## **Probability Distribution**

1 Binomial (n,p),  $(x=0,1,\ldots,n)$  R:xbinom  $f(x) = \binom{n}{x} p^x (1-p)^{n-x}$   $M(t) = (pe^t + 1-p)^n$  E(X) = np, Var(X) = np(1-p)

Note:  $n \to \infty$  with  $\mu = np \Rightarrow \operatorname{Poisson}(\mu)$   $F_{\operatorname{Bin}(n,p)}(x) = F_{\operatorname{Beta}(n-x,x+1)}(1-p)$   $F_{\operatorname{Bin}(n,p)}(r-1) = 1 - F_{\operatorname{NegBin}(r,p)}(n-r)$  $X_i \sim \operatorname{Bin}(n_i,p) \Rightarrow \sum X_i \sim \operatorname{Bin}\left(\sum n_i,p\right)$ 

 $\begin{array}{ll} \text{R:}x\mathsf{geom} \\ f(x) = p(1-p)^x, & (x=0,1,\ldots) \\ F(t) = 1 - (1-p)^{\lfloor t \rfloor + 1}, \text{ where } \lfloor t \rfloor = \max\{m \in \mathbb{Z} : m \leq t\} \\ S(t) = P[X \geq t] = (1-p)^{\lceil t \rceil} \text{ with } \lceil t \rceil = \min\{m \in \mathbb{Z} : m \geq t\} \\ R(t) = P[X > t] = (1-p)^{\lfloor t \rfloor + 1} \\ M_X(t) = \frac{p}{1 - (1-p)e^t}, & (t < -\log(1-p)) \\ E(X) = \frac{1-p}{p}, \operatorname{Var}(X) = \frac{1-p}{n^2} \\ \end{array}$ 

$$\begin{split} f(y) &= p(1-p)^{y-1}, \quad (y=1,2,\ldots) \\ F(t) &= 1 - (1-p)^{\lfloor t \rfloor} \\ S(t) &= P[Y \geq t] = (1-p)^{\lceil t \rceil - 1} \\ R(t) &= P[Y > t] = (1-p)^{\lfloor t \rfloor} \\ M_Y(t) &= \frac{pe^t}{1 - (1-p)e^t}, \quad (t < -\log(1-p)) \\ E(Y) &= \frac{1}{p}, \operatorname{Var}(Y) = \frac{1-p}{p^2} \end{split}$$

Note:  $\min_{1 \le i \le n} (X_i) \sim \text{Geo}(1 - (1 - p)^n)$ : self-reproducing

 $\begin{aligned} & & & \text{Hypergeometric}\left(N,M,n\right) & & & \text{R:}x \\ & & & & f(x) = \frac{\binom{M}{x}\binom{N-M}{n-x}}{\binom{N}{n}}, \\ & & & & & \left(\max(0,M-(N-n)) \leq x \leq \min(n,M)\right) \\ & & & & E(X) = n\left(\frac{M}{N}\right), \operatorname{Var}(X) = n\frac{M}{N}\frac{(N-M)(N-n)}{N(N-1)} \end{aligned}$ 

4 Negative Binomial (r,p)  $f(x) = {r+x-1 \choose x} p^r (1-p)^x, (x=0,1,\ldots)$   $M_X(t) = \left(\frac{p}{1-(1-p)e^t}\right)^r$   $E(X) = r\frac{1-p}{p}, \ \mathrm{Var}(X) = r\frac{1-p}{p^2}$ 

$$f(y) = {y-1 \choose r-1} p^r (1-p)^{y-r}, (y=r,r+1,...)$$

$$M_Y(t) = \left(\frac{pe^t}{1-(1-p)e^t}\right)^r$$

$$E(Y) = r\frac{1}{p}, \text{ Var}(Y) = r\frac{1-p}{p^2}$$

NOTE:  $X = V_1 + \dots + V_r$ ,  $(V_i \sim \text{Geometric}(p))$   $(1-x)^{-n} = \sum_{k=0}^{\infty} {n+k-1 \choose k} x^k \quad (|x| < 1)$   $\mu = r(1-p), r \to \infty \Rightarrow \text{Poisson}(\mu)$   $F_{\text{Bin}(n,p)}(r-1) = 1 - F_{\text{NegBin}(r,p)}(n-r)$  $X_i \sim \text{NB}(r_i, p) \Rightarrow \sum X_i \sim \text{NB}(\sum r_i, p)$ 

