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Supplement to the paper "On some known derivations and new ones for the Wishart distribution: A didactic"

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This article supplements Ogasawara (2023) with the second proof and associated remarks for Lemma 1.

Lemma 1 (Ogasawara, 2023). Suppose that each of 2m variables X_{ik} and X_{jk} $(i \neq j; k = 1,...,m; m = 1,2,...)$ independently follows $N(0,1) \equiv N_1(0,1)$. Then, the distribution of $\sum_{k=1}^{m} X_{ik} X_{jk}$ is the same as that of $X_{il} \sqrt{\sum_{k=1}^{m} X_{jk}^2}$ $(i \neq j; l = 1,...,m)$.

Proof 2. In this proof, the pdf of the chi-distribution is used with associated mgf's. Let $X = \sqrt{\sum_{k=1}^{m} X_{jk}^2}$ and $Y = X_{il}$. Then, X is chi-distributed with m df. The pdf of X at x denoted by $f_{\chi}(x \mid m)$ is given by that of the chi-square distributed $U = X^2$ at u with m df i.e., $f_{\chi^2}(u \mid m) = \frac{u^{(m/2)-1}}{2^{m/2}\Gamma(m/2)} \exp(-u/2)$ with the Jacobian du / dx = 2x, yielding

$$f_{\chi}(x \mid m) = \frac{u^{(m/2)-1}}{2^{m/2} \Gamma(m/2)} \exp(-u/2) \frac{du}{dx}$$
$$= \frac{u^{(m/2)-1} \exp(-u/2)}{2^{m/2} \Gamma(m/2)} 2x$$
$$= \frac{x^{m-1} \exp(-x^2/2)}{2^{(m/2)-1} \Gamma(m/2)}.$$

Then, the joint pdf of X and Y becomes $\frac{x^{m-1}\exp(-x^2/2)}{2^{(m/2)-1}\Gamma(m/2)}\frac{\exp(-y^2/2)}{\sqrt{2\pi}}.$

Consider the variable transformation Z = XY with unchanged X. Since the Jacobian is $J(Y \rightarrow Z) = x^{-1}$, the joint pdf of X and Z is

$$\frac{x^{m-1}\exp(-x^2/2)}{2^{(m/2)-1}\Gamma(m/2)}\frac{\exp(-y^2/2)}{\sqrt{2\pi}}x^{-1} = \frac{x^{m-2}\exp\{-(x^2+z^2x^{-2})/2\}}{2^{(m-1)/2}\pi^{1/2}\Gamma(m/2)},$$

which gives the pdf of Z as $f_Z(z \mid m) = \frac{\int_0^\infty x^{m-2}\exp\{-(x^2+z^2x^{-2})/2\}\,\mathrm{d}x}{2^{(m-1)/2}\pi^{1/2}\Gamma(m/2)}.$ The

mgf of Z is

$$\begin{split} & \frac{\int_{-\infty}^{\infty} \int_{0}^{\infty} x^{m-2} \exp\{-(x^{2}+z^{2}x^{-2})/2\} \exp(tz) \, dx \, dz}{2^{(m-1)/2} \pi^{1/2} \Gamma(m/2)} \\ &= \frac{\int_{0}^{\infty} x^{m-2} \exp\{-(1-t^{2})x^{2}/2\} \int_{-\infty}^{\infty} \exp\{-(z-tx^{2})^{2}x^{-2}/2\} \, dz \, dx}{2^{(m-1)/2} \pi^{1/2} \Gamma(m/2)} \\ &= \frac{\int_{0}^{\infty} x^{m-2} \exp\{-(1-t^{2})x^{2}/2\} x(2\pi)^{1/2} \, dx}{2^{(m-1)/2} \pi^{1/2} \Gamma(m/2)} \\ &= \frac{\int_{0}^{\infty} (1-t^{2})^{m/2} x^{m-1} \exp\{-(1-t^{2})x^{2}/2\} \{2^{(m/2)-1} \Gamma(m/2)\}^{-1} \, dx}{(1-t^{2})^{m/2}} \\ &= \frac{1}{(1-t^{2})^{m/2}}, \end{split}$$

where the integrand of the last integral is the density of the scaled chidistributed variable with the scale parameter $(1-t^2)^{-1/2}$ $(t^2 < 1)$.

Noting that the distribution of each $X_{ik}X_{jk}$ in $\sum_{k=1}^{m} X_{ik}X_{jk}$ is equal to that of $|X_{ik}|X_{jk}$, which is distributed as Z given earlier when m = 1, the mgf of $X_{ik}X_{jk}$ becomes $(1-t^2)^{-/2}$. Since the mgf of $\sum_{k=1}^{m} X_{ik}X_{jk}$ is equal to that of $\sum_{k=1}^{m} |X_{ik}|X_{jk}$ with the *m* terms being i.i.d., the mgf of $\sum_{k=1}^{m} X_{ik}X_{jk}$ is given by $(1-t^2)^{-m/2}$ as obtained earlier for $\sqrt{\sum_{k=1}^{m} X_{jk}^2} Y_{il}$ showing their same distributions. Q.E.D.

