RELATIONS AMONG CERTAIN RANGES OF VECTOR MEASURES

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1. Introduction and definitions. The purpose of the present paper is to prove certain measure theoretical results concerning ranges of measures. One of our results (the closure and convexity result implied by Theorem 4) may be regarded as a generalization of a theorem of Liapounoff [5]. The results obtained here have applications to statistics and the theory of games.

Throughout this paper $\{x\} = X$ denotes an arbitrary space, and $\{S\} \equiv \mathbb{S}$ denotes a Borel field of subsets of X; that is, \mathbb{S} is a nonempty family of subsets of X which is closed with respect to the operations of complementation (with respect to X) and countable union. The phrase, S is measurable, will be used as synonymous with $S \in \mathbb{S}$.

A real-valued countably additive set function defined for all measurable sets will be called a *measure*. Thus we admit measures assuming negative or infinite values. A measure cannot, however, assume the value $+\infty$ for one measurable set and $-\infty$ for another such set, since in such a case additivity cannot be defined satisfactorily. A measure is called *finite* if it assumes finite values for all measurable sets. It is called *nonnegative* if it assumes nonnegative values for all such sets.

We say that f(x) is a measurable function if it is real-valued, defined for all $x \in X$, and if, moreover, the set f_c of all $x \in X$ for which f(x) < c is measurable for every real number c. A step function is a measurable function which assumes only a finite number of values.

If n is a positive integer and $\eta_j(x)$ ($j=1,\cdots,n$) are nonnegative measurable functions satisfying

(1)
$$\eta_1(x) + \cdots + \eta_n(x) = 1$$
 for every $x \in X$,

then $\eta(x) = [\eta_1(x), \dots, \eta_n(x)]$ will be called a probability n-vector. The functions $\eta_i(x)$ are called the components of this vector. If all the components of

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 $\eta(x)$ are step functions, then $\eta(x)$ is called a (probability) step n-vector. We shall occasionally denote such vectors by $\eta^0(x)$. If, in particular, all components of $\eta(x)$ assume only the two values zero and one, that is, if for every $x \in X$ one $\eta_j(x)$ is equal to one and all others vanish, then $\eta(x)$ is called a pure n-vector. Such vectors will be denoted by $\eta^*(x)$. If the jth component $(j = 1, \dots, n)$ of $\eta^*(x)$ is considered as the characteristic function of a set S_j , then the sets S_1, \dots, S_n are measurable and disjoint and their union is X. Conversely, if S_1, \dots, S_n is a decomposition of X into n disjoint measurable sets, and $\eta_j^*(x)$ is the characteristic function of S_j , then $\eta^*(x) = [\eta_1^*(x), \dots, \eta_n^*(x)]$ is a pure n-vector. We therefore call $\eta^*(x)$ also a decomposition n-vector or, more specifically, a decomposition n-vector corresponding to the decomposition $X = S_1 \cup \dots \cup S_n$.

Let $\mu_k(S)$ $(k=1,\dots,p)$ be a finite set of measures, and let $\eta(x)$ be a probability *n*-vector. We denote by $v(\eta) = v(\eta; \mu_1, \dots, \mu_p)$ the *np* dimensional vector (or point in *np* space),

$$\left[\int_X \eta_1(x) \ d\mu_1(x), \cdots, \int_X \eta_1(x) \ d\mu_p(x), \right.$$

$$\left. \int_X \eta_2(x) \ d\mu_1(x), \cdots, \int_X \eta_p(x) \ d\mu_p(x) \right].$$

The set of all points $v(\eta) = v(\eta; \mu_1, \dots, \mu_p)$ corresponding to all probability n-vectors $\eta(x)$ is called the n-range of μ_1, \dots, μ_p and is further denoted by $V_n(\mu_1, \dots, \mu_p)$ or, more concisely, by V_n . In the same way we define the step n-range of μ_1, \dots, μ_p as the set of all points $v(\eta^0) = v(\eta^0; \mu_1, \dots, \mu_p)$ corresponding to all step n-vectors $\eta^0(x)$, and denote it by $V_n^0(\mu_1, \dots, \mu_p)$ or V_n^0 . Similarly V_n^* or $V_n^*(\mu_1, \dots, \mu_p)$ denotes the set of all points

$$v(\eta^*) = v(\eta^*; \mu_1, \cdot \cdot \cdot, \mu_p)$$

corresponding to all pure *n*-vectors $\eta^*(x)$ and is called the *decomposition n-range* of μ_1, \dots, μ_p . When no confusion is possible we replace *n*-range in the above terms by range.

