

a)

$$Y_{it} = \mu + \tau_i + \epsilon_{it}, \quad \epsilon_{it} \sim N(0, \sigma^2)$$
$$i = A, B, C$$
$$t = 1, \dots, r_i; \quad r_A = r_B = r_C = 2$$

b)

$$\overline{Y}_{B}$$
. $= \frac{5-1}{2} = 4$
 \overline{Y}_{B} . $\sim N\left(4, \frac{\sigma^2}{2}\right)$

c)

$$\begin{split} \overline{Y}_{A} &= \frac{-14-4}{2} = -9 \\ \overline{Y}_{B} &= \frac{5-1}{2} = 2 \\ \overline{Y}_{C} &= \frac{-2+6}{2} = 2 \\ \overline{Y}_{..} &= \frac{-14-4+5-1-2+6}{6} = \frac{-5}{3} \\ \text{SSE} &= \sum_{i=1}^{\nu} \sum_{t=1}^{r_{i}} \left(\overline{Y}_{it} - \overline{Y}_{i.} \right)^{2} \\ &= (-14+9)^{2} + (-4+9)^{2} + (5-2)^{2} + (-1-2)^{2} + (-2-2)^{2} + (6-2)^{2} \\ &= 100 \\ \text{SST} &= \sum_{i=1}^{\nu} r_{i} \left(\overline{Y}_{i.} - \overline{Y}_{..} \right)^{2} \\ &= 2 \left(-9 - \frac{-5}{3} \right)^{2} + 2 \left(2 - \frac{-5}{3} \right)^{2} + 2 \left(2 - \frac{-5}{3} \right)^{2} \\ &= \frac{484}{3} \approx 161.\overline{333} \\ \text{SSTOT} &= \sum_{i=1}^{\nu} \sum_{t=1}^{r_{i}} \left(\overline{Y}_{it} - \overline{Y}_{i.} \right)^{2} \\ &= \left(-14 - \frac{-5}{3} \right)^{2} + \left(-4 - \frac{-5}{3} \right)^{2} + \left(5 - \frac{-5}{3} \right)^{2} + \left(-1 - \frac{-5}{3} \right)^{2} + \left(-2 - \frac{-5}{3} \right)^{2} + \left(6 - \frac{-5}{3} \right)^{2} \\ &= \frac{784}{3} \approx 261.\overline{333} \end{split}$$

d)

$$\widehat{\sigma^2} \approx \frac{\text{SSE}}{n-\nu} = \frac{756}{6-3} = 252$$

e)

$$\widehat{\Delta}_{AC} = \overline{Y}_{A\cdot} - \overline{Y}_{C\cdot} = -9 - 2 = -11$$

f)

$$\overline{Y}_{A}$$
. $-\overline{Y}_{C}$. $\sim N(\mu + \tau_A, \sigma^2) + N(-\mu - \tau_C, \sigma^2) = N(0, 2\sigma^2)$

g)

$$\frac{\overline{Y}_{A \cdot} - \overline{Y}_{C \cdot}}{K(\sigma)} \sim N(0, 2 \left(\frac{1}{K(\sigma)}\right)^2 \sigma^2)$$

$$K(\sigma) = \sqrt{2}\sigma$$

h)

$$\frac{\rm SSE}{\sigma^2} \sim \chi^2_{n-\nu}$$

i)

$$\frac{\left(\overline{Y}_{A}.-\overline{Y}_{C}.\right)^{2}/\left(K(\sigma)\right)^{2}}{\mathsf{SSE}/\left[\left(n-\nu\right)\sigma^{2}\right]}\sim F_{1,(n-\nu)}$$

j)

Under the null hypothesis of H_0 : $\tau_A = \tau_B = \tau_C$, we use the test statistic $T^* = \frac{\text{SST}/(v-1)}{\text{SSE}/(n-v)}$ where $T^* \sim F_{(v-1),(n-v)}$.

