



UNIVERSITY OF
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Statistical Sciences

DoSS Summer Bootcamp Probability Module 10

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Recap

Learnt in last module:

- Convergence of functions of random variables
 - ▷ Slutsky's theorem
 - ▷ Continuous mapping theorem
- Laws of large numbers
 - ▷ WLLN
 - ▷ SLLN
 - ▷ Glivenko-Cantelli theorem
- Central limit theorem

Outline

- Limit Theorems and Counterexamples
 - ▷ Law of Large Numbers
 - ▷ Monotone Convergence Theorem
 - ▷ Dominated Convergence Theorem
 - ▷ More about CLT

Limit Theorems and Counterexamples

integrable.

Recall: For the law of large numbers to hold, the assumption $E|X| < \infty$ is crucial.

Law of Large Numbers fail for infinite mean i.i.d. random variables

If X_1, X_2, \dots are i.i.d. to X with $E|X_i| = \infty$, then for $S_n = X_1 + \dots + X_n$,
 $P(\lim_{n \rightarrow \infty} S_n/n \in (-\infty, \infty)) = 0$.

Proof: Omitted

Probability of limit of S_n/n exists is zero

$\Leftrightarrow S_n/n$ doesn't converge a.s.

Limit Theorems and Counterexamples

Monotone Convergence Theorem

If $X_n \geq c$ and $X_n \nearrow X$, then $EX_n \nearrow EX$

Usage: Let X_n be $P(X_n = \frac{1}{n^2}) = p = 1 - P(X_n = 0)$

$$\text{Note } 0 \leq X_n \leq \frac{1}{n^2}, \quad EX_n = \frac{p}{n^2}$$

Let $S_n = \sum_{i=1}^n X_i$. Then S_n is monotone increasing since $X_n \geq 0$.

Also, $S_n \geq 0$.

$$\text{Furthermore } S_n \leq \sum_{i=1}^n \frac{1}{i^2} \leq \frac{\pi^2}{6} = \frac{1}{1} + \frac{1}{2^2} + \frac{1}{3^2} + \frac{1}{4^2} + \dots$$

$$\text{Then for } S_n \nearrow \underbrace{\quad}_{\text{r.v.}} \leq \frac{\pi^2}{6}$$

Do $ES_n \rightarrow ES$?

$S_n \geq 0$, $S_n \nearrow S$ so we can use the monotone convergence theorem.

$$\Rightarrow \lim_{n \rightarrow \infty} ES_n = ES.$$

$$ES_n = \sum_{i=1}^n EX_i = \sum_{i=1}^n \frac{p}{c^2} = p \sum_{i=1}^n \frac{1}{c^2}.$$

$$\therefore ES = p \lim_{n \rightarrow \infty} \sum_{i=1}^n \frac{1}{c^2} = p \cdot \frac{\pi^2}{6}.$$

Limit Theorems and Counterexamples

Monotone Convergence Theorem

If $X_n \geq 0$ and $X_n \nearrow X$, then $EX_n \nearrow EX$

Counterexample when X_n is not lower bounded:

$X_0 = 0$ otherwise.

Let X_0 be $P(X_0 = -2^i) = 2^{-i}$ for $i = 1, 2, \dots$

Then X_0 is not lower bounded and $EX_0 = -\infty$, $X_0 < 0$

Let $X_n = n^{-1} X_0$.

Then X_n is monotone increasing since

$$X_{n+1} - X_n = \frac{X_0}{n+1} - \frac{X_0}{n} = -\frac{X_0}{n(n+1)} > 0$$

Furthermore, $\lim_{n \rightarrow \infty} X_n = 0$ since $\lim_{n \rightarrow \infty} X_n = \lim_{n \rightarrow \infty} \frac{X_0}{n} = 0$

However, $E X_n = n^{-1} E X_0 = -\infty$

Then for, $\lim_{n \rightarrow \infty} E X_n = -\infty \neq 0 = E \lim_{n \rightarrow \infty} X_n$

Limit Theorems and Counterexamples

$$E|Y| < \infty$$

Dominated Convergence Theorem

If $X_n \rightarrow X$ a.s. and $|X_n| \leq Y$ a.s. for all n and Y is integrable, then $EX_n \rightarrow EX$

Usage:

X_n is dominated by
integrable Y .

Let $M(t) = E e^{xt}$ moment generating function of X .

Suppose $M(t) < \infty$ for any $t \in [-\epsilon, \epsilon]$.

Then $\left. \frac{d}{dt} M(t) \right|_{t=0} = EX$

(Proof) For $h \in (-\frac{\epsilon}{2}, \frac{\epsilon}{2})$,

$$\frac{M(h) - M(0)}{h} = E \left(\frac{e^{hx} - 1}{h} \right)$$

Note that $\lim_{h \rightarrow 0} \frac{e^{hx} - 1}{h} = X$.

