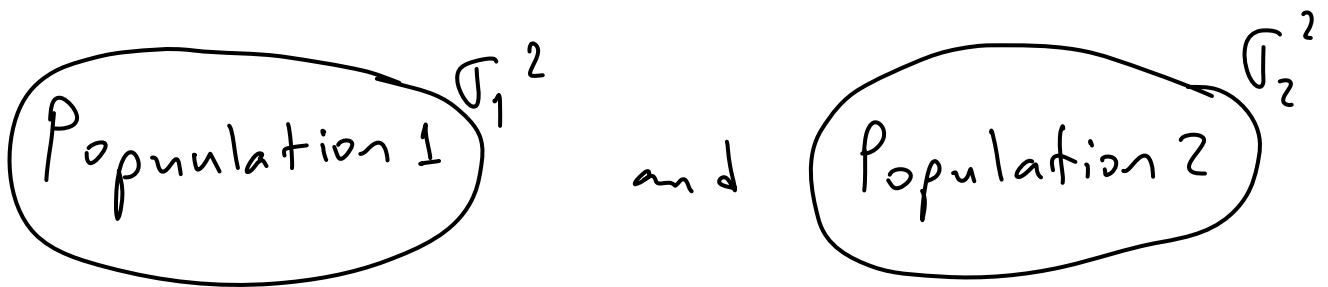


Testing the Equality of Variances (F-Distribution)

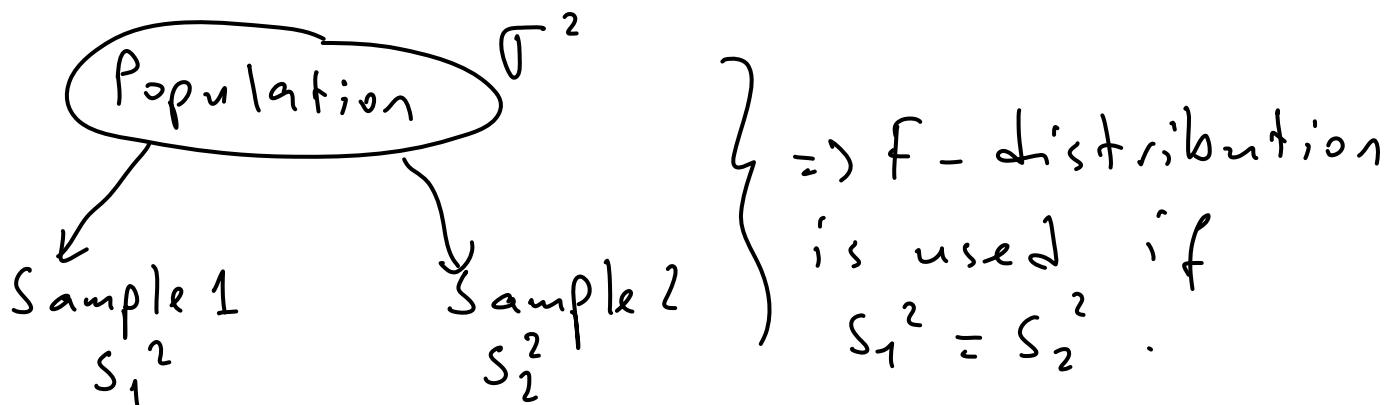
The F-distribution can be used to test the hypothesis that the variances σ_1^2 and σ_2^2 of two normally distributed populations are equal.

It is not necessary to assume that the two populations have equal means.

