

## THE ITÔ-CLIFFORD INTEGRAL. IV: A RADON-NIKODYM THEOREM AND BRACKET PROCESSES

C. BARNETT, R. F. STREATER and I. F. WILDE

### 0. INTRODUCTION

The construction and various properties of the Itô-Clifford stochastic integral have been discussed in [1, 2, 3]. In particular, it was shown in [1] that any centred

$L^2$ -martingale is given as an Itô-Clifford stochastic integral;  $X_t = \int_0^t \tilde{X}(s) d\Psi_s$ ,

where  $\Psi_s \equiv \Psi(\chi_{[0,s]})$  is the Fermi-field. It was also shown in [1] that stochastic integrals of the form  $\int f dX$  can be defined as elements of  $L^2(\mathcal{C})$ , the non-commutative  $L^2$ -space associated with the Clifford probability gage space. We consider the relationship between the stochastic integral with respect to  $\Psi$  and that with respect to  $X$ . Specifically, we prove a Radon-Nikodym theorem in the form:  $\int f dX = \int f \tilde{X} d\Psi$ .

Using the Doob-Meyer decomposition of the submartingale  $X_t^* X_t$ , given in [1], we define the pointed-bracket  $L^1$ -process  $\langle X_t, Y_t \rangle$  associated with  $L^2$ -martingales  $(X_t)$  and  $(Y_t)$ . The stochastic integral  $\int f dX$  is shown to be characterized, as a process, in terms of pointed-bracket processes. These results parallel those of standard (i.e. commutative) probability theory (see, for example [8,9]).

In Sections 1 and 2 we review and generalize some of the results from [1]. The stochastic integral  $\int f dX$  is defined in Section 3 — this being a simplified version of that in [1]. The Radon-Nikodym theorem is presented in Section 4, and in Section 5 an analogous result for stochastic integrals with respect to Wick martingales is proved. The pointed-bracket process is considered in Section 6, together with a characterization of the stochastic integral as a process.

Finally, in Section 7, we give a summary of the analogous results valid for “left” rather than “right” integrals.

A Doob-Meyer decomposition and stochastic integration with respect to martingales over an arbitrary probability space is considered in [4].

## 1. FOCK SPACE AND THE CLIFFORD ALGEBRA

We recall some notation and definitions from [1]. Let  $\Lambda(L^2(\mathbf{R}^+))$  denote the antisymmetric Fock space over  $L^2(\mathbf{R}^+)$ , and, for  $u \in L^2(\mathbf{R}^+)$ , let  $C(u)$  and  $A(u) = C(u)^*$  denote the creation and annihilation operators on  $\Lambda(L^2(\mathbf{R}^+))$ . The fermion field is defined as  $\Psi(u) = C(u) + A(\bar{u})$ , where  $\bar{u}$  is the complex-conjugate of  $u$  in  $L^2(\mathbf{R}^+)$ . The fermion fields satisfy the canonical anticommutation relations

$$(1.1) \quad \Psi(u)\Psi(v) + \Psi(v)\Psi(u) = 2(\bar{v}, u)\mathbf{1}.$$

For each  $t \geq 0$ ,  $\mathcal{C}_t$  denotes the von Neumann algebra generated by the fields  $\Psi(u)$  for  $u \in L^2(\mathbf{R}^+)$  with  $\text{supp } u \subseteq [0, t]$ , and  $\mathcal{C}$  is the von Neumann algebra generated by the increasing family  $\{\mathcal{C}_t : t \geq 0\}$ .  $\mathcal{C}$  is the weakly closed Clifford algebra over  $L^2(\mathbf{R}^+)$  [6, 11].

Let  $m$  denote the vector state  $m(x) = (\Omega, x\Omega)$ ,  $x \in \mathcal{C}$ , where  $\Omega$  is the Fock vacuum vector. Then  $m$  is a faithful, central state on  $\mathcal{C}$ .

For  $1 \leq p < \infty$ ,  $L^p(\mathcal{C})$  is the completion of  $\mathcal{C}$  with respect to the norm  $\|x\|_p = m(|x|^p)^{1/p}$ ,  $x \in \mathcal{C}$ , and  $L^\infty(\mathcal{C})$  is  $\mathcal{C}$  equipped with its  $C^*$ -norm.

The elements of  $L^p(\mathcal{C})$  can be identified with closed (possibly unbounded) operators on  $\Lambda(L^2(\mathbf{R}^+))$  [7,10]. In fact,  $L^p(\mathcal{C})$  consists of those closed operators  $Y$  on  $\Lambda(L^2(\mathbf{R}^+))$  affiliated to  $\mathcal{C}$  such that  $\Omega$  is in the domain of  $|Y|^{p/2}$ . Similarly, one defines  $L^p(\mathcal{C}_t)$  for  $t \geq 0$  and  $1 \leq p \leq \infty$ , so that  $L^p(\mathcal{C}_t)$  is a closed subspace of  $L^p(\mathcal{C})$ . Indeed, if  $\mathcal{B}$  is any von Neumann subalgebra of  $\mathcal{C}$ ,  $L^p(\mathcal{B})$  is a closed subspace of  $L^p(\mathcal{C})$ . The conditional expectation given  $\mathcal{B}$  is a contraction of  $L^p(\mathcal{C})$  onto  $L^p(\mathcal{B})$  for all  $1 \leq p \leq \infty$ , and is denoted  $m(\cdot|\mathcal{B})$ . We will write  $M_t$  for  $m(\cdot|\mathcal{C}_t)$ ,  $t \geq 0$ .

**DEFINITION 1.1.** An  $L^p(\mathcal{C})$ -martingale is a family  $\{X_t : t \geq 0\}$  with  $X_t \in L^p(\mathcal{C})$  for all  $t \geq 0$  and such that  $M_t X_s = X_t$  for all  $0 \leq t \leq s$ . It follows that  $X_t \in L^p(\mathcal{C}_t)$ ,  $t \geq 0$ .

It is convenient to define here the parity operator  $\beta$  and establish some of its properties. Let  $Q^0$  denote the linear space of even polynomials in the fields  $\Psi(u)$ ,  $u \in L^2(\mathbf{R}^+)$ , and let  $Q$  denote the von Neumann subalgebra of  $\mathcal{C}$  generated by  $Q^0$ .

**DEFINITION 1.2.** The *parity operator*  $\beta$  is the map  $\beta : L^p(\mathcal{C}) \rightarrow L^p(\mathcal{C})$ ,  $1 \leq p \leq \infty$ , given by  $\beta(f) = 2m(f|Q) - f$ ,  $f \in L^p(\mathcal{C})$ .

Evidently,  $\beta: L^p(\mathcal{C}) \rightarrow L^p(\mathcal{C})$  is continuous, and  $\beta(f)^* = \beta(f^*)$  for  $f \in L^p(\mathcal{C})$ . Moreover, using the fact that  $m(m(\cdot|Q)|Q) = m(\cdot|Q)$ , we see that  $\beta^2(f) = f$  for  $f \in L^p(\mathcal{C})$ . Furthermore, since  $m(\cdot|Q)$  defines the orthogonal projection of  $L^2(\mathcal{C})$  onto  $L^2(Q)$ , it follows that  $\beta: L^2(\mathcal{C}) \rightarrow L^2(\mathcal{C})$  is self-adjoint and unitary.

