

Likelihood Ratio Test

Want to test

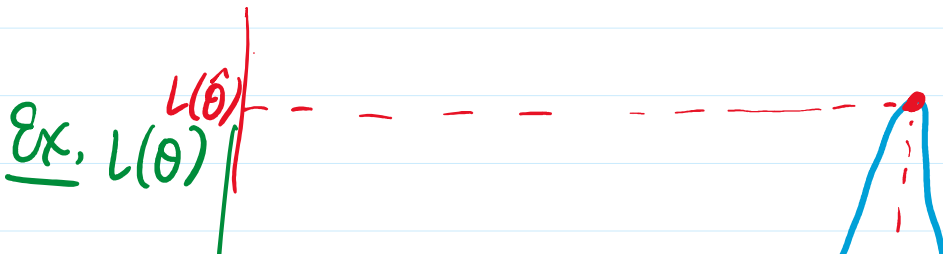
$$H_0: \theta \in \Theta_0 \quad \text{v.} \quad H_a: \theta \in \Theta_a$$

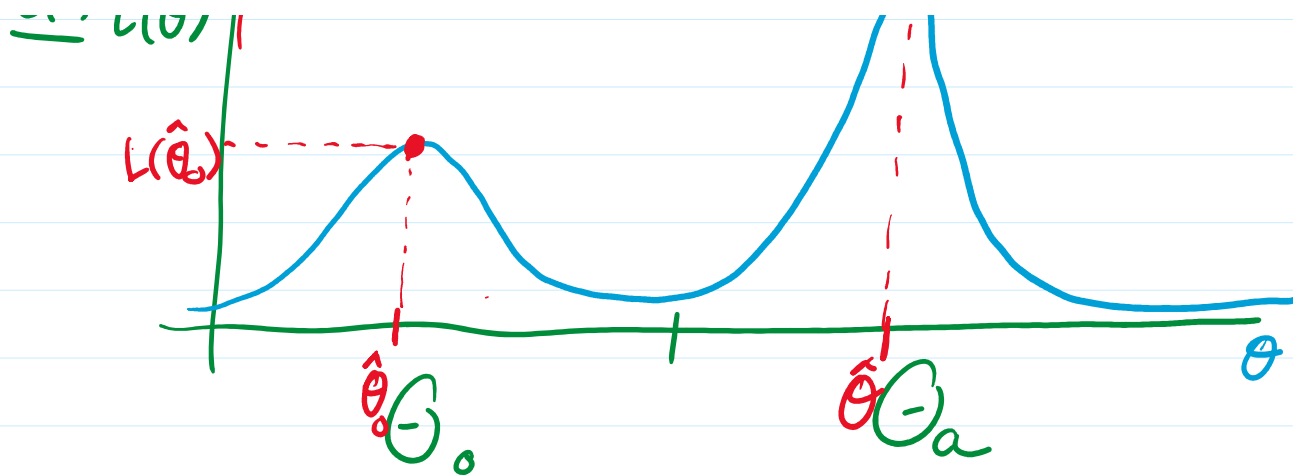
We define the likelihood ratio test (LRT) statistic as

$$\begin{aligned} \lambda(\tilde{x}) &= \frac{\max_{\theta \in \Theta_0} L(\theta)}{\max_{\theta \in \Theta} L(\theta)} = \frac{\text{max of } L \text{ over } \Theta_0}{\text{max of } L \text{ overall}} \\ &= \frac{L(\hat{\Theta}_0)}{L(\hat{\Theta})} \leq 1 \end{aligned}$$