It is shown in Section 2 that if μ_1, \dots, μ_p are finite measures then the ranges $V_n(\mu_1, \dots, \mu_p)$ is compact and convex and coincides with the step-range $V_n^0(\mu_1, \dots, \mu_p)$. Actually a stronger result is proved; this states that the points $v(\eta; \mu_1, \dots, \mu_p)$ for which the components of $\eta(x)$ assume at most 2^{np-p+1}

different values already fill V_n . Applying a theorem of Liapounoff we deduce, in Section 3, the result that if the measures are atomless then the decomposition range $V_n^*(\mu_1, \dots, \mu_p)$ is identical with $V_n(\mu_1, \dots, \mu_p)$. This result is extended in Section 4 to arbitrary (not necessarily finite) atomless measures. Applications of these results to statistics and the theory of games are briefly indicated in Section 5.

2. Identity of the step range and the range for finite measures. First, we prove the following result.

THEOREM 1. If μ_1, \dots, μ_p are finite measures, then for every n, the range $V_n(\mu_1, \dots, \mu_p)$ is a compact and convex set in Euclidean np dimensional space.*

Proof. Let $A=v(\eta)$ and $A'=v(\eta')$ be any two points of V_n . Then every point of the segment joining them is represented vectorially by cA+(1-c)A', with 0 < c < 1. But such a point is clearly $v\left[c\eta+(1-c)\eta'\right]$ and, since $c\eta+(1-c)\eta'$ is a probability *n*-vector, the point also belongs to the range. Thus V_n is convex.

The proof of compactness is more difficult. We start by establishing a lemma on sequences of measures.

Lemma 1. Let $\{B\} = \mathcal{B}$ be a Borel field of subsets of X generated by countably many sets. Let μ^t $(t = 1, 2, \cdots)$ and μ be measures over \mathcal{B} satisfying, for all $B \in \mathcal{B}$,

$$(2) 0 \leq \mu^{t}(B) \leq \mu(B) < \infty (t = 1, 2, \cdots).$$

Then there exists a measure ν over \mathcal{B} satisfying

(3)
$$0 < \nu(B) < \mu(B) \quad \text{for all } B \in \mathfrak{B},$$

and a sequence of integers t_q $(q = 1, 2, \cdot \cdot \cdot)$ satisfying

$$(4) 0 < t_1 < t_2 < \cdots < t_q < t_{q+1} < \cdots$$

such that

(5)
$$\lim_{q=\infty} \mu^{t_q}(B) = \nu(B) \qquad \text{for every} \quad B \in \mathcal{B}.$$

^{*}For the special case when X is a finite-dimensional Euclidean space and all the μ_k are absolutely continuous, this theorem follows from Theorems 3.1 and 3.2a of [6].

The proof of the lemma proceeds as follows. Let $B_0 = X$, B_1 , \cdots , $B_n \cdots$ be a countable basis of \mathfrak{B} . Then, according to the well-known diagonal procedure of Cantor, there exists a sequence (4) for which

$$\beta_r = \lim_{q=\infty} \mu^{t_q}(B_r)$$

exists for $r=0,1,2,\cdots$. To prove the existence of the limit in (5) for every B, and the fact that this limit is a measure, it suffices to show that: (a) if $\mu^t q(B)$ tends to a limit as $q \longrightarrow \infty$, then so does $\mu^t q(\overline{B})$ where \overline{B} is the complement of B with respect to X; and (b) if $B^s(s=1,2,\cdots)$ are disjoint sets of $\mathcal B$ for which $\lim_{q\to\infty} \mu^t q(B^s)$ exists for $s=1,2,\cdots$, then we have also

(6)
$$\lim_{q=\infty} \mu^{t_q} \begin{pmatrix} \omega \\ 0 \\ s=1 \end{pmatrix} = \sum_{s=1}^{\infty} \lim_{q=\infty} \mu^{t_q}(B^s).$$

Now (a) follows immediately when we write $\mu^t(\overline{B}) = \mu^t(X) - \mu^t(B)$ and observe that $\mu^t q(X)$ has the limit β_0 . To prove (6) it is sufficient to observe that the functions μ^t are countably additive, that by (2) we have $\mu^t q(B^s) \leq \mu(B^s)$, and that $\sum_{s=1}^{\infty} \mu(B^s)$ is a convergent series of nonnegative terms. (This is the standard bounded convergence argument.) Since (2) and (5) obviously imply (3), the proof of Lemma 1 is completed.

Let now $\eta^t(x)$ $(t=1,2,\cdots)$ be any sequence of probability *n*-vectors. The compactness of V_n will be proved if we show that there exist a probability *n*-vector and a sequence (4) satisfying

(7)
$$\lim_{q=\infty} \int_{X} \eta_{j}^{t_{q}}(x) d\mu_{k}(x) = \int_{X} \eta_{j}(x) d\mu_{k}(x) \qquad (j = 1, \dots, n; k = 1, \dots, p).$$

Denote by $B_{j,\rho}^t(t=1,2,\cdots;j=1,\cdots,n;\ \rho$ rational with $0\leq\rho\leq 1)$ the set of all x for which $\eta_j^t(x)\leq\rho$, and let $\{B\}=\mathfrak{B}\subset\mathfrak{S}$ be the smallest Borel field containing these sets. Write $|\mu_k|$ for the absolute measure* associated with μ_k . Put $\mu(B)=|\mu_1|(B)+\cdots+|\mu_p|(B)$ for every $B\in\mathfrak{B}$. Then \mathfrak{B} , μ , and

^{*}That is, $|\mu_k|(S) = \sup [|\mu_k(S')| + |\mu_k(S'')|]$ for all decompositions of S into two disjoint measurable sets S' and S''.

