Marketing Agencies and Collusive Bidding* in Online Ad Auctions †

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Abstract

The transition of the advertising market from traditional media to the internet has induced a proliferation of marketing agencies specialized in bidding in the auctions that are used to sell ad space on the web. We analyze how collusive bidding can emerge from bid delegation to a common marketing agency and how this can undermine the revenues and allocative efficiency of both the Generalized Second Price auction (GSP, used by Google and Microsoft-Bing and Yahoo!) and of the VCG mechanism (used by Facebook). We find that, despite its well-known susceptibility to collusion, the VCG mechanism outperforms the GSP auction both in terms of revenues and efficiency.

Keywords: Collusion, Digital Marketing Agencies, Facebook, Google, GSP, Internet Auctions, Online Advertising, VCG. JEL: C72, D44, L81.

*The use of the word collusion in this essay is unrelated to any assessment of the legal implications of agencies or advertisers behavior under the competition laws of the US or other countries.
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1 Introduction

Online advertising is the main source of revenues for important firms such as Google, Facebook, Twitter, etc., and it represents one of the largest and fastest growing industries in the US: in 2017, for instance, the value of advertising on search engines alone exceeded 50 billion dollars in the U.S., with an annual growth of nearly 10% (PwC 2017). Almost all online ads are sold through auctions, in which bidders compete for the adjudication of one of a given number of ‘slots’ available in various online venues, such as search engine result pages, social networks feeds, and so on. The most important auction formats in this market are the Generalized Second Price (GSP) auction (used, for instance, by Google, Microsoft-Bing, Yahoo!, etc.) and the Vickerey-Clarke-Groves (VCG) mechanism (used by Facebook and by Google for its contextual ads). The VCG is a classic and widely studied mechanism: it involves fairly complex payments that price externalities, but it has the advantage of being strategy-proof and efficient. In contrast, the GSP auction – whose study was pioneered by Varian (2007) and Edelman, Ostrovsky and Schwarz (EOS) (2007) – has very simple rules (the $k$-highest bidder obtains the $k$-highest slot at a price-per-click equal to the $(k + 1)$-highest bid), but it gives rise to complex strategic interactions. Both auction formats have been studied extensively. With few exceptions, however, existing models have largely ignored a major trend in this market: the rise of intermediaries operating on the bidding platforms.\(^1\)

At least since 2011, an increasing number of advertisers are delegating their bidding campaigns to specialized digital marketing agencies (DMAs), many of which belong to a handful of agency networks (seven in the US) that conduct all bidding activities through centralized agency trading desks (ATDs). In recent years, these agency networks have expanded their activities and contributed to a major increase in the market concentration, reaching global revenues that compare well even with those of tech giants like Google\(^2\). As a result, with increasing frequency, the same entity (be it DMA or ATD) bids in the same auction on behalf of different advertisers. But this clearly changes the strategic interaction, as these agencies have the opportunity to lower their payments by coordinating the bids of their clients. This not only affects advertisers’ optimal bidding strategies, but it also has the potential to alter the very functioning of these auction formats.

This paper proposes a theoretical analysis of the impact of agency bidding on the two main auction formats. We find that the agency’s equilibrium bids are akin to implementing a certain form of collusion (even if none of its clients explicitly attempt it), and that in this situation the VCG outperforms the GSP both in terms of revenues and efficiency. This is a

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\(^1\)Mansour, Muthukrishnan and Nisan (2012) first pointed at the potential risk of collusive bidding that intermediaries posed for online ad auctions. That paper focused on the ad exchanges used for display ads. Balseiro and Candogan (2017) is an important attempt to look at optimal contracts for intermediaries, showing that under such contracts intermediaries can bolster revenues for budget-constrained advertisers, while also increasing the overall market efficiency. McAfee (2011) studies how intermediaries can help solving problems of limited information in ad exchanges. The importance of information and learning in display auction is also stressed in Despotakis et al. (2016).

\(^2\)In 2016, the total revenues of these seven agency networks amounted to one third of those of Alphabet.
strong result because the VCG is typically considered to be highly susceptible to collusion (e.g., Ausubel and Milgrom (2006)), but it is especially noteworthy if one considers the sheer size of transactions currently occurring under the GSP. It also suggests a rationale for why Facebook’s recent adoption of the VCG mechanism was so successful, despite the early surprise it provoked (e.g., Wired (2015)), and for why the last few years have recorded a steady decline in ad prices. The fragility of the GSP auction which we uncover suggests that further changes are likely to occur in this industry, raising important questions from both a market-design and an antitrust perspective.

A satisfactory model of agency bidding needs to satisfy at least two desiderata: First, it must allow collusive and competitive behavior to coexist, because agencies in these auctions typically operate side by side with independent advertisers. Second, it needs to be sufficiently tractable and amenable to direct comparisons to the existing benchmarks in the literature. To achieve these goals, we modify EOS and Varian’s baseline model by introducing a marketing agency, which we model as a player choosing bids for its clients in order to maximize the total profits. Bidders that do not belong to the agency are referred to as ‘independents’, and have the usual objectives. To overcome the curse of multiplicity in the GSP auction, and ensure a meaningful comparison with the competitive benchmark, we introduce a refinement of bidders’ best responses that distills the individual-level underpinnings of EOS equilibrium, and assume that independents place their bids accordingly. This stratagem enables us to maintain the logic of EOS refinement for the independents, even if their equilibrium is not defined in the game with collusion. The marketing agency in turn makes a proposal of a certain profile of bids to its clients. The proposal is implemented if it is ‘recursively stable’ in the sense that, anticipating the bidding strategies of others, and taking into account the possible unraveling of the rest of the coalition, no client has an incentive to abandon the agency and bid as an independent. Hence, the clients’ outside options are equilibrium objects themselves, and implicitly incorporate the restrictions entailed by the underlying coalition formation game.

We consider different models of collusive bidding within this general framework. First, we assume that the agency is constrained to placing bids that cannot be distinguished from EOS competitive equilibria by an external observer (e.g., the auction platform or an antitrust authority). We show that, under this constraint, the GSP auction is efficient and its revenues are identical to those obtained if the same agency bid in a VCG auction. We then relax this ‘undistinguishability constraint’, and show that, even in the absence of allocative distortions, the GSP’s revenues are lower than those obtained in the VCG mechanism with the same agency configuration. Furthermore, once the ‘undistinguisha-

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3Google, for instance, reports passing from a positive growth rate in its average cost-per-click of about 4 percent per year in the four years before 2012, to a negative growth rate in each year since then, with an average yearly decline of 9 percent. Source: 10-k filings of Alphabet inc.
4The problem of ‘partial cartels’ is acknowledged as a major difficulty in the literature (e.g., Hendricks, Porter and Tan (2008)). We discuss this point and the connection with that literature in Section 5.
5This approach, which involves both equilibrium and recursive stability restrictions, is closely related to the equilibrium binding agreements of (Ray and Vohra, 1997), further discussed in Section 5.
bility constraint’ is lifted, efficiency is no longer guaranteed by the GSP. Since the VCG is well-known to be highly susceptible to collusion, finding that it outperforms the GSP both in terms of revenues and efficiency is remarkably negative for the GSP auction.

The source of the GSP’s fragility, and the complexity of agency bidding in this context, can be understood thinking about an agency that controls the first, second, and fourth highest bidders in an auction. The agency in this case can lower the highest bidder’s payment by shading the bid of the second, without necessarily affecting either his position or his payment. Given the rules of the GSP auction, the agency can benefit from this simple strategy only if two of her members occupy adjacent positions. But due to the GSP’s complex equilibrium effects, the agency can do more than that. For instance, suppose that the agency shades the bid of her lowest member, with no direct impact on her other clients’ payments. Intuitively, if this bid is kept persistently lower, then the logic of EOS’ refinement suggests that the third highest bidder, who is an independent, would eventually lower his bid. But not only would this lower the second bidder’s payment, it would also give the agency extra leeway to lower the second highest bid, to the greater benefit of the highest bidder. Revenues in this case diminish for both the direct effect (lowering the 2-nd highest bid lowers the highest bidder’s payment) and for the indirect effect (lowering the 4-th highest bid induces a lower bid for the independent, which in turn lowers the second bidder’s payment). Hence, even a small coalition may have a large impact on total revenues. Our general results show that this impact is larger if the agency includes members which occupy low or adjacent positions in the ranking of valuations, but it also depends on the rate at which click-through-rates vary from one position to another, and on how independents’ valuations compare to those of the coalition members.

We also explore whether these concerns on the GSP auction may be mitigated by competition between agencies. Although multiple agencies each with multiple bidders in the same auction are not the typical case in the data, the question has theoretical relevance because the phenomenon may become more common in the future. If an increase in agency competition restored the good properties of these auctions, then the diffusion of marketing agencies need not lead to major structural changes in this industry. Our results, however, suggest otherwise: for certain coalition structures, agency competition as expected mitigates the revenue losses in both mechanisms (while preserving their relative performance); but for other coalition structures, it has a particularly perverse impact on both mechanisms. That is because, from the viewpoint of an agency bidding for multiple clients, these auction mechanisms have a flavor of a first-price auction: even holding positions constant, the price paid depends on the agency’s own bids. With multiple agencies, this feature of agency bidding may lead to non-existence of pure equilibria, very much like the case of competitive (non-agency) bidding in a Generalized First Price (GFP) auction. But as seen in the early days of this industry, when the GFP was adopted, lack of pure equilibria may generate bidding cycles which eventually lead to a different form of collusion. In fact, these bidding cycles are often cited as the primary cause for
the transition, in the early ’00s, from the GFP to the GSP auction (cf. Edelman and Ostrovsky (2007)). Hence, not only does agency competition not solve the problems with these auctions, but it appears likely to exacerbate them.

The rest of the paper is organized as follows: Section 2 reviews the competitive benchmark; Section 3 introduces the agency model, and Section 4 presents the main results; Section 5 discusses the related literature and some extensions; Section 6 concludes.

2 Competitive Bidding in Online Ad Auctions

Online ad auctions are mechanisms to assign agents $i \in I = \{1, \ldots, n\}$ to slots $s = 1, \ldots, S$, $n \geq S$ where for simplicity we assume $n = S+1$ (the extension to $n \geq S$ is straightforward). In our case, agents are advertisers, and slots are positions for ads on a webpage (e.g., on a social media’s newsfeed for a certain set of cookies, on a search-engine result page for a given keyword, etc.). Slot $s = 1$ corresponds to the highest/best position, and so on until $s = S$, which is the slot in the lowest/worst position. For each $s$, we let $x_s$ denote the ‘click-through-rate’ (CTR) of slot $s$, that is the number of clicks that an ad in position $s$ is expected to receive, and assume that $x_1 > x_2 > \cdots > x_S > 0$. We also let $x^t = 0$ for all $t > S$. Finally, we let $v_i$ denote the per-click-valuation of advertiser $i$, and we label advertisers so that $v_1 > v_2 > \cdots > v_n$. As in Varian (2007) and EOS, we maintain that valuations and CTRs are common knowledge. This complete information environment is the main benchmark for the literature on the GSP auction.

2.1 Rules of the auctions

Both in the VCG and in the GSP auction, advertisers submit bids $b_i \in \mathbb{R}_+$, and slots are assigned according to their ranking: first slot to the highest bidder, second slot to the second-highest bidder, and so on. We denote bid profiles by $b = (b_i)_{i=1,\ldots,n}$ and $b_{-i} = (b_j)_{j \neq i}$. For any profile $b$, we let $\rho(i;b)$ denote the rank of $i$’s bid in $b$ (ties are broken according to bidders’ labels).\footnote{Formally, $\rho(i;b) := |\{j : b_j > b_i\} \cup \{j : b_j = b_i \text{ and } j < i\}| + 1$. This tie-breaking rule is convenient for the analysis of coordinated bidding. It can be relaxed at the cost of added technicalities.} When $b$ is clear from the context, we omit it and write simply $\rho(i)$. For any $t = 1, \ldots, n$ and $b$ or $b_{-i}$, we let $b^t$ and $b^t_{-i}$ denote the $t$-highest component of the vectors $b$ and $b_{-i}$, respectively. Hence, with this notation, for any profile $b$, in either mechanism bidder $i$ obtains position $\rho(i)$ if $\rho(i) \leq S$, and no position otherwise. The resulting payoff, ignoring payments, is thus $v_i x^{\rho(i)}$.

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A notable exception is Gomes and Sweeney (2014), which provide a thorough analysis of competitive bidding in the GSP auction with independent private values, with a much more pessimistic outlook on both the allocative and revenue properties of the GSP auction. Borgers et al. (2013) maintain the complete information assumption, but consider a more general model of CTRs and valuations. Also, similar to EOS, our baseline model abstracts from quality scores, which in practice are often used to adjust advertisers’ bids in determining their position and payments. Athey and Nekipelov (2014) introduced uncertainty over quality scores in a model with competitive bids. We discuss quality scores in Section 5.

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The GSP and VCG mechanisms only differ in their payment rule. In the GSP mechanism, the $k$-highest bidder gets position $k$ and pays a price-per-click equal to the $(k+1)$-th highest bid. Using our notation, given a profile of bids $b$, agent $i$ obtains position $\rho(i)$ and pays a price-per-click equal to $b^{\rho(i)+1}$. Bidder $i$’s payoff in the GSP auction, given a bids profile $b \in \mathbb{R}_+^n$, can thus be written as $u_i^G(b) = (v_i - b^{\rho(i)+1}) x^{\rho(i)}$.

In the VCG auction, an agent pays the total allocation externality he imposes on others. In this setting, if the advertiser in position $k$ were removed from the auction, all bidders below him would climb up one position. Hence, if other bidders are bidding truthfully (i.e., $b_j = v_j$, as will be the case in equilibrium), the total externality of the $k$-highest bidder is equal to $\sum_{t=k+1}^{S+1} b^t(x_t - x^t)$. We can thus write $i$’s payoff in the VCG mechanism, given a bids profile $b \in \mathbb{R}_+^n$, as $u_i^V(b) = v_i x^{\rho(i)} - \sum_{t=\rho(i)+1}^{S+1} b^t(x_t - x^t)$.

In the rest of this section we review known results on the competitive benchmarks for these two mechanisms. The only original result will be Lemma\[1\] which provides an alternative characterization of EOS’ lowest envy-free equilibrium of the GSP auction.

### 2.2 Equilibria

Despite the relative complexity of its payment rule, bidding behavior in the VCG is very simple, as truthful bidding (i.e., $b_i = v_i$) is a dominant strategy in this auction. In the resulting equilibrium, advertisers are efficiently assigned to positions. The VCG mechanism therefore is efficient and strategy-proof.

Equilibrium behavior in the GSP auction is much more complex. To see this, first note that a generic profile of bids for $i$’s opponents, $b_{-i} = (b_j)_{j \neq i}$, partitions the space of $i$’s bids into $S + 1$ intervals of payoff-equivalent bids for bidder $i$. So, for each $b_{-i} \in \mathbb{R}_{+}^{n-1}$, let $\pi_i(b_{-i})$ denote $i$’s favorite position, given $b_{-i}$. Then, $i$’s best-response correspondence $BR_i : \mathbb{R}_{+}^{n-1} \rightrightarrows \mathbb{R}_+$ is such that, for every $b_{-i}$, $BR_i(b_{-i}) = (\pi_{\rho_i(b_{-i})}, \pi_{\rho_i(b_{-i})} - 1)$.

