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# MATHEMATICS

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**Abstract**

**Full Text**

## MATHEMATICS

O. V. SARMANOV

# PROPER CORRELATION FUNCTIONS AND THEIR APPLICATIONS IN THE THEORY OF STATIONARY MARKOV PROCESSES

*(Presented by Academician A. N. Kolmogorov on 13 II 1960)*

1. Consider a stationary real process with continuous parameter  $\{x_t; -\infty < t < \infty\}$  and suppose that the random variables  $x_t$  have means equal to zero and variances equal to one. Denote  $x_{t_1}$  by  $x$ ,  $x_{t_1+t}$  by  $y$ ; by stationarity,  $x$  and  $y$  have identical a priori distributions.

**Definition 1.**  $r_1(t)$  is called the **maximal correlation function** of the process  $x_t$ , if for  $t \neq 0$

$$|r_1(t)| = \sup_{f,g} |\mathbf{M}(f(x)g(y))| \quad (1)$$

in the class of functions  $f, g$  satisfying the conditions

$$\mathbf{M}f(x) = \mathbf{M}g(y) = 0; \quad \mathbf{M}f^2(x) = \mathbf{M}g^2(y) = 1, \quad (2)$$

where  $\mathbf{M}$  denotes mathematical expectation. (For  $t = 0$ ,  $r_1(t)$  is completed by the condition  $r_1(0) = 1$ .)

**Remark.** If the functions  $f_1$  and  $g_1$  at which the upper bound (1) is attained are linear, then  $r_1(t)$  coincides with the ordinary correlation function  $r(t)$ , where

$$r(t) = \mathbf{M}(xy), \quad t \neq 0. \quad (3)$$

The properties of  $r_1(t)$  are analogous to the properties of the maximal correlation coefficient <sup>(1,2)</sup>.

2. We now considerably narrow the class of processes under consideration, assuming that the law of the joint distribution of  $x$  and  $y$  has density  $p(t, x, y)$ , that this density is symmetric,

$$p(t, x, y) = p(t, y, x) \quad (4)$$

in the region  $\Omega = [a \leq x; y \leq b]$  and for all  $t \neq 0$  satisfies the boundedness condition

$$\iint_{(\Omega)} \frac{p^2(t, x, y)}{p(x)p(y)} dx dy < \infty, \quad (5)$$

where

$$p(x) = \int_a^b p(t, x, y) dy, \quad p(y) = \int_a^b p(t, x, y) dx. \quad (6)$$

By virtue of (4) and (5), the upper bound (1) is attained on identical functions, i.e.  $g_1(y) = f_1(y)$ .

**Definition 2.** The sequence of eigenvalues  $r_k(t)$ ,  $k = 1, 2, \dots$ , of the kernel

$$\frac{p(t, x, y)}{\sqrt{p(x)p(y)}} \quad (7)$$

is called the sequence of intrinsic correlation functions of the process  $x_{t_1}$ ,  $t \neq 0$ .

**Definition 3.** If the eigenfunctions of the kernel (7)  $\{\varphi_k(x), \varphi_k(y)\}$  do not depend on  $t$ , then the process is called maximally stationary.

**Definition 4.** The process  $x_{t_1}$  is called  $C$ -continuous if all  $r_k(t)$  are continuous; in particular,

$$\lim_{t \rightarrow 0} r_k(t) = 1, \quad k = 1, 2, \dots \quad (8)$$

**3.** Suppose that  $p(t, x, y)$ , in addition to the symmetry condition (4) and the constraint (5), satisfies the Markov equation

$$p(t_1 + t_2, x, y) = \int_a^b \frac{p(t_1, x, z)p(t_2, z, y)}{p(z)} dz; \quad (9)$$

then this density completely determines a stationary Markov process.

A consequence of the definitions introduced and equation (9) is

**Theorem 1.** In order that a symmetric two-dimensional density  $p(t, x, y)$ , satisfying the constraint (5), be able to define a continuous Markov process, it is necessary and sufficient that the following conditions be fulfilled:

- a) the eigenfunctions of the kernel (7) do not depend on  $t$ , i.e. the process is maximally stationary;
- b) the intrinsic correlation functions have the form (for  $k = 1, 2, \dots$ )

$$r_k(t) = e^{-\lambda_k t}, \quad t \geq 0, \quad 0 < \lambda_1 \leq \lambda_2 \leq \dots \quad (10)$$

4. Suppose now that the limits exist

$$\begin{aligned} A(x) &= \lim_{t \rightarrow 0} \frac{1}{t} \int_a^b (y-x) \frac{p(t, x, y)}{p(x)} dy, \\ B(x) &= \lim_{t \rightarrow 0} \frac{1}{t} \int_a^b (y-x)^2 \frac{p(t, x, y)}{p(x)} dy \end{aligned} \quad (11)$$

and that the transition probability density

$$f(t, x, y) = \frac{p(t, x, y)}{p(x)} \quad (12)$$

satisfies the well-known equations of A. N. Kolmogorov <sup>(3)</sup>

$$\frac{\partial f}{\partial t} = A(x) \frac{\partial f}{\partial x} + \frac{1}{2} B(x) \frac{\partial^2 f}{\partial x^2}, \quad (I)$$

$$\frac{\partial f}{\partial t} = -\frac{\partial}{\partial y} [A(y)f] + \frac{1}{2} \frac{\partial^2}{\partial y^2} [B(y)f]. \quad (II)$$

