

Notation

y_t	value of series in period t
$Y_t = \{y_1, y_2, \dots, y_t\}$	sample of size t
ε_t	innovation in period t
Δ	difference operator
L	lag operator
$\phi(L), \theta(L), \psi(L)$	polynomial lag functions
ϕ	damping factor
σ^2	variance
τ_t	level of trend in period t
c_t	level of transitory component in period t

Innovations State Space Model

Measurement equation

$$y_t = z'x_{t-1} + \varepsilon_t$$

Transition equations

$$x_t = Fx_{t-1} + g\varepsilon_t$$

Stochastic Assumptions

$$\varepsilon_t \sim \text{NID}(0, \sigma^2)$$

The vectors z , g and the matrix F potentially depend on an unknown vector of parameters θ .

Exponential Smoothing

The one-step ahead prediction errors needed for likelihood evaluation is generated with the following algorithm (originally proposed by Box and Jenkins)

Given trial values of the seed state x_0 and the parameter vectors θ to start the algorithm, repeat the following steps in each period t :

1. Prediction step

$$\hat{y}_t = z'x_{t-1}$$

2. Error step

$$e_t = y_t - \hat{y}_t$$

3. Revision step

$$x_t = Fx_{t-1} + ge_t$$

This is used as a simple alternative to the Kalman filter, which performs essentially the same function in traditional approaches.

References

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