

SYMMETRIC, COHERENT, CHOQUET CAPACITIES

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Choquet capacities are a generalization of probability measures that arise in robustness, decision theory and game theory. Many capacities that arise in robustness are symmetric or can be transformed into symmetric capacities. We characterize the extreme points of the set of upper distribution functions corresponding to coherent, symmetric Choquet capacities on $[0, 1]$. We also show that the set of 2-alternating capacities is a simplex and we give a Choquet representation of this set.

1. Introduction. A *Choquet capacity* on a measurable space (Ω, \mathcal{B}) is a mapping $C: \mathcal{B} \rightarrow [0, 1]$ such that $C(\emptyset) = 0$. C is *coherent* if there exists a nonempty set of probability measures M such that $C(A) = \sup_{P \in M} P(A)$ for every $A \in \mathcal{B}$. Coherent capacities are also called *upper probabilities* [Walley (1991), Fine (1988), Dempster (1967, 1968), and Smith (1961)] or *upper envelopes* [Anger and Lembcke (1985) and Denneberg (1994)]. Let $\Omega = [0, 1]$, let \mathcal{B} be the Borel subsets of Ω and let μ be Lebesgue measure. C is *symmetric* if $C(A) = C(B)$ whenever $\mu(A) = \mu(B)$. As we shall show, it is possible to say exactly when a symmetric capacity is coherent.

Many robustness models used in statistics involve symmetric, coherent capacities or can be transformed into the same by a smooth, one-to-one mapping [Buja (1986), Huber and Strassen (1973), Wasserman and Kadane (1990) and Fortini and Ruggeri (1994)]. For example, the upper probability for an ε -contamination neighborhood around a probability measure P [Berger (1984, 1990) and Huber (1973, 1981)] generates a symmetric capacity once the set of probabilities is transformed to the unit interval under the inverse integral transform corresponding to P . This is true for many neighborhoods. Capacities are also used in decision theory [Gilboa (1987) and Schmeidler (1989)] and game theory [Shapley (1971)]. Symmetric Choquet integrals, which are related to symmetric capacities, have been studied by Armstrong (1990) and Talagrand (1978). Symmetric capacities were studied in Wasserman and Kadane (1992) under the additional assumption that M consisted of nonatomic probabilities with bounded densities. Despite the ubiquity of capacities, there is little in the way of simple characterizations for capacities as

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there are for probabilities. This paper is concerned with such characterizations. In particular, we are interested in the following question: what are the extreme points in the set of all distribution functions corresponding to symmetric capacities?

Our interest in the extreme points is twofold. First, in Bayesian robustness, where heavy use is made of sets of priors, the extreme points play a crucial role. For example, one is typically interested in bounding posterior expectations. These bounds occur at the extreme points. Thus, much attention in Bayesian robustness has focused on extreme points. Second, it is very difficult to form an intuitive picture of the set M since it is typically infinite dimensional. It is our hope that our characterization of the extreme points, which has a simple geometric interpretation (Corollary 3.1), will cast light on the structure of these sets.

A capacity is *2-alternating* if

$$(1) \quad C(A \cup B) \leq C(A) + C(B) - C(A \cap B)$$

for all $A, B \in \mathcal{B}$. Many capacities used in statistics are 2-alternating. Furthermore, the 2-alternating condition is crucial for many important results. For example, a particular generalization of the Neyman–Pearson lemma holds if and only if the capacity generated by the underlying models is 2-alternating [Huber and Strassen (1973)]. Similarly, a particular generalization of Bayes' theorem for capacities holds if and only if the capacity is 2-alternating [Wasserman and Kadane (1990)]. In game theory, 2-alternating capacities represent certain convex games [Shapley (1971)]. Most work on coherent capacities has focused on the 2-alternating case. Little is known about the non-2-alternating case. Some work on non-2-alternating and noncoherent capacities is contained in Papamarcou and Fine (1986) and Sadrolhefazi and Fine (1994). We shall consider the general case in Sections 2 and 3 and the 2-alternating case in Section 4.

The following is an outline of the paper and serves as a summary of the main contributions of this paper. In Section 2 we give a majorization representation of symmetric capacities (Theorem 2.1) which generalizes a theorem in Wasserman and Kadane (1992). In Section 3, which is the main section of the paper, we study the distribution functions of symmetric capacities. There we establish (Lemmas 3.1 and 3.2) a correspondence between distribution functions of symmetric capacities and functions α taking $[0, 1]$ to $[0, 1]$ that satisfy

$$\lim_{\omega \rightarrow 1} \int_{1/2}^{\omega} \frac{\alpha(u) - u}{u(1-u)} du = -\infty.$$

This correspondence allows us to characterize the extreme points in the set of all distribution functions for symmetric capacities (Theorem 3.2). This is the main theorem of the paper and, loosely, it says that F is extreme if and only if the corresponding α function takes values 0 and 1 almost everywhere. In Section 4 we identify the extreme points of the set of distribution functions for 2-alternating capacities (Theorem 4.1) and we give a Choquet representa-

tion for this set by identifying the unique mixing measure over the extreme points (Theorem 4.2). Closing remarks are contained in Section 5.

2. A characterization of symmetric, coherent capacities. Let \mathcal{P} be the set of all probability measures on \mathcal{B} , let \mathcal{P}_a be the set of all $P \in \mathcal{P}$ that are absolutely continuous with respect to μ and let \mathcal{P}_s be the set of all probability measures that are singular with respect to μ . Suppose that $P, Q \in \mathcal{P}_a$ and let $p = dP/d\mu$ and $q = dQ/d\mu$. We write $p \sim q$ if

$$(2) \quad \mu(\{\omega; p(\omega) > t\}) = \mu(\{\omega; q(\omega) > t\})$$

for all real t . If (2) holds we say that p and q are *equimeasurable*. We shall also say that P and Q are equimeasurable and we will write $P \sim Q$. Given any $P \in \mathcal{P}_a$ with $p = dP/d\mu$, there exists a unique, nonincreasing, right-continuous function p^* , called the *decreasing rearrangement* of p , such that $p \sim p^*$ [Ryff (1965)]. We call the corresponding probability measure P^* the decreasing rearrangement of P .

