AN EXTENSION TO THE STRONG DOMINATION MARTINGALE INEQUALITY

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ABSTRACT. For each $1 , there exists a positive constant <math>c_p$, depending only on p, such that the following holds. Let (d_k) , (e_k) be real-valued martingale difference sequences. If for for all bounded nonnegative predictable sequences (s_k) and all positive integers k we have

$$E[s_k \vee |e_k|] \le E[s_k \vee |d_k|]$$

then for all positive integers n we have

$$\left\| \sum_{k=1}^{n} e_k \right\|_p \le c_p \left\| \sum_{k=1}^{n} d_k \right\|_p.$$

1. Introduction

Let (Ω, \mathcal{F}, P) be a probability space, and let (\mathcal{F}_k) be a filtration on (Ω, \mathcal{F}, P) . (We will suppose that $\mathcal{F}_0 = \{\emptyset, \Omega\}$.) If an adapted sequence (d_k) is a real-valued martingale difference sequence, Burkholder's inequality [3] shows that for any $1 , if <math>(v_k)$ is a predictable sequence bounded in absolute value by 1, then there exists a positive constant c_p , depending only on p, such that such that for all positive integers n

$$\left\| \sum_{k=1}^{n} v_k d_k \right\|_p \le c_p \left\| \sum_{k=1}^{n} d_k \right\|_p.$$

Later Burkholder [4] extended this result to *subordination* martingales: if (d_k) , (e_k) are two martingale difference sequences such that (e_k) is subordinate to (d_k) , that is, for all $k \geq 1$,

$$(1) |e_k| \le |d_k|$$

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then there exists a positive constant c_p , depending only on p, such that for all positive integers n

(2)
$$\left\| \sum_{k=1}^{n} e_k \right\|_p \le c_p \left\| \sum_{k=1}^{n} d_k \right\|_p.$$

A different approach to this inequality was proposed by Kwapień and Woycziński [9] (see also [10]). Two adapted sequence (d_k) and (e_k) are said to be tangent if for each $k \geq 1$, we have that the law of d_k conditionally on \mathcal{F}_{k-1} is the same as the law of e_k conditionally on \mathcal{F}_{k-1} , that is,

(3)
$$P(d_k > \lambda | \mathcal{F}_{k-1}) = P(e_k > \lambda | \mathcal{F}_{k-1})$$

for all real numbers λ . Answering a conjecture of Kwapień and Woycziński [9], it was proved by Hitczenko [6] and Zinn [15] that for $1 that there exists a positive constant <math>c_p$, depending only on p, such that if (d_k) and (e_k) are martingale difference sequences and (d_k) , (e_k) are tangent, then for all positive integers n we have equation (2).

Given two adapted sequences, (e_k) is said to be strongly dominated by (d_k) if for each $k \geq 1$,

(4)
$$P(|e_k| > \lambda | \mathcal{F}_{k-1}) \le P(|d_k| > \lambda | \mathcal{F}_{k-1})$$

for all $\lambda \geq 0$. It is obvious that the case of (1) and the case of (3) are contained in the cases of (4). Thus the following result of Kwapień and Woycziński [9] is a common generalization of these two results: if (d_k) , (e_k) are two martingale difference sequences such that (e_k) is strongly dominated by (d_k) , then there exists a positive constant c_p , depending only on p, such that for all positive integers n equation (2) holds.

The purpose of this paper is to use a different approach to provide another common generalization of those two results, an even a further extension to Kwapień and Woycziński's result.

Theorem 1. For each $1 , there exists a positive constant <math>c_p$, depending only on p, such that the following holds. Let (d_k) , (e_k) be real-valued martingale difference sequences. If for for all bounded nonnegative predictable sequence (s_k) and all positive integers k we have

(5)
$$E[s_k \vee |e_k|] \le E[s_k \vee |d_k|]$$

then for all positive integers n we have equation (2).

Remark (a). We have that (5) is equivalent to

(6)
$$E[(\lambda \vee |e_k|)|\mathcal{F}_{k-1}] \leq E[(\lambda \vee |d_k|)|\mathcal{F}_{k-1}]$$

for all $\lambda \geq 0$. This is because for any $A_k \in \mathcal{F}_{k-1}$ and $a \geq 0$ we have that $(a\chi_{A_k^c} \vee \lambda)$ is predictable, and hence

$$E[(a\chi_{A_k^c} \vee \lambda) \vee |e_k| - a\chi_{A_k^c}] \leq E[(a\chi_{A_k^c} \vee \lambda) \vee |d_k| - a\chi_{A_k^c}]$$

When a intends to infinity, we obtain

$$E[(\lambda \vee |e_k|)\chi_{A_k}] \leq E[(\lambda \vee |d_k|)\chi_{A_k}]$$

which is equivalent to (6).

Remark (b). To see that Theorem 1 is really an extension to Kwapień and Woycziński's result, we just simply observe that (4) is equivalent to

$$P(\{|e_k| > \lambda\} \cap A_k) \le P(\{|d_k| > \lambda\} \cap A_k),$$

and (6) is equivalent to

$$\int_{\lambda}^{\infty} P(\{|e_k| > t\} \cap A_k) dt \le \int_{\lambda}^{\infty} P(\{|d_k| > t\} \cap A_k) dt$$

for all $A_k \in \mathcal{F}_{k-1}$.

