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# GENERALIZED NORMAL CORRELATION

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**Abstract**

**Full Text**

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*MATHEMATICS*

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## GENERALIZED NORMAL CORRELATION AND TWO-DIMENSIONAL FRÉCHET CLASSES

*(Presented by Academician S. N. Bernstein on 17 VIII 1965)*

1. Consider the class of functions  $N\{f(x, y)\}$  representable by bilinear expansions in Hermite polynomials

$$f(x, y) = \frac{\exp\left[-\frac{x^2+y^2}{2}\right]}{2\pi} \left[ 1 + \sum_{k=1}^{\infty} c_k H_k(x) H_k(y) \right]. \quad (1)$$

The series (1) is assumed to converge in the mean, which is equivalent to the condition

$$\sum_{k=1}^{\infty} c_k^2 < \infty, \quad (2)$$

where  $H_k(x)$  are orthogonal and normal with weight  $e^{-x^2/2}/\sqrt{2\pi}$  on the whole real axis, i.e.

$$H_k(x) = \frac{(-1)^k}{\sqrt{k!}} e^{x^2/2} \frac{d^k}{dx^k} (e^{-x^2/2}) = \frac{1}{\sqrt{k!}} \left[ x^k - \frac{k(k-1)}{2} x^{k-2} + \dots \right]. \quad (3)$$

It follows from (1) that

$$\int_{-\infty}^{\infty} f(x, y) dx = \frac{e^{-y^2/2}}{\sqrt{2\pi}}; \quad \int_{-\infty}^{\infty} f(x, y) dy = \frac{e^{-x^2/2}}{\sqrt{2\pi}}. \quad (4)$$

**Definition.** When the sum of the series (1) is nonnegative for all real  $x$  and  $y$ , we shall say that there is a generalized normal correlation between  $x$  and  $y$ .

**2. Theorem.** *In order that the sum of the series (1) be the density of a generalized normal correlation, it is necessary and sufficient that the coefficients  $c_k$*

be the moments of some probability distribution concentrated inside the interval  $[-1, 1]$ .

**Necessity.** Let the sum of the series (1) be nonnegative for all  $x$  and  $y$ . Find the conditional mathematical expectation of the quantity  $(y/x)^k$  for fixed  $x \neq 0$ . Express  $y^k$  through the polynomials (3):

$$y^k = \sqrt{k!}H_k(y) + a_{k-2}H_{k-2}(y) + a_{k-4}H_{k-4}(y) + \dots,$$

where  $a_{k-2}, a_{k-4}, \dots$  are fully determined constants.

$$\begin{aligned} M_x y^k &= \sqrt{k!}c_k H_k(x) + a_{k-2}c_{k-2}H_{k-2}(x) + \dots \\ &= c_k x^k + b_{k-2}x^{k-2} + b_{k-4}x^{k-4} + \dots, \end{aligned}$$

where  $b_{k-2}, b_{k-4}, \dots$  are also certain constants. Thus,

$$M_x (y/x)^k = c_k + b_{k-2}/x^2 + b_{k-4}/x^4 + \dots. \quad (5)$$

and for sufficiently large  $x$ ,  $c_k$  differs arbitrarily little from the  $k$ -th moment of the variable  $y/x$ ; moreover, by virtue of (2),  $c_k$  cannot be a moment of a variable that, with positive probability, goes outside the interval  $[-1, 1]$  or assumes the values  $\pm 1$ .

**Sufficiency.** Let  $\{c_k\}$  form the sequence of moments of some variable  $\xi$ , concentrated inside the interval  $[-1, 1]$  and given by the distribution function  $G(\xi)$ ; then for each fixed  $|\xi| < 1$  consider the well-known expansion for the density of normal correlation with correlation coefficient  $\xi$

$$\frac{\exp\left[-\frac{x^2+y^2-2\xi xy}{2(1-\xi^2)}\right]}{2\pi\sqrt{1-\xi^2}} = \frac{\exp\left[-\frac{x^2+y^2}{2}\right]}{2\pi} \left[1 + \sum_{k=1}^{\infty} \xi^k H_k(x)H_k(y)\right], \quad (6)$$

where, by assumption,

$$\int_{-1}^1 \xi^k dG(\xi) = c_k, \quad k = 1, 2, \dots \quad (7)$$

From (6) and (7) it follows that

$$\frac{\exp\left[-\frac{x^2+y^2}{2}\right]}{2\pi} \left[1 + \sum_{k=1}^{\infty} c_k H_k(x)H_k(y)\right] =$$

$$= \frac{1}{2\pi} \int_{-1}^1 \frac{\exp\left[-\frac{x^2+y^2-2\xi xy}{2(1-\xi^2)}\right]}{\sqrt{1-\xi^2}} dG(\xi) > 0, \quad (8)$$

which completes the proof.