Every $P \in \mathcal{P}$ may be written in terms of its Lebesgue decomposition $P = \alpha P_a + \bar{\alpha} P_s$, where $P_a \in \mathcal{P}_a$, $P_s \in \mathcal{P}_s$, $\alpha \in [0, 1]$ and $\bar{\alpha} = 1 - \alpha$. We define the decreasing rearrangement of P by $P^* = \alpha P_a^* + \bar{\alpha} \delta_0$, where δ_0 is a point mass at 0. Note that, if $P \in \mathcal{P}_a$, then this agrees with the earlier definition of decreasing rearrangement. We say that P is *majorized* by Q , denoted by $P \prec Q$, if $P^*([0, t]) \leq Q^*([0, t])$ for every real t . Majorization has been studied in discrete settings [Marshall and Olkin (1979)] and continuous settings [Ryff (1963, 1965, 1967, 1970)]. Our definition is slightly different from previous definitions to allow for probabilities with both absolutely continuous and singular components.

Let $M \subset \mathcal{P}$ be nonempty and let $C(A) = \sup_{P \in M} P(A)$. Throughout the rest of the paper we restrict attention to symmetric, coherent capacities. We say that P is *dominated* by C , written $P \triangleleft C$, if $P(A) \leq C(A)$ for all $A \in \mathcal{B}$.

THEOREM 2.1. *The following two statements are equivalent:*

- (i) C is symmetric.
- (ii) $P \triangleleft C$ and $Q \prec P$ imply that $Q \triangleleft C$.

Before proving the theorem, we establish some lemmas.

LEMMA 2.1. *For every $P \in \mathcal{P}_a$ and every $t \in [0, 1]$, there exists A_t such that (i) $\mu(A_t) = t$, (ii) $P^*([0, t]) = P(A_t)$ and (iii) $P(A_t) \geq P(B)$ for every B for which $\mu(B) = t$.*

PROOF. The lemma is obvious for $t = 0$ and $t = 1$ so assume $0 < t < 1$. Let $p = dP/d\mu$ and $p^* = dP^*/d\mu$. Define $A_0 = \{\omega; p(\omega) > p^*(t)\}$ and $A_{00} = \{\omega; p(\omega) \geq p^*(t)\}$. Then $\mu(A_0) \leq t \leq \mu(A_{00})$ and $A_0 \subset A_{00}$. Let $u = \mu(A_0)$. Choose B such that $A_0 \subset B \subset A_{00}$ and $\mu(B) = t - u$. Let $A_t = A_0 \cup B$.

Hence, $\mu(A_t) = t$ and $P(A_t) = P(A_0) + P(B) = P^*([0, u]) + p^*(u)(t - u) = P^*([0, t])$. Now let K be such that $\mu(K) = t$. Then

$$\begin{aligned} P(A_t) - P(K) &= P(A_t - K) - P(K - A_t) \\ &\geq \mu(A_t - K) \operatorname{ess\,inf}_{A_t - K} p(\omega) - \mu(K - A_t) \operatorname{ess\,sup}_{K - A_t} p(\omega) \\ &= \mu(A_t - K) \left(\operatorname{ess\,inf}_{A_t - K} p(\omega) - \operatorname{ess\,sup}_{K - A_t} p(\omega) \right) \geq 0. \quad \square \end{aligned}$$

LEMMA 2.2. For every $P \in \mathcal{P}$ and every $A \in \mathcal{B}$, $P(A) \leq P^*([0, \mu(A)])$.

PROOF. Let $t = \mu(A)$. Then $P^*([0, \mu(A)]) = \alpha P_a^*([0, \mu(A)]) + \bar{\alpha} = \alpha P_a(A_t) + \bar{\alpha} \geq \alpha P_a(A) + \bar{\alpha} \geq \alpha P_a(A) + \bar{\alpha} P_s(A) = P(A)$, where A_t is as defined in Lemma 2.1. The second equality and the inequality that follows it are both due to Lemma 2.1. \square

LEMMA 2.3. If C is symmetric and $P \triangleleft C$, then $P^*([0, t]) \leq C([0, t])$ for every t .

PROOF. Let $P = \alpha P + \bar{\alpha} P_s$ and let S be the support of P_s . $P^*([0, t]) = \alpha P_a^*([0, t]) + \bar{\alpha} = \alpha P_a(A_t) + \bar{\alpha} = \alpha P_a(A_t \cup S) + \bar{\alpha} P_s(A_t \cup S) = P(A_t \cup S) \leq C(A_t \cup S) = C([0, t])$. The second equality follows from Lemma 2.1. The last equality follows since C is symmetric and $\mu(A_t \cup S) = \mu([0, t])$. \square

PROOF OF THEOREM 2.1. (i) implies (ii). Let $P \triangleleft C$ and $Q \prec P$. For any A , $Q(A) \leq Q^*([0, \mu(A)]) \leq P^*([0, \mu(A)]) \leq C([0, \mu(A)]) = C(A)$. This follows from, respectively, Lemma 2.2, $Q \prec P$, Lemma 2.3 and (i). Thus, $Q \triangleleft C$.