On the other hand, according to Theorem 1, for  $f(t, x, y)$  one obtains the bilinear expansion

$$f(t, x, y) = p(y) \left[ 1 + \sum_{k=1}^{\infty} \varphi_k(x) \varphi_k(y) e^{-\lambda_k t} \right]. \quad (13)$$

Substituting (13) into (I) and (II) and using the orthogonality of  $\varphi_k(x)$  with weight  $p(x)$ , we obtain the ordinary differential equations

$$\frac{1}{2} B(x) \varphi_k''(x) + A(x) \varphi_k'(x) + \lambda_k \varphi_k(x) = 0, \quad k = 1, 2, \dots; \quad (14)$$

$$B(x) p'(x) + [B'(x) - 2A(x)] p(x) = 0. \quad (15)$$

From (15) we find an explicit expression for  $p(x)$ :

$$p(x) = \frac{\gamma}{B(x)} \exp \left[ \int \frac{2A(x)}{B(x)} dx \right], \quad (16)$$

where  $\gamma$  is a normalizing constant. (In particular, the process can be Gaussian only if  $A(x)$  is linear and  $B(x) = \text{const.}$ )

5. Suppose, for example, that  $\varphi_k(x)$  is a polynomial of degree  $k$ ,  $k = 1, 2, \dots$ , and that the domain  $\Omega$  is infinite. In this case, using (13) and the orthogonality of the eigenfunctions, we immediately obtain

$$\begin{aligned} A(x) &= -\lambda_1 x, \\ B(x) &= (2\lambda_1 - \lambda_2)x^2 + c(\lambda_2 - \lambda_1)x + \lambda_2, \end{aligned} \quad (17)$$

where

$$c = Mx^3 \quad (18)$$

is the coefficient of asymmetry of the a priori distribution of  $x_{t_1}$  (since  $Mx_{t_1} = 0$ , and  $Mx_{t_1}^2 = 1$ ).

Since  $B(x) \geq 0$  on an infinite domain, the condition

$$\lambda_2 \leq 2\lambda_1 \quad (19)$$

is necessary.

On the other hand, substituting into (14), instead of  $\varphi_k(x)$ , a polynomial of degree  $k$  with undetermined coefficients, and instead of  $A(x)$  and  $B(x)$  their expressions (17), we find

$$\lambda_k = k \left[ \frac{k-1}{2} \lambda_2 - (k-2) \lambda_1 \right]. \quad (20)$$

Expression (20), when  $\lambda_2 < 2\lambda_1$ , becomes negative for sufficiently large  $k$ ; consequently, in inequality (19) only the equality sign is possible, and then in general

$$\lambda_k = k\lambda_1, \quad k = 1, 2, \dots, \quad (21)$$

and equation (14) is reduced to the form

$$\left( \frac{c}{2}x + 1 \right) \varphi_k''(x) - x\varphi_k'(x) + k\varphi_k(x) = 0, \quad k = 1, 2, \dots, \quad (22)$$

where

$$p(x) = \frac{\gamma_1}{cx+2} \exp \left[ - \int \frac{2x}{cx+2} dx \right], \quad cx+2 \geq 0, \quad c \geq 0, \quad \gamma_1 = \text{const.} \quad (23)$$

6. Let  $c = 0$ ; then  $p(x) = \frac{1}{\sqrt{2\pi}} e^{-x^2/2}$ , and  $\varphi_k(x)$  are Hermite polynomials, and the process is Gaussian, while for  $t > 0$

$$p(t, x, y) = \frac{1}{2\pi\sqrt{1-e^{-2\lambda_1 t}}} \exp \left[ - \frac{x^2 + y^2 - 2e^{-\lambda_1 t} xy}{2(1-e^{-2\lambda_1 t})} \right]. \quad (24)$$

is the density of a normal correlation with positive correlation coefficient  $R = e^{-\lambda_1 t}$ .

7. For  $c > 0$ , equation (22) defines generalized Laguerre polynomials, orthogonal on the interval  $-\frac{2}{c} \leq x < \infty$  with weight

$$p(x) = \gamma_2 e^{-\frac{2}{c}x} \left( x + \frac{2}{c} \right)^{\frac{4}{c^2}-1}, \quad \gamma_2 = \text{const.} \quad (25)$$

The corresponding density of the joint probability distribution has been studied in detail in paper (4); we note here only that the characteristic function of this distribution has the form

$$\varphi_c(\tau_1, \tau_2) = \frac{\exp \left[ - \frac{2i}{c} (\tau_1 + \tau_2) \right]}{\left[ 1 - \frac{c}{2} i (\tau_1 + \tau_2) - \frac{c^2}{4} (1 - e^{-\lambda_1 t}) \tau_1 \tau_2 \right]^{\frac{4}{c^2}}}, \quad (26)$$

and as  $c \rightarrow 0$  it converges to the characteristic function of the normal distribution (24).

**Remark.** If  $c < 0$ , then the corresponding correlation dependence between  $x$  and  $y$  is specified in the quadrant  $-\infty < x$ ;  $y \leq \frac{2}{c}$ , and a simultaneous change of signs of  $x$  and  $y$  reduces this case to the one already considered.

8. In items 5-7 the following assertion has been proved.

**Theorem 2.** *The normal correlation (24) and the correlation with characteristic function (26), constructed from generalized Laguerre polynomials, exhaust the class of densities determining a continuous stationary Markov process and such that the eigenfunctions of the kernel (7) constitute a complete system of polynomials orthogonal on an infinite interval.*

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- <sup>4</sup> O. V. Sarmanov, DAN, **132**, No. 2 (1960).

*Note: Figure translations are in progress. See original paper for figures.*

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