

The Measurability of a Stochastic Process of Second Order and its Linear Space*

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1 The Measurability of a Stochastic Process of Second Order

Let T be a separable metric space and $\mathcal{B}(T)$ the σ -algebra of Borel sets of T , and let X_t , $t \in T$, be a real stochastic process on the probability space (Ω, \mathcal{F}, P) . X_t , $t \in T$, is called measurable if the map $(t, \omega) \rightarrow X_t(\omega)$ is $\mathcal{B}(T) \times \mathcal{F}$ -measurable. A process Y_t , $t \in T$, on (Ω, \mathcal{F}, P) is called a modification of X_t , $t \in T$, if $P\{X_t = Y_t\} = 1$ for all t in T . X_t , $t \in T$, is of second order if $E(X_t^2) < +\infty$ for all t in T , and then its autocorrelation R is defined by

$$R(t, s) = E(X_t X_s) \quad \text{for all } t, s \in T. \quad (1)$$

It is clear from Fubini's theorem that if a second order process X_t , $t \in T$, has a measurable modification then R is $\mathcal{B}(T) \times \mathcal{B}(T)$ -measurable. That the measurability of R is not sufficient for the existence of a measurable modification of X_t , $t \in T$, is demonstrated in Remark 2. It is thus of interest to find a condition which along with the measurability of R would imply the existence of a measurable modification of X_t , $t \in T$. This question is answered in Theorem 1, where in fact necessary and sufficient conditions are given for a second order process to have a measurable modification. A remarkable consequence of these conditions is that the existence of a measurable modification of a second order process is a second order property.

The proof of Theorem 1 is based on the necessary and sufficient conditions for a process (not necessarily of second order) to have a measurable modification given in [5], which are expressed as follows (here the terminology of [6] is followed). Let M be the space of all real random variables on (Ω, \mathcal{F}, P) with the topology of convergence in probability, where random variables that are equal a.e. $[P]$ are considered identical. If ξ is a real random variable, its class in M is denoted by $[\xi]$. Then X_t , $t \in T$, has a measurable modification if and only if the

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map from T to M taking t to $[X_t]$ is measurable and has separable range [5, 6]. Moreover, the measurable modification can be taken to be separable and also progressively measurable, the latter if T is an interval and a nondecreasing family \mathcal{F}_t , $t \in T$, of sub- σ -algebras of \mathcal{F} is given.

For a second order process X_t , $t \in T$, we denote by $H(X)$ the closure in $L_2(\Omega, \mathcal{F}, P)$ of the linear space of the random variables $\{X_t, t \in T\}$ and we call it the linear space of the process. We also denote by $R(K)$ the reproducing kernel Hilbert space of a real, symmetric, nonnegative definite function K on $T \times T$. It is well known that $R(R)$ consists of all functions f on T of the form $f(t) = E(\xi X_t)$, $t \in T$, for some $\xi \in H(X)$, and that the map $\xi \rightarrow E(\xi X_t)$ defines an inner product preserving isomorphism between $H(X)$ and $R(R)$ [16].

Theorem 1. *Let X_t , $t \in T$, be a real, second order process with autocorrelation R . The following are equivalent:*

- (i) X_t , $t \in T$, has a measurable modification.
- (ii) R is $\mathcal{B}(T) \times \mathcal{B}(T)$ -measurable and $H(X)$ (or $R(R)$) is separable.

Proof. (a) We first show that ((ii)) implies ((i)). It suffices to verify the conditions of [5, 6]; the construction of a measurable modification is the same as in [5] or in [6].

