
A Guide to Hypothesis Testing in Linear Regression Models

Tests of *One Coefficient Restriction: One Restriction on One Coefficient*

H_0 specifies *only one equality* restriction on *one coefficient*.

□ *Two-Tail Tests of One Restriction on One Coefficient*

• **Example:** $H_0: \beta_j = b_j$ versus $H_1: \beta_j \neq b_j$ where b_j is a specified constant.

• **Use:** either a *two-tail t-test* or an **F-test**.

• **Test Statistics:**

$$\Rightarrow t(\hat{\beta}_j) = \frac{\hat{\beta}_j - \beta_j}{\widehat{se}(\hat{\beta}_j)} \sim t[N - k] \quad \Rightarrow \text{sample value} = t_0(\hat{\beta}_j) = \frac{\hat{\beta}_j - b_j}{\widehat{se}(\hat{\beta}_j)}.$$

$$\Rightarrow F(\hat{\beta}_j) = \frac{(\hat{\beta}_j - \beta_j)^2}{\widehat{Var}(\hat{\beta}_j)} \sim F[1, N - k] \quad \Rightarrow \text{sample value} = F_0(\hat{\beta}_j) = \frac{(\hat{\beta}_j - b_j)^2}{\widehat{Var}(\hat{\beta}_j)}.$$

Note: $[t(\hat{\beta}_j)]^2 = F(\hat{\beta}_j)$ or $t(\hat{\beta}_j) = \sqrt{F(\hat{\beta}_j)}$ and $[t_{\alpha/2}[N - k]]^2 = F_\alpha[1, N - k]$.

• **Decision Rules:**

Reject H_0 if $|t_0| > t_{\alpha/2}[N - k]$ or **two-tail p-value** for $t_0 = \Pr(|t| > |t_0|) < \alpha$;
 $F_0 > F_\alpha[1, N - k]$ or **p-value** for $F_0 = \Pr(F > F_0) < \alpha$.

Retain H_0 if $|t_0| \leq t_{\alpha/2}[N - k]$ or **two-tail p-value** for $t_0 = \Pr(|t| > |t_0|) \geq \alpha$;
 $F_0 \leq F_\alpha[1, N - k]$ or **p-value** for $F_0 = \Pr(F > F_0) \geq \alpha$.

□ **One-Tail Tests of One Restriction on One Coefficient**

- **Examples:** $H_0: \beta_j = b_j$ (or $\beta_j \geq b_j$) versus $H_1: \beta_j < b_j$ a **left-tail test**
 $H_0: \beta_j = b_j$ (or $\beta_j \leq b_j$) versus $H_1: \beta_j > b_j$ a **right-tail test**

- **Use:** a **one-tail t-test**.

- **Test Statistic:**

$$\Rightarrow t(\hat{\beta}_j) = \frac{\hat{\beta}_j - \beta_j}{\widehat{\text{se}}(\hat{\beta}_j)} \sim t[N - k] \Rightarrow \text{sample value} = t_0(\hat{\beta}_j) = \frac{\hat{\beta}_j - b_j}{\widehat{\text{se}}(\hat{\beta}_j)}.$$

- **Decision Rules -- left-tail t-test:**

Reject H_0 if $t_0 < -t_{\alpha}[N - k]$ or **one-tail p-value** for $t_0 = \Pr(t < t_0) < \alpha$;

Retain H_0 if $t_0 \geq -t_{\alpha}[N - k]$ or **one-tail p-value** for $t_0 = \Pr(t < t_0) \geq \alpha$.

- **Decision Rules -- right-tail t-test:**

Reject H_0 if $t_0 > t_{\alpha}[N - k]$ or **one-tail p-value** for $t_0 = \Pr(t > t_0) < \alpha$;

Retain H_0 if $t_0 \leq t_{\alpha}[N - k]$ or **one-tail p-value** for $t_0 = \Pr(t > t_0) \geq \alpha$.

