

Fixed Interval Smoothing for Nonlinear Continuous Time Systems*

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An equation is derived for the probability density of the state of a nonlinear dynamical system, conditioned on measurements over a fixed interval. In deriving the equation, the conditional Fokker Planck equation yielding the probability density of the filtering problem is used several times in a novel way.

1. INTRODUCTION

We consider the nonlinear system

$$dx = f(x_t, t) dt + G(x_t, t) dv, \quad t \geq 0, \quad (1)$$

with measurements

$$dz = h(x_t, t) dt + dw, \quad t \geq 0. \quad (2)$$

Here, dv/dt and dw/dt are independent, zero mean, gaussian white noise processes, with covariances $I\delta(t - \tau)$ and $R(t)\delta(t - \tau)$, respectively. The matrix $R(t)$ is positive definite for all t . An a priori distribution for the density of x_0 is assumed known, and it is assumed that f , G , h and R all have sufficient smoothness properties to guarantee the usual existence and uniqueness requirements on solutions of (1) and (2), together with such other quantities as will be introduced. In particular, we assume that the conditional density $p(x_t | Z_{(0,t)})$ of x_t , given the measurements z_t over $[0, t]$, exists and satisfies the conditional Fokker-Planck equation; see, e.g., Jazwinski (1970).

In this paper, we aim to give a differential equation for the probability density $p(x_t | Z_{(0,T)})$, with T fixed. This is the probability density associated with the fixed-interval smoothing problem.

* Work supported by the Australian Research Grants Committee.

Earlier results on fixed-interval smoothing may be found in Striebel (1965) (which are less complete than our results, and a good deal more formal), and Leondes *et al.* (1970). We derive the same basic result as Leondes *et al.* (1970), with, however, much greater economy. Part of this economy is achieved through use of the conditional Fokker-Planck equation, for the conditional filter density $p(x_t | Z_{t_0,t})$. Fixed point smoothing is discussed in Lee (1971).

In Section 2, we review the conditional filtering equation, and use it to prove two helpful lemmas. In Section 3, the main result is proved, and we also indicate an equation for the evolution of the mean of an arbitrary function of x_t . Section 4 contains some concluding remarks.

2. THE CONDITIONAL FOKKER-PLANCK EQUATION

The conditional Fokker-Planck equation is derived in Jazwinski (1970); for an original reference; see, e.g., Kushner (1962). Let us define the operator $\mathcal{L}(\cdot)$ by

$$\mathcal{L}\{\phi(x_t, t)\} = -\sum_i \frac{\partial}{\partial x_i} (\phi f) + \frac{1}{2} \sum_{i,j} \frac{\partial^2}{\partial x_i \partial x_j} [\phi (GG')_{ij}] \quad (3)$$

with the superscript prime denoting matrix transposition. Also, henceforth let us write

$$p_f(v) = p_{x_t}(v | Z_{t_0,t}) \quad (4)$$

and

$$p_s(v) = p_{x_t}(v | Z_{t_0,t_1}). \quad (5)$$

Quantities such as $p_{x_{t+\Delta t}}(\eta | Z_{t_0,t+\Delta t})$ will not be abbreviated. Finally, for arbitrary $\phi(x_t, t)$, let

$$\bar{\phi}_f(t) = E[\phi | p_f]$$

with $\bar{\phi}_s$ defined obviously.

With this notation, the conditional Fokker-Planck equation then becomes

$$dp_f = \mathcal{L}(p_f) dt + (h_t - \bar{h}_f)' R^{-1} (dx - \bar{h}_f dt) p_f, \quad (7)$$

where dp_f and dx have obvious interpretations.

For our derivation of the smoothing equation, we shall require knowledge

of the density $p_{x_{t+dt}}(\eta | x_t = \nu, dx)$. As shown in the proof of the following lemma, this density follows from (7):

LEMMA 1. *With quantities as defined earlier,*

$$p_{x_{t+dt}}(\eta | x_t = \nu, dx) = \delta(\eta - \nu) + \mathcal{L}_\eta(\delta(\eta - \nu)) dt \quad (8)$$

Proof. Let us apply (7), taking as the initial time t and initial density $p_{x_t}(\eta) = \delta(\eta - \nu)$. Then (7) gives immediately

$$\begin{aligned} p_{x_{t+dt}}(\eta | p_{x_t}(\eta) = \delta(\eta - \nu), dx) - p_{x_t}(\eta) \\ = \mathcal{L}_\eta(p_{x_t}(\eta)) dt + [h(\eta) - E(h(\eta))] R^{-1}[dx - E(h(\eta))] p_{x_t}(\eta) \end{aligned}$$

or

$$\begin{aligned} p_{x_{t+dt}}(\eta | x_t = \nu, dx) - \delta(\eta - \nu) \\ = \mathcal{L}_\eta[\delta(\eta - \nu)] dt + [h(\eta) - h(\nu)] R^{-1}[dx - h(\nu)] \delta(\eta - \nu) \\ = \mathcal{L}_\eta[\delta(\eta - \nu)] dt. \end{aligned}$$

The second lemma is concerned with the density $p_{x_t}(\nu | Z_{[0,t]}, dx)$, which is a smoothed density, because of the appearance of dx in the conditioning variables.

