#### 1 Clark87

clark0

SSM

The UC model of Clark (1987) is a generalisation of the HP–Filter (a local linear trend model) that can be expressed as an SSM taking the following form:

 $y_t = y_t^* + \tilde{y}_t \tag{1a}$ 

$$\Delta y_t^* = g_{t-1} + \sigma_1 \varepsilon_{1t} \tag{1b}$$

$$\Delta g_t = \sigma_2 \varepsilon_{2t} \tag{1c}$$

$$a(L)\tilde{y}_t = \sigma_3 \varepsilon_{3t},\tag{1d}$$

where the shocks  $\{\varepsilon_{it}\}_{i=1}^3$  are assumed to be *i.i.d.* standard normal and mutually uncorrelated, with standard deviation  $\sigma_i$  and a(L) is commonly assumed to be a stable AR(2), so that  $a(L) = (1 - a_1L - a_2L^2)$ . The only observable is  $y_t$  (generally 100 times) the log of real GDP and with the cycle (denoted by  $\tilde{y}_t$ ) now allowed to be serially correlated, following a stationary AR(2) process. There are 3 shocks in the model.

The 'numbered' shock to 'named' shock mapping is:

$$\begin{bmatrix} \varepsilon_{1t} \\ \varepsilon_{2t} \\ \varepsilon_{3t} \end{bmatrix} = \begin{bmatrix} \varepsilon_t^{y^*} \\ \varepsilon_t^{g} \\ \varepsilon_t^{\tilde{y}} \end{bmatrix}. \tag{2}$$

# 2 Shock recovery SSM

## 2.1 SSM with lagged states

Kurz's (2018) SSM has the following general from:

Measurement: 
$$Z_t = D_1 X_t + D_2 X_{t-1} + R \varepsilon_t$$
 (3a) ssm1

State: 
$$X_t = AX_{t-1} + C\varepsilon_t$$
, (3b) ssm2

where  $\varepsilon_t \sim MN(0, I_m)$ ,  $D_1$ ,  $D_2$ , A, R are C are conformable system matrices,  $Z_t$  the observed variable and  $X_t$  the latent state variable.

## 2.2 Clark87 SSM for shock recovery

To assess recovery, re-write the model in (1) in 'shock recovery' State Space Form (SSF). That is, collect all observables in  $Z_t$  and all shocks (and other state variables) in  $X_t$  to yield:

$$\Delta^{2} y_{t} = \Delta^{2} y_{t}^{*} + \Delta^{2} \tilde{y}_{t}$$

$$= \sigma_{1} \Delta \varepsilon_{1t} + \Delta g_{t-1} + \Delta^{2} \tilde{y}_{t}$$

$$= \sigma_{1} \Delta \varepsilon_{1t} + \sigma_{2} \varepsilon_{2t-1} + a(L)^{-1} \Delta^{2} \sigma_{3} \varepsilon_{3t}$$

$$\Leftrightarrow a(L) \Delta^{2} y_{t} = \sigma_{1} a(L) \Delta \varepsilon_{1t} + \sigma_{2} a(L) \varepsilon_{2t-1} + \sigma_{3} \Delta^{2} \varepsilon_{3t}$$

$$Z_{t} = \sigma_{1} \left[ \varepsilon_{1t} - (2 + a_{1}) \varepsilon_{1t-1} + (1 + 2a_{1} - a_{2}) \varepsilon_{1t-2} + (2a_{2} - a_{1}) \varepsilon_{1t-3} - a_{2}\varepsilon_{1t-4} \right]$$

$$+ \sigma_{2} (\varepsilon_{2t-1} - a_{1}\varepsilon_{2t-2} - a_{2}\varepsilon_{2t-3})$$

$$+ \sigma_{3} (\varepsilon_{3t} - 2\varepsilon_{3t-1} + \varepsilon_{3t-2}),$$

$$(4) z$$

where I have made use of the fact:

$$a(L)\Delta^{2} = (1 - a_{1}L - a_{2}L^{2})(1 - L)^{2}$$

$$= 1 - (2 + a_{1})L + (1 + 2a_{1} - a_{2})L^{2} + (2a_{2} - a_{1})L^{3} - a_{2}L^{4},$$
(5) expal

and  $Z_t = a(L)\Delta^2 y_t$  is the only observed variable.

The Measurement and State equations corresponding to (4) are:

KOSSM

Measurement :  $Z_t = D_1 X_t + D_2 X_{t-1} + R \varepsilon_t$ 

$$Z_{t} = \underbrace{\begin{bmatrix} \sigma_{1} & 0 & \sigma_{3} & -\sigma_{1} (2+a_{1}) & \sigma_{1} (1+2a_{1}-a_{2}) & \sigma_{1} (2a_{2}-a_{1}) & 0 & 0 & 0 \end{bmatrix}}_{D_{1}} \underbrace{\begin{bmatrix} \varepsilon_{1t} \\ \varepsilon_{2t} \\ \varepsilon_{3t} \\ \varepsilon_{1t-1} \\ \varepsilon_{1t-2} \\ \varepsilon_{1t-3} \\ \varepsilon_{2t-1} \\ \varepsilon_{2t-2} \\ \varepsilon_{3t-1} \end{bmatrix}}_{X_{t}}$$

