

1 Double-sided Crystal Ball

PDF and CDF definitions:

$$\int_{-\infty}^{\infty} f(x) dx = 1 \quad (1)$$

$$F(x) = \int_{-\infty}^x f(x') dx' \quad (2)$$

$$y \equiv F(x) \rightarrow F^{-1}(y) = x \quad (3)$$

Parameters:

$$\vec{p} = (\mu, \sigma, a_L, n_L, a_R, n_R) \quad (4)$$

$$d_L = n_L/a_L \quad (5)$$

$$d_R = n_R/a_R \quad (6)$$

Parameter conditions:

$$n_L, n_R > 1 \quad (7)$$

$$a_L, a_R > 0 \quad (8)$$

Normalization:

$$N = \frac{1}{\sigma \left[\frac{d_L}{n_L-1} \cdot \exp\left(-\frac{a_L^2}{2}\right) + \sqrt{\frac{\pi}{2}} \left(\operatorname{erf}\left(\frac{a_L}{\sqrt{2}}\right) + \operatorname{erf}\left(\frac{a_R}{\sqrt{2}}\right) \right) + \frac{d_R}{n_R-1} \cdot \exp\left(-\frac{a_R^2}{2}\right) \right]} \quad (9)$$

Probability density function:

$$f(x; \vec{p}) = N \cdot \begin{cases} \exp\left(-\frac{a_L^2}{2}\right) \cdot \left[\frac{1}{d_L} \left(d_L - a_L - \frac{x-\mu}{\sigma} \right) \right]^{-n_L} & \text{for } \frac{x-\mu}{\sigma} \leq -a_L \\ \exp\left(-\frac{1}{2} \left(\frac{x-\mu}{\sigma} \right)^2\right) & \text{for } -a_L < \frac{x-\mu}{\sigma} < a_R \\ \exp\left(-\frac{a_R^2}{2}\right) \cdot \left[\frac{1}{d_R} \left(d_R - a_R + \frac{x-\mu}{\sigma} \right) \right]^{-n_R} & \text{for } \frac{x-\mu}{\sigma} \geq a_R \end{cases} \quad (10)$$

Cumulative distribution function:

$$F(x; \vec{p}) = \sigma N \cdot \begin{cases} \frac{d_L}{n_L-1} \exp\left(-\frac{a_L^2}{2}\right) \left[\frac{1}{d_L} \left(d_L - a_L - \frac{x-\mu}{\sigma}\right)\right]^{-n_L+1} & \text{for } \frac{x-\mu}{\sigma} \leq -a_L \\ \frac{d_L}{n_L-1} \exp\left(-\frac{a_L^2}{2}\right) + \sqrt{\frac{\pi}{2}} \operatorname{erf}\left(\frac{a_L}{\sqrt{2}}\right) + \sqrt{\frac{\pi}{2}} \operatorname{erf}\left(\frac{x-\mu}{\sigma\sqrt{2}}\right) & \text{for } -a_L < \frac{x-\mu}{\sigma} < a_R \\ \frac{d_L}{n_L-1} \exp\left(-\frac{a_L^2}{2}\right) + \sqrt{\frac{\pi}{2}} \operatorname{erf}\left(\frac{a_L}{\sqrt{2}}\right) \\ + \sqrt{\frac{\pi}{2}} \operatorname{erf}\left(\frac{a_R}{\sqrt{2}}\right) + \frac{d_R}{n_R-1} \exp\left(-\frac{a_R^2}{2}\right) \\ + \frac{d_R}{1-n_R} \exp\left(-\frac{a_R^2}{2}\right) \left[\frac{1}{d_R} \left(d_R - a_R + \frac{x-\mu}{\sigma}\right)\right]^{-n_R+1} & \text{for } \frac{x-\mu}{\sigma} \geq a_R \end{cases} \quad (11)$$

$$= \sigma N \cdot \begin{cases} B_L \left[\frac{1}{d_L} \left(d_L - a_L - \frac{x-\mu}{\sigma}\right)\right]^{-n_L+1} & \text{for } \frac{x-\mu}{\sigma} \leq -a_L \\ A_L + C_L + \sqrt{\frac{\pi}{2}} \left(1 - \operatorname{erfc}\left(\frac{x-\mu}{\sigma\sqrt{2}}\right)\right) & \text{for } -a_L < \frac{x-\mu}{\sigma} < a_R \\ A_L + C_L + C_R + A_R \\ + B_R \left[\frac{1}{d_R} \left(d_R - a_R + \frac{x-\mu}{\sigma}\right)\right]^{-n_R+1} & \text{for } \frac{x-\mu}{\sigma} \geq a_R \end{cases} \quad (12)$$

Inverse cumulative distribution function:

$$x = \begin{cases} \mu + \sigma \left(-d_L \left[\frac{y}{\sigma N} / B_L\right]^{-\frac{1}{-n_L+1}} - a_L + d_L\right) & \text{for } y < \sigma N A_L \\ \mu + \sigma \sqrt{2} \operatorname{erfc}^{-1} \left[1 - \sqrt{\frac{2}{\pi}} \left(\frac{y}{\sigma N} - A_L - C_L\right)\right] & \text{for } \sigma N A_L \leq y \leq \sigma N (A_L + C_L + C_R) \\ \mu + \sigma \left(d_R \left[\frac{\frac{y}{\sigma N} - A_L - C_L - C_R - A_R}{B_R}\right]^{-\frac{1}{-n_R+1}} + a_R - d_R\right) & \text{for } y > \sigma N (A_L + C_L + C_R) \end{cases} \quad (13)$$