

Noise Robust Core-stable Coalitions of Hedonic Games Supplementary Material

Prashant Trivedi
IEOR, Indian Institute of Technology Bombay

TRIVEDI.PRASHANT15@IITB.AC.IN

Nandyala Hemachandra
IEOR, Indian Institute of Technology Bombay

NH@IITB.AC.IN

Editors: Emtiyaz Khan and Mehmet Gönen

1. n agents 2 support partial information noise model

In a two support noise model we have $\mathcal{N}_{sp} = \{1, \alpha\}$ with $\alpha > 1$, such that for any coalition $S \subseteq N$, $\mathbb{P}[\alpha(S) = \alpha] = p = 1 - \mathbb{P}[\alpha(S) = 1]$. We derive the agreement probability, $f_T(p, \alpha)$ in the following lemma. Note that this lemma serves as the base case in the Mathematical induction based proof of the Theorem 11 in the main paper.

Lemma 1 *Let $\tilde{\pi}$ be $\tilde{\epsilon}$ -PAC stable partition of noisy game $(N, \tilde{\mathbf{v}})$, and let $\tilde{\pi}$ be a ϵ -PAC stable outcome of the noise-free game (N, \mathbf{v}) , where ϵ is identified in Theorem 5 of the paper. Then the agreement probability $f_T(p, \alpha)$ is given by*

$$f_T(p, \alpha) = \begin{cases} 1, & \text{if } \tilde{\pi}(i) = T, \forall i \in T \\ p + (1-p)^{|\mathcal{R}(T)|+1-|\mathcal{I}(\alpha, T)|}, & \text{otherwise} \end{cases}$$

where $\mathcal{I}(\alpha, T) = \left\{ \tilde{\pi}(i) \in \mathcal{R}(T) \mid \frac{\tilde{v}_i(\tilde{\pi}(i))}{\tilde{v}_i(T)} \geq \alpha \right\}$.

Proof Recall from Theorem 5 in main paper we have the following

$$\mathbb{P}_{T \sim \tilde{\mathcal{D}}}[\bigcup_{i \in T} v_i(\tilde{\pi}(i)) \geq v_i(T)] \geq (1 - \tilde{\epsilon})f_T(\mathbf{p}, \boldsymbol{\alpha}).$$

Also, recall that the agreement event is defined as

$$M(\tilde{\pi}, T) := \{(\{\alpha(\tilde{\pi}(i))\}_{\tilde{\pi}(i) \in \mathcal{R}(T)}, \alpha(T)) : \bigcap_{i \in T} \{v_i(\tilde{\pi}(i)) \geq v_i(T) \cap \alpha(\tilde{\pi}(i))v_i(\tilde{\pi}(i)) \geq \alpha(T)v_i(T)\}\},$$

and $f_T(p, \alpha) = \mathbb{P}[M(\tilde{\pi}, T)]$ is the probability of agreement event. Moreover,

$$\mathcal{R}(T) := \{\tilde{\pi}(i) \mid i \in T\}; \quad \mathcal{I}(\alpha, T) = \left\{ \tilde{\pi}(i) \in \mathcal{R}(T) \mid \frac{\tilde{v}_i(\tilde{\pi}(i))}{\tilde{v}_i(T)} \geq \alpha \right\}.$$

To find the agreement probability, $f_T(\mathbf{p}, \boldsymbol{\alpha})$ we consider two cases $\mathcal{I}(\alpha, T) = \emptyset$, and $\mathcal{I}(\alpha, T) \neq \emptyset$. For these cases we identify the possible noise values $\{\alpha(\tilde{\pi}(i))\}_{\tilde{\pi}(i) \in \mathcal{R}(T)}, \alpha(T)$ that are element of $M(\tilde{\pi}, T)$.

- **Case 01:** $[\mathcal{I}(\alpha, T) = \emptyset]$. In this case, we have following elements in $M(\tilde{\pi}, T)$.

- $\alpha(\tilde{\pi}(i)) = 1, \forall \tilde{\pi}(i) \in \mathcal{R}(T)$ and $\alpha(T) = 1$. The probability of such choice of α 's is

$$(1-p)^{|\mathcal{R}(T)|+1}. \quad (1)$$

- $\alpha(\tilde{\pi}(i)) = \alpha$ for **exactly one** $\tilde{\pi}(i) \in \mathcal{R}(T)$, and $\alpha(\tilde{\pi}(i)) = 1$ for remaining coalitions in $\mathcal{R}(T)$, and $\alpha(T) = \alpha$. Probability of such choice of α 's is $(p \times (1-p)^{|\mathcal{R}(T)|-1}) \times p$. And there are $\binom{|\mathcal{R}(T)|}{1}$ ways of selecting **exactly one** coalition $\tilde{\pi}(i) \in \mathcal{R}(T)$. Thus, the probability of above α 's is $\binom{|\mathcal{R}(T)|}{1} p(1-p)^{|\mathcal{R}(T)|-1} p$.

In general, for any $k \in \{0, 1, \dots, |\mathcal{R}(T)|\}$ coalitions $\tilde{\pi}(i) \in \mathcal{R}(T)$, take $\alpha(\tilde{\pi}(i)) = \alpha$. Moreover, $\alpha(\tilde{\pi}(i)) = 1$ for remaining $|\mathcal{R}(T)| - k$ coalitions and take $\alpha(T) = \alpha$. Further, we have $\binom{|\mathcal{R}(T)|}{k}$ similar choices. So, the probability of the above choice of α 's is

$$\begin{aligned} \sum_{k=0}^{|\mathcal{R}(T)|} \left\{ \binom{|\mathcal{R}(T)|}{k} p^k (1-p)^{|\mathcal{R}(T)|-k} \right\} \times p &= p \times \left(\sum_{k=0}^{|\mathcal{R}(T)|} \binom{|\mathcal{R}(T)|}{k} p^k (1-p)^{|\mathcal{R}(T)|-k} \right) \\ &= p. \end{aligned} \quad (2)$$

This is because for any coalition S , we have $\mathbb{P}[\alpha(S) = \alpha] = p = 1 - \mathbb{P}[\alpha(S) = 1]$ and the fact that binomial probabilities summed up to 1.

- **Case 02:** $[\mathcal{I}(\alpha, T) \neq \emptyset]$. Then, in addition to the above possible cases, we will have a few other cases, which are:

- $\alpha(\tilde{\pi}(i)) = \alpha$ for **exactly one** $\tilde{\pi}(i) \in \mathcal{I}(\alpha, T)$, $\alpha(\tilde{\pi}(i)) = 1$ for remaining coalitions in $\mathcal{R}(T)$ and $\alpha(T) = 1$. Probability of such choice of α 's is $p(1-p)^{|\mathcal{R}(T)|-1}(1-p) = p(1-p)^{|\mathcal{R}(T)|}$. And there are $\binom{|\mathcal{I}(\alpha, T)|}{1}$ ways of choosing **exactly one** coalition $\tilde{\pi}(i) \in \mathcal{I}(\alpha, T)$. Thus the overall probability is $\binom{|\mathcal{I}(\alpha, T)|}{1} p(1-p)^{|\mathcal{R}(T)|}$.

In general, we have $\alpha(\tilde{\pi}(i)) = \alpha$ for any $k \in \{1, 2, \dots, |\mathcal{I}(\alpha, T)|\}$ coalitions $\tilde{\pi}(i) \in \mathcal{I}(\alpha, T)$. Moreover, $\alpha(\tilde{\pi}(i)) = 1$ for remaining $|\mathcal{R}(T)| - k$ coalitions, and $\alpha(T) = 1$. Probability of such choice of α 's is $p^k (1-p)^{|\mathcal{R}(T)|-|\mathcal{I}(\alpha, T)|} (1-p)$. And there are $\binom{|\mathcal{I}(\alpha, T)|}{k}$ ways of selecting k coalitions $\tilde{\pi}(i) \in \mathcal{I}(\alpha, T)$. Thus the overall probability is

$$\sum_{k=1}^{|\mathcal{I}(\alpha, T)|} \binom{|\mathcal{I}(\alpha, T)|}{k} p^k (1-p)^{|\mathcal{R}(T)|-|\mathcal{I}(\alpha, T)|} (1-p). \quad (3)$$

The probability of event $M(\tilde{\pi}, T)$, i.e., $\mathbb{P}[M(\tilde{\pi}, T)]$ is obtained by adding probabilities given in Equations (1), (2) and (3).

$$\begin{aligned} \mathbb{P}[M(\tilde{\pi}, T)] &= (1-p)^{|\mathcal{R}(T)|+1} + p + \sum_{k=1}^{|\mathcal{I}(\alpha, T)|} \binom{|\mathcal{I}(\alpha, T)|}{k} p^k (1-p)^{|\mathcal{R}(T)|-|\mathcal{I}(\alpha, T)|} (1-p) \\ &= (1-p)^{|\mathcal{R}(T)|+1} + p + (1-p)^{|\mathcal{R}(T)|-|\mathcal{I}(\alpha, T)|+1} \left[1 - (1-p)^{|\mathcal{I}(\alpha, T)|} \right] \end{aligned}$$

$$= p + (1 - p)^{|\mathcal{R}(T)| - |\mathcal{I}(\alpha, T)| + 1}.$$

This ends the proof. ■

If $\tilde{\pi}(i) \neq T$ for at least one $i \in T$, then $f_T(p, \alpha) = 1$, $\forall \alpha$ if and only if $p = 0$ or $p = 1$. That is, if the value of all the coalitions are retained, or if values of all of them are inflated by α , then for all $i \in T$, and for all $\tilde{\pi}(i) \in \mathcal{R}(T)$, one has $\tilde{\pi}(i) \succeq_i T$, and $\tilde{\pi}(i) \succeq'_i T$. Thus, $\tilde{\pi}$ is ϵ -PAC stable outcome of *unknown noise-free* game and hence $\tilde{\pi}$ is noise-robust.

