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SLLN :  $\frac{S_n}{n} \rightarrow \mu$  w/ prob. 1, That is

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

Proof Suppose  $\mu=0$  (Redefine  $X_i$  by  $X_i - \mu$ ) and

$E[X_i^4] < \infty$  (can be removed by "truncation argument")

Borel-Cantelli Lemma:  $P\left(\lim_{n \rightarrow \infty} Y_n = Y\right) = 1$

$$\Leftrightarrow P\left(\lim_{n \rightarrow \infty} Y_n \neq Y\right) = 0$$

$$\Leftrightarrow \text{For } \varepsilon > 0, P(|Y_n - Y| > \varepsilon \text{ infinitely many } n) = 0$$

$$\Leftrightarrow \text{For } \varepsilon > 0, P(\forall m, \exists n \geq m \text{ s.t. } |Y_n - Y| > \varepsilon) = 0.$$

$$\Leftrightarrow \forall \varepsilon > 0, P\left(\bigcap_{m=1}^{\infty} \bigcup_{n \geq m} \{|Y_n - Y| > \varepsilon\}\right) = 0$$

$$\Leftrightarrow \forall \varepsilon > 0, \lim_{m \rightarrow \infty} P\left(\bigcup_{n \geq m} \{|Y_n - Y| > \varepsilon\}\right) = 0$$

If  $\sum_{m=1}^{\infty} P(|Y_m - Y| > \varepsilon) < \infty$  then

$$P\left(\bigcup_{n \geq m} \{|Y_n - Y| > \varepsilon\}\right) \leq \sum_{n=m}^{\infty} P(|Y_n - Y| > \varepsilon) \rightarrow 0$$

By this, it suffices to show that

$$\sum_{n=1}^{\infty} P\left(\left|\frac{S_n}{n}\right| > \varepsilon\right) < \infty$$

To this end,  $P\left(\left|\frac{S_n}{n}\right| > \varepsilon\right) \leq \frac{E S_n^4}{n^4 \varepsilon^4} \leq \frac{C}{n^2 \varepsilon^4}$ . □

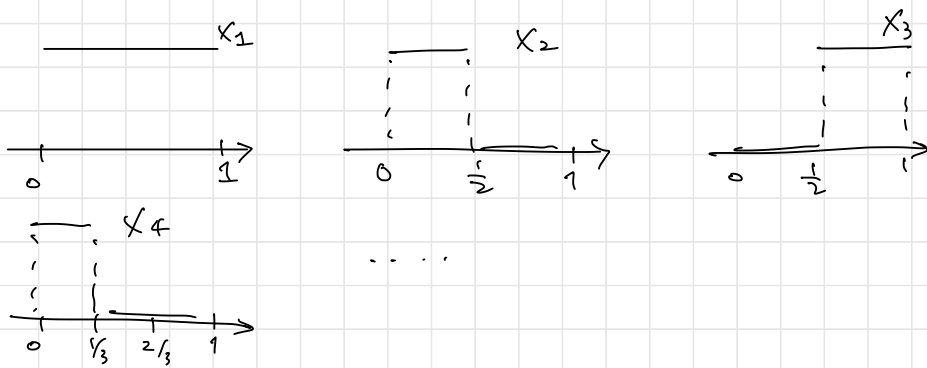
Def  $X_n \rightarrow X$  in prob. if  $P(|X_n - X| > \epsilon) \rightarrow 0 \forall \epsilon.$

$X_n \rightarrow X$  a.s. if  $P(\lim_{n \rightarrow \infty} X_n = X) = 1$

Prop  $X_n \rightarrow X$  a.s. implies  $X_n \rightarrow X$  in prob.

Proof 
$$\lim_{m \rightarrow \infty} P\left(\bigcup_{n \geq m} \{|X_n - X| > \epsilon\}\right) = 0$$
$$\geq P(|X_m - X| > \epsilon).$$

Note The converse is not true.



$X_n \rightarrow 0$  in prob.

$X_n \not\rightarrow 0$  a.s.

### Central Limit Thm

Let  $X_1, X_2, \dots$  be a sequence of independent, identically distributed random variables. with  $\mu = EX_1$ ,  $\sigma^2 = \text{Var}(X_1) < \infty$ . Let  $S_n = X_1 + X_2 + \dots + X_n$ , then

$$\lim_{n \rightarrow \infty} P\left(\frac{S_n - E[S_n]}{\sqrt{\text{Var}(S_n)}} \leq x\right) = P(Z \leq x) = \Phi(x), \quad \forall x \in \mathbb{R}$$

Let  $\mu = EX_i$ ,  $\sigma^2 = \text{Var}(X_i)$  then  $E[S_n] = \mu n$ ,  $\text{Var}(S_n) = n\sigma^2$

Thus

$$P\left(\frac{S_n - E[S_n]}{\sqrt{\text{Var}(S_n)}} \leq x\right) = P\left(\frac{S_n - \mu \cdot n}{\sigma \sqrt{n}} \leq x\right)$$