An abstract decomposition of measures and its many applications

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Abstract

We consider a little-known abstract decomposition result, due to Dellacherie, and show that it yields many decompositions of measures, several of which are new. Then, we investigate how the outputs of the decomposition depend on its inputs, in particular characterising the two elements of the decomposition as projections in the sense of Riesz spaces and of metric spaces.

1 Introduction

In his book [9, Theorem T14, page 30], Dellacherie introduced the following theorem.

Theorem 1.1. Let $\mathcal{F}$ be a $\sigma$-algebra on $\Omega$, $\mu$ be a (positive and $\sigma$-additive) $\sigma$-finite measure on $(\Omega, \mathcal{F})$, and $\mathcal{G} \subseteq \mathcal{F}$ be closed under countable unions. Then there exist a set $\bar{G} \in \mathcal{G}$ and two measures $\mu_{\mathcal{G}}$ and $\mu_{\perp}^\mathcal{G}$ such that

$$\mu = \mu_{\mathcal{G}} + \mu_{\perp}^\mathcal{G}, \tag{1.1}$$

where $\mu_{\mathcal{G}}(\bar{G}^c) = 0$, and $\mu_{\perp}^\mathcal{G}(G) = 0$ for every $G \in \mathcal{G}$. Moreover, the decomposition (1.1) is unique.

We find this decomposition result to be very neat, and we agree with Dellacherie that it deserves to be classical. It is however a vast understatement to say that it has not become classical. Indeed, we were not able to find any

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This being our translation of the original French text ‘qui mériterait d’être classique’.
mention of it anywhere in the literature; even more curiously, Dellacherie himself, after introducing Theorem 1.1, never uses it once in his whole book, even if his [9, Theorem T41, page 58] can easily be proved using Theorem 1.1, as we will show in our Corollary 3.4.

The main contribution of this paper is then to show how Theorem 1.1, though easy to prove, can very profitably be applied to obtain many decompositions of measures (and processes), some well-known and some new; we hope our effort may help pluck Dellacherie’s nice theorem from obscurity. Our additional contribution is to study the dependence of \( G \) and \( \mu_G \) on \( G \) and \( \mu \).

Here an outline of the paper. In Section 2 we feature Theorem 2.3 (a slight refinement of Theorem 1.1). In Section 3 we give many examples of decompositions of positive measures which can be obtained using Theorem 2.3. In Section 4, we vastly extend the range of applicability of Theorem 2.3 by easily establishing a variant which holds for controlled vector measures, i.e. essentially all vector measures (including all Banach-valued and all \( L^0 \)-valued ones), which we use to derive the Hanh-Jordan decomposition of a real measure. In Section 5 we apply such extension to the stochastic integral with respect to a semimartingale, and obtain a decomposition of semimartingales due to Bichteler, but with a much more elementary proof and a sharper statement. In Section 6 we prove that any spectral measure is controlled, so our decomposition theorem can be applied to it. Finally, in Section 7 we investigate the monotonicity and order continuity of \( \bar{G} \) and \( \mu_G \) as functions of \( G \); and we show that the maps \( \mu \mapsto \mu_G \), \( \mu \mapsto \mu_G^\perp \), defined on the Banach lattice of real-measures, are the projections in the sense of Riesz spaces (i.e. they are band projections) and of metric spaces (i.e. they are unique nearest point maps).

2 Dellacherie’s theorem for positive measures

In this section we consider a theorem which slightly generalises Theorem 1.1. This variant brings several advantages: (i) it provides a simple characterisation of \( \bar{G} \in \mathcal{G} \) which will prove useful in suggesting a proof and, more importantly, in enabling us to extend the decomposition theorem to vector measures in Section 4; (ii) it allows us to consider finitely-additive (and not only countably-additive) measures; (iii) it outlines the case when there is a uniquely distinguished set \( \bar{G} \) (the largest one) in its equivalence class\(^2\), which is important as this sometimes occurs in practice. Moreover, we point out that an alternative and essentially trivial proof of such decomposition result can be obtained by applying the concept of essential supremum. This has the considerable advantage that, while Dellacherie’s proof relies on several non-trivial properties\(^3\) of the (pointwise)\(^2\)

\(^2\)Notice that \( \bar{G} \) in Theorem 1.1 is always unique up to \( \mu \)-null sets, as it trivially follows from the uniqueness of the decomposition (1.1).

\(^3\)These properties: (i) any family \( M \) of measures bounded from above by \( \mu \) admits a supremum \( \nu := \sup M = \sup M_c \) for some countable \( \{\nu_k\}_{k \in \mathbb{N}} \), \( M_c \) \( \subseteq \) \( M \), where \( \{\nu_k\}_{k \in \mathbb{N}} \) can w.l.o.g. be chosen to be increasing; (ii) if \( \{\nu_k\}_{k \in \mathbb{N}} \) is a sequence of measures then
order\(^4\) between measures (with no explicit mention or reference), the concept of \(\mu\)-essential supremum (i.e. of supremum on the space \(L^0(\mu)\)) is well known and its properties are easily proved.

We want to point out that, though hardly apparent at first sight, behind the curtains the two proofs are actually closely related, as follows. Consider w.l.o.g. the case of finite \(\mu\). While Dellacherie considers the order on the space \(\mathcal{M}\) of real measures, one can rely on the (easier to deal with) order on the space \(L^1(\mu)\) of (equivalence classes of) \(\mu\)-integrable functions, because all measures considered in Dellacherie’s proof are absolutely continuous with respect to \(\mu\) (i.e. of the form\(^5\) \(f \cdot \mu\)), and so one can identify the measure \(f \cdot \mu\) with the function \(f\); moreover, the map \(f \mapsto f \cdot \mu\), from \(L^1(\mu)\) to \(\{\nu \in \mathcal{M} : \nu \ll \mu\}\) is an isomorphism of Banach lattices (see [2, Theorem 13.19]).

We now introduce some standard notations and definitions; in particular we recall the definition of \(\mu\)-essential supremum, and an important property thereof.

Given a measurable space \((\Omega, \mathcal{F})\), \(A^c\) denotes the complement of \(A \subseteq \Omega\), i.e., \(A^c := \Omega \setminus A\). If \(\lambda\) is a positive or a vector\(^6\) measure (finitely or countably additive) on \((\Omega, \mathcal{F})\), we will say that \(F \in \mathcal{F}\) is a \(\lambda\)-null set (or a null set for \(\lambda\)) if the restriction \(\lambda|_F := \lambda(F \cap \cdot)\) of \(\lambda\) to \(F \in \mathcal{F}\) is the zero measure, i.e., if \(\lambda(G) = 0\) for all \(G \subseteq F, G \in \mathcal{F}\); in this case we will say that \(\lambda\) is concentrated on \(F^c\). Notice that if \(\lambda\) is a positive measure then \(F \in \mathcal{F}\) is \(\lambda\)-null if and only if \(\lambda(F) = 0\), so our definition of null set extends the usual definition given for positive measures. If a family of measures \((\lambda_i)_{i \in I}\) are concentrated on pairwise disjoint sets \((A_i)_{i \in I}\), we will say that they are (mutually) singular. From now on, when we say simply ‘measure’, without further qualifications, we will mean a countably-additive positive measure, i.e., any countably-additive map \(\mu : \mathcal{F} \to [0, \infty]\).

Remark 2.1. Let \(\mu\) be a positive, finite (countably-additive) measure. The \(\mu\)-essential supremum of a family \(\mathcal{H}\) of random variables is the supremum (i.e. the smallest majorant) of \(\mathcal{H}\) seen as a subset of the Riesz space \(L^0(\mu)\) of (equivalence classes of) random variables, endowed with the order \(\leq_\mu\) defined by: \(X \leq_\mu Y\) if \(X \leq Y\) \(\mu\)-a.e., i.e., if \(\mu\{\{X > Y\}\} = 0\). In other words, we say that a random variable \(X\) is the \(\mu\)-essential supremum of a family \((X_i)_{i \in I}\) of random variables if

1. \(X_i \leq X\) \(\mu\)-a.e. for every \(i \in I\),
2. \(X_i \leq Y\) \(\mu\)-a.e. for every \(i \in I\) implies \(X \leq Y\) \(\mu\)-a.e..

It is easy to prove (see, e.g., [15, Proposition 4.1.1]) the existence of the \(\mu\)-essential supremum \(X\) of any family \(\mathcal{H} \subseteq L^0(\mu)\) which is bounded above\(^7\), and

\(^4\)\(\sup_k \nu_k(F) = (\sup_k \nu_k(F))\) holds if \(\nu_k\)\(k \in \mathbb{N}\) is increasing (the first supremum being on the space of measures, the second on the set of real numbers).

\(^5\)We use the notation \(f \cdot \mu\) to denote the measure \((f \cdot \mu)(F) := \int_F f \, d\mu\) for all \(F \in \mathcal{F}\).

\(^6\)I.e., a measure with values in a vector space, as defined in Section 4.

\(^7\)I.e. \(\exists Y \in L^0(\mu)\) such that \(H \leq Y\) \(\mu\)-a.e. for all \(H \in \mathcal{H}\).
the fact that \( X \) is (the equivalence class of) \( \sup_{n \in \mathbb{N}} X_n \) for some sequence of random variables \( X_n \in \mathcal{H} \). It easily follows that the \( \mu \)-essential supremum of a family of indicators (i.e., a \( \{0,1\} \)-valued random variable) is an indicator (see, e.g., [15, Proposition 4.1.2]), and so one can talk of \( \mu \)-essential supremum of a family of (measurable) sets. In other words, if we endow \( \mathcal{F} \) with the order \( \leq \mu \) of inclusion \( \mu \)-a.e. (\( A \leq \mu B \) means \( A \subseteq B \) \( \mu \)-a.e., i.e., \( \mu(A \setminus B) = 0 \)), then any family of sets \( \mathcal{G} \subseteq \mathcal{F} \) admits a \( \leq \mu \)-supremum \( G \), and \( \exists \) a sequence \( G_n \in \mathcal{G}, n \in \mathbb{N} \) such that \( \cup_n G_n = G \) \( \mu \)-a.e.

**Remark 2.2.** The (pointwise) supremum of an uncountable family of random variables is in general not measurable. So, in measure theory the concept of supremum of families of random variables has no importance; instead one always considers the \( \mu \)-essential supremum of families of equivalence classes of random variables. The two concepts of course coincide when applied to countable families of random variables, since the countable union of sets of measure 0 has measure 0.

**Theorem 2.3.** Let \( \mu \) be a finitively-additive positive measure on \((\Omega, \mathcal{F})\) and \( \mathcal{G} \subseteq \mathcal{F} \). Then,

1. The following two statements are equivalent:
   
   (a) There exist a set \( G_\mu \in \mathcal{G} \) such that \( G \setminus G_\mu \) is a \( \mu \)-null set for every \( G \in \mathcal{G} \).
   
