# TMA4255 Applied Statistics Solution to Exercise 8

#### Problem 1

#### a) Two-sample T-test:

We assume that

- $X_1, \ldots, X_n, Y_1, \ldots, Y_m, X_j \sim N(\mu_X, \sigma_X^2), Y_j \sim N(\mu_Y, \sigma_Y^2),$ n = 10, m = 8.
- $\sigma_X^2 = \sigma_Y^2$

Two-Sample T-Test and CI: X\_i; Y\_i

Two-sample T for X\_i vs Y\_i

N Mean StDev SE Mean X\_i 10 5201,3 10,2 3,2 Y\_i 8 5182,0 19,6 6,9

Difference = mu (X\_i) - mu (Y\_i)
Estimate for difference: 19,3000
95% CI for difference: (4,1579; 34,4421)
T-Test of difference = 0 (vs not =): T-Value = 2,70 P-Value = 0,016 DF = 16
Both use Pooled StDev = 15,0584

#### Explanation of the result from Minitab:

- N: The number of observations in each column.
- MEAN: average=  $\frac{1}{N} \sum_{j=1}^{N} X_j = \bar{X}$ .
- <u>STDEV</u>:  $S = \sqrt{\frac{1}{n-1} \sum_{j=1}^{N} (X_j \bar{X})^2}$ .
- SE MEAN: standard deviation for  $\bar{X}$ , this is equal to  $\frac{S}{\sqrt{N}}$ . (correspondingly for Y.)
- 95 PCT CI: 95 % confidence interval for  $(\mu_X \mu_Y)$ .

The T-statistic is given by

$$T = \frac{\bar{X} - \bar{Y} - (\mu_X - \mu_Y)}{\sqrt{\frac{S^2}{n} + \frac{S^2}{m}}} \sim T_{n+m-2} = T_{16}$$

(Student-T-distributed with 16 degrees of freedom.) Here  $S^2$  is pooled-stdev (see page 308) i.e. estimated variance under the assumption that the two samples have the same

variance:

$$S = \sqrt{\frac{(n-1)S_X^2 + (m-1)S_Y^2}{n+m-2}}$$
$$= \sqrt{\frac{\sum_{j=1}^n (X_j - \bar{X})^2 + \sum_{j=1}^n (Y_j - \bar{Y})^2}{n+m-2}} = 15.1$$

To find a 95% confidence interval we set ut:

$$P\left(-t_{0.025,16} \leqslant \frac{\bar{X} - \bar{Y} - (\mu_X - \mu_Y)}{\sqrt{\frac{1}{n} + \frac{1}{m}}S} \leqslant -t_{0.025,16}\right) = 1 - 0.05 = 0.95$$

$$P\left(\bar{X} - \bar{Y} - t_{0.025,16}\sqrt{\frac{1}{n} + \frac{1}{m}}S \leqslant \mu_X - \mu_Y \leqslant \bar{X} - \bar{Y} + t_{0.025,16}\sqrt{\frac{1}{n} + \frac{1}{m}}S\right) = 0.95$$

The confidence interval is therefore given by:

$$\bar{X} - \bar{Y} \pm t_{0.025,16} \sqrt{\frac{1}{n} + \frac{1}{m}} S = 5201.3 - 5182.0 \pm 2.12 \sqrt{\frac{1}{10} + \frac{1}{8}} 15.1$$
  
= [4.2, 34.4]

• <u>TTEST</u>: Here we test  $H_0$ :  $\mu_X = \mu_Y$  against  $H_1$ :  $\mu_X \neq \mu_Y$  The test is based on the same T-statistic:

$$T_{0 \text{ obs}} = \frac{5201.3 - 5182.0}{\sqrt{\frac{1}{10} + \frac{1}{8}}15.1} = 2.7$$

(We write  $T_{0 \text{ obs}}$  to indicate that we observe T under  $H_0$ , i.e.  $\mu_X - \mu_Y = 0$ .)

• P: p-value,

$$p = P(T_{16} \ge 2.7) + P(T_{16} \le -2.7) = 2P(T_{16} \ge 2.7) = 0.016$$

(Two sided test and symmetric T-distribution.)

