

Integer Order Gaussian Moments

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In this work, integer order moments of Gaussian random variables are deduced.

I. GAUSSIAN MARGINAL MOMENT

Suppose a Gaussian random variable X , with mean a_X and variance σ_X^2 , and the zero mean variate $\tilde{X} = X - a_X$. Then, given $l \in \mathbb{N}$, it follows that

$$E \{X^l\} = \sum_{j=0}^l \left[\binom{l}{j} a_X^{l-j} E \{\tilde{X}^j\} \right]. \quad (1)$$

The expected value of \tilde{X}^j , $j \in \mathbb{N}$, is given in [1, equação (5-73)]:

$$E \{\tilde{X}^j\} = \begin{cases} (j-1)!! \sigma_X^j & j \text{ even} \\ 0 & j \text{ odd} \end{cases}, \quad (2a)$$

where

$$(j-1)!! = \begin{cases} (j-1) \cdot (j-3) \cdot \dots \cdot 6 \cdot 4 \cdot 2 & (j-1) \text{ even} \\ (j-1) \cdot (j-3) \cdot \dots \cdot 5 \cdot 3 \cdot 1 & (j-1) \text{ odd} \end{cases}. \quad (2b)$$

From (1) and (2),

$$E \{X^l\} = a_X^l \sum_{j=0}^{\lfloor l/2 \rfloor} \left[\binom{l}{2j} \frac{(2j-1)!!}{k_X^j} \right], \quad (3a)$$

where $k_X = a_X^2/\sigma_X^2$, and $\lfloor \cdot \rfloor$ is the largest integer less than or equal to the input value. Alternatively, using $(2j-1)!! = (2j)!/(2^j j!)$, $j \in \mathbb{N}$,

$$E \{X^l\} = a_X^l l! \sum_{j=0}^{\lfloor l/2 \rfloor} \left[\frac{1}{(2k_X)^j (l-2j)! j!} \right]. \quad (3b)$$

Then, the ratio

$$\gamma_1(l; k_X) = \frac{E \{X^{2l}\}}{\sigma_X^{2l}} \quad (4a)$$

is

$$\gamma_1(l; k_X) = \frac{(2l)!}{2^l} \sum_{j=0}^l \left[\frac{(2k_X)^j}{(2j)!(l-j)!} \right]. \quad (4b)$$

II. GAUSSIAN JOINT MOMENTS

A. Two Random Variables

Suppose the joint Gaussian random variables X and Y , respectively with means a_X and a_Y , variances σ_X^2 and σ_Y^2 , and correlation coefficient ν , and also the zero mean variates $\tilde{X} = X - a_X$ and $\tilde{Y} = Y - a_Y$. Then, given $l_i \in \mathbb{N}$, $i = 1, 2$, it follows that

$$E \{ X^{2l_1} Y^{2l_2} \} = \sum_{j_1=0}^{2l_1} \sum_{j_2=0}^{2l_2} \left[\binom{2l_1}{j_1} \binom{2l_2}{j_2} a_X^{2l_1-j_1} a_Y^{2l_2-j_2} E \{ \tilde{X}^{j_1} \tilde{Y}^{j_2} \} \right]. \quad (5)$$

Similarly, the ratios

$$\gamma_2(\mathbf{l}; \mathbf{k}; \nu) = \frac{E \{ X^{2l_1} Y^{2l_2} \}}{(\sigma_X^{2l_1} \sigma_Y^{2l_2})} \quad (6a)$$

and

$$\varrho_2(\mathbf{j}; \nu) = \frac{E \{ \tilde{X}^{j_1} \tilde{Y}^{j_2} \}}{(\sigma_X^{j_1} \sigma_Y^{j_2})}, \quad (6b)$$

where $\mathbf{l} = [l_1 \ l_2] \in \mathbb{N}^2$, $\mathbf{j} = [j_1 \ j_2] \in \mathbb{N}^2$, and $\mathbf{k} = [k_X \ k_Y] = [a_X^2/\sigma_X^2 \ a_Y^2/\sigma_Y^2]$, are related by

$$\gamma_2(\mathbf{l}; \mathbf{k}; \nu) = k_X^{l_1} k_Y^{l_2} \sum_{j_1=0}^{2l_1} \sum_{j_2=0}^{2l_2} \left[\binom{2l_1}{j_1} \binom{2l_2}{j_2} \frac{\varrho_2(\mathbf{j}; \nu)}{(k_X^{j_1} k_Y^{j_2})^{1/2}} \right]. \quad (6c)$$