The GSP auction has many Nash equilibria (fixed-points of the $x_{i \in I} BR_i$). For this reason, EOS introduced a refinement of the equilibrium correspondence, the lowest-revenue locally envy-free equilibrium, which was crucial to cut through the complexity of the GSP auction.\[8\] As EOS showed, such equilibria induce the same allocations and payments as truthful bidding in the VCG, and they are fully characterized by the following conditions: $b_1 > b_2, b_i = v_i$ for all $i > S$, and for all $i = 2, \ldots, S$,

$$b_i = v_i - \frac{x_i}{x_{i-1}} (v_i - b_{i+1}).$$

But EOS refinement is not defined when agencies are present. We thus consider instead

\[8\] Allowing ties in individuals’ bids or non-generic indifferences complicates the notation, without affecting the results and the main insights. See Appendix A.1 for details on this.

\[9\] A Nash equilibrium $(b_i)_{i \in I}$ is locally envy-free if $x^{\rho(i)} (v_i - b^{\rho(i)+1}) \geq x^{\rho(i)-1} (v_i - b^{\rho(i)})$ for every $i$. EOS refinement is the lowest-revenue Nash equilibrium which satisfies this condition. This refinement is especially important because it conforms with the search engines’ tutorials on how to bid in these auctions. See, for instance, the Google AdWord tutorial in which Hal Varian teaches how to maximize profits by following this bidding strategy: http://www.youtube.com/watch?v=jRx7AMb6rZ0
a refinement of the individual best response correspondences, which distills the individual-level underpinnings of EOS refinement. Formally: for any $b_{-i} \in \mathbb{R}^{n-1}$, let

$$BR^*_i(b_{-i}) = \left\{ b^*_i \in BR_i(b_{-i}) : \left( v_i - b^*_{\pi_i}(b_{-i}) \right) x^{\pi_i(b_{-i})} = (v_i - b^*_i) x^{\pi_i(b_{-i})-1} \right\}. \quad (2)$$

In words, of the many $b_i \in BR_i(b_{-i})$ that would grant player $i$ his favorite position $\pi_i(b_{-i})$, he chooses the bid $b^*_i$ that makes him indifferent between occupying the current position and climbing up one position paying a price equal to $b^*_i$. The set of fixed points of the $\times_{i\in I} BR^*_i$ correspondence, given valuations $v$, are denoted as $E^*(v)$.

**Lemma 1** For any $v = (v_i)_{i=1,\ldots,n}$, $b \in E^*(v)$ if and only if $b$ is an EOS equilibrium.

This lemma shows that EOS’ equilibrium – originally defined as a refinement of the Nash equilibrium correspondence – can be equivalently defined as the fixed point of a refinement of individual best responses. Hence, $BR^*_i$ provides a model of individual behavior which is consistent with EOS’ equilibrium, and which is well-defined in our setting even if EOS’ equilibrium is not. The next example will be used repeatedly throughout the paper to illustrate the relative performance of the GSP and VCG mechanisms:

**Example 1** Consider an auction with four slots and five bidders, with the following valuations: $v = (5, 4, 3, 2, 1)$. The CTRs for the five positions are the following: $x = (20, 10, 5, 2, 0)$. In the VCG mechanism, bids are $b_i = v_i$ for every $i$, which induces total expected revenues of 96. Bids in EOS’ lowest envy-free equilibrium of the GSP auction instead are as follows: $b_5 = 1$, $b_4 = 1.6$, $b_3 = 2.3$ and $b_2 = 3.15$. The highest bid $b_1 > b_2$ is not uniquely determined, but it does not affect the revenues, which in this equilibrium are exactly the same as in the VCG mechanism: 96. Clearly, also the allocation is the same in the two mechanisms, and efficient. □

**3 A Model of Agency Bidding**

Our analysis of marketing agencies focuses on their opportunity to coordinate the bids of different advertisers. We thus borrow the language of cooperative game theory and refer to the clients of the agency as ‘members of a coalition’ and to the remaining bidders as ‘independents’. In this Section we focus on environments with a single agency, and postpone the analysis of the multiple agency case to Section 4.3.

Modelling coordinated bidding, it may seem natural to consider standard solution concepts such as strong Nash [Aumann, 1959] or coalition proof equilibrium [Bernheim and Whinston, 1987]. Unfortunately, these concepts have no bite in the GSP auction, as it can be shown that EOS’ equilibrium satisfies both refinements.

We model the marketing agency as a player that makes proposals of binding agreements to its members, subject to certain stability constraints. The independents then play the game which ensues from taking the bids of the agency as given. The agency’s proposals,
however, can only be implemented if they are stable in two senses: (\textbf{S.1}) first, if they are consistent with the independents’ equilibrium behavior; (\textbf{S.2}) second, if no individual member of the coalition has an incentive to abandon it and bid as an independent. We also assume that, when considering such deviations, coalition members are farsighted in the sense that they anticipate the impact of their deviation on both the independents and the remaining members of the coalition [Ray and Vohra, 1997]. Hence, given a coalition \( C \), the outside option for each member \( i \in C \) is his equilibrium payoff in the game with coalition \( C \setminus \{i\} \), in which \( i \) bids as an independent. The constraint for coalition \( C \) thus depends on the solutions to the problems of all the subcoalitions \( C' \subseteq C \), and hence the solution concept for the game with the agency will be defined recursively. We thus call it the ‘Recursively-Stable Agency Equilibrium’ (\textit{RAE}).

We will also consider a third constraint, (\textbf{R}), which we formalize as a set \( R(C) \subseteq A \), to accommodate the possibility that the agency exogenously discards certain bids. For instance, we will consider the case of an agency whose primary concern is not being identified as inducing collusion (Section 4.2.1) or to induce efficient outcomes (Section 4.2.2). In those cases, \( R(C) \) would be comprised respectively of only those profiles that are ‘undistinguishable’ to an external observer as collusive, or efficient.

### 3.1 The Recursively Stable Agency Equilibrium

Let \( G = (A_i, u_i)_{i=1, \ldots, n} \) denote the baseline game (without a coalition) generated by the underlying mechanism (e.g., GSP or VCG). We let \( \mathcal{C} \) denote the collection of all sets \( C \subseteq I \) such that \( |C| \geq 2 \). For any \( C \in \mathcal{C} \), we let \( C \) denote the agency, and we refer to advertisers \( i \in C \) as ‘members of the coalition’ and to \( i \in I \setminus C \) as ‘independents’. The coalition chooses a vector of bids \( b_C = (b_j)_{j \in C} \in \times_{j \in C} A_j \). Given \( b_C \), the independents \( i \in I \setminus C \) simultaneously choose bids \( b_i \in A_i \). We let \( b_{-C} := (b_j)_{j \in I \setminus C} \) and \( A_{-C} := \times_{j \in I \setminus C} A_j \). Finally, given profiles \( b \) or \( b_{-C} \), we let \( b_{-i,-C} := (b_j)_{j \in I \setminus C, j \neq i} \) denote the subprofile of bids of all independents other than \( i \). We assume that the agency maximizes the sum of its members’ payoffs\footnote{This is a simplifying assumption, which can be justified in a number of ways. From a theoretical viewpoint, our environment satisfies the informational assumptions of Bernheim and Whinston [1985] and Bernheim and Whinston [1986]. Hence, as long as the agency is risk-neutral, this particular objective function may be the result of an underlying common agency problem. More relevant from an empirical viewpoint, the agency contracts most commonly used in this industry specify a lump-sum fee per advertiser and per campaign. Thus, the agency’s ability to generate surplus for its clients is an important determinant of its long run profitability.} denoted by \( u_C(b) := \sum_{i \in C} u_i(b) \), under the three constraints (\textbf{S.1}) (\textbf{S.2}) and (\textbf{R}) discussed above, which we formally introduce next:

(\textbf{R}): \( R = \{R(C)\}_{C \in \mathcal{C}} \) denotes the collection of exogenous restrictions for all possible coalitions, and for each \( C \), \( R_C \subseteq A_C \) denotes the coalition bids consistent with \( R(C) \):

\[
R_C := \{b_C \in A_C : \exists b_{-C} \in A_{-C} \text{ s.t. } (b_C, b_{-C}) \in R(C)\}.
\]  

(\textbf{S.1}): For any \( i \in I \setminus C \), let \( BR^*_i : A_{-i} \Rightarrow A_i \) denote some refinement of \( i \)'s best
response correspondence in the baseline game $G$ (e.g., truthful bidding in the VCG, or (2) in the GSP). Define the independents’ equilibrium correspondence $BR^*_C : A_C \Rightarrow A_{−C}$ as

$$BR^*_C (b_C) = \{b_{−C} \in A_{−C} : \forall j \in I \setminus C, \ b_j \in BR^*_C (b_C, b_{−j,−C})\}.$$  

(4)

If the agency proposes a profile $b_C$ that is not consistent with the equilibrium behavior of the independents (as specified by $BR^*_C$), then that proposal does not induce a stable agreement. We thus incorporate this stability constraint into the agency’s optimization, and assume that the agency can only choose bid profiles from the set

$$S_C = \{b_C \in A_C : \exists b_{−C} \text{ s.t. } b_C \in BR^*_C (b_C)\}.$$ \[5\]

**(S.2):** The agency forms conjectures about how its bids $b_C$ the independents’ bids in the continuation game. Let $\beta : S_C \to A_{−C}$ represent such conjectures, and define the set of conjectures which are consistent with the independents playing an equilibrium:

$$B^* = \{\beta \in A^*_{−C} : \beta (b_C) \in BR^*_C (b_C) \text{ for all } b_C \in S_C\}.$$ \[6\]

The second stability condition requires that, given conjectures $\beta$, no coalition member $i \in C$ has an incentive to leave and bid as an independent in the game with coalition $C \setminus \{i\}$. This constraint thus requires a recursive definition. To this end, first let $E^* \equiv \{b \in \mathbb{R}_+^n : b_i \in BR^*_C (b_{−i}) \text{ for all } i \in I\}$ denote the set of equilibria in the game without coalition (given refinement $BR^*_C$). Then, letting $E^R (C')$ denote the set of Recursively Stable Agency Equilibrium (RAE) outcomes of the game with coalition $C'$, given restrictions $\mathcal{R}$ (and refinement $BR^*_C$), we initialize the recursion setting $E^R (C') = E^*$ if $|C'| = 1$ (that is, if an agency controls only one bidder, then the RAE are the same as the competitive equilibria). Suppose next that $E^R (C')$ has been defined for all subcoalitions $C' \subset C$. For each $i \in C$, and $C' \subseteq C \setminus \{i\}$, let $\bar{u}_i^C = \min_{b \in E^R (C')} u_i (b)$. Then, recursively:

**Definition 1** A Recursively Stable Agency Equilibrium (RAE) of the game $G$ with coalition $C$, given restrictions $\mathcal{R} = \{R (C)\}_{C \subseteq C}$ and refinement $BR^*_C$, is a profile of bids and conjectures $(b^*, \beta^*) \in A_C \times B^*$ such that \[12\]

1. The independents play a best response: for all $i \in I \setminus C$, $b^*_i \in BR^*_C (b^*_{−i})$.

2. The conjectures of the agency are correct and consistent with the exogenous restrictions: $\beta^* (b^*_C) = b^*_{−C}$ and $(b_C, \beta^* (b_C)) \in R (C)$ for all $b_C \in R_C$.

3. The agency best responds to conjectures $\beta^*$, subject to the exogenous restrictions

\[11\]The strength of constraint \[5\] clearly depends on the underlying game $G$ and on the particular correspondence $BR^*_C$. This restriction is conceptually important, and needed to develop a general framework for arbitrary mechanisms, but it plays no role in our results, since \[5\] will be either vacuous (Theorem 1) or a redundant constraint (Theorems 2 and 3).

\[12\]Note that, by requiring $\beta^* \in B^*$, this equilibrium rules out the possibility that the coalition’s bids are sustained by ‘incredible’ threats of the independents.
(R) and the stability restrictions (S.1) and (S.2):

\[ b^*_C \in \arg \max_{b_C} u_C(b_C, \beta^*(b_C)) \]

subject to: (R) \( b_C \in R_C \)

- (S.1) \( b_C \in S_C \)

- (S.2) for all \( i \in C \), \( u_i(b_C, \beta^*(b_C)) \geq u^C_{i}(i) \)

The set of (R-constrained) RAE outcomes for the game with coalition \( C \) is:

\[ E^R(C) = \{ b^* \in A : \exists \beta^* \text{ s.t. } (b^*, \beta^*) \text{ is a RAE} \} . \] (7)

We will refer to the case in which \( R(C) = A \) for all \( C \in C \) as the ‘unconstrained’ case, and denote the set of unconstrained RAE outcomes as \( E(C) \).

Before moving to the general results of the next section, we first illustrate the logic of this definition in the context of a simple example. In the example, as well as in some results in Section 4, equilibrium bids will sometimes be such that \( b_i = b_{i+1} \) for some \( i \). Since ties are broken according to bidders’ labels (cf. footnote 7), in that case bidder \( i \) obtains the position above \( i + 1 \). To emphasize this, we will write \( b_i = b^+_{i+1} \).

Example 2 Consider an environment with five bidders who compete for the allocation of four slots sold through the VCG mechanism. Bidders’ valuations are \( v = (40, 25, 20, 10, 9) \), and the CTRs are \( x = \{20, 10, 9, 1, 0\} \). As discussed in Section 2, in this mechanism advertisers bid truthfully in the competitive benchmark, and hence equilibrium payoffs for the five bidders are \( u^{\text{Comp}} = (441, 141, 91, 1, 0) \).

Now consider a setting in which bidders 1 and 5 belong to the same agency, \( C' = \{1, 5\} \), and everyone else is an independent. Bidding truthfully remains a dominant strategy for the independents, but clearly this is not the case for the agency: since 1’s payment is strictly decreasing in \( b_5 \), the optimal solution for the agency is to lower \( b_5 \) as much as possible, while ensuring that 1 keeps the first position. Hence, any profile \( b' = (b'_1, 25, 20, 10, 0) \) such that \( b'_1 > 25 \) is an (unconstrained) RAE when \( C' = \{1, 5\} \), and the resulting payoffs are \( u' = (450, 150, 100, 10, 0) \), with a total 450 for the coalition. Comparing \( u' \) with \( u^{\text{Comp}} \), it is also clear that no member of the coalition would rather bid as an independent.

Next, suppose that the coalition also includes bidder 2: \( C'' = \{1, 2, 5\} \). In this case, the (unconstrained) RAE-bids are \( b'' = (b''_1, 20^+, 20, 10, 0) \), where \( b''_1 > 20 \), which induce payoffs \( u'' = (500, 150, 100, 10, 0) \) and a total of 650 for the coalition. To see that this is

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13 Without the tie-breaking rule embedded in \( \rho \) (footnote 7), the agency’s best replies may be empty valued. In that case, our analysis would go through assuming that bids are placed from an arbitrarily fine discrete grid (i.e., \( A_i = (R_i \cap \varepsilon \mathbb{Z}) \) where \( \varepsilon \) is the minimum bid increment). In that setting, \( b_i = b^+_{i+1} \) can be thought of as \( i \) bidding the lowest feasible bid higher than \( b_{i+1} \), i.e. \( b_i = b_{i+1} + \varepsilon \). All our results would hold in such a discrete model, once the equilibrium bids in the theorems are interpreted as the limit of the equilibria in the discrete model, letting \( \varepsilon \to 0 \) (the notation \( b^+_{i+1} \) is thus reminiscent of this alternative interpretation, as the right-hand limit \( b^+_{i+1} := \lim_{\varepsilon \to 0} (b_{i+1} + \varepsilon) \)).
a RAE, recall that truthful bidding is still dominant for the independent bidders. The argument for keeping \( b''_5 = 0 \) and \( b''_1 > 20 \) are the same as above. As for \( b_2 \), first note that, if the agency set \( b_2 = 10^+ \), pushing bidder 2 down to the third slot, then the coalition payoff would be 655, which is higher than 650. But, in such a profile, 2’s payoff would be 145, which is lower than \( u'_2 = 150 \), the payoff he could obtain if he bid as an independent in the game with \( C' = \{1, 5\} \). Hence, lowering \( b_2 \) to the point of obtaining a lower position would increase the overall coalition payoff (by decreasing bidder 1’s payment), but would violate the stability constraint (S.2) for bidder 2. Hence, the optimal \( b''_2 \) is the lowest bid which ensures that bidder 2 maintains the second position. □

Note that the recursive definition of the outside option matters in this example: If outside options were defined with respect to the competitive case, bidder 2 would remain in the coalition even when forced to take the lower position, since his payoff in the competitive benchmark is \( u^{comp}_2 = 141 < 145 \). But we find it unreasonable to model 2’s outside option this way: why would an agency client assume that, were he to abandon the agency, the entire coalition would be disrupted and full competition restored? The recursivity of the (S.2) constraint reflects these considerations. Finally, the example also shows that RAE outcomes in general are not Nash equilibria of the baseline game, nor of the game in which the coalition is replaced by a single player. Similar to Ray and Vohra (1997) and Ray and Vohra (2014) equilibrium binding agreements (which we discuss in Section 5), the stability restrictions affect the set of equilibrium outcomes, not merely as a refinement.

