or lost the auction. And hence, how many points you have won and lost, respectively. Let us, for example, look at the first auction. Here you can see how much you bid and how much your co-player bid. [[FSP: Let us, for example, look at the seventh auction. If this auction will be drawn for payoff, you would lose and earn 0 points.]] [[SNPAP: Let us, for example, look at the ninth auction. If this auction will be drawn for payoff, you would lose and pay 3 points.]] Here you can see the auctions ones more graphically. The red dots represent those auctions you have lost. The green dots represent those auctions you have won. The blue crosses represent, in every auction, the bids of your co-player. If you click on "done", you will be directed to a new round, in which you will play again 10 auctions with a new player. If this task is chosen for payoff you will earn, in addition to the 3.50€ for participating in the experiment and the 2.50€ for answering the questionnaire, 7€. Plus the payoff of one auction out of the 15 rounds. Note that you can win but you can also lose those auctions. If you have any further questions please press the red button on your keyboard and we will come to you. Otherwise, we wish you good luck.

Text to the video: Market (Kimbrough-Reiss)

Let us now come to task D. Please note: At the end of this video you will answer 5 control questions to check whether you have understood this task. In this task, you play one round in which you will play 10 auctions. Only one of the auctions will be randomly paid out. In this sub-experiment, every point is worth 1 cent. Every auction consists of the following parts: In every auction, two players take part who bid for a prize. In this example player A and player B. Both players value the prize randomly differently. Hence, player A values the prize with valuation A and player B values the prize with valuation B. E.g. valuation corresponds to how worth the prize is to one player. Both submit a bid according to their valuation. Let us assume that the bid of player A is higher than the bid of player B. In this case player A wins the auction and his payoff is: The valuation of player A minus the bid of the loser- in this case player B. Player B loses the auction, e.g. he is not getting any payoff and his payment is 0 points. Let us now come to the procedure in this sub-experiment. This sub-experiment consists of four parts. Let us come to the decision. You play 10 auctions with one randomly assigned player. You will decide for all possible valuations how much you want to bid. Out of all possible valuations, 10 valuations will be drawn randomly by the computer and you will bid according to your decision. Here you see the interface in task D. The red circle shows here your possible valuations. Here you have to indicate how much you would bid if your valuation would be 500, 550, 600 etc. If you click on "draw" you can see how much you would bid if your randomly drawn valuation is a number between 500 and 550 or between

550 and 600 and so on. Every number between 500 and 1000 can be randomly picked by the computer to be your valuation. On the horizontal axis you see your valuations and on the vertical axis you see your bids according to your input. Let us assume your random valuation is 870. Then you would bid according to your input 600 points. If you are happy with your input please click on "done". Let us now come to the second part of the task: the result. Here you see the 10 auctions. Here you can see your random valuations for each of the auctions. The red circle shows here how you bid according to your input. Here you can see whether the bid of your co-player was smaller or higher than your bid. Here you can see whether you won or lost the auction. In those auctions in which you lost you can see the bid of your co-player. The payoff of one auction is calculated as the following: If you win the auction you gain your valuation minus the bid of the loser, in this case your co-player. If you lose the auction you obtain 0 points as your payoff. Let us consider the following example: let us assume your valuation is 650 points. And the bid of your co-player for his, to you unknown, valuation is 540. If you have bid for example 600 points, then you win the auction, and you obtain your valuation minus the bid of the loser as payoff. In this case 650, your valuation, minus 540, the bid of your co-player. Hence, 110 points which equates to 1.10€. If you have bid for example 530 points, then you lose the auction, as you bid less than your co-player. Hence, you obtain a payoff of 0 points. In case both bid the same one player will be randomly announced the winner and the other the loser. Let us now come to the third part of task D: the adaptation. In the adaptation you can increase your bid, in those auctions you won. You can also increase your bid in those auctions you lost. However, you cannot overbid your co-player. E.g. if you have lost an auction it will stay this way. Here you can see the interface for the adaptation. Here you can see your bids. The green lines mark your bids in those auctions you have won, and the red lines mark your bids in those auctions you have lost. The red circle marks here the bids of your co-player, if he has won the auctions. Here you can view your new bids. You can view the increased bids in those auctions you have won and you can view the increased bids in those auctions you have lost. You can adapt your bids by moving the ruler in the marked circle. At the bottom, you can see the same information once more. You can see your valuations. Your former bids and your new bids. Note that the adaptation is not implemented for every player. Whether your adaptation is implemented depends on a further bid. You can do that in the fourth part of task D: the implementation. Here you bid for the adaptation. For that purpose, you will be assigned a new partner. You decide how much you are willing to pay for implementing the adaptation. If your new partner bids more than you, his adaptation will be implemented and yours will not. However, he will need to pay for this implementation as much as you were willing to pay for the adaptation. If you bid more than your new partner, your adaptation will be implemented and his will not. However, you will need to pay for this implementation as much as he was willing to pay for the adaptation. The player, whose adaptation is not implemented, does not need to pay his bid for the adaptation. Note: As you and your co-player are assigned a new player it might happen that the adaptation of both players is implemented. It can, however, also happen that no adaptation or only one of the adaptations is implemented. Here you see the interface for the implementation. Here you type in how much you are willing to pay to adapt the bid in those auctions you lost. Here you type in how much you are willing to pay to adapt the bid in those auctions in which you are the highest bidder. The payoff in this task, after adaptation,

