

# AN EXTENSION OF THE SLUTZKY-FRÉCHET THEOREM.

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## § 0. Notation and conventions.

In this paper German capital letters denote Euclidean vector spaces of finite dimensionality. Small German letters denote point sets in these spaces; and  $\mathfrak{z} - \mathfrak{z}'$  denotes the (perhaps empty) set of all points which belong to  $\mathfrak{z}$  and not to  $\mathfrak{z}'$ . Script letters denote classes of point sets. Clarendon type denotes points (or vectors) of a Euclidean space. Ordinary italic type is reserved for scalar quantities. The symbol  $\Rightarrow$  denotes implication, the arrow pointing from the premiss to the conclusion; and the double-headed arrow  $\Leftrightarrow$  means 'implies and is implied by'. Two statements I and II, which together imply a third III, are linked by an ampersand: — 'I & II  $\Rightarrow$  III'.

## § 1. Introduction.

Let  $y(x)$  be a continuous one-valued function of  $x$ , and consider the equations

$$\lim_{\nu \rightarrow \infty} x_\nu = x, \quad (1.1)$$

$$\lim_{\nu \rightarrow \infty} (x_\nu - x) = 0, \quad (1.2)$$

$$\lim_{\nu \rightarrow \infty} \{y(x_\nu) - y(x)\} = 0, \quad (1.3)$$

$$\lim_{\nu \rightarrow \infty} y(x_\nu) = y(x). \quad (1.4)$$

When  $x$  and  $x_\nu$  are real variables, it is familiar that

$$(1.1) \Leftrightarrow (1.2) \Rightarrow (1.3) \Leftrightarrow (1.4). \quad (1.5)$$

For random variables, the position is different. Slutsky (4) proved

$$(1.2) \Rightarrow (1.3) \quad (1.6)$$

when  $x_\nu$  is a random variable and  $x$  a real variable; while Fréchet (1) proved (1.6)

in case  $x_r$  and  $x$  were both random variables. It is an immediate consequence of the definition of 'lim' for random variables that

$$(1.1) \Leftarrow (1.2) \quad \text{and} \quad (1.3) \Rightarrow (1.4) \tag{1.7}$$

but the converse statements

$$(1.1) \Rightarrow (1.2) \quad \text{and} \quad (1.3) \Leftarrow (1.4) \tag{1.8}$$

are generally false. It is, however, easy to find special cases in which (1.8) is true for certain specific random variables; and then the question naturally arises whether, given *any* random variables satisfying (1.1), we can *always* find at least one special case such that (1.2) is also true. In Theorem 1 I shall give an affirmative answer to this question: so that, combining Theorem 1 with the Slutsky-Fréchet theorem (1.6) and with the second part of (1.7), we shall have established

$$(1.1) \Rightarrow (1.4) \tag{1.9}$$

for random variables. However (1.9) is insufficient for certain practical applications; and I shall prove a generalisation of it in Theorem 3: namely, that (1.6) and (1.9) remain true for almost-certainly-continuous many-valued vector functions of a vector variable.

A number of authors have discussed, in a few special cases, the distribution of the zeros of a random polynomial. I hope to show elsewhere how the extended form of (1.9) provides a general solution to this problem.

## § 2. One-valued random variables and their limits.

Let  $\mathcal{X}$  denote an  $n$ -dimensional Euclidean space. A *probability set function*  $F[\mathfrak{z}]$  is any one-valued real non-negative completely-additive set function defined for all Borel sets  $\mathfrak{z}$  of  $\mathcal{X}$  and satisfying  $F[\mathcal{X}] = 1$ . If  $\mathfrak{z}$  is the particular set of all points, whose coordinates do not exceed the corresponding coordinates of a given point  $\mathbf{x}$  of  $\mathcal{X}$ , we write  $F[\mathfrak{z}] = F(\mathbf{x})$  and call  $F(\mathbf{x})$  a *cumulative distribution function*. Obviously  $F[\mathfrak{z}]$  uniquely determines  $F(\mathbf{x})$ , and the converse is a consequence of Lebesgue's theory of integration. A cumulative distribution function is monotone increasing and everywhere continuous on the right. For the purposes of axiomatic theory it is permissible to identify a *one-valued random variable*  $\mathbf{x}^*$  with a probability set function. Asterisks will hereinafter denote random variables. If the functional form of  $F$ , either as a probability set function or as a cumulative distribution function, is supposed given we say that  $F$  *determines* the random variable  $\mathbf{x}^*$  identified with it.

This corresponds to saying that a real variable  $\mathbf{x}$  is determined when the numerical values of its coordinates are supposed given. A *random constant*  $\mathbf{a}^*$  is the random variable identified with that probability set function  $F[\mathfrak{r}]$  which equals 1 or 0 according as the fixed point  $\mathbf{a}$  belongs to  $\mathfrak{r}$  or not.

Let  $\mathfrak{X}_i, i = 1, 2, \dots, m$ , be an  $n_i$ -dimensional Euclidean space in which  $\mathfrak{r}_i$  is a typical Borel set. Let  $\mathbf{x}_i^*$  be a one-valued random variable in  $\mathfrak{X}_i$  determined by  $F_i[\mathfrak{r}_i]$ . In the direct product space  $\mathfrak{X}_1 \times \mathfrak{X}_2 \times \dots \times \mathfrak{X}_m$  any probability set function  $G$  is called a *joint determination* of  $\mathbf{x}_1^*, \mathbf{x}_2^*, \dots, \mathbf{x}_m^*$  if it satisfies

$$G[\mathfrak{X}_1 \times \mathfrak{X}_2 \times \dots \times \mathfrak{X}_{i-1} \times \mathfrak{r}_i \times \mathfrak{X}_{i+1} \times \dots \times \mathfrak{X}_m] = F_i[\mathfrak{r}_i] \tag{2.1}$$

for all values of  $i$  and all Borel sets  $\mathfrak{r}_i$  of  $\mathfrak{X}_i$ . The random variable identified with  $G$  is written  $\mathbf{x}_1^* \times \mathbf{x}_2^* \times \dots \times \mathbf{x}_m^*$ . We say that the  $\mathbf{x}_i^*$  are *independently distributed* if a stronger form of (2.1) holds, namely

$$G[\mathfrak{r}_1 \times \mathfrak{r}_2 \times \dots \times \mathfrak{r}_m] = F_1[\mathfrak{r}_1] F_2[\mathfrak{r}_2] \dots F_m[\mathfrak{r}_m] \tag{2.2}$$

for all Borel sets  $\mathfrak{r}_i \subseteq \mathfrak{X}_i$ .

