

Convergence of random variables

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Last time: Class activity

Suppose that $X_1, X_2, \dots \stackrel{iid}{\sim} \text{Uniform}(0, 1)$, and let $X_{(n)} = \max\{X_1, \dots, X_n\}$. Then $X_{(n)} \xrightarrow{P} 1$.

PF : wts $\forall \varepsilon > 0$, $P(|X_{(n)} - 1| \geq \varepsilon) \rightarrow 0$ as $n \rightarrow \infty$

Let $0 < \varepsilon < 1$

$$\begin{aligned} P(|X_{(n)} - 1| \geq \varepsilon) &= P(X_{(n)} \leq 1 - \varepsilon) + \underbrace{P(X_{(n)} \geq 1 + \varepsilon)}_{=0 \text{ (} X_{(n)} \leq 1 \text{)}} \\ &= P(X_{(n)} \leq 1 - \varepsilon) \\ &= P(X_1 \leq 1 - \varepsilon, X_2 \leq 1 - \varepsilon, \dots, X_n \leq 1 - \varepsilon) \\ &= P(X_1 \leq 1 - \varepsilon) P(X_2 \leq 1 - \varepsilon) \dots P(X_n \leq 1 - \varepsilon) \\ &= (1 - \varepsilon)^n \\ &\rightarrow 0 \quad \text{as } n \rightarrow \infty \end{aligned}$$

if $\varepsilon \geq 1$

$$P(|X_{(n)} - 1| \geq \varepsilon) \leq P(|X_{(n)} - 1| \geq 1) \rightarrow 0$$

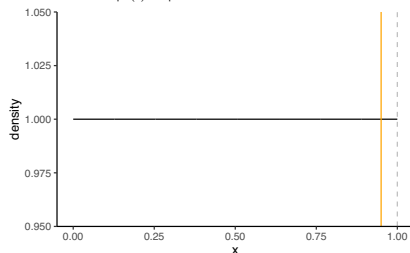
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Last time: Class activity

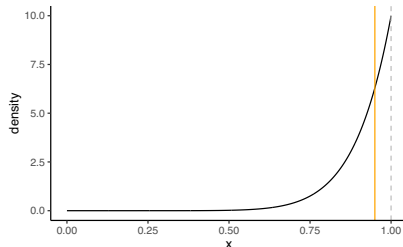
pdf: $\sim x^{n-1}$

Suppose that $X_1, X_2, \dots \stackrel{iid}{\sim} \text{Uniform}(0, 1)$. Then $X_{(n)} \sim \text{Beta}(n, 1)$

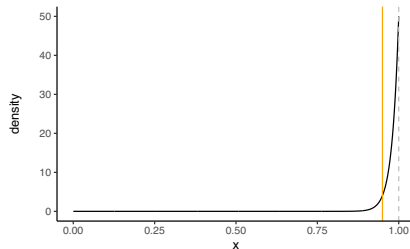
$n = 1, P(|X_{(n)} - 1| \geq 0.05) = 0.95$



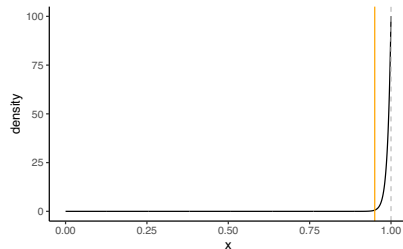
$n = 10, P(|X_{(n)} - 1| \geq 0.05) = 0.6$



$n = 50, P(|X_{(n)} - 1| \geq 0.05) = 0.077$



$n = 100, P(|X_{(n)} - 1| \geq 0.05) = 0.006$



Warmup

Work on the warmup activity (handout), then we will discuss as a class.

