

A STATISTICAL TREATMENT OF THE INTENSITY FLUCTUATION OF RANDOM NOISE CURRENT*

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ABSTRACT

As a method of treating statistically many correlative physical quantities fluctuating only in positive region, the joint characteristic function in the form of Hankel transform was introduced. Then it was demonstrated that this function is more effective for the analyses of the noise current than the functions of Fourier or Laplace type which have been used so far. When the system of correlative random currents are led into an energy detector and the energy output fluctuation is considered, it can be reduced to a probability problem of distance in N -dimensional signal space, and further can correspond to a problem of random walks in N -dimensional space. From this point of view, the explicit expressions for the probability distribution of intensity fluctuation of noise current were derived by means of the Hankel type characteristic function, in connection with the random walk problem. These expressions are more general than the well-known expressions due to Rice, Rayleigh and Watson, because they include the latter ones as special cases.

1. Introduction

The probability variables defined over positive region in a functional space are fundamental quantities in physics, and the probability density distribution of distance in the multi-dimensional space (1, 2) gives statistically the intensity fluctuation of random noise current, if we take the distance as the mean energy. We are well aware of the fact that the energy fluctuation comes to the front in almost all of noise measurements (3).

In this paper, the energy fluctuation of random noise in the multi-dimensional space is particularly considered, by treating the generalized problem of random walks in multi-dimensional vector space (1, 4, 5, 6). First, we shall introduce a characteristic function in the form of the Hankel transform (1, 7) applicable as a postern to probability problems of K correlative physical quantities fluctuating only in positive region, and then, prove that the above characteristic function of Hankel type is more useful for calculating the joint-probability distribution of energy fluctuation of random noise than the usual characteristic function of

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Fourier (or Laplace) type. As applications, we shall calculate explicitly the probability density distribution of intensity fluctuation of random noise current: $P(R)$ or $P(E)$, and derive the relation between the resultant vector and respective element vectors affecting parameters of the resultant distribution.

Between the statistical theory of random noise and the information theory of message which are two main currents in general communication theory, this paper falls under the former, as S. O. Rice's way of treating (8).

2. The N -dimensional representation of the random intensity fluctuation

Now, we consider along time axis a set of the general stochastic processes:

$$G_h(t) \quad (h = 1, 2, \dots, K),$$

which are respectively composed of the regular component $S_h(t)$ and the random component $N_h(t)$ ($\langle N_h(t) \rangle = 0$), and especially the integral energies:

$$E_h = \int_{t_1}^{t_2} G_h^2(t) \omega(t) dt,$$

such as the output from energy detector, where $G_h^2(t)$ corresponds to the instantaneous energy fluctuation and $\omega(t)$ is a weight function, since our observation coincides essentially with the weighted mean of energy, in that an information which we can obtain for the crude random phenomenon through our observation is a certain mean image. The form of the mean operation (especially the weight function) will be successively improved with the steady progress in science. However, at the stage where the effect of observation on the crude random fluctuation is unknown, a special form:

$$\frac{1}{T} \int_{t_1}^{t_2} G_h^2(t) dt$$

taking $\omega(t) = \frac{1}{T}$, $T = t_2 - t_1$, of the mean operation (1, 2, 5, 9) seems to be a most natural one to take.

Generally, in investigating properties of $G_h(t)$ ($= S_h(t) + N_h(t)$) in a finite time interval, say (t_1, t_2) , it is convenient to expand it in terms of an orthonormal set of functions, $\varphi_i(t)$ ($i = 1, 2, \dots$) with a weight function $\omega(t)$. That is, we can write:

$$\left. \begin{aligned} G_h(t) &= S_h(t) + N_h(t) = \sum_{i=0}^{\infty} g_{hi} \varphi_i(t), \\ g_{hi} &= s_{hi} + n_{hi} = \int_{t_1}^{t_2} \omega(t) G_h(t) \varphi_i(t) dt, \\ s_{hi} &= \int_{t_1}^{t_2} \omega(t) S_h(t) \varphi_i(t) dt, \\ n_{hi} &= \int_{t_1}^{t_2} \omega(t) N_h(t) \varphi_i(t) dt, \end{aligned} \right\} \quad (1)$$

where

$$\int_{t_1}^{t_2} \omega(t) \varphi_i(t) \varphi_j(t) dt = \delta_{ij} = \begin{cases} 1 & \text{when } i = j \\ 0 & \text{when } i \neq j. \end{cases}$$

On the other hand, by use of Khintchine-Wiener's theorem (10), auto- and cross-correlation functions of $G_h(t)$:

$$\rho_{hl}(t, t') = \langle (G_h(t) - \langle G_h(t) \rangle) (G_l(t') - \langle G_l(t') \rangle) \rangle,$$

assuming

$$\rho_{hl}(t, t') = \rho_{hl}(\tau), \quad \tau = |t - t'|,$$

are derived from its auto- and cross-spectral densities *a priori* assigned.

In the above expression (1) of $G_h(t)$, it is very convenient to choose, as the orthonormal set $\varphi_i(t)$ ($i=1, 2, \dots$) with $\omega(t)$, the solutions (eigenfunctions) of the homogeneous Fredholm integral equation:

$$A_{hl} \int_{t_1}^{t_2} \sqrt{\omega(t')} \varphi_j(t') dt' = \int_{t_1}^{t_2} K_{hl}(t, t') \sqrt{\omega(t)} \varphi_j(t) dt \quad (t_1 \leq t \leq t_2) \quad (2)$$

with the symmetric kernel:

$$K_{hl}(t, t') = \rho_{hl}(t, t') \sqrt{\omega(t) \omega(t')}.$$

Then, from the Parseval equation, our weighted average power E_h can be expressed by:

$$E_h = \int_{t_1}^{t_2} \omega(t) G_h^2(t) dt = \sum_{i=1}^N x_{hi}^2, \quad (x_{hi} = g_{hi}). \quad (3)$$

By finding eigenvalues A_{hi} 's and eigenfunctions $\varphi_i(t)$'s of the above integral equation and letting N be the effective number of eigenvalues (therefore, $i=1, 2, \dots, N$), it can easily be shown that the mean value and the covariance for the expansion coefficients:

$$g_{hi}, \quad (h = 1, 2, \dots, K; i = 1, 2, \dots, N)$$

become:

$$\left. \begin{aligned} \langle g_{hi} \rangle &= \langle s_{hi} \rangle = \int_{t_1}^{t_2} \omega(t) \langle S_h(t) \rangle \varphi_i(t) dt (\equiv a_{hi}), \\ \langle (g_{hi} - \langle g_{hi} \rangle) (g_{lj} - \langle g_{lj} \rangle) \rangle &= A_{hl} \cdot \delta_{ij} (\equiv \sigma_{hl}), \\ \sigma_{g_{hi}}^2 &= \langle (g_{hi} - \langle g_{hi} \rangle)^2 \rangle = A_{hh} = \sigma_{s_{hi}}^2 + \sigma_{n_{hi}}^2, \end{aligned} \right\}$$

where s_{hi} 's are independent of n_{hi} 's and we must notice that our eigenvalues and eigenfunctions are affected by $\omega(t)$.