Let  $G$  in (2.1) be a joint determination of  $\mathbf{x}_1^*, \mathbf{x}_2^*, \dots, \mathbf{x}_m^*$ . Let  $\mathbf{y} = \mathbf{y}(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_m)$  be a one-valued Borel-measurable mapping of  $\mathfrak{X}_1 \times \mathfrak{X}_2 \times \dots \times \mathfrak{X}_m$  into a Euclidean space  $\mathfrak{Y}$ . Let  $\mathfrak{h}$  be a Borel set of  $\mathfrak{Y}$ , and let  $\mathfrak{r}(\mathfrak{h})$  be the set of all points  $(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_m)$  in  $\mathfrak{X}_1 \times \mathfrak{X}_2 \times \dots \times \mathfrak{X}_m$  for which  $\mathbf{y}(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_m) \in \mathfrak{h}$ . Since  $\mathbf{y}(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_m)$  is a Borel-measurable function,  $\mathfrak{r}(\mathfrak{h})$  is a Borel set. The *function of several jointly determined random variables*

$$\mathbf{y}^* = \mathbf{y}(\mathbf{x}_1^*, \mathbf{x}_2^*, \dots, \mathbf{x}_m^*) = \mathbf{y}(\mathbf{x}_1^* \times \mathbf{x}_2^* \times \dots \times \mathbf{x}_m^*)$$

is defined to be the random variable identified with

$$H[\mathfrak{h}] = G[\mathfrak{r}(\mathfrak{h})],$$

it being easy to verify that  $H[\mathfrak{h}]$  so defined is a probability set function. Indeed this is a particular case of some more general theory discussed by Hammersley (3).

In a Euclidean space  $\mathfrak{X}$ , let  $\mathbf{x}^*$  be a random variable determined by the cumulative distribution function  $F(\mathbf{x})$  and let  $\mathbf{x}_\nu^*, \nu = 1, 2, \dots$ , be a sequence of random variables respectively determined by  $F_\nu(\mathbf{x})$ . If, as  $\nu \rightarrow \infty$ ,  $F_\nu(\mathbf{x})$  tends to  $F(\mathbf{x})$  at every point of continuity of  $F(\mathbf{x})$ , we say that  $\mathbf{x}_\nu^*$  *converges in distribution* to  $\mathbf{x}^*$ , and we write

$$\text{dlim}_{\nu \rightarrow \infty} \mathbf{x}_\nu^* = \mathbf{x}^*. \tag{2.3}$$

Let  $\mathbf{a}$  be a constant vector, and let  $\mathfrak{r}(\mathbf{a})$  denote the set of points  $\mathbf{x}$  satisfying

$|\mathbf{x} - \mathbf{a}| < \delta$ , where  $\delta > 0$  is any prescribed positive number. If to every prescribed pair of positive numbers  $\delta > 0$  and  $\varepsilon > 0$  we can find a positive integer  $\nu_0 = \nu_0(\delta, \varepsilon)$  such that the probability set functions  $F_\nu[\mathfrak{X}]$  of  $\mathbf{x}_\nu^*$  satisfy

$$F_\nu[\mathfrak{X}(\mathbf{a})] > 1 - \varepsilon, \quad \nu \geq \nu_0(\delta, \varepsilon)$$

we say that  $\mathbf{x}_\nu^*$  converges in probability to  $\mathbf{a}$ , and we write

$$\text{plim}_{\nu \rightarrow \infty} \mathbf{x}_\nu^* = \mathbf{a}. \quad (2.4)$$

It is not difficult to see that

$$\text{plim}_{\nu \rightarrow \infty} \mathbf{x}_\nu^* = \mathbf{a} \Leftrightarrow \text{dlim}_{\nu \rightarrow \infty} \mathbf{x}_\nu^* = \mathbf{a}^*. \quad (2.5)$$

If, for each value of  $\nu$ ,  $\mathbf{x}_\nu^*$  and  $\mathbf{x}^*$  are jointly determined by some given  $G_\nu$ , and if the function  $\mathbf{x}_\nu^* - \mathbf{x}^*$  of such a pair of jointly determined random variables converges in probability to the zero vector as  $\nu \rightarrow \infty$ , we say that  $\mathbf{x}_\nu^*$  converges in probability to  $\mathbf{x}^*$ , and write

$$\text{plim}_{\nu \rightarrow \infty} \mathbf{x}_\nu^* = \mathbf{x}^*. \quad (2.6)$$

Thus

$$\text{plim}_{\nu \rightarrow \infty} (\mathbf{x}_\nu^* - \mathbf{x}^*) = \mathbf{0} \Leftrightarrow \text{plim}_{\nu \rightarrow \infty} \mathbf{x}_\nu^* = \mathbf{x}^* \Leftrightarrow \text{dlim}_{\nu \rightarrow \infty} (\mathbf{x}_\nu^* - \mathbf{x}^*) = \mathbf{0}^* \quad (2.7)$$

when  $\mathbf{x}_\nu^*$  and  $\mathbf{x}^*$  are jointly determined; and it is quite simple to show that (2.7)  $\Rightarrow$  (2.3). This is a fuller explanation of the first part of (1.7). On the other hand, the truth of '(2.3)  $\Rightarrow$  (2.7)' depends upon the form of the joint determination of  $\mathbf{x}^*$  and  $\mathbf{x}_\nu^*$ . We shall now prove in Theorem 1 that, amongst the class of all joint determinations of any given pair of individually determined random variables  $\mathbf{x}^*$  and  $\mathbf{x}_\nu^*$ , there is always at least one joint determination such that (2.3)  $\Rightarrow$  (2.7).

**Theorem 1.** *If  $\mathbf{x}^*$  is a given one-valued random variable, and if  $\mathbf{x}_\nu^*$ ,  $\nu = 1, 2, \dots$ , is a sequence of given one-valued random variables satisfying*

$$\text{dlim}_{\nu \rightarrow \infty} \mathbf{x}_\nu^* = \mathbf{x}^*, \quad (2.8)$$

*then, for each value of  $\nu$ , there exists a joint determination of  $\mathbf{x}^*$  and  $\mathbf{x}_\nu^*$  such that*

$$\text{dlim}_{\nu \rightarrow \infty} (\mathbf{x}_\nu^* - \mathbf{x}^*) = \mathbf{0}^*. \quad (2.9)$$

Take  $\mathfrak{X}$  to be the Euclidean space in which  $\mathbf{x}^*$  is defined; and write  $\mathbf{x} = \{x_1, x_2, \dots, x_n\}$  for a typical point of  $\mathfrak{X}$ , and  $\mathfrak{X}$  for a typical Borel set of  $\mathfrak{X}$ . Suppose that  $F[\mathfrak{X}]$  and  $F_\nu[\mathfrak{X}]$  are the given probability set functions which determine

$\mathbf{x}^*$  and  $\mathbf{x}_v^*$  respectively, and that  $F(\mathbf{x})$  and  $F_v(\mathbf{x})$  are the corresponding cumulative distribution functions. Let  $\delta > 0$  and  $\varepsilon > 0$  be any pair of prescribed positive numbers.