Since convergence in $L_2(\Omega, \mathcal{F}, P)$ implies convergence in probability, the separability of $H(X)$ as a subset of $L_2(\Omega, \mathcal{F}, P)$ implies its separability as a subset of M . Thus its subset $\{[X_t], t \in T\}$ is separable in M . To complete the proof it suffices to show that the map $X : T \rightarrow M$ defined by $X(t) = [X_t]$ is measurable. The metric ρ on M defined by $\rho(\xi, \eta) = E \left[\frac{|\xi - \eta|}{1 + |\xi - \eta|} \right]$, $\xi, \eta \in M$ metrizes the topology of convergence in probability. Thus for the measurability of X it suffices to show that $X^{-1}(B) \in \mathcal{B}(T)$ for every set B in M of the form $B = \{Y \in M : \rho(Y, Y_0) \leq r\}$, where $Y_0 \in M$ and $r > 0$. Since $X^{-1}(B) = \{t \in T : \rho([X_t], Y_0) < r\}$, it suffices to prove that the real function $\rho([X_t], Y_0)$ on T is $\mathcal{B}(T)$ -measurable for all $Y_0 \in M$.

Let $\{\xi_n\}_{n=1}^\infty$ be a complete orthonormal sequence in $H(X)$ (which exists because $H(X)$ is separable). Then for all $t \in T$ we have

$$X_t = \sum_{n=1}^{\infty} a_n(t) \xi_n \quad \text{in } L_2(\Omega, \mathcal{F}, P), \quad (2)$$

where

$$a_n(t) = E(\xi_n X_t). \quad (3)$$

Thus $a_n \in R(R)$, and in fact $\{a_n\}_{n=1}^\infty$ is a complete orthonormal sequence in $R(R)$. If for every $t \in T$ we let

$$X_t^{(N)} = \sum_{n=1}^N a_n(t) \xi_n, \quad (4)$$

then $X_t^{(N)}$ converges to X_t in $L_2(\Omega, \mathcal{F}, P)$ and thus in probability. Let $Y_0 \in M$ be fixed and $Y_t = X_t - Y_0$ and $Y_t^{(N)} = X_t^{(N)} - Y_0$ for all $t \in T$. Then $Y_t^{(N)}$ converges to Y_t in probability,

i.e., $\rho([Y_t^{(N)}], [Y_t]) \rightarrow 0$ as $N \rightarrow \infty$. Dropping the index t for simplicity we have

$$\rho([X^{(N)}], [X]) \leq \frac{E \left[\frac{|X^{(N)} - X|}{1 + |X^{(N)} - X|} \right]}{E \left[\frac{|X|}{1 + |X|} \right]}. \quad (5)$$

Thus

$$\rho([X_t], Y_0) = \lim_{N \rightarrow \infty} E \left[\frac{|X_t^{(N)} - Y_0|}{1 + |X_t^{(N)} - Y_0|} \right] \quad \text{for all } t \in T. \quad (6)$$

Note that every function in $R(R)$ is either a finite linear combination of the functions $\{R(t, \cdot), t \in T\}$ or a pointwise limit on T of such functions. Hence, since R is $\mathcal{B}(T) \times \mathcal{B}(T)$ -measurable, $R(t, \cdot)$ is $\mathcal{B}(T)$ -measurable for all $t \in T$, and it follows that every f in $R(R)$ is $\mathcal{B}(T)$ -measurable. Consequently $Y_t^{(N)} = \sum_{n=1}^N a_n(t) \xi_n - Y_0$ is $\mathcal{B}(T) \times \mathcal{F}$ -measurable. By Fubini's theorem $E \left[\frac{|X_t^{(N)} - Y_0|}{1 + |X_t^{(N)} - Y_0|} \right]$ is $\mathcal{B}(T)$ -measurable, and thus so is $\rho([X_t], Y_0)$, which completes the proof.

(b) We now show that ((i)) implies ((ii)). The measurability of R follows from Fubini's theorem and ((i)). In order to prove the separability of $H(X)$ we first assume that R is uniformly bounded on T :

$$R(t, t) \leq C < +\infty \quad \text{for all } t \in T. \quad (7)$$

We will show that this implies the uniform integrability of the family of random variables $\{X_t, t \in T\}$. Indeed we have for all $a > 0$,

$$\int_{\{|X_t| > a\}} |X_t| dP \leq [P\{|X_t| > a\} \cdot E(X_t^2)]^{1/2} \leq \left[\frac{E(X_t^2)}{a^2} \cdot E(X_t^2) \right]^{1/2} = \frac{R(t, t)}{a}.$$

Thus

$$\lim_{a \rightarrow \infty} \sup_{t \in T} \int_{\{|X_t| > a\}} |X_t| dP = 0, \quad (8)$$

and $\{X_t, t \in T\}$ is uniformly integrable.

