

# TEMA 6: CONTRASTE DE HIPÓTESIS

10-05-2017

$X$  v.a.  $\{F_\theta: \theta \in \Theta\}$

PASO 1  $\rightarrow$  establecer hipótesis

$H_0: \theta \in \Theta_0$  —  $\theta = \theta_0 \vee \theta_1$   
 $H_1: \theta \in \Theta_1$  —  $\theta_0 \cap \theta_1 = \emptyset$

} Exhaustivas excluyentes

$H_0 \rightarrow$  hipótesis nula

$H_1 \rightarrow$  hipótesis alternativa

$H_0: \theta = \theta_0$  hipótesis simple

$\theta \geq \theta_0$   
 $\theta \leq \theta_0$   
 $\theta \neq \theta_0$

} hipótesis compuestas

$H_0: \theta = \theta_0$        $H_0: \theta = \theta_0$

$H_1: \theta = \theta_1$        $H_1: \theta \neq \theta_0$

2 COLAS

$H_0: \theta \leq \theta_0$        $H_0: \theta \geq \theta_0$

$H_1: \theta > \theta_0$        $H_1: \theta < \theta_0$

$H$  compuesta / compuesta una cola

PASO 2.1

Elegir el estadístico que resume adecuadamente la información

PASO 2.2.

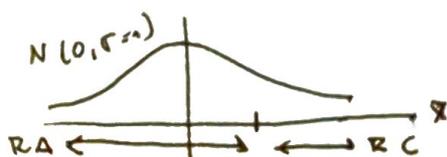
Se divide el espacio muestral en dos regiones

$\rightarrow$  R.A. Región de aceptación

$\rightarrow$  R.C. Región crítica o región de rechazo

$\rightarrow$  Def: Región crítica: Región del espacio muestral  $\mathcal{X}$  tal que si el punto muestral  $\vec{x} = (x_1, \dots, x_n) \in R.C.$  rechazaremos

$H_0$  (la hipótesis nula) en favor de  $H_1$  (hip. alternativa)



PASO 2.3.

Se toma una m.a.s. de tamaño  $n$ , se calcula el valor del estadístico para muestra

PASO 2.4.

Se resuelve el contraste.

→ Si el valor numérico cae en la región crítica, rechazamos  $H_0$  en favor de  $H_1$ .

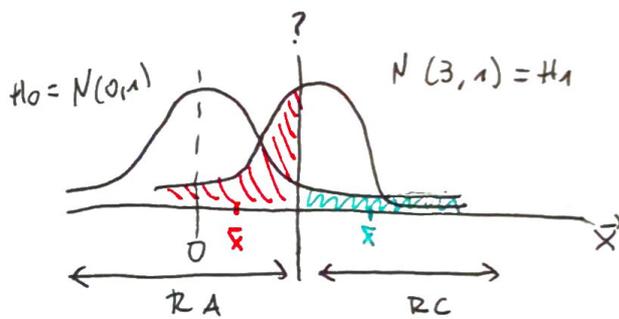
→ Si el valor numérico cae en la RA no se rechaza, no tengo evidencia para rechazar la hipótesis nula.

\* Ejemplo

$X \sim N(\theta, 1)$

$\begin{cases} H_0 : \theta = 0 \\ H_1 : \theta = 3 \end{cases}$

$T(\vec{x}) = \vec{x}$



$X = R.A. \cup R.C$

$\bar{x}, T(\bar{x}) = \bar{x}$

TEST. DE LA REGIÓN CRÍTICA

$\bar{x} \in RA \quad H_0 \checkmark$

$\bar{x} \in RC \quad H_1 \checkmark$

→ Puede ser que lo acerto

$H_0 \quad \theta \in \theta_0$

$H_1 \quad \theta \in \theta_1$

PRO

$\bullet \bar{x} \in RC$

$\bullet \bar{x} \in RA$

ERROR

TIPO I ← Rechazo hip. nula siendo

TIPO II ← Aceptar hip. nula siendo

→  $P_\theta$  (Error Tipo I) = Más grave Tipo I que Tipo II

$P_\theta$  (Rechazar  $H_0$  |  $H_0$  verdadera)  $\theta \in \theta_0$  } Inocente en la cárcel

→  $P_\theta$  (Error Tipo II) =

$P_\theta$  (Aceptar  $H_0$  |  $H_0$  falsa)  $\theta \in \theta_1$  } ← culpable en la calle

# FUNCIÓN POTENCIA DE UN TEST

Tamaño test

$$\beta(\theta) = P_{\theta}(R_c) \quad \theta \in \Theta = \Theta_0 \cup \Theta_1$$

$$\sup_{\theta \in \Theta_0} P(R_c)$$

→ Si  $\theta \in \Theta_0$ ,  $\beta(\theta) = P_{\theta}(R_c) = P(\text{Error tipo I})$

Quiero que  $\beta(\theta)$  sea pequeña cuando  $\theta \in \Theta_0$  ← Priorizo este error

→ Si  $\theta \in \Theta_1$ ,  $\beta(\theta) = 1 - P_{\theta}(\text{Error tipo II})$

Quiero que  $\beta(\theta)$  sea máximo para que  $P_{\theta}(\text{Error II})$  sea mínima

