

Read H.O.3, Chp 3 in Design (ANOVA) book.

- 1) (a) Write a model for this study and explain all the terms in your model

$$Y = XB + \epsilon \quad \Leftrightarrow \quad y_{ij} = \mu_i + \epsilon_{ij}$$

See H.O.3
pg. 19

- (b) State the assumptions needed to perform an analysis of variance in this study.

- The ϵ_{ij} 's $\stackrel{\text{iid}}{\sim} N(0, \sigma^2)$
- The $y_{ij} \sim N(\mu_i, \sigma^2)$ under each treatment i
- The y_{ij} 's are independent

H.O.3 pg. 21

- (c) Produce the ANOV table for this study

Source of variation	df	Sum of Squares	Mean Square	F	P-value
Treatment	4	48569	12142	78.08	6.48×10^{-12}
Error	20	3110	156		
Total	24	51679			

See code
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see code
pg. 24

- (d) Compute the least squares estimates of the treatment means and their estimated standard errors

• Use `lsmeans()` function in R

$$\underline{\mu} = \begin{bmatrix} \text{Prestit} \\ \text{Fasting} \\ \text{Wogbon} \\ \text{Sogbon} \\ \text{Laying Martin} \end{bmatrix} = [86.4, 112.9, 206.2, 110.7, 88.0]'$$

$$SE(\underline{\mu}) = [5.58, 5.58, 5.58, 5.58, 5.58]'$$

- (e) Compute 95% CI for the treatment means:

$$\mu_1: (74.7, 98.0)$$

$$\mu_2: (101.2, 124.5)$$

$$\mu_3: (194.5, 217.8)$$

$$\mu_4: (99.1, 122.3)$$

$$\mu_5: (76.3, 99.6)$$

- (f) Is there significant evidence at the $\alpha = 0.05$ level that the average TB level differs across the 5 stages?

- Yes, we can see from the ANOVA table above that we have significant evidence, $P(F \geq 78.08) \leq 0.05$

2) (c) Produce the ANOVA table for this study.

Source of Variation	df	Sum of Squares	Mean Square	F	P-value
Treatment	4	24121	6030	54.06	4.29×10^{-9}
Error	16	1795	112		
Total	20				

(d) Compute the least squares estimates of the treatment means; their estimated std errors.

$$\underline{\mu} = [89.4, 112.9, 194.0, 110.7, 89.2]'$$

$$SE(\underline{\mu}) = [5.28, 4.72, 6.10, 4.72, 5.28]'$$

(e) Compute 95% confidence intervals for the treatment means.

$$\mu_1: (18.2, 101)$$

$$\mu_2: (102.8, 123)$$

$$\mu_3: (191.1, 207)$$

$$\mu_4: (106.7, 121)$$

$$\mu_5: (18.0, 100)$$

(d) Yes, we can see from the above table that we have significant evidence $P(F > 54.06) \leq 0.05$

- 3.) What is the number of intersections, r , that would be needed for each of the three signaling devices so that an $\alpha = 0.01$ test would have a probability of 90% to detect that the true mean traffic delays were $\mu_1 = 20$, $\mu_2 = 18$, & $\mu_3 = 16$ seconds per vehicle? Use $\hat{\sigma}_e^2 = 12$ in your solution.

• using power. anova. test in R: (see H.O.3 pg. 34)

- means = c(20, 18, 16)
 - power.anova.test (groups = 3, n = , between.var = var(means), within.var = 12, sig.level = 0.01, power = 0.90)
- $\Rightarrow \boxed{r = 28}$

4.)

Refer to problem 1. What is the minimum number of children, r , that the researcher should assign to each of the 5 treatments in order that an $\alpha = 0.05$ test would have probability of at least 90% to detect a difference of at least $30 \text{ ng/dl} \times 10^{-1}$ unit of T3 serum between any pair of treatments? Use $\hat{\sigma}_e^2 = 150$ in your solution.

