

# Parametric Statistics

## Large Random Samples

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# Lecture Summary

6.3 The central limit theorem

6.4 Correction for continuity

## Last time

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.

## Last time

### Markov Inequality

Let  $X$  be a random variable such that  $P(X \geq 0) = 1$ . Then for every real number  $t$ ,

$$P(X \geq t) \leq \frac{E(X)}{t}.$$

### Chebysev Inequality

Let  $X$  be a random variable for which  $Var(X)$  exists. Then for every number  $t > 0$ ,

$$P(|X - E(X)| \geq t) \leq \frac{Var(X)}{t^2}.$$

# Compute the probability of Uniform

## Example

Suppose that a random sample of size  $n = 12$  is taken from the uniform distribution on the interval  $[0, 1]$ . We shall approximate the value of  $\Pr(|\bar{X}_n - \frac{1}{2}| \leq 0.1)$

- ▶  $E(X_i) = \frac{1}{2}, E(\bar{X}_n) = \frac{1}{2}$
- ▶  $Var(X_i) = \frac{1}{12}, Var(\bar{X}_n) = \frac{1}{12n}$

# Central Limit Theorem

## Lindberg and Lévy

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  $x$ ,

$$\lim_{n \rightarrow \infty} \Pr \left[ \frac{\bar{X}_n - \mu}{\sigma/n^{1/2}} \leq x \right] = \Phi(x),$$

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

## Stable variance

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

# The Delta Method

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

## Delta Method for Average of a Random Sample.

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.

## Example

What is the asymptotic limit of  $\sqrt{n}(\bar{X}_n^2 - \mu^2)$  ?

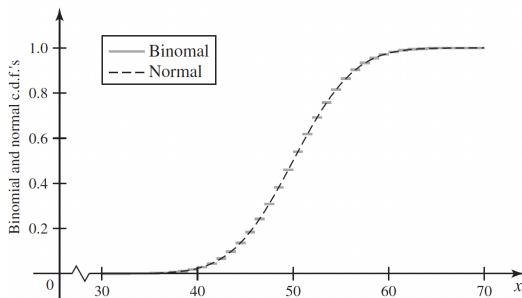


## Correction for continuity

- ▶  $Y \sim \text{Binomial}(n = 100, p = \frac{1}{2})$ . We are interested in  $P(Y \leq 50)$ .
- ▶ We know that a Binomial  $(n = 100, p = \frac{1}{2})$  can be written as the sum of  $n$  i.i.d. Bernoulli ( $p$ ) random variables.
- ▶ Using CLT:

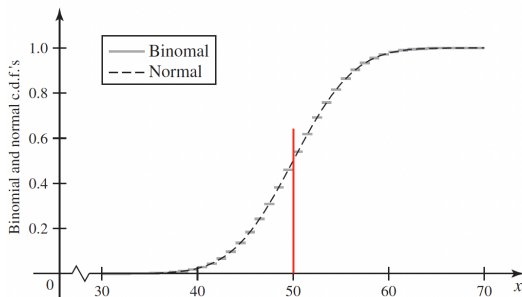
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- ▶ Using CLT:  $P(Y \leq 50) = \Phi(0) = 0.5$
- ▶ Using Binomial:  $P(Y \leq 50) = 0.539$



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- ▶ Using CLT:  $P(Y \leq 50) = \Phi(0) = 0.5$
- ▶ Using Binomial:  $P(Y \leq 50) = 0.539$
- ▶ Instead of  $\Pr(Y \leq x)$  use  $\Pr(Y \leq x + 0.5)$ , which is larger and usually closer to  $\Pr(X \leq x)$ .
- ▶  $\Pr(Y = x) = P(x - 0.5 \leq Y \leq x + 0.5)$

## Application of the CLT

- ▶ You are doing a poll on “ratio of people who believe in climate change”.
- ▶ True ratio:  $p$ , estimate  $\bar{X}_n$ .
- ▶ Your boss wants a guarantee that your prediction will be “off” by more than 1% with probability at most 5%

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- ▶ True ratio:  $p$ , estimate  $\bar{X}_n$ .
- ▶ Your boss wants a guarantee that your prediction will be “off” by more than 1% with probability at most 5%
- ▶ No guarantee for finding exactly  $p$ , so

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

- ▶ Apply Chebysev inequality with  $t = 0.01$ :
- ▶ Apply CLT:

# Group Exercise

- ▶ Form teams of two.
- ▶ Go to <https://rolladie.net/roll-4-dice>.
- ▶ Complete the exercise sheet given to you and answer the questions.