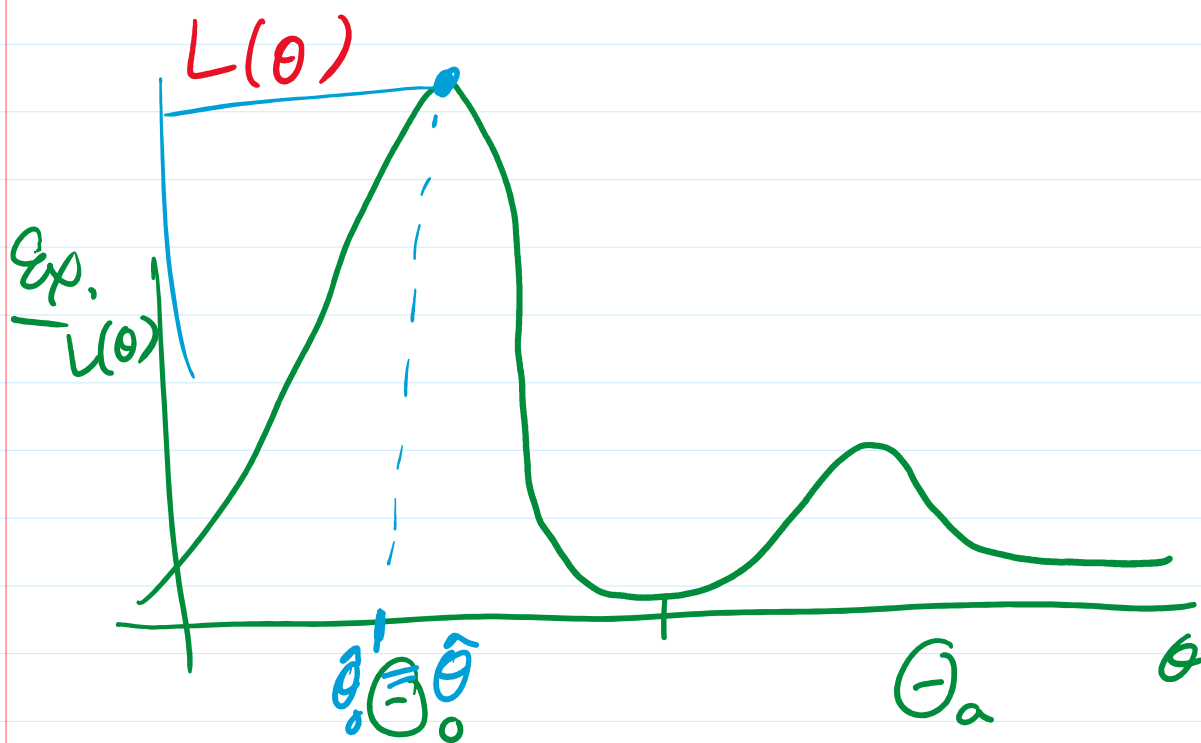
where $\hat{\Theta}_0$ = MLE restricted to Θ_0

$\hat{\Theta}$ = MLE



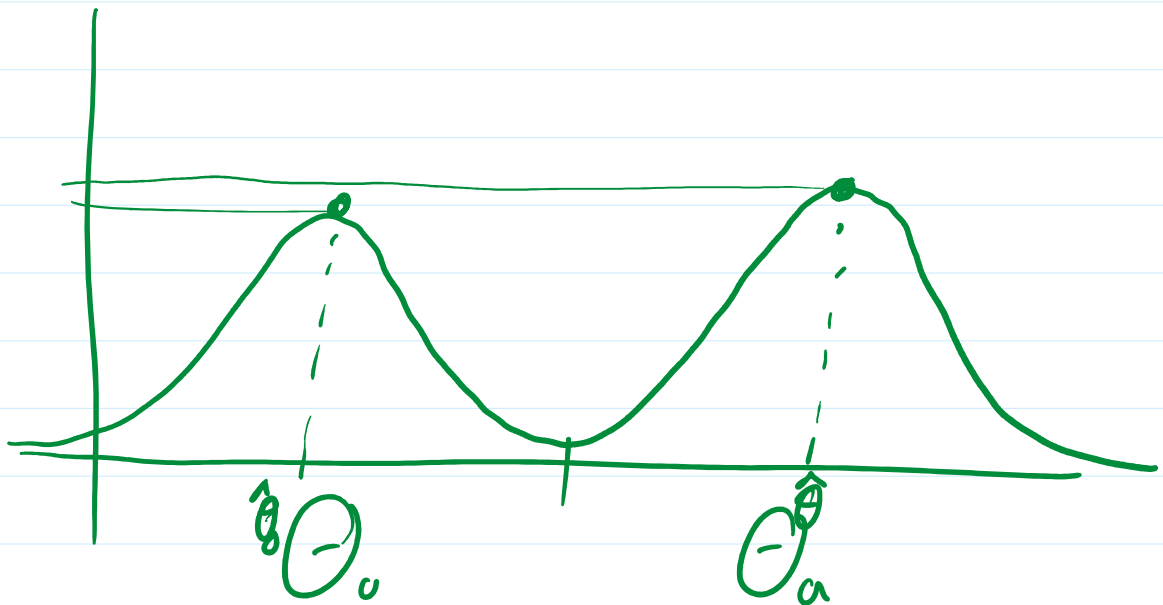


$$\lambda = \frac{L(\hat{\theta}_0)}{L(\hat{\theta})} \ll 1 \leadsto \text{probably reject}$$



