

Adaptive Expectations Model

$$(*) \quad y_i = \alpha + \beta x_i^* + \varepsilon_i$$

Adaptive exp. hypothesis: \swarrow speed of adjustment
 $(1-\lambda) = \lambda$

$$x_i^* - x_{i-1}^* = \lambda (x_i - x_{i-1}^*)$$

" Δx_i^*

$$x_i^* = \lambda x_i + (1-\lambda) x_{i-1}^* \quad (1)$$

$$\left\{ \begin{array}{l} \text{plug in} \\ \text{lagged (1)} \end{array} \right\} = \lambda (x_i + (1-\lambda) x_{i-1} + (1-\lambda)^2 x_{i-2} + \dots)$$

$$= \lambda \sum_{j=0}^{\infty} (1-\lambda)^j x_{i-j}$$

Plug in (1) to (*)

$$y_i = \alpha + \beta (\lambda x_i + (1-\lambda) x_{i-1}^*) + \varepsilon_i$$

$$= \alpha + \beta \sum_{j=0}^{\infty} w_j x_{i-j}$$

$w_j = \lambda (1-\lambda)^j$

ADL - form:

$$y_i = \alpha_0 + \beta_0 x_i + (1-\lambda) y_{i-1} + v_i$$


$$\alpha_0 = \alpha \lambda$$

$$\beta_0 = \beta \lambda$$

$$v_i = \varepsilon_i - \lambda \varepsilon_{i-1}$$

SR effect : $\beta = \beta\lambda$

LR effect :

$$\bar{y} = \alpha + \beta\lambda\bar{x} + (1-\lambda)\bar{y}$$


$$\lambda\bar{y} = \lambda\alpha + \lambda\beta\bar{x} \Rightarrow \text{LR effect}$$

$$\bar{y} = \alpha + \beta\bar{x}$$

(AFM) $\leftrightarrow x_t^*$

$$y_t = \alpha + \beta x_t^e + \varepsilon_t$$

$$\Delta x_t^e = \lambda(x_t - x_{t-1}^e) \quad 0 < \lambda \leq 1$$

↑ revision of expectation

λ - speed of adjustment

$$\begin{aligned} x_t^e &= x_{t-1}^e + \lambda x_t - \lambda x_{t-1}^e \\ &= (1-\lambda)x_{t-1}^e + \lambda x_t \end{aligned}$$

$\lambda \approx 1 \Rightarrow$ revision fast

$\lambda \approx 0 \Rightarrow$ revision slow

(P6)

$$y_t = \beta_1 + \beta_2 x_{t+1}^e + u_t \quad (1)$$

$$x_{t+1}^e - x_t^e = \lambda(x_t - x_t^e) \quad (2)$$

new indexes

$$x_t^e \rightarrow x_{t+1}^e$$

$$y_t = \beta_1 + \beta_2 \lambda x_t + \beta_2 \lambda (1-\lambda) x_{t-1} + \dots$$

$$+ \beta_2 \lambda (1-\lambda)^{s-1} x_{t-s+1} + \beta_2 (1-\lambda)^s x_{t-s+1}^e + u_t$$

(P7)

from (2) express $x_{t+1}^e \rightarrow (1)$

$$(3) \quad y_t = \beta_1 + \beta_2 (\lambda x_t + (1-\lambda) x_t^e) + u$$

$$\dots = \beta_1 + \beta_2 \lambda x_t + \beta_2 \lambda (1-\lambda) x_{t-1} + \dots$$

$$+ \beta_2 \lambda (1-\lambda)^{s-1} x_{t-s+1} + \boxed{\beta_2 (1-\lambda)^s x_{t-s+1}^e} + u_t$$

Same Using Lag operator:

$$x_{t+1}^e = \lambda x_t + (1-\lambda) x_t^e$$

$$\lambda L x_{t+1}^e = \lambda x_t$$

$$x_{t+1}^e = (\lambda L)^{-1} \lambda x_t \quad \left\{ \lambda = 1 - \tilde{\lambda} \right\}$$

$$x_{t+1}^e = (1 - \tilde{\lambda} L)^{-1} (1 - \tilde{\lambda}) x_t$$

$$= (1 - \tilde{\lambda}) \sum_{k=0}^{\infty} \tilde{\lambda}^k x_{t+k}$$

$$= \sum_{k=0}^{\infty} \beta_k x_{t-k}$$

 \Rightarrow plug in (1)

$$\beta_k = (1 - \tilde{\lambda}) \tilde{\lambda}^k = \lambda (1 - \lambda)^k$$

(P8)

