

Some Conclusions and Implications: The Seemingly Unrelated Regression (SUR) Model

$$1) \underline{y}_i = X_i \underline{\beta}_i + \underline{\mu}_i, i=1, 2, E\underline{\mu}_i = 0, E\underline{\mu}_i \underline{\mu}_i' = \sigma_{ii}, E\underline{\mu}_i \underline{\mu}_m' = \sigma_{im}, i, m=1, 2, i=1, \dots, n.$$

A. OLS Equation-by-Equation is feasible when there are no cross equation restrictions and can be a useful starting point even with cross-equation restrictions.

$$2) \hat{\underline{\beta}}_i = (X_i' X_i)^{-1} X_i' y_i, \text{Var}(\hat{\underline{\beta}}_i) = \sigma_{ii} (X_i' X_i)^{-1}, \hat{\sigma}_{im} = (y_i - X_i \hat{\underline{\beta}}_i)' (y_m - X_m \hat{\underline{\beta}}_m) / n \quad (\text{or } n-k)$$

B. If no cross-equation restrictions and no tests of coefficients across equations, the potential exists to fit SUR model as OLS equation by equation or by the stacked 2-equation model.

i) If  $X_1 = X_2$ , there is no gain over OLS EQBE in going to the stacked model

ii) If  $X_1 \neq X_2$  and  $\mu_1$  and  $\mu_2$  are heteroscedastic  $\sigma_{11} \neq \sigma_{22}$  and/or correlated  $\sigma_{12} \neq 0$ , then there is potential for efficiency gain by fitting the two equation model as one stacked model by GLS or FGLS.

iii) If  $\mu_1$  and  $\mu_2$  have the same variance and are uncorrelated, then perform OLS EQBE.

NOTE: SSE for the stacked model unrestricted:  $SSE(\Omega_Y) = SSE(\Omega_1) + SSE(\Omega_2)$ .

C. If have cross-equation restrictions, then the SUR model must be fitted by the stacked model.

Also, this must be used if one is testing or imposing cross-equation restrictions.  $\underline{Y} = \underline{X}\underline{B} + \underline{U}$  SL.  $\underline{CB} = \underline{U}$

i) If  $\mu_1$  and  $\mu_2$  are homoscedastic and uncorrelated across equations, estimate the stacked model by OLS subject to a set of  $q$  linear restrictions.

Remark: This could happen if you have two different data sets on the same econometric model. By fitting the stacked model, you can perform a test that  $\underline{\beta}_1 = \underline{\beta}_2$ , i.e., the regressions coefficients are the same in the two data sets.

ii) If  $\mu_1$  and  $\mu_2$  are heteroscedastic and/or correlated across equations, then estimate the stacked model by GLS/FGLS.

Remark: With cross-equation restrictions, the same transformation must be performed on both equations in the stack; FGLS subject to  $q$  linear restrictions.

Remark: Testing here uses the  $t$  and  $F$  statistics for the stacked model.

$$\underline{Y} = \begin{bmatrix} y_1 \\ y_2 \end{bmatrix}, \underline{X} = \begin{bmatrix} X_1 & 0 \\ 0 & X_2 \end{bmatrix}, \underline{U} = \begin{bmatrix} \mu_1 \\ \mu_2 \end{bmatrix}, \underline{B} = \begin{bmatrix} \beta_1 \\ \beta_2 \end{bmatrix}, n_1 = n_2 = n$$

$2n \times 1$        $2n \times 2k$        $2n \times 1$        $2k \times 1$

$$F^* = \frac{[SSE(\hat{\Omega}) - SSE(\hat{\Omega}_Y)]/q}{SSE(\hat{\Omega})/(2n-2k)} \sim F_{q, 2n-2k}$$