**PROPOSITION 1.3.** (i)  $\beta(f) = f$  for all  $f \in L^p(Q)$ ,  $1 \leq p \leq \infty$ .

(ii)  $\beta(x) = -x$  for any odd polynomial  $x$  in the fields  $\Psi(u)$ ,  $u \in L^2(\mathbf{R}^+)$ .

(iii)  $\beta: \mathcal{C} \rightarrow \mathcal{C}$  is  $\sigma$ -weakly continuous.

(iv)  $\beta(hg) = \beta(h)\beta(g)$  for  $h \in L^p(\mathcal{C})$ ,  $g \in L^q(\mathcal{C})$ , with  $1/p + 1/q = 1$ .

(v)  $\beta: L^p(\mathcal{C}) \rightarrow L^p(\mathcal{C})$  is isometric for  $1 \leq p \leq \infty$ .

(vi) If  $f \in L^1(\mathcal{C})$  and  $f \geq 0$ , then  $\beta(f) \geq 0$ .

(vii)  $\beta: L^p(\mathcal{C}_t) = L^p(\mathcal{C}_t)$  for all  $t \geq 0$ ,  $1 \leq p \leq \infty$ .

*Proof.* (i) Trivial using  $m(f|Q) = f$  for  $f \in L^p(Q)$ .

(ii) If  $x$  is an odd polynomial in the fields and  $y \in Q^0$ , one sees that  $y\Omega$  and  $x\Omega$  are orthogonal in  $A(L^2(\mathbf{R}^+))$ . Hence  $m(y^*x) = 0$ . By continuity it follows that  $m(y^*x) = 0$  for all  $y \in Q$ , and so  $m(x|Q) = 0$ . Thus  $\beta(x) = -x$ .

(iii) By continuity and self-adjointness of  $\beta$  on  $L^2(\mathcal{C})$ , we have  $m(g\beta(h)) = -m(\beta(g)h)$  for  $h \in L^p(\mathcal{C})$ ,  $g \in L^q(\mathcal{C})$  with  $1/p + 1/q = 1$ . In particular, this holds for  $q = 1$ ,  $p = \infty$ . But  $L^1(\mathcal{C})$  is the predual of  $\mathcal{C}$  [10] under the pairing  $L^1(\mathcal{C}) \times L^\infty(\mathcal{C}) \rightarrow \mathbf{C}$ ,  $(g, h) \mapsto m(gh)$ , and so we deduce that  $\beta: L^\infty(\mathcal{C}) \rightarrow L^\infty(\mathcal{C})$  is  $\sigma$ -weakly continuous.

(iv) Any polynomial in the fields can be written as a linear combination of an odd and an even polynomial. Now if  $x'$ ,  $x''$  are odd and  $y'$ ,  $y''$  are even polynomials in the fields, we have that  $x'x''$ ,  $y'y''$  are even and  $x'y''$  and  $x''y'$  are odd. Hence, using (i) and (ii),

$$\begin{aligned} \beta((x' + y')(x'' + y'')) &= \beta(x'x'' + x'y'' + y'x'' + y'y'') = \\ &= x'x'' - x'y'' - y'x'' + y'y'' = \\ &= \beta(x' + y')\beta(x'' + y''). \end{aligned}$$

By continuity, it follows that  $\beta(xy) = \beta(x)\beta(y)$  for any  $x, y \in \mathcal{C}$  and, again by continuity, the result follows.

(v) From (iv), and the fact that  $\beta(x) = 0$  implies that  $\beta^2(x) = x = 0$ , we see that  $\beta$  is an automorphism of  $\mathcal{C}$ . Hence  $\beta: \mathcal{C} \rightarrow \mathcal{C}$  is isometric. By duality it follows that  $\beta: L^1(\mathcal{C}) \rightarrow L^1(\mathcal{C})$  is a contraction, and so by interpolation [7]  $\beta: L^p(\mathcal{C}) \rightarrow L^p(\mathcal{C})$  is a contraction for  $1 \leq p < \infty$ . Now, for  $f \in L^p(\mathcal{C})$ ,  $1 \leq p < \infty$ ,

$$\|f\|_p = \|\beta^2(f)\|_p \leq \|\beta(f)\|_p \leq \|f\|_p.$$

It follows that  $\beta: L^p(\mathcal{C}) \rightarrow L^p(\mathcal{C})$  is an isometry, for  $1 \leq p \leq \infty$ .

(vi) If  $f \in L^1(\mathcal{C})$  with  $f \geq 0$ , then, there is  $g \in L^2(\mathcal{C})$  such that  $f = g^*g$ . Indeed, we can take  $g = f^{1/2}$ . Then, by (iv),  $\beta(f) = \beta(g^*g) = \beta(g)^*\beta(g) \geq 0$ .

(vii) This follows from (i), (ii), (iii) and (v). Q.E.D.

**REMARK 1.4.** Properties (i) and (ii) imply that the definition of  $\beta$  given here agrees with that in [1]. Indeed, the above results are readily obtained using the definition of  $\beta$  given in [1], but the proofs given here seem to be interesting in their own right.

## 2. THE ITÔ-CLIFFORD INTEGRAL

**DEFINITION 2.1.** An  $L^p(\mathcal{C})$ -valued process on  $[0, t]$  is a map  $f: [0, t] \rightarrow L^p(\mathcal{C})$  such that  $f(s) \in L^p(\mathcal{C}_s)$  for  $0 \leq s \leq t$ . Such a process is said to be *elementary* if it is of the form  $f = g\chi_{[r, t]}$  for some  $0 \leq r < \tau \leq t$  and  $g \in L^p(\mathcal{C})$ . Note that  $f(s) \in L^p(\mathcal{C}_s)$  implies that  $g \in L^p(\mathcal{C}_r)$ . An  $L^p$ -process on  $[0, t]$  is said to be *simple* if, on  $[0, t]$ , it is a finite sum of elementary  $L^p$ -processes. Denote by  $\mathcal{S}([0, t], L^p(\mathcal{C}))$  the linear space of simple  $L^p(\mathcal{C})$ -valued processes on  $[0, t]$ .

For  $u \in L^2_{\text{loc}}(\mathbf{R}^+)$ , set  $\Psi_t(u) = \Psi(u\chi_{[0, t]})$ . Then it was shown in [1] that  $\{\Psi_t(u) : t \geq 0\}$  is an  $L^\infty$ -martingale adapted to the filtration  $\{\mathcal{C}_t : t \geq 0\}$ .