4 Agency Bidding in VCG and GSP: Results

In this Section we specialize the general notion of RAE to the GSP and VCG mechanisms:

**Definition 2 (RAE in the GSP and VCG)** Given a set of exogenous restrictions \( R \), the \( R \)-constrained RAE of the GSP and VCG mechanisms are obtained from Definition 1 letting \( G \) denote the corresponding game, and \( BR^*_i \) be defined, respectively, as in (2) for the GSP and as the dominant (i.e., truthful) strategy in the VCG.

We first present the analysis of the VCG mechanism (Section 4.1), and then proceed to the GSP auction (Section 4.2). Our main conclusion is that the VCG outperforms the GSP both in terms of revenues and allocative efficiency, thereby uncovering a striking fragility of the GSP with respect to agency bidding.

4.1 Agency Bidding in the VCG mechanism

Our first result characterizes the unconstrained RAE of the VCG mechanism: it shows that they are unique up to the bid of the highest coalition member, and that in all such equilibria advertisers are assigned to positions efficiently, independents’ bids are equal to their valuations and all the coalition members (except possibly the highest) bid the lowest possible value that ensures their efficient position. Formally:
Theorem 1 (RAE in the VCG) For any $C$, let $E(C)$ denote the unconstrained RAE of the VCG. Then: $\hat{b} \in E(C)$ if and only if

$$\begin{align*}
\hat{b}_i &= v_i & \text{if } i \in I \setminus C; \\
\hat{b}_i &= \hat{b}_{i+1}^+ & \text{if } i \in C \setminus \{\min(C)\} \text{ and } i \leq S; \\
&\in \left(\hat{b}_{i+1}^+, v_{i-1}\right) & \text{if } i = \min(C) \text{ and } i \leq S.
\end{align*}$$

(8)

where we denote $v_0 := \infty$ and $\hat{b}_{n+1} := 0$.

The unconstrained RAE of the VCG mechanism therefore are efficient, with generally lower revenues than in the VCG’s competitive benchmark. The efficiency result is due to the stability restrictions in RAE, which limits the agency’s freedom to place bids. Restriction (S.2), in particular, requires that the agency’s proposal gives no member of the coalition an incentive to abandon it and bid as an independent. Similar to the illustrative example 2, a recursive argument further shows that the payoff that any coalition member can attain from abandoning the coalition is bounded below by the equilibrium payoffs in the baseline (coalition-less) game, in which assignments are efficient. The ‘Pigouvian’ logic of the VCG payments in turn implies that such (recursive) participation constraints can only be satisfied by the efficient assignment of positions. As shown by example 2 the recursive stability restriction (S.2) is key to this efficiency result.

Whereas the presence of an agency does not alter the allocation of the VCG mechanism, it does affect its revenues: in any RAE of the VCG mechanism, the agency lowers the bids of its members (except possibly the one with the highest valuation) as much as possible, within the constraints posed by the efficient ranking of bids. Since, in the VCG mechanism, lowering the $i$-th bid affects the price paid for all slots $s = 1, \ldots, \min\{S + 1, i - 1\}$, even a small coalition can have a significant impact on the total revenues. On the other hand, the VCG’s strategy-proofness ensures that the agency has no impact on the independents, which continue to use their dominant strategy and bid truthfully. Hence, while an agency may have a large ‘direct effect’ on revenues, it has no ‘indirect effect’ in this mechanism.

Example 3 Consider the environment in Example 2 and suppose that $C = \{1, 3\}$. Then, applying the formula in (8), the RAE of the VCG mechanism is $\hat{b} = \left(\hat{b}_1, 4, 2^+, 2, 1\right)$. The resulting revenues are 86, as opposed to 96 of the competitive benchmark. □

4.2 Agency Bidding in the GSP auction

We begin our analysis of the GSP auction by characterizing the RAE when the agency is constrained to placing bids which, to an external observer, are undistinguishable from a (competitive) EOS equilibrium – the ‘Undistinguishable (from EOS) Coordination’ (UC) restriction. Theorem 2 shows that the equilibrium outcomes of the GSP with this restriction are exactly the same as the unrestricted RAE of the VCG mechanism. We lift the

\footnote{This property also ensures that $S_C = A_C$. Hence, constraint (S.1) in Def. 1 plays no role in the result.}
UC-restriction in Section [12.2], and show that the GSP’s RAE may be inefficient and induce strictly lower revenues than their VCG counterparts. Moreover, the revenue ranking holds even if the agency is restrained from inducing allocative distortions (Theorem 3).

4.2.1 ‘Undistinguishable Coordination’: A VCG-Equivalence Result

Consider the following set of exogenous restrictions: for any $C \in C$,

$$R_{UC}(C) := \left\{ b \in A : \exists v'_C \in \mathbb{R}^{|C|}_+ \text{ s.t. } b \in E^* \left( v'_C, v_{-C} \right) \right\}. \tag{9}$$

In words, $R_{UC}(C)$ is comprised of all bid profiles that could be observed as part of an EOS equilibrium in the GSP auction, given the valuations of the independents $v_{-C} = (v_j)_{j \in I \setminus C}$. For instance, consider an external observer (e.g., the search engine or the antitrust authority) who can only observe the bid profile, but not the valuations $(v_i)_{i \in C}$. Then, $R_{UC}(C)$ characterizes the bid profiles that ensure the agency’s bidding strategy could not be distinguished from an EOS equilibrium (and, hence, detected as ‘collusive’), even if the independents had revealed their own valuations to the external observer.\footnote{This formulation of the UC-constraint is consistent with our choice to use EOS equilibrium as the competitive benchmark, as it has become standard in the literature. The definition in (9), however, may easily accommodate alternative benchmarks too. As explained in Section 3, alternative models of competitive behavior could be accommodated in the definition of RAE by replacing (2) with the corresponding refinement of individual best-responses. In that case, the set $E^* \left( v'_C, v_{-C} \right)$ in (9) would consists of the fixed points of such individual best responses, i.e. the Nash-equilibria of the GSP taken as benchmark of competitive bidding. Hence, whatever refinement of Nash equilibria is taken as model of competitive behavior in the GSP – and, hence, embedded in the definition of RAE – the set $R_{UC}$ denotes the set of bids profile which cannot be distinguished from that competitive benchmark.}

The next result characterizes the RAE of the GSP under these restrictions, and shows its revenue and allocative equivalence to the unrestricted RAE of the VCG:

**Theorem 2** For any $C$, let $v_{i+1} = 0$, and for each $i = n, ..., 1$, let $v'_i := v_{i+1}^f$ if $i \in C$ and $v'_i = v_i$ if $i \notin C$. Then, in any RAE of the GSP auction under the ‘undistinguishable coordination’ (UC) restriction, the bids profile $\hat{b}$ are such that, for every $i$,

$$\hat{b}_i = \begin{cases} v'_i - \frac{x_i}{x_i - 1} (v'_i - \hat{b}_{i+1}), & \text{if } i \neq 1 \text{ and } i \neq \min(C); \\ \in \left[ v'_i - \frac{x_i}{x_i - 1} (v'_i - \hat{b}_{i+1}), \hat{b}_{i-1} \right), & \text{otherwise} \end{cases}, \tag{10}$$

where $\hat{b}_0 := \infty$ and $x_i/x_{i-1} := 0$ whenever $i > S$. Moreover, in each of these equilibria, advertisers are assigned to positions efficiently, and advertisers’ payments are the same as in the corresponding unrestricted RAE of the VCG mechanism (Theorem 7).
equilibrium behavior: intuitively, in order to satisfy the UC-restriction, the agency’s bids for each of its members should mimic the behavior of an independent advertiser in the competitive benchmark, for some valuation. The agency’s problem therefore boils down to ‘choosing’ a \textit{feigned valuation} for each of its members, and bid accordingly. The optimal choice of the feigned valuation is the one which, given others’ bids, and the bidding strategy of an independent, induces the lowest bid consistent with \(i\) obtaining the \(i\)-th position in the competitive equilibrium of the model with feigned valuations, which is achieved by \(v_i^f = v_{i+1}^f\). Note that the fact that bidder \(i\) cannot be forced to a lower position is not implicit in the UC-restriction, but the result of the equilibrium restrictions.\(^{16}\) The resulting allocation is efficient, and it yields the same individual payments (and hence total revenues) as the unrestricted RAE of the VCG mechanism.

To understand the implications of this equilibrium, note that, in the GSP auction, the \(i\)-th bid only affects the payment of the \((i-1)\)-th bidder. Hence, the ‘direct effect’ of bids manipulation is weaker in the GSP than in the VCG mechanism, where the payments for all positions above \(i\) are affected. Unlike the VCG mechanism, however, manipulating the bid of coalition member \(i\) also has an ‘indirect effect’ on the bids of all the independents placed above \(i\), who lower their bids according to the recursion in (10).

\textbf{Example 4} Consider the environment of Example 3 with \(C = \{1, 3\}\). Then, applying the formula in (10), the UC-RAE is \(\hat{b} = (\hat{b}_1, 2.9, 1.8, 1.6, 1)\), which results in revenues 86. These are the same as in the VCG mechanism (Example 3), and 10 less than in the non-agency case (Example 1). Note that the bid \(\hat{b}_3 = 1.8\) obtains setting \(v_3^f = v_4 = 2\), and then applying the same recursion as for the independents. Also note that the ‘direct effect’, due to the reduction in \(\hat{b}_3\), is only equal to \((b_3^{EOS} - \hat{b}_3) \cdot x_2 = 5\) (where \(b_3^{EOS}\) denotes 3’s bid in the non-agency benchmark). Thus, 50% of the revenue loss in this example is due to the agency’s ‘indirect effect’ on the independents. \(\square\)

Thus, despite the simplicity of the payment rule in the GSP auction, the equilibrium effects in (10) essentially replicate the complexity of the VCG payments: once the \textit{direct} and \textit{indirect effects} are combined, the resulting revenue loss is the same in the two mechanisms. This result also enables us to simplify the analysis of the impact of agency bidding on the GSP, by studying the comparative statics of the unconstrained RAE in the VCG mechanism. We can thus obtain some qualitative insights for this complex problem.

\textbf{Corollary 1} \textit{Hold the agency configuration, }\(C\), \textit{constant. Then, in both the unconstrained RAE of the VCG and in the UC-RAE of the GSP auction, the revenue losses due to agency bidding are larger if: (i) the differences \((x_{i-1} - x_i)\) associated to the agency’s clients }\(i \in C\) \textit{are larger; or if (ii) the difference in valuations between the agency’s clients and the independents immediately below them in the ranking of valuations are larger.}

\(^{16}\)The reason is similar to that discussed for Theorem 1, only here is more complicated due to the fact that, in the GSP auction, the bids of the agency alter the bids placed by the independents.
To understand this Corollary, recall that the price-per-click for position \( s \) in the VCG, given a profile \( b \), is equal to \( \sum_{t=s+1}^{S+1} b^t (x^{t-1} - x^t) \). By Theorem 1 in the RAE of the VCG the agency lowers the bids of its members as much as possible, while preserving the efficient ranking of bids. Hence, holding \( C \) and \((v_i)_{i \in I}\) constant, it is clear that the revenue losses due agency bidding are larger if the terms \((x^{t-1} - x^t)\) associated to agency members are larger, which is part (i) of the Corollary. To understand part (ii), let \( i \) be an agency member such that \( i+1 \) is an independent. Since independents bid truthfully in the VCG, we have \( b_{i+1} = v_{i+1} \), and hence the efficient ranking can be maintained only if \( b_i \geq v_{i+1} \). Hence, the lower \( v_{i+1} \), the stronger the impact of agency bidding.

The next comparative statics refer to the agency composition. Besides the obvious statement that an agency’s impact is stronger if it includes more bidders, the impact of different coalitions in general depends on the exact CTRs and valuations. To isolate the position effects from the comparative statics in Corollary 1 which were driven by the differences \((x_s - x_{s+1})\) and \((v_s - v_{s+1})\), we assume that they are constant in \( s \).

**Corollary 2** Assume that \( \Delta_s(x) := (x_s - x_{s+1}) \) and \( \Delta_s(v) := (v_s - v_{s+1}) \) are constant in \( s \). Then, in both the RAE of the VCG and in the UC-RAE of the GSP, the revenue losses due to agency bidding are larger if the agency includes members that occupy adjacent or lower positions in the ranking of valuations.

To understand this result, note that if an agency has no two ‘adjacent’ members, then \( i+1 \) is an independent for all \( i \in C \), and hence for the above explanation the lower bound to \( i \)’s bid equals \( v_{i+1} \). But if instead \( i+1 \) also belongs to the agency, then the lower bound drops to the valuation of the next lower independent. The rest of the Corollary follows directly from the fact that a given reduction of a bid in the VCG has a larger impact if it’s lower in the ranking, because it affects the payments for all positions above. The latter point is particularly interesting, since one might have expected that the agency would have a larger impact if she controlled the high-valuation bidders. We find that, in fact, the opposite is true when one controls for the increments \( \Delta_s(x) \) and \( \Delta_s(v) \).

### 4.2.2 Lifting the UC-Restriction: Revenue Losses and Inefficiency

As discussed in Section 4.1 even a small coalition of bidders may have a large impact on revenues in the VCG. Theorem 2 therefore already entails a fairly negative outlook on the GSP’s revenues when an agency is active, even if it cannot be detected as collusive, because it is undistinguishable from an EOS equilibrium. The next example shows that, when the undistinguishability constraint is lifted, an agency may induce larger revenue losses as well as inefficient allocations in the GSP auction.

**Example 5** Consider an environment with 8 bidders and 7 slots, with valuations \( v = (12, 10.5, 10.4, 10.3, 10.2, 10.1, 10.0, 0) \) and CTRs \( x = (50, 40, 30, 20, 10, 2, 1, 0) \). Let the coalition be \( C = \{5, 6\} \). The unrestricted RAE is essentially unique (up to the highest
overall bid) and inefficient, with the coalition bidders obtaining slots 4 and 6. Equilibrium bids (rounding off to the second decimal) are \( b = (b_1, 9.91, 9.76, 9.12, 9.5, 7.94, 5.5, 1) \). Note that \( b_4 = 9.12 < 9.5 = b_5 \), which induces an inefficient allocation. The inefficiency arises as follows. Suppose that the agency drastically lowers \( b_6 \) to benefit the other member. If \( b_6 \) is very low, it creates incentives for the independents \( i < 5 \) to move down to the position just above bidder 6, in order to appropriate some of the rents generated by its lower bid. Hence, if efficiency were to be preserved, 5’s bid would also have to be reduced, to make the higher positions more attractive. But the reduction of 6’s bid in this example is large enough that 4’s undercut is sufficiently low that the coalition prefers to give up position 5. Thus, the coalition does not benefit directly from the reduction of 6’s bid, but indirectly, by attracting 4 to the lower position. □

Hence, unlike the VCG mechanism, the unrestricted RAE of the GSP auction can be inefficient. In light of this result, it may appear that the unconstrained-RAE in the GSP allows an implausible degree of freedom to the agency, and that this alone is the cause of the low revenues of the GSP auction. To see whether this is the case, we consider next exogenous restrictions that force the agency to induce efficient allocations. Theorem 3 shows that, even with this restriction, the GSP’s revenues are no higher than in the unrestricted RAE of the VCG mechanism. Formally, let \( R^{EFF} = \{ R^{EFF}(C) \}_{C \in C} \) be such that, for each non-trivial coalition \( C \in C \),

\[
R^{EFF}(C) := \{ b \in A : \rho(i; b) = i \ \forall i \in I \}.
\]

**Definition 3** An efficiency-constrained RAE of the GSP auction is a RAE of the GSP auction where the exogenous restrictions are given by \( R = R^{EFF} \).

**Theorem 3** Efficiency-constrained RAE of the GSP auction exist; in any such RAE: (i) the agency’s payoff is at least as high as in any RAE of the VCG mechanism, and (ii) the auctioneer’s revenue is no higher than in the corresponding equilibrium of the VCG auction. Furthermore, there exist parameter values under which both orderings are strict.

