

# Limit theorems

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Probability - MA 416

Mostly taken from *A first course in probability*  
by S. Ross

# Outline

- 1 Weak law of large numbers
- 2 Central limit theorem

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# Weak law of large numbers

## Theorem 1.

Consider *independent and identically distributed*

- A sequence  $\{X_i; i \geq 1\}$  of *i.i.d* random variables
- Write  $\mathbf{E}[X_i] = \mu$
- Set

*empirical mean*

$$\bar{X}_n = \frac{1}{n} \sum_{i=1}^n X_i$$

Then for any  $\varepsilon > 0$  we have

$$\lim_{n \rightarrow \infty} \mathbf{P}(|\bar{X}_n - \mu| > \varepsilon) = 0$$

*As  $n \rightarrow \infty$ , empirical mean is close to the theoretical mean*

Application We flip a fair coin  
 $n = 400$  times. We let

$$X_i = \mathbb{1}(\text{i-th flip is H})$$

Hyp:  $X_i$ 's are iid with

$$X_i \sim \mathcal{B}(p = \frac{1}{2}), \quad \mathbb{E}[X_i] = p = \frac{1}{2}$$

Then LLN says

$$\mathbb{P}\left( \left| \frac{1}{400} \sum_{i=1}^{400} X_i - \frac{1}{2} \right| > 0.05 \right)$$

is "small"  $\rightarrow$  Pb: Not a very precise statement

With CLT: We will get statements like

$$P\left(\sum_{i=1}^{400} X_i > 210\right) \approx \text{a probability that can be computed}$$

# Outline

1 Weak law of large numbers

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*→  $\equiv$  important*

# DeMoivre-Laplace theorem (repeated)

## Theorem 2.

Let

- $n \geq 1, p \in (0, 1)$
- $X_n \sim \text{Bin}(n, p)$
- $a < b$

Rmk :  $X_n = \sum_{i=1}^n Y_i$ , with  
 $Y_i$  i.i.d with  $Y_i \sim \mathcal{B}(p)$

Then

$$\lim_{n \rightarrow \infty} \mathbf{P} \left( a < \frac{X_n - \widehat{np}}{(np(1-p))^{1/2}} < b \right) = \Phi(b) - \Phi(a)$$

$= \mathbb{E}[X_n]$

$\downarrow$   $\text{Var}(X_n)$



# Another way to write De Moivre's theorem

## Theorem 3.

Consider

- A sequence  $\{Y_i; i \geq 1\}$  of indep.  $\mathcal{B}(p)$  random variables
- We have  $\mathbf{E}[Y_i] = p$  and  $\mathbf{Var}(Y_i) = p(1 - p) \equiv \sigma^2$
- Set

$$\bar{Y}_n = \frac{1}{n} \sum_{i=1}^n Y_i = \frac{X_n}{n}$$

Then for any  $\varepsilon > 0$  we have

$$\lim_{n \rightarrow \infty} \mathbf{P} \left( a < \sqrt{n} \left( \frac{\bar{Y}_n - p}{\sigma} \right) < b \right) = \Phi(b) - \Phi(a)$$

# Central limit theorem

## Theorem 4.

Consider

- A sequence  $\{X_i; i \geq 1\}$  of i.i.d random variables
- Write  $\mathbf{E}[X_i] = \mu$  and  $\mathbf{Var}(X_i) = \sigma^2$
- Set

$$\bar{X}_n = \frac{1}{n} \sum_{i=1}^n X_i$$

Then for any  $\varepsilon > 0$  we have

$$\lim_{n \rightarrow \infty} \mathbf{P} \left( a < \sqrt{n} \left( \frac{\bar{X}_n - \mu}{\sigma} \right) < b \right) = \Phi(b) - \Phi(a)$$

Interpretation of CLT. As  $n \rightarrow \infty$

$$\sqrt{n} \left( \frac{\bar{X}_n - \mu}{\sigma} \right) \approx z, \quad z \sim \mathcal{U}(0,1)$$

$$\Rightarrow \bar{X}_n - \mu \approx \frac{\sigma z}{\sqrt{n}}$$

$$\Rightarrow \bar{X}_n \approx \mu + \frac{\sigma z}{\sqrt{n}}$$

$\bar{X}_n = \mathbb{E}[X_i] +$  Gauss. fluctuations of order  $\frac{1}{\sqrt{n}}$

## Problem 8.5 (1)

**Situation:** We have

- Fifty numbers rounded off to the nearest integer and then summed
- The individual round-off errors are uniformly distributed over  $(-0.5, 0.5)$

**Question:**

Approximate the probability that the resultant sum differs from the exact sum by more than 3.