**DEFINITION 2.2.** If  $f = g\chi_{[r, t]}$ ,  $0 \leq r < \tau \leq t$ , is an elementary  $L^\infty$ -valued process on  $[0, t]$ , the *Itô-Clifford stochastic integral* of  $f$  with respect to  $\Psi_s(u)$  is defined to be

$$\int_0^t f(s) d\Psi_s(u) = g(\Psi_t(u) - \Psi_r(u)).$$

The integral  $\int_0^t f(s) d\Psi_s(u)$  for  $f \in \mathcal{S}([0, t], L^\infty)$  is defined by linearity.

Let  $\mathfrak{H}([0, t], |u(s)|^2 ds)$  denote the subspace of processes in  $L^2([0, t], |u(s)|^2 ds; L^2(\mathcal{C}))$ . Then as in [1], one sees that  $\mathcal{S}([0, t], L^\infty)$  is dense in  $\mathfrak{H}([0, t], |u(s)|^2 ds)$ , and that the Itô-Clifford stochastic integral is well-defined, by continuity, as an element of  $L^2(\mathcal{C})$  for  $f \in \mathfrak{H}([0, t], |u(s)|^2 ds)$  and satisfies the isometry property:

$$(2.1) \quad \left\| \int_0^t f(s) d\Psi_s(u) \right\|_2^2 = \int_0^t \|f(s)\|_2^2 |u(s)|^2 ds.$$

Let  $\mathfrak{H}_{\text{loc}}(\mathbf{R}^+, |u(s)|^2 ds)$  denote those maps (i.e. classes of maps)  $f$  such that the restriction of  $f$  to each  $[0, t]$ ,  $t \geq 0$ , belongs to  $\mathfrak{H}([0, t], |u(s)|^2 ds)$ . Then it was shown

in [1] that for each  $f \in \mathfrak{H}_{\text{loc}}(\mathbf{R}^+, |u(s)|^2 ds)$ ,  $\left\{ \int_0^t f(s) d\Psi_s(u) : t \geq 0 \right\}$  is a centred  $L^2$ -martingale.

If  $u(s) = 1$  for  $s \in \mathbf{R}^+$ , write  $\Psi_t$  for  $\Psi_t(u)$ . Using Theorem 4.1 of [1], we have the following representation theorem.

**THEOREM 2.3.** *Let  $(X_t)$  be an  $L^2$ -martingale (adapted to the family  $\{\mathcal{C}_t : t \geq 0\}$ ). Then there is a unique element  $\tilde{X}$  of  $\mathfrak{H}_{\text{loc}}(\mathbf{R}^+, ds)$  such that*

$$X_t = X_0 + \int_0^t \tilde{X}(s) d\Psi_s \quad \text{for } t \geq 0.$$

*Proof.* Since  $(X_t - X_0)$  is centred, the existence of  $\tilde{X}$  follows from Theorem 4.1 of [1]. The uniqueness follows immediately from the isometry property equation (2.1).

Q.E.D.

Our first aim is to generalize this theorem.

**LEMMA 2.4.** *For  $u \in L^2_{\text{loc}}(\mathbf{R}^+)$  and  $h \in \mathfrak{H}([0, t], |u(s)|^2 ds)$ , we have*

$$(2.2) \quad \int_0^t h(s) d\Psi_s(u) = \int_0^t h(s) u(s) d\Psi_s.$$

*Proof.* We first note that  $hu \in \mathfrak{H}([0, t], ds)$  and so the right-hand side is well-defined. If  $h$  is elementary and  $u = \chi_{[s_1, s_2]}$  for some  $0 \leq s_1 \leq s_2 \leq t$ , the result is clear. Now, for any  $v \in L^2(\mathbf{R}^+)$ ,  $\|\Psi(v)\|_\infty = \|v\|$ , so  $v \mapsto \Psi(v)$  is a linear continuous map from  $L^2(\mathbf{R}^+)$  into  $L^\infty(\mathcal{C})$ . Hence, by linearity and continuity, the result holds for  $h$  elementary and  $u \in L^2_{\text{loc}}(\mathbf{R}^+)$ . But then again by linearity and continuity the result holds for arbitrary  $u \in L^2_{\text{loc}}(\mathbf{R}^+)$  and  $h \in \mathfrak{H}([0, t], |u(s)|^2 ds)$ . Q.E.D.

**REMARK.** This is a Radon-Nikodym theorem.

**THEOREM 2.5.** *Let  $u \in L^2_{\text{loc}}(\mathbf{R}^+)$  and let  $(X_t)$  be an  $L^2$ -martingale. There is  $g \in \mathfrak{H}_{\text{loc}}(\mathbf{R}^+, |u(s)|^2 ds)$  such that  $X_t = X_0 + \int_0^t g(s) d\Psi_s(u)$ ,  $t \geq 0$ , if and only if  $\{s : \tilde{X}(s) \neq 0\} \subseteq \{s : u(s) \neq 0\}$  up to a set of (Lebesgue) measure zero. If such  $g$  exists, it is unique.*

*Proof.* Let  $E = \{s \in \mathbf{R}^+ : u(s) \neq 0\}$ , and set  $v = u + \chi_E c$ . Then  $v \neq 0$  (Lebesgue) almost everywhere, and so  $h(s) = v(s)^{-1} \tilde{X}(s)$  is well-defined (Lebesgue) almost everywhere, and determines an element of  $\mathfrak{H}_{\text{loc}}(\mathbf{R}^+, |v(s)|^2 ds)$ . If

$\{s : \tilde{X}(s) \neq 0\} \subseteq \{s : u(s) \neq 0\}$  up to a set of Lebesgue measure zero, we have  $\tilde{X} = \chi_E \tilde{X} = \chi_E v h = \chi_E u h$  (Lebesgue) almost everywhere. Putting  $g = \chi_E h$ , we see that  $g \in \mathfrak{H}_{loc}(\mathbf{R}^+, |u(s)|^2 ds)$ , and by Lemma 2.4, for  $t \geq 0$ ,

$$X_t - X_0 = \int_0^t \tilde{X}(s) d\Psi_s = \int_0^t \chi_E(s) u(s) h(s) d\Psi_s = \int_0^t g(s) d\Psi_s(u).$$

The uniqueness of  $g$  follows from the isometry property, equation 2.1.

Conversely, suppose that there is  $g \in \mathfrak{H}_{loc}(\mathbf{R}^+, |u(s)|^2 ds)$  such that  $X_t =: X_0 + \int_0^t g(s) d\Psi_s(u)$ ,  $t \geq 0$ . Then, by Lemma 2.4,

$$X_t - X_0 = \int_0^t g(s) u(s) d\Psi_s, \quad t \geq 0.$$

Hence, since  $gu \in \mathfrak{H}_{loc}(\mathbf{R}^+, ds)$ , it follows from Theorem 2.3 that  $\tilde{X} = gu$ , and the proof is complete. Q.E.D.

REMARK. As is customary, for convenience (and brevity) we have treated elements of  $L_{loc}^2(\mathbf{R}^+)$  and  $\mathfrak{H}$  as if they were maps rather than equivalence classes of maps.

