

HW6

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Problem 1)

```
data <- read.table('pearls.txt', header=T, colClasses = c('numeric', 'factor','factor','factor'))

mu <- mean(data$value)
mu_i <- tapply(data$value, data$no_coats, mean)
mu_j <- tapply(data$value, data$batch, mean)

tau_b <- mu - mu_j
tau_a <- mu - mu_i
```

(a)

$$Y_{ijk} = \mu_{..} + \alpha_i + \beta_j + (\alpha\beta)_{ij} + \epsilon_{ijk}$$

the terms $\alpha_i, (\alpha\beta)_{ij}, \epsilon_{ijk}$ are random and $\mu_{..}, \beta_j$ are fixed.

(b)

| Mean sq | df | Expected MS |
|---------|----------------------|---|
| MSA | $a - 1 = 2$ | $\sigma^2 + n\sigma_{\alpha\beta}^2 + \frac{bn}{a-1} \sum_j \alpha_i^2$ |
| MSB | $b - 1 = 3$ | $\sigma^2 + an\sigma_{\beta}^2$ |
| MSAB | $(a - 1)(b - 1) = 6$ | $\sigma^2 + n\sigma_{\alpha\beta}^2$ |
| MSE | $ab(n - 1) = 36$ | σ^2 |

$$\sigma^2 = 4.832$$

$$\sigma_{\alpha\beta}^2 = -1.1285$$

$$\sigma_{\beta}^2 = 3.844$$

$$E\{MSAB\} = 4.823$$

$$E\{MSB\} = 50.951$$

$$E\{MSA\} = 81.2488333$$

(c)

```
fit2 <- lm(value ~ no_coats + batch + no_coats:batch, data)
anova(fit2)
```

```
## Analysis of Variance Table
```

```
##
## Response: value
##           Df Sum Sq Mean Sq F value    Pr(>F)
## no_coats    2 150.388   75.194   15.591 1.327e-05 ***
## batch       3 152.852   50.951   10.564 3.984e-05 ***
## no_coats:batch 6   1.852    0.309    0.064  0.9988
## Residuals   36 173.625    4.823
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

$$H_0 : \sigma_{\alpha\beta}^2 = 0$$

$$H_a : \sigma_{\alpha\beta}^2 > 0$$

$$F^* = \frac{MSAB}{MSE} = 0.0627666 < F = 2.363751$$

concluding null, interaction term is not significant.

d)

$$H_0 : \sigma_{\alpha}^2 = 0$$

$$H_a : \sigma_{\alpha}^2 > 0$$

$$F^* = \frac{75.194}{0.309} = 243.3462783$$

$$F = 5.1432528$$

$F^* > F$ concluding H_a the effect of coat factor is significant.

e)

| Mean sq | df | Expected MS |
|---------|----------------------|--|
| MSA | $a - 1 = 2$ | $\sigma^2 + bn\sigma_{\alpha}^2$ |
| MSB | $b - 1 = 3$ | $\sigma^2 + n\sigma_{\alpha\beta}^2 + \frac{an}{b-1} \sum_j \beta_j^2$ |
| MSAB | $(a - 1)(b - 1) = 6$ | $\sigma^2 + n\sigma_{\alpha\beta}^2$ |
| MSE | $ab(n - 1) = 36$ | σ^2 |

$$\sigma^2 = 4.832$$

$$\sigma_{\alpha\beta}^2 = -1.1285$$

$$\sigma_{\alpha} = 4.3981875$$

$$E\{MSAB\} = 4.823$$

$$E\{MSA\} = 75.194$$

$$E\{MSB\} = 72.7570741$$

f)

```
mu = tapply(data$value, data$no_coat, mean)
l1 = mu[2] - mu[1]
l2 = mu[3] - mu[1]
```

