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A note on time-dependent additive functionals

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Abstract. This note develops shortly the theory of time-inhomogeneous additive functionals and is a useful support for the analysis of time-dependent Markov processes and related topics. It is a significant tool for the analysis of BSDEs in law. In particular we extend to a non-homogeneous setup some results concerning the quadratic variation and the angular bracket of Martingale Additive Functionals (in short MAF) associated to a homogeneous Markov processes.

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1 Introduction

The notion of Additive Functional of a general Markov process is due to E.B Dynkin and has been studied since the early '60s by the Russian, French and American schools of probability, see for example [8], [16], [4]. A mature version of the homogeneous theory may be found for example in [7], Chapter XV. In that context, given a probability μ on some state space E, P^{μ} denotes the law of a time-homogeneous Markov process with initial law μ . An **Additive Functional** is a process $(A_t)_{t>0}$ defined on a canonical space, adapted to the canonical filtration such that for any $s \leq t$ and μ , $A_{s+t} = A_s + A_t \circ \theta_s$ P^{μ}-a.s., where θ is the usual shift operator on the canonical space.

If moreover A is under any law P^{μ} a martingale, then it is called a Martingale Additive Functional (MAF). The quadratic variation and angular bracket

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of a MAF were shown to be AFs in [7]. We extend this type of results to a more general definition of an AF which is closer to the original notion of Additive Functional associated to a stochastic system introduced by E.B. Dynkin, see [9] for instance.

Our setup will be the following. We consider a canonical Markov class $(P^{s,x})_{(s,x)\in[0,T]\times E}$ with time index $[0, T]$ and state space E being a Polish space. For any $(s, x) \in$ $[0, T] \times E$, P^{s,x} corresponds to the probability law (defined on some canonical filtered space $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t \in [0,T]})$ of a Markov process starting from point x at time s. On (Ω, \mathcal{F}) , we define a non-homogeneous Additive Functional (shortened by AF) as a real-valued random-field $A := (A_u^t)_{0 \le t \le u \le T}$ verifying the two following conditions.

- 1. For any $0 \le t \le u \le T$, A_u^t is $\mathcal{F}_{t,u}$ -measurable;
- 2. for any $(s, x) \in [0, T] \times E$, there exists a real cadlag $(\mathcal{F}^{s,x}_t)_{t \in [0,T]}$ -adapted process $A^{s,x}$ (taken equal to zero on $[0,s]$ by convention) such that for any $x \in E$ and $s \le t \le u$, $A_u^t = A_u^{s,x} - A_t^{s,x}$ P^{s,x} a.s.

Where $\mathcal{F}_{t,u}$ is the σ -field generated by the canonical process between time t and u, and $\mathcal{F}^{s,x}_t$ is obtained by adding the P^{s,x} negligible sets to \mathcal{F}_t . $A^{s,x}$ will be called the **cadlag version of** A under $P^{s,x}$. If for any (s, x) , $A^{s,x}$ is a $(P^{s,x}, (\mathcal{F}_t)_{t\in[0,T]})$ -square integrable martingale then A will be called a square integrable Martingale Additive Functional (in short, square integrable MAF).

The main contributions of the paper are essentially the following. In Section 3, we recall the definition and prove some basic results concerning canonical Markov classes. In Section 4, we start by defining an AF in Definition 4.1. In Proposition 4.2, we show that if $(M_u^t)_{0 \le t \le u \le T}$ is a square integrable MAF, then there exists an AF $([M]^t_u)_{0 \le t \le u \le T}$ which for any $(s, x) \in [0, T] \times E$, has $[M^{s,x}]$ as cadlag version under $P^{s,x}$. Corollary 4.9 states that given two square integrable MAFs $(M_u^t)_{0 \le t \le u \le T}$, $(N_u^t)_{0 \le t \le u \le T}$, there exists an AF, denoted by $(\langle M, N\rangle_u^t)_{0 \le t \le u \le T}$, which has $\langle M^{s,x}, N^{s,x} \rangle$ as cadlag version under $P^{s,x}$. Finally, we prove in Proposition 4.14 that if M or N is such that for $P^{s,x}$, its cadlag version under $P^{s,x}$, its angular bracket is absolutely continuous with respect to some continuous non-decreasing function V , then there exists a Borel function v such that for any (s, x) , $\langle M^{s,x}, N^{s,x} \rangle = \int_s^{\langle \vee s} v(r, X_r) dV_r$.

The present note constitutes a support for the authors, in the analysis of deterministic problems related to Markovian type backward stochastic differential equations where the forward process is given in law, see e.g. [2]. Indeed, when the forward process of the BSDE does not define a stochastic flow (typically if it is not the strong solution of an SDE but only a weak solution), we cannot exploit the mentioned flow property to show that the solution of the BSDE is a function of the forward process, as it is usually done, see Remark 5.35 (ii) in [17] for instance.

2 Preliminaries

In the whole paper we will use the following notions, notations and vocabulary.

A topological space E will always be considered as a measurable space with its Borel σ -field which shall be denoted $\mathcal{B}(E)$ and if S is another topological space equipped with its Borel σ -field, $\mathcal{B}(E, S)$ will denote the set of Borel functions from E to S.

Let (Ω, \mathcal{F}) , (E, \mathcal{E}) be two measurable spaces. A measurable mapping from (Ω, \mathcal{F}) to (E, \mathcal{E}) shall often be called a **random variable** (with values in E), or in short r.v. If T is some set, an indexed set of r.v. with values in $E, (X_t)_{t\in\mathbb{T}}$ will be called a **random field** (indexed by T with values in E). In particular, if T is an interval included in R_+ , $(X_t)_{t\in\mathcal{T}}$ will be called a **stochastic process** (indexed by T with values in E). Given a stochastic process, if the mapping

$$
\begin{array}{rcl}\n(t,\omega) & \longmapsto & X_t(\omega) \\
(T \times \Omega, \mathcal{B}(T) \otimes \mathcal{F}) & \longrightarrow & (E,\mathcal{E})\n\end{array}
$$

is measurable, then the process $(X_t)_{t\in\mathcal{T}}$ will be called a **measurable process** (indexed by T with values in E).

On a fixed probability space (Ω, \mathcal{F}, P) , for any $p \geq 1$, L^p will denote the set of random variables with finite p -th moment. Two random fields (or stochastic processes) $(X_t)_{t\in\mathbb{T}}$, $(Y_t)_{t\in\mathbb{T}}$ indexed by the same set and with values in the same space will be said to be modifications (or versions) of each other if for every $t \in T$, $P(X_t = Y_t) = 1$.

A probability space equipped with a right-continuous filtration $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t \in \mathbb{T}}, P)$ will be called called a stochastic basis and will be said to fulfill the usual conditions if the probability space is complete and if \mathcal{F}_0 contains all the Pnegligible sets.

Concerning spaces of stochastic processes, in a fixed stochastic basis $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t \in \mathbb{T}}, P)$, we will use the following notations and vocabulary. M will be the space of cadlag martingales.

For any $p \in [1,\infty]$ H^p will denote the subset of M of elements M such that $\sup_{t\in\mathcal{T}}|M_t| \in L^p$ and in this set we identify indistinguishable elements. It is a

Banach space for the norm $||M||_{\mathcal{H}^p} = \mathbb{E}[\sup_{t \in \mathcal{T}} M_t|^p]^{\frac{1}{p}}$, and \mathcal{H}_0^p will denote the Banach subspace of \mathcal{H}^p containing the elements starting at zero.

If $T = [0, T]$ for some $T \in \mathbb{R}^*_+$, a stopping time will be defined as a random variable with values in $[0, T] \cup \{+\infty\}$ such that for any $t \in [0, T]$, $\{\tau \leq t\} \in \mathcal{F}_t$. We define a localizing sequence of stopping times as an increasing sequence of stopping times $(\tau_n)_{n>0}$ such that there exists $N \in N$ for which $\tau_N = +\infty$. Let Y be a process and τ a stopping time, we denote Y^{τ} the process $t \mapsto Y_{t \wedge \tau}$ which we call stopped process. If C is a set of processes, we define its localized class \mathcal{C}_{loc} as the set of processes Y such that there exist a localizing sequence $(\tau_n)_{n\geq 0}$ such that for every n, the stopped process Y^{τ_n} belongs to C.

For any $M, N \in \mathcal{M}$, we denote $[M]$ (resp. $[M, N]$) the **quadratic variation** of M (resp. quadratic covariation of M, N). If $M, N \in \mathcal{H}_{loc}^2, \langle M, N \rangle$ (or simply $\langle M \rangle$ if $M = N$) will denote their (predictable) **angular bracket**. \mathcal{H}_0^2 will be equipped with scalar product defined by $(M, N)_{\mathcal{H}^2} := \mathbb{E}[M_T N_T] = \mathbb{E}[\langle M, N \rangle_T]$ which makes it a Hilbert space. Two elements M, N of $\mathcal{H}_{0,loc}^2$ will be said to be strongly orthogonal if $\langle M, N \rangle = 0$.

If A is an adapted process with bounded variation then $Var(A)$ (resp. $Pos(A)$, $Neg(A)$) will denote its total variation (resp. positive variation, negative variation), see Proposition 3.1, chap. 1 in [15]. In particular for almost all $\omega \in \Omega$, $t \mapsto Var_t(A(\omega))$ is the total variation function of the function $t \mapsto A_t(\omega)$.

3 Markov classes

We recall here some basic definitions and results concerning Markov processes. For a complete study of homogeneous Markov processes, one may consult [7], concerning non-homogeneous Markov classes, our reference was Chapter VI of [10].

3.1 Definition and basic results

The first definition refers to the canonical space that one can find in [14], see paragraph 12.63.

Notation 3.1. In the whole section E will be a fixed Polish space (a separable completely metrizable topological space), and $\mathcal{B}(E)$ its Borel σ -field. E will be called the state space.

We consider $T \in \mathbb{R}_+^*$. We denote $\Omega := \mathcal{D}(E)$ the Skorokhod space of functions from $[0, T]$ to E right-continuous with left limits and continuous at time T (e.g. cadlag). For any $t \in [0, T]$ we denote the coordinate mapping $X_t : \omega \mapsto \omega(t)$, and we introduce on Ω the σ -field $\mathcal{F} := \sigma(X_r|r \in [0,T])$.

Remark 3.2. All the results of the present paper remain valid if Ω is the space of continuous functions from $[0, T]$ to E, and to a time index equal to R_+ .