### 3. EXTENSION OF THE ITÔ-CLIFFORD INTEGRAL

We shall construct stochastic integrals with respect to an arbitrary  $L^2$ -martingale and, in the next section, relate them to the Itô-Clifford integral by means of a Radon-Nikodym theorem. The construction here is a simplified version of that given in § 7 of [1].

DEFINITION 3.1. For an  $L^2$ -martingale  $(X_t)$ , let  $\mu_X$  denote the Borel measure on  $\mathbf{R}^+$  given by  $\mu_X([0, t]) = \int_0^t |\tilde{X}(s)|^2 ds$ , where  $\tilde{X}$  is given by Theorem 2.3.

We shall sometimes use  $\lambda$  to denote Lebesgue measure on  $\mathbf{R}^+$ .

DEFINITION 3.2. Let  $f = g\chi_{[r, t]}$  be an elementary  $L^\infty$ -valued process on  $[0, t]$ .

The *stochastic integral of  $f$  with respect to the  $L^2$ -martingale  $(X_t)$*  is  $\int_0^t f(s) dX_s = g(X_t - X_r) \in L^2(\mathcal{C})$ .  $\int_0^t f(s) dX_s$  for  $f \in \mathcal{S}([0, t], L^\infty)$  is defined by linearity.

We shall extend the definition of the stochastic integral to more general integrands using a contraction property. First we recall the following result from [1].

**THEOREM 3.3.** *If  $(X_t)$  is an  $L^2$ -martingale, then  $Z_t = X_t^* X_t - \int_0^t |\beta(\tilde{X}(s))|^2 ds$  is an  $L^1$ -martingale.*

*Proof.* Use Theorem 2.3 and [1, Theorem 3.18]. Q.E.D.

**THEOREM 3.4** (Contraction property). *For  $f \in \mathcal{S}([0, t], L^\infty)$ , we have*

$$(3.1) \quad \left\| \int_0^t f(s) dX_s \right\|_2^2 \leq \int_0^t \|f(s)\|_\infty^2 d\mu_X(s).$$

*Proof.* Suppose that  $f = \sum_{k=1}^n h_{k-1} \chi_{[t_{k-1}, t_k)}$  on  $[0, t]$ , with  $0 \leq t_0 \leq \dots \leq t_n = t$ , is a simple  $L^\infty$ -valued process. Then, writing  $\Delta X_k$  for  $X_{t_k} - X_{t_{k-1}}$ , we have

$$\begin{aligned} & \left\| \int_0^t f(s) dX_s \right\|_2^2 = \sum_{k,j} m(\Delta X_k^* h_{k-1}^* h_{j-1} \Delta X_j) = \sum_k m(\Delta X_k^* |h_{k-1}|^2 \Delta X_k) = \\ & \quad (\text{using the martingale property}) \\ & = \sum_k m(|h_{k-1}|^2 \Delta X_k \Delta X_k^*) \leq \sum_k \|h_{k-1}\|_\infty^2 \|\Delta X_k^*\|_2^2 = \\ & = \sum_k \|h_{k-1}\|_\infty^2 \|\Delta X_k\|_2^2 = \sum_k \|h_{k-1}\|_\infty^2 (\|X_{t_k}\|_2^2 - \|X_{t_{k-1}}\|_2^2) = \quad (\text{using the martingale property}) \\ & = \sum_k \|h_{k-1}\|_\infty^2 \int_{t_{k-1}}^{t_k} \|\beta \tilde{X}(s)\|_2^2 ds = \quad (\text{using Theorem 23.3}) \\ & = \sum_k \|h_{k-1}\|_\infty^2 \int_{t_{k-1}}^{t_k} \|\tilde{X}(s)\|_2^2 ds = \quad (\text{since } \beta : L^2 \rightarrow L^2 \text{ is isometric}) \\ & = \int_0^t \|f(s)\|_\infty^2 d\mu_X(s). \end{aligned}$$

Q.E.D.

**DEFINITION 3.5.** Let  $\mathcal{K}([0, t], \mu_X)$  denote the closure of  $\mathcal{S}([0, t], L^\infty)$  in  $L^2([0, t], d\mu_X; L^\infty(\mathcal{C}))$ .

**COROLLARY 3.6.** *For any  $f \in \mathcal{K}([0, t], \mu_X)$ , let  $(g_n)$  be a sequence in  $\mathcal{S}([0, t], L^\infty)$  such that  $g_n \rightarrow f$  in  $\mathcal{K}([0, t], \mu_X)$ . Then there exists  $L^2(\mathcal{C})$ -lim <sub>$n$</sub>   $\int_0^t g_n(s) dX_s$ . This limit is independent of the particular sequence  $(g_n)$  converging to  $f$ ; and is denoted  $\int_0^t f(s) dX_s$ .*

Furthermore

$$\left\| \int_0^t f(s) dX_s \right\|_2^2 \leq \int_0^t \|f(s)\|_\infty^2 d\mu_X(s).$$

*Proof.* This is an immediate consequence of Theorem 3.4. Q.E.D.

**REMARK 3.7.** It is not clear whether  $\mathcal{K}([0, t], \mu_X)$  is the set of all processes in  $L^2([0, t], d\mu_X; L^\infty(\mathcal{C}))$  or not. The obstruction to the analogue of the proof in [1] for  $\mathfrak{H}$  concerns the continuity of the map  $s \mapsto M_s g$  for fixed  $g \in L^\infty(\mathcal{C})$ . This is continuous as a map:  $\mathbf{R}^+ \rightarrow L^2(\mathcal{C})$  but we do not know whether or not it is also continuous as a map:  $\mathbf{R}^+ \rightarrow L^\infty(\mathcal{C})$ . However, using the continuity of  $v \mapsto \Psi(v) \in L^\infty(\mathcal{C})$ , it is not difficult to see that  $s \mapsto M_s g$  is continuous as a map from  $\mathbf{R}^+$  into  $\mathfrak{A}$  for any  $g \in \mathfrak{A}$ , the  $C^*$ -algebra generated by the fields  $\Psi(v)$ ,  $v \in L^2(\mathbf{R}^+)$ . Thus, by copying the proof of Theorem 3.9 in [1], we see that  $\mathcal{K}([0, t], \mu_X)$  contains the set of processes in  $L^2([0, t], d\mu_X; \mathfrak{A})$ .

If  $f: \mathbf{R}^+ \rightarrow L^\infty(\mathcal{C})$  is such that the restriction of  $f$  to  $[0, t]$  belongs to  $\mathcal{K}([0, t], \mu_X)$  for each  $t \geq 0$ , then  $f\tilde{X} \in \mathfrak{H}_{loc}(\mathbf{R}^+, ds)$  and it is easy to see [1] that  $\int_0^t f dX$  is a centred  $L^2(\mathcal{C})$ -martingale.

