

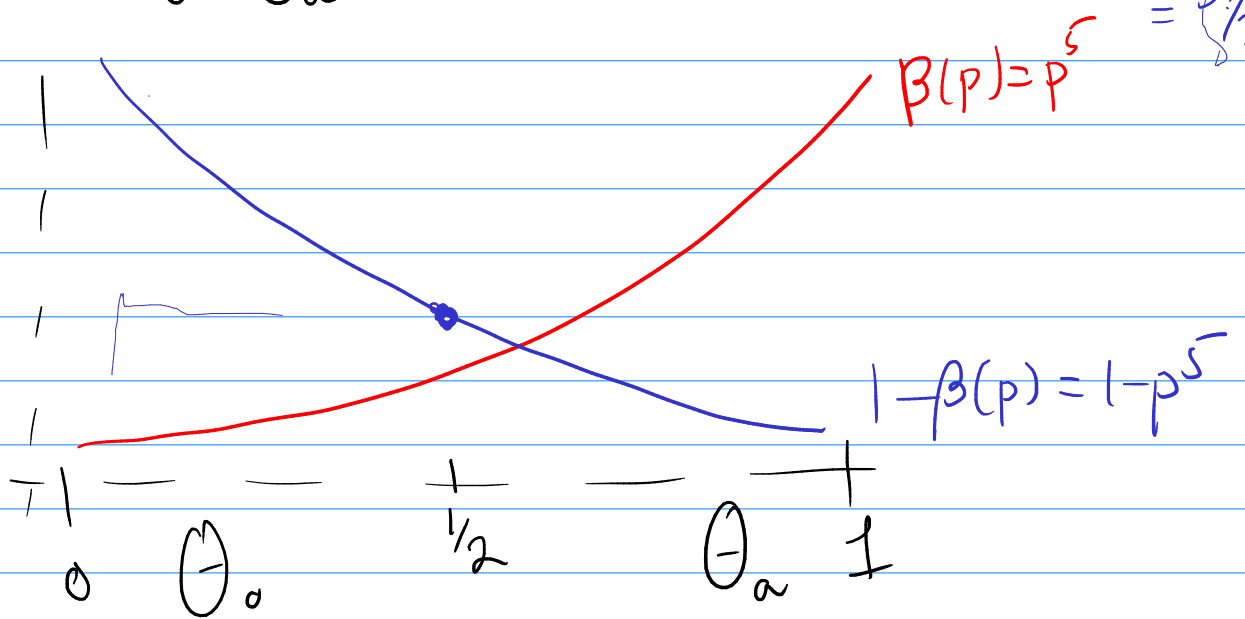
## Lecture 16

What's max type II error prob?

When  $\theta \in (-)_{\alpha}$  then  $1 - \beta(\theta) = \text{prob. type II}$

so

$$\max_{\theta \in (-)_{\alpha}} 1 - \beta(\theta) = \max_{p > 1/2} 1 - p^5 = 1 - \left(\frac{1}{2}\right)^5 = \frac{31}{32}$$



Consider another test:

