

Our likelihood ratio test has rejection region

$$C = \left\{ x: \frac{h_1(x)}{h_0(x)} > k \right\}$$

Define $\phi_c(x) = \begin{cases} 1 & \text{if } x \in C \\ 0 & \text{if } x \notin C \end{cases}$

Let D be the critical region for any other test of the same hypotheses of size α .

Define $\phi_D(x) = \begin{cases} 1 & \text{if } x \in D \\ 0 & \text{if } x \notin D \end{cases}$

$$\text{If } \phi_c(x) = 1 \Rightarrow \frac{h_1(x)}{h_0(x)} > k$$

$$\Rightarrow h_1(x) - k h_0(x) > 0$$

$$\text{If } \phi_c(x) = 0 \Rightarrow h_1(x) - k h_0(x) \leq 0$$

Consider

$$(\phi_c(x) - \phi_D(x))(h_1(x) - k h_0(x)) \geq 0$$

$\forall x.$

(always the product of two terms with the same sign)

$h_1(\underline{x}) = f_1(\underline{x}) \leftarrow$ joint density of \underline{X} under H_1

$h_0(\underline{x}) = f_0(\underline{x}) \leftarrow$ joint density of \underline{X} under H_0 .

$$\int_{\mathcal{X}} (\phi_c(\underline{x}) - \phi_D(\underline{x})) (h_1(\underline{x}) - k h_0(\underline{x})) d\underline{x} \geq 0$$

$$\int_{\mathcal{X}} \phi_c(\underline{x}) h_1(\underline{x}) d\underline{x} - \int_{\mathcal{X}} \phi_D(\underline{x}) h_1(\underline{x}) d\underline{x}$$

$$- k \left(\int_{\mathcal{X}} \phi_c(\underline{x}) h_0(\underline{x}) d\underline{x} - \int_{\mathcal{X}} \phi_D(\underline{x}) h_0(\underline{x}) d\underline{x} \right) \geq 0$$

$$\int_{\mathcal{X}} \phi_c(\underline{x}) h_1(\underline{x}) d\underline{x} - \int_{\mathcal{X}} \phi_D(\underline{x}) h_1(\underline{x}) d\underline{x}$$

$$- k [\cancel{\alpha} - \overset{0}{\alpha}] \geq 0$$

$$\underbrace{\int_{\mathcal{X}} \phi_c(\underline{x}) h_1(\underline{x}) d\underline{x}} - \int_{\mathcal{X}} \phi_D(\underline{x}) h_1(\underline{x}) d\underline{x} \geq 0$$

$$IP(\underline{X} \in C | H_1) - IP(\underline{X} \in D | H_1) \geq 0$$

Power of test with
crit. region C

Power of test
with crit. region D

\therefore Most powerful test of size α has critical region C.

$$H_0: \mu = \mu_0 \quad \text{vs.} \quad H_1: \mu = \mu_1, \quad (\mu_1 > \mu_0)$$

$$\text{Let } X_1, \dots, X_n \quad X_i \sim \mathcal{N}(\mu, \sigma^2)$$

Under H_0

$$X_i \sim \mathcal{N}(\mu_0, \sigma^2)$$

Likelihood f'n

$$\begin{aligned} L(\mu_0 | \underline{X}) &= \prod_{i=1}^n \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left\{-\frac{1}{2\sigma^2} (X_i - \mu_0)^2\right\} \\ &= (2\pi\sigma^2)^{-n/2} \exp\left\{-\frac{1}{2\sigma^2} \sum_{i=1}^n (X_i - \mu_0)^2\right\} \end{aligned}$$

Under H_1 :

$$X_i \sim \mathcal{N}(\mu_1, \sigma^2)$$

$$L(\mu_1 | \underline{X}) = (2\pi\sigma^2)^{-n/2} \exp\left\{-\frac{1}{2\sigma^2} \sum_{i=1}^n (X_i - \mu_1)^2\right\}$$