```
s = sqrt((0.309/(4*4))*(2))
t = qt(1- 0.1/(2*2), 3*4*3)
```

$$B = 2.028094$$

$$s\{\hat{D}\} = 0.1965324$$

$$\mu_6 - \mu_8 = 3.6875$$

$$\mu_6 - \mu_{10} = 3.81875$$

$$3.2889137 \leq \mu_8 - \mu_6 \leq 4.0860863$$

$$3.4201637 \leq \mu_8 - \mu_{10} \leq 4.2173363$$

Problem 2)

a)

$$Y_{ijk} = \mu_{..} + \alpha_i + \beta_{j(i)} + \epsilon_{ijk}$$

$\mu_{..}$ constant variable

α_i : constant related to factor A and subject to $\sum \alpha_i = 0$

$\beta_{j(i)}$: random normal variable with mean 0

ϵ_{ijk} independent $N(0, \sigma^2)$

$$i = 1, \dots, 2; j = 1, \dots, 3; k = 1, \dots, 5$$

b)

$$SSA = 0.01825$$

$$SSB(A) = 0.01153 + 0.44249 = 0.45402$$

$$SSE = 0.29020$$

| Source | SS | df | MS | $E\{MS\}$ |
|----------------|---------|------------|-----------|---|
| Factor A | 0.01825 | a-1=1 | 0.01825 | $\sigma^2 + bn \frac{\sum \alpha_i^2}{a-1} + n\sigma_\beta^2$ |
| Factor B(in A) | 0.45402 | a(b-1)=4 | 0.113505 | $\sigma^2 + n\sigma_\beta^2$ |
| Error | 0.29020 | ab(n-1)=24 | 0.0120917 | σ^2 |
| Total | 0.47227 | abn-1=29 | 0.0162852 | |

assuming $\alpha = 0.05$

for factor A:

$$F^* = MSA/MSB(A) = 0.1607859$$

$$F = 7.7086474$$

$$F^* < F$$

concluding H_0 , there's no considerable variability in different sites.

for factor B:

$$F^* = MSB(A)/MSE = 9.3870434$$

$$F = 2.7762893$$

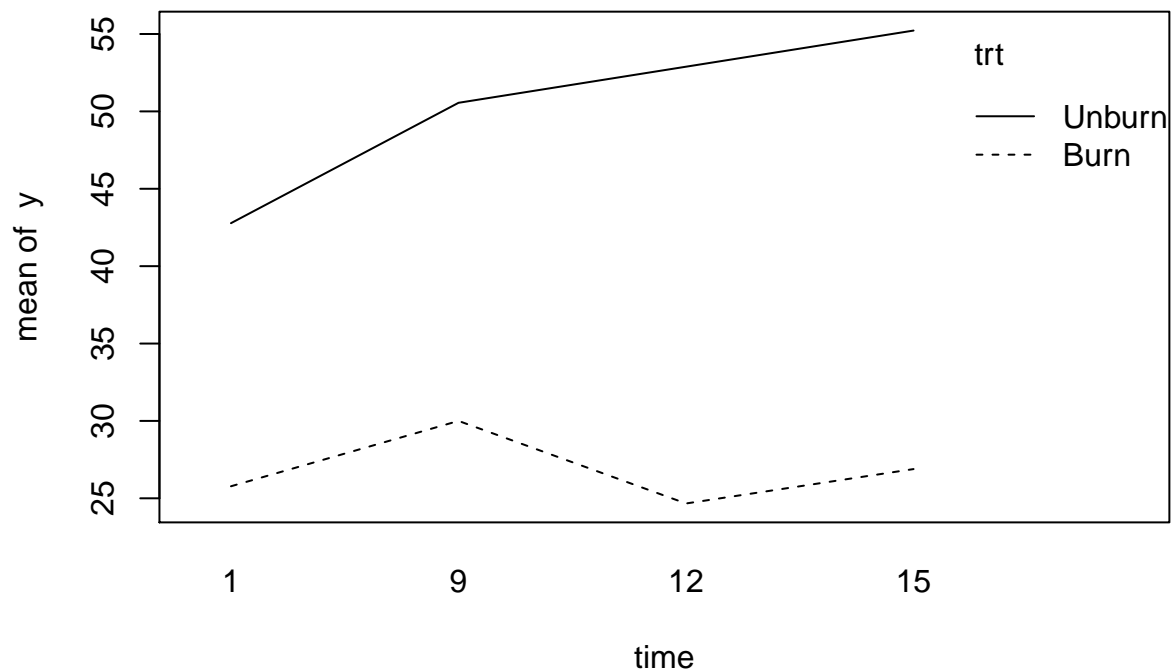
$F^* > F$ concluding H_a , there is considerable batch to batch difference.

Problem 3)

