

(Un-)Common Preferences, Ambiguity, and Coordination*

Simone Cerreia-Vioglio^a, Roberto Corrao^b, Giacomo Lanzani^b

^aUniversità Bocconi and ^bMIT

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Abstract

We study the “common prior” assumption and its implications when agents have differential information and preferences beyond subjective expected utility (SEU). We consider consequentialist interim preferences that are consistent with respect to the same ex-ante evaluation and characterize the latter in terms of extreme limits of higher-order expectations. Notably, agents are mutually dynamic consistent with respect to the same ex-ante evaluation if and only if all the limits of higher-order expectations are the same, extending beyond SEU the classical characterization of the common prior assumption in Samet [55]. Within this framework, we characterize the properties of equilibrium prices in financial beauty contests (and other coordination games) in terms of the agents’ private information, coordination motives, and attitudes toward uncertainty. Differently from the SEU case, the limit price does not coincide in general with the common ex-ante expectation. Moreover, when the agents share the same benchmark probabilistic model, high-coordination motives make their concern for misspecification disappear in equilibrium, exposing them to a divergence between the market price and the fundamental value of the security.

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1 Introduction

The common prior assumption is one of the most used and debated concepts in economic theory.¹ When the agents are subjective expected utility (SEU) maximizers, this assumption captures the idea of mutual ex-ante agreement on the preferences over uncertain prospects. However, preferences that do not reduce uncertainty to a single probability are normatively convincing and consistent with experimental findings.² Notably, these departures are consistent with the rationality of decision makers that acknowledge their *ambiguity* (or “Knightian uncertainty”, see Gilboa and Marinacci [23]) about an objective probabilistic model and have nonneutral attitudes toward it. Therefore, it is crucial to understand whether the ex-ante mutual agreement can be expressed independently of agents’ attitudes toward ambiguity and, in this case, to study the implications for the agents’ interim preferences and behavior of this mutual agreement. This paper answers these questions by formalizing increasing degrees of mutual ex-ante agreement among agents with differential information and rational preferences (cf. Cerreia-Vioglio et al. [8]) such as SEU, Maxmin expected utility, Choquet expected utility, variational preferences, and, in general, uncertainty averse preferences.³

We first impose restrictions on the agents’ interim preferences that guarantee the existence of a *single* ex-ante preference that is *jointly* “consistent” for all the agents. Next, we show that, as for the baseline SEU case, all these restrictions can be fully characterized by properties of the *higher-order interim preferences* of the agents. In turn, this directly allows us to study the implications of the ex-ante agreement for strategic reasoning and market behavior, which we address in the second and third parts of the paper, respectively.

We then embed rational preferences in standard coordination games (e.g., beauty contests and price competitions) and we derive a complete characterization of equilibrium behavior in the high-coordination limit in terms of the agents’ higher-order preferences without any ex-ante agreement restriction. However, when we impose some ex-ante agreement, we find a striking result: the desire for coordination considerably tames the attitudes toward uncertainty, and the limit equilibrium behavior in some critical cases is indistinguishable from the ones obtained under SEU. Finally, we provide necessary and sufficient conditions for ex-ante agreement in terms of no trade, highlighting a gap between these two that is specific to non-SEU preferences.

¹See, for example, Morris [49] and Bonanno and Nehring [5] for complete discussions on the foundation and the role of the common prior assumption in economics.

²See, for example, the survey by Trautmann and Van De Kuilen [64].

³Maxmin expected utility and Choquet expected utility were introduced respectively by Gilboa and Schmeidler [24] and Schmeidler [57] by relaxing the independence axiom of SEU. For Maxmin preferences, c-independence and uncertainty aversion replace the standard independence axiom. The variational preferences of Maccheroni et al. [44] are a generalization of Maxmin preferences generated by an even weaker form of independence and maintain uncertainty aversion. Finally, the uncertainty averse preferences of Cerreia-Vioglio et al. [10] completely relax independence while still maintaining uncertainty aversion.

Common ex-ante preferences and beyond First, we generalize the notion of conditional expectation for preferences that are not necessarily SEU, but just rational. We start with a pair of *ex-ante* and *interim expectations*, modeling the preferences of the agent before and after the arrival of information, and require them to be “consistent” in the sense that they jointly exhibit attitudes for gradual and one-shot resolution of uncertainty. These consistency properties, that are satisfied by several existing updating rules for non SEU preferences (e.g., full Bayesian updating and proxy updating for maxmin preferences), give rise respectively to the notion of *lower* and *upper conditional expectations*. When both consistency properties are satisfied, that is, when the agent has neutral attitudes for the timing of resolution of uncertainty, we define the notion of *nonlinear conditional expectation*. Here, we maintain the standard properties of consequentialism and dynamic consistency of conditional SEU, while relaxing linearity.

Armed with this novel taxonomy, we move to analyze a multi-agent setting with differential information. We extend the notion of (*nonlinear*) *higher-order expectations* to rational preferences capturing the idea of preferences over acts formed by the evaluations attached by other agents to an original act. This notion is essential for our analysis and is illustrated through a simple asset-pricing model where agents care about the willingness to pay of the other traders rather than the fundamental value of an asset. Our first result shows that, under a full-support condition and the presence of null public information, the higher-order expectations over sequences of agents converge to a state-independent limit, provided that all the agents appear infinitely often in the sequence. This seemingly technical result greatly generalizes the equivalent result of Samet [55] beyond SEU and lies the foundation of our analysis. Moreover, it is easily illustrated in our asset pricing example by implying the existence of a well-defined and state-independent equilibrium price that, in principle, only depends on the order of trades among agents.

Next, we say that agents share a *lower* (resp. *upper*) *common ex-ante expectation* if their conditional preferences exhibit attitudes for gradual (resp. one-shot) resolution of uncertainty mutually with respect to the same ex-ante expectation. The interpretation is that, before observing their private information, the agents share the same perceived ambiguity about the probabilistic model and the same attitude toward it. Then, in the interim stage, the agents’ preferences may differ, but only insofar the nature of their private information was different. Therefore, our consistency properties impose restrictions between periods for each individual as well as restrictions across all individuals. For every profile of interim expectations, there always exist a lower and an upper common ex-ante preferences exhibiting the minimal degrees of aversion and attraction for one-shot resolution of uncertainty. Under the deterministic convergence property highlighted above, these ex-ante preferences are characterized via the extreme limits of higher-order expectations of the agents.

As before, when all the agents have neutral attitudes for timing of resolution of uncertainty, we say that they share a *nonlinear common ex-ante expectation*: all the agents are dynamically

consistent with respect to the same unconditional preference.⁴. In other words, we weaken the assumption of mutual agreement about an objective probabilistic model to that of *mutual dynamic consistency* with respect to a common ex-ante rational preference. We provide a characterization of the existence of a nonlinear common ex-ante expectation that purely concerns the interim preferences of the agents. There is a nonlinear common ex-ante expectation if and only if all the interim higher-order (nonlinear) expectations of the agents converge to the same limit, which coincides with the nonlinear common ex-ante expectation. On the one hand, this result significantly generalizes the characterization of the common prior assumption in Samet [55]. On the other hand, it points out that it is the *invariance property* of dynamic consistency that allows us to characterize mutual ex-ante agreement in terms of interim higher-order beliefs, as opposed to the probabilistic nature of beliefs. However, dynamic consistency with respect to all the information structures of the agents is very restrictive under ambiguity averse preferences, as pointed for example by Ellis [14]. Therefore, our result implies that the order of traders in our asset pricing example is generally relevant for the equilibrium price, and more so with ambiguity aversion.

Finally, for variational preferences, we consider an intermediate form of mutual consistency where the agents only share some ex-ante benchmark (i.e., most trusted) probabilistic models, but their interim preferences are otherwise unrelated. This further generalization allows us to consider coordination and market models where the agents share a common perception of the uncertain data-generating process(es) they face but have a heterogeneous level of confidence in it.

Coordination and ambiguity We next move to the implications for coordination games of the assumptions on an ex-ante agreement under variational preferences. We first consider an application of our result to network beauty contests in asset markets under incomplete information. Here, we show that, under connectedness of the network structure, the (bid) prices in the unique equilibrium become independent of the state and agent as the coordination motives prevail. Notably, we provide bounds on the equilibrium price dispersion that only depends on the joint connectivity of the network and information structure.

Next, we analyze the unique equilibrium price in the limit for strong coordination motives. In general, this limit is characterized by a worst-case weighted average of the benchmark interim expectations of the agent. With this result, we can already see that a significant part of the ambiguity aversion of the agents disappears in the limit equilibrium, as all the probabilistic models that are not maximally trusted become irrelevant. Moreover, we can provide bounds on the limit evaluation of the asset in terms of the ex-ante preferences that we presented, thereby

⁴Importantly, the information structures of the agents are fixed throughout the entire analysis. The assumption of dynamic consistency concerning only a fixed information structure is weak enough to include a much richer set of preferences than SEU.

assessing the price effect of interim information.

Our theorem implies that whenever the agents share the same unique ex-ante benchmark probability model, the limit equilibrium price collapses to the expectation of the value of the asset for this unique benchmark. This establishes a strong irrelevance result: as coordination motives prevail, the limit price is unaffected by ambiguity aversion, the information structure, or the network structure.⁵ In turn, this has important implications for our financial beauty contest application. If the common benchmark probability model of the agents is misspecified, then our result implies a mispricing with respect to the true fundamental value of the asset, despite agents that are concerned for misspecification. Intuitively, the agents attach a much higher value to coordination than to the fundamental value of the asset, hence, in equilibrium, they have no reason to reduce their willingness to pay due to the concern for misspecification of the *shared* benchmark probability model.

In general, if a nonlinear common ex-ante expectation exists, then the limit price can lie strictly above the ex-ante preference, pointing out a key difference with the limit result under SEU of Golub and Morris [27]. However, this wedge exist only if the agents are ambiguous with respect to each other information structure. Indeed, when agents are unambiguous about the aggregate information, the standard limit equivalence of the SEU case is restored. Notably, in this case, agents might still perceive ambiguity about the fundamental, and their full-coordination limit price decreases in their ambiguity aversion.

The previous results depend only on the best-response structure of the game. In particular, we can derive the same best-response functions from different games with strong coordination motives. An example is a price-competition game where firms produce partially differentiated goods under incomplete information about the demand function.

No-trade implications Finally, we study the relation between the existence of a nonlinear common ex-ante expectation and no-trade implications, which are usually used to characterize the common-prior assumption under SEU (cf. Morris [48], Samet [56], Feinberg [18], Gizatulina and Hellman [25]). Here, we first show that if two agents with variational interim preferences are willing to trade the same asset in any state, then they cannot be mutually dynamically consistent with respect to the same ex-ante preference. As already established, the exact converse of this statement does not hold in general (cf. Dow and Werlang [13]). However, the existence of a common prior is implied by the following stronger no-trade condition. Suppose there is no endowment economy with two large populations of agents, each characterized by one of the primitive interim expectations, where trade can create a Pareto gain in every state. In that case, these interim expectations are consistent with respect to the same ex-ante preference.

⁵The irrelevance of the latter two aspects was established by [27] for a similar class of beauty contests.

2 Nonlinear conditional expectations

In this section, we introduce nonlinear conditional expectations. We start by recalling the usual notion of (linear) conditional expectation. This will set the stage to formalize our main theoretical questions: when the expectations of different agents holding different private information can be seen as generated by a common perception of uncertainty and how this affects equilibrium prices in asset pricing beauty contest models. As in Samet [55], we consider a finite state space Ω endowed with the power set $\mathcal{P}(\Omega)$. We denote by Δ the set of all probabilities over Ω . We let Π denote a partition of Ω , and for every $\omega \in \Omega$, we let $\Pi(\omega)$ denote the unique element of Π that contains ω . Finally, we endow \mathbb{R}^Ω with the supnorm.

2.1 Linear case

Consider a probability $\mu \in \Delta$ and denote by $\mathbb{E}_\mu : \mathbb{R}^\Omega \rightarrow \mathbb{R}$ the functional $\mathbb{E}_\mu(f) = \int f d\mu$. If Π is a partition of Ω , then a map $p_\mu : \Omega \times \mathcal{P}(\Omega) \rightarrow [0, 1]$ is a regular conditional probability of μ given Π if and only if: (i) For each $\omega \in \Omega$ the function $p_\mu(\omega, \cdot) \in \Delta$; (ii) For each $F \in \mathcal{P}(\Omega)$ the function $p_\mu(\cdot, F) : \Omega \rightarrow [0, 1]$ is a version of the conditional probability of F given Π .

Since Ω is finite, any probability μ on Ω admits a regular conditional probability p_μ . Moreover, the function $V_\mu : \Omega \times \mathbb{R}^\Omega \rightarrow \mathbb{R}$, defined by

$$V_\mu(\omega, f) = \int f dp_\mu(\omega, \cdot) \quad \forall \omega \in \Omega, \forall f \in \mathbb{R}^\Omega,$$

is a regular conditional expectation and has the following properties:

- a. For each $\omega \in \Omega$ the functional $V_\mu(\omega, \cdot) : \mathbb{R}^\Omega \rightarrow \mathbb{R}$ is normalized, monotone, and linear;⁶
- b. For each $f \in \mathbb{R}^\Omega$ the function $V_\mu(\cdot, f) : \Omega \rightarrow \mathbb{R}$ is Π -measurable and satisfies

$$\mathbb{E}_\mu(f) = \mathbb{E}_\mu(V_\mu(\cdot, f)) \text{ and } V_\mu(\omega, f 1_{\Pi(\omega)} + h 1_{\Pi(\omega)^c}) = V_\mu(\omega, f) \quad \forall \omega \in \Omega, \forall h \in \mathbb{R}^\Omega. \quad (1)$$

In words, (1) contains two properties: the law of iterated expectations and that the update of μ assigns probability one to the realized partition cell. The functionals \mathbb{E}_μ and V_μ can be axiomatized as the conditional representation of a subjective expected utility (SEU) decision maker who then satisfies dynamic consistency and consequentialism.

2.2 Nonlinear case

Mimicking what we discussed above, we consider two functions $\bar{V} : \mathbb{R}^\Omega \rightarrow \mathbb{R}$ and $V : \Omega \times \mathbb{R}^\Omega \rightarrow \mathbb{R}$. In terms of interpretation, the functional $\bar{V}(f)$ is the unconditional expectation of f while $V(\cdot, f)$ describes its conditional expectation.

⁶ A functional $T : \mathbb{R}^\Omega \rightarrow \mathbb{R}$ is normalized if and only if $T(k 1_\Omega) = k$ for all $k \in \mathbb{R}$.

Definition 1. Let $\bar{V} : \mathbb{R}^\Omega \rightarrow \mathbb{R}$. We say that \bar{V} is an *ex-ante* (generalized) expectation if and only if \bar{V} is normalized and monotone.

This definition amounts to say that the preference \succsim represented by an *ex-ante* expectation \bar{V} is *rational* as in [8]. Monotonicity is a conceptual (although mild) requirement implying that the agents prefer larger monetary outcomes, whereas normalization requires that the representing \bar{V} is the certainty equivalent for the preference. Moreover, under normalization, the comparative notion of ambiguity aversion of Ghirardato and Marinacci [22] is easily characterized: \bar{V} is more ambiguity averse than \bar{V}' if and only if $\bar{V}(f) \leq \bar{V}'(f)$ for all $f \in \mathbb{R}^\Omega$.⁷

Definition 2. Fix a partition Π and $V : \Omega \times \mathbb{R}^\Omega \rightarrow \mathbb{R}$. We say that (V, Π) is an *interim* (generalized) expectation if and only if for each $\omega \in \Omega$ the functional $V(\omega, \cdot) : \mathbb{R}^\Omega \rightarrow \mathbb{R}$ is normalized, monotone, and continuous and the function $V(\cdot, f) : \Omega \rightarrow \mathbb{R}$ is Π -measurable and satisfies

$$V(\omega, f1_{\Pi(\omega)} + h1_{\Pi(\omega)^c}) = V(\omega, f) \quad \forall \omega \in \Omega, \forall f, h \in \mathbb{R}^\Omega. \quad (2)$$

A conditional expectation is a pair formed by an *ex-ante* (generalized) expectation and an *interim* (generalized) expectation that satisfies some consistency properties.

Definition 3. Let (V, Π) be an *interim* expectation.

1. We say that (V_\circ, V, Π) is a *lower conditional expectation* if and only if V_\circ is an *ex-ante* expectation such that

$$V_\circ(f) \leq V_\circ(V(\cdot, f)) \quad \forall f \in \mathbb{R}^\Omega. \quad (3)$$

2. We say that (V°, V, Π) is an *upper conditional expectation* if and only if V° is an *ex-ante* expectation such that

$$V^\circ(f) \geq V^\circ(V(\cdot, f)) \quad \forall f \in \mathbb{R}^\Omega. \quad (4)$$

3. We say that (\bar{V}, V, Π) is a *nonlinear conditional expectation* if and only if it is both a lower and an upper conditional expectation.

Compared to standard conditional expectations, a nonlinear conditional expectation only relaxes the assumption of linearity from both \bar{V} and V , indeed point 3 implies that

$$\bar{V}(f) = \bar{V}(V(\cdot, f)) \quad \forall f \in \mathbb{R}^\Omega. \quad (5)$$

From a preferential viewpoint, this is tantamount to weaken the assumption of independence, but still retain consequentialism and dynamic consistency. Consequentialism takes care of (2),

⁷Compared to the standard definition of rational preferences, we are implicitly assuming that the utility index $u : \mathbb{R} \rightarrow \mathbb{R}$ coincides to the identity function. In the multi-agent setting of Section 3, this assumption is without loss of generality as long as the risk preferences of all agents are homogeneous, since we can always interpret each $f \in \mathbb{R}^\Omega$ as a utility act.

while dynamic consistency is the main axiom behind the law of iterated expectations in (5).⁸ However, it is well known that full-fledged dynamic consistency is restrictive outside the realm of subjective expected utility, especially with uncertainty averse preferences (see for example [2], [4], [15], [20], and [60]). Therefore, in points 1 and 2, we consider ex-ante preferences that are consistent with the interim expectation yet possibly exhibiting a strict preference for gradual or one-shot resolution of uncertainty.

Whenever the agent has a lower conditional expectation, her interim preference can be rationalized by V_\circ provided that it exhibits preferences for gradual resolution of uncertainty (cf. Dillenberger [12] and Strzalecki [62]). Equation (3) is also equivalent to require that the agent attaches a positive value to her information Π .⁹ Moreover, such condition is satisfied by existing updating rules for preferences under uncertainty as we next show.

Example 1 (Choquet expected utility with proxy updating). We analyze the class of preferences and updating rule recently proposed by Gul and Pesendorfer [28]. Consider a totally monotone capacity $\nu : 2^\Omega \rightarrow [0, 1]$ and a partition Π .¹⁰ In the ex-ante stage, the agent evaluates every act $f \in \mathbb{R}^\Omega$ with the Choquet integral of f with respect to ν , denoted as $V_\circ(f)$. Recall that the core of ν is defined as

$$\text{core}(\nu) = \{\mu \in \Delta(\Omega) : \forall E \in 2^\Omega, \mu(E) \geq \nu(E)\}$$

and that $V_\circ(f) = \min_{\mu \in \text{core}(\nu)} \mathbb{E}_\mu(f)$. We let $\mu_\nu \in \Delta(\Omega)$ denote the Shapley value corresponding to ν . The interim expectations at state ω are:

$$V(\omega, f) = \min_{\mu \in \text{core}_\circ(\nu)} \mathbb{E}_{p_\mu(\omega, \cdot)}(f) \quad \forall f \in \mathbb{R}^\Omega$$

where $p_\mu(\omega, \cdot)$ is the conditional probability of μ given Π and

$$\text{core}_\circ(\nu) = \{\mu \in \text{core}(\nu) : \forall E \in \Pi, \mu(E) = \mu_\nu(E)\}. \quad (6)$$

In words, each agent updates her preferences with full Bayesian updating but starting from the restricted set $\text{core}_\circ(\nu)$. In this case, the results in [28, Axiom C.4 and Theorem 1] imply that (V_\circ, V, Π) is a lower conditional expectation. \blacktriangle

Instead, an upper conditional expectation rationalizes the interim expectation of the agent provided that it exhibits preferences for one-shot resolution of uncertainty. The next example shows that with maxmin expected utility functionals (see Gilboa and Schmeidler [24]) and full Bayesian updating, we obtain an upper conditional expectation.

⁸Online Appendix F contains a simple axiomatic foundation for nonlinear conditional expectations.

⁹Formally, equation (3) is equivalent to assume that, for each finite set of acts $A \subseteq \mathbb{R}^\Omega$, $V_\circ(\max_{f \in A} V(\cdot, f)) \geq \max_{f \in A} V_\circ(f)$.

¹⁰A capacity ν is totally monotone if and only, for all $k \geq 2$ and all $E_1, \dots, E_k \in 2^\Omega$,

$$\nu(\cup_{i=1}^n E_i) \geq \sum_{\{J: \emptyset \neq J \subseteq \{1, \dots, k\}\}} (-1)^{|J|+1} \nu(\cap_{j \in J} E_j).$$

Example 2 (Maxmin expected utility with full Bayesian updating). Let C be a compact and convex set of probabilities over Ω and let Π be a partition. Define

$$V_C^\circ(f) = \min_{\mu \in C} \mathbb{E}_\mu(f) \quad \forall f \in \mathbb{R}^\Omega \quad (7)$$

and

$$V_C(\omega, f) = \min_{p \in C_\omega} \mathbb{E}_p(f) \quad \forall \omega \in \Omega, \forall f \in \mathbb{R}^\Omega, \quad (8)$$

where, for all $\omega \in \Omega$, $C_\omega = \{p_\mu(\omega, \cdot) : \mu \in C\}$ and p_μ is the regular conditional probability of μ given Π . Then (V_C°, V_C, Π) is an upper conditional expectation. Moreover, it is well known that if C is rectangular for Π (see Epstein and Schneider [16]),¹¹ then (V_C°, V_C, Π) is a nonlinear conditional expectation. Linear expectations are obtained when C is a singleton and rectangularity in that case is trivially satisfied. \blacktriangle

The observation that Bayesian updating induces an upper common ex-ante expectation also holds for the class of divergence preferences introduced in [44]. We illustrate this class in Example 4 below where we consider multiplier preferences a la [32].

We close this section with a useful, yet technical, notion of full support for rational preferences. Indeed, as [55], we mostly focus on the case of full support which we next discuss.¹² Given states $\bar{\omega}, \omega \in \Omega$, we say that $\bar{\omega}$ is \bar{V} -essential (resp., $V(\omega, \cdot)$ -) if and only if there exists an $\varepsilon > 0$ such that for each $f \in \mathbb{R}^\Omega$ and for each $\delta \geq 0$

$$\bar{V}(f + \delta 1_{\{\bar{\omega}\}}) - \bar{V}(f) \geq \varepsilon \delta \quad (\text{resp., } V(\omega, f + \delta 1_{\{\bar{\omega}\}}) - V(\omega, f) \geq \varepsilon \delta). \quad (9)$$

In the linear case, we clearly have that $\bar{\omega}$ belongs to the support of μ (resp., $p_\mu(\omega, \cdot)$) if and only if $\bar{\omega}$ is \bar{V} -essential (resp., $V(\omega, \cdot)$ -essential). For the general case, we say that \bar{V} (resp., $V(\omega, \cdot)$) has *full support* if and only if each $\bar{\omega} \in \Omega$ (resp., each $\bar{\omega} \in \Pi(\omega)$) is \bar{V} -essential (resp., $V(\omega, \cdot)$ -essential). Moreover, we say that an interim expectation (V, Π) has full support if and only if $V(\omega, \cdot)$ has full support for all $\omega \in \Omega$.

3 (Un-)common ex-ante preferences

We now consider a finite set of agents $I = \{1, \dots, n\}$, each endowed with an interim expectation (V_i, Π_i) . Given the collection of partitions $\{\Pi_i\}_{i \in I}$ for the agents, that is, an *information struc-*

¹¹ C is rectangular if and only if $C = \left\{ \sum_{l=1}^L p_{\mu_l}(E_l, \cdot) \mu(E_l) : \mu, \mu_1, \dots, \mu_L \in C \right\}$, where $\Pi = \{E_1, \dots, E_L\}$ and we denote the update on the E_l cell by $p_\mu(E_l, \cdot)$.

¹²Corollary 1, our extension of Samet's Theorem [55], does not rely on the full-support assumption per se but rather on a regularity condition of the sequences of higher-order beliefs (cf. Definition 4). Our full-support condition, paired with the absence of non-trivial public information, implies that the regularity condition holds. However, this can be verified directly and independently of the full-support assumption (cf. Example 4 and Remark 1).

ture, we denote by Π_{sup} and Π_{inf} the *meet* and the *join* of the partitions.¹³ They respectively correspond to the public information among agents and the aggregate information collectively held by the agents.

In a multi-agent setting, it might be convenient to view V_i as an operator from \mathbb{R}^Ω to \mathbb{R}^Ω . In this case, the j -th component of this operator is $V_i(\omega_j, f)$ for all $f \in \mathbb{R}^\Omega$. With a small abuse of notation, we will still denote this operator by V_i . This rewriting turns out to be useful in order to formally discuss higher-order expectations. For instance, given two agents $i, j \in I$ and an act $f \in \mathbb{R}^\Omega$, the expectation of agent i at state ω about the evaluation of act f by agent j is $V_i(\omega, V_j(f))$. Moreover, if we do not fix a state $\omega \in \Omega$, we obtain the second-order evaluation (of i through j) $V_i \circ V_j : \mathbb{R}^\Omega \rightarrow \mathbb{R}^\Omega$. We next illustrate the relevance of this concept in a stylized asset-pricing model.

Example 3 (Forecasting the forecaster). Consider a state-contingent asset $f \in \mathbb{R}^\Omega$ in a discrete-time economy with $t \in \mathbb{N}$ periods. Each index $i \in I$ represents a continuum of speculative traders with the same interim expectations (V_i, Π_i) . Let $(i_1, \dots, i_t) \in I^t$, with $t \in \mathbb{N}$, be a finite sequence of agents' classes. At period 0, an external agent is endowed with the asset. At period 1 she has to sell the asset to one of the agents in class i_1 . The price is determined by Bertrand competition among the potential buyers. At period 2, the agent of class i_1 holding the asset has to sell it to an agent in class i_2 according to the same procedure as above. This scheme proceeds until period t when the agent of class i_t holding the asset is paid its realized value.¹⁴

We can easily solve for the unique equilibrium by backward induction. At period t , the willingness to pay for the asset of agent in class i_t , and therefore the (state-contingent) equilibrium price, is exactly $V_{i_t}(f)$. Given Bertrand competition among potential buyers, for an agent in class i_{t-1} , the (state-contingent) value of the asset is then $V_{i_{t-1}} \circ V_{i_t}(f)$. Iterating this backward reasoning up to period 1, the initial (state-contingent) price of the asset is $V_{i_1} \circ V_{i_2} \circ \dots \circ V_{i_{t-1}} \circ V_{i_t}(f) \in \mathbb{R}^\Omega$. Observe that the initial price is a random variable that is measurable with respect to the information of agent i_1 . This highlights the importance of the higher-order expectations in market interactions.¹⁵ ▲

Following Samet [55], we call a sequence $(i_t)_{t \in \mathbb{N}}$ in I an *I-sequence* if and only if for each individual $i \in I$, $i = i_t$ for infinitely many t indexes.

Definition 4. We say that a collection $\{(V_i, \Pi_i)\}_{i \in I}$ of interim expectations exhibits convergence to a deterministic limit if and only if for all *I-sequences* $\iota = (i_t)_{t \in \mathbb{N}}$ and for all $f \in \mathbb{R}^\Omega$,

¹³That is, Π_{sup} is the finest among all partitions that are coarser than each Π_i , and Π_{inf} is the coarsest among all partitions that are finer than each Π_i .

¹⁴This model is a variation of classical models of sequential speculative trading such as Harrison and Kreps [33] and Morris [50], that also allows for non SEU preferences of the traders.

¹⁵Toward pointing out the direct role of higher-order expectations, we assumed that the agents know the class of the potential buyers (and hence their interim expectations). In Section 4, we characterize the equilibrium of the related beauty-contest game where the relevant class of buyers is uncertain.

there exists $k_{f,\iota} \in \mathbb{R}$ such that

$$\lim_{t \rightarrow \infty} V_{i_t} \circ V_{i_{t-1}} \circ \dots \circ V_{i_2} \circ V_{i_1} (f) = k_{f,\iota} 1_\Omega.$$

In this case, for each I -sequence $\iota = (i_t)_{t \in \mathbb{N}} \in I^\mathbb{N}$ define $\bar{V}_\iota : \mathbb{R}^\Omega \rightarrow \mathbb{R}$ by $\bar{V}_\iota (f) = k_{f,\iota}$.

If there is convergence to a deterministic limit, then the sequences of higher-order expectations of the agents converge to a limit whose value is necessarily common knowledge. In this case, each limit functional \bar{V}_ι clearly inherits the properties of the interim expectations, hence it is an ex-ante expectation.

Our first result shows that there is convergence to a deterministic limit, provided that all the interim expectations of the agents have full support and there is no non-trivial public event. Moreover, the rate of convergence is *quasi-exponential*, that is, it is exponential in the number of times that all the agents have been repeated in the sequence. This technical result is the foundation for our analysis.

Theorem 1. *If $\{(V_i, \Pi_i)\}_{i \in I}$ is a collection of full support interim expectations such that $\Pi_{\text{sup}} = \{\Omega\}$, then $\{(V_i, \Pi_i)\}_{i \in I}$ exhibits convergence to a deterministic limit. Moreover, there exist $\varepsilon \in (0, 1)$ and $C \in \mathbb{R}_+$ such that for each I -sequence $(i_m)_{m \in \mathbb{N}}$ and for each $\tau, t \in \mathbb{N}$, if every $i \in I$ appears at least τ times in (i_1, \dots, i_t) , then*

$$\|\bar{V}_\iota (f) 1_\Omega - V_{i_t} \circ \dots \circ V_{i_1} (f)\| \leq C \varepsilon^\tau \|f\|_\infty \quad \forall f \in \mathbb{R}^\Omega.$$

Quasi-exponential convergence provides a bound on the approximation error for computing the limit higher-order expectation of f given ι using the t -th order expectation. In particular, the bound improves in t only if additional expectations of *all* the agents are involved. We next illustrate the meaning of quasi-exponential convergence to a deterministic limit in the asset-pricing example.