The most powerful test ^{of size α} of H_0 vs. H_1 has rejection region given by:

$$\frac{L_1(\mu_1 | \underline{X})}{L_0(\mu_0 | \underline{X})} > k$$

$$\Rightarrow \frac{(2\pi\sigma^2)^{-n/2} \exp\left\{-\frac{1}{2\sigma^2} \sum_{i=1}^n (X_i - \mu_1)^2\right\}}{(2\pi\sigma^2)^{-n/2} \exp\left\{-\frac{1}{2\sigma^2} \sum_{i=1}^n (X_i - \mu_0)^2\right\}} > k$$

$$\Rightarrow \exp \left\{ -\frac{1}{2\sigma^2} \sum_{i=1}^n (X_i - \mu_1)^2 + \frac{1}{2\sigma^2} \sum_{i=1}^n (X_i - \mu_0)^2 \right\} > k$$

$$\exp \left\{ -\frac{1}{2\sigma^2} \left[\sum_{i=1}^n (\cancel{X_i^2} - 2\mu_1 X_i + \mu_1^2) \right. \right. \\ \left. \left. \rightarrow -\sum_i (\cancel{X_i^2} - 2\mu_0 X_i + \mu_0^2) \right] \right\} > k$$

$$\exp \left\{ -\frac{1}{2\sigma^2} \left[-2[\mu_1 + \mu_0] \sum_{i=1}^n X_i + n(\mu_1^2 - \mu_0^2) \right] \right\} > k \quad (t)$$

This is a monotonic increasing function

$$\sum_{i=1}^n X_i$$

$$\therefore (t) \Rightarrow \sum_{i=1}^n X_i > c$$

$$\Rightarrow \bar{X} > c'$$

Under H_0 :

$$X_i \sim \mathcal{N}(\mu_0, \sigma^2)$$

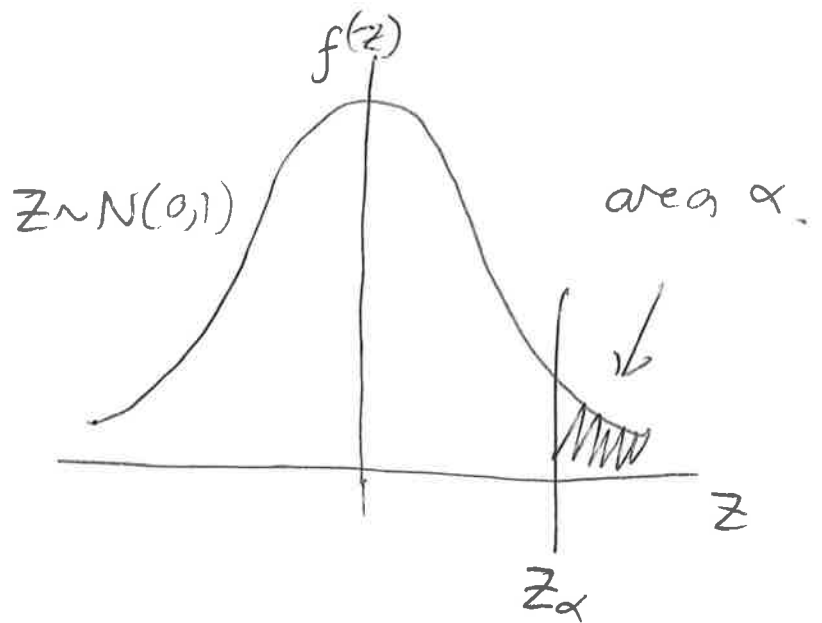
$$\Rightarrow \bar{X} \sim \mathcal{N}\left(\mu_0, \frac{\sigma^2}{n}\right)$$

$$\therefore P(\bar{X} > c' | H_0) = \alpha$$

$$\Rightarrow P\left(\frac{\bar{X} - \mu_0}{\sigma/\sqrt{n}} > \frac{c' - \mu_0}{\sigma/\sqrt{n}}\right) = \alpha$$

$$\frac{c' - \mu_0}{\sigma/\sqrt{n}} = z_\alpha$$

$$c' = \mu_0 + z_\alpha \frac{\sigma}{\sqrt{n}}$$



\therefore Reject H_0 if

$$\bar{x} > \mu_0 + z_\alpha \frac{\sigma}{\sqrt{n}}$$

$$C = \left\{ \underline{x} : \bar{x} > \mu_0 + z_\alpha \frac{\sigma}{\sqrt{n}} \right\}$$

(b)

Test has size α ✓

Is the test most powerful for all $\mu_1 > \mu_0$?

