

Discrete and continuous univariate distributions

$X$ (r.v.)	Values	$P(X = x)$ or $f_X(x)$	$E(X)$	$V(X)$	$M_X(t)$ or $E(X^k)$	$P_X(s)$
Binomial( $n, p$ )	$\{0, \dots, n\}$	$\binom{n}{x} p^x (1-p)^{n-x}$	$np$	$np(1-p)$	$[pe^t + (1-p)]^n$	$(1-p+ps)^n$
HyperG( $N, M, n$ )	$\{\max\{0, n-N+M\}, \dots, \min\{n, M\}\}$	$\frac{\binom{M}{x} \binom{N-M}{n-x}}{\binom{N}{n}}$	$n \frac{M}{N}$	$n \frac{M}{N} \frac{N-M}{N-1}$	not interesting	not interesting
Geometric( $p$ )	$\mathbb{N}$	$(1-p)^{x-1} p$	$\frac{1}{p}$	$\frac{1-p}{p^2}$	$\frac{pe^t}{1-(1-p)e^t}$	$\frac{ps}{1-(1-p)s}$
Geometric <sup>*</sup> ( $p$ )	$\mathbb{N}_0$	$(1-p)^x p$	$\frac{1-p}{p}$	$\frac{1-p}{p^2}$	$\frac{pe^t}{1-(1-p)e^t}$	$\frac{ps}{1-(1-p)s}$
NegativeBin( $r, p$ )	$\{r, r+1, \dots\}$	$\binom{x-1}{r-1} p^r (1-p)^{x-r}$	$\frac{r}{p}$	$\frac{r(1-p)}{p^2}$	$\left[ \frac{pe^t}{1-(1-p)e^t} \right]^r$	$\left[ \frac{ps}{1-(1-p)s} \right]^r$
NegativeBin <sup>*</sup> ( $r, p$ )	$\{0, 1, \dots\}$	$\binom{y+r-1}{r-1} p^r (1-p)^y$	$\frac{r(1-p)}{p}$	$\frac{r(1-p)}{p^2}$	$\left[ \frac{pe^t}{1-(1-p)e^t} \right]^r$	$\left[ \frac{ps}{1-(1-p)s} \right]^r$
Poisson( $\lambda$ )	$\mathbb{N}_0$	$e^{-\lambda} \frac{\lambda^x}{x!}$	$\lambda$	$\lambda$	$e^{\lambda(e^t-1)}$	$e^{-\lambda(1-s)}$
Uniform ( $\{1, \dots, n\}$ )	$\{1, \dots, n\}$	$\frac{1}{n}$	$\frac{n+1}{2}$	$\frac{n^2-1}{12}$	$\frac{e^t(1-e^{tn})}{n(1-e^t)}$	$\frac{s(1-s^n)}{n(1-s)}$
Beta( $\alpha, \beta$ )	$[0, 1]$	$\frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha-1} (1-x)^{\beta-1}$	$\frac{\alpha}{\alpha+\beta}$	$\frac{\alpha\beta}{(\alpha+\beta)^2(\alpha+\beta+1)}$	$1 + \sum_{k=1}^{+\infty} \left( \prod_{r=0}^{k-1} \frac{\alpha+r}{\alpha+\beta+r} \right) \frac{t^k}{k!}$	—
Cauchy( $\mu, \sigma$ )	$\mathbb{R}$	$\frac{1}{\pi\sigma} \frac{1}{1+(\frac{x-\mu}{\sigma})^2}$	nonexistent	nonexistent	nonexistent	—
$\chi_{(n)}^2$	$\mathbb{R}_0^+$	$\frac{(\frac{n}{2})^{\frac{n}{2}}}{\Gamma(\frac{n}{2})} x^{\frac{n}{2}-1} e^{-\frac{x}{2}}$	$n$	$2n$	$\left( \frac{1-t}{2} \right)^{\frac{n}{2}}, t < \frac{1}{2}$	—
Exponential( $\lambda$ )	$\mathbb{R}_0^+$	$\lambda e^{-\lambda x}$	$\frac{1}{\lambda}$	$\frac{1}{\lambda^2}$	$\frac{\lambda}{\lambda-t}, t < \lambda$	—
Gamma( $\alpha, \lambda$ )	$\mathbb{R}_0^+$	$\frac{\lambda^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{-\lambda x}$	$\frac{\alpha}{\lambda}$	$\frac{\alpha}{\lambda^2}$	$\left( \frac{\lambda}{\lambda-t} \right)^\alpha, t < \lambda$	—
LogNormal( $\mu, \sigma^2$ )	$\mathbb{R}^+$	$\frac{1}{\sqrt{2\pi\sigma x}} e^{-\frac{(\log x - \mu)^2}{2\sigma^2}}$	$e^{\mu+\frac{\sigma^2}{2}}$	$(e^{\sigma^2}-1)e^{2\mu+\sigma^2}$	$E(X^k) = e^{k\mu+\frac{k^2\sigma^2}{2}}$	—
Normal( $\mu, \sigma^2$ )	$\mathbb{R}$	$\frac{1}{\sqrt{2\pi\sigma}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$	$\mu$	$\sigma^2$	$e^{\mu t + \frac{(t\sigma)^2}{2}}$	—
Rayleigh( $\sigma$ )	$\mathbb{R}_0^+$	$\frac{x}{\sigma^2} e^{-\frac{x^2}{2\sigma^2}}$	$\sigma\sqrt{\frac{\pi}{2}}$	$\frac{4-\pi}{2}\sigma^2$	$E(X^k) = (\sqrt{2}\sigma)^k \Gamma(1+\frac{k}{2})$	—
Uniform( $a, b$ )	$[a, b]$	$\frac{1}{b-a}$	$\frac{a+b}{2}$	$\frac{(b-a)^2}{12}$	$\frac{e^{bt}-e^{at}}{(b-a)t}, t \neq 0$	—
Weibull( $\alpha, \beta$ )	$\mathbb{R}_0^+$	$\frac{\beta}{\alpha} \left(\frac{x}{\alpha}\right)^{\beta-1} e^{-\left(\frac{x}{\alpha}\right)^\beta}$	$\alpha\Gamma\left(1+\frac{1}{\beta}\right)$	$\alpha^2 \left[ \Gamma\left(1+\frac{2}{\beta}\right) - \Gamma\left(1+\frac{1}{\beta}\right)^2 \right]$	$E(X^k) = \alpha^k \Gamma\left(1+\frac{k}{\beta}\right)$	—