): With significance level  $\alpha = 0.01$  we can *not* reject the hypothesis because  $p > \alpha$ , i.e. we can not assume unequal strength in the copper wires.

### b) Variance analysis of one-way grouping:

Rename the variable (to get the same notation as in the book)

$$X_1, X_2, \dots, X_n \to X_{11}, X_{12}, \dots, X_{1n_1}$$
  
 $Y_1, Y_2, \dots, Y_n \to X_{21}, X_{22}, \dots, X_{2n_2}$ 

and we have that  $n_1 = 10$  and  $n_2 = 8$ .  $N = n_1 + n_2 = 18$ . (Total number of observations) Assumptions:

$$E(X_{1j}) = \mu_1, \quad j = 1, \dots, n_1$$
  
 $E(X_{2j}) = \mu_2, \quad j = 1, \dots, n_2$   
 $Var(X_{ij}) = \sigma^2, \quad i = 1, 2$ 

(i.e. the number of groups=2). We follow the notation from the book

$$\mu_i = \mu + \alpha_i,$$

og  $\mu = \frac{n_1 \mu_1 + n_2 \mu_2}{N}$  is "grand mean". We call  $\alpha_i$  the effect of an observation coming from group i

Model:

$$X_{ij} = \mu + \alpha_i + \epsilon_{ij}$$
, der  $\epsilon_{ij}$  er tilfeldige feil.

#### Variance table:

One-way ANOVA: X\_i; Y\_i

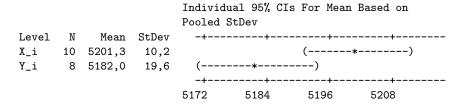
 Source
 DF
 SS
 MS
 F
 P

 Factor
 1
 1656
 1656
 7,30
 0,016

 Error
 16
 3628
 227

 Total
 17
 5284

S = 15,06 R-Sq = 31,33% R-Sq(adj) = 27,04%



Pooled StDev = 15,1

The test that has been done is:

$$H_0$$
:  $\mu_1 = \mu_2$  (1)

$$H_1 \qquad : \qquad \mu_1 \neq \mu_2. \tag{2}$$

Under  $H_0$   $\mu_1 = \mu_2 = \mu$  so that an equivalent test is:

$$H_0 \qquad : \qquad \alpha_1 = \alpha_2 \tag{3}$$

$$H_1$$
:  $\alpha_1 \neq 0$  eller  $\alpha_2 \neq 0$ . (4)

p-verdi:

$$p = P(F_{r-1,N-r} \geqslant F_{\text{obs}}) = 1 - P(F_{1,16} \leqslant 7.30) = 0.016$$

): We have  $p=0.016>\alpha=0.01,$  i.e. we do not reject  $H_0.$ 

The p-value is the same as for the test in a) because

$$T_{\nu}^2 = F_{1,\nu}$$

# Problem 2

a)

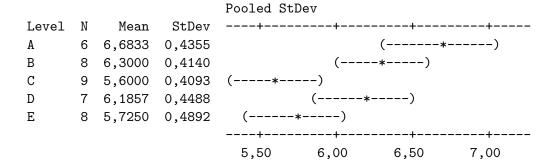
Results for: lympho.MTW

One-way ANOVA: count versus drug

Source DF SS MS F P drug 4 5,703 1,426 7,38 0,000 Error 33 6,372 0,193 Total 37 12,075

