

POPOLAZIONE NORMALE			
$\left(\bar{X} - z_{\alpha/2} \frac{\sigma}{\sqrt{n}}, \bar{X} + z_{\alpha/2} \frac{\sigma}{\sqrt{n}}\right)$	Media: varianza nota	$\left(\bar{X} - t_{\alpha/2;n-1} \frac{S}{\sqrt{n}}, \bar{X} + t_{\alpha/2;n-1} \frac{S}{\sqrt{n}}\right)$	Media: varianza sconosciuta
$\left(\bar{X}_1 - \bar{X}_2 - z_{\alpha/2} \sqrt{\frac{\sigma_1^2 + \sigma_2^2}{n+m}}, \bar{X}_1 - \bar{X}_2 + z_{\alpha/2} \sqrt{\frac{\sigma_1^2 + \sigma_2^2}{n+m}}\right)$	Differenza tra le medie, campioni indip: varianza nota	$(\bar{X}_1 - \bar{X}_2 - t_{\alpha/2;n+m-2} S_p \delta, \bar{X}_1 - \bar{X}_2 + t_{\alpha/2;n+m-2} S_p \delta)$ $\delta = \sqrt{\frac{1}{n} + \frac{1}{m}}, S_p^2 = \frac{(n-1)S_1^2 + (m-1)S_2^2}{n+m-2}$	Differenza tra le medie, campioni indip: varianze uguali ma sconosciute
$\bar{x}_1 - \bar{x}_2 \pm t'_{\alpha/2} \sqrt{\frac{S_1^2}{n_1} + \frac{S_2^2}{n_2}}$ $t'_{\alpha/2} = \frac{\frac{S_1^2}{n_1} t_1 + \frac{S_2^2}{n_2} t_2}{\frac{S_1^2}{n_1} + \frac{S_2^2}{n_2}}$ $t_1 = t_{\alpha/2} \quad (n_1 - 1) \text{ g.l.}$ $t_2 = t_{\alpha/2} \quad (n_2 - 1) \text{ g.l.}$		Formula approssimata: differenza tra le medie, campioni indip. varianze diverse	
$\left(\frac{\sum_{i=1}^n (X_i - \mu)^2}{\chi_{\alpha/2;n}^2}, \frac{\sum_{i=1}^n (X_i - \mu)^2}{\chi_{1-\alpha/2;n}^2}\right)$	Varianza: media nota	$\left(\frac{(n-1)S^2}{\chi_{\alpha/2;n-1}^2}, \frac{(n-1)S^2}{\chi_{1-\alpha/2;n-1}^2}\right)$	Varianza: media sconosciuta
$\left(\frac{1}{F_{\alpha/2;n-1,m-1}} \frac{S_1^2}{S_2^2}, F_{\alpha/2;m-1,n-1} \frac{S_1^2}{S_2^2}\right)$	Rapporto tra le varianze		
POPOLAZIONE DI BERNOULLI			
$\left(\bar{X} - z_{\alpha/2} \sqrt{\frac{\bar{X}(1-\bar{X})}{n}}, \bar{X} + z_{\alpha/2} \sqrt{\frac{\bar{X}(1-\bar{X})}{n}}\right)$	Campioni grandi	$\bar{X}_1 - \bar{X}_2 \pm z_{\alpha/2} \sqrt{\frac{\bar{X}_1(1-\bar{X}_1)}{n_1} + \frac{\bar{X}_2(1-\bar{X}_2)}{n_2}}$	Campioni grandi: differenza tra proporzioni
POPOLAZIONE ESPONENZIALE			
$\left(\frac{2}{\chi_{\alpha/2;2n}^2} \sum_{i=1}^n X_i, \frac{2}{\chi_{1-\alpha/2;2n}^2} \sum_{i=1}^n X_i\right)$	Parametro θ di una popolazione con media θ		
QUANTILI			
Normale	Chi quadro	T di Student	Fisher
$P(Z \geq z_\alpha) \equiv \int_{z_\alpha}^{\infty} f_Z(x) dx = \alpha$	$P(X \geq \chi_{\alpha,v}^2) \equiv \int_{\chi_{\alpha,v}^2}^{\infty} f_X(x) dx = \alpha$	$P(T \geq t_{\alpha,v}) \equiv \int_{t_{\alpha,v}}^{\infty} f_T(x) dx = \alpha$	$P(Y \geq F_{\alpha;\nu_2,\nu_1}) = \alpha$ $F_{1-\alpha;\nu_1,\nu_2} = \frac{1}{F_{\alpha;\nu_2,\nu_1}}$
NORMALE MULTIVARIATA			
DENSITA' DI PROBABILITA'		MOMENTI CONDIZIONATI	
$C = \begin{pmatrix} \sigma_1^2 & \sigma_{12} \\ \sigma_{12} & \sigma_2^2 \end{pmatrix}, \quad \sigma_i^2 = Var(X_i), \quad \sigma_{12} = cov(X_1, X_2), \quad \rho = \frac{\sigma_{12}}{\sigma_1 \sigma_2}$		$E(X_1 X_2 = x_2) = m_1 + \rho \frac{\sigma_1}{\sigma_2} (x_2 - m_2)$	