#### 4. A RADON-NIKODYM THEOREM

In this section, we shall relate the stochastic integral  $\int_0^t f dX$  to the Itô-Clifford stochastic integral. Indeed, Theorem 2.3 suggests that formally  $dX = \tilde{X}d\Psi$ , and so we would expect  $\int_0^t f dX = \int_0^t f \tilde{X} d\Psi$ . We shall see that this is the case.

**LEMMA 4.1.** *Let  $(X_t)$  be an  $L^2$ -martingale, and let  $f \in \mathcal{S}([0, t], L^\infty(\mathcal{C}))$ . Then*

$$\int_0^t f(s) dX_s = \int_0^t f(s) \tilde{X}(s) d\Psi_s$$

where  $\tilde{X}$  is given by Theorem 2.3.

*Proof.* By linearity, we may suppose that  $f$  is elementary: say,  $f = g\chi_{[r, \tau]}$ ,  $0 \leq r < \tau \leq t$ ,  $g \in L^\infty(\mathcal{C})$ . Then

$$\int_0^t f(s) dX_s = g(X_\tau - X_r) = g \int_r^\tau \tilde{X}(s) d\Psi_s = \int_r^\tau g \tilde{X}(s) d\Psi_s =$$

(since left-multiplication by an element of  $L^\infty$  is continuous from  $L^2(\mathcal{C}) \rightarrow L^2(\mathcal{C})$ )

$$= \int_0^t f(s) \tilde{X}(s) d\Psi_s. \quad \text{Q.E.D.}$$

**THEOREM 4.2** (Radon-Nikodym theorem for stochastic integrals). *Let  $(X_t)$  be an  $L^2$ -martingale and let  $f \in \mathcal{K}([0, t], \mu_X)$ . Then  $f\tilde{X} \in \mathfrak{H}([0, t], ds)$  and*

$$\int_0^t f(s) dX_s = \int_0^t f(s) \tilde{X}(s) d\Psi_s.$$

*Proof.* Let  $(g_n)$  be a sequence in  $\mathcal{S}([0, t], L^\infty(\mathcal{C}))$  such that  $g_n \rightarrow f$  in  $\mathcal{K}([0, t], \mu_X)$ . By passing to a subsequence if necessary we may suppose that  $(g_n(s) - f(s))\|\tilde{X}(s)\|_2^2 \rightarrow 0$  in  $L^\infty(\mathcal{C})$   $\lambda$  a.e., and hence  $g_n \tilde{X} \rightarrow f\tilde{X}$  in  $L^2(\mathcal{C})$   $\lambda$  a.e. on  $[0, t]$ .

Now

$$\begin{aligned} \int_0^t f dX &= L^2\text{-lim} \int_0^t g_n dX = \\ &= L^2\text{-lim} \int_0^t g_n \tilde{X} d\Psi, \quad \text{by Lemma 4.1.} \end{aligned}$$

In particular,  $(g_n \tilde{X})$  is a Cauchy sequence in  $\mathfrak{H}([0, t], ds)$  and thus there is  $G$  such that  $g_n \tilde{X} \rightarrow G$  in  $\mathfrak{H}([0, t], ds)$ .

Again, by passing to a subsequence if necessary, we may suppose that  $g_n \tilde{X} \rightarrow G$  in  $L^2(\mathcal{C})$   $\lambda$  a.e. on  $[0, t]$ .

Hence  $f\tilde{X} = G$   $\lambda$  a.e. on  $[0, t]$ ; i.e.  $f\tilde{X} \in \mathfrak{H}([0, t], ds)$  and  $g_n \tilde{X} \rightarrow f\tilde{X}$  in  $\mathfrak{H}([0, t], ds)$ .

By the isometry property, equation (2.1), it follows that  $\int_0^t g_n \tilde{X} d\Psi \rightarrow \int_0^t f \tilde{X} d\Psi$

in  $L^2(\mathcal{C})$  and the proof is complete.

Q.E.D.

**COROLLARY 4.3.** *Let  $(X_t)$  be an  $L^2$ -martingale and let  $g \in \mathcal{K}([0, t], \mu_X)$  for each  $t \geq 0$ . Then if  $Y_t = \int_0^t g(s) dX_s$ ,  $t \geq 0$ , we have*

$$\int_0^t f(s) dY_s = \int_0^t f(s)g(s)\tilde{X}(s) d\Psi_s$$

for any  $f \in \mathcal{K}([0, t], \mu_Y)$ .

*Proof.* By the theorem,  $\int f dY = \int f \tilde{Y} d\Psi$ . But  $Y = \int g dX = \int g \tilde{X} d\Psi$ , again by the theorem. Hence  $\tilde{Y} = g \tilde{X}$  and the result follows. Q.E.D.

## 5. STOCHASTIC INTEGRALS WITH RESPECT TO WICK MONOMIAL MARTINGALES

For given real  $u_1, \dots, u_n \in L^2_{\text{loc}}(\mathbf{R}^+)$ , the Wick monomial martingale  $W(u_1, \dots, u_n; s)$ ,  $s \geq 0$  is defined as  $W(u_1, \dots, u_n; s) = : \Psi(u_1 \chi_{[0, s]}) \dots \Psi(u_n \chi_{[0, s]}) :$ , where  $: \dots :$  denotes Wick ordering. It was shown in [1] that  $W_s = W(u_1, \dots, u_n; s)$ ,  $s \geq 0$  is an  $L^\infty$ -martingale, and that stochastic integrals  $\int_0^t f(s) dW_s$  can be constructed as elements of  $L^2(\mathcal{C})$  for  $f \in \mathfrak{H}([0, t], dv)$ , the set of processes in  $L^2([0, t], dv; L^2(\mathcal{C}))$ , where  $v$  is the measure on  $\mathbf{R}^+$  given by  $v([s, t]) = a_t - a_s$  and  $a_s \in \mathbf{R}$  is  $a_s = W_s^* W_s = W_s W_s^*$ .

Writing  $W_t = \int_0^t \tilde{W}(s) d\Psi_s$ , we have, by the isometry property

$$a_t = m(W_t^* W_t) = \|W_t\|_2^2 = \int_0^t \|\tilde{W}(s)\|_2^2 ds.$$

Hence  $v$  is the measure  $\mu_W$ .

The closure of  $\mathcal{S}([0, t], L^\infty(\mathcal{C}))$  in  $L^2([0, t], d\mu_W; L^2(\mathcal{C}))$  is equal to  $\mathfrak{H}([0, t], d\mu_W)$  and so Theorem 4.3 has a sharper analogue here.

**THEOREM 5.1.** *Let  $f \in \mathfrak{H}([0, t], d\mu_W)$ . Then  $f \tilde{W} \in \mathfrak{H}([0, t], ds)$  and*

$$\int_0^t f(s) dW_s = \int_0^t f(s) \tilde{W}(s) d\Psi_s.$$

*Proof.* This is analogous to that of Theorem 4.2. Q.E.D.