The deduction of $\varrho_2(\mathbf{j}; \nu)$ is performed departing from the JPDP of \tilde{X} and \tilde{Y} ,

$$f_{\tilde{X}\tilde{Y}}(\tilde{x}, \tilde{y}) = \frac{1}{2\pi\sigma_X\sigma_Y\sqrt{1-\nu^2}} \exp \left[-\frac{1}{2(1-\nu^2)} \left(\frac{\tilde{x}^2}{\sigma_X^2} + \frac{\tilde{y}^2}{\sigma_Y^2} - 2\nu \frac{\tilde{x}\tilde{y}}{\sigma_X\sigma_Y} \right) \right], \quad (7a)$$

and from the definition of joint moment,

$$E \{ \tilde{X}^{j_1} \tilde{Y}^{j_2} \} = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \tilde{x}^{j_1} \tilde{y}^{j_2} f_{\tilde{X}\tilde{Y}}(\tilde{x}, \tilde{y}) d\tilde{x} d\tilde{y}. \quad (7b)$$

Then, substituting (7a) into (7b), using $\varrho_2(\mathbf{j}; \nu) = E \{ \tilde{X}^{j_1} \tilde{Y}^{j_2} \} / (\sigma_X^{j_1} \sigma_Y^{j_2})$, and making some algebraic

manipulations:

$$\varrho_2(\mathbf{j}; \nu) = \frac{1}{\sqrt{2\pi}\sigma_X^{j_1}\sigma_Y^{j_2+1}} \int_{-\infty}^{\infty} \tilde{y}^{j_2} M_{\tilde{X}}(\tilde{y}) \exp\left(-\frac{\tilde{y}^2}{\sigma_Y^2}\right) d\tilde{y}, \quad (8a)$$

where

$$M_{\tilde{X}}(\tilde{y}) = \frac{1}{\sqrt{2\pi(1-\nu^2)}\sigma_X} \int_{-\infty}^{\infty} \tilde{x}^{j_1} \exp\left[-\frac{1}{2(1-\nu^2)\sigma_X^2} \left(\tilde{x} - \nu\frac{\sigma_X}{\sigma_Y}\tilde{y}\right)^2\right] d\tilde{x}. \quad (8b)$$

Observing that $M_{\tilde{X}}(\tilde{y})$ is the moment of order $j_1 \in \mathbb{N}$ of a Gaussian random variable with mean $\nu(\sigma_X/\sigma_Y)\tilde{y}$ and variance $\sqrt{(1-\nu^2)}\sigma_X$, and using (3b),

$$M_{\tilde{X}}(\tilde{y}) = j_1! \nu^{j_1} \sigma_X^{j_1} \sum_{n=0}^{\lfloor j_1/2 \rfloor} \left[\frac{\sigma_Y^{2n-j_1} \tilde{y}^{j_1-2n}}{2^n n! (j_1-2n)!} \left(\frac{1-\nu^2}{\nu^2}\right)^n \right]. \quad (9)$$

Replacing (9) into (8a), it follows that

$$\varrho_2(\mathbf{j}; \nu) = j_1! \nu^{j_1} \sum_{n=0}^{\lfloor j_1/2 \rfloor} \left[\frac{\sigma_Y^{2n-j_1-j_2} M_{\tilde{Y}}}{2^n n! (j_1-2n)!} \left(\frac{1-\nu^2}{\nu^2}\right)^n \right], \quad (10a)$$

where

$$M_{\tilde{Y}} = \frac{1}{\sqrt{2\pi}\sigma_Y} \int_{-\infty}^{\infty} \tilde{y}^{j_1+j_2-2n} \exp\left(-\frac{\tilde{y}^2}{\sigma_Y^2}\right) d\tilde{y}. \quad (10b)$$

