

Supplement to “Subjective Dynamic Information Constraints”

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April 2016

ERID Working Paper Number 215

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Supplement to

Subjective Dynamic Information Constraints

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APRIL 2016

All references to definitions and results in this Supplement refer to Dillenger, Krishna, and Sadowski (2016, henceforth DKS) unless otherwise specified. This supplement is organized as follows. Section 1 establishes the Abstract Static Representation that is the starting point for our derivations in Appendix C of DKS. Section 2 reviews relevant notions from convex analysis. Section 3 provides a preference independent notion of minimality on the space of rics, which is referred to in Section 6 of DKS. Section 4 provides a metric on the space of partitions as referred to in Appendix A.3 of DKS. Section 5 extends the existence of the RAA representation, which is established in Krishna and Sadowski (2014) for finite prize spaces, to our domain with a compact set of prizes, as discussed in Appendix A.7 of DKS. Finally, Section 6 provides a detailed proof of the partitional representation introduced in Appendix C.1 of DKS.

1. Abstract Static Representation

Let Y be a compact metric space. Then, $\Delta(Y)$ is the space of probability measures on Y . For compact metric spaces Y_1, \dots, Y_n , we will consider the product space $Z := \Delta(Y_1) \times \dots \times \Delta(Y_n)$. We are interested in the space of closed subsets of Z , $\mathcal{K}(Z)$ (endowed with the Hausdorff metric), and also in the space of closed and convex subsets $\mathcal{K}_c(Z)$. It is well known that $\mathcal{K}_c(Z)$ is a closed subset of $\mathcal{K}(Z)$.

The convex hull of a set A (in the relevant ambient vector space) is denoted by $\text{ch } A$. If the ambient vector space has a topology, then $\text{cch } A$ denotes the closed convex hull of A .

Recall that $\mathbf{C}(Y_i)$ is the space of all uniformly continuous functions on Y_i and for $\alpha_i \in \Delta(Y_i)$ and $u_i \in \mathbf{C}(Y_i)$, $u_i(\alpha_i) := \int_{Y_i} u_i(y_i) d\alpha_i(y_i) =: \langle \alpha_i, u_i \rangle$; endowed with the supremum norm, $\mathbf{C}(Y_i)$ is a Banach space. For each $s \in S$, let $L_s \subset \Delta(Y_s)$ be a closed

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subset, and define $L := \times_{s \in S} L_s$. Fix $\ell_s^\dagger \in L_s$, and define $\mathfrak{U}_{Y_s, \ell_s^\dagger} := \{u_s \in \mathbf{C}(Y_s) : u_s(\ell_s^\dagger) = 0, \|u\|_\infty = 1\}$. Finally, define $\mathfrak{U} := \{(p_1 u_1, \dots, p_n u_n) : u_s \in \mathfrak{U}_{Y_s, \ell_s^\dagger}, p_s \geq 0, \sum_s p_s = 1\}$. The space \mathfrak{U} will serve as our *subjective state space* below. It is useful to reconsider \mathfrak{U} as $\mathfrak{U} := \{(p, u) : p := (p_1, \dots, p_n) \in \Delta(S), u := (u_1, \dots, u_n) \in \times_{s \in S} \mathfrak{U}_{Y_s, \ell_s^\dagger}\}$.

Specifically, if we consider the domain X , then each $Y_s := C \times X$, which then results in a corresponding definition of \mathfrak{U} .

Theorem 1. *Let \succsim be a binary relation on X . Then, the following are equivalent:*

- (a) \succsim satisfies Basic Properties (Axiom 1) and L-Independence (Axiom 2 (a)).
- (b) There exists a metric space of continuous functions \mathfrak{U} (as defined above) and a minimal set \mathfrak{M} of finite, normal, and positive charges¹ on \mathfrak{U} that is weak* compact such that
 - [i] For all $\ell \in L$ and $s \in S$, $\int_{\mathfrak{U}} p_s u_s(\ell_s) d\mu(p, u)$ is independent of $\mu \in \mathfrak{M}$, and
 - [ii] The function $V : X \rightarrow \mathbb{R}$ given by

$$\spadesuit \quad V(x) := \max_{\mu \in \mathfrak{M}} \left[\int_{\mathfrak{U}} \max_{\alpha \in x} \sum_s p_s u_s(\alpha_s) d\mu(p, u) \right]$$

represents \succsim .

The proof of Theorem 1 follows immediately from Propositions 1.10, 1.11, and 1.12 below.

1.1. Algebraic Representation

Recall that our domain is $X \simeq \mathcal{K}(\mathcal{F}(\Delta(C \times X)))$. We shall first show that under our assumptions, every closed subset is indifferent to its closed convex hull.

Lemma 1.1. *If \succsim satisfies Axiom 1, then for each $x \in \mathcal{K}(Z)$, $x \sim \text{cch}(x)$.*

Proof. First consider $x \in X$ that is finite and follow Ergin and Sarver (2010a, Lemma 2). Notice that $\text{cch}(x) \succsim x$ by Monotonicity (Axiom 1(d)). Let $x^0 := x$, and for each $k \geq 1$, define $x^k := \frac{1}{2}x^{k-1} + \frac{1}{2}x^{k-1}$. Then, by Aversion to Randomization (Axiom 1(e)), $x^{k-1} \succsim x^k$. In other words, by Order (Axiom 1(a)), $x \succsim x^k$ for all $k \geq 1$. But notice that $d(x^k, \text{cch}(x)) \rightarrow 0$ as $k \rightarrow \infty$. Therefore, by Continuity (Axiom 1(b)), it follows that $x \succsim \text{cch}(x)$, which proves that $x \sim \text{cch}(x)$ for all finite subsets of X .

Now consider the general case, where $x \in X$ is arbitrary. Then, there exists a sequence of finite sets (x_m) such that (i) $x_m \subset x$ for all m , and (ii) $d(x_m, x) \rightarrow 0$ (in the Hausdorff metric). But each $x_m \sim \text{cch}(x_m)$. It is also easy to see that $d(\text{cch}(x), \text{cch}(x_m)) \rightarrow 0$ as $m \rightarrow \infty$. Continuity (Axiom 1(b)) now implies that $x \sim \text{cch}(x)$, which proves the claim. \square

In light of lemma 1.1, in what follows, we may restrict attention to the space $\mathcal{H}_c(X)$.

(1) A charge is a finitely additive measure.

Lemma 1.2. If \succsim satisfies Continuity (Axiom 1(b)) and L-Independence (Axiom 2(a)), then there exists a continuous and affine function $\zeta : L \rightarrow \mathbb{R}$ such that ζ represents $\succsim|_L$, ie, for all $\ell, \ell' \in L$, $\ell \succsim \ell'$ if, and only if, $\zeta(\ell) \geq \zeta(\ell')$.

Proof. Independence and Continuity hold on L , so by the Expected Utility Theorem, the claim follows. \square

Corollary 1.3. If \succsim satisfies Axiom 1, there exist $\ell^\#, \ell_\# \in L$ such that $\ell^\# \succ \ell_\#$.

Proof. Consider $\ell^\#, \ell_\# \in L$ that exist by Lipschitz continuity (Axiom 1(c)). Set $x = y = \{\ell^\#\}$ and $\alpha = \frac{1}{2}$. Lipschitz continuity then implies $\ell^\# \succ \frac{1}{2}\ell^\# + \frac{1}{2}\ell_\#$. Similarly, let $x = y = \{\ell_\#\}$ and $\alpha = \frac{1}{2}$, so Lipschitz continuity implies $\frac{1}{2}\ell^\# + \frac{1}{2}\ell_\# \succ \ell_\#$. It follows immediately that $\ell^\# \succ \ell_\#$. \square

Lemma 1.4. Given the function $\zeta : L \rightarrow \mathbb{R}$ from lemma 1.2 above, there exists $V : X \rightarrow \mathbb{R}$ such that

- (a) $x \succsim y$ if, and only if, $V(x) \geq V(y)$ for all $x, y \in X$,
- (b) for all $\ell \in L$, $V(\ell) = \zeta(\ell)$, and
- (c) V is continuous.

Proof. By Corollary 1.3, $\ell^* \succ \ell_*$. First, consider the case where $x \in X$ is such that $\ell^* \succsim x \succsim \ell_*$. By Continuity (Axiom 1(b)), there exists $a \in [0, 1]$ such that $x \sim a\ell^* + (1-a)\ell_*$. Define $V(x) := \zeta(a\ell^* + (1-a)\ell_*) = a\zeta(\ell^*) + (1-a)\zeta(\ell_*)$. It is easy to see that for all $\ell \in L$, $V(\ell) = \zeta(\ell)$.

Next, consider the case where $x \succ \ell^*$. By Continuity, for any $\ell \in L$, there exists $a \in [0, 1]$ such that $ax + (1-a)\ell_* \sim \ell$. Now, set $V(x) = [V(\ell) - (1-a)V(\ell_*)]/a$.

To see that $V(x)$ is independent of the choice of ℓ , suppose $\ell' \in L$ and $a' \in [0, 1]$ are such that $\ell \succsim \ell'$ and $a'x + (1-a')\ell_* \sim \ell'$, so that $V(x) = [V(\ell') - (1-a')V(\ell_*)]/a'$. Because $ax + (1-a)\ell_* \sim \ell$, for all $b \in [0, 1]$, $b(ax + (1-a)\ell_*) + (1-b)\ell_* \sim b\ell + (1-b)\ell_*$. Now, choose b such that $b\ell + (1-b)\ell_* \sim \ell'$. Then, $b(ax + (1-a)\ell_*) + (1-b)\ell_* \sim \ell'$, which implies $ba = a'$. Using the fact that $V(\ell') = bV(\ell) + (1-b)V(\ell_*)$, we see that

$$\begin{aligned} V(x) &= \frac{V(\ell') - (1-a')V(\ell_*)}{a'} \\ &= \frac{[bV(\ell) + (1-b)V(\ell_*)] - (1-ba)V(\ell_*)}{ba} \\ &= \frac{V(\ell) - (1-a)V(\ell_*)}{a} \end{aligned}$$

which is independent of the choice of b , or equivalently, the choice of ℓ' .

We can deal with case where $\ell_* \succ x$ in a similar fashion. The continuity of V follows immediately from the continuity of \succsim and from the continuity of ζ , which completes the proof. \square

Lemma 1.5. If $tx + (1-t)\ell \succ ty + (1-t)\ell$ then $x \succ y$.

Proof. Suppose not. Then, by L-Independence, there are x, y, ℓ , and t such that $x \sim y$ and $tx + (1-t)\ell \succ ty + (1-t)\ell$. By Lipschitz Continuity (Axiom 1(c)), and because $d(x, x) = 0$, we have $t'x + (1-t')\ell^\# \succ t'x + (1-t')\ell_\#$ for all $t' > 0$. Observe that by Negative Transitivity of the strict relation \succ , it must be that for all t' , either $t'x + (1-t')\ell^\# \succ x$ or $x \succ t'x + (1-t')\ell_\#$ holds, and the same for y . There are three cases to consider.

Case 1: For all $\varepsilon > 0$ there is $(1-t') < \varepsilon$ with $x \succ t'x + (1-t')\ell_\#$. Then, since $x \sim y$, L-Independence implies that $ty + (1-t)\ell \succ t(t'x + (1-t')\ell_\#) + (1-t)\ell$ for all such $(1-t') > 0$. At the same time, by continuity, we can pick $(1-\bar{t}) > 0$ small enough, such that by replacing x with $\bar{t}x + (1-\bar{t})\ell_\#$, $t(\bar{t}x + (1-\bar{t})\ell_\#) + (1-t)\ell \succ ty + (1-t)\ell$ still holds. Taking $\varepsilon \leq (1-\bar{t})$ establishes a contradiction.

Case 2: For all $\varepsilon > 0$ there is $(1-t') < \varepsilon$ with $t'y + (1-t')\ell^\# \succ y$. This case is analogous to case 1.

Case 3: There is $\varepsilon > 0$ such that for all $(1-t') < \varepsilon$, both $t'x + (1-t')\ell_\# \succsim x$ and $y \succsim t'y + (1-t')\ell^\#$. We claim that this case can never occur. To see this, first observe that by continuity, if $t'x + (1-t')\ell_\# \succsim x$ for all $(1-t') < \varepsilon$ then $\ell_\# \succsim x$; and if $y \succsim t'y + (1-t')\ell^\# \succsim x$ for all $(1-t') < \varepsilon$ then $y \succsim \ell^\#$. But then we have $y \succsim \ell^\# \succ \ell_\# \succsim x$, which contradicts the premise that $x \sim y$. \square

Corollary 1.6. It follows immediately from L-Independence and Lemma 1.5 that $tx + (1-t)\ell \succ ty + (1-t)\ell$ if, and only if, $x \succ y$.

Lemma 1.7. $\ell \succ \ell'$ if, and only if, $tx + (1-t)\ell \succ tx + (1-t)\ell'$.

Proof. If $x \succ \ell_*$, by continuity there are $\alpha \in (0, 1)$ and $\bar{\ell} \in L$ with $\alpha x + (1-\alpha)\ell_* \sim \bar{\ell}$. Applying Corollary 1.6 repeatedly yields that $\ell \succ \ell'$ if, and only if, $t'[\alpha x + (1-\alpha)\ell_*] + (1-t')\ell \sim t'\bar{\ell} + (1-t')\ell \succ t'\bar{\ell} + (1-t')\ell' \sim t'[\alpha x + (1-\alpha)\ell_*] + (1-t')\ell'$ for all $t' \in (0, 1)$. Again by Corollary 1.6, and for $t' = \frac{t}{\alpha + t(1-\alpha)}$, this is equivalent to $tx + (1-t)\ell \succ tx + (1-t)\ell'$. The case where $\ell^* \succ x$ is similar and hence omitted. \square

Lemma 1.8. The function V defined in the proof of Lemma 1.4 has the following properties:

- (a) V is monotone, ie, $V(x \cup y) \geq V(x)$ for all $x, y \in X$;
- (b) V is L -affine, ie, for all $x \in X, \ell \in L$ and $a \in [0, 1]$, $V(ax + (1-a)\ell) = aV(x) + (1-a)V(\ell)$;
- (c) V is midpoint convex, ie, $V(\frac{1}{2}x_1 + \frac{1}{2}x_2) \leq \frac{1}{2}V(x_1) + \frac{1}{2}V(x_2)$;
- (d) V is convex.

Proof. To ease notational burden, we shall assume only in this part of the proof, and without loss of generality, that $V(\ell^*) = 1$ while $V(\ell_*) = 0$. We prove the claims in turn.

- (a) V represents \succsim , so it is clear that it is monotone.

(b) Let $x \in X$ and $\ell \in L$. Consider first the case where $\ell^* \succsim x \succsim \ell_*$. Then, there exists $\ell_x \in L$ such that $x \sim \ell_x$. Then, by L -Independence, for all $a \in (0, 1]$, $ax + (1-a)\ell \sim a\ell_x + (1-a)\ell$. Therefore, $V(ax + (1-a)\ell) = V(a\ell_x + (1-a)\ell) = aV(\ell_x) + (1-a)V(\ell) = aV(x) + (1-a)V(\ell)$, as required.