Corollary 2 *When $\tilde{\pi} = N$, i.e., the grand coalition is $\tilde{\epsilon}$ -PAC stable outcome in the noisy game, then $\mathcal{R}(T) = \{N\}$ for any coalition T . Thus, $\mathcal{I}(\alpha, T) = \emptyset$, or $\mathcal{I}(\alpha, T) = \{N\}$. Therefore, $f_T(p, \alpha)$ simplifies to*

$$f_T(p, \alpha) = \begin{cases} 1, & \text{if } \mathcal{I}(\alpha, T) = \{N\} \\ (1 - p)^2 + p, & \text{if } \mathcal{I}(\alpha, T) = \emptyset. \end{cases} \quad (4)$$

2. n agents 2 support partial information noisy games without core

Suppose $\tilde{\pi}$ is not $\tilde{\epsilon}$ -PAC stable partition for the noisy game $(N, \tilde{\mathbf{v}})$. Moreover, let the noise support be $\mathcal{N}_{sp} = \{1, \alpha\}$, the following lemma provides the expression of $h_T(p, \alpha)$. Note that this lemma serves as the base case for the Mathematical induction based proof of Theorem 15 in the main paper.

Lemma 3 *Suppose $\tilde{\pi}$ is not a $\tilde{\epsilon}$ -PAC stable outcome of the noisy game $(N, \tilde{\mathbf{v}})$, then the agreement probability $h_T(p, \alpha)$ for noise support $\mathcal{N}_{sp} \in \{1, \alpha\}$ is given by*

$$h_T(p, \alpha) = \begin{cases} 1, & \text{if } \tilde{\pi}(i) = T, \forall i \in T \\ (1 - p) + p^{|\mathcal{R}(T)| + 1 - |\mathcal{J}(\alpha, T)|}, & \text{otherwise,} \end{cases} \quad (5)$$

where $\mathcal{J}(\alpha, T) := \left\{ \tilde{\pi}(i) \in \mathcal{R}(T) \mid \frac{\tilde{v}_i(\tilde{\pi}(i))}{\tilde{v}_i(T)} \geq \frac{1}{\alpha} \right\}$.

Proof From Theorem 13 of the main paper, we have the following

$$\mathbb{P}[\cup_{i \in T} v_i(\tilde{\pi}(i)) \geq v_i(T)] \geq (1 - \tilde{\epsilon})h_T(p, \alpha).$$

To get $h_T(p, \alpha) := \mathbb{P}[F(T, \tilde{\pi})]$ we consider two cases viz. $\mathcal{J}(\alpha, T) = \emptyset$, and $\mathcal{J}(\alpha, T) \neq \emptyset$. For these cases, we identify the possible noise values elements of $F(T, \tilde{\pi})$.

- **Case 01:** $[\mathcal{J}(\alpha, T) = \emptyset]$. In this case, we have the following possibilities:

- $\alpha(\tilde{\pi}(i)) = \alpha$, $\forall \tilde{\pi}(i) \in \mathcal{R}(T)$, and $\alpha(T) = \alpha$. Probability of such a choice of α 's is

$$p^{|\mathcal{R}(T)| + 1}. \quad (6)$$

- $\alpha(\tilde{\pi}(i)) = 1$ for $k \in \{0, 1, \dots, |\mathcal{R}(T)|\}$ coalitions $\tilde{\pi}(i) \in \mathcal{R}(T)$, and $\alpha(\tilde{\pi}(i)) = \alpha$ for remaining $|\mathcal{R}(T)| - k$ coalitions. Moreover, $\alpha(T) = 1$. Probability of such choice of α 's is $(1-p)^k p^{|\mathcal{R}(T)|-k} (1-p)$. Further, there are $\binom{|\mathcal{R}(T)|}{k}$ ways of selecting k coalitions $\tilde{\pi}(i)$ from $\mathcal{R}(T)$. Thus, the overall probability is

$$\sum_{k=0}^{|\mathcal{R}(T)|} \binom{|\mathcal{R}(T)|}{k} (1-p)^k p^{|\mathcal{R}(T)|-k} (1-p) = 1-p. \quad (7)$$

- **Case 02:** $[\mathcal{J}(\alpha, T) \neq \emptyset]$. In addition to the above possible cases, we have a few other cases:

- $\alpha(\tilde{\pi}(i)) = 1$ for any $k \in \{1, 2, \dots, |\mathcal{J}(\alpha, T)|\}$ coalitions $\tilde{\pi}(i) \in \mathcal{J}(\alpha, T)$. Moreover, $\alpha(\tilde{\pi}(i)) = \alpha$ for remaining coalitions in $\mathcal{R}(T)$. Also, $\alpha(T) = \alpha$. Probability of such choice of α 's is $(1-p)^k p^{|\mathcal{R}(T)|-k} p = (1-p)^k p^{|\mathcal{R}(T)|-k+1}$. And there are $\binom{|\mathcal{J}(\alpha, T)|}{k}$ ways of selecting k coalitions $\tilde{\pi}(i) \in \mathcal{J}(\alpha, T)$. Thus the overall probability is

$$\sum_{k=1}^{|\mathcal{J}(\alpha, T)|} \binom{|\mathcal{J}(\alpha, T)|}{k} (1-p)^k p^{|\mathcal{R}(T)|-k+1}. \quad (8)$$

The probability $\mathbb{P}[F(T, \tilde{\pi})]$ is obtained by adding probabilities given in Equations (6), (7) and (8).

$$\begin{aligned} \mathbb{P}[F(T, \tilde{\pi})] &= p^{|\mathcal{R}(T)|+1} + (1-p) + \sum_{k=1}^{|\mathcal{J}(\alpha, T)|} \binom{|\mathcal{J}(\alpha, T)|}{k} (1-p)^k p^{|\mathcal{R}(T)|-k+1} \\ &= p^{|\mathcal{R}(T)|+1} + (1-p) + p^{|\mathcal{R}(T)|-|\mathcal{J}(\alpha, T)|+1} \left[1 - p^{|\mathcal{J}(\alpha, T)|} \right] \\ &= (1-p) + p^{|\mathcal{R}(T)|-|\mathcal{J}(\alpha, T)|+1}. \end{aligned}$$

This ends the proof. ■

If $\tilde{\pi}(i) \neq T$ for at least one $i \in T$, then $h_T(p, \alpha) = 1$, $\forall \alpha$ if $p = 0$ or $p = 1$. That is, if the value of all coalitions are retained, or if value of all of them are inflated by α , then coalition $T \succeq_i \tilde{\pi}(i)$, and $T \succeq'_i \tilde{\pi}(i)$ for all $i \in T$. Thus, neither noise-free nor noisy game will have $\tilde{\pi}$ as PAC stable outcome. Moreover, if we allow $h_T(p, \alpha) = \eta$ for some user-given satisfaction η , we get a noise set in accordance to the Remark 14 in the main paper. In this case, the noise set also depends on $|\mathcal{R}(T)|$, and $|\mathcal{J}(\alpha, T)|$ for coalition T . Hence, the partition is η noise-robust non core-stable for the noise set $I^*(T, \eta)$.

3. Proof of Theorem 15 of main paper

Theorem: For n agent noisy hedonic game (N, \tilde{v}) with $\mathcal{N}_{sp} = \{\alpha_1, \alpha_2, \dots, \alpha_l\}$, the agreement probability $h_T(\mathbf{p}, \boldsymbol{\alpha})$ is given by:

$$h_T(\mathbf{p}, \boldsymbol{\alpha}) = \begin{cases} 1, & \text{if } \tilde{\pi}(i) = T, \forall i \in T, \\ \sum_{r,s \in [l]: \alpha_r > \alpha_s} p_r^{|\mathcal{R}(T)| - |\mathcal{J}(\alpha_r, \alpha_s, T)| + 1} \times \{(p_s + p_r)^{|\mathcal{J}(\alpha_r, \alpha_s, T)|} - p_r^{|\mathcal{J}(\alpha_r, \alpha_s, T)|}\} \\ + \sum_{a=1}^l p_a \left(\sum_{b=a}^l p_b \right)^{|\mathcal{R}(T)|}, & \text{otherwise.} \end{cases}$$

Proof We will prove this via Mathematical induction on the noise support $l \geq 2$. Clearly, this is true for $l = 2$ (from Lemma 3 above). Let us assume that it is true for $l = k$, i.e.; there are sets

$$\mathcal{J}(\alpha_r, \alpha_s, T) = \left\{ \tilde{\pi}(i) \in \mathcal{R}(T) \mid \frac{\tilde{v}_i(\tilde{\pi}(i))}{\tilde{v}_i(T)} \geq \frac{\alpha_s}{\alpha_r} \right\},$$

such that the support $\alpha(S) = \{\alpha_1, \dots, \alpha_k\}$, $\forall S \subseteq N$ where $\alpha_s < \alpha_r$, $\forall 1 \leq s < r \leq k$. For this k we have $f_T(p_j, \alpha_j : j \in [k]) =: h_T(\mathbf{p}, \boldsymbol{\alpha})$ (by assumption)

$$h_T(\mathbf{p}, \boldsymbol{\alpha}) = \sum_{a=1}^k p_a \left(\sum_{b=a}^k p_b \right)^{|\mathcal{R}(T)|} + \sum_{r,s \in [k]: \alpha_r > \alpha_s} p_r^{|\mathcal{R}(T)| - |\mathcal{J}(\alpha_r, \alpha_s, T)| + 1} ((p_r + p_s)^{|\mathcal{J}(\alpha_r, \alpha_s, T)|} - p_r^{|\mathcal{J}(\alpha_r, \alpha_s, T)|}).$$

