

Some Useful Expected Values with Multivariate Normal Distribution and Uniform Distribution on Sphere

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1 Useful values with multivariate normally distributed variables

Let $\epsilon \in \mathbb{R}^n$ be a set of variables following multivariate normal distribution: $\epsilon \sim \mathcal{N}(0_{n \times 1}, I_{n \times n})$. Let a be an arbitrary constant vector in \mathbb{R}^n . We will derive the following expectations.

$$\mathbb{E}_{\epsilon \sim \mathcal{N}(0, I)} (\epsilon^\top \epsilon)^k = \frac{(n + 2k - 2)!!}{(n - 2)!!} = n(n + 2)(n + 4) \cdots (n + 2k - 2) \quad (1a)$$

$$\mathbb{E}_{\epsilon \sim \mathcal{N}(0, I)} (\epsilon^\top \epsilon)^k \epsilon \epsilon^\top = \frac{(n + 2k)!!}{n!!} I = (n + 2)(n + 4) \cdots (n + 2k) I \quad (1b)$$

$$\mathbb{E}_{\epsilon \sim \mathcal{N}(0, I)} (a^\top \epsilon)^{2k} = (a^\top a)^k \quad (1c)$$

$$\mathbb{E}_{\epsilon \sim \mathcal{N}(0, I)} (a^\top \epsilon)^{2k_1} (\epsilon^\top \epsilon)^{k_2} = (2k_1 - 1)!! \frac{(n + 2k_1 + 2k_2 - 2)!!}{(n + 2k_1 - 2)!!} (a^\top a)^{k_1} \quad (1d)$$

There is also an attempted derivation of

$$\mathbb{E}_{\epsilon \sim \mathcal{N}(0, I)} \mathbb{E}_{\bar{\epsilon} \sim \mathcal{N}(0, I)} (a^\top \epsilon)^{2k_1} (a^\top \bar{\epsilon})^{2k_2} (\epsilon^\top \bar{\epsilon})^{k_3}$$

with k_1, k_2, k_3 being natural numbers, but I was unable to get the result. Instead, the direct approach was used to calculate the following two expectations used in the paper:

$$\begin{aligned} & \mathbb{E}_{\epsilon \sim \mathcal{N}(0, I)} \mathbb{E}_{\bar{\epsilon} \sim \mathcal{N}(0, I)} (a^\top \epsilon)^2 (a^\top \bar{\epsilon})^2 (\epsilon^\top \bar{\epsilon})^2 \\ & \mathbb{E}_{\epsilon \sim \mathcal{N}(0, I)} \mathbb{E}_{\bar{\epsilon} \sim \mathcal{N}(0, I)} (a^\top \epsilon)^3 (a^\top \bar{\epsilon}) (\epsilon^\top \bar{\epsilon}). \end{aligned}$$

To calculate the above values, we need to first notice ϵ_i and ϵ_j , the i th and j th element of ϵ , each follows standard normal distribution independently for any $i \neq j$. In addition, any odd moment of a normally distributed variable is 0, i.e., for any $k \geq 0$ that is odd

$$\mathbb{E}_{\epsilon_i \sim \mathcal{N}(0, 1)} \epsilon_i^k = 0. \quad (2)$$

$$\mathbb{E}_{\epsilon \sim \mathcal{N}(0, I)} (\epsilon^\top \epsilon)^k = n(n+2)(n+4) \cdots (n+2k-2)$$

$$\mathbb{E}_{\epsilon \sim \mathcal{N}(0, I)} (\epsilon^\top \epsilon)^k \epsilon \epsilon^\top = (n+2)(n+4) \cdots (n+2k) I$$

Notice that $\mathbb{E}_{\epsilon \sim \mathcal{N}(0, I)} (\epsilon^\top \epsilon)^k$ is the k th moment of a Chi-square distributed random variable, so we can get this expectation by simply looking it up in Wikipedia.