Thus, a joint-distribution density of x_{hi} 's can be expressed in the form of product of K -dimensional normal distributions (4, 6, 11), that is,

$$\begin{aligned}
 & P(x_{11}, x_{21}, \dots, x_{K1}; x_{12}, x_{22}, \dots, x_{K2}; \dots, x_{1N}, x_{2N}, \dots, x_{KN}) \\
 &= \prod_{i=1}^N \left(\frac{1}{\sqrt{2\pi}} \right)^K \frac{1}{\sqrt{|\sigma_{(i)}^{hl}|}} \exp \left\{ -\frac{1}{2} \sum_h \sum_l \sigma_{(i)}^{hl} (x_{hi} - a_{hi})(x_{li} - a_{li}) \right\}, \quad (4)
 \end{aligned}$$

where $[\sigma_{(i)}^{hl}]$ is an inverse-symmetric matrix of the variance matrix $[\sigma_{hl}]_{(i)}$.

After all, the problem is: What is the joint-probability density of energies $P(E_1, E_2, \dots, E_K)$ for K correlative random output fluctuations, from Eqs. (3) and (4)?

On the other hand, we shall take up the generalized problem of K correlative series of random walks $\mathbf{R}_h = \sum_{j=1}^{S_h} \mathbf{r}_{hj}$ ($h=1, 2, \dots, K$) as shown in Fig. 1, where $\mathbf{r}_{hj} = \sum_{i=1}^{N_h} e_i x_{hji}$, e_i being a unit vector along x_i -axis in the N -dimensional vector space. Now, in a special case where $N_1 = N_2 = \dots = N_K = N$ ($=N'$),

$$R_h^2 = \sum_{i=1}^N x_{hi}^2; \quad x_{hi} = \sum_{j=1}^{S_h} x_{hji}. \quad (5)$$

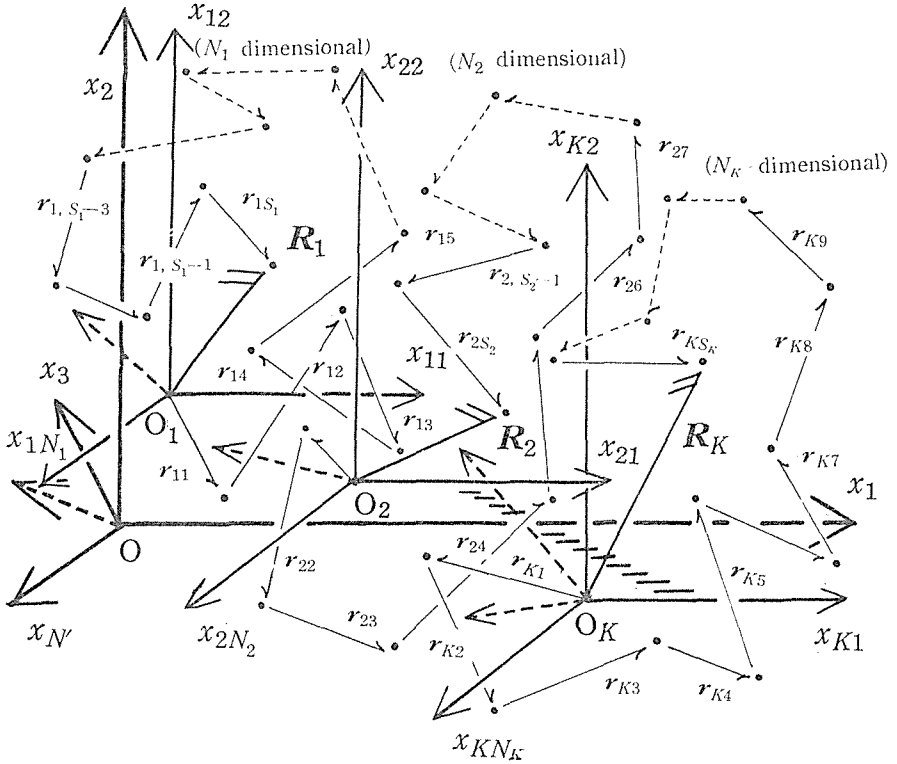


Fig. 1. K series of random walks in an N -dimensional space, ($\max [N_1, N_2, \dots, N_K] \leq N' \leq N_1 + N_2 + \dots + N_K$).

Thus, the problem can be transformed into the same form as Eq. (3) by

letting $E_k = R_k^2$. Then, we are able to regard the above random process as a probability problem in multi-dimensional vector space (1, 7).

We may mention several interesting cases.

(i) As shown in Fig. 2, the problem of the resultant of several random waves such as the signal and the noise, corresponds to the random walk problem in the Gaussian plane with $N=2$, where we can consider the noise character as the amplitude fluctuation or the phase fluctuation (1, 12).

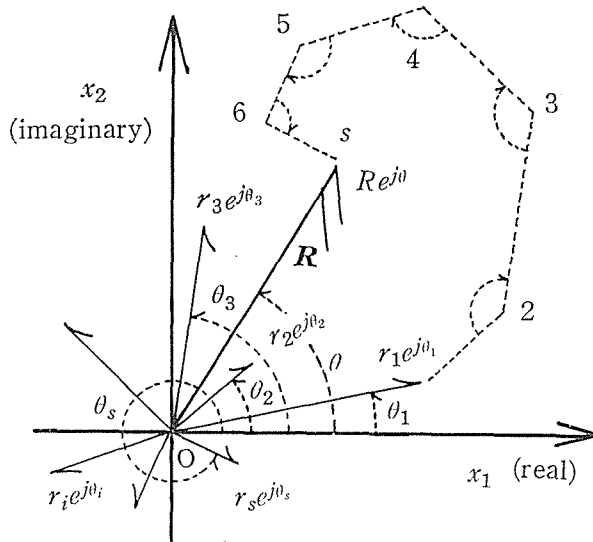


Fig. 2. The relation between random waves and random walks.

(ii) A probability calculation of average power of the random signals limited within the frequency interval W and time interval T can be regarded, following Shannon's sampling theorem (13), as a problem of the random process in the form of Eq. (3) in the functional space with $N=2TW$ (1, 7, 11).

3. The application of Hankel transform to random processes

In the preceding paragraph, it has been confirmed that we can regard the actual 3-dimensional random processes along time axis as a certain problem of random walks in an N -dimensional vector space, when the energy fluctuation is concerned. We are also well aware of the effectiveness of using the moment generating function of Laplace type: $m(t) = \langle e^{-tE} \rangle = \int_{-\infty}^{\infty} e^{-tE} P(E) dE$ or characteristic function of Fourier type as a postern for the probability calculation in a scalar field.

The problem is: What is the postern for the probability calculation in an N -dimensional vector space?