Now ((i)) implies that $\{[X_t], t \in T\}$ is separable in M . Thus there exists a countable subset M' of $\{[X_t], t \in T\}$ such that for every $t \in T$, $[X_t]$ is the limit in probability of a sequence in M' , and hence also in $L_2(\Omega, \mathcal{F}, P)$, since M' is uniformly integrable [14]. It follows that $H(X)$ equals the $L_2(\Omega, \mathcal{F}, P)$ closure of the linear span of M' and, since M' is countable, $H(X)$ is separable.

We now consider the general case and define for $N = 1, 2, \dots$

$$T_N = \{t \in T : R(t, t) \leq N\}. \quad (9)$$

Since R is measurable, $T_N \in \mathcal{B}(T)$ and by ((i)) $\{X_t, t \in T_N\}$ has a measurable modification. It follows by what has been proven that the $L_2(\Omega, \mathcal{F}, P)$ closure of the linear span of the random variables $\{X_t, t \in T_N\}$, $H_N(X)$, is separable. Since X_t is of second order, R is finite valued and thus $\bigcup_{N=1}^{\infty} T_N = T$. It follows that $H(X)$ is the $L_2(\Omega, \mathcal{F}, P)$ closure of $\bigcup_{N=1}^{\infty} H_N(X)$ and thus $H(X)$ is separable.

Thus a $\mathcal{B}(T) \times \mathcal{B}(T)$ -measurable, symmetric, nonnegative definite, real function R on $T \times T$ is the autocorrelation of a measurable process if and only if $R(R)$ is separable. \square

Remark 1. The mean m and the covariance C of a real second order process X_t , $t \in T$, are defined by $m(t) = E(X_t)$ and $C(t, s) = E([X_t - m(t)][X_s - m(s)])$ for all $t, s \in T$. Then $R(t, s) = m(t)m(s) + C(t, s)$. In connection with ((ii)) of Theorem 1 it should be noted that R is $\mathcal{B}(T) \times \mathcal{B}(T)$ -measurable if and only if m is $\mathcal{B}(T)$ -measurable and C is $\mathcal{B}(T) \times \mathcal{B}(T)$ -measurable.

The “if” part is obvious. The “only if” part is shown as follows. We have $m(t) = E(X_t I_\Omega)$ for all $t \in T$, where I_Ω is the indicator function. Denote by ξ the projection of $I_\Omega \in L_2(\Omega, \mathcal{F}, P)$ onto the subspace $H(X)$. Then $m(t) = E(X_t \xi)$ for all $t \in T$ and $\xi \in H(X)$, and thus $m \in R(R)$. Since R is $\mathcal{B}(T) \times \mathcal{B}(T)$ -measurable, m is $\mathcal{B}(T)$ -measurable (see part (a) of the proof of Theorem 1) and $C(t, s) = R(t, s) - m(t)m(s)$ is $\mathcal{B}(T) \times \mathcal{B}(T)$ -measurable.

