

ECON 709 - PS 5

Alex von Hafften*

10/11/2020

1. For the following sequences, show $a_n \rightarrow 0$ as $n \rightarrow \infty$:

(a) $a_n = 1/n$

Fix $\varepsilon > 0$. Choose $\bar{n} > \frac{1}{\varepsilon}$. For all $n \geq \bar{n}$,

$$|1/n - 0| = |1/n| = \varepsilon.$$

Thus, $a_n = 1/n \rightarrow 0$ as $n \rightarrow \infty$.

(b) $a_n = \frac{1}{n} \sin\left(\frac{n\pi}{2}\right)$

Fix $\varepsilon > 0$. Notice that $|\sin(x)| \leq 1 \forall x$. Choose $\bar{n} > \frac{1}{\varepsilon}$. For all $n \geq \bar{n}$,

$$\left| \frac{1}{n} \sin\left(\frac{n\pi}{2}\right) - 0 \right| = \left| \frac{1}{n} \sin\left(\frac{n\pi}{2}\right) \right| \leq |1| \left| \frac{1}{n} \right| = \left| \frac{1}{n} \right| \leq \varepsilon$$

Thus, $a_n = \frac{1}{n} \sin\left(\frac{n\pi}{2}\right) \rightarrow 0$ as $n \rightarrow \infty$.

*I worked on this problem set with a study group of Michael Nattinger, Andrew Smith, and Ryan Mather. I also discussed problems with Emily Case, Sarah Bass, and Danny Edgel.

2. Consider a random variable X^n with the probability function

$$X_n = \begin{cases} -n, & \text{with probability } 1/n \\ 0, & \text{with probability } 1 - 2/n \\ n, & \text{with probability } 1/n \end{cases}$$

(a) Does $X_n \rightarrow_p 0$ as $n \rightarrow \infty$?

Fix $\varepsilon > 0$. Choose $\bar{n} > \varepsilon$. For $n \geq \bar{n}$,

$$P(|X_n| \geq \varepsilon) \leq P(|X_n| \geq n) = P(X_n = -n) + P(X_n = n) = 1/n + 1/n = 2/n$$

Since $1/n \rightarrow 0$, $2/n \rightarrow 0$. Thus, $X_n \rightarrow_p 0$ as $n \rightarrow \infty$.

(b) Calculate $E(X_n)$.

$$E(X_n) = \sum_{x \in \text{Supp}(X)} \pi(x)x = (1/n) * (-n) + (1 - 2/n)(0) + (1/n)(n) = -1 + 1 = 0.$$

(c) Calculate $\text{Var}(X_n)$.

$$\text{Var}(X_n) = E(X_n^2) - E(X_n)^2 = E(X_n^2) = \sum_{x \in \text{Supp}(X)} \pi(x)x^2 = (1/n)*(-n)^2 + (1-2/n)(0)^2 + (1/n)(n)^2 = n + n = 2n.$$

(d) Now suppose the distribution is

$$X_n = \begin{cases} 0, & \text{with probability } 1 - 1/n \\ n, & \text{with probability } 1/n \end{cases}$$

Calculate $E(X_n)$.

$$E(X_n) = \sum_{x \in \text{Supp}(X)} \pi(x)x = (1 - 1/n)(0) + (1/n)(n) = 0 + 1 = 1$$

(e) Conclude that $X_n \rightarrow_p 0$ is not sufficient for $E(X_n) \rightarrow 0$.

Fix $\varepsilon > 0$. Choose $\bar{n} > \varepsilon$. For $n > \bar{n}$

$$P(|X_n| \geq \varepsilon) \leq P(|X_n| \geq n) = P(X_n = n) = 1/n$$

Since $1/n \rightarrow 0$, $X_n \rightarrow_p 0$ as $n \rightarrow \infty$. Thus, $X_n \rightarrow_p 0$ is not sufficient for $E(X_n) \rightarrow 0$.

3. A weighted sample mean takes the form $\bar{Y}^* = \frac{1}{n} \sum_{i=1}^n w_i Y_i$ for some non-negative constants w_i satisfying $\frac{1}{n} \sum_{i=1}^n w_i = 1$. Assume that $Y_i : i = 1, \dots, n$ are i.i.d.

