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Continuous Previsions^{*}

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Abstract. We define strong monads of *continuous (lower, upper) previsions*, and of *forks*, modeling both probabilistic and non-deterministic choice. This is an elegant alternative to recent proposals by Mislove, Tix, Keimel, and Plotkin. We show that our monads are sound and complete, in the sense that they model exactly the interaction between probabilistic and (demonic, angelic, chaotic) choice.

1 Introduction

Moggi’s computational λ -calculus [17] has proved useful to define various notions of computations on top of the lambda-calculus: side-effects, input-output, continuations, non-determinism [26], probabilistic computation [20] in particular. But mixing monads is hard, and finding the “right” monad that would combine both non-determinism and probabilistic choice has taken quite some effort. (We review recent progress below.)

The purpose of this paper is to introduce simple monads that do the job well. These are monads of *continuous previsions*, which can be seen as continuation-style monads. The idea of considering previsions comes from economics and statistics [5, 12].

Outline. After stating some required preliminaries in Section 2, we recall the notion of *game* introduced in [6], arguing why these are natural extensions of notions of continuous valuations (\sim measures) that also accommodate demonic and angelic non-deterministic choice. These notions induce functors on *Top*, *Cpo*, *Pcpo*, but fail to yield monads. We analyze this failure in Section 4 by moving, through a Riesz-like representation theorem, to the new notions of collinear previsions, and previsions. We then show that indeed previsions yield strong monads, suitable to give semantics to a rich λ -calculus [17] with both probabilistic and non-deterministic choice. Finally, we show in Section 5 that our monad model is not only sound but complete.

This work is a summary of most of Chapters 10-12 of [7]. We however give the proofs of the main results in the Appendix.

Related Work. Finding a monad combining both probabilistic and non-deterministic choice can be done by using general monad combination principles. The right way to combine monads in general is open to discussion. Lüth [11] proposes to combine monads by taking their coproduct in the category of monads. This coproduct exists under relatively mild assumptions [10]. However, in general the coproduct of two monads is an inscrutable object. A simpler, explicit description can be found in specific cases. For

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example, when taking coproducts of two *ideal* monads [3]. In particular, combining *non-blocking* non-determinism and probabilistic choice falls into this case. The resulting monad is relatively unenlightening, though: it is the monad of all sequences of choices, both probabilistic and non-deterministic [3, exemple 4.3].

Varacca [25] also proposed a monad combining non-determinism with probabilistic choice. Ghani and Uustalu [3] note that the above coproduct monad is close to Varacca's synchronization trees. The works closer to ours in computer science are those of Mislove [16] and Tix [23, 24]. While this won't be entirely obvious from our definitions, we will establish a formal connection between their models and ours (Section 5). Outside computer science, previsions have their roots in economics and statistics [27]. However, we consider previsions on topological spaces, not just on sets.

This paper can be seen also be seen as a followup to [6], inasmuch as previsions are strongly tied to notions of convex and concave games.

2 Preliminaries

We assume the reader to be familiar with (point-set) topology, in particular topology of T_0 but not necessarily Hausdorff spaces, as encountered in domain theory. See [4, 1, 15] for background. Let $\text{int}(A)$ denote the interior of A , $\text{cl}(A)$ its closure.

The *Scott topology* on a poset X , with ordering \leq , has as opens the upward-closed subsets U (i.e., $x \in U$ and $x \leq y$ imply $y \in U$) such that for every directed family $(x_i)_{i \in I}$ having a least upper bound $\sup_{i \in I} x_i$ inside U , some x_i is already in U . The *way-below* relation \ll is defined by $x \ll y$ iff for any directed family $(z_i)_{i \in I}$ with a least upper bound z such that $y \leq z$, then $x \leq z_i$ for some $i \in I$. A poset is *continuous* iff $\downarrow y = \{x \in X \mid x \ll y\}$ is directed, and has x as least upper bound. Then every open U can be written $\bigcup_{x \in U} \uparrow x$, where $\uparrow x = \{y \in X \mid x \ll y\}$.

Every topological space X has a specialization quasi-ordering \leq , defined by: $x \leq y$ iff every open that contains x contains y . X is T_0 iff \leq is a (partial) ordering. That of the Scott topology of a quasi-ordering \leq is \leq itself. A subset $A \subseteq X$ is *saturated* if and only if A is the intersection of all opens that contain it; alternatively, iff A is upward-closed in \leq . Every open is upward-closed. Let $\uparrow A$ denote the upward-closure of A under a quasi-ordering \leq , $\downarrow A$ its downward-closure. A T_0 space is *sober* iff every irreducible closed subset is the closure $\text{cl}\{x\} = \downarrow x$ of a (unique) point x . The Hofmann-Mislove Theorem implies that every sober space is *well-filtered* [9]: given any filtered family of saturated compacts $(Q_i)_{i \in I}$ in X , and any open U , $\bigcap_{i \in I} Q_i \subseteq U$ iff $Q_i \subseteq U$ for some $i \in I$. In particular, $\bigcap_{i \in I} Q_i$ is saturated compact. X is *locally compact* iff whenever $x \in U$ (U open) there is a saturated compact Q such that $x \in \text{int}(Q) \subseteq Q \subseteq U$. Every continuous cpo is sober and locally compact in its Scott topology. X is *coherent* iff the intersection of any two compacts is compact. A coherent, well-filtered locally compact space is called *stably locally compact*. *Stably compact* spaces are those that are additionally compact, and have a wonderful theory (see, e.g., [9]). We shall consider the space \mathbb{R} of all reals with the Scott topology of its natural ordering \leq . Its opens are \emptyset , \mathbb{R} , and the intervals $(t, +\infty)$, $t \in \mathbb{R}$. \mathbb{R} is a stably locally compact, continuous cpo. Because we equip \mathbb{R} with the Scott topology, our *continuous*

functions $f : X \rightarrow \mathbb{R}$ are those which are usually called *lower semi-continuous* in the mathematical literature.

We call *capacity* on X any function ν from $\mathcal{O}(X)$, the set of all opens of X , to \mathbb{R}^+ , such that $\nu(\emptyset) = 0$ (a.k.a., a *set function*). A *game* ν is a *monotonic capacity*, i.e., $U \subseteq V$ implies $\nu(U) \leq \nu(V)$ ¹. A *valuation* is a *modular game* ν , i.e., one such that $\nu(U \cup V) + \nu(U \cap V) = \nu(U) + \nu(V)$ for every opens U, V . A game is *continuous* iff $\nu(\bigcup_{i \in I} U_i) = \sup_{i \in I} \nu(U_i)$ for every directed family $(U_i)_{i \in I}$ of opens, and *normalized* iff $\nu(X) = 1$. Continuous valuations have a nice theory that fits topology well [8, 9].

The *Dirac valuation* δ_x at $x \in X$ is the continuous valuation mapping each open U to 1 if $x \in U$, to 0 otherwise. Continuous valuations are canonically ordered by $\nu \leq \nu'$ iff $\nu(U) \leq \nu'(U)$ for every open U of X .

A *monad* on a category \mathbb{C} may be presented in several different ways. One is based on triples $(\mathbf{T}, \boldsymbol{\eta}, \boldsymbol{\mu})$ of an endofunctor on \mathbb{C} , a unit, and a multiplication natural transformation. A presentation that is easier to grasp is in terms of Kleisli triples [14]. A *Kleisli triple* is a triple $(\mathbf{T}, \boldsymbol{\eta}, \dashv)$, where \mathbf{T} maps objects X of \mathbb{C} to objects $\mathbf{T}X$ of \mathbb{C} , $\boldsymbol{\eta}_X$ is a morphism from X to $\mathbf{T}X$ for each X , and f^\dagger (the *extension of f*) is a morphism from $\mathbf{T}X$ to $\mathbf{T}Y$ for each morphism $f : X \rightarrow Y$, satisfying: (1) $\boldsymbol{\eta}_X^\dagger = \text{id}_{\mathbf{T}X}$; (2) for every $f : X \rightarrow Y$, $f^\dagger \circ \boldsymbol{\eta}_X = f$; (3) for every $g : X \rightarrow Y$, $f : Y \rightarrow Z$, then $f^\dagger \circ g^\dagger = (f^\dagger \circ g)^\dagger$. Kleisli triples and monads are equivalent.

3 Continuous Games, Convexity, Concavity

We follow [6]. A game ν on X is *convex* iff $\nu(U \cup V) + \nu(U \cap V) \geq \nu(U) + \nu(V)$ for every opens U, V . It is *concave* if the opposite inequality holds. Convex games are a cornerstone of economic theory [5, 18].

One fundamental example of a game that is not a valuation is the *unanimity game* u_A ($A \neq \emptyset$), defined by $u_A(U) = 1$ if $A \subseteq U$, $u_A(U) = 0$ otherwise. As we argue in [6], u_A is a natural “probability-like” description of demonic non-deterministic choice, in the sense that drawing “at random” according to u_A means that some malicious adversary C will choose an element of A for you. This is perhaps best conveyed by a thought experiment. You, the honest player P , would like to draw some element x from X with distribution ν (a game). Imagine you would like to know your chances of getting one from some (open) subset U of X . If ν is a probability distribution, then your chances will be equal to $\nu(U)$. This is standard. For general ν , continue to define your chances as $\nu(U)$. If $\nu = u_A$, and U does not contain A , then $\nu(U) = 0$, and your chances are zero: intuitively, C will pick an element in A , but outside U —on purpose. The only case where C is forced to pick an element in A which will suit P (i.e., be in U , too), is when $A \subseteq U$ —and then P will be pleased with probability one.

It is clear that u_A is convex. It is in fact more. Call a game ν *totally convex* iff:

$$\nu\left(\bigcup_{i=1}^n U_i\right) \geq \sum_{I \subseteq \{1, \dots, n\}, I \neq \emptyset} (-1)^{|I|+1} \nu\left(\bigcap_{i \in I} U_i\right) \quad (1)$$

¹ The name “game” is unfortunate, as there is no obvious relationship between this and games as they are usually handled in computer science, in particular with stochastic games. The notion stems from (cooperative) games in economics, where X is the set of players, not of states.

for every finite family $(U_i)_{i=1}^n$ of opens ($n \geq 1$), where $|I|$ denotes the cardinality of I . A *belief function* is a totally convex game. The dual notion of *total concavity* is obtained by replacing \cup by \cap and conversely in (1), and turning \geq into \leq . A *plausibility* is a totally concave game. If \geq is replaced by $=$ in (1), then we retrieve the familiar inclusion-exclusion principle from statistics. In particular any (continuous) valuation is a (continuous) belief function. Clearly, any belief function is a convex game. The converses of both statements fail: On $X = \{1, 2, 3\}$ with the discrete topology, $u_{\{1,2\}}$ is a belief function but not a valuation, and $\frac{1}{2}(u_{\{1,2\}} + u_{\{1,3\}} + u_{\{2,3\}} - u_{\{1,2,3\}})$ is a convex game but not a belief function.

Every game of the form $\sum_{i=1}^n a_i u_{Q_i}$, with $a_i \in \mathbb{R}^+$, and Q_i compact saturated and non-empty, is a continuous belief function, which we call *simple* belief function in [6]. When $\sum_{i=1}^n a_i = 1$, drawing an element from X “at random” (in the sense illustrated above) according to the simple belief function $\nu = \sum_{i=1}^n a_i u_{Q_i}$ intuitively corresponds to drawing one compact Q_i at random with probability a_i , then to let the malicious adversary C draw some element, demonically, from Q_i [6].

Let us turn to integration. Let ν be a game on X , and f be continuous from X to \mathbb{R}^+ (i.e., lower semi-continuous: \mathbb{R}^+ comes with the Scott topology). Assume f bounded, too, i.e., $\sup_{x \in X} f(x) < +\infty$. The *Choquet integral* of f along ν is:

$$\oint_{x \in X} f(x) d\nu = \int_0^{+\infty} \nu(f^{-1}(t, +\infty)) dt \quad (2)$$

where the right hand side is an improper Riemann integral. This is well-defined, since $f^{-1}(t, +\infty)$ is open for every $t \in \mathbb{R}^+$ by assumption, and ν measures opens. Also, since f is bounded, the improper integrals above really are ordinary Riemann integrals over some closed intervals. The function $t \mapsto \nu(f^{-1}(t, +\infty))$ is decreasing, and every decreasing (even non-continuous, in the usual sense) function is Riemann-integrable, therefore the definition makes sense.

Alternatively, any *step function* $\sum_{i=0}^n a_i \chi_{U_i}$, where $a_0 \in \mathbb{R}^+$, $a_1, \dots, a_n \in \mathbb{R}^+$, $X = U_0 \supseteq U_1 \supseteq \dots \supseteq U_n$ is a decreasing sequence of opens, and χ_U denotes the indicator function of U ($\chi_U(x) = 1$ if $x \in U$, $\chi_U(x) = 0$ otherwise) is continuous: its integral along ν then equals $\sum_{i=0}^n a_i \nu(U_i)$ —for *any* game ν . It is well-known that every bounded continuous function f can be written as the least upper bound of a sequence of step functions $f_K = a + \frac{1}{2^K} \sum_{k=1}^{\lfloor (b-a)2^K \rfloor} \chi_{f^{-1}(a + \frac{k}{2^K}, +\infty)}(x)$, $K \in \mathbb{N}$, where $a = \inf_{x \in X} f(x)$, $b = \sup_{x \in X} f(x)$. Then the integral of f along ν is the least upper bound of the increasing sequence of the integrals of f_K along ν .

The main properties of Choquet integration are as follows. First, the integral is increasing in its function argument: if $f \leq g$ then the integral of f along ν is less than or equal to that of g along ν . If ν is continuous, then integration is also Scott-continuous in its function argument. The integral is also monotonic and Scott-continuous in the game ν . Integration is linear in the game, too, so integrating along $\sum_{i=1}^n a_i \nu_i$ is the same as taking the integrals along each ν_i , and computing the obvious linear combination. However, Choquet integration is *not* linear in the function integrated, unless the game ν is a valuation. Still, it is *positively homogeneous*: integrating αf for $\alpha \in \mathbb{R}^+$ yields α

times the integral of f . And it is additive on *comonotonic* functions $f, g : X \rightarrow \mathbb{R}$ (i.e., there is no pair $x, x' \in X$ such that $f(x) < f(x')$ and $g(x) > g(x')$).

Returning to the example of a simple belief function $\nu = \sum_{i=1}^n a_i u_{Q_i}$, the properties above imply that the integral of f along ν is $\sum_{i=1}^n a_i \min_{x \in Q_i} f(x)$ [6, Proposition 1]. (Note that $f(x)$ indeed attains its minimum over Q_i , which is compact.) Another way to read this is as follows. Imagine P publishes how much money, $f(x)$, she would earn if you picked x . When $\sum_{i=1}^n a_i = 1$, it is legitimate to say that the integral of f along ν should be some form of expected income. The formula above states that, when ν is a simple belief function, your expected income is exactly what you would obtain on average by drawing Q_i at random with probability a_i , then letting the malicious adversary C pick some element of Q_i for you—minimizing your earnings $f(x)$. In other words, integrating along a simple belief function computes *average min-payoffs*.

This can be generalized to all continuous, not just simple, belief functions [6, Theorem 4]. More precisely, the space $\mathbf{Cd}_{\leq 1}(X)$ of all continuous belief functions ν on X such that $\nu(X) \leq 1$ is isomorphic to the space $\mathbf{V}_{\leq 1}(\mathcal{Q}(X))$ of continuous valuations ν^* (of total mass at most 1) over the *Smyth powerdomain* $\mathcal{Q}(X)$ of X , provided X is well-filtered and locally compact. $\mathcal{Q}(X)$ is the cpo of non-empty compact saturated subsets of X , ordered by reverse inclusion \supseteq , and is a model of demonic non-determinism. (A similar result holds for *normalized* games and valuations ν , i.e., such that $\nu(X) = 1$: $\nu \mapsto \nu^*$ is again an isomorphism from $\mathbf{Cd}_1(X)$ to $\mathbf{V}_1(\mathcal{Q}(X))$.) The construction of ν^* from ν is relatively difficult, however it is noteworthy that when $\nu = \sum_{i=1}^n a_i u_{Q_i}$ is simple, then ν^* is exactly the simple valuation $\sum_{i=1}^n a_i \delta_{Q_i}$, which describes the choice of an element Q_i at random with probability a_i , as intuition would have it.

Similarly, the space $\mathbf{Pb}_{\leq 1}(X)$ of all *continuous plausibilities* (which are all concave) ν with $\nu(X) \leq 1$ is isomorphic to $\mathbf{V}_{\leq 1}(\mathcal{H}_u(X))$ when X is stably locally compact, and where $\mathcal{H}_u(X)$ is the topological Hoare powerdomain. The latter is used to model angelic non-determinism. The corresponding *simple* plausibilities are of the form $\sum_{i=1}^n a_i \epsilon_{F_i}$, where F_i is a non-empty closed subset of X (an element of $\mathcal{H}_u(X)$), and the *example game* ϵ_F is defined so that $\epsilon_F(U) = 1$ if F meets U , $\epsilon_F(U) = 0$ otherwise: in this case C tries to help you, by finding some element in U that would also be in F , if possible.

Recall that every belief function is convex. One may show that Choquet integration along ν is *super-additive* (the integral of $f + g$ is at least that of f plus that of g) when ν is convex, and *sub-additive* (the integral of $f + g$ is at most that of f plus that of g) when ν is concave. See [5] for the finite case, [7, chapitre 4] for the topological case.

