A NOTE ON UNIFORM CONVERGENCE OF
STOCHASTIC PROCESSES

JOHN B. WALSH

Let $T$ be a compact metrizable space and let \( \{ X_n(t), t \in T \} \)
\( n = 1, 2, \cdots \) be independent continuous (w.p.1.) stochastic processes
with mean zero. We apply an abstract martingale convergence
theorem to show that if \( \sum X_n(t) \) converges pointwise with probability
one to a continuous stochastic process with an integrable stochastic norm, it
also converges uniformly in $t$ with probability one. Among the im-
mediate consequences of this theorem is a generalization of Paley and
Wiener's result on the uniform convergence of the Fourier expansion
of Brownian motion.

1. Let $\mathfrak{K}$ be a Banach space with norm $\| \|$, and let \( (\Omega, \mathfrak{F}, \mathbb{P}) \) be a
probability space. If $X$ is a strong random variable with values in $\mathfrak{K}$
such that $E \{ \| X \| \} < \infty$, and if $\mathfrak{F}_1 \subset \mathfrak{F}$ is a Borel field, we will denote
the strong conditional expectation of $X$ given $\mathfrak{F}_1$ by $E \{ X | \mathfrak{F}_1 \}$. (For
definitions and properties of strong random variables and strong con-
ditional expectations, see [6].)

The following theorem is due to Chatterji [1, Theorems 1 and 4].

THEOREM A (CHATTERJI). Let $X$ be a strong random variable with
values in $\mathfrak{K}$ and let \( \{ X_n, \mathfrak{F}_n, n \geq 1 \} \) be a martingale, where $X_n = E \{ X | \mathfrak{F}_n \}$,
\( n = 1, 2, \cdots \). If for some $p$, $1 \leq p < \infty$, $E \{ \| X \|^p \} < \infty$, then there is a
strong random variable $Y$ such that

\[
\text{(A1) } \| X_n - Y \| \to 0 \quad \text{w.p.1.} \\
\text{(A2) } E \{ \| X_n - Y \|^p \} \to 0
\]

and $Y = E \{ X | \mathfrak{F}_\infty \}$ where $\mathfrak{F}_\infty$ is the Borel field generated by $\cup \mathfrak{F}_n$.

2. In what follows, $T$ is compact and metrizable and $C(T)$ is the
Banach space of continuous functions on $T$ with the sup-norm. If
\( \{ X(t), t \in T \} \) is a continuous stochastic process on \( (\Omega, \mathfrak{F}, \mathbb{P}) \), it deter-
mines a random variable $X$ with values in $C(T)$. It is a straightfor-
ward consequence of the separability of $T$ and $C(T)$ coupled with
the measurability of $X(t)$ for each $t$ that $X$ is a strong random
variable.

3. Let \( \{ X(t), t \in T \} \) be a continuous stochastic process satisfying
$E \{ \sup_{t \in T} |X(t)| \} < \infty$ and let \( \{ X_n(t), t \in T \} \)
\( n = 1, 2, \cdots \) be a sequence of independent continuous stochastic processes with mean

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zero. Let \( S_n(t) = \sum_{j} X_j(t) \) and let \( \mu \) be a finite regular Borel measure on \( T \) whose smallest closed support is \( T \) itself. Then we have

**Theorem B.** The following are equivalent:

(B1) \( S_n(t) \to X(t) \) uniformly in \( t \) w.p.1.

(B2) For some strictly monotone increasing function \( \Phi \) on \([0, \infty)\), continuous at zero, with \( \Phi(0) = 0 \):

\[
\int_T \Phi \left( \frac{|S_n(t) - X(t)|}{d\mu(t)} \right) \, d\mu(t) \to 0 \quad \text{w.p.1.}
\]

(B3) For each \( t \) in some countable dense subset \( T_0 \subset T \), \( S_n(t) \to X(t) \) w.p.1.

**Proof.** (B1) \( \Rightarrow \) (B2) is obvious for any such \( \Phi \). (B2) \( \Rightarrow \) (B3) follows from an application of Fubini's theorem and the fact that a sum of independent random variables which converges in measure also converges w.p.1. The only nontrivial part is (B3) \( \Rightarrow \) (B1).

For each \( n \), let \( \mathcal{F}_n \) be the Borel field generated by \( \{X_1(t), \ldots, X_n(t), t \in T\} \) and consider

\[
E\left\{ X(t) - S_n(t) \, | \, \mathcal{F}_n \right\} = 0.
\]

\( X(t) \) is absolutely integrable since \( \sup_{t \in T} |X(t)| \) is, and by (B3) \( X(t) \) is the a.e. limit of a sum of independent mean zero random variables, hence \( E\left\{ X(t) \right\} = 0 \) [3, Theorem 5.1, p. 373].

Further, \( X(t) - S_n(t) = \sum_{j=1}^{n} X_j(t) \), which is independent of \( \mathcal{F}_n \) so that

\[
E\left\{ X(t) - S_n(t) \, | \, \mathcal{F}_n \right\} = 0;
\]

or, since \( S_n(t) \) is measurable \( \mathcal{F}_n \),

\[
E\left\{ X(t) \, | \, \mathcal{F}_n \right\} = S_n(t).
\]

