

PREDICTION THEORY FOR NON-STATIONARY GENERALIZED STOCHASTIC PROCESSES

BY JOSÉ L. ABREU

Let $D(\mathbf{R})$ denote the space of infinitely differentiable complex valued functions on the real line \mathbf{R} with compact support, equipped with the topology of L. Schwartz [1]. In this paper we give some results concerning the linear least squares prediction theory of generalized stochastic processes which are actually continuous linear maps from $D(\mathbf{R})$ into some Hilbert space H (see [2, 3 and 4] for background).

An m -dimensional generalized random field $X = (X_1, \dots, X_m)$ is a continuous linear map $X: D(\mathbf{R}) \rightarrow H^m$ where H is a Hilbert space and H^m denotes the cartesian product of m copies of H . X_1, \dots, X_m are the components of X and $X_k: D(\mathbf{R}) \rightarrow H$ is a generalized stochastic process for each $k = 1, \dots, m$. Let $\langle \cdot, \cdot \rangle$ denote the inner product and $\| \cdot \|$ the norm of H . We define the covariance matrix of X as $B = (B_{ij})$ where $B_{ij}(f \otimes \bar{g}) = \langle X_{ij}f, X_{ij}g \rangle$ for $i, j = 1, \dots, m; f, g \in D(\mathbf{R})$. (Here \otimes denotes tensor product (see [5]) and $\bar{\cdot}$ denotes complex conjugation). The existence of the covariance distributions $B_{ij} \in D'(\mathbf{R}^2)$ is guaranteed by the theorem of kernels of L. Schwartz [1, 3].

For every $t \in \mathbf{R}$, let D_t be the subspace of $D(\mathbf{R})$ consisting of those functions with support contained in $(-\infty, t]$. Let $H_t =$ closed linear span $\{X_k f: f \in D_t; k = 1, \dots, m\}$ for each $t \in \mathbf{R}$, and also let $H_\infty =$ closed linear span $\{X_k f: f \in D(\mathbf{R}); k = 1, \dots, m\}$.

For $t \in \mathbf{R}$ and also for $t = \infty$, let $P_t: H \rightarrow H_t$ denote the orthogonal projections. X is said to be deterministic if $H_t = H_\infty$ for every $t \in \mathbf{R}$; X is called linearly free if $\bigcap_{t \in \mathbf{R}} H_t = H_\infty = \{0\}$. (The characterization of deterministic and linearly free generalized random fields on \mathbf{R} in terms of their covariance matrices has been studied by Deo [4].)

The prediction problem for X consists in expressing the projections $P_t X_k f$ in terms of linear combinations of $X_j g$ with $j = 1, \dots, m$ and $g \in D_t$.

THEOREM 1. *Let $X = (X_1, \dots, X_m)$ be a generalized random field on \mathbf{R} with covariance matrix $B = (B_{ij})$. Then for every t in \mathbf{R} there exist functions f_{ij} in D_t , $j = 1, \dots, m; i = 1, 2, \dots$ such that for each $n = 1, \dots, m$,*

$$(1) \quad P_t X_n f = \sum_{i=1}^{\infty} \sum_{j=1}^m \sum_{k=1}^m B_{nj}(f \otimes \bar{f}_{ij}) X_k(f_{ij})$$

for every $f \in D(\mathbf{R})$.

Proof. First note that for each $t \in \mathbf{R}$,

$$H_t = \text{c.l.s.} \{ \sum_{i=1}^m X_i f_i: f_i \in D_t, i = 1, \dots, m \}.$$

Now let D^m be the cartesian product of m copies of $D(\mathbf{R})$, and for each t in \mathbf{R} , let D_t^m be the cartesian product of m copies of D_t . We introduce a continuous

linear map $\tilde{X} = D^m \rightarrow H$ given by $\tilde{X}(f) = \sum_{n=1}^m X_n f_n$ for each $f = (f_1, \dots, f_m) \in D^m$. We also introduce an inner product C on D^m given by $C(f, g) = \langle \tilde{X}f, \tilde{X}g \rangle$ for each f, g in D^m . Define $\|f\|_C = C(f, f)^{1/2}$, and let $\ker C = \{f \in D^m : \|f\|_C = 0\}$. Let K be the Hilbert space obtained by completing the quotient space $D^m/\ker C$ with respect to the norm $\|\cdot\|_C$. Finally define $V:D^m \rightarrow K$ to be the canonical map and let $K(t)$ be the closure of $V(D_t^m)$ in K .

It is possible to find f_1, f_2, \dots in D_t^m such that $V(f_1), V(f_2), \dots$ is an orthonormal basis for $K(t)$. Then it is clear that $\tilde{X}(f_1), \tilde{X}(f_2), \dots$ is an orthonormal basis for H_t and therefore

$$(2) \quad P_t \tilde{X}(f) = \sum_{i=1}^{\infty} C(f, f_i) \tilde{X}(f_i)$$

for every f in D^m . Now choosing $f = (f_1, \dots, f_m)$ such that $f_k = 0$ if $k \neq n$ and $f_n = f$, we obtain equation (1) with $f_i = (f_{i1}, \dots, f_{im})$ for $i = 1, 2, \dots$ Q.E.D.

