

Discrete Distributions

Bernoulli

$$0 < p < 1$$

$$\begin{aligned} f(x) &= p^x(1-p)^{1-x}, & x = 0, 1 \\ M(t) &= 1 - p + pe^t, & -\infty < t < \infty \\ \mu &= p, & \sigma^2 = p(1-p) \end{aligned}$$

Binomial

$$b(n, p)$$

$$0 < p < 1$$

$$\begin{aligned} f(x) &= \frac{n!}{x!(n-x)!} p^x(1-p)^{n-x}, & x = 0, 1, 2, \dots, n \\ M(t) &= (1 - p + pe^t)^n, & -\infty < t < \infty \\ \mu &= np, & \sigma^2 = np(1-p) \end{aligned}$$

Geometric

$$0 < p < 1$$

$$\begin{aligned} f(x) &= (1-p)^{x-1}p, & x = 1, 2, 3, \dots \\ M(t) &= \frac{pe^t}{1 - (1-p)e^t}, & t < -\ln(1-p) \\ \mu &= \frac{1}{p}, & \sigma^2 = \frac{1-p}{p^2} \end{aligned}$$

Hypergeometric

$$N_1 > 0, N_2 > 0$$

$$N = N_1 + N_2$$

$$\begin{aligned} f(x) &= \frac{\binom{N_1}{x} \binom{N_2}{n-x}}{\binom{N}{n}}, & x \leq n, x \leq N_1, n-x \leq N_2 \\ \mu &= n \left(\frac{N_1}{N} \right), & \sigma^2 = n \left(\frac{N_1}{N} \right) \left(\frac{N_2}{N} \right) \left(\frac{N-n}{N-1} \right) \end{aligned}$$

Negative Binomial

$$0 < p < 1$$

$$r = 1, 2, 3, \dots$$

$$\begin{aligned} f(x) &= \binom{x-1}{r-1} p^r (1-p)^{x-r}, & x = r, r+1, r+2, \dots \\ M(t) &= \frac{(pe^t)^r}{[1 - (1-p)e^t]^r}, & t < -\ln(1-p) \\ \mu &= r \left(\frac{1}{p} \right), & \sigma^2 = \frac{r(1-p)}{p^2} \end{aligned}$$

Poisson

$$\lambda > 0$$

$$\begin{aligned} f(x) &= \frac{\lambda^x e^{-\lambda}}{x!}, & x = 0, 1, 2, \dots \\ M(t) &= e^{\lambda(e^t-1)}, & -\infty < t < \infty \\ \mu &= \lambda, & \sigma^2 = \lambda \end{aligned}$$

Uniform

$$m > 0$$

$$\begin{aligned} f(x) &= \frac{1}{m}, & x = 1, 2, \dots, m \\ \mu &= \frac{m+1}{2}, & \sigma^2 = \frac{m^2-1}{12} \end{aligned}$$

Continuous Distributions

Beta

$$\alpha > 0$$

$$\beta > 0$$

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

$$\mu = \frac{\alpha}{\alpha + \beta}, \quad \sigma^2 = \frac{\alpha\beta}{(\alpha + \beta + 1)(\alpha + \beta)^2}$$

Chi-square

$$\chi^2(r)$$

$$r = 1, 2, \dots$$

$$f(x) = \frac{1}{\Gamma(r/2)2^{r/2}} x^{r/2-1} e^{-x/2}, \quad 0 < x < \infty$$

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

$$\mu = r, \quad \sigma^2 = 2r$$

Exponential

$$\theta > 0$$

$$f(x) = \frac{1}{\theta} e^{-x/\theta}, \quad 0 \leq x < \infty$$

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

$$\mu = \theta, \quad \sigma^2 = \theta^2$$

Gamma

$$\alpha > 0$$

$$\theta > 0$$

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

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

$$\mu = \alpha\theta, \quad \sigma^2 = \alpha\theta^2$$

Normal

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

$$-\infty < \mu < \infty$$

$$\sigma > 0$$

$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-(x-\mu)^2/2\sigma^2}, \quad -\infty < x < \infty$$

