



Stochastic Processes

Interpretation of CLT, What CLT is Not, Local CLT

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Weak Law of Large Numbers (WLLN)

Fix a probability space $(\Omega, \mathcal{F}, \mathbb{P})$. Let $\{X_n\}_{n=1}^\infty$ be defined w.r.t. \mathcal{F} .

Theorem (Weak Law of Large Numbers)

Let $\{X_n\}_{n=1}^\infty$ be **i.i.d.** with $\mathbb{E}[|X_1|] < +\infty$. Further, let $\mathbb{E}[X_1] = \mu$. Let

$$S_n = \sum_{i=1}^n X_i.$$

Then,

$$\frac{S_n}{n} \xrightarrow{\text{p.}} \mu.$$

More formally, for every $\varepsilon > 0$,

$$\lim_{n \rightarrow \infty} \mathbb{P} \left(\left| \frac{S_n}{n} - \mu \right| > \varepsilon \right) = 0.$$

Strong Law of Large Numbers (SLLN)

Fix a probability space $(\Omega, \mathcal{F}, \mathbb{P})$. Let $\{X_n\}_{n=1}^\infty$ be defined w.r.t. \mathcal{F} .

Theorem (Strong Law of Large Numbers)

Let $\{X_n\}_{n=1}^\infty$ be **i.i.d.** with $\mathbb{E}[|X_1|] < +\infty$. Further, let $\mathbb{E}[X_1] = \mu$. Let

$$S_n = \sum_{i=1}^n X_i.$$

Then,

$$\frac{S_n}{n} \xrightarrow{\text{a.s.}} \mu.$$

More formally,

$$\mathbb{P} \left(\lim_{n \rightarrow \infty} \frac{S_n}{n} = \mu \right) = 1.$$

Central Limit Theorem (CLT)

Fix a probability space $(\Omega, \mathcal{F}, \mathbb{P})$. Let $\{X_n\}_{n=1}^\infty$ be defined w.r.t. \mathcal{F} .

Theorem (Central Limit Theorem)

Let $\{X_n\}_{n=1}^\infty$ be **i.i.d.** with mean $\mathbb{E}[X_1] = \mu \in \mathbb{R}$ and $\text{Var}(X_1) = \sigma^2 < +\infty$. Let $S_n = \sum_{i=1}^n X_i$. Then,

$$\frac{S_n - \mathbb{E}[S_n]}{\sqrt{\text{Var}(S_n)}} = \frac{S_n - n\mu}{\sigma \sqrt{n}} \xrightarrow{d} X, \quad X \sim \mathcal{N}(0, 1).$$

More formally,

$$\lim_{n \rightarrow \infty} \mathbb{P} \left(\frac{S_n - \mathbb{E}[S_n]}{\sqrt{\text{Var}(S_n)}} \leq x \right) = \int_{-\infty}^x \frac{1}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt \quad \forall x \in \mathbb{R}.$$

Proof of CLT

Let

$$Z_i = \frac{X_i - \mathbb{E}[X_i]}{\sqrt{\text{Var}(X_1)}}, \quad U_n = \frac{\sum_{i=1}^n Z_i}{\sqrt{n}}$$

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$$C_{U_n}(s) = \left(C_{Z_1} \left(\frac{s}{\sqrt{n}} \right) \right)^n = \left(1 - \frac{s^2}{2n} + o \left(\frac{s^2}{n} \right) \right)^n$$

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$$\begin{aligned} \lim_{n \rightarrow \infty} C_{U_n}(s) &= \lim_{n \rightarrow \infty} \left(1 - \frac{s^2}{2n} + o \left(\frac{s^2}{n} \right) \right)^n \\ &= e^{\frac{-s^2}{2}} \end{aligned}$$

Some Notes about CLT

- From SLLN, we know that

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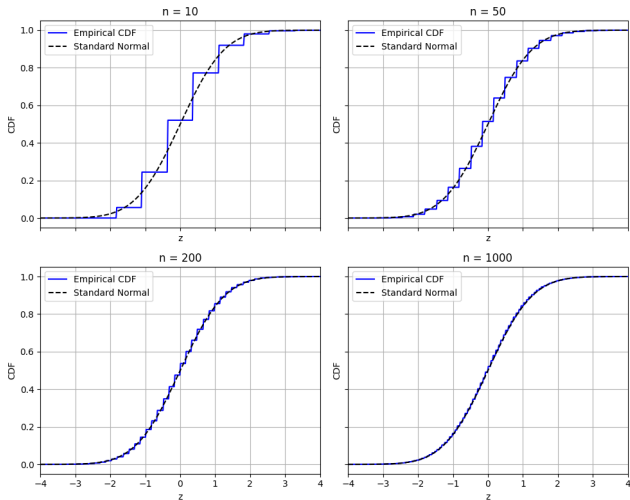
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- CLT: the distribution of $S_n - \mathbb{E}[S_n]$, when **divided by \sqrt{n}** , is non-degenerate for large n
- According to CLT, for large n ,

