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Distribution of the product of two normal variables. A state of the Art

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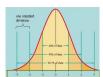


Outline

- INTRODUCTION
- FIRST APPROACHES
- ROHATGI'S THEOREM
- COMPUTATIONAL TECHNIQUES
- **5** RECENT ADVANCES

Introduction

- Normal distribution: the most common in Theory of Probability.
- Applications to the real world: biology, psychology, physics, economics,... .
- Density function (PDF): $f(x) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp{-\frac{(x-\mu)^2}{2\sigma^2}}$, where μ is the mean and σ is the standard deviation (σ^2 is the variance).
- Distribution function (CDF): $F(x) = \frac{1}{2} \left[1 + \operatorname{erf}\left(\frac{x-\mu}{\sigma\sqrt{2}}\right) \right]$, where the error function is: $\operatorname{erf}(t) = \frac{2}{\sqrt{\pi}} \int_0^t e^{-y^2} \mathrm{d}y$.



Normal Distribution N(0,1)



Abraham de Moivre (1667-1754)



Carl F. Gauss (1777-1855)

Introduction

- Several distributions are derived from normal distribution: Chi-square or t distribution are the most famous.
- Relation with other distributions (exponential, uniform, ...) is known.
- Let X and Y be two normally distributed variables with means μ_X and μ_Y and variances σ_X^2, σ_Y^2 ,
- Sum X+Y is normally distributed with mean $\mu_x + \mu_y$ and variance $\sigma_x^2 + \sigma_y^2$, when there is no correlation.
- When there exists correlation (ρ), variance of the sum is $\sigma_x^2 + \sigma_y^2 + 2\rho\sigma_x\sigma_y$.
- The product of two variables was not be able to characterize like the sum and remains like an open problem.



First Historical Approach

- Wishart and Bartlett (1932): The product of two independent normal variables is directly
 proportional to a second class Bessel function with a zero-order pure imaginary argument
 [WB32]
- Craig (1936): Let be two normal variables $X \sim N(\mu_X, \sigma_X)$ and $Y \sim N(\mu_Y, \sigma_Y)$, and correlation coefficient ρ_{xy} and the inverse of the variation coefficient: $r_X = \frac{\mu_X}{\sigma_X}$ and $r_Y = \frac{\mu_Y}{\sigma_Y}$. Then we could deduce the moment-generating function.[Cra36]

$$M_{xy}(t) = \frac{\exp\left[\frac{(r_x^2 + r_y^2 - 2\rho_{xy}r_xr_y)t^2 + 2r_xr_yt}{2(1 - (1 + \rho_{xy})t)(1 - (1 - \rho_{xy})t))}\right]}{((1 - (1 + \rho_{xy})t)(1 - (1 - \rho_{xy})t))^{1/2}}$$
(1)

- The product of two normal variables might be a non-normal distribution
- Skewness is $(-2\sqrt{2}, +2\sqrt{2})$, maximum kurtosis value is 12
- The function of density of the product is proportional to a Bessel function and its graph is asymptotical at zero.

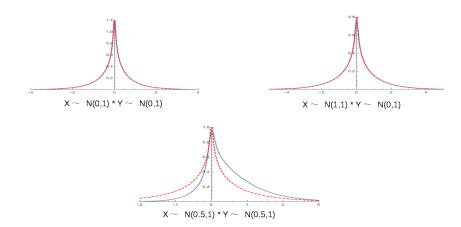


Figure: Examples of the product of two Normal Variables with $\rho=0$ Craig (red -dashed) and MonteCarlo Simulation (blue)



Advances in 50's in 20th Century

- Aroian (1947): Type III Pearson function or Gram-Charlier Type A series ([Aro47]) .
- Limitations: $\rho = 0$, the Type III Pearson requires $\mu_X \neq 0$ or $\mu_Y \neq 0$, Gram-Charlier approach has a very limited range of applicability.
- Advantages: There is no discontinuity at zero.

Theorem ([ATC78], p. 167)

Let X and Y be two normally distributed variables with mean μ_x, μ_y , variances σ_x^2, σ_y^2 and correlation coefficient ρ . Let be $r_x = \frac{\mu_x}{\sigma_x}$ and $r_y = \frac{\mu_y}{\sigma_y}$. Distribution function of $Z = \frac{xy}{\sigma_x\sigma_y}$ is

$$F_Z(z) = \frac{1}{2} + \frac{1}{\pi} \int_0^\infty \phi(z, r_x, r_y, \rho, t) dt,$$
 (2)

$$\begin{split} & \text{where } \phi(z, r_x, r_y, \rho, t) = \\ & \frac{1}{t} \frac{1}{G} \exp\left(-\frac{(H + 4\rho r_x r_y)t^2 + (1 - \rho^2)Ht}{2G^2}\right) * \left\{ \left[\left(\frac{G + I}{2}\right)^{1/2} \sin A \right] - \left[\left(\frac{G - I}{2}\right)^{1/2} \cos A \right] \right\}, \text{ with } \\ & A = \left(t \left(y - \frac{r_1 r_2 I - \rho Ht^2}{G^2}\right)\right), G^2 = (1 + (1 - \rho^2)t^2)^2 + 4\rho^2 t^2, \ H = r_1^2 + r_2^2 - 2\rho r_1 r_2 \text{ and } \\ & I = 1 + (1 - \rho^2)t^2. \end{split}$$

