

ECO227Y5 Tutorial 18

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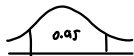
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Question 8.90

Question: Do SAT scores for high school students differ depending on the students' intended field of study? Fifteen students who intended to major in engineering were compared with 15 students who intended to major in language and literature. Given in the accompanying table are the means and standard deviations of the scores on the verbal and mathematics portion of the SAT for the two groups of students:

	Verbal	Math
Engineering	$\bar{y} = 446, s = 42$	$\bar{y} = 548, s = 57$
Language/literature	$\bar{y} = 534, s = 45$	$\bar{y} = 517, s = 52$

- (a) Construct a 95% confidence interval for the difference in average verbal scores of students majoring in engineering and of those majoring in language/literature.



Let \bar{x} be $\bar{y}_1 - \bar{y}_2$ $n_1 = n_2 = 15$, t statistic, $t_{df=28}$
- How spread out your estimate would be

$$\bar{x} = 446 - 534 = -88$$

$$S.E. = \sqrt{\frac{42^2}{15} + \frac{45^2}{15}} = 15.84, \quad t_{28}$$

$$P(-a \leq \frac{-88 - (m_1 - m_2)}{15.84} \leq a) = 0.95 \Rightarrow a = 2.048$$

95% C.I. for $\mu_1 - \mu_2$: $[-88 - 2.048 \cdot 15.84, -88 + 2.048 \cdot 15.84] = [-120.59, -55.46]$



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Language/literature	$\bar{y} = 534, s = 45$	$\bar{y}_2 = 517, s = 52$

- (b) Construct a 95% confidence interval for the difference in average math scores of students majoring in engineering and of those majoring in language/literature.

t_{n_1-1}, t_{n_2} again
Let $\bar{x} = \bar{y}_1 - \bar{y}_2$. $\bar{x} = 548 - 517 = 31$ $SE = \sqrt{\frac{s_1^2}{15} + \frac{s_2^2}{15}} = 19.92$

$$P\left(-a \leq \frac{\bar{x} - (\mu_1 - \mu_2)}{19.92} \leq a\right) = 0.95 \Rightarrow a = 2.048$$

$$95\% \text{ C.I. for } \mu_1 - \mu_2: \left[31 - 2.048 \cdot 19.92, 31 + 2.048 \cdot 19.92 \right] = [-9.79, 71.80]$$

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- (c) Interpret the results obtained in parts (a) and (b).
- (d) What assumptions are necessary for the methods used previously to be valid?

c.) Since the 95% C.I. for verbal is all negative, language/literature students are likely to score higher. For math, the interval contains 0, so there is a chance that there is no difference between scores.

d.) Independent samples

Question 8.93

Question: A factory operates with two machines of type A and one machine of type B. The weekly repair costs X for type A machines are normally distributed with mean μ_1 and variance σ^2 . The weekly repair costs Y for machines of type B are also normally distributed but with mean μ_2 and variance $3\sigma^2$. The expected repair cost per week for the factory is thus $2\mu_1 + \mu_2$. If you are given a random sample X_1, X_2, \dots, X_n on costs of type A machines and an independent random sample Y_1, Y_2, \dots, Y_m on costs for type B machines, show how you would construct a 95% confidence interval for $2\mu_1 + \mu_2$

(a) if σ^2 is known.

a.) want C.I. for $2\mu_1 + \mu_2$

$$V(\bar{X} + \bar{Y}) = 4 \frac{\sigma^2}{n} + \frac{3\sigma^2}{m} \Rightarrow \text{S.E.} = \sigma \sqrt{\frac{4}{n} + \frac{3}{m}}, \text{ on } \bar{X} + \bar{Y}$$

(b) if σ^2 is not known.