On the measurable space (Ω, \mathcal{F}) , we introduce the **canonical process**

$$
X: \begin{array}{ccc} (t,\omega)&\longmapsto&\omega(t)\\ ([0,T]\times \Omega,\mathcal{B}([0,T])\otimes\mathcal{F})&\longrightarrow&(E,\mathcal{B}(E)),\end{array}
$$

and the right-continuous filtration $(\mathcal{F}_t)_{t \in [0,T]}$ where $\mathcal{F}_t := \bigcap_{s \in [t,T]}$ $\sigma(X_r|r \leq s)$ if $t < T$, and $\mathcal{F}_T := \sigma(X_r | r \in [0, T]) = \mathcal{F}$.

 $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t \in [0,T]})$ will be called the **canonical space** (associated to T and E).

For any $t \in [0,T]$ we denote $\mathcal{F}_{t,T} := \sigma(X_r | r \geq t)$, and for any $0 \leq t \leq u < T$ we will denote $\mathcal{F}_{t,u} := \bigcap_{n \geq 0} \sigma(X_r | r \in [t, u + \frac{1}{n}]).$

We recall that since E is Polish, then $D(E)$ can be equipped with a Skorokhod distance which makes it a Polish metric space (see Theorem 5.6 in Chapter 3 of [11]), and for which the Borel σ -field is $\mathcal F$ (see Proposition 7.1 in Chapter 3 of [11]). This in particular implies that $\mathcal F$ is separable, as the Borel σ -field of a separable metric space.

Remark 3.3. The above σ -fields fulfill the properties below.

- 1. For any $0 \leq t \leq u \leq T$, $\mathcal{F}_{t,u} = \mathcal{F}_u \cap \mathcal{F}_{t,T}$;
- 2. for any $t > 0$, $\mathcal{F}_t \vee \mathcal{F}_{t,T} = \mathcal{F}$;
- 3. for any $(s, x) \in [0, T] \times E$, the two first items remain true when considering the $P^{s,x}$ -closures of all the σ -fields;
- 4. for any $t \geq 0$, $\Pi := \{F = F_t \cap F_T^t | (F_t, F_T^t) \in \mathcal{F}_t \times \mathcal{F}_{t,T}\}$ is a π -system generating F , i.e. it is stable with respect to the intersection.

Definition 3.4. The function

$$
P: \begin{array}{ccc} (s, t, x, A) & \longmapsto & P_{s,t}(x, A) \\ [0, T]^2 \times E \times \mathcal{B}(E) & \longrightarrow & [0, 1], \end{array}
$$

will be called **transition kernel** if, for any s, t in [0, T], $x \in E$, $A \in \mathcal{B}(E)$, it verifies the following.

- 1. $P_{s,t}(\cdot, A)$ is Borel,
- 2. $P_{s,t}(x, \cdot)$ is a probability measure on $(E, \mathcal{B}(E)),$
- 3. if $t \leq s$ then $P_{s,t}(x, A) = \mathbb{1}_A(x)$,
- 4. if $s < t$, for any $u > t$, $\int_E P_{s,t}(x, dy) P_{t,u}(y, A) = P_{s,u}(x, A)$.

The latter statement is the well-known Chapman-Kolmogorov equation.

Definition 3.5. A transition kernel P for which the first item is reinforced supposing that $(s, x) \mapsto P_{s,t}(x, A)$ is Borel for any t, A, will be said to be measurable in time.

Remark 3.6. Let P be a transition kernel which is measurable in time. By approximation by simple functions, one can easily show that, for any Borel function ϕ from E to R then $(s, x) \mapsto \int \phi(y)P_{s,t}(x, dy)$ is Borel, provided previous integral makes sense. In this paper we will only consider transition kernels which are measurable in time.

Definition 3.7. A canonical Markov class associated to a transition kernel P is a set of probability measures $(P^{s,x})_{(s,x)\in[0,T]\times E}$ defined on the measurable space (Ω, \mathcal{F}) and verifying for any $t \in [0, T]$ and $A \in \mathcal{B}(E)$

$$
P^{s,x}(X_t \in A) = P_{s,t}(x, A),
$$
\n(3.1)

and for any $s \le t \le u$

$$
\mathbf{P}^{s,x}(X_u \in A | \mathcal{F}_t) = P_{t,u}(X_t, A) \quad \mathbf{P}^{s,x} \quad a.s. \tag{3.2}
$$

Remark 3.8. Formula 1.7 in Chapter 6 of [10] states that for $(s, x) \in [0, T] \times E$, $t \geq s$ and $F \in \mathcal{F}_{t,T}$ yields

$$
P^{s,x}(F|\mathcal{F}_t) = P^{t,X_t}(F) = P^{s,x}(F|X_t) \quad P^{s,x} a.s.
$$
\n(3.3)

Property (3.3) will be called **Markov property**.

For the rest of this section, we are given a canonical Markov class $(P^{s,x})_{(s,x)\in[0,T]\times E}$ which transition kernel is measurable in time.

Proposition A.10 in [3] states the following.

Proposition 3.9. For any event $F \in \mathcal{F}$, $(s, x) \longmapsto P^{s,x}(F)$ is Borel. For any random variable Z, if the function $(s, x) \rightarrow E^{s,x}[Z]$ is well-defined (with possible values in $[-\infty, \infty]$, then it is Borel.

Definition 3.10. For any $(s, x) \in [0, T] \times E$ we will consider the (s, x) -**completion** $(\Omega, \mathcal{F}^{s,x}, (\mathcal{F}^{s,x}_t)_{t\in[0,T]}, \mathbf{P}^{s,x})$ of the stochastic basis $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t\in[0,T]}, \mathbf{P}^{s,x})$ by defining $\mathcal{F}^{s,x}$ as the $P^{s,x}$ -completion of $\mathcal F$, by extending $P^{s,x}$ to $\mathcal{F}^{s,x}$ and finally by defining $\mathcal{F}_t^{s,x}$ as the $P^{s,x}$ -closure of \mathcal{F}_t , for every $t \in [0,T]$.

We remark that, for any $(s, x) \in [0, T] \times E$, $(\Omega, \mathcal{F}^{s,x}, (\mathcal{F}^{s,x}_t)_{t \in [0,T]}, P^{s,x})$ is a stochastic basis fulfilling the usual conditions, see 1.4 in [15] Chapter I.

We recall the following simple consequence of Remark 32 in [5] Chapter II.

Proposition 3.11. Let G be a sub-σ-field of F, P a probability on (Ω, \mathcal{F}) and \mathcal{G}^{Γ} the P-closure of \mathcal{G} . Let Z^{Γ} be a real \mathcal{G}^{Γ} -measurable random variable. There exists a G-measurable random variable Z such that $Z = Z^{\Gamma}$ P-a.s.

From this we can deduce the following.

Proposition 3.12. Let $(s, x) \in [0, T] \times E$ be fixed, Z be a random variable and $t \in [s, T]$. Then $\mathbb{E}^{s,x}[Z|\mathcal{F}_t] = \mathbb{E}^{s,x}[Z|\mathcal{F}_t^{s,x}]$ $\mathbb{P}^{s,x}$ a.s.

Proof. $E^{s,x}[Z|\mathcal{F}_t]$ is \mathcal{F}_t -measurable and therefore $\mathcal{F}_t^{s,x}$ -measurable. Moreover, let $G^{s,x} \in \mathcal{F}_t^{s,x}$, by Remark 32 in [5] Chapter II, there exists $G \in \mathcal{F}_t$ such that $P^{s,x}(G \cup G^{s,x}) = P^{s,x}(G \backslash G^{s,x})$ implying $1\!\!1_G = 1\!\!1_{G^{s,x}}$ $P^{s,x}$ a.s. So

$$
E^{s,x}[1_{G^{s,x}}E^{s,x}[Z|\mathcal{F}_t]] = E^{s,x}[1_{G}E^{s,x}[Z|\mathcal{F}_t]]
$$

=
$$
E^{s,x}[1_{G}Z]
$$

=
$$
E^{s,x}[1_{G^{s,x}}Z],
$$

where the second equality occurs because of the definition of $E^{s,x}[Z|\mathcal{F}_t]$. \Box

In particular, under the probability $P^{s,x}$, $(\mathcal{F}_t)_{t\in[0,T]}$ -martingales and $(\mathcal{F}_t^{s,x})_{t\in[0,T]}$ martingales coincide.

We now show that in our setup, a canonical Markov class verifies the Blumenthal 0-1 law in the following sense.

Proposition 3.13. Let $(s, x) \in [0, T] \times E$ and $F \in \mathcal{F}_{s,s}$. Then $P^{s,x}(F)$ is equal to 1 or to 0; In other words, $\mathcal{F}_{s,s}$ is $\mathbf{P}^{s,x}$ -trivial.

Proof. Let $F \in \mathcal{F}_{s,s}$ as introduced in Notation 3.1. Since by Remark 3.3, $\mathcal{F}_{s,s} = \mathcal{F}_s \cap \mathcal{F}_{s,T}$, then F belongs to \mathcal{F}_s so by conditioning we get

$$
\begin{array}{rcl}\n\mathbf{E}^{s,x}[\mathbbm{1}_F] & = & \mathbf{E}^{s,x}[\mathbbm{1}_F \mathbbm{1}_F] \\
& = & \mathbf{E}^{s,x}[\mathbbm{1}_F \mathbf{E}^{s,x}[\mathbbm{1}_F | \mathcal{F}_s]] \\
& = & \mathbf{E}^{s,x}[\mathbbm{1}_F \mathbf{E}^{s,X_s}[\mathbbm{1}_F]],\n\end{array}
$$

where the latter equality comes from (3.3) because $F \in \mathcal{F}_{s,T}$. But $X_s = x$, $P^{s,x}$ a.s., so

$$
\begin{array}{lll} \mathbf{E}^{s,x}[1\hskip-1.5pt1_F] & = & \mathbf{E}^{s,x}[1\hskip-1.5pt1_F\mathbf{E}^{s,x}[1\hskip-1.5pt1_F]] \\ & = & \mathbf{E}^{s,x}[1\hskip-1.5pt1_F]^2. \end{array}
$$

 \Box

3.2 Examples of canonical Markov classes

We will list here some well-known examples of canonical Markov classes and some more recent ones.