S = 0,4394 R-Sq = 47,23% R-Sq(adj) = 40,83%

Individual 95% CIs For Mean Based on



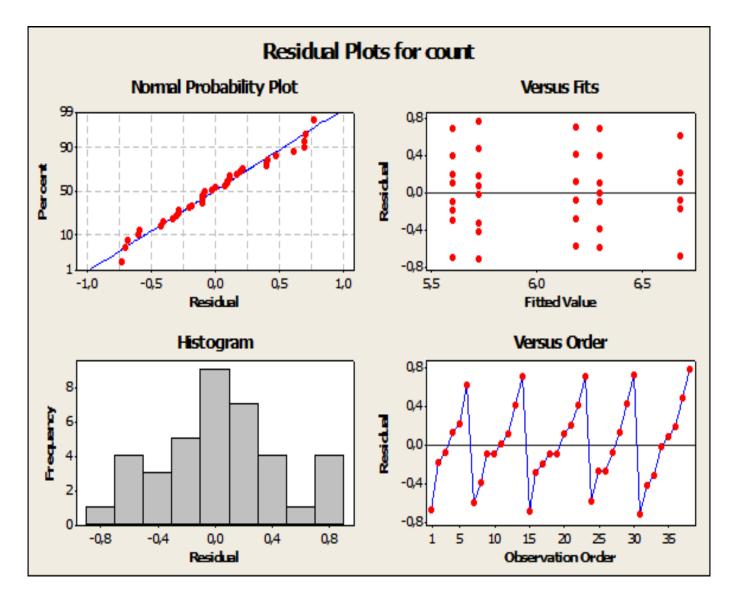
Pooled StDev = 0,4394

The effect of drug is significant.

Bartlett's Test (Normal Distribution)
Test statistic = 0,29; p-value = 0,991

Levene's Test (Any Continuous Distribution)
Test statistic = 0,11; p-value = 0,978

The variances for the different groups are not found to be different. Residual plots show an adequate model fit wrt normality of error terms.



**b)** Using the method of Bonferroni to perform four given comparisons we will use significance level 0.05/4 = 0.125. This was given in the call to Fisher method in MINITAB (meaning that significance level 0.125 is used below).

### Grouping Information Using Fisher Method

```
Grouping
drug
      N
           Mean
      6
         6,6833
Α
                  Α
                  A B
В
         6,3000
      7
         6,1857
                  A B
D
                    ВС
Ε
         5,7250
С
         5,6000
                      С
```

Means that do not share a letter are significantly different.

This means that when we ONLY compare A vs B, B vs C, C vs D and D vs E, we find that

- A and B does not differ,
- B and C differs
- C and D differs
- D and E does not differ.
- c) We now study all pairwise comparisions with the method of Tukey.

Grouping Information Using Tukey Method

drug	N	Mean	Grouping
Α	6	6,6833	A
В	8	6,3000	A B
D	7	6,1857	A B C
E	8	5,7250	ВС
C	9	5,6000	C

Means that do not share a letter are significantly different.

Tukey 95% Simultaneous Confidence Intervals All Pairwise Comparisons among Levels of drug

Individual confidence level = 99,32%

Using Tukeys method we conclude that A is different from both C and E, and B is different from C, but the finding from b) above (C and D differ) is not now significan when more tests are performed.

# Problem 3

$$A_1, \ldots, A_4 =$$
workers (added as  $1, \ldots, 4$  in C2)  $M_1, \ldots, M_4 =$ machines (added as  $1, \ldots, 4$  in C3)

a) We assume that the skills of the workers do not influence the production units. This means we have one-way grouping, and we assume the model

$$Y_{ij} = \mu + \alpha_j + \epsilon_{ij}, \quad \sum_j \alpha_j = 0$$

Here:

- $Y_{ij}$ : number of produced units by machine j and worker i.
- $E(Y_{ij}) = \mu + \alpha_i$ .
- $\epsilon_{ij}$  assumed independent and  $\sim N(0, \sigma^2) \ \forall i, j.$
- $\alpha_j$  is a factor which is special for machine j.
- $\mu$ : "average effect"

Wish to test wether the machines have different capacities:

 $H_0$ :  $\alpha_1 = \alpha_2 = \alpha_3 = \alpha_4 = 0$  $H_1$ : at least one not equal.

The total variation in the data  $SS_{\text{tot}} = \sum_{j=1}^{4} \sum_{i=1}^{4} (Y_{ij} - \bar{Y}_{..})^2$ , can be written as a sum of two sums of squares:[Theorem. 13.1]

$$SS_{\text{tot}} = SS_A + SS_E = \sum_{i=1}^4 4(\bar{Y}_{.i} - \bar{Y}_{..})^2 + \sum_{i=1}^4 \sum_{i=1}^4 (Y_{ij} - \bar{Y}_{.j})^2$$

It can be shown that [Teorem 13.2]

$$E(SS_A) = (4-1)\sigma^2 + \sum_{i=1}^4 4\alpha_i^2 = 3\sigma^2 + 4\sum_i \alpha_i^2$$

$$E(SS_E) = (16-4)\sigma^2$$

$$F = \frac{MS_A}{MS_E} = \frac{SS_A/(4-1)}{SS_E/(16-4)} \sim F_{(4-1),(16-4)} = F_{3,12}$$

We see that if  $H_0$  is correct, we can expect an  $F_{0 \text{ obs}}$  of about 1. If  $H_0$  is wrong, we can expect a big value of  $F_{0 \text{ obs}}$ .

