

## Lecture 12:

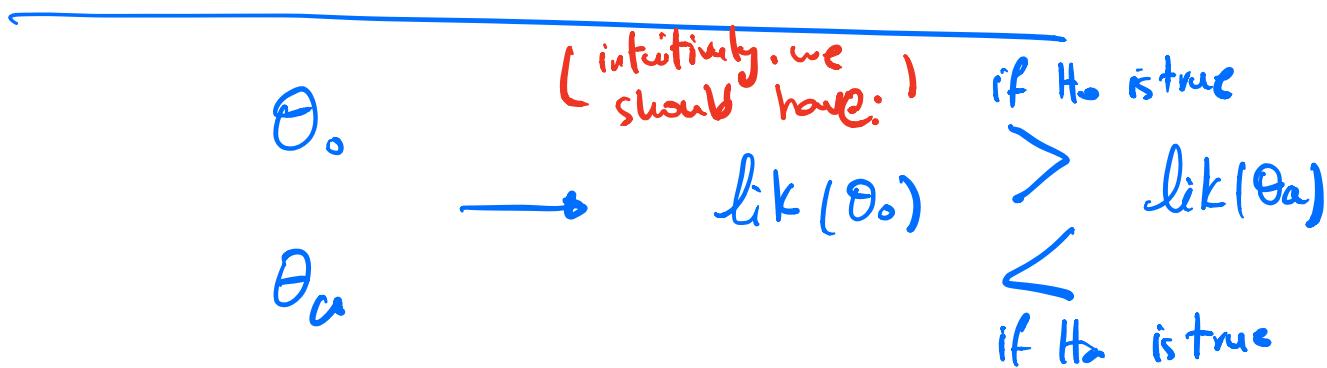
$$x_1, \dots, x_n \sim f(x|\theta)$$

Example of a more general hypothesis testing problem:

$$\left\{ \begin{array}{l} H_0 : \theta = \theta_0 \\ H_a : \theta = \theta_a \end{array} \right.$$

For such problems, the likelihood function is the main tool to design appropriate tests.

$$\begin{aligned} \text{lik}(\theta) &= f(x_1, \dots, x_n | \theta) \\ &= f(x_1 | \theta) \times \dots \times f(x_n | \theta) \end{aligned} \quad \leftarrow$$

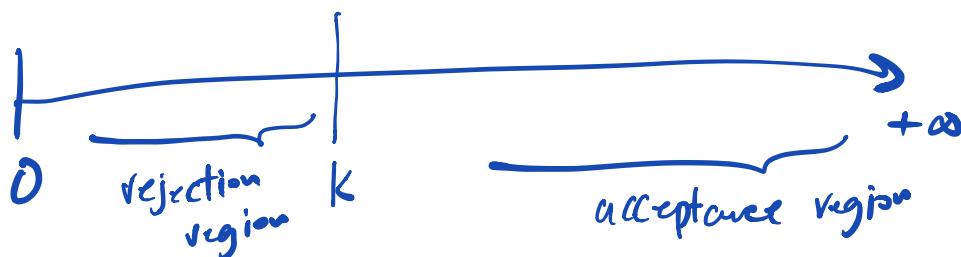


Test:

$$T(x_1, \dots, x_n) = \frac{\text{lik}(\theta_0)}{\text{lik}(\theta_a)}$$

$$\text{lik}(\theta) = f(x_1, \dots, x_n | \theta)$$

outcome of  $T(\cdot)$



$$\frac{\text{lik}(\theta_0)}{\text{lik}(\theta_a)} \begin{cases} > K \\ < K \end{cases}$$

$K$  is determined based on the significance level  $\alpha$ .

Example: Assume  $X_1, \dots, X_n \sim f(x|\lambda)$

$$f(x|\lambda) = \lambda e^{-\lambda x} \mathbf{1}_{\{x \geq 0\}}$$

$\underbrace{\quad}_{\text{exp}(\lambda) \text{ dist.}}$

$$\left\{ \begin{array}{l} H_0 = \lambda = \lambda_0 \\ H_a = \lambda = \lambda_a \end{array} \right. \quad \begin{array}{l} \text{(Assume} \\ \text{for simplicity)} \\ \lambda_a > \lambda_0 \end{array}$$

significance level  $\underline{\alpha}$  is also given.