$$\text{So } \lambda = \frac{L(\hat{\theta}_0)}{L(\hat{\theta})} = 1 \leadsto \text{probably don't reject}$$

hard case:



$$\lambda = \frac{L(\hat{\theta}_0)}{L(\hat{\theta})} < 1 \text{ but } \approx 1$$

The LRT says I should reject
when

$$\lambda = \frac{L(\hat{\theta}_0)}{L(\hat{\theta})} \leq c$$

$\uparrow \quad 0 \leq c \leq 1$

c is a thresh I choose to balance
type I and II errors.

TP is small don't reject easily

If c small, don't reject easily,
less type I, more type II

If c is large, reject more easily,
more type I, less type II.

The rejection region for LRT is

$$R = \{x : \lambda(x) \leq c\}.$$

Ex. $X_n \stackrel{iid}{\sim} N(\theta, \sigma^2)$ \uparrow Known

$$H_0: \theta \leq a \quad \text{v.} \quad H_a: \theta > a$$

Let's derive the LRT

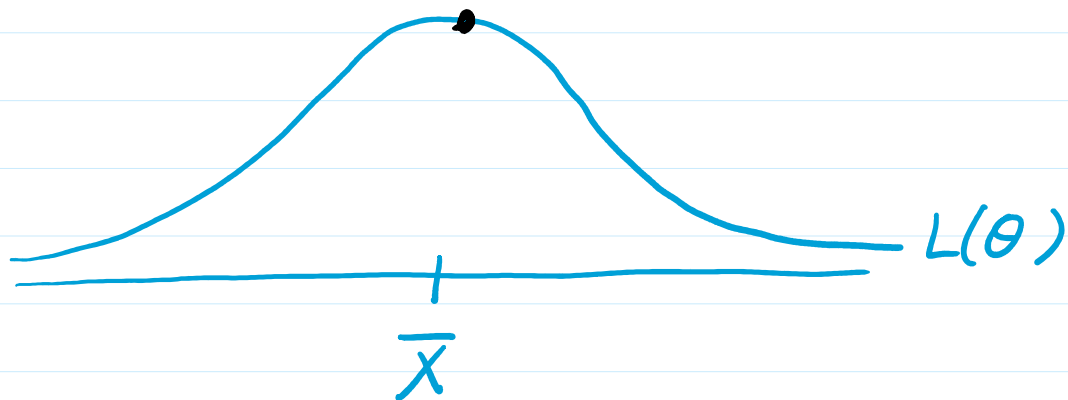
$$L(\theta) = \prod_n \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{1}{2\sigma^2} (x_n - \theta)^2\right)$$

$$= \dots (-1)^{-N/2} (-1)^{-N/2} \left(-\frac{1}{2} \int (x-0)^2 \right)$$

$$= (2\pi)^{-N/2} (\sigma^2)^{-N/2} \exp \left(-\frac{1}{2\sigma^2} \sum_n (\chi_n - \theta)^2 \right)$$

quadratic in θ

looks like $e^{-(\theta - \bar{x})^2}$



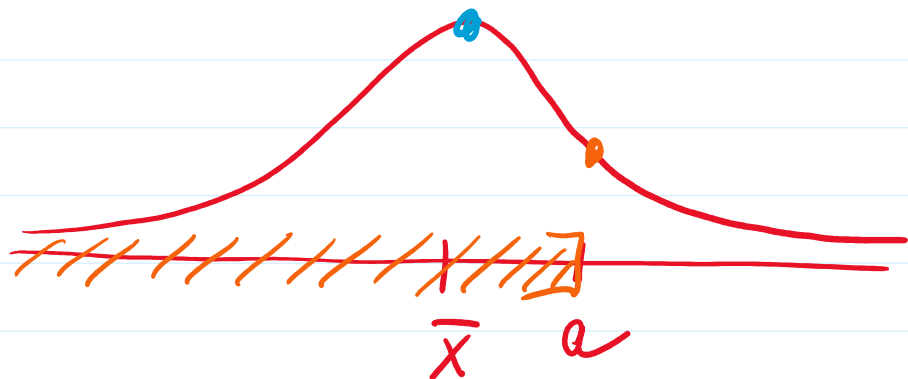
Clearly, $\hat{\theta} = \bar{x}$.