In the sequel, let $\mathbf{J}(X)$, $\nabla \mathbf{J}(X)$, $\Delta \mathbf{J}(X)$ be the spaces of plain, convex and concave continuous games respectively (“plain” meaning with no added property).

4 Continuous Previsions

For any space X , let $\langle X \rightarrow \mathbb{R}^+ \rangle$ be the space of all bounded continuous functions from X to \mathbb{R}^+ , with the Scott topology. Each continuous game ν on X gives rise to a functional $\alpha_c(\nu)$ from $\langle X \rightarrow \mathbb{R}^+ \rangle$ to \mathbb{R}^+ , mapping f to its Choquet integral along ν .

Think of $f(x)$ again as defining how much money if x is chosen from X by some computation process. We intentionally leave the notion of computation process unde-

fined. This may be the process of drawing “at random” along a game ν , as in Section 3. In the sequel, we shall explore the view that x is the output of an arbitrary program, defined in some non-deterministic and probabilistic functional language. I.e., any program returns a value x (\perp on non-termination, say), and if so P earns $f(x)$. For purely probabilistic programs (no non-deterministic choice), a prevision F is essentially a function mapping earning functions f to their average value $F(f)$, over all possible executions. Slightly more generally, for any belief function ν , there is a prevision $\alpha_e(\nu)$ that maps each $f \in \langle X \rightarrow \mathbb{R}^+ \rangle$ to the average min-payoff we get when our final earnings are given by f .

Milking out the properties of $\alpha_e(\nu)$, we arrive at:

Definition 1 (Prevision). A prevision is a functional F from $\langle X \rightarrow \mathbb{R}^+ \rangle$ to \mathbb{R}^+ such that F is positively homogeneous (for every $\alpha \geq 0$, $F(\alpha f) = \alpha F(f)$), and monotonic (if $f \leq g$ [pointwise], then $F(f) \leq F(g)$).

F is a lower prevision if moreover F is super-additive, i.e., $F(f+g) \geq F(f)+F(g)$. F is an upper prevision iff F is sub-additive: $F(f+g) \leq F(f)+F(g)$. F is collinear iff F is additive on comonotonic pairs, i.e., if whenever f and g are comonotonic, then $F(f+g) = F(f)+F(g)$. A prevision F is linear iff $F(f+g) = F(f)+F(g)$ for every $f, g \in \langle X \rightarrow \mathbb{R}^+ \rangle$.

Finally, F is continuous iff it is Scott-continuous: for every directed family $(f_i)_{i \in I}$ of bounded continuous functions with least upper bound f , $F(\sup_{i \in I} f_i) = \sup_{i \in I} F(f_i)$.

We write $\mathbf{P}(X)$, $\nabla \mathbf{P}(X)$, $\Delta \mathbf{P}(X)$ respectively the spaces of all continuous previsions, of continuous lower previsions, of continuous upper previsions equipped with the Scott topology of the pointwise ordering \leq . The spaces $\mathbf{P}^*(X)$, $\nabla \mathbf{P}^*(X)$, $\Delta \mathbf{P}^*(X)$ will be the subspaces of those that are collinear.

We do not quite follow standard naming conventions. Standardly [27], a lower prevision is just a real-valued functional. *Coherent* lower previsions (taking a more readable definition from [13]) are those F such that $F(f) \geq \sum_{i=1}^n \lambda_i F(f_i) + \lambda_0$ whenever $f \geq \sum_{i=1}^n \lambda_i f_i + \lambda_0$, $\lambda_i > 0$, $\lambda_0 \in \mathbb{R}$. In our case, we reserve the “lower” adjective, so as to have a dual notion of *upper* prevision.

It is clear that any continuous game ν defines a continuous collinear prevision $\alpha_e(\nu)$. Moreover, if ν is convex, then $\alpha_e(\nu)$ is lower, and if ν is concave, then $\alpha_e(\nu)$ is upper. The following isomorphism result, akin to Riesz’ Representation Theorem, is known as Schmeidler’s Theorem for convex games on discrete topologies. Let $\gamma_e(F)$, for any prevision F , be the capacity ν such that $\nu(U) = F(\chi_U)$ for every open U of X . Order previsions pointwise, then:

Theorem 1. $\alpha_e \dashv \gamma_e$ is a Galois injection from (plain, convex, concave) games into (plain, lower, upper) collinear previsions. That is, α_e and γ_e are monotonic, $\alpha_e(\gamma_e(F)) \leq F$ for every collinear prevision F , and $\gamma_e(\alpha_e(\nu)) = \nu$ for every game ν .

Moreover, when restricted to continuous previsions and games, α_e and γ_e define an isomorphism between $\mathbf{J}(X)$ and $\mathbf{P}^*(X)$, between $\nabla \mathbf{J}(X)$ and $\nabla \mathbf{P}^*(X)$, between $\Delta \mathbf{J}(X)$ and $\Delta \mathbf{P}^*(X)$.

Proof. That $\gamma_e(F)$ is a game for any prevision is easy. When F is lower, note that $\chi_{U \cup V}$ and $\chi_{U \cap V}$ are comonotonic, and $\chi_{U \cup V} + \chi_{U \cap V} = \chi_U + \chi_V$. So $\gamma_e(F)(U \cup V) \geq \gamma_e(F)(U) + \gamma_e(F)(V)$.

$V) + \gamma_e(F)(U \cap V) = F(\chi_{U \cup V} + \chi_{U \cap V})$ (since F is collinear) $= F(\chi_U + \chi_V) \geq F(\chi_U) + F(\chi_V)$ (since F is super-additive) $= \gamma_e(F)(U) + \gamma_e(F)(V)$. Similarly, $\gamma_e(F)$ is concave if F is upper.

For the converse, we first show that: (A) for any collinear prevision F on X , for any step function f , written $a + \sum_{i=1}^m a_i \chi_{U_i}$ with $U_1 \supseteq \dots \supseteq U_m$, $a \in \mathbb{R}$, $a_1, \dots, a_m \in \mathbb{R}^+$, then the Choquet integral of f along $\gamma_e(F)$ equals $F(f)$. This is an easy exercise as soon as one realizes that $\sum_{i=0}^{k-1} a_i \chi_{U_i}$ and $a_k \chi_{U_k}$ are comonotonic for every k , $1 \leq k \leq m$. The equality $\gamma_e(\alpha_e(\nu))(U) = \nu(U)$ is obvious, α_e and γ_e are clearly monotonic. To show that $\alpha_e(\gamma_e(F)) \leq F$, we must show that the Choquet integral of f along $\gamma_e(F)$ is less than or equal to $F(f)$. Using the step functions f_K , $K \in \mathbb{N}$, by (A) the Choquet integral of f_K is less than or equal to $F(f_K)$. The least upper bound of the Choquet integrals of f_K , $K \in \mathbb{N}$ is that of f , and the least upper bound of $F(f_K)$ is at most $F(f)$. So $\alpha_e(\gamma_e(F))(f) \leq F(f)$. When F is continuous, the least upper bound of $F(f_K)$ is exactly $F(f)$, whence $\alpha_e(\gamma_e(F)) = F$. \square

One easy, well-known consequence of this is that α_e and γ_e define an order isomorphism between the space $\mathbf{V}(X)$ of continuous valuations and that $\mathbf{P}^\Delta(X)$ of continuous linear previsions ([8, Theorem 6.2], [22, Satz 4.16]). Intuitively, any continuous game ν gives rise to a continuous collinear prevision $\alpha_e(\nu)$ that computes a generalized form of expectation along ν , and every continuous collinear prevision arises this way.

It is easy to check that $\mathbf{J}, \nabla \mathbf{J}, \Delta \mathbf{J}, \mathbf{V}, \mathbf{P}^*, \nabla \mathbf{P}^*, \Delta \mathbf{P}^*, \mathbf{P}^\Delta$ define functors \mathbf{T} from \mathbf{Top} to \mathbf{Top} , where \mathbf{Top} is the category of topological spaces.

To define a monad structure on \mathbf{T} , we need a *unit*

$\eta_X : X \rightarrow \mathbf{T}X$, natural in X . This is defined by $\eta_X(x) = \delta_x$. However, there is general no extension $f^\dagger(\nu)(V) = \oint_{x \in X} f(x)(V) d\nu$ of $f : X \rightarrow \mathbf{T}Y$. The natural candidate is:

when \mathbf{T} is a game functor ($\mathbf{J}, \nabla \mathbf{J}, \Delta \mathbf{J}, \mathbf{V}$), or $f^\dagger(F)(h) = F(\lambda x \in X \cdot f(x)(h))$ when \mathbf{T} is a prevision functor ($\mathbf{P}^*, \nabla \mathbf{P}^*, \Delta \mathbf{P}^*, \mathbf{P}^\Delta$). While this indeed works when $\mathbf{T} = \mathbf{V}$ [8, Section 4.2], or when $\mathbf{T} = \mathbf{P}^\Delta$ using the isomorphism between \mathbf{V} and the latter, it fails for the other functors. To understand why, take $\mathbf{T} = \nabla \mathbf{P}^*$, and consider $X = \{1, 2\}$, $Y = \{*_11, *_12, *_21, *_22\}$ (with their discrete topologies), $F = \alpha_e \mathbf{u}_{\{1,2\}}$, i.e., $F(h) = \min(h(1), h(2))$ for every $h : Y \rightarrow \mathbb{R}^+$, $f : X \rightarrow \mathbf{T}Y$ defined by $f(1) = \alpha_e(3/4\delta_{*_11} + 1/4\delta_{*_12})$ and $f(2) = \alpha_e(1/3\delta_{*_21} + 2/3\delta_{*_22})$, so that $f(1)(h) = 3/4h(*_11) + 1/4h(*_12)$ and $f(2)(h) = 1/3h(*_21) + 2/3h(*_22)$ for every $h : Y \rightarrow \mathbb{R}^+$. Let h and h' be defined by: $h(*_11) = 0.3$, $h(*_12) = h(*_22) = 0.1$, $h(*_21) = 0.7$, $h'(*_11) = 0.5$, $h'(*_12) = h'(*_22) = 0$, $h'(*_21) = 0.7$, then $f^\dagger(F)(h) = 0.25$, $f^\dagger(F)(h') = 0.233\dots$, $f^\dagger(F)(h+h') = 0.533\dots$, but $f^\dagger(F)(h) + f^\dagger(F)(h') = 0.4833\dots \neq f^\dagger(F)(h+h')$, although h and h' are comonotonic. In other words, $_^\dagger$ does not preserve collinearity.

In everyday terms, collinear previsions, or more specifically belief functions represent a process where P draws at random first, then C chooses non-deterministically [6]. The example above is a non-deterministic choice (among $\{1, 2\}$) followed by probabilistic choices. In other words, the non-deterministic player C plays first, then only the probabilistic player P. But it is well-known that you cannot permute non-deterministic and probabilistic choices, and the example above only serves to restate this.

Our cure is simple: drop the collinearity condition. We shall therefore consider monads of continuous (plain, lower, upper) prevision. Let \mathbf{Posc} be the category of posets with Scott-continuous maps, \mathbf{Cpo} its full subcategory of cpos. We consider posets equipped with their Scott topology, whence these two categories are full subcategories of \mathbf{Top} . Note that $\mathbf{P}(X)$, $\nabla \mathbf{P}(X)$, $\Delta \mathbf{P}(X)$ are only posets, not cpos.

Theorem 2. *Define \mathbf{TX} as $\mathbf{P}(X)$, resp. $\nabla \mathbf{P}(X)$, resp. $\Delta \mathbf{P}(X)$. Let $\eta_X(x) = \lambda h \in \langle X \rightarrow \mathbb{R}^+ \rangle \cdot h(x)$, and $f^\dagger(F)(h) = F(\lambda x \in X \cdot f(x)(h))$ for every $f : X \rightarrow \mathbf{TY}$. Then \mathbf{T} is a monad on \mathbf{Top} , i.e., $(\mathbf{T}, \eta, _^\dagger)$ is a Kleisli triple. On \mathbf{Posc} , \mathbf{T} is a strong monad: $\mathbf{t}_{X,Y} : X \times \mathbf{TY} \rightarrow \mathbf{T}(X \times Y)$ defined as $\mathbf{t}_{X,Y}(x, F)(h) = F(\lambda y \in Y \cdot h(x, y))$ is a tensorial strength.*

Proof. We must first show that, for every $f : X \rightarrow \mathbf{TY}$, f^\dagger is indeed a continuous map from \mathbf{TX} to \mathbf{TY} . Foremost, we must make sure that for every continuous (plain, lower, upper) prevision F on X , $f^\dagger(F)$ is a continuous (plain, lower, upper) prevision on Y . This is easy, but relatively tedious verification. Now note that the formulae defining η , $_^\dagger$, \mathbf{t} are exactly the formulae defining the *continuation monad* [17]. It follows that the Kleisli triple axioms also hold in our case.

Contrarily to what might be expected, $\mathbf{t}_{X,Y}$ is not defined on all of \mathbf{Top} —it may fail to be continuous. On \mathbf{Posc} , this is repaired by the fact that a function of two arguments is continuous iff it is continuous in each argument separately (a fact that fails in \mathbf{Top}). The tensorial strength equations [17] are checked as for the continuation monad. \square

That the formulae for unit, extension, and tensorial strength are the same as for the continuation monad is no accident. Imagine $F \in \mathbf{TX}$ is the semantics of a (probabilistic and non-deterministic) program expected to return a result x of type X . As we have already argued, when $F = \alpha_e(P)$, with P a continuous valuation, then $F(h)$ is the *average* payoff, defined as the (Choquet) integral of $h(x)$ along P . When $F = \alpha_e(\nu)$ with ν a continuous belief function, then $F(h)$ is the average min-payoff, where minima are taken over (demonically) non-deterministic choices. When F is not collinear, then more complicated “averaging” processes are involved. In particular, we allow taking means of mins of means of mins. . . representing plays where P, C, P, C, . . . take turns. The fact that arbitrarily many turns can be chained in a (not necessarily collinear) prevision will be a consequence of the fact that prevision functors define monads, and in particular have a well-defined multiplication. This is standard in the monadic approach to side-effects [17]: multiplication is the key to defining sequential composition—here, of plays.

More explicitly, take n continuous functions $f_1 : X_0 \rightarrow \mathbf{TX}_1$, $f_2 : X_1 \rightarrow \mathbf{TX}_2$, . . . , $f_n : X_{n-1} \rightarrow \mathbf{TX}_n$. Then, when \mathbf{T} is a monad, $f_n^\dagger \circ f_{n-1}^\dagger \circ \dots \circ f_2^\dagger \circ f_1 : X_0 \rightarrow \mathbf{TX}_n$ is the sequential composition of $f_1, f_2, \dots, f_{n-1}, f_n$ in this order: given $x_0 \in X_0$, the process $f_1(x_0)$ computes some element $x_1 \in X_1$ (in our case, by drawing it “at random”, say; deterministic computations are of course allowed, too), then $f_2(x_1)$ computes some $x_2 \in X_2$, etc. The monad laws then typically say that composing with the idle process $\eta_X : X \rightarrow \mathbf{TX}$ does nothing, and that sequential composition is associative.

While Theorem 2 then establishes a form of soundness (which we shall make more precise below), the goal of the next sections will be to show that the prevision axioms are

complete, in the sense that there is no junk: every continuous (lower, upper) prevision is a mix of (demonic, angelic) non-deterministic and probabilistic choices.

One may wonder what the equivalent of *normalized* games ($\nu(X) = 1$) and *sub-normalized* games ($\nu(X) \leq 1$) would be through the correspondence of Theorem 1. Requiring $F(\chi_X)$ to equal (resp. less than or equal to) 1 is the obvious choice. However, this is not preserved by $_^\dagger$ when F is not collinear. So we define:

Definition 2. A prevision F on X is *normalized*, resp. *sub-normalized*, iff for every $f \in \langle X \rightarrow \mathbb{R}^+ \rangle$, for every $a \in \mathbb{R}^+$, $F(a+f) = a+F(f)$ (resp. $F(a+f) \leq a+F(f)$).

We let $\mathbf{J}_1(X)$, $\nabla \mathbf{P}_1^*(X)$, $\nabla \mathbf{P}_1(X)$, \dots , be the subspaces of normalized games/previsions, and $\mathbf{J}_{\leq 1}(X)$, $\nabla \mathbf{P}_{\leq 1}^*(X)$, $\nabla \mathbf{P}_{\leq 1}(X)$, \dots , those of sub-normalized games/previsions.

Proposition 1. Theorem 1 again holds for normalized (continuous) games and previsions, and for sub-normalized (continuous) games and previsions.

Now the spaces of sub-normalized and normalized continuous previsions are cpos. The spaces of sub-normalized continuous previsions are *pointed*, i.e., they have a least element \perp , the constant 0 function. If X is itself pointed, then the spaces of normalized continuous previsions are pointed, too, with least element $\alpha_{\mathbb{C}}(\delta_\perp)$ (a continuous linear prevision). The latter maps $h \in \langle X \rightarrow \mathbb{R}^+ \rangle$ to $h(\perp)$. Let \mathbf{Cpo} the category of cpos, \mathbf{Pcpo} that of pointed cpos. It follows:

Proposition 2. Let \mathbf{TX} be defined as $\mathbf{P}_{\leq 1}(X)$, $\nabla \mathbf{P}_{\leq 1}(X)$, $\Delta \mathbf{P}_{\leq 1}(X)$, $\mathbf{P}_1(X)$, $\nabla \mathbf{P}_1(X)$, or $\Delta \mathbf{P}_1(X)$. $(\mathbf{T}, \boldsymbol{\eta}, \boldsymbol{\mu}, \boldsymbol{\iota})$ is a strong monad on \mathbf{Cpo} and on \mathbf{Pcpo} .