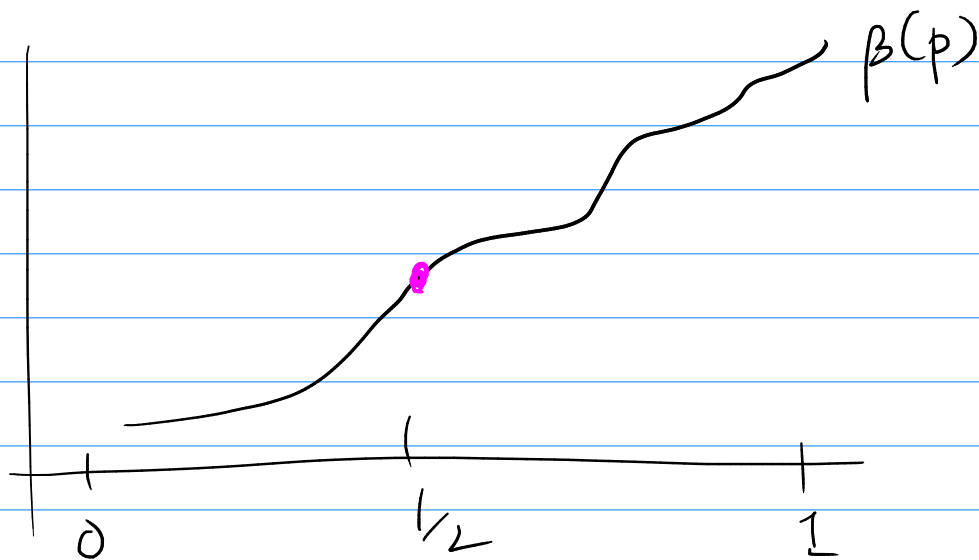
$$\begin{aligned} \hat{R} &= \{ \underline{x} \mid T \geq 3 \} \\ &= \{ \underline{x} \mid T/5 = \bar{X} \geq 1/2 \} \end{aligned}$$

What's the power function?

$$\beta(p) = P_p(\underline{x} \in \hat{R}) = P(T \geq 3)$$

$$\begin{aligned}
 &= P(T=3) + P(T=4) + P(T=5) \\
 &= \binom{5}{3} p^3 (1-p)^2 + \binom{5}{4} p^4 (1-p) + \binom{5}{5} p^5 \\
 &= p^3 (6p^2 - 15p + 10)
 \end{aligned}$$

Notice:-  $\frac{\partial \beta}{\partial p} = 30p^2(p-1)^2 > 0$   
 So  $\beta$  is increasing.



- ① max type I err prob. =  $\beta(1/2)$
- ② max type II err. prob. =  $1 - \beta(1/2)$

Defn: Size and level  $\alpha$  test

A test is size  $\alpha \in [0, 1]$  if

$$\alpha = \max_{\theta \in \Theta_0} \beta(\theta) = \text{max type I err prob.}$$