## \* Ejercicio

$$N(\mu, \sigma^2 = 125)$$

n.a.s tamaño  $n = 25$

Test:  $R_c$

$$\begin{cases} H_0: \mu = 60 & \theta \in \Theta_0 \\ H_1: \mu = 65 & \theta \in \Theta_1 \end{cases}$$

$$R_c = \{ \bar{x} : \bar{X} \geq \lambda \}$$

$$\lambda = 61,437$$

Calcula la función potencia del test

$$R_c = \{ \bar{x} \geq 61,437 \}$$

$$\beta(\mu) = P_{\mu}(\bar{x} \geq 61,437) = 1 - P(\bar{X} \leq 61,437) = 1 - P\left(\frac{\bar{X} - \mu}{\frac{\sigma}{\sqrt{n}}} \leq \frac{61,437 - \mu}{\frac{\sigma}{\sqrt{n}}}\right) = \dots$$

para  $\mu = 60; 65$

$$\bar{X} \sim N\left(\mu, \frac{\sigma^2}{n}\right) \quad \text{Tipificación}$$

$$= 1 - \phi\left(\frac{61,437 - \mu}{\sqrt{\frac{125}{25}}}\right)$$

↑  
test normal

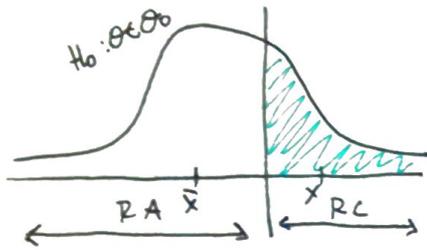
•  $\theta \in \Theta_0$   $\mu = 60$

$$\beta(60) = 1 - \phi(0,64) = 0,26$$

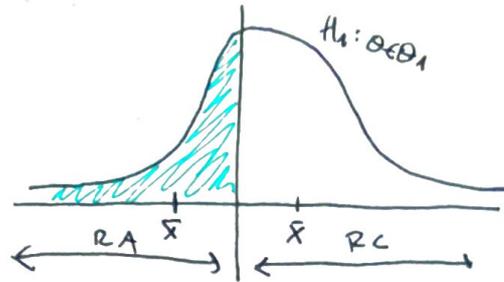
•  $\theta \in \Theta_1$   $\mu = 65$

$$\beta(65) = 1 - P(\text{TIPO II}) = 0,99$$

## ERROR TIPO I



## ERROR TIPO II



## FUNCIÓN POTENCIA (BIS)

Defino una función que me calcula probabilidad

$$\beta(\theta) = P(R.C.) \quad ; \quad \theta \in \theta_0 \cup \theta_1$$

$$\bullet \text{ Si } \theta \in \theta_0, \quad \beta(\theta) = P(R.C.) = P(\text{Error tipo I})$$

$$\bullet \text{ Si } \theta \in \theta_1, \quad \beta(\theta) = 1 - P(R.C.)$$

## \* Ejercicio

Construir R.C. con un nivel de significación de  $\alpha \in (0,1)$

$$\begin{cases} H_0: \mu = 60 \\ H_1: \mu = 65 \end{cases}$$

Una cota para los errores de tipo I

Establezco un límite (no quiero rechazar críticas que me den un error mayor que  $\alpha$ )

$$\boxed{P_{\theta}(R.C.) \leq \alpha} \quad \begin{array}{l} \text{Nivel} \\ \text{de} \\ \text{significación} \end{array}$$

$$\forall \theta \in \theta_0$$

$$R.C. = \{ \bar{x} \geq \lambda \} \quad \text{nivel de significación de } 0,1$$

$$\forall \theta \in \theta_0$$

$$\rightarrow \mu = 60 \quad P(R.C.) = P(\bar{x} \geq \lambda) \leq \alpha = 0,1$$

Tipificamos

↳

Tipificamos

$$\bar{X} \sim N(\mu=60, \sigma^2/n = \frac{125}{25} = 5)$$

$$P(\bar{X} \geq \lambda) \leq 0,1$$

$$P\left(\frac{\bar{X}-60}{\sqrt{5}} \geq \frac{\lambda-60}{\sqrt{5}}\right) \leq 0,1 \rightarrow P\left(Z \geq \frac{\lambda-60}{\sqrt{5}}\right) \leq 0,1$$

$\downarrow$   $\downarrow$   
 $N(0,1)$   $N(0,1)$

C  
A  
M  
B  
I  
O  
A

$$1 - \Phi\left(\frac{\lambda-60}{\sqrt{5}}\right) \leq 0,1 \quad ; \quad \Phi \text{ dist } N(0,1)$$

OTRA TABLA

$$0,9 \leq \Phi\left(\frac{\lambda-60}{\sqrt{5}}\right) \quad \text{miramos en la tabla}$$