Observing that $M_{\tilde{Y}}$ is the moment of order $(j_1+j_2-2n) \in \mathbb{N}$ of a zero mean Gaussian random variable with variance σ_Y , and using (2),

$$M_{\tilde{Y}} = \begin{cases} (j_1+j_2-2n-1)!! \sigma_Y^{j_1+j_2-2n} & (j_1+j_2) \text{ even} \\ 0 & (j_1+j_2) \text{ odd} \end{cases}. \quad (11)$$

Thus, from (10a) and (11), it results that

$$\varrho_2(\mathbf{j}; \nu) = \begin{cases} j_1! \nu^{j_1} \sum_{n=0}^{\lfloor j_1/2 \rfloor} \left[\frac{(j_1+j_2-2n-1)!!}{2^n n! (j_1-2n)!} \left(\frac{1-\nu^2}{\nu^2}\right)^n \right] & (j_1+j_2) \text{ even} \\ 0 & (j_1+j_2) \text{ odd} \end{cases}. \quad (12)$$

B. Four Random Variables

Suppose the joint Gaussian random variables X_1 , Y_1 , X_2 , and Y_2 , respectively with means a_{X_1} , a_{Y_1} , a_{X_2} , and a_{Y_2} , and variances $\sigma_{X_1}^2$, $\sigma_{Y_1}^2$, $\sigma_{X_2}^2$, and $\sigma_{Y_2}^2$. Besides, assume X_i uncorrelated of Y_i , $i = 1, 2$, and

the correlation coefficients

$$\nu_1 = \frac{Cov\{X_1, X_2\}}{\sigma_{X_1}\sigma_{X_2}} = \frac{Cov\{Y_1, Y_2\}}{\sigma_{Y_1}\sigma_{Y_2}} \quad (13a)$$

and

$$\nu_2 = \frac{Cov\{X_1, Y_2\}}{\sigma_{X_1}\sigma_{Y_2}} = -\frac{Cov\{Y_1, X_2\}}{\sigma_{Y_1}\sigma_{X_2}}. \quad (13b)$$

Now, consider $\tilde{X}_1 = X_1 - a_{X_1}$, $\tilde{Y}_1 = Y_1 - a_{Y_1}$, $\tilde{X}_2 = X_2 - a_{X_2}$, and $\tilde{Y}_2 = Y_2 - a_{Y_2}$. Then, given $l_i \in \mathbb{N}$, $i = 1, 2, 3, 4$, it follows that

$$\begin{aligned} E\{X_1^{2l_1}Y_1^{2l_2}X_2^{2l_3}Y_2^{2l_4}\} &= \sum_{j_1=0}^{2l_1} \sum_{j_2=0}^{2l_2} \sum_{j_3=0}^{2l_3} \sum_{j_4=0}^{2l_4} \left[\binom{2l_1}{j_1} \binom{2l_2}{j_2} \binom{2l_3}{j_3} \binom{2l_4}{j_4} \right. \\ &\quad \left. \times a_{X_1}^{2l_1-j_1} a_{Y_1}^{2l_2-j_2} a_{X_2}^{2l_3-j_3} a_{Y_2}^{2l_4-j_4} E\{\tilde{X}_1^{j_1}\tilde{Y}_1^{j_2}\tilde{X}_2^{j_3}\tilde{Y}_2^{j_4}\} \right]. \end{aligned} \quad (14)$$

Analogously, the relation between the ratios

$$\gamma_4(\mathbf{l}; \mathbf{k}; \boldsymbol{\nu}) \triangleq \frac{E\{X_1^{2l_1}Y_1^{2l_2}X_2^{2l_3}Y_2^{2l_4}\}}{(\sigma_{X_1}^{2l_1}\sigma_{Y_1}^{2l_2}\sigma_{X_2}^{2l_3}\sigma_{Y_2}^{2l_4})} \quad (15a)$$

and

$$\varrho_4(\mathbf{j}; \boldsymbol{\nu}) \triangleq \frac{E\{\tilde{X}_1^{j_1}\tilde{Y}_1^{j_2}\tilde{X}_2^{j_3}\tilde{Y}_2^{j_4}\}}{(\sigma_{X_1}^{j_1}\sigma_{Y_1}^{j_2}\sigma_{X_2}^{j_3}\sigma_{Y_2}^{j_4})}, \quad (15b)$$

