

1. Introduction

Welcome to the series of E-learning modules on Confidence Intervals for the ratio of variances of two independent normal populations. In this module, we are going to cover the Interval Estimation - procedure to estimate the ratio of two population variances or standard deviations when the population means are known and unknown.

By the end of this session, you will be able to:

- Explain the role of F distribution in an Interval Estimation
- Explain the confidence interval for the ratio of variances or standard deviations of two independent normal populations when the population means are known
- Explain the confidence interval for the ratio of variances or standard deviations of two independent normal populations when the population means are unknown

We saw several situations in which we wanted to compare the means of two different populations. Recall, we did this by looking at the difference of the means of the two samples drawn from those populations.

Like the importance of a single variance differences in variances to an experimenter, it also important to compare the two population variances.

We might need to compare the precision of one measuring device with that of another and also the stability of one manufacturing process with that of another or even the variability in the grading procedure of one college professor with that of another.

However, when we want to compare the variances of two populations rather than looking at the difference of the two sample variances, it turns out to be more convenient if we look at their ratio.

For example, a prominent sociologist at a large Midwestern University wants to estimate ratio of variability in the earnings of college graduates to the earnings of those who did not attend college.

Hence, a way to compare the variances of two normally distributed populations is to use the variance ratio, σ_1^2 by σ_2^2 .

Consider two independent normal populations X and Y having unknown variances σ_1^2 and σ_2^2 respectively. We wish to construct a confidence interval for the ratio of variances σ_1^2 by σ_2^2 . This ratio can be initially estimated by the ratio of sample variances $(s_x)^2$ by $(s_y)^2$ obtained from independent random samples from X and Y having sizes m and n respectively.

If $(s_x)^2$ by $(s_y)^2$ is nearly equal to 1, we will find little evidence to indicate that σ_1^2 and σ_2^2 are unequal. On the other hand, a very large or very small value of $(s_x)^2$ by $(s_y)^2$ provides evidence of a difference in the population variances. How large or small must $(s_x)^2$ by $(s_y)^2$ be found by studying the distribution of $(s_x)^2$ by $(s_y)^2$ in repeated sampling.

Sampling distribution

The sampling distribution of $(s_x)^2$ by $(s_y)^2$ is $(s_x)^2$ by $(s_y)^2$.

is used. Since the population variances are usually not known, the sample variances are used. The assumptions are that s_x^2 and s_y^2 are computed from independent samples of size m and n , respectively, drawn from two normally distributed populations. If the assumptions are met, $(s_x^2 / \sigma_x^2) / (s_y^2 / \sigma_y^2)$ follows a distribution known as the F distribution with two values used for degrees of freedom.

2. F - Distribution

F-Distribution

The confidence interval for the ratio of two variances requires the use of the probability distribution known as the *F*-distribution. So, let us spend a few minutes learning the definition and characteristics of the *F*-distribution.

Characteristics of the *F*-Distribution

F-distributions are generally skewed. The shape of an *F*-distribution depends on the values of m and n , the numerator and denominator degrees of freedom respectively.

The *F*-Table

One of the primary ways that we will need to interact with an *F*-distribution is by knowing either (1) an *F*-value, or (2) the probabilities associated with an *F*-random variable in order to complete a statistical analysis

Degrees of freedom

The *F* distribution uses two values for degrees of freedom. The numerator degree of freedom is the value of m minus 1, which is used in calculating s_x square. The denominator degree of freedom is the value of n minus 1, which is used in calculating s_y square.

Reading *F* tables

F tables come in denominations based on $F(1 - \alpha/2)$ which are *F* point nine nine five, *F* point nine nine, *F* point nine seven five etc or with $F_{\alpha/2}$ must be calculated to give values of *F* point zero five, *F* point zero two five and *F* point zero zero five.

Here are the steps for using the *F*-table to find a *F*-value.