Remark S.1 A byproduct of Proof 2 is the pdf of Z using an integral expression. A slightly different derivation of the pdf is given by the variable transformation Z = XY with unchanged Y rather than X. Since the Jacobian is $J(X \rightarrow Z) = |y|^{-1}$, the joint pdf of Y and Z is

$$\frac{x^{m-1} \exp(-x^2/2)}{2^{(m/2)-1} \Gamma(m/2)} \frac{\exp(-y^2/2)}{\sqrt{2\pi}} |y|^{-1}$$

$$= \frac{(z/y)^{m-1} \exp\{-(z/y)^2/2\}}{2^{(m/2)-1} \Gamma(m/2)} \frac{\exp(-y^2/2)}{\sqrt{2\pi}} |y|^{-1}$$

$$= \frac{(z/y)^{m-1} |y|^{-1} \exp[-\{(z/y)^2 + y^2\}/2]}{2^{(m-1)/2} \pi^{1/2} \Gamma(m/2)},$$

where $z / y \ge 0$ by definition. The above result gives another expression of the pdf for Z

$$f_{Z}(z \mid m) = \frac{\int_{-\infty}^{\infty} (z \mid y)^{m-1} \mid y \mid^{-1} \exp[-\{(z \mid y)^{2} + y^{2}\} / 2] dy}{2^{(m-1)/2} \pi^{1/2} \Gamma(m \mid 2)}$$

which is not simpler than that given earlier and can be shown to be equal to the previous one using x = z / y and $J(Y \rightarrow X) = |dy/dx| = |z|x^{-2}$.

Remark S.2 The derivation of the pdf of Z suggests the corresponding pdf when X and Y are correlated. Let $Y = \rho X + (1 - \rho^2)^{1/2} U(\rho^2 \le 1)$, where U is standard normally distributed and uncorrelated with X. Then, the correlation coefficient of X and Y becomes ρ . Consider the transformation from U to $Z = XY = X\{\rho X + (1 - \rho^2)^{1/2}U\}$ with $J(U \rightarrow Z) = |x|^{-1} (1 - \rho^2)^{-1/2}$. Since the joint pdf of X and Z is $(2\pi)^{-1} \exp\{-(x^2 + u^2)/2\}J(U \rightarrow Z)$ $= (2\pi)^{-1} \exp[-\{x^2 + (zx^{-1} - \rho x)^2(1 - \rho^2)^{-1}\}/2]|x|^{-1} (1 - \rho^2)^{-1/2}$ $= (2\pi)^{-1}(1 - \rho^2)^{-1/2} \exp\left\{-\frac{x^2 + z^2x^{-2} - 2z\rho}{2(1 - \rho^2)}\right\}|x|^{-1}$,

we have

$$f_{Z}(z) = (2\pi)^{-1} (1-\rho^{2})^{-1/2} \int_{-\infty}^{\infty} \exp\left\{-\frac{x^{2}+z^{2}x^{-2}-2z\rho}{2(1-\rho^{2})}\right\} |x|^{-1} dx$$
$$= \pi^{-1} (1-\rho^{2})^{-1/2} \int_{0}^{\infty} \exp\left\{-\frac{x^{2}+z^{2}x^{-2}-2z\rho}{2(1-\rho^{2})}\right\} x^{-1} dx,$$

which is seen as a special case of Pearson, Jeffery and Elderton (1929, Equation (iv)), Wishart and Bartlett (1932, Equation (12)), and Craig (1936, pp. 3-4) though these authors use the Bessel function of the second kind with imaginary argument (see McKay,1932; Watson, 1944/1995). It is found that when $\rho = 0$, the pdf becomes equal to that obtained earlier when m = 1. The mgf of Z is

The above result becomes $(1-t^2)^{-1/2}$ when $\rho = 0$ as obtained earlier. An algebraically equal expression $\{1-2\rho t - (1-\rho^2)t^2\}^{-1/2}$ was given by Wishart and Bartlett (1932, Equation (9)), which supports the validity of $f_Z(z)$ given earlier.