Now consider the case where $x \succ \ell^*$, the case where $\ell_* \succ x$ being analogous. Because $\ell \succsim \ell_*$, Lemma 1.7 yields $t\ell_* + (1-t)\ell \succsim \ell_*$, and then, by Corollary 1.6, $tx + (1-t)\ell \succ \ell_*$. By continuity, there are $\alpha \in (0, 1)$ and $\bar{\ell}$, such that $\ell^* \succ \alpha(tx + (1-t)\ell) + (1-\alpha)\ell_* \sim \bar{\ell} \succ \ell_*$. Further, let $\beta \in [0, 1]$ be such that $\ell \sim \beta\ell^* + (1-\beta)\ell_*$ (so that $V(\ell) = \beta$), and let $\gamma \in (0, 1)$ be such that $\bar{\ell} \sim \gamma\ell^* + (1-\gamma)\ell_*$. First, from Corollary 1.6 and the definition of V it is easy to verify that $V(tx + (1-t)\ell) = \frac{\gamma}{\alpha}$ (independent of whether $tx + (1-t)\ell \succsim \ell^*$ or not). Next, by Lemma 1.7, $tx + (1-t)\ell \sim tx + (1-t)(\beta\ell^* + (1-\beta)\ell_*)$. Then, by Corollary 1.6,

$$\alpha(tx + (1-t)(\beta\ell^* + (1-\beta)\ell_*)) + (1-\alpha)\ell_* \sim \gamma\ell^* + (1-\gamma)\ell_*$$

or

$$\alpha tx + \alpha(1-t)\beta\ell^* + [1 - \alpha t - \alpha(1-t)\beta]\ell_* \sim \gamma\ell^* + (1-\gamma)\ell_*$$

Because $x \succ \ell^*$, Corollary 1.6 and Lemma 1.7 further imply that $\alpha(1-t)(1-\beta) + (1-\alpha) > (1-\gamma)$ or $\gamma - \alpha(1-t)\beta > \alpha t > 0$. This implies that $\gamma > \alpha(1-t)\beta$. Corollary 1.6 then yields that

$$\frac{\alpha t}{D_1}x + \frac{1 - \alpha t - \alpha(1-t)\beta}{D_1}\ell_* \sim \frac{\gamma - \alpha(1-t)\beta}{D_1}\ell^* + \frac{1 - \gamma}{D_1}\ell_*$$

where $D_1 = \gamma - \alpha(1-t)\beta + (1-\gamma) = 1 - \alpha(1-t)\beta$.

It follows that $1 - \gamma < 1 - \alpha t - \alpha(1-t)\beta$, and hence, again by Corollary 1.6,

$$\frac{\alpha t}{D_2}x + \frac{1 - \alpha t - \alpha(1-t)\beta - (1-\gamma)}{D_2}\ell_* \sim \ell^*$$

where $D_2 = \alpha t + 1 - \alpha t - \alpha(1-t)\beta - (1-\gamma) = \gamma - \alpha(1-t)\beta$.

Hence, $\frac{\alpha t}{\gamma - \alpha(1-t)\beta}x + \left[1 - \frac{\alpha t}{\gamma - \alpha(1-t)\beta}\right]\ell_* \sim \ell^*$, so that $V(x) = \frac{\gamma - \alpha(1-t)\beta}{\alpha t}$. Putting everything together establishes the lemma, ie,

$$tV(x) + (1-t)V(\ell) = \frac{\gamma}{\alpha} = V(tx + (1-t)\ell)$$

(c) Suppose first that $x_1 \sim x_2$. Then, by Aversion to Randomization (Axiom 1 (e)), $x_1 \succsim \frac{1}{2}x_1 + \frac{1}{2}x_2$, from which it follows immediately that $V(\frac{1}{2}x_1 + \frac{1}{2}x_2) \leq \frac{1}{2}V(x_1) + \frac{1}{2}V(x_2)$. Let us now suppose that $x_1 \succ x_2$ and consider the case where $\ell^* \succ x_1$. By continuity, there exists $\lambda \in (0, 1)$ such that $y := \lambda x_2 + (1-\lambda)\ell^* \sim x_1$. Notice that because V is L -affine, $V(y) = \lambda V(x_2) + (1-\lambda)V(\ell^*) = V(x_1)$. Let $\bar{x} := \frac{\lambda}{1+\lambda}x_1 + \frac{1}{1+\lambda}y = \frac{2\lambda}{1+\lambda}(\frac{1}{2}x_1 + \frac{1}{2}x_2) + \frac{1-\lambda}{1+\lambda}\ell^*$, so that $V(\bar{x}) = \frac{2\lambda}{1+\lambda}V(\frac{1}{2}x_1 + \frac{1}{2}x_2) + \frac{1-\lambda}{1+\lambda}V(\ell^*)$, where we have

used the L -affinity of V . But notice also that $V(\bar{x}) \leq \frac{\lambda}{1+\lambda}V(x_1) + \frac{1}{1+\lambda}V(y)$ by Aversion to Randomization (Axiom 1 (e)) because $x_1 \sim y$. We also have $\frac{\lambda}{1+\lambda}V(x_1) + \frac{1}{1+\lambda}V(y) = \frac{\lambda}{1+\lambda}(V(x_1) + V(x_2)) + \frac{1-\lambda}{1+\lambda}V(\ell^*)$. Substituting in the value of $V(\bar{x})$ obtained above, we see that $V(\frac{1}{2}x_1 + \frac{1}{2}x_2) \leq \frac{1}{2}V(x_1) + \frac{1}{2}V(x_2)$, as claimed.

Now consider the case where $x_1 \succ x_2$ but $x_1 \succ \ell^*$. Then, by continuity, there exists $a \in [0, 1]$ such that $y = ax_1 + (1-a)\ell^* \sim x_2$. Therefore, $V(y) = aV(x_1) + (1-a)V(\ell^*) = V(x_1)$. Set $\bar{x} = \frac{a}{1+a}x_2 + \frac{1}{1+a}y = \frac{2a}{1+a}(\frac{1}{2}x_1 + \frac{1}{2}x_2) + \frac{1-a}{1+a}\ell^*$. Then, using the L -affinity of V , we obtain $V(\bar{x}) = \frac{2a}{1+a}V(\frac{1}{2}x_1 + \frac{1}{2}x_2) + \frac{1-a}{1+a}V(\ell^*)$.

But notice that $x_2 \sim y$, so that by Aversion to Randomization (Axiom 1 (e)), $V(\bar{x}) \leq \frac{a}{1+a}V(x_2) + \frac{1}{1+a}V(y)$. We also have $\frac{a}{1+a}V(x_1) + \frac{1}{1+a}V(y) = \frac{a}{1+a}(V(x_1) + V(x_2)) + \frac{1-a}{1+a}V(\ell^*)$. Substituting in the value of $V(\bar{x})$ obtained above, we see that $V(\frac{1}{2}x_1 + \frac{1}{2}x_2) \leq \frac{1}{2}V(x_1) + \frac{1}{2}V(x_2)$, as claimed.

(d) As noted above, V is continuous, and because it is midpoint convex, it is convex. \square

Recall that V is Lipschitz if there exists a constant $K > 0$ such that for all $x, y \in X$, $|V(x) - V(y)| \leq Kd(x, y)$, where $d(\cdot, \cdot)$ is the metric on X .

Lemma 1.9. If \succsim satisfies Lipschitz continuity (Axiom 1(c)) and is represented by a continuous and L -affine V , then V is Lipschitz. Conversely, if V is Lipschitz, non-trivial, L -affine, and represents \succsim , then it satisfies Lipschitz continuity.

Proof. Let $N > 0$ be as given in Lipschitz continuity. Fix $\beta \in (0, 1)$ such that $N\beta < 1$. First consider the case where $x, y \in X$ are such that $0 < d(x, y) \leq \beta$ and let $\alpha = Nd(x, y)$. Then, by Lipschitz Continuity, $(1 - \alpha)x + \alpha\ell^\# \succ (1 - \alpha)y + \alpha\ell_\#$. By the L -affinity of V , it follows that $V(y) - V(x) < \frac{\alpha}{1 - \alpha}[V(\ell^\#) - V(\ell_\#)]$. But notice that $\alpha/N \leq \beta$, so setting $K = N/(1 - N\beta)[V(\ell^\#) - V(\ell_\#)]$, we find that

$$\begin{aligned} V(y) - V(x) &< \frac{\alpha}{1 - \alpha}[V(\ell^\#) - V(\ell_\#)] \\ &< \frac{N}{1 - \alpha}[V(\ell^\#) - V(\ell_\#)]d(x, y) \\ &< Kd(x, y) \end{aligned}$$

We now follow Dekel et al. (2007) and remove the restriction on the x and y . For arbitrary $x, y \in X$, let $0 =: \lambda_0 < \lambda_1 < \dots < \lambda_{J+1} = 1$ such that $(\lambda_{j+1} - \lambda_j)d(x, y) \leq \beta$ for all $j = 0, \dots, J+1$. Define $x_j := \lambda_j x + (1 - \lambda_j)y$, so $d(x_{j+1}, x_j) = (\lambda_{j+1} - \lambda_j)d(x, y) < \beta$. From the result established above, we see that $V(x_{j+1}) - V(x_j) \leq Kd(x_{j+1}, x_j) = K(\lambda_{j+1} - \lambda_j)d(x, y)$. Summing over j , we find $V(y) - V(x) \leq Kd(x, y)$. Interchanging the roles of x and y , it follows that $|V(x) - V(y)| \leq Kd(x, y)$, as claimed. The converse is as in Dekel et al. (2007) and is omitted. \square

In sum, we have proven that (a) implies (b) in the following representation result.

Proposition 1.10. Let \succsim be a binary relation. Then, the following are equivalent.

- (a) \succsim satisfies Basic Properties (Axiom 1) and L -Independence (Axiom 2(a)).
- (b) There exists a function $V : X \rightarrow \mathbb{R}$ that represents \succsim and is L -affine, Lipschitz Continuous, and convex. Moreover, any such representation of \succsim is unique up to a positive affine transformation.

The proof that (b) implies (a) is standard and is omitted.

1.2. Abstract Convex and Monotone Representation

Every $\alpha \in \mathcal{F}(\Delta(C \times X))$ is a product lottery of the form $\alpha_1 \times \cdots \times \alpha_n$. A function $u \in \mathcal{U}$ acts on $\mathcal{F}(\Delta(C \times X))$ as follows: $u(\alpha) := \sum_i p_i u_i(\alpha_i)$. For any $x \in \mathcal{K}_c(\mathcal{F}(\Delta(C \times X)))$, define its *support function* $H_x : \mathcal{U} \rightarrow \mathbb{R}$ as $H_x(u) := \max_{\alpha \in x} u(\alpha)$. The *extended* support function of $x \in \mathcal{K}_c(\mathcal{F}(\Delta(C \times X)))$ is the unique extension of the support function H_x to $\text{span}(\mathcal{U})$ by positive homogeneity. Theorem 5.102 and Corollary 6.27 of Aliprantis and Border (1999) imply that a function defined on $\text{span}(\mathcal{U})$ is sublinear, norm continuous, and positively homogeneous if, and only if, it is the extended support function of some weak* closed, convex subset of $\mathcal{F}(\Delta(C \times X))$. Therefore, a function $H : \mathcal{U} \rightarrow \mathbb{R}$ is a support function if its unique extension to $\text{span}(\mathcal{U})$ by positive homogeneity is sublinear and norm continuous.

Given a function $H : \mathcal{U} \rightarrow \mathbb{R}$ whose extension to $\text{span}(\mathcal{U})$ by positive homogeneity is sublinear and norm continuous, we may define $x_H := \{\alpha \in \text{aff}(Z) : u(\alpha) \leq H(u) \text{ for all } u \in \mathcal{U}\}$. Support functions enjoy the following duality: For any weak* compact, convex subset x of $\text{aff}(Z)$, $x_{H_x} = x$, and for any function H as defined above, $H_{x_H} = H$.

For weak* compact, convex subsets x and x' of X , support functions exhibit the following properties: (i) $x \subset x'$ if, and only if, $H_x \leq H_{x'}$, (ii) $H_{tx+(1-t)x'} = tH_x + (1-t)H_{x'}$ for all $t \in (0, 1)$, (iii) $H_{x \cap x'} = H_x \wedge H_{x'}$, and (iv) $H_{\text{ch}(x \cup x')} = H_x \vee H_{x'}$. (By Lemma 5.14 of Aliprantis and Border (1999), $\text{ch}(x \cup x')$ is compact because x and x' are compact, which ensures that $H_{\text{ch}(x \cup x')}$ is well defined.) Finally, observe that for $\ell^\dagger := \ell_1^\dagger \times \cdots \times \ell_n^\dagger$, $H_{\ell^\dagger} = \mathbf{0}$.

Proposition 1.11. Let $V : X \rightarrow \mathbb{R}$ be Lipschitz, convex, and L -affine. Then, there exists a minimal set \mathfrak{M} of finite normal charges on \mathcal{U} so that V can be written as

$$[\bullet] \quad V(x) = \max_{\mu \in \mathfrak{M}} \left[\int_{\mathcal{U}} \max_{\alpha \in x} \sum_i p_i u_i(\alpha_i) d\mu(p, u) \right]$$

where the set $\mathfrak{M} \subset ba_n(\mathcal{U})$ is weak* compact and $\int_{\mathcal{U}} \max_{\alpha \in x} \sum_i p_i u_i(\alpha_i) d\mu(p, u)$ is independent of μ for all $x \in L$.² Moreover, for a dense set of points in X , there is a unique $\mu \in \mathfrak{M}$ that achieves the maximum in $[\bullet]$.

(2) Recall that $ba_n(\mathcal{U})$ is the space of finite normal charges on \mathcal{U} .

In Proposition 1.11 above, $ba_n(\mathfrak{U})$ is the space of bounded additive (or finitely additive) measures (ie, charges) on \mathfrak{U} that are also normal (ie, inner and outer regular). The last part of the proposition reflects the fact that V is linear on L . The set \mathfrak{M} is minimal in the sense that if $\mathcal{N} \subset \mathfrak{M}$ is compact, then there exists $x \in X$ such that $V(x) > \max_{\mu \in \mathcal{N}} \left[\int_{\mathfrak{U}} \max_{\alpha \in x} \sum_i p_i u_i(\alpha_i) d\mu(p, u) \right]$.

Proof. By Lemma 1.1, for every $x \in \mathcal{K}(\mathcal{F}(\Delta(C \times X)))$, $V(x) = V(\text{cch}(x))$. Therefore, we may restrict attention to convex menus.

Let $\Psi : \mathcal{K}_c(\mathcal{F}(\Delta(C \times X))) \rightarrow \mathbf{C}_b(\mathfrak{U})$ be the map that associates each compact, convex subset x of $\mathcal{F}(\Delta(C \times X))$ with its support function, $\Psi : x \mapsto H_x$. Note that Ψ is invertible. Moreover, Ψ is an isometry because $d(x, x') = \|H_x - H_{x'}\|_\infty$ for all $x, x' \in \mathcal{K}_c(\mathcal{F}(\Delta(C \times X)))$. Thus Ψ is an affine isometric embedding of $\mathcal{K}_c(\mathcal{F}(\Delta(C \times X)))$ in $\mathbf{C}_b(\mathfrak{U})$. Moreover, $\Psi(\{\ell^*\}) = \mathbf{0}$. In sum, $\Psi(\mathcal{K}_c(\mathcal{F}(\Delta(C \times X))))$ is a compact and convex subset of $\mathbf{C}_b(\mathfrak{U})$ that contains the origin.

Let $\bar{V} : \Psi(\mathcal{K}_c(\mathcal{F}(\Delta(C \times X)))) \rightarrow \mathbb{R}$ be defined as follows: $\bar{V}(H) := V(x)$ where $H = H_x$ for some x . Because Ψ is injective, it follows that \bar{V} is well defined. Thus, \bar{V} is Lipschitz, convex, and $\Psi(L)$ -affine. Recall that by definition, $V(\{\ell^*\}) = 0 = \bar{V}(H_{\{\ell^*\}})$, and $\Psi(\{\ell^*\}) = \mathbf{0}$. Therefore, \bar{V} is positively homogeneous. Extending \bar{V} to $\text{cone}(\Psi(\mathcal{K}_c(\mathcal{F}(\Delta(C \times X))))$ by positive homogeneity, it follows by Proposition 2.4 below that \bar{V} (and hence V) has the desired representation. \square

Proposition 1.12. Let $V : \mathcal{K}(\mathcal{F}(\Delta(C \times X))) \rightarrow \mathbb{R}$ be as in [•]. Then, the following are equivalent.