1. Find the column that corresponds to the relevant numerator degrees of freedom m .
2. Find the row that corresponds to the relevant denominator degrees of freedom n and $\alpha/2$
3. Determine the *F*-value, where the m column and the probability row identified in the row corresponding to n degree of freedom and $\alpha/2$ or $1 - \alpha/2$ level of significance intersect

Confidence Interval for variance

Confidence intervals are not just for means or difference of means. You are already familiar with confidence intervals for the difference of means of two populations. The idea of a confidence interval is very general, and you can express the precision of any computed value as a ninety five percent or ninety-nine percent confidence interval (CI).

Very often practical problems lead to estimation of sigma square, the variance of the population and the ratio of variances of two independent populations. A confidence interval for the ratio of two population variances specifies a range of values within which the unknown population parameter, in this case the ratio of variances may lie.

The ninety five percent confidence interval for the ratio of variances

The sample variances are the values that we compute from a samples of data, which gives point estimates of the variances of two populations. It is not done often, but it is certainly

possible to compute a confidence interval for the ratio of two independent population variances.

If you assume that your data were randomly and independently sampled from a Gaussian (Normal) distribution, you can be ninety-five percent sure that the confidence interval computed from the sample variances contains the ratio of two true population variances.

An interval estimator is a rule for computing two numbers say A and B . This is to create an interval that contains the parameter of interest that is the ratio of two population variances σ_1^2 by σ_2^2 .

Hence, a confidence interval gives an estimated range of values for the ratio of two population variances or standard deviations, which is likely to include a ratio of two unknown population variances or standard deviations, the estimated range being calculated from a given set of sample data.

3. Result 1

Result 1:

Confidence Intervals for the ratio of variances of two independent populations when the population means are known

Assumptions:

- Population Means are known
- Populations are Normally distributed
- If population is not normal, use large sample

Suppose $x_1, x_2, \dots, x_m$ and $y_1, y_2, \dots, y_n$ are the random samples of size m and n drawn from the populations of size N_1 and N_2 with means μ_1 and μ_2 and variances σ_1^2 and σ_2^2 , we know that $\bar{x}$ is equal to $\sum x_i / m$ and $\bar{y}$ is equal to $\sum y_i / n$

X_i follows Normal distribution with mean μ_1 and variance σ_1^2 and Y_i follows Normal distribution with mean μ_2 and variance σ_2^2

$\bar{X}$ follows Normal distribution with mean μ_1 and variance σ_1^2 / m and $\bar{y}$ follows Normal distribution with mean μ_2 and variance σ_2^2 / n

$\sum (x_i - \mu_1)^2 / \sigma_1^2$ follows chi square with m degrees of freedom and

$\sum (y_i - \mu_2)^2 / \sigma_2^2$ follows chi square with n degrees of freedom

Then, $F(m, n)$ is equal to $\chi^2(m) / m$ divided by $\chi^2(n) / n$ which is equal to $\sum (x_i - \mu_1)^2 / \sigma_1^2$ into m divided by $\sum (y_i - \mu_2)^2 / \sigma_2^2$ into n

Which is equal to $n \sum (x_i - \mu_1)^2 / \sigma_1^2$ into σ_2^2 by $m \sum (y_i - \mu_2)^2$

Which is equal to σ_2^2 into n into $\sum (x_i - \mu_1)^2$ by σ_1^2 into m into $\sum (y_i - \mu_2)^2$

From the table of probabilities of Snedecor's F distribution, we can always find two quantities A and B such that

Probability of $A \leq F(m, n) \leq B$ is equal to $1 - \alpha$

Probability of $A \leq \sigma_2^2 / n \sum (x_i - \mu_1)^2 / \sigma_1^2 \leq B$ is equal to $1 - \alpha$

Probability of $A \leq \sum (y_i - \mu_2)^2 / \sigma_2^2 \leq B$ is equal to $1 - \alpha$