Theorem 2 allows us to give a semantics to a λ -calculus with both probabilistic and non-deterministic choices. Consider the syntax of terms and types:

$M, N, P ::= x$	variable	
c	constant	
MN	application	
$\lambda x \cdot M$	abstraction	$\tau ::= \alpha$ base types
$()$	empty tuple	\mathbf{u} type of $()$
(M, N)	pair	$\tau \times \tau$ product
$\text{fst } M$	first projection	$\tau \rightarrow \tau$ function types
$\text{snd } M$	second projection	$T\tau$ computation types
$\text{val } M$	trivial computation	
$\text{let val } x = M \text{ in } N$	let-expression	

The typing rules, as well as the categorical semantics in a let-CCC, are standard [17]. Note that \mathbf{Cpo} and \mathbf{Pcpo} are Cartesian-closed. Together with the strong monads of Proposition 2, they form let-CCCs. The typing rules for computation types are: if $\Gamma \vdash M : \tau$ then $\Gamma \vdash \text{val } M : T\tau$; and if $\Gamma \vdash M : T\tau_1$ and $\Gamma, x : \tau_1 \vdash N : T\tau_2$ then $\Gamma \vdash \text{let val } x = M \text{ in } N : T\tau_2$.

As should be expected, the semantics has a strong continuation flavor. For each term M of type τ in context $\Gamma = x_1 : \tau_1, \dots, x_n : \tau_n$, $\llbracket M \rrbracket$ is a morphism (a continuous map) from $\llbracket \Gamma \rrbracket = \llbracket \tau_1 \rrbracket \times \dots \times \llbracket \tau_n \rrbracket$ to $\llbracket \tau \rrbracket$. The cases for val and let are given by: $\llbracket \text{val } M \rrbracket (v_1, \dots, v_n) = \lambda h \in \langle \llbracket \tau \rrbracket \rightarrow \mathbb{R}^+ \rangle \cdot h(\llbracket M \rrbracket (v_1, \dots, v_n))$, and $\llbracket \text{let val } x = M \text{ in } N \rrbracket (v_1, \dots, v_n) = \lambda h \in \langle \llbracket \tau_2 \rrbracket \rightarrow \mathbb{R}^+ \rangle \cdot \llbracket M \rrbracket (v_1, \dots, v_n)(\lambda v \in$

$\llbracket \tau_1 \rrbracket \cdot \llbracket N \rrbracket (v_1, \dots, v_n, v)(h)$. Let `bool` be a base type, with $\llbracket \text{bool} \rrbracket = \mathbb{S}$, where $\mathbb{S} = \{0, 1\}$ is Sierpiński space ($0 < 1$). Constants c may include a least fixpoint operator in **Pcpo**, the Boolean constants `false`, `true`, a case construct `case` : `bool` \times $\tau \times \tau \rightarrow \tau$ with $\llbracket \text{case} \rrbracket (0, v_0, v_1) = v_0$ and $\llbracket \text{case} \rrbracket (1, v_0, v_1) = v_1$. The interpretation of T as a monad of previsions allows us, additionally, to give meaning to a coin-flipping operator `flip` : $T\text{bool}$, with $\llbracket \text{flip} \rrbracket = \alpha_e(1/2\delta_0 + 1/2\delta_1) = \lambda h \in \langle \mathbb{S} \rightarrow \mathbb{R}^+ \rangle \cdot 1/2(h(0) + h(1))$, and a non-deterministic choice operator `amb` : $T\text{bool}$. When T is $\nabla \mathbf{P}_1$, `amb` is the demonic choice (of a Boolean): $\llbracket \text{amb} \rrbracket = \alpha_e(u_{\{0,1\}}) = \lambda h \in \langle \mathbb{S} \rightarrow \mathbb{R}^+ \rangle \cdot \min(h(0), h(1))$ (the chosen Boolean x is the one that minimizes payoff $h(x)$). When T is $\Delta \mathbf{P}_1$, we get angelic choice: $\llbracket \text{amb} \rrbracket = \alpha_e(u_{\{0,1\}}) = \lambda h \in \langle \mathbb{S} \rightarrow \mathbb{R}^+ \rangle \cdot \max(h(0), h(1))$ (maximize payoff).

One might think that letting T be \mathbf{P}_1 would lead to chaotic choice. This certainly accommodates both demonic (min) and angelic choice (max). However, \mathbf{P}_1 is a very large space, and seems to contain objects that do not correspond to any mixture of probabilistic and non-deterministic choice. The right notion is suggested by [7, section 7.5].

Definition 3 (Fork). A fork on X is any pair $F = (F^-, F^+)$ where F^- is a lower prevision, F^+ is an upper prevision, and for any $h, h' \in \langle X \rightarrow \mathbb{R}^+ \rangle$,

$$F^-(h + h') \leq F^-(h) + F^+(h') \leq F^+(h + h') \quad (3)$$

F is continuous, resp. normalized, sub-normalized, collinear, whenever both F^- and F^+ are.

While the above definition was found from purely mathematical arguments, Walley [27, Section 2] defines essentially the same notion in finance. However, we allow any pair (F^-, F^+) satisfying these conditions to be a fork. Walley only observes that whenever F^- is a coherent prevision (in his sense), on a discrete space, then letting $F^+(h) = -F^-(-h)$ yields a fork (F^-, F^+) .

One may think of F^- as the pessimistic part of F , which will give us the least expected payoff, while F^+ is the optimistic part, yielding the greatest expected payoff. Taking $h' = 0$ in (3) shows indeed that $F^-(h) \leq F^+(h)$ for each h . Let $\mathbf{F}(X)$ be the space of continuous forks on X , ordered by $\leq \times \leq$. The subspaces $\mathbf{F}_1(X)$ and $\mathbf{F}_{\leq 1}(X)$ of normalized and sub-normalized forks are cpos. The latter is pointed (with least element $(0, 0)$) and the former is as soon as X is (with least element $(\alpha_e(\delta_\perp), \alpha_e(\delta_\perp))$). The semantics is essentially the pairing of two continuation semantics, e.g., $\llbracket \text{val } M \rrbracket (v_1, \dots, v_n) = (F^-, F^+)$, where $F^- = F^+ = \lambda h \in \langle \llbracket \tau \rrbracket \rightarrow \mathbb{R}^+ \rangle \cdot h(\llbracket M \rrbracket (v_1, \dots, v_n))$ (a linear prevision); $\llbracket \text{let val } x = M \text{ in } N \rrbracket (v_1, \dots, v_n) = (\lambda h \in \langle \llbracket \tau_2 \rrbracket \rightarrow \mathbb{R}^+ \rangle \cdot F^-(\lambda v \in \llbracket \tau_1 \rrbracket \cdot F_v^-(h)), \lambda h \in \langle \llbracket \tau_2 \rrbracket \rightarrow \mathbb{R}^+ \rangle \cdot F^+(\lambda v \in \llbracket \tau_1 \rrbracket \cdot F_v^+(h)))$, where $(F^-, F^+) = \llbracket M \rrbracket (v_1, \dots, v_n)$ and $(F_v^-, F_v^+) = \llbracket N \rrbracket (v_1, \dots, v_n, v)$. The constants with the most interesting semantics are `amb`, where $\llbracket \text{amb} \rrbracket = (\lambda h \in \langle \mathbb{S} \rightarrow \mathbb{R}^+ \rangle \cdot \min(h(0), h(1)), \lambda h \in \langle \mathbb{S} \rightarrow \mathbb{R}^+ \rangle \cdot \max(h(0), h(1)))$ (i.e., it computes both pessimistic and optimistic outcomes), and `flip`, where $\llbracket \text{flip} \rrbracket = (F^-, F^+)$ with $F^- = F^+ = \lambda h \in \langle \mathbb{S} \rightarrow \mathbb{R}^+ \rangle \cdot 1/2(h(0) + h(1))$. For the rest of the language, we rely on [17] and:

Proposition 3. Let $\mathbf{T}X$ be defined as $\mathbf{F}(X)$, $\mathbf{F}_{\leq 1}(X)$, or $\mathbf{F}_1(X)$. Let $\eta_X(x) = (F^-, F^+)$ with $F^- = F^+ = \lambda h \in \langle X \rightarrow \mathbb{R}^+ \rangle \cdot h(x)$, and for every $f : X \rightarrow \mathbf{T}Y$, let $f^\dagger(F^-, F^+) = (\lambda h \in \langle Y \rightarrow \mathbb{R}^+ \rangle \cdot F^-(\lambda x \in X \cdot f^-(x)(h)), \lambda h \in \langle Y \rightarrow \mathbb{R}^+ \rangle \cdot$

$F^+(\lambda x \in X \cdot f^+(x)(h))$), where by convention $f(x) = (f^-(x), f^+(x))$. Then $(\mathbf{T}, \boldsymbol{\eta}, \boldsymbol{\mu})$ is a monad on \mathbf{Top} . Together with $\mathbf{t}_{X,Y} : X \times \mathbf{T}Y \rightarrow \mathbf{T}(X \times Y)$ defined by $\mathbf{t}_{X,Y}(x, (F^-, F^+)) = (\lambda h \in \langle Y \rightarrow \mathbb{R}^+ \rangle \cdot F^-(\lambda y \in Y \cdot h(x, y)), \lambda h \in \langle Y \rightarrow \mathbb{R}^+ \rangle \cdot F^+(\lambda y \in Y \cdot h(x, y)))$, it forms a strong monad on \mathbf{Cpo} and \mathbf{Pcpo} .

Proof. That the strong monad laws are satisfied is obvious. The core of the proof is in showing that unit, extension, and tensorial strength are well-defined. We deal with extension. Recall that $f^\dagger(F^-, F^+) = (F'^-, F'^+)$, where $F'^- = \lambda h \in \langle Y \rightarrow \mathbb{R}^+ \rangle \cdot F^-(\lambda x \in X \cdot f^-(x)(h))$ and $F'^+ = \lambda h \in \langle Y \rightarrow \mathbb{R}^+ \rangle \cdot F^+(\lambda x \in X \cdot f^+(x)(h))$. Then $F'^-(h + h') = \lambda h \in \langle Y \rightarrow \mathbb{R}^+ \rangle \cdot F^-(\lambda x \in X \cdot f^-(x)(h + h')) \leq \lambda h \in \langle Y \rightarrow \mathbb{R}^+ \rangle \cdot F^-(\lambda x \in X \cdot f^-(x)(h) + f^-(x)(h'))$ (since $f(x) = (f^-(x), f^+(x)) \in \mathbf{T}Y$ and F^- is monotonic) $\leq \lambda h \in \langle Y \rightarrow \mathbb{R}^+ \rangle \cdot F^-(\lambda x \in X \cdot f^-(x)(h)) + F^-(\lambda x \in X \cdot f^-(x)(h'))$ (since $(F^-, F^+) \in \mathbf{T}X = F'^-(h) + F'^+(h')$). Similarly, $F'^-(h) + F'^+(h') \leq F'^+(h + h')$. \square

5 Hearts and Skins

One of the fundamental theorems of the theory of cooperative games is Shapley's Theorem, which states that every convex game ν has a non-empty core (on finite discrete X)—the core $\text{Core}(\nu)$ being the set of measures p such that $\nu \leq p$ and $\nu(X) = p(X)$. A refinement of this is Rosenmuller's Theorem, which states that a game ν is convex iff its core is non-empty and for every function $f : X \rightarrow \mathbb{R}^+$, the integral of f along ν is the minimum of all integrals of f along p , $p \in \text{Core}(\nu)$. In particular, there is a measure p such that $\nu \leq p$, $\nu(X) = p(X)$, and integrating f along p gives the same result as integrating it along ν [5]².

We show that the same results hold in the continuous case in [7, chapitre 10]. Remember that games correspond to collinear previsions. Our purpose here is to show that similar theorems hold on previsions that need not be collinear (see [7, chapitre 11] for a more complete development). The analogue of measures will be linear previsions. We drop the analogue of the $\nu(X) = p(X)$ condition, however we concentrate on normalized games and previsions, because the technical treatment is slightly easier. We call the analogue of cores hearts, and the dual notion skin.

Definition 4 (Heart, Skin). For any function F from $\langle X \rightarrow \mathbb{R}^+ \rangle$ to \mathbb{R}^+ , its heart $\text{Coeur}(F)$ is the set of linear functionals G such that $F \leq G$. Its continuous heart $\text{CCoeur}(F)$ is the subset of those G s that are continuous. Its skin $\text{Peau}(F)$ is the set of linear functionals G such that $G \leq F$. Its continuous skin $\text{CPeau}(F)$ is the subset of those functionals G that are continuous.

² An anonymous referee for a previous version of this paper asked whether this had anything to do with a theorem due to Shannon, stating that for any distribution p and function f , there is another distribution p' such that f has the same mean relative to p and p' , and which maximizes entropy. While there is a similar flavor to it, I must confess that I don't see any relationship. Moreover, Shannon's Theorem, contrarily to Rosenmuller's, does not extend to the continuous case, if only because entropy is only defined on finite spaces (relative entropy is another matter). I won't include this footnote in the final version of this paper.

Again, we let $\text{Coeur}_1(F)$, $\text{CCoeur}_1(F)$, \dots , be the subsets of the corresponding spaces consisting of normalized previsions only, and similarly $\text{Coeur}_{\leq 1}(F)$, \dots , for those consisting of sub-normalized previsions.

Most of the developments below rest on Roth's Sandwich Theorem ([21], [24, Theorem 3.1]), which states that on every ordered cone C , for every positively homogeneous super-additive function $q : C \rightarrow \overline{\mathbb{R}}^+$ and every positively homogeneous sub-additive function $p : C \rightarrow \overline{\mathbb{R}}^+$ such that $a \leq b$ implies $q(a) \leq p(b)$ (e.g., when $q \leq p$ and either q or p is monotonic), then there is a monotonic linear function $f : C \rightarrow \overline{\mathbb{R}}^+$ such that $q \leq f \leq p$. $\overline{\mathbb{R}}^+$ is \mathbb{R}^+ plus an extra point at infinity $+\infty$. A *cone* is a set C , together with a binary operation $+$ turning it into a commutative monoid and a scalar multiplication \cdot from $\mathbb{R}^+ \times C$ to C , such that $1 \cdot a = a$, $0 \cdot a = 0$, $(rs) \cdot a = r \cdot (s \cdot a)$, $r \cdot (a + b) = r \cdot a + r \cdot b$, and $(r + s) \cdot a = r \cdot a + s \cdot a$. An *ordered cone* is equipped in addition with a partial ordering \leq making $+$ and \cdot monotonic. We only use Roth's Theorem on ordered cones of the form $\langle X \rightarrow \mathbb{R}^+ \rangle$. Our key result is:

Theorem 3. *Let X be a stably locally compact space, F a continuous lower prevision, and f a bounded continuous function from X to \mathbb{R}^+ . Then there is a continuous linear functional G from $\langle X \rightarrow \mathbb{R}^+ \rangle$ to $\overline{\mathbb{R}}^+$ such that $F \leq G$ and $F(f) = G(f)$.*

Proof. Let F be a lower prevision on X , and $f \in \langle X \rightarrow \mathbb{R}^+ \rangle$. Define \widetilde{F}_f by $\widetilde{F}_f(g) = \inf_{\substack{\lambda \in \mathbb{R}^+ \\ \lambda f \geq g}} \left[F(\lambda f) - \sup_{\substack{h \in \langle X \rightarrow \mathbb{R}^+ \rangle \\ g+h \leq \lambda f}} F(h) \right]$, taking this to be $+\infty$ if there is no $\lambda \in \mathbb{R}^+$ such that $\lambda f \geq g$. One checks that \widetilde{F}_f is monotonic, positively homogeneous, sub-additive, above F ($\widetilde{F}_f(g) \geq F(g)$ for all g), touches F at f ($\widetilde{F}_f(f) = F(f)$). Apply Roth's Sandwich Theorem gives us a monotonic linear functional G_0 such that $F \leq G_0$ and $F(f) = G_0(f)$. However, G_0 may fail to be continuous. One now observes that $\langle X \rightarrow \mathbb{R}^+ \rangle$ is a continuous poset, with a basis B consisting of step functions. By Scott's Formula, the functional G defined by $G(f) = \sup_{g \in B, g \ll f} G_0(g)$ is continuous; in fact, the largest continuous functional below G_0 . It follows that $F \leq G$ and $F(f) = G(f)$. The most difficult part of the proof is showing that G is linear. This rests on the fact that \ll is multiplicative i.e., for any $a > 0$, $f \ll g$ iff $a \cdot f \ll a \cdot g$, and additive, i.e., if $h, f, g \in \langle X \rightarrow \mathbb{R}^+ \rangle$ are such that $h \ll f + g$, then $h \leq f' + g'$ for some $f', g' \in B$ with $f' \ll f$, $g' \ll g$; and conversely, $f' \ll f$ and $g' \ll g$ imply $f' + g' \ll f + g$. \square

Note that G may take the value $+\infty$. We can refine this in the case of normalized previsions (for sub-normalized previsions, see [7, section 11.4]):

Theorem 4. *Let X be a stably locally compact space, F a normalized continuous lower prevision on X , and f a bounded continuous function from X to \mathbb{R}^+ . Then there is a normalized continuous linear prevision G such that $F \leq G$ and $F(f) = G(f)$.*

Proof. Similar to Theorem 3. However, it may be that \widetilde{F}_f reaches $+\infty$. Refine this by letting $\widetilde{\widetilde{F}}_f(g) = \inf_{\epsilon \in \mathbb{R}^+} \widetilde{F}_{f+\epsilon}(g)$, and using $\widetilde{\widetilde{F}}_f$ instead of \widetilde{F}_f . One checks that, since F is normalized, $\widetilde{\widetilde{F}}_{f+\epsilon}$ is antitone in ϵ . Then $\widetilde{\widetilde{F}}_f$ is again monotonic, positively homogeneous, sub-additive (using antitony in ϵ), above F , and touches F at f . Moreover,

it is easy to see that $\check{F}_f(\chi_X) = 1$. We build G_0 , then G from \check{F}_f , as in Theorem 3. Additionally, we need X to be compact so as to establish that $G(\chi_X) = 1$. Since G is linear, it follows that G is normalized. \square

One can deal with upper previsions instead, see [7, section 11.5], using a notion we call convex-concave duality to reduce to the above. We then obtain [7, théorème 11.5.22] that, when X is stably compact, F is a normalized continuous upper prevision on X , there is a normalized continuous linear prevision G on X such that $G \leq F$. Moreover, for every $f \in \langle X \rightarrow \mathbb{R}^+ \rangle$, $F(f) = \sup_{G \in CPeau_1(F)} G(f)$.