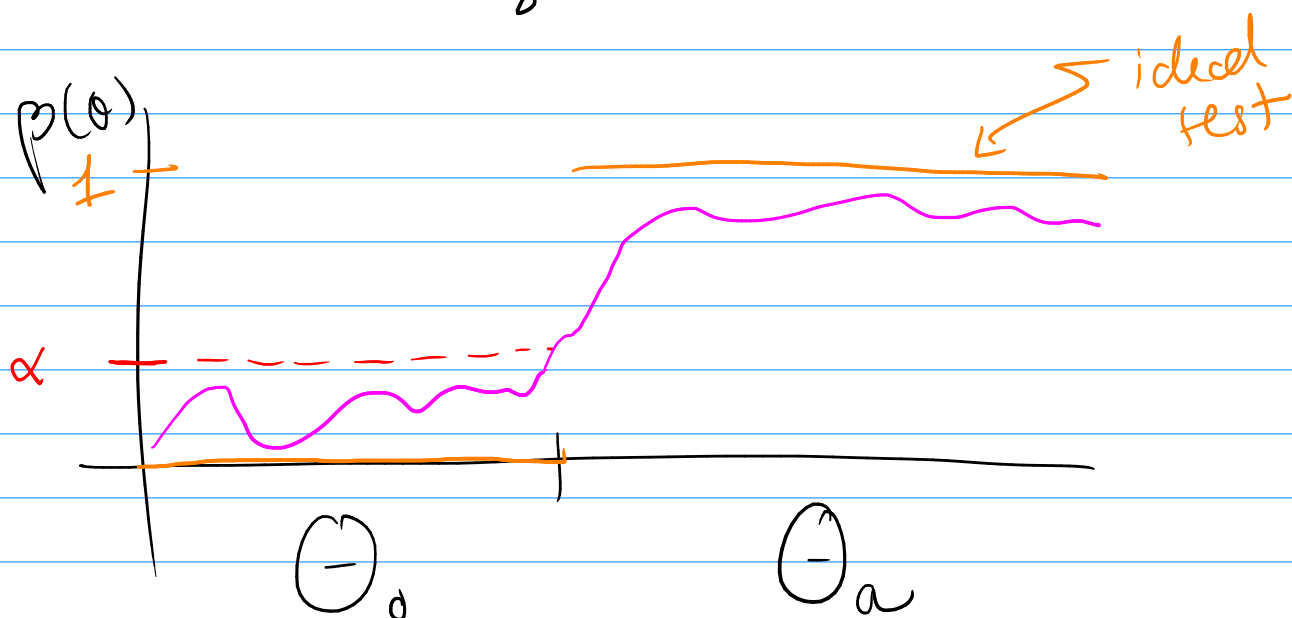
a level  $\alpha$  test is one where

$$\max_{\theta \in \Theta_0} p(\theta) \leq \alpha$$

Game: try to find test w/ maximum power

$$\max p(\theta) \text{ when } \theta \in \Theta_a$$

s.t. it is size  $\alpha$  or level  $\alpha$ .



## Likelihood Ratio Test

Want to test

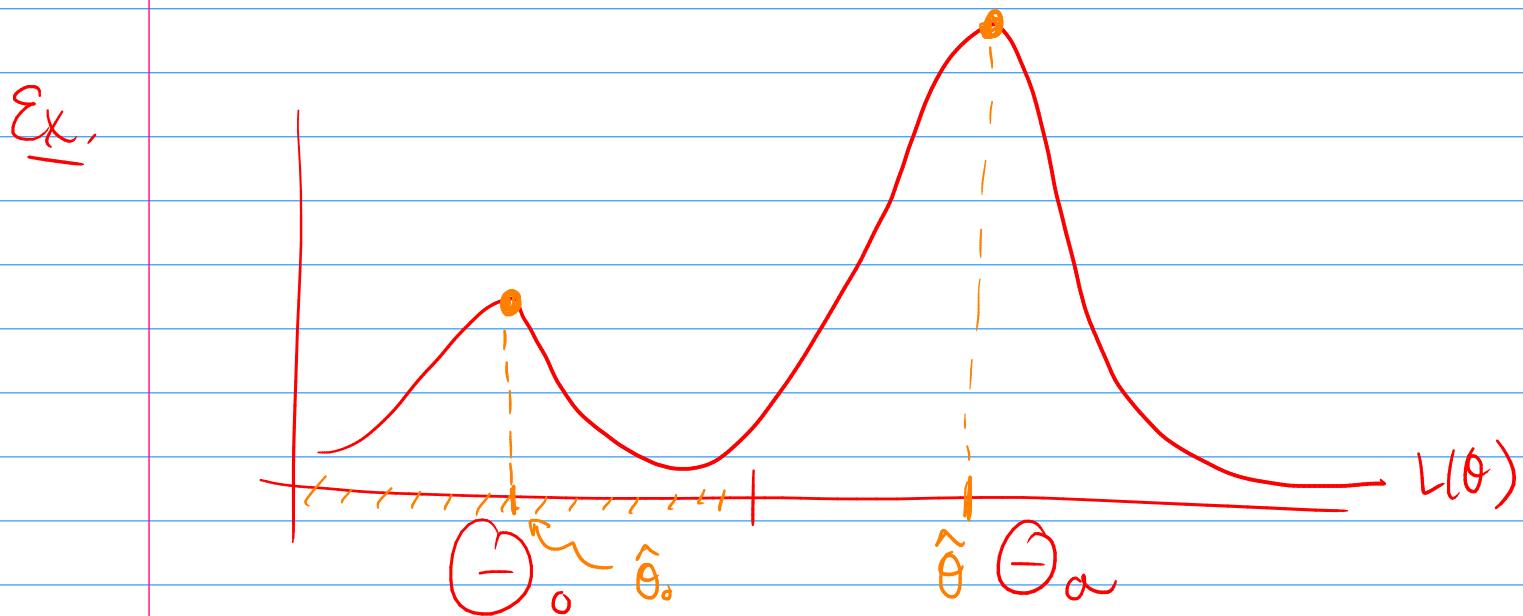
$$H_0: \theta \in \Theta_0 \quad \text{v.} \quad H_a: \theta \in \Theta_a$$