Remark (c). Once we have Theorem 1, we can obtain that for $\kappa \geq 1$, if

$$P(|e_k| > \lambda | \mathcal{F}_{k-1}) \le \kappa P(|d_k| > \lambda | \mathcal{F}_{k-1}),$$

we have

(7)
$$\left\| \sum_{k=1}^{n} e_k \right\|_p \le \kappa c_p \left\| \sum_{k=1}^{n} d_k \right\|_p.$$

This is because

$$\int_{\lambda}^{\infty} P(\{|e_{k}| > t\} \cap A_{k}) dt \leq \kappa \int_{\lambda}^{\infty} P(\{|d_{k}| > t\} \cap A_{k}) dt$$

$$= \kappa \int_{\frac{\lambda}{\kappa}}^{\infty} P\left(\{|d_{k}| > \frac{t}{\kappa}\} \cap A_{k}\right) d\left(\frac{t}{\kappa}\right)$$

$$\leq \int_{\lambda}^{\infty} P(\{\kappa|d_{k}| > t\} \cap A_{k}) dt$$

Hence

$$E[(\lambda \vee |e_k|)|\mathcal{F}_{k-1}] \leq E[(\lambda \vee \kappa |d_k|)|\mathcal{F}_{k-1}]$$

and equation (7) follows.

Let us give an application of Theorem 1. In fact this application is essentially equivalent to Theorem 1, and indeed will play a large role in its proof. We will consider the probability space $[0,1]^{\mathbb{N}}$ equipped with the product Lebesgue measure \mathcal{L} , and consider the filtration (\mathcal{L}_k) , where \mathcal{L}_k is the minimal σ -field for which the first k coordinate functions of

 $[0,1]^{\mathbb{N}}$ are measurable. Then two sequences (d_k) and (e_k) are tangent if

$$e_k(x_1,\ldots,x_k) = d_k(x_1,\ldots,x_{k-1},\phi_k(x_1,\ldots,x_k))$$

where $(\phi_k : [0,1]^k \to [0,1])$ is a sequence of measurable functions such that $\phi_k(x_1,\ldots,x_{k-1},\cdot)$ is a measure preserving map for almost all x_1,\ldots,x_{k-1} .

We will consider a more general situation. Suppose we have a sequence of linear operators $(T_k(x_1,\ldots,x_{k-1}))$, depending measurably upon $(x_k) \in [0,1]^{\mathbb{N}}$, that are bounded operators on both $L_1([0,1])$ and $L_{\infty}([0,1])$ with norm 1. Then consider the condition

(8)
$$e_k(x_1,\ldots,x_{k-1},\cdot) = [T_k(x_1,\ldots,x_{k-1})]d_k(x_1,\ldots,x_{k-1},\cdot).$$

Theorem 2. For each $1 , there exists a positive constant <math>c_p$, depending only on p, such that the following holds. If (d_k) , (e_k) and (T_k) are as above satisfying (8), then for all positive integers n we have equation (2).

We will also need the following intermediate result. For any random variable f, let $f^{\#}$ be the decreasing rearrangement of |f|, that is,

$$f^{\#}(t) = \sup\{s \in \mathbb{R} : P(|f| < s) < t\}.$$

Theorem 3. For each $1 , there exists a positive constant <math>c_p$, depending only on p, such that the following holds. Let (d_k) , (e_k) be martingale difference sequences on $[0,1]^{\mathbb{N}}$ with respect to (\mathcal{L}_k) . Suppose that for each positive integer k

$$\int_0^t (e_k(x_1, \dots, x_{k-1}, \cdot))^{\#}(s)ds \le \int_0^t (d_k(x_1, \dots, x_{k-1}, \cdot))^{\#}(s)ds$$

for all $t \in [0,1]$ and almost all x_1, \ldots, x_{k-1} . Then for all positive integers n we have equation (2).

2. The Discrete Type Case

In this section we will prove Theorems 2 and 3 in a special discrete situation, which we now describe. For any positive integer N, let Σ_N be the σ -field generated by the partition $\{[\frac{i-1}{N}, \frac{i}{N}) : i = 1, 2, \dots, N\}$. Define a filtration (\mathcal{F}_k) on $[0,1]^{\mathbb{N}}$ by $\mathcal{F}_k = \mathcal{L}_{k-1} \otimes \Sigma_N$. Suppose (d_k) , (e_k) are (\mathcal{F}_k) -adapted. Then for each k and for each x_1, \dots, x_{k-1} , we see that $d_k(x_1, \dots, x_{k-1}, \cdot)$ and $e_k(x_1, \dots, x_{k-1}, \cdot)$ are Σ_N -measurable simple functions on [0,1). Therefore d_k and e_k can be written as N-dimensional vectors and $T_k(x_1, \dots, x_{k-1})$ can be represented by a $N \times N$

