Protocol Invariance and the Timing of Decisions in Dynamic Games*

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Abstract

The timing of decisions is an essential ingredient into modelling many strategic situations and often matters crucially for equilibrium behavior. We characterize a class of dynamic stochastic games that we call separable dynamic games with noisy transitions and establish that these widely used models are protocol invariant provided that periods are sufficiently short. Protocol invariance means that the set of Markov perfect equilibria is nearly the same irrespective of the order in which players are assumed to move within a period. We also show that the equilibria have a remarkably simple structure.

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1 Introduction

The timing of decisions is an essential ingredient into modelling many strategic situations. Asynchronous
decisions are a type of commitment, and being able to move first and thus set the stage for rivals can
confer a considerable advantage on a player. Synchronous decisions, in contrast, take away the ability to
commit as players are neither leaders nor followers. From the basically static models in Cournot (1838) and
von Stackelberg (1934) to the genuinely dynamic models in Cyert and DeGroot (1970), Maskin and Tirole
(1987, 1988a, 1988b), Cahuc and Kempf (1997), Noel (2008), and Iskhakov, Rust, and Schjerning (2016) and
the anti-folk theorems in Rubinstein and Wolinsky (1995) and Lagunoff and Matsui (1997, 2001), a long and
distinguished literature has pointed out cases where the protocol of moves matters crucially for equilibrium
behavior.

Our paper provides a counterpoint to this literature. We show that a fairly general and widely used class
of dynamic models is protocol invariant provided that periods are sufficiently short and moves are therefore
sufficiently frequent. Protocol invariance means that the set of equilibria of a model is nearly the same
irrespective of the order in which players are assumed to move within a period, including—and extending
beyond—simultaneous, alternating, and sequential moves.

We focus on infinite-horizon dynamic stochastic games and their stationary Markov perfect equilibria
(henceforth Markov perfect equilibria for short). Dating back to Shapley (1953), dynamic stochastic games
have a long tradition in economics and are central to the analysis of strategic interactions among forward-
looking players in dynamic environments. The main contribution of this paper is to characterize a class of
dynamic stochastic games that we call separable dynamic games with noisy transitions and to establish that
these models are protocol invariant provided that periods are sufficiently short. In addition, we show that
the Markov perfect equilibria of separable dynamic games with noisy transitions have a remarkably simple
structure.

Separability and noisy transitions are assumptions that restrict per-period payoffs and state-to-state
transitions in a dynamic stochastic game. Many models in the literature are amenable to these assumptions.
Examples include investment games (Spence 1979, Fudenberg and Tirole 1983, Hanig 1986, Reynolds 1987,
Reynolds 1991, Dockner 1992), R&D races (Reinganum 1982, Lippman and McCardle 1987), models of
industry dynamics (Ericson and Pakes 1995), and dynamic public contribution games (Marx and Matthews
2000, Compte and Jehiel 2004, Georgiadis 2014). The recent continuous-time stochastic games with moves
at random times (Arcidiacono, Bayer, Blevins, and Ellickson 2015, Ambrus and Lu 2015, Calcagno, Kamada,
Lovo, and Sugaya 2014) also satisfy these assumptions.

A dynamic stochastic game is a dynamic system that can be in different states at different times. The
evolution of the state from one period to the next is governed by a discrete-time Markov process that players
can influence through their actions. Each player strives to maximize the expected net present value of his
stream of payoffs. While per-period payoffs and state-to-state transitions in a general dynamic stochastic game
depend arbitrarily on the state and on players’ actions, in a separable dynamic game per-period payoffs and
state-to-state transitions depend on players’ actions in an additive manner: to a first-order approximation, they are built from parts that depend on the actions taken by subsets of players. As a consequence, the strategic situation that the $N$ players face in a given state of the dynamic system, holding fixed the value of continued play, is akin to $N$ independent optimization problems. While at first glance this may seem to trivialize the game, the separability assumption does not restrict how per-period payoffs and state-to-state transitions depend on the state. Because players are forward looking, this allows strategic interactions to be channeled through continuation values. Indeed, as the many examples in the literature show, the separability assumption is not overly onerous.

The assumption of noisy transitions precludes that there is an action that a player can take to guarantee a change in the state. This assumption reflects the view that models are only an approximation to reality, so that there typically is at least some residual uncertainty. To impose noisy transitions we model the evolution of the state by a discrete-time approximation to a continuous-time Markov process in which the time spent in a state has an exponential distribution with a finite hazard rate, in line with the fact that in many real-world settings players often take time to finalize their actions (Iijima and Kasahara 2015).

Our first main result, Theorem 1, is that separable dynamic games with noisy transitions are protocol invariant provided that periods are sufficiently short. To provide intuition, consider a prototypical investment game between two firms. A firm can undertake a risky investment project to increase its capital stock. A firm’s per-period payoff increases in its own capital stock and decreases in its rival’s capital stock. The separability assumption is satisfied, as whether its rival invests affects directly neither the firm’s per-period payoff nor the probability that the firm succeeds in increasing its capital stock. Moreover, transitions from one state to another are noisy due to the risky nature of the investment project.

Now contrast two protocols of moves. When firms move alternatingly, a forward-looking firm deciding whether to invest understands that its rival’s capital stock remains constant for (at least) the period. In contrast, when firms move simultaneously, the firm has to take into account the probability that its rival’s capital stock increases over the course of the period. This probability, however, becomes negligible as periods become short because transitions are noisy. It follows that the protocol of moves is almost immaterial to the firm’s decision.

The proof of Theorem 1 proceeds from the equilibrium conditions for a separable dynamic game with noisy transitions. We fix an arbitrary protocol of moves and take the limit as periods become short and we therefore pass from discrete to continuous time. We observe that the limit conditions are independent of the protocol of moves used to pass to the limit. Hence, when the limit conditions admit a unique solution, our protocol-invariance theorem is immediate. However, the limit conditions may admit multiple solutions. To handle the resulting difficulties, we introduce differential topology tools to study the limit conditions. Drawing on ideas in Harsanyi (1973a, 1973b) for normal-form games and in Doraszelski and Escobar (2010) for dynamic stochastic games, we prove that generically the limit conditions have a finite number of solutions and that all solutions can be approximated by the Markov perfect equilibria of a separable dynamic game.
with noisy transitions and an arbitrary protocol of moves provided that periods are sufficiently short. To the best of our knowledge, our paper is the first attempt to use differential topology tools to explore the robustness of a class of dynamic models to the timing of decisions.

We show that the assumptions of separability and noisy transitions are tight in the sense that counterexamples to protocol invariance can be constructed if any one of them is relaxed. Moreover, we show that protocol invariance does not extend beyond Markov perfect equilibria to other equilibrium concepts.

While we mostly treat the limit conditions as a technical device, they are of interest by themselves. The limit conditions can be interpreted as the equilibrium conditions for a continuous-time stochastic game. We also provide an equivalence result showing that the limit conditions are identical to the equilibrium conditions for a dynamic stochastic game in which in any period one player is randomly selected to make a decision. In this game with random moves, the fact that a player can revise his decision only at random times confers a kind of commitment power on the player similar to that in the games with alternating moves in Maskin and Tirole (1988a, 1988b) and Lagunoff and Matsui (1997). Our equivalence result therefore underscores the richness of the class of separable dynamic games with noisy transitions and clarifies the strategic implications of these restrictions on per-period payoffs and state-to-state transitions.

Our second main result, Theorem 2, shows that the Markov perfect equilibria of a separable dynamic game with noisy transitions have a remarkably simple structure provided that periods are sufficiently short. In particular, we show that the number of actions that a player uses with positive probability in a given state cannot exceed the number of players in the game. The proof of Theorem 2 capitalizes on the analytic tractability of the limit conditions and on Theorem 1 to extend the analysis to separable dynamic games with noisy transitions.

Our main results facilitate and inform applied work in a number of ways. First and perhaps most important, determining the protocol of moves that is most realistic and appropriate for the application at hand may be amongst the most difficult choices a modeler has to make. In empirical work, in particular, the timing of decisions and the ability to commit is typically not observable to the researcher. Hence, we may be suspicious of any implication or prediction from a model that is driven by the protocol of moves that the modeler has chosen to impose, a point that has been made forcefully by Rosenthal (1991) and van Damme and Hurkens (1996) for normal-form games and by Kalai (2004) for large Bayesian games. Protocol invariance alleviates this concern and the burden of determining the protocol of moves for the class of separable dynamic game with noisy transitions by ensuring that equilibrium behavior is independent of the timing of decisions provided that periods are sufficiently short. Second, our main results caution against the presumption that imposing asynchronous instead of synchronous decisions on a dynamic stochastic game reduces the number of equilibria. Third, because the timing of decisions and the ability to commit is typically not observable, empirical work sometimes averages over different protocols of moves (Einav 2010). This average depends on the assumed probability distribution over protocols of moves and may be difficult to interpret if it does not correspond to an equilibrium of any game. Protocol invariance renders averaging unnecessary. Fourth,
Dynamic stochastic games are often not very tractable analytically and thus call for the use of numerical methods. Doraszelski and Judd (2007) show that the computational burden can vary by orders of magnitude with the protocol of moves. For the class of separable dynamic game with noisy transitions, protocol invariance justifies imposing the protocol of moves that is most convenient from a computational perspective.

We apply and extend our main results in three ways. First, we provide a new justification for focusing on Markov perfect equilibria. Provided that periods are sufficiently short, we show that if a strict finite-memory equilibrium payoff profile in a separable dynamic game with noisy transitions and simultaneous moves is protocol invariant, then it is arbitrarily close to a Markov perfect equilibrium payoff profile. In this sense, Markov perfect equilibria are the only equilibria that are robust to changes in the protocol of moves. This result adds to the literature providing foundations for Markov perfect equilibria (Maskin and Tirole 2001, Bhaskar and Vega-Redondo 2002, Bhaskar, Mailath, and Morris 2013, Bohren 2014). Second, as Doraszelski and Judd (2012) argue, the limit conditions that arise as we pass from discrete to continuous time are particularly easy to solve numerically. We provide a justification for doing so by showing that the solutions to the limit conditions almost coincide with the Markov perfect equilibria of separable dynamic games with noisy transitions and arbitrary protocols of moves provided that periods are sufficiently short. We also show that restricting attention to games with simultaneous moves, we can dispense with the separability assumption. Third, we shed light on the numerous examples in the literature in which the protocol of moves matters crucially for equilibrium behavior. We show that there is a discontinuity in the set of Markov perfect equilibria as hazard rates become large and moves become frequent. Hence, caution is warranted in working with games with infinite hazard rates and arbitrarily frequent moves.

Our paper is related to two strands of literature. First, our notion of protocol invariance builds on and extends the notion of a commitment robust equilibrium in Rosenthal (1991) and van Damme and Hurkens (1996) from two-player normal-form games to $N$-player dynamic stochastic games. Rosenthal (1991) defines a Nash equilibrium of a two-player normal-form game to be commitment robust if it is also a subgame perfect equilibrium outcome of each of the two extensive-form games in which one of the players moves first, and provides a series of illustrative examples. In contrast to the notion of a commitment robust equilibrium, our notion of protocol invariance pertains to the entire set of equilibria of a fairly general class of dynamic models. Our work is also related to Kalai (2004), who shows that the Nash equilibria of large anonymous Bayesian games are approximately robust to variations in the extensive-form version of the game. The driving force behind Kalai’s (2004) result is the vanishing impact that a player’s action has on other players’ payoffs as the number of players grows large. In our setting, the impact that a player’s action has on other players’ payoffs vanishes as periods become short. From a more technical perspective, Kalai (2004) allows for $\epsilon$-equilibria, while we impose exact equilibrium and establish our results for generic payoffs.

Second, previous attempts to exposit dynamic games where the protocol of moves does not matter for equilibrium behavior are few and far between and confined to very specific models. Abreu and Gul (2000) study bilateral bargaining and show that independent of the bargaining protocol the same limit is reached as
the time between offers becomes short. Caruana and Einav (2008) study a model in which players repeatedly announce an action but only the final announced action is relevant for payoffs. While players can revise their announcements, they pay a cost each time they do so; in this way, announcements play the role of an imperfect commitment device. Caruana and Einav (2008) show that the order in which players make announcements does not matter as long as the time between announcements is sufficiently short. In contrast to Abreu and Gul (2000) and Caruana and Einav (2008), we do not presuppose that the limit conditions admit a unique solution. Because our framework is much less tightly specified, we require differential topology tools to analyse the limit conditions.

The remainder of this paper is organized as follows. Section 2 introduces separable dynamic games with noisy transitions. Sections 3 and 4 develop our main results. Section 5 discusses a number of applications and extensions of our main results and Section 6 concludes. The proofs of our main results are in the Appendix. An Online Appendix provides further examples and proofs.

2 Separable Dynamic Games with Noisy Transitions

We focus on dynamic stochastic games with finite sets of players, states, and actions. Time \( t = 0, \Delta, 2\Delta, \ldots \) is discrete and measured in units of \( \Delta > 0 \). We refer to \( \Delta \) as the length of a period; as \( \Delta \to 0 \), moves become frequent. The time horizon is infinite. Let \( \{1, 2, \ldots, N\} \) denote the set of players, \( \Omega \) the set of states, and \( \mathcal{A}_i(\omega) \) the set of actions of player \( i \) in state \( \omega \). Each player strives to maximize the expected net present value of his stream of payoffs and discounts future payoffs using a discount rate \( \rho > 0 \). Monitoring is perfect.

The protocol of moves determines which players can take an action at time \( t \) and which players cannot. We allow for a general protocol of moves that encompasses—and goes beyond—simultaneous, alternating, and sequential moves. To this end, we allow the set of players who have the move to change from one period to the next. The set of players \( J^t = 0 \subseteq \{1, 2, \ldots, N\} \) who have the move at time \( t \) thus becomes part of the state of the system, and we refer to it as the “protocol” state to distinguish it from the familiar “physical” state \( \omega^t \in \Omega \). In contrast to the physical state, for simplicity we assume that the protocol state evolves independently of players’ actions. In the Online Appendix, we show that our protocol-invariance theorem remains valid without this simplifying assumption.

The game proceeds as follows. It starts at time \( t = 0 \) from an initial state \((\omega^t=0, J^t=0)\). After observing \((\omega^{t=0}, J^{t=0})\), the players \( j \in J^{t=0} \) who have the move choose their actions \( a^{t=0}_{J^{t=0}} = (a^{t=0}_j)_{j \in J^{t=0}} \) simultaneously and independently from each other. Now two things happen, depending on the state \((\omega^{t=0}, J^{t=0})\) and the actions \( a^{t=0}_{J^{t=0}} \). First, player \( i \) receives a payoff \( u^\Delta_i(\omega^{t=0}, J^{t=0}, a^{t=0}_{J^{t=0}}) \). Second, the system transits from state \((\omega^{t=0}, J^{t=0})\) to state \((\omega^{t=\Delta}, J^{t=\Delta})\). Independent of each other, the transition from \( \omega^{t=0} \) to \( \omega^{t=\Delta} \) happens with probability \( \Pr(\omega^{t=\Delta} | \omega^{t=0}, J^{t=0}, a^{t=0}_{J^{t=0}}) \) and that from \( J^{t=0} \) to \( J^{t=\Delta} \) with probability \( \Pr(J^{t=\Delta} | J^{t=0}) \). While player \( i \) receives a payoff irrespective of whether he has the move \((i \in J^{t=0})\) or not \((i \notin J^{t=0})\), the exact amount depends on the actions \( a^{t=0}_{J^{t=0}} \) of the players who have the move, as do the state-to-state transitions.
In the next round at time $t = \Delta$, after observing $(\omega^t = \Delta, J^t = \Delta)$, the players $j \in J^t = \Delta$ who have the move choose their actions $a^t_{J^t = \Delta}$. Then player $i$ receives a payoff $u^i(\omega^t = \Delta, J^t = \Delta, a^t_{J^t = \Delta})$ and the state changes again from $(\omega^t = \Delta, J^t = \Delta)$ to $(\omega^{t+\Delta} = 2\Delta, J^{t+\Delta} = 2\Delta)$. The game goes on in this way ad infinitum.

To allow for a general protocol of moves, we partition the set of players $\{1, 2, \ldots, N\}$ and assume that the set of players $J'$ who have the move at time $t$ evolves according to a Markov process that is defined over this partition as follows:

**Assumption 1 (Protocol of Moves)** Let $J$ be a partition of $\{1, 2, \ldots, N\}$ and $P = (\Pr (J'|J))_{J,J' \in \mathcal{J}}$ a $|\mathcal{J}| \times |\mathcal{J}|$ transition matrix. $P$ is irreducible and its unique stationary distribution is uniform on $J$.

In stating Assumption 1 and throughout the remainder of the paper we omit the time superscript whenever possible and use a prime to distinguish future from current values.

We denote the protocol of moves as $<J, P>$ in what follows. Because $J$ is a partition of $\{1, 2, \ldots, N\}$, Assumption 1 ensures that player $i$ always has the move in conjunction with the same rivals. By requiring the transition matrix $P$ to have a unique stationary distribution that is uniform on $J$, Assumption 1 further ensures that all players have the move with the same frequency over a sufficiently large number of periods.

Assumption 1 accommodates synchronous and asynchronous decisions and thus encompasses most dynamic stochastic games in the literature, including games with simultaneous moves (Shapley 1953, Ericson and Pakes 1995), games with alternating moves, (Maskin and Tirole 1987, Maskin and Tirole 1988b, Maskin and Tirole 1988a, Lagunoff and Matsui 1997), and games with random moves (Doraszelski and Judd 2007).

In games with simultaneous moves, the partition is $J = \{\{1, \ldots, N\}\}$ with the trivial $1 \times 1$ transition matrix $P$; in games with alternating moves the partition is $J = \{\{1\}, \ldots, \{N\}\}$ with the $N \times N$ transition matrix $P$ with entries $\Pr(\{\text{mod } N(i + 1)\}|\{i\}) = 1$.

In games with asynchronous moves, $J = \{\{1\}, \ldots, \{N\}\}$ and the identity of the player who has the move in a given period may follow a deterministic sequence as in games with alternating moves or it may be stochastic. Games with random moves are another special case of games with asynchronous moves. In these games, the probability that a player has the move in a given period may follow a deterministic sequence as in games with alternating moves or it may be stochastic. Games with random moves are another special case of games with asynchronous moves. In these games, the probability that a player has the move in a given period is uniform across players and periods. Finally, Assumption 1 accommodates more than one—but less than all—players having the move in a given period and thus settings where decisions are partially synchronous.

In the Online Appendix, we show that Assumption 1 can be relaxed in several ways. First, we show that our protocol-invariance theorem remains valid if the evolution of the protocol state $J$ depends on players’ actions $a_J$ and the physical state $\omega$. Second, we show that the uniform stationary distribution in Assumption 1 can be replaced by a non-uniform stationary distribution. Third, we provide a partial extension of our protocol-invariance theorem that does not require $J$ to be a partition of the set of players.