Need to find $\hat{\theta}_0 = \max_{\theta \in (-\infty, a]} L(\theta)$

θ_0

Two cases:

$\bar{x} \leq a$

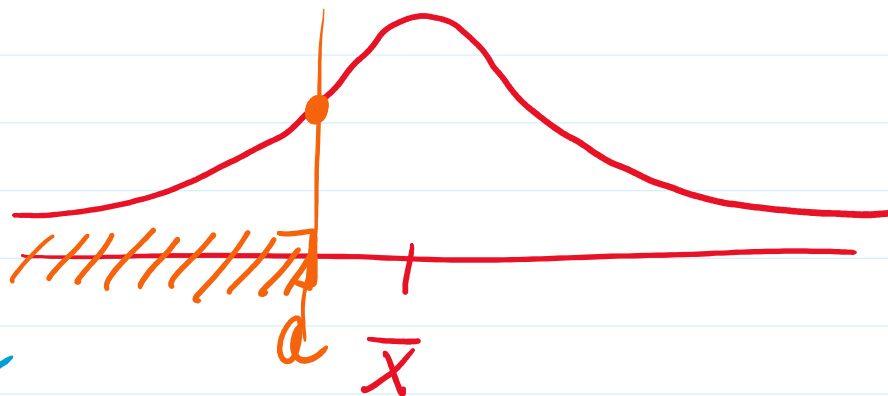


Since $\bar{x} \leq a$ then $\hat{\theta}_0 = \bar{x}$

Since $\lambda = 0$ then $\hat{\theta}_0 = \lambda$

$$\bar{x} > a$$

$$S_0 \quad \hat{\theta}_0 = a$$



Overall then

$$\hat{\theta}_0 = \begin{cases} \bar{x} & \text{if } \bar{x} \leq a \\ a & \text{if } \bar{x} > a \end{cases}$$

$$H_0: \theta \leq a$$

$$\lambda(\bar{x}) = \frac{L(\hat{\theta}_0)}{L(\hat{\theta})} = \begin{cases} \frac{L(\bar{x})}{L(\bar{x})} = 1 & \text{never reject } \bar{x} \leq a \\ \frac{L(a)}{L(\bar{x})} & \bar{x} > a \end{cases}$$

LRT says reject when $\lambda \leq c < 1$

i.e. reject when

$$L(a) < c [L(a) \leq c L(\bar{x})]$$

$$\frac{L(a)}{L(\bar{x})} \leq C \quad \left[L(a) \leq C L(\bar{x}) \right]$$

and $\bar{x} > a.$

$$\lambda = \frac{L(a)}{L(\bar{x})} = \frac{\cancel{(2\pi\sigma^2)^{-N/2}} \exp\left(-\frac{1}{2\sigma^2} \sum_n (x_n - a)^2\right)}{\cancel{(2\pi\sigma^2)^{-N/2}} \exp\left(-\frac{1}{2\sigma^2} \sum_n (x_n - \bar{x})^2\right)}$$

$$= \frac{\exp\left(-\frac{1}{2\sigma^2} \sum_n (x_n^2 - 2ax_n + a^2)\right)}{\exp\left(-\frac{1}{2\sigma^2} \sum_n (x_n^2 - 2x_n\bar{x} + \bar{x}^2)\right)}$$

$$\rightarrow \exp\left(-\frac{1}{2\sigma^2} \sum_n (x_n^2 - 2x_n\bar{x} + \bar{x}^2)\right)$$

$$\frac{e^{a+b}}{e^{a+c}} = \frac{e^a e^b}{e^a e^c}$$

$$= \exp\left(-\frac{1}{2\sigma^2} (-2aN\bar{x} + Na^2 + 2N\bar{x}^2 - N\bar{x}^2)\right)$$

