

## Discrete Distributions

	Bernoulli with parameter $p$	Binomial with parameters $n$ and $p$
<b>p.f.</b>	$f(x) = p^x(1-p)^{1-x},$ for $x = 0, 1$	$f(x) = \binom{n}{x} p^x(1-p)^{n-x},$ for $x = 0, \dots, n$
<b>Mean</b>	$p$	$np$
<b>Variance</b>	$p(1-p)$	$np(1-p)$
<b>m.g.f.</b>	$\psi(t) = pe^t + 1 - p$	$\psi(t) = (pe^t + 1 - p)^n$

	Uniform on the integers $a, \dots, b$	Hypergeometric with parameters $A, B$ , and $n$
<b>p.f.</b>	$f(x) = \frac{1}{b-a+1},$ for $x = a, \dots, b$	$f(x) = \frac{\binom{A}{x}\binom{B}{n-x}}{\binom{A+B}{n}},$ for $x = \max\{0, n-b\}, \dots, \min\{n, A\}$
<b>Mean</b>	$\frac{b+a}{2}$	$\frac{nA}{A+B}$
<b>Variance</b>	$\frac{(b-a)(b-a+1)}{12}$	$\frac{nAB}{(A+B)^2} \frac{A+B-n}{A+B-1}$
<b>m.g.f.</b>	$\psi(t) = \frac{e^{(b+1)t} - e^{at}}{(e^t - 1)(b-a+1)}$	Nothing simpler than $\psi(t) = \sum_x f(x)e^{tx}$

	Geometric with parameter $p$	Negative binomial with parameters $r$ and $p$
<b>p.f.</b>	$f(x) = p(1-p)^x,$ for $x = 0, 1, \dots$	$f(x) = \binom{r+x-1}{x} p^r(1-p)^x,$ for $x = 0, 1, \dots$
<b>Mean</b>	$\frac{1-p}{p}$	$\frac{r(1-p)}{p}$
<b>Variance</b>	$\frac{1-p}{p^2}$	$\frac{r(1-p)}{p^2}$
<b>m.g.f.</b>	$\psi(t) = \frac{p}{1-(1-p)e^t},$ for $t < \log(1/[1-p])$	$\psi(t) = \left( \frac{p}{1-(1-p)e^t} \right)^r,$ for $t < \log(1/[1-p])$

	Poisson with mean $\lambda$	Multinomial with parameters $n$ and $(p_1, \dots, p_k)$
<b>p.f.</b>	$f(x) = e^{-\lambda} \frac{\lambda^x}{x!},$ for $x = 0, 1, \dots$	$f(x_1, \dots, x_k) = \binom{n}{x_1, \dots, x_k} p_1^{x_1} \cdots p_k^{x_k},$ for $x_1 + \dots + x_k = n$ and all $x_i \geq 0$
<b>Mean</b>	$\lambda$	$E(X_i) = np_i,$ for $i = 1, \dots, k$
<b>Variance</b>	$\lambda$	$\text{Var}(X_i) = np_i(1-p_i), \text{Cov}(X_i, X_j) = -np_i p_j,$ for $i, j = 1, \dots, k$
<b>m.g.f.</b>	$\psi(t) = e^{\lambda(e^t - 1)}$	Multivariate m.g.f. can be defined, but is not defined in this text.

# Continuous Distributions

	Beta with parameters $\alpha$ and $\beta$	Uniform on the interval $[a, b]$
<b>p.d.f.</b>	$f(x) = \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha-1} (1-x)^{\beta-1},$ for $0 < x < 1$	$f(x) = \frac{1}{b-a},$ for $a < x < b$
<b>Mean</b>	$\frac{\alpha}{\alpha+\beta}$	$\frac{a+b}{2}$
<b>Variance</b>	$\frac{\alpha\beta}{(\alpha+\beta)^2(\alpha+\beta+1)}$	$\frac{(b-a)^2}{12}$
<b>m.g.f.</b>	Not available in simple form	$\psi(t) = \frac{e^{-at}-e^{-bt}}{t(b-a)}$