REMARK 5.2. The isometry property for  $f \in \mathfrak{H}([0, t], d\mu_W)$  can be written as

$$\left\| \int_0^t f dW \right\|_2^2 = \int_0^t \|f(s)\|_2^2 \|\tilde{W}(s)\|_2^2 ds.$$

This follows because  $dv = d\mu_W = \|\tilde{W}(s)\|_2^2 ds$ . We see from the theorem that

$$\left\| \int_0^t f dW \right\|_2^2 = \int_0^t \|f(s)\tilde{W}(s)\|_2^2 ds$$

and so

$$\int_0^t \|f(s)\tilde{W}(s)\|_2^2 ds = \int_0^t \|f(s)\|_2^2 \|\tilde{W}(s)\|_2^2 ds.$$

We can formulate an analogue of Theorem 3.18 of [1].

THEOREM 5.3. Let  $f \in \mathfrak{H}([0, t], d\mu_W)$ . Then  $Z_t = \left| \int_0^t f dW \right|^2 - \int_0^t |\beta(f(s)\tilde{W}(s))|^2 ds$

is a centred  $L^1$ -martingale (on  $[0, t]$ ).

*Proof.* Use Theorem 5.1 together with [1, Theorem 3.18]. Q.E.D.

As a corollary, we note that  $\left( \int_0^t f dW \right)$  has a Doob-Meyer type decomposition

for any  $f \in \mathfrak{H}_{loc}(\mathbb{R}^+, d\mu_W)$ .

## 6. THE POINTED-BRACKET PROCESS

We shall define the so-called pointed-bracket process corresponding to a pair of  $L^2$ -martingales, and give a characterization of the process given by the Itô-Clifford integral.

If  $f, g \in L^2_{loc}(\mathbb{R}^+, d\lambda; L^2(\mathcal{C}))$ , then  $fg \in L^1_{loc}(\mathbb{R}^+, d\lambda; L^1(\mathcal{C}))$ , and so we may make the following definition.

DEFINITION 6.1. Let  $X_t = X_0 + \int_0^t \tilde{X} d\Psi$  and  $Y_t = Y_0 + \int_0^t \tilde{Y} d\Psi$ ,  $t \geq 0$ , be

$L^2$ -martingales. The *pointed-bracket* between  $(X_t)$  and  $(Y_t)$  is the  $L^1(\mathcal{C})$ -process

$$\langle X_t, Y_t \rangle = \int_0^t \beta(\tilde{Y}(s))^* \beta(\tilde{X}(s)) ds, \quad t \geq 0.$$

Notice that  $\langle X_t, X_t \rangle$  is the increasing  $L^1$ -process given in the Doob-Meyer decomposition of  $X_t^* X_t$  by Theorem 3.3.

Clearly,  $\langle \cdot, \cdot \rangle$  is a sesquilinear map from  $\mathfrak{M} \times \mathfrak{M}$  into the set of processes in  $L_{\text{loc}}^1(\mathbf{R}^+, d\lambda; L^1(\mathcal{C}))$ , where  $\mathfrak{M}$  denotes the set of  $L^2$ -martingales.

For any bounded Borel set  $E$  in  $\mathbf{R}^+$ , set  $\langle X, Y \rangle(E) = \int_E \beta(\tilde{Y}(s))^* \beta(\tilde{X}(s)) ds$ .

Then  $\langle X, Y \rangle$  defines an  $L^1(\mathcal{C})$ -valued countably additive Borel vector measure on any bounded interval  $[0, t]$  in  $\mathbf{R}^+$ .

**LEMMA 6.2.** *For  $X, Y \in \mathfrak{M}$ ,  $\langle X, Y \rangle$  has bounded variation on any interval  $[0, t]$  in  $\mathbf{R}^+$ , and is continuous with respect to Lebesgue measure on the Borel sets in  $[0, t]$ .*

*Proof.* By polarization, it is enough to show that  $\langle X, X \rangle$  has bounded variation on  $[0, t]$ . Now, the variation of  $\langle X, X \rangle$  on  $[0, t]$  is defined to be the set function

$$E \mapsto |\langle X, X \rangle|(E) = \sup_i \sum_i \|\langle X, X \rangle(E_i)\|_1$$

where  $E$  is any Borel set in  $[0, t]$ , and the supremum is taken over all partitions of  $E$  into a finite number of disjoint Borel sets.

But for any such partition  $\{E_i\}$  of a Borel set  $E$  in  $[0, t]$ , we have

$$\begin{aligned} \sum_i \|\langle X, X \rangle(E_i)\|_1 &= \sum_i \left\| \int_{E_i} \beta(\tilde{X}(s))^* \beta(\tilde{X}(s)) ds \right\|_1 = \\ &= \sum_i m \left( \int_{E_i} |\beta(\tilde{X}(s))|^2 ds \right) = \\ &\quad \left( \text{since } \int_{E_i} \beta(\tilde{X}(s))^* \beta(\tilde{X}(s)) ds \text{ is a non-negative element of } L^1(\mathcal{C}) \right), \\ &= m \left( \sum_i \int_{E_i} |\beta(\tilde{X}(s))|^2 ds \right) = \left\| \int_E |\beta(\tilde{X}(s))|^2 ds \right\|_1 = \|\langle X, X \rangle(E)\|_1. \end{aligned}$$

Hence, for any  $t \geq 0$ ,  $|\langle X, X \rangle|([0, t]) = \|\langle X, X \rangle([0, t])\|_1$ , and so  $\langle X, X \rangle$  has finite variation on  $[0, t]$ . The same is then true of  $\langle X, Y \rangle$ .

If  $E$  is a Borel set in  $[0, t]$  with Lebesgue measure  $\lambda(E) = 0$ , then

$$\langle X, Y \rangle(E) = \int_E \beta(\tilde{Y}(s))^* \beta(\tilde{X}(s)) ds = 0. \quad \text{Q.E.D.}$$

For suitable  $L^\infty(\mathcal{C})$ -valued maps  $f$ , one can define the  $L^1(\mathcal{C})$ -valued Bartle-integral  $\int f d\langle X, Y \rangle$ .

**LEMMA 6.3.** *Let  $t \geq 0$ , and suppose that  $f: [0, t] \rightarrow L^\infty(\mathcal{C})$  is the limit  $\lambda$  a.e. on  $[0, t]$  of a uniformly bounded sequence  $(g_n)$  in  $\mathcal{S}([0, t], L^\infty(\mathcal{C}))$ . Then, for any  $X, Y \in \mathfrak{M}$ ,  $f$  is integrable with respect to  $\langle X, Y \rangle$  and*

$$L^1(\mathcal{C})\text{-lim}_{\mathcal{E}} \int_{\mathcal{E}} g_n d\langle X, Y \rangle = \int_{\mathcal{E}} f d\langle X, Y \rangle$$

*exists, where  $\int_{\mathcal{E}} f d\langle X, Y \rangle$  is the Bartle integral.*

*Proof.* This follows immediately from Bartle's bounded convergence theorem [5]. Q.E.D.

**DEFINITION 6.4.** Let  $\mathcal{P}[0, t]$  denote the set of maps  $f: [0, t] \rightarrow L^\infty(\mathcal{C})$  satisfying the hypothesis of Lemma 6.3.

