

FORMULARIO DE ESTADÍSTICA

MODELOS DE PROBABILIDAD				
Nombre	Símbolo	F. de probabilidad o densidad	Media	Varianza
Bernoulli	$B(1, p)$	$P(x = r) = p^r q^{1-r}; r = 0, 1$	p	pq
Binomial	$B(n, p)$	$P(x = r) = \binom{n}{r} p^r q^{n-r}; r = 0, 1, \dots, n$	np	npq
Geométrica	$G(p)$	$P(x = r) = pq^{r-1}; r = 1, 2, \dots$	$\frac{1}{p}$	$\frac{q}{p^2}$
Poisson	$P(\lambda)$	$P(x = r) = \frac{\lambda^r e^{-\lambda}}{r!}; r = 0, 1, \dots$	λ	λ
Uniforme (cont.)	$U(a, b)$	$f(x) = \frac{1}{b-a}; a \leq x \leq b$	$\frac{b+a}{2}$	$\frac{(b-a)^2}{12}$
Exponencial	$\text{Exp}(\lambda)$	densidad: $f(x) = \lambda e^{-\lambda x}; x > 0$ distribución: $F(x) = 1 - e^{-\lambda x}$	$1/\lambda$	$1/\lambda^2$
Normal	$N(\mu, \sigma^2)$	$f(x) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp \left\{ -\frac{(x-\mu)^2}{2\sigma^2} \right\}$	μ	σ^2
Weibull	densidad distribución tasa de fallos	$f(t) = \frac{\beta}{\alpha^\beta} t^{\beta-1} \exp \left\{ -\left(\frac{t}{\alpha}\right)^\beta \right\}; t > 0$ $F(t) = 1 - \exp \left\{ -\left(\frac{t}{\alpha}\right)^\beta \right\}$ $\lambda(t) = \frac{\beta}{\alpha} \left(\frac{t}{\alpha}\right)^{\beta-1}$	Media $\mu = \alpha \Gamma \left(1 + \frac{1}{\beta} \right)$ con $\Gamma(r) = (r-1)!$	

Gráficos de Control			
Gráfico de	Dispersión	Línea Central	Límites de Control
Medias $\bar{x}_i$	$\bar{s}$	$\bar{\bar{x}}$	$\bar{\bar{x}} \pm 3 \frac{\bar{s}}{c_2 \sqrt{n}}$
Medias $\bar{x}_i$	Rangos	$\bar{\bar{x}}$	$\bar{\bar{x}} \pm 3 \frac{R}{d_2 \sqrt{n}}$
D. típicas s_i	$\bar{s}$	$\bar{s}$	$B_3 \bar{s}; B_4 \bar{s}$
Rangos R_i	Rangos	$\bar{R}$	$D_3 \bar{R}; D_4 \bar{R}$
Proporciones p_i		$\bar{p}$	$\bar{p} \pm 3 \sqrt{\frac{\bar{p}(1-\bar{p})}{n_i}}$
$\bar{s}$ = promedio de las desv. típicas s_i del estudio inicial, con k muestras de tamaño n			

Regresión simple: $\hat{y} = a + bx$; $a = \bar{y} - b\bar{x}$; $b = \text{cov}(x, y) / \text{var}(x)$

