

During my doctoral research, I had used the software Mathematica by (Stephen) Wolfram. My visits to the Wolfram Mathworld webpage gradually became rare events. I visited it recently and landed at their pages of [probability](#) & [statistics](#). And I couldn't help exclaim, 'how beautifully demonstrated!'

Maybe I was too distracted back then to have not noticed, or I was too much into differential equations [PDEs](#) 😊

✅ <https://mathworld.wolfram.com/topics/ProbabilityandStatistics.html>

➡ The French Mathematician, A. de Moivre developed 'normal distribution' of a variate (with mean & variance) as an approximation to 'binomial distribution'.

<https://mathworld.wolfram.com/BinomialDistribution.html>

The normal distribution becomes a 'standard normal distribution' of a variate with zero mean & unity variance.

<https://mathworld.wolfram.com/NormalDistribution.html>

➡ The ratio of two standard normal variates gets us a Cauchy variate.

<https://mathworld.wolfram.com/CauchyDistribution.html>

➡ The square of a standard normal variate gets us a Chi-squared variate.

<https://mathworld.wolfram.com/Chi-SquaredDistribution.html>

➡ The addition/sum of Chi-squared variates gets us a gamma variate (another Chi-squared variate) and the ratio of two independent gamma variates gets us a beta variate.

<https://mathworld.wolfram.com/GammaDistribution.html>

➡ The beta distribution, for both shape parameters equal to unity that is, beta (1, 1) becomes a standard uniform distribution.

<https://mathworld.wolfram.com/UniformDistribution.html>

➡ One can use Box-Muller transformation on two independent standard uniform variates to arrive at two standard normal variates.

<https://mathworld.wolfram.com/Box-MullerTransformation.html>

I dug up the web deeper and found this illustration of the relationships between distributions:

<https://www.math.wm.edu/~leemis/chart/UDR/UDR.html>