

Table A.1 Continuous distributions

Distribution	Notation	Parameters
Uniform	$\theta \sim U(\alpha, \beta)$ $p(\theta) = U(\theta \alpha, \beta)$	boundaries $\alpha, \beta$ with $\beta > \alpha$
Normal	$\theta \sim N(\mu, \sigma^2)$ $p(\theta) = N(\theta \mu, \sigma^2)$	location $\mu$ scale $\sigma > 0$
Lognormal	$\theta \sim \text{lognormal}(\mu, \sigma^2)$ $p(\theta) = \text{lognormal}(\theta \mu, \sigma^2)$	location $\mu$ log-scale $\sigma > 0$
Multivariate normal	$\theta \sim N(\mu, \Sigma)$ $p(\theta) = N(\theta \mu, \Sigma)$ (implicit dimension $d$ )	symmetric, pos. definite, $d \times d$ variance matrix $\Sigma$
Gamma	$\theta \sim \text{Gamma}(\alpha, \beta)$ $p(\theta) = \text{Gamma}(\theta \alpha, \beta)$	shape $\alpha > 0$ inverse scale $\beta > 0$
Inverse-gamma	$\theta \sim \text{Inv-gamma}(\alpha, \beta)$ $p(\theta) = \text{Inv-gamma}(\theta \alpha, \beta)$	shape $\alpha > 0$ scale $\beta > 0$
Chi-square	$\theta \sim \chi_{\nu}^2$ $p(\theta) = \chi_{\nu}^2(\theta)$	degrees of freedom $\nu > 0$
Inverse-chi-square	$\theta \sim \text{Inv-}\chi_{\nu}^2$ $p(\theta) = \text{Inv-}\chi_{\nu}^2(\theta)$	degrees of freedom $\nu > 0$
Scaled inverse-chi-square	$\theta \sim \text{Inv-}\chi^2(\nu, s^2)$ $p(\theta) = \text{Inv-}\chi^2(\theta \nu, s^2)$	degrees of freedom $\nu > 0$ scale $s > 0$
Exponential	$\theta \sim \text{Expon}(\beta)$ $p(\theta) = \text{Expon}(\theta \beta)$	inverse scale $\beta > 0$
Laplace (double-exponential)	$\theta \sim \text{Laplace}(\mu, \sigma)$ $p(\theta) = \text{Laplace}(\theta \mu, \sigma)$	location $\mu$ scale $\sigma > 0$
Weibull	$\theta \sim \text{Weibull}(\alpha, \beta)$ $p(\theta) = \text{Weibull}(\theta \alpha, \beta)$	shape $\alpha > 0$ scale $\beta > 0$
Wishart	$W \sim \text{Wishart}_{\nu}(S)$ $p(W) = \text{Wishart}_{\nu}(W S)$ (implicit dimension $k \times k$ )	degrees of freedom $\nu$ symmetric, pos. definite $k \times k$ scale matrix $S$
Inverse-Wishart	$W \sim \text{Inv-Wishart}_{\nu}(S^{-1})$ $p(W) = \text{Inv-Wishart}_{\nu}(W S^{-1})$ (implicit dimension $k \times k$ )	degrees of freedom $\nu$ symmetric, pos. definite $k \times k$ scale matrix $S$
LKJ correlation	$\Sigma \sim \text{LkjCorr}(\eta)$ $p(\Sigma) = \text{LkjCorr}(\Sigma \eta)$ (implicit dimension $k \times k$ )	shape $\eta > 0$