Note that $\left| \frac{e^{hx} - 1}{h} \right| = \left| \frac{hx \cdot e^{\zeta x}}{h} \right|$

by mean-value theorem
where ζ is between

0 and h .

$$= |x| e^{\zeta x}$$

Note that $|u| \leq e^a + e^{-a}$

Therefore,

$$|x| e^{\zeta x} = \frac{\epsilon}{2} \cdot \frac{\epsilon}{2} |x| \cdot e^{\zeta x}$$

$$\leq \frac{\epsilon}{2} \cdot (e^{\frac{\epsilon}{2}x} + e^{-\frac{\epsilon}{2}x}) \cdot e^{\zeta x}$$

$$= \frac{\epsilon}{2} (e^{(3+\frac{\epsilon}{2})x} + e^{(3-\frac{\epsilon}{2})x})$$

Note that $\zeta \pm \frac{\epsilon}{2} \in (-\epsilon, \epsilon)$

$$\leq C \cdot \frac{\epsilon}{2} (e^{\epsilon x} + e^{-\epsilon x})$$

integrable.

Therefore, we can use the dominated convergence theorem to $\frac{e^{hx} - 1}{h}$.

$$\Rightarrow \lim_{h \rightarrow 0} E \frac{e^{hx} - 1}{h} = E \lim_{h \rightarrow 0} \frac{e^{hx} - 1}{h} = EX.$$

$$\lim_{h \rightarrow 0} \frac{M(h) - M(0)}{h}$$

"

$$M'(0)$$

Limit Theorems and Counterexamples

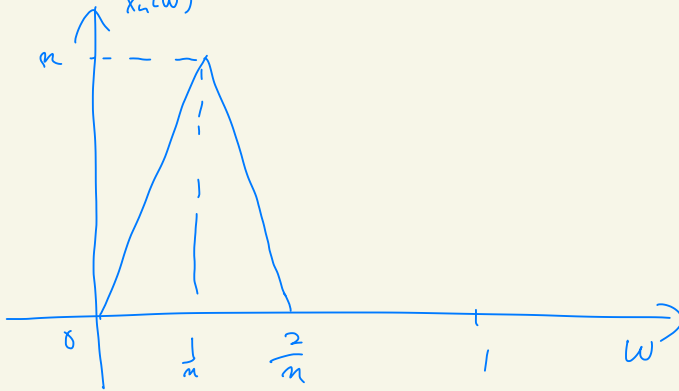
Dominated Convergence Theorem

If $X_n \rightarrow X$ a.s. and $|X_n| \leq Y$ a.s. for all n and Y is integrable, then $EX_n \rightarrow EX$

Counterexample when X_n is not dominated by an integrable random variable:

- 1) the counterexample for monotone convergence theorem.
- 2) Let $\Omega = (0, 1)$ with $\mathbb{P}(w \in (a, b]) = b - a$ if $0 < a < b < 1$
(uniform measure)

Then define $X_n(\omega)$ as follows.



for any $\omega \in (0, 1) = \Omega$, $\lim_{n \rightarrow \infty} X_n(\omega) = 0$

while $E X_n =$ the area of triangle $= \frac{1}{2} \cdot n \cdot \frac{2}{n} = 1$

Therefore $\lim_{n \rightarrow \infty} E X_n = 1 \neq 0 = E \lim_{n \rightarrow \infty} X_n$

Limit Theorems and Counterexamples

More about CLT: Delta method

Suppose X_n are i.i.d. random variables with $EX_n = 0$, $VAR(X_n) = \sigma^2 > 0$. Let g be a measurable function that is differentiable at 0 with $g'(0) \neq 0$. Then

$$\sqrt{n} \left(g \left(\frac{\sum_{k=1}^n X_k}{n} \right) - g(0) \right) \rightarrow N(0, \sigma^2 g'(0)^2) \text{ weakly.}$$

Proof under stronger assumption: Here, we suppose g is continuously differentiable on \mathbb{R} . If you are interested in a general proof refer to Robert Keener's *Theoretical Statistics*.

$$\sqrt{n} (g(\bar{X}) - g(0)) \xrightarrow{d} N(0, \sigma^2 g'(0)^2)$$

By the mean-value theorem, there exists C_n s.t.

$$g(\bar{X}) - g(0) = g'(C_n) \bar{X} \quad \text{where } C_n \text{ is between } 0 \text{ and } \bar{X}.$$

By the strong law of large numbers, $\lim_{n \rightarrow \infty} \bar{X} = 0$ a.s.

Since C_n is between 0 and \bar{X} , $\lim_{n \rightarrow \infty} C_n = 0$ a.s.

By the continuity of g' , $\lim_{n \rightarrow \infty} g'(C_n) = g'(0)$ almost surely.

$$\sqrt{n} (g(\bar{X}) - g(0)) = \underbrace{g'(C_n)}_{\substack{\rightarrow g'(0) \\ \text{a.s.}}} \cdot \underbrace{\sqrt{n} \bar{X}}_{\substack{\rightarrow N(0, \sigma^2) \\ \text{by CLT}}}$$

by Slutsky's theorem.

$$\xrightarrow{d} N(0, \sigma^2 g'(0)^2)$$