(ii) implies (i). Let A and B be such that $\mu(A) = \mu(B)$. There exists $P \triangleleft C$ such that $P(A) = C(A)$. Write $P = \alpha P_a + \bar{\alpha} P_s$. Define

$$R_a(\cdot) = \frac{P_a(A)}{\mu(A)} \mu(\cdot \cap B) + \frac{P_a(A^c)}{\mu(A^c)} \mu(\cdot \cap B^c)$$

and define $R = \alpha R_a + \bar{\alpha} R_s$, where R_s is any singular measure such that $R_s(B) = P_s(A)$. By construction, $R(B) = P(A)$. Furthermore, $R \prec P$ so that, by (ii), $R \triangleleft C$. Thus, $C(A) = P(A) = R(B) \leq C(B)$. By a similar argument, $C(B) \leq C(A)$. \square

Now we consider some examples of symmetric capacities.

EXAMPLE 2.1 (ε -contamination). Let $M = \{(1 - \varepsilon)\mu + \varepsilon Q; Q \in \mathcal{P}\}$ where $\varepsilon \in [0, 1]$. Then $C(A) = (1 - \varepsilon)\mu(A) + \varepsilon$ if $A \neq \emptyset$. This model is used extensively in robustness [Huber (1973, 1981) and Berger (1984)].

EXAMPLE 2.2 (Total variation). Let $d(P, Q) = \sup_{A \in \mathcal{B}} |P(A) - Q(A)|$ be the total variation distance between P and Q . Let $M = \{P; d(\mu, P) \leq \varepsilon\}$. Then $C(A) = \min\{P(A) + \varepsilon, 1\}$. This is also common in robustness.

EXAMPLE 2.3 (Density bounded class). Let $a < 1 < b$ and let M be the set of all P with densities p (with respect to μ) such that $a \leq p(\omega) \leq b$ for μ -almost all ω . Then $C(A) = \min\{b\mu(A), 1 - a\mu(A^c)\}$. Lavine (1991a, b) uses this class in Bayesian robustness.

EXAMPLE 2.4 (Density ratio class). Let $k \geq 1$ and let M be the set of all P with densities p (with respect to μ) such that $\text{ess sup } p(\omega)/\text{ess inf } p(\omega) \leq k$. It turns out that $C(A) = k\mu(A)[k\mu(A) + \mu(A^c)]^{-1}$. This is a specialization of a class used by DeRobertis and Hartigan (1981) in Bayesian robustness.

3. Distribution functions for capacities. The distribution function for a symmetric capacity C is defined by $F(\omega) = C([0, \omega])$. Conversely, a function $F: [0, 1] \rightarrow [0, 1]$ defines a symmetric capacity by way of $C(A) = F(\mu(A))$. Distribution functions for capacities were used by Buja (1986) and Bednarski (1981) in a different context.

EXAMPLES 2.1–2.4 (Continued). The distribution functions for these examples are, respectively, $F(\omega) = (1 - \varepsilon)\omega + \varepsilon$, $F(\omega) = \min\{\omega + \varepsilon, 1\}$, $F(\omega) = \min\{b\omega, 1 - a(1 - \omega)\}$ and $F(\omega) = k\omega[k\omega + (1 - \omega)]^{-1}$.

Let $\text{gr } F = \{(\omega, y) \in [0, 1] \times [0, 1]; y \leq F(\omega)\}$. If $a = (a_1, a_2)$ and $b = (b_1, b_2)$ are two points in the plane with $a_1 < b_1$, let $L_{a,b}(\omega) = a_2 + (\omega - a_1)(b_2 - a_2)/(b_1 - a_1)$ and $\mathcal{L}(a, b) = \{(\omega, L_{a,b}(\omega)); a_1 \leq \omega \leq b_1\}$. We say that F is *doubly star-shaped* if $a \in \text{gr } F$ implies that $\mathcal{L}(\mathbf{0}, a) \subset \text{gr } F$ and $\mathcal{L}(a, \mathbf{1}) \subset \text{gr } F$, where $\mathbf{0} = (0, 0)$ and $\mathbf{1} = (1, 1)$.

In general, it is difficult to know whether a capacity is coherent; see Anger and Lembcke (1985) for example. The next theorem characterizes coherent, symmetric capacities in terms of their distribution functions.

THEOREM 3.1. *If C is symmetric and coherent, then its distribution F is doubly star-shaped. Conversely, if $F: [0, 1] \rightarrow [0, 1]$ is doubly star-shaped, then there exists a symmetric, coherent capacity C such that $F(\omega) = C([0, \omega])$ for every $\omega \in \Omega$.*

PROOF. Let C be symmetric. Fix $A = [0, t]$, $t \in (0, 1)$. There exists $P \triangleleft C$ such that $P(A) = C(A)$. Define $R(\cdot) = P(A)\mu(\cdot \cap A)/\mu(A) + P(A^c)\mu(\cdot \cap A^c)/\mu(A^c)$. Then $R \triangleleft P$; hence $R \triangleleft C$. So, for every $\omega \in \Omega$, $G(\omega) \equiv R([0, \omega]) \leq C([0, \omega]) = F(\omega)$. Since G is piecewise linear this implies that $\mathcal{L}(\mathbf{0}, a) \in \text{gr } F$ and $\mathcal{L}(a, \mathbf{1}) \in \text{gr } F$, where $a = (\omega, F(\omega))$. It follows that $\text{gr } F$ is doubly star-shaped.