Remark 2. Let $T = [0, 1]$ and $R(t, s) = 1$ for $t = s$ in T and $R(t, s) = 0$ for $t \neq s$ in T . Since R is symmetric and nonnegative definite, there exists a probability space (Ω, \mathcal{F}, P) and a real process X_t , $t \in T$, on it with autocorrelation R . R is clearly $\mathcal{B}(T) \times \mathcal{B}(T)$ -measurable, but since the values of X_t are orthogonal in $L_2(\Omega, \mathcal{F}, P)$, $E(X_t X_s) = 0$ for $t \neq s$ in T , $H(X)$ is not separable and by Theorem 1, X_t , $t \in T$, does not have a measurable modification. This can be also shown without using Theorem 1. Indeed, assume that X_t , $t \in T$, has a measurable modification Y_t , $t \in T$. Then

$$\int_T R(t, t) dt = 1 < \infty \quad (10)$$

implies that $\int_T Y_t^2 dt < +\infty$ a.e. $[P]$. If $\{\Phi_n\}_{n=1}^\infty$ is a complete orthonormal set in $L_2(T, \mathcal{B}(T), \text{Leb})$, then

$$Y_t = \sum_{n=1}^\infty \Phi_n(t) \int_T Y_s \Phi_n(s) ds \quad \text{a.e. } [P], \quad (11)$$

where the convergence is in $L_2(T)$ a.e. $[P]$, and thus

$$E \left[\int_T Y_t^2 dt \right] = \sum_{n=1}^\infty \int_T \int_T R(t, s) \Phi_n(t) \Phi_n(s) dt ds = 0 \quad \text{a.e. } [P],$$

which contradicts $E \left[\int_T Y_t^2 dt \right] = 1$. It follows that X_t , $t \in T$, does not have a measurable modification.

Remark 3. For Gaussian processes it can be easily shown that ((ii)) implies ((i)) without relying on the results of [5]; this is done in [15].

Corollary 1. Let R be a symmetric, nonnegative definite, real function on $T \times T$. If $R(R)$ is separable the following are equivalent:

- (i) $R(t, \cdot)$ is $\mathcal{B}(T)$ -measurable for all $t \in T$.
- (ii) R is $\mathcal{B}(T) \times \mathcal{B}(T)$ -measurable.

Proof. It suffices to show that ((i)) implies ((ii)). Since R is symmetric, nonnegative definite and real, there exists a probability space (Ω, \mathcal{F}, P) and a real process X_t , $t \in T$, on it with autocorrelation R . It is clear from part (a) of the proof of Theorem 1 that the separability

of $R(R)$ and ((i)) imply the existence of a measurable modification of X_t , $t \in T$, and thus ((ii)). This result can be shown in a simpler way without using an associated process. Indeed, if $\{f_n\}_{n=1}^\infty$ is a complete orthonormal set in $R(R)$, then it is easily seen that $R(t, s) = \sum_{n=1}^\infty f_n(t)f_n(s)$ for all $t, s \in T$. Now ((i)) implies as in part (a) of the proof of Theorem 1 that every f_n is $\mathcal{B}(T)$ -measurable and thus ((ii)) holds. \square

Corollary 2. *A second order process X_t , $t \in T$, which satisfies any of the following conditions has a measurable modification (in ((iii)) also progressively measurable):*

- (i) X_t , $t \in T$, is weakly continuous on T .
- (ii) T is an arbitrary interval and X_t , $t \in T$, has orthogonal increments.
- (iii) T is an arbitrary interval and X_t , $t \in T$, is a martingale.

Proof. ((i)): Since T is separable and X_t weakly continuous on T , $H(X)$ is separable [16]. By the weak continuity of X_t , $R(t, \cdot)$ is continuous, hence $\mathcal{B}(T)$ -measurable, for all $t \in T$. The conclusion follows from Corollary 1 and Theorem 1.

((ii)): It is known that $H(X)$ is separable [8]. Also, X_t has left and right $L_2(\Omega, \mathcal{F}, P)$ limits on T and that except on a countable subset of T , $X_{t-} = X_t = X_{t+}$. This implies the measurability of R and the result follows from Theorem 1.

