

Probability and Statistics
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Lecture - 64
Confidence Intervals – IV

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Ex To compare the gripping strengths of left hand and right hand of left handed persons the measurements are made on 10 persons

Person	1	2	3	4	5	6	7	8	9	10
Left hand	140	90	125	130	95	121	85	97	131	110
Right hand	138	87	110	132	96	120	86	90	129	100

Confidence interval for $\mu_1 - \mu_2$

d_i | 2 | 3 | 15 | -2 | -1 | 1 | -1 | 7 | 2 | 10

$\bar{d} = 3.6, s_d = \frac{1}{9} \sum d_i^2 - \bar{d}^2$

$0.05, 9 \quad 1.833 = \frac{1}{9} (4+9+225+4+1+1+1+49+4+100) - (3.6)^2$

$(3.6 \pm \frac{1.833}{\sqrt{9}})$

Let me take up some examples here to illustrate the situations. So, to compare the gripping a strengths of left hand and right hand of 10 handed of left handed persons the measurements are made on 10 persons and the data is observed. So, left hand and right hand, and we have of persons 1, 2, 3, 4, 5, 6, 7, 8, 9 and 10. The gripping strengths are measured as 140, 90, 125, 130, 95, 121, 85, 97, 131, 110; for the right hand it is 138, 87, 110, 132, 96, 120, 86, 90, 129, 100.

So, we need the confidence interval for say $\mu_1 - \mu_2$. Now observe here that this is the data related with the correlated observations. So, we will need here the means of; so let me call this as the first set. So, this is x_i data this is y_i data. So, we will look at d_i 's the differences here. So, the differences here is 2 3 15 minus 2 minus 1 1 minus 1 7 2 and 10. So, we look at the \bar{d} value here which is the mean of this that is 1 t, so 17, 24, 26, 36 that is 3.6. Similarly we calculate s_d that will be equal to 1 by 9 sigma d_i square minus \bar{d} square. So, once again it can be easily evaluated it is 4 plus 9 plus 225 plus 4

plus 1 plus 1 plus 1 plus 49 plus 4 plus 100 minus 3.6 square. So, this value can be evaluated.

Now, we look at the value of t on; suppose we want a 90 percent confidence interval. So, we need 0.059 that is equal to 1.833. So, we get the confidence interval as 3.6 plus minus sd by root 10 into 1.833; that will be the confidence interval for the difference in the gripping strengths of the left hand and the right hand of the left handed persons.

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Ex To compare age at marriage of women in two ethnic groups a random sample of 100 women is taken from each group.

$\bar{x} = 18.5, \bar{y} = 20.7, s_1 = 5.8, s_2 = 6.3$

$$S_p^2 = \frac{(m-1)S_1^2 + (n-1)S_2^2}{m+n-2} = \frac{99S_1^2 + 99S_2^2}{198}$$

$$= \frac{(5.8)^2 + (6.3)^2}{2}$$

$\bar{x} - \bar{y} \pm \sqrt{\frac{mn}{mn}} S_p \cdot t_{0.05, 198}$

$(18.5 - 20.7 \pm \frac{\sqrt{200}}{100} \cdot 1.645)$

90%
 $t_{0.05, 198} = 1.645$

Let me take another example here: to compare age at marriage of women in two ethnic groups, a random sample of 100 women is taken and we observed that x bar is equal to 18.5 years, y bar is equal to 20.7 years and s 1 is equal to 5.8, s 2 is equal to 6.3 and we want say a confidence interval for this. So, we calculate here that we may use the model for S p square. So, S p square is equal to m minus 1 s 1 square plus n minus 1 s 2 square divided by m plus n minus 2 that is equal to 99 s 1 square plus 99 s 2 square by 198 that is equal to 5.8 square plus 6.3 square by 2. So, this value can be evaluated.

In a similar way we have the confidence interval as x bar minus y bar minus root m plus n by m n that is and then s p into t alpha by 2; suppose I want 90 percent confidence interval so 0.05 and the degrees of freedom will be m plus n minus 2. So, this value we can see t 0.05 198; just almost as a normal distribution 1645. So, we substitute these values here 18.5 minus 20.7 minus this is 100 plus 100 that is 200 by 100. So, that is root

2 by 10 into 1.645. So, plus minus that gives the confidence interval for the difference in ages at marriage of women in two ethnic groups.

So, here we have you would the pooled formula. We may actually do a testing of hypothesis for sigma s 5 is equal to sigma 2 square and if sigma 1 square is equal to sigma 2 square is accepted then we may go for this formula.

