

## نمایش فرایندهای گاوسی با خاصیت تقابل بوسیله مارتینگل‌ها

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### چکیده

در این مقاله مفهوم خاصیت تقابل فرایندهای گاوسی در فضای هیلبرت معرفی و به کمک تکنیک‌های موجود در این فضا نمایشی بوسیله مارتینگل‌ها برای فرایندهای گاوسی با خاصیت تقابل ارائه می‌گردد. ثابت می‌شود که این نمایش منحصر بفرد و مشخص‌کننده فرایندهای گاوسی با خاصیت تقابل است.

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## A Martingale Representation of Gaussian Reciprocal Processes.

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### Abstract

In This paper the notion of reciprocity in the Hilbert space is considered, and using techniques in a Hilbert space we introduce a martingale representation for the Gaussian mean zero reciprocal processes. It is shown that the representation is unique and completely characterizes the Gaussian reciprocal processes.

## Introduction

Let  $\{X_t ; t \in [a, b]\}$  be a mean zero Gaussian process on a complete probability space  $(S, F, P)$ . It is well known (see for example A.N. Shiriyayev, 1984) that for each  $t$ ,  $X_t$  can be considered as an element of the Hilbert space

$$H = \overline{Sp} \{X_t ; t \in [a, b]\}$$

where "Sp" stands for the span closure under the inner product given by

$$(X_t, X_s) = E(X_t \cdot X_s) = \int_s X_t \cdot X_s dP.$$

Thus, the probabilistic independence of two random variables is equivalent to their orthogonality as elements of  $H$ . Because of this property the counterpart of the conditional expectation would be the orthogonal projections on the subspaces of  $H$ . (J.Neveu, 1975). Hence the definition of Markov, martingale and reciprocal mean zero Gaussian processes, which are based on the conditional expectation, can be stated as orthogonal projections on the subspaces of  $H$ . In Section 1 we give these definitions and provide some of their elementary property.

In Section 2 we give the martingale representation of reciprocal processes. In this section we give a modified result of Dang-Ngoc and Royer (1978) which provides a condition under which a reciprocal process will be a Markov one. The converse is always true, i.e. a Markov process is always reciprocal.

The notion of reciprocal processes was introduced first by Jamison (1970). A reciprocal process is a process that have Markov property on finite intervals. The Markov property on sets (in particular on finite intervals) is discussed by

V.Mandrekar (1976). Jamison classified the covariance function of stationary mean zero Gaussian reciprocal processes.

The technique of Hilbert space that we have considered in this paper is an alternative approach to the study of reciprocal processes.

The representation that we give in this paper will provide the use of stochastic calculus theory as well as other properties of martingales in the area of reciprocal processes.

## 1. Preliminaries and notations

Throughout this paper we consider only mean zero real valued Gaussian processes on a complete probability space  $(S, F, P)$ .

Let  $\{X_t\}$  be a process and consider the Hilbert space generated by  $\left\{ \sum_{i=1}^n a_i X_{t_i} : n \in \mathbb{N}, a_i \in \mathbb{R} \right\}$  with the inner product given by

$$(X_t, X_s) = E(X_t \cdot X_s) = \int_s X_t \cdot X_s dP.$$

We denote this space by  $H = H(X) = Sp\{X_t\}$ .

For a real  $t$  we define the following subspaces of  $H$ :

$$\begin{aligned} H_t^- (X) &= \overline{Sp}\{X_s : s \leq t\} && \text{"Past"} \\ H_t^+ (X) &= \overline{Sp}\{X_s : s \geq t\} && \text{"Future"} \\ L_t (X) &= \overline{Sp}\{X_t\} && \text{"Present"} \end{aligned}$$

and for the reals  $u, v$ ;  $u < v$  let:

$$\begin{aligned} H_{u,v}^- (X) &= \overline{Sp}\{X_s : s \leq u \text{ or } s \geq v\} \\ H_{u,v}^+ (X) &= \overline{Sp}\{X_s : s \in [u, v]\} \\ L_{u,v} (X) &= \overline{Sp}\{X_u, X_v\}. \end{aligned}$$

If there is no ambiguity we will write

$H_t^-, H_t, L_t^+, H_{u,v}^-, H_{u,v}^+$  and  $L_{u,v}$  instead of the above notations.