e.g. 0.05

Test:  $\frac{\text{lik}(\lambda_0)}{\text{lik}(\lambda_a)} \stackrel{D}{=} T(X_1, \dots, X_n)$

$$T(X_1, \dots, X_n) \stackrel[H_0]{\geq}^{\stackrel[H_a]{<}} k$$

$$lik(\lambda) = f(x_1, \dots, x_n | \lambda)$$

$$= \prod_{i=1}^n f(x_i | \lambda)$$

$$= \prod_{i=1}^n \lambda e^{-\lambda x_i}$$

$$= \lambda^n e^{-\lambda \sum_{i=1}^n x_i}$$

$$T(x_1, \dots, x_n) = \frac{lik(\lambda_0)}{lik(\lambda_a)} = \frac{\lambda_0^n e^{-\lambda_0 \sum_{i=1}^n x_i}}{\lambda_a^n e^{-\lambda_a \sum_{i=1}^n x_i}}$$

$$= \left( \frac{\lambda_0}{\lambda_a} \right)^n e^{-(\lambda_0 - \lambda_a) \sum_{i=1}^n x_i}$$

$$\rightarrow \left( \frac{\lambda_0}{\lambda_a} \right)^n e^{-(\lambda_0 - \lambda_a) \sum_{i=1}^n x_i}$$

$\begin{cases} H_0 & \geq K \\ H_a & \end{cases}$

$$-(\lambda_0 - \lambda_a) \sum_{i=1}^n x_i + n \log \frac{\lambda_0}{\lambda_a} \stackrel{H_0}{>} \stackrel{H_a}{<} \log k$$

$$\Leftrightarrow \overbrace{(\lambda_a - \lambda_0) \sum x_i}^{> 0} \stackrel{H_0}{>} \stackrel{H_a}{<} \log k - n \log \frac{\lambda_0}{\lambda_a}$$

$$\Leftrightarrow \sum_{i=1}^n x_i \stackrel{H_0}{>} \stackrel{H_a}{<} \frac{\log k - n \log \frac{\lambda_0}{\lambda_a}}{(\lambda_a - \lambda_0)}$$

$$\Leftrightarrow \sum_{i=1}^n x_i \stackrel{H_0}{>} \stackrel{H_a}{<} K'$$

we have to find  
K' in terms of  $\alpha$ .

To find  $K'$ :

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

$$\Leftrightarrow \Pr \left\{ H_0 \text{ is rejected} \mid H_0 \text{ is true} \right\} = \alpha$$

$$\Leftrightarrow \Pr \left\{ \sum_{i=1}^n x_i < k' \mid x_i \sim \exp(\lambda_0) \right\} = \alpha$$

If  $x_i \sim \exp(\lambda_0) \Rightarrow \sum_{i=1}^n x_i \stackrel{\text{CLT}}{\sim} \frac{n}{\lambda_0} + \frac{\sqrt{n}}{\lambda_0} N(0, 1)$

$$\Pr \left\{ \frac{n}{\lambda_0} + \frac{\sqrt{n}}{\lambda_0} N(0, 1) < k' \right\} = \alpha$$

CLT:

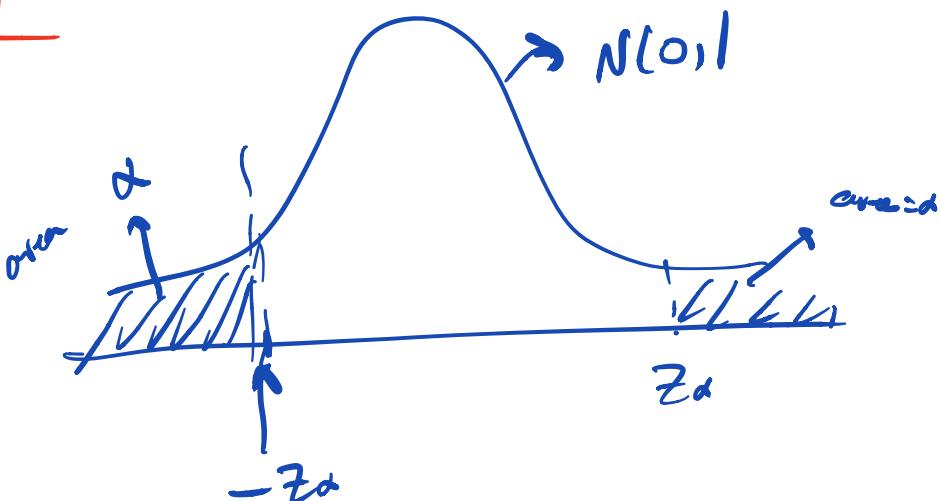
if  $x_i \sim \text{dist}(\mu, \sigma^2)$