By imposing efficiency as an exogenous constraint, Theorem 3 shows that the fragility of the GSP’s revenues is independent of the allocative distortions it may generate. The intuition behind Theorem 3 is simple, in hindsight: in the VCG mechanism, truthful bidding is dominant for the independents, and hence the agency’s manipulation of its members’ bids only has a direct effect on revenues. In the GSP auction, in contrast, the agency has both a direct and an indirect effect. Under the UC-restrictions, the two effects combined induce just the same revenue-loss as in the VCG mechanism, but lifting that restriction tilts the balance, to the disadvantage of the GSP.\(^{17}\)

\(^{17}\)Since the UC-RAE induce efficient allocations, it may seem that Theorem 3 follows immediately from the efficiency constraint being weaker than the UC-restriction. This intuition is incorrect for two reasons. First, the UC-constraint requires the existence of feigned valuations which can rationalize the observed...
Table 1: Summary of Results in Examples

<table>
<thead>
<tr>
<th>Valuations</th>
<th>VCG</th>
<th>GSP (EOS)</th>
<th>RAEX in VCG</th>
<th>UC-RAEX in GSP</th>
<th>(Eff.) RAEX in GSP</th>
</tr>
</thead>
<tbody>
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<td>5</td>
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</tr>
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</tr>
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<td>Revenues</td>
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<td>96</td>
<td>86</td>
<td>86</td>
<td>82</td>
</tr>
</tbody>
</table>

Summary of results in Examples 1, 3, 4 and 6. Coalition members' bids and valuations are in bold. The VCG and GSP columns represent the competitive equilibria in the two mechanisms as described in example 1. The RAEX in VCG and the revenue equivalent UC-RAEX in the GSP are from Examples 3 and 4 respectively. The last column denotes both the Efficient RAEX and the unrestricted RAEX of the GSP auction, which coincide in Example 6.

**Example 6** Consider the environment of Examples 3 and 4 with \(C = \{1, 3\}\). The efficiency-constrained RAEX is \(\hat{b} = (\hat{b}_1, 2.8, 1.6^+, 1.6, 1)\), which results in revenues 82, which are lower than the RAEX in VCG mechanism (86). Note that, relative to the UC-RAEX in Example 4, the coalition lowers \(b_3\) to the lowest level consistent with the efficient ranking. This in turn induces independent bidder 2 to lower his bids, hence the extra revenue loss is due to further direct and indirect effects. We note that the efficiency restriction is not binding in this example, and hence the Eff-RAEX and the unconstrained RAEX coincide.

(Table 1 summarizes and compares the equilibria illustrated in our running examples.)

Summing up, since – under the efficiency restriction – the GSP auction induces the same allocation as the VCG mechanism, the two mechanisms are ranked in terms of revenues purely due to the agency’s effect on prices. Obviously, if allocative inefficiencies were introduced, they might provide a further, independent source of revenue reduction. As already noted, this is not the case in Example 6, in which the efficiency constraint is not binding, but it is possible in general (see Example 5).

The unrestricted RAEX of the GSP are difficult to characterize, and (as shown with Ex. 5) allow perhaps too much freedom to the agency\(^{18}\). The extra restrictions enabled by the Eff-RAEX or UC-RAEX may thus prove be more fruitful from an applied perspective, since bid profile, but does not require that they preserve the ranking of the true valuations. Second, when the exogenous restrictions \(R = (R_C)_{C \subseteq C}\) are changed, they change for all coalitions: hence, even if \(R_C\) is weaker for any given \(C\), the fact that it is also weaker for the subcoalitions may make the stability constraint (S.2) more stringent.

\(^{18}\)Short of a characterization, one could consider whether there are clear revenue rankings for the GSP’s unrestricted RAEX. Meaningful revenue comparisons, however, require normalizing the criterion to break the agency’s indifference over her highest bid – which does not affect the agency’s payoff (beyond the position it ensures), but does affect the revenues. This indeterminacy does not create problems for the results above, since it is preserved uniformly across the mechanisms and restrictions we consider; but when all restrictions are lifted, the mere breaking of indifference may impact revenues asymmetrically under different mechanisms and restrictions. Standard tie-breaking criteria, however, do enable natural revenue comparisons. For instance, if one applied the same logic used to refine the independents’ indifference (i.e., the locally envy-free criterion implicit in (2)), or if one considered the lowest-revenue selections among the bids which maximize the agency’s payoff, then it could be shown that the GSP’s revenues in an inefficient RAEX are never higher (and typically lower) than in the Eff-RAEX.
they represent outcomes that are computationally easier to attain for an agency.

4.3 Agency Competition

Multiple agencies competing in the same auction appears rarely in the data (Decarolis, Goldmanis and Penta [2018], but for the reasons explained in the introduction, it is nevertheless interesting to assess whether competition may soften the impact of agency bidding on online ad auctions. This is a reasonable conjecture, but the results we present in this section suggest a more nuanced view on this point. In particular, for certain coalition structures, our earlier results extend to the case with multiple agencies essentially unchanged: the revenue losses will be less pronounced when the same set of coordinating bidders is divided into two (or more) competing coalitions, but they would still be substantial, and preserve the relative performance of the VCG and GSP auctions. But, for other coalition structures, equilibria in pure strategies will not exist, and hence bidding cycles are likely to emerge.\footnote{Hence, while competition between agencies may indeed mitigate the agencies’ on the platforms’ revenues, it may also impair the working of the current mechanisms in a more fundamental way.}

For simplicity, we consider the case with two agencies (the extension to more than two agencies is cumbersome but straightforward). We also assume that agencies break indifferences over bids in the same way that independents do. This implies that the highest bidder in any coalition bids as if he were an independent. With the formal definitions given in Appendix A.3 the following result holds.

**Theorem 4**

1. If no members of different coalitions occupy adjacent positions in the ordering of valuations, then the UC-RAE of the GSP with multiple coalitions is unique. In this equilibrium, the allocation is efficient and the search engine revenues are weakly higher than those of the UC-RAE in which all members of the different coalitions bid under the same agency, but lower than under full competition. Moreover, both the allocation and the associated revenues are identical to those resulting in the unconstrained RAE of the VCG mechanism with the same agency configuration.

2. If non-top members of different coalitions occupy adjacent positions in the ranking of valuations, no unconstrained RAE of the VCG and no UC-RAE of the GSP exist.

The first part of the theorem extends Theorems 1 and 2 to the case of multiple agencies. The result therefore shows that competition between agencies may mitigate, but not solve, the revenue losses due to coordinated bidding. If coalitions have bidders in adjacent positions (part 2 of the Theorem), further problems arise, such as non-existence of pure-strategy equilibria and bidding cycles. We illustrate both these points in the context of our workhorse example.
Example 7 Consider the environment of the examples in Table 1. Table 2 reports EOS’ equilibrium bids (second column) as well as the bids under different coalition structures. We first look at the case of a single coalition \( C = \{1, 2, 4, 5\} \). According to our earlier results, in the UC-RAE with this agency configuration the bottom two bidders bid zero. This has an indirect effect on the independent bidder (3), who lowers his bid from 2.3 to 1.5, thereby lowering the payments and bids for bidders 1 and 2. If we split this coalition into two separate coalitions, however, things will change depending on the way we do it. If we split \( C \) as in the fourth column of the table, \( C_1 = \{1, 4\} \) and \( C_2 = \{2, 5\} \), we obtain two coalitions with no adjacent members, as in part 1 of Theorem 4. With this coalition structure, equilibrium revenues amount to 88, which is above the single coalition case (60), but still well below the competitive benchmark (96). If we split \( C \) as in the last column of Table 2, \( C_1 = \{1, 4\} \) and \( C_2 = \{2, 5\} \), pure equilibria would cease to exist. To see this, note that \( C_2 \) would ideally like to set \( b_5 = 0 \), and given this \( C_1 \) would ideally like to set \( b_4 = 0 \). This, however, is incompatible with an equilibrium because once \( b_4 = 0 \), \( C_2 \) would find it profitable to increase \( b_5 \) so as to obtain a higher position, with a negligible increase in its payments. On the other hand, if \( b_4 \) is set so high that \( C_2 \) does not find this deviation profitable, then \( C_2 \)’s optimal response is to set \( b_5 = 0 \). But then, a strictly positive \( b_4 \) cannot be optimal for \( C_1 \). Hence, a pure equilibrium does not exist. □

Part 2 of Theorem 4 shows that this phenomenon emerges whenever two coalitions have non-top members which occupy contiguous positions in the ordering of valuations. It is interesting to note that the behaviors behind this phenomenon is nearly identical to that explained by Edelman and Ostrovsky (2007) in their characterization of the original Generalized First Price (GFP) auction, under which the market started, to explain the bidding cycles observed in the data. As discussed earlier, such bidding cycles are considered the GSP. (For a discussion of bidding cycles in the Overture’s GFP, see Edelman and Ostrovsky (2007); Ottaviani (2003) provides an early assessment of the transition from the GFP to the GSP.)

Note that, if the highest placed member of the lower coalition (i.e., the bidder with a value of 2 in this example) were to slightly increase/decrease his bid, his coalition’s payoffs would not change, but the revenues of the other coalition would correspondingly decrease/increase. Hence, without the assumption that top coalition members behave as independents, a multiplicity of equilibria might arise. Different selections from the best-response correspondence may thus be used to model other forms of behavior, such as spiteful bidding (cf., Levin and Skrzypacz (2016)).
to be the main cause for the shift from the GFP to the GSP auction. The fact that a similar phenomenon emerges here with multiple agencies may thus be seen as a troubling result for the existing mechanisms, in that it suggests that agency competition, instead of mitigating the impact of agency bidding, could exacerbate the system’s instability.

From an empirical perspective, these results suggest further directions of research, since they imply that bidding cycles are more likely to be observed as agency competition spreads in this market, and especially so for certain configurations of agency membership.

5 Discussion: Related Literature and Extensions

RAE and EBA. Our notion of RAE is closely related to the ‘Equilibrium Binding Agreements’ of Ray and Vohra (1997, RV hereafter). Given a certain coalition structure, RV postulate that binding agreements are possible within a coalition. The objective is to endogenize the collection of agreements such that no subcoalition has an incentive to break the agreement and separate from the original coalition. Moreover, when considering such deviations, the subcoalition is ‘farsighted’ in the sense that it does not take the behavior of the other coalitions as given, nor does she assume that the remaining members of the coalition will band together. Instead, it tries to predict the coalition structure and the agreements that would ultimately arise as a result of its deviation. In equilibrium, such predictions are required to be correct. Because of the ‘farsightedness assumption’, RV’s equilibrium is defined recursively, as is our RAE.

RV’s and our approach share the same fundamental philosophy. Like RV, we also maintain that binding agreements are only possible within the coalition, but the interaction between the agency and the independents, as well as among the independents, is fully non-cooperative. As in RV, the agency in our model is a proposer of a binding agreement, subject to certain stability constraints, which crucially incorporate RV’s farsightedness assumption. Relative to RV, our approach differs mainly in that our stability restriction (S.2) only allows agency proposals to be blocked by individual members, whereas RV allow for any joint deviation of coalition members. That advertisers can make binding agreements outside the agency, and jointly block its proposals, seems unrealistic in this context. Hence, a direct application of their concept to this setting seems inappropriate. Also, unlike RV (in which the non-cooperative interaction is based on Nash equilibrium), our definition also allows for refinements. As already explained, this is crucial here, especially for the analysis of GSP auction.

Bidding Rings and Partial Cartels. The literature on bidding rings in auctions (e.g., Graham and Marshall (1987), Mailath and Zemski (1991), McAfee and McMillan (1992)
Hendricks, Porter and Tan (2008) also addresses related phenomena, but from a very different perspective. In particular, the main focus of this literature is on whether members of the coalition may be incentivized to share their private information so as to implement collusion, a moot point under EOS’ complete information assumption. Maintaining EOS and Varian’s complete information setting, we implicitly abstract away the information extraction problem within the coalition. Furthermore, we don’t allow transfers between members of the coalition. Other mechanisms for collusion have been considered, for instance, by Harrington and Skrzypacz (2007) and Harrington and Skrzypacz (2011). More importantly, a key feature of our setting is the co-presence of coordinated and independent bidding. Combining cooperative and non-cooperative interaction is a well-known challenge in this literature, which either considered mechanisms in which non-cooperative behavior is straightforward (e.g., second price auctions with private values, as in Mailath and Zemsky (1991)), or has assumed that the coalition includes all bidders (as in the first price auctions of McAfee and McMillan (1992), and Hendricks, Porter and Tan (2008), or in the dynamic auctions of Ortner and Chassang (2018) in a different setting). The notion of RAE enables us to combine cooperative and non-cooperative interaction in general mechanisms, even if non-cooperative behavior is complex. The results above perhaps suggest that the general concept of RAE (or other concepts based on RV’s approach) may provide a valuable methodological contribution from a broader theoretical perspective, to overcome some of the difficulties involved with modeling partial cartels in auctions.

Alternative Competitive Benchmarks. The prior literature has shown that equilibria in the GSP auction, without using the spite move refinement concept of EOS, can be worse in terms of revenue and efficiency than in the VCG auction. Varian (2007) presents the EOS equilibrium as a lower bound of a class of NE that he refers to as ‘symmetric’ NE. He shows that the lower bound on revenues among all NE is generally less than the revenue bound for the symmetric NE (i.e., the EOS revenues). Borgers et al. (2013) go a step further and show that inefficient NE will typically exist. Given these negative results on the GSP outside the EOS equilibrium, one might wonder whether the poor performance of the GSP auction is due to collusive agency bidding or simply due to the fact that the equilibrium concept changed somewhat and the spite refinement has no bite anymore.

The response to this concern is that collusive agency bidding is the main driver of the poor performance of the GSP. First, the behaviors associated with the RAE equilibrium in the GSP will typically lead to coalition bids that are below what could be sustained in the revenue minimizing equilibrium of the corresponding competitive game. This is immediately clear when considering, for instance, the situation of coalition bidders occupying

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22Allowing transfers would relax constraint (S.2) in the definition of RAE, and affect our results (for instance, it may induce inefficiencies even in the VCG mechanism, cf. Example 2). That different advertisers make side-payments to each other seems implausible in this market. If indirect transfers could be implemented through dynamic effects (e.g., swapping bids for some of its members) or across different keywords, distinct strategic issues might arise, which would best be studied considering a richer model.
adjacent positions: their bids will typically be so low that the lowest among these coalition members, if he were to act as an independent, would find it individually profitable to raise his bid and jump to a higher position. The associated revenue loss therefore is directly due to agency bidding, and not merely to the possibility of low revenues among non-EOS competitive equilibria. Second, as we showed in Lemma 1, our notion of RAE maintains the same individual-level underpinning of EOS refinement, and in particular the features of EOS concepts which make the GSP ‘work well’. The EOS allocation and payments are embedded in our model as the end point of the recursion which defines the outside options of the coalition members. From this viewpoint, the fact that the VCG outperforms the GSP under the same conditions which – absent agency coordination – make the GSP perform better than it would if other refinements were considered, strengthens the result on the GSP’s fragility. Obviously, our characterizations do exploit the specific properties of our EOS-based refinement. Results based on alternative competitive benchmarks would require altogether different proofs. This we think would be an interesting enterprise for future research, but seems beyond the scope of the present paper. Nonetheless, as we explained in Section 3, the general notion of RAE lends itself to this kind of exercises, since it provides a tool to study the effect of agency bidding using different baseline refinements as plug-in (see also footnote 15).

Endogenous Participation and Alternative Approaches. An obvious extension to our approach would be to model bidders’ choice to join the agency explicitly. This would also be useful from an empirical viewpoint, as it would generate extra restrictions to further identify bidders’ valuations. Once again, however, the structure of the GSP auction raises non trivial challenges. First, it is easy to see that without an exogenous cost of joining the agency, the only outcome of a standard coalition formation game would result in a single agency consisting of the grand-coalition of players. Thus, the ‘obvious’ extension of the model would not be capable of explaining the lack of grand coalitions in the data. At a minimum, some cost of joining the coalition should be introduced. Clearly, there are many possible ways in which participation costs could be modeled (e.g., costs associated to information leakage, management practices, agency contracts, etc.). But given the still incomplete understanding of digital marketing agencies, it is not obvious which should be preferable. More empirical work is needed on this subject.