Theorem 1 provides an explicit solution to the prediction problem of X in terms of a sequence of values of the past of X .

The generalized random field X is stationary if its covariance matrix $B = (B_{ij})$ satisfies: $B_{ij}(f \otimes \bar{g}) = B_{ij}(T_s f \otimes \overline{T_s g})$ for every s in \mathbf{R} ; $i, j = 1, \dots, m$ and f, g in $D(\mathbf{R})$. Here T_s denotes the translation operator in $D(\mathbf{R})$, i.e. $T_s f(t) = f(t - s)$ for $f \in D(\mathbf{R})$; $s, t \in \mathbf{R}$.

COROLLARY 1. *If $X = (X_1, \dots, X_m)$ is a stationary generalized random field on \mathbf{R} with covariance matrix $B = (B_{ij})$ then there exist functions f_{ij} in D_0 for $i = 1, 2, \dots; j = 1, \dots, m$ such that for every $t \in \mathbf{R}, f \in D(\mathbf{R}), n = 1, \dots, m$,*

$$(3) \quad P_t X_n(f) = \sum_{i=1}^{\infty} \sum_{j=1}^m \sum_{k=1}^m B_{nj}(f \otimes \overline{T_t f_{ij}}) X_k(T_t f_{ij})$$

Proof. It is easy to see that if $\{Vf_i\}$ is an orthonormal basis for $K(0)$, then $\{V(T_t f_i)\}$ is an orthonormal basis for $K(t)$. Therefore putting $f_i = (f_{i1}, \dots, f_{im})$ for every $i = 1, 2, \dots$ for some sequence $\{f_i\} \subset D_0$ such that $\{Vf_i\}$ is an orthonormal basis for $K(0)$, we obtain (3). Q.E.D.

If $m = 1$, theorem 1 and corollary 1 give the solution to the prediction problem of generalized stochastic processes. The stationary case was treated in a different way by Rozanov [6].

Now we give some results which characterize deterministic covariance distributions. (The covariance matrix of a generalized stochastic process consists of a single distribution $B \in D'(\mathbf{R}^2)$ which is called the covariance distribution of the process).

It is not difficult to see that every covariance distribution B has representations of the form $B = \sum_{n=1}^{\infty} \bar{T}_n \otimes T_n$ with $T_n \in D'(\mathbf{R})$ for $n = 1, 2, \dots$; $\bar{T}(f) = \overline{T(\bar{f})}$ for every $f \in D(\mathbf{R}), T \in D'(\mathbf{R})$; and $T_n(f) = C(f, g_n)$ for every $f \in D(\mathbf{R})$, where $\{g_n\}$ is an orthonormal basis for the Hilbert space K obtained by completion of $D(\mathbf{R})/\ker C$ with respect to the norm $\|\cdot\|_C$. (C and $\|\cdot\|_C$ were defined in the proof of Theorem 1). Such a representation is called an orthonormal representation of B .

THEOREM 2. *If $B = \sum_{n=1}^{\infty} \bar{T}_n \otimes T_n$ is an orthonormal representation of the covariance distribution B , then B is deterministic if and only if whenever c_n are complex numbers such that $0 < \sum_{n=1}^{\infty} |c_n|^2 < \infty$, the support of the distribution $\sum_{n=1}^{\infty} c_n T_n$ extends to $-\infty$, i.e. it is not contained in (t, ∞) for any real number t .*

Proof. For each $t \in \mathbf{R}$ let $K(t)$ be the closure in K of $V(D_t)$ where $V: D(\mathbf{R}) \rightarrow K$ is the natural map. Let $W: K \rightarrow \ell^2$ be the map given by $W(f) = \{\bar{T}_n f\}$. W is a unitary isomorphism. The map $W \circ V: D(\mathbf{R}) \rightarrow \ell^2$ is a generalized stochastic process with covariance distribution B . Hence B is deterministic if and only if $W(K(t)) = \ell^2$ for every real t ; i.e. B is deterministic if and only if for every non-zero sequence $\{c_n\} \in \ell^2$, $\{c_n\}$ is not orthogonal to $W(K(t))$. Since $V(D_t)$ is dense in $K(t)$, we conclude that B is deterministic if and only if whenever $\{c_n\}$ is a non-zero element of ℓ^2 , and t is a real number, there exists $f \in D_t$ such that $\sum_{n=1}^{\infty} c_n T_n f \neq 0$. Q.E.D.