We define the likelihood ratio test statistic  
(LRT)  
as

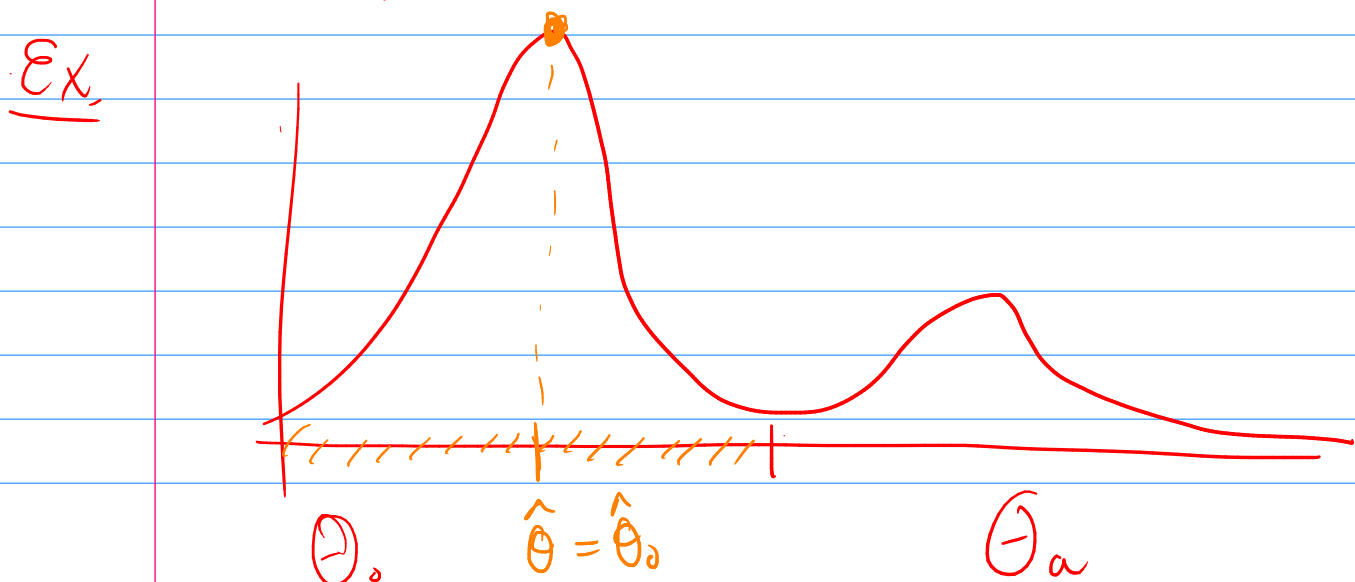
$$\lambda(\underline{x}) = \frac{\max_{\theta \in \Theta_0} L(\theta)}{\max_{\theta \in \Theta} L(\theta)} = \frac{\text{max val. of } L(\theta), \theta \text{ in null}}{\text{max val. of } L(\theta) \text{ overall}}$$

$$\hat{\theta}_0 = \text{MLE restricted to } \Theta_0 \quad = \frac{L(\hat{\theta}_0)}{L(\hat{\theta})}$$