• Let $E := \mathbb{R}^d$ for some $d \in \mathbb{N}^*$. We are given $b \in \mathcal{B}_b(\mathbb{R}_+ \times \mathbb{R}^d, \mathbb{R}^d)$, $\alpha \in C_b(R_+ \times R^d, S^*_+(R^d))$ (where $S^*_+(R^d)$ is the space of symmetric strictly positive definite matrices of size d) and K a Lévy kernel (this means that for every $(t, x) \in \mathbb{R}_+ \times \mathbb{R}^d$, $K(t, x, \cdot)$ is a σ -finite measure on $\mathbb{R}^d \setminus \{0\}$, $\sup_{t,x} \int \frac{\|y\|^2}{1+\|y\|^2} K(t, x, dy) < \infty$ and for every Borel set $A \in \mathcal{B}(\mathbb{R}^d \setminus \{0\}),$ $(t, x) \longmapsto \int_A$ $\frac{\|y\|^2}{1+\|y\|^2}K(t, x, dy)$ is Borel) such that for any $A \in \mathcal{B}(\mathbb{R}^d \setminus \{0\}),$ $(t, x) \longmapsto \int_A \frac{y}{1 + \|y\|^2} K(t, x, dy)$ is bounded continuous.

Let a denote the operator defined on some $\phi \in C_b^{1,2}(\mathbb{R}_+ \times \mathbb{R}^d)$ by

$$
\partial_t \phi + \frac{1}{2} Tr(\alpha \nabla^2 \phi) + (b, \nabla \phi) + \int \left(\phi(\cdot, \cdot + y) - \phi - \frac{(y, \nabla \phi)}{1 + ||y||^2} \right) K(\cdot, \cdot, dy)
$$
\n(3.4)

In [19] (see Theorem 4.3 and the penultimate sentence of its proof), the following is shown.

For every $(s, x) \in \mathbb{R}_+ \times \mathbb{R}^d$, there exists a unique probability $\mathbb{P}^{s,x}$ on the canonical space (see Definition 3.1) such that $\phi(\cdot, X) - \int_s^{\cdot} a(\phi)(r, X_r) dr$ is a local martingale for every $\phi \in C_b^{1,2}(\mathbb{R}_+ \times \mathbb{R}^d)$ and $P^{s,x}(X_s = x) = 1$.

Moreover $(P^{s,x})_{(s,x)\in\mathbb{R}_+\times\mathbb{R}^d}$ defines a canonical Markov class and its transition kernel is measurable in time.

- The case $K = 0$ was studied extensively in the celebrated book [20] in which it is also shown that if b, α are bounded and continuous in the second variable, then there exists a canonical Markov class with transition kernel measurable in time $(P^{s,x})_{(s,x)\in\mathbf{R}_+\times\mathbf{R}^d}$ such that $\phi(\cdot,X_\cdot) - \int_s^t a(\phi)(r,X_r) dr$ is a local martingale for any $\phi \in C_b^{1,2}(\mathbb{R}_+ \times \mathbb{R}^d)$.
- In [18], a canonical Markov class whose transition kernel is the weak fundamental solution of a parabolic PDE in divergence form is exhibited.
- In [13], diffusions on manifolds are studied and shown to define canonical Markov classes.
- Solutions of PDEs with distributional drift are exhibited in [12] and shown to define canonical Markov classes.

Some of previous examples were only studied as homogeneous Markov processes but can easily be shown to fall in the non-homogeneous setup of the present paper as it was illustrated in [3].

4 Martingale Additive Functionals

We now introduce the notion of non-homogeneous Additive Functional that we use in the paper. This looks to be a good compromise between the notion of Additive Functional associated to a stochastic system introduced by E.B. Dynkin (see for example [9]) and the more popular notion of homogeneous Additive Functional studied extensively, for instance by C. Dellacherie and P.A. Meyer in [7] Chapter XV. This section consists in extending some essential results stated in [7] Chapter XV to our setup.

Definition 4.1. We denote $\Delta := \{(t, u) \in [0, T]^2 | t \leq u\}$. On (Ω, \mathcal{F}) , we define a non-homogeneous Additive Functional (shortened AF) as a random-field $A := (A_u^t)_{(t,u) \in \Delta}$ indexed by Δ with values in R, verifying the two following conditions.

- 1. For any $(t, u) \in \Delta$, A_u^t is $\mathcal{F}_{t, u}$ -measurable;
- 2. for any $(s, x) \in [0, T] \times E$, there exists a real cadlag $\mathcal{F}^{s,x}$ -adapted process $A^{s,x}$ (taken equal to zero on $[0,s]$ by convention) such that for any $x \in E$ and $s \le t \le u$, $A_u^t = A_u^{s,x} - A_t^{s,x}$ **P**_s,x a.s.

 $A^{s,x}$ will be called the **cadlag version of** A under $P^{s,x}$.

An AF will be called a non-homogeneous square integrable Martingale **Additive Functional** (shortened square integrable MAF) if under any $P^{s,x}$ its cadlag version is a square integrable martingale. More generally an AF will be said to verify a certain property (being non-negative, increasing, of bounded variation, square integrable, having L^1 terminal value) if under any $P^{s,x}$ its cadlag version verifies it.

Finally, given an increasing $AF A$ and an increasing function V , A will be said to be absolutely continuous with respect to V if for any $(s, x) \in [0, T] \times E$, $dA^{s,x} \ll dV$ in the sense of stochastic measures.

In this section for a given MAF $(M_u^t)_{(t,u)\in\Delta}$ we will be able to exhibit two AF, denoted respectively by $([M]_u^t)_{(t,u)\in\Delta}$ and $(\langle M\rangle_u^t)_{(t,u)\in\Delta}$, which will play respectively the role of a quadratic variation and an angular bracket of it. Moreover we will show that the Radon-Nikodym derivative of the mentioned angular bracket of a MAF with respect to our reference function V is a time-dependent function of the underlying process.

Proposition 4.2. Let $(M_u^t)_{(t,u)\in\Delta}$ be a square integrable MAF, and for any $(s, x) \in [0, T] \times E$, $[M^{s,x}]$ be the quadratic variation of its cadlag version $M^{s,x}$ under $P^{s,x}$. Then there exists an AF which we will call $([M]_u^{\dagger})_{(t,u)\in\Delta}$ and which, for any $(s, x) \in [0, T] \times E$, has $[M^{s,x}]$ as cadlag version under $P^{s,x}$.

Proof. We adapt Theorem 16 Chapter XV in [7] to a non homogeneous set-up but the reader must keep in mind that our definition of Additive Functional is different from the one related to the homogeneous case.

For the whole proof $t < u$ will be fixed. We consider a sequence of subdivisions of $[t, u]$: $t = t_1^k < t_2^k < \cdots < t_k^k = u$ such that $\min_{i \le k} (t_{i+1}^k - t_i^k) \longrightarrow_{k \to \infty} 0$. Let $(s, x) \in [0, t] \times E$ with corresponding probability $P^{s,x}$. For any k, we have \sum $i< k$ $\left(M^{t^k_i}_{t^k_{i+1}}\right)^2 = \sum_{i \leq k}$ $_{i < k}$ $(M^{s,x}_{t^{k}_{i+1}}-M^{s,x}_{t^{k}_{i}})^2$ P^{s,x} a.s., so by definition of quadratic variation we know that

$$
\sum_{i < k} \left(M_{t_{i+1}^k}^{t_i^k} \right)^2 \xrightarrow[k \to \infty]{\mathbf{P}^{s,x}} [M^{s,x}]_u - [M^{s,x}]_t. \tag{4.1}
$$

In the sequel we will construct an $\mathcal{F}_{t,u}$ -measurable random variable $[M]_u^t$ such that for any $(s, x) \in [0, t] \times E$, $\sum_{i \leq k} \left(M_{t_{i+1}^k}^{t_i^k} \right)^2 \xrightarrow[k \to \infty]{P^{s,x}} [M]_u^t$. In that case $[M]_u^t$ will then be $P^{s,x}$ a.s. equal to $[M^{s,x}]_u - [M^{s,x}]_t$.

Let $x \in E$. Since M is a MAF, for any k, \sum $i< k$ $\left(M_{t_{i+1}^k}^{t_i^k}\right)^2$ is $\mathcal{F}_{t,u}$ -measurable and therefore $\mathcal{F}^{t,x}_{t,u}$ -measurable. Since $\mathcal{F}^{t,x}_{t,u}$ is complete, the limit in probability of this sequence, $[M^{t,x}]_u - [M^{t,x}]_t$, is still $\mathcal{F}_{t,u}^{t,x}$ -measurable. By Proposition 3.11, there is an $\mathcal{F}_{t,u}$ -measurable variable which depends on (t, x) , that we call $a_t(x, \omega)$ such that

$$
a_t(x,\omega) = [M^{t,x}]_u - [M^{t,x}]_t, \mathbb{P}^{t,x} \text{ a.s.}
$$
 (4.2)

We will show below that there is a jointly measurable version of $(x, \omega) \mapsto$ $a_t(x,\omega)$.

For every integer $n \geq 0$, we set $a_t^n(x, \omega) := n \wedge a_t(x, \omega)$ which is in particular limit in probability of $n \wedge \sum$ $i \leq k$ $\left(M_{t_{i+1}^k}^{t_i^k}\right)^2$ under $\mathbf{P}^{t,x}$.

For any integers k, n and any $x \in E$, we define the finite positive measures $\mathbf{Q}^{k,n,x}$, $\mathbf{Q}^{n,x}$ and \mathbf{Q}^x on $(\Omega, \mathcal{F}_{t,u})$ by

1.
$$
Q^{k,n,x}(F) := E^{t,x} \left[\mathbb{1}_F \left(n \wedge \sum_{i < k} \left(M_{t_{i+1}^k}^{t_i^k} \right)^2 \right) \right];
$$

2.
$$
Q^{n,x}(F) := E^{t,x}[\mathbb{1}_F(a_t^n(x,\omega))];
$$

$$
3. \ Qx(F) := \mathcal{E}^{t,x}[\mathbb{1}_F(a_t(x,\omega))].
$$

When k and n are fixed, for any fixed F , by Proposition 3.9,

 $x \longmapsto E^{t,x} \left[F \left(n \wedge \sum \right) \right]$ $i< k$ $\left(M_{t_{k+1}^{k}}^{t_{i}^{k}}\right)^{2}\right)$, is Borel. Then $n \wedge \sum$ $i< k$ $\left(M_{t_{i+1}^{k}}^{t_{i}^{k}}\right)^{2}\underset{k\rightarrow\infty}{\overset{\mathbf{p}_{t,x}}{\longrightarrow}}a_{t}^{n}(x,\omega),$ and this sequence is uniformly bounded by the constant n, so the convergence takes place in L^1 , therefore $x \mapsto Q^{n,x}(F)$ is also Borel as the pointwise limit in k of the functions $x \longmapsto Q^{k,n,x}(F)$. Similarly, $a_t^n(x,\omega) \xrightarrow[n \to \infty]{a.s.} a_t(x,\omega)$ and is non-decreasing, so by monotone convergence theorem, being a pointwise limit in n of the functions $x \mapsto Q^{n,x}(F)$, the function $x \mapsto Q^x(F)$ is Borel. We recall that F is separable. The just two mentioned properties and the fact that, for any x , we also have (by item 3. above) $Q^x \ll P^{t,x}$, allows to show (see Theorem 58 Chapter V in [6]) the existence of a jointly measurable (for $\mathcal{B}(E) \otimes \mathcal{F}_{t,u}$) version of $(x,\omega) \mapsto a_t(x,\omega)$, that we recall to be densities of Q^x with respect to $P^{t,x}$. That version will still

We can now set $[M]_u^t(\omega) = a_t(X_t(\omega), \omega)$, which is a correctly defined $\mathcal{F}_{t,u}$ measurable random variable. For any x, since $P^{t,x}(X_t = x) = 1$, we have the equalities

$$
[M]_u^t = a_t(x, \cdot) = [M^{t,x}]_u - [M^{t,x}]_t \ \mathbf{P}^{t,x} \text{a.s.}
$$
 (4.3)

We will moreover prove that

be denoted by the same symbol.

$$
[M]_u^t = [M^{s,x}]_u - [M^{s,x}]_t \quad \mathbf{P}^{s,x} \text{ a.s.},\tag{4.4}
$$

holds for every $(s, x) \in [0, t] \times E$, and not just in the case $s = t$ that we have just established in (4.3).

