

## MARKOVIAN EXTENSIONS AND REDUCTIONS OF A FAMILY OF $\sigma$ -ALGEBRAS

JELENA BULATOVIC GILL

Department of Statistics and Probability  
Michigan State University  
East Lansing, Michigan 48824

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**ABSTRACT.** In the first part of this paper different properties of two special realizations of a stochastic dynamic system, as well as relationships between the realization problem and a problem derived from it are investigated. In the second part, solutions of the following two problems (that follow directly from the realization problem) are given: how to find the minimal (resp. maximal) markovian flow of information (understood as a family of  $\sigma$ -algebras) that contains (resp. is contained in) given output of a system, and is such that each of these two flows of information is equally predictable by the other one.

**KEY WORDS AND PHRASES:** *Markovian family of  $\sigma$ -algebras; equal predictability; stochastic dynamic system; extension and reduction.*

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### 1. INTRODUCTION.

Let  $(\Omega, S, P)$  be an arbitrary probability space and let  $H = (H_t)_{t \in \mathbb{R}}$  be a family of  $\sigma$ -subalgebras of  $S$ ; new families  $P_t, P_{t-}, F_t, F_{t+}$  are defined in the usual way:  $P_t = \bigvee_{s \leq t} H_s, P_{t-} = \bigvee_{s < t} H_s, F_t = \bigvee_{s \geq t} H_s, F_{t+} = \bigvee_{s > t} H_s$ . One can think about  $P_{t-}$  and  $P_{t+}$  (resp.  $P_t$  and  $F_t$ ) as about real past and real future (resp. past and future) at the moment  $t$  of a flow of information carried by  $H$ .

The fact that a real-valued random variable  $X$  with finite expectation, defined on  $(\Omega, S, P)$ , is measurable with respect to some  $\sigma$ -algebra  $K$  will be written as  $X \in K$ . If  $K_1, K_2$  are arbitrary sub- $\sigma$ -algebras of  $S$ , then the smallest  $\sigma$ -algebra with respect to which all conditional expectations  $E(X|K_2), X \in K_1$ , are measurable, will be denoted as  $E(K_1|K_2)$ , [1].

The smallest  $\sigma$ -algebra that contains  $K_1$  and  $K_2$  will be denoted  $K_1 \vee K_2$ . If  $K_1$  and  $K_2$  are independent, we shall write  $K_1 \underline{\vee} K_2$  instead of  $K_1 \vee K_2$ .

If for every event  $A_2 \in K_2$  there exists an event  $A_1 \in K_1$  such that  $P(A_1 \Delta A_2) = 0$ , then it will be said that  $K_1 \supseteq K_2$  a.s. [P]. If  $K_1 \supseteq K_2$  a.s. [P] and  $K_2 \supseteq K_1$  a.s. [P], then it will be said that  $K_1 = K_2$  a.s. [P] (compare with [2]). All relationships indicated among  $\sigma$ -algebras will hold only a.s. [P].

Family  $H$  will be called markovian if

$$E(F_t | P_t) = H_t, t \in R$$

It will be said that  $H$  is the minimal (resp. maximal) family having some certain property if any other family  $H^*$  with the same property is such that  $H_t^* \supseteq H_t$  (resp.  $H_t^* \subseteq H_t$ ),  $t \in R$ .

Let  $A, A_1, A_2$  be  $\sigma$ -subalgebras of  $S$ .  $A_1$  and  $A_2$  are said to be conditionally independent given  $A$  if

$$E(X_1 X_2 | A) = E(X_1 | A) E(X_2 | A) \text{ for all } X_1 \in A_1, X_2 \in A_2; \quad (1.1)$$

we write  $A_1 \perp\!\!\!\perp A_2 | A$  and say, also, that  $A$  is splitting for  $A_1$  and  $A_2$  (compare with [1] and [3]).

Let  $A$  and  $B$  be  $\sigma$ -subalgebras of  $S$  such that  $A \subseteq B$ . It is said that  $A_{i,B}$  is the independent complement of  $A$  in  $B$  if  $A_{i,B} \subseteq B$  and

(i)  $A$  and  $A_{i,B}$  are independent  $\sigma$ -algebras;

(ii)  $A \vee A_{i,B} = B$ .