UNA POBLACIÓN				
Población	Contrastes	Estadísticos de contraste	Región de rechazo (p-valor < α)	Intervalo de confianza IC(1 - α)
Cualquiera (μ, σ^2) con $n \rightarrow \infty$	(1)- $H_0 : \mu = \mu_0$; $H_1 : \mu \neq \mu_0$ (2)- $H_0 : \mu \geq \mu_0$; $H_1 : \mu < \mu_0$ (3)- $H_0 : \mu \leq \mu_0$; $H_1 : \mu > \mu_0$	(a) $z_0 = \frac{x - \mu_0}{\sigma / \sqrt{n}}$ (b) $t_0 = \frac{\bar{x} - \mu_0}{\hat{s} / \sqrt{n}}$	(1-a) $ z_0 > z_{\alpha/2}$ (1-b) $ t_0 > z_{\alpha/2}$ (2-a) $z_0 < -z_\alpha$ (2-b) $t_0 < -z_\alpha$ (3-a) $z_0 > z_\alpha$ (3-b) $t_0 > z_\alpha$	$\left[\begin{array}{l} \mu \in \left\{ \bar{x} \pm z_{\alpha/2} \sigma / \sqrt{n} \right\} \\ \mu \in \left\{ \bar{x} \pm z_{\alpha/2} \hat{s} / \sqrt{n} \right\} \end{array} \right]$
$N(\mu, \sigma^2)$	(1)- $H_0 : \mu = \mu_0$; $H_1 : \mu \neq \mu_0$ (2)- $H_0 : \mu \geq \mu_0$; $H_1 : \mu < \mu_0$ (3)- $H_0 : \mu \leq \mu_0$; $H_1 : \mu > \mu_0$	(a) $z_0 = \frac{x - \mu_0}{\sigma / \sqrt{n}}$ (b) $t_0 = \frac{\bar{x} - \mu_0}{\hat{s} / \sqrt{n}}$	(1-a) $ z_0 > z_{\alpha/2}$ (1-b) $ t_0 > t_{n-1; \alpha/2}$ (2-a) $z_0 < -z_\alpha$ (2-b) $t_0 < -t_{n-1; \alpha}$ (3-a) $z_0 > z_\alpha$ (3-b) $t_0 > t_{n-1; \alpha}$	$\left[\begin{array}{l} \mu \in \left\{ \bar{x} \pm z_{\alpha/2} \sigma / \sqrt{n} \right\} \\ \mu \in \left\{ \bar{x} \pm t_{n-1; \alpha/2} \hat{s} / \sqrt{n} \right\} \end{array} \right]$
$B(n, p)$ con $n \rightarrow \infty$	(1)- $H_0 : p = p_0$; $H_1 : p \neq p_0$ (2)- $H_0 : p \geq p_0$; $H_1 : p < p_0$ (3)- $H_0 : p \leq p_0$; $H_1 : p > p_0$	$z_0 = \frac{\hat{p} - p_0}{\sqrt{p_0 q_0 / n}}$	(1) $ z_0 > z_{\alpha/2}$ (2) $z_0 < -z_\alpha$ (3) $z_0 > z_\alpha$	$p \in \left\{ \hat{p} \pm z_{\alpha/2} \sqrt{\frac{\hat{p}\hat{q}}{n}} \right\}$
$N(\mu, \sigma^2)$	(1)- $H_0 : \sigma^2 = \sigma_0^2$; $H_1 : \sigma^2 \neq \sigma_0^2$ (2)- $H_0 : \sigma^2 \geq \sigma_0^2$; $H_1 : \sigma^2 < \sigma_0^2$ (3)- $H_0 : \sigma^2 \leq \sigma_0^2$; $H_1 : \sigma^2 > \sigma_0^2$	$X_0 = \frac{(n-1)\hat{s}^2}{\sigma_0^2}$ $= \frac{ns^2}{\sigma_0^2}$	(1) $\left[\begin{array}{l} X_0 > \chi_{n-1; \alpha/2}^2 \\ \text{ó } X_0 < \chi_{n-1; 1-\alpha/2}^2 \end{array} \right]$ (2) $X_0 < \chi_{n-1; 1-\alpha}^2$ (3) $X_0 > \chi_{n-1; \alpha}^2$	$\sigma^2 \in \left[\frac{(n-1)\hat{s}^2}{\chi_{n-1; \alpha/2}^2}; \frac{(n-1)\hat{s}^2}{\chi_{n-1; 1-\alpha/2}^2} \right]$
Cualquiera θ (por Máxima Ver.) con $n \rightarrow \infty$	(1)- $H_0 : \theta = \theta_0$; $H_1 : \theta \neq \theta_0$ (2)- $H_0 : \theta \geq \theta_0$; $H_1 : \theta < \theta_0$ (3)- $H_0 : \theta \leq \theta_0$; $H_1 : \theta > \theta_0$	$t_0 = \frac{\hat{\theta}_{MV} - \theta_0}{\sqrt{\widehat{\text{Var}}(\hat{\theta}_{MV})}}$ $\widehat{\text{Var}}(\hat{\theta}_{MV}) = - \left[\frac{\partial^2 L(\theta)}{\partial \theta^2} \right]_{\theta = \hat{\theta}_{MV}}^{-1}$	(1) $ t_0 > z_{\alpha/2}$ (2) $t_0 < -z_\alpha$ (3) $t_0 > z_\alpha$	$\theta \in \left\{ \hat{\theta}_{MV} \pm z_{\alpha/2} \sqrt{\widehat{\text{Var}}(\hat{\theta}_{MV})} \right\}$