Theorem 4 allows us to state a form of Rosenmuller's Theorem:

Theorem 5. *Let X be stably locally compact, F a continuous normalized prevision on X . Then F is lower iff $CCoeur_1(F) \neq \emptyset$ and for every $f \in \langle X \rightarrow \mathbb{R}^+ \rangle$, $F(f) = \inf_{G \in CCoeur_1(F)} G(f)$. In this case, the inf is attained: $F(f) = \min_{G \in CCoeur_1(F)} G(f)$.*

There is, of course, a dual theorem on upper previsions and their skins [7, théorème 11.7.4]; infs are replaced by sups, which need not be attained.

To go further, we need to consider another topology on spaces of previsions: the *weak topology* is the coarsest that makes the function $F \mapsto F(f)$ continuous, for each $f \in \langle X \rightarrow \mathbb{R}^+ \rangle$. The Scott topology is in general finer. Write $\nabla \mathbf{P}_{1\ wk}(X)$ the space $\nabla \mathbf{P}_1(X)$ with the weak topology, and similarly for other spaces. Then:

Proposition 4. *Let X be stably compact, F a normalized continuous lower prevision, then $CCoeur_1(F)$ is a non-empty saturated compact convex subset of $\mathbf{P}_{1\ wk}^\Delta(X)$.*

Compactness can be deduced from Plotkin's version of the Banach-Alaoglu Theorem [19], while convexity (i.e., $\alpha F + (1 - \alpha)F'$ is in $CCoeur_1(F)$ as soon as F and F' are, $\alpha \in [0, 1]$) is clear. It is much easier to show that the continuous skin $CPeau_1(F)$ of a normalized continuous upper prevision F is closed:

Proposition 5. *Let X be a topological space, F a normalized continuous upper prevision, then $CPeau_1(F)$ is a closed convex subset of $\mathbf{P}_{1\ wk}^\Delta(X)$. It is non-empty as soon as X is stably compact.*

Finally, call a *lens* of a space X any non-empty intersection $L = Q \cap F$ of a saturated compact Q and a closed subset F . Then:

Proposition 6. *Let X be a stably compact space. The continuous normalized body $CCorps_1(F) = CCoeur_1(F^-) \cap CPeau_1(F^+)$ of a continuous normalized fork $F = (F^-, F^+)$ on X is a lens. Moreover, $CCoeur_1(F^-) = \uparrow CCorps_1(F)$ and $CPeau_1(F^+) = \downarrow CCorps_1(F)$.*

Proof. We show that: (*) whenever $G \in CCoeur_1(F^-)$, there is some $G' \in CCoeur_1(F^-) \cap CPeau_1(F^+)$ such that $G' \leq G$. Let $F'(h) = \inf_{\substack{f, g \in \langle X \rightarrow \mathbb{R}^+ \rangle \\ f+g \geq h}} (F^+(f) + G(g))$. One

checks that $F^- \leq F' \leq G$, that F' is an upper prevision, so by Roth's Sandwich Theorem, there is a linear monotonic functional G_0 such that $F^- \leq G_0 \leq F'$. Since $G_0 \leq F'$, G_0 does not take the value $+\infty$. Build G from G_0 using Scott's Formula, as before. It is easy to see that G is a continuous, normalized, linear prevision. Since

$F^- \leq G', G' \in CCoeur_1(F^-)$. Since $G' \leq F' \leq F^+$, $G' \in CPeau_1(F^+)$. Since $F' \leq F' \leq G, G' \leq G$.

By (*), $CCoeur_1(F^-) \cap CPeau_1(F^+)$ is non-empty. That $CCoeur_1(F^-) = \uparrow(CCoeur_1(F^-) \cap CPeau_1(F^+))$ is another easy consequence of (*). That $CPeau_1(F^+) = \downarrow(CCoeur_1(F^-) \cap CPeau_1(F^+))$ can be shown in a similar way, by defining $F''(h) = \sup_{\substack{f, g \in (X \rightarrow \mathbb{R}^+) \\ f+g \leq h}} (F^-(f) + G(g))$, where $G \in CPeau_1(F^-)$, and using F'' to show that there is some $G' \in CCoeur_1(F^-) \cap CPeau_1(F^+)$ such that $G \leq G'$. \square

The last three propositions state that any normalized continuous lower prevision, resp. upper prevision, resp. fork F gives rise to an element $CCoeur_1(F)$, resp. $CPeau_1(F)$, resp. $CCorps_1(F)$ of the Smyth powerdomain $\mathcal{Q}(\mathbf{P}_{1\ wk}^\Delta(X))$ (demonic non-deterministic choice of a probability distribution—remember that $\mathbf{P}_1^\Delta(X) \cong \mathbf{V}_1(X)$), resp. the Hoare powerdomain $\mathcal{H}(\mathbf{P}_{1\ wk}^\Delta(X))$ over $\mathbf{P}_{1\ wk}^\Delta(X)$ (angelic), resp. the Plotkin powerdomain over $\mathbf{P}_{1\ wk}^\Delta(X)$ (chaotic). This is a form of *completeness*: our spaces of previsions and of forks contain no junk, and really are no more than mixes of non-deterministic and probabilistic choice.

In the converse direction, still assuming X stably compact, there is a map $\sqcap : \mathcal{Q}(\mathbf{P}_{1\ wk}^\Delta(X)) \rightarrow \nabla \mathbf{P}_1(X)$ defined by $\sqcap K(f) = \min_{G \in K} G(f)$, and $CCoeur_1 \dashv \sqcap$ is a Galois injection consisting of continuous maps [7, théorème 11.7.10], while there is a continuous map $\sqcup : \mathcal{H}(\mathbf{P}_{1\ wk}^\Delta(X)) \rightarrow \nabla \mathbf{P}_1(X)$ defined by $\sqcup C(f) = \sup_{G \in C} G(f)$, so that $\sqcup \dashv CPeau_1$ is a Galois surjection.

We conclude by noticing that, when X is a continuous cpo with a least element, $\mathbf{P}_{1\ wk}^\Delta(X)$ is homeomorphic to $\mathbf{V}_1(X)$ with the weak topology, and the latter coincides then with the Scott topology [9]. Apart from spurious details (e.g., we bound our valuations by 1 instead of $+\infty$), there is therefore a strong connection with the models of Mislove [16] and Tix [23, 24]. The question whether the Galois connections above can be turned into isomorphisms remains open.

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A Proofs of Theorems

Theorem 1. $\alpha_{\mathcal{E}} \dashv \gamma_{\mathcal{E}}$ is a Galois injection from (plain, convex, concave) games into (plain, lower, upper) collinear previsions. That is, $\alpha_{\mathcal{E}}$ and $\gamma_{\mathcal{E}}$ are monotonic, $\alpha_{\mathcal{E}}(\gamma_{\mathcal{E}}(F)) \leq F$ for every collinear prevision F , and $\gamma_{\mathcal{E}}(\alpha_{\mathcal{E}}(\nu)) = \nu$ for every game ν .

Moreover, when restricted to continuous previsions and games, $\alpha_{\mathcal{E}}$ and $\gamma_{\mathcal{E}}$ define an isomorphism between $\mathbf{J}(X)$ and $\mathbf{P}^*(X)$, between $\nabla \mathbf{J}(X)$ and $\nabla \mathbf{P}^*(X)$, between $\Delta \mathbf{J}(X)$ and $\Delta \mathbf{P}^*(X)$.

Proof. Let F be a prevision. Then $\gamma_{\mathcal{E}}(F)(\emptyset) = 0$ because F is positively homogeneous (take $\alpha = 0$); $\gamma_{\mathcal{E}}(F)$ is monotonic because F is. So $\gamma_{\mathcal{E}}(F)$ is a game.

If F is a lower prevision, then notice that $\chi_{U \cup V}$ and $\chi_{U \cap V}$ are comonotonic: assume $\chi_{U \cup V}(x) < \chi_{U \cup V}(x')$ and $\chi_{U \cap V}(x) > \chi_{U \cap V}(x')$, then necessarily $x \notin U \cup V$ and $x \in U \cap V$, which is impossible. Next, note that $\chi_{U \cup V} + \chi_{U \cap V} = \chi_U + \chi_V$. So $\gamma_{\mathcal{E}}(F)(U \cup V) + \gamma_{\mathcal{E}}(F)(U \cap V) = F(\chi_{U \cup V} + \chi_{U \cap V})$ (since F is collinear) = $F(\chi_U + \chi_V) \geq F(\chi_U) + F(\chi_V)$ (since F is super-additive) = $\gamma_{\mathcal{E}}(F)(U) + \gamma_{\mathcal{E}}(F)(V)$. Similarly, $\gamma_{\mathcal{E}}(F)$ is concave if F is upper.

In the sequel, we shall need the following claim.

Claim A. Let F be a collinear prevision on X , and $f = a + \sum_{i=1}^m a_i \chi_{U_i}$ a step function, with $U_1 \supseteq \dots \supseteq U_m$, $a \in \mathbb{R}$, $a_1, \dots, a_m \in \mathbb{R}^+$. Then:

$$\oint_{x \in X} f(x) d\gamma_{\mathcal{E}}(F) = F(f)$$

Proof. Let $U_0 = X$ and $a_0 = a$, to make notation uniform. Then:

$$\begin{aligned} \oint_{x \in X} f(x) d\gamma_{\mathcal{E}}(F) &= a\gamma_{\mathcal{E}}(F)(X) + \sum_{i=1}^m a_i \gamma_{\mathcal{E}}(F)(U_i) \\ &= aF(\chi_X) + \sum_{i=1}^m a_i F(\chi_{U_i}) = F(a\chi_X) + \sum_{i=1}^m F(a_i \chi_{U_i}) \\ &= \sum_{i=0}^m F(a_i \chi_{U_i}) \end{aligned}$$

Note that $U_0 \supseteq U_1 \supseteq \dots \supseteq U_m$. Wlog., assume that $a_1, \dots, a_m > 0$. For each k , $1 \leq k \leq m$, the functions $\sum_{i=0}^{k-1} a_i \chi_{U_i}$ and $a_k \chi_{U_k}$ are comonotonic. Indeed, assume $(\sum_{i=0}^{k-1} a_i \chi_{U_i})(x) > (\sum_{i=0}^{k-1} a_i \chi_{U_i})(x')$ and $a_k \chi_{U_k}(x) < a_k \chi_{U_k}(x')$. The latter implies that $x \notin U_k$, and $x' \in U_k$. Since $U_0 \supseteq U_1 \supseteq \dots \supseteq U_m$, x' is in every U_i , $0 \leq i \leq k-1$, so $(\sum_{i=0}^{k-1} a_i \chi_{U_i})(x') = \sum_{i=0}^{k-1} a_i \geq (\sum_{i=0}^{k-1} a_i \chi_{U_i})(x)$, a contradiction.

Since F is collinear, it follows:

$$\sum_{i=0}^m F(a_i \chi_{U_i}) = F\left(\sum_{i=0}^{k-1} a_i \chi_{U_i}\right) + \sum_{i=k}^m F(a_i \chi_{U_i})$$

for every k , $1 \leq k \leq m + 1$, by induction on k . So, for $k = m + 1$,

$$\oint_{x \in X} f(x) d\gamma_e(F) = F\left(\sum_{i=0}^m a_i \chi_{U_i}\right) = F(f)$$

□

We now show that $\gamma_e(\alpha_e(\nu)) = \nu$ for every game ν :

$$\gamma_e(\alpha_e(\nu))(U) = \alpha_e(\nu)(\chi_U) = \oint_{x \in X} \chi_U(x) d\nu = \nu(U)$$

Once we know $\alpha_e \dashv \gamma_e$ is a Galois connection, this will imply it is a Galois *injection*. To show that $\alpha_e \dashv \gamma_e$ is a Galois connection, it remains to show that α_e is monotonic (which is clear), that γ_e is monotonic (clear since previsions are monotonic), and that $\alpha_e(\gamma_e(F)) \leq F$ for every collinear prevision F . First,

$$\alpha_e(\gamma_e(F))(f) = \oint_{x \in X} f(x) d\gamma_e(F)$$

Using the step functions f_K , $K \in \mathbb{N}$, by Claim A:

$$\oint_{x \in X} f_K(x) d\gamma_e(F) = F(f_K) \tag{4}$$

The least upper bound of the left-hand side is the Choquet integral of f along $\gamma_e(F)$, i.e., $\alpha_e(\gamma_e(F))(f)$. Since F is monotonic, the right-hand side is less than or equal to $F(f)$, so $\alpha_e(\gamma_e(F))(f) \leq F(f)$.

Let us turn to continuous games and continuous previsions. If ν is a continuous game, then $\alpha_e(\nu)$ is continuous, since Choquet integration is Scott-continuous in its function argument. Conversely, if F is continuous, then for every directed family $(U_i)_{i \in I}$ of opens, $\gamma_e(F)(\bigcup_{i \in I} U_i) = F(\chi_{\bigcup_{i \in I} U_i}) = F(\sup_{i \in I} \chi_{U_i}) = \sup_{i \in I} F(\chi_{U_i}) = \sup_{i \in I} \gamma_e(F)(U_i)$, so $\gamma_e(F)$ is continuous. Now if F is continuous, then the least upper bound of the right-hand side of (4) is $F(f)$, since $f = \sup_{K \in \mathbb{N}} f_K$ and F is continuous; while the left-hand side is the Choquet integral of f along $\gamma_e(F)$, i.e., $\alpha_e(\gamma_e(F))(f)$. So $\alpha_e(\gamma_e(F))(f) = F(f)$, whence the isomorphism. □

Theorem 2. Define $\mathbf{T}X$ as $\mathbf{P}(X)$, resp. $\nabla \mathbf{P}(X)$, resp. $\Delta \mathbf{P}(X)$. Let $\eta_X(x) = \lambda h \in \langle X \rightarrow \mathbb{R}^+ \rangle \cdot h(x)$, and $f^\dagger(F)(h) = F(\lambda x \in X \cdot f(x)(h))$ for every $f : X \rightarrow \mathbf{T}Y$. Then \mathbf{T} is a monad on \mathbf{Top} , i.e., $(\mathbf{T}, \eta, \dagger)$ is a Kleisli triple. On \mathbf{Posc} , \mathbf{T} is a strong monad: $\mathbf{t}_{X,Y} : X \times \mathbf{T}Y \rightarrow \mathbf{T}(X \times Y)$ defined as $\mathbf{t}_{X,Y}(x, F)(h) = F(\lambda y \in Y \cdot h(x, y))$ is a tensorial strength.

Proof. We must first show that, for every $f : X \rightarrow \mathbf{T}Y$, f^\dagger is indeed a continuous map from $\mathbf{T}X$ to $\mathbf{T}Y$. Foremost, we must make sure that for every continuous (plain, lower, upper) prevision F on X , $f^\dagger(F)$ is a continuous (plain, lower, upper) prevision on Y . Positive homogeneity: $f^\dagger(F)(\alpha h) = F(\lambda x \in X \cdot f(x)(\alpha h)) = F(\lambda x \in X \cdot \alpha f(x)(h))$ (since $f(x) \in \mathbf{T}Y$ is positively homogeneous) = $F(\alpha \lambda x \in X \cdot f(x)(h)) = \alpha F(\lambda x \in X \cdot f(x)(h))$ (since $F \in \mathbf{T}X$ is positively homogeneous) = $\alpha f^\dagger(F)(h)$. Monotonicity:

assume $h \leq h'$, then for each $x \in X$, $f(x)(h) \leq f(x)(h')$ since $f(x) \in \mathbf{TY}$ is monotonic, so $f^\dagger(F)(h) = F(\lambda x \in X \cdot f(x)(h)) \leq F(\lambda x \in X \cdot f(x)(h')) = f^\dagger(F)(h')$, since $F \in \mathbf{TX}$ is monotonic. In the case $\mathbf{T} = \nabla \mathbf{P}$, F and every $f(x)$ is super-additive, so $f^\dagger(F)(h + h') = F(\lambda x \in X \cdot f(x)(h + h')) \geq F(\lambda x \in X \cdot f(x)(h) + f(x)(h')) \geq F(\lambda x \in X \cdot f(x)(h)) + F(\lambda x \in X \cdot f(x)(h')) = f^\dagger(F)(h) + f^\dagger(F)(h')$, so $f^\dagger F$ is super-additive, too. Similarly when $\mathbf{T} = \triangle \mathbf{P}$. Continuity: let $(h_i)_{i \in I}$ be a directed family of bounded continuous functions from Y to \mathbb{R}^+ with h as least upper bound. Then $f^\dagger(F)(\sup_{i \in I} h_i) = F(\lambda x \in X \cdot f(x)(\sup_{i \in I} h_i)) = F(\lambda x \in X \cdot \sup_{i \in I} f(x)(h_i))$ (since $f(x) \in \mathbf{TY}$ is continuous) $= \sup_{i \in I} F(\lambda x \in X \cdot f(x)(h_i))$ (since $F \in \mathbf{TX}$ is continuous).

