Code documentation

César Garro-Marín*

May 30, 2023

Estimation 1

We estimate the three-equation model:

$$\Delta \overline{\ln S_{ijet}^e} = \sum_k \theta_{ke} \overline{S_{kejt}} \pi_{kjt} - \pi_{ijt}$$
 (1)

$$1 = \sum_{k} \theta_k^e \overline{S_{kejt}} \tag{2}$$

$$\Delta \left[\ln \frac{q_{eJt}}{q_{e'Jt}} \right] = \beta_j \left[\sum_k \left(\theta_k^e \overline{S_{kJet}} - \theta_k^{e'} \overline{S_{kJe't}} \right) \pi_{kjt} \right] + \gamma_{e,e't}$$
 (3)

where $\pi_{kjt} = \frac{\sigma_j}{\sigma_j - 1} \Delta \ln A_{kjt}$ and $\beta_j = \frac{\sigma_j - 1}{\sigma_j}$. Let ε_l be vector of stacked errors from equation j and let ε the vector of stacked errors of the full model ($\varepsilon = (\varepsilon_1, \varepsilon_2, \varepsilon_3)'$). We estimate the model parameters μ by GMM, minimizing the quadratic form:

$$Q = \varepsilon(\mu)' Z' Z \varepsilon(\mu) \tag{4}$$

where Z is the vector of instruments.

Instruments 1.1

The matrix of instruments Z is a block-diagonal matrix with diagonal elements given by the matrix of instruments for each equation (Z_l) .

• Equation 1 instruments: we instrument the skill levels $\overline{S_{kejt}}$ with the size-weighted skill level of the other education groups in that job:

$$\overline{S_{k-ejt}} = w_{e'jt}\overline{S_{ke'jt}} + w_{e''jt}\overline{S_{ke''jt}}$$

where $w_{e'it}$ is the employment share of education e' in that job and year. We also include as instruments the appropriate set of –negative– job by skill by year dummies. Therefore, Z_2 is a $12JT \times (12 + 4JT)$ matrix.

^{*}Boston University, email: cesarlgm@bu.edu

- Equation 2: instruments for this equation are the skill levels by education level. Therefore, Z_2 is a $3JT \times 12$ matrix.
- Equation 3 instruments: there are three types of instruments for this equation: education pair dummies, job by year dummies, and job by education dummies. Thus, Z_3 is a $2JT \times (6 + JT + JE)$ matrix.

1.2 Identification

The GMM model is overidentified. In total we have 30+5JT+JE instruments to identify 18+4JT+J parameters.

1.3 Standard errors

The GMM estimates are distributed as follows:

$$\sqrt{N}(\hat{\mu} - \mu) \to N(0, V)$$

where $(D'D)^{-1}D'VD(D'D)^{-1}$, with D the model gradient, and

$$\hat{V} = \frac{1}{N} \sum_{n=1}^{N} \varepsilon_n(\hat{\mu}) \varepsilon_n(\hat{\mu})'$$

1.4 Calculating the gradient

Reminder:

$$\frac{\partial Q}{\partial x} = x'(A + A')$$

I can write the quadratic from as:

$$Q(\gamma) = \frac{1}{N} \left(u(\gamma)' Z(Z'Z)^{-1} Z' u(\gamma) \right)$$

then,

$$\frac{\partial Q(\gamma)}{\partial \gamma} \ = \ \frac{2}{N} \left(\frac{\partial u(\gamma)}{\partial \gamma} \right)' Z(Z'Z)^{-1} Zu(\gamma)$$

now, I start element by element:

$$\begin{array}{lcl} \frac{\partial \Delta \ln S_{iejt}}{\partial \ln A_{kjt}} & = & \theta_{ke} S_{kejt} \\ \frac{\partial \Delta \ln S_{iejt}}{\partial \ln A_{ijt}} & = & \theta_{ie} S_{iejt} - 1 \\ \frac{\partial \Delta \ln S_{iejt}}{\partial \ln \theta_{ke}} & = & S_{kjt} \Delta \ln A_{kjt} \end{array}$$

for the sum to 1 restrictions we have:

$$\frac{\partial g_{ejt}(\gamma)}{\partial \theta_{ke}} = -S_{kejt}$$

next thing to do: write the gradient