Example (Forecasting the forecaster continued). Assume that the assumptions of Theorem 1 are satisfied. Then, rather than looking at a fixed-length sequence, we consider an infinite sequence of classes $(i_t)_{t \in \mathbb{N}}$. We can focus on I -sequences as, if the identity of classes are iid draws with full support on I , then with probability 1 an I -sequence is realized. With this, Theorem 1 guarantees that, for a truncation $(i_1, \dots, i_{\bar{t}})$ of $(i_t)_{t \in \mathbb{N}}$ such that each agent appears sufficiently many times, the dependence of the initial equilibrium price on the realized state of the world is arbitrarily (and exponentially) small. Intuitively, the willingness to pay of an agent in class i_1 does not significantly depend on the state as she knows that the selling value depends on a large number of subsequent transactions. This and the assumption $\Pi_{\text{sup}} = \{\Omega\}$ imply that many of the subsequent buyers will care about the value of the asset also in states that are ruled out by the information of i_1 . ▲

3.1 Common ex-ante expectations

A natural question that emerges in this setting is whether the interim preferences of the agents are consistent with a common ex-ante expectation.

Definition 5. We say that V_\circ (resp. V°) is a lower (resp. upper) common ex-ante expectation for $\{(V_i, \Pi_i)\}_{i \in I}$ if and only if (V_\circ, V_i, Π_i) (resp. (V°, V_i, Π_i)) is a lower (resp. upper) conditional expectation for all $i \in I$. When \bar{V} is both a lower and an upper common ex-ante expectation for $\{(V_i, \Pi_i)\}_{i \in I}$, we say that \bar{V} is a nonlinear common ex-ante expectation for $\{(V_i, \Pi_i)\}_{i \in I}$.

We let \mathcal{V}_\circ and \mathcal{V}° denote respectively the sets of lower and upper common ex-ante expectations for $\{(V_i, \Pi_i)\}_{i \in I}$. Clearly, their intersection is the set of nonlinear common ex-ante expectations for $\{(V_i, \Pi_i)\}_{i \in I}$. It is plain that in the case each $V_i(\omega, \cdot)$ is SEU the nonemptiness of this intersection amounts to the existence of a *common prior*. We next illustrate these new concepts via some examples.

Example (Choquet expected utility continued). In the same setting of Example 1, consider now a set of n agents $I = \{1, \dots, n\}$ and assume that each agent has an information partition Π_i . Let V_\circ be defined as before and, for every i , let the interim expectation (V_i, Π_i) be defined as

$$V_i(\omega, f) = \min_{\mu \in \text{core}_i(\nu)} \mathbb{E}_{p_{\mu,i}(\omega, \cdot)}(f) \quad \forall f \in \mathbb{R}^\Omega$$

where $p_{\mu,i}(\omega, \cdot)$ is the conditional probability of μ given Π_i and $\text{core}_i(\nu)$ is defined as in (6) by replacing Π with Π_i . The results in [28, Axiom C.4 and Theorem 1] imply that V_\circ is a lower common ex-ante expectation for $\{(V_i, \Pi_i)\}_{i \in I}$. \blacktriangle

Example (Maxmin expected utility continued). In the same setting of Example 2, consider now a set of n agents $I = \{1, \dots, n\}$ and assume that each agent has an information partition Π_i which is coarser than Π . For every i , the interim expectation $(V_{i,C}, \Pi_i)$ is defined as in (8) where the collection of sets of probabilities $(C_{\omega,i})_{\omega \in \Omega, i \in I}$ are computed with Π_i in place of Π . In this case, V_C° defined as in (7) is a upper common ex-ante expectation for $\{(V_{i,C}, \Pi_i)\}_{i \in I}$. If in addition, C is composed only of full support probabilities and is rectangular for Π , then V_C° is a nonlinear common ex-ante expectation for $\{(V_{i,C}, \Pi_i)\}_{i \in I}$. Note that in this case a state $\bar{\omega} \in \Omega$ is V_C° -essential if and only if $\mu(\bar{\omega}) > 0$ for all $\mu \in C$. A similar reasoning holds for $V_{i,C}(\omega, \cdot)$ and C_ω . Therefore, V_C° and each $(V_{i,C}, \Pi_i)$ have full support. \blacktriangle

Example 4 (Multiplier expectations and misspecification aversion). Here, we consider multiplier preferences functionals (see Hansen and Sargent [32], axiomatized in Strzalecki [61]). Let $\mu \in \Delta(\Omega)$ have full support and let Π be a partition. Define

$$\bar{V}_{\lambda, \mu}(f) = \min_{\mu' \in \Delta} \{\mathbb{E}_{\mu'}(f) + \lambda R(\mu' || \mu)\} \quad \forall f \in \mathbb{R}^\Omega \quad (10)$$

and

$$V_{\lambda,\mu}(\omega, f) = \min_{p \in \Delta: p(\Pi(\omega))=1} \{ \mathbb{E}_p(f) + \lambda R(p || p_\mu(\omega, \cdot)) \} \quad \forall \omega \in \Omega, \forall f \in \mathbb{R}^\Omega \quad (11)$$

where $\lambda > 0$ and $R(\cdot || \cdot)$ is the relative entropy. Compared to the previous examples the agent has a probability model of reference μ , but she does not fully trust it. She is willing to consider other models μ' , nevertheless the farther they are in terms of relative entropy from μ (resp., its update) the less plausible they are, and the smaller role they play in the minimization (10) (resp., (11)). Here, λ is a parameter that captures the decision maker aversion to the potential misspecification of μ : the lower λ the more the decision maker considers other probability models p . It is well known that $(\bar{V}_{\lambda,\mu}, V_{\lambda,\mu}, \Pi)$ is a nonlinear conditional expectation (see Maccheroni, Marinacci, and Rustichini [45, Section 5.2]). Linear expectations are obtained as the limit case when $\lambda \uparrow \infty$ (see Maccheroni, Marinacci, and Rustichini [44, Proposition 12]). Next, assume that each agent has an information partition Π_i and her conditional interim expectations $(V_{i,\lambda,\mu}, \Pi_i)$ are computed according to (11) with respect to Π_i . In this case, the common prior μ uniquely defines (as in (7)) the nonlinear common ex-ante expectation $\bar{V}_{\lambda,\mu}$.

It is possible to generalize multiplier expectation so to take in account aversion to misspecification as in Cerreia-Vioglio et al. [9]. Formally, rather than a single model, let us fix a set $\mathcal{M} \subseteq \Delta(\Omega)$ of probabilities with full support over Ω and a partition Π . In particular, assume that $\mu|_\Pi = \mu'|_\Pi$ for all $\mu, \mu' \in \mathcal{M}$, that is, there is no model uncertainty with respect to the events that are Π -measurable. Next, define $\bar{V}_{\lambda,\mathcal{M}}(f) = \min_{\mu \in \mathcal{M}} \bar{V}_{\lambda,\mu}(f)$ for all $f \in \mathbb{R}^\Omega$. Similarly as before, assume that each agent has an information partition Π_i and her conditional interim expectation (V_i, Π_i) is $V_{i,\lambda,\mathcal{M}}(\omega, f) = \min_{\mu \in \mathcal{M}} V_{i,\lambda,\mu}(\omega, f)$. For every $i \in I$, if Π_i is coarser than Π , then (\bar{V}, V_i, Π_i) is a nonlinear conditional expectation. The interpretation is that the agents are uncertain about the probabilistic model beyond their aggregate information Π_{\inf} . Moreover, the agents are averse to misspecification both about the model restricted on Π_{\inf} and the set of models assigning likelihoods to events that are finer than Π_{\inf} .¹⁶ \blacktriangle

Clearly, the sets \mathcal{V}_\circ and \mathcal{V}° might contain multiple elements. However, we focus on two selections with a clear interpretation in terms of attitudes toward resolution of uncertainty. Let

$$V_*(f) = \sup_{V_\circ \in \mathcal{V}_\circ} V_\circ(f) \quad \text{and} \quad V^*(f) = \inf_{V^\circ \in \mathcal{V}^\circ} V^\circ(f) \quad \forall f \in \mathbb{R}^\Omega.$$

denote the maximal and minimal elements of \mathcal{V}_\circ and \mathcal{V}° . By construction, the lower (resp. upper) common ex-ante expectation V_* (resp. V^*), has the minimal attraction (resp. aversion) for gradual resolution of uncertainty among the elements in \mathcal{V}_\circ (resp. \mathcal{V}°), provided it exists. We now show that both V_* and V^* are always well defined and provide a characterization of them in terms of the higher-order expectations of the agents.

¹⁶Note that the interim expectations in Example 4 do not satisfy the full support assumption on \mathbb{R}^Ω . However, it satisfies the full support assumption on K^Ω for some closed interval with nonempty interior $K \subseteq \mathbb{R}$. In Remark 1 below, we argue how all our results extend to preferences defined over bounded domains K^Ω , provided that the representation functionals satisfy translation invariance, as in the case of Example 4.

Proposition 1. *Let $\{(V_i, \Pi_i)\}_{i \in I}$ be a collection of interim expectations. Both V_* and V^* are well-defined and respectively a lower and an upper common ex-ante expectation for $\{(V_i, \Pi_i)\}_{i \in I}$. Moreover, if $\{(V_i, \Pi_i)\}_{i \in I}$ exhibits convergence to a deterministic limit, then, for every $f \in \mathbb{R}^\Omega$,*

$$V_*(f) = \inf_{\iota \in I^\mathbb{N}: \iota \text{ is an } I\text{-sequence}} \bar{V}_\iota(f) \quad \text{and} \quad V^*(f) = \sup_{\iota \in I^\mathbb{N}: \iota \text{ is an } I\text{-sequence}} \bar{V}_\iota(f).$$

The interpretation is that by looking at the lowest (resp. highest) limit of the iterated expectations, we exactly identify the minimal attraction (resp. aversion) to gradual resolution of uncertainty needed to jointly rationalize the interim preferences of the agents.¹⁷ In turn, this implies that $V_*(f) \leq V^*(f)$ for all $f \in \mathbb{R}^\Omega$, that is, the ex-ante preferences V_* and V^* are ranked in terms of their uncertainty aversion.

Example (Forecasting the forecaster continued). In the setting of Example 3, fix an I -sequence $\iota = (i_n)_{n \in \mathbb{N}}$ and recall that the equilibrium initial price of asset f , for the game with length t , is equal to the random variable $V_{i_t} \circ V_{i_{t-1}} \circ \dots \circ V_{i_2} \circ V_{i_1}(f)$. In this case, by Theorem 1, as we let t go to infinity, the limit price is deterministic and equal to $\bar{V}_\iota(f)$. Moreover, by Proposition 1, the limit initial price satisfies

$$V_\circ(f) \leq \bar{V}_\iota(f) \leq V^\circ(f) \tag{12}$$

for all upper and lower common ex-ante expectations $V_\circ \in V_\circ$ and $V^\circ \in V^\circ$, and, more accurately, $\bar{V}_\iota(f) \in [V_*(f), V^*(f)]$. For example, equation (12) implies that, if the traders are maximin agents and share the same set of ex-ante probabilistic models $C \subseteq \Delta(\Omega)$, then, under full Bayesian updating, the limit initial price with private information $\bar{V}_\iota(f)$ is smaller than the common ex-ante evaluation $V^\circ(f) = \min_{p \in C} \int f dp$. Indeed, the initial equilibrium price is the result of a compounded pessimistic evaluation due to full Bayesian updating and iterated minimization across all the updated probabilistic models. \blacktriangle

Combining our previous results we get a characterization for the existence of a nonlinear common ex-ante expectation \bar{V} for $\{(V_i, \Pi_i)\}_{i \in I}$: under convergence to a deterministic limit, there exists a common ex-ante expectation if and only if the deterministic limit of all the I -sequences of higher-order expectations is the same. This generalizes the main result of Samet [55] to the class of rational preferences.

Corollary 1. *Let $\{(V_i, \Pi_i)\}_{i \in I}$ be a collection of interim expectations that exhibits convergence to a deterministic limit. The following statements are equivalent:*

- (i) *There exists a nonlinear common ex-ante expectation \bar{V} for $\{(V_i, \Pi_i)\}_{i \in I}$;*

¹⁷In Online Appendix H, we provide an algorithm to compute V_* and V^* starting from the interim preferences of the agents, which are, in principle, observable.

(ii) For each $f \in \mathbb{R}^\Omega$ there exists $k_f \in \mathbb{R}$ such that for each I -sequence $(i_t)_{t \in \mathbb{N}}$

$$\lim_{t \rightarrow \infty} V_{i_t} \circ V_{i_{t-1}} \circ \dots \circ V_{i_2} \circ V_{i_1} (f) = k_f 1_\Omega.$$

(iii) We have $V_* = V^*$.

In this case, for each $f \in \mathbb{R}^\Omega$, we have $V_*(f) = V^*(f) = \bar{V}(f) = k_f$.

As an immediate consequence of Theorem 1 and Corollary 1, we get that our characterization of the nonlinear common ex-ante expectation holds provided that agents' interim preferences have full support and there is no public information. Next we illustrate the (asset-pricing) equilibrium implications of the existence of a nonlinear common ex-ante expectation.

Example (Forecasting the forecaster continued). Assume that the agents have a nonlinear common ex-ante expectation $\bar{V} : \mathbb{R}^\Omega \rightarrow \mathbb{R}$. For a sufficiently long truncation $(i_1, \dots, i_{\bar{t}})$ of $(i_t)_{t \in \mathbb{N}}$, the initial equilibrium price is approximately state-independent and equal to the common ex-ante evaluation $\bar{V}(f)$ of the asset. In words, under a nonlinear common ex-ante expectation, the particular order of trades does not affect the initial price. Conversely, for any two arbitrary I -sequences truncated at $\bar{t} \in \mathbb{N}$, we can falsify the existence of a nonlinear common ex-ante expectation by checking whether the corresponding equilibrium prices are sufficiently different.¹⁸

▲

On the one hand, Corollary 1 provides sufficient condition for the existence of a nonlinear common ex-ante expectation, as well as a way to compute it. On the other hand, mutual dynamic consistency with respect to all the information structures of the agents is very restrictive under ambiguity averse preferences, as pointed out by Ellis [14] and Gumen and Savochkin [29]. Therefore, our corollary implies that the order of sequential traders in Example 3 is generally relevant for the equilibrium price, and more so with ambiguity aversion. This creates scope for the manipulation of the frequency with which agents with a given information structure can trade to affect the equilibrium price. In other words, without SEU, for example under ambiguity averse preferences, there is scope for optimal design of the sequential trading protocol.

Finally, Proposition 1 and Corollary 1 provide a new characterization of the common prior assumption in the setting of Samet [55]. In particular, there exists a common prior if and only if both the extreme ex-ante preferences of the agents are neutral with respect to the timing of resolution of uncertainty.

¹⁸Formally, consider two I -sequences with truncations $(i_1, \dots, i_{\bar{t}})$ and $(\tilde{i}_1, \dots, \tilde{i}_{\bar{t}})$ in which each $i \in I$ appears at least τ times. By inspection of the proof of Theorem 1, we have explicit expressions for the constant $C \in \mathbb{R}_+$ and $\varepsilon \in (0, 1)$ of Definition 3. With this, we can say that a common rational preference does not exist if

$$\|V_{i_{\bar{t}}} \circ \dots \circ V_{i_1} (f) - V_{\tilde{i}_{\bar{t}}} \circ \dots \circ V_{\tilde{i}_1} (f)\| > 2C\varepsilon^\tau.$$

Corollary 2. Let $\{(V_i, \Pi_i)\}_{i \in I}$ be a collection of full support interim expectations such that $\Pi_{\text{sup}} = \{\Omega\}$ and such that V_i is SEU for all $i \in I$. There exists a common prior $p \in \Delta(\Omega)$ for $\{(V_i, \Pi_i)\}_{i \in I}$ if and only both V_* and V^* are SEU.

3.2 Dynamic consistency of local subjective beliefs

In this section, we consider an intermediate notion of mutual dynamic consistency that only involves the most trusted probabilistic models. We show below that it is strictly linked to the concept of *subjective beliefs at an act* introduced by Rigotti et al. [54, Definition 1 and Proposition 3] to study Pareto optimal allocations under ambiguity. To formalize this concept we need to restrict ourselves to the class of variational preferences (cf. [44]).

A collection of interim expectations $\{(V_i, \Pi_i)\}_{i \in I}$ is *variational* if and only if for every $i \in I$ and $\omega \in \Omega$, there exists a lower semicontinuous, grounded, and convex cost function $c_{i,\omega} : \Delta(\Omega) \rightarrow [0, \infty]$ such that

$$V_i(\omega, f) = \min_{p \in \Delta(\Omega)} \{\mathbb{E}_p(f) + c_{i,\omega}(p)\} \quad (13)$$

for all $f \in \mathbb{R}^\Omega$.¹⁹ Variational interim expectations exhibit violations of subjective expected utility due to aversion to ambiguity, a widely documented trait. The interpretation is that each agent considers the evaluation of the act under many probabilistic models and $c_{i,\omega}$ penalizes more the models (subjectively) deemed less plausible. In particular, the probabilistic models p for which $c_{i,\omega}(p) = 0$ represent the ones that i trusts the most in state ω . All the examples of preferences we have introduced are variational.²⁰

Define the following set which captures a minimal extent of mutual dynamic consistency among the agents:

$$\Theta = \bigcap_{i \in I} \text{co} \{p \in \Delta(\Omega) : \exists \omega \in \Omega, c_{i,\omega}(p) = 0\}.$$

In words, Θ contains all the ex-ante probabilistic models that, when updated, are among the most trusted by every agent in every state, that is, those that minimize the interim cost function. Following Ghirardato and Marinacci [22], we call these probability measures *benchmark models*.²¹ Incidentally, Θ also coincides with the set of ex-ante probabilistic models that are consistent with a selection from the subjective beliefs of the interim preferences of the agents at any constant act (cf. [54, Definition 1 and Proposition 3], [39], [46]).

Definition 6. We say that a variational collection of interim expectations $\{(V_i, \Pi_i)\}_{i \in I}$ has a

¹⁹A cost function c is grounded if and only if $\min_{p \in \Delta(\Omega)} c(p) = 0$.

²⁰Imposing the representation in equation (13) is equivalent to assume that each $V_i(\omega, \cdot)$ is concave and translation invariant, that is, $V_i(\omega, f + ke) = V_i(\omega, f) + k$ for all $f \in \mathbb{R}^\Omega$ and $k \in \mathbb{R}$.

²¹These probability measures correspond to SEU preferences that are less ambiguity averse than the interim preference of the agent as formally showed in [44].

common local subjective belief if and only if $\Theta \neq \emptyset$. In this case, we define $V^\Theta : \mathbb{R}^\Omega \rightarrow \mathbb{R}$ as

$$V^\Theta(f) = \min_{\mu \in \Theta} \mathbb{E}_\mu(f).$$

In words, V^Θ is a caution evaluation of acts that only relies on the benchmark ex-ante probabilistic models. In the next result, we relate these intermediate notions of common preferences with the ones we have already studied.

Proposition 2. *Let $\{(V_i, \Pi_i)\}_{i \in I}$ be a variational collection of interim expectations that exhibits convergence to a deterministic limit. The following facts are true:*

1. *If $\{(V_i, \Pi_i)\}_{i \in I}$ has a common local subjective belief, then V^Θ is a upper common ex-ante expectation for $\{(V_i, \Pi_i)\}_{i \in I}$, hence $V^\Theta \geq V^*$.*
2. *If there exists a nonlinear common ex-ante expectation \bar{V} for $\{(V_i, \Pi_i)\}_{i \in I}$, then $\{(V_i, \Pi_i)\}_{i \in I}$ has a common local subjective belief, hence $V^\Theta \geq \bar{V}$.*

Not surprisingly, the new notion of ex-ante expectation introduced V^Θ is less ambiguity averse than the previous ones. The reason is that each $\mu \in \Theta$ is obtained by mixing interim SEU preferences that are less ambiguity averse than the agents' ones.

We conclude this section with a remark discussing the relevance of the unbounded-domain assumption that we have implicitly made for the results presented so far.

Remark 1. For expositional purposes, we assumed that the agents' expectations are defined over the entire space of available acts \mathbb{R}^Ω . The cost of such choice is that certain decision models satisfy a property just over bounded domains and not over unbounded ones. For example, the multiplier preferences model satisfies the full support assumption (a key property for us) over bounded domains, but not over \mathbb{R}^Ω (see Example 4). That said, this limitation can be easily circumvented for a very large class of decision models and, in particular, those in our examples. Indeed, we can generalize all of our results to the class of *translation invariant* interim expectations. In fact, consider a translation invariant interim expectation (V_i, Π_i) where V_i is merely defined over the space $\Omega \times K^\Omega$ and $K \subseteq \mathbb{R}$ is a closed interval with nonempty interior. In this case, it is not hard to show that V_i admits an extension $\tilde{V}_i : \Omega \times \mathbb{R}^\Omega \rightarrow \mathbb{R}$ such that (\tilde{V}_i, Π_i) is a translation invariant interim expectation.²² Moreover, we have that \tilde{V}_i inherits the properties of V_i needed for our results. For example, \tilde{V}_i has full support (resp., is concave)

²²First, observe that $V_i(\omega, \cdot) : K^\Omega \rightarrow \mathbb{R}$, being monotone and translation invariant, is Clarke differentiable on $\text{int } K^\Omega$ for all $\omega \in \Omega$. We denote by $\text{Rg}(\partial_C V_i(\omega, \cdot))$ the range of Clarke's differential for all $\omega \in \Omega$. The extension considered is then defined by

$$\tilde{V}_i(\omega, f) = \sup_{g \in \text{int } K^\Omega} \left\{ V_i(\omega, g) + \inf_{\mu \in \text{Rg}(\partial_C V_i(\omega, \cdot))} \int (f - g) d\mu \right\} \quad \forall \omega \in \Omega, \forall f \in \mathbb{R}^\Omega.$$

The formal arguments are available upon request.

if and only if V_i shares the same property on the restricted domain. Thus, as mentioned, all our formal results extend to limited domains. This is because either the proofs never rely on unboundedness or we can resort to the extension above. \blacktriangle

4 Equilibrium and (un-)common ex-ante preferences

In this section, we consider the equilibrium implications of our previous analysis for a class of coordination games. In each of the following applications, the equilibrium $\sigma^\beta = \left(\sigma_i^\beta\right)_{i \in I} \in (\mathbb{R}^\Omega)^n$ is described by the following fixed-point condition:

$$\sigma_i^\beta(\omega) = V_i \left(\omega, (1 - \beta) \hat{f} + \beta \sum_{j \in I} w_{ij} \sigma_j^\beta \right) \quad \forall \omega \in \Omega, \forall i \in I. \quad (14)$$

Here, $\hat{f} \in \mathbb{R}^\Omega$ is a payoff-relevant fundamental, $\beta \in (0, 1)$ parametrizes the relative importance of *coordination* with other agents over *adaptation* to the fundamental, and $W = \{w_{ij}\}_{i,j \in I} \in \mathbb{R}^{n \times n}$ is a stochastic matrix where each w_{ij} captures the relative importance of agent j for i .²³

The interpretation is that the equilibrium outcome for agent i coincides with her (generalized) expectation of a combination of the fundamental and the equilibrium outcomes of the other players. These kind of fixed-point conditions are ubiquitous in models of asset pricing with beauty-contests (cf. Morris and Shin [51]), networks of financial institutions (cf. Jackson and Pernoud [41]), and price competition (cf. Angeletos and Pavan [3]) as we show below. In the SEU case, the high-coordination limit ($\beta \rightarrow 1$) of the equilibrium strategies is used to select an equilibrium of the pure-coordination games and can be related to the common prior expectation of the asset (cf. Shin and Williamson [59] and Golub and Morris [27]). Analogously, the characterization of this limit and its relation to the ex-ante preferences we have defined will be the main focus of our analysis.

4.1 Beauty contests: coordination and equilibrium

As a leading application, we consider a beauty-contest model with random matching and private information (as in [27]) that generalizes the leading example of Section 3. Each $i \in I$ represents a continuum of agents sharing the same information partition Π_i . Time is discrete and there is a random variable $\hat{f} \in \mathbb{R}^\Omega$ denoting the only asset in this economy which is sequentially traded with random matching. Let $\beta \in (0, 1)$. At every period $t \in \mathbb{N}$, if an agent in class i holds the asset, with probability $(1 - \beta)$ she has to liquidate the asset and obtain its fundamental (uncertain) value \hat{f} . With complementary probability β , she privately has to sell the asset to an agent from a randomly selected class and then leaves the game. The matching probabilities, conditional on not liquidating the asset, are described by a stochastic matrix W , where w_{ij}

²³A matrix $W \in \mathbb{R}^{n \times n}$ is stochastic if and only if $w_{ij} \geq 0$ for all $i, j \in I$ and $\sum_{j \in I} w_{ij} = 1$ for all $i \in I$.

is the probability with which an agent in class i is matched to class j . In particular, the random matching is independent of the state $\omega \in \Omega$ and plays the role of objective lotteries a la Anscombe and Aumann in our setting.²⁴ After the realization of the matched class j , the agents in j compete a la Bertrand offering a price to the asset holder in i who decides to whom to sell the asset. This mechanism implies that in equilibrium the offered price is equal to the (common) willingness to pay for the asset of the agents in class j . If an agent in class j acquires the asset, then the game continues to period $t + 1$.²⁵ We study the equilibria of this game under the following maintained assumption:

Assumption 1 The collection of interim expectations $\{(V_i, \Pi_i)\}_{i \in I}$ has full support, is such that $\Pi_{\text{sup}} = \{\Omega\}$, and is variational (see equation (13)) and W is strongly connected.

A (Markov) *strategy* for an agent in class $i \in I$ is a random variable $\sigma_i \in \mathbb{R}^\Omega$ that is measurable with respect to the information structure Π_i . In particular, from the point of view of agents in i , the strategies $\sigma_j \in \mathbb{R}^\Omega$ of agents in any class j are state-dependent offers that can be evaluated through their interim preferences V_i as standard acts. Let Σ_i and Σ denote respectively the set of strategies for agents in class i and the set of profiles of strategies for the n classes of agents. For every $\beta \in (0, 1]$, if we fix a profile of strategies $\sigma = (\sigma_j)_{j \in I} \in \Sigma$, then the corresponding (state-dependent) willingness to pay for asset \hat{f} of any agent in class $i \in I$ is:

$$S_{\beta,i}(\sigma) = V_i \left((1 - \beta) \hat{f} + \beta \sum_{j \in I} w_{ij} \sigma_j \right) \quad \forall \omega \in \Omega. \quad (15)$$

The *equilibria* of this game correspond to the fixed points of the map $S_\beta(\cdot) : \Sigma \rightarrow \Sigma$, that is, $\sigma^\beta \in \Sigma$ is a equilibrium if and only if it satisfies equation (14).

Proposition 3. *For every $\beta \in (0, 1)$, there exists a unique equilibrium $\sigma^\beta \in \Sigma$ of the game. Moreover, there exists $C \in \mathbb{R}_+$ such that, for every $\beta \in (0, 1)$,*

$$\max_{i,j \in I, \omega, \omega' \in \Omega} \left| \sigma_i^\beta(\omega) - \sigma_j^\beta(\omega') \right| \leq (1 - \beta) C \max_{\omega, \omega' \in \Omega} \left| \hat{f}(\omega) - \hat{f}(\omega') \right|. \quad (16)$$

The inequality in equation (16) gives a bound on the maximum level of disagreement among the equilibrium asset evaluations. First, we observe that the right hand side is monotonically decreasing in β and *linearly* vanishes as we let coordination become more important, that is $\beta \rightarrow 1$. This implies that the price of the asset becomes constant across states and agents in the limit. Second, the speed of this convergence is disciplined by C which can be linked to the preferences, information, and network primitives, as we explain in the next remark.

²⁴In other words, the matching probabilities are used to take convex linear combinations of acts in \mathbb{R}^Ω .

²⁵Observe that there is no relevant learning over time since the past owners of the asset have left the game. Moreover, conditional on non liquidation, even if the asset holder would learn something about the state $\omega \in \Omega$ from the offers of the agents in j , accepting the highest offer is still a dominant strategy given the absence of outside options.

Remark 2. In a companion paper [7], we further elaborate on the estimate on the range of the fixed points of equations like (14) and find an explicit expression for the estimate in Proposition 3 in terms of the properties of S_β . In the current setting, this translates in the following way. Define the adjacency matrix $A \in \{0, 1\}^{(I \times \Omega) \times (I \times \Omega)}$ over $(I \times \Omega)$ by letting, for all $i, j \in I$ and $\omega', \omega \in \Omega$, $a_{(i, \omega')(j, \omega)} = 1$ if and only if $w_{ij} > 0$ and $\omega \in \Pi_i(\omega')$. Also, for all $i \in I$, $\omega' \in \Omega$, and $\omega \in \Pi_i(\omega')$, let $\varepsilon_{i, \omega, \omega'} > 0$ denote the ε satisfying the full-support equation (9) for agent i at state ω' with respect the essential state ω . Next, let

$$\underline{\varepsilon} = \min_{i, j \in I, \omega, \omega' \in \Omega: a_{(i, \omega')(j, \omega)} = 1} \varepsilon_{i, \omega, \omega'} w_{ij}$$

and with this define the bound

$$C = \sum_{\tau=0}^{d-1} \left(1 + \frac{1}{\underline{\varepsilon}}\right)^{2d-\tau}$$

where d is the diameter of the graph corresponding to A . The number of connections in A depends on both the number of connections in the network among agents W as well as on the dependence of their information structures. In turn, increasing the number of connections in A has two contrasting effects: first it reduces the diameter of the graph, making C smaller, second it reduces the maximal possible magnitude of $\underline{\varepsilon}$, making C larger. For example, the diameter is low when all the agents are connected and the information structure has a product form, i.e., whenever $\Pi_i(\omega) \cap \Pi_j(\omega') \neq \emptyset$ for all $i, j \in I$ and $\omega', \omega \in \Omega$, and is high under a circular information structure, i.e., whenever for every i and $\omega \in \Omega$, $\Pi_i(\omega)$ has nonempty intersection only with two partition cells of the coplayers. \blacktriangle

4.2 Beauty contests: coordination and misspecification neutrality

In this section, we characterize the unique equilibrium σ^β as coordination becomes more and more important, i.e., $\beta \rightarrow 1$. Define the set of interim benchmark beliefs

$$Q = \left\{ q \in \Delta(\Omega)^{I \times \Omega} : \forall (i, \omega) \in I \times \Omega, \forall \omega' \in \Pi_i(\omega), q_{i, \omega'} = q_{i, \omega}, c_{i, \omega}(q_{i, \omega}) = 0 \right\}.$$

Each $q \in Q$ is a collection of interim beliefs for all the agents and states that are (i) measurable with respect to the information of the corresponding agents and (ii) most trusted in the given state. This can be combined with the network structure W to obtain an *interaction structure* $W^q \in \mathbb{R}_+^{(I \times \Omega) \times (I \times \Omega)}$ among agent-state pairs capturing both the interim beliefs of the agents as well as the strength of their links. Formally, we let

$$w_{(i, \omega)(j, \omega')}^q = w_{ij} q_{i, \omega}(\omega') \quad \forall i, j \in I, \forall \omega, \omega' \in \Omega. \quad (17)$$

Under SEU interim preferences, there is a unique interaction structure (introduced by Golub and Morris [27]) pinned down by the network W and the posterior beliefs of the agents. In the present setting, model uncertainty translates into a multiplicity of interim relevant beliefs,

hence into a multiplicity of interaction structures. However, this multiplicity is disciplined by both the information and the interim preferences of the agents.

