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**Name:** Using the Generalized Schur form to Solve a Multivariate Linear Rational Expectations Model

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## Model

### The problem

$$A\mathbb{E}[x_{t+1}|\mathcal{F}_t] = Bx_t + Cz_t$$

- *Stochastic Space:*  $(\Omega, \mathcal{F}, P)$
- *Filtration:*  $\tilde{\mathcal{F}} = \{\mathcal{F}_t; t = 0, 1, 2, \dots\}$
- *$\tilde{\mathcal{F}}$ -adapted stochastic process:*  $z = \{z_t; t = 0, 1, \dots\}$ , exogenously given, with  $n_z$  dimensional

## Mathematic Background

■ **Generalized Eigenvalues:** Let  $P : \mathbb{C} \rightarrow \mathbb{C}^{n \times n}$  be a matrix-valued function of a complex variable (a matrix pencil). Then the set of its generalized eigenvalues is defined via

$$\lambda(P) = \{z \in \mathbb{C} : |P(z)| = 0\}$$

- When  $P(z) = Az - B$ , then the set of generalized eigenvalues  $\lambda(A, B)$
- When exists  $x \in \mathbb{C}^n$ , such that  $Bx = \lambda Ax$ , then  $\lambda \in \mathbb{C}$  is a generalized eigenvalue

■ **Regular:** Let  $P(z)$  be a matrix pencil. Then P is said to be regular if there is  $z \in \mathbb{C}$  such that  $|P(z)| \neq 0$



## Complex Generalized Schur Form:

Let  $A$  and  $B$  be  $n \times n$  matrices of complex numbers such that  $P(z) = Az - B$  is a regular matrix pencil. Then there exist unitary  $n \times n$  matrices of complex numbers  $Q$  and  $Z$  such that

🔥 There exist following decompositions, here both  $T$  and  $S$  are **upper triangular**

$$QAZ = S, \quad QBZ = T$$

🔥 The corresponding diagonal elements are

- For each  $i$ ,  $s_{ii}$  and  $t_{ii}$  are not both zero
- The pairs  $(s_{ii}, t_{ii})$ ,  $i = 1, 2, \dots, n$  can be arranged in any order.

🔥 The generalized eigenvalues are

$$\lambda(A, B) = \left\{ \frac{t_{ii}}{s_{ii}} : s_{ii} \neq 0 \right\}$$

- Unstable eigenvalues  $\begin{cases} \text{Finite Unstable} \\ \text{Infinite} \end{cases}$ , means  $|\lambda| > 1$
- Stable eigenvalues means  $|\lambda| < 1$

## Algorithm

## Assumption

- **Stability:** Let  $x$  be a stochastic process with values in  $\mathbb{R}^n$ ,  $x$  is stable if there is an  $M$  such that

$$\begin{aligned} \|x_t\|_{max} &\leq M \\ \downarrow \\ \|x_t\|_{max} &= \max_i \sqrt{E[|x_i|]} \end{aligned}$$

The unconditionally expected values of the moduli of the elements of  $x_t$  do not blow up as  $t$  increases beyond all bounds.

- **Backward-looking:** Let  $(\Omega, \mathcal{F}, P, \tilde{\mathcal{F}})$  be a filtered probability space. A process  $k$  is called backward-looking if
  1. The prediction error  $\xi$  defined via  $\xi_{t+1} \equiv k_{t+1} - E[k_{t+1}|\mathcal{F}_t]$  is an exogenous martingale difference process, and
    - ☀  **$(P, \tilde{\mathcal{F}})$ -Martingale Difference Process:** Let  $(\Omega, \mathcal{F}, P, \tilde{\mathcal{F}})$  be a filtered probability space. A vector process  $\xi$  is called a martingale difference process if
      1.  $\xi$  is adapted to  $\tilde{\mathcal{F}}$
      2.  $E[\xi_{t+1}|\tilde{\mathcal{F}}_t] = 0$  for each  $t = 0, 1, \dots$
  2.  $k_0 \in \mathcal{F}_0$  is exogenously given

### ✓ Assumptions:

- **Assumption 1:** The exogenous  $n_z$ -dimensional process  $z$  is stable and adapted to the given filtration  $\tilde{\mathcal{F}}$ .

$$z_t = \phi(z_{t-1}) + \epsilon_t$$

- **Assumption 2:**  $k_0$  is an exogenously given  $\tilde{\mathcal{F}}_0$ -measurable random variable and

$$k_{t+1} - E[k_{t+1}|\mathcal{F}_t] = \xi_{t+1}$$

Here  $\xi$  is a Martingale Difference Process

- **Assumption 3:** There exists a  $z \in \mathbb{C}$  such that  $|Az - B| \neq 0$
- **Assumption 4:** There is no  $z \in \mathbb{C}$  such that  $|Az - B| = 0$  and  $|z| = 1$
- **Assumption 5:**  $Z_{11}$  is square and invertible.

