50.034 – Introduction to Probability and Statistics

January–May Term, 2019

Homework Set 7

Due by: Week 11 Cohort Class (11 Apr 2019 or 12 Apr 2019)

Note: The tables of values for the χ^2 distribution and the t distribution can be found at the back of the course textbook.

Question 1. Let $\{X_1,\ldots,X_n\}$ be a random sample of n observable normal random variables with mean μ and variance σ^2 . Assume that $n \geq 3$, and that $\sigma^2 \neq 0$. Let $\hat{\sigma}^2$ denote the biased sample variance of this random sample $\{X_1,\ldots,X_n\}$. For each of the two inequalities given below, determine the smallest possible value for the sample size n such that the inequality is satisfied:

- (i) $\Pr\left(\frac{\hat{\sigma}^2}{\sigma^2} \le 1.2\right) \ge 0.8.$ (ii) $\Pr\left(|\hat{\sigma}^2 0.77\sigma^2| \le 0.37\sigma^2\right) \ge 0.725.$

(i) The random variable $\frac{n\hat{\sigma}^2}{\sigma^2}$ has a χ^2 distribution with (n-1) degrees of freedom. Note that $\Pr\left(\frac{\hat{\sigma}^2}{\sigma^2} \leq 1.2\right) \geq 0.8$ is equivalent to $\Pr\left(\frac{n\hat{\sigma}^2}{\sigma^2} \leq 1.2n\right) \geq 0.8$. To find the smallest value for n satisfying this inequality, it suffices to find an integer value for n such that $\Pr\left(\frac{n\hat{\sigma}^2}{\sigma^2} \leq 1.2n\right) \geq 0.8$; and $\Pr\left(\frac{(n-1)\hat{\sigma}^2}{\sigma^2} \leq 1.2(n-1)\right) < 0.8$.

From the table of values for the χ^2 distribution (at the back of the course textbook), we

- $\Pr\left(\frac{19\hat{\sigma}^2}{\sigma^2} \le 1.2 \times (19) = 22.8\right) > \Pr\left(\frac{19\hat{\sigma}^2}{\sigma^2} \le 22.76\right) = 0.8.$ $\Pr\left(\frac{18\hat{\sigma}^2}{\sigma^2} \le 1.2 \times (18) = 21.6\right) < \Pr\left(\frac{18\hat{\sigma}^2}{\sigma^2} \le 21.61\right) = 0.8.$

Therefore, the smallest sample size n satisfying $\Pr\left(\frac{\hat{\sigma}^2}{\sigma^2} \le 1.2\right) \ge 0.8$ is n = 19.

(ii) Recall that the random variable $\frac{n\hat{\sigma}^2}{\sigma^2}$ has a χ^2 distribution with (n-1) degrees of freedom. Note that

$$\Pr\left(|\hat{\sigma}^2 - 0.77\sigma^2| \le 0.37\sigma^2\right) \ge 0.725 \iff \Pr\left(|\frac{\hat{\sigma}^2}{\sigma^2} - 0.77| \le 0.37\right) \ge 0.725$$

$$\iff \Pr\left(-0.37 \le \frac{\hat{\sigma}^2}{\sigma^2} - 0.77 \le 0.37\right) \ge 0.725$$

$$\iff \Pr\left(0.4 \le \frac{\hat{\sigma}^2}{\sigma^2} \le 1.14\right) \ge 0.725$$

$$\iff \Pr\left(0.4n \le \frac{n\hat{\sigma}^2}{\sigma^2} \le 1.14n\right) \ge 0.725.$$

To find the smallest value for n satisfying this inequality, it suffices to find an integer value for n such that

- Pr $\left(0.4n \le \frac{n\hat{\sigma}^2}{\sigma^2} \le 1.14n\right) \ge 0.725$; and Pr $\left(0.4(n-1) \le \frac{(n-1)\hat{\sigma}^2}{\sigma^2} \le 1.14(n-1)\right) < 0.725$.

From the table of values for the χ^2 distribution, we find that

$$\Pr(0.4(17) \le \frac{17\hat{\sigma}^2}{\sigma^2} \le 1.14(17)) = \Pr(6.8 \le \frac{17\hat{\sigma}^2}{\sigma^2} \le 19.38)$$

$$= \Pr(\frac{17\hat{\sigma}^2}{\sigma^2} \le 19.38) - \Pr(\frac{17\hat{\sigma}^2}{\sigma^2} \le 6.8)$$

$$> \Pr(\frac{17\hat{\sigma}^2}{\sigma^2} \le 19.37) - \Pr(\frac{17\hat{\sigma}^2}{\sigma^2} \le 6.908)$$