Let us fix $s < t$ and $x \in E$. We show that under any $P^{s,x}$, $[M]_u^t$ is the limit in probability of \sum $i< k$ $\left(M_{t_{i+1}^{k}}^{t_{i}^{k}}\right)^{2}$. Indeed, let $\epsilon > 0$: the event $\left\{\left|\right. \right.$ \sum $i< k$ $\left(M_{t_{i+1}^{k}}^{t_{i}^{k}}\right) ^{2}-[M]_{u}^{t}$ $\begin{array}{c} \begin{array}{c} \begin{array}{c} \end{array} \\ \begin{array}{c} \end{array} \end{array} \end{array}$ $>\epsilon$ belongs to $\mathcal{F}_{t,T}$ so by conditioning and using the Markov property (3.3) we have

$$
\begin{split} &\mathbf{P}^{s,x}\left(\left|\sum_{i& \epsilon\right) \\ =& \mathbf{E}^{s,x}\left[\mathbf{P}^{s,x}\left(\left|\sum_{i \epsilon\right|\mathcal{F}_t\right)\right] \\ =& \mathbf{E}^{s,x}\left[\mathbf{P}^{t,X_t}\left(\left|\sum_{i \epsilon\right)\right]. \end{split}
$$

For any fixed y, by (4.1) and (4.3), $P^{t,y}$ $\left(\right)$ \sum $_{i < k}$ $\left(M_{t_{i+1}^{k}}^{t_{i}^{k}}\right) ^{2}-[M]_{u}^{t}$ $\begin{array}{c} \hline \end{array}$ $>\epsilon$ tends to zero when k goes to infinity, for every realization ω , it yields $P^{t, X_t} \left(\left| \sum_{i \in I} \right| \right)$ tends to zero when k goes to infinity. Since this sequence is dominated by the $i< k$ $\left(M_{t_{i+1}^{k}}^{t_{i}^{k}}\right) ^{2}-[M]_{u}^{t}$ $\begin{array}{c} \begin{array}{c} \begin{array}{c} \end{array} \\ \begin{array}{c} \end{array} \end{array} \end{array}$ constant 1, that convergence still holds under the expectation with respect to

 $> \epsilon$

the probability the probability $P^{s,x}$, thanks to the dominated convergence theorem.

So we have built an $\mathcal{F}_{t,u}$ -measurable variable $[M]_u^t$ such that under any $P^{s,x}$ with $s \leq t$, $[M^{s,x}]_u - [M^{s,x}]_t = [M]^t_u$ a.s. and this concludes the proof. \Box

We will now extend the result about quadratic variation to the angular bracket of MAFs. The next result can be seen as an extension of Theorem 15 Chapter XV in [7] to a non-homogeneous context.

Proposition 4.3. Let $(B_u^t)_{(t,u)\in\Delta}$ be an increasing AF with L^1 terminal value, for any $(s, x) \in [0, T] \times E$, let $B^{s,x}$ be its cadlag version under $P^{s,x}$ and let $A^{s,x}$ be the predictable dual projection of $B^{s,x}$ in $(\Omega, \mathcal{F}^{s,x}, (\mathcal{F}^{s,x}_t)_{t\in[0,T]}, \mathbf{P}^{s,x})$. Then there exists an increasing AF with L^1 terminal value $(A_u^t)_{(t,u)\in\Delta}$ such that under any $P^{s,x}$, the cadlag version of A is $A^{s,x}$.

Proof. The first half of the demonstration will consist in showing that

$$
\forall (s, x) \in [0, t] \times E, \ (A_u^{s, x} - A_t^{s, x}) \text{ is } \mathcal{F}_{t, u}^{s, x} - \text{measurable.} \tag{4.5}
$$

We start by recalling a property of the predictable dual projection which we will have to extend slightly.

Let us fix (s, x) and the corresponding stochastic basis $(\Omega, \mathcal{F}^{s,x}, (\mathcal{F}^{s,x}_t)_{t \in [0,T]}, \mathbf{P}^{s,x}).$ For any $F \in \mathcal{F}^{s,x}$, let $N^{s,x,\overline{F}}$ be the cadlag version of the martingale, $r \mapsto$ $E^{s,x}[\mathbb{1}_F | \mathcal{F}_r].$ Then for any $0 \le t \le u \le T$, the predictable projection of the

process $r \mapsto \mathbb{1}_F \mathbb{1}_{[t,u]}(r)$ is $r \mapsto N_r^{s,x,F} \mathbb{1}_{[t,u]}(r)$, see the proof of Theorem 43 Chapter VI in [6]. Therefore by definition of the dual predictable projection (see Definition 73 Chapter VI in [6]) we have

$$
\mathcal{E}^{s,x} \left[\mathbb{1}_F (A_u^{s,x} - A_t^{s,x}) \right] = \mathcal{E}^{s,x} \left[\int_t^u N_r^{s,x,F} d B_r^{s,x} \right],\tag{4.6}
$$

for any $F \in \mathcal{F}^{s,x}$.

We will now prove some technical lemmas which in a sense extend this property, and will permit us to operate with a good common version of the random variable $\int_t^u N_{r-}^{s,x,F} dB_r^{s,x}$ not depending on (s, x) .

For the rest of the proof, $0 \le t \le u \le T$ will be fixed.

Notation 4.4. Let $F \in \mathcal{F}_{t,T}$. We denote for any $r \in [t,T], \omega \in \Omega$, $N_r^F(\omega) :=$ $\mathrm{P}^{t, X_t(\omega)}(F).$

It is clear that N^F previously introduced is an $(\mathcal{F}_{t,r})_{r \in [t,T]}$ -adapted process which does not depend on (s, x) , which takes values in [0, 1] for all r, ω and by Remark 3.8, for any $(s, x) \in [0, t] \times E$, $N^{s,x,F}$ is on $[t, T]$ a $P^{s,x}$ -version of N^F .

Lemma 4.5. Let $F \in \mathcal{F}_{t,T}$. There exists an $\mathcal{F}_{t,u}$ -measurable random variable which we will denote $\int_{t_1}^{u} N_r^F dB_r$ such that for any $(s, x) \in [0, t] \times E$, $\int_t^u N_{r-}^F dB_r = \int_t^u N_{r-}^{s,x,\bar{F}} dB_r^{s,x} \; P^{s,x} \; a.s.$

Remark 4.6. By definition, the process N^F introduced in Notation 4.4 and the r.v. $\int_t^u N_{r-}^F dB_r$ will not depend on any (s, x) .

Proof. In some sense we wish to integrate $r \mapsto N_{r^-}^F$ against B^t for fixed ω . However first we do not know a priori if the paths $r \mapsto N_r^F$ and $r \mapsto B_r^t$ are measurable, second $r \mapsto N_r^F$ may not have a left limit and B^t may be not of bounded variation. So it is not clear if $\int_{t_0}^{u} N_r^F dQ_r^t$ makes sense for any ω . Moreover under a certain $P^{s,x}$, $N^{F,s,x}$ and $B^{s,x} - B^{s,x}_t$ are only versions of N^F and B^t and not indistinguishable to them. Even if we could compute the overmentioned integral, it would not be clear if $\int_t^u N_{r-}^F dB_r^t = \int_t^u N_{r-}^{s,\tilde{x},F} dB_r^{s,x}$ P_{s,x} a.s.

We start by some considerations about B, setting $W_{tu} := \{ \omega : \sup_{r \in [t,u] \cap \mathbf{Q}}$ $B_r^t < \infty$ which is $\mathcal{F}_{t,u}$ -measurable, and for $r \in [t, u]$

$$
\bar{B}^t_\pi(\omega) := \begin{cases} \sup_{t \le \psi < \gamma} B^t_v(\omega) \text{ if } \omega \in \\ \max_{\omega \in \Omega} \end{cases}
$$

$$
\bar{B}_r^t(\omega) := \begin{cases}\n\sup_{\substack{t \le v < r \\ v \in \mathbb{Q}}} B_v^t(\omega) \text{ if } \omega \in W_{tu} \\
0 \text{ otherwise.} \n\end{cases}
$$

 \bar{B}^t is an increasing, finite (for all ω) process. In general, it is neither a measurable nor an adapted process; however for any $r \in [t, u]$, \overline{B}_r^t is still $\mathcal{F}_{t, u}$ measurable. Since it is increasing, it has right and left limits at each point for every ω , so we can define the process \tilde{B}^t indexed on $[t, u]$ below:

$$
\tilde{B}_r^t := \lim_{\substack{v \downarrow r \\ v \in \mathbf{Q}}} \bar{B}_v^t, r \in [t, u],\tag{4.7}
$$

when $u \in]t, T[$ and $\tilde{B}^t_T := B^t_T$ if $u = T$. Therefore \tilde{B}^t is an increasing, cadlag process. It is constituted by $\mathcal{F}_{t,u}$ -measurable random variables, and by Theorem 15 Chapter IV of [5], \tilde{B}^t is a also a measurable process (indexed by $[t, u]$).

