

Answers

Week 1

Theory

- Population mean: \bar{x} , Sample mean, μ
- H_0 is what we expect to see happen based on the knowledge we have, H_a is what we suppose happens if the null hypothesis is not true
- The p-value is the probability that you find what you find
- The limit after which we state that the null hypothesis is not true and we assume the alternative hypothesis as probable.

Application

2 and 5 are correct

Week 2

- Event A alters event B or vice versa.
- Event A and B cannot exist or occur at the same time.
- They cannot.
- 1.** $P(A/B) = P(A) + P(B) - P(A\&B)$ **2.** $P(A/B) = P(A) + P(B)$ **3.** $1 - P(A) = P(B)$ **4.** $P(A\&B) = P(A) \times P(B|A)$ **5.** $P(A\&B) = P(A) \times P(B)$
- There is no chance of you knowing the outcome
- $P(x) = F(x)/n$
- $\mu_{\bar{x}} = \sum \bar{x}_n \times p_n$
- $\sigma_{\bar{x}}^2 = \sum (x_i - \mu_x)^2 \times p_i$
- $\mu_{x+y} = \mu_x + \mu_y$
- $\sigma_{x+y}^2 = \sigma_x^2 + \sigma_y^2 + 2p_{xy} \times \sigma_x \times \sigma_y$

Week 3

- Numerical data can only take certain, definite values, while categorical values can be anything.
- When they don't alter or influence each other's results in any way.
- Goodness of fit, Independence and homogeneity
- Number of cells that can be filled in freely with only restrictions of marginal totals
- Number of cells that can be filled in freely with only restrictions of marginal totals - 1
- $f_e(A \text{ and } B) = f(A) \times f(B) / n$
- $\chi^2 = \sum (\text{Observed} - \text{Expected}) / \text{Expected}$

Week 4

- H_0 is regular
- S_x
- σ_x
- $T = \bar{x} - \mu / S_x$

- e. $z = \bar{X} - \mu / \sigma_x$
- f. One-sided

Week 5

- a. $H_0: \mu_d = 0$ and (if one-sided) $H_a: \mu_d </> 0$ or (if two-sided) $H_a: \mu_d \neq 0$
- b. $H_0: \mu_1 - \mu_2 = 0$ and (if one-sided) $H_a: \mu_1 - \mu_2 </> 0$ or (if two-sided) $H_a: \mu_1 - \mu_2 \neq 0$
- c. 1. Df = number of pairs observed - 1. 2. Df = The smallest of the two n's - 1. 3. $n_1 + n_2 - 2$.
- d. $t = \bar{X}_1 - \bar{X}_2 / se$

$$se_{\bar{X}_1 - \bar{X}_2} = \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}$$

- e. Because most of the time $\mu_1 - \mu_2$ is equal to 0
- f. $t = D/se$ $se_d = S_d / \sqrt{n}$

Week 6

Theory

- a. That in 95% of all samples, the mean falls within two standard errors of the population mean.
- b. $CI_{1-\alpha} = \bar{X} \pm t^* \times SE$
 - a. $t^* = t_{\alpha/2}(df)$
 - b. $SE = s/\sqrt{n}$
- c. $CI_{1-\alpha} = (\bar{X}_1 - \bar{X}_2) \pm t^* \times SE_{\bar{X}_1 - \bar{X}_2}$
- d. Effect size: $\eta^2 = t^2 / (t^2 + df)$
- e. Cohen's d: $d = \bar{X} - \mu_0 / s$

Application

D is correct, a is technically true as well but has nothing to do with the interpretation of the 95% CI

Week 7

Theory

- a. Distribution-free data that does not, or rarely makes claims about a population, as well as for resampling tests or ordinal data
- b. When dealing with ordinal data or with continuous data that has a small n and a skewed sample distribution, in the case when you don't know anything about the population.
- c. Paired = signed rank, Independent = rank sum
- d. You take the average.
- e. $\mu = n(n+1) / 4$ $\sigma = n(n+1)(2n+1)/24$
- f. $\mu = n_1(n_1 + n_2 + 1) / 2$ $\sigma = n_1 \times n_2(n_1 + n_2 + 1)/12$
- g. $z = T - \mu_T / \sigma_T$ $T = W - \mu_W / \sigma_W$

Application

a. 4, 4, 5, 6, 6, 7, 8, 9, 9, 9, 10, 14, 14, 15, 15, 17, 18, 19, 22, 36

4	4	5	6	6	7	8	9	9	9	10	14	14	15	15	17	18	19	22	36
1.5	1.5	3	4.5	4.5	6	7	9	9	9	11	12.5	12.5	14.5	14.5	16	17	18	19	26

No brain damage (n=9) = 53 Mean is 115,5

brain damage (n=11) = 157 Mean is 94.5

Standard dev. = 13.16

Z = -3.153

P = 0.016

H₀ = rejected

Week 8

a. $\phi = \sqrt{(x^2/n)}$

b. $V = \sqrt{(x^2/n(k-1))}$