

# ISE 315: Engineering Statistics

*Supplementary Material: Finding P-Value Bounds from Statistical Tables*

Instructor: Mansur M. Arief, PhD  
Industrial and Systems Engineering, KFUPM

Office: 22-219 — Email: [mansur.arief@kfupm.edu.sa](mailto:mansur.arief@kfupm.edu.sa)

*Based on Montgomery & Runger, Applied Statistics and Probability for Engineers, 6th Ed.*

# Supplementary Material

Finding P-Value Bounds from the  $t$ ,  $\chi^2$ , and  $F$  Tables

## Why Do We Need P-Value Bounds?

---

Software gives **exact p-values** (e.g.,  $p = 0.0297$ ), but statistical tables only give **critical values** at specific  $\alpha$  levels.

From tables, we can only **bound** the p-value between two  $\alpha$  levels.

On exams and quizzes, you are expected to find p-value bounds using:

- The  **$t$ -table** (Appendix Table V)
- The  **$\chi^2$ -table** (Appendix Table IV)
- The  **$F$ -table** (Appendix Table VI)

**Key idea:** Locate your test statistic between two critical values, then read off the corresponding  $\alpha$  values as your bounds.

## The General Strategy: Three Steps

---

### Step 1: Identify the correct table row/section

- For  $t$ : Go to the row matching your degrees of freedom
- For  $\chi^2$ : Go to the row matching your degrees of freedom
- For  $F$ : Go to the page matching your  $\alpha$ , then find  $(df_1, df_2)$

### Step 2: Bracket your test statistic between two critical values

- Find  $t_{\alpha_1} < |t_0| < t_{\alpha_2}$  (or similarly for  $\chi^2, F$ )

### Step 3: Convert to a p-value bound

- **One-tailed:**  $\alpha_2 < p\text{-value} < \alpha_1$
- **Two-tailed:**  $2\alpha_2 < p\text{-value} < 2\alpha_1$

## Part 1: P-Value Bounds from the $t$ -Table

---

### How the $t$ -table is organized:

- Rows = degrees of freedom ( $\nu = n - 1$ )
- Columns = upper-tail area  $\alpha$ : 0.25, 0.10, 0.05, 0.025, 0.01, 0.005
- Entry =  $t_{\alpha,\nu}$  such that  $P(T > t_{\alpha,\nu}) = \alpha$

### Example: Row for $\nu = 14$ :

$\nu \backslash \alpha$	0.25	0.10	0.05	0.025	0.01	0.005
14	0.692	1.345	1.761	2.145	2.624	2.977

Reading:  $P(T_{14} > 2.145) = 0.025$  and  $P(T_{14} > 2.624) = 0.01$ .

## $t$ -Table Example 1

---

A petroleum engineer measures the hardness (HRC) of  $n = 15$  drill bits from a new supplier. The sample gives  $\bar{x} = 52.0$  and  $s = 3.2$ . Test whether the mean hardness differs from the specification of  $\mu_0 = 50.0$  at  $\alpha = 0.05$ . Estimate the  $p$ -value.

**Hypotheses:**  $H_0 : \mu = 50$  vs  $H_1 : \mu \neq 50$ . Two-tailed  $t$ -test,  $df = 14$ .

**Test statistic:**  $t_0 = \frac{52.0 - 50.0}{3.2/\sqrt{15}} = \frac{2.0}{0.826} = 2.42$

**P-value bounds:** Row  $\nu = 14$ :  $t_{0.025,14} = 2.145 < 2.42 < 2.624 = t_{0.01,14}$ .

Upper tail:  $0.01 < P(T > 2.42) < 0.025$ . Two-tailed:  $0.02 < p\text{-value} < 0.05$

**Decision:** Since  $p < 0.05 = \alpha$ , we **reject**  $H_0$ . The mean hardness differs from spec.

## *t*-Table Example 2

---

A chemical engineer claims a new catalyst increases mean yield above 90 units. A sample of  $n = 20$  batches gives  $\bar{x} = 91.5$  and  $s = 3.5$ . Test at  $\alpha = 0.05$ . Estimate the  $p$ -value.