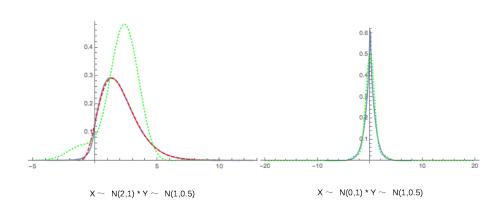


Figure: Examples of the Product of two normal variables no correlated: Gram-Charlier (green - pointed), Pearson Type III (red - dashed) y MonteCarlo simulation (blue)

Rohatgi's Theorem [Roh76]

Theorem ([GLD04], p.452-453)

Let X be a continuous random variable with PDF f(x) definite and positive in (a,b), with $0 < a < b < \infty$. Let Y be a random variable with PDF g(y), definite and positive in (c,d), with $0 < c < d < \infty$. Then, PDF of Z = XY is

■ When ad < bc:</p>

$$h(z) = \left\{ \begin{array}{l} \int_{a}^{z/c} g\left(\frac{z}{x}\right) f(x) \frac{1}{x} dx & ac < z < ad \\ \int_{z/d}^{z/c} g\left(\frac{z}{x}\right) f(x) \frac{1}{x} dx & ad < z < bc \\ \int_{z/d}^{b} g\left(\frac{z}{x}\right) f(x) \frac{1}{x} dx & bc < z < bd \end{array} \right.$$

● When ad = bc

$$h(z) = \begin{cases} \int_a^{z/c} g\left(\frac{z}{x}\right) f(x) \frac{1}{x} dx & ac < z < ad \\ \int_{z/d}^b g\left(\frac{z}{x}\right) f(x) \frac{1}{x} dx & ad < z < bd \end{cases}$$

When ad > bc

$$h(z) = \left\{ \begin{array}{ll} \int_a^{z/c} g\left(\frac{z}{x}\right) f(x) \frac{1}{x} dx & ac < z < ad \\ \int_a^b g\left(\frac{z}{x}\right) f(x) \frac{1}{x} dx & bc < z < ad \\ \int_{z/d}^b g\left(\frac{z}{x}\right) f(x) \frac{1}{x} dx & ad < z < bd \end{array} \right.$$

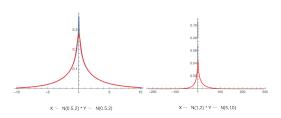
Application Rohatgi's Theorem

- Only for PDF of random variables in first quadrant, but generalization to other quadrants is straightforward.
- The PDF of the product is not defined at zero.
- Range for normal distribution must be bounded.
- Very good approach for the product of two independent N(0,1) distributions:

$$h(z) = \begin{cases} \frac{\kappa_0(-z)}{\pi} & -\infty < z < 0\\ \frac{\kappa_0(z)}{\pi} & 0 < z < \infty \end{cases}$$

where $K_0(\cdot)$ is the modified second class Bessel function.





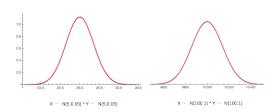


Figure: Examples of Product of two independent normal distributions: Rohatgi's Theorem Approach (blue) and MonteCarlo Simulation (red)

Advances in early 21st Century

Ware and Lad (2003): A bivariate independent normal distribution [WL03]:

$$\left[\begin{array}{c} X \\ Y \end{array}\right] \sim N \left(\left[\begin{array}{c} \mu_{x} \\ \mu_{y} \end{array}\right], \left[\begin{array}{cc} \sigma_{x}^{2} & 0 \\ 0 & \sigma_{y}^{2} \end{array}\right] \right]$$

Marginal density f(z) would be:

$$f(z) = \int_{-\infty}^{\infty} f(z|y)f(y)dy = \int_{-\infty}^{\infty} f(z,y)dy$$
 (3)

- Approach using numerical integration: Newton-Cotes
- Simulation with MonteCarlo method
- Analytical approach using normal distribution: Moment-generating Function:

$$\mu_{z} = \mu_{x}\mu_{y} + \rho\sigma_{x}\sigma_{y} \tag{4}$$

$$\sigma_z^2 = \mu_x^2 \sigma_y^2 + \mu_y^2 \sigma_x^2 + \sigma_x^2 \sigma_y^2 + 2\rho \mu_x \mu_y \sigma_x \sigma_y + \rho^2 \sigma_x^2 \sigma_y^2$$
 (5)

