

uniform distr: $f(x; k) = 1/k$ $\mu = \frac{\sum_{i=1}^k x_i}{k}$ $\sigma^2 = \frac{\sum_{i=1}^k (x_i - \mu)^2}{k}$

binomial: $b(x; n, p) = \binom{n}{x} p^x q^{n-x}$ $B(x; n, p) = \sum_{i=0}^x b(x; n, p)$
 $\mu = np$ $\sigma^2 = npq$

multinomial
 $f(x_1, x_2, \dots, x_k; p_1, p_2, \dots, p_k) = \binom{n}{x_1, x_2, \dots, x_k} p_1^{x_1} p_2^{x_2} \dots p_k^{x_k}$

hypergeometric $h(x; N, n, k) = \frac{\binom{k}{x} \binom{N-k}{n-x}}{\binom{N}{n}}, \quad x = 0, 1, 2, \dots, n$
 $\mu = \frac{nk}{N}$ $\sigma^2 = \left(\frac{N-n}{N-1}\right) \left(\frac{nk}{N}\right) \left(1 - \frac{k}{N}\right)$

hypergeometric multinomial: $f(x_1, x_2, \dots, x_k; a_1, a_2, \dots, a_k, N, n) = \frac{\binom{a_1}{x_1} \binom{a_2}{x_2} \dots \binom{a_k}{x_k}}{\binom{N}{n}}$

negative binomial (first k successes): $b^*(x; k, p) = \binom{x-1}{k-1} p^k q^{x-k}, \quad x = k, k+1, k+2, \dots$

geometric (first success) $g(x, p) = pq^{x-1}, \quad x = 1, 2, 3, \dots$ $\mu = \frac{1}{p}$ $\sigma^2 = \frac{1-p}{p^2}$

Poisson $p(x; \lambda t) = \frac{e^{-\lambda t} (\lambda t)^x}{x!}, \quad x = 0, 1, 2, \dots, \quad P(r; \lambda t) = \sum_{x=0}^r p(x; \lambda t) \quad \text{see Table A.2}$

uniform distr: $f(x; A, B) = \frac{1}{B-A}, \quad \text{where } A \leq x \leq B$
 $= 0 \quad \text{elsewhere}$ $\mu = \frac{A+B}{2}$ $\sigma^2 = \frac{(B-A)^2}{12}$

Normal

$Z = \frac{X - \mu}{\sigma}$ $P(x_1 < X < x_2) = \frac{1}{\sigma\sqrt{2\pi}} \int_{x_1}^{x_2} e^{-0.5\left\{\frac{(x-\mu)}{\sigma}\right\}^2} dx = \frac{1}{\sigma\sqrt{2\pi}} \int_{z_1}^{z_2} e^{-0.5\{z^2\}} dz = \int_{z_1}^{z_2} n(z; 0, 1) dz$

Normal approx of binomial $\mu = np$ $\sigma^2 = npq$ $Z = \frac{X - np}{\sqrt{npq}}$

Note: remember 1/2 offset

gamma

$\Gamma(x) = \int_0^{\infty} x^{\alpha-1} e^{-x} dx, \quad \alpha > 0$ $f(x = \text{time}) = \begin{cases} \frac{x^{\alpha-1} e^{-x/\beta}}{\beta^\alpha \Gamma(\alpha)} & x > 0 \\ 0 & \text{elsewhere} \end{cases}$ $f(t) = \frac{e^{-t/\beta}}{\beta}$

$\mu = \alpha\beta \quad \sigma^2 = \alpha\beta^2$

Exponential $f(x) = \begin{cases} \frac{e^{-x/\beta}}{\beta} & x > 0 \\ 0 & \text{elsewhere} \end{cases}$ $\mu = \alpha\beta$ $\sigma^2 = \alpha\beta^2$

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

log-normal: $f(x) = \frac{e^{-0.5\left\{\frac{(\ln[x]-\mu)^2}{\sigma^2}\right\}}}{\sigma\sqrt{2\pi}}, \quad x \geq 0$

$$Var(X) = e^{2\mu + \sigma^2} (e^{\sigma^2} - 1)$$

Z-values $s^2 = \frac{\sum_{i=1}^n (X_i - \bar{X})^2}{n-1}$ $s^2 = \frac{n \sum_{i=1}^n (X_i)^2 - \left(\sum_{i=1}^n X_i\right)^2}{n(n-1)}$ $Z = \frac{\bar{X} - \mu}{\sigma/\sqrt{n}}$

quantile plots: $f_i = \frac{i - (3/8)}{n + (1/4)}$

$$q_{\mu,\sigma}(f_i) = \mu + \sigma \{4.91 [f_i^{0.14} - (1-f_i)^{0.14}]\}$$

Example, for std normal:

$$q_{0,1}(f_i) = 0 + 1 \{4.91 [f_i^{0.14} - (1-f_i)^{0.14}]\}$$

difference in means, normal

$$\mu_{\bar{X}_1 - \bar{X}_2} = \mu_1 - \mu_2 \quad \sigma_{\bar{X}_1 - \bar{X}_2}^2 = \frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2} \quad Z = \frac{(\bar{X}_1 - \bar{X}_2) - (\mu_1 - \mu_2)}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}}$$

Chi-squared $\chi^2 = \frac{(n-1)S^2}{\sigma^2} = \sum_{i=1}^n \frac{(X_i - \bar{X})^2}{\sigma^2}, \quad \nu = n-1$

Student t $T = \frac{\bar{X} - \mu}{S/\sqrt{n}}$

F-test: $F = \frac{S_1^2/\sigma_1^2}{S_2^2/\sigma_2^2}$ Note: remember that typically assume $\sigma_1 = \sigma_2$ $F = \frac{S_1^2}{S_2^2}$

$$f_{1-\alpha, \nu_1, \nu_2} = \frac{1}{f_{\alpha, \nu_2, \nu_1}}$$