5 Poisson  $(\mu)$  R:xpois  $f(x) = \frac{e^{-\mu}\mu^x}{x!}$ , (x = 0, 1, ...)  $M(t) = \exp(\mu(e^t - 1))$   $E(X) = \mu$ ,  $Var(X) = \mu$  Note:  $X_i \sim Poi(\mu_i) \Rightarrow \sum X_i \sim Poi(\sum \mu_i)$ 

$$P_{\text{Poi}(\mu)}[X \ge n] = F_{\Gamma(n,1)}(\mu) = F_{\Gamma(n,\mu)}(1)$$
  
=  $F_{\chi^2_{2n}}(2\mu)$ 

1 Beta  $(\alpha, \beta)$ ,  $(\alpha > 0, \beta > 0)$  R:xbeta  $f(x) = \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha)\Gamma(\beta)}x^{\alpha-1}(1-x)^{\beta-1}$ , (0 < x < 1)  $M(t) = 1 + \sum_{k=1}^{\infty} \left(\prod_{r=1}^{k-1} \frac{\alpha+r}{\alpha+\beta+r}\right) \frac{t^k}{k!}$   $E(X) = \frac{\alpha}{\alpha+\beta}$ ,  $Var(X) = \frac{\alpha\beta}{(\alpha+\beta+1)(\alpha+\beta)^2}$  Note:  $F_{\text{Beta}(n-x,x+1)}(1-p) = F_{\text{Bin}(n,p)}(x)$  2 BS  $(\alpha, \beta)$ ,  $(\alpha > 0, \beta > 0)$ : Birnbaum-Saunders

2 **BS** 
$$(\alpha, \beta)$$
,  $(\alpha > 0, \beta > 0)$ : Birnbaum-Saunders  $f(t) = \frac{1}{2\alpha\beta} \sqrt{\frac{\beta}{t}} \left( 1 + \frac{\beta}{t} \right) \phi \left[ \frac{1}{\alpha} \left( \sqrt{\frac{t}{\beta}} - \sqrt{\frac{\beta}{t}} \right) \right], (t > 0)$   $F(t) = \Phi \left[ \frac{1}{\alpha} \left( \sqrt{\frac{t}{\beta}} - \sqrt{\frac{\beta}{t}} \right) \right]$ 

$$\begin{split} F^{-1}(p) &= \frac{1}{4} \Big[ \alpha \sqrt{\beta} \Phi^{-1}(p) + \sqrt{\alpha^2 \beta \{\Phi^{-1}(p)\}^2 + 4\beta} \Big]^2 \\ &= \beta \Big\{ 1 + \gamma(p)^2 + \gamma(p) \sqrt{\gamma(p)^2 + 2} \Big\}, \\ \text{where } \gamma(p) &= \alpha \Phi^{-1}(p) / \sqrt{2} \end{split}$$

$$\begin{split} E(T) &= \beta (1 + \tfrac{1}{2}\alpha^2), \quad \mathrm{Var}(T) = (\alpha\beta)^2 (1 + \tfrac{5}{4}\alpha^2) \\ \text{NOTE:} \quad & \mathrm{median}(T) = \beta, \, cT \sim \mathrm{BS}(\alpha, c\beta), \, T^{-1} \sim \mathrm{BS}(\alpha, \beta^{-1}) \\ X &= \left(\sqrt{\tfrac{T}{\beta}} - \sqrt{\tfrac{\beta}{T}}\right) \sim N(0, \alpha^2) \\ & \log T \sim \mathrm{sinh-Normal}(\log \beta, \alpha) \end{split}$$

