

35

a

$$\begin{aligned}
f_A &= \frac{MSA}{MSE} \\
&= \frac{SSA/(I-1)}{SSE/((I-1)(J-1))} \\
&= \frac{30.6/4}{59.2/12} \\
&= 1.55 < F_{.05,4,12}
\end{aligned}$$

We don't reject H_{0A}

b

$$\begin{aligned}
f_B &= \frac{MSB}{MSE} \\
&= \frac{SSB/(J-1)}{SSE/((I-1)(J-1))} \\
&= \frac{44.1/3}{59.2/12} \\
&= 2.98 < F_{.05,3,12} = 3.49
\end{aligned}$$

We don't reject H_{0B}

38

```

exe11.38=read.table(
  "http://www.uio.no/studier/emner/matnat/math/STK2120/v16/exe11-38.txt",
  header=T, sep=",");
fit.aov=aov(Response~factor(Paint)+factor(Roller), data=exe11.38);
summary(fit.aov)

```

```

#           Df Sum Sq Mean Sq F value Pr(>F)
#factor(Paint)  3 159.58   53.19   7.848 0.0169 *
#factor(Roller)  2   38.00   19.00   2.803 0.1381
#Residuals     6   40.67    6.78

```

#b) Paint is significant for $\alpha=0.05$.
#c) Roller is not.

```

tukey.Paint=TukeyHSD(fit.aov, "factor(Paint)")
png('exe1138tuk.png')
plot(tukey.Paint)
dev.off()
print(tukey.Paint)