   (b) There exist a set \( G \in \mathcal{G} \) and two measures \( \mu_G, \mu_G^c \) on \((\Omega, \mathcal{F})\) such that \( \mu = \mu_G + \mu_G^c \), \( \mu_G \) is concentrated on \( \bar{G} \in \mathcal{G} \), every \( G \in \mathcal{G} \) is \( \mu_G^c \)-null. (2.1)

   Moreover, if the above statements hold then \( G_\mu \) and \( \bar{G} \) are unique up to a \( \mu \)-null set, \( G_\mu = \bar{G} \) up to a \( \mu \)-null set, \( \mu_G = \mu(G_\mu \cap \cdot) \) and \( \mu_G^c = \mu(G_{\mu}^c \cap \cdot) \); in particular, such \( \mu_G, \mu_G^c \) are unique.

2. If \( \mu \) is \( \sigma \)-finite and countably-additive, and \( \mathcal{G} \) is non-empty and closed under countable unions, then there exists a set \( G_\mu \in \mathcal{G} \) as in item (1a).

Proof. We first prove item 1. If (1b) holds then \( \bar{G}^c \) is \( \mu_G \)-null and \( \bar{G} \) is \( \mu_G^c \)-null, and so \( \mu_G = \mu(G \cap \cdot) \) and \( \mu_G^c = \mu(G^c \cap \cdot) \), which also implies that \( G \cap G^c \) is \( \mu \)-null for every \( G \in \mathcal{G} \), i.e., (1a) holds with \( G_\mu := \bar{G} \). Conversely, if \( G_\mu \) is as in (1a), then (1b) trivially holds with \( \bar{G} := G_\mu \), \( \mu_G = \mu(\bar{G} \cap \cdot) \), \( \mu_G^c = \mu(\bar{G}^c \cap \cdot) \). Such \( G_\mu \) is unique up to \( \mu \)-null sets, since if \( G_\mu \in \mathcal{G} \) is also such that \( G \setminus G_\mu \) is \( \mu \)-null for every \( G \in \mathcal{G} \), then both \( G_\mu \setminus G_n \) and \( G_\mu \setminus G_\mu \) are \( \mu \)-null.

We now prove item 2, using Remark 2.1. If \( \mu \) is a finite measure then the \( \mu \)-essential supremum \( G \) of \( \mathcal{G} \) exists; moreover, \( G \) is the maximum of \( \mathcal{G} \) under \( \leq \mu \) (indeed \( G \in \mathcal{G} \) because \( \mathcal{G} \) is closed under countable unions and \( \exists \) a sequence \( G_n \in \mathcal{G}, n \in \mathbb{N} \) such that \( \cup_n G_n = G \) \( \mu \)-a.e.); in other words, \( G \) satisfies item (1a).

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8To be precise \( \leq \mu \) is an order not really on \( \mathcal{F} \), but rather on the family \( \mathcal{F}^\mu \) of equivalence classes of sets in \( \mathcal{F} \) which differ by a \( \mu \)-null set.
If $\mu$ is $\sigma$-finite but not finite then there exists $f > 0$ such that $f \in L^1(\mu)$ (see, e.g., [32, Lemma 6.9]). Therefore, we can apply the same argument as above to the finite positive measure $\nu = f \cdot \mu$, which is equivalent to $\mu$, and obtain the desired result.

If $\mu, G, G_\mu$ are as in Theorem 2.3, in analogy with the case discussed later in Corollary 3.2, we will use the following terminology: we will call $\mu_G, \mu_G^\perp$, and the equivalence class (up to equality $\mu$-a.e.) $G_\mu$ of $G_\mu$, respectively, the $G$-atomic part of $\mu$, the $G$-diffuse part of $\mu$, and the $G$-atomic support of $\mu$. Notice that, as $G_\mu$ is unique up to $\mu$-null sets, $G_\mu$ is unique; as usual, we will often conveniently (though slightly inappropriately) disregard the distinction between $G_\mu$ and $G_\mu^\perp$, for example calling $G_\mu$ the $G$-atomic support of $\mu$, or writing $G_\mu$ to mean one of its representatives.

Remark 2.4. As we will see, it sometimes happens that one is interested in a family of sets $G$ which admits a maximum $G^*$ with respect to the pointwise order $\leq$, i.e. the order of inclusion of sets (for which $A \leq B$ means $A \subseteq B$, i.e. the indicator functions satisfy $1_A \leq 1_B$ at every point); equivalently, $G^* \in \mathcal{G}$ satisfies $G \setminus G^* = \emptyset$ for all $G \in \mathcal{G}$. In this case, though any set $G$ which differs from $G^*$ by a $\mu$-null set will still satisfy (2.1), $G^*$ has the distinguished property of being the largest of the sets $\bar{G}$ such that (2.1) holds (since the maximum is always unique).

Moreover, in this case we can apply item 1 of Theorem 2.3 to conclude that $\mu_G, \mu_G^\perp$ as in (2.1) exist even if $\mu$ is only finitely-additive.

3 Examples of decompositions

In this section we apply Theorem 2.3 to obtain many decompositions of measures. We first highlight three well known decompositions, and a recently-discovered closely related one. Then we briefly list many other ones, several of which are new. The reader will notice how many choices of $G$ are of the form ‘all the subsets of $\Omega$ of size at most $x$', where the ‘size’ could be meant in different ways: as measure, as dimension, as Baire category.

Recall that a measure $\mu$ is said to be absolutely continuous with respect to a measure $\nu$ if $\nu(A) = 0$ implies $\mu(A) = 0$ for $A \in \mathcal{F}$; in this case we write $\mu \ll \nu$.

Corollary 3.1 (Lebesgue decomposition). Let $\mu$ and $\nu$ be two $\sigma$-finite measures on $(\Omega, \mathcal{F})$. There exist two $\sigma$-finite measures $\mu^{ac}, \mu^s$ such that

$$\mu = \mu^{ac} + \mu^s,$$

where $\mu^{ac}$ is absolutely continuous with respect to $\nu$, and $\mu^s$ and $\nu$ are singular. Moreover, the decomposition (3.1) is unique.

Proof. Let $G := \{A \in \mathcal{F} : \nu(A) = 0\}$, which contains $\emptyset$ and is closed under countable unions. Thus, Theorem 2.3 applied to $\mu$ and $G$ provides the unique decomposition $\mu = \mu_G + \mu_G^\perp$ where $\mu_G$ is concentrated on a $\nu$-null set and every
null set is a $\mu_G^\perp$-null set. Hence, $\mu^a := \mu_G^\perp$ and $\mu^d := \mu_G$ satisfy the required decomposition (3.1).

Recall that a set $A \in \mathcal{F}$ is an atom of the measure $\mu$ if $\mu(A) > 0$ and for every $B \in \mathcal{F}$, $B \subseteq A$ either $\mu(B) = 0$ or $\mu(B) = \mu(A)$; restated in the language of Remark 2.1, the atoms of $\mu$ are the (equivalence classes of) minimal sets of $(\mathcal{F}, \leq_{\mu})$ which are not the minimum (i.e. which are not $\mu$-a.e. = $\emptyset$). We say that a measure $\mu$ is (purely) atomic if every measurable set of strictly positive measure contains an atom of $\mu$, and diffuse (or non-atomic) if it gives measure zero to every atom of $\mu$. We refer to [22] for further information on atomic and non-atomic measures.

**Corollary 3.2** (Decomposition into atomic and diffuse components). Let $\mu$ be a $\sigma$-finite measure on $(\Omega, \mathcal{F})$. There exist two $\sigma$-finite measures $\mu^a$, $\mu^d$ such that

$$
\mu = \mu^a + \mu^d,
$$

where $\mu^a$ is purely atomic and $\mu^d$ is diffuse. Moreover, the decomposition (3.2) is unique.

**Proof.** Let $\mathcal{G}$ be the family of all the countable unions of atoms, which is trivially closed under countable unions. If $\mathcal{G} = \emptyset$, take $\mu = \mu^d$; otherwise, apply Theorem 2.3 to $\mu$ and $\mathcal{G}$ to get the unique decomposition $\mu = \mu_G + \mu_G^\perp$ where $\mu_G$ is concentrated on a countable union of atoms $\bar{G} \in \mathcal{G}$ and $\mu_G^\perp$ gives null measure to every atom. Hence, $\mu^a := \mu_G$ and $\mu^d := \mu_G^\perp$ satisfy the required decomposition (3.2).

**Remark 3.3.** While Corollary 3.2 applies in any measurable space, a more precise statement can be given when each atom of $\mu$ is (the equivalence class of) a singleton or, more generally, an atom of $\mathcal{F}$. An atom of $\mathcal{F}$ is defined as a set $A \in \mathcal{F}$ such that $B \in \mathcal{F}$, $B \subseteq A$ imply either $B = \emptyset$ or $B = A$; restated in the language of Remark 2.4, an atom of $\mathcal{F}$ is a minimal set of $(\mathcal{F}, \leq)$ which is not a minimum (i.e. which is non-empty).

Indeed, in this case the set of atoms of $\mathcal{F}$ with strictly positive measure $\mu$ is at most countable (because $\mu$ is $\sigma$-finite), and thus the unions of all such sets is the pointwise maximum $G^*$ of the family $\mathcal{G}$ of countable unions of (some) such sets. This gives an example of a set $G^*$ which admits a maximum for the pointwise order, see Remark 2.4.

We now show two applications of Theorem 2.3 to stopping and random times: the first one leads to the decomposition $\tau = \tau_A \wedge \tau_{A^c}$ of a stopping time $\tau$ into its accessible part $\tau_A$ and totally inaccessible part $\tau_{A^c}$; the second one leads to the decomposition $\tau = \tau_B \wedge \tau_{B^c}$ of a random time $\tau$ into its thin part $\tau_B$ and thick part $\tau_{B^c}$. For the former decomposition, we refer to [29, Ch. III, Sec. 2] where the definitions of accessible and totally inaccessible stopping times are given and the result is presented as Theorem 3. For the latter decomposition, we refer to [1, Theorem 5.5] and we will recall later the non-standard definitions of thin and thick random times. Here, we fix an underlying filtered probability
space \((\Omega, \mathcal{F}, (\mathcal{F}_t), \mathbb{P})\) and, for a set \(A \in \mathcal{F}\) and a stopping time \(\tau\), we use the standard notations
\[
\tau_A(\omega) := \begin{cases} \tau(\omega) & \text{if } \omega \in A \\ \infty & \text{if } \omega \notin A. \end{cases}
\]
(3.3)
and point out that the identity
\[
[A] := \{(\omega, t) \in \Omega \times [0, \infty): \tau(\omega) = t\}.
\]
and thus
\[
[\tau_A] = (A \times [0, \infty)) \cap [\tau] \quad \text{(3.4)}
\]
trivially holds for any stopping time \(\tau\) and \(A \in \mathcal{F}\).

**Corollary 3.4.** If \(\tau\) is a stopping time, there exists \(A \in \mathcal{F}\) such that \(A \subseteq \{\tau < \infty\}\), \(\tau_A\) is accessible, \(\tau_{A^c}\) is totally inaccessible and \(\tau = \tau_A \wedge \tau_{A^c}\) \(\mathbb{P}\)-a.s.

