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REMARKS ON FINITE DIMENSIONAL NONLINEAR ESTIMATION

by

R.W. BROCKETT

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I.- INTRODUCTION.

In this paper we consider the problem of estimating the values taken on by a random process y(t) of the form

(1)
$$dx = f(x) dt + g(x) dw ; dy(t) = h(x) dt + dv$$

where w and v are independent Wiener processes and the differential equation is to be interpreted as an Ito equation in \mathbb{R}^n . In particular, we investigate the existence of "recursive estimators", i.e. differential equations of the form

$$dz = a(z) dt + b(z) dy$$

$$\hat{\mathbf{y}} = \mathbf{c}(\mathbf{z})$$

where \hat{y} may be for example, the conditional mean of y and z is finite dimensional. The main ideas involve the conditional density equation which, in unnormalized, Ito form is

(4)
$$d\hat{\rho}(t, x) = L_{\rho} \hat{\rho}(t, x) dt + L_{\rho} \hat{\rho}(t, x) dy$$

where $\hat{\rho}$ is, apart from a scale factor, the conditional density for x, given y(s) for $0 \le s \le t$. Of course conditional expectations, etc. can be expressed in terms of $\hat{\rho}$ as

(5)
$$\hat{\mathbf{y}}(t) = \int \boldsymbol{\psi}(\mathbf{x}) \, \boldsymbol{\beta}(t, \mathbf{x}) \, \mathrm{d}\mathbf{x} \left[\int \boldsymbol{\beta} \, \mathrm{d}\mathbf{x} \right]^{-1}$$

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Earlier work on the relationship between Lie-theoretic ideas and estimation theory appears in [1]. The main points of this paper involve the exploration of the following two ideas, both of which seem to be novel.

(a) If \mathcal{L} is the Lie algebra generated by the operators $L_0 - \frac{1}{2}L_1^2$ and L_1 and \mathfrak{F} is the Lie algebra of vector fields generated by $a - \frac{1}{2}b'b$, and bthen under appropriate hypothesis \mathfrak{F} will be a homomorphic image of \mathfrak{L} .

Conversely any homomorphism of \mathfrak{L} onto a Lie algebra of complete vector fields on a finite dimensional manifold permits one to obtain some information about the conditional density by propagating the solution of a finite dimensional set of equations.

(b) Under appropriate hypothesis the input-output map defined by (4)-(5) is characterized by a Volterra series. This Volterra series may or may not have kernels which are separable in the sense of [2]. A necessary condition for the existance of a finite dimensional nonlinear estimator is that the kernels be separable.

Because of space limitations we can only sketch the basic ideas in this paper.

II. - GENERALITIES.

Consider the following notation. If $f: \mathbb{R}^n \to \mathbb{R}^n$ then we associate with f a vector field on \mathbb{R}^n according to

$$L_{f} = \sum_{i=1}^{n} f_{i} \frac{\delta}{\delta x_{i}}$$

Write for the adjoint

$$L_{f}^{\bigstar} = \sum_{i=1}^{n} \frac{\delta}{\delta x_{i}} f_{i}.$$

We can express equation (4) in this notation as

(4)
$$d\rho(t, x) = L_{\hat{f}}^* dt + \frac{1}{2} (L_g^*)^2 \hat{\rho} dy$$

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where

$$\hat{\mathbf{f}} = \mathbf{f} - \frac{1}{2} \left(\frac{\partial \mathbf{g}}{\partial \mathbf{x}} \right) \mathbf{g}$$
.

