

# List of Probability Density Function and Cumulative Distribution Function for Common Continuous Random Variable

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## 1. Normal Distribution

$$f(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-(x-\mu)^2/2\sigma^2} \quad (\sigma > 0)$$

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

where  $\operatorname{erf}(x) = \frac{2}{\sqrt{\pi}} \int_0^x e^{-t^2} dt$ .

## 2. Log-Normal Distribution

$$f(x) = \frac{1}{x\sigma\sqrt{2\pi}} \exp \left( -\frac{(\ln x - \mu)^2}{2\sigma^2} \right) \quad (\sigma > 0)$$

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

where  $\operatorname{erf}(x) = \frac{2}{\sqrt{\pi}} \int_0^x e^{-t^2} dt$ .

## 3. Skew-Normal Distribution

$$f(x) = \frac{2}{\omega\sqrt{2\pi}} e^{-\frac{(x-\xi)^2}{2\omega^2}} \int_{-\infty}^{\alpha(\frac{x-\xi}{\omega})} \frac{1}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt \quad (\omega > 0)$$

$$F(x) = \Phi \left( \frac{x-\xi}{\omega} \right) - 2T \left( \frac{x-\xi}{\omega}, \alpha \right)$$

with  $\Phi(x) = \int_{-\infty}^x \phi(t)dt = \frac{1}{2} \left[ 1 + \operatorname{erf} \left( \frac{x}{\sqrt{2}} \right) \right]$  and  $T(h, a) = \frac{1}{2\pi} \int_0^a \frac{e^{-\frac{1}{2}h^2(1+x^2)}}{1+x^2} dx \quad (-\infty < h, a < +\infty)$  is Owen's  $T$  function.

## 4. Alpha-Skew-Normal Distribution

$$f(x) = \frac{(1-\alpha x)^2 + 1}{2 + \alpha^2} \phi(x)$$

where  $\phi(\cdot)$  and  $\Phi(\cdot)$  are p.d.f. and c.d.f. of the normal distribution with mean  $\mu$  and variance  $\sigma^2$ .

$$F(x) = \Phi(x) + \alpha \left( \frac{2 - \alpha x}{2 + \alpha^2} \right) \phi(x)$$

## 5. Log-Skew-Normal Distribution

$$f(x) = \frac{2}{x} \phi(\ln x) \Phi(\lambda \ln x)$$

where  $\phi(\cdot)$  and  $\Phi(\cdot)$  are p.d.f. and c.d.f. of the normal distribution with mean  $\mu$  and variance  $\sigma^2$ .

## 6. Generalized-Normal Distribution

$$f(x) = \frac{\beta}{2\alpha\Gamma(1/\beta)} e^{-(|x-\mu|/\alpha)^\beta} \quad (\alpha > 0; \beta > 0)$$

where  $\Gamma(z) = \int_0^\infty t^{z-1} e^{-t} dt$  denotes the gamma function.

$$F(x) = \frac{1}{2} + \text{sign}(x - \mu) \frac{1}{2\Gamma(1/\beta)} \gamma \left( \frac{x - \mu}{\alpha}, \left| \frac{x - \mu}{\alpha} \right|^\beta \right)$$

where  $\beta$  is a shape parameter,  $\alpha$  is a scale parameter and  $\gamma(s, x) = \int_0^x t^{s-1} e^{-t} dt$  is the unnormalized incomplete lower gamma function.

