

STAR SMOOTH TRANSITION AUTO-REGRESSION MODEL.

All smoothing models are
based by probability

Considers y_t to follow a ^{univariate case} STAR ^{2 Regimes}

model such that

$$y_t = \phi_1 x_t + [1 - G(s_t, c_t, \gamma)] \phi_2' x_t + e_t$$

$\downarrow$ Degree of Smoothing
 $\downarrow$ Threshold parameter
 $\downarrow$ Transition parameter

where $x_t = [1, \hat{x}_2']$ such that

$$\phi_{ik} = [\phi_{i0}, \phi_{i1}, \phi_{i2}, \dots, \phi_{ip}]$$

$$\phi_2' x_t + G(s_t, c_t, \gamma) \phi_1' x_t + e_t$$

$$\hat{x}_t = [y_{t-1}, y_{t-2}, \dots, y_{t-n}]$$

This model can be extended to include other exogenous variables.

STAR-X is given as

$$\textcircled{II} \quad y_t = \phi_1' x_t [1 - G(s_t, c_t, \gamma)] + \phi_2' x_t [G(s_t, c_t, \gamma)] +$$

$$z_{1t} + z_{2t} + \dots + z_{kt} + e_t$$

$$E(y_t / \mathcal{I}_{t-1}) = 0, \quad E(e_t / \mathcal{I}_{t-1}) = 0$$

Information set

$$s_t = y_{t-d}, \quad d > 0$$

$$s_t = y_t, \quad d = 0$$

$$s_t = t$$

$$z_t = [z_{1t}, z_{2t}, \dots, z_{kt}]$$

SETAR

$$s_t = z_{1t}$$

+ two regimes if

$$G = 0 - 1$$

STAR-

M Regimes if G
0-1

assumes continues value between

(Regime)

What Define G-Function Behave

It defines the regime
and value of transition
function.

WE WILL SEE HOW WE
PROB. DISTRIBUTION

$$G(S_t, C_{t+1}, \gamma) = \left[1 + \exp \left(-\gamma (S_t - C) \right) \right]^{-1} \quad (111)$$

if $\gamma > 0$, $C \approx 0$, $S_t = Y_{t-1}$

G is said to follow 1st order logistic function. such that

Equation 1 and 3 gives us

Equation 2 and 3 gives us

DETERMINE THE CHOICE OF

$$\left[1 + \exp \left(-\gamma (S_t - C) \right) \right]^{-1} = (111)$$

LOGISTIC SMOOTHING TRANSITION

LSTAR

LSTAR-X

$$\left[1 + \exp \left(-\gamma (S_t - C) \right) \right]^{-1} = (111)$$

$$G(S_t, C, \gamma) = [1 + \exp(-\gamma (S_t - c_2))]^{-1} \quad (iv)$$

It is just C and not C_t

If $C_1 \neq C_2$, $C = (C_1, C_2)$

Combine (i) and (iv) | 2nd
(ii) and (iv) |

STAR

STAR-X

This can be generalised to a function such that

$$G(S_t, C, \gamma) = [1 + \exp(-\gamma \prod_{i=1}^n (S_t - c_i))]^{-1}$$

$$C_1 \leq C_2 \leq C_3 \dots \leq C_n$$

$$C = (C_1, C_2, \dots, C_n)$$

summation of the likelihood prob. distribution

likelihood function

The condition that γ Gamma > 0

is a must

Condition otherwise

the model will collapse.

ANOTHER FUNCTION

$$G(S_t, C, \gamma) = \left[1 - \exp\left[-\gamma (S_t - C)^2\right] \right]$$

This is an
exponential
form.

I and V	- E	STAR
II and V		STAR-X

(A) If $G = 1$ $S_t \rightarrow -\infty$ to $+\infty$

$$\int_{-\infty}^{+\infty} S_t$$

(B) $G = 0$ $S_t = C$

③ If $\gamma = 0$ or $\gamma \rightarrow \infty$

the function collapse to constant and the model become linear

MULTIPLE REGIMES

STAR

① if M regimes are characterised by one transition variable S_t

CASE

② if M regimes are characterised by many transition variables i.e. $(S_{1t}, S_{2t}, \dots, S_{Mt})$

CASE 1

Consider 2 regimes STAR Second regime

$$y_t = \phi'_1 x_t + (\phi'_2 - \phi'_1) x_t G_t(S_t, c_1, \gamma) + e_t$$

2 Regime

STAR MODEL

If you have to go for 3 regime STAR

$$y_t = \underbrace{\phi_1'}_{\text{linear}} x_t + (\phi_2' - \phi_1') x_t G_1$$

If G_1 is exponential then G_2 must also be exponential.