Independent of these modeling choices, however, the cost of joining the agency would

23Varian’s (2007) upper bound symmetric NE has also a recursive structure and, hence, it could be plugged-in as an alternative to the EOS in our analysis. However, while Varian’s (2007) offers insights on the type of individual logic that might support behaviors leading to this upper bound, this logic is less compelling than the one upon which the spite move of EOS is based. The upper bound would be reached if everyone thinks defensively of squeezing the profit of the player placed right above, but only up to the point where the player above does not prefer to jump down one position. Other refinements need not have that structure, and hence may entail significantly different strategy of proofs.

24Moreover, costs need not be symmetric, and hence it may be that an advertisers is willing to join the coalition, but current members are better-off without him. Whereas the decision to abandon an agency is unilateral, the decision to join it is not, raising further modeling questions.
ultimately have to be traded-off against the benefit, which in turn presumes solving for the equilibrium for a given coalition structure. Our work can thus be seen as a necessary first step in developing a full-blown model of agency formation. Exploring different specifications of such costs, and empirically assessing their relative merits, is thus an important direction for future research in this area.

Our formulation of the agency problem is also related to the literature on mediators in games, introduced by Monderer and Tennenholtz (2009) for complete information and extended by Ashlagi, Monderer and Tennenholtz (2009) to incomplete information, with an application to position auctions. Within this context, the issue of participation has been discussed, for instance, by Kalai (2010) and Roth and Shorrer (2018).

**Quality Scores.** In the variant of the GSP auction run by Google or Microsoft-Bing (but not, for instance, by Taobao), ‘quality scores’ concur in determining the assignment of advertisers to slots and prices: advertisers are ranked by the product of their bid and quality score, and pay a price equal to the minimum bid consistent with keeping that position. EOS and Varian (2007) showed how to extend their equilibrium characterization when quality scores are introduced, and assuming that they are common knowledge. Quality scores could be introduced in our model of collusive bidding in a way similar to EOS and Varian’s, delivering analogous characterizations of the results above. We pursue such an extension in Decarolis, Goldmanis and Penta (2018), to develop a criterion to detect various forms of collusion based on the variables which are typically contained in the datasets available to the auction platforms.

### 6 Conclusions

This is the first study to focus on the impact of coordinated bidding through intermediaries in the search auctions. It therefore contributes to the growing need of understanding both how firms operate on the platforms where online ad space is sold and how these platforms should be designed. Our results uncover a striking fragility of the GSP auction to bid coordination. Aside from its theoretical interest, this is a first order finding since most of the online marketing is still passing through GSP auctions. Our findings may also provide a rationale for why Facebook has recently adopted the VCG and Google is said to be considering the transition. Shifts between one mechanism and the other are

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25 While an in depth discussion of potential applications to data is beyond the scope of this paper, we refer to Ghose and Yang (2009) for an empirical model of the Google or Microsoft-Bing type of search auctions and to Hsieh, Shum and Yang (2015) for the case of Taobao auctions. Neither these two papers nor others we are aware of in the literature develop methods for estimating an empirical model of the search auctions in the presence of collusive bidding. This type of analysis is presented in a paper related to ours, Decarolis, Goldmanis and Penta (2018).

26 Competitive bidding with quality scores has also been studied by Athey and Nekipelov (2014), who introduced introducing uncertainty over quality scores.

27 The empirical analysis in Decarolis, Goldmanis and Penta (2018) shows that even the small two-bidder coalitions frequently observed in the data can have large effects on revenues.
important both for the large stakes involved and because the proper functioning of this
market is essential for both advertisers to reach consumers and for consumers to learn
about products.

From a methodological perspective, we note that the notion of RAE has been key to
obtain clear results in the complex GSP auction, and more broadly to accommodate the
coeexistence of competitive and coordinated bidding. This suggests that our approach,
which combines cooperative and non-cooperative ideas, may be fruitful to address the
important problem of partial cartels, an outstanding challenge in the literature.

Our results are also interesting from a market design perspective. While beyond the
scope of this paper, our analysis suggests some possible guidelines for research in this
area. For instance, our analysis of the GSP auction with ‘undistinguishable coordination'
constraints implicitly suggests a way of deriving reservation prices to limit the impact of
bid coordination. This kind of intervention would thus reinforce the resilience of the GSP
auction, without necessarily entailing major changes in the mechanism. The design of
auction formats more robust to collusion is a challenging task and the new formats that
might emerge could have profound implications on the profitability of one of today’s most
important industries.

Finally, our results have implications for competition policy. For competition author-
ities, ad auctions might be worth investigations for potential violations of the antitrust
laws, especially in those jurisdictions where price coordination is a violation per se, re-
gardless of any welfare implication. In fact, the multiple activities that DMAs undertake
beyond bid coordination make a priori ambiguous their overall effects on consumers’ wel-
fare. Furthermore, an additional complication is that, in the context of the ad auctions,
bid coordination by a DMA simply requires it to use bid algorithms that optimize joint
profits of its clients, without the need of any explicit communication. This poses a chal-
lenge for those authorities operating under jurisdictions that only sanction explicit (as
opposed to tacit) collusion. In this respect, our analysis offers a clear application of the
novel problems that algorithmic pricing poses for the enforcement of competition policy.

A Appendix

A.1 Technical Details

As discussed in Section 2 any generic profile \( b_{-i} = (b_j)_{j \neq i} \) in the GSP auction partitions the
space of \( i \)'s bids, \( \mathbb{R}_+ \), into \( S + 1 \) intervals: \( [0, b^{S_i}_1), [b^{S}_1, b^{S-1}_1), \ldots, [b^{1}_1, \infty) \). Letting \( b^{0}_i = \infty \) and \( b^{S_i+1}_i \equiv 0 \), if bidder \( i \) bids \( b_i \in (b^{S}_i, b^{S-1}_i) \), then he obtains slot \( t = 1, \ldots, S+1 \) at per-
click-price \( b^t \). If \( b_i \) is placed at one extreme of such intervals, the allocation is determined by
the tie-breaking rule embedded in the function \( \rho \). The function \( \pi_i \) introduced in section 2
can be seen as a correspondence \( \pi_i : \mathbb{R}_+^{n-1} \Rightarrow \{1, \ldots, S + 1\} \) such that for each \( b_{-i} \in \mathbb{R}_+^{n-1} \),
\[ \pi_i(b_{-i}) = \arg\max_{t=1,...,s+1} (v_i - b_{-i}^t) x^t \]

To allow for the possibility of ties in the bids profiles, it is necessary to generalize some of these concepts. In particular, if some of \(i\)'s opponents place equal bids (i.e., \(b_{-i} = (b_{j}')_{j \neq i}\) is such that \(b_j = b_k\) for some \(j \neq k\)), then, depending on the tie-breaking rule embedded in \(\rho\), some of the \(S + 1\) positions may be precluded to player \(i\) (e.g., if \(i = 1\), and \(b_2 = b_3\), if the tie-breaking rule is specified as in footnote 7, position \(s = 2\) is precluded to player \(i\)). In that case, the argmax in the definition of \(\pi_i\) should be taken over the set of positions that are actually accessible to \(i\). Formally: for any \(b_{-i} \in \mathbb{R}_{n+1}^n\), let

\[ S(b_{-i}) = \{ s = 1,..., S + 1 : \exists b_i \text{ s.t. } \rho(i; b_i, b_{-i}) = s \} . \]

Then, we redefine the function \(\pi_i : \mathbb{R}_{n+1}^n \to \{1,...,S + 1\}\) as follows: for every \(b_{-i} \in \mathbb{R}_{n+1}^n\)

\[ \pi_i(b_{-i}) \in \arg\max_{s \in S(b_{-i})} \left( v_i - b_{-i}^s \right) x^t. \]

Since \(S(b_{-i})\) is always non-empty and finite, the best responses \(BR_i : \mathbb{R}_{n+1}^n \Rightarrow \mathbb{R}_+\) defined in Section 2 is well-defined, and so is \(BR_i^* : \mathbb{R}_{n+1}^n \Rightarrow \mathbb{R}_+\) in (2). With these changes to the definition of \(\pi_i\), the rest of the analysis also extends to the case of ties in bids.

### A.2 Proofs of the Main Results

All the results are proven for the case in which \(n = S + 1\). The extension to the general case is straightforward but requires more cumbersome notation.

#### A.2.1 Proof of Lemma 1

Let \(\hat{b} \in E^s(v)\). By definition, for any \(i\), \(\rho(i) = s\) implies \(\pi_i(\hat{b}_{-i}) = s\) if \(s \leq S\) and \(\pi_i(\hat{b}_{-i}) = S + 1\) if \(s > S\). Hence, \(\pi_i(\hat{b}_{-i}) = \hat{b}^{s+1}\) whenever \(s \leq S\). Now, for any \(i\) such that \(\rho(i) \leq S\) and \(j\) s.t. \(\rho(j) = \rho(i) + 1\), the following must hold:

by the optimality of \(\hat{b}_i\):

\[ (v_i - \hat{b}^{\rho(i)+1}) x^{\rho(i)} \geq (v_i - \hat{b}^{\rho(i)+2}) x^{\rho(i)+1}; \quad (11) \]

by the condition in (2) for \(j\):

\[ (v_j - \hat{b}^{\rho(i)+2}) x^{\rho(i)+1} = (v_j - \hat{b}^{\rho(i)+1}) x^{\rho(i)}. \quad (12) \]

Rearranging, we obtain

\[ v_i \cdot (x^{\rho(i)} - x^{\rho(i)+1}) \geq \hat{b}^{\rho(i)+1} x^{\rho(i)} - \hat{b}^{\rho(i)+2} x^{\rho(i)+1} = v_j \cdot (x^{\rho(i)} - x^{\rho(i)+1}), \]

which implies that \(v_i > v_j\) (since, by assumption, \(x^s > x^{s+1}\) for all \(s \leq S\) and \(v_i \neq v_j\) for all \(i \neq j\)). Hence, in equilibrium, the top \(S\) bidders are ranked efficiently among themselves.

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28 This correspondence is always non-empty valued, and multi-valued only if \(i\) is indifferent between two positions. We can ignore this case here (for instance, assuming that such ties are always broken in favor of the lower position) and treat \(\pi_i : \mathbb{R}_{n+1}^n \to \Pi\) as a function (if not, \(\pi_i\) should be thought of as a selection from the correspondence above).
For the others, for any \(i\) such that \(\rho(i) > S\), eq. (2) requires that 0 = \((v_i - \hat{b}_i) x^S\), hence \(v_i = \hat{b}_i\) whenever \(\rho(i) > S\). It follows that \(\hat{b'} = \hat{b}_i\) for all \(i\) (agents bids are efficiently ranked) and \(\hat{b}_i = v_i\) for all \(i \geq S + 1\). The equilibrium bid, \(b_i = v_i - \frac{x^t}{x^{t+1}} (v_i - b_{i+1})\), then follows immediately, applying eq. (2) for all \(i = 2, ..., S\) with initial condition \(b_{S+1} = v_{S+1}\). The only restriction this entails on \(\hat{b}_1\) is that \(\hat{b}_1 > \hat{b}_2\). Finally, note that the equilibrium bid coincides with EOS' lowest envy free equilibrium (EOS, Theorem 2), and with Varian’s lower-bound symmetric Nash Equilibrium (Varian (2007), eq.9).

A.2.2 Proof of Theorem 1

We prove the statement by induction on the size of the coalition. The induction basis is the non-collusive benchmark (i.e., \(|C| = 1\)). In this case all players use their dominant strategies, \(b_i = v_i\) for each \(i\), which clearly ensures \(v_i \in (b_{i+1}, v_{i-1})\) for all \(i\), and the equilibrium bids profile is as claimed in the Theorem.

For the inductive step, suppose we have shown that the result holds for all coalitions \(C'\) such that \(C' \subseteq C\). We want to show that it also holds for \(C\). Let \(i\) be the lowest bidder in the coalition, and let \(r\) denote his position. Then, his payoff is equal to:

\[
u_i = v_i x^r - \sum_{t=r+1}^{S+1} b_t (x^{t-1} - x^t).
\]

It is useful to introduce notation to rank independent among themselves, based on their valuation. Let \(v_{I \setminus C} = (v_j)_{j \in I \setminus C}\), and let \(v_{I \setminus C} (k) = v_{I \setminus C}^{1|C|+1-k}\) denote the valuation of the \(k\)-th lowest value independent: for \(k = 1\), \(v_{I \setminus C} (1) = v_{I \setminus C}^{1|C|}\) is the lowest valuation among the independents, \(v_{I \setminus C} (2) = v_{I \setminus C}^{1|C|-1}\) is the second lowest valuation among the independents, and so on. Now, if \(i\) is the lowest-bidding member of the coalition, all players placing lower bids are independents, and therefore bid according to their dominant strategy, \(b_j = v_j\). This in turn implies that bids in positions \(t = r + 1, ..., S + 1\) are ranked efficiently between themselves, but it does not guarantee that \(b^t = v_t\) for each \(t \geq r + 1\), unless all \(j \in C\) are such that \(j \leq r\). Thus, we conclude that bids \(b^t\) for \(t = r + 1, ..., S + 1\) are placed by the \(S + 1 - r\) lowest-valued independents. Hence,

\[
u_i = v_i x^r - \sum_{t=r+1}^{S+1} v_{I \setminus C} (S + 2 - t) (x^{t-1} - x^t).
\]

Let us consider the function \(\tilde{u}_i (k)\) of \(i\)'s payoff, as a function of the position \(k\) he occupies, given that he is the lowest-bidding member of the coalition. Let \(u^*_i := \max_k \tilde{u}_i (k)\). Clearly, \(u^*_i \geq u_i\). We show next that, if \(i \neq \max \{j : j \in C\}\), then \(u^*_i < u^*_{C \setminus \{i\}}\) (the payoff \(i\) would obtain by leaving the coalition). Hence, the coalition is stable only if the lowest bidding member is also the member with the lowest valuation.

First we show that \(\tilde{u}_i\) is maximized only if \(i\) is placed efficiently with respect to the
among themselves, for any \( j \in I \setminus C, j < i \) if and only if \( \rho(j) < r \). We proceed by contradiction: suppose that there exist \( j \in I \setminus C \) such that either \( j < i \) and \( \rho(j) > r \), or \( j > i \) and \( \rho(j) < r \). Consider the first case: Since independents are ranked efficiently among themselves, for any \( j, l \in I \setminus C, l < j \) if and only if \( \rho(l) < \rho(j) \). It follows that if there exists \( j \in I \setminus C : j < i \) and \( \rho(j) > r \), such \( j \) can be chosen so that \( j = r + 1 \), i.e. \( j \) occupies the position immediately following \( i \)'s. We next show that, in this case, \( i \)'s payoff would increase if he dropped one position down. To see this, notice that

\[
\hat{u}_i (r + 1) - \hat{u}_i (r) = v_i \left( x^{r+1} - x^r \right) + v_{I \setminus C} (S + 1 - r) \left( x^r - x^{r+1} \right)
= (v_{I \setminus C} (S + 1 - r) - v_i) \left( x^r - x^{r+1} \right),
\]

where \( v_{I \setminus C} (S + 1 - r) = v_{r+1} \) is the valuation of the highest independent if \( i \) occupies position \( r \). Since, by assumption, \( x^r > x^{r+1} \), it follows that

\[
\text{sign} \left( \hat{u}_i (r + 1) - \hat{u}_i (r) \right) = \text{sign} \left( v_{I \setminus C} (S + 1 - r) - v_i \right).
\]

Under the absurd hypothesis, \( v_{I \setminus C} (S + 1 - r) > v_i \), hence \( u_i \) increases dropping one position down. A similar argument shows that in the second case of the absurd hypothesis, i.e. if there exists \( j \in I \setminus C : j > i \) and \( \rho(j) < \rho(i) \), \( u_i \) could be increased climbing one position up, from \( r \) to \( (r - 1) \). The result obtains considering the difference

\[
u_i (r) - u_i (r - 1) = (b^{r-1} - v_i) \left( x^{r-1} - x^r \right) \leq (v_{I \setminus C} (S + 2 - r) - v_i) \left( x^{r-1} - x^r \right), \tag{14}\]

which holds because all bids for positions from \( \rho(j) \) down are no higher than \( b_j = v_{I \setminus C} (S + 2 - r) \). The final expression is negative under the absurd hypothesis.