**Proof.** Since the constant \(\infty\) is a predictable stopping time, the family
\[
\mathcal{G} := \{A \in \mathcal{F}: A \subseteq \{\tau < \infty\} \text{ and } \tau_A \text{ is accessible}\}
\]
contains \(\emptyset\), and it is not empty. Moreover, \(\mathcal{G}\) in closed under countable unions, as we now show. If \((A_n)_{n \in \mathbb{N}} \subseteq \mathcal{G}\) and \(A = \bigcup_n A_n\), then (3.4) shows that \([\tau_A] = \bigcup_n [\tau_{A_n}]\) and \(A \subseteq \{\tau < \infty\}\). Since \(\tau_{A_n}\) is accessible, there exists predictable stopping times \((\tau_{n,m})_m\) such that
\[
[\tau_A] = \bigcup_n [\tau_{A_n}] \subseteq \bigcup_{n,m} [\tau_{n,m}] \quad \mathbb{P}\text{-a.s.},
\]
and so \(\tau_A\) is accessible, thus \(A \in \mathcal{G}\). Hence, we can apply Theorem 2.3 to \(\mathbb{P}\) and \(\mathcal{G}\) and obtain \(A := G_\mathbb{P} \in \mathcal{G}\) such that \(\mathbb{P}(A^c \cap B) = 0\) for every \(B \in \mathcal{G}\). Since \(A \in \mathcal{G}\), \(\tau_A\) is accessible; let us prove that \(\tau_{A^c}\) is totally inaccessible. Indeed, if \(\sigma\) be a predictable stopping time and
\[
B := A \cup \{\tau = \sigma < \infty\},
\]
then \([\tau_B] \subseteq [\tau_A] \cup [\sigma] \quad \mathbb{P}\text{-a.s.},\) and so \(B \in \mathcal{G}\) and thus \(\mathbb{P}(A^c \cap B) = 0\). Since
\[
B \cap A^c = \{\tau = \sigma < \infty\} \cap A^c = \{\tau_{A^c} = \sigma < \infty\},
\]
we get that
\[
0 = \mathbb{P}(A^c \cap B) = \mathbb{P}(\{\tau_{A^c} = \sigma < \infty\}),
\]
and so \(\tau_{A^c}\) is totally inaccessible. \(\square\)

**Remark 3.5.** As suggested by the proofs of Corollary 3.4 given in [10, Theorem in Chapter 4, no. 80] and [9, Theorem T41, page 58], there are other possible choices for the family \(\mathcal{G}\) to be chosen in the proof of Corollary 3.4. For example, if \(\mathcal{S}(\tau)\) denotes the set of increasing sequences \((\sigma_n)_n\) of stopping times bounded from above by \(\tau\), and for \((\sigma_n)_n \in \mathcal{S}(\tau)\) we define
\[
A[(\sigma_n)_n] := \left\{ \lim_n \sigma_n = \tau, \sigma_n < \tau \text{ for all } n \right\} \cup \{\tau = 0\},
\]
then one can choose \(\mathcal{G}\) to be \(\{A[(\sigma_n)_n] : (\sigma_n)_n \in \mathcal{S}(\tau)\}\). Alternatively, one could take for \(\mathcal{G}\) the set of countable unions of sets of the form \(\{\sigma = \tau < \infty\}\), where \(\sigma\) ranges across all *predictable* stopping times.
To obtain an analogous decomposition for random times, we recall the definitions of thin random time and thick random time. A random time \( \tau \) (i.e., a \([0, \infty]\)-valued, \( F \)-measurable random variable) is thin if its graph is a thin set, i.e., if there exists a sequence of stopping times \((\tau_n)_{n=1}^{\infty}\) such that \( [\tau] \subseteq \bigcup_{n=1}^{\infty} [\tau_n] \). A random time is thick if it avoids any stopping time \( \sigma \), i.e., if \( P(\tau = \sigma < \infty) = 0 \) for every stopping time \( \sigma \).

**Corollary 3.6.** If \( \tau \) is a random time, there exists \( B \in F \) such that \( B \subseteq \{ \tau < \infty \} \), \( \tau_B \) is thin, \( \tau_B^c \) is thick and \( \tau = \tau_B \wedge \tau_B^c \) \( P \)-a.s.

**Proof.** Since the constant \( \infty \) is a thin (and a thick) time, the family \( G := \{ B \in F : B \subseteq \{ \tau < \infty \} \text{ and } \tau_B \text{ is thin} \} \)
contains \( \emptyset \), and so it is not empty. The rest of proof is the same as that of Corollary 3.4, mutatis mutandis.

**Remark 3.7.** Of course we could, as it is done in the literature, prove that \( A \) in Corollary 3.4 is \( P \)-a.s. unique and belongs to \( F_{\tau-} \), and that the decomposition of \( \tau \) as \( \alpha \wedge \beta \), where \( \alpha \) is an accessible stopping time, \( \beta \) is a totally inaccessible stopping time, and \( \alpha \vee \beta = \infty \) \( P \)-a.s., \( \beta \) is \( P \)-a.s. unique; we do not do this, as we have nothing new to add in this regard. Analogously for Corollary 3.6.

**Remark 3.8.** Notice that, to obtain the decomposition of Corollary 3.6, the authors in [1, Section 5] needs to prove some lemmas and theorems about dual optional projections, whereas we can more simply obtain it as a direct application of Theorem 2.3.

We will now list several more notable choices of \( G \) in Theorem 2.3, each one leading to a decomposition of a measure \( \mu \) into its \( G \)-atomic and its \( G \)-diffuse part.

1. Let \( F \) be the Borel \( \sigma \)-algebra on \( \mathbb{R}^n \) and \( d_H \) be the Hausdorff dimension. Moreover, let \( G := \{ G \in F : d_H(G) \in A \} \), where \( A \subseteq [0, n] \) is a set closed under supremum. Then, the family \( G \subseteq F \) is closed under countable unions due to the property of countable stability \( d_H(\bigcup_k A_k) = \sup_k d_H(A_k) \) for all \( A_k \in F \) of the Hausdorff dimension (see, e.g., [16, shortly before Proposition 2.3]). In particular, taking \( A = [0, n-1] \) we obtain that the \( G \)-diffuse part of \( \mu \) gives zero mass to every set of Hausdorff dimension at most \( n-1 \); such measures are considered in the theory of optimal transport, e.g. [35, Theorem 2.12], where such sets are called ‘small sets’.

2. Let \( F \) be the Borel \( \sigma \)-algebra on a locally compact Hausdorff space \( \Omega \), then the family \( G \) of Baire sets is closed under countable unions, and the \( G \)-atomic part of \( \mu \) is the ‘Baire contraction’ \( \nu \) of \( \mu \) of (see [18, Section 52, Theorem H]).
3. Let $F$ be the Borel $\sigma$-algebra of a locally compact Hausdorff space $\Omega$, then the family $G$ of outer regular sets with respect to a measure $\mu$ is closed under countable unions (see [18, Section 52, Theorem C]).

Now we point out that there are two operations which output suitable choices of $G$, i.e. a family of sets which is closed under countable unions. Then we list several choices of $G$ arising from the first, and then the second, such operation.

(i) Given a collection $\{G_i\}_{i \in I}$ of $G_i \subseteq F$ all closed under countable unions, then also $G := \cap_i G_i$ is closed under countable unions.

(ii) Given a family $H \subseteq F$, then the family $G := H^\sigma$ of all the countable unions of sets in $H$ is trivially closed under countable unions.

Here some choices of $G$ obtained via intersections.

4. Let $\{\mu_i\}_{i \in I}$ be measures and $G_i$ be the family of $\mu_i$-null sets, then $G := \cap_i G_i$ is the family of the polar sets of $\{\mu_i\}_{i \in I}$ (see, e.g., the definition in [6, Section 1.1]).

5. Let $\nu$ be a measure on the Borel $\sigma$-algebra $F$ of a separable topological space $\Omega$, $G_1 \subseteq F$ be the family of open sets and $G_2 \subseteq F$ be the family of $\nu$-null sets. Then, by setting $G := G_1 \cap G_2$ we find that $G$ is closed under countable unions, and the separability assumption implies that $G$ admits a maximum $\bar{G} \in G$ for the pointwise order (see Remark 2.4); $\bar{G}^c$ is thus the topological support of $\nu$ (see, e.g., [2, Section 12.3]).

Here some choices of $G$ obtained taking countable unions.

6. Let $F$ be the Borel $\sigma$-algebra of a topological space $\Omega$, and $H \subseteq F$ be the family of closed sets, then $G := H^\sigma$ is the family of the so called $F_\sigma$ sets, i.e., countable unions of closed sets.

7. Let $F$ be the Borel $\sigma$-algebra of a topological space $\Omega$, and $H \subseteq F$ be the family of nowhere dense sets, then $G := H^\sigma$ is the family of meagre sets (also called sets of first category; see, e.g., [2, Section 3.11]).

8. Let $F$ be the Borel $\sigma$-algebra on a metric space $\Omega$, $m \in \mathbb{N}$ and $H \subseteq F$ be the family of $m$-rectifiable sets, then $G := H^\sigma$ is the family of countably $m$-rectifiable sets (see, e.g., [17, Section 3.2.14]).

9. Let $H$ be the collection of graphs of strictly decreasing Lipschitz functions, then the family $G := H^\sigma$ is closed under countable unions, so the $G$-diffuse part of $\mu$ gives zero mass to the graphs of strictly decreasing Lipschitz functions. Such measures are considered in [25, Theorem 4.5].

We have also come across the following instances where a family $G$ closed under countable unions, and which thus admits a maximum for $\leq_\mu$, has been considered.

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(i) In the proof of [7, Lemma 2.3], the authors show that the family $\mathcal{B}$ is closed under countable unions, and conclude that there is a set $\Omega_u$ of maximal measure in $\mathcal{B}$, which is unique up to $P$-null sets. Clearly $\Omega_u$ is the set $\bar{G}$ of Theorem 2.3 when $\mu = P$ and $G = \mathcal{B}$.

(ii) The proof of [24, Proposition 2.4] states that there exists a set $\Omega_w \in \mathcal{F}$ satisfying the two dotted properties therein. Clearly $\Omega_w$ is the set $\bar{G}$ of Theorem 2.3 when $\mu = P$ and $G := \{G \in \mathcal{F} : P(G \cap \{Y > 0\}) = 0, \forall Y \in \mathcal{K}\}$: indeed, the first property immediately follows from the fact that $\bar{G} := \Omega_w \in G$, and the second property follows from the fact that $P(A \cap \bar{G}^c) = 0$ for all $A \in G$.

4 The decomposition theorem for vector measures

In this section we show how Theorem 2.3 can be extended to a very large class of vector measures. While, once identified the correct statement, the corresponding theorem is easy to prove, it is important because it will allow us to consider all of the most important and naturally occurring types of vector measures: the stochastic integral, the spectral measure, as well as any Banach-valued or $L^0(P)$-valued vector measure. In turn, this will allow us to derive an interesting decomposition of semimartingales in Section 5. We remark that while the notion of order between measures underlies Theorem 2.3 in the case of real-measures, this is not so for vector measures, since (as far as we know) one cannot define (even for $\mathbb{R}^2$-valued measures) an order on them such that two measures are singular with respect to such order if and only if they are concentrated on disjoint sets.

