

**A Comment on “Can Relaxation of Beliefs Rationalize the Winner’s Curse?:  
An Experimental Study”**

by

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**Supplementary Material**

In this note, we provide (i) the derivation of the cursed equilibrium and analogy-based expectation equilibrium (which are the same under our specification) and (ii) the derivation of  $L1$ 's bids with a discrete bid space (which is analogous to ILN's Proposition 3).

## A Analogy-Based Expectation Equilibrium and Cursed Equilibrium

The set of signals is  $X = \{0, 1, \dots, 10\}$ . Let  $\omega_{x_1, x_2}$  correspond to the state where player  $i$ 's signal is  $x_i$  ( $i = 1, 2$ ). There are 121 ( $= 11 \times 11$ ) possible states,  $\omega_{x_1, x_2} \in \Omega$ . The item's value at  $\omega_{x_1, x_2}$  is  $\max\{x_1, x_2\}$ . Each state is equally possible. We have

$$\begin{aligned} \mathcal{P}_1 &= \{ \{ \omega_{x_1, 0}, \omega_{x_1, 1}, \omega_{x_1, 2}, \omega_{x_1, 3}, \omega_{x_1, 4}, \omega_{x_1, 5}, \omega_{x_1, 6}, \omega_{x_1, 7}, \omega_{x_1, 8}, \omega_{x_1, 9}, \omega_{x_1, 10} \}_{x_1 \in X} \} \\ \mathcal{P}_2 &= \{ \{ \omega_{0, x_2}, \omega_{1, x_2}, \omega_{2, x_2}, \omega_{3, x_2}, \omega_{4, x_2}, \omega_{5, x_2}, \omega_{6, x_2}, \omega_{7, x_2}, \omega_{8, x_2}, \omega_{9, x_2}, \omega_{10, x_2} \}_{x_2 \in X} \}. \end{aligned}$$

where  $\mathcal{P}_i$  is the partition of the states from player  $i$ 's point of view.

Let  $\mathcal{A}_i$  be the analogy partitions of the states from player  $i$ 's point of view ( $i = 1, 2$ ). For the analogy based expectation equilibrium, we assume that  $\mathcal{A}_i = \mathcal{P}_i$  for  $i = 1, 2$ , i.e., the *private information analogy partition* (Jehiel and Koessler (2008, p. 538)). This is visualized in Figure 1. As Jehiel and Koessler (2008, p.539) and Eyster and Rabin (2005, p.1634) note, this specification coincides with the *fully* cursed equilibrium (i.e.,  $\chi = 1$ , in Eyster and Rabin (2005)  $\chi$ -cursed equilibrium). Thus, the analogy-based expectation equilibrium that we construct is also a fully cursed equilibrium.

Note that  $E[X^{max} | 10] = 10$ . For  $x_i \in X \setminus \{10\}$ , the expected value of the item is

$$\begin{aligned} E[X^{max} | x_i] &= \left(\frac{1}{11}\right) 10 + \dots + \left(\frac{1}{11}\right) (x_i + 1) + \left(\frac{x_i + 1}{11}\right) x_i \\ &= \left(\frac{1}{11}\right) \sum_{l=x_i+1}^{10} l + \left(\frac{x_i + 1}{11}\right) x_i \end{aligned}$$