Test is uniformly most powerful as C does not depend on μ_1 .

$$H_0: \theta = \theta_0 \quad \text{vs.} \quad H_1: \theta \neq \theta_0$$

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \text{Bern}(\theta)$$

$$L(\theta | \underline{X}) = \theta^{\sum X_i} (1 - \theta)^{n - \sum X_i}$$

Log-likelihood:

$$\ell(\theta | \underline{X}) = (\sum X_i) \log \theta + (n - \sum X_i) \log(1 - \theta)$$

$$\text{MLE: } \hat{\theta} = \frac{\sum_{i=1}^n X_i}{n}$$

Test statistic. Let $R = \sum_{i=1}^n X_i$ (no. of successes)

$$\begin{aligned} X^2 &= 2 \left[\ell(\hat{\theta} | \underline{X}) - \ell(\theta_0 | \underline{X}) \right] \\ &= 2 \left\{ R \log \left(\frac{R}{n} \right) + (n - R) \log \left(1 - \frac{R}{n} \right) \right. \\ &\quad \left. - R \log \theta_0 - (n - R) \log (1 - \theta_0) \right\} \\ &= 2 \left\{ R \left[\log \left(\frac{R}{n} \right) - \log \theta_0 \right] \right. \\ &\quad \left. + (n - R) \left[\log \left(1 - \frac{R}{n} \right) - \log (1 - \theta_0) \right] \right\} \end{aligned}$$

Under H_0 : $X^2 \sim \chi_1^2$

\therefore For a test of size α , reject H_0 if $X^2 > \chi_1^2(1 - \alpha)$

(b) $r = 18$ voters who will vote for the mayor

$$n = 80$$

$$H_0: \theta = 0.3 \text{ vs. } H_1: \theta \neq 0.3$$

$$\therefore \theta_0 = 0.3$$

$$\text{Test statistic: } 2.2645$$

$$\chi^2_{1, (0.05)} = 3.84$$

$$\therefore \chi^2 = 2.2645 < 3.84$$

\Rightarrow Don't reject H_0 .

$$H_0: \theta = \theta_0 \text{ vs. } H_1: \theta \neq \theta_0$$

$$\ell(\theta | \underline{X}) = R \log \theta + (n - R) \log (1 - \theta)$$

$$\frac{\partial \ell}{\partial \theta} = \frac{R}{\theta} - \frac{(n - R)}{(1 - \theta)}$$

$$\frac{\partial^2 \ell}{\partial \theta^2} = -\frac{R}{\theta^2} - \frac{(n - R)}{(1 - \theta)^2}$$

$$\mathcal{I}(\theta) = \mathbb{E} \left[-\frac{\partial^2 \ell}{\partial \theta^2} \right] \quad \mathbb{E}[R] = n\theta.$$

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

For a Wald test ^{of size α} , reject H_0 if

$$\mathcal{I}(\theta_0) (\hat{\theta} - \theta_0)^2 > \chi^2_1 (1 - \alpha)$$

$$W = \frac{n}{\theta_0(1 - \theta_0)} \left(\frac{R}{n} - \theta_0 \right)^2 > \chi^2_1 (1 - \alpha).$$

Election example

$$R = 18$$

$$n = 80$$

$$\theta_0 = 0.3$$

$$W = 2.14857 < 3.84$$

Retain H_0 .

$$H_0: \theta = \theta_0$$

vs.

$$H_1: \theta \neq \theta_0$$

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \text{Bern}(\theta)$$

$$\begin{aligned} U(\theta; X) &= \frac{\partial \ell}{\partial \theta} \\ &= \frac{R}{\theta} - \frac{(n-R)}{1-\theta} \end{aligned}$$

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

\therefore Score test statistic:

$$\begin{aligned} X^2 &= \frac{[U(\theta_0; X)]^2}{I(\theta_0)} \\ &= \frac{\theta_0(1-\theta_0)}{n} \left[\frac{R}{\theta_0} - \frac{(n-R)}{1-\theta_0} \right]^2 \sim \chi_1^2 \text{ under } H_0. \end{aligned}$$

$$\begin{aligned} (b) \quad r &= 18 \\ n &= 80 \\ \theta_0 &= 0.3 \end{aligned}$$

$$X^2 = 2.14857 < 3.84 \Rightarrow \text{Retain } H_0.$$