We first introduce the notion of vector measure on a measurable space $(\Omega, \mathcal{F})$. Let $V$ be a (Hausdorff) Topological Vector Space (TVS); a $V$-valued finitely-additive (vector) measure is a set function $\Theta : \mathcal{F} \to V$ such that $\Theta(A \cup B) = \Theta(A) + \Theta(B)$ for any disjoint $A, B \in \mathcal{F}$ (so in particular $\Theta(\emptyset) = 0$). A $V$-valued (vector) measure is a set function $\Theta : \mathcal{F} \to V$ which is countably additive, i.e., such that the identity

$$\Theta(\bigcup_{n=1}^{\infty} A_n) = \sum_{n=1}^{\infty} \Theta(A_n), \quad (4.1)$$

holds for any sequence $A_n \in \mathcal{F}, n \in \mathbb{N}$ of pairwise disjoint sets, where $\sum_{n=1}^{\infty} \Theta(A_n)$ is the limit of $\sum_{n=1}^{k} \Theta(A_n)$ as $k \to \infty$. The simplest interesting case of a vector measure is obtained taking for $V$ the real line $\mathbb{R}$, in which case we will speak of a real-measure. Working component by component, one can use them to treat $\mathbb{R}^n$-valued, and complex-valued, measures. Of special interest are Banach-valued measures, whose theory (both for finitely-additive and countably-additive measures) is presented in [14, 12]; aspects of the (much more complicated) case
of measures with values in a general TVS are considered in [13], whereas [30] considers the important case of metric linear spaces (like \( L^0(\mathcal{P}) \)).

**Definition 4.1.** We will say that a finite positive measure \( \theta \) on \((\Omega, \mathcal{F})\) is a **control** measure for a vector measure \( \Theta \) on \((\Omega, \mathcal{F})\) if \( \Theta \ll \theta \), i.e., if every \( \theta \)-null set is a \( \Theta \)-null set. We will say that \( \theta \) is equivalent to \( \Theta \) if they have the same null sets.

We warn the reader that the definition of control measure can change slightly from one source to another. Though we will not need it, we should anyway mention the fact that every \( \theta \)-null set is a \( \Theta \)-null set is equivalent to asking that \( \lim_{\theta(E)\to 0} \Theta(E) = 0 \) if \( \Theta \) is Banach-valued (see [12, Chapter 1, Section 2, Theorem 1]); it is probably true in general, but we were not able to locate a corresponding reference.

Given the definition of control measure, we can introduce a simple extension of Theorem 2.3 to vector measures.

**Theorem 4.2.** Let \( \mathcal{G} \subseteq \mathcal{F} \), and \( V \) be any (Hausdorff) topological vector space. Then,

1. Item 1 of Theorem 2.3 holds if the positive measures \( \mu, \mu_G, \mu_{\perp G} \) are replaced by finitely-additive \( V \)-valued (vector) measures \( \Theta, \Theta_G, \Theta_{\perp G} \).

2. If the \( V \)-valued countably-additive measure \( \Theta \) admits a control measure \( \theta \), and \( \mathcal{G} \) is non-empty and closed under countable unions, then there exists a set \( G_\theta \in \mathcal{G} \) such that \( G \setminus G_\theta \) is a \( \Theta \)-null set for every \( G \in \mathcal{G} \) and, up to \( \Theta \)-null sets, \( G_{\theta} \) is unique and equals \( G_\theta \).

**Proof.** The proof of item 1 is absolutely identical to the case of positive measures. As for the other item, applying Theorem 2.3 to \( \mathcal{G} \) and the control measure \( \theta \), we obtain a set \( G_\theta \in \mathcal{G} \) such that, for every \( G \in \mathcal{G} \), \( G_\theta \cap G = G \setminus G_\theta \) is a \( \theta \)-null set; as \( \theta \) is a control measure for \( \Theta \), by taking \( G_{\theta} := G_\theta \), we obtain the existence of \( G_{\theta} \), which is trivially unique up to \( \Theta \)-null sets. \( \square \)

**Remark 4.3.** Of course, for Theorem 4.2 to be of any use, we need to show that many vector measures admit a control measure; this is indeed the case, as we will now explain. The variation of a measure is an equivalent control measure for any real measure. More generally, an equivalent control measure (with additional properties) exists for any Banach-valued measure: see e.g. [12, Chapter 1, Section 2, Corollary 6] or [14, Chapter 4, Section 10, Lemma 5]. More generally still, [28, Corollary 2] shows that every measure with values in a metrizable locally-convex space admits an equivalent control measure. Moreover, even if the TVS \( L^0(\mathcal{P}) \) is neither locally convex nor locally bounded, it is true (though hard to prove) that any \( L^0 \)-valued measure is convexly-bounded, and using this one can prove that any such \( \Theta \) admits a control measure (see [26, Appendix B, Theorem B.2.2], or [34, Theoreme B]). This allows to apply, in Section 5, our Theorem 4.2 to the special case where the \( L^0 \)-valued measure is the stochastic integral with respect to a semimartingale (in this case a control measure can
actually be built somewhat explicitly, so its existence is easier to prove, as we will explain in Section 5). Finally, we will prove in Section 6 that any spectral measure admits a control measure.

To showcase the use of Theorem 4.2, we will now use it to recover Hahn-Jordan’s decomposition of a real measure \( \mu \); the proof is of course related to the classic proof found in [18, Section 29], since essentially we break up their proof in parts, as follows. Our upcoming Lemma 4.4 states that, given any set \( A \) of strictly positive measure, the family \( A_\geq \) of its non-trivial subsets of \( \mu \)-positive measure is non-empty; since \( A_\geq \) is closed under countable unions, Theorem 4.2 then shows that \( A_\geq \) has a maximum with respect to \( \leq_\mu \) (see Remark 2.1), from which it is then easy to prove Corollary 4.5.

Given a real measure \( \mu \) on a measurable space \((\Omega, \mathcal{F})\), we say that \( A \in \mathcal{F} \) is a positive set for \( \mu \) (or a \( \mu \)-positive set) if, for every \( B \in \mathcal{F} \) such that \( B \subseteq A \), we have \( \mu(B) \geq 0 \). Analogously, we say that \( A \in \mathcal{F} \) is a negative set for \( \mu \) if it is a positive set for \(-\mu\).

**Lemma 4.4.** Let \( \mu \) be a real measure on \((\Omega, \mathcal{F})\) and \( A \in \mathcal{F} \) such that \( \mu(A) > 0 \). Then, there exists \( \bar{A} \in \mathcal{F} \) such that \( \bar{A} \subseteq A \), \( \bar{A} \) is \( \mu \)-positive, and is not \( \mu \)-null (i.e., \( \mu(\bar{A}) > 0 \)).

**Proof.** If \( A_0 := A \) is positive then simply take \( \bar{A} := A_0 \). Otherwise, we proceed by induction. If \( A_{k-1}, k \in \mathbb{N}^* := \mathbb{N} \setminus \{0\} \) is not positive, let \( n_k \in \mathbb{N}^* \) be the smallest integer for which there exists \( B_k \in \mathcal{F} \) such that \( B_k \subseteq A_{k-1} \) and \( \mu(B_k) < -1/n_k \). Choose one such \( B_k \), and define \( A_k := A \setminus \bigcup_{i=1}^k B_i \); notice that \( A_k \subseteq A_{k-1} \) implies \( n_k \geq n_{k-1} \) for every \( k \in \mathbb{N} \). If there exists the smallest \( K \in \mathbb{N} \) such that \( A_K \) is positive we can take \( \bar{A} := A_K \), and the thesis follows since the \( \{B_k\}_k \) are disjoint and so

\[
\mu(\bar{A}) = \mu(A) - \sum_{k=1}^K \mu(B_k) > \mu(A) > 0. \tag{4.2}
\]

Assume no \( \{A_k\}_k \in \mathbb{N} \) is positive; then the thesis follows taking \( \bar{A} := A_\infty := \cap_{k=0}^\infty A_k \), as we will now prove. Since \( \{B_k\}_k \) are disjoint, the series \( \sum_{k=1}^\infty \mu(B_k) \) converges, so (4.2) holds with \( K = \infty \), and \( \lim_{k \to \infty} \mu(B_k) = 0 \) and thus \( \lim_{k \to \infty} 1/n_k = 0 \). To show that \( \bar{A} \) is positive, assume by contradiction that there exists \( N \subseteq \bar{A} \) such that \( \mu(N) < 0 \). Since \( \lim_{k \to \infty} 1/n_k = 0 \), there exists \( k \in \mathbb{N} \) such that \( \mu(N) < -1/(n_k - 1) \), and since \( N \subseteq \bar{A} \subseteq A_{k-1} \), the minimality property of \( n_k \) is violated by \( n_k - 1 \), a contradiction. \( \square \)

**Corollary 4.5** (Hahn-Jordan decomposition). If \( \mu \) is a real measure on \((\Omega, \mathcal{F})\), there exist a \( \mu \)-positive set \( G \in \mathcal{F} \) whose complement \( G^c \) is a \( \mu \)-negative set; such \( G \) is unique, up to \( \mu \)-null sets. In particular, setting

\[
\mu^+ := \mu(\cdot \cap G), \quad \mu^- := -\mu(\cdot \cap G^c)
\]

one obtains two finite positive mutually singular measures \( \mu^+ \) and \( \mu^- \) such that

\( \mu = \mu^+ - \mu^- \); such a decomposition of \( \mu \) is unique.
Proof. If $\mu$ or $-\mu$ is positive the theorem is trivially true, so let us assume otherwise. Let $\mathcal{G}$ be the family of positive sets, which is closed under countable unions; by Lemma 4.4, $\mathcal{G}$ is non-empty. Thus, Theorem 4.2 provides a positive set $\tilde{G} \in \mathcal{G}$ and unique decomposition $\mu = \mu_{\tilde{G}} + \mu_{\tilde{G}^c}$ where $\mu^+ := \mu_{\tilde{G}} := \mu(\cdot \cap \tilde{G})$ and $\mu^- := \mu_{\tilde{G}^c} := \mu(\cdot \cap \tilde{G}^c)$, such that $\mu^{-}(G) = 0$ for every $G \in \mathcal{G}$. Since $G$ is positive, $\mu^-$ is a positive measure; we will now show that $\tilde{G}^c$ is negative, concluding the proof of existence. Assume by contradiction that there exists $A \in \mathcal{F}$ such that $\mu^{-}(A) > 0$ then, by Lemma 4.4, there exists $\bar{A} \in \mathcal{F}$ such that $\bar{A} \subseteq A$, $\bar{A}$ is positive (i.e., $A \in \mathcal{G}$) and $\mu^{-}(\bar{A}) > 0$, a contradiction.

