

Estimación de un modelo de equilibrio general dinámico y estocástico

- **El modelo DSGE:**

$$\underset{\{c_t, h_t, k_{t+1}\}}{\text{Max}} E_0 \sum_{t=0}^{\infty} \beta^t [\ln c_t - \gamma h_t] a_t$$

$$\text{sujeto a: } \theta_t k_t^\alpha h_t^{1-\alpha} = c_t + k_{t+1} - (1 - \delta)k_t$$

$$k_0 \text{ dado}$$

donde a_t y θ_t son dos shocks estructurales, el primero es un shock en preferencias (que llamaremos shock de demanda) y el otro es un shock en la productividad total de los factores (que llamaremos shock de oferta).

Condiciones del equilibrio

$$y_t = \theta_t k_t^\alpha h_t^{1-\alpha} \quad (1)$$

$$\ln \theta_t = (1-\rho) \ln \bar{\theta} + \rho \ln \theta_{t-1} + \varepsilon_t, \quad (2)$$

$$y_t = c_t + i_t, \quad (3)$$

$$k_{t+1} = (1-\delta)k_t + i_t, \quad (4)$$

$$\gamma c_t h_t = (1-\alpha) y_t, \quad (5)$$

$$\frac{a_t}{c_t} = \beta E_t \left[\frac{a_{t+1}}{c_{t+1}} \left(\alpha \frac{y_{t+1}}{k_{t+1}} + 1 - \delta \right) \right], \quad (6)$$

$$\ln a_t = \rho_a \ln a_{t-1} + \zeta_t, \quad (7)$$

Estado estacionario

$$c_{ss} = \frac{1-\alpha}{\gamma} \bar{\theta} \left[\frac{\bar{\theta} \alpha}{(1/\beta) - 1 + \delta} \right]^{\frac{\alpha}{1-\alpha}},$$

$$k_{ss} = \frac{\alpha c_{ss}}{(1/\beta) - 1 + \delta - \alpha \delta},$$

$$h_{ss} = k_{ss} \left[\frac{(1/\beta) - 1 + \delta}{\bar{\theta} \alpha} \right]^{\frac{1}{1-\alpha}},$$

$$y_{ss} = \bar{\theta} k_{ss}^{\alpha} h_{ss}^{1-\alpha}$$

Log-linearización

Sea $\hat{x}_t = \ln(x_t / x_{ss})$, $x = y, c, k, i, h, a, \theta$

$$\hat{y}_t = \hat{\theta}_t + \alpha \hat{k}_t + (1 - \alpha) \hat{h}_t, \quad (1)$$

$$\hat{\theta}_t = \rho \hat{\theta}_{t-1} + \varepsilon_t, \quad (2)$$

$$\left[(1/\beta) - 1 + \delta \right] \hat{y}_t = \left[(1/\beta) - 1 + \delta - \alpha \delta \right] \hat{c}_t + \alpha \delta \hat{i}_t, \quad (3)$$

$$\hat{k}_{t+1} = (1 - \delta) \hat{k}_t + \delta \hat{i}_t, \quad (4)$$

$$\hat{c}_t + \hat{h}_t = \hat{y}_t, \quad (5)$$

$$(1/\beta) \hat{a}_t + (1/\beta) \hat{c}_t - (1/\beta) E_t \hat{c}_{t+1} + \left[(1/\beta) - 1 + \delta \right] \left(E_t \hat{y}_{t+1} - \hat{k}_{t+1} \right) - (1/\beta) E_t \hat{a}_{t+1}, \quad (6)$$

$$\hat{a}_t = \rho_a \hat{a}_{t-1} + \zeta_t \quad (7)$$

- **Transformando el modelo en una representación en espacio de los estados**

- *Ecuaciones contemporáneas*

$$\underbrace{\begin{bmatrix} 1 & 0 & \alpha - 1 \\ \frac{1}{\beta} - 1 + \delta & -\alpha\delta & 0 \\ 1 & 0 & -1 \end{bmatrix}}_A \underbrace{\begin{bmatrix} \hat{y}_t \\ \hat{i}_t \\ \hat{h}_t \end{bmatrix}}_{f_t^0} = \underbrace{\begin{bmatrix} \alpha & 0 \\ 0 & \frac{1}{\beta} - 1 + \delta - \alpha\delta \\ 0 & 1 \end{bmatrix}}_B \underbrace{\begin{bmatrix} \hat{k}_t \\ \hat{c}_t \end{bmatrix}}_{s_t^0} + \underbrace{\begin{bmatrix} 1 & 0 \\ 0 & 0 \\ 0 & 0 \end{bmatrix}}_C \underbrace{\begin{bmatrix} \hat{\theta}_t \\ \hat{a}_t \end{bmatrix}}_{\vartheta_t} \Rightarrow$$