We have thus proved that, in equilibrium, for all \( j \in I \setminus C, j < i \) if and only if \( \rho(j) < r \). Hence, the lowest coalition bidder is placed efficiently with respect to the independents, and only independents are below him. Letting \( J = \{ j \in C : j > i \} \) denote the set of coalition members with values lower than \( v_i \), the lowest coalition bidder \( i \) therefore occupies position \( i + |J| \). (Clearly, \( i \) occupies the \( i \)-th position if and only if \( J = \emptyset \), i.e. if \( i \), the lowest bidding member of the coalition, also has the lowest value in the coalition.) But then, setting \( r = i + J \) in eq. \((13)\), we have that

\[
u_i^* = v_i x^{i + |J|} - \sum_{t = i + |J| + 1}^{S + 1} v_{I \setminus C} (S + 2 - t) \left( x^{t-1} - x^t \right). \tag{15}\]

We show next that \( J \neq \emptyset \) implies \( u_i^* < u_i^{C \setminus \{i\}} \). For any \( k \), let \( \tilde{b}_k \) denote \( k \)'s bid in the equilibrium with coalition \( C \setminus \{i\} \). Since, under the inductive hypothesis, the equilibrium
with coalition $C \setminus \{i\}$ is efficient, $\bar{b}_k = \bar{b}^k$ for any $k$, and hence

$$u_i^{C \setminus \{i\}} = v_i x^i - \sum_{k=i+1}^{S+1} \bar{b}_k \left( x^{k-1} - x^k \right).$$

By the inductive hypothesis, the equilibrium with this smaller coalition is as in the Theorem’s statement. Hence, $\bar{b}_k < v_{k-1}$ for all $k \in I$ (if $k$ is an independent, because he bids $\bar{b}_k = v_k < v_{k-1}$; if he’s the highest-value member of the coalition, because $\bar{b}_k \in (b_{k+1}^1, v_{k-1})$, otherwise $\bar{b}_k = b_{k+1}^* < v_{k-1}$). We also show that $\bar{b}_k \leq v_{I \setminus C}(S+2-k)$ for all $k$. To this end, observe that all $k < \max \{J\}$ are independents (both before and after $i$ drops out), so that for all $k \geq \max \{J\}$, $\bar{b}_k = v_k = v_{I \setminus C}(S+2-k)$: these are the lowest bidding and lowest-value bidders, hence also the lowest independents. For $k < \max \{J\}$, at least one of the $S + 2 - k$ elements of the set $\{k, k+1, \ldots, S+1\}$ is a member of the coalition. It follows that the valuation of the $(S + 2 - k)$-th lowest independent is higher than $v_k$, hence $v_{I \setminus C}(S+2-k) \geq v_{k-1}$, which in turn implies $v_{I \setminus C}(S+2-k) > \bar{b}_k$. Overall, we have that $\bar{b}_k < v_{k-1}$ and $\bar{b}_k \leq v_{I \setminus C}(S+2-k)$ for all $k \in I$. Using the first inequality for $k \leq i + |J|$ and the second inequality otherwise, we see that if $J \neq \emptyset$,

$$
\begin{align*}
    u_i^{C \setminus \{i\}} &= v_i x^i - \sum_{k=i+1}^{i+|J|} \bar{b}_k \left( x^{k-1} - x^k \right) - \sum_{k=i+|J|+1}^{S+1} \bar{b}_k \left( x^{k-1} - x^k \right) \\
    &> v_i x^i - \sum_{k=i+1}^{i+|J|} v_{k-1} \left( x^{k-1} - x^k \right) - \sum_{k=i+|J|+1}^{S+1} v_{I \setminus C}(S+2-k) \left( x^{k-1} - x^k \right) \quad (16)
\end{align*}
$$

Combining (15) and (16), we get

$$
\begin{align*}
    u_i^{C \setminus \{i\}} - u_i^* &> v_i \left( x^i - x^{i+|J|} \right) - \sum_{k=i+1}^{i+|J|} v_{k-1} \left( x^{k-1} - x \right) \\
    &\geq v_i \left( x^i - x^{i+|J|} \right) - v_i \left( x^i - x^{i+|J|} \right) = 0,
\end{align*}
$$

where the latter inequality follows because $v_{k-1} \leq v_i$ for all $k \geq i + 1$. Hence, whenever $J \neq \emptyset$, we obtain $u_i < u_i^{C \setminus \{i\}}$: that is, the recursive stability condition (S.2) is violated for bidder $i$. $J = \emptyset$ therefore is a necessary condition for equilibrium. Hence, in any equilibrium, the lowest coalition bidder also has the lowest valuation in the coalition. Moreover, if $J = \emptyset$, $u_i^* = u_i^{C \setminus \{i\}}$ (by equations (15) and (16)), hence in equilibrium $u_i = u_i^*$ and $i = \rho(i)$:

$$u_i = v_i x^i - \sum_{k=i+1}^{S+1} v_k \left( x^{k-1} - x^k \right) = u_i^{C \setminus \{i\}}. \quad (17)$$

Furthermore, since the payment of coalition members above $i$ is strictly decreasing in $b_i$ and positions are independent of $b_i$ (as long as $b_i \in (b_{i+1}, b_{i-1})$), the coalition will set $b_i$ as low as possible to ensure $i$’s efficient position. That is, $b_i = b_{i+1}^* = v_{i+1}^*$. 

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We have determined the positions and bids of all bidders $k \geq i$. We know that the remaining coalition members are positioned above these bidders and do not affect $u_i$. Thus, the remaining task for the coalition is to choose bids $(b_j)_{j \in C \setminus \{i\}}$ in order to maximize $\sum_{j \in C \setminus \{i\}} u_j$, subject to the constraint that $b_j > b_i$ for all $j \in C \setminus \{i\}$. We now need to look separately at two cases: $|C| = 2$ and $|C| > 2$.

First, if $|C| = 2$, the task is simply to maximize the payoff of the other member of the coalition, $j$, by determining his position relative to the remaining independents. But this, by the usual argument, is achieved when $j$ is placed efficiently with respect to these independents. This is achieved if and only if $b_j \in (b_j + 1, v_j - 1)$.

Second, if $|C| > 2$, note that even when one of the members $j \in C \setminus \{i\}$ drops out, $i$ still remains a non-top member of the coalition. Hence, its bid does not change. Naturally, the bids of all $k > i$ (who are independents) do not change either. Hence, the payoffs of all bidders $k < i$ both before and after one of the coalition members (other than $i$) drops out are shifted by the same constant relative to a game in which the bidders $k \geq i$ (and the corresponding slots) are removed: thus, the presence of these bidders has no effect on either the payoffs or the outside options. It follows that the problem we are solving at this stage is exactly equivalent to finding the equilibrium in the VCG game played between coalition $C \setminus \{i\}$ and independents $\{j \in I \setminus C : j > i\}$ with slots $x_1, \ldots, x_{i-1}$. This game has coalition size $C - 1$, so the solution follows by the inductive hypothesis.

A.2.3 Proof of Theorem 2

Since the UC-restrictions imply the stability restriction (S.1), the agency’s problem in the GSP auction with the feigned values restriction reduces to:

$$\max_{b_C} u_C (b_C, \beta^* (b_C))$$

subject to: (R) $\exists v_C' \in \mathbb{R}_+^{[C]}$ s.t. $(b_C, \beta^* (b_C)) \in E^* (v_C', v_C)$

: (S.2) $\forall i \in C, u_i (b_C, \beta^* (b_C)) \geq \bar{u}_i^C \setminus \{i\}$.

where the equilibrium conjectures $\beta^*$ are such that,

$$\forall b_C, \beta^* (b_C) \in \left\{ b_{-C}^* \in \mathbb{R}_+^{n-|C|} : \forall i \in I \setminus C, b_i^* \in BR_i^* (b_C, b_{-i,C}^*) \right\}.$$ 

Let $\sim$ be an equivalence relation on $\mathbb{R}_+^n$ such that $v \sim v'$ (resp., $b \sim b'$) if and only if $v$ and $v'$ only differ in the highest valuation (resp., highest bid), but not in the identity of the highest valuation individual (bidder).\(^{29}\) For any $v \in \mathbb{R}_+^n$, let $[v]$ (resp., $[v]$) denote the equivalence class of $v$ (resp., $b$) under this equivalence relation, and let $\mathbb{V}^\sim$ (resp., $\mathbb{B}^\sim$) denote the set of such equivalence classes. Next, consider the competitive equilibrium

\(^{29}\) Formally: $v \sim v'$ if and only if the following two conditions hold: (1) $\arg \max_{i \in I} v_i = \arg \max_{i \in I} v'_i$; (2) $v_i = v'_i$ for all $i \neq \arg \max_{i \in I} v_i$. 

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exists one equivalence class of UC-RAE of the GSP, \((\beta, b^*)\). However, because the VCG is efficient (Theorem 1), so is the ordering of bidders in the the UC-RAE of the GSP.

\[ u_i^V(v) = u_i^Q(E^*(v)) \]

Since \(E^*\) is a well-defined function on the equivalence classes of \(~\), the profile of valuation \(v'_C\) in the restriction (R) uniquely pins down \((b_C, b^*_C) \in E^*(v'_C, v-C)\) up to the highest overall bid. That is, \((b_C, b^*_C), (b'_C, b'_C) \in E^*(v'_C, v-C)\) if and only if \((b_C, b^*_C) \sim (b'_C, b'_C)\). Together with (18), this implies that \(u_i^Q(b_C, b^*_C) = u_i^Q(v'_C, v-C)\), so that also \(u_i^Q(b_C, b^*_C) = u_i^Q(v'_C, v-C)\). As a result, we can now easily recast the coalition’s problem as one of choosing \(v'_C\) (the coalition’s ‘feigned valuations’):

\[
\max_{v'_C} u_i^V(v'_C, v-C) \\
\text{subject to : (S.2) } \forall i \in C, u_i^V(v'_C, v-C) \geq \bar{u}_i^C \setminus \{i\}.
\]

(Notice that the restriction (R) and the restriction that \(\beta^*(b_C)\) always be in the set \(BR_{-C}\) are both built in this formulation of the problem.) In the following, we let \(\bar{u}_i^C\) denote bidder \(i\)’s payoff when the coalition is \(C\) in the GSP game being studied, while \(\bar{u}_i^{C,V}\) denotes the same object in the corresponding VCG game. With this in mind, note that \(\bar{u}_i^C = \bar{u}_i^{C,V}\) for all \(i\) when \(|C| = 1\), and the recursion defining \(\bar{u}_i^C\) is identical to that defining \(\bar{u}_i^{C,V}\). It follows that the coalition’s problem is now equivalent to its problem in the VCG game. By Theorem 1, the solution \(v'^*_C\) is unique up to the report of the highest coalition member, \(v'^*_{\min(C)}\).

Finally, by (R), the UC-RAE of the GSP satisfies \((b'^*_C, \beta^*(b^*_C)) \in E^*(v'^*_C, v-C)\). Hence all bidders’ positions and payoffs in this GSP equilibrium are the same as in the unrestricted RAE of the VCG, \((v'^*_C, v-C)\). Because the ordering of bidders in the RAE of the VCG is efficient (Theorem 1), so is the ordering of bidders in the UC-RAE of the GSP. However, because \(v'^*\) is unique only up to the highest coalition bid, \((b'^*_C, \beta^*(b^*_C))\) is not uniquely defined: there exists a continuum of equilibria differing in the payments of all bidders above the highest coalition bidder: for each \(v'^*_{\min(C)} \in (v'^*_{\min(C)+1}, v'^*_{\min(C)-1})\), there exists one equivalence class of UC-RAE of the GSP, \([b'^*_C, \beta^*(b^*_C)]\). Because \(E^*\) is unique only up to the highest overall bid, there also exist a continuum of equilibria yielding the same payoffs and positions, but differing in the highest overall bid, within each \([b^*_C]\). In this sense, the equilibrium is unique up to the highest coalition and overall bids.
A.2.4 Proof of Theorem \[3\]

The claim about the possibility of strict ordering in revenues is proven by Example \[6\] in the text. Here we prove the general claims about existence, uniqueness and weak ordering. The proof is by construction, and it is based on the following intermediate result.

**Lemma 2** Fix \(C \subseteq I\), and let \(K\) be a finite index set. Let \(\{b^{(k)}\}_{k \in K}\) be a collection of bid profiles such that, for each \(k \in K\), \(b^{(k)}_C \in BR^*_C(b^{(k)}_C)\) and \(\rho(i; b^{(k)}) = i\) for each \(i \in I\). Define \(L(\{b^{(k)}\}_{k \in K}) \equiv \hat{b} \in \mathbb{R}^n_+\) as follows:

\[
\hat{b}_i = \begin{cases} 
    \hat{b}_i = \min_{k \in K} b^{(k)}_i & \text{if } i \in C \\
    \hat{b}_i = v_{S+1} & \text{if } i = S + 1 \notin C \\
    \frac{1}{x^{i-1}} \left[ \sum_{j = i}^{\hat{c}(i) - 1} v_j (x_j^{i-1} - x_i^j) + \hat{b}_{\hat{c}(i)} x^{\hat{c}(i)-1} \right] & \text{otherwise}
\end{cases}
\]

where \(\hat{c}(i) := \min \{ j < i | \rho(i; \hat{b}) = j \}\) if \(i \in \max C\) and \(\hat{c}(i) = S + 1\) otherwise.

Then: (i) \(\rho(i; \hat{b}) = i\) \(\forall i \in I\); (ii) \(u_i(\hat{b}) \geq u_i(b^{(k)})\) for all \(i \in I\) and for all \(k \in K\), with strict inequality whenever \(\hat{b}_{\hat{c}(i)} \neq b^{(k)}_{\hat{c}(i)}\); (iii) \(u_C(\hat{b}) \geq u_C(b^{(k)})\) for all \(k \in K\), with strict inequality whenever \(\exists i \in C \setminus \min C\) such that \(\hat{b}^{(k)}_i \neq \hat{b}_i\); (iv) \(\hat{b}_C \in BR^*_C(\hat{b}_C)\).

**Proof of Lemma 2**

We begin by noting that because for each \(k \in K\), \(b^{(k)}_C \in BR^*_C(b^{(k)}_C)\) and \(\rho(i; b^{(k)}) = i\) for each \(i \in I\), we have that \(\forall k \in K, i \notin C\) s.t. \(i \neq S + 1\),

\[
b^{(k)}_i = \frac{1}{x^{i-1}} \left[ \sum_{j = i}^{\hat{c}(i) - 1} v_j (x_j^{i-1} - x_i^j) + b^{(k)}_{\hat{c}(i)} x^{\hat{c}(i)-1} \right],
\]

and \(\hat{b}^{(k)}_i = v_{S+1}\) if \(i = S + 1 \notin C\) (\(\hat{c}(i)\) is defined in the statement in the Lemma.)

The following two key observations are now immediate:

1. For every \(k \in K\) and for every \(i \in I\), \(\hat{b}_i \leq b^{(k)}_i\): For \(i \in C\), \(\hat{b}_i \leq b^{(k)}_i\) by the definition of coalition bids in the statement of the lemma. For \(i = S + 1 \notin C\), \(\hat{b}_i = v_{S+1} = b^{(k)}_i\) (the second equality is because the Lemma requires \(b^{(k)}_C \in BR^*_C(b^{(k)}_C)\)). Finally, for \(i \notin C\) s.t. \(i \neq S + 1\),

\[
\hat{b}_i = \frac{1}{x^{i-1}} \left[ \sum_{j = i}^{\hat{c}(i) - 1} v_j (x_j^{i-1} - x_i^j) + b^{(k)}_{\hat{c}(i)} x^{\hat{c}(i)-1} \right] \leq \frac{1}{x^{i-1}} \left[ \sum_{j = i}^{\hat{c}(i) - 1} v_j (x_j^{i-1} - x_i^j) + b^{(k)}_{\hat{c}(i)} x^{\hat{c}(i)-1} \right] = b^{(k)}_i,
\]

where the inequality follows because, by definition, \(\hat{c}(i) \in C \cup \{S + 1\}\) and hence \(\hat{b}^{(k)}_{\hat{c}(i)} \leq b^{(k)}_{\hat{c}(i)}\). Note that the inequality is strict whenever \(\hat{b}^{(k)}_{\hat{c}(i)} \neq b^{(k)}_{\hat{c}(i)}\).