**1.1. Definitions.** The process  $\{X\}$  is a

(i) Markov process if for each  $t$ , the orthogonal projection of  $H_t^+$  on  $H_t^-$  is exactly the same as

the orthogonal projection of  $H_t^+$  on  $L_t$ . . In symbols:

$$P_{H_t^-}^{H_t^+} = P_{L_t}^{H_t^+}, \text{ (P stands for orthogonal projection).}$$

For convenience we write  $(H_t^+ | H_t^-)$  for  $P_{H_t^-}^{H_t^+}$ , and use similar notations in similar cases.

(ii) Reciprocal process, if for each  $u \leq v$  we have:

$$(H_{u,v}^+ | H_{u,v}^-) = (H_{u,v}^+ | L_{u,v})$$

(iii) Martingale, if for each  $t \geq s$  we have:

$$(X_t | H_s^-) = X_s.$$

We have the following symmetry property for these notions.

**1.2. Theorem.** If  $\{X\}$  is a

(i) Markov-process, then for each  $t$ ;

$$(H_t^- | H_t^+) = (H_t^- | L_t)$$

(ii) reciprocal process, then for each  $u < v$ ;

$$(H_{u,v}^- | H_{u,v}^+) = (H_{u,v}^- | L_{u,v})$$

(iii) martingale, then for each  $t < s$ ;

$$(X_t | H_s^+) = X_s.$$

*Proof.* We give a proof for (ii). The other parts follow by a similar technique.

Let  $t \in (u, v)$  and  $Y_t = (X_t | H_{u,v}^+)$ . In order to show the equality in (ii) it suffices to show that  $Y_t \in L_{u,v}$ , or equivalently  $(Y_t | L_{u,v}) = Y_t$ . To show this, we observe that

$$\begin{aligned} (Y_t | L_{u,v}) &= \left[ (X_t | H_{u,v}^+) | L_{u,v} \right] \\ &= (X_t | L_{u,v}) \quad (L_{u,v} \subset H_{u,v}^+) \\ &= (X_t | H_{u,v}^+) = Y_t \quad (\{X_t\} \text{ is reciprocal}), \end{aligned}$$

which completes the proof.

The following theorem states that reciprocal property is weaker than Markov property.

**1.3. Theorem.** A Markov process is reciprocal.

*Proof.* Let  $\{X_t\}$  be a Markov process and  $t \in (u, v)$ ,

we have:

$$\begin{aligned} H_{u,v}^- &= H_u^- \oplus (H_{u,v}^- \ominus H_u^-) \\ &= H_u^+ \oplus (H_{u,v}^- \ominus H_u^-). \end{aligned}$$

Thus

$$\begin{aligned} (X_t | H_{u,v}^-) &= (X_t | H_u^-) + (X_t | H_{u,v}^- \ominus H_u^-) \\ &= (X_t | L_u) + (X_t | H_{u,v}^- \ominus H_u^-). \end{aligned}$$

Similarly

$$(X_t | H_{u,v}^-) = (X_t | L_v) + (X_t | H_{u,v}^- \ominus H_v^+).$$

Subtracting the two sides of the above equality

we get:

$$(X_t | L_u) - (X_t | L_v) = (X_t | H_{u,v}^- \ominus H_v^+) - (X_t | H_{u,v}^- \ominus H_u^-)$$

