

UTMD-039

# Unprecedented

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#### Abstract

This paper considers a dynamic game in which each player can take a new action only if either she privately learns it or the opponent takes it. The new action profile is a Nash equilibrium, and is Pareto dominated by the default action profile. Under the assumptions that taking the new action is an irreversible choice and moves are asynchronous, we show that there is a unique perfect Bayesian equilibrium when the probability of private learning is low and the players are patient. In the unique equilibrium, the new action is never taken, i.e., the new action remains unprecedented. This is the case even though, after many periods, it is almost common knowledge among the players that they have learned the new action.

JEL Classification: C73; D80

*Keywords:* precedents; asynchronous moves; unique equilibrium; almost common knowledge; repeated game

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## 1 Introduction

There are many situations where an action can be taken only when there is a precedent. Absence of a precedent is often the public sector's favorite excuse to avoid doing something new, and this "precedentism" is a target of criticism not only by mass media and the general public but also by academics (Merton, 1940, 1968).

If an action cannot be taken without a precedent, then that action can never be taken. In real life, however, some people may accidentally find out a way to take a new action or is granted a right to take the action for some exogenous reason. For example, a firm may invent a promotion strategy to effectively attract customers, and once this firm employs such a strategy, it may be used by any other firms.<sup>1</sup> Technological innovations by firms or development of war weapons by countries are also an example of an action that can be taken after seeing a precedent.<sup>2</sup> Even when a city hall has been declining the request to provide data to researchers because there is no precedent, a researcher may accidentally get a connection to the city mayor, which enables her to get a permission to obtain data from the city hall. Once she uses this right and obtains data, it becomes a precedent and other researchers can also obtain data. In sports, once an athlete acquires a new technique and performs it in a competition, other athletes will start mimicking that technique. The "Tomoa Skip" in the speed category of sport climbing and the "Fosbury Flop" in the high jump are now used by almost all athletes in the respective competitions.<sup>3</sup>

In this paper, we are interested in the incentives of players when a precedent makes a new action feasible. Specifically, we consider the situation in which a  $2 \times 2$  stage game is repeatedly played by two players. In the stage game, one action is the "old action" that the players can play from the initial period and the other action is the "new action" that a player can only take when either she has accidentally learned it privately or the opponent has chosen it in the past. We consider a particular class of stage games in which the old action profile Pareto dominates the new action profile (as in the case of technological innovations that are harmful for the climate or development of war weapons) while the latter is a Nash equilibrium. This class

<sup>&</sup>lt;sup>1</sup>For instance, since American Airlines had launched the world's first mileage-based frequent-flyer program (AAdvantage) in 1981, more than 130 airlines have issued miles as of 2005 (The Economist, 2005).

<sup>&</sup>lt;sup>2</sup>See, for instance, Pandya (2019).

 $<sup>^{3}</sup>$ There are many other examples of performance breakthroughs in sports that have become new standards. See Veenendaal (2018).

includes prisoner's dilemma and coordination games.

Under the assumptions to be discussed shortly, we show that the model has a unique perfect Bayesian equilibrium when the probability of private learning is small and the players are sufficiently patient. In the unique equilibrium, players keep playing the old action even if both of them have privately learned the new action. Hence, *the new action will remain unprecedented*.

There are two assumptions in obtaining this conclusion: *irreversibility* and *asynchronicity*. Irreversibility means that once a player chooses the new action, she cannot switch back to the old action. Asynchronicity means that the two players alternate in choosing actions.<sup>4</sup> We show that if one of these assumptions fails, there are multiple equilibria.

Why will the new action remain unprecedented? To understand the intuition, suppose that the probability of private learning is small and players use pure strategies. If a patient player happens to privately learn the new action in the first period, she does not want to play it right away because it is unlikely that the opponent has already learned it, and thus keeping the new action a "secret" is a good strategy. This argument may seem to break down when the time passes and the probability that each player has learned the new action has become large. Suppose that player i is supposed to play the new action at (a large) time t. If she followed this strategy, then she might obtain an instantaneous gain, but from the next period on the two players would be playing the new action forever. If, however, she deviates and plays the old action, that would substantially change the opponent j's belief about the likelihood of player i has an incentive to deviate and play the old action. This argument suggests that the new action will never be played in any equilibrium. The actual proof is more contrived as we consider mixed strategies as well.

The irreversibility and asynchronicity play key roles in the preceding argument. We used irreversibility when claiming that once a player takes the new action, the players will be playing the new action forever. Without irreversibility, the game

<sup>&</sup>lt;sup>4</sup>We do not claim that these are the most natural assumptions in all applications, but they may be natural in some settings. For an example of irreversibility, a firm may have a reputation concern and may not be able to switch back to providing no promotion once a promotion strategy is launched. Also, irreversibility could seem natural if we view the problem as that of an optimal stopping problem. For an example of asynchronicity, the decision times of competing firms are typically not coordinated with each other.

after the new action is played is the standard repeated game, so the folk theorem generically holds in this continuation game. Players can condition their continuation play on who has played the new action first, so as to incentivize players to play the new action as soon as they learn it. This results in the existence of an equilibrium in which players play the new action as soon as they can. Also, we used asynchronicity when discussing the consequence of a deviation to play the old action. If moves are synchronous, the opponent may be playing the new action at the same time as the player takes the new action: When it is likely that the opponent has learned the new action and hence plays it with high probability, it is risky to take the old action. This implies the existence of an equilibrium in which the new action is played on the equilibrium path.<sup>5</sup>

Our objective is to understand the strategic implications of the role of new actions and our predictions may sometimes be at odds with reality. Indeed, we acknowledge that, in some situations in reality, new actions are taken due to various factors. In aforementioned examples of sports, athletes' career is typically short and the benefit from using a new technique may be high in some particular competitions.<sup>6</sup> In other situations, however, new actions are unprecedented, being consistent with our prediction. For example, nuclear weapons were not used between the United States and the Soviet Union during the Cold War.<sup>7</sup> While one could explain this by considering the grim trigger strategy in standard models, we show that it is in fact the unique equilibrium outcome in our model. Another example is a firm launching a new social product. The customers' decision problem can be described by a coordination game in which they choose the existing or the new product and want to match with the peers.<sup>8</sup> Our uniqueness result strengthens our common sense that the firm cannot just hope for the customers to coordinate on their new product if it is worse than the existing one.

The insight that a sharper prediction can be obtained when moves are asyn-

<sup>&</sup>lt;sup>5</sup>For some parameter range, we use a public randomization device to obtain the existence of such an equilibrium. Our uniqueness result under asynchronicity, which is our main theorem, holds even if a public randomization device exists.

<sup>&</sup>lt;sup>6</sup>Indeed, Fosbury Flop was first used in the Olympics and led the athlete to the gold medal.

<sup>&</sup>lt;sup>7</sup>Irreversibility is satisfied by regarding the first period in which the new action is taken as the day on which a nuclear weapon is used (instead of imagining that a nuclear weapon is used every day). After that day, the effect of radiations continues and the victim country would have a less international pressure not to use a nuclear weapon.

<sup>&</sup>lt;sup>8</sup>Irreversibility would capture an expensive adoption cost.

chronous appears in the literature, e.g., Maskin and Tirole (1987, 1988a,b) and Lagunoff and Matsui (1997, 2001). The key factor in these papers is that a player can commit to an action while the opponent chooses an action. The key factor in our model is rather that a player can take a certain action to influence the opponent's belief update while the opponent is not choosing an action.

In our unique equilibrium, the new action remains unprecedented even when it is very close to common knowledge that the players have learned the new action. However, if it were common knowledge that both players have learned the new action and the stage game was a prisoner's dilemma, there would exist an equilibrium in which the new action would be taken. This may look at odds with the result of Monderer and Samet (1989) that shows certain continuity of the equilibrium set with respect to the degree of common beliefs. Section 4.3 discusses how to reconcile this seeming contradiction, where we argue that in our model, there is an action that can change what is almost common knowledge, which contributes to the difference.

We assume that a new action is initially unavailable to the players. This assumption can be interpreted as that of a feasibility constraint, but another possible interpretation is that the players are initially unaware of the new action, and they learn it through private learning or a precedent. Defining strategies and equilibria in dynamic games with unawareness is a tricky task because of the issue that arises when a player's strategy conditions on what she is unaware of. In our model, however, the interpretation based on unawareness is still valid as long as there are only two actions (as in our main model). This is because a player's strategy is only relevant when the player has choices between multiple actions, and this happens only when the player has already learned the new action.<sup>9</sup> There is a burgeoning literature, such as Feinberg (2004), Chung and Fortnow (2016), Jéhiel and Newman (2019), Heifetz et al. (2021), and Schipper (2021), which studies dynamic games that involve unawareness about actions. In particular, Feinberg (2004) considers a finitely repeated prisoner's dilemma in which one player may be unaware of the "defect" action, while she can play it if the other player—who knows both actions—plays it. He showed that some cooperation will be achieved in equilibrium. His result uses the finite horizon of the game and is reminiscent of the reputation argument where the probability of

<sup>&</sup>lt;sup>9</sup>This argument breaks down if there are three or more actions as in Section 4.4. Whether player i's strategy (defined in that section) is a best response when i has a choice between two actions can depend on what the opponent j would do after j has learned the third action which i is currently unaware of.

unawareness is fixed. We, in contrast, consider infinite horizon and the probability of unawareness converges to zero due to continual private learning.

The paper is structured as follows. Section 2 formulates our model and Section 3 provides our main uniqueness result. Section 4 discusses several topics. Section 5 provides concluding remarks. The proofs are relegated to the Appendix. The Online Supplementary Appendix contains additional results and discussions.

## 2 Model

There are two players, 1 and 2. Time is discrete and the horizon is infinite:  $t = 1, 2, \ldots$ 

Player 1 chooses her action  $a_{1,t}$  from set  $A_{1,t} \subseteq \{O, N\}$  at every period t that is odd.<sup>10</sup> Player 2 chooses her action  $a_{2,t}$  from set  $A_{2,t} \subseteq \{O, N\}$  at every period t that is even. At each period in which player i moves, before i makes a choice of her action, she privately learns action N with probability  $p_i \in [0, 1]$ , independently across players and periods, if she has not learned it yet.

To define the players' action sets at every period, we define player *i*'s private history  $h_{i,t}$  at her moving time t as the following form:

$$h_{i,t} = ((a^{\tau})_{\tau \in \{1,\dots,t-1\}}, s) \in (\{O,N\} \times \{O,N\})^{t-1} \times \{1,\dots,t+1\},$$

where the interpretation is that each  $a^{\tau}$  is the action profile taken at period  $\tau$  in the past, and s is the period at which player i has privately learned action N if she has already privately learned it before or at time t, and s = t + 1 otherwise. If we can write  $h_{i,t}$  in the above form, we write  $s(h_{i,t}) := s$ .

Player *i*'s action set  $A_{i,t}$  is recursively defined according to the following rule. Initially, only action O is available to *i*. Action N becomes available to *i* if either *i* privately learns it or the opponent plays it in the past (i.e., learning through a precedent). Once player *i* chooses N, N becomes the only choice for *i*.

Formally, the set of actions available to the moving player i at period t can be

 $<sup>^{10}\</sup>mathrm{The}$  notation O stands for an "old" or "original" action, and N stands for a "new" action.

	O	N
0	1, 1	$-\ell_1, x_2$
N	$x_1, -\ell_2$	0,0

Table 1: The payoff matrix of the stage game

written as:

$$A_{i,t} = A(h_{i,t}) = \begin{cases} \{O\} & \text{if } s(h_{i,t}) = t+1 \text{ and } a_{-i,\tau} = O \text{ for all } \tau \le t-1 \\ \{N\} & \text{if } a_{i,\tau} = N \text{ for some } \tau \le t-2 \\ \{O, N\} & \text{otherwise} \end{cases}$$

Let  $H_{i,t}$  be the set of all private histories of player *i* at her moving time *t* that happen with positive probability under the above-specified rule for the action evolution. Let  $H_1 := \bigcup_{\tau \in \mathbb{N}} H_{1,2\tau-1}$  and  $H_2 := \bigcup_{\tau \in \mathbb{N}} H_{2,2\tau}$ .

Player *i*'s strategy is a mapping  $\sigma_i : H_i \to \Delta(\{O, N\})$  such that  $\sigma_i(h_{i,t})(A(h_{i,t})) =$ 1 for every  $h_{i,t} \in H_i$ . Let  $\Sigma_i$  be the set of all strategies of player *i*, and let  $\Sigma = \Sigma_1 \times \Sigma_2$ .

Player *i*'s stage-game payoff at time *t* under action profile  $(a_1^t, a_2^t)$ , which we denote by  $u_i(a_1^t, a_2^t)$ , is given by the payoff matrix as in Table 1, where  $\ell_i > 0$  and  $x_i \in \mathbb{R}$  for each i = 1, 2.