Lemma 1. *For each $q \in Q$, there exists a unique $\gamma^q \in \Delta(I \times \Omega)$ such that $\gamma^q = \gamma^q W^q$.*

This is a consequence of the connectedness properties of each W^q implied by $\Pi_{\inf} = \{\Omega\}$, full support of $\{V_i, \Pi_i\}_{i \in I}$, and that W is strongly connected. We are now ready to state the main result of this section.

Theorem 2. *For all $i \in I$ and $\omega \in \Omega$,*

$$V_* \left(\hat{f} \right) \leq \lim_{\beta \rightarrow 1} \sigma_i^\beta(\omega) = \min_{q \in Q} \sum_{(i, \omega) \in I \times \Omega} \gamma_{i, \omega}^q \mathbb{E}_{q_{i, \omega}} \left(\hat{f} \right) \leq \inf_{\mu \in \Theta} \mathbb{E}_\mu \left(\hat{f} \right). \quad (18)$$

Moreover, if there exists a nonlinear common ex-ante expectation \bar{V} for $\{(V_i, \Pi_i)\}_{i \in I}$, then, for all $i \in I$ and $\omega \in \Omega$,

$$\lim_{\beta \rightarrow 1} \sigma_i^\beta(\omega) \in \left[\bar{V} \left(\hat{f} \right), V^\Theta \left(\hat{f} \right) \right].$$

First, we observe that, in the limit where the coordination motive prevails, the equilibrium price is independent of the realized state and the agent's identity. In particular, the limit selects an equilibrium of the pure coordination game where the asset is payoff irrelevant. This generalizes a well-known fact under subjective expected utility (cf. [27] and [59]).

Second, the constant limit price equals the most cautious average of the benchmark evaluations of \hat{f} that are consistent with the network structure. Notably, the cautious selection of the benchmark models q from Q induced by the market interaction has two roles. While selecting beliefs that evaluate the asset in a cautious way (i.e., to keep the first-order evaluations $\mathbb{E}_{q_{i, \omega}} \left(\hat{f} \right)$ low), it also determines how the heterogeneous evaluations are aggregated through the eigenvector centrality γ^q of the interaction structure.

Third, our formula points out that the strong coordination motives in the market attenuate the ambiguity concern exhibited by the equilibrium evaluation. Intuitively, the asymmetric information of the traders combined with their coordination motive imply that the equilibrium prices are less variable across states than the fundamental itself. Therefore, the uncertainty averse traders evaluate the asset more favorably than the fundamental. More formally, we have

$$\lim_{\beta \rightarrow 1} \sigma_i^\beta(\omega) \geq V_i \left(\omega, \hat{f} \right) \quad \forall i \in I, \forall \omega \in \Omega,$$

since each collection of beliefs $q \in Q$ satisfy $c_{i, \omega}(q_{i, \omega}) = 0$ for all $i \in I$ and $\omega \in \Omega$. In turn, this immediately yields the lower bound in equation (18) and, when there exists a common ex-ante evaluation, we actually have $V_* \left(\hat{f} \right) = \bar{V} \left(\hat{f} \right)$, implying that the equilibrium price is higher than the *shared* ex-ante evaluation. This is a sharp difference with respect to the case of SEU interim preferences where, under a common prior, the limit equilibrium price coincides with the prior expectation.

Fourth, the equilibrium price cannot be higher than the evaluation of the fundamental under any *ex-ante* probabilistic model that all the agents consistently trust in all the states. Importantly, while the specific value of the limit equilibrium price depends on the network structure, the two bounds we have just described are robust as they hold across all the strongly connected network structures. Moreover, as we next show, the upper bound is attained in several important cases.

Corollary 3. *Assume that, for all $i \in I$ and $\omega \in \Omega$, it holds $\arg \min_{p \in \Delta(\Omega)} c_{i,\omega}(p) = \{q_{i,\omega}^*\}$. For all $i \in I$ and $\omega \in \Omega$,*

$$\lim_{\beta \rightarrow 1} \sigma_i^\beta(\omega) = \sum_{(i,\omega) \in I \times \Omega} \gamma_{i,\omega}^{q^*} \mathbb{E}_{q_{i,\omega}^*}(\hat{f}).$$

Moreover, if $\{(V_i, \Pi_i)\}_{i \in I}$ has a common local subjective belief, then $\Theta = \{\mu^\}$ and*

$$\lim_{\beta \rightarrow 1} \sigma_i^\beta(\omega) = \mathbb{E}_{\mu^*}(\hat{f}).$$

This result characterizes an extreme form of ambiguity-aversion reduction. Indeed, whenever each interim preference has a unique benchmark model (e.g., all the agents have divergence preferences in the interim), the limit equilibrium price is equal to an ex-ante SEU evaluation of the asset, implying that only the interim benchmark models matter as the importance of coordination grows. This reduction is particularly stark when the agents share a common local subjective belief μ^* . In this case, the ex-ante evaluation of the asset according to this probabilistic model is the limit price equilibrium and this limit is the same regardless of the ambiguity attitudes and the network structure. Therefore, whenever μ^* is highly misspecified with respect to the “objective” probability model ν^* , then there is a divergence between the limit market price $\lim_{\beta \rightarrow 1} \sigma_i^\beta(\omega)$ and the rational-expectations value $\mathbb{E}_{\nu^*}(\hat{f})$.

In the next example, we illustrate this phenomenon within the class of multiplier preferences with Bayesian updating from a common prior.

Example 5. Suppose that, in the ex-ante stage, the agents share the same unique benchmark model $\mu^* \in \Delta(\Omega)$, but they are averse to misspecification with possibly different attitudes: each $i \in I$ evaluates \hat{f} as

$$\min_{p \in \Delta} \left\{ \mathbb{E}_p(\hat{f}) + \lambda_i R(p || \mu^*) \right\}$$

where $(\lambda_i)_{i \in I} \in \mathbb{R}_{++}^n$ is a profile of misspecification fear indexes. After having observed their own private information, the agents update the benchmark model to $p_{\mu^*,i}(\omega, \cdot)$. Therefore, the interim evaluation of i at ω is

$$V_i(\omega, f) = \min_{p \in \Delta} \left\{ \mathbb{E}_p(f) + \lambda_i R(p || p_{\mu^*,i}(\omega, \cdot)) \right\} \quad \forall f \in \mathbb{R}^\Omega.$$

In this case, Corollary 3 implies that

$$\lim_{\beta \rightarrow 1} \sigma_i^\beta(\omega) = \mathbb{E}_{\mu^*}(\hat{f}) \quad \forall i \in I, \forall \omega \in \Omega.$$

That is, the ambiguity is completely washed out in the limit and the price converges to the expected evaluation of the asset, independently of the attitudes towards misspecification. If these attitudes are homogeneous, i.e., $\lambda_i = \lambda$ for all $i \in I$, then there exists a common ex-ante expectation

$$\bar{V}(f) = \min_{p \in \Delta} \{\mathbb{E}_p(f) + \lambda R(p || \mu^*)\} \quad \forall f \in \mathbb{R}^\Omega$$

and a wedge between $\bar{V}(\hat{f})$ and $\lim_{\beta \rightarrow 1} \sigma_i^\beta(\omega)$ arises whenever the asset pays a different amount in each state. More generally, this wedge remains present between V_* and $\lim_{\beta \rightarrow 1} \sigma_i^\beta(\omega)$ even when the misspecification attitudes are heterogeneous. \blacktriangle

The simple Example 8 in Online Appendix I illustrates how the ambiguity-attenuating effect of the market interaction becomes relevant already at intermediate levels of coordination, i.e., for β far from 1.

Even beyond the case of interim preferences with single benchmark models, Theorem 2 has important implications for games with incomplete information with existing updating rules. For example, by equation (18), if all the traders share the same set $C \subseteq \Delta(\Omega)$ of ex-ante probability models, are maxmin, and update with full Bayesian updating, then the equilibrium price is lower than $\min_{p \in C} \mathbb{E}_p(\hat{f})$, the common ex-ante willingness to pay, hence the price of the asset without information. Therefore, the limit equilibrium price is affected by the ex-ante ambiguity of the agents even if some of them have unambiguous beliefs at the interim stage. In the next example, we illustrate this effect and we interpret it as *contagion of interim ambiguity*.

Example 6 (Contagion of interim ambiguity). Consider two traders $I = \{1, 2\}$ that are uncertain about an asset $\hat{f} \in \mathbb{R}^\Omega$ with $\Omega = \{l, m, h\}$ and $\hat{f}(l) < \hat{f}(m) < \hat{f}(h)$. The agents are endowed with the following information structures $\Pi_1 = \{\{l\}, \{m, h\}\}$ and $\Pi_2 = \{\{l, m\}, \{h\}\}$. Fix $\psi \in (0, 1)$ and $\varepsilon \in (0, 1/2)$, and assume that the agents have a common set of ex-ante probabilistic models

$$C = \{\alpha \delta_l + (1 - \alpha)(\psi \delta_m + (1 - \psi) \delta_h) : \alpha \in [\varepsilon, 1 - \varepsilon]\}.$$

In the interim stage, conditional on each ω , each agent i has maxmin preferences with respect to $C_{i,\omega}$ obtained via full Bayesian updating. In particular, we have

$$C_{1,l} = \{\delta_l\} \quad \text{and} \quad C_{1,m} = C_{1,h} = \{\psi \delta_m + (1 - \psi) \delta_h\},$$

and

$$C_{2,l} = C_{2,m} = \left\{ \alpha \delta_l + (1 - \alpha) \delta_m : \alpha \in \left[\frac{\varepsilon}{\varepsilon + \psi(1 - \varepsilon)}, \frac{(1 - \varepsilon)}{(1 - \varepsilon) + \psi \varepsilon} \right] \right\} \quad \text{and} \quad C_{2,h} = \{\delta_h\}.$$

In the interim stage, only agent 2 conditional on $\omega \in \{l, m\}$ perceives ambiguity. For every β , it is easy to guess and verify that the equilibrium strategy satisfies $\sigma_1^\beta(l) \leq \sigma_1^\beta(m) = \sigma_1^\beta(h)$.

Therefore, conditional on $\omega \in \{l, m\}$, agent 2 behaves as if her probabilistic model assigns the highest possible probability to l , that is, $\alpha = \frac{(1-\varepsilon)}{(1-\varepsilon)+\psi\varepsilon}$. With some tedious algebra, this observation allows us to compute the equilibrium in closed form and obtain the limit equilibrium

$$\lim_{\beta \rightarrow 1} \sigma_i^\beta(\omega) = \min_{p \in C} \mathbb{E}_p(\hat{f}) = (1 - \varepsilon) \hat{f}(l) + \varepsilon \psi \hat{f}(m) + \varepsilon (1 - \psi) \hat{f}(h).$$

In words, the ambiguity aversion of agent 2 conditional on $\omega \in \{l, m\}$ is strong enough to infect both types of agent 1 as well as her type when she observes h . This effect leads to full coordination on the ex-ante ambiguity averse evaluation. This is particularly sharp as we increase the ex-ante ambiguity of the players by letting $\varepsilon \rightarrow 0$. In this case, in the high-coordination limit, the unique price will converge to the lowest evaluation possible $\hat{f}(l)$ at every state. \blacktriangle

The previous two examples may suggest that, whenever $\lim_{\beta \rightarrow 1} \sigma_i^\beta(\omega)$ is well defined and Θ is nonempty, we have $\lim_{\beta \rightarrow 1} \sigma_i^\beta(\omega) = \min_{\mu \in \Theta} \mathbb{E}_\mu(\hat{f}) = V^\Theta$, that is, the upper bound in Theorem 2 is always achieved even beyond the scope of Corollary 3. However, the next simple example shows that this is not always the case under these assumptions.

Example 7. Let $I = \{1, 2\}$, $\hat{f} \in \mathbb{R}^\Omega$, and endow the two traders with no information, that is, $\Pi_1 = \Pi_2 = \{\Omega\}$. In the ex-ante stage, both the agents have maxmin preferences with corresponding sets of probabilistic models $C_1, C_2 \subseteq \Delta(\Omega)$ such that $C_1 \neq C_2$ and $C_1 \cap C_2 \neq \emptyset$. In this case, we have $\Theta = C_1 \cap C_2$ given that both agents have no information. Moreover, for every $\beta \in (0, 1)$, the unique equilibrium σ^β is given by

$$\sigma_i^\beta(\omega) = \frac{\min_{p \in C_i} \left\{ \mathbb{E}_p(\hat{f}) \right\} + \beta \min_{p \in C_{-i}} \left\{ \mathbb{E}_p(\hat{f}) \right\}}{1 + \beta} \quad \forall i \in I, \forall \omega \in \Omega.$$

With this, the high-coordination limit price is given by

$$\lim_{\beta \rightarrow 1} \sigma_i^\beta(\omega) = \frac{\min_{p \in C_1} \left\{ \mathbb{E}_p(\hat{f}) \right\} + \min_{p \in C_2} \left\{ \mathbb{E}_p(\hat{f}) \right\}}{2} \leq \min_{p \in C_1 \cap C_2} \mathbb{E}_p(\hat{f}),$$

and, in general, the previous inequality may be strict.²⁶ \blacktriangle

The previous example with maxmin preferences and full Bayesian updating crucially relies on the nonexistence of a common ex-ante expectation \bar{V} . Indeed, the next corollary of Theorem 2 shows that, in this setting, if \bar{V} exists, then the lower and upper bound collapses and are equal to the limit price, regardless of the network structure.

²⁶To see this concretely, let $\Omega = \{L, H\}$, $C_1 = \{p \in \Delta(\Omega) : p(H) \in [1/4, 1/2]\}$, $C_2 = \{p \in \Delta(\Omega) : p(H) \in [1/3, 1/2]\}$, and $\hat{f}(L) = 1 - \hat{f}(H) = 0$.

Corollary 4. *Let $\{(V_i, \Pi_i)\}_{i \in I}$ be a collection of maxmin (cf. Example 2) interim expectations. If there exists a nonlinear common ex-ante expectation \bar{V} for $\{(V_i, \Pi_i)\}_{i \in I}$, then \bar{V} is a maxmin ex-ante expectation and, for all $i \in I$ and $\omega \in \Omega$,*

$$\bar{V}(\hat{f}) = \lim_{\beta \rightarrow 1} \sigma_i^\beta(\omega) = V^\Theta(\hat{f}).$$

In stark contrast with Corollary 3 and Example 5, under maxmin preferences the perception of and aversion to ambiguity is still present in the high-coordination limit. Moreover, the higher the ex-ante ambiguity about the underlying fundamental (i.e., the lower \bar{V}), the lower the equilibrium price. However, all the ex-ante ambiguity is preserved in the limit since full Bayesian updating is an overly cautious updating rule under maxmin preferences.²⁷ We next illustrate our results in the less cautious proxy updating of [28] introduced in Example 1. In this case, recall that the *lower common ex-ante expectation* $V_o = \min_{\mu \in \text{core}(\nu)} \mathbb{E}_\mu$ describes the ex-ante preferences of the agents. Moreover, by Theorem 2, we have that, for every network structure, the equilibrium price in the high-coordination limit belongs to

$$\left[\min_{\mu \in \text{core}(\nu)} \mathbb{E}_\mu(\hat{f}), \min_{\mu \in \cap_{i \in I} \text{core}_i(\nu)} \mathbb{E}_\mu(\hat{f}) \right]$$

as $\cap_{i \in I} \text{core}_i(\nu)$ is included in Θ and this intersection is always nonempty since it contains the Shapley value μ_ν . Importantly, if the probabilities in $\text{core}(\nu)$ agree on the events that are Π_{inf} -measurable, so that the information structures are unambiguous, then the two bounds collapse since $\text{core}_i(\nu) = \text{core}(\nu)$ for all $i \in I$. In the next section, we generalize this result to the whole class of variational models.

4.3 Beauty contests: unambiguous information structure

Here, we consider an important particular case: the agents are unambiguous with respect to the information structure while still possibly perceiving ambiguity about the fundamental \hat{f} , i.e., there is no *strategic ambiguity*. In this case, the first-order expectations of the agents exhibit perceived ambiguity and ambiguity aversion, whereas the higher-order expectations do not, that is, they are SEU. Formally, we say that the *information structure is unambiguous* if and only if for every $i \in I$, V_i is Π_{inf} -affine, that is

$$V_i(\omega, (1 - \alpha)h + \alpha g) = (1 - \alpha)V_i(\omega, h) + \alpha V_i(\omega, g)$$

for all $\alpha \in (0, 1)$, for all $\omega \in \Omega$, and for all $g, h \in \mathbb{R}^\Omega$ where g is Π_{inf} -measurable. This implies that V_i is linear over the vector space of elements $g \in \mathbb{R}^\Omega$ that are Π_{inf} -measurable. This restriction is reasonable, for instance, in games where the agents repeatedly interact and can observe the actions of the coplayers after each interaction. In this case, if the agents are correctly specified, then their beliefs will converge to the true distribution on Π_{inf} .

²⁷The existence of a common ex-ante expectation implies that each V_i is obtained via full Bayesian updating from \bar{V} (cf. [16]).

Proposition 4. *For all $i \in I$ and $\omega \in \Omega$,*

$$\lim_{\beta \rightarrow 1} \sigma_i^\beta(\omega) \in \left[V_* \left(\hat{f} \right), V^* \left(\hat{f} \right) \right].$$

Moreover, if there exists a nonlinear common ex-ante expectation \bar{V} for $\{(V_i, \Pi_i)\}_{i \in I}$, then, for all $i \in I$ and $\omega \in \Omega$,

$$\lim_{\beta \rightarrow 1} \sigma_i^\beta(\omega) = \bar{V} \left(\hat{f} \right).$$

Whenever the traders are not ambiguous regarding events in their information structures, the extreme ex-ante preferences give both an upper and lower bound for any possible equilibrium selection. In particular, observe that the upper bound given by the previous proposition improves on the one of Theorem 2 since V^Θ is a common upper ex-ante preference. Next, observe that, whenever a nonlinear common ex-ante expectation exists, the identity $\bar{V} = V_* = V^*$ implies that the limit equilibrium $\lim_{\beta \rightarrow 1} \sigma^\beta$ is well defined and equal to the ex-ante evaluation. This is an implication of the common prior assumption under SEU (cf. [27]) that we extend to the unambiguous-information case. Finally, comparing the second parts of Theorem 2 and of Proposition 4, we observe that the only ambiguity the market interaction can tame is the one about the information structures of the agents.

4.4 Additional application: price competition

As mentioned above, the previous analysis only depends on the equilibrium equation (14) regardless of the specifics of the underlying games. Here, we provide an alternative foundation of (14) based on a price-competition model. Concretely, n firms are competing on prices. We fix a random variable $\hat{f} \in \mathbb{R}^\Omega$ representing the state of the economy and we let $y \in \mathbb{R}$ denote its realization. The interpretation is that there is aggregate uncertainty about the state y . Each firm i chooses the price $x_i \in \mathbb{R}$ for its good, has 0 production costs, and its payoff function $u_i : \mathbb{R}^n \times \mathbb{R} \rightarrow \mathbb{R}$ depends on the state y as well as the entire profile of prices $x \in \mathbb{R}^n$:

$$u_i(x, y) = D_i(x, y) x_i$$

where $D_i : \mathbb{R}^n \times \mathbb{R} \rightarrow \mathbb{R}$ is the demand function faced by firm i and is defined as

$$D_i(x, y) = \beta \sum_{j \in I} w_{ij} x_j + (1 - \beta) y - \frac{x_i}{2}$$

for some $\beta \in (0, 1)$ and a stochastic and strongly connected matrix W with $w_{jj} = 0$ for all $j \in I$. The demand faced by firm i negatively depends on its own price and positively depends on the state of the economy and on the prices of the other firms, respectively, with coefficients $(1 - \beta)$ and β . As usual, the interpretation is that the firms compete on the same market with partially differentiated products and w_{ij} captures the similarity of products i and j . For the rest of this section we strengthen Assumption 1 by letting $\{(V_i, \Pi_i)\}_{i \in I}$ be a collection of

maxmin (cf. Example 2) interim preferences. In particular, let $C_{i,\omega} \subseteq \Delta(\Omega)$ denote the set of interim probabilistic models of agent i at state ω .

As before, a strategy $\sigma_i \in \Sigma_i$ of agent i is measurable with respect to Π_i . Given a strategy profile $\sigma_{-i} \in \prod_{j \neq i} \Sigma_j$ for the co-players of i , the problem faced by i given state $\omega \in \Omega$ is

$$\max_{x_i \in \mathbb{R}} \min_{p \in C_{i,\omega}} \mathbb{E}_p \left(\left((1 - \beta) \hat{f} + \beta \sum_{j \in I} w_{ij} \sigma_j \right) x_i - \frac{x_i^2}{2} \right).$$

With this, the first-order condition characterizing the equilibrium σ^β for every $\beta \in (0, 1)$ is

$$\sigma_i^\beta(\omega) = \min_{p \in C_{i,\omega}} \mathbb{E}_p \left((1 - \beta) \hat{f} + \beta \sum_{j \in I} w_{ij} \sigma_j^\beta \right) \quad \forall \omega \in \Omega, \forall i \in I, \quad (19)$$

which is just a particular case of equation (14).

5 No trade and betting implications

In this section, we give both necessary and sufficient conditions, in terms of interim trade and betting behavior, for the existence of a common ex-ante expectation.²⁸ For simplicity, we let $I = \{1, 2\}$ and assume that the only feasible acts are $f \in F = [-k, k]^\Omega$, $k \in \mathbb{R}_{++}$. The additional restriction we impose with respect to Section 3 is translation invariance of the interim preferences. The class of variational preferences, considered in Section 4 and in all our examples, satisfies this property.

First, we show that if there exist an asset $f \in F$ and a price $r \in \mathbb{R}$ such that in each state $\omega \in \Omega$, if endowed with the asset player 2 would like to sell it, while player 1 would like to buy it, then there is no nonlinear common ex-ante expectation. Formally, we say that there exists an *interim Pareto improving transaction* if there exists $f \in F$ and $r \in \mathbb{R}$ such that, for all $\omega \in \Omega$, we have $V_1(\omega, f) > r > V_2(\omega, f)$.

Proposition 5. *Let $\{(V_i, \Pi_i)\}_{i \in \{1, 2\}}$ be a collection of full support and translation invariant interim expectations such that $\Pi_{\text{sup}} = \{\Omega\}$. If there is an interim Pareto improving transaction, then there is no nonlinear common ex-ante expectation \bar{V} for $\{(V_i, \Pi_i)\}_{i \in \{1, 2\}}$.*

This result clarifies that common dynamic consistency, even without purely probabilistic beliefs, already implies the absence of trade between the agents. This does not come as a surprise since, as shown by Kajii and Ui [39] for maxmin preferences, and by Martins-da-Rocha [46] for more general preferences, the absence of interim trade is equivalent to the existence of a common local subjective belief $\Theta \neq \emptyset$. In turn, by Proposition 2, the latter is always implied

²⁸Here, we do not consider the interim no-trade characterizations of the existence of V_* and V^* , as well as the existence of a common local subjective belief, i.e., $\Theta \neq \emptyset$. Indeed, the former always exist as shown in Proposition 1, whereas the no-trade implications of the latter have been extensively studied in [39] and [46].

by the existence of a common ex-ante expectation \bar{V} . However, the existence of \bar{V} is in general much stronger than $\Theta \neq \emptyset$, thereby establishing an important difference with the SEU case, where common dynamic consistency is equivalent to the absence of interim Pareto improving transactions (cf. [56]).

The sufficient conditions for the existence of an ex-ante expectation can instead be expressed in terms of the existence of an interim Pareto gain in a large economy with a unit mass of agents endowed with the interim expectation (V_1, Π_1) and a unit mass of agents endowed with the interim expectation (V_2, Π_2) . Formally, a *two-population endowment economy* $\{(\chi_i, V_i, \Pi_i)\}_{i \in \{1,2\}}$ is composed by a pair of measurable functions with finite range $(\chi_1, \chi_2) \in F^{[0,1]} \times F^{[0,1]}$ and a pair of interim expectations $\{(V_i, \Pi_i)\}_{i \in \{1,2\}}$. Here $\chi_i(x)$ is the initial asset position of agent x of population i . We say that an endowment economy is interim Pareto improvable if there exists measurable $(\chi'_1, \chi'_2) \in F^{[0,1]} \times F^{[0,1]}$ such that

1. Market clearing:

$$\int_{[0,1]} \chi_1(x)(\omega) dx + \int_{[0,1]} \chi_2(x)(\omega) dx = \int_{[0,1]} \chi'_1(x)(\omega) dx + \int_{[0,1]} \chi'_2(x)(\omega) dx \quad \forall \omega \in \Omega;$$

2. Interim Pareto improvement:

$$V_i(\omega, \chi_i(x)) < V_i(\omega, \chi'_i(x)) \quad \forall i \in \{1, 2\}, \forall x \in [0, 1], \forall \omega \in \Omega.$$

Theorem 3. *Let $\{(V_i, \Pi_i)\}_{i \in \{1,2\}}$ be a set of full support and translation invariant interim expectations such that $\Pi_{\text{sup}} = \{\Omega\}$. If there is no two-population endowment economy $\{(\chi_i, V_i, \Pi_i)\}_{i \in \{1,2\}}$ that is interim Pareto improvable, then there exists a translation invariant nonlinear common ex-ante expectation \bar{V} for $\{(V_i, \Pi_i)\}_{i \in \{1,2\}}$.*

There are two reasons behind the gap between the necessary and sufficient conditions for the existence of a nonlinear common ex-ante expectation. First, for non-SEU agents the value of shortening a position, $V_i(\omega, -f)$, is in general different from the negative of the value of the position, $-V_i(\omega, f)$. Therefore, to guarantee the existence of a common prior the absence of profitable trade must be verified at every initial asset position, and it is not enough to look at neutral initial positions. Moreover, the nonadditivity of V_i over the different assets implies that ruling out bilateral improvements is not enough, and instead joint transfers between multiple agents must be considered.

6 Related literature

Our work lies at the intersection of several strands of literature, including decision theory, game theory, and information economics. Our Theorem 1 and Corollary 1 generalize to rational preferences the common-prior characterization of Samet [55]. In the case of SEU, the latter has

been previously extended to compact spaces of uncertainty in Hellman [35], and to more general payoff-relevant spaces in Golub and Morris [26]. More generally, both Samet's (for SEU) and our characterization (for rational preferences) can be used to study the implication of mutual ex-ante agreement of the agents. The support condition in Lipman [43] and the critical-path theorem in Kajii and Morris [36] are two standard examples of implications of the common prior assumption. Our work is a first step that provides the framework to obtain similar results in the more general case of rational preferences.

More recently, the existence of a nonlinear common ex-ante expectation for non-ambiguity-neutral preferences but under both dynamic consistency and consequentialism has been studied by Ellis [14]. This paper shows that if the agents' information has a product structure in addition to the previous properties, then their interim preferences cannot exhibit violations of Savage's sure-thing principle for acts that are measurable with respect to the aggregate information.²⁹ However, the following facts limit the implications of this critical result for our analysis: (i) We also consider and characterize weaker versions of common dynamic consistency, which allow for violations of Savage's sure-thing principle (ii) We never impose a product structure for the information of the agents which in turn would rule out hard evidence about the interim types of the opponents (e.g., E-mail-game like information structures have such hard evidence) (iii) For the class of games that we consider in Section 4, even the residual ambiguity about the fundamental state is relevant for the equilibrium outcomes (iv) Even if we impose consequentialism throughout the entire analysis, the latter can be dispensed with for the characterization of existence of a common ex-ante expectation of Corollary 1, provided that the interim preferences satisfy full support.³⁰

Our applications generalize the standard beauty-contest settings in Shin and Williamson [59], Allen et al. [1], or Golub and Morris [27] by allowing for ambiguity aversion and obtaining notable equilibrium implications.³¹ More in general, our work proposes a viable theory for games under incomplete information without SEU. In this regard, Epstein and Wang [17] introduce a universal type space for a class of preferences very similar to the rational one analyzed in the current paper. We improve on this work by characterizing the collections of finite type spaces that admit some degree of ex-ante mutual agreement within this universal type space. Relatedly, we improve on the analysis of incomplete-information games under uncertainty of Kajii and Ui [37] by considering variational preferences and deriving equilibrium properties for a specific class of coordination games. Moreover, we focus here on simultaneous-move games rather than analyzing the effect of ambiguity aversion in multistage-games such as Battigalli et al. [4], and Hanany et al. [31], which in turn provide a very different set of results.

²⁹More in detail, the agents' information has a product structure if each interim type of each player cannot rule out any interim type of the opponents. In [14], this property is implied by Assumption 3 (Full support), which, in general, is not implied by our full-support assumption.

³⁰This fact can be verified by inspection of the proof of Corollary 1.

³¹For analogous models directly phrased as speculative trading see Harrison and Kreps [33] and Morris [50].

Our results in the last part of the paper are related and complementary to the extended literature on no-trade results without SEU. On the one hand, Rigotti et al. [54] and Strzalecki and Werner [63] study efficient allocations under ambiguity with public information, as opposed to the private-information setting of the current paper. On the other hand, Kajii and Ui [38] and [39], and Martins-da-Rocha [46] provide no-trade characterizations of the existence of common ex-ante benchmark beliefs without analyzing the case of full mutual dynamic consistency as we do in Section 5.

Finally, our work is related to the extended literature on updating non-SEU preferences under (relaxations of) consequentialism and dynamic consistency as in Ghirardato [20], Pires [53], Epstein and Schneider [16], Maccheroni et al. [45], Hanany and Klibanoff [30], and Gumen and Savochkin [29]. However, we take an interim approach rather than deriving or studying a given updating rule as in the works above. We derive the ex-ante preferences that are consistent with the given interim ones. This allows us to connect our results to existing updating rules by comparing the prescribed ex-ante preferences with the ones we obtain from the interim preferences and derive new insights into their implications in strategic interactions.

7 Conclusion

The results of this paper can also be used as a stepping stone for further analysis of games beyond SEU. Here we highlight some open questions and future research avenues.

