Example Used in Exemplifying the Marginalized (Rao-Blackwellized) Particle Filter

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In order to illustrate the marginalized (Rao-Blackwellized) particle filter, we provide the MATLAB code for solving the following examples,

$$x_{t+1}^{n} = \arctan x_{t}^{n} + \begin{pmatrix} 1 & 0 & 0 \end{pmatrix} x_{t}^{l} + w_{t}^{n},$$
(1a)

$$x_{t+1}^{l} = \begin{pmatrix} 1 & 0.3 & 0\\ 0 & 0.92 & -0.3\\ 0 & 0.3 & 0.92 \end{pmatrix} x_{t}^{l} + w_{t}^{l},$$
(1b)

$$y_t = \begin{pmatrix} 0.1(x_t^{n})^2 \operatorname{sgn}(x_t^{n}) \\ 0 \end{pmatrix} + \begin{pmatrix} 0 & 0 & 0 \\ 1 & -1 & 1 \end{pmatrix} x_t^{l} + e_t,$$
(1c)

where

$$w_t = \begin{pmatrix} w_t^{\rm n} \\ w_t^{\rm l} \end{pmatrix} \sim \mathcal{N}\left(0, \ 0.01I_{4\times 4}\right),\tag{1d}$$

$$e_t \sim \mathcal{N}(0, 0.1I_{2\times 2}),$$
 (1e)
 $x_0^n = \mathcal{N}(0, 1)$ (1f)

$$x_0^{n} = \mathcal{N}(0, 1), \tag{1f}$$

$$x_0^1 = \mathcal{N}(0_{3 \times 1}, 0_{3 \times 3}),$$
 (1g)

(1h)

Looking at the notation used in Model 3 in [2], that is the model specified in

(18) and (19) of [2], we have,

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$$f_t^{\mathbf{n}}(x_t^{\mathbf{n}}) = \arctan x_t^{\mathbf{n}},\tag{2a}$$

$$A_t^{n}(x_t^{n}) = \begin{pmatrix} 1 & 0 & 0 \end{pmatrix},$$
(2b)

$$G_t^{n}(x_t^{n}) = 1,$$
(2c)

$$f_t^{\mathbf{l}}(x_t^{\mathbf{n}}) = \begin{pmatrix} 0 & 0 & 0 \end{pmatrix}^T, \qquad (2d)$$

$$A_t^{\rm l}(x_t^{\rm n}) = \begin{pmatrix} 1 & 0.3 & 0\\ 0 & 0.92 & -0.3\\ 0 & 0.3 & 0.92 \end{pmatrix},$$
(2e)

$$G_t^{\mathbf{l}}(x_t^{\mathbf{n}}) = I_{3\times 3},\tag{2f}$$

$$h_t(x_t^{\mathbf{n}}) = \begin{pmatrix} 0.1(x_t^{\mathbf{n}})^2 \operatorname{sgn}(x_t^{\mathbf{n}}) \\ 0 \end{pmatrix}, \qquad (2g)$$

$$C_t(x_t^{n}) = \begin{pmatrix} 0 & 0 & 0\\ 1 & -1 & 1 \end{pmatrix},$$
 (2h)

$$Q_t = 0.01 I_{4 \times 4},$$
 (2i)

$$R_t = 0.1I_{2\times 2},\tag{2J}$$

$$\bar{x}_0 = \begin{pmatrix} 0 & 0 & 0 \end{pmatrix}^T,$$
(2k)
 $\bar{P}_0 = 0_{3\times 3}.$
(2l)

$$_{0} = 0_{3 \times 3}. \tag{21}$$

It is worth noting that we also used this example in [1] in order to illustrate how to perform maximum likelihood identification in mixed linear/nonlinear models.

References

- [1] F. Lindsten and T. B. Schön. Maximum likelihood identification in mixed linear/nonlinear state-space models. In Proceedings of the 49th IEEE Conference on Decision and Control (CDC), Atlanta, USA, December 2010.
- [2] T. Schön, F. Gustafsson, and P.-J. Nordlund. Marginalized particle filters for mixed linear/nonlinear state-space models. IEEE Transactions on Signal Processing, 53(7):2279–2289, July 2005.