 μ^t , defined by

$$\mu^{t}(B) = \int_{B} \eta_{1}^{t}(x) d\mu(x) \qquad (t = 1, 2, \cdots, B \in \mathcal{B}),$$

satisfy the conditions of Lemma 1. Hence there exists a nonnegative measure ν_1 over \mathcal{B} and a sequence (4), for which

(8)
$$\lim_{q=\infty} \int_{B} \eta_{1}^{t_{q}}(x) d\mu(x) = \nu_{1}(B)$$

for every $B \in \mathcal{B}$.

Again applying Lemma 1, we can extract from the sequence t_q a further subsequence for which (8) holds with the subscript 1 replaced by 2. Repeating this n-1 times, and again denoting, for simplicity of writing, the final subsequence by t_q , we see that there exist nonnegative measures ν_1, \dots, ν_n over \mathcal{B} and a sequence (4) satisfying

(9)
$$\lim_{q \to \infty} \int_{B} \eta_{j}^{tq}(\mathbf{x}) d\mu(\mathbf{x}) = \nu_{j}(B) \qquad (j = 1, \dots, n)$$

for every $B \in \mathfrak{B}$. Clearly, we have

(10)
$$\nu_1(B) + \cdots + \nu_n(B) = \mu(B) \qquad (B \in \mathcal{B}).$$

By the Radon-Nikodym theorem there exist \mathcal{B} -measurable functions $f_j(x)$ $(j=1,\cdots,n)$ such that

(11)
$$\nu_j(B) = \int_B f_j(\mathbf{x}) d\mu(\mathbf{x}) \qquad (j = 1, \dots, n)$$

for every $B \in \mathcal{B}$. Since the ν_j are nonnegative measures, we may assume that the f_j are nonnegative functions; and, because of (10), we may further assume that $f_1(x) + \cdots + f_n(x) = 1$ for every x. The f_j are \mathcal{B} -measurable and are, a fortiori, \mathcal{E} -measurable; hence $[f_1(x), \cdots, f_n(x)]$ is a probability n-vector. We denote this vector by $\eta(x)$ and proceed to show that (7) holds with this η and the above constructed sequence (4) satisfying (9).

Let $g_k(x)(k=1,\dots,p)$ denote a \mathcal{B} -measurable Radon-Nikodym derivative

 $d\mu_k(x)/d\mu(x)$. Then, replacing f_i in (11) by η_i , we have

$$\begin{split} \int_{X} \, \eta_{j}(x) \, d\mu_{k}(x) &= \int_{X} \, \eta_{j}(x) \, g_{k}(x) d\mu(x) \\ \\ &= \int_{X} \, g_{k}(x) \, d \int \eta_{j}(x) \, d\mu(x) = \int_{X} \, g_{k}(x) \, d\nu_{j}(x) \, . \end{split}$$

Similarly, the left side of (7) may be rewritten as

$$\lim_{q=\infty} \int_{X} g_{k}(x) \ d \int \eta_{j}^{t_{q}}(x) \ d \mu(x) ,$$

and thus (7) follows from (9). This completes the proof of Theorem 1.

For any compact convex set C in a Euclidean space, we designate as extreme points of C, all those points of C which are not interior points of any segment lying in C. Our next result is the following.

THEOREM 2. If the measures μ_1, \dots, μ_p are finite, and $v(\eta)$ is an extreme point of V_n , then the set of x for which $0 < \eta_j(x) < 1$ for at least one $j(j = 1, \dots, n)$ is a null-set* for each of the measures μ_1, \dots, μ_p . In particular, all extreme points of V_n belong to the decomposition range V_n^* .

Proof. Let Y denote the set of x defined in the theorem. If Y is not a null-set for μ_{k_0} with $1 \le k_0 \le p$, then there exist integers j_0 , j_1 with $1 \le j_0 < j_1 \le n$, a number $\delta > 0$, and a measurable set $Z \subset Y$, such that

(12)
$$\delta < \eta_I(x) < 1 - \delta$$
 for $x \in Z$ and $j = j_0, j_1$,

and

$$\mu_{k_0}(Z) \neq 0$$
.

Let $\zeta = \zeta(x) = [\zeta_1(x), \cdots, \zeta_n(x)]$ be the vector defined as follows:

$$\zeta_{j_0}(x) = -\zeta_{j_1}(x) = \begin{cases} \delta & \text{if } x \in Z \\ 0 & \text{if } x \notin Z \end{cases}$$

^{*}A measurable S is a null-set for the measure μ if $\mu(S') = 0$ for every measurable $S' \subset S$.

and all other components vanish identically.