$$= \exp\left(-\frac{N}{2\sigma^2} (-2a\bar{x} + a^2 + \bar{x}^2)\right)$$

$$\lambda = \exp\left(-\frac{N}{2\sigma^2} (\bar{x} - a)^2\right)$$

... \ 2\sigma^2 ... /

and LRT says reject when $\lambda \leq C$

$$\Leftrightarrow \exp\left(-\frac{N}{2\sigma^2}(\bar{X}-a)^2\right) \leq C$$

$$\Leftrightarrow -\frac{N}{2\sigma^2}(\bar{X}-a)^2 \leq \log C$$

$$\Leftrightarrow \frac{N}{\sigma^2}(\bar{X}-a)^2 \geq -2 \log C$$

when $\bar{X} > a$ then $\bar{X} - a > 0$

↓

$$\Leftrightarrow \frac{\sqrt{N}}{\sigma}(\bar{X}-a) \geq \underbrace{\sqrt{-2 \log C}}_{c^*}$$

$$\Leftrightarrow \boxed{\frac{\bar{X}-a}{\sigma/\sqrt{N}} \geq c^*}$$

$$\Leftrightarrow \boxed{\bar{X} \geq a + \frac{\sigma}{\sqrt{N}} c^*}$$

$$\begin{array}{|l} H_0: \theta \leq a \\ H_a: \theta > a \end{array}$$

LRT \Leftrightarrow reject when \bar{X} is
 $a + c^* \text{Sid.}(\bar{X})$

$$a + c^* \text{sid.}(\bar{X})$$

What's a reasonable value for c^* ?

Maybe want to choose c^* so that this is a size α test

i.e.

$$\max \text{ type I error} = \alpha$$

i.e.

$$\max_{\theta \in \Theta_0} P_{\theta}(\text{reject}) = \alpha$$

$\underbrace{\hspace{10em}}_{\beta(\theta)}$

$(-\infty, a]$ \nearrow

$$\beta(\theta) = P_{\theta}(\lambda \leq c) = P_{\theta}\left(\frac{\bar{X} - a}{\sigma/\sqrt{N}} \geq c^*\right)$$

$$= P_{\theta}\left(\underbrace{\frac{\bar{X} - a}{\sigma/\sqrt{N}} + \frac{a - \theta}{\sigma/\sqrt{N}}}_{\sim N(0,1)} \geq c^* + \frac{a - \theta}{\sigma/\sqrt{N}}\right)$$

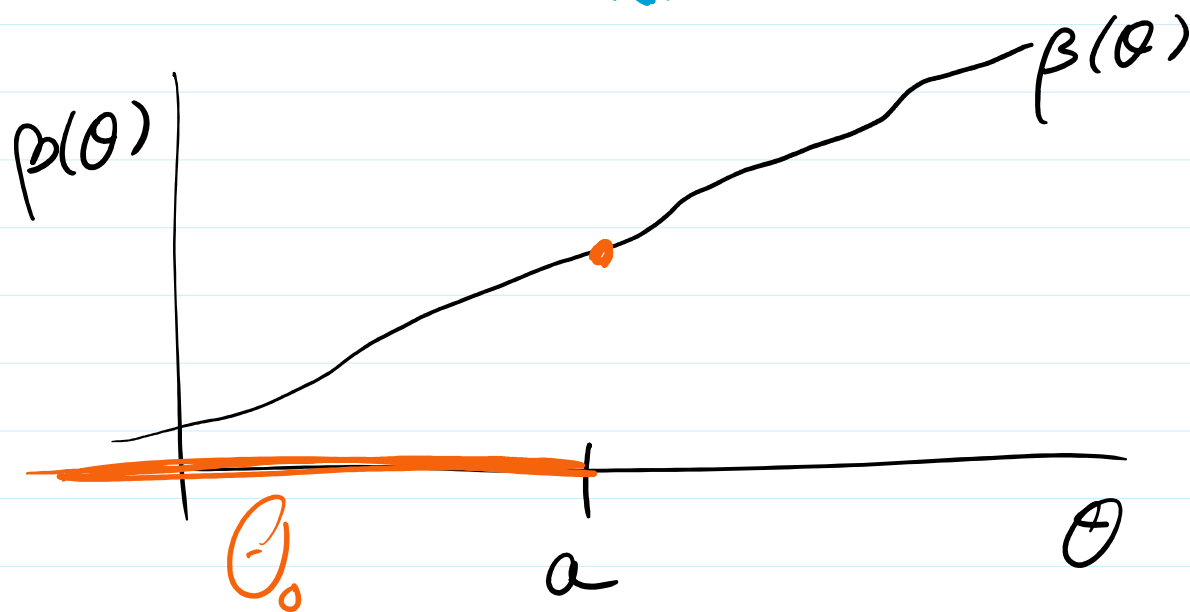
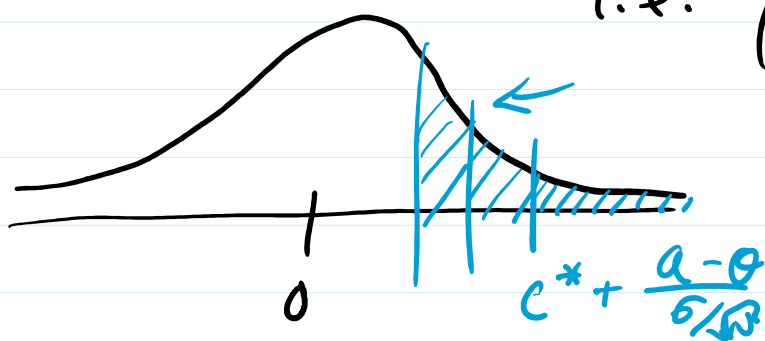
$$\bar{X} \sim N(\theta, \sigma^2/N) \quad Z = \frac{\bar{X} - \theta}{\sigma/\sqrt{N}} \sim N(0,1)$$