Minitab gives us:

One-way Analysis of Variance Analysis of Variance for Data

Source	DF	SS	MS	F	P			
M	3	72,0	24,0	1,58	0,245			
Error	12	182,0	15,2					
Total	15	254,0						
				Individual 95% CIs For Mean				
				Based on	Pooled	StDev		
Level	N	Mean	StDev	-+	+	+		-
1	4	72,000	2,944	(	*	)		
2	4	75,000	3,162	(		*	)	
3	4	77,000	4,243		(	*-	)	
4	4	72,000	4,899	(	*	)		
				-+	+	+		-
Pooled St	Dev =	3,894	6	8,0 7	2,0	76,0	80,0	

Here we have that:

$$F_{0 \text{ obs}} = \frac{SS_A/3}{SS_E/12} = \frac{24.0}{15.2} = 1.58$$

the p-value:

$$p = P(F_{3.12} > F_{0 \text{ obs}}) = P(F_{3.12} > 1.58) = 0.245$$

): p is larger than any reasonable significance level  $\alpha$ , which means we can not reject  $H_0$ , and claim that there is a difference between the machines.

b) Now we assume that skills of the workers have an influence. Model:

$$X_{ij} = \mu + \alpha_j + \beta_i + \epsilon_{ij}, \quad \sum_j \alpha_j = \sum_i \beta_i = 0$$

We have:

- $X_{ij}$ : The number of produced units with machine j and worker i.
- $\epsilon_{ij}$  assumed independent and  $\sim N(0, \sigma^2) \ \forall i, j$ .
- $\alpha_j$  is a factor which is special for machine j.
- $\beta_i$  is a factor which is special for worker i.
- $\mu$ : "average effect"

We have the same hypothesis test as in **a**):  $H_0$ :  $\alpha_1 = \alpha_2 = \alpha_3 = \alpha_4 = 0$  against  $H_1$ : at least one is different.

We split the total variation into three sums of squares

$$SS_{\text{tot}} = SS_{\text{mask}} + SS_{\text{arb}} + SS_{E}$$

$$\updownarrow$$

$$\sum_{j=1}^{4} \sum_{i=1}^{4} (X_{ij} - \bar{X}_{\cdot \cdot})^{2} = 4 \sum_{j=1}^{4} (\bar{X}_{\cdot j} - \bar{X}_{\cdot \cdot})^{2} + \sum_{i=1}^{4} (\bar{X}_{i \cdot} - \bar{X}_{\cdot \cdot})^{2}$$

$$+ \sum_{j=1}^{4} \sum_{i=1}^{4} (X_{ij} - \bar{X}_{i \cdot} - \bar{X}_{\cdot j} + \bar{X}_{\cdot \cdot})^{2}$$

The same type of argument as in a) tells us that we can expect a big value of  $F_{0 \text{ obs}}$  if  $H_0$  is wrong.

Here

$$F = \frac{SS_{\text{mask}}/(4-1)}{SS_E/((4-1)(4-1))} \sim F_{4-1,(4-1)(4-1)} = F_{3,9}$$

Minitab gives:

Two-way Analysis of Variance

Analysis of Variance for Data

Source	DF	SS	MS	F	P
A	3	160,00	53,33	21,82	0,000
M	3	72,00	24,00	9,82	0,003
Error	9	22,00	2,44		
Total	15	254,00			

And we see that

$$F_{0 \text{ obs}} = \frac{72.0/3}{22.0/9} = 9.82,$$

and this gives p-value:  $p = P(F_{3,9} > 9.82) = 0.003$ .