First, as already stressed, despite our analysis following an interim approach, our results can be used in games of incomplete information with general preferences under uncertainty and a *given* set of updating rules. Indeed, the disagreement bound in Proposition 3 and the limit characterization in Theorem 2 did not put any intertemporal restriction on the agents' preferences. So, for example, if all the agents are maxmin, share the same ex-ante set of probability models, and update their beliefs with full Bayesian updating, then our results give tools to study how the equilibrium outcomes change with respect to the agents' private information. Therefore, our results can be a stepping stone toward an information design model in beauty contests under non-SEU preferences.

Second, our framework enables us to revisit some classical results for SEU agents on incomplete information games to understand whether they carry on with more general preferences. An example is a result established in [52] that if a stochastically monotone function (often interpreted as the price of an asset) of the beliefs is common knowledge across the players, their beliefs coincide. The result extends if the information structure is unambiguous but may fail more generally.

Third, our framework and results are the first step toward a general analysis of approximate common knowledge under model uncertainty. The standard analysis based on p -belief operators of Monderer and Samet [47] can be extended to a setting with multiple interim beliefs, for

example, by requiring that all the interim probability models assign probability p to an event. In particular, within this richer framework, we can also ask about the strategic implications of approximate common unambiguity of an event, that is, for each agent, all the interim belief of that agent assigns the same probability to that event. Our examples suggest that this might well have significant strategic consequences such as contagion or taming of ambiguity aversion among agents.

Relatedly, approximate common knowledge has been recently studied in a learning setting under SEU by Frick et al. [19]. Their common-learning results complement Samet's convergence result on higher-order expectations by showing that their KL-divergence relative to the prior distribution decreases monotonically along any sequence. On the one hand, our Theorem 1 offers an alternative distance between higher-order expectations and the prior expectations. On the other hand, our setting can be used to analyze common learning in the presence of ambiguity, for example, by resorting to the learning rules studied in Lanzani [42].

Finally, our analysis is a stepping stone to obtaining sharper equilibrium refinements in complete information games. Indeed, in the SEU world, Kajii and Morris [36] pioneered a robust approach that selects only the subset of equilibria that are limit points of *every* sequence of incomplete information games approximating the original complete information game. An even sharper refinement would only select equilibria that are limit points, including elaborations under incomplete information and non-SEU preferences.

A Appendix: Mathematical preliminaries

In this preliminary section, we introduce and analyze the mathematical tools that we use in the proof of Theorem 1. Our main technical result is Theorem 4 which in turn yields Theorem 1 when applied in the setting of the current paper.

Since Ω is finite, with a small abuse of notation, we equivalently view Ω as the set $J = \{1, \dots, \bar{n}\}$. In this way, \mathbb{R}^Ω is isomorphic to the set of vectors $\mathbb{R}^{\bar{n}}$, where both are endowed with the supnorm. We also denote the elements of the canonical basis of $\mathbb{R}^{\bar{n}}$ by e^j for all $j \in J$. Finally, we denote the vector whose components are all 1s by e : it corresponds to the function 1_Ω in \mathbb{R}^Ω .

In this section, we focus on operators $T : \mathbb{R}^{\bar{n}} \rightarrow \mathbb{R}^{\bar{n}}$. In what follows any such operator will be assumed to be normalized, monotone, and continuous with the exception of Definition 7 and Lemma 2. The composition of normalized, monotone, and continuous operators is an operator which shares the same properties. A normalized, monotone, and continuous operator $T : \mathbb{R}^{\bar{n}} \rightarrow \mathbb{R}^{\bar{n}}$ is linear if and only if there exists a stochastic $\bar{n} \times \bar{n}$ matrix M such that $T(f) = Mf$ for all $f \in \mathbb{R}^{\bar{n}}$. All products of $\bar{n} \times \bar{n}$ matrices are to be intended backward/left, that is, $\Pi_{l=1}^{k+1} M_l = M_{k+1} \Pi_{l=1}^k M_l = M_{k+1} \dots M_1$ for all $k \in \mathbb{N}$. Define $I_{\bar{n}}$ to be the $\bar{n} \times \bar{n}$ identity matrix. Given $j, j' \in J$ we say that j is *strongly monotone with respect to* j' (under T) if and

only if there exists $\varepsilon_{jj'} \in (0, 1)$ such that for each $f \in \mathbb{R}^\Omega$ and for each $\delta \geq 0$

$$T_j(f + \delta e^{j'}) - T_j(f) \geq \varepsilon_{jj'} \delta. \quad (20)$$

We also say that j is constant with respect to j' if and only if for each $f \in \mathbb{R}^\Omega$ and for each $\delta \geq 0$

$$T_j(f + \delta e^{j'}) - T_j(f) = 0. \quad (21)$$

Given T and $j, j' \in J$, it might be the case that neither j is strongly monotone with respect to j' nor j is constant with respect to j' . Therefore, we say that T is *dichotomic* if and only if for each $j, j' \in J$, j is either strongly monotone with respect to j' or constant.

Definition 7. Let T be a monotone operator. We say that $A(T)$ is the indicator matrix of T if and only if its jj' -th entry is such that

$$a_{jj'} = \begin{cases} 1 & j \text{ is strongly monotone wrt } j' \\ 0 & \text{otherwise} \end{cases} \quad \forall j, j' \in J.$$

The indicator matrix $A(M)$ of an $\bar{n} \times \bar{n}$ nonnegative matrix M is defined to be such that $a_{jj'} = 1$ if and only if $m_{jj'} > 0$ and $a_{jj'} = 0$ if and only if $m_{jj'} = 0$. We say that $A(T)$ is *nontrivial* if and only if for each $j \in J$ there exists $j' \in J$ such that $a_{jj'} = 1$. The indicator matrix $A(T)$ of a monotone operator T induces a natural partition of J . Recall that given a nonnegative $\bar{n} \times \bar{n}$ matrix A with *nonnull rows*, we can partition the set $J = \{1, \dots, \bar{n}\}$ with the partition $\{J_l(A)\}_{l=1}^{m_A+1}$ of essential and inessential indexes of A . The first m_A sets consist of the essential classes while $J_{m_A+1}(A)$ consists of all inessential indexes and it might be empty. This is the case if A is symmetric, that is, $a_{jj'} = a_{j'j}$ for all $j, j' \in J$. Instead, there always exists at least a nonempty class of essential indexes $J_1(A)$.³² We call $\Pi(A) = \{J_l(A)\}_{l=1}^{m_A+1}$ the partition of A . When $A = A(T)$ where T is normalized, monotone, and continuous and $A(T)$ is nontrivial, we denote by $\Pi(T)$ the partition $\Pi(A(T))$.

Lemma 2. Let $\{B_k\}_{k \in \{1, \dots, K\}}$ be a finite collection of $\bar{n} \times \bar{n}$ nonnegative matrices such that $b_{k,jj} > 0$ for all $k \in \{1, \dots, K\}$ and for all $j \in J$. If $A(B_k)$ is symmetric for all $k \in \{1, \dots, K\}$, then $A(B_K \dots B_1) \geq A(B_k)$ for all $k \in \{1, \dots, K\}$ and $\Pi(A(B_K \dots B_1))$ is coarser than $\Pi(B_k)$ for all $k \in \{1, \dots, K\}$.

³²We follow Seneta [58]. Denote by $a_{jj'}^{(t)}$ the jj' -th entry of A^t . We write $j \xrightarrow{A} j'$ if and only if $a_{jj'}^{(t)} > 0$ for some $t \in \mathbb{N}$. It is immediate to see that if $j \xrightarrow{A} j'$ and $j' \xrightarrow{A} j''$, then $j \xrightarrow{A} j''$. We also write $j \xleftarrow{A} j'$ if and only if $j \xrightarrow{A} j'$ and $j' \xrightarrow{A} j$. In this case, clearly, we have that $j \xrightarrow{A} j$, that is, $a_{jj}^{(t)} > 0$ for some $t \in \mathbb{N}$. Next, we classify each index $j \in J$ as essential or inessential. An index $j \in J$ is *essential* if and only if for each $j' \in J$

$$j \xrightarrow{A} j' \implies j \xleftarrow{A} j'.$$

We say that j is inessential if and only if it is not essential. Note that there always exists at least one essential index (see Seneta [58, Lemma 1.1]). For each essential $j \in J$, define $[j] = \{j' \in J : j \xleftarrow{A} j'\}$. Note that given two essential indexes j and j' in J we have that either $[j] = [j']$ or $[j] \cap [j'] = \emptyset$. Moreover, given $j, j' \in J$ such that $j \xleftarrow{A} j'$, j is essential if and only if j' is.

We already observed that a normalized, monotone, and continuous operator T is linear if and only if $T(f) = Mf$ for all $f \in \mathbb{R}^{\bar{n}}$ where M is an $\bar{n} \times \bar{n}$ stochastic matrix. Intuitively, the next two results show that dropping the linearity assumption allows M to depend on f . The first result will not impose much discipline on the replicating matrices $M(f)$ while the second one will connect the indicator matrix of $M(f)$ to the one of T .

Lemma 3. *If $T : \mathbb{R}^{\bar{n}} \rightarrow \mathbb{R}^{\bar{n}}$ is normalized, monotone, and continuous, then there exists a compact and convex set $\mathcal{M}(T)$ of $\bar{n} \times \bar{n}$ stochastic matrices such that for each $f \in \mathbb{R}^{\bar{n}}$ there exists $M(f) \in \mathcal{M}(T)$ such that $T(f) = M(f)f$. Moreover, if j is constant with respect to j' , then $m_{jj'} = 0$ for all $M \in \mathcal{M}(T)$.*

The next result builds on [6, Proposition 8].

Proposition 6. *If $T : \mathbb{R}^{\bar{n}} \rightarrow \mathbb{R}^{\bar{n}}$ is normalized, monotone, continuous, and such that $A(T)$ is nontrivial, then there exists a compact and convex set $\mathcal{M}(T)$ of $\bar{n} \times \bar{n}$ stochastic matrices such that $A(M) \geq A(T)$ for all $M \in \mathcal{M}(T)$ and for each $f \in \mathbb{R}^{\bar{n}}$ there exists $M(f) \in \mathcal{M}(T)$ such that $T(f) = M(f)f$. Moreover, if T is dichotomic, then $\mathcal{M}(T)$ can be chosen to be such that $A(M) = A(T)$ for all $M \in \mathcal{M}(T)$.*

Proof. For each $j, j' \in J$ if j is strongly monotone with respect to j' , consider $\varepsilon_{jj'} \in (0, 1)$ as in (20) otherwise let $\varepsilon_{jj'} = 1/2$. Define \tilde{M} to be such that $\tilde{m}_{jj'} = a_{jj'}\varepsilon_{jj'}$ for all $j, j' \in J$ where $a_{jj'}$ is the jj' -th entry of $A(T)$. Since each row of $A(T)$ is not null, for each $j \in J$ there exists $j' \in J$ such that $a_{jj'} = 1$ and, in particular, $\tilde{m}_{jj'} > 0$. This implies that $\sum_{l=1}^{\bar{n}} \tilde{m}_{jl} > 0$ for all $j \in J$. Define also $\varepsilon = \min \left\{ \min_{j \in J} \sum_{l=1}^{\bar{n}} \tilde{m}_{jl}, 1/2 \right\} \in (0, 1)$. Define the stochastic matrix \bar{M} to be such that $\bar{m}_{jj'} = \tilde{m}_{jj'} / \sum_{l=1}^{\bar{n}} \tilde{m}_{jl}$ for all $j, j' \in J$. Clearly, we have that for each $j, j' \in J$, $\bar{m}_{jj'} > 0$ if and only if $\tilde{m}_{jj'} > 0$ if and only if $a_{jj'} = 1$. This yields that $A(\bar{M}) = A(T)$. Next, consider $f, g \in \mathbb{R}^{\bar{n}}$ such that $f \geq g$. Define $g^0 = g$. For each $j' \in \{1, \dots, \bar{n} - 1\}$ define $g^{j'} \in \mathbb{R}^{\bar{n}}$ to be such that $g_j^{j'} = f_j$ for all $j \leq j'$ and $g_j^{j'} = g_j$ for all $j \geq j' + 1$. Define $g^{\bar{n}} = f$. Note that $f = g^{\bar{n}} \geq \dots \geq g^1 \geq g^0 = g$. It follows that

$$\begin{aligned} T_j(f) - T_j(g) &= \sum_{j'=1}^{\bar{n}} \left[T_j(g^{j'}) - T_j(g^{j'-1}) \right] \geq \sum_{j'=1}^{\bar{n}} a_{jj'} \varepsilon_{jj'} (g_{j'}^{j'} - g_{j'}^{j'-1}) \\ &= \sum_{j'=1}^{\bar{n}} \tilde{m}_{jj'} (f_{j'} - g_{j'}) = \left(\sum_{l=1}^{\bar{n}} \tilde{m}_{jl} \right) \left(\sum_{j'=1}^{\bar{n}} \frac{\tilde{m}_{jj'}}{\sum_{l=1}^{\bar{n}} \tilde{m}_{jl}} (f_{j'} - g_{j'}) \right) \\ &= \left(\sum_{l=1}^{\bar{n}} \tilde{m}_{jl} \right) \left(\sum_{j'=1}^{\bar{n}} \bar{m}_{jj'} (f_{j'} - g_{j'}) \right) \geq \varepsilon \sum_{j'=1}^{\bar{n}} \bar{m}_{jj'} (f_{j'} - g_{j'}) \quad \forall j \in J. \end{aligned}$$

This implies that

$$f \geq g \implies T(f) - T(g) \geq \varepsilon \bar{M}(f - g) = \varepsilon (\bar{M}f - \bar{M}g). \quad (22)$$

Define $S : \mathbb{R}^{\bar{n}} \rightarrow \mathbb{R}^{\bar{n}}$ by $S(f) = \frac{T(f) - \varepsilon \bar{M}f}{1 - \varepsilon}$ for all $f \in \mathbb{R}^{\bar{n}}$. By definition of S and (22) and since \bar{M} is a stochastic matrix and T is normalized, monotone, and continuous, it is immediate to see that S is normalized, monotone, and continuous. We can rewrite T to be such that

$$T(f) = \varepsilon \bar{M}f + (1 - \varepsilon) S(f) \quad \forall f \in \mathbb{R}^{\bar{n}}. \quad (23)$$

Consider the set $\mathcal{M}(S)$ of Lemma 3. Define $\mathcal{M}(T) = \varepsilon \bar{M} + (1 - \varepsilon) \mathcal{M}(S)$. Since $\mathcal{M}(S)$ is compact and convex, $A(T) = A(\bar{M})$, and $\varepsilon \in (0, 1)$, it follows that $\mathcal{M}(T)$ is compact and convex and $A(M) \geq A(\bar{M}) = A(T)$ for all $M \in \mathcal{M}(T)$. By (23) and since for each $f \in \mathbb{R}^{\bar{n}}$ there exists $\hat{M}(f) \in \mathcal{M}(S)$ such that $S(f) = \hat{M}(f)f$, for each $f \in \mathbb{R}^{\bar{n}}$ we have that $T(f) = M(f)f$ where $M(f) = \varepsilon \bar{M} + (1 - \varepsilon) \hat{M}(f) \in \mathcal{M}(T)$.

Finally, consider $j, j' \in J$. Since $A(M) \geq A(T)$, if the jj' -entry of $A(T)$ is 1 so is the one of $A(M)$ for all $M \in \mathcal{M}(T)$. Assume that the jj' -entry of $A(T)$ is 0. Since $A(T) = A(\bar{M})$, the jj' -entry of $A(\bar{M})$ is 0 too. Since T is dichotomic, it follows that for each $f \in \mathbb{R}^{\bar{n}}$ and for each $\delta \geq 0$

$$\begin{aligned} \varepsilon \sum_{l=1}^{\bar{n}} \bar{m}_{jl} f_l + (1 - \varepsilon) S_j(f + \delta e^{j'}) &= \varepsilon \sum_{l=1}^{\bar{n}} \bar{m}_{jl} (f_l + \delta e_l^{j'}) + (1 - \varepsilon) S_j(f + \delta e^{j'}) \\ &= T_j(f + \delta e^{j'}) = T_j(f) = \varepsilon \sum_{l=1}^{\bar{n}} \bar{m}_{jl} f_l + (1 - \varepsilon) S_j(f). \end{aligned}$$

Since $\varepsilon \in (0, 1)$, we can conclude that $S_j(f + \delta e^{j'}) = S_j(f)$ for all $f \in \mathbb{R}^{\bar{n}}$ and for all $\delta \geq 0$, that is, j is constant with respect to j' under S . By Lemma 3, we have that $m_{jj'} = 0$ for all $M \in \mathcal{M}(S)$. Since $\mathcal{M}(T) = \varepsilon \bar{M} + (1 - \varepsilon) \mathcal{M}(S)$ and $\bar{m}_{jj'} = 0$, we can conclude that the jj' -entry of $A(M)$ is 0 for all $M \in \mathcal{M}(T)$. Since j and j' were arbitrarily chosen, we can conclude that $A(M) = A(T)$ for all $M \in \mathcal{M}(T)$. \blacksquare

The next lemma is an extension to our framework of Lemma 2 of Samet [55]. In order to discuss it, we need to introduce some notation. Given a stochastic matrix M , we denote by $\delta(M) = \min_{j, j' \in J: m_{jj'} > 0} m_{jj'}$ and $d(M) = \min_{j \in J} m_{jj}$.

Lemma 4. *Let M and \bar{M} be two $\bar{n} \times \bar{n}$ stochastic matrices. If $A(\bar{M})$ is symmetric and $0 < d(\bar{M})$, then we have that $A(\bar{M}M) \geq A(M)$ and*

1. $\delta(\bar{M}M) \geq \delta(M)$, provided $A(\bar{M}M) = A(M)$.
2. $\delta(\bar{M}M) \geq \delta(M) \delta(\bar{M})$, provided $A(\bar{M}M) > A(M)$.

Moreover, if $\{M_k\}_{k=1}^{\infty}$ is a sequence of $\bar{n} \times \bar{n}$ stochastic matrices such that $A(M_k)$ is symmetric, $\delta(M_k) \geq \delta > 0$, and $d(M_k) > 0$ for all $k \in \mathbb{N}$, then

$$\delta\left(\prod_{k=1}^m M_k\right) \geq \delta^{\bar{n}^2} \quad \forall m \in \mathbb{N}. \quad (24)$$

Theorem 4. Let $\{T_i\}_{i \in I}$ be a finite collection of normalized, monotone, and continuous dichotomic operators. If

1. $A(T_i)$ is symmetric for all $i \in I$,
2. $a_{i,jj} = 1$ for all $i \in I$ and for all $j \in J$,
3. the meet of the partitions $\{\Pi(T_i)\}_{i \in I}$ is $\{\Omega\}$,

then for each I -sequence $(i_m)_{m \in \mathbb{N}}$ and for each $f \in \mathbb{R}^{\bar{n}}$ we have that $\lim_{m \rightarrow \infty} T_{i_m} \circ \dots \circ T_{i_1}(f)$ exists and is a constant vector. Moreover, for each I -sequence $(i_m)_{m \in \mathbb{N}}$ and for each $\tau, t \in \mathbb{N}$, if i appears at least τ times in (i_1, \dots, i_t) for all $i \in I$, then

$$\left\| \lim_{m \rightarrow \infty} T_{i_m} \circ \dots \circ T_{i_1}(f) - T_{i_t} \circ \dots \circ T_{i_1}(f) \right\|_{\infty} \leq \left(1 - \delta^{2^{\bar{n}^2} \bar{n}^2}\right)^{\tau 2^{\bar{n}^2} \bar{n}^2 - 1} \|f\|_{\infty},$$

where $\delta = \inf_{i \in I, M \in \mathcal{M}(T_i)} \delta(M) > 0$.

Proof. Define $\hat{t} = 2^{\bar{n}^2}$. By Proposition 6, we have that $I_{\bar{n}} \leq A(T_i) = A(M)$ for all $M \in \mathcal{M}(T_i)$ and for all $i \in I$. Since $\mathcal{M}(T_i)$ is compact for all $i \in I$ and I is finite, this implies that $\delta = \inf_{i \in I, M \in \mathcal{M}(T_i)} \delta(M) > 0$. Define $\hat{\delta} = \delta^{\hat{t} \bar{n}^2} > 0$. Consider $f \in \mathbb{R}^{\bar{n}}$ and an I -sequence $(i_t)_{t \in \mathbb{N}}$. Define $f_t = T_{i_t} \circ \dots \circ T_{i_1}(f) \in \mathbb{R}^{\bar{n}}$ for all $t \in \mathbb{N}$ and set $f_0 = f$. By Proposition 6, there exists a sequence $\{M_t\}_{t \in \mathbb{N}}$ of $\bar{n} \times \bar{n}$ stochastic matrices such that $M_t \in \mathcal{M}(T_{i_t})$ and $T_{i_t}(f_{t-1}) = M_t f_{t-1}$ for all $t \in \mathbb{N}$. Set $t_0 = 0$. Define recursively the following subsequence

$$t_{h+1} = \min \{m > t_h : \{i_{t_h+1}, \dots, i_m\} \supseteq I\} \quad \forall h \geq 0.$$

We next proceed by steps.

Step 1: $A\left(\Pi_{t=t_h+1}^{t_{h+1}} M_t\right) \geq I_{\bar{n}}$ and $\Pi\left(A\left(\Pi_{t=t_h+1}^{t_{h+1}} M_t\right)\right) = \{\Omega\}$ for all $h \in \mathbb{N}_0$.

Proof of the Step. Fix $h \in \mathbb{N}_0$. Since $I_{\bar{n}} \leq A(T_{i_t}) = A(M_t)$ for all $t \in \{t_h + 1, \dots, t_{h+1}\}$, we have that $A(M_t)$ has a strictly positive diagonal and it is symmetric for all $t \in \{t_h + 1, \dots, t_{h+1}\}$. By Lemma 2 and since $\{t_h + 1, \dots, t_{h+1}\} \supseteq I$ and the meet of the partitions $\{\Pi(T_i)\}_{i \in I}$ is $\{\Omega\}$, so is the meet of the partitions $\{\Pi(M_t)\}_{t=t_h+1}^{t_{h+1}}$, yielding that $\Pi(A(M_{t_{h+1}} \dots M_{t_h+1})) = \{\Omega\}$. By Lemma 2, we also have that $A\left(\Pi_{t=t_h+1}^{t_{h+1}} M_t\right) \geq A(M_t) \geq I_{\bar{n}}$ for all $t \in \{t_h + 1, \dots, t_{h+1}\}$. \square

Step 2: $\delta\left(\Pi_{t=t_h+1}^{t_{h+1}} M_t\right) \geq \delta^{\bar{n}^2}$ for all $h \in \mathbb{N}_0$.

Proof of the Step. Fix $h \in \mathbb{N}_0$. By Lemma 4 and since $A(M_t) = A(T_{i_t})$ is symmetric, $\delta(M_t) \geq \delta > 0$, and $d(M_t) > 0$ for all $t \in \mathbb{N}$, the statement follows. \square

Define $\bar{M}_h = \Pi_{t=t_h+1}^{t_{h+1}} M_t$ for all $h \in \mathbb{N}_0$. By Steps 1 and 2 and [58, Lemma 4.8 and Theorem 4.19], we have that $\Pi_{h=0}^m \bar{M}_h$ converges to a stochastic matrix M whose rows coincide to each other and, in particular, that

$$\|M - \Pi_{h=0}^{\tau-1} \bar{M}_h\|_{\infty} \leq \left(1 - \hat{\delta}\right)^{\frac{\tau}{\bar{n}^2} - 1} \quad \forall \tau \in \mathbb{N}.$$

This implies that $\Pi_{l=1}^m M_l \rightarrow M$ and, in particular, that for each $\tau, t \in \mathbb{N}$, if i appears at least τ times in (i_1, \dots, i_t) for all $i \in I$, then

$$\|M - \Pi_{l=1}^t M_t\|_\infty \leq \|M - \Pi_{h=0}^{\tau-1} \bar{M}_h\|_\infty \leq (1 - \hat{\delta})^{\frac{\tau}{t}-1}.$$

Finally, it follows that

$$\lim_{m \rightarrow \infty} T_{i_m} \circ \dots \circ T_{i_1}(f) = \lim_{m \rightarrow \infty} \Pi_{l=1}^m M_l f = Mf,$$

and, in particular, that for each $\tau, t \in \mathbb{N}$, if i appears at least τ times in (i_1, \dots, i_t) for all $i \in I$, then

$$\begin{aligned} \left\| \lim_{m \rightarrow \infty} T_{i_m} \circ \dots \circ T_{i_1}(f) - T_{i_t} \circ \dots \circ T_{i_1}(f) \right\|_\infty &= \|Mf - (\Pi_{l=1}^t M_t) f\|_\infty \leq (1 - \hat{\delta})^{\frac{\tau}{t}-1} \|f\|_\infty \\ &= (1 - \delta^{2^{\bar{n}^2}})^{\tau 2^{\bar{n}^2}-1} \|f\|_\infty \end{aligned}$$

proving the statement. ■

B Appendix: Section 3

In this section, we prove the main results stated in Section 3. The proofs of the secondary results in Section 3 and the proofs of the ancillary results stated here are relegated to Online Appendix E. Here we use the mathematical preliminaries previously discussed. For such a reason, we equivalently refer to \mathbb{R}^Ω and $\mathbb{R}^{\bar{n}}$, since they are isomorphic.

Lemma 5. *If $\bar{V} : \mathbb{R}^\Omega \rightarrow \mathbb{R}$ is an ex-ante expectation, then it is continuous at constant functions.*

Lemma 6. *Let (V, Π) be an interim expectation with full support. The following statements are equivalent:*

- (i) $a_{jj'} = 1$;
- (ii) $\Pi(\omega_j) = \Pi(\omega_{j'})$.

In particular, $A(V)$ is symmetric, $a_{jj} = 1$ for all $j \in J$, $\Pi(V) = \Pi$, and V is dichotomic.

Proof of Theorem 1. By Lemma 6 and since $\{(V_i, \Pi_i)\}_{i \in I}$ is a finite set of full support interim expectations, we have that $A(V_i)$ is symmetric, $\Pi(V_i) = \Pi_i$, and V_i is dichotomic for all $i \in I$. Moreover, we have that $a_{i,jj} = 1$ for all $j \in J$ and for all $i \in I$. By Theorem 4 and since the meet of $\{\Pi(V_i)\}_{i \in I}$ is $\{\Omega\}$, we can conclude that for each I -sequence $\iota = (i_t)_{t \in \mathbb{N}}$ and for each $f \in \mathbb{R}^\Omega$ we have that $\lim_{m \rightarrow \infty} V_{i_m} \circ \dots \circ V_{i_1}(f) = k_{\iota,f} 1_\Omega$ for some $k_{\iota,f} \in \mathbb{R}$. Moreover, there exist

$\hat{\delta} = \left(\inf_{i \in I, M \in \mathcal{M}(V_i)} \delta(M) \right)^{2^{\bar{n}^2}} \in (0, 1)$ and $\hat{t} = 2^{\bar{n}^2} \in \mathbb{N}$ such that for each I -sequence $(i_m)_{m \in \mathbb{N}}$ and for each $\tau, t \in \mathbb{N}$, if i appears at least τ times in (i_1, \dots, i_t) for all $i \in I$, then

$$\|k_{f,\iota} 1_\Omega - V_{i_t} \circ \dots \circ V_{i_1}(f)\|_\infty \leq \left(1 - \hat{\delta}\right)^{\frac{\tau}{\hat{t}} - 1} \|f\|_\infty.$$

Finally, the last part of the statement follows from the previous claim by setting $C = \frac{1}{1-\hat{\delta}}$ and $\varepsilon = \left(1 - \hat{\delta}\right)^{\frac{1}{\hat{t}}}$. ■

Lemma 7. *The sets \mathcal{V}_\circ and \mathcal{V}° are nonempty and V_\star and V^\star are well defined and respectively a lower and an upper common ex-ante expectation for $\{(V_i, \Pi_i)\}_{i \in I}$.*

Before proving Proposition 1, we define $V_\star : \mathbb{R}^\Omega \rightarrow \mathbb{R}$ and $V^\star : \mathbb{R}^\Omega \rightarrow \mathbb{R}$ by

$$V_\star(f) = \inf_{\iota \in I^\mathbb{N} : \iota \text{ is an } I\text{-sequence}} \bar{V}_\iota(f) \text{ and } V^\star(f) = \sup_{\iota \in I^\mathbb{N} : \iota \text{ is an } I\text{-sequence}} \bar{V}_\iota(f) \quad \forall f \in \mathbb{R}^\Omega.$$

Clearly, we have that $V_\star \leq V^\star$.

Proof of Proposition 1. The first part of the statement immediately follows by Lemma 7. Since V_\star (resp. V^\star) is a pointwise infimum (resp. supremum) of normalized and monotone functionals, so is V_\star (resp. V^\star). Fix $f \in \mathbb{R}^\Omega$ and $i \in I$. Consider also an I -sequence ι' . Since $\{(V_i, \Pi_i)\}_{i \in I}$ exhibits convergence to a deterministic limit, we have $k_{V_i(f), \iota'} 1_\Omega = \lim_{t \rightarrow \infty} V_{i'_t} \circ V_{i'_{t-1}} \circ \dots \circ V_{i'_2} \circ V_{i'_1}(V_i(f)) = \lim_{t \rightarrow \infty} V_{i'_t} \circ V_{i'_{t-1}} \circ \dots \circ V_{i'_2} \circ V_{i'_1}(f) = k_{f, \iota''} 1_\Omega$ where ι'' is the I -sequence such that $\iota''_1 = i$ and $\iota''_t = \iota'_{t-1}$ for all $t \in \mathbb{N} \setminus \{1\}$. This implies that

$$k_{V_i(f), \iota'} = k_{f, \iota''} \geq \inf_{\iota \in I^\mathbb{N} : \iota \text{ is an } I\text{-sequence}} k_{f, \iota} = \inf_{\iota \in I^\mathbb{N} : \iota \text{ is an } I\text{-sequence}} \bar{V}_\iota(f) = V_\star(f).$$

Since ι' was arbitrarily chosen, this implies that

$$V_\star(V_i(f)) = \inf_{\iota \in I^\mathbb{N} : \iota \text{ is an } I\text{-sequence}} \bar{V}_\iota(V_i(f)) = \inf_{\iota \in I^\mathbb{N} : \iota \text{ is an } I\text{-sequence}} k_{V_i(f), \iota} \geq V_\star(f),$$

proving that $V_\star \in \mathcal{V}_\circ$. Next, consider $V' \in \mathcal{V}_\circ$ and suppose by contradiction that $V'(g) > V_\star(g)$ for some $g \in \mathbb{R}^\Omega$. Since $V'(g) > V_\star(g)$, there exists an I sequence ι' such that $V'(g) 1_\Omega > \lim_{t \rightarrow \infty} V_{i'_t} \circ V_{i'_{t-1}} \circ \dots \circ V_{i'_2} \circ V_{i'_1}(g) = k_{g, \iota'} 1_\Omega$. Since V' is normalized and continuous at $k_{g, \iota'} 1_\Omega$ by Lemma 5,

$$\begin{aligned} V'(g) &= V'(V'(g) 1_\Omega) > V'(k_{g, \iota'} 1_\Omega) = V' \left(\lim_{t \rightarrow \infty} V_{i'_t} \circ V_{i'_{t-1}} \circ \dots \circ V_{i'_2} \circ V_{i'_1}(g) \right) \\ &= \lim_{t \rightarrow \infty} V' \circ V_{i'_t} \circ V_{i'_{t-1}} \circ \dots \circ V_{i'_2} \circ V_{i'_1}(g) \geq V'(g), \end{aligned}$$

a contradiction. This proves that $V_\star = V_\star$. A symmetric argument shows that $V^\star = V^\star$. ■

Denote by P the set of permutations of agents, that is, bijections $\rho : \{1, \dots, n\} \rightarrow \{1, \dots, n\}$. Given $\rho \in P$, we denote by $V_\rho : \mathbb{R}^\Omega \rightarrow \mathbb{R}^\Omega$ the operator defined by

$$V_\rho = V_{\rho(1)} \circ V_{\rho(2)} \circ \dots \circ V_{\rho(n)}. \tag{25}$$

As usual, we also denote by V_ρ^t the composition $\underbrace{V_\rho \circ \dots \circ V_\rho}_{t\text{-times}}$ for all $t \in \mathbb{N}$ and for all $\rho \in P$.

