

In the previous lecture, you learned the concept of **Maximum Likelihood Estimation (MLE)** and showed that the **sample proportion** is the maximum likelihood estimator of a **population proportion**.

In this lecture, you will learn how to use the **likelihood function** to test a claim about a population proportion.

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Introduction to Likelihood Ratio and its Properties

Example:

A random sample of 100 students is selected from a college, and 60 of them use an iPhone. We want to use the sample data to determine whether the proportion of all students who use an iPhone is different from 50%.

The null and alternative hypotheses are:

$$H_0: p = 0.5$$

$$H_a: p \neq 0.5$$

To perform the test, we calculate how "likely" your observed data is under both hypotheses. Then compare these two likelihoods using a ratio (R):

$$R = \frac{\text{Likelihood under } H_0}{\text{Maximum Likelihood under } H_a}$$

The Likelihood Ratio Test asks the following question:

Does the null hypothesis make your observed data nearly as likely to occur as the maximum likelihood under the alternative hypothesis?

Properties of Likelihood Ratio

1. The likelihood ratio R is always less than or equal to 1, because the maximum likelihood under the alternative hypothesis is always larger than or equal to the likelihood under the null hypothesis.
2. If the null hypothesis makes the observed data nearly as likely as the maximum likelihood under the alternative hypothesis, then R is close to 1.

This indicates that the observed data are consistent with H_0 , and we do not reject H_0 .

3. If the null hypothesis does not make the observed data nearly as likely as the maximum likelihood under the alternative hypothesis, then R is close to 0.

This indicates that the observed data are not consistent with H_0 , and we reject H_0 .

4. As the likelihood ratio R gets closer to 0, the evidence against the null hypothesis becomes stronger.

Performing the Likelihood Ratio Test (LRT)

From the last lecture, we learned the likelihood function of the unknown population proportion is given by

$$L(p) = \text{constant} * p^x (1 - p)^{n-x}$$

where

- p represents the unknown value of the population proportion (e.g. the proportion of all students who use an iPhone),
- x represents the number of individuals who fall into the category of interest (e.g. the number of students who use an iPhone),
- n represents the sample size.

Note: The *constant* makes no contribution to the Likelihood Ratio Test because it cancels out when forming the likelihood ratio. Therefore, it can be safely omitted from the calculations.

In this example, we sampled **100 students**, and **60 of them use an iPhone**.

Then, $n = 100$ and $x = 60$. The likelihood function is:

$$L(p) =$$

Under the null hypothesis, $p = 0.5$. We substitute 0.5 into the likelihood function.

The likelihood under H_0 is

$$L(0.5) =$$

Under the alternative hypothesis, $p \neq 0.5$. So, there are many possible choices of p .

We will choose the value of p that maximizes the likelihood function.

The value of p that maximizes the likelihood function is the **sample proportion**, \bar{p} .

$$\text{The sample proportion is: } \bar{p} = \frac{x}{n} = \frac{60}{100} = 0.6$$

Under the alternative hypothesis, $p = 0.6$ We substitute 0.6 into the likelihood function.

The likelihood under H_a is

$$L(0.6) =$$

Therefore, the likelihood ratio is by:

$$\begin{aligned} R &= \frac{L(0.5)}{L(0.6)} = \\ &= \frac{0.5^{100}}{0.6^{60} * 40^{40}} \\ &\approx 0.1335137 \end{aligned}$$

As we can see, the likelihood ratio is **very close to zero**. This indicates that the observed data are **not consistent with H_0** and provide **strong evidence against the null hypothesis**.

To determine **how strong this evidence is**, we calculate the **p-value**.