If $G$ and $\bar{G}$ are positive sets with negative complements, the sets $\tilde{G} \setminus \bar{G}$ and $\tilde{G} \setminus G$ are both positive and negative, and thus they are $\mu$-null, proving the uniqueness of $\bar{G}$, and thus that of $\mu = \mu^+ - \mu^-$ (since a $\mu^{-}$-null set on which $\mu^+$ is concentrated is $\mu$-positive and has $\mu$-positive complement). \hfill \square

5 The stochastic integral, and a decomposition of semimartingales

The most important example of a vector measure is represented by the stochastic integral with respect to a semimartingale, whose theory is developed, e.g., in the books [5], [29], [11]. Such integral $I_X : H \mapsto (H \cdot X)_\infty$ is a $L^0(\mathbb{P})$-valued measure, since it satisfies the (stochastic) dominated convergence theorem; conversely, the celebrated theorem by Bichteler and Dellacherie (for an elementary proof see [3]) essentially states that if the integral with respect to a process $X$ is a $L^0$-valued measure, then $X$ is a semimartingale. Given a semimartingale $X$, the stochastic integral can also be considered as the map $I_X : H \mapsto (H \cdot X)$, which takes values in the space $\mathcal{S}$ of semimartingales (so that $J_X(H)_\infty = I_X(H)$), and since $J_X$ also satisfies the dominated convergence theorem, it is also a ($\mathcal{S}$-valued) vector measure.

If $p \in [1, \infty)$, the previous statements admit the following much less well-known variant (see [36, Theorem 3.1]): the stochastic integral $I_X$ with respect to $X$ is a $L^p$-valued measure if and only if $X$ is a $\mathcal{S}^p$-semimartingale. In this case it is easy\footnote{By applying the Burkholder-David-Gundy inequality to the local martingale part $M$, this boils down to showing that $1_P : [M] = 0$ if and only if $1_P \cdot [M]^\gamma = 0$, where $P$ is a predictable set (see [36, Theorem 3.4]). That this holds can be proved easily, since $t \mapsto t^{1/2}$ is strictly increasing on $[0, \infty)$; otherwise one can apply the more delicate [27, Lemma 2.26], since $t \mapsto t^{1/2}$ and $t \mapsto t^2$ are absolutely continuous on every $[a, b] \subseteq [0, \infty)$.} to see that

$$\theta(ds \times d\omega) := (d|M|^\gamma_s + d|[A]_s) d\mathbb{P}(\omega)$$

is an equivalent control measure (for both $I_X$ and $J_X$, though now considered as having values in $L^p$ and $\mathcal{S}^p$), where $X = M + A$ is the canonical decomposition of $X$. This allows to easily construct somewhat explicitly a control measure for an arbitrary semimartingale $X$, as follows: by [29, Chapter 4, Theorem 34] there exists $Q \sim \mathbb{P}$ such that $X$ is an $\mathcal{S}^2(Q)$-semimartingale. Thus, $X$ can be
canonically decomposed as $X = M + A$ under $\mathbb{Q}$, and

\[
\theta(ds \times d\omega) := (d[M]_s \mathcal{P} + d[A]_s)d\mathbb{Q}(\omega)
\]

is clearly an equivalent control measure for both $I_X$ and $J_X$.

Among the possible applications of Theorem 4.2 to the stochastic integral, we show how to use it to obtain a decomposition that appears in [5, Theorem 4.4.5] in a more elementary way; we will also see how to obtain a slightly stronger result (see Remark 5.3). We need to introduce some definitions. As usual, we always consider not sets, but rather their equivalence classes: any set $F$ is identified with the process $1_F$, and so sets $F, \tilde{F}$ are identified (and we write $F = \tilde{F}$) if $1_F = 1_{\tilde{F}}$ a.s. for all $t$ i.e., if the set $\{1_F \neq 1_{\tilde{F}}\}$, which can more explicitly be written as

\[
\{\omega : 1_F(\omega, t) \neq 1_{\tilde{F}}(\omega, t) \text{ for some } t \geq 0\} = \{\omega : (\omega, t) \in F\Delta \tilde{F} \text{ for some } t \geq 0\},
\]

is $\mathbb{P}$-null; analogously $F \subseteq \tilde{F}$ actually means that $1_F \leq 1_{\tilde{F}}$ holds (i.e., it holds a.s. for all $t$), and this we call the pointwise order for processes (and so for subsets of $\Omega \times [0, \infty)$). Given a semimartingale $X$, we can then consider the $X$-a.e. order given by $F \leq_X \tilde{F}$ if the semimartingale $(1_F - \tilde{F}) \cdot X$ equals 0, and identity sets $F, \tilde{F}$ which are equal $X$-a.e., i.e. such that $\{1_F \neq 1_{\tilde{F}}\}$ is a $\mathbb{P}$-null set.

We will say that a set $F \subseteq \Omega \times [0, \infty)$ is thin if $F = \bigcup_n [\tau_n]$ for some stopping times $(\tau_n)_{n \in \mathbb{N}}$; we call such $F$ a thin predictable set if $F = \bigcup_n [\tau_n]$ for some predictable stopping times $(\tau_n)_{n \in \mathbb{N}}$, and a thin totally-inaccessible set if $F = \bigcup_n [\tau_n]$ for some totally-inaccessible stopping times $(\tau_n)_{n \in \mathbb{N}}$. We will use that the set of jumps $F := \{\Delta X \neq 0\}$ of a cadlag adapted process $X$ is a thin set, and it is a thin predictable set if $X$ is predictable (see, e.g., [19, Proposition 1.32] or [33, Theorem 3.1]). We will say that a cadlag process $X$ has predictable jump times if $\{\Delta X \neq 0\}$ is thin predictable, and that it has totally-inaccessible jump times if $\{\Delta X \neq 0\}$ is thin totally-inaccessible (this is easily seen to be equivalent to $X$ being quasi-left-continuous).

**Lemma 5.1.** If $Z$ is a semimartingale and $\tau$ a predictable stopping time then $1_{[\tau]} : Z = \Delta Z \cdot 1_{[\tau, \infty)}$

**Proof.** Let $(\tau_n)_{n \in \mathbb{N}}$ be an increasing sequence of stopping times converging to $\tau$ and such that $\tau_n < \tau$ on $\{\tau > 0\}$ for all $n$ (it exists: see [33, Corollary 2.1]). Applying to $\tau_n$ and to $\tau$ the well known identity $1_{[0, \sigma]} \cdot Z = Z_{\sigma\wedge \tau}$, valid for any stopping time $\sigma$, and subtracting the two we get $1_{[\tau_n, \tau]} \cdot Z = Z_{\tau\wedge \tau_n} - Z_{\tau_n\wedge \tau}$, and taking the limit as $n \to \infty$ by the stochastic dominated converge theorem we get the thesis. \hfill $\Box$

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10This terminology is justified by the fact that a set $F$ is thin and predictable if and only if it is a thin predictable set (we will not need this fact).
Corollary 5.2. Given a semimartingale $X$, there exists a thin predictable set $P$ such that
\[ X = Y + Z, \]  
where $Y$ satisfies $Y = 1_P \cdot Y$, and $Z$ has totally-inaccessible jump times. Moreover, the decomposition (5.2) is unique.

Proof. Let $G$ be the family of thin predictable sets; this is obviously non-empty and closed under countable unions. Since the vector measure $J_X$ admits a control measure (recall (5.1)), we can apply Theorem 4.2 and obtain the 'largest' thin predictable set $P := G_{J_X} \in G$, such that $1_{P \cap G} \cdot X = 0$ for every $G \in G$.

Let $Y := 1_P \cdot X$, $Z := 1_{P^c} \cdot X$. Since $1_P 1_P = 1_P$, and the stochastic integral satisfies $H \cdot (K \cdot X) = (HK) \cdot X$, we get
\[ Y = 1_P \cdot X = 1^2_P \cdot X = 1_P \cdot (1_P \cdot X) = 1_P \cdot Y; \]
moreover, if $\tau$ is a predictable stopping time then $[\tau] \in G$ and so $1_{P \cap [\tau]} \cdot X = 0$. Thus,
\[ 0 = 1_{P \cap [\tau]} \cdot X = (1_P 1_{[\tau]} - 1_P) \cdot X = 1_{[\tau]} (1_P \cdot X) = 1_{[\tau]} \cdot Z. \]

Lemma 5.1 gives $\Delta Z_{\tau} 1_{[\tau, \infty)} = 0$, or equivalently $\Delta Z_{\tau} = 0$, i.e., the set $J$ of jumps of $Z$ satisfies $J \cap [\tau] = \emptyset$. Writing $J = \cup_n [\tau_n]$ for some stopping times $(\tau_n)_{n \in \mathbb{N}}$ we get
\[ [\tau_n] \cap [\tau] \subseteq J \cap [\tau] = \emptyset \quad \text{for all } n \in \mathbb{N}, \]
so every $\tau_n$ is totally-inaccessible, i.e., $J$ is thin totally-inaccessible. The decomposition $X = Y + Z$ is unique, because given another one $X = \tilde{Y} + \tilde{Z}$ we get that the semimartingale $W := Z - \tilde{Z} = \check{Y} - Y$ is quasi-left-continuous (and so $1_P \cdot W = 0$), and satisfies $W = 1_P \cdot W$, so $W = 0$. \hfill \square

Remark 5.3. The thin set $P \in \mathcal{P}$ in Corollary 5.2 is only unique up to $X$-null sets. However, if $X$ is predictable then (it has finite variation and) the set of its jumps is a thin predictable set, which is obviously the smallest possible choice for the set $P$ (smallest with respect to the inclusion of sets; to be precise, we always consider equivalence classes of sets). It turns out that even if $X$ is a local martingale, or more generally any semimartingale, there always exists the smallest choice $\bar{P}$ for a set $P$ as in Corollary 5.2, and this observation is (we believe) new. This choice involves the concept of predictable support, developed in [19, Chapter 1, Section 2d], as we explain now.

Let $X$ be a general semimartingale, and consider its set of jumps $J := \{ \Delta X \neq 0 \}$. Write $J$ as $J = \cup_n [\tau_n]$ for some stopping times $(\tau_n)_{n \in \mathbb{N}}$. By Corollary 3.4, each $\tau_n$ admits an a.s. unique decomposition $\tau_n = \tau_n^a \wedge \tau_n^t$, where $\tau_n^a$ is an accessible stopping time and $\tau_n^t$ is a totally-inaccessible stopping time. Therefore, $J = J^a \cup J^t$, where $J^a := \cup_n [\tau_n^a]$ and $J^t := \cup_n [\tau_n^t]$ are disjoint (i.e., $1_{J^a \cap J^t} = 0$). The predictable support $J^P$ of the thin set $J$ is the smallest thin predictable set that contains $J^a$, whose existence is proved in [19, Proposition 2.34]. Clearly $J^P$ is the $\preceq$-smallest choice for the set $P$ in Corollary 5.2, i.e., if $\bar{P}$ is another choice for $P$ then $J^P \subseteq \bar{P}$ a.s. for all $t$. 

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Remark 5.4. Applying Corollary 5.2 to a purely-discontinuous process of finite variation $V = X$ we obtain two purely-discontinuous processes of finite variation $V^{jpr} := Y, V^{jti} := Z$, where $V^{jpr}$ has predictable jump times and $V^{jti}$ has totally-inaccessible jump times. Notice that, since a stopping time $\tau$ can be identified with the purely-discontinuous process of finite variation $V = 1_{[0,\tau]}$, which is predictable/accessible/totally inaccessible iff $\tau$ is such, this decomposition of $V$ subsumes the decomposition of $\tau$ that we obtained in Corollary 3.4.