((iii)): Define the function F by $F(t) = E(X_t^2)$ for all $t \in T$. By the martingale property, with respect to the nondecreasing family \mathcal{F}_t , $t \in T$, of sub- σ -algebras of \mathcal{F} , we have for all $s < t$ in T ,

$$E(X_t X_s) = E[E(X_t X_s | \mathcal{F}_s)] = E[X_s E(X_t | \mathcal{F}_s)] = E(X_s^2)$$

and thus

$$E(\{X_t - X_s\}^2) = F(t) - F(s).$$

It follows from this relationship, as in [8] and in ((ii)), that $H(X)$ is separable and R is $\mathcal{B}(T) \times \mathcal{B}(T)$ -measurable. \square

Remark 4. *Let X_t , $t \in T$, T an arbitrary interval, be a real separable process of second order with autocorrelation R . If X_t is mean square differentiable on T and $\frac{\partial R(t,s)}{\partial t}$ and $\frac{\partial R(t,s)}{\partial s}$ are locally Lebesgue integrable in T and in t, s respectively, then with probability one the paths of X_t , $t \in T$, are absolutely continuous on every compact subinterval of T . This is shown in [10] with the additional assumption that the mean square derivative \dot{X}_t of X_t has a measurable modification, which is always satisfied because of Theorem 1. Indeed, since X_t is mean square differentiable on T , it is mean square continuous on T . Thus $H(X)$ is separable and the continuity of R implies the measurability of R . Since \dot{X}_t is the autocorrelation of X_t and since $H(\dot{X}) \subset H(X)$, the conclusion follows from Theorem 1.*

Corollary 3. *If a real process X_t , $t \in T$, has a measurable modification, then $\mathcal{F}(X)$ coincides mod 0 with a separable σ -algebra.*

Proof. Since X_t , $t \in T$, has a measurable modification, $\{[X_t], t \in T\}$ is a separable subset of M . Thus there exists a countable subset $M' = \{[X_t], t \in S\}$ of $\{[X_t], t \in T\}$ (S is a countable subset of T) such that for every $t \in T$, $[X_t]$ is the limit in probability of a sequence from M' , and thus X_t is the a.e. $[P]$ limit of a sequence from $\{X_t, t \in S\}$. If \mathcal{F}' is the sub- σ -algebra of \mathcal{F} generated by the random variables $\{X_t, t \in S\}$, then $\mathcal{F}' \subset \mathcal{F}(X)$, \mathcal{F}' is separable and $\mathcal{F}(X)$ coincides with \mathcal{F}' mod 0. \square

2 On the Separability of the Linear Space of a Second Order Process

The linear space $H(X)$ of a second order process $X_t, t \in T$, plays an important role in the structure of such processes and in a variety of problems in statistical inference. If $H(X)$ is separable then X_t admits series representations and also integral representations (Theorem 2) that can be effectively used in problems such as linear mean square estimation. Also the separability of $H(X)$ is the only condition needed for a second order process to have the Hida-Cramér representation [11]. It is thus of interest that a measurable second order process has a separable linear space. $H(X)$ is known to be separable when the process $X_t, t \in T$, is weakly continuous [16], has orthogonal increments [8], or is a martingale (Corollary 2((iii))). In Theorem 2 necessary and sufficient conditions are given for $H(X)$ to be separable in terms of integral representations of X_t .