## 7. Mixture of Two Normal

$$f(x, p) = pf_1(x) + (1 - p)f_2(x)$$

with  $0 < p < 1$ , for  $i = 1, 2$

$$f_i(x) = \frac{1}{\sigma_i \sqrt{2\pi}} e^{-(x - \mu_i)^2 / 2\sigma_i^2}$$

## 8. Beta-Normal Distribution

$$F(x) = G \left[ \Phi \left( \frac{x - \mu}{\sigma} \right) \right] = \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha)\Gamma(\beta)} \left[ \Phi \left( \frac{x - \mu}{\sigma} \right) \right]^{\alpha-1} \times \left[ 1 - \Phi \left( \frac{x - \mu}{\sigma} \right) \right]^{\beta-1} \sigma^{-1} \phi \left( \frac{x - \mu}{\sigma} \right)$$

where

$$G(x) = \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha)\Gamma(\beta)} \int_0^{\Phi(x)} t^{\alpha-1} (1-t)^{\beta-1} dt, \quad 0 < \alpha, \quad \beta < \infty$$

and

$$g(x) = \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha)\Gamma(\beta)} \Phi(x)^{\alpha-1} (1 - \Phi(x))^{\beta-1} \Phi'(x)$$

and  $\Phi(x)$  is the c.d.f. of normal distribution with mean  $\mu$  and variance  $\sigma^2$ .

## 9. Arcsine Gaussian Distribution

$$f(x) = \frac{\phi(x)}{\pi \sqrt{\Phi(x)[1 - \Phi(x)]}}$$

where  $\phi(\cdot)$  and  $\Phi(\cdot)$  are p.d.f. and c.d.f. of the normal distribution with mean  $\mu$  and variance  $\sigma^2$ .

#### 10. Double Gaussian (Two-Piece Normal) Distribution

$$f(x) = \begin{cases} A \exp [-(x - \mu)^2 / 2\sigma_1^2], & x \leq \mu, \\ A \exp [-(x - \mu)^2 / 2\sigma_2^2], & x \geq \mu, \end{cases}$$

where  $A = (\sqrt{2\pi} (\sigma_1 + \sigma_2) / 2)^{-1}$ .

#### 11. Inverse Gaussian Distribution

$$f(x) = \sqrt{\frac{\lambda}{2\pi x^3}} \exp \left[ -\frac{\lambda(x - \mu)^2}{2\mu^2 x} \right]$$

$$F(x) = \Phi \left( \sqrt{\frac{\lambda}{x}} \left( \frac{x}{\mu} - 1 \right) \right) + \exp \left( \frac{2\lambda}{\mu} \right) \Phi \left( -\sqrt{\frac{\lambda}{x}} \left( \frac{x}{\mu} + 1 \right) \right)$$

and  $\Phi(x)$  is the c.d.f. of standard normal distribution.

#### 12. Laplace Distribution

$$f(x) = \frac{1}{2\sigma} \exp \left( -\frac{|x - \mu|}{\sigma} \right) \quad (\sigma > 0)$$

$$F(x) = \begin{cases} \frac{1}{2} \exp \left( \frac{x - \mu}{\sigma} \right) & \text{if } x \leq \mu \\ 1 - \frac{1}{2} \exp \left( -\frac{x - \mu}{\sigma} \right) & \text{if } x \geq \mu \end{cases}$$

#### 13. Beta Laplace Distribution

$$f(x) = \begin{cases} \{2^a B(a, b)\}^{-1} \exp(-|x|) \exp\{-|x|(a - 1)\} \{1 - \exp(-|x|)/2\}^{b-1}, & x < 0, \\ \{2^b B(a, b)\}^{-1} \exp(-|x|) \exp\{-|x|(b - 1)\} \{1 - \exp(-|x|)/2\}^{a-1}, & x \geq 0 \end{cases}$$

where  $a > 0$  and  $b > 0$

$$F(x) = \begin{cases} I_{\exp(x)/2}(a, b), & x < 0, \\ I_{1-\exp(-x)/2}(a, b), & x \geq 0. \end{cases}$$

#### 14. Double Exponential (Skew Laplace) Distribution

$$f(x) = \begin{cases} \sqrt{\frac{2}{\pi}} \frac{1}{\sigma_1 + \sigma_2} e^{-\frac{x^2}{2\sigma_2^2}}, & x < 0 \\ \sqrt{\frac{2}{\pi}} \frac{1}{\sigma_1 + \sigma_2} e^{-\frac{x^2}{2\sigma_1^2}}, & x \geq 0, \end{cases}$$

where  $\sigma_1 > 0$  and  $\sigma_2 > 0$ .