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Two machines are used to fill plastic bottles with dishwashing detergent. The s.d. of fill volume are known to be $\sigma_1 = 0.15$ fluid ounces and $\sigma_2 = 0.12$ fluid ounces for the two machines. Two random samples of $n_1 = 12$ bottles from machine 1 and $n_2 = 10$ bottles from machine 2 are selected and the observations are $\bar{x}_1 = 30.87$, $\bar{x}_2 = 30.68$. Find 90% C.I. for $\mu_1 - \mu_2$.

$$\bar{x} - \bar{y} \pm \sqrt{\frac{\sigma_1^2}{m} + \frac{\sigma_2^2}{n}} z_{0.05}$$

$$30.87 - 30.68 \pm \sqrt{\frac{(0.15)^2}{12} + \frac{(0.12)^2}{10}} (1.645)$$

(0.095, 0.285) is 90% C.I. for $\mu_1 - \mu_2$.

Let me take another example here: two machines are used to fill plastic bottles with dishwashing detergent the standard deviations of fill volume are known to be sigma 1 is equal to 0.15 fluid ounces and sigma 2 is equal to 0.12 fluid ounces, for the two machines. Now two random samples of n 1 is equal to 12 bottles from machine 1 and n 2 is equal to 10 bottles from machine 2 are selected. And the observations are: x 1 bar is equal to 30.87, x bar is equal to 30.68. So, find 95 percent confidence interval for mu 1 minus mu 2.

So here we can see; we can look at the confidence interval as x bar minus y bar plus minus square root sigma 1 square by m plus sigma 2 square by n z 0.05. Now z 0.05 we can see from the tables of normal distribution it is 1.645. So, this interval becomes 30.87 minus 30.68 plus minus square root. Now 1 sigma square is 0.15 square by 12 plus sigma 2 square is 0.12 square by n, n is 10 multiplied by 1.645. So, after simplification this value is turn out to be 0.095 to 0.285. So, this is 90 percent confidence interval for the mean difference that is mu 1 minus mu 2.

So, here the variances were known σ^2_1 and σ^2_2 so we have adopted a procedure where the formula for known variances is utilized here.

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Example. The capacities of batteries (ampere-hours) are distributed as $N(\mu, \sigma^2)$. The records for 10 batteries are 140, 136, 150, 144, 148, 152, 138, 141, 143, 151.

99% C.I. for σ^2

$$\left(\frac{(n-1)s^2}{\chi^2_{0.005,9}}, \frac{(n-1)s^2}{\chi^2_{0.995,9}} \right)$$

$\chi^2_{0.005,9} = 23.59$
 $\chi^2_{0.995,9} = 1.73$

$s^2 = 32.23$
 (12.30, 167.21)

Let me take another example here: the capacities of batteries, so these are measured in say ampere hours they are distributed as normal μ σ^2 . The records for 10 batteries are say 140, 136, 150, 144, 148, 152, 138, 141, 143, 151; we want 99 percent confidence interval for σ^2 .

So, now here we will make use of the fact that μ is unknown. So, if μ is unknown then the formula for confidence interval for σ^2 is based on chi square on $n - 1$ degrees of freedom. The formula is $(n - 1)s^2$ by chi square, so 0.005 n is 10 so this is 9 to $n - 1$ s^2 by chi square 0.005 9; so 0.995.

So, this values we see from the tables of the chi square distribution that is chi square 0.005 9 it is 23.59 and chi square 0.995 on 9 degrees of freedom is 1.73. So, s^2 we calculate here it is turning out to be 32.23. So, after substitution of these values the confidence interval turns out to be 12.30 to 167.21, which is pretty large confidence interval but that will be there because we are considering for σ^2 and the variability of the original sample itself is large; this is s^2 is 32.23 here.

If we reduce the confidence level, suppose we make it 90 percent then this will be shrinking. Since we have made a very a high confidence level that is why the confidence interval is very large which look likely in practical also.

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Comparing Variances

(i) μ_1 and μ_2 are known

$$\frac{\sum (X_i - \mu_1)^2}{\sigma_1^2} \sim \chi_m^2, \quad \frac{\sum (Y_j - \mu_2)^2}{\sigma_2^2} \sim \chi_n^2$$

$$\frac{m \sum (X_i - \mu_1)^2 \sigma_2^2}{n \sigma_1^2 \sum (Y_j - \mu_2)^2} \sim F_{m,n}$$

$U \psi \sim F_{m,n}$

$$P\left(\frac{f_{1-\alpha/2, m, n}}{\psi} \leq U \psi \leq \frac{f_{\alpha/2, m, n}}{\psi}\right) = 1 - \alpha$$

$\psi = \frac{\sigma_2^2}{\sigma_1^2}$

$$U = \frac{m \sum (X_i - \mu_1)^2}{n \sum (Y_j - \mu_2)^2}$$

The graph shows the F-distribution curve with a central area of $1 - \alpha$ bounded by $f_{1-\alpha/2, m, n}$ and $f_{\alpha/2, m, n}$.