## Using the Generalized Schur Form

➤ **Triangularizing the system:** to find an upper triangular system of expectational difference equations in the auxiliary variables  $y_t$  defined via

$$y_t \equiv Z^H x_t = Z^H \begin{bmatrix} k_t \\ d_t \end{bmatrix} = \begin{bmatrix} s_t \\ u_t \end{bmatrix}$$

- $k_t$ : backward looking variables
- $s_t$ : stable variables
- $u_t$ : unstable variables

⇓

$$\begin{aligned} \star \quad & \mathbb{S}\mathbb{E}[y_{t+1} | \mathcal{F}_t] = T y_t + Q C z_t \\ & \Rightarrow \begin{bmatrix} S_{11} & S_{12} \\ 0 & S_{22} \end{bmatrix} \mathbb{E} \left\{ \begin{bmatrix} s_{t+1} \\ u_{t+1} \end{bmatrix} \middle| \mathcal{F}_t \right\} = \begin{bmatrix} T_{11} & T_{12} \\ 0 & T_{22} \end{bmatrix} \begin{bmatrix} s_t \\ u_t \end{bmatrix} + \begin{bmatrix} Q_1 \\ Q_2 \end{bmatrix} C z_t \end{aligned}$$

➤ **Solving the triangular system:**

📍 **Solving for  $u_t$ :**

1. When  $n_u$  is not large, Hint:  $u_\infty = 0$  to get a solution

### Forward Iteration for Unstable Variables

$$u_t = -T_{22}^{-1} \sum_{k=0}^{\infty} [T_{22}^{-1} S_{22}]^k Q_2 C \mathbb{E}[z_{t+k} | \mathcal{F}_t]$$

When  $z$  is a stationary VAR process with autocorrelation matrix  $\Phi$

$$\begin{aligned} u_t &= M z_t \\ \downarrow \\ \text{vec}(M) &= [\Phi^T \otimes S_{22} - I_{n_z} \otimes T_{22}]^{-1} \text{vec}[Q_2 C] \end{aligned}$$

**Hints:**

$$\left. \begin{aligned} S &= \sum_{k=0}^{\infty} \Phi^k A \Psi^k = A + \sum_{k=1}^{\infty} \Phi^k A \Psi^k \\ \Phi S \Psi &= \sum_{k=1}^{\infty} \Phi^k A \Psi^k \end{aligned} \right\} \Rightarrow S - \Phi S \Psi = A$$

$$\Rightarrow \text{vec}(S) - (\Psi^T \otimes \Phi) \text{vec}(S) = \text{vec}(A)$$

$$\Rightarrow \text{vec}(S) = [I - \Psi^T \otimes \Phi]^{-1} \text{vec}(A)$$

and

$$\mathbb{E}_t z_{t+k} = \Phi^k z_t$$

2. When  $n_u$  is large, calculating row by row.

📍 **Solving for  $s_t$ :**

$$\begin{aligned}
& \star \\
& \Downarrow \\
\mathbb{E}[s_{t+1}|\mathcal{F}_t] &= S_{11}^{-1}T_{11}s_t + S_{11}^{-1}\{T_{12}u_t - S_{12}\mathbb{E}[u_{t+1}|\mathcal{F}_t] + Q_1Cz_t\} \\
& \Downarrow (u_t = Mz_t, \mathbb{E}z_{t+1} = \Phi z_t) \\
\mathbb{E}[s_{t+1}|\mathcal{F}_t] &= S_{11}^{-1}T_{11}s_t + S_{11}^{-1}\{T_{12}M - S_{12}M\Phi + Q_1C\}z_t
\end{aligned}$$

$$\begin{aligned}
\underbrace{\begin{bmatrix} k_{t+1} \\ d_{t+1} \end{bmatrix} = \begin{bmatrix} Z_{11} & Z_{12} \\ Z_{21} & Z_{22} \end{bmatrix} \begin{bmatrix} s_{t+1} \\ u_{t+1} \end{bmatrix}}_{\Downarrow} \quad k_{t+1} - \mathbb{E}[k_{t+1}|\mathcal{F}_t] = \xi_{t+1} \\
Z_{11}(s_{t+1} - \mathbb{E}[s_{t+1}|\mathcal{F}_t]) + Z_{12}(u_{t+1} - \mathbb{E}[u_{t+1}|\mathcal{F}_t]) = \xi_{t+1} \\
\Downarrow (z_t = \phi(z_{t-1}) + \epsilon_t) \\
s_{t+1} - \mathbb{E}[s_{t+1}|\mathcal{F}_t] = Z_{11}^{-1}\xi_{t+1} - Z_{11}^{-1}Z_{12}M\epsilon_{t+1}
\end{aligned}$$

$$\begin{aligned}
s_{t+1} &= S_{11}^{-1}T_{11}s_t + S_{11}^{-1}\{T_{12}M - S_{12}M\Phi + Q_1C\}z_t \\
&+ Z_{11}^{-1}\xi_{t+1} - Z_{11}^{-1}Z_{12}M\epsilon_{t+1}
\end{aligned}$$

➤ Eliminating the auxiliary process  $y_t$