(see [2, Part II]). If  $A = H_t$ , we shall write  $H_{t,i,B}$  instead of  $(H_t)_{i,B}$ .

If  $A_{i,A_k \vee A}$  exists for  $k=1,2$ , then (1.1) is equivalent to the independence of  $A_{i,A_1 \vee A}$  and  $A_{i,A_2 \vee A}$ .

We shall say that families  $H$  and  $H^*$  are equally predictable by each other if

$$E(F_t^* | P_t) = E(F_t | P_t^*), t \in R. \quad (1.2)$$

If  $H_t^* \subseteq H_t$ ,  $t \in R$ , equality (1.2) becomes

$$E(F_t^* | P_t) = E(F_t^* | P_t^*) = E(F_t | P_t^*), t \in R. \quad (1.3)$$

Following [4, p. 181], we might say that the first of these equalities means that  $H$  does not anticipate  $H^*$ ; in that case, the second equality can be interpreted to mean that  $H$  is not anticipated by  $H^*$ .

If  $H^*$ ,  $H_t^* \subseteq H_t$ ,  $t \in R$ , is a markovian family, then (1.3) becomes

$$E(F_t^* | P_t) = H_t^* = E(F_t | P_t^*), t \in R. \quad (1.4)$$

It is easy to see that in this case (1.4) is equivalent to

$$F_t^* \perp\!\!\!\perp P_t | H_t^*, F_t \perp\!\!\!\perp P_t^* | H_t^*, t \in R, \quad (1.5)$$

which proves the following result.

**LEMMA 1.** Let  $H^*$  be a markovian family such that  $H_t^* \subseteq H_t$ ,  $t \in R$ . Then  $H$  and  $H^*$  are equally predictable by each other if and only if  $H_t^*$  is splitting for  $F_t^*$  and  $P_t$  as well as for  $F_t$  and  $P_t^*$  for every  $t$ .  $\square$

If  $H_t^* \supseteq H_t$ ,  $t \in R$ , then (1.2) becomes

$$E(F_t^* | P_t) = E(F_t | P_t) = E(F_t | P_t^*), t \in R;$$

thus,  $H^*$  is not anticipated by  $H$  and does not anticipate it. The following simple result will later be needed.

**LEMMA 2.** Let  $H^*$  be a markovian family such that  $H_t^* \supseteq H_t$ ,  $t \in R$ . If  $H^*$  does not anticipate  $H$ , then  $H^* \supseteq E(F_t | P_t)$ ,  $t \in R$ .

**PROOF.** Markovian property of  $H^*$  and  $F_t \subseteq F_t^*$  imply

$$H_t^* = E(F_t^* | P_t^*) \supseteq E(F_t | P_t^*) = E(F_t | P_t). \quad \square$$

In this paper we shall be concerned with the following two problems. The first one is establishing a relationship between the equal predictability of two families by each other and realizations of a stochastic dynamic system (which is, as we shall see, closely related to the problem of determining a  $\sigma$ -algebras splitting for two given  $\sigma$ -algebras) whose output is represented by one of these two families. The second one consists of finding, when H is given, the maximal (resp. minimal) markovian family  $H^*$ , such that H and  $H^*$  are equally predictable by each other and  $H_t^* \subseteq H_t$  (resp.  $H_t^* \supseteq H_t$ ),  $t \in R$ ; those two families will be called, respectively, maximal markovian reduction and minimal markovian extension of family H.

2. PREDICTABILITY AND STOCHASTIC DYNAMIC SYSTEM.

A family  $H = (H_t)_{t \in R}$  of  $\sigma$ -algebras from S can be considered as a flow of information representing output of a stochastic dynamic system. More precisely, a stochastic dynamic system consists of two families, H and  $H^*$ , which satisfy condition

$$P_{t-} \vee P_t^* \parallel F_{t+} \vee F_t^* | H_t^*, t \in R; \tag{2.1}$$

here, H represents outputs and  $H^*$  states of that system. For a given family of outputs H, any family  $H^*$  satisfying (2.1), is called a realization of the system with those outputs. In practice, two additional families  $H^1 = (H_t^1)_{t \in R}$  and  $H^2 = (H_t^2)_{t \in R}$ ,  $H_t^1 \subseteq H_t^2$ ,  $t \in R$ , are given and it is required that a realization  $H^*$  is to be such that

$$H_t^1 \subseteq H_t^* \subseteq H_t^2, t \in R. \tag{2.2}$$

(for detailed definitions of a stochastic dynamic system and the realization problem see e.g. [ 1]).