[A] Derivation of the joint characteristic function of Hankel type

First, we shall introduce a joint characteristic function $F(\lambda_1, \lambda_2, \dots, \lambda_K)$ in the form of the Hankel transform applicable to respective probability problems of K correlative physical quantities R_h 's ($h=1, 2, \dots, K$) fluctuating only in positive region, and express the joint-probability density $P(R_1, R_2, \dots, R_K)$ (1, 2, 7).

In general, the joint-probability distribution function :

$$\mathcal{P}(R_1, R_2, \dots, R_K) = \int_0^{R_1} \int_0^{R_2} \dots \int_0^{R_K} P(R_1, R_2, \dots, R_K) dR_1 dR_2 \dots dR_K$$

can be expressed by

$$\left. \begin{aligned} \mathcal{P}(R_{10}, R_{20}, \dots, R_{K0}) &= \int_0^\infty \dots \int_0^\infty P(R_1, R_2, \dots, R_K) \prod_{h=1}^K D_h(R_h) dR_h, \\ D_h(R_h) &= \begin{cases} 1 & (R_h < R_{h0}), \\ 0 & (R_h > R_{h0}). \end{cases} \end{aligned} \right\} \quad (6)$$

By using, instead of $D_h(R_h)$, the discontinuous integral due to Weber-Schafheitlin (14) :

$$R_{h0}^{m_h} \int_0^\infty J_{m_h}(R_{h0}\lambda_h) \frac{J_{m_h-1}(R_h\lambda_h)}{R_h^{m_h-1}} d\lambda_h = \begin{cases} 1 & (R_h < R_{h0}), \\ 0 & (R_h > R_{h0}), \end{cases} \quad (7)$$

we obtain :

$$\left. \begin{aligned} \mathcal{P}(R_1, R_2, \dots, R_K) &= \left(\frac{1}{2}\right)^{\sum_{h=1}^K m_h - K} \\ &\times \prod_{h=1}^K \frac{R_{h0}^{m_h}}{\Gamma(m_h)} \int_0^\infty \int_0^\infty \dots \int_0^\infty \left\{ \prod_{h=1}^K \lambda_h^{m_h-1} J_{m_h}(\lambda_h R_h) \right\} F(\lambda_1, \lambda_2, \dots, \lambda_K) d\lambda_1 d\lambda_2 \dots d\lambda_K, \\ P(R_1, R_2, \dots, R_K) &= \left(\frac{1}{2}\right)^{\sum_{h=1}^K m_h - K} \\ &\times \prod_{h=1}^K \frac{R_h^{m_h}}{\Gamma(m_h)} \int_0^\infty \int_0^\infty \dots \int_0^\infty \left\{ \prod_{h=1}^K \lambda_h^{m_h} J_{m_h-1}(\lambda_h R_h) \right\} F(\lambda_1, \lambda_2, \dots, \lambda_K) d\lambda_1 d\lambda_2 \dots d\lambda_K, \\ F(\lambda_1, \lambda_2, \dots, \lambda_K) &= \left\langle \prod_{h=1}^K 2^{m_h-1} \Gamma(m_h) \frac{J_{m_h-1}(\lambda_h R_h)}{(\lambda_h R_h)^{m_h-1}} \right\rangle, \end{aligned} \right\} \quad (8)$$

with $m_h \geq 1/2$ ($h=1, 2, \dots, K$), or, more explicitly,

$$\left. \begin{aligned} \mathcal{P}(R_1, R_2, \dots, R_K) &= \left(\prod_{h=1}^K R_{h0}^{m_h} \right) \int_0^\infty \int_0^\infty \dots \int_0^\infty \left\{ \prod_{h=1}^K J_{m_h}(\lambda_h R_h) \right\} \\ &\quad \times \left\langle \prod_{h=1}^K R_h^{1-m_h} J_{m_h-1}(\lambda_h R_h) \right\rangle d\lambda_1 d\lambda_2 \dots d\lambda_K, \\ P(R_1, R_2, \dots, R_K) &= \left(\prod_{h=1}^K R_h^{m_h} \right) \int_0^\infty \int_0^\infty \dots \int_0^\infty \left\{ \prod_{h=1}^K \lambda_h J_{m_h-1}(\lambda_h R_h) \right\} \\ &\quad \times \left\langle \prod_{h=1}^K R_h^{1-m_h} J_{m_h-1}(\lambda_h R_h) \right\rangle d\lambda_1 d\lambda_2 \dots d\lambda_K. \end{aligned} \right\} \quad (9)$$

But $F(\lambda_1, \lambda_2, \dots, \lambda_K) = \prod_{h=1}^K F(\lambda_h)$ when $P(R_1, R_2, \dots, R_K) = \prod_{h=1}^K P(R_h)$, i.e., R_1, R_2, \dots, R_K are statistically independent.

Especially, when we consider the probability problem given in the form $R_h^2 = \sum_{i=1}^{N_h} x_{hi}^2$ of Eq. (5), we can also derive the characteristic function $F(\lambda_1, \lambda_2, \dots, \lambda_K)$ in Eq. (8) under the additional condition: $m_h = N_h/2$, as follows:

$$\left. \begin{aligned}
 F(\lambda_1, \lambda_2, \dots, \lambda_K) &= \frac{1}{\prod_{h=1}^K S_{(N_h)}} \int_{S_{(N_1)}} \dots \int_{S_{(N_2)}} \dots \int_{S_{(N_K)}} \dots \int_{h=1}^K dS_{(N_h)} \\
 &\quad \times [F(\mu_{11}, \mu_{12}, \dots, \mu_{1N_1}; \dots; \mu_{K1}, \mu_{K2}, \dots, \mu_{KN_K})] \Big|, \quad (10) \\
 &\quad (\mu_{h1}, \mu_{h2}, \dots, \mu_{hN_h}) \rightarrow (\lambda_h, \varphi_{h1}, \dots, \varphi_{h, N_h-1}) (\forall h) \\
 F(\mu_{11}, \mu_{12}, \dots, \mu_{1N_1}; \dots; \mu_{K1}, \mu_{K2}, \dots, \mu_{KN_K}) &= \langle \exp(i[\sum_{h=1}^K \sum_{j=1}^{N_h} \mu_{hj} x_{hj}]) \rangle,
 \end{aligned} \right\}$$

where

$$S_{(N_h)} = \frac{(\sqrt{\pi})^{N_h} N_h}{\Gamma\left(\frac{N_h}{2} + 1\right)} \quad \text{and} \quad dS_{(N_h)} = \prod_{j=1}^{N_h-1} (\sin \varphi_{hj})^{N_h-1-j} d\varphi_{hj} \quad (h = 1, 2, \dots, K)$$

mean respectively a surface area and a surface element of an N_h -dimensional unit hypersphere, and $(\mu_{h1}, \mu_{h2}, \dots, \mu_{hN_h}) \rightarrow (\lambda_h, \varphi_{h1}, \dots, \varphi_{h, N_h-1})$ denotes the transformation to the polar coordinates. After a somewhat troublesome calculation, we can find that Eq. (10) agrees with $F(\lambda_1, \lambda_2, \dots, \lambda_K)$ in Eq. (8), by use of an inversion formula of the Hankel transformation, and also $P(R_1, R_2, \dots, R_K)$ in Eq. (8) can be derived from Eq. (10) by application of the transformation to an N_h -dimensional polar coordinates, $(x_{h1}, x_{h2}, \dots, x_{hN_h}) \rightarrow (R_h, \theta_{h1}, \theta_{h2}, \dots, \theta_{h, N_h-1})$.