- (a) V is monotone, in the sense that $x \subset x'$ implies $V(x) \leq V(x')$.
- (b) Every charge $\mu \in \mathfrak{M}$ is *positive*, ie, $\mu(E) \geq 0$ for all (Borel) measurable $E \subset \mathfrak{U}$.

Proof. That (b) implies (a) is easy to see. That (a) implies (b) follows from Theorem S.2 of Ergin and Sarver (2010b) after observing that \bar{V} (defined in the proof of 1.12) is monotone. We note that a similar statement is contained in the proof of Lemma 3.5 of Gilboa and Schmeidler (1989). \square

The following corollary follows immediately from Lemma 2.5.

Corollary 1.13. Let $V : \mathcal{K}(\mathcal{F}(\Delta(C \times X))) \rightarrow \mathbb{R}$ have a representation as in [•]. Suppose $E \subset \mathcal{K}(\mathcal{F}(\Delta(C \times X)))$ is convex and $V|_E$ is linear. Then, there exists $\mu \in \mathfrak{M}$ such that $V(x) = \int_{\mathfrak{U}} \max_{\alpha \in x} \sum_i p_i u_i(\alpha_i) d\mu(p, u)$ for all $x \in E$.

2. Convex Duality

We review some notions from convex analysis. Our review follows Ekeland and Turnbull (1983).

Let X be a Banach space, X^* its norm dual, $C \subset X$, and $f : C \rightarrow \mathbb{R}$ a convex and Lipschitz function. The *subdifferential* of f at $x \in C$ is $\partial f(x) := \{x^* \in X^* : \langle y - x, x^* \rangle \leq f(y) - f(x) \text{ for all } y \in C\}$. A necessary and sufficient condition for the existence of a subdifferential at $x \in C$ is that there exists $K \geq 0$ such that for all $y \in X$, $f(x) - f(y) \leq K \|y - x\|$. To see this, recall that the set $\text{epi}(f) := \{(x, t) \in X \times \mathbb{R} : t \geq f(x)\}$, the *epigraph* of the function f , is a convex set (if, and only if, f is a convex function). For each $x \in C$, we define $A(x) := \{(y, t) \in X \times \mathbb{R} : f(x) - t > K \|y - x\|\}$. It is easy to see that the set $A(x)$ is (i) nonempty, (ii) convex, and (iii) open. It is also easy to show that $\text{epi}(f) \cap A(x) = \emptyset$, so there exists a non-vertical hyperplane that separates the two sets. Following the arguments in Gale (1967), we can conclude that $\partial f(x) \neq \emptyset$, and moreover, there exists $x^* \in \partial f(x)$ such that $\|x^*\| \leq K$. This is the content of the Duality Theorem of Gale (1967). (Indeed, Gale (1967) also shows that local Lipschitzness is a necessary condition for $\partial f(x)$ to be nonempty.) We will rely on the following result in the sequel.

Proposition 2.1 (Duality Theorem in Gale (1967)). Let $C \subset X$ be convex and suppose $f : C \rightarrow \mathbb{R}$ is convex and Lipschitz of rank K . Then, there exists $x^* \in \partial f(x)$ such that $\|x^*\| \leq K$.

In what follows, we will denote by $\partial_K f(x) := \{x^* \in \partial f(x) : \|x^*\| \leq K\}$. For each $x^* \in X^*$ and $a \in \mathbb{R}$, we can define the continuous affine functional $\varphi(\cdot, x^*) : X \rightarrow \mathbb{R}$ as $\varphi(y; x^*) := \langle y, x^* \rangle - a$. The function $\varphi \leq f$ for all $y \in C$ if, and only if, $\langle y, x^* \rangle - a \leq f(y)$, and is *exact* at $x \in C$ if $\varphi(x; x^*) = f(x)$. If φ is exact, the value of a which makes it so is given by $-a(x^*) := f(x) - \langle x, x^* \rangle$. Therefore, $x^* \in \partial f(x)$ if, and only if, the continuous affine functional $\varphi(y; x^*) = f(x) + \langle y - x, x^* \rangle \leq f(y)$ for all $y \in C$ with $\varphi(x; x^*) = f(x)$. In other words, $x^* \in \partial f(x)$ if, and only if, $\varphi(y; x^*) = f(x) + \langle y - x, x^* \rangle$ is a supporting hyperplane for the graph of f at x .

Notice that for any intercept $a \geq a(x^*)$, $\langle x, x^* \rangle - a < \langle x, x^* \rangle - a(x^*)$, so $a(x^*) = \inf\{a \in \mathbb{R} : f(x) \geq \langle x, x^* \rangle - a\} = \sup\{x \in C : \langle x, x^* \rangle - f(x)\}$. This smallest intercept is the *Fenchel conjugate* of f , and is denoted by $f^* : X^* \rightarrow \mathbb{R} \cup \{-\infty, +\infty\}$, and is given by

$$f^*(x^*) := \sup_{x \in C} [\langle x, x^* \rangle - f(x)]$$

Proposition 2 of Ekeland and Turnbull (1983) shows that $x^* \in \partial f(x)$ if, and only if, $f(x) + f^*(x^*) = \langle x, x^* \rangle$.

By Proposition 2.1, it follows that for Lipschitz f , the conjugate function is given by $f^*(x^*) := \max_{x \in C} [\langle x, x^* \rangle - f(x)]$. We now show that for positively homogeneous functions, the conjugate function f^* is identically 0.

Proposition 2.2. Let $C \subset X$ be a convex cone, and let $f : C \rightarrow \mathbb{R}$ be convex and Lipschitz. Then, the following are equivalent:

- (a) f is positively homogeneous, ie, $f(\lambda x) = \lambda f(x)$ for all $\lambda > 0$;
- (b) $f^*(x^*) \in \mathbb{R}$ implies $f^*(x^*) = 0$.

Proof. Suppose $f^* = 0$. Fix $x \in C$, and recall that because f is convex and Lipschitz, there exists $x^* \in \partial f(x)$. This implies $f(x) = \langle x, x^* \rangle$. It is easy to see that $x^* \in \partial f(\lambda x)$ for all $\lambda > 0$, so that $f(\lambda x) = \lambda f(x)$. That is, f is positively homogeneous.

Now suppose f is positively homogeneous. Fix $x \in C$ and suppose $x^* \in \partial f(x)$. We will first show that for any $\lambda > 0$, $x^* \in \partial f(\lambda x)$. Then, by the definition of ∂f , for any $y \in C$, $\langle y - x, x^* \rangle \leq f(y) - f(x)$. Now let $\lambda > 0$ and let $y \in C$ be arbitrary. Because C is a cone, there exists $z \in C$ such that $\lambda z = y$. This implies $\langle y - \lambda x \rangle = \lambda \langle z - x^* \rangle \leq \lambda [f(z) - f(x)] = f(y) - f(\lambda x)$, which proves that $x^* \in \partial f(x)$ implies $x^* \in \partial f(\lambda x)$ for all $\lambda > 0$.

Now suppose x^* is such that $f^*(x^*) \in \mathbb{R}$. Because f is positively homogeneous, we have $f(0) = 0$. (To see this, note that $f(0) = f(2 \times 0) = 2f(0)$ which implies $f(0) = 0$.) Therefore, $f^*(x^*) \geq \langle 0, x^* \rangle - f(0) = 0$. Now suppose $f^*(x^*) > 0$. Then, for any $\varepsilon \in (0, f^*(x^*))$, there exists $x \in C$ such that $f^*(x^*) - \varepsilon = \langle x, x^* \rangle - f(x) > 0$. But then we can choose $\lambda > 0$ such that $\langle \lambda x, x^* \rangle - f(\lambda x) > f^*(x^*)$, which is a contradiction. Therefore, it must be that $f^*(x^*) = 0$. \square

This allows us to establish the following corollary.

Corollary 2.3. Let $C \subset X$ be a convex cone, and $f \in \mathbb{R}^C$ be convex, Lipschitz, and positively homogeneous. Then, there exists a weak* compact set $\mathfrak{M} \subset X^*$ such that $f(x) = \max[\langle x, x^* \rangle : x^* \in \mathfrak{M}]$.

Proof. We have already established that for each $x \in C$, there exists $x^* \in \partial f(x)$ such that $\|x^*\| \leq K$, where K is the Lipschitz constant of f . We have also established that $x^* \in \partial f(\lambda x)$ for all $\lambda \geq 0$. Therefore, $f(y) \geq \langle y, x^* \rangle$ for all $y \in C$. Letting $\mathfrak{M} = \text{cl}(\{x^* \in \partial f(x) : x \in C, \|x^*\| \leq K\})$ (in the weak* topology) establishes the claim. \square

If C is convex and $A \subset C$ is also convex, then $f : C \rightarrow \mathbb{R}$ is A -affine if for all $x \in C$, $a \in A$, and $t \in (0, 1)$, we have $f(tx + (1-t)a) = tf(x) + (1-t)f(a)$.

For a fixed $x \in C$, notice that f is affine on the set $\text{ch}(\{x\} \cup A)$. Let \mathcal{E}_x be the collection of all (convex) subsets of C such that if $E \in \mathcal{E}_x$ then (i) $x \in E$ and (ii) $f|_E$ is affine. A simple application of Zorn's lemma shows that for each $x \in C$, there is a largest set E_x that contains x and where $f|_{E_x}$ is affine.

Notice that there exist $x \in X$ such that this maximal set E_x is not unique. Indeed, for any $a \in A$, and $x, y \in C$ such that f is not affine on $[x, y]$ (the closed line segment joining x and y), then $a \in E_x \cap E_y$, but $E_x \cup E_y$ (or its convex hull) is not a member of \mathcal{E}_a .

If f is Lipschitz continuous (as we shall assume below), then it is easy to see that the set E_x must be closed as well.

Proposition 2.4. Let $C \subset X$ be a convex set, and $f \in \mathbb{R}^C$ be convex and Lipschitz of rank K . Let $A \subset C$ be convex and suppose that $\mathbf{0} \in A$, $f(\mathbf{0}) = 0$, and that f is A -affine. Then, for each x , there exists $x^* \in X^*$ such that $x^* \in \partial f_K(y)$ for all $y \in E_x$ where

E_x is defined above. Moreover, there exists a weak* compact set $\mathfrak{M}_f \subset X^*$ such that $f(x) = \max[\langle x, x^* \rangle : x^* \in \mathfrak{M}_f]$ and $\langle a, x^* \rangle$ is independent of $x^* \in \mathfrak{M}_f$ for all $a \in A$.

Proof. Fix $x \in C$, let $y_1, \dots, y_n \in E_x$, and define $y := \frac{1}{n} \sum_i y_i$. Then, by Proposition 2.1, there exists $y^* \in \partial_K f(y)$. Recall the affine function $\varphi(\cdot, y^*) : X \rightarrow \mathbb{R}$ given by

$$\varphi(x; y^*) := \langle x - y, y^* \rangle + f(y)$$

The affine function φ satisfies the following two properties:

- $f(x) \geq \varphi(x; y^*)$ for all $x \in C$, and
- $f(y) = \varphi(y; y^*)$.

The first requirement implies that $f(y_i) \geq \varphi(y_i; y^*)$ for all $i = 1, \dots, n$. Summing up and dividing by n , we see that $\frac{1}{n} \sum_i f(y_i) \geq \frac{1}{n} \sum_i \varphi(y_i; y^*)$. However, f restricted to E_x is affine which implies $\frac{1}{n} \sum_i f(y_i) = f(y)$; similarly, φ is affine, which implies $\frac{1}{n} \sum_i \varphi(y_i; y^*) = \varphi(y; y^*)$.

But we have noted above that $f(y) = \varphi(y; y^*)$, which is possible if, and only if, $f(y_i) = \varphi(y_i; y^*)$ for all $i = 1, \dots, n$. But this is equivalent to saying that $y^* \in \partial_K f(y_i)$.

For any $y \in E_x$, $\partial_K f(y)$ is a (nonempty) closed (and hence compact) subset of $\{x^* \in X^* : \|x^*\| \leq K\}$.³ Thus, $(\partial_K f(y))_{y \in E_x}$ is a collection of closed subsets of the compact set $\{x^* \in X^* : \|x^*\| \leq K\}$. But we have just established that for any $y_1, \dots, y_n \in E_x$, $\bigcap_{i=1}^n \partial_K f(y_i) \neq \emptyset$. In other words, the collection of closed sets $(\partial_K f(y))_{y \in E_x}$ has the finite intersection property. The compactness of $\{x^* \in X^* : \|x^*\| \leq K\}$ then implies that $\bigcap_{y \in E_x} \partial_K f(y) \neq \emptyset$. Thus, there exists $\zeta_x \in \bigcap_{y \in E_x} \partial_K f(y)$ which proves the first part.

Fix this ζ_x and notice that $\varphi(y; \zeta_x) = f(y)$ for all $y \in E_x$. Because $\mathbf{0} \in A$, this implies $\varphi(\mathbf{0}; \zeta_x) = 0$. In other words, $f^*(\zeta_x) = 0$. (In geometric terms, the supporting hyperplane determined by ζ_x passes through the origin.) Now, let $\mathfrak{M}_f := \text{cl}(\{\zeta_x \in X^* : x \in C\})$. It is immediate that \mathfrak{M}_f is closed. Because $f(a) = \langle a, \zeta_x \rangle$ for all $x \in C$, it follows that the same holds for all $x^* \in \mathfrak{M}_f$, which completes the proof. \square

We end with an easy observation.

Lemma 2.5. Let $C \subset X$ be a convex set, and $f \in \mathbb{R}^C$, and \mathfrak{M}_f a weak* compact subset of X^* such that for all $x \in C$, $f(x) = \max[\langle x, x^* \rangle : x^* \in \mathfrak{M}_f]$. (This implies f is convex and Lipschitz of rank K for some K .) Let $C_0 \subset C$ be convex. Then, the following are equivalent.

- (a) The function $f|_{C_0}$ is linear.
- (b) There exists $x_0^* \in \mathfrak{M}_f$ such that $x_0^* \in \bigcap_{x \in C_0} \partial_K f(x)$ (which is equivalent to saying that $f(x) = \langle x, x_0^* \rangle$ for all $x \in C_0$).

Proof. It is easy to see that (b) implies (a). To prove that (a) implies (b), we shall prove the contrapositive. So, suppose $\bigcap_{x \in C_0} \partial_K f(x) = \emptyset$. Then, there exist $x_1, \dots, x_n \in C_0$ such that $\bigcap_{i=1}^n \partial_K f(x_i) = \emptyset$. Let $\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i$.

Then, for all $x^* \in \mathfrak{M}_f$ we have

(3) By the Banach-Alaoglu Theorem — see, for instance, Theorem 6.25 of Aliprantis and Border (1999) — the set $\{x^* \in X^* : \|x^*\| \leq K\}$ is a weak* compact subset of the dual X^* .

- $\langle x_i, x^* \rangle \leq \langle x_i, x_i^* \rangle = f(x_i)$ for all $i = 1, \dots, n$, and
- $\langle x_i, x^* \rangle < \langle x_i, x_i^* \rangle = f(x_i)$ for some $i \in \{1, \dots, n\}$

This implies $\frac{1}{n} \sum_i \langle x_i, x^* \rangle = \langle \bar{x}, x^* \rangle < \frac{1}{n} \sum_i f(x_i)$. Since this is true for all $x^* \in \mathfrak{M}_f$, and because \mathfrak{M}_f is compact, it follows that $f(\bar{x}) = \max[\langle \bar{x}, x^* \rangle : x^* \in \mathfrak{M}_f] < \frac{1}{n} \sum_i f(x_i)$, which proves that f is not linear on C_0 , as claimed. \square

3. Minimal RICs

Let $\hat{\Omega}_n$ be defined for all $n \in \mathbb{N}$ as in Appendix A.5. Define inclusion for $n = 0$ as follows: for $\omega_0, \omega'_0 \in \hat{\Omega}_0$, $\omega_0 \subset_0 \omega'_0$ if $(P, \hat{\omega}) \in \omega_0$ implies $(P, \hat{\omega}) \in \omega'_0$.