We will now show that this is true for $l = k + 1$. To this end define $\mathcal{J}(\alpha_{k+1}, \alpha_s, T)$ for all $s \in [k]$ such that $\alpha_{k+1} > \alpha_s$

$$\mathcal{J}(\alpha_{k+1}, \alpha_s, T) = \left\{ \tilde{\pi}(i) \in \mathcal{R}(T) \mid \frac{\tilde{v}_i(\tilde{\pi}(i))}{\tilde{v}_i(T)} \geq \frac{\alpha_s}{\alpha_{k+1}} \right\}.$$

Now, there are two cases, $\mathcal{J}(\alpha_{k+1}, \alpha_s, T) = \emptyset$, $\forall \alpha_s, s \in [k]$, or $\mathcal{J}(\alpha_{k+1}, \alpha_s, T) \neq \emptyset$ for at least for one $s \in [k]$.

Case 01: $[\mathcal{J}(\alpha_{k+1}, \alpha_s, T) = \emptyset, \forall \alpha_s, s \in [k]]$. Apart from the existing $\{\alpha(\tilde{\pi}(i))\}_{\tilde{\pi}(i) \in \mathcal{R}(T)}$ and $\alpha(T)$ for k support case, with this extra $k + 1$, it will also have $\alpha(T) = \alpha_{k+1}$ and $\alpha(\tilde{\pi}(i)) = \alpha_{k+1}$, $\forall \tilde{\pi}(i) \in \mathcal{R}(T)$. The probability of such extra α 's is $p_{k+1} \left(\sum_{b=k+1}^{k+1} p_b \right)^{|\mathcal{R}(T)|}$. Therefore, the overall probability is

$$\sum_{a=1}^k p_a \left(\sum_{b=a}^k p_b \right)^{|\mathcal{R}(T)|} + p_{k+1} \left(\sum_{b=k+1}^{k+1} p_b \right)^{|\mathcal{R}(T)|} = \sum_{a=1}^{k+1} p_a \left(\sum_{b=a}^{k+1} p_b \right)^{|\mathcal{R}(T)|}.$$

Case 02: $[\mathcal{J}(\alpha_{k+1}, \alpha_s, T) \neq \emptyset \text{ for at least one } s \in [k]]$. In this case, apart from all $\alpha(T)$ and $\{\alpha(\tilde{\pi}(i))\}_{\tilde{\pi}(i) \in \mathcal{J}(\alpha_r, \alpha_s, T)}$, we have $\{\alpha(\tilde{\pi}(i))\}_{\forall \tilde{\pi}(i) \in \mathcal{J}(\alpha_{k+1}, \alpha_r, T), \alpha(T)}$. For this set, the possible pairs are such that $\alpha(\tilde{\pi}(i)) = \alpha_r$, $\forall \tilde{\pi}(i) \in \mathcal{R}(T) \setminus \mathcal{J}(\alpha_{k+1}, \alpha_r, T)$, and $\alpha(T) = \alpha_{k+1}$. Thus, their combined probability is $p_{k+1}^{|\mathcal{R}(T)| - |\mathcal{J}(\alpha_{k+1}, \alpha_s, T)| + 1} ((p_{k+1} - p_s)^{|\mathcal{J}(\alpha_{k+1}, \alpha_r, T)|} - p_{k+1}^{|\mathcal{J}(\alpha_{k+1}, \alpha_r, T)|})$. Hence for $k + 1$ support, the probability is

$$\sum_{r,s \in [k]: \alpha_r > \alpha_s} p_r^{|\mathcal{R}(T)| - |\mathcal{J}(\alpha_r, \alpha_r, T)| + 1} ((p_r + p_s)^{|\mathcal{J}(\alpha_r, \alpha_s, T)|} - p_r^{|\mathcal{J}(\alpha_r, \alpha_s, T)|})$$

$$+ p_{k+1}^{|\mathcal{R}(T)|-|\mathcal{J}(\alpha_{k+1},\alpha_s,T)|+1} \left((p_{k+1} + p_s)^{|\mathcal{I}(\alpha_{k+1},\alpha_s,T)|} - p_{k+1}^{|\mathcal{J}(\alpha_{k+1},\alpha_s,T)|} \right).$$

From case 01 and case 02 with $k + 1$ support, we have

$$\begin{aligned} h_T(p_j, \alpha_j; j \in [k + 1]) &= \sum_{r,s \in [k+1]; \alpha_r > \alpha_s} p_r^{|\mathcal{R}(T)|-|\mathcal{J}(\alpha_r,\alpha_s,T)|+1} \left((p_r + p_s)^{|\mathcal{J}(\alpha_r,\alpha_s,T)|} - p_r^{|\mathcal{J}(\alpha_r,\alpha_s,T)|} \right) \\ &\quad + \sum_{a=1}^{k+1} p_a \left(\sum_{b=a}^{k+1} p_b \right)^{|\mathcal{R}(T)|}. \end{aligned}$$

Furthermore, it is true for $k + 1$ support. Thus, from the principle of Mathematical induction, this is true for any $l \geq 2$. \blacksquare

4. 2 agent 2 support model

In this Section, we will provide further details about the 2 agents' full information noisy game with 2 support of the noise distribution. First, we consider the following noisy game.

$$\tilde{v}_1(12) > \tilde{v}_1(1); \quad \tilde{v}_2(12) > \tilde{v}_2(2). \quad (\text{game 1})$$

We also consider the other possible noisy games with 2 agents in later subsections.

4.1. Proof of Lemma 18 of main paper

Lemma: For noisy [game 1](#) with complete information on \tilde{v} and $\mathcal{N}_{sp} = \{1, \alpha\}$ we have

$$\mathbb{P}[\pi = \tilde{\pi} \mid \text{game 1}] = \begin{cases} 1 - p(1 - p^2), & \text{if } \alpha \geq \bar{r} \\ 1 - p(1 - p), & \text{if } \underline{r} \leq \alpha < \bar{r} \\ 1, & \text{if } \alpha < \underline{r}, \end{cases} \quad (9)$$

where $\bar{r} = \max \left\{ \frac{\tilde{v}_1(12)}{\tilde{v}_1(1)}, \frac{\tilde{v}_2(12)}{\tilde{v}_2(2)} \right\}$, and $\underline{r} = \min \left\{ \frac{\tilde{v}_1(12)}{\tilde{v}_1(1)}, \frac{\tilde{v}_2(12)}{\tilde{v}_2(2)} \right\}$.

Also, this prediction probability $\mathbb{P}[\pi = \tilde{\pi} \mid \text{game 1}]$ is convex in p . So, while the minimal value for $\mathbb{P}[\pi = \tilde{\pi} \mid \text{game 1}]$ occurs for noise probabilities around $p = 0.5$ (depending on α, \bar{r} and \underline{r}), the maximal value of it is 1 at $p = 0$ and $p = 1$.

Proof For noisy [game 1](#), we have $\tilde{\pi} = N$. Now, consider the noise support $\mathcal{N}_{sp} = \{1, \alpha\}$, where $\alpha > 1$ such that $\mathbb{P}[\alpha(S) = \alpha] = p = 1 - \mathbb{P}[\alpha(S) = 1]$, for some fixed and unknown p . Given noisy [game 1](#), there are 8 possible combinations of α 's (because each coalition has two options). We will now enumerate all such possibilities:

1. $\alpha(1) = 1; \alpha(2) = 1; \alpha(12) = 1$. The probability of such alpha is $(1 - p)^3$. Thus, the noise-free values are $v_1(1) = \tilde{v}_1(1); v_2(2) = \tilde{v}_2(2), v_1(12) = \tilde{v}_1(12)$ and $v_2(12) = \tilde{v}_2(12)$. Therefore, The noise-free game is:

$$v_1(12) > v_1(1); \quad v_2(12) > v_2(2).$$

From this game we have $\pi = \tilde{\pi} = N$.

2. $\alpha(1) = 1; \alpha(2) = 1; \alpha(12) = \alpha$ Probability of such alpha's is $p(1-p)^2$. Thus the actual values are $v_1(1) = \tilde{v}_1(1); v_2(2) = \tilde{v}_2(2), v_1(12) = \frac{\tilde{v}_1(12)}{\alpha}$ and $v_2(12) = \frac{\tilde{v}_2(12)}{\alpha}$. Therefore, the actual preferences will depend on the relative values of α and \tilde{v} . If α and \tilde{v} 's are such that $\frac{\tilde{v}_1(12)}{\alpha} > \tilde{v}_1(1)$ and $\frac{\tilde{v}_2(12)}{\alpha} > \tilde{v}_2(2)$, then $\pi = N$, otherwise $\pi = \{\{1\}, \{2\}\}$.