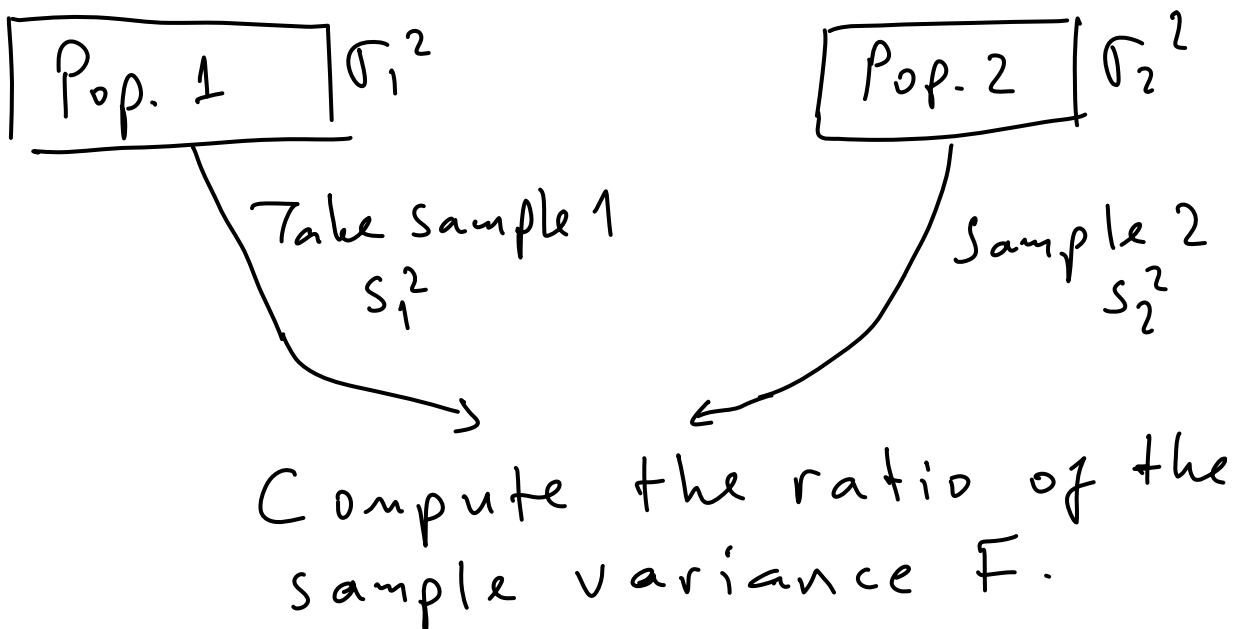


F-distribution is used to test if $\sigma_1^2 = \sigma_2^2$ or $(\sigma_1 = \sigma_2)$.

OR



In general



⊗ When independent random samples have been drawn from the respective populations, the ratio

$F = \frac{S_1^2}{S_2^2}$ could be used to make inferences about the ratio σ_1^2 / σ_2^2 .

$$F_{\text{calc}} = \frac{\text{Larger sample variance}}{\text{Smaller sample variance}}$$

V_1 : The numerator degrees of freedom.

$$V_1 = n_1 - 1$$

V_2 : The denominator degrees of freedom

$$V_2 = n_2 - 1$$

⊗ If the null hypothesis (H_0) were true ($\sigma_1^2 = \sigma_2^2$), then, we would expect the ratio of the sample variances to be close to 1.0.

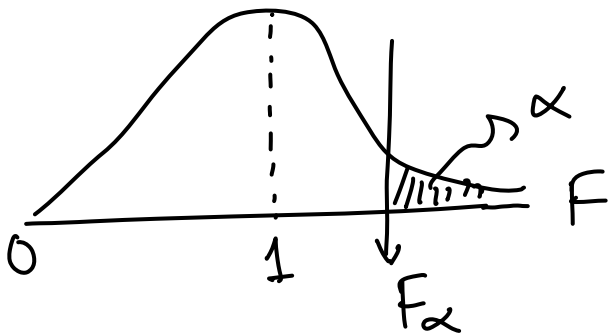
$$F = \frac{S_1^2}{S_2^2} \approx 1 \Rightarrow \sigma_1^2 = \sigma_2^2 : H_0$$

⊗ If $\frac{S_1^2}{S_2^2} \approx 0$ or $\frac{S_1^2}{S_2^2} \gg 1.0 \Rightarrow$

the population variances are different ($\sigma_1^2 \neq \sigma_2^2$)
 i.e., reject H_0 .

Properties of F-Distribution

- 1) Unlike t or Z, F can assume only positive values.
- 2) It is nonsymmetrical
- 3) The F values are fixed from tables by α , ν_1 , ν_2 .
- 4) There are different F tables for various α values ($F_{0.10}$, $F_{0.05}$, $F_{0.025}$, $F_{0.01}$ tables)
- 5) Mean for F-distribution is 1.0



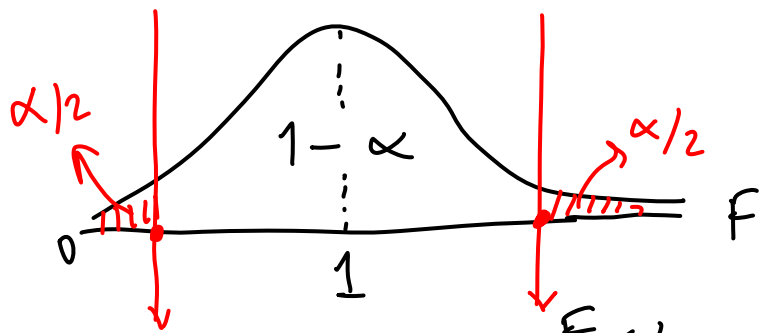
Confidence Interval For $F = \frac{\sigma_1^2}{\sigma_2^2}$

$$P \left[F_{(1 - \frac{\alpha}{2})}(\nu_1, \nu_2) < F < F_{\alpha/2}(\nu_1, \nu_2) \right] = 1 - \alpha$$

cannot be fixed (read) from F tables.

can be fixed (read) from the F tables easily.