$$\Gamma(\alpha) = \int_0^{+\infty} x^{\alpha-1} e^{-x} dx, \alpha > 0; \quad \Gamma(n) = (n-1)!, n \in \mathbb{N}; \quad \Gamma(\alpha+1) = \alpha\Gamma(\alpha), \alpha > 0; \quad \Gamma\left(\frac{1}{2}\right) = \sqrt{\pi}$$

## Relating c.d.f.

$$F_{NegativeBin(r,p)}(x) = 1 - F_{Binomial(x,p)}(r-1)$$

$$F_{Erlang(n,\lambda)}(x) = 1 - F_{Poisson(\lambda x)}(n-1)$$

$$F_{Gamma(\alpha,\beta)}(x) = F_{\chi^2_{(2\alpha)}}(2\beta x)$$

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

## Moment/probability generating function; moments

$$M_X(t) = E(e^{tX})$$

$$E(X^k) = \left. \frac{d^k M_X(t)}{dt^k} \right|_{t=0}$$

$$P_X(s) = E(s^X); \quad P(X=k) = \left. \frac{1}{k!} \times \frac{d^k P_X(s)}{ds^k} \right|_{s=0}$$

$$E[X(X-1)\cdots(X-k+1)] = \left. \frac{d^k P_X(s)}{ds^k} \right|_{s=1}, \quad k \in \mathbb{N}$$

$$E(X) = \int_0^{+\infty} [1 - F_X(x)] dx, \quad \text{for } X \geq 0$$

$$E(X^k) = \int_0^{+\infty} kx^{k-1} [1 - F_X(x)] dx, \quad \text{for } X \geq 0$$

$$SC(X) = \frac{E\{[X-E(X)]^3\}}{[SD(X)]^3}$$

$$KC(X) = \frac{E\{[X-E(X)]^4\}}{[SD(X)]^4} - 3$$

## Multinomial distribution

$$P(N_1 = n_1, \dots, N_d = n_d) = \frac{n!}{\prod_{i=1}^d n_i!} \times \prod_{i=1}^d p_i^{n_i}$$

