

Discrete	Description	Mean $E(X) = \mu$	Variance $Var(X) = \sigma^2$	PMF $P(X = x)$	CDF $F(X) = P(X \leq x)$
Uniform $X \sim \text{Discrete Uni}(a, b)$	Each of the n outcomes has an equal probability.	$\frac{a+b}{2}$	$\frac{(b-a+1)^2-1}{12}$	$\frac{1}{b-a+1}, a \leq x \leq b$	$\frac{\lfloor x \rfloor - a + 1}{b - a + 1}, a \leq x \leq b$
Bernoulli $X \sim \text{Ber}(p)$	A single binary outcome, success (1) with probability p and failure (0) with probability $1-p$.	p	$p(1-p)$	$p^x(1-p)^{1-x}, x \in \{0, 1\}$	$\begin{cases} 0, & x < 0 \\ 1-p, & 0 \leq x < 1 \\ 1, & x \geq 1 \end{cases}$
Binomial $X \sim \text{Bin}(n, p)$	The sum of n independent Bernoulli trials with success probability p .	np	$np(1-p)$	$\binom{n}{x} p^x (1-p)^{n-x}$	$\sum_{k=0}^x P(X = k)$
Negative Binomial $X \sim \text{NegBin}(r, p)$	The number of trials required to achieve r successes in independent Bernoulli trials with success probability p .	$\frac{r}{p}$	$\frac{r(1-p)}{p^2}$	$\binom{x-1}{r-1} p^r (1-p)^{x-r}$	$\sum_{k=r}^x P(X = k)$
Geometric $X \sim \text{Geo}(p)$	The number of trials until the first success in independent Bernoulli trials with success probability p .	$\frac{1}{p}$	$\frac{1-p}{p^2}$	$p(1-p)^{x-1}$	$1 - (1-p)^x$
Hypergeometric $X \sim \text{HypGeo}(n, N, M)$	The number of successes in a sample of size n drawn without replacement from a population of size N containing M successes.	$n \frac{M}{N}$	$n \frac{M}{N} \left(1 - \frac{M}{N}\right) \frac{N-n}{N-1}$	$\frac{\binom{M}{x} \binom{N-M}{n-x}}{\binom{N}{n}}$	$\sum_{k=0}^x P(X = k)$
Poisson $X \sim \text{Poi}(\lambda)$	Models the number of events occurring in a fixed interval of time or space with a known average rate λ .	λ	λ	$\frac{\lambda^x e^{-\lambda}}{x!}$	$\sum_{k=0}^x P(X = k)$
Continuous	Description	Mean $E(X) = \mu$	Variance $Var(X) = \sigma^2$	PDF $f(x)$	CDF $F(X) = P(X \leq x)$
Uniform $X \sim \text{Uni}(a, b)$	A continuous distribution where all outcomes in the range $[a, b]$ are equally likely.	$\frac{a+b}{2}$	$\frac{(b-a)^2}{12}$	$\frac{1}{b-a}, a \leq x \leq b$	$\frac{x-a}{b-a}, a \leq x \leq b$
Beta $X \sim \text{Beta}(\alpha, \beta)$	A family of distributions defined on $[0, 1]$, often used in Bayesian statistics.	$\frac{\alpha}{\alpha+\beta}$	$\frac{\alpha\beta}{(\alpha+\beta)^2(\alpha+\beta+1)}$	$x^{\alpha-1}(1-x)^{\beta-1} \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)}$	$\frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} \int_0^x t^{\alpha-1}(1-t)^{\beta-1} dt$
Normal $X \sim \mathcal{N}(\mu, \sigma^2)$	A symmetric bell-shaped distribution used extensively in statistics.	μ	σ^2	$\frac{1}{\sigma\sqrt{2\pi}} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right)$	$\Phi\left(\frac{x-\mu}{\sigma}\right)$
Lognormal $X \sim \text{Lognormal}(\mu, \sigma^2)$	A distribution where the logarithm of the variable follows a normal distribution.	$e^{\mu+\sigma^2/2}$	$e^{2\mu+\sigma^2}(e^{\sigma^2}-1)$	$\frac{1}{x\sigma\sqrt{2\pi}} \exp\left(-\frac{(\ln x - \mu)^2}{2\sigma^2}\right)$	$\Phi\left(\frac{\ln x - \mu}{\sigma}\right)$
Gamma $X \sim \text{Gamma}(\alpha, \beta)$	A distribution modelling waiting times and generalising the exponential distribution.	$\alpha\beta$	$\alpha\beta^2$	$\frac{x^{\alpha-1} e^{-x/\beta}}{\Gamma(\alpha)\beta^\alpha}, x > 0$	$\frac{1}{\Gamma(\alpha)\beta^\alpha} \int_0^x t^{\alpha-1} e^{-t/\beta} dt$
Exponential $X \sim \text{Exp}(\lambda) \sim \text{Exp}(\beta)$	A special case of the Gamma distribution modelling the time until an event occurs.	$\frac{1}{\lambda} = \beta$	$\frac{1}{\lambda^2} = \beta^2$	$\lambda e^{-\lambda x} = \frac{1}{\beta} e^{-x/\beta}$	$1 - e^{-\lambda x} = 1 - e^{-x/\beta}$