**Hypotheses:**  $H_0 : \mu = 90$  vs  $H_1 : \mu > 90$ . One-tailed upper  $t$ -test,  $df = 19$ .

**Test statistic:**  $t_0 = \frac{91.5 - 90.0}{3.5/\sqrt{20}} = \frac{1.5}{0.783} = 1.92$

**P-value bounds:** Row  $\nu = 19$ :  $t_{0.05,19} = 1.729 < 1.92 < 2.093 = t_{0.025,19}$ .

One-tailed:  $0.025 < p\text{-value} < 0.05$

**Decision:** At  $\alpha = 0.05$ :  $p < 0.05$ , so **reject**  $H_0$ . Evidence supports the claim.

At  $\alpha = 0.01$ :  $p > 0.01$ , so **fail to reject**. Not significant at 1%.

## Key Difference: One-Tailed vs. Two-Tailed from the $t$ -Table

---

---

### One-Tailed Test

Read  $\alpha$  values directly from the column headers bracketing  $|t_0|$

If  $t_{0.05} < |t_0| < t_{0.025}$ :

$\Rightarrow 0.025 < p < 0.05$

### Two-Tailed Test

Double the column headers to get the  $p$ -value bounds

If  $t_{0.025} < |t_0| < t_{0.01}$ :

$\Rightarrow 0.02 < p < 0.05$

---

**This is the #1 mistake on exams:** forgetting to multiply by 2 for two-tailed tests, or multiplying when the test is one-tailed.

## Part 2: P-Value Bounds from the $\chi^2$ -Table

---

### How the $\chi^2$ -table is organized:

- Rows = degrees of freedom ( $\nu = n - 1$ )
- Columns = **upper-tail area**  $\alpha$
- The table includes **both small and large**  $\alpha$  values:

$0.995, 0.99, 0.975, 0.95, 0.90$        $0.10, 0.05, 0.025, 0.01, 0.005$   
These give **left-tail** critical values      These give **right-tail** critical values

### Important difference from the $t$ -table:

- The  $\chi^2$  distribution is **not symmetric**
- For **lower-tail** tests, use columns with large  $\alpha$  values (0.95, 0.975, ...)
- For **upper-tail** tests, use columns with small  $\alpha$  values (0.05, 0.025, ...)

## Reading the $\chi^2$ -Table: Upper vs. Lower Tail

---

**Example row:**  $\nu = 19$

$\nu \backslash \alpha$	Left-tail critical values					Right-tail critical values				
	0.995	0.99	0.975	0.95	0.90	0.10	0.05	0.025	0.01	0.005
19	6.844	7.633	8.907	10.117	11.651	27.204	30.144	32.852	36.191	38.582

**Key readings:**

- $\chi_{0.05,19}^2 = 30.14$  means  $P(\chi_{19}^2 > 30.14) = 0.05$  (right tail)
- $\chi_{0.95,19}^2 = 10.12$  means  $P(\chi_{19}^2 > 10.12) = 0.95$ , so  $P(\chi_{19}^2 < 10.12) = 0.05$  (left tail)

## $\chi^2$ -Table Example 1

---

A bottling plant specifies  $\sigma^2 = 0.020 \text{ mL}^2$  for fill volume. A sample of  $n = 20$  bottles gives  $s^2 = 0.032 \text{ mL}^2$ . Test whether the variance differs from the standard at  $\alpha = 0.05$ . Estimate the p-value.

