

Bayesian inference in NDLM: Known variances

$$y_t = \mathbf{F}_t' \boldsymbol{\theta}_t + \nu_t, \quad \nu_t \sim N(0, v_t), \quad (1)$$

$$\boldsymbol{\theta}_t = \mathbf{G}_t \boldsymbol{\theta}_{t-1} + \boldsymbol{\omega}_t, \quad \boldsymbol{\omega}_t \sim N(0, \mathbf{W}_t), \quad (2)$$

with $\mathbf{F}_t, \mathbf{G}_t, v_t$ and \mathbf{W}_t known. We also assume a prior distribution of the form $(\boldsymbol{\theta}_0 | \mathcal{D}_0) \sim N(\mathbf{m}_0, \mathbf{C}_0)$, with $\mathbf{m}_0, \mathbf{C}_0$ known.

Smoothing

For $t < T$, we have that

$$(\boldsymbol{\theta}_t | \mathcal{D}_T) \sim N(\mathbf{a}_T(t - T), \mathbf{R}_T(t - T)),$$

where

$$\begin{aligned} \mathbf{a}_T(t - T) &= \mathbf{m}_t - \mathbf{B}_t[\mathbf{a}_{t+1} - \mathbf{a}_T(t - T + 1)], \\ \mathbf{R}_T(t - T) &= \mathbf{C}_t - \mathbf{B}_t[\mathbf{R}_{t+1} - \mathbf{R}_T(t - T + 1)]\mathbf{B}_t', \end{aligned}$$

for $t = (T - 1), (T - 2), \dots, 0$, with $\mathbf{B}_t = \mathbf{C}_t \mathbf{G}_{t+1}' \mathbf{R}_{t+1}^{-1}$, and $\mathbf{a}_T(0) = \mathbf{m}_T$, $\mathbf{R}_T(0) = \mathbf{C}_T$. Here $\mathbf{a}_t, \mathbf{m}_t, \mathbf{R}_t$, and \mathbf{C}_t are obtained using the filtering equations as explained before.

Forecasting

For $h \geq 0$ it is possible to show that

$$(\boldsymbol{\theta}_{t+h}|\mathcal{D}_t) \sim N(\boldsymbol{a}_t(h), \boldsymbol{R}_t(h)),$$

$$(y_{t+h}|\mathcal{D}_t) \sim N(f_t(h), q_t(h)),$$

with

$$\boldsymbol{a}_t(h) = \boldsymbol{G}_{t+h}\boldsymbol{a}_t(h-1), \quad \boldsymbol{R}_t(h) = \boldsymbol{G}_{t+h}\boldsymbol{R}_t(h-1)\boldsymbol{G}_{t+h}' + \boldsymbol{W}_{t+h},$$

$$f_t(h) = \boldsymbol{F}_{t+h}'\boldsymbol{a}_t(h), \quad q_t(h) = \boldsymbol{F}_{t+h}'\boldsymbol{R}_t(h)\boldsymbol{F}_{t+h} + v_{t+h},$$

$$\boldsymbol{a}_t(0) = \boldsymbol{m}_t, \boldsymbol{R}_t(0) = \boldsymbol{C}_t.$$