

Table A.1	Continuous distributions	
Distribution	Notation	Parameters
Uniform	$\theta \sim U(\alpha, \beta)$ $p(\theta) = U(\theta \alpha, \beta)$	boundaries α, β with $\beta > \alpha$
Normal	$\theta \sim N(\mu, \sigma^2)$ $p(\theta) = N(\theta \mu, \sigma^2)$	location μ scale $\sigma > 0$
Lognormal	$\theta \sim \text{lognormal}(\mu, \sigma^2)$ $p(\theta) = \text{lognormal}(\theta \mu, \sigma^2)$	location μ 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 Σ
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	$\theta \sim \text{Inv-}\chi_\nu^2(\nu, s^2)$ $p(\theta) = \text{Inv-}\chi_\nu^2(\theta \nu, s^2)$	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 μ 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 ν 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 ν 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
$p(\theta) = \frac{1}{\beta - \alpha}, \quad \theta \in [\alpha, \beta]$	$E(\theta) = \frac{\alpha + \beta}{2}$ $\text{var}(\theta) = \frac{(\beta - \alpha)^2}{12}$ no mode
$p(\theta) = \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{1}{2\sigma^2}(\theta - \mu)^2\right)$	$E(\theta) = \mu$ $\text{var}(\theta) = \sigma^2$ $\text{mode}(\theta) = \mu$
$p(\theta) = (\sqrt{2\pi}\sigma\theta)^{-1} \exp(-\frac{1}{2\sigma^2}(\log \theta - \mu)^2)$	$E(\theta) = \exp(\mu + \frac{1}{2}\sigma^2)$, $\text{var}(\theta) = \exp(2\mu + \sigma^2)(\exp(\sigma^2) - 1)$ $\text{mode}(\theta) = \exp(\mu - \sigma^2)$
$p(\theta) = (2\pi)^{-d/2} \Sigma ^{-1/2} \times \exp\left(-\frac{1}{2}(\theta - \mu)^T \Sigma^{-1}(\theta - \mu)\right)$	$E(\theta) = \mu$ $\text{var}(\theta) = \Sigma$ $\text{mode}(\theta) = \mu$
$p(\theta) = \frac{\beta^\alpha}{\Gamma(\alpha)} \theta^{\alpha-1} e^{-\beta\theta}, \quad \theta > 0$	$E(\theta) = \frac{\alpha}{\beta}$ $\text{var}(\theta) = \frac{\alpha}{\beta^2}$ $\text{mode}(\theta) = \frac{\alpha-1}{\beta}, \text{ for } \alpha \geq 1$
$p(\theta) = \frac{\beta^\alpha}{\Gamma(\alpha)} \theta^{-(\alpha+1)} e^{-\beta/\theta}, \quad \theta > 0$	$E(\theta) = \frac{\beta}{\alpha-1}, \text{ for } \alpha > 1$ $\text{var}(\theta) = \frac{\beta^2}{(\alpha-1)^2(\alpha-2)}, \alpha > 2$ $\text{mode}(\theta) = \frac{\beta}{\alpha+1}$
$p(\theta) = \frac{2^{-\nu/2}}{\Gamma(\nu/2)} \theta^{\nu/2-1} e^{-\theta/2}, \quad \theta > 0$ same as Gamma($\alpha = \frac{\nu}{2}, \beta = \frac{1}{2}$)	$E(\theta) = \nu$ $\text{var}(\theta) = 2\nu$ $\text{mode}(\theta) = \nu - 2, \text{ for } \nu \geq 2$
$p(\theta) = \frac{2^{-\nu/2}}{\Gamma(\nu/2)} \theta^{-(\nu/2+1)} e^{-1/(2\theta)}, \quad \theta > 0$ same as Inv-gamma($\alpha = \frac{\nu}{2}, \beta = \frac{1}{2}$)	$E(\theta) = \frac{1}{\nu-2}, \text{ for } \nu > 2$ $\text{var}(\theta) = \frac{2}{(\nu-2)^2(\nu-4)}, \nu > 4$ $\text{mode}(\theta) = \frac{1}{\nu+2}$