Since  $(X_t | L_u)$  is in the space  $L_u$  it is of the form  $AX_u$ , where  $A$  is a real which depends on  $t$  and  $u$ , similarly  $(X_t | L_v)$  is of the form  $BX_v$ ; thus

$$AX_u - BX_v = (X_t | H_{u,v}^- \ominus H_v^+) - (X_t | H_{u,v}^- \ominus H_u^-).$$

We project the two sides of the above equation on  $H_v^+$  and get

$$A(X_u | H_v^+) - BX_v = (X_t | H_{u,v}^- \ominus H_u^-);$$

the above equality is based on the facts that

$X_v \in H_v^+$  and that  $(X_t | H_{u,v}^- \ominus H_v^+)$  is orthogonal to  $H_v^+$ . Again  $(X_u | H_v^+)$  is of the form  $CX_v$  for some constant  $C$ . We have thus

$$(X_t | H_{u,v}^- \ominus H_u^-) = \lambda X_v \quad (\text{for some real } \lambda).$$

Similarly

$$(X_t | H_{u,v}^- \ominus H_v^+) = BX_u$$

Therefore

$$(X_t | H_{u,v}^-) = \lambda X_v + BX_u$$

which states that the projection of  $X_t$  on  $H_{u,v}^-$  is in  $L_{u,v}$ , thus the proof is complete.

In Section 2 we refer to the following theorem.

The proof is omitted.

**1.4. Theorem** (V. Mandrekar (1974)).

Let  $\{X_t\}$  be a Markov process, then there exists a unique martingale  $M$  adapted to the filter  $F_t = \sigma\{X_s : s \leq t\}$  and a unique deterministic never

vanishing function  $f$  such that for each  $t$

$$X_t = f(t)M_t.$$

The following theorem is a result which provides a condition that a reciprocal process will be a Markov one.

**1.5. Theorem.** Let  $\{X_t : t \in [a, b]\}$  be a mean zero Gaussian reciprocal process. If either  $X_a$  or  $X_b$  is constant then the process is a Markov one.

*Proof.* Assume that  $x$  is constant (since  $EX_b = 0$ , we will have  $X_b \equiv 0$ ). Let  $a < s < t < b$ , we have

$$\begin{aligned} (X_t | H_s^-) &= (X_t | H_s^- \oplus \overline{Sp}\{X_b\}) \\ &= (X_t | H_{s,b}^-) \quad (X_t \text{ is reciprocal}) \\ &= (X_t | \overline{Sp}\{X_s, X_b\}) \\ &= (X_t | \overline{Sp}\{X_s\}) \\ &= (X_t | L_s) \end{aligned}$$

i.e.  $\{X_t\}$  is a Markov process.

## 2. A representation of reciprocal processes

Let  $\{X_t : t \in [a, b]\}$  be a mean zero Gaussian reciprocal process. Let

$$Z_t = (X_t | \overline{Sp}\{X_a, X_b\}) = \alpha(t)X_a + B(t)X_b, \quad Y_t = X_t - Z_t$$

It is clear that  $\{Y_t\}$  is orthogonal to  $\{Z_t\}$ . Now we have the following theorem:

**2.1. Theorem.** The above introduced  $\{Y_t\}$  is a Markov process.

*Proof.* First we observe that for each  $u < v$

$$H_{u,v}^+(X_t) = H_{u,v}^+(Y_t + Z_t) + H_{u,v}^+(Y_t) + H_{u,v}^+(Z_t)$$

Now we show that  $\{Y_t\}$  is a reciprocal process. Let

$a < u < v < b$  and  $t \in (u, v)$ , we have:

$$\begin{aligned} &(Y_t | H_{u,v}^-(Y)) \\ &= (Y_t | H_{u,v}^-(X)) \quad (\{Z_t\} \text{ and } \{Y_t\} \text{ are orthogonal}), \\ &= (X_t - Z_t | H_{u,v}^-(X)) \\ &= (X_t | H_{u,v}^-(X)) - Z_t \quad (Z_t \in H_{u,v}^-(X)) \\ &= (X_t | L_{u,v}(X)) - Z_t \quad (X_t \text{ is reciprocal}) \\ &= AX_u + BX_v - Z_t \quad (\text{for same reals } A, B) \end{aligned}$$