Proof of Corollary 1. We prove the equivalence between (i) and (ii). The equivalence between (ii) and (iii) immediately follows by Proposition 1.

(i) implies (ii). By assumption, for each I -sequence $\iota = (\iota_t)_{t \in \mathbb{N}}$ and for each $f \in \mathbb{R}^\Omega$ we have that $\lim_{m \rightarrow \infty} V_{\iota_m} \circ \dots \circ V_{\iota_1}(f) = k_{\iota, f} 1_\Omega$ for some $k_{\iota, f} \in \mathbb{R}$. By Lemma 5 and since \bar{V} is an ex-ante expectation and (\bar{V}, V_i, Π_i) is a nonlinear conditional expectation, we have that

$$\begin{aligned} k_{\iota, f} &= \bar{V}(k_{\iota, f} 1_\Omega) = \bar{V}\left(\lim_{m \rightarrow \infty} V_{\iota_m} \circ \dots \circ V_{\iota_1}(f)\right) = \lim_{m \rightarrow \infty} \bar{V}(V_{\iota_m} \circ \dots \circ V_{\iota_1}(f)) \\ &= \lim_{m \rightarrow \infty} \bar{V}(V_{\iota_{m-1}} \circ \dots \circ V_{\iota_1}(f)) = \dots = \lim_{m \rightarrow \infty} \bar{V}(V_{\iota_1}(f)) = \bar{V}(f), \end{aligned}$$

proving the implication.

(ii) implies (i). Fix a permutation $\bar{\rho} \in P$. Define the I -sequence $(i_k)_{k \in \mathbb{N}}$ by $i_k = \bar{\rho}(k \bmod n)$ for all $k \in \mathbb{N}$ such that $k \bmod n \neq 0$ and $i_k = \bar{\rho}(n)$ for all $k \in \mathbb{N}$ such that $k \bmod n = 0$. Define $\hat{V} : \mathbb{R}^\Omega \rightarrow \mathbb{R}^\Omega$ by $\hat{V}(f) = \lim_{\tau \rightarrow \infty} V_{\bar{\rho}}^\tau(f)$ for all $f \in \mathbb{R}^\Omega$. By assumption, we have that \hat{V} is well defined and $\hat{V}(f)$ is a constant function for all $f \in \mathbb{R}^\Omega$. Since $V_{\bar{\rho}}$ is the composition of normalized, monotone, and continuous operators, so is $V_{\bar{\rho}}^\tau$ for all $\tau \in \mathbb{N}$ and, by passing to the limit, \hat{V} is normalized and monotone. By assumption, we also have that $\hat{V}(f) = \lim_{\tau \rightarrow \infty} V_{\bar{\rho}}^\tau(f)$ for all $f \in \mathbb{R}^\Omega$ and for all $\rho \in P$. Since \hat{V} is normalized and monotone and $\hat{V}(f)$ is a constant function for all $f \in \mathbb{R}^\Omega$, we also have that $\hat{V}(\hat{V}(f)) = \hat{V}(f)$ for all $f \in \mathbb{R}^\Omega$, that is, $\hat{V} \circ \hat{V} = \hat{V}$. Define also $\bar{V} : \mathbb{R}^\Omega \rightarrow \mathbb{R}$ by $\bar{V}(f) = \hat{V}_1(f)$ for all $f \in \mathbb{R}^\Omega$. Since $\hat{V} \circ \hat{V} = \hat{V}$, it is immediate to see that \bar{V} is an ex-ante expectation such that $\bar{V} \circ \hat{V} = \bar{V}$. This implies that for each $f \in \mathbb{R}^\Omega$ and for each $\rho \in P$

$$\bar{V}(V_\rho(f)) = \bar{V}(\hat{V}(V_\rho(f))) = \bar{V}\left(\lim_{\tau \rightarrow \infty} V_\rho^\tau(V_\rho(f))\right) = \bar{V}\left(\lim_{\tau \rightarrow \infty} V_\rho^{\tau+1}(f)\right) = \bar{V}(\hat{V}(f)) = \bar{V}(f). \quad (26)$$

Consider $i \in I$. Consider any permutation such that $\tilde{\rho}(1) = i$. By (26), we have that $\bar{V} \circ V_{\tilde{\rho}} \circ V_i = \bar{V} \circ V_i$. Consider the permutation $\hat{\rho}$ such that $\hat{\rho}(i') = \tilde{\rho}(i' + 1)$ for all $i' \in \{1, \dots, n-1\}$ and $\hat{\rho}(n) = i$. Define also $\tilde{V} = \bar{V} \circ V_i$. It follows that \tilde{V} is an ex-ante expectation. Since $\bar{V} \circ V_{\tilde{\rho}} \circ V_i = \bar{V} \circ V_i$, we can conclude that $\tilde{V} \circ V_{\tilde{\rho}} = \bar{V} \circ V_i \circ V_{\tilde{\rho}} = \bar{V} \circ V_{\tilde{\rho}} \circ V_i = \bar{V} \circ V_i = \tilde{V}$. By induction, this implies that $\tilde{V} \circ V_{\tilde{\rho}}^\tau = \bar{V} \circ V_i = \tilde{V}$ for all $\tau \in \mathbb{N}$. By (26) and Lemma 5 and since \tilde{V} is an ex-ante expectation, $\bar{V} \circ \hat{V} = \bar{V}$, and $\tilde{V} \circ V_{\tilde{\rho}}^\tau = \bar{V} \circ V_i = \tilde{V}$ for all $\tau \in \mathbb{N}$, we can conclude that

$$\begin{aligned} \bar{V}(f) &= \bar{V}(\hat{V}(f)) = \bar{V}(V_i(\hat{V}(f))) = \tilde{V}(\hat{V}(f)) = \tilde{V}\left(\lim_{\tau \rightarrow \infty} V_{\tilde{\rho}}^\tau(f)\right) \\ &= \lim_{\tau \rightarrow \infty} \tilde{V}(V_{\tilde{\rho}}^\tau(f)) = \bar{V}(V_i(f)) \quad \forall f \in \mathbb{R}^\Omega, \end{aligned}$$

yielding that $\bar{V} \circ V_i = \bar{V}$. Since i was arbitrarily chosen, the statement follows. \blacksquare

C Appendix: Section 4

In this section, we prove the main results stated in Section 4. The proofs of the secondary results in Section 4 and the proofs of the ancillary results stated here are relegated to Online Appendix E.

The elements of $(\mathbb{R}^\Omega)^n$ are vectors of n components, \mathbf{f} , where each component i , f_i , is an element of \mathbb{R}^Ω . We endow $(\mathbb{R}^\Omega)^n$ with the norm $\|\cdot\|_* : (\mathbb{R}^\Omega)^n \rightarrow [0, \infty)$ defined by $\|\mathbf{f}\|_* = \sup_{i \in I} \|f_i\|_\infty$ for all $\mathbf{f} \in (\mathbb{R}^\Omega)^n$. Define $\hat{\mathbf{f}} \in (\mathbb{R}^\Omega)^n$ as $\hat{f}_i = \hat{f}$ for all $i \in I$.

Lemma 8. *Let $S_\beta : (\mathbb{R}^\Omega)^n \rightarrow (\mathbb{R}^\Omega)^n$ be defined by equation (15). If $\beta \in (0, 1]$, then S_β is a β -contraction. In particular, for each $\beta \in (0, 1)$, there exists a unique $\sigma^\beta \in (\mathbb{R}^\Omega)^n$ such that*

$$S_\beta^\tau(\hat{\mathbf{f}}) \xrightarrow{\|\cdot\|_*} \sigma^\beta, S_\beta(\sigma^\beta) = \sigma^\beta, \text{ and } \|\sigma^\beta\|_* \leq \|\hat{f}\|_\infty.$$

Lemma 9. *Let $S_\beta : (\mathbb{R}^\Omega)^n \rightarrow (\mathbb{R}^\Omega)^n$ be defined by equation (15) and $\mathbf{f} \in (\mathbb{R}^\Omega)^n$. The following statements are equivalent:*

- (i) $S_1(\mathbf{f}) = \mathbf{f}$;
- (ii) *There exists $m \in \mathbb{R}$ such that $f_i = f_{i'} = m1_\Omega$ for all $i, i' \in I$.*

Recall that $\bar{n} = |\Omega|$. For every monotone operator $R : (\mathbb{R}^\Omega)^n \rightarrow (\mathbb{R}^\Omega)^n$ define the adjacency matrices $\underline{A}(R), \bar{A}(R) \in \{0, 1\}^{(n \times \bar{n}) \times (n \times \bar{n})}$ as follows. For every $i, j \in I$ and $\omega, \omega' \in \Omega$, we set $\underline{a}_{(i, \omega)(j, \omega')}(R) = 1$ if and only if there exists $\varepsilon_{(i, \omega)(j, \omega')} > 0$ such that for each $\mathbf{f} \in (\mathbb{R}^\Omega)^n$ and $\delta \geq 0$,

$$R_{i, \omega}(\mathbf{f} + \delta e^{j, \omega'}) - R_{i, \omega}(\mathbf{f}) \geq \varepsilon_{(i, \omega)(j, \omega')} \delta,$$

and we set $\bar{a}_{(i, \omega)(j, \omega')}(R) = 1$ if and only if there exist $\mathbf{f} \in (\mathbb{R}^\Omega)^n$ and $\delta \geq 0$ such that

$$R_{i, \omega}(\mathbf{f} + \delta e^{j, \omega'}) - R_{i, \omega}(\mathbf{f}) > 0.$$

Moreover, we say that a class of indices $Z, \emptyset \neq Z \subseteq I \times \Omega$, is closed and strongly connected with respect to an adjacency matrix $A \in \{0, 1\}^{(n \times \bar{n}) \times (n \times \bar{n})}$ if and only if (i) for each $z, z' \in Z$ there exists a path $\{z_l\}_{l=1}^K \subseteq Z$ such that $a_{z_l z_{l+1}} = 1$ for all $l \in \{1, \dots, K-1\}$, $z_1 = z$ and $z_K = z'$; (ii) for each $z \in Z$, $a_{zz'} = 1$ implies $z' \in Z$.

Lemma 10. *Let $S_\beta : (\mathbb{R}^\Omega)^n \rightarrow (\mathbb{R}^\Omega)^n$ be defined by equation (15). There exists a unique class of indices $Z, \emptyset \neq Z \subseteq I \times \Omega$, that is closed and strongly connected with respect to $\underline{A}(S_1)$ and, in addition, every row of $\underline{A}(S_1)$ is not null.*

Proof of Proposition 3. By Lemma 8, it follows that, for every $\beta \in (0, 1)$, S_β is a contraction with respect to the supnorm and it admits a unique fixed point $\sigma^\beta \in \Sigma$. With this, the result follows by Lemma 10 and applying [7, Theorem 2] with $T = S_1$. \blacksquare

Next, let $\mathcal{W} \subseteq \mathbb{R}_+^{(n \times \bar{n}) \times (n \times \bar{n})}$ denote the set of stochastic matrices over $I \times \Omega$ and define

$$\partial S_1(0) = \left\{ \hat{W} \in \mathcal{W} : \forall (i, \omega) \in I \times \Omega, w_{i,\omega} \in \partial S_{1,i,\omega}(0) \right\},$$

where $\partial S_{1,i,\omega}(0) \subseteq \Delta(I \times \Omega)$ is the superdifferential of the concave functional $S_{1,i,\omega}$ at 0.

Lemma 11. *Let $S_\beta : (\mathbb{R}^\Omega)^n \rightarrow (\mathbb{R}^\Omega)^n$ be defined by equation (15). For each $\beta \in (0, 1)$, we have*

$$V_* \left(S_{\beta,i}^\tau(\hat{\mathbf{f}}) \right) \geq V_* \left(\hat{f} \right) \quad \forall i \in I, \forall \tau \in \mathbb{N},$$

where $\hat{\mathbf{f}} \in (\mathbb{R}^\Omega)^n$ is such that $\hat{f}_i = \hat{f}$ for all $i \in I$. Moreover, we have $V_* \left(\sigma_i^\beta \right) \geq V_* \left(\hat{f} \right)$ for all $i \in I$ and for all $\beta \in (0, 1)$.

Lemma 12. *If a collection of variational interim expectations $\{(V_i, \Pi_i)\}_{i \in I}$ has full support, and is such that $\Pi_{\text{sup}} = \{\Omega\}$ then $\Theta \subseteq \text{int}(\Delta(\Omega))$.*

Let $s \in \text{int}(\Delta(I))$ denote the unique probability vector that satisfies $s = sW$, where uniqueness and strict positivity follow from the fact that W is strongly connected.

Proof of Theorem 2. First, recall that S_1 is normalized, monotone, translation invariant, concave and, by Lemma 9, $S_1(\mathbf{f}) = \mathbf{f}$ if and only if there exists $m \in \mathbb{R}$ such that $f_i = f_{i'} = m1_\Omega$ for all $i, i' \in I$. With this, for all $(i, \tilde{\omega}) \in I \times \Omega$,

$$\lim_{\beta \rightarrow 1} \sigma_i^\beta(\tilde{\omega}) = \min_{\{\eta \in \Delta(I \times \Omega) : \exists q \in Q, \eta = \eta W^q\}} \sum_{(i,\omega) \in I \times \Omega} \eta_{i,\omega} \hat{f}(\omega) = \min_{q \in Q} \sum_{(i,\omega) \in I \times \Omega} \gamma_{i,\omega}^q \hat{f}(\omega),$$

where the first equality follows by [7, Corollary 2] and the second equality follows by Lemma 1. Next, fix $q \in Q$ and observe that

$$\begin{aligned} \sum_{(i,\omega) \in I \times \Omega} \gamma_{i,\omega}^q \mathbb{E}_{q_{i,\omega}}(\hat{f}) &= \sum_{(i,\omega) \in I \times \Omega} \gamma_{i,\omega}^q \mathbb{E}_{q_{i,\omega}} \left(\sum_{j=1}^n w_{ij} \hat{f} \right) = \sum_{(i,\omega) \in I \times \Omega} \gamma_{i,\omega}^q \left[\sum_{(j,\omega') \in I \times \Omega} q_{i,\omega}(\omega') w_{ij} \hat{f}(\omega) \right] \\ &= \sum_{(i,\omega) \in I \times \Omega} \gamma_{i,\omega}^q \left[\sum_{(j,\omega') \in I \times \Omega} w_{(i,\omega)(j,\omega')}^q \hat{f}(\omega) \right] = \sum_{(i,\omega) \in I \times \Omega} \gamma_{i,\omega}^q \hat{f}(\omega), \end{aligned}$$

where the third equality follows from the definition of W^q and the last equality follows from the fact that $\gamma^q = \gamma^q W^q$. This proves the equality in (18).

We now prove the left inequality in (18). Fix $\bar{i} \in I$. By the previous part, we know that there exists $m \in \mathbb{R}$ such that $\lim_{\beta \rightarrow 1} \sigma_{\bar{i}}^\beta(\omega) = m$ for all $(i, \omega) \in I \times \Omega$. By contradiction, assume that $V_* \left(\hat{f} \right) > m$. By Lemmas 5 and 11, we can conclude that

$$m = V_*(m1_\Omega) = \lim_{\beta \rightarrow 1} V_* \left(\sigma_{\bar{i}}^\beta \right) \geq V_* \left(\hat{f} \right) > m$$

yielding a contradiction.

Next, we prove the right inequality in (18). First, observe that, if $\Theta = \emptyset$, then $\inf_{\mu \in \Theta} \mathbb{E}_\mu(\hat{f}) = \infty$ and the right inequality in (18) trivially holds. Next, assume that $\Theta \neq \emptyset$ and fix $\mu \in \Theta$. In particular, we have $\mu \in \text{int}(\Delta(\Omega))$ by Lemma 12, hence $p_{\mu,i}(\omega, \cdot)$ is uniquely defined for all $(i, \omega) \in I \times \Omega$. By definition of Θ we have that $p_{\mu,i}(\omega, \cdot) = p_{\mu,i}(\omega', \cdot)$ and $c_{i,\omega}(p_{\mu,i}(\omega, \cdot)) = 0$ for all $(i, \omega) \in I \times \Omega$ and $\omega' \in \Pi_i(\omega)$. With this, define $q^\mu \in \Delta(\Omega)^{I \times \Omega}$ as $q_{i,\omega}^\mu = p_{\mu,i}(\omega, \cdot)$ for all $(i, \omega) \in I \times \Omega$ and observe that $q^\mu \in Q$ by construction. With this, by Lemma 1 there exists a unique probability vector $\gamma^{q^\mu} \in \Delta(I \times \Omega)$ such that $\gamma^{q^\mu} = \gamma^{q^\mu} W^{q^\mu}$. Now, define $\gamma^\mu \in \Delta(I \times \Omega)$ as $\gamma_{i,\omega}^\mu = s_i \mu(\omega)$ for all $(i, \omega) \in I \times \Omega$. Observe that, for all $(i, \omega) \in I \times \Omega$, we have

$$\begin{aligned} \sum_{(j,\omega') \in I \times \Omega} \gamma_{j,\omega'}^\mu w_{(j,\omega')(i,\omega)}^{q^\mu} &= \sum_{(j,\omega') \in I \times \Omega} s_j \mu(\omega') w_{ji} q_{j,\omega'}^\mu(\omega) = \sum_{j \in I} s_j w_{ji} \sum_{\omega' \in \Omega} \mu(\omega') p_{\mu,j}(\omega', \omega) \\ &= \mu(\omega) \sum_{j \in I} s_j w_{ji} = \mu(\omega) s_i = \gamma_{i,\omega}^\mu. \end{aligned}$$

This show that $\gamma^\mu = \gamma^\mu W^{q^\mu}$, proving that $\gamma^\mu = \gamma^{q^\mu}$. With this, we have

$$\begin{aligned} \mathbb{E}_\mu(\hat{f}) &= \sum_{i \in I} s_i \mathbb{E}_\mu(\hat{f}) = \sum_{(i,\omega) \in I \times \Omega} s_i \mu(\omega) \mathbb{E}_{p_{\mu,i}(\omega, \cdot)}(\hat{f}) = \sum_{(i,\omega) \in I \times \Omega} \gamma_{i,\omega}^\mu \mathbb{E}_{q_{i,\omega}^\mu}(\hat{f}) \\ &= \sum_{(i,\omega) \in I \times \Omega} \gamma_{i,\omega}^{q^\mu} \mathbb{E}_{q_{i,\omega}^\mu}(\hat{f}) \geq \min_{q \in Q} \sum_{(i,\omega) \in I \times \Omega} \gamma_{i,\omega}^q \mathbb{E}_{q_{i,\omega}^q}(\hat{f}). \end{aligned}$$

Given that $\mu \in \Theta$ was arbitrarily chosen, the right inequality in (18) follows.

The second part of the statement directly follows by the first part and by Theorem 1 and Corollary 1 (left inequality) and Proposition 2 (right inequality). \blacksquare

Lemma 13. *Let $\{(V_i, \Pi_i)\}_{i \in I}$ be a collection of interim expectations that exhibits convergence to a deterministic limit. The following facts are true*

1. *If V_i is concave for all $i \in I$, then V_* is concave. If in addition V_i is positive homogeneous (resp. translation invariant) for all $i \in I$, then V_* is positive homogeneous (resp. translation invariant).*
2. *If V_i is Π_{inf} -affine for all $i \in I$, then $V_*((1-\lambda)f + \lambda g) \geq (1-\lambda)V_*(f) + \lambda V_*(g)$ and $V^*((1-\lambda)f + \lambda g) \leq (1-\lambda)V^*(f) + \lambda V^*(g)$ for all $\lambda \in (0, 1)$ and for all $f, g \in \mathbb{R}^\Omega$ where g is Π_{inf} -measurable.*

Proof of Proposition 4. Fix $\beta \in (0, 1)$. By Lemma 8, we have that $\sigma_i^\beta = S_{\beta,i}(\sigma^\beta) = V_i((1-\beta)\hat{f} + \beta \sum_{l=1}^n w_{il} \sigma_l^\beta)$ for all $i \in I$. This implies that σ_i^β is Π_i -measurable and, in particular, Π_{inf} -measurable for all $i \in I$. Since V_i is Π_{inf} -affine, this implies that

$$\sigma_i^\beta = V_i\left((1-\beta)\hat{f} + \beta \sum_{l=1}^n w_{il} \sigma_l^\beta\right) = (1-\beta)V_i(\hat{f}) + \beta \sum_{l=1}^n w_{il} V_i(\sigma_l^\beta) \quad \forall i \in I. \quad (27)$$

By Lemma 13, since V_i is Π_{\inf} -affine for every $i \in I$, we have that V_* is such that

$$V_*((1 - \alpha)h + \alpha g) \geq (1 - \alpha)V_*(h) + \alpha V_*(g) \quad (28)$$

and V^* is such that

$$V^*((1 - \alpha)h + \alpha g) \leq (1 - \alpha)V^*(h) + \alpha V^*(g) \quad (29)$$

for all $\alpha \in (0, 1)$ and for all $g, h \in \mathbb{R}^\Omega$ where g is Π_{\inf} -measurable. By (27), (28), (29) and since each $V_i(\hat{f})$ is Π_i -measurable, hence Π_{\inf} -measurable, we have that, for each $i \in I$,

$$V_*(\sigma_i^\beta) = V_*\left((1 - \beta)V_i(\hat{f}) + \beta \sum_{l=1}^n w_{il}V_i(\sigma_l^\beta)\right) \geq (1 - \beta)V_*(\hat{f}) + \beta \sum_{l=1}^n w_{il}V_*(\sigma_l^\beta),$$

and

$$V^*(\sigma_i^\beta) = V^*\left((1 - \beta)V_i(\hat{f}) + \beta \sum_{l=1}^n w_{il}V_i(\sigma_l^\beta)\right) \leq (1 - \beta)V^*(\hat{f}) + \beta \sum_{l=1}^n w_{il}V^*(\sigma_l^\beta).$$

Define $x_* \in \mathbb{R}^n$ to be such that $x_{*i} = V_*(\sigma_i^\beta) - V_*(\hat{f})$ for all $i \in I$. We can conclude that $x_* \geq \beta W x_*$. Assume by contradiction that $x_{*i'} = \min_{i \in I} x_{*i} < 0$. Since W is a stochastic matrix, we have $x_{*i'} \leq (W x_*)_{i'}$. Since $\beta \in (0, 1)$ was arbitrarily chosen, it follows that $x_{*i'} < \beta (W x_*)_{i'}$, yielding the contradiction

$$x_{*i'} < \beta (W x_*)_{i'} \leq x_{*i'}.$$

Therefore, we must have $V_*(\sigma_i^\beta) \geq V_*(\hat{f})$ for all $i \in I$ and for all $\beta \in (0, 1)$. By taking the limit for $\beta \rightarrow 1$ in the previous inequality and by Lemma 5 and Theorem 2, we get $\lim_{\beta \rightarrow 1} \sigma_i^\beta(\omega) \geq V_*(\hat{f})$ for all $\omega \in \Omega$ and for all $i \in I$. Analogous steps yield that $\lim_{\beta \rightarrow 1} \sigma_i^\beta(\omega) \leq V^*(\hat{f})$ for all $\omega \in \Omega$ and for all $i \in I$. The second part of the statement follows from the first part, Theorem 1, and Corollary 1. \blacksquare

D Appendix: Section 5

In this section, we prove the results stated in Section 5. The proofs of the ancillary results stated here are relegated to Online Appendix E.

Let $\mathcal{M}(F)$ denote the set of countably additive measures over F , and as $\mathcal{M}_0(F)$ the subset of measures in $\mathcal{M}(F)$ with finite support.

Proof of Proposition 5. Suppose there exists a common ex-ante expectation \bar{V} . By definition, we have that

$$\bar{V}(f) = \bar{V}(V_1(f)) > r > \bar{V}(V_2(f)) = \bar{V}(f),$$

yielding a contradiction. \blacksquare

Lemma 14. Let $\{(V_i, \Pi_i)\}_{i \in \{1,2\}}$ be a collection of full support and translation invariant interim expectations such that $\Pi_{\text{sup}} = \{\Omega\}$. If there is no $\nu \in \mathcal{M}_0(F)$ such that

$$\int_F V_1(\omega, f) d\nu(f) > 0 > \int_F V_2(\omega, f) d\nu(f) \quad \forall \omega \in \Omega,$$

then there exists a translation invariant ex-ante expectation \bar{V} such that (\bar{V}, V_i, Π_i) is a non-linear conditional expectation for all $i \in \{1, 2\}$.

Proof of Theorem 3. We show that if there is $\nu \in \mathcal{M}_0(F)$ such that

$$\int_F V_1(\omega, f) d\nu(f) > 0 > \int_F V_2(\omega, f) d\nu(f) \quad \forall \omega \in \Omega, \quad (30)$$

then there exists a two-population endowment economy $\{(\chi_i, V_i, \Pi_i)\}_{i \in \{1,2\}}$ that is interim Pareto improvable. By Lemma 14, this proves the statement. We define the endowment economy in the following way. Let ν^+ and ν^- be the positive and negative components of ν , and enumerate their respective finite supports as $S^+ = (f_1, \dots, f_m)$ and $S^- = (g_1, \dots, g_l)$. Observe that at least one between $\nu^+(F)$ and $\nu^-(F)$ is strictly larger than 0. We prove the case in which $\nu^-(F) > 0$, the proof of the other case being analogous. There are two cases $\frac{\nu^+(F)}{\nu^-(F)} \leq 1$ and $\frac{\nu^+(F)}{\nu^-(F)} > 1$. In the first case, define (c_1, \dots, c_{l-1}) as

$$c_1 = \frac{\nu^-(g_1)}{\nu^-(F)} \text{ and } c_{i+1} = c_i + \frac{\nu^-(g_{i+1})}{\nu^-(F)} \text{ for all } i \in \{2, \dots, l-1\}$$

and (d_1, \dots, d_{m-1}) as

$$d_1 = \frac{\nu^+(f_1) \nu^+(F)}{\nu^-(F)} \text{ and } d_{i+1} = d_i + \frac{\nu^+(f_{i+1}) \nu^+(F)}{\nu^-(F)} \text{ for all } i \in \{2, \dots, m-1\}.$$

Let

$$\chi_1(x) = g_i \quad x \in [c_{i-1}, c_i), i \in \{1, \dots, l\}$$

and

$$\chi_2(x) = \begin{cases} f_i & x \in [d_{i-1}, d_i), i \in \{1, \dots, m\} \\ 0 & x \in [d_m, 1) \end{cases}.$$

We now show that the two-population endowment economy $\{(\chi_i, V_i, \Pi_i)\}_{i \in \{1,2\}}$ is interim Pareto improvable. Indeed, if let $(\chi'_1, \chi'_2) \in F^{[0,1]} \times F^{[0,1]}$ be given by

$$\chi'_1(x) = \chi_2(x) + V_1(\cdot, \chi_1(x)) - V_1(\cdot, \chi_2(x)) + \hat{f}(\cdot)$$

where

$$\hat{f}(\omega) = \int_F V_1(\omega, f) d\frac{\nu(f)}{\nu^-(F)} - \int_F V_2(\omega, f) d\frac{\nu(f)}{\nu^-(F)}.$$

Notice that by construction (χ'_1, χ'_2) satisfies market clearing, and it is an interim Pareto improvement if $\hat{f}(\omega) > 0$ for all $\omega \in \Omega$. But by equation (30), this is indeed the case. The second case is entirely symmetric and therefore omitted. ■

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E Online appendix: Omitted proofs

E.1 Omitted proofs of statements in the main text

Proof of Corollary 2. (i) \implies (ii) By Theorem 1 and Corollary 1, we have that $\mathbb{E}_p(f) = V_*(f) = V^*(f)$ for all $f \in \mathbb{R}^\Omega$. This immediately implies (ii) via Proposition 1.

(ii) \implies (i) Given that both V_* and V^* are SEU, there exist $p_*, p^* \in \Delta(\Omega)$ such that $V_*(f) = \mathbb{E}_{p_*}(f)$ and $V^*(f) = \mathbb{E}_{p^*}(f)$ for all $f \in \mathbb{R}^\Omega$. By Proposition 1, it follows that

$$\mathbb{E}_{p_*}(f) = V_*(f) \leq V^*(f) = \mathbb{E}_{p^*}(f) \quad \forall f \in \mathbb{R}^\Omega.$$

Therefore, we have that $p_* = p^*$, implying that $V_* = V^*$. By Theorem 1 and Corollary 1, it follows that there exists an ex-ante expectation \bar{V} such that (\bar{V}, V_i, Π_i) is a nonlinear conditional expectation for all $i \in I$ and such that $\bar{V} = \mathbb{E}_{p_*} = \mathbb{E}_{p^*}$, proving (i). \blacksquare

Proof of Proposition 2. 1. It is immediate to see that V^Θ is monotone and normalized provided that $\Theta \neq \emptyset$. Fix $i \in I$, $f \in \mathbb{R}^\Omega$, and $\mu \in \Theta$. For every $\omega \in \Omega$, we have

$$\mathbb{E}_{p_{\mu,i}(\omega, \cdot)}(f) \geq \min_{p \in \Delta} \{\mathbb{E}_p(f) + c_{i,\omega}(p)\} = V_i(\omega, f).$$

In particular, we have $\mathbb{E}_\mu(f) \geq \mathbb{E}_\mu(V_i(f))$. Given that μ was arbitrarily chosen, it follows that $V^\Theta(f) = \min_{\mu \in \Theta} \mathbb{E}_\mu(f) \geq \min_{\mu \in \Theta} \mathbb{E}_\mu(V_i(f)) = V^\Theta(V_i(f))$. Given that i and f were arbitrarily chosen, it follows that $V^\Theta \in \mathcal{V}^\circ$.

