
Dynamic Nelson Siegel Model

Daeyoung Lim

1 Model

- Initial model

$$y_t(\tau) = L_t + \frac{1 - e^{-\tau\lambda}}{\tau\lambda} S_t + \left(\frac{1 - e^{-\tau\lambda}}{\tau\lambda} - e^{-\tau\lambda} \right) C_t + \eta_t(\tau) \quad (1.1)$$

- State space model

$$\begin{pmatrix} y_t(\tau_1) \\ y_t(\tau_2) \\ \vdots \\ y_t(\tau_N) \end{pmatrix} = \begin{pmatrix} 1 & (1 - e^{-\tau_1\lambda})(\lambda\tau_1)^{-1} & (1 - e^{-\tau_1\lambda})(\lambda\tau_1)^{-1} - e^{\lambda\tau_1} \\ 1 & (1 - e^{-\tau_2\lambda})(\lambda\tau_2)^{-1} & (1 - e^{-\tau_2\lambda})(\lambda\tau_2)^{-1} - e^{\lambda\tau_2} \\ \vdots & \vdots & \vdots \\ 1 & (1 - e^{-\tau_N\lambda})(\lambda\tau_N)^{-1} & (1 - e^{-\tau_N\lambda})(\lambda\tau_N)^{-1} - e^{\lambda\tau_N} \end{pmatrix} \begin{pmatrix} L_t \\ S_t \\ C_t \end{pmatrix} + \begin{pmatrix} \eta_t(\tau_1) \\ \eta_t(\tau_2) \\ \vdots \\ \eta_t(\tau_N) \end{pmatrix} \quad (1.2)$$

The evolution of the factors (state equation):

$$\begin{pmatrix} L_t \\ S_t \\ C_t \end{pmatrix} = \begin{pmatrix} \mu_{1,t} \\ \mu_{2,t} \\ \mu_{3,t} \end{pmatrix} + \begin{pmatrix} \phi_{11} & \phi_{12} & \phi_{13} \\ \phi_{21} & \phi_{22} & \phi_{23} \\ \phi_{31} & \phi_{32} & \phi_{33} \end{pmatrix} \begin{pmatrix} L_{t-1} \\ S_{t-1} \\ C_{t-1} \end{pmatrix} + \begin{pmatrix} \epsilon_{1,t} \\ \epsilon_{2,t} \\ \epsilon_{3,t} \end{pmatrix} \quad (1.3)$$

- Compact notation (Dynamic linear model)

$$\begin{aligned} y_t &= \Gamma \boldsymbol{\beta}_t + \eta_t, \quad \eta_t \sim N(\mathbf{0}, \Sigma) \\ \boldsymbol{\beta}_t &= \mu + \mathbf{G} \boldsymbol{\beta}_{t-1} + \epsilon_t, \quad \epsilon_t \sim N(\mathbf{0}, \Omega) \end{aligned} \tag{1.4}$$

2 Forward Filtering Backward Sampling (FFBS) algorithm

2.1 Kalman Filter

Algorithm 1 Kalman Filter

```

1: procedure KALMANFILTER
2:    $f_{0|0} \leftarrow (I_3 - \mathbf{G})^{-1} \mu \quad (\because E(\boldsymbol{\beta}) = \mu + \mathbf{G} E(\boldsymbol{\beta}))$ 
3:    $\text{vec}(P_{0|0}) \leftarrow (I_9 - \mathbf{G} \otimes \mathbf{G})^{-1} \text{vec}(\Omega) \quad (\because \text{Var}(\boldsymbol{\beta}) = \mathbf{G} \text{Var}(\boldsymbol{\beta}) \mathbf{G}^T + \Omega)$ 
4:   for  $t = 1 : T$  do
5:      $f_{t|t-1} \leftarrow \mu + \mathbf{G} f_{t-1|t-1}$ 
6:      $P_{t|t-1} \leftarrow \mathbf{G} P_{t-1|t-1} \mathbf{G}^T + \Omega$ 
7:      $K_t \leftarrow P_{t|t-1} \Gamma^T (\Gamma P_{t|t-1} \Gamma^T + \Sigma)^{-1}$ 
8:      $\boldsymbol{\beta}_{t|t} \leftarrow f_{t|t-1} + K_t (y_t - \Gamma f_{t|t-1})$ 
9:      $P_{t|t} \leftarrow (I_3 - K_t \Gamma) P_{t|t-1}$ 

```

Note: $\text{vec}(ABC) = (C^T \otimes A) \text{vec}(B)$

2.2 Backward Sampling

Algorithm 2 Backward Sampling

```

1: procedure BACKWARD SAMPLING
2:    $\boldsymbol{\beta}_T \sim N(f_{T|T}, P_{T|T})$ 
3:   for  $t = (T - 1) : 1$  do
4:      $f_{t+1|t} \leftarrow \mu + \mathbf{G} f_{t|t}$ 
5:      $P_{t+1|t} = \mathbf{G} P_{t|t} \mathbf{G}^T + \Omega$ 
6:      $f_{t|t, \beta_{t+1}} \leftarrow f_{t|t} + P_{t|t} \mathbf{G}^T P_{t+1|t}^{-1} (\boldsymbol{\beta}_{t+1} - f_{t+1|t})$ 
7:      $P_{t|t, \beta_{t+1}} \leftarrow P_{t|t} - P_{t|t} \mathbf{G}^T P_{t+1|t}^{-1} \mathbf{G} P_{t|t}$ 
8:      $\boldsymbol{\beta}_t \sim N(f_{t|t, \beta_{t+1}}, P_{t|t, \beta_{t+1}})$ 

```

$$\begin{pmatrix} \boldsymbol{\beta}_t \\ \boldsymbol{\beta}_{t+1} \end{pmatrix} | Y^t \sim N \left(\begin{pmatrix} f_{t|t} \\ f_{t+1|t} \end{pmatrix}, \begin{pmatrix} P_{t|t} & P_{t|t} \mathbf{G}^T \\ \mathbf{G} P_{t|t} & P_{t+1|t} \end{pmatrix} \right) \quad (2.1)$$