

# “ON THE CONVERGENCE OF PROJECTED DISTRIBUTIONS ”\*

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## 1. Introduction

Let  $X_n, Y_n$  ( $n=1, 2, \dots$ ) be real random variables with “laws” or cumulative distribution functions (c.d.f.’s)  $L(X_n), L(Y_n)$  and characteristic functions (c.f.’s)  $\varphi_{X_n}(t), \varphi_{Y_n}(t)$  respectively. Suppose, as we shall throughout, that there are c.d.f.’s  $F(x), G(x)$  with corresponding c.f.’s  $f(t), g(t)$  such that

$$(1) \quad L(X_n) \rightarrow F, \quad L(Y_n) \rightarrow G,$$

i.e.,  $X_n$  and  $Y_n$  converge in law or distribution. If  $*$  denotes convolution, we seek conditions under which  $L(cX_n + Y_n) \rightarrow F\left(\frac{x}{c}\right) * G(x)$ , for some fixed positive  $c$  (mutatis mutandis for  $c < 0$ ) or equivalently for any fixed  $c \neq 0$ ,

$$(2) \quad \lim_{n \rightarrow \infty} \varphi_{cX_n + Y_n}(t) = f(ct)g(t)$$

for all real  $t$ . It is well known that the independence of  $X_n$  and  $Y_n$  for all  $n$  or the degeneracy<sup>1)</sup> of  $F$  or  $G$  is ample to guarantee this for all real  $c$ . In fact, if  $X_n$  and  $Y_n$  are merely asymptotically independent, that is

$$\lim_{n \rightarrow \infty} E e^{it_1 X_n + it_2 Y_n} = f(t_1)g(t_2)$$

then (2) is the immediate consequence of the substitution  $t_1 = ct, t_2 = t$ . As customary,  $E$  is the expected value operator.

Note that the distribution of  $cX_n + Y_n$  arises from the projection of the 2-dimensional mass distribution of  $X_n, Y_n$  on the ray through the origin with slope  $1/c$ .

REMARK. Clearly, (2) holds when  $\varphi_{cX_n + Y_n}(t) = \varphi_{X_n}(ct)\varphi_{Y_n}(t)$ , hence when  $X_n$  and  $Y_n$  are independent ( $n=1, 2, \dots$ ). On the other hand, if

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1) A c.d.f.  $F(x)$  is called degenerate if it has only one point of increase (with unit saltus).

$c=1$ ,  $Y_n=X_n$  then (2) holds if and only if  $f(2t)=f^2(t)$ , a severe limitation. If, further,  $\int_{-\infty}^{\infty} x^2 dF(x) < \infty$ , then  $F$  is degenerate.

For if  $X$  and  $Y$  are independent random variables, each having c.f.  $f(t)$ , the prior functional equation asserts that  $2X$  and  $X+Y$  are equidistributed; hence, from a result of Linnik (Dokl. Akad. Nauk SSSR 89 (1953) pp. 9-11),  $X$  is normally distributed, but then the functional equation coerces  $f(t)$  to equal  $e^{iat}$  for some real  $a$ .

## 2. Convergence<sup>2)</sup>

As the point of departure, we recall the unusual definition of the independence of two random variables proffered by Wintner. In [3, p. 5],  $X$  and  $Y$  are defined to be statistically independent if

$$(3) \quad L(X+Y) = L(X) * L(Y)$$

That this is not equivalent to the standard definition (in terms of product measure) is shown by an example of Cramér [see section 3]. Note, however, that if  $EX^k < \infty$ ,  $EY^k < \infty$  for some integer  $k > 1$ , (3) implies

$$\sum_{j=0}^k \frac{(it)^j}{j!} E(X+Y)^j + o(t^k) = \left[ \sum_{j=0}^k \frac{(it)^j}{j!} EX^j + o(t^k) \right] \left[ \sum_{j=0}^k \frac{(it)^j}{j!} EY^j + o(t^k) \right]$$

which in turn requires that

$$\sum_{j=1}^{r-1} \binom{r}{j} \text{Cov}(X^j, Y^{r-j}) = 0, \quad r=2, 3, \dots, k$$

where  $\text{Cov}(X, Y) = E[X-EX][Y-EY]$ . In particular ( $k=2$ ), if it is meaningful to speak of the existence or lack of correlation between  $X$  and  $Y$ , i.e.,  $E[X-EX]^2$  and  $E[Y-EY]^2$  are non-zero finite quantities, (3) implies that  $X$  and  $Y$  are uncorrelated.