Thus, we consider the class of stage games with an inefficient Nash equilibrium: While action profile (N, N) is a Nash equilibrium, action profile (O, O) Paretodominates (N, N). When  $x_i > 1$  for each i = 1, 2, the stage game is the prisoner's dilemma. When  $x_i < 1$  for each i = 1, 2, the stage game is a coordination game (in this case, (O, O) is also a stage-game Nash equilibrium).<sup>11</sup> In other words, the class of stage games we consider contains the prisoner's dilemma and coordination games.

The discount factor is  $\delta_i \in (0,1)$  for each *i*. Given the action sequence  $(a^t)_{t=1}^{\infty}$ , player *i*'s payoff in the supergame is  $(1 - \delta_i) \sum_{t=1}^{\infty} \delta_i^{t-1} u_i(a^t)$ . We sometimes use the unnormalized payoff in the supergame, and it is  $\sum_{t=1}^{\infty} \delta_i^{t-1} u_i(a^t)$ .

Given *i*'s private history  $h_{i,t} \in H_i$  and strategy profile  $\sigma \in \Sigma$ , we can define player *i*'s continuation payoff  $\pi_i(\sigma|h_{i,t})$ .

Our solution concept is a perfect Bayesian equilibrium (PBE).

<sup>&</sup>lt;sup>11</sup>In particular, when  $x_i \in [0, 1)$ , the stage game is a stag hunt game.

## 3 Main Result: Unique Equilibrium

Let  $\sigma^{G}$  be the grim trigger strategy profile, that is, each player chooses action O if and only if no one has chosen N in the past. The following is the main theorem of this paper.

**Theorem 1.** There exist  $\overline{p} \in (0,1)$  and  $\overline{\delta} \in (0,1)$  such that if  $p_i < \overline{p}$  and  $\delta_i > \overline{\delta}$  for all i = 1, 2, then the grim trigger strategy profile  $\sigma^{G}$  is a unique PBE.

That is, in the unique equilibrium, action N keeps being unprecedented when private learning is rare and the players are patient.<sup>12</sup> This result is of particular interest because it provides certain discontinuity of the set of equilibria between exact and almost common knowledge. In our model, on the path of equilibrium play, after many periods, it is "almost common knowledge" among the players that they have learned action N. In this continuation game, the strategy profile in which each player plays N whenever she learns action N, which we denote by  $\sigma^N$ , does not constitute an equilibrium. However, when  $x_i > 1$  for each  $i = 1, 2, \sigma^N$  is an equilibrium in the game in which the "almost common knowledge" is replaced with the exact common knowledge.<sup>13</sup> We will discuss this issue in more depth in Section 4.3 after explaining the proof.

To explain the proof of our main result, we introduce some notation. Define  $\pi_i(\delta_i)$  as

 $\pi_i(\delta_i) := p_{-i}[1 + \delta_i(-\ell_i) + \delta_i^2 \cdot 0] + (1 - p_{-i})[1 + \delta_i \cdot 1 + \delta_i^2 x_i].$ 

To interpret this, suppose that the players play  $\sigma^N$  and, at period t, the moving player i privately learns action N while -i has not. Then,  $\pi_i(\delta_i)$  is i's unnormalized payoff when she makes a one-period deviation to play O at t (and follows the original strategy  $\sigma_i^N$  later on).

For each *i*, for a given  $p_{-i} \in [0, 1]$ , we define

$$\overline{\delta}_i(p_{-i}) := \min\{\delta'_i \in [0,1] \mid x_i < \pi_i(\delta_i) \text{ for all } \delta_i \in (\delta'_i,1)\}, \text{ and} \\ \underline{\delta}_i(p_{-i}) := \max\{\delta'_i \in [0,1] \mid x_i < \pi_i(\delta_i) \text{ for all } \delta_i \in [0,\delta'_i)\},$$

<sup>&</sup>lt;sup>12</sup>One could make the model satisfy those conditions by assuming short period lengths so that private learning occurs with small probability and the players discount only little during any given period.

<sup>&</sup>lt;sup>13</sup>In our context, common knowledge is equivalent to common certainty (common belief with probability one).

where we sometimes omit the dependence on  $p_{-i}$ . The interpretation of these variables is the following. Suppose that the players play  $\sigma^N$  and, at period t, the moving player i privately learns action N while -i has not. The left-hand side in the defining inequality of  $\overline{\delta}_i(p_{-i})$ ,  $x_i$ , is i's unnormalized payoff when i follows this strategy. The right-hand side,  $\pi_i(\delta_i)$ , is her unnormalized payoff when she makes a one-period deviation to play O at t (and follows the original strategy  $\sigma_i^N$  later on). Thus, the discount factor  $\overline{\delta}_i(p_{-i})$  is the minimum one above which such one-period deviation is always profitable for player i. Likewise, the discount factor  $\underline{\delta}_i(p_{-i})$  is the maximum one below which such one-period deviation is always profitable for player i.

We also define

$$\hat{\delta}_i := \min\{\delta_i \in [0,1] \mid (1-\delta_i)x_i \le 1\}.$$

To interpret this, consider the strategy profile  $\sigma^{\text{G}}$ . Suppose that at period t, the moving player i privately learns action N while -i has not. The left-hand side in the defining inequality of  $\hat{\delta}_i$ ,  $(1 - \delta_i)x_i$ , is i's payoff when i makes a one-period deviation to play N at t (and, by the assumption on the evolution of the action sets, play N forever). The right-hand side, 1, is her payoff when she follows strategy  $\sigma_i^{\text{G}}$ . The discount factor  $\hat{\delta}_i$  is the minimum one for which such one-period deviation is not profitable for player i. Note that if  $x_i > 1$  then  $\hat{\delta}_i = \frac{x_i - 1}{x_i}$  and  $\hat{\delta}_i = 0$  otherwise.

Now, we are ready to give a proof of Theorem 1.

*Proof of Theorem 1.* We prove the theorem through a series of lemmas, whose formal proofs we provide in the Appendix. Here we provide the formal statements of the lemmas as well as their intuitions.

First, the standard argument shows that the grim trigger strategy profile  $\sigma^{G}$  is an equilibrium when the players are sufficiently patient.

**Lemma 1.** Suppose  $\delta_i \geq \hat{\delta}_i$  for each i = 1, 2. Then, for any  $p_1, p_2 \in [0, 1]$ ,  $\sigma^G$  is a *PBE*.

Next, we show that  $\sigma^{G}$  is the only possibility for an equilibrium under which action N is not observed on the path of play.

**Lemma 2.** For any  $\delta_1, \delta_2 \in (0, 1)$  and  $p_1, p_2 \in [0, 1]$ , fix any PBE  $\sigma$  in which, on the path of play, there is ex ante probability zero that some player plays N. Then,  $\sigma = \sigma^{G}$ . This lemma is straightforward as well because *i*'s playing N is a unique best response once -i chooses N. This follows from  $-\ell_i < 0$  and the assumption that -i will not be able to change his action from N.

Now, notice that at time t, the moving player i needs to make a non-trivial choice from multiple actions only at private histories of the form  $h_{i,t}$  such that  $h_{i,t}$  involves no observation of action N and she has already privately learned action N (i.e.,  $s(h_{i,t}) \leq t$ ). Let the set of such histories be  $\overline{H}_{i,t}$ .

Fix  $p_1, p_2 \in [0, 1]$  and  $\delta_1, \delta_2 \in (0, 1)$ . Suppose for a contradiction that there is a (potentially mixed) PBE other than  $\sigma^G$ , and fix an arbitrary one of them,  $\sigma^*$ . Let  $t_1$  be the first period at which some player's strategy  $\sigma_i^*$  assigns positive probability to action N after some history:

$$t_1 := \min\{t \in \mathbb{N} \mid \sigma_i^*(h_{i,t})(N) > 0 \text{ for some } i \in \{1,2\} \text{ and some } h_{i,t} \in \overline{H}_{i,t}\},\$$

with a convention that  $t_1 := \infty$  if there is no player *i*, time *t* and private history  $h_{i,t} \in \overline{H}_{i,t}$  such that  $\sigma_i^*(h_{i,t})(N) > 0$ . Since we have assumed  $\sigma^* \neq \sigma^G$ , Lemma 2 implies  $t_1 < \infty$ . Now we recursively define  $t_k$ . First, if  $t_{k-1} < \infty$ , let  $t_k$  be the *k*-th period at which some player *i*'s strategy  $\sigma_i^*$  assigns positive probability to action N after some history: For each  $k \in \mathbb{N}$  with  $k \geq 2$ ,

$$t_k := \min\{t \in \mathbb{N} \mid t > t_{k-1} \text{ and } \sigma_i^*(h_{i,t})(N) > 0 \text{ for some } i \in \{1,2\} \text{ and some } h_{i,t} \in \overline{H}_{i,t}\}$$

with the same convention as before that  $t_k := \infty$  if there is no player *i*, time *t* with  $t > t_{k-1}$  and her private history  $h_{i,t} \in \overline{H}_{i,t}$  such that  $\sigma_i^*(h_{i,t})(N) > 0$ . If  $t_{k-1} = \infty$ , then we let  $t_k = \infty$ .

We next show that  $t_k$ 's are consecutive.

**Lemma 3.** Fix i = 1, 2. If  $\delta_i > \hat{\delta}_i$ , then for all  $k \in \mathbb{N}$  such that i moves at  $t_k, t_k < \infty$  implies  $t_{k+1} = t_k + 1$ .

We also show that, for any time t and any private histories in  $\overline{H}_{i,t}$ , the moving player i has the same continuation payoff.

**Lemma 4.** For any i = 1, 2, t such that i moves, and  $h_{i,t}, h'_{i,t} \in \overline{H}_{i,t}$ , we have  $\pi_i(\sigma^*|h_{i,t}) = \pi_i(\sigma^*|h'_{i,t}).$ 

Hence, at period t that is no earlier than  $t_1$ , the moving player i's continuation

payoff given any  $h_{i,t} \in \overline{H}_{i,t}$  must be equal to the payoff from her playing action N, which is  $(1 - \delta_i)x_i$ .

The next lemma provides a condition that is useful in what follows.

**Lemma 5.** We have  $\overline{\delta}_i \geq \hat{\delta}_i$  for each i = 1, 2.

We next show that at no  $t_k$ , the moving player *i* plays a pure action.

**Lemma 6.** Fix i = 1, 2. Suppose  $\delta_i > \hat{\delta}_i$  and  $\delta_{-i} \in (0, \underline{\delta}_{-i}) \cup (\overline{\delta}_{-i}, 1)$ . Then, for any  $k \in \mathbb{N}$  such that i moves at  $t_k$ ,  $\sigma_i^*(h_{i,t_k})(N) < 1$  holds for some  $h_{i,t_k} \in \overline{H}_{i,t_k}$ .

The intuition behind this result is as follows. If *i* assigned probability 0 to action O at some  $t_k$  when she had privately learned N, then conditional on *i*'s playing O, -i at the next period would become sure that *i* has not learned action N yet. This would make it the unique best response for -i to play O in order to make action N a "secret," instead of playing N which teaches player *i* the existence of action N. This is a contradiction to our earlier conclusion in Lemma 3 which would imply that N must be a best response for -i at period  $t_k + 1$ . Condition  $\delta_i > \hat{\delta}_i$  ensures that the one-period gain from playing O is not worthwhile for *i* at  $t_k$ , and condition  $\delta_{-i} \in (0, \underline{\delta}_{-i}) \cup (\overline{\delta}_{-i}, 1)$  ensures that -i is either (i) impatient enough so that the impact of the one-period loss from playing N is too large  $(0 < \underline{\delta}_{-i}$  holds only when  $x_{-i} < 1$  or (ii) patient enough so that sticking to O is strictly better than playing N to obtain a one-period gain in the payoff.

For any time t at which i moves, for any private history  $h_{i,t} \in \overline{H}_{i,t}$ , let  $\mu(h_{i,t})$  be the posterior probability that i at  $h_{i,t}$  assigns to the event that -i knows action N at the next period that -i plays (that is, after the private-learning stage in the next period). Notice that  $\mu(h_{i,t}) = \mu(h'_{i,t})$  must hold for any  $h_{i,t}, h'_{i,t} \in \overline{H}_{i,t}$  since private learning happens independently across players and  $\mu(\cdot)$  is derived from Bayes rule because, under  $\sigma^*$ , the event under which player -i chooses action O at every period in the past is assigned a strictly positive probability (as it is possible that -i has never privately learned action N at any past periods). So, simply denote  $\mu_t := \mu(h_{i,t})$ for any  $h_{i,t} \in \overline{H}_{i,t}$ .

**Lemma 7.** Suppose  $p_j \in [0,1)$  for each j = 1,2 and suppose  $\delta_i > \hat{\delta}_i$  and  $\delta_{-i} \in (0, \underline{\delta}_{-i}) \cup (\overline{\delta}_{-i}, 1)$ . Then, there exist j = 1, 2 and  $\alpha_j^* > 1$  such that, for any  $t \ge t_1$  such that j moves, we have  $1 - \mu_{t+2} = \alpha_j^*(1 - \mu_t)$ .