We model the evolution of the physical state by a discrete-time approximation to a continuous-time Markov process in order to impose that transitions are noisy:

**Assumption 2 (Noisy Transitions)** The transition probability $\Pr(\omega' | \omega, J, a_J)$ is differentiable in $\Delta$ and $1$

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The notation $\text{mod } N(x)$ refers to the modulo $N$ congruence.
can be written as

\[
\Pr^\Delta(\omega'|\omega, J, a_J) = \begin{cases} 
1 - q_J(\omega, a_J)\Delta + O(\Delta^2) & \text{if } \omega' = \omega, \\
q_J(\omega, a_J)p_J(\omega'|\omega, a_J)\Delta + O(\Delta^2) & \text{if } \omega' \neq \omega,
\end{cases}
\]

where \( q_J : \{(\omega, (a_j)_{j \in J}) | a_j \in A_j(\omega)\} \to \mathbb{R}^+ \cup \{0\}, \) \( p_J : \{(\omega, (a_j)_{j \in J}) | a_j \in A_j(\omega)\} \to P(\Omega), \) and \( P(\Omega) \) is the set of probability distributions over \( \Omega. \) We normalize \( p_J(\omega | \omega, a_J) = 0. \)

Without loss of generality, we decompose the transition probability \( \Pr^\Delta(\omega'|\omega, J, a_J) \) into a probability that the state changes in a given period—or that a jump occurs in the lingo of stochastic processes—and a probability distribution over successor states conditional on the state changing. The probability that the state changes is \( q_J(\omega, a_J)\Delta \) in proportion to the length of a period \( \Delta \) and, conditional on the state changing, the probability that it changes from \( \omega \) to \( \omega' \) is \( p_J(\omega'|\omega, a_J). \) Normalizing \( p_J(\omega | \omega, a_J) = 0 \) amounts to ignoring a jump from a state to itself and adjusting the hazard rate \( q_J(\omega, a_J) \) of a jump occurring accordingly. Importantly, Assumption 2 restricts the model to have a finite hazard rate in a given state so that there is no action that a player can take to guarantee a change in the state.

We finally assume that per-period payoffs and state-to-state transitions have an additively separable structure:

**Assumption 3 (Separability)** The per-period payoff \( u^\Delta_i(\omega, J, a_J) \) is differentiable in \( \Delta \) and can be written as

\[
u^\Delta_i(\omega, J, a_J) = |J| \sum_{j \in J} u_{i,j}(\omega, a_j)\Delta + O(\Delta^2),\]

where \( u_{i,j} : \{(\omega, a_j) | a_j \in A_j(\omega)\} \to \mathbb{R}. \) The hazard rate \( q_J(\omega, a_J) \) and transition probability \( p_J(\omega' | \omega, a_J) \) can be written as

\[
q_J(\omega, a_J) = |J| \sum_{j \in J} q_j(\omega, a_j)
\]

and

\[
q_J(\omega, a_J)p_J(\omega' | \omega, a_J) = |J| \sum_{j \in J} q_j(\omega, a_j)p_j(\omega' | \omega, a_j),
\]

where \( q_j : \{(\omega, a_j) | a_j \in A_j(\omega)\} \to \mathbb{R}^+ \cup \{0\} \) and \( p_j : \{(\omega, a_j) | a_j \in A_j(\omega)\} \to P(\Omega). \)

To a first-order approximation, Assumption 3 builds up the per-period payoff \( u^\Delta_i(\omega, J, a_J) \) of player \( i \) from the flow payoff \( u_{i,j}(\omega, a_j) \) by summing over the players \( j \in J \) who have the move. By taking action \( a_j \) in state \( \omega, \) player \( j \) “contributes” \( |J| u_{i,j}(\omega, a_j)\Delta \) to the per-period payoff of player \( i \) in proportion to the length of a period \( \Delta. \) This restricts complementarities between players’ actions and other non-separabilities to the higher-order term \( O(\Delta^2). \)

\[\text{We discuss below our reason for scaling by the number of elements of the partition } J.\]

\[\text{More explicitly, we assume that there exists } \bar{c} > 0 \text{ and } \bar{\Delta} > 0 \text{ such that } \|u^\Delta_i(\omega, J, a_J) - |J| \sum_{j \in J} u_{i,j}(\omega, a_j)\Delta\| \leq \bar{c}\Delta^2 \text{ for all } \Delta < \bar{\Delta}.\]
Assumption 3 in conjunction with Assumption 2 also builds up the components of the transition probability \( \Pr^\Delta (\omega' | \omega, J, a_j) \) from the player-specific hazard rate \( q_j(\omega, a_j) \) and the transition probability \( p_j(\omega' | \omega, a_j) \) by summing over the players \( j \in J \) who have the move. Because it imposes a competing hazards model on the transition probability, a change in the state is with high probability due to the action taken by one of the players having the move.

In what follows, we denote the above game by \( \Gamma = \langle \Delta, J, \mathcal{P}, u, p, q, \rho \rangle \). We view the function \( u_{i,j} : \{(\omega, a_j) \mid a_j \in A_j(\omega)\} \rightarrow \mathbb{R} \) as a vector \( u_{i,j} \in \mathbb{R}^{\sum_{\omega \in \Omega} |A_j(\omega)|} \) and denote \( u_i = (u_{i,j})_{j=1}^N \in \mathbb{R}^{\sum_{j=1}^N \sum_{\omega \in \Omega} |A_j(\omega)|} \) and \( u = (u_i)_{i=1}^N \in \mathbb{R}^{N \sum_{j=1}^N \sum_{\omega \in \Omega} |A_j(\omega)|} \). We further denote the collection of hazard rates and transition probabilities \( q = (q_j(\omega, a_j))_{\omega \in \Omega, j=1,\ldots,N, a_j \in A_j(\omega)} \) and \( p = (p_j(\omega' | \omega, a_j))_{\omega \in \Omega, j=1,\ldots,N, a_j \in A_j(\omega)} \).

A stationary Markovian strategy for player \( i \) is a function \( \sigma_i : \Omega \rightarrow \cup_{\omega \in \Omega} \mathbb{P}(A_i(\omega)) \) with \( \sigma_i(\omega) \in \mathbb{P}(A_i(\omega)) \) for all \( \omega \), where \( \mathbb{P}(A_i(\omega)) \) is the set of probability distributions over \( A_i(\omega) \). Because \( J \) is a partition of \( \{1, 2, \ldots, N\} \), Assumption 1 ensures that player \( i \) always has the move in conjunction with the same rivals. Hence, while the state of the system comprises both the physical state \( \omega \) and the protocol state \( J \), it suffices to consider \( \Omega \) as the domain of \( \sigma_i \). We use \( \sigma_i(a_i \mid \omega) \) to denote the probability that action \( a_i \in A_i(\omega) \) is played in state \( \omega \).

From hereon, we denote by \( \Sigma_i \) the set of stationary Markovian strategies for player \( i \) and \( \Sigma = \prod_{i=1}^N \Sigma_i \) the set of strategy profiles. To account for mixed strategies, we extend the flow payoff \( u_{i,j}(\omega, \sigma_j(\omega)) = \sum_{a_j \in A_j(\omega)} u_{i,j}(\omega, a_j) \sigma_j(a_j \mid \omega) \) and transition probability

\[
\Pr^\Delta (\omega' | \omega, J, \sigma_j(\omega)) = \sum_{a_j \in \prod_{i \in J} A_i(\omega)} \left( \Pr^\Delta (\omega' | \omega, J, a_j) \prod_{j \in J} \sigma_j(a_j \mid \omega) \right).
\]

A profile of stationary Markovian strategies \( \sigma = (\sigma_i)_{i=1}^N \) is a stationary Markov perfect equilibrium if it is a subgame perfect equilibrium of the game \( \Gamma \). The set of Markov perfect equilibria of the game \( \Gamma \) is denoted \( \text{Equil}(\Gamma) \). This set is nonempty (Shapley 1953)\(^3\).

Our main interest is to compare equilibrium behavior under different protocols of moves. Assumption 1 further ensures that a player’s action brings about the same payoffs and chances of changing the state in the two models. To see this, contrast a game with simultaneous moves \( \Gamma \) with a game with alternating moves \( \tilde{\Gamma} \). In the game with simultaneous moves \( \Gamma \), player \( j \) takes an action \( a_j \) every \( \Delta \) units of time, yielding the payoff \( u_{i,j}(\omega, a_j) \Delta \) and the hazard rate \( q_j(\omega, a_j) \Delta \) (neglecting the higher-order term \( \mathcal{O}(\Delta^2) \)). Over a stretch of \( N\Delta \) units of time, the action \( a_j \) thus yields the payoff \( u_{i,j}(\omega, a_j) N \Delta \) and the hazard rate \( q_j(\omega, a_j) N\Delta \). In the game with alternating moves \( \tilde{\Gamma} \), in contrast, player \( j \) has the move only once every \( N\Delta \) units of time. According to Assumption 3 if player \( j \) takes an action \( a_j \), then this yields the

\(^3\)Shapley (1953) establishes existence for dynamic stochastic games with simultaneous moves. To apply his result, we view the game \( \Gamma \) as a dynamic stochastic game with simultaneous moves in which the players that do not have the move have no impact on per-period payoffs and state-to-state transitions.
payoff $[\mathcal{J}] u_{i,j}(\omega, a_j) \Delta = N u_{i,j}(\omega, a_j) \Delta$ and the hazard rate $[\mathcal{J}] q_{j}(\omega, a_j) \Delta = N q_{j}(\omega, a_j) \Delta$. Hence, per-period payoffs and state-to-state transitions in the game with alternating moves $\Gamma$ are comparable to those in the game with simultaneous moves $\Gamma$.\footnote{Instead of scaling by the number of elements of the partition $\mathcal{J}$ in Assumption 3, we can assume that interactions occur at time $t = 0, \Delta/|\mathcal{J}|, 2\Delta/|\mathcal{J}|, \ldots$. This alternative formulation ensures that a player has the move on average once every $\Delta$ units of time. Our results immediately carry over.}

2.1 Examples

In the remainder of this section we discuss how prominent examples of dynamic stochastic games from the literature can be cast as special cases of our model.

Example 1 (Entry Games and R&D Races) Consider $N = 2$ firms that may enter a new market. To enter the market, firm $i$ must complete $K$ steps. For example, to build a cement plant and enter the market, a firm needs to find a location, design the plant, obtain environmental permits, negotiate with contractors, etc. Alternatively, consider an R&D race in which a firm gradually discovers an invention and obtains a patent through a series of intermediate steps (Fudenberg, Gilbert, Stiglitz, and Tirole 1983, Grossman and Shapiro 1987, Harris and Vickers 1987).

Let $K \geq 1$ be the number of required steps and $\omega_i \in \Omega_i = \{0, 1, \ldots, K\}$ the number of steps that firm $i$ has already completed. The state of the game is $\omega = (\omega_1, \omega_2) \in \Omega_1 \times \Omega_2 = \Omega$. To take the next step, firm $i$ can make an investment, denoted by $a_i = 1$, at cost $c_i > 0$. Action $a_i \in A_i(\omega) = \{0, 1\}$ induces the hazard rate

$$q_i(\omega, a_i) = \begin{cases} a_i & \text{if } \omega_i \leq K - 1, \\ 0 & \text{if } \omega_i = K. \end{cases}$$

The transition probability is\footnote{Note that conditional on a jump occurring we specify a transition from $\omega_i = K$ to $\omega'_i = 0$ with probability one. This is immaterial, however, because no jump occurs as $q_i(\omega, a_i) = 0$ if $\omega_i = K$.}

$$p_i(\omega' | \omega, a_i) = \begin{cases} 1 & \text{if } \omega'_i = \omega_i + 1, \omega'_{-i} = \omega_{-i}, \omega_i \leq K - 1, \\ 1 & \text{if } \omega'_i = 0, \omega'_{-i} = \omega_{-i}, \omega_i = K, \\ 0 & \text{otherwise}. \end{cases}$$

Once firm $i$ has completed all steps it enters the new market (or obtains the patent) and, depending on whether its rival has also completed all steps, obtains the monopoly profit $B_i > 0$ or the duopoly profit $b_i$ with $b_i < B_i$. Its flow payoff is

$$u_{i,i}(\omega, a_i) = \begin{cases} B_i - c_i a_i & \text{if } \omega_i = K, \omega_{-i} \leq K - 1, \\ b_i - c_i a_i & \text{if } \omega_i = \omega_{-i} = K, \\ -c_i a_i & \text{otherwise} \end{cases}$$

and $u_{i,j}(\omega, a_j) = 0$ if $j \neq i$. Assumptions 2 and 3 are satisfied.
Lippman and McCardle (1987) study the continuous-time limit of this game as $\Delta \to 0$. Under suitable parameter restrictions, the equilibrium is generically unique and equilibrium behavior exhibits a pattern of increasing dominance. In standard entry games, in contrast, multiple equilibria often arise and potentially complicate empirical research (Bresnahan and Reiss 1990, Berry 1992, Quint and Einav 2005).

Example 2 (Industry Dynamics) Ericson and Pakes (1995) develop a discrete-time model of industry dynamics. In their model and the large literature following it (see Doraszelski and Pakes (2007) for a survey), incumbent firms decide on investment and exit and compete in the product market; potential entrants decide on entry. Depending on the application, firm $i$’s state variable $\omega_i \in \Omega_i$ encodes its current product quality, production capacity, marginal cost, etc. It further encodes whether firm $i$ is currently an incumbent firm that competes in the product market or a potential entrant. The state of the game is $\omega = (\omega_1, \omega_2, \ldots, \omega_N) \in \prod_{i=1}^N \Omega_i = \Omega$.

Incumbent firm $i$ earns a profit $\pi_i(\omega)$ from competing in the product market (price or quantity competition, depending on the application) that, following the literature, we treat as a reduced-form input into the model. While $\pi_i(\omega)$ depends on the current state of the game $\omega$, it does not depend on the current investment and exit decisions. The cost of investment $c_i(\omega, a_i)$ as well as any cost or benefit pertaining to exit are simply added to $\pi_i(\omega)$. As a result, per-period payoffs are separable in the sense of Assumption 3.

In many applications of the Ericson and Pakes (1995) model, firm $i$ has exclusive control over the evolution of $\omega_i$ through its investment, exit, and entry decisions (e.g., Besanko and Doraszelski 2004, Chen 2009, Doraszelski and Markovich 2007). Because the decisions of firm $i$ affect its own state variable but not its rivals’ state variables, the transition probabilities are separable in the sense of Assumption 3. In other applications, there is in addition a common shock such as an increase in the quality of the outside good or an industry-wide depreciation shock (e.g., Berry and Pakes 1993, Gowrisankaran 1999, Fershtman and Pakes 2000, de Roos 2004, Markovich 2008). Assumption 3 accommodates a common shock because transitions effected by “nature” can be subsumed into those effected by one of the players.

Because investment may or may not result in a favorable outcome, transitions due to investment decisions are noisy as required by Assumption 3. Transitions due to entry and exit decisions present a difficulty because in the Ericson and Pakes (1995) model, an incumbent firm can exit the industry for sure and a potential entrant can enter the industry for sure. Doraszelski and Judd (2012) show how to formulate exit and entry with finite hazard rates either by way of exit and entry intensities or by way of randomly drawn, privately observed scrap values and setup costs (as in Doraszelski and Satterthwaite 2010). Their formulation satisfies Assumption 3.

Example 3 (Continuous-Time Stochastic Games with Moves at Random Times) Arcidiacono, Bayer, Blevins, and Ellickson (2015) develop a continuous-time stochastic game in which a player is given the move at random times. Decisions are asynchronous as the probability that more than one player has the move at

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6One may alternatively represent nature by an extra player 0.
a given time is zero. Ambrus and Lu (2015), Ambrus and Ishii (2015), and Calcagno, Kamada, Lovo, and Sugaya (2014) develop closely related continuous-time stochastic games with moves at random times.

Arcidiacono, Bayer, Blevins, and Ellickson (2015) endow player $i$ with a Poisson process with a constant hazard rate $\lambda$. The time between jumps in this process is therefore exponentially distributed. If process $i$ is the first of the $N$ processes to jump, then player $i$ is given the move and chooses an action $a_i$. The state of the game then changes from $\omega$ to $\omega'$ with probability $l_i(\omega' | \omega, a_i)$, with $l_i(\cdot | \omega, a_i) \in P(\Omega)$.

We can formulate this process in our framework by defining the hazard rate $q_i(\omega, a_i) = \lambda(1 - l_i(\omega | \omega, a_i))$ and the transition probability

$$p_i(\omega' | \omega, a_i) = \begin{cases} \frac{1}{(1-l_i(\omega | \omega, a_i))} l_i(\omega' | \omega, a_i) & \text{if } \omega' \neq \omega, \\ 0 & \text{if } \omega' = \omega. \end{cases}$$

Finally, the flow payoff of player $i$ is

$$u_{i,j}(\omega, a_j) = \begin{cases} s_i(\omega) + \lambda \pi_i(\omega, a_i) & \text{if } i = j, \\ 0 & \text{if } i \neq j, \end{cases}$$

where $s_i(\omega)$ is a baseline payoff and $\pi_i(\omega, a_i)$ an additional payoff that player $i$ receives if he is given the move. To account for the likelihood that player $i$ is given the move, $\pi_i(\omega, a_i)$ is multiplied by $\lambda$ in the flow payoff. The flow payoff and transition probability in Arcidiacono, Bayer, Blevins, and Ellickson (2015) clearly conform to Assumptions 2 and 3.

Example 4 (Dynamic Public Contribution Games) Consider $N$ players that contribute towards completing a public project (Marx and Matthews 2000, Compte and Jehiel 2004, Georgiadis 2014). Completing the project requires $K$ steps and $\omega \in \Omega = \{0, 1, \ldots, K\}$ indicates the number of steps that have been completed. Player $i$’s contribution $a_i \in A_i(\omega) \subseteq \mathbb{R}$ induces a hazard rate $q_i(\omega, a_i)$ which is strictly increasing in $a_i$ if $\omega \neq K$, while $q_i(\omega, a_i) = 0$ if $\omega = K$. The transition probability is

$$p_i(\omega' | \omega, a_i) = \begin{cases} 1 & \text{if } \omega' = \omega + 1, \omega \leq K - 1, \\ 1 & \text{if } \omega' = 0, \omega = K, \\ 0 & \text{otherwise}. \end{cases}$$

The public project is completed once state $\omega = K$ is reached and results in flow payoffs $B_i$ for player $i$. The

\footnote{If $l_i(\omega | \omega, a_i) = 1$, then $p_i(\cdot | \omega, a_i)$ can be defined arbitrarily.}
cost of contribution is \(c_i(\omega, a_i)\) for player \(i\). We therefore specify its flow payoff as

\[
u_{i,i}(\omega, a_i) = \begin{cases} B_i - c_i(\omega, a_i) & \text{if } \omega = K, \\ -c_i(\omega, a_i) & \text{otherwise}, \end{cases}
\]

and \(u_{i,j}(\omega, a_j) = 0\) if \(j \neq i\). Assumptions 2 and 3 are satisfied.

