

Structural VARs

W_0 matrix

$$\underline{Y}_t = \begin{bmatrix} Y_{1t} \\ \vdots \\ Y_{kt} \end{bmatrix} \left. \vphantom{\underline{Y}_t} \right\} \text{endogenous variables}$$

$$\bullet W_0 \underline{Y}_t = W_1 \underline{Y}_{t-1} + \dots + W_p \underline{Y}_{t-p} + u_t$$

$$\bullet E[u_t] = 0$$

$$\bullet E[u_t u_s'] = 0 \quad s \neq t$$

$$\bullet E[u_t u_t'] = \Omega, \quad E[u_{it} u_{jt}] = 0$$

Cannot use OLS, because we have endogeneity

$$\underbrace{\underline{Y}_t}_{\text{reduced form VAR}} = W_0^{-1} W_1 \underline{Y}_{t-1} + W_0^{-1} W_2 \underline{Y}_{t-2} + \dots + W_0^{-1} W_p \underline{Y}_{t-p} + W_0^{-1} u_t$$

$$\equiv A_1 \underline{Y}_{t-1} + \dots + A_p \underline{Y}_{t-p} + e_t$$

$$\bullet E[e_t] = E[W_0^{-1} u_t] = W_0^{-1} E[u_t] = 0$$

$$\bullet E[e_t e_t'] = E[W_0^{-1} u_t u_t' (W_0^{-1})'] \\ = W_0^{-1} \Omega (W_0^{-1})'$$

$$\bullet E[e_t e_s'] = W_0^{-1} \underbrace{E[u_t u_s']}_{=0} (W_0^{-1})' = 0$$

$$\text{but } E[e_{jt} e_{jt}] \neq 0$$

$$\begin{aligned} Y_t &= A_1 Y_{t-1} + \dots + A_p Y_{t-p} + \dots + e_t \\ &= A_1 L Y_t + \dots + A_p L^p Y_t + \dots + e_t \end{aligned}$$

$$\Rightarrow (I - A_1 L - \dots - A_p L^p - \dots) Y_t = e_t$$

$$\Rightarrow Y_t = A(L) e_t \quad \text{MA}(\infty) \text{ process}$$

$$\text{also, } Y_t = B(L) \bar{Y}_{t-1} + e_t \quad \text{VAR(1) process}$$

$$\begin{aligned} \text{since } Y_t &= A_1 Y_{t-1} + A_2 L Y_{t-1} + \dots + A_p L^{p-1} Y_{t-p} + \dots + e_t \\ &= (A_1 + A_2 L + \dots + A_p L^{p-1} + \dots) \bar{Y}_{t-1} + e_t \end{aligned}$$

Covariance stationarity

$E[Y_t]$ independent of t

$\text{Var}(Y_t)$ constant

$E[Y_t, Y_s]$ is a function of $(t-s)$.

$$\text{If } Y_t = \alpha Y_{t-1} + \beta Y_{t-2} + \phi_1 e_t + \phi_2 e_{t-1} \quad \text{ARMA}(2,1)$$

(*) Read first chapter of Hamilton

$$\begin{aligned} Y_t &= C(L) M_t \\ M_t &= Q e_t \end{aligned} \quad \left. \vphantom{\begin{aligned} Y_t &= C(L) M_t \\ M_t &= Q e_t \end{aligned}} \right\} \text{want this}$$

$$A(L)^{-1} Y_t = e_t$$

$$\Rightarrow \underbrace{Q A(L)^{-1}}_{\equiv C(L)^{-1}} Y_t = Q e_t = \eta_t$$

$$\Rightarrow Y_t = C(L) \eta_t$$

want orthogonal errors: $E[\eta_{it} \eta_{jt}] = 0 \quad \forall i \neq j$

$$\begin{aligned} \circ E[\eta_t \eta_t'] &= E[Q e_t e_t' Q'] = Q E[e_t e_t'] Q' \\ &= \Sigma = \omega_0^{-1} \Omega (\omega_0^{-1})' \\ &= Q \Sigma Q' = I \end{aligned}$$

want $\Sigma = Q^{-1} (Q^{-1})'$

◦ can use the Choleski decomposition:
 $A = LL'$, L lower triangular

◦ this is not unique, because any R s.t.
 $RR' = I$ gives us: $RQ \Sigma Q' R'$
 $= R I R' = R R' = I$

$C(L)$ restrictions to pick "right" Q matrix

$$\circ C(L) = A(L) Q^{-1}$$

◦ short-run restrictions: $C(0)$: $Y_t = C_0 \eta_t + C_1 \eta_{t-1} + \dots$

$$C_0 = \begin{bmatrix} c_{11}^0 & c_{12}^0 \\ c_{21}^0 & c_{22}^0 \end{bmatrix}$$

◦ want this lower triangular want to restrict some of these elements to be zero (i.e. $c_{12}^0 = 0$)

$$\Rightarrow C(\lambda) = A(\lambda)Q^{-1}$$

$$\Rightarrow A(\lambda)^{-1}C(\lambda) = Q^{-1}$$

$$\text{Also, } C(\lambda)C(\lambda)' = A(\lambda)Q^{-1}(Q^{-1})'A(\lambda)'$$

Long-run restrictions: Blanchard, Quah (1989)

$$\begin{bmatrix} \Delta Y_t \\ \Delta M_t \end{bmatrix} = C(L)\eta_t$$

• impose restrictions on $C(L)$

• call these restrictions (LR)

$$C(L) = \sum_{j=0}^{\infty} c_j L^j$$

$$c_0 + c_1 L + c_2 L^2 + \dots$$

Read Blanchard and Quah.