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STAT 245

HOMEWORK 3

Confidence intervals

Let X_1, \dots, X_n be independent $N(\mu, \sigma^2)$ random variables.

- (a) Determine random variables $L_{\sigma^2}(\alpha)$ and $U_{\sigma^2}(\alpha)$ such that the interval $[L_{\sigma^2}(\alpha), U_{\sigma^2}(\alpha)]$ is a $(1 - \alpha)$ confidence interval for σ^2 . In doing this ensure that

$$\Pr(L_{\sigma^2} > \sigma^2) = \Pr(U_{\sigma^2} < \sigma^2)$$

Consider two estimators for σ^2

$$s^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2$$

and

$$\hat{\sigma}^2 = \frac{1}{n} \sum_{i=1}^n (X_i - \bar{X})^2.$$

Since

$$\frac{(n-1)s^2}{\sigma^2} \sim \chi_{n-1}^2.$$

then

$$\frac{n\hat{\sigma}^2}{\sigma^2} \sim \chi_{n-1}^2$$

and

$$\Pr(\chi_{n-1}^2(1 - \alpha/2) \leq \frac{n\hat{\sigma}^2}{\sigma^2} \leq \chi_{n-1}^2(\alpha/2)) = 1 - \alpha$$

where $\chi_m^2(\alpha)$ denotes the point beyond which the chi-squared distribution with m degrees of freedom has probability α . With some manipulation

$$\Pr\left(\frac{n\hat{\sigma}^2}{\chi_{n-1}^2(\alpha/2)} \leq \sigma^2 \leq \frac{n\hat{\sigma}^2}{\chi_{n-1}^2(1 - \alpha/2)}\right) = 1 - \alpha$$

hence the interval for σ^2 is

$$\left[\frac{n\hat{\sigma}^2}{\chi_{n-1}^2(\alpha/2)}, \frac{n\hat{\sigma}^2}{\chi_{n-1}^2(1 - \alpha/2)} \right].$$

- (b) Find $(1 - \alpha)100\%$ confidence interval for μ using t-statistic.

We can assume that $\hat{\mu} \sim N(\mu, \frac{s^2}{n})$, approximately from CLT. Since

$$\frac{\sqrt{n}(\bar{X} - \mu)}{s} \sim t_{n-1}$$

where s^2 is as defined in part (a) then we can build an interval for \bar{X} without knowing σ^2 . Let $t_m(\alpha)$ denote that point beyond which

the t-distribution with m degrees of freedom has probability α . By symmetry of t

$$\Pr(-t_{n-1}(\alpha/2) \leq \frac{\sqrt{n}(\bar{X} - \mu)}{s} \leq t_{n-1}(\alpha/2)) = 1 - \alpha$$

which can be manipulated to

$$\Pr(\bar{X} - \frac{s}{\sqrt{n}}t_{n-1}(\alpha/2) \leq \mu \leq \bar{X} + \frac{s}{\sqrt{n}}t_{n-1}(\alpha/2)) = 1 - \alpha$$

forming the interval

$$[\bar{X} \pm \frac{s}{\sqrt{n}}t_{n-1}(\alpha/2)].$$

- (c) What is the probability that both $\mu \in [L_\mu(\alpha), U_\mu(\alpha)]$ and $\sigma^2 \in [L_{\sigma^2}(\alpha), U_{\sigma^2}(\alpha)]$.

Since s^2 appears in both pairs of endpoints, there is no reason to think that the intervals are independent. The event that $\mu \in [L_\mu(\alpha), U_\mu(\alpha)]$ and $\sigma^2 \in [L_{\sigma^2}(\alpha), U_{\sigma^2}(\alpha)]$ can be represented as the region in $\bar{X} - s$ plane bound by the following lines

$$\begin{aligned} L_1 : s &= \sqrt{\sigma^2 \chi_{n-1}^2(\alpha/2)/(n-1)} \\ L_2 : s &= \sqrt{\sigma^2 \chi_{n-1}^2(1-\alpha/2)/(n-1)} \\ L_3 : \bar{X} &= \mu + \frac{s}{\sqrt{n}}t_{n-1}(\alpha/2) \\ L_4 : \bar{X} &= \mu + \frac{s}{\sqrt{n}}t_{n-1}(1-\alpha/2). \end{aligned}$$

Then

$$\begin{aligned} \Pr(\mu \in [L_\mu(\alpha), U_\mu(\alpha)], \sigma^2 \in [L_{\sigma^2}(\alpha), U_{\sigma^2}(\alpha)]) \\ &= 1 - \Pr(\mu \notin [L_\mu(\alpha), U_\mu(\alpha)] \text{ or } \sigma^2 \notin [L_{\sigma^2}(\alpha), U_{\sigma^2}(\alpha)]) \\ &\geq 1 - \Pr(\mu \in [L_\mu(\alpha), U_\mu(\alpha)]) - \Pr(\sigma^2 \in [L_{\sigma^2}(\alpha), U_{\sigma^2}(\alpha)]) \\ &\geq 1 - 2\alpha \end{aligned}$$

- (d) Certainly there exist many confidence intervals for σ^2 , and the answer in (a) is one of them. Find the confidence interval which has minimum length.