**Example 5 (Asynchronously Repeated Games)** Maskin and Tirole (1988a) and Lagunoff and Matsui (1997) study discrete-time repeated games with asynchronous moves. Restricting Example 3 by setting \(N = 2\), \(\Omega = \Omega_1 \times \Omega_2\), \(A_i(\omega) = \Omega_i\), \(l_i(\omega' \mid \omega, a_i) = 1\) if and only if \(a_i = \omega_i'\), and \(u_{i,j}(\omega, a_j) = s_i(\omega)\), we obtain a game in which the state \(\omega_i\) of player \(i\) is simply a record of the last chosen action. This game is similar to the discrete-time repeated games in Maskin and Tirole (1988a) and Lagunoff and Matsui (1997) in that changes in the payoff-relevant state do not occur at the same time. One difference is that in Maskin and Tirole (1988a) and Lagunoff and Matsui (1997) the player who has the move is sure that his rival has the next move, whereas in our model a player may have consecutive chances of changing the payoff-relevant state. Yet, as Lagunoff and Matsui (1997) point out, what matters for their results is that moves are asynchronous, “rather than the specific structure of asynchronous choice” (p. 1473).

### 3 Protocol-Invariance Theorem

Consider the separable dynamic game with noisy transitions \(\Gamma =< \Delta, \mathcal{J}, \mathcal{P}, u, p, q, \rho >\). We are interested in exploring how the set of Markov perfect equilibria \(\text{Equil}(\Gamma)\) of the game \(\Gamma\) changes as we change the protocol of moves \(< \mathcal{J}, \mathcal{P} >\).

We endow the set of all flow payoffs \(u \in \mathbb{R}^N \sum_{i=1}^{N} \sum_{\omega \in \Omega} |A_i(\omega)|\) with the Lebesgue measure and say that a property is generic if it does not hold at most on a closed subset of measure zero. In this case we say that the property holds for almost all \(u \in \mathbb{R}^N \sum_{i=1}^{N} \sum_{\omega \in \Omega} |A_i(\omega)|\).

The first main result of the paper is a protocol-invariance theorem:

**Theorem 1 (Protocol-Invariance Theorem)** Fix \(p, q, \) and \(\rho\). For almost all \(u\), all \(< \mathcal{J}, \mathcal{P} >\) and \(< \mathcal{J'}, \mathcal{P} >\), and all \(\varepsilon > 0\), there exists \(\Delta > 0\) such that for all \(\Delta < \Delta\) and \(\sigma \in \text{Equil}(< \Delta, \mathcal{J}, \mathcal{P}, u, p, q, \rho >)\), there exists \(\bar{\sigma} \in \text{Equil}(< \Delta, \mathcal{J'}, \mathcal{P}, u, p, q, \rho >)\) such that \(\|\sigma - \bar{\sigma}\| < \varepsilon\).

In words, for any Markov perfect equilibrium \(\sigma\) of a game with a protocol of moves \(< \mathcal{J}, \mathcal{P} >\), the game with another protocol \(< \mathcal{J'}, \mathcal{P} >\) has a Markov perfect equilibrium \(\bar{\sigma}\) that is arbitrarily close to \(\sigma\) provided that periods are sufficiently short. Theorem 1 thus shows that the set of Markov perfect equilibria of separable dynamic games with noisy transitions is generically almost independent of the protocol of moves.

The intuition for Theorem 1 is best seen by contrasting two protocols of moves. In a game with alternating moves, if player \(i\) has the move, then to choose an action \(a_i\) he must consider the contribution \(u_{i,i}(\omega, a_i)\Delta\)
to his per-period payoff that his action yields and the impact his action has on state-to-state transitions through \( q_i(\omega, a_i)p_i(\omega'|\omega, a_i)\Delta \) (neglecting the higher-order term \( O(\Delta^2) \)). In the game with simultaneous moves, two additional considerations arise. First, player \( i \) must consider how his rivals’ actions change the contribution to his per-period payoff that his action yields and the impact his action has on state-to-state transitions. However, because complementarities between players’ actions and other non-separabilities in per-period payoffs and state-to-state transitions are restricted to the higher-order term \( O(\Delta^2) \), player \( i \) can neglect his rivals’ actions if the period length \( \Delta \) is sufficiently small. Second, player \( i \) must consider the possibility that his rivals’ actions further change the state of the game. The probability that two or more players cause the state to change is, however, negligible if the period length \( \Delta \) is sufficiently small. Assumption 1 finally ensures that irrespective of the protocol of moves all players move with the same frequency over a sufficiently large number of periods. Thus, player \( i \) faces the same tradeoff between current and future payoffs. As a result, provided that periods are sufficiently short, the protocol of moves ceases to matter for equilibrium behavior.

To establish Theorem 1 consider a Markov perfect equilibrium \( \sigma^\Delta = (\sigma_j^\Delta)_{j=1}^N \) of the separable dynamic game with noisy transitions \( \Gamma = < \Delta, J, \mathcal{P}, u, p, q, \rho > \). Let \( V_i^\Delta(\omega, J) \) be the continuation value of player \( i \) if players \( J \in J \) have the move and the state is \( \omega \in \Omega \). The discrete-time Bellman equation is

\[
V_i^\Delta(\omega, J) = u_i^\Delta(\omega, J, \sigma_j^\Delta(\omega)) + \exp(-\rho\Delta) \sum_{\omega' \in \Omega} \sum_{J' \in J} V_i^\Delta(\omega', J') \Pr(J'|J) \Pr(\omega'|\omega, J, \sigma_j^\Delta(\omega)),
\]

where the player discounts payoffs accruing in the subsequent period by \( \exp(-\rho\Delta) \) and \( \sigma_j^\Delta(\omega) = (\sigma_j^\Delta(\omega))_{j \in J} \). Under Assumptions 2 and 3 this becomes

\[
V_i^\Delta(\omega, J) = |J| \sum_{j \in J} u_i(\omega, \sigma_j^\Delta(\omega))\Delta + \exp(-\rho\Delta) \left\{ \sum_{J' \in J} V_i^\Delta(\omega, J') \Pr(J'|J) \left[ 1 - |J| \sum_{j \in J} q_j(\omega, \sigma_j^\Delta(\omega))\Delta \right] \right. \\
+ \left. \sum_{\omega' \neq \omega} \sum_{J' \in J} V_i^\Delta(\omega', J') \Pr(J'|J) \left[ |J| \sum_{j \in J} \varphi_j(\omega'|\omega, \sigma_j^\Delta(\omega))\Delta \right] \right\} + O(\Delta^2), \tag{3.1}
\]

where we use the shorthand notation \( \varphi_j(\omega'|\omega, a_j) = q_j(\omega, a_j)p_j(\omega'|\omega, a_j) \) and \( \varphi_j(\omega'|\omega, \sigma_j(\omega)) = \sum_{a_j \in A_j(\omega)} \varphi(\omega'|\omega, a_j) \).

Let \( V^\Delta = (V_i^\Delta)_{i=1}^N \) be the profile of value functions corresponding to the Markov perfect equilibrium \( \sigma^\Delta \). Consider a sequence \( (\sigma^\Delta, V^\Delta) \) indexed by the period length \( \Delta \). Assuming that \( (\sigma^\Delta, V^\Delta) \rightarrow (\sigma^0, V^0) \) (where convergence is possibly through a subsequence \( \Delta_n \)) and taking the limit of equation (3.1) as \( \Delta \rightarrow 0 \), we deduce that

\[
V_i^0(\omega, J) = \sum_{J' \in J} V_i^0(\omega, J') \Pr(J'|J).
\]

Stacking this equation for all \( J \in J \) yields the system of linear equations \( Px = x \), where \( x \) is a \(|J|\)-dimensional
column vector with entries $V^0_i(\omega, J)$. Assumption 1 implies that $V^0_i(\omega, J) = V^0_i(\omega, J')$ for all $J, J' \in J$. This means that in equilibrium the continuation value of player $i$ is almost independent of the identity of the players who have the move and equals $V^0_i(\omega)$: having the move does not imply a higher or lower payoff. From hereon, let $V^0_i : \Omega \to \mathbb{R}$ be the value function of player $i$ and $V^0 = (V^0_i)_{i=1}^N$ be the profile of value functions in the limit as $\Delta \to 0$.

Equation (3.1) can equivalently be written as

$$
\frac{1}{\Delta} V^\Delta_i(\omega, J) - \frac{\exp(-\rho \Delta)}{\Delta} \sum_{J' \in J} V^\Delta_i(\omega, J') \Pr(J'|J) = |J| \sum_{j \in J} u_{i,j}(\omega, \sigma^\Delta_j(\omega))
$$

$$
+ \exp(-\rho \Delta) |J| \sum_{j \in J} \left( \sum_{\omega' \neq \omega} \sum_{J' \in J} V^\Delta_i(\omega', J') \Pr(J'|J) \varphi_j(\omega'|\omega, \sigma^\Delta_j(\omega)) - \sum_{J' \in J} V^\Delta_i(\omega, J') \Pr(J'|J) q_j(\omega, \sigma^\Delta_j(\omega)) \right) + \mathcal{O}(\Delta).
$$

Summing this equation for all $J \in J$ yields

$$
\frac{1 - \exp(-\rho \Delta)}{\Delta} \sum_{J \in J} V^\Delta_i(\omega, J) = |J| \sum_{j \in J} u_{i,j}(\omega, \sigma^\Delta_j(\omega))
$$

$$
+ \exp(-\rho \Delta) |J| \sum_{j \in J} \left( \sum_{\omega' \neq \omega} \sum_{J' \in J} V^\Delta_i(\omega', J') \Pr(J'|J) \varphi_j(\omega'|\omega, \sigma^\Delta_j(\omega)) - \sum_{J' \in J} V^\Delta_i(\omega, J') \Pr(J'|J) q_j(\omega, \sigma^\Delta_j(\omega)) \right) + \mathcal{O}(\Delta^2),
$$

where we use the fact that, under Assumption 1, $\sum_{J \in J} \Pr(J'|J) = 1$. Taking the limit as $\Delta \to 0$ yields the continuous-time Bellman equation

$$
\rho V^0_i(\omega) = \sum_{j \in J} \sum_{\omega'} \sum_{J' \in J} u_{i,j}(\omega, \sigma^0_j(\omega)) + \sum_{J' \in J} \varphi_j(\omega'|\omega, \sigma^0_j(\omega)) - V^0_i(\omega) q_j(\omega, \sigma^0_j(\omega))
$$

$$
= \sum_{j=1}^N u_{i,j}(\omega, \sigma^0_j(\omega)) + \sum_{j=1}^N \left( V^0_i(\omega') - V^0_i(\omega) \right) \varphi_j(\omega'|\omega, \sigma^0_j(\omega)),
$$

(3.2)

where the last inequality uses that, under Assumption 1, there exists a unique $J \in J$ such that $j \in J$ and the fact that $\sum_{\omega' \neq \omega} p_j(\omega'|\omega, a_i) = 1$. Importantly, condition (3.2) is independent of the protocol of moves $< J, P >$ used to pass from discrete to continuous time.

The discrete-time optimality condition for a period length of $\Delta$ is

$$
\sigma^\Delta(a_i | \omega) > 0 \Rightarrow a_i \in \arg \max_{a_i \in K_i(\omega)} u^\Delta_i(\omega, J, a_i, \sigma^\Delta_{i,i}(\omega)) + \exp(-\rho \Delta) \sum_{\omega' \in \Omega} \sum_{J' \in J} V^\Delta_i(\omega', J') \Pr(J'|J) \Pr^\Delta(\omega'|\omega, J, a_i, \sigma^\Delta_{i,i}(\omega)).
$$

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8 The vector $y = (1, \ldots, 1)'$ is always a right eigenvector since $P$ is stochastic. Since $P$ is irreducible, the Perron-Frobenious theorem implies that both the left and right eigenvectors associated to the eigenvalue 1 are unique, up to scalar multiplication. It follows that for any solution to the system $Px = x$, $x_i = x_j$ for all $i$ and $j$.

9 Recall that if the transition matrix $P$ is irreducible and its unique stationary distribution is uniform on $J$, then $P$ is doubly stochastic.
Since $\sigma^\Delta \to \sigma^0$, $\sigma^0_\Delta(a_i | \omega) > 0$ implies $\sigma^\Delta_\Delta(a_i | \omega) > 0$ if the period length $\Delta$ is sufficiently small. Dividing by $\Delta$, rearranging terms, and taking the limit as $\Delta \to 0$ (as we did in the previous paragraph) thus yields the continuous-time optimality condition

$$\sigma^0_i(a_i | \omega) > 0 \Rightarrow a_i \in \arg \max_{\tilde{a}_i \in \tilde{A}_i(\omega)} \left( V^0_i(\omega') - V^0_i(\omega) \right) \varphi_i(\omega' | \omega, \tilde{a}_i).$$ \tag{3.3}$$

Condition (3.3) is again independent of the protocol of moves. It formalizes the intuition that player $i$ faces the same tradeoff between current and future payoffs under any protocol of moves $< J, P >$ and that this tradeoff is not directly affected by his rivals’ actions. Holding fixed the value of continued play, the strategic situation that the $N$ players face in a given state is thus akin to $N$ independent optimization problems.

Conditions (3.2) and (3.3) are the limit as $\Delta \to 0$ of the equilibrium conditions for the separable dynamic game with noisy transitions $\Gamma$. We provide economic interpretations of these conditions in Section 3.1. Here we merely observe that they impose restrictions on the limit strategy and continuation value profiles $(\sigma^0, V^0)$. Noting that the limit conditions (3.2) and (3.3) may admit multiple solutions and that $V^0$ is entirely determined by $\sigma^0$ using condition (3.2), we denote the set of strategy profiles $\sigma^0 \in \Sigma$ satisfying condition (3.3) as $\text{Equil}^0(<u, p, q, \rho>)$. This set does not depend on the protocol of moves $< J, P >$ used to pass to the limit.

We summarize the above discussion in a lemma:

**Lemma 1** Consider a sequence $(\sigma^\Delta)$ with $\sigma^\Delta \in \text{Equil}(< \Delta, J, P, u, p, q, \rho >)$. If $\sigma^\Delta \to \sigma^0$, then $\sigma^0 \in \text{Equil}^0(<u, p, q, \rho>)$.

Unfortunately, Theorem 1 cannot be established by simply taking the limit of the equilibrium conditions as $\Delta \to 0$ because conditions (3.2) and (3.3) may admit multiple solutions. In other words, Theorem 1 is not implied by Lemma 1 when the limit system has several solutions. To overcome this difficulty, we use tools from differential topology to analyze the limit conditions. We first restrict attention to solutions $\sigma^0 \in \text{Equil}^0(< u, p, q, \rho >)$ that are regular. The formal definition of regularity is in the Appendix; here we just note that $\sigma^0$ is regular if it is strict, i.e., if the maximization problem in condition (3.3) admits a unique solution. Intuitively, a regular solution $\sigma^0$ can be approximated by a Markov perfect equilibrium of a separable dynamic game with noisy transitions and an arbitrary protocol of moves if the period length $\Delta$ is sufficiently small. The key technical point is that for almost all flow payoffs $u$, the restriction to regular solutions is without loss of generality.

**Lemma 2** Fix $p, q,$ and $\rho$. For almost all $u$, all $\sigma^0 \in \text{Equil}^0(< u, p, q, \rho >)$, all $< J, P >$, and all $\varepsilon > 0$, there exists $\Delta > 0$ such that for all $\Delta < \Delta$, there exists $\sigma \in \text{Equil}(< \Delta, J, P, u, p, q, \rho >)$ such that $||\sigma - \sigma^0|| < \varepsilon$.

The proof of Lemma 2 draws on ideas in Harsanyi (1973a, 1973b) and Doraszelski and Escobar (2010). We note that the results in Doraszelski and Escobar (2010) do not directly apply because separable dynamic games with noisy transitions restrict per-period payoffs and state-to-state transitions and are therefore a
subset of measure zero of the dynamic stochastic games considered in Doraszelski and Escobar (2010).

Lemmas 1 and 2 combine to yield Theorem 1. Our proofs also show that Equil\(^R(<u,p,q,\rho>)\) consists of a finite number of isolated solutions. This generalizes results on the generic finiteness of the set of Markov perfect equilibria in Haller and Lagunoff (2000) and Doraszelski and Escobar (2010) to continuous-time stochastic games.

### 3.1 Interpretations of Limit Conditions

We offer two economic interpretations of the limit conditions (3.2) and (3.3). First, they can be interpreted as the equilibrium conditions for a continuous-time stochastic game along the lines of Doraszelski and Judd (2012). In this game, the state follows a continuous-time Markov process that players can influence through their actions. Properly defining mixed strategies in continuous time is, however, subtle because it requires working with a continuum of independent and identically distributed random variables that satisfy a law of large numbers. As in Bolton and Harris (1999), we can use time to “purify” these strategies and avoid the continuum of independent and identically distributed random variables. Beyond this observation, we follow the literature and alert the reader that a rigorous foundation for mixed strategies in continuous time is an open problem (Bolton and Harris 1999, Faingold and Sannikov 2011).

Second, the limit conditions (3.2) and (3.3) can be interpreted as the equilibrium conditions for a dynamic stochastic game with random moves. The following construction, known as uniformization (Serfozo 1979), is adapted from single-agent decision problems. Fix any \(B > N\max_{j=1,\ldots,N,\omega\in\Omega, a_j\in A_j(\omega)}q_j(\omega, a_j)\). Define the per-period payoff \(\tilde{u}_{i,j}(\omega, a_j) = \frac{N}{\rho + B}u_{i,j}(\omega, a_j)\), the discount factor \(\beta = \frac{B}{\rho + B} < 1\), and the transition probability

\[
\tilde{\varphi}_j(\omega' | \omega, a_j) = \begin{cases} 
\frac{N}{B} \varphi_j(\omega' | \omega, a_j) & \text{if } \omega' \neq \omega, \\
1 - \frac{N}{B} q_j(\omega, a_j) & \text{if } \omega' = \omega.
\end{cases}
\]

Note that \(\tilde{\varphi}_j(\cdot | \omega, a_j) \in P(\Omega)\) by construction of \(B\). Now formulate a dynamic stochastic game with random moves in which in any period one player \(j \in \{1, \ldots, N\}\) is randomly and uniformly selected to make a decision \(a_j \in A_j(\omega)\). Each player strives to maximize the expected net present value of his stream of payoffs and discounts future payoffs using the discount factor \(\beta\). Denote by Equil\(^R(\tilde{u}, \tilde{\varphi}, \beta)\) the set of Markov perfect equilibria of this game.

