

## Likelihood Ratio Tests

For simple vs. simple, N-P says to compare likelihoods.  
So perhaps this is a good idea in general?

LRT: For testing  $H_0: \theta \in \omega_1$  vs.  $H_1: \theta \in \omega_2$   
where  $\omega_1 \cup \omega_2 = \Omega$

$$\text{let } \lambda(x) = \frac{\sup_{\theta \in \omega_1} L(\theta|x)}{\sup_{\theta \in \omega_2} L(\theta|x)}$$

The LRT of size  $\alpha$  is the test that rejects when  $\lambda \leq c$ , where  $\sup_{\theta \in \omega_1} P(\lambda \leq c | \theta) = \alpha$ .

Ex:  $X_1, \dots, X_n \stackrel{iid}{\sim} U[0, \theta]$ .

$H_0: \theta \leq 1$  vs.  $H_1: \theta > 1$ .

Assume  $\max\{X_i\} \leq 1$  (otherwise, clearly reject).

$$\lambda = \frac{\sup_{\theta \leq 1} \frac{1}{\theta^n} I_{\{\max X_i \leq \theta\}}}{\sup_{\theta > 1} \frac{1}{\theta^n} I_{\{\max X_i \leq \theta\}}} = \frac{\frac{1}{(\max X_i)^n}}{\frac{1}{(1)^n}}$$

LRT is to reject  $H_0$  when  $\frac{1}{(\max X_i)^n} \leq c$

i.e. when  $\max\{X_i\} \geq c'$ .

$$\begin{aligned} \sup_{\theta \leq 1} P(\max\{X_i\} \geq c' | \theta) &= P(\max\{X_i\} \geq c' | \theta = 1) \\ &= 1 - (c')^n = \alpha \quad c' = (1 - \alpha)^{1/n} \end{aligned}$$

Ex:  $X_1, \dots, X_n \stackrel{iid}{\sim} \text{Geom}(p)$   $f_X(x|p) = p(1-p)^x$   $\hat{p} = \frac{1}{\bar{x}+1}$

$H_0: p = \frac{1}{2}$

$H_1: p \neq \frac{1}{2}$

$$\lambda = \frac{\sup_{p=\frac{1}{2}} p^n (1-p)^{\sum X_i}}{\sup_{p \neq \frac{1}{2}} p^n (1-p)^{\sum X_i}} = \frac{(\frac{1}{2})^n (\frac{1}{2})^{\sum X_i}}{(\frac{1}{\bar{x}+1})^n (1 - \frac{1}{\bar{x}+1})^{\sum X_i}}$$

reject when  $\lambda = \frac{(\frac{1}{2})^{n(\bar{x}+1)}}{(\frac{1}{\bar{x}+1})^{n(\bar{x}+1)} (\bar{x})^{n\bar{x}}} < c$

$$= \frac{(\frac{\bar{x}+1}{2})^{n(\bar{x}+1)}}{(\bar{x})^{n\bar{x}}} = \left(\frac{\bar{x}+1}{2\bar{x}}\right)^{n\bar{x}} \left(\frac{\bar{x}+1}{2}\right)^n < c$$

$$\Rightarrow \left(\frac{1}{2} + \frac{1}{2\bar{x}}\right)^{\bar{x}} \left(\frac{\bar{x}+1}{2}\right) < c'$$

$$y^y = e^{y \log y}$$

$$\frac{dy^y}{dy} = e^{y \log y} \left( \frac{y}{y} + \log y \right) = y^y (1 + \log y)$$

$$f(y) = \left( \frac{1}{2} + \frac{1}{2y} \right)^y \left( \frac{y+1}{2} \right) = e^{y \log \left( \frac{1}{2} + \frac{1}{2y} \right)} \left( \frac{y+1}{2} \right)$$

$$f'(y) = \left( \frac{y+1}{2} \right) e^{y \log \left( \frac{1}{2} + \frac{1}{2y} \right)} \left( \log \left( \frac{1}{2} + \frac{1}{2y} \right) + \frac{y}{\frac{1}{2} + \frac{1}{2y}} \left( -\frac{1}{2y^2} \right) \right) + \frac{1}{2} e^{y \log \left( \frac{1}{2} + \frac{1}{2y} \right)}$$

$$= \left( \frac{y+1}{2} \right) \left( \frac{1}{2} + \frac{1}{2y} \right)^y \left( \log \left( \frac{1}{2} + \frac{1}{2y} \right) - \frac{1}{y+1} \right) + \frac{1}{2} \left( \frac{1}{2} + \frac{1}{2y} \right)^y$$

$$= \left( \frac{y+1}{2} \right) \left( \frac{1}{2} + \frac{1}{2y} \right)^y \left( \log \left( \frac{1}{2} + \frac{1}{2y} \right) - \frac{1}{y+1} + \frac{1}{y+1} \right)$$

$$= \left( \frac{y+1}{2} \right) \left( \frac{1}{2} + \frac{1}{2y} \right)^y \log \left( \frac{1}{2} + \frac{1}{2y} \right)$$

max ~~min~~ when  $f'(y) = 0$ , i.e. when  $\log \left( \frac{1}{2} + \frac{1}{2y} \right) = 0$   
 or when  $y = 1$   
 $y = \bar{x} = 1 \Rightarrow \hat{p} = \frac{1}{2}$

$$f(y) < 0 \text{ for } y > 1$$

$$f(y) > 0 \text{ for } y < 1$$

So reject when  $\hat{p}$  is away from  $\frac{1}{2}$   
 reject when  $\bar{x}$  is outside some  
 asymmetric interval around 1.

Thm: Under certain regularity conditions  
 (satisfied by exponential families)

$$-2 \log \lambda(\bar{x}) \xrightarrow{D} \chi_p^2$$

$p = \#$  free parameters in  $H_1$   
 $- \#$  free parameters in  $H_0$