If a special case $K=1$ is taken up, we have

$$\left. \begin{aligned}
 P(R) &= \frac{R^m}{2^{m-1} \Gamma(m)} \int_0^\infty F(\lambda) \lambda^m J_{m-1}(\lambda R) d\lambda, \\
 F(\lambda) &= \left\langle 2^{m-1} \Gamma(m) \frac{J_{m-1}(\lambda R)}{(\lambda R)^{m-1}} \right\rangle \\
 &= 1 + \sum_n \frac{(-1)^n \Gamma(m) \langle R^{2n} \rangle}{2^{2n} n! \Gamma(m+n)} \lambda^{2n},
 \end{aligned} \right\} \quad (11)$$

or,

$$\left. \begin{aligned}
 F(\lambda) &= \frac{1}{S_{(N)}} \int_0^{2\pi} \int_0^\pi \dots \int_0^\pi [F(\mu_1, \mu_2, \dots, \mu_N)] |dS_{(N)}|, \\
 &\quad (\mu_1, \mu_2, \dots, \mu_N) \rightarrow (\lambda, \varphi_1, \dots, \varphi_{N-1}) \\
 F(\mu_1, \mu_2, \dots, \mu_N) &= \left\langle \exp(i \sum_{K=1}^N x_K \mu_K) \right\rangle,
 \end{aligned} \right\} \quad (12)$$

corresponding to Eq. (8) or Eq. (10).

[B] **The effectiveness of the characteristic function of Hankel type**

The method using the moment generating function $m(t)$ of Laplace type is indeed useful for calculating the probability density $P(E)$ for the sum: $E = \sum_{i=1}^S e_i$ of independent s scalar functions $e_i(t)$, since we have simply

$$m(t) = \prod_{i=1}^S m_i(t), \quad m_i(t) = \langle e^{-te_i} \rangle.$$

The method using the characteristic function of Hankel type: $F(\lambda) = \left\langle 2^{m-1} \Gamma(m) \times \frac{J_{m-1}(\lambda R)}{(\lambda R)^{m-1}} \right\rangle$ (see Eq. (11)) is, however, generally effective in calculating the probability density $P(R)$ for the sum $\mathbf{R} = \sum_{i=1}^S \mathbf{r}_i$ ($R = |\mathbf{R}|$) of independent S vector functions $\mathbf{r}_i(t)$ in an N -dimensional vector space, when the magnitude $r_i (= |\mathbf{r}_i|)$ is independent of the direction of \mathbf{r}_i .

Consequently, we shall prove Eq. (11) under the special condition that $F(\lambda) = \prod_{i=1}^S F_i(\lambda)$ and $m = N/2$, where $F_i(\lambda) = \left\langle 2^{m-1} \Gamma(m) \frac{J_{m-1}(\lambda r_i)}{(\lambda r_i)^{m-1}} \right\rangle$, more explicitly,

$$\left. \begin{aligned} \mathcal{P}(R) &= \Gamma\left(\frac{N}{2}\right)^{S-1} \left(\frac{1}{2}\right)^{(N/2)-1} R \int_0^\infty (\lambda R)^{(N/2)-1} J_{N/2}(\lambda R) \\ &\quad \times \prod_{j=1}^S \left\langle \frac{J_{(N/2)-1}(\lambda r_j)}{\left(\frac{1}{2} \lambda r_j\right)^{(N/2)-1}} \right\rangle d\lambda, \\ P(R) &= \Gamma\left(\frac{N}{2}\right)^{S-1} \left(\frac{1}{2}\right)^{(N/2)-1} \int_0^\infty (\lambda R)^{N/2} J_{(N/2)-1}(\lambda R) \\ &\quad \times \prod_{j=1}^S \left\langle \frac{J_{(N/2)-1}(\lambda r_j)}{\left(\frac{1}{2} \lambda r_j\right)^{(N/2)-1}} \right\rangle d\lambda. \end{aligned} \right\} \quad (13)$$

The proof consists in answering our question given at the beginning of this paragraph. It is sufficient to verify that $F(\lambda) = F_1(\lambda)F_2(\lambda)$ holds when $\mathbf{R} = \mathbf{r}_1 + \mathbf{r}_2$ (i.e., $S=2$) in an N -dimensional vector space. But the way of verification differs between the next two cases: $N \neq 2$ and $N=2$.

(i) Case when $N \neq 2$ ($m \neq 1$)

We shall take up the problem of random walks with two steps in an N -dimensional vector space as shown in Fig. 3.

Since $R^2 = r_1^2 + r_2^2 - 2r_1r_2 \cos \theta$, we shall apply to $F(\lambda)$ an addition-formula for Bessel function of order $\nu (= m-1)$ (14):

$$\left. \begin{aligned} \rho^{-\nu} J_\nu(\rho) &= 2^\nu \Gamma(\nu) \sum_{k=0}^{\infty} (\nu+k) \frac{J_{\nu+k}(Z)}{Z^\nu} \frac{J_{\nu+k}(X)}{X^\nu} C_k^\nu(\cos \theta), \\ \rho &= \sqrt{Z^2 + X^2 - 2ZX \cos \theta} \quad (\nu \neq 0, -1, -2, \dots). \end{aligned} \right\} \quad (14)$$

Then, our characteristic function $F(\lambda)$ in Eq. (11) can be expressed by

$$F(\lambda) = 4^{m-1} \Gamma(m) \Gamma(m-1) \times \sum_{k=0}^{\infty} (m-1+k) \left\langle \frac{J_{m-1+k}(\lambda r_1)}{(\lambda r_1)^{m-1}} \right\rangle \left\langle \frac{J_{m-1+k}(\lambda r_2)}{(\lambda r_2)^{m-1}} \right\rangle \left\langle C_k^{m-1}(\cos \theta) \right\rangle. \quad (15)$$

Now, we take, along the first vector r_1 , the negative direction of x_1 -axis of an N -dimensional rectangular coordinate-system and express the second vector r_2 by N -dimensional polar coordinates, say $r_2 = [r_2, \theta_1, \theta_2, \dots, \theta_{N-1}]$ as in Fig. 3, taking $\theta_1 = \theta$. Under the assumption that the magnitude of each vector is independent of its direction, we have $P(r_2, \theta_1, \theta_2, \dots, \theta_{N-1}) = P(r_2)P(\theta_1, \theta_2, \dots, \theta_{N-1})$. Further, from the randomness, i.e., the isotropic property of each vector, the probability that r_2 should face to any surface element $dS_{(N)}$ of the N -dimensional hypersphere (radius r_2 , surface area $S_{(N)}$) is constant.