The following proposition shows that the Markov perfect equilibria of the dynamic stochastic game with random moves constructed above are the solutions of the limit conditions (3.2) and (3.3):

**Proposition 1** Equil\(^R(\tilde{u}, \tilde{\varphi}, \beta)\) = Equil\(^R(<u,p,q,\rho>)\).

The proof of Proposition 1 is simple and illustrative. The equilibrium conditions for \(\sigma \in \text{Equil}^R(<\tilde{u}, \tilde{\varphi}, \beta>)\) are

\[
V_i(\omega) = \sum_{j=1}^{N} \frac{1}{N} \left( \tilde{u}_{i,j}(\omega, \sigma_j(\omega)) + \beta \sum_{\omega' \in \Omega} V_i(\omega) \tilde{\varphi}_j(\omega' | \omega, \sigma_j(\omega)) \right)
\]  
(3.4)
and

$$\sigma_i(a_i \mid \omega) > 0 \Rightarrow a_i \in \arg \max_{a_i \in A_i(\omega)} \tilde{u}_{i,i}(\omega, \tilde{a}_i) + \beta \sum_{\omega' \in \Omega} V_i(\omega') \tilde{\varphi}_i(\omega' \mid \omega, \tilde{a}_i).$$  \hspace{1cm} (3.5)$$

These conditions can be equivalently written as

$$V_i(\omega) = \frac{1}{N} \left\{ \frac{N}{\rho + B} u_{i,j}(\omega, \sigma_j(\omega)) + B \left( \sum_{\omega' \neq \omega} V_i(\omega') \frac{N}{B} \varphi_j(\omega' \mid \omega, \sigma_j(\omega)) + (1 - \frac{N}{B} q_{j}(\omega, \sigma_j(\omega)) \right) \right\}$$

and

$$\sigma_i(a_i \mid \omega) > 0 \Rightarrow a_i \in \arg \max_{a_i \in A_i(\omega)} \frac{N}{\rho + B} u_{i,i}(\omega, \tilde{a}_i) + B \left( \sum_{\omega' \neq \omega} V_i(\omega') \frac{N}{B} \varphi_i(\omega' \mid \omega, \tilde{a}_i) + (1 - \frac{N}{B} q_i(\omega, \tilde{a}_i) V_i(w)) \right).$$

Rearranging terms, the limit conditions (3.2) and (3.3) are therefore identical to the equilibrium conditions (3.4) and (3.5) for the dynamic stochastic game with random moves constructed above.

While dynamic stochastic games with random moves are sparsely used, several important papers study repeated games with alternating moves. For example, Maskin and Tirole (1988a) explore a repeated Bertrand game with alternating moves and show how Edgeworth cycles can arise. Lagunoff and Matsui (1997) show how players can coordinate on the efficient outcome in a dynamic coordination game with alternating moves. These results are driven by the fact that a player remains committed to his previously chosen action over a stretch of time. The dynamic stochastic game with random moves constructed above shares this feature. Similarly rich dynamic phenomena thus appear in the continuous-time stochastic game that we obtain as we pass to the limit and, by Theorem 1, in separable dynamic games with noisy transitions and arbitrary protocols of moves provided that periods are sufficiently short.

3.2 Discussion of Assumptions

To illustrate the tightness of our assumptions, we provide a series of examples showing that protocol invariance may fail if any one of them is relaxed.

Example 6 (Separability) The literature provides a number of examples in which complementarities between players’ actions and other non-separabilities in per-period payoffs preclude protocol invariance. Our example with non-separable per-period payoffs is inspired by Lagunoff and Matsui (1997) and Wen (2002). In the Online Appendix we present a closely related example with non-separable state-to-state transitions.

Consider a coordination game with the following payoff matrix:

<table>
<thead>
<tr>
<th></th>
<th>E</th>
<th>O</th>
</tr>
</thead>
<tbody>
<tr>
<td>E</td>
<td>2,2</td>
<td>0,0</td>
</tr>
<tr>
<td>O</td>
<td>0,0</td>
<td>1,1</td>
</tr>
</tbody>
</table>

Denote by \(b(a) = (b_1(a), b_2(a))\) the payoff profile given the action profile \(a = (a_1, a_2) \in \{E, O\}^2\).
We construct a dynamic stochastic game with a trivial state space $|\Omega| = 1$ (that we omit along with specifying the transition probability) and contrast the set of Markov perfect equilibria under simultaneous and alternating moves. In the game with simultaneous moves, the per-period payoff of player $i$ is $u^\Delta_i(\omega, \{1,2\}, a) = b_i(a)\Delta$. Irrespective of the period length $\Delta$, there are two Markov perfect equilibria, namely $\sigma_1(E) = \sigma_2(E) = 1$ and $\sigma_1(O) = \sigma_2(O) = 1$.

In the game with alternating moves, in violation of Assumption 3, the per-period payoff of player $i$ is $u^\Delta_i(\omega, \{1\}, a_1) = 0$ and $u^\Delta_i(\omega, \{2\}, a) = b_i(a)\Delta$, meaning that payoffs “materialize” after player 2 moves. Since player 1’s action $a_1$ is payoff relevant for player 2, a Markovian strategy for player 2 includes $a_1$ as a state variable. Irrespective of the period length $\Delta$, the unique Markov perfect equilibrium is $\sigma_1(E) = 1$ and $\sigma_2(a_1 \mid a_1) = 1$ for $a_1 \in \{E,O\}$.

**Example 7 (Noisy Transitions)** Consider the entry game in Example 1 with $K = 1$. We further restrict $b_i < 0$ so that a duopolist incurs a loss.

We change Example 1 by assuming that if firm $i$ takes action $a_i = 1$, then its state changes for sure from $\omega_i = 0$ to $\omega_i' = 1$. Irrespective of the protocol of moves, given a set of players $J \subseteq \{1,2\}$ who have the move, the transition probability takes the form

$$
\Pr^\Delta(\omega' \mid \omega, J, a_J) = \begin{cases} 
1 & \text{if } \omega_j = a_j, \omega = (0,0), \\
1 & \text{if } \omega' = \omega, \omega_i = 1 \text{ for some } i, \\
0 & \text{otherwise},
\end{cases}
$$

and does not satisfy Assumption 3. In the game with alternating moves, if $\Delta$ is sufficiently small, then the unique Markov perfect equilibrium outcome is that the firm that moves first enters whereas its rival never enters. In contrast, in the game with simultaneous moves, there exists a Markov perfect equilibrium in which both firms enter with positive probability.$^{10}$

Finally, we show that protocol invariance does not extend beyond Markov perfect equilibria to more general equilibrium concepts. For a separable dynamic game with noisy transitions $\Gamma = < \Delta, J, P, u, p, q, \rho >$, we say that a strategy $\sigma^T_i$ for player $i$ has finite memory $T \geq 0$ if $\sigma^T_i(h) = \sigma^T_i(h)$ for any two histories $h$ and $\hat{h}$ (perhaps of different length) that coincide in the current state and the outcomes of the previous $T$ rounds of interactions between players. If $T = 0$, then we recover the definition of a stationary Markovian strategy in Section 2.

$^{10}$Iskhakov, Rust, and Schjerning (2016) study an investment game that satisfies Assumption 3. Contrary to Assumption 2, however, by investing a firm can guarantee increasing its capital stock by exactly one unit. Iskhakov, Rust, and Schjerning (2016) show that their game with alternating moves has a unique Markov perfect equilibrium whereas they find an enormous number of equilibria in the game with simultaneous moves. There is also a well-known literature drawing subtle connections between discrete- and continuous-time stochastic games with infinite hazard rates (Fudenberg and Tirole 1985, Simon and Stinchcombe 1989). Fudenberg and Tirole (1985), in particular, show that passing to the limit is non-trivial in games with infinite hazard rates even if strategies are restricted to be Markovian.
Example 8 (Markov Perfect Equilibrium) Consider a partnership game and construct a separable dynamic game with noisy transitions and a trivial state space (that we again omit). There are $N = 2$ players. The set of actions of player $i$ is $A_i = \{0, 1\}$, his flow payoff is

$$u_{i,j}(a_j) = \begin{cases} -a_j & \text{if } j = i, \\ 2a_j & \text{if } j \neq i, \end{cases}$$

and the discount rate is $\rho$. Irrespective of the protocol of moves $<J,P>$ and the period length $\Delta$, the unique Markov perfect equilibrium of this game is $\sigma_1(0) = \sigma_2(0) = 1$ and has players repeating $(0,0)$.

We show that this not the case for strict subgame perfect equilibria in finite memory strategies. In the game with simultaneous moves, consider a finite memory strategy $\sigma^T_i$ with $T \geq 1$ for player $i$ such that player $i$ chooses $a_{1,t} = 1$ in period $t$ if $t = 0$ or if the players have chosen the same action over the last $\min\{T,t\}$ rounds: $a_{1,t} = a_{1,\tilde{t}}$ for all $\tilde{t} \in \{t-1, \ldots, t-\min\{T,t\}\}$. The strategy profile $\sigma^T = (\sigma^T_1, \sigma^T_2)$ is a strict subgame perfect equilibrium if $1 - e^{-\rho \Delta T} > e^{\rho \Delta}$. This condition holds if $T \geq 2$ and the period length $\Delta$ is sufficiently small. Hence, there exists a strict subgame perfect equilibrium in finite memory strategies in which players repeatedly play $(1,1)$.

Turning to the game with alternating moves, consider a finite memory strategy $\sigma^T_i$ with $T \geq 1$ for player $i$. We argue that for any strategy profile $\sigma^T$ to be a strict subgame perfect equilibrium it must be a Markov perfect equilibrium. Hence, the unique strict subgame perfect equilibrium in finite memory strategies is the Markov perfect equilibrium in which players repeat $(0,0)$.

To complete the argument, suppose player $i$ moves in round $t$. Because player $-i$ moves after player $i$ and conditions his decision on the previous $T$ periods of interactions, the continuation value of player $i$ depends on $a^t_1$ and the previous $T - 1$ periods of interactions. The current payoff of player $i$ moreover depends only on $a^t_1$. Since $\sigma^T$ is a strict subgame perfect equilibrium, the maximization problem of player $i$ admits a unique solution which depends, at most, on the previous $T - 1$ rounds of interactions. This means that $\sigma^T_i$ actually conditions on the previous $T - 1$ periods of interactions. Continuing iteratively, we deduce that the strategy profile $\sigma^T$ cannot condition on any previous interactions.

4 Complexity Theorem

The limit conditions (3.2) and (3.3) are analytically more tractable than the equilibrium conditions for the separable dynamic game with noisy transitions from which they are derived. This allows us to characterize the solutions to the limit conditions in more detail and to use Theorem 1 to extend this characterization to separable dynamic games with noisy transitions and frequent moves.

The following lemma shows that the solutions to the limit conditions (3.2) and (3.3) have a remarkably simple structure.
Lemma 3 Fix $p$, $q$, and $\rho$. For almost all $u$, all $\sigma \in \text{Equil}^0(<u,p,q,\rho>)$, all $i$, and all $\omega$, $\sigma_i(\cdot | \omega)$ puts positive probability on at most $N$ actions.

The proof of Lemma 3 adapts and extends some results for so-called additive-reward, additive-transition dynamic stochastic games (Raghavan, Tijs, and Vrieze 1985). While Lemma 3 bounds the complexity of mixed strategies, in the Online Appendix we provide an example showing that the limit conditions (3.2) and (3.3) in general may not admit a solution in pure strategies.

The proof of Lemma 3 proceeds by contradiction. Note that a player’s strategy determines his rivals’ expected continuation values in a given state as a convex combination of the finite number of the continuation values associated with the actions that the player chooses with positive probability. Because of Assumption 3, some actions can be dispensed with if the player randomizes over more than $N$ actions without affecting his rivals’ expected continuation values. We can therefore construct a new solution to the limit conditions (3.2) and (3.3) which puts positive probability on a smaller number of actions. But this new solution cannot be regular because at least one player uses some but not all of his best replies. Lemma 3 follows by noting that generically all solutions are regular, as established in Section 3.

Lemma 3 and Theorem 1 combine to yield the second main result of the paper:

Theorem 2 (Complexity Theorem) Fix $p$, $q$, and $\rho$. For almost all $u$, there exists $\bar{\Delta} > 0$ such that for all $\Delta < \bar{\Delta}$, for any $\sigma^\Delta \in \text{Equil}(<\Delta,J,P,u,p,q,\rho>)$, all $i$, and all $\omega$, $\sigma^\Delta_i(\cdot | \omega)$ puts positive probability on at most $N$ actions.

We note that Theorem 2 applies although the separable dynamic game with noisy transitions $<\Delta,J,P,u,p,q,\rho>$ may admit complementarities between players’ actions and other non-separabilities in the higher-order term $O(\Delta^2)$ that make a direct proof difficult. This again shows the usefulness of the limit conditions (3.2) and (3.3).

Theorem 2 may be contrasted with a result for discrete-time repeated games with alternating moves in Haller and Lagunoff (2010). Haller and Lagunoff (2010) show that these games generically do not possess completely mixed Markov perfect equilibria. As Example 5 shows, asynchronously repeated games are a special case of our model. When $|A_i(\omega)| \geq 3$ for all $i$ and all $\omega$, our Theorem 2 is sharper than Haller and Lagunoff’s (2010) result because it shows that for a broader class of two-player dynamic models, irrespective of the protocol of moves, all Markov perfect equilibria are not completely mixed and put (strictly) positive weight on at most two actions per state.\textsuperscript{11}

\textsuperscript{11}To be clear, the Haller and Lagunoff’s (2010) result is not a corollary to Theorem 2 because of the differences between the models discussed in Example 5.
5 Applications and Extensions

5.1 Justification of Markov Perfect Equilibria

We apply our main results to provide a new justification for focusing on Markov perfect equilibria in a class of dynamic stochastic games. Provided that periods are sufficiently short and a robustness requirement is imposed, we show that the set of Markov perfect equilibrium payoffs in separable dynamic games with noisy transitions and simultaneous moves almost coincides with the set of payoffs that can be attained under more general equilibrium concepts.

We focus on strict subgame perfect equilibria in finite memory strategies. By definition, a strict equilibrium involves only pure strategies. Strictness is a natural robustness requirement. In repeated public monitoring games only strict subgame perfect equilibria in finite memory strategies are robust to private monitoring (Mailath and Morris 2002, Mailath and Samuelson 2006, Bhaskar, Mailath, and Morris 2013). Equilibria that fail to be strict are also fragile to perturbations of payoffs and information (Harsanyi 1973a, Harsanyi 1973b, Doraszelski and Escobar 2010).

As we change the protocol of moves \( \langle J, P \rangle \) of a separable dynamic game with noisy transitions \( \Gamma = \langle \Delta, J, P, u, p, q, \rho \rangle \), the sets of histories change and are therefore difficult to compare. To circumvent this difficulty, we explore how the set of payoff profiles \( \text{Payoffs}^F(\Gamma) \subseteq \mathbb{R}^N \) associated with strict subgame perfect equilibria in finite memory strategies changes as we change the protocol of moves. We also define the set of payoff profiles \( \text{Payoffs}^M(\Gamma) \subseteq \mathbb{R}^N \) corresponding to the set of Markovian perfect equilibria \( \text{Equil}(\Gamma) \).

Let \( \Gamma_{sim} = \langle \Delta, J_{sim}, P_{sim}, u, p, q, \rho \rangle \) denote a separable dynamic game with noisy transitions under a protocol of simultaneous moves \( \langle J_{sim}, P_{sim} \rangle \), with \( J_{sim} = \{\{1, \ldots, N\}\} \). We say that the payoff profile \( v \in \text{Payoffs}^F(\Gamma_{sim}) \) is approachable if for all \( \epsilon > 0 \) there exists some protocol of asynchronous moves \( \langle J^{asy}, P^{asy} \rangle \), with \( J^{asy} = \{\{1\}, \{2\}, \ldots, \{N\}\} \), and a payoff profile \( w \in \text{Payoffs}^F(\langle \Delta, J^{asy}, P^{asy}, u, p, q, \rho \rangle) \) such that \( \|v - w\| < \epsilon \). In words, focusing on strict subgame perfect equilibria in finite memory strategies, an equilibrium payoff profile of the game with simultaneous moves is approachable if there exists a nearby equilibrium payoff profile of the game for some asynchronous protocol of moves. An approachable equilibrium payoff profile can therefore be obtained as a limit of equilibrium payoffs of games with asynchronous moves.

The following proposition shows that an approachable equilibrium payoff profile of the game with simultaneous moves almost coincides with a payoff profile corresponding to a Markov perfect equilibrium provided that periods are sufficiently short:

**Proposition 2** Fix \( p, q, \) and \( \rho \). For almost all \( u \), and all \( \epsilon > 0 \), there exists \( \bar{\Delta} > 0 \) such that for all \( \Delta < \bar{\Delta} \), if \( v \in \text{Payoffs}^F(\Gamma_{sim}) \) is approachable, then there exists \( w \in \text{Payoffs}^M(\Gamma_{sim}) \) such that \( \|v - w\| < \epsilon \).

Proposition 2 shows that generically any approachable equilibrium payoff profile of a separable dynamic game with noisy transition and simultaneous moves can be approximated by Markov perfect equilibrium payoff profile provided that periods are sufficiently short. This means that there is virtually no loss in restricting
attention to Markov perfect equilibria and thus Proposition 2 provides a rationale for focusing on Markov perfect equilibria in separable dynamic games with noisy transitions and simultaneous moves.

To prove Proposition 2, we build on related results for dynamic stochastic games with asynchronous moves by Bhaskar and Vega-Redondo (2002) and Bhaskar, Mailath, and Morris (2009) and combine them with our Theorem 1. The proof of Proposition 2 draws on the insight from Example 8 that while some payoff profiles can be attained with strict subgame perfect equilibria in finite memory strategies when moves are simultaneous, these payoff profiles cannot be attained when moves are alternating.

Proposition 2 complements several arguments in favor of Markov perfect equilibria given for a variety of dynamic models (Maskin and Tirole 2001, Bhaskar and Vega-Redondo 2002, Bhaskar, Mailath, and Morris 2013, Bohren 2014). Our approachability restriction is conceptually similar to the purifiability restriction in Bhaskar, Mailath, and Morris (2013) in that both are robustness requirements: approachability says that equilibrium payoffs should survive changes in the protocol of moves, whereas purifiability says that equilibrium strategies should survive the introduction of private information. We show that only Markov perfect equilibria are approachable in our separable dynamic games with noisy transitions and simultaneous moves, whereas Bhaskar, Mailath, and Morris (2013) show that only Markov perfect equilibria are purifiable in dynamic stochastic games with asynchronous moves.

Proposition 2 also puts limits on possible extensions of Theorem 1. Indeed, by showing that an equilibrium payoff that is robust to alternative specifications of the protocol of moves must be a Markov perfect equilibrium payoff, Proposition 2 implies that restricting to Markov perfect equilibria is not only sufficient (as shown in Theorem 1) but also necessary for a protocol-invariance theorem.

5.2 Computation of Markov Perfect Equilibria

Dynamic stochastic games are often not very tractable analytically and thus call for the use of numerical methods. Our main results have a number of implications for computing Markov perfect equilibria.

First, Doraszelski and Judd (2007) show that the computational burden can vary by orders of magnitude with the protocol of moves. For separable dynamic games with noisy transitions and frequent moves, protocol invariance justifies imposing the protocol of moves that is most convenient from a computational perspective.

Second, Assumption 3 facilitates numerically solving the limit conditions (3.2) and (3.3) that arise as we pass from discrete to continuous time. Recall that holding fixed the value of continued play, the strategic situation that the $N$ players face in a given state is akin to $N$ independent optimization problems. Without Assumption 3, the strategic situation is, in contrast, akin to a normal-form game with $N$ players that is harder to solve than $N$ independent optimization problems (see McKelvey and McLennan (1996) and the

---

Note, however, that Proposition 2 applies only when strategies have finite memory. In the tightly specified model in Example 8 under arbitrary protocols of moves there exists a subgame perfect equilibrium with unbounded memory in which $a_t^i = 1$ for all $i$ and after all on-path histories. Without restrictions on strategies, the properties of the set of equilibrium payoffs as $\Delta \to 0$ are generally not well understood in the literature. The existing results consider either the limit $\Delta \to 0$ with simultaneous moves (Peski and Wiseman 2015) or the limit $\rho \to 0$ (Dutta 1995, Yoon 2001, Hörner, Sugaya, Takahashi, and Vieille 2011).
Our main results justify solving the limit conditions (3.2) and (3.3) by showing that their solutions almost coincide with the Markov perfect equilibria of separable dynamic games with noisy transitions and arbitrary protocols of moves provided that periods are sufficiently short.