The intuition behind this lemma is as follows. Take any  $t \ge t_1$  and let the moving player be *i*. First, Lemma 3 shows that *N* is a best response for *i* at some history in  $\overline{H}_{i,t}$ , where we use Lemma 5 to show the condition necessary for Lemma 3, i.e.,  $\delta_{-i} > \hat{\delta}_{-i}$ , is satisfied. Also, Lemma 6 implies that *O* is a best response for *i* at some history in  $\overline{H}_{i,t}$ . Hence, Lemma 4 implies that both *O* and *N* are a best response for *i* at every history in  $\overline{H}_{i,t}$ .

Now, the fact that *i* is indifferent at every history in  $H_{i,t}$  at every  $t_k$  has implications on the evolution of the belief  $\mu_t$ . To see this, first note that since *i* is indifferent at  $t_k$ , her payoff from playing *O* at  $t_k$  is  $(1 - \delta_i)x_i$ , which is the payoff she gets when she plays *N*. Suppose *i* plays *O* at  $t_k$ . If -i plays *N* at  $t_k + 1$ , then *i* receives a payoff of  $(1 - \delta_i)(-\ell_i)$  from period  $t_k + 1$  onward. The probability of this event is  $\mu_t r_{t+1}$ , where  $r_{t+1}$  is the probability that -i plays *N* at period  $t_k + 1$ , conditional on his having privately learned *N* at  $t_k + 1$ . If -i plays *O* at  $t_k + 1$ , then *i* receives a payoff of  $(1 - \delta_i)x_i$  from period  $t_k + 2$  onward by the indifference at  $t_k + 2$ . The probability of this event is  $1 - \mu_t r_{t+1}$ . Summarizing, by the indifference at period  $t_k$ , we have that  $(1 - \delta_i)x_i$  is equal to a convex combination of some constants, where the weights are given by  $\mu_t r_{t+1}$  and  $1 - \mu_t r_{t+1}$ . This implies that  $\mu_t r_{t+1}$  is a constant across all *t*. Note that  $\mu_t r_{t+1} < 1$  must hold if  $p_{-i} < 1$ .

Finally, when  $p_{-i} < 1$ , the Bayes rule implies

$$1 - \mu_{t+2} = \frac{(1 - p_{-i})(1 - \mu_t)}{1 - \mu_t r_{t+1}}$$

and since  $\mu_t r_{t+1}$  is a constant, we have  $1 - \mu_{t+2} = \alpha_i (1 - \mu_t)$ , where  $\alpha_i := \frac{1 - p_{-i}}{1 - \mu_t r_{t+1}}$  is a constant. With some more work, we can show  $\alpha_j > 1$  for some j = 1, 2.

Now, suppose that there is i = 1, 2 such that  $\delta_i > \hat{\delta}_i$  and  $\delta_{-i} > \overline{\delta}_{-i}(p_i)$ . Lemma 7 implies there is j = 1, 2 such that for any  $t \ge t_1, 1 - \mu_{t+2} = \alpha_j^*(1 - \mu_t)$  holds with  $\alpha_j^* > 1$  whenever j moves at t. Hence, there must exist  $\overline{t}$  such that  $\mu_{\overline{t}} < 0$ . But this cannot hold because  $\mu_t$  must be a probability for any t. This is a contradiction, and thus there cannot exist a (potentially mixed) PBE other than  $\sigma^{G}$ .

**Lemma 8.** There exists  $\overline{p} > 0$  such that  $\overline{\delta}_i(p_{-i}) < 1$  for all i = 1, 2 and  $p_{-i} \in (0, \overline{p})$ .

This lemma is straightforward because of the continuity of the function  $\pi_i$  that defines  $\overline{\delta}_i(p_{-i})$ : We have  $\pi_i(\delta_i) = 2 + x_i > x_i$  when  $\delta_i = 1$  and  $p_{-i} = 0$ , and  $\pi_i$  is continuous in  $\delta_i$  and  $p_{-i}$  at these values.

Now, take  $\overline{p}$  as in Lemma 8 and let  $\overline{\delta} = \max(\overline{\delta}_1, \overline{\delta}_2)$ . The proof of Theorem 1 is then complete.

## 4 Discussion

This section discusses several topics. First, Sections 4.1 and 4.2 discuss the models that violate irreversibility or asynchronicity and show that multiple equilibria exist. Section 4.3 discusses the relationship of our result with the one in Monderer and Samet (1989). Section 4.4 extends our model to the case with more than two actions, and Section 4.5 considers the case when players may not be able to learn the new action privately. Section 4.6 discusses the cases of low discount factors.

### 4.1 Reversible Case

Our main model assumes irreversibility of actions: Once player *i* chooses *N*, she will never be able to take action *O* again. This subsection argues that some type of irreversibility assumption is necessary for the conclusion of our main result to hold. To that end, consider the following model that is the same as our main model except that the action set of player *i* evolves in a different manner<sup>14</sup>: Player *i* initially knows *O* only, and privately learns *N* with probability  $p_i$  each period she moves if she has not learned it yet. Once she privately learns *N* or -i chooses *N*, *i*'s action set becomes  $\{O, N\}$  and will never change. That is, the action set includes *N* under the same condition as in our main model, while *O* is always in the action set. Call this model the *reversible model*. Notice that for any  $p_1, p_2 \in [0, 1]$ ,  $\sigma^{G}$  is a PBE in this model if  $\delta_1, \delta_2 < 1$  are sufficiently high. A stage game is a *pure coordination game* if  $x_i = -\ell_{-i}$  for each i = 1, 2. We can show the following:

**Proposition 1.** Suppose that the stage game is not a pure coordination game. For any  $p_1, p_2 \in (0, 1]$ , there exists  $\delta' < 1$  such that if  $\delta_i \in (\delta', 1)$  for each i = 1, 2, then the reversible model has a PBE in which action N is played with probability 1.

To show this, we construct an equilibrium strategy profile. In this equilibrium, the first player who privately learns N plays N. Let this first player be j. In the

<sup>&</sup>lt;sup>14</sup>We note that the same argument as in this subsection can be made for the case when the moves are synchronous, which is analyzed in the next subsection (Section 4.2).

continuation game after j has played N, an equilibrium  $\sigma^{(j)}$  will be played. The strategy profiles  $\sigma^{(1)}$  and  $\sigma^{(2)}$  are set up so that the implied equilibrium payoff profiles,  $(u_1^{(1)}, u_2^{(1)})$  and  $(u_1^{(2)}, u_2^{(2)})$ , satisfy  $u_1^{(1)} > u_1^{(2)} > 0$  and  $u_2^{(2)} > u_2^{(1)} > 0$ . Since the stage game is not a pure coordination game, such equilibria can be easily constructed when the discount factors are high by a standard argument of folk theorems on asynchronous moves (e.g., Dutta, 1995; Yoon, 2001). This concludes the proof of the theorem.<sup>15</sup>

The key is that, in the continuation game after some player j has chosen N, the players' flexibility to switch back and forth between the two actions—such a back-and-forth happens in  $\sigma^{(j)}$ —makes it possible to set up a "reward" for j and a punishment for -j, and this incentivizes players to choose N as soon as possible.

### 4.2 Synchronous Case

Our main model assumes asynchronous moves. To clarify the role of asynchronicity, this section considers a model with synchronous moves, and shows that there are multiple equilibria when the players are patient.

Formally, suppose that at each period t = 1, 2, ..., the two players simultaneously make a choice. As in the main model, each player *i* initially knows only *O*, but she privately learns *N* with probability  $p_i$  each period. Player *i* can choose action *N* if either she has privately learned it or the opponent has taken it. Once *i* takes action *N*, that will be the only choice for her thereafter.<sup>16</sup> Suppose also that at each period, a public randomization device is available.<sup>17</sup> Specifically, at the beginning of each period, a real number  $q_t$  is drawn (independently over time) according to the uniform distribution over [0, 1], and both players observe its realization. Call this model the synchronous model.

The set of private histories  $H_{i,t}$ , strategies, and PBE are defined analogously to Section 2 (with the standard measurability requirement). Consider the following strategy  $\sigma_i^{(T,\bar{q})}$ : At any  $h_{i,t} \in H_{i,t}$ , *i* plays *N* if and only if (i) some player has already taken *N*, or (ii) *i* has privately learned *N*,  $t = n \cdot T$  for some n = 1, 2, ..., and  $q_t \leq \bar{q}$ .

 $<sup>1^{15}</sup>$  If  $p_i > 0 = p_{-i}$ , then the conclusion of Proposition 1 holds if and only if  $x_i \ge 1$ . In this case, we set  $u_i^{(i)} \ge 1$  in the proof.

 $<sup>^{16}</sup>$ We note that the same argument as in this subsection can be made for the case when the action set evolves as in the previous subsection (Section 4.1).

<sup>&</sup>lt;sup>17</sup>The public randomization device can be anything, such as a sunspot, that the players can commonly observe but does not directly affect the payoffs. Our main theorem (Theorem 1) holds even when a public randomization device exists, which strengthens the uniqueness result.

**Proposition 2.** For any  $p_1, p_2 \in (0, 1]$  and  $\overline{q} \in [0, 1)$ , there exist  $\delta' \in (0, 1)$  and  $T' < \infty$  such that if  $\delta_i \in (\delta', 1)$  for each i = 1, 2 and T > T', then  $\sigma^{(T,\overline{q})}$  is a PBE in the synchronous model.

The intuition behind this result hinges on synchronous moves. For any  $p_{-i}$ , if we take T sufficiently high, then the posterior probability that the opponent has learned N is sufficiently high at period T under  $\sigma_i^{(T,\overline{q})}$ . Hence, when  $q_t \leq \overline{q}$ , i has an incentive to play N because she attaches probability close to 1 to the event that the opponent plays N as well. The opponent is in the same situation and hence plays N. This argument would break down if moves are asynchronous because -i would not switch his action when i plays, and thus i could take O to manipulate -i's belief in the next period, without receiving the payoff of  $-\ell_i$ .

When  $x_i > 1$  for some i = 1, 2, we use the public randomization device to incentivize players to play O at period T - 1: Given  $\overline{q} < 1$ , there is a positive probability that players continue playing O for at least additional T periods which is long, and thus players are incentivized to take O instead of receiving an instantaneous gain of  $x_i$ .

We note that public randomization is not necessary for multiple equilibria if  $x_i \leq 1$  for each *i*. Moreover, it is not necessary even if  $x_i > 1$  holds for some i = 1, 2 for a certain parameter range such that there is a unique PBE under the main model of asynchronous moves. Details of these points are investigated in the Online Supplementary Appendix.

Given that T' in Proposition 2 can be taken uniformly across  $\delta$ , the payoff from  $\sigma^{(T,\bar{q})}$  can be shown to approach 0 as  $\delta_i \to 1$  for any  $\bar{q} \in (0, 1)$ .

**Proposition 3.** For any  $p_1, p_2 \in (0,1]$  and  $\overline{q} \in (0,1)$ , there exist  $\delta' \in (0,1)$  and  $T' < \infty$  such that, for any sequence  $((\delta_i^k)_{i=1,2})_{k=1}^{\infty}$  such that  $\delta_i^k \in (\delta',1)$  for each i and k and  $\delta_i^k \to 1$  as  $k \to \infty$  for each i, the following hold:

1.  $\sigma^{(T,\bar{q})}$  is a PBE of the synchronous model with parameters  $(p_1, p_2, \bar{q}, \delta_1^k, \delta_2^k)$ ; and

2. Each i's expected payoff from  $\sigma^{(T,\overline{q})}$  converges to 0 as  $k \to \infty$ .

## 4.3 Almost Common Knowledge of the Unprecedented Action

The unique equilibrium  $\sigma^{G}$  has the feature that, for any fixed  $(p_{1}, p_{2})$  and  $\varepsilon > 0$ , there is  $T < \infty$  such that at period T, it is common  $(1 - \varepsilon)$ -belief among players that they have privately learned action N. Note that if  $x_{i} > 1$  for each i = 1, 2, and if common  $(1 - \varepsilon)$ -belief was replaced with common knowledge, then  $\sigma^{N}$  would also be an equilibrium in the continuation game starting at period T. One may wonder how this seeming discontinuity can be reconciled with the result of Monderer and Samet (1989), which provides a certain continuity property of the equilibrium set with respect to the degree of common beliefs. We assume  $x_{i} > 1$  for each i = 1, 2throughout this subsection because this is the case where the seeming discontinuity arises.

Recall that Monderer and Samet (1989) consider a game  $\Gamma$  in which each player obtains some information about which of the multiple possible games  $(\Gamma_1, \ldots, \Gamma_m)$  is the true game, where those games share a common action set for each player. They show that for any  $\varepsilon > 0$ , if it is common *p*-belief in  $\Gamma$  that players are playing game  $\Gamma_1$ with sufficiently high *p*, then for any Nash equilibrium *s* of  $\Gamma_1$ , "players play as in *s* when the true game is in fact  $\Gamma_1$ " is part of an  $\varepsilon$ -Nash equilibrium of the original game  $\Gamma$  where  $\varepsilon$ -best response is required conditional on the information each player has.<sup>18</sup> In our context, we could let  $\Gamma$  be the continuation game starting at period *T* with action profile (O, O), and imagine there are four possible games,  $\Gamma_1, \ldots, \Gamma_4$ : Both players have learned *N* in  $\Gamma_1$ , only player 1 has learned it in  $\Gamma_2$ , only 2 has learned it in  $\Gamma_3$ , and no one has learned it in  $\Gamma_4$ . Each player knows whether she herself has learned *N* or not. For any  $\varepsilon > 0$ , we can take *T* large enough so that it is common *p*-belief with high *p* that the game is  $\Gamma_1$ .<sup>19</sup> By the result of Monderer and Samet (1989), since  $\sigma^N$  is a Nash equilibrium of  $\Gamma_1$ , there must exist an  $\varepsilon$ -Nash equilibrium of the continuation game  $\Gamma$  in which each player plays *N* as soon as possible at  $\Gamma_1$ .