⇒ Modification is needed.



$$F(1 - \frac{\alpha}{2})$$

↓

$$F(1 - \frac{\alpha}{2})(v_1, v_2)$$

$F_{\alpha/2}(v_1, v_2) =$ upper critical limit (point).

↓

$$\frac{1}{F_{\alpha/2}(v_2, v_1)} = \text{lower critical limit (point)}$$

$$F_{\alpha/2}(v_2, v_1)$$

↓

new numerator d.f.

↪ new denominator d.f.

$$\frac{1}{F_{\alpha/2}(v_2, v_1)} < F < F_{\alpha/2}(v_1, v_2) : CI$$

multiply both sides by $F_{\alpha/2}(v_2, v_1) =$

$$1 < F < F_{\alpha/2}(v_1, v_2) \times F_{\alpha/2}(v_2, v_1) \text{ and}$$

divide both sides by $F_{\alpha/2}(v_1, v_2) =$

$$\frac{1}{F_{\alpha/2}(V_1, V_2)} < F < F_{\alpha/2}(V_2, V_1)$$

multiply both sides by $\frac{s_1^2}{s_2^2} \Rightarrow$

CI: $\frac{s_1^2}{s_2^2} \times \frac{1}{F_{\alpha/2}(V_1, V_2)} < F < \frac{s_1^2}{s_2^2} \times F_{\alpha/2}(V_2, V_1)$

σ_1^2 / σ_2^2

Example: Determine the upper and lower critical limits for an F distribution with $V_1=6, V_2=3$ and $\alpha=0.10$.

Solution: $\alpha=0.10 \Rightarrow F_{\alpha/2} = F_{0.10/2} = F_{0.05} \Rightarrow$

Use $F_{0.05}$ table

denominator d.f. V_2	V_1 (numerator d.f.)						
	1	2	3	4	5	6	7 ...
1							
2							
3						8.94	
4							
5							
6			4.76				
7							
⋮							

Upper limit: $F_{\alpha/2}(V_1, V_2) = F_{0.05}(6, 3) = 8.94$

$$\text{Lower limit: } \frac{1}{F_{\alpha/2}(v_2, v_1)} = \frac{1}{F_{0.05}(3, 6)} = \frac{1}{4.76} = 0.21$$

Example: What is $F_{0.95}(6, 10)$ value ?

Solution: $F_{0.95}(6, 10) \equiv \frac{1}{F_{0.05}(10, 6)} = \frac{1}{4.06} = 0.246$

$\xrightarrow{1.0}$
 num. d.f. den. d.f.

$F_{0.05}$ table \Rightarrow

↓ v_2	...	10	...
⋮			
⋮		↓	
⋮		4.06	
⋮		←	
⋮		6	

Example: Assume we have 25 boy and 16 girl students. Average grade for the boys is 82 with a standard deviation of 8 and average grade for the girls is 78 with a standard deviation of 7. Find 98 % confidence interval for $\frac{\sigma_1^2}{\sigma_2^2}$ and $\frac{\sigma_1}{\sigma_2}$.

Solution: $\alpha = 0.02$, $n_1 = 25$, $n_2 = 16$, $s_1 = 8$, $s_2 = 7$

$F_{\alpha/2} = F_{0.02/2} = F_{0.01} \Rightarrow$ Use $F_{0.01}$ table.

$$v_1 = n_1 - 1 = 25 - 1 = 24$$

$$v_2 = n_2 - 1 = 16 - 1 = 15$$

$$F_{\alpha/2}(v_1, v_2) = F_{0.01}(24, 15) = 3.29$$

$$F_{\alpha/2}(v_2, v_1) = F_{0.01}(15, 24) = 2.89$$

98% CI \Rightarrow

$$\frac{64}{49} \cdot \left(\frac{1}{3.29}\right) < \frac{\sigma_1^2}{\sigma_2^2} < \frac{64}{49} \times (2.89)$$

$$0.397 < \frac{\sigma_1^2}{\sigma_2^2} < 3.775 \rightarrow \text{CI for } \frac{\sigma_1^2}{\sigma_2^2}$$

take $\sqrt{\quad}$ of both sides \Rightarrow

$$0.630 < \frac{\sigma_1}{\sigma_2} < 1.943 \rightarrow \text{CI for } \frac{\sigma_1}{\sigma_2}$$

