

Computing Cluster-Unbiased Standard Errors

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This document is written in support of the MATLAB function `KSS_SE` which permits to compute the leave-out clustered SEs introduced by [Kline et al. \(2020\)](#)—KSS henceforth—in a linear regression model.

1 Introduction

Consider a regression equation of the form

$$y_{ij} = x'_{ij}\beta + \varepsilon_{ij} \quad j = 1, \dots, J; \quad i = 1, \dots, n_j;$$

where i indexes a particular observation which belongs to a cluster j and we have $N = \sum_j M_j$ total observations; x_{ij} is a vector of regressors of dimension $K \times 1$ and y_{ij} is the outcome of interest. The error terms, ε_{ij} , are assumed to be heteroskedastic and potentially correlated across observations belonging to the same cluster j with a block-diagonal variance-covariance matrix given by

$$\Omega = \begin{bmatrix} \Omega_1 & 0 & 0 & 0 \\ 0 & \Omega_2 & 0 & 0 \\ 0 & 0 & \ddots & 0 \\ 0 & 0 & 0 & \Omega_J \end{bmatrix}$$

Cluster robust standard errors for the OLS estimator of β , $\hat{\beta}$, in most software packages (e.g. `reghdfe`) is based on the following well-known formula

$$\tilde{V}[\hat{\beta}] = d \left(\sum_{i,j} x_{ij} x'_{ij} \right)^{-1} \left[\sum_{j=1}^J \left(\sum_i^{n_j} x_{ij} \hat{\varepsilon}_{ij} \right) \left(\sum_i^{n_j} x_{ij} \hat{\varepsilon}_{ij} \right)' \right] \left(\sum_{i,j} x_{ij} x'_{ij} \right)^{-1},$$

where d is some degrees of freedom adjustment and $\hat{\varepsilon}_{ij} = y_{ij} - x_{ij}\hat{\beta}$ is the OLS residual.¹

KSS introduces an unbiased estimate of the variance of Ω_j given by

$$\hat{\Omega}_j = \mathbf{y}_j(\mathbf{y}_j - \mathbf{x}_j\hat{\beta}_{-j})',$$

where $\hat{\beta}_{-j}$ is the OLS estimate of β obtained after fitting (1) leaving cluster j out; \mathbf{y}_j is a $n_j \times 1$ vector that stacks the outcome variable y_{ij} for observations belonging in cluster j ; similarly, \mathbf{x}_j is a $n_j \times K$ matrix that stacks the regressors x_{ij} for the observations belonging to cluster j .

¹For instance, `reghdfe` sets $d = \frac{J}{J-1} \frac{N}{N-K}$.

Let $\hat{\eta}_{ij}$, represent the leave-cluster out residual, i.e. $\hat{\eta}_{ij} \equiv y_{ij} - x_{ij}\hat{\beta}_{-j}$. KSS shows that the following is unbiased estimate of the sampling variability of $\hat{\beta}$

$$\hat{\mathbb{V}}[\hat{\beta}] = \left(\sum_{i,j} x_{ij} x'_{ij} \right)^{-1} \left[\sum_{j=1}^J \left(\sum_i^{n_j} x_{ij} y_{ij} \right) \left(\sum_i^{n_j} x_{ij} \hat{\eta}_{ij} \right)' \right] \left(\sum_{i,j} x_{ij} x'_{ij} \right)^{-1},$$

The software described in this vignette, `KSS_SE`, computes $\hat{\mathbb{V}}[\hat{\beta}]$. The readme in package illustrates how this function works in an example where one is interested in estimating an event-study design using a two-way fixed effects regression.

References

Kline, P., R. Saggio, and M. S¸olvsten (2020). Leave-out estimation of variance components. *Econometrica* 88(5), 1859–1898.