Since the Chi-squared distribution is not symmetric for small values of m , you need find points on the distribution a, b such that $f(a) = f(b)$ and $F(b) - F(a) = 0$. There is some R code in the directory that will solve if degrees of freedom are provided.

To obtain a different confidence interval choose $0 < a < b$ such that

$$\Pr(a < \frac{(n-1)s^2}{\sigma^2} < b) = 1 - \alpha \rightarrow (*).$$

Therefore

$$\left(\frac{(n-1)s^2}{b}, \frac{(n-1)s^2}{a} \right)$$

is a $100(1-\alpha)\%$ confidence interval for σ^2 with length of

$$(n-1)s^2 \left(\frac{1}{a} - \frac{1}{b} \right).$$

We need to find a choice a and b that minimizes $\frac{1}{a} - \frac{1}{b}$ subject to the constraint (*). Since (*) determines b as a function of a we can use implicit differentiation:

$$\frac{d}{da} \int_a^b f_{n-1}(x) dx = \frac{d}{da} (1-\alpha)$$

where f_{n-1} is the density χ_{n-1}^2 . Then

$$-f_{n-1}(a) + f_{n-1}(b) \frac{db}{da} = 0 \Rightarrow \frac{db}{da} = \frac{f_{n-1}(a)}{f_{n-1}(b)}.$$

Now

$$\frac{d}{da} \left(\frac{1}{a} - \frac{1}{b} \right) = -\frac{1}{a^2} + \frac{1}{b^2} \frac{f_{n-1}(a)}{f_{n-1}(b)}$$

which equals 0 when $a^2 f_{n-1}(a) = b^2 f_{n-1}(b)$. This result can be verified as a minimum thus an interval of minimum length when a, b such that

$$a^2 f_{n-1}(a) = b^2 f_{n-1}(b)$$

2. Likelihood ratio test

Consider i.i.d. observations $X_1, \dots, X_n \sim p_\theta$. One needs to test

$$H_0 : \theta = \theta_0, H_1 : \theta = \theta_1.$$

The likelihood ratio test is

$$\mathbb{I} \left\{ \prod_{i=1}^n \frac{p_{\theta_1}(X_i)}{p_{\theta_0}(X_i)} > C \right\},$$

where C is some number such that the Type-I error is α .

- (a) Find the α -level likelihood ratio test for $H_0 : \theta = 5$ vs $H_1 : \theta = 10$ with observations $X_1, \dots, X_n \sim N(\theta, 1)$.

Reject H_0 when $\bar{X} > C$, e.g. large values of \bar{X} should be considered evidence for H_1 .

Let

$$\begin{aligned} \alpha &= \Pr(\bar{X} > C \mid H_0) \\ &= \Pr\left(\frac{\sqrt{n}(\bar{X} - \mu_0)}{\sigma_0} > \frac{\sqrt{n}(C - \mu_0)}{\sigma_0} \mid H_0 \right) \\ &= \Pr\left(Z > \frac{\sqrt{n}(C - \mu_0)}{\sigma_0} \right) \end{aligned}$$

Stigler 6-6 derives where this idea comes from.

Where $Z \sim N(0, 1)$. To find C , observe

$$z_{1-\alpha} = \frac{\sqrt{n}(C - \mu_0)}{\sigma_0} \Rightarrow C = \mu_0 + z_{1-\alpha} \left(\frac{\sigma_0}{\sqrt{n}} \right).$$

We know $\mu_0 = 5$ and $\sigma_0 = \sigma = 1$ so

$$C = 5 + z_{1-\alpha} \frac{1}{\sqrt{n}}.$$

The test will reject H_0 when $\bar{X} > C$ for observed X_1, \dots, X_n and given α .

(b) What is the power of the test when $n = 10, 50, 100$? Power is

$$\pi = \Pr(\bar{X} > C \mid H_1) = 1 - \beta$$

where

$$\begin{aligned} \beta &= \Pr(\bar{X} \leq C \mid H_1) \\ &= \Pr(\bar{X} - \mu_1 \leq C - \mu_1) \\ &= \Pr\left(Z \leq \left(5 + z_{1-\alpha} \frac{1}{\sqrt{n}}\right) - 10\right) \\ &= \Pr\left(Z \leq -5 + z_{1-\alpha} \left(\frac{1}{\sqrt{n}}\right)\right) \end{aligned}$$

Where $\beta \rightarrow \text{pnorm}(-5)$ as $n \rightarrow \infty$ so Power increases as n increases.