Then we must show that f^\dagger is continuous. Since \mathbf{TX} and \mathbf{TY} are posets with the Scott topology, it is enough to show that for any directed family $(F_i)_{i \in I}$ in \mathbf{TX} , $f^\dagger(\sup_{i \in I} F_i) = \sup_{i \in I} f^\dagger(F_i)$. But this is obvious from the definition.

We now check the Kleisli triple axioms. This is in fact automatic, since \mathbf{T} is defined as a continuation-style monad. (1) $\eta_X^\dagger(F)(h) = F(\lambda x \in X \cdot \eta_X(x)(h)) = F(\lambda x \in X \cdot h(x)) = F(h)$, so $\eta_X^\dagger = \text{id}_X$. (2) Let $f : X \rightarrow \mathbf{TY}$, then $(f^\dagger \circ \eta_X)(x)(h) = f^\dagger(\eta_X(x))(h) = \eta_X(x)(\lambda x' \in X \cdot f(x')(h)) = f(x)(h)$, so $f^\dagger \circ \eta_X = f$. (3) Let $g : X \rightarrow \mathbf{TY}$, $f : Y \rightarrow \mathbf{TZ}$. On the one hand, $(f^\dagger \circ g^\dagger)(F)(h) = f^\dagger(g^\dagger(F))(h) = g^\dagger(F)(\lambda y \in Y \cdot f(y)(h)) = F(\lambda x \in X \cdot g(x)(\lambda y \in Y \cdot f(y)(h)))$. On the other hand, $(f^\dagger \circ g)^\dagger(F)(h) = F(\lambda x \in X \cdot (f^\dagger \circ g)(x)(h)) = F(\lambda x \in X \cdot f^\dagger(g(x))(h)) = F(\lambda x \in X \cdot g(x)(\lambda y \in Y \cdot f(y)(h)))$, whence $f^\dagger \circ g^\dagger = (f^\dagger \circ g)^\dagger$.

Contrarily to what might be expected, $\mathbf{t}_{X,Y}$ is not defined on all of \mathbf{Top} . The reason is that it may fail to be continuous. On \mathbf{Posc} , this is repaired by the fact that a function of two arguments is continuous iff it is continuous in each argument separately (a fact that fails in \mathbf{Top}). Let us be more precise. Let h be any bounded continuous function from $X \times Y$ to \mathbb{R}^+ . For any fixed x , the function $\lambda y \in Y \cdot h(x, y)$ is bounded and continuous, so $F(\lambda y \in Y \cdot h(x, y))$ makes sense. It is clear that the function $\mathbf{t}_{X,Y}(x, F)$ mapping $h \in \langle X \times Y \rightarrow \mathbb{R}^+ \rangle$ to $F(\lambda y \in Y \cdot h(x, y))$ is a continuous (plain, lower, upper) prevision, since F is. Now $\mathbf{t}_{X,Y}(x, F)$ is obviously Scott-continuous in F (x fixed), and also in $x \in X$ (F fixed), since h is continuous, and F is continuous. So $\mathbf{t}_{X,Y}$ is a morphism in \mathbf{Posc} .

We need to check the tensorial strength equations [17]. These are in fact obvious, as again $\mathbf{t}_{X,Y}$ is defined exactly as for the continuation monad. \square

Proposition 1. *Theorem 1 again holds for normalized (continuous) games and previsions, and for sub-normalized (continuous) games and previsions.*

Proof. It is enough to show that α_e maps normalized (resp. sub-normalized) games to normalized (resp. sub-normalized) previsions, and that γ_e goes the other way around. The essential point is that: (*) Choquet integration is linear on comonotonic functions, and any constant a is comonotonic with any function f . Now if ν is normalized, i.e.,

$\nu(X) = 1$, then:

$$\begin{aligned}\alpha_{\mathcal{C}}(\nu)(a + f) &= \oint_{x \in X} a + f(x) d\nu \\ &= \oint_{x \in X} a d\nu + \oint_{x \in X} f(x) d\nu \quad \text{by } (*) \\ &= a + \alpha_{\mathcal{C}}(\nu)\end{aligned}$$

When ν is sub-normalized, the last line is an inequality \leq instead.

Conversely, if F is a normalized prevision, then $\gamma_{\mathcal{C}}(F)(X) = F(\chi_X) = F(1 + f)$ where f is the zero function. Since F is normalized, this equals $1 + F(f) = 1$, since $F(f) = 0$ by positive homogeneity with $\alpha = 0$. Similarly, if F is sub-normalized, then $\gamma_{\mathcal{C}}(F)(X) = F(1 + f) \leq 1 + F(f) = 1$. \square

Proposition 3. Let \mathbf{TX} be defined as $\mathbf{F}(X)$, $\mathbf{F}_{\leq 1}(X)$, or $\mathbf{F}_1(X)$. Let $\eta_X(x) = (F^-, F^+)$ with $F^- = F^+ = \lambda h \in \langle X \rightarrow \mathbb{R}^+ \rangle \cdot h(x)$, and for every $f : X \rightarrow \mathbf{TY}$, let $f^\dagger(F^-, F^+) = (\lambda h \in \langle Y \rightarrow \mathbb{R}^+ \rangle \cdot F^-(\lambda x \in X \cdot f^-(x)(h)), \lambda h \in \langle Y \rightarrow \mathbb{R}^+ \rangle \cdot F^+(\lambda x \in X \cdot f^+(x)(h)))$, where by convention $f(x) = (f^-(x), f^+(x))$. Then (\mathbf{T}, η, μ) is a monad on \mathbf{Top} . Together with $\mathbf{t}_{X,Y} : X \times \mathbf{TY} \rightarrow \mathbf{T}(X \times Y)$ defined by $\mathbf{t}_{X,Y}(x, (F^-, F^+)) = (\lambda h \in \langle Y \rightarrow \mathbb{R}^+ \rangle \cdot F^-(\lambda y \in Y \cdot h(x, y)), \lambda h \in \langle Y \rightarrow \mathbb{R}^+ \rangle \cdot F^+(\lambda y \in Y \cdot h(x, y)))$, it forms a strong monad on \mathbf{Cppo} and \mathbf{Pcpo} .

Proof. That the strong monad laws are satisfied is obvious: this is just the product of two strong monads as described in Theorem 2. The only thing to check is that unit, extension, and tensorial strength are well defined, i.e., (3) holds for objects meant to be in some space \mathbf{TZ} .

Unit. Let $F^- = F^+ = \lambda h \in \langle X \rightarrow \mathbb{R}^+ \rangle \cdot h(x) = \alpha_{\mathcal{C}}(\delta_x)$. This is a linear prevision, hence (3) is trivial.

Extension. Recall that $f^\dagger(F^-, F^+) = (F'^-, F'^+)$, where $F'^- = \lambda h \in \langle Y \rightarrow \mathbb{R}^+ \rangle \cdot F^-(\lambda x \in X \cdot f^-(x)(h))$ and $F'^+ = \lambda h \in \langle Y \rightarrow \mathbb{R}^+ \rangle \cdot F^+(\lambda x \in X \cdot f^+(x)(h))$. Then:

$$\begin{aligned}F'^-(h + h') &= \lambda h \in \langle Y \rightarrow \mathbb{R}^+ \rangle \cdot F^-(\lambda x \in X \cdot f^-(x)(h + h')) \\ &\leq \lambda h \in \langle Y \rightarrow \mathbb{R}^+ \rangle \cdot F^-(\lambda x \in X \cdot f^-(x)(h) + f^-(x)(h')) \\ &\quad \text{since } f(x) = (f^-(x), f^+(x)) \in \mathbf{TY} \text{ and } F^- \text{ is monotonic} \\ &\leq \lambda h \in \langle Y \rightarrow \mathbb{R}^+ \rangle \cdot F^-(\lambda x \in X \cdot f^-(x)(h)) + F^+(\lambda x \in X \cdot f^+(x)(h')) \\ &\quad \text{since } (F^-, F^+) \in \mathbf{TX} \\ &= F'^-(h) + F'^+(h')\end{aligned}$$

We show similarly that $F'^-(h) + F'^+(h') \leq F'^+(h + h')$.

Tensorial strength. Recall that $\mathbf{t}_{X,Y}(x, (F^-, F^+)) = (F'^-, F'^+)$ where $F'^* = \lambda h \in \langle Y \rightarrow \mathbb{R}^+ \rangle \cdot F^*(\lambda y \in Y \cdot h(x, y))$ (* being - or +). So:

$$\begin{aligned}F'^-(h + h') &= F^-(\lambda y \in Y \cdot h(x, y) + h'(x, y)) \\ &\leq F^-(\lambda y \in Y \cdot h(x, y)) + F^+(\lambda y \in Y \cdot h'(x, y)) = F'^-(h) + F'^+(h')\end{aligned}$$

since $(F^-, F^+) \in \mathbf{T}(X \times Y)$; and similarly $F'^-(h) + F'^+(h') \leq F'^+(h + h')$. \square

Theorem 3. Let X be a stably locally compact space, F a continuous lower prevision, and f a bounded continuous function from X to \mathbb{R}^+ . Then there is a continuous linear functional G from $\langle X \rightarrow \mathbb{R}^+ \rangle$ to $\overline{\mathbb{R}}^+$ such that $F \leq G$ and $F(f) = G(f)$.

Proof. Recall that a *cone* is a set C with two binary operations $+$: $C \times C \rightarrow C$ and \cdot : $\mathbb{R}^+ \times C \rightarrow C$, and a constant $0 \in C$ such that $(C, +, 0)$ is a commutative monoid, and \cdot defines an action of (\mathbb{R}^+, \times) of \mathbb{R}^+ on C such that additionally $(r + s) \cdot a = r \cdot a + s \cdot a$. An *ordered cone* is a cone with a partial ordering \leq such that $+$ and \cdot are monotonic in all their arguments. \mathbb{R}^+ , and $\overline{\mathbb{R}}^+ = \mathbb{R}^+ \cup \{+\infty\}$ are ordered cones. It is clear that $\langle X \rightarrow \mathbb{R}^+ \rangle$ is an ordered cone, too. A function $p : C \rightarrow \overline{\mathbb{R}}^+$ is *sub-linear* iff $p(r \cdot a) = rp(a)$ for every $r \in \mathbb{R}^+$ and $p(a + b) \leq p(a) + p(b)$, for all $a, b \in C$. It is *super-linear* iff $p(a + b) \geq p(a) + p(b)$ instead, and that is *linear* if equality holds.

Let F be a lower prevision on X , and $f \in \langle X \rightarrow \mathbb{R}^+ \rangle$. Let, for every $g \in \langle X \rightarrow \mathbb{R}^+ \rangle$:

$$\widetilde{F}_f(g) = \inf_{\substack{\lambda \in \mathbb{R}^+ \\ \lambda f \geq g}} \left[F(\lambda f) - \sup_{\substack{h \in \langle X \rightarrow \mathbb{R}^+ \rangle \\ g+h \leq \lambda f}} F(h) \right]$$

where we take the convention that this is equal to $+\infty$ if there is no $\lambda \in \mathbb{R}^+$ such that $\lambda f \geq g$.

We shall abbreviate this as $\widetilde{F}_f(g) = \inf_{\lambda/\lambda f \geq g} [F(\lambda f) - \sup_{h \leq \lambda f - g} F(h)]$. Note that we cannot in general write $F(\lambda f - g)$ instead of $\sup_{h \leq \lambda f - g} F(h)$, since $\lambda f - g$ is not in general continuous from X to \mathbb{R}^+ (with its Scott topology).

Claim B. $\widetilde{F}_f(0) = 0$.

Proof.

$$\widetilde{F}_f(0) = \inf_{\lambda \in \mathbb{R}^+} \left[F(\lambda f) - \sup_{h \leq \lambda f} F(h) \right] = \inf_{\lambda \in \mathbb{R}^+} [F(\lambda f) - F(\lambda f)] = 0$$

□

Claim C. \widetilde{F}_f is monotonic.

Proof. Let $g, g' \in \langle X \rightarrow \mathbb{R}^+ \rangle$, with $g \leq g'$. Fix $\lambda \in \mathbb{R}^+$ such that $\lambda f \geq g'$. For every $h' \leq \lambda f - g'$ in $\langle X \rightarrow \mathbb{R}^+ \rangle$, there is an $h \leq \lambda f - g$ in $\langle X \rightarrow \mathbb{R}^+ \rangle$ such that $F(h) \geq F(h')$, namely h' itself. So $\sup_{h \leq \lambda f - g} F(h) \geq \sup_{h' \leq \lambda f - g'} F(h')$. So $F(\lambda f) - \sup_{h \leq \lambda f - g} F(h) \leq F(\lambda f) - \sup_{h' \leq \lambda f - g'} F(h')$. By making λ vary,

$$\inf_{\lambda/\lambda f \geq g'} \left[F(\lambda f) - \sup_{h \leq \lambda f - g} F(h) \right] \leq \inf_{\lambda/\lambda f \geq g'} \left[F(\lambda f) - \sup_{h' \leq \lambda f - g'} F(h') \right]$$

Since $\lambda f \geq g'$ implies $\lambda f \geq g$, the left-hand side of the above inequality is at least:

$$\inf_{\lambda/\lambda f \geq g} \left[F(\lambda f) - \sup_{h \leq \lambda f - g} F(h) \right]$$

that is, $\widetilde{F}_f(g)$. The right-hand side is by definition $\widetilde{F}_f(g')$, so $\widetilde{F}_f(g) \leq \widetilde{F}_f(g')$. □

Claim D. Let F be a lower prevision on X . \widetilde{F}_f is convex: for every real α , $0 \leq \alpha \leq 1$, for every $g \in \langle X \rightarrow \mathbb{R}^+ \rangle$,

$$\widetilde{F}_f(\alpha g + (1 - \alpha)g') \leq \alpha \widetilde{F}_f(g) + (1 - \alpha) \widetilde{F}_f(g') \quad (5)$$

Proof. The inequality is clear if $\alpha = 0$ or $\alpha = 1$. So assume $0 < \alpha < 1$. If there is no $\lambda \in \mathbb{R}^+$ such that $\lambda f \geq g$, or if there is no $\lambda' \in \mathbb{R}^+$ such that $\lambda' f \geq g'$, the right-hand side of (5) is $+\infty$, so the inequality is vacuously true. So let us assume that for some $\lambda \in \mathbb{R}^+$, $\lambda f \geq g$ and for some $\lambda' \in \mathbb{R}^+$, $\lambda' f \geq g'$. Let us fix λ and λ' for now. To ease reading, define the following abbreviations: $g'' = \alpha g + (1 - \alpha)g'$, and let $\lambda'' = \alpha\lambda + (1 - \alpha)\lambda'$. For every $h \leq \lambda f - g$ in $\langle X \rightarrow \mathbb{R}^+ \rangle$, for every $h' \leq \lambda' f - g'$ in $\langle X \rightarrow \mathbb{R}^+ \rangle$, let $h'' = \alpha h + (1 - \alpha)h'$. The map h'' is continuous, since α and $1 - \alpha$ are non-negative, multiplication by non-negative reals is Scott-continuous, and addition is also Scott-continuous. Also, $h'' \leq \lambda'' f - g''$. Finally, since F is a lower prevision, F is concave, i.e., $\alpha F(h) + (1 - \alpha)F(h') \leq F(h'')$. We have just shown that for every $h \leq \lambda f - g$ et $h' \leq \lambda' f - g'$, there is an $h'' \leq \lambda'' f - g''$ such that $\alpha F(h) + (1 - \alpha)F(h') \leq F(h'')$. It follows:

$$\sup_{h'' \leq \lambda'' f - g''} F(h'') \geq \alpha \sup_{h \leq \lambda f - g} F(h) + (1 - \alpha) \sup_{h' \leq \lambda' f - g'} F(h')$$

Since $\lambda'' = \alpha\lambda + (1 - \alpha)\lambda'$,

$$\begin{aligned} F(\lambda'' f) - \sup_{h'' \leq \lambda'' f - g''} F(h'') &\leq \alpha \left[F(\lambda f) - \sup_{h \leq \lambda f - g} F(h) \right] \\ &\quad + (1 - \alpha) \left[F(\lambda' f) - \sup_{h' \leq \lambda' f - g'} F(h') \right] \end{aligned}$$

By making λ and λ' , we obtain:

$$\begin{aligned} &\inf_{\substack{\lambda, \lambda' \in \mathbb{R}^+ \\ \lambda f \geq g, \lambda' f \geq g' \\ \lambda'' = \alpha\lambda + (1 - \alpha)\lambda'}} \left[F(\lambda'' f) - \sup_{h'' \leq \lambda'' f - g''} F(h'') \right] \quad (6) \\ &\leq \inf_{\substack{\lambda, \lambda' \in \mathbb{R}^+ \\ \lambda f \geq g, \lambda' f \geq g'}} \left[\alpha \left[F(\lambda f) - \sup_{h \leq \lambda f - g} F(h) \right] + (1 - \alpha) \left[F(\lambda' f) - \sup_{h' \leq \lambda' f - g'} F(h') \right] \right] \end{aligned}$$

Clearly $\lambda'' \in \mathbb{R}^+$ and $\lambda'' f \geq g''$. Recall that there exist $\lambda \in \mathbb{R}^+$ such that $\lambda f \geq g$ and $\lambda' \in \mathbb{R}^+$ such that $\lambda' f \geq g'$. The right hand side of (6) therefore equals $\alpha \widetilde{F}_f(g) + (1 - \alpha) \widetilde{F}_f(g')$. For the left hand side, observe that for every $\lambda, \lambda' \in \mathbb{R}^+$ such that $\lambda f \geq g, \lambda' f \geq g'$, the quantity $\lambda'' = \alpha\lambda + (1 - \alpha)\lambda'$ is such that $\lambda'' \in \mathbb{R}^+$ and $\lambda'' f \geq g''$: the left hand side of (6) is in particular greater than or equal to the greatest lower bound, over all $\lambda'' \in \mathbb{R}^+$ such that $\lambda'' f \geq g''$, of $F(\lambda'' f) - \sup_{h'' \leq \lambda'' f - g''} F(h'')$. But this greatest lower bound is exactly $\widetilde{F}_f(g'')$. So $\widetilde{F}_f(g'') \leq \alpha \widetilde{F}_f(g) + (1 - \alpha) \widetilde{F}_f(g')$. \square

Claim E. Let F be a lower prevision on X . \widetilde{F}_f is positively homogeneous: for every $\alpha \geq 0$, $\widetilde{F}_f(\alpha g) = \alpha \widetilde{F}_f(g)$.

