Probability Distribution

1 **Binomial** (n, p), (x = 0, 1, ..., n)

R:xbinom

$$f(x) = \binom{n}{x} p^x (1-p)^{n-x}$$

$$M(t) = (pe^t + 1 - p)^n$$

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 $E(X) = np, Var(X) = np(1 - p)$

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

 $X_i \sim \text{Bin}(n_i, p) \Rightarrow \sum X_i \sim \text{Bin}(\sum n_i, p)$

R:xgeom

2 **Geometric** (p)

$$f(x) = p(1-p)^x, \quad (x = 0, 1, \dots)$$

$$F(t) = 1 - (1-p)^{\lfloor t \rfloor + 1}, \text{ where } \lfloor t \rfloor = \max\{m \in \mathbb{Z} : m \le t\}$$

$$S(t) = P[X \ge t] = (1-p)^{\lceil t \rceil} \text{ with } \lceil t \rceil = \min\{m \in \mathbb{Z} : m \ge t\}$$

$$R(t) = P[X > t] = (1-p)^{\lfloor t \rfloor + 1}$$

$$M_X(t) = \frac{p}{1 - (1-p)e^t}, \quad (t < -\log(1-p))$$

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

$$\begin{split} f(y) &= p(1-p)^{y-1}, \quad (y=1,2,\ldots) \\ F(t) &= 1 - (1-p)^{\lfloor t \rfloor} \\ S(t) &= P[Y \geq t] = (1-p)^{\lceil t \rceil - 1} \\ R(t) &= P[Y > t] = (1-p)^{\lfloor t \rfloor} \end{split}$$

$$M_Y(t) = \frac{pe^t}{1 - (1 - p)e^t}, \ (t < -\log(1 - p))$$

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

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

3 Hypergeometric (N, M, n)

R:xhyper

$$f(x) = \frac{\binom{M}{x} \binom{N - M}{n - x}}{\binom{N}{n}},$$

$$(\max(0, M - (N - n)) \le x \le \min(n, M))$$

$$E(X) = n\left(\frac{M}{N}\right), \operatorname{Var}(X) = n\frac{M}{N}\frac{(N - M)(N - n)}{N(N - 1)}$$

4 Negative Binomial (r, p)

R:xnbinom

$$f(x) = {r+x-1 \choose x} p^r (1-p)^x, (x = 0, 1, ...)$$

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

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

$$\frac{f(y) = {y-1 \choose r-1} p^r (1-p)^{y-r}, (y=r,r+1,\ldots)}{M_Y(t) = \left(\frac{pe^t}{1-(1-p)e^t}\right)^r}$$

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

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

5 Poisson (μ)

R:xpois

$$f(x) = \frac{e^{-\mu}\mu^x}{x!}, \quad (x = 0, 1, ...)$$

$$M(t) = \exp(\mu(e^t - 1))$$

$$E(X) = \mu, \operatorname{Var}(X) = \mu$$

Note: $X_i \sim \text{Poi}(\mu_i) \Rightarrow \sum X_i \sim \text{Poi}(\sum \mu_i)$

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

= $F_{\chi^2_{2n}}(2\mu)$

1 Beta (α, β) , $(\alpha > 0, \beta > 0)$

R:xbeta

$$f(x) = \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha - 1} (1 - x)^{\beta - 1}, \ (0 < x < 1)$$

$$M(t) = 1 + \sum_{k=1}^{\infty} \left(\prod_{r=1}^{k-1} \frac{\alpha + r}{\alpha + \beta + r} \right) \frac{t^k}{k}$$

$$M(t) = 1 + \sum_{k=1}^{\infty} \left(\prod_{r=1}^{k-1} \frac{\alpha+r}{\alpha+\beta+r} \right) \frac{t^k}{k!}$$

$$E(X) = \frac{\alpha}{\alpha+\beta}, \quad \text{Var}(X) = \frac{\alpha\beta}{(\alpha+\beta+1)(\alpha+\beta)^2}$$