$$P(-a \leq \frac{(\bar{X} + \bar{Y}) - (2\mu_1 + \mu_2)}{\sigma \sqrt{\frac{4}{n} + \frac{3}{m}}} \leq a) = 0.95 \Rightarrow a = 1.96$$

b.) If σ^2 not known, we can use pooled

estimator instead:

$$s_p^2 = \frac{(n-1)s_x^2 + (m-1)s_y^2}{n+m-2}, \text{ t.d.f. } n+m-2$$

$$P(-a \leq \frac{(\bar{X} + \bar{Y}) - (2\mu_1 + \mu_2)}{s_p \sqrt{\frac{4}{n} + \frac{3}{m}}} \leq a) = 0.95 \Rightarrow a = 2.048$$

$$\Rightarrow 95\% \text{ C.I. for } 2\mu_1 + \mu_2 : \left(\bar{X} + \bar{Y} - 2.048 \cdot s_p \sqrt{\frac{4}{n} + \frac{3}{m}}, \bar{X} + \bar{Y} + 2.048 \cdot s_p \sqrt{\frac{4}{n} + \frac{3}{m}} \right)$$

$$\Rightarrow 95\% \text{ C.I. for } 2\mu_1 + \mu_2 : \left(\bar{X} + \bar{Y} - 1.96 \cdot \sigma \sqrt{\frac{4}{n} + \frac{3}{m}}, \bar{X} + \bar{Y} + 1.96 \cdot \sigma \sqrt{\frac{4}{n} + \frac{3}{m}} \right)$$

Question 8.97

Question: Suppose that S^2 is the sample variance based on a sample of size n from a normal population with unknown mean and variance. Derive a $100(1 - \alpha)\%$

(a) upper confidence bound for σ^2 .

(b) lower confidence bound for σ^2 .

Recall: $\frac{(n-1)S^2}{\sigma^2} \sim \chi_{n-1}^2$

a.) $P\left(\frac{(n-1)S^2}{\sigma^2} \geq \chi_{\alpha}^2\right) = 1 - \alpha$

$\Rightarrow P\left(\frac{1}{\sigma^2} \geq \frac{\chi_{\alpha}^2}{(n-1)S^2}\right) = 1 - \alpha$

$\Rightarrow P\left(\sigma^2 \leq \frac{(n-1)S^2}{\chi_{\alpha}^2}\right) = 1 - \alpha$

so $100(1-\alpha)\%$ upper C.I.: $\left(0, \frac{(n-1)S^2}{\chi_{\alpha}^2}\right)$

b.) $P\left(\frac{(n-1)S^2}{\sigma^2} \leq \chi_{1-\alpha}^2\right) = 1 - \alpha$

$\Rightarrow P\left(\frac{1}{\sigma^2} \leq \frac{\chi_{1-\alpha}^2}{(n-1)S^2}\right) = 1 - \alpha \Rightarrow P\left(\sigma^2 \geq \frac{(n-1)S^2}{\chi_{1-\alpha}^2}\right)$

$\Rightarrow 100(1-\alpha)\%$ lower bound C.I.: $\left(\frac{(n-1)S^2}{\chi_{1-\alpha}^2}, \infty\right)$

Question 9.3

Question: Let Y_1, Y_2, \dots, Y_n denote a random sample from the uniform distribution on the interval $(\theta, \theta + 1)$. Let

$$f(y) = \begin{cases} 1, & y \in (\theta, \theta + 1) \\ 0, & \text{elsewhere} \end{cases} \quad \hat{\theta}_1 = \bar{Y} - \frac{1}{2} \quad \text{and} \quad \hat{\theta}_2 = Y_{(n)} - \frac{n}{n+1}.$$

- (a) Show that both $\hat{\theta}_1$ and $\hat{\theta}_2$ are unbiased estimators of θ .
 (b) Find the efficiency of $\hat{\theta}_1$ relative to $\hat{\theta}_2$.