Third, Doraszelski and Judd (2012) contrast the burden of computing Markov perfect equilibria in discrete- and continuous-time stochastic games with simultaneous moves. They argue that, under widely used laws of motion for the evolution of the state, computing the expectation over successor states \( \omega' \) in a continuous-time stochastic game does not suffer from the curse of dimensionality that plagues the discrete-time stochastic game, and that this can reduce the computational burden by orders of magnitude. Even without Assumption 3, the techniques we develop allow us to clarify the relationship between discrete- and continuous-time stochastic games.

Consider a dynamic stochastic game with noisy transitions and simultaneous moves. The per-period payoff is \( u_i(\omega, \{1, \ldots, N\}, a) = u_i(\omega, a)\Delta + O(\Delta^2) \). Hence, while Assumptions 1 and 2 are satisfied, Assumption 3 is not. Overloading notation, let \( \text{Equil}(\langle \Delta, u, p, q, \rho \rangle) \) be the set of Markov perfect equilibria of this game. Analogously, let \( \text{Equil}^0(\langle u, p, q, \rho \rangle) \) be the set of solutions to the limit conditions (3.2) and (3.3).

**Proposition 3** Fix \( p, q, \) and \( \rho \). For almost all \( u \), \( \lim_{\Delta \to 0} \text{Equil}(\langle \Delta, u, p, q, \rho \rangle) = \text{Equil}^0(\langle u, p, q, \rho \rangle) \).

In words, provided that periods are sufficiently short the Markov perfect equilibria of the discrete-time stochastic game with simultaneous moves almost coincide with those of the continuous-time stochastic game, although the latter are much easier to compute than the former. We note that Proposition 3 does not carry over from simultaneous to alternating moves. We also note that with a continuum of actions, a version of Proposition 3 (and of Theorem 1) can be obtained by considering approximate equilibria as in Fudenberg and Levine (1986).

### 5.3 Games with Infinite Hazard Rates and Frequent Moves

We apply and extend our main results to shed light on the numerous examples in the literature in which the protocol of moves matters crucially for equilibrium behavior. We use the canonical model of Lagunoff and Matsui (1997) to expose a discontinuity in the set of Markov perfect equilibria as hazard rates become large and moves become frequent.

Consider a two-player game as the one in Example 5. The set of states is \( \Omega = \Omega_1 \times \Omega_2 \), with \( |\Omega_i| \geq 2 \) for all \( i \). The set of actions of player \( i \) is \( A_i(\omega) = \Omega_i \). We now assume that the flow payoffs of the players coincide:

\[ s_i(\omega, a_j) = \frac{1}{2} \pi(\omega). \]

Given a sequence \((A_\nu)\) indexed by \( \nu \in \mathbb{N} \), with \( A_\nu \subseteq \mathbb{R}^n \), we define

\[
\lim \inf_{\nu \to \infty} A_\nu = \{ x \in \mathbb{R}^n \mid \limsup_{\nu \to \infty} d(x, A_\nu) = 0 \}
\quad \text{and} \quad
\limsup_{\nu \to \infty} A_\nu = \{ x \in \mathbb{R}^n \mid \liminf_{\nu \to \infty} d(x, A_\nu) = 0 \},
\]

where \( d(x, A) = \inf \{ \| y - x \| \mid y \in A \} \). If both limits coincide, we denote their common value by \( \lim_{\nu \to \infty} A_\nu \).
Denote $\omega^* = \arg\max_{\omega \in \Omega} \pi(\omega)$ and assume it is unique. Use the function $\pi: \Omega \to \mathbb{R}$ to define the payoff matrix of a normal-form game in which players simultaneously choose actions $a_i \in \Omega_i$. We assume that this normal-form game has a Nash equilibrium $\omega^{NE}$ such that $\omega^{NE}_i \neq \omega^*_i$ for all $i$. Because $\omega^*$ is the unique maximizer of the function $\pi$, $\omega^*$ is also a Nash equilibrium. An example of such game is the payoff matrix in Example [3] with $\Omega_i = \{E, O\}$ and $\pi(E, E) = 2$, $\pi(O, O) = 1$, and $\pi(\omega) = 0$ otherwise. In this example, $\omega^* = (E, E)$ and $\omega^{NE} = (O, O)$.

Our Theorem 1 shows that irrespective of the protocol of moves $<J, P>$, the set of Markov perfect equilibria $\text{Equil}(<\Delta, J, P, u, p, q, \rho>)$ converges to $\text{Equil}^0(<u, p, q, \rho>)$ as $\Delta \to 0$. We can further characterize the limit as $\lambda \to \infty$:

**Proposition 4** There exists a strategy profile $\sigma^*$ such that

$$
\lim_{\lambda \to \infty} \lim_{\Delta \to 0} \text{Equil}(<\Delta, J, P, u, p, q, \rho>) = \{\sigma^*\}.
$$

Moreover, $\sigma^*_i(\omega) = \omega^*_i$ for all $i$ and all $\omega$.

Up to the fact that the limit as $\Delta \to 0$ is a continuous-time stochastic game, the logic of Proposition 4 follows from Lagunoff and Matsui (1997). Note that if player $i$’s state is $\omega_i = \omega^*_i$, then player $-i$ has an incentive to choose $a_{-i} = \omega^*_{-i}$ to obtain $\pi(\omega^*)$. In a given state $\omega$, player $i$ thus knows that if his state changes to $\omega^*_i$, then his rival will switch to action $a_{-i} = \omega^*_{-i}$ relatively soon as long as $\lambda$ is sufficiently large. The unique limit solution is therefore $\sigma^*$.

Proposition 4 shows that when the hazard rate goes to infinite, the set of Markov perfect equilibria does not depend on the protocol of moves if moves are frequent. This seemingly contradicts the celebrated results in Maskin and Tirole (1988a) and Lagunoff and Matsui (1997) that the set of Markov perfect equilibria depends critically on the protocol of moves when hazard rates are infinite.

The explanation is that there is a discontinuity in the joint limit as $\lambda \to \infty$ and $\Delta \to 0$. Hence, caution is warranted in working with games with infinite hazard rates and arbitrarily frequent moves, as the order of limits matters. To establish the discontinuity, consider a protocol of simultaneous moves $<J^{sim}, P^{sim}>$. Imposing $\lambda \Delta = 1$, players determine the state with probability one when they move. It is relatively simple to show that

$$
\{\sigma^*, \sigma^{NE}\} \subseteq \text{Equil}(<\Delta, J^{sim}, P^{sim}, u, p, q, \rho>)
$$

where $\sigma^{NE}_i(\omega) = \omega^{NE}_i$ for all $i$ and all $\omega$. Intuitively, if $\lambda \Delta = 1$, then players can coordinate on one of the two Nash equilibria $\omega^{NE}$ and $\omega^*$. We conclude that

$$
\lim_{\lambda \to \infty} \lim_{\Delta \to 0} \text{Equil}(<\Delta, J, P, u, p, q, \rho>) \subseteq \lim_{\lambda \Delta = 1, \Delta \to 0} \text{Equil}(<\Delta, J^{sim}, P^{sim}, u, p, q, \rho>)
$$

Moreover, the limit of the set of Markov perfect equilibria as $\Delta \to 0$ keeping $\lambda \Delta$ constant depends on the protocol of moves, in line with Maskin and Tirole (1988a) and Lagunoff and Matsui (1997). To illustrate,
consider a protocol of alternating moves $\langle \mathcal{J}^{alt}, \mathcal{P}^{alt} \rangle$. Imposing $\lambda \Delta = 1/2$, a player determines his state with probability one when he moves. From Theorem 1 in Lagunoff and Matsui (1997), if the period length $\Delta$ is sufficiently small, then

$$\text{Equil}(\langle \Delta, \mathcal{J}^{alt}, \mathcal{P}^{alt}, u, p, q, \rho \rangle) = \{\sigma^*\}.$$ 

This implies that

$$\lim_{\lambda \Delta = 1/2, \Delta \to 0} \text{Equil}(\langle \Delta, \mathcal{J}^{alt}, \mathcal{P}^{alt}, u, p, q, \rho \rangle) \subseteq \lim_{\lambda \Delta = 1, \Delta \to \infty} \inf \, \text{Equil}(\langle \Delta, \mathcal{J}^{sim}, \mathcal{P}^{sim}, u, p, q, \rho \rangle).$$

In this sense, the results in Maskin and Tirole (1988a) and Lagunoff and Matsui (1997) can be obtained by taking the joint limit $\lambda \Delta \to 1$ and $\Delta \to 0$, but this is just one out of many ways of taking the joint limit in our setting.

6 Conclusions

The timing of decisions is an essential ingredient into modeling many strategic situations. Yet, determining the protocol of moves that is most realistic and appropriate for the application at hand can be challenging. While the literature abounds with examples in which the protocol of moves matters crucially for equilibrium behavior, our paper is a first attempt to show that the implications and predictions of a fairly general and widely used class of dynamic models are independent of the timing of decisions and thus more robust for the purposes of applied work.

In particular, we introduce separable dynamic games with noisy transitions and establish that they are protocol invariant provided that periods are sufficiently short and moves are therefore sufficiently frequent. Protocol invariance means that the set of Markov perfect equilibria of these games is nearly the same irrespective of the order in which players are assumed to move within a period, including—and extending beyond—simultaneous, alternating, and sequential moves. We also show that the Markov perfect equilibria of separable dynamic stochastic games with noisy transitions have a remarkably simple structure.

In addition to alleviating the burden of determining the most realistic and appropriate protocol of moves, our main results have a number of implications for applied work. They provide a new justification for focusing on Markov perfect equilibria in dynamic stochastic games and facilitate computing these equilibria. They further point to a discontinuity in the set of Markov perfect equilibria as hazard rates become large and moves become frequent, thereby shedding light on the examples in the literature in which the protocol of moves matters crucially for equilibrium behavior.

\[\text{Recall from Assumption 3 that the hazard rates are scaled by } |\mathcal{J}|. \text{ This is the reason we take } \lambda \Delta = 1 \text{ with simultaneous moves and } \lambda \Delta = 1/2 \text{ with alternating moves.}\]

\[\text{The restriction to a discount factor close to one in Lagunoff and Matsui (1997) translates into a period length } \Delta \text{ close to zero in our setting.}\]
Appendix

This Appendix consists of three parts. Appendix A.1 provides the proof of Theorem 1, Appendix A.2 provides the proof of Theorem 2, and Appendix A.3 provides the proofs for Section 5.

A.1 Proof of Theorem 1

A.1.1 Notation and Preliminary Definitions

Enumerate the state space as $\Omega = \{\omega^1, \ldots, \omega^{|\Omega|}\}$ and the set of actions for player $i$ as $A_i(\omega) = \{a_i^1, \ldots, a_i^{|A_i(\omega)|}\}$. Given a strategy profile $\sigma = (\sigma_i)_{i=1}^N \in \Sigma$, define the matrix $P_\sigma \in \mathbb{R}^{|\Omega| \times \sum_{\omega \in \Omega} \sum_{i=1}^N |A_i(\omega)|}$ as

$$
\begin{pmatrix}
\sigma_1(a_i^1 | \omega^1) & \ldots & \sigma_1(a_i^{|A_i(\omega)|} | \omega^1) & \sigma_2(a_i^1 | \omega^1) & \ldots & \sigma_N(a_N^1 | \omega^1) & 0 & \ldots & 0 & \ldots \\
0 & \ldots & \ldots & 0 & \ldots & \sigma_1(a_i^1 | \omega^2) & \ldots & \sigma_N(a_N^1 | \omega^2) & \ldots
\end{pmatrix}
$$

Define the matrix $Q \in \mathbb{R}^{|\Omega| \times \sum_{\omega \in \Omega} \sum_{i=1}^N |A_i(\omega)|}$ as

$$
\begin{pmatrix}
q_1(a_i^1, \omega) & 0 & \ldots \\
q_1(a_i^1, \omega^1) & 0 & \ldots \\
q_N(a_N^1, \omega^1) & 0 & \ldots \\
0 & q_1(a_i^1, \omega^2) & 0 & \ldots \\
\vdots & \vdots & \vdots
\end{pmatrix}
$$

and the matrix $P \in \mathbb{R}^{|\Omega| \times \sum_{\omega \in \Omega} \sum_{i=1}^N |A_i(\omega)|}$ as

$$
P_{(i,a_i,\omega,\omega')} = \begin{cases} 
\varphi_i(\omega' | a_i, \omega) & \text{if } \omega' \neq \omega, \\
0 & \text{if } \omega' = \omega.
\end{cases}
$$

Given a player $i \in \{1, \ldots, N\}$, limit condition (3.2) can be written as

$$
(\rho \mathbb{I} + P_\sigma(Q - P)) V_i^0 = P_\sigma u_i,
$$

where $\mathbb{I}$ is the identity matrix, $V_i^0 \in \mathbb{R}^{|\Omega|}$, and $u_i \in \mathbb{R}^{|A_i(\omega)| \times \sum_{\omega \in \Omega} |A_i(\omega)|}$. The matrix $\rho \mathbb{I} + P_\sigma(Q - P)$ is strictly dominant diagonal and therefore invertible.\footnote{A strictly dominant diagonal matrix $X$ is a square matrix with entries $X_{ij}$ such that $|X_{ii}| > \sum_{j \neq i} |X_{ij}|$ for all $i$.} We emphasize the dependence of the unique solution to limit condition (3.2) by writing $V_i^0(\cdot) = V_i^0(\cdot, \sigma)$. This solution is

$$
V_i^0(\cdot, \sigma) = \left(\rho \mathbb{I} + P_\sigma(Q - P)\right)^{-1} P_\sigma u_i.
$$
Given \( i \in \{1, \ldots, N\} \) and \( u_i \in \mathbb{R}^{\sum_{j=1}^N \sum_{\omega \in \Omega} |A_j(\omega)|} \), consider the vector

\[
u_i + (P - Q) V_i^0(\cdot; \sigma) = u_i + (P - Q)(\rho I + \mathcal{P}_\sigma(Q - P))^{-1} \mathcal{P}_\sigma u_i
\]

\[
= u_i + (P - Q) \frac{1}{\rho} \left( 1 - \frac{1}{\rho} \mathcal{P}_\sigma(Q - P) + \frac{1}{\rho^2}(\mathcal{P}_\sigma(Q - P))^2 \cdots \right) \mathcal{P}_\sigma u_i
\]

\[
= \left( 1 - \frac{1}{\rho}(Q - P) \mathcal{P}_\sigma + \frac{1}{\rho^2}((Q - P) \mathcal{P}_\sigma)^2 - \frac{1}{\rho^3}((Q - P) \mathcal{P}_\sigma)^3 + \cdots \right) u_i
\]

\[
= \left( 1 + \frac{1}{\rho}(P - Q) \mathcal{P}_\sigma \right)^{-1} u_i,
\]

where the inversion is justified by strict diagonal dominance. The map

\[
u_i \in \mathbb{R}^{\sum_{j=1}^N \sum_{\omega \in \Omega} |A_j(\omega)|} \mapsto \left( 1 + \frac{1}{\rho}(P - Q) \mathcal{P}_\sigma \right)^{-1} u_i \in \mathbb{R}^{\sum_{j=1}^N \sum_{\omega \in \Omega} |A_j(\omega)|}
\]

is invertible.

The above results have been presented for a given strategy profile \( \sigma \in \Sigma \). Following Appendix A.1 in Doraszelski and Escobar (2010), we construct an open set \( \Sigma^\varepsilon \subset \mathbb{R}^{\sum_{j=1}^N \sum_{\omega \in \Omega} |A_j(\omega)|} \) that strictly contains \( \Sigma \) such that all the preceding operations are valid for any \( \sigma \in \Sigma^\varepsilon \).

### A.1.2 Regularity

We begin by providing a formal definition of regularity and establishing the key technical point that for almost all flow payoffs \( u \in \mathbb{R}^{N \sum_{j=1}^N \sum_{\omega \in \Omega} |A_j(\omega)|} \), the restriction to regular solutions is without loss of generality.

Given \( i \in \{1, \ldots, N\} \), \( \omega \in \Omega \), \( a_i \in A_i(\omega) \), and \( \sigma \in \Sigma^\varepsilon \), define the function

\[U_i(\omega, a_i, \sigma) = u_{i,i}(\omega, a_i) + \sum_{\omega' \neq \omega} (V_i^0(\omega', \sigma) - V_i^0(\omega, \sigma)) \varphi_i(\omega' | \omega, a_i).
\]

In light of limit condition (3.3), we interpret \( U_i(a_i, \omega, \sigma) \) as the objective function that player \( i \in \{1, \ldots, N\} \) maximizes over \( a_i \in A_i(\omega) \) given state \( \omega \in \Omega \) and continuation play \( \sigma \in \Sigma^\varepsilon \).

Consider \( \bar{\sigma} \in \text{Equil}^0(<u, p, q, \rho>) \). Choose \( a_i^\omega \) such that \( \bar{\sigma}_i(a_i^\omega | \omega) > 0 \) for all \( i \in \{1, \ldots, N\} \) and all \( \omega \in \Omega \). Given \( a_i \neq a_i^\omega \) and \( \sigma \in \Sigma^\varepsilon \), define

\[f_{i,a_i,\omega}^{\sigma}(\sigma) = \sigma_i(a_i | \omega)(U_i(a_i, \omega, \sigma) - U_i(a_i^\omega, \omega, \sigma))
\]

while

\[f_{i,a_i^\omega,\omega}^{\sigma}(\sigma) = \sum_{a_i \in A_i(\omega)} \sigma_i(a_i | \omega) - 1.\]

By definition, \( f(\bar{\sigma}) = 0 \). In this subsection, we sometimes emphasize the dependence of \( f \) on \( u \) by writing \( f(\sigma, u) \). Note that \( f: \Sigma^\varepsilon \times \mathbb{R}^{N \sum_{\omega \in \Omega} \sum_{j=1}^N |A_j(\omega)|} \rightarrow \mathbb{R}^{\sum_{\omega \in \Omega} \sum_{j=1}^N |A_j(\omega)|} \) is continuously differentiable.
Definition 1: \( \bar{\sigma} \in \text{Equil}^0(<u, p, q, \rho>) \) is regular if the Jacobian of \( f \) with respect to \( \bar{\sigma} \), \( \frac{\partial f}{\partial \bar{\sigma}}(\bar{\sigma}) \), has full rank \( \sum_{j=1}^{N} \sum_{\omega \in \Omega} |A_j(\omega)| \).

We present two preliminary lemmas. We say that a strategy profile \( \sigma \in \Sigma \varepsilon \) is completely mixed if \( \sigma_i(a_i | \omega) > 0 \) for all \( i \in \{1, \ldots, N\} \), \( \omega \in \Omega \), and all \( a_i \in A_i(\omega) \).