It can be extended to

$$y_t = \phi_1' x_t + (\phi_2' - \phi_1') x_t G_1(s_t, c_1, \gamma) + (\phi_3' - \phi_2') x_t G_2(s_t, c_2, \gamma)$$

$$+ \dots + (\phi_m' - \phi_{m-1}') x_t G_{m-1}(s_t, c_{m-1}, \gamma) + e_t$$

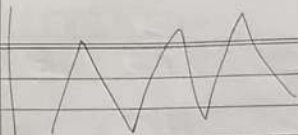
Threshold point

$$(s_t, c_1, \gamma) + (\phi_3' - \phi_2') x_t$$

$$G_2(s_t, c_1, \gamma) + e_t$$

multiple regimes. (Generalised case)

$$+ \dots + (\phi_m' - \phi_{m-1}') x_t G_{m-1}(s_t, c_{m-1}, \gamma) + e_t$$



CASE 2

TWO REGIMES

$$y_t = \left[\phi_1' x_t (1 - G_1(s_{1t}, \gamma_1, c_1)) + \phi_2' x_t G_1(s_{1t}, \gamma_1, c_1) \right]$$

Within the 2 regimes there are 2 transitions.

either the process is driven by first transition or 2nd transition

$$+ \left[1 - G_2(s_{2t}, \gamma_2, c_2) \right] \text{ transition}$$

Beyond this bar is the second regime

$$+ \left[\phi_3' x_t (1 - G_1(s_{1t}, \gamma_1, c_1)) + \phi_4' x_t G_1(s_{1t}, \gamma_1, c_1) \right]$$

$$+ G_2(s_{2t}, \gamma_2, c_2) + e_t$$

Smoothing transition VAR

OR
Vector STAR

OR
Multivariate STAR

Next week.

26

VECTOR STAR

OR

SMOOTHING TRANSITION

VAR MODEL.

Consider y_t to be K -dimensional vector of stationary variables such that

$$y_t = (y_{1t}, y_{2t}, \dots, y_{Kt})'$$

classmate

$$y_t = \left(\phi_{1,0} + \phi_{1,1} y_{t-1} + \dots + \phi_{1,p} y_{t-p} \right) (1 - G(s_t, \gamma, c))$$

$K \times 1$ $K \times K$

slope of first lag of first regime

$$+ \left(\phi_{2,0} + \phi_{2,1} y_{t-1} + \dots + \phi_{2,p} y_{t-p} \right) (G(s_t, \gamma, c)) + e_t$$

$\phi_{i,0}$ $i=1, 2$ $K \times 1$ vectors

ϕ_{ij} $i=1, 2$ and $j=1, 2$ $K \times K$ matrix

two-regime ST-VAR

$$e_t = (e_{1t}, e_{2t}, \dots, e_{kt}) \quad K \times 1$$

The error is white noise with 0 mean and positive definite covariance matrix Σ
 (idiosyncratic error term)

TRANSITION FUNCTION

$$G_1(S_{1t}, Y_{1t}, C_{1t}) \dots \dots \dots$$

$$\dots \dots \dots G_K(S_{Kt}, Y_{Kt}, C_{Kt}).$$

Each equation will be guided by a specific transition function.

Alternatively you can give each equation a specific transition function

THIS CAN BE EXTENDED TO MULTIPLE REGIMES

$$Y_t = (\phi_{1,0} + \phi_{1,1} Y_{t-1} + \dots + \phi_{1,p})$$

$$(\phi_{1,0} + \phi_{1,1} Y_{t-1} + \dots + \phi_{1,p}) G_1(S_t) + (\phi_{2,0} + \phi_{2,1} Y_{t-1}$$

$$+ \dots + \phi_{2,p}) G_2(S_t) + \dots$$

$$\dots + (\phi_{m,0} + \phi_{m,1} Y_{t-1} + \dots + \phi_{m,p}) G_m(S_t) + e_t$$

where

$$G_j(s_t) = (\xi_{jt}, c_j, \gamma_j),$$

$$j=1, \dots, m.$$