Next we look at the confidence intervals for variances, so comparing variances. Again we have two cases that is mu 1 and mu 2 are known, if mu 1 and mu 2 are known then we make use of sigma xi minus mu 1 square by sigma 1 square following chi square distributions on m degrees of freedom and sigma y j minus mu 2 square by sigma 2 square follows chi square distribution on n degrees of freedom. So, if you take the ratios here sigma xi minus mu 1 square by sigma 1 square pi m, so that is m here divided by sigma y j minus mu 2 square by sigma 2 square so that will come in the numerator divided by n. That will have chi square f distribution on m and n degrees of freedom.

So, if we look at this quantity if mu 1 and mu 2 are known then here the ratio sigma 2 square by sigma 1 square is coming. Let us denote it by say psi that is sigma 2 square by sigma 1 square. So, we are having and let me use the notation say U as m sigma xi minus mu 1 square divided by n sigma y j minus mu 2 square. So if we look at this one, then we are having U psi following f distribution on m n degrees of freedom. So, if we make use of the tables of f distributions that is f on m and n degrees of freedom here and f 1 minus alpha by 2 on m n degrees of freedom this is alpha by 2 and this is 1 minus alpha by 2, so this is 1 minus alpha.

So, probability of $f_{1-\frac{\alpha}{2}, m, n}$ less than or equal to $\frac{U}{\psi}$ less than or equal to $f_{\frac{\alpha}{2}, m, n}$ that is equal to $1 - \alpha$.

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$$P\left(f_{1-\frac{\alpha}{2}, m, n} \leq \frac{U}{\psi} \leq f_{\frac{\alpha}{2}, m, n}\right) = 1 - \alpha$$

So a $(1 - \alpha)$ C.I. for σ_1^2 / σ_2^2

$$U f_{1-\frac{\alpha}{2}, m, n} \leq \frac{\sigma_1^2}{\sigma_2^2} \leq U f_{\frac{\alpha}{2}, m, n}$$

$\frac{1}{F_{m, n}} = F_{n, m}$

So, we can write probability of $\frac{U}{\psi}$, so divided by $f_{1-\frac{\alpha}{2}, m, n}$ less than or equal to $\frac{\sigma_1^2}{\sigma_2^2}$ that is $1 - \alpha$ it becomes $\frac{\sigma_1^2}{\sigma_2^2}$ less than or equal to $U f_{1-\frac{\alpha}{2}, m, n}$; that is equal to $1 - \alpha$. So, we have a $1 - \alpha$ confidence interval for $\frac{\sigma_1^2}{\sigma_2^2}$. This can also be written as $U f_{1-\frac{\alpha}{2}, n, m}$ to $U f_{\frac{\alpha}{2}, n, m}$ by using the reciprocal property of the F distribution. Because we know that $\frac{1}{F_{m, n}}$ is equal to $F_{n, m}$, so this property can be utilized here.

Let me give one example here for confidence interval for the ratios.

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Example. Two brands of cough medicine are given and the response times are measured in days. COPY SLIDE

$\mu_1 = 3, \mu_2 = 5$

$m = 10, n = 12, s = 1$

$x \rightarrow 2 \ 3 \ 2 \ 4 \ 2 \ 5 \ 6 \ 7 \ 1 \ 2$

$y \rightarrow 3 \ 4 \ 6 \ 8 \ 3 \ 2 \ 9 \ 5 \ 11 \ 7 \ 2 \ 1$

$\bar{x} = \dots, \bar{y} = \dots$

$\frac{\sum (x_i - \mu_1)^2}{\sigma_1^2} \sim \chi_9^2$

$\frac{\sum (y_j - \mu_2)^2}{\sigma_2^2} \sim \chi_{11}^2$

$F_{0.05, 10, 12}$

So, two brands of say cough medicine are given and the response times are measured in days. So, here we are having the data m is equal to say 10, n is equal to 12, $s = 1$ and we are getting the observations as x is equal to say 2 3 2 4 2 5 6; so 7 and then 1 2. So, we have 10 data here and for y we have the data say 3 4 6 8 3 2 9 5 11 7 2 1. Now based on this we calculate \bar{x} , \bar{y} , and we calculate $\sum (x_i - \mu_1)^2$. So, it is given that μ_1 is say 3 and μ_2 is equal to 5. So, if we are looking at $\sum (x_i - \mu_1)^2$ divided by σ_1^2 then that will follow chi square on 9 and this divided by $\sum (y_j - \mu_2)^2$ divided by σ_2^2 will follow chi square on 11 degrees of freedom.