Let us inductively define a partial order representing inclusion for all $n \geq 0$: for $\omega_{n+1}, \omega'_{n+1} \in \hat{\Omega}_{n+1}$, let $\omega_{n+1} \subset_{n+1} \omega'_{n+1}$ if $(P, \omega_n) \in \omega_{n+1}$ implies there exists $(P, \omega'_n) \in \omega'_{n+1}$ such that $\omega_{n,s} \subset_n \omega'_{n,s}$ for all $s \in S$.

In analogy with Lemma A.2, it can be shown that $\subset_{n+1} |_{\hat{\Omega}_n} = \subset_n$. As before, then, for $\omega, \omega' \in \hat{\Omega}$, let $\omega \subset^* \omega'$ if $\omega \subset_n \omega'$ for some n with $\omega, \omega' \in \hat{\Omega}_n$.

By definition of $\hat{\Omega}$, there is some n such that $\omega, \omega' \in \hat{\Omega}_n$, and because \subset_n extends faithfully, the precise choice of n is immaterial. Thus, \subset^* is a well defined partial order on $\hat{\Omega}$. We now show that \subset^* has a recursive definition as well.

Proposition 3.1. For any $\omega, \omega' \in \hat{\Omega}$, the following are equivalent.

- $\omega \subset^* \omega'$.
- for all $(P, \tilde{\omega}) \in \omega$, there exists $(P, \tilde{\omega}') \in \omega'$ such that $\tilde{\omega}_s \subset^* \tilde{\omega}'_s$ for all $s \in S$.

Therefore, \subset^* is the *unique* partial order for inclusion on $\hat{\Omega}$ defined as $\omega \subset^* \omega'$ if (b) holds.

The *proof* of Proposition 3.1 is analogous to the proof of Proposition A.3, and so is omitted. Finally, just as in Proposition A.4, \subset^* has a unique continuous extension to Ω . Thus, \subset^* is the unique partial order on Ω that signifies inclusion. Moreover, for $\omega, \omega' \in \Omega$, let $\omega \cap^* \omega'$ represent the \subset^* -greatest lower bound of both ω and ω' . Naturally, \cap^* then represents recursive set intersection.

For $\omega, \omega' \in \Omega$, let $\omega_n := \text{proj}_n \omega$ and $\omega'_n := \text{proj}_n \omega'$. The following is an easy corollary.

Corollary 3.2. For $\omega, \omega' \in \Omega$, $\bar{\omega} := \omega \cap^* \omega'$ if, and only if, $\bar{\omega}_n := \text{proj}_n \bar{\omega} = \omega_n \cap^* \omega'_n$ for all $n \in \mathbb{N}$.

Proof. The ‘only if’ part is straightforward. The ‘if’ part follows from the continuity of \subset^* . \square

Let \approx denote the symmetric part of \succsim , the recursive Blackwell order, and note that \approx is transitive. Then, $\omega \approx \omega'$ if, and only if, ω and ω' recursively Blackwell dominate each other.

Lemma 3.3. Let $\omega_0, \omega'_0 \in \hat{\Omega}_0$ such that $\omega_0 \approx \omega'_0$. Then, $\bar{\omega}_0 \approx \omega_0$.

Proof. It is easy to see that $\bar{\omega}_0 \subset^* \omega_0$, and so $\omega_0 \gtrsim \bar{\omega}_0$ (and similarly for ω'_0). All that remains is to show that $\bar{\omega}_0 \gtrsim \omega_0$.

Towards this end, let $(P^{(0)}, \hat{\omega}) \in \omega_0$, and suppose $(P^{(0)}, \hat{\omega}) \notin \bar{\omega}_0$. Then, because $\omega'_0 \gtrsim \omega_0$, there exists $(P^{(1)}, \hat{\omega}) \in \omega'_0$ such that $P^{(1)}$ is (strictly) finer than $P^{(0)}$. But now because $\omega_0 \gtrsim \omega'_0$, either $(P^{(1)}, \hat{\omega}) \in \omega_0$ and hence $\bar{\omega}_0$, in which case we are done, or there exists $(P^{(2)}, \hat{\omega}) \in \omega_0$ where $P^{(2)}$ is strictly finer than $P^{(1)}$. Continuing in this fashion, we get a sequence $(P^{(j)})$ of strictly finer partitions, where the even members belong to ω_0 (in the obvious sense) and the odd members belong to ω'_0 . But this sequence is finite, and so the final member must belong to both ω_0 as well as ω'_0 , otherwise we would contradict the assumption that $\omega_0 \approx \omega'_0$. Let $P^{(n)}$ be this final member of the sequence. Then, $(P^{(n)}, \hat{\omega}) \in \bar{\omega}_0$, so that $\bar{\omega}_0 \gtrsim \omega$, which proves the claim. \square

A similar result holds for all $\hat{\Omega}_n$.

Lemma 3.4. Let $\omega_n, \omega'_n \in \hat{\Omega}_n$. Then, for all $n \geq 0$, $\omega_n \approx \omega'_n$ implies $\bar{\omega}_n \approx \omega_n$.

Proof. It is easy to see that $\bar{\omega}_n \subset^* \omega_n$, and so $\omega_n \gtrsim \bar{\omega}_n$ (and similarly for ω'_n). All that remains is to show that $\bar{\omega}_n \gtrsim \omega_n$. We shall establish the proof by induction. Suppose that $\omega_n \approx \omega'_n$, and that the result is true for $n - 1$.

Let $(P^{(0)}, \omega_{n-1}^{(0)}) \in \omega_n$, and suppose $(P^{(0)}, \cdot) \notin \bar{\omega}_n$. Then, there exists $(P^{(1)}, \omega_{n-1}^{(1)}) \in \omega'_n$ such that $P^{(1)}$ is finer than $P^{(0)}$ and $\omega_{n-1,s}^{(1)} \gtrsim \omega_{n-1,s}^{(0)}$. Continuing just as we did in Lemma 3.3, we note that there exists a sequence $(P^{(j)}, \omega_{n-1}^{(j)})$ where $P^{(j)}$ is strictly finer than $P^{(j-1)}$ and $\omega_{n-1,s}^{(j)} \gtrsim \omega_{n-1,s}^{(j-1)}$ for all $s \in S$, and where the even members belong to ω_n (in the sense of \subset^*) and the odd members belong to ω'_n . But this sequence is finite, and there must be eventual members of this sequence where $(P^{(m)}, \omega_{n-1}^{(m)}) \in \omega_n$ and $(P^{(m)}, \omega_{n-1}^{(m+1)}) \in \omega'_n$, and $\omega_{n-1,s}^{(m)} \approx \omega_{n-1,s}^{(m+1)}$ for all $s \in S$, because by hypothesis, $\omega_n \approx \omega'_n$. Moreover, we must also have that $P^{(m)}$ is strictly finer than $P^{(0)}$ and $\omega_{n-1,s}^{(m)} \gtrsim \omega_{n-1,s}^{(0)}$.

Then, $(P^{(m)}, \bar{\omega}_{n-1}^{(m)}) \in \bar{\omega}_n$, where $\bar{\omega}_{n-1}^{(m)} := \omega_{n-1,s}^{(m)} \cap \omega_{n-1,s}^{(m+1)}$. But, by the induction hypothesis, $\bar{\omega}_{n-1}^{(m)} \approx \omega_{n-1,s}^{(m)}$. This implies, $\bar{\omega}_n \gtrsim \omega_n$, as claimed. \square

We can now show that the recursive intersection of two recursively Blackwell equivalent RICS is also in the same equivalence class.

Proposition 3.5. For $\omega, \omega' \in \Omega$, $\omega \approx \omega'$ implies $\omega \cap^* \omega =: \bar{\omega} \approx \omega$.

Proof. As in Appendix A.5, let $\omega_n := \text{proj}_n \omega$ and $\omega'_n := \text{proj}_n \omega'$. By Corollary A.5, $\omega_n \approx \omega'_n$ for all $n \geq 0$. Corollary 3.2 implies that $\bar{\omega}_n := \omega_n \cap^* \omega'_n$, and Lemma 3.4 implies $\bar{\omega}_n \approx \omega_n$ for all $n \geq 0$. Corollary A.5 now implies $\bar{\omega} \approx \omega$, as claimed. \square

Let $[\omega] := \{\omega' : \omega' \approx \omega\}$ denote the \approx -equivalence class of ω . Note that $[\omega]$ is a closed (and hence compact) subset of Ω because \gtrsim is continuous. For each $\omega' \in [\omega]$, define the set $\mathcal{D}(\omega') := \{\tilde{\omega} : \tilde{\omega} \subset^* \omega'\}$. We are now ready to prove the existence of \subset^* -minimal RICS.

Proposition 3.6. Each $[\omega]$ has a unique \subset^* -minimal element given by

$$\bigcap_{\omega' \in [\omega]} \mathcal{D}(\omega')$$

Proof. Recall that, by construction, \subset^* is a continuous partial order. Therefore, the set $\mathcal{D}(\omega') := \{\tilde{\omega} : \tilde{\omega} \subset^* \omega'\}$ is closed for each $\omega' \in [\omega]$. Moreover, for any finite collection $\omega^1, \dots, \omega^m \in [\omega]$, the intersection $\bigcap_{i=1}^m \mathcal{D}(\omega^i)$ is non-empty by Proposition 3.5. Thus, the collection of closed sets $(\mathcal{D}(\omega'))_{\omega' \in [\omega]}$ has the finite intersection property. Because Ω is compact, the intersection

$$\bigcap_{\omega' \in [\omega]} \mathcal{D}(\omega')$$

is non-empty. By Proposition 3.5, this intersection must have a unique element, which proves the claim. \square

4. A Metric on the Space of Partitions

In this section, we define a natural metric on the space of partitions that is related to the informational content of the partitions. The metric we introduce is fairly standard. However, we have been unable to find a formulation suitable for our purposes, so we prove that our proposed metric is indeed a metric. It is also worth noting that all the results in this section remain valid if the state space S is an arbitrary countable set, μ is a countably additive measure on S , and \mathcal{P} represents the space of all partitions of S with countably many (measurable) cells.

Let S be a finite set, and \mathcal{P} be the space of all partitions of S . Let μ be a probability measure on S . Define the *entropy* of the partition $P \in \mathcal{P}$ as

$$H(P) := - \sum_{J \in P} \mu(J) \log \mu(J)$$

Let \geq be a partial order on \mathcal{P} , wherein $P \geq Q$ if P is coarser than Q (or equivalently, Q is finer than P). We shall say that $P > Q$ if $P \geq Q$ and $P \neq Q$.

We may also define the coarsest refinement of P and Q , denoted by $P \wedge Q$. If $P = (I_m)$ and $Q = (J_n)$, then $P \wedge Q = (I_m \cap J_n)_{m,n}$, so

$$H(P \wedge Q) = - \sum_m \sum_n \mu(I_m \cap J_n) \log (\mu(I_m \cap J_n))$$

Similarly, $P \vee Q$ is the finest partition coarser than P and Q . Then, $(\mathcal{P}, \geq, \vee, \wedge)$ is a *lattice*, with greatest (coarsest) element $\{S\}$, and least (finest) element $\{\{s\} : s \in S\}$. Notice that $H(\{S\}) = 0$, while $H(P) > 0$ for all other partitions P . Define the *conditional entropy* $H(P | Q)$ as

$$H(P | Q) := H(P \wedge Q) - H(Q)$$

It is easy to see that

$$H(P | Q) = - \sum_n \mu(J_n) \sum_m \frac{\mu(I_m \cap J_n)}{\mu(J_n)} \log \left(\frac{\mu(I_m \cap J_n)}{\mu(J_n)} \right)$$

We now come to the main result of this section.

Proposition 4.1. The function

$$d(P, Q) := 2H(P \wedge Q) - H(P) - H(Q) = H(P | Q) + H(Q | P)$$

is a metric on \mathcal{P} .

We begin with some lemmata.

Lemma 4.2. H is anti-monotone, ie, $P \geq Q$ implies $H(Q) \geq H(P)$. Moreover, H is strictly anti-monotone, ie, $P > Q$ implies $H(Q) > H(P)$.

The proof is trivial and is omitted.

Lemma 4.3. The function $H(P | Q)$ is anti-monotone in P , and is monotone in Q .

Proof. Notice that if $P' \geq P$, then $P' \wedge Q \geq P \wedge Q$, so the anti-monotonicity of H implies that $H(P | Q)$ is anti-monotone in P . We say that Q is an *elementary refinement* of Q' if $Q' = \{J_1, \dots, J_N\}$ and $Q = \{\tilde{J}_1, \dots, \tilde{J}_{n-1}, \tilde{J}_N, \tilde{J}_{N+1}\}$, where $\tilde{J}_n := J_n$ for all $n = 1, \dots, N-1$, while $J_N = \tilde{J}_N \cup \tilde{J}_{N+1}$. In other words, Q and Q' are identical except that there exists a cell $J_N \in Q'$ that is the union of exactly two cells in Q .

Let $Q' \geq Q$. Then, there exist $Q_1, \dots, Q_k \in \mathcal{P}$ such that $Q' = Q_k \geq Q_{k-1} \geq \dots \geq Q_1 = Q$, and where Q_i is an elementary refinement of Q_{i+1} . Thus, in order to show that $H(P | Q)$ is monotone in Q , it suffices to consider Q and Q' where Q is an elementary refinement of Q' .

Let $P = \{I_1, \dots, I_M\}$ and Q and Q' be as above. In what follows, we shall let $\eta(x) = x \log x$ for all $x > 0$ and $\eta(0) = 0$. Then $\eta \in \mathbb{R}^{\mathbb{R}^+}$ is strictly convex and continuous on its domain. Let

$$\Lambda = - \sum_{n=1}^{N-1} \mu(J_n) \sum_m \eta \left(\frac{\mu(I_m \cap J_n)}{\mu(J_n)} \right)$$

This allows us to write

$$\begin{aligned}
H(P \mid Q) &= - \sum_n \mu(\tilde{J}_n) \sum_m \frac{\mu(I_m \cap \tilde{J}_n)}{\mu(\tilde{J}_n)} \log \left(\frac{\mu(I_m \cap \tilde{J}_n)}{\mu(\tilde{J}_n)} \right) \\
&= \Lambda - \sum_{n=N, N+1} \mu(\tilde{J}_n) \sum_m \eta \left(\frac{\mu(I_m \cap \tilde{J}_n)}{\mu(\tilde{J}_n)} \right) \\
&= \Lambda - \mu(J_N) \sum_m \sum_{n=N, N+1} \frac{\mu(\tilde{J}_n)}{\mu(J_N)} \eta \left(\frac{\mu(I_m \cap \tilde{J}_n)}{\mu(\tilde{J}_n)} \right) \\
&\leq \Lambda - \mu(J_N) \sum_m \eta \left(\sum_{n=N, N+1} \frac{\mu(I_m \cap \tilde{J}_n)}{\mu(J_N)} \right) \\
&= \Lambda - \mu(J_N) \sum_m \eta \left(\frac{\mu(I_m \cap J_N)}{\mu(J_N)} \right) \\
&= H(P \mid Q')
\end{aligned}$$

where we have used the fact that $-\eta$ is concave to establish the inequality. \square

Lemma 4.4. The function H is submodular, ie, $H(P \wedge Q) + H(P \vee Q) \leq H(P) + H(Q)$.

Proof. Fix P and Q , and let $Q \leq Q'$. We shall use the fact that the function $H(P \mid Q)$ is anti-monotone in P and monotone in Q . Then, $H(P \wedge Q) - H(Q) = H(P \mid Q) \leq H(P \mid Q') = H(P \wedge Q') - H(Q')$. Now set, $Q' := P \vee Q$, so that $P \wedge (P \vee Q) = P$, which implies $H(P \wedge Q) - H(Q) \leq H(P) - H(P \vee Q)$. Therefore, H is submodular. \square

We now list some properties of the lattice $(\mathcal{P}, \geq, \vee, \wedge)$.