LEMMA 2. *With quantities as defined earlier,*

$$p_{x_t}(\nu | Z_{[0,t]}, dx) = [1 + (h - h_f) R^{-1}(dx - h_f dt)] p_{x_t}(\nu | Z_{[0,t]}) \quad (9)$$

Proof. Let t be a fixed variable and τ a temporary running variable and suppose temporarily that for $\tau \geq t$, (1) is replaced by

$$dx_\tau = 0.$$

Then $x_{\tau+dt} = x_\tau$, and so $p_{x_\tau}(\nu | Z_{[0,\tau]}, dx)$ will be the same as $p_{x_{\tau+dt}}(\nu | Z_{[0,\tau]}, dx)$ which is a filtering density. The conditional Fokker-Planck equation then yields (recall that now $f(x_\tau, t)$ and $G(x_\tau, t)$ are zero)

$$p_{x_{\tau+dt}}(\nu | Z_{[0,\tau]}, dx) - p_{x_\tau}(\nu | Z_{[0,\tau]}) = (h - h_f) R^{-1}(dx - h_f dt) p_{x_\tau}(\nu | Z_{[0,\tau]})$$

Equation (9) is immediate.

3. DERIVATION OF MAIN RESULTS

The first result which we prove is stated in the following theorem:

THEOREM. *With quantities as defined previously,*

$$dp_s = \left[\mathcal{L}(p_s) \frac{p_s}{p_s} - p_s \mathcal{L}^0 \left(\frac{p_s}{p_s} \right) \right] dt, \tag{10}$$

where the operator \mathcal{L}^0 is the formal adjoint of \mathcal{L} , i.e.,

$$\mathcal{L}^0[\phi(x_i, t)] = \sum_i f_i \frac{\partial \phi}{\partial x_i} + \frac{1}{2} \sum_{i,j} (GG^T)_{ij} \frac{\partial^2 \phi}{\partial x_i \partial x_j}. \tag{11}$$

We shall prove this theorem in several stages.

(1) *A Chapman-Kolmogorov equation.* It is clear that

$$p_{x_t}(v | Z_{[0,t]}) = \int p_{x_{t+\Delta t}}(v | x_{t+\Delta t} = \eta, Z_{[0,t]}) p_{x_{t+\Delta t}}(\eta | Z_{[0,t]}) d\eta,$$

or, using the Markov nature of the x process,

$$p_x(v) = \int \left[p_{x_t}(v | x_{t+\Delta t} = \eta, Z_{[0,t]}, dx) \left(p_x(\eta) + \frac{\partial p_x(\eta)}{\partial t} dt \right) \right] d\eta. \tag{12}$$

(2) *Evaluation of the integrand in (12).* An expression for $p_{x_t}(v | x_{t+\Delta t} = \eta, Z_{[0,t]}, dx)$ in terms of more readily manageable quantities follows straightforwardly via Bayes' rule. Evidently,

$$\begin{aligned} p_{x_t}(v | x_{t+\Delta t} = \eta, Z_{[0,t]}, dx) &= \frac{p_{x_{t+\Delta t}-x_t}(\eta, v | Z_{[0,t]}, dx)}{p_{x_{t+\Delta t}}(\eta | Z_{[0,t+\Delta t]})} \\ &= \frac{p_{x_{t+\Delta t}}(\eta | x_t = v, Z_{[0,t]}, dx) p_{x_t}(v | Z_{[0,t]}, dx)}{p_{x_{t+\Delta t}}(\eta | Z_{[0,t+\Delta t]})} \\ &= \frac{p_{x_{t+\Delta t}}(\eta | x_t = v, dx) p_{x_t}(v | Z_{[0,t]}, dx)}{p_{x_{t+\Delta t}}(\eta | Z_{[0,t+\Delta t]})}. \end{aligned} \tag{13}$$

The third equality follows on using the Markovian nature of x . Notice that the two densities in the numerator are replaceable by certain expressions

stated in the lemmas of the last section. Notice too that the denominator is expressible in terms of $p_f(\eta)$, using the conditional Fokker-Planck equation

(3) *Evaluation of the right-hand side of (12).* The manipulations, though intricate, are straightforward.