Now we construct a class M that generates a symmetric capacity C with distribution F . For every measurable set A such that $0 < \mu(A) < 1$, define $P_A(\cdot) = F(\mu(A))\mu(\cdot \cap A)/\mu(A) + (1 - F(\mu(A)))\mu(\cdot \cap A^c)/\mu(A^c)$. For

nonempty A such that $\mu(A) = 0$, define $P_A(\cdot) = F(0)\delta_{\omega_A} + (1 - F(0))\mu(\cdot)$, where $\omega_A \in A$. For A with $\mu(A) = 1$, define $P_A = \mu$. Let $M = \{P_A; A \in \mathcal{B}, A \neq \emptyset\}$ and $C(A) = \sup_{P \in M} P(A)$. We claim that $C(A) = P_A(A)$. First, suppose that $A = [0, t]$, $t \in (0, 1)$. For any C , $P_A(A) \geq P_C(A) \geq P_C(A)$, where $\hat{C} = [0, \mu(C)]$. The first inequality is from the fact that F is doubly star-shaped. The second follows from how P_A is defined. Hence, $P_A(A) = C(A)$. Now consider any A such that $\mu(A) > 0$. Let C be any set, let $\hat{A} = [0, \mu(A)]$ and let D be such that $\mu(D) = \mu(C)$ and $\mu(D \cap \hat{A}) = \mu(C \cap A)$. Then $P_A(A) = P_{\hat{A}}(\hat{A}) \geq P_D(\hat{A}) = P_C(A)$. Hence, $P_A(A) = C(A)$. It follows that C is symmetric and that $C([0, \omega]) = F(\omega)$. \square

Let \mathcal{F} be the set of doubly star-shaped functions taking $[0, 1]$ to $[0, 1]$. The following propositions record some basic properties of doubly star-shaped functions. The proof of Proposition 3.1 is straightforward and is omitted.

PROPOSITION 3.1. *If $F \in \mathcal{F}$, then:*

- (i) $F(1) = 1$;
- (ii) $F(\omega) \geq \omega$ for every $\omega \in \Omega$;
- (iii) F is strictly increasing on $\{\omega; F(\omega) < 1\}$;
- (iv) F is continuous.

Given a function F and a point $\omega \in (0, 1)$, define

$$\begin{aligned}
 v^\omega(y) &= \left(\frac{1 - F(\omega)}{1 - \omega} \right) y + \left(\frac{F(\omega) - \omega}{1 - \omega} \right), \\
 \lambda^\omega(y) &= \frac{F(\omega)}{\omega} y, \\
 \xi^\omega(y) &= \lambda^\omega(y)I_{[0, \omega)}(y) + v^\omega(y)I_{[\omega, 1]}(y).
 \end{aligned}
 \tag{3}$$

When $\omega = 1$, define $v^\omega(y) \equiv 1$ and $\lambda^\omega(y) = y$.

PROPOSITION 3.2. *Let $F: \Omega \rightarrow \Omega$. The following four statements are equivalent:*

- (i) $F \in \mathcal{F}$.
- (ii) For all $\omega \in (0, 1]$ and all $0 \leq y \leq \omega$, $\lambda^\omega(y) \leq F(y) \leq v^\omega(y)$.
- (iii) F is a continuous function such that $F(1) = 1$ and is strictly increasing on $\{\omega; F(\omega) < 1\}$. Hence, the derivative exists almost everywhere. Furthermore, for almost all $\omega \in (0, 1]$,

$$\frac{1 - F(\omega)}{1 - \omega} \leq F'(\omega) \leq \frac{F(\omega)}{\omega}.
 \tag{4}$$

- (iv) For every $\omega \in \Omega$, $F(\omega) = \sup_{G \in \mathcal{G}} G(\omega)$, where \mathcal{G} is a nonempty set of functions such that each $G \in \mathcal{G}$ satisfies (a) $G: \Omega \rightarrow \Omega$, (b) $G(1) = 1$, (c) G is concave and (d) $G(\omega) \leq F(\omega)$ for all ω .

REMARK. Statement (ii) shows that λ^x and v^x provide a lower and upper bound for F .

PROOF. (i) implies (ii): straightforward.

(ii) implies (iii): straightforward.

(iii) implies (iv): fix $\omega \in (0, 1]$. Suppose there exists $y \in (0, \omega]$ such that $F(y) < \xi^\omega(y)$.

Let $y_0 = \inf\{z \in [y, \omega]; F(z) \geq \xi^\omega(z)\}$. Thus, $F(y_0) = \xi^\omega(y_0)$ and $F(x) < \xi^\omega(x)$ for $x \in [y, y_0)$. Hence, $F(\omega)/\omega = (\xi^\omega(y_0) - \xi^\omega(y))/(y_0 - y) < (F(y_0) - F(y))/(y_0 - y)$. From this, together with (4) we derive the following contradiction:

$$\begin{aligned} \frac{F(\omega)}{\omega} &< \frac{F(y_0) - F(y)}{y_0 - y} = \frac{\int_y^{y_0} F'(s) ds}{y_0 - y} \\ &\leq \frac{\int_y^{y_0} F(s)/s ds}{y_0 - y} < \frac{\int_y^{y_0} \xi^\omega(s)/s ds}{y_0 - y} \\ &= \frac{\int_y^{y_0} F(\omega)/\omega ds}{y_0 - y} = \frac{F(\omega)}{\omega}. \end{aligned}$$

So we conclude that $F(y) \geq \xi^\omega(y)$ for all $y \in (0, \omega]$. A similar argument shows that $F(y) \geq \xi^\omega(y)$ for all $y \in [\omega, 1]$. Now take $\mathcal{S} = \{\xi^\omega; x \in [0, 1]\}$.