$$A f_t^0 = B s_t^0 + C \vartheta_t \Rightarrow f_t^0 = A^{-1} B s_t^0 + A^{-1} C \vartheta_t, \quad (8)$$

$$\text{Sea } P = \begin{bmatrix} \rho & 0 \\ 0 & \rho_a \end{bmatrix}; \text{ entonces } E_t \mathfrak{Y}_{t+1} = P \mathfrak{Y}_t$$

- *Ecuaciones dinámicas (excepto los shocks estructurales)*

$$\underbrace{\begin{bmatrix} 1 & 0 \\ \frac{1}{\beta} - 1 + \delta & \frac{1}{\beta} \end{bmatrix}}_D \underbrace{\begin{bmatrix} \hat{k}_{t+1} \\ E_t \hat{c}_{t+1} \end{bmatrix}}_{E_t s_{t+1}^0} + \underbrace{\begin{bmatrix} 0 & 0 & 0 \\ \frac{1}{\beta} - 1 + \delta & 0 & 0 \end{bmatrix}}_F \underbrace{\begin{bmatrix} E_t \hat{y}_{t+1} \\ E_t \hat{i}_{t+1} \\ E_t \hat{h}_{t+1} \end{bmatrix}}_{E_t f_{t+1}^0} = \underbrace{\begin{bmatrix} 1 - \delta & 0 \\ 0 & \frac{1}{\beta} \end{bmatrix}}_G \underbrace{\begin{bmatrix} \hat{k}_t \\ \hat{c}_t \end{bmatrix}}_{s_t^0} + \\
 \underbrace{\begin{bmatrix} 0 & \delta & 0 \\ 0 & 0 & 0 \end{bmatrix}}_H \underbrace{\begin{bmatrix} \hat{y}_t \\ \hat{i}_t \\ \hat{h}_t \end{bmatrix}}_{f_t^0} + \underbrace{\begin{bmatrix} 0 & 0 \\ 0 & \frac{1}{\beta}(1 - \rho_a) \end{bmatrix}}_J \underbrace{\begin{bmatrix} \hat{\theta}_t \\ \hat{a}_t \end{bmatrix}}_{\mathfrak{I}_t} \Rightarrow$$

$$DE_t s_{t+1}^0 + FE_t f_{t+1}^0 = Gs_t^0 + Hf_t^0 + J\mathfrak{I}_t \quad , \quad (9)$$

De (8) y (9) tenemos que:

$$E_t s_{t+1}^0 = K s_t^0 + L \vartheta_t, \quad (10)$$

$$\text{donde } K \equiv (D + FA^{-1}B)^{-1} (G + HA^{-1}B)$$

$$L \equiv (D + FA^{-1}B)^{-1} (J + HA^{-1}C - FA^{-1}CP)$$

Suponemos que K admite la descomposición de Jordan:

$$K = M \Lambda M^{-1}; \quad \text{Sea } M^{-1} = \begin{bmatrix} u_1 & v_1 \\ u_2 & v_2 \end{bmatrix}, \Lambda = \begin{bmatrix} \mu_1 & 0 \\ 0 & \mu_2 \end{bmatrix}, |\mu_1| < 1, |\mu_2| > 1$$

El sistema de ecuaciones (10) puede ser expresado como sigue:

$$\begin{bmatrix} u_1 & v_1 \\ u_2 & v_2 \end{bmatrix} \begin{bmatrix} \hat{k}_{t+1} \\ E_t \hat{c}_{t+1} \end{bmatrix} = \begin{bmatrix} \mu_1 & 0 \\ 0 & \mu_2 \end{bmatrix} \begin{bmatrix} u_1 & v_1 \\ u_2 & v_2 \end{bmatrix} \begin{bmatrix} \hat{k}_t \\ \hat{c}_t \end{bmatrix} + \underbrace{\begin{bmatrix} u_1 & v_1 \\ u_2 & v_2 \end{bmatrix} \begin{bmatrix} L_{11} & L_{12} \\ L_{21} & L_{22} \end{bmatrix}}_Q \begin{bmatrix} \hat{\theta}_t \\ \hat{a}_t \end{bmatrix}$$