Because of (12), $\eta(x) + \theta \zeta(x)$ is a probability *n*-vector whenever $-1 \le \theta \le 1$. Since

$$\int_{X} \left[\eta_{j_{0}}(x) + \zeta_{j_{0}}(x) \right] d\mu_{k_{0}}(x) - \int_{X} \left[\eta_{j_{0}}(x) - \zeta_{j_{0}}(x) \right] d\mu_{k_{0}}(x)$$

$$=2\delta\mu_{k_0}(Z)\neq 0,$$

the points $v(\eta + \zeta)$ and $v(\eta - \zeta)$ are different. But clearly as θ increases from -1 to +1 the point $v(\eta + \theta\zeta)$ moves from $v(\eta - \zeta)$ to $v(\eta + \zeta)$ along the segment connecting them. Moreover, $v(\eta)$ is the middle point of this segment and thus it cannot be an extreme point of V_n .

If $v(\eta)$ is an extreme point, then Y is a null-set for all μ_k . Therefore, if we put $\eta^*(x) = \eta(x)$ for $x \notin Y$ and, say, $\eta_1^*(x) = 1$ for $x \in Y$, the decomposition vector $\eta^*(x)$ thus defined satisfies $v(\eta^*) = v(\eta)$. This proves the last assertion of Theorem 2.

Theorem 3. If the measures μ_1 , \cdot \cdot , μ_p are finite, then the step-range V_n^0 coincides with the range V_n . More precisely, every point of V_n may be represented as $v(\eta^0)$, where η^0 is a step n-vector whose components assume not more than 2^{np-p+1} different values.

Proof. According to Theorem 1, V_n is a compact convex set in Euclidean n_p -dimensional space. However, because of the p equations

$$\sum_{j=1}^{n} \int_{X} \eta_{j}(x) \ d\mu_{k}(x) = \mu_{k}(x) \qquad (k = 1, \dots, p)$$

 V_n lies in an N=np-p dimensional linear subspace. Hence, according to well-known facts on convex bodies, every point P of V_n may be represented vectorially by $P=c_1P_1 + \cdots + c_NP_N + c_{N+1}P_{N+1},$

where P_1, \dots, P_{N+1} are extreme points of V_n and c_1, \dots, c_{N+1} are nonnegative constants whose sum is 1. According to Theorem 2, we have $P_r = v(\eta^*r)$ with η^*r a decomposition *n*-vector $(r = 1, 2, \dots, N+1)$.

Hence, putting $\eta^0 = \sum_{r=1}^{N+1} c_r \, \eta^{*r}$, we have $P = v(\eta^0)$. Clearly, for every x, every component of $\eta^0(x)$ equals $\sum_{r \in K} c_r$, where K is a subset of $\{1, 2, \dots, N+1\}$. There being 2^{N+1} such subsets, Theorem 3 is proved.

3. Identity of the range and the decomposition range for finite atomless measures. A measurable set S is called an atom of the measure μ if $\mu(S) \neq 0$ and if, moreover, for every measurable $S' \subset S$ we have either $\mu(S') = 0$ or $\mu(S') = \mu(S)$. If the measure $\mu(S)$ has no atoms it is called atomless.

For atomless measures we can improve on Theorem 3 by establishing the following result.

Theorem 4. If μ_1, \dots, μ_p are finite atomless measures then, for every n, the range V_n and the decomposition range V_n^* are identical.

According to Theorem 1, the common range is convex and compact.

Proof. In view of Theorem 3 it suffices to prove that, in the present case, $V_n^* = V_n^0$.

For this purpose we shall use the following fact: If μ_1, \dots, μ_p are finite and atomless, then, given $0 \le c \le 1$, there exists a measurable set S for which

(13)
$$\mu_k(S) = c \mu_k(X) \qquad (k = 1, \dots, p).$$

The existence of such a set S follows immediately from a result of Liapounoff [5] (see also [3]) according to which, under the above stated conditions, the set of points $\mu_1(S), \dots, \mu_p(S)$ in Euclidean p-space corresponding to all measurable S is convex. Indeed, the empty set Λ and X are certainly measurable and (1-c) $\mu_k(\Lambda) + c\mu_k(X) = c\mu_k(X)$ for all k.

To complete the proof of Theorem 4, we use the following lemma.