matrix, that is,

$$\begin{bmatrix} e_k(1) \\ e_k(2) \\ \vdots \\ e_k(N) \end{bmatrix} = \begin{bmatrix} a_k(1,1), & \dots & , a_k(1,N) \\ a_k(2,1), & \dots & , a_k(2,N) \\ \vdots & & & \vdots \\ a_k(N,1), & \dots & , a_k(N,N) \end{bmatrix} \begin{bmatrix} d_k(1) \\ d_k(2) \\ \vdots \\ d_k(N) \end{bmatrix}$$

where

$$d_k(i) = d_k(x_1, \dots, x_{k-1}, i) = d_k(x_1, \dots, x_k) \text{ if } x_k \in \left[\frac{i-1}{N}, \frac{i}{N}\right)$$

$$e_k(i) = e_k(x_1, \dots, x_{k-1}, i) = e_k(x_1, \dots, x_k) \text{ if } x_k \in \left[\frac{i-1}{N}, \frac{i}{N}\right)$$

$$T_k = T_k(x_1, \dots, x_{k-1}) = [(a_k(x_1, \dots, x_{k-1}))(i, j)]_{N \times N} = [a_k(i, j)]_{N \times N}$$

The condition of being martingale difference sequences implies that

$$\sum_{i=1}^{N} d_k(i) = \sum_{i=1}^{N} e_k(i) = 0$$

Proposition 4. Theorem 2 holds in the case that (d_k) and (e_k) are adapted to the filtration (\mathcal{F}_k) described above.

In this discrete case, the boundedness of $||T_k||_{L_1([0,1])}$ and $||T_k||_{L_\infty([0,1])}$ by 1 is equivalent to the condition that $\sum_{j=1}^N |a_k(i,j)| \le 1$ for all i and $\sum_{i=1}^N |a_k(i,j)| \le 1$ for all j. We claim that without loss of generality, we can assume that every row sum and column sum of T_k is 0, that is,

$$\sum_{j=1}^{N} a_k(i,j) = \sum_{i=1}^{N} a_k(i,j) = 0$$

for all i and j. Suppose the i^{th} row sum $\sum_{j=1}^{N} a_k(i,j) = R_k(i)$. Let T'_k be the liner operator defined by

$$T'_{k} = \left[a_{k}(i,j) - \frac{R_{k}(i)}{N} \right]_{N \times N}$$

It is clear that every row sum of T'_k is 0 and

$$(T'_k d_k)(i) = \sum_{j=1}^{N} \left(a_k(i,j) - \frac{R_k(i)}{N} \right) d_k(j)$$

$$= \sum_{j=1}^{N} a_k(i,j) d_k(j) - \frac{R_k(i)}{N} \sum_{j=1}^{N} d_k(j)$$

$$= e_k(i)$$

Now we can assume that every row sum of T_k is 0. Similarly suppose the j^{th} column sum $\sum_{i=1}^{N} a_k(i,j) = C_k(j)$. Let T''_k be the linear operator defined by

$$T_k'' = \left[a_k(i,j) - \frac{C_k(j)}{N}\right]_{N \times N}$$

Again it is clear that every row sum and column sum of T''_k is 0 and

$$(T_k''d_k)(i) = \sum_{j=1}^N \left(a_k(i,j) - \frac{C_k(j)}{N} \right) d_k(j)$$

$$= \sum_{j=1}^N a_k(i,j) d_k(j) - \frac{1}{N} \sum_{j=1}^N C_k(j) d_k(j)$$

$$= e_k(i)$$

since

$$\sum_{i=1}^{N} e_k(i) = \sum_{j=1}^{N} C_k(j) d_k(j) = 0$$

After adjusting T_k , it is easy to check that the norms of T_k may be enlarged up to 4. Of course, we can pick up $T_k/4$ instead and absorb the 4 into the constant c_p .

A nonnegative real matrix is said to be *doubly stochastic* if each of its row and column sum is 1. A sub-doubly stochastic matrix means that each of its row and column sum is less than or equal to 1. Therefore we can change the assumption in Proposition 4 to be that: "for almost all x_1, \ldots, x_{k-1} , every row sum and column sum of the matrix from T_k is 0, and the matrix from $|T_k|$ is sub-doubly stochastic for each positive integer k"

One of the fundamental results in the theory of doubly stochastic matrices was introduced by Birkhoff [1] (or see for example [12]).

Theorem A. If M is a doubly stochastic matrix, then

$$M = \sum_{i=1}^{S} \theta_i P_i$$

where P_i are permutation matrices, and the θ_i are nonnegative numbers satisfying $\sum_{i=1}^{S} \theta_i = 1$.

Lemma 5. If M is a $n \times n$ sub-doubly stochastic matrix, then there exists a $2n \times 2n$ doubly stochastic matrix such that its upper left $n \times n$ sub-matrix is M.

