## Lecture 12: Relationship among convergence modes and uniform integrability

**Theorem 1.8.** Let  $X, X_1, X_2, \ldots$  be random k-vectors.

- (i) If  $X_n \to_{a.s.} X$ , then  $X_n \to_p X$ . (The converse is not true.)
- (ii) If  $X_n \to_{L_r} X$  for an r > 0, then  $X_n \to_p X$ . (The converse is not true.)
- (iii) If  $X_n \to_p X$ , then  $X_n \to_d X$ . (The converse is not true.)
- (iv) (Skorohod's theorem). If  $X_n \to_d X$ , then there are random vectors  $Y, Y_1, Y_2, ...$  defined on a common probability space such that  $P_Y = P_X$ ,  $P_{Y_n} = P_{X_n}$ , n = 1, 2, ..., and  $Y_n \to_{a.s.} Y$ . (A useful result; a conditional converse of (i)-(iii).)
- (v) If, for every  $\epsilon > 0$ ,  $\sum_{n=1}^{\infty} P(\|X_n X\| \ge \epsilon) < \infty$ , then  $X_n \to_{a.s.} X$ .
- (A conditional converse of (i):  $P(||X_n X|| \ge \epsilon)$  tends to 0 fast enough.)
- (vi) If  $X_n \to_p X$ , then there is a subsequence  $\{X_{n_j}, j = 1, 2, ...\}$  such that  $X_{n_j} \to_{a.s.} X$  as  $j \to \infty$ . (A partial converse of (i).)
- (vii) If  $X_n \to_d X$  and P(X = c) = 1, where  $c \in \mathcal{R}^k$  is a constant vector, then  $X_n \to_p c$ . (A conditional converse of (i).)
- (viii) Suppose that  $X_n \to_d X$ . Then, for any r > 0,

$$\lim_{n \to \infty} E \|X_n\|_r^r = E \|X\|_r^r < \infty \tag{1}$$

if and only if  $\{\|X_n\|_r^r\}$  is uniformly integrable in the sense that

$$\lim_{t \to \infty} \sup_{n} E\left( \|X_n\|_r^r I_{\{\|X_n\|_r > t\}} \right) = 0.$$
 (2)

(A conditional converse of (ii).)

Discussion on uniform integrability

If there is only one random vector, then (2) is

$$\lim_{t \to \infty} E\left( \|X\|_r^r I_{\{\|X\|_r > t\}} \right) = 0,$$

which is equivalent to the integrability of  $||X||_r^r$  (dominated convergence theorem). Sufficient conditions for uniform integrability:

$$\sup_{r} E \|X_n\|_r^{r+\delta} < \infty \quad \text{for a } \delta > 0$$

This is because

$$\lim_{t \to \infty} \sup_{n} E\left(\|X_n\|_r^r I_{\{\|X_n\|_r > t\}}\right) \le \lim_{t \to \infty} \sup_{n} E\left(\|X_n\|_r^r I_{\{\|X_n\|_r > t\}} \frac{\|X_n\|_r^{\delta}}{t^{\delta}}\right)$$

$$\le \lim_{t \to \infty} \frac{1}{t^{\delta}} \sup_{n} E\left(\|X_n\|_r^{r+\delta}\right)$$

$$= 0$$

Exercises 117-120.

**Proof of Theorem 1.8.** (i) The result follows from Lemma 1.4.

(ii) The result follows from Chebyshev's inequality with  $\varphi(t) = |t|^r$ .

(iii) Assume k = 1. (The general case is proved in the textbook.) Let x be a continuity point of  $F_X$  and  $\epsilon > 0$  be given. Then

$$F_X(x - \epsilon) = P(X \le x - \epsilon)$$

$$\le P(X_n \le x) + P(X \le x - \epsilon, X_n > x)$$

$$\le F_{X_n}(x) + P(|X_n - X| > \epsilon).$$

Letting  $n \to \infty$ , we obtain that

$$F_X(x - \epsilon) \le \liminf_n F_{X_n}(x).$$

Switching  $X_n$  and X in the previous argument, we can show that

$$F_X(x+\epsilon) \ge \limsup_n F_{X_n}(x).$$

Since  $\epsilon$  is arbitrary and  $F_X$  is continuous at x,  $F_X(x) = \lim_{n \to \infty} F_{X_n}(x)$ .

(iv) The proof of this part can be found in Billingsley (1986, pp. 399-402).

(v) Let  $A_n = \{||X_n - X|| \ge \epsilon\}$ . The result follows from Lemma 1.4, Lemma 1.5(i), and Proposition 1.1(iii).

(vi)  $X_n \to_p X$  means  $\lim_{n\to\infty} P(\|X_n - X\| > \epsilon) = 0$  for every  $\epsilon > 0$ .

That is, for every  $\epsilon > 0$ ,  $P(||X_n - X|| > \epsilon) < \epsilon$  for  $n > n_{\epsilon}$  ( $n_{\epsilon}$  is an integer depending on  $\epsilon$ ). For every j = 1, 2, ..., there is a positive integer  $n_j$  such that

$$P(||X_{n_j} - X|| > 2^{-j}) < 2^{-j}.$$

For any  $\epsilon > 0$ , there is a  $k_{\epsilon}$  such that for  $j \geq k_{\epsilon}$ ,  $P(\|X_{n_j} - X\| > \epsilon) < P(\|X_{n_j} - X\| > 2^{-j})$ . Since  $\sum_{j=1}^{\infty} 2^{-j} = 1$ , it follows from the result in (v) that  $X_{n_j} \to_{a.s.} X$  as  $j \to \infty$ .

