## Solution: Homework 1

19.5. The table of means is given below. You will also find the main effects of factors A and B and the interaction effects (in the brackets) for Problems 19.7 and 19.8.

> Table of means Factor B Factor A  $B_1$  $B_2$  $B_3$  $B_4$  $\mu_i$  $\alpha_i$  $A_1$ 250 265 268 269 263 -6275 $A_2$ 288 273 270 269 6  $\mu_{..} = 269$ 269 269 269 269  $\mu_{\cdot j}$  $\beta_i$ 0 0 0 0

| Table of interactions |          |       |       |       |
|-----------------------|----------|-------|-------|-------|
|                       | Factor B |       |       |       |
| Factor A              | $B_1$    | $B_2$ | $B_3$ | $B_4$ |
| $A_1$                 | -13      | 2     | 5     | 6     |
| $A_2$                 | 13       | -2    | -5    | -6    |

- (a) Factor B main effects are all zero: no main effect of factor B.
- (b) Interactions effects are present since the curves are not parallel.
- (c) Table of  $\{\ln \mu_{ii}\}$

|          | Factor B |        |        |        |  |
|----------|----------|--------|--------|--------|--|
| Factor A | $B_1$    | $B_2$  | $B_3$  | $B_4$  |  |
| $A_1$    | 5.5215   | 5.5797 | 5.5910 | 5.5947 |  |
| $A_2$    | 5.6630   | 5.6095 | 5.5948 | 5.5947 |  |

Once again, the graph of  $\{\ln \mu_{ij}\}$  indicate presence of interactions. 19.8. (a)

$$E(MSE) = \sigma^2 = 4^2 = 16$$
  
 $E(MSAB) = \sigma^2 + n \sum_{ij} \sum_{j=1}^{n} (\alpha \beta)_{ij}^2 / [(a-1)(b-1)] = 952$ 

(b) The ratio E(MSAB)/E(MSE) = 59.5 is quite large, i.e., E(MSAB) is much larger than E(MSE). Since MSAB/MSE fluctuates about E(MSAB)/E(MSE), any F-test for a data set from the populations with the means and  $\sigma^2$  as given here will very likely result in rejecting  $H_0: (\alpha \beta)_{ij} = 0$  for all i, j (against  $H_1:$ not all  $(\alpha\beta)_{ij}$  are equal to zero) since the p-value of the F-test is likely to be very small.

(a) The fitted values are  $\hat{Y}'_{ijk} = \bar{Y}'_{ij}$ . The values of the estimated means are  $\bar{Y}'_{11.} = 0.44348, \ \bar{Y}'_{12.} = 0.80997, \ \bar{Y}'_{13.} = 1.10670, \ \bar{Y}'_{21.} = 0.39823, \ \bar{Y}'_{22.} = 0.58096, \ \bar{Y}'_{23.} = 0.86639.$  The residuals  $\{e'_{ijk}\}$  are

|       | j = 1               | j = 2              | j = 3               |
|-------|---------------------|--------------------|---------------------|
|       | -0.4435 0.0336      | -0.3329 -0.1110    | $0.0974\ 0.1238$    |
|       | 0.0336 $-0.4435$    | -0.1110 -0.2079    | -0.0653 -0.2036     |
| i = 1 | $-0.1425 \ 0.3347$  | 0.0931 $-0.5089$   | $-0.1525 \ 0.3847$  |
|       | $0.1586 \ 0.4016$   | 0.3040 $-0.0318$   | -0.3286 -0.5046     |
|       | $-0.4435 \ 0.5108$  | $0.3942 \ 0.5123$  | $0.3083 \ \ 0.3405$ |
|       | -0.3982 0.0790      | 0.1972 - 0.2799    | $0.1750 \ 0.3377$   |
|       | $-0.0972 \ 0.5049$  | $0.0211 \ 0.0211$  | 0.0879 - 0.1674     |
| i=2   | -0.0972  0.3007     | $-0.1038 \ 0.2641$ | $0.2476 \ 0.1336$   |
|       | -0.3982 -0.3982     | $-0.5810 \ 0.3221$ | -0.2643 -0.0213     |
|       | $0.3007 \ \ 0.2038$ | -0.2799 0.4190     | 0.0367 $-0.5654$    |

- (b) The dotplot show that the residuals are roughly equally aligned with reasonable symmetry about 0. Thus it seems that the assumption of equal variance is reasonable here and the distribution of the error terms may be symmetric about zero.
- (c) The normal probability plot does not indicate any obvious departure from normality. Corr=0.987.

19.19

(a) Estimated mean plot shows almost parallel lines indicating that interactions effects may be negligible. Factor B means seem to be quite different indicating the presence of factor B main effects. However, factor A means do not seem to be very different and thus it is unclear if factor A main effects are present.