Lemma 4: If \( \sigma \in \Sigma \varepsilon \) is completely mixed, then the Jacobian of \( f \) with respect to \( (\sigma, u) \), \( \frac{\partial f}{\partial (\sigma, u)}(\sigma, u) \), has full rank \( \sum_{j=1}^{N} \sum_{\omega \in \Omega} |A_j(\omega)| \).

Proof. Define the matrix \( M(\sigma, i) \in \mathbb{R}^{\sum_{\omega \in \Omega}(|A_i(\omega)| - 1) \times \sum_{j=1}^{N} \sum_{\omega \in \Omega} |A_j(\omega)|} \) such that, for all \( a_i \neq \bar{a}_i^\omega \), its \( (i, a_i, \omega) \) row equals 0 in all components save for the \( (i, a_i, \omega) \) column, where we write \( \sigma_i(a_i | \omega) \), and for the \( (i, \bar{a}_i^\omega, \omega) \) column, where we write \( -\sigma_i(a_i | \omega) \). The function \( f \) can be expressed as

\[
\begin{align*}
    f_i(\sigma, u) &= \begin{cases} 
        \sum_{a_i \in A_i(\omega)} \sigma_i(a_i | \omega^i) - 1, \\
        \vdots \\
        \sum_{a_i \in A_i(\omega)} \sigma_i(a_i | \omega^{\Omega}) - 1, \\
        M(\sigma, i) \left( 1 + \frac{1}{\rho} (P - Q) P_\sigma \right)^{-1} u_i. 
    \end{cases}
\end{align*}
\]

Up to permutation (which are irrelevant to determine the rank of the Jacobian), we can write

\[
\frac{\partial f(\sigma, u)}{\partial (\sigma, u)} = \begin{pmatrix}
    \sigma_1 & \sigma_2 & \cdots & \sigma_N & u_1 & u_2 & \cdots & u_N \\
    X_1 & 0 & 0 & 0 & 0 & 0 & \cdots & 0 \\
    0 & X_2 & 0 & 0 & 0 & 0 & \cdots & 0 \\
    \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \ddots & \vdots \\
    0 & 0 & 0 & X_N & 0 & 0 & \cdots & 0 \\
    Y_1 & Y_2 & \cdots & Y_N & \vdots & \vdots & \ddots & \vdots \\
    \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\
    0 & 0 & \cdots & Z_N
\end{pmatrix},
\]

where \( X_i \) equals

\[
\begin{pmatrix}
    \sigma_i(\cdot | \omega^1) & \sigma_i(\cdot | \omega^2) & \cdots & \sigma_i(\cdot | \omega^{\Omega}) \\
    1 & \ldots & 1 & 0 & \ldots & 0 & \cdots & 0 & \cdots & 0 \\
    0 & \ldots & 0 & 1 & \ldots & 1 & \cdots & 0 & \cdots & 0 \\
    \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \ddots & \vdots \\
    0 & \ldots & 0 & 0 & \ldots & 0 & \cdots & 1 & \cdots & 1
\end{pmatrix}
\]

and has rank \( |\Omega| \), while \( Z_i = M(\sigma, i) \left( 1 + \frac{1}{\rho} (P - Q) P_\sigma \right)^{-1} \). Since \( M(\sigma, i) \) has full rank \( \sum_{\omega \in \Omega} (|A_i(\omega)| - 1) \)

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and \((1 + \frac{1}{\rho}(P - Q)P_{\sigma})^{-1}\) has full rank \(\sum_{\omega \in \Omega} \sum_{j=1}^{N} |A_{j}(\omega)|\), \(Z_{i}\) has rank \(\sum_{\omega \in \Omega}(|A_{i}(\omega)| - 1)\). We deduce that the Jacobian has full rank \(\sum_{j=1}^{N} \sum_{\omega \in \Omega} |A_{j}(\omega)|\) and the lemma follows. \(\blacksquare\)

Given \(\sigma \in \Sigma, i \in \{1, \ldots, N\}\), and \(\omega \in \Omega\), define the best reply as

\[ B_{i}(\sigma, \omega) = \arg \max_{a_{i} \in A_{i}(\omega)} U_{i}(\omega, a_{i}, \sigma) \]

and the carrier as

\[ C_{i}(\sigma, \omega) = \left\{ a_{i} \in A_{i}(\omega) \mid \sigma_{i}(a_{i} | \omega) > 0 \right\}. \]

Using this notation, \(\sigma \in \text{Equil}^{0}(<u, p, q, \rho>)\) if and only if \(C_{i}(\sigma, \omega) \subseteq B_{i}(\sigma, \omega)\) for all \(i \in \{1, \ldots, N\}\). We say that \(\sigma \in \text{Equil}^{0}(<u, p, q, \rho>)\) is quasi-strict if \(C_{i}(\sigma, \omega) = B_{i}(\sigma, \omega)\) for all \(i \in \{1, \ldots, N\}\).

**Lemma 5** For almost all \(u \in \mathbb{R}^{N} \sum_{j=1}^{N} \sum_{\omega \in \Omega} |A_{j}(\omega)|\), any \(\sigma \in \text{Equil}^{0}(<u, p, q, \rho>)\) is quasi-strict.

**Proof.** Given \(i \in \{1, \ldots, N\}\), consider correspondences \(B_{i}^{*} : \Omega \to \bigcup_{\omega \in \Omega} A_{i}(\omega)\) and \(C_{i}^{*} : \Omega \to \bigcup_{\omega \in \Omega} A_{i}(\omega)\), with \(C_{i}^{*}(\omega) \subseteq B_{i}^{*}(\omega) \subseteq A_{i}(\omega)\) for all \(\omega \in \Omega\). Define \(G(B^{*}, C^{*})\) as the set of all \(u\) having some \(\sigma \in \text{Equil}^{0}(<u, p, q, \rho>)\) with best replies \(B^{*} = (B_{i}^{*})_{i=1}^{N}\) and carriers \(C^{*} = (C_{i}^{*})_{i=1}^{N}\). Formally,

\[ G(B^{*}, C^{*}) = \left\{ u \mid \text{there exists } \sigma \in \text{Equil}^{0}(<u, p, q, \rho>) \text{ with } B_{i}(\sigma, \cdot) = B_{i}^{*} \text{ and } C_{i}(\sigma, \cdot) = C_{i}^{*} \text{ for all } i = 1, \ldots, N \right\}. \]

Consider first \(\bar{\sigma} \in \text{Equil}^{0}(<\bar{u}, p, q, \rho>)\) such that \(B_{i}(\bar{\sigma}, \omega) = B_{i}^{*}(\omega)\) for all \(\omega \in \Omega\). Fix \(a_{i}^{\omega}\) such that \(\bar{\sigma}_{i}(a_{i}^{\omega} | \omega) > 0\) and note that the indifference condition \(U_{i}(a_{i}, \omega, \bar{\sigma}) - U_{i}(a_{i}^{\omega}, \omega, \bar{\sigma}) = 0\) holds for all \(i \in \{1, \ldots, N\}\) and all \(a_{i} \in B_{i}^{*}(\omega)\). For all \(\omega \in \Omega\) and all \(i \in \{1, \ldots, N\}\), define the matrix \(P_{i}(\sigma) \in \mathbb{R}^{\sum_{\omega \in \Omega} |B_{i}^{*}(\omega)| - 1 \times \sum_{\omega \in \Omega} |A_{i}(\omega)|}\), such that for all \(a_{i} \in B_{i}^{*}(\omega)\), its \((\omega, a_{i})\) row equals 0 save for the \((\omega, a_{i})\) component, where it equals 1, and the \((\omega, a_{i}^{\omega})\) component, where it equals -1. We can therefore stack all the indifference conditions by writing

\[ M(\sigma, u) = \begin{pmatrix} P_{1}(\sigma) \left( 1 + \frac{1}{\rho}(P - Q)P_{\sigma} \right)^{-1} u_{1} \\ \vdots \\ P_{N}(\sigma) \left( 1 + \frac{1}{\rho}(P - Q)P_{\sigma} \right)^{-1} u_{N} \end{pmatrix} \]

and note that \(M(\bar{\sigma}, \bar{u}) = 0\). The Jacobian \(\frac{\partial M}{\partial \sigma}(\sigma, u)\) can be computed as

\[ \frac{\partial M}{\partial u}(\sigma, u) = \begin{pmatrix} P_{1}(\sigma) \left( 1 + \frac{1}{\rho}(P - Q)P_{\sigma} \right)^{-1} & 0 & \cdots & 0 \\ 0 & P_{2}(\sigma) \left( 1 + \frac{1}{\rho}(P - Q)P_{\sigma} \right)^{-1} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & \cdots & 0 & P_{N}(\sigma) \left( 1 + \frac{1}{\rho}(P - Q)P_{\sigma} \right)^{-1} \end{pmatrix}. \]

Since \(P_{1}(\sigma)\) has full rank \(\sum_{\omega \in \Omega} |B_{i}^{*}(\omega)| - 1\), the Jacobian \(\frac{\partial M}{\partial u}(\sigma, u)\) has rank \(\sum_{i=1}^{N} \sum_{\omega \in \Omega} (|B_{i}^{*}(\omega)| - 1)\).
In particular, since \( M(\bar{\sigma}, \bar{u}) = 0 \), we can construct open sets \( \mathcal{N}, \mathcal{N}_1 \subseteq \mathbb{R}^{\sum_{i=1}^{N} \sum_{\omega \in \Omega} (|A_i(\omega)| - |B_i^*(\omega)| - 1)} \), \( \mathcal{N}_2 \subseteq \mathbb{R}^{\sum_{i=1}^{N} \sum_{\omega \in \Omega} (|B_i^*(\omega)| - 1)} \), with \( \bar{\sigma} \in \mathcal{N} \) and \( \bar{u} \in \mathcal{N}_1 \times \mathcal{N}_2 \), and a continuously differentiable function \( \Phi \) such that for all \( (\sigma, u_1) \in \mathcal{N} \times \mathcal{N}_1 \) there exists a unique \( u_2 = \Phi(\sigma, u_1) \in \mathcal{N}_2 \) which is a solution to \( M(\sigma, (u_1, u_2)) = 0 \). Without loss, all these open sets are balls with rational centers and radii and we emphasize their dependence on \((\bar{\sigma}, \bar{u})\) by writing \( \mathcal{N}_1^{\bar{\sigma}, \bar{u}}, \mathcal{N}_2^{\bar{\sigma}, \bar{u}}, \) and \( \mathcal{N}^{\bar{\sigma}, \bar{u}} \).

Now take \( C^* \) such that for some \( i \in \{1, \ldots, N\} \) and some \( \omega \in \Omega \), \( C_i^*(\omega) \subseteq B_i^*(\omega) \). Consider the set

\[
R^{\bar{\sigma}, \bar{u}}(B^*, C^*) = \left\{ u \in N_1^{\bar{\sigma}, \bar{u}} \times N_2^{\bar{\sigma}, \bar{u}} \mid \text{there exists } (\sigma, u^1) \in (N^\sigma \cap A(C^*)) \times N_1^{\bar{\sigma}, \bar{u}} \text{ such that } u_2 = \Phi(\sigma, u_1) \right\}
\]

where \( A(C^*) = \{ \sigma \in \Sigma \mid C_i(\cdot, \omega) = C_i^*(\omega) \text{ for all } i = 1, \ldots, N \} \). Note that the dimension of \((N^\sigma \cap A(C^*)) \times N_1^{\bar{\sigma}, \bar{u}}\) equals \( N \sum_{\omega \in \Omega} \sum_{j=1}^{N} |A_j(\omega)| \). Therefore, \( M^{\bar{\sigma}, \bar{u}}(B^*, C^*) \) has measure zero. Since we are choosing the neighborhoods from a countable set, it follows that \( G(B^*, C^*) \subseteq \cup_{n \in \mathbb{N}} Q^n \), where \( Q^n = R^{\bar{\sigma}, \bar{u}}(B^*, C^*) \), has measure zero as well.

The following is the main result of this subsection.

**Proposition 5** For almost all \( u \in \mathbb{R}^{N \sum_{j=1}^{N} \sum_{\omega \in \Omega} |A_j(\omega)|} \), all \( \bar{\sigma} \in \text{Equil}^0(<u, p, q, \rho>) \) are regular.

**Proof.** From Lemma 5 we can rule out games \( u \in \mathbb{R}^{N \sum_{j=1}^{N} \sum_{\omega \in \Omega} |A_j(\omega)|} \) having non-quasi-strict solutions and focus on games having only quasi-strict solutions. Since there is a finite number of correspondences \( B_i^* : \Omega \to A_i \), it is enough to prove that the set of games having a non-regular equilibrium \( \sigma \) with

\[
B_i(\sigma, \omega) = C_i(\sigma, \omega) = B_i^*(\omega)
\]

for all \( i \in \{1, \ldots, N\} \) and all \( \omega \in \Omega \) has measure zero. Considering the submatrix \( \bar{J}(\sigma, u) \) obtained from \( \frac{\partial f}{\partial \sigma}(\sigma, u) \) by crossing out all rows and columns corresponding to components \((a_i, \omega)\) with \( a_i \notin B_i^*(\omega) \), it follows that \( \bar{J}(\sigma) \) has full rank if and only if so does \( \frac{\partial J}{\partial \sigma}(\sigma, u) \). Noting that \( J(\sigma, u) \) is the Jacobian of a completely mixed solution, without loss of generality we can therefore assume that \( B^*(\omega) \) does not depend on \( \omega \) and restrict attention to completely mixed solutions. Using Lemma 4 and the transversality theorem (see the discussion in Section 7.1.2 in Doraszelski and Escobar 2010), we deduce that for almost all games, all completely mixed equilibria are regular.

**A.1.3 Establishing Lemma 2**

Fix a game \( u \in \mathbb{R}^{N \sum_{\omega \in \Omega} \sum_{j=1}^{N} |A_j(\omega)|} \) and a regular solution \( \sigma^0 \in \text{Equil}^0(<u, p, q, \rho>) \). Let \( \mathcal{J}, \mathcal{P} > \) be a protocol of moves and \( \Delta > 0 \) the period length. We establish that the regular solution \( \sigma^0 \) can be approximated by a Markov perfect equilibrium of a separable dynamic game with noisy transitions and an arbitrary protocol of moves if the period length \( \Delta \) is sufficiently small. To do so, we apply a version of the implicit function theorem to the limit conditions.
Proof of Lemma 2. In the separable dynamic game with noisy transitions \( \Gamma = (\Delta, J, \mathcal{P}, u, p, q, \rho) \), write the continuation value of player \( i \in \{1, \ldots, N\} \) if players \( J \in J \) have the move and the state is \( \omega \in \Omega \) as \( V_i^\Delta(\omega, J) \). Note that the value function \( V_i^\Delta : \Omega \times J \to \mathbb{R} \) is uniquely determined by the strategy profile \( \sigma \in \Sigma \). We therefore write \( V_i^\Delta(\cdot, \cdot) = V_i^\Delta(\cdot, \cdot, \sigma) \). Note that \( V_i^\Delta(\cdot, \cdot, \sigma) \) is a continuous function of \((\sigma, \Delta)\) and its differential with respect to \( \sigma \) at \( \Delta = 0 \) exists. In particular, for all \( J \in J \) and all \( \sigma \in \Sigma, V_i^\Delta(\cdot, J, \sigma) \to V_i^0(\cdot, \sigma) \) as \( \Delta \to 0 \).

A strategy profile \( \sigma^\Delta \) is a Markov perfect equilibrium of the separable dynamic game with noisy transitions \( \Gamma = (\Delta, J, \mathcal{P}, u, p, q, \rho) \) if for all \( i = 1, \ldots, N, \omega, \in \Omega, \) and all \( a_i \in A_i(\omega) \)

\[
\sigma^\Delta_i(a_i \mid \omega) > 0 \Rightarrow a_i \in \arg \max_{\tilde{a}_i \in A_i(\omega)} \mathcal{U}_i^\Delta(\omega, \tilde{a}_i, \sigma^\Delta)
\]

with

\[
\mathcal{U}_i^\Delta(\omega, a_i, \sigma^\Delta) = u_{i,i}(\omega, a_i) + \exp(-\rho\Delta) \sum_{\omega' \neq \omega} \sum_{J' \in J} \left( V_i^\Delta(\omega', J', \sigma^\Delta) - V_i^\Delta(\omega, J', \sigma^\Delta) \right) \psi_i(\omega' \mid \omega, a_i) \Pr(J' \mid J) + O(\Delta)
\]

and \( J \in J \) is such that \( i \in J \).

Consider the profile \((a_i^\omega)_{i=1,\ldots,N,\omega \in \Omega}\) that is used in the construction of the function \( f \) in Appendix \ref{sec:proof-lemma2} for which \( \sigma^0 \) is regular. Abusing notation, construct the function \( f : [0, 1] \times \Sigma^r \to \mathbb{R}^{\sum_{i=1}^N |A_i(\omega)|} \) such that for all \( a_i \neq a_i^\omega \)

\[
f_{i,a_i,\omega}(\Delta, \sigma) = \sigma_i(a_i, \omega) \left( U_i^\Delta(\omega, a_i, \sigma) - U_i^\Delta(\omega, a_i^\omega, \sigma) \right)
\]

while

\[
f_{i,a_i^\omega,\omega}(\Delta, \sigma) = \sum_{a_i \in A_i(\omega)} \sigma_i(a_i \mid \omega) - 1.
\]

Observe that \( f(\Delta, \sigma) \) is a continuous function, with a well-defined differential with respect to \( \sigma, D_\sigma f(0, \sigma) \), at \((0, \sigma)\). Moreover, \( f(0, \sigma^0) = 0 \) and \( D_\sigma f(0, \sigma^0) \) has full rank \( \sum_{j=1}^N \sum_{\omega \in \Omega} |A_i(\omega)| \). A version of the implicit function theorem (see Lemma \ref{lem:implicit-function-theorem} below) implies that for all \( r > 0 \) there exists \( \Delta > 0 \) such that for all \( \Delta < \Delta \), there exists \( \sigma^\Delta \in \Sigma^r \) with \( \|\sigma^0 - \sigma^\Delta\| < r \) such that \( f(\Delta, \sigma^\Delta) = 0 \). Moreover, we can take \( \Delta \) and \( r \) small enough so that (i) \( \sigma^\Delta_i(a_i, \omega) > 0 \) whenever \( \sigma^0_i(a_i, \omega) > 0 \), and (ii) \( U_i^\Delta(\omega, a_i, \sigma^\Delta) < U_i^\Delta(\omega, a_i^\omega, \sigma^\Delta) \) whenever \( U_i(\omega, a_i, \sigma^0) < U_i(\omega, a_i^\omega, \sigma^0) \).