(c) Find the α -level likelihood ratio test for $H_0 : \theta = 0$ vs $H_1 : \theta = n^{-1/2}$ with observations $X_1, \dots, X_n \sim N(\theta, 1)$, What is the power of the test when $n = 10, 50, 100$?

Observe, where $n^{-1/2} = \frac{1}{\sqrt{n}}$ we reject H_0 when $\bar{X} > C$ where

$$C = z_{1-\alpha} \left(\frac{1}{\sqrt{n}} \right).$$

The power of this test is $1 - \beta$ where

$$\begin{aligned} \beta &= \Pr(\bar{X} \leq C \mid H_1) \\ &= \Pr\left(Z \leq \frac{z_{1-\alpha}}{\sqrt{n}} - \frac{1}{\sqrt{n}}\right) \\ &= \Pr\left(Z \leq \frac{1}{\sqrt{n}}(z_{1-\alpha} - 1)\right). \end{aligned}$$

As $n \rightarrow \infty$ $\beta \rightarrow .5$.

(d) Discuss your discovery.

Power and H_1 both depend on n sample size. When n increases, power decreases to .5 because it becomes harder to discern H_0 from H_1 .

3. Chi-squared

Consider i.i.d. observations $X_1, \dots, X_n \sim N(\mu, \sigma^2)$. The mean μ is unknown. Construct an α -level test for $H_0 : \sigma^2 = \sigma_0^2$ vs $H_1 : \sigma^2 \neq \sigma_0^2$.

The test statistic will be

$$\frac{(n-1)s^2}{\sigma_0^2} \sim \chi_{n-1}^2.$$

Reject H_0 if $\chi_{n-1}^2 \in W$ where $W = [0, z_1) \cup (z_2, \infty]$ and $z_1, z_2 \in \mathbb{R}$. To determine critical values, z_1, z_2 use Chi-squared table with $n-1$ degrees of freedom to satisfy

$$\alpha = \Pr\left(\frac{(n-1)s^2}{\sigma_0^2} \in W\right)$$

alternatively

$$1 - \alpha = \Pr\left(\frac{(n-1)s^2}{\sigma_0^2} \notin W\right)$$

which is similar to confidence interval found in question 1.

$$1 - \alpha = \Pr\left(\frac{(n-1)s^2}{\chi_{n-1}^2(\alpha/2)} \leq \sigma_0^2 \leq \frac{(n-1)s^2}{\chi_{n-1}^2(1-\alpha/2)}\right).$$

Which can be written as confidence interval

$$\left[\frac{(n-1)s^2}{\chi_{n-1}^2(\alpha/2)}, \frac{(n-1)s^2}{\chi_{n-1}^2(1-\alpha/2)} \right].$$

Therefore, our test rejects H_0 if σ_0^2 not in the CI.

Specifically stated, reject the null when

$$\frac{(n-1)s^2}{\sigma_0^2} < \chi_{n-1}^2(\alpha/2) \text{ or } \frac{(n-1)s^2}{\sigma_0^2} > \chi_{n-1}^2(1-\alpha/2)$$

where $\chi_{n-1}^2(a)$ is the a percentile of a Chi-squared distribution with $n-1$ degree of freedom.

4. Problem 1 on Page 362

A coin is thrown independently 10 times to test the hypothesis that the probability of heads is $\frac{1}{2}$ versus the alternative that the probability is not $\frac{1}{2}$. The test rejects if either 0 or 10 heads are observed.

(a) What is the significance level of the test?

$$\begin{aligned}\alpha &= \Pr(\text{Reject } H_0 \mid H_0) \\&= \Pr(X = 0 \text{ or } X = 10 \mid H_0) \\&= \Pr(X = 0) + \Pr(X = 10) \\&= \binom{10}{0}.5^0(.5)^{10} + \binom{10}{10}.5^{10}(1 - .5)^0 \\&= .00195\end{aligned}$$

(b) If in fact the probability of heads is .1, what is the power of the test?

$$\begin{aligned}\pi &= 1 - \beta = \Pr(\text{Reject } H_0 \mid p = .1) \\&= \Pr(X = 0 \text{ or } X = 10 \mid p = .1) \\&= \binom{10}{0}.1^0(1 - .1)^{10} + \binom{10}{10}.1^{10}(1 - .1)^0 \\&= .349\end{aligned}$$