Proof. Suppose that R(i) is the i^{th} row sum of M, C(j) is the j^{th} column sum and S is the sum of all entries. Let

$$A = \begin{bmatrix} \frac{1-R(1)}{n}, & \dots & , \frac{1-R(1)}{n} \\ \vdots & & \vdots \\ \frac{1-R(n)}{n}, & \dots & , \frac{1-R(n)}{n} \end{bmatrix}_{n \times n}$$

$$B = \begin{bmatrix} \frac{1-C(1)}{n}, & \dots & , \frac{1-C(n)}{n} \\ \vdots & & \vdots \\ \frac{1-C(1)}{n}, & \dots & , \frac{1-C(n)}{n} \end{bmatrix}_{n \times n}$$

$$C = \text{Diag} \begin{bmatrix} \frac{S}{n}, & \dots & , \frac{S}{n} \end{bmatrix}_{n \times n}$$

Then define

$$M' = \left[\begin{array}{cc} M & A \\ B & C \end{array} \right]_{2n \times 2n}$$

It is easy to check that M' is a doubly stochastic matrix.

Lemma 6. If M is a sub-doubly stochastic matrix, then there exists a sub-doubly stochastic matrix N such that M + N is doubly stochastic.

Proof. Let M' be the $2n \times 2n$ doubly stochastic matrix such that its upper left $n \times n$ sub-matrix is M. By Theorem A,

$$M' = \sum_{i=1}^{S} \theta_i P_i'$$

where P'_i are $2n \times 2n$ permutation matrices and $\sum_{i=1}^{S} \theta_i = 1$. Suppose that P_i is the upper left $n \times n$ sub-permutation matrix of P'_i , then

$$M = \sum_{i=1}^{S} \theta_i P_i$$

Let Q_i be a $n \times n$ sub-permutation matrix such that $P_i + Q_i$ is a permutation matrix, say R_i . Define

$$N = \sum_{i=1}^{S} \theta_i Q_i$$

thus

$$M + N = \sum_{i=1}^{S} \theta_i R_i$$

which is a doubly stochastic matrix.

Lemma 7. Let M be an $n \times n$ matrix. If every row sum and column sum of M is 0 and |M| is sub-doubly stochastic, then

$$M = \sum_{i=1}^{S} \theta_i P_i$$

where P_i are permutation matrices, $\sum_{i=1}^{S} \theta_i = 0$ and $\sum_{i=1}^{S} |\theta_i| = 1$

Proof. Let

$$A = \frac{|M| + M}{2}$$
$$B = \frac{|M| - M}{2}$$

so A and B are nonnegative, and 2A and 2B are sub-doubly stochastic. By Lemma 6, there exists a sub-doubly stochastic matrix C such that 2(A+C) is a doubly stochastic. But A and B have the same row sums and column sums, and hence 2(B+C) is also a doubly stochastic. By applying Theorem A, we have

$$2(A+C) = \sum_{i=1}^{m} \lambda_i Q_i$$

$$2(B+C) = \sum_{i=1}^{m'} \lambda_i' Q_i'$$

where Q_i , Q_i' are permutation matrices, and the λ_i , λ_i' are nonnegative numbers satisfying $\sum_{i=1}^{m} \lambda_i = \sum_{i=1}^{m'} \lambda_i' = 1$. Then the result follows because

$$M = (A + C) - (B + C) = \sum_{i=1}^{m} \frac{\lambda_i}{2} Q_i - \sum_{i=1}^{m'} \frac{\lambda'_i}{2} Q'_i$$

Proof of Proposition 4. From Lemma 7, we know that for each $k \geq 1$ and almost all x_1, \ldots, x_{k-1}

$$T_k(x_1, \dots, x_{k-1}) = \sum_{i_k=1}^{S_k} \theta_{k, i_k}(x_1, \dots, x_{k-1}) \cdot P_{k, i_k}(x_1, \dots, x_{k-1})$$

where P_{k,i_k} are permutation matrices, $\sum_{i=1}^{S_k} \theta_{k,i_k} = 0$, and $\sum_{i=1}^{S_k} |\theta_{k,i_k}| = 1$. Let

(9)
$$h_{k,i_k}(x_1,\ldots,x_{k-1},\cdot) = [P_{k,i_k}(x_1,\ldots,x_{k-1})]d_k(x_1,\ldots,x_{k-1},\cdot).$$

Then

$$e_k = \left[\sum_{i_k=1}^{S_k} \theta_{k,i_k} P_{k,i_k}\right] d_k$$
$$= \sum_{i_k=1}^{S_k} |\theta_{k,i_k}| \varepsilon_{k,i_k} h_{k,i_k}$$

where $\varepsilon_{k,i_k} = \operatorname{sgn}(\theta_{k,i_k})$.

Now we need to consider the probability space $\Omega_1 \times \Omega_2$, where $\Omega_1 = \Omega_2 = [0,1]^{\mathbb{N}}$. We consider all of the previous random variables as random variables on this new probability space, depending only upon the first coordinate ω_1 . We define a filtration (\mathcal{G}_k) where $\mathcal{G}_k = \mathcal{F}_k \otimes \mathcal{L}_{k+1}$.