Warmup

Suppose that $X_1, X_2, \dots \stackrel{iid}{\sim} \text{Uniform}(0, 1)$, $X_{(n)} = \max\{X_1, \dots, X_n\}$, and let $Y_n = n(1 - X_{(n)})$.

$$\begin{aligned} F_{Y_n}(t) &= P(Y_n \leq t) = P(n(1 - X_{(n)}) \leq t) \\ &= P(1 - X_{(n)} \leq \frac{t}{n}) \\ &= P(X_{(n)} \geq 1 - \frac{t}{n}) \\ &= 1 - P(X_{(n)} \leq 1 - \frac{t}{n}) \\ &= 1 - (P(X_i \leq 1 - \frac{t}{n}))^n \\ &= 1 - (1 - \frac{t}{n})^n \end{aligned}$$

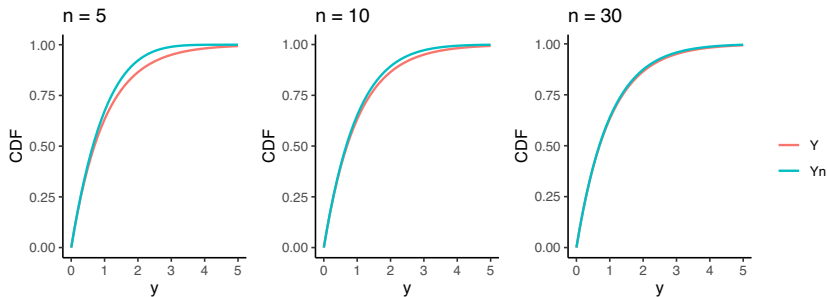
$$\begin{aligned} \lim_{n \rightarrow \infty} F_{Y_n}(t) &= \lim_{n \rightarrow \infty} 1 - (1 - \frac{t}{n})^n \\ &= 1 - e^{-t} \end{aligned}$$

$$(1 + \frac{x}{n})^n \rightarrow e^x$$

(Weibull distribution)

Warmup

$$F_{Y_n}(t) = 1 - \left(1 - \frac{t}{n}\right)^n \rightarrow F_Y(t) = 1 - e^{-t}$$



Convergence in distribution

Definition: A sequence of random variables X_1, X_2, \dots *converges in distribution* to a random variable X if

$$\lim_{n \rightarrow \infty} F_{X_n}(x) = F_X(x)$$

at all points where $F_X(x)$ is continuous. We write $X_n \xrightarrow{d} X$.

Class activity

Work on the class activity (handout), then we will discuss as a class.

Class activity

Suppose that X_1, X_2, \dots are iid random variables with cdf

$$F(x) = \begin{cases} 1 - \left(\frac{1}{x}\right)^\alpha & x \geq 1 \\ 0 & x < 1 \end{cases}$$

Let $X_{(n)} = \max\{X_1, \dots, X_n\}$, and $Y_n = n^{-1/\alpha} X_{(n)}$.

$$\begin{aligned} F_{Y_n}(t) &= P(n^{-1/\alpha} X_{(n)} \leq t) = P(X_{(n)} \leq n^{\frac{1}{\alpha}} t) \\ &= \left(1 - \left(n^{\frac{1}{\alpha}} t\right)^{-\alpha}\right)^n \\ &= \left(1 - \frac{1}{t^\alpha n}\right)^n \\ &= \left(1 - \frac{t^{-\alpha}}{n}\right)^n \\ &\rightarrow e^{-t^{-\alpha}} \end{aligned}$$

(Frechet
distribution)

Convergence in distribution: Central Limit Theorem

Let X_1, X_2, \dots be iid random variables, with $\mu = \mathbb{E}[X_i]$ and $0 < \sigma^2 = \text{Var}(X_i) < \infty$. Then

$$\frac{\sqrt{n}(\bar{X}_n - \mu)}{\sigma} \xrightarrow{d} Z$$

where $Z \sim N(0, 1)$.

what if we don't know σ^2 ?

$$\hat{\sigma} = \sqrt{\frac{1}{n} \sum_{i=1}^n (X_i - \bar{X})^2}$$

Does
$$\frac{\sqrt{n}(\bar{X}_n - \mu)}{\hat{\sigma}} \xrightarrow{d} Z \quad ?$$

Yes!

Key results:

$$\hat{\sigma}^2 \xrightarrow{P} \sigma^2$$

Continuous

mapping

$$\begin{aligned} X_n &\xrightarrow{P} X \\ X_n &\xrightarrow{P} X \end{aligned}$$

(HW)

theorem: if g is continuous,

$$\begin{aligned} X_n &\xrightarrow{P} X \Rightarrow g(X_n) \xrightarrow{P} g(X) \\ X_n &\xrightarrow{P} X \Rightarrow g(X_n) \xrightarrow{P} g(X) \end{aligned}$$