$$M(t) = e^{\mu t + \sigma^2 t^2/2}, \quad -\infty < t < \infty$$

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

Uniform

$$U(a, b)$$

$$-\infty < a < b < \infty$$

$$f(x) = \frac{1}{b-a}, \quad a \leq x \leq b$$

$$M(t) = \frac{e^{tb} - e^{ta}}{t(b-a)}, \quad t \neq 0; \quad M(0) = 1$$

$$\mu = \frac{a+b}{2}, \quad \sigma^2 = \frac{(b-a)^2}{12}$$

Confidence Intervals

Parameter	Assumptions	Endpoints
μ	$N(\mu, \sigma^2)$ or n large, σ^2 known	$\bar{x} \pm z_{\alpha/2} \frac{\sigma}{\sqrt{n}}$
μ	$N(\mu, \sigma^2)$ σ^2 unknown	$\bar{x} \pm t_{\alpha/2}(n-1) \frac{s}{\sqrt{n}}$
$\mu_X - \mu_Y$	$N(\mu_X, \sigma_X^2)$ $N(\mu_Y, \sigma_Y^2)$ σ_X^2, σ_Y^2 known	$\bar{x} - \bar{y} \pm z_{\alpha/2} \sqrt{\frac{\sigma_X^2}{n} + \frac{\sigma_Y^2}{m}}$
$\mu_X - \mu_Y$	Variances unknown, large samples	$\bar{x} - \bar{y} \pm z_{\alpha/2} \sqrt{\frac{s_x^2}{n} + \frac{s_y^2}{m}}$
$\mu_X - \mu_Y$	$N(\mu_X, \sigma_X^2)$ $N(\mu_Y, \sigma_Y^2)$ $\sigma_X^2 = \sigma_Y^2$, unknown	$\bar{x} - \bar{y} \pm t_{\alpha/2}(n+m-2) s_p \sqrt{\frac{1}{n} + \frac{1}{m}},$ $s_p = \sqrt{\frac{(n-1)s_x^2 + (m-1)s_y^2}{n+m-2}}$
$\mu_D = \mu_X - \mu_Y$	X and Y normal, but dependent	$\bar{d} \pm t_{\alpha/2}(n-1) \frac{s_d}{\sqrt{n}}$
p	$b(n, p)$ n is large	$\frac{y}{n} \pm z_{\alpha/2} \sqrt{\frac{(y/n)[1 - (y/n)]}{n}}$
$p_1 - p_2$	$b(n_1, p_1)$ $b(n_2, p_2)$	$\frac{y_1}{n_1} - \frac{y_2}{n_2} \pm z_{\alpha/2} \sqrt{\frac{\hat{p}_1(1 - \hat{p}_1)}{n_1} + \frac{\hat{p}_2(1 - \hat{p}_2)}{n_2}},$ $\hat{p}_1 = y_1/n_1, \hat{p}_2 = y_2/n_2$

Tests of Hypotheses

Hypotheses	Assumptions	Critical Region
$H_0: \mu = \mu_0$ $H_1: \mu > \mu_0$	$N(\mu, \sigma^2)$ or n large, σ^2 known	$z = \frac{\bar{x} - \mu_0}{\sigma/\sqrt{n}} \geq z_\alpha$
$H_0: \mu = \mu_0$ $H_1: \mu > \mu_0$	$N(\mu, \sigma^2)$ σ^2 unknown	$t = \frac{\bar{x} - \mu_0}{s/\sqrt{n}} \geq t_\alpha(n-1)$
$H_0: \mu_X - \mu_Y = 0$ $H_1: \mu_X - \mu_Y > 0$	$N(\mu_X, \sigma_X^2)$ $N(\mu_Y, \sigma_Y^2)$ σ_X^2, σ_Y^2 known	$z = \frac{\bar{x} - \bar{y} - 0}{\sqrt{(\sigma_X^2/n) + (\sigma_Y^2/m)}} \geq z_\alpha$
$H_0: \mu_X - \mu_Y = 0$ $H_1: \mu_X - \mu_Y > 0$	Variances unknown, large samples	$z = \frac{\bar{x} - \bar{y} - 0}{\sqrt{(s_X^2/n) + (s_Y^2/m)}} \geq z_\alpha$
$H_0: \mu_X - \mu_Y = 0$ $H_1: \mu_X - \mu_Y > 0$	$N(\mu_X, \sigma_X^2)$ $N(\mu_Y, \sigma_Y^2)$ $\sigma_X^2 = \sigma_Y^2$, unknown	$t = \frac{\bar{x} - \bar{y} - 0}{s_p \sqrt{(1/n) + (1/m)}} \geq t_\alpha(n+m-2)$ $s_p = \sqrt{\frac{(n-1)s_X^2 + (m-1)s_Y^2}{n+m-2}}$
$H_0: \mu_D = \mu_X - \mu_Y = 0$ $H_1: \mu_D = \mu_X - \mu_Y > 0$	X and Y normal, but dependent	$t = \frac{\bar{d} - 0}{s_d/\sqrt{n}} \geq t_\alpha(n-1)$
$H_0: p = p_0$ $H_1: p > p_0$	$b(n, p)$ n is large	$z = \frac{(y/n) - p_0}{\sqrt{p_0(1-p_0)/n}} \geq z_\alpha$
$H_0: p_1 - p_2 = 0$ $H_1: p_1 - p_2 > 0$	$b(n_1, p_1)$ $b(n_2, p_2)$	$z = \frac{(y_1/n_1) - (y_2/n_2) - 0}{\sqrt{\left(\frac{y_1 + y_2}{n_1 + n_2}\right)\left(1 - \frac{y_1 + y_2}{n_1 + n_2}\right)\left(\frac{1}{n_1} + \frac{1}{n_2}\right)}} \geq z_\alpha$