We write for the various terms in (13) their expansions as given from Section 2, and insert (13) into (12). With suppression of the argument t , we obtain

$$\begin{aligned}
 p_s(v) &= \int \frac{\{\delta(\eta - v) + \mathcal{L}_0^s(\delta(\eta - v))\} dt [1 + (h(v) - h_f) R^{-1}(dx - h_f dt)] \int \frac{\times p_s(v) [p_s(\eta) + (\partial p_s(\eta) / \partial t) dt] d\eta}{[1 + \mathcal{L}_0^s(\cdot) dt + (h(\eta) - h_f) R^{-1}(dx - h_f dt)] p_f(\eta)} \\
 &= \int \frac{\{\delta(\eta - v) [1 + (h(v) - h_f) R^{-1}(dx - h_f dt)] \times p_s(v) [p_s(\eta) + (\partial p_s(\eta) / \partial t) dt] d\eta}{[1 + \mathcal{L}_0^s(\cdot) dt + (h(\eta) - h_f) R^{-1}(dx - h_f dt)] p_f(\eta)} \\
 &\quad + p_s(v) \int \frac{\mathcal{L}_0^s(\delta(\eta - v)) p_s(\eta)}{p_f(\eta)} d\eta dt + o(dt). \tag{14}
 \end{aligned}$$

To evaluate the first integral, replace the term

$$[1 + (h(v) - h_f) R^{-1}(dx - h_f dt)] p_s(v)$$

in the numerator of the integrand by the difference of the two terms $[1 + \mathcal{L}_0^s(\cdot) dt + (h(v) - h_f) R^{-1}(dx - h_f dt)] p_s(v)$ and $\mathcal{L}_0^s(p_s(v)) dt$. The first integral then becomes the difference of two integrals, readily computable to $o(dt)$, and is

$$p_s(v) + \frac{\partial p_s(v)}{\partial t} dt - \frac{p_s(v)}{p_f(v)} \mathcal{L}_0^s(p_s(v)) dt.$$

Further, the definition of \mathcal{L}_0^s leads to ready evaluation of the second integral:

$$\begin{aligned}
 p_s(v) \int \frac{\mathcal{L}_0^s(\delta(\eta - v)) p_s(\eta)}{p_f(\eta)} d\eta dt \\
 = p_s(v) \int \delta(\eta - v) \mathcal{L}_0^s \left[\frac{p_s(\eta)}{p_f(\eta)} \right] d\eta dt = p_s(v) \mathcal{L}_0^s \left[\frac{p_s(v)}{p_f(v)} \right] dt.
 \end{aligned}$$

Accordingly, (14) becomes

$$p_s(v) = p_s(v) + \frac{\partial p_s(v)}{\partial t} dt - \frac{p_s(v)}{p_f(v)} \mathcal{L}_0^s(p_s(v)) + p_s(v) \mathcal{L}_0^s \left[\frac{p_s(v)}{p_f(v)} \right].$$

$$dp_t = \left[\frac{p_t}{p_t} \mathcal{L}(p_t) - p_t \mathcal{L}^a \left(\frac{p_t}{p_t} \right) \right] dt. \quad (10)$$

The boundary condition for this equation is obtained by observing that the smoothing density associated with x_T is the same as the filtered density. In other words, one must solve the conditional Eq. (7) forwards in time, using the measurements and with boundary condition the prescribed density of x_0 , and then solve (10) backwards in time, using stored values of the filtered density. Solution of (10) does not however require re-use of the measurements.

From Eq. (10), it is straightforward to obtain an exact equation for the evolution of the conditional mean \bar{x}_t of an arbitrary function $\phi(x)$. Multiply (10) by $\phi(x)$ and integrate over the whole of the x_t -space. The result is

$$\begin{aligned} d\bar{x}_t &= \left[\int \frac{\phi(x)p_t(x)}{p_t(x)} \mathcal{L}_2(p_t(x)) dx - \int p_t(x) \phi(x) \mathcal{L}_2^a \left(\frac{p_t(x)}{p_t(x)} \right) dx \right] dt \\ &= \left[\overline{\frac{\phi}{p_t} \mathcal{L}(p_t)} - \phi \mathcal{L}^a \left(\frac{p_t}{p_t} \right) \right] dt \end{aligned} \quad (15)$$

Alternative forms follow by noticing that

$$\overline{\frac{\phi}{p_t} \mathcal{L}(p_t)} = \int \mathcal{L}^a \left(\frac{\phi p_t}{p_t} \right) p_t dx = \mathcal{L}^a \left(\frac{\phi p_t}{p_t} \right), \quad (16)$$

and

$$\phi \mathcal{L}^a \left(\frac{p_t}{p_t} \right) = \int \mathcal{L}(p_t \phi) \frac{p_t}{p_t} dx = \overline{\mathcal{L}(p_t \phi)}. \quad (17)$$

4. CONCLUDING REMARKS

As with nonlinear filtering, the exact equations of nonlinear smoothing are impractical to use. Storage of $p_t(x_t)$ for all x_t , together with its derivatives (which are required in (10)) is a task which would be well nigh impossible. In Leondes *et al.* (1970), approximate equations are derived for the mean and covariance of x , under the assumptions that (1) $f(x_t, t)$ can be represented by a Taylor series expansion around $E(x_t | Z_{1:t,T})$ up to the quadratic term, (2) the filtered density is Gaussian, with known mean and covariance and (3) third central moments associated with the smoothed estimate are

negligible. These equations become exact for the standard linear-gaussian problem. A problem which suggests itself is application of the approximation technique of Kushner (1967), to the smoothing problem; the first two assumptions noted above are common in filtering problems and, as pointed out in Kushner (1967), may be quite inadequate.

RECEIVED: June 21, 1971

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