

Testing General Linear Constraints on the Regression Coefficient Vector: A Note

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Consider a linear model with n observations and k explanatory variables:

$$(1) \mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \mathbf{u}, \quad \mathbf{u} \sim N(\mathbf{0}, \sigma^2 \mathbf{I}_n).$$

We assume that the model satisfies the ideal conditions. Consider the general linear constraints on regression coefficient vector:

$$(2) \mathbf{R}\boldsymbol{\beta} = \mathbf{r},$$

where \mathbf{R} and \mathbf{r} are known matrices of orders $q \times k$ and $q \times 1$ respectively, and the rank of \mathbf{R} is $q < k$. We also assume $n > k + q$. Without loss of generality \mathbf{R} can be partitioned as $\mathbf{R} = (\mathbf{R}_1 \ \mathbf{R}_2)$, where \mathbf{R}_2 is a $q \times q$ nonsingular matrix so that

$$\mathbf{R}\boldsymbol{\beta} = (\mathbf{R}_1 \ \mathbf{R}_2) \begin{bmatrix} \boldsymbol{\beta}_1 \\ \boldsymbol{\beta}_2 \end{bmatrix} = \mathbf{R}_1\boldsymbol{\beta}_1 + \mathbf{R}_2\boldsymbol{\beta}_2 = \mathbf{r},$$

or,

$$(3) \boldsymbol{\beta}_2 = -\mathbf{R}_2^{-1}\mathbf{R}_1\boldsymbol{\beta}_1 + \mathbf{R}_2^{-1}\mathbf{r}.$$

Therefore, under the constraints (2) or (3), (1) can be written as

$$\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \mathbf{u} = (\mathbf{X}_1 \ \mathbf{X}_2) \begin{bmatrix} \boldsymbol{\beta}_1 \\ \boldsymbol{\beta}_2 \end{bmatrix} + \mathbf{u} = \mathbf{X}_2\mathbf{R}_2^{-1}\mathbf{r} + \mathbf{X}\mathbf{R}^*\boldsymbol{\beta}_1 + \mathbf{u},$$

or,

$$(4) \mathbf{y}^* = \mathbf{X}^*\boldsymbol{\beta}_1 + \mathbf{u},$$

where

$$\mathbf{R}^* = \begin{bmatrix} \mathbf{I}_{k-q} \\ -\mathbf{R}_2^{-1}\mathbf{R}_1 \end{bmatrix}$$

$$\mathbf{y}^* = \mathbf{y} - \mathbf{X}_2\mathbf{R}_2^{-1}\mathbf{r}$$

and $\mathbf{X}^* = \mathbf{X}\mathbf{R}^*$.

Now, we have the following lemma:

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Lemma: Let R^* be a $k \times k_1$ matrix with full column rank $k_1 < k$, and let $X^* = XR^*$, where X is an $n \times k$ matrix with full column rank. Define $M = I - X(X'X)^{-1}X'$ and $M^* = I - X^*(X'^*X^*)^{-1}X'^*$. Then M and $M^* - M$ are symmetric idempotent matrices with ranks $n - k$ and $k - k_1$, respectively, and $M \cdot (M^* - M) = 0$. (See Fisher [1].)

Proof: Notice

$$\begin{aligned} X^*(X'^*X^*)^{-1}X'^*X(X'X)^{-1}X' &= X^*(X'^*X^*)^{-1}R'^*X'X(X'X)^{-1}X' \\ &= X^*(X'^*X^*)^{-1}R'^*X' = X^*(X'^*X^*)^{-1}X'^*, \end{aligned}$$

then the proof is straightforward.

We can easily verify that X , R^* and X^* in (4) satisfy the conditions of the lemma with $k_1 = k - q$, or,

$q = k - k_1$. Therefore,

$$(5) \quad y'^*My^* = y'My = u'Mu$$

is distributed as $\sigma^2\chi^2(n - k)$, and

$$(6) \quad y'^*(M^* - M)y^* = u'(M^* - M)u$$

is independently distributed as $\sigma^2\chi^2(q)$ under the constraints (2). (The equalities in (5) and (6) follow from the fact that $MX = 0$, and $MX_2 = 0$.)

Therefore

$$(7) \quad F = \frac{y'^*(M^* - M)y^*/q}{y'^*My^*/(n - k)}$$

is distributed as $F(q, n - k)$ under the constraints (2). If we denote the LS residual vectors, My of (1) and M^*y^* of (4), by e and e^* , respectively, then (7) becomes

$$(7)' \quad F = \frac{(e'^*e^* - e'e)/q}{e'e/(n - k)}$$

Therefore, using (7) or (7)', we can test any hypothesis given by, or reducible to, the constraints (2).

REFERENCES

- [1] Fisher, F.M., "Tests of Equality between Sets of Coefficients in Two Linear Regressions: An Expository Note," *Econometrica*, 38 (1970), 361-366.
 [2] Theil, H., *Principles of Econometrics*. New York; Wiley, 1971.