```

#Paint Brand 1 seems preferable

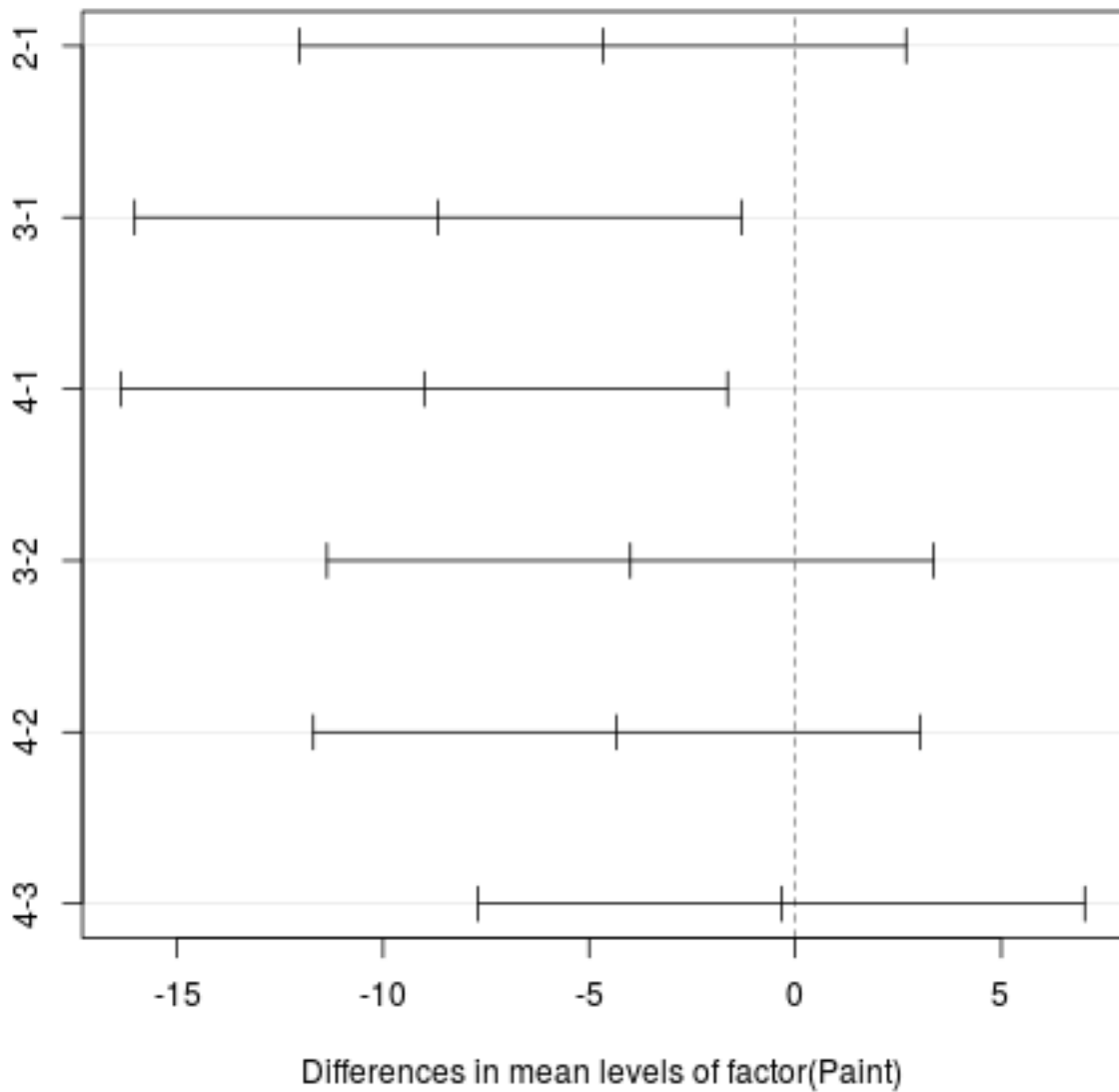
```

#e
png('exe1138norm.png')
qqnorm(fit.aov$residuals)

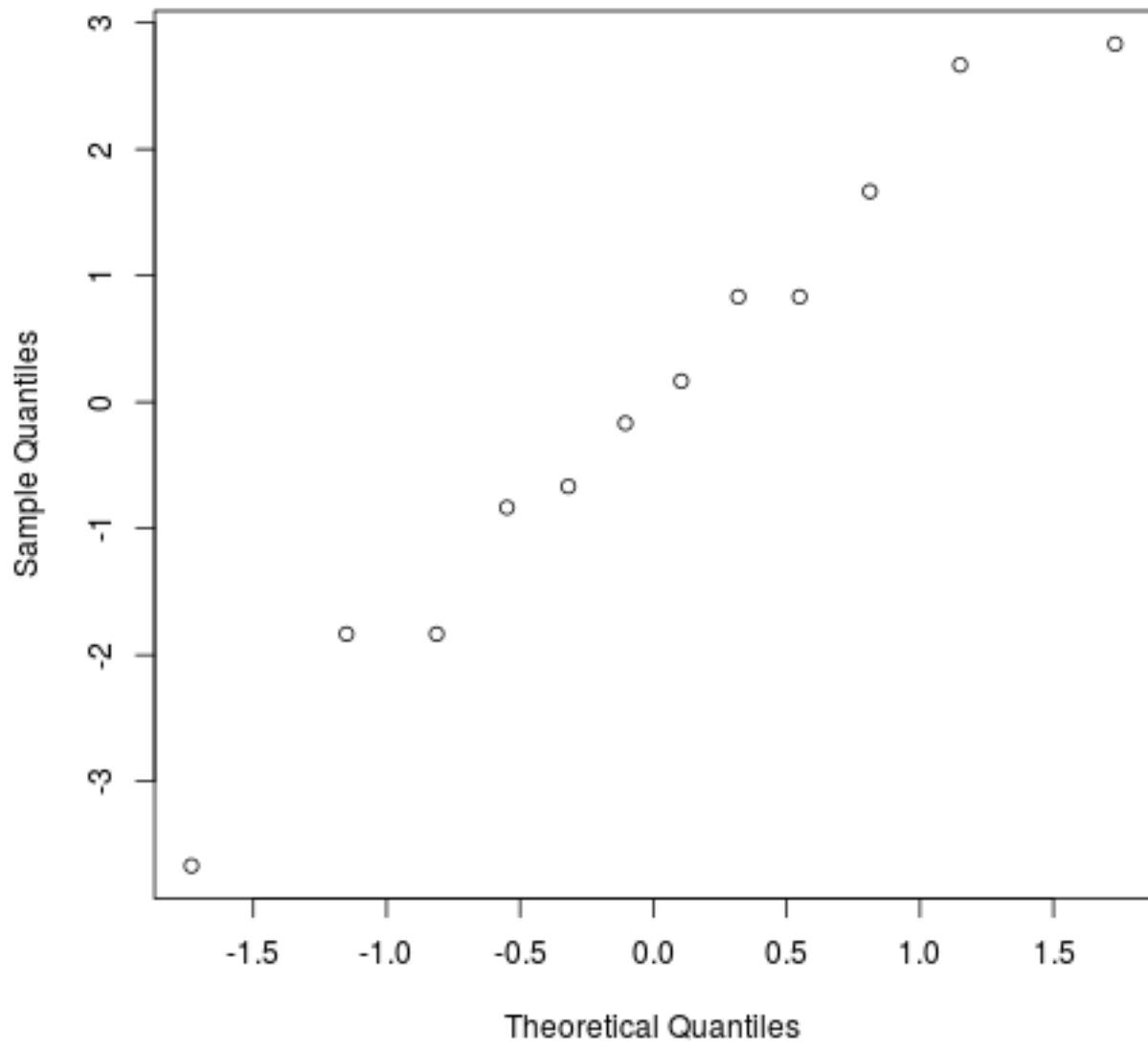
```