(iv) implies (i): for $\omega \in [0, 1]$ and $\varepsilon > 0$, find $G_\varepsilon \in \mathcal{S}$ such that $G(\omega) \geq F(\omega) - \varepsilon$. Let $S_\varepsilon(\omega)$ be defined as ξ^ω is defined in (3) except with G_ε in place of F . The concavity of G_ε implies that $S_\varepsilon(y) \leq F(y)$ for all y . Note that $S_\varepsilon \uparrow \xi^\omega$ as $\varepsilon \downarrow 0$. It follows that $\xi^\omega \leq F$. This holds for every ω and (i) follows. \square

Note that, by Proposition 3.2(iii), if $F: \Omega \rightarrow \Omega$ is in \mathcal{F} , then there exists a measurable function α on Ω such that $0 \leq \alpha(\omega) \leq 1$ almost everywhere and

$$(5) \quad F'(\omega) = \frac{\alpha(\omega)F(\omega)}{\omega} + \frac{(1 - \alpha(\omega))(1 - F(\omega))}{(1 - \omega)}$$

almost everywhere. From this observation, we are motivated to consider the transform from F to the corresponding α .

Define the function α_1 by $\alpha_1(\omega) = 1$ for all $\omega \in \Omega$. Let $\mathcal{A} = \{\alpha_1\} \cup \mathcal{A}_\infty$, where \mathcal{A}_∞ is the set of all measurable functions on Ω such that $0 \leq \alpha(\omega) \leq 1$ for μ -almost all ω and

$$(6) \quad \lim_{\omega \rightarrow 1} \int_{1/2}^{\omega} \frac{\alpha(u) - u}{u(1 - u)} du = -\infty.$$

Define the function F_1 by $F_1(\omega) = \omega$. Given a function F , define $\alpha = \mathcal{F}(F)$ and $c = \mathcal{U}(F)$ by

$$(7) \quad \alpha(\omega) = \begin{cases} \frac{F'(\omega) - (1 - F(\omega))/(1 - \omega)}{F(\omega)/\omega - (1 - F(\omega))/(1 - \omega)}, & \text{if } F \neq F_1, \\ 1, & \text{if } F = F_1 \end{cases}$$

and $c = F(1/2) - 1/2$.

LEMMA 3.1. *Let $F \in \mathcal{F}$. If $\alpha = \mathcal{F}(F)$ and $c = \mathcal{U}(F)$, then $\alpha \in \mathcal{A}$ and $0 \leq c \leq 1/2$. Furthermore, $c = 0$ if and only if $\alpha = \alpha_1$.*

PROOF. If $F = F_1$, then $\alpha = \alpha_1 \in \mathcal{A}$ and $c = 0$. Suppose that $F \neq F_1$. Since $F \in \mathcal{F}$, we have by Proposition 3.2(iii) that

$$\frac{1 - F(\omega)}{1 - \omega} \leq F'(\omega) \leq \frac{F(\omega)}{\omega}$$

for almost all ω . Thus, for some function β such that $0 \leq \beta(\omega) \leq 1$ (almost all ω), we can write

$$F'(\omega) = \beta(\omega) \frac{F(\omega)}{\omega} + (1 - \beta(\omega)) \frac{1 - F(\omega)}{1 - \omega}.$$

Solving this equation and comparing it to (7), we see that $\beta(\omega) = \alpha(\omega)$ which shows that $0 \leq \alpha(\omega) \leq 1$ almost everywhere. Now we show that (6) holds. Note that $\alpha(\omega)/(\omega(1 - \omega)) = (F'(\omega) - 1)/(F(\omega) - \omega)$. Since this last expression is equal to $d \log(F(\omega) - \omega)/d\omega$, it follows that

$$\lim_{\omega \rightarrow 1} \int_{1/2}^{\omega} \frac{\alpha(u) - u}{u(1 - u)} du = \lim_{\omega \rightarrow 1} \left(\log(F(\omega) - \omega) - \log\left(F\left(\frac{1}{2}\right) - \frac{1}{2}\right) \right) = -\infty,$$

since $F(1) = 1$ and $F(1/2) > 1/2$. Thus, $\alpha \in \mathcal{A}$. The remarks about c follow immediately from the fact that $1/2 \leq F(1/2) \leq 1$ and the fact that $F(1/2) = 1/2$ if and only if $F = F_1$. \square

Given a function α and a real number c , define a function $F = \mathcal{R}(\alpha, c)$ by

$$(8) \quad F(\omega) = \omega + c \exp\left\{ \int_{1/2}^{\omega} \frac{\alpha(u) - u}{u(1 - u)} du \right\}.$$

Also, let

$$c_{\alpha} = \inf_{\omega} (1 - \omega) \exp\left\{ - \int_{1/2}^{\omega} \frac{\alpha(u) - u}{u(1 - u)} du \right\}.$$

LEMMA 3.2. *Let $\alpha \in \mathcal{A}$ and $c \in \mathbb{R}$ be given. Let $F = \mathcal{R}(\alpha, c)$. If $0 \leq c \leq c_{\alpha}$, then $F \in \mathcal{F}$. Furthermore, $\mathcal{F}(F) = \alpha$ and $\mathcal{U}(F) = c$.*

PROOF. If $\alpha = \alpha_1$, then $c = 0$ and the claim follows easily. So assume that $c > 0$. From (6) we deduce that $F(1) = 1$. Clearly, $F(\omega) \geq 0$ for ω and F is continuous. That $F(\omega) \leq 1$ for all ω follows from the condition on c . Now differentiate F and use the fact that $0 \leq \alpha(\omega) \leq 1$ for almost all ω to conclude that

$$\frac{1 - F(\omega)}{1 - \omega} \leq F'(\omega) \leq \frac{F(\omega)}{\omega}$$

for almost all ω . Since $F(\omega) \leq 1$ for all ω , this implies that F' is nonnegative almost everywhere. Furthermore, on the set $\{\omega; F(\omega) < 1\}$, F' is strictly positive wherever the derivative is defined. Thus, F is strictly increasing on this set. By Proposition 3.2(iii) it follows that $F \in \mathcal{F}$. That $\mathcal{S}(F) = \alpha$ and $\mathcal{Z}(F) = c$ follow from direct calculation. \square

Finally, we are in a position to characterize the extreme points of \mathcal{F} . Denote the set of extreme points by \mathcal{E} .