First, we note that Monderer and Samet (1989)'s result only implies  $\varepsilon$ -equilibrium, as opposed to the exact equilibrium. Thus, the fact that  $\sigma^N$  is not an exact equilibrium

<sup>&</sup>lt;sup>18</sup>This version of the statement corresponds to their result that requires "ex post"  $\varepsilon$ -equilibrium.

<sup>&</sup>lt;sup>19</sup>Precisely speaking, there are more states because a player can privately learn N at one of many periods in the past, but this variation is not essential to our discussion here. For this reason, this subsection (and only this subsection) assumes that the players' actions do not condition on the period in which they privately learned N.

in  $\Gamma$  is not a contradiction *per se*. We now discuss why " $\sigma^N$  at  $\Gamma_1$ " can only be part of an  $\varepsilon$ -equilibrium of  $\Gamma$  (it indeed is), and the nature of a deviation that a player would like to make.

To see where the approximation appears in the entire game  $\Gamma$ , note first that the moving player *i* at *T* is taking a best response given that the opponent plays  $\sigma_{-i}^{N}$ . This is because, if the opponent will take *N* with high probability in the next period, *i* wants to take *N* at the current period. A suboptimal action is taken, however, by -i at period T + 1 after *i* has played *O* at *T*. Although  $\sigma^{N}$  dictates that -i plays *N* even under such a history, playing *N* is suboptimal because -i has updated his belief to assign only a small probability to the event that *i* would have learned *N* at the next period. Since the event that *i* plays *O* at *T* happens with a very small probability under  $\sigma^{N}$  (it happens only when *i* has not learned *N* yet), what -i plans to do after such an event affects -i's expected payoff only minimally, and thus -i is still taking an  $\varepsilon$ -best response under  $\sigma^{N,20}$  Conditional on the event that *i* plays *O* at *T*, however,  $\sigma_{-i}^{N}$  is not only  $\varepsilon$  away from a best response at T + 1 but is substantially suboptimal.

The key is that there is an action that can change what is almost common knowledge: The event that players have learned N by T was almost common knowledge at T but, once player i takes O at T, it is no longer almost common knowledge. Monderer and Samet (1989)'s model, in contrast, does not allow for such an action because their model is static.

Fudenberg and Tirole (1991) prove two results that are worth noting here. First, they present a corollary of Monderer and Samet (1989), which says that if there is a strict Nash equilibrium in each  $\Gamma_k$ , then there is an exact (not  $\varepsilon$ -) equilibrium of the entire game  $\Gamma$  that plays *s* when the true game is in fact  $\Gamma_1$  (which is common *p*-belief). In our context, the hypothesis of this corollary does not apply because  $\sigma^N$ is not a strict equilibrium in  $\Gamma_1$ : Player -i assigns probability 0 to *i*'s playing *O* at *T*, and hence -i is indifferent among any actions under such a history at T + 1. Indeed, as we have explained, the equilibrium we construct is not exact because -iwas playing suboptimally under such a history.

Second, Fudenberg and Tirole (1991) prove that there is an (exact) Nash equilibrium of  $\Gamma$  that is close to playing a Nash equilibrium at each game  $\Gamma_k$ . Their proof

 $<sup>^{20}</sup>$ Note that we are still talking about the notion of *ex post* equilibrium here in the sense defined in Monderer and Samet (1989) because the game has already started.

uses the theorem by Wu and Jiang (1962) which assumes finite strategic-form games (as Fudenberg and Tirole (1991) has clarified in their statement). Since our strategy space is infinite, the theorem does not apply.

### 4.4 The Case with More Than Two Actions

In the main analysis, we assumed that each player has two actions in the stage game. In this section, we consider the case that allows for more than two actions.

Formally, each player *i*'s action set is  $\{a^1, \ldots, a^K\}$ . We suppose the following:

(A) For each  $k \in \{1, \ldots, K-1\}$ , an action profile  $(a^k, a^k)$  Pareto dominates  $(a^{k+1}, a^{k+1})$ .

(B) 
$$u_i(a^l, a^k) < u_i(a^k, a^k)$$
 for  $l < k$ .

Note that, by (B), action profile  $(a^K, a^K)$  is a Nash equilibrium of the stage game.

The initial action set is  $\{a^1\}$  for each i = 1, 2. Once player i chooses  $a^k$ , then she cannot take actions in  $\{a^1, \ldots, a^{k-1}\}$ . For each  $k, k' = 1, \ldots, K$  with k < k', there is  $p_{k,k'}^i \in [0,1]$ . In each period player i moves, when her current action set is  $\{a^l, \ldots, a^k\}$  (where  $1 \le l \le k \le K$ ) with k < k', (i) i privately learns  $a^{k'}$  with probability  $p_{k,k'}^i$  and (ii) she learns  $a^{k'}$  if her opponent's latest action is  $a^{k'}$ . Under any one of these events, her action set becomes  $\{a^l, \ldots, a^{k'}\}$ . In other words, learning  $a^{k'}$  makes all actions  $a^m$  with k < m < k' available as well. We call this model the ordered-action model.

The set of private histories  $H_{i,t}$ , strategies, and PBE are defined analogously to Section 2. We extend the notion of grim-trigger strategy profile  $\sigma^{G}$  as follows: At a history  $h_{i,t} \in H_{i,t}$  at which each player j's latest action is  $a^{k_j}$ , player i plays  $\sigma_i^{G}(h_{i,t}) = a^l$  where  $l = \max\{k_1, k_2\}$ . Note that, in the two-action case, this reduces to the grim-trigger strategy profile  $\sigma^{G}$  that we defined in Section 3.

**Theorem 2.** Consider the ordered-action model. There exist  $\overline{p} \in (0, 1)$  and  $\overline{\delta} \in (0, 1)$ such that for any  $(p_{k,k'}^i)_{i,k,k'}$  with  $p_{k,k'}^i < \overline{p}$  for each k, k' such that k < k' and each i = 1, 2 and  $\delta_i > \overline{\delta}$  for each  $i = 1, 2, \sigma^G$  is a unique PBE.

*Proof Sketch of Theorem 2.* The proof of this theorem can be done by induction. To do this, consider the following claim:

**Claim** k. There exists  $\hat{\delta}^{(k)} \in (0,1)$  such that for all  $\delta_1, \delta_2 > \hat{\delta}^{(k)}$  and all  $k_1, k_2 \ge k$ , at a history  $h_{i,t} \in H_{i,t}$  at which the latest action profile is  $(a^{k_1}, a^{k_2})$ , player i plays  $a^l$ where  $l = \max(k_1, k_2)$  under any PBE of the ordered-action model. In Appendix A.2.3, we show that for any  $k = 1, \dots, K-1$ , Claim k holds if Claim l holds for every  $l = k + 1, \dots, K$ .

By induction, this shows that Claim k holds for every k.

To see that the assumptions on payoffs and the evolution of action sets are essential, the Supplementary Appendix provides two examples in which the assumptions are violated and the conclusion of Theorem 2 does not hold. The first example features a violation of the assumption on payoffs and has two Pareto-unranked Nash equilibria in the stage game without the original action  $(a^1)$ . We show that there is an equilibrium in which each player wants to play a certain action other than  $a^1$ as soon as possible to settle into the Nash equilibrium that she prefers before the opponent does so. The game in the second example violates the assumption on the evolution of action sets and multiple actions are available after an action other than  $a^1$  is taken. In such a case, there are multiple equilibria in the "subgame" after an action other than  $a^1$  is taken, just as in the reversible case in Section 4.1. This enables us to construct an equilibrium in which each player wants to play a certain action other than  $a^1$  as soon as possible to obtain a better payoff in the subgame.

### 4.5 Possibility of No Private Learning of a Precedent

In our main analysis in Section 3, each player privately learns action N with positive probability p in each period (if she has not learned it yet). In reality, it may be the case that, with certain probability, no player may be able to learn the existence of action N at all. For another instance, with certain probability for each player, the only way in which a player learns action N is through the observation of the opponent's action.

This subsection shows that the conclusion of our main result (Theorem 1) is robust to such possibilities. Formally, we consider the following two variants of our main model. In the first model, there are two states, which realize with positive probability before the start of the game. In the first state, action N does not exist. That is, no player privately learns action N. In the second state, action N exists and the players learn N as in our main model. That is, whether there is a possibility of private learning is common across players. Call this model the *common-possibility model*. In this model, the characterization of the players' incentives is the same as our main model. This is because, in order to consider a player's incentive to take action N, we only need to consider histories at which the player has learned N and both players have been taking O. Conditional on such histories, the player knows that action Nexists, and our analysis reduces to the main one in Section 3. Thus, our main result (Theorem 1) carries over to this model. In fact, our full analysis presented in Section 4.6 also carries over to this extension.

In the second model, for each player, there exists a probability-q event (which occurs independently across the players) under which the player cannot privately learn action N on her own (i.e., she learns N only when the opponent plays it). That is, whether there is a possibility of private learning is independently determined. Call this model the *independent-possibility model*. We argue that, in this second model, there exist  $\overline{\delta}_1, \overline{\delta}_2 < 1$  such that if  $\delta_i > \overline{\delta}_i$  for each i = 1, 2, then for any  $p_1, p_2 \in [0, 1]$ ,  $\sigma^{\rm G}$  is a unique PBE. In order to consider player *i*'s incentive to take action N, it is sufficient to consider histories at which player *i* has learned N and both players have been taking O. Suppose that player *i*'s strategy assigns positive probability to playing N at such a history at some finite t. Then, *i*'s continuation payoff from such a strategy is  $(1 - \delta_i)x_i$ . In contrast, if she obeys strategy  $\sigma_i^{\rm G}$ , then her continuation payoff is at least

$$q \cdot 1 + (1 - q) \left( (1 - \delta_i) \cdot 1 + \delta_i (1 - \delta_i) (-\ell_i) \right).$$

As  $\delta_i$  converges to 1, the former and the latter converge to 0 and q > 0, respectively. Thus, there exists  $\overline{\delta}_i \in (0,1)$  such that if  $\delta_i > \overline{\delta}_i$  then  $\sigma_i^{G}$  is a unique equilibrium strategy for player *i*.

We summarize the findings of this section as follows.

**Proposition 4.** In both the common-possibility model and the independent-possibility model, the conclusion of Theorem 1 holds.

### 4.6 $2 \times 2$ Full Analysis

Our main theorem pertains to the case when  $p_i$  is small and  $\delta_i$  is high. For  $2 \times 2$  games, we provide an equilibrium characterization for any  $p_i$  and  $\delta_i$ . The following lemma provides basic properties of best responses.

**Lemma 9.** Fix  $p_1, p_2 \in [0, 1]$ . For each i = 1, 2, the following hold.

1. The strategy  $\sigma_i^N$  is a unique best response against  $\sigma_{-i}^N$  if and only if  $x_i > \pi_i(\delta_i)$ .

		$\hat{\delta}_2 > \delta_2$	$\hat{\delta}_2 < \delta_2$	
		$x_2 > \pi_2(\delta_2)$		$x_2 < \pi_2(\delta_2)$
$\delta_1 > \hat{\delta}_1$	$\begin{array}{l} x_1 < \\ \pi_1(\delta_1) \end{array}$	Unique $(\sigma_1^{\mathrm{G}}, \sigma_2^N)$ (Part 2)		Unique $\sigma^{\rm G}$ (Part 1)
	$x_1 >$		$\sigma^{\mathrm{G}} \text{ and } \sigma^{N}$ (Part 3)	
$\delta_1 < \hat{\delta}_1$	$\pi_1(\delta_1)$	Unique $\sigma^N$ (Part 4)		Unique $(\sigma_1^N, \sigma_2^G)$ (Part 2)

Figure 1: Illustration of Theorem 3. For example, "Unique  $\sigma^{G}$ " means that  $\sigma^{G}$  is a unique PBE when the parameters satisfy the coditions in the corresponding region. Also, " $\sigma^{G}$  and  $\sigma^{N}$ " means that both  $\sigma^{G}$  and  $\sigma^{N}$  are a PBE when the parameters satisfy the coditions in the corresponding region.

- 2. The strategy  $\sigma_i^{G}$  is a unique best response against  $\sigma_{-i}^{N}$  if and only if  $x_i < \pi_i(\delta_i)$ .
- 3. If  $\delta_i \in (0, \hat{\delta}_i)$ , then player *i*'s strategy must be  $\sigma_i^N$  in any PBE.

Using this lemma, we obtain the following theorem.