We can show that \tilde{B}^t is $P^{s,x}$ -indistinguishable from $B^{s,x}_t - B^{s,x}_t$ for any $(s, x) \in [0, t] \times E$. Indeed, let (s, x) be fixed. Since $B^{s,x} - B^{s,x}_t$ is a version of B^t and Q being countable, there exists a P^{s,x}-null set N such that for all $\omega \in \mathcal{N}^c$ and $r \in \mathbb{Q} \cap [t, u], B_r^{s,x}(\omega) - B_t^{s,x}(\omega) = B_r^t(\omega)$. Therefore for any $\omega \in \mathcal{N}^c$ and $r \in [t, u],$

$$
\tilde{B}^t_r(\omega) = \lim_{\substack{v \downarrow r \\ v \in \mathcal{Q}^t}} \sup_{\substack{w \in \omega \\ w \in \mathcal{Q}}} B^t_w(\omega) = \lim_{\substack{v \downarrow r \\ v \in \mathcal{Q}^t}} \sup_{\substack{w \in \omega \\ w \in \mathcal{Q}}} B^{s,x}(\omega)_w - B^{s,x}(\omega)_t = B^{s,x}(\omega)_r - B^{s,x}(\omega)_t,
$$

where the latter equality comes from the fact that $B^{s,x}(\omega)$ is cadlag and increasing. So we have constructed an increasing finite cadlag (for all ω) process and so the path $r \mapsto \tilde{B}^t(\omega)$ is a Lebesgue integrator on $[t, u]$ for each ω .

We fix now $F \in \mathcal{F}_{t,T}$ and we discuss some issues related to N^F . Since it is positive, we can start defining the process \overline{N} , for index values $r \in [t, T]$ by $\bar{N}_r^F := \liminf_{v \downarrow r_\Omega} N_v^F$, and setting $\bar{N}_T^F := N_T^F$. This process is (by similar v∈^Q

arguments as for \tilde{B}^t defined in (4.7)), $P^{s,x}$ -indistinguishable to $N^{s,x,F}$ for all $(s, x) \in [0, t] \times E$. For any $r \in [t, T]$, N_r^F (see Notation 4.4) is $\mathcal{F}_{t,r}$ -measurable, so \bar{N}_r^F will also be $\mathcal{F}_{t,r}$ -measurable for any $r \in [t,T]$ by right-continuity of $\mathcal{F}_{t,r}$. (see Definition 3.1). However, \bar{N}^F is not necessarily cadlag for every ω , and also not necessarily a measurable process.

We subsequently define

 $W'_{tu} := \{ \omega \in \Omega | \text{there exists a cadlag function } f \text{ such that } \overline{N}^F(\omega) = f \text{ on } [t, u] \cap \mathbb{Q} \}.$

By Theorem 18 b) in Chapter IV of [5], W'_{tu} is $\mathcal{F}_{t,u}$ -measurable so we can define on $[t, u]$ $\tilde{N}_r^F := \bar{N}_r^F 1_{W'_{tu}}$. \tilde{N}_r^F is no longer $(\mathcal{F}_t)_{t \in [0,T]}$ -adapted, however, it is now cadlag for all ω and therefore a measurable process by Theorem 15 Chapter IV of [5]. The r.v. \tilde{N}_r^F are still $\mathcal{F}_{t,u}$ -measurable, and \tilde{N}^F is still $P^{s,x}$ indistinguishable to $N^{s,x,F}$ on $[t, u]$ for any $(s, x) \in [0, t] \times E$.

Finally we can define $\int_t^u N_{r-}^F dB_r := \int_t^u \tilde{N}_{r-}^F d\tilde{B}_r^t$ which is $P^{s,x}$ a.s. equal to $\int_t^u N_{r-}^{s,x,F} dB_r^{s,x}$ for any $(s, x) \in [0, t] \times E$.

Moreover, since \tilde{N}^F and \tilde{B} are both measurable with respect to $\mathcal{B}([t,u])\otimes \mathcal{F}_{t,u}$, then $\int_t^u N^F_{r^-}dB_r$ is $\mathcal{F}_{t,u}$ -measurable.

 \Box

The lemma below is a conditional version of the property (4.6).

Lemma 4.7. For any $(s, x) \in [0, t] \times E$ and $F \in \mathcal{F}_{t, T}^{s, x}$ we have $P^{s, x}$ -a.s.

$$
\mathcal{E}^{s,x}\left[\mathbb{1}_F(A^{s,x}_u - A^{s,x}_t)\big|\mathcal{F}_t\right] = \mathcal{E}^{s,x}\left[\int_t^u N_{r-}^F dB_r \bigg|\mathcal{F}_t\right].
$$

Proof. Let s, x, F be fixed. By definition of conditional expectation, we need to show that for any $G \in \mathcal{F}_t$ we have

$$
\mathcal{E}^{s,x}\left[\mathbb{1}_G \mathbb{1}_F(A_u^{s,x} - A_t^{s,x})\right] = \mathcal{E}^{s,x}\left[\mathbb{1}_G \mathcal{E}^{s,x}\left[\int_t^u N_{r-}^F dB_r \bigg| \mathcal{F}_t\right]\right] \text{ a.s.}
$$

For $r \in [t, u]$ we have $\mathbf{E}^{s,x}[\mathbb{1}_{F \cap G} | \mathcal{F}_r] = \mathbb{1}_G \mathbf{E}^{s,x}[\mathbb{1}_F | \mathcal{F}_r]$ a.s. therefore the cadlag versions of those processes are indistinguishable on $[t, u]$ and the random variables $\int_t^u N_r^{G \cap F} dB_r$ and $1_G \int_t^u N_{r-}^F dB_r$ as defined in Lemma 4.5 are a.s. equal. So by the non conditional property of dual predictable projection (4.6) we have

$$
\begin{array}{rcl}\n\mathbf{E}^{s,x}\left[\mathbbm{1}_G \mathbbm{1}_F (A^{s,x}_u - A^{s,x}_t)\right] & = & \mathbf{E}^{s,x} \left[\int_t^u N^{G \cap F}_{t} dB_r\right] \\
& = & \mathbf{E}^{s,x} \left[\mathbbm{1}_G \int_t^u N^F_{r-} dB_r\right] \\
& = & \mathbf{E}^{s,x} \left[\mathbbm{1}_G \mathbf{E}^{s,x} \left[\int_t^u N^F_{r-} dB_r | \mathcal{F}_t\right]\right],\n\end{array}
$$

which concludes the proof.

Lemma 4.8. For any $(s, x) \in [0, t] \times E$ and $F \in \mathcal{F}_{t,T}$ we have $P^{s,x}$ -a.s.,

$$
\mathcal{E}^{s,x} \left[\mathbb{1}_F (A_u^{s,x} - A_t^{s,x}) | \mathcal{F}_t \right] = \mathcal{E}^{s,x} \left[\mathbb{1}_F (A_u^{s,x} - A_t^{s,x}) | X_t \right].
$$

Proof. By Lemma 4.7 we have

$$
\mathcal{E}^{s,x}\left[\mathbb{1}_F(A^{s,x}_u - A^{s,x}_t)\big|\mathcal{F}_t\right] = \mathcal{E}^{s,x}\left[\int_t^u N_{r-}^F dB_r \bigg|\mathcal{F}_t\right].
$$

By Lemma 4.5, $\int_t^u N_{r-}^F dB_r$ is $\mathcal{F}_{t,T}$ measurable so the Markov property (3.3) implies

$$
\mathbf{E}^{s,x}\left[\int_t^u N^F_{r^-} dB_r \bigg| \mathcal{F}_t\right] = \mathbf{E}^{s,x}\left[\int_t^u N^F_{r^-} dB_r \bigg| X_t\right],
$$

therefore $E^{s,x} \left[\mathbb{1}_F(A^{s,x}_u - A^{s,x}_t)|\mathcal{F}_t\right]$ is a.s. equal to a $\sigma(X_t)$ -measurable r.v and so is a.s. equal to $E^{s,x} [\mathbb{1}_F(A_u^{s,x} - A_t^{s,x})|X_t]$.

We are now able to prove (4.5) which is the first important issue of the proof of Proposition 4.3, which states that By definition, a predictable dual projection is adapted so we already know that $(A_u^{s,x} - A_t^{s,x})$ is $\mathcal{F}_u^{s,x}$ -measurable, therefore by Remark 3.3, it is enough to show that it is also $\mathcal{F}_{t,T}^{s,x}$ -measurable. So we are going to show that

$$
A_{u}^{s,x} - A_{t}^{s,x} = \mathbf{E}^{s,x} \left[A_{u}^{s,x} - A_{t}^{s,x} | \mathcal{F}_{t,T} \right] \mathbf{P}^{s,x} \text{ a.s.}
$$
\n(4.8)

口

For this we will show that

$$
\mathcal{E}^{s,x} \left[\mathbb{1}_F (A_u^{s,x} - A_t^{s,x}) \right] = \mathcal{E}^{s,x} \left[\mathbb{1}_F \mathcal{E}^{s,x} \left[A_u^{s,x} - A_t^{s,x} | \mathcal{F}_{t,T} \right] \right],\tag{4.9}
$$

for any $F \in \mathcal{F}$. We will prove (4.9) for $F \in \mathcal{F}$ event of the form $F = F_t \cap F_{t,T}$ with $F_t \in \mathcal{F}_t$ and $F_{t,T} \in \mathcal{F}_{t,T}$.

By item 4. of Remark 3.3, such events form a π -system Π which generates \mathcal{F} .

Consequently, by the monotone class theorem, (4.9) will remain true for any $F \in \mathcal{F}$ and even in $\mathcal{F}^{s,x}$ since $\mathbb{P}^{s,x}$ -null set will not impact the equality. This will imply (4.8) so that $A^{s,x}_u - A^{s,x}_t$ is $\mathcal{F}^{s,x}_{t,T}$ -measurable.

At this point, as we have anticipated, we prove (4.9) for a fixed $F = F_t \cap F_{t,T} \in \Pi$. By Lemma 4.8 we have

$$
\begin{array}{rcl}\n\mathbf{E}^{s,x}\left[\mathbbm{1}_F(A^{s,x}_u - A^{s,x}_t)\right] & = & \mathbf{E}^{s,x}\left[\mathbbm{1}_{F_t}\mathbf{E}^{s,x}\left[\mathbbm{1}_{F_{t,T}}(A^{s,x}_u - A^{s,x}_t)|\mathcal{F}_t\right]\right] \\
& = & \mathbf{E}^{s,x}\left[\mathbbm{1}_{F_t}\mathbf{E}^{s,x}\left[\mathbbm{1}_{F_{t,T}}(A^{s,x}_u - A^{s,x}_t)|X_t\right]\right] \\
& = & \mathbf{E}^{s,x}\left[\mathbbm{1}_{F_t}\mathbf{E}^{s,x}\left[\mathbf{E}^{s,x}\left[\mathbbm{1}_{F_{t,T}}(A^{s,x}_u - A^{s,x}_t)|\mathcal{F}_{t,T}\right]|X_t\right]\right],\n\end{array}
$$

where the latter equality holds since $\sigma(X_t) \subset \mathcal{F}_{t,T}$.

