The expectation of a product of Gaussian random variables

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October 16, 2007

Let X_1, X_2, \ldots, X_{2n} be a collection of random variables which are jointly Gaussian. Assume that each X_j has mean zero and variance one. In this note, we will derive a formula for the expectation of their product in terms of their pairwise covariances. To state the formula, we introduce the following notation. A partition of a set S is a collection of pairwise disjoint, nonempty subsets of S whose union is S. Let A be a partition of $\{1, 2, \ldots, 2n\}$. We will call A a pair-partition if each set $A \in A$ has exactly two elements. (In particular, this means A consists of exactly n sets.) Let $A = A^n$ denote the set of all pair-partitions of $\{1, 2, \ldots, 2n\}$. We shall show that

$$E\left[\prod_{j=1}^{2n} X_j\right] = \sum_{\mathcal{A} \in \mathbb{A}} \prod_{\{i,j\} \in \mathcal{A}} \rho_{ij},$$

where $\rho_{ij} = E[X_i X_j]$. For example,

$$E[X_1X_2X_3X_4] = \rho_{12}\rho_{34} + \rho_{13}\rho_{24} + \rho_{14}\rho_{23},$$

and

$$E[X_1X_2X_3X_4X_5X_6] = \rho_{12}\rho_{34}\rho_{56} + \rho_{12}\rho_{35}\rho_{46} + \rho_{12}\rho_{36}\rho_{45}$$

$$+ \rho_{13}\rho_{24}\rho_{56} + \rho_{13}\rho_{25}\rho_{46} + \rho_{13}\rho_{26}\rho_{45}$$

$$+ \rho_{14}\rho_{23}\rho_{56} + \rho_{14}\rho_{25}\rho_{36} + \rho_{14}\rho_{26}\rho_{35}$$

$$+ \rho_{15}\rho_{23}\rho_{46} + \rho_{15}\rho_{24}\rho_{36} + \rho_{15}\rho_{26}\rho_{34}$$

$$+ \rho_{16}\rho_{23}\rho_{45} + \rho_{16}\rho_{24}\rho_{35} + \rho_{16}\rho_{25}\rho_{34}.$$

To prove this formula, we begin with the following lemma.

Lemma 1 If X_1, \ldots, X_{2n} are jointly Gaussian with $EX_j = 0$ and $EX_j^2 = 1$, then

$$E\left[\prod_{j=1}^{2n} X_j\right] = \frac{1}{n} \sum_{i < j} \rho_{ij} E\left[\prod_{k \notin \{i,j\}} X_k\right],$$

where $\rho_{ij} = E[X_i X_j]$.

Proof. Let W_1, \ldots, W_{2n} be dependent Brownian motions with covariation $[W_i, W_j]_t = \rho_{ij}t$. Using Itô's formula, we have

$$E[X_1 \cdots X_{2n}] = E[W_1(1) \cdots W_{2n}(1)]$$

$$= \sum_{i < j} \int_0^1 E\left[\prod_{k \notin \{i,j\}} W_k(s)\right] \rho_{ij} ds$$

$$= \sum_{i < j} \int_0^1 (\sqrt{s})^{2n-2} E\left[\prod_{k \notin \{i,j\}} X_k\right] \rho_{ij} ds$$

$$= \sum_{i < j} \frac{1}{n} E\left[\prod_{k \notin \{i,j\}} X_k\right] \rho_{ij},$$

completing the proof.

We can now prove the general formula.

Theorem 2 If X_1, \ldots, X_{2n} are jointly Gaussian with $EX_j = 0$ and $EX_j^2 = 1$, then

$$E\left[\prod_{j=1}^{2n} X_j\right] = \sum_{\mathcal{A} \in \mathbb{A}} \prod_{\{i,j\} \in \mathcal{A}} \rho_{ij},\tag{1}$$

where $\rho_{ij} = E[X_i X_j]$.