We can find a finite number  $U = U(\varepsilon) > 0$  such that

- (i)  $F(\mathbf{x})$  is continuous on each of the hyperplanes  $\mathfrak{h}_{ij}$ , ( $i = 1, 2, \dots, n$ ;  $j = 1, 2$ ), where  $\mathfrak{h}_{i1}$  is the hyperplane  $x_i = +U$  and  $\mathfrak{h}_{i2}$  is the hyperplane  $x_i = -U$ ; and
- (ii)  $F[\mathfrak{X}_0] < \frac{1}{2}\varepsilon$ , where  $\mathfrak{X}_0$  is the set of all points which violate at least one of the  $n$  inequalities  $-U < x_i \leq +U$ ,  $i = 1, 2, \dots, n$ .

We can now find a finite sequence of numbers  $u_k$ ,  $k = 1, 2, \dots, m$ , where  $m = m(\delta, \varepsilon)$ , such that

- (iii)  $-U = u_1 < u_2 < \dots < u_m = +U$ ; and
- (iv)  $u_{k+1} - u_k < \delta/Vn$ ,  $k = 1, 2, \dots, m - 1$ ; and
- (v)  $F(\mathbf{x})$  is continuous on the hyperplanes  $\mathfrak{h}^{ik}$ , ( $i = 1, 2, \dots, n$ ;  $k = 1, 2, \dots, m$ ), where  $\mathfrak{h}^{ik}$  is the hyperplane  $x_i = u_k$ .

Write  $M = M(\delta, \varepsilon) = (m - 1)^n$ ; and let  $\mathfrak{X}_p$ ,  $p = 1, 2, \dots, M$ , denote the half-open finite intervals in  $\mathfrak{X}$

$$u_{k(i)} < x_i \leq u_{k(i)+1}, \quad i = 1, 2, \dots, n \tag{2.10}$$

enumerated in some specific order, where  $k(i)$  denotes an integer (depending upon  $i$ ) selected from the integers  $1, 2, \dots, m - 1$ . Consider the non-negative numbers

$$a_p = F[\mathfrak{X}_p], \quad b_p = F_v[\mathfrak{X}_p], \quad p = 0, 1, 2, \dots, M, \tag{2.11}$$

where  $b_p$  is a function of  $v$ . Since  $\mathfrak{X}_0, \mathfrak{X}_1, \dots, \mathfrak{X}_M$  are mutually disjoint and cover  $\mathfrak{X}$  completely

$$\sum_p a_p = \sum_p b_p = 1. \tag{2.12}$$

Let  $\delta_{pq}$  denote the Kronecker delta ( $\delta_{pq} = 1$  or  $0$  according as  $p = q$  or  $p \neq q$ ); and let  $\Lambda(\theta) = \theta$  if  $\theta \neq 0$  while  $\Lambda(0) = 1$ . Define for  $p, q = 0, 1, 2, \dots, M$

$$c_{pq} = \frac{(a_p + b_p - |a_p - b_p|)\delta_{pq}}{2} + \frac{(|a_p - b_p| + a_p - b_p)(|a_q - b_q| - a_q + b_q)}{2\Lambda(\sum_p |a_p - b_p|)}. \tag{2.13}$$

In view of (2.12) and  $a_p \geq 0$ ,  $b_p \geq 0$ , we find without difficulty

$$c_{pq} \geq 0, \quad \sum_p c_{pq} = b_q, \quad \sum_q c_{pq} = a_p, \quad \sum_p c_{pp} = 1 - \frac{1}{2} \sum_p |a_p - b_p|. \tag{2.14}$$

Let  $\mathfrak{J}$  denote the  $2n$ -dimensional space  $\mathfrak{X} \times \mathfrak{X}$ , and let  $\mathfrak{z}'$  denote any Borel set of  $\mathfrak{J}$  which can be expressed in the form

$$\mathfrak{z}' = \mathfrak{r}' \times \mathfrak{r}'', \quad \mathfrak{r}' \subseteq \mathfrak{X}_p \text{ and } \mathfrak{r}'' \subseteq \mathfrak{X}_q \text{ for some } p, q, \tag{2.15}$$

$\mathfrak{r}'$  and  $\mathfrak{r}''$  being Borel sets. Define  $G'$ , depending upon  $\nu$ , by

$$G'[\mathfrak{z}'] = F[\mathfrak{r}'] F_\nu[\mathfrak{r}''] c_{pq} / \Lambda(a_p b_q), \quad (2.16)$$

where the values of  $p$  and  $q$  are those appearing in (2.15). It is easy to see that  $G'[\mathfrak{z}']$  is a non-negative completely-additive set function for all sets  $\mathfrak{z}'$  satisfying (2.15) for any fixed pair  $p, q$ . Now the intervals  $\mathfrak{r}_p \times \mathfrak{r}_q$  are mutually disjoint and cover  $\mathfrak{z}$  completely; and any Borel set of  $\mathfrak{z}$  can be built up from an enumerable number of sets of the form  $\mathfrak{r}' \times \mathfrak{r}''$ . Therefore we may uniquely define  $G[\mathfrak{z}]$  as that non-negative completely-additive set function of Borel sets  $\mathfrak{z} \subseteq \mathfrak{z}$  such that  $G[\mathfrak{z}'] = G'[\mathfrak{z}']$  for all sets of the type  $\mathfrak{z}'$ . Let  $\mathfrak{x}$  be any Borel set of  $\mathfrak{X}$ . There is a unique decomposition

$$\mathfrak{x} = \sum_p \mathfrak{x}^p, \quad \mathfrak{x}^p \subseteq \mathfrak{r}_p,$$

namely  $\mathfrak{x}^p = \mathfrak{x} \cdot \mathfrak{r}_p$ . Now

$$a_p = 0 \Rightarrow F[\mathfrak{x}^p] = 0; \quad b_q = 0 \Rightarrow c_{pq} = 0;$$

and so (2.14) establishes

$$\begin{aligned} G[\mathfrak{x} \times \mathfrak{X}] &= \sum_{p,q} F[\mathfrak{x}^p] F_\nu[\mathfrak{r}_q] c_{pq} / \Lambda(a_p b_q) \\ &= \sum_p \frac{F[\mathfrak{x}^p]}{\Lambda(a_p)} \sum_q \frac{c_{pq} b_q}{\Lambda(b_q)} = \sum_p \frac{F[\mathfrak{x}^p] a_p}{\Lambda(a_p)} = \sum_p F[\mathfrak{x}^p] = F[\mathfrak{x}]. \end{aligned} \quad (2.17)$$

Similarly

$$G[\mathfrak{X} \times \mathfrak{x}] = F_\nu[\mathfrak{x}]. \quad (2.18)$$

Hence

$$G[\mathfrak{z}] = G[\mathfrak{X} \times \mathfrak{X}] = F[\mathfrak{X}] = 1;$$

so that  $G$  is a probability set function. Whereupon (2.17) and (2.18) show that  $G$  jointly determines  $\mathbf{x}^*$  and  $\mathbf{x}_\nu^*$ .