If  $H^*$  is a realization of a stochastic dynamic system, that is if (2.1) holds, then the following relations are true:

$$P_{t-} \parallel F_t^* | H_t^*, P_t^* \parallel F_{t+} | H_t^*, P_t^* \parallel F_t^* | H_t^*; \tag{2.3}$$

the third of these relations simply means that family  $H^*$  is markovian. It is clear that, generally, (2.1) and (2.3) are not equivalent.

In the previous section, in connection with equal predictability of H and  $H^*$  by each other, we were, when H was given, considering cases  $H_t^* \subseteq H_t$  and  $H_t^* \supseteq H_t$ ,  $t \in R$ . Corresponding cases concerning a realization  $H^*$  of a stochastic dynamic system with outputs H are connected with families  $H^1$  and  $H^2$  from (2.2) in the following ways:

$$H_t^1 = \{\phi, \Omega\}, H_t^2 = H_t, t \in R; \tag{2.4}$$

$$H_t^1 = H_t, H_t^2 = S, t \in R. \tag{2.5}$$

Obviously, (2.4) means  $H_t^* \subseteq H_t$  and (2.5) means  $H_t^* \supseteq H_t$ ,  $t \in R$ .

If family H is given, then family  $H^*$  that satisfies (2.3) is a markovian family which is not anticipated by the real past of H and does not anticipate the real future of H. It is clear that, if H and  $H^*$  are equally predictable by each other,  $H_t^* \subseteq H_t$ ,  $t \in R$ , and  $H^*$  is markovian, then (2.3) holds, although the converse is not always true.

In this section we shall investigate relationships between (1.2), (2.1) and (2.3), when families  $H^1$  and  $H^2$  from (2.2) are given by (2.4) and (2.5). Some of the results obtained will lead to the main results, given in the next section.

Family  $H^*$ , defined by  $H_t^* = P_t$ ,  $t \in R$ , satisfies (1.2), (2.1), (2.3) and (2.5). However, family  $H^*$ , defined by  $H_t^* = \{\phi, \Omega\}$ , satisfies (1.2), (2.3) and (2.4), but not

(2.1). It is clear that, just as it is of interest to look for the minimal realization of a dynamic system, in the case of a family  $H^*$  satisfying (1.2) (or (2.3)), it is of interest to find maximal among such families in case (2.4), and minimal among them in case (2.5).

LEMMA 3. Let  $H^*$  be a family satisfying (2.1). If  $H_t^* \subseteq H_t$  and  $H_t \perp\!\!\!\perp F_{t+} | P_{t-} \vee H_t^*$ ,  $t \in R$ , then family  $H$  itself is markovian.

PROOF. From (2.1) it follows  $P_{t-} \perp\!\!\!\perp F_{t+} | H_t^*$ , which is equivalent to  $E(F_{t+} | P_{t-} \vee H_t^*) \subseteq H_t^*$ , so that, because  $H_t^* \subseteq P_{t-} \vee H_t^*$ ,

$$E(F_{t+} | P_{t-} \vee H_t^*) = E(F_{t+} | H_t^*), \quad t \in R. \quad (2.6)$$