3. $\alpha(1) = 1; \alpha(2) = \alpha; \alpha(12) = 1$. The probability of such alpha is $p(1-p)^2$. Thus, the actual values are $v_1(1) = \tilde{v}_1(1); v_2(2) = \frac{\tilde{v}_2(2)}{\alpha}, v_1(12) = \tilde{v}_1(12)$ and $v_2(12) = \tilde{v}_2(12)$. Since, $\tilde{v}_2(12) > \tilde{v}_2(2) > \frac{\tilde{v}_2(2)}{\alpha}$. The noise-free game is:

$$v_1(12) > v_1(1); v_2(12) > v_2(2).$$

So, we have $\pi = \tilde{\pi} = N$.

4. $\alpha(1) = \alpha; \alpha(2) = 1; \alpha(12) = 1$. Probability of such alpha's is $p(1-p)^2$. Thus, the actual values are $v_1(1) = \frac{\tilde{v}_1(1)}{\alpha}; v_2(2) = \tilde{v}_2(2), v_1(12) = \tilde{v}_1(12)$ and $v_2(12) = \tilde{v}_2(12)$. Since, $\tilde{v}_1(12) > \tilde{v}_1(1) > \frac{\tilde{v}_1(1)}{\alpha}$. Therefore The noise-free game is:

$$v_1(12) > v_1(1); v_2(12) > v_2(2).$$

From this game we have $\pi = \tilde{\pi} = N$.

5. $\alpha(1) = 1; \alpha(2) = \alpha; \alpha(12) = \alpha$. The probability of this alpha is $p^2(1-p)$. Thus, the actual values are $v_1(1) = \tilde{v}_1(1); v_2(2) = \frac{\tilde{v}_2(2)}{\alpha}, v_1(12) = \frac{\tilde{v}_1(12)}{\alpha}$ and $v_2(12) = \frac{\tilde{v}_2(12)}{\alpha}$. The actual preferences will depend on the relative values of α and \tilde{v} . If α and \tilde{v} 's are such that $\frac{\tilde{v}_1(12)}{\alpha} > \tilde{v}_1(1)$, then $\pi = N$, otherwise $\pi = \{\{1\}, \{2\}\}$.

6. $\alpha(1) = \alpha; \alpha(2) = 1; \alpha(12) = \alpha$. The probability of such alpha is $p^2(1-p)$. Thus, the actual values are $v_1(1) = \frac{\tilde{v}_1(1)}{\alpha}, v_2(2) = \tilde{v}_2(2); v_1(12) = \frac{\tilde{v}_1(12)}{\alpha}$ and $v_2(12) = \frac{\tilde{v}_2(12)}{\alpha}$. The actual preferences will depend on the relative values of α and \tilde{v} . If α and \tilde{v} 's are such that $\frac{\tilde{v}_2(12)}{\alpha} > \tilde{v}_2(2)$, then $\pi = N$ otherwise $\pi = \{\{1\}, \{2\}\}$.

7. $\alpha(1) = \alpha; \alpha(2) = \alpha; \alpha(12) = 1$. Probability of such alpha's is $p^2(1-p)$. Thus, the actual values are $v_1(1) = \frac{\tilde{v}_1(1)}{\alpha}, v_2(2) = \frac{\tilde{v}_2(2)}{\alpha}; v_1(12) = \tilde{v}_1(12)$ and $v_2(12) = \tilde{v}_2(12)$. Since, $\tilde{v}_1(12) > \tilde{v}_1(1) > \frac{\tilde{v}_1(1)}{\alpha}$. and, $\tilde{v}_2(12) > \tilde{v}_2(2) > \frac{\tilde{v}_2(2)}{\alpha}$. The noise-free game is:

$$v_1(12) > v_1(1); v_2(12) > v_2(2).$$

From this game we have $\pi = \tilde{\pi} = N$.

8. $\alpha(1) = \alpha; \alpha(2) = \alpha; \alpha(12) = \alpha$. The probability of such alpha is p^3 . Thus, the actual values are $v_1(1) = \frac{\tilde{v}_1(1)}{\alpha}, v_2(2) = \frac{\tilde{v}_2(2)}{\alpha}; v_1(12) = \frac{\tilde{v}_1(12)}{\alpha}$ and $v_2(12) = \frac{\tilde{v}_2(12)}{\alpha}$. Therefore, the noise-free game is:

$$v_1(12) > v_1(1); v_2(12) > v_2(2).$$

From this game it is clear that $\pi = \tilde{\pi} = N$.

Recall, $\bar{r} = \max \left\{ \frac{\tilde{v}_1(12)}{\tilde{v}_1(1)}, \frac{\tilde{v}_2(12)}{\tilde{v}_2(2)} \right\}$, and $\underline{r} = \min \left\{ \frac{\tilde{v}_1(12)}{\tilde{v}_1(1)}, \frac{\tilde{v}_2(12)}{\tilde{v}_2(2)} \right\}$. Out of 8 cases there are 5 cases (case 1,3,4,7,8) in which the grand coalition $\pi = \tilde{\pi} = N$ is formed in noise-free game. In these conditions, the relative value of $\tilde{v}_1(\cdot), \tilde{v}_2(\cdot)$ should satisfy $\alpha \geq \bar{r}$, and this constitute the first expression $p^3 + p^2(1-p) + 2p(1-p)^2 + (1-p)^3$ of $\mathbb{P}[\pi = \tilde{\pi} \mid \text{game 1}]$. Apart from this, if the inflation interval is $\underline{r} \leq \alpha < \bar{r}$, then $\pi = \tilde{\pi} = N$ is also possible from case (6) with probability $p^2(1-p)$. Thus, $p^2(1-p)$ will be added to the above prediction probability. So, we have $\mathbb{P}[\pi = \tilde{\pi} \mid \text{game 1}]$ corresponding to it. Moreover, finally, if $\alpha < \underline{r}$, all cases are allowable, and hence the grand coalition will always form in the noise-free game. Thus,

$$\mathbb{P}[\pi = \tilde{\pi} \mid \text{game 1}] = \begin{cases} p^3 + p^2(1-p) + 2p(1-p)^2 + (1-p)^3, & \text{if } \alpha \geq \bar{r} \\ p^3 + 2p^2(1-p) + 2p(1-p)^2 + (1-p)^3, & \text{if } \underline{r} \leq \alpha < \bar{r} \\ 1, & \text{if } \alpha < \underline{r}. \end{cases} \quad (10)$$

Simplifying these polynomials, we have

$$\mathbb{P}[\pi = \tilde{\pi} \mid \text{game 1}] = \begin{cases} 1 - p(1-p^2), & \text{if } \alpha \geq \bar{r} \\ 1 - p(1-p), & \text{if } \underline{r} \leq \alpha < \bar{r} \\ 1, & \text{if } \alpha < \underline{r}. \end{cases} \quad (11)$$

This ends the proof. ■

If we allow some user given satisfaction ζ on the prediction probability, i.e., $\mathbb{P}[\pi = \tilde{\pi} \mid \text{game 1}] = \zeta$, we get the following noise interval

$$I^*(\zeta = 0.9) = \begin{cases} [0, 0.101] \cup [0.946, 1], & \text{if } \alpha \geq \bar{r}; \\ [0, 0.113] \cup [0.887, 1], & \text{if } \underline{r} \leq \alpha < \bar{r} \\ 1, & \text{if } \alpha < \underline{r}. \end{cases} \quad (12)$$

4.2. Details of the other 2 agent noisy games

Here we will give the prediction probabilities for other possible noisy games with 2 agents and 2 noise support.

4.2.1. BOTH AGENTS PREFER STAYING ALONE IN NOISY GAME

As opposed to the noisy [game 1](#), in noisy [game 2](#) both agents prefer to stay alone. The noisy preferences of agents are as follows:

$$\tilde{v}_1(1) > \tilde{v}_1(12); \quad \tilde{v}_2(2) > \tilde{v}_2(12). \quad (\text{game 2})$$

Clearly $\tilde{\pi} = \{\{1\}, \{2\}\} \neq N$ is the core-stable outcome. The following lemma provides prediction probability, $\mathbb{P}[\pi = \tilde{\pi} \mid \text{game 2}]$ for noisy [game 2](#).

Lemma 4 *For noisy [game 2](#) with full information of \tilde{v} 's, the prediction probability that unknown noise-free game has $\pi = \tilde{\pi}$ as a core-stable outcome is*

$$\mathbb{P}[\pi = \tilde{\pi} \mid \text{game 2}] = \begin{cases} 1 - p^2(1-p), & \text{if } \frac{1}{\alpha} < \underline{r} \\ 1, & \text{if } \frac{1}{\alpha} \geq \underline{r}. \end{cases} \quad (13)$$

Moreover, the minimal and maximal values of above prediction probability are 0.85 (when $p = 2/3$), and 1, respectively.

Similar to [game 1](#), the probability of formation of partition $\pi = \{\{1\}, \{2\}\}$ in an *unknown* noise-free game is always more than 0.85. So, the safety value is 0.85. The prediction probability is 1 when $\frac{1}{\alpha} \geq \underline{r}$ for any noise probability p . Moreover, for some user-given satisfaction ζ , we obtain the corresponding p by setting $\mathbb{P}[\pi = \{\{1\}, \{2\}\} \mid \text{game 2}] = \zeta$. In particular, we have

$$I^*(\zeta = 0.9) = \begin{cases} [0, 0.413] \cup [0.867, 1], & \text{if } \frac{1}{\alpha} < \underline{r} \\ [0, 1], & \frac{1}{\alpha} \geq \underline{r}. \end{cases} \quad (14)$$

It is easy to see that the allowable p is larger than the interval given in Equation (12) for [game 1](#). So, the partition $\tilde{\pi} = \{\{1\}, \{2\}\}$ is noise robust for larger number of inflation probabilities p . Again the noise set will shrink if we increase the satisfaction ζ .