Probability of $n \sum (x_i - \mu_1)^2 / \sigma_1^2 \leq A$ is greater than σ_2^2 by A into $m \sum (y_i - \mu_2)^2$ is greater than $n \sum (x_i - \mu_1)^2$ by B into $m \sum (y_i - \mu_2)^2$

μ_2^2 whole square is equal to $1 - \alpha$

Probability of $\frac{n \sum (x_i - \mu_1)^2}{B \sum (y_i - \mu_2)^2} \leq \frac{\sigma_1^2}{\sigma_2^2}$ is less than or equal to $1 - \alpha$

Therefore, $(1 - \alpha)$ percent Confidence Interval for the ratio of variances of two populations with known means as μ_1 and μ_2 is given by

$\left[\frac{n \sum (x_i - \mu_1)^2}{B \sum (y_i - \mu_2)^2}, \frac{n \sum (x_i - \mu_1)^2}{A \sum (y_i - \mu_2)^2} \right]$

4. Result 2

Result 2:

Confidence Interval for the ratio of two variances of two populations with unknown means

Assumptions:

- Populations means are unknown
- Population is normally distributed
- If population is not normal, use large sample

Suppose $x_1, x_2, \dots, x_m$ and $y_1, y_2, \dots, y_n$ are the random samples of size m and n drawn from the populations of size N_1 and N_2 with means μ_1 and μ_2 and variances σ_1^2 and σ_2^2 ,

We know that $\bar{x}$ is equal to $\sum x_i / m$ and $\bar{y}$ is equal to $\sum y_i / n$

X_i follows Normal distribution with mean μ_1 and variance σ_1^2 and Y_i follows Normal distribution with mean μ_2 and variance σ_2^2

$\bar{X}$ follows Normal distribution with mean μ_1 and variance σ_1^2 / m and $\bar{y}$ follows Normal distribution with mean μ_2 and variance σ_2^2 / n

$(m-1) s_1^2 / \sigma_1^2$ follows Chi square with $(m-1)$ degrees of freedom and $(n-1) s_2^2 / \sigma_2^2$ follows Chi square with $(n-1)$ degrees of freedom

$F_{(m-1, n-1)}$ is equal to $\text{Chi square } (m-1) / (m-1) \text{ whole divided by Chi square } (n-1) / (n-1)$

Which is equal to $(m-1) s_1^2 / \sigma_1^2 \text{ into } (m-1) \text{ whole divided by } (n-1) s_2^2 / \sigma_2^2 \text{ into } (n-1)$

Which is equal to $s_1^2 / \sigma_1^2 \text{ by } s_2^2 \text{ into } \sigma_1^2 \text{ which is equal to } \sigma_2^2 \text{ into } s_1^2 \text{ by } \sigma_1^2 \text{ into } s_2^2$

From the table of probabilities of Snedecor's F distribution, we can always find two quantities A and B such that

Probability of A less than or equal to $F_{(m-1, n-1)}$ less than or equal to B is equal to $1 - \alpha$

Probability of A less than or equal to $s_2^2 / \sigma_2^2 \text{ into } s_1^2 / \sigma_1^2 \text{ into } s_2^2$ less than or equal to B is equal to $1 - \alpha$

Probability of A into $s_2^2 \text{ by } s_1^2$ less than or equal to $\sigma_2^2 \text{ by } \sigma_1^2$ less than or equal to B into $s_2^2 \text{ by } s_1^2$ is equal to $1 - \alpha$