**LEMMA 6.5.** *For  $f \in \mathcal{P}[0, t]$  and  $X \in \mathfrak{M}$ , we have*

- (i)  $\beta(f) \in \mathcal{P}[0, t]$ ;
- (ii)  $\beta(f)\tilde{X} \in \mathfrak{H}([0, t], ds; L^2(\mathcal{C}))$ ;
- (iii)  $\int_0^t \tilde{X}(s) d\Psi_s = \left( \int_0^t \beta(\tilde{X}(s))^* d\Psi_s \right)^*$ .

*Proof.* (i) If  $(g_n)$  is a uniformly bounded sequence in  $\mathcal{S}([0, t], L^\infty(\mathcal{C}))$  such that  $\|g_n(s) - f(s)\|_\infty \rightarrow 0$   $\lambda$  a.e. on  $[0, t]$ , then, using Proposition 1.3, it follows that  $(\beta(g_n))$  is also a uniformly bounded sequence in  $\mathcal{S}([0, t], L^\infty(\mathcal{C}))$  and  $\beta(g_n) \rightarrow \beta(f)$  in  $L^\infty(\mathcal{C})$   $\lambda$  a.e. on  $[0, t]$ . In other words,  $\beta(f) \in \mathcal{P}[0, t]$ .

(ii) There is a sequence  $(h_n)$  in  $\mathcal{S}([0, t], L^\infty(\mathcal{C}))$  such that  $h_n \rightarrow \tilde{X}$  in  $\mathfrak{H}([0, t], ds; L^2(\mathcal{C}))$ . Then with the notation of part (i) above, and passing to a subsequence if necessary, we see that  $\beta(g_n)h_n \rightarrow \beta(f)\tilde{X}$  in  $L^2(\mathcal{C})$   $\lambda$  a.e. on  $[0, t]$ . Hence  $\beta(f)\tilde{X}$  is a  $\lambda$  measurable  $L^2$ -process on  $[0, t]$ .

It is easy to see that  $\beta(g_n)h_n$  is a Cauchy sequence in  $\mathfrak{H}([0, t], ds)$  and so the result follows.

(iii) Using the canonical anticommutation relations, equation (1.1), and Proposition 1.3, one sees [1, Lemma 3.15] that for any  $0 \leq r \leq s$  and  $g \in L^\infty(\mathcal{C})$ ,

$$(g\Psi(\chi_{(r, s]}))^* = \Psi(\chi_{(r, s]})g^* = \beta(g)^*\Psi(\chi_{(r, s]}).$$

That is,  $(g(\Psi_s - \Psi_r))^* = \beta(g)^*(\Psi_s - \Psi_r)$ .

From the definition of the stochastic integral and the continuity of the adjoint  $*: L^2(\mathcal{C}) \rightarrow L^2(\mathcal{C})$ , it follows that  $\left( \int_0^t h(s) d\Psi_s \right)^* = \int_0^t \beta(h(s))^* d\Psi_s$  for any  $t \geq 0$  and  $h \in \mathfrak{H}([0, t], ds)$ . Q.E.D.

**THEOREM 6.6.** *Let  $(X_t), (Y_t)$  be  $L^2(\mathcal{C})$ -martingales. Then for any  $t \geq 0$  and  $f \in \mathcal{K}([0, t], \mu_X) \cap \mathcal{K}([0, t], \mu_Y)$  we have*

$$\left\langle X_t, \int_0^t f(s) dY_s \right\rangle = \left\langle \int_0^t f(s)^* dX_s, Y_t \right\rangle.$$

*Proof.* For  $f \in \mathcal{K}([0, t], \mu_X) \cap \mathcal{K}([0, t], \mu_Y)$ , we have  $\int_0^t f dY = \int_0^t f \tilde{Y} d\Psi$  and

$$\int_0^t f^* dX = \int_0^t f^* \tilde{X} d\Psi, \text{ by Theorem 4.2. Hence, by definition}$$

$$\begin{aligned} \left\langle X_t, \int_0^t f dY \right\rangle &= \int_0^t \beta(f(s) \tilde{Y}(s))^* \beta(\tilde{X}(s)) ds = \\ &= \int_0^t \beta(\tilde{Y}(s))^* \beta(f(s)^* \tilde{X}(s)) ds = \left\langle \int_0^t f^* dX, Y_t \right\rangle. \end{aligned} \quad \text{Q.E.D.}$$

**THEOREM 6.7.** *Let  $(X_t), (Y_t)$  be  $L^2(\mathcal{C})$ -martingales and let  $f \in \mathcal{P}[0, t]$ . Then*

$$(i) \int_0^t f d\langle X, Y \rangle = \left\langle X_t, \left( \int_0^t f(s) dY_s^* \right)^* \right\rangle,$$

and

$$(ii) \int_0^t f d\langle X, Y \rangle^* = \left\langle \left( \int_0^t f(s) dX_s^* \right)^*, Y_t \right\rangle^*.$$

*Proof.* Suppose that  $f = g\chi_{[r, t]}$  is elementary, where  $0 \leq r \leq \tau \leq t$  and  $g \in L^\infty(\mathcal{C}_r)$ . Then

$$\begin{aligned} \int_0^t f d\langle X, Y \rangle &= g \int_r^t \beta(\tilde{Y}(s))^* \beta(\tilde{X}(s)) ds = \\ &= \int_0^t g\chi_{[r, t]} \beta(\tilde{Y}(s))^* \beta(\tilde{X}(s)) ds = \int_0^t f(s) \beta(\tilde{Y}(s))^* \beta(\tilde{X}(s)) ds. \end{aligned}$$

By linearity and continuity, this equality holds for any  $f \in \mathcal{P}[0, t]$ .

Now, by Lemma 6.5 (iii),

$$Y_t^* = Y_0^* + \left( \int_0^t \tilde{Y}(s) d\Psi_s \right)^* = Y_0^* + \int_0^t \beta(\tilde{Y}(s))^* d\Psi_s.$$

But, for  $h \in L^2(\mathcal{C})$ ,  $\|\beta(h)^*\|_2 = \|h\|_2$  and so  $\mu_Y = \mu_{Y^*}$ . Moreover,  $\mathcal{P}[0, t] \subseteq \mathcal{K}([0, t], \mu_Y)$  and thus  $\int_0^t f dY^* = \int_0^t f(s) \beta(\tilde{Y}(s))^* d\Psi_s$ . Thus, again by Lemma 6.5 (iii), we have

$$\left( \int_0^t f dY^* \right)^* = \int_0^t \beta \{ \beta(\tilde{Y}(s)) f(s)^* \} d\Psi_s = \int_0^t \tilde{Y}(s) \beta(f(s))^* d\Psi_s.$$

Hence

$$\left\langle X_t, \left( \int_0^t f dY^* \right)^* \right\rangle = \int_0^t \beta(\tilde{Y}(s) \beta(f(s))^*)^* \beta(\tilde{X}(s)) ds = \int_0^t f(s) \beta(\tilde{Y}(s))^* \beta(\tilde{X}(s)) ds.$$

This proves (i).