Estimation using Koyck distribution

Approximation (3)

$$\begin{aligned} y_t &= \beta_1 + \beta_2 (\lambda x_t + (1-\lambda) x_t^e) + u_t \\ \dots &= \beta_1 + \beta_2 \lambda x_t + \beta_2 \lambda (1-\lambda) x_{t-1} + \dots \\ &\quad + \beta_2 \lambda (1-\lambda)^{s-1} x_{t-s+1} + u_t \end{aligned}$$

Estimation : fix $\lambda \Rightarrow$ no non-linearity
 $= x_t^{(*)}$

$$y_t = \beta_1 + \beta_2 (\lambda x_t + \dots)$$

Hence we use OLS

Calc. RSS

\Rightarrow min RSS

(P11)

Est. using Koyck transform:

(with Koyck distribution)

$$(3) \quad y_t = \beta_1 + \beta_2 \left(\lambda x_t + (1-\lambda) x_t^e \right) + u_t \\ = \beta_1 + \beta_2 \lambda \sum_{j=0}^{\infty} (1-\lambda)^j x_{t-j} + u_t$$

KT: $(3) - \lambda y_{t-1}$

$$(1-\lambda) y_{t-1} = \beta_1 + \beta_2 \lambda \sum_{j=0}^{\infty} (1-\lambda)^j x_{t-j-1} + u_{t-1}$$

$$y_t - (1-\lambda) y_{t-1} = \lambda \beta_1 + \beta_2 \lambda x_t + u_t - (1-\lambda) u_{t-1}$$

$$\Rightarrow y_t = (1-\lambda) y_{t-1} + \lambda \beta_1 + \beta_2 \lambda x_t + u_t - (1-\lambda) u_{t-1}$$

$$\Rightarrow \text{ARD}(1,0)$$

$$1 - \hat{\lambda} = 0,85$$

$$\hat{\lambda} = 0,15 \Rightarrow \text{rather slow revision}$$

(P12)

Est. using Koyck transform:

(without Koyck distribution)

$$\left\{ \begin{array}{l} y_t = \beta_1 + \beta_2 \underbrace{x_{t+1}^e}_{\cdot} + u_t \end{array} \right. \quad (*)$$

$$\left\{ \begin{array}{l} x_{t+1}^e - x_t^e = \lambda (x_t - x_t^e) \end{array} \right. \quad (**)$$

KT: lag (*)

$$y_{t-1} = \beta_1 + \beta_2 x_t^e + u_{t-1}$$

$$\beta_2 x_t^e = -\beta_1 + y_{t-1} - u_{t-1}$$

$$y_t = \beta_1 + \beta_2 \overset{\nwarrow}{x_t^e} + \beta_2 (\lambda (x_t - \overset{\nearrow}{x_t^e})) + u_t = \dots$$

$$\dots = \lambda \beta_1 + \beta_2 \lambda x_t + (1 - \lambda) y_{t-1} +$$

$$u_t - (1 - \lambda) u_{t-1}$$

(ECM)

Error Correction Model

$$(1) \quad y_t^* = \alpha + \beta x_t + \varepsilon_t$$

ECM hypothesis:

$$(2) \quad \underbrace{y_t - y_{t-1}}_{\Delta y_t} = (1 - \tilde{\gamma}) \underbrace{(y_t^* - y_{t-1}^*)}_{\Delta y_t^* - \text{change potential value}} + (1 - \tilde{\lambda}) \underbrace{(y_{t-1}^* - y_{t-1})}_{\text{previous disequilibrium value}}$$

$$0 < \gamma < 1, \quad 0 < \lambda < 1$$

$$\gamma = \lambda \Rightarrow \text{PAM}$$

Plugging (2) to (1)

$$y_t - y_{t-1} = (1 - \gamma) (\beta (x_t - x_{t-1}) + \varepsilon_t - \varepsilon_{t-1}) + (1 - \lambda) (\alpha + \beta x_{t-1} + \varepsilon_{t-1} - y_{t-1})$$

$$y_t = \alpha_0 + \beta_0 x_t + \beta_1 x_{t-1} + \lambda y_{t-1} + N_t$$

$$\alpha_0 = (1 - \lambda)\alpha \quad \beta_0 = (1 - \gamma)\beta \quad \beta_1 = (\gamma - \lambda)\beta$$

$$N_t = (1 - \gamma)\varepsilon_t + (\gamma - \lambda)\varepsilon_{t-1}$$

$$SR: (1 - \gamma) \beta$$

$$LR: \frac{\beta^0 + \beta^1}{1 - \gamma} = \beta \frac{\beta - \gamma \beta + \gamma \beta - \gamma \beta}{1 - \gamma} = \beta$$