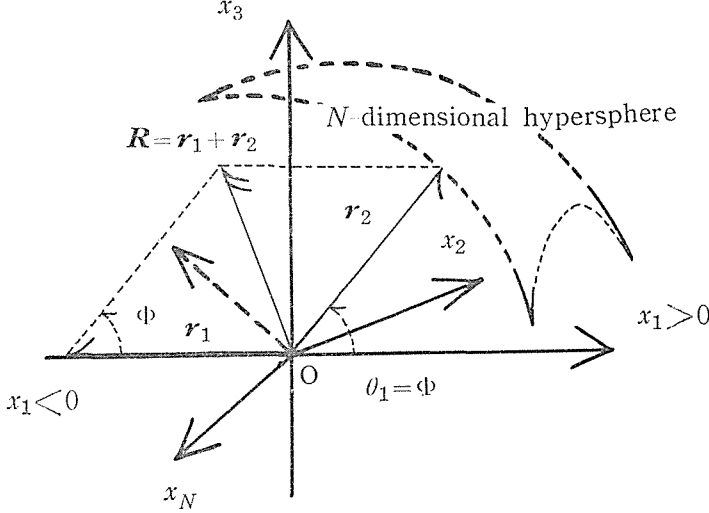


Fig. 3. Random walks with two steps in an N -dimensional space.

Thus, we have the following relation :

$$P(\theta_1, \theta_2, \dots, \theta_{N-1}) d\theta_1 d\theta_2 \dots d\theta_{N-1} = \frac{1}{S_{(N)}} dS_{(N)} \\ = \frac{\Gamma(m+1)}{2^m (\sqrt{\pi})^{2m}} \prod_{i=1}^{N-1} (\sin \theta_i)^{N-1-i} d\theta_1 d\theta_2 \dots d\theta_{N-1}, \quad (2m = N), \quad (16)$$

where

$$S_{(N)} = r_2^{N-1} \frac{\Gamma\left(\frac{1}{2}\right)^N 2^m}{\Gamma(m+1)} \quad \text{and} \quad dS_{(N)} = r_2^{N-1} \prod_{i=1}^{N-1} (\sin \theta_i)^{N-1-i} d\theta_i.$$

Accordingly, from Eq. (16), the mean value of the Gegenbauer function (14) : $\langle C_k^{m-1}(\cos \theta) \rangle$ in Eq. (15) becomes :

$$\begin{aligned}
\langle C_k^{m-1}(\cos \vartheta) \rangle &= \int_0^\pi \cdots \int_0^\pi \int_0^{2\pi} C_k^{m-1}(\cos \theta_1) P(\theta_1, \theta_2, \dots, \theta_{N-1}) d\theta_1 d\theta_2 \cdots d\theta_{N-1} \\
&= \int_0^\pi C_k^{m-1}(\cos \theta_1) (\sin \theta_1)^{N-2} d\theta_1 \int_0^\pi (\sin \theta_2)^{N-3} d\theta_2 \int_0^\pi \cdots \int_0^\pi \cdots \int_0^{2\pi} d\theta_{N-1} \frac{\Gamma(m+1)}{2m(\sqrt{\pi})^{2m}}. \quad (17)
\end{aligned}$$

Hereupon, if we use an orthogonality relation for the Gegenbauer function (14) :

$$\begin{aligned}
&\int_0^\pi C_k^{m-1}(\cos \theta_1) C_0^{m-1}(\cos \theta_1) \sin^{2(m-1)} \theta_1 d\theta_1 \\
&= \delta_{k0} \frac{\pi \Gamma(2m-2)}{2^{2m-3} (m-1) \Gamma(m-1)^2}, \quad (18)
\end{aligned}$$

under the particular condition $m=N/2$, we can derive :

$$\begin{aligned}
F(\lambda) &= \left\langle 2^{m-1} \Gamma(m) \frac{J_{m-1}(\lambda r_1)}{(\lambda r_1)^{m-1}} \right\rangle \left\langle 2^{m-1} \Gamma(m) \frac{J_{m-1}(\lambda r_2)}{(\lambda r_2)^{m-1}} \right\rangle \\
&= F_1(\lambda) F_2(\lambda). \quad (19)
\end{aligned}$$

(ii) Case when $N=2$ ($m=1$)

In this case, since $m=1$, we cannot use the same addition-formula for Bessel functions as in the preceding case. Therefore, we had better use instead the following addition-formula for Bessel functions (14) :

$$J_\nu(\rho) \cos \nu \Psi = \sum_{k=-\infty}^{\infty} J_{\nu+k}(Z) J_k(X) \cos k\Phi, \quad (20)$$

as illustrated in Fig. 4.

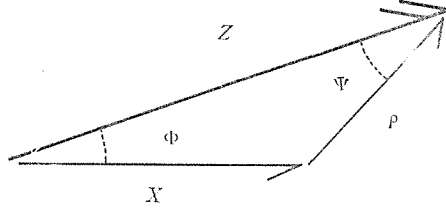


Fig. 4. Addition-formula for Bessel functions.

Thus, $F(\lambda)$ in the case $m=1$ becomes :

$$F(\lambda) = \langle J_0(\lambda R) \rangle = \langle J_0(\lambda r_1) \rangle \langle J_0(\lambda r_2) \rangle + 2 \sum_{k=1}^{\infty} \langle J_k(\lambda r_1) J_k(\lambda r_2) \rangle \langle \cos k\Phi \rangle. \quad (21)$$

Since $P(\vartheta) = \frac{1}{2\pi}$ corresponding to the same isotropic property as in the case when $N \neq 2$, we find

$$\langle \cos k\Phi \rangle = 0 \quad (k \neq 0).$$

Finally, we have

$$F(\lambda) = \langle J_0(\lambda r_1) \rangle \langle J_0(\lambda r_2) \rangle = F_1(\lambda) F_2(\lambda). \quad (22)$$

Thus, the proof has been completed.

[C] **The transformation between Hankel and Laplace type characteristic functions.**

As stated in [B], it goes without saying that our joint-characteristic function of Hankel type: $F(\lambda_1, \lambda_2, \dots, \lambda_K)$ in Eq. (8) is also effective in calculating the joint-probability: $P(R_1, R_2, \dots, R_K)$ in the vector field, in contrast to the usual joint moment generating function: $m(t_1, t_2, \dots, t_K) = \langle \exp(-\sum_{h=1}^K t_h E_h) \rangle$ being effective in calculating the joint-probability: $P(E_1, E_2, \dots, E_K)$ in the scalar field.