Analogously, applying Corollary 5.2 to a purely-discontinuous local martingale $M = X$ gives two purely-discontinuous local martingales $M^{jpr} := Y, M^{jti} := Z$, where $M^{jpr}$ has predictable jump times and $M^{jti}$ has totally-inaccessible jump times. Applying Corollary 5.2 to a continuous semimartingale $X$ obviously gives $Y = 0, Z = X$. Thus, using that processes of finite variation and local martingales admit a decomposition in continuous + purely discontinuous parts, one obtains the decompositions in [23, Proposition 22.17] and [4, Theorem 7.9] as a corollary of Theorem 5.2.

6 The spectral measure

We now briefly introduce spectral measures; for more details we refer to [8, Chapter 9, Section 1], or to [31, Chapter 12, Section 17] (which calls them resolution of the identity). We then prove that they always admit an equivalent control measure, and thus Theorem 4.2 applies to them.

Given a separable Hilbert space $(\mathcal{H}, (\cdot, \cdot)_\mathcal{H})$ on the field $\mathbb{C}$ of complex numbers, let $\mathcal{B}(\mathcal{H})$ be the set of bounded linear transformations from $\mathcal{H}$ into $\mathcal{H}$. Given a measurable space $(\Omega, \mathcal{F})$, a spectral measure for $(\Omega, \mathcal{F}, \mathcal{H})$ is a function $E : \mathcal{F} \rightarrow \mathcal{B}(\mathcal{H})$ such that:

1. for each $\Delta \in \mathcal{F}$, $E(\Delta)$ is an orthogonal projection on a closed subspace of $\mathcal{H}$;
2. $E(\emptyset)$ equals the projection 0 on the origin $\{0_\mathcal{H}\}$, and $E(\Omega)$ equals the identity $1 = Id_\mathcal{H}$ on $\mathcal{H}$;
3. $E(\Delta_1 \cap \Delta_2) = E(\Delta_1) E(\Delta_2)$ for each $\Delta_1$ and $\Delta_2$ in $\mathcal{F}$;
4. For every $x \in H$ and $y \in H$, the set function $E_{x,y} : \mathcal{F} \rightarrow \mathbb{C}$ defined by

$$E_{x,y}(\cdot) := (E(\cdot)x, y)_{\mathcal{H}}$$

is a complex measure on $\mathcal{F}$.

Let us now point out the relationships between spectral measures and vector measures. Item 4 states that $E_{x,y}$ is a $\mathbb{C}$-valued measure, i.e., if $\{\Delta_n\}_{n \in \mathbb{N}}$ are pairwise disjoint sets of $\mathcal{F}$ then

$$E_{x,y} \left( \bigcup_{n \in \mathbb{N}} \Delta_n \right) = \sum_{n \in \mathbb{N}} E_{x,y}(\Delta_n). \quad (6.1)$$
It follows\textsuperscript{11} that, for each \(x \in \mathcal{H}\), \(\Delta \mapsto E(\Delta)x\) is a (countably additive) \(\mathcal{H}\)-valued (vector) measure on \((\Omega, \mathcal{F})\). Notice that \(E\) is \textit{not} a vector measure with values in the Banach space \(\mathcal{B}(\mathcal{H})\), since given pairwise disjoint sets \(\{\Delta_n\}_{n \in \mathbb{N}}\) of \(\mathcal{F}\), the series

\[
\sum_{n \in \mathbb{N}} E(\Delta_n)
\]

(6.2)
does \textit{not}\textsuperscript{12} generally converge \textit{in the norm topology} of \(\mathcal{H}\). On the other hand, (6.1) states that the series in (6.2) converges \textit{in the Strong Operator Topology}\textsuperscript{13} (SOT); thus \(E\) is a (special type of) \(\mathcal{B}(\mathcal{H})\)-valued measure, if \(\mathcal{B}(\mathcal{H})\) is endowed with the SOT (as such, it is a locally-convex TVS).

The definitions of null set, concentration, mutual singularity, and control measure, for a spectral measure \(E\) on \((\Omega, \mathcal{F}, \mathcal{H})\) are the same as for a vector measure; in particular, \(F \in \mathcal{F}\) is an \(E\)-null set if \(E(G) = 0\) for all \(G \subseteq F, G \in \mathcal{F}\), and a finite positive measure \(\mu\) is a control for \(E\) if every \(\mu\)-null set is an \(E\)-null set.

**Lemma 6.1.** Any spectral measure \(E\) admits an equivalent control measure, and \(\Delta \in \mathcal{F}\) is an \(E\)-null set if and only if \(E(\Delta) = 0\).

**Proof.** If \(E\) is a spectral measure for \((\Omega, \mathcal{F}, \mathcal{H})\), and \(\Delta \in \mathcal{F}\), each \(E(\Delta)\) is an orthogonal projection, so \(x - E(\Delta)x\) is orthogonal to \(E(\Delta)x\) for each \(x \in \mathcal{H}\), leading to the identity

\[
\|E(\Delta)x\|^2_{\mathcal{H}} = E_{x,x}(\Delta), \quad x \in \mathcal{H}.
\]

(6.3)

Equation (6.3) shows that the complex measure \(E_{x,y}\) is a finite positive measure when \(y = x\), and that \(\Delta\) is a \(E\)-null set if and only if \(E(\Delta) = 0\), i.e., if and only if \(E_{x,x}(\Delta) = 0\) for all \(x \in \mathcal{H}\). Let \(D := \{x_n\}_{n \in \mathbb{N}}\) be a dense subset of \(\mathcal{H}\). Then, since \((x, y) \mapsto E_{x,y}(\Delta)\) is continuous, \(E_{x,x}(\Delta)\) equals 0 for all \(x \in \mathcal{H}\) if and only if \(E_{x_n,x_n}(\Delta) = 0\) for all \(n \in \mathbb{N}\). Since \(E_{x,y}\) has total variation at most \(\|x\|_{\mathcal{H}}\|y\|_{\mathcal{H}}\) (see \[8, Chapter 9, Section 1, Lemma 1.9\]), the formula

\[
\mu := \sum_{n \in \mathbb{N}} t_n E_{x_n,x_n}, \quad \text{where } t_n := \frac{2^{-(n+1)}}{1 + \|x_n\|^2_{\mathcal{H}}} > 0,
\]

defines a positive finite measure (of mass at most 1). Since \(\mu(\Delta) = 0\) if and only if \(E_{x_n,x_n}(\Delta) = 0\) for all \(n \in \mathbb{N}\), \(\mu\) is an equivalent control measure for \(E\). \(\Box\)

**Remark 6.2.** It follows from Lemma 6.1 that, if \(E\) is a spectral measure for \((\Omega, \mathcal{F}, \mathcal{H})\), then Theorem 4.2 applies to \(\Theta := E\); of course the resulting \(\Theta(G_0 \cap \cdot)\)

\textsuperscript{11}Because \(E_{x,y}\) is a complex measure, and item 3 implies that the projections \(E(\Delta_1)\) and \(E(\Delta_2)\) commute (which is equivalent to saying that they are orthogonal, i.e., their ranges are orthogonal subspaces).

\textsuperscript{12}It only converges if all but finitely many of the \(E(\Delta_n)\) are 0 (since the norm of any projection is either 0 or 1, the partial sums of the series in (6.2) cannot form a Cauchy sequence).

\textsuperscript{13}The SOT is the topology on \(\mathcal{B}(\mathcal{H})\) generated by the family of seminorms \(\{p_x\}_{x \in \mathcal{H}}\), where \(p_x(T) := \|Tx\|_{\mathcal{H}}\) for \(T \in \mathcal{B}(\mathcal{H})\).
and \( \Theta(G \cap \cdot) \) are not spectral measures, since they do not satisfy the property that, once applied to \( \Omega \), they equal the identity (i.e., the projection on the whole of \( \mathcal{H} \)). They do however satisfy all the other properties which define a spectral measure.

**Remark 6.3.** The crucial result about spectral measures, which clarifies their importance, is the spectral theorem, which ‘can be used to answer essentially every question about normal operators’ [8, Page 255].

If \( N \) is a normal operator on a separable Hilbert space \( \mathcal{H} \), and \( E^N \) is the associated spectral measure given by the spectral theorem, then the fact that \( E^N \) admits a control measure \( \mu \) is stated and proved in [8, Chapter 9, Proposition 8.3]. We chose to present Lemma 6.1 anyway, since it is much more elementary.

## 7 Dependence of \( \mu_G \) and \( \mathcal{G}_\mu \) on \( \mu \) and \( \mathcal{G} \)

In this section we study the dependence on \( \mathcal{G} \subseteq \mathcal{F} \) (a non-empty family closed under countable unions) and \( \mu \) (a real measure) of \( \mathcal{G}_\mu \) (the \( \mathcal{G} \)-atomic support of \( \mu \)), \( \mu_G \) and \( \mu_\perp \) (the \( \mathcal{G} \)-atomic and \( \mathcal{G} \)-diffuse parts of \( \mu \)).

Let \( \mathcal{M} \) be the space of signed measures (i.e., real-valued measures) on \((\Omega, \mathcal{F})\) and \( \mathcal{M}_+ \subseteq \mathcal{M} \) be the convex cone of positive finite measures. We endow \( \mathcal{M} \) with the setwise (partial) order \( \geq \) and with the total variation norm \( \| \cdot \|_\mathcal{M} \) defined, for \( \mu, \nu \in \mathcal{M} \), as

\[
\mu \geq \nu \quad \text{if} \quad \mu(A) \geq \nu(A) \quad \text{for every} \quad A \in \mathcal{F} \quad \text{(i.e., we take} \quad \mathcal{M}_+ \quad \text{as the positive cone)}
\]

and

\[
\| \mu \|_\mathcal{M} := |\mu|(\Omega), \quad \text{where} \quad |\mu| \in \mathcal{M}_+ \quad \text{is the variation of} \quad \mu \in \mathcal{M}.
\]