To prove that \( \sigma^\Delta \) is a Markov perfect equilibrium of the separable dynamic game with noisy transitions \( \Gamma = (\Delta, J, \mathcal{P}, u, p, q, \rho) \), consider first \( a_i \in A_i(\omega) \) and \( \omega \in \Omega \) such that \( \sigma^0_i(a_i \mid \omega) = 0 \). Since \( \sigma^0 \) is regular, it is also quasi-strict and therefore \( U_i(\omega, a_i, \sigma^0) < U_i(\omega, a_i^\omega, \sigma^0) \). From (ii), \( U_i^\Delta(\omega, a_i, \sigma^\Delta) < U_i^\Delta(\omega, a_i^\omega, \sigma^\Delta) \) and \( \sigma^\Delta(a_i \mid \omega) = 0 \). Next consider \( a_i \in A_i(\omega) \) and \( \omega \in \Omega \) such that \( \sigma^0_i(a_i \mid \omega) > 0 \). We can use (i) to deduce that \( \sigma^\Delta(a_i \mid \omega) > 0 \) and, since \( f(\Delta, \sigma^\Delta) = 0 \), \( U_i^\Delta(\omega, a_i, \sigma^\Delta) = U_i^\Delta(\omega, a_i^\omega, \sigma^\Delta) \). All of these observations prove that whenever \( \sigma^\Delta(a_i, \omega) > 0 \), \( a_i \) solves \( \max_{\tilde{a}_i \in A_i(\omega)} U_i^\Delta(\omega, \tilde{a}_i, \sigma^\Delta) \). Therefore, \( \sigma^\Delta \) is a Markov perfect equilibrium of the separable dynamic game with noisy transitions \( \Gamma = (\Delta, J, \mathcal{P}, u, p, q, \rho) \).
Since for almost all $u \in \mathbb{R}^N \sum_{\omega \in \Omega}^{N} |A_j(\omega)|$, all $\sigma^0 \in \text{Equil}^0(<u,p,q,\rho>)$ are regular, Lemma 2 follows.

It remains to prove the implicit function theorem we used above. The textbook presentation of the implicit function theorem (Section M.E in Mas-Colell, Whinston, and Green 1995) applies to continuously differentiable functions defined on open sets. In our setup, the set of parameters $\Delta \in [0,1]$ is closed and, moreover, we are interested in the boundary case $\Delta = 0$. The following result is a modification of Theorem A in Halkin (1974).

**Lemma 6 (Implicit Function Theorem)** Assume $f : [0,1] \times \Sigma^\varepsilon \to \mathbb{R}^{\sum_{\omega \in \Omega}^{N} |A_j(\omega)|}$ is a continuous function such that its differential with respect to $\sigma \in \Sigma^\varepsilon$ at $\Delta = 0$, $D_{\sigma} f(0,\sigma)$, exists. Let $\sigma^0 \in \Sigma$ be such that $f(0,\sigma^0) = 0$ and $D_{\sigma} f(0,\sigma^0)$ has full rank $\sum_{j=1}^{N} \sum_{\omega \in \Omega} |A_j(\omega)|$. Then, for all $r > 0$ there exists $\Delta > 0$ such that for all $\Delta < \tilde{\Delta}$, there exists $\sigma^\Delta$ such that $\|\sigma^0 - \sigma^\Delta\| < r$ and $f(\Delta, \sigma^\Delta) = 0$.

**Proof.** Consider the function $\varphi(\sigma, \Delta) = \sigma - [D_{\sigma} f(0,\sigma^0)]^{-1} f(\Delta, \sigma)$ and note that the problem of finding $\sigma^\Delta$ such that $f(\Delta, \sigma^\Delta) = 0$ reduces to the problem of finding a fixed point of $\varphi(\cdot, \Delta)$. Note that $D_{\sigma} \varphi(0,\sigma^0) = 0$ and therefore we can assume, without loss, that $r > 0$ is small enough so that for all $\|\sigma - \sigma^0\| < r$, $\sigma \in \Sigma^\varepsilon$ and

\[
\frac{\|\varphi(\sigma,0) - \varphi(\sigma^0,0)\|}{\|\sigma - \sigma^0\|} < \frac{1}{2}.
\]

Since $\varphi(\sigma^0,0) = \sigma^0$, we can therefore deduce that for all $\|\sigma - \sigma^0\| \leq r$, $\|\varphi(\sigma,0) - \sigma^0\| \leq r/2$.

Define now $m(\Delta) = \max_{\|\sigma - \sigma^0\| \leq r} \|\varphi(\sigma, \Delta) - \varphi(\sigma,0)\|$. Berge’s maximum theorem (Theorem 17.31 in Aliprantis and Border 2006) implies that $m$ is continuous in $\Delta \in [0,1]$. Since $m(0) = 0$, there exists $\Delta > 0$ such that for all $\Delta < \tilde{\Delta}$, $m(\Delta) < r/2$. We thus deduce that for all $\sigma$ such that $\|\sigma - \sigma^0\| \leq r$ and $\Delta < \tilde{\Delta}$

\[
\|\varphi(\sigma, \Delta) - \sigma^0\| \leq \|\varphi(\sigma, \Delta) - \varphi(\sigma,0)\| + \|\varphi(\sigma,0) - \sigma^0\|
\]

\[
\leq m(\Delta) + \|\varphi(\sigma, \Delta) - \sigma^0\|
\]

\[
\leq r.
\]

It follows that for all $\Delta < \tilde{\Delta}$, the continuous function $\varphi(\cdot, \Delta)$ maps the convex and compact set $\{\sigma \mid \|\sigma - \sigma^0\| \leq r\}$ into itself. For any such $\Delta < \tilde{\Delta}$, Brouwer’s fixed point theorem (Theorem M.I.1 in Mas-Colell, Whinston, and Green 1995) implies the existence of $\sigma^\Delta$ within distance $r$ of $\sigma^0$ such that $\varphi(\sigma^\Delta, \Delta) = \sigma^\Delta$. ■

**A.1.4 Proof of Theorem 1**

**Proof of Theorem 1.** From Lemma 2 take one of the generic flow payoffs $u \in \mathbb{R}^N \sum_{\omega \in \Omega}^{N} |A_j(\omega)|$ and any two protocols of moves as in the statement of Theorem 1. From Lemma 1 there exists $\tilde{\Delta} > 0$ such that for all $\Delta < \tilde{\Delta}$ and all $\sigma^\Delta \in \text{Equil}(<\Delta, J, \mathcal{P}, u, p, q, \rho>)$, there exists $\sigma^0 \in \text{Equil}^0(<u, p, q, \rho>)$ such that $\|\sigma^\Delta - \sigma^0\| < \varepsilon/2$. From Lemma 2 we can find $\tilde{\Delta} > 0$ such that for all $\Delta < \tilde{\Delta}$, there exists
\( \tilde{\sigma} \in \text{Equil}(\Delta, \tilde{J}, \tilde{P}, u, p, q, \rho >) \) such that \( \| \tilde{\sigma} - \sigma^0 \| < \varepsilon/2 \). Taking \( \tilde{\Delta} = \min\{\tilde{\Delta}, \Delta\} > 0 \), Theorem 1 follows from the triangle inequality. ■

### A.2 Proof of Theorem 2

It is useful to begin by establishing the following lemma:

**Lemma 7** Let \( \sigma \in \text{Equil}(u, p, q, \rho >) \) with corresponding profile of value functions \( V \). Then there exists \( \tilde{\sigma} \in \text{Equil}(u, p, q, \rho >) \) with corresponding profile of value functions \( \tilde{V} \) such that for all \( \omega \in \Omega \) and all \( i \in \{1, \ldots, N\} \),

a. \( \tilde{V}_i(\omega) = V_i(\omega) \);  
b. \( \{a_i \in A_i(\omega) \mid \tilde{\sigma}_i(a_i \mid \omega) > 0\} \subseteq \{a_i \in A_i(\omega) \mid \sigma_i(a_i \mid \omega) > 0\} \);  
c. \( \{|a_i \in A_i(\omega) \mid \tilde{\sigma}_i(a_i \mid \omega) > 0\}| \leq N \).

In particular, if \( \{a_i \in A_i(\omega) \mid \sigma_i(a_i \mid \omega) > 0\} \geq N + 1 \) for some \( i \in \{1, \ldots, N\} \) and some \( \omega \in \Omega \), then \( \tilde{\sigma} \) is not regular.

The proof of this result extends some of the ideas in the proof of Theorem 3.1 in Raghavan, Tijs, and Vrieze (1985).

**Proof of Lemma 7** Write limit condition (3.2) as

\[
\rho V_i(\omega) = \sum_{j=1}^{N} \sum_{a_j \in A_j(\omega)} \left( u_{i,j}(\omega, a_j) + \sum_{\omega' \in \Omega} (V_i(\omega') - V_i(\omega)) \varphi_j(\omega' \mid \omega, a_j) \right) \sigma_j(a_j \mid \omega).
\]

Fix player 1 and state \( \omega \in \Omega \). Consider the finite set

\[
C_1(\omega) = \left\{ v = (v_2, \ldots, v_N) \in \mathbb{R}^{N-1} \mid \text{there exists } a_1 \in A_1(\omega) \text{ such that } v_i = u_{1,1}(\omega, a_1) + \sum_{\omega' \in \Omega} (V_i(\omega') - V_i(\omega)) \varphi_j(\omega' \mid \omega, a_1) \text{ for all } i = 2, \ldots, N \right\}.
\]

Observe that the vector \( \tilde{v} = (\tilde{v}_2, \ldots, \tilde{v}_N) \in \mathbb{R}^{N-1} \) defined by

\[
\tilde{v}_i = \sum_{a_1 \in A_1(\omega)} \left( u_{1,1}(\omega, a_1) + \sum_{\omega' \in \Omega} (V_i(\omega') - V_i(\omega)) \varphi_j(\omega' \mid \omega, a_1) \right) \sigma_1(a_1 \mid \omega)
\]

belongs to the convex hull of \( C_1(\omega) \). Caratheodory’s theorem (Theorem 17.1 in Rockafellar 1970) implies that there exists a distribution \( \tilde{\sigma}_1(\cdot \mid \omega) \in \mathcal{P}(A_1(\omega)) \) with \( \{a_1 \in A_1(\omega) \mid \tilde{\sigma}_1(a_1 \mid \omega) > 0\} \subseteq \{a_1 \in A_1(\omega) \mid \sigma_1(a_1 \mid \omega) > 0\} \) such that

\[
\tilde{v} = \left( \sum_{a_1 \in A_1(\omega)} u_{1,1}(\omega, a_1) + \sum_{\omega' \in \Omega} (V_i(\omega') - V_i(\omega)) \varphi_j(\omega' \mid \omega, a_1) \right) \tilde{\sigma}_1(a_1 \mid \omega) \in \mathbb{R}^{N-1}
\]
and \(|\{a_i \in A_i(\omega) \mid \sigma_i(a_i \mid \omega) > 0\}| \leq N\). Do this procedure for all players (not just player 1) and obtain a strategy profile \(\tilde{\sigma} = (\tilde{\sigma}_i)_{i=1}^N\) that satisfies (b) and (c) for all \(i = 1, \ldots, N\) and all \(\omega \in \Omega\). Also note that, by construction, \(V_i\) is the value function for player \(i\) and \(\sigma_i(\cdot, \omega)\) is putting positive weight only on actions \(a_i \in A_i(\omega)\) that maximize

\[
u_i(\omega, a_i) + \sum_{\omega' \neq \omega} (V_i(\omega') - V_i(\omega)) \varphi_i(\omega' \mid \omega, a_i).
\]

We therefore conclude that \((\tilde{\sigma}, V)\) is a solution to limit condition (3.2) and (3.3) having properties (a), (b), and (c).

**Proof of Lemma 3** Proposition 3 shows that all solutions are regular for almost all flow payoffs \(u \in \mathbb{R}^N \sum_{i=1}^N \sum_{\omega \in A_i(\omega)}\). Take such flow payoffs \(u\) and a solution \(\sigma\). If for some \(i = 1, \ldots, N\) and some \(\omega \in \Omega\), \(|\{a_i \in A_i(\omega) \mid \sigma_i(a_i \mid \omega)\}| \geq N + 1\), we could use Lemma 1 to find the non-regular solution \(\tilde{\sigma}\).

**Proof of Theorem 2** From the proof of Theorem 1, for almost all flow payoffs \(u \in \mathbb{R}^N \sum_{i=1}^N \sum_{\omega \in A_i(\omega)}\), all \(\epsilon > 0\) and all protocols of moves \(< J, P >\), we can find \(\Delta > 0\) such that for all \(\Delta < \Delta\), the carrier of \(\sigma^o \in \text{Equil}^0(< u, p, q, \rho >)\) and \(\sigma^\Delta \in \text{Equil}(\Delta, J, P, u, p, q, \rho)\) coincide and \(||\sigma^\Delta - \sigma^o|| < \epsilon\). Theorem 2 thus follows from Lemma 3.

### A.3 Proofs for Section 5

**Proof of Proposition 2** Take any \(u\) as in Theorem 1. Define \(\text{Payoffs}^0(< u, p, q, \rho >) \subseteq \mathbb{R}^N\) to be the set of payoff profiles associated with solutions in pure strategy profiles to the limit conditions (3.2) and (3.3):

\[
\text{Payoffs}^0(< u, p, q, \rho >) = \left\{ V^0(\omega^{t=0}) \in \mathbb{R}^N \mid (V^0, \sigma^0) \text{ solves (3.2) and (3.3) for some pure strategy profile } \sigma^0 \right\},
\]

where \(\omega^{t=0} \in \Omega\) is the initial state of the game. For \(\epsilon > 0\) take \(\tilde{\Delta}\) such that for all \(\Delta < \tilde{\Delta}\), and all protocols \(< J, P >\), the Hausdorff distance between \(\text{Payoffs}^0(< u, p, q, \rho >)\) and \(\text{Payoffs}^\Delta(< \Delta, < J, P, u, p, q, \rho >)\) is less than \(\epsilon/3\).

Take now \(v \in \text{Payoffs}^\Delta(\Gamma^{sim})\) which is approachable. Then, for all \(n \geq 1\), there exists an asynchronous protocol \(< J^n, P^n >\) and \(w^n \in \text{Payoffs}^F(< \Delta, J^n, P^n, u, p, q, \rho >)\) such that \(||v - w^n|| < 1/n\). Restrict the sequence such that \(1/n < \epsilon/3\). From Bhaskar, Mailath, and Morris (2013), we can actually take \(w^n \in \text{Payoffs}^M(< \Delta, J^n, P^n, u, p, q, \rho >)\). By construction, for any such \(w^n\) we can find \(\tilde{w}^n \in \text{Payoffs}^0(< u, p, q, \rho >)\) such that \(||w^n - \tilde{w}^n|| < \epsilon/3||. Since \(\text{Payoffs}^0(< u, p, q, \rho >)\) has a finite number of elements, we can assume that \(\tilde{w}^n = \tilde{w}\) does not depend on \(n\) (perhaps, by taking a subsequence). Now, take \(w \in \text{Payoffs}^M(< \Delta, J^{sim}, P^{sim}, u, p, q, \rho >)\) such that \(||w - \tilde{w}|| < \epsilon/3\). It follows that

\[
||v - w|| \leq ||v - w^n|| + ||w^n - \tilde{w}|| + ||\tilde{w} - w|| < \frac{\epsilon}{3} + \frac{\epsilon}{3} + \frac{\epsilon}{3}
\]

which proves the result.
Proof Sketch of Proposition 3. The proof follows from the analysis in Appendix A.1 and arguments in Doraszelski and Escobar (2010). Details are available upon request. To provide a sketch, consider the analog to the limit conditions (3.2) and (3.3) that arise without Assumption 3:

\[ \rho V_i(\omega) = u_i(\omega, \sigma(\omega)) + \sum_{\omega' \neq \omega} (V_i(\omega') - V_i(\omega))\varphi(\omega' | \omega, \sigma(\omega)) \]

and

\[ \sigma_i(a_i | \omega) > 0 \Rightarrow a_i \in \arg \max_{\tilde{a}_i \in A_i(\omega)} u_i(\omega, \tilde{a}_i, \sigma_{-i}(\omega)) + \sum_{\omega' \neq \omega} (V_i(\omega') - V_i(\omega))\varphi(\omega' | \omega, \tilde{a}_i, \sigma_{-i}(\omega)). \]

From these limit conditions, we can construct a function \( f \) (as we did in Appendix A.1) such that all solutions are zeros of \( f \) and, moreover, for almost all flow payoffs \( u \), all solutions are regular. We then apply the implicit function theorem to deduce the result. ■

Proof of Proposition 4. The results follow directly from the analysis in Section 3 in Lagunoff and Matsui (1997) and Theorem 1 in their working paper. ■

References


Online Appendix

This Online Appendix consists of four parts. Appendix OA.1 provides a counterexample complementing Example 6. Appendix OA.2 shows that the limit conditions (3.2) and (3.3) may not admit a solution in pure strategies, and Appendices OA.3 and OA.4 generalize our notion of a protocol of moves and provide extensions of Theorem 1.

OA.1 Non-Separable State-to-State Transitions

Consider a dynamic stochastic game with $N = 2$ players, $\Omega = \{0, 1\}$, and $A_i = \{0, 1\}$. The hazard rate in state $\omega = 0$ is $q(0, a_1, a_2) = 1$ if and only if $a_1 = a_2 = 1$, and $q(0, a_1, a_2) = 0$ otherwise, whereas in state $\omega = 1$, $q(1, a) = 0$. The transition probability satisfies $p(1 \mid 0, (a_1, a_2)) = 1$. State $\omega = 1$ is thus absorbing. Flow payoffs do not depend on actions and take the form $u_{i,i}(\omega) = \omega$. In the game with simultaneous moves, it is simple to see that there exist two Markov perfect equilibria in pure strategies. In one of them, the state is stuck in $\omega = 0$. In contrast, in the game with alternating moves and transitions “materializing” only once both players have made a decision (in a violation of Assumption 3 similar to Example 6), the unique Markov perfect equilibrium is $\sigma_i^* = 1, \sigma_2^*(a_1) = a_1$, and the state eventually jumps to $\omega' = 1$.

OA.2 Non-Existence of Solution in Pure Strategies

Consider a separable dynamic game with noisy transitions, $N = 2$ players, $\Omega = \{1, 2\}$, $A_i(\omega) = \{1, 2\}$ if $\omega = i$, and $A_i(\omega) = \{1\}$ if $\omega \neq i$. This means that player $i$ makes a nontrivial decision only when $\omega = i$. Flow payoffs are $u_{i,i}(\omega, a_i) = 0$ if $\omega = i$ for all $a_i \in \{1, 2\}$, while

$$u_{1,2}(a_2, 2) = \begin{cases} 10 & \text{if } a_2 = 1, \\ -10 & \text{if } a_2 = 2 \end{cases}$$

and

$$u_{2,1}(a_1, 1) = \begin{cases} -10 & \text{if } a_1 = 1, \\ 10 & \text{if } a_1 = 2. \end{cases}$$

Hence, the flow payoff of player $i$ is 0 when $\omega = i$, but his decision determines whether the flow payoff of player $-i$ is 10 or -10. Transition probabilities are determined as in Example 3 with $\lambda = 1$ and

$$l_1(1 \mid 1, a_1) = \begin{cases} 0 & \text{if } a_1 = 1, \\ 1 & \text{if } a_1 = 2. \end{cases}$$
and

\[ l_2(1 \mid 2, a_2) = \begin{cases} 1 & \text{if } a_2 = 1, \\ 0 & \text{if } a_2 = 2, \end{cases} \]

while \( l_i(i \mid i, a_{-i}) = 1 \). This means that in state \( \omega = i \), player \( i \) (and only player \( i \)) determines a probability distribution over the successor state \( \omega' \). (Note that given \( \lambda, l_1, \) and \( l_2 \), we can construct \( p \) and \( q \) as we did in Example 3.)