Remark S.3 We deal with the sum of the products of correlated variables $\sum_{i=1}^{m} X_i Y_i$, $(m \ge 2)$, where X_i and Y_i are standard normally distributed with $E(X_i Y_i) = \rho$ $(-1 \le \rho \le 1)$ (i = 1, ..., m) and independent of X_j and Y_j $(i \ne j)$. Redefine $X = \sum_{i=1}^{m} X_i^2$, $Y = \sum_{i=1}^{m} Y_i^2$ and $Z = \sum_{i=1}^{m} X_i Y_i$ with the random matrix $\mathbf{S}^* = \begin{pmatrix} X & Z \\ Z & Y \end{pmatrix}$. Since \mathbf{S}^* follows the Wishart distribution with the scale matrix $\mathbf{\Sigma} = \begin{pmatrix} 1 & \rho \\ \rho & 1 \end{pmatrix}$ and *m* df, the density of \mathbf{S}^* at $\mathbf{S} = \begin{pmatrix} x & z \\ Z & y \end{pmatrix}$ with *p* = 2 becomes

$$w_{p}(\mathbf{S} \mid \mathbf{\Sigma}, m) = \frac{\exp\{-\operatorname{tr}(\mathbf{\Sigma}^{-1}\mathbf{S})/2\} \mid \mathbf{S} \mid^{(m-p-1)/2}}{2^{mp/2} \mid \mathbf{\Sigma} \mid^{m/2} \Gamma_{p}(m/2)}$$
$$= \exp\left\{-\frac{x+y-2\rho z}{2(1-\rho^{2})}\right\} \frac{(xy-z^{2})^{(m-3)/2}}{2^{m}(1-\rho^{2})^{m/2} \Gamma_{2}(m/2)}.$$

Consider the variable transformation from Y to $U = XY - Z^2 \ge 0$ with $J(Y \rightarrow U) = x^{-1}$. Then, the joint pdf of X, U and Z is

$$\exp\left\{-\frac{x+y-2\rho z}{2(1-\rho^2)}\right\} \frac{(xy-z^2)^{(m-3)/2}}{2^m(1-\rho^2)^{m/2}\Gamma_2(m/2)} J(Y \to U)$$

=
$$\exp\left\{-\frac{x+(u+z^2)x^{-1}-2\rho z}{2(1-\rho^2)}\right\} \frac{u^{(m-3)/2}x^{-1}}{2^m(1-\rho^2)^{m/2}\Gamma_2(m/2)}$$

=
$$\exp\left\{-\frac{u}{2(1-\rho^2)x}\right\} \frac{u^{(m-3)/2}x^{-1}}{2^m(1-\rho^2)^{m/2}\Gamma_2(m/2)} \exp\left\{-\frac{x+z^2x^{-1}-2\rho z}{2(1-\rho^2)}\right\},$$

which gives the marginal density of X and Z as

$$\int_{0}^{\infty} \exp\left\{-\frac{u}{2(1-\rho^{2})x}\right\} \frac{u^{(m-3)/2}x^{-1}}{2^{m}(1-\rho^{2})^{m/2}\Gamma_{2}(m/2)} du \exp\left\{-\frac{x+z^{2}x^{-1}-2\rho z}{2(1-\rho^{2})}\right\}$$
$$=\frac{\{2(1-\rho^{2})x\}^{(m-1)/2}\Gamma\{(m-1)/2\}x^{-1}}{2^{m}(1-\rho^{2})^{m/2}\Gamma_{2}(m/2)} \exp\left\{-\frac{x+z^{2}x^{-1}-2\rho z}{2(1-\rho^{2})}\right\}$$
$$=\frac{x^{(m-3)/2}}{2^{(m+1)/2}(1-\rho^{2})^{1/2}\pi^{1/2}\Gamma(m/2)} \exp\left\{-\frac{x+z^{2}x^{-1}-2\rho z}{2(1-\rho^{2})}\right\}.$$

From this result, the pdf of Z is derived as

$$f_Z(z \mid m) = \int_0^\infty \frac{x^{(m-3)/2}}{2^{(m+1)/2} (1-\rho^2)^{1/2} \pi^{1/2} \Gamma(m/2)} \exp\left\{-\frac{x+z^2 x^{-1}-2\rho z}{2(1-\rho^2)}\right\} dx.$$

Let $X = V^2$ with $J(X \to V) = 2v$. Then, the above result becomes

$$f_{Z}(z \mid m) = \int_{0}^{\infty} \frac{x^{(m-3)/2}}{2^{(m+1)/2}(1-\rho^{2})^{1/2}\pi^{1/2}\Gamma(m/2)} \exp\left\{-\frac{x+z^{2}x^{-1}-2\rho z}{2(1-\rho^{2})}\right\} 2v dv$$

$$= \int_{0}^{\infty} \frac{v^{m-2}}{2^{(m-1)/2}(1-\rho^{2})^{1/2}\pi^{1/2}\Gamma(m/2)} \exp\left\{-\frac{v^{2}+z^{2}v^{-2}-2\rho z}{2(1-\rho^{2})}\right\} dv$$

$$= \int_{0}^{\infty} \frac{x^{m-2}}{2^{(m-1)/2}(1-\rho^{2})^{1/2}\pi^{1/2}\Gamma(m/2)} \exp\left\{-\frac{x^{2}+z^{2}x^{-2}-2\rho z}{2(1-\rho^{2})}\right\} dx,$$

where the last result is given by the redefinition of X = V. Note that in the above density $m \ge 2$ is assumed. Though $\Gamma_2(m/2)$ when m = 1 is not defined, it is found that the derived density when m = 1 becomes

$$f_Z(z \mid m=1) = \frac{1}{\pi (1-\rho^2)^{1/2}} \int_0^\infty \exp\left\{-\frac{x^2 + z^2 x^{-2} - 2\rho z}{2(1-\rho^2)}\right\} x^{-1} dx$$

as obtained earlier for the product of the correlated standard normal variables.

