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## **convergence in probability is preserved under continuous transformations**

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**Theorem 1.** *Let  $g: \mathbb{R}^k \rightarrow \mathbb{R}^l$  be a continuous function. If  $\{X_n\}$  are  $\mathbb{R}^k$ -valued random variables converging to  $X$  in probability, then  $\{g(X_n)\}$  converge in probability to  $g(X)$  also.*

*Proof.* Suppose first that  $g$  is uniformly continuous. Given  $\epsilon > 0$ , there is  $\delta > 0$  such that  $\|g(X_n) - g(X)\| < \epsilon$  whenever  $\|X_n - X\| < \delta$ . Therefore,

$$\mathbb{P}(\|g(X_n) - g(X)\| \geq \epsilon) \leq \mathbb{P}(\|X_n - X\| \geq \delta) \rightarrow 0$$

as  $n \rightarrow \infty$ .

Now suppose  $g$  is not necessarily uniformly continuous on  $\mathbb{R}^k$ . But it will be uniformly continuous on any compact set  $\{x \in \mathbb{R}^k: \|x\| \leq m\}$  for  $m \geq 0$ . Consequently, if  $X_n$  and  $X$  are bounded (by  $m$ ), then the proof just given is applicable. Thus we attempt to reduce the general case to the case that  $X_n$  and  $X$  are bounded.

Let

$$f_m(x) = \begin{cases} x, & \|x\| \leq m \\ mx/\|x\|, & \|x\| \geq m \end{cases}$$

Clearly,  $f_m: \mathbb{R}^k \rightarrow \mathbb{R}^k$  is continuous; in fact, it can be verified that  $f_m$  is uniformly continuous on  $\mathbb{R}^k$ . (This is geometrically obvious in the one-dimensional case.)

Set  $X_n^m = f_m(X_n)$  and  $X^m = f_m(X)$ , so that  $X_n^m$  converge to  $X^m$  in probability for each  $m \geq 0$ .

We now show that  $g(X_n)$  converge to  $g(X)$  in probability by a four-step estimate. Let  $\epsilon > 0$  and  $\delta > 0$  be given. For any  $m \geq 0$  (which we will later),

$$\mathbb{P}(\|g(X_n) - g(X)\| \geq \delta) \leq \mathbb{P}(\|g(X_n^m) - g(X^m)\| \geq \delta) + \mathbb{P}(\|X_n\| \geq m) + \mathbb{P}(\|X\| \geq m).$$

Choose  $M$  such that for  $m \geq M$ ,

$$\mathbb{P}(\|X\| \geq m) \leq \mathbb{P}(\|X\| \geq M) < \frac{\epsilon}{4}.$$

(This is possible since  $\lim_{m \rightarrow \infty} \mathbb{P}(\|X\| \geq m) = \mathbb{P}(\bigcap_{m=0}^{\infty} \{\|X\| \geq m\}) = \mathbb{P}(\emptyset) = 0$ .)

In particular, let  $m = M + 1$ . Since  $X_n^m$  converge in probability to  $X^m$  and  $X_n^m, X^m$  are bounded,  $g(X_n^m)$  converge in probability to  $g(X^m)$ . That means for  $n$  large enough,

$$\mathbb{P}(\|g(X_n^m) - g(X^m)\| \geq \delta) < \frac{\epsilon}{4}.$$

Finally, since  $\|X_n\| \leq \|X_n - X\| + \|X\|$ , and  $X_n$  converge to  $X$  in probability, we have

$$\mathbb{P}(\|X_n\| \geq m = M + 1) \leq \mathbb{P}(\|X_n - X\| \geq 1) + \mathbb{P}(\|X\| \geq M) < \frac{\epsilon}{4} + \frac{\epsilon}{4}$$

for large enough  $n$ .

Collecting the previous inequalities together, we have

$$\mathbb{P}(\|g(X_n) - g(X)\| \geq \delta) < \epsilon$$

for large enough  $n$ .

□