Let  $I$  be the class of all infinitely differentiable c.f.'s; let  $I_1$  be the subclass such that  $\varphi \in I$ ,  $\varphi_1 \in I_1$ ,  $D_k[\varphi(t)]_{t=0} = \varphi^{(k)}(0) = \varphi_1^{(k)}(0)$  ( $k=1, 2, \dots$ ) imply  $\varphi(t) = \varphi_1(t)$ . For instance,  $\varphi \in I_1$  if  $\varphi \in I$  and  $\sum_{j=0}^{\infty} |\varphi^{(2j)}(0)|^{-1/2j} = \infty$  (Carleman). This is true in turn when  $\varphi(t)$  is analytic at  $t=0$ , e.g., when the corresponding random variable has a finite range.

**LEMMA 1.** *Let  $\varphi(t) \in I_1$ ,  $\varphi_n(t) \in I$  ( $n=1, 2, \dots$ ). In order that*

<sup>2)</sup> The writer cordially thanks his colleagues Dr. J. H. B. Kemperman and Dr. J. I. Rosenblatt for helpful suggestions.

$$(4) \quad \lim_{n \rightarrow \infty} \varphi_n(t) = \varphi(t), \quad -\infty < t < \infty$$

it is sufficient that

$$(5) \quad \lim_{n \rightarrow \infty} \varphi_n^{(k)}(0) = \varphi^{(k)}(0)$$

for all positive integers  $k$ . If, for infinitely many even  $k$ , the sequence  $\{\varphi_n^{(k)}(0)\}$  is bounded, then a necessary condition for (4) is that (5) holds for all  $k=1, 2, \dots$ .

PROOF. cf. [2, p. 184-185].

When it is unnecessary or inconvenient to emphasize the random variable in question, we shall denote  $\varphi_{X_n}(t)$ ,  $\varphi_{Y_n}(t)$ ,  $\varphi_{cX_n+Y_n}(t)$  by  $f_n(t)$ ,  $g_n(t)$  and  $h_n(t)$  respectively.

THEOREM 1. Suppose that for each  $n=1, 2, \dots$ ,  $f_n(t) \in I$ ,  $g_n(t) \in I$ ,  $f(ct)g(t) \in I_1$ . Moreover, suppose that for each  $k=1, 2, \dots$  the sequences  $\{f_n^{(k)}(0)\}$  and  $\{g_n^{(k)}(0)\}$  are bounded. Then, in order that (2) holds, it is sufficient that

$$(6) \quad \lim_{n \rightarrow \infty} \sum_{j=1}^{r-1} \binom{r}{j} c^j \text{Cov}(X_n^j, Y_n^{r-j}) = 0$$

for all positive integers  $r$ . Conversely, (2) implies that (6) holds for every integer  $r \geq 2$ .

PROOF. For each  $k$ ,  $\{D_k[f_n(ct) \cdot g_n(t)]_{t=0}\}$  is a bounded sequence. Further, from (1),  $\lim_{n \rightarrow \infty} f_n(ct)g_n(t) = f(ct)g(t)$ .

Hence, from lemma 1, for each  $k$

$$(7) \quad \lim_{n \rightarrow \infty} D_k[f_n(ct)g_n(t)]_{t=0} = D_k[f(ct)g(t)]_{t=0}.$$

Also, the sequence  $\{h_n^{(k)}(0)\} = \{i^k E(cX_n + Y_n)^k\}$  is easily seen to be bounded ( $k=1, 2, \dots$ ). Again, using lemma 1, it follows that (2) holds if

$$(8) \quad \lim_{n \rightarrow \infty} h_n^{(k)}(0) = D_k[f(ct)g(t)]_{t=0}.$$

for all positive  $k$  and only if (8) holds for all  $k$ . The assertion of the theorem now follows from (7), (8) and

$$(9) \quad h_n^{(k)}(0) - D_k[f_n(ct)g_n(t)]_{t=0} = i^k \sum_{j=1}^{k-1} \binom{k}{j} c^j \text{Cov}(X_n^j, Y_n^{k-j}).$$