**Remark.** In normal correlation,  $\xi = R$  with probability 1, and  $c_k = R^k$ . If  $\xi$  assumes the two values  $\pm R$  with probabilities  $1/2$ , so that  $c_{2k} = R^{2k}$ ,  $c_{2k-1} = 0$ ,  $k = 1, 2, \dots$ , the corresponding density was considered in <sup>(1)</sup> and was given the name pseudonormal correlation; finally, if all  $c_k = 0$ , then  $x$  and  $y$  are independent.

**Corollary 1.** If the sum (1) has constant sign and the variables  $x$  and  $y$  are dependent, then  $c_{2k} > 0$  for all  $k = 1, 2, \dots$

**Corollary 2.** It is easy to verify that the characteristic function of the density (8) has the form

$$\varphi(t_1, t_2) = \exp\left[-\frac{t_1^2 + t_2^2}{2}\right] \int_{-1}^1 \exp[-Rt_1 t_2] dG(R). \quad (9)$$

In particular, if it is known that the correlation between  $x$  and  $y$  is normal, but the correlation coefficient is not known exactly and is uniformly distributed on the interval  $[R - \varepsilon, R + \varepsilon]$ ,  $R - \varepsilon > -1$ ,  $R + \varepsilon < 1$ , then the characteristic function of such a generalized normal correlation is expressed with the aid of (9) as follows:

$$\varphi(t_1, t_2, \varepsilon) = \exp\left[-\frac{t_1^2 + t_2^2 + 2Rt_1 t_2}{2}\right] \frac{\text{sh}(\varepsilon t_1 t_2)}{\varepsilon t_1 t_2}. \quad (10)$$

2. M. Fréchet in <sup>(2, 3)</sup> posed the following general problem: to find the class of two-dimensional distribution functions  $\{P(x, y)\}$  having given partial (extreme or marginal) one-dimensional distributions  $F(x)$  and  $F_1(y)$ . In the preceding section an important part of the normal Fréchet class has been found.

Morgenstern and Gumbel <sup>(4)</sup> considered the following example of functions belonging to the Fréchet class  $K\{F(x), F_1(y)\}$  for arbitrary given  $F(x)$  and  $F_1(y)$ :

$$P(x, y) = F(x)F_1(y) [1 + \lambda(1 - F(x))(1 - F_1(y))], \quad -1 < \lambda < 1. \quad (11)$$

If, however, densities exist (denoted by the corresponding lowercase letters), then

$$p(x, y) = f(x)f_1(y) [1 + \lambda(2F(x) - 1)(2F_1(y) - 1)]; \quad (12)$$

in particular, in the normal case,

$$p(x, y) = \frac{\exp\left[-\frac{x^2+y^2}{2}\right]}{2\pi} [1 + 4\lambda\Phi(x)\Phi(y)], \quad (13)$$

where

$$\Phi(x) = \frac{1}{\sqrt{2\pi}} \int_0^x e^{-t^2/2} dt.$$

**3.** Let us consider the following more general method of constructing functions from the Fréchet class  $k\{f(x), f_1(y)\}$ , specified by the densities  $f(x)$  and  $f_1(y)$  on  $L = [a \leq x \leq b]$  and  $L_1 = [a_1 \leq y \leq b_1]$ . Let  $\varphi(x)$  be an arbitrary function (not equal to a constant) bounded on  $L$ , and let  $\varphi_1(y)$  be an arbitrary function bounded on  $L_1$ . Then the quantities

$$\begin{aligned} m &= \int_a^b \varphi(x)f(x) dx, & m_1 &= \int_{a_1}^{b_1} \varphi_1(y)f_1(y) dy, \\ \sigma^2 &= \int_a^b [\varphi(x) - m]^2 f(x) dx, & \sigma_1^2 &= \int_{a_1}^{b_1} [\varphi_1(y) - m_1]^2 f_1(y) dy \end{aligned} \quad (14)$$

are meaningful.

Denote

$$h = \sup_{x \in L} \frac{|\varphi(x) - m|}{\sigma}, \quad h_1 = \sup_{y \in L_1} \frac{|\varphi_1(y) - m_1|}{\sigma_1}. \quad (15)$$

The one-parameter family of densities

$$p_1(x, y, \lambda) = f(x)f_1(y) \left[ 1 + \frac{\lambda}{hh_1} \frac{[\varphi(x) - m][\varphi_1(y) - m_1]}{\sigma\sigma_1} \right], \quad (16)$$

where  $-1 \leq \lambda \leq 1$ , belongs to the Fréchet class  $k\{f(x), f_1(y)\}$ , since  $\varphi(x) - m$  and  $\varphi_1(y) - m_1$  are orthogonal to unity with weights  $f(x)$  and  $f_1(y)$ , and the sum (16) is nonnegative.