We must, however, point out that $F(\lambda_1, \lambda_2, \dots, \lambda_K)$ and $m(t_1, t_2, \dots, t_K)$ are closely connected with each other (11). When setting $E_h = R_h^2$, by use of the integral formula (15):

$$\int_0^\infty e^{-tR^2} R^m J_{m-1}(\lambda R) dR = \frac{\lambda^{m-1}}{(2t)^m} e^{-(\lambda^2/4t)}, \quad (23)$$

we can find that the transformability between the above two characteristic functions reduces itself to a K -dimensional Laplace transform as follows:

$$M(S_1, S_2, \dots, S_K) = \int_0^\infty \dots \int_0^\infty e^{-S_1 A_1 + S_2 A_2 + \dots + S_K A_K} \times G(A_1, A_2, \dots, A_K) dA_1 dA_2 \dots dA_K, \quad (24)$$

where

$$\left. \begin{aligned} M(S_1, S_2, \dots, S_K) &= m(t_1, t_2, \dots, t_K) \prod_{h=1}^K 2^{2m_h} \Gamma(m_h) t_h^{m_h}, \\ &\quad \left(S_h = \frac{1}{4t_h} \right), \\ G(A_1, A_2, \dots, A_K) &= F(\lambda_1, \lambda_2, \dots, \lambda_K) \prod_{h=1}^K \lambda_h^{2(m_h-1)} \\ &\quad (A_h = \lambda). \end{aligned} \right\} \quad (25)$$

Similarly, our $F(\lambda_1, \lambda_2, \dots, \lambda_K)$ is essentially connected with a usual characteristic function $\varphi(t_1, t_2, \dots, t_K) = \langle \exp(i \sum_{h=1}^K t_h E_h) \rangle$ by a K -dimensional Fourier transform.

4. The integral representation of probability density of K correlative series of random walks.

We have indicated that the probability density $P(R)$ for one series (i.e., $R = \sum_{i=1}^S r_i$) of the random walks in an N -dimensional space can be given by Eq. (11) in a form of integral expression. Of course, when the magnitude and the direction of each vector are independent of all the preceding ones, we have $F(\lambda) = \prod_{i=1}^S F_i(\lambda)$ (cf. Eq. (13)). A case of special interest when $P(r_i) = \delta(r_i - r_{i0})$ arises when the magnitude of each vector is respectively constant, say $|r_i| = r_{i0}$.

On this occasion, a probability $\mathcal{P}(R) (= \int_0^R P(R) dR)$ in Eq. (13) agrees with Watson's results (15), and also, using an integral formula (15):

$$\int_0^{\infty} t^{1+2\nu-\gamma m} \prod_{n=1}^{S+1} J_{\nu}(a_n t) dt = 0 \quad (a_1 > \sum_{n=2}^{S+1} a_n, \operatorname{Re}(\nu) > -1), \quad (26)$$

we can find that $P(R)=0$ if $R > \sum_{i=1}^S r_{i0}$.

Generalizing the above, we shall now consider K correlative series (i.e. $h=1, 2, \dots, K$) of random walks of this kind, and assume that the h th series of random walks $\mathbf{R}_h = \sum_{j=1}^{S_h} \mathbf{r}_{hj}$ exists in an N_h -dimensional vector space, as shown in Fig. 1. Then, corresponding to Eq. (11), we can here use the same integral expression as in Eq. (8) or (9), under the special additional condition: $m_h = N_h/2$, by altering the interpretation (1, 4, 6).

5. The explicit representation of intensity distribution of random noise current

[i] First, one series of random walks is considered, consisting of the regular components $\mathbf{I}_j (j=1, 2, \dots, q)$ and the random components $\mathbf{r}_j (j=q+1, q+2, \dots, S)$ in an N -dimensional vector space as shown in Fig. 5. That is,

$$\mathbf{R} = \sum_{j=1}^q \mathbf{I}_j + \sum_{j=q+1}^S \mathbf{r}_j.$$

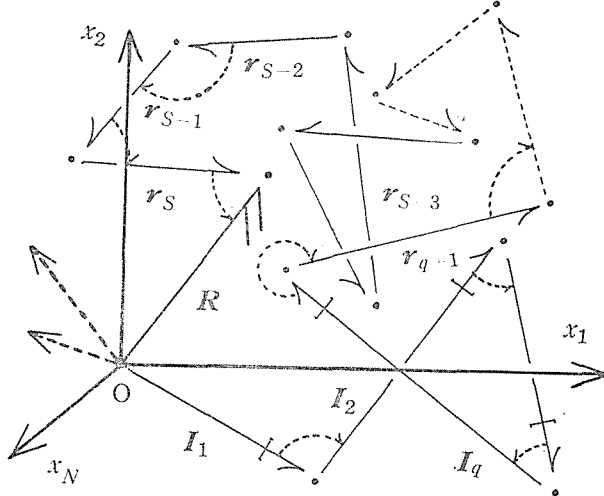


Fig. 5. Random walks accompanied by regular vectors in an N -dimensional space.

We shall now obtain the general expression of probability density $P(R)$ for very large values of $(S-q)$ under no special assumption concerning the distribu-

tion of the different walks except that all the $P(r_j)$'s ($r_j = |r_j|$) are nearly of the same functional form.

Since $\prod_{j=q+1}^S F_j(\lambda) \simeq e^{-\langle \Omega_0/4m \rangle \lambda^2}$ ($\Omega_0 = \sum_{j=q+1}^S \langle r_j^2 \rangle$) in Eq. (11) by use of Laplace's process or the method of steepest descent (14, 15) we can rewrite Eq. (13) in a form as follows:

$$P(R) = (2^{m-1} \Gamma(m))^{q-1} R^m \int_0^\infty \lambda^m J_{m-1}(\lambda R) e^{-\langle \Omega_0/4m \rangle \lambda^2} \times \prod_{j=1}^q \frac{1}{(\lambda I_j)^{m-1}} J_{m-1}(\lambda I_j) d\lambda, \quad (I_j = |I_j|). \quad (27)$$

In the special case of interest when $N=2$ ($m=1$) corresponding to the random wave problem in the example (i) of §1 [B], i.e.,

$$I = \sum_{j=1}^q I_j \cos \omega_j t \text{ (signal) } + I_N \text{ (noise)},$$

the probability density for an envelope amplitude of I agrees with S. O. Rice's expression (8) as follows:

$$P(R) = R \int_0^\infty \lambda J_0(\lambda R) e^{-\langle \Psi_0/4 \rangle \lambda^2} \prod_{j=1}^q J_0(\lambda I_j) d\lambda, \quad (\Psi_0 = \langle I_N^2 \rangle). \quad (28)$$

Further, when $\mathbf{R} = \mathbf{I}_1 + \sum_{j=2}^S \mathbf{r}_j$ (i.e., $q=1$) in the above, Eq. (27) can be expressed explicitly by

$$P(R) = \frac{NR^{N/2}}{\Omega_0} \left(\frac{1}{I_1} \right)^{(N/2)-1} e^{-\langle N(R^2 + I_1^2)/2\Omega_0 \rangle} I_{(N/2)-1} \left(\frac{NI_1}{\Omega_0} R \right), \quad (\text{Bessel-distribution}), \quad (29)$$

where $\Omega_0 = \sum_{j=2}^S \langle r_j^2 \rangle$ (1, 7, 13, 14, 16). And if q regular vectors I_j ($j=1, 2, \dots, q$) are given instead of one regular vector I_1 , it can be proved that Eq. (29) approximately expresses $P(R)$ in this case, under the substitutions $I_1 = \sqrt{\sum_{j=1}^q I_j^2}$ and $\Omega_0 = \sum_{j=q+1}^S \langle r_j^2 \rangle$ by applying to Eq. (13) the property in the vicinity of $\lambda=0$ for the following addition-formula for Bessel functions (14):

$$\frac{J_\nu(\sqrt{Z^2 + Y^2})}{(\sqrt{Z^2 + Y^2})^\nu} = 2^\nu \sum_{m=0}^\infty (-1)^m (\nu + 2m) \frac{\Gamma(\nu + m)}{m!} \times \frac{J_{\nu+2m}(Z)}{Z^\nu} \frac{J_{\nu+2m}(Y)}{Y^\nu}, \quad (Z = \lambda I_i, Y = \lambda I_j, i \neq j), \quad (30)$$

using the method of steepest descent.