## Density function

## Mean, variance, and mode

$$p(\theta) = \frac{1}{\beta - \alpha}, \quad \theta \in [\alpha, \beta]$$

$$\begin{aligned} E(\theta) &= \frac{\alpha + \beta}{2} \\ \text{var}(\theta) &= \frac{(\beta - \alpha)^2}{12} \\ \text{no mode} \end{aligned}$$

$$p(\theta) = \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{1}{2\sigma^2}(\theta - \mu)^2\right)$$

$$\begin{aligned} E(\theta) &= \mu \\ \text{var}(\theta) &= \sigma^2 \\ \text{mode}(\theta) &= \mu \end{aligned}$$

$$p(\theta) = (\sqrt{2\pi}\sigma\theta)^{-1} \exp\left(-\frac{1}{2\sigma^2}(\log \theta - \mu)^2\right)$$

$$\begin{aligned} E(\theta) &= \exp(\mu + \frac{1}{2}\sigma^2), \\ \text{var}(\theta) &= \exp(2\mu + \sigma^2)(\exp(\sigma^2) - 1) \\ \text{mode}(\theta) &= \exp(\mu - \sigma^2) \end{aligned}$$

$$\begin{aligned} p(\theta) &= (2\pi)^{-d/2} |\Sigma|^{-1/2} \\ &\times \exp\left(-\frac{1}{2}(\theta - \mu)^T \Sigma^{-1}(\theta - \mu)\right) \end{aligned}$$

$$\begin{aligned} E(\theta) &= \mu \\ \text{var}(\theta) &= \Sigma \\ \text{mode}(\theta) &= \mu \end{aligned}$$

$$p(\theta) = \frac{\beta^\alpha}{\Gamma(\alpha)} \theta^{\alpha-1} e^{-\beta\theta}, \quad \theta > 0$$

$$\begin{aligned} E(\theta) &= \frac{\alpha}{\beta} \\ \text{var}(\theta) &= \frac{\alpha}{\beta^2} \\ \text{mode}(\theta) &= \frac{\alpha-1}{\beta}, \text{ for } \alpha \geq 1 \end{aligned}$$

$$p(\theta) = \frac{\beta^\alpha}{\Gamma(\alpha)} \theta^{-(\alpha+1)} e^{-\beta/\theta}, \quad \theta > 0$$

$$\begin{aligned} E(\theta) &= \frac{\beta}{\alpha-1}, \text{ for } \alpha > 1 \\ \text{var}(\theta) &= \frac{\beta^2}{(\alpha-1)^2(\alpha-2)}, \alpha > 2 \\ \text{mode}(\theta) &= \frac{\beta}{\alpha+1} \end{aligned}$$

$$\begin{aligned} p(\theta) &= \frac{2^{-\nu/2}}{\Gamma(\nu/2)} \theta^{\nu/2-1} e^{-\theta/2}, \quad \theta > 0 \\ &\text{same as Gamma}(\alpha = \frac{\nu}{2}, \beta = \frac{1}{2}) \end{aligned}$$

$$\begin{aligned} E(\theta) &= \nu \\ \text{var}(\theta) &= 2\nu \\ \text{mode}(\theta) &= \nu - 2, \text{ for } \nu \geq 2 \end{aligned}$$

$$\begin{aligned} p(\theta) &= \frac{2^{-\nu/2}}{\Gamma(\nu/2)} \theta^{-(\nu/2+1)} e^{-1/(2\theta)}, \quad \theta > 0 \\ &\text{same as Inv-gamma}(\alpha = \frac{\nu}{2}, \beta = \frac{1}{2}) \end{aligned}$$

$$\begin{aligned} E(\theta) &= \frac{1}{\nu-2}, \text{ for } \nu > 2 \\ \text{var}(\theta) &= \frac{2}{(\nu-2)^2(\nu-4)}, \nu > 4 \\ \text{mode}(\theta) &= \frac{1}{\nu+2} \end{aligned}$$

$$\begin{aligned} p(\theta) &= \frac{(\nu/2)^{\nu/2}}{\Gamma(\nu/2)} s^\nu \theta^{-(\nu/2+1)} e^{-\nu s^2/(2\theta)}, \quad \theta > 0 \\ &\text{same as Inv-gamma}(\alpha = \frac{\nu}{2}, \beta = \frac{\nu}{2}s^2) \end{aligned}$$

