MATH 392 Problem Set 5 (Corrected)

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9 March 2018

6.39

 $\{\hat{\beta}_n\}$ is consistent if and only if $\lim_{n\to\infty} P(|\hat{\beta}_n - \beta| < \varepsilon) = 1 \ \forall \ \varepsilon > 0$.

We know that for any n, the distribution of X_{max} follows the distribution

$$F_{X_{max}}(x) = \begin{cases} 0ifx > \beta \\ (\frac{x}{\beta})^n if0 < x < \beta \\ 0ifx < 0 \end{cases}$$

Thus

$$\lim_{n \to \infty} P(|\hat{\beta}_n - \beta| < \varepsilon) = \lim_{n \to \infty} P(\beta - \varepsilon < \hat{\beta}_n < \beta + \varepsilon)$$

$$= \lim_{n \to \infty} P(\beta - \varepsilon < X_{max} < \beta)$$

$$= \lim_{n \to \infty} (\frac{x}{\beta})^n |_{\beta - \varepsilon}^{\beta}$$

$$= 1 - \lim_{n \to \infty} (\frac{\beta - \varepsilon}{\beta})^n$$

$$= 1$$

6.40

 $\{\hat{\sigma}_n^2\}$ is consistent if and only if $\lim_{n\to\infty} P(|\hat{\sigma}_n^2 - \sigma^2| < \varepsilon) = 1 \ \forall \ \varepsilon > 0$. From Problem Set 4,

$$E[\hat{\sigma}_n^2] - \sigma^2 = \frac{n-1}{n}\sigma^2 - \sigma^2 = \frac{-1}{n}\sigma^2$$

It suffices to show that $\lim_{n\to\infty} \left|\frac{n-1}{n}-1\right| < \varepsilon \ \forall \varepsilon > 0.$

Let $\varepsilon > 0$. Then $\exists m \in \mathbb{N}$ st $\forall n \geq m, , \frac{1}{n} < \varepsilon$. Thus, $\forall n \geq m,$

$$\left|\frac{1}{n}\right|<\varepsilon\Rightarrow\left|\frac{1}{n}\right|<\varepsilon\Rightarrow\left|(\frac{n-1}{n}-\frac{n}{n})\right|<\varepsilon\Rightarrow\left|\frac{n-1}{n}-1\right|<\varepsilon\Rightarrow\left|E[\hat{\sigma}_n^2]-\sigma^2\right|<\varepsilon.$$

Thus, $\lim_{n\to\infty} P(|\hat{\sigma}_n^2 - \sigma^2| < \varepsilon) = 1 \forall \varepsilon > 0.$

7.3

 \mathbf{a}

$$X \sim N(185, 50^2), n = 100, \bar{X} = 210$$

$$P(q_{.05} < \frac{\bar{X} - \mu}{\sigma / \sqrt{n}} < q_{.95}) = 1 - \alpha = .9$$

Solve for μ :

$$P(\bar{X} - q_{.95} \frac{\sigma}{\sqrt{n}} < \mu < \bar{X} - q_{.05} \frac{\sigma}{\sqrt{n}}) = .9$$

$$(210 - 1.645 \frac{50}{\sqrt{100}} < \mu < 210 + .1.645 \frac{50}{\sqrt{100}}) = (201.8 < \mu < 218.8)$$

Thus the 90% confidence interval is (201.8, 218.8).

b

Solve for n:

$$\bar{X} - LB_{.05} = \bar{X} - (\bar{X} - 1.96(\frac{50}{\sqrt{n}})) \le 10$$

$$1.96\frac{50}{10} \le \sqrt{n}$$

$$n \ge 96.04$$

A sample size of at least 97 is needed.

 \mathbf{c}

Like b,

$$2.576\frac{50}{10} \leq \sqrt{n}$$

$$n \ge 165.9$$

A sample size of at least 166 is needed.

7.8

For n = 30, the code generating the simulation is below, along with the proportion of times the confidence interval missed μ . Compared to the next three numbers for n=60, n=100, and n=250, the frequency of missing μ converges to the thoretical frequency of .05.

```
# Run simulation

low <- 0

high <- 0

n <- 30

q <- qt(.975, n-1)

N <- 10^5
```

```
for(i in 1:N){
    x <- rgamma(n,shape = 5, rate = 2)
    xbar <- mean(x)
    s <- sd(x)
    L <- xbar - q*s/sqrt(n)
    U <- xbar + q*s/sqrt(n)
    if(U<5/2){
        low <- low + 1
    }
    if(5/2<L){
        high <- high + 1
    }
}</pre>
(high+low)/N
```

[1] 0.05396

[1] 0.05194

[1] 0.05169

[1] 0.05009

7.34

We are given

$$X \sim Gamma(2, \lambda), f_X(x) = \frac{xe^{-x/\lambda}}{\Gamma(2)\lambda^2}$$

$$2\lambda x \sim \chi_{df=4}^2, f_{2\lambda X}(2\lambda x) = \frac{(2\lambda x)^{4/2-1} e^{(-2\lambda x)^2/2}}{2^{4/2} \Gamma(4/2)}$$

Thus, the quantiles will be calculated in terms of $2\lambda x$, using the $\chi^2_{df=4}$ distribution.

$$.95 = P(q_{1,x} < \lambda < q_{2,x}) = P(q_{1,2\lambda x} < 2\lambda X < q_{2,2\lambda x})$$

$$= P(\frac{q_{1,2\lambda x}}{2X} < \lambda < \frac{q_{2,2\lambda x}}{2X})$$

$$= P(\frac{.2422}{X} < \lambda < \frac{5.572}{X})$$

Thus the 95% confidence interval for λ is $(\frac{.2422}{X},\frac{5.572}{X}).$