CONTRASTES- DOS POBLACIONES			
Dos poblaciones x_1, x_2	Contrastes	Estadísticos de contraste	Región de rechazo (p-valor< α)
Cualesquiera con $E(x_1) = \mu_1; \text{var}(x_1) = \sigma_1^2$ $E(x_2) = \mu_2; \text{var}(x_2) = \sigma_2^2$	(1)-H ₀ : $\mu_1 = \mu_2$; H ₁ : $\mu_1 \neq \mu_2$ (2)-H ₀ : $\mu_1 \geq \mu_2$; H ₁ : $\mu_1 < \mu_2$ (3)-H ₀ : $\mu_1 \leq \mu_2$; H ₁ : $\mu_1 > \mu_2$	$(a) z_0 = \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}}$ $(b) t_0 = \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{\frac{\hat{s}_1^2}{n_1} + \frac{\hat{s}_2^2}{n_2}}}$	(1-a) $ z_0 > z_{\alpha/2}$ (1-b) $ t_0 > z_{\alpha/2}$ (2-a) $z_0 < -z_{\alpha}$ (2-b) $t_0 < -z_{\alpha}$ (3-a) $z_0 > z_{\alpha}$ (3-b) $t_0 > z_{\alpha}$
Datos apareados $d = x_1 - x_2$	(1)-H ₀ : $\mu_d = 0$; H ₁ : $\mu_d \neq 0$ (2)-H ₀ : $\mu_d \geq 0$; H ₁ : $\mu_d < 0$ (3)-H ₀ : $\mu_d \leq 0$; H ₁ : $\mu_d > 0$	$t_0 = \frac{\bar{d}}{\hat{s}_d / \sqrt{n}}$	(1) $ t_0 > z_{\alpha/2}$ (1) $ t_0 > t_{n-1; \alpha/2}$ (2) $t_0 < -z_{\alpha}$ (2) $t_0 < -t_{n-1; \alpha}$ (3) $t_0 > z_{\alpha}$ (3) $t_0 > t_{n-1; \alpha}$ si $n \rightarrow \infty$ Sólo si hay normalidad
Cualesquiera con $E(x_1) = \mu_1; \text{var}(x_1) = \sigma^2$ $E(x_2) = \mu_2; \text{var}(x_2) = \sigma^2$	(1)-H ₀ : $\mu_1 = \mu_2$; H ₁ : $\mu_1 \neq \mu_2$ (2)-H ₀ : $\mu_1 \geq \mu_2$; H ₁ : $\mu_1 < \mu_2$ (3)-H ₀ : $\mu_1 \leq \mu_2$; H ₁ : $\mu_1 > \mu_2$	$(a) z_0 = \frac{\bar{x}_1 - \bar{x}_2}{\sigma \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}}$ $(b) t_0 = \frac{\bar{x}_1 - \bar{x}_2}{\hat{s}_T \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}}$ con $\hat{s}_T^2 = \frac{(n_1-1)\hat{s}_1^2 + (n_2-1)\hat{s}_2^2}{n_1 + n_2 - 2}$	(1) $ z_0 , t_0 > z_{\alpha/2}$ (1-b) $ t_0 > t_{n_1+n_2-2; \alpha/2}$ (2) $z_0, t_0 < -z_{\alpha}$ (2-b) $t_0 < -t_{n_1+n_2-2; \alpha}$ (3) $z_0, t_0 > z_{\alpha}$ (3-b) $t_0 > t_{n_1+n_2-2; \alpha}$ Sólo si hay normalidad
Binomiales $x_1 \sim B(n_1, p_1)$ $x_2 \sim B(n_2, p_2)$	(1)-H ₀ : $p_1 = p_2$; H ₁ : $p_1 \neq p_2$ (2)-H ₀ : $p_1 \geq p_2$; H ₁ : $p_1 < p_2$ (3)-H ₀ : $p_1 \leq p_2$; H ₁ : $p_1 > p_2$	$z_0 = \frac{\hat{p}_1 - \hat{p}_2}{\sqrt{\hat{p}_0 \hat{q}_0 \left(\frac{1}{n_1} + \frac{1}{n_2} \right)}}$ con $\hat{p}_0 = \frac{n_1 \hat{p}_1 + n_2 \hat{p}_2}{n_1 + n_2}$	(1) $ z_0 > z_{\alpha/2}$ (2) $z_0 < -z_{\alpha}$ (3) $z_0 > z_{\alpha}$ Siempre que $n_1, n_2 \rightarrow \infty$
Normales $x_1 \sim N(\mu_1, \sigma_1^2)$ $x_2 \sim N(\mu_2, \sigma_2^2)$	(1)-H ₀ : $\sigma_1^2 = \sigma_2^2$; H ₁ : $\sigma_1^2 \neq \sigma_2^2$ (2)-H ₀ : $\sigma_1^2 \geq \sigma_2^2$; H ₁ : $\sigma_1^2 < \sigma_2^2$ (3)-H ₀ : $\sigma_1^2 \leq \sigma_2^2$; H ₁ : $\sigma_1^2 > \sigma_2^2$	$F_0 = \frac{\hat{s}_1^2}{\hat{s}_2^2}$	(1) $F_0 > F_{n_1-1; n_2-1; \alpha/2}$ ó $F_0 < F_{n_1-1; n_2-1; 1-\alpha/2}$ donde $F_{n_1-1; n_2-1; 1-\alpha/2} = 1 / F_{n_2-1; n_1-1; \alpha/2}$ (2) $F_0 < F_{n_1-1; n_2-1; 1-\alpha}$ (3) $F_0 > F_{n_1-1; n_2-1; \alpha}$

