

## Formulário

$P(X = x) = \binom{n}{x} p^x (1-p)^{n-x}$ $x = 0, 1, \dots, n$ $E(X) = np \quad \text{Var}(X) = np(1-p)$	$P(X = x) = \frac{e^{-\lambda} \lambda^x}{x!}$ $x = 0, 1, \dots$ $E(X) = \text{Var}(X) = \lambda$	$P(X = x) = p(1-p)^{x-1}$ $x = 1, 2, \dots$ $E(X) = \frac{1}{p} \quad \text{Var}(X) = \frac{(1-p)}{p^2}$
$P(X = x) = \binom{M}{x} \binom{N-M}{n-x} / \binom{N}{n}$ $x = \max\{0, n-N+M\}, \dots, \min\{n, M\}$ $E(X) = n \frac{M}{N} \quad \text{Var}(X) = n \frac{M}{N} \frac{N-M}{N} \frac{N-n}{N-1}$	$f_X(x) = \frac{1}{b-a}, \quad a \leq x \leq b$ $E(X) = \frac{b+a}{2} \quad \text{Var}(X) = \frac{(b-a)^2}{12}$	
$f_X(x) = \frac{1}{\sqrt{2\pi}\sigma^2} \exp\left\{-\frac{(x-\mu)^2}{2\sigma^2}\right\}, \quad x \in \mathbb{R}$ $E(X) = \mu \quad \text{Var}(X) = \sigma^2$	$f_X(x) = \lambda e^{-\lambda x}, \quad x \geq 0$ $E(X) = \frac{1}{\lambda} \quad \text{Var}(X) = \frac{1}{\lambda^2}$	
$\frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \sim N(0, 1)$	$\frac{\bar{X} - \mu}{S/\sqrt{n}} \sim t_{(n-1)}$	$\frac{(\bar{X}_1 - \bar{X}_2) - (\mu_1 - \mu_2)}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}} \sim N(0, 1)$
$\frac{\bar{X} - \mu}{S/\sqrt{n}} \stackrel{a}{\sim} N(0, 1)$	$S^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2$	
$\frac{(\bar{X}_1 - \bar{X}_2) - (\mu_1 - \mu_2)}{\sqrt{\frac{S_1^2}{n_1} + \frac{S_2^2}{n_2}}} \stackrel{a}{\sim} N(0, 1)$	$\frac{(\bar{X}_1 - \bar{X}_2) - (\mu_1 - \mu_2)}{\sqrt{\frac{(n_1-1)S_1^2 + (n_2-1)S_2^2}{n_1+n_2-2} \left(\frac{1}{n_1} + \frac{1}{n_2}\right)}} \sim t_{(n_1+n_2-2)}$	
$\frac{(n-1)S^2}{\sigma^2} \sim \chi_{(n-1)}^2$	$\sum_{i=1}^k \frac{(O_i - E_i)^2}{E_i} \stackrel{a}{\sim} \chi_{(k-\beta-1)}^2$	$\sum_{i=1}^r \sum_{j=1}^s \frac{(O_{ij} - E_{ij})^2}{E_{ij}} \stackrel{a}{\sim} \chi_{(r-1)(s-1)}^2$
$Y_i = \beta_0 + \beta_1 x_i + \varepsilon_i$	$\hat{\beta}_0 = \bar{Y} - \hat{\beta}_1 \bar{x}$	$\hat{\beta}_1 = \frac{\sum_{i=1}^n x_i Y_i - n \bar{x} \bar{Y}}{\sum_{i=1}^n x_i^2 - n \bar{x}^2}$
$\hat{\sigma}^2 = \frac{1}{n-2} \sum_{i=1}^n (Y_i - \hat{Y}_i)^2, \quad \hat{Y}_i = \hat{\beta}_0 + \hat{\beta}_1 x_i$		$\hat{\sigma}^2 = \frac{1}{n-2} \left[ \left( \sum_{i=1}^n Y_i^2 - n \bar{Y}^2 \right) - (\hat{\beta}_1)^2 \left( \sum_{i=1}^n x_i^2 - n \bar{x}^2 \right) \right]$
$\frac{\hat{\beta}_0 - \beta_0}{\sqrt{\left(\frac{1}{n} + \frac{\bar{x}^2}{\sum x_i^2 - n \bar{x}^2}\right) \hat{\sigma}^2}} \sim t_{(n-2)}$	$\frac{\hat{\beta}_1 - \beta_1}{\sqrt{\frac{\hat{\sigma}^2}{\sum x_i^2 - n \bar{x}^2}}} \sim t_{(n-2)}$	$\frac{(\hat{\beta}_0 + \hat{\beta}_1 x_0) - (\beta_0 + \beta_1 x_0)}{\sqrt{\left(\frac{1}{n} + \frac{(\bar{x} - x_0)^2}{\sum x_i^2 - n \bar{x}^2}\right) \hat{\sigma}^2}} \sim t_{(n-2)}$
$R^2 = \frac{\left( \sum_{i=1}^n x_i Y_i - n \bar{x} \bar{Y} \right)^2}{\left( \sum_{i=1}^n x_i^2 - n \bar{x}^2 \right) \times \left( \sum_{i=1}^n Y_i^2 - n \bar{Y}^2 \right)}$		