We note that, by virtue of (14),  $h \geq 1$  and  $h_1 \geq 1$ , and

$$\rho = \lambda/hh_1 \quad (17)$$

is the maximal correlation coefficient (5) between  $x$  and  $y$ .

Integrating (16), we pass to the distribution functions

$$P_1(x, y, \lambda) = F(x)F_1(y) + \frac{\lambda}{hh_1} \int_a^x \frac{\varphi(u) - m}{\sigma} dF(u) \int_{a_1}^y \frac{\varphi_1(v) - m_1}{\sigma_1} dF_1(v). \quad (18)$$

Since  $F(x)$  and  $F_1(y)$  are bounded functions, one may put  $\varphi(x) \equiv F(x)$ , and  $\varphi_1(y) \equiv F_1(y)$ ; then, for continuous  $F(x)$  and  $F_1(y)$ , we obtain that

$$\frac{1}{2} = \int_a^b F(x) dF(x) = m = m_1 = \sup_{x \in L} |F(x) - m| = \sup_{y \in L_1} |F_1(y) - m_1|,$$

$$\begin{aligned} P_1(x, y, \lambda) &= F(x)F_1(y) + \lambda(F^2(x) - F(x))(F_1^2(y) - F_1(y)) = \\ &= F(x)F_1(y)[1 + \lambda(F(x) - 1)(F_1(y) - 1)], \end{aligned} \quad (19)$$

as in example (11). The assumption of continuity is not a restriction, for, if it is proved that (19) is a distribution function for

for arbitrary discontinuous  $F(x)$  and  $F_1(y)$ , then (19) gives a distribution function also for arbitrary distribution functions  $F(x)$  and  $F_1(y)$ .

4. If the densities  $p_k(x, y)$  belong to some Fréchet class, then a linear combination of them of the form  $\sum_k \varepsilon_k p_k(x, y)$ , where  $\varepsilon_k > 0$  and  $\sum_k \varepsilon_k = 1$ , also belongs to the same class. With the help of this remark and (16), one can construct new examples of densities and distribution functions belonging to a given class (the possibility of similar bilinear constructions was indicated in <sup>6</sup>).

Let us give several examples

$$p(x, y) = f(x)f_1(y) \left[ 1 + \sum_{k=1}^{\infty} \varepsilon_k \cos 2k\pi(F(x) - 1/2) \cos 2k\pi(F_1(y) - 1/2) \right], \quad (20)$$

where  $\sum_{k=1}^{\infty} |\varepsilon_k| \leq 1$ , and, integrating (20), we obtain the corresponding family of distribution functions

$$P(x, y) = F(x)F_1(y) + \frac{1}{4\pi^2} \sum_{k=1}^{\infty} \frac{\varepsilon_k}{k^2} \sin 2\pi k(F(x) - 1/2) \sin 2\pi k(F_1(y) - 1/2). \quad (21)$$

If in (18) we put  $\varphi(x) = F^k(x)$ ,  $\varphi_1(y) = F_1^k(y)$ , and then consider the linear combination of the functions obtained, putting  $\varepsilon_k = \varepsilon^k$ ,  $k = 1, 2, \dots$ ,  $0 \leq \varepsilon \leq 1/2$ ,

then we arrive at the following expression, also giving functions from the Fréchet class:

$$P_2(x, y, \varepsilon) = F(x)F_1(y) \left[ 1 + \frac{l(\varepsilon)}{\varepsilon} \right] - \frac{F(x)}{\varepsilon} l[\varepsilon F_1(y)] - \frac{F_1(y)}{\varepsilon} l[\varepsilon F(x)] + \frac{1}{\varepsilon} l[\varepsilon F(x)F_1(y)], \quad (22)$$

where

$$l(z) = - \int_0^z \frac{\ln(1-t)}{t} dt = \sum_{k=1}^{\infty} \frac{z^k}{k^2}, \quad \text{and} \quad P_2(x, y, 0) = F(x)F_1(y).$$

In conclusion, let us give one example of discontinuous densities from the normal Fréchet class

$$f(x, y, \lambda) = \begin{cases} \frac{\exp \left[ -\frac{x^2 + y^2}{2} \right]}{2\pi} (1 + \lambda), & \text{for } xy > 0, \\ \frac{\exp \left[ -\frac{x^2 + y^2}{2} \right]}{2\pi} (1 - \lambda), & \text{for } xy \leq 0, \\ -1 \leq \lambda \leq 1. \end{cases} \quad (23)$$

For the distribution (23),  $\lambda$  is the maximal correlation coefficient.

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

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