Now write  $\mathbf{z} = \{z_1, z_2, \dots, z_{2n}\}$  for a typical point of  $\mathfrak{z}$ , and let  $\mathfrak{z}_0$  be the set of all points  $\mathbf{z}$  which satisfy all the inequalities

$$|z_i - z_{n+i}| < \delta/\nu, \quad i = 1, 2, \dots, n.$$

From (iv) and (2.10)

$$\mathfrak{z}_0 \supseteq \sum_{p=1}^M \mathfrak{r}_p \times \mathfrak{r}_p;$$

and therefore by (2.14)

$$\begin{aligned} G[\mathfrak{z}_0] &\geq G\left[\sum_{p=1}^M \mathfrak{r}_p \times \mathfrak{r}_p\right] = \sum_{p=1}^M G[\mathfrak{r}_p \times \mathfrak{r}_p] = \sum_{p=1}^M c_{pp} = 1 - c_{00} - \frac{1}{2} \sum_{p=0}^M |a_p - b_p| \\ &\geq 1 - a_0 - \frac{1}{2} \sum_{p=0}^M |a_p - b_p| > 1 - \frac{1}{2} \varepsilon - \frac{1}{2} \sum_{p=0}^M |a_p - b_p|, \end{aligned} \quad (2.19)$$

where in the final step we have employed condition (ii). Now each of the numbers  $a_p$  and  $b_p$  can be expressed as the sum or difference of  $2^n$  quantities of the form  $F(\mathbf{x})$  or  $F_v(\mathbf{x})$  where  $\mathbf{x}$  is an intersection of fixed hyperplanes  $\mathfrak{h}^{i^k}$ . Consequently (2.8) and condition (v) show that we can determine  $\nu_0 = \nu_0(\delta, \varepsilon)$  such that, for each  $p$ ,  $|a_p - b_p| < \varepsilon/(M + 1)$ ,  $\nu \geq \nu_0$ . On substitution into (2.19) we get

$$G[\mathfrak{z}_0] > 1 - \varepsilon, \quad \nu \geq \nu_0(\delta, \varepsilon),$$

which establishes (2.9) and completes the proof.

### § 3. Almost-certainly-continuous many-valued vector functions.

Suppose that, to each point  $\mathbf{x}$  of an  $n$ -dimensional Euclidean space  $\mathfrak{X}$ , there corresponds a system  $\mathbf{y}(\mathbf{x})$  of  $p$  points (not necessarily distinct) in a  $q$ -dimensional Euclidean space  $\mathfrak{Y}$ . We call  $\mathbf{y}(\mathbf{x})$  a  $p$ -valued  $q$ -dimensional vector function of  $\mathbf{x}$ . If there are defined a system of  $p$  one-valued functions of  $\mathbf{x}$

$$\mathbf{y}_1(\mathbf{x}), \mathbf{y}_2(\mathbf{x}), \dots, \mathbf{y}_p(\mathbf{x}) \tag{3.1}$$

such that, having due regard to multiple points, the points (3.1) coincide with the points  $\mathbf{y}(\mathbf{x})$  for each  $\mathbf{x}$  in  $\mathfrak{X}$ , then we call the functions (3.1) an *indexing* of  $\mathbf{y}(\mathbf{x})$ . If  $\mathbf{y}(\mathbf{x})$  possesses at least one indexing (3.1) such that  $\mathbf{y}_j(\mathbf{x})$  is a Borel-measurable function for each fixed  $j = 1, 2, \dots, p$ , then  $\mathbf{y}(\mathbf{x})$  is called a *many-valued Borel-measurable function*. In this paper we shall only be concerned with Borel-measurable  $\mathbf{y}(\mathbf{x})$ ; and we shall therefore assume that (3.1) is an indexing for which  $\mathbf{y}_j(\mathbf{x})$  is Borel-measurable for each fixed  $j$ .

We say that  $\mathbf{y}(\mathbf{x})$  is *continuous in a Borel set*  $\mathfrak{x}_0$  if, for every prescribed  $\varepsilon > 0$  and all points  $\mathbf{x} \in \mathfrak{x}_0$ , there exists  $\eta = \eta(\varepsilon, \mathbf{x}) > 0$  and at least one permutation  $1', 2', \dots, p'$  (possibly depending on  $\varepsilon, \mathbf{x}, \mathbf{x}'$ ) of the integers  $1, 2, \dots, p$  such that

$$\mathbf{x} \in \mathfrak{x}_0 \ \& \ |\mathbf{x} - \mathbf{x}'| < \eta \Rightarrow |\mathbf{y}_j(\mathbf{x}) - \mathbf{y}_{j'}(\mathbf{x}')| < \varepsilon, \quad j = 1, 2, \dots, p. \tag{3.2}$$

If further  $F[\mathfrak{x}_0] = 1$ , where  $F$  determines a random variable  $\mathbf{x}^*$ , we say that  $\mathbf{y}(\mathbf{x})$  is *almost-certainly-continuous with respect to*  $\mathbf{x}^*$ .

**Theorem 2.** *If  $\varepsilon > 0$  and  $\theta > 0$  are prescribed, and if  $F[\mathfrak{X}]$  is a probability set function, and if  $\mathbf{y}(\mathbf{x})$  is a  $p$ -valued Borel-measurable vector function, continuous in a Borel set  $\mathfrak{x}_0$ , then we can find a Borel set  $\hat{\mathfrak{x}}$ , satisfying  $F[\hat{\mathfrak{x}}] \geq (1 - \theta)F[\mathfrak{x}_0]$ , and a number  $\eta = \eta(\varepsilon, \theta)$ , independent of  $\mathbf{x}$ , such that*

$$\mathbf{x} \in \hat{\mathfrak{x}} \ \& \ |\mathbf{x} - \mathbf{x}'| < \eta \Rightarrow |\mathbf{y}_j(\mathbf{x}) - \mathbf{y}_{j'}(\mathbf{x}, \mathbf{x}')| < \varepsilon, \quad j = 1, 2, \dots, p$$

where  $y_j(\mathbf{x}, \mathbf{x}')$  is a Borel-measurable function of  $\mathbf{x}$  and  $\mathbf{x}'$  for each fixed  $j = 1, 2, \dots, p$ , and the set  $y_1(\mathbf{x}, \mathbf{x}'), y_2(\mathbf{x}, \mathbf{x}'), \dots, y_p(\mathbf{x}, \mathbf{x}')$  is a permutation (depending on  $\mathbf{x}$ ) of the set  $y_1(\mathbf{x}'), y_2(\mathbf{x}'), \dots, y_p(\mathbf{x}')$ .