$$\hat{\theta} = \text{MLE} \leq 1$$



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

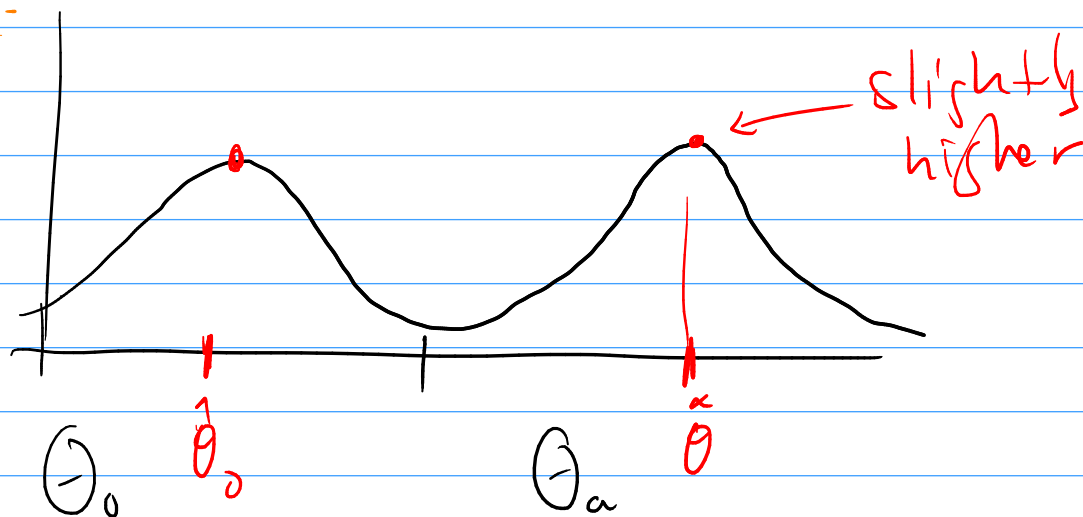


so

$$\lambda = \frac{L(\hat{\theta}_0)}{L(\hat{\theta})} = \frac{L(\hat{\theta})}{L(\hat{\theta})} = 1$$

↗ don't reject

hard case:



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

The LRT says I should reject if

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

↗  $0 \leq c \leq 1$

$c$  is some threshold I choose to trade off Type I and Type II err probs.

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

$c$  large, reject easily, more type I  
less type II

rejection region for LRT

$$R = \{ \gamma_{\sim} : \chi(\gamma_{\sim}) \leq c \}$$

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

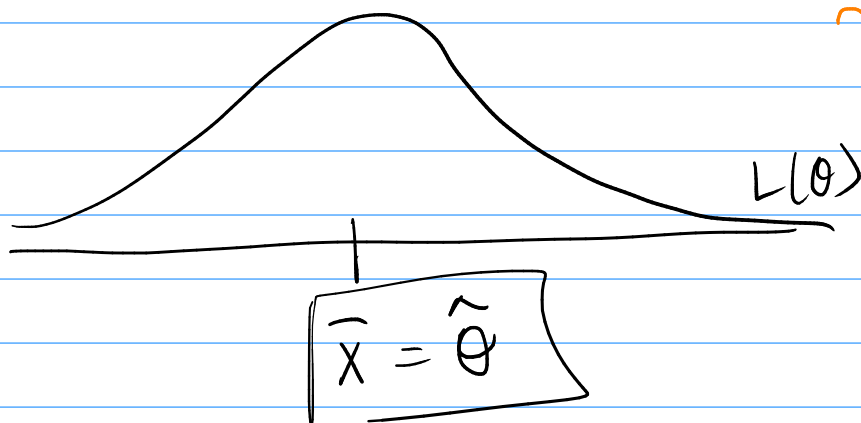
$$H_0: \theta \leq a \quad \text{v.} \quad H_a: \theta > a$$

Let's derive the LRT

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

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

kinda quadratic in  $\theta$   
 $\sim e^{-\theta^2}$



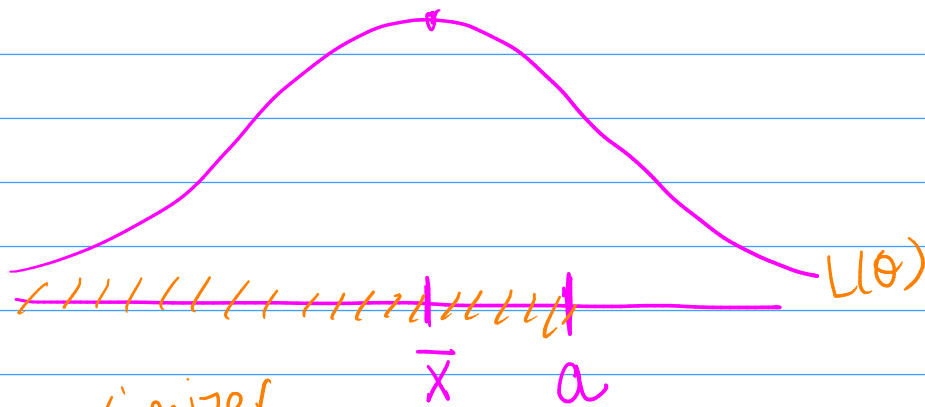
Need to find  $\hat{\theta}_0$ ,  $\Theta_0 = [-\infty, a]$

