

# Lecture 16

## Central Limit Theorem

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Manju M. Johny

STAT 330 - Iowa State University

## Central Limit Theorem (CLT)

Suppose  $X_1, X_2, \dots, X_n$  are iid random variables. For  $i = 1, \dots, n$ ,

$$X_i \stackrel{iid}{\sim} \text{distribution}$$

Any function of  $\{X_i\}$  is also a random variable. Specifically,

- $S_n = \sum_{i=1}^n X_i$  is a R.V (with some distribution)
- $\bar{X}_n = \frac{\sum_{i=1}^n X_i}{n}$  is a R.V (with some distribution)

For large sample size  $n$ , the distribution of  $S_n$  and  $\bar{X}$  both follow **normal distributions!**

Even without knowing the distribution of  $\{X_i\}$ , we can calculate probabilities for its sample mean and sample sum using the normal distribution. (extremely useful for real life problems)!

# Central Limit Theorem (CLT)

- Sums and averages of RVs from *any* distribution have approximately normal distributions for large sample sizes

## Central Limit Theorem (CLT)

Suppose  $X_1, X_2, \dots, X_n$  are iid random variables with  $E(X_i) = \mu$  and  $Var(X_i) = \sigma^2$  for  $i = 1, \dots, n$ .

Define:

1. sample mean:  $\bar{X}_n = \frac{\sum_{i=1}^n X_i}{n}$
2. sample sum:  $S_n = \sum_{i=1}^n X_i$

Then, for *large*  $n$ ,

$$\bar{X}_n \sim N\left(\mu, \frac{\sigma^2}{n}\right)$$
$$S_n \sim N(n\mu, n\sigma^2)$$

## How to Use CLT for Means

- For large  $n$ ,

$$\bar{X}_n \sim N\left(\mu, \frac{\sigma^2}{n}\right)$$

- How to calculate probabilities involving  $\bar{X}_n$  ?
- Standardize  $\bar{X}_n$  to turn it into a standard normal random variable  $Z$ , and use the  $z$ -table! (lecture notes 14)
- Standardize any normal random variable by subtracting its mean, and dividing by its standard deviation.

$$Z = \frac{\bar{X}_n - \mu}{\sigma/\sqrt{n}}$$

$$Z \sim N(0, 1)$$

## How to Use CLT for Means Cont.

- Ex:  $P(a < \bar{X}_n < b) = ?$
- Standardize all of the quantities involved in the above probability. Then use Z-table to obtain probabilities.

$$\begin{aligned}P(a < \bar{X}_n < b) &= P\left(\frac{a - \mu}{\sigma/\sqrt{n}} < \frac{\bar{X}_n - \mu}{\sigma/\sqrt{n}} < \frac{b - \mu}{\sigma/\sqrt{n}}\right) \\&= P\left(\frac{a - \mu}{\sigma/\sqrt{n}} < Z < \frac{b - \mu}{\sigma/\sqrt{n}}\right) \\&= P\left(Z < \frac{b - \mu}{\sigma/\sqrt{n}}\right) - P\left(Z < \frac{a - \mu}{\sigma/\sqrt{n}}\right) \\&= \Phi\left(\frac{b - \mu}{\sigma/\sqrt{n}}\right) - \Phi\left(\frac{a - \mu}{\sigma/\sqrt{n}}\right)\end{aligned}$$

# How to Use CLT for Sums

- For large  $n$ ,

$$S_n \sim N(n\mu, n\sigma^2)$$

- Standardize  $S_n$  by subtracting its mean, and dividing by its standard deviation.

$$Z = \frac{S_n - n\mu}{\sqrt{n\sigma^2}} = \frac{S_n - n\mu}{\sigma\sqrt{n}}$$

$$Z \sim N(0, 1)$$

- Then, use the  $Z$ -table to obtain desired probabilities.
- Ex:

$$\begin{aligned} P(S_n < a) &= P\left(\frac{S_n - n\mu}{\sigma\sqrt{n}} < \frac{a - n\mu}{\sigma\sqrt{n}}\right) \\ &= P\left(Z < \frac{a - n\mu}{\sigma\sqrt{n}}\right) \\ &= \Phi\left(\frac{a - n\mu}{\sigma\sqrt{n}}\right) \end{aligned}$$

## Examples

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## Examples

Example 1: The time you spend waiting for the bus each day has a uniform distribution between 2 minutes and 5 minutes. Suppose you wait for the bus every day for a month (30 days).

1. Let  $X_i =$  time spent waiting for the bus on the  $i^{\text{th}}$  day for  $i = 1, \dots, 30$ .

What is the distribution of each  $X_i$ ?

What is its expected value and variance?



## Examples

2. Let  $\bar{X}_n$  be the average time spent waiting for the bus over the month.  $\bar{X}_n = \frac{\sum_{i=1}^n X_i}{n} = \frac{\sum_{i=1}^{30} X_i}{30}$

What is the (approximate) probability that the average time you spent waiting for the bus is less than 4 min?

# Examples

## Examples

3. How much time do you expect to spend waiting for the bus in total for a month?
4. What is the (approximate) probability that you spend more than 2 hours waiting for a bus in total for a month?



## Examples

Example 2: Suppose an image has an expected size 1 megabyte with a standard deviation of 0.5 megabytes. A disk has 330 megabytes of free space. Is this disk likely to be sufficient for 300 independent images?



## Examples

Example 3: An astronomer wants to measure the distance,  $d$ , from the observatory to a star. The astronomer plans to take  $n$  measurements of the distance and use the sample mean to estimate the true distance. From past records of these measurements the astronomer knows the standard deviation of a single measurement is 2 parsecs. How many measurements should the astronomer take so that the chance that his estimate differs by  $d$  by more than 0.5 parsecs is at most 0.05?





# Examples