When  $p = 1$  this theorem reduces to one on uniform continuity over the 'non-trivial' part of a probability set. Surprisingly enough, the standard textbooks on topological measure theory do not mention even this special case, which therefore seems new.<sup>1</sup>

Let  $\pi_k, k = 1, 2, \dots, p!$ , denote the permutations of  $p$  objects, and let  $(1k), (2k), \dots, (pk)$  denote the result of applying  $\pi_k$  to the integers  $1, 2, \dots, p$ . Let  $m$  be a positive integer. Let  $C_m$  denote the class (containing at least the empty Borel set) of all Borel sets  $\mathfrak{z}$  of  $\mathfrak{X}$  with the property

$$\mathbf{x} \in \mathfrak{z} \ \& \ |\mathbf{x} - \mathbf{x}'| < m^{-1} \Rightarrow |y_j(\mathbf{x}) - y_{(jk)}(\mathbf{x}')| < \varepsilon; \quad j = 1, 2, \dots, p; \text{ some } k; \quad (3.3)$$

where 'some  $k$ ' means that there is at least one value of  $k$  (possibly depending on  $\mathbf{x}$  and  $\mathbf{x}'$ ) such that (3.3) holds for all  $j$  with this fixed  $k$ . We notice first that  $C_m, m = 1, 2, \dots$ , is a monotone increasing collection of  $\sigma$ -rings [Halmos (2)]: that is to say

$$\mathfrak{z} \in C_m \ \& \ \mathfrak{z}' \in C_m \Rightarrow \mathfrak{z} - \mathfrak{z}' \in C_m; \quad (3.4)$$

$$\mathfrak{z}_s \in C_m, \quad s = 1, 2, \dots \Rightarrow \sum_{s=1}^{\infty} \mathfrak{z}_s \in C_m; \quad (3.5)$$

$$m < m' \Rightarrow C_m \subseteq C_{m'}. \quad (3.6)$$

Since  $F$  is a probability set function there exists  $M_m$ , the least upper bound of  $F[\mathfrak{z}]$  for  $\mathfrak{z} \in C_m$ . We have

$$\mathfrak{z} \in C_m \Rightarrow F[\mathfrak{z}] \leq M_m. \quad (3.7)$$

Moreover we can find  $\mathfrak{z}_{m\alpha} \in C_m, \alpha = 1, 2, \dots$ , such that  $F[\mathfrak{z}_{m\alpha}] \geq M_m - \alpha^{-1}$ . Write  $\mathfrak{z}_m = \sum_{\alpha=1}^{\infty} \mathfrak{z}_{m\alpha}$  and notice that  $\sum_{\alpha=1}^{\beta} \mathfrak{z}_{m\alpha}, \beta = 1, 2, \dots$ , is a monotone increasing sequence of Borel sets. Then

$$F[\mathfrak{z}_m] = F\left[\lim_{\beta \rightarrow \infty} \sum_{\alpha=1}^{\beta} \mathfrak{z}_{m\alpha}\right] = \lim_{\beta \rightarrow \infty} F\left[\sum_{\alpha=1}^{\beta} \mathfrak{z}_{m\alpha}\right] \geq \sup_{\beta} F[\mathfrak{z}_{m\beta}] \geq \sup_{\beta} (M_m - \beta^{-1}) = M_m. \quad (3.8)$$

But (3.5), (3.7), and (3.8) now show

$$\mathfrak{z}_m \in C_m, \quad F[\mathfrak{z}_m] = M_m. \quad (3.9)$$

When we have thus found  $\mathfrak{z}_m$  to satisfy (3.9) for each  $m = 1, 2, \dots$  we define

$$\mathfrak{z}^m = \sum_{\mu=1}^m \mathfrak{z}_{\mu}, \quad m = 1, 2, \dots, \infty. \quad (3.10)$$

<sup>1</sup> Professor Kac has remarked to me in conversation that the special case  $p = 1$  can be deduced from Lusin's theorem.

Now (3.5), (3.6), and (3.7) show  $F[\xi^m] \leq M_m$ ; and (3.9) and (3.10) show  $F[\xi^m] \geq M_m$ . So we have

$$\xi^m \in C_m, F[\xi^m] = M_m. \tag{3.11}$$

The definition of  $M_m$ , the fact that  $F$  is a probability set function, and (3.6) demonstrate

$$M_m \leq M_{m'} \leq 1, \quad m < m'; \tag{3.12}$$

so  $M = \lim_{m \rightarrow \infty} M_m$  exists. Further  $\xi^m, m = 1, 2, \dots$ , is a monotone increasing sequence of Borel sets. Thus

$$F[\xi^\infty] = F[\lim_{m \rightarrow \infty} \xi^m] = \lim_{m \rightarrow \infty} F[\xi^m] = \lim_{m \rightarrow \infty} M_m = M. \tag{3.13}$$

Since  $\xi_0$  is a Borel set by hypothesis,  $\xi_0 - \xi^\infty$  is a Borel set (perhaps empty). We shall show that the supposition

$$F[\xi_0 - \xi^\infty] > 0 \tag{3.14}$$

leads to a contradiction.

In the  $n$ -dimensional space  $\mathcal{X}$ , in which  $F[\xi]$  is defined, a bounded half-open set of all points  $\mathbf{x} = \{x_1, x_2, \dots, x_n\}$  which satisfy all the inequalities

$$a_i - h < x_i \leq a_i + h, \quad a_i \text{ and } h \text{ finite, } i = 1, 2, \dots, n \tag{3.15}$$

is called a hypercube. Given a hypercube (3.15), the set of all points satisfying, for each value of  $i$ , one or other (but not both) of the inequalities

$$a_i - h < x_i \leq a_i \quad \text{or} \quad a_i < x_i \leq a_i + h$$

is called a first hyperquadrant of the hypercube (3.15). We then inductively define a  $(q + 1)$ th hyperquadrant of (3.15) as a first hyperquadrant of a  $q$ th hyperquadrant of (3.15). The unqualified term 'hyperquadrant' will mean a  $q$ th hyperquadrant for some unspecified positive integer  $q$ .

If (3.14) holds, we can find a hypercube  $\mathfrak{h}_0$  such that

$$F[(\xi_0 - \xi^\infty) \cdot \mathfrak{h}_0] > 0, \tag{3.16}$$

because  $F$  is a probability set function. Let  $\mathfrak{h}'$  denote the union of all hyperquadrants  $\mathfrak{h}$  of  $\mathfrak{h}_0$  which satisfy

$$F[(\xi_0 - \xi^\infty) \cdot \mathfrak{h}] = 0. \tag{3.17}$$

The set of hyperquadrants  $\mathfrak{h}$  satisfying (3.17) is at most enumerable, because it is a subset of the enumerable set of all hyperquadrants of  $\mathfrak{h}_0$ . Hence  $\mathfrak{h}'$  is a Borel set, and

$$F[(\xi_0 - \xi^\infty) \cdot \mathfrak{h}'] = 0.$$

Consequently, from (3.16)

$$F[(\xi_0 - \xi^\infty) \cdot (\eta_0 - \eta')] > 0. \quad (3.18)$$

Now (3.18) implies that  $(\xi_0 - \xi^\infty) \cdot (\eta_0 - \eta')$  is not empty. So we can choose a point  $\mathbf{x}_0$  (hereafter fixed) such that

$$\mathbf{x}_0 \in (\xi_0 - \xi^\infty) \cdot (\eta_0 - \eta'). \quad (3.19)$$

Since (3.19) implies  $\mathbf{x}_0 \in \eta_0$ , we may define  $\eta_q$  to be the  $q$ th hyperquadrant of  $\eta_0$  such that  $\mathbf{x}_0 \in \eta_q$ . This definition is unique, because, for each fixed  $q$ , the several  $q$ th hyperquadrants of  $\eta_0$  are mutually disjoint. Further

$$F[(\xi_0 - \xi^\infty) \cdot \eta_q] > 0, \quad q = 1, 2, \dots, \quad (3.20)$$

for otherwise  $\mathbf{x}_0 \in \eta_q \subseteq \eta'$  in contradiction to (3.19).