$$\begin{aligned} E(\theta) &= \frac{\nu}{\nu-2} s^2 \\ \text{var}(\theta) &= \frac{2\nu^2}{(\nu-2)^2(\nu-4)} s^4 \\ \text{mode}(\theta) &= \frac{\nu}{\nu+2} s^2 \end{aligned}$$

$$\begin{aligned} p(\theta) &= \beta e^{-\beta\theta}, \quad \theta > 0 \\ &\text{same as Gamma}(\alpha = 1, \beta) \end{aligned}$$

$$\begin{aligned} E(\theta) &= \frac{1}{\beta} \\ \text{var}(\theta) &= \frac{1}{\beta^2} \\ \text{mode}(\theta) &= 0 \end{aligned}$$

$$p(\theta) = \frac{1}{2\sigma} \exp\left(-\frac{|x-\mu|}{\sigma}\right)$$

$$\begin{aligned} E(\theta) &= \mu \\ \text{var}(\theta) &= 2\sigma^2 \\ \text{mode}(\theta) &= \mu \end{aligned}$$

$$p(\theta) = \frac{\alpha}{\beta^\alpha} \theta^{\alpha-1} \exp(-(\theta/\beta)^\alpha), \quad \theta > 0$$

$$\begin{aligned} E(\theta) &= \beta \Gamma(1 + \frac{1}{\alpha}) \\ \text{var}(\theta) &= \beta^2 [\Gamma(1 + \frac{2}{\alpha}) - (\Gamma(1 + \frac{1}{\alpha}))^2] \\ \text{mode}(\theta) &= \beta(1 - \frac{1}{\alpha})^{1/\alpha} \end{aligned}$$

$$\begin{aligned} p(W) &= \left(2^{\nu k/2} \pi^{k(k-1)/4} \prod_{i=1}^k \Gamma\left(\frac{\nu+1-i}{2}\right)\right)^{-1} \\ &\times |S|^{-\nu/2} |W|^{-(\nu-k-1)/2} \\ &\times \exp\left(-\frac{1}{2} \text{tr}(S^{-1}W)\right), W \text{ pos. definite} \end{aligned}$$

$$E(W) = \nu S$$

$$\begin{aligned} p(W) &= \left(2^{\nu k/2} \pi^{k(k-1)/4} \prod_{i=1}^k \Gamma\left(\frac{\nu+1-i}{2}\right)\right)^{-1} \\ &\times |S|^{\nu/2} |W|^{-(\nu+k+1)/2} \\ &\times \exp\left(-\frac{1}{2} \text{tr}(SW^{-1})\right), W \text{ pos. definite} \end{aligned}$$

$$E(W) = (\nu - k - 1)^{-1} S$$

$$\begin{aligned} p(\Sigma) &= \det(\Sigma)^{\eta-1} \\ &\times 2^{\sum_{i=1}^k (2\eta - 2 + k - i)(k - i)} \\ &\times \prod_{i=1}^k \left(B\left(\frac{i+1}{2}, \frac{i+1}{2}\right)\right)^k \end{aligned}$$

$$E(\Sigma) = I_k,$$

**Table A.1** Continuous distributions *continued*

Distribution	Notation	Parameters
t	$\theta \sim t_\nu(\mu, \sigma^2)$ $p(\theta) = t_\nu(\theta \mu, \sigma^2)$ $t_\nu$ is short for $t_\nu(0, 1)$	degrees of freedom $\nu > 0$ location $\mu$ scale $\sigma > 0$
Multivariate t	$\theta \sim t_\nu(\mu, \Sigma)$ $p(\theta) = t_\nu(\theta \mu, \Sigma)$ (implicit dimension $d$ )	degrees of freedom $\nu > 0$ location $\mu = (\mu_1, \dots, \mu_d)$ symmetric, pos. definite $d \times d$ scale matrix $\Sigma$
Beta	$\theta \sim \text{Beta}(\alpha, \beta)$ $p(\theta) = \text{Beta}(\theta \alpha, \beta)$	'prior sample sizes' $\alpha > 0, \beta > 0$
Dirichlet	$\theta \sim \text{Dirichlet}(\alpha_1, \dots, \alpha_k)$ $p(\theta) = \text{Dirichlet}(\theta \alpha_1, \dots, \alpha_k)$	'prior sample sizes' $\alpha_j > 0; \alpha_0 \equiv \sum_{j=1}^k \alpha_j$
Logistic	$\theta \sim \text{Logistic}(\mu, \sigma)$ $p(\theta) = \text{Logistic}(\theta \mu, \sigma)$	location $\mu$ scale $\sigma > 0$
Log-logistic	$\theta \sim \text{Log-logistic}(\alpha, \beta)$ $p(\theta) = \text{Log-logistic}(\theta \alpha, \beta)$	scale $\alpha > 0$ shape $\beta > 0$