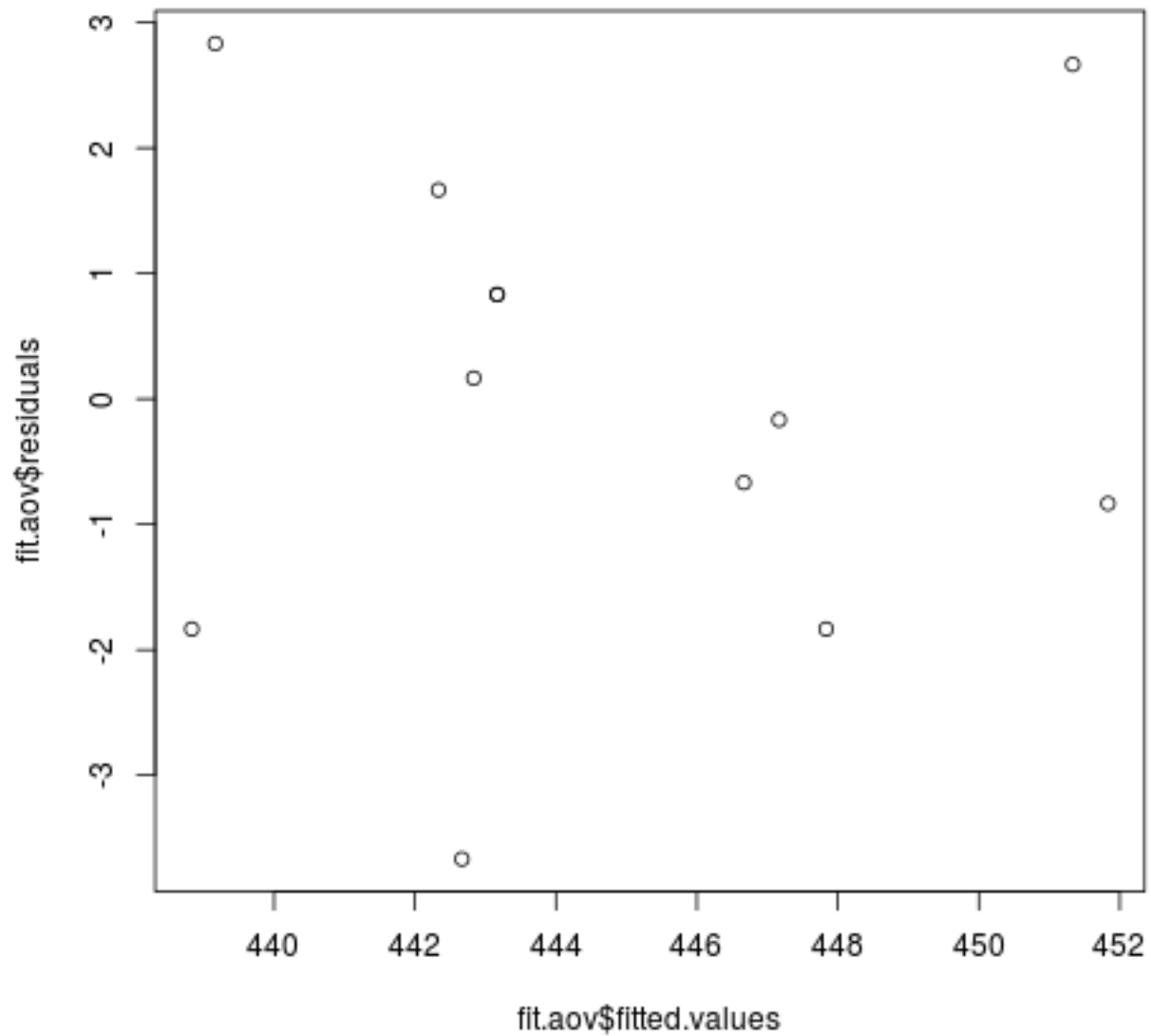
```
dev.off()
#Normality is reasonable
png('exe1138res.png')
plot(fit.aov$fitted.values, fit.aov$residuals)
dev.off()
#Constant variance is reasonable
```