2. For each \(i \in I\), there exists \(k \in K\) such that \(b_i = b^{(k)}_i\). For \(i \in C\) this is immediate from the definition. For \(i = S + 1 \notin C\), \(\hat{b}_i = v_{S+1} = b^{(k)}_i\) for all \(k\) (cf. previous
point). For $i \notin C$ s.t. $i \neq S + 1$, the result follows because $\hat{c}(i) \in C \cup \{S + 1\}$, hence there exists $k \in K$ such that $\hat{b}_{\ell(i)} = b_{\ell(i)}^{(k)}$, so that

$$
\hat{b}_i = \frac{1}{x^{i-1}} \left[ \sum_{j=1}^{\ell(i)-1} v_j (x^{j-1} - x^j) + \hat{b}_{\ell(i)} x^{\ell(i)-1} \right] = \frac{1}{x^{i-1}} \left[ \sum_{j=i}^{\ell(i)-1} v_j (x^{j-1} - x^j) + b_{\ell(i)}^{(k)} x^{\ell(i)-1} \right] = b_i^{(k)}.
$$

We can now establish the lemma’s results:

(i) $\rho(i; \hat{b}) = i$ for all $i \in I$: Let $i, j \in I$ be s.t. $i < j$. We show that $\hat{b}_i > \hat{b}_j$. By point 2 above, there exists $k \in K$ such that $\hat{b}_i = b_i^{(k)}$. Because, by assumption, $b^{(k)}$ is ordered efficiently, $b_i^{(k)} > b_j^{(k)}$. By point 1, $b_j^{(k)} \geq \hat{b}_j$. Hence, $\hat{b}_i = b_i^{(k)} > b_j^{(k)} \geq \hat{b}_j$, as desired.

(ii) $u_i(\hat{b}) \geq u_i(b^{(k)})$ for all $i \in I$ and all $k \in K$, with strict inequality if $\hat{b}_{\ell(i)} \neq b_{\ell(i)}^{(k)}$: Because $i$ obtains its efficient position under both $\hat{b}$ (established in (i)) and $b^{(k)}$ (given),

$$u_i(\hat{b}) = (v_i - \hat{b}_{i+1}) x^i \geq (v_i - b_{i+1}^{(k)}) x^i = u_i(b^{(k)}),$$

where the inequality holds because $\hat{b}_{i+1} \leq b_{i+1}^{(k)}$ by point 1 above, with strict inequality if $\hat{b}_{\ell(i)} \neq b_{\ell(i)}^{(k)}$, as noted at the end of point 1.

(iii) $u_C(\hat{b}) \geq u_C(b^{(k)})$ for all $k \in K$, with strict inequality whenever $\exists i \in C \setminus \min C$ such that $b_i^{(k)} \neq \hat{b}_i$: The weak inequality follows immediately from part (ii). Now, suppose $b_i^{(k)} \neq \hat{b}_i$ for some $i \in C \setminus \min C$, and let $j = \max \{k \in C | k < i\}$ be the coalition member directly above $i$ in the ranking of valuations. Then $\hat{c}(j) = i$, so that by the strict inequality part of result (ii), $u_j(b^{(k)}) < u_j(\hat{b})$. Since $u_j(b^{(k)}) \leq u_j(\hat{b})$ for all other terms in the sums defining $u_C(\cdot)$, this completes the proof for strict inequality.

(iv) $\hat{b}_{-C} \in BR^*_C(\hat{b}_{-C})$: The LREF condition holds by construction. We must simply prove the Nash condition, i.e., that each $i \notin C$ (weakly) prefers position $i$ to position $j$ for all $j \neq i$. Define $j' = j + 1$ if $j > i$ and $j' = j$ if $j < i$. Note that if bidder $i$ deviates to position $j \neq i$ under bid profile $\hat{b}$, it gets payoff $(v_i - \hat{b}_{j'}) x^j$. By the observation in point 2 above, there exists some $k$ such that $\hat{b}_{j'} = b_{j'}^{(k)}$, so that $(v_i - \hat{b}_{j'}) x^j = (v_i - b_{j'}^{(k)}) x^j$. Because $b_{j'}^{(k)} \in BR_{-C}(b_{-C}^{(k)})$ and $\rho(i; b^{(k)}) = i$, $i$ cannot profitably deviate from position $i$ to position $j \neq i$ under bid profile $b^{(k)}$, i.e. $(v_i - b_{j'}^{(k)}) x^j \leq (v_i - b_{i+1}^{(k)}) x^i$. Finally, by point 1 above, $b_{i+1}^{(k)} \geq \hat{b}_{i+1}$, so that $(v_i - b_{i+1}^{(k)}) x^i \leq (v_i - \hat{b}_{i+1}) x^i$. Putting these results together,

$$
(v_i - \hat{b}_{i+1}) x^i \geq (v_i - b_{i+1}^{(k)}) x^i \geq (v_i - b_{j'}^{(k)}) x^j = (v_i - b_{j'}) x^j.
$$

That is, bidder $i$ cannot profitably deviate to position $j \neq i$ under bid profile $\hat{b}$, as desired. This concludes the proof of the Lemma.
the LREF equilibrium by definition.

For the **inductive step**, we fix \( C \) and suppose that for all coalitions of size \( |C| - 1 \) Eff-RAE exist, then we show that Eff-RAE also exists for \( C \), and that in each of these RAE the coalition’s surplus is no lower than in any RAE of the VCG mechanism, while the auctioneer’s revenue is no higher than in a corresponding RAE of the VCG mechanism.

Fix \( C \), and let \( b^{UC} \in \mathbb{R}^n_+ \) be the bids in the UC-RAE of the GSP auction with the same coalition \( C \), in which the top coalition member is placing the highest possible bid (this exists, it is efficient and unique by Theorem 2). Observe that because of the bijection between UC-RAE of the GSP auction and unconstrained RAE of the VCG mechanism (established in Theorem 2), we can use the coalition’s surplus in the GSP auction with bids \( b^{UC} \) as our reference point. Next, note that, for any \( b_c \), the beliefs \( \beta^*(b_c) \) are uniquely determined by the Varian/EOS recursion. Hence, a complete Eff-RAE, \((b^*, \beta^*) \in \mathbb{R}^n_+ \times B^*\), if it exists, is in fact fully determined by \( b^*_C \in \mathbb{R}^n_+ \).

We now proceed to prove that such a \( b^*_C \) exists by constructing a candidate profile.

For each \( i \in C \), let \( b^{(i)} \) be the bids in an Eff-RAE with coalition \( C \setminus \{i\} \) (these exist under the inductive hypothesis). Let \( b^{(0)} = b^{UC} \). Let \( \hat{b} = \mathcal{L} \left( \{b^{(i)}\}_{i \in C \cup \{0\}} \right) \), where \( \mathcal{L} \) is as defined in Lemma 2. Now, by results (i) and (iv) of Lemma 2, we have \( \rho(i; \hat{b}) = i \) for all \( i \in I \) and \( \hat{b}_c \in BR^*_{c}(\hat{b}_C) \). It follows that \( \hat{b}_C \in R_{C}^{Eff} \). By result (ii) of Lemma 2, \( u_i(\hat{b}) \geq u_i(b^{(k)}) \) for each \( i \). Moreover, by construction, \( u_i(b^{(k)}) = \hat{u}^{C \setminus \{i\}}_i \) for each \( i \in C \), hence profile \( \hat{b} \) satisfies the recursive stability condition. It follows that \( \hat{b}_C \) is a valid bid vector for coalition \( C \) trying to achieve an Eff-RAE and that \( \hat{b}_c = \beta^*(\hat{b}_C) \), where \( \beta^* \) are the unique beliefs consistent with Eff-RAE. Maintaining the assumption of finite bid increments, as in Theorems 1 and 2, the coalition is therefore maximizing over a non-empty, finite set of valid bid vectors, so that a maximum, \( b^*_C \), exists. Thus, an efficiency constrained RAE for coalition \( C \) exists (and is equal to \( ((b^*_C, \beta^*(b^*_C)), \beta^*) \)).

Now the weak ordering of coalition surplus is immediate: Result (iii) of Lemma 2 implies \( u_C(\hat{b}) \geq u_C(b^{UC}) \), and clearly the optimal bid profile \((b^*_C, \beta^*(b^*_C)) \) must satisfy \( u_C(b^*_C, \beta^*(b^*_C)) \geq u_C(\hat{b}) \). It follows that \( u_C(b^*_C, \beta^*(b^*_C)) \geq u_C(b^{UC}) \).

Next, we establish the ordering for the auctioneer’s revenues. We first show that, in the Eff-RAE \((b^*, \beta^*)\), the bid of coalition members other than the highest-value is weakly lower than in \( \hat{b} \). To this end, suppose that there exists some \( i \in C \setminus \min C \) such that \( b^*_i > \hat{b}_i \). Let \( b' = \mathcal{L}(\{b^*, \hat{b}\}) \). By part (i) of Lemma 2, \( b'_C \) is still a valid bid vector for the coalition, whereas part (iii) implies \( u_C(b'_C, \beta^*(b'_C)) > u_C(b^*_C, \beta^*(b^*_C)) \) which contradicts the optimality of \( b^*_C \). We thus conclude that \( b^*_i \leq \hat{b}_i \) for all \( i \in C \setminus \min C \).

Because the independents’ bids are fixed by the recursion under both \( \hat{b} \) and \( b^* \), we know that in fact \( b^*_i \leq \hat{b}_i \) for all \( i > \min C \). Because by construction \( \hat{b}_i \leq b^{UC}_i \) for all \( i \in I \), we thus have \( b^*_i \leq b^{UC}_i \) for all \( i > \min C \). If \( \min C = 1 \), this completes the proof that the auctioneer’s revenues are weakly lower under \( b^* \) than under \( b^{UC} \). If \( \min C > 1 \), we need to show that even the top coalition bidder in \( b^* \) cannot bid more than this bidder’s maximum possible UC-RAE bid. Because \( b^{UC}_{\min C} \) is the maximum bid that the top coalition bidder
can place in a UC-RAE, it is equal to (cf. Theorem 2)

\[ b_{\text{min}C}^{UC} = v_{\text{min}C-1} - \frac{x_{\text{min}C}}{x_{\text{min}C-1}} (v_{\text{min}C-1} - b_{\text{min}C}^{UC}) \, . \]

If \( b_{\text{min}C}^* > b_{\text{min}C}^{UC} \), then the independent above the top coalition member obtains a payoff

\[ U_0 = (v_{\text{min}C-1} - b_{\text{min}C}^*)x_{\text{min}C-1} < (v_{\text{min}C-1} - b_{\text{min}C}^{UC})x_{\text{min}C-1} = (v_{\text{min}C-1} - b_{\text{min}C}^{UC+1})x_{\text{min}C} \, , \]

where the last inequality follows by substituting in the expression for \( b_{\text{min}C}^{UC} \) from above.

Dropping one position down this independent would obtain

\[ U' = (v_{\text{min}C-1} - b_{\text{min}C+1}^*)x_{\text{min}C} \geq (v_{\text{min}C-1} - b_{\text{min}C+1}^{UC})x_{\text{min}C} > U_0 \, , \]

where the first inequality follows because \( b_{i}^* \leq b_{i}^{UC} \) for all \( i > \text{min}C \), as established above. Thus this independent has a profitable deviation; a contradiction. We conclude that \( b_{\text{min}C}^* \leq b_{\text{min}C}^{UC} \). But then, by the independents’ recursion, we also have \( b_{i}^* \leq b_{i}^{UC} \) for all \( i \leq \text{min}C \). Because we already knew that the \( b_{i}^* \leq b_{i}^{UC} \) for all \( i > \text{min}C \), we have established that all bids in \( b^* \) are weakly lower than in \( b^{UC} \), which completes the claim about the auctioneer’s revenues.

Next, we show that the Eff-RAE is unique up to the highest coalition bid. To this end, fix some coalition \( C \subseteq I \) and let \( b^{R1} \) and \( b^{R2} \) be two (possibly equal) Eff-RAE for \( C \). Let \( \hat{b} := \mathcal{L}(\{b^{R1}, b^{R2}\}) \). By results (i), (iii) and (iv) of Lemma 2, \( \hat{b} \) is still efficiently ordered and \( \hat{b}_{\text{min}C} \in BR^{*\text{C}}(\hat{b}_C) \), so that \( \hat{b}_C \) is in the set of permitted bids for the coalition in the efficiency-constrained problem without the recursive stability restriction, with \( \hat{b}_{\text{min}C} \in \beta^{*\text{C}}(\hat{b}_C) \). Furthermore, by result (ii) of Lemma 2, each coalition member is at least as well off under \( \hat{b} \) as under \( b^{R1} \) and \( b^{R2} \). Therefore, the fact that \( b^{R1} \) and \( b^{R2} \) satisfy the recursive stability condition implies that so does \( \hat{b} \). The optimality of \( b^{C1}_C \) and \( b^{C2}_C \) in this set therefore implies that \( u_C(\hat{b}) \leq u_C(b^{Rk}_C) \forall k \in \{1, 2\} \). But result (iii) of Lemma 2 then implies that \( \hat{b}_i = b^{R1}_i = b^{R2}_i \) for all \( i \in C \setminus \text{min}C \).

Combining these results yields \( b^{R1}_i = b^{R2}_i = \hat{b}_i \) for all \( i \in C \setminus \text{min}C \). Because coalition bids also uniquely determine independents’ bids, the Eff-RAE is thus unique up to the highest coalition bid. This completes the proof.

### A.3 Multiple Agencies

#### A.3.1 Formal definition

We consider the case with two SEMAs, which coordinate the bids of subsets \( C_1, C_2 \subseteq I \) of bidders, s.t. \( C_1 \cap C_2 = \emptyset \). Similar to the baseline notion with a single SEMA, the definition of RAE with multiple agencies is recursive, with the outside option of coalition member \( i \in C_1 \) being defined as his equilibrium payoff in the game with coalitions \( (C_1 \setminus \{i\}, C_2) \).

Hence, the recursion in the RAE with multiple coalitions involves, for every \( C \), a recursion
similar to the one for the single SEMA, but with initial condition set by the RAE in which 
\( C_{-g} \) is the only coalition.

Let \( G(v) = (A_i, u_i)_{i=1,...,n} \) denote the baseline game (e.g., GSP or the VCG), given the 
profile of valuations \( v = (v_i)_{i \in I} \). For any \( C_1, C_2 \subseteq I \) with \( |C_g| \geq 2 \) and \( C_1 \cap C_2 = \emptyset \), we let 
\( C := C_1 \cup C_2 \). For each \( g = 1,2 \), coalition \( C_g \) chooses a vector of bids \( b_{C_g} = (b_j)_{j \in C_g} \in \times_{j \in C_g} A_j \), and let \( b_C = (b_{C_1}, b_{C_2}) \). Given \( b_C \), independents \( i \in I \setminus C \) simultaneously choose bids \( b_i \in A_i \). We let \( b_{-C} := (b_j)_{j \in I \setminus C} \) and \( A_{-C} := \times_{j \in I \setminus C} A_j \). Given profiles \( b \) or \( b_{-C} \), we let \( b_{-i,C} := (b_j)_{j \in I \setminus C; j \neq i} \). As above, each SEMA maximizes the sum of the payoffs of 
its members, \( u_{C_g}(b) := \sum_{i \in C_g} u_i(b) \), under the three constraints from the single-agency 
model, given conjectures about both the independents and the other coalition.