Next (3.19) implies  $\mathbf{x}_0 \in \xi_0$ ; so that  $\mathbf{y}(\mathbf{x})$  is continuous at  $\mathbf{x}_0$  by hypothesis. Therefore,  $\mathbf{x}_0$  being fixed, we can find a positive integer  $r = r(\mathbf{x}_0, \varepsilon) = r(\varepsilon)$  such that

$$|\mathbf{x} - \mathbf{x}_0| < 2r^{-1} \rightarrow |y_j(\mathbf{x}_0) - y_{(j,k)}(\mathbf{x})| < \frac{1}{2}\varepsilon; \quad j = 1, 2, \dots, p; \quad \text{some } k. \quad (3.21)$$

Let  $\delta$  denote the set of all points  $\mathbf{x}$  satisfying  $|\mathbf{x} - \mathbf{x}_0| < r^{-1}$ . If the value of  $h$  [see (3.15)] for  $\eta_q$  is  $h_q$ ,  $h_q = 2^{-q}h_0 \rightarrow 0$  as  $q \rightarrow \infty$  because  $h_0$  is finite. Hence we can choose a value of  $q$ , say  $q = t$ , so that  $\eta_t \subseteq \delta$ . Now let  $\mathbf{x}$  and  $\mathbf{x}'$  be any two points satisfying

$$\mathbf{x} \in (\xi_0 - \xi^\infty) \cdot \eta_t \quad \text{and} \quad |\mathbf{x} - \mathbf{x}'| < r^{-1}. \quad (3.22)$$

Then

$$\mathbf{x} \in \eta_t \subseteq \delta \Rightarrow |\mathbf{x} - \mathbf{x}_0| < r^{-1} \Rightarrow |\mathbf{x}' - \mathbf{x}_0| < 2r^{-1}.$$

*A fortiori*

$$|\mathbf{x} - \mathbf{x}_0| < 2r^{-1} \quad \text{and} \quad |\mathbf{x}' - \mathbf{x}_0| > 2r^{-1};$$

so that (3.21) shows that there exist integers  $k'$  and  $k''$  with  $1 \leq k', k'' \leq p!$  such that

$$|y_j(\mathbf{x}_0) - y_{(j,k')}(\mathbf{x})| < \frac{1}{2}\varepsilon, \quad |y_j(\mathbf{x}_0) - y_{(j,k'')}(\mathbf{x}')| < \frac{1}{2}\varepsilon, \quad j = 1, 2, \dots, p.$$

Whereupon

$$|y_{(j,k')}(\mathbf{x}) - y_{(j,k'')}(\mathbf{x}')| < \varepsilon, \quad j = 1, 2, \dots, p.$$

Now apply the inverse permutation  $\pi_k^{-1}$  to these last inequalities, and there results

$$|y_j(\mathbf{x}) - y_{(j,k)}(\mathbf{x}')| < \varepsilon; \quad j = 1, 2, \dots, p; \quad \text{some } k. \quad (3.23)$$

Since (3.22)  $\Rightarrow$  (3.23), we have from (3.3)

$$(\xi_0 - \xi^\infty) \cdot \eta_t \in C_r;$$

and therefore by (3.5)

$$\xi^r + (\xi_0 - \xi^\infty) \cdot \eta_t \in C_r. \tag{3.24}$$

Now  $\xi^r \subseteq \xi^\infty$ ; so  $\xi^r$  and  $(\xi_0 - \xi^\infty) \cdot \eta_t$  are mutually disjoint. Therefore, by (3.24), (3.7), (3.20), and (3.11)

$$M_r \geq F[\xi^r + (\xi_0 - \xi^\infty) \cdot \eta_t] = F[\xi^r] + F[(\xi_0 - \xi^\infty) \cdot \eta_t] > F[\xi^r] = M_r,$$

which is the required contradiction. So we must abandon the supposition (3.14); and there only remains the possibility

$$F[\xi_0 - \xi^\infty] = 0.$$

Consequently

$$F[\xi_0] \leq F[\xi^\infty] = M = \lim_{m \rightarrow \infty} F[\xi^m].$$

This last equation shows that we can find an integer  $s$  such that

$$(1 - \theta) F[\xi_0] \leq F[\xi^s]. \tag{3.25}$$

We now choose  $\eta = \eta(\varepsilon, \theta)$  to satisfy  $0 < \eta < s^{-1}$ , and put  $\hat{\xi} = \xi^s$ . We have

$$F[\hat{\xi}] \geq (1 - \theta) F[\xi_0], \tag{3.26}$$

and, as a stronger case of (3.3)

$$\mathbf{x} \in \hat{\xi} \ \& \ |\mathbf{x} - \mathbf{x}'| < \eta \Rightarrow |y_j(\mathbf{x}) - y_{(j)k}(\mathbf{x}')| < \varepsilon; \quad j = 1, 2, \dots, p; \quad \text{some } k. \tag{3.27}$$

Let  $\mathfrak{B} = \mathfrak{X} \times \mathfrak{X}'$  denote the space of points  $\mathbf{z} = (\mathbf{x}, \mathbf{x}')$ ; and let  $\mathbf{R}_\lambda, \lambda = 1, 2, \dots$ , be an enumeration of the rational points of  $\mathfrak{Y}$ . For each value of  $k = 1, 2, \dots, p!$ , let  $\mathfrak{z}_k$  denote the (possibly empty) set of points  $\mathbf{z} = (\mathbf{x}, \mathbf{x}')$  which satisfy all the inequalities

$$|y_j(\mathbf{x}) - y_{(j)k}(\mathbf{x}')| < \varepsilon, \quad j = 1, 2, \dots, p; \tag{3.28}$$

and let  $\mathfrak{z}$  denote the set of points for which (3.28) holds for some  $k$  (perhaps depending on  $\mathbf{x}$  and  $\mathbf{x}'$ ). Then