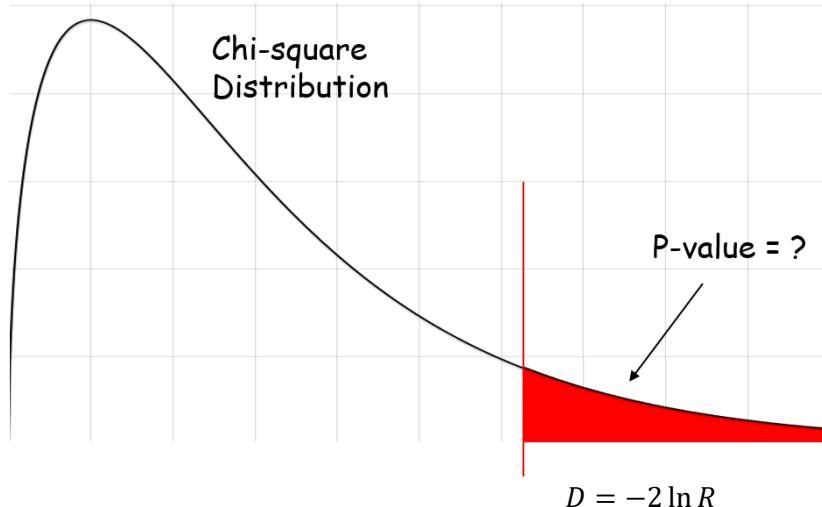
Calculate the P-value of the Likelihood Ratio Test (LRT)

To do this, we first convert the likelihood ratio R into a **Chi-Square test statistic** (also called the **deviance**):

$$D = -2 \ln R \quad \text{or} \quad -2 \ln \left(\frac{\text{Likelihood under } H_0}{\text{Maximum Likelihood under } H_a} \right)$$

For a **large sample**, the statistic D follows a **Chi-Square distribution with 1 degree of freedom**.

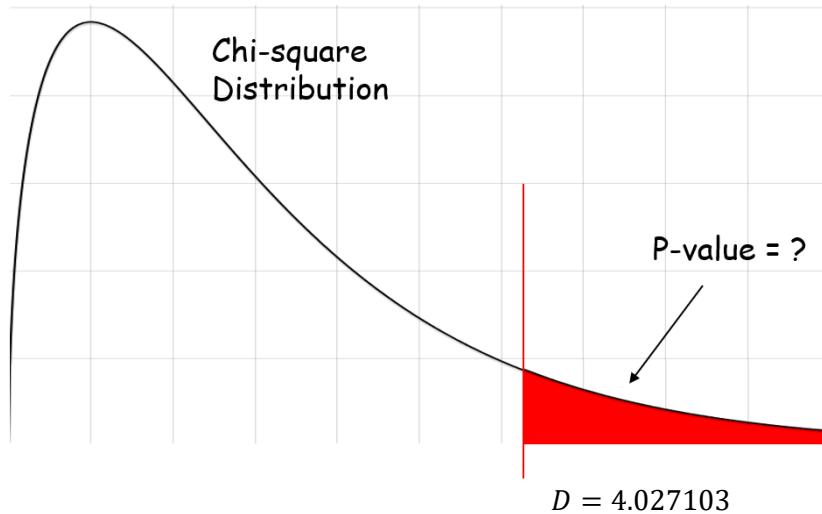
The **p-value** is then calculated as the area on the right under the Chi-Square distribution ($\text{df} = 1$) bounded by D .



Let's get the p-value in our example.

$$\begin{aligned} R &= \frac{0.5^{100}}{0.6^{60} 0.4^{40}} \\ \rightarrow D &= -2 \ln \left(\frac{0.5^{100}}{0.6^{60} 0.4^{40}} \right) \end{aligned}$$

$$\approx 4.027103$$



The p-value of D is calculated from the **Chi-Square distribution with 1 degree of freedom**.

In R, you can obtain it using the following command:

```
r = 0.5^100/ ((0.6)^60 * 0.4^40)
d = -2 * log( r )

#Calculate the area on the left under the Chi-Square curve
#using pchisq(...) function.
#The pchisq function takes two inputs:
# (1) the chi-square value and (2) the degree of freedom(df)
area.left.tail = pchisq( d, df=1)
pvalue = 1-area.left.tail
```

```
R Console

> r = 0.5^100/ ((0.6)^60 * 0.4^40)
> d = -2 * log( r )
> area.left.tail = pchisq( d, df=1)
> pvalue = 1-area.left.tail
> pvalue
[1] 0.04477477
```

Since the p -value (0.04477) is less than the significance level (0.05), we reject H_0 .

