

## Forecast error decomposition

Useful for calculating the proportion of movements attributable a variable's own shocks versus other variables.

recall the VMA( $\infty$ ) representation from previous note:

$$x_t = \mu + \sum_{i=0}^{\infty} \phi_i \varepsilon_{t-i} \rightarrow E(x) = \mu + \sum_{i=h}^{\infty} \phi_i \varepsilon_{t+h-i}$$

$$x_{t+h} - E_t(x_{t+h}) = \sum_{i=0}^{h-1} \phi_i \varepsilon_{t+h-i} \rightarrow \text{the } h\text{-period ahead forecast error}$$

$$1 = \frac{\sigma_y^2}{\sigma_y^2(h)} \sum_{i=0}^{h-1} \phi_{11}^2(i) + \frac{\sigma_z^2}{\sigma_y^2(h)} \sum_{i=0}^{h-1} \phi_{12}^2(i)$$

Prop. of  $\sigma_y^2(h)$  that is due to shocks in the  $\{\varepsilon_{1t}\}$  sequence

Prop. of  $\sigma_y^2(h)$  that is due to shocks in the  $\{\varepsilon_{2t}\}$  sequence

notes:

- In practice, forecast error typically driven by own shocks in SR & other variables in LR
- Just like impulse response analysis, variance decomp reqs the estimation of some matrix which requires restriction. we can again rely on the Cholesky decomposition.
- IF the VAR system is stable, and  $h$  is increasing, variance decomp should be approaching some constant.