We define a predictable sequence of random variables (I_k) so that for each $\omega_1 \in \Omega_1$, the random variable $I_k(\omega_1, \cdot)$ takes the value i with probability $|\theta_{k,i}(\omega_1)|$. Then we see that

$$e_k = E[\varepsilon_{k,I_k} h_{k,I_k} | \mathcal{L} \otimes \{\emptyset, \Omega_2\}].$$

Hence, since conditional expectation is a contraction on L_p

$$\left\| \sum_{k=1}^{n} e_k \right\|_p \le \left\| \sum_{k=1}^{n} \varepsilon_{k,I_k} h_{k,I_k} \right\|_p.$$

Now we see that (ε_{k,I_k}) is a predictable sequence bounded by 1. Hence by Burkholder's inequality, we see that

$$\left\| \sum_{k=1}^n \varepsilon_{k,I_k} h_{k,I_k} \right\|_p \le c_p \left\| \sum_{k=1}^n h_{k,I_k} \right\|_p.$$

Next, observing (9), since P_{k,i_k} are permutation matrices, for each $k \ge 1$, $i_k = 1, 2, ..., S_k$, h_{k,i_k} is just an x_k -rearrangement of d_k . that is

$$h_{k,i_k}(x_1,\ldots,x_{k-1},j)=d_k(x_1,\ldots,x_{k-1},\pi_{k,i_k}(j))$$

for some permutation π_{k,i_k} . Thus for any sequence (i_k) we have that (h_{k,i_k}) and (d_k) are tangent sequences. But then we see that (h_{k,I_k}) and (d_k) are tangent sequences. Hence there exists a positive constant c_p such that

$$\left\| \sum_{k=1}^{n} h_{k,I_k} \right\|_{p} \le c_p \left\| \sum_{k=1}^{n} d_k \right\|_{p}.$$

The result follows.

Proposition 8. Theorem 3 holds in the case that (d_k) and (e_k) are adapted to the filtration (\mathcal{F}_k) described above.

This will follow immediately from the following well-known result [11].

Theorem B. $f = (f_1, f_2, ..., f_N), g = (g_1, g_2, ..., g_N)$ are N-dimensional real-valued vectors. $f^{\#} = (f_1^{\#}, f_2^{\#}, ..., f_N^{\#})$ is the decreasing rearrangement of $|f| = (|f_1|, |f_2|, ..., |f_N|)$. Then

$$\sum_{k=1}^{n} g_k^{\#} \le \sum_{k=1}^{n} f_k^{\#}$$

for all n = 1, 2, ..., N if and only if there exists a matrix $T = [a_{ij}]_{N \times N}$ such that Tf = g, $\sum_{i=1}^{N} |a_{ij}| \le 1$ and $\sum_{j=1}^{N} |a_{ij}| \le 1$.

3. The General Case

The following theorem was proved by Crowe, Zweibel and Rosenbloom [5].

Theorem C. Suppose f, g are random variables on [0,1], then for $1 \le p \le \infty$,

$$||f^{\#} - g^{\#}||_{p} \le ||f - g||_{p}$$

Lemma 9. $1 \leq p < \infty$. (Ω, \mathcal{F}, P) is a probability space. Let $d(\omega, x), e(\omega, x) \in L_p(\Omega \times [0, 1])$ be two random variables such that for almost every $\omega \in \Omega$ that

$$\int_0^t (e(\omega, \cdot))^\# \le \int_0^t (d(\omega, \cdot))^\#$$

and

$$\int_0^1 d(\omega, \cdot) = \int_0^1 d(\omega, \cdot) = 0.$$

Then given $\epsilon > 0$, there exists a positive integer N and $d', e' \in L_p(\Omega \times [0,1])$ that are measurable with respect to $\mathcal{F} \otimes \Sigma_N$ such that $\|d-d'\|_p$, $\|e-e'\| \leq \epsilon$,

$$\int_0^t (e'(\omega,\cdot))^\# \le \int_0^t (d'(\omega,\cdot))^\#$$

and

$$\int_0^1 d'(\omega, \cdot) = \int_0^1 d'(\omega, \cdot) = 0.$$

Proof. For every $\epsilon > 0$, pick $0 < \gamma < \min\{\epsilon/[7(\|d\|_p \lor \|e\|_p)], 1/3\}$. Fix $\omega \in \Omega$, and regard the functions as functions of only one variable x on [0,1]. Hence there exist simple functions

$$\bar{d} = \sum_{i=1}^{S} \bar{\alpha}_i \chi_{A_i}$$

$$\bar{e} = \sum_{i=1}^{S} \bar{\beta}_i \chi_{B_i}$$

such that

$$\|\bar{d} - d\|_{L_p([0,1])} \le \gamma \|d\|_{L_1([0,1])}.$$

$$\|\bar{e} - e\|_{L_p([0,1])} \le \gamma \|e\|_{L_1([0,1])}.$$

We may suppose without loss of generality that the sets A_i and B_i are the sets of the form $[r_1, s_1)$, where the r_i and s_i are rational numbers. Furthermore, we will suppose that $A_{i_1} \cap A_{i_2} = \emptyset$ and $B_{i_1} \cap B_{i_2} = \emptyset$ for $i_1 \neq i_2$.