• using code from H.O.3 pg. 45: (Approach 5)

we get $r = n_i = 7 \Rightarrow \boxed{35 \text{ total children}}$

see discussion of
approaches on H.O.3
pgs 37-41
(see code at pg. 45)

5.) For an experiment w/ four treatments and number of reps given by:

$n_1 = 3, n_2 = 4, n_3 = 5, n_4 = 3$; the following models were proposed:

• Cell means model: $y_{ij} = \mu_i + e_{ij}$

• Effects Model: $y_{ij} = \mu + \tau_i + e_{ij}$

Answer the following questions:

(a) write out the design matrix for the cell means model

$$X = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

X_1 : a column starting w/ 3 (1)'s then 12 (0)'s

X_2 : a column starting w/ 3 (0)'s, 4 (1)'s then 8 (0)'s

X_3 : a column starting w/ 7 (0)'s, then 5 (1)'s followed by 3 (0)'s

X_4 : a column starting w/ 1

(b) write out the design matrix for the effects model w/ No constraints.

$$X = \begin{bmatrix} 1 & 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 & 1 \end{bmatrix}$$

(c) write out the design matrix for the effects model w/ constraints: $\tau_4 = 0$

$$X = \begin{bmatrix} 1 & 1 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 \\ 1 & 0 & 0 & 1 \\ 1 & 0 & 0 & 1 \\ 1 & 0 & 0 & 1 \\ 1 & 0 & 0 & 1 \\ 1 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix}$$

see #0.3
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→ (d) For the effects model w/ no constraints express τ_i in terms of μ, μ_i

$$\mu_i = E[Y_{ij}] = \mu + \tau_i \Leftrightarrow \boxed{\tau_i = \mu_i - \mu \quad i=1,2,3,4}$$

see #0.3
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→ (e) For the effects model w/ the constraint $\tau_4 = 0$, express τ_i in terms of μ, μ_i

$$\tau_i = \mu_i - (\mu + \tau_4) = \boxed{\tau_i = \mu_i - \mu \quad i=1,2,3}$$

6.) The cell means model, $y_{ij} = \mu_i + \epsilon_{ij}$ has as one of its assumptions that the y_{ij} 's are random, independent observations from the treatment populations.

(a) What could the statistician do during the conduct of the experiment in order to ensure that the condition of random, independent observations is reasonably valid.

- To ensure the condition of random independent observations is reasonably valid, I would randomly assign our EVs to the treatments.

see HOS
p. 19

~~(b)~~ What condition is required of the y_{ij} 's in order for the least squares estimates of μ_i to be linear unbiased estimators?

- The ϵ_{ij} 's are iid w/ a normal distribution.

(c) What conditions are required of the y_{ij} 's in order to validly use the F-test in ANOVA hypothesis and to place confidence intervals on the μ_{ij} 's?

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p. 20

→ • $y_{ij} \sim N(\mu_i, \sigma_e^2)$ and are independent.

~~(d)~~ If the condition from part c is not valid. How could you test for differences in the treatment means?

- We could increase our sample size. Then, by the CLT, each μ_{ij} would be approximately normally distributed.

see p. 20 → 7.) (a) Interpret each of the 4 regression parameters

β_0 : the mean response of treatment group A.

β_1 : the difference between the treatment mean of group B and treatment mean of group A

β_2 : "

"C"

"

β_3 : "

"D"

"

see p. 29

→ (b) compute a 95% CI for the mean response in treatment group B?

$$\mu_B = \beta_0 + \beta_1 \Rightarrow \underline{y}' \beta \text{ w/ } \underline{y}' = [1 \ 1 \ 0 \ 0]$$

$$\text{Var}(\underline{y}' \beta) = \sigma_e^2 \underline{y}' (X'X)^{-1} \underline{y} = \sigma_e^2 (0.02) = 19.45^2 (0.02) = 0.389$$

$$\begin{aligned} \text{95\% CI: } \mu_B \pm t_{0.025, 196} \frac{\hat{\sigma}_e}{\sqrt{50}} &= (37.5 - 11.5) \pm t_{0.025, 196} \frac{19.45}{\sqrt{50}} \\ &= (20.57534, 31.42466) \end{aligned}$$

see p. 29

(c) Compute a 95% CI for the mean difference in response between treatment groups B & A, that is the difference $\mu_B - \mu_A$?

$$[\text{Note: } \mu_B - \mu_A = \beta_1]$$

$$\begin{aligned} \text{95\% CI: } \mu_B - \mu_A \pm t_{0.025, 196} \sqrt{\frac{\hat{\sigma}_e^2}{50} + \frac{\hat{\sigma}_e^2}{50}} &= \mu_B - \mu_A \pm t_{0.025, 196} \frac{19.45}{5} \\ &= (-11.5) \pm t_{0.025, 196} (19.45/5) \\ &= (-19.207124, -2.792872) \end{aligned}$$

Note:

$$\mu_B - \mu_A = \beta_0 + \beta_1 - \beta_0 = \beta_1$$