where $\mathbf{l} = [l_1 \ l_2 \ l_3 \ l_4] \in \mathbb{N}^4$, $\mathbf{k} = [k_{X_1} \ k_{Y_1} \ k_{X_2} \ k_{Y_2}] = [a_{X_1}^2/\sigma_{X_1}^2 \ a_{Y_1}^2/\sigma_{Y_1}^2 \ a_{X_2}^2/\sigma_{X_2}^2 \ a_{Y_2}^2/\sigma_{Y_2}^2]$, $\boldsymbol{\nu} = [\nu_1 \ \nu_2]$, and $\mathbf{j} = [j_1 \ j_2 \ j_3 \ j_4] \in \mathbb{N}^4$, is

$$\gamma_4(\mathbf{l}; \mathbf{k}; \boldsymbol{\nu}) = k_{X_1}^{l_1} k_{Y_1}^{l_2} k_{X_2}^{l_3} k_{Y_2}^{l_4} \sum_{j_1=0}^{2l_1} \sum_{j_2=0}^{2l_2} \sum_{j_3=0}^{2l_3} \sum_{j_4=0}^{2l_4} \left[\binom{2l_1}{j_1} \binom{2l_2}{j_2} \binom{2l_3}{j_3} \binom{2l_4}{j_4} \frac{\varrho_4(\mathbf{j}; \boldsymbol{\nu})}{(k_{X_1}^{j_1} k_{Y_1}^{j_2} k_{X_2}^{j_3} k_{Y_2}^{j_4})^{1/2}} \right]. \quad (16)$$

The deduction of $\varrho_4(\mathbf{j}; \boldsymbol{\nu})$ is performed departing from the JPDPF of $\tilde{\mathbf{X}} = [\tilde{X}_1 \ \tilde{X}_2]$ and $\tilde{\mathbf{Y}} = [\tilde{Y}_1 \ \tilde{Y}_2]$,

$$\begin{aligned} f_{\tilde{\mathbf{X}}\tilde{\mathbf{Y}}}(\tilde{\mathbf{x}}, \tilde{\mathbf{y}}) &= \frac{1}{4\pi^2\sigma_{X_1}\sigma_{Y_1}\sigma_{X_2}\sigma_{Y_2}(1-\rho^2)} \exp \left[-\frac{1}{2(1-\rho^2)} \left(\frac{\tilde{x}_1^2}{\sigma_{X_1}^2} + \frac{\tilde{y}_1^2}{\sigma_{Y_1}^2} + \frac{\tilde{x}_2^2}{\sigma_{X_2}^2} + \frac{\tilde{y}_2^2}{\sigma_{Y_2}^2} \right. \right. \\ &\quad \left. \left. - 2\nu_1 \frac{\tilde{x}_1\tilde{x}_2}{\sigma_{X_1}\sigma_{X_2}} - 2\nu_1 \frac{\tilde{y}_1\tilde{y}_2}{\sigma_{Y_1}\sigma_{Y_2}} - 2\nu_2 \frac{\tilde{x}_1\tilde{y}_2}{\sigma_{X_1}\sigma_{Y_2}} + 2\nu_2 \frac{\tilde{y}_1\tilde{x}_2}{\sigma_{Y_1}\sigma_{X_2}} \right) \right], \end{aligned} \quad (17a)$$

where $\tilde{\mathbf{x}} = [\tilde{x}_1 \ \tilde{x}_2]$, $\tilde{\mathbf{y}} = [\tilde{y}_1 \ \tilde{y}_2]$, and $\rho = \sqrt{\nu_1^2 + \nu_2^2}$, and from the definition of joint moment,

$$E \left\{ \tilde{X}_1^{j_1} \tilde{Y}_1^{j_2} \tilde{X}_2^{j_3} \tilde{Y}_2^{j_4} \right\} = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \tilde{x}_1^{j_1} \tilde{y}_1^{j_2} \tilde{x}_2^{j_3} \tilde{y}_2^{j_4} f_{\tilde{\mathbf{X}}\tilde{\mathbf{Y}}}(\tilde{\mathbf{x}}, \tilde{\mathbf{y}}) d\tilde{x}_1 d\tilde{y}_1 d\tilde{x}_2 d\tilde{y}_2. \quad (17b)$$