The expectation $\mathbb{E}_{\epsilon \sim \mathcal{N}(0, I)} (\epsilon^\top \epsilon)^k \epsilon \epsilon^\top$ is an $n \times n$ matrix. Let's denote this matrix as A , and the element on its i th row and j th column as A_{ij} . We can see

$$A_{ij} = (\epsilon^\top \epsilon)^k \epsilon_i \epsilon_j = \left(\sum_{t=1}^n \epsilon_t^2 \right)^k \epsilon_i \epsilon_j.$$

By expanding $(\sum_{t=1}^n \epsilon_t^2)^k$, A_{ij} can be written as the sum of n^k terms. Let $\epsilon_{t_1}^2 \epsilon_{t_2}^2 \cdots \epsilon_{t_k}^2 \epsilon_i \epsilon_j$ be one of those terms, where $(t_1, t_2, \dots, t_k) \in \{1, 2, \dots, n\}^k$. Considering (2), the only way $\mathbb{E} \epsilon_{t_1}^2 \epsilon_{t_2}^2 \cdots \epsilon_{t_k}^2 \epsilon_i \epsilon_j$ is not zero is when $i \neq j$. Thus $A_{ij} = 0$ for all (i, j) such that $i \neq j$.

Now we know $A = \mathbb{E}_{\epsilon \sim \mathcal{N}(0, I)} (\epsilon^\top \epsilon)^k \epsilon \epsilon^\top$ is a diagonal matrix. Also notice

$$\sum_{i=1}^n A_{ii} = \sum_{i=1}^n \mathbb{E}_{\epsilon \sim \mathcal{N}(0, I)} (\epsilon^\top \epsilon)^k \epsilon_i^2 = \mathbb{E}_{\epsilon \sim \mathcal{N}(0, I)} (\epsilon^\top \epsilon)^{k+1}.$$

We should also have $A_{ii} = A_{jj}$ for all $(i, j) \in \{1, 2, \dots, n\}$ by symmetry, so

$$A = \frac{1}{n} \mathbb{E}_{\epsilon \sim \mathcal{N}(0, I)} (\epsilon^\top \epsilon)^{k+1} I = (n+2)(n+4) \cdots (n+2k) I.$$

$$\mathbb{E}_{\epsilon \sim \mathcal{N}(0, I)} (a^\top \epsilon)^{2k} = (a^\top a)^k$$

The moment generating function of $a^\top \epsilon$ is

$$\begin{aligned} M_{a^\top \epsilon}(t) &= \mathbb{E} \exp(t a^\top \epsilon) = \mathbb{E} \prod_{i=1}^n \exp(t a_i \epsilon_i) \\ &= \prod_{i=1}^n \mathbb{E} \exp(t a_i \epsilon_i) = \prod_{i=1}^n \exp\left(\frac{1}{2} t^2 a_i^2\right) = \exp\left(\frac{1}{2} t^2 a^\top a\right) \\ &= \sum_{i=1}^{\infty} \frac{1}{i!} \left(\frac{1}{2} t^2 a^\top a\right)^i. \end{aligned}$$

Thus we have

$$\begin{aligned} &\mathbb{E}_{\epsilon \sim \mathcal{N}(0, I)} (a^\top \epsilon)^{2k} \\ &= \frac{d^{2k}}{dt^{2k}} M_{a^\top \epsilon}(t) = \frac{d^{2k}}{dt^{2k}} \frac{1}{k!} \left(\frac{1}{2} t^2 a^\top a\right)^k \\ &= \frac{(2k)!}{2^k k!} (a^\top a)^k = (2k-1)!! (a^\top a)^k \end{aligned}$$

$$\mathbb{E}_{\epsilon \sim \mathcal{N}(0, I)} (a^\top \epsilon)^{2k_1} (\epsilon^\top \epsilon)^{k_2} = (2k_1 - 1)!! \frac{(n+2k_1+2k_2-2)!!}{(n+2k_1-2)!!} (a^\top a)^{k_1}$$