Y vemos que  $X = 1,2816$  ;  $P(R.C) = 0,1$   
 $\theta \in \theta_0$

$$\rightarrow X = 1,29 \Rightarrow R.C. = \{ \bar{X} \geq 62,88 \}$$

$$X = \frac{\lambda-60}{\sqrt{5}} \Rightarrow \lambda = 62,8845$$

$$\rightarrow X = 1,28 \Rightarrow R.C. = \{ \bar{X} \geq 62,86 \}$$

$$X = \frac{\lambda-60}{\sqrt{5}} \Rightarrow \lambda = 62,8621$$

0,08	0,09	
0,847	0,905	1,2

$\lambda \geq 62,88$  Nivel de ~~tip~~ significación de 0,1

$\alpha = \sup_{\theta \in \theta_0} P_{\theta}(R.C)$

 Tamaño del test

**Def:** **P-valor** lo proporciona la muestra para un determinado contraste.

$$H_0: \theta \in \theta_0$$

$$H_1: \theta \in \theta_1$$

$$R.C. = \{ \bar{x} : T(\bar{x}) \geq c \}$$

T estadístico; c. conocida

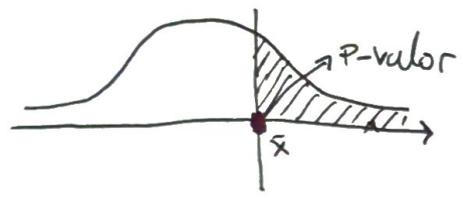
$$T(\bar{x}) = \bar{x}$$

$$c = T(\bar{x}) = \bar{x}$$

$$P_{\theta \in \theta_0} (R.C.) \leq \alpha$$

nivel de significación

$$\sup_{\theta \in \theta_0} P_{\theta} (T(\bar{x}) \geq T(\bar{x})) = P(x_1, \dots, x_n) \rightarrow \text{p-valor}$$



→ p-valor cualquier otra muestra  $\bar{x}'$  sea mayor igual que  $\bar{x}$   
 → Probabilidad de que  $\bar{x}$  sea mayor igual que el valor de mi muestra  $\bar{x}$  es muy pequeña

### TEST DE LA RAZÓN DE VEROSIMILITUDES

(Método para construir las R.C. del contraste)

Para contrastar  $H_0: \theta \in \theta_0$  frente a  $H_1: \theta \in \theta_1$  con m.a.s de tamaño n de una población  $f_{\theta}(x)$

$$R.C. = \{ \bar{x} : \lambda(\bar{x}) \leq k \}$$

$$\lambda(\bar{x}) = \frac{\sup_{\theta \in \theta_0} f_{\theta}(\bar{x})}{\sup_{\theta \in \theta_0 \cup \theta_1} f_{\theta}(\bar{x})}$$

k constante,  $0 \leq k \leq 1$

$\alpha$  valor fijo  $\alpha \in [0, 1]$

$$\sup_{\theta \in \theta_0} P(R.C.) \leq \alpha \text{ nivel de significación}$$

## \* Ejemplo

Binoulli ( $\theta$ )  $\theta \in [0, 1]$

$$\begin{cases} H_0: \theta \leq \theta_0 \\ H_1: \theta > \theta_0 \end{cases}$$

Test de la r.v. dado  $\alpha$

$$f_{\theta}(\vec{x}) = \theta^{\sum x_i} (1-\theta)^{n - \sum x_i}$$

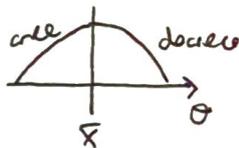
$$\ln f_{\theta}(\vec{x}) = (\sum x_i) \ln \theta + (n - \sum x_i) \ln (1-\theta)$$

$$\frac{\partial}{\partial \theta} \ln f_{\theta}(\vec{x}) = \frac{\sum x_i}{\theta} - (n - \sum x_i) \frac{1}{1-\theta} = 0 \quad \leftarrow \text{busca el máximo}$$

$$\frac{(\sum x_i)}{\theta} = (n - \sum x_i) \frac{1}{1-\theta}$$

Despejando:

$$\rightarrow \hat{\theta} = \bar{X}$$



• Numerador

$$\theta \leq \theta_0 ; \theta \leq \bar{X}$$

$$\theta > \bar{X} \rightarrow \text{Maximo } \bar{X}$$

$$\theta < \bar{X} \rightarrow \text{Maximo } \bar{\theta}$$

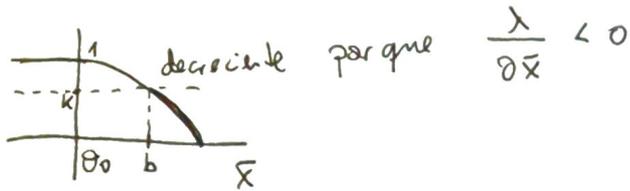
• Denominador

Siempre  $\bar{X}$

$$\rightarrow \lambda(x) = \begin{cases} 1 & \text{si } \theta_0 \geq \bar{X} \\ \frac{\theta_0^{n\bar{X}} (1-\theta_0)^{n-n\bar{X}}}{\bar{X}^{n\bar{X}} (1-\bar{X})^{n-n\bar{X}}} & \text{si } \theta_0 < \bar{X} \end{cases}$$



Estudio  $\lambda(\bar{x})$



$$R.C. \{ \bar{x} : \lambda(\bar{x}) \leq k \} \longrightarrow R.C. = \{ \bar{x} : \bar{x} \geq b \}$$

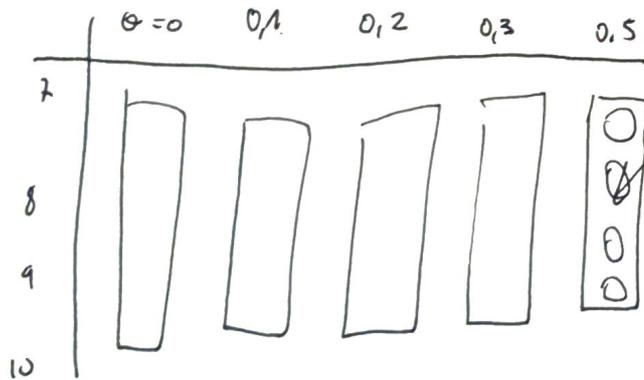
\* pregunta ej anterior

Caso  $\theta_0 = 1/2$ , 7 éxitos, 10 experimentos Bernoulli

$$p\text{-valor} = P_{\theta \in \theta_0} (\bar{X} \geq \bar{x}) = P_{\substack{\theta \in \theta_0 \\ n=10}} (\sum X_i \geq \sum x_i) = P(\underbrace{\sum X_i}_{\text{Binomial}(10, \theta)} \geq 7) =$$