$$\mathfrak{z} = \sum_{k=1}^{p!} \mathfrak{z}_k.$$

The set of points  $\mathbf{x}$ , satisfying  $|y_j(\mathbf{x}) - \mathbf{R}_\lambda| < \frac{1}{2}\varepsilon$  for fixed  $j$  and fixed  $\lambda$ , is a Borel set  $\xi_{j\lambda}$  since  $y_j(\mathbf{x})$  is a Borel-measurable function of  $\mathbf{x}$ . Similarly the set of points  $\mathbf{x}'$ , satisfying  $|y_{(j)k}(\mathbf{x}') - \mathbf{R}_\lambda| < \frac{1}{2}\varepsilon$  for fixed  $j$  and fixed  $k$  and fixed  $\lambda$ , is a Borel set  $\xi'_{(j)k\lambda}$ . Consequently

$$\mathfrak{z}_k = \prod_{j=1}^p \left\{ \sum_{\lambda=1}^{\infty} (\xi_{j\lambda} \times \xi'_{(j)k\lambda}) \right\}$$

is a Borel set. We define

$$z^\kappa = z_\kappa - \sum_{k=1}^{\kappa-1} z_k, \quad \kappa = 2, 3, \dots, p!, \quad z^1 = z - \sum_{\kappa=2}^{p!} z^\kappa.$$

Then  $z^\kappa$ ,  $\kappa = 1, 2, \dots, p!$ , are mutually disjoint Borel sets covering  $\mathfrak{B}$ ; so that their characteristic functions

$$\chi_\kappa(\mathbf{x}, \mathbf{x}') = \begin{cases} 1 & \text{if } \mathbf{z} = (\mathbf{x}, \mathbf{x}') \in z^\kappa \\ 0 & \text{if } \mathbf{z} = (\mathbf{x}, \mathbf{x}') \notin z^\kappa \end{cases}$$

are Borel-measurable functions of  $\mathbf{x}$  and  $\mathbf{x}'$ . The theorem is now proved by taking

$$y'_j(\mathbf{x}, \mathbf{x}') = \sum_{\kappa=1}^{p!} \chi_\kappa(\mathbf{x}, \mathbf{x}') y_{(j\kappa)}(\mathbf{x}'), \quad j = 1, 2, \dots, p. \quad (3.29)$$

The case of Theorem 2 which will interest us in this paper arises when  $\mathbf{y}(\mathbf{x})$  is almost-certainly-continuous, and we have  $F[\hat{\mathfrak{z}}] \geq 1 - \theta$ . The counter-example

$$y(x) = x^{-1}, \quad F(x) = \begin{cases} 0 & \text{if } x < 0 \\ x & \text{if } 0 \leq x \leq 1 \\ 1 & \text{if } x > 1 \end{cases}$$

shows that Theorem 2 would be false were  $\theta = 0$  permitted.

#### § 4. Many-valued random variables.

An unordered set of  $p$  probability set functions determines a  $p$ -valued random variable,  $\mathbf{x}^*$ . Suppose that, for any given Borel set  $\mathfrak{z}$ , these  $p$  probability set functions are arranged in an arbitrary order and then denoted by  $F_j[\mathfrak{z}]$ ,  $j = 1, 2, \dots, p$ . This indexing may depend in general upon the set  $\mathfrak{z}$  chosen; but the symmetric sum

$$F[\mathfrak{z}] = p^{-1} \sum_{j=1}^p F_j[\mathfrak{z}]$$

is evidently independent of the indexing. It is moreover easy to verify that  $F[\mathfrak{z}]$  is a probability set function, which we call the *condensed probability set function* of  $\mathbf{x}^*$ . The corresponding *condensed cumulative distribution function* is

$$F(\mathbf{x}) = p^{-1} \sum_{j=1}^p F_j(\mathbf{x})$$

where  $F_j(\mathbf{x})$  are specified by an indexing of the  $p$  cumulative distribution functions

of  $\mathbf{x}^*$ . Further,  $F[\mathfrak{X}]$  and  $F(\mathbf{x})$  determine the same one-valued random variable, which we call the *condensation* of  $\mathbf{x}^*$  and denote by  ${}^c\mathbf{x}^*$ .

If  $\mathbf{x}^*$  is a one-valued random variable, and  $\mathbf{y}(\mathbf{x})$  is a  $p$ -valued Borel-measurable function with an indexing (3.1) of one-valued Borel-measurable functions, then

$$y_j^* = y_j(\mathbf{x}^*), \quad j = 1, 2, \dots, p$$

will be  $p$  one-valued random variables determined by  $H_j[\mathfrak{Y}]$  say. Whereupon

$$H[\mathfrak{Y}] = p^{-1} \sum_{j=1}^p H_j[\mathfrak{Y}]$$

will be the condensed probability set function of the many-valued random variable  $\mathbf{y}(\mathbf{x}^*)$ , and will determine a condensation denoted by  ${}^c\mathbf{y}(\mathbf{x}^*)$ .

**Theorem 3.** *If  $\mathbf{x}^*$  and  $\mathbf{x}_\nu^*$ ,  $\nu = 1, 2, \dots$ , are one-valued random variables satisfying*

$$\text{dlim}_{\nu \rightarrow \infty} \mathbf{x}_\nu^* = \mathbf{x}^*, \tag{4.1}$$

*and if  $\mathbf{y}(\mathbf{x})$  is a many-valued Borel-measurable function which is almost-certainly-continuous with respect to  $\mathbf{x}^*$ , then*

$$\text{dlim}_{\nu \rightarrow \infty} {}^c\mathbf{y}(\mathbf{x}_\nu^*) = {}^c\mathbf{y}(\mathbf{x}^*). \tag{4.2}$$

It will be noticed that  $\mathbf{y}(\mathbf{x})$  need not be almost-certainly-continuous with respect to  $\mathbf{x}_\nu^*$  for any value of  $\nu$  at all.

Suppose that  $F[\mathfrak{X}]$  and  $F_\nu[\mathfrak{X}]$  are the probability set functions determining  $\mathbf{x}^*$  and  $\mathbf{x}_\nu^*$  respectively. Let  $\varepsilon > 0$  and  $\theta > 0$  be any pair of prescribed positive numbers. Since  $\mathbf{y}(\mathbf{x})$  is almost-certainly-continuous with respect to  $\mathbf{x}^*$  we can find a Borel set  $\hat{\mathfrak{X}}$ , satisfying

$$F[\hat{\mathfrak{X}}] \geq 1 - \frac{1}{2}\theta, \tag{4.3}$$

and a number  $\eta = \eta(\varepsilon, \theta)$  such that

$$\mathbf{x} \in \hat{\mathfrak{X}} \ \& \ |\mathbf{x} - \mathbf{x}'| < \eta \Rightarrow |y_j(\mathbf{x}) - y_j(\mathbf{x}, \mathbf{x}')| < \varepsilon, \quad j = 1, 2, \dots, p \tag{4.4}$$

where  $y_j(\mathbf{x}, \mathbf{x}')$  is defined by (3.29).