The assumption of Lemma 3 is equivalent to  $E(F_{t+} | P_t) \subseteq P_{t-} \vee H_t^*$ , which, because of  $P_{t-} \vee H_t^* \subseteq P_t$  and (2.6), gives

$$E(F_{t+} | P_t) = E(F_{t+} | P_{t-} \vee H_t^*) = E(F_{t+} | H_t^*), \quad t \in R$$

that is, equivalently,

$$F_{t+} \perp\!\!\!\perp P_t | E(F_{t+} | H_t^*), \quad t \in R. \quad (2.7)$$

Since  $E(F_{t+} | H_t^*) \subseteq H_t$ , (2.7) implies  $F_{t+} \perp\!\!\!\perp P_t | H_t$ , that is

$$F_t \perp\!\!\!\perp P_t | H_t, \quad t \in R,$$

which is, because  $H_t \subseteq F_t$ ,  $H_t \subseteq P_t$ , equivalent to the fact that  $H$  is a markovian family.  $\square$

LEMMA 4. Let family  $H$  be an output of a stochastic dynamic system. The minimal realization  $H^m$  of that system, satisfying condition

$$H_t \subseteq H_t^m \subseteq P_t, \quad t \in R, \quad (2.8)$$

is defined by

$$H_t^m = E(F_t | P_t), \quad t \in R.$$

PROOF. It can be shown that  $H_t^*$  is splitting for  $P_{t-}$  and  $F_{t+}$  if and only if it is splitting for  $P_{t-} \vee \bar{H}_t$  and  $F_{t+} \vee \hat{H}_t$ , where  $\bar{H}_t \subseteq H_t^*$  and  $\hat{H}_t \subseteq H_t^*$  (see [1]). Thus, because of (2.8),  $H_t^*$  must be splitting for  $P_t$  and  $F_t$ . But, the minimal  $\sigma$ -algebra from  $P_t$  that splits  $P_t$  and  $F_t$  is defined by  $E(F_t | P_t)$ ,  $t \in R$  (see [5]). Put  $H_t^m = E(F_t | P_t)$ ,  $t \in R$ . From  $H_s \subseteq H_s^* \subseteq P_s \subseteq P_t$  for  $s \leq t$ , it follows  $P_t = P_t^m$ , which gives

$$P_{t-} \vee P_t^m \perp\!\!\!\perp F_t | H_t^m, \quad t \in R.$$

It is easy to check that  $H^m$  is a markovian family, which, because  $F_t \subseteq F_t^m$ , implies

$$P_{t-} \vee P_t^m \perp\!\!\!\perp F_{t+} \vee F_t^m | H_t^m, \quad t \in R. \quad \square$$

LEMMA 5. Let  $A = (A_t)_{t \in R}$  be a family of  $\sigma$ -algebras such that  $A_t \subseteq P_t$  and  $A_t$  is independent of  $H_t$  for every  $t \in R$ . If  $H^*$  is a markovian family of  $\sigma$ -algebras such that  $H_t^* \subseteq H_t$ ,  $t \in R$  and that (1.5) is satisfied, then  $H_t^*$  is independent of  $E(F_u | A_u)$  for all  $t, u \in R$ .

PROOF. The second equality in (1.5) and  $A_u \subseteq P_u$  imply  $E(E(F_u | A_u) | P_u^*) \subseteq E(F_u | P_u^*) = H_u^*$ , but because  $H_u^* \subseteq H_u$ , this gives  $E(E(F_u | A_u) | P_u^*) = \{\phi, \Omega\}$ . Thus,  $H_t^*$  is independent of  $E(F_u | A_u)$  for all  $t \leq u$ .

By using the first equality in (1.5) one gets  $E(F_u^* | E(F_u | A_u)) \subseteq E(F_u^* | P_u) = H_u^*$ , which,

because  $H_u^*$  and  $A_u$  are independent, means that  $E(F_u^* | E(F_u | A_u)) = \{\phi, \Omega\}$ . Thus,  $H^*$  is independent of  $E(F_u | A_u)$  for all  $t \geq u$ .  $\square$

COROLLARY 1. Let  $H$  be a family of  $\sigma$ -algebras such that  $H_{t,i,P_t}$  exists for every  $t \in R$ . If  $H^*$  is a markovian family of  $\sigma$ -algebras such that  $H_t^* \subseteq H_t$ ,  $t \in R$ , and (1.5) is satisfied, then  $H_t^*$  is independent of  $E(F_u | H_{u,i,P_u})$  for all  $t, u \in R$ .