We recall that \((\mathcal{M}, \geq, \| \cdot \|_\mathcal{M})\) is an AL-space (see [2, Theorems 9.51 and 10.56]), and it is order complete ([2, Theorems 10.53, 10.56, Lemma 8.14]); in particular \( |\mu| = \mu \vee (-\mu) \), and there exists the supremum \( \bigvee_{i \in I} \mu^i \) of any net \( \{\mu_i\}_{i \in I} \subseteq \mathcal{M} \) bounded from above; moreover, [2, Theorem 10.53, item (2)] shows that if \( \{\mu_i\}_{i \in I} \) is increasing then

\[
(\bigvee_{i \in I} \mu^i)(A) = \sup_{i \in I} (\mu^i(A)), \quad A \in \mathcal{F}; \tag{7.1}
\]

obviously (7.1) fails without the monotonicity assumption, since for general measures \( \nu_1, \nu_2 \) the supremum \( \nu_1 \vee \nu_2 := \sup\{\nu_1, \nu_2\} \) satisfies

\[
(\nu_1 \vee \nu_2)(F) = \sup\{\nu_1(B) + \nu_2(F \setminus B) : B \in \mathcal{F}, B \subseteq F\},
\]

and so does not normally equal \( \sup\{\nu_1(F), \nu_2(F)\} \). Define the set of \( \mathcal{G} \)-atomic measures to be

\[
\mathcal{A}(\mathcal{G}) := \{\mu \in \mathcal{M} : \exists G \in \mathcal{G} \text{ such that } G^c \text{ is } \mu\text{-null}\}, \tag{7.2}
\]

so that \( \mu \in \mathcal{M} \) belongs to \( \mathcal{A}(\mathcal{G}) \) iff \( \mu \) is concentrated on some \( G \in \mathcal{G} \). Then, define the set of \( \mathcal{G} \)-diffuse measures to be

\[
\mathcal{D}(\mathcal{G}) := \{\mu \in \mathcal{M} : \text{every } G \in \mathcal{G} \text{ is } \mu\text{-null}\}, \tag{7.3}
\]

and recall that \( F \in \mathcal{F} \) is \( \mu \)-null iff \( |\mu|(F) = 0 \).
We now recall some facts about projections, in three different contexts. A linear operator $P$ on a vector space $B$ is called a projection if $P^2 = P$, i.e., if $Pb = b$ for every $b$ in the range of $P$. Two subspaces $V, W$ of $B$ are said to be complemented if $V + W = B$ and $V \cap W = \{0\}$, in which case we write $B = V \oplus W$, and each $b \in B$ decomposes as $v + w$ for a unique $(v, w) \in V \times W$, which defines the projections $b \mapsto v$ and $b \mapsto w$ of $B$ on $V$ and $W$. We say that $x$ in a Banach space $(B, \| \cdot \|_B)$ is orthogonal to $y \in B$ (in the sense of James, see [20, 21]), and write $x \perp y$, if $\|x\|_B = \inf\{\|x + ty\|_B : t \in \mathbb{R}\}$; if $C, D \subseteq B$, then we say that they are orthogonal, and write $C \perp D$, if $c \perp d$ for each $c \in C, d \in D$. Given a subset $C$ of a metric space $(B, d_B)$, if for every $b \in B$ there exists a unique minimizer $P_C(b)$ of the distance $C \ni c \mapsto d_B(b, c)$ from $C$, then $P_C : B \to C$ is called the unique nearest-point map of $C$ (often called the metric projection on $C$), and $C$ is said to be a Chebyshev set.

A subset $S$ of a Riesz space $R$ is called solid if $|y| \leq |x|$ and $x \in S$ imply $y \in S$. A solid vector subspace of a Riesz space is called an ideal. A net $(x_i)_{i \in I}$ in a Riesz space $R$ converges in order (or is order convergent) to some $x \in R$, written $x_i \xrightarrow{\text{ord}} x$, if there is a net $(y_i)_{i \in I}$ (with the same directed set $I$) satisfying $y_i \downarrow 0$ and $|x_i - x| \leq y_i$ for each $i$. A function $f : R \to S$ between two Riesz spaces is order continuous if $x_i \xrightarrow{\text{ord}} x$ in $R$ implies $f(x_i) \xrightarrow{\text{ord}} f(x)$ in $S$. A subset $S \subseteq R$ is order closed if $(x_i)_{i \in I} \subseteq S$ and $x_i \xrightarrow{\text{ord}} x$ imply $x \in S$. An order closed ideal is called a band. If $S \subseteq R$ is non-empty, then its disjoint complement is

$$S^d := \{x \in R : |x| \wedge |y| = 0 \text{ for all } y \in S\},$$

and it is always a band, as it follows immediately from the order continuity of the lattice operations. A band $B$ in a Riesz space $R$ is a projection band if $R = B \oplus B^d$, in which case the projections on $B$ and $B^d$ are called band projections.

**Proposition 7.1.** If $\mathcal{G} \subseteq \mathcal{F}, \mathcal{G} \neq \emptyset$ is closed under countable unions, then $\mathcal{A}(\mathcal{G}), \mathcal{D}(\mathcal{G}) \subseteq \mathcal{M}$ are closed complemented and orthogonal subspaces, Chebyshev sets, and projection bands, each one the disjoint complement of the other.

The maps $A : \mu \mapsto \mu_{\mathcal{G}}$ and $D : \mu \mapsto \mu_{\mathcal{D}}$ are the projections of $\mathcal{M}$ on $\mathcal{A}(\mathcal{G})$ and $\mathcal{D}(\mathcal{G})$, and they are positive linear contractions, unique nearest-point maps, and order-continuous homomorphisms of Riesz spaces, and satisfy $|A\mu| \leq |\mu|, |D\mu| \leq |\mu|$ for every $\mu \in \mathcal{M}$.

Throughout our proofs, we will sometimes denote with $G_\alpha \in \mathcal{G}_\alpha$ a representative of the equivalence class $\mathcal{G}_\alpha$ (for any $\alpha \in \mathcal{M}$): while it is normally best to ignore the difference between the two (e.g. to write $G_\mu \subseteq H_\mu$ $\mu$-a.e. to mean $G_\mu \subseteq H_\mu$ $\mu$-a.e., or equivalently $G_\mu \leq \mu$ $H_\mu$, i.e. $G_\mu \subseteq H_\mu$), we will occasionally be picky about this difference, when it makes things clearer.

**Proof of Proposition 7.1.** Obviously two measures $\mu, \nu \in \mathcal{M}$ are singular iff $\exists F \in \mathcal{F}$ such that $|\mu|(F^c) = 0 = |\nu|(F)$, and this is equivalent to $|\mu| \wedge |\nu| = 0$. Thus $\nu \in \mathcal{D}(\mathcal{G})$ iff $\nu$ is singular with every measure concentrated on a set of $\mathcal{G}$, i.e. $\mathcal{D}(\mathcal{G}) = \mathcal{A}(\mathcal{G})^d$; in particular $\mathcal{D}(\mathcal{G})$ is a band.
That \( \mathcal{A}(\mathcal{G}) \) is a vector subspace follows simply from the fact that, if \( \mu, \nu \) are concentrated on \( G_\mu, G_\nu \in \mathcal{G} \), then \( \mu + \nu \) is concentrated on \( G_\mu \cup G_\nu \in \mathcal{G} \), and if \( t \in \mathbb{R} \) then \( t \mu \) is concentrated on \( G_\mu \). Analogously, \( \mathcal{D}(\mathcal{G}) \) is a vector subspace, since \( |\mu|(G) = |\nu|(G) \) and \( t \in \mathbb{R} \) imply \( |\mu + \nu|(G) = |t \mu|(G) \). Since \( \mathcal{A}(\mathcal{G}) \) and \( \mathcal{D}(\mathcal{G}) \) are vector subspaces, \( A, D \) are linear. If \( \mu_n \to \mu \) and \( \mu_n \) is concentrated on \( G_n \in \mathcal{G} \) then \( \mu \) is concentrated on \( G := \bigcup_n G_n \in \mathcal{G} \), and so \( \mathcal{A}(\mathcal{G}) \) is (topologically) closed. If \( \mu_n \to \mu \) then \( |\mu_n|(G) = 0 \) for every \( n \) implies \( |\mu|(G) = 0 \), and so \( \mathcal{D}(\mathcal{G}) \) is (topologically) closed. That \( \mathcal{A}(\mathcal{G}) \) and \( \mathcal{D}(\mathcal{G}) \) are complemented follows from Theorem 2.3, since the existence of the decomposition for every \( \mu \in \mathcal{M} \) shows that \( \mathcal{M} = \mathcal{A}(\mathcal{G}) + \mathcal{D}(\mathcal{G}) \), and its uniqueness shows that \( \mathcal{A}(\mathcal{G}) \cap \mathcal{D}(\mathcal{G}) = \{0\} \). Since \( \mathcal{A}(\mathcal{G}), \mathcal{D}(\mathcal{G}) \) are closed complemented subspaces of \( \mathcal{M} \), then \( A, D \) are continuous (see [2, Theorem 6.47]).

Since \( \mu \geq 0 \) implies \( \mu G, \mu^2 G \geq 0 \), then \( A, D \) are positive; since they are linear, they are increasing. To prove that \( \mathcal{D}(\mathcal{G})^d = \mathcal{A}(\mathcal{G}) \), recall that in every order complete Riesz space every band \( B \) satisfies \( (B^d)^d = B \) (see [2, Lemma 8.4, Theorem 8.19]); since by definition \( \mathcal{D}(\mathcal{G}) = \mathcal{A}(\mathcal{G})^d \), to conclude that \( \mathcal{D}(\mathcal{G})^d = \mathcal{A}(\mathcal{G}) \) we only need to show that \( \mathcal{A}(\mathcal{G}) \) is a band (and thus \( \mathcal{D}(\mathcal{G}), \mathcal{A}(\mathcal{G}) \) are projection bands). Since it is clearly an ideal of \( \mathcal{M} \), we only need to show that \( 0 \leq \mu^i \uparrow \mu \) and \( \{\mu^i\}_{i \in I} \subseteq \mathcal{A}(\mathcal{G}) \) imply \( \mu \in \mathcal{A}(\mathcal{G}) \), as this implies that \( \mathcal{A}(\mathcal{G}) \) is order closed (see [2, First 6 lines of section 8.9]) and thus a band. Since \( \nu \in \mathcal{M} \) is in \( \mathcal{A}(\mathcal{G}) \) iff \( \nu = \mu G, \) and \( A, D \) are increasing, we can equivalently show that \( 0 \leq \mu^i \uparrow \mu \) implies \( \bigvee_{i \in I} \mu^i = \mu G \) (and so \( 0 \leq \mu^i \uparrow \mu \)). So, assume \( 0 \leq \mu^i \uparrow \mu \); as \( \mu \geq \mu^i \) implies \( \mu G \geq \mu^i G \), then \( \{\mu^i G\}_{i \in I} \) is bounded from above, so it admits a supremum, which by definition satisfies

\[
\mu G \geq \bigvee_{i \in I} \mu^i G. \tag{7.4}
\]

To prove the opposite inequality, since \( G_{\mu^i} \) is the \( \leq_{\mu^i} \)-maximum of \( \mathcal{G} \) and \( G_{\mu} \in \mathcal{G} \), we have that

\[
\mu^i G(F) = \mu^i(F \cap G_{\mu^i}) \geq \mu^i(F \cap G_{\mu}), \quad \forall \ F \in F, \tag{7.5}
\]

so that

\[
(\bigvee_{i \in I} \mu^i G)(F) \geq \mu^i G(F) \geq \mu^i(F \cap G_{\mu}), \quad \forall \ F \in F, \tag{7.6}
\]

and taking the supremum over \( i \in I \) and using (7.1) we conclude that, for every \( F \in F \),

\[
(\bigvee_{i \in I} \mu^i G)(F) \geq \sup_{i \in I} (\mu^i(F \cap G_{\mu})) = (\bigvee_{i \in I} \mu^i)(F \cap G_{\mu}) = \mu(F \cap G_{\mu}) = \mu G(F), \tag{7.7}
\]

proving that indeed \( \bigvee_{i \in I} \mu^i G = \mu G \), and so \( \mu^i G \uparrow \mu G \).

