

THE MULTIVARIATE NORMAL AND CHI-SQUARE DISTRIBUTIONS

Let Z_1, Z_2, \dots, Z_n be independent $N(0, 1)$ random variables. When treated as the coordinates of a point in \mathbb{R}^n they define a random vector \mathbf{Z} , whose (joint) density function is

$$f(\mathbf{z}) = (2\pi)^{-n/2} \exp\left(-\frac{1}{2} \sum_i z_i^2\right) = (2\pi)^{-n/2} \exp\left(-\frac{1}{2} \|\mathbf{z}\|^2\right).$$

Such a random vector is said to have a *spherical normal distribution*.

The *chi-square*, χ_n^2 , is defined as the distribution of the sum of squares $Z_1^2 + \dots + Z_n^2$ of independent $N(0, 1)$ random variables. The *noncentral chi-square*, $\chi_n^2(\gamma)$, with noncentrality parameter $\gamma \geq 0$ is defined as the distribution of the sum of squares $(Z_1 + \gamma)^2 + Z_2^2 + \dots + Z_n^2$.

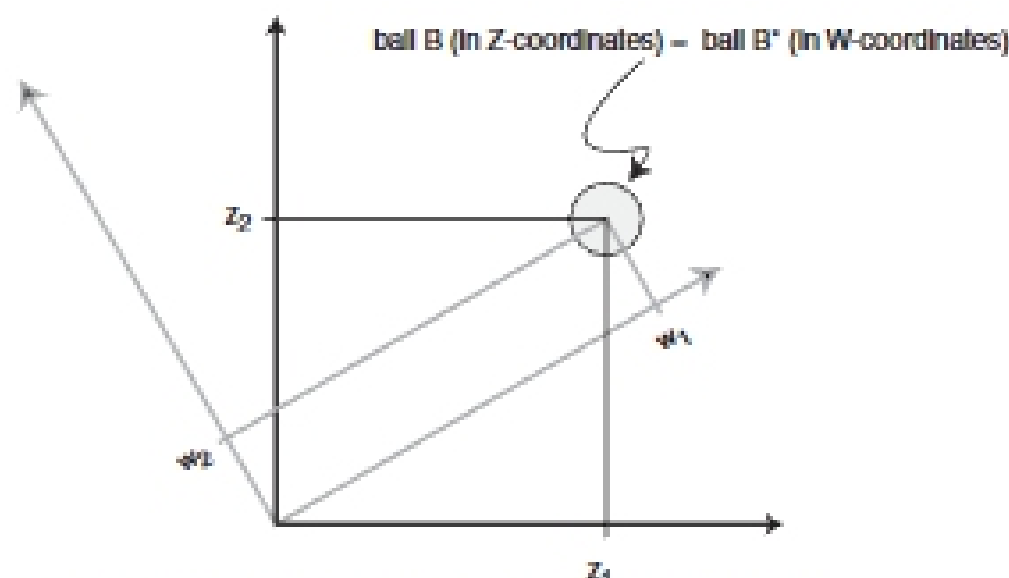
The spherical symmetry of the density $f(\cdot)$ is responsible for an important property of multivariate normals. Let $\mathbf{q}_1, \dots, \mathbf{q}_n$ be a new orthonormal basis for \mathbb{R}^n , and let

$$\mathbf{Z} = W_1 \mathbf{q}_1 + \dots + W_n \mathbf{q}_n$$

be the representation for \mathbf{Z} in the new basis.

<1> **Theorem.** *The W_1, \dots, W_n are also independent $N(0, 1)$ distributed random variables.*

If you know about multivariate characteristic functions this is easy to establish using the matrix representation $\mathbf{Z} = \mathbf{Q}\mathbf{W}$, where \mathbf{Q} is the orthogonal matrix with columns $\mathbf{q}_1, \dots, \mathbf{q}_n$.



A more intuitive explanation is based on the approximation

$$\mathbf{P}\{\mathbf{Z} \in B\} \approx f(\mathbf{z})(\text{volume of } B)$$

for a small ball B centered at \mathbf{z} . The transformation from \mathbf{Z} to \mathbf{W} corresponds to a rotation, so

$$\mathbf{P}\{\mathbf{Z} \in B\} = \mathbf{P}\{\mathbf{W} \in B^*\},$$

where B^* is a ball of the same radius, but centered at the point $\mathbf{w} = (w_1, \dots, w_n)$ for which $w_1 \mathbf{q}_1 + \dots + w_n \mathbf{q}_n = \mathbf{z}$. The last equality implies $\|\mathbf{w}\| = \|\mathbf{z}\|$, from which we get

$$\mathbf{P}\{\mathbf{W} \in B^*\} \approx (2\pi)^{-n/2} \exp\left(-\frac{1}{2} \|\mathbf{w}\|^2\right)(\text{volume of } B^*).$$

That is, \mathbf{W} has the asserted spherical normal density.