$$X \sim N(\theta, \sigma^2/N) \quad Z = \frac{X - \theta}{\sigma/\sqrt{N}} \sim N(0,1)$$

$$\beta(\theta) = P_{\theta} \left(Z \geq c^* + \frac{a - \theta}{\sigma/\sqrt{N}} \right)$$

as $\theta \uparrow$ then $\beta(\theta) \uparrow$

i.e. β is inc. in θ

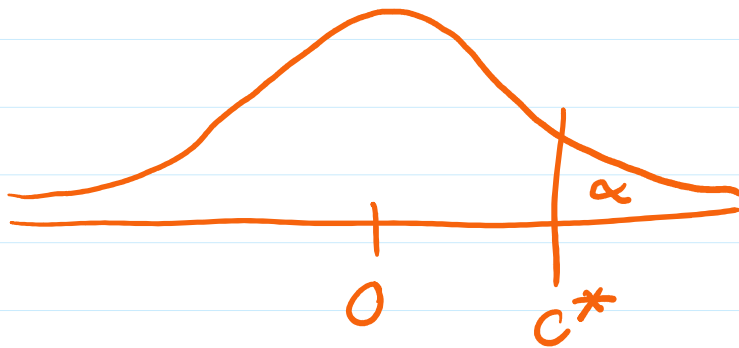


$$\begin{aligned} \max_{\theta \in \Theta_0} \beta(\theta) &= \beta(a) = P_a \left(Z \geq c^* + \frac{a - a}{\sigma/\sqrt{N}} \right) \\ &= P_a(Z \geq c^*) \end{aligned}$$

$$= P_a(Z \geq c^*)$$

So should choose c^* so that

$$P(Z \geq c^*) = \alpha$$



If F_Z is the CDF of a $N(0, 1)$ then this condition is

$$1 - F_Z(c^*) = \alpha$$

i.e. $c^* = F_Z^{-1}(1 - \alpha)$

Defn: Uniformly Most Powerful (UMP) test

If \mathcal{C} is a class of tests for the

If \mathcal{C} is a class of tests for the hypothesis

$$H_0: \theta \in \Theta_0 \text{ v. } H_a: \theta \in \Theta_a$$

then the test w/ power function β^* is called the UMP test for this class if

$$\beta^*(\theta) \geq \beta(\theta) \quad \forall \theta \in \Theta_a$$

for any other test in \mathcal{C} w/ power β .

UMP level α test

the UMP among all tests where

$$\max_{\theta \in \Theta_0} \beta(\theta) \leq \alpha. \quad [\text{level } \alpha \text{ test}]$$

$\beta(\theta)$