Lemma 4.5. For $P, Q, R \in \mathcal{P}$, the following hold:

- (a) $R \geq (P \wedge R) \vee (Q \wedge R)$.
- (b) $(P \vee Q) \wedge R \geq (P \wedge R) \vee (Q \wedge R)$.

Proof. Note that $R \geq P \wedge R$ and $R \geq Q \wedge R$, so $R \geq (P \wedge R) \vee (Q \wedge R)$, which establishes (a). To see (b), note that $P \geq P \wedge R$, while $Q \geq Q \wedge R$. Therefore, $P \vee Q \geq (P \wedge R) \vee (Q \wedge R)$. But we also have that $R \geq (P \wedge R) \vee (Q \wedge R)$, from (a). The definition of \wedge then implies that $(P \vee Q) \wedge R \geq (P \wedge R) \vee (Q \wedge R)$, as required. \square

Proof of Proposition 4.1. The proof relies on the fact that conditional entropy $H(P \mid Q)$ is anti-monotone (Lemmas 4.2 and 4.3) and submodular (Lemma 4.4). Because H is anti-monotone (Lemma 4.2), $d(P, Q) \geq 0$ for all P, Q . We have already established that $P < Q$ implies $H(P) > H(Q)$. If P and Q are distinct, then $P \wedge Q$ is distinct from either P or Q , so that $d(P, Q) > 0$.

It is easy to see that $d(P, Q) \leq d(P, R) + d(R, Q)$ if, and only if,

$$[\heartsuit] \quad H(P \wedge Q) + H(R) \leq H(P \wedge R) + H(Q \wedge R)$$

By lemma 4.5, we see that $R \geq (P \wedge R) \vee (Q \wedge R)$ and $(P \vee Q) \wedge R \geq (P \wedge R) \vee (Q \wedge R)$. Set $P' = P \wedge R$ and $Q' = Q \wedge R$. The submodularity of H implies $H(P' \vee Q') + H(P' \wedge Q') \leq H(P') + H(Q')$. That is, $H((P \wedge R) \vee (Q \wedge R)) + H(P \wedge Q \wedge R) \leq H(P \wedge R) + H(Q \wedge R)$. But $R \geq (P \wedge R) \vee (Q \wedge R)$, so $H(R) \leq H((P \wedge R) \vee (Q \wedge R))$. Similarly, $P \wedge Q \geq P \wedge Q \wedge R$, which implies $H(P \wedge Q) \leq H(P \wedge Q \wedge R)$. These observations imply $[\heartsuit]$, so that d is a metric. \square

5. Consumption Streams and the RAA Representation

To see that $L \simeq \mathcal{F}(\Delta(C \times L))$, note that we can define $L^{(1)} := \mathcal{F}(\Delta(C))$ and then recursively define $L^{(n)} := \mathcal{F}(\Delta(C \times L^{(n-1)}))$ as the space of consumption streams of length n . Just as with the definition of the space of RACPS X in Appendix A.2, we say that L is the space of all *consistent* sequences in $\times_{n=1}^{\infty} L^{(n)}$.

The *support* of a consumption stream $\ell \in L$ is a set $B \subset C$ such that at any date and in any state, the realized consumption lies in B . A consumption stream has *finite support* if its support in C is finite. For any finite set $B \subset C$, we can define L_B as the space of all consumption streams with prizes in B . Formally, $L_B \simeq \mathcal{F}(\Delta(C \times L_B))$. Let L_0 be the space of all consumption streams with finite support. That is, $L_0 := \bigcup \{L_B : B \subset C, B \text{ finite}\}$.

Recall the consumption stream $\ell^\dagger \in L$ which delivers $c^\dagger(s)$ in state s at every date. Clearly, the support of ℓ^\dagger is finite. Analogous to L_0 , we can define $L_0^{(n)}$ as the space of consumption streams of length n with finite support. For any $\ell^{(n)} \in L_0^{(n)}$, $\ell^{(n)} \diamond \ell^\dagger \in L_0$, where $\ell^{(n)} \diamond \ell^\dagger$ is the concatenation of ℓ^\dagger to $\ell^{(n)}$. In other words, each $L^{(n)}$ is naturally embedded in L_0 .

Proposition 5.1. The space L_0 is dense in L .

Proof. Because probability measures on C with finite support are dense in $\Delta(C)$, it follows that for all $n \geq 1$, $L_0^{(n)}$ is dense in $L^{(n)}$. (The metrics defined on $L^{(n)}$ make this clear — see Appendix A.2 for a formal definition.) By the definition of the product metric (see Appendix A.2), this means that for any $\ell \in L$ and $\varepsilon > 0$, there exists an n and an $\ell^{(n)} \in L^{(n)}$ such that $d(\ell, \ell^{(n)} \diamond \ell^\dagger) < \varepsilon$, where $\ell^{(n)} \diamond \ell^\dagger$ is the concatenation of ℓ^\dagger to $\ell^{(n)}$. This completes the proof. \square

It follows immediately from Lipschitz Continuity (Axiom 1(c)) that $\succsim|_L$ is non-trivial, see Corollary 1.3. We now show that \succsim_s (as defined in Section 3.1) is also non-trivial for each $s \in S$.

Lemma 5.2. Let $\ell^0, \ell^1 \in L$. Then, $\ell^0(s) \sim_s \ell^1(s)$ for all $s \in S$ implies $\ell^0 \sim|_L \ell^1$.

Proof. By definition of \succsim_s , $\ell^0(s) \sim_s \ell^1(s)$ if, and only if, $\ell^0 \oplus_{(1, S \setminus s)} \ell_* \sim |_L \ell^1 \oplus_{(1, S \setminus s)} \ell_*$. Repeatedly applying L-Independence, we find

$$\frac{1}{n} \ell^0 + \frac{n-1}{n} \ell_* = \frac{1}{n} \sum_{s \in S} \ell^0 \oplus_{(1, S \setminus s)} \ell_* \sim |_L \frac{1}{n} \sum_{s \in S} \ell^1 \oplus_{(1, S \setminus s)} \ell_* = \frac{1}{n} \ell^1 + \frac{n-1}{n} \ell_*$$

By L-Independence, we find $\ell^0 \sim |_L \ell^1$. (More precisely, this follows immediately once we note that, by the Mixture Space Theorem, $\succsim |_L$ has an affine representation.) \square

Lemma 5.3. There exists $s \in S$ such that $\ell^*(s) \approx_s \ell_*(s)$. For all $s \in S$, there exists $s' \in S$ such that $(c, \ell^* \oplus_{(1, S \setminus s)} \ell_*) \approx_{s'} (c, \ell_*)$.

Proof. Corollary 1.3 says that $\ell^* \succ |_L \ell_*$. Therefore, by (the contrapositive to) Lemma 5.2, there must exist an s such that $\ell^*(s) \approx_s \ell_*(s)$. In particular, then, $\ell^* \oplus_{(1, S \setminus s)} \ell_* \approx |_L \ell_*$.

To see the second part, let us suppose by way of contradiction that for all $s' \in S$, $(c, \ell^* \oplus_{(1, S \setminus s)} \ell_*) \not\sim_{s'} (c, \ell_*)$. Now, set ℓ^0, ℓ^1 such that $\ell^0(s') = (c, \ell^* \oplus_{(1, S \setminus s)} \ell_*)$, while $\ell^1(s') = (c, \ell_*)$. It follows from Lemma 5.2 that $(c, \ell^* \oplus_{(1, S \setminus s)} \ell_*) \sim |_L (c, \ell_*)$.

Now, L-Stationarity (Axiom 2) and the fact that $\ell^* \oplus_{(1, S \setminus s)} \ell_* \approx |_L \ell_*$ imply that we have $(c, \ell^* \oplus_{(1, S \setminus s)} \ell_*) \approx |_L (c, \ell_*)$, which yields the desired contradiction. \square

Proposition 5.4. For all $s \in S$, \succsim_s is non-trivial.

Proof. Lemma 5.3 and (the contrapositive to) L-History Independence (Axiom 2) imply $(c, \ell^* \oplus_{(1, S \setminus s)} \ell_*) \approx_{s''} (c, \ell_*)$ for all $s'' \in S$, as claimed. \square

Proposition 5.5. The preference $\succsim |_L$ on L has a standard RAA representation. Moreover, Π and δ are unique and the collection $(u_s)_{s \in S}$ is unique up to a common positive scaling.

As described in Section 3.1, for each $s \in S$, \succsim_s is an induced preference over $\Delta(C \times L)$. Let \succsim_s^C denote the induced preference over $\Delta(C)$ in state s . It is clear that \succsim_s^C is well defined, continuous on $\Delta(C)$, and satisfies Independence. These properties imply there exist \succsim_s^C -maximal and -minimal lotteries that are degenerate; denote them by $c^*(s)$ and $c_*(s)$. Let F_0 be the finite set of consumption defined as

$$F_0 := \{c_*(s), c^\dagger(s), c^*(s) : s \in S\}$$

Lemma 5.6. For any finite set $B \subset C$, the induced preference $\succsim |_{L_B}$ satisfies the Axioms stated in Corollary 5 of Krishna and Sadowski (2014, henceforth KS).

Proof. It follows from Proposition 5.4 that each \succsim_s is non-trivial. That is, $\succsim |_L$ is state-wise nontrivial. In addition, $\succsim |_L$ is continuous, satisfies Independence, and is separable in ℓ_1 and ℓ_2 , thereby satisfying Axioms 2, 3, and 5 in KS. Axioms 6, 7, and 9 in KS correspond to properties (c), (d), and (b) of L-Properties (Axiom 2). \square

We now proceed to the proof of Proposition 5.5.

Proof of Proposition 5.5. Let $B \subset C$ be finite. By Lemma 5.6, $\succsim|_{L_B}$ satisfies the Axioms in Corollary 5 of KS. This implies there exists a tuple $((u_s^B)_{s \in S}, \delta^B, \Pi^B)$ that is an RAA representation of $\succsim|_{L_B}$. If $F_0 \subset B$, then we may assume, without loss of generality, that $u_s^B(c^\dagger(s)) = 0$ for all $s \in S$. Then, Corollary 5 in KS says that the collection of utilities (u_s^B) is uniquely identified up to a joint scaling, and that Π^B and δ^B are also uniquely determined.

Now, consider any other finite set D such that $F_0 \subset B \subset D$. By Lemma 5.6, $\succsim|_{L_D}$ also has an RAA representation $((u_s^D)_{s \in S}, \delta^D, \Pi^D)$. As before, if we set $u_s^D(c^\dagger(s)) = 0$ for all $s \in S$, then the collection of utilities (u_s^D) is identified up to a common scaling. Now, because $B \subset D$, we have $L_B \subset L_D$. Therefore, the RAA representation $((u_s^D)_{s \in S}, \delta^D, \Pi^D)$ of $\succsim|_{L_D}$ when restricted to L_B , is also a representation of $\succsim|_{L_B}$. And this representation has the feature that $u_s^D(c^\dagger(s)) = 0$ for all $s \in S$. Once again, the uniqueness of the RAA representation implies that a single joint scaling of the collection (u_s^D) results in $u_s^D|_B = u_s^B$ for all $s \in S$, $\Pi^B = \Pi^D$, and $\delta^B = \delta^D$.

Recall that $c^*(s) \succsim_s^C \alpha \succsim_s^C c_*(s)$ for all $\alpha \in \Delta(C)$. Because u_s^B and u_s^D represent, respectively, $\succsim_s^C|_{\Delta(B)}$ and $\succsim_s^C|_{\Delta(D)}$, it must be that $\lambda^*(s) := u_s^j(c^*(s))$ and $\lambda_*(s) := u_s^j(c_*(s))$ for $j = B, D$. Since B and D are arbitrary, it follows that it holds for all finite B that contains F_0 . In other words, the Markov transition operator Π has been identified uniquely, as has the discount factor $\delta \in (0, 1)$.

Let $u_s \in \mathbf{C}(C)$ be a vN-M utility representation of \succsim_s^C such that $u_s(c^\dagger(s)) = 0$. Both $u_s|_{\Delta(B)}$ and u_s^B are vN-M representations of $\succsim_s^C|_{\Delta(B)}$ and by the Mixture Space Theorem, differ at most by a positive affine transformation. Because they agree on $c^\dagger(s)$, they differ at most by a positive scaling. Therefore, if we scale u_s so that $u_s(c^*(s)) = \lambda^*(s)$, we must necessarily have $u_s(c_*(s)) = \lambda_*(s)$ for all $s \in S$.

Consider, now, the tuple $((u_s), \Pi, \delta)$, and the functional $W_0 : L \rightarrow \mathbb{R}$ defined as $W_0(\ell) := \sum_s \pi_0(s)W(\ell, s)$, where

$$W(\ell, s) := \sum_{s'} \Pi(s, s') [u_{s'}(\ell_1(s')) + \delta W(\ell_2(s'), s')]$$

It is easy to see that the function W_0 ⁴ is uniquely determined by the tuple $((u_s)_{s \in S}, \delta, \Pi)$. As established above, W_0 represents $\succsim|_{L_B}$ for every finite B . In other words, W_0 represents $\succsim|_{L_0}$. Proposition 5.1 says that L_0 is dense in L , and because W_0 is (uniformly) continuous, it also represents \succsim on L . The uniqueness of the RAA representation of $\succsim|_L$ (given our normalizations) follows immediately, which concludes the proof. \square

6. Partitional Representation – Proof Details

In this section, we prove Proposition C.1 in Appendix C.1 of DKS. We begin with some lemmas.

(4) As always, W_0 also denotes the linear extension of W_0 to $\Delta(L)$.

Let $\tilde{\mathcal{E}}_x := \{\xi \in \mathcal{E}_x : \mathcal{F}(\xi) \sim x\}$. By IICC (Axiom 4), $\tilde{\mathcal{E}}_x$ is non-empty. It follows from the definition of $\tilde{\mathcal{E}}_x$ that for each $\xi \in \tilde{\mathcal{E}}_x$, there exist $f_1, \dots, f_m \in x$ such that for each $i = 1, \dots, m$, $f_i = \xi(s)$ for some $s \in S$. The collection $\{f_1, \dots, f_m\}$ denotes a set of *generators* of the set x according to ξ . We shall also say that $\{f_1, \dots, f_m\}$ *generates* x according to ξ .

Lemma 6.1. For $x \in X^*$, let $\{f_1, \dots, f_m\}$ generate x according to $\xi \in \tilde{\mathcal{E}}_x$. Then, $x \sim \{f_1, \dots, f_m\}$.

Proof. Notice that

$$\begin{array}{ll} x \succsim \{f_1, \dots, f_m\} & \text{by Monotonicity (Axiom 1(d))} \\ \succsim \mathcal{F}(\xi) & \text{by IICC(a) and Continuity} \\ \sim x & \text{by IICC(b)} \end{array}$$

which establishes the claim. \square

Definition 6.2. A menu x is *nice* if $x \in X^*$ and there is a unique $\xi \in \tilde{\mathcal{E}}_x$. X_0 denotes the space of nice menus. A menu x is *minimal* if $x \succ x \setminus \{f\}$ for all $f \in x$.

Let x be a nice menu, $\xi \in \tilde{\mathcal{E}}_x$, and f_1, \dots, f_m the corresponding generators of x . Each such ξ induces a partition J_1, \dots, J_m of S wherein $\xi(s) = f_k$ if, and only if, $s \in J_k$. In this case, we shall say that f_k is *active* in state $s \in J_k$, so that J_k denotes all the states where f_k is active.

Proposition 6.3. The space X_0 is dense in X .