a.) $E(\hat{\theta}_1) = E(\bar{Y}) - \frac{1}{2} = \theta + \frac{1}{2} - \frac{1}{2} = \theta$ is unbiased $f(y) = y - \theta$

$$\begin{aligned} P(Y_{(n)} \leq y) &= P(Y_1 \leq y, \dots, Y_n \leq y) = F(y)^n = (y - \theta)^n \Rightarrow f(y_{(n)}) = n(y - \theta)^{n-1} \\ E(Y_{(n)}) &= \int_{\theta}^{\theta+1} n y (y - \theta)^{n-1} dy = n \int_{\theta}^{\theta+1} y (y - \theta)^{n-1} dy = n \int_{\theta}^{\theta+1} (\theta + y - \theta)(y - \theta)^{n-1} dy \\ &= n\theta \int_{\theta}^{\theta+1} (y - \theta)^{n-1} dy + n \int_{\theta}^{\theta+1} (y - \theta)^n dy = \theta + \frac{n}{n+1} \Rightarrow E(\hat{\theta}_2) = \theta + \frac{n}{n+1} - \frac{n}{n+1} = \theta \end{aligned}$$

b.) $\frac{\text{var}(\hat{\theta}_2)}{\text{var}(\hat{\theta}_1)} =$

$$\begin{aligned} \text{var}(\hat{\theta}_1) &= \text{var}(\bar{Y}), \quad \text{var}(Y_i) = \frac{1}{12} \Rightarrow \text{var}(\bar{Y}) = \frac{1}{12n} \\ \text{var}(\hat{\theta}_2) &= \text{var}(Y_{(n)}) = E(Y_{(n)}^2) - [E(Y_{(n)})]^2 \end{aligned}$$

Continued on next slide

Question 9.3

Question: Let Y_1, Y_2, \dots, Y_n denote a random sample from the uniform distribution on the interval $(\theta, \theta + 1)$. Let

$$\hat{\theta}_1 = \bar{Y} - \frac{1}{2} \quad \text{and} \quad \hat{\theta}_2 = Y_{(n)} - \frac{n}{n+1}.$$

- (a) Show that both $\hat{\theta}_1$ and $\hat{\theta}_2$ are unbiased estimators of θ .
- (b) Find the efficiency of $\hat{\theta}_1$ relative to $\hat{\theta}_2$.

$$b) E(Y_{(n)}) = \int_{\theta}^{\theta+1} y^2 n(y-\theta)^{n-1} dy \quad \begin{array}{l} u = y - \theta \quad y = u + \theta \\ du = dy \end{array}$$

$$= n \int_0^1 (\theta^2 + 2\theta u + u^2) u^{n-1} du$$

$$= n \int_0^1 (\theta^2 u^{n-1} + 2\theta u^n + u^{n+1}) du = \theta^2 + \frac{2n\theta}{n+1} + \frac{n}{n+2}$$

$$v(\hat{\theta}_2) = \left(\theta^2 + \frac{2n\theta}{n+1} + \frac{n}{n+2} \right) - \left(\theta + \frac{n}{n+1} \right)^2 = \left(\theta^2 + \frac{2n\theta}{n+1} + \frac{n}{n+2} \right) - \left(\theta^2 + \frac{2n\theta}{n+1} + \frac{n^2}{(n+1)^2} \right)$$

$$v(\hat{\theta}_2) = \frac{n}{n+2} - \frac{n^2}{(n+1)^2} = \frac{n}{(n+1)^2(n+2)} \quad \text{Eff}(\hat{\theta}_1 \text{ rel. } \hat{\theta}_2) = \frac{v(\hat{\theta}_2)}{v(\hat{\theta}_1)} = \frac{\frac{n}{(n+1)^2(n+2)}}{\frac{1}{12n}} = \frac{12n^2}{(n+1)^2(n+2)}$$

notice as $n \rightarrow \infty$ Eff $\rightarrow 0$ so $\hat{\theta}_2$ is more efficient.

Question 9.15

Question: Refer to Exercise 9.3. Show that both $\hat{\theta}_1$ and $\hat{\theta}_2$ are consistent estimators for θ .

