

Auctioning Keywords in Online Search

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Web Appendix

Derivation of the Expected Revenue. Denote $V(v)$ as the equilibrium payoff of an advertiser of type v , and $P_j(v)$ as the equilibrium probability for him to win the j th share.

So

$$(W1) \quad V(v) = v \sum_{j=1}^n Q(s_j)P_j(v) - E[\text{equilibrium payment of type } v]$$

First notice that

$$(W2) \quad V(v) \geq V(\tilde{v}) + (v - \tilde{v}) \sum_{j=1}^n Q(s_j)P_j(\tilde{v}),$$

where the right-hand side is the payoff of a type- v bidder if he follows instead an equilibrium strategy of a type- \tilde{v} bidder.

Therefore we have

$$(W3) \quad V(v) \geq V(v + dv) + (-dv) \sum_{j=1}^n Q(s_j)P_j(v + dv)$$

$$(W4) \quad V(v + dv) \geq V(v) + dv \sum_{j=1}^n Q(s_j)P_j(v)$$

Reorganizing the above two equations yields

$$(W5) \quad \sum_{j=1}^n Q(s_j)P_j(v + dv) \geq \frac{V(v + dv) - V(v)}{dv} \geq \sum_{j=1}^n Q(s_j)P_j(v)$$

Taking the limit of dv to zero, we obtain $\frac{dV(v)}{dv} = \sum_{j=1}^n Q(s_j)P_j(v)$. Moving dv to the right hand side, integrating both sides from \underline{v} to v , and assuming $V(\underline{v}) = 0$ (the lowest type gets zero payoff), we get

$$(W6) \quad V(v) = \sum_{j=1}^n Q(s_j) \int_{\underline{v}}^v P_j(x) dx, \text{ for } v \in [\underline{v}, \bar{v}].$$

Notice the expected payment from an advertiser of type v is $v \sum_{j=1}^n Q(s_j)P_j(v) - V(v)$ by (W1). So the expected payment from one bidder is

$$(W7) \quad \begin{aligned} E \left[v \sum_{j=1}^n Q(s_j)P_j(v) - V(v) \right] &= \int_{\underline{v}}^{\bar{v}} \left[v \sum_{j=1}^n Q(s_j)P_j(v) - \sum_{j=1}^n Q(s_j) \int_{\underline{v}}^v P_j(t) dt \right] f(v) dv \\ &= \int_{\underline{v}}^{\bar{v}} \left[v \sum_{j=1}^n Q(s_j)P_j(v) f(v) - (1 - F(v)) \sum_{j=1}^n Q(s_j)P_j(v) \right] dv \\ &= \sum_{j=1}^n Q(s_j) \int_{\underline{v}}^{\bar{v}} P_j(v) \left(v - \frac{1 - F(v)}{f(v)} \right) f(v) dv \end{aligned}$$

The total expected revenue from all advertisers is n times the above. With strictly increasing bidding functions, the equilibrium probability for an advertiser of type v to win the j th share is the likelihood that $j - 1$ of his competitors have higher valuation than his and the rest of the competitors have lower valuation, and thus $P_j(v)$ can be specified as in (7). ■

Proof of Lemma 1. (a) For $j = 1, 2, \dots, n - 1$,

$$(W8) \quad \begin{aligned} \alpha_j &= n \int_{\underline{v}}^{\bar{v}} P_j(v) [vf(v) - (1 - F(v))] dv = n \int_{\underline{v}}^{\bar{v}} P_j(v) d[-v(1 - F(v))] \\ &= n \int_{\underline{v}}^{\bar{v}} v(1 - F(v)) dP_j(v) \\ &= n \int_{\underline{v}}^{\bar{v}} \binom{n-1}{n-j} F(v)^{n-j-1} (1 - F(v))^{j-1} [(n - j) - (n - 1)F(v)] vf(v) dv \end{aligned}$$

where the third step is due to integration by parts. We can easily verify $\alpha_1 > 0$.

For $j = 2, 3, \dots, n - 1$,

$$(W9) \quad \alpha_1 - \alpha_j = n \int_{\underline{v}}^{\bar{v}} \{(n-1)F(v)^{n-2}(1-F(v)) - \binom{n-1}{n-j}F(v)^{n-j-1}(1-F(v))^{j-1}[(n-j) - (n-1)F(v)]\}vf(v)dv$$