Tests of *One Linear Restriction on Two or More Coefficients*

H_0 specifies *only one* linear restriction on *two or more* regression coefficients.

□ Two-Tail Tests of *One Linear Restriction on Two Coefficients*

• **Example:** $H_0: c_j\beta_j + c_h\beta_h = c_0$ versus $H_1: c_j\beta_j + c_h\beta_h \neq c_0$.

• **Use:** either a *two-tail t-test* or an **F-test**.

• **Test Statistics:**

$$\Rightarrow t(c_j\hat{\beta}_j + c_h\hat{\beta}_h) = \frac{(c_j\hat{\beta}_j + c_h\hat{\beta}_h) - (c_j\beta_j + c_h\beta_h)}{\hat{s}e(c_j\hat{\beta}_j + c_h\hat{\beta}_h)} \sim t[N - k] = t[N - k_1]$$

$$\text{where: } \hat{s}e(c_j\hat{\beta}_j + c_h\hat{\beta}_h) = \sqrt{\hat{V}ar(c_j\hat{\beta}_j + c_h\hat{\beta}_h)}$$

$$\hat{V}ar(c_j\hat{\beta}_j + c_h\hat{\beta}_h) = c_j^2\hat{V}ar(\hat{\beta}_j) + c_h^2\hat{V}ar(\hat{\beta}_h) + 2c_jc_h\hat{C}ov(\hat{\beta}_j, \hat{\beta}_h).$$

$$\text{sample value} = t_0(c_j\hat{\beta}_j + c_h\hat{\beta}_h) = \frac{(c_j\hat{\beta}_j + c_h\hat{\beta}_h) - c_0}{\hat{s}e(c_j\hat{\beta}_j + c_h\hat{\beta}_h)}.$$

$$\Rightarrow F(c_j\hat{\beta}_j + c_h\hat{\beta}_h) = \frac{[(c_j\hat{\beta}_j + c_h\hat{\beta}_h) - (c_j\beta_j + c_h\beta_h)]^2}{\hat{V}ar(c_j\hat{\beta}_j + c_h\hat{\beta}_h)} \sim F[1, N - k] = F[1, N - k_1]$$

$$\text{where: } \hat{V}ar(c_j\hat{\beta}_j + c_h\hat{\beta}_h) = c_j^2\hat{V}ar(\hat{\beta}_j) + c_h^2\hat{V}ar(\hat{\beta}_h) + 2c_jc_h\hat{C}ov(\hat{\beta}_j, \hat{\beta}_h).$$

$$\text{sample value} = F_0(c_j\hat{\beta}_j + c_h\hat{\beta}_h) = \frac{[(c_j\hat{\beta}_j + c_h\hat{\beta}_h) - c_0]^2}{\hat{V}ar(c_j\hat{\beta}_j + c_h\hat{\beta}_h)}$$

• **Decision Rules:**

Reject H_0 if $|t_0| > t_{\alpha/2}[N - k]$ or **two-tail p-value** for $t_0 = \Pr(|t| > |t_0|) < \alpha$;
 $F_0 > F_\alpha[1, N - k]$ or **p-value** for $F_0 = \Pr(F > F_0) < \alpha$.

Retain H_0 if $|t_0| \leq t_{\alpha/2}[N - k]$ or **two-tail p-value** for $t_0 = \Pr(|t| > |t_0|) \geq \alpha$;
 $F_0 \leq F_\alpha[1, N - k]$ or **p-value** for $F_0 = \Pr(F > F_0) \geq \alpha$.

□ **One-Tail Tests of One Linear Restriction on Two Coefficients**

- **Examples:** $H_0: c_j\beta_j + c_h\beta_h = c_0$ vs. $H_1: c_j\beta_j + c_h\beta_h < c_0$ a **left-tail test**
 $H_0: c_j\beta_j + c_h\beta_h = c_0$ vs. $H_1: c_j\beta_j + c_h\beta_h > c_0$ a **right-tail test**

- **Use:** a **one-tail t-test**.