Proof. When $\alpha = 0$, this is by Claim B. For all $\alpha > 0$,

$$\begin{aligned} \widetilde{F}_f(\alpha g) &= \inf_{\lambda \in \mathbb{R}^+, \lambda f \geq \alpha g} \left[F(\lambda f) - \sup_{h \leq \lambda f - \alpha g} F(h) \right] \\ &= \inf_{\lambda' \in \mathbb{R}^+, \lambda' f \geq g} \left[F(\alpha \lambda' f) - \sup_{h \leq \alpha \lambda' f - \alpha g} F(h) \right] \quad (\text{where } \lambda' = \lambda/\alpha) \\ &= \inf_{\lambda' \in \mathbb{R}^+, \lambda' f \geq g} \left[F(\alpha \lambda' f) - \sup_{h' \leq \lambda' f - g} F(\alpha h') \right] \quad (\text{where } h' = h/\alpha) \\ &= \inf_{\lambda' \in \mathbb{R}^+, \lambda' f \geq g} \left[\alpha F(\lambda' f) - \sup_{h' \leq \lambda' f - g} \alpha F(h') \right] \end{aligned}$$

since F is positively homogeneous. But this is exactly $\alpha \widetilde{F}_f(g)$. \square

Claim F. Let F be a lower prevision on X . For every $g \in \langle X \rightarrow \mathbb{R}^+ \rangle$, $\widetilde{F}_f(g) \geq F(g)$.

Proof. For every $\lambda \in \mathbb{R}^+$ such that $\lambda f \geq g$, for every $h \leq \lambda f - g$, $F(\lambda f) \geq F(h) + F(g)$. Indeed, $F(h) + F(g) \leq F(h + g)$ since F is super-additive, and $F(h + g) \leq F(\lambda f)$ by assumption.

So $F(\lambda f) \geq \sup_{h \leq \lambda f - g} F(h) + F(g)$, i.e., $F(\lambda f) - \sup_{h \leq \lambda f - g} F(h) \geq F(g)$. We conclude by taking greatest lower bounds over all the $\lambda \in \mathbb{R}^+$ such that $\lambda f \geq g$. \square

Claim G. Let F be a lower prevision on X . Then $\widetilde{F}_f(f) = F(f)$.

Proof. If f is the 0 function, then $\widetilde{F}_f(f) = F(f) = 0$. Otherwise, the smallest $\lambda \in \mathbb{R}^+$ such that $\lambda f \geq f$ is 1, so $\widetilde{F}_f(f) = \inf_{\lambda/\lambda f \geq f} [F(\lambda f) - \sup_{h \leq \lambda f - f} F(h)] = F(1 \cdot f) - F(0) = F(f)$. \square

Although \widetilde{F}_f may take have $+\infty$ as value, it is not the case if f is bounded away from 0 from below:

Claim H. Let F be a lower prevision on X . $\widetilde{F}_f(\chi_X) \geq F(\chi_X)$. Moreover, if $\inf_{x \in X} f(x) > 0$, then $\widetilde{F}_f(\chi_X) \leq \frac{1}{\inf_{x \in X} f(x)} F(f)$.

Proof.

$$\begin{aligned} \widetilde{F}_f(\chi_X) &= \inf_{\lambda/\lambda f \geq \chi_X} \left[F(\lambda f) - \sup_{h \leq \lambda f - \chi_X} F(h) \right] \\ &= \inf_{\lambda/\lambda f \geq \chi_X} [F(\lambda f) - F(\lambda f - \chi_X)] \end{aligned}$$

We observe indeed that $h = \lambda f - \chi_X \equiv \lambda f - 1$ is continuous. Since F is concave, $F(\lambda f) \geq F(\lambda f - \chi_X) + F(\chi_X)$, so $\widetilde{F}_f(\chi_X) \geq F(\chi_X)$.

As far as the second inequality is concerned, since $F(\lambda f - \chi_X) \geq 0$, we obtain $\widetilde{F}_f(\chi_X) \leq \inf_{\lambda/\lambda f \geq \chi_X} F(\lambda f)$. Let $a_0 = \inf_{x \in X} f(x)$, and note that $\lambda f \geq \chi_X$ iff $\lambda \geq 1/a_0$. So $\widetilde{F}_f(\chi_X) \leq 1/a_0 F(f)$. \square

We can then use Roth's Sandwich Theorem to conclude that there is a linear G such that $F \leq G$ and satisfying some added conditions. For now, G may take the value $+\infty$, and is not necessarily continuous. In a picturesque way, such a G will be called an *upper tangent*.

Claim I. *Let F be a lower prevision on X , and f a bounded continuous function from X to \mathbb{R}^+ . There is a functional G from $\langle X \rightarrow \mathbb{R}^+ \rangle$ to $\overline{\mathbb{R}}^+$, which is linear and monotonic, and such that $F \leq G$ and $F(f) = G(f)$. Furthermore, if $\inf_{x \in X} f(x) \neq 0$, then G is a linear prevision.*

Proof. By Claim E and Claim D, \widetilde{F}_f is sub-linear. By assumption, F is super-linear. Moreover, F and \widetilde{F}_f are monotonic and $F \leq \widetilde{F}_f$ by Claim F. By Roth's Sandwich Theorem, there is a functional G from $\langle X \rightarrow \mathbb{R}^+ \rangle$ to $\overline{\mathbb{R}}^+$, which is monotonic and linear, and such that $F \leq G \leq \widetilde{F}_f$. By Claim G, $F(f) \leq G(f) \leq \widetilde{F}_f(f) = F(f)$, so $G(f) = F(f)$.

If moreover $\inf_{x \in X} f(x) \neq 0$, by Claim H, $G(\chi_X) \leq \widetilde{F}_f(\chi_X) \leq 1/\inf_{x \in X} f(x)F(f) < +\infty$. So for every $g \in \langle X \rightarrow \mathbb{R}^+ \rangle$, $G(g) \leq G(\sup_{x \in X} g(x)\chi_X) < +\infty$. Since G takes its values in \mathbb{R}^+ , no longer in $\overline{\mathbb{R}}^+$, G is a linear prevision. \square

To attack *continuous* previsions, we first need to explore the structure of the space $\langle X \rightarrow \mathbb{R}^+ \rangle$. This is probably well-known. In doubt, I preferred to produce proofs. Let \Subset be the way-below relation on the set $\mathcal{O}(X)$ of opens of X , ordered by inclusion. X is a *core compact* space iff $\mathcal{O}(X)$ is a continuous cpo [2]. This is in particular the case when X is locally compact, where $U \Subset V$ iff $U \subseteq Q \subseteq V$ for some saturated compact subset Q of X .

Claim J. *Let \ll be the way-below relation of $\langle X \rightarrow \mathbb{R}^+ \rangle$, and \ll_1 that of $\langle X \rightarrow [0, 1] \rangle$ of all continuous functions f bounded from above by 1.*

Let $f = \sum_{i=1}^n a_i \chi_{U_i}$ a step function from X to \mathbb{R}^+ (resp. $[0, 1]$), $U_1 \supseteq \dots \supseteq U_n$, $a_1, \dots, a_n \in \mathbb{R}^+ \setminus \{0\}$. Let g a continuous function from X to \mathbb{R}^+ (resp. $[0, 1]$). Then $f \ll g$ (resp. $f \ll_1 g$) iff for every i , $1 \leq i \leq n$, $U_i \Subset g^{-1}(\sum_{j=1}^i a_j, +\infty)$.

Proof. **Step 1.** The condition is necessary. Let us indeed fix i , $1 \leq i \leq n$, let t_i be $\sum_{j=1}^i a_j$, and consider any directed family $(V_k)_{k \in I}$ of opens such that $g^{-1}(t_i, +\infty) \subseteq \bigcup_{k \in I} V_k$. Let $W_k = V_k \cap g^{-1}(t_i, +\infty)$. For each $k \in I$ and each real r , $0 < r < 1$, let $f_{r,k}$ be the function $r \cdot (\max(\min(t_i, g), \chi_{W_k} \cdot g))$. This is continuous, as a composition of continuous functions. Note in particular that \max , \min and multiplication by a non-negative scalar are Scott-continuous. The only thing to check really is that $\chi_{W_k} \cdot g$ is continuous: the inverse image of $(t, +\infty)$ is X if $t < 0$, $W_k \cap g^{-1}(t, +\infty)$ otherwise, which is indeed open. Note also that, in case we consider functions in $\langle X \rightarrow [0, 1] \rangle$, and $f \ll_1 g$, then $f_{r,k}$ is again in $\langle X \rightarrow [0, 1] \rangle$. If $x \in W_k \subseteq g^{-1}(t_i, +\infty)$, $f_{r,k}(x) = r \cdot g(x)$; if $x \in g^{-1}(t_i, +\infty) \setminus W_k$, $f_{r,k}(x) = r \cdot t_i$; if $x \notin g^{-1}(t_i, +\infty)$ (in particular $x \notin W_k$), $f_{r,k}(x) = r \cdot g(x)$.

Note that if $r \leq r'$ and $V_k \subseteq V_{k'}$ (so $W_k \subseteq W_{k'}$), then $f_{r,k} \leq f_{r',k'}$: if $x \in W_k$, then $f_{r,k}(x) = r.g(x) \leq r'.g(x) = f_{r',k'}(x)$, since $x \in W_{k'}$; if $x \in W_{k'} \setminus W_k$, then $x \in g^{-1}]t_i, +\infty[\setminus W_k$, so $f_{r,k}(x) = r.t_i < r'.g(x) = f_{r',k'}(x)$; if $x \in g^{-1}(t_i, +\infty) \setminus W_{k'}$, $f_{r,k}(x) = r.t_i \leq r'.t_i = f_{r',k'}(x)$; and is $x \notin g^{-1}(t_i, +\infty)$, then $f_{r,k}(x) = r.g(x) \leq r'.g(x) = f_{r',k'}(x)$. It follows that the family $(f_{r,k})_{\substack{0 < r < 1 \\ k \in I}}$ is directed, since $f_{r,k}$ and $f_{r',k'}$ are both less than or equal to $f_{\max(r,r'),k''}$, where k'' is such that $V_k, V_{k'} \subseteq V_{k''}$.

Moreover, the least upper bound of this family is exactly g : for every $x \in X$, either $x \in g^{-1}(t_i, +\infty) = \bigcup_{k \in I} W_k$, so there is $k \in I$ with $x \in W_k$, and then $f_{r,k}(x) = r.g(x)$; or $x \notin g^{-1}(t_i, +\infty)$ and then $f_{r,k}(x) = r.g(x)$ again; but then $\sup_{0 < r < 1} r.g(x) = g(x)$.

If $f \ll g$, necessarily $f \leq f_{r,k}$ for some $r, 0 < r < 1$, and some $k \in I$. Then, for every $x \in U_i$, $f(x) \geq t_i$, so $f_{r,k}(x) \geq t_i$. By definition of $f_{r,k}$, this entails $\max(\min(t_i, g(x)), \chi_{W_k}(x).g(x)) \geq t_i/r$. Since $\min(t_i, g(x)) \leq t_i < t_i/r$, necessarily $\chi_{W_k}(x).g(x) \geq t_i/r$, so $x \in W_k$ and $g(x) \geq t_i/r$.

In particular, $U_i \subseteq W_k \subseteq V_k$. Since the family $(V_k)_{k \in I}$ is arbitrary, $U_i \in g^{-1}(t_i, +\infty)$.

Step 2. Conversely, the condition is sufficient. Indeed, assume that for every $i, 1 \leq i \leq n$, $U_i \in g^{-1}(t_i, +\infty)$, where, as above, $t_i = \sum_{j=1}^{i-1} a_j$.

Let us show that $f \ll g$ (resp. $f \ll_1 g$). Let $(f_k)_{k \in I}$ a directed family of continuous functions from X to \mathbb{R}^+ (resp. $[0, 1]$) such that $g \leq \sup_{k \in I} f_k$.

For every $t \in \mathbb{R}^+$, $g^{-1}(t, +\infty) = \{x \in X \mid g(x) > t\} \subseteq \{x \in X \mid \sup_{k \in I} f_k(x) > t\} = \{x \in X \mid \exists k \in I. f_k(x) > t\} = \bigcup_{k \in I} f_k^{-1}(t, +\infty)$. The family $(f_k^{-1}(t, +\infty))_{k \in I}$ is directed, since $(f_k)_{k \in I}$ is directed and $f_k \leq f_{k'}$ implies $f_k^{-1}(t, +\infty) \subseteq f_{k'}^{-1}(t, +\infty)$.

Since $U_i \in g^{-1}(t_i, +\infty) \subseteq \bigcup_{k \in I} f_k^{-1}(t_i, +\infty)$, there is a $k_i \in I$ such that $U_i \subseteq f_{k_i}^{-1}(t_i, +\infty)$. Since $(f_k)_{k \in I}$ is directed, there is a $k \in I$ such that $f_{k_i} \leq f_k$ for every $i, 1 \leq i \leq n$. So $U_i \subseteq f_{k_i}^{-1}(t_i, +\infty) \subseteq f_k^{-1}(t_i, +\infty)$.

Let $U_0 = X, U_{n+1} = \emptyset$. For every $x \in X$, let i be the unique natural number between 0 and n such that $x \in U_i \setminus U_{i+1}$. If $i = 0$, then $f(x) = 0$, so $f(x) \leq f_k(x)$.

If $i \neq 0$, then $f(x) = \sum_{j=1}^i a_j = t_i$, and since $x \in U_i, x \in f_k^{-1}(t_i, +\infty)$, i.e., $f_k(x) > t_i$. In other words, $f_k(x) > f(x)$. Since x is arbitrary, $f_k \geq f$. So $f \ll g$. \square

Recall that a basis B of a continuous poset Y is any subset of Y such that every element of x can be written as a directed least upper bound of elements in B .

Claim K. Let X be a core compact space. Let B the set of step functions of the form $1/2^K \sum_{k=1}^N \chi_{U_k}$, $K, N \in \mathbb{N}$, where $(U_k)_{k=1}^{2^K}$ is a decreasing sequence of opens of X .

Then $\langle X \rightarrow \mathbb{R}^+ \rangle$ is a continuous poset, with basis B . The space $\langle X \rightarrow [0, 1] \rangle$ is a continuous cpo, with basis $B_1 = B \cap \langle X \rightarrow [0, 1] \rangle$.

Proof. Let $f \in \langle X \rightarrow \mathbb{R}^+ \rangle$, resp. $f \in \langle X \rightarrow [0, 1] \rangle$. Consider the set D of all functions of the form $1/2^K \sum_{k=1}^N \chi_{U_k}$, $K, N \in \mathbb{N}$, where $U_k \in f^{-1}(k/2^K, +\infty)$ for every $k, 1 \leq k \leq N$. Clearly $D \subseteq B$, resp. $D \subseteq B_1$.

We first show that D is directed. D is non-empty: take $N = 0$. Then, if $1/2^K \sum_{k=1}^N \chi_{U_k}$ and $1/2^{K'} \sum_{k=1}^{N'} \chi_{V_k}$ are in D , we may assume wlog. that $K = K'$ and $N = N'$.