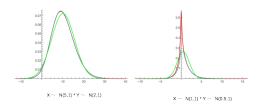
For the case of two independent normally distributed variables, the limit distribution of the product is normal. These approach follows the evolution of ratio (mean/standard deviation), but some important questions remain open [WL03]:

- When the ratio mean/standard deviation is enough to guarantee the normal approach for the product.
- Approximation to normality is more sensitive for individual ratios or combined ratio.
- How is the evolution of the skewness of the product, when is null? Is there a level for skewness and normality of product

Approach to the Product of Two Normal Variables

Let X and Y be two variables normales with parameter: μ_x, σ_x^2 and $r_x = \frac{\mu_x}{\sigma_x}$ and μ_y, σ_y^2 and $r_y = \frac{\mu_y}{\sigma_y}$. Then [SMO12] :

- When two variables have unit variance ($\sigma^2 = 1$), with different mean, normal approach is a good option for means greater than 1. But, when the mean is lower, normal approach is not correct.
- When two variables have unit mean $(\mu = 1)$, with different variance, normal approach requires that, at least, one variable has a variance lower than 1.
- When, at least, one of the inverse of the variation coefficient δ_x or δ_y is high, then normal approach is correct.
- When two normal distributions have same variance $\sigma_x^2 = \sigma_y^2 = \sigma^2$, we define combined ratio as $\frac{\mu_x \mu_y}{\sigma}$, then a high value for combined ratio produce a good normal approach for product, but when combined ratio is lower than 1, the normal approach fails [OOSM13].



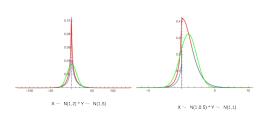


Figure: Examples of Product of two independent normal distributions: Numerical Integration (blue), MonteCarlo Simulation (red), Normal Approach (green)

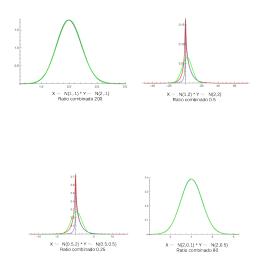


Figure: Examples of Product of two independent normal distributions: Numerical Integration (blue), MonteCarlo Simulation (red), Normal Approach (green)

Recent Publications

Theorem ([NP16] p. 202)

Let (X, Y) be a bivariate normal distribution random vector with mean zero and variance one and correlation coefficient ρ . Then, PDF of Z = XY is

$$f_Z(z) = \frac{1}{\pi\sqrt{1-\rho^2}} \exp\left[\frac{\rho z}{1-\rho^2}\right] K_0\left(\frac{|z|}{1-\rho^2}\right)$$
 (6)

for $-\infty < z < \infty$, where $K_0(\cdot)$ is second class zero order modified Bessel function.

Recent Publications

Theorem ([Cui+16], pp.1662-1663)

Let X and Y two real Gaussian random variables $X \sim N(\mu_x, \sigma_x)$ and $Y \sim N(\mu_y, \sigma_y)$ with ρ the correlation coefficient. Then the exact PDF $f_Z(z)$ of the product Z = XY is given by:

$$\exp\left\{-\frac{1}{2(1-\rho^{2}}\left(\frac{\mu_{x}^{2}}{\sigma_{x}^{2}} + \frac{\mu_{y}^{2}}{\sigma_{y}^{2}} - \frac{2\rho(x + \mu_{x}\mu_{y})}{\sigma_{x}\sigma_{y}}\right)\right\} \times \sum_{n=0}^{\infty} \sum_{m=0}^{2n} \frac{x^{2n-m}|x|^{m-n}\sigma_{x}^{m-n-1}}{\pi(2n)!(1-\rho^{2})^{2n+1/2}\sigma_{y}^{m-n+1}}\left(\frac{\mu_{x}}{\sigma_{x}^{2}} - \frac{\rho\mu_{y}}{\sigma_{x}\sigma_{y}}\right)^{m} \\ \binom{2n}{m} \times \left(\frac{\mu_{y}}{\sigma_{y}^{2}} - \frac{\rho\mu_{x}}{\sigma_{x}\sigma_{y}}\right)^{2n-m} K_{m-n}\left(\frac{|x|}{(1-\rho^{2})\sigma_{x}\sigma_{y}}\right)$$

where $K_{\nu}(\cdot)$ denotes the modified Bessel function of the second kind and order ν .

Final Summary

- First Approaches: Bessel Function Product of two independent standard normal distributions.
- Moment-generating function of the product
- New Options: Pearson Function Type III Gram-Charlier Series Type
 A.
- Rohatgi's Theorem.
- Alternatives approaches:
 - Approach using functions: Bessel, Pearson, Gram-Charlier Series, ...
 - Approach to normal distribution: mean and variance of the product, skewness and kurtosis.
 - Approach using numerical integration methods.
- Future: Alternative distributions: Skew-Normal, Extended Skew-Normal, ...



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Thank you for your attention

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