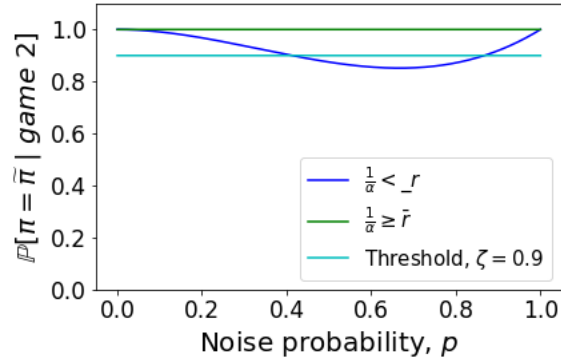


Figure 1: The prediction probability $\mathbb{P}[\pi = \tilde{\pi} \mid \text{game 2}]$. For $\zeta = 0.9$, the noise regimes are given in Equation (14).

4.2.2. AGENT 1 PREFERS TO STAY ALONE AND AGENT 2 PREFERS GRAND COALITION IN NOISY GAME

Now, we consider a noisy game where agent 1 prefers to stay alone, whereas agent 2 prefers the grand coalition. In particular, the preferences in the noisy game are

$$\tilde{v}_1(1) > \tilde{v}_1(12); \quad \tilde{v}_2(12) > \tilde{v}_2(2). \quad (\text{game 3})$$

Again $\tilde{\pi} = \{\{1\}, \{2\}\} \neq N$ is noisy core-stable outcome. The prediction probability, $\mathbb{P}[\pi = \tilde{\pi} \mid \text{game 3}]$ is given in the Lemma below.

Lemma 5 For noisy [game 3](#) with full information of \tilde{v} 's, the prediction probability that unknown noise-free game has $\pi = \tilde{\pi}$ as a core-stable outcome is given by:

$$\mathbb{P}[\pi = \tilde{\pi} \mid \text{game 3}] = \begin{cases} 1 - p(1 - p), & \text{if } \frac{1}{\alpha} < \frac{\tilde{v}_1(12)}{\tilde{v}_1(1)} \\ 1, & \text{if } \frac{1}{\alpha} \geq \frac{\tilde{v}_1(12)}{\tilde{v}_1(1)}. \end{cases} \quad (15)$$

Moreover, the minimal and maximal values of above prediction probability are 0.75 (when $p = 0.5$), and 1, respectively.

Similar to [game 1](#) and [game 2](#) the probability of formation of partition $\pi = \{\{1\}, \{2\}\}$ in an *unknown* noise-free game is always more than 0.75 that is the safety value for [game 3](#). The prediction probability is 1 when $\frac{1}{\alpha} \geq \frac{\tilde{v}_1(12)}{\tilde{v}_1(1)}$ for any noise probability p . Moreover, for some user-given satisfaction, ζ we obtain the corresponding p by setting $\mathbb{P}[\pi = \{\{1\}, \{2\}\} \mid \text{game 3}] = \zeta$. In particular,

$$I^*(\zeta = 0.9) = \begin{cases} [0, 0.113] \cup [0.887, 1], & \text{if } \frac{1}{\alpha} < \frac{\tilde{v}_1(12)}{\tilde{v}_1(1)} \\ [0, 1], & \text{if } \frac{1}{\alpha} \geq \frac{\tilde{v}_1(12)}{\tilde{v}_1(1)}. \end{cases} \quad (16)$$

The following figure shows the prediction probabilities for [game 3](#).

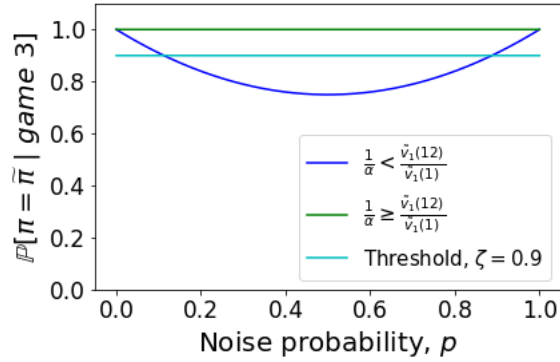


Figure 2: The prediction probability $\mathbb{P}[\tilde{\pi} = \pi \mid \text{game 3}]$. For $\zeta = 0.9$, we obtain the noise regimes as given in Equation (16).

4.2.3. AGENT 1 PREFERS GRAND COALITION AND AGENT 2 PREFERS TO STAY ALONE

Finally, consider a noisy game symmetric to [game 3](#). Here agent 1 prefers a grand coalition, and agent 2 prefers to stay alone. In particular, we have the following preferences.

$$\tilde{v}_1(12) > \tilde{v}_1(1); \quad \tilde{v}_2(2) > \tilde{v}_2(12). \quad (\text{game 4})$$

Again $\tilde{\pi} = \{\{1\}, \{2\}\} \neq N$ is a noisy core-stable outcome. In the following lemma, we find the prediction probability when noisy [game 4](#) is considered.

Lemma 6 *For noisy [game 4](#) with full information of \tilde{v} 's, the prediction probability that noise-free game has $\pi = \tilde{\pi}$ as as core-stable outcome is given by:*

$$\mathbb{P}[\pi = \tilde{\pi} \mid \text{game 4}] = \begin{cases} 1 - p(1 - p), & \text{if } \frac{1}{\alpha} < \frac{\tilde{v}_2(12)}{\tilde{v}_2(2)} \\ 1, & \text{if } \frac{1}{\alpha} \geq \frac{\tilde{v}_2(12)}{\tilde{v}_2(2)}. \end{cases} \quad (17)$$

So, the minimal and maximal values of above prediction probability are 0.75 (when $p = 0.5$) and 1 respectively.

In this case also, the noise regime can be obtained using $\mathbb{P}[\pi = \tilde{\pi} \mid \text{game 4}] = \zeta$. In particular,

$$I^*(\zeta = 0.9) = \begin{cases} [0, 0.113] \cup [0.887, 1], & \text{if } \frac{1}{\alpha} < \frac{\tilde{v}_2(12)}{\tilde{v}_2(2)} \\ [0, 1], & \text{if } \frac{1}{\alpha} \geq \frac{\tilde{v}_2(12)}{\tilde{v}_2(2)}. \end{cases} \quad (18)$$

Figure 3 shows the prediction probabilities for [game 4](#).

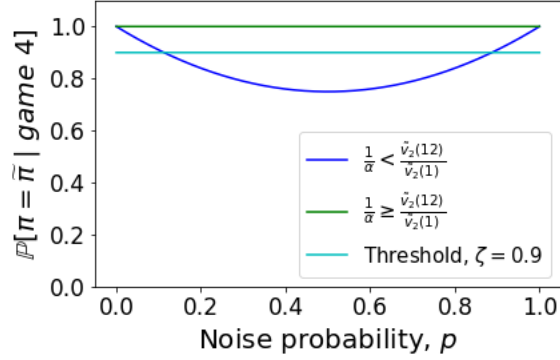


Figure 3: The prediction probability $\mathbb{P}[\pi = \tilde{\pi} \mid \text{game 4}]$. For $\zeta = 0.9$, we obtain the noise regimes as given in Equation (18).

5. 2 agents 3 support noise model

In this section, we consider two player noisy hedonic game with three support noise model, i.e., $\mathcal{N}_{sp} = \{1, \alpha_1, \alpha_2\}$, with $\alpha_1 > 1$, and $\alpha_2 < 1$. Note that $\alpha_1, \alpha_2 > 0$. Let $\mathbb{P}[\alpha(S) = \alpha_1] = p_1$; $\mathbb{P}[\alpha(S) = \alpha_2] = p_2$; and $\mathbb{P}[\alpha(S) = 1] = 1 - p_1 - p_2$. That is the value of each coalition is either inflated with probability p_1 , or deflated with probability p_2 or retained with probability $1 - p_1 - p_2$. The following lemma provides the prediction probability for [game 1](#).

5.1. Proof of Lemma 20 of main paper

Lemma: For the 3 support noise model the prediction probability $\mathbb{P}[\pi = \tilde{\pi} \mid \text{game 1}]$ is

$$\mathbb{P}[\pi = \tilde{\pi} \mid \text{game 1}] = \begin{cases} g(p_1, p_2), & \text{if } \alpha_1 \geq \bar{r} ; \frac{1}{\alpha_2} \geq \bar{r} ; \frac{\alpha_1}{\alpha_2} \geq \bar{r} \\ 1, & \text{if } \alpha_1 < \underline{r} ; \frac{1}{\alpha_2} < \underline{r} ; \frac{\alpha_1}{\alpha_2} < \underline{r} \end{cases} \quad (19)$$

where $g(p_1, p_2) = p_1^3 + p_2^3 + 2(p_1(1 - p_1 - p_2)^2 + p_2^2(1 - p_1 - p_2) + p_1p_2(1 - p_1 - p_2) + p_1p_2^2) + p_1^2p_2 + p_1^2(1 - p_1 - p_2) + p_2(1 - p_1 - p_2)^2 + (1 - p_1 - p_2)^3$.