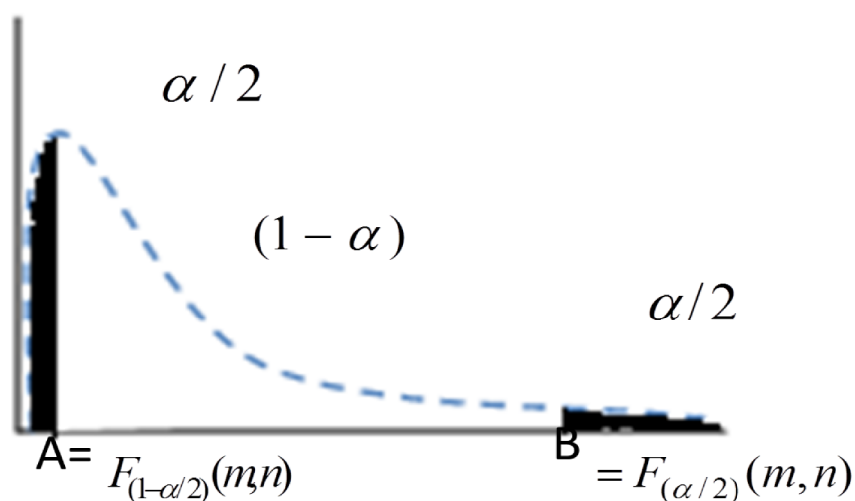
Probability of $s_1^2 \text{ by } A \text{ into } s_2^2 \text{ greater than } \sigma_1^2 \text{ by } \sigma_2^2 \text{ square greater than } s_1^2 \text{ by } B \text{ into } s_2^2$ is equal to $1 - \alpha$

Probability of $s_1^2 \text{ by } B \text{ into } s_2^2$ less than or equal to $\sigma_1^2 \text{ by } \sigma_2^2$ less than or equal to $s_1^2 \text{ by } A \text{ into } s_2^2$ is equal to $1 - \alpha$

Therefore, hundred into $(1 - \alpha)$ percent confidence Interval for the ratio of variances of two populations with unknown means is given by

$[s_1^2 \text{ by } B \text{ into } s_2^2, s_1^2 \text{ by } A \text{ into } s_2^2]$

Figure 1



As shown in the graph, here there are two critical values for each level of confidence. The value of B is equal to F (alpha by two) with (m, n) degrees of freedom represents the right-tail critical value and A is equal to F (1 minus alpha by two) with (m, n) degrees of freedom represents the left-tail critical value. Each area in the F- table represents the region under F curve to the RIGHT of the critical value.

We can use the critical values B is equal to F (alpha by two) with (m, n) degrees of freedom and A is equal to F (1 minus alpha by two) with (m, n) degrees of freedom to construct confidence intervals for the ratio of population variances and standard deviations when means are known.

5. Steps to Obtain Confidence Interval for the Ratio of two Variances

Steps to obtain confidence interval for the ratio of two variances

- First, obtain the point estimates of μ_1 and μ_2 that is, the sample mean $\bar{x}$ and $\bar{y}$ if population means are not known otherwise consider the value of μ_1 and μ_2
- Select a confidence level. The most common level is ninety-five percent or ninety-nine percent
- Obtain the critical values of F probabilities for $\alpha/2$ and $1 - \alpha/2$ level of significance with (m, n) degrees of freedom if μ_1 and μ_2 are known ($(m - 1, n - 1)$ degrees of freedom if μ_1 and μ_2 are unknown)
- Let B is equal to $F(\alpha/2)$ with (m, n) degrees of freedom and A is equal to $F(1 - \alpha/2)$ with (m, n) degrees of freedom
- Then, the confidence interval for the ratio of variances when means are known as μ_1 and μ_2 is computed by making use of the given derived formula in result one
- When the population means are unknown, Let B is equal to $F(\alpha/2)$ with $(m - 1, n - 1)$ degrees of freedom and A is equal to $F(1 - \alpha/2)$ with $(m - 1, n - 1)$ degrees of freedom
- Then, the confidence interval for the ratio of variances when means are unknown is computed by making use of the above-derived formula in result two

Here's a summary of our learning in this session, where we understood:

- The derivation for confidence limits for the ratio of the two unknown population variances when means are known
- The derivation for confidence limits for the ratio of two unknown population variances when the means are unknown