Let us show that we may require $K = K'$: if, say, $K' < K$, one can rewrite the second function as $1/2^K \sum_{k=1}^{2^{K-K'}N'} \chi_{W_k}$, where $W_k = V_{\lfloor (k+2^{K-K'}-1)/2^{K-K'} \rfloor}$ for every k , $1 \leq k \leq 2^{K-K'}N'$. It follows from $K = K'$ that we may also require $N = N'$: if $N' < N$, for example, then $1/2^{K'} \sum_{k=1}^{N'} \chi_{V_k} = 1/2^{K'} \sum_{k=1}^N \chi_{V_k}$, where we let $V_k = \emptyset$ for each k , $N' < k \leq N$. Given any two functions of the form $1/2^K \sum_{k=1}^N \chi_{U_k}$ and $1/2^K \sum_{k=1}^N \chi_{V_k}$ in D , then $1/2^K \sum_{k=1}^N \chi_{U_k \cup V_k}$ is again in D , since $U_k \cup V_k \in f^{-1}(k/2^K, +\infty)$. (In any core compact space, $U \cup V \in W$ as soon as $U \in W$ and $V \in W$ [2].) So D is directed.

Now f is the least upper bound of the sequence $f_K = 1/2^K \sum_{k=1}^N \chi_{f^{-1}(k/2^K, +\infty]}$, $K \in \mathbb{N}$, $N = \lfloor a2^K \rfloor$, where $a = \sup_{x \in X} f(x)$. Since $\mathcal{O}(X)$ is a continuous cpo, $f^{-1}(k/2^K, +\infty)$ is the directed union of all opens $U \in f^{-1}(k/2^K, +\infty)$, so $\chi_{f^{-1}(k/2^K, +\infty)} = \sup_{U \in f^{-1}(k/2^K, +\infty)} \chi_U$. So f_K is a least upper bound of elements in D . Since $f = \sup_{K \in \mathbb{N}} f_K$, f is also a directed least upper bound of elements in D . By Claim J, every element in D is way-below f . \square

In particular, whenever $f \ll g$ then there is a step function $f' \in B$ such that $f \ll f' \ll g$. This is the *interpolation property*, valid in any continuous poset [15].

The following states that \ll is *multiplicative*:

Claim L. *Let f be any step function from X to \mathbb{R}^+ , g a continuous function from X to \mathbb{R}^+ , and $a > 0$ a real such that $a \geq \sup_{x \in X} f(x)$, $a \geq \sup_{x \in X} g(x)$. Then $f \ll g$ iff $f/a \ll_1 g/a$.*

Proof. Let $f = \sum_{i=1}^n a_i \chi_{U_i}$. Then $f \ll g$ iff for every i , $1 \leq i \leq n$, $U_i \in g^{-1}(\sum_{j=1}^i a_j, +\infty)$ by Claim J. Again by Claim J, $f/a \ll_1 g/a$ iff for every i , $1 \leq i \leq n$, $U_i \in (g/a)^{-1}(\sum_{j=1}^i a_i/a, +\infty)$. Now note that $(g/a)^{-1}(\sum_{j=1}^i a_i/a, +\infty) = g^{-1}(\sum_{j=1}^i a_i, +\infty)$. So $f \ll g$ is equivalent to $f/a \ll_1 g/a$. \square

Let us say that X is *core coherent* iff for all opens U, V_1, V_2 of X , if $U \in V_1$ and $U \in V_2$ then $U \in V_1 \cap V_2$. We call X *stably core compact* iff it is both core compact and core coherent. Every stably locally compact space X is stably core compact, and the converse holds if X is sober.

The following states that \ll is *additive*, provided X is stably core compact.

Claim M. *Let X be core compact, and f, g two bounded continuous functions from X to \mathbb{R}^+ . If $h \ll f + g$, then for some $f', g' \in B$ we have $h \leq f' + g'$, $f' \ll f$ and $g' \ll g$.*

Proof. Let B_f the set of all functions $f' \in B$ such that $f' \leq f$, B_g that of all functions $g' \in B$ such that $g' \leq g$. By Claim K, $f = \sup_{f' \in B_f} f'$, $g = \sup_{g' \in B_g} g'$. Since addition is Scott-continuous, $f + g = \sup_{f' \in B_f, g' \in B_g} f' + g'$. Moreover, B_f and B_g are directed, so $B_f \times B_g$ as well. Since $h \ll f + g$, there are $f' \in B_f$ et $g' \in B_g$ such that $h \leq f' + g'$. \square

We now need an explicit formula for sums of functions in B .

Claim N. *Let $f' = 1/2^K \sum_{i=1}^N \chi_{A_i}$, $g' = 1/2^K \sum_{j=1}^{N'} \chi_{B_j}$ two functions from X to \mathbb{R} , with $X \supseteq A_1 \supseteq A_2 \supseteq \dots \supseteq A_N$ and $X \supseteq B_1 \supseteq B_2 \supseteq \dots \supseteq B_{N'}$. By*

extension, let $A_0 = B_0 = X$, and $A_i = \emptyset$ for every $i > N$, $B_j = \emptyset$ for every $j > N'$. Then $f' + g' = 1/2^K \sum_{k=1}^{N+N'} \chi_{W_k}$, where $W_k = \bigcup_{i+j=k}^{i \in \mathbb{N}, j \in \mathbb{N}} (A_i \cap B_j)$.

Proof. As a side remark, note that choosing the same K for f' and g' incurs no loss in generality. For every $k \geq 1$, $f'(x) + g'(x) \geq k/2^K$ iff there are two indices i, j such that $i + j = k$, $f'(x) \geq i/2^K$, and $g'(x) \geq j/2^K$, i.e., such that $i + j = k$, and $x \in A_i \cap B_j$. So $f' + g' = 1/2^K \sum_{k \geq 1} \chi_{W_k}$. But, for $k > N + N'$, whatever i and j such that $i + j = k$, either $i > N$ or $j > N'$, so either $U_i = \emptyset$ or $V_j = \emptyset$, i.e. $W_k = \emptyset$. The sum therefore indeed stop at $k = N + N'$ at the latest. \square

Claim O. *Let X be a core coherent space. For every bounded continuous functions f, g from X to \mathbb{R}^+ , and every functions f', g' in B , if $f' \ll f$ and $g' \ll g$, then $f' + g' \ll f + g$.*

Proof. Write $f' = 1/2^K \sum_{i=1}^N \chi_{U_i}$, $g' = 1/2^K \sum_{j=1}^N \chi_{V_j}$. As in Claim K, we may assume that f' and g' are written with the same K and the same N . By extension, let $U_0 = V_0 = X$, and $U_i = V_i = \emptyset$ for every $i > N$. By Claim N, $f' + g' = 1/2^K \sum_{k=1}^{2N} \chi_{W_k}$, where $W_k = \bigcup_{i+j=k} (U_i \cap V_j)$.

By assumption, $f' \ll f$, so by Claim J, $U_i \in f^{-1}(i/2^K, +\infty)$ for every i , $1 \leq i \leq 2^K$. This again holds when $i > 2^K$, since then U_i is empty, therefore way-below any open. Similarly, since $g' \ll g$, $V_j \in g^{-1}(j/2^K, +\infty)$ pour tout $j \geq 1$.

It follows that $U_i \cap V_j \in (f + g)^{-1}(k/2^K, +\infty)$ for every $i, j \geq 1$ such that $i + j = k$. Indeed, $U_i \cap V_j \subseteq U_i \in f^{-1}(i/2^K, +\infty)$ and $U_i \cap V_j \subseteq V_j \in g^{-1}(j/2^K, +\infty)$, so $U_i \cap V_j \in f^{-1}(i/2^K, +\infty) \cap g^{-1}(j/2^K, +\infty)$, since X is core coherent. But, for every $x \in f^{-1}(i/2^K, +\infty) \cap g^{-1}(j/2^K, +\infty)$, $f(x) > i/2^K$ and $g(x) > j/2^K$, so $f(x) + g(x) > k/2^K$, i.e., $x \in (f + g)^{-1}(k/2^K, +\infty)$. So $U_i \cap V_j \in f^{-1}(i/2^K, +\infty) \cap g^{-1}(j/2^K, +\infty) \subseteq (f + g)^{-1}(k/2^K, +\infty)$.

We also have $U_i \cap V_j \in (f + g)^{-1}(k/2^K, +\infty)$ when $i + j = k \geq 1$ but i or j is zero. If for example $i = 0$, then $j = k$, $U_i \cap V_j = X \cap V_k = V_k \in g^{-1}(k/2^K, +\infty) \subseteq (f + g)^{-1}(k/2^K, +\infty)$.

So $U_i \cap V_j \in (f + g)^{-1}(k/2^K, +\infty)$ for every i, j such that $i + j = k$, $k \geq 1$. It follows easily that $W_k = \bigcup_{i+j=k} U_i \cap V_j \in (f + g)^{-1}(k/2^K, +\infty)$ for every $k \geq 1$. By Claim J, $f' + g' \ll f + g$. \square

Claim P. *Let X be compact. For every $h' \in B$, for every $a > 0$, $h' \ll a\chi_X$ iff h' is of the form $\sum_{i=1}^n a_i \chi_{U_i}$, $U_1 \supseteq \dots \supseteq U_n$, $a_1, \dots, a_n \in \mathbb{R}^+$, with $\sum_{i=1}^n a_i < a$.*

In particular, $h' \ll a\chi_X$ iff there is $a' < a$ with $h' \leq a'\chi_X$.

Proof. For every $h' \in B$, write $h' = \sum_{i=1}^n a_i \chi_{U_i}$, $U_1 \supseteq \dots \supseteq U_n$, $a_1, \dots, a_n \in \mathbb{R}^+ \setminus \{0\}$. By Claim J., $h' \ll a\chi_X$ iff $U_i \in \emptyset$ for every i such that $\sum_{j=1}^i a_j \geq a$, and $U_i \in X$ for every i such that $\sum_{j=1}^i a_j < a$. This amounts to requiring that $U_i = \emptyset$ for every i such that $\sum_{j=1}^i a_j \geq a$; since X is compact, $U_i \in X$. The functions h' such that $h' \ll a\chi_X$ are therefore exactly those of the form $\sum_{i=1}^n a_i \chi_{U_i}$, $U_1 \supseteq \dots \supseteq U_n$ with $\sum_{i=1}^n a_i < a$.

If $h' \ll a\chi_X$, we have $h' \leq a'\chi_X$, where a' is any real strictly between $\sum_{i=1}^n a_i$ and a . Conversely, if $h' \leq a'\chi_X$ with $a' < a$, then $h' \ll a\chi_X$ since $a'\chi_X \ll a\chi_X$ by Claim J. \square

We now have enough material to apply Scott's formula on the continuous poset $\langle X \rightarrow \mathbb{R}^+ \rangle$, with basis B . For any functional F from $\langle X \rightarrow \mathbb{R}^+ \rangle$ to $\overline{\mathbb{R}}^+$, let $\tau(F)$ be the function defined by:

$$\tau(F)(f) = \sup_{g \in B, g \ll f} F(g)$$

Then $\tau(F)$ is a continuous functional from $\langle X \rightarrow \mathbb{R}^+ \rangle$ to $\overline{\mathbb{R}}^+$. In fact, $\tau(F)$ is the greatest continuous functional below F . We observe:

Claim Q. *Let X be stably core compact. If F is a (plain, lower, upper, linear) prevision then $\tau(F)$, too. If moreover X is compact and F is normalized, then $\tau(F)$ is normalized.*

Proof. The function $\tau(F)$ takes its values in \mathbb{R}^+ , since if $a = \sup_{x \in X} f(x)$, then $\tau(F)(f) = \sup_{g \in B, g \ll f} F(g) \leq F(a\chi_X) < +\infty$. It is continuous, and is positively homogeneous by Claim L. It is lower as soon as F is, since:

$$\begin{aligned} \tau(F)(f + g) &= \sup_{h \in B, h \ll f + g} F(h) \\ &= \sup_{h', h'' \in B, h' \ll f, h'' \ll g} F(h' + h'') \quad \text{by Claim M and Claim O} \\ &\geq \sup_{h', h'' \in B, h' \ll f, h'' \ll g} (F(h') + F(h'')) \\ &= \sup_{h' \in B, h' \ll f} F(h') + \sup_{h'' \in B, h'' \ll g} F(h'') = \tau(F)(f) + \tau(F)(g) \end{aligned}$$

Similarly, if F is upper then $\tau(F)$ is upper too, and if F is linear then so is $\tau(F)$. Finally, if F is normalized and X is compact, then for every $a > 0$,

$$\begin{aligned} \tau(F)(a\chi_X + f) &= \sup_{h \in B, h \ll a\chi_X + f} F(h) \\ &= \sup_{h', h'' \in B, h' \ll a\chi_X, h'' \ll f} F(h' + h'') \\ &= \sup_{a' < a, h'' \in B, h'' \ll f} F(a'\chi_X + h'') \quad \text{(by Claim P)} \\ &= \sup_{a' < a, h'' \in B, h'' \ll f} (a' + F(h'')) \quad \text{(since } F \text{ is normalized)} \\ &= a + \tau(F)(f) \end{aligned}$$

□

By Claim I, there is a monotonic linear functional G_0 such that $F \leq G_0$ and $F(f) = G_0(f)$. Let $G = \tau(G_0)$. By Claim Q, G is a continuous linear functional. Since it is the greatest one below G_0 $F \leq G$. Finally, $G \leq G_0$, so $G(f) \leq F(f)$; since $F \leq G$, $F(f) \leq G(f)$, so $F(f) = G(f)$. □

Theorem 4. *Let X be a stably locally compact space, F a normalized continuous lower prevision on X , and f a bounded continuous function from X to \mathbb{R}^+ . Then there is a normalized continuous linear prevision G such that $F \leq G$ and $F(f) = G(f)$.*

Proof. Recall the function \check{F}_f from the proof of Theorem 3, and define:

$$\check{\check{F}}_f(g) = \inf_{\epsilon \in \mathbb{R}^+} \check{F}_{f+\epsilon}(g)$$

This is well defined, and is always a non-negative real, never $+\infty$. Indeed, by Claim H and Claim E, whenever $\epsilon > 0$, $\check{F}_{f+\epsilon}(g) \leq \sup_{x \in X} g(x) \cdot \frac{1}{\epsilon + \inf_{x \in X} f(x)} F(f+\epsilon) < +\infty$.

Claim R. $\check{\check{F}}_f(0) = 0$. $\check{\check{F}}_f$ is monotonic. $\check{\check{F}}_f$ is positively homogeneous. For every $g \in \langle X \rightarrow \mathbb{R}^+ \rangle$, $\check{\check{F}}_f(g) \geq F(g)$.

Proof. By Claim B, Claim C, Claim E, Claim F. □

Claim S. Let F be a lower prevision on X . If F is normalized, then $\check{F}_{f+\epsilon}$ is antitone in ϵ .

Proof. Assume $\epsilon < \epsilon'$. If $\check{F}_{f+\epsilon}(g) = +\infty$ (which entails $\epsilon = 0$), clearly $\check{F}_{f+\epsilon'}(g) \leq \check{F}_{f+\epsilon}(g)$. So let us assume that $\check{F}_{f+\epsilon}(g) < +\infty$. Recall that:

$$\check{F}_{f+\epsilon}(g) = \inf_{\lambda/\lambda(f+\epsilon) \geq g} \left[F(\lambda(f+\epsilon)) - \sup_{h \leq \lambda(f+\epsilon)-g} F(h) \right]$$

Now $F(\lambda(f+\epsilon')) = \lambda(\epsilon' - \epsilon) + F(\lambda(f+\epsilon))$ since F is normalized. Moreover, if $g \leq \lambda(f+\epsilon)$, then:

$$\begin{aligned} \lambda(\epsilon' - \epsilon) + \sup_{h \leq \lambda(f+\epsilon)-g} F(h) &= \sup_{h \leq \lambda(f+\epsilon)-g} F(h + \lambda(\epsilon' - \epsilon)) \\ &\quad \text{since } F \text{ is normalized} \\ &\leq \sup_{h' \leq \lambda(f+\epsilon')-g} F(h') \end{aligned}$$

since for every $h \leq \lambda(f+\epsilon) - g$, $h' = h + \lambda(\epsilon' - \epsilon)$ is less than or equal to $\lambda(f+\epsilon') - g$. Therefore:

$$\begin{aligned} \check{\check{F}}_{f+\epsilon}(g) &= \inf_{\lambda/\lambda(f+\epsilon) \geq g} \left[F(\lambda(f+\epsilon')) - \left[\lambda(\epsilon' - \epsilon) + \sup_{h \leq \lambda(f+\epsilon)-g} F(h) \right] \right] \\ &\geq \inf_{\lambda/\lambda(f+\epsilon) \geq g} \left[F(\lambda(f+\epsilon')) - \sup_{h' \leq \lambda(f+\epsilon')-g} F(h') \right] \\ &\geq \inf_{\lambda/\lambda(f+\epsilon') \geq g} \left[F(\lambda(f+\epsilon')) - \sup_{h' \leq \lambda(f+\epsilon')-g} F(h') \right] \\ &\quad \text{since if } g \leq \lambda(f+\epsilon) \text{ then } g \leq \lambda(f+\epsilon') \\ &= \check{\check{F}}_{f+\epsilon'}(g) \end{aligned}$$

□

Claim T. Let F be a lower prevision on X . If F is normalized, then \check{F}_f is convex.