**Hypotheses:**  $H_0 : \sigma^2 = 0.020$  vs  $H_1 : \sigma^2 \neq 0.020$ . Two-tailed  $\chi^2$ -test,  $df = 19$ .

**Test statistic:**  $\chi_0^2 = \frac{(n-1)s^2}{\sigma_0^2} = \frac{19 \times 0.032}{0.020} = 30.4$  (falls in upper tail)

**P-value bounds:** Row  $\nu = 19$ :  $\chi_{0.05,19}^2 = 30.14 < 30.4 < 32.85 = \chi_{0.025,19}^2$ .

Upper tail:  $0.025 < P(\chi^2 > 30.4) < 0.05$ . Two-tailed:  $0.05 < p\text{-value} < 0.10$

**Decision:** Since  $p > 0.05 = \alpha$ , we **fail to reject**  $H_0$ . Not enough evidence.

## $\chi^2$ -Table Example 2

---

A manufacturer claims a new CNC machine achieves lower variance than the current standard of  $\sigma_0^2 = 0.50 \text{ mm}^2$ . A sample of  $n = 25$  parts gives  $s^2 = 0.30 \text{ mm}^2$ . Test at  $\alpha = 0.05$ . Estimate the p-value.

**Hypotheses:**  $H_0 : \sigma^2 = 0.50$  vs  $H_1 : \sigma^2 < 0.50$ . One-tailed lower  $\chi^2$ -test,  $df = 24$ .

**Test statistic:**  $\chi_0^2 = \frac{24 \times 0.30}{0.50} = 14.4$  (falls in lower tail)

**P-value bounds:** Row  $\nu = 24$ , use **large- $\alpha$**  columns:

$$\chi_{0.95,24}^2 = 13.85 < 14.4 < 15.66 = \chi_{0.90,24}^2.$$

$P(\chi^2 < 14.4)$ : between  $1 - 0.95 = 0.05$  and  $1 - 0.90 = 0.10$ .

$0.05 < p\text{-value} < 0.10$
--------------------------------

**Decision:** Since  $p > 0.05 = \alpha$ , we **fail to reject**  $H_0$ . Insufficient evidence.

## Key Trick: Lower-Tail $\chi^2$ Bounds

---

**For a lower-tail  $\chi^2$  test:**

If  $\chi_{\alpha_1, \nu}^2 < \chi_0^2 < \chi_{\alpha_2, \nu}^2$  where  $\alpha_1 > \alpha_2$  (both large values like 0.90, 0.95), then:

$$P(\chi^2 < \chi_0^2) \text{ is between } (1 - \alpha_1) \text{ and } (1 - \alpha_2)$$

$$\alpha_1 = 0.95, \alpha_2 = 0.90 \Rightarrow p\text{-value between } 0.05 \text{ and } 0.10$$

**Think of it this way:**

- The column header  $\alpha$  always means  $P(\chi^2 > \text{value}) = \alpha$
- So  $P(\chi^2 < \text{value}) = 1 - \alpha$
- You just subtract from 1

## Part 3: P-Value Bounds from the $F$ -Table

---

**How the  $F$ -table is organized (different from  $t$  and  $\chi^2$ !):**

- **Separate page for each  $\alpha$ :**  $\alpha = 0.25, 0.10, 0.05, 0.025, 0.01$
- On each page: rows =  $df_2$  (denominator), columns =  $df_1$  (numerator)
- Entry =  $F_{\alpha, df_1, df_2}$  such that  $P(F > F_{\alpha, df_1, df_2}) = \alpha$

**To find p-value bounds:**

- Check your test statistic across **multiple pages** (one per  $\alpha$ )
- For each  $\alpha$  page, look up  $F_{\alpha, df_1, df_2}$
- Find which two consecutive  $\alpha$  values bracket your  $f_0$

**Tip:** You are reading **across pages** instead of across columns.

## F-Table Example 1

---

An engineer suspects Line A is more variable than Line B. Samples:  $n_1 = 10$  from Line A with  $s_1^2 = 0.057$ , and  $n_2 = 15$  from Line B with  $s_2^2 = 0.020$ . Test at  $\alpha = 0.05$ . Estimate the p-value.

**Hypotheses:**  $H_0 : \sigma_1^2 = \sigma_2^2$  vs  $H_1 : \sigma_1^2 > \sigma_2^2$ . One-tailed  $F$ -test.  
 $df_1 = 9$ ,  $df_2 = 14$ .

