

COMMUTATION OF VARIATION AND DUAL PROJECTION

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ABSTRACT. For a raw process of integrable variation V , taking values in a Banach space E having the Radon-Nikodym property, the variation of the predictable (optional) dual projection is the predictable (optional) dual projection of the variation. An analogous result holds for the associated stochastic measures. The result is applied to the stochastic integral of a real, optional process H with respect to V when V is adapted.

1. INTRODUCTION

In a series of articles [4–7], Dinculeanu has detailed the theory of Banach-valued stochastic processes. In particular, given a raw process of integrable variation V , the existence of predictable and optional dual projections, denoted by V^p and V^o respectively, is established [6, Theorems 14 and 15]. The variations of these processes satisfy the inequalities $|V^p| \leq |V|^p$ and $|V^o| \leq |V|^o$ [6].

In this article, we use an alternate construction of the dual projection to show that in fact $|V^p| = |V|^p$ and $|V^o| = |V|^o$. We obtain analogous equalities for the associated stochastic measures μ_p and μ_o , which improves upon the inequalities in Lemma 1 from [7]. Lastly, we apply this commutation result to provide a sufficient condition for a real optional process H to be stochastically integrable with respect to an adapted process of integrable variation V .

Throughout, we let (Ω, \mathcal{F}, P) be a probability space with a filtration (\mathcal{F}_t) which satisfies the usual conditions. We let E be a Banach space which has the Radon-Nikodym property and let V be an E -valued, raw process of integrable variation which is also separably valued.

2. THE VARIATION OF THE PROJECTION

We proceed immediately with the result on commutation of variation and dual projection.

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Theorem 1. *The variation of the predictable (resp. optional) dual projection of V equals the predictable (resp. optional) dual projection of the variation. That is, $|V^p| = |V|^p$ (resp. $|V^o| = |V|^o$).*

Proof. We prove the result for the predictable projection. The other case is analogous. We define a stochastic measure μ on the predictable σ -algebra by $\mu(\cdot) = E[\int_0^\infty 1_{(\cdot)} dV_s]$ and note that its variation is given by $|\mu|(\cdot) = E[\int_0^\infty 1_{(\cdot)} d|V|_s]$ [4, Theorem 3.2]. Since μ is absolutely continuous with respect to $|\mu|$, we let $Q = d\mu/d|\mu|$ be an E -valued predictable density such that $\|Q\| = 1$, $|\mu|$ almost everywhere. Then for all predictable sets B ,

$$E \left[\int_0^\infty 1_B dV_s \right] = E \left[\int_0^\infty 1_B Q_s d|V|_s \right].$$

We now let $W = |V|^p$. Then for every real, bounded, measurable process H , with predictable projection ${}^p H$,

$$\begin{aligned} E \left[\int_0^\infty {}^p H_s dV_s \right] &= E \left[\int_0^\infty {}^p H_s Q_s d|V|_s \right] \\ &= E \left[\int_0^\infty {}^p (H_s Q_s) d|V|_s \right] \\ &= E \left[\int_0^\infty H_s Q_s dW_s \right]. \end{aligned}$$

Thus, $V_t^p = \int^t Q_s dW_s$. However, since W is an increasing process, dW_s defines a scalar-valued measure on $B(R^+)$ (for each w). Hence, the total variation of V^p is given by $|V^p|_t = \int^t \|Q_s\| dW_s$ [3, §10.9, Theorem 6, p. 186].

On the other hand, since $\| \|Q_s\| - 1 \|$ is a predictable process,

$$\begin{aligned} E \left[\int_0^\infty \| \|Q_s\| - 1 \| dW_s \right] &= E \left[\int_0^\infty \| \|Q_s\| - 1 \| d|V|_s \right] \\ &= \int \| \|Q_s\| - 1 \| d|\mu| = 0; \end{aligned}$$

hence, $\int_0^\infty \| \|Q_s\| - 1 \| dW_s = 0$ a.e. It follows that $\int^t \|Q_s\| dW_s$ and $\int^t 1 dW_s$ are indistinguishable. Hence, $|V^p| = \int^t dW_s = |V|^p$.