Hypothesis Testing Using F-Distribution

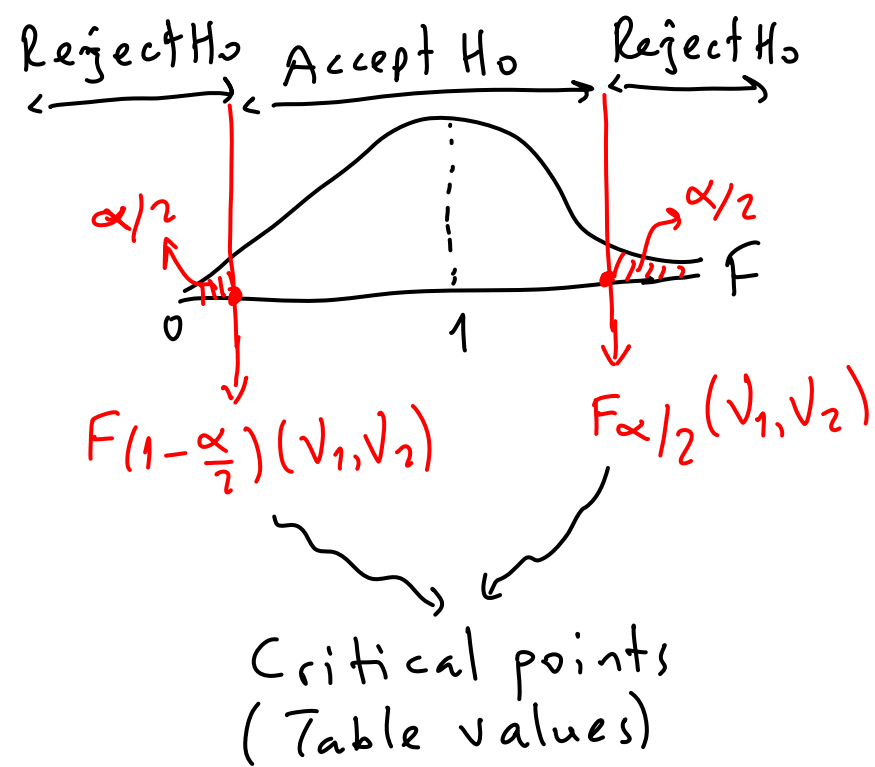
a) Two-Tailed Test

$$H_0: \frac{\sigma_1^2}{\sigma_2^2} = 1 \text{ OR } \sigma_1^2 = \sigma_2^2$$

$$H_A: \frac{\sigma_1^2}{\sigma_2^2} \neq 1 \text{ OR } \sigma_1^2 \neq \sigma_2^2$$

Test statistic: f_{calc}

$$f_{\text{calc}} = \frac{\text{Larger Variance}}{\text{Smaller Variance}}$$



$$F_{calc} = \frac{S_1^2}{S_2^2} \text{ when } S_1^2 > S_2^2$$

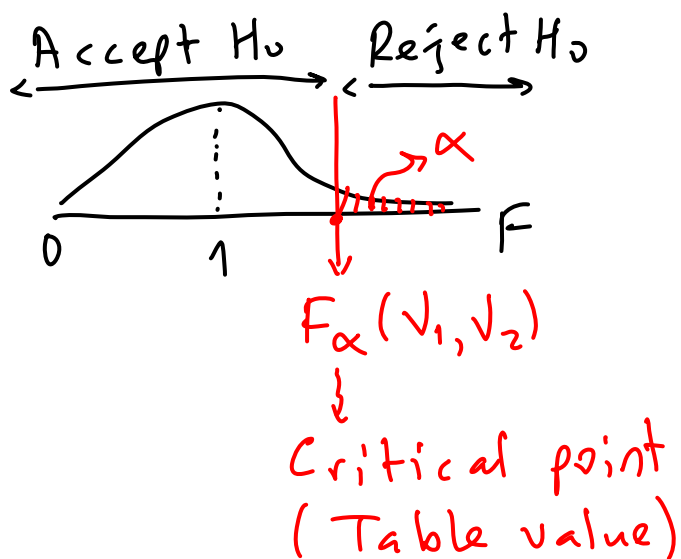
OR

$$F_{calc} = \frac{S_2^2}{S_1^2} \text{ when } S_2^2 > S_1^2$$

b) One-Tailed Test (Right Hand Sided)