2. We first prove an ancillary claim.

Claim 1. *If there exists a nonlinear common ex-ante expectation \bar{V} for $\{(V_i, \Pi_i)\}_{i \in I}$, then there is no $(k_i)_{i \in I} \in \mathbb{R}_{++}^n$ and $(f_i)_{i \in I} \in (\mathbb{R}^\Omega)^n$ such that*

$$\begin{aligned} \min_{i \in I, \omega \in \Omega} \{k_i + f_i(\omega)\} &\geq 0, \\ \sum_{i \in I} f_i(\omega) &= 0 \quad \forall \omega \in \Omega, \\ \min_{i \in I, \omega \in \Omega} \{V_i(\omega, k_i e + f_i) - V_i(\omega, k_i e)\} &> 0. \end{aligned}$$

Proof. Suppose by contradiction that there exist $(k_i)_{i \in I} \in \mathbb{R}_{++}^n$ and $(f_i)_{i \in I} \in (\mathbb{R}^\Omega)^n$ as in the statement. By Lemma 13, \bar{V} is concave and therefore we have

$$\begin{aligned} \frac{1}{n} \sum_{i \in I} k_i &= \bar{V} \left(\frac{1}{n} \sum_{i \in I} k_i e + \frac{1}{n} \sum_{i \in I} f_i \right) \geq \frac{1}{n} \sum_{i \in I} \bar{V}(k_i e + f_i) = \frac{1}{n} \sum_{i \in I} \bar{V}(V_i(k_i e + f_i)) \\ &\geq \frac{1}{n} \sum_{i \in I} \bar{V} \left(V_i(k_i e) + \left(\min_{j \in I, \omega \in \Omega} \{V_j(\omega, k_j e + f_j) - V_j(\omega, k_j e)\} \right) e \right) \\ &= \frac{1}{n} \left[\sum_{i \in I} \bar{V}(V_i(k_i e)) \right] + \min_{j \in I, \omega \in \Omega} \{V_j(\omega, k_j e + f_j) - V_j(\omega, k_j e)\} \\ &= \frac{1}{n} \sum_{i \in I} k_i + \min_{j \in I, \omega \in \Omega} \{V_j(\omega, k_j e + f_j) - V_j(\omega, k_j e)\} > \frac{1}{n} \sum_{i \in I} k_i \end{aligned}$$

yielding a contradiction. \square

We are now ready to prove the statement. By the previous claim, [46, Theorem 6.2], and [4, Corollary 5] if there exists a nonlinear common ex-ante expectation \bar{V} for $\{(V_i, \Pi_i)\}_{i \in I}$, then

$$\bigcap_{i \in I} co \{p \in \Delta(\Omega) : \exists \omega \in \Omega, p \in \partial V_i(\omega, e)\} \neq \emptyset,$$

where $\partial V_i(\omega, e)$ denotes the superdifferential of the concave functional $V_i(\omega, \cdot)$ evaluated at $e \in \mathbb{R}^\Omega$. Finally, by [44, Lemma 32], we have that $\Theta \neq \emptyset$. \blacksquare

Proof of Lemma 1. By Lemma 10, there exists a unique class of indices Z , $\emptyset \neq Z \subseteq I \times \Omega$, that is closed and strongly connected with respect to $\underline{A}(S_1)$ and, in addition, every row of $\underline{A}(S_1)$ is not null. Given that S_1 is concave, it follows easily from the definition of $\partial S_1(0)$ that, for each $\hat{W} \in \partial S_1(0)$, Z is the unique closed and strongly connected class of indices with respect to $\underline{A}(\hat{W})$.³³ Fix $q \in Q$. By Lemma 17, $W^q \in \partial S_1(0)$, so that Z is the unique closed and strongly connected class of indices with respect to $\underline{A}(W^q)$. Next, observe that, for each $\gamma \in \Delta(I \times \Omega)$, we have $\gamma = \frac{1}{2}\gamma I + \frac{1}{2}\gamma W^q = \gamma \left(\frac{I+W^q}{2}\right)$ if and only if $\gamma = \gamma W^q$. In addition, given that $\underline{A}\left(\frac{I+W^q}{2}\right) \geq \underline{A}(W^q)$, it follows by [6, Corollaries 8.1 and 8.2] and [8, Theorem 2.2.5] that there exists a unique $\gamma^q \in \Delta(I \times \Omega)$ such that $\gamma^q = \gamma^q \left(\frac{I+W^q}{2}\right)$. By the previous claim, γ^q is also the unique probability vector such that $\gamma^q = \gamma^q W^q$. Given that $q \in Q$ was arbitrarily chosen, the statement follows. \blacksquare

Proof of Corollary 3. The first part of the statement follows from Theorem 2 and from the fact that, by assumption, $Q = \{q^*\}$. Next, assume that $\Theta \neq \emptyset$. Observe that, by Lemma 12, for each $\mu \in \Theta$, we have that $p_{\mu,i}(\omega, \cdot) = p_{\mu,i}(\omega', \cdot)$ and $c_{i,\omega}(p_{\mu,i}(\omega, \cdot)) = 0$ for all $(i, \omega) \in I \times \Omega$ and $\omega' \in \Pi_i(\omega)$, so that $p_{\mu,i}(\omega, \cdot) = q_{i,\omega}^*$ for all $(i, \omega) \in I \times \Omega$. Assume by contradiction that there exist $\mu, \mu' \in \Theta$ with $\mu \neq \mu'$ and consider the collection $\{(\mathbb{E}_{q_i^*}, \Pi_i)\}_{i \in I}$ of interim expectations. This collection has full support by Lemma 12. Therefore, $\{(\mathbb{E}_{q_i^*}, \Pi_i)\}_{i \in I}$ exhibits convergence to a deterministic limit by Theorem 1. In particular, both \mathbb{E}_μ and $\mathbb{E}_{\mu'}$ are common ex-ante expectations for $\{(\mathbb{E}_{q_i^*}, \Pi_i)\}_{i \in I}$ by construction, yielding a contradiction with Corollary 1. Therefore, we obtain $\Theta = \{\mu^*\}$ for some $\mu^* \in \Delta(\Omega)$. Moreover, by Lemma 1, there exists a unique probability vector $\gamma^{q^*} \in \Delta(I \times \Omega)$ such that $\gamma^{q^*} = \gamma^{q^*} W^{q^*}$. Now, for each $(i, \omega) \in I \times \Omega$, define $\gamma^{\mu^*} \in \Delta(I \times \Omega)$ as $\gamma_{i,\omega}^{\mu^*} = s_i \mu^*(\omega)$ and observe that

$$\begin{aligned} \sum_{(j,\omega') \in I \times \Omega} \gamma_{i,\omega}^{\mu^*} w_{(j,\omega')(i,\omega)}^{q^*} &= \sum_{(j,\omega') \in I \times \Omega} s_j \mu^*(\omega') w_{ji} q_{j,\omega'}^*(\omega) = \sum_{j \in I} s_j w_{ji} \sum_{\omega' \in \Omega} \mu^*(\omega') p_{\mu^*,j}(\omega', \omega) \\ &= \mu^*(\omega) \sum_{j \in I} s_j w_{ji} = \mu^*(\omega) s_i = \gamma_{i,\omega}^{\mu^*}. \end{aligned}$$

This show that $\gamma^{\mu^*} = \gamma^{\mu^*} W^{q^*}$, proving that $\gamma^{q^*} = \gamma^{\mu^*}$. Finally, we have

$$\sum_{(i,\omega) \in I \times \Omega} \gamma_{i,\omega}^{q^*} \mathbb{E}_{q_{i,\omega}^*}(\hat{f}) = \sum_{(i,\omega) \in I \times \Omega} \gamma_{i,\omega}^{\mu^*} \mathbb{E}_{q_{i,\omega}^*}(\hat{f}) = \sum_{(i,\omega) \in I \times \Omega} s_i \mu^*(\omega) \mathbb{E}_{p_{\mu^*,i}(\omega, \cdot)}(\hat{f}) = \mathbb{E}_{\mu^*}(\hat{f}),$$

proving the second part of the statement. \blacksquare

³³Here, with an abuse of notation we identify the linear operator induced by the matrix \hat{W} with \hat{W} itself.

Proof of Corollary 4. By Lemma 13, we have that V_* is a maxmin ex-ante expectation. By Theorem 1 and Corollary 1, it follows that \bar{V} is a maxmin ex-ante expectation as well. Let $\bar{C} \subseteq \Delta(\Omega)$ denote the set of probabilities such that $\bar{V}(f) = \min_{\mu \in \bar{C}} \mathbb{E}_\mu(f)$ for all $f \in \mathbb{R}^\Omega$. Fix $\mu \in \bar{C}$, $i \in I$, and $\omega \in \Omega$. Since \bar{V} and V_i satisfy dynamic consistency, it follows by [16] that \bar{C} is Π_i -rectangular. With this, we have that $p_{\mu,i}(\omega, \cdot) \in C_{i,\omega}$ where $C_{i,\omega}$ is the set of probabilities such that $V_i(\omega, f) = \min_{p \in C_{i,\omega}} \mathbb{E}_p(f)$ for all $f \in \mathbb{R}^\Omega$. Hence, it follows that $c_{i,\omega}(p_{\mu,i}(\omega, \cdot)) = 0$. Given that $i \in I$ and $\omega \in \Omega$ were arbitrarily chosen, it follows that $\mu \in \Theta$. This in turn proves that $\bar{C} \subseteq \Theta$, hence that $\bar{V}(\hat{f}) \leq V^\Theta(\hat{f})$. Finally, the result follows by the second part of Theorem 2. \blacksquare

E.2 Omitted proofs of statements in the main appendix

We start with an ancillary result that elaborates on how the indicator matrix of the composition of a finite collection of operators $\{T_h\}_{h \in \{1, \dots, H\}}$ is related to the product of their indicator matrices.

Lemma 15. *Let $S, T : \mathbb{R}^{\bar{n}} \rightarrow \mathbb{R}^{\bar{n}}$ be monotone and define $\hat{A} = A(T \circ S)$, $\tilde{A} = A(S)$, and $A = A(T)$. If there exists $k \in J$ such that $a_{jk} > 0$ and $\tilde{a}_{kj'} > 0$, then $\hat{a}_{jj'} > 0$. In particular, we have that:*

1. *If $\{T_h\}_{h \in \{1, \dots, H\}}$ is a collection of monotone operators from $\mathbb{R}^{\bar{n}}$ to $\mathbb{R}^{\bar{n}}$ and the jj' -th entry of $\Pi_{h=1}^H A(T_h)$ is strictly positive, then the jj' -th of $A(T_H \circ \dots \circ T_1)$ is strictly positive.*
2. *If $t \in \mathbb{N}$ and the jj' -th entry of $A(T)^t$ is strictly positive, then the jj' -th of $A(T^t)$ is strictly positive.*

Proof. By assumption, there exists $k \in \{1, \dots, \bar{n}\}$ such that $a_{jk}, \tilde{a}_{kj'} > 0$, that is, there exist $\varepsilon_{jk}, \varepsilon_{kj'} \in (0, 1)$ such that for each $f \in \mathbb{R}^{\bar{n}}$ and for each $\delta \geq 0$

$$S_k(f + \delta e^{j'}) - S_k(f) \geq \varepsilon_{kj'} \delta \text{ and } T_j(f + \delta e^k) - T_j(f) \geq \varepsilon_{jk} \delta.$$

Since S is monotone, this implies that $S(f + \delta e^{j'}) \geq S(f) + \varepsilon_{kj'} \delta e^k$ for all $f \in \mathbb{R}^{\bar{n}}$ and for all $\delta \geq 0$. Since T is monotone, this yields that for each $f \in \mathbb{R}^{\bar{n}}$ and for each $\delta \geq 0$

$$T_j(S(f + \delta e^{j'})) \geq T_j(S(f) + \varepsilon_{kj'} \delta e^k) \geq T_j(S(f)) + \varepsilon_{jk} \varepsilon_{kj'} \delta.$$

Since $\varepsilon_{jk} \varepsilon_{kj'} \in (0, 1)$, this proves that, under $T \circ S$, j is strongly monotone with respect to j' , proving that $\hat{a}_{jj'} > 0$ and the main part of the statement.

1. Consider a collection of H monotone operators from $\mathbb{R}^{\bar{n}}$ to $\mathbb{R}^{\bar{n}}$: $\{T_h\}_{h \in \{1, \dots, H\}}$. We prove by finite induction the statement that, for each $l \in \{1, \dots, H\}$, if the jj' -th entry of $\Pi_{h=1}^l A(T_h)$ is strictly positive, then the jj' -th of $A(T_l \circ \dots \circ T_1)$ is strictly positive.

Initial step. Assume $l = 1$. In this case, we have that $A(T_1) = \Pi_{h=1}^1 A(T_h)$. This proves that if the jj' -th entry of $\Pi_{h=1}^1 A(T_h)$ is strictly positive, so is the jj' -th entry of the indicator matrix of the composition.

Inductive step. Assume the statement is true for l . We prove it is true for $l+1$. Define $S = T_l \circ \dots \circ T_1$ and $T = T_{l+1}$. As before, set $\tilde{A} = A(S)$, $A = A(T)$, and $\hat{A} = A(T \circ S) = A(T_{l+1} \circ \dots \circ T_1)$.

Finally, define by $a_{jj'}^{(l)}$ (resp., $a_{jj'}^{(1)}$ and $a_{jj'}^{(l+1)}$) the generic jj' -th entry of $\Pi_{h=1}^l A(T_h)$ (resp., $A(T_{l+1})$ and $\Pi_{h=1}^{l+1} A(T_h)$). Observe that

$$a_{jj'}^{(l+1)} = \sum_{k=1}^{\bar{n}} a_{jk}^{(1)} a_{kj'}^{(l)}.$$

If the jj' -th entry of $\Pi_{h=1}^{l+1} A(T_h)$ is strictly positive, then $a_{jj'}^{(l+1)} > 0$, yielding that $a_{jk}^{(1)} a_{kj'}^{(l)} > 0$ for some $k \in J$, that is, $a_{jk}^{(1)}, a_{kj'}^{(l)} > 0$ for some $k \in J$. By inductive hypothesis, we have that $a_{kj'}^{(l)} > 0$ implies that $\tilde{a}_{kj'} > 0$ as well as $a_{jk} > 0$. By the main part of the statement, we can conclude that $\hat{a}_{jj'} > 0$, proving the inductive step.

The statement follows by finite induction.

2. By point 1, the statement trivially follows by considering the collection $\{T_h\}_{h=1}^H$ where $H = t$ and $T_h = T$ for all $h \in \{1, \dots, H\}$. ■

Proof of Lemma 2. Define $B = \Pi_{k=1}^K B_k$. By induction, we prove that $A(\Pi_{k=1}^m B_k) \geq A(B_k) \geq I_{\bar{n}}$ for all $k \in \{1, \dots, m\}$ and for all $m \in \{1, \dots, K\}$. By definition and since $b_{1,jj} > 0$ for all $j \in J$, if $m = 1$, then $A(\Pi_{k=1}^1 B_k) = A(B_1) \geq I_{\bar{n}}$. By point 1 of Lemma 15 and inductive hypothesis and since $b_{k,jj} > 0$ for all $k \in \{1, \dots, K\}$ and for all $j \in J$, if $m, m+1 \in \{1, \dots, K\}$, then $A(B_{m+1}) A(\Pi_{k=1}^m B_k) \geq I_{\bar{n}} A(B_k)$ and $A(\Pi_{k=1}^{m+1} B_k) = A(B_{m+1} \Pi_{k=1}^m B_k) \geq A(A(B_{m+1}) A(\Pi_{k=1}^m B_k)) \geq A(I_{\bar{n}} A(B_k)) = A(B_k) \geq I_{\bar{n}}$ for all $k \in \{1, \dots, m\}$. By point 1 of Lemma 15 and inductive hypothesis, we also have that $A(B_{m+1}) A(\Pi_{k=1}^m B_k) \geq A(B_{m+1}) I_{\bar{n}}$ and $A(\Pi_{k=1}^{m+1} B_k) = A(B_{m+1} \Pi_{k=1}^m B_k) \geq A(A(B_{m+1}) A(\Pi_{k=1}^m B_k)) \geq A(A(B_{m+1}) I_{\bar{n}}) = A(B_{m+1}) \geq I_{\bar{n}}$. The statement follows by finite induction. In particular, this yields that

$$A(B_K \dots B_1) \geq A(B_k) \geq I_{\bar{n}} \quad \forall k \in \{1, \dots, K\}.$$

Consider $k \in \{1, \dots, K\}$. Since $A(B_k)$ is symmetric, any index $j \in J$ is essential under B_k . Let $l \in \{1, \dots, m_{B_k}\}$ and $j \in J_l(B_k)$. We have two cases:

1. $j \in J_{l'}(A(B))$ for some $l' \in \{1, \dots, m_{A(B)}\}$. Consider $j' \in J_l(B_k)$. It follows that $j \xleftrightarrow{B_k} j'$. Since $A(B) \geq A(B_k)$, we have that $j \xleftrightarrow{A(B)} j'$, yielding that $j' \in J_{l'}(A(B))$. This implies that $J_l(B_k) \subseteq J_{l'}(A(B))$.
2. $j \in J_{m_{B+1}}(A(B))$. Consider $j' \in J_l(B_k)$. It follows that $j \xleftrightarrow{B_k} j'$. Since $A(B) \geq A(B_k)$, we have that $j \xleftrightarrow{A(B)} j'$, yielding that $j' \in J_{m_{B+1}}(A(B))$. Otherwise, since $j \xleftrightarrow{A(B)} j'$, if $j' \notin J_{m_{B+1}}(A(B))$, then j' would be essential under $A(B)$ and so would be j , a contradiction. This implies that $J_l(B_k) \subseteq J_{m_{B+1}}(A(B))$. ■

Proof of Lemma 3. Before starting, we denote by $\langle \cdot, \cdot \rangle$ the inner product of $\mathbb{R}^{\bar{n}}$. Let $j \in J$. Define the binary relation \succsim_j^* on \mathbb{R}^{Ω} by $f \succsim_j^* g$ if and only if $T_j(\lambda f + (1-\lambda)h) \geq T_j(\lambda g + (1-\lambda)h)$ for all $\lambda \in (0, 1]$ and $h \in \mathbb{R}^{\bar{n}}$. By [3] and since T_j is normalized, monotone, and continuous, we have that there exists a compact and convex set C_j of $\Delta_{\bar{n}}$ such that

$$f \succsim_j^* g \iff \langle f, p \rangle \geq \langle g, p \rangle \quad \forall p \in C_j \quad (31)$$

and

$$T_j(f) = \alpha_j(f) \min_{p \in C_j} \langle f, p \rangle + (1 - \alpha_j(f)) \max_{p \in C_j} \langle f, p \rangle \quad \forall f \in \mathbb{R}^{\bar{n}} \quad (32)$$

where $\alpha_j : \mathbb{R}^{\bar{n}} \rightarrow [0, 1]$ and $\Delta_{\bar{n}}$ is the collection of all vectors in $\mathbb{R}_+^{\bar{n}}$ whose entries sum up to 1. Observe also that if j is constant with respect to j' , then $e^{j'} \sim_j^* 0$. By (31), it follows that

$$p_{j'} = 0 \quad \forall p \in C_j. \quad (33)$$

Since C_j is compact, for each $f \in \mathbb{R}^{\bar{n}}$ define $p_{\min, f}, p_{\max, f} \in C_j$ such that $\langle f, p_{\min, f} \rangle = \min_{p \in C_j} \langle f, p \rangle$ and $\langle f, p_{\max, f} \rangle = \max_{p \in C_j} \langle f, p \rangle$. By (32) and since C_j is convex, it follows that $p_{j, f} = \alpha_j(f) p_{\min, f} + (1 - \alpha_j(f)) p_{\max, f} \in C_j$ such that $T_j(f) = \langle f, p_{j, f} \rangle$ for all $f \in \mathbb{R}^{\bar{n}}$. Fix $f \in \mathbb{R}^{\bar{n}}$. Since j was arbitrarily chosen, define $M(f)$ to be the matrix whose j -th row entries correspond to the entries of $p_{j, f}$. It follows that $T(f) = M(f)f$. Moreover, $M(f)$ belongs to the set $\mathcal{M}(T)$ of matrices M whose j -th row belongs to C_j . Since each of these sets is compact and convex, so is $\mathcal{M}(T)$. Since f was arbitrarily chosen, the statement follows. By construction of $\mathcal{M}(T)$ and (33), it follows that if j is constant with respect to j' , then $m_{jj'} = 0$ for all $M \in \mathcal{M}(T)$. \blacksquare

Proof of Lemma 4. Since $d(\bar{M}) > 0$, it follows that $\bar{m}_{jj} > 0$ for all $j \in J$. This implies that the jj -th entry of $A(\bar{M})$ is 1 for all $j \in J$, and, in particular, if the jj' -th entry of $A(M)$ is strictly positive, so is the one of $A(\bar{M})A(M)$. By point 1 of Lemma 15, we can conclude that $A(\bar{M}M) \geq A(M)$. We have two cases:

1. $A(\bar{M}M) = A(M)$. Set $\hat{M} = \bar{M}M$ and consider $\hat{m}_{jj'} > 0$. We next prove that for each $l \in \{1, \dots, \bar{n}\}$

$$m_{lj'} = 0 \implies \bar{m}_{jl} = 0. \quad (34)$$

By contradiction, assume that there exists $\bar{l} \in \{1, \dots, \bar{n}\}$ such that $m_{\bar{l}j'} = 0$ and $\bar{m}_{j\bar{l}} > 0$. Since $A(\hat{M}) = A(\bar{M}M) = A(M)$ and $\hat{m}_{jj'} > 0$ and $m_{\bar{l}j'} = 0$, we would have that $m_{jj'} > 0$ and $\hat{m}_{\bar{l}j'} = 0$. Since $A(\bar{M})$ is symmetric, we would also have that $\bar{m}_{\bar{l}j} > 0$, yielding that $\hat{m}_{\bar{l}j'} \geq \bar{m}_{\bar{l}j} m_{jj'} > 0$, a contradiction with $\hat{m}_{\bar{l}j'} = 0$. By (34), we can conclude that $\hat{m}_{jj'} = \sum_{l=1}^{\bar{n}} \bar{m}_{jl} m_{lj'} \geq \sum_{l=1}^{\bar{n}} \bar{m}_{jl} \delta(M) = \delta(M)$, proving the statement.

2. $A(\bar{M}M) > A(M)$. Set $\hat{M} = \bar{M}M$. In this case, if $\hat{m}_{jj'} > 0$, then $\bar{m}_{j\bar{l}} m_{\bar{l}j'} > 0$ for some $\bar{l} \in \{1, \dots, \bar{n}\}$ and, in particular, $\bar{m}_{j\bar{l}}, m_{\bar{l}j'} > 0$. It follows that $\hat{m}_{jj'} = \sum_{l=1}^{\bar{n}} \bar{m}_{jl} m_{lj'} \geq \bar{m}_{j\bar{l}} m_{\bar{l}j'} \geq \delta(\bar{M}) \delta(M)$, proving the statement.

Consider a sequence $\{M_k\}_{k=1}^{\infty}$ of $\bar{n} \times \bar{n}$ stochastic matrices such that $A(M_k)$ is symmetric, $\delta(M_k) \geq \delta > 0$, and $d(M_k) > 0$ for all $k \in \mathbb{N}$. By induction and the previous part, we have that $A\left(\prod_{k=1}^{m+1} M_k\right) = A\left(M_{m+1} \prod_{k=1}^m M_k\right) \geq A\left(\prod_{k=1}^m M_k\right)$ for all $m \in \mathbb{N}$. Define $f : \mathbb{N} \rightarrow \{0, 1\}$ by $f(1) = 1$ and

$$f(m+1) = \begin{cases} 1 & \text{if } A\left(\prod_{k=1}^{m+1} M_k\right) > A\left(\prod_{k=1}^m M_k\right) \\ 0 & \text{if } A\left(\prod_{k=1}^{m+1} M_k\right) = A\left(\prod_{k=1}^m M_k\right) \end{cases} \quad \forall m \in \mathbb{N}.$$

By induction, we prove that

$$\delta \left(\prod_{k=1}^m M_k \right) \geq \delta^{\sum_{k=1}^m f(k)} \quad \forall m \in \mathbb{N}. \quad (35)$$

Initial step. Assume $m = 1$. Since $f(1) = 1$, $\delta \left(\prod_{k=1}^m M_k \right) = \delta(M_1) \geq \delta = \delta^{\sum_{k=1}^m f(k)}$.

Inductive step. Assume the statement is true for $m \in \mathbb{N}$. We prove it is true for $m + 1$. Since $A \left(\prod_{k=1}^{m+1} M_k \right) \geq A \left(\prod_{k=1}^m M_k \right)$, we have two cases:

1. $A \left(\prod_{k=1}^{m+1} M_k \right) > A \left(\prod_{k=1}^m M_k \right)$. In this case, we have that $f(m+1) = 1$. By the first part of the statement and inductive hypothesis, we have that

$$\delta \left(\prod_{k=1}^{m+1} M_k \right) = \delta \left(M_{m+1} \prod_{k=1}^m M_k \right) \geq \delta(M_{m+1}) \delta \left(\prod_{k=1}^m M_k \right) \geq \delta \delta^{\sum_{k=1}^m f(k)} = \delta^{\sum_{k=1}^{m+1} f(k)}.$$

2. $A \left(\prod_{k=1}^{m+1} M_k \right) = A \left(\prod_{k=1}^m M_k \right)$. In this case, we have that $f(m+1) = 0$. By the first part of the statement and inductive hypothesis, we have that

$$\delta \left(\prod_{k=1}^{m+1} M_k \right) = \delta \left(M_{m+1} \prod_{k=1}^m M_k \right) \geq \delta \left(\prod_{k=1}^m M_k \right) \geq \delta^{\sum_{k=1}^m f(k)} = \delta^{\sum_{k=1}^{m+1} f(k)}.$$

Thus, (35) follows by induction. Since $\left\{ A \left(\prod_{k=1}^m M_k \right) \right\}_{m \in \mathbb{N}}$ is an increasing sequence with upper bound the $\bar{n} \times \bar{n}$ square matrix whose entries are all 1s, we observe that $f(k) = 1$ for at most \bar{n}^2 indices, yielding that $\sum_{k=1}^m f(k) \leq \bar{n}^2$ for all $m \in \mathbb{N}$, proving (24). \blacksquare

Proof of Lemma 5. Consider $k \in \mathbb{R}$ and a sequence of functions $\{f_m\}_{m \in \mathbb{N}} \subseteq \mathbb{R}^\Omega$ such that $f_m \rightarrow k1_\Omega$. Since $f_m \rightarrow k1_\Omega$ and Ω is finite, we have that $\lim_{m \rightarrow \infty} \min_{\omega \in \Omega} f_m(\omega) = k = \lim_{m \rightarrow \infty} \max_{\omega \in \Omega} f_m(\omega)$. Since \bar{V} is normalized and monotone, we also have that $\min_{\omega \in \Omega} f_m(\omega) \leq \bar{V}(f_m) \leq \max_{\omega \in \Omega} f_m(\omega)$ for all $m \in \mathbb{N}$. By passing to the limit and since \bar{V} is normalized, we have that

$$\lim_{m \rightarrow \infty} \bar{V}(f_m) = k = \bar{V}(k1_\Omega),$$

proving continuity at $k1_\Omega$. \blacksquare

Proof of Lemma 6. (i) implies (ii). Let $j, j' \in J$. Since $a_{jj'} = 1$, we have that j is strongly monotone with respect to j' . By contradiction, assume that $\Pi(\omega_j) \neq \Pi(\omega_{j'})$. Since Π is a

partition, it follows that $\Pi(\omega_j) \cap \Pi(\omega_{j'}) = \emptyset$. Since (V, Π) is an interim expectation and j is strongly monotone with respect to j' , we thus have that there exists $\varepsilon_{jj'} \in (0, 1)$ such that

$$\begin{aligned} 0 &= V\left(\omega_j, 01_{\Pi(\omega_j)} + 1_{\{\omega_{j'}\}}1_{\Pi(\omega_j)^c}\right) - V(\omega_j, 0) \\ &= V\left(\omega_j, 1_{\{\omega_{j'}\}}1_{\Pi(\omega_j)} + 1_{\{\omega_{j'}\}}1_{\Pi(\omega_j)^c}\right) - V(\omega_j, 0) = V\left(\omega_j, 1_{\{\omega_{j'}\}}\right) - V(\omega_j, 0) \geq \varepsilon_{jj'} > 0, \end{aligned}$$

a contradiction.

(ii) implies (i). Note that $\Pi(\omega_j) = \Pi(\omega_{j'})$ only if $\omega_{j'} \in \Pi(\omega_j)$. Since (V, Π) is an interim expectation with full support, we have that each $\bar{\omega} \in \Pi(\omega_j)$ is $V(\omega_j, \cdot)$ -essential and, in particular, so is $\omega_{j'}$, yielding that $a_{jj'} = 1$.