3. MAIN RESULTS.

Let  $H$  be a given family of  $\sigma$ -algebras. The solution of the problem of finding its minimal markovian extension, is given by the following result.

THEOREM 1. If a family  $H^m = (H_t^m)_{t \in R}$  is such that  $H_t^m \supseteq H_t$ ,  $t \in R$ , then  $H^m$  is the minimal markovian extension of  $H$  if and only if  $H^m$  is the minimal realization (satisfying (2.2), where  $H^1$  and  $H^2$  are as in (2.5)) of a stochastic dynamic system whose output is  $H$ .

PROOF. Let  $H^m$  be the minimal realization (satisfying (2.2) and (2.5)) of a stochastic dynamic system whose output is  $H$ . According to Lemma 4,  $H^m$  is defined by  $H_t^m = E(F_t | P_t)$ ,  $t \in R$ . From obvious equality  $P_t^m = P_t$ , it follows  $E(F_t | P_t) = E(F_t | P_t^m)$ , and, because  $H^m$  is markovian,  $E(F_t^m | P_t) = (F_t^m | P_t^m) = H_t^m = E(F_t | P_t)$ , which proves that  $H$  and  $H^m$  are equally predictable by each other.

If  $H^*$  is some other markovian extension of  $H$  (such that  $H^*$  and  $H$  are equally predictable by each other), then, according to Lemma 2, it follows that  $H_t^* \supseteq E(F_t | P_t) = H_t^m$ , which proves the minimality of  $H^m$ .

The other half of the proof is obvious.  $\square$

THEOREM 2. Let  $H$  be a family of  $\sigma$ -algebras such that  $H_{t,i,P_t}$  exists for every  $t \in R$ , and let  $A$  be a  $\sigma$ -algebra from  $P$  that is independent of  $E(F_t | H_{t,i,P_t})$  for each  $t \in R$ . Then family  $H^* = (H_t^*)_{t \in R}$ , defined by

$$H_t^* = E(A | H_t), \quad t \in R, \tag{3.1}$$

is a markovian reduction of  $H$ . If independent complement  $\sum_{t,i,P}$  of  $\sum_t = E(F_t | H_{t,i,P_t})$  exists for every  $t \in R$ , and if  $A$  from (3.1) is replaced by

$$A^M = \bigcap_{t \in R} \sum_{t,i,P}, \tag{3.2}$$

then family  $H^*$  is the maximal markovian reduction of  $H$ .

PROOF. Equation (3.1) implies  $H_t^* \subseteq H_t$ ,  $t \in R$ , so that  $H^*$  is a reduction. If we define  $A_t$  by  $A_t = E(A | P_t)$ , then we get  $H_t^* \subseteq A_t$ , which gives  $E(A | H_t) \subseteq E(H_t | A_t)$ . But, on the other side, if an event  $A \in H_t$  is independent of  $H_t^*$ , then it is independent of  $A$ , so that  $E(H_t | A_t) \subseteq H_t^*$ , which, together with the previous inclusion, gives

$$H_t^* = E(H_t | A_t), \quad t \in R. \tag{3.3}$$

Let  $s, t$  be such that  $s \geq t$ . The markovian property of  $H^*$  is implied by (3.3) and  $E(H_s^* | P_t^*) \subseteq E(H_s^* | A_t) = E(E(H_s | A_s) | A_t) = E(H_s | A_t) = E(E(H_s | P_t) | A_t) \subseteq E(H_t | A_t) = H_t^*$ .

For  $s \geq t$ , we get  $E(H_s | P_t^*) \subseteq H_t^*$ , which implies

$$E(F_t | P_t^*) = H_t^*. \tag{3.4}$$

Also, we have  $E(H_s^* | P_t) = E(H_s^* | H_t) \subseteq E(A | H_t) = H_t^*$ , which, together with (3.4), prove

that  $H$  and  $H^*$  are equally predictable by each other. Thus, family  $H^*$  is a markovian reduction of  $H$ .