Corollary. *Let E be separable and let μ be an E -valued P -measure on $B(R^+) \times \mathcal{F}$ with finite variation $|\mu|$, with predictable (resp. optional) projection denoted by μ_p (resp. μ_o). Then, $|\mu_p| = |\mu|_p$ (resp. $|\mu_o| = |\mu|_o$).*

Proof. We let V be the associated raw process of integrable variation associated with μ which satisfies the following [4, Theorems 4.1, 5.1]:

$$\mu(\cdot) = E \left[\int_0^\infty 1_{(\cdot)} dV_s \right], \quad |\mu|(\cdot) = E \left[\int_0^\infty 1_{(\cdot)} d|V|_s \right].$$

Then since V^p is the process associated with μ_p , we have

$$\begin{aligned} |\mu_p|(\cdot) &= E \left[\int_0^\infty 1_{(\cdot)} d|V^p|_s \right] \\ &= E \left[\int_0^\infty 1_{(\cdot)} d|V|_s^p \right] = |\mu|_p(\cdot). \end{aligned}$$

The argument is analogous for the optional projection. \square

3. OPTIONAL STOCHASTIC INTEGRATION

As an application of Theorem 1, we shall prove a sufficient condition for a real-valued optional process H to be stochastically integrable, as defined in [8] with respect to an adapted process of integrable variation V . We also describe the form of the stochastic integral $H \cdot V$.

Theorem 2. *Let H be a real-valued optional process such that ${}^p H$ exists and let V be an E -valued, adapted process of integrable variation. If $E[\int^\infty |H_s| d|V|_s] < \infty$ and $E[\int^\infty {}^p |H_s| d|V|_s] < \infty$, then $H \in \mathcal{L}_1(I_V)$ and the optional stochastic integral of H with respect to V is given by*

$$(H \cdot V)_t = \int^t H_s dV_s - \left(\int^{\cdot} H_s dV_s \right)_t^p + \int^t {}^p H_s dV_s^p.$$

Proof. From [8], the measure I_V is defined on the optional σ -algebra \mathcal{O} by $I_V(B) = \int^\infty 1_B dV_s - (\int^{\cdot} 1_B dV_s)_\infty^p + \int^\infty {}^p(1_B) dV_s^p$, where $B \in \mathcal{O}$, and hence

$$(1_B \cdot V)_t = \int 1_B 1_{[0,t]} dI_V = \int^t 1_B dV_s - \left(\int^{\cdot} 1_B dV_s \right)_t^p + \int^t {}^p(1_B) dV_s^p.$$

By additivity, the result holds for the generating class of processes H of the form $1_{[S_1, T_1]} + \dots + 1_{[S_n, T_n]}$, where S_i, T_i are stopping times with $S_1 \leq T_1 \leq \dots \leq S_n \leq T_n$. We shall use a monotone class argument to first show the result for bounded H , since if H is bounded then ${}^p H$ exists and is also bounded [2, VI.43]. Consequently, both $E[\int^\infty |H_s| d|V|_s]$ and $E[\int^\infty {}^p |H_s| d|V|_s]$ are finite since V is of integrable variation.

If the theorem holds for a nonnegative sequence $\{H^n\}$ which is uniformly bounded and increases to H , or for a sequence converging uniformly to H , then by dominated convergence $\{H^n\}$ will also converge to H in the Lebesgue space $L_1(I_V)$. Thus, the $L_1(E)$ -valued stochastic integrals $\{(H^n \cdot V)_t\}$ will converge for each t to $(H \cdot V)_t$ since

$$\begin{aligned} \|(H^n \cdot V)_t - (H \cdot V)_t\|_{L_1(E)} &= \left\| \int (H^n - H) 1_{[0,t]} dI_V \right\|_{L_1(E)} \\ &\leq \|(H^n - H)\|_{L_1(I_V)}. \end{aligned}$$

We see that all three pieces of $H^n \cdot V$ will converge in $L_1(E)$ to the appropriate pieces of $H \cdot V$. For if we let $W_t^n = \int^t (H_s^n - H_s) dV_s$, then $E[\|W_t^n\|] \leq E[\int^\infty |H_s^n - H_s| d|V|_s] \rightarrow 0$, by dominated convergence. Also,

$$\begin{aligned} E[\|(W^n)_t^p\|] &\leq E \left[\int^\infty d|(W^n)_s^p| \right] \\ &= E \left[\int^\infty d|W^n_s|^p \right] \\ &= E \left[\int^\infty d|W^n_s| \right] \\ &\leq E \left[\int^\infty |H_s^n - H_s| d|V|_s \right]; \end{aligned}$$

hence, $(W^n)_t^p \rightarrow 0$ in $L_1(E)$. Moreover, since V^p is also of integrable variation,

$$\begin{aligned} E \left[\left\| \int^t {}^p(H^n - H)_s dV_s^p \right\| \right] &\leq E \left[\int^\infty |{}^p(H^n - H)_s| d|V^p|_s \right] \\ &\leq E \left[\int^\infty {}^p|H_s^n - H_s| d|V^p|_s \right] \\ &= E \left[\int^\infty |H_s^n - H_s| d|V^p|_s \right] \\ &\rightarrow 0. \end{aligned}$$