By the previous part of the proof and since $\Pi(\omega_j) = \Pi(\omega_j)$ for all $j \in J$ and $A(V)$ is $\{0, 1\}$ -valued, we thus have that both $a_{jj'} = 1$ and $a_{j'j} = 1$ hold if and only if $\Pi(\omega_j) = \Pi(\omega_{j'})$, proving that $A(V)$ is symmetric, $a_{jj} = 1$ for all $j \in J$, and $\Pi(V) = \Pi$. Finally, for all $j, j' \in J$, if j is not strongly monotone with respect to j' , we can conclude that $a_{jj'} = 0$ and $\omega_{j'} \notin \Pi(\omega_j)$. Since $V(\omega, f1_{\Pi(\omega)} + h1_{\Pi(\omega)^c}) = V(\omega, f)$ for all $\omega \in \Omega$ and for all $f, h \in \mathbb{R}^\Omega$, this implies that

$$\begin{aligned} V\left(\omega_j, f + \delta 1_{\{\omega_{j'}\}}\right) &= V\left(\omega_j, f1_{\Pi(\omega_j)} + \delta 1_{\{\omega_{j'}\}}1_{\Pi(\omega_j)^c} + 01_{\Pi(\omega_j)^c}\right) \\ &= V\left(\omega_j, f1_{\Pi(\omega_j)} + 01_{\Pi(\omega_j)^c}\right) = V(\omega_j, f) \end{aligned}$$

for all $f \in \mathbb{R}^\Omega$ and for all $\delta \geq 0$, yielding that j is constant with respect to j' . This implies that V is dichotomic. \blacksquare

Proof of Lemma 7. Define

$$V_\circ(f) = \min_{\omega \in \Omega} f(\omega) \quad \text{and} \quad V^\circ(f) = \max_{\omega \in \Omega} f(\omega) \quad \forall f \in \mathbb{R}^\Omega.$$

It is immediate to see that both V_\circ and V° are ex-ante expectations. Next, fix $f \in \mathbb{R}^\Omega$, and observe that given

$$V_i(\omega, f) \in \left[\min_{\omega' \in \Omega} f(\omega'), \max_{\omega' \in \Omega} f(\omega') \right] \quad \forall \omega \in \Omega, \forall i \in I,$$

we have that

$$V_\circ(V_i(f)) = \min_{\omega \in \Omega} V_i(\omega, f) \geq \min_{\omega' \in \Omega} f(\omega') = V_\circ(f) \quad \forall i \in I$$

and

$$V^\circ(V_i(f)) = \max_{\omega \in \Omega} V_i(\omega, f) \leq \max_{\omega' \in \Omega} f(\omega') = V^\circ(f) \quad \forall i \in I.$$

This proves that V_\circ and V° are respectively lower and upper common ex-ante expectations for $\{(V_i, \Pi_i)\}_{i \in I}$, hence that \mathcal{V}_\circ and \mathcal{V}° are nonempty. We next show that V_* and V^* are well defined lower and upper common ex-ante expectations for $\{(V_i, \Pi_i)\}_{i \in I}$. First, observe that

$$V_*(ke) = \sup_{V_\circ \in \mathcal{V}_\circ} V_\circ(ke) = \sup_{V_\circ \in \mathcal{V}_\circ} k = k \quad \forall k \in \mathbb{R}$$

and that, for all $f, g \in \mathbb{R}^\Omega$ with $f \geq g$, we have

$$V_*(f) = \sup_{V_\circ \in \mathcal{V}_\circ} V_\circ(f) \geq \sup_{V_\circ \in \mathcal{V}_\circ} V_\circ(g) = V_*(g),$$

where the inequality follows from monotonicity of each $V_o \in \mathcal{V}_o$. With this, V_* is an ex-ante expectation. With exactly the same steps we can show that V^* is also an ex-ante expectation. Next, fix $f \in \mathbb{R}^\Omega$ and $V_o \in \mathcal{V}_o$. For each $i \in I$, we have

$$V_o(f) \leq V_o(V_i(f)) \leq \sup_{V'_o \in \mathcal{V}_o} V'_o(V_i(f)) = V_*(V_i(f)).$$

Given that $V_o \in \mathcal{V}_o$ was arbitrarily chosen, it follows that

$$V_*(f) = \sup_{V_o \in \mathcal{V}_o} V_o(f) \leq V_*(V_i(f))$$

proving that V_* is a lower common ex-ante expectation. With exactly the same steps we can show that V^* is also an upper common ex-ante expectation. \blacksquare

Proof of Lemma 8. Given that $\{(V_i, \Pi_i)\}_{i \in I}$ is a variational collection of interim expectations, it follows that $V_i(\omega, \cdot)$ is concave and translation invariant for all $i \in I$ and for all $\omega \in \Omega$. Therefore, by [11, p. 346], we have that

$$\begin{aligned} \|S_{\beta,i}(\mathbf{f}) - S_{\beta,i}(\mathbf{g})\|_\infty &= \left\| V_i \left((1-\beta) \hat{f} + \beta \sum_{l=1}^n w_{il} f_l \right) - V_i \left((1-\beta) \hat{f} + \beta \sum_{l=1}^n w_{il} g_l \right) \right\|_\infty \\ &\leq \left\| (1-\beta) \hat{f} + \beta \sum_{l=1}^n w_{il} f_l - (1-\beta) \hat{f} - \beta \sum_{l=1}^n w_{il} g_l \right\|_\infty \\ &= \left\| \beta \sum_{l=1}^n w_{il} (f_l - g_l) \right\|_\infty \leq \beta \sum_{l=1}^n w_{il} \|f_l - g_l\|_\infty \\ &\leq \beta \sum_{l=1}^n w_{il} \|\mathbf{f} - \mathbf{g}\|_* \leq \beta \|\mathbf{f} - \mathbf{g}\|_* \quad \forall i \in I, \forall \mathbf{f}, \mathbf{g} \in (\mathbb{R}^\Omega)^n, \end{aligned}$$

proving that $\|S_\beta(\mathbf{f}) - S_\beta(\mathbf{g})\|_* = \sup_{i \in I} \|S_{\beta,i}(\mathbf{f}) - S_{\beta,i}(\mathbf{g})\|_\infty \leq \beta \|\mathbf{f} - \mathbf{g}\|_*$ for all $\mathbf{f}, \mathbf{g} \in (\mathbb{R}^\Omega)^n$.

By the Banach contraction principle, for each $\beta \in (0, 1)$ we have that $S_\beta^\tau(\hat{\mathbf{f}}) \xrightarrow{\|\cdot\|_*} \sigma^\beta$ as well as $S_\beta(\sigma^\beta) = \sigma^\beta$ where σ^β is the unique fixed point of S_β for all $\beta \in (0, 1)$. Finally, by [11, p. 346] and since V_i is normalized, we have that

$$\begin{aligned} \|S_{\beta,i}(\mathbf{f})\|_\infty &= \left\| V_i \left((1-\beta) \hat{f} + \beta \sum_{l=1}^n w_{il} f_l \right) \right\|_\infty \leq \left\| (1-\beta) \hat{f} + \beta \sum_{l=1}^n w_{il} f_l \right\|_\infty \\ &\leq (1-\beta) \|\hat{f}\|_\infty + \beta \sum_{l=1}^n w_{il} \|f_l\|_\infty \quad \forall i \in I, \forall \mathbf{f} \in (\mathbb{R}^\Omega)^n. \end{aligned}$$

By induction, this implies that $\|S_\beta^\tau(\hat{\mathbf{f}})\|_* \leq \|\hat{f}\|_\infty$ for all $\tau \in \mathbb{N}$. By passing to the limit, the statement follows. \blacksquare

Proof of Lemma 9. (i) implies (ii). Before starting, since Ω is finite, we enumerate its elements $\Omega = \{\omega_1, \dots, \omega_{\bar{n}}\}$ and set as before $J = \{1, \dots, \bar{n}\}$. By assumption, we have that $f_i = V_i(\sum_{l=1}^n w_{il} f_l)$ for all $i \in I$. By Proposition 6 and Lemma 6, for each $i \in I$ there exists an $\bar{n} \times \bar{n}$

stochastic matrix M_i whose diagonal is strictly positive and it is such that: 1) $A(V_i) = A(M_i)$ is symmetric, 2) $\Pi(M_i) = \Pi_i$, and 3) $V_i(\sum_{l=1}^n w_{il}f_l) = M_i(\sum_{l=1}^n w_{il}f_l) = \sum_{l=1}^n w_{il}M_i f_l$. It follows that \mathbf{f} is also a fixed point of the operator $\tilde{S} : (\mathbb{R}^\Omega)^n \rightarrow (\mathbb{R}^\Omega)^n$ where

$$\tilde{S}_i(\mathbf{g}) = \sum_{l=1}^n w_{il}M_i g_l \quad \forall i \in I.$$

We next show that $\tilde{S}(\mathbf{f}) = \mathbf{f}$ only if there exists $m \in \mathbb{R}$ such that $f_i = f_{i'} = m1_\Omega$ for all $i, i' \in I$. By contradiction, assume that there exist $\bar{i}, \bar{i}' \in I$ and $\omega_{\bar{j}}, \omega_{\bar{j}'} \in \Omega$ such that $f_{\bar{i}}(\omega_{\bar{j}}) = \max_{i \in I} \max_{j \in J} f_i(\omega_j) > \min_{i \in I} \min_{j \in J} f_i(\omega) = f_{\bar{i}'}(\omega_{\bar{j}'}).$ We begin with an observation. For each $t \in \mathbb{N}$ denote by I^t the set of (finite) sequences in I with t elements, that is, $\mathbf{i} \in I^t$ if and only if $\mathbf{i} = (i_1, \dots, i_t)$ with $i_l \in I$ for all $l \in \{1, \dots, t\}$. By induction, note that for each $t \in \mathbb{N}$

$$\tilde{S}_i^t(\mathbf{g}) = \sum_{\mathbf{i} \in I^{t+1}: i_1=i} w_{i_1 i_2} \dots w_{i_t i_{t+1}} M_{i_1} \dots M_{i_t} g_{i_{t+1}} \quad \forall \mathbf{g} \in (\mathbb{R}^\Omega)^n$$

and

$$w_{i_1 i_2} \dots w_{i_t i_{t+1}} \geq 0 \text{ for all } \mathbf{i} \in I^{t+1} \text{ such that } i_1 = i \text{ and } \sum_{\mathbf{i} \in I^{t+1}: i_1=i} w_{i_1 i_2} \dots w_{i_t i_{t+1}} = 1.$$

Since W is strongly connected, there exists a sequence of agents $(\bar{i}_1, \dots, \bar{i}_{\bar{t}+1})$ such that $\bar{t} \in \mathbb{N}$, $\{\bar{i}_1, \dots, \bar{i}_{\bar{t}}\} \supseteq I$, and $\bar{i}_1 = \bar{i}_{\bar{t}+1} = \bar{i}$ with $w_{\bar{i}_l \bar{i}_{l+1}} > 0$ for all $l \in \{1, \dots, \bar{t}\}$. By Lemma 2 and since $\{\bar{i}_1, \dots, \bar{i}_{\bar{t}}\} \supseteq I$, we have that $\Pi(A(M_{\bar{i}_1} \dots M_{\bar{i}_{\bar{t}}}))$ is coarser than $\Pi(M_i) = \Pi_i$ for all $i \in I$. Since $\Pi_{\text{sup}} = \{\Omega\}$, we can conclude that $\Pi(A(M_{\bar{i}_1} \dots M_{\bar{i}_{\bar{t}}})) = \{\Omega\}$, yielding that $M_{\bar{i}_1} \dots M_{\bar{i}_{\bar{t}}}$ is strongly connected. By Lemma 2 and since the diagonal of each $M_{\bar{i}_l}$ is strictly positive, we also have that $M_{\bar{i}_1} \dots M_{\bar{i}_{\bar{t}}}$ has a strictly positive diagonal. This implies that $M_{\bar{i}_1} \dots M_{\bar{i}_{\bar{t}}}$ is primitive, that is, there exists $\tau \in \mathbb{N}$ such that each entry of $(M_{\bar{i}_1} \dots M_{\bar{i}_{\bar{t}}})^\tau$ is strictly positive. Since W is strongly connected there exists a sequence of agents $(\hat{i}_1, \dots, \hat{i}_{\hat{t}+1})$ such that $\hat{t} \in \mathbb{N}$, $\hat{i}_1 = \bar{i}$, and $\hat{i}_{\hat{t}+1} = \bar{i}'$ with $w_{\hat{i}_l \hat{i}_{l+1}} > 0$ for all $l \in \{1, \dots, \hat{t}\}$. Next, recall that by Euclid's algorithm for each $l \in \{1, \dots, \tau\bar{t} + 1\}$ there exists unique $q_l \in \mathbb{N}_0$ and $r'_l \in \{0, \dots, \bar{t} - 1\}$ such that

$$l = q_l \bar{t} + r'_l.$$

We define $r_l = r'_l$ if $r'_l \in \{1, \dots, \bar{t} - 1\}$ and $r_l = \bar{t}$ if $r'_l = 0$. Finally, consider the sequence of agents $(\tilde{i}_1, \dots, \tilde{i}_{\tau\bar{t} + \hat{t} + 1})$ where $\tilde{i}_l = \bar{i}_{r_l}$ for all $l \in \{1, \dots, \tau\bar{t} + 1\}$ and $\tilde{i}_l = \hat{i}_{l - \tau\bar{t}}$ for all $l \in \{\tau\bar{t} + 1, \dots, \tau\bar{t} + 1 + \hat{t}\}$. By construction, we have that $w_{\tilde{i}_l \tilde{i}_{l+1}} > 0$ for all $l \in \{1, \dots, \tau\bar{t} + 1 + \hat{t}\}$. Since \mathbf{f} is a fixed point of \tilde{S} , note that $\tilde{S}^\tau(\mathbf{f}) = \mathbf{f}$ for all $\tau \in \mathbb{N}$ and

$$f_{\bar{i}} = \tilde{S}_{\bar{i}}^{\tau\bar{t} + \hat{t}}(\mathbf{f}) = \sum_{\mathbf{i} \in I^{\tau\bar{t} + \hat{t} + 1}: i_1 = \bar{i}} w_{i_1 i_2} \dots w_{i_{\tau\bar{t} + \hat{t}} i_{\tau\bar{t} + \hat{t} + 1}} M_{i_1} \dots M_{i_{\tau\bar{t} + \hat{t}}} f_{i_{\tau\bar{t} + \hat{t} + 1}}.$$

Define $f^{\mathbf{i}} = M_{i_1} \dots M_{i_{\tau\bar{t} + \hat{t}}} f_{i_{\tau\bar{t} + \hat{t} + 1}}$ for all $\mathbf{i} \in I^{\tau\bar{t} + \hat{t} + 1}$ such that $i_1 = \bar{i}$. We have that

$$f_{\bar{i}} = \sum_{\mathbf{i} \in I^{\tau\bar{t} + \hat{t} + 1}: i_1 = \bar{i}} w_{i_1 i_2} \dots w_{i_{\tau\bar{t} + \hat{t}} i_{\tau\bar{t} + \hat{t} + 1}} f^{\mathbf{i}}. \quad (36)$$

Since each M_i is an $\bar{n} \times \bar{n}$ stochastic matrix and $\max_{j \in J} f_i(\omega_j) \leq f_{\bar{i}}(\omega_{\bar{j}})$ for all $i \in I$, we have that $\max_{j \in J} f^{\mathbf{i}}(\omega_j) \leq f_{\bar{i}}(\omega_{\bar{j}})$ for all $\mathbf{i} \in I^{\tau\bar{t} + \hat{t} + 1}$ such that $i_1 = \bar{i}$. We focus on the summand

$$w_{\tilde{i}_1 \tilde{i}_2} \dots w_{\tilde{i}_{\tau\bar{t} + \hat{t}} \tilde{i}_{\tau\bar{t} + \hat{t} + 1}} M_{\tilde{i}_1} \dots M_{\tilde{i}_{\tau\bar{t} + \hat{t}}} f_{\tilde{i}_{\tau\bar{t} + \hat{t} + 1}}^{\tilde{i}} = w_{\tilde{i}_1 \tilde{i}_2} \dots w_{\tilde{i}_{\tau\bar{t} + \hat{t}} \tilde{i}_{\tau\bar{t} + \hat{t} + 1}} f^{\tilde{i}}.$$

By construction, we have that $w_{\tilde{i}_1\tilde{i}_2}\dots w_{\tilde{i}_{\tau\tilde{i}+\tilde{i}}\tilde{i}_{\tau\tilde{i}+\tilde{i}+1}} > 0$ and

$$M_{\tilde{i}_1}\dots M_{\tilde{i}_{\tau\tilde{i}+\tilde{i}}} f_{\tilde{i}_{\tau\tilde{i}+\tilde{i}+1}} = (M_{\tilde{i}_1}\dots M_{\tilde{i}_{\tau\tilde{i}}})^\tau M_{\tilde{i}_1}\dots M_{\tilde{i}_{\tau\tilde{i}}} f_{\tilde{i}_{\tau\tilde{i}+1}}.$$

Set $g = M_{\tilde{i}_1}\dots M_{\tilde{i}_{\tau\tilde{i}+1}} f_{\tilde{i}_{\tau\tilde{i}+1}} = M_{\tilde{i}_1}\dots M_{\tilde{i}_{\tau\tilde{i}}} f_{\tilde{i}_{\tau\tilde{i}+1}}$. Since each $M_{\tilde{i}_l}$ is an $\bar{n} \times \bar{n}$ stochastic matrix with strictly positive diagonal, so is $M_{\tilde{i}_1}\dots M_{\tilde{i}_{\tau\tilde{i}}}$. Since $\max_{j \in J} f_{\tilde{i}'}(\omega_j) \leq f_{\tilde{i}}(\omega_{\tilde{j}})$ and $f_{\tilde{i}'}(\omega_{\tilde{j}'}) < f_{\tilde{i}}(\omega_{\tilde{j}})$, this implies that $\min_{\omega \in \Omega} g(\omega) \leq g(\omega_{\tilde{j}'}) < f_{\tilde{i}}(\omega_{\tilde{j}})$ and $\max_{\omega \in \Omega} g(\omega) \leq f_{\tilde{i}}(\omega_{\tilde{j}})$. Since each entry of $(M_{\tilde{i}_1}\dots M_{\tilde{i}_{\tau\tilde{i}}})^\tau$ is strictly positive and $f^{\tilde{i}} = (M_{\tilde{i}_1}\dots M_{\tilde{i}_{\tau\tilde{i}}})^\tau g$, we can conclude that $f^{\tilde{i}}(\omega) < f_{\tilde{i}}(\omega_{\tilde{j}})$ for all $\omega \in \Omega$. By (36) and since $w_{\tilde{i}_1\tilde{i}_2}\dots w_{\tilde{i}_{\tau\tilde{i}+\tilde{i}}\tilde{i}_{\tau\tilde{i}+\tilde{i}+1}} > 0$ and $\max_{j \in J} f^{\mathbf{i}}(\omega_j) \leq f_{\tilde{i}}(\omega_{\tilde{j}})$ for all $\mathbf{i} \in I^{\tau\tilde{i}+\tilde{i}+1}$, this implies that

$$\begin{aligned} 0 &= \sum_{\mathbf{i} \in I^{\tau\tilde{i}+\tilde{i}+1}: i_1 = \tilde{i}} w_{i_1 i_2} \dots w_{i_{\tau\tilde{i}+\tilde{i}} i_{\tau\tilde{i}+\tilde{i}+1}} [f^{\mathbf{i}}(\omega_{\tilde{j}}) - f_{\tilde{i}}(\omega_{\tilde{j}})] \\ &\leq w_{\tilde{i}_1 \tilde{i}_2} \dots w_{\tilde{i}_{\tau\tilde{i}+\tilde{i}} \tilde{i}_{\tau\tilde{i}+\tilde{i}+1}} [f^{\tilde{i}}(\omega_{\tilde{j}}) - f_{\tilde{i}}(\omega_{\tilde{j}})] < 0, \end{aligned}$$

a contradiction.

(ii) implies (i). Since each V_i is normalized and W is a stochastic matrix, the statement is trivial. \blacksquare

Lemma 16. Let $S_\beta : (\mathbb{R}^\Omega)^n \rightarrow (\mathbb{R}^\Omega)^n$ be defined by equation (15). Fix $i, j \in I$ and $\omega, \omega' \in \Omega$. The following are equivalent:

- (i) $w_{ij} > 0$ and $\omega' \in \Pi_i(\omega)$;
- (ii) $\underline{a}_{(i,\omega)(j,\omega')}(S_1) = 1$;
- (iii) $\bar{a}_{(i,\omega)(j,\omega')}(S_1) = 1$.

Proof. (i) implies (ii). By Lemma 6, there exists $\varepsilon > 0$ such that for each $f \in \mathbb{R}^\Omega$ and for each $\delta \geq 0$

$$V_i(\omega, f + \delta e^{\omega'}) - V_i(\omega, f) \geq \varepsilon \delta.$$

Next, fix $\mathbf{f} = (f_l)_{l=1}^n \in (\mathbb{R}^\Omega)^n$ and $\delta \geq 0$, and observe that

$$S_{1,i,\omega}(\mathbf{f} + \delta e^{j,\omega'}) - S_{1,i,\omega}(\mathbf{f}) = V_i\left(\omega, \sum_{l=1}^n w_{il} f_l + w_{ij} \delta e^{\omega'}\right) - V_i\left(\omega, \sum_{l=1}^n w_{il} f_l\right) \geq \varepsilon w_{ij} \delta$$

proving the statement by setting $\varepsilon_{(i,\omega)(j,\omega')} = \varepsilon w_{ij}$.

(ii) implies (iii). Immediate.

(iii) implies (i). We prove the statement by contradiction. Fix $\mathbf{f} = (f_l)_{l=1}^n \in (\mathbb{R}^\Omega)^n$ and $\delta \geq 0$ and observe that

$$S_{1,i,\omega}(\mathbf{f} + \delta e^{j,\omega'}) - S_{1,i,\omega}(\mathbf{f}) = V_i\left(\omega, \sum_{l=1}^n w_{il} f_l + w_{ij} \delta e^{\omega'}\right) - V_i\left(\omega, \sum_{l=1}^n w_{il} f_l\right).$$

Therefore, if either $w_{ij} = 0$ or $\omega' \notin \Pi_i(\omega)$, then $S_{1,i,\omega}(\mathbf{f} + \delta e^{j,\omega'}) = S_{1,i,\omega}(\mathbf{f})$. Given that \mathbf{f} and δ were arbitrarily chosen, we obtain a contradiction. \blacksquare

Proof of Lemma 10. We have that $S_\beta(\mathbf{f}) = S_1((1-\beta)\hat{\mathbf{f}} + \beta\mathbf{f})$ for all $\beta \in (0,1)$ and recall that S_1 is normalized, monotone, and translation invariant. Fix $\lambda \in (0,1)$ and define $S_1^\lambda = \lambda I + (1-\lambda)S_1$. Clearly, we have that, for each $\mathbf{f} \in (\mathbb{R}^\Omega)^n$,

$$S_1^\lambda(\mathbf{f}) = \mathbf{f} \iff S_1(\mathbf{f}) = \mathbf{f}.$$

Therefore, by Lemma 9, $S_1^\lambda(\mathbf{f}) = \mathbf{f}$ if and only if there exists $m \in \mathbb{R}$ such that $f_i = f_{i'} = m1_\Omega$ for all $i, i' \in I$. By [2, Corollary 1 and part 2 of Proposition 2], it follows that there exists a unique class of indices $Z', \emptyset \neq Z' \subseteq I \times \Omega$, that is closed and strongly connected with respect to $\bar{A}(S_1^\lambda)$. It is easy to see that every row of $\bar{A}(S_1)$ is not null and that Z' is also closed and strongly connected with respect to $\bar{A}(S_1)$. In addition, by Lemma 16, every row of $\underline{A}(S_1)$ is not null and Z' is closed and strongly connected with respect to $\underline{A}(S_1)$. Finally, the statement follows by setting $Z = Z'$. \blacksquare

Lemma 17. Let $S_\beta : (\mathbb{R}^\Omega)^n \rightarrow (\mathbb{R}^\Omega)^n$ be defined by equation (15). We have $\{W^q \in \mathcal{W} : q \in Q\} \subseteq \partial S_1(0)$.

Proof. For every $(i, \omega) \in I \times \Omega$, by [5, Theorem 2.3.9], we have that

$$\partial S_{1,i,\omega}(0) = \{\rho \in \Delta(I \times \Omega) : \exists \tilde{q}_{i,\omega} \in \partial V_i(\omega, 0), \rho(j, \omega') = w_{ij}\tilde{q}_{i,\omega}(\omega')\}$$

where $\partial V_i(\omega, 0)$ denotes the superdifferential of the concave functional $V_i(\omega, \cdot)$ evaluated at $0 \in \mathbb{R}^\Omega$. With this, the statement follows by the definition of $\partial S_1(0)$, the definition of each W^q in equation (17), and by [44, Theorem 18]. \blacksquare

Proof of Lemma 11. Fix $\beta \in (0,1)$. By Theorem 1, $\{(V_i, \Pi_i)\}_{i \in I}$ exhibits convergence to a deterministic limit, hence, by Lemma 13, V_* is concave. This implies that

$$\begin{aligned} V_*(S_{\beta,i}(\mathbf{f})) &= V_*\left(V_i\left((1-\beta)\hat{f} + \beta \sum_{l=1}^n w_{il}f_l\right)\right) \geq V_*\left((1-\beta)\hat{f} + \beta \sum_{l=1}^n w_{il}f_l\right) \\ &\geq (1-\beta)V_*(\hat{f}) + \beta \sum_{l=1}^n w_{il}V_*(f_l) \quad \forall i \in I, \forall \mathbf{f} \in (\mathbb{R}^\Omega)^n. \end{aligned}$$

We now prove the statement for $\tau = 1$. We have that

$$V_*\left(S_{\beta,i}^1(\hat{\mathbf{f}})\right) = V_*\left(S_{\beta,i}(\hat{\mathbf{f}})\right) \geq (1-\beta)V_*(\hat{f}) + \beta \sum_{l=1}^n w_{il}V_*(\hat{f}_l) = V_*(\hat{f}) \quad \forall i \in I.$$

Assume that the statement is true for $\tau \in \mathbb{N}$. Observe that for each $i \in I$

$$V_*\left(S_{\beta,i}^{\tau+1}(\hat{\mathbf{f}})\right) = V_*\left(S_{\beta,i}\left(S_{\beta}^\tau(\hat{\mathbf{f}})\right)\right) \geq (1-\beta)V_*(\hat{f}) + \beta \sum_{l=1}^n w_{il}V_*\left(S_{\beta,l}^\tau(\hat{\mathbf{f}})\right) \geq V_*(\hat{f}).$$

The statement follows by induction. By Lemma 8, the previous part of the proof, and since by Lemma 13 V_* is a continuous ex-ante expectation, we have that

$$V_*\left(\sigma_i^\beta\right) = V_*\left(\lim_{\tau} S_{\beta,i}^\tau(\hat{\mathbf{f}})\right) = \lim_{\tau} V_*\left(S_{\beta,i}^\tau(\hat{\mathbf{f}})\right) \geq V_*(\hat{f}) \quad \forall i \in I, \forall \beta \in (0,1),$$

proving the statement. ■

Proof of Lemma 12. If $\Theta = \emptyset$ then the statement is trivially true, so fix $\mu \in \Theta$. We next show that $\mu \in \text{int}(\Delta(\Omega))$. First, observe that the full-support assumption on $\{(V_i, \Pi_i)\}_{i \in I}$ implies that, for all $i \in I$, $\omega' \in \Omega$, $\omega \in \Pi_i(\omega')$, and $p \in \arg \min_{\tilde{p} \in \Delta(\Omega)} c_{i, \omega'}(\tilde{p})$, we have $p(\omega) > 0$.³⁴ Second, let $\text{supp} \mu = E$ and assume by contradiction that $E \neq \Omega$. Since $\Pi_{\text{sup}} = \{\Omega\}$, we have that there exists $\omega \in \Omega \setminus E$, $i \in I$, and $\omega' \in E$ such that $\omega \in \Pi_i(\omega')$. Given that $\mu(\omega) = 0$ and $\mu(\omega') > 0$, we obtain

$$p_{\mu, i}(\omega', \omega) = 0,$$

yielding a contradiction with the fact that $p_{\mu, i}(\omega', \cdot) \in \arg \min_{\tilde{p} \in \Delta(\Omega)} c_{i, \omega'}(\tilde{p})$. ■

Proof of Lemma 13. 1. Consider an I -sequence $\iota = (i_k)_{k \in \mathbb{N}} \in I^{\mathbb{N}}$. Consider $f, g \in \mathbb{R}^{\Omega}$ and $\lambda \in (0, 1)$. Since each V_{i_1} is concave, we have that

$$V_{i_1}(\lambda f + (1 - \lambda)g) \geq \lambda V_{i_1}(f) + (1 - \lambda)V_{i_1}(g).$$

By induction, assume that

$$V_{i_k} \circ V_{i_{k-1}} \circ \dots \circ V_{i_2} \circ V_{i_1}(\lambda f + (1 - \lambda)g) \geq \lambda V_{i_k} \circ V_{i_{k-1}} \circ \dots \circ V_{i_2} \circ V_{i_1}(f) + (1 - \lambda)V_{i_k} \circ V_{i_{k-1}} \circ \dots \circ V_{i_2} \circ V_{i_1}(g).$$

Since $V_{i_{k+1}}$ is a concave interim expectation, we have that

$$\begin{aligned} V_{i_{k+1}} \circ V_{i_k} \circ V_{i_{k-1}} \circ \dots \circ V_{i_2} \circ V_{i_1}(\lambda f + (1 - \lambda)g) &= V_{i_{k+1}}(V_{i_k} \circ V_{i_{k-1}} \circ \dots \circ V_{i_2} \circ V_{i_1}(\lambda f + (1 - \lambda)g)) \\ &\geq V_{i_{k+1}}(\lambda V_{i_k} \circ V_{i_{k-1}} \circ \dots \circ V_{i_2} \circ V_{i_1}(f) + (1 - \lambda)V_{i_k} \circ V_{i_{k-1}} \circ \dots \circ V_{i_2} \circ V_{i_1}(g)) \\ &\geq \lambda V_{i_{k+1}} \circ V_{i_k} \circ V_{i_{k-1}} \circ \dots \circ V_{i_2} \circ V_{i_1}(f) + (1 - \lambda)V_{i_{k+1}} \circ V_{i_k} \circ V_{i_{k-1}} \circ \dots \circ V_{i_2} \circ V_{i_1}(g). \end{aligned}$$

By passing to the limit, we obtain that

$$\bar{V}_{\iota}(\lambda f + (1 - \lambda)g) 1_{\Omega} \geq \lambda \bar{V}_{\iota}(f) 1_{\Omega} + (1 - \lambda) \bar{V}_{\iota}(g) 1_{\Omega},$$

proving that \bar{V}_{ι} is concave. Since ι was arbitrarily chosen, we have that \bar{V}_{ι} is concave for every I -sequence ι . Finally, given that, by Proposition 1, we have

$$V_*(f) = \inf_{\iota \in I^{\mathbb{N}}: \iota \text{ is an } I\text{-sequence}} \bar{V}_{\iota}(f) \quad \forall f \in \mathbb{R}^{\Omega},$$

it follows that V_* is concave. With similar steps we can prove the second part of the first item.