Let us prove that, if  $H_t^M = E(A^M | H_t)$ , where  $A^M$  is defined by (3.2), then  $H^M$  is the maximal such family. Let  $H^1$  be any markovian reduction of  $H$  such that  $H^1$  and  $H$  are equally predictable by each other. From Corollary 1 it follows  $H_t^1 \subseteq \bigcap_{u,i,p} H_{t,u,i,p}$  for all  $t,u$ , that is  $H_t^1 \subseteq A^M$ , which implies  $H_t^1 \subseteq E(A^M | H_t) = H_t^M$ , so that maximality of  $H^M$  is proved.  $\square$

#### 4. EXAMPLES AND COMMENTS.

Let  $H = (H_t)_{t=1,2,3}$  be a family of three arbitrary  $\sigma$ -algebras, and let families  $H^* = (H_t^*)_{t=1,2,3}$  and  $H^{**} = (H_t^{**})_{t=1,2,3}$  be defined by

$$H_1^* = H_1, H_2^* = H_1 \vee H_2, H_3^* = H_3$$

$$H_1^{**} = H_1, H_2^{**} = H_2 \vee H_3, H_3^{**} = H_3$$

It is easy to see that both families  $H^*$  and  $H^{**}$  are markovian and contain information carried by  $H$  at any given  $t=1,2,3$ . However, there is no way in which  $H^*$  and  $H^{**}$  could be compared and, also, neither one of them and  $H$  are equally predictable by each other. Both of these families are realizations of a stochastic dynamic system with output  $H$ , but they are not minimal. According to Theorem 1, family  $H^m$ , defined by  $H_1^m = H_1, H_2^m = E(H_2 \vee H_3 | H_1 \vee H_2), H_3^m = H_3$  is the minimal realization of that system and the minimal markovian extension of  $H$ .  $\square$

Let output  $H = (H_t)_{t=1,2,3}$  of a stochastic dynamic system be defined by  $H_1 = B, H_2 = C, H_3 = B$ , where  $C \subseteq B$ . Realization of this system that satisfies (2.4) does not exist. But the maximal markovian reduction  $H^M$  of  $H$  is given by  $H_1^M = H_2^M = H_3^M = A$ . Note that families  $H^*$  and  $H^{**}$ , defined by  $H_1^* = H_2^* = A, H_3^* = B$ , and  $H_1^{**} = B, H_2^{**} = H_3^{**} = A$  are also markovian reductions but neither one of them and  $H$  are equally predictable by each other.  $\square$

Let us now suppose that  $H = (H_t)_{t=1,2,3,4}$  is a family of  $\sigma$ -algebras which are all mutually independent except  $H_1$  and  $H_3$ . If  $\bigcap = E(H_3 | H_1)$  and if  $\bigcap_{i,p}$  exists, then family  $H^*$ , defined by

$$H_1^* = \{\phi, \Omega\}, H_2^* = H_2, H_3^* = E(H_3 | \bigcap_{i,p}), H_4^* = H_4,$$

is a markovian reduction of  $H$ , but it is not the maximal one. By using Theorem 2, we can show that the maximal reduction is defined by

$$H_1^M = E(H_3 | H_1)_{i,H_1}, H_2^M = H_2, H_3^M = E(H_3 | E(H_3 | H_1)_{i,H_1 \vee H_3}), H_4^M = H_4.$$

Note that dynamic system with output  $H$  does not have realization satisfying (2.4).

#### REFERENCES

1. PUTTEN, C. VAN, SCHUPPEN, J. H. VAN. On Stochastic Dynamical Systems. To appear in the Proceedings of the Fourth International Symposium on the Mathematical Theory of Networks and Systems, July 3-6, 1979, Delft, Netherlands.
2. RAMACHANDRAN, D. Perfect Measures, I and II, ISI Lecture Notes No. 7, MacMillan Company of India Limited, 1979.
3. RUCKEBUSCH, G. Théorie géométrique de la Représentation Markovienne. Ann. Inst. Henri Poincaré, XVI, no. 3 (1980) 225-297.
4. ROZANOV, Yu. A. Markov Random Fields. Springer-Verlag, 1982.
5. SKIBINSKY, M. Adequate Subfields and Sufficiency. Ann. Math. Statist. 38 (1967) 155-161.