Probability Distributions

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

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

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

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

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

Note: $n \to \infty$ with $\mu = np \Rightarrow \text{Poisson}(\mu)$

$$F_{\operatorname{Bin}(n,p)}(x) = F_{\operatorname{Beta}(n-\lfloor x\rfloor,\lfloor x\rfloor+1)}(1-p)$$

$$F_{\operatorname{Bin}(n,p)}(r-1) = 1 - F_{\operatorname{NegBin}(r,p)}(n-r)$$

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

2 **Geometric** (p)

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

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

$$F(t) = 1 - (1 - p)^{\lfloor t \rfloor}$$
, where $\lfloor t \rfloor = \max\{m \in \mathbb{Z} : m \le t\}$
 $S(t) = P[Y > t] - (1 - p)^{\lfloor t \rfloor}$ with $\lfloor t \rfloor = \min\{m \in \mathbb{Z} : m > t\}$

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

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

$$R(t) = P[X \ge t] = (1 - p) \text{ with } |t| = \min$$

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

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

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

$$f(y) = p(1-p)^{y-1}, \quad (y = 1, 2, ...)$$

 $F(t) = 1 - (1-p)^{\lfloor t \rfloor}$

$$F(t) = 1 - (1 - n)^{\lfloor t \rfloor}$$

$$S(t) = P[Y \ge t] = (1-p)^{\lceil t \rceil - 1}$$

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

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

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

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

3 Hypergeometric (N, M, n)

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

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

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

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

4 Negative Binomial (r, p)

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

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

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

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

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

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

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

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

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

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

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

$$(1-x) = \sum_{k=0}^{\infty} {n+k \choose k} x^k (|x| < u = r(1-p), r \to \infty \Rightarrow \text{Poisson}(u)$$

$$\mu = r(1-p), r \to \infty \Rightarrow \text{Poisson}(\mu)$$

 $X_i \sim \text{NB}(r_i, p) \Rightarrow \sum X_i \sim \text{NB}(\sum r_i, p)$

$$F_{NB(r,p)}(x) = F_{Beta(r,\lfloor x\rfloor+1)}(p)$$

$$F$$
 (r , p) (w) Beta(r , $\lfloor x \rfloor + 1$) (P)

$$F_{NB(r,p)}(n-r) = 1 - F_{Bin(n,p)}(r-1)$$

5 Poisson (μ)

$$-\mu$$

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

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

$$E(X) = \mu$$
, $Var(X) = \mu$

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

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

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

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

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

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

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

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

NOTE:
$$f_{\text{Beta}(\alpha,\beta)}(x) = f_{\text{Beta}(\beta,\alpha)}(1-x)$$

$$F_{\text{Beta}(\alpha,\beta)}(x) = 1 - F_{\text{Beta}(\beta,\alpha)}(1-x)$$

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

$$\begin{split} X &= \left(\sqrt{\frac{T}{\beta}} - \sqrt{\frac{\beta}{T}}\right) \sim N(0, \alpha^2) \\ \log T &\sim \text{sinh-Normal}(\log \beta, \alpha) \end{split}$$

3 Cauchy (α, β)

R:xchisq

R:xexp

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

Cauchy
$$(\alpha, \beta)$$

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

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

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

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

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 $\binom{n}{1}$

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

$$E(X) = n$$
, $Var(X) = 2n$

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

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

$$E(X^{-}) = 2^{-1} (m + n/2)/1 (n/2)$$

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

$$\chi^2(2) = \text{Exponential}(\beta = 2)$$

$$X_i \sim \chi^2(n_i) \Rightarrow \sum X_i \sim \chi^2(\sum n_i)$$

$$X_i \sim \chi^2(n_i) \Rightarrow \sum X_i \sim \chi^2(\sum n_i)$$

$$X_i \sim N(0, 1) \Rightarrow \sum_{i=1}^n X_i^2 \sim \chi^2(n)$$

$$F_{\chi^{2}(p+2)}(x) = F_{\chi^{2}(p)}(x) - 2xf_{\chi^{2}(p)}(x)/p$$

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

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

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

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

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

$$F_{\chi^2(3)}(x) \equiv 2\Psi(\sqrt{x}) - 1 - 2\sqrt{x}\varphi(\sqrt{x})$$

5 Exponential (θ)

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

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

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

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

$$E(X) = \theta, Var(X) = \theta$$

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

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

$$\min_{1 \le i \le n} (X_i) \sim \text{Exponential}(\theta/n)$$
: 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 $\mathbf{F}(m,n)$

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

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

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

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

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
$$(\alpha, \theta)$$

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

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

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

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

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

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

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

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

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

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

$$\frac{H'(t) - 1 - (\sigma t)^2, (|t| < \sigma)}{1 - (\sigma t)^2, (|t| < \sigma)}$$

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

9 Logistic,
$$(\mu, \beta)$$

$$f(x) = \frac{e^{-(x-\mu)/\beta}}{\beta[1 + e^{-(x-\mu)/\beta}]^2} = \frac{1}{2\beta \left[1 + \cosh\left(\frac{x-\alpha}{\beta}\right)\right]}$$

$$\beta[1 + e^{-(x-\mu)/\beta}]^2 = 2\beta \left[1 + \cosh\left(\frac{x-\beta}{\beta}\right)\right]$$

$$E(x) = \frac{1}{\beta}$$

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

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

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

R:xnorm

R:xlogis

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

$$E(X^k) = e^{k\mu + k^2\sigma^2/2}, Var(X) = e^{2(\mu + \sigma^2)} - e^{2\mu + \sigma^2}$$

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
$$(\mu, \sigma^2)$$

$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} \exp(-\frac{1}{2}(\frac{x-\mu}{\sigma})^2)$$

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

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

$$\begin{split} F_{N(0,1)}(x) &= \frac{1}{2} + \frac{1}{2} \mathrm{sign}(x) F_{\Gamma(1/2,1)}(\frac{1}{2}x^2) \\ &= \frac{1}{2} + \frac{1}{2} \mathrm{sign}(x) F_{\chi^2(1)}(x) \end{split}$$

$$\phi'(z) = -z\phi(z), \, \phi''(z) = (z^2 - 1)\phi(z)$$

$$E(X^3) = \mu^3 + 3\mu\sigma^2, \, E(X^4) = \mu^4 + 6\mu^2\sigma^2 + 3\sigma^4$$

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)

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

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 (μ, λ) : 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:
$$\lambda(X - \mu)^2/(\mu^2 X) \sim \chi^2(1)$$

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

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

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

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

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