THEOREM 3.2 (Characterization of extreme F). *Suppose that $F \in \mathcal{F}$. Then $F \in \mathcal{E}$ if and only if:*

- (i) $\mu(\{\omega; 0 < \mathcal{S}(F)(\omega) < 1\}) = 0$;
- (ii) $\mathcal{Z}(F) \in \{0, c_\alpha\}$.

PROOF. Suppose first that (ii) fails to hold. Choose $0 < \varepsilon < \min\{c, c_\alpha - c\}$. Define $c_1 = c + \varepsilon$ and $c_2 = c - \varepsilon$. Let $F_i = \mathcal{R}(\alpha, c_i)$, $i = 1, 2$. By Lemma 3.2, $F \in \mathcal{F}$, $i = 1, 2$. Clearly, F_1, F_2 and F are distinct and $F = (1/2)F_1 + (1/2)F_2$ so F is not extreme. Suppose now that (i) fails to hold. Let A be a set of positive measure such that α_F is strictly between 0 and 1. We may find $\delta > 0$, $\omega_0 \in (0, 1)$ and $a > 0$ such that $\delta < \text{ess inf } \alpha(\omega) \leq \text{ess sup } \alpha(\omega) < 1 - \delta$ for $\delta \in I = (\omega_0 - a, \omega_0 + a) \subset (0, 1)$. Fix $\Delta \in \mathbb{R}$. Define

$$\alpha_1(\omega) = \alpha(\omega) + \frac{\omega(1 - \omega)f'(\omega)}{1 + f(\omega)}$$

and

$$\alpha_2(\omega) = \alpha(\omega) - \frac{\omega(1 - \omega)f'(\omega)}{1 - f(\omega)},$$

where $f(\omega) = \Delta(\omega - \omega_0)$ for $\omega \in I$ and $f(\omega) = 0$ otherwise. Let $F_i = \mathcal{R}(\alpha_i, c)$, $i = 1, 2$. Now

$$F_1(\omega) - F_2(\omega) = 2f(\omega) \exp \left\{ \int_{1/2}^\omega \frac{\alpha(u) - u}{u(1 - u)} \right\},$$

so F_1 and F_2 are distinct. It is easy to see that $F = (1/2)F_1 + (1/2)F_2$. Finally, we claim that, for $\Delta > 0$ sufficiently small, $F_1, F_2 \in \mathcal{F}$. For Δ small and from the fact that α is almost surely between δ and $1 - \delta$ on I , we conclude that α_1 and α_2 are almost surely between 0 and 1. It is easy to

verify that (6) holds for both α_1 and α_2 . Also, for $i = 1, 2$, $c_{\alpha_i} \rightarrow c$ as $\Delta \rightarrow 0$ so that, for small Δ , $0 \leq c \leq \min\{c_{\alpha_1}, c_{\alpha_2}\}$. Hence, Lemma 3.2 implies that $F_1, F_2 \in \mathcal{F}$. Hence, F is not extreme.

Now suppose that (i) and (ii) hold. Let $F = pF_1 + (1 - p)F_2$ with $p \in (0, 1)$. Let $\alpha = \mathcal{A}(F)$ and let $\alpha_i = \mathcal{A}(F_i)$, $i = 1, 2$. Let $A_1 = \{\omega; \alpha(\omega) = 1\}$. Suppose there exists $A \subset A_1$ such that $\mu(A) > 0$ and $\alpha_1(\omega) < 1$ on A . From (5), together with the fact that $F'_1(\omega) \leq F_1(\omega)/\omega$ for almost all ω , we conclude that $F'_1(\omega) < F_1(\omega)/\omega$ for almost all $\omega \in A$. Thus, for almost all $\omega \in A$, we have

$$\begin{aligned} F'(\omega) &= pF'_1(\omega) + (1 - p)F'_2(\omega) < pF_1(\omega)/\omega + (1 - p)F_2(\omega)/\omega \\ &= F(\omega)/\omega. \end{aligned}$$

But since $\alpha(\omega) = 1$ on A we have that $F'(\omega) = F(\omega)/\omega$ which is a contradiction. Hence, $\alpha_1(\omega) = 1$ for almost all ω in A_1 and similarly for α_2 . By a similar argument, $\alpha_1(\omega) = \alpha_2(\omega) = 0$ for almost all $\omega \in A_0 = \{\omega; \alpha(\omega) = 0\}$. Since $\mu(A_0 \cup A_1) = 1$, $\alpha = \alpha_1 = \alpha_2$ almost everywhere. It follows that $c_{\alpha_1} = c_{\alpha_2} = c_\alpha$. Let $c_i = \mathcal{Z}(F_i)$, $i = 1, 2$. Since $c = 0$ or $c = c_\alpha$ it follows that $c = c_1 = c_2$. Thus, $F = F_1 = F_2$. \square

Now we consider another characterization of the extreme points. Given $F \in \mathcal{F}$, define

$$\begin{aligned} \Gamma_0(F) &= \{\omega; \lambda^\omega(y) < F(y) < v^\omega(y) \text{ for all } y < \omega\}, \\ \Gamma_1(F) &= \{\omega; \text{there exists some } y < x \text{ such that } F(z) = v^\omega(y) \text{ for all} \\ &\qquad\qquad\qquad z \in [y, \omega]\}, \\ \Gamma_2(F) &= \{\omega; \text{there exists some } y < x \text{ such that } F(z) = \lambda^\omega(y) \text{ for all} \\ &\qquad\qquad\qquad z \in [y, \omega]\}. \end{aligned}$$