**Stability-1: (Stability w.r.t. Independents)** For any \( i \in I \setminus C \), let \( B^*_i : A_{-i} \rightrightarrows A_i \), 
\( BR^*_C : A_C \rightrightarrows A_{-C} \) and \( S_C \) be defined as in the single-agency case (except now \( C = C_1 \cup C_2 \)). For each agency \( C_g \), we let 
\[
S_{C_g} = \{ b_{C_g} \in A_{C_g} : \exists b_{C_{-g}} \in A_{C_{-g}} \text{ s.t. } (b_{C_g}, b_{C_{-g}}) \in S_C \},
\]

**Stability-2: ((Recursive) Stability w.r.t. Coalition Members)** Let \( B^* \) be defined as in the single-agency case. Letting \( E^R(C_1, C_2) \) denote the set of \textit{Recursively Stable} 
Agency Equilibrium \( \text{(RAE)} \) outcomes of the game with coalitions \( C_1 \) and \( C_2 \), given 
restrictions \( \mathcal{R} \) (and refinement \( BR^*_1 \)), we initialize the recursion setting \( E^R(C_g', C_{-g}) = 
E^R(C_{-g}) \) if \( |C_g'| = 1 \) (that is, if an agency controls only one bidder, then the RAE are 
the same as when there exists only the other agency). Suppose next that \( E^R(C_g', C_{-g}) \) 
has been defined for all subcoalitions \( C_g' \subseteq C_g \). For each \( i \in C_g \), and \( C_g' \subseteq C_g \setminus \{i\} \), 
let \( u_{i}^{C_g', C_{-g}} = \min_{b \in E^R(C_g', C_{-g})} u_i(b) \). The second stability requirement therefore requires 
\( u_i \geq u_{i}^{C_g', C_{-g}} \). Finally, we define the set of ‘Rational Conjectures’ about the Opponent 
Coalition as \( B^*_g = \{ \beta_g \in (A_{C_{-g}})^{\mathcal{F}}_{C_g} : \beta_g(b_{C_g}) \in BR^C_{-g}(b_{C_g}) \text{ for all } b_{C_g} \in S_{C_g} \} \), where 
\( \mathcal{F}_{C_g} = \{ b_{C_g} \in S_{C_g} : BR^C_{-g}(b_{C_g}) \neq \emptyset \} \), and 
\[
BR^C_{-g}(b_{C_g}) = \arg \max_{b_{C_{-g}}} u_{C_{-g}}(b_{C_g}, b_{C_{-g}}, \beta(b_{C_g}, b_{C_{-g}})) 
\]
subject to: (R) \( (b_{C_g}, b_{C_{-g}}) \in R_C \) 
: (S.1) \( (b_{C_g}, b_{C_{-g}}) \in S_C \) 
: (S.2) for all \( i \in C_{-g} \), \( u_i(b_{C_g}, b_{C_{-g}}, \beta(b_{C_g}, b_{C_{-g}})) \geq u_{i}^{C_{-g}\setminus\{i\}, C_g} \)

**Definition 4** A Recursively Stable Agency Equilibrium \( \text{(RAE)} \) of the game \( G \) with 
coalition structure \( (C_1, C_2) \), given restrictions \( \mathcal{R} \) and independents’ equilibrium refinement 
\( BR^*_1 \), is a profile of bids and conjectures \( (b^*, \beta^*, \beta^*_1, \beta^*_2) \in A_C \times B^* \times B^*_1 \times B^*_2 \) such that:

1. The independents play a mutual best response: for all \( i \in I \setminus C \), \( b^*_i \in BR^*_i(b^*_{-i}) \).
2. The conjectures of the agencies are correct and consistent with the exogenous restric-
tions: \( \beta^* (b^*_C) = b^*_C \), and, for each \( g \in \{1, 2\} \), \( \beta^*_g \left( b^*_C \right) = b^*_C - g \), and \( (b^*_C, \beta^*_g(b^*_C), \beta^* (b^*_C, \beta^*_g(b^*_C))) \in R(C) \) for all \( b^*_C \in R_C \).

3. Each agency best responds to the conjectures \( \beta^* \) and \( \beta^*_g \), given the exogenous restrictions \( (R) \) and the stability restrictions about the independents and the coalition members \( (S.1 \text{ and } S.2, \text{ respectively}) \): For each \( g = 1, 2 \)

\[
b^*_C \in \arg \max_{b^*_C} u_{C_g} \left( b^*_C, \beta^*_g(b^*_C), \beta^* (b^*_C, \beta^*_g(b^*_C)) \right)
\]

subject to: \( (R) \) \((b^*_C, \beta^*_g(b^*_C), \beta^* (b^*_C, \beta^*_g(b^*_C))) \in R_C \)

\( (S.1) \) \((b^*_C, \beta^*_g(b^*_C), \beta^* (b^*_C, \beta^*_g(b^*_C))) \in S_C \)

\( (S.2) \) for all \( i \in C_g \), \( u_i \left( b^*_C, \beta^*_g(b^*_C), \beta^* (b^*_C, \beta^*_g(b^*_C)) \right) \geq u^C_i \left( C_1, C_2 \right) \)

The set of RAE outcomes for the game with coalitions \( (C_1, C_2) \) (given \( BR^* \) and \( R_C \)) is:

\[
E^R (C_1, C_2) = \{ b^* \in A : \exists \beta^*, \beta^*_1, \beta^*_2 \text{ s.t. } (b^*, \beta^*, \beta^*_1, \beta^*_2) \text{ is a RAE} \}.
\]

Note that the definition above does not uniquely pin down the bid of the top bidder of the “lower” coalition. To remove this ambiguity, in the following we break these ties by making this coalition member bid as if it were an independent, whenever such bids are still in the optimal set.

A.3.2 Proof of Theorem 4

We prove the theorem by providing a precise characterization of the RAE in the VCG and the UC-RAE of the GSP. That is, we show that with two coalitions, \( C_1 \) and \( C_2 \), the following statements hold:

1. If in the overall value ranking no member of one coalition is adjacent to a member of the other coalition, then:

   (a) There exists a unique RAE of the VCG mechanism. In this equilibrium, the bid profile \( \hat{b}^V \) is such that

   \[
   \hat{b}^V_i = \begin{cases} 
   v_i & \text{if } i \in (I \setminus C) \cup \min C_1 \cup \min C_2; \\
   (\hat{b}^V_{i+1}^+) & \text{if } i \in C \setminus \{ \min C_1 \cup \min C_2 \} \text{ and } i \leq S;
   \end{cases}
   \]

   where \( v_0 := \infty \) and \( \hat{b}^V_{n+1} := 0 \).

   (b) There exists a unique UC-constrained RAE of the GSP auction. In this equilibrium, for every \( i \),

   \[
   \hat{b}^G_i = v_i^f - \frac{x_i^f}{x_i^f - x_{i+1}^f} \left( v_{i+1}^f - \hat{b}_{i+1} \right),
   \]
where \( v_i^f \) is equal to bidder \( i \)'s bid (reported value) in the VCG mechanism (as described in Statement 1 above): \( v_i^f = \hat{b}_i^V \).

2. If in the overall value ranking a non-top member of one coalition is directly above a non-top member of the other coalition (i.e., there exist \( i \) and \( i + 1 \), such that \( i \in C_j \), \( i + 1 \in C_j', j \neq j' \), \( i \neq \min C_j \), and \( i + 1 \neq \min C_j' \)), then no unconstrained RAE of the VCG and no UC-RAE of the GSP exist.

Below, we prove the results for the VCG (statement 1(a) and the VCG part of statement 2 above). The proofs of the GSP results are analogous.

First we show that, regardless of whether there are or are not adjacencies in the value rankings, an arrangement like that in statement 1(a) is the only possible RAE of the VCG. We then show that this candidate is in fact an equilibrium when there are no adjacencies, but not when there are adjacencies involving non-top bidders.\(^{30}\)

Before proceeding to the proof, it pays to make two observations about the best-response correspondences \( BR_g \):

**Observation 1:** The best-response function of any coalition requires that each non-top member of the coalition bid just above the bid below. Formally, let \( i \in C_g \setminus \min C_g \) and let \( b \) be such that \( b_{C_g} \in BR_g^C(b_{C_g}) \). Then \( b_i = (b^{\rho(i)} + 1)^+ \).

*Proof of Observation 1:* Suppose \( b_i \neq (b^{\rho(i)} + 1)^+ \), and let \( \delta = b_i - b^{\rho(i)} \). Now note that in the definition of \( BR_g \), coalition \( g \) takes the bids of the other coalition (and the independents) as fixed. Thus, lowering \( b_i \) to \( b^{\rho(i)} + \delta / 2 < b_i \) does not change the allocation, but reduces the payments of all higher-ranked members of \( C_g \) by \( \delta/2)(x^{\rho(i)} - x^{\rho(i-1)} > 0 \), and is therefore a profitable deviation for \( C_g \), a contradiction. \( \square \)

**Observation 2:** The best-response function of any coalition requires that no member of the coalition (top or non-top) be placed above a bidder bidding higher than this member’s value. Formally, if \( i \in C_g \) and \( b_{C_g} \in BR_g^C(b_{C_g}) \), then \( v_i \geq b^{\rho(i)} + 1 \).

*Proof of Observation 2:* Suppose \( v_i < b^{\rho(i)} + 1 \), and consider the deviation where \( C_g \) bids \( (b^{\rho(i)})^- \). Note that this deviation improves \( i \)'s individual payoff by \( (b^{\rho(i)} + 1 - v_i)(x^{\rho(i)} - x^{\rho(i)} + 1) > 0 \). Also observe that the deviation decreases the payments of higher-ranked coalition members (if any) by \( (b_i - b^{\rho(i)} + 1)(x^{\rho(i)} + 1 - x^{\rho(i)}) > 0 \). Thus, the deviation is unambiguously profitable for the coalition. \( \square \)

With these observations in hand, we proceed to the proof of Theorem 4.

As in Theorem 1, the proof is by recursion on the overall size of the coalition, \(|C| = |C_1| + |C_2| \). The induction basis is the case of no coalitions (\(|C| = 2 \), i.e., \(|C_1| = |C_2| = 1 \),

\(^{30}\)Compared to the single agency case, the part of the proof that parallels Theorem 1 has two complications. First, the placement of the highest bidder of the the coalition that does not have the top overall bidder requires some additional technicality, as this placement is not only relative to independents but also relative to the other coalition’s bidders. Second, the candidate equilibrium produced by the recursion still needs to be verified, because the recursive procedure does not guarantee that a coalition’s bidders are best-responding to those bidders of the rival coalition that are placed below them. It is precisely this verification step that will yield the fundamental difference between the cases with and without members from different coalitions that are adjacent in the value ranking.
for which the result holds trivially, by EOS. For the inductive step, we first look at the overall lowest placed coalition bidder, $i$. The same argument as in the proof of Theorem 1 shows that, due to the recursive stability condition, this bidder is in fact the lowest-valued bidder among all coalition bidders ($i = \max(C_1 \cup C_2)$) and that it must occupy its efficient position ($\rho(i) = i$). The rationale is the same as in Theorem 1: because there are only independents below this bidder, $j$ cannot be compensated by the rest of the coalition for taking an inefficient position (which the individually bidder prefers). Furthermore, by Observation 1 above, $b_i = v^+_{i+1}$.

Just as in the proof of Theorem 1, after fixing the lowest coalition bidder’s bid, we can essentially remove this bidder and all lower-valued independents from the analysis and proceed to the next-lowest placed coalition bidder. Unless this bidder is the top bidder of a coalition, the same argument as in the proof of Theorem 1 again applies to show the bidder is placed in its efficient position. In addition, by Observation 1, it is bidding just above the value of the bidder just below. We then move to the next-lowest-placed coalition bidder.

Now, suppose we reach the top bidder of some coalition, bidder $i$. As in Theorem 1, this bidder must simply set its bid so as to maximize its own payoff (as there are no other coalition members above, whose payoffs it would affect). As in Theorem 1, this bidder cannot be placed directly above a higher-valued independent or directly below a lower-valued independent, by the standard EOS argument (e.g., when placed directly above $j$ with $v_j > v_i$, $i$ can increase its payoff by $\Delta x(v_j - v_i) > 0$ if it drops one position down). Unlike Theorem 1, however, this does not necessarily guarantee the efficient placement of $i$, as $i$ could be placed directly below a lower-valued member of the other coalition ($i$ cannot be placed above a higher-valued member of the other coalition, because, by construction, $i$ is the lowest-placed remaining member of $C$, with all previous members placed in their efficient positions).

To rule out this remaining possibility, suppose $i$ is placed directly below the other coalition’s member $j$, with $v_i > v_j$. By Observation 1, this means that $b_i < v_j < v_i$. But consider the deviation where bidder $i$’s bid is changed to $b_i' = v^+_{j} > b_i$ (note also that $b_i' < v_i$ because $v_i > v_j$). By Observation 2, this deviation causes the other coalition to move bidder $j$ (and any other members with with values below $b_i'$) below bidder $i$, reducing their bids to no more than $b_i'$. Consequently, bidder $i$ gains at least one position, which happens at a price that is less than $v_i$. Therefore, bidder $i$’s payoff increases by (at least) $(v_i - b_i'\rho(i) - x_{\rho(i)}) > 0$. The deviation is thus profitable.

This completes the proof that the top bidder of each coalition must occupy its efficient position and will therefore bid its true value (by the assumed equilibrium selection).

We now can repeat the above arguments for all remaining coalition bidders until all of their bids are fixed. We have thus proved that the only possible equilibrium has all bidders placed efficiently, with bids as specified in the theorem statement.

We next verify that this candidate is in fact an equilibrium when no members of
different coalitions are adjacent. Note that for the top bidders of both coalitions this is equivalent to checking that they do not have any individually profitable deviations (because their bids and positions relative to bidders outside of their coalition do not affect the payoffs of the other members off their coalition), and for non-top bidders any deviation must also be weakly profitable individually, as they are already held to their outside options in the candidate equilibrium. Also, because inefficient reversals within a coalition are never profitable for the coalition, we need to consider only deviations that preserve ranking within a coalition. Now, for deviations upward consider any coalition bidder $i$ such that the bidder directly above is not a member of the same coalition. If $i$ is its coalition’s top bidder, then $b_i = v_i$ and hence $b_j > b_i = v_i$ for all bidders above $i$. Then the standard EOS argument shows that $i$ does not have a profitable deviation upwards. If $i$ is not a top bidder, then, by assumption, the bidder directly above $i$ (that is, bidder $i - 1$) is a higher-valued independent, so $b_{i-1} = v_{i-1} > v_i$, and again $b_j \geq b_{i-1} > v_i$ for all bidders above $i$. The standard EOS argument again shows that $i$ does not have a profitable deviation upwards. For deviations downward consider any coalition bidder $i$ such that the bidder directly below is not a member of the same coalition. If $i$ is its coalition’s top bidder, then $b_i = v_i$ and hence $b_j < b_i = v_i$ for all bidders below $i$. Then the standard EOS argument shows that $i$ does not have a profitable deviation downwards. If $i$ is not a top bidder, then, by assumption, the bidder directly below $i$ (that is, bidder $i + 1$) is a lower-valued independent, so $b_{i+1} = v_{i+1} < v_i$, and again $b_j \leq b_{i+1} < v_i$ for all bidders below $i$. The standard EOS argument again shows that $i$ does not have a profitable deviation downwards. This completes the proof of the theorem.

Finally, we show that there is no equilibrium if there are any cases where non-top members of different coalitions are adjacent to each other. That is, suppose that $v_i \in C_j$ and $v_{i+1} \in C_k \neq C_j$, with $v_i \neq \min C_j$ and $v_{i+1} \neq \min C_k$. By the first part of the proof, we know that the only candidate equilibrium has $i$ and $i + 1$ placed in their efficient positions, with $b_{i+1} = b_{i+2}^+ < v_{i+1}$ and $b_i = b_{i+1}^+ < v_{i+1}$ (recall that the statement about the magnitudes of the bids follows from Observation 1 about the best-response functions). However, it is obvious that $b_{i+1}$ is not a (static) best response to $b_i$: if, holding $b_j$ fixed, $C_k$ deviates to setting $b_{i+1}' = b_i^+$, $i + 1$’s individual payoff increases by $(v_{i+1} - b_i)(x^i - x^{i+1}) > 0$, without perceptibly increasing the payoff of other members of $C_k$. Thus, $b_{i+1} \notin BR_k(b_{C_j})$, i.e., we are not in a RAE.

**References**


Aumann, Robert J. 1959. “Acceptable Points in General Cooperative n-Person Games.” Contributions to the Theory of Games IV.