$p(\theta) = \frac{(\nu/2)^{\nu/2}}{\Gamma(\nu/2)} s^\nu \theta^{-(\nu/2+1)} e^{-\nu s^2/(2\theta)}, \quad \theta > 0$ same as Inv-gamma($\alpha = \frac{\nu}{2}, \beta = \frac{\nu}{2} s^2$)	$E(\theta) = \frac{2\nu}{\nu-2} s^2$ $\text{var}(\theta) = \frac{2\nu^2}{(\nu-2)^2(\nu-4)} s^4$ $\text{mode}(\theta) = \frac{1}{\nu+2} s^2$
$p(\theta) = \beta e^{-\beta\theta}, \quad \theta > 0$ same as Gamma($\alpha = 1, \beta$)	$E(\theta) = \frac{1}{\beta}$ $\text{var}(\theta) = \frac{1}{\beta^2}$ $\text{mode}(\theta) = 0$
$p(\theta) = \frac{1}{2\sigma} \exp\left(-\frac{ \theta - \mu }{\sigma}\right)$	$E(\theta) = \mu$ $\text{var}(\theta) = 2\sigma^2$ $\text{mode}(\theta) = \mu$
$p(\theta) = \frac{\alpha}{\beta^\alpha} \theta^{\alpha-1} \exp(-(\theta/\beta)^\alpha), \quad \theta > 0$	$E(\theta) = \beta \Gamma(1 + \frac{1}{\alpha})$ $\text{var}(\theta) = \beta^2 [\Gamma(1 + \frac{2}{\alpha}) - (\Gamma(1 + \frac{1}{\alpha}))^2]$ $\text{mode}(\theta) = \beta(1 - \frac{1}{\alpha})^{1/\alpha}$
$p(W) = \left(\frac{2^{\nu k/2} \pi^{k(k-1)/4}}{\Gamma(\nu/2) \nu^{d/2} \pi^{d/2}} \prod_{i=1}^k \Gamma\left(\frac{\nu+1-i}{2}\right) \right)^{-1} \times S ^{-\nu/2} W ^{-(\nu-k-1)/2} \times \exp\left(-\frac{1}{2} \text{tr}(S^{-1}W)\right), W \text{ pos. definite}$	$E(W) = \nu S$
$p(W) = \left(\frac{2^{\nu k/2} \pi^{k(k-1)/4}}{\Gamma(\nu/2) \nu^{d/2} \pi^{d/2}} \prod_{i=1}^k \Gamma\left(\frac{\nu+1-i}{2}\right) \right)^{-1} \times S ^{ \nu/2 } W ^{-(\nu-k+1)/2} \times \exp\left(-\frac{1}{2} \text{tr}(SW^{-1})\right), W \text{ pos. definite}$	$E(W) = (\nu - k - 1)^{-1} S$
$p(\Sigma) = \det(\Sigma)^{\eta-1} \times 2 \sum_{i=1}^n (2\eta-2+k-i)(k-i) \times \prod_{i=1}^k \left(B\left(\frac{i+1}{2}, \frac{i+1}{2}\right)\right)^k$	$E(\Sigma) = I_k,$

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_ν is short for $t_\nu(0, 1)$	degrees of freedom $\nu > 0$ location μ 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 Σ
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 μ 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
$p(\theta) = \frac{\Gamma((\nu+1)/2)}{\Gamma(\nu/2)\sqrt{\nu\pi}\sigma} \left(1 + \frac{1}{\nu} \left(\frac{\theta - \mu}{\sigma}\right)^2\right)^{-(\nu+1)/2}$	$E(\theta) = \mu, \text{ for } \nu > 1$ $\text{var}(\theta) = \frac{\nu}{\nu-2} \sigma^2, \text{ for } \nu > 2$ $\text{mode}(\theta) = \mu$
$p(\theta) = \frac{\Gamma((\nu+d)/2)}{\Gamma(\nu/2)\nu^{d/2}\pi^{d/2}} \Sigma ^{-1/2} \times \left(1 + \frac{1}{\nu} (\theta - \mu)^T \Sigma^{-1} (\theta - \mu)\right)^{-(\nu+d)/2}$	$E(\theta) = \mu, \text{ for } \nu > 1$ $\text{var}(\theta) = \frac{\nu}{\nu-2} \Sigma, \text{ for } \nu > 2$ $\text{mode}(\theta) = \mu$