Then, substituting (17a) into (17b), using $\varrho_4(\mathbf{j}; \boldsymbol{\nu}) = E \left\{ \tilde{X}_1^{j_1} \tilde{Y}_1^{j_2} \tilde{X}_2^{j_3} \tilde{Y}_2^{j_4} \right\} / (\sigma_{X_1}^{j_1} \sigma_{Y_1}^{j_2} \sigma_{X_2}^{j_3} \sigma_{Y_2}^{j_4})$, and making some algebraic manipulations:

$$\varrho_4(\mathbf{j}; \boldsymbol{\nu}) = \frac{(\sigma_{X_1}^{j_1} \sigma_{Y_1}^{j_2} \sigma_{X_2}^{j_3} \sigma_{Y_2}^{j_4})^{-1}}{2\pi \sigma_{X_2} \sigma_{Y_2}} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \tilde{x}_2^{j_3} \tilde{y}_2^{j_4} M_{\tilde{X}_1}(\tilde{x}_2, \tilde{y}_2) M_{\tilde{Y}_1}(\tilde{x}_2, \tilde{y}_2) \exp\left(-\frac{\tilde{x}_2^2}{\sigma_{X_2}^2} - \frac{\tilde{y}_2^2}{\sigma_{Y_2}^2}\right) d\tilde{x}_2 d\tilde{y}_2, \quad (18a)$$

where

$$M_{\tilde{X}_1}(\tilde{x}_2, \tilde{y}_2) = \frac{1}{\sqrt{2\pi(1-\rho^2)}\sigma_{X_1}} \int_{-\infty}^{\infty} \tilde{x}_1^{j_1} \exp\left[-\frac{1}{2(1-\rho^2)\sigma_{X_1}^2} \left(\tilde{x}_1 - \nu_1 \frac{\sigma_{X_1}}{\sigma_{X_2}} \tilde{x}_2 - \nu_2 \frac{\sigma_{X_1}}{\sigma_{Y_2}} \tilde{y}_2\right)^2\right] d\tilde{x}_1 \quad (18b)$$

$$M_{\tilde{Y}_1}(\tilde{x}_2, \tilde{y}_2) = \frac{1}{\sqrt{2\pi(1-\rho^2)}\sigma_{Y_1}} \int_{-\infty}^{\infty} \tilde{y}_1^{j_2} \exp\left[-\frac{1}{2(1-\rho^2)\sigma_{Y_1}^2} \left(\tilde{y}_1 - \nu_1 \frac{\sigma_{Y_1}}{\sigma_{Y_2}} \tilde{y}_2 + \nu_2 \frac{\sigma_{Y_1}}{\sigma_{X_2}} \tilde{x}_2\right)^2\right] d\tilde{y}_1. \quad (18c)$$

Note that $M_{\tilde{X}_1}(\tilde{x}_2, \tilde{y}_2)$ and $M_{\tilde{Y}_1}(\tilde{x}_2, \tilde{y}_2)$ are moments, respectively of orders $j_1 \in \mathbb{N}$ and $j_2 \in \mathbb{N}$, of Gaussian random variables, respectively with means $[\nu_1(\sigma_{X_1}/\sigma_{X_2})\tilde{x}_2 + \nu_2(\sigma_{X_1}/\sigma_{Y_2})\tilde{y}_2]$ and $[\nu_1(\sigma_{Y_1}/\sigma_{Y_2})\tilde{y}_2 - \nu_2(\sigma_{Y_1}/\sigma_{X_2})\tilde{x}_2]$ and variances $\sqrt{(1-\rho^2)}\sigma_{X_1}$ and $\sqrt{(1-\rho^2)}\sigma_{Y_1}$. Then, using (3b) for $M_{\tilde{X}_1}(\tilde{x}_2, \tilde{y}_2)$ and $M_{\tilde{Y}_1}(\tilde{x}_2, \tilde{y}_2)$,