Proof For [game 1](#), with $l = 3$ support of noise there are 27 possible cases for α 's. Since there are 3 coalitions, each coalition's value can either be retained, inflated by α_1 , or deflated by α_2 . We will now enumerate all of them:

1. $\alpha(1) = 1; \alpha(2) = 1; \alpha(12) = 1$ Probability of such alpha's is $(1 - p_1 - p_2)^3$. Thus, the actual values are $v_1(1) = \tilde{v}_1(1); v_2(2) = \tilde{v}_2(2), v_1(12) = \tilde{v}_1(12)$ and $v_2(12) = \tilde{v}_2(12)$. The noise-free game is: $v_1(12) > v_1(1); v_2(12) > v_2(2)$. So, $\pi = \tilde{\pi}$ in this case.
2. $\alpha(1) = 1; \alpha(2) = 1; \alpha(12) = \alpha_1$. Probability of such alpha's is $p_1(1 - p_1 - p_2)^2$. Thus, the actual values are $v_1(1) = \tilde{v}_1(1); v_2(2) = \tilde{v}_2(2), v_1(12) = \frac{\tilde{v}_1(12)}{\alpha_1}$ and $v_2(12) = \frac{\tilde{v}_2(12)}{\alpha_1}$. The noise-free game preferences are unclear; they will depend on the relative values of α_1 and \tilde{v} . If α_1 and \tilde{v} 's are such that $\frac{\tilde{v}_1(12)}{\alpha_1} > \tilde{v}_1(1)$ and $\frac{\tilde{v}_2(12)}{\alpha_1} > \tilde{v}_2(2)$ then $\pi = N$, otherwise $\pi = \{\{1\}, \{2\}\}$.
3. $\alpha(1) = 1; \alpha(2) = \alpha_1; \alpha(12) = 1$. Probability of such alpha's is $p_1(1 - p_1 - p_2)^2$. Thus, the actual values are $v_1(1) = \tilde{v}_1(1); v_2(2) = \frac{\tilde{v}_2(2)}{\alpha_1}, v_1(12) = \tilde{v}_1(12)$ and $v_2(12) = \tilde{v}_2(12)$. Since $\tilde{v}_2(12) > \tilde{v}_2(2) > \frac{\tilde{v}_2(2)}{\alpha_1}$. The noise-free game is: $v_1(12) > v_1(1); v_2(12) > v_2(2)$. So, $\pi = \tilde{\pi}$ in this case.
4. $\alpha(1) = \alpha_1; \alpha(2) = 1; \alpha(12) = 1$. Probability of such alpha's is $p_1(1 - p_1 - p_2)^2$. Thus, the actual values are $v_1(1) = \frac{\tilde{v}_1(1)}{\alpha_1}; v_2(2) = \tilde{v}_2(2), v_1(12) = \tilde{v}_1(12)$ and $v_2(12) = \tilde{v}_2(12)$. Since, $\tilde{v}_1(12) > \tilde{v}_1(1) > \frac{\tilde{v}_1(1)}{\alpha_1}$. The noise-free game is: $v_1(12) > v_1(1); v_2(12) > v_2(2)$. So, $\pi = \tilde{\pi}$ in this case.
5. $\alpha(1) = 1; \alpha(2) = \alpha_1; \alpha(12) = \alpha_1$. Probability of such alpha's is $p_1^2(1 - p_1 - p_2)$. Thus, the actual values are $v_1(1) = \tilde{v}_1(1); v_2(2) = \frac{\tilde{v}_2(2)}{\alpha_1}, v_1(12) = \frac{\tilde{v}_1(12)}{\alpha_1}$ and $v_2(12) = \frac{\tilde{v}_2(12)}{\alpha_1}$. The noise-free game preferences are unclear; they will depend on the relative values of α_1 and \tilde{v} . If α_1 and \tilde{v} 's are such that $\frac{\tilde{v}_1(12)}{\alpha_1} > \tilde{v}_1(1)$, then $\pi = N$, otherwise $\pi = \{\{1\}, \{2\}\}$.
6. $\alpha(1) = \alpha_1; \alpha(2) = 1; \alpha(12) = \alpha_1$. Probability of such alpha's is $p_1^2(1 - p_1 - p_2)$. Thus, the actual values are $v_1(1) = \frac{\tilde{v}_1(1)}{\alpha_1}, v_2(2) = \tilde{v}_2(2); v_1(12) = \frac{\tilde{v}_1(12)}{\alpha_1}$ and $v_2(12) = \frac{\tilde{v}_2(12)}{\alpha_1}$. The noise-free game preferences are unclear; they will depend on the relative values of α_1 and \tilde{v} . If α_1 and \tilde{v} 's are such that $\frac{\tilde{v}_2(12)}{\alpha_1} > \tilde{v}_2(2)$, then $\pi = N$, otherwise $\pi = \{\{1\}, \{2\}\}$.
7. $\alpha(1) = \alpha_1; \alpha(2) = \alpha_1; \alpha(12) = 1$. Probability of such alpha's is $p_1^2(1 - p_1 - p_2)$. Thus, the actual values are $v_1(1) = \frac{\tilde{v}_1(1)}{\alpha_1}, v_2(2) = \frac{\tilde{v}_2(2)}{\alpha_1}; v_1(12) = \tilde{v}_1(12)$ and $v_2(12) = \tilde{v}_2(12)$. Since, $\tilde{v}_1(12) > \tilde{v}_1(1) > \frac{\tilde{v}_1(1)}{\alpha_1}$. and, $\tilde{v}_2(12) > \tilde{v}_2(2) > \frac{\tilde{v}_2(2)}{\alpha_1}$. The noise-free game is: $v_1(12) > v_1(1); v_2(12) > v_2(2)$. So, $\pi = \tilde{\pi}$ in this case.
8. $\alpha(1) = \alpha_1; \alpha(2) = \alpha_1; \alpha(12) = \alpha_1$. The probability of such alpha is p_1^3 . Thus, the actual values are $v_1(1) = \frac{\tilde{v}_1(1)}{\alpha_1}, v_2(2) = \frac{\tilde{v}_2(2)}{\alpha_1}; v_1(12) = \frac{\tilde{v}_1(12)}{\alpha_1}$ and $v_2(12) = \frac{\tilde{v}_2(12)}{\alpha_1}$. The noise-free game is: $v_1(12) > v_1(1); v_2(12) > v_2(2)$. So, $\pi = \tilde{\pi}$ in this case.
9. $\alpha(1) = 1; \alpha(2) = 1; \alpha(12) = \alpha_2$. Probability of such alpha's is $p_2(1 - p_1 - p_2)^2$. Thus, the actual values are $v_1(1) = \tilde{v}_1(1); v_2(2) = \tilde{v}_2(2), v_1(12) = \frac{\tilde{v}_1(12)}{\alpha_2}$ and $v_2(12) = \frac{\tilde{v}_2(12)}{\alpha_2}$. Since $\alpha_2 < 1$, thus $\frac{\tilde{v}_1(12)}{\alpha_2} > \tilde{v}_1(12) > \tilde{v}_1(1) = v_1(1)$. Similarly, $\frac{\tilde{v}_2(12)}{\alpha_2} > \tilde{v}_2(12) > \tilde{v}_2(2) = v_2(2)$. The noise-free game is: $v_1(12) > v_1(1); v_2(12) > v_2(2)$. So, $\pi = \tilde{\pi}$ in this case.

