

Tues Jan 14

(21)

Continuous mapping If $g: \mathbb{R}^k \rightarrow \mathbb{R}^p$ is continuous at c

$$\underline{X}_n \xrightarrow{p} c \text{ then } g(\underline{X}_n) \xrightarrow{p} g(c)$$

Law of Large numbers

Let $\underline{X}_1, \dots, \underline{X}_n$ be independent random vectors with expected value $\underline{\mu}$.

$$\bar{X}_n \xrightarrow{p} \underline{\mu}$$

Degenerate RV $P(X_n = a_n) = 1$

Theorem If the constants a_1, a_2, \dots satisfy

$\lim_{n \rightarrow \infty} a_n = a$, then $a_n \xrightarrow{p} a$ As degenerate RVs.

Proof Need to show for all $\varepsilon > 0$ $\lim_{n \rightarrow \infty} P\{|a_n - a| < \varepsilon\} = 1$

By def of limit for all $\varepsilon > 0 \exists N$ s.t. if $n > N$ $|a_n - a| < \varepsilon$

There is N such that for all $n > N$

$$|a_n - a| < \varepsilon \Rightarrow P(|a_n - a| < \varepsilon) = 1 \text{ And}$$

$$\lim_{n \rightarrow \infty} P(|a_n - a| < \varepsilon) = \lim_{n \rightarrow \infty} 1 = 1 \quad \square$$

$$S_n^2 = \frac{\sum_{i=1}^n (X_i - \bar{X}_n)^2}{n-1} = \underbrace{\left(\frac{n}{n-1}\right)}_{a_n} \frac{1}{n} \sum_{i=1}^n (X_i - \bar{X}_n)^2 \quad (2.2)$$

$$= \left(\frac{n}{n-1}\right) \cdot \frac{1}{n} \sum_{i=1}^n (X_i^2 - 2X_i\bar{X}_n + \bar{X}_n^2)$$

$$= \left(\frac{n}{n-1}\right) \cdot \frac{1}{n} \left(\sum_{i=1}^n X_i^2 - 2\bar{X}_n \sum_{i=1}^n X_i + \sum_{i=1}^n \bar{X}_n^2 \right)$$

$$= \frac{n}{n-1} \left(\frac{1}{n} \sum_{i=1}^n X_i^2 - 2\bar{X}_n + \frac{n\bar{X}_n^2}{n} \right)$$

$$= \left(\frac{n}{n-1}\right) \left(\frac{1}{n} \sum_{i=1}^n X_i^2 - \bar{X}_n^2 \right)$$

Since $\lim_{n \rightarrow \infty} \frac{n}{n-1} = 1$, $\frac{n}{n-1} \xrightarrow{p} 1$

By LLN $\frac{1}{n} \sum_{i=1}^n X_i^2 \xrightarrow{p} E(X^2) = \sigma^2 + \mu^2$

$$\text{Var}(X) = E(X^2) - (E(X))^2$$

$$\sigma^2 = E(X^2) - \mu^2$$

$$\Rightarrow E(X^2) = \sigma^2 + \mu^2$$

$\bar{X}_n \xrightarrow{p} \mu$ by LLN & so by continuous mapping
 $\bar{X}_n^2 \xrightarrow{p} \mu^2$

Finally, the function $g(x, y, z) = x(y-z)$ is continuous, so

$$S_n^2 = g\left(\frac{n}{n-1}, \frac{1}{n} \sum_{i=1}^n X_i^2, \bar{X}_n^2\right) \rightarrow g(1, \sigma^2 + \mu^2, \mu^2)$$

$$= 1 \cdot (\sigma^2 + \mu^2 - \mu^2) = \sigma^2 \text{ consistent}$$