$$M_{\tilde{X}_1}(\tilde{x}_2, \tilde{y}_2) = j_1! \sigma_{X_1}^{j_1} \sum_{n_1=0}^{\lfloor j_1/2 \rfloor} \left[\frac{(1-\rho^2)^{n_1}}{2^{n_1} n_1! (j_1 - 2n_1)!} \left(\nu_1 \frac{\tilde{x}_2}{\sigma_{X_2}} + \nu_2 \frac{\tilde{y}_2}{\sigma_{Y_2}} \right)^{j_1 - 2n_1} \right] \quad (19a)$$

$$M_{\tilde{Y}_1}(\tilde{x}_2, \tilde{y}_2) = j_2! \sigma_{Y_1}^{j_2} \sum_{n_2=0}^{\lfloor j_2/2 \rfloor} \left[\frac{(1-\rho^2)^{n_2}}{2^{n_2} n_2! (j_2 - 2n_2)!} \left(\nu_1 \frac{\tilde{y}_2}{\sigma_{Y_2}} - \nu_2 \frac{\tilde{x}_2}{\sigma_{X_2}} \right)^{j_2 - 2n_2} \right]. \quad (19b)$$

Applying the Newton's Binomial formula,

$$\left(\nu_1 \frac{\tilde{x}_2}{\sigma_{X_2}} + \nu_2 \frac{\tilde{y}_2}{\sigma_{Y_2}} \right)^{j_1 - 2n_1} = \sum_{n_3=0}^{j_1 - 2n_1} \left[\binom{j_1 - 2n_1}{n_3} \nu_1^{n_3} \nu_2^{j_1 - 2n_1 - n_3} \frac{\tilde{x}_2^{n_3} \tilde{y}_2^{j_1 - 2n_1 - n_3}}{\sigma_{X_2}^{n_3} \sigma_{Y_2}^{j_1 - 2n_1 - n_3}} \right], \quad (20a)$$

$$\left(\nu_1 \frac{\tilde{y}_2}{\sigma_{Y_2}} - \nu_2 \frac{\tilde{x}_2}{\sigma_{X_2}} \right)^{j_2 - 2n_2} = \sum_{n_4=0}^{j_2 - 2n_2} \left[\binom{j_2 - 2n_2}{n_4} (-1)^{n_4} \nu_1^{j_2 - 2n_2 - n_4} \nu_2^{n_4} \frac{\tilde{x}_2^{n_4} \tilde{y}_2^{j_2 - 2n_2 - n_4}}{\sigma_{X_2}^{n_4} \sigma_{Y_2}^{j_2 - 2n_2 - n_4}} \right]. \quad (20b)$$

From (18), (19), and (20),

$$\begin{aligned} \varrho_4(\mathbf{j}; \boldsymbol{\nu}) &= j_1! j_2! \nu_1^{j_2} \nu_2^{j_1} \sum_{n_1=0}^{\lfloor j_1/2 \rfloor} \sum_{n_2=0}^{\lfloor j_2/2 \rfloor} \sum_{n_3=0}^{j_1-2n_1} \sum_{n_4=0}^{j_2-2n_2} \left[\binom{j_1-2n_1}{n_3} \binom{j_2-2n_2}{n_4} \frac{1}{(j_1-2n_1)!(j_2-2n_2)!} \right. \\ &\quad \left. \times \frac{M_{\tilde{X}_2} M_{\tilde{Y}_2}}{\sigma_{X_2}^{j_3+n_3+n_4} \sigma_{Y_2}^{j_1+j_2+j_4-2n_1-2n_2-n_3-n_4}} \frac{(-1)^{n_4} (1-\rho^2)^{n_1+n_2}}{2^{n_1+n_2} n_1! n_2! \nu_1^{2n_2} \nu_2^{2n_1}} \left(\frac{\nu_1}{\nu_2} \right)^{n_3-n_4} \right], \end{aligned} \quad (21a)$$

where

$$M_{\tilde{X}_2} = \frac{1}{\sqrt{2\pi}\sigma_{X_2}} \int_{-\infty}^{\infty} \tilde{x}_2^{j_3+n_3+n_4} \exp\left(-\frac{\tilde{x}_2^2}{\sigma_{X_2}^2}\right) d\tilde{x}_2 \quad (21b)$$

$$M_{\tilde{Y}_2} = \frac{1}{\sqrt{2\pi}\sigma_{Y_2}} \int_{-\infty}^{\infty} \tilde{y}_2^{j_1+j_2+j_4-2n_1-2n_2-n_3-n_4} \exp\left(-\frac{\tilde{y}_2^2}{\sigma_{Y_2}^2}\right) d\tilde{y}_2. \quad (21c)$$

