

Table 1: Formulae Sheet

$\bar{x} = \sum_{i=1}^n \frac{x_i}{n} = \frac{x_1+x_2+\dots+x_n}{n}$
$\sigma^2 = \sum_{i=1}^n \frac{(x_i-\bar{x})^2}{n-1}, \sigma = \sqrt{\sigma^2}$
${}_n P_r = n(n-1)\dots(n-r+1) = \frac{n!}{(n-r)!}$
$\binom{n}{r} = \frac{n!}{r!(n-r)!}$
$\binom{n}{n_1, n_2, \dots, n_r} = \frac{n!}{n_1!n_2!\dots n_r!}, (where\ n_1 + n_2 + \dots + n_r = n)$
$C(n, r) = \binom{n}{r} = \frac{n!}{r!(n-r)!}$
$P(A \cup B) = P(A) + P(B) - P(A \cap B)$
$P(B A) = \frac{P(A \cap B)}{P(A)}$
$P(B A) = P(B)$ or $P(A B) = P(A)$ .
$P(A_1 \cap A_2 \cap \dots \cap A_k) = P(A_1)P(A_2 A_1)P(A_3 A_1 \cap A_2)$
$P(A) = \sum_{i=1}^k P(B_i \cap A) = \sum_{i=1}^k P(B_i)P(A B_i)$
$P(B_r A) = \frac{P(B_r \cap A)}{\sum_{i=1}^k P(B_i \cap A)} = \frac{P(B_r)P(A B_r)}{\sum_{i=1}^k P(B_i)P(A B_i)}$
$P(B_r A) = P(B_r \cap A)/P(A)$
$F(x) = P(X \leq x) = \sum_{t \leq x} f(t),$ for $-\infty < x < \infty$
$F(x) = P(X \leq x) = \int_{-\infty}^x f(t)dt$ for $-\infty < x < \infty$
$g(x) = \sum_y f(x, y)$ and $h(y) = \sum_x f(x, y)$
$g(x) = \int_{-\infty}^{\infty} f(x, y)dy$ and $h(y) = \int_{-\infty}^{\infty} f(x, y)dx$
$f(y x) = \frac{f(x, y)}{g(x)}, g(x) > 0$
$f(x y) = \frac{f(x, y)}{h(y)}, h(y) > 0$
$P(a < X < b Y = y) = \sum_x f(x y),$ for the discrete case
$P(a < X < b Y = y) = \int_a^b f(x y),$ for the continuous case
$\begin{cases} \mu = E(X) = \sum_x x f(x) \text{ if } X \text{ is discrete} \\ \mu = E(X) = \int_{-\infty}^{\infty} x f(x) dx \text{ if } X \text{ is continuous} \end{cases}$
$\begin{cases} \mu_{g(X)} = E[g(X)] = \sum_x g(x) f(x) \text{ if } X \text{ is discrete} \\ \mu_{g(X)} = E[g(X)] = \int_{-\infty}^{\infty} g(x) f(x) dx \text{ if } X \text{ is continuous} \end{cases}$
$\begin{cases} \sigma^2 = E(X - \mu)^2 = \sum_x (x - \mu)^2 f(x), \text{ if } X \text{ is discrete} \\ \sigma^2 = E(X - \mu)^2 = \int_{-\infty}^{\infty} (x - \mu)^2 f(x) dx, \text{ if } X \text{ is continuous} \end{cases}$
$\begin{cases} \sigma_{g(X)}^2 = E\{[g(X) - \mu_{g(X)}]^2\} = \sum_x [g(X) - \mu_{g(X)}]^2, \\ \text{if } X \text{ is discrete} \\ \sigma_{g(X)}^2 = E\{[g(X) - \mu_{g(X)}]^2\} = \int_{-\infty}^{\infty} [g(X) - \mu_{g(X)}]^2 f(x) dx, \\ \text{if } X \text{ is continuous} \end{cases}$
$\sigma_{XY} = E(XY) - \mu_X \mu_Y$
$\rho_{XY} = \frac{\sigma_{XY}}{\sigma_X \sigma_Y}, -1 \leq \rho_{XY} \leq 1$
$P(\mu - k\sigma < X < \mu + k\sigma) \geq 1 - \frac{1}{k^2}$

Table 2: Formulae Sheet Cont.