Proof. It is easy to see that the space X^* is dense in X . Therefore, it will suffice to show that X_0 is dense in X^* . For any $x \in X^*$, it is easy to see that IICC (Axiom 4) implies the existence of a minimal set of generators, $\{f_1, \dots, f_m\}$. Let $x_\varepsilon := (1 - \varepsilon)x + \varepsilon\ell_*$ and $y := \{f_1, \dots, f_m\} \cup x_\varepsilon$. By Monotonicity (Axiom 1(d)), $y \succsim x$. Obviously $d(y, x) \rightarrow 0$ as $\varepsilon \rightarrow 0$. Because $x \in X^*$ and $\varepsilon > 0$ are arbitrary, it suffices to establish that (some perturbation of) $y \in X_0$.

Because $x \in X^*$, we also have $x_\varepsilon \in X^*$ and, because $\{f_1, \dots, f_m\} \subset x$, also $\{f_1, \dots, f_m\} \in X^*$, which then implies $y \in X^*$. We now show that there must be a unique $\xi \in \tilde{\mathcal{E}}_y$ (perhaps after further perturbing y) to establish the proposition.

Suppose there is $\xi \in \tilde{\mathcal{E}}_y$ with generator set

$$\{f'_1, \dots, f'_j, (1 - \varepsilon)f'_{j+1} + \varepsilon\ell_*, \dots, (1 - \varepsilon)f'_k + \varepsilon\ell_*\} \sim y$$

(indifference follows from Lemma 6.1) where $f'_a \in x$ for all $a \in \{1, \dots, k\}$. Consider, now, $\mathcal{F}(\xi)$ and note that it can be generated inductively from $y_0 := \{f'_1, \dots, f'_k\}$ as follows, where

the induction is over the set of states $S = \{s_1, \dots, s_n\}$. For $i \in \{1, \dots, n\}$, let $e_i : y \rightarrow [0, 1]$ be defined by

$$e_i(f) := \begin{cases} 0 & \text{if } f = \xi(s_i) \text{ and } f \in \{f_1, \dots, f_m\} \\ \varepsilon & \text{if } (1 - \varepsilon)f + \varepsilon \ell_* = \xi(s_i) \notin \{f_1, \dots, f_m\} \\ 1 & \text{otherwise} \end{cases}$$

Given y_i , let

$$y_{i+1} := y_i \oplus_{(e_{i+1}, s_{i+1})} \ell_*$$

Observe that, indeed, $y_n = \mathcal{F}(\xi)$. Note, further, that by IICC (part a) and Continuity (Axiom 1(b)), $y_i \succsim y_{i+1}$, with $y_i \succ y_{i+1}$ if $\xi(s_i) \in x_\varepsilon$. Suppose, now, that $k > j$. In that case, $y_0 \succ y_n = \mathcal{F}(\xi) \sim y$. By Monotonicity, $x \succsim y_0$, and hence $x \succ y$, which contradicts the observation above that $y \succsim x$. Therefore, $m = j$. But then $y \succsim x$ and the minimality of $\{f_1, \dots, f_m\}$ implies that the generator set that corresponds to ξ must be $\{f_1, \dots, f_m\}$. Because ξ was chosen arbitrarily among the $\xi \in \tilde{\mathcal{E}}_y$, any such ξ must have generator set $\{f_1, \dots, f_m\}$.

Suppose, then, that there are $\xi, \xi' \in \tilde{\mathcal{E}}_y$ with the same generator set $\{f_1, \dots, f_m\}$, and $f_b = \xi(s) \neq \xi'(s)$ for some $s \in S$ and $b \in \{1, \dots, m\}$. Let

$$\hat{f}_b(s') := \begin{cases} f_b(s') & s' \neq s \\ (1 - t)f_b + t\ell_* & s' = s \end{cases}$$

Note that, by Continuity, for $t > 0$ small enough, $\{f_1, \dots, \hat{f}_b, \dots, f_m\}$ remains the unique generator set for $\hat{y} := [y \setminus \{f_b\}] \cup \{\hat{f}_b\}$. Let $\hat{\xi} \in \tilde{\mathcal{E}}_{\hat{y}}$ be the contingent plan with

$$\hat{\xi}(s') := \begin{cases} \hat{f}_b(s') & \xi(s') = f_b \\ \xi(s') & \text{otherwise} \end{cases}$$

and analogously for $\hat{\xi}'$ and ξ' . Then IICC (part a) implies that $y \succ \mathcal{F}(\hat{\xi})$. At the same time $\mathcal{F}(\hat{\xi}') = \mathcal{F}(\xi') \sim y$. It is also clear that, for $t > 0$ small enough and by Continuity, for any $\xi'' \in \tilde{\mathcal{E}}_y$ with $\mathcal{F}(\xi'') \approx y$, also $\mathcal{F}(\hat{\xi}'') \approx \hat{y}$, where $\hat{\xi}''$ is again defined analogously. Hence, $\tilde{\mathcal{E}}_{\hat{y}}$ has at least one element less than $\tilde{\mathcal{E}}_y$. In finitely many steps we arrive at an (arbitrarily small) perturbation of y that is in X_0 . This establishes the proposition. \square

A (static) *strategy* for DM at a menu x given $\mu \in \mathfrak{M}$ is a mapping $\zeta_x^\mu : \mathfrak{U} \rightarrow x$. The strategy ζ_x^μ is *partitional* if there is a finite partition (E_i) of \mathfrak{U} , such that for each E_i there exists $f_i \in x$ with $\zeta_x^\mu(E_i) = f_i$. The value of this strategy is

$$V(x, \mu, \zeta_x^\mu) = \sum_i \int_{E_i} \sum_s p_s u_s(f_i(s)) d\mu(p, u)$$

A strategy ζ_x^μ is *optimal* at x if there is no other strategy that gives a higher payoff. A *partitional optimal strategy* ζ_x^μ is an optimal strategy that is partitional, ie, one where

$$\begin{aligned} V(x, \mu, \zeta_x^\mu) &= \sum_i \int_{E_i} \langle (p, u), f_i \rangle d\mu(p, u) \\ &= \max_{\mu \in \mathfrak{M}} \left[\int_{\mathfrak{U}} \max_{f \in x} \langle (p, u), f \rangle d\mu(p, u) \right] \end{aligned}$$

where $\langle (p, u), f \rangle = \sum_s p_s u_s (f_i(s))$. Notice that if a partitional strategy ζ_x^μ is optimal at x and if f_i is the act chosen in the cell E_i , we must necessarily have, for all $(p, u) \in E_i$, $\langle (p, u), f_i \rangle \geq \langle (p, u), f \rangle$ for all $f \in x$.

In the sequel, ζ_x^μ denotes an optimal partitional strategy when one exists. It is easy to see that for a finite x , an optimal strategy is always partitional, though there may be many such strategies that are optimal. If ζ_x^μ induces the partition (E_i) , we refer to (E_i) as an optimal partition for μ at x .

Definition 6.4. Let $\{f_1, \dots, f_m\}$ be a set of generators of x , and let $(E_i)_{i=1}^m$ be a partition of \mathfrak{U} . Then, (E_i) is a *partition of \mathfrak{U} consistent with $\{f_1, \dots, f_m\}$* if $(p, u) \in E_i$ implies $\langle (p, u), f_i \rangle \geq \langle (p, u), f_j \rangle$ for all $j = 1, \dots, m$.

Intuitively, a partition (E_i) of \mathfrak{U} is consistent with $\{f_1, \dots, f_m\}$ if there is some optimal μ such that it is optimal to choose f_i when $(p, u) \in E_i$. As in Appendix C of DKS, $\mathcal{Y} : X \rightrightarrows \mathfrak{M}$ is the mapping selecting the maximizing μ for each x ; that is, $\mathcal{Y}(x) := \arg \max_{\mu \in \mathfrak{M}} V(x, \mu)$. The following lemma implies that finite menus always have consistent partitions.

Lemma 6.5. Let $x \in X$ be finite and suppose $\{f_1, \dots, f_m\}$ is a set of generators of x . Then, $\mu \in \mathcal{Y}(\{f_1, \dots, f_m\})$ implies $\mu \in \mathcal{Y}(x)$.

Proof. Consider the following string of inequalities:

$$\begin{aligned} V(x) &= V(\{f_1, \dots, f_m\}) && \text{because } \{f_1, \dots, f_m\} \text{ generates } x \\ &= V(\{f_1, \dots, f_m\}, \mu) && \text{definition of } \mu \\ &\leq V(x, \mu) && V(\cdot, \mu) \text{ is monotone} \\ &\leq V(x) && \text{definition of } V \end{aligned}$$

which proves that $\mu \in \mathcal{Y}(x)$, as claimed. \square

Lemma 6.6. Let x be finite. For any $\ell \in L$ and $\varepsilon > 0$, (i) $\mathcal{Y}(x) = \mathcal{Y}((1 - \varepsilon)x + \varepsilon\ell)$, (ii) if x is nice, then $(1 - \varepsilon)x + \varepsilon\ell$ is also nice, and (iii) if $\mu \in \mathcal{Y}(x)$ and (E_i) is an optimal partition for μ at x , then it is also an optimal partition for μ at $(1 - \varepsilon)x + \varepsilon\ell$.

Proof. Let x be finite and $\mu \in \mathfrak{M}(x)$. Then, $V(x) = V(x, \mu) \geq V(x, \mu')$ for all $\mu' \in \mathfrak{M}$. We also have

$$\begin{aligned} V((1-\varepsilon)x + \varepsilon\ell, \mu) &= (1-\varepsilon)V(x, \mu) + \varepsilon V(\ell, \mu) \\ &\geq (1-\varepsilon)V(x, \mu') + \varepsilon V(\ell, \mu) \\ &= (1-\varepsilon)V(x, \mu') + \varepsilon V(\ell, \mu') \\ &= V((1-\varepsilon)x + \varepsilon\ell, \mu') \end{aligned}$$

where the inequality uses the fact that $V(x, \mu) \geq V(x, \mu')$ and the second equality follows because $V(\ell, \mu) = V(\ell, \mu')$ for all $\mu, \mu' \in \mathfrak{M}$ and $\ell \in L$. This proves part (i). Part (ii) follows immediately from the definition.

To see part (iii), let ζ_x^μ be a partitional optimal strategy with optimal partition (E_i) . Then,

$$V(x) = V(x, \mu, \zeta_x^\mu) = \sum_i \int_{E_i} \langle (p, u), f_i \rangle d\mu(p, u)$$

For the menu $(1-\varepsilon)x + \varepsilon\ell$, consider the strategy $\zeta_{(1-\varepsilon)x + \varepsilon\ell}^\mu(E_i) = (1-\varepsilon)f_i + \varepsilon\ell$. Then,

$$\begin{aligned} &V((1-\varepsilon)x + \varepsilon\ell, \mu, \zeta_{(1-\varepsilon)x + \varepsilon\ell}^\mu) \\ &= (1-\varepsilon) \sum_i \int_{E_i} \langle (p, u), f_i \rangle d\mu(p, u) + \varepsilon \sum_i \int_{E_i} \langle (p, u), \ell \rangle d\mu(p, u) \\ &= (1-\varepsilon)V(x) + \varepsilon V(\ell) \\ &\geq V((1-\varepsilon)x + \varepsilon\ell, \mu') \end{aligned}$$

for all $\mu' \in \mathfrak{M}$ where the second equality follows from part (i). This proves that $\zeta_{(1-\varepsilon)x + \varepsilon\ell}^\mu$ is a partitional optimal strategy at the menu x given the optimal $\mu \in \mathfrak{M}$ and completes the proof. \square

For a fixed partition (E_i) of \mathfrak{U} , $\mu \in \mathfrak{M}$, and $s \in S$, consider the map

$$(\mu, E_i, s) \mapsto \int_{E_i} p_s u_s(\cdot) d\mu(p, u)$$

Each tuple (μ, E_i, s) induces a continuous and linear preference functional $\int_{E_i} p_s u_s(\cdot) d\mu(p, u)$ on $\Delta(C \times X)$. By the Expected Utility Theorem, this linear functional has a vN-M utility representation which we denote by $\bar{p}_i(s) \bar{u}_{i,s}(\cdot)$, where $\|\bar{u}_{i,s}\|_\infty = 1$. Thus, for all $\alpha \in \Delta(C \times X)$, we have

$$\bar{p}_i(s) \bar{u}_{i,s}(\alpha) = \int_{E_i} p(s) u_s(\alpha) d\mu(p, u)$$

Then, $\bar{p}_i(s) \bar{u}_{i,s}$ is a *local EU* representation of μ on E_i for state s . We do not index $\bar{p}_i(s) \bar{u}_{i,s}$ by the relevant (E_i) and μ because these should be clear from the context.

Definition 6.7. Let $\mu \in \mathfrak{M}$ and (E_i) a partition of \mathfrak{U} . Then,

- A measure μ is *Type Ia* on E_i in state s if $\bar{p}_i(s)\bar{u}_{i,s} = \mathbf{0}$, ie, if $\bar{p}_i(s)\bar{u}_{i,s}$ is trivial.
- A measure μ is *Type Ib* on E_i in state s if $\bar{p}_i(s)\bar{u}_{i,s}$ is non-trivial, $\bar{p}_i(s)\bar{u}_{i,s}$ is constant on $\Delta(C \times L)$, and ℓ_* maximizes $\bar{p}_i(s)\bar{u}_{i,s}$ on $\Delta(C \times X)$.
- A measure μ is *Type IIa* on E_i in state s if $\bar{p}_i(s)\bar{u}_{i,s}$ is non-trivial and not constant on $\Delta(C \times L)$.
- A measure μ is *Type IIb* on E_i in state s if $\bar{p}_i(s)\bar{u}_{i,s}$ is non-trivial, constant on $\Delta(C \times L)$, and there exists $\alpha \in \Delta(C \times X)$ such that $\bar{p}_i(s)\bar{u}_{i,s}(\alpha) > \bar{p}_i(s)\bar{u}_{i,s}(\beta)$ for some (and hence all) $\beta \in \Delta(C \times L)$.

It is easy to see that the above taxonomy of measures is both mutually exclusive and exhaustive. Analogous to the definition in Section 3.1 of DKS (and abusing notation), for any $\alpha \in \Delta(C \times X)$ we define

$$(f \oplus_{\varepsilon,s} \alpha)(s') := \begin{cases} (1 - \varepsilon)f(s) + \varepsilon\alpha & \text{if } s' = s \\ f(s) & \text{otherwise} \end{cases}$$

Lemma 6.8. Let x be a finite menu, $\mu \in \mathcal{Y}(x)$, and suppose there is a partitional optimal strategy ζ_x^μ with optimal partition (E_i) , where $\zeta_x^\mu(E_i) = f_i \in x$. Suppose μ is Type II (a or b) on some E_i in state $s \in S$ and there exists $\alpha \in \Delta(C \times X)$ such that

$$\int_{E_i} p(s)u_s(\alpha - f_i(s)) d\mu(p, u) > 0$$

Then, the menu $z := x \setminus \{f_i\} \cup \{f_i \oplus_{\varepsilon,s} \alpha\}$ is such that $V(z) > V(x)$ for all $\varepsilon > 0$.

Proof. Let $\mu \in \mathcal{Y}(x)$ so that $V(x) = V(x, \mu)$. If

$$\int_{E_i} p(s)u_s(\alpha - f_i(s)) d\mu(p, u) > 0$$

then it must necessarily be that $\mu(E_i) > 0$. The measure μ and the set E_i induce the functional

$$V_i(x, \mu, E_i) := \int_{E_i} \max_{f \in x} \sum_s p(s)u_s(f(s)) d\mu(p, u)$$

on X . Let V_i^0 denote the restriction of V_i to $\mathcal{F}(\Delta(C \times X))$. By construction,

$$V_i^0(f) = \int_{E_i} \sum_s p(s)u_s(f(s)) d\mu(p, u)$$

and because $\mu(E_i) > 0$, V_i^0 is non-trivial. By hypothesis, we have $V_i^0(f \oplus_{\varepsilon,s} \alpha) > V_i^0(f_i)$.