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

$$f(t) = \frac{1}{2\alpha\beta} \sqrt{\frac{\beta}{t}} \left(1 + \frac{\beta}{t} \right) \phi \left[\frac{1}{\alpha} \left(\sqrt{\frac{t}{\beta}} - \sqrt{\frac{\beta}{t}} \right) \right], \ (t > 0)$$

$$F(t) = \Phi\left[\frac{1}{\alpha}\left(\sqrt{\frac{t}{\beta}} - \sqrt{\frac{\beta}{t}}\right)\right]$$

$$F^{-1}(p) = \frac{1}{4} \left[\alpha \sqrt{\beta} \Phi^{-1}(p) + \sqrt{\alpha^2 \beta \{\Phi^{-1}(p)\}^2 + 4\beta} \right]^2$$
$$= \beta \left\{ 1 + \gamma(p)^2 + \gamma(p) \sqrt{\gamma(p)^2 + 2} \right\},$$
where $\gamma(p) = \alpha \Phi^{-1}(p) / \sqrt{2}$

$$E(T) = \beta(1 + \frac{1}{2}\alpha^2), \quad Var(T) = (\alpha\beta)^2(1 + \frac{5}{4}\alpha^2)$$

NOTE:
$$\operatorname{median}(T) = \beta, cT \sim \operatorname{BS}(\alpha, c\beta), T^{-1} \sim \operatorname{BS}(\alpha, \beta^{-1})$$

$$X = \left(\sqrt{\frac{T}{\beta}} - \sqrt{\frac{\beta}{T}}\right) \sim N(0, \alpha^2)$$

$$\log T \sim \operatorname{sinh-Normal}(\log \beta, \alpha)$$

3 Cauchy (α, β)

R:xcauchy

R:xchisq

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

$$F(x) = \frac{1}{\pi} \left[\arctan\left(\frac{x - \alpha}{\beta}\right) + \frac{\pi}{2} \right]$$

$$\phi(t) = \exp(it\alpha - \beta|t|)$$

NOTE: Cauchy
$$(0, 1) = t(1)$$

$$X, Y \sim N(0, 1) \Rightarrow X/Y \sim \text{Cauchy}(0, 1)$$

$$cX \sim \text{Cauchy}(c\alpha, c\beta)$$

$$X_i \sim \text{Cauchy}(\alpha, \beta) \Rightarrow \frac{1}{n} \sum X_i \sim \text{Cauchy}(\alpha, \beta)$$

4 Chi Squared (n)

$$f(x) = \frac{1}{\Gamma(n/2)2^{n/2}} x^{n/2 - 1} e^{-x/2}, (x \ge 0)$$

$$\Gamma(n/2)2^{n/2}$$

$$E(X) = n, \operatorname{Var}(X) = 2n$$

$$M(t) = \left(\frac{1}{1-2t}\right)^{-1}, \quad (t < \frac{1}{2})$$

$$M(t) = \left(\frac{1}{1-2t}\right)^{n/2}, \quad (t < \frac{1}{2})$$

$$E(X^m) = 2^n \Gamma(m + n/2) / \Gamma(n/2)$$

Note:
$$\chi^2(n) = \text{Gamma}(n/2, 2)$$

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

$$F_{\chi^2(2)}(x) = 1 - \sqrt{2\pi}\phi(\sqrt{x})$$

$$F_{\chi^2(3)}(x) = 1 - \sqrt{2\pi}\phi(\sqrt{x})$$

 $F_{\chi^2(3)}(x) = 2\Phi(\sqrt{x}) - 1 - 2\sqrt{x}\phi(\sqrt{x})$

5 Exponential (θ)

R:xexp

$$f(x) = \frac{1}{\theta} e^{-x/\theta} \ (x \ge 0, \ \theta > 0)$$

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

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

$$M(t) = \frac{1}{1 - \theta t}, (t < \frac{1}{\theta})$$

$$E(X) = \theta, Var(X) = \theta^{2}$$

$$cX \sim \text{Exponential}(c\theta)$$

$$\sum X_i \sim \text{Gamma}(n,\theta)$$

$$\min_{1 \le i \le n} (X_i) \sim \text{Exponential}(\theta/n): \text{ self-reproducing}$$