): We have a small p and we reject  $H_0$ .

c) Expected number of produced units from machine  $M_2$ :

$$\mu_{\cdot 2} = E(X_{\cdot 2}) = \mu + \alpha_2$$

Estimator:  $\hat{\mu}_{\cdot 2} = \frac{1}{4} \sum_{i=1}^{4} X_{i2}$ . This gives the point estimate:  $\hat{\mu}_{\cdot 2} = \frac{1}{4} (77 + 71 + 78 + 74) = 75$ . We have:

$$E(\hat{\mu}_{\cdot 2}) = \frac{1}{4} \sum_{i=1}^{4} E(X_{i2}) = \frac{1}{4} \sum_{i=1}^{4} (\mu + \alpha_2 + \beta_i) = \mu + \alpha_2 + \frac{1}{4} \sum_{i=1}^{4} \beta_i = \mu + \alpha_2$$

and

$$Var(\hat{\mu}_{\cdot 2}) = E[((\hat{\mu}_{\cdot 2}) - E(\hat{\mu}_{\cdot 2}))^{2}] = E\left[\left(\frac{1}{4}\sum_{i=1}(Y_{i2} - E(Y_{i2}))\right)^{2}\right]$$

$$= E\left[\left(\frac{1}{4}\sum_{i=1}^{4}\epsilon_{i2}\right)^{2}\right] = \frac{1}{16}\sum_{i=1}^{4}E(\epsilon_{i2})^{2} = \left(\frac{1}{16}\right)^{2}\sum_{i=1}^{4}Var(\epsilon_{i2})$$

$$= \frac{1}{4}\sigma^{2}$$

Therefore we get  $\hat{\mu}_{\cdot 2} \sim N(\mu + \alpha_2, \frac{1}{4}\sigma^2) \Rightarrow \frac{\hat{\mu}_{\cdot 2} - (\mu + \alpha_2)}{\sigma/2} \sim N(0, 1)$ .  $\sigma^2$  is estimated in **b**) as  $S^2 = \frac{1}{9}SS_E$ .

Now we have:

$$\frac{\hat{\mu}_{\cdot 2} - (\mu + \alpha_2)}{S/2} \sim T_9$$

(same number of degrees of freedom as  $SS_E$ ).

 $(1-\alpha)\cdot 100$  % confidence interval:

$$P\left(-t_{\alpha/2,9} \leqslant \frac{\hat{\mu}_{\cdot 2} - (\mu + \alpha_2)}{S/2} \leqslant t_{\alpha/2,9}\right) = 1 - \alpha$$

$$\updownarrow$$

$$P(\hat{\mu}_{\cdot 2} - t_{\alpha/2,9}S/2 \leqslant \mu + \alpha_2 \leqslant \hat{\mu}_{\cdot 2} + t_{\alpha/2,9}S/2) = 1 - \alpha$$

With numbers:  $\hat{\mu}_{\cdot 2} = 75$ ,  $\alpha = 0.1$ ,  $t_{0.05,4} = 1.83$ ,  $S^2 = 2.444$ .

): 90 % confidence interval  $\mu + \alpha_2$ : [73.6, 76.4].

### Problem 4

Assume  $Y_{ijk}$  to be independent and normally distributed  $\sim N(\mu_{ij}, \sigma^2)$ .

- i indicates cottontype, i = 1, 2, 3.
- j indicates silktype, j = 1, 2, 3, 4.
- k indicates trial. k with combination ij, k = 1, 2.

$$E(Y_{ijk}) = \mu_{ij} = \mu + \alpha_i + \beta_j + \gamma_{ij}, \ \sum_i \alpha_i = \sum_j \beta_j = \sum_i \gamma_{ij} = \sum_j \gamma_{ij} = 0.$$

**a**)

- $\mu$  is an "average effect".
- $\alpha_i$  is a factor that is special for cottontype i.
- $\beta_j$  is a factor that is special for silktype j.
- $\gamma_{ij}$  is a factor that is special for the interaction between cottontype i and silktype j.

Estimators for these four parameters: This is done using maximum likelihood here (optional), can also be done based on intuition.