10. $\alpha(1) = 1; \alpha(2) = \alpha_2; \alpha(12) = 1$. Probability of these alpha's is $p_2(1 - p_1 - p_2)^2$. Thus, the actual values are $v_1(1) = \tilde{v}_1(1); v_2(2) = \frac{\tilde{v}_2(2)}{\alpha_2}, v_1(12) = \tilde{v}_1(12)$ and $v_2(12) = \tilde{v}_2(12)$. The noise-free game preferences are unclear; they will depend on the relative values of α_2 and \tilde{v} . If α_2 and \tilde{v} 's are such that $\tilde{v}_2(12) > \frac{\tilde{v}_2(2)}{\alpha_2}$ then $\pi = N$, otherwise $\pi = \{\{1\}, \{2\}\}$.
11. $\alpha(1) = \alpha_2; \alpha(2) = 1; \alpha(12) = 1$. Probability of such alpha's is $p_2(1 - p_1 - p_2)^2$. Thus, the actual values are $v_1(1) = \frac{\tilde{v}_1(1)}{\alpha_2}; v_2(2) = \tilde{v}_2(2), v_1(12) = \tilde{v}_1(12)$ and $v_2(12) = \tilde{v}_2(12)$. The noise-free game preferences are unclear; they will depend on the relative values of α_2 and \tilde{v} . If α_2 and \tilde{v} 's are such that $\tilde{v}_1(12) > \frac{\tilde{v}_1(1)}{\alpha_2}$ then $\pi = N$, otherwise $\pi = \{\{1\}, \{2\}\}$.
12. $\alpha(1) = 1; \alpha(2) = \alpha_2; \alpha(12) = \alpha_2$. probability of such alpha's is $p_2^2(1 - p_1 - p_2)$. Thus, the actual values are $v_1(1) = \tilde{v}_1(1); v_2(2) = \frac{\tilde{v}_2(2)}{\alpha_2}, v_1(12) = \frac{\tilde{v}_1(12)}{\alpha_2}$ and $v_2(12) = \frac{\tilde{v}_2(12)}{\alpha_2}$. Since $\alpha_2 < 1$, thus $\frac{\tilde{v}_1(12)}{\alpha_2} > \tilde{v}_1(12) > \tilde{v}_1(1) = v_1(1)$, and $\frac{\tilde{v}_2(12)}{\alpha_2} > \frac{\tilde{v}_2(2)}{\alpha_2}$. The noise-free game is: $v_1(12) > v_1(1); v_2(12) > v_2(2)$. So, $\pi = \tilde{\pi}$ in this case.
13. $\alpha(1) = \alpha_2; \alpha(2) = 1; \alpha(12) = \alpha_2$. Probability of such alpha's is $p_2^2(1 - p_1 - p_2)$. Thus, the actual values are $v_1(1) = \frac{\tilde{v}_1(1)}{\alpha_2}, v_2(2) = \tilde{v}_2(2); v_1(12) = \frac{\tilde{v}_1(12)}{\alpha_2}$ and $v_2(12) = \frac{\tilde{v}_2(12)}{\alpha_2}$. Since $\alpha_2 < 1$ thus $\frac{\tilde{v}_2(12)}{\alpha_2} > \tilde{v}_2(12) > \tilde{v}_2(2) = v_2(2)$, and $\frac{\tilde{v}_1(12)}{\alpha_2} > \frac{\tilde{v}_1(1)}{\alpha_2}$. The noise-free game is: $v_1(12) > v_1(1); v_2(12) > v_2(2)$. So, $\pi = \tilde{\pi}$ in this case.
14. $\alpha(1) = \alpha_2; \alpha(2) = \alpha_2; \alpha(12) = 1$. Probability of such alpha's is $p_2^2(1 - p_1 - p_2)$. Thus, the actual values are $v_1(1) = \frac{\tilde{v}_1(1)}{\alpha_2}, v_2(2) = \frac{\tilde{v}_2(2)}{\alpha_2}; v_1(12) = \tilde{v}_1(12)$ and $v_2(12) = \tilde{v}_2(12)$. The noise-free game preferences are unclear; they will depend on the relative values of α_2 and \tilde{v} . If α_2 and \tilde{v} 's are such that $\tilde{v}_1(12) > \frac{\tilde{v}_1(1)}{\alpha_2}$ and $\tilde{v}_1(12) > \frac{\tilde{v}_2(2)}{\alpha_2}$ then $\pi = N$, otherwise $\pi = \{\{1\}, \{2\}\}$.
15. $\alpha(1) = 1; \alpha(2) = \alpha_1; \alpha(12) = \alpha_2$. Probability of such alpha's is $p_1 p_2(1 - p_1 - p_2)$. Thus, the actual values are $v_1(1) = \tilde{v}_1(1), v_2(2) = \frac{\tilde{v}_2(2)}{\alpha_1}; v_1(12) = \frac{\tilde{v}_1(12)}{\alpha_2}$ and $v_2(12) = \frac{\tilde{v}_2(12)}{\alpha_2}$. Since $\frac{\tilde{v}_1(12)}{\alpha_2} > \tilde{v}_1(12) > \tilde{v}_1(1)$ and $\frac{\tilde{v}_2(12)}{\alpha_2} > \tilde{v}_2(12) > \tilde{v}_2(2) > \frac{\tilde{v}_2(2)}{\alpha_1}$. The noise-free game is: $v_1(12) > v_1(1); v_2(12) > v_2(2)$. So, $\pi = \tilde{\pi}$ in this case.
16. $\alpha(1) = 1; \alpha(2) = \alpha_2; \alpha(12) = \alpha_1$. Probability of such alpha's is $p_1 p_2(1 - p_1 - p_2)$. Thus, actual values are $v_1(1) = \tilde{v}_1(1), v_2(2) = \frac{\tilde{v}_2(2)}{\alpha_2}; v_1(12) = \frac{\tilde{v}_1(12)}{\alpha_1}$ and $v_2(12) = \frac{\tilde{v}_2(12)}{\alpha_1}$. The noise-free game preferences are unclear; it will depend on the relative values α_1, α_2 and \tilde{v} . If α_1, α_2 and \tilde{v} 's are such that $\frac{\tilde{v}_1(12)}{\alpha_1} > \tilde{v}_1(1)$ and $\frac{\tilde{v}_2(12)}{\alpha_1} > \frac{\tilde{v}_2(2)}{\alpha_2}$ then $\pi = N$, otherwise $\pi = \{\{1\}, \{2\}\}$.
17. $\alpha(1) = \alpha_1; \alpha(2) = 1; \alpha(12) = \alpha_2$. Probability of such alpha's is $p_1 p_2(1 - p_1 - p_2)$. Thus, the actual values are $v_1(1) = \frac{\tilde{v}_1(1)}{\alpha_1}, v_2(2) = \tilde{v}_2(2); v_1(12) = \frac{\tilde{v}_1(12)}{\alpha_2}$ and $v_2(12) = \frac{\tilde{v}_2(12)}{\alpha_2}$. Since $\frac{\tilde{v}_1(12)}{\alpha_2} > \tilde{v}_1(12) > \tilde{v}_1(1)$ and $\frac{\tilde{v}_2(12)}{\alpha_2} > \tilde{v}_2(12) > \tilde{v}_2(2)$. The noise-free game is: $v_1(12) > v_1(1); v_2(12) > v_2(2)$. So, $\pi = \tilde{\pi}$ in this case.
18. $\alpha(1) = \alpha_2; \alpha(2) = 1; \alpha(12) = \alpha_1$. Probability of such alpha's is $p_1 p_2(1 - p_1 - p_2)$. Thus, the actual values are $v_1(1) = \frac{\tilde{v}_1(1)}{\alpha_2}, v_2(2) = \tilde{v}_2(2); v_1(12) = \frac{\tilde{v}_1(12)}{\alpha_1}$ and $v_2(12) = \frac{\tilde{v}_2(12)}{\alpha_1}$.

The noise-free game preferences are unclear; it will depend on the relative values of α_1 , α_2 , and \tilde{v} . If α_1 , α_2 and \tilde{v} 's are such that $\frac{\tilde{v}_1(12)}{\alpha_1} > \frac{\tilde{v}_1(1)}{\alpha_2}$ and $\frac{\tilde{v}_2(12)}{\alpha_1} > \tilde{v}_2(2)$ then $\pi = N$, otherwise $\pi = \{\{1\}, \{2\}\}$.

