

## 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  $H_t^1 \subseteq \bigcap_{u,i,p} \mathcal{L}_{u,i,p}$  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  $H_1 = \mathcal{B}$ ,  $H_2 = C$ ,  $H_3 = \mathcal{B}$ , where  $C \subseteq \mathcal{B}$ . 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  $H_1^* = H_2^* = A$ ,  $H_3^* = \mathcal{B}$ , and  $H_1^{**} = \mathcal{B}$ ,  $H_2^{**} = H_3^{**} = A$  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

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

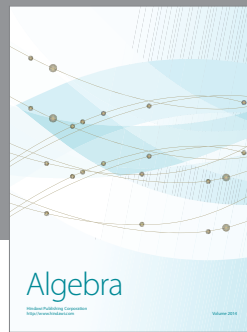
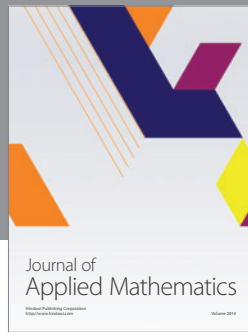
is a markovian reduction of  $H$ , but it is not the maximal one. By using Theorem 2, we can show that the maximal reduction is defined by

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

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

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