The moment generating function for $(a^\top \epsilon)(\epsilon^\top \epsilon)$ is

$$\begin{aligned} & M_{a^\top \epsilon, \epsilon^\top \epsilon}(t_1, t_2) \\ &= \mathbb{E} \exp(t_1(a^\top \epsilon)) \exp(t_2(\epsilon^\top \epsilon)) = \prod_{i=1}^n \mathbb{E} \exp(t_1 a_i \epsilon_i + t_2 \epsilon_i^2) \\ &= \prod_{i=1}^n \int_{\mathbb{R}^n} \exp(t_1 a_i u_i + t_2 u_i^2) \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{u_i^2}{2}\right) du_i \\ &= \prod_{i=1}^n \frac{1}{\sqrt{1-2t_2}} \exp\left(\frac{(t_1 a_i)^2}{2(1-2t_2)}\right) \int_{\mathbb{R}^n} \frac{\sqrt{1-2t_2}}{\sqrt{2\pi}} \exp\left(-\frac{1-2t_2}{2} \left(u_i - \frac{t_1 a_i}{1-2t_2}\right)^2\right) du_i \\ &= \prod_{i=1}^n \frac{1}{\sqrt{1-2t_2}} \exp\left(\frac{(t_1 a_i)^2}{2(1-2t_2)}\right) = (1-2t_2)^{-\frac{n}{2}} \exp\left(\frac{t_1^2 a^\top a}{2(1-2t_2)}\right) \\ &= (1-2t_2)^{-\frac{n}{2}} \sum_{i=1}^{\infty} \frac{1}{i!} \left(\frac{t_1^2 a^\top a}{2(1-2t_2)}\right)^i. \end{aligned}$$

Taking the $2k_1$ th derivative of $M_{a^\top \epsilon, \epsilon^\top \epsilon}(t_1, t_2)$ with respect to t_1 yields

$$\begin{aligned} & \frac{\partial^{2k_1}}{\partial t_1^{2k_1}} M_{a^\top \epsilon, \epsilon^\top \epsilon}(t_1, t_2) \\ &= (1-2t_2)^{-\frac{n}{2}} \frac{1}{k_1!} \left(\frac{a^\top a}{2(1-2t_2)}\right)^{k_1} (2k_1)! \\ &= (1-2t_2)^{-\frac{n+2k_1}{2}} (2k_1 - 1)!! (a^\top a)^{k_1}, \end{aligned}$$

and further take the k_2 th derivative with respect to t_2 yields

$$\begin{aligned} & \frac{\partial^{k_2}}{\partial t_2^{k_2}} \frac{\partial^{2k_1}}{\partial t_1^{2k_1}} M_{a^\top \epsilon, \epsilon^\top \epsilon}(t_1, t_2) \\ &= \frac{\partial^{k_2}}{\partial t_2^{k_2}} (1-2t_2)^{-\frac{n+2k_1}{2}} (2k_1 - 1)!! (a^\top a)^{k_1} \\ &= \frac{(n+2k_1+2k_2-2)!!}{(n+2k_1-2)!!} (2k_1 - 1)!! (a^\top a)^{k_1}, \end{aligned}$$

where the last equality can be easily shown with induction.

$$\mathbb{E}_{\epsilon \sim \mathcal{N}(0, I)} \mathbb{E}_{\bar{\epsilon} \sim \mathcal{N}(0, I)} (a^\top \epsilon)^{2k_1} (a^\top \bar{\epsilon})^{2k_2} (\epsilon^\top \bar{\epsilon})^{k_3}$$

The moment generating function:

$$\begin{aligned} & \mathbb{E} \exp(t_1 a^\top \epsilon) \exp(t_2 a^\top \bar{\epsilon}) \exp(t_3 \epsilon^\top \bar{\epsilon}) \\ &= \mathbb{E} \exp\left(\sum_{i=1}^n t_1 a_i \epsilon_i + t_2 a_i \bar{\epsilon}_i + t_3 \epsilon_i \bar{\epsilon}_i\right) \\ &= \prod_{i=1}^n \mathbb{E} \exp(t_1 a_i \epsilon_i + t_2 a_i \bar{\epsilon}_i + t_3 \epsilon_i \bar{\epsilon}_i) \\ &= \prod_{i=1}^n \int_{\mathbb{R}^2} \exp(t_1 a_i u + t_2 a_i v + t_3 uv) \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{u^2}{2}\right) \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{v^2}{2}\right) d(u, v) \\ &= \prod_{i=1}^n \int_{\mathbb{R}^2} \frac{1}{2\pi} \int_{\mathbb{R}^2} \exp\left(-\frac{1}{2}u^2 - \frac{1}{2}v^2 + t_1 a_i u + t_2 a_i v + t_3 uv\right) d(u, v); \end{aligned}$$

