Moments of the univariate truncated normal distribution

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December 5, 2017

The PDF of the truncated normal distribution is¹:

$$f\left(x;\mu,\sigma,a,b\right) = \begin{cases} \frac{1}{\sigma} \frac{\phi\left(\frac{x-\mu}{\sigma}\right)}{\Phi\left(\frac{b-\mu}{\sigma}\right) - \Phi\left(\frac{a-\mu}{\sigma}\right)} & a \leq x \leq b \\ 0 & \text{otherwise} \end{cases}$$

where μ, σ are the mean and standard deviation of the untruncated normal, and

$$\phi(x) = \frac{1}{\sqrt{2\pi}} e^{-x^2/2}, \quad \Phi(x) = \frac{1}{2} \left(1 + \text{erf}\left(x/\sqrt{2}\right) \right)$$
$$(x - a)^2 = x^2 - 2ax + a^2$$

The mean and variance of this distribution are:

$$\langle x \rangle = \mu + \frac{\phi(\alpha) - \phi(\beta)}{\Phi(\beta) - \Phi(\alpha)} \sigma$$

$$\operatorname{var} x = \left[1 + \frac{\alpha \phi(\alpha) - \beta \phi(\beta)}{\Phi(\beta) - \Phi(\alpha)} - \left(\frac{\phi(\alpha) - \phi(\beta)}{\Phi(\beta) - \Phi(\alpha)} \right)^{2} \right] \sigma^{2}$$

where

$$\alpha = \frac{a - \mu}{\sigma}, \quad \beta = \frac{b - \mu}{\sigma}$$

The numerical evaluation of $\frac{\phi(\alpha)-\phi(\beta)}{\Phi(\beta)-\Phi(\alpha)}$ and $\frac{\alpha\phi(\alpha)-\beta\phi(\beta)}{\Phi(\beta)-\Phi(\alpha)}$ can be problematic. We describe our approach below.

Numerical evaluation

Let us define the functions:

$$F_{1}\left(x,y\right) =rac{{{\mathrm{e}}^{-{{x}^{2}}}-{{\mathrm{e}}^{-{{y}^{2}}}}}}{{\mathrm{erf}\left(y\right) -\mathrm{erf}\left(x\right) }},\quad F_{2}\left(x,y\right) =rac{x{{\mathrm{e}}^{-{{x}^{2}}}-y{{\mathrm{e}}^{-{{y}^{2}}}}}}{{\mathrm{erf}\left(y\right) -\mathrm{erf}\left(x\right) }}$$

¹See https://en.wikipedia.org/wiki/Truncated normal distribution.

The problem arises when x, y have the same sign and both are large in magnitude. Then $\operatorname{erf}(y) - \operatorname{erf}(x)$ is the difference of two numbers close to ± 1 , but typically their difference itself is very small. Subtraction of two numbers that are both large relative to their difference is a classic source of floating-point errors (known as *catastrophic cancellation*).

Recall a few useful properties of the error and related functions:

$$\operatorname{erf}(-x) = \operatorname{erf}(x), \quad \operatorname{erfc}(x) = 1 - \operatorname{erf}(x) = e^{-x^{2}} \operatorname{erfcx}(x)$$

The functions erf (x), erfc (x), erfc (x) are all available in numerical packages. Using these, we can make sure that the error function difference is never catastrophic. Let $\epsilon = e^{x^2-y^2}$. We propose:

$$\begin{split} F_1\left(x,y\right) & F_2\left(x,y\right) \\ &= F_1\left(y,x\right), & = F_2\left(y,x\right), & \text{if } |x| > |y| \\ &= P_1\left(x,y-x\right), & = P_2\left(x,y-x\right), & \text{if } |x-y| < 10^{-7} \\ &= \frac{1-\epsilon}{\epsilon \operatorname{erfcx}\left(-y\right) - \operatorname{erfcx}\left(-x\right)} & = \frac{x-y\epsilon}{\epsilon \operatorname{erfcx}\left(-y\right) - \operatorname{erfcx}\left(-x\right)} & \text{if } x,y \le 0 \\ &= \frac{1-\epsilon}{\operatorname{erfcx}\left(x\right) - \epsilon \operatorname{erfcx}\left(y\right)} & = \frac{x-y\epsilon}{\operatorname{erfcx}\left(x\right) - \epsilon \operatorname{erfcx}\left(y\right)} & \text{if } x,y \ge 0 \\ &= \frac{(1-\epsilon)\operatorname{e}^{-x^2}}{\operatorname{erf}\left(y\right) - \operatorname{erf}\left(x\right)} & = \frac{(x-y\epsilon)\operatorname{e}^{-x^2}}{\operatorname{erf}\left(y\right) - \operatorname{erf}\left(x\right)} & \text{otherwise} \end{split}$$

When $x \approx y$, we employ a Taylor expansion of $F_i(x, x + \epsilon)$ in powers of ϵ ,

$$P_{1}(x,\epsilon) = \sqrt{\pi}x + \frac{1}{2}\sqrt{\pi}\epsilon - \frac{1}{6}\sqrt{\pi}x\epsilon^{2} - \frac{1}{12}\sqrt{\pi}\epsilon^{3} + \frac{1}{90}\sqrt{\pi}x(x^{2} + 1)\epsilon^{4}$$

$$P_{2}(x,\epsilon) = \frac{1}{2}\sqrt{\pi}(2x^{2} - 1) + \sqrt{\pi}x\epsilon - \frac{1}{3}\sqrt{\pi}(x^{2} - 1)\epsilon^{2} - \frac{1}{3}\sqrt{\pi}x\epsilon^{3} + \frac{1}{90}\sqrt{\pi}(2x^{4} + 3x^{2} - 8)\epsilon^{4}$$

Mean and variance

Finally observe that,

$$\frac{\phi(\alpha) - \phi(\beta)}{\Phi(\beta) - \Phi(\alpha)} = \sqrt{\frac{2}{\pi}} F_1\left(\frac{\alpha}{\sqrt{2}}, \frac{\beta}{\sqrt{2}}\right), \quad \frac{\alpha\phi(\alpha) - \beta\phi(\beta)}{\Phi(\beta) - \Phi(\alpha)} = \frac{2}{\sqrt{\pi}} F_2\left(\frac{\alpha}{\sqrt{2}}, \frac{\beta}{\sqrt{2}}\right)$$
$$\frac{\alpha\phi(\alpha) - \beta\phi(\beta)}{\Phi(\beta) - \Phi(\alpha)} - \left(\frac{\phi(\alpha) - \phi(\beta)}{\Phi(\beta) - \Phi(\alpha)}\right)^2 = \frac{2}{\sqrt{\pi}} F_2\left(\frac{\alpha}{\sqrt{2}}, \frac{\beta}{\sqrt{2}}\right) - \frac{2}{\pi} \left[F_1\left(\frac{\alpha}{\sqrt{2}}, \frac{\beta}{\sqrt{2}}\right)\right]^2$$