19. $\alpha(1) = \alpha_1; \alpha(2) = \alpha_2; \alpha(12) = 1$. Probability of such alpha's is $p_1 p_2 (1 - p_1 - p_2)$. Thus, the actual values are $v_1(1) = \frac{\tilde{v}_1(1)}{\alpha_1}$, $v_2(2) = \frac{\tilde{v}_2(2)}{\alpha_2}$; $v_1(12) = \tilde{v}_1(12)$ and $v_2(12) = \tilde{v}_2(12)$. The noise-free game preferences are unclear; it will depend on the relative values of α_1 , α_2 , and \tilde{v} . If α_1 , α_2 and \tilde{v} 's are such that $\tilde{v}_2(12) > \frac{\tilde{v}_2(2)}{\alpha_2}$ then $\pi = N$ otherwise $\pi = \{\{1\}, \{2\}\}$.
20. $\alpha(1) = \alpha_2; \alpha(2) = \alpha_1; \alpha(12) = 1$. Probability of such alpha's is $p_1 p_2 (1 - p_1 - p_2)$. Thus, the actual values are $v_1(1) = \frac{\tilde{v}_1(1)}{\alpha_2}$, $v_2(2) = \frac{\tilde{v}_2(2)}{\alpha_1}$; $v_1(12) = \tilde{v}_1(12)$ and $v_2(12) = \tilde{v}_2(12)$. The noise-free game preferences are unclear; it will depend on the relative values of α_1 , α_2 , and \tilde{v} . If α_1 , α_2 and \tilde{v} 's are such that $\tilde{v}_1(12) > \frac{\tilde{v}_1(1)}{\alpha_2}$ then $\pi = N$, otherwise $\pi = \{\{1\}, \{2\}\}$.
21. $\alpha(1) = \alpha_1; \alpha(2) = \alpha_1; \alpha(12) = \alpha_2$. Probability of such alpha's is $p_1^2 p_2$. Thus, the actual values are $v_1(1) = \frac{\tilde{v}_1(1)}{\alpha_1}$, $v_2(2) = \frac{\tilde{v}_2(2)}{\alpha_1}$; $v_1(12) = \frac{\tilde{v}_1(12)}{\alpha_2}$ and $v_2(12) = \frac{\tilde{v}_2(12)}{\alpha_2}$. Since, $\frac{\tilde{v}_1(12)}{\alpha_2} > \tilde{v}_1(12) > \tilde{v}_1(1) > \frac{\tilde{v}_1(1)}{\alpha_1}$, and $\frac{\tilde{v}_2(12)}{\alpha_2} > \tilde{v}_2(12) > \tilde{v}_2(2) > \frac{\tilde{v}_2(2)}{\alpha_1}$. The noise-free game is: $v_1(12) > v_1(1)$; $v_2(12) > v_2(2)$. So, $\pi = \tilde{\pi}$ in this case.
22. $\alpha(1) = \alpha_1; \alpha(2) = \alpha_2; \alpha(12) = \alpha_1$. Probability of such alpha's is $p_1^2 p_2$. Thus, the actual values are $v_1(1) = \frac{\tilde{v}_1(1)}{\alpha_1}$, $v_2(2) = \frac{\tilde{v}_2(2)}{\alpha_2}$; $v_1(12) = \frac{\tilde{v}_1(12)}{\alpha_1}$ and $v_2(12) = \frac{\tilde{v}_2(12)}{\alpha_1}$. The noise-free game preferences are unclear; it will depend on the relative values of α_1 , α_2 , and \tilde{v} . If α_1 , α_2 and \tilde{v} 's are such that $\frac{\tilde{v}_2(12)}{\alpha_1} > \frac{\tilde{v}_2(2)}{\alpha_2}$ then $\pi = N$ otherwise $\pi = \{\{1\}, \{2\}\}$.
23. $\alpha(1) = \alpha_1; \alpha(2) = \alpha_2; \alpha(12) = \alpha_2$. Probability of such alpha's is $p_1 p_2^2$. Thus, the actual values are $v_1(1) = \frac{\tilde{v}_1(1)}{\alpha_1}$, $v_2(2) = \frac{\tilde{v}_2(2)}{\alpha_2}$; $v_1(12) = \frac{\tilde{v}_1(12)}{\alpha_2}$ and $v_2(12) = \frac{\tilde{v}_2(12)}{\alpha_2}$. Since, $\frac{\tilde{v}_1(12)}{\alpha_2} > \tilde{v}_1(12) > \tilde{v}_1(1) > \frac{\tilde{v}_1(1)}{\alpha_1}$. The noise-free game is: $v_1(12) > v_1(1)$; $v_2(12) > v_2(2)$. So, $\pi = \tilde{\pi}$ in this case.
24. $\alpha(1) = \alpha_2; \alpha(2) = \alpha_1; \alpha(12) = \alpha_1$. Probability of such alpha's is $p_1^2 p_2$. Thus, the actual values are $v_1(1) = \frac{\tilde{v}_1(1)}{\alpha_2}$, $v_2(2) = \frac{\tilde{v}_2(2)}{\alpha_1}$; $v_1(12) = \frac{\tilde{v}_1(12)}{\alpha_1}$ and $v_2(12) = \frac{\tilde{v}_2(12)}{\alpha_1}$. Clearly, the preferences in the noise-free game are not clear; it will depend on the relative values of α_1 , α_2 and \tilde{v} . If α_1 , α_2 and \tilde{v} 's are such that $\frac{\tilde{v}_1(12)}{\alpha_1} > \frac{\tilde{v}_1(1)}{\alpha_2}$ then $\pi = N$ otherwise $\pi = \{\{1\}, \{2\}\}$.
25. $\alpha(1) = \alpha_2; \alpha(2) = \alpha_1; \alpha(12) = \alpha_2$. Probability of such alpha's is $p_1 p_2^2$. Thus, the actual values are $v_1(1) = \frac{\tilde{v}_1(1)}{\alpha_2}$, $v_2(2) = \frac{\tilde{v}_2(2)}{\alpha_1}$; $v_1(12) = \frac{\tilde{v}_1(12)}{\alpha_2}$ and $v_2(12) = \frac{\tilde{v}_2(12)}{\alpha_2}$. Since, $\frac{\tilde{v}_2(12)}{\alpha_2} > \tilde{v}_2(12) > \tilde{v}_2(2) > \frac{\tilde{v}_2(2)}{\alpha_1}$. The noise-free game is: $v_1(12) > v_1(1)$; $v_2(12) > v_2(2)$. So, $\pi = \tilde{\pi}$ in this case.
26. $\alpha(1) = \alpha_2; \alpha(2) = \alpha_2; \alpha(12) = \alpha_1$. Probability of such alpha's is $p_1 p_2^2$. Thus, the actual values are $v_1(1) = \frac{\tilde{v}_1(1)}{\alpha_2}$, $v_2(2) = \frac{\tilde{v}_2(2)}{\alpha_2}$; $v_1(12) = \frac{\tilde{v}_1(12)}{\alpha_1}$ and $v_2(12) = \frac{\tilde{v}_2(12)}{\alpha_1}$. The noise-free game preferences are unclear; it will depend on the relative values α_1 ,

α_2 and \tilde{v} . If α_1, α_2 and \tilde{v} 's are such that $\frac{\tilde{v}_1(12)}{\alpha_1} > \frac{\tilde{v}_1(1)}{\alpha_2}$ and $\frac{\tilde{v}_2(12)}{\alpha_1} > \frac{\tilde{v}_2(2)}{\alpha_2}$ then $\pi = N$ otherwise $\pi = \{\{1\}, \{2\}\}$.

27. $\alpha(1) = \alpha_2; \alpha(2) = \alpha_2; \alpha(12) = \alpha_2$. The probability of such alpha is p_2^3 . Thus, the actual values are $v_1(1) = \frac{\tilde{v}_1(1)}{\alpha_2}$, $v_2(2) = \frac{\tilde{v}_2(2)}{\alpha_2}$; $v_1(12) = \frac{\tilde{v}_1(12)}{\alpha_2}$ and $v_2(12) = \frac{\tilde{v}_2(12)}{\alpha_2}$. The noise-free game is: $v_1(12) > v_1(1)$; $v_2(12) > v_2(2)$. So, $\pi = \tilde{\pi}$ in this case.

Since $\bar{r} = \max \left\{ \frac{\tilde{v}_1(12)}{\tilde{v}_1(1)}, \frac{\tilde{v}_2(12)}{\tilde{v}_2(2)} \right\}$, and $\underline{r} = \min \left\{ \frac{\tilde{v}_1(12)}{\tilde{v}_1(1)}, \frac{\tilde{v}_2(12)}{\tilde{v}_2(2)} \right\}$. From above cases, we see that in 14 out of 27 cases (case 1,3,4,7,8,9,12,13,15,17,21,23,25,27) we have $\pi = \tilde{\pi} = N$ in noise-free game. In these cases, the relative value of $\tilde{v}_1(\cdot), \tilde{v}_2(\cdot)$ should satisfy $\alpha_1 \geq \bar{r}, \frac{1}{\alpha_2} \geq \bar{r}, \frac{\alpha_1}{\alpha_2} \geq \bar{r}$. The prediction probability in this case is given below as $g(p_1, p_2)$. Whereas if we allow for the cases, say $\alpha_1 < \underline{r}; \frac{1}{\alpha_2} < \underline{r}; \frac{\alpha_1}{\alpha_2} < \underline{r}$, then the prediction probability is 1. So, these are the two extreme cases. However, if we take any other range of α 's, the prediction probability will be more than $g(p_1, p_2)$ and less than 1. Thus,

$$\mathbb{P}[\pi = \tilde{\pi} \mid \text{game 1}] = \begin{cases} g(p_1, p_2), & \text{if } \alpha_1 \geq \bar{r}; \frac{1}{\alpha_2} \geq \bar{r}; \frac{\alpha_1}{\alpha_2} \geq \bar{r} \\ 1, & \text{if } \alpha_1 < \underline{r}; \frac{1}{\alpha_2} < \underline{r}; \frac{\alpha_1}{\alpha_2} < \underline{r}, \end{cases} \quad (20)$$

where $g(p_1, p_2) = p_1^3 + p_2^3 + 2(p_1(1 - p_1 - p_2))^2 + p_2^2(1 - p_1 - p_2) + p_1p_2(1 - p_1 - p_2) + p_1p_2^2 + p_1^2p_2 + p_1^2(1 - p_1 - p_2) + p_2(1 - p_1 - p_2)^2 + (1 - p_1 - p_2)^3$. ■

5.2. Safety value via global minima for 2 agents and 3 support noise model

Here we will show that the above prediction probability given in Equation (20) can be non-convex in p_1, p_2 . So, the global minima are difficult to hope for.

Note that $\frac{\partial g(p_1, p_2)}{\partial p_1} = 3p_1^2 - (p_2 - 1)^2$ and $\frac{\partial g(p_1, p_2)}{\partial p_2} = -2p_1(p_2 - 1) - 3p_2^2 + 6p_2 - 2$. Hence, we have $\frac{\partial^2 g(p_1, p_2)}{\partial^2 p_1} = 6p_1$, $\frac{\partial^2 g(p_1, p_2)}{\partial p_1 p_2} = \frac{\partial^2 g(p_1, p_2)}{\partial p_2 p_1} = -2(p_2 - 1)$, and $\frac{\partial^2 g(p_1, p_2)}{\partial p_2^2} = -2p_1 - 6p_2 + 6$. Thus, the Hessian of $g(p_1, p_2)$ is

$$H(g(p_1, p_2)) = \begin{bmatrix} 6p_1 & -2(p_2 - 1) \\ -2(p_2 - 1) & -2p_1 - 6p_2 + 6 \end{bmatrix}.$$

For $p_1 = 0.3$ and $p_2 = 0.5$, we have

$$H(g(p_1, p_2)) = \begin{bmatrix} 0.18 & 1 \\ 1 & 2.4 \end{bmatrix}.$$

The eigenvalues are $\lambda_1 = 2.78$, and $\lambda_2 = -0.20$. So, $g(p_1, p_2)$ is not a convex function. Therefore, finding the global minima is difficult.

Though the above prediction probability is non-convex, one can get the noise set such that the prediction probability is more than a given satisfaction ζ . Similar to the 2 support cases, where the prediction probability was a convex function, but the noise regimes were disjoint intervals, in 3 support cases also, we get disjoint sets. However, computing the exact safety value is problematic because it is the global minima of the non-convex prediction

probability function. Note that the safety value is a fundamental limit such that below a user-given satisfaction ζ , the partition is noise robust in the entire noise probability simplex.

As earlier, in the noise regimes where the prediction probability is more than ζ , a partition $\tilde{\pi}$ that is core-stable in a noisy game will remain core-stable in a noise-free game.