Now since $E^{s,x} \left[\mathbb{1}_{F_{t,T}}(A^{s,x}_u - A^{s,x}_t)\big|\mathcal{F}_{t,T}\right]$ is $\mathcal{F}_{t,T}$ -measurable, the Markov property (3.3) allows us to substitute the conditional σ -field $\sigma(X_t)$ with \mathcal{F}_t and obtain

$$
\begin{array}{rcl} \mathbf{E}^{s,x} \left[\mathbbm{1}_F (A^{s,x}_u - A^{s,x}_t) \right] & = & \mathbf{E}^{s,x} \left[\mathbbm{1}_{F_t} \mathbf{E}^{s,x} \left[\mathbbm{1}_{F_{t,T}} (A^{s,x}_u - A^{s,x}_t) | \mathcal{F}_{t,T} \right] | \mathcal{F}_t \right] \right] \\ & = & \mathbf{E}^{s,x} \left[\mathbbm{1}_{F_t} \mathbf{E}^{s,x} \left[\mathbbm{1}_{F_{t,T}} (A^{s,x}_u - A^{s,x}_t) | \mathcal{F}_{t,T} \right] \right] \\ & = & \mathbf{E}^{s,x} \left[\mathbbm{1}_{F_t} \mathbbm{1}_{F_{t,T}} \mathbf{E}^{s,x} \left[(A^{s,x}_u - A^{s,x}_t) | \mathcal{F}_{t,T} \right] \right] \\ & = & \mathbf{E}^{s,x} \left[\mathbbm{1}_{F} \mathbf{E}^{s,x} \left[(A^{s,x}_u - A^{s,x}_t) | \mathcal{F}_{t,T} \right] \right]. \end{array}
$$

This concludes the proof of (4.9), therefore (4.8) holds so that $A_u^{s,x} - A_t^{s,x}$ is $\mathcal{F}_{t,u}^{s,x}$ -measurable and so (4.5) is established. This concludes the first part of the proof of Proposition 4.3.

We pass to the second part of the proof of Proposition 4.3 where we will show that for given $0 < t < u$ there is an $\mathcal{F}_{t,u}$ -measurable r.v. A_u^t such that for every $(s, x) \in [0, t] \times E$, $(A_u^{s,x} - A_t^{s,x}) = A_u^t$ $P^{s,x}$ a.s.

Similarly to what we did with the quadratic variation in Proposition 4.2, we start by noticing that for any $x \in E$, since $(A_u^{t,x} - A_t^{t,x})$ is $\mathcal{F}_{t,u}^{t,x}$ -measurable, there exists by Proposition 3.11 an $\mathcal{F}_{t,u}$ -measurable r.v. $a(x,\omega)$ such that

$$
a(x,\omega) = A_u^{t,x} - A_t^{t,x} \quad \mathbf{P}^{t,x} \text{ a.s.}
$$
 (4.10)

As in the proof of Proposition 4.2, we will show the existence of a jointlymeasurable version of $(x, \omega) \mapsto a(x, \omega)$.

For every $x \in E$ we define on $\mathcal{F}_{t,u}$ the positive measure

$$
\mathbf{Q}^x : F \longmapsto \mathbf{E}^{t,x} \left[\mathbb{1}_F (A_t^{t,x} - A_t^{t,x}) \right] = \mathbf{E}^{t,x} \left[\mathbb{1}_F a(x,\omega) \right]. \tag{4.11}
$$

By Lemma 4.5, and (4.6), for every $F \in \mathcal{F}_{t,u}$ we have

$$
\mathbf{Q}^x(F) = \mathbf{E}^{t,x} \left[\int_t^u N_{r^-}^F dB_r \right],\tag{4.12}
$$

and we recall that $\int_t^u N_r^F dB_r$ does not depend on x. So by Proposition 3.9 $x \longmapsto Q^x(F)$ is Borel for any F. Moreover, for any $x, Q^x \ll P^{t,x}$. Again by Theorem 58 Chapter V in [6], there exists a version $(x, \omega) \mapsto a(x, \omega)$ measurable for $\mathcal{B}(E) \otimes \mathcal{F}_{t,u}$ of the related Radon-Nikodym densities.

We can now set $A_u^t(\omega) := a(X_t(\omega), \omega)$ which is then an $\mathcal{F}_{t,u}$ -measurable r.v. Since $P^{t,x}(X_t = x) = 1$ and (4.10) hold, we have

$$
A_u^t = a(X_t, \cdot) = a(x, \cdot) = A_u^{t, x} - A_t^{t, x} \quad P^{t, x} \text{ a.s.}
$$
 (4.13)

We now set $s < t$ and $x \in E$ and we want to show that we still have $A_u^t = A_u^{s,x} - A_t^{s,x}$ P_s,x a.s. So, as above, we consider $F \in \mathcal{F}_{t,u}$ and, thanks to (4.6) we compute

$$
E^{s,x} \left[\mathbb{1}_F(A_u^{s,x} - A_t^{s,x}) \right] = E^{s,x} \left[\int_t^u N_F^F dB_r \right]
$$

\n
$$
= E^{s,x} \left[\begin{matrix} \int_t^u N_F^F dB_r \end{matrix} \right]
$$

\n
$$
= E^{s,x} \left[\begin{matrix} \int_t^u N_F^F dB_r \end{matrix} \right]
$$

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$$
= E^{s,x} \left[\begin{matrix} \int_t^u N_F^F dB_r \end{matrix} \right]
$$

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$$
= E^{s,x} \left[\begin{matrix} \int_t^u N_F^F dB_r \end{matrix} \right]
$$

\n
$$
= E^{s,x} \left[\begin{matrix} \int_t^u N_F^F dB_r \end{matrix} \right]
$$

\n
$$
= E^{s,x} \left[\begin{matrix} \int_t^u K_u^L \end{matrix} \right]
$$

\n
$$
= E^{s,x} \left[\begin{matrix} \int_t^u K_u^L \end{matrix} \right].
$$

\n(4.14)

Indeed, concerning the fourth equality we recall that, by (4.11) , (4.12) and (4.13), we have $E^{t,x} \left[\int_t^u N_r^F dB_r \right] = E^{t,x} \left[1_F A_u^t \right]$ for all x, so this equality becomes an equality whatever random variable we plug into x . The third and fifth equalities come from the Markov property (3.3) since $\int_t^u N_r^F dB_r$ and A_u^t are $\mathcal{F}_{t,T}$ -measurable.

Then, adding $P^{s,x}$ -null sets does not change the validity of (4.14) , so we have for any $F \in \mathcal{F}_{t,u}^{s,x}$ that $E^{s,x} [\mathbb{1}_F(A_u^{s,x} - A_t^{s,x})] = E^{s,x} [\mathbb{1}_F A_u^t]$.

Finally, since we had shown in the first half of the proof that $A_u^{s,x} - A_t^{s,x}$ is $\mathcal{F}_{t,u}^{s,x}$ -measurable, and since A_u^t also has, by construction, the same measurability property, we can conclude that $A_u^{s,x} - A_t^{s,x} = A_u^t P^{s,x}$ a.s.

Since this holds for every $t \leq u$ and $(s, x) \in [0, t] \times E$, $(A_u^t)_{(t, u) \in \Delta}$ is the desired AF, which ends the proof of Proposition 4.3. \Box

Corollary 4.9. Let M, M' be two square integrable MAFs, let $M^{s,x}$ (respectively $M'^{s,x}$) be the cadlag version of M (respectively M') under $P^{s,x}$. Then there exists a bounded variation AF with L^1 terminal condition denoted $\langle M, M' \rangle$ such that under any $P^{s,x}$, the cadlag version of $\langle M, M' \rangle$ is $\langle M^{s,x}, M'^{s,x} \rangle$. If $M = M'$ the AF $\langle M, M' \rangle$ will be denoted $\langle M \rangle$ and is increasing.

Proof. If $M = M'$, the corollary comes from the combination of Propositions 4.2 and 4.3, and the fact that the angular bracket of a square integrable martingale is the dual predictable projection of its quadratic variation.

Otherwise, it is clear that $M + M'$ and $M - M'$ are square integrable MAFs, so we can consider the increasing MAFs $\langle M - M' \rangle$ and $\langle M + M' \rangle$. We introduce the AF

$$
\langle M, M' \rangle = \frac{1}{4} (\langle M + M' \rangle - \langle M - M' \rangle),
$$

which by polarization has cadlag version $\langle M^{s,x}, M'^{s,x} \rangle$ under $P^{s,x}$. $\langle M, M' \rangle$ is therefore a bounded variation AF with L^1 terminal condition. \Box

We are now going to study the Radon-Nikodym derivative of an increasing continuous AF with respect to some measure. The next result can be seen as an extension of Theorem 13 Chapter XV in [7] in a non-homogeneous setup.

Proposition 4.10. Let A be a positive, non-decreasing AF absolutely continu $ous with respect to some continuous non-decreasing function V , and for every$ $(s, x) \in [0, T] \times E$ let $A^{s,x}$ be the cadlag version of A under $P^{s,x}$. There exists a Borel function $h \in \mathcal{B}([0,T] \times E, R)$ such that for every $(s,x) \in [0,T] \times E$, $A^{s,x} = \int_{s}^{t\vee s} h(r, X_r) dV_r$, in the sense of indistinguishability.

Proof. We set

$$
C_u^t = A_u^t + (V_u - V_t) + (u - t), \tag{4.15}
$$

which is an AF with cadlag versions

$$
C_t^{s,x} = A_t^{s,x} + V_t + t,\t\t(4.16)
$$

and we start by showing the statement for A and C instead of A and V . We introduce the intermediary function C so that for any $u > t$ that $\frac{A_v^{s,x} - A_t^{s,x}}{C_u^{s,x} - C_t^{s,x}} \in [0,1]$; that property will be used extensively in connections with the application of dominated convergence theorem.

Since $A^{s,x}$ is non-decreasing for any $(s,x) \in [0,T] \times E$, A can be taken positive (in the sense that $A_u^t(\omega) \geq 0$ for any $(t, u) \in \Delta$ and $\omega \in \Omega$) by considering A^+ (defined by $(A^+)_u^t(\omega) := A_u^t(\omega)^+$) instead of A.