Proof. For every $\epsilon', \epsilon'' \geq 0$, there is $\epsilon \geq 0$ such that $\check{F}_{f+\epsilon}(g) + \check{F}_{f+\epsilon}(g') \leq \check{F}_{f+\epsilon'}(g) + \check{F}_{f+\epsilon''}(g')$. Indeed, by Claim S, it is enough to take $\epsilon = \max(\epsilon', \epsilon'')$. So $\inf_{\epsilon \geq 0} [\check{F}_{f+\epsilon}(g) + \check{F}_{f+\epsilon}(g')] \leq \inf_{\epsilon' \geq 0} \check{F}_{f+\epsilon'}(g) + \inf_{\epsilon'' \geq 0} \check{F}_{f+\epsilon''}(g')$. It follows:

$$\begin{aligned} \check{F}_f(g + g') &= \inf_{\epsilon \geq 0} \check{F}_{f+\epsilon}(g + g') \\ &\leq \inf_{\epsilon \geq 0} [\check{F}_{f+\epsilon}(g) + \check{F}_{f+\epsilon}(g')] \quad \text{by Claim D and Claim E} \\ &\leq \inf_{\epsilon' \geq 0} \check{F}_{f+\epsilon'}(g) + \inf_{\epsilon'' \geq 0} \check{F}_{f+\epsilon''}(g') \\ &= \check{F}_f(g) + \check{F}_f(g') \end{aligned}$$

\check{F}_f is therefore sub-linear, hence convex since it is positively homogeneous by Claim R. \square

Claim U. Let F be a lower prevision on X . If F is normalized, then $\check{F}_f(f) = F(f)$.

Proof. Clearly $\check{F}_f(g) \leq \check{F}_f(g)$ (take $\epsilon = 0$). When $g = f$, by Claim G $\check{F}_f(f) = F(f)$. So $\check{F}_f(f) \leq F(f)$. The converse inequality follows by Claim U. \square

Claim V. Let F be a lower prevision on X . If F is normalized, then $\check{F}_f(\chi_X) = 1$.

Proof. If there is no λ such that $1 \leq \lambda(f + \epsilon)$,

$$\begin{aligned} \check{F}_{f+\epsilon}(\chi_X) &= \inf_{\lambda/1 \leq \lambda(f+\epsilon)} \left[F(\lambda(f + \epsilon)) - \sup_{h \leq \lambda(f+\epsilon)-1} F(h) \right] \\ &= \inf_{\lambda/1 \leq \lambda(f+\epsilon)} \left[F(\lambda(f + \epsilon) - 1) + 1 - \sup_{h \leq \lambda(f+\epsilon)-1} F(h) \right] \end{aligned}$$

since F is normalized. Since $\sup_{h \leq \lambda(f+\epsilon)-1} F(h) = F(\lambda(f+\epsilon)-1)$, $\check{F}_{f+\epsilon}(\chi_X) = 1$. On the other hand, if there is no λ such that $1 \leq \lambda(f+\epsilon)$, then $\check{F}_{f+\epsilon}(\chi_X) = +\infty$. This can only happen when $\epsilon = 0$, and provided $\inf_{x \in X} f(x) = 0$. In this case, $\check{F}(\chi_X)$ is the greatest lower bound of 1 and $+\infty$; otherwise, $\check{F}(\chi_X)$ already equals 1, directly. \square

Let us prove the Theorem. Since F is super-linear, \check{F}_f is sub-linear by Claim T, F is monotonic and \check{F}_f also by Claim R, we can apply Roth's Sandwich Theorem: there is a functional G_0 from $\langle X \rightarrow \mathbb{R}^+ \rangle$ to $\overline{\mathbb{R}}^+$, which is monotonic and linear, and such that $F \leq G_0 \leq \check{F}_f$. By Claim U, $F(f) \leq G_0(f) \leq \check{F}_f(f) = F(f)$, so $G_0(f) = F(f)$. Next, \check{F}_f takes its values in \mathbb{R}^+ , so G_0 too. In particular, G_0 is a prevision. G_0 is normalized, since for every bounded continuous function g , for every $a \in \mathbb{R}^+$,

$G_0(a+g) = aG_0(\chi_X) + G_0(g)$ (since G_0 is linear) $= a + G_0(g)$. Indeed, $1 = F(\chi_X) \leq G_0(\chi_X) \leq \check{F}_f(\chi_X) = 1$ by Claim V.

So G_0 is a normalized linear prevision such that $F \leq G_0$ et $F(f) = G_0(f)$. (And this holds on any topological space X .) Now X is stably locally compact, hence stably core compact, so let $G = \mathfrak{r}(G_0)$. By Claim Q, G is a continuous linear prevision. Since G is the greatest continuous functional below G_0 and $F \leq G_0$, it follows that $F \leq G$. Since $G \leq G_0$, $G(f) \leq G_0(f) = F(f) \leq G(f)$, so $G(f) = F(f)$. Finally, we show that G is normalized—and we do not need X to be compact, as Claim Q would suggest. We claim indeed that $G(a+g) = a + G(g)$: by linearity, $G(a+g) = aG(\chi_X) + G(g)$, and $G(\chi_X) = 1$ since $G(\chi_X) \leq G_0(\chi_X) = 1$ on the one hand, and $G(\chi_X) \geq F(\chi_X) = 1$ on the other hand. \square

Theorem 5. *Let X be stably locally compact, F a continuous normalized prevision on X . Then F is lower iff $CCoeur_1(F) \neq \emptyset$ and for every $f \in \langle X \rightarrow \mathbb{R}^+ \rangle$, $F(f) = \inf_{G \in CCoeur_1(F)} G(f)$. In this case, the inf is attained: $F(f) = \min_{G \in CCoeur_1(F)} G(f)$.*

Proof. If F is lower then $CCoeur_1(F)$ is non-empty by Theorem 4. Moreover, for every $f \in \langle X \rightarrow \mathbb{R}^+ \rangle$, there is a $G \in CCoeur_1(F)$ such that $F(f) = G(f)$, so clearly $F(f) = \inf_{G \in CCoeur_1(F)} G(f)$ and the inf is attained.

Conversely, if $CCoeur_1(F)$ is non-empty and $F(f) = \inf_{G \in CCoeur_1(F)} G(f)$ for every $f \in \langle X \rightarrow \mathbb{R}^+ \rangle$, then F is super-additive, hence lower. Indeed:

$$\begin{aligned} F(f + f') &= \inf_{G \in CCoeur_1(F)} G(f + f') = \inf_{G \in CCoeur_1(F)} (G(f) + G(f')) \\ &\quad \text{since } G \text{ is additive} \\ &\geq \inf_{G \in CCoeur_1(F)} G(f) + \inf_{G \in CCoeur_1(F)} G(f') = F(f) + F(f') \end{aligned}$$

\square

Proposition 4. *Let X be stably compact, F a normalized continuous lower prevision, then $CCoeur_1(F)$ is a non-empty saturated compact convex subset of $\mathbf{P}_{1wk}^\Delta(X)$.*

Proof. Convexity is obvious, non-emptiness is by Theorem 5, and saturation is clear. Compactness is almost a consequence of [19, Corollary 2]. We would need to check that $C = \langle X \rightarrow \mathbb{R}^+ \rangle$ is a so-called continuous d-cone, with an additive way-below relation. (Note that what Plotkin calls weak*-Scott what we simply call the weak topology.) Claim M indeed shows that \ll is additive, however every (continuous) d-cone is a (continuous) cpo, which $\langle X \rightarrow \mathbb{R}^+ \rangle$ fails to be. (One might argue we could have defined previsions as taking maps from $\langle X \rightarrow \overline{\mathbb{R}}^+ \rangle$ to $\overline{\mathbb{R}}^+$ instead. However, we wanted to provide a nice link with the theory of games and belief functions, and the latter works more smoothly provided we exclude $+\infty$. Moreover, in applications we only really need sub-normalized or normalized games/previsions.)

However, here is a direct argument, inspired by [9]. We need to recall that the *co-compact topology* of a stably compact space X is the collection of all complements of saturated compacts of X . The space X with its co-compact topology is its *de Groot dual* X^d . Then X^d is also stably compact, its specialization ordering is the converse

\geq of that of X , $X^{dd} = X$. The coarsest topology containing both the original and the co-compact topology is the *patch topology*. X equipped with the patch topology is a compact pospace, which we write X' [9]. Note that the product $Y = \prod_{i \in I} Y_i$ of stably compact spaces is stably compact, that $Y' = \prod_{i \in I} Y'_i$, and that the specialization ordering of Y is the component-wise ordering.

The space $[0, 1]$ is stably compact, and $[0, 1]'$ is equipped with the usual, metric topology. Let X be stably compact, and consider $Z = \prod_{\substack{f \in \langle X \rightarrow \mathbb{R}^+ \rangle \\ \sup_{x \in X} f(x) = 1}} [0, 1]$, which is also stably compact. Let $P(X)$ be any of the spaces of sub-normalized previsions we consider. There is an obvious map $e : P(X) \rightarrow Z$ that sends F to the family of all $F(f)$, $f \in \langle X \rightarrow \mathbb{R}^+ \rangle$, $\sup_{x \in X} f(x) = 1$. Conversely, for any family $z = (z_f)_{\substack{f \in \langle X \rightarrow \mathbb{R}^+ \rangle \\ \sup_{x \in X} f(x) = 1}}$ of elements of $[0, 1]$, one may define a positively homogeneous functional $m(z)$ from $\langle X \rightarrow \mathbb{R}^+ \rangle$ to \mathbb{R}^+ by: $m(z)(0) = 0$, $m(z)(f) = az_{f/a}$ when $a = \sup_{x \in X} f(x) > 0$.

We then check that the subspace of Z of those z such that $m(z)$ is a (plain, lower, upper, sub-normalized, normalized) prevision is patch-closed in Z , hence stably compact. (See [7, lemme 11.6.1].) The idea is that this subspace is defined as the set of objects z satisfying a collection of *patch-continuous* equations, i.e., equations of the form $f(z_{f_1}, \dots, z_{f_n}) = g(z_{f_1}, \dots, z_{f_n})$, where f_1, \dots, f_n are fixed, and f and g are continuous functions from $[0, 1]^n$ to $[0, 1]$. E.g., that $m(z)$ is monotonic is equivalent to the fact that $a \times z_{f/a} \leq b \times z_{g/b}$ for all $f, g \in \langle X \rightarrow \mathbb{R}^+ \rangle$ that are not identically zero and such that $f \leq g$, where $a = \sup_{x \in X} f(x)$, $b = \sup_{x \in X} g(x)$; in turn, an inequality $u \leq v$ is an abbreviation for the equation $\max(u, v) = v$. Since every intersection of patch-closed subsets is patch-closed, the given subspace is patch-closed.

We then check that, if $Z_{\leq 1}$ is the subspace of Z consisting of all z such that $m(z)$ is a sub-normalized prevision, then e defines a homeomorphism of the space $P_{\leq 1} wk(X)$ of all sub-normalized previsions onto $Z_{\leq 1}$, with inverse m . In particular, $P_{\leq 1} wk(X)$ is stably compact.

The technique of patch-continuous equations implies that given any conjunction of properties among “lower”, “upper”, “linear” and “normalized”, the subspace of those previsions in $P_{\leq 1} wk(X)$ satisfying these properties is also patch-closed in it, hence stably compact.

Next, we observe that $\tau : P_{\leq 1} wk(X) \rightarrow \mathbf{P}_{\leq 1} wk(X)$ (see Claim Q and later) and the obvious inclusion $\mathfrak{s} : \mathbf{P}_{\leq 1} wk(X) \rightarrow P_{\leq 1} wk(X)$ forms a retraction-section pair. I.e., that they are continuous and $\tau(\mathfrak{s}(F)) = F$ for all F . This allows us to conclude that $\mathbf{P}_{\leq 1} wk(X)$ is stably compact: by Lawson’s Lemma (see again [9]), any retract of a stably compact space is stably compact. We need X to be compact additionally, in the case of normalized previsions, to be able to apply Claim Q and show that $\tau : P_1 wk(X) \rightarrow \mathbf{P}_1 wk(X)$.

Finally, we note that $CCoeur_1(X)$ is the intersection of $\uparrow F$ (the set of all normalized continuous previsions F' such that $F \leq F'$) and $\mathbf{P}_{1}^{\Delta} wk(X)$. The former is trivially compact (every set $\uparrow x$ is compact in any topological space), hence patch-closed in $\mathbf{P}_1 wk(X)$. The technique of patch-continuous equations shows that $\mathbf{P}_{1}^{\Delta} wk(X)$ is again patch-closed in $\mathbf{P}_1 wk(X)$, so $CCoeur_1(X)$ is patch-closed in $\mathbf{P}_1^{\Delta} wk(X)$. It is easy to see that a patch-closed subset of a stably compact space is compact. \square

Proposition 5. *Let X be a topological space, F a normalized continuous upper prevision, then $CPeau_1(F)$ is a closed convex subset of $\mathbf{P}_{1\ wk}^\Delta(X)$. It is non-empty as soon as X is stably compact.*

Proof. Convexity is again obvious. We have already mentioned that $CPeau_1(F)$ would be non-empty if X is stably compact (details in [7, théorème 11.7.4]). Then we notice that:

$$\begin{aligned} CPeau_1(X) &= \{G \in \mathbf{P}_{1\ wk}^\Delta(X) \mid G \leq F\} = \bigcap_{f \in (X \rightarrow \mathbb{R}^+)} \{G \in \mathbf{P}_{1\ wk}^\Delta(X) \mid G(f) \leq F(f)\} \\ &= \bigcap_{f \in (X \rightarrow \mathbb{R}^+)} \left(\mathbf{P}_{1\ wk}^\Delta(X) \setminus [f > F(f)] \right) \end{aligned}$$

which is therefore closed in the weak topology. \square

Proposition 6. *Let X be a stably compact space. The continuous normalized body $CCorps_1(F) = CCoeur_1(F^-) \cap CPeau_1(F^+)$ of a continuous normalized fork $F = (F^-, F^+)$ on X is a lens. Moreover, $CCoeur_1(F^-) = \uparrow CCorps_1(F)$ and $CPeau_1(F^+) = \downarrow CCorps_1(F)$.*

Proof. We show that: (*) whenever $G \in CCoeur_1(F^-)$, there is some $G' \in CCoeur_1(F^-) \cap CPeau_1(F^+)$ such that $G' \leq G$.

Let $F'(h) = \inf_{\substack{f, g \in (X \rightarrow \mathbb{R}^+) \\ f+g \geq h}} (F^+(f) + G(g))$. This is well-defined since e.g., we may take $f = 0$ and $g = h$. This also implies that $F'(h) \leq G(h)$. Clearly, $F^- \leq F' \leq G$: we have just shown $F' \leq G$, and for the other inequality, we note that for every f, g with $f + g \geq h$, $F^+(f) + G(g) \geq F^+(f) + F^-(g)$ (by assumption) $\geq F^-(f + g)$ (by (3)) $\geq F^-(h)$.

Now we observe that F' is an upper prevision. Indeed, $F'(0) = 0$ (taking $f = g = 0$), and when $\alpha > 0$, $F'(\alpha h) = \inf_{f+g \geq \alpha h} (F^+(f) + G(g)) = \inf_{f'+g' \geq h} (F^+(\alpha f') + G(\alpha g)) = \alpha F'(h)$; so F' is positively homogeneous. F' is clearly monotonic, while:

$$\begin{aligned} F'(h) + F'(h') &= \inf_{f+g \geq h, f'+g' \geq h'} (F^+(f) + G(g) + F^+(f') + G(g')) \\ &\geq \inf_{f+g \geq h, f'+g' \geq h'} (F^+(f + f') + G(g + g')) \\ &\quad \text{(because } F^+ \text{ is upper and } G \text{ is linear)} \\ &\geq \inf_{f''+g'' \geq h+h'} (F^+(f'') + G(g'')) = F'(h + h') \end{aligned}$$

So F' is upper.

Using Roth's Sandwich Theorem again, there is a linear monotonic functional G_0 such that $F^- \leq G_0 \leq F'$. Because $G_0 \leq F'$, G_0 does not take the value $+\infty$, so G_0 is a linear prevision. Let $G' = \tau(G_0)$. By Claim Q, G' is a continuous linear prevision, and $F^- \leq G' \leq F'$, as above. In particular, $1 = F^-(\chi_X) \leq G'(\chi_X) \leq F'(\chi_X) = 1$, so $G'(\chi_X) = 1$; using the fact that G' is linear, G' is then normalized. Since $F^- \leq G'$, $G' \in CCoeur_1(F^-)$. Since $G' \leq F'$ and clearly $F' \leq F^+$, $G' \in CPeau_1(F^+)$. Finally, since $G' \leq F'$ and $F' \leq G$, $G' \leq G$. So (*) obtains.

It follows in particular that $CCoeur_1(F^-) \cap CPeau_1(F^+)$ is non-empty. That $CCoeur_1(F^-) = \uparrow (CCoeur_1(F^-) \cap CPeau_1(F^+))$ is an easy consequence of (*). That $CPeau_1(F^+) = \downarrow (CCoeur_1(F^-) \cap CPeau_1(F^+))$ can be shown in a similar way, by defining $F''(h) = \sup_{\substack{f,g \in \langle X \rightarrow \mathbb{R}^+ \rangle \\ f+g \leq h}} (F^-(f) + G(g))$, where $G \in CPeau_1(F^-)$, and using F'' to show that there is some $G' \in CCoeur_1(F^-) \cap CPeau_1(F^+)$ such that $G \leq G'$. \square