- **Test Statistic:**

$$\Rightarrow t(c_j\hat{\beta}_j + c_h\hat{\beta}_h) = \frac{(c_j\hat{\beta}_j + c_h\hat{\beta}_h) - (c_j\beta_j + c_h\beta_h)}{\hat{s}e(c_j\hat{\beta}_j + c_h\hat{\beta}_h)} \sim t[N-k] = t[N-k_1]$$

$$\text{where } \hat{s}e(c_j\hat{\beta}_j + c_h\hat{\beta}_h) = \sqrt{\text{V}\hat{a}r(c_j\hat{\beta}_j + c_h\hat{\beta}_h)}$$

$$\text{V}\hat{a}r(c_j\hat{\beta}_j + c_h\hat{\beta}_h) = c_j^2 \text{V}\hat{a}r(\hat{\beta}_j) + c_h^2 \text{V}\hat{a}r(\hat{\beta}_h) + 2c_jc_h \text{C}\hat{o}v(\hat{\beta}_j, \hat{\beta}_h).$$

$$\text{sample value} = t_0(c_j\hat{\beta}_j + c_h\hat{\beta}_h) = \frac{(c_j\hat{\beta}_j + c_h\hat{\beta}_h) - c_0}{\hat{s}e(c_j\hat{\beta}_j + c_h\hat{\beta}_h)}.$$

- **Decision Rules -- left-tail t-test:**

Reject H_0 if $t_0 < -t_\alpha[N-k]$ or **one-tail p-value** for $t_0 = \Pr(t < t_0) < \alpha$;

Retain H_0 if $t_0 \geq -t_\alpha[N-k]$ or **one-tail p-value** for $t_0 = \Pr(t < t_0) \geq \alpha$.

- **Decision Rules -- right-tail t-test:**

Reject H_0 if $t_0 > t_\alpha[N-k]$ or **one-tail p-value** for $t_0 = \Pr(t > t_0) < \alpha$;

Retain H_0 if $t_0 \leq t_\alpha[N-k]$ or **one-tail p-value** for $t_0 = \Pr(t > t_0) \geq \alpha$.

Tests of *Two or More* Linear Coefficient Restrictions

H_0 specifies *two or more* linear coefficient restrictions.

- **Example:** $H_0: \beta_2 = \beta_4$ and $\beta_3 = \beta_5$
 $H_1: \beta_2 \neq \beta_4$ and/or $\beta_3 \neq \beta_5$
- **Use:** a general F-test; only an F-test can be used to test jointly *two or more* coefficient restrictions.
- **Test Statistics:** Either of the following two general F-statistics.

$$\Rightarrow F = \frac{(RSS_0 - RSS_1)/(df_0 - df_1)}{RSS_1/df_1} = \frac{(RSS_0 - RSS_1)/(k - k_0)}{RSS_1/(N - k)}$$

$$\Rightarrow F = \frac{(R_U^2 - R_R^2)/(df_0 - df_1)}{(1 - R_U^2)/df_1} = \frac{(R_U^2 - R_R^2)/(k - k_0)}{(1 - R_U^2)/(N - k)}$$

Null distribution: $F \sim F[df_0 - df_1, df_1] = F[k - k_0, N - k]$.

- **Sample value** of F-statistic under $H_0 = F_0$.

- **Decision Rules:**

Reject H_0 if $F_0 > F_\alpha[df_0 - df_1, df_1] = F_\alpha[k - k_0, N - k]$ or
p-value for $F_0 = \Pr(F > F_0) < \alpha$.

Retain H_0 if $F_0 \leq F_\alpha[df_0 - df_1, df_1] = F_\alpha[k - k_0, N - k]$ or
p-value for $F_0 = \Pr(F > F_0) \geq \alpha$.