Recall: If $\hat{\theta}$ is unbiased, and

$$\text{if } \lim_{n \rightarrow \infty} V(\hat{\theta}) = 0$$

Then $\hat{\theta} \xrightarrow{p} \theta$ (consistent)

faster
than definition
method

$$\hat{\theta}_1 = \bar{y} - \frac{1}{2}, \quad V(\hat{\theta}_1) = \frac{1}{12n}, \quad \lim_{n \rightarrow \infty} \frac{1}{12n} = 0$$

$$\hat{\theta}_2 = Y(n) - \frac{n}{n+1}, \quad V(\hat{\theta}_2) = \frac{12n^2}{(n+1)^2(n+2)}, \quad \lim_{n \rightarrow \infty} \frac{12n^2}{(n+1)^2(n+2)} = 0$$

\leftarrow degree 2
 \uparrow degree 3

Question 9.19

Question: Let Y_1, Y_2, \dots, Y_n denote a random sample from the probability density function

$$f(y) = \begin{cases} \theta y^{\theta-1}, & 0 < y < 1, \\ 0, & \text{elsewhere,} \end{cases}$$

where $\theta > 0$. Show that \bar{Y} is a consistent estimator of $\frac{\theta}{\theta+1}$.

$$\bar{Y} = \frac{1}{n} \sum_{i=1}^n Y_i = \frac{1}{n} \cdot n \left(\frac{\theta}{\theta+1} \right) = \frac{\theta}{\theta+1} \quad \text{unbiased}$$

$$E(Y_i) = \int_0^1 \theta y^\theta dy = \frac{\theta y^{\theta+1}}{\theta+1} \Big|_0^1 = \frac{\theta}{\theta+1}$$

$$E(Y_i^2) = \int_0^1 \theta y^{2\theta} dy = \frac{\theta}{2\theta+1}$$

$$V(\bar{Y}) = V\left(\frac{1}{n} \sum_{i=1}^n Y_i\right) = \frac{1}{n^2} \cdot n \cdot V(Y_i) = \frac{1}{n} \left(\frac{\theta}{2\theta+1} - \frac{\theta^2}{(\theta+1)^2} \right) \quad \text{clearly as } n \rightarrow \infty, V(\bar{Y}) \rightarrow 0$$

so consistent

$$V(Y_i) = E(Y_i^2) - E(Y_i)^2 = \frac{\theta}{2\theta+1} - \frac{\theta^2}{(\theta+1)^2}$$

Question 9.20

Question: If Y has a binomial distribution with n trials and success probability p , show that Y/n is a consistent estimator of p .

$$E\left(\frac{Y}{n}\right) = \frac{E(Y)}{n} = \frac{np}{n} = p \quad \text{unbiased} \quad Y \sim \text{Bin}(n, p)$$
$$V\left(\frac{Y}{n}\right) = \frac{V(Y)}{n^2} = \frac{np(1-p)}{n^2} = \frac{p(1-p)}{n} \rightarrow 0 \text{ as } n \rightarrow \infty \text{ so consistent}$$

Question 9.23

Question: Refer to Exercise 9.21. Suppose that Y_1, Y_2, \dots, Y_n is a random sample of size n from a population for which the first four moments are finite. That is, $m_1 = E(Y_1) < \infty$, $m_2 = E(Y_1^2) < \infty$, $m_3 = E(Y_1^3) < \infty$, and $m_4 = E(Y_1^4) < \infty$. (Note: This assumption is valid for the normal and Poisson distributions in Exercises 9.21 and 9.22, respectively.) Again, assume that $n = 2k$ for some integer k . Consider

\uparrow
if even

$$\hat{\sigma}^2 = \frac{1}{2k} \sum_{i=1}^k (Y_{2i} - Y_{2i-1})^2.$$

(a) Show that $\hat{\sigma}^2$ is an unbiased estimator for σ^2 .