Let $N_0 = N_0(\omega)$ be the least common denominator of all these rational numbers. For each ω , since $d(\omega, \cdot)^{\#}$, $e(\omega, \cdot)^{\#}$ are Reimann integrable as a function of x, there is a number $N_1 = N_1(\omega)$ that is a multiple of N_0 and such that for all $n \geq N_1$ that

$$||E[d(\omega,\cdot)^{\#}|\Sigma_n] - d(\omega,\cdot)^{\#}||_{L_p([0,1])} \leq \gamma ||d||_{L_1([0,1])}.$$

$$||E[e(\omega,\cdot)^{\#}|\Sigma_n] - e(\omega,\cdot)^{\#}||_{L_p([0,1])} \leq \gamma ||e||_{L_1([0,1])}.$$

Now let $d_n=d\chi_{\{N_1(\omega)\leq n\}}$ and $e_n=e\chi_{\{N_1(\omega)\leq n\}}$. Then $d_n\to d$ and $e_n\to e$ in $L_p(\Omega\times[0,1])$. So pick N such that

$$||d_N - d||_{L_p(\Omega \times [0,1])} < \epsilon/7,$$

 $||e_N - e||_{L_p(\Omega \times [0,1])} < \epsilon/7.$

For each fixed $\omega \in \{N_1(\omega) \leq N\}$, $[\frac{i-1}{N}, \frac{i}{N})$ is either contained in some A_j or disjoint to all A_j . Let $\alpha_i = \bar{\alpha}_j$ if $[\frac{i-1}{N}, \frac{i}{N}) \subset A_j$ for some j, and $\alpha_i = 0$ otherwise. Let $\chi_i = \chi_{\lceil \frac{i-1}{N}, \frac{i}{N} \rangle}$. Thus

(10)
$$\left\| \sum_{i=1}^{N} \alpha_i \chi_i - d_N \right\|_{L_p([0,1])} \le \gamma \|d\|_{L_1([0,1])}$$

and

$$\left(\sum_{i=1}^{N} \alpha_i \chi_i\right)^{\#} = \sum_{i=1}^{N} \varepsilon_{\sigma(i)} \alpha_{\sigma(i)} \chi_i$$

for some permutation σ , where $\varepsilon_i = \operatorname{sgn}(\alpha_i)$. By Theorem C,

(11)
$$\left\| \sum_{i=1}^{N} \varepsilon_{\sigma(i)} \alpha_{\sigma(i)} \chi_{i} - d_{N}^{\#} \right\|_{L_{p}([0,1])} \leq \gamma \|d\|_{L_{1}([0,1])}$$

and also the analogous statement holds for e.

Now if we set

$$E[d_N^{\#}|\Sigma_N] = \sum_{i=1}^N \alpha_i'' \chi_i$$

then

(12)
$$\left\| \sum_{i=1}^{N} \alpha_i'' \chi_i - d_N^{\#} \right\|_{L_p([0,1])} \le \gamma \|d\|_{L_1([0,1])}$$

Note that in this case that

$$\int_{0}^{t} \sum_{i=1}^{N} \alpha_{i}'' \chi_{i} = \int_{0}^{t} d_{N}^{\#}$$

if $t = \frac{j}{N}$, j = 0, 1, 2, ..., N. Then by (11) and (12),

$$\left\| \sum_{i=1}^{N} \alpha_i'' \chi_i - \sum_{i=1}^{N} \varepsilon_{\sigma(i)} \alpha_{\sigma(i)} \chi_i \right\|_{L_p([0,1])} \le 2\gamma \|d\|_{L_1([0,1])}$$

By doing the reverse process of taking decreasing rearrangement of $|\sum_{i=1}^{N} \alpha_i \chi_i|$, and setting

$$\hat{\alpha}_i = \varepsilon_{\sigma^{-1}(i)} \alpha''_{\sigma^{-1}(i)}$$

we have

(13)
$$\left\| \sum_{i=1}^{N} \hat{\alpha}_{i} \chi_{i} - \sum_{i=1}^{N} \alpha_{i} \chi_{i} \right\|_{L_{p}([0,1])} \leq 2\gamma \|d\|_{L_{1}([0,1])}$$

From (10) and (13),

$$\left\| \sum_{i=1}^{N} \hat{\alpha}_{i} \chi_{i} - d_{N} \right\|_{L_{p}([0,1])} \le 3\gamma \|d\|_{L_{1}([0,1])}$$

For $t = \frac{j}{N}$, $j = 0, 1, 2, \dots, N$, it is clear that

$$\int_0^t \left(\sum_{i=1}^N \hat{\alpha}_i \chi_i \right)^\# = \int_0^t \sum_{i=1}^N \alpha_i'' \chi_i = \int_0^t d_N^\#.$$

Furthermore, if we set

$$\zeta = E\left[\sum_{i=1}^{N} \hat{\alpha}_i \chi_i\right]$$

then

$$|\zeta| \le 3\gamma ||d||_{L_1([0,1])}.$$

We can also perform this same construction for e, the analogues of $\hat{\alpha}_i$ and ζ being $\hat{\beta}_i$ and η . Thus we see that for $t = \frac{j}{N}, \ j = 0, 1, 2, \dots, N$

that

(14)
$$\int_{0}^{t} \left(\sum_{i=1}^{N} (\hat{\alpha}_{i} - \zeta) \chi_{i} \right)^{\#}$$

$$\leq \int_{0}^{t} \left(\sum_{i=1}^{N} (|\hat{\alpha}_{i}| + |\zeta|) \chi_{i} \right)^{\#}$$