LEMMA 2. If μ_1, \dots, μ_p are finite and atomless and c_1, \dots, c_n are non-negative numbers satisfying $c_1 + \dots + c_n = 1$, then there exists a decomposition of X into n disjoint measurable sets S_1, \dots, S_n having the property that

(14)
$$\mu_k(S_j) = c_j \mu_k(X) \qquad (j = 1, \dots, n; k = 1, \dots, p).$$

Indeed, according to (13) there exists a measurable S_1 satisfying (14) for j=1. Similarly, there exists a measurable $S_2 \subset X-S_1$ satisfying

$$\mu_k(S_2) = \frac{c_2}{c_2 + \cdots + c_n} \ \mu_k(X - S_1) = c_2 \mu_k(X) ,$$

where we interpret

$$\frac{c_2}{c_2 + \cdots + c_n}$$

as zero if $c_2 = \cdots = c_n = 0$. That is, S_2 satisfies (14) for j = 2. In the same manner $S_j \subset X - \bigcup_{i=1}^{j-1} S_i$ satisfying (14) may be obtained for $j = 1, \cdots, n-1$. But then

$$\mu_k \left(X - \bigcup_{j=1}^{n-1} S_j \right) = 1 - (c_1 + \dots + c_{n-1}) \, \mu_k(X) = c_n \mu_k(X) ;$$

thus

$$S_n = X - \bigcup_{j=1}^{n-1} S_j$$

satisfies (14) for j = n as required. Hence, Lemma 2 holds.

The proof of Theorem 4 can now easily be completed. Let $\eta^0(x)$ be any step *n*-vector. Then X can be decomposed into a finite number of disjoint measurable subsets Y_t over each of which all the components of $\eta^0(x)$ are constant. According to Lemma 2, Y_t may be decomposed into n disjoint measurable sets $S_{1,t}, \dots, S_{n,t}$ such that

(15)
$$\mu_k(S_{j,t}) = \int_{Y_t} \eta_j^0(x) \ d\mu_k(x) \qquad (j = 1, \dots, n; k = 1, \dots, p) .$$

Putting $S_j = \bigcup_t S_{j,t} (j = 1, \dots, n)$ we have, from (15),

$$\int_{X} \eta_{j}^{0}(x) d\mu_{k}(x) = \mu_{k}(S_{j}) \qquad (j = 1, \dots, n; k = 1, \dots, p).$$

Thus the point $v(\eta^0; \mu_1, \dots, \mu_p)$ coincides with $v(\eta^*; \mu_1, \dots, \mu_p)$, where $\eta_j^*(x) = 1$ if $x \in S_j$ and zero otherwise. In other words, $V_n^0 \subset V_n^*$. Since the converse inclusion is obvious, Theorem 4 is proved.

Remarks. (a) Liapounoff [5] proved that if the conditions of Theorem 4 are satisfied then the set of all points $[\mu_1(S), \dots, \mu_p(S)]$ in Euclidean p-space

corresponding to all measurable S is convex and compact. This result is clearly implied by the convexity and compactness of V_n^* ; thus the convexity and compactness of V_n^* may be considered as a generalization of Liapounoff's theorem. If we put $\overline{S} = X - S$, then Liapounoff's result is easily seen to be equivalent to the statement that the set of all points $[\mu_1(S), \cdots, \mu_p(S), \mu_1(\overline{S}), \cdots, \mu_p(\overline{S})]$ in Euclidean 2p dimensional space is convex and compact. But this amounts precisely to the assertion that V_n^* is convex and compact for n = 2. That this assertion remains valid also for n > 2 is precisely the generalization of Liapounoff's result contained in Theorem 4.

We used in our proof the convexity part of Liapounoff's result. This is, however, the easier part (cf. Halmos [3]), and thus our method furnishes also a new proof of Liapounoff's theorem.

- (b) The values 0 and 1 are among those which the components of η^0 in Theorem 3 are allowed to assume. Hence, on combining the results of Theorems 3 and 4 we see that, if all but p' of the measures μ_1, \dots, μ_p are atomless, we may replace p by p' in the exponent of 2 in Theorem 3. This estimate is again independent of the number of atoms.
- (c) If the measures μ_1, \dots, μ_p in Theorem 4 are not assumed to be atomless, then of course V_n^* need not be convex. It is, however, compact as can easily be seen on decomposing into atomless and purely atomic parts and dealing separately with each (see, for example, [3]).
- (d) For some applications the following is of importance: If η is a probability n-vector, then there exists a decomposition n-vector η^* with $v(\eta^*) = v(\eta)$ having the further property that, for every $x \in X$ and $j = 1, \dots, n$, the vanishing of $\eta_j(x)$ implies that of $\eta_j^*(x)$. This assertion follows easily from Theorem 4. Indeed, X may be decomposed into a finite number of measurable sets Y with the following property: If $\eta_j(x) = 0$ for some $x \in Y$, then $\eta_j(x) = 0$ for all $x \in Y$. Let j_1, \dots, j_m be those j for which $\eta_j(x) > 0$ when $x \in Y$. We may now define $\eta^*(x)$ for $x \in Y$ by applying Theorem 4 (with X replaced by Y and n by m) to the m-vector formed by these components, and putting $\eta_j(x) = 0$ for all other j and $x \in Y$. Combining these definitions for all sets Y, we obtain an η^* with the required property.
- 4. Extension to arbitrary atomless measures. The assumption of finiteness in Theorem 4 is unnecessary. Indeed, we shall prove the following result.