**THEOREM 2.** Let  $X, Y$  be real random variables with joint cf.  $\varphi(z_1, z_2)$ . Suppose that  $\psi_1(z) = \varphi(z, 0)$  and  $\psi_2(z) = \varphi(0, z)$  are analytic for  $|z| \leq r_1, |z| \leq r_2$ , respectively. Then  $\varphi(z_1, z_2)$  is analytic for  $|\vartheta(z_1)| < 1/2 r_1, |\vartheta(z_2)| < 1/2 r_2$ .

**PROOF.** Since  $\sum_{k=0}^{\infty} |\phi_j^{(k)}(0)| r_j^k / k!$  is convergent, we have  $|\phi_j^{(k)}(0)| \leq C_j k! r_j^{-k}$  for  $j=1, 2$  where  $C_1, C_2$  are absolute constants. Further,

$$\begin{aligned} |\varphi^{(k,m)}(0, 0)| &\leq E|X|^k |Y|^m \leq \sqrt{E|X|^{2k} E|Y|^{2m}} \\ &\leq C_1 C_2 r_1^{-k} r_2^{-m} \sqrt{(2k)! (2m)!} \\ &\leq C (1/2 r_1)^{-k} (1/2 r_2)^{-m} k! m! \end{aligned}$$

since  $\lim_{n \rightarrow \infty} \frac{(2n)!}{(2^n n!)^2} = 0$ . The assertion now follows from a straight-forward generalization of a standard procedure (cf. [1, p. 177], [2, p. 212]).

**REMARK.** Taking  $X \equiv Y$ , we have  $\varphi(t_1, t_2) = \psi_1(t_1 + t_2)$  showing the stated region of analyticity cannot be improved very much.

**COROLLARY.** Under the hypothesis of Theorem 2,  $\varphi(z_1, z_2)$  defines an analytic function for  $|\vartheta(z_1)| < \theta r_1, |\vartheta(z_2)| < (1 - \theta)r_2$  for any  $\theta$  in  $(0, 1)$ .

**PROOF.** Now  $E e^{r_1 |X|} < \infty, E e^{r_2 |Y|} < \infty$  (cf. [2, p. 212]). Hence, by Holder's inequality,  $E e^{\theta r_1 |X| + (1-\theta)r_2 |Y|} < \infty$ . Thus, the integral

$$E[\exp \{iz_1 X + iz_2 Y\}] = \varphi(z_1, z_2)$$

converges and defines an analytic function in  $|\vartheta(z_1)| < \theta r_1, |\vartheta(z_2)| < (1 - \theta)r_2$ .

The optimum choice seems  $\theta = \frac{r_2}{r_1 + r_2}$ .

We are now in a position to prove:

**THEOREM 3.** Suppose that  $f_n(z), g_n(z)$  are analytic in  $|z| \leq r_1, |z| \leq r_2$  respectively and that  $f(ct) g(t)$  has a unique extension from an  $\frac{r_1 r_2}{|c|r_1 + r_2} \sqrt{1+c^2}$  neighborhood of the origin. Let

$$a_{n,k}(c) = \sum_{j=1}^{k-1} \frac{c^j}{j! (k-j)!} \text{Cov}(X_n^j, Y_n^{k-j})$$

and suppose further that for some positive number  $\varepsilon > \frac{r_1 r_2 \sqrt{1+c^2}}{|c|r_1 + r_2}$ ,

$$\lim_{n \rightarrow \infty} \sum_1^{\infty} |a_{nk}(c)| \varepsilon^k = 0 .$$

Then (2) holds.

PROOF. From [2, p. 211], it suffices to prove that (2) holds in a neighborhood of zero. Then, according to (1), it suffices to demonstrate that

$$\lim_{n \rightarrow \infty} [h_n(t) - f_n(ct)g_n(t)] = 0$$

holds in a neighborhood of the origin. However, from the corollary to Theorem 2  $h_n(t)$  is analytic (Take  $(1-\theta)r_2 = |c|\theta r_1$ ) for  $|t| < \frac{r_1 r_2}{|c|r_1 + r_2} \sqrt{1+c^2} = \delta$  (say). From (9), for  $|t| < \delta$

$$(10) \quad h_n(t) - f_n(ct)g_n(t) = \sum_1^{\infty} a_{nk}(it)^k ;$$

hence, the left hand side of (10) tends to zero for  $|t| < \delta$ .