Consider the menu z and the strategy which entails the choice of f_j for $(p, u) \in E_j$ when $j \neq i$, and the choice of $f_i \oplus_{\varepsilon,s} \alpha$ when $(p, u) \in E_i$. This strategy delivers utility

bounded above by $V(z, \mu)$, ie,

$$\begin{aligned}
V(z, \mu) &\geq \sum_{j \neq i} \left[\int_{E_j} \sum_s p(s) u_s f_j(s) \, d\mu(p, u) \right] + \int_{E_i} \sum_s p(s) u_s (f_i(s)) \, d\mu(p, u) \\
&\quad + \varepsilon \int_{E_i} p(s) u_s (\alpha - f_i(s)) \, d\mu(p, u) \\
&= V(x) + \varepsilon \int_{E_i} p(s) u_s (\alpha - f_i(s)) \, d\mu(p, u) \\
&> V(x)
\end{aligned}$$

because $\int_{E_i} p(s) u_s (\alpha - f_i(s)) \, d\mu(p, u) > 0$ by hypothesis. Noting that $V(z) \geq V(z, \mu)$ by the definition of V completes the proof. \square

Let $\mathfrak{M}_0 := \{\mathcal{Y}(\{f_1, \dots, f_m\}) : \{f_1, \dots, f_m\} \text{ generates } x \text{ for some } x \in X\}$. It follows from Lemma 6.5 that for all finite x ,

$$\max_{\mu \in \mathfrak{M}_0} V(x, \mu) = \max_{\mu \in \mathfrak{M}} V(x, \mu)$$

In what follows, we shall restrict attention to finite menus and, therefore, it suffices to consider the set \mathfrak{M}_0 . Let $\mathcal{Y}_0 : X_0 \rightrightarrows \mathfrak{M}_0$ be defined as $\mathcal{Y}_0(x) = \mathcal{Y}(x) \cap \mathfrak{M}_0$ (where \mathcal{Y} is defined in Appendix C of DKS).

Lemma 6.9. Let $x_0 := \{f_1, \dots, f_m\}$ be the generator set for some nice menu x , and suppose $\mu \in \mathcal{Y}(x_0)$. Let J_i denote the states where f_i is active, and also let (E_i) represent an optimal partitional strategy (for μ) at x so that act f_i is chosen in the cell E_i . Then, μ is not Type II (a or b) at E_i in state s for all $i = 1, \dots, m$ and $s \in J_i^c$.

Proof. Let $\mu \in \mathcal{Y}(x_0)$ so that $V(x) = V(x_0) = V(x_0, \mu)$ and suppose μ is Type II (a or b) at E_i in state $s \in J_i^c$. Note also that because x is nice, there is a unique $\xi \in \mathcal{E}_x$ such that $x \sim \mathcal{F}(\xi)$, and the generator of x is unique.

Case 1: First consider the case where $f_i(s)$ is not a maximizer for $\bar{p}_i(s) \bar{u}_{i,s}(\cdot)$ on $\Delta(C \times X)$. Let f_i^* be the act such that (i) $f_i^*(s') = f_i(s')$ for all $s' \neq s$, and (ii) $f_i^*(s)$ maximizes $\bar{p}_i(s) \bar{u}_{i,s}(\cdot)$ on $\Delta(C \times X)$, so that $\bar{p}_i(s) \bar{u}_{i,s}(f_i^*(s)) > \bar{p}_i(s) \bar{u}_{i,s}(f_i(s))$. An act satisfying (ii) exists because μ is Type II at E_i in state s .

Now, consider the menu $x_{i,\varepsilon} := \{f_1, \dots, (1 - \varepsilon)f_i + \varepsilon f_i^*, \dots, f_m\}$. By Lemma 6.8, $V(x_{i,\varepsilon}) > V(x)$ for all $\varepsilon > 0$. Notice also that $x_{i,\varepsilon} \rightarrow x$ as $\varepsilon \rightarrow 0$.

For any $\varepsilon > 0$, consider $\mathcal{E}_{x_{i,\varepsilon}}$, and notice that the set-valued map $\varepsilon \mapsto \mathcal{E}_{x_{i,\varepsilon}}$ is a continuous, closed, and compact valued correspondence. By Axiom ICC (Axiom 4), there exists $\xi \in \tilde{\mathcal{E}}_{x_{i,\varepsilon}}$. Consider the maximization problem (parametrized by ε)

$$[\mathbf{P1}] \quad W(\varepsilon) := \max V(\mathcal{F}(\xi)) \quad \text{s.t. } \xi \in \mathcal{E}_{x_{i,\varepsilon}}$$

Notice that $W(0) = V(x)$ and that because $\mathcal{E}_{i,\varepsilon}$ is finite, a solution to [P1] always exists. We claim that for any $\varepsilon > 0$, the value of problem [P1] is precisely the value of $x_{i,\varepsilon}$, ie, $W(\varepsilon) = V(x_{i,\varepsilon})$.

To see this, notice that from the proof of Lemma 6.1, it follows that $V(x_{i,\varepsilon}) \geq V(\mathcal{F}(\xi))$ for all $\xi \in \mathcal{E}_{i,\varepsilon}$. By Axiom IICC (Axiom 4), there exists $\xi \in \tilde{\mathcal{E}}_{x_{i,\varepsilon}}$ such that $V(\mathcal{F}(\xi)) = V(x_{i,\varepsilon})$. Therefore, $W(\varepsilon) \geq V(x_{i,\varepsilon})$. Combining the two inequalities establishes that $W(\varepsilon) = V(x_{i,\varepsilon})$ for all $\varepsilon > 0$.

By the Theorem of the Maximum — see for instance, Ok (2007, p306) — W is continuous in ε . The Theorem of the Maximum also implies that the maximizer correspondence is upper hemicontinuous, and therefore for any ξ_ε^* that is optimal for the problem [P1], the limit $\xi_0^* := \lim_{\varepsilon \rightarrow 0} \xi_\varepsilon^*$ is also a maximizer. (The limit always exists because $\mathcal{E}_{x_{i,\varepsilon}}$ is a continuous, closed, and compact valued correspondence.) The continuity of W then implies that $W(0) = V(\mathcal{F}(\xi_0^*))$.

There are two possibilities now. The first is that for all $\varepsilon^\circ > 0$, there exists $\varepsilon \in (0, \varepsilon^\circ)$ such that $\xi_\varepsilon^*(s) = (1 - \varepsilon)f_i + \varepsilon f_i^*$ is active in state s . Because $\xi_0^* = \lim_{\varepsilon \rightarrow 0} \xi_\varepsilon^*$, it follows that $\xi_0^*(s) = f_i$, ie, f_i is active in state s . In other words, $\xi_0^* \neq \xi$. But we have already established that $W(0) = V(x) = V(\mathcal{F}(\xi_0^*))$, which contradicts the assumption that x is nice, which rules out this first possibility.

The other possibility is that there exists an $\varepsilon_\circ > 0$ such that for all $\varepsilon < \varepsilon_\circ$, the act $(1 - \varepsilon)f_i + \varepsilon f_i^*$ is *inactive* in every such state $s \in J_i^c$, ie, $\xi_\varepsilon^*(s) \neq (1 - \varepsilon)f_i + \varepsilon f_i^*$. In this case, for all $\varepsilon < \varepsilon_\circ$, we have $\xi_0^* = \xi_\varepsilon^*$. Because x is nice, it must necessarily be that $\xi_0^* = \xi$. This implies that for all such ε , $V(x_{i,\varepsilon}) = W(\varepsilon) = W(0) = V(x)$. But this contradicts our earlier observation (which follows from Lemma 6.8) that $V(x_{i,\varepsilon}) > V(x)$ if μ is Type II at E_i in state s whenever f_i is active in state $s \in J_i$. This contradiction rules out the second possibility, and completes the proof of the first case.

Case 2: Suppose that $f_i(s)$ is a maximizer for $\bar{p}_i(s)\bar{u}_{i,s}(\cdot)$ on $\Delta(C \times X)$. If μ is of Type IIa on E_i in state $s \in J_i^c$, then $\bar{p}_i(s)\bar{u}_{i,s}(\cdot)$ is not constant on $\Delta(C \times L)$. If μ is of Type IIb on E_i in state $s \in J_i^c$, then $\bar{p}_i(s)\bar{u}_{i,s}(\cdot)$ is constant on $\Delta(C \times L)$. However, in either case, there exists $\ell \in L$ such that $\bar{p}_i(s)\bar{u}_{i,s}(f_i(s)) > \bar{p}_i(s)\bar{u}_{i,s}(\ell(s))$. (Such an ℓ exists because $f_i(s)$ is a maximizer of $\bar{p}_i(s)\bar{u}_{i,s}(\cdot)$ and by hypothesis that μ is of Type II, there exists some $\beta \in \Delta(C \times L)$ that is *not* a maximizer.)

Consider the menu $\frac{1}{2}x + \frac{1}{2}\ell$. By Lemma 6.6, we see that $\mu \in \mathcal{Y}(x)$ implies $\mu \in \mathcal{Y}(\frac{1}{2}x + \frac{1}{2}\ell)$. Because x is nice, x_0 , which satisfies $V(x_0) = V(x)$, is the unique generator set of x . L-Independence now implies that $V(\frac{1}{2}x_0 + \frac{1}{2}\ell) = V(\frac{1}{2}x + \frac{1}{2}\ell)$. Moreover, Lemma 6.6 says that $\frac{1}{2}x + \frac{1}{2}\ell$ is nice. It follows immediately that $\frac{1}{2}x_0 + \frac{1}{2}\ell$ is a generator set for $\frac{1}{2}x + \frac{1}{2}\ell$.

Now consider the nice menu $\frac{1}{2}x + \frac{1}{2}\ell$ with generator $\frac{1}{2}x_0 + \frac{1}{2}\ell$, and let $\mu \in \mathcal{Y}(\frac{1}{2}x_0 + \frac{1}{2}\ell)$. By construction, $\frac{1}{2}f_i(s) + \frac{1}{2}\ell(s)$ is not a maximizer of $\bar{p}_i(s)\bar{u}_{i,s}$ on $\Delta(C \times X)$ (although $f_i(s)$ is), which means that we now satisfy the hypotheses of Case 1. Lemma 6.6 ensures that $\mathcal{Y}(\frac{1}{2}x + \frac{1}{2}\ell) \cap \mathcal{Y}(x) \neq \emptyset$ and that a partitional optimal strategy at x is also optimal at

$\frac{1}{2}x + \frac{1}{2}\ell$. These facts allow us to establish that even in this case, μ cannot be of Type II, which completes the proof. \square

Let x be nice and let $\mu \in \mathcal{Y}_0(x)$. Let $(E_i^{\mu,x})$ be the partition induced by an optimal strategy (for instance, one coming from the generators of x) given μ and consider the mapping

$$(\mu, E_i^{\mu,x}, s) \mapsto \bar{p}_i(s)\bar{u}_{i,s}(\cdot) = \int_{E_i^{\mu,x}} p(s)u_s(\cdot) d\mu(p, u)$$

Let $\{f_1, \dots, f_k\}$ be the unique generator set of x , and let J_i denote the set of states where f_i is active so (J_i) is a partition of S . Now define

$$\begin{aligned} \gamma_{\mu,x}^i &:= \sum_{s \in J_i} \bar{p}_i(s) \\ \clubsuit \quad p_i(s) &:= \begin{cases} \bar{p}_i(s)/\gamma_{\mu,x}^i & \text{if } s \in J_i \\ 0 & \text{otherwise} \end{cases} \\ \hat{u}_s &:= \gamma_{\mu,x}^i \bar{u}_{i,s} \quad \text{where } i \text{ is such that } s \in J_i \end{aligned}$$

and let

$$\hat{\mathfrak{M}} := \{\hat{\mu} \in \Delta(\mathfrak{U}) : \text{supp}(\hat{\mu}) = \{(p_i, \hat{u}) : i = 1, \dots, k \text{ where } k \leq n = |S|\}\}$$

Note that $\gamma_{\mu,x}^i \neq 0$ so that p_i is well defined. To see this, suppose that $\gamma_{\mu,x}^i = 0$. Then, $\bar{p}_i(s) = 0$ for all $s \in J_i$. This implies that $\bar{p}_i(s)\bar{u}_{i,s}(f) = 0$ for all acts f , which implies that $\{f_1, \dots, f_k\} \sim \{f_1, \dots, f_k\} \setminus \{f_i\}$. That is, we can drop the act f_i from the set $\{f_1, \dots, f_k\}$ without any loss in utility, contradicting the assumption that $\{f_1, \dots, f_k\}$ is the unique generator set of x .

Consider the mapping

$$\mathfrak{D}(\mu, x, (E_i^{\mu,x})) \mapsto \hat{\mu} \in \hat{\mathfrak{M}}$$

where $\text{supp } \hat{\mu} = \{(p_i, \hat{u}) : i = 1, \dots, k\}$, p_i for $i = 1, \dots, k$ and \hat{u} are defined in \clubsuit , and $\hat{\mu}$ itself is defined as

$$\hat{\mu}((p_i, \hat{u})) = \mu(E_i^{\mu,x})$$

Let $\hat{\mathfrak{M}}_p \subset \hat{\mathfrak{M}}$ be the image of \mathfrak{D} . (The domain of \mathfrak{D} is easily defined, but notationally cumbersome, and because omitting it will not cause any confusion in the sequel, we refrain from a formal definition.)

A collection of probability measures $\{p_1, \dots, p_k\}$ on S (so each $p_i \in \Delta(S)$) forms a *partitional system* if (i) for all $s \in S$, $p_i(s) > 0$ implies $p_j(s) = 0$ for all $j \neq i$, and (ii) for all s , $\sum_{i=1}^k p_i(s) > 0$. In other words, every state s is supported by exactly one p_i in the collection.

A positive measure μ on \mathfrak{U} is *elementary* if its support is Dirac (degenerate) on $\mathfrak{U}_{s,\ell^+(s)}$ (see Appendix C of DKS for a definition) and the support on $\Delta(S)$ is a partitional system of

probability measures on S . In other words, there exist $p_1, \dots, p_k \in \Delta(S)$ and $u_s \in \mathfrak{U}_{s, \ell^\dagger(s)}$ for all s such that μ is supported on the finite collection $(p_1, u), \dots, (p_k, u)$ where $u = (u_s)_{s \in S}$. Rather than saying that the marginal of μ on $\Delta(S)$ has support $\{p_1, \dots, p_k\}$, we will often say in the sequel that μ supports the partitional system (p_i) .

With these definitions, it is clear that each $\hat{\mu} \in \hat{\mathfrak{M}}_p$ is elementary. The following proposition says that it is without loss of generality to restrict attention to elementary measures. Towards this end, let us define $\hat{V} : X_0 \rightarrow \mathbb{R}$ as

$$\hat{V}(x) := \sup_{\mu \in \hat{\mathfrak{M}}_p} \left[\sum_i \left[\max_{f \in x} \sum_s p_i(s) u_s(f(s)) \right] \mu(p_i, u) \right]$$

Proposition 6.10. For all nice x , $\hat{V}(x) = V(x)$. Moreover, the supremum in the definition of \hat{V} is attained.

