

# On a Paper of Ramesh Gangolli

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**Introduction.** In this note we construct an infinite vector-valued weakly stationary stochastic process such that

- (i) it has an absolutely continuous spectrum,
- (ii) it has a nontrivial deterministic part, and
- (iii) its prediction error matrix is non-singular.

A special case of our example will show that certain results of [4] are in error. We shall state correct versions of these results. In [3], R. G. Douglas constructed a counter example showing that a certain result in [5] was in error. Our counter example was suggested after looking at his paper, but our method of construction differs from his.

**Section 1.** Let  $\mathcal{H}$  be a Hilbert space which we take to be  $L_2$  of some probability space  $(\Omega, \mathfrak{B}, P)$ . We assume that  $\mathcal{H}$  is infinite-dimensional. We shall need the following:

**Lemma.** Let  $\varphi, \varphi_1, \varphi_2, \dots$  be a set of orthonormal vectors in  $\mathcal{H}$ . Then  $\varphi$  lies in the subspace spanned by  $\{\varphi + \varphi_1, \varphi + \varphi_2, \varphi + \varphi_3, \dots\}$ .

*Proof.*

$$\left\| \varphi - \frac{1}{n} \sum_{i=1}^n (\varphi + \varphi_i) \right\|^2 = \frac{1}{n} \rightarrow 0 \quad \text{as } n \rightarrow \infty. \quad \text{Q.E.D.}$$

Now we construct our example having properties (i), (ii) and (iii) stated in the introduction. Let  $H, H_0, H_{-1}, H_1, H_{-2}, H_2, \dots$  be infinite-dimensional mutually orthogonal subspaces of  $\mathcal{H}$ . In each  $H_k$  choose a one-dimensional purely non-deterministic weakly stationary stochastic process  $(Y_n^k), -\infty < n < \infty$ . Further for each  $k$  choose  $\|Y_0^k\| = 1$ . Let  $f_k$  be the spectral density of  $(Y_n^k), -\infty < n < \infty$ . In  $H$  choose a one-dimensional weakly stationary stochastic process  $(Y_n), -\infty < n < \infty$ , with absolutely continuous spectrum. Let  $g$  be the spectral density of  $(Y_n), -\infty < n < \infty$ . Let  $(Y_n), -\infty < n < \infty$ , be so normalized that  $\|Y_n\| = 1$ . Consider the infinite vector  $\tilde{X}_n = [Y_n^k + Y_{k+n}], -\infty < k < \infty$ . Then  $(\tilde{X}_n), -\infty < n < \infty$ , is an infinite-dimensional weakly stationary stochastic process.

**Theorem.**

- (i)  $(\bar{X}_n)$ ,  $-\infty < n < \infty$ , has an absolutely continuous spectrum.  
(ii)  $(\bar{X}_n)$ ,  $-\infty < n < \infty$ , has a non-trivial deterministic part.  
(iii) The one step prediction error matrix of  $(\bar{X}_n)$ ,  $-\infty < n < \infty$ , is a diagonal matrix with each term of the diagonal non-zero.

*Proof.* (i) Due to the orthogonality of  $H$  with the  $H_k$ 's we have

$$\begin{aligned} (X_n^k, X_0^l) &= (Y_n^k + Y_{k+n}, Y_0^l + Y_l) = (Y_n^k, Y_0^l) + (Y_{n+k}, Y_l) \\ &= \frac{1}{2\pi} \int_0^{2\pi} e^{inu} \delta_{k,l} f_k(u) du + \frac{1}{2\pi} \int_0^{2\pi} e^{i(k+n-l)u} g(u) du \\ &= \frac{1}{2\pi} \int_0^{2\pi} e^{inu} f_{k,l}(u) du, \end{aligned}$$

where  $f_{k,l}(u) = \delta_{k,l} f_k(u) + e^{i(k-l)u} g(u)$ . Thus  $(\bar{X}_n)$ ,  $-\infty < n < \infty$ , is a process with an absolutely continuous spectrum and its spectral density matrix  $f(\cdot)$  has  $(k, l)$ th entry given by  $f_{k,l}(u) = \delta_{k,l} f_k(u) + e^{i(k-l)u} g(u)$ .

(ii) Let  $\mathfrak{M}_{-1}$  be the subspace spanned by  $\{X_n^k : -\infty < k < \infty, n \leq -1\}$ . We show that each  $Y_p \in \mathfrak{M}_{-1}$ ,  $-\infty < p < \infty$ . For each  $l > 0$ ,

$$X_{-l}^{p+l} = Y_{-l}^{p+l} + Y_{p+l-l} = Y_p + Y_{-l}^{p+l} \in \mathfrak{M}_{-1}.$$

Further  $Y_p, Y_{-1}^{p+1}, Y_{-2}^{p+2}, \dots$  are orthonormal vectors. Hence by our lemma,  $Y_p$  lies in the closed subspace spanned by  $\{X_{-l}^{p+l} : l > 0\} \subset \mathfrak{M}_{-1}$ . Thus  $Y_p \in \mathfrak{M}_{-1}$ . The same reasoning shows that for every  $p, Y_p \in \mathfrak{M}_n$  for every  $n$ . Hence  $(\bar{X}_n)$ ,  $-\infty < n < \infty$ , has a non-trivial deterministic part.