Now substituting for  $X_u$  and  $X_v$  in terms of  $Y$  and

$Z$  we get:

$$\begin{aligned} AX_u + BX_v &= A(Y_u + Z_u) + B(Y_v + Z_v) \\ &= AY_u + BY_v + AZ_u + BZ_v \\ &= AY_u + BY_v + (AX_u + BX_v | \overline{Sp}\{X_a, X_b\}) \\ &= AY_u + BY_v + ((X_t | L_{u,v}(X)) | \overline{Sp}\{X_a, X_b\}) \\ &= AY_u + BY_v + ((X_t | H_{u,v}^-(X)) | \overline{Sp}\{X_a, X_b\}) \\ &= AY_u + BY_v + (X_t | \overline{Sp}\{X_a, X_b\}) \\ &= AY_u + BY_v + Z_t. \end{aligned}$$

Thus

$$\begin{aligned} (Y_t | H_{u,v}^-(Y)) &= AY_u + BY_v + Z_t - Z_t \\ &= AY_u + BY_v. \end{aligned}$$

Hence

$$\begin{aligned} (Y_t | \overline{Sp}\{Y_u, Y_v\}) &= ((Y_t | H_{u,v}^-) | \overline{Sp}\{Y_u, Y_v\}) \\ &= (AY_u + BY_v | \overline{Sp}\{Y_u, Y_v\}) \\ &= AY_u + BY_v. \end{aligned}$$

$$\text{i.e. } (Y_t | H_{u,v}^-(Y)) = (Y_t | \overline{Sp}\{Y_u, Y_v\})$$

which completes the proof of reciprocity of  $\{Y_t\}$ .

In order to show that  $\{Y_t\}$  is a Markov process, by Theorem 1.5 it suffices to show that the "remote past" of the process is trivial. In other words, it suffices to show that

$$\bigcap_{u < v} \overline{Sp}\{Y_t : t \notin (u, v)\} = \{0\}$$

To show this we observe that

$$\begin{aligned} \overline{Sp}\{Y_t : t \notin (u, v)\} &= \overline{Sp}\{X_t : t \notin (u, v)\} \ominus \overline{Sp}\{Z_t : t \notin (u, v)\} \\ &= \overline{Sp}\{X_t : t \notin (u, v)\} \ominus \overline{Sp}\{X_a, X_b\} \end{aligned}$$

Therefore

$$\begin{aligned} \bigcap_{u < v} \overline{Sp}\{Y_t : t \notin (u, v)\} &= \left\{ \bigcap_{u < v} \overline{Sp}\{X_t : t \notin (u, v)\} \ominus \overline{Sp}\{X_a, X_b\} \right\} \\ &= \overline{Sp}\{X_a, X_b\} - \overline{Sp}\{X_a, X_b\} \end{aligned}$$

which completes the proof.

Now we have the following main result which provides a representation for the mean zero Gaussian reciprocal processes.

**2.2.Theorem** (main result).

Let  $\{X_t, t \in [a, b]\}$  be a mean zero Gaussian process.  $\{X_t\}$  is reciprocal if and only if it has the following representation

$$X_t = Y_t + Z_t$$

where  $\{Y_t\}$  is a Markov process,  $\{Z_t\}$  is orthogonal to  $\{Y_t\}$  and lies in the "remote past" of  $\{X_t\}$ .

Moreover this representation is unique, in the sense that if

$$X_t = Y_t' + Z_t',$$

where  $\{Y_t'\}$  and  $\{Z_t'\}$  have the same property as  $\{Y_t\}$  and  $\{Z_t\}$  respectively, then

$$Y_t' = Y_t \text{ and } Z_t' = Z_t$$

*Proof.* If  $\{X_t\}$  is reciprocal then by Theorem 2.1  $X_t$  has the required form.