Before stating the theorem we mention a few basic facts about random measures, that can be found for instance in [7, 8]. Let (V, \mathcal{V}) be a measurable space. A random measure Z on (V, \mathcal{V}) is a countably additive map from \mathcal{V} to $L_2(\Omega, \mathcal{F}, P)$; i.e., whenever A is the disjoint union of the sets $A_k \in \mathcal{V}$, $Z(A) = \sum_{k=1}^{\infty} Z(A_k)$ in $L_2(\Omega, \mathcal{F}, P)$. (Here we consider the case where Z is defined on the entire σ -algebra \mathcal{V}). To each random measure Z on V there corresponds a finite signed measure μ on $V \times V$ by $\mu(A \times B) = E[Z(A)Z(B)]$, $A, B \in \mathcal{V}$. μ is symmetric and nonnegative definite on the measurable rectangles of $V \times V$. A random measure Z is called orthogonal if $\mu(A \times B) = 0$ whenever A and B are disjoint, and to each orthogonal random measure there corresponds a finite nonnegative measure ν on V by $\nu(A) = E[Z^2(A)]$, $A \in \mathcal{V}$. Let $H(Z)$ be the closure in $L_2(\Omega, \mathcal{F}, P)$ of the linear span of $\{Z(A), A \in \mathcal{V}\}$, and let $\Lambda_2(\mu)$ be the Hilbert space of real, \mathcal{V} -measurable functions on V with inner product $\langle f, g \rangle_{\Lambda_2(\mu)} = \int_V \int_V f(u)g(v)d\mu(u, v)$ (of course $\Lambda_2(\mu)$ consists of equivalence classes of functions, two functions f and g considered identical if $\langle f-g, f-g \rangle_{\Lambda_2(\mu)} = 0$). There is an inner product preserving isomorphism between $\Lambda_2(\mu)$ and $H(Z)$, denoted by \leftrightarrow , such that $I_A \leftrightarrow Z(A)$, $A \in \mathcal{V}$, and integration of functions $f \in \Lambda_2(\mu)$ with respect to Z is defined by $\int_V f(u)dZ(u) \leftrightarrow f$. If Z is orthogonal, there is an inner product preserving isomorphism between $L_2(\nu) = L_2(V, \mathcal{V}, \nu)$ and $H(Z)$, denoted again by \leftrightarrow , such that $I_A \leftrightarrow Z(A)$, $A \in \mathcal{V}$, and integration of functions in $L_2(\nu)$ with respect to Z is defined by $\int_V f(u)dZ(u) \leftrightarrow f$.

Theorem 2. *Let $X_t, t \in T$, be a second order process.*

- (i) *If $H(X)$ is separable then for every finite measure space (V, \mathcal{V}, ν) such that $L_2(\nu) = L_2(V, \mathcal{V}, \nu)$ is separable and infinite dimensional, X_t has a representation*

$$X_t = \int_V f(t, u)dZ(u) \quad \text{for all } t \in T, \quad (12)$$

where Z is an orthogonal measure on V with corresponding measure ν and $f(t, \cdot) \in L_2(\nu)$ for all $t \in T$. Conversely, if X_t has such a representation, $H(X)$ is separable.

- (ii) *If $H(X)$ is separable, then for every measurable space (V, \mathcal{V}) and every finite signed measure μ on $V \times V$ which is symmetric and nonnegative definite on the measurable*

rectangles of $V \times V$, and such that $\Lambda_2(\mu)$ is separable and infinite dimensional, X_t has a representation

$$X_t = \int_V f(t, u) dZ(u) \quad \text{for all } t \in T, \quad (13)$$

where Z is a random measure on V with corresponding measure μ and $f(t, \cdot) \in \Lambda_2(\mu)$ for all $t \in T$. Conversely, if X_t has such a representation, $H(X)$ is separable.

Proof. ((ii)) being a particular case of ((i)), we will prove only ((ii)). We start with the second claim. If X_t has such a representation then $X_t \in H(Z)$ for all $t \in T$, hence $H(X) \subset H(Z)$ and the conclusion follows from the isomorphism between $H(Z)$ and $\Lambda_2(\mu)$ and the separability of the latter.

We now prove the first claim. Assume that $H(X)$ is separable and let $\{\xi_n\}_{n=1}^\infty$ be a complete orthonormal set. Then for all $t \in T$,

$$X_t = \sum_{n=1}^\infty a_n(t) \xi_n \quad \text{in } L_2(\Omega, \mathcal{F}, P), \quad (14)$$

where $a_n(t) = E(X_t \xi_n)$. Let $\{f_n\}_{n=1}^\infty$ be a complete orthonormal set in $\Lambda_2(\mu)$. Since μ is finite, $I_A \in \Lambda_2(\mu)$ for all $A \in \mathcal{V}$. Then