$$E(\hat{\sigma}^2) = \frac{1}{2k} \sum_{i=1}^k E((Y_{2i} - Y_{2i-1})^2) = \frac{1}{2k} \sum_{i=1}^k E(Y_{2i}^2 - 2Y_{2i}Y_{2i-1} + Y_{2i-1}^2)$$

- independent

$$= \frac{1}{2k} \sum_{i=1}^k (E(Y_{2i}^2) - 2E(Y_{2i})E(Y_{2i-1}) + E(Y_{2i-1}^2))$$

Recall: $E(X^2) = V(X) + E(X)^2$

But $Y_j \sim \dots$, so $E(Y_{2i}) = E(Y_{2i-1})$
etc

$$= \frac{1}{2k} \sum_{i=1}^k (V(Y_{2i}) + E(Y_{2i})^2 - 2E(Y_{2i})E(Y_{2i-1}) + V(Y_{2i-1}) + E(Y_{2i-1})^2)$$

$$= \frac{1}{2k} \sum_{i=1}^k (\sigma^2 + \mu^2 - 2\mu^2 + \sigma^2 + \mu^2) = \frac{1}{2k} \sum_{i=1}^k 2\sigma^2$$

$$= \frac{1}{k} \sum_{i=1}^k \sigma^2 = \frac{k\sigma^2}{k} = \sigma^2$$

Question 9.23

Question: Refer to Exercise 9.21. Suppose that Y_1, Y_2, \dots, Y_n is a random sample of size n from a population for which the first four moments are finite. That is, $m_1 = E(Y_1) < \infty$, $m_2 = E(Y_1^2) < \infty$, $m_3 = E(Y_1^3) < \infty$, and $m_4 = E(Y_1^4) < \infty$. (Note: This assumption is valid for the normal and Poisson distributions in Exercises 9.21 and 9.22, respectively.) Again, assume that $n = 2k$ for some integer k . Consider

$$\hat{\sigma}^2 = \frac{1}{2k} \sum_{i=1}^k (Y_{2i} - Y_{2i-1})^2.$$

(b) Show that $\hat{\sigma}^2$ is a consistent estimator for σ^2 .

(c) Why did you need the assumption that $m_4 = E(Y_1^4) < \infty$?

b.) Define $\hat{\sigma}^2 = \frac{1}{2k} \sum_{i=1}^k J_i$, $J_i = (Y_{2i} - Y_{2i-1})^2$, J_i iid
and $E(J_i) = 2\sigma^2$. $\text{var}(\hat{\sigma}^2) = \text{var}\left(\frac{1}{2k} \sum_{i=1}^k J_i\right) = \frac{1}{4k} \cdot \text{var}(J_i) = \frac{\text{var}(J_i)}{4k}$ so if $\text{var}(J_i) < \infty$

c.) To make limit of $\text{var}(\hat{\sigma}^2)$ as $n \rightarrow \infty$ become 0.

Question 9.24

Question: Let $Y_1, Y_2, Y_3, \dots, Y_n$ be independent standard normal random variables.

- (a) What is the distribution of $\sum_{i=1}^n Y_i^2$?
- (b) Let $W_n = \frac{1}{n} \sum_{i=1}^n Y_i^2$. Does W_n converge in probability to some constant? If so, what is the value of the constant?

a.) $Y_i \stackrel{iid}{\sim} N(0, 1)$

$$\sum_{i=1}^n Y_i^2 \sim \chi_n^2$$

$$V(Y_i) = 1 = E(Y_i^2) - E(Y_i)^2 \Rightarrow E(Y_i^2) = 2$$

b.) $E(W_n) = \frac{1}{n} \sum_{i=1}^n E(Y_i^2) = \frac{n \cdot 2}{n} = 2$

$$V(W_n) = \frac{1}{n^2} \sum_{i=1}^n V(Y_i^2) = \frac{2}{n} \rightarrow 0 \text{ as } n \rightarrow \infty \text{ by Law of large numbers}$$

and consistency