$$\sum_{i=1}^n x_i = n\mu + \sigma \cdot \sqrt{n} N(0, 1)$$

$x_i \sim \exp(\lambda_0)$

$$\mu = \frac{1}{\lambda_0}, \sigma^2 = \frac{1}{\lambda_0^2}$$

$$\Leftrightarrow \Pr \left\{ N(0, 1) < \frac{k' - \frac{n}{\lambda_0}}{\frac{\sqrt{n}}{\lambda_0}} \right\} = \alpha$$



$$\Leftrightarrow \frac{k' - \frac{n}{\lambda_0}}{\frac{\sqrt{n}}{\lambda_0}} = -z_\alpha$$

$$\Rightarrow k' = -\frac{z_\alpha \cdot \sqrt{n}}{\lambda_0} + \frac{n}{\lambda_0}$$

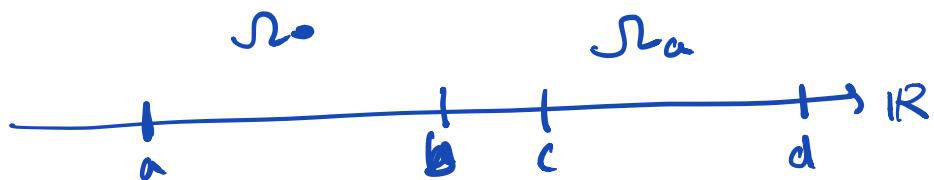
$$\Rightarrow \sum_{i=1}^n x_i - \frac{n}{\lambda_0} \geq H_a - \frac{z_\alpha \sqrt{n}}{\lambda_0}$$

$$x_1, \dots, x_n \sim f(x|\theta)$$

$$\left\{ \begin{array}{l} H_0 : \theta = \theta_0 \\ H_a : \theta = \theta_a \end{array} \right. \xrightarrow{\text{General version}} \left\{ \begin{array}{l} H_0 : \theta \in \mathcal{R}_0 \\ H_a : \theta \in \mathcal{R}_a \end{array} \right.$$

e.g.  $\mathcal{R}_0 = \{\theta_0\}$  and  $\mathcal{R}_a = \{\theta_a\}$  ← solved using the likelihood test

e.g.



e.g.



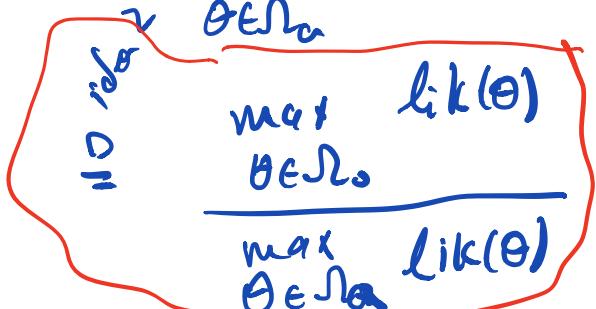
$$\theta \in \mathbb{R}^2$$

Generalized Likelihood Test:

$$x_1, \dots, x_n \sim f(x|\theta)$$

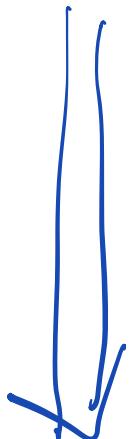
$$\left. \begin{array}{l} H_0 : \theta \in \mathcal{S}_0 \\ H_{\alpha} : \theta \in \mathcal{S}_{\alpha} \end{array} \right\}$$

$$\text{Test: } T(x_1, \dots, x_n) \stackrel{\text{ideally}}{\approx} \frac{\int_{\theta \in \mathcal{S}_0} \text{lik}(\theta) d\theta}{\int_{\theta \in \mathcal{S}_{\alpha}} \text{lik}(\theta) d\theta} \stackrel{\mathcal{S}_0}{=} \frac{\max_{\theta \in \mathcal{S}_0} \text{lik}(\theta)}{\max_{\theta \in \mathcal{S}_{\alpha}} \text{lik}(\theta)}$$



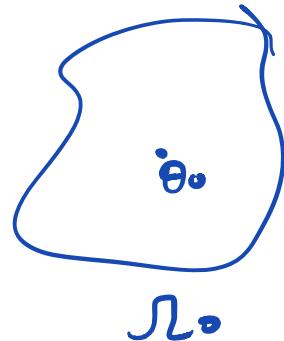
intuition:

Let's assume that  $H_0$  is true.