Intuition will give us that:

- $\mu$  can be estimated using the overall mean,  $\hat{\mu} = \bar{Y}_{...}$ ,
- $\alpha_i$  by the difference between the mean for cotton group i and the overall mean,  $\hat{\alpha}_i = \bar{Y}_{i..} \bar{Y}_{...}$ ,
- $\beta_j$  by the difference between the mean for silk group j,  $\hat{\beta}_j = \bar{Y}_{.j.} \bar{Y}_{...}$ ,
- $\gamma_{ij}$  by the difference between the mean for the combined cotton and slik group and the mean over cotton, silk and the overall mean,  $\hat{\gamma}_{ij} = \bar{Y}_{ij} \bar{Y}_{i..} \bar{Y}_{.j.} + \bar{Y}_{...}$

The following is optional, but should be possible to follow: Probability density of one  $Y_{ijk}$ :

$$f_Y(Y_{ijk}) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2\sigma^2}(Y_{ijk} - \mu_{ij})^2}$$

Find the joint density:

$$l(\mu, \alpha_i, \beta_j, \gamma_{ij}, \sigma^2 | Y_{ijk}) = \prod_{i,j,k} f_Y(Y_{ijk})$$

$$= \left[ \frac{1}{\sqrt{2\pi}\sigma} \right]^{3+4+2} e^{-\frac{1}{2\sigma^2} \sum_i \sum_j \sum_k (Y_{ijk} - \mu - \alpha_i - \beta_j - \gamma_{ij})^2}$$

In is a strictly increasing function, and  $\ln l$  therefore has the same maximum points as l.

$$L = \ln l = 9 \cdot \ln \frac{1}{\sqrt{2\pi}\sigma} - \frac{1}{2\sigma^2} \sum_{i} \sum_{j} \sum_{k} (Y_{ijk} - \mu - \alpha_i - \beta_j - \gamma_{ij})^2$$

Maximizing L wrt. the unknown parameters:

$$\begin{split} \frac{\partial L}{\partial \mu} &= -\frac{1}{2\sigma^2} 2 \sum_i \sum_j \sum_k (Y_{ijk} - \mu - \alpha_i - \beta_j - \gamma_{ij})(-1) \\ &= \frac{1}{2\sigma^2} 2 \sum_i \sum_j \sum_k (Y_{ijk} - \mu) \\ \frac{\partial L}{\partial \mu} &= 0 \\ &\Rightarrow \sum_i \sum_j \sum_k (Y_{ijk} - \hat{\mu}) = 0 \\ &\Rightarrow \hat{\mu} = \bar{Y}... \end{split}$$

$$\frac{\partial L}{\partial \alpha_i} = \frac{1}{\sigma^2} \sum_j \sum_k (Y_{ijk} - \mu - \alpha_i - \beta_j - \gamma_{ij})$$

$$= \frac{1}{\sigma^2} \sum_j \sum_k (Y_{ijk} - \mu - \alpha_i)$$

$$\frac{\partial L}{\partial \alpha_i} = 0$$

$$\Rightarrow \frac{1}{\sigma^2} \sum_j \sum_k (Y_{ijk} - \hat{\mu} - \hat{\alpha}_i) = 0$$

$$\Rightarrow \hat{\alpha}_i = \bar{Y}_{i..} - \hat{\mu} = \bar{Y}_{i..} - \bar{Y}_{...}$$

$$\frac{\partial L}{\partial \beta_i} = 0 \Rightarrow \hat{\beta}_j = \bar{Y}_{.j.} - \hat{\mu} = \bar{Y}_{.j.} - \bar{Y}_{...}$$
 (5)

$$\begin{split} \frac{\partial L}{\partial \gamma_{ij}} &= \frac{1}{\sigma^2} \sum_k (Y_{ijk} - \mu - \alpha_i - \beta_j - \gamma_{ij}) \\ \frac{\partial L}{\partial \gamma_{ij}} &= 0 \\ \Rightarrow \sum_k Y_{ijk} - k\hat{\mu} - k\hat{\alpha}_i - k\hat{\beta}_j - k\hat{\gamma}_{ij} &= 0 \\ \Rightarrow \hat{\gamma}_{ij} &= \frac{1}{k} \sum_k Y_{ijk} - \hat{\mu} - \hat{\alpha}_i - \hat{\beta}_j \\ &= \bar{Y}_{ij} - \bar{Y}_{\cdots} - (\bar{Y}_{i\cdots} - \bar{Y}_{\cdots}) - (\bar{Y}_{\cdot j} - \bar{Y}_{\cdots}) \\ &= \bar{Y}_{ij} - \bar{Y}_{i\cdots} - \bar{Y}_{\cdot j} + \bar{Y}_{\cdots} \end{split}$$