$$\leq \int_{0}^{t} \left(\sum_{i=1}^{N} \hat{\alpha}_{i} \chi_{i} \right)^{\#} + 3\gamma \|d\|_{L_{1}([0,1])} \cdot t$$

$$= \int_{0}^{t} d_{N}^{\#} + 3\gamma \|d\|_{L_{1}([0,1])} \cdot t$$

$$\leq (1 + 3\gamma) \int_{0}^{t} d_{N}^{\#}$$

and similarly

(15)
$$\int_{0}^{t} \left(\sum_{i=1}^{N} \left(\hat{\beta}_{i} - \eta \right) \chi_{i} \right)^{\#} \geq \int_{0}^{t} e_{N}^{\#} - 3\gamma \|e\|_{L_{1}([0,1])} \cdot t$$
$$\geq (1 - 3\gamma) \int_{0}^{t} e_{N}^{\#}$$

Thus, we are ready to define d' and e'. Let

$$d' = (1 + 3\gamma) \sum_{i=1}^{N} (\hat{\alpha}_i - \zeta) \chi_i$$
$$e' = (1 - 3\gamma) \sum_{i=1}^{N} (\hat{\beta}_i - \eta) \chi_i$$

It is clear that E[d']=E[e']=0. Combining (14) and (15), we have for $t=\frac{j}{N},\ j=0,1,2,\ldots,N$

$$\int_0^t (e')^\# \le \int_0^t e_N^\# = \int_0^t d_N^\# \le \int_0^t (d')^\#.$$

But then by linear interpolation, this follows for all $t \in [0, 1]$. Now an easy argument shows that

$$||d' - d||_{L_p(\Omega \times [0,1])} \le 6\gamma ||d||_{L_1(\Omega \times [0,1])} + \epsilon/7$$

$$||e' - e||_{L_p(\Omega \times [0,1])} \le 6\gamma ||e||_{L_1(\Omega \times [0,1])} + \epsilon/7$$

and we are done.

Proof of Theorem 3. For each $1 \le k \le n$, apply Lemma 9, there exists an integer N_k and functions d'_k , e'_k satisfying $\|d'_k - d_k\|_p$, $\|e'_k - e_k\|_p \le \epsilon$ such that (d'_k) and (e'_k) are adapted to $(\mathcal{L}_{k-1} \otimes \Sigma_N)$, where N is the least common multiple of N_k , keep the martingale property, and

$$\int_0^t (e'_k(x_1, \dots, x_{k-1}, \cdot))^{\#}(s)ds \le \int_0^t (d'_k(x_1, \dots, x_{k-1}, \cdot))^{\#}(s)ds$$

for all $t \in [0,1]$. By Proposition 8, there exist a positive constant c_p such that

$$\left\| \sum_{k=1}^{n} e_k' \right\|_p \le c_p \left\| \sum_{k=1}^{n} d_k' \right\|_p$$

 $||d_k - d_k'||_p \to 0$ and $||e_k - e_k'||_p \to 0$ as $\epsilon \to 0$. The result follows. \square

Proof of Theorem 2. If f is a random variable on (Ω, \mathcal{F}, P) , $1 \leq p < \infty$, $0 \leq t \leq 1$, we define the K-functional by

$$K(t, f; L_p, L_{\infty}) = \inf_{f_0 + f_1 = f} \{ ||f_0||_p + t ||f_1||_{\infty} \}.$$

J. Peetre [14] has shown that

$$K(t, f; L_1, L_\infty) = \int_0^t f^{\#}(s) ds.$$

Hence it follows that if T is an operator on both $L_1([0,1])$ and $L_{\infty}([0,1])$ with norm bounded by 1, then for $t \geq 0$

$$\int_0^t (Tf)^{\#}(s)ds \le \int_0^t f^{\#}(s)ds.$$

Thus the result follows from Theorem 3.

Lemma 10. Let f and g be real-valued random variables on (Ω, \mathcal{F}, P) . Then

(16)
$$E\left[\lambda \vee |g|\right] \le E\left[\lambda \vee |f|\right]$$

for all nonnegative number λ if and only if

$$\int_0^t g^{\#}(s)ds \le \int_0^t f^{\#}(s)ds$$

for all $t \in [0, 1]$.

Proof. Equation (16) is equivalent to $E[\lambda \vee g^{\#}] \leq E[\lambda \vee f^{\#}]$. For the "if" part, let

$$\alpha = \sup \left\{ t : f^{\#}(t) \ge \lambda \right\}$$

$$\beta = \sup \left\{ t : g^{\#}(t) \ge \lambda \right\}.$$

Then

$$E \left[\lambda \vee f^{\#} \right] = \int_{0}^{\alpha} f^{\#} + (1 - \alpha)\lambda$$

$$= \int_{0}^{\beta} f^{\#} + (1 - \beta)\lambda + \int_{\beta}^{\alpha} (f^{\#} - \lambda)$$

$$\geq \int_{0}^{\beta} g^{\#} + (1 - \beta)\lambda + \int_{\beta}^{\alpha} (f^{\#} - \lambda)$$

$$= E \left[\lambda \vee g^{\#} \right] + \int_{\beta}^{\alpha} (f^{\#} - \lambda).$$

If $\alpha \leq \beta$, then for all $x \in (\alpha, \beta)$ we have $f^{\#}(x) \leq \lambda$, and if $\beta \leq \alpha$, then for all $x \in (\beta, \alpha)$ we have $f^{\#}(x) \geq \lambda$. Either way, we see that $\int_{\beta}^{\alpha} (f^{\#} - \lambda) \geq 0$, and the result follows.