- *Condición de estabilidad (resolviendo la segunda ecuación de (Ω) hacia adelante, aplicando la ley de expectativas iteradas)*

$$\hat{c}_t = -\frac{u_2}{v_2} \hat{k}_t + \frac{Q_{21}/v_2}{\rho - \mu_2} \hat{\theta}_t + \frac{Q_{22}/v_2}{\rho_a - \mu_2} \hat{a}_t$$

- *Ecuación de estado (sustituyendo la condición de estabilidad en la primera ecuación de (Ω))*

$$\hat{k}_{t+1} = \mu_1 \hat{k}_t + \underbrace{\left[\frac{Q_{11} - \frac{\rho - \mu_1}{\rho - \mu_2} \frac{v_1}{v_2} Q_{21}}{u_1 - \frac{v_1}{v_2} u_2} \right]}_{N_1} \hat{\theta}_t + \underbrace{\left[\frac{Q_{12} - \frac{\rho_a - \mu_1}{\rho_a - \mu_2} \frac{v_1}{v_2} Q_{22}}{u_1 - \frac{v_1}{v_2} u_2} \right]}_{N_2} \hat{a}_t$$

- *Representación en el espacio de los estados*

$$\begin{bmatrix} \hat{k}_{t+1} \\ \hat{\theta}_{t+1} \\ \hat{a}_{t+1} \end{bmatrix} = \begin{bmatrix} \mu_1 & N_1 & N_2 \\ 0 & \rho & 0 \\ 0 & 0 & \rho_a \end{bmatrix} \begin{bmatrix} \hat{k}_t \\ \hat{\theta}_t \\ \hat{a}_t \end{bmatrix} + \begin{bmatrix} 0 & 0 \\ 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \varepsilon_{t+1} \\ \zeta_{t+1} \end{bmatrix},$$

$$\hat{c}_t = \begin{bmatrix} -\frac{u_2}{v_2} & \frac{Q_{21}/v_2}{\rho - \mu_2} & \frac{Q_{22}/v_2}{\rho_a - \mu_2} \end{bmatrix} \begin{bmatrix} \hat{k}_t \\ \hat{\theta}_t \\ \hat{a}_t \end{bmatrix}$$

Además,

$$\left. \begin{aligned} \hat{y}_t &= \hat{\theta}_t + \alpha \hat{k}_t + (1 - \alpha) \hat{h}_t \\ \hat{y}_t = \hat{c}_t + \hat{h}_t &= -\frac{u_2}{v_2} \hat{k}_t + \frac{Q_{21}/v_2}{\rho - \mu_2} \hat{\theta}_t + \frac{Q_{22}/v_2}{\rho_a - \mu_2} \hat{a}_t + \hat{h}_t \end{aligned} \right\} \Rightarrow$$

$$\underbrace{\begin{bmatrix} 1 & \alpha - 1 \\ 1 & -1 \end{bmatrix}}_{A1} \begin{bmatrix} \hat{y}_t \\ \hat{h}_t \end{bmatrix} = \underbrace{\begin{bmatrix} \alpha & 1 & 0 \\ -\frac{u_2}{v_2} & \frac{Q_{21}/v_2}{\rho - \mu_2} & \frac{Q_{22}/v_2}{\rho_a - \mu_2} \end{bmatrix}}_{B1} \begin{bmatrix} \hat{k}_t \\ \hat{\theta}_t \\ \hat{a}_t \end{bmatrix} \Rightarrow$$

$$\begin{bmatrix} \hat{y}_t \\ \hat{h}_t \end{bmatrix} = \underbrace{(A1)^{-1} B1}_{BT1} \begin{bmatrix} \hat{k}_t \\ \hat{\theta}_t \\ \hat{a}_t \end{bmatrix}$$