Also, in the case of no regular vectors ($q=0$), we have:

$$P(R) = \frac{2 \left(\frac{N}{2} \right)^{N/2} R^{N-1}}{\Gamma \left(\frac{N}{2} \right) \Omega^{N/2}} e^{-\langle NR^2/2\Omega_0 \rangle}, \quad \left(P(E) = P(R) \left| \frac{dR}{dE} \right| : \Gamma\text{-distribution} \right). \quad (31)$$

Of course, in the above-mentioned expressions, we must take $\Omega_0 = (S-q)r_0^2$ when all the r_j 's have equal lengths, say $r_j = r_0$ ($j = q+1, q+2, \dots, S$). In order to improve the approximation, we had better use the following asymptotic expression (1, 2, 16) :

$$P(R) = \frac{2\left(\frac{N}{2}\right)^{N/2} R^{N-1}}{\Gamma\left(\frac{N}{2}\right)\Omega_0^{N/2}} e^{-(NR^2/2\Omega_0)} \left\{ 1 - \frac{2}{(N+2)S} L_2^{(N/2)-1}\left(\frac{N}{2Sr_0^2}R^2\right) + O\left(\frac{1}{S^2}\right) \right\} \quad (\Omega_0 = Sr_0^2), \quad (32)$$

instead of Eq. (31).

[ii] We shall again consider putting $p(r_j) = \delta(r_j - r_{j0})$ in Eq. (13) in the same way as we have done in §4, under a special condition $S=2$ corresponding to the special case $q=0, S=2$ in the above [i]. Then, we obtain the following explicit expression :

$$\left. \begin{aligned} P(R) &= \frac{1}{2^{N-3}} \left(\frac{1}{r_{10}r_{20}}\right)^{N-2} \frac{\Gamma(N/2)}{\sqrt{\pi}\Gamma\left(\frac{N-1}{2}\right)} \\ &\quad \times R [R^2 - (r_{10} - r_{20})^2]^{(N-3)/2} [(r_{10} + r_{20})^2 - R^2]^{(N-3)/2} \\ &\quad \text{for } |r_{10} - r_{20}| < R < |r_{10} + r_{20}|, \\ &= 0 \quad \text{for } 0 < R < |r_{10} - r_{20}| \text{ or } r_{10} + r_{20} < R < \infty. \end{aligned} \right\} \quad (33)$$

When we especially fix $N=2$, this expression evidently means an interference effect between two random phase waves and becomes explicitly

$$\left. \begin{aligned} P(R) &= \frac{2}{\pi} \frac{R}{\sqrt{[R^2 - (r_{10} - r_{20})^2][(r_{10} + r_{20})^2 - R^2]}}, \\ &\quad (|r_{10} - r_{20}| < R < r_{10} + r_{20}) \\ \mathcal{P}(R) &= \int_0^R P(R) dR = 0, \quad (0 < R < |r_{10} - r_{20}|) \\ &= \frac{1}{\pi} \cos^{-1} \frac{r_{10}^2 + r_{20}^2 - R^2}{2r_{10}r_{20}}, \quad (|r_{10} - r_{10}| < R < r_{10} + r_{20}) \\ &= 1, \quad (r_{10} + r_{20} < R < \infty). \end{aligned} \right\} \quad (34)$$

[iii] In the next place, we consider the case of very large S_h in two correlative series of such random walks: $\mathbf{R}_h = \sum_{j=1}^{S_h} \mathbf{r}_{hj}$ ($h=1, 2$) having no regular vectors (i.e., $q=0$) in an N -dimensional space as shown in Fig. 6. Then, we can derive the joint-probability density of Eq. (31) from Eq. (8) under the particular conditions :

$\Omega'_{h0} = \frac{1}{m} \Omega_{h0} = \frac{2}{N} \sum_{j=1}^{S_h} \langle r_{hj}^2 \rangle$ ($h=1, 2$) and $N_1 = N_2 = N = 2m$ as follows :

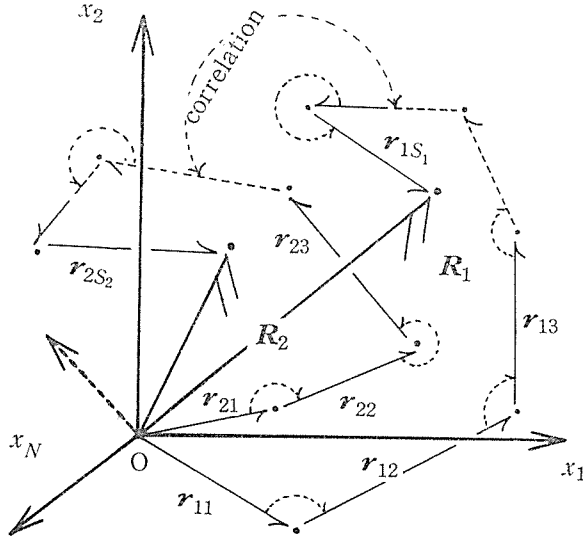


Fig. 6. Two correlative series of random walks in an N -dimensional space.