So, we can construct $\frac{\sum (x_i - \mu_1)^2}{\sigma_1^2}$ by $\frac{\sum (y_j - \mu_2)^2}{\sigma_2^2}$. And then we need to look at the tables of F on say 0.05 10 and 12 degrees of freedom.

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(ii) μ_1 and μ_2 are unknown.

$$\frac{(m-1)S_1^2}{\sigma_1^2} \sim \chi_{m-1}^2, \quad \frac{(n-1)S_2^2}{\sigma_2^2} \sim \chi_{n-1}^2$$

indep.

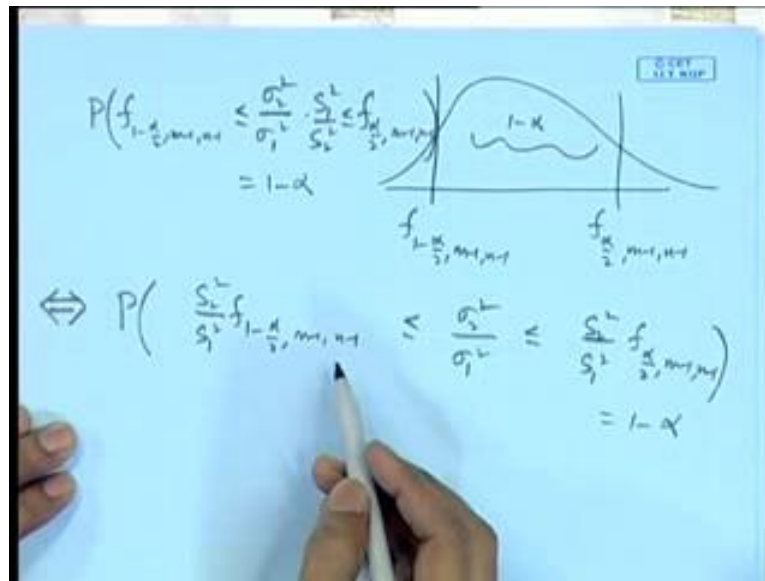
$$\frac{\frac{(m-1)S_1^2}{\sigma_1^2(m-1)}}{\frac{(n-1)S_2^2}{\sigma_2^2(n-1)}} \sim F_{m-1, n-1}$$

$$\Rightarrow \frac{\sigma_2^2}{\sigma_1^2} \frac{S_1^2}{S_2^2} \sim F_{m-1, n-1}$$

Now, another situation may occur when μ_1 and μ_2 are unknown. If μ_1 and μ_2 are unknown then we will be not able to make use of the formula that we derived earlier because they are in the confidence interval μ_1 and μ_2 are actually appearing. So, what we do we make use of s_1^2 and s_2^2 , we have $m-1$ s_1^2 follows chi square distribution on $m-1$ degrees of freedom and $n-1$ s_2^2 by σ_2^2 follows chi square distribution on $n-1$ degree of freedom.

Furthermore, these two random variables are independent. So, we can make use of the ratios $\frac{m-1}{\sigma_1^2} \frac{s_1^2}{m-1}$ divided by $\frac{n-1}{\sigma_2^2} \frac{s_2^2}{n-1}$ that will follow F distribution on $m-1$ $n-1$ degree of freedom; which is reducing to $\frac{\sigma_2^2}{\sigma_1^2} \frac{s_1^2}{s_2^2}$ this follows F distribution on $m-1$ $n-1$ degrees of freedom.

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So, making the use of distribution of f that is we have $f_{1-\alpha/2, m, n}$ and $f_{\alpha/2, m, n}$, intermediate probability $1-\alpha$. So, probability that $f_{1-\alpha/2, m, n} \leq \frac{\sigma_1^2}{\sigma_2^2} \cdot \frac{S_1^2}{S_2^2} \leq f_{\alpha/2, m, n}$ is less than are equal to $1-\alpha$. So, we make use of this and adjust the coefficient as probability that $\frac{S_1^2}{S_2^2} f_{1-\alpha/2, m, n} \leq \frac{\sigma_1^2}{\sigma_2^2} \leq \frac{S_1^2}{S_2^2} f_{\alpha/2, m, n}$ is less than are equal to $1-\alpha$.