PROPOSITION 1. *A necessary and sufficient condition for a representation $B = \sum_{n=1}^{\infty} \bar{T}_n \otimes T_n$ of a covariance distribution B to be an orthonormal representation is that for every sequence $\{c_n\} \in \ell^2$, $\sum_{n=1}^{\infty} c_n T_n = 0$ necessarily implies $c_n = 0$ for $n = 1, 2, \dots$ (i.e. the T_n 's are Hilbert-free).*

Proof. The necessity is obvious. To prove the sufficiency suppose the T_n 's are Hilbert-free. There exists a sequence $\{g_n\} \subset K$ such that $\bar{T}_n f = C(Vf, g_n)$ for $n = 1, 2, \dots$; $f \in D(\mathbf{R})$. Furthermore $\{\sum_{i=1}^m x_i g_i; x_i \in \mathbf{C}, i = 1, \dots, m; m = 1, 2, \dots\}$ is dense in K , because the g_n 's are Hilbert-free. Let ℓ_0^2 be the subspace of ℓ^2 consisting of those sequences which have only finitely many non-zero elements. The natural map $U: \ell_0^2 \rightarrow K$ given by $U\{x_i\} = \sum_{i=1}^{\infty} x_i g_i$ is weakly bounded. Indeed, if $\{x_i\} \in \ell_0^2$ and $\sum_{i=1}^{\infty} |x_i|^2 \leq 1$, then $|\sum_{i=1}^{\infty} x_i T_i f|^2 \leq \sum_{i=1}^{\infty} |x_i|^2 \sum_{i=1}^{\infty} |T_i f|^2 \leq B(f \otimes \bar{f})$ for every $f \in D(\mathbf{R})$. Thus the image of the unit ball of ℓ_0^2 under U is weakly bounded, and by Mackey's theorem (see theorem 3.5.3 of [5]) it is also strongly bounded. Therefore U may be continuously extended to ℓ^2 . Since the g_n 's are Hilbert-free, U is injective; and since the closed linear subspace generated by $\{g_n\}$ coincides with K , it follows that U is onto. Now given $f \in D(\mathbf{R})$, $\{\bar{T}_i f\}$ belongs to ℓ^2 and for every $g \in D(\mathbf{R})$ we have

$$\begin{aligned} C(U\{\bar{T}_i f\}, Vg) &= C(\sum_{i=1}^{\infty} (\bar{T}_i f) g_i, Vg) = \sum_{i=1}^{\infty} \bar{T}_i f C(g_i, Vg) = \sum_{i=1}^{\infty} \bar{T}_i f T_i \bar{g} \\ &= B(f \otimes \bar{g}) = C(Vf, Vg) \end{aligned}$$

Hence $U\{\bar{T}_i f\} = Vf$ for every $f \in D(\mathbf{R})$. Therefore for $f, g \in D(\mathbf{R})$

$$C(U\{\bar{T}_i f\}, U\{\bar{T}_i g\}) = C(Vf, Vg) = \sum_{i=1}^{\infty} \bar{T}_i f T_i \bar{g} = \sum_{i=1}^{\infty} \bar{T}_i f (\bar{T}_i g)$$

Since $V(D(\mathbf{R}))$ is dense in K we conclude that for every $\{x_i\}, \{y_i\} \in \ell^2$, $C(U\{x_i\}, U\{y_i\}) = \sum_{i=1}^{\infty} x_i \bar{y}_i$.

We just proved that U is a unitary isomorphism from ℓ^2 onto K , and $\{g_n\}$ is the image under U of the canonical orthonormal basis in ℓ^2 . Hence $\{g_n\}$ is an orthonormal basis for K . Q.E.D.

We say that a covariance distribution B is degenerate if the corresponding space K defined above is finite dimensional. It follows that B is degenerate if and only if $B = \sum_{n=1}^N \bar{T}_n \otimes T_n$ for some integer N and $T_n \in D'(\mathbf{R})$ for $n = 1, 2, \dots, N$.

COROLLARY 2. $B = \sum_{n=1}^N \bar{T}_n \otimes T_n$ is an orthonormal representation of the degenerate covariance distribution B if and only if the T_n 's are linearly independent. If $B = \sum_{n=1}^N \bar{T}_n \otimes T_n$ is an orthonormal representation of B , then B is deterministic if and only if the restrictions of the T_n 's to D_t are linearly independent for any real number t .

The proof follows easily from theorem 2 and proposition 1.

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UNIVERSIDAD NACIONAL AUTÓNOMA DE MÉXICO

REFERENCES

- [1] L. SCHWARTZ, *Théorie des Distributions*, Hermann, Paris, 1966.
- [2] K. ITO, *Stationary random distributions*, Mem. of the College of Science, University of KYOTO, Ser. A, 28, No. 3 (1953), 209-23.
- [3] I. M. GELFAND AND N. YA. VILENKIN, *Generalized Functions*, Vol. 4, Applications of Harmonic Analysis, Academic Press, New York and London, 1964.
- [4] CH. M. DEO, *Prediction theory for non-stationary processes*, Sankyā: The Indian Journal of Statistics, 27 (1965) 113-32.
- [5] J. HORVÁTH, *Topological Vector Spaces and Distributions*, Vol. 1, Addison-Wesley, Reading, Mass., 1966.
- [6] YU. A. ROZANOV, *On the extrapolation of generalized stationary random processes*. Theor. Probability Appl., 4 (1959)—(Translated from Russian), 426-31.