If, in Theorem 1, we take  $X_n = X$ ,  $Y_n = Y$  ( $n = 1, 2, \dots$ ), we obtain the following:

COROLLARY. If  $f(t), g(t) \in I$  and  $f(ct)g(t) \in I$ , then  $cX + Y$  has the characteristic function  $f(ct)g(t)$  if

$$(11) \quad 0 = \sum_{j=1}^{r-1} \binom{r}{j} c^j \text{Cov}(X^j, Y^{r-j}) = P_r(c) \text{ (say)}$$

for all positive integers  $r$  and only if (11) holds for all integers  $r \geq 2$  (in particular,  $X$  and  $Y$  are uncorrelated).

Now, (11) will certainly hold if

$$(12) \quad \text{Cov}(X^j, Y^k) = 0 \quad \text{for all } j, k = 1, 2, \dots .$$

Suppose this to be the case and in addition that  $f(t)$  and  $g(t)$  are analytic, say for  $|t| < 2r_1$  and  $|t| < 2r_2$  respectively. Then (11) is valid for all integers  $r \geq 2$  and all real  $c$ , whence by the corollary just cited,  $\varphi_{cX+Y}(t) = \varphi_X(ct)\varphi_Y(t)$ , for all real  $c$ . But then if  $(\theta, t)$  is the polar coordinate representation of a point  $(t_1, t_2)$  of the real plane and  $c = \cot \theta$ ,  $\theta \neq 0$ ,

$$\begin{aligned} \varphi_{X,Y}(t_1, t_2) &= \varphi_{X \cos \theta, Y \sin \theta}(t, t) \\ &= \varphi_{X \cos \theta + Y \sin \theta}(t) = \varphi_{cX+Y}(t \sin \theta) \\ &= \varphi_{cX}(t \sin \theta) \varphi_Y(t \sin \theta) = \varphi_X(t_1) \varphi_Y(t_2) . \end{aligned}$$

Consequently,  $X$  and  $Y$  are independent in this case. An alternative argument, due to J. H. B. Kemperman, consists in noting that by Theorem 2

$$\varphi(t_1, t_2) = \sum_{j,k} E(X^j Y^k) \frac{(it_1)^j (it_2)^k}{j! k!}$$

is analytic in a neighborhood of  $(0, 0)$ , that according to (12),  $EX^j Y^k = EX^j \cdot EY^k$  and hence  $\varphi(t_1, t_2) = f(t_1)g(t_2)$ , first, in a neighborhood of the origin and then for all real  $(t_1, t_2)$  by analytic continuation.

If (11) obtains and (12) does not, let  $r$  be the smallest positive integer with  $\text{Cov}(X^j, Y^{r-j}) \neq 0$  for some (and therefore at least two such)  $j=1, 2, \dots, r-1$ . Again, by the corollary, since  $P_r(c)$  has constant term zero,

$$(13) \quad \varphi_{cX+Y}(t) = \varphi_X(ct) \cdot \varphi_Y(t), \quad \text{all real } t$$

for at most  $r-2$  non-zero values of  $c$ . We have thus proved:

**THEOREM 4.** *If  $X$  and  $Y$  have analytic characteristic functions and  $\text{Cov}(X^j, Y^k) = 0$  for all integers  $j, k \geq 1$ , then  $X$  and  $Y$  are independent random variables. Conversely, if  $r \geq 2$  is the smallest integer with  $\text{Cov}(X^j, Y^{r-j}) \neq 0$  for some  $j=1, 2, \dots, r-1$ , then (13) obtains for at most  $r-2$  non-zero values of  $c$ .*

### 3. Extension and Examples

Under the hypothesis of Theorem 1, we have even  $\lim_{n \rightarrow \infty} \varphi_{c_n X_n + Y_n}(t) = f(ct)g(t)$  where  $c_n$  is a sequence of real constants approaching  $c$ .

