# Chapter 37: Central Limit Theorem

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### **Learning Objectives**

- 1. Calculate probability of a sample mean using a population mean and variance with unknown distribution
- 2. Use the Central Limit Theorem to construct the Normal approximation of the Binomial and Poisson distributions

### The Central Limit Theorem

#### Theorem 1: Central Limit Theorem (CLT)

Let  $X_i$  be iid rv's with common mean  $\mu$  and variance  $\sigma^2$ , for  $i=1,2,\ldots,n$ . Then

$$\sum_{i=1}^n X_i \longrightarrow \mathbb{N}(n\mu, n\sigma^2)$$

- · Xi's do NOT need to be normally distrib.
- · do NOT need to know dist of X
- when nis large, we can use the Normal approx

Chapter 37 Slides

> converges in distribution as

### **Extension of the CLT**

#### Corollary 1

Let  $X_i$  be iid rv's with common mean  $\mu$  and variance  $\sigma^2$  , for  $i=1,2,\ldots,n$ . Then

$$\overline{X} = rac{\sum_{i=1}^n X_i}{n} o \mathrm{N}igg(\mu, rac{\sigma^2}{n}igg)$$

## Example of Corollary in use

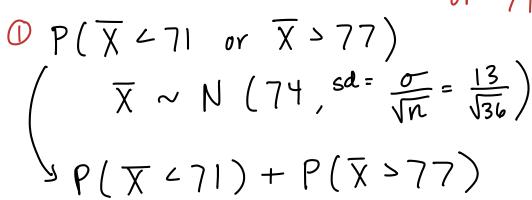
#### Example :

According to a large US study, the mean resting heart rate of adult women is about 74 beats per minutes (bpm), with standard deviation 13 bpm (NHANES 2003-2004).

- 1. Find the probability that the average resting heart rate for a random sample of 36 adult women is more than 3 bpm away from the mean.
- 2. Repeat the previous question for a single adult woman.



$$n = 30$$
, then can Normal approx  
 $\mu = 74$   $\sigma = 13$   $n = 36$   $\overline{\chi} = 77$   
or  $71$ 



= 
$$2 \cdot pnorm(x=71, mean = 74, sa=13/sqrt(36))$$

$$= 0.166$$

= 2 P(X < 71)

### Example of CLT for exponential distribution

Example 2

$$E(X_{i}) = \frac{1}{\lambda} = \mu$$

Let  $X_{i} \sim Exp(\lambda)$  be iid RVs for  $i = 1, 2, ..., n$ . Then  $\forall Var(X_{i}) = \frac{1}{\lambda^{2}} = 0^{\lambda}$ 

Converges in distribution as  $n \rightarrow \infty$ 

$$n = \frac{n}{\lambda}$$

$$n = \frac{n}{\lambda^{2}}$$

$$\sum_{i=1}^{n} X_{i} \rightarrow N \left(\frac{n}{\lambda}, \frac{n}{\lambda^{2}}\right)$$

$$\sum_{i=1}^{n} X_{i} \rightarrow N \left(\frac{n}{\lambda}, \frac{n}{\lambda^{2}}\right)$$

### **CLT for Discrete RVs**

$$ullet X = \sum_{i=1}^n \underline{X_i}, ext{where}\, \underline{X_i} ext{ are iid} \overline{\operatorname{Bernoulli}(p)}$$

• Rule of thumb:  $\underline{np} \geq 10$  and  $n(1-p) \geq 10$  to use Normal approximation

$$\sum_{i=1}^{n} X_{i} \longrightarrow N(np, np(1-p))$$

$$\hat{p} \rightarrow N(p, \frac{p(1-p)}{N})$$

2. Poisson rv's: Let  $X \sim Poisson(\lambda)$ 

$$ullet X = \sum_{i=1}^n X_i, ext{where } X_i ext{ are iid } ext{Poiss}(1)$$

- ullet Recall from Chapter 18 that if  $X_i \sim Poiss(\lambda_i)$  and  $X_i$  independent, then  $\sum_{i=1}^n X_i \sim Poiss(\sum_{i=1}^n \lambda_i)$
- Rule of thumb:  $\lambda \geq 10$  to use Normal approximation

$$\sum_{i=1}^{n} X_{i} \rightarrow N(\lambda, \sigma = \lambda)$$

$$\sqrt{\alpha r} = \lambda^{2}$$

### At home example

#### Example 3

Suppose that the probability of developing a specific type of breast cancer in women aged 40-49 is 0.001. Assume the occurrences of cancer are independent. Suppose you have data from a random sample of 20,000 women aged 40-49.

- 1. How many of the 20,000 women would you expect to develop this type of breast cancer, and what is the standard deviation?
- 2. Find the exact probability that more than 15 of the 20,000 women will develop this type of breast cancer.
- 3. Use the CLT to find the **approximate** probability that more than 15 of the 20,000 women will develop this type of breast cancer.
- 4. Use the CLT to approximate the following probabilities, where X is the number of women that will develop this type of breast cancer.

a. 
$$\mathbb{P}(15 \leq X \leq 22)$$

b. 
$$\mathbb{P}(X>20)$$

c. 
$$\mathbb{P}(X < 20)$$

- 5. Find the approximate probability that more than 15 of the 20,000 women will develop this type of breast cancer not using the CLT!
- 6. Use the CLT to approximate the approximate probability in the previous question!