$$\{(n_1, \dots, n_d) \in \mathbb{N}_0^d : \sum_{i=1}^d n_i = n\}$$

$$M_{N_1, \dots, N_{d-1}}(t_1, \dots, t_{d-1}) = \left[ \left( \sum_{i=1}^{d-1} p_i e^{t_i} \right) + p_d \right]^n$$

$$N_i \sim \text{Binomial}(n, p_i); \quad \text{Cov}(N_i, N_j) = -n p_i p_j, \quad i \neq j$$

$$M_{\underline{X}}(\underline{t}) = E[\exp(\sum_{i=1}^n t_i X_i)]$$

$$E\left(\prod_{i=1}^n X_i^{k_i}\right) = \left. \frac{\partial^{\sum_{i=1}^n k_i} M_{\underline{X}}(\underline{t})}{\partial t_1^{k_1} \dots \partial t_n^{k_n}} \right|_{\underline{t}=0}$$

## Functions of r.v.

$$F_{Y=g(X)}(y) = P[X \in g^{-1}((-\infty, y])]$$

$$f_{Y=g(X)}(y) = f_X[g^{-1}(y)] \times \left| \frac{dg^{-1}(y)}{dy} \right|$$

## Hierarchical models resulting from mixtures

$$P(X=x) = \sum_y P(X=x|Y=y) \times P(Y=y)$$

$$P(X=x) = \int_{\mathbb{R}_Y} P(X=x|Y=y) \times f_Y(y) dy$$

$$E[g(X)] = E\{E[g(X)|Y]\}$$

$$V[g(X)] = V\{E[g(X)|Y]\} + E\{V[g(X)|Y]\}$$

## Functions of random vectors

$$F_{\underline{Y}=g(\underline{X})}(\underline{y}) = P[\underline{X} \in \underline{g}^{-1}(\prod_{i=1}^m (-\infty, y_i])]$$

$$f_{\underline{Y}=g(\underline{X})}(\underline{y}) = f_{\underline{X}}[\underline{g}^{-1}(\underline{y})] \times |J(\underline{y})|$$

$$J(\underline{y}) = \begin{vmatrix} \frac{\partial g_1^{-1}(\underline{y})}{\partial y_1} & \dots & \frac{\partial g_1^{-1}(\underline{y})}{\partial y_n} \\ \vdots & \dots & \vdots \\ \frac{\partial g_n^{-1}(\underline{y})}{\partial y_1} & \dots & \frac{\partial g_n^{-1}(\underline{y})}{\partial y_n} \end{vmatrix}$$

$$f_{X+Y}(z) = \int_{-\infty}^{+\infty} f_{X,Y}(z-y, y) dy$$

$$f_{X-Y}(u) = \int_{-\infty}^{+\infty} f_{X,Y}(u+y, y) dy$$

$$f_{XY}(v) = \int_{-\infty}^{+\infty} f_{X,Y}(v/y, y) \times \frac{1}{|y|} dy$$

$$f_{X/Y}(w) = \int_{-\infty}^{+\infty} f_{X,Y}(wy, y) \times |y| dy$$

## Order statistics

$$P[X_{(n-k+1)} > x] = 1 - F_{Binomial(n,1-F_X(x))}(k-1)$$

$$f_{X_{(1), \dots, X_{(n)}}}(x_{(1)}, \dots, x_{(n)}) = n! \times \prod_{i=1}^n f_X(x_{(i)})$$

$$F_{X_{(i)}}(x) = 1 - F_{Binomial(n, F_X(x))}(i-1)$$

$$f_{X_{(i)}}(x) = \frac{n!}{(i-1)!(n-i)!} [F_X(x)]^{i-1} [1 - F_X(x)]^{n-i} f_X(x)$$

$$f_{(X_{(i)}, X_{(j)})}(x, y) = \frac{n!}{(i-1)!(j-i-1)!(n-j)!} [F_X(x)]^{i-1} [F_X(y) - F_X(x)]^{j-i-1} [1 - F_X(y)]^{n-j} f_X(x) f_X(y), \quad x < y$$