$$H_0: \frac{\sigma_1^2}{\sigma_2^2} = 1 \text{ or } \sigma_1^2 = \sigma_2^2$$

$$H_A: \frac{\sigma_1^2}{\sigma_2^2} > 1 \text{ or } \sigma_1^2 > \sigma_2^2$$



Test statistic value: F_{calc}

$$F_{calc} = \frac{\text{Larger Variance}}{\text{Smaller Variance}}$$

$$F_{\text{calc}} = \frac{S_1^2}{S_2^2} \text{ when } S_1^2 > S_2^2 \text{ or}$$

$$F_{\text{calc}} = \frac{S_2^2}{S_1^2} \text{ when } S_2^2 > S_1^2$$

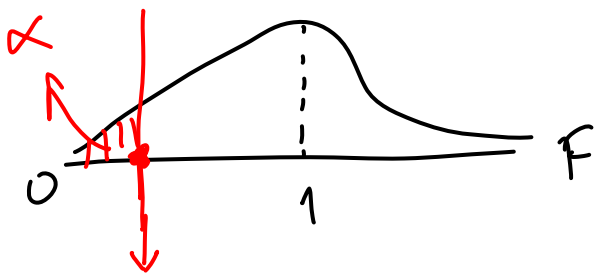
c) One-Tailed Test (Left Hand Sided)

$$H_0: \frac{\sigma_1^2}{\sigma_2^2} = 1 \text{ or } \sigma_1^2 = \sigma_2^2$$

$$H_A: \frac{\sigma_1^2}{\sigma_2^2} < 1 \text{ or } \sigma_1^2 < \sigma_2^2$$

Test statistic value: F_{calc}

$$F_{\text{calc}} = \frac{\text{Smaller Variance}}{\text{Larger Variance}}$$



$$F_{(1-\alpha)}(v_1, v_2)$$

↓
Critical point

$$F_{(1-\alpha)}(v_1, v_2) \equiv \frac{1}{F_{\alpha}(v_2, v_1)}$$

Example: If $s_1^2 = 48.7$, $n_1 = 5$, $s_2^2 = 3.7$ and $n_2 = 6$, then, test whether the two populations have the same variabilities at $\alpha = 0.10$.

Solution: Two-tailed test.

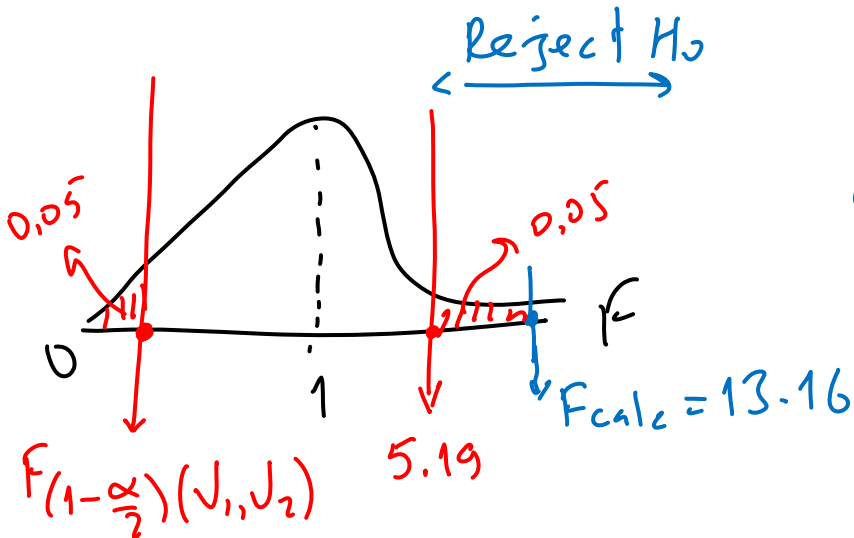
$\times H_0: \sigma_1^2 = \sigma_2^2$ OR $S_1^2 = S_2^2$

$\checkmark H_A: \sigma_1^2 \neq \sigma_2^2$ OR $S_1^2 \neq S_2^2$

$$F_{calc} = \frac{\text{Larger Var.}}{\text{Smaller Var.}} = \frac{48.7}{3.7} = 13.16$$

Table values: $V_1 = 5 - 1 = 4$, $V_2 = 6 - 1 = 5$

$$F_{\alpha/2}(V_1, V_2) = F_{0.05}(4, 5) = 5.19$$



Decision: Reject H_0

Conclusion: The two populations have different variabilities.