by doing the replacement

$$\begin{aligned} \begin{bmatrix} u \\ v \end{bmatrix} &= \begin{bmatrix} \sqrt{2}/2 & \sqrt{2}/2 \\ -\sqrt{2}/2 & \sqrt{2}/2 \end{bmatrix} \begin{bmatrix} u' \\ v' \end{bmatrix} \\ &= \prod_{i=1}^n \int_{\mathbb{R}^2} \frac{1}{2\pi} \int_{\mathbb{R}^2} \exp\left(-\frac{1+t_3}{2}u'^2 - \frac{1-t_3}{2}v'^2 + \frac{\sqrt{2}(t_1-t_2)}{2}a_i u' + \frac{\sqrt{2}(t_1+t_2)}{2}a_i v'\right) d(u', v') \\ &= \prod_{i=1}^n \left[\exp\left(\frac{a_i^2(t_1-t_2)^2}{4(1+t_3)}\right) \frac{1}{\sqrt{1+t_3}} \int_{\mathbb{R}} \frac{\sqrt{1+t_3}}{\sqrt{2\pi}} \exp\left(-\frac{1+t_3}{2}\left(u - \frac{\sqrt{2}a_i(t_1-t_2)}{2(t_3+1)}\right)^2\right) \right. \\ & \quad \left. \cdot \exp\left(\frac{a_i^2(t_1+t_2)^2}{4(1-t_3)}\right) \frac{1}{\sqrt{1-t_3}} \int_{\mathbb{R}} \frac{\sqrt{1-t_3}}{\sqrt{2\pi}} \exp\left(-\frac{1-t_3}{2}\left(v + \frac{\sqrt{2}a_i(t_1+t_2)}{2(t_3-1)}\right)^2\right) \right] \\ &= \prod_{i=1}^n \exp\left(\frac{a_i^2(t_1-t_2)^2}{4(1+t_3)}\right) \frac{1}{\sqrt{1+t_3}} \exp\left(\frac{a_i^2(t_1+t_2)^2}{4(1-t_3)}\right) \frac{1}{\sqrt{1-t_3}} \\ &= (1-t_3^2)^{-\frac{n}{2}} \exp\left(\left[\frac{(t_1-t_2)^2}{4(1+t_3)} + \frac{(t_1+t_2)^2}{4(1-t_3)}\right] a^\top a\right) \end{aligned}$$

The next step should be calculating

$$\frac{\partial^{2k_1}}{\partial t_1^{2k_1}} \frac{\partial^{2k_2}}{\partial t_2^{2k_2}} \frac{\partial^{k_3}}{\partial t_3^{k_3}} \mathbb{E} \exp(t_1 a^\top \epsilon) \exp(t_2 a^\top \bar{\epsilon}) \exp(t_3 \epsilon^\top \bar{\epsilon}).$$

However I could not do it.

$$\mathbb{E}_{\epsilon \sim \mathcal{N}(0, I)} \mathbb{E}_{\bar{\epsilon} \sim \mathcal{N}(0, I)} (a^\top \epsilon)^2 (a^\top \bar{\epsilon})^2 (\epsilon \bar{\epsilon})^2 = (n+8)(a^\top a)^2$$

$$\begin{aligned} & \mathbb{E}_{\epsilon \sim \mathcal{N}(0, I)} \mathbb{E}_{\bar{\epsilon} \sim \mathcal{N}(0, I)} (a^\top \epsilon)^2 (a^\top \bar{\epsilon})^2 (\epsilon \bar{\epsilon})^2 \\ &= \sum_{p=1}^n \sum_{q=1}^n \sum_{r=1}^n \sum_{s=1}^n \sum_{t=1}^n \sum_{w=1}^n (a_p u_p)(a_q u_q)(a_r \bar{u}_r)(a_s \bar{u}_s)(u_t \bar{u}_t)(u_w \bar{u}_w) \end{aligned}$$