(b) ANOVA table

| Source      | df             | SS      | MS      |
|-------------|----------------|---------|---------|
| Duration    | a - 1 = 1      | 0.44129 | 0.44129 |
| Weight Gain | b - 1 = 2      | 3.20098 | 1.60049 |
| Interaction | (a-1)(b-1) = 2 | 0.11989 | 0.05995 |
| Error       | (n-1)ab = 54   | 5.46770 | 0.10125 |
| Totral      | nab - 1 = 59   | 9.22987 |         |

Most of the variability in the response seem to be explained by the main effect of factor B (weight gain)

- (c)  $H_0: (\alpha\beta)_{ij} = 0$  for all i and j,  $H_1:$  not all  $(\alpha\beta)_{ij}$  equal zero. Decision rule; reject  $H_0$  if  $F^* = MSAB/MSE > F(0.95; 2, 54) = 3.17$ . Since  $F^* = MSAB/MSE = 0.59 < F(0.95; 2, 54)$ , we cannot reject  $H_0$ .
- (d)  $H_0: \alpha_i = 0$  for all i,  $H_1:$  not all  $\alpha_i$  equal 0. Decision rule: reject  $H_0$  if  $F^* = MSA/MSE > F(0.95; 1, 54) = 4.02$ . Since  $F^* = MSA/MSE = 4.36$ , we reject  $H_0$ .  $H_0: \beta_j = 0$  for all j,  $H_1:$  not all  $\beta_j$  equal 0.

Decision rule: reject  $H_0$  if  $F^* = MSB/MSE > F(0.95; 2, 54) = 3.17$ 

Since  $F^* = MSB/MSE = 15.81$ , we reject  $H_0$ .

(e) The overall level of significance, using Kimball inequality, has a bound

$$\alpha \le 1 - (0.95)(0.95)(0.95) = 0.143.$$

(f) The results from parts (c) and (d) confirm the preliminary assessments in part (a)

(a)  $s(\bar{Y}'_{22}.)=0.1006,\ t(0.975;54)=2.005.A$  95% confidence interval for  $\mu_{22}$  is  $\bar{Y}'_{22}.\pm t(0.975;54)\ s(\bar{Y}'_{22}.),$  i.e., (0.3793,0.7827).

We have a 95% confidence that 
$$\mu_{22}$$
 is in the interval (0.3793, 0.7827). (b)  $\hat{D} = \bar{Y}'_{23}$ .  $-\bar{Y}'_{21}$ . = 0.4682,  $s(\hat{D}) = \sqrt{(2/n)MSE} = 0.1423$ ,  $t(0.975; 54) = 2.005$ .

A 95% confidence interval for *D* is  $\hat{D} \pm t(0.975; 54)s(\hat{D})$ , i.e., (0.1829, 7535). Since this interval does not include 0, we can conclude that  $\mu_{23}$  and  $\mu_{21}$  are different at a 0.05 level of significance.

(c) We have 
$$\bar{Y}'_{1..} = 0.7867$$
,  $\bar{Y}'_{2..} = 0.6152$ ;  $\bar{Y}'_{1.1} = 0.4209$ ,  $\bar{Y}'_{2..} = 0.6955$ ,  $\bar{Y}'_{3..} = 0.9866$ 

The plots are not given here. But the values given above suggest that factor B main effects are present, whereas it is unclear if factor A main effects are

(d) Total number of intervals to be constructed is 4 - one for pairwise comparisons for factor A and 3 for pairwise comparisons for factor B. Since  $\alpha = 0.1$ , the multipliers are

Bonferroni: 
$$B = t(1 - \alpha/8; 54) = t(0.9875; 54) = 2.306,$$
  
Tukey: factor A,  $q(1 - \alpha/2; 2, 54) = q(0.95; 2, 54) = 2.84, T = 2.008$   
Tukey: factor B,  $q(1 - \alpha/2; 3, 54) = q(0.95; 3, 54) = 3.41, T = 2.411,$   
Scheffe :  $S = \sqrt{(a + b - 2)F(1 - \alpha; a + b - 2; 54)} = \sqrt{3F(0.95; 3, 54)} = 2.569.$ 

Bonferroni seems to be the preferred method. Even though Tukey method seems to be the best for factor A comparisons, but it suffers from having a higher multiplier than Bonferroni for factor B comparisons.

(e) Let

$$D_1 = \mu_{1} - \mu_{2}$$
,  $D_2 = \mu_{1} - \mu_{2}$ ,  $D_3 = \mu_{1} - \mu_{3}$ ,  $D_4 = \mu_{2} - \mu_{3}$ ,

Note that  $s(\hat{D}_1) = 0.0822$  and  $s(\hat{D}_i) = 0.1006$ , i = 2, 3, 4. The Bonferroni multiplier is B-2.306. Simultaneous 90% confidence intervals (Bonferroni) are

$$\begin{array}{lll} D_1 & : & 0.1715 \pm (2.306)(0.0822), \text{ i.e., } (-0.018, 0.361), \\ D_2 & : & -0.2746 \pm (2.306)(0.1006), \text{ i.e., } (-0.507, -0.043), \\ D_3 & : & -0.5657 \pm (2.306)(0.1006), \text{ i.e., } (-0.798, -0.334), \\ D_4 & : & -0.2911 \pm (2.306)(0.1066), \text{ i.e., } (-0.523, -0.059). \end{array}$$

Since the first interval contains 0, presence of factor A main effect is in doubt. The last three intervals all exclude 0 and hence the evidence points to all the three factor B means are different. These are consistent with the preliminary evidence in part (c).

