
STAT 461: Homework 4

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PROBLEM 1

a)

This model is estimable where: $b_1 = b_2 = 1; b_3 = -2$

$$\begin{aligned}\sum_{n=1}^3 b_i(\mu + \tau_i) &= 1(\mu + \tau_1) + 1(\mu + \tau_2) - 2(\mu + \tau_3) \\ &= 2\mu - 2\mu + \tau_1 + \tau_2 - 2\tau_3 \\ &= \tau_1 + \tau_2 - 2\tau_3 \\ \sum_{n=1}^3 b_i \bar{Y}_i &= \bar{Y}_1 + \bar{Y}_2 - 2\bar{Y}_3\end{aligned}$$

b)

This model is estimable where: $b_1 = b_2 = 0; b_3 = 1$

$$\begin{aligned}\sum_{n=1}^3 b_i(\mu + \tau_i) &= 0(\mu + \tau_1) + 0(\mu + \tau_2) + 1(\mu + \tau_3) \\ &= \mu + \tau_3 \\ \sum_{n=1}^3 b_i \bar{Y}_i &= 0\bar{Y}_1 + 0\bar{Y}_2 + 1\bar{Y}_3 \\ &= \bar{Y}_3\end{aligned}$$

c)

This model is not estimable for any real values of b_i

d)

This model is estimable where: $b_1 = b_2 = b_3 = \frac{1}{3}$

$$\begin{aligned}\sum_{n=1}^3 b_i(\mu + \tau_i) &= \frac{1}{3}(\mu + \tau_1) + \frac{1}{3}(\mu + \tau_2) + \frac{1}{3}(\mu + \tau_3) \\ &= 3 * \frac{1}{3}\mu + \frac{1}{3}\tau_1 + \frac{1}{3}\tau_2 + \frac{1}{3}\tau_3 \\ &= \mu + \frac{1}{3}(\tau_1 + \tau_2 + \tau_3) \\ \sum_{n=1}^3 b_i \bar{Y}_i &= \frac{1}{3}\bar{Y}_1 + \frac{1}{3}\bar{Y}_2 + \frac{1}{3}\bar{Y}_3\end{aligned}$$

PROBLEM 2

a)

$$Y_{it} = \mu + \tau_i + \epsilon_{it}; \quad i = 1, 2, 3; \quad t = 1, 2, 3, 4$$

$$\epsilon_{it} \stackrel{iid}{\sim} N(0, \sigma^2)$$

1 = Regular 2 = Deodorant 3 = Moisturizing

b)

For future calculations:

$$\bar{Y}_i = \bar{Y} + \hat{\tau}_i$$

$$\hat{\tau}_i = \bar{Y}_i - \bar{Y}$$

$$\bar{Y} = \frac{1}{12}(-.3 - .1 - .14 + .40 + 2.63 + 2.61 + 2.41 + 3.15 + 1.86 + 2.03 + 2.26 + 1.82) = 1.5525\bar{3}$$

$$\bar{Y}_1 = \frac{1}{4}(-.3 - .1 - .14 + .40) = -0.035$$

$$\bar{Y}_2 = \frac{1}{4}(2.63 + 2.61 + 2.41 + 3.15) = 2.7$$

$$\bar{Y}_3 = \frac{1}{4}(1.86 + 2.03 + 2.26 + 1.82) = 1.9925$$

$$\hat{\tau}_1 = \bar{Y}_1 - \bar{Y} = -0.035 - 1.5525\bar{3} \approx -1.5875$$

$$\hat{\tau}_2 = \bar{Y}_2 - \bar{Y} = 2.7 - 1.5525\bar{3} \approx 1.1475$$

$$\hat{\tau}_3 = \bar{Y}_3 - \bar{Y} = 1.9925 - 1.5525\bar{3} \approx 0.44$$

$$\sum_{n=1}^3 b_i \bar{Y}_i = b_1 \bar{Y}_1 + b_2 \bar{Y}_2 + b_3 \bar{Y}_3$$

$$\sum_{n=1}^3 b_i \bar{Y}_i = -0.035b_1 + 2.7b_2 + 1.9925b_3$$

The LSE for a bar of Deodorant Soap is where: $b_1 = b_3 = 0$; $b_2 = 1$

$$\begin{aligned} \sum_{n=1}^3 b_i \bar{Y}_i &= -0.035 * 0 + 2.7 * 1 + 1.9925 * 0 \\ &= 2.7 \end{aligned}$$

c)