Denoting $A(v) \equiv (n-1)F(v)^{j-1} - \binom{n-1}{n-j}(1-F(v))^{j-2}[(n-j) - (n-1)F(v)]$, we can rewrite (W9) as $\alpha_1 - \alpha_j = n \int_{\underline{v}}^{\bar{v}} [1-F(v)]F(v)^{n-j-1}A(v)vf(v)dv$. We argue that $A(v)$ single-crosses zero from below on $[\underline{v}, \bar{v}]$. To see, let v^0 be the solution to $(n-j) - (n-1)F(v) = 0$. We can verify that $A(\underline{v}) < 0$, $A(v)$ increases in v for $v \leq v^0$, and $A(v)$ is positive for all $v > v^0$. Thus $A(v)$ crosses zero only once from below, implying $[1-F(v)]F(v)^{n-j-1}A(v)f(v)$ also single-crosses zero from below on (\underline{v}, \bar{v}) . Denoting the crossing point of the latter as v^c , we have

$$(W10) \quad \begin{aligned} \alpha_1 - \alpha_j &= n \int_{\underline{v}}^{\bar{v}} (1-F(v))F(v)^{n-j-1}A(v)f(v)vdv \\ &> nv^c \int_{\underline{v}}^{\bar{v}} (1-F(v))F(v)^{n-j-1}A(v)f(v)dv \\ &= nv^c \int_{\underline{v}}^{\bar{v}} [P_2(v) - jP_{j+1}(v) + (j-1)P_j(v)]f(v)dv \end{aligned}$$

where the last equality results from substituting the definition of $A(v)$ and rearranging terms. The right side of (W10) is zero because for $j = 1, \dots, n$,

$$(W11) \quad \begin{aligned} \int_{\underline{v}}^{\bar{v}} P_j(v)f(v)dv &= \binom{n-1}{n-j} \int_{\underline{v}}^{\bar{v}} F(v)^{n-j}[1-F(v)]^{j-1}dF(v) \\ &= \binom{n-1}{n-j} \int_0^1 x^{n-j}(1-x)^{j-1}dx \\ &= \binom{n-1}{n-j} \binom{n-1}{n-j}^{-1} \frac{1}{n} = \frac{1}{n} \end{aligned}$$

where the second step is due to integration by substitution and the third step is due to repeated integration by parts. Therefore, $\alpha_1 - \alpha_j > 0$ for $j = 2, 3, \dots, n - 1$.

We next show that $\alpha_1 - \alpha_n > 0$.

$$\begin{aligned}
\alpha_1 - \alpha_n &= n \int_{\underline{v}}^{\bar{v}} \{F(v)^{n-1} - [1 - F(v)]^{n-1}\} d[-v(1 - F(v))] \\
&= -n\underline{v} + n \int_{\underline{v}}^{\bar{v}} v [1 - F(v)] (n-1) [F(v)^{n-2} + (1 - F(v))^{n-2}] f(v) dv \\
&> -n\underline{v} + n\underline{v} \int_{\underline{v}}^{\bar{v}} [1 - F(v)] (n-1) [F(v)^{n-2} + (1 - F(v))^{n-2}] f(v) dv \\
&= -n\underline{v} + n\underline{v} \left(\frac{1}{n} + \frac{n-1}{n} \right) = 0
\end{aligned}$$

where the second step is due to integration by parts and the fourth step is due to (W11).

(b) Denote $h_j(x) \equiv nP_j(x) f(x)$. By (W11), $\int_{\underline{v}}^{\bar{v}} h_j(x) dx = 1$. Thus we can regard $h_j(x)$ as a probability density function. We next show that for $j = 1, 2, \dots, n-1$, $h_j(x)$ first-order stochastically dominates $h_{j+1}(x)$.

$$\begin{aligned}
&h_j(x) - h_{j+1}(x) \\
&= nf(x) \left\{ \binom{n-1}{n-j} F(x)^{n-j} [1 - F(x)]^{j-1} - \binom{n-1}{n-j-1} F(x)^{n-j-1} [1 - F(x)]^j \right\} \\
&= \binom{n}{j} f(x) F(x)^{n-j-1} [1 - F(x)]^{j-1} [nF(x) - (n-j)]
\end{aligned}$$