$p(\theta) = \frac{\Gamma(\alpha)\Gamma(\beta)}{\Gamma(\alpha+\beta)} \theta^{\alpha-1} (1-\theta)^{\beta-1}, \quad \theta \in [0, 1]$	$E(\theta) = \frac{\alpha}{\alpha+\beta}$ $\text{var}(\theta) = \frac{\alpha\beta}{(\alpha+\beta)^2(\alpha+\beta+1)}$ $\text{mode}(\theta) = \frac{\alpha-1}{\alpha+\beta-2}$
$p(\theta) = \frac{\Gamma(\alpha_1+\dots+\alpha_k)}{\Gamma(\alpha_1)\dots\Gamma(\alpha_k)} \theta_1^{\alpha_1-1} \dots \theta_k^{\alpha_k-1}, \quad \theta_1, \dots, \theta_k \geq 0; \sum_{j=1}^k \theta_j = 1$	$E(\theta_j) = \frac{\alpha_j}{\alpha_0}$ $\text{var}(\theta_j) = \frac{\alpha_j(\alpha_0 - \alpha_j)}{\alpha_0^2(\alpha_0 + 1)}$ $\text{cov}(\theta_i, \theta_j) = -\frac{\alpha_i\alpha_j}{\alpha_0^2(\alpha_0 + 1)}$ $\text{mode}(\theta_j) = \frac{\alpha_j - 1}{\alpha_0 - k}$
$p(\theta) = \frac{\exp\left(-\frac{\theta - \mu}{\sigma}\right)}{\sigma(1 + \exp\left(-\frac{\theta - \mu}{\sigma}\right))}$	$E(\theta) = \mu$ $\text{var}(\theta) = \frac{1}{3} \sigma^2 \pi^2$ $\text{mode}(\theta) = \mu$
$p(\theta) = \frac{\frac{\beta}{\alpha} \left(\frac{\alpha}{\beta}\right)^{\beta-1}}{\left[1 + \left(\frac{\alpha}{\beta}\right)^\beta\right]^\alpha}, \quad \theta > 0$	$E(\theta) = \frac{1}{1 + \left(\frac{\alpha}{\beta}\right)^\beta}$ $\text{var}(\theta) = \alpha^2 \frac{2\pi/\beta}{\sin(2\pi/\beta)}, \quad \beta > 2$ $\text{mode}(\theta) = \alpha \left(\frac{\beta-1}{\beta+1}\right)^{\frac{1}{\beta}}, \quad \beta > 1$
Density function	Mean, variance, and mode
$p(\theta) = \frac{1}{\theta!} \lambda^\theta \exp(-\lambda)$ $\theta = 0, 1, 2, \dots$	$E(\theta) = \lambda, \text{ var}(\theta) = \lambda$ $\text{mode}(\theta) = \lfloor \lambda \rfloor$
$p(\theta) = \binom{n}{\theta} p^\theta (1-p)^{n-\theta}$ $\theta = 0, 1, 2, \dots, n$	$E(\theta) = np$ $\text{var}(\theta) = np(1-p)$ $\text{mode}(\theta) = \lfloor (n+1)p \rfloor$
$p(\theta) = \binom{n}{\theta_1, \theta_2, \dots, \theta_k} p_1^{\theta_1} \dots p_k^{\theta_k}$ $\theta_j = 0, 1, 2, \dots, n; \sum_{j=1}^k \theta_j = n$	$E(\theta_j) = np_j$ $\text{var}(\theta_j) = np_j(1-p_j)$ $\text{cov}(\theta_i, \theta_j) = -np_i p_j$
$p(\theta) = \binom{\theta + \alpha - 1}{\alpha - 1} \left(\frac{\beta}{\beta + 1}\right)^\alpha \left(\frac{1}{\beta + 1}\right)^\theta$ $\theta = 0, 1, 2, \dots$	$E(\theta) = \frac{\alpha}{\beta}$ $\text{var}(\theta) = \frac{\alpha}{\beta^2} (\beta + 1)$
$p(\theta) = \frac{\Gamma(n+1)}{\Gamma(\theta+1)\Gamma(n-\theta+1)} \frac{\Gamma(\alpha+\theta)\Gamma(n+\beta-\theta)}{\Gamma(\alpha+\beta)} \times \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)}, \quad \theta = 0, 1, 2, \dots, n$	$E(\theta) = n \frac{\alpha}{\alpha+\beta}$ $\text{var}(\theta) = n \frac{\alpha\beta}{(\alpha+\beta)^2(\alpha+\beta+1)}$