$f(x_1, x_2) = \frac{1}{2\pi\sigma_1\sigma_2\sqrt{1-\rho^2}} \exp\left\{-\frac{1}{2(1-\rho^2)} \left[\left(\frac{x-m_1}{\sigma_1}\right)^2 + \left(\frac{x-m_2}{\sigma_2}\right)^2 - 2\rho \left(\frac{x-m_1}{\sigma_1}\right) \left(\frac{x-m_2}{\sigma_2}\right) \right]\right\}$		$Var(X_1 X_2 = x_2) = \sigma_1^2 (1 - \rho^2) \beta$	

	Regressione Y=aX+b	
$S_{xx} = \sum_{i=1}^n (x_i - \bar{x})^2$ $S_{xy} = \sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y}) = \sum_{i=1}^n x_i y_i - n\bar{x}\bar{y}$	$\hat{a} = \bar{y} - \hat{b}\bar{x}$ $\hat{b} = \frac{S_{xy}}{S_{xx}}$ $\hat{\sigma}^2 = \frac{1}{n} \sum_{i=1}^n \left[y_i - (\hat{a} + \hat{b}x_i) \right]^2 = \frac{S_{yy} - \hat{b}S_{xy}}{n} = \frac{SS_R}{n}$	$\hat{y} = \bar{y} + \frac{S_{xy}}{S_{xx}}(x - \bar{x})$
Intervalli confidenza	$\left(\hat{A} - t_{\alpha/2,n-2} \sqrt{\frac{SS_R}{(n-2)} \left(\frac{1}{n} + \frac{\bar{x}^2}{S_{xx}} \right)}, \hat{A} + t_{\alpha/2,n-2} \sqrt{\frac{SS_R}{(n-2)} \left(\frac{1}{n} + \frac{\bar{x}^2}{S_{xx}} \right)} \right)$ $\left(\hat{B} - t_{\alpha/2,n-2} \sqrt{\frac{SS_R}{(n-2)S_{xx}}}, \hat{B} + t_{\alpha/2,n-2} \sqrt{\frac{SS_R}{(n-2)S_{xx}}} \right)$	E(Y(x₀) x₀)): $\left(\hat{A} + \hat{B}x_0 \pm t_{\alpha/2,n-2} \sqrt{\frac{SS_R}{(n-2)} \left[\frac{1}{n} + \frac{(x_0 - \bar{x})^2}{S_{xx}} \right]} \right)$ Y(x₀) $\left(\hat{A} + \hat{B}x_0 \pm t_{\alpha/2,n-2} \sqrt{\frac{SS_R}{(n-2)} \left[1 + \frac{1}{n} + \frac{(x_0 - \bar{x})^2}{S_{xx}} \right]} \right)$
Test	$T_{a_0} = \frac{(\hat{A} - a_0)}{\sqrt{\frac{S_R}{(n-2)} \left(\frac{1}{n} + \frac{\bar{x}^2}{S_{xx}} \right)}}$ $T_{b_0} = (\hat{B} - b_0) \sqrt{\frac{(n-2)S_{xx}}{SS_R}}$	$R^2 = \frac{S_{xy}^2}{S_{xx} S_{yy}} = 1 - \frac{SS_R}{S_{yy}}$ $U = \frac{1}{2} \ln \frac{1+R}{1-R} \underset{\text{approssimativamente}}{\approx} N\left(\frac{1}{2} \ln \frac{1+\rho}{1-\rho}, \frac{1}{n-3}\right)$
	ANOVA	
a 1 via	$\frac{n}{\sigma^2} \sum (\bar{X}_{i\bullet} - \bar{X}_{\bullet\bullet})^2 \equiv \frac{SS_F}{\sigma^2} \approx \chi^2_{k-1}$	$\sum_{i=1}^k \sum_{j=1}^n (X_{ij} - \bar{X}_{i\bullet})^2 / \sigma^2 = SS_E / \sigma^2 \approx \chi^2_{k(n-1)}$
a 2 vie	$SS_A = n \sum_{i=1}^k (\bar{X}_{i\bullet} - \bar{X}_{\bullet\bullet})^2$	$SS_B = k \sum_{j=1}^n (\bar{X}_{\bullet j} - \bar{X}_{\bullet\bullet})^2$
	$F_A = (n-1) \frac{SS_A}{SS_E} \approx F_{\alpha;(k-1),(n-1)(k-1)}$	$F_B = (k-1) \frac{SS_B}{SS_E} \approx F_{\alpha;(n-1),(n-1)(k-1)}$
	Test del χ^2	
	Bontà di un fit $\chi^2 = \sum_{i=1}^k \frac{(n_i - n\pi_i)^2}{n\pi_i} \approx \chi^2_{\alpha,k-1}$	
	Indipendenza p _i note $\Xi = \sum_{i=1}^k \sum_{m=1}^l \frac{(N_{im} - np_i p_{.m})^2}{np_i p_{.m}} \underset{n \text{ grande}}{\approx} \chi^2_{kl-1}$	
	Indipendenza p _i sconosciute $\Xi = \sum_{i=1}^k \sum_{m=1}^l \frac{(N_{im} - n\hat{p}_i \hat{p}_{.m})^2}{n\hat{p}_i \hat{p}_{.m}} \approx \chi^2_{\alpha,(k-1)(l-1)}$	