For $t \in [0, T]$ we set

$$
K_{t} := \liminf_{n \to \infty} \frac{A_{t+\frac{1}{n}}^{t}}{A_{t+\frac{1}{n}}^{t} + \frac{1}{n} + (V_{t+\frac{1}{n}} - V_{t})}
$$

\n
$$
= \lim_{n \to \infty} \inf_{p \ge n} \frac{A_{t+\frac{1}{p}}^{t}}{A_{t+\frac{1}{p}}^{t} + \frac{1}{p} + (V_{t+\frac{1}{p}} - V_{t})}
$$

\n
$$
= \lim_{n \to \infty} \lim_{n \to \infty} \min_{p \le n} \frac{A_{t+\frac{1}{p}}^{t}}{A_{t+\frac{1}{p}}^{t} + \frac{1}{p} + (V_{t+\frac{1}{p}} - V_{t})}.
$$
\n(4.17)

By positivity, this liminf always exists and belongs to $[0, 1]$ since the sequence belongs to [0, 1]. For every $(s, x) \in [0, T] \times E$, since for all $t \geq s$ and $n \geq 0$,

 $A_{t+\frac{1}{n}}^t = A_{t+\frac{1}{n}}^{s,x} - A_t^{s,x}$ P_s,x a.s., then $K^{s,x}$ defined by $K_t^{s,x} := \liminf_{n \to \infty}$ $A^{s,x}_{t+\frac{1}{n}}-A^{s,x}_{t}$ $C^{s,x}_{t+\frac{1}{n}}-C^{s,x}_{t}$ is a P^{s,x}-version of K, for $t \in [s, T]$.

By Lebesgue Differentiation theorem (see Theorem 12 Chapter XV in [7] for a version of the theorem with a general atomless measure), for any (s, x) , for $P^{s,x}$ almost all ω , since $dC^{s,x}(\omega)$ is absolutely continuous with respect to $dA^{s,x}(\omega)$, $K^{s,x}(\omega)$ is a density of $dA^{s,x}(\omega)$ with respect to $dC^{s,x}(\omega)$.

We now show that there exists a Borel function k in $\mathcal{B}([0,T]\times E, R)$ such that under any $P^{s,x}, k(t, X_t)$ is on [s, T[a version of K (and therefore of $K^{s,x}$). For every $t \in [0, T[, K_t \text{ is measurable with respect to } \bigcap_{n \geq 0} \mathcal{F}_{t,t+\frac{1}{n}} = \mathcal{F}_{t,t}$ by construction, taking into account Notation 3.1. So for any $(t, x) \in [0, T] \times E$, by Proposition 3.13, there exists a constant which we denote $k(t, x)$ such that

$$
K_t = k(t, x), \quad P^{t, x} \text{a.s.}
$$
 (4.18)

For any integers (n, m) , we define $k^{n,m}$ by

$$
(t,x) \mapsto \mathcal{E}^{t,x} \left[\min_{n \le p \le m} \frac{A_{t+\frac{1}{p}}^t}{A_{t+\frac{1}{p}}^t + \frac{1}{p} + (V_{t+\frac{1}{p}} - V_t)} \right],
$$

and k^n by

$$
(t, x) \mapsto \mathcal{E}^{t, x} \left[\inf_{p \ge n} \frac{A_{t + \frac{1}{p}}^t}{A_{t + \frac{1}{p}}^t + \frac{1}{p} + (V_{t + \frac{1}{p}} - V_t)} \right],
$$
\n(4.19)

We start showing that $\tilde{k}^{n,m}$ defined by

$$
(s, x, t) \longrightarrow \mathbf{E}^{s \wedge t, x} \left[\min_{n \le p \le m} \frac{A_{t + \frac{1}{p}}^t}{A_{t + \frac{1}{p}}^t + V_{t + \frac{1}{p}} - V_t} \right],
$$
\n
$$
[0, T] \times E \times [0, T] \longrightarrow [0, 1],
$$
\n
$$
(4.20)
$$

is jointly Borel.

If we fix t, then by Proposition 3.9 $(s, x) \longmapsto E^{s,x} \left[\min_{n \leq p \leq m} \right]$ $\frac{A_{t+\frac{1}{p}}^t}{A_{t+\frac{1}{p}}^t+\frac{1}{p}+(V_{t+\frac{1}{p}}-V_t)}$ 1 is a Borel function, so by composing with $(s, x) \mapsto (s \wedge t, x)$, then

 $(s, x) \mapsto \tilde{k}^{n,m}(s, x, t)$ is Borel. Moreover, if we fix $(s, x) \in [0, T] \times E$ we show below that $t \mapsto \tilde{k}^{n,m}(s, x, t)$ is continuous, which by Lemma 4.51 in [1] implies the joint measurability of $\tilde{k}^{n,m}$.

To show that mentioned continuity property, we first remark that $\tilde{k}^{n,m}(s, x, \cdot)$ is constant on [0, s]; moreover $A^{s,x}$ is continuous $\mathbb{P}^{s,x}$ a.s. V is continuous, and the minimum of a finite number of continuous functions remains continuous. Let $t_q \longrightarrow t$ be a converging sequence in [s, T[. Then $\min_{n \leq p \leq m}$ $A^{s,x}_{t_q+\frac{1}{p}}-A^{s,x}_{t_q}$ $A^{s,x}_{tq+\frac{1}{p}}-A^{s,x}_{tq}+\frac{1}{p}+(V_{tq+\frac{1}{p}}-V_{tq})$ tends a.s. to $\min_{n \leq p \leq m}$ $A^{s,x}_{t+\frac{1}{p}}-A^{s,x}_{t}$ $A_{t+\frac{1}{p}}^{s,x} - A_t^{s,x} + \frac{1}{p} + (V_{t+\frac{1}{p}} - V_t)$ when q tends to infinity. Since for

any
$$
s \le t \le u
$$
, $A_u^t = A_u^{s,x} - A_t^{s,x}$ P^{s,x} a.s., then
$$
\frac{A_{t_q+\frac{1}{p}}^{t_q}}{A_{t_q+\frac{1}{p}}^{t_q+\frac{1}{p}+\frac{1}{r}+(V_{t_q+\frac{1}{p}}-V_{t_q})}}
$$
tends a.s.

to $A_{t+\frac{1}{p}}^t$ $A_{t+\frac{1}{p}}^t$ + $\frac{1}{r}$ + $\frac{1}{r}(V_{t+\frac{1}{p}}-V_t)$. All those terms being smaller than one, by dominated convergence theorem, the mentioned convergence also holds under the expecta-

tion, hence the announced continuity related to $\tilde{k}^{n,m}$ is established.

Since $k^{n,m}(t, y) = \tilde{k}^{n,m}(t, t, y)$, by composition we can deduce that for any n, m , $k^{n,m}$ is Borel. By the dominated convergence theorem, $k^{n,m}$ tends pointwise to k^n (which was defined in (4.19), when m goes to infinity so k^n are also Borel for every *n*. Finally, keeping in mind (4.17) nd (4.18) we have $P^{t,x}$ a.s.

$$
k(t,x) = K_t = \lim_{n \to \infty} \inf_{p \ge n} \frac{A_{t+\frac{1}{p}}^t}{A_{t+\frac{1}{p}}^t + \frac{1}{p} + (V_{t+\frac{1}{p}} - V_t)}.
$$

Taking the expectation and again by the dominated convergence theorem, k^n (defined in (4.19)) tends pointwise to k when n goes to infinity so k is Borel.

We now show that, for any $(s, x) \in [0, T] \times E$, $k(\cdot, X)$ is a P^{s,x}-version of K on $[s, T]$.

Since $P^{t,x}(X_t = x) = 1$, we know that for any $t \in [0,T]$, $x \in E$, we have $K_t = k(t, x) = k(t, X_t) P^{t, x}$ -a.s., and we prove below that for any $t \in [0, T]$, $(s, x) \in [0, t] \times E$, we have $K_t = k(t, X_t)$ P^{s,x}-a.s.

Let $t \in [0, T]$ be fixed. Since A is an AF, for any n , $\frac{A_{t+\frac{1}{p}}^t}{A_{t+\frac{1}{p}}^t + \frac{1}{n} + (V_{t+\frac{1}{n}} - V_t)}$ is

 $\mathcal{F}_{t,t+\frac{1}{n}}$ -measurable.

So the event $\begin{cases} \liminf_{n \to \infty} \end{cases}$ $\frac{A_{t+\frac{1}{n}}^t}{A_{t+\frac{1}{n}}^t + \frac{1}{n} + (V_{t+\frac{1}{n}} - V_t)} = k(t, X_t)$ belongs to $\mathcal{F}_{t,T}$ and by Markov property (3.3), for any $(s, x) \in [0, t] \times E$, we get

$$
P^{s,x}(K_t = k(t, X_t)) = E^{s,x}[P^{s,x}(K_t = k(t, X_t)|\mathcal{F}_t)]
$$

=
$$
E^{s,x}[P^{t,X_t}(K_t = k(t, X_t))]
$$

= 1.

For any (s, x) , the process $k(\cdot, X)$ is therefore on $[s, T[$ a $P^{s,x}$ -modification of K and therefore of $K^{s,x}$. However it is not yet clear if provides another density of $dA^{s,x}$ with respect to $dC^{s,x}$, which was defined at (4.16).

Considering that $(t, u, \omega) \mapsto V_u - V_t$ also defines a positive non-decreasing AF absolutely continuous with respect to C , defined in (4.15) , we proceed similarly as at the beginning of the proof, replacing the AF A with V .