ESTIMADORES-DOS POBLACIONES		
Dos poblaciones x_1, x_2	Parámetro	Intervalo de confianza, IC(1 - α)
Cualesquiera con $E(x_1) = \mu_1; \text{var}(x_1) = \sigma_1^2$ $E(x_2) = \mu_2; \text{var}(x_2) = \sigma_2^2$	$\mu_1 - \mu_2$	$\mu_1 - \mu_2 \in \left\{ \bar{x}_1 - \bar{x}_2 \pm z_{\alpha/2} \sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}} \right\}$ Si no hay normalidad, sólo si $n_1, n_2 \rightarrow \infty$ $\mu_1 - \mu_2 \in \left\{ \bar{x}_1 - \bar{x}_2 \pm z_{\alpha/2} \sqrt{\frac{\hat{s}_1^2}{n_1} + \frac{\hat{s}_2^2}{n_2}} \right\}$ Siempre que $n_1, n_2 \rightarrow \infty$
Cualesquiera con $E(x_1) = \mu_1; \text{var}(x_1) = \sigma^2$ $E(x_2) = \mu_2; \text{var}(x_2) = \sigma^2$	$\mu_1 - \mu_2$	$\mu_1 - \mu_2 \in \left\{ \bar{x}_1 - \bar{x}_2 \pm z_{\alpha/2} \sigma \sqrt{\frac{1}{n_1} + \frac{1}{n_2}} \right\}$ Si no hay normalidad, sólo si $n_1, n_2 \rightarrow \infty$ $\mu_1 - \mu_2 \in \left\{ \bar{x}_1 - \bar{x}_2 \pm z_{\alpha/2} \hat{s}_T \sqrt{\frac{1}{n_1} + \frac{1}{n_2}} \right\}$ Siempre que $n_1, n_2 \rightarrow \infty$ $\mu_1 - \mu_2 \in \left\{ \bar{x}_1 - \bar{x}_2 \pm t_{n_1+n_2-2; \alpha/2} \hat{s}_T \sqrt{\frac{1}{n_1} + \frac{1}{n_2}} \right\}$ Sólo si hay normalidad
Normales con $E(x_1) = \mu_1; \text{var}(x_1) = \sigma_1^2$ $E(x_2) = \mu_2; \text{var}(x_2) = \sigma_2^2$	$\frac{\sigma_1^2}{\sigma_2^2}$	$\frac{\sigma_1^2}{\sigma_2^2} \in \left[\frac{\hat{s}_1^2}{\hat{s}_2^2} F_{n_2-1; n_1-1; 1-\alpha/2}; \frac{\hat{s}_1^2}{\hat{s}_2^2} F_{n_2-1; n_1-1; \alpha/2} \right]$ donde $F_{n_2-1; n_1-1; 1-\alpha/2} = 1 / F_{n_1-1; n_2-1; \alpha/2}$
Binomiales con $x_1 \sim B(n_1, p_1); x_2 \sim B(n_2, p_2)$	$p_1 - p_2$	$p_1 - p_2 \underset{n_1, n_2 \rightarrow \infty}{\in} \left\{ \hat{p}_1 - \hat{p}_2 \pm z_{\alpha/2} \sqrt{\frac{\hat{p}_1 \hat{q}_1}{n_1} + \frac{\hat{p}_2 \hat{q}_2}{n_2}} \right\}$