$$= P(Z=7) + P(Z=8) + P(Z=9) + P(Z=10)$$

$H_0: \theta \leq 1/2$  miro en la tabla



cojo estos porque tienen el máximo y lo sumo

## \* Ejemplo

$$\text{Bernoulli } (n) \quad \begin{cases} H_0: \theta \in \theta_0 \\ \theta \in [0, 1] \\ \text{Test R.V.} \end{cases} \quad \begin{cases} f_{\theta}(\vec{x}) = \theta^{\sum_{i=1}^n x_i} (1-\theta)^{n - \sum_{i=1}^n x_i} \\ \ln f_{\theta}(\vec{x}) = \left(\sum_{i=1}^n x_i\right) \ln \theta + \left(n - \sum_{i=1}^n x_i\right) \ln(1-\theta) \end{cases}$$

$$\frac{\partial}{\partial \theta} \ln f_{\theta}(\vec{x}) = \frac{\sum x_i}{\theta} - \frac{(n - \sum x_i)}{1-\theta} = 0 \Rightarrow \hat{\theta} = \bar{x}$$

Máxim  
global  
derivada

→ Numerador  $\sup_{\theta \in \theta_0} f_{\theta}(\vec{x}) \quad \theta = \theta_0$

$$f_{\theta_0}(\vec{x}) = \theta_0^{\sum x_i} (1-\theta_0)^{n - \sum x_i}$$

→ Denominador  $\sup_{\theta \in \theta_0 \cup \theta_1} f_{\theta}(\vec{x})$  (máximos para  $\bar{x}$ )

$$\text{Si } \bar{x} = \theta_0 \Rightarrow \lambda(\bar{x}) = 1$$

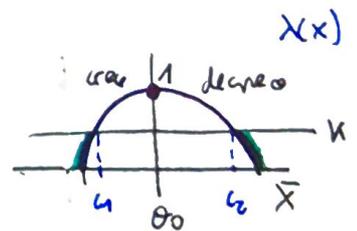
$$\text{Si } \bar{x} \neq \theta_0 \Rightarrow \lambda(\bar{x}) = \left(\frac{\theta_0}{\bar{x}}\right)^{n\bar{x}} \left(\frac{1-\theta_0}{1-\bar{x}}\right)^{n-n\bar{x}}$$

→ R.L. =  $\{\vec{x} : \lambda(\vec{x}) \leq k\}$

$$\ln \lambda(\vec{x}) = n\bar{x} (\ln \theta_0 - \ln \bar{x}) + n(1-\bar{x}) \cdot (\ln(1-\theta_0) - \ln(1-\bar{x}))$$

$$\frac{\partial}{\partial \bar{x}} \ln \lambda(\vec{x}) = n \cdot \ln\left(\frac{\theta_0}{1-\theta_0}\right) - n \ln\left(\frac{\bar{x}}{1-\bar{x}}\right) \leq 0 \Rightarrow n \ln\left(\frac{\theta_0}{1-\theta_0}\right) \leq n \ln\left(\frac{\bar{x}}{1-\bar{x}}\right)$$

$$\frac{\theta_0}{1-\theta_0} \leq \frac{\bar{x}}{1-\bar{x}} \Leftrightarrow \frac{1-\bar{x}}{\bar{x}} \leq \frac{1-\theta_0}{\theta_0} \Rightarrow \theta_0 \leq \bar{x}$$



$$\text{R.L.} = \{\bar{x} : \bar{x} \leq c_1, \bar{x} \geq c_2\}$$

→ Der  $c_1, c_2$

→ calcular  $k$

$$k = \lambda(c_1) = \lambda(c_2) \text{ constantes del test}$$

## \* Ejercicio

Calcular  $k$  para que el test tenga tamaño  $\alpha = 0,05$ ;  $\theta_0 = 1/2$ ;  $n = 1000$

Tamaño test

$$\hookrightarrow \sup_{\theta \in \Theta_0} P(R.C.)$$

Para calcular con una binomial

$$R.C. = \{ \bar{x} : \bar{x} \leq c_1, \bar{x} \geq c_2 \} = \{ \bar{x} : \underbrace{\sum_{i=1}^n x_i \leq nc_1}_{N(0,1)}, \sum_{i=1}^n x_i \geq nc_2 \}$$

por tanto  $C$  simétrico?

$$\rightarrow 0,05 = \sup_{\theta \in \Theta_0} P_{\theta}(R.C.) = P_{\theta=1/2} (|\bar{X} - 1/2| \geq |c - 1/2|)$$