To prove results about the spherical normal it is often merely a matter of transforming to an appropriate orthonormal basis.

<2> **Theorem.** Let \mathcal{X} be an m -dimensional subspace of \mathbb{R}^n . Let \mathbf{Z} be a vector of independent $N(0, 1)$ random variables, and $\boldsymbol{\mu}$ be a vector of constants. Then

- (i) the projection $\hat{\mathbf{Z}}$ of \mathbf{Z} onto \mathcal{X} is independent of the projection $\mathbf{Z} - \hat{\mathbf{Z}}$ of \mathbf{Z} onto \mathcal{X}^\perp , the orthogonal complement of \mathcal{X} .
- (ii) $\|\hat{\mathbf{Z}}\|^2$ has a χ_m^2 distribution.
- (iii) $\|\mathbf{Z} + \boldsymbol{\mu}\|^2$ has a noncentral $\chi_n^2(\gamma)$ distribution, with $\gamma = \|\boldsymbol{\mu}\|$.
- (iv) $\|\hat{\mathbf{Z}} + \boldsymbol{\mu}\|^2$ has a noncentral $\chi_m^2(\gamma)$ distribution, with $\gamma = \|\boldsymbol{\mu}\|$.

Proof. Let $\mathbf{q}_1, \dots, \mathbf{q}_n$ be an orthonormal basis of \mathbb{R}^n such that $\mathbf{q}_1, \dots, \mathbf{q}_m$ span the space \mathcal{X} and $\mathbf{q}_{m+1}, \dots, \mathbf{q}_n$ span \mathcal{X}^\perp . If $\mathbf{Z} = W_1\mathbf{q}_1 + \dots + W_n\mathbf{q}_n$ then

$$\begin{aligned}\hat{\mathbf{Z}} &= W_1\mathbf{q}_1 + \dots + W_m\mathbf{q}_m, \\ \mathbf{Z} - \hat{\mathbf{Z}} &= W_{m+1}\mathbf{q}_{m+1} + \dots + W_n\mathbf{q}_n, \\ \|\mathbf{Z}\|^2 &= W_1^2 + \dots + W_m^2,\end{aligned}$$

from which the first two asserted properties follow.

For the third and fourth assertions, choose the basis so that $\boldsymbol{\mu} = \gamma\mathbf{q}_1$. Then

$$\begin{aligned}\mathbf{Z} + \boldsymbol{\mu} &= (W_1 + \gamma)\mathbf{q}_1 + W_2\mathbf{q}_2 + \dots + W_n\mathbf{q}_n \\ \hat{\mathbf{Z}} + \boldsymbol{\mu} &= (W_1 + \gamma)\mathbf{q}_1 + W_2\mathbf{q}_2 + \dots + W_m\mathbf{q}_m\end{aligned}$$

□ from which we get the noncentral chi-squares.

Fact about the multivariate normal

If Z is an $n \times 1$ vector of independent $N(0, 1)$ random variables, if $\boldsymbol{\mu}$ is an $m \times 1$ vector of constants, and if A is an $m \times n$ matrix of constants, then the random vector $X = \boldsymbol{\mu} + AZ$ has expected value $\boldsymbol{\mu}$ and variance matrix $V = AA'$, and moment generating function

$$\mathbb{E} \exp(t'X) = \exp(t'\boldsymbol{\mu} + t'AA't/2)$$

In particular, the distribution of X depends only on $\boldsymbol{\mu}$ and V . The random vector X has a $N(\boldsymbol{\mu}, V)$ distribution. If $\boldsymbol{\gamma}$ is a $k \times 1$ vector of constants and B is a $k \times m$ matrix of constants then

$$\boldsymbol{\gamma} + BX = (\boldsymbol{\gamma} + B\boldsymbol{\mu}) + BAZ$$

has a $N(\boldsymbol{\gamma} + B\boldsymbol{\mu}, BV B')$ distribution.

Standard distributions

Suppose

$$\begin{aligned}Z &\text{ has a } N(0, 1) \text{ distribution} \\ S_k^2 &\text{ has a } \chi_k^2 \text{ distribution} \\ S_\ell^2 &\text{ has a } \chi_\ell^2 \text{ distribution}\end{aligned}$$

with all random variables independent of each other. Then, by definition,

$$\frac{Z}{\sqrt{S_k^2/k}} \text{ has a } t\text{-distribution on } k \text{ degrees of freedom } (t_k)$$

and

$$\frac{S_\ell^2/\ell}{S_k^2/k} \text{ has an } F\text{-distribution on } \ell \text{ and } k \text{ degrees of freedom } (F_{\ell,k})$$