

## GAUSSIAN PROCESSES WITH MARKOVIAN COVARIANCES

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**ABSTRACT.** We show that any Gaussian process can be derived in a simple manner from a Markov process if it has zero mean and covariance identical to the covariance of a real valued function of a temporally homogeneous Markov process.

Suppose that  $M_T$ ,  $T$  being either the nonnegative integers or the nonnegative real numbers, is a temporally homogeneous Markov process on a measurable space  $(S, \Sigma)$  with initial distribution  $P(\cdot)$  and transition probability function  $P_t(\cdot, \cdot)$ . Let  $f$  be a mapping of  $T \times S$  into the real numbers which is square integrable with respect to the measure  $P_t(a, \cdot)$  for each  $t \in T$  and  $a \in S$ .

Suppose now that  $X_T$  is a real Gaussian process with zero expectations and covariance

$$\Gamma_{st} = \int_S P(du) \int_S f(s, v) P_s(u, dv) \int_S f(t - s, w) P_{t-s}(v, dw)$$

identical to the covariance of  $f(t, M_t)$ ,  $t \in T$ . Let  $X_T^*(\Sigma)$  be the generalized Gaussian random field (see [2]) on  $T \times \Sigma$  with zero expectations and covariance function

$$\Gamma_{st}^*(U, V) = \int_S P(du) \int_U P_s(u, dv) \int_V P_{t-s}(v, dw)$$

identical to the covariance of the random field

$$I_U(M_t), \quad t \in T, \quad U \in \Sigma$$

where  $I_U$  is the indicator function of  $U$ . Then if  $F(\Sigma)$  is the set of all functions mapping  $\Sigma$  into the real numbers we have the following

**THEOREM.** *Under the above conditions  $X_T^*$  is a Markov process on  $F(\Sigma)$  and*

$$X_T = \int_S f(t, u) X_t^*(du)$$

where the integral is the Wiener-Ito stochastic integral.

**Proof.** Since conditioning and projections are the same for a Gaussian field (see [1] or [3]), it follows that  $X_T$  is a Markov process on  $F(\Sigma)$  if for each  $s$  and

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$t$  in  $T$  and  $V \in \Sigma$ , the projection  $E_2(X_{s+t}^*(V) | X_s^*(U), U \in \Sigma)$  of  $X_{s+t}^*$  on the closed linear span of the functions  $X_s^*(U), U \in \Sigma$  is equal to the projection  $E_2(X_{s+t}^*(V) | X_r^*(U), r \leq s, U \in \Sigma)$  of  $X_{s+t}^*$  on the closed linear span of the functions  $X_r^*(U), r \leq s$  and  $U \in \Sigma$ . To do this we show that for each  $V \in \Sigma$ ,

$$\begin{aligned} E_2(X_{s+t}^*(V) | X_s^*(U), U \in \Sigma) &= \int_S P_t(u, V) X_s^*(du) \\ &= E_2(X_{s+t}^*(V) | X_r^*(U), r \leq s, U \in \Sigma) \end{aligned}$$

That is, we show that for each  $r \leq s, U \in \Sigma$  and  $V \in \Sigma$ ,

$$X_r^*(U) \perp X_{s+t}^*(V) - \int_S P_t(w, V) X_s^*(dw).$$

But

$$\begin{aligned} EX_r^*(U) \left[ X_{s+t}^*(V) - \int_S P_t(w, V) X_s^*(dw) \right] &= EX_r^*(U) X_{s+t}^*(V) - \int_S P_t(w, V) EX_r^*(U) X_s^*(dw) \\ &= \int_S P(du) \int_U P_r(u, dv) \int_V P_{s+t-r}(v, dw) - \int_S P_t(w, V) \\ &\quad \times \int_S P(du) \int_U P_r(u, dv) P_{s-r}(v, dw) \\ &= \int_S P(du) \int_U P_r(u, dv) P_{s+t-r}(v, V) - \int_S P(du) \int_U P_r(u, dv) \\ &\quad \times \int_S P_{s-r}(v, dw) P_t(w, V) = \int_S P(du) \int_U P_r(u, dv) P_{s+t-r}(v, V) \\ &\quad - \int_S P(du) \int_U P_r(u, dv) P_{s+t-r}(v, V) = 0 \end{aligned}$$

and so  $X_T^*$  is a Markov process on  $F(\Sigma)$ .

To complete the proof, we need only show that

$$X_t = \int_S f(t, u) X_t^*(du), \quad t \in T.$$

Clearly both  $X_t$  and  $\int_S f(t, u) X_t^*(du)$  are Gaussian with zero expectation. To show that they are equal in distribution we need only show that their covariances are

equal. Since the covariance of  $X_t$  is  $\Gamma_s^t$  and since

$$\begin{aligned} E \int_S f(s, v) X_s^*(dv) \int_S f(t, w) X_t^*(dw) \\ &= \int_S \int_S f(s, v) f(t, w) E X_s^*(dv) X_t^*(dw) \\ &= \int_S \int_S f(s, v) f(t, w) \int_S P_s(du) P(u, dv) P_{t-s}(v, dw) \\ &= \int_S P(du) \int_S f(s, v) P_s(u, dv) \int_S f(t, w) P_{t-s}(v, dw) = \Gamma_{st} \end{aligned}$$

the theorem is proved.

#### REFERENCES

1. J. L. Doob, *Stochastic Processes*, Wiley, New York, 1953.
2. I. M. Gel'fand and N. Ja. Vilenkin, *Generalized functions*. Vol. 4; Some applications of harmonic analysis, Fizmatgiz, Moscow, 1961; English transl., Academic Press, New York, 1964.
3. M. Loeve, *Probability Theory*, 3rd ed. Van Nostrand, Princeton, New Jersey, 1963.

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