Given an initial value for β the pair of equations (4)-(5) then define what is called in control theory, an input-output map. If for a given initial value of z the pair (2)-(3) generate the same input-output map then we can deduce certain relationships concerning a, b, L_0 and L_1 . In order to facilitate these comparisons, which, for the most part involve Lie algebraic constructions, it is more natural to convert the two differential equations (2) and (4) to Fisk-Strato-novich form. In this way we can avoid the use of the somewhat unintuitive Ito calculus. [See, e.g.[3]). As is well known, in Fisk-Stratonovich form equations (2) and (4) are (d distinguishes the Fisk-Stratonovich differentials from Ito differentials)

(2')
$$\mathbf{dx} = (\mathbf{a} - \frac{1}{2} \frac{\partial \mathbf{b}}{\partial \mathbf{x}} \mathbf{b}) \mathbf{dt} + \mathbf{b} \mathbf{dy}$$

(4')
$$\frac{d}{d\rho} = (L_{f}^{*} - \frac{1}{2}(L_{g}^{*})^{2} - \frac{1}{2}(h(x))^{2}\rho dt + h(x)\rho dy$$

Appending (3) to (2') and (5) to (4') we obtain, for each assignment of initial data, input-output systems.

Let us suppose now that the vectors a and b entering the differential equation (2') are real analytic maps of \mathbb{R}^n into \mathbb{R}^n . Also, suppose that the pair (2')-(3) defines a <u>minimal system</u> in the sense that the Lie algebra of vector fields generated by $a - \frac{1}{2} b'b$, and b acts transitively at each point in \mathbb{R}^n and that no two distinct initial states for x give rise to the same response for all smooth inputs. These assumptions have the effect of insuring that there is no redundancy in the pair (2')-(3).

The following observation is now appropriate. (cf.[4]-[5]).

<u>Remark</u>: Suppose that there exists for $L_0 - \frac{1}{2}L_1^2$ and L_1 a common set of analytic vectors \mathcal{B} . If for some choice of initial condition z_0 and $\rho(0) \in \mathcal{B}$ the pair (2)-(3) and the pair (4')-(5) generate the same input output map for all smooth inputs with (2)-(3) analytic and minimal then the map

$$\Phi : L_{0} - \frac{1}{2} \quad L_{1}^{2} \rightarrow a - \frac{1}{2} b' b$$

$$\Phi$$
: $L_1 \rightarrow b$

extends to a Lie algebra homomorphism from the Lie algebra of operators on \mathcal{P} generated by $L_0 - \frac{1}{2}L_1^2$ and L_1 to the Lie algebra of vector fields generated by $a - \frac{1}{2}b'b$ and b.

Now of course relevance of this remark stems from the fact that under suitable assumptions it has been shown (see, e.g.[6],[7],[8]) the behaviour of a stochastic differential equation with white noise inputs can be deduced from its behaviour on smooth functions.

Reference [1] discusses a context in which this program is particularly easy to carry out in detail. It is the case where the stochastic process x takes on only a finite number of values and the operators L_0 and L_1 act on finite dimensional spaces. In this case results available on the representation of Lie groups together with the above remark provide considerable guidance about the design of finite dimensional nonlinear filters.

III. - THE LINEAR PROBLEM.

We now illustrate these ideas in one nontrivial case. Even though the example is a linear problem our techniques give important new information about the role of "all pass" factors in simplifying the Kalman-Bucy filter.

The unnormalized conditional density equation associated with the estimation problem described by n simultaneous Ito equations

$$d\mathbf{x}_{i} = \Sigma \mathbf{a}_{ij} \mathbf{x}_{j} dt + \mathbf{b}_{i} dw ; dy = \mathbf{c}_{i} \mathbf{x}_{i} dt + d\mathbf{v}$$

is given in Ito form by

(*)
$$d\rho = \widetilde{L}_{0}\rho dt + L_{1}\rho dy$$

where

$$\widetilde{L}_{o} = \frac{\delta}{\delta x_{i}} a_{ij} x_{j}^{+} \frac{1}{2} \Sigma \frac{\delta^{2}}{\delta x_{i} \delta x_{j}} b_{i} b_{j}$$

and

$$L_1 = \Sigma c_i x_i$$
.