3 Cauchy  $(\alpha, \beta)$  R:x cauchy  $f(x) = \frac{\beta}{\pi} \frac{1}{\beta^2 + (x - \alpha)^2}$   $F(x) = \frac{1}{\pi} \left[ \arctan\left(\frac{x - \alpha}{\beta}\right) + \frac{\pi}{2} \right]$   $\phi(t) = \exp(it\alpha - \beta|t|)$  Note: Cauchy(0, 1) = t(1)

NOTE: Cauchy
$$(0,1) = t(1)$$
  
 $X, Y \sim N(0,1) \Rightarrow X/Y \sim \text{Cauchy}(0,1)$   
 $cX \sim \text{Cauchy}(c\alpha, c\beta)$   
 $X_i \sim \text{Cauchy}(\alpha, \beta) \Rightarrow \frac{1}{n} \sum X_i \sim \text{Cauchy}(\alpha, \beta)$ 

R:xchisq

$$\begin{split} & \text{4 Chi Squared }(n) \\ & f(x) = \frac{1}{\Gamma(n/2)2^{n/2}} x^{n/2-1} e^{-x/2}, \ (x \geq 0) \\ & E(X) = n, \ \text{Var}(X) = 2n \\ & M(t) = \left(\frac{1}{1-2t}\right)^{n/2}, \quad (t < \frac{1}{2}) \\ & E(X^m) = 2^n \Gamma(m+n/2)/\Gamma(n/2) \end{split}$$

Note: 
$$\chi^2(n) = \text{Gamma}(n/2, 2)$$
  
 $\chi^2(2) = \text{Exponential}(\beta = 2)$   
 $X_i \sim \chi^2(n_i) \Rightarrow \sum X_i \sim \chi^2(\sum n_i)$   
 $X_i \sim N(0, 1) \Rightarrow \sum_{i=1}^n X_i^2 \sim \chi^2(n)$   
 $F_{\chi^2(p+2)}(x) = F_{\chi^2(p)}(x) - 2x f_{\chi^2(p)}(x)/p$   
 $F_{\chi^2(1)}(x) = 2\Phi(\sqrt{x}) - 1$   
 $F_{\chi^2(2)}(x) = 1 - \sqrt{2\pi}\phi(\sqrt{x})$   
 $F_{\chi^2(3)}(x) = 2\Phi(\sqrt{x}) - 1 - 2\sqrt{x}\phi(\sqrt{x})$ 

$$F_{\chi^2(2)}(x) = 1 - \sqrt{2\pi}\phi(\sqrt{x})$$
 
$$F_{\chi^2(3)}(x) = 2\Phi(\sqrt{x}) - 1 - 2\sqrt{x}\phi(\sqrt{x})$$
 
$$5 \text{ Exponential } (\beta)$$
 
$$f(x) = \frac{1}{\beta}e^{-x/\beta} \ (x \ge 0, \ \beta > 0)$$
 
$$F(x) = 1 - e^{-x/\beta}$$
 
$$M(t) = \frac{1}{1 - \beta t}, \ (t < \frac{1}{\beta})$$
 
$$E(X) = \beta, \text{Var}(X) = \beta^2$$

Note: Memoryless property 
$$cX \sim \text{Exponential}(c\beta)$$
 
$$\sum X_i \sim \text{Gamma}(n,\theta)$$
 
$$\min_{1 \leq i \leq n} (X_i) \sim \text{Exponential}(\beta/n): \text{ self-reproducing}$$
 
$$Y = X^{1/\alpha} \sim \text{Weibull}(\alpha,\beta)$$
 
$$Y = \sqrt{2X/\beta} \sim \text{Rayleigh}(1)$$
 
$$Y = \alpha - \gamma \log(X/\beta) \sim \text{Gumbel}(\alpha,\gamma)$$

6  $\mathbf{F}(m,n)$ 

$$F(m,n) = \frac{\chi_m^2/m}{\chi_n^2/n}$$

$$E(X) = \frac{n}{n-2} \quad (n > 2),$$

$$Var(X) = 2\left(\frac{n}{n-2}\right)^2 \frac{m+n-2}{m(n-4)} \quad (n > 4)$$

NOTE: 
$$[F(m,n)]^{-1} = F(n,m)$$
  
 $F_{1-\alpha}(m,n) = [F_{\alpha}(n,m)]^{-1}$   
 $F(1,k) = t^2(k)$   
If  $X \sim F(m,n), mX/(mX+n) \sim \text{Beta}(\frac{m}{2}, \frac{n}{2}).$ 