(f) Here

$$L = 0.3\mu_{.1} + 0.4\mu_{.2} + 0.3\mu_{.3}, \hat{L} = 0.0415, t(0.975; 54) = 2.005...$$

95% confidence interval for L is  $\hat{L} + t(0.975; 54)s(\hat{L})$ , i.e., (0.6172, 0.7836). In original units this interval is (3.142, 5.076). Yes, the evidence points that the mean number of days is less than 7.

19.41. We would like to have  $Bs(\hat{D}_i) \leq 0.2, i = 1, ..., 4$ . Since a = 2 and b = 3, we have  $s(\hat{D}_1) \leq s(\hat{D}_i), i = 2, 3, 4$ . Thus we need to have  $Bs(\hat{D}_i) \leq 0.2$ , i.e.,  $B^2s^2(\hat{D}_i) \leq (0.2)^2, i = 2, 3, 4$ , i.e., (taking  $s^2(\hat{D}_i) \approx 2\sigma^2/(na)$ )

$$B^2 \frac{2\sigma^2}{na} \le (0.2)^2$$
, i.e.,  $B^2 \frac{(2)\sigma^2}{n(2)} \le (0.2)^2$ ,  
i.e.,  $n \ge (B\sigma/0.2)^2 = 13.61$ .

So we we should have  $n \geq 14$ .

19.47

Note that

$$\bar{Y}_{ij} - \bar{Y}_{...} = \hat{\alpha}_i + \hat{\beta}_j + (\widehat{\alpha\beta})_{ij}.$$

Now

$$SSTR = n \sum \sum (\bar{Y}_{ij.} - \bar{Y}_{...})^2,$$

and hence

$$\sum \sum (\bar{Y}_{ij} - \bar{Y}_{...})^{2} = \sum \sum \left[ \hat{\alpha}_{i} + \hat{\beta}_{j} + (\widehat{\alpha \beta})_{ij} \right]^{2}$$

$$= \sum \sum \hat{\alpha}_{i}^{2} + \sum \sum \hat{\beta}_{j}^{2} + \sum \sum (\widehat{\alpha \beta})_{ij}^{2}$$

$$+2 \sum \sum \hat{\alpha}_{i}\hat{\beta}_{j} + 2 \sum \sum \hat{\alpha}_{i}(\widehat{\alpha \beta})_{ij} + 2 \sum \sum \hat{\beta}_{j}(\widehat{\alpha \beta})_{ij}.$$

Note that each of three cross product terms are zero and thus we have

$$\sum \sum (\bar{Y}_{ij}. - \bar{Y}...)^2 = \sum \sum \hat{\alpha}_i^2 + \sum \sum \hat{\beta}_j^2 + \sum \sum (\widehat{\alpha\beta})_{ij}^2$$
$$= b \sum \hat{\alpha}_i^2 + a \sum \hat{\beta}_j^2 + \sum \sum (\widehat{\alpha\beta})_{ij}^2.$$

The result now follows since

$$SSTR = n \sum_{i} \sum_{j} (\bar{Y}_{ij.} - \bar{Y}_{...})^{2}$$

$$= nb \sum_{i} \hat{\alpha}_{i}^{2} + na \sum_{j} \hat{\beta}_{j}^{2} + n \sum_{j} \sum_{i} (\widehat{\alpha \beta})_{ij}^{2}$$

$$= S \cdot SA + SSB + SSAB.$$

19.48. Note that

$$\begin{split} \bar{Y}_{\cdot j \cdot} &= \mu_{\cdot \cdot} + \beta_j + \bar{\varepsilon}_{\cdot j \cdot} = \mu_{\cdot j} + \bar{\varepsilon}_{\cdot j \cdot \cdot}, \\ \hat{L} &= \sum_{\cdot} c_j \bar{Y}_{\cdot j \cdot} = \sum_{\cdot} c_j \mu_{\cdot j} + \sum_{\cdot} c_j \bar{\varepsilon}_{\cdot j \cdot} = L + \sum_{\cdot} c_j \bar{\varepsilon}_{\cdot j \cdot}. \end{split}$$

Since  $\{\bar{\varepsilon}_{.j.}\}$  are iid  $N(0, \sigma^2/(na))$ , we have

$$E(\hat{L}) = L$$
, and 
$$Var(\hat{L}) = \sum c_j^2 Var(\bar{\epsilon}_{\cdot j \cdot}) = \sum c_j^2 \sigma^2/(na) = \frac{\sigma^2}{na} \sum c_j^2.$$