At the 5% significance level, the sample data provide sufficient evidence to conclude that the proportion of all students who use an iPhone is different from 50%.

Conditions for a Valid Likelihood Ratio Test of a Population Proportion

As mentioned earlier, the sample size must be sufficiently large. But how large is enough?

For the likelihood ratio test of a population proportion, it is recommended that:

- The expected number of individuals who fall into the category of interest and
- The expected number of individuals who **do not** fall into the category of interest

are both **at least 5** in the sample. The expected numbers are calculated based on the null hypothesis.

In our example, under H_0 , we expect 50% of 100 sampled students use an iPhone,

So, the expected numbers are:

| | Expected Number |
|---------------------|--------------------------|
| Using an iPhone | 50% of 100 = 50 ≥ 5 |
| Not using an iPhone | $100 - 50 = 50 \geq 5$ |

Since both numbers are at least 5, the sample is sufficiently large.

The conclusion drawn from the test should be valid.

Likelihood Ratio Test vs One-Sample Z Test

We can also perform the Z-test to determine whether the proportion of all students who use an iPhone is different from 50%. The null and alternative hypotheses are shown:

$$H_0: p = 0.5$$

$$H_a: p \neq 0.5$$

The z-statistic and p-value are shown below.

p_0 is the population proportion of iPhone users under $H_0 \rightarrow 0.5$

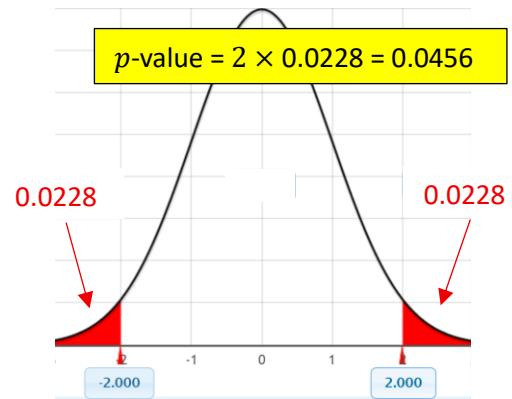
$$z = \frac{\bar{p} - p_0}{\sqrt{\frac{p_0(1-p_0)}{n}}}$$

$$p\text{-value} = 2 \times 0.0228 = 0.0456$$

\bar{p} is the sample proportion of iPhone users $\rightarrow 0.6$

$$= \frac{0.6 - 0.5}{\sqrt{\frac{0.5(1-0.5)}{100}}} = 2$$

n is the sample size $\rightarrow 100$



The results of the Likelihood Ratio Test and the Z-Test are summarized below:

| | Likelihood Ratio Test | Z-Test |
|---------|-----------------------|--------|
| p-value | 0.0448 | 0.0456 |

The p-values obtained from the two tests are nearly identical. This indicates a strong agreement between the Likelihood Ratio Test and the Z-Test.

In fact, the chi-square statistic (often referred to as the deviance) used in the Likelihood Ratio Test can be approximated by the following expression.

Likelihood Ratio Test

$$-2 \ln \left(\frac{\text{Likelihood under } H_0}{\text{Maximum Likelihood under } H_a} \right) \approx \left(\frac{\bar{p} - p_0}{\sqrt{\frac{\bar{p}(1 - \bar{p})}{n}}} \right)^2$$

The quantity inside the square, $\frac{\bar{p} - p_0}{\sqrt{\frac{\bar{p}(1 - \bar{p})}{n}}}$,

is similar the z-statistic under the Z-test $\frac{\bar{p} - p_0}{\sqrt{\frac{p_0(1 - p_0)}{n}}}$.

Mathematically, both two quantities follow a standard normal distribution (with mean of 0 and SD of 1).

As a result, the Likelihood Ratio Test and the Z-test produce very similar p-values.