**Table A.2** Discrete distributions

Distribution	Notation	Parameters
Poisson	$\theta \sim \text{Poisson}(\lambda)$ $p(\theta) = \text{Poisson}(\theta \lambda)$	'rate' $\lambda > 0$
Binomial	$\theta \sim \text{Bin}(n, p)$ $p(\theta) = \text{Bin}(\theta n, p)$	'sample size' $n$ (positive integer) 'probability' $p \in [0, 1]$
Multinomial	$\theta \sim \text{Multin}(n; p_1, \dots, p_k)$ $p(\theta) = \text{Multin}(\theta n; p_1, \dots, p_k)$	'sample size' $n$ (positive integer) 'probabilities' $p_j \in [0, 1]; \sum_{j=1}^k p_j = 1$
Negative binomial	$\theta \sim \text{Neg-bin}(\alpha, \beta)$ $p(\theta) = \text{Neg-bin}(\theta \alpha, \beta)$	shape $\alpha > 0$ inverse scale $\beta > 0$
Beta-binomial	$\theta \sim \text{Beta-bin}(n, \alpha, \beta)$ $p(\theta) = \text{Beta-bin}(\theta n, \alpha, \beta)$	'sample size' $n$ (positive integer) 'prior sample sizes' $\alpha > 0, \beta > 0$

## Density function

## Mean, variance, and mode

$$p(\theta) = \frac{\Gamma((\nu+1)/2)}{\Gamma(\nu/2)\sqrt{\nu\pi}\sigma} (1 + \frac{1}{\nu}(\frac{\theta-\mu}{\sigma})^2)^{-(\nu+1)/2}$$

E( $\theta$ ) =  $\mu$ , for  $\nu > 1$   
var( $\theta$ ) =  $\frac{\nu}{\nu-2}\sigma^2$ , for  $\nu > 2$   
mode( $\theta$ ) =  $\mu$

$$p(\theta) = \frac{\Gamma((\nu+d)/2)}{\Gamma(\nu/2)\nu^{d/2}\pi^{d/2}} |\Sigma|^{-1/2} \times (1 + \frac{1}{\nu}(\theta - \mu)^T \Sigma^{-1} (\theta - \mu))^{-(\nu+d)/2}$$

E( $\theta$ ) =  $\mu$ , for  $\nu > 1$   
var( $\theta$ ) =  $\frac{\nu}{\nu-2}\Sigma$ , for  $\nu > 2$   
mode( $\theta$ ) =  $\mu$

$$p(\theta) = \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} \theta^{\alpha-1} (1-\theta)^{\beta-1} \quad \theta \in [0, 1]$$

E( $\theta$ ) =  $\frac{\alpha}{\alpha+\beta}$   
var( $\theta$ ) =  $\frac{\alpha\beta}{(\alpha+\beta)^2(\alpha+\beta+1)}$   
mode( $\theta$ ) =  $\frac{\alpha-1}{\alpha+\beta-2}$

$$p(\theta) = \frac{\Gamma(\alpha_1+\dots+\alpha_k)}{\Gamma(\alpha_1)\dots\Gamma(\alpha_k)} \theta_1^{\alpha_1-1} \dots \theta_k^{\alpha_k-1} \quad \theta_1, \dots, \theta_k \geq 0; \sum_{j=1}^k \theta_j = 1$$

