

## Review

• The central limit Thm

Let  $X_1, \dots, X_n, \dots$  be an i.i.d. sequence of r.v.'s, each having finite mean  $\mu$  and variance  $\sigma^2$ .

Then  $\forall a \in \mathbb{R}$ ,

$$P\left\{ \frac{X_1 + \dots + X_n - n\mu}{\sqrt{n}\sigma} \leq a \right\} \rightarrow \Phi(a) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^a e^{-\frac{x^2}{2}} dx,$$

as  $n \rightarrow \infty$ .

Example 1 If 10 fair dice are rolled, find the approximate prob. that the sum obtained is between 30 and 40.

Solution: Let  $X_i$  be the value obtained in the  $i$ -th roll,  $i=1, 2, \dots, 10$ .

We need to calculate

$$P\left\{ 29.5 \leq X_1 + \dots + X_{10} \leq 40.5 \right\}$$

Notice that  $\mu = E[X_i] = \frac{1}{6}(1+2+\dots+6)$   
 $= 7/2$

$$E[X_i^2] = \frac{1}{6} (1^2 + 2^2 + \dots + 6^2)$$

$$= \frac{1}{6} \cdot \frac{6 \times 7 \times 13}{6}$$

$$\sigma^2 = \text{Var}(X_i) = E[X_i^2] - E[X_i]^2$$

$$= \frac{35}{12}$$

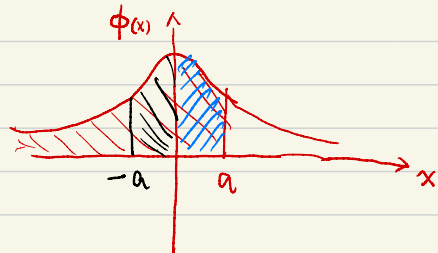
Hence  $P\{29.5 \leq X_1 + \dots + X_{10} \leq 40.5\}$

$$= P\left\{ \frac{29.5 - 10 \times \frac{7}{2}}{\sqrt{350/12}} \leq \frac{X_1 + \dots + X_{10} - 10 \times \frac{7}{2}}{\sqrt{10} \cdot \sqrt{\frac{35}{12}}} \leq \frac{40.5 - 10 \times \frac{7}{2}}{\sqrt{350/12}} \right\}$$

$$\approx P\{-1.018 \leq Z \leq 1.018\}$$

$$= 2 \cdot \Phi(1.018) - 1$$

$$\approx 0.692$$



Thm 2 (The strong law of large numbers).

Let  $X_1, \dots, X_n, \dots$ , be an i.i.d. sequence of r.v.'s with a finite mean  $\mu$ . Then with prob. 1,

$$\frac{X_1 + \dots + X_n}{n} \rightarrow \mu \quad \text{as } n \rightarrow \infty.$$

In other word,

$$P\left\{ \lim_{n \rightarrow \infty} \frac{X_1 + \dots + X_n}{n} = \mu \right\} = 1.$$

The proof of Thm 2 is based on two Lemmas.

Lem 3. Assume  $X$  is a non-negative r.v. which may take the value  $+\infty$ . Suppose  $E[X] < \infty$ .

$$\text{Then } P\{X < \infty\} = 1.$$

Pf. For any positive integer  $n$ , by the Markov inequality,

$$P\{X = \infty\} \leq P\{X \geq n\} \leq \frac{E[X]}{n}.$$

Letting  $n \rightarrow \infty$  gives  $P\{X = \infty\} = 0$ .

$$\text{Hence } P\{X < \infty\} = 1 - P\{X = \infty\} = 1. \quad \square$$

Lem 4. Let  $X$  be a r.v. Then

$$E[X^4] \geq (E[X^2])^2.$$

pf. Notice that

$$\text{Var}(X^2) = E[X^4] - (E[X^2])^2.$$

Since  $\text{Var}(X^2) \geq 0$ , it follows that  $E[X^4] \geq (E[X^2])^2$ .  $\square$

Pf of Thm 2.

We will prove the thm under an additional assumption

$$E[X_i^4] = K < \infty.$$

WLOG, assume  $\mu = 0$ .

$$\text{Write } S_n = X_1 + \dots + X_n.$$

We will estimate

$$E[S_n^4] = E[(X_1 + \dots + X_n)^4].$$

Expand  $(X_1 + \dots + X_n)^4$  in terms of

$$X_i^4, X_i^3 X_j, X_i^2 X_j^2, X_i^2 X_j X_k, X_i X_j X_k X_l$$

with distinct  $i, j, k, l$ .

Notice that  $E[X_i^3 X_j] = E[X_i^3] E[X_j] = 0$ .

$$E[X_i^2 X_j X_k] = E[X_i^2] E[X_j] E[X_k] = 0.$$

$$E[X_i X_j X_k X_l] = 0.$$

Hence

$$E[S_n^4] = E[(X_1 + \dots + X_n)^4]$$

$$= n E[X_i^4]$$

$$+ \binom{n}{2} \binom{4}{2} E[X_i^2 X_j^2]$$

$$= n E[X_i^4] + 6 \binom{n}{2} E[X_i^2] E[X_j^2]$$

$$\leq n E[X_i^4] + 6 \binom{n}{2} E[X_i^4] \quad (\text{Using Lem 4})$$

$$= (3n^2 - 2n) K$$

$$\leq 3n^2 K, \quad \text{where } K = E[X_i^4] < \infty.$$

$$\text{So } E\left[\frac{S_n^4}{n^4}\right] \leq \frac{3K}{n^2}.$$

Hence

$$\sum_{n=1}^{\infty} E\left[\left(\frac{S_n}{n}\right)^4\right] \leq \sum_{n=1}^{\infty} \frac{3K}{n^2} < \infty.$$

$$\text{Thus } E\left[\sum_{n=1}^{\infty} \left(\frac{S_n}{n}\right)^4\right] = \sum_{n=1}^{\infty} E\left[\left(\frac{S_n}{n}\right)^4\right] < \infty.$$

Let  $X = \sum_{n=1}^{\infty} \left(\frac{S_n}{n}\right)^4$ . Then  $X$  is a r.v, non-negative  
(may take the value  $\infty$ )

However  $E[X] < \infty$

By Lem 3,  $P\{X < \infty\} = 1$ .

Hence  $P\left\{\sum_{n=1}^{\infty} \left(\frac{S_n}{n}\right)^4 < \infty\right\} = 1$

However  $\sum_{n=1}^{\infty} \left(\frac{S_n}{n}\right)^4 < \infty \Rightarrow \frac{S_n}{n} \rightarrow 0$  as  $n \rightarrow \infty$ .

Hence  $P \left\{ \lim_{n \rightarrow \infty} \frac{S_n}{n} = 0 \right\} = 1$

Thus with Prob. 1,

$$\frac{S_n}{n} = \frac{X_1 + \dots + X_n}{n} \rightarrow 0.$$

If  $\mu \neq 0$ , then letting  $\widetilde{X}_n = X_n - \mu$   
applying the SLLN to  $(\widetilde{X}_n)$  gives

$$\frac{\widetilde{X}_1 + \dots + \widetilde{X}_n}{n} \rightarrow 0 \quad \text{almost sure.}$$

$$\Leftrightarrow \frac{X_1 + \dots + X_n}{n} \rightarrow \mu \quad \text{almost sure.}$$