Let  $\mathfrak{Z} = \mathfrak{X} \times \mathfrak{X}'$  be the space of points  $\mathbf{z} = (\mathbf{x}, \mathbf{x}')$ . By (4.1), we can take the probability set function  $G[\mathfrak{Z}]$  defined in Theorem 1 to be a joint determination of  $\mathbf{x}^*$  and  $\mathbf{x}_\nu^*$ . We put the quantity  $\delta$  of Theorem 1 equal to  $\eta(\varepsilon, \theta)$ , and the quantity  $\varepsilon$  of Theorem 1 equal to the quantity  $\frac{1}{2}\theta$  of the present theorem. Then, with  $\mathfrak{z}_0$  denoting the set of points  $\mathbf{z} = (\mathbf{x}, \mathbf{x}')$  satisfying  $|\mathbf{x} - \mathbf{x}'| < \eta$ , we have from Theorem 1

$$G[\mathfrak{X} \times \mathfrak{X}] = F[\mathfrak{X}]; \quad G[\mathfrak{X} \times \mathfrak{X}'] = F_\nu[\mathfrak{X}']; \quad G[\mathfrak{z}_0] > 1 - \frac{1}{2}\theta, \quad \nu \geq \nu'_0(\varepsilon, \theta); \quad (4.5)$$

where  $\nu'_0(\varepsilon, \theta) = \nu_0\{\eta(\varepsilon, \theta), \frac{1}{2}\theta\}$ . Then, with  $\hat{\mathfrak{z}} = \hat{\mathfrak{x}} \times \mathfrak{X}'$  we have

$$G[\hat{\mathfrak{z}}] = F[\hat{\mathfrak{x}}] \geq 1 - \frac{1}{2}\theta. \quad (4.6)$$

Then (4.5) and (4.6) show that

$$\begin{aligned} G[\mathfrak{z}_0 \cdot \hat{\mathfrak{z}}] &= 1 - G[\mathfrak{z} - (\mathfrak{z}_0 \cdot \hat{\mathfrak{z}})] = 1 - G[(\mathfrak{z} - \mathfrak{z}_0) + (\mathfrak{z} - \hat{\mathfrak{z}})] \\ &\geq 1 - G[\mathfrak{z} - \mathfrak{z}_0] - G[\mathfrak{z} - \hat{\mathfrak{z}}] = -1 + G[\mathfrak{z}_0] + G[\hat{\mathfrak{z}}] > 1 - \theta, \end{aligned} \quad (4.7)$$

while (4.4) becomes

$$\mathbf{z} = (\mathbf{x}, \mathbf{x}') \in \mathfrak{z}_0 \cdot \hat{\mathfrak{z}} \Rightarrow |\mathbf{y}_j(\mathbf{x}) - \mathbf{y}'_j(\mathbf{x}, \mathbf{x}')| < \varepsilon, \quad j = 1, 2, \dots, p. \quad (4.8)$$

Now, since  $\varepsilon$  and  $\theta$  are arbitrary, (4.7) and (4.8) imply

$$\text{plim}_{\nu \rightarrow \infty} \{\mathbf{y}_j(\mathbf{x}^*) - \mathbf{y}'_j(\mathbf{x}^*, \mathbf{x}_*^*)\} = \mathbf{0}, \quad j = 1, 2, \dots, p$$

and, since (2.7)  $\Rightarrow$  (2.3), we have

$$\text{dlim}_{\nu \rightarrow \infty} \mathbf{y}'_j(\mathbf{x}^*, \mathbf{x}_*^*) = \mathbf{y}_j(\mathbf{x}^*), \quad j = 1, 2, \dots, p.$$

Summing the corresponding cumulative distribution functions over all values of  $j$ , and remembering that a distribution function has at most an enumerable number of discontinuities, we deduce without difficulty

$$\text{dlim}_{\nu \rightarrow \infty} \mathbf{y}'(\mathbf{x}^*, \mathbf{x}_*^*) = \mathbf{y}(\mathbf{x}^*). \quad (4.9)$$

We complete the proof by showing that  $\mathbf{y}'(\mathbf{x}^*, \mathbf{x}_*^*)$  has the same condensed probability set function as  $\mathbf{y}(\mathbf{x}^*)$ . Let  $\eta$  denote a typical Borel set of the space  $\mathfrak{Y}$  of points  $\mathbf{y}$ . Let  $\mathfrak{z}'_j(\eta)$  denote the set of all points  $\mathbf{z} = (\mathbf{x}, \mathbf{x}')$  such that  $\mathbf{y}'_j(\mathbf{x}, \mathbf{x}') \in \eta$ ; and let  $\mathfrak{z}_j(\eta)$  denote the set of all  $\mathbf{z}$  such that  $\mathbf{y}_j(\mathbf{x}') \in \eta$ . Since  $\mathbf{y}_j(\mathbf{x}')$  and  $\mathbf{y}'_j(\mathbf{x}, \mathbf{x}')$  are Borel-measurable functions  $\mathfrak{z}_j(\eta)$  and  $\mathfrak{z}'_j(\eta)$  are Borel sets. The condensed probability set function of  $\mathbf{y}'(\mathbf{x}^*, \mathbf{x}_*^*)$  is

$$p^{-1} \sum_{j=1}^p G[\mathfrak{z}'_j(\eta)] = p^{-1} \sum_{j=1}^p \sum_{\kappa=1}^{p!} G[\mathfrak{z}^{\kappa} \cdot \mathfrak{z}'_j(\eta)], \quad (4.10)$$

where  $\mathfrak{z}^{\kappa}$ ,  $\kappa = 1, 2, \dots, p!$ , are the disjoint Borel sets covering  $\mathfrak{z}$  defined in Theorem 2. If  $\mathbf{z} \in \mathfrak{z}^{\kappa} \cdot \mathfrak{z}'_j(\eta)$ , (3.29) shows  $\mathbf{y}'_j(\mathbf{x}, \mathbf{x}') = \mathbf{y}_{(j \kappa)}(\mathbf{x}')$ ; and so

$$\mathfrak{z}^{\kappa} \cdot \mathfrak{z}'_j(\eta) = \mathfrak{z}^{\kappa} \cdot \mathfrak{z}_{(j \kappa)}(\eta).$$

Substituting into (4.10), we get

$$\begin{aligned}
 p^{-1} \sum_{j=1}^p G[\beta_j'(\eta)] &= p^{-1} \sum_{\kappa=1}^{p!} \sum_{j=1}^p G[\beta^{\kappa} \cdot \beta_{(j \kappa)}(\eta)] = p^{-1} \sum_{\kappa=1}^{p!} \sum_{j=1}^p G[\beta^{\kappa} \cdot \beta_j(\eta)] \\
 &= p^{-1} \sum_{j=1}^p \sum_{\kappa=1}^{p!} G[\beta^{\kappa} \cdot \beta_j(\eta)] = p^{-1} \sum_{j=1}^p G[\beta_j(\eta)],
 \end{aligned}$$

which is the condensed probability set function of  $\mathbf{y}(\mathbf{x}^*)$ .

### § 5. References.

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