To show the "only if", for any $\alpha \in [0, 1]$, let

$$\lambda = f^{\#}(\alpha)$$

$$\beta = \inf \left\{ t : g^{\#}(t) \ge \lambda \right\}.$$

Then

$$\int_0^{\alpha} g^{\#} = \int_0^{\beta} g^{\#} + \int_{\beta}^{\alpha} (g^{\#} - \lambda) + \lambda(1 - \beta) + \lambda(\alpha - 1)$$

$$= E \left[\lambda \vee g^{\#}\right] + \lambda(\alpha - 1) + \int_{\beta}^{\alpha} (g^{\#} - \lambda)$$

$$\leq E \left[\lambda \vee f^{\#}\right] + \lambda(\alpha - 1) + \int_{\beta}^{\alpha} (g^{\#} - \lambda)$$

$$= \int_0^{\alpha} f^{\#} + \int_{\beta}^{\alpha} (g^{\#} - \lambda).$$

Arguing as above, we see that $\int_{\beta}^{\alpha} (g^{\#} - \lambda) \leq 0$, and again the result follows.

Given a random variable f and a sigma field \mathcal{G} , we will say that f is nowhere constant with respect to \mathcal{G} if P(f=g)=0 for every \mathcal{G} measurable function g. The following theorem [13] shows a concrete representation of a sequence of random variables.

Theorem D. Let (f_n) be a sequence of random variables takeing values in a separable sigma filed (S, \mathcal{S}) . Then there exists a sequence of measurable functions $(g_n : [0,1]^n \to S)$ that has the same law as (f_n) . If further we have that f_{n+1} is nowhere constant with respect to $\sigma(f_1, \ldots, f_n)$ for all $n \geq 0$, then we may suppose that $\sigma(g_1, \ldots, g_n) = \mathcal{L}_n$ for all $n \geq 0$.

Proof of Theorem 1. We will prove this theorem under the assumption (6). Consider the map $D_k = (d_k, e_k, f_k) : \Omega \times [0, 1]^{\mathbb{N}} \to \mathbb{R}^3$ by $(\omega, (x_k)) \mapsto (d_k(\omega), e_k(\omega), x_k)$. It is clear that D_k is nowhere constant with respect to $\sigma(D_1, \ldots, D_{k-1})$. Apply the previous theorem to get $\widetilde{D}_k = (\widetilde{d}_k, \widetilde{e}_k, \widetilde{f}_k) : [0, 1]^k \to \mathbb{R}^3$ such that (\widetilde{D}_k) has the same law as (D_k) and $\sigma(\widetilde{D}_1, \ldots, \widetilde{D}_k) = \mathcal{L}_k$.

Next, we show that for almost every x_1, \ldots, x_{k-1} and $\lambda \geq 0$ that

$$\int_0^1 \lambda \vee |\widetilde{e}_k(x_1, \dots, x_k)| \, dx_k \leq \int_0^1 \lambda \vee |\widetilde{d}_k(x_1, \dots, x_k)| \, dx_k$$

which will follow from showing that for any bounded non-negative measurable function $\phi_k : [0,1]^{k-1} \to [0,\infty)$ that

$$E[\phi_k \vee |\widetilde{e}_k|] \leq E[\phi_k \vee |\widetilde{d}_k|].$$

But then there exists a bounded Borel measurable function $\theta_k : \mathbb{R}^{3(k-1)} \to [0, \infty)$ such that $\phi = \theta(\widetilde{D}_1, \dots, \widetilde{D}_{k-1})$ almost everywhere in $[0, 1]^{k-1}$. Thus

$$\int_{[0,1]^k} \phi_k \vee |\widetilde{e}_k| = \int_{[0,1]^k} \theta(\widetilde{D}_1, \dots, \widetilde{D}_{k-1}) \vee |\widetilde{e}_k|$$

$$= E[\theta(D_1, \dots, D_{k-1}) \vee |e_k|]$$

$$\leq E[\theta(D_1, \dots, D_{k-1}) \vee |d_k|]$$

$$= \int_{[0,1]^k} \theta(\widetilde{D}_1, \dots, \widetilde{D}_{k-1}) \vee |\widetilde{d}_k|$$

$$= \int_{[0,1]^k} \phi_k \vee |\widetilde{d}_k|$$

Also to show that $E[\widetilde{d}_k|\mathcal{L}_{k-1}] = E[\widetilde{e}_k|\mathcal{L}_{k-1}] = 0$, it is sufficient to show that for any bounded measurable function $\phi_k : [0,1]^{k-1} \to \mathbb{R}$ that $E[\phi_k \widetilde{d}_k] = E[\phi_k \widetilde{e}_k] = 0$. Thus follows by a very similar argument to that above.

The result then follows from Lemma 10 and Theorem 3. \square

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