

More Complex State Space Models

observation
equation

$$y_t = \tilde{a}' \tilde{x}_t + u_t, \quad u_t \stackrel{iid}{\sim} N(0, \sigma_u^2)$$

state
equation

$$\tilde{x}_t = \tilde{A} \tilde{x}_{t-1} + \tilde{v}_t, \quad \tilde{v}_t \stackrel{iid}{\sim} N(0, \Sigma_v)$$

VAR(1) model for \tilde{x}_t

initial
conditions

$$\tilde{x}_1 \text{ or } \tilde{x}_0 = \mu_{\tilde{x}}$$

Even though we can write a model like this down, it is tough to say if the parameters are identified, don't usually use in practice

Multivariate state space models have the form

observation equation $y_t \text{ rx1} = A_{\text{rxs}} \underset{\sim}{x}_t \text{ sx1} + u_t \text{ rx1}, \quad u_t \underset{\sim}{\sim} N(0, \Sigma_u)$

state equation $\underset{\sim}{x}_t \text{ sx1} = \Gamma_{\text{sxs}} \underset{\sim}{x}_{t-1} \text{ sx1} + v_t \text{ sx1}, \quad v_t \underset{\sim}{\sim} N(0, \Sigma_v)$

Vector autoregressive process of order 1

Parameters:

initial conditions, $\underset{\sim}{x}_1$ or $\underset{\sim}{x}_0 = \underset{\sim}{\mu}$

$$\text{rxs} + \frac{\text{r}(\text{r}-1)}{2} + \text{s} + \text{sxs} + \frac{\text{s}(\text{s}-1)}{2}$$

* Could have $S=r \Rightarrow$ our state process is also multivariate with same dim. as observed

* We could have $S < r$, e.g. $S=1$

much simpler, lower dimensional processes than in VARMA

if $S=1$

Multivariate state space models have the form

observation equation $y_{t \times 1} = \underset{\sim}{a}_{\times 1} x_t + \underset{\sim}{u}_{t \times 1}, \quad u_t \stackrel{iid}{\sim} N(0, \Sigma_u)$

state equation $x_t = \alpha_1 x_{t-1} + v_t \quad v_t \stackrel{iid}{\sim} N(0, \sigma_v^2)$

autoregressive process of order 1

initial condition x_1 or $x_0 = \mu$

$p(y | A, \Sigma_u, \Gamma, \Sigma_v, \mu)$ may not be convex.

$$x_1 \sim N\left(0, \frac{\sigma_v^2}{1 - \gamma_{11}^2}\right) \quad \text{initial condition}$$

ARM A in state space framework...

First, we'll consider AR(1)

$$u_t, u_{t-1} \stackrel{iid}{\sim} N(0, \sigma_u^2)$$

$$\sigma_u^2 = 0$$

$$\begin{pmatrix} y_t \\ y_{t-1} \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} x_t \\ x_{t-1} \end{pmatrix} + \begin{pmatrix} u_t \\ u_{t-1} \end{pmatrix}$$

$$\begin{pmatrix} v_t \\ v_{t-1} \end{pmatrix} \sim N(0, \Sigma_v), \quad \sigma_{v22} = \sigma_{12} = \sigma_{21} = 0, \quad \sigma_{v11}^2 \text{ freely varying}$$

$$\begin{pmatrix} x_t \\ x_{t-1} \end{pmatrix} = \begin{pmatrix} \gamma_{11} & 0 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} x_{t-1} \\ x_{t-2} \end{pmatrix} + \begin{pmatrix} v_t \\ v_{t-1} \end{pmatrix}$$

$$\begin{pmatrix} x_t \\ x_{t-1} \end{pmatrix} = \begin{pmatrix} \gamma_{11} x_{t-1} \\ x_{t-1} \end{pmatrix} + \begin{pmatrix} v_t \\ v_{t-1} \end{pmatrix}$$

Biggest omission:

GARCH, ARCH, Stochastic
Volatility