Model For  $i=1, \dots, n$ ,  $n=50$  let

$X_i = i$ -th error

Hyp  $X_i$ 's are iid,  $X_i \sim U([-0.5, 0.5])$

Recall: for  $Y \sim U([a, b])$  then

$$f_Y(y) = \frac{1}{b-a} \mathbb{1}_{[a,b]}(y)$$

$$\mathbb{E}[Y] = \frac{a+b}{2} \quad V(Y) = \frac{(b-a)^2}{12}$$

Here

$$\mathbb{E}[X_i] = 0 \quad V(Y) = \frac{1}{12} = \sigma^2 \\ = \mu$$

We wish to compute

$$n=50$$

$$\mu=0$$
$$\sigma = \frac{1}{\sqrt{2}}$$

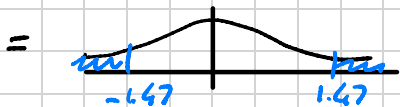
$$\begin{aligned} & \mathbb{P}\left(\left|\sum_{i=1}^n X_i\right| > 3\right) \text{ in CLT: } \sqrt{n} \left(\frac{\bar{X}_n - \mu}{\sigma}\right) \\ &= 1 - \mathbb{P}\left(-3 \leq \sum_{i=1}^n X_i \leq 3\right) \\ &= 1 - \mathbb{P}\left(-3/n \leq \bar{X}_n \leq 3/n\right) \\ &= 1 - \mathbb{P}\left(\frac{(-3/n - \mu)\sqrt{n}}{\sigma} \leq \sqrt{n} \left(\frac{\bar{X}_n - \mu}{\sigma}\right) \leq \frac{(3/n - \mu)\sqrt{n}}{\sigma}\right) \\ &\stackrel{\text{CLT}}{\approx} 1 - \mathbb{P}\left(\frac{-3}{\sqrt{n}\sigma} \leq z \leq \frac{3}{\sqrt{n}\sigma}\right) \\ &= 1 - \mathbb{P}\left(\frac{-3\sqrt{2}}{\sqrt{50}} \leq z \leq \frac{3\sqrt{2}}{\sqrt{50}}\right) \\ &= 1 - \mathbb{P}(-1.47 \leq z \leq 1.47) \end{aligned}$$

Summary : We have found

$$P\left(\left|\sum_{i=1}^n X_i\right| > 3\right)$$

$$\approx 1 - P(-1.47 \leq Z \leq 1.47)$$

$$= P(|Z| > 1.47)$$



$$= 2(1 - \Phi(1.47))$$

$$= 0.14$$

## Problem 8.5 (2)

Model: Set

$X_i \equiv i$ -th error

We wish to find

$$\mathbf{P} \left( \left| \sum_{i=1}^n X_i \right| > 3 \right)$$

Law of  $X_i$ : We have

- $X_i$ 's i.i.d
- $X_i \sim \mathcal{U}([-0.5, 0.5])$
- $\mathbf{E}[X_i] = 0$  , and  $\mathbf{Var}(X_i) = \sigma^2 = \frac{1}{12}$



## Problem 8.5 (3)

Application of CLT: Write

$$\begin{aligned} \mathbf{P} \left( \left| \sum_{i=1}^n X_i \right| > 3 \right) &= \mathbf{P} \left( \left| \sqrt{n} \bar{X}_n \right| > \frac{3}{\sqrt{50}} \right) \\ &= \mathbf{P} \left( \left| \sqrt{n} \frac{\bar{X}_n}{\sigma} \right| > \frac{3\sqrt{12}}{\sqrt{50}} \right) \\ &\stackrel{CLT}{\approx} \mathbf{P} (|Z| > 1.47) \\ &= .14 \end{aligned}$$