

lecture6: t and F-distribution

20 Mart 2025 Perşembe 11:23

t-distribution

Example 8.11: A chemical engineer claims that the population mean yield of a certain batch process is 500 grams per milliliter of raw material. To check this claim he samples 25 batches each month. If the computed t-value falls between $-t_{0.05}$ and $t_{0.05}$, he is satisfied with this claim. What conclusion should he draw from a sample that has a mean $\bar{x} = 518$ grams per milliliter and a sample standard deviation $s = 40$ grams? Assume the distribution of yields to be approximately normal.

$H_0: \mu = 500$
 $t = \frac{\bar{x} - \mu}{\frac{s}{\sqrt{n}}} = \frac{518 - 500}{\frac{40}{\sqrt{25}}} = 2.25$

Since $t = 2.25 > t_{0.05}$, the claim is not true.

$v = n - 1 = 24$

$$t = \frac{\bar{x}_1 - \bar{x}_2 - (\mu_1 - \mu_2)}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}}$$

Population I

$$\mu_1$$
$$\sigma_1^2$$

Population II

$$\mu_2$$
$$\sigma_2^2$$

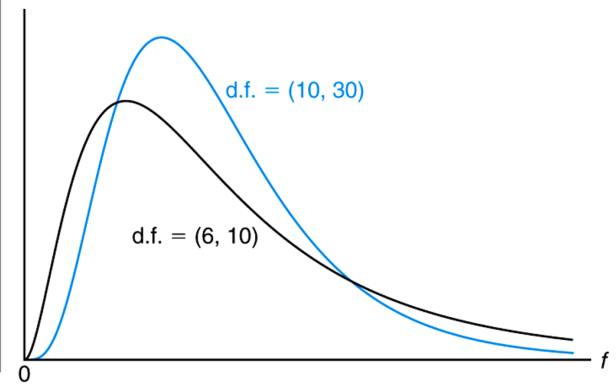
$$\bar{x}_1 - \bar{x}_2 \sim N(\mu_1 - \mu_2, \sigma_{\bar{x}_1 - \bar{x}_2}^2)$$
$$\sigma_{\bar{x}_1 - \bar{x}_2}^2 = \frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}$$
$$\text{var}(\bar{x}_1 - \bar{x}_2) = \text{var}(\bar{x}_1) + \text{var}(\bar{x}_2) = \frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}$$

F-distribution

Let U and V be two independent random variables having chi-squared distributions with v_1 and v_2 degrees of freedom, respectively. Then the distribution of the random variable $F = \frac{U/v_1}{V/v_2}$ is given by the density function

$$h(f) = \begin{cases} \frac{\Gamma((v_1+v_2)/2) \Gamma(v_1/2) \Gamma(v_2/2)}{\Gamma(v_1/2) \Gamma(v_2/2)} \frac{f^{(v_1/2)-1}}{(1+v_1 f/v_2)^{(v_1+v_2)/2}}, & f > 0, \\ 0, & f \leq 0. \end{cases}$$

This is known as the **F-distribution** with v_1 and v_2 degrees of freedom (d.f.).



Writing $f_\alpha(v_1, v_2)$ for f_α with v_1 and v_2 degrees of freedom, we obtain

$$f_{1-\alpha}(v_1, v_2) = \frac{1}{f_\alpha(v_2, v_1)}$$

therefore, we can find $f_{0.95}$ from $\alpha = 0.05$ Table

$f_{0.95}(8, 9) = \frac{1}{f_{0.05}(9, 8)} = \frac{1}{3.39}$

$a = f_{1-\alpha}(v_1, v_2)$
 $b = f_\alpha(v_1, v_2)$
assume they are symmetrical

If S_1^2 and S_2^2 are the variances of independent random samples of size n_1 and n_2 taken from normal populations with variances σ_1^2 and σ_2^2 , respectively, then

$$F = \frac{S_1^2/\sigma_1^2}{S_2^2/\sigma_2^2} = \frac{\sigma_2^2 S_1^2}{\sigma_1^2 S_2^2}$$

has an F -distribution with $v_1 = n_1 - 1$ and $v_2 = n_2 - 1$ degrees of freedom.