If in (6),  $X_n$  and  $Y_n$  are replaced by  $\log X_n$  and  $\log Y_n$  (where  $X_n$  and  $Y_n$  are now presumed to be positive r.v.'s), we obtain a condition for  $L(X_n^{c_n} \cdot Y_n^b) \rightarrow H$  where  $b \neq 0$  is an arbitrary constant and

$$H(z) = \int_{a^{c_0}} \int_{y^b < z} dF(x) dG(y), \quad z > 0$$

$$= 0, \quad z \leq 0.$$

Results analogous to those of section 2 may be obtained for the case of a sequence of  $k$  dimensional vectors  $X_n = (X_{n1}, X_{n2}, \dots, X_{nk})$  where it is known that  $L(X_{nj}) \rightarrow F_j$ ,  $j=1, 2, \dots, k$  and it is desired that

$$L\left(\sum_{j=1}^k c_j X_{nj}\right) \rightarrow F_1\left(\frac{x}{c_1}\right) * F_2\left(\frac{x}{c_2}\right) * \dots * F_k\left(\frac{x}{c_k}\right).$$

In the following example of Cramér ( $\theta=1$ ) already alluded to,  $r=4$  while  $X+Y$  and  $-X+Y$  are precisely the two linear combinations with the property (13):

$$f_\theta(x, y) = \frac{1}{4\theta^2} [1 + xy(x^2 - y^2)], \quad \text{if } |x| \leq \theta, |y| \leq \theta \\ = 0, \quad \text{otherwise.}$$

Here, both  $X$  and  $Y$  are uniformly distributed in  $[-\theta, \theta]$ . Theorem 1 may be trivially exemplified for  $c=1$  or  $-1$  by assigning  $X_n$  and  $Y_n$  the joint density function  $f_{\theta_n}(x, y)$  where  $\lim_{n \rightarrow \infty} \theta_n = \theta > 0$ .

The question naturally poses itself whether for every integer  $r > 2$  and arbitrary constants  $c_1, c_2, \dots, c_s$  ( $s=r-2$ ), there exists a joint distribution such that (13) obtains for these and only these values of  $c$  (excluding the trivial value zero). The answer is affirmative.

Let  $q(x) = \sin x$ ,  $-\pi \leq x \leq \pi$

$$0, \quad |x| > \pi$$

Its Fourier transform, say  $Q(t)$  is

$$Q(t) = i \left[ \frac{\sin \pi(1-t)}{1-t} - \frac{\sin \pi(1+t)}{1+t} \right]$$

an entire function of exponential order 1 with  $Q(0)=0$ . Define

$$\Gamma(t_1, t_2) = Q(t_1) Q(t_2) \prod_{i=1}^s Q(t_1 - c_i t_2)$$

where  $c_1, c_2, \dots, c_s$  are the previously specified constants. Then

$$\Gamma(t, 0) = \Gamma(0, t) = \Gamma(c_i t, t) = 0 \quad \text{for all real } t, \quad i=1, 2, \dots, s$$

and since  $\Gamma(t_1, t_2)$  is of exponential order, entire, its inverse Fourier transform, say  $\gamma(x, y)$  vanishes outside some bounded region  $A$  of the  $x, y$  plane. [cf. 4, p. 13] Finally, since  $\gamma(x, y)$  is the convolution of functions one of which is continuous, it is bounded in  $A$ . Thus, if  $f_1(x), f_2(x)$  are probability density functions with spectrum  $(-\infty, \infty)$  e.g.,  $f_1(x) = f_2(x) = \frac{1}{\sqrt{2\pi}} e^{-x^2/2}$ , then for sufficiently small  $\epsilon$ ,

$$f(x, y) = f_1(x) f_2(y) + \epsilon \gamma(x, y)$$

is a probability density function whose c.f.

$$\varphi(t_1, t_2) = \varphi_1(t_1)\varphi_2(t_2) + \varepsilon\Gamma(t_1, t_2)$$

where  $\varphi_j(t) = \int_{-\infty}^{\infty} e^{itz} f_j(x) dx$ , Clearly,

$$\varphi(c_j t, t) = \varphi_1(c_j t)\varphi_2(t) \quad \text{for all real } t$$

and  $j=1, 2, \dots, s$ , and no other non-zero values of  $c$ .

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