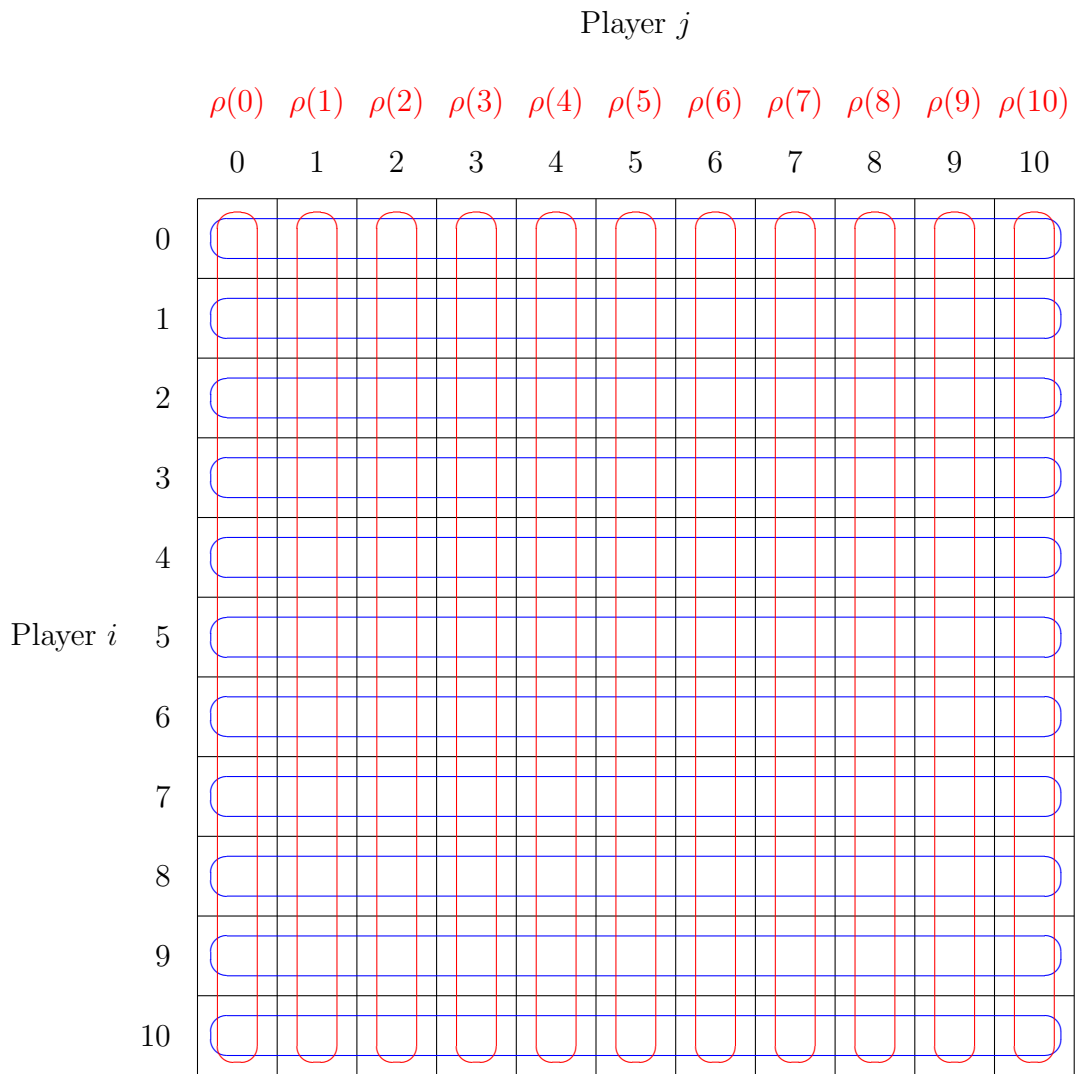


Figure 1: Partitions of States

$x_i$	0	1	2	3	4	5	6	7	8	9	10
$E[X^{max}   x_i]$	5	$\frac{56}{11}$	$\frac{58}{11}$	$\frac{61}{11}$	$\frac{65}{11}$	$\frac{70}{11}$	$\frac{76}{11}$	$\frac{83}{11}$	$\frac{91}{11}$	$\frac{100}{11}$	10
$b^l(x_i)$	5	5.09	5.27	5.54	5.90	6.36	6.90	7.54	8.27	9.09	10
$b^h(x_i)$	5	5.10	5.28	5.55	5.91	6.37	6.91	7.55	8.28	9.10	10
$\rho(x_i)$	-	$\frac{10}{11}$	$\frac{8}{11}$	$\frac{5}{11}$	$\frac{1}{11}$	$\frac{7}{11}$	$\frac{1}{11}$	$\frac{5}{11}$	$\frac{8}{11}$	$\frac{10}{11}$	-

Table 1:  $x_i$ ,  $E[X^{max} | x_i]$  and Equilibrium Strategy

$b$	5	5.09	5.10	5.27	5.28	5.54	5.55	5.90	5.91	6.36
$\bar{\rho}(b)$	$\frac{1}{11}$	$\frac{10}{121}$	$\frac{1}{121}$	$\frac{8}{121}$	$\frac{3}{121}$	$\frac{5}{121}$	$\frac{6}{121}$	$\frac{1}{121}$	$\frac{10}{121}$	$\frac{7}{121}$
$b$	6.37	6.90	6.91	7.54	7.55	8.27	8.28	9.09	9.10	10
$\bar{\rho}(b)$	$\frac{4}{121}$	$\frac{1}{121}$	$\frac{10}{121}$	$\frac{5}{121}$	$\frac{6}{121}$	$\frac{8}{121}$	$\frac{3}{121}$	$\frac{10}{121}$	$\frac{1}{121}$	$\frac{1}{11}$

