

Integrals involving erf

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Abstract

I compute some integrals involving erf and apply them to some problems related to the maximum of Gaussian RVs.

1 Introduction

Let

$$f_{\mu, \sigma^2}(x) = \frac{1}{\sqrt{2\pi\sigma}} e^{-(x-\mu)^2/2\sigma^2}$$

be the $N(\mu, \sigma^2)$ Gaussian density, and

$$F_{\mu, \sigma^2}(x) = \frac{1}{2} \left[1 + \operatorname{erf} \left(\frac{x-\mu}{\sqrt{2}\sigma} \right) \right]$$

be its cumulative distribution. Here

$$\operatorname{erf}(x) = \frac{2}{\sqrt{\pi}} \int_0^x e^{-u^2} du.$$

2 Known integrals

Most of these are not useful but are given for reference. Some obvious conditions on α, β etc. are not stated. The first integral will be used in section 3.

$$\begin{aligned} \int_{-\infty}^{\infty} e^{-(\alpha x + \beta)^2} \operatorname{erf}(\gamma x + \delta) dx &= \frac{\sqrt{\pi}}{\alpha} \operatorname{erf} \left[\frac{\alpha\delta - \beta\gamma}{\sqrt{\alpha^2 + \gamma^2}} \right] \\ \int_0^{\infty} e^{-\alpha^2 x^2} \operatorname{erf}(\beta x) dx &= \frac{\arctan(\beta/\alpha)}{\alpha\sqrt{\pi}} \\ \int_0^{\infty} e^{-x^2} \operatorname{erf}(x) dx &= \frac{\sqrt{\pi}}{4} \\ \int_0^{\infty} e^{-x^2} \operatorname{erf}^3(x) dx &= \frac{\sqrt{\pi}}{8} \\ \int e^{-x^2} \operatorname{erf}^n(x) dx &= \frac{\sqrt{\pi} \operatorname{erf}^{n+1}(x)}{2(n+1)} \\ \int_0^{\infty} x e^{-\alpha^2 x^2} \operatorname{erf}(\beta x) dx &= \frac{\beta}{2\alpha^2 \sqrt{\alpha^2 + \beta^2}} \\ \int \operatorname{erf}^2(x) dx &= \frac{\sqrt{\pi} x \operatorname{erf}^2 x + 2 e^{-x^2} \operatorname{erf}(x)}{\sqrt{\pi}} - \frac{2 \operatorname{erf}(\sqrt{2}x)}{\sqrt{2\pi}} \end{aligned}$$

3 $\Pr[x_1 < x_2]$

Let $x_1 \sim N(\mu_1, \sigma_1^2)$ and $x_2 \sim N(\mu_2, \sigma_2^2)$ be two independent Gaussian RVs. I would like to calculate $\Pr[x_1 < x_2]$. This is given by

$$\begin{aligned}\Pr[x_1 < x_2] &= \int_{-\infty}^{\infty} \int_{x_1}^{\infty} f_{\mu_1, \sigma_1^2}(x_1) f_{\mu_2, \sigma_2^2}(x_2) dx_2 dx_1 \\ &= \frac{1}{2} - \frac{1}{2\sigma_2\sqrt{2\pi}} \int_{-\infty}^{\infty} \exp\left[-\frac{(x_1 - \mu_1)^2}{2\sigma_1^2}\right] \operatorname{erf}\left[\frac{x_1 - \mu_2}{\sqrt{2}\sigma_2}\right] dx_1 \\ &= \frac{1}{2} \left[1 + \operatorname{erf}\left(\frac{\mu_2 - \mu_1}{\sqrt{2(\sigma_1^2 + \sigma_2^2)}}\right) \right]\end{aligned}$$

using the first integral of the last section.

The same result can be obtained more easily - the sum of two Gaussians is Gaussian, so the result is immediate from $\Pr[x_1 < x_2] = \Pr[x_2 - x_1 > 0]$.

4 The mean of the maximum of Gaussians

I would like to compute the mean of the maximum of $n = 2, 3, 4, \dots$ iid $N(0,1)$ variates. This is given by

$$\mu(n) = n \int_{-\infty}^{\infty} x f(x) F^n(x) dx = \frac{n2^{3-n}}{\sqrt{2\pi}} \int_0^{\infty} u e^{-u^2} \sum_{k=1, k \text{ odd}}^{n-1} \binom{n-1}{k} \operatorname{erf}^k(u) du.$$

We have $\mu(2) = 1/\sqrt{\pi}$, $\mu(3) = 3/(2\sqrt{\pi})$, and it seems that $\cos(\mu(4)\pi^{3/2}) = 23/27$, though I cannot prove this.

Using results from David *Order statistics* pp. 32-38:

$$\begin{aligned}
 \mu(1,1) &= 0 \\
 \mu(1,2) &= -\pi^{-1/2} \\
 \mu(2,2) &= \pi^{-1/2} \\
 \mu(1,3) &= -\frac{3}{2}\pi^{-1/2} \\
 \mu(2,3) &= 0 \\
 \mu(3,3) &= \frac{3}{2}\pi^{-1/2} \\
 \mu(1,4) &= -6\pi^{-3/2}\arctan(2^{1/2}) \\
 \mu(2,4) &= ? \\
 \mu(3,4) &= ? \\
 \mu(4,4) &= -\mu(1,4) \\
 \mu(1,5) &= -\mu(5,5) \\
 \mu(2,5) &= -\mu(4,5) \\
 \mu(3,5) &= 0 \\
 \mu(4,5) &= \frac{5}{4}\pi^{-1/2} - 15\pi^{-3/2}\arcsin(1/3) \\
 \mu(5,5) &= \frac{5}{4}\pi^{-1/2} + \frac{15}{2}\pi^{-3/2}\arcsin(1/3)
 \end{aligned}$$

Here $\mu(r,n)$ is the expectation value of the r th order statistic for a sample of n $N(0,1)$ RVs.