R:xgamma

$$7 \quad \begin{array}{l} \mathbf{Gamma} \left( \alpha, \theta \right) \\ f(x) = \frac{1}{\Gamma(\alpha) \theta^{\alpha}} x^{\alpha - 1} \exp(-x/\theta), \ \left( 0 < x < \infty \right) \\ M(t) = \left( 1 - \theta t \right)^{-\alpha} \ \left( t < 1/\theta \right) \\ E(X) = \alpha \theta, \ \mathrm{Var}(X) = \alpha \theta^2 \end{array}$$

Note: 
$$X \sim \text{Gamma}(n, \theta) = \text{Erlang}(n, \theta)$$
  
 $X = V_1 + \dots + V_n, \ (V_i \sim \text{Exponential}(\theta))$   
 $2X/\theta \sim \chi^2(2n)$ 

$$F_X(x) = 1 - \sum_{k=0}^{n-1} (\frac{x}{\theta})^k \frac{e^{-x/\theta}}{k!}$$
$$= \sum_{k=n}^{\infty} (\frac{x}{\theta})^k \frac{e^{-x/\theta}}{k!}$$

$$\begin{aligned} &\operatorname{Gamma}(n/2,2) = \chi^2(n), \ \Gamma(1/2) = \sqrt{\pi} \\ &X_i \sim \operatorname{Gamma}(\alpha_i,\theta) \Rightarrow \sum_{i=1}^r X_i \sim \operatorname{Gamma}(\sum_{i=1}^r \alpha_i,\theta) \\ &X/d \sim \operatorname{Gamma}(\alpha,\theta/d) \\ &E[X^c] = \Gamma(\alpha+c)\theta^c/\Gamma(\alpha) \ (c>-\alpha) \end{aligned}$$

8 Laplace 
$$(\mu, \sigma)$$
: Double Exponential 
$$f(x) = \frac{1}{2\sigma} e^{-|x-\mu|/\sigma}$$
 
$$F(x) = \begin{cases} \frac{1}{2} e^{-|x-\mu|/\sigma} & (x < \mu) \\ 1 - \frac{1}{2} e^{-|x-\mu|/\sigma} & (x \ge \mu) \end{cases}$$
 
$$M(t) = \frac{e^{\mu t}}{1 - (\sigma t)^2}, (|t| < \frac{1}{\sigma})$$
 
$$E(X) = \text{median}(X) = \mu, \text{Var}(X) = 2\sigma^2$$

R:xlogis

$$\begin{split} 9 \ & \operatorname{Logistic}_{\cdot}(\mu,\beta) \\ f(x) &= \frac{e^{-(x-\mu)/\beta}}{\beta[1+e^{-(x-\mu)/\beta}]^2} = \frac{1}{2\beta\left[1+\cosh\left(\frac{x-\alpha}{\beta}\right)\right]} \\ F(x) &= \frac{1}{1+e^{-(x-\mu)/\beta}} \\ M(t) &= e^{\mu t}\Gamma(1-\beta t)\Gamma(1+\beta t), \ |t| < \frac{1}{\beta} \end{split}$$

$$M(t) = e^{\mu t} \Gamma(1 - \beta t) \Gamma(1 + \beta t), |t| < \frac{1}{\beta}$$

$$E(X) = \operatorname{median}(X) = \mu, \operatorname{Var}(X) = \frac{\pi^2 \beta^2}{3}$$

$$\begin{split} &10 \ \ \underset{}{\text{Lognormal}} \left(\mu,\sigma^2\right) \\ & f(x) = \frac{1}{\sigma\sqrt{2\pi}x} \exp(-\frac{1}{2}(\frac{\log(x)-\mu}{\sigma})^2) \\ & E(X^k) = e^{k\mu+k^2\sigma^2/2}, \text{Var}(X) = e^{2(\mu+\sigma^2)} - e^{2\mu+\sigma^2} \end{split}$$

