

A NOTE ON INFINITELY DIVISIBLE RANDOM VECTORS

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The purpose of this note is to show a characterization of infinitely divisible random vectors whose projection onto a subspace is normally distributed.

In this paper, \mathbf{x} will denote a k -dimensional vector of real numbers, \mathbf{x}' its transpose, $|\mathbf{x}|$ its length, $\xi' \mathbf{x}$ the inner product of the vectors ξ , \mathbf{x} , R^k the k -dimensional Euclidean space. By a normal distribution we understand either a proper or a singular normal distribution.

A nonnegative measure M defined on R^k is said to be canonical if it has no atom at the origin and $\int_{R^k} [1 + |\mathbf{x}|^2]^{-1} dM(\mathbf{x}) < \infty$.

For \mathbf{X} to be an infinitely divisible random vector, it is necessary and sufficient that there exist a canonical measure M , a vector of real numbers \mathbf{b} , and a nonnegative definite symmetric matrix $\Sigma = (\sigma_{ij})$ such that

$$(1) \quad \log E e^{i\xi' \mathbf{X}} = \int_{R^k} \frac{e^{i\xi' \mathbf{x}} - 1 - i\xi' \tau(\mathbf{x})}{|\mathbf{x}|^2} dM(\mathbf{x}) + i\xi' \mathbf{b} - \frac{1}{2} \xi' \Sigma \xi$$

where

$$\begin{aligned} \tau(x) &= -1 & x \leq -1 \\ &= x & -1 \leq x \leq +1, \\ &= 1 & 1 \leq x \end{aligned} \quad [\tau(\mathbf{x})]' = (\tau(x_1), \dots, \tau(x_k)).$$

This representation is unique (see [1] page 559).

Notice that if \mathbf{X} is normal, $\log E e^{i\xi' \mathbf{X}} = i\xi' \mathbf{b} - \frac{1}{2} \xi' \Sigma \xi$, and thus for normally distributed random vectors M is identically zero.

THEOREM. *Let $\mathbf{X}' = (X_1, \dots, X_k)$ be an infinitely divisible random vector, and $\xi_0 \neq \mathbf{0}$ an arbitrary vector in k -space. The distribution of $\xi_0' \mathbf{X}$ is one-dimensional normal if and only if the canonical measure M which corresponds to \mathbf{X} in the representation (1) is carried by the subspace $\{\mathbf{x} | \xi_0' \mathbf{x} = 0\}$.*

PROOF. We need only prove that if $\xi_0' \mathbf{X}$ is normal, then M is carried by $\{\mathbf{x} | \xi_0' \mathbf{x} = 0\}$.

Since for every infinitely divisible random vector \mathbf{X} and every matrix A , $A\mathbf{X}$ is infinitely divisible, we may assume without loss of generality that $\xi_0' = (1, 0, \dots, 0)$. Then:

$$(2) \quad \log E e^{it\xi_0' \mathbf{X}} = ib_1 t - \frac{1}{2} t^2 \sigma_{11} + \int_{R^k} \frac{e^{itx_1} - 1 - it\tau(x_1)}{|\mathbf{x}|^2} dM(\mathbf{x})$$

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$$\begin{aligned}
 (3) \quad \int_{R^k} \frac{e^{itx_1} - 1 - it\tau(x_1)}{|\mathbf{x}|^2} dM(\mathbf{x}) &= \int_{R^k} \frac{e^{itx_1} - 1 - it\tau(x_1)}{x_1^2} \frac{x_1^2}{|\mathbf{x}|^2} dM(\mathbf{x}) \\
 &= \int_{R^1} \frac{e^{itx_1} - 1 - it\tau(x_1)}{x_1^2} dM_1(x_1)
 \end{aligned}$$

where

$$(4) \quad dM_1(x_1) = \int_{x_1 \times R^{k-1}} \frac{x_1^2}{|\mathbf{x}|^2} dM(\mathbf{x}).$$

Clearly,

$$\begin{aligned}
 \int_{R^1} \frac{1}{1+x_1^2} dM_1(x_1) &= \int_{R^k} \frac{x_1^2}{1+x_1^2} \frac{1}{|\mathbf{x}|^2} dM(\mathbf{x}) \\
 &\leq M\{\mathbf{x} \mid \max_{1 \leq i \leq k} |x_i| \leq 1\} \\
 &\quad + \int_{R^{k-|\mathbf{x} \mid \max_{1 \leq i \leq k} |x_i| \leq 1}} \frac{1}{|\mathbf{x}|^2} dM(\mathbf{x}) \\
 &< \infty.
 \end{aligned}$$

Define:

$$(5) \quad M_2(S) = M_1(S - \{0\}), \quad S \text{ being any measurable subset of } R^1.$$

Thus, M_2 is a canonical measure defined on R^1 . From (2) and (3) we have:

$$\begin{aligned}
 \log Ee^{it\epsilon_0'x} &= ib_1t - \frac{1}{2}\sigma_{11}t^2 + \int_{R^1} \frac{e^{itx_1} - 1 - it\tau(x_1)}{x_1^2} dM_1(x_1) \\
 &= ib_1t - \frac{1}{2}(\sigma_{11} + M_1(0))t^2 + \int_{R^1} \frac{e^{itx_1} - 1 - it\tau(x_1)}{x_1^2} dM_2(x_1).
 \end{aligned}$$

Now, if $\epsilon_0'X$ is normal, it follows from the uniqueness of (1) that M_2 is identically zero; and thus from (5) and (4) it follows that for $x_1 \neq 0$, $dM(\mathbf{x}) = 0$. Or: M is carried by the set $\{\mathbf{x} \mid 0 = (1, 0, \dots, 0)\mathbf{x}\}$. \square

COROLLARY 1. *Let X be an infinitely divisible random k -dimensional vector. Then there exists a (trivial or non-trivial) linear subspace L of R^k , such that every projection of X onto a linear subspace of L has a normal distribution, and every projection of X onto a linear subspace in $R^k - L$ is not normally distributed.*

PROOF. This follows when one takes the orthogonal complement of L to be the minimal subspace which carries the measure M .

Note. It follows from Corollary 1 that an infinitely divisible random vector is normally distributed if and only if its coordinates are normally distributed.

The array $\{X_{i,n}\}_{i=1, \dots, n; n=1, 2, \dots}$ is said to be a null triangular array if for each n , $X_{1,n} \dots X_{n,n}$ are independent, and for every $\epsilon > 0$, $\max_{1 \leq i \leq n} \text{Prob}(|X_{i,n}| > \epsilon) \rightarrow_{n \rightarrow \infty} 0$.

Exactly as in the one-dimensional case, a limit distribution of the sum

$S_n = \sum_{i=1}^n X_{i,n}$ is necessarily infinitely divisible. Combining this with the previous remark, we obtain:

COROLLARY 2. *Let $\{X_{i,n}\}_{i=1,\dots,n;n=1,2,\dots} = \{(X_{i,n}^{(1)}, \dots, X_{i,n}^{(k)})\}_{i=1,\dots,n;n=1,2,\dots}$ be a null triangular array, such that the distribution of $S_n = \sum_{i=1}^n X_{i,n}$ tend to the distribution of a random vector U . A necessary and sufficient condition for U to be (k -dimensional) normal is that the distributions of the sums $\sum_{i=1}^n X_{i,n}^{(j)}$, ($j = 1, \dots, k$), tend to a normal (one-dimensional) distribution.*

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REFERENCE

- [1] FELLER, WILLIAM (1966). *An Introduction to Probability Theory and its Applications 2*. Wiley, New York.