Let the process K' be defined by

$$
K'_{t} = \liminf_{n \to \infty} \frac{V_{t + \frac{1}{n}} - V_{t}}{A_{t + \frac{1}{n}}^{t} + \frac{1}{n} + (V_{t + \frac{1}{n}} - V_{t})},
$$

and for any (s, x) , let $K'^{s,x}$ be defined on $[s, T]$ by

$$
K_t^{s,x} = \liminf_{n \to \infty} \frac{V_{t+\frac{1}{n}} - V_t}{A_{t+\frac{1}{n}}^{s,x} - A_t^{s,x} + \frac{1}{n} + (V_{t+\frac{1}{n}} - V_t)}.
$$

Then, for any (s, x) , $K'^{s,x}$ on $[s, T]$ is a P^{s,x}-version of K', and it constitutes a density of $dV(\omega)$ with respect to $dC^{s,x}(\omega)$ on [s, T], for almost all ω . One shows then the existence of a Borel function k' such that for any (s, x) , $k'(\cdot, X)$ is a $P^{s,x}$ -version of K' and a modification of $K'^{s,x}$ on $[s,T[$. So for any (s, x) , under $P^{s,x}$, we can write

$$
\left\{ \begin{array}{ccc} A^{s,x} & = & \int_s^{\cdot \vee s} K^{s,x}_r d C^{s,x}_r \\ V_{\cdot \vee s} - V_s & = & \int_s^{\cdot \vee s} K'^{s,x}_r d C^{s,x}_r \end{array} \right.
$$

Now since $dA^{s,x} \ll dV$, for a fixed ω , the set $\{r \in [s,T]|K_r'^{s,x}(\omega) = 0\}$ is negligible with respect to dV so also for $dA^{s,x}(\omega)$ and therefore we can write

$$
\begin{array}{rcl} A^{s,x} & = & \int_s^{\cdot \vee s} K^{s,x}_{r} dC^{s,x}_{r} \\ & = & \int_s^{\cdot \vee s} \frac{K^{s,x}_{r}}{K^{r,x,x}_{r}} 1_{\{K^{r',x}_{r} \neq 0\}} K^{s,x}_{r} dC^{s,x}_{r} \\ & + & \int_s^{\cdot \vee s} 1_{\{K^{r',x}_{r} = 0\}} dA^{s,x}_{r} \\ & = & \int_s^{\cdot \vee s} \frac{K^{s,x}_{s,x}}{K^{r',x,x}_{r}} 1_{\{K^{r',s,x}_{r} \neq 0\}} dV_{r}, \end{array}
$$

where we use the convention that for any two functions ϕ, ψ then $\frac{\phi}{\psi} 1_{\psi \neq 0}$ is defined by by

$$
\frac{\phi}{\psi} 1\!\!1_{\{\psi \neq 0\}}(x) = \begin{cases} \frac{\phi(x)}{\psi(x)} & \text{if } \psi(x) \neq 0\\ 0 & \text{if } \psi(x) = 0. \end{cases}
$$

We set now $h := \frac{k}{k'} \mathbb{1}_{\{k'_r \neq 0\}}$ which is Borel, and clearly for any (s, x) , $h(t, X_t)$ is a P^{s,x}-version of $H^{s,x} := \frac{K^{s,x}}{K^{s,x}} 1_{\{K^{s,x}\neq 0\}}$ on [s, T[. So by Lemma 5.12 in [2], $H_t^{s,x} = h(t, X_t) dV \otimes dP^{s,x}$ a.e. and finally we have shown that under any $P^{s,x}$, $A^{s,x} = \int_s^{\sqrt{s}} h(r, X_r) dV_r$ on [0, T[. Without change of notations we extend h to $[0, T] \times \mathbb{E}$ by zero for $t = T$. Since $A^{s,x}$ is continuous $P^{s,x}$ -a.s. previous equality extends to T .

 \Box

Proposition 4.11. Let $(A_u^t)_{(t,u)\in\Delta}$ be an AF with bounded variation and taking L^1 values. Then there exists an increasing AF which we denote $(Pos(A)_u^t)_{(t,u)\in \Delta}$ (resp. $(Neg(A)_u^t)_{(t,u)\in \Delta}$) and which, for any $(s, x) \in [0, T] \times E$, has $Pos(A^{s,x})$ (resp. $Neg(A^{s,x})$) as cadlag version under $P^{s,x}$.

Proof. By definition of the total variation of a bounded variation function, the following holds. For every $(s, x) \in [0, T] \times E$, $s \le t \le u \le T$ for $P^{s,x}$ almost all $\omega \in \Omega$, and any sequence of subdivisions of $[t, u]$: $t = t_1^k < t_2^k < \cdots < t_k^k = u$ such that $\min_{i \leq k} (t_{i+1}^k - t_i^k) \underset{k \to \infty}{\longrightarrow} 0$ we have

$$
\sum_{i< k} |A^{s,x}_{t^{k}_{i+1}}(\omega) - A^{s,x}_{t^{k}_i}(\omega)| \underset{k \to \infty}{\longrightarrow} Var(A^{s,x})_u(\omega) - Var(A^{s,x})_t(\omega), \tag{4.21}
$$

taking into account the considerations of the end of Section 2. By Proposition 3.3 in [15] Chapter I, we have $Pos(A^{s,x}) = \frac{1}{2}(Var(A^{s,x}) + A^{s,x})$ and $Neg(A^{s,x}) =$ $\frac{1}{2}(Var(A^{s,x}) - A^{s,x})$. Moreover, for any $x \in \mathbb{R}$ we know that $x^+ = \frac{1}{2}(|x| + x)$ and $x^- = \frac{1}{2}(|x| - x)$, so we also have

$$
\begin{cases}\n\sum_{i < k} (A^{s,x}_{t^{k}_{i+1}}(\omega) - A^{s,x}_{t^{k}_{i}}(\omega))^{+} & \longrightarrow_{k \to \infty} \operatorname{Pos}(A^{s,x})_{u}(\omega) - \operatorname{Pos}(A^{s,x})_{t}(\omega) \\
\sum_{i < k} (A^{s,x}_{t^{k}_{i+1}}(\omega) - A^{s,x}_{t^{k}_{i}}(\omega))^{-} & \longrightarrow_{k \to \infty} \operatorname{Neg}(A^{s,x})_{u}(\omega) - \operatorname{Neg}(A^{s,x})_{t}(\omega), \\
(4.22)\n\end{cases}
$$

for $P^{s,x}$ almost all ω . Since the convergence a.s. implies the convergence in probability, for every $(s, x) \in [0, T] \times E$, $s \le t \le u$ and any sequence of subdivisions of $[t, u]$: $t = t_1^k < t_2^k < \cdots < t_k^k = u$ such that $\min_{i \le k} (t_{i+1}^k - t_i^k) \longrightarrow_{k \to \infty} 0$, we have

$$
\begin{cases}\n\sum_{i < k} \left(A_{t_{i+1}^k}^{t_i^k} \right)^+ \xrightarrow[k \to \infty]{\text{P.s.}} \text{Pos}(A^{s,x})_u - \text{Pos}(A^{s,x})_t \\
\sum_{i < k} \left(A_{t_{i+1}^k}^{t_i^k} \right)^- \xrightarrow[k \to \infty]{\text{P.s.}} \text{Neg}(A^{s,x})_u - \text{Neg}(A^{s,x})_t.\n\end{cases} \tag{4.23}
$$

The proof can now be performed according to the same arguments as in the proof of Proposition 4.2, replacing M with A , the quadratic increments with the positive (resp. negative) increments, and the quadratic variation with the positive (resp. negative) variation of an adapted process. \Box

We recall a definition and a result from [2]. We assume for now that we are given a fixed stochastic basis fulfilling the usual conditions, and a non-decreasing function V .

Notation 4.12. We denote $\mathcal{H}^{2,V} := \{ M \in \mathcal{H}^2_0 | d \langle M \rangle \ll dV \}$ and $\mathcal{H}^{2,\perp V} :=$ $\{M \in \mathcal{H}_0^2 | d\langle M \rangle \perp dV\}.$

Proposition 3.6 in [2] states the following.

Proposition 4.13. $\mathcal{H}^{2,V}$ and $\mathcal{H}^{2,\perp V}$ are orthogonal sub-Hilbert spaces of \mathcal{H}^2_0 and $\mathcal{H}_0^2 = \mathcal{H}^{2,V} \oplus^{\perp} \mathcal{H}^{2,\perp V}$. Moreover, any element of $\mathcal{H}_{loc}^{2,V}$ is strongly orthogonal to any element of $\mathcal{H}_{loc}^{2,\perp V}$.

For any $M \in \mathcal{H}_0^2$, we denote by M^V its projection on $\mathcal{H}^{2,V}$.

We can now finally establish the main result of the present note.

Proposition 4.14. Let V be a continuous non-decreasing function. Let M, N be two square integrable MAFs, and assume that the AF $\langle N \rangle$ is absolutely continuous with respect to V. There exists a function $v \in \mathcal{B}([0,T] \times E, R)$ such that for any (s, x) , $\langle M^{s,x}, N^{s,x} \rangle = \int_s^{\sqrt{s}} v(r, X_r) V_r$.

Proof. By Corollary 4.9, there exists a bounded variation AF with L^1 values denoted $\langle M, N \rangle$ such that under any P^{s,x}, the cadlag version of $\langle M, N \rangle$ is $\langle M^{s,x}, N^{s,x} \rangle$.

By Proposition 4.11, there exists an increasing AF with $L¹$ values denoted $Pos(\langle M, N \rangle)$ (resp. $Neg(\langle M, N \rangle)$) such that under any P^{s,x}, the cadlag version of $Pos(\langle M, N \rangle)$ (resp. $Neg(\langle M, N \rangle)$) is $Pos(\langle M^{s,x}, N^{s,x} \rangle)$ (resp. $Neg(\langle M^{s,x}, N^{s,x} \rangle)$). We fix some (s, x) and the associated probability $P^{s,x}$. Since $\langle N \rangle$ is absolutely continuous with respect to V , comparing Definition 4.1 and Notation 4.12 we have $N^{s,x} \in \mathcal{H}^{2,V}$. Therefore by Proposition 4.13 we have

$$
\langle M^{s,x}, N^{s,x} \rangle = \langle (M^{s,x})^V, N^{s,x} \rangle
$$

= $\frac{1}{4} \langle (M^{s,x})^V + N^{s,x} \rangle - \frac{1}{4} \langle (M^{s,x})^V - N^{s,x} \rangle.$ (4.24)

Since both processes $\frac{1}{4} \langle (M^{s,x})^V + N^{s,x} \rangle$, $\frac{1}{4} \langle (M^{s,x})^V - N^{s,x} \rangle$ are increasing and starting at zero, we have $Pos(\langle M^{s,x}, N^{s,x}\rangle) = \frac{1}{4}\langle (M^{s,x})^V + N^{s,x}\rangle$ and $Neg(\langle M^{s,x}, N^{s,x}\rangle) = \frac{1}{4}\langle (M^{s,x})^V - N^{s,x}\rangle$. Now since $(M^{s,x})^V + N^{s,x}$ and $(M^{s,x})^V - N^{s,x}$ belong to $\mathcal{H}^{2,V}$, we have shown that $dPos(\langle M^{s,x}, N^{s,x}\rangle) \ll dV$ and $dNeq(\langle M^{s,x}, N^{s,x}\rangle) \ll dV$ in the sense of stochastic measures.

Since this holds for all (s, x) Proposition 4.10 insures the existence of two functions v_+, v_- in $\mathcal{B}([0,T] \times E, R)$ such that for any (s, x) , $Pos(\langle M^{s,x}, N^{s,x} \rangle) =$ $\int_s^{\cdot \vee s} v_+(r, X_r) dV_r$ and $Neg(\langle M^{s,x}, N^{s,x}\rangle) = \int_s^{\cdot \vee s} v_-(r, X_r) dV_r$. The conclusion now follows setting $v = v_+ - v_-$. \Box

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