2. Consider an I -sequence $\iota = (i_k)_{k \in \mathbb{N}} \in I^{\mathbb{N}}$. Consider $f, g \in \mathbb{R}^{\Omega}$ where g is Π_{inf} -measurable, and $\lambda \in (0, 1)$. Since each V_i is Π_{inf} -affine, we have that

$$V_{i_1}(\lambda f + (1 - \lambda)g) = \lambda V_{i_1}(f) + (1 - \lambda)V_{i_1}(g).$$

³⁴Indeed, for every $i \in I$, the operator $V_i : \mathbb{R}^{\Omega} \rightarrow \mathbb{R}^{\Omega}$ is monotone and such that its indicator matrix $A(V_i)$ (cf. Definition 7) satisfies

$$\omega \in \Pi_i(\omega') \implies a_{\omega' \omega} = 1 \quad \forall \omega, \omega' \in \Omega.$$

In particular, this implies that

$$\arg \min_{\tilde{p} \in \Delta(\Omega)} c_{i, \omega'}(\tilde{p}) = \partial V_i(\omega, 0) \subseteq \text{int}(\Delta(\Pi_i(\omega'))),$$

where the first equality follows from [7, Lemma 32]. The inclusion follows from concavity and the definition of the superdifferential.

By induction, assume that

$$V_{i_k} \circ V_{i_{k-1}} \circ \dots \circ V_{i_2} \circ V_{i_1} (\lambda f + (1 - \lambda) g) = \lambda V_{i_k} \circ V_{i_{k-1}} \circ \dots \circ V_{i_2} \circ V_{i_1} (f) + (1 - \lambda) V_{i_k} \circ V_{i_{k-1}} \circ \dots \circ V_{i_2} \circ V_{i_1} (g).$$

Since $V_{i_{k+1}}$ is Π_{inf} -affine and $V_{i_k} \circ V_{i_{k-1}} \circ \dots \circ V_{i_2} \circ V_{i_1} (g)$ is Π_{inf} -measurable, we have that

$$\begin{aligned} V_{i_{k+1}} \circ V_{i_k} \circ V_{i_{k-1}} \circ \dots \circ V_{i_2} \circ V_{i_1} (\lambda f + (1 - \lambda) g) &= V_{i_{k+1}} (V_{i_k} \circ V_{i_{k-1}} \circ \dots \circ V_{i_2} \circ V_{i_1} (\lambda f + (1 - \lambda) g)) \\ &= V_{i_{k+1}} (\lambda V_{i_k} \circ V_{i_{k-1}} \circ \dots \circ V_{i_2} \circ V_{i_1} (f) + (1 - \lambda) V_{i_k} \circ V_{i_{k-1}} \circ \dots \circ V_{i_2} \circ V_{i_1} (g)) \\ &= \lambda V_{i_{k+1}} \circ V_{i_k} \circ V_{i_{k-1}} \circ \dots \circ V_{i_2} \circ V_{i_1} (f) + (1 - \lambda) V_{i_{k+1}} \circ V_{i_k} \circ V_{i_{k-1}} \circ \dots \circ V_{i_2} \circ V_{i_1} (g). \end{aligned}$$

By Theorem 1 we can pass to the limit and obtain that

$$\bar{V}_\iota (\lambda f + (1 - \lambda) g) 1_\Omega = \lambda \bar{V}_\iota (f) 1_\Omega + (1 - \lambda) \bar{V}_\iota (g) 1_\Omega,$$

proving that \bar{V}_ι is Π_{inf} -affine. Since ι was arbitrarily chosen, we have that \bar{V}_ι is Π_{inf} -affine for every I -sequence ι . Finally, given that, by Proposition 1, we have

$$V_*(f) = \inf_{\iota \in I^\mathbb{N}: \iota \text{ is an } I\text{-sequence}} \bar{V}_\iota (f) \quad \forall f \in \mathbb{R}^\Omega,$$

it follows that

$$\begin{aligned} V_*((1 - \lambda) f + \lambda g) &= \inf_{\iota \in I^\mathbb{N}: \iota \text{ is an } I\text{-sequence}} \bar{V}_\iota ((1 - \lambda) f + \lambda g) \\ &= \inf_{\iota \in I^\mathbb{N}: \iota \text{ is an } I\text{-sequence}} \{ \lambda \bar{V}_\iota (f) + (1 - \lambda) \bar{V}_\iota (g) \} \\ &\geq \lambda \inf_{\iota \in I^\mathbb{N}: \iota \text{ is an } I\text{-sequence}} \bar{V}_\iota (f) + (1 - \lambda) \inf_{\iota \in I^\mathbb{N}: \iota \text{ is an } I\text{-sequence}} \bar{V}_\iota (g) \\ &= (1 - \lambda) V_*(f) + \lambda V_*(g). \end{aligned}$$

for all $\lambda \in (0, 1)$ and for all $f, h \in \mathbb{R}^\Omega$ where g is Π_{inf} -measurable. The statement for V^* follows from completely symmetric steps. \blacksquare

We first need some ancillary objects for the proof of Lemma 14. Define

$$\mathcal{V} = \{ \bar{V} \in C(F) : \bar{V} \text{ is normalized, monotone, translation invariant} \}.$$

Given that F is compact and each $\bar{V} \in \mathcal{V}$ is 1-Lipschitz continuous, (see, [11, p. 346]) it follows by Arzelà-Ascoli Theorem that \mathcal{V} is compact in the topology of uniform convergence.

Given the interim expectations $\{(V_i, \Pi_i)\}_{i \in I}$ and, for every $i \in I$, define

$$\mathcal{V}_i = \{ \bar{V} \in \mathcal{V} : \forall f \in F, \bar{V}(f) = \bar{V}(V_i(f)) \}.$$

Proof of Lemma 14. We first prove an ancillary claim.

Claim 2. *For every $i \in N$, $\mathcal{V}_i \subseteq C(F)$ is convex and compact in the topology of uniform convergence.*

Proof. Fix $i \in N$ and consider $\bar{V}, \bar{V}' \in \mathcal{V}_i$ as well as $\lambda \in [0, 1]$. Fix $f \in F$ and note that

$$\begin{aligned} (\lambda \bar{V} + (1 - \lambda) \bar{V}') (f) &= \lambda \bar{V}(f) + (1 - \lambda) \bar{V}'(f) \\ &= \lambda \bar{V}(V_i(f)) + (1 - \lambda) \bar{V}'(V_i(f)) = (\lambda \bar{V} + (1 - \lambda) \bar{V}') (V_i(f)), \end{aligned}$$

showing that $\lambda \bar{V} + (1 - \lambda) \bar{V}' \in \mathcal{V}_i$. Next, consider a sequence $\{\bar{V}_n\}_{n \in \mathbb{N}} \subseteq \mathcal{V}_i$ such that $\bar{V}_n \rightarrow \bar{V}$. Given that uniform convergence implies pointwise convergence, it is standard to show that \bar{V} is normalized, monotone, translation invariant and such that, for every $f \in F$, $\bar{V}(f) = \bar{V}(V_i(f))$. Therefore, \mathcal{V}_i is closed, hence compact, in the topology of uniform convergence. \square

Suppose that there exists no ex-ante expectation \bar{V} such that (\bar{V}, V_i, Π_i) is a nonlinear conditional expectation for all $i \in \{1, 2\}$, that is,

$$\mathcal{V}_1 \cap \mathcal{V}_2 = \emptyset.$$

By the Hahn-Banach separation theorem, there exists a linear continuous functional $L : C(F) \rightarrow \mathbb{R}$ and $c \in \mathbb{R}$, such that

$$L(\bar{V}_1) > c > L(\bar{V}_2)$$

for all $\bar{V}_1 \in \mathcal{V}_1$ and $\bar{V}_2 \in \mathcal{V}_2$. By the Riesz representation theorem, there exists $\nu \in \mathcal{M}(F)$ such that

$$L(V) = \int_F V(f) d\nu(f) \quad \forall f \in F.$$

Therefore,

$$\int_F \bar{V}_1(f) d\nu(f) > c > \int_F \bar{V}_2(f) d\nu(f)$$

for all $\bar{V}_1 \in \mathcal{V}_1$ and $\bar{V}_2 \in \mathcal{V}_2$. Fix $a \in [-k, k]$ such that $a \neq 0$ and define the measure $\nu_c \in \mathcal{M}(F)$ as

$$\nu_c = \nu - \frac{c}{a} \delta_{\{ae\}}$$

where $\delta_{\{ae\}}$ is the Dirac measure on ae . For every $\bar{V}_1 \in \mathcal{V}_1$ and $\bar{V}_2 \in \mathcal{V}_2$, it follows that

$$\int_F \bar{V}_1(f) d\nu_c(f) = \int_F \bar{V}_1(f) d\nu(f) - \frac{c}{a} \bar{V}_1(ae) = \int_F \bar{V}_1(f) d\nu(f) - c > 0.$$

Symmetrically, we also have

$$\int_F \bar{V}_2(f) d\nu_c(f) < 0.$$

Therefore,

$$\int_F \bar{V}_1(f) d\nu_c(f) > 0 > \int_F \bar{V}_2(f) d\nu_c(f),$$

for all $\bar{V}_1 \in \mathcal{V}_1$ and $\bar{V}_2 \in \mathcal{V}_2$. Given that for every $i \in I$ and $\omega \in \Omega$

$$V_i(\omega, \cdot) \in \mathcal{V}_i,$$

we obtain

$$\int_F V_1(\omega, f) d\nu_c(f) > 0 > \int_F V_2(\omega, f) d\nu_c(f) \quad \forall \omega \in \Omega,$$

as desired. Moreover, by [1, Corollary 5.108] $\mathcal{M}_0(F)$ is dense in $\mathcal{M}(F)$ endowed with the weak*-topology, and since Ω is finite and $V_i(\omega, \cdot)$ is continuous for all $i \in \{1, 2\}$ and for all $\omega \in \Omega$, the implication follows. \blacksquare

F Online appendix: An axiomatic foundation

In this section, we consider a single decision maker with preferences over monetary acts or utility profiles, that is, \mathbb{R}^Ω . We model the decision maker preferences via a binary relation \succsim on \mathbb{R}^Ω . We next list four important properties:

A 1 (Weak order). *The binary relation \succsim is complete and transitive.*

A 2 (Certainty equivalent). *For each $f \in \mathbb{R}^\Omega$ there exists $k \in \mathbb{R}$ such that $f \sim k1_\Omega$.*

A 3 (Continuity). *For each $f, g, h \in \mathbb{R}^\Omega$ the sets*

$$\{\lambda \in [0, 1] : \lambda f + (1 - \lambda)g \succsim h\} \text{ and } \{\lambda \in [0, 1] : h \succsim \lambda f + (1 - \lambda)g\}$$

are closed.

A 4 (Monotonicity). *For each $f, g \in \mathbb{R}^\Omega$ and for each $h, k \in \mathbb{R}$*

$$\begin{aligned} f \geq g &\implies f \succsim g \\ &\text{and} \\ h > k &\implies h1_\Omega \succ k1_\Omega. \end{aligned}$$

On the one hand, transitivity and monotonicity are common assumptions of rationality while completeness reflects the burden of choice the decision maker faces. On the other hand, continuity is a technical assumption which will allow us to represent preferences through a continuous utility function. The assumption of certainty equivalent shares both features. It allows us to show that preferences admit a utility function, possibly not continuous, yet it takes a clear behavioral interpretation: the decision maker for each random variable admits an equivalent amount which received with certainty makes her indifferent to the random prospect. The above axioms define the following two nested class of preferences.

Definition 8. *Let \succsim be a binary relation on \mathbb{R}^Ω . We say that \succsim is a rational preference if and only if it satisfies weak order, certainty equivalent, and monotonicity. We say that \succsim is a continuous rational preference if and only if it satisfies weak order, continuity, and monotonicity.*

It is easy to show that continuous rational preferences are rational preferences. Continuous rational preferences were studied by Cerreia-Vioglio, Ghirardato, Maccheroni, Marinacci, and Siniscalchi [8]. The next result is a version of their Proposition 1.

Proposition 7. *Let \succsim be a binary relation on \mathbb{R}^Ω . The following statements are equivalent:*

- (i) \succsim is a rational preference;
- (ii) There exists a normalized and monotone functional $\tilde{V} : \mathbb{R}^\Omega \rightarrow \mathbb{R}$ such that

$$f \succsim g \iff \tilde{V}(f) \geq \tilde{V}(g). \quad (37)$$

Moreover, we have that:

1. The functional \tilde{V} is continuous if and only if \succsim is a continuous rational preference.
2. The functional \tilde{V} is the unique normalized functional satisfying (37).

Proof. (ii) implies (i). It is routine.

(i) implies (ii). Since \succsim satisfies certainty equivalent, for each $f \in \mathbb{R}^\Omega$ define k_f to be such that $k_f 1_\Omega \sim f$. Since \succsim satisfies weak order and monotonicity, we have that k_f is unique. Define $\tilde{V} : \mathbb{R}^\Omega \rightarrow \mathbb{R}$ by $\tilde{V}(f) = k_f$ for all $f \in \mathbb{R}^\Omega$. Since \succsim satisfies weak order and monotonicity, we have that

$$f \succsim g \iff k_f 1_\Omega \succsim k_g 1_\Omega \iff k_f \geq k_g \iff \tilde{V}(f) \geq \tilde{V}(g),$$

proving (37). Clearly, if $f = k 1_\Omega$ for some $k \in \mathbb{R}$, we have that $\tilde{V}(k 1_\Omega) = \tilde{V}(f) = k_f = k$, proving that \tilde{V} is normalized. Finally, since \succsim satisfies monotonicity, if $f \geq g$, then $f \succsim g$ and $\tilde{V}(f) \geq \tilde{V}(g)$, proving that \tilde{V} is monotone.

1. The “Only if” is routine. “If”. Since \succsim satisfies weak order, continuity, and monotonicity, we have that \succsim satisfies certainty equivalent. It follows that \tilde{V} as defined above represents \succsim . Since \succsim satisfies continuity, it follows that for each $f, g \in \mathbb{R}^\Omega$ and for each $c \in \mathbb{R}$

$$\begin{aligned} \left\{ \lambda \in [0, 1] : \tilde{V}(\lambda f + (1 - \lambda)g) \leq c \right\} &= \left\{ \lambda \in [0, 1] : \tilde{V}(\lambda f + (1 - \lambda)g) \leq \tilde{V}(c 1_\Omega) \right\} \\ &= \{ \lambda \in [0, 1] : c 1_\Omega \succsim \lambda f + (1 - \lambda)g \} \end{aligned}$$

where the latter set is closed. By [10, Lemma 42], we have that \tilde{V} is lower semicontinuous. By [10, Appendix A.3], upper semicontinuity follows similarly.

2. Assume that \hat{V} is normalized and satisfies (37). We have that for each $f \in \mathbb{R}^\Omega$

$$\hat{V}(f) = \hat{V}(\hat{V}(f) 1_\Omega) \implies f \sim \hat{V}(f) 1_\Omega \implies \tilde{V}(f) = \tilde{V}(\hat{V}(f) 1_\Omega) = \hat{V}(f),$$

proving that $\hat{V} = \tilde{V}$. ■

We can now discuss conditional preferences. We assume that there are two periods 0 and 1. At 0, the decision maker has no information and has also preferences over \mathbb{R}^Ω . At time 1, the decision maker observes an event E from a partition Π of Ω and updates her preferences. We model this by a pair $(\succsim, \{\succsim_\omega\}_{\omega \in \Omega})$.

A 5 (Rationality). *The binary relation \succsim is a rational preference and \succsim_ω is a continuous rational preference for all $\omega \in \Omega$.*

A 6 (Conditional preferences). *For each $\omega, \omega' \in \Omega$*

$$\Pi(\omega) = \Pi(\omega') \implies \succsim_\omega = \succsim_{\omega'}.$$

We thus assume that original and updated preferences are rational, where the latter are also assumed to be continuous. At the same time, we assume that if two states belong to the same event, then the corresponding updated preferences must be the same, incorporating exactly nothing more than the information embedded in Π .

A 7 (Consequentialism). *For each $f \in \mathbb{R}^\Omega$ and for each $\omega \in \Omega$*

$$f 1_{\Pi(\omega)} + h 1_{\Pi(\omega)^c} \sim_\omega f \quad \forall h \in \mathbb{R}^\Omega.$$

A 8 (Dynamic consistency). For each $f, g \in \mathbb{R}^\Omega$

$$f \succsim_\omega g \quad \forall \omega \in \Omega \implies f \succsim g.$$

On the one hand, consequentialism imposes that updated preferences over are only influenced by the states that are still relevant/possible. On the other hand, dynamic consistency is a form of monotonicity and it states that if interim f is weakly better than g , no matter which event realized in Π , then f is weakly better than g also at time 0.

Definition 9. Let $(\succsim, \{\succsim_\omega\}_{\omega \in \Omega})$ be a collection of binary relations on \mathbb{R}^Ω . We say that $(\succsim, \{\succsim_\omega\}_{\omega \in \Omega})$ is a dynamic rational preference if and only if it satisfies the properties of rationality, conditional preferences, consequentialism, and dynamic consistency.

The next result provides a behavioral foundation for nonlinear conditional expectations.

Proposition 8. Let $(\succsim, \{\succsim_\omega\}_{\omega \in \Omega})$ be a collection of binary relations on \mathbb{R}^Ω . The following statements are equivalent:

- (i) $(\succsim, \{\succsim_\omega\}_{\omega \in \Omega})$ is a dynamic rational preference;
- (ii) There exist two functions $\bar{V} : \mathbb{R}^\Omega \rightarrow \mathbb{R}$ and $V : \Omega \times \mathbb{R}^\Omega \rightarrow \mathbb{R}$ such that (\bar{V}, V, Π) is a nonlinear conditional expectation and for each $\omega \in \Omega$

$$f \succsim_\omega g \iff V(\omega, f) \geq V(\omega, g) \quad \text{and} \quad f \succsim g \iff \bar{V}(f) \geq \bar{V}(g).$$

Proof. (ii) implies (i). It is routine.

(i) implies (ii). By Proposition 7 and since $(\succsim, \{\succsim_\omega\}_{\omega \in \Omega})$ satisfies rationality, we have that there exists a normalized and monotone function $\bar{V} : \mathbb{R}^\Omega \rightarrow \mathbb{R}$ and a collection of normalized, monotone, and continuous functions $\{V_\omega\}_{\omega \in \Omega}$ from \mathbb{R}^Ω to \mathbb{R} such that \bar{V} represents \succsim and V_ω represents \succsim_ω for all $\omega \in \Omega$. Define $V : \Omega \times \mathbb{R}^\Omega \rightarrow \mathbb{R}$ by $V(\omega, f) = V_\omega(f)$ for all $(\omega, f) \in \Omega \times \mathbb{R}^\Omega$. By point 2 of Proposition 7 and since $(\succsim, \{\succsim_\omega\}_{\omega \in \Omega})$ satisfies conditional preferences, we have that for each $\omega, \omega' \in \Omega$

$$\Pi(\omega) = \Pi(\omega') \implies \succsim_\omega = \succsim_{\omega'} \implies V(\omega, \cdot) = V(\omega', \cdot),$$

proving that $V(\cdot, f)$ is Π -measurable for all $f \in \mathbb{R}^\Omega$. Since $(\succsim, \{\succsim_\omega\}_{\omega \in \Omega})$ satisfies consequentialism, we have that for each $\omega \in \Omega$ and for each $f, h \in \mathbb{R}^\Omega$

$$f1_{\Pi(\omega)} + h1_{\Pi(\omega)^c} \sim_\omega f \implies V(\omega, f1_{\Pi(\omega)} + h1_{\Pi(\omega)^c}) = V(\omega, f).$$

Finally, for each $f \in \mathbb{R}^\Omega$ define $\bar{f} \in \mathbb{R}^\Omega$ by $\bar{f}(\omega) = V(\omega, f)$ for all $\omega \in \Omega$. It follows that $\bar{f} \sim_\omega \bar{f}1_{\Pi(\omega)} \sim_\omega f$ for all $\omega \in \Omega$ and for all $f \in \mathbb{R}^\Omega$. Since $(\succsim, \{\succsim_\omega\}_{\omega \in \Omega})$ satisfies dynamic consistency, we can conclude that $\bar{f} \sim f$ and, in particular, $\bar{V}(f) = \bar{V}(\bar{f}) = \bar{V}(V(\cdot, f))$ for all $f \in \mathbb{R}^\Omega$. \blacksquare

Clearly, in Proposition 8, linear conditional expectations are obtained by requiring in (i) \succsim and each \succsim_ω to satisfy the axiom of independence. Similarly, maxmin conditional expectations, as in Example 2, are obtained by imposing c-independence.

G Online appendix: Partial dynamic consistency

Consider the following weaker notion of common ex-ante expectation. As before, let $\{(V_i, \Pi_i)\}_{i \in I}$ be a profile of interim expectations. Fix any partition Π' that is finer than Π_{\inf} . We say that the agents have a Π' -nonlinear common ex-ante expectation if there exists an ex-ante expectation \bar{V} that satisfies

$$\bar{V}(f) = \bar{V}(V_i(\cdot, f))$$

for all $i \in I$ and for all $f \in \mathbb{R}^\Omega$ that are Π' -measurable. By inspection of the proof of Corollary 1, it is easy to see that, if $\{(V_i, \Pi_i)\}_{i \in I}$ exhibits convergence to a deterministic limit, then the existence of this weaker form of common ex-ante expectation is equivalent to the following:

(ii') For each Π' -measurable $f \in \mathbb{R}^\Omega$ there exists $k_f \in \mathbb{R}$ such that for each I -sequence $(i_t)_{t \in \mathbb{N}}$

$$\lim_{t \rightarrow \infty} V_{i_t} \circ V_{i_{t-1}} \circ \dots \circ V_{i_2} \circ V_{i_1}(f) = k_f 1_\Omega.$$

Moreover, as in Corollary 1, this common limit coincides with the common ex-ante evaluation for every Π' -measurable act f , that is, $\bar{V}(f) = k_f$.

Next, define the ex-ante preference $\bar{V}_i = \bar{V} \circ V_i$ for every $i \in I$.³⁵ One can show that, for every $i \in I$, the functional \bar{V}_i is the unique ex-ante preference that coincides with \bar{V} on the Π' -measurable acts and is individually dynamically consistent in the sense that

$$\bar{V}_i(g) = \bar{V}_i(V_i(\cdot, g)) \quad \forall g \in \mathbb{R}^\Omega.$$

Moreover, for each I -sequence $(i_t)_{t \in \mathbb{N}}$,

$$\lim_{t \rightarrow \infty} V_{i_t} \circ V_{i_{t-1}} \circ \dots \circ V_{i_2} \circ V_{i_1}(g) = \bar{V}_{i_1}(g) 1_\Omega \quad \forall g \in \mathbb{R}^\Omega,$$

that is, the ex-ante expectation of i_1 corresponds to the limit for the higher order expectations of every I -sequence where the first-order expectation is the one of i_1 .

H Online appendix: Algorithm to construct extreme ex-ante expectations

In this section, we propose an algorithm to compute V_* . Consider any ex-ante expectation $\hat{V} : \mathbb{R}^\Omega \rightarrow \mathbb{R}$ such that $\hat{V} \geq V_*$. For example, one can choose

$$\hat{V}(f) = \max_{\omega \in \Omega} f(\omega) \quad \forall f \in \mathbb{R}^\Omega.$$

Define recursively the sequence $\{\hat{V}^\tau\}_{\tau \in \mathbb{N}}$ of real-valued functionals over \mathbb{R}^Ω by $\hat{V}^1 = \hat{V}$ and

$$\hat{V}^{\tau+1}(f) = \min_{i \in I} \hat{V}^\tau(V_i(f)) \quad \forall f \in \mathbb{R}^\Omega, \forall \tau \in \mathbb{N}.$$

³⁵Observe that each \bar{V}_i is well defined since, for every $g \in \mathbb{R}^\Omega$, $V_i(g)$ is Π' -measurable, hence we can evaluate through \bar{V} .

By induction, we have that each \hat{V}^τ is an ex-ante expectation. Fix $f \in \mathbb{R}^\Omega$. Since each V_i is an interim expectation, if $\tau \geq 2$, then we have that

$$\begin{aligned}\hat{V}^{\tau+1}(f) &= \min_{i \in I} \hat{V}^\tau(V_i(f)) = \min_{i \in I} \min_{i' \in I} \hat{V}^{\tau-1}(V_{i'}(V_i(f))) \leq \min_{i \in I} \hat{V}^{\tau-1}(V_i(V_i(f))) \\ &= \min_{i \in I} \hat{V}^{\tau-1}(V_i(f)) = \hat{V}^\tau(f).\end{aligned}$$

Since f was arbitrarily chosen, this implies that $\hat{V}^{\tau+1} \leq \hat{V}^\tau$ for all $\tau \in \mathbb{N} \setminus \{1\}$. Define $\hat{V}^\infty : \mathbb{R}^\Omega \rightarrow \mathbb{R}$ by $\hat{V}(f) = \lim_\tau \hat{V}^\tau(f)$ for all $f \in \mathbb{R}^\Omega$.

Proposition 9. *For every ex-ante expectation $\hat{V} : \mathbb{R}^\Omega \rightarrow \mathbb{R}$ such that $\hat{V} \geq V_*$, we have $\hat{V}^\infty = V_*$.*

Proof of Proposition 9. Since $\{\hat{V}^\tau(f)\}_{\tau \in \mathbb{N}}$ is an eventually decreasing sequence bounded from below by $\min_{\omega \in \Omega} f(\omega)$, \hat{V}^∞ is a well defined ex-ante expectation. By construction, we have that

$$\hat{V}^{\tau+1}(f) \leq \hat{V}^\tau(V_i(f)) \quad \forall f \in \mathbb{R}^\Omega, \forall i \in I.$$

By passing to the limit, we obtain that $\hat{V}^\infty(f) \leq \hat{V}^\infty(V_i(f))$ for all $f \in \mathbb{R}^\Omega$ and for all $i \in I$, which in turn yields that $\hat{V}^\infty \leq V_*$ by definition of V_* . Conversely, note that

1. Since $\hat{V}^1 = \hat{V} \geq V_*$, if $\tau = 1$, then $\hat{V}^{\tau+1}(f) = \min_{i \in I} \hat{V}^\tau(V_i(f)) \geq \min_{i \in I} V_*(V_i(f)) \geq V_*(f)$ for all $f \in \mathbb{R}^\Omega$.
2. By induction assume that $\hat{V}^\tau \geq V_*$. It follows that

$$\hat{V}^{\tau+1}(f) = \min_{i \in I} \hat{V}^\tau(V_i(f)) \geq \min_{i \in I} V_*(V_i(f)) \geq V_*(f) \quad \forall f \in \mathbb{R}^\Omega,$$

proving the inductive step.

By induction, we conclude that $\hat{V}^\tau \geq V_*$ for all $\tau \in \mathbb{N}$, yielding that $\hat{V}^\infty \geq V_*$ and, in particular, $\hat{V}^\infty = V_*$. ■

I Online appendix: Additional results and examples

We first state a corollary of Theorem 1 and Proposition 1 bounding the difference between the iterated expectations along two different I -sequences without assuming the existence of a nonlinear common ex-ante expectation.

Corollary 5. *Let $\{(V_i, \Pi_i)\}_{i \in I}$ be a collection of full support interim expectations such that $\Pi_{\text{sup}} = \{\Omega\}$. There exist $\varepsilon \in (0, 1)$ and $C \in \mathbb{R}_+$ such that for every pair of I -sequence $\iota = (i_m)_{m \in \mathbb{N}}$ and $\iota' = (i'_m)_{m \in \mathbb{N}}$, and for each $\tau, t \in \mathbb{N}$, if every $i \in I$ appears at least τ times in both (i_1, \dots, i_t) and (i'_1, \dots, i'_t) , then*

$$\|V_\iota^t(f) - V_{\iota'}^t(f)\|_\infty \leq \|V_*(f) - V^*(f)\|_\infty + C\varepsilon^\tau \|f\|_\infty \quad \forall f \in \mathbb{R}^\Omega.$$

Observe that, in the two-agent case, when there exists a nonlinear common ex-ante expectation, the previous result gives a bound on the higher-order disagreement between agents, by getting rid of the first term on the right-hand side.

Next we present an example that complements the analysis in Example 5.

Example 8 (Extreme information asymmetry). Consider two traders $I = \{1, 2\}$ that are uncertain about an asset $\hat{f} \in \mathbb{R}^\Omega$ and are endowed respectively with full-information $\Pi_1 = 2^\Omega$ and no-information $\Pi_2 = \{\Omega\}$. In this case, the interim expectation of an act f by player 1 in each state $\omega \in \Omega$ must coincide with $f(\omega)$. As player 2 does not receive any information, both her ex-ante and interim expectations are variational and given by

$$V_2(f) = \min_{p \in \Delta(\Omega)} \{\mathbb{E}_p(f) + c(p)\}.$$

With this, the interim expectations of the players admit a common ex-ante expectation which must coincide with the preference of player 2, that is $\bar{V} = V_2$.³⁶ Next, for all $\beta \in (0, 1)$, the equilibrium strategy of player 2 does not depend on the realized state

$$\sigma_2^\beta = \min_{p \in \Delta(\Omega)} \left\{ \mathbb{E}_p \left((1 - \beta) \hat{f} + \beta \sigma_1^\beta \right) + c(p) \right\},$$

while the equilibrium strategy of player 1 is adapted to the realized state

$$\sigma_1^\beta(\omega) = (1 - \beta) \hat{f}(\omega) + \beta \sigma_2^\beta \quad \forall \omega \in \Omega.$$

By simple substitution, we get

$$\sigma_2^\beta = \min_{p \in \Delta(\Omega)} \left\{ \mathbb{E}_p(\hat{f}) + \frac{1}{(1 - \beta^2)} c(p) \right\} \geq \bar{V}(\hat{f}),$$

that is, the equilibrium willingness to pay of player 2 coincides with a less ambiguity-averse version of the ex-ante common expectation. In the high-coordination limit, the ambiguity of the players is restricted only among the least penalized probabilistic models:

$$\lim_{\beta \rightarrow 1} \sigma_2^\beta = \lim_{\beta \rightarrow 1} \sigma_1^\beta(\omega) = \min_{\mu \in \Theta} \mathbb{E}_\mu(\hat{f}) \geq \bar{V}(\hat{f}) \quad \forall \omega \in \Omega,$$

where $\Theta = \arg \min_{p \in \Delta(\Omega)} c(p)$.³⁷ In words, the equilibrium price is converging to a cautious evaluation consistent with the most trusted probabilistic models, i.e., $p \in \Delta(\Omega)$ such that $c(p) = 0$. In the maxmin model, where c is the (convex-analysis) indicator function of a set $C \subseteq \Delta(\Omega)$, this cautious evaluation coincides with the common ex-ante expectation, so that $\lim_{\beta \rightarrow 1} \sigma_i^\beta(\omega) = \bar{V}(\hat{f})$. ▲

³⁶Observe that, given the extreme nature of the information structures considered, there is no need to specify an updating rule for the preferences of the agents.

³⁷This last step follows by [7, Proposition 12].

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