

# Lecture Summary

## 6.2 The central limit theorem

# Convergence of Random Variables

Let  $X_1, X_2, \dots$  be a sequence of random variables, let  $X$  be a random variable.

## Convergence in Probability

$X_1, X_2, \dots$  converges in probability to  $X$  if

$$\forall \epsilon > 0, \lim_{n \rightarrow \infty} P(|X_n - X| \geq \epsilon) = 0$$

## Convergence in Distribution

$X_1, X_2, \dots$  converges in distribution to  $X$  if

$$\forall \epsilon > 0, \lim_{n \rightarrow \infty} F_{X_n}(x) = F_X(x)$$

for all  $x$  where  $F_X$  is continuous.

# Convergence of Random Variables

Let  $X_1, X_2, \dots$  be a sequence of random variables, let  $X$  be a random variable.

## Almost Sure Convergence

$X_1, X_2, \dots$  converges almost surely to  $X$  if

$$\forall \epsilon > 0, P(\lim_{n \rightarrow \infty} |X_n - X| \geq \epsilon) = 0$$

- ▶ Almost sure convergence implies convergence in probability.
- ▶ Convergence in probability implies convergence in distribution.
- ▶ Convergence in distribution does not imply similarity of the random variables, just their distributions.

# Central Limit Theorem

## Theorem (Central Limit Theorem)

If the random variables  $X_1, \dots, X_n$  form a random sample of size  $n$  from a given distribution with mean  $\mu$  and variance  $\sigma^2$  ( $0 < \sigma^2 < \infty$ ), then for each fixed number  $x$

$$\lim_{n \rightarrow \infty} P\left(\sqrt{n} \frac{\bar{X}_n - \mu}{\sigma}\right) = \Phi(x),$$

where  $\Phi$  denotes the c.d.f. of the standard normal distribution.

# Central Limit Theorem

- ▶  $S_n = \sum_{i=1}^n X_i$ , mean  $n\mu$ , variance  $n\sigma^2$ .
- ▶  $\bar{X}_n = \frac{S_n}{n}$ , mean  $\mu$  variance  $\frac{\sigma^2}{n}$ .
- ▶  $\frac{S_n}{\sqrt{n}}$ , mean  $\mu\sqrt{n}$ , variance  $\sigma^2$ .
- ▶  $Z_n = \frac{S_n - n\mu}{\sqrt{n}\sigma} = \frac{\sqrt{n}(\bar{X}_n - \mu)}{\sigma}$ , mean 0, variance 1.

## Example

- ▶ You are doing a poll on "ratio of people who believe it was a proper penalty".
- ▶ True ratio:  $p$ , estimate  $\bar{X}_n$
- ▶ No guarantee for finding exactly  $p$ , so

$$P(|\bar{X}_n - p| \geq 0.01) \leq 0.05$$

- ▶ Apply Chebysev inequality with  $t = 0.01$ :  $n = 50,000$ .
- ▶ Apply CLT: ?

# The Delta Method

Sometimes we are interested in the asymptotic behavior of a function of the sample mean.

## Theorem (The Delta Method)

*Let  $X_1, X_2, \dots$  be a sequence of i.i.d. random variables from a distribution with mean  $\mu$  and finite variance  $\sigma^2$ . Let  $\alpha$  be a function with continuous derivative such that  $\alpha(\mu)' \neq 0$ . Then the asymptotic distribution of*

$$\frac{n^{1/2}}{\sigma\alpha'(\mu)}[\alpha(\bar{X}_n) - \alpha(\mu)]$$

*is the standard normal distribution.*

## Practice Exercises

6.2 2, 6, 8

6.3 3, 10, 13