Let \( x'_t \) be the evaluational functional on \( C(T) \), i.e. if \( \xi \in C(T) \), \( x'_t(\xi) = \xi(t) \). Let \( X \) and \( S_n \) be the random variables with values in \( C(T) \) determined by \( \{X(t), t \in T\} \) and \( \{S_n(t), t \in T\} \) respectively. \( E\{\|X\|\} < \infty \) by hypothesis so that \( E\left\{ X \, | \, \mathcal{F}_n \right\} \) exists and we have for \( t \in T_0 \):

\[
x'_t \left( E\left\{ X \, | \, \mathcal{F}_n \right\} - S_n \right) = \frac{1}{2} E\left\{ x'_t(X) \, | \, \mathcal{F}_n \right\} - S_n(t) = 0 \quad \text{w.p.1.}
\]

where we have used the fact that a continuous linear functional commutes with the strong conditional expectation operator [6, p. 357]. \( T_0 \) is countable so (3.2) holds w.p.1. simultaneously for all \( t \in T_0 \). \( T_0 \) is dense in \( T \), so the set \( \{x'_t, t \in T_0\} \) separates points of \( C(T) \), hence

\[
E\left\{ X \, | \, \mathcal{F}_n \right\} = S_n \quad \text{w.p.1.}
\]
Therefore by Theorem A, the sequence \((S_n)_{n \in \mathbb{N}}\) converges strongly w.p.1. The only possible limit is \(X\), so \(S_n(t) \to X(t)\) uniformly in \(t\) w.p.1.

**Note.** Let \(y_1, y_2, \ldots\) be independent mean zero random variables and let \(S_n = \sum_1^n y_n\). If for some subsequence \((n_k)\), \((S_{n_k})\) converges w.p.1. to an integrable random variable \(x\), the whole sequence converges to \(x\). Using this remark, we see that we can replace the convergence in any or all of (B1), (B2), and (B3) with convergence along some subsequence and the theorem still holds.

If we apply the mean convergence part of Theorem A we get

**Corollary 1.** If \(E\left\{ \sup |X(t)|^p : t \in T \right\} < \infty \) for some \( p \), \( 1 \leq p < \infty \), then any of (B1), (B2) or (B3) implies

\[
E\left\{ \sup |X(t) - S_n(t)|^p : t \in T \right\} \to 0 \quad \text{as} \quad n \to \infty.
\]

4. Applications—Brownian motion. Let \(\eta_1, \eta_2, \ldots\) be a sequence of independent normal random variables with mean zero and variance one. Let \(a > 0\) and let \(\phi_1, \phi_2, \ldots\) be any complete orthonormal sequence in \(L^2[0, a]\). Define

\[
\Phi_j(t) = \int_0^t \phi_j(u) du \quad j = 1, 2, \ldots
\]

L. A. Shepp [7] has proved that

(4.1) for each \(t\) the series \(\sum_1^\infty \eta_j \Phi_j(t)\) converges w.p.1. to \(Z(t)\), where \(\{Z(t) : 0 \leq t \leq a\}\) is separable Brownian motion from the origin, and McKean has shown that

(4.2) the convergence in (4.1) is even uniform in \(t\) w.p.1. We have an immediate independent proof of (4.2) since \(Z(t)\) is continuous w.p.1. [3, p. 393], and by the reflection principle [3, p. 392], \(\sup_{0 \leq t \leq a} |Z(t)|^p\) is integrable for any \(p > 0\). For each \(j\), \(\{\eta_j \phi_j(t), 0 \leq t \leq a\}\) is a continuous process with mean zero and by (4.1) the partial sums converge for each \(t\) w.p.1. so by Theorem B and Corollary 1, we get:

**Corollary 2.** \(\sum_1^\infty \eta_j \Phi_j(t)\) converges uniformly in \(t\) w.p.1., and moreover \(E\left\{ \sup_{0 \leq t \leq a} |Z(t) - \sum_1^\infty \eta_j \Phi_j(t)|^p \right\} \to 0\) as \(n \to \infty\) for all \(p > 0\).

This includes as a special case the result of Paley and Wiener [5]: If \(\eta_0, \eta_1, \ldots\) and \(\eta'_0, \eta'_1, \ldots\) are independent normal mean zero and variance one random variables, then

\[
\eta_0t + \frac{1}{2\pi} \sum_1^\infty \frac{1 - \cos 2\pi nt}{n} \eta_n + \frac{1}{2\pi} \sum_1^\infty \frac{\sin 2\pi nt}{n} \eta'_n,
\]

converges uniformly to \(Z(t)\) in \([0, 1]\). Paley and Wiener actually
proved only that a subsequence of partial sums converges, whereas we see that the whole sequence converges, a fact also proved by Delporte [2].

5. Eigenfunction expansions. Let \( \{ X(t), t \in [0, a] \} \) be a continuous gaussian process with mean zero and covariance function \( \Gamma(s, t) \), \( x, t \in [0, a] \). Consider the normalized eigenfunctions \( \phi_k \) and eigenvalues \( \lambda_k \) of the equation

\[
\lambda \phi(s) = \int_0^a \Gamma(s, t) \phi(t) dt.
\]

\( X(t) \) has the expansion

\[
(5.1) \quad X(t) = \sum_{i=1}^{\infty} \eta_i \lambda_k^{1/2} \phi_k(t) \quad 0 \leq t \leq a
\]

where \( \eta_1, \eta_2, \cdots \) is a sequence of independent normal mean zero variance one random variables. This expansion converges for each \( t \) w.p.1. and also converges in the \( L^2[0, a] \) mean w.p.1., i.e.,

\[
\int_0^1 \left| X(t) - \sum_{i=1}^{n} \eta_i \lambda_k^{1/2} \phi_k(t) \right|^2 dt \to 0 \quad \text{as} \quad n \to \infty.
\]

(This is proved in [4] for the case where \( X(t) \) is stationary; it is well known that the results remain true in the nonstationary case.) Now each \( \phi_k \) is continuous so using either (B2) ⇒ (B1) or (B3) ⇒ (B1) of Theorem B, we have:

**Corollary 3.** If \( E \left\{ \sup_{0 \leq t \leq a} |X(t)| \right\} < \infty \), the convergence in (5.1) is uniform in \( t \) w.p.1.

**References**


**University of Illinois**