THEOREM 5. If the measures μ_1, \dots, μ_p are atomless, then, for every n,

the range V_n and the decomposition range V_n^* are identical.

Since the measures are now allowed to assume infinite values, the components of $V(\eta)$ are no longer necessarily finite and one should look upon V_n and V_n^* as imbedded in Euclidean space extended by allowing each coordinate to assume also infinite values.

Before proceeding to the proof we establish the following lemmas.

Lemma 3. If μ is a nonnegative atomless measure with $\mu(X) = \infty$, and u is any finite positive number, then there exists a measurable set T with $\mu(T) = u$.

Proof. Since μ is nonnegative and atomless, there exists a set S with $0 < \mu(S) < \infty$. We first show that $\alpha = \sup \mu$ for all such sets S is infinite. Indeed, assume α finite; then, for every integer m, there exists a measurable S_m with $\mu(S_m) > \alpha - 1/m$. Put $S' = \bigcup_{m=1}^{\infty} S_m$; then $\mu(S') = \alpha$. But $\mu(X - S') = \infty$; hence, there exists a measurable $S'' \subset X - S'$ with $0 < \mu(S'') = b < \infty$. Thus $\alpha < \mu(S' \cup S'')$, contradicting the assumption that α is finite.

Therefore, given u there exists a measurable T' with $u < \mu(T') < \infty$. But then, according to the intermediary values theorem of Sierpinski (see, for example, [2,52]), or the one dimensional case of Liapounoff's theorem, there exists a measurable $T \subset T'$ with $\mu(T) = u$.

Lemma 4. If μ is a nonnegative atomless measure with $\mu(X) = \infty$, and q is any positive integer, then X may be decomposed into q measurable disjoint sets X_1, \dots, X_q with $\mu(X_1) = \dots = \mu(X_q) = \infty$.

Proof. According to Lemma 3, there exist a set T_1 with $\mu(T_1)=1$, a set $T_2 \subset X-T_1$ with $\mu(T_2)=1$, a set $T_3 \subset X-(T_1 \cup T_2)$ with $\mu(T_3)=1$, and so on. Putting $X_i=\bigcup_{n=0}^{\infty}T_{qn+i}$ for $i=1,\cdots,q-1$ and $X_q=X-\bigcup_{i=1}^{q-1}X_i$ we obtain the required result.

Lemma 5. If ν_1, \dots, ν_m are nonnegative atomless measures with $\nu_1(X) = \dots \nu_m(X) = \infty$, and q is any positive integer, then X may be decomposed into q measurable sets X_1, \dots, X_q satisfying $\nu_i(X_1) = \dots = \nu_i(X_q) = \infty$ for $i = 1, \dots, m$.

Proof. For m = 1, this lemma reduces to the preceding one. Assume m > 1 and the lemma proved for m - 1. According to Lemma 4, X is the union of m disjoint

measurable sets Y_1 , \cdots , Y_m with $\nu_m(Y_1) = \cdots = \nu_m(Y_m) = \infty$. For every $i (i = 1, \cdots, m-1)$ let i' denote the smallest integer for which $\nu_i(Y_{i'}) = \infty$. (Since $\nu_i(X) = \infty$ we have $1 \leq i' \leq m$.) Put $Y' = \bigcup_{i=1}^{m-1} Y_{i'}$ and Y'' = X - Y'. Then $\nu_m(Y'') = \infty$ and Y'' is the union of disjoint measurable sets Y_1'' , \cdots , Y_q'' with $\nu_m(Y_1'') = \cdots = \nu_m(Y_q'') = \infty$. Also $\nu_i(Y'') = \infty$ for $i = 1, \cdots, m-1$ and hence, by the assumption of induction, it can be decomposed into measurable sets Y_1' , \cdots , Y_q' with $\nu_i(Y_1') = \cdots = \nu_i(Y_q')$ for $i = 1, \cdots, m-1$. Putting $X_1 = Y_1' \cup Y_1''$, \cdots , $X_q = Y_q' \cup Y_q''$, we obtain the required decomposition.

Lemma 6. Let μ , ν be nonnegative atomless measures with $\mu(X) < \infty$, $\nu(X) = \infty$. Then either X may be decomposed into countably many measurable sets, each having finite ν measure, or there exists a measurable set T with $\mu(T) = 0$, $\nu(T) = \infty$.

Proof. For every positive integer t consider the measure μ_t defined by

$$\mu_t(S) = \nu(S) - t\mu(S) .$$

According to Hahn (see for example [2,p.18] or [4,p.121]) X may be decomposed into two disjoint measurable sets Y_t and \overline{Y}_t with $\mu_t(S) \leq 0$ for every measurable $S \subset Y_t$ and $\mu_t(S) > 0$ for every measurable $S \subset \overline{Y}$. Clearly,

$$\nu(Y_t) < t\mu(Y_t) < t\mu(X) < \infty$$
.