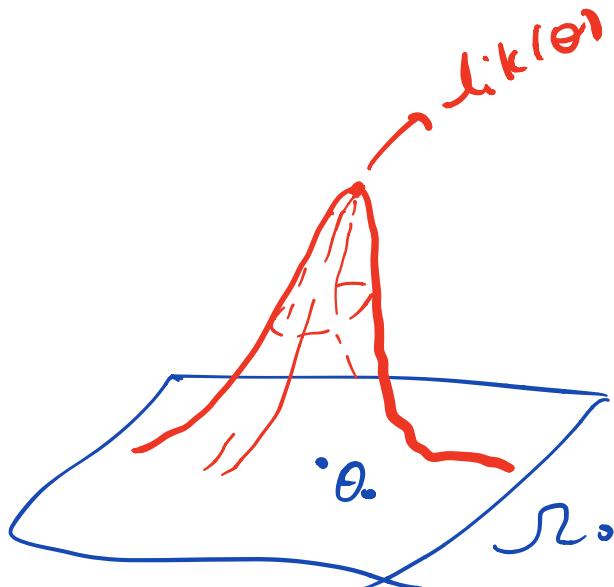


$$x_1, \dots, x_n \sim f(x|\theta_0)$$

$$\theta_0 \in \Omega_0$$



$$\Rightarrow \text{lik}(\theta_0) = \max_{\theta \in \Omega_0} \text{lik}(\theta)$$



if  $\theta_0 \in \Omega_0$  is  
the true parameter

the

$$\max_{\theta \in \Omega_0} \text{lik}(\theta) = \text{lik}(\theta_0)$$

- Generalized Likelihood Test:

$$T(x_1, \dots, x_n) \stackrel{?}{=} \frac{\max_{\theta \in \Omega_0} \text{lik}(\theta)}{\max_{\theta \in \Omega} \text{lik}(\theta)}$$

- Acceptance/Rejection regions:

$$T(x_1, \dots, x_n) \begin{matrix} \xrightarrow{H_0} \\ \geq \\ \xrightarrow{H_a} \end{matrix} K$$

↓  
is determined  
in terms of  $\alpha$ .

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Hypothesis testing problem → design  $T$   
→ acc/rej regions

$\alpha$  → Type I error  
 $\beta$  → Type II error

- Remember that our goal is to  
design a test + acc/rej regions such that  
} (1)  $\alpha$  is less than a given significance level  
} (2)  $\beta$  is as small as possible

- Definition Power of a test  $\triangleq 1 - \beta$ .
  - Among all the tests + acc/reject regions with  $\alpha$  less than a given level, our goal should to find the most powerful one (i.e. smallest  $\beta$ ).
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Neyman - Pearson Lemma:

Consider the following hypothesis testing problem:

$$X_1, \dots, X_n \sim f(x|\theta)$$

$$\left. \begin{array}{l} H_0: \theta = \theta_0 \\ H_a: \theta = \theta_a \end{array} \right\}$$

$$\left. \begin{array}{l} H_0: \theta = \theta_0 \\ H_a: \theta = \theta_a \end{array} \right\}$$

Significance level  $= \alpha$

Among all the tests that we can design for the hypothesis problem above, the likelihood test is the most powerful one.

$$5l \rightarrow 4l$$

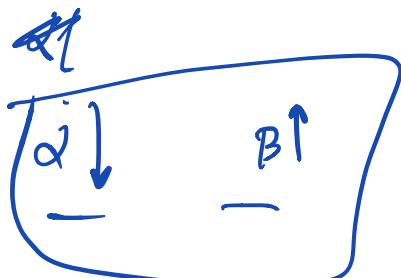
$$H_0 : \beta = 0.$$

$$H_a : \beta > 50\%.$$

$$\min \beta(s)$$

over all the solutions

$$\underline{\alpha}(s) < \alpha_0 = 0.01$$



$\theta \in \mathcal{N}_0$

$\theta \in \mathcal{N}_a$

$$\rightarrow \theta = \theta_0$$

$$\rightarrow \theta \neq \theta_0$$