#### 15. Asymmetric Laplace Distribution

$$f(x) = \frac{\lambda}{\kappa + 1/\kappa} \begin{cases} \exp((\lambda/\kappa)(x - \mu)) & \text{if } x < \mu \\ \exp(-\lambda\kappa(x - \mu)) & \text{if } x \geq \mu \end{cases}$$

where,  $\mu$  is a location parameter,  $\lambda > 0$  is a scale parameter, and  $\kappa$  is an asymmetry parameter.

$$F(x) = \begin{cases} \frac{\kappa^2}{1+\kappa^2} \exp((\lambda/\kappa)(x-\mu)) & \text{if } x \leq \mu \\ 1 - \frac{1}{1+\kappa^2} \exp(-\lambda\kappa(x-\mu)) & \text{if } x > \mu \end{cases}$$

16. Exponential Distribution

$$\begin{aligned} f(x) &= \lambda e^{-\lambda x} \quad (\lambda > 0) \\ F(x) &= 1 - e^{-\lambda x} \end{aligned}$$

17. Gamma Distribution

$$\begin{aligned} f(x) &= \frac{\beta^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{-\beta x} \quad (\alpha > 0; \beta > 0) \\ F(x) &= \frac{1}{\Gamma(\alpha)} \gamma(\alpha, \beta x) \end{aligned}$$

where  $\gamma(s, x) = \int_0^x t^{s-1} e^{-t} dt$  is the unnormalized incomplete lower gamma function.

18. Beta Distribution

$$f(x) = \frac{x^{\alpha-1}(1-x)^{\beta-1}}{B(\alpha, \beta)} \quad (\alpha > 0; \beta > 0)$$

where  $B(\alpha, \beta) = \frac{\Gamma(\alpha)\Gamma(\beta)}{\Gamma(\alpha+\beta)}$  and  $\Gamma$  is the Gamma function.

$$F(x) = \frac{B(x; \alpha, \beta)}{B(\alpha, \beta)}$$

where  $B(x; a, b) = \int_0^x t^{a-1}(1-t)^{b-1} dt$  is the incomplete beta function.

19. Beta Prime Distribution

$$\begin{aligned} f(x) &= \frac{(x/\sigma)^{\alpha-1}(1+x/\sigma)^{-\alpha-\beta}}{B(\alpha, \beta)\sigma} \quad (\alpha > 0; \beta > 0) \\ F(x) &= I_{\frac{x/\sigma}{1+x/\sigma}}(\alpha, \beta) \end{aligned}$$

where  $I_x(\alpha, \beta) = \frac{B(x; a, b)}{B(a, b)}$  is the incomplete beta function.

20. Logistic Distribution

$$\begin{aligned} f(x) &= \frac{e^{-(x-\mu)/s}}{s(1+e^{-(x-\mu)/s})^2} \quad (s > 0) \\ F(x) &= \frac{1}{1+e^{-(x-\mu)/s}} \end{aligned}$$

21. Double Logistic Distribution

$$f(x) = \begin{cases} \frac{1}{\sigma_1+\sigma_2} \frac{2e^{\frac{x}{\sigma_2}}}{\left(1+e^{\frac{x}{\sigma_2}}\right)^2}, & x < 0 \\ \frac{1}{\sigma_1+\sigma_2} \frac{2e^{-\frac{x}{\sigma_1}}}{\left(1+e^{-\frac{x}{\sigma_1}}\right)^2}, & x \geq 0, \end{cases}$$

where  $\sigma_1 > 0$  and  $\sigma_2 > 0$ .