95% family-wise confidence level



Normal Q-Q Plot





42

```
exe11.42=read.table(
  "http://www.uio.no/studier/emner/matnat/math/STK2120/v16/exe11-42.txt",
  header=T,sep=" ")
```

```
fit.aov=aov(response~factor(subject)+factor(Stool),data=exe11.42);
summary(fit.aov)
```

```
#           Df Sum Sq Mean Sq F value    Pr(>F)
#factor(subject)  8  66.50   8.312   6.866 0.000106 ***
#factor(Stool)    3  81.19  27.065  22.356 3.93e-07 ***
#Residuals      24  29.06   1.211
#-----
```

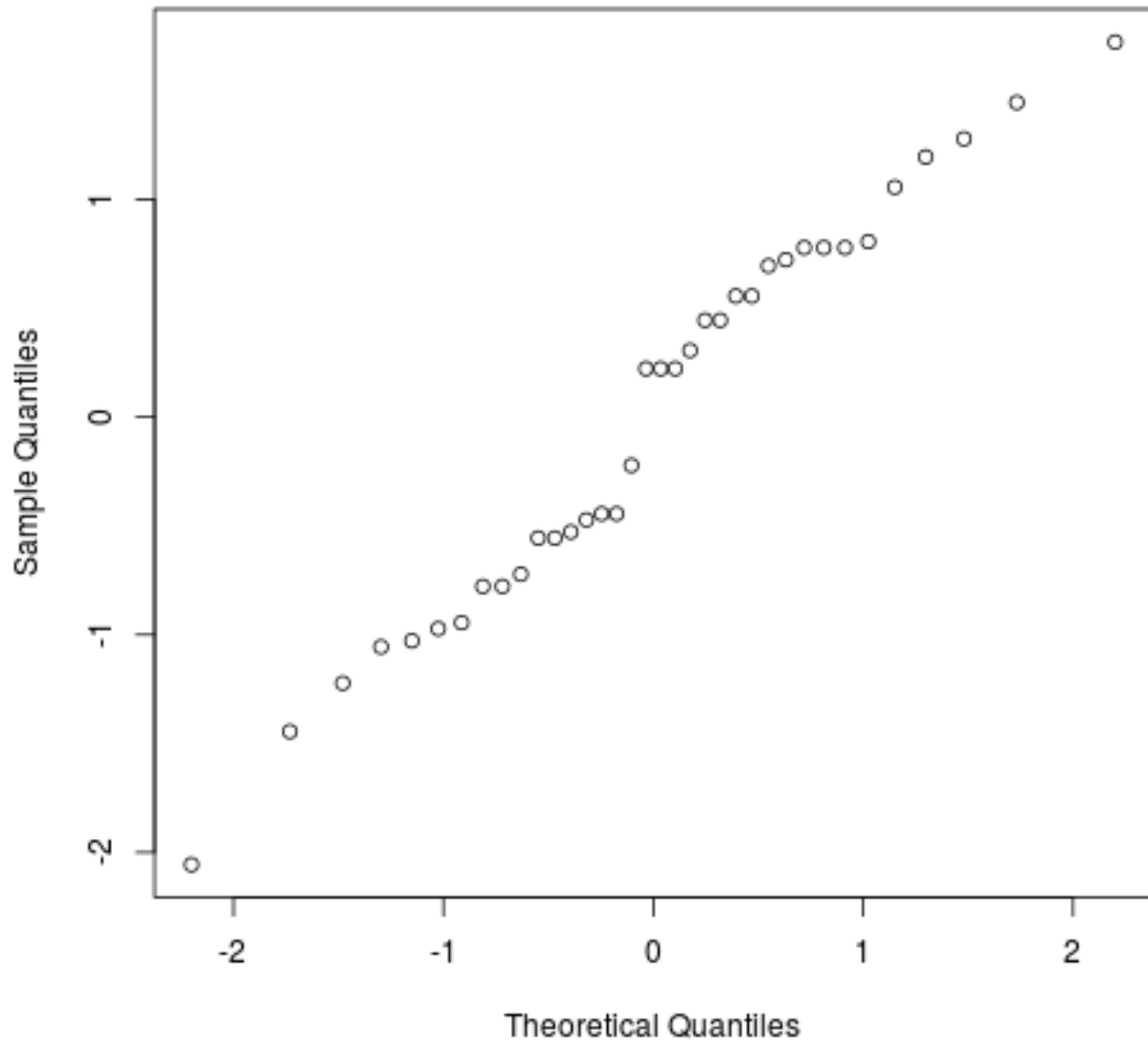
```
png('exe1142norm.png')
qqnorm(fit.aov$residuals)
dev.off()