(iii) It is easy to see that the prediction error matrix of  $(\bar{X}_n)$ ,  $-\infty < n < \infty$ , is a diagonal matrix, with diagonal terms

$$\lambda_k = \exp \left( \frac{1}{2\pi} \int_0^{2\pi} \log f_k(u) du \right). \quad \text{Q.E.D.}$$

We now consider a special case.

**Example A.** In the above example choose  $(Y_n^k)$ ,  $-\infty < n < \infty$ , uncorrelated. Choose  $(Y_n)$ ,  $-\infty < n < \infty$ , also uncorrelated. Set  $\bar{X}_n = [Y_n^k + Y_{k+n}]$ ,  $-\infty < k < \infty$ . Now the spectral density matrix  $f(\cdot)$  of  $(\bar{X}_n)$ ,  $-\infty < n < \infty$ , has entries  $f_{k,l}(\cdot)$  given by  $f_{k,l}(u) = \delta_{k,l} + e^{i(k-l)u}$ . Clearly  $\sum_{k=-\infty}^{\infty} |f_{k,l}(u)|^2 = \infty$  for every  $u$ . Hence  $f(u)$  does not define any operator (bounded or unbounded) on  $l^2$  with values in  $l^2$ . (Here  $l^2$  is the Hilbert space of all bisequences  $(x_n)$ ,  $-\infty < n < \infty$ , such that  $\sum |x_n|^2 < \infty$ .) The prediction error matrix of this process is the identity matrix. It has a non-trivial deterministic part.

Example A contradicts certain results in [4], as we now show. Let  $(\bar{X}_n)$ ,  $-\infty < n < \infty$ , be as in this example. For each  $x \in l^2$ ,  $x = (x_k)$ ,  $-\infty < k < \infty$ , define

$$F_n(x) = \sum_{k=-\infty}^{\infty} x_k X_n^k.$$

Then  $F_n$ ,  $-\infty < n < \infty$ , is a weakly stationary sequence of distributions in the sense of [4]. According to considerations in paragraph two of page 903 of [4], the matrix  $f(\cdot)$  with entries given by  $f_{k,l}(u) = \delta_{k,l} + e^{i(k-l)u}$  defines a matrix of an operator (bounded or unbounded) on  $l^2$  with values in  $l^2$ . This is obviously false. Hence all one can say is that the entity  $W'$  of paragraph two on page 903 of [4] is a non-negative definite matrix with summable entries such that  $\int_0^{2\pi} W'(\theta) d\theta$  is a bounded operator.

Next, Example A shows that Theorem 5.3 of [4] is in error. This theorem asserts in effect that a full rank process (infinite-dimensional) with an absolutely continuous spectrum is purely non-deterministic. Our Example A contradicts this. Theorem 7.3 of [4] is also in doubt since the proof suggested for it relies on Theorem 5.3 of [4]. With the notation and terminology of [4] a correct version of Theorem 5.3 of [4] can be stated as follows:

*Suppose that  $(F_n)_{-\infty}^{\infty}$  has full rank. Let  $W = W_p + W_s$  be its spectral distribution function with  $W_p$  and  $W_s$  as in Theorem 5.2 of [4]. Let  $W_{abs}$  and  $W_{sing}$  be the absolutely continuous and singular parts of  $W$ . Assume that  $W'(\theta)$  is a bounded and invertible operator for almost every  $\theta$ . Then  $W_p = W_{abs}$ ,  $W_s = W_{sing}$ .*

We omit the proof of this theorem. It can be derived from a result on page 494 of [1], or it can be derived by extending method of Doob ([2], p. 597) where the analogous result is proved for  $n \times n$  matrix-valued spectral distribution functions.

**Remark.** We would like to mention that [6] contains results with the help of which the condition of invertibility and boundedness of  $W'$  can be relaxed. [6] also contains an analysis of the situation where the prediction error matrix is not a bounded invertible matrix.

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#### BIBLIOGRAPHY

- [1] DEVINATZ, ALLEN, The factorization of operator-valued functions, *Ann. Math.*, **73** (1961) 458-495.
- [2] DOOB, J. L., *Stochastic Processes*, John Wiley, New York, 1953.
- [3] DOUGLAS, R. G., On factoring positive operator functions, *Journal of Math. and Mech.*, **16** (1966) 119-126.
- [4] GANGOLLI, RAMESH, Wide-sense stationary sequence of distributions on Hilbert space and the factorization of operator-valued functions, *Journal of Math. and Mech.*, **12** (1963) 893-910.
- [5] LOWDENSLAGEN, DAVID, On factoring matrix valued functions, *Ann. Math.* **78** (1963) 450-454.
- [6] NADKARNI, M. G., *Vector-valued weakly stationary stochastic processes and factorability of matrix valued functions . . . .*, Ph.D. Thesis, Brown University, Providence, Rhode Island, 1965.

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