Now there are the following scenarios:

1. $p = q = r = s = t = w$

$$\begin{aligned} & \mathbb{E}(a_p u_p)(a_q u_q)(a_r \bar{u}_r)(a_s \bar{u}_s)(u_t \bar{u}_t)(u_w \bar{u}_w) \\ &= \mathbb{E}(a_p u_p)^2 (a_p \bar{u}_p)^2 (u_p \bar{u}_p)^2 \\ &= 9a_p^4; \end{aligned}$$

2. $p = q = r = s, t = w$ (and $s \neq t$ is implied here and similarly below)

$$\begin{aligned} & \mathbb{E}(a_p u_p)(a_q u_q)(a_r \bar{u}_r)(a_s \bar{u}_s)(u_t \bar{u}_t)(u_w \bar{u}_w) \\ &= \mathbb{E}(a_p u_p)^2 (a_p \bar{u}_p)^2 (u_t \bar{u}_t)^2 \\ &= a_p^4; \end{aligned}$$

3. $p = q, r = s, t = w$

$$\begin{aligned} & \mathbb{E}(a_p u_p)(a_q u_q)(a_r \bar{u}_r)(a_s \bar{u}_s)(u_t \bar{u}_t)(u_w \bar{u}_w) \\ &= \mathbb{E}(a_p u_p)^2 (a_r \bar{u}_r)^2 (u_t \bar{u}_t)^2 \\ &= a_p^2 a_r^2; \end{aligned}$$

4. $p = q = t = w, r = s$ (or $r = s = t = w, p = q$)

$$\begin{aligned} & \mathbb{E}(a_p u_p)(a_q u_q)(a_r \bar{u}_r)(a_s \bar{u}_s)(u_t \bar{u}_t)(u_w \bar{u}_w) \\ &= \mathbb{E}(a_p u_p)^2 (a_r \bar{u}_r)^2 (u_p \bar{u}_p)^2 \\ &= 3a_p^2 a_r^2; \end{aligned}$$

5. $p = r = t, q = s = w$ (or $p = s = t, q = r = w$ or $q = r = t, p = s = w$ or $q = s = t, p = r = w$)

$$\begin{aligned} & \mathbb{E}(a_p u_p)(a_q u_q)(a_r \bar{u}_r)(a_s \bar{u}_s)(u_t \bar{u}_t)(u_w \bar{u}_w) \\ &= \mathbb{E}(a_p u_p)(a_p \bar{u}_p)(u_p \bar{u}_p)(a_q u_q)(a_q \bar{u}_q)(u_q \bar{u}_q) \\ &= a_p^2 a_q^2. \end{aligned}$$

All remain cases have expectation 0 as they will fall into the situation of (2). Thus the final

expectation is

$$\begin{aligned}
& \mathbb{E}_{\epsilon \sim \mathcal{N}(0, I)} \mathbb{E}_{\bar{\epsilon} \sim \mathcal{N}(0, I)} (a^\top \epsilon)^2 (a^\top \bar{\epsilon})^2 (\epsilon \bar{\epsilon})^2 \\
&= 9 \sum_{p=1}^n a_p^4 + \sum_{p=1}^n \sum_{t \neq p}^n a_p^4 + \sum_{p=1}^n \sum_{r \neq p}^n \sum_{t \neq p, t \neq r}^n a_p^2 a_r^2 + 2 \sum_{p=1}^n \sum_{r \neq p}^n 3a_p^2 a_r^2 + 4 \sum_{p=1}^n \sum_{q \neq p}^n a_p^2 a_q^2 \\
&= (n+8) \sum_{p=1}^n a_p^4 + (n+8) \sum_{p=1}^n \sum_{r \neq p}^n a_p^2 a_r^2 \\
&= (n+8)(a^\top a)^2.
\end{aligned}$$