Hence, in $L_1(E)$,

$$\begin{aligned} (H \cdot V)_t &= \lim_{n \rightarrow \infty} (H^n \cdot V)_t = \lim_{n \rightarrow \infty} \left[\int^t H_s^n dV_s - \left(\int^{\cdot} H_s^n dV_s \right)_t^p + \int^t {}^p H_s^n dV_s^p \right] \\ &= \int^t H_s dV_s - \left(\int^{\cdot} H_s dV_s \right)_t^p + \int^t {}^p H_s dV_s^p. \end{aligned}$$

We thus obtain equality outside of a null set for each t ; but by right continuity, we obtain indistinguishability. Thus, H itself satisfies the theorem and by the monotone class theorem, the result holds for all bounded H .

Now suppose H satisfies the hypotheses of the theorem. We let $f \in \mathcal{L}_\infty(E^*)$ and let Y be a cadlag version of the martingale $E(f|\mathcal{F}_t)$. We denote the $L_1(E)$ -valued measure I_V by m . The real-valued measure m_f defined on \mathcal{O} is given by $m_f(B) = \langle f, m(B) \rangle = E[f \cdot m(B)]$. From [1, p. 360], the $L_1(I_V)$ norm of H is given by

$$\|H\|_{L_1(I_V)} = \sup_{\|f\| \leq 1} \int |H| d|m_f|,$$

where $\|f\|$ is the $L_\infty(E^*)$ norm of f . Thus, $H \in \mathcal{L}_1(I_V)$, and hence is stochastically integrable, if and only if this norm is finite. However, letting $A_t = \int^t 1_B dV_s$, for a fixed $B \in \mathcal{O}$, then

$$\begin{aligned} m_f(B) &= E \left[f \left(A_\infty - A_\infty^p + \int^\infty {}^p(1_B) dV_s^p \right) \right] \\ &= E \left[\int^\infty f dA_s \right] - E \left[\int^\infty f dA_s^p \right] + E \left[\int^\infty f {}^p(1_B) dV_s^p \right] \\ &= E \left[\int^\infty {}^o(f) dA_s \right] - E \left[\int^\infty {}^p(f) dA_s \right] + E \left[\int^\infty {}^p(f) {}^p(1_B) dV_s^p \right] \\ &= E \left[\int^\infty \Delta Y_s dA_s \right] + E \left[\int^\infty Y_{s-} {}^p(1_B) dV_s^p \right] \\ &= E \left[\int^\infty \Delta Y_s 1_B dV_s \right] + E \left[\int^\infty {}^p(Y_{s-} 1_B) dV_s^p \right] \\ &= E \left[\int^\infty \Delta Y_s 1_B dV_s \right] + E \left[\int^\infty Y_{s-} 1_B dV_s^p \right]. \end{aligned}$$

We see that m_f can be written as the sum of two stochastic P -measures: $m_f(B) = \sigma(B) + \tau(B)$. Hence, $|m_f| \leq |\sigma| + |\tau|$. But since we take $\|f\| \leq 1$,

$\|Y\| \leq 1$ also; hence, $|\sigma|(B) \leq E[\int^\infty \|\Delta Y_s\| 1_B d|V|_s] \leq E[\int^\infty 2 1_B |dV_s|]$ and $|\tau|(B) \leq E[\int^\infty 1_B d|V^p|_s] = E[\int^\infty 1_B d|V|_s^p]$, for all f .

Finally, we obtain

$$\begin{aligned} \|H\|_{L_1(I_V)} &\leq \int |H| d|\sigma| + \int |H| d|\tau| \\ &\leq 2E \left[\int^\infty |H_s| d|V|_s \right] + E \left[\int^\infty |H_s| d|V|_s^p \right] \\ &= 2E \left[\int^\infty |H_s| d|V|_s \right] + E \left[\int^\infty |H_s|^p d|V|_s \right] \\ &< \infty. \end{aligned}$$

Lastly, if we let $H^n = H 1_{\{|H| \leq n\}}$, then $\{H^n\}$ are bounded, converge to H in $L_1(I_V)$, and $\{H^n \cdot V\}$ satisfies the stated form of the theorem. By taking limits as in the monotone class argument, we see that $H \cdot V$ also satisfies the theorem. \square

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