We show that the limit conditions (3.2) and (3.3) do not admit a solution in pure strategies. The intuition is similar to the non-existence of a Nash equilibrium in pure strategies in matching pennies. Consider a solution \( \sigma^* \in \text{Equil}^0(< u, p, q, \rho >) \) in pure strategies. If \( \sigma^*_i(\cdot \mid 1) = (1, 0) \), then it must be the case that player 2 is choosing \( \sigma^*_2(1 \mid 2) = 1 \) for otherwise player 1 makes a loss in state \( \omega = 2 \) while he can secure 0 by playing \( a_1 = 2 \). But this would mean that player 2 is willing to make a loss in state \( \omega = 1 \), while he can secure 0 by playing \( a_2 = 2 \). Similarly, it cannot be that \( \sigma^*_1(\cdot \mid 1) = (0, 1) \). Thus, the limit conditions (3.2) and (3.3) do not admit a solution in pure strategies. They do, however, admit a solution in mixed strategies in which player \( i \) chooses \( \sigma_i(\cdot \mid \omega) = (1/2, 1/2) \) when \( \omega = i \).

### OA.3 Generalized Protocol of Moves

We relax Assumption 1 by generalizing our notion a protocol of moves. We allow the evolution of the protocol state \( J \) to depend on players’ actions \( a_J \) and the physical state \( \omega \). We maintain that \( J \) is a partition of the set of players, but allow for a non-uniform stationary distribution. We show that Theorem 1 remains valid.

**Assumption 4 (Generalized Protocol of Moves)** Let \( J \) be a partition of \( \{1, 2, \ldots, N\} \) and \( \mathcal{P} = (\Pr(J' \mid J, \omega, a_J))_{J,J' \in J} \) a \( |J| \times |J| \) transition matrix for all \( \omega \in \Omega \) and all selections \( J \mapsto a_J \in \prod_{j \in J} A_j(\omega) = A_J(\omega) \). Assume that

\[
\mathcal{P}(\cdot \mid \cdot, \omega, \sigma_J) = \sum_{a_J \in \prod_{j \in J} A_j(\omega)} \mathcal{P}(\cdot \mid \cdot, \omega, a_J) \prod_{j \in J} \sigma_j(a_j)
\]

is irreducible for all \( \omega \in \Omega \) and all selections \( J \mapsto \sigma_J \in \prod_{j \in J} \Sigma_j = \Sigma_J \) and its unique stationary distribution \( \pi = (\pi(J))_{J \in J} \in \mathcal{P}(J) \) is independent of \( \omega \) and \( \sigma_J \).

We call \( < J, \mathcal{P} > \) a generalized protocol of moves. Under a generalized protocol of moves, the current physical state and action profile may make a transition from one protocol state to another more likely, but on average all protocol states are visited with frequencies that are independent of physical states and action profiles. The remaining aspects of the model are unchanged.

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17 This example can be easily adapted to show that additive-reward, additive-transition dynamic stochastic games may not admit a Markov perfect equilibrium in pure strategies. It is much simpler than other examples in the literature.
Denote the above game by \( \Gamma = < \Delta, \mathcal{J}, \mathcal{P}, u, p, q, \rho > \) and consider a Markov perfect equilibrium \( \sigma^\Delta = (\sigma^\Delta_i)_{i=1}^N \). Under Assumptions 2 and 3, the discrete-time Bellman equation for a period length of \( \Delta \) is

\[
V_i^\Delta(\omega, J) = |J| \sum_{j \in J} u_{i,j}(\omega, \sigma^\Delta_j(\omega)) \Delta \\
+ \exp(-\rho \Delta) \left\{ \sum_{J' \in J} V_i^\Delta(\omega, J') \sum_{a_j \in A_j(\omega)} \sigma^\Delta_j(a_j \mid \omega) \Pr(J' \mid J, \omega, a_j) \left( 1 - |J| \sum_{j \in J} q_j(\omega, a_j) \Delta \right) \\
+ \sum_{\omega' \neq \omega} \sum_{J' \in J} V_i^\Delta(\omega', J') \sum_{a_j \in A_j(\omega)} \sigma^\Delta_j(a_j \mid \omega) \Pr(J' \mid J, \omega, a_j) \left( |J| \sum_{j \in J} \varphi_j(\omega' \mid \omega, a_j) \Delta \right) \right\} + \mathcal{O}(\Delta^2). \tag{OA.1}
\]

Taking the limit as \( \Delta \to 0 \), we deduce that

\[
V_i^0(\omega, J) = \sum_{J \in J} V_i^0(\omega, J') \Pr(J' \mid J, \omega, \sigma^0_j(\omega)).
\]

Assumption 4 implies that the transition matrix \( \Pr(J' \mid J, \omega, \sigma^0_j) \) has a unique (and uniform) right eigenvector so that \( V_i^0(\omega, J) = V_i^0(\omega, J') \) for all \( J, J' \in J \). Let \( V_i^0 : \Omega \to \mathbb{R} \) be the value function of player \( i \) and \( V^0 = (V_i^0)_{i=1}^N \) be the profile of value functions in the limit as \( \Delta \to 0 \).

The Bellman equation can equivalently be written as

\[
\frac{1}{\Delta} V_i^\Delta(\omega, J) - \frac{\exp(-\rho \Delta)}{\Delta} \sum_{J' \in J} \sum_{a_j \in A_j(\omega)} \sigma^\Delta_j(a_j \mid \omega) V_i^\Delta(\omega, J') \Pr(J' \mid J, \omega, a_j) = |J| \sum_{j \in J} u_{i,j}(\omega, \sigma^\Delta_j(\omega)) \\
+ \exp(-\rho \Delta) |J| \sum_{j \in J} \left\{ \sum_{\omega' \neq \omega} \sum_{J' \in J} \sum_{a_j \in A_j(\omega)} \sigma^\Delta_j(a_j \mid \omega) V_i^\Delta(\omega', J') \Pr(J' \mid J, \omega, a_j) \varphi_j(\omega' \mid \omega, a_j) \\
- \sum_{J' \in J} \sum_{a_j \in A_j(\omega)} \sigma^\Delta_j(a_j \mid \omega) V_i^\Delta(\omega, J') \Pr(J' \mid J, \omega, a_j) q_j(\omega, a_j) \right\} + \mathcal{O}(\Delta).
\]

Multiplying by \( \pi(J) \) and summing over \( J \in \mathcal{J} \) yields

\[
\frac{1}{\Delta} \sum_{J \in \mathcal{J}} \pi(J) V_i^\Delta(\omega, J) - \frac{\exp(-\rho \Delta)}{\Delta} \sum_{J \in \mathcal{J}} \pi(J) \sum_{a_j \in A_j(\omega)} \sigma^\Delta_j(a_j \mid \omega) V_i^\Delta(\omega, J') \Pr(J' \mid J, \omega, a_j) \\
= |J| \sum_{J \in \mathcal{J}} \pi(J) \sum_{j \in J} u_{i,j}(\omega, \sigma^\Delta_j(\omega)) \\
+ \exp(-\rho \Delta) |J| \sum_{J \in \mathcal{J}} \pi(J) \left\{ \sum_{\omega' \neq \omega} \sum_{J' \in \mathcal{J}} V_i^\Delta(\omega', J') \Pr(J' \mid J, \omega, \sigma^\Delta_j(\omega)) \varphi_j(\omega' \mid \omega, \sigma^\Delta_j(\omega)) \\
- \sum_{J' \in \mathcal{J}} V_i^\Delta(\omega, J') \Pr(J' \mid J, \omega, \sigma^\Delta_j(\omega)) q_j(\omega, \sigma^\Delta_j(\omega)) \right\} + \mathcal{O}(\Delta^2).
\]
Using the facts that $\sum_{J \in \mathcal{J}} \Pr(J'|J, \omega, \sigma_J(\omega)) \pi(J) = \pi(J')$ and $\sum_{J' \in \mathcal{J}} \Pr(J'|J, \omega, \sigma_J(\omega)) = 1$, and taking the limit as $\Delta \to 0$, we obtain the continuous-time Bellman equation

$$
\rho V^0_i(\omega) = |\mathcal{J}| \sum_{J \in \mathcal{J}} \pi(J) \sum_{j \in J} u_{i,j}(\omega, \sigma^0_j(\omega)) + |\mathcal{J}| \sum_{J' \in \mathcal{J}} \pi(J) \sum_{\omega' \neq \omega} \left( \sum_{\omega'' \neq \omega} V^0_i(\omega'') \varphi_{i,j}(\omega' | \omega, \sigma^0_j(\omega)) - V^0_i(\omega) q_j(\omega, \sigma^0_j(\omega)) \right).
$$

(OA.2)

The discrete-time optimality condition for a period length $\Delta$ is

$$
\sigma^\Delta_i(a_i | \omega) > 0 \Rightarrow a_i \in \arg \max_{\hat{a}_i \in A_i(\omega)} u^\Delta_i(\omega, J, \hat{a}_i, \sigma^\Delta_{J \setminus \{i\}}(\omega)) + \exp(-\rho \Delta) \sum_{\omega' \in \Omega} \sum_{J' \in \mathcal{J}} \sum_{a_{J' \setminus \{i\}}} \sigma_{J' \setminus \{i\}}(a_{J' \setminus \{i\}} | \omega) V^\Delta_i(\omega', J') \Pr(J'|J, \omega, a_i, a_{J' \setminus \{i\}}) \Pr^\Delta(\omega' | \omega, J, \hat{a}_i, a_{J \setminus \{i\}}).
$$

Dividing by $\Delta$, rearranging terms, and taking the limit as $\Delta \to 0$, we deduce the continuous-time optimality condition

$$
\sigma^0_i(a_i | \omega) > 0 \Rightarrow a_i \in \arg \max_{\hat{a}_i \in A_i(\omega)} u_{i,j}(\omega, \hat{a}_i) + \sum_{\omega' \neq \omega} \left( V^0_i(\omega'') - V^0_i(\omega) \right) \varphi_i(\omega' | \omega, \hat{a}_i). \quad \text{(OA.3)}
$$

Conditions (OA.2) and (OA.3) are the analogs of conditions (3.2) and (3.3) for a generalized protocol of moves.

Consider the generalized protocols of moves $< \mathcal{J}_1, \mathcal{P}_1 >$ and $< \mathcal{J}_2, \mathcal{P}_2 >$ with stationary distributions $\pi_1$ and $\pi_2$. For all $j = 1, \ldots, N$, define $J_1(j)$ to be the unique element in $\mathcal{J}_1$ such that $j \in J_1(j)$. Define $J_2(j)$ analogously. We say that the generalized protocols of moves $< \mathcal{J}_1, \mathcal{P}_1 >$ and $< \mathcal{J}_2, \mathcal{P}_2 >$ are comparable if $|\mathcal{J}_1| \pi_1(J_1(j)) = |\mathcal{J}_2| \pi_2(J_2(j))$ for all $j = 1, \ldots, N$. Given a protocol $< \mathcal{J}, \mathcal{P} >$, a player $j$ moves a fraction $\pi(J(j))$ of the time and has an impact on payoffs and transitions which is scaled by $|\mathcal{J}|$. Thus, comparability means that the total impact of a player’s strategy on payoffs and transitions does not depend on the particular protocol that we use in the model. All protocols that satisfy Assumption 1 are comparable.

Theorem 1 remains valid for generalized protocols of moves that are comparable \(^{18}\).

**Theorem 3 (Generalized Protocol-Invariance Theorem)** Fix $p$, $q$, and $\rho$. For almost all $u$, all generalized protocols of moves $< \mathcal{J}, \mathcal{P} >$ and $< \overline{\mathcal{J}}, \overline{\mathcal{P}} >$ that are comparable, and all $\varepsilon > 0$, there exists $\Delta > 0$ such that for all $\Delta < \Delta$ and $\sigma \in \text{Equil}(< \Delta, \mathcal{J}, \mathcal{P}, u, p, q, \rho >)$, there exists $\overline{\sigma} \in \text{Equil}(< \Delta, \overline{\mathcal{J}}, \overline{\mathcal{P}}, u, p, q, \rho >)$ such that $||\sigma - \overline{\sigma}|| < \varepsilon$.

**OA.4 Non-Partition Protocol of Moves**

We relax Assumption 1 and assume that $\mathcal{J}$ is not a partition of the set of players but contains subsets $J \subseteq \{1, \ldots, N\}$ such that for all $i = 1, \ldots, N$, there exists $J \in \mathcal{J}$ such that $i \in J$. This allows player $i$

\(^{18}\)Strictly speaking, here we only show that Lemma 1 remains valid. The proof that Lemma 2 remains valid is available upon request.
to have the move in conjunction with different sets of rivals. To simplify the exposition, we assume that |
\{J \in \mathcal{J} \mid i \in J\}| = \kappa for all i = 1, \ldots, N. As before, there is an irreducible Markov chain \( \mathcal{P} \) defined on \( \mathcal{J} \) that has a unique stationary distribution that is uniform on \( \mathcal{J} \). We call \( <\mathcal{J}, \mathcal{P}> \) a non-partition protocol of moves.

With a non-partition protocol of moves \( <\mathcal{J}, \mathcal{P}> \), the per-period payoff \( u_i^\Delta(\omega, J, a_J) \) is written as

\[
u_i^\Delta(\omega, J, a_J) = \frac{|J|}{\kappa} \sum_{j \in J} u_{i,j}(\omega, a_j) \Delta + O(\Delta^2),
\]

and the hazard rate \( q_J(\omega, a_J) \) and transition probability \( p_J(\omega' \mid \omega, a_J) \) are written as

\[
q_J(\omega, a_J) = \frac{|J|}{\kappa} \sum_{j \in J} q_j(\omega, a_j)
\]

and

\[
q_J(\omega, a_J)p_J(\omega' \mid \omega, a_J) = \frac{|J|}{\kappa} \sum_{j \in J} q_j(\omega, a_j)p_j(\omega' \mid \omega, a_j),
\]

where \( q_j : \{ (\omega, a_j) \mid a_j \in A_j(\omega) \} \rightarrow \mathbb{R}^+ \cup \{0\} \) and \( p_j : \{ (\omega, a_j) \mid a_j \in A_j(\omega) \} \rightarrow \mathcal{P}(\Omega) \). The remaining aspects of the model are unchanged.

With a non-partition protocol of moves \( <\mathcal{J}, \mathcal{P}> \), the identity of the players that have the move in conjunction with player \( i \) is a state variable. Thus, a Markovian strategy for player \( i \) is a function \( \sigma_i : \Omega \times \{ J \in \mathcal{J} \mid i \in J \} \rightarrow \cup_{\omega \in \Omega} \mathcal{P}(A_i(\omega)) \). Overloading notation, we use \( \text{Equil}(<\Delta, \mathcal{J}, \mathcal{P}, u, p, q, \rho>) \) to denote the set of Markov perfect equilibria. We say that a Markov perfect equilibrium \( \sigma \) is simple if \( \sigma_i(a_i \mid \omega, J) = \sigma_i(a_i \mid \omega, \hat{J}) \) for all \( i = 1, \ldots, N, \omega \in \Omega, a_i \in A_i, \) and all \( J, \hat{J} \in \mathcal{J} \). In this case, we write \( \sigma_i(a_i \mid \omega) \).

The following proposition partially extends Theorem 1 to a non-partition protocol of moves:

**Proposition 6** Assume \( \text{Equil}^0(<u, p, q, \rho>) \) only contains strict solutions. Then there exists \( \bar{\Delta} > 0 \) such that for all \( \Delta < \bar{\Delta} \), \( \sigma \in \text{Equil}(<\Delta, \mathcal{J}, \mathcal{P}, u, p, q, \rho>) \) is simple and

\[
\text{Equil}(<\Delta, \mathcal{J}, \mathcal{P}, u, p, q, \rho>) = \text{Equil}^0(<u, p, q, \rho>).
\]

In contrast to Theorem 1, Proposition 6 restricts attention to strict and thus pure solutions. When mixed solutions are considered, the limit conditions may have a continuum of solutions if players use the payoff-irrelevant realization of \( J \) to randomize over actions and our differential topology tools therefore cannot be directly applied.

**Proof.** Consider a sequence \( (\sigma^\Delta) \) with \( \sigma^\Delta \in \text{Equil}(<\Delta, \mathcal{J}, \mathcal{P}, u, p, q, \rho>) \) and \( \sigma^\Delta \rightarrow \sigma^0 \) (possibly through a subsequence) as \( \Delta \rightarrow 0 \). Let \( V^\Delta \) be the profile of value functions corresponding to \( \sigma^\Delta \) and assume it converges to \( V^0 \). Similar to Section 3, we can deduce that \( V^0(\omega, J) \) does not depend on \( J \in \mathcal{J} \) and simply
write $V^0(\omega)$. We can also follow Section 3 to deduce that

$$
\rho V_i^0(\omega) = \frac{1}{\kappa} \sum_{J \in \mathcal{J}} \sum_{j \in J} \left( u_{i,j}(\omega, \sigma_j(\omega, J)) + \sum_{\omega' \neq \omega} (V_i^0(\omega') - V_i^0(\omega)) \varphi_j(\omega' | \omega, \sigma_j(\omega, J)) \right)
$$

and

$$
\sigma_i^0(a_i | \omega, J) > 0 \Rightarrow a_i \in \arg \max_{\tilde{a}_i \in A_i(\omega)} u_{i,i}(\omega, \tilde{a}_i) + \sum_{\omega' \neq \omega} (V_i^0(\omega') - V_i^0(\omega)) \varphi_i(\omega' | \omega, \tilde{a}_i).
$$

(4.4)

Define $\tilde{\sigma}_i(\cdot | \omega) = \frac{1}{\kappa} \sum_{J \in \mathcal{J}} \sigma_i^0(\cdot | \omega, J)$ for all $i = 1, \ldots, N$ and all $\omega \in \Omega$, and note that $(\tilde{\sigma}, V^0)$ is a solution to the limit conditions (3.2) and (3.3). Since Equil$^0(< u, p, q, \rho >)$ only contains strict solutions, the profile $\tilde{\sigma} = (\tilde{\sigma}_i)_{i=1}^N$ must be a strict solution and thus the maximization problem in equation (4.4) has a unique solution. Therefore $\sigma_i^0(a_i | \omega, J)$ does not depend on $J$ and $\sigma^0$ is simple. In particular, $\sigma^0 \in \text{Equil}^0(< u, p, q, \rho >)$ and therefore there exists $\Delta > 0$ such that for all $\Delta < \Delta$, $\sigma^0 \in \text{Equil}^0(< u, p, q, \rho >)$. To see the converse, note that Equil$^0(< u, p, q, \rho >)$ has a finite number of pure solutions that are all strict. For any solution $(\sigma^0, V^0)$ to the limit conditions (3.2) and (3.3), $\sigma^0$ satisfies the conditions for a (simple) Markov perfect equilibrium of a separable dynamic game with noisy transitions and the non-partition protocol of moves $< J, P >$ since the continuation values in such a game converge to $V^0$ and, as a result, the incentive constraints are satisfied if $\Delta > 0$ is sufficiently small. ■