Remark S.4 The mgf of the sum of the products of correlated variables $Z = \sum_{i=1}^{m} X_i Y_i \ (m \ge 2)$ as defined in Remark S.3 is obtained. Using the pdf of *Z*, we have

$$\begin{split} & \mathsf{M}_{Z}(t \mid m) \\ &= \int_{-\infty}^{\infty} \int_{0}^{\infty} \frac{x^{m-2}}{2^{(m-1)/2} (1-\rho^{2})^{1/2} \pi^{1/2} \Gamma(m/2)} \exp\left\{-\frac{x^{2}+z^{2}x^{-2}-2\rho z}{2(1-\rho^{2})}\right\} \exp(tz) \, \mathrm{d}x \, \mathrm{d}z \\ &= \int_{0}^{\infty} \frac{x^{m-2}}{2^{(m-1)/2} (1-\rho^{2})^{1/2} \pi^{1/2} \Gamma(m/2)} \int_{-\infty}^{\infty} \exp\left\{-\frac{x^{2}+z^{2}x^{-2}-2\rho z-2(1-\rho^{2})tz}{2(1-\rho^{2})}\right\} \, \mathrm{d}z \, \mathrm{d}x \\ &= \int_{0}^{\infty} \frac{x^{m-2}}{2^{(m-1)/2} (1-\rho^{2})^{1/2} \pi^{1/2} \Gamma(m/2)} \\ &\times \int_{-\infty}^{\infty} \exp\left(-\frac{[z-\{\rho+(1-\rho^{2})t\}x^{2}]^{2}+x^{4}-\{\rho+(1-\rho^{2})t\}^{2}x^{4}}{2(1-\rho^{2})x^{2}}\right) \, \mathrm{d}z \, \mathrm{d}x \\ &= \int_{0}^{\infty} \frac{x^{m-2}}{2^{(m-1)/2} (1-\rho^{2})^{1/2} \pi^{1/2} \Gamma(m/2)} (2\pi)^{1/2} (1-\rho^{2})^{1/2} x \\ &\times \exp\left(-\frac{[1-\{\rho+(1-\rho^{2})t\}^{2}]x^{4}}{2(1-\rho^{2})x^{2}}\right) \, \mathrm{d}x \\ &= \int_{0}^{\infty} \frac{x^{m-1}}{2^{(m-2)/2} \Gamma(m/2)} \exp\left(-\frac{[1-\{\rho+(1-\rho^{2})t\}^{2}]y^{2}}{2(1-\rho^{2})}\right) \frac{y^{-1/2}}{2} \, \mathrm{d}y \\ &= \int_{0}^{\infty} \frac{y^{(m-1)/2}}{2^{(m-2)/2} \Gamma(m/2)} \exp\left(-\frac{[1-\{\rho+(1-\rho^{2})t\}^{2}]y}{2(1-\rho^{2})}\right) \, \mathrm{d}y \\ &= \int_{0}^{\infty} \frac{y^{(m-2)/2}}{2^{m/2} \Gamma(m/2)} \exp\left(-\frac{[1-\{\rho+(1-\rho^{2})t\}^{2}]y}{2(1-\rho^{2})}\right) \, \mathrm{d}y \\ &= \int_{0}^{\infty} \frac{y^{(m-2)/2}}{2^{m/2} \Gamma(m/2)} \exp\left(-\frac{[1-\{\rho+(1-\rho^{2})t\}^{2}]y}{2(1-\rho^{2})}\right) \, \mathrm{d}y \\ &= \int_{0}^{\infty} \frac{y^{(m-2)/2}}{2^{m/2} \Gamma(m/2)} \exp\left(-\frac{[1-\{\rho+(1-\rho^{2})t\}^{2}]y}{2(1-\rho^{2})}\right) \, \mathrm{d}y \end{aligned}$$

which is expected since $Z = \sum_{i=1}^{m} X_i Y_i$ is the sum of *m* independent identically distributed terms, where the mgf of each term was obtained as

$$\frac{(1-\rho^2)^{1/2}}{\left[1-\{\rho+(1-\rho^2)t\}^2\right]^{1/2}}.$$

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