## 22. Generalized Extreme Value Distribution

$$f(s) = \begin{cases} \exp(-s) \exp(-\exp(-s)) & \text{for } \xi = 0 \\ (1 + \xi s)^{-(1+1/\xi)} \exp(-(1 + \xi s)^{-1/\xi}) & \text{for } \xi \neq 0 \text{ and } \xi s > -1 \\ 0 & \text{otherwise} \end{cases}$$

where  $s = (x - \mu)/\sigma$  and  $\sigma > 0$ .

$$F(s) = \begin{cases} \exp(-\exp(-s)) & \text{for } \xi = 0 \\ \exp(-(1 + \xi s)^{-1/\xi}) & \text{for } \xi \neq 0 \text{ and } \xi s > -1 \\ 0 & \text{for } \xi > 0 \text{ and } \xi s \leq -1 \\ 1 & \text{for } \xi < 0 \text{ and } \xi s \leq -1 \end{cases}$$

## 23. Weibull Distribution

$$f(x) = \frac{k}{\sigma} \left( \frac{x}{\sigma} \right)^{k-1} e^{-(x/\sigma)^k} \quad (x > 0; k > 0; \sigma > 0)$$

$$F(x) = 1 - e^{-(x/\sigma)^k}$$

## 24. Gumbel Distribution

$$f(x) = \frac{1}{\sigma} e^{-(z+e^{-z})}$$

where  $z = \frac{x-\mu}{\sigma}$

$$F(x) = e^{-e^{-(x-\mu)/\sigma}}$$

## 25. Student's t Distribution

$$f(x) = \frac{\Gamma(\frac{\nu+1}{2})}{\sqrt{\nu\pi}\Gamma(\frac{\nu}{2})} \left(1 + \frac{x^2}{\nu}\right)^{-\frac{\nu+1}{2}} \quad (\nu > 0)$$

$$F(x) = \frac{1}{2} + x\Gamma\left(\frac{\nu+1}{2}\right) \times \frac{{}_2F_1\left(\frac{1}{2}, \frac{\nu+1}{2}; \frac{3}{2}; -\frac{x^2}{\nu}\right)}{\sqrt{\pi\nu}\Gamma(\frac{\nu}{2})}$$

where  ${}_2F_1(a, b; c; z) = \sum_{n=0}^{\infty} \frac{(a)_n(b)_n}{(c)_n n!} z^n$  is the hypergeometric function.

## 26. Skew t Distribution

$$f(x) = 2g(x; v)G\left(\lambda x \sqrt{\frac{1+v}{x^2+v}}; v+1\right)$$

where  $g(x; v)$  and  $G(x; v)$  are the p.d.f. and c.d.f. of the usual Student's-  $t$  distribution with  $v$  degrees of freedom.

## 27. Chi-square Distribution

$$f(x) = \frac{1}{2^{k/2}\Gamma(k/2)} x^{k/2-1} e^{-x/2}$$

where  $x > 0$  if  $k = 1$ , otherwise  $x \geq 0$

$$F(x) = \frac{1}{\Gamma(k/2)} \gamma\left(\frac{k}{2}, \frac{x}{2}\right)$$

28. F Distribution

$$f(x) = \frac{\sqrt{\frac{(d_1 x)^{d_1} d_2^{d_2}}{(d_1 x + d_2)^{d_1 + d_2}}}}{x B\left(\frac{d_1}{2}, \frac{d_2}{2}\right)} \quad (d_1 > 0; d_2 > 0)$$

where  $x > 0$  if  $d_1 = 1$ , otherwise  $x \geq 0$

$$F(x) = I_{\frac{d_1 x}{d_1 x + d_2}}\left(\frac{d_1}{2}, \frac{d_2}{2}\right)$$

where  $I_x(a, b) = \frac{B(x; a, b)}{B(a, b)}$  and  $B(x; a, b) = \int_0^x t^{a-1} (1-t)^{b-1} dt$ .

29. Pareto Distribution

$$f(x) = \frac{\alpha x_m^\alpha}{x^{\alpha+1}} \quad (x \geq x_m; x_m > 0; \alpha > 0)$$

$$F(x) = 1 - \left(\frac{x_m}{x}\right)^\alpha$$