**Test statistic:**  $F_0 = \frac{s_1^2}{s_2^2} = \frac{0.057}{0.020} = 2.85$

## F-Table Example 1

---

**P-value bounds:** Look up  $F_{\alpha,9,14}$  across pages:

$\alpha$	0.25	0.10	<b>0.05</b>	<b>0.025</b>	0.01
$F_{\alpha,9,14}$	1.47	2.12	<b>2.65</b>	<b>3.21</b>	4.03

Bracket:  $F_{0.05} = 2.65 < 2.85 < 3.21 = F_{0.025}$ .

$$0.025 < p\text{-value} < 0.05$$

**Decision:**  $p < 0.05$ , so **reject**  $H_0$ . Line A has significantly greater variance.

## F-Table Example 2

---

A process engineer tests whether two catalyst batches have **different** variances. Samples:  $n_1 = 8$  with  $s_1 = 4.8$ , and  $n_2 = 10$  with  $s_2 = 2.7$ . Test at  $\alpha = 0.05$ . Estimate the p-value.

**Hypotheses:**  $H_0 : \sigma_1^2 = \sigma_2^2$  vs  $H_1 : \sigma_1^2 \neq \sigma_2^2$ . Two-tailed  $F$ -test.  
 $df_1 = 7$ ,  $df_2 = 9$ .

**Test statistic:** 
$$F_0 = \frac{s_1^2}{s_2^2} = \frac{4.8^2}{2.7^2} = \frac{23.04}{7.29} = 3.16$$

## F-Table Example 2

---

**P-value bounds:** Look up  $F_{\alpha,7,9}$  across pages:

$\alpha$	0.25	<b>0.10</b>	<b>0.05</b>	0.025	0.01
$F_{\alpha,7,9}$	1.60	<b>2.51</b>	<b>3.29</b>	4.20	5.61

Upper tail:  $0.05 < P(F > 3.16) < 0.10$ . Two-tailed:  $0.10 < p\text{-value} < 0.20$

**Decision:**  $p > 0.05$ , so **fail to reject**  $H_0$ . No significant difference in variances.

## Quick Reference: P-Value Bounds Cheat Sheet

---

Test	One-Tailed Bound	Two-Tailed Bound	Table Tip
$t$ -test	Read $\alpha$ values directly from columns bracketing $ t_0 $	Multiply both bounds by 2	Read across columns in one row
$\chi^2$ -test (upper)	Read $\alpha$ from columns bracketing $\chi_0^2$	Multiply both bounds by 2	Use small- $\alpha$ columns
$\chi^2$ -test (lower)	Subtract column $\alpha$ from 1	Multiply by 2 after subtracting	Use large- $\alpha$ columns
$F$ -test	Read $\alpha$ values from pages bracketing $f_0$	Multiply both bounds by 2	Read across <b>pages</b>

## Common Mistakes to Avoid

---

- **Forgetting to multiply by 2 for two-tailed tests.**

The table gives you the *one-tail* area only. Two-tailed p-value =  $2 \times$  one-tail area.

- **Wrong direction for  $\chi^2$  lower-tail tests.**

Column header  $\alpha$  means  $P(\chi^2 > \text{value})$ . For lower tail:  $p = 1 - \alpha$ .

- **Reading the wrong  $F$ -table page.**

Each page corresponds to a different  $\alpha$ . Also:  $df_1$  is columns,  $df_2$  is rows.

- **Reversing the inequality direction.**

If  $t_{0.025} < |t_0| < t_{0.01}$ , then  $0.01 < \text{tail area} < 0.025$ .

The *larger* critical value corresponds to the *smaller*  $\alpha$ .

## Summary

---

- The  **$t$ -table** and  **$\chi^2$ -table** are read across columns for a given df row.
- The  **$F$ -table** is read across pages (each page = different  $\alpha$ ).
- For **one-tailed** tests, the bounds come directly from the table.
- For **two-tailed** tests, multiply the one-tail bounds by 2.
- For  **$\chi^2$  lower-tail** tests, subtract the column header from 1.
- Always state your bounds as:  $\alpha_{\text{small}} < p\text{-value} < \alpha_{\text{large}}$ .

*Practice with the textbook appendix tables until this becomes second nature. These bounds are helpful on every exam and quiz!*