$$\begin{aligned}
 P(R_1, R_2) &= \frac{4(R_1 R_2)^{N/2}}{\Gamma\left(\frac{N}{2}\right) \Omega'_{10} \Omega'_{20} (1 - \rho_E)} \left\{ \frac{1}{\Omega'_{10} \Omega'_{20} \rho_E} \right\}^{(N/2)-1} \\
 &\quad \times e^{-\frac{1}{1 - \rho_E} \left\{ \frac{R_1^2}{\Omega'_{10}} + \frac{R_2^2}{\Omega'_{20}} \right\}} I_{(N/2)-1} \left(\frac{2\sqrt{\rho_E} R_1 R_2}{(1 - \rho_E) \sqrt{\Omega'_{10} \Omega'_{20}}} \right) \\
 &= P(R_1) P(R_2) \left\{ 1 + \sum_{n=1}^{\infty} L_n^{(N/2)-1} \left(\frac{R_1^2}{\Omega'_{10}} \right) L_n^{(N/2)-1} \left(\frac{R_2^2}{\Omega'_{20}} \right) \rho_E^n \right. \\
 &\quad \left. \times nB\left(n, \frac{N}{2}\right) \right\} \quad (R_1, R_2 > 0), \\
 &= 0 \quad (R_1, R_2 \leq 0), \\
 F(\lambda_1, \lambda_2) &= \frac{\Gamma\left(\frac{N}{2}\right) 2^{N-2} (\lambda_1 \lambda_2)^{1-N/2}}{(\sqrt{\rho_E} \Omega'_{10} \Omega'_{20})^{(N/2)-1}} e^{-\frac{1}{2} [\Omega'_{10} \lambda_1^2 + \Omega'_{20} \lambda_2^2]} \\
 &\quad \times I_{(N/2)-1} \left(\frac{1}{2} \sqrt{\rho_E \Omega'_{10} \Omega'_{20}} \lambda_1 \lambda_2 \right), \quad (36)
 \end{aligned}$$

where $P(R_h)$'s express respectively

$$P(R_h) = \frac{2\left(\frac{N}{2}\right)^{N/2}}{\Gamma\left(\frac{N}{2}\right) \Omega_{h0}^{N/2}} R_h^{N-1} e^{-(NR_h^2/2\Omega_{h0})} \quad (h = 1, 2),$$

following Eq. (31), and ρ_E denotes the correlation coefficient of R_1^2 and R_2^2 . In the above derivation, we have used the following integrals (14):

$$\left. \begin{aligned}
 & \int_0^\infty \lambda_1 e^{-(\Omega/4)\lambda_1^2} I_{m-1} \left(\frac{C}{2} \lambda_1 \lambda_2 \right) J_{m-1}(\lambda R_1) d\lambda_1 \\
 & \quad = \frac{2}{\Omega} \exp \left\{ \frac{1}{\Omega} \left[\left(\frac{C}{2} \lambda_2 \right)^2 - R_1^2 \right] \right\} J_{m-1} \left(\frac{C}{\Omega} \lambda_2 R_1 \right), \\
 & \int_0^\infty x e^{-Ax^2} J_\nu(Bx) J_\nu(Rx) dx \\
 & \quad = \frac{1}{2A} \exp \left\{ -\frac{1}{4A} (B^2 + R^2) \right\} I_\nu \left(\frac{1}{2A} BR \right).
 \end{aligned} \right\} \quad (37)$$

Now, to apply Eq. (35) to the detector output such as considered in § 2, Eqs. (3) and (5) are taken into account. Then, we can obtain the joint energy distribution $P(E_1, E_2)$ of detector output noise as follows:

$$P(E_1, E_2) = \frac{D_1 D_2}{\Gamma(m)(1-\rho_E)} \left(\sqrt{\frac{D_1 D_2 E_1 E_2}{\rho_E}} \right)^{m-1} \exp \left\{ -\frac{1}{1-\rho_E} (D_1 E_1 + D_2 E_2) \right\} \\
 \times I_{m-1} \left(\frac{2\sqrt{\rho_E D_1 D_2 E_1 E_2}}{1-\rho_E} \right), \quad (38)$$

where

$$E_h = R_h^2, \quad \sigma_{E_h}^2 = \langle (E_h - \langle E_h \rangle)^2 \rangle, \\
 D_h = m / \langle E_h \rangle = \langle E_h \rangle / \sigma_{E_h}^2,$$

and

$$\langle E_h \rangle = \Omega_{h0} \quad (h = 1, 2).$$

By use of the integral formula (14):

$$\int_0^\infty e^{-p^2 t^2} t^{\nu+1} J_\nu(at) dt = \frac{a^\nu}{(2p^2)^{\nu+1}} e^{-(a^2/4p^2)} \quad (\text{Re}(\nu) > -1), \quad (39)$$

the moment generating function $m(t_1, t_2)$ of Eq. (38) can be derived as follows:

$$m(t_1, t_2) = \left[\left(1 + \frac{t_1}{D_1} \right) \left(1 + \frac{t_2}{D_2} \right) - \rho_E \frac{t_1 t_2}{D_1 D_2} \right]^{-m}. \quad (40)$$

When Eq. (39) is used again, we can easily confirm that the above $m(t_1, t_2)$ is essentially connected with $F(\lambda_1, \lambda_2)$ in Eq. (36) by the 2-dimensional Laplace transform from Eqs. (24) and (25) ($K=2$).

[iv] We shall again take up the generalized problem of K correlative series of random walks, $\mathbf{R}_h = \sum_{j=1}^{S_h} \mathbf{r}_{hj}$ ($\mathbf{r}_{hj} = \sum_{i=1}^N e_i x_{hji} = [x_{hj1}, x_{hj2}, \dots, x_{hjN}]$) which is the h th series of such random walks in the same N -dimensional vector space. When we are specially interested in the asymptotic expression of the joint-probability for a large number of steps corresponding to Eqs. (31) and (35), we had better fix our eyes upon the form given by Eq. (3), that is, treat the probability problem given in the same form as Eq. (5).

Since S_h 's are very large, by use of the central limit theorem in NK dimensions, we can find that x_{hi} 's ($h=1, 2, \dots, K; i=1, 2, \dots, N$) are distributed according

to the NK -dimensional Gaussian distribution. If the coordinate components of an N -dimensional Euclidean space are statistically independent of each other, the joint distribution density of x_{hi} 's can be expressed in the form of such product of K -dimensional Gaussian distribution as seen in Eq. (4).

This property completely agrees with the starting point of §IV, §V and §VI in a previous paper (11). Accordingly, by thinking each energy E_h as the square of magnitude of each resultant vector \mathbf{R}_h , we can obtain many different explicit solutions to the generalized problem of random walks from separate energy distribution in the previous paper (11).

Moreover, if $N \rightarrow \infty \left(m = \frac{N}{2} = \frac{\langle E_h \rangle^2}{\sigma_{E_h}^2} \right)$ according to an estimation by the method of moment (2, 7), by using the central limit theorem in N dimensions, we can also find that the joint-probability density of E_h ($=R_h^2$, $h=1, 2, \dots, K$) is distributed asymptotically according to the K -dimensional Gaussian distribution.

6. Conclusion

In this paper, we have shown that the intensity fluctuation of random noise current can be investigated as a problem of many correlative series of random walks in multi-dimensional space, and from such a point of view, when we pay our attention to the energy fluctuation of white noise, the characteristic function of Hankel type is very useful for calculating its probability distribution. Then, for typical cases, we have derived several explicit expressions for the intensity distribution of white noise.

The following properties of random noise were derived:

(1) The number of dimensions for the used space is closely connected with the statistical characters of random noise and the characters of the observational device.

(2) The fluctuation in the output noise is reasonably described by one parameter m involved in the energy distribution expressions.

(3) The above parameter m is approximately equivalent to TW (W : equivalent noise band-width, T : time-interval of observation) appearing in the sampling theorem.

The method described in this paper is also applicable to the other fields of measurement on the random phenomena, since the mean energy is a conservative physical quantity.

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