In general:

	DF	Sum Sq	Mean Sq	F-Value
Treatment	v-1	SST	SST/(v-1)	$\frac{SST/(\nu-1)}{SSE/(n-\nu)}$
Error	n-v	SSE	SSE/(n-v)	NA
Total	n-1	SSTOT	NA	NA

For our case:

	DF	Sum Sq	Mean Sq	F-Value
Treatment	2	<u>484</u> 3	<u>242</u> 3	121 50
Error	3	100	100 3	
Total	5	$\frac{784}{3}$	NA	NA

k)

Based on the below output from R, we can conclude that there is no significant difference in the response of the three populations. In this case, H_0 should not be rejected.

```
1 Analysis of Variance Table
2
3 Response: delta
4 Df Sum Sq Mean Sq F value Pr(>F)
5 meds 2 161.33 80.667 2.42 0.2367
```

I)

Examining the pairwise comparisons for the variables, we can see that none of the p-values are significant for any of the contrasts. This confirms that there is no significant difference between the treatments of the 3 populations.

```
H_{0}: \tau_{A} = \tau_{B} \quad H_{A}: \tau_{A} \neq \tau_{B}
H_{0}: \tau_{A} = \tau_{C} \quad H_{A}: \tau_{A} \neq \tau_{C}
H_{0}: \tau_{B} = \tau_{C} \quad H_{A}: \tau_{B} \neq \tau_{C}
T^{*} = \frac{\sqrt{\frac{r_{i} + r_{j}}{r_{i}r_{j}}} \left(\overline{Y}_{i} - \overline{Y}_{j}\right)}{\sqrt{SSE/(n - v)}}
T^{*} \sim t_{n - v} = t_{3}
```

```
1 contrast estimate SE df t.ratio p.value

2 A - B -1.100000e+01 5.773503 3 -1.905 0.2817

3 A - Control -1.100000e+01 5.773503 3 -1.905 0.2817

4 B - Control -1.776357e-15 5.773503 3 0.000 1.0000
```

PROBLEM 2

```
\begin{split} H_0: &\tau_{reg} = \tau_{deo} & H_A: \tau_{reg} \neq \tau_{deo} \\ H_0: &\tau_{reg} = \tau_{moi} & H_A: \tau_{reg} \neq \tau_{moi} \\ H_0: &\tau_{deo} = \tau_{moi} & H_A: \tau_{deo} \neq \tau_{moi} \\ \\ &T^* = \frac{\sqrt{\frac{r_i + r_j}{r_i r_j} \left(\overline{Y}_i - \overline{Y}_j\right)}}{\sqrt{SSE/(n-\nu)}} \\ &T^* \sim t_{n-\nu} = t_9 \end{split}
```

Examining the ANOVA table, one can see that there is a significant difference between at least one pair of treatment. Therefore, we reject H_0 and must examine further to determine which treatments are significant.

Looking at the pairwise comparisons, we can observe that the deodorant soap lost less than the regular soap, and that the moisturizing soap also lost less than the regular soap. However, there was no significant difference between the moisturizing and deodorant soaps.

```
Analysis of Variance Table

Response: loss

Df Sum Sq Mean Sq F value Pr(>F)

type 2 16.1220 8.0610 104.45 5.914e-07 ***

Residuals 9 0.6946 0.0772

---

Signif. codes: 0 $***$ 0.001 $**$ 0.01 $*$ 0.05 $.$ 0.1 $$ 1

contrast estimate SE df t.ratio p.value

deo - moi 0.7075 0.196437 9 3.602 0.0143

deo - reg 2.7350 0.196437 9 13.923 <.0001

moi - reg 2.0275 0.196437 9 10.321 <.0001
```

$$H_{0}:\tau_{0} = \tau_{1} \quad H_{A}:\tau_{0} \neq \tau_{1}$$

$$H_{0}:\tau_{0} = \tau_{2} \quad H_{A}:\tau_{0} \neq \tau_{2}$$

$$H_{0}:\tau_{0} = \tau_{3} \quad H_{A}:\tau_{0} \neq \tau_{3}$$

$$H_{0}:\tau_{1} = \tau_{2} \quad H_{A}:\tau_{1} \neq \tau_{2}$$

$$H_{0}:\tau_{1} = \tau_{3} \quad H_{A}:\tau_{1} \neq \tau_{3}$$

$$H_{0}:\tau_{2} = \tau_{3} \quad H_{A}:\tau_{2} \neq \tau_{3}$$

$$T^{*} = \frac{\sqrt{\frac{r_{i} + r_{j}}{r_{i}r_{j}}} \left(\overline{Y}_{i} - \overline{Y}_{j}\right)}{\sqrt{SSE/(n - v)}}$$