$$I_A = \sum_{n=1}^\infty \langle I_A, f_n \rangle f_n \quad \text{in } \Lambda_2(\mu), \quad (15)$$

where $\langle I_A, f_n \rangle = \int_V \int_V f_n(v) d\mu(u, v)$. Throughout the proof we will write $\langle \cdot, \cdot \rangle$ for $\langle \cdot, \cdot \rangle_{\Lambda_2(\mu)}$. Thus for all n , λ_n is a finite signed measure on (V, \mathcal{V}) . We also have

$$\langle I_A, I_A \rangle = \sum_{n=1}^\infty \lambda_n^2(A) = \mu(A \times A) < +\infty. \quad (16)$$

Hence

$$Z(A) = \sum_{n=1}^\infty \lambda_n(A) \xi_n \quad (17)$$

defines a function from \mathcal{V} to $L_2(\Omega, \mathcal{F}, P)$ (the convergence being in $L_2(\Omega, \mathcal{F}, P)$). We will show that Z is a random measure with corresponding measure μ . The latter is clear since for all $A, B \in \mathcal{V}$ we have

$$E[Z(A)Z(B)] = \sum_{n=1}^\infty \lambda_n(A)\lambda_n(B) = \langle I_A, I_B \rangle = \mu(A \times B). \quad (18)$$

For the countable additivity of Z let $A = \bigcup_{k=1}^\infty A_k$ where $\{A_k\}_{k=1}^\infty$ is a disjoint sequence of sets in \mathcal{V} . Then

$$E \left[\left(Z(A) - \sum_{k=1}^K Z(A_k) \right)^2 \right] = \mu \left(\left(\bigcup_{k=K+1}^\infty A_k \right) \times \left(\bigcup_{k=K+1}^\infty A_k \right) \right) \rightarrow 0 \quad \text{as } K \rightarrow \infty. \quad (19)$$

Thus $Z(A) = \sum_{k=1}^{\infty} Z(A_k)$.

We now show that for every $g \in \Lambda_2(\mu)$,

$$\int_V g(u) dZ(u) = \sum_{n=1}^{\infty} \langle g, f_n \rangle \xi_n \quad \text{in } L_2(\Omega, \mathcal{F}, P). \quad (20)$$

This is true for indicator functions by definition of Z , and therefore also for simple functions. Since $H(Z)$ is defined as the $L_2(\Omega, \mathcal{F}, P)$ closure of the linear space of $\{Z(A), A \in \mathcal{V}\}$, it follows by the isomorphism between $\Lambda_2(\mu)$ and $H(Z)$ that the linear span of $\{I_A, A \in \mathcal{V}\}$ is dense in $\Lambda_2(\mu)$. Thus every $g \in \Lambda_2(\mu)$ is the limit of a sequence of simple functions. The result follows from

$$E \left[\left(\int_V g dZ - \sum_{n=1}^{\infty} \langle g, f_n \rangle \xi_n \right)^2 \right] = \langle g - g_k, g - g_k \rangle \rightarrow 0, \quad (21)$$

which implies that $H(Z) = H(X)$.

In particular we have

$$\int_V f(t, u) dZ(u) = \sum_{n=1}^{\infty} a_n(t) \xi_n. \quad (22)$$

Now since $f(t) \in \Lambda_2(\mu)$ for all $t \in T$ and $f(t) \leftrightarrow X_t$, we can define $f(t, \cdot) \in \Lambda_2(\mu)$ for all $t \in T$ by

$$f(t, u) = \sum_{n=1}^{\infty} a_n(t) f_n(u), \quad (23)$$

where the convergence is in $\Lambda_2(\mu)$. It follows from the property of the integral just proven that for all $t \in T$ we have the following equality in $L_2(\Omega, \mathcal{F}, P)$,