	Exponential with parameter $\beta$	Gamma with parameters $\alpha$ and $\beta$
<b>p.d.f.</b>	$f(x) = \beta e^{-\beta x},$ for $x > 0$	$f(x) = \frac{\beta^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{-\beta x},$ for $x > 0$
<b>Mean</b>	$\frac{1}{\beta}$	$\frac{\alpha}{\beta}$
<b>Variance</b>	$\frac{1}{\beta^2}$	$\frac{\alpha}{\beta^2}$
<b>m.g.f.</b>	$\psi(t) = \frac{\beta}{\beta-t},$ for $t < \beta$	$\psi(t) = \left(\frac{\beta}{\beta-t}\right)^\alpha,$ for $t < \beta$

	Normal with mean $\mu$ and variance $\sigma^2$	Bivariate normal with means $\mu_1$ and $\mu_2$ , variances $\sigma_1^2$ and $\sigma_2^2$ , and correlation $\rho$
<b>p.d.f.</b>	$f(x) = \frac{1}{(2\pi)^{1/2}\sigma} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right)$	Formula is too large to print here. See Eq. (5.10.2) on page 338.
<b>Mean</b>	$\mu$	$E(X_i) = \mu_i,$ for $i = 1, 2$
<b>Variance</b>	$\sigma^2$	Covariance matrix: $\begin{pmatrix} \sigma_1^2 & \rho\sigma_1\sigma_2 \\ \rho\sigma_1\sigma_2 & \sigma_2^2 \end{pmatrix}$
<b>m.g.f.</b>	$\psi(t) = \exp\left(\mu t + \frac{t^2\sigma^2}{2}\right)$	Bivariate m.g.f. can be defined, but is not defined in this text.

Continuous Distributions

	Lognormal with parameters $\mu$ and $\sigma^2$	$F$ with $m$ and $n$ degrees of freedom
<b>p.d.f.</b>	$f(x) = \frac{1}{(2\pi)^{1/2}\sigma x} \exp\left(-\frac{(\log(x)-\mu)^2}{2\sigma^2}\right),$ for $x > 0$	$f(x) = \frac{\Gamma\left[\frac{1}{2}(m+n)\right]m^{m/2}n^{n/2}}{\Gamma\left(\frac{1}{2}m\right)\Gamma\left(\frac{1}{2}n\right)} \cdot \frac{x^{(m/2)-1}}{(mx+n)^{(m+n)/2}},$ for $x > 0$
<b>Mean</b>	$e^{\mu+\sigma^2/2}$	$\frac{n}{n-2},$ if $n > 2$
<b>Variance</b>	$e^{2\mu+2\sigma^2}[e^{\sigma^2} - 1]$	$\frac{2n^2(m+n-2)}{m(n-2)^2(n-4)},$ if $n > 4$
<b>m.g.f.</b>	Not finite for $t > 0$	Not finite for $t > 0$

  

	$t$ with $m$ degrees of freedom	$\chi^2$ with $m$ degrees of freedom
<b>p.d.f.</b>	$f(x) = \frac{\Gamma\left(\frac{m+1}{2}\right)}{(m\pi)^{1/2}\Gamma\left(\frac{m}{2}\right)} \left(1 + \frac{x^2}{m}\right)^{-(m+1)/2}$	$f(x) = \frac{1}{2^{m/2}\Gamma(m/2)} x^{(m/2)-1} e^{-x/2},$ for $x > 0$
<b>Mean</b>	0, if $m > 1$	$m$
<b>Variance</b>	$\frac{m}{m-2},$ if $m > 2$	$2m$
<b>m.g.f.</b>	Not finite for $t \neq 0$	$\psi(t) = (1 - 2t)^{-m/2},$ for $t < 1/2$

  

	Cauchy centered at $\mu$	Pareto with parameters $x_0$ and $\alpha_0$
<b>p.d.f.</b>	$f(x) = \frac{1}{\pi(1+[x-\mu]^2)}$	$f(x) = \frac{\alpha x_0^\alpha}{x^{\alpha+1}},$ for $x > x_0$
<b>Mean</b>	Does not exist	$\frac{\alpha x_0}{\alpha-1},$ if $\alpha > 1$
<b>Variance</b>	Does not exist	$\frac{\alpha x_0^2}{(\alpha-1)^2(\alpha-2)},$ if $\alpha > 2$
<b>m.g.f.</b>	Not finite for $t \neq 0$	Not finite for $t > 0$