(maksimum because  $\frac{\partial^2 L}{\partial \gamma_{ij}^2} < 0.$ )

#### **b)** Analysis of variance table:

Kilde	DF(DF)	Sum of squares (SS)
Rows $(A)$	a-1	$bn \sum_{i} (\bar{X}_{i\cdots} - \bar{X}_{\cdots})^2$
Columns $(B)$	b-1	$an \sum_{j} (\bar{X}_{\cdot j \cdot} - \bar{X}_{\cdot \cdot \cdot})^2$
Interaction $(AB)$	(a-1)(b-1)	$n \sum_{i} \sum_{j} (\bar{X}_{ij\cdot} - \bar{X}_{i\cdot\cdot} - \bar{X}_{\cdot j\cdot} + \bar{X}_{\cdot\cdot})^2$
Error	ab(n-1)	$\sum_{i}\sum_{j}\sum_{k}(X_{ijk}-\bar{X}_{ij.})^{2}$
Total	abn-1	$\sum_{i}\sum_{j}\sum_{k}(X_{ijk}-\bar{X}_{})^{2}$

#### Here:

- a = number of cottontypes = 3.
- b = number of silktypes = 4.
- n = number of replicates for each combination(ij) = 2.
- $X_{ijk} = k$ -th observation with cotton type i and silktype j.
- $\bar{X}_{ij}$  = average value of the *n* observations in "cell (i,j)".
- $\bar{X}_{i..}$  = average value of the  $b \cdot n$  observations of cotton type i.
- $\bar{X}_{.j.}$  = average value of the  $a \cdot n$  observations of silk type j.
- $\bar{X}$ ... = average value of all observations.

In general we have that

$$MS = SS/DF = \frac{\text{sum of squares}}{DF}$$

Hypothesis test:

 $H_0$ : No interaction  $H_1$ : Interaction

Under  $H_0$ 

$$F = \frac{MS_{\mathrm{interaction}}}{MS_{\mathrm{error}}} \sim F_{6,12}.$$

From the Minitab results in c) we have a p-value:  $P(F_{6,12} > 2.31) = 0.103$ .

): We use  $\alpha = 0.05$  and can not reject  $H_0$ . We conclude that there is not evidence to believe that an interaction term is present.

## c) Hypothesis test for A:

 $H_0$ : A has no effect, i.e.  $\alpha_1 = \alpha_2 = \alpha_3 = 0$ 

 $H_1: A \text{ has effect}$ 

Under  $H_0$ 

$$F_{\mathrm{obs}} = \frac{MS_A}{MS_{\mathrm{error}}} = \frac{SS_A/2}{SS_{\mathrm{error}}/12} \sim F_{2,12}$$

Minitab gives:

Two-way Analysis of Variance Analysis of Variance for Response

Source	DF	SS	MS	F	P
A	2	434,2	217,1	15,70	0,000
В	3	200,3	66,8	4,83	0,020
Interaction	6	191,4	31,9	2,31	0,103
Error	12	166,0	13,8		
Total	23	992,0			

p-verdi:

$$p = P(F_{2,12} > F_{\text{obs}}) = P(F_{2,12} > 15.70) = 0.000$$

): We reject  $H_0$  and claim that A har effect.

Hypothesis test for B:

 $H_0$ : B has no effekt, i.e.  $\beta_1 = \beta_2 = \beta_3 = 0$ 

 $H_1: B \text{ has effect}$ 

Just like above we get:

$$F_{\rm obs} = \frac{MS_B}{MS_{\rm error}} \sim F_{3,12}$$

p-verdi:

$$p = P(F_{3,12} > F_{\text{obs}}) = P(F_{3,12} > 4.83) = 0.02$$

): We use  $\alpha = 0.05$  and we can therefore reject  $H_0$  and say that B has effect.

Conclusion: Both cotton type and silk type influence the quality.