PROPOSITION 3.3. *If $F \in \mathcal{F}$, then $\Gamma_0(F)$, $\Gamma_1(F)$ and $\Gamma_2(F)$ partition $(0, 1]$.*

PROOF. Suppose that $F(y) = v^\omega(y)$ for some $y < \omega$. Let $a = (y, v(y))$ and $b = (x, F(x))$. For all $x \in [y, \omega]$, the double star-shaped condition implies that $F(z) \geq \mathcal{L}_{a,b}(z) = v^\omega(z)$. By Proposition 3.2(ii), $F(z) \leq v^\omega(z)$. Hence, $F(z) = v^\omega(z)$ and $\omega \in \Gamma_1(F)$. By a similar argument, if $F(y) = \lambda^\omega(y)$ for some $y < \omega$, then it can be shown that $\omega \in \Gamma_2(F)$. If $F(y)$ is never equal to $v^\omega(y)$ or $\lambda^\omega(y)$ for any $y < z$, then $\omega \in \Gamma_3(F)$. \square

COROLLARY 3.1. *F is an extreme point in \mathcal{F} if and only if $\Gamma_0(F) = \emptyset$.*

PROOF. Let $\alpha = \mathcal{A}(F)$. Suppose that $\Gamma_0(F) \neq \emptyset$. Choose $\omega \in \Gamma_0(F)$. Then α is strictly between 0 and 1 in some interval $(\omega - \delta, \omega]$. Hence, by the previous theorem, F is not extreme. Now suppose that F is not extreme.

Thus, either $0 < c < c_\alpha$ or α is strictly between 0 and 1 on a set of positive measure. First suppose that $0 < c < c_\alpha$. It is easy to see that $1 \in \Gamma_0(F)$ so that $\Gamma_0(F) \neq \emptyset$. Now suppose that α is strictly between 0 and 1 on a set of positive measure. In particular, this is true for some open interval (a, b) . Thus, $b \in \Gamma_0$ so that $\Gamma_0(F) \neq \emptyset$. \square

The latter corollary gives a geometric interpretation to the extreme points: they are piecewise linear functions that oscillate between the upper and lower supporting lines v^ω and λ^ω .

4. The 2-alternating case. In the previous section we identified the extreme points of \mathcal{F} . In this section we identify the extreme points of the subset \mathcal{F}_2 of distribution functions corresponding to 2-alternating Choquet capacities. It turns out that \mathcal{F}_2 is a simplex so that each $F \in \mathcal{F}_2$ has a unique Choquet representation as a mixture of the extreme points. We identify the mixing distribution explicitly.

For every $t \in (0, 1]$, define $F_t(\omega) = \min\{\omega/t, 1\}$. Also, define $F_0(\omega) \equiv 1$. Let \mathcal{E}_2 be the extreme points of \mathcal{F}_2 .

THEOREM 4.1. $\mathcal{E}_2 = \{F_t; 0 \leq t \leq 1\}$.

PROOF. It is easy to establish that each F_t is an extreme point. That this set exhausts all extreme points follows from Theorem 4.2. \square

THEOREM 4.2 (Choquet representation for symmetric, 2-alternating capacities). *For every $F \in \mathcal{F}_2$ there exists a probability measure Q on $([0, 1], \mathcal{B})$ such that, for all $\omega \in \Omega$,*

$$(9) \quad F(\omega) = \int_0^1 F_t(\omega) Q(dt).$$

Furthermore, Q is unique and satisfies $dQ/d\nu = r(\omega)$, where

$$r(\omega) = F(0)I_0 - \omega F''(\omega)I_B(\omega) - \omega(F'(\omega^+) - F'(\omega^-))I_{B^c}(\omega),$$

$\nu = \mu + \tilde{\mu}$, $\tilde{\mu}$ is counting measure on B^c and B is the set where F'' exists. By convention, we set $F'(1^+) = 0$ and we include 0 and 1 in B^c .

PROOF. Since $F \in \mathcal{F}_2$, it is concave by a straightforward generalization of Theorem 5 of Wasserman and Kadane (1992). Thus, F'' exists on a set B such that B^c is countable. Define $G(\omega) = \int_{[0, \omega] \cap B} r(s)\nu(ds)$. Now G is non-negative, nondecreasing and right-continuous and so is a distribution function for a measure Q . We need only show that (9) holds. This will establish that Q is a probability measure since $F(\omega) = 1$.

Let $a(t) = F'(\omega^+) - F'(\omega^-)$. First suppose that B^c is finite. We can thus write $B^c = \{\omega_0, \dots, \omega_n\}$, where $0 = \omega_0 < \omega_1 < \dots < \omega_n = 1$. Suppose that $\omega_j < \omega_{j+1}$, say. Then