This model is estimable where: $b_1 = 1$; $b_2 = b_3 = -\frac{1}{2}$

$$\sum_{n=1}^3 b_i(\mu + \tau_i) = 1(\mu + \tau_1) - \frac{1}{2}(\mu + \tau_2) - \frac{1}{2}(\mu + \tau_3)$$

$$= \mu - 2\left(\frac{1}{2}\mu\right) + \tau_1 - \frac{1}{2}\tau_2 - \frac{1}{2}\tau_3$$

$$= \tau_1 - \frac{1}{2}(\tau_2 + \tau_3)$$

$$\sum_{n=1}^3 b_i \bar{Y}_i = \bar{Y}_1 - \frac{1}{2}\bar{Y}_2 - \frac{1}{2}\bar{Y}_3$$

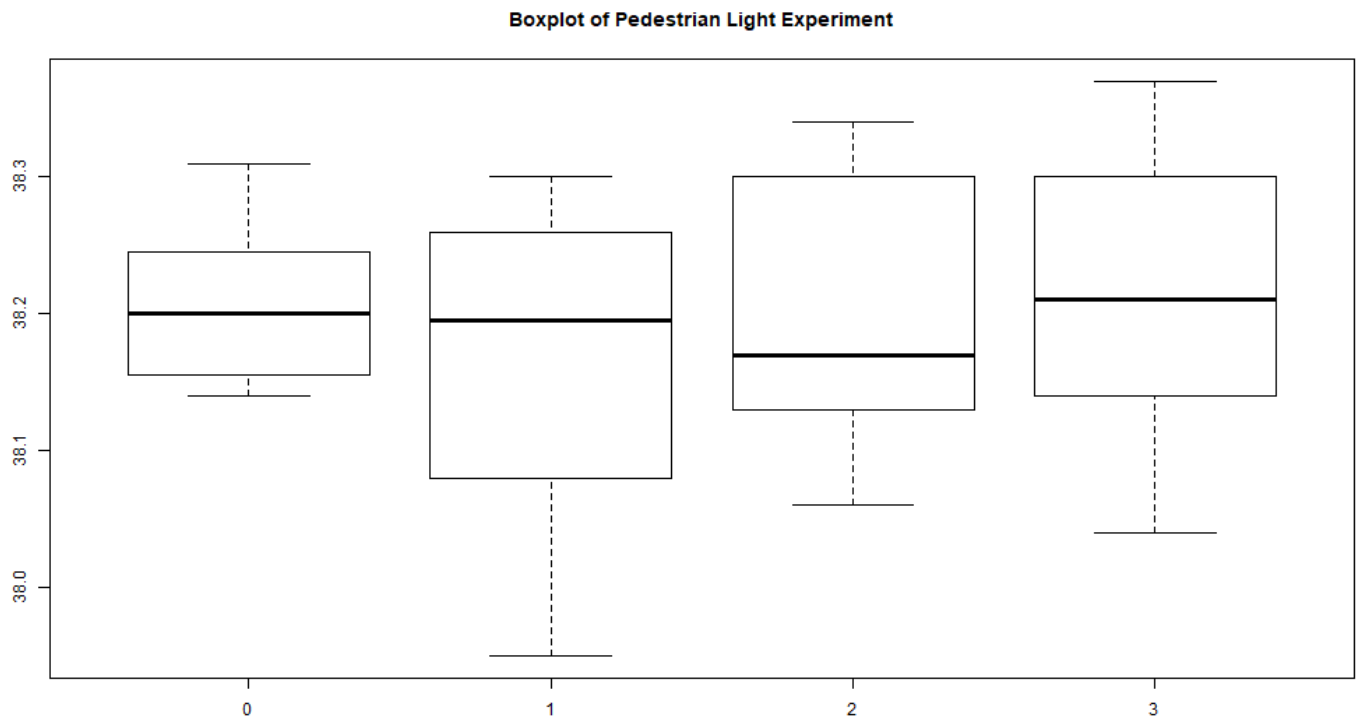
$$= -0.035 - \frac{1}{2}(2.7 + 1.9925) = -2.38125$$

d)

```
1 > #Part D
2 > reg_mean = mean(loss[type=='reg'])
3 > deo_mean = mean(loss[type=='deo'])
4 > moi_mean = mean(loss[type=='moi'])
5 >
6 > #Part B Re-calculation:
7 > deo_mean
8 [1] 2.7
9 >
10 > #Part C Re-calculation:
11 > reg_mean -(deo_mean + moi_mean)/2
12 [1] -2.38125
```

PROBLEM 3

a)



b)

$$Y_{it} = \mu + \tau_i + \epsilon_{it}; \quad i = 0, 1, 2, 3; \quad t = 1, \dots, r_i$$

$$r_1 = 7, r_1 = r_2 = 10, r_3 = 5$$

$$\epsilon_{it} \stackrel{iid}{\sim} N(0, \sigma^2)$$

c)

```
1 > mean_0 = mean(time[presses == '0'])
2 > mean_1 = mean(time[presses == '1'])
3 > mean_2 = mean(time[presses == '2'])
4 > mean_3 = mean(time[presses == '3'])
5 > mean_0
6 [1] 38.20714
7 > mean_1
8 [1] 38.171
9 > mean_2
10 [1] 38.194
11 > mean_3
12 [1] 38.212
```

d)