Table 2: Strategy of player  $j$  perceived by player  $i$

$$\begin{aligned}
&= \left(\frac{1}{11}\right) \left(55 - \frac{x_i(x_i + 1)}{2}\right) + \frac{x_i(x_i + 1)}{11} \\
&= \frac{x_i(x_i + 1) + 110}{22}.
\end{aligned}$$

Table 1 shows the value of  $E[X^{max} | x_i]$  for each  $x_i \in X$ .

Suppose that player  $i$  with  $x_i$  chooses  $b^l(x_i) = \frac{[100E[X^{max}|x_i]]}{100}$  with probability  $\rho(x_i) \in (0, 1)$  and  $b^h(x_i) = \frac{[100E[X^{max}|x_i]]}{100}$  with  $1 - \rho(x_i)$  where

$$\rho(x_i) = \frac{b^h(x_i) - E[X^{max} | x_i]}{b^h(x_i) - b^l(x_i)} \in (0, 1)$$

for each  $x_i \in X$ .<sup>1</sup> Table 1 shows their values for each  $x_i \in X$ . It is important to note that for each  $x_i$ , (i)  $b^l(x_i) < E[X^{max} | x_i] < b^h(x_i)$  and (ii)  $\rho(x_i)b^l(x_i) + (1 - \rho(x_i))b^h(x_i) = E[X^{max} | x_i]$ . Let

$$B^p = \{b \in B \mid \text{there exists } x_i \text{ such that } b = b^l(x_i) \text{ or } b = b^h(x_i)\}.$$

Given the strategy specified above, Table 2 shows *the strategy for player  $j$  perceived by player  $i$* ,  $\bar{\rho}$ . Note that  $\bar{\rho}$  is independent of the state.

Given  $\bar{\rho}$ , the expected payoff for player  $i$  with  $x_i$  from  $b_i \in B$  is computed as follows: If

<sup>1</sup>That is, we choose two numbers in  $B$  closest to  $E[X^{max} | x_i]$ .

player  $i$  chooses  $b_i \in B \setminus B^p$ ,

$$\left(\frac{1}{11}\right) \sum_{x_j \in X} \sum_{b < b_i, b \in B^p} \bar{\rho}(b) \{\max\{x_i, x_j\} - b\} = \sum_{b < b_i, b \in B^p} \bar{\rho}(b) \{E[X^{max} | x_i] - b\}$$

while if player  $i$  chooses  $b_i \in B^p$ ,

$$\begin{aligned} & \left(\frac{1}{11}\right) \sum_{x_j \in X} \left[ \sum_{b < b_i, b \in B^p} \bar{\rho}(b) \{\max\{x_i, x_j\} - b\} + \bar{\rho}(b_i) \{\max\{x_i, x_j\} - b_i\} \left(\frac{1}{2}\right) \right] \\ &= \sum_{b < b_i, b \in B^p} \bar{\rho}(b) \{E[X^{max} | x_i] - b\} + \bar{\rho}(b_i) \{E[X^{max} | x_i] - b_i\} \left(\frac{1}{2}\right). \end{aligned}$$

Note (i) that  $b_i = 5$  is a best response for  $x_i = 0$ , (ii)  $b_i = 10$  is a best response for  $x_i = 10$ , and (iii) that every  $b_i < b^l(x_i)$  and  $b_i > b^h(x_i)$  cannot be a best response for each  $x_i \in X \setminus \{0, 10\}$ , meaning that the only remaining bids are  $\{b^l(x_i), b^h(x_i)\}$  for each  $x_i \in X \setminus \{0, 10\}$ .<sup>2</sup> Given  $x_i \in \{0, 10\}$ , the expected payoff from  $b_i = b^l(x_i)$  is

$$\sum_{b < b^l(x_i), b \in B^p} \bar{\rho}(b) \{E[X^{max} | x_i] - b\} + \bar{\rho}(b^l(x_i)) \{E[X^{max} | x_i] - b^l(x_i)\} \left(\frac{1}{2}\right) \quad (1)$$

while the expected payoff from  $b_i = b^h(x_i)$  is

$$\sum_{b < b^h(x_i), b \in B^p} \bar{\rho}(b) \{E[X^{max} | x_i] - b\} + \bar{\rho}(b^h(x_i)) \{E[X^{max} | x_i] - b^h(x_i)\} \left(\frac{1}{2}\right). \quad (2)$$