Note:  $F_{logN(\mu,\sigma^2)}(x) = F_{N(\mu,\sigma^2)}(\log x) = \Phi(\frac{\log x - \mu}{\sigma})$ Self-reproducing under multiplication and division

R:xnorm

$$\begin{array}{l} 11 \ \ \operatorname{Normal}\left(\mu,\sigma^2\right) \\ f(x) = \frac{1}{\sigma\sqrt{2\pi}} \exp(-\frac{1}{2}(\frac{x-\mu}{\sigma})^2) \\ M(t) = \exp(\mu t + \frac{1}{2}\sigma^2 t^2) \end{array}$$

$$M(t) = \exp(\mu t + \frac{1}{2}\sigma^2 t^2)$$

NOTE: If  $X \sim N(\mu, \sigma^2)$ ,  $Y = e^X \sim log N(\mu, \sigma^2)$ 

$$F_{N(0,1)}(x) = \frac{1}{2} + \frac{1}{2}\operatorname{sign}(x)F_{\Gamma(1/2,1)}(\frac{1}{2}x^2)$$
$$= \frac{1}{2} + \frac{1}{2}\operatorname{sign}(x)F_{\chi^2(1)}(x)$$

$$\phi'(z) = -z\phi(z), \ \phi''(z) = (z^2 - 1)\phi(z)$$
  
 
$$E(X^3) = \mu^3 + 3\mu\sigma^2, \ E(X^4) = \mu^4 + 6\mu^2\sigma^2 + 3\sigma^4$$

12 Rayleigh  $(\beta)$ 

R:xf

$$f(x) = \frac{x}{\beta^2} \exp\left(-\frac{x^2}{2\beta^2}\right)$$

$$F(x) = 1 - \exp\left(-\frac{x^2}{2\beta^2}\right)$$

$$E(X) = \beta\sqrt{\pi/2}, E(X^k) = (\sqrt{2}\beta)^k \frac{k}{2}\Gamma(\frac{k}{2})$$

$$\text{median}(X) = \beta\sqrt{2\ln 2}$$

$$\text{Var}(X) = (2 - \pi/2)\beta^2$$

NOTE: Rayleigh(
$$\beta$$
) = Weibull(2,  $2\beta^2$ )
$$cX \sim \text{Rayleigh}(c\beta)$$

$$(X/\beta)^2 \sim \chi^2(2)$$

$$X^2 \sim \text{Exponential}(2\beta^2) = \text{Gamma}(1, 2\beta^2)$$

$$\min_{1 \leq i \leq n} (X_i) \sim \text{Rayleigh}(\beta/\sqrt{n}): \text{ self-reproducing}$$

13 Slash  $(\alpha, \beta)$ 

$$f(x) = \frac{1}{\sqrt{2\pi}(x-\alpha)^2/\beta^2} \left( 1 - \exp\left(-\frac{1}{2}\left(\frac{x-\alpha}{\beta}\right)^2\right) \right)$$
  
$$F(x) = \Phi\left(\frac{x-\alpha}{\beta}\right) - \left(\frac{x-\alpha}{\beta}\right) f_{\text{Slash}(0,1)}\left(\frac{x-\alpha}{\beta}\right)$$

Note: 
$$X = \alpha + \beta \frac{Z}{U},$$
 where  $Z \sim N(0,1)$  and  $U \sim \text{Uniform}(0,1).$ 

14 Student t(k)

Find the expectation of the following states 
$$f(x) = \frac{\Gamma((k+1)/2)}{\Gamma(k/2)} \frac{1}{\sqrt{k\pi}} \left(1 + \frac{x^2}{k}\right)^{-(k+1)/2}, (k \ge 1)$$

$$E(X) = 0 \ (k > 1), \ Var(X) = k/(k-2) \ (k > 2)$$

Note: 
$$X \sim \frac{N(0,1)}{\sqrt{\chi_k^2/k}}, \quad F_{\alpha}(1,k) = t_{\alpha/2}(k)^2$$

$$F_{t(k)}(x) = 1 - \frac{1}{2} F_{\text{Beta}(k/2,1/2)}(\frac{k}{k+x^2}) \qquad (x \ge 0)$$
$$= \frac{1}{2} F_{\text{Beta}(k/2,1/2)}(\frac{k}{k+x^2}) \qquad (x < 0)$$