Example

Let S_1^2 denote the sample variance for a random sample of size 10 from Population I and let S_2^2 denote the sample variance for a random sample of size 8 from Population II. The variance of Population I is assumed to be three times the variance of Population II. Find two numbers a and b such that $P(a \leq S_1^2/S_2^2 \leq b) = 0.90$ assuming S_1^2 to be independent of S_2^2 .

Population I

$$n = 10$$
$$\sigma_1^2$$
$$S_1^2$$

Population II

$$n = 8$$
$$\sigma_2^2 = 3\sigma_1^2$$
$$S_2^2$$

$P(\frac{a}{3} \leq \frac{S_1^2}{S_2^2} \cdot \frac{1}{3} \leq \frac{b}{3}) = 0.9$
 $P(\frac{a}{3} \leq F \leq \frac{b}{3}) = 0.9$
 $f_{0.05}(9, 7) = \frac{a}{3}$
 $f_{0.05}(9, 7) = \frac{b}{3}$
 3.68
 $3 \times 3.68 = b \rightarrow b = 11.04$
 $\frac{1}{3.29} = \frac{1}{f_{0.05}(7, 9)} = f_{0.95}(7, 9) = \frac{a}{3}$
 $a = 0.91$

Cutoff points for the F Distribution ($\alpha = 0.01$)

	1	2	3	4	5	6	7	8	9	10	12	15
1	4052	5000	5403	5625	5764	5859	5928	5981	6022	6056	6106	6157
2	98.50	99.00	99.17	99.25	99.30	99.33	99.36	99.37	99.39	99.40	99.42	99.43
3	34.12	30.82	29.46	28.71	28.24	27.91	27.67	27.49	27.35	27.23	27.05	26.87
4	21.20	18.00	16.69	15.98	15.52	15.21	14.98	14.80	14.66	14.55	14.37	14.20
5	16.26	13.27	12.06	11.39	10.97	10.67	10.46	10.29	10.16	10.05	9.89	9.72
6	13.75	10.92	9.78	9.15	8.75	8.47	8.26	8.10	7.98	7.87	7.72	7.56
7	12.25	9.55	8.45	7.85	7.46	7.19	6.99	6.84	6.72	6.62	6.47	6.31
8	11.26	8.65	7.59	7.01	6.63	6.37	6.18	6.03	5.91	5.81	5.67	5.52
9	10.56	8.02	6.99	6.42	6.06	5.80	5.61	5.47	5.35	5.26	5.11	4.96
10	10.04	7.56	6.55	5.99	5.64	5.39	5.20	5.06	4.94	4.85	4.71	4.56
11	9.65	7.21	6.22	5.67	5.32	5.07	4.89	4.74	4.63	4.54	4.40	4.25
12	9.33	6.93	5.95	5.41	5.06	4.82	4.64	4.50	4.39	4.30	4.16	4.01

$\alpha = P(F > 4.89) = f_\alpha(7, 11)$

Example

If S_1^2 and S_2^2 represent the variances of independent random samples of size $n_1 = 8$ and $n_2 = 12$, taken from normal populations with equal variances, find $P(S_1^2/S_2^2 < 4.89)$

$F = \frac{\sigma_1^2 S_1^2}{\sigma_1^2 S_2^2} = \frac{S_1^2}{S_2^2}$
since equal

$P(F_{v_1, v_2} < 4.89) = ?$
 $v_1 = 7, v_2 = 11$
 $n_1 = 8, n_2 = 12$
 $P(\frac{S_1^2}{S_2^2} < 4.89) = ?$
 $\sigma_1^2 = \sigma_2^2$

Since this value is exactly at $\alpha = 0.01$ table, we use that table

$F > 4.89 \Rightarrow f_\alpha(7, 11) \quad \alpha = 0.01$
 $f_{1-\alpha}(7, 11) \Rightarrow \alpha = 0.99$

example: (chi-square)

For a chi-square distribution, find χ_α^2 such that;

* $P(19.68 < x^2 < \chi_\alpha^2) = 0.04$, when $v = 11$