If \( P \) is any projection on a projection band of a Riesz space \( R \), then \( P \) is an order-continuous homomorphism of Riesz spaces, and satisfies \( |Pr| \leq |r| \) for every \( r \in B \), and so such are \( A, D \); the inequalities \( |A\mu| \leq |\mu|, |D\mu| \leq |\mu| \) trivially hold anyway, and they imply that \( A, D \) are contractions. Since \( |\mu| \geq |\mu| \) implies \( |\mu + \nu| = |\mu| + |\nu| \) (see [2, Theorem 8.12]), and the norm on \( \mathcal{M} \) is given by \( \|\mu\|_{\mathcal{M}} = |\mu|(\Omega) \), we get that \( \|\mu + \nu\|_{\mathcal{M}} = \|\mu\|_{\mathcal{M}} + \|\nu\|_{\mathcal{M}} \) for
every \( \mu \in A(\mathcal{G}) \), \( \nu \in D(\mathcal{G}) \), and so \( A, D \) are the nearest point projections on \( A(\mathcal{G}), D(\mathcal{G}) \), and these are Chebyshev sets which are orthogonal.

We now study the order properties of \( \mathcal{G} \mapsto \mu_\mathcal{G} \) and \( \mathcal{G} \mapsto \mathcal{G}_\mu \). Let \( \mathbb{S} := \{ \mathcal{G} : \mathcal{G} \subseteq \mathcal{F} \} \) be the set of all subfamilies of \( \mathcal{F} \), and define its subset

\[
\mathcal{G} := \{ \mathcal{G} \in \mathbb{S} : (G_n)_{n \in \mathbb{N}} \subseteq \mathcal{G} \Rightarrow \bigcup_{n \in \mathbb{N}} G_n \in \mathcal{G} \}
\]  

(7.8)

of all subfamilies of \( \mathcal{F} \) which are closed under countable unions. We endow \( \mathbb{S} \) and \( \mathcal{G} \) with the order of inclusion of sets. Let \( \mathcal{G} \in \mathbb{S} \). Since the intersection of families closed under countable unions is closed under countable unions, the intersection of every \( \mathcal{H} \in \mathcal{G} \) such that \( \mathcal{G} \subseteq \mathcal{H} \) is the smallest element of \( \mathcal{G} \) that contains \( \mathcal{G} \), which we denote by \( \mathcal{G}^\sigma \). Notice that \( \mathcal{G}^\sigma \) can also be built ‘from the inside’, as

\[
\mathcal{G}^\sigma = \{ \bigcup_{n \in \mathbb{N}} G_n : G_n \in \mathcal{G}, n \in \mathbb{N} \},
\]  

(7.9)

and

\[
\mathcal{G} = \{ \mathcal{G} \in \mathbb{S} : \mathcal{G} = \mathcal{G}^\sigma \}.
\]  

(7.10)

For an arbitrary family \( \{ \mathcal{G}^i \}_{i \in I} \subseteq \mathcal{G} \), its infimum \( \wedge_{i \in I} \mathcal{G}^i \) and supremum \( \vee_{i \in I} \mathcal{G}^i \) in \( \mathcal{G} \) are given by

\[
\wedge_{i \in I} \mathcal{G}_i = \bigcap_{i \in I} \mathcal{G}^i
\]  

(7.11)

\[
\vee_{i \in I} \mathcal{G}_i = (\bigcup_{i \in I} \mathcal{G}^i)^\sigma.
\]  

(7.12)

Since (7.9) expresses \( \mathcal{G}^\sigma \) as the family of countable unions of sets in \( \mathcal{G} \), we get that

\[
(\bigcup_{i \in I} \mathcal{G}^i)^\sigma = \{ \bigcup_{i \in J} G_i : J \subseteq I \text{ is countable}, J \neq \emptyset \text{ and } G_i \in \mathcal{G}^i \},
\]  

(7.13)

and so

\[
\vee_{i \in I} \mathcal{G}_i = \{ \bigcup_{i \in J} G_i : J \subseteq I \text{ is countable}, J \neq \emptyset \text{ and } G_i \in \mathcal{G}^i \}.
\]  

(7.14)

For instance, if \( \mathcal{G}, \mathcal{H} \in \mathcal{G} \) then

\[
\mathcal{G} \lor \mathcal{H} = \{ G, H, G \cup H : G \in \mathcal{G}, H \in \mathcal{H} \};
\]  

(7.15)

and in particular if \( \emptyset \in \mathcal{G} \cap \mathcal{H} \) then \( \mathcal{G} \lor \mathcal{H} = \{ G \cup H : G \in \mathcal{G}, H \in \mathcal{H} \} \).

**Proposition 7.2.** If \( \mu \in \mathcal{M} \), the map \( \mathcal{G} \mapsto \mathcal{G}_\mu \) is increasing, i.e., if \( \mathcal{G} \subseteq \mathcal{H} \) then \( \mathcal{G}_\mu \subseteq \mathcal{H}_\mu \), \( \mu \)-a.e. and it satisfies the following property: if \( \{ \mathcal{G}^i \}_{i \in I} \subseteq \mathcal{G} \), then there exists \( J \subseteq I \) countable such that

\[
(\vee_{i \in I} \mathcal{G}^i)_\mu = (\bigvee_{i \in J} \mathcal{G}^i)_\mu = \bigcup_{i \in J} \mathcal{G}^i_\mu.
\]  

(7.16)

**Proof.** Since \( \mathcal{G}_\mu = \mathcal{G}[\mu] \), we can assume w.l.o.g. that \( \mu \in \mathcal{M}_+ \). Monotonicity is straightforward: as \( \mathcal{H}_\mu \) is the \( \leq_\mu \)-maximum of \( \mathcal{H} \) and \( \mathcal{G}_\mu \in \mathcal{G} \subseteq \mathcal{H} \), then also \( \mathcal{G}_\mu \subseteq \mathcal{H}_\mu \), \( \mu \)-a.e. To prove (7.16) we first show that \( (\vee_{i \in I} \mathcal{G}^i)_\mu = \bigcup_{i \in I} \mathcal{G}^i_\mu \).
holds for any countable set $J$. Indeed (7.14) shows that $\bigcup_{i\in J} G^i_{\mu} \subseteq \bigvee_{i\in J} G^i_{\mu}$, and that if $G \in \bigvee_{i\in J} G^i_{\mu}$ then $G = \bigcup_{i\in J} G_i$ with $J_i \subseteq J$ and $G_i \in G^i_{\mu}$ and so $G \subseteq \bigcup_{i\in J} G^i_{\mu} \subseteq \bigvee_{i\in J} G^i_{\mu}$, $\mu$-a.e. To complete the proof, we now want to prove that there exists $J \subseteq I$ countable such that $(\bigvee_{i\in J} G^i_{\mu}) = \bigcup_{i\in J} G^i_{\mu}$. By (7.14), since the set $(\bigvee_{i\in J} G^i_{\mu})$ belongs to $\bigvee_{i\in J} G^i_{\mu}$, it is of the form $\bigcup_{i\in J} G_i$ for some $G_i \in G^i_{\mu}$ and countable $J \subseteq I$. Since $G \mapsto G^0$ and $G \mapsto G_{\mu}$ are increasing, (7.11) and $J \subseteq I$ imply that $(\bigvee_{i\in J} G^i_{\mu}) \geq (\bigvee_{i\in J} G^i_{\mu})_{\mu}$. Since $J$ is countable, $(\bigvee_{i\in J} G^i_{\mu})_{\mu} = \bigcup_{i\in J} G^i_{\mu}$, and putting everything together gives

$$\bigcup_{i\in J} G^i_{\mu} = (\bigvee_{i\in J} G^i_{\mu})_{\mu} \geq (\bigvee_{i\in J} G^i_{\mu})_{\mu} = \bigcup_{i\in J} G^i_{\mu},$$

implying that the inclusion $\supseteq$ holds with equality. \hfill \Box

**Proposition 7.3.** If $\mu \in M_+$, the map $G \in \mathcal{G} \mapsto \mu_G \in M_+$ is

(i) increasing, i.e., if $G \subseteq H$ then $\mu_G \geq \mu_H$;

(ii) continuous from below with respect to the order, i.e., if $\{G^i\}_{i\in I} \subseteq \mathcal{G}$ then

$$\mu_{\bigvee_{i\in I} G^i} = \bigvee_{i\in I} \mu_G^i. \quad (7.17)$$

**Proof.** To prove monotonicity, recall that $\mu_G(\cdot) := \mu(\cdot \cap G)$ and $\mu_H(\cdot) := \mu(\cdot \cap H)$ with $G \in \mathcal{G}$ and $H \in \mathcal{H}$. Moreover, $H$ is the maximum in $\mathcal{H}$ and so $\mu(F \cap H) = \mu(F \cap H)$ for every $F \in \mathcal{F}$ and $H \in \mathcal{H}$, and in particular for $H = G \in \mathcal{G}, \subseteq \mathcal{H}$; thus $\mu_G \leq \mu_H$.

Let us prove (7.17). By the monotonicity of $G \mapsto G$ we have that

$$\mu_{\bigvee_{i\in I} G^i} \geq \bigvee_{i\in I} \mu_G^i =: \nu;$$

let us prove the opposite inequality. Notice that, since $\nu \ll \mu$ holds (e.g., because $\nu \leq \mu$), we do not need to be careful about $\mu$-null sets, so we can improperly write $F \subseteq G^i_{\mu}$ etc. without causing problems. By Proposition 7.2, there exists $J \subseteq I$ countable and such that $(\bigvee_{i\in J} G^i_{\mu}) = \bigcup_{i\in J} G^i_{\mu}$; let $(j_n)_{n\in \mathbb{N}}$ be an enumeration of $J$. Since

$$\mu_{\bigvee_{i\in J} G^i} := \mu(\cdot \cap (\bigvee_{i\in J} G^i)) = \mu(\cdot \cap (\bigcup_{i\in J} G^i)), \quad \nu \geq \mu_{\bigvee_{i\in J} G^i}$$

showing $\nu \geq \mu_{\bigvee_{i\in J} G^i}$ is equivalent to showing that $\nu(F) \geq \mu(F)$ for every $F \subseteq \bigcup_{i\in J} G^i_{\mu}, F \in \mathcal{F}$. Notice that we can write an arbitrary $F \subseteq (\bigvee_{i\in J} G^i_{\mu})_\mu = \bigcup_{n\in \mathcal{N}} G^i_{\mu} \cap F \in \mathcal{F}$ as the countable union of disjoint sets $F^n \subseteq G^i_{\mu} \cap F$, $F \in \mathcal{F}$ defined as

$$F^0 := F \cap G^i_{\mu}, \quad F^{n+1} := (F \cap G^i_{\mu}) \setminus (\bigcup_{k=1}^n G^i_{\mu}), \quad n \in \mathbb{N}. \quad (7.18)$$

Hence, since $\nu \geq \mu_{\bigvee_{i\in J} G^i} = \mu(\cdot \cap G^i_{\mu})$ gives $\nu(F) \geq \mu(F)$ for all $F \subseteq G^i_{\mu}, F \in \mathcal{F}$, we get

$$\nu(F) = \sum_{n=1}^\infty \nu(F^n) \geq \sum_{n=1}^\infty \mu(F^n) = \mu(F), \quad (7.19)$$

and the proof is complete. \hfill \Box


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