$$= 0.75 - 0.025 = 0.725.$$

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$$\Pr(0.4(16) \le \frac{16\hat{\sigma}^2}{\sigma^2} \le 1.14(16)) = \Pr(6.4 \le \frac{16\hat{\sigma}^2}{\sigma^2} \le 18.24)$$

$$= \Pr(\frac{16\hat{\sigma}^2}{\sigma^2} \le 18.24) - \Pr(\frac{16\hat{\sigma}^2}{\sigma^2} \le 6.4)$$

$$< \Pr(\frac{16\hat{\sigma}^2}{\sigma^2} \le 18.25) - \Pr(\frac{16\hat{\sigma}^2}{\sigma^2} \le 6.262)$$

$$= 0.75 - 0.025 = 0.725.$$

Therefore, the smallest sample size n satisfying $\Pr(|\hat{\sigma}^2 - 0.77\sigma^2| \le 0.37\sigma^2) \ge 0.725$ is n = 17.

Question 2. Consider a statistical model consisting of observable exponential random variables X_1, \ldots, X_n that are conditionally iid given the parameter θ . Let $\hat{\theta}$ be the maximum likelihood estimator of θ . What is the conditional probability density function of the random variable $n(\hat{\theta})^{-1}$ given $\theta = 5$? (Hint: Theorem 5.7.7 of the course textbook may be useful.)

Solution. As covered in Lecture 16 (slide 21), the maximum likelihood estimator of θ is $\hat{\theta} = \frac{1}{\overline{X_n}} = \frac{n}{X_1 + \dots + X_n}$, where \overline{X}_n denotes the sample mean of $\{X_1, \dots, X_n\}$. We are given that each X_i has the exponential distribution with parameter θ , or equivalently, each X_i has the gamma distribution with parameters 1 and θ . Since X_1, \dots, X_n are independent, it follows from Theorem 5.7.7 of the course textbook that $X_1 + \dots + X_n$ has the gamma distribution with parameters n and θ . Since $n(\hat{\theta})^{-1} = X_1 + \dots + X_n$, it follows that the conditional probability density function of $n(\hat{\theta})^{-1}$ given $\theta = 5$, which we denote by $f_{n(\hat{\theta})^{-1}|\theta}(x|5)$, is the probability density function of a gamma random variable with parameters n and n. Therefore, using $\Gamma(n) = (n-1)!$ (for any positive intege n), we get

$$f_{n(\hat{\theta})^{-1}|\theta}(x|5) = \begin{cases} \frac{5^n}{(n-1)!} x^{n-1} e^{-5x}, & \text{if } x \ge 0. \\ 0, & \text{if } x < 0; \end{cases}$$

Question 3. An automated juice vending machine dispenses orange juice in cups. A total of 25 dispensed cups of orange juice are collected, and the amount of juice in each cup is measured. Consider a statistical model consisting of a random sample $\{X_1, \ldots, X_{25}\}$ of normal random variables with unknown mean μ and unknown variance σ^2 , where each X_i represents the amount of juice (in ml) in the *i*-th dispensed cup. Let \overline{X}_{25} and s_{25}^2 be the sample mean and the unbiased sample variance respectively of $\{X_1, \ldots, X_{25}\}$.

- (i) Find an exact 90% confidence interval for μ in terms of \overline{X}_{25} and s_{25}^2 .
- (ii) After measuring the amount of juice in these 25 dispensed cups, it was noticed that the 25 cups have a mean amount of 251 ml and a standard deviation of 4 ml. Using this information, find an observed value of the 90% confidence interval that you found in the previous part.

Solution. (i) Note that $Z = \frac{\sqrt{25}(\overline{X}_{25} - \mu)}{s_{25}} = \frac{5(\overline{X}_{25} - \mu)}{s_{25}}$ has the *t*-distribution with 24 degrees of freedom. Let F(z) be the cumulative distribution function of Z. Then for any real number r > 0,

$$\Pr(\overline{X}_{25} - r < \mu < \overline{X}_{25} + r) = \Pr(-r < \overline{X}_{25} - \mu < r) = \Pr\left(-\frac{5r}{s_{25}} < \frac{5(\overline{X}_{25} - \mu)}{s_{25}} < \frac{5r}{s_{25}}\right)$$

$$= F\left(\frac{5r}{s_{25}}\right) - F\left(-\frac{5r}{s_{25}}\right) = F\left(\frac{5r}{s_{25}}\right) - \left(1 - F\left(\frac{5r}{s_{25}}\right)\right)$$

$$= 2 \cdot F\left(\frac{5r}{s_{25}}\right) - 1.$$

Thus $\Pr(\overline{X}_{25} - r < \mu < \overline{X}_{25} + r) = 0.9$ if and only if $2 \cdot F(\frac{5r}{s_{25}}) - 1 = 0.9$, or equivalently, if and only if $F(\frac{5r}{s_{25}}) = 0.95$. From the table of values for the *t*-distribution (at the back of the course textbook), the

From the table of values for the t-distribution (at the back of the course textbook), the closest value of z (that we can find from the table) satisfying F(z) = 0.95 for 24 degrees of freedom is z = 1.711, hence $\frac{5r}{s_{25}} \approx 1.711$, which implies $r \approx 0.3422s_{25}$. Therefore an exact 90% confidence interval for μ is $(\overline{X}_{25} - 0.3422s_{25}, \overline{X}_{25} + 0.3422s_{25})$.