→ TEOREMA CENTRAL LÍMITE (muestra grande  $n \gg 30$ )

$$\frac{\bar{X} - E[\bar{X}]}{\sqrt{V[\bar{X}]}} \xrightarrow{d} N(0,1)$$

$$E[\bar{X}] = \theta$$

$$V[\bar{X}] = \frac{1}{n} V(X) = \frac{1}{n} \theta(1-\theta) \Rightarrow \sqrt{V(\bar{X})} = 0,0158$$

$$P_{\theta=1/2} \left( \underbrace{\frac{|\bar{X} - 0,5|}{0,0158}}_{N(0,1)} \geq \frac{|c - 0,5|}{0,0158} \right) = 1 - P \left( \frac{|\bar{X} - 0,5|}{0,0158} \leq \frac{\overbrace{|c - 0,5|}^d}{0,0158} \right) = 1 - P(-d \leq N(0,1) \leq d)$$

$$= 1 - [P(N(0,1) \leq d) - \underbrace{P(N(0,1) \leq -d)}_{\substack{P(N(0,1) \geq d) \\ (1 - P(N(0,1) \leq d))}}] = 1 - 2P(N(0,1) \leq d)$$

$$= 1 - 2P(N(0,1) \leq d) + 1 = -2P(N(0,1) \leq d) + 2 = 0,05$$

$$P(N(0,1) \leq d) = 0,975 \Rightarrow \Phi(d) = 0,975 ; \boxed{d = 1,96}$$

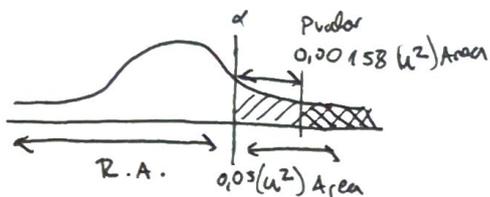
$$d = \frac{c - 0,5}{0,0158} \Rightarrow c = 0,53$$

$$\boxed{R.C. = \{ \bar{x} \mid \bar{x} \leq -0,53, \bar{x} \geq 0,53 \}}$$

→ Calcular el P-valor

Calcular el p-valor

$$\begin{aligned}
 \text{p-valor} &= \sup_{\theta \in \Theta_0} P_{\theta}(|\bar{X} - 0.5| \geq |\bar{X} - 0.5|) = P_{\sigma=1/2} \left( \frac{|\bar{X} - 0.5|}{0.0158} \geq \frac{|\bar{X} - 0.5|}{0.0158} \right) = \\
 &= P(|N(0,1)| \geq 3.16) = 2 - 2P(N(0,1) \leq 3.16) = \\
 &= 2 - \Phi(3.16) = 2 - 2 \cdot 0.99921 = 0.00158
 \end{aligned}$$



Hay mucha distancia entre  $\alpha$  y el p-valor eso es bueno para rechazar la hipótesis "Tenemos mucha evidencia de que podemos rechazar"

**Ejemplo**

$H_0: \theta \leq 1$  Mas.  $n=1$   $x=3$  Poisson ( $\theta$ )

$H_1: \theta > 1$

$\theta \in (0, 2)$

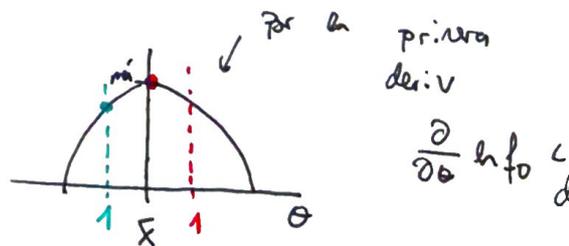
Función densidad poisson

$$f_{\theta}(\vec{x}) = \frac{e^{-n\theta} \theta^{\sum x_i}}{\prod_{i=1}^n x_i} \quad x=0,1,2,\dots$$

$$\ln f_{\theta}(\vec{x}) = -\ln \prod x_i - n\theta + (\sum x_i) \ln \theta$$

$$\frac{\partial}{\partial \theta} \ln f_{\theta}(\vec{x}) = -n + \frac{\sum x_i}{\theta} = 0 \Rightarrow n = \frac{\sum x_i}{\theta} \Rightarrow \hat{\theta} = \bar{x}$$

$$\frac{\partial^2}{\partial \theta^2} \ln f_{\theta}(\vec{x}) = -\frac{\sum x_i}{\theta^2} < 0 \quad \hat{\theta} = \bar{x} \text{ máximo}$$



Numerador:  $\sup_{\theta \in \Theta_0} f_{\theta}(\vec{x}) =$

$$\begin{cases}
 \theta \leq 1 \\
 (0, 1]
 \end{cases}
 \begin{cases}
 \Sigma: \bar{x} > 1 \text{ max es } 1 \\
 \Sigma: \bar{x} < 1 \text{ máx es } \bar{x}
 \end{cases}$$

Denominador:  $\bar{x}$

Construimos  $\lambda(x)$

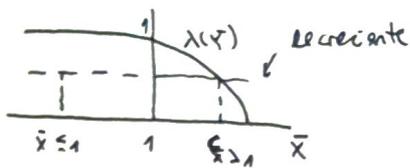
$$\lambda(x) = \begin{cases} \text{S: } \bar{x} \leq 1 & 1 \\ \text{S: } \bar{x} > 1 & \frac{e^{-n \cdot 1} (2^{x_i})}{e^{-n \bar{x}} \bar{x} (2^{x_i})} \end{cases}$$

substituto por  $\frac{1}{\bar{x}}$  na f.c.

$$\lambda(\bar{x}) = \begin{cases} \text{S: } \bar{x} < 1 & 1 \\ \text{S: } \bar{x} \geq 1 & \frac{e^{-n(\bar{x}-1)}}{\bar{x}^{n\bar{x}}} \end{cases}$$

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Estudamos  $\lambda(\bar{x})$ :



$$R_C = \{ \bar{x} / \lambda(\bar{x}) \leq k \} = \{ \bar{x} / \bar{x} \geq c \} = \{ X : X \geq c \}$$