**Theorem 3.** Fix  $p_1, p_2 \in [0, 1]$ . Let  $\delta_1, \delta_2 \in (0, 1)$ .

- 1. If  $\delta_i > \hat{\delta}_i$  and  $x_{-i} < \pi_{-i}(\delta_{-i})$ , then  $\sigma^{\rm G}$  is a unique PBE.
- 2. If  $\delta_i < \hat{\delta}_i$  and  $x_{-i} < \pi_{-i}(\delta_{-i})$ , then  $(\sigma_i^N, \sigma_{-i}^G)$  is a unique PBE.
- 3. If  $\delta_j > \hat{\delta}_j$  and  $x_j > \pi_j(\delta_j)$  for each j = 1, 2, then both  $\sigma^{\rm G}$  and  $\sigma^{\rm N}$  are a PBE.
- 4. If  $\delta_i < \hat{\delta}_i$  and  $x_{-i} > \pi_{-i}(\delta_{-i})$ , then  $\sigma^N$  is a unique PBE.

Figure 1 illustrates Theorem 3. The Online Supplementary Appendix uses Theorem 3 to provide an alternative characterization of PBE. Here, we provide graphical illustrations of this characterization. Figure 2 illustrates PBE for different values of  $(\delta_1, \delta_2)$ . We note that, if  $p_{-j}$  is small, then  $\overline{\delta}_j(p_{-j}) < 1$  in the left and right panels and  $\underline{\delta}_j(p_{-j}) = 1$  in the central panel. Thus,  $\sigma^G$  is indeed a unique PBE when the discount factors are high as shown in Theorem 1. For the symmetric cases, Figure 3 illustrates the characterization for different values of  $(p, \delta)$ . These figures, in particular, demonstrate that there may exist multiple equilibria when private learning is likely or the players are impatient. The Online Supplementary Appendix provides additional graphical illustrations to cover the cases of asymmetric parameter values.



Figure 2: The alternative characterization of PBE. For example, "Unique  $\sigma^{G}$ " means that  $\sigma^{G}$  is a unique PBE in the interior of the corresponding region. Also, " $\sigma^{G}$ and  $\sigma^{N}$ " means that both  $\sigma^{G}$  and  $\sigma^{N}$  are a PBE when the parameters satisfy the coditions in the corresponding region. *Left*: The case with  $x_i > 1$  for each i = 1, 2. *Center*: The case with  $x_i < 1$  for each i = 1, 2. The graph depicts the case in which  $\underline{\delta}_i(p_{-i}) \leq \overline{\delta}_i(p_{-i})$  (otherwise,  $\underline{\delta}_i(p_{-i}) = 1$  and  $\overline{\delta}_i(p_{-i}) = 0$ , and thus  $\sigma^{G}$  is a unique PBE in the entire region. This happens when  $p_{-i}$  is low.) for each i = 1, 2. In this case,  $\overline{\delta}_i(p_{-i}) = 1$  for each i. *Right*: The case with  $x_i > 1 > x_{-i}$ . The graph depicts the case in which  $\underline{\delta}_{-i}(p_i) \leq \overline{\delta}_{-i}(p_i)$  (otherwise,  $\underline{\delta}_{-i}(p_i) = 1$  and  $\overline{\delta}_{-i}(p_i) = 0$ , and thus only  $\sigma^{G}_{-i}$  is played in the entire region).



Figure 3: The alternative characterization of PBE for the symmetric cases  $(p := p_1 = p_2, \delta := \delta_1 = \delta_2, x := x_1 = x_2$ , and  $\ell := \ell_1 = \ell_2$ ). For example, "Unique  $\sigma^{G}$ " means that  $\sigma^G$  is a unique PBE in the interior of the corresponding region. Also, " $\sigma^G$  and  $\sigma^N$ " means that both  $\sigma^G$  and  $\sigma^N$  are a PBE when the parameters satisfy the coditions in the corresponding region. Left: The case with x > 1. Right: The case with x < 1. We let  $\overline{\delta}(p) = \overline{\delta}_1(p) = \overline{\delta}_2(p), \ \underline{\delta}(p) = \underline{\delta}_1(p) = \underline{\delta}_2(p)$ , and  $\hat{\delta} = \hat{\delta}_1 = \hat{\delta}_2$ .

## 5 Conclusion

This paper studied a dynamic game in which players are not able to take certain actions without a precedent, i.e., unless some other player has already taken them. For  $2 \times 2$  stage games in which the new actions constitute a Nash equilibrium that is Pareto dominated by the initial default action profile, we showed that, in the perfect Bayesian equilibrium which we show is unique, each player takes the default action as long as the opponent has only taken it, when each player is patient and privately learns the new action with low probability.

Our model is simple and provides various avenues for future research. As the first paper to consider the effect of precedents in a dynamic interaction, however, those extensions of interest were beyond its scope, and we only suggest some possibilities here. First, it may be interesting to consider extensions of the no-private learning models as in Section 4.5 to an environment with a larger action set. In such extensions, the probability of no private learning for an action could depend on the action itself or the current action set (recall the specification of probability  $p_{k,k'}^i$  in Section 4.4). This probability would affect players' incentives to take a new action, and would complicate the analysis. Second, we showed that the new action keeps being unprecedented even though it becomes almost common knowledge among the players in our model. One can consider a different type of almost common knowledge à la Rubinstein (1989): The players are either informed or uninformed in the initial period, and messages are sent back and forth between them every period with some failure. It would be interesting to see if the players take the new action in an equilibrium of such a model. Third, one can consider a situation in which a player can only imperfectly monitor the new action taken by the opponent. The new action could generate signals and a certain combination of those signals would enable the player to take the new action. Fourth, as a special case of imperfect monitoring, we find it worthwhile to investigate the role of a precedent in a community enforcement setting. While the contagious (cooperative) equilibrium would survive as in Kandori (1992) when the feasibility of the new action is as in our model, it is of interest to study whether it would be a unique equilibrium. We expect the insights obtained in this paper to be useful in studying those extensions.

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## A Proofs

### A.1 Proofs for Section 3

Proof of Lemma 1. Under  $\sigma^{G}$ , suppose that player *i* moves at time *t* and has not played *N*.

Suppose first that her opponent -i has already played N. Then, if *i* chooses O at time t, her payoff is at most  $(1 - \delta_i) \cdot (-\ell_i) + \delta_i \cdot 0 < 0$ . If *i* instead chooses N at time t, her payoff is 0. Hence, it is a best response to choose N at time t.

Next, suppose that -i has not played N yet. Then, if i chooses O at time t, her payoff is 1. If i instead chooses N at time t, her payoff is  $(1-\delta_i)\cdot x_i + \delta_i \cdot 0 = (1-\delta_i)x_i$ . This is no greater than 1 if  $(1-\delta_i)x_i \leq 1$ , or  $\hat{\delta}_i \leq \delta_i$ . Hence, it is a best response to choose O at time t. This completes the proof.

Proof of Lemma 2. Fix  $\sigma$  that satisfies the condition stated in the lemma. Then, suppose that *i* moves at time *t* given some history in which -i has already played action *N*. Then, the same proof as in the proof of Lemma 1 shows that it is the unique best response to choose action *N* at time *t*. Hence, action *N* is chosen under  $\sigma$  whenever the opponent has chosen action *N* in the past. Since  $\sigma$  coincides with  $\sigma^{\rm G}$  in histories where the opponent has not chosen action *N*, this completes the proof.  $\Box$ 

Proof of Lemma 3. Fix  $\delta_i > \hat{\delta}_i$ . Suppose to the contrary that there exists k such that  $t_k + 1 < t_{k+1}$ , where i is the moving player at time  $t_k$ . Fix an arbitrary private history  $h_{i,t_k} \in \overline{H}_{i,t_k}$  at time  $t_k$ . Player i's payoff from playing action N at  $h_{i,t_k}$  is

$$(1-\delta_i)x_i + \delta_i \cdot 0 = (1-\delta_i)x_i.$$

On the other hand, if she plays action O at  $h_{i,t_k}$  and then plays action N at time  $t_k + 2$ , her payoff is

$$(1 - \delta_i) \cdot 1 + \delta_i((1 - \delta_i) \cdot 1 + \delta_i(1 - \delta_i)x_i) = (1 - \delta_i^2) \cdot 1 + \delta_i^2(1 - \delta_i)x_i$$

because player -i will not play action N at time  $t_k + 1$  by our supposition. Note that

$$(1 - \delta_i)x_i < (1 - \delta_i^2) \cdot 1 + \delta_i^2(1 - \delta_i)x_i$$

because  $\delta_i > \hat{\delta}_i$  implies  $(1 - \delta_i)x_i < 1$ . Hence, player *i* assigns probability 0 to action N at  $h_{i,t_k}$ . Since  $h_{i,t_k}$  was an arbitrary history in  $\overline{H}_{i,t_k}$ , this contradicts the definition of  $t_k$ . Thus, we have shown that  $t_{k+1} = t_k + 1$  for every  $k \in \mathbb{N}$  such that *i* moves at  $t_k$ .

Proof of Lemma 4. We first provide two definitions. First, for any  $h_{i,t} \in \overline{H}_{i,t}$  and  $h'_{i,t'} \in \overline{H}_{i,t'}$ , write  $h_{i,t} \sqsubseteq h'_{i,t'}$  if  $t \le t'$ ,  $s(h_{i,t}) = s(h'_{i,t'}) \le t$ , and the action profiles in  $h_{i,t}$  appear as the first t elements of  $h'_{i,t'}$  in the same order.

Second, for any  $h_{i,t} \in \overline{H}_{i,t}$  and  $h'_{i,t'} \in \overline{H}_{i,t'}$  such that  $t \leq t'$ , write  $h_{i,t} \circ h'_{i,t'}$  for a history such that  $s(h_{i,t} \circ h'_{i,t'}) = s(h_{i,t})$ , the action profiles in  $h_{i,t}$  appear as the first t elements of  $h_{i,t} \circ h'_{i,t'}$  in the same order and the last t' - t action profiles in  $h'_{i,t'}$  appear as the next t' - t elements of  $h_{i,t} \circ h'_{i,t'}$  in the same order.

With these definitions in hand, take  $h_{i,t}, h'_{i,t} \in \overline{H}_{i,t}$ . Suppose that  $\pi_i(\sigma^*|h_{i,t}) > \pi_i(\sigma^*|h'_{i,t})$ . Then, consider a strategy  $\sigma'_i$  such that, for  $t' \ge t$ ,

$$\sigma'_{i}(\tilde{h}_{i,t'}) = \begin{cases} \sigma^{*}_{i}(h_{i,t} \circ \tilde{h}_{i,t'}) & \text{if } h'_{i,t} \sqsubseteq \tilde{h}_{i,t'} \\ \sigma^{*}_{i}(\tilde{h}_{i,t'}) & \text{otherwise} \end{cases}$$

Then, at history  $h'_{i,t}$ , player *i* has a profitable deviation to play  $\sigma'_i$ . This contradicts our assumption that  $\sigma^*$  is a PBE (so  $\sigma^*_i$  must be a best response to  $\sigma^*_{-i}$ ). The case when  $\pi_i(\sigma^*|h_{i,t}) < \pi_i(\sigma^*|h'_{i,t})$  is symmetric. Thus, we must have  $\pi_i(\sigma^*|h_{i,t}) = \pi_i(\sigma^*|h'_{i,t})$ .  $\Box$ 

Proof of Lemma 5. Fix i = 1, 2. First, suppose  $x_i \leq 1$ . Since  $\hat{\delta}_i = 0$ , we have  $\overline{\delta}_i \geq \hat{\delta}_i$ . Next, consider the case in which  $x_i > 1$ . In this case,  $\hat{\delta}_i = \frac{x_i - 1}{x_i} < 1$  is the positive solution of

$$x_i = 1 + \delta_i \cdot 1 + \delta_i^2 x_i. \tag{1}$$

Notice that the left-hand side is constant in  $\delta_i$ , while the right-hand side is 1 at  $\delta_i = 0$ and is increasing in  $\delta_i \ge 0$ .

If  $\overline{\delta}_i = 1$ , then we are done. Suppose  $\overline{\delta}_i < 1$ . Recall

$$\pi_i(\delta_i) = p_{-i}[1 + \delta_i(-\ell_i) + \delta_i^2 \cdot 0] + (1 - p_{-i})[1 + \delta_i \cdot 1 + \delta_i^2 x_i].$$

If  $p_{-i} = 1$ , then  $\pi_i(0) = 1$  and  $\pi_i(\delta_i)$  is decreasing in  $\delta_i$ . Thus, by the definition of  $\overline{\delta}_i$ , we have  $\overline{\delta}_i = 1$ . Hence, we have  $p_{-i} < 1$ . Then,  $\pi_i(\delta_i)$  is quadratic in  $\delta_i$  with a positive coefficient, so it is convex in  $\delta_i$  and diverges to infinity as  $\delta_i$  goes to infinity. Thus, by the definition of  $\overline{\delta}_i$ ,  $\overline{\delta}_i$  is the unique positive solution of  $x_i = \pi_i(\delta_i)$ .