This model is estimable where: $b_0 = -1; b_1 = 1; b_2 = b_3 = 0$

$$\begin{aligned}\sum_{n=0}^3 b_i(\mu + \tau_i) &= b_0(\mu + \tau_0) + b_1(\mu + \tau_1) + b_2(\mu + \tau_2) + b_3(\mu + \tau_3) \\ &\approx \sum_{n=0}^3 b_i \bar{Y}_i \\ \sum_{n=0}^3 b_i \bar{Y}_i &= -\bar{Y}_0 + \bar{Y}_1 + 0\bar{Y}_2 + 0\bar{Y}_3 \\ &= 38.171 - 38.20714 = -0.03614286\end{aligned}$$

```
1 > # Part D
2 > mean_1 - mean_0
3 [1] -0.03614286
```

e)

This model is estimable where: $b_0 = -1; b_1 = b_2 = b_3 = \frac{1}{3}$

$$\begin{aligned}\sum_{n=0}^3 b_i(\mu + \tau_i) &= b_0(\mu + \tau_0) + b_1(\mu + \tau_1) + b_2(\mu + \tau_2) + b_3(\mu + \tau_3) \\ &\approx \sum_{n=0}^3 b_i \bar{Y}_i \\ \sum_{n=0}^3 b_i \bar{Y}_i &= -\bar{Y}_0 + \frac{1}{3}\bar{Y}_1 + \frac{1}{3}\bar{Y}_2 + \frac{1}{3}\bar{Y}_3 \\ &= \frac{1}{3}(38.171 + 38.194 + 38.212) - 38.20714 = -0.01480952\end{aligned}$$

```
1 > # Part E
2 > (1/3)*(mean_1 + mean_2 + mean_3) - mean_0
3 [1] -0.01480952
```

CODE APPENDIX

```
1 #####
2 #### Setup
3 #####
4 ## Install and load libraries
5 # ipak function taken from: https://gist.github.com/stevenworthington/3178163
6 ipak <- function(pkg) {
7   new.pkg <- pkg[!(pkg %in% installed.packages()[, "Package"])]
8   if (length(new.pkg))
9     install.packages(new.pkg, dependencies = TRUE)
10   sapply(pkg, require, character.only = TRUE)
11 }
12 packages <- c("ggplot2", "reshape2", "gridExtra", "TSA", "astsa", "orcutt",
13              "nlme", "fGarch", "vars", "lsmeans")
14 ipak(packages)
15
16 #####
17 #### Problem 2
18 #####
19 # From HW1
20 loss = c(-.3, -.1, -.14, .4, 2.63, 2.61, 2.41, 3.15, 1.86, 2.03, 2.26, 1.82)
21 type = c(rep('reg', 4), rep('deo', 4), rep('moi', 4))
22 losses = data.frame(loss, type)
23
24 # Part D
25 mean_reg = mean(loss[type=='reg'])
26 mean_deo = mean(loss[type=='deo'])
27 mean_moi = mean(loss[type=='moi'])
28 ybar = mean(loss)
29 tau1 = mean_reg - ybar
30 tau2 = mean_deo - ybar
31 tau3 = mean_moi - ybar
32
33 # Part B Re-calculation:
34 mean_deo
35
36 # Part C Re-calculation:
37 mean_reg -(mean_deo + mean_moi)/2
38
39 #####
40 #### Problem 3
41 #####
42 time = c(38.14, 38.20, 38.31, 38.14, 38.29, 38.17, 38.20,
43          38.28, 38.17, 38.08, 38.25, 38.18, 38.03, 37.95, 38.26, 38.30, 38.21,
44          38.17, 38.13, 38.16, 38.30, 38.34, 38.34, 38.17, 38.18, 38.09, 38.06,
45          38.14, 38.30, 38.21, 38.04, 38.37)
46 presses = c(rep('0', 7), rep('1', 10), rep('2', 10), rep('3', 5))
47 lights = data.frame(time, presses)
48
49 # Part A
50 png("./figures/p3.png", width = 1024, height = 576)
51 boxplot(time ~ presses, main="Boxplot of Pedestrian Light Experiment")
52 dev.off()
53
54 # Part C
55 mean_0 = mean(time[presses == '0'])
56 mean_1 = mean(time[presses == '1'])
57 mean_2 = mean(time[presses == '2'])
58 mean_3 = mean(time[presses == '3'])
59
60 # Part D
61 mean_1 - mean_0
62
63 # Part E
64 (1/3)*(mean_1 + mean_2 + mean_3) - mean_0
```