Poisson( $\theta$ )

$$\lambda(c) = k$$

General:  $\sum x_i \sim \text{Poisson}(n\theta)$

$x_i \sim \text{Poisson}(\theta)$

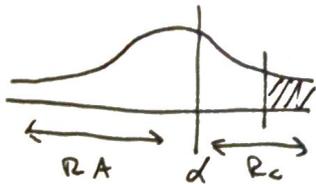
Função  $\alpha$ ,  $\alpha = \sup_{\theta \in \Theta_0} P(R_C)$  ; p-value:  $\sup_{\theta \in \Theta_0} P(X \geq x) = \sup_{\theta \in (0,1]} P(X \geq 3)$

$$P(X \geq 3) = 1 - P(X \leq 3) = 1 - [P(X=0) + P(X=1) + P(X=2)]$$

$$\sup_{\theta \in (0,1]} : P(X=0 | \theta=1) = 0.3679$$

$$P(X=1 | \theta=1) = 0.3679$$

$$P(X=2 | \theta=1) = 0.1839$$



$$\rightarrow \sup_{\theta \in \Theta_0} P(X \geq 3) = 0.0803$$

## \* Ejemplo

$x_1, \dots, x_n$  población gamma ( $\alpha, p=1$ )  $a \in (0, \infty)$

Test ZV

$$\begin{cases} H_0: a = a_0 \\ H_1: a \neq a_0 \end{cases}$$

$$Z.C = \{ \bar{x} / \chi(\bar{x}) \leq k \}$$

$$\lambda(\bar{x}) = \frac{\sup_{\theta \in \Theta} f_{\theta}(\bar{x})}{\sup_{\theta \in \Theta} f_{\theta}(\bar{x})}$$

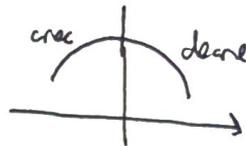
$$f_{\theta}(\bar{x}) = \frac{a^n}{(\Gamma(p))^n} e^{-a \sum_1^n x_i} \prod_1^n x_i^{p-1}$$

$$\ln f_{\theta}(\bar{x}) = n \ln a - n \ln(\Gamma(p)) - a \sum_1^n x_i + (p-1) \sum_1^n \ln x_i$$

$$\frac{\partial}{\partial a} \ln f_{\theta}(\bar{x}) = \frac{n}{a} - \left( \sum_1^n x_i \right) = 0 \Rightarrow \frac{n}{a} = \sum_1^n x_i \Rightarrow a = \frac{1}{\bar{x}}$$

$$\frac{n}{a} < \sum_1^n x_i \quad \text{decreciente}$$

$$\frac{1}{\bar{x}} < a \quad \text{decreciente}$$



• Numerador:  $a = a_0$

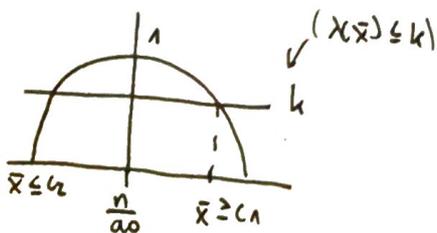
• Denominador:  $1/\bar{x}$

$$a_0 \neq \frac{1}{\bar{x}} \rightarrow \lambda(\bar{x}) = \frac{a_0^n e^{-a_0 \bar{x}} \prod_1^n x_i^{p-1}}{\left(\frac{1}{\bar{x}}\right)^n e^{-\sum_1^n x_i} \prod_1^n x_i^{p-1}} = (a_0 \bar{x})^n e^{n - a_0 \bar{x}}$$

$$\Rightarrow \lambda(\bar{x}) = \begin{cases} 1 & \text{si } 1/\bar{x} = a_0 \\ (a_0 \bar{x})^n e^{n - a_0 \bar{x}} & \text{si } 1/\bar{x} \neq a_0 \end{cases}$$

$$\ln \lambda(\bar{x}) = n \ln(a_0 \bar{x}) + n - a_0 \bar{x}$$

$$\frac{\partial}{\partial \bar{x}} \ln \lambda(\bar{x}) = \frac{n a_0}{a_0 \bar{x}} - a_0 \Rightarrow \frac{n}{\bar{x}} = a_0 \Rightarrow \bar{x} = \frac{n}{a_0}$$



$$Z.C = \{ \bar{x} : \bar{x} \geq c_1, \bar{x} \leq c_2 \}$$

$$X_i \sim \text{gamma}(a, p=1)$$

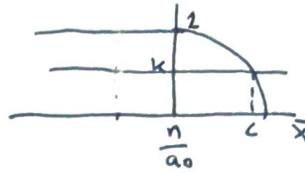
$$\sum X_i \sim \text{gamma}(a, n)$$

$$\frac{1}{n} \sum X_i \sim \text{gamma}(na, n)$$

\* Ejemplo

$$\begin{cases} H_0: a \geq a_0 \\ H_1: a < a_0 \end{cases}$$

$$\begin{cases} \text{Si } a_0 \leq \frac{n}{\bar{x}}, & \lambda(\bar{x}) = 1 \\ \text{Si } a_0 > \frac{n}{\bar{x}}, & \lambda(\bar{x}) = (a_0 \bar{x})^n e^{-a_0 \bar{x}} \end{cases}$$



$$R.C. = \{ \bar{x} : \bar{x} \geq c \}$$

CONTRASTES SOBRE LA MEDIA POBLACIONAL NORMAL:

\* Ejemplo

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

$$\begin{cases} H_0: \mu = \mu_0 \\ H_1: \mu \neq \mu_0 \end{cases}$$

$$f_{\theta}(\vec{x}) = \frac{1}{(\sqrt{2\pi})^n \sigma^n} e^{-\frac{1}{2\sigma^2} \sum_{i=1}^n (x_i - \mu)^2}$$

$$\bar{x} \sim N(\mu, \sigma^2/n)$$

$$\leftarrow \ln f_{\theta}(\vec{x}); \frac{\partial}{\partial \mu} \ln(f_{\theta}(\vec{x})) = 0 \rightarrow \hat{\theta} = \bar{x}$$

$$\frac{\bar{x} - \mu}{\sigma/\sqrt{n}} \sim N(0, 1)$$

$$\lambda(x) = \begin{cases} 1 & \text{Si } \bar{x} = \mu_0 \\ e^{-\frac{n}{2\sigma^2} (\mu_0 - \bar{x})^2} & \text{Si } \bar{x} \neq \mu_0 \end{cases}$$

$$\ln \lambda(\bar{x}) = -\frac{n}{2\sigma^2} (\mu_0 - \bar{x})^2$$

$$\frac{\partial}{\partial x} \ln \lambda(\bar{x}) = -\frac{n}{\sigma^2} (\mu_0 - \bar{x}) = 0 \Rightarrow \frac{n\mu_0}{\sigma^2} = \frac{n\bar{x}}{\sigma^2} \Rightarrow \mu_0 = \bar{x}$$

$$R.C. = \{ \bar{x} : |\bar{x} - \mu_0| \geq |c - \mu_0| \}$$

$$1 = \sup_{\theta \in \theta_0} P(R.C.) = \sup_{\theta \in \theta_0} P\left( \underbrace{\sqrt{n} \frac{(\bar{x} - \mu_0)}{\sigma}}_z \geq \underbrace{\frac{|c - \mu_0| \sqrt{n}}{\sigma}}_d \right) = P(|Z| \geq d) = 1 - P(|Z| \leq d)$$

$Z \sim N(0, 1)$

$$= 1 - [P(-d \leq Z \leq d)] = 2 - 2\phi(d) \quad ; \quad \phi(d) = 1 - \alpha/2 \quad \cancel{1 - \alpha/2}$$

$d = Z_{\alpha/2}$

\* Ejemplo

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

$$\begin{cases} H_0: \mu = \mu_0 \\ H_1: \mu \neq \mu_0 \end{cases}$$

$$\hat{\theta} = \bar{x}$$

$$\lambda(\bar{x}) = \begin{cases} 1 & \text{si } \bar{x} = \mu_0 \\ e^{-\frac{n(\mu_0 - \bar{x})^2}{2\sigma^2}} & \text{si } \bar{x} \neq \mu_0 \end{cases}$$

$$\frac{\bar{x} - \mu}{\sigma/\sqrt{n}} \sim N(0, 1)$$

$$(n-1) \frac{s^2}{\sigma^2} \sim \chi^2_{n-1}$$

$$\frac{\bar{x} - \mu}{s/\sqrt{n}} \sim t_{n-1}$$

$$R.C. = \left\{ \bar{x} : \sqrt{n} \frac{|\bar{x} - \mu_0|}{s} \geq \frac{|c - \mu_0|}{s} \right\}$$

Igual que anterior con  $d = t_{n-1, \alpha/2}$

\* Ejercicio

Utilizar el test de verosimilitud para hallar la R.C. dado población  $N(\theta, \sigma)$  m.a.s. Test RV para contrastar

23-05-2017

$$\begin{cases} H_0: \theta = 0 \\ H_1: \theta > 0 \end{cases}$$

$$f_{\theta}(x) = \frac{1}{(\sqrt{2\pi})^n} e^{-\frac{1}{2} \sum_1^n (x_i - \theta)^2}$$

$$\ln f_{\theta}(\bar{x}) = -n \ln(\sqrt{2\pi}) - \frac{1}{2} \sum (x_i - \theta)^2$$

$$\frac{\partial \ln f_{\theta}(\bar{x})}{\partial \theta} = -\frac{1}{2} \sum_1^n 2(x_i - \theta) = \sum_1^n x_i - n\theta = 0$$

$$\hat{\theta} = \frac{\sum_1^n x_i}{n} = \bar{x}$$

• Numerador:  $\sigma = 0$

$$\frac{1}{\sqrt{2\pi}} n e^{-\frac{1}{2} \sum x_i^2}$$

• Denominador  $\frac{1}{\sqrt{2\pi}} n e^{-\frac{1}{2} \sum (x_i - \bar{x})^2}$

$$\lambda(\bar{x}) = \begin{cases} 1 & \text{si } \bar{x} = 0 \\ e^{-\frac{1}{2} \sum x_i^2 - 2(x_i - \bar{x})^2} & \text{si } \bar{x} \neq 0 \end{cases}$$

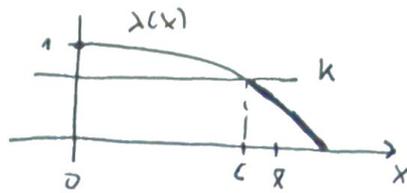
$$= e^{-n\bar{x}^2/2}$$

↗

✶

$$\lambda(x) = e^{-\frac{1}{2} n \bar{x}^2}$$

$$\ln \lambda(x) = -\frac{1}{2} n \bar{x}^2$$



$$\frac{\partial}{\partial x} \ln \lambda(x) = -\frac{2n}{2} \bar{x} \Rightarrow -n \bar{x} < 0; \bar{x} > 0 \text{ decreciente}$$

$$R_c = \{ \bar{x} / \lambda(\bar{x}) \leq k \} = \{ \bar{x}^2 / \bar{x} \geq c \} \rightarrow \lambda(c) = k$$