Then, we have

$$\begin{aligned} & (2) - (1) \\ &= \bar{\rho}(b^l(x_i)) \{E[X^{max} | x_i] - b^l(x_i)\} \left(\frac{1}{2}\right) + \bar{\rho}(b^h(x_i)) \{E[X^{max} | x_i] - b^h(x_i)\} \left(\frac{1}{2}\right) \\ &= \left(\frac{1}{22}\right) \left[ \rho(x_i) \{E[X^{max} | x_i] - b^l(x_i)\} + (1 - \rho(x_i)) \{E[X^{max} | x_i] - b^h(x_i)\} \right] \\ &= \left(\frac{1}{22}\right) \left\{ E[X^{max} | x_i] - [\rho(x_i)b^l(x_i) + (1 - \rho(x_i))b^h(x_i)] \right\} \\ &= 0 \end{aligned}$$

where the last equality comes from the fact that  $\rho(x_i)b^l(x_i) + (1 - \rho(x_i))b^h(x_i) = E[X^{max} | x_i]$  for each  $x_i \in X$ . This shows that both  $b^l(x_i)$  and  $b^h(x_i)$  are best responses for each  $x_i \in X \setminus \{0, 10\}$ . The strategy specified above hence constitutes an analogy-based expectation

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<sup>2</sup>More generally, any  $b_i < 5.09$  is a best response for  $x_i = 0$  and any  $b_i > 9.10$  is a best response for  $x_i = 10$ .

equilibrium with private information analogy partition, and hence is also a fully cursed equilibrium.

## B ILN's Proposition 3 with a Discrete Bid Space

We derive *random L1*'s best-responses for *BL/SBF* under the assumption that *random L0* randomizes uniformly over  $B^0 = \{0, 0.01, \dots, 9.99, 10\} \subset B$ . Given signal  $x_i \in X$ , the following expression is the expected payoff for player  $i$  by bidding  $b_i \in B$ . Note that we use "cents" instead of "dollars".

$$\begin{aligned}
& \left( \frac{1}{11} \right) \sum_{x_j \in X} \left[ \left( \frac{1}{|B^0|} \right) \sum_{b < b_i, b \in B^0} \{\max\{x_i, x_j\} - b\} + \left( \frac{1}{|B^0|} \right) \{\max\{x_i, x_j\} - b_i\} \left( \frac{1}{2} \right) \right] \\
&= \left( \frac{1}{|B^0|} \right) \sum_{b < b_i, b \in B^0} \{E[X^{max} | x_i] - b\} + \left( \frac{1}{|B^0|} \right) \{E[X^{max} | x_i] - b_i\} \left( \frac{1}{2} \right) \\
&= \left( \frac{b_i}{|B^0|} \right) E[X^{max} | x_i] - \left( \frac{1}{|B^0|} \right) \left( \frac{(b_i - 1)b_i}{2} \right) + \left( \frac{1}{|B^0|} \right) \{E[X^{max} | x_i] - b_i\} \left( \frac{1}{2} \right) \\
&= \left( \frac{b_i + \frac{1}{2}}{|B^0|} \right) E[X^{max} | x_i] - \left( \frac{1}{|B^0|} \right) \left[ \frac{(b_i - 1)b_i}{2} + \frac{b_i}{2} \right] \\
&= \left( \frac{1}{|B^0|} \right) \left\{ \left( b_i + \frac{1}{2} \right) E[X^{max} | x_i] - \left( \frac{b_i^2}{2} \right) \right\}.
\end{aligned}$$

Given that  $b_i \in B^0$  and the expression above is quadratic, the expected payoff is maximized at either  $E[X^{max} | x_i] - 0.01$  or  $E[X^{max} | x_i] + 0.01$  (in dollars). We numerically computed the values of the expected payoff at these points, and selected the bid with the highest expected payoff as the best-response.