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Converting (*) into Fisk-Stratonovich form we get

(**)
$$d\rho = L_0 \rho dt + L_1 \rho dy$$

where $L_0 = is \tilde{L} - \frac{1}{2}L_1^2$.

The main point of [1] is the observation that the structure of the Lie algebra generated by L_0 and L_1 is of overriding importance in understanding the estimation problem. (See also the work of Mitter [9]).

<u>Lemma</u>: <u>The Lie algebra generated by</u> L_0 <u>and</u> L_1 <u>has a basis</u> $L_0, L_1, \ldots, L_{2n}, L_{2n+1}$ <u>and commutation relations</u>

$$L_{i} = ad_{L_{0}}^{i-1}(L_{1}) \qquad i = 1, 2, ..., 2n$$
$$[L_{i}, L_{j}] = \beta_{ij} L_{2n+1} \qquad i, j = 1, 2, ..., 2n$$
$$[L_{2n+1}, L_{i}] = 0 \qquad i = 0, 1, ..., 2n+1$$

provided that $g(s) = c(Is-A)^{-1}b$ is of Mc Millan degree n and g(s)g(-s) is of Mc Millan degree 2n.

<u>Remark</u>: One easily verifies that if one regards the Lie algebra as a complex Lie algebra then the derived algebra in this case is isomorphic to the Heisenberg algebra \mathbb{H}^n .

<u>Remark</u>: The conditions on g(s) corresponds to the requirement that the filter be minimal and that g(s) have no factors of the form (s-a)/(s+a). The contribution of such "all pass" factors to degeneracy in filtering problems is considerably clarified by this lemma.

 as defined above certainly contains $\{L_0, L_l\}_{LA}$ in their linear span. The only remaining question is that of the independence of the L_i . The key calculation is this. Write

$$L = \sum_{i=1}^{n} (\alpha_i \frac{\delta}{\delta \mathbf{x}_i} + \beta_i \mathbf{x}_i) .$$

Then a calculation verifies that

$$[L_{o}, L] = \sum_{i=1}^{n} (\widetilde{\alpha}_{i} \frac{\delta}{\delta x_{i}} + \widetilde{\beta}_{i} x_{i})$$

where

$$\begin{bmatrix} A & cc' \\ -bb' & -A' \end{bmatrix} \begin{bmatrix} \alpha \\ \beta \end{bmatrix} = \begin{bmatrix} \widetilde{\alpha} \\ \widetilde{\beta} \end{bmatrix}$$

Thus the independence of L_1, L_2, \ldots, L_{2n} is determined by the independence of the vectors

$$\begin{bmatrix} \mathbf{A} & \mathbf{c}\mathbf{c'} \\ -\mathbf{b}\mathbf{b'} & -\mathbf{A'} \end{bmatrix} \begin{bmatrix} \mathbf{c} \\ \mathbf{c} \\ \mathbf{0} \end{bmatrix} ; \mathbf{k} = 0, 1, \dots, 2\mathbf{n}-1 .$$

Now with a little work one can verify that the following generating function identity

$$\sum_{i=0}^{\infty} [0, c] \begin{bmatrix} A & cc' \\ b & \\ -bb' & -A' \end{bmatrix}^{1} \begin{bmatrix} c \\ 0 \end{bmatrix} s^{-i-1} = \frac{g(s) g(-s)}{1+g(s) g(-s)}$$

We see from standard results in realization theory [10] that the Lemma holds.

Incidentally, this Lie algebra is isomorphic to the Lie algebra which appears in the recent Cauchy-Riemann theory and it has the Heisenberg algebra as its derived algebra (see [11]).