$$Y = X^{1/\alpha} \sim \text{Weibull}(\alpha, \theta)$$

$$Y = \sqrt{2X/\theta} \sim \text{Rayleigh}(1)$$

$$Y = \alpha - \gamma \log(X/\theta) \sim \text{Gumbel}(\alpha, \gamma)$$

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

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

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

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

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

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 $X \sim F(m,n), mX/(mX+n) \sim \text{Beta}(\frac{m}{2}, \frac{n}{2}).$

7 Gamma
$$(\alpha, \theta)$$

$$f(x) = \frac{1}{\Gamma(\alpha)\theta^{\alpha}} x^{\alpha-1} \exp(-x/\theta), \ (0 < x < \infty)$$

$$M(t) = (1 - \theta t)^{-\alpha} \ (t < 1/\theta)$$

$$M(t) = (1 - \theta t) \quad (t < 1/\theta)$$

$$E(X) = \alpha \theta, \text{Var}(X) = \alpha \theta^2$$

Note:
$$X \sim \text{Gamma}(n, \theta) = \text{Erlang}(n, \theta)$$

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

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

$$\operatorname{Gamma}(n/2, 2) = \chi^{2}(n), \ \Gamma(1/2) = \sqrt{\pi}$$

$$X_{i} \sim \operatorname{Gamma}(\alpha_{i}, \theta) \Rightarrow \sum_{i=1}^{r} X_{i} \sim \operatorname{Gamma}(\sum_{i=1}^{r} \alpha_{i}, \theta)$$

$$X/d \sim \operatorname{Gamma}(\alpha, \theta/d)$$

$$E[X^{c}] = \Gamma(\alpha + c)\theta^{c}/\Gamma(\alpha) \ (c > -\alpha)$$

8 Laplace
$$(\mu, \sigma)$$
: Double Exponential
$$f(x) = \frac{1}{2\sigma} e^{-|x-\mu|/\sigma}$$

$$F(x) = \begin{cases} \frac{1}{2} e^{-|x-\mu|/\sigma} & (x < \mu) \\ 1 - \frac{1}{2} e^{-|x-\mu|/\sigma} & (x \ge \mu) \end{cases}$$

$$M(t) = \frac{e^{\mu t}}{1 - (\sigma t)^2}, (|t| < \frac{1}{\sigma})$$

$$E(X) = \text{median}(X) = \mu, \text{Var}(X) = 2\sigma^2$$

9
$$\begin{aligned} & \textbf{Logistic}, (\mu, \beta) \\ & f(x) = \frac{e^{-(x-\mu)/\beta}}{\beta[1 + e^{-(x-\mu)/\beta}]^2} = \frac{1}{2\beta\left[1 + \cosh\left(\frac{x-\alpha}{\beta}\right)\right]} \end{aligned} \end{aligned}$$
 R: $x \log i s$

$$F(x) = \frac{1}{1 + e^{-(x-\mu)/\beta}}$$

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

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

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

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

11 Normal
$$(\mu, \sigma^2)$$
 R: x norm $f(x) = \frac{1}{\sigma\sqrt{2\pi}} \exp(-\frac{1}{2}(\frac{x-\mu}{\sigma})^2)$

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

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 (β)

R:xf

R:xgamma

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

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

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

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

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

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 (α, β)

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

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

14 Student t(k)

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

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

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

$$E(X) = \mu, \operatorname{Var}(X) = \mu^3/\lambda$$

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

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

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

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

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 wei $f(x) = \frac{\alpha}{\theta} \left(\frac{x}{\theta}\right)^{\alpha-1} \exp\left(-(x/\theta)^{\alpha}\right), (x \ge 0, \alpha > 0, \theta > 0)$

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

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

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

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