Similar technique can be used to derive the following expectations:

$$\mathbb{E}_{\epsilon \sim \mathcal{N}(0, I)} \mathbb{E}_{\bar{\epsilon} \sim \mathcal{N}(0, I)} (a^\top \epsilon) (a^\top \bar{\epsilon}) (\epsilon^\top \bar{\epsilon}) = a^\top a$$

$$\mathbb{E}_{\epsilon \sim \mathcal{N}(0, I)} \mathbb{E}_{\bar{\epsilon} \sim \mathcal{N}(0, I)} (a^\top \epsilon)^3 (a^\top \bar{\epsilon}) (\epsilon^\top \bar{\epsilon}) = 3(a^\top a)^2$$

2 Useful values with variables uniformly distributed on unit sphere

Uniform distribution on unit sphere is closely related to multivariate normal distribution because for $\epsilon \sim \mathcal{N}(0, I)$, $\epsilon/\|\epsilon\| \sim U(S(0, 1))$. Here $U(\cdot)$ stands for uniform distribution on a set, and $S(0, 1)$ stands for a sphere centered at the all zero vector and has radius of 1.

One important thing worth noticing is that the angle and the magnitude of an $\epsilon \sim \mathcal{N}(0, I)$ are independent, so we can have

$$\begin{aligned} \mathbf{E}_{\epsilon \sim \mathcal{N}(0, I)} \{(\epsilon^\top \epsilon)^2 \epsilon \epsilon^\top\} &= \mathbf{E}_{\epsilon \sim \mathcal{N}(0, I)} \left\{ \frac{(\epsilon^\top \epsilon)^2 \epsilon \epsilon^\top}{\|\epsilon\|^6} \|\epsilon\|^6 \right\} \\ &= \mathbf{E}_{\epsilon \sim U(S(0, 1))} \{(\epsilon^\top \epsilon)^2 \epsilon \epsilon^\top\} \mathbf{E}_{X \sim \chi_n^2} \{X^3\} \end{aligned}$$

with ϵ and X being independent from each other.

Then we have, for example,

$$\begin{aligned} \mathbf{E}_{\epsilon \sim U(S(0, 1))} \{(\epsilon^\top \epsilon)^2 \epsilon \epsilon^\top\} &= \mathbf{E}_{\epsilon \sim \mathcal{N}(0, I)} \{(\epsilon^\top \epsilon)^2 \epsilon \epsilon^\top\} / \mathbf{E}_{X \sim \chi_n^2} \{X^3\} \\ &= \frac{(n+2)(n+4)}{n(n+2)(n+4)} I \\ &= \frac{1}{n} I. \end{aligned}$$

Similarly,

$$\mathbf{E}_{\epsilon \sim U(S(0, 1))} (a^\top \epsilon)^2 \epsilon \epsilon^\top = \frac{a^\top a I + 2a a^\top}{n(n+2)} \quad (3a)$$

$$\mathbf{E}_{\epsilon \sim U(S(0, 1))} (\epsilon^\top \epsilon) \epsilon \epsilon^\top = \frac{(n+2)I}{n(n+2)} = \frac{1}{n} I \quad (3b)$$

$$\mathbf{E}_{\epsilon \sim U(S(0, 1))} (\epsilon^\top \epsilon)^2 \epsilon \epsilon^\top = \frac{(n+2)(n+4)I}{n(n+2)(n+4)} = \frac{1}{n} I \quad (3c)$$

$$\mathbf{E}_{\epsilon \sim U(S(0, 1))} (\epsilon^\top \epsilon)^3 \epsilon \epsilon^\top = \frac{(n+2)(n+4)(n+6)I}{n(n+2)(n+4)(n+6)} = \frac{1}{n} I, \quad (3d)$$

where a is any vector in \mathbb{R}^n independent from ϵ .