$$\int_V f(t, u) dZ(u) = \sum_{n=1}^{\infty} a_n(t) \xi_n = X_t, \quad (24)$$

which concludes the proof. \square

Remark 5. We assume throughout this remark that $H(X)$ is separable. Then it is clear that the first claim in ((i)) and ((ii)) is valid provided the dimensionality of $L_2(\nu)$ and $\Lambda_2(\mu)$ is no less than the dimensionality of the integers. Also, one can take $(V, \mathcal{V}) = (T, \mathcal{B}(T))$ or as V any interval and \mathcal{V} its Borel sets; in the latter case ν may be taken the Lebesgue measure or one absolutely continuous to it, and μ may be taken absolutely continuous to the Lebesgue measure on $V \times V$. If a series (respectively, integral) representation of X_t is known then one can obtain integral (respectively, series) representations of X_t as indicated in the proof of Theorem 2. These representations will be explicitly obtained if one can find complete orthonormal sets in the spaces $L_2(\nu)$ and $\Lambda_2(\mu)$. If V is an interval and \mathcal{V} its Borel sets, complete orthonormal sets in $L_2(\nu)$ are given in [2] (see also [13]), and complete sets in $\Lambda_2(\mu)$ are given in [3] (In [3] the case where V is the entire real line is treated and the case where V is an interval can be treated similarly). If neither an integral nor a series representation of X_t is available, the problem arises how to obtain explicitly such a representation (in terms of the process X_t , $t \in T$, and its autocorrelation R). This problem is solved in [4] for weakly continuous processes X_t , $t \in T$, and T an arbitrary interval.

Remark 6. Theorem 2 may also be stated in terms of integral representation of the auto-correlation R , which for (i) and (ii) are respectively

$$R(t, s) = \int_V f(t, u)f(s, u)d\nu(u) \quad \text{for all } t, s \in T, \quad (25)$$

and

$$R(t, s) = \int_V \int_V f(t, u)f(s, v)d\mu(u, v) \quad \text{for all } t, s \in T. \quad (26)$$

Remark 7. In [12] a second order process $X_t, t \in \mathbb{R} = (-\infty, +\infty)$, is called oscillatory if it has a representation

$$X_t = \int_{-\infty}^{\infty} e^{itu} a_t(u) dZ(u) \quad \text{for all } t \in \mathbb{R}, \quad (27)$$

where Z is an orthogonal random measure on $(\mathbb{R}, \mathcal{B}(\mathbb{R}))$ with corresponding measure ν and $a_t(\cdot) \in L_2(\nu)$ for all $t \in T$ (this is a generalization of a concept introduced by Priestley). If $X_t, t \in \mathbb{R}$, is oscillatory then $H(X)$ is separable, since $L_2(\mathbb{R}, \mathcal{B}(\mathbb{R}), \nu)$ is separable. Conversely, if $H(X)$ is separable it follows by Theorem 2(i) that for any finite measure ν on $(\mathbb{R}, \mathcal{B}(\mathbb{R}))$ we have $X_t = \int_{-\infty}^{\infty} f(t, u) dZ(u)$ for all $t \in T$, where Z is an orthogonal random measure on $(\mathbb{R}, \mathcal{B}(\mathbb{R}))$ with corresponding measure ν and $f(t, \cdot) \in L_2(\nu)$ for all $t \in T$. If we define $a_t(u) = e^{-itu} f(t, u)$, it becomes clear that $X_t, t \in \mathbb{R}$, is oscillatory. Thus a second order process is oscillatory if and only if its linear space is separable.

Remark 8. Some simple sufficient conditions for $H(X)$ to be separable are as follows. If $X_t, t \in T$, is a linear operation on a second order process $Y_s, s \in S$, with separable linear space, then $H(X) \subset H(Y)$ and the separability of $H(X)$ follows from that of $H(Y)$. Also, because of the isomorphism between $H(X)$ and $R(R)$, $H(X)$ is separable if there is a symmetric, nonnegative definite function K on $T \times T$ such that $R(R) \subset R(K)$ and $R(K)$ is separable. A sufficient condition for $R(R) \subset R(K)$ is that $K - R$ be nonnegative definite [1].

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