Proof. Let x be nice, $\mu \in \mathcal{Y}_0(x)$, and $\{f_1, \dots, f_k\}$ the unique generator set of x . Let us first prove that $V(x) \leq \hat{V}(x)$. Let $(E_i^{\mu, x})$ be an optimal partition for μ at x , and let $\hat{\mu} = \mathfrak{D}(\mu, x, (E_i^{\mu, x}))$. Then,

$$\begin{aligned} V(x, \mu) &= \sum_i \max_{f \in x} \left[\sum_s \int_{E_i^{\mu, x}} p(s) u_s(f(s)) d\mu(p, u) \right] \\ &= \sum_i \max_{f \in x} \sum_s \bar{p}_i(s) \bar{u}_{i,s}(f(s)) \end{aligned}$$

Lemma 6.9 says that μ cannot be of Type II (a or b) if $s \in J_i^c$, and hence must be either Type Ia or Type Ib. In either case, $\bar{p}_i(s) \bar{u}_{i,s}(f(s)) \leq 0 = \bar{p}_i(s) \bar{u}_{i,s}(\ell^\dagger(s))$ for all $s \in J_i^c$. Therefore, it must be that

$$V(x) = V(x, \mu) \leq \sum_i \max_{f \in x} \sum_s p_i(s) \hat{u}_s(f(s)) = \hat{V}(x, \hat{\mu}) \leq \hat{V}(x)$$

We now prove that $\hat{V}(x) \leq V(x)$ for all nice x . Suppose, by way of contradiction, that $\hat{V}(x, \hat{\mu}) > V(x)$ for some nice x and $\hat{\mu} \in \hat{\mathfrak{M}}_p$. Suppose the optimal strategy here is to choose $f_i \in x$ whenever the ‘interim information’ is (p_i, u) .

Now recall that $\hat{\mu} = \mathfrak{D}(\mu, y, (E_i^{\mu, y}))$ for some $\mu \in \mathcal{Y}_0$ and $y \in X_0$. Consider the strategy ζ^μ that is constant on $E_i^{\mu, y}$, ie, satisfies $\zeta^\mu(E_i^{\mu, y}) = f_i \in x$ for each i (where f_i is the optimal choice when presented with the interim information (p_i, u)). The value of this strategy, $V(x, \mu, \zeta^\mu)$, is given by

$$\begin{aligned} V(x, \mu, \zeta^\mu) &= \sum_i \left[\sum_s \int_{E_i^{\mu, y}} p(s) u_s(f_i(s)) d\mu(p, u) \right] \\ &= \sum_i \left[\sum_s \bar{p}_i(s) \bar{u}_{i,s}(f_i(s)) \right] \end{aligned}$$

It follows from Lemma 6.9 that μ is not Type II (a or b) at $E_i^{\mu,y}$ in state s for all $s \in J_i^c$. (Note that the partition (J_i) is generated by the unique $\xi \in \tilde{\mathcal{E}}_y$. Thus, (J_i) does not depend on x .) Therefore, for all such $s \in J_i^c$, it must be that $\bar{p}_i(s)\bar{u}_{i,s}(f_i(s)) \leq 0$. For such an $s \in J_i^c$, if we replace $f_i(s)$ by ℓ_* , we obtain the new menu x' , which has the property that $V(x', \mu, \zeta_{x'}^\mu) = \hat{V}(x, \hat{\mu})$. But this implies $V(x') \geq \hat{V}(x, \hat{\mu}) > V(x)$, where the strict inequality follows from our hypothesis. This violates Axiom IICC (Axiom 4) and Continuity because x' is obtained from x by replacing payoffs in acts in x by ℓ_* , so that $x \succsim x'$. This proves that $\hat{V}(x) = V(x)$ for all nice x .

Now, to show that the maximum is achieved in the definition of $\hat{V}(x)$, observe that for each nice x , there exists $\mu \in \mathcal{Y}_0(x)$, so that

$$\begin{aligned} V(x) &= V(x, \mu) && \text{definition of } \mu \\ &\leq \hat{V}(x, \hat{\mu}) && \text{from proof of } V(x) \leq \hat{V}(x) \text{ above} \\ &\leq \hat{V}(x) && \text{definition of } \hat{V} \\ &\leq V(x) && \text{because } \hat{V}(x) \leq V(x) \text{ as proved above} \end{aligned}$$

where $\hat{\mu} = \mathfrak{D}(\mu, x, (E_i^{\mu,x}))$, $\mu \in \mathcal{Y}_0(x)$, and $(E_i^{\mu,x})$ is an optimal partition strategy for μ at x . Therefore, $\hat{\mu}$ is \hat{V} -optimal for x , as claimed. \square

Because V is Lipschitz, it follows immediately that \hat{V} is also Lipschitz on X_0 . By Proposition 6.3, X_0 is dense in X , so that \hat{V} uniquely extends to X . It is easy to see that in the representation of \hat{V} , this amounts to replacing $\hat{\mathfrak{M}}_p$ with its closure. In what follows, we shall therefore assume that $\hat{\mathfrak{M}}_p$ is closed and that \hat{V} is defined on X .

Thus far, we have shown that \succsim is represented by a function $V : X \rightarrow \mathbb{R}$ that has the form

$$[6.1] \quad V(x) = \max_{\mu \in \mathfrak{M}} V(x, \mu)$$

where

- each $\mu \in \mathfrak{M}$ is a positive elementary measure,
- $V(x, \mu) = \left[\sum_{p \in \Delta(S)} \left(\max_{f \in x} \sum_{s \in S} p(s) u_s(f(s)) \right) \mu(p; u) \right]$, and
- $V(\ell; \mu) = V(\ell; \mu')$ for all $\mu, \mu' \in \mathfrak{M}$ and $\ell \in L$.

Our first result establishes that we can replace an elementary measure by an elementary probability measure.

Lemma 6.11. Let μ be an elementary measure. Then, there exists an elementary probability measure $\hat{\mu}$ such that for all $x \in X$, $V(x, \mu) = V(x, \hat{\mu})$.

Proof. Let μ be supported on $(p_1, u), \dots, (p_k, u)$, and let $\|\mu\|_1$ be the total weight of μ . (That is, $\|\mu\|_1 := \sum_i \mu((p_i, u))$.) For any $s \in S$, define $\hat{u}_s := \|\mu\|_1 u_s$, and for any $p \in \Delta(S)$, let

$\hat{\mu}(p, \hat{u}) := \mu(p, u) / \|\mu\|_1$ where $\hat{u} = (\hat{u}_s)_{s \in S}$. It is easy to see that $\hat{\mu}$ so defined is elementary and is also a probability measure.

Moreover, we have

$$\begin{aligned} V(x, \hat{\mu}) &= \sum_p \max_{f \in x} \sum_s \hat{\mu}(p, \hat{u}) p(s) \hat{u}_s(f(s)) \\ &= \sum_p \max_{f \in x} \sum_s \frac{\mu(p, u)}{\|\mu\|_1} p(s) \|\mu\|_1 u_s(f(s)) \\ &= V(x, \mu) \end{aligned}$$

which establishes the claim. \square

Two partitional systems of probability measures $\{p_1, \dots, p_k\}$ and $\{q_1, \dots, q_k\}$ are *similar* if for all $i = 1, \dots, k$, $\text{supp}(p_i) = \text{supp}(q_i)$.

Every elementary probability measure μ on $\Delta(S)$ supports a partitional system. We now show that we can replace, ie, without affecting utility considerations, μ by another elementary probability measure $\hat{\mu}$ that supports another partitional system that is similar to the partitional system supported by μ .

Lemma 6.12. Let μ be an elementary probability measure whose support is $(p_1, u), \dots, (p_k, u)$. Let $\{\tilde{p}_1, \dots, \tilde{p}_k\}$ be a partitional system on $\Delta(S)$ that is similar to $\{p_1, \dots, p_k\}$. Then, there exists an elementary probability measure $\tilde{\mu}$ with support $(\tilde{p}_1, \tilde{u}), \dots, (\tilde{p}_k, \tilde{u})$ such that for all $x \in X$ we have $V(x, \mu) = V(x, \tilde{\mu})$.

Proof. Define $\tilde{u}_s := (p_i(s) / \tilde{p}_i(s)) u_s$, and set $\mu(p_i, u) = \tilde{\mu}(\tilde{p}_i, \tilde{u})$, where $\tilde{u} = (\tilde{u}_s)_{s \in S}$. Then, we have

$$\begin{aligned} V(x, \tilde{\mu}) &= \sum_i \max_{f \in x} \sum_s \tilde{\mu}(\tilde{p}_i, \tilde{u}) \tilde{p}_i(s) \tilde{u}_s(f(s)) \\ &= \sum_i \max_{f \in x} \sum_s \mu(p_i, u) p_i(s) u_s(f(s)) \\ &= V(x, \mu) \end{aligned}$$

which completes the proof. \square

Let μ be an elementary probability measure and define $\pi_\mu \in \Delta(S)$ as

$$\pi_\mu(s) := \sum_p \mu(p) p(s)$$

Let $\pi_0 \in \Delta(S)$ and $P := (J_i)$ be a partition of S . Then, the conditional probability induced by J_i is $q_i(\cdot, \pi_0 | J_i)$ where

$$q_i(s; \pi_0 | J_i) := \pi_0(s | J_i)$$

for all $J_i \in P$. It is easy to see that $(q_i(\cdot, \pi_0 | J_i))$ is a partitional system of probabilities on S . Conversely, let μ be an elementary measure that supports the partitional system (p_i) . This induces the partition $P_\mu := (J_i)$ of S where $J_i := \text{supp}(p_i)$.

Lemma 6.13. Let $\pi_0 \in \Delta(S)$, μ an elementary probability measure that supports the partitional system (p_i) , and let (J_i) be the partition of S induced by (p_i) . Then, there exists an elementary probability measure $\hat{\mu}$ such that

- (a) μ^* supports the partitional system $(q_i(\cdot, \pi_0 \mid J_i))$,
- (b) $\pi_{\mu^*} = \pi_0$, and
- (c) $V(x, \mu) = V(x, \mu^*)$ for all $x \in X$.

Proof. Let μ and π_0 be as hypothesized and consider the induced partitional system $(q_i(\cdot; \pi_0 \mid J_i))$. By Lemma 6.12, there exists an elementary probability measure $\tilde{\mu}$ that supports $(q_i(\cdot; \pi_0 \mid J_i))$ while keeping utilities unaltered.

For each s , define the utility function

$$u_s^* := \left[\frac{\sum_i \tilde{\mu}(q_i(s; \pi_0 \mid J_i), \tilde{u}) \mathbb{1}_{\{s \in J_i\}}}{\sum_i \pi_0(J_i) \mathbb{1}_{\{s \in J_i\}}} \right] \tilde{u}_s$$

and observe that in the sums in both the numerator and denominator, only one term is non-zero. Now, define the elementary probability measure μ^* as follows: If s is supported by $q_i(\cdot; \pi_0 \mid J_i)$, set

$$\mu^*(q_i(\cdot; \pi_0 \mid J_i), u^*) := \pi_0(J_i)$$

and 0 otherwise, which proves (a). With this definition, $\pi_{\mu^*}(s) = \sum_i \mu^*(q_i(\cdot; \pi_0 \mid J_i), u^*) \cdot q_i(s; \pi_0 \mid J_i) = \pi_0(s)$, as desired for the proof of (b). To see (c), notice that we have

$$\begin{aligned} V(x, \mu^*) &= \sum_i \max_{f \in x} \sum_s \mu^*(q_i(\cdot; \pi_0 \mid J_i), u^*) q_i(s; \pi_0 \mid J_i) u_s^*(f(s)) \\ &= \sum_i \max_{f \in x} \sum_s \pi_0(J_i) q_i(s; \pi_0 \mid J_i) \frac{\tilde{\mu}(q_i(\cdot; \pi_0 \mid J_i), \tilde{u})}{\pi_0(J_i)} \tilde{u}_s(f(s)) \\ &= \sum_i \max_{f \in x} \sum_s q_i(s; \pi_0 \mid J_i) \tilde{\mu}(q_i(\cdot; \pi_0 \mid J_i), \tilde{u}) \tilde{u}_s(f(s)) \\ &= V(x, \tilde{\mu}) = V(x, \mu) \end{aligned}$$

which completes the proof. □

We are now in a position to prove Proposition C.1 of DKS.

Proof of Proposition C.1. We shall first prove (a) implies (b). We have shown that given the representation $[\diamond]$ in Theorem 1 and ICC (Axiom 4), V has the form in [6.1], where every $\mu \in \mathfrak{M}$ is an elementary (positive, but finite) measure. Lemma 6.11 shows that it is without loss of generality to consider μ that are elementary *probability* measures. Consider such a μ and suppose it supports the partitional system (p_i) . Let $J_i = \text{supp}(p_i)$, and notice that (J_i) is a partition of S . Lemma 6.13 says that it is without loss of generality to assume that every μ supports the partitional system $(q_i(\cdot; \pi_0 \mid J_i))$ (recall that $q_i(s; \pi_0 \mid J_i) = \pi_0(s \mid J_i)$) and

also has the feature that $\pi_\mu(s) := \sum_i \mu(q_i(s; \pi_0 | J_i)) q_i(s; \pi_0 | J_i) = \pi_0(s)$ for all s . (To ease notational burden, in what follows we shall write $q_i(s; \pi_0 | J_i)$ as $q_i(s)$.)

In particular, this last property implies that $\mu(q_i, u) = \pi_0(J_i)$ and $\mu(q_i, u) q_i(s) = \pi_0(J_i) \pi_0(s | J_i)$. This implies

$$\begin{aligned}
V(x, \mu) &:= \sum_i \left[\max_{f \in x} \sum_s q_i(s) u_s(f(s)) \right] \mu(q_i, u) \\
&= \sum_{J_i \in P} \left[\max_{f \in x} \sum_s \pi_0(s | J_i) u_s(f(s)) \right] \pi_0(J_i) \\
&= \sum_{J_i \in P} \left[\max_{f \in x} \sum_{s \in J_i} \pi_0(s | J_i) u_s(f(s)) \right] \pi_0(J_i) \\
&=: V'(x, \pi_0, (P, u))
\end{aligned}$$

In other words, the informational content of the elementary probability measure μ is now encoded into the prior π_0 , the partition $P = (J_i)$, and the utility functions $u = (u_s)$. Let \mathfrak{M}' be the collection of all such pairs (P, u) induced by elementary probability measures in \mathfrak{M} . Then, we can write

$$\begin{aligned}
V(x) &= \max_{\mu} V(x, \mu) \\
&= \max_{(P, u) \in \mathfrak{M}'} V'(x, \pi_0, (P, u)) \\
&=: V'(x)
\end{aligned}$$

where $V'(x) = V(x)$ for all $x \in X$; this proves the representation part.

Observe now — see [6.1] — that for all $\ell \in L$ and $\mu, \mu' \in \mathfrak{M}$, we have $V(\ell, \mu) = V(\ell, \mu')$. This implies that, for all $\ell \in L$ and $(P, u), (P', u') \in \mathfrak{M}'$, we have $V'(\ell, \pi_0, (P, u)) = V'(\ell, \pi_0, (P', u'))$.

Recall that $\ell^\dagger \in L$ is such that $u_s(\ell^\dagger(s)) = 0$ for all $s \in S$. For any $\alpha \in \Delta(C \times L)$, define $\hat{\ell}_\alpha^s \in L$ as

$$\hat{\ell}_\alpha^s(s') = \begin{cases} \alpha & \text{if } s' = s \\ \ell^\dagger(s') & \text{otherwise} \end{cases}$$

For all $(P, u), (P', u') \in \mathfrak{M}'$, we then have $V(\hat{\ell}_\alpha^s, \pi_0, (P, u)) = V(\hat{\ell}_\alpha^s, \pi_0, (P', u'))$. Notice that $V(\hat{\ell}_\alpha^s, \pi_0, (P, u)) = \pi_0(s) u_s(\alpha) = \pi_0(s) u'_s(\alpha) = V(\hat{\ell}_\alpha^s, \pi_0, (P', u'))$. Since this is true for all $\alpha \in \Delta(C \times L)$, it follows that u_s and u'_s are identical on $C \times L$ for all $(P, u), (P', u') \in \mathfrak{M}'$. This proves that (a) implies (b).

That (b) implies (a) follows immediately from Lemma 6.13 which shows how to construct an elementary measure μ given the prior π_0 , the partition $P_\mu = (J_i)$, and the utility function $u = (u_s)$. \square

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