15 Uniform (a, b)

R:xt

$$f(x) = \frac{1}{b-a}$$

$$M(t) = \frac{e^{tb} - e^{ta}}{t(b-a)}, (t \neq 0)$$

$$E(X) = \text{median}(X) = \frac{a+b}{2}, \text{Var}(X) = \frac{(b-a)^2}{12}$$

Note:  $X \sim \text{Uniform}(0,1), -\log X \sim \text{Exponential}(1)$ 

16 Wald  $(\mu, \lambda)$ : Inverse Gaussian (IG) R:xinvgauss{statmod} 
$$\begin{split} f(x;\mu,\lambda) &= \sqrt{\frac{\lambda}{2\pi x^3}} \exp\left[-\lambda \frac{(x-\mu)^2}{2\mu^2 x}\right] = \sqrt{\frac{\lambda}{x^3}} \phi\left(\sqrt{\frac{\lambda}{x}} \frac{x-\mu}{\mu}\right) \\ E(X) &= \mu, \, \mathrm{Var}(X) = \mu^3/\lambda \end{split}$$

$$F(x) = \Phi\left(\sqrt{\frac{\lambda}{x}} \frac{x - \mu}{\mu}\right) + \exp\left(\frac{2\lambda}{\mu}\right) \Phi\left(-\sqrt{\frac{\lambda}{x}} \frac{x + \mu}{\mu}\right)$$
$$M(t) = \exp\left[\frac{\lambda}{\mu}\left(1 - \sqrt{1 - 2\mu^2 t/\lambda}\right)\right]$$

$$M(t) = \exp\left[\frac{\lambda}{\mu}\left(1 - \sqrt{1 - 2\mu^2 t/\lambda}\right)\right]$$

$$\hat{\mu}_{\text{mle}} = \bar{X}, \ \hat{\lambda}_{\text{mle}} = \left[\frac{1}{n} \sum \left\{ X_i^{-1} - \bar{X}^{-1} \right\} \right]^{-1}.$$

Note: 
$$\lambda(X - \mu)^2/(\mu^2 X) \sim \chi^2(1)$$
  
 $X_i \sim \mathrm{IG}(\mu, \lambda) \Rightarrow k X_i \sim \mathrm{IG}(k\mu, k\lambda),$   
 $\sum X_i \sim \mathrm{IG}(n\mu, n^2 \lambda), \ n\lambda/\hat{\lambda}_{\mathrm{mle}} \sim \chi^2(n-1)$ 

17 Weibull 
$$(\alpha, \theta)$$
 R: $x$ we 
$$f(x) = \frac{\alpha}{\theta} \left(\frac{x}{\theta}\right)^{\alpha - 1} \exp\left(-\left(x/\theta\right)^{\alpha}\right), \ (x \ge 0, \alpha > 0, \theta > 0)$$

$$F(x) = 1 - \exp\left(-\left(x/\theta\right)^{\alpha}\right)$$

$$E(X^k) = \theta^k \Gamma(1 + \frac{k}{\alpha}), \quad \text{median}(X) = \theta \left(\ln 2\right)^{1/\alpha}$$

$$\text{Var}(X) = \theta^2 \left[\Gamma(1 + \frac{2}{\alpha}) - \Gamma^2(1 + \frac{1}{\alpha})\right]$$

NOTE: 
$$\theta = 1 - e^{-1} \approx 63.2\%$$
 percentile  $Y = X^{\alpha} \sim \text{Exponential}(\theta^{\alpha})$  Weibull $(1, \theta) = \text{Exponential}(\theta)$  Weibull $(2, \sqrt{2}\theta) = \text{Rayleigh}(\theta)$  
$$\min_{1 \le i \le n} (X_i) \sim \text{Weibull}(\alpha, \theta/n^{1/\alpha}): \text{ self-reproducing}$$