Finally, the right-hand side of equation (1) is no less than  $\pi_i(\delta_i)$  because  $-\ell_i < 1$ and  $0 < x_i$ .

Combining these facts, we obtain  $\overline{\delta}_i \geq \hat{\delta}_i$  when  $x_i > 1$ . In sum, we obtain  $\overline{\delta}_i \geq \hat{\delta}_i$  for all  $x_i$ .

Proof of Lemma 6. Fix  $\delta_i > \hat{\delta}_i$  and  $\delta_{-i} \in (0, \underline{\delta}_{-i}) \cup (\overline{\delta}_{-i}, 1)$ . Suppose, to the contrary, that there exists  $k \in \mathbb{N}$  such that *i* moves at  $t_k$  and assigns probability one to action N under every history in  $\overline{H}_{i,t_k}$ . At  $t_k + 1$ , consider any history  $h_{-i,t_k+1} \in \overline{H}_{-i,t_k+1}$ . If the moving player -i takes action N, her continuation payoff is  $(1 - \delta_{-i})x_{-i}$ . If,

instead, player -i takes action O at  $h_{-i,t_k+1}$ , then her continuation payoff is at least<sup>21</sup>

$$(1 - \delta_{-i}) \cdot 1 + \delta_{-i} \left( p_i ((1 - \delta_{-i})(-\ell_{-i}) + \delta_{-i} \cdot 0) + (1 - p_i)((1 - \delta_{-i}) \cdot 1 + \delta_{-i}(1 - \delta_{-i})x_{-i}) \right)$$

By dividing both expressions by  $1 - \delta_{-i}$ , we have that the former being strictly smaller than the latter if and only if

$$x_{-i} < \pi_{-i}(\delta_{-i}) = p_i [1 + \delta_{-i}(-\ell_{-i}) + \delta_{-i}^2 \cdot 0] + (1 - p_i) [1 + \delta_{-i} \cdot 1 + \delta_{-i}^2 x_{-i}].$$
(2)

The left-hand side and the right-hand side of (2), respectively, are the left-hand side and the right-hand side of the inequality defining  $(\underline{\delta}_{-i}, \overline{\delta}_{-i})$ . Thus, -i's continuation payoff from taking N is strictly smaller than the one from taking O at  $h_{-i,t_k+1}$  if and only if  $\delta_{-i} \in (0, \underline{\delta}_{-i}) \cup (\overline{\delta}_{-i}, 1)$ . Hence, at  $h_{-i,t_k+1}$ , player -i assigns probability 0 to action N. Since the choice of  $h_{-i,t_k+1}$  was arbitrary from  $\overline{H}_{-i,t_k+1}$ , this implies that  $t_{k+1} > t_k + 1$ . But this contradicts Lemma 3 whose conclusion (i.e.,  $t_{k+1} = t_k + 1$ ) must hold as  $\delta_i > \hat{\delta}_i$ . Therefore, we conclude that there exists no  $k \in \mathbb{N}$  such that the moving player i at time  $t_k$  assigns probability one to action N under some history in  $\overline{H}_{i,t_k}$ .

Proof of Lemma 7. We start with showing that any  $t \ge t_1$  satisfies  $t = t_k$  for some  $k \in \mathbb{N}$ . To see this, first,  $\delta_i > \hat{\delta}_i$  by assumption. Second, if  $\underline{\delta}_{-i} > 0$ , then  $x_{-i} \le 1$ . This is because if  $x_{-i} > 1$ , then  $x_{-i} > \pi_{-i}(0)$  because  $\pi_{-i}(0) = 1$ , and thus  $\underline{\delta}_{-i} = 0$ . Since  $x_{-i} \le 1$  implies  $\hat{\delta}_{-i} = 0$ , this second conclusion shows that  $\underline{\delta}_{-i} > 0$  implies that  $\hat{\delta}_{-i} = 0$ . Third, Lemma 5 implies  $\overline{\delta}_{-i} \ge \hat{\delta}_{-i}$ . Hence, we have  $\delta_{-i} > \hat{\delta}_{-i}$ , which enables us to invoke Lemma 3. The lemma implies that any  $t \ge t_1$  satisfies  $t = t_k$  for some  $k \in \mathbb{N}$ .

Now, for each  $t_k$  such that i moves, by definition, there is a history  $h'_{i,t_k} \in \overline{H}_{i,t_k}$ such that N is i's best response at  $t_k$ . Also, Lemma 6 implies that there is a history  $h''_{i,t_k} \in \overline{H}_{i,t_k}$  such that O is i's best response at  $t_k$ . Thus, Lemma 4 implies that at any history  $h_{i,t_k} \in \overline{H}_{i,t_k}$ , both O and N are best responses, and thus in particular they induce the same continuation payoff. Then, since we have shown that any  $t \ge t_1$ satisfies  $t = t_k$  for some  $k \in \mathbb{N}$ , it follows that, at any  $t \ge t_1$ , for any history  $h_{i,t} \in \overline{H}_{i,t}$ 

<sup>&</sup>lt;sup>21</sup>If player *i* takes action *N* at  $t_k + 2$  (after learning action *N*), player -i's continuation payoff from time  $t_k + 2$  on is  $(1 - \delta_{-i})(-\ell_{-i}) + \delta_{-i} \cdot 0 = (1 - \delta_{-i})(-\ell_{-i})$ . If instead player *i* takes action *O* (even after learning action *N*) then player -i can secure a continuation payoff of  $(1 - \delta_{-i})(-\ell_{-i}) + \delta_{-i} \cdot 0 = 1 - \delta_{-i}$ . Since  $(1 - \delta_{-i})(-\ell_{-i}) \leq 1 - \delta_{-i}$ , player -i can secure a payoff of  $(1 - \delta_{-i})(-\ell_{-i}) + \delta_{-i} \cdot 0$ .

of the moving player i, actions O and N induce the same continuation payoff.

Letting the conditional probability that  $\sigma_{-i}^*$  assigns to action N at time t + 1 be  $r_{t+1}$  (conditional on  $h_{i,t}$  and *i*'s taking O at time t),<sup>22</sup> *i*'s indifference condition at time t implies

$$(1 - \delta_i)x_i = (1 - \delta_i) \cdot 1 + \delta_i [\mu_t r_{t+1}((1 - \delta_i)(-\ell_i) + \delta_i \cdot 0) + (1 - \mu_t r_{t+1})((1 - \delta_i) \cdot 1 + \delta_i (1 - \delta_i) x_i)].$$
(3)

Note that the left-hand side does not depend on t. The right-hand side depends on t only through  $\mu_t r_{t+1}$ . Thus,  $\mu_t r_{t+1}$  does not depend on t.

Now, notice that, conditional on history  $h_{i,t}$  and i playing O at time t, the probability that the history is in  $\overline{H}_{i,t+2}$  at time t+2 is  $1-\mu_t r_{t+1}$  because  $\mu_t r_{t+1}$  is the probability that -i will take action N at time t+1. Also, under the same conditioning, the probability that -i does not know action N at time t+3 is  $(1-p_{-i})(1-\mu_t)$  because -i does not know N with probability  $1-\mu_t$  at t+1 (when he chooses an action), and he does not learn action N at time t+3 with probability  $1-p_{-i}$ . Hence, by the Bayes rule, we have

$$1 - \mu_{t+2} = \frac{(1 - p_{-i})(1 - \mu_t)}{1 - \mu_t r_{t+1}} = \alpha_i (1 - \mu_t),$$

where  $\alpha_i := \frac{1-p_{-i}}{1-\mu_t r_{t+1}}$  is a well-defined constant. It is well defined because  $p_{-i} < 1$  implies  $\mu_t < 1$ , and it is a constant, i.e., it does not depend on t, because  $\mu_t r_{t+1}$  does not depend on t.

Now, define

$$f_i(z) := (1 - \delta_i) \cdot 1 + \delta_i \left( z((1 - \delta_i)(-\ell_i) + \delta_i \cdot 0) + (1 - z)((1 - \delta_i) \cdot 1 + \delta_i(1 - \delta_i)x_i) \right).$$

Note that  $f_i$  is strictly decreasing in z because

$$(1 - \delta_i)(-\ell_i) + \delta_i \cdot 0 < 0 < (1 - \delta_i) \cdot 1 + \delta_i(1 - \delta_i)x_i.$$
$$r_{t+1} = \sum_{\substack{h_{-i,t+1} \in \overline{H}_{-i,t+1}}} \operatorname{Prob}(h_{-i,t+1}|h_{i,t})\sigma_{-i}^*(h_{-i,t+1})(N).$$

 $^{22}\mathrm{That}$  is,

We note that 
$$\operatorname{Prob}(h_{-i,t+1}|h_{i,t}) = \operatorname{Prob}(h_{-i,t+1}|h'_{i,t})$$
 for any  $h_{i,t}, h'_{i,t} \in \overline{H}_{i,t}$  and  $h_{-i,t+1} \in \overline{H}_{-i,t+1}$  for a reason analogous to the one presented in the first paragraph of the current proof, and thus  $r_{t+1}$  does not depend on  $h_{i,t}$  as long as  $h_{i,t} \in \overline{H}_{i,t}$ .

Also, equation (3) shows that  $f_i(\mu_t r_{t+1}) = (1 - \delta_i)x_i$ .

- **Case 1.** Suppose that  $\delta_i \in (\hat{\delta}_i, \underline{\delta}_i) \cup (\overline{\delta}_i, 1)$ . The proof of Lemma 6 shows that  $f_i(p_{-i}) > (1 \delta_i)x_i$  in this case. Therefore,  $\mu_t r_{t+1} > p_{-i}$ . Hence,  $\alpha_i = \frac{1 p_{-i}}{1 \mu_t r_{t+1}} > 1$ . Letting j = i and  $\alpha_i^* = \alpha_i$  completes the proof for this case.
- **Case 2.** Suppose that  $\delta_i \notin (\hat{\delta}_i, \underline{\delta}_i) \cup (\overline{\delta}_i, 1)$ . The proof of Lemma 6 shows that  $f_i(p_{-i}) < (1-\delta_i)x_i$  in this case. Therefore,  $\mu_t r_{t+1} < p_{-i}$ . Hence,  $\alpha_i = \frac{1-p_{-i}}{1-\mu_t r_{t+1}} < 1$ . This implies that  $\mu_t < \mu_{t+2}$  for all  $t \ge t_1$  such that i moves. Since  $\mu_t r_{t+1}$  is independent of t and  $r_{t+1}$  is a probability, we have  $r_{t+1} \in [0,1)$  for all  $t \ge t_1$  such that i moves. Moreover, substituting  $r_{t+1} = 0$  into (3) implies  $\delta_i = \hat{\delta}_i$ , but we assumed  $\delta_i > \hat{\delta}_i$ . Hence, we have  $r_{t+1} \in (0,1)$ . This implies that there are histories  $h'_{-i,t+1}, h''_{-i,t+1} \in \overline{H}_{-i,t+1}$  such that O and N are -i's best responses, respectively. Thus, Lemma 4 implies that at any history  $h_{-i,t+1} \in \overline{H}_{-i,t+1}$ , both O and N are best responses, and thus in particular they induce the same continuation payoff. Then we can follow the same argument from the second paragraph of the current proof (i.e., the proof of Lemma 7) to Case 1, with the roles of players i and -i switched, to show that  $\mu_{t+1}r_{t+2} > p_i$  and thus  $\alpha_{-i} = \frac{1-p_i}{1-\mu_{t+1}r_{t+2}} > 1$ . Letting j = -i and  $\alpha^*_{-i} = \alpha_{-i}$  completes the proof for this case as well.

Proof of Lemma 8. Recall

$$\pi_i(\delta_i, p_{-i}) = p_{-i}[1 + \delta_i(-\ell_i) + \delta_i^2 \cdot 0] + (1 - p_{-i})[1 + \delta_i \cdot 1 + \delta_i^2 x_i],$$

where we wrote  $p_{-i}$  as an argument of  $\pi_i$  to make the dependence clear. Then,

$$\pi_i(1,0) = 2 + x_i > x_i.$$

The conclusion of the lemma follows because  $\pi_i$  is a continuous function of  $(\delta_i, p_{-i})$ .

### A.2 Proofs for Section 4

#### A.2.1 Proofs for Section 4.1

Proof of Proposition 1. The proof is provided in the main text.

#### A.2.2 Proofs for Section 4.2

Proof of Proposition 2. If  $\overline{q} = 0$ , then  $\sigma^{(T,\overline{q})}$  is the grim-trigger strategy  $\sigma^{G}$  for any T. Thus, similarly to Lemma 1, there exists  $\delta' \in (0,1)$  such that  $\sigma^{(T,\overline{q})}$  is a PBE if  $\delta_i \in (\delta', 1)$  for each i = 1, 2. Henceforth, assume  $\overline{q} \in (0, 1)$ .

Let  $\overline{H}_{i,t}$  be the set of *i*'s private histories at period *t* such that she has privately learned *N* while no player has taken it.

