PO91Q: Fundamentals in Quantitative Research Methods

Worksheet Week 4 - Solutions

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Confidence Intervals

Conceptual

- 1. A researcher is analysing individuals' relative fear of being a victim of burglary on a 1-100 scale. A random sample of 9 individuals found a mean score of 47 on the scale with a sample variance of 158.76 for fear of being burgled.
 - a. What distribution would be used to calculate an 80% confidence interval around this mean?

A t-distribution as we don't know the population standard deviation and n is

b. Construct that interval.

$$\bar{x} = 47$$

$$n = 9$$

t from tables = 1.397

$$s = \sqrt{158.76}$$

$$s = 12.6$$

Confidence interval formula

$$\bar{x} \pm t \times \frac{s}{c}$$

 $\bar{x} \pm t \times \frac{s}{\sqrt{n}}$ Lower bound

$$47 - 1.397 \times \frac{12.6}{\sqrt{9}} = 47 - 5.867 = 41.13$$

Upper bound

$$47 + 1.397 \times \frac{12.6}{\sqrt{9}} = 47 + 5.867 = 52.87$$



- 2. We are investigating the height of men in the UK. For this we have obtained a random sample of 100 UK men and found they had a mean height of 180cm with a standard deviation of 10cm.
 - a. Construct a 95% confidence interval for the mean height of UK males.

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\bar{x} = 180
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s = 10

n = 100

As the population standard deviation is not known, the t distribution and t need to be used.

Find the t-score for a 95% confidence interval in the t-table with 99 df.

$$t = 1.984$$

Confidence interval:

$$\bar{x} \pm t \times \frac{s}{\sqrt{n}}$$

Lower bound:

$$180 - 1.984 \times \frac{10}{\sqrt{100}} = 180 - 1.984 = 178.02$$

Upper bound:

$$180 + 1.984 \times \frac{10}{\sqrt{100}} = 181.98$$

- b. Select all true statements concerning the constructed confidence interval and justify your choice for each statement.
 - i. The probability of the population mean being within the upper and lower bounds is 95%.

FALSE - The population mean is fixed but unknown and therefore can either be inside the bounds or outside. The Probability is therefore 50%.

ii. 95% of men's heights fall between the upper and lower bound.

FALSE - The distribution calculated is not the distribution of men's height, but the sampling distribution of the mean male height.

- iii. 95% of the cases in the sample fall between the upper and lower bound.
 - FALSE The distribution calculated is not of men's height in this sample, but the sampling distribution of the mean male height.
- iv. On average 95% of confidence intervals constructed would contain the population mean.

TRUE

- v. On average 95% of the means of samples with 100 respondents will fall within the upper and lower bands.
 - FALSE This confidence interval is not making statements about various sample means but rather about the population mean.
- vi. On average 95% of the sample means equal the population mean.

 FALSE The confidence interval is a range and does not make claims about where the population mean is exactly.

Significance Testing

Conceptual – Working with the App

- 1. Same effect size, different n (mapping t, n, and \hat{d})
 - a. With t=1.6 and n=16, $\hat{d}=t/\sqrt{n}=1.6/4=0.40$. The power at n=16 for $\hat{d}\approx 0.40$ and $\alpha=0.05$ will be modest (typically well below 0.80).
 - b. Keeping \hat{d} fixed at 0.40 and raising n to 64 requires $t = \hat{d}\sqrt{n} = 0.40 \times 8 = 3.2$. The app will show the same \hat{d} but a higher power at n = 64.
 - c. As n increases while the underlying effect stays fixed, the standard error shrinks: SE = s/\sqrt{n} . The test statistic is $t = \frac{\bar{x} \mu_0}{\text{SE}}$, so a smaller SE makes |t| larger on average. For a fixed α , the critical cutoff (e.g., $t_{\alpha/2,df}$ for a two-sided test) is essentially fixed, so larger typical |t| increases the chance that $|t| > t_{\alpha/2,df}$. Therefore, bigger $n \Rightarrow$ less sampling noise \Rightarrow tighter estimates \Rightarrow higher power.
- 2. Planning with a SESOI and α sensitivity
 - a. For t=2.0 and n=30, $\hat{d}=t/\sqrt{n}\approx 2.0/\sqrt{30}\approx 0.37$. The app will report the power at n=30 using this \hat{d} (typically moderate at $\alpha=0.05$).
 - b. Turning on the SESOI with d=0.5 switches the curve to a fixed target effect. The orange marker shows the n giving 80% power at $\alpha=0.05$; for d=0.5 this is typically in the few-dozen range for a one-sample t-test (on the order of the 30s).
 - c. Increasing α to 0.10 lowers the critical threshold and reduces the required n for a given power; decreasing α to 0.01 raises the threshold and increases the required n. Formally, stricter α increases the critical t value, so a larger n is needed for the same probability of exceeding it under the alternative.
- 3. Was the study well powered? Post-hoc check and replication planning
 - a. With t=2.1 and n=25, $\hat{d}=2.1/\sqrt{25}=2.1/5=0.42$. The app will show the power at n=25 for $\hat{d}\approx 0.42$ and $\alpha=0.05$; this is typically only moderate, not comfortably high.
 - b. For replication planning, turn on SESOI and choose the closest option to \hat{d} (e.g., d = 0.5). The orange marker will give the n needed for 80% power at the chosen α ; expect a sample size in the few-dozen range for d = 0.5 at α = 0.05.
 - c. A significant result with low power can be fragile because small shifts in sampling may miss the effect and estimates are noisier. Picking a SESOI (a target *d* tied to practical importance) and planning for 80% power helps ensure a replication has adequate sensitivity to detect a meaningfully sized effect.

Summarising Conclusions

- Power is the probability your test will detect a real effect (reject H₀ when the effect truly exists).
- The curve fixes an effect size d and shows how power increases with n.
- If you base d on your observed test ($\hat{d} = t/\sqrt{n}$), the curve answers: "If the effect really was the size we observed, how likely would the test be to detect that effect at other sample sizes n?"
- If you base *d* on a meaningful minimum (SESOI), the curve answers: "How large should *n* be to reliably detect an effect that actually matters?"
- The 0.80 (80%) line is a convention (not a law). Many fields use it as a planning minimum; higher power (e.g., 90%) is desirable when stakes are high.
- Rule of thumb: because $t = d\sqrt{n}$, halving d requires about 4× the sample size to keep power roughly the same.

Why do many plots show an 80% power line?

- 80% power means choosing β = 0.20 (so power = 1 β = 0.80). This is a planning convention, not a mathematical law.
- The convention became common through methodological guidance (e.g., Cohen's power analysis texts) and is echoed in many applied fields.
- In higher-stakes settings (e.g., confirmatory clinical trials), targets of 80–90% power are typical; 90

Applied Exercises

See RScript in the Online Companion