The proof of (ii) follows from (i) using  $\langle X_t, Y_t \rangle^* = \langle Y_t, X_t \rangle$ . Q.E.D.

This theorem has as a corollary a characterization of the stochastic integral in terms of pointed-bracket processes.

**THEOREM 6.8.** *Let  $(X_t)$  be an  $L^2(\mathcal{C})$ -martingale, and let  $f: \mathbb{R}^+ \rightarrow L^\infty(\mathcal{C})$  be such that the restriction of  $f$  to  $[0, t]$  belongs to  $\mathcal{P}[0, t]$  for all  $t \geq 0$ . Then  $\left( \int_0^t f dX \right)^*$  is the unique centred  $L^2(\mathcal{C})$ -martingale,  $Z_t$ , say, such that*

$$\int_0^t f d\langle Y, X^* \rangle = \langle Y_t, Z_t \rangle$$

for any  $L^2(\mathcal{C})$ -martingale  $(Y_t)$ .

*Proof.* By Theorem 6.7,  $(Z_t)$  satisfies

$$\int_0^t f d\langle Y, X^* \rangle = \langle Y_t, Z_t \rangle.$$

If  $(Z'_t)$  is a centred  $L^2$ -martingale also satisfying this equation, then  $\langle Y_t, (Z_t - Z'_t) \rangle = 0$  for all  $t \geq 0$  and for all  $Y \in \mathfrak{M}$ . Hence

$$\langle Z_t - Z'_t, Z_t - Z'_t \rangle = 0 \quad \text{for all } t \geq 0,$$

which implies that  $\int_0^t |\beta(\tilde{Z}(s) - \tilde{Z}'(s))|^2 ds = 0$  for  $t \geq 0$ . Thus  $\tilde{Z} = \tilde{Z}'$  in  $\mathfrak{H}_{\text{loc}}(\mathbb{R}^+, ds)$ .

Since  $(Z_t)$  and  $(Z'_t)$  are both centred, it follows that they are equal. Q.E.D.

## 7. THE LEFT STOCHASTIC INTEGRAL

Results analogous to those of the preceding sections can be obtained for the left stochastic integral  $\int dX f$  which is defined similarly to the right stochastic integral. Indeed, if  $(X_t)$  is an  $L^2$ -martingale, then, in terms of the left stochastic integral, Theorem 2.3 becomes

$$X_t = X_0 + \int_0^t \tilde{X}(s) d\Psi_s = X_0 + \int_0^t d\Psi_s \beta(\tilde{X}(s)), \quad t \geq 0.$$

The analogues of Theorem 3.4 and Corollary 3.6 yield the contraction property:

$$\left\| \int_0^t dX_s f(s) \right\|_2^2 \leq \int_0^t \|f(s)\|_\infty^2 d\mu_X(s)$$

for  $f \in \mathcal{K}([0, t], \mu_X)$ .

The analogue of Theorem 4.2 is:

**THEOREM 7.1.** *Let  $(X_t)$  be an  $L^2$ -martingale and let  $f \in \mathcal{K}([0, t], \mu_X)$ . Then  $\beta(\tilde{X})f \in \mathfrak{H}([0, t], ds)$  and*

$$\int_0^t dX_s f(s) = \int_0^t d\Psi_s \beta(\tilde{X}(s))f(s).$$

For left integrals, Theorem 6.6 becomes as follows.

**THEOREM 7.2.** *Let  $(X_t)$ ,  $(Y_t)$  be  $L^2(\mathcal{C})$ -martingales. Then for  $t \geq 0$  and  $f \in \mathcal{K}([0, t], \mu_X) \cap \mathcal{K}([0, t], \mu_Y)$  we have*

$$\left\langle \left( \int_0^t dY_s f(s) \right)^*, X_t \right\rangle = \left\langle Y_t^*, \left( \int_0^t dX_s^* f(s)^* \right)^* \right\rangle.$$

**THEOREM 7.3.** *Let  $(X_t)$ ,  $(Y_t)$  be  $L^2(\mathcal{C})$ -martingales and let  $f \in \mathcal{P}[0, t]$ . Then*

$$\int_0^t d\langle X, Y \rangle f = \left\langle \int_0^t dX_s f(s), Y_t \right\rangle.$$

*Proof.* As for that of Theorem 6.7.

**THEOREM 7.4.** *Let  $(X_t)$  be an  $L^2(\mathcal{C})$ -martingale, and suppose that  $f \in \mathcal{P}[0, t]$  for all  $t \geq 0$ . Then  $\left( \int_0^t dXf \right)$  is the unique centred  $L^2(\mathcal{C})$ -martingale,  $N_t$ , say, such that*

$$\int_0^t d\langle X, Y \rangle f = \langle N_t, Y_t \rangle.$$

*Proof.* Just as for that of Theorem 6.8.

#### REFERENCES

1. BARNETT, C.; STREATER, R. F.; WILDE, I. F., The Itô-Clifford integral, *J. Functional Analysis*, **48**(1982), 172–212.
2. BARNETT, C.; STREATER, R. F.; WILDE, I. F., The Itô-Clifford integral. II: Stochastic differential equations, *J. London Math. Soc.*, **27**(1983), 373–384.
3. BARNETT, C.; STREATER, R. F.; WILDE, I. F., The Itô-Clifford integral. III: Markov property of solutions to stochastic differential equations, *Commun. Math. Phys.*, **89**(1983), 13–17.
4. BARNETT, C.; STREATER, R. F.; WILDE, I. F., Stochastic integrals in an arbitrary probability space, *Math. Proc. Cambridge Phil. Soc.*, **94**(1983), 541–551.
5. BARTLE, R. G., A general bilinear vector integral, *Studia Math.*, **15**(1956), 337–352.
6. GROSS, L., Existence and uniqueness of physical ground states, *J. Functional Analysis*, **10**(1972), 52–109.
7. KUNZE, R. A.,  $L_p$  Fourier transforms on locally compact unimodular groups, *Trans. Amer. Math. Soc.*, **89**(1958), 519–540.
8. MEYER, P. A., *Probability and potentials*, Blaisdell, Waltham, Mass., 1966.
9. RAO, K. M., On decomposition theorems of Meyer, *Math. Scand.*, **24**(1969), 66–78.
10. SEGAL, I. E., A non-commutative extension of abstract integration, *Ann. of Math.*, **57**(1953), 401–457; Correction to “A non-commutative extension of abstract integration”, *Ann. of Math.*, **58**(1953), 595–596.
11. SEGAL, I. E., Tensor algebras over Hilbert spaces, *Ann. of Math.*, **63**(1956), 160–175.

*C. BARNETT, R. F. STREATER and I. F. WILDE*

*Department of Mathematics,*

*Bedford College,*

*Regent's Park,*

*London, N.W.1,*

*England.*

Received December 9, 1982.