Conversely let  $X_t = Y_t + Z_t$ ,  $\{Y_t\}$  and  $\{Z_t\}$  satisfy the condition mentioned in the theorem. Let  $t \in (u, v)$ ,  $u < v$ ; then

$$\begin{aligned} & (X_t | H_{u,v}^-(X)) \\ &= (Y_t + (X_t | \overline{Sp}\{X_a, X_b\}) | H_{u,v}^-(X)) \\ &= (Y_t | H_{u,v}^-(X)) + (X_t | \overline{Sp}\{X_a, X_b\}) \\ &= (Y_t | H_{u,v}^-(Y) \oplus H_{u,v}^-(Z)) + (X_t | \overline{Sp}\{X_a, X_b\}) \\ &= (Y_t | H_{u,v}^-(Y)) + (X_t | \overline{Sp}\{X_a, X_b\}) \end{aligned}$$

the last equality is due to the orthogonality of  $\{Y_t\}$  and  $\{Z_t\}$ . By Markov property of  $\{Y_t\}$  we have:

$$\begin{aligned} & (X_t | H_{u,v}^-(X)) \\ &= (Y_t | \overline{Sp}\{Y_u, Y_v\}) + (X_t | \overline{Sp}\{X_a, X_b\}) \\ &= AY_u + BY_v + (X_t | \overline{Sp}\{X_a, X_b\}) \end{aligned}$$

for some constants A and B.

On the other hand:

$$\begin{aligned} & (X_t | H_{u,v}^-(X)) \\ &= (X_t | (H_{u,v}^-(X) \ominus \overline{Sp}\{X_a, X_b\}) \oplus \overline{Sp}\{X_a, X_b\}) \\ &= (X_t | H_{u,v}^-(X) \ominus \overline{Sp}\{X_a, X_b\}) + (X_t | \overline{Sp}\{X_a, X_b\}) \end{aligned}$$

Comparing the two values of  $(X_t | H_{u,v}^-(X))$

we conclude that:

$$\begin{aligned} & (X_t | H_{u,v}^-(X) \ominus \overline{Sp}\{X_a, X_b\}) \\ &= AX_u + BX_v - (AX_u + BX_v | \overline{Sp}\{X_a, X_b\}) \end{aligned}$$

which implies that

$$(X_t | H_{u,v}^-(X)) = AX_u + BX_v.$$

By (ii) of Theorem 1.2. we conclude that  $\{Y_t\}$  is reciprocal.

Uniqueness: Since  $\overline{Sp}\{Y_a, Y_b\} = (0)$  and Y is orthogonal to  $\overline{Sp}\{Z_a, Z_b\}$  we get  $\overline{Sp}\{X_a, X_b\} = \overline{Sp}\{Z_a, Z_b\} = \overline{Sp}\{Z_a', Z_b'\}$ .

So

$$\begin{aligned} (X_t | \overline{Sp}\{X_a, X_b\}) &= (Y_t + Z_t | \overline{Sp}\{X_a, X_b\}) = Z_t \\ (X_t | \overline{Sp}\{X_a, X_b\}) &= (Y_t' + Z_t' | \overline{Sp}\{X_a, X_b\}) = Z_t' \end{aligned}$$

Therefore  $Z_t = Z_t'$  and  $Y_t = Y_t'$  and this proves the uniqueness of the representation.

Combining Theorems 2.2 and 1.4 we have the following theorem.

**2.3. Theorem.**  $\{X_t\}$  is a reciprocal process if and only if it has the following representation:

$$X_t = f(t) M_t + Z_t$$

where

$f(\cdot)$  is a deterministic function.

$\{M_t\}$  is a  $\{X_t\}$  adapted martingale

$\{Z_t\}$  is orthogonal to  $\{M_t\}$  and lies in the remote past of  $\{X_t\}$ .

Moreover this representation is unique.

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