

# ANOVA F-test for SLR model usefulness

## Use case

We have a single simple linear regression model (SLR) and want to assess its usefulness. Does the model provide useful predictions about our variable of interest?

The population regression line is defined by

$\mu_y = \beta_0 + \beta_1 x$ , and the least-squares regression line is

$\hat{y} = b_0 + b_1 x$ .

The ANOVA approach uses the following quantities to assess how much of the data's variation is explained by our model.

$$\text{Total SS (SST)} = \sum (y_i - \bar{y})^2 \quad \text{df} = n - 1$$

$$\text{Regression SS (SSR)} = \sum (\hat{y}_i - \bar{y})^2 \quad \text{df} = 1$$

$$\text{Residual SS (SSE)} = \sum (y_i - \hat{y}_i)^2 \quad \text{df} = n - 2$$

$$\text{Note: SST} = \text{SSR} + \text{SSE}$$

Using these values, we can readily compute the *coefficient of determination*,

$r^2$ .

$$r^2 = \frac{\text{SSR}}{\text{SST}}$$

Finally, we compute the *mean squares* values before we can calculate the test statistic.

$$\text{Total MS (MST)} = \frac{\text{SST}}{\text{df}_{\text{SST}}} = \frac{\text{SST}}{n - 1}$$

$$\text{Regression MS (MSR)} = \frac{\text{SSR}}{\text{df}_{\text{SSR}}} = \frac{\text{SSR}}{1}$$

$$\text{Residual MS (MSE)} = \frac{\text{SSE}}{\text{df}_{\text{SSE}}} = \frac{\text{SSE}}{n - 2}$$

Note: In general,  $\text{SST} \neq \text{SSR} + \text{SSE}$  (mean squares are not additive)

## Preconditions

For an ANOVA F-test to be valid, we should verify that we already meet the preconditions for the validity of SLR. (If SLR is inappropriate, an F test on your model would be pointless.)

In our case, this means that we should verify that residuals are scattered randomly around zero with uniform variation. This satisfies the *linearity*, *independence*, and *constant variance* assumptions that underpin SLR.

## Hypotheses

**Null hypothesis,**

$H_0$ : the model is not useful (

$\beta_1 = 0$ )

**Alternative hypothesis,**

$H_a$ : the model is useful (

$\beta_1 \neq 0$ )

## Test statistic

$$F = \frac{\text{MSR}}{\text{MSE}}$$

## p-value

We compute our p value as follows, where

$F(k_1, k_2)$  indicates a random variable distributed according to the

$F(k_1, k_2)$  distribution.

$$p = P(F(k_1, k_2) \geq F), \text{ where } k_1 = 1 \text{ and } k_2 = n - 2$$

## p-value (Python)

```
p = 1 - stats.f.cdf(F, 1, n-2)
```