E( $\theta_j$ ) =  $\frac{\alpha_j}{\alpha_0}$   
var( $\theta_j$ ) =  $\frac{\alpha_j(\alpha_0-\alpha_j)}{\alpha_0^2(\alpha_0+1)}$   
cov( $\theta_i, \theta_j$ ) =  $-\frac{\alpha_i\alpha_j}{\alpha_0^2(\alpha_0+1)}$   
mode( $\theta_j$ ) =  $\frac{\alpha_j-1}{\alpha_0-k}$

$$p(\theta) = \frac{\exp(-\frac{x-\mu}{\sigma})}{\sigma(1+\exp(-\frac{x-\mu}{\sigma}))}$$

E( $\theta$ ) =  $\mu$   
var( $\theta$ ) =  $\frac{1}{3}\sigma^2\pi^2$   
mode( $\theta$ ) =  $\mu$

$$p(\theta) = \frac{\frac{\beta}{\alpha} \left(\frac{x}{\alpha}\right)^{\beta-1}}{\left[1 + \left(\frac{x}{\alpha}\right)^{\beta}\right]^2}, \quad \theta > 0$$

E( $\theta$ ) =  $\frac{1}{1 + \left(\frac{x}{\alpha}\right)^{-\beta}}$   
var( $\theta$ ) =  $\alpha^2 \frac{2\pi/\beta}{\sin(2\pi/\beta)}$ ,  $\beta > 2$   
mode( $\theta$ ) =  $\alpha \left(\frac{\beta-1}{\beta+1}\right)^{\frac{1}{\beta}}$ ,  $\beta > 1$

## Density function

## Mean, variance, and mode

$$p(\theta) = \frac{1}{\theta!} \lambda^\theta \exp(-\lambda) \quad \theta = 0, 1, 2, \dots$$

E( $\theta$ ) =  $\lambda$ , var( $\theta$ ) =  $\lambda$   
mode( $\theta$ ) =  $\lfloor \lambda \rfloor$

$$p(\theta) = \binom{n}{\theta} p^\theta (1-p)^{n-\theta} \quad \theta = 0, 1, 2, \dots, n$$

E( $\theta$ ) =  $np$   
var( $\theta$ ) =  $np(1-p)$   
mode( $\theta$ ) =  $\lfloor (n+1)p \rfloor$

$$p(\theta) = \binom{n}{\theta_1 \theta_2 \dots \theta_k} p_1^{\theta_1} \dots p_k^{\theta_k} \quad \theta_j = 0, 1, 2, \dots, n; \sum_{j=1}^k \theta_j = n$$

E( $\theta_j$ ) =  $np_j$   
var( $\theta_j$ ) =  $np_j(1-p_j)$   
cov( $\theta_i, \theta_j$ ) =  $-np_i p_j$

$$p(\theta) = \binom{\theta+\alpha-1}{\alpha-1} \left(\frac{\beta}{\beta+1}\right)^\alpha \left(\frac{1}{\beta+1}\right)^\theta \quad \theta = 0, 1, 2, \dots$$

E( $\theta$ ) =  $\frac{\alpha}{\beta}$   
var( $\theta$ ) =  $\frac{\alpha}{\beta^2}(\beta+1)$

$$p(\theta) = \frac{\Gamma(n+1)}{\Gamma(\theta+1)\Gamma(n-\theta+1)} \frac{\Gamma(\alpha+\theta)\Gamma(n+\beta-\theta)}{\Gamma(\alpha+\beta+n)} \times \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)}, \quad \theta = 0, 1, 2, \dots, n$$

E( $\theta$ ) =  $n \frac{\alpha}{\alpha+\beta}$   
var( $\theta$ ) =  $n \frac{\alpha\beta(\alpha+\beta+n)}{(\alpha+\beta)^2(\alpha+\beta+1)}$