Furthermore, let  $H_i^{1-\overline{q}}$  be the set of private histories of the form  $h_{i,t} \in \overline{H}_{i,t}$  such that  $t = n \cdot T$  for some  $n = 1, 2, \ldots$  and  $q_t > \overline{q}$ . Let the continuation payoff under  $\sigma^{(T,\overline{q})}$  at *i*'s history  $h_{i,t} \in H_i^{1-\overline{q}}$  be  $V_i(h_{i,t})$ . Then, we have

$$V_{i}(h_{i,t}) \leq (1 - \delta_{i}^{T}) \cdot 1 + \delta_{i}^{T} \left( \overline{q}(1 - \delta_{i}) \max\{x_{i}, 0\} + (1 - \overline{q}) \sup_{h_{i,t}' \in H_{i}^{1 - \overline{q}}} V_{i}(h_{i,t}') \right).$$

Since this inequality must hold for every  $h_{i,t} \in H_i^{1-\overline{q}}$ , we have

$$\sup_{h'_{i,t}\in H_i^{1-\overline{q}}} V_i(h'_{i,t}) \le (1-\delta_i^T) \cdot 1 + \delta_i^T \left( \overline{q}(1-\delta_i) \max\{x_i, 0\} + (1-\overline{q}) \sup_{h'_{i,t}\in H_i^{1-\overline{q}}} V_i(h'_{i,t}) \right).$$

This implies that

$$\sup_{h'_{i,t} \in H_i^{1-\overline{q}}} V_i(h'_{i,t}) \le \frac{(1-\delta_i^T) + \delta_i^T \overline{q}(1-\delta_i) \max\{x_i, 0\}}{1-\delta_i^T (1-\overline{q})}$$

Note that

$$\frac{(1-\delta_i^T)+\delta_i^T \overline{q}(1-\delta_i) \max\{x_i,0\}}{1-\delta_i^T (1-\overline{q})} \leq \frac{1-\delta_i}{\overline{q}} \left(\frac{(1-\delta_i^T)}{1-\delta_i}+\delta_i^T \overline{q} \max\{x_i,0\}\right)$$
$$\leq \frac{1-\delta_i}{\overline{q}} \left(T+\delta_i^T \overline{q} \max\{x_i,0\}\right),$$

where the first inequality follows from  $1 - \delta_i^T (1 - \overline{q}) \ge \overline{q}$  and the second from  $\frac{1 - \delta_i^T}{1 - \delta_i} \le T$ .

Thus, we have

$$\sup_{h'_{i,t}\in H_i^{1-\overline{q}}} V_i(h'_{i,t}) \le \frac{1-\delta_i}{\overline{q}} \left(T + \delta_i^T \overline{q} \max\{x_i, 0\}\right).$$
(4)

Now we consider two (exhaustive) types of histories in  $\overline{H}_{i,t}$  in turn. First, consider any history  $h_{i,t} \in \overline{H}_{i,t}$  such that  $t = n \cdot T$  for some n = 1, 2, ... and  $q_t \leq \overline{q}$ . If *i* plays *N*, her continuation payoff is at least  $(1 - p_{-i})^T (1 - \delta_i) \min\{x_i, 0\}$ . If instead she plays *O*, her continuation payoff is at most

$$(1 - (1 - p_{-i})^{T})(1 - \delta_{i})(-\ell_{i}) + (1 - p_{-i})^{T} \left( (1 - \delta_{i}^{T}) \cdot 1 + \delta_{i}^{T} \left( \overline{q}(1 - \delta_{i}) \max\{x_{i}, 0\} + (1 - \overline{q}) \sup_{h_{i,t}' \in H_{i}^{1 - \overline{q}}} V_{i}(h_{i,t}') \right) \right),$$

which is no greater than

$$(1 - (1 - p_{-i})^{T})(1 - \delta_{i})(-\ell_{i}) + (1 - p_{-i})^{T}(1 - \delta_{i})\left(T + \left(\overline{q}\max\{x_{i}, 0\} + \frac{1 - \overline{q}}{\overline{q}}\left(T + \overline{q}\max\{x_{i}, 0\}\right)\right)\right)$$

by equation (4). This is equal to:

$$(1 - (1 - p_{-i})^T)(1 - \delta_i)(-\ell_i) + \frac{(1 - p_{-i})^T}{\overline{q}}(1 - \delta_i)(T + \overline{q}\max\{x_i, 0\}).$$

Thus, the continuation payoff from playing N is no less than the one from playing O if

$$(1 - p_{-i})^T \min\{x_i, 0\} \ge (1 - (1 - p_{-i})^T)(-\ell_i) + \frac{(1 - p_{-i})^T}{\overline{q}} \left(T + \overline{q} \max\{x_i, 0\}\right).$$
(5)

Since  $(1 - p_{-i})^T T \to 0$  as  $T \to \infty$ , it follows that as  $T \to \infty$ , the left-hand side of (5) converges to 0 while its right-hand side converges to  $-\ell_i$ , which is strictly less than 0. Hence, there exists  $T' < \infty$  such that for each i = 1, 2 and for all T > T', N is a best response at any  $h_{i,t} \in \overline{H}_{i,t}$  such that  $t = n \cdot T$  for some  $n = 1, 2, \ldots$  and  $q_t \leq \overline{q}$ . Note that  $T' < \infty$  can be taken uniformly for  $\delta$  because (5) is independent of  $\delta$ . Second, consider any history  $h_{i,t} \in \overline{H}_{i,t}$  such that  $t \neq n \cdot T$  for any n = 1, 2, ... or  $q_t > \overline{q}$ . Let  $s \in \{1, ..., T-1\}$  be such that the current period is t = nT - s for some n = 1, 2, ...

If *i* plays *N*, her continuation payoff is  $(1 - \delta_i)x_i$ . If instead she plays *O*, her continuation payoff is at least

$$(1-\delta_i^s)\cdot 1+\delta_i^s\left(\overline{q}\cdot(1-\delta_i)\min\{x_i,0\}+(1-\overline{q})((1-\delta_i^T)\cdot 1+\delta_i^T(1-\delta_i)\min\{x_i,0\})\right),$$

which is equal to

$$(1-\delta_i^s) + \delta_i^s (1-\overline{q})(1-\delta_i^T) + \delta_i^s (1-\delta_i) \min\{x_i, 0\} (\overline{q} + (1-\overline{q})\delta_i^T).$$

The continuation payoff from playing O is no less than the one from playing N if

$$(1 - \delta_i)x_i \le (1 - \delta_i^s) + \delta_i^s (1 - \overline{q})(1 - \delta_i^T) + \delta_i^s (1 - \delta_i) \min\{x_i, 0\}(\overline{q} + (1 - \overline{q})\delta_i^T),$$

which is equivalent to:

$$x_i \le \frac{1-\delta_i^s}{1-\delta_i} + \delta_i^s (1-\overline{q}) \frac{1-\delta_i^T}{1-\delta_i} + \delta_i^s \min\{x_i, 0\} (\overline{q} + (1-\overline{q})\delta_i^T).$$
(6)

Note that (6) holds if

$$|x_i| \le \frac{1 - \delta_i^s}{1 - \delta_i} + \delta_i^s (1 - \overline{q}) \frac{1 - \delta_i^T}{1 - \delta_i}$$

Since the right-hand side is a convex combination of  $\frac{1}{1-\delta_i}$  and  $(1-\overline{q})\frac{1-\delta_i^T}{1-\delta_i}$  and since  $\frac{1}{1-\delta_i} \ge (1-\overline{q})\frac{1-\delta_i^T}{1-\delta_i}$ , the above inequality holds if

$$|x_i + |x_i| \le (1 - \overline{q}) \frac{1 - \delta_i^T}{1 - \delta_i}.$$
(7)

Now, we show that there exist  $\delta' \in (0,1)$  and  $T' < \infty$  such that (7) holds at  $(\delta_i, T) = (\delta', T')$  for each i = 1, 2. In fact, for any  $T' < \infty$ , the right-hand side of (7) converges to  $(1 - \overline{q})T'$  as  $\delta_i \to 1$ . Thus, noting that  $\overline{q} < 1$  and taking T' that satisfies  $\frac{x_i + |x_i|}{1 - \overline{q}} < T'$  for each i = 1, 2, we can choose  $\delta' < 1$  satisfying (7) for each i = 1, 2. Since the right-hand side of (7) is non-decreasing in  $\delta_i$  and T, it follows that if  $\delta_i \in (\delta', 1)$  and T > T', *i*'s continuation payoff from playing O is no less than the

one from playing N at any  $h_{i,t} \in \overline{H}_{i,t}$  such that  $t \neq n \cdot T$  for any  $n = 1, 2, \ldots$  or  $q_t > \overline{q}$ .

Proof of Proposition 3. Fix any  $p_1, p_2 \in (0, 1]$  and  $\overline{q} \in (0, 1)$ . We have shown in the proof of Proposition 2 that there exist  $\delta' \in (0, 1)$  and  $T' < \infty$  such that  $\sigma^{(T,\overline{q})}$  is a PBE of the synchronous model if  $\delta_i \in (\delta', 1)$  for each i = 1, 2 and T > T'.

Let  $H_i^1$  be the set of histories at period  $t = n \cdot T + 1$  for some n = 0, 1, ...,where the realization of private learning or the public randomization device has not occurred yet.<sup>23</sup> For any  $h_{i,t} \in H_i^1$ , denoting by  $V_i(h_{i,t})$  player *i*'s continuation payoff at  $h_{i,t}$  under  $\sigma^{(T,\bar{q})}$ , we have:

$$V_{i}(h_{i,t}) \leq (1 - \delta_{i}^{T-1}) \cdot 1 + \delta_{i}^{T-1} \bigg( \overline{q} (1 - (1 - p_{1})^{T} (1 - p_{2})^{T}) (1 - \delta_{i}) \max\{x_{i}, 0\} + (\overline{q} (1 - p_{1})^{T} (1 - p_{2})^{T} + 1 - \overline{q}) ((1 - \delta_{i}) + \delta_{i} \sup_{\substack{h_{i,t}' \in H_{i}^{1}}} V_{i}(h_{i,t}')) \bigg).$$

Since this inequality must hold for every  $h_{i,t} \in H_i^1$ , we have

$$\sup_{\substack{h'_{i,t} \in H_i^1}} V_i(h'_{i,t}) \le (1 - \delta_i^{T-1}) \cdot 1 + \delta_i^{T-1} \bigg( \overline{q} (1 - (1 - p_1)^T (1 - p_2)^T) (1 - \delta_i) \max\{x_i, 0\} + (\overline{q} (1 - p_1)^T (1 - p_2)^T + 1 - \overline{q}) ((1 - \delta_i) + \delta_i \sup_{\substack{h'_{i,t} \in H_i^1}} V_i(h'_{i,t})) \bigg)$$

This is equivalent to:

$$(1 - \delta_i^{T-1}(\overline{q}(1 - p_1)^T (1 - p_2)^T + 1 - \overline{q})\delta_i) \sup_{\substack{h'_{i,t} \in H_i^1\\h'_{i,t} \in H_i^1}} V_i(h'_{i,t}) \le (1 - \delta_i^{T-1}) \cdot 1 + \delta_i^{T-1} \left(\overline{q}(1 - (1 - p_1)^T (1 - p_2)^T)(1 - \delta_i) \max\{x_i, 0\} + (\overline{q}(1 - p_1)^T (1 - p_2)^T + 1 - \overline{q})(1 - \delta_i)\right).$$
(8)

For any fixed  $T < \infty$ , notice that the limit superior of the left-hand side of (8) as  $\delta_i \to 1$  is

$$(1 - (\overline{q}(1 - p_1)^T (1 - p_2)^T + 1 - \overline{q})) \limsup_{\delta_i \to 1} \sup_{h'_{i,t} \in H_i^1} V_i(h_{i,t})$$

<sup>&</sup>lt;sup>23</sup>Strictly speaking, these are not "histories" as defined in Section 2, and thus this involves an abuse of terminology. Nonetheless, the payoff is still well defined.

Notice that  $1 - (\overline{q}(1-p_1)^T(1-p_2)^T + 1 - \overline{q}) > 0$  because  $\overline{q} > 0, p_1, p_2 > 0$ , and T > 0. On the other hand, the right-hand side of (8) goes to 0 as  $\delta_i \to 1$ . This implies that

$$\limsup_{\delta_i \to 1} \sup_{\substack{h'_{i,t} \in H_i^1}} V_i(h'_{i,t}) \le 0.$$

Finally, notice that we must have

$$\liminf_{\delta_i \to 1} \inf_{h_{i,t} \in H_i^1} V_i(h_{i,t}) \ge 0$$

because the limit payoff of  $\sigma_i^N$  as  $\delta_i \to 1$  is no less than 0 for any strategy of the opponent. Hence, for any  $h_{i,t} \in H_i^1$ , *i*'s expected payoff under  $\sigma^{(T,\bar{q})}$  at that history converges to 0 as  $\delta_i \to 1$ . By setting n = 0, we conclude that *i*'s expected payoff of the game under  $\sigma^{(T,\bar{q})}$  converges to 0 as  $\delta_i \to 1$ .