(vii) The proof for this part is left as an exercise.

(viii) First, by part (iv), we may assume that  $X_n \to_{a.s.} X$  (why?).

Proof of (2) implies (1)

Note that (2) (the uniform integrability of  $\{\|X_n\|_r^r\}$ ) implies that  $\sup_n E\|X_n\|_r^r < \infty$  (why?) By Fatou's lemma (Theorem 1.1(i)),  $E\|X\|_r^r \leq \liminf_n E\|X_n\|_r^r < \infty$ .

Hence, (1) follows if we can show that

$$\lim_{r} \sup_{n} E \|X_n\|_r^r \le E \|X\|_r^r. \tag{3}$$

For any  $\epsilon > 0$  and t > 0, let  $A_n = \{ \|X_n - X\|_r \le \epsilon \}$  and  $B_n = \{ \|X_n\|_r > t \}$ . Then

$$E||X_n||_r^r = E(||X_n||_r^r I_{A_n^c \cap B_n}) + E(||X_n||_r^r I_{A_n^c \cap B_n^c}) + E(||X_n||_r^r I_{A_n})$$

$$\leq E(||X_n||_r^r I_{B_n}) + t^r P(A_n^c) + E||X_n I_{A_n}||_r^r.$$

For  $r \leq 1$ ,  $||X_n I_{A_n}||_r^r \leq (||X_n - X||_r^r + ||X||_r^r)I_{A_n}$  and

$$E||X_n I_{A_n}||_r^r \le E[(||X_n - X||_r^r + ||X||_r^r)I_{A_n}] \le \epsilon^r + E||X||_r^r.$$

For r > 1, an application of Minkowski's inequality leads to

$$E\|X_{n}I_{A_{n}}\|_{r}^{r} = E\|(X_{n} - X)I_{A_{n}} + XI_{A_{n}}\|_{r}^{r}$$

$$\leq E[\|(X_{n} - X)I_{A_{n}}\|_{r} + \|XI_{A_{n}}\|_{r}]^{r}$$

$$\leq \{[E\|(X_{n} - X)I_{A_{n}}\|_{r}^{r}]^{1/r} + [E\|XI_{A_{n}}\|_{r}^{r}]^{1/r}\}^{r}$$

$$\leq \{\epsilon + [E\|X\|_{r}^{r}]^{1/r}\}^{r}.$$

In any case, since  $\epsilon$  is arbitrary,  $\limsup_n E \|X_n I_{A_n}\|_r^r \leq E \|X\|_r^r$ . This result and the previously established inequality imply that

$$\limsup_{n} E \|X_{n}\|_{r}^{r} \leq \limsup_{n} E(\|X_{n}\|_{r}^{r} I_{B_{n}}) + t^{r} \lim_{n \to \infty} P(A_{n}^{c})$$

$$+ \lim_{n} \sup_{n} E \|X_{n} I_{A_{n}}\|_{r}^{r}$$

$$\leq \sup_{n} E(\|X_{n}\|_{r}^{r} I_{\{\|X_{n}\|_{r} > t\}}) + E \|X\|_{r}^{r},$$

since  $P(A_n^c) \to 0$ . Since  $\{||X_n||_r^r\}$  is uniformly integrable, letting  $t \to \infty$  we obtain (3).

Proof of (1) impies (2)

Let  $\xi_n = ||X_n||_r^r I_{B_n^c} - ||X||_r^r I_{B_n^c}$ . Then  $\xi_n \to_{a.s.} 0$  and  $|\xi_n| \le t^r + ||X||_r^r$ , which is integrable. By the dominated convergence theorem,  $E\xi_n \to 0$ ; this and (1) imply that

$$E(||X_n||_r^r I_{B_n}) - E(||X||_r^r I_{B_n}) \to 0.$$

From the definition of  $B_n$ ,  $B_n \subset \{||X_n - X||_r > t/2\} \cup \{||X||_r > t/2\}$ . Since  $E||X||_r^r < \infty$ , it follows from the dominated convergence theorem that

$$\lim_{r \to \infty} E(\|X\|_r^r I_{\{\|X_n - X\|_r > t/2\}}) = 0$$

Hence

$$\limsup_{n} E(\|X_{n}\|_{r}^{r} I_{B_{n}}) \leq \limsup_{n} E(\|X\|_{r}^{r} I_{B_{n}}) \leq E(\|X\|_{r}^{r} I_{\{\|X\|_{r} > t/2\}}).$$

Letting  $t \to \infty$ , it follows from the dominated convergence theorem that

$$\lim_{t \to \infty} \limsup_{n} E(\|X_n\|_r^r I_{B_n}) \le \lim_{t \to \infty} E(\|X\|_r^r I_{\{\|X\|_r > t/2\}}) = 0.$$

This proves (2).