$$\begin{aligned}
 & \int_0^1 F_t(\omega) Q(dt) \\
 &= \int_0^1 F_t(\omega) \frac{dQ}{d\nu}(t) \nu(dt) \\
 &= \int_0^1 F_t(\omega) r(t) \mu(dt) + \sum_{t \in B^c} F_t(\omega) r(t) \\
 &= - \int_0^1 F_t(\omega) t F''(t) \mu(dt) - \sum_{t \in B^c} F_t(\omega) t a(t) \\
 &= - \int_0^\omega t F''(t) \mu(dt) - \omega \int_\omega^1 F''(t) \mu(dt) \\
 &\quad - \sum_{t \in B^c \cap [0, \omega]} t a(t) - \omega \sum_{t \in B^c \cap (\omega, 1]} a(t) \\
 &= - \sum_{i=0}^{j-1} (\omega_{i+1} F'(\omega_{i+1}^-) - \omega_i F'(\omega_i^+)) + F(\omega_j) - F(0) \\
 &\quad - \omega F''(\omega) + \omega_j F'(\omega_j^+) + F(\omega) - F(\omega_j) - \omega F'(\omega_{j+1}^-) \\
 &\quad + \omega F'(\omega) - \omega \sum_{i=j+1}^{n-1} (F'(\omega_{i+1}^-) - F'(\omega_i^+)) \\
 &\quad - \sum_{t \in B^c \cap [0, \omega]} t a(t) - \omega \sum_{t \in B^c \cap (\omega, 1]} a(t) \\
 &= F(\omega) - F(0) + \sum_{i=1}^j \omega_i (F'(\omega_i^-) - F'(\omega_i^+)) \\
 &\quad + \omega \sum_{i=j+1}^n (F'(\omega_i^-) - F'(\omega_i^+)) \\
 &\quad - \sum_{t \in B^c \cap [0, \omega]} t a(t) - \omega \sum_{t \in B^c \cap (\omega, 1]} a(t) \\
 &= F(\omega).
 \end{aligned}$$

A similar calculation can be used when $\omega \in B^c$.

Now suppose that B^c is denumerable. Fix $\varepsilon > 0$. Choose $A = \{\omega_0, \dots, \omega_n\} \subset B^c$ such that $0 = \omega_0 < \omega_1 < \dots < \omega_n = 1$, and $-\sum_{\omega \in A} \omega a(\omega) > -\sum_{\omega \in B^c} \omega a(\omega) - \varepsilon$. Let $W_1 = [0, \omega_1]$, $W_2 = (\omega_1, \omega_2]$, \dots , $W_n = (\omega_{n-1}, 1]$. We can find $F_\varepsilon \in \mathcal{F}_2$ such that $\sup_{\omega \in \Omega} |F_\varepsilon(\omega) - F(\omega)| < \varepsilon$, $\sup_{\omega \in B} |F''_\varepsilon(\omega) - F''(\omega)| < \varepsilon$ and

$$\sum_{\omega \in A} ([F'(\omega^+) - F'(\omega^-)] - [F''_\varepsilon(\omega^+) - F''_\varepsilon(\omega^-)]) < \varepsilon.$$

Let Q_ε be the representing measure for F_ε and let $r_\varepsilon = dQ_\varepsilon/d\nu$. Now $r_\varepsilon \rightarrow r$ uniformly in ω as $\varepsilon \rightarrow 0$. It follows that Q_ε converges weakly to Q . Now

$$F(\omega) - \int_0^1 F_t(\omega)Q(dt) = (F(\omega) - F_\varepsilon(\omega)) + \left(F_\varepsilon(\omega) - \int_0^1 F_t(\omega)Q_\varepsilon(dt) \right) \\ + \left(\int_0^1 F_t(\omega)Q_\varepsilon(dt) - \int_0^1 F_t(\omega)Q(dt) \right).$$

The first term tends to 0 as $\varepsilon \rightarrow 0$ since F_ε approximates F . The second term is identically 0. The third term tends to 0 as $\varepsilon \rightarrow 0$ since F_t is bounded and since Q_ε converges weakly to Q . Taking the limit of both sides of the equation as $\varepsilon \rightarrow 0$ establishes (9).

Finally, we show the uniqueness of Q . Suppose there exists another representing probability R . Note that $R \ll \nu$. Let $q = dQ/d\nu$ and $r = dR/d\nu$. Since, for all ω , $F(\omega) = \int_0^1 F_t(\omega)Q(dt) = \int_0^1 F_t(\omega)R(dt)$ we conclude that $\int_0^1 (1 - F_t(\omega))Q(dt) = \int_0^1 (1 - F_t(\omega))R(dt)$ which implies that

$$\int_0^\omega \frac{q(t) - r(t)}{1 - t^{-1}} d\nu(t) = 0$$

for all ω . Differentiating with respect to ω , we conclude that $q(\omega) = r(\omega)$ for almost all ω . \square

Since \mathcal{F}_2 is convex and since each F can be represented as a unique mixture of extreme points, it follows that \mathcal{F}_2 is a simplex [Choquet (1969), Section 28]. The class \mathcal{F}_2 might be pictured as a triangle inscribed in a circle, being representable as a unique mixture of the extreme points of $\mathcal{F}_2 \subset \mathcal{F}$ but, of course, representable as many different mixtures of the elements of \mathcal{E} . Finally, we remark that 2-alternating is equivalent to

$$\alpha'(\omega) \leq \frac{\alpha(\omega)(1 - \alpha(\omega))}{\omega(1 - \omega)}$$

for almost all ω .

5. Conclusion. In this paper we gave a characterization of the extreme points of the set of distribution functions for symmetric capacities. With the appropriate transform, these correspond to a subset of the 0–1 functions on the unit interval. In the 2-alternating case, each distribution function has a Choquet representation as a unique mixture of the extreme points. Given the special role that 2-alternating capacities play in robustness theory, it would be interesting to develop a measure of the degree to which a capacity fails to be 2-alternating. An open question is whether the characterizations given in this paper can be used to develop such a measure. If they can, then it might be possible to quantify the degree to which the Huber–Strassen [Huber and Strassen (1973)] theorem and the Bayes theorem for Choquet capacities [Wasserman and Kadane (1990)] fail when a capacity is not 2-alternating.

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