```
data = read.table('floral.dat',header=T,sep='\t')
trt = as.factor(data$trt)
plot = as.factor(data$plot)
time = as.factor(data$time)
y = data$resp
```

a)

```
interaction.plot(time, trt, y)
```



The mean of y for unburned region is higher compared to the burned region. The mean of Y increases with time for unburned region. For the burned region the the mean of y increases with time between 1-9 and decreases between 9-12 and again increases between 12-15.

b)

$$Y_{ijk} = \mu_{...} + \rho_{i(j)} + \alpha_j = \beta_k + (\alpha\beta)_{jk} + \epsilon_{ijk}$$

$\mu_{...}$ is constant

$\rho_{i(j)}$ random $\sim N(0, \sigma_p^2)$, nested with in factor A.

α_j burn/no burn, fixed

β_k : Time factor, fixed

c)

That the covariance for the same plot has compound symmetry and all covariances should be similar in magnitude. Yes.

d)

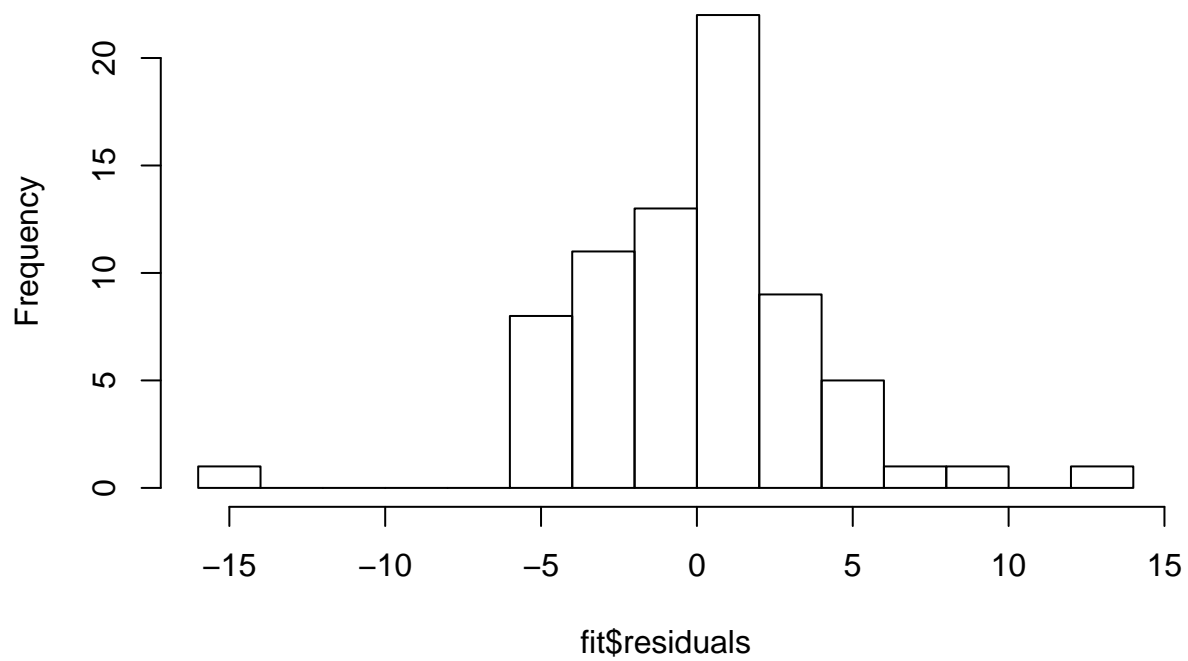
Yes, except for the two extreme points in left and right the rest of the points seems to be normal.