$$T^{*} \sim t_{n - v} = t_{28}$$

The p-value obtained from the ANOVA is extremely large, indicating that H_0 should not be rejected. Continuing to look at the pairwise comparisons, the p-values for each contrast are also very large, indicating that there is no difference between the treatments.

```
1 Analysis of Variance Table
2
3 Response: time
4 Df Sum Sq Mean Sq F value Pr(>F)
5 presses 3 0.008047 0.0026824 0.2455 0.8638
6 Residuals 28 0.305953 0.0109269
8
9
10
                                 SE df t.ratio p.value
   contrast
                estimate
11
   0 - 1
             0.036142857 \ 0.05151381 \ 28 \ 0.702 \ 0.8956
12 0 - 2
             0.013142857 0.05151381 28
                                         0.255
                                                0.9940
13 0 - 3
            -0.004857143 \ 0.06120753 \ 28 \ -0.079
                                                0.9998
14 1 - 2
            -0.023000000 \ 0.04674802 \ 28 \ -0.492
                                                0.9602
15 1 - 3
            -0.041000000 \ 0.05725440 \ 28 \ -0.716
                                                0.8899
16 2 - 3
            -0.018000000 \ 0.05725440 \ 28 \ -0.314 \ 0.9890
```

CODE APPENDIX

```
2 #### Setup
4 ## Install and load libraries
5 # ipak function taken from: https://gist.github.com/stevenworthington/3178163
6 ipak <- function(pkg) {
   new.pkg <- pkg[!(pkg %in% installed.packages()[, "Package"])]</pre>
   if (length (new.pkg))
9
     install.packages(new.pkg, dependencies = TRUE)
sapply(pkg, require, character.only = TRUE)
11 }
12 packages <- c("ggplot2", "reshape2", "gridExtra", "TSA", "astsa", "orcutt",
             "nlme", "fGarch", "vars", "lsmeans")
13
14 ipak (packages)
15
19 meds = c(rep('A',2),rep('B',2),rep('Control',2))
20 delta = c(-14, -4, 5, -1, -2, 6)
21 bpData = data.frame(delta, meds)
22
23 # Calculate ANOVA table and Pairwise Comparison
24 bpModel = aov(delta ~ meds, data=bpData)
25 anova (bpModel)
26 bpLSM = lsmeans(bpModel, ~ meds)
27 contrast (bpLSM, method='pairwise')
28
30 #### Problem 2
32 # From HW1
33 loss = c(-.3, -.1, -.14, .4, 2.63, 2.61, 2.41, 3.15, 1.86, 2.03, 2.26, 1.82)
34 type = c(rep('reg',4), rep('deo',4), rep('moi',4))
35 losses = data.frame(loss, type)
37 # Calculate ANOVA table and Pairwise Comparison
38 lossModel = aov(loss~type, data=losses)
39 anova (lossModel)
40 lossLSM = lsmeans(lossModel, ~type)
41 contrast (lossLSM, method='pairwise')
42
43
45 #### Problem 3
47 # From HW4
48 time = c(38.14, 38.20, 38.31, 38.14, 38.29, 38.17, 38.20,
         38.28, 38.17, 38.08, 38.25, 38.18, 38.03, 37.95, 38.26, 38.30, 38.21,
50
         38.17, 38.13, 38.16, 38.30, 38.34, 38.34, 38.17, 38.18, 38.09, 38.06,
         38.14, 38.30, 38.21, 38.04, 38.37)
52 presses = c(rep('0',7), rep('1',10), rep('2',10), rep('3',5))
53 lights = data.frame(time, presses)
55 # Calculate ANOVA table and Pairwise Comparison
56 lightModel = aov(time~presses, data=lights)
57 anova(lightModel)
58 lightLSM = lsmeans(lightModel, ~presses)
59 contrast (lightLSM, method='pairwise')
```