The equations of motion for the conditional density in this case can be written in vector/matrix notation as

$$dz = (A - P(\tau) cc') zdt + P(\tau) dy; z(t) \in \mathbb{R}^{n}$$
$$\frac{d\tau}{dt} = 1$$

Here $P(\tau)$ is an **n** by n matrix which satisfies the following differential equation

$$\frac{\mathrm{d}}{\mathrm{d}\tau} \mathbf{P}(\tau) = \mathbf{A}\mathbf{P}(\tau) + \mathbf{P}(\tau)\mathbf{A}' + \mathbf{b}\mathbf{b}' - \mathbf{P}(\tau)\mathbf{c}\mathbf{c}'\mathbf{P}(\tau) \ .$$

We see that

$$\widetilde{\mathbf{f}} = \begin{bmatrix} (\mathbf{A} - \mathbf{P}(\tau) \mathbf{c} \mathbf{c}') \mathbf{x} \\ 1 \end{bmatrix} \quad \mathbf{and} \quad \widetilde{\mathbf{g}} = \begin{bmatrix} \mathbf{P} \mathbf{c} \\ 0 \end{bmatrix}$$

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are the local-coordinate descriptions of vector fields on \mathbb{R}^{n+1} . The vector fields in the Lie algebra which they generate are, with the exception of \tilde{T} itself, of the form

$$h = \begin{bmatrix} Pn_1 + n_2 \\ 0 \end{bmatrix}$$

with n_1 and n_2 in \mathbb{R}^n . Clearly $\{\widetilde{f}, \widetilde{g}\}_{LA}$ is then a Lie algebra of dimension 2n+1 or less. Additional calculations, which we omit here, show that it is of dimension 2n+1 under the hypothesis of the lemma.

To illustrate these ideas in the simple case we consider the basic example

$$dx = dw$$
; $dy = xdt + dv$

The filtering equations for the conditional mean z , given a gaussian initial distribution for $\, x \,$ with variance $\, k_{\, Q} \,$ are

$$dz = -kzdt + kdy$$
$$\dot{k} = -k^2 + 1$$

Introduce the vector fields

$$F = -kz \frac{\partial}{\partial z} + (1 - k^2) \frac{\partial}{\partial k}$$
$$G = k \frac{\partial}{\partial z} .$$

A calculation shows that

$$[F,G] = \frac{\partial}{\partial z} \stackrel{\text{def}}{=} H$$

and that

$$[F,H] = k \frac{\partial}{\partial z} = G$$
.

On the other hand, for the conditional density equation we have in this case

$$d\hat{\rho} = L_0 \hat{\rho} dt + L_1 \hat{\rho} dy$$

where

$$L_{o} = \frac{1}{2} \left(\frac{\delta^{2}}{\delta z^{2}} - z^{2} \right) , \quad L_{1} = z .$$

Now the Lie algebra of operators $\{L_0, L_l\}_{LA}$ satisfies

$$\begin{bmatrix} \mathbf{L}_{o}, \mathbf{L}_{1} \end{bmatrix} = \frac{\delta}{\delta z} \stackrel{\text{def}}{=} \mathbf{L}_{2}$$
$$\begin{bmatrix} \mathbf{L}_{o}, \mathbf{L}_{2} \end{bmatrix} = z = \mathbf{L}_{1}$$
$$\begin{bmatrix} \mathbf{L}_{1}, \mathbf{L}_{2} \end{bmatrix} = 1 \stackrel{\text{def}}{=} \mathbf{L}_{4} \quad .$$

There is a Lie algebra homomorphism from \pounds onto \mathcal{J} which sends L_0 into F, L_1 into G and has the operator L_4 in its kernel.

The Volterra series referred to in (b) of the introduction may, in this case, be computed to be t

$$\hat{y}(t) = e^{\alpha(t)} \hat{y}(0) + \int_{0}^{t} \int_{0}^{c} \beta(\rho) d\rho dy(\rho)$$

which is separable and has no nonlinear terms. The expression for α and β involves solving for k above.

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R.W. BROCKETT Division of Engineering and Applied Physics Harvard University, Pierce Hall CAMBRIDGE, MASSACHUSSETS 02138