Two cases

$$\hat{\theta}_0 = \arg\max_{\theta \leq a} L(\theta)$$

$\bar{X} \leq a$

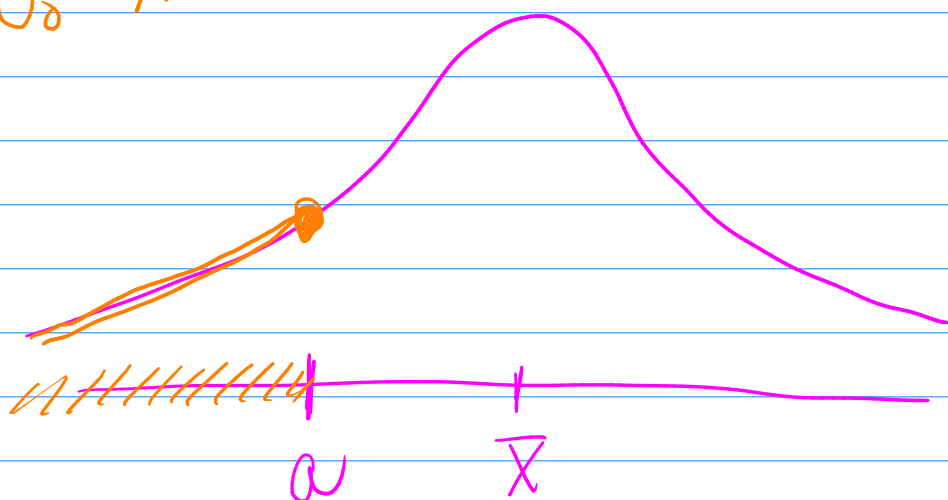
Since  $\bar{X} \leq a$



and it is the over-all maximizer,  
then  $\hat{\theta}_0 = \bar{X}$

$\bar{X} > a$

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



Overall,

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

So

$$\lambda(\underline{x}) = \frac{L(\hat{\theta}_0)}{L(\hat{\theta})} = \begin{cases} \frac{L(\bar{X})}{L(\bar{X})} = 1 & \bar{X} \leq a \\ \frac{L(a)}{L(\bar{X})} & \bar{X} > a \end{cases}$$

never reject

$H_0: \theta \leq a$

LRT says, reject if  $\lambda \leq c$

Case  $\bar{X} > a$

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

$$= \frac{\exp\left(-\frac{1}{2\sigma^2} \sum_n (X_n^2 - 2aX_n + a^2)\right)}{\exp\left(-\frac{1}{2\sigma^2} \sum_n (X_n^2 - 2\bar{X}X_n + \bar{X}^2)\right)}$$

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

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

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

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

LRT says reject when  $\lambda \leq c$

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

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

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



↓ assuming  $\bar{x} > a \Leftrightarrow \bar{x} - a > 0$

$$\Leftrightarrow \frac{\sqrt{N}}{\sigma} (\bar{x} - a) \geq \sqrt{-2 \log c}$$

$$\Leftrightarrow \frac{\bar{x} - a}{\sigma/\sqrt{N}} \geq \sqrt{-2 \log c} \quad c^*$$

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

$$H_0: \theta \leq a$$

$$H_a: \theta > a$$

reject if  $\bar{x} \leq a + c^* \text{ s.d.s}(\bar{x})$