(ii) We are given that some specific observed values $X_1=x_1,\ldots,X_{25}=x_{25}$ satisfy $\overline{x}_{25}=x_{25}$ $\frac{x_1 + \dots + x_{25}}{25} = 251$, as well as

$$\frac{1}{25} \sum_{i=1}^{25} (x_i - \overline{x}_{25})^2 = 4^2 = 16,$$

which is the observed value of the biased sample variance $\hat{\sigma}_{25}^2(X_1,\ldots,X_{25})$. Note that

$$s_{25}^2(X_1,\ldots,X_{25}) = \frac{25}{24}\hat{\sigma}_{25}^2(X_1,\ldots,X_{25}),$$

hence $s_{25}(x_1,\ldots,x_{25})=\sqrt{\frac{25}{24}}\hat{\sigma}_{25}(x_1,\ldots,x_{25})=\frac{5}{\sqrt{24}}(4)=\frac{20}{\sqrt{24}}.$ Therefore, an observed value of the 90% confidence interval found in the previous part is

the open interval

$$\left(251 - 0.3422\frac{20}{\sqrt{24}}, 251 + 0.3422\frac{20}{\sqrt{24}}\right) \approx \left(249.603, 252.397\right).$$

Question 4. At a Fuji apple farm in Fujisaki, Japan, 10 Fuji apples are randomly selected and weighed. Consider a statistical model consisting of a random sample $\{X_1,\ldots,X_{10}\}$ of normal random variables with unknown mean μ and unknown variance σ^2 , where each X_i represents the weight (in grams) of the *i*-th selected apple. Let \overline{X}_{10} and s_{10}^2 be the sample mean and the unbiased sample variance respectively of $\{X_1, \ldots, X_{10}\}$.

- (i) Find an exact 95% confidence interval for μ in terms of \overline{X}_{10} and s_{10}^2 .
- (ii) The measurements of all 10 weights (in grams) are indicated below:

Using these values, find an observed value of the 95% confidence interval for μ that you found in the previous part.

tion. (i) Note that $Z = \frac{\sqrt{10}(\overline{X}_{10} - \mu)}{s_{10}}$ has the t-distribution with 9 degrees of freedom. Let F(z) be the cumulative distribution function of Z. Then for any real number r > 0,

$$\begin{split} \Pr(\overline{X}_{10} - r < \mu < \overline{X}_{10} + r) &= \Pr(-r < \overline{X}_{10} - \mu < r) = \Pr\left(-\frac{\sqrt{10}r}{s_{10}} < \frac{\sqrt{10}(\overline{X}_{10} - \mu)}{s_{10}} < \frac{\sqrt{10}r}{s_{10}}\right) \\ &= F\left(\frac{\sqrt{10}r}{s_{10}}\right) - F\left(-\frac{\sqrt{10}r}{s_{10}}\right) = F\left(\frac{\sqrt{10}r}{s_{10}}\right) - \left(1 - F\left(\frac{\sqrt{10}r}{s_{10}}\right)\right) \\ &= 2 \cdot F\left(\frac{\sqrt{10}r}{s_{10}}\right) - 1. \end{split}$$

Thus $\Pr(\overline{X}_{10}-r<\mu<\overline{X}_{10}+r)=0.95$ if and only if $2\cdot F(\frac{\sqrt{10}r}{s_{10}})-1=0.95$, or equivalently, if and only if $F(\frac{\sqrt{10}r}{s_{10}}) = 0.975$. From the table of values for the t-distribution (at the back of the course textbook), the

closest value of z (that we can find from the table) satisfying F(z) = 0.975 for 9 degrees of freedom is z=2.262, hence $\frac{\sqrt{10}r}{s_{10}}\approx 2.262$, which implies $r\approx 0.7153s_{10}$. Therefore an exact 95% confidence interval for μ is $(\overline{X}_{10} - 0.7153s_{10}, \overline{X}_{10} + 0.7153s_{10})$.