Note that $M_{\tilde{X}_2}$ and $M_{\tilde{Y}_2}$ are moments, respectively of orders $(j_3 + n_3 + n_4) \in \mathbb{N}$ and $(j_1 + j_2 + j_4 - 2n_1 - 2n_2 - n_3 - n_4) \in \mathbb{N}$, of zero mean Gaussian random variables, respectively with variances σ_{X_2} and σ_{Y_2} . Therefore, using (2) for $M_{\tilde{X}_2}$ and $M_{\tilde{Y}_2}$

$$M_{\tilde{X}_2} = \begin{cases} (j_3 + n_3 + n_4 - 1)!! \sigma_{X_2}^{j_3+n_3+n_4} & (j_3 + n_3 + n_4) \text{ even} \\ 0 & (j_3 + n_3 + n_4) \text{ odd} \end{cases} \quad (22a)$$

$$M_{\tilde{Y}_2} = \begin{cases} \frac{(j_1 + j_2 + j_4 - 2n_1 - 2n_2 - n_3 - n_4 - 1)!!}{\sigma_{Y_2}^{-(j_1+j_2+j_4-2n_1-2n_2-n_3-n_4)}} & (j_1 + j_2 + j_4 - n_3 - n_4) \text{ even} \\ 0 & (j_1 + j_2 + j_4 - n_3 - n_4) \text{ odd} \end{cases}. \quad (22b)$$

Thus, the product $M_{\tilde{X}_2} M_{\tilde{Y}_2}$ is

$$M_{\tilde{X}_2} M_{\tilde{Y}_2} = \frac{(j_3 + n_3 + n_4 - 1)!! (j_1 + j_2 + j_4 - 2n_1 - 2n_2 - n_3 - n_4 - 1)!!}{\sigma_{X_2}^{-(j_3+n_3+n_4)} \sigma_{Y_2}^{-(j_1+j_2+j_4-2n_1-2n_2-n_3-n_4)}} \quad (23a)$$

if both $(j_1 + j_2 + j_3 + j_4)$ and $(j_3 + n_3 + n_4)$ are even, and

$$M_{\tilde{X}_2} M_{\tilde{Y}_2} = 0 \quad (23b)$$

otherwise.

From (21) and (23),

$$\begin{aligned} \varrho_4(\mathbf{j}; \boldsymbol{\nu}) &= j_1! j_2! \nu_1^{j_2} \nu_2^{j_1} \sum_{n_1=0}^{\lfloor j_1/2 \rfloor} \sum_{n_2=0}^{\lfloor j_2/2 \rfloor} \sum_{n_3=0}^{j_1-2n_1} \sum_{n_4=0}^{j_2-2n_2} \left[\frac{(j_1 + j_2 + j_4 - 2n_1 - 2n_2 - n_3 - n_4 - 1)!!}{(j_1 - 2n_1 - n_3)!(j_2 - 2n_2 - n_4)!} \right. \\ &\quad \left. \times \frac{[(-1)^{j_3+n_3} + (-1)^{n_4}] (j_3 + n_3 + n_4 - 1)!! (1-\rho^2)^{n_1+n_2}}{2^{n_1+n_2+1} n_1! n_2! n_3! n_4!} \frac{(1-\rho^2)^{n_1+n_2}}{\nu_1^{2n_2} \nu_2^{2n_1}} \left(\frac{\nu_1}{\nu_2} \right)^{n_3-n_4} \right] \end{aligned} \quad (24a)$$

for $(j_1 + j_2 + j_3 + j_4)$ even, and

$$\varrho_4(\mathbf{j}; \boldsymbol{\nu}) = 0 \quad (24b)$$

for $(j_1 + j_2 + j_3 + j_4)$ odd.

REFERENCES

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