Put now $Y_t'=Y_1\cup\cdots\cup Y_t$ and $Z_1=Y_1'$, $Z_t=Y_t'-Y_{t-1}'$ for $t=2,3,\cdots$, and denote by Z_0 the complement of $\bigcup_{t=1}^{\infty}Z_t$. Then $X=\bigcup_{t=0}^{\infty}Z_t$ and $\nu(Z_t)<\infty$ for $t\geq 1$. If $\nu(Z_0)<\infty$ then this is a decomposition of X into countably many sets of finite ν measure. If, on the other hand, $\nu(Z_0)=\infty$ then, by Lemma 3, there exists for every integer u a measurable $T_u\subset Z_0$ with $\nu(T_u)=u$. Moreover, $\mu(T_u)=0$ since, according to the construction of Z_0 , $\mu(S)>0$ for $S\subset Z_0$ implies $\nu(S)=\infty$. Thus $T=\bigcup_{u=1}^{\infty}T_u$ has the properties required in Lemma 6.

Proof of Theorem 5. Since every measure is the difference between two nonnegative measures, we may assume throughout the proof that the measures $\mu_k(k=1,\cdots,p)$ are nonnegative.

Let η be any probability *n*-vector. For every $j(j=1,\dots,n)$ we denote by $Y_{j,0}$ the set of x for which $\eta_j(x)=0$ and by $Y_{j,t}(t=1,2,\dots)$ the set of x for which

$$\frac{1}{t+1} < \eta_j(x) < \frac{1}{t}.$$

We use Y to denote any set of the form

$$\bigcap_{j=1}^{n} Y_{jt_{j}} \quad \text{with } t_{j} = 0, 1, 2, \cdots \qquad (j = 1, \cdots, n) .$$

The space X is thus decomposed into countably many sets Y having the following property: There exists a nonempty subset J = J(Y) of $\{1, \dots, n\}$ and a positive $\delta = \delta(Y)$ such that for all $x \in Y$ we have

(16)
$$\eta_j(x) > \delta > 0$$
 if $j \in J$, $\eta_j(x) = 0$ if $j \notin J$.

Let Y be any such set and consider the subset K' of $\{1, \cdots, p\}$ consisting of all those k for which Y can be decomposed into countably many sets, all having finite μ_k measure. If K' is empty, we call Y final, if not we decompose Y into countably many measurable sets Y' with $\mu_k(Y') < \infty$ for $k \in K'$. Let Y' be any such set and denote by K'' the subset of $\{1, \cdots, p\}$ consisting of all k for which Y' can be decomposed into countably many sets, all having finite μ_k measure. Clearly, $K' \subset K''$. If K' = K'' we call Y' final, if not we decompose it into countably many Y'' with $\mu_k(Y'') < \infty$ for $k \in K''$. Again a $K''' \supset K''$ is defined and Y'' is called final if K'' = K''', and so on. After not more than p steps we always end with a final set Z.

We have thus decomposed Y, and hence X, into countably many sets Z having the following property: To every Z there corresponds a decomposition of $\{1,2,\cdots,p\}$ into two disjoint sets K and K such that $\mu_k(Z)<\infty$ if $k\in K$, while if $k\in \overline{K}$ then Z cannot be decomposed into countably many sets, all having finite μ_k measure. Furthermore, since Z is contained in some Y, (16) holds for all $x\in Z$.

Next, we show how to decompose Z into disjoint measurable sets Z_1, \dots, Z_n satisfying

(17)
$$\mu_k(Z_j) = \int_Z \eta_j(x) \ d\mu_k(x) \qquad (j = 1, \dots, n; k = 1, \dots, p) .$$

(If $\eta_j(x) = 0$ for all $x \in Z$, the right side of (17) is understood to be 0 even when $\mu_k(Z) = \infty$.)

If K is empty, then the possibility of such a decomposition is assured by

If K is empty then, by (16), the integral in (17) is infinite if $j \in J$ and is zero

otherwise. By Lemma 5 it is possible to decompose Z into sets Z_j $(j \in J)$ with

$$\mu_k(Z_1) = \infty$$
 for $k = 1, \dots, p$.