$$(6a)$$

$$+\underbrace{\begin{bmatrix}0 & \sigma_{2} & -2\sigma_{3} & 0 & 0 & -\sigma_{1}a_{2} & -\sigma_{2}a_{1} & -\sigma_{2}a_{2} & \sigma_{3}\end{bmatrix}}_{D_{2}}\underbrace{\begin{bmatrix}\varepsilon_{1t-1}\\\varepsilon_{2t-1}\\\varepsilon_{3t-1}\\\varepsilon_{1t-2}\\\varepsilon_{1t-3}\\\varepsilon_{1t-4}\\\varepsilon_{2t-2}\\\varepsilon_{2t-3}\\\varepsilon_{3t-2}\end{bmatrix}}_{X_{t-1}}+\underbrace{\begin{bmatrix}0 & 0 & 0\end{bmatrix}}_{R}\underbrace{\begin{bmatrix}\varepsilon_{1t}\\\varepsilon_{2t}\\\varepsilon_{2t}\\\varepsilon_{3t}\end{bmatrix}}_{\varepsilon_{t}}$$

$$(6b)$$

State:  $X_t = AX_{t-1} + C\varepsilon_t$ ,

#### Alternative formulation 3

Try to reduce the state vector  $X_t$  by using changes:

$$\Delta^2 y_t = \Delta^2 y_t^* + \Delta^2 \tilde{y}_t \tag{7}$$

$$\Delta^2 y_t^* = \sigma_2 \varepsilon_{2t-1} + \sigma_1 \Delta \varepsilon_{1t} \tag{8}$$

$$\Delta^2 \tilde{y}_t = a(L)^{-1} \sigma_3 \Delta^2 \varepsilon_{3t},\tag{9}$$

so that

$$\Delta^{2} y_{t} = \Delta^{2} y_{t}^{*} + \Delta^{2} \tilde{y}_{t}$$

$$= \sigma_{2} \varepsilon_{2t-1} + \sigma_{1} \Delta \varepsilon_{1t} + a(L)^{-1} \sigma_{3} \Delta^{2} \varepsilon_{3t}$$

$$\Leftrightarrow \underline{a(L)} \Delta^{2} y_{t} = \sigma_{2} a(L) \varepsilon_{2t-1} + \sigma_{1} a(L) \Delta \varepsilon_{1t} + \sigma_{3} \Delta^{2} \varepsilon_{3t}$$

$$Z_{t} = \sigma_{1} a(L) \Delta \varepsilon_{1t} + \sigma_{2} a(L) \varepsilon_{2t-1} + \sigma_{3} \Delta^{2} \varepsilon_{3t}$$

$$= \sigma_{1} \Delta \varepsilon_{1t} - a_{1} \sigma_{1} \Delta \varepsilon_{1t-1} - a_{2} \sigma_{1} \Delta \varepsilon_{1t-2}$$

$$+ \sigma_{2} \varepsilon_{2t-1} - a_{1} \sigma_{2} \varepsilon_{2t-2} - a_{2} \sigma_{2} \varepsilon_{2t-3}$$

$$+ \sigma_{3} \Delta \varepsilon_{3t} - \sigma_{3} \Delta \varepsilon_{3t-1}$$

This is correct now:

This is correct now: 
$$\begin{bmatrix} \varepsilon_{1t} \\ \varepsilon_{2t} \\ \varepsilon_{3t} \\ \varepsilon_{2t-2} \\ \Delta \varepsilon_{1t-1} \\ \Delta \varepsilon_{3t} \end{bmatrix} + \begin{bmatrix} -\sigma_1 & \sigma_2 & 0 & -a_2\sigma_2 & -a_2\sigma_1 & -\sigma_3 \end{bmatrix} \begin{bmatrix} \varepsilon_{1t-1} \\ \varepsilon_{2t-1} \\ \varepsilon_{3t-1} \\ \varepsilon_{2t-3} \\ \Delta \varepsilon_{1t-2} \\ \Delta \varepsilon_{3t-1} \end{bmatrix} = \sigma_2 \varepsilon_{2t-1} - \sigma_1 \varepsilon_{t-1} + \sigma_1 \varepsilon_{t} + \Delta \sigma_3 \varepsilon_{3t} - \Delta \sigma_3 \varepsilon_{3t-1} - \sigma_2 a_1 \varepsilon_{2t-2} - \sigma_2 a_2 \varepsilon_{2t-3} - \Delta \sigma_1 a_1 \varepsilon_{t-1} - \Delta \sigma_1 a_2 \varepsilon_{t-2}$$

Measurement :  $Z_t = D_1 X_t + D_2 X_{t-1} + R \varepsilon_t$ 

$$Z_{t} = \underbrace{\begin{bmatrix} \sigma_{1} & 0 & \sigma_{3} & -\sigma_{1} (2+a_{1}) & \sigma_{1} (1+2a_{1}-a_{2}) & \sigma_{1} (2a_{2}-a_{1}) & 0 & 0 & 0 \end{bmatrix}}_{D_{1}} \underbrace{\begin{bmatrix} \varepsilon_{1t} \\ \varepsilon_{2t} \\ \varepsilon_{3t} \\ \varepsilon_{1t-1} \\ \varepsilon_{1t-2} \\ \varepsilon_{1t-3} \\ \varepsilon_{2t-1} \\ \varepsilon_{2t-2} \\ \varepsilon_{3t-1} \end{bmatrix}}_{X_{t}}$$

$$+ \underbrace{\begin{bmatrix} 0 & \sigma_{2} & -2\sigma_{3} & 0 & 0 & -\sigma_{1}a_{2} & -\sigma_{2}a_{1} & -\sigma_{2}a_{2} & \sigma_{3} \end{bmatrix}}_{D_{2}} \underbrace{\begin{bmatrix} \varepsilon_{1t} \\ \varepsilon_{2t} \\ \varepsilon_{3t-1} \\ \varepsilon_{2t-1} \\ \varepsilon_{3t-1} \\ \varepsilon_{1t-2} \\ \varepsilon_{2t-3} \\ \varepsilon_{2$$

(11)