So, we are getting $100(1-\alpha)$ percent confidence interval for $\frac{\sigma_1^2}{\sigma_2^2}$. We can reverse if we want for $\frac{\sigma_2^2}{\sigma_1^2}$ then we interchange the roles here we put $\frac{\sigma_2^2}{\sigma_1^2}$ and the degrees of freedom will get reverse it will become $n-1, m-1$.

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Handwritten derivation on a blue background:

$$\Leftrightarrow P\left(\frac{\sum_{i=1}^m f_{1-\frac{\alpha}{2}, m, n-1}}{s_1^2} \leq \frac{\sigma_2^2}{\sigma_1^2} \leq \frac{\sum_{i=1}^m f_{\frac{\alpha}{2}, m, n-1}}{s_2^2}\right) = 1 - \alpha$$

So $\left(\frac{\sum_{i=1}^m f_{1-\frac{\alpha}{2}, m, n-1}}{s_1^2}, \frac{\sum_{i=1}^m f_{\frac{\alpha}{2}, m, n-1}}{s_2^2}\right)$ is $100(1-\alpha)\%$ C.I. for σ_2^2/σ_1^2 .

We give one example here. So, s_2^2 square by s_1^2 square $f_{1-\alpha/2, m, n-1}$ to s_2^2 square by s_1^2 square $f_{\alpha/2, m, n-1}$ is $100(1-\alpha)$ percent confidence interval for σ_2^2 square by σ_1^2 square.

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Handwritten example on a blue background:

Viscosity of two brands of oil used in cars is measured and the following data is recorded:

Brand 1: 10.62 10.58 10.33 10.72 10.44

Brand 2: 10.50 10.52 10.62 10.53

$$s_1^2 = 0.02362 \quad \frac{s_2^2}{s_1^2} = 8.36$$

$$s_2^2 = 0.02825$$

$$f_{0.05, 4, 3} = 9.1172$$

$$f_{0.95, 4, 3} = 0.1517$$

90% C.I. for $\sigma_2^2/\sigma_1^2 = (8.36 \times 0.1517, 0.2649 \times 9.1172)$

So, say viscosity of two brands of oil used in car is measured and the following data is recorded. So, from brand 1 you have 10.62 10.58 10.33 10.72 10.44; for brand 2 it is 10.50 10.52 10.62 10.53. Suppose we want a confident interval for σ_2^2 square by σ_1^2 square. So, we will calculate the values here s_1^2 square s_2^2 square. So, s_1^2 square

turns out be 0.02362, s_2^2 is equal to 0.002825. You can see here there is a 10 times difference here. So, the f value that s_2^2 by or you can say s_1^2 by s_2^2 will be equal to 8.36.

So, if we look at the f value on 0.05 say 1 2 3 4 5, so 4 three degrees of freedom that is equal to 9.1172 and f value 0.9543 that is equal to 0.1517. So, a 90 percent confident interval for σ_1^2 by σ_2^2 that will be equal to 8.36 into 0.1517 to 8.36 into 9.1172. So, this is the confidence interval for the ratio of the variances here.

So, in a given practical situation we need to analyze that what is the model that will be applicable and accordingly we make use of the formulae. So for example, when we are looking at the confidence interval for $\mu_1 - \mu_2$ then we worry about that what is the status of the variances. If the variances are known then we have some formula, if that is based on the z that is normal distribution. If we have variance is unknown but equal then we have a formula which is based on a t distribution based on the pooling of the concept, pooling the variances. And if we have variances to be completely unknown then in that case we have another approximate t distribution formula and we make use of that.

On the other hand if the data is correlated then we make use of pairing and pair t formula is used. Similarly when we are worry about the confidence interval for the σ_1^2 and σ_2^2 then we look at the knowledge about the means. If the means are known then we have a formula based on f distributions on the total degree of freedom m and n. If the means are unknown then we have another formula which is based on s_1^2 and s_2^2 and the degree of freedom are slightly reduce to m minus 1 n minus 1.

Now, these formulae are quite standard because they are making use of sampling distributions from the normal populations. When we do not have normal populations then in that case we may have to look for appropriate sampling distribution; for example if we are dealing with uniform distribution, if we are dealing with from exponential populations, then we look at from the description that what is the sufficient statistics from where we find out the pivoting quantity if we are able to derive the sampling distribution of that.

So, the techniques for that and also for the propulsions are available and one can work out various formulae for confidence interval from other populations as well. So, that is part of another course that is statistical inference that will be doing later on.