- **Filtro de Kalman**

Sistema general en espacio de los estados:

$$\underset{(r \times 1)}{\xi_{t+1}} = \underset{(r \times r)}{F} \underset{(r \times 1)}{\xi_t} + \underset{(r \times s)}{U} \underset{(s \times 1)}{v_{t+1}}, \quad \text{ecuación de estado}$$

$$\underset{(n \times 1)}{y_t} = \underset{(n \times k)}{A'} \underset{(k \times 1)}{x_t} + \underset{(n \times r)}{H'} \underset{(r \times 1)}{\xi_t} + \underset{(n \times 1)}{w_t}, \quad \text{ecuación de control}$$

$$E(v_t v_\tau') = \begin{cases} Q, & \text{para } t = \tau \\ 0, & \text{para } t \neq \tau \end{cases} \quad \underset{(s \times s)}$$

$$E(w_t w_\tau') = \begin{cases} R, & \text{para } t = \tau \\ 0, & \text{para } t \neq \tau \end{cases} \quad \underset{(n \times n)}$$

Equivalencias con nuestro modelo teórico:

$$\xi_t \equiv [\hat{k}_t, \hat{\theta}_t, \hat{a}_t]', \upsilon_t \equiv [\varepsilon_t, \zeta_t], F \equiv \begin{bmatrix} \mu_1 & N_1 & N_2 \\ 0 & \rho & 0 \\ 0 & 0 & \rho_a \end{bmatrix}, U \equiv \begin{bmatrix} 0 & 0 \\ 1 & 0 \\ 0 & 1 \end{bmatrix},$$

$$y_t \equiv [\hat{c}_t, \hat{h}_t]' \quad \text{ó} \quad y_t \equiv [\hat{y}_t, \hat{h}_t],$$

$$H' \equiv \begin{bmatrix} -\frac{u_2}{v_2} & \frac{Q_{21}/v_2}{\rho - \mu_2} & \frac{Q_{22}/v_2}{\rho_a - \mu_2} \\ BT_{21} & BT_{22} & BT_{23} \end{bmatrix}, \quad \text{ó} \quad H' \equiv BT,$$

$$A' \equiv 0, x_t \equiv 0, w_t \equiv 0, \forall t, Q \equiv \begin{bmatrix} \sigma_\varepsilon^2 & 0 \\ 0 & \sigma_\zeta^2 \end{bmatrix}, R \equiv 0$$

Derivación del filtro:

$$\text{Sean } \hat{\xi}_{t+1|t} \equiv \hat{E}(\xi_{t+1} | \Omega_t), P_{t+1|t} \equiv \hat{E} \left[\left(\xi_{t+1} - \hat{\xi}_{t+1|t} \right) \left(\xi_{t+1} - \hat{\xi}_{t+1|t} \right)' \right],$$

$$\text{donde } \Omega_t \equiv \left(y_t', y_{t-1}', \dots, y_1', x_t', x_{t-1}', \dots, x_1' \right)'$$

Condiciones iniciales:

$$\hat{\xi}_{1|0} = E(\xi_1) = 0$$

$$\text{Media incondicional: } E(\xi_{t+1}) = FE(\xi_t) \xrightarrow{E(\xi_{t+1})=E(\xi_t)} (I_r - F)E(\xi_t) = 0 \xRightarrow{\text{si } \det(I_r - F) \neq 0} E(\xi_t) = 0$$

$$P_{1|0} = E \left\{ \left[\xi_1 - E(\xi_1) \right] \left[\xi_1 - E(\xi_1) \right]' \right\} \rightarrow \text{vec}(P_{1|0}) = \left[I_{r^2} - (F \otimes F)^{-1} \right] \text{vec}(UQU').$$