Please answer the following questions When entering numbers please insert integers only	
A bids 528 and B bids 739, who wins the auction?	<input style="width: 60px; height: 20px;" type="text"/>
If the valuation of A is 650 and the bid of B is 550, how much payoff would A obtain, if A bids 700?	<input style="width: 60px; height: 20px;" type="text"/>
If the valuation of A is 650 and the bid of B is 550, how much payoff would A obtain, if A bids 500?	<input style="width: 60px; height: 20px;" type="text"/>
If the valuation of A is 520 and the bid of B is 550, how much payoff would A obtain, if A bids 580?	<input style="width: 60px; height: 20px;" type="text"/>
If a wins the adaptation of his bids, can it be that also the co-player of player A wins the adaptation?	<input style="width: 60px; height: 20px;" type="text"/>
<input style="width: 40px; height: 20px; background-color: red; color: white; border: none;" type="button" value="OK"/>	

Figure 18: Control questions in the spite measure (Kimbrough-Reiss).

is calculated as follows: If you win the auction you obtain as payoff your valuation minus the bid of the loser. At that, you have to pay either the old bid of the loser or the new one, dependent on whether the adaption of your co-player was implemented. In addition, you pay the amount you are willing to pay for the adaptation of those auctions you won. If you lose the auction, you have to pay, dependent on whether your adaption was implemented or not, the amount for the adaption. If this task is chosen for payoff you will earn, in addition to the 3.50€ for participating in the experiment and the 2.50€ for answering the questionnaire, 7€. Plus the payoff of one auction. Note that you can win but you can also lose those auctions. If you have any further questions please press the red button on your keyboard and we will come to you. Otherwise, we wish you good luck.

F.3. Control questions

To check and enhance the understanding of subjects, subjects had to solve the following two sets of control questions. Subjects had seven attempts to solve these questions. If subjects were not able to solve them after seven attempts they were presented the correct answers. Questions are shown in Figures 18 and 19.

Please answer the following questions
When entering numbers please insert integers only

If A bids 16 and B bids 12, who wins the auction?

If the valuation of A is 18 and the bid of B is 24, how much must A bid to have the smallest loss? (Tips: A number out of (0/11/18/24))

If the valuation of A is 18 and the bid of B is 10, how much must A bid to have the highest (safe) payoff ? (Tips: A number out of (0/10/11))

Figure 19: Control questions in the second-price all-pay auction.