Proof. If n = 1, the claim is trivial. Suppose the formula holds for n - 1. For each i and j, define $\mathbb{A}_{ij} = \{A \in \mathbb{A}^n : \{i, j\} \in A\}$ and $\mathbb{B}_{ij} = \{A \setminus \{\{i, j\}\}\} : A \in \mathbb{A}_{ij}\}$. Note that \mathbb{B}_{ij} is simply the set of pair-partitions of $\{1, \ldots, 2n\} \setminus \{i, j\}$. By Lemma 1 and the inductive hypothesis,

$$E\left[\prod_{j=1}^{2n} X_j\right] = \frac{1}{n} \sum_{i < j} \rho_{ij} E\left[\prod_{k \notin \{i, j\}} X_k\right]$$
$$= \frac{1}{n} \sum_{i < j} \rho_{ij} \sum_{\mathcal{A} \in \mathbb{B}_{ij}} \prod_{\{m, n\} \in \mathcal{A}} \rho_{mn}$$
$$= \frac{1}{n} \sum_{i < j} \sum_{\mathcal{A} \in \mathbb{A}_{ij}} \prod_{\{m, n\} \in \mathcal{A}} \rho_{mn}.$$

Since each partition in \mathbb{A}^n is counted exactly n times in the above double sum, this completes the proof.

Corollary 3 If X_1, \ldots, X_{2n} are jointly Gaussian with $EX_j = 0$ and $EX_j^2 = 1$, then

$$E\left[\prod_{j=1}^{2n} X_{j}\right] = \sum_{k=2}^{2n} \rho_{1k} E\left[\prod_{\ell \notin \{1,k\}} X_{\ell}\right],$$

where $\rho_{1k} = E[X_1 X_k]$.

Proof. By Theorem 2,

$$E\left[\prod_{j=1}^{2n} X_j\right] = \sum_{\mathcal{A} \in \mathbb{A}} \prod_{\{i,j\} \in \mathcal{A}} \rho_{ij}$$

$$= \sum_{k=2}^{2n} \sum_{\mathcal{A} \in \mathbb{A}_{1k}} \prod_{\{i,j\} \in \mathcal{A}} \rho_{ij}$$

$$= \sum_{k=2}^{2n} \rho_{1k} \sum_{\mathcal{A} \in \mathbb{B}_{1k}} \prod_{\{i,j\} \in \mathcal{A}} \rho_{ij}.$$

Since \mathbb{B}_{1k} is the set of pair-partitions of $\{1, \ldots, 2n\} \setminus \{1, k\}$, a second application of Theorem 2 completes the proof.

As an example of the use of Corollary 3, we compute the following:

$$E[X_1 X_2^2 X_3 X_4^2] = 2\rho_{12} E[X_2 X_3 X_4^2] + \rho_{13} E[X_2^2 X_4^2] + 2\rho_{14} E[X_2^2 X_3 X_4]$$

$$= 2\rho_{12}(\rho_{23} + 2\rho_{24}\rho_{34}) + \rho_{13}(1 + 2\rho_{24}^2) + 2\rho_{14}(\rho_{34} + 2\rho_{23}\rho_{24})$$

$$= 2\rho_{12}\rho_{23} + 4\rho_{12}\rho_{24}\rho_{34} + \rho_{13} + 2\rho_{13}\rho_{24}^2 + 2\rho_{14}\rho_{34} + 4\rho_{14}\rho_{23}\rho_{24}.$$

As can be seen, Corollary 3 provides an iterative method for computing this expectation. The assumption that each X_j has variance 1 is unnecessary.

Corollary 4 If Y_1, \ldots, Y_{2n} are jointly Gaussian, then

$$E\left[\prod_{j=1}^{2n} Y_j\right] = \sum_{k=2}^{2n} E[Y_1 Y_k] E\left[\prod_{\ell \notin \{1,k\}} Y_\ell\right],$$

provided $EY_j = 0$ for all j.

Proof. Let
$$\sigma_j^2 = EY_j^2$$
, $X_j = \sigma_j^{-1}Y_j$, and $\rho_{ij} = E[X_iX_j]$. Apply Corollary 3.

Acknowledgments

Thanks go to Wenbo Li for bringing Equation (1) to my attention.