Nótese que $\Sigma \equiv E(\xi_{t+1} \xi_{t+1}') = E \left[(F\xi_t + U\upsilon_{t+1})(F\xi_t + U\upsilon_{t+1})' \right] = FE(\xi_t \xi_t')F' + \dots$

$$UE(\upsilon_{t+1} \upsilon_{t+1}')U'$$

$$= F\Sigma F' + UQU' \Rightarrow$$

$$\text{vec}(\Sigma) = (F \otimes F) \text{vec}(\Sigma) + \text{vec}(UQU')$$

Calculando $\hat{\xi}_{t+1|t}$

$$K_t = FP_{t|t-1}H(H'P_{t|t-1}H + R)^{-1},$$

$$\hat{\xi}_{t+1|t} = F\hat{\xi}_{t|t-1} + K_t \left(y_t - A'x_t - H'\hat{\xi}_{t|t-1} \right),$$

$$P_{t+1|t} = FP_{t|t-1}F' - K_tH'P_{t|t-1}F' + UQU'$$

- **Estimación**

Si el estado inicial (ξ_1) y las innovaciones $\{w_t, v_t\}_{t=1}^T$ son gaussianas multivariantes, entonces $\hat{\xi}_{t+1|t}$ y $\hat{y}_{t+1|t}$ calculadas a través del filtro de Kalman, son óptimas y la distribución de y_t condicional sobre (x_t, Ω_{t-1}) es gaussiana:

$$y_t | (x_t, \Omega_{t-1}) \sim N\left(\left(A'x_t + H'\xi_{t|t-1}\right), \left(H'P_{t|t-1}H + R\right)\right)$$

Por tanto,

$$f_{Y_t|(X_t, \Omega_{t-1})} = (2\pi)^{-n/2} \left| H' P_{t|t-1} H + R \right|^{-1/2} \times \\ \exp \left\{ -\frac{1}{2} \left(y_t - A' x_t - H' \hat{\xi}_{t|t-1} \right)' \left(H' P_{t|t-1} H + R \right)^{-1} \left(y_t - A' x_t - H' \hat{\xi}_{t|t-1} \right) \right\}$$

Así, el logaritmo de la verosimilitud muestral es:

$$lfn = \sum_{t=1}^T \ln f_{Y_t|(X_t, \Omega_{t-1})}$$

El objetivo será encontrar los valores de los parámetros estructurales del modelo teórico que maximice el logaritmo de la verosimilitud muestral:

$\underset{\{\Theta\}}{Min} -lfn$, donde Θ es el vector que contiene los parámetros estructurales.

- **Descomposición de la varianza**

Sea el modelo empírico:

$$s_{t+1} = As_t + B\varepsilon_{t+1}, \quad (1)$$

$$d_t = Cs_t, \quad (2)$$

$$V \equiv E(\varepsilon_{t+1}\varepsilon'_{t+1}), \text{ diagonal}, \quad (3)$$

De (1): $(I - AL)s_{t+1} = B\varepsilon_{t+1} \Rightarrow s_t = \sum_{j=0}^{\infty} A^j B\varepsilon_{t-j}$

Así, $s_{t+k} = \sum_{j=0}^{\infty} A^j B\varepsilon_{t+k-j}$, $E_t(s_{t+k}) = \sum_{j=k}^{\infty} A^j B\varepsilon_{t+k-j}$, $\underbrace{s_{t+k} - E_t(s_{t+k})}_{\text{error de previsión a horizonte } k} = \sum_{j=0}^{k-1} A^j B\varepsilon_{t+k-j}$

$\underbrace{\Sigma_k^{(s)}}_{\text{varianza del error de previsión a horizonte } k} = E \left[\left(s_{t+k} - E_t(s_{t+k}) \right) \left(s_{t+k} - E_t(s_{t+k}) \right)' \right] =$

$$= BVB' + ABVB'A' + A^2BVB'(A^2)' + \dots + A^{k-1}BVB'(A^{k-1})'.$$

Además, $\lim_{k \rightarrow \infty} \Sigma_k^{(s)} = \Sigma^{(s)}$, donde $vec(\Sigma^{(s)}) = \left[I_{size(s_t)} - A \otimes A \right]^{-1} vec(BVB')$

$$\begin{aligned}
\text{De (2): } \Sigma_k^{(d)} &= E_t \left[(d_{t+k} - E_t d_{t+k})(d_{t+k} - E_t d_{t+k})' \right] \\
&= E_t \left[C(s_{t+k} - E_t s_{t+k})(s_{t+k} - E_t s_{t+k})' C' \right] \\
&= C \Sigma_k^{(s)} C'
\end{aligned}$$

$$\text{Además, } \Sigma^{(d)} = \lim_{k \rightarrow \infty} \Sigma_k^{(d)} = C \Sigma^{(s)} C'$$