(ii) We are given the vector of observed values

$$(x_1, \ldots, x_{10}) = (148, 150, 155, 154, 152, 148, 155, 160, 152, 145).$$

We calculate that $\overline{x}_{10} = \frac{x_1 + \dots + x_{10}}{10} = 151.9$. We also calculate that the observed value of the unbiased sample variance is

$$\frac{1}{9} \sum_{i=1}^{10} (x_i - \overline{x}_{10})^2 = \frac{1}{9} \Big[(-3.9)^2 + (-1.9)^2 + (3.1)^2 + (2.1)^2 + (0.1)^2 + (-3.9)^2 + (3.1)^2 + (8.1)^2 + (0.1)^2 + (-6.9)^2 \Big]$$
$$= \frac{170.9}{9} \approx 18.989.$$

Therefore, an observed value of the 95% confidence interval found in the previous part is the open interval

$$\left(151.9 - 0.7153\sqrt{\frac{170.9}{9}}, 151.9 + 0.7153\sqrt{\frac{170.9}{9}}\right) \approx \left(148.783, 155.017\right).$$

Question 5. At a steel factory, the tensile strength of 20 randomly cut steel sample pieces are measured. Consider a statistical model consisting of a random sample $\{X_1, \ldots, X_{20}\}$ of normal random variables with unknown mean μ and unknown variance σ^2 , where each X_i represents the tensile strength (in MPa) of the *i*-th steel sample piece. Let \overline{X}_{20} and s_{20}^2 be the sample mean and the unbiased sample variance respectively of $\{X_1, \ldots, X_{20}\}$.

- (i) Find an exact 95% confidence interval for the variance σ^2 in terms of \overline{X}_{20} and s_{20}^2 .
- (ii) After the 20 measurements have been collected, it was noticed that the 20 steel sample pieces have a mean tensile strength of 355 MPa and a standard deviation of 25 MPa. Using this information, find an observed value of the 95% confidence interval that you found in the previous part.

Solution. (i) Let $\hat{\sigma}_{20}^2$ be the unbiased sample variance of $\{X_1, \ldots, X_{20}\}$. Note that $Z = \frac{20\sigma_{20}^2}{\sigma^2}$ has the χ^2 distribution with 19 degrees of freedom. Let F(z) be the cumulative distribution function of Z. Then for any real number $c_1, c_2 > 0$,

$$\Pr(c_1\hat{\sigma}_{20}^2 < \sigma^2 < c_2\hat{\sigma}_{20}^2) = \Pr\left(\frac{1}{c_2\hat{\sigma}_{20}^2} < \frac{1}{\sigma^2} < \frac{1}{c_1\hat{\sigma}_{20}^2}\right) = \Pr\left(\frac{20}{c_2} < \frac{20\hat{\sigma}_{20}^2}{\sigma^2} < \frac{20}{c_1}\right)$$
$$= F\left(\frac{20}{c_1}\right) - F\left(\frac{20}{c_2}\right).$$

We want to find values for c_1 , c_2 such that $\Pr(c_1\hat{\sigma}_{20}^2 < \sigma^2 < c_2\hat{\sigma}_{20}^2) = F(\frac{20}{c_1}) - F(\frac{20}{c_2}) = 0.95$. There are infinitely many such possible pairs of values for c_1 , c_2 . We shall find one specific pair.

Note that 0.975-0.025=0.95. From the table of values for the χ^2 -distribution (at the back of the course textbook), we have F(32.85)=0.975 and F(8.907)=0.025 (for 19 degrees of freedom), so we could set $\frac{20}{c_1}=32.85$, $\frac{20}{c_2}=8.907$, or equivalently, set $c_1\approx 0.6088$, $c_2\approx 2.2454$. Thus, an exact 95% confidence interval for the variance σ^2 is $(0.6088\hat{\sigma}_{20}^2<\sigma^2<2.2454\hat{\sigma}_{20}^2)$.

Finally, since we are asked to express the exact 95% confidence interval in terms of \overline{X}_{20} and s_{20}^2 , we shall use the fact that $\hat{\sigma}_{20}^2 = \frac{19}{20}s_{20}^2$ to get the following exact 95% confidence interval for σ^2 :

$$(0.57836s_{20}^2 < \sigma^2 < 2.1331s_{20}^2).$$

(ii) We are given that some specific observed values $X_1 = x_1, \ldots, X_{20} = x_{20}$ satisfy $\overline{x}_{20} = \frac{x_1 + \cdots + x_{20}}{20} = 355$, as well as $\frac{1}{20} \sum_{i=1}^{20} (x_i - \overline{x}_{20})^2 = 25^2 = 625$, which is the observed value of the biased sample variance $\hat{\sigma}_{20}^2(X_1, \ldots, X_{20})$.

Therefore, an observed value of the 95% confidence interval found in the previous part is the open interval

$$\left(0.6088\hat{\sigma}_{20}^2 < \sigma^2 < 2.2454\hat{\sigma}_{20}^2\right) \approx \left(380.5, 1403.4\right).$$