```
fit <- lm(y ~ trt/plot + trt + time + trt:time)
anova(fit)
```

```
## Analysis of Variance Table
##
## Response: y
##          Df Sum Sq Mean Sq  F value    Pr(>F)
## trt         1 9964.0   9964.0 446.7287 < 2.2e-16 ***
## time        3  496.0    165.3   7.4132 0.0003540 ***
## trt:plot    16 3564.1    222.8   9.9870 2.43e-10 ***
## trt:time    3  434.6    144.9   6.4949 0.0008879 ***
## Residuals  48 1070.6     22.3
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

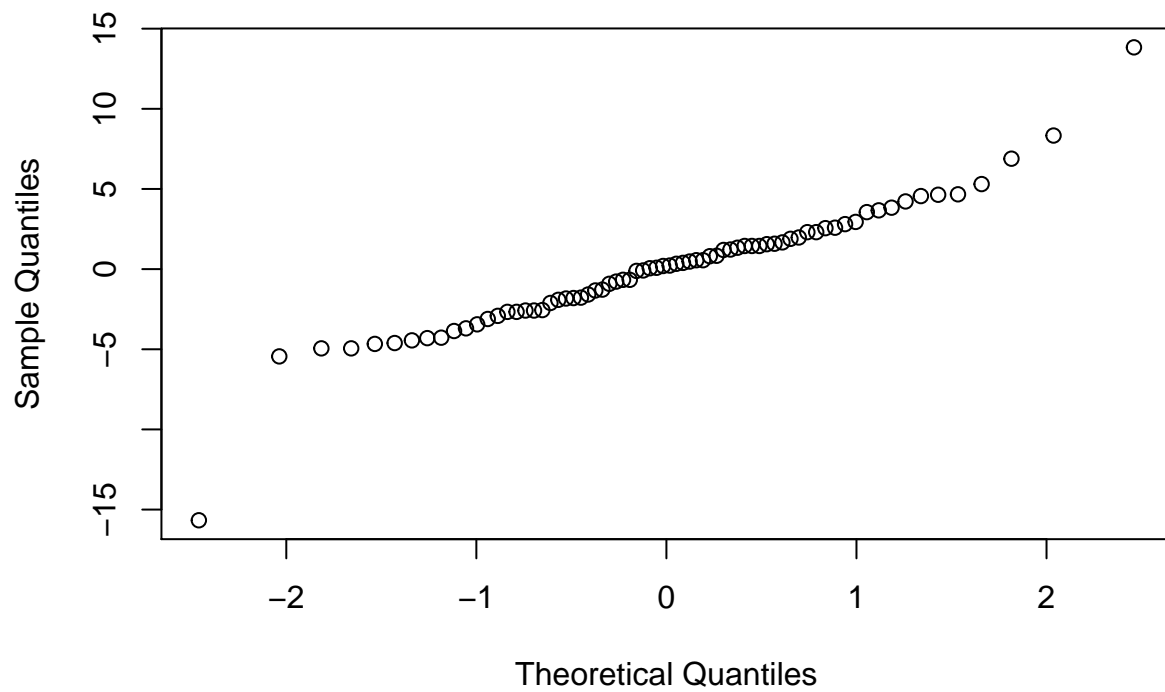
hist(fit$residuals, 20)
```

Histogram of fit\$residuals



```
qqnorm(fit$residuals)
```

Normal Q-Q Plot



e)

$H_0 : \text{all } \alpha_j = 0$ $H_a : \text{not all } \alpha_j \text{ equal zero}$

$$F^* = \frac{MSA}{MSS(A)} = 44.7217235$$

$$F = 4.4939985$$

$F^* > F$ we conclude H_a . That treatment effect exist and burned areas are significantly different.