

## A Moment Inequality and Its Applications\*

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Marcikewicz-Zygmund inequality plays a very important role in various proofs. However it is well-known that M-Z inequality focus on independent random variables. In this paper we establish a more useful inequality for  $\varphi$ -mixing sequence follows.

**Theorem 1** Let  $X_1, \dots, X_n$  be a sequence of  $B$ -valued random vectors and let  $S_i = \sum_{j=1}^i X_j$ . Suppose there exist integer  $p \geq 1$  and constants  $A > 0$  and  $\varepsilon > 0$  such that

$\varphi_n(p) + \max_{p \leq m \leq n} P(\|S_n - S_m\| \geq A) < \varepsilon$  where  $\varphi_n(p) = \max_{1 \leq k \leq n-p} \varphi(F_1^k, F_{k+p}^n)$ ,  $\varphi(F_1^k, F_{k+p}^n) = \sup_{C \in F_1^k, B \in F_{k+p}^n} |P(B|C) - P(B)|$ ,  $F_a^b = \sigma(x_i, a \leq i \leq b)$ . Then for every  $q > 0$  we have

$$E \max_{1 \leq i \leq n} \|S_i\|^q \leq (1 - \varepsilon - 4^q \varepsilon)^{-1} ((8A)^q + 2(4p)^q E \max_{1 \leq i \leq n} \|X_i\|^q)$$

provided  $4^q \varepsilon < 1 - \varepsilon$ .

**Corollary** Let  $\{X_k, k \geq 1\}$  be a  $\varphi$ -mixing sequence of real-valued random variables and let  $S_k(n) = \sum_{i=k+1}^{k+n} X_i$  ( $k \geq 0, n \geq 1$ ). Suppose there exists a sequence of positive numbers  $\{C_n\}$  such that for every  $k \geq 0, n \geq 1, m \leq n$   $ES_k^2(m) \leq C_n$ . Then for every  $q \geq 2$ , there is a constant  $K$  depending only on  $\varphi(\cdot)$  and  $q$  such that for every  $k \geq 0, n \geq 1$

$$E \max_{1 \leq i \leq n} |S_k(i)|^q \leq K(C_n^{q/2} + E \max_{k < i \leq k+n} |X_i|^q)$$

As applications we have

**Theorem 2** Let  $\{X_n\}$  be a  $\varphi$ -mixing sequence of identically distributed random variables. Then the following conditions are equivalent:

- (i)  $E |X_1|^{r'} < \infty, EX_1 = 0$ ;
- (ii)  $\sum_{n=1}^{\infty} n^{r-2} P(\max_{i \leq n} |S_i| > \varepsilon n^{1/t}) < \infty \quad \varepsilon > 0$
- (iii)  $\sum_{n=1}^{\infty} n^{r-2} P(\sup_{k \geq n} |S_k|/k^{1/t} > \varepsilon) < \infty \quad \varepsilon > 0$

where  $r > 1$  and  $1 \leq t < 2$ . (to 23)

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that the condition is sufficient .

If  $\hat{\sigma} \in L^2(\hat{G}, H)$ , then the mapping

$$T: L^2(G, H) \rightarrow C, f \mapsto \langle \hat{f}, \hat{\sigma} \rangle$$

defines a continuous linear functional on  $L^2(G, H)$  because

$$|Tf| = |\langle \hat{f}, \hat{\sigma} \rangle| \leq \|\hat{f}\|_{L^2} \|\hat{\sigma}\|_{L^2} = \|f\|_{L^2} \|\sigma\|_{L^2}.$$

Thus there is unique  $g \in L^2(G, H)$  such that

$$\langle \hat{f}, \hat{\sigma} \rangle = Tf = \langle f, g \rangle$$

for all  $f \in L^2(G, H)$ , using the Plancherel formula in  $L^2(G, H)$  (see VI.2) and the property of vector-valued pseudomeasure we conclude that for all  $f \in L^2(G, H)$  the equality  $\langle f, \sigma \rangle = \langle (\hat{f})^\wedge, \sigma \rangle = \langle \hat{f}, \hat{\sigma} \rangle = \langle f, g \rangle$  holds. Since  $A_C(G, H)$  is dense in  $L^2(G, H)$ , obviously,  $\langle f, \sigma \rangle = \langle f, g \rangle$  for all  $f \in A_C(G, H)$  By definition in VI.3 we have  $\sigma \in L^2(G, H)$ .

The proof of theorem is complete.

### References

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**Theorem 3** Let  $\{X_n\}$  be a strictly stationary  $\varphi$ -mixing sequence of random variables with  $EX_1 = 0, EX_1^2 < \infty$  and  $\sigma_n^2 = ES_n^2 \rightarrow \infty$  as  $n \rightarrow \infty$ . If  $\sum_{n=1}^{\infty} \varphi^{1/2}(2^n) < \infty$ , then, without changing the distribution of  $\{S(t)\}$  we can redefine  $\{S(t)\}$  on a richer probability space together with a standard Wiener process  $\{W(t), t \geq 0\}$  such that

$$S(t) - W(\sigma_t^2) = o((\sigma_t^2 \log \log t)^{1/2}) \quad \text{a.s.}$$

as  $t \rightarrow \infty$ . Where  $S(t) = \sum_{i \leq t} X_i$  and  $\sigma_t^2 = ES^2(t)$ .