Denoting the empty set, for $j \notin J$, by Z_j , we have a decomposition satisfying (17). Finally, assume both K and \overline{K} nonempty. We define a nonnegative measure μ by $\mu(S) = \sum_{k \in K} \mu_k(S)$. Clearly, μ is atomless and $\mu(Z) < \infty$. According to Lemma 6 there exists, for every $k \in \overline{K}$, a measurable $T_k \subset Z$ with $\mu(T_k) = 0$, $\mu_k(T_k) = \infty$. Let T be the union of $T_k(k \in \overline{K})$. Then (see the treatment of the case when K is empty) it is possible to decompose T into disjoint measurable sets Z_1' , \cdots , Z_n' so that Z_j' is empty for $j \notin J$, while for all $j \in J$ and $k \in \overline{K}$ we have $\mu_k(Z_j') = \infty$. Since $\mu(T) = 0$ we have, for all j, $\mu_k(Z_j') = 0$ whenever $k \in K$. Let T' = Z - T; then it is possible, by Theorem 4, to decompose T' into disjoint measurable sets T_1' , \cdots , T_n' such that T_j' is empty for $j \notin J$, while for $j \in J$ and $k \in K$ we have

$$\mu_k(Y_j) = \int_{T'} \eta_j(x) \ d\mu_k(x) = \int_{Z} \eta_j(x) \ d\mu_k(x).$$

Putting $Z_j = T_j' \cup Z_j'$ for $j = 1, \dots, n$, we have a decomposition satisfying (17).

We now define the decomposition n-vector η^* as follows: For $x \in Z$, put $\eta_j^*(x) = 1$ if $x \in Z_j$, and $\eta_j^*(x) = 0$ for all other $x \in Z$. Because of the countable additivity of the measures and the integrals, (17) implies $v(\eta^*) = v(\eta)$ and the proof is completed.

Remarks. (a) The last remark after Theorem 4 applies also here. Indeed, our construction in the proof of Theorem 5 yields a vector having the properties required of η^* in that remark.

- (b) In applications usually X can be decomposed into countably many sets of finite μ_k measure $(k = 1, \dots, p)$. For this special case Theorem 5 is, of course, an immediate consequence of Theorem 4.
- 4. Application to statistics and the theory of games.* Theorem 4 (together with its extension mentioned in the last remark of the preceding section) has

^{*}A more detailed discussion and other results, including a discussion of the sequential statistical decision problem, are contained in our paper, Elimination of randomization in certain statistical decision procedures and zero-sum two-person games, Annals of Mathematical Statistics, 22, No. 1, March, 1951. A brief discussion of these applications was also given in an earlier publication [1].

immediate applications to the following statistical decision problem: Let $y = \{y_1, \cdots, y_t\}$ be a random vector with t components, where t is a given positive integer. For every point $x = (x_1, \cdots, x_t)$ of the t-dimensional Euclidean space X, let F(x) denote the probability that $y_i < x_i$ for $i = 1, \cdots, t$; that is, F(x) is the distribution function of y. The distribution function F(x) is assumed to be unknown. It is known, however, that F(x) is one of the distribution functions $F_1(x), \cdots, F_m(x)$. An observation x is made on y and according to the observed value x the statistician may adopt any one of x decisions x denote the loss sustained by the statistician when x is the true distribution of x, x is the observed value of x, and the x decision is adopted. y is assumed to be a finite nonnegative and measurable function of x. If the statistician, on observing the value x, adopts the various decisions with probabilities y is assumed to be a ronnegative measurable function satisfying (1), then the risk, or expected loss, when x is the true distribution function, is given by

$$r_{i}(\eta) = \sum_{j=1}^{n} \int_{X} W_{i,j}(x) \, \eta_{j}(x) \, dF_{i}(x)$$
.

The decision function $\eta_j(x)$ is said to be nonrandomized if for every x all but one of the $\eta_j(x)$ vanish. Theorem 4 yields without difficulty the following result: If the distribution functions $F_i(x)$ ($i=1,\cdots,m$) are atomless then, given any decision function $\eta(x)$, there exists a nonrandomized decision function $\eta^*(x)$ such that $r_i(\eta) = r_i(\eta^*)$ ($i=1,\cdots,m$).

Similar application can be made to the theory of games. In fact, the above described statistical decision problem may be interpreted as a zero-sum two-person game as follows: Player 1 has a finite number of pure strategies $i (i = 1, \dots, m)$, while a pure strategy of Player 2 is a nonrandomized decision function $\eta^*(x)$ (decomposition n-vector). If i is the pure strategy of Player 1 and $\eta^*(x)$ the pure strategy of Player 2, the outcome is defined by

$$R[i,\eta^*(x)] = r_i(\eta^*).$$

A mixed strategy of Player 1 is represented by a vector $\xi = (\xi_1, \dots, \xi_m)$ with nonnegative components whose sum is one, while a mixed strategy of Player 2 is given by a probability *n*-vector $\eta(x)$. The expected value of the outcome corresponding to the mixed strategies ξ and $\eta(x)$ is given by

$$R[\xi,\eta(x)] = \sum_{i=1}^{m} \xi_i r_i(\eta)$$
.

The above stated result for the statistical decision problem can be restated in game terminology as follows: If the distribution functions $F_i(x)$ ($i=1,\dots,m$) are atomless, then given any mixed strategy $\eta(x)$ of Player 2, there exists a pure strategy $\eta^*(x)$ such that $R[\xi, \eta^*(x)] = R[\xi, \eta(x)]$ for all strategies ξ of Player 1.

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