(a) Show that \bar{Y}^* is unbiased for $\mu = E(Y_i)$.

$$E(\bar{Y}^*) = E\left(\frac{1}{n} \sum_{i=1}^n w_i Y_i\right) = \frac{1}{n} \sum_{i=1}^n w_i E(Y_i) = \frac{1}{n} \sum_{i=1}^n w_i \mu = (1)\mu = \mu$$

(b) Calculate $Var(\bar{Y}^*)$.

$$Var(\bar{Y}^*) = Var\left(\frac{1}{n} \sum_{i=1}^n w_i Y_i\right) = \frac{1}{n^2} \sum_{i=1}^n w_i^2 Var(Y_i)$$

(c) Show that a sufficient condition for $\bar{Y}^* \rightarrow_p \mu$ is that $\frac{1}{n^2} \sum_{i=1}^n w_i^2 \rightarrow 0$. (Hint: use the Markov's or Chebyshev's Inequality).

Fix $\varepsilon > 0$. Because $\frac{1}{n^2} \sum_{i=1}^n w_i^2 \rightarrow 0$, there exists \bar{n} such that for $n \geq \bar{n}$,

$$\left| \frac{1}{n^2} \sum_{i=1}^n w_i^2 \right| \leq \varepsilon$$

From (a) we know that $E(\bar{Y}^*) = \mu$ and from (b) we know that $Var(\bar{Y}^*) = \frac{1}{n^2} \sum_{i=1}^n w_i^2 Var(Y_i)$, so by Chebyshev's Inequality,

$$P(|\bar{Y}^* - \mu| \geq \lambda) \leq \frac{Var(\bar{Y}^*)}{\lambda^2} = \frac{\frac{1}{n^2} \sum_{i=1}^n w_i^2 Var(Y_i)}{\lambda^2} \leq \frac{\varepsilon Var(Y_i)}{\lambda^2} = \frac{\varepsilon Var(Y_i)}{\left(\sqrt{Var(Y_i)}\right)^2} = \varepsilon$$

where $\lambda = \sqrt{Var(Y_i)}$. Thus $\bar{Y}^* \rightarrow_p \mu$.

(d) Show that the sufficient condition for the condition in part (c) is $\max_{i \leq n} w_i/n \rightarrow 0$.

Fix $\varepsilon > 0$. Let $\delta = \sqrt{\frac{\varepsilon}{n}}$. Because $\max_{i \leq n} w_i/n \rightarrow 0$, there exists a \bar{n} such that for $n \geq \bar{n}$,

$$\begin{aligned} \left| \max_{i \leq n} w_i/n \right| \leq \delta &\implies |w_i/n| \leq \delta \quad \forall i \in \{1, \dots, n\} \\ &\implies (w_i/n)^2 \leq \delta^2 \quad \forall i \in \{1, \dots, n\} \\ &\implies \sum_{i=1}^n \frac{w_i^2}{n^2} \leq n\delta^2 \\ &\implies \sum_{i=1}^n \frac{w_i^2}{n^2} \leq n \left(\sqrt{\frac{\varepsilon}{n}} \right)^2 \\ &\implies \sum_{i=1}^n \frac{w_i^2}{n^2} \leq \varepsilon \end{aligned}$$

Thus, $\frac{1}{n^2} \sum_{i=1}^n w_i^2 \rightarrow 0$.

4. Take a random sample $\{X_1, \dots, X_n\}$. Which statistic converges in probability by the weak law of large numbers and continuous mapping theorem, assuming the moment exists?

(a) $\frac{1}{n} \sum_{i=1}^n X_i^2$

(b) $\frac{1}{n} \sum_{i=1}^n X_i^3$

(c) $\max_{i \leq n} X_i$

(d) $\frac{1}{n} \sum_{i=1}^n X_i^2 - (\frac{1}{n} \sum_{i=1}^n X_i)^2$

(e) $\frac{\sum_{i=1}^n X_i^2}{\sum_{i=1}^n X_i}$ assuming $\mu = E(X_i) > 0$.

(f) $1(\frac{1}{n} \sum_{i=1}^n X_i > 0)$ where

$$1(a) = \begin{cases} 1 & \text{if } a \text{ is true} \\ 0 & \text{if } a \text{ is not true} \end{cases}$$

is called the indicator function of event a .