

Inference in regression: confidence intervals

ST552 Lecture 9

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Today

- ▶ Finish last time's slides and exercises from last time
- ▶ Another view of F-tests
- ▶ Confidence intervals for single parameters
- ▶ Confidence intervals for linear combinations of parameters
- ▶ Confidence intervals for parameters jointly

Last time...

Certain hypotheses of interest can be set up as competing models. A full model and a simpler model (nested in the full model). A.K.A testing models.

Identify the models of interest. Fit both. Check fit of full model. Find F-statistic, and answer questions of interest.

Another way to set up F-tests A.K.A testing linear parametric functions

Assuming the regression model:

$$Y = X\beta + \epsilon, \quad \epsilon \sim N(0, \sigma^2 I)$$

Consider the hypotheses:

$$H_0 : K^T \beta = m$$

$$H_1 : K^T \beta \neq m$$

where $K_{k \times p}$ matrix with $\text{rank}(K) = k$.

Then under the null hypothesis,

$$F = \frac{\left((K^T \beta - m)^T \left(K^T (X^T X)^{-1} K \right)^{-1} (K^T \beta - m) \right) / k}{\text{RSS} / (n - p)} \sim F_{k, n-p}$$

This alternative is equivalent to the model testing set up we considered. Every null hypothesis of the form $K^T\beta = m$ is comparing a full and reduced model and vice versa.

For example, consider

$$K = \begin{pmatrix} 0 \\ \vdots \\ 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix}, \quad m = \mathbf{0}$$

Your turn

What are K and m for exercises 1 and 5 from the handout from last time?

Confidence intervals for individual β_j and linear combinations

The t-test for an individual parameter can be flipped around to give $100(1 - \alpha)\%$ confidence intervals of the form

$$\hat{\beta}_j \pm t_{n-p}^{(\alpha/2)} \text{SE}(\hat{\beta}_j)$$

(Remember $\text{SE}(\hat{\beta}_j)$ is coming from the diagonal entry of the estimated variance-covariance matrix.)

Similarly, confidence intervals for a linear combination of the parameters, $c^T \beta$ where $c_{p \times 1}$, can be formed

$$c^T \hat{\beta} \pm t_{n-p}^{(\alpha/2)} \sqrt{\hat{\sigma}^2 c^T (X^T X)^{-1} c}$$

Joint confidence regions

A joint $100(1 - \alpha)\%$ confidence for the vector β can be formed using,

$$(\hat{\beta} - \beta)^T X^T X (\hat{\beta} - \beta) \leq p \hat{\sigma}^2 F_{p, n-p}^{(\alpha)}$$

and results in p -dimensional ellipsoids (very hard to visualise, but essential for communicating joint uncertainty when the parameters are correlated).

The 2D ellipsoid example