→ Función potencia del test

$$\beta(\theta) = P_{\theta \in \Theta} (R_c) = P_{\theta \in \Theta} (\bar{X} \geq c) \stackrel{\text{for las tablas}}{=} 1 - P_{\theta \in \Theta} (\bar{X} \leq c) =$$

$$\bar{X} \sim N(\theta, \frac{1}{n})$$

$$= 1 - P_{\theta \in \Theta} ((\bar{X} - \theta) \sqrt{n} \leq (c - \theta) \sqrt{n}) =$$

$\downarrow$   
 $N(0,1)$

$$= 1 - \Phi((c - \theta) \sqrt{n}) \rightarrow \Phi \notin N(0,1)$$

$$\beta(0.1) = 0.95$$

$$\beta(-0.1) = 0.05$$

$$\theta = 0.1 \rightarrow 0.95 = 1 - \Phi((c - 0.1) \sqrt{n}) \Rightarrow \Phi((c - 0.1) \sqrt{n}) = 0.05$$

$$(c - 0.1) \sqrt{n} = -1.6449$$

$$\theta = -0.1 \rightarrow 0.05 = 1 - \Phi((c + 0.1) \sqrt{n}) \Rightarrow \Phi((c + 0.1) \sqrt{n}) = 0.95$$

$$(c + 0.1) \sqrt{n} = 1.6449$$

$$\Rightarrow 2c \sqrt{n} = 0 \Rightarrow c = 0$$

$$\begin{cases} \lambda(c) = k \\ k = 1 \end{cases}$$

\* Ejercicio

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

Fijado nivel de significación  $\alpha$

$H_0 = \sigma^2 = 64,25$

$(n-1) S^2 / \sigma_0^2 \sim \chi^2_{n-1}$

$H_1 : \sigma^2 \neq 64,25$

$n = 20, \bar{x} = 81,42$

$S = 87,47$

$f_{\theta}(\vec{x})$

$\ln f_{\theta}(\vec{x}) = -n \ln \sqrt{2\pi} - n \ln \sigma - \frac{1}{2\sigma^2} \sum (x_i - \mu)^2$

$\frac{\partial}{\partial \sigma} \ln f_{\theta}(\vec{x}) = \frac{-1}{\sigma} + \frac{1}{\sigma^3} \sum (x_i - \mu)^2 = 0 \Rightarrow \hat{\sigma}^2 = \frac{(n-1)}{n} S^2$

$\frac{n}{\sigma} \leq \frac{1}{\sigma^3} \sum (x_i - \mu)^2$  decreciente

$\sigma^2 \geq \frac{1}{n} (n-1) S^2 = \hat{\sigma}^2$



• Numerador:  $\theta = 64,25 \rightarrow f_{\theta} = 64,25(\vec{x})$

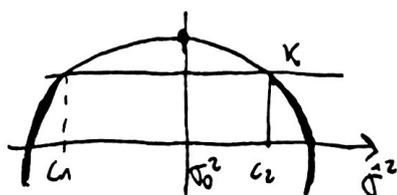
• Denominador:  $\theta = \hat{\sigma}^2 = f_{\theta} = \hat{\sigma}^2(\vec{x})$

$\lambda(\hat{\sigma}^2) = \begin{cases} 1 & \text{si } \hat{\sigma}^2 = 64,25 \\ \left(\frac{\sigma_0^2}{\hat{\sigma}^2}\right)^n e^{-\frac{1}{2} \left(\frac{\hat{\sigma}^2}{\sigma_0^2} - 1\right)} & \text{si } \hat{\sigma}^2 \neq \sigma_0^2 \end{cases}$

$\ln \lambda(\hat{\sigma}^2)$

$\frac{\partial}{\partial \hat{\sigma}^2} \ln \lambda(\hat{\sigma}^2) = \frac{n}{\hat{\sigma}^2} - n \frac{\hat{\sigma}^2}{\hat{\sigma}^4} = 0 \Rightarrow \hat{\sigma}^2 = \sigma_0^2$

decreciente  $\sigma_0^2 \leq \hat{\sigma}^2$



$$R.C = \{ \bar{x} : \hat{\sigma}^2 \leq L_1, \hat{\sigma}^2 \geq L_2 \}$$

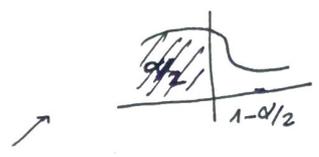
→  $\alpha$  tamaño test

$$\sup_{\theta \in \theta_0} P(R.C)$$

$$\theta = 64.25$$

$$P(\hat{\sigma}^2 \leq L_1, \hat{\sigma}^2 \geq L_2)$$

$$\begin{cases} P(\hat{\sigma}^2 \leq L_1) = \alpha/2 \\ P(\hat{\sigma}^2 \geq L_2) = \alpha/2 \end{cases} \quad \left\{ \begin{array}{l} \text{como no dice} \\ \text{nada se toma} \\ \text{colas iguales} \end{array} \right.$$



$$P\left( (n-1) \frac{s^2}{\sigma_0^2} \leq \frac{L_1}{\sigma_0^2} \right) = \alpha/2 \quad \frac{n L_1}{\sigma_0^2} = \chi_{n-1, 1-\alpha/2}^2$$

$\underbrace{\frac{n L_1}{\sigma_0^2}}_{\chi_{n-1}^2}$  *en paso aquí*

(Mirar tablas si hubiera valores)