$$\mathbb{P}\left(\frac{S_n - \mathbb{E}[S_n]}{\sigma \sqrt{n}} > t\right) \approx \mathbb{P}(X > t), \quad X \sim \mathcal{N}(0, 1)$$

Demonstration of CLT



What CLT is Not

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- If X_1, X_2, \dots are discrete random variables, then $\frac{S_n - n\mu}{\sigma\sqrt{n}}$ is a discrete random variable, and hence does not admit any PDF.
- Even if X_1, X_2, \dots are continuous random variables, and $\frac{S_n - n\mu}{\sigma\sqrt{n}}$ admits a PDF, CLT does not make any claim about the convergence of these PDFs to the Gaussian PDF

Example

Suppose that $X_1, X_2, \dots \stackrel{\text{i.i.d.}}{\sim} \text{Unif}([-\sqrt{3}, +\sqrt{3}])$

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Example

Suppose that $X_1, X_2, \dots \stackrel{\text{i.i.d.}}{\sim} \text{Unif}([- \sqrt{3}, + \sqrt{3}])$

- $\mathbb{E}[X_1] = 0$
- $\text{Var}(X_1) = 1$
- $C_{X_1}(s) =$

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- $\mathbb{E}[X_1] = 0$
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- $C_{X_1}(s) = \frac{\sin(s\sqrt{3})}{s\sqrt{3}}, \quad s \in \mathbb{R}.$

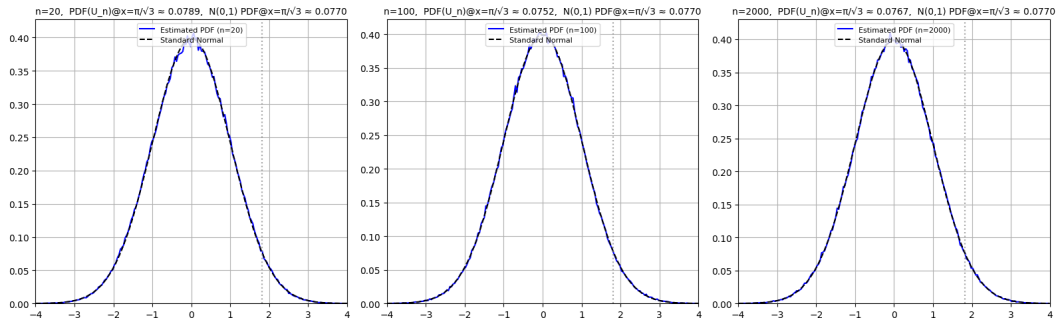
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- $\mathbb{E}[X_1] = 0$
- $\text{Var}(X_1) = 1$
- $C_{X_1}(s) = \frac{\sin(s\sqrt{3})}{s\sqrt{3}}, \quad s \in \mathbb{R}.$
- As per CLT,

$$C_{\frac{s_n}{\sqrt{n}}}(s) \xrightarrow{n \rightarrow \infty} e^{-\frac{s^2}{2}} \quad \forall s \in \mathbb{R}.$$

No Convergence of PDFs for Previous Example



Local CLT

- In many practical examples where X_1, X_2, \dots are continuous, one may observe convergence of PDFs of $\frac{S_n - n\mu}{\sigma\sqrt{n}}$ to the Gaussian PDF, but this is NOT to be interpreted as a consequence of the CLT

This may be a consequence of some stronger property playing in hindsight

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Theorem (Local Central Limit Theorem)

Suppose that $X_1, X_2, \dots \stackrel{\text{i.i.d.}}{\sim} f_X$. W.l.o.g., let $\mathbb{E}[X_1] = 0$ and $\text{Var}(X_1) = 1$.

Suppose that there exists $r \in \mathbb{N}$ such that

$$\int_{-\infty}^{\infty} |C_{X_1}(s)|^r ds < +\infty.$$

Then,

$$f_{\frac{S_n}{\sqrt{n}}}(x) \xrightarrow{n \rightarrow \infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} \quad \forall x \in \mathbb{R}.$$