$b(x; n, p) = \binom{n}{x} p^x q^{n-x}, \quad x = 0, 1, 2, \dots, n \quad B(r; n, p) = \sum_{x=0}^r b(x; n, p)$ $\mu = np \text{ and } \sigma^2 = npq$	
$h(x; N, n, k) = \frac{\binom{k}{x} \binom{N-k}{n-x}}{\binom{N}{n}} \quad \mu = \frac{nk}{N} \text{ and } \sigma^2 = \frac{N-n}{n-1} * n * \frac{k}{N} * \left(1 - \frac{k}{N}\right)$	
$f(x_1, x_2, \dots, x_k; a_1, a_2, \dots, a_k, N, n) = \frac{\binom{a_1}{x_1} \binom{a_2}{x_2} \dots \binom{a_k}{x_k}}{\binom{N}{n}}$	
$b^*(x; k, p) = \binom{x-1}{k-1} p^k q^{x-k}, \quad x = k, k+1, k+2, \dots$ $g(x; p) = b^*(x; 1, p) = pq^{x-1}, \quad x = 1, 2, 3, \dots \quad \mu = \frac{1}{p}, \quad \sigma^2 = \frac{1-p}{p^2}$	
$p(x; \lambda t) = \frac{e^{-\lambda t} (\lambda t)^x}{x!}$	
$n(x; \mu, \sigma) = \frac{1}{\sqrt{2\pi\sigma}} e^{-\frac{1}{2\sigma^2}(x-\mu)^2}, \quad -\infty < x < \infty \quad Z = \frac{X-\mu}{\sigma} \quad Z = \frac{X-np}{\sqrt{npq}}$	
$f(x) = \begin{cases} \frac{1}{\beta^\alpha \Gamma(\alpha)} x^{\alpha-1} e^{-x/\beta}, & x > 0 \\ 0, & \text{elsewhere} \end{cases} \quad \Gamma(n+1) = n\Gamma(n) = n!$ $\mu = \alpha\beta \text{ and } \sigma^2 = \alpha\beta^2$	
$f(x) = \begin{cases} \frac{1}{\beta} e^{-x/\beta}, & x > 0 \\ 0, & \text{elsewhere} \end{cases} \quad \mu = \beta \text{ and } \sigma^2 = \beta^2$	
$f(x; \nu) = \begin{cases} \frac{1}{2^{\nu/2} \Gamma(\nu/2)} x^{\nu/2-1} e^{-x/2}, & x > 0 \\ 0, & \text{elsewhere} \end{cases} \quad \mu = \nu \text{ and } \sigma^2 = 2\nu$	
$f(x) = \begin{cases} \frac{1}{\sqrt{2\pi\sigma x}} e^{-\frac{1}{2}[(\ln(x)-\mu)/\sigma]^2}, & x \geq 0 \\ 0, & x < 0 \end{cases} \quad \mu = e^{\mu+\sigma^2/2} \text{ and } \sigma^2 = e^{2\mu+\sigma^2} * (e^{\sigma^2} - 1)$	
$E(\bar{X}) = \mu_{\bar{X}} = \frac{\mu+\mu+\dots+\mu}{n} = \mu \quad \sigma_{\bar{X}}^2 = \frac{\sigma^2+\sigma^2+\dots+\sigma^2}{n^2} = \frac{\sigma^2}{n} \text{ (or } \frac{\sigma^2}{n} \left(\frac{N-n}{N-1}\right)) \quad Z = \frac{X-\mu}{\sigma/\sqrt{n}}$	
$\mu_{\bar{X}_1 - \bar{X}_2} = \mu_1 - \mu_2 \quad \sigma_{\bar{X}_1 - \bar{X}_2}^2 = \frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2} \quad Z = \frac{(\bar{X}_1 - \bar{X}_2) - (\mu_1 - \mu_2)}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}}$	
$h(t) = \frac{\Gamma[(\nu+1)/2]}{\Gamma(\nu/2)\sqrt{\pi\nu}} \left(1 + \frac{t^2}{\nu}\right)^{-(\nu+1)/2}, \quad -\infty < t < \infty \quad T = \frac{Z}{\sqrt{V/\nu}}$	
$h(f) = \begin{cases} \frac{\Gamma[(\nu_1+\nu_2)/2] (\nu_1/\nu_2)^{\nu_1/2}}{\Gamma(\nu_1/2)\Gamma(\nu_2/2)} \frac{f^{\nu_1/2-1}}{(1+\nu_1 f/\nu_2)^{(\nu_1+\nu_2)+1}}, & f > 0 \\ 0, & f \leq 0 \end{cases} \quad F = \frac{U/\nu_1}{V/\nu_2} = \frac{S_1^2/\sigma_1^2}{S_2^2/\sigma_2^2} = \frac{\sigma_2^2 S_1^2}{\sigma_1^2 S_2^2}$	
$\int u dv = uv - \int v du$	