#### A.2.3 Proofs for Section 4.4

Proof of Theorem 2. Fix  $k \leq K-1$ . Suppose, as an induction hypothesis, that Claim l holds for every  $l = k + 1, \dots, K$ . We show that Claim k holds. For this purpose, assume  $\delta > \hat{\delta}^{(k+1)}$  and fix a PBE  $\sigma$ . Fix i and t. The proof consists of three steps.

For the first step, consider a history  $h_{i,t} \in H_{i,t}$  at which player *i*'s current latest action is  $a^k$  and player -i's current latest action is  $a^{k'}$ , where k' > k. The induction hypothesis implies the following.

1. If i plays  $a^l$  such that l > k', then her payoff is

$$(1-\delta_i)u_i(a^l, a^{k'}) + \delta_i u_i(a^l, a^l).$$

Due to condition (A), there is  $\hat{\delta}_i < 1$  such that for all  $\delta_i \in (\hat{\delta}_i, 1)$ , this is strictly less than  $u_i(a^{k'}, a^{k'})$ .

2. If i plays  $a^l$  such that l = k', then her payoff is

$$u_i(a^{k'}, a^{k'})$$

3. If i plays  $a^l$  such that k < l < k', then her payoff is at most

$$(1 - \delta_i^2)u_i(a^l, a^{k'}) + \delta_i^2 u_i(a^{k'}, a^{k'}).$$

Due to condition (B), this is strictly less than  $u_i(a^{k'}, a^{k'})$ .

4. If *i* plays  $a^l$  such that l = k, then her payoff is a convex combination of  $u_i(a^k, a^{k'})$ and at most  $u_i(a^{k'}, a^{k'})$  when  $\delta_i \in (\hat{\delta}_i, 1)$  from cases 1–3 above, with a strictly positive weight on the former payoff. Due to condition (B), such a convex combination is strictly less than  $u_i(a^{k'}, a^{k'})$ .

Overall, we find that it is *i*'s unique best response to play  $a^{k'}$  if k' > k, provided  $\delta_i > \max(\hat{\delta}^{(k+1)}, \hat{\delta}_i) < 1$ .

For the second step, consider a history  $h_{i,t} \in H_{i,t}$  at which player *i*'s current latest action is  $a^{k'}$  and player -i's current latest action is  $a^k$ , where k' > k. Then, if *i* plays  $a^l$  such that  $l \ge k'$ , we know from the first step that -i plays  $a^l$  at t + 1, and by the induction hypothesis, the action profile  $(a^l, a^l)$  will be played thereafter at every period. Hence, the payoff is

$$(1 - \delta_i)u_i(a^l, a^k) + \delta_i u_i(a^l, a^l).$$

Due to condition (A), there is  $\hat{\delta}'_i < 1$  such that for all  $\delta \in (0, \hat{\delta}'_i)$ , this is maximized when l = k'. Hence, we find that it is *i*'s unique best response to play  $a^{k'}$  if k' > k, provided  $\delta_i > \max(\hat{\delta}^{(k+1)}, \hat{\delta}_{-i}, \hat{\delta}'_i) < 1$ .

For the third step, consider a history  $h_{i,t} \in H_{i,t}$  at which player *i*'s current latest action is  $a^k$  and player -i's current latest action is also  $a^k$ .

We start with showing that for all time  $t' \ge t$ , for all history  $h_{j,t'}$  where j is the moving player at t' such that the current action profile is  $(a^k, a^k)$ , j does not play  $a^l$  with k + 1 < l. To see this, note that our first step implies that the payoff from playing  $a^{l'}$  with k < l' is

$$(1 - \delta_j)u_j(a^{l'}, a^k) + \delta_j u_j(a^{l'}, a^{l'}).$$

Due to condition (A), there is  $\hat{\delta}''_j < 1$  such that for all  $\delta_j \in (\hat{\delta}''_j, 1)$ , this is maximized when l' = k + 1 among  $l' \in \{k + 1, ..., K\}$ . Since action  $a^{k+1}$  is available to a player when action  $a^{l'}$  with k + 1 < l' is available to that player, this implies that when the current action profile is  $(a^k, a^k)$ , the moving player j will never take any action  $a^l$ with k + 1 < l if  $\delta > \hat{\delta}''_j$ .

Note that the first step, the second step, and the induction hypothesis pin down the players' actions under  $\sigma$  when the current action profile is  $(a^k, a^{k+1}), (a^{k+1}, a^k),$  and  $(a^{k+1}, a^{k+1})$ , respectively, and they imply that the chosen action is  $a^{k+1}$  with probability 1 in either case. This implies that at every time  $t' \geq t$  such that the current action profile is  $(a^k, a^k)$ , the moving player's optimization problem is the same as the one for  $2 \times 2$  games. Thus, by letting  $\hat{\delta}^{(k)} := \max(\hat{\delta}^{(k+1)}, \hat{\delta}_1, \hat{\delta}_2, \hat{\delta}'_1, \hat{\delta}'_2, \hat{\delta}''_1, \hat{\delta}''_2) < 1$ , Theorem 1 implies that it is *i*'s unique best response to play  $a^k$  if the current latest action profile is  $(a^k, a^k)$ , provided  $\delta_i > \hat{\delta}^{(k)}$ . This concludes the third step.

Overall, we find that Claim k holds. Since Claim k holds if Claim l holds for all  $l \in \{k + 1, ..., K\}$ , by induction, Claim k holds for every k. Thus, we conclude that  $\sigma^{G}$  is a unique candidate for a PBE if  $\delta > \max_k \hat{\delta}^{(k)} (= \hat{\delta}^{(1)})$ . The same argument shows that  $\sigma^{G}$  is indeed a PBE under the same condition. The proof is complete.  $\Box$ 

#### A.2.4 Proofs for Section 4.6

Proof of Lemma 9. Fix i = 1, 2.

1. First, notice that  $\sigma_i^N$  designates a unique best response at any history such that the opponent has already chosen N in the past.

So, consider a history in  $\overline{H}_{i,t}$ . If she takes action N, her continuation payoff is  $(1 - \delta_i)x_i$ . If, instead, she takes action O, then her continuation payoff is

$$(1 - \delta_i) \cdot 1 + \delta_i \left( p_{-i}((1 - \delta_i)(-\ell_i) + \delta_i \cdot 0) + (1 - p_{-i})((1 - \delta_i) \cdot 1 + \delta_i(1 - \delta_i)x_i) \right) + \delta_i \left( p_{-i}((1 - \delta_i)(-\ell_i) + \delta_i \cdot 0) + (1 - p_{-i})((1 - \delta_i) \cdot 1 + \delta_i(1 - \delta_i)x_i) \right) + \delta_i \left( p_{-i}((1 - \delta_i)(-\ell_i) + \delta_i \cdot 0) + (1 - p_{-i})((1 - \delta_i) \cdot 1 + \delta_i(1 - \delta_i)x_i) \right) + \delta_i \left( p_{-i}((1 - \delta_i)(-\ell_i) + \delta_i \cdot 0) + (1 - p_{-i})((1 - \delta_i) \cdot 1 + \delta_i(1 - \delta_i)x_i) \right) + \delta_i \left( p_{-i}((1 - \delta_i)(-\ell_i) + \delta_i \cdot 0) + (1 - p_{-i})((1 - \delta_i) \cdot 1 + \delta_i(1 - \delta_i)x_i) \right) + \delta_i \left( p_{-i}((1 - \delta_i) - \delta_i)x_i \right) + \delta_i \left( p_{-i}((1 - \delta_i) - \delta_i \right) + \delta_i \left( p_{-i}((1 - \delta_i) - \delta_i)x_i \right) + \delta_i \left( p_{-i}((1 - \delta_i) - \delta_i \right) + \delta_i \left( p_{-i}((1 - \delta_i) - \delta_i)x_i \right) + \delta_i \left( p_{-i}((1 - \delta_i) - \delta_i \right) + \delta_i \left( p_{-i}((1 - \delta_i) - \delta_i \right) + \delta_i \left( p_{-i}((1 - \delta_i) - \delta_i)x_i \right) + \delta_i \left( p_{-i}((1 - \delta_i) - \delta_i \right) + \delta_i \left( p_{-i}((1 - \delta_i) - \delta_i \right) + \delta_i \left( p_{-i}((1 - \delta_i) - \delta_i \right) + \delta_i \left( p_{-i}((1 - \delta_i) - \delta_i \right) + \delta_i \left( p_{-i}((1 - \delta_i) - \delta_i \right) + \delta_i \left( p_{-i}((1 - \delta_i) - \delta_i \right) +$$

which is  $(1 - \delta_i)\pi_i(\delta_i)$ . Then, we have the former being strictly greater than the latter if and only if  $x_i > \pi_i(\delta_i)$ .<sup>24</sup> Thus,  $\sigma_i^N$  is a unique best response against  $\sigma_{-i}^N$  whenever  $x_i > \pi_i(\delta_i)$ .

2. First, notice that  $\sigma_i^{\text{G}}$  designates a unique best response at any history such that the opponent has already chosen N in the past.

So, consider a history in  $\overline{H}_{i,t}$ . If she takes action N, her continuation payoff is  $(1 - \delta_i)x_i$ . If, instead, she takes action O, then her continuation payoff is

$$(1 - \delta_i) \cdot 1 + \delta_i \left( p_{-i} ((1 - \delta_i)(-\ell_i) + \delta_i \cdot 0) + (1 - p_{-i})((1 - \delta_i) \cdot 1 + \delta_i (1 - \delta_i) x_i) \right) + \delta_i (1 - \delta_i) (1 - \delta_i) \cdot 1 + \delta_i \cdot 1 + \delta$$

which is  $(1 - \delta_i)\pi_i(\delta_i)$ . Then, we have the latter being strictly greater than the

 $<sup>^{24}\</sup>mathrm{Note}$  that this calculation is essentially the same as in the proof of Lemma 6.

former if and only if  $x_i < \pi_i(\delta_i)$ .<sup>25</sup> Thus,  $\sigma_i^{\rm G}$  is a unique best response against  $\sigma_{-i}^{N}$  whenever  $x_i < \pi_i(\delta_i)$ .

3. Fix any PBE. Note first that, if player *i* chooses *N* at time *t*, then -i chooses *N* at time t + 1. Consider a history  $h_{i,t} \in \overline{H}_{i,t}$  at time *t* in which player *i* moves. Notice that if *i* plays *N* at *t*, her payoff is  $(1 - \delta_i) \cdot x_i + \delta_i \cdot 0 = (1 - \delta_i)x_i$ . If she plays *O* at *t* but -i plays *N* at time t + 1, then *i*'s payoff is  $(1 - \delta_i) \cdot 1 + \delta_i[(1 - \delta_i) \cdot (-\ell_i) + \delta_i \cdot 0] = (1 - \delta_i)(1 - \delta_i\ell_i)$ . If no one plays *N* at any time, *i*'s continuation payoff is 1. Hence, *i*'s continuation payoff at  $h_{i,t}$  is in the following set:

$$\operatorname{co}\left(\left\{\delta_i^{2n}(1-\delta_i)x_i|n=0,1,\ldots\right\}\cup\left\{\delta_i^{2n+1}(1-\delta_i)(1-\delta_i\ell_i)|n=0,1,\ldots\right\}\cup\left\{1\right\}\right),$$

where co(S) denotes the convex hull of the set S. If  $\delta_i < \frac{x_i-1}{x_i}$ , the maximum element in this set is  $(1 - \delta_i)x_i$  because  $(1 - \delta_i)x_i > 1$  and  $(1 - \delta_i)(1 - \delta_i\ell_i) < 1$ . Hence, any action that induces the continuation payoff of  $(1 - \delta_i)x_i$  is a best response at  $h_{i,t}$ . Hence, choosing action N is a best response at  $h_{i,t}$ . Since there is an action that induces the payoff of  $(1 - \delta_i)x_i$  at  $h_{i,t}$ , no other action is a best response. Hence, at  $h_{i,t}$ , choosing action N is a unique best response and thus, i chooses N at  $h_{i,t}$  under the PBE we fixed.

Since this was true for any choice of  $h_{i,t}$  from  $\overline{H}_{i,t}$ , this shows that if  $\delta_i < \frac{x_i-1}{x_i}$ , in any PBE, *i* chooses action N at any history in  $\overline{H}_{i,t}$ . This completes the proof.

#### Proof of Theorem 3.

- 1. This part follows from Lemmas 1-7.
- 2. By part 3 of Lemma 9, in any PBE, player *i*'s equilibrium strategy is  $\sigma_i^N$ . Then, by part 2 of Lemma 9, player -i's unique best response is  $\sigma_{-i}^G$ . Thus, the result obtains.
- 3. Lemma 1 shows that  $\sigma^{G}$  is a PBE. Also, by part 2 of Lemma 9,  $\sigma^{N}$  is a PBE.

 $<sup>^{25}\</sup>mathrm{Again},$  this calculation is essentially the same as in the proof of Lemma 6.

4. By part 3 of Lemma 9, in any PBE, player *i*'s equilibrium strategy is  $\sigma_i^N$ . Then, by part 1 of Lemma 9, player -i's unique best response is  $\sigma_{-i}^N$ . Thus, the result obtains.