Denote v_j^c as the solution to $nF(x) - (n-j) = 0$. Because $h_j(x) < h_{j+1}(x)$ for any $x \in (\underline{v}, v_j^c)$, $\int_{\underline{v}}^v h_j(x) dx < \int_{\underline{v}}^v h_{j+1}(x) dx$ for $v \in (\underline{v}, v_j^c)$. Because $h_j(x) > h_{j+1}(x)$ for any $x \in (v_j^c, \bar{v})$, $\int_v^{\bar{v}} h_j(x) dx > \int_v^{\bar{v}} h_{j+1}(x) dx$ for $v \in (v_j^c, \bar{v})$, which implies $\int_{\underline{v}}^v h_j(x) dx < \int_{\underline{v}}^v h_{j+1}(x) dx$ for $v \in (v_j^c, \bar{v})$ (note that $\int_{\underline{v}}^v h_j(x) dx = 1 - \int_v^{\bar{v}} h_j(x) dx$). In all, we have $\int_{\underline{v}}^v h_j(x) dx < \int_{\underline{v}}^v h_{j+1}(x) dx$ for any $v \in (\underline{v}, \bar{v})$, implying that $h_j(x)$ first-order stochastically dominates $h_{j+1}(x)$. According to the property of first-order stochastic dominance (e.g., Proposition 6.D.1 at page 195 of Mas-Colell, Whinston, and Green (1995)), if $J(x)$ is an increasing function of x , $\int_{\underline{v}}^{\bar{v}} h_j(x) J(x) dx > \int_{\underline{v}}^{\bar{v}} h_{j+1}(x) J(x) dx$. Therefore $\alpha_j > \alpha_{j+1}$.

■

Proof of Lemma 2. Assume the optimal share structure is $(s_1^*, s_2^*, \dots, s_n^*)$. Denote $\sum_{j=j_k+1}^{j_{k+1}} s_j^* \equiv \sigma$ and notice that $s_{j_k}^* \geq \frac{1}{j_{k+1}-j_k} \sigma \geq s_{j_{k+1}+1}^* \geq 0$ because of the size-order constraint.

$(s_{j_{k+1}}^*, s_{j_{k+2}}^*, \dots, s_{j_{k+1}}^*)$ must be the solution to the following maximization problem:

$$(W12) \quad \max \sum_{j=j_k+1}^{j_{k+1}} \alpha_j Q(s_j), \text{ subject to: } s_{j_{k+1}} \geq \dots \geq s_{j_{k+1}} \text{ and } \sum_{j=j_k+1}^{j_{k+1}} s_j \leq \sigma$$

$$(W13) \quad s_{j_k}^* \geq s_{j_{k+1}} \text{ and } s_{j_{k+1}} \geq s_{j_{k+1}+1}^*$$

We will work on the related maximization problem without constraint (W13) and check (W13) later. The Lagrangian function then can be written as

$$L = \sum_{j=j_k+1}^{j_{k+1}} \alpha_j Q(s_j) + \mu \left(\sigma - \sum_{j=j_k+1}^{j_{k+1}} s_j \right) + \sum_{j=j_k+1}^{j_{k+1}-1} \gamma_j (s_j - s_{j+1})$$

where μ and γ_j are Lagrange multipliers. Hence, the Kuhn-Tucher conditions are (let $\gamma_{j_k} \equiv 0$, $\gamma_{j_{k+1}} \equiv 0$)

$$(W14) \quad \alpha_j Q'(s_j) - \mu + \gamma_j - \gamma_{j-1} = 0, \text{ for } j = j_k + 1, \dots, j_{k+1}$$

Averaging (W14) for the first l shares and the remaining shares, respectively, we have

$$(W15) \quad \frac{1}{l} \left(\sum_{j=j_k+1}^{j_k+l} \alpha_j Q'(s_j) + \gamma_{j_k+l} \right) = \frac{1}{j_{k+1} - j_k - l} \left(\sum_{j=j_k+l+1}^{j_{k+1}} \alpha_j Q'(s_j) - \gamma_{j_k+l} \right).$$

By definition, j_{k+1} is the maximizer for the average return factor starting from $j_k + 1$, so

$$(W16) \quad \frac{1}{l} \sum_{j=j_k+1}^{j_k+l} \alpha_j \leq \frac{1}{j_{k+1} - j_k - l} \sum_{j=j_k+l+1}^{j_{k+1}} \alpha_j$$

Also note that $Q'(s_j)$ is nondecreasing in j . Therefore, we have $\frac{1}{l} \sum_{j=j_k+1}^{j_k+l} \alpha_j Q'(s_j) \leq \frac{1}{j_{k+1}-j_k-l} \sum_{j=j_k+l+1}^{j_{k+1}} \alpha_j Q'(s_j)$. If $\gamma_{j_k+l} = 0$, (W15) can hold only if (W16) holds in equality and $s_{j_k+1} = \dots = s_{j_{k+1}}$. In other words, if any $\gamma_j = 0$ ($j_k < j < j_{k+1}$), we must have $s_{j_k+1} = \dots = s_{j_{k+1}}$. Otherwise, we have $\gamma_j > 0$ for all $j_k < j < j_{k+1}$, which implies $s_{j_k+1} = \dots = s_{j_{k+1}}$.

by the Kuhn-Tucker condition. So, regardless, we have $s_{j_{k+1}} = \dots = s_{j_{k+1}} = \frac{1}{j_{k+1}-j_k}\sigma$, which naturally satisfies constraint (W13). ■

Proof of Lemma 3. It is sufficient to show that if $\hat{s}_j \geq s_j$ then $\hat{s}_{j+1} \geq s_{j+1}$, for any j .

If j and $j + 1$ are located in the same plateau, $\hat{s}_j = \hat{s}_{j+1}$ and $s_j = s_{j+1}$, and $\hat{s}_{j+1} \geq s_{j+1}$ holds trivially. So we assume j and $j + 1$ are located in plateau k and $k + 1$, respectively.

If $s_{j+1} = 0$, the result holds trivially. Suppose $s_{j+1} > 0$ (so that $\hat{s}_j \geq s_j > s_{j+1} > 0$). By Proposition 2, we have $\hat{\alpha}_k Q'(\hat{s}_j) \geq \hat{\alpha}_{k+1} Q'(\hat{s}_{j+1})$ and $\bar{\alpha}_k Q'(s_j) = \bar{\alpha}_{k+1} Q'(s_{j+1})$. Together with condition (17), we have

$$(W17) \quad \frac{Q'(\hat{s}_j)}{Q'(\hat{s}_{j+1})} \geq \frac{\hat{\alpha}_{k+1}}{\hat{\alpha}_k} \geq \frac{\bar{\alpha}_{k+1}}{\bar{\alpha}_k} = \frac{Q'(s_j)}{Q'(s_{j+1})}$$

which implies $\frac{Q'(\hat{s}_j)}{Q'(s_j)} \geq \frac{Q'(\hat{s}_{j+1})}{Q'(s_{j+1})}$. Note that $\frac{Q'(\hat{s}_j)}{Q'(s_j)} \leq 1$ by concavity of $Q(\cdot)$ and $\hat{s}_j \geq s_j$. So we have $\frac{Q'(\hat{s}_{j+1})}{Q'(s_{j+1})} \leq 1$, which implies $\hat{s}_{j+1} \geq s_{j+1}$. ■

Lemma W1 (Ranking of $\alpha_j(v_0)$) *Under the MHR condition, for any marginal type $v_0 \in (\underline{v}, \bar{v})$, if $\alpha_j(v_0) > 0$, $\alpha_j(v_0) > \alpha_{j+1}(v_0)$; if $\alpha_j(v_0) \leq 0$, $\alpha_{j+1}(v_0) < 0$.*

Proof. For the case $v_0 \in (\underline{v}, \bar{v})$, define $h_j(x|x \geq v_0) \equiv \frac{h_j(x)}{\int_{v_0}^{\bar{v}} h_j(t) dt}$. Following steps in the proof of Lemma 1(b), we can similarly show that $h_i(x|x \geq v_0)$ first-order stochastically dominates $h_{i+1}(x|x \geq v_0)$. Thus,

$$(W18) \quad \int_{v_0}^{\bar{v}} h_j(x|x \geq v_0) J(x) dx > \int_{v_0}^{\bar{v}} h_{j+1}(x|x \geq v_0) J(x) dx, \text{ for } j = 1, 2, \dots, n-1.$$

Substituting $h_j(x|x \geq v_0)$ with $\frac{h_j(x)}{\int_{v_0}^{\bar{v}} h_j(t) dt}$ and rearranging, we have

$$(W19) \quad \alpha_j(v_0) = \int_{v_0}^{\bar{v}} h_j(x) J(x) dx > \frac{\int_{v_0}^{\bar{v}} h_j(t) dt}{\int_{v_0}^{\bar{v}} h_{j+1}(t) dt} \int_{v_0}^{\bar{v}} h_{j+1}(x) J(x) dx = \frac{\int_{v_0}^{\bar{v}} h_j(t) dt}{\int_{v_0}^{\bar{v}} h_{j+1}(t) dt} \alpha_{j+1}(v_0).$$

Suppose $\alpha_j(v_0) > 0$. If $\alpha_{j+1}(v_0) \geq 0$, from $\int_{v_0}^{\bar{v}} h_j(t) dt > \int_{v_0}^{\bar{v}} h_{j+1}(t) dt > 0$ (because $h_j(t)$ first-order stochastically dominates $h_{j+1}(t)$), we have $\alpha_j(v_0) > \alpha_{j+1}(v_0)$. If $\alpha_{j+1}(v_0) < 0$, it is easy to see $\alpha_j(v_0) > \alpha_{j+1}(v_0)$.

If $\alpha_j(v_0) \leq 0$, (W19) implies $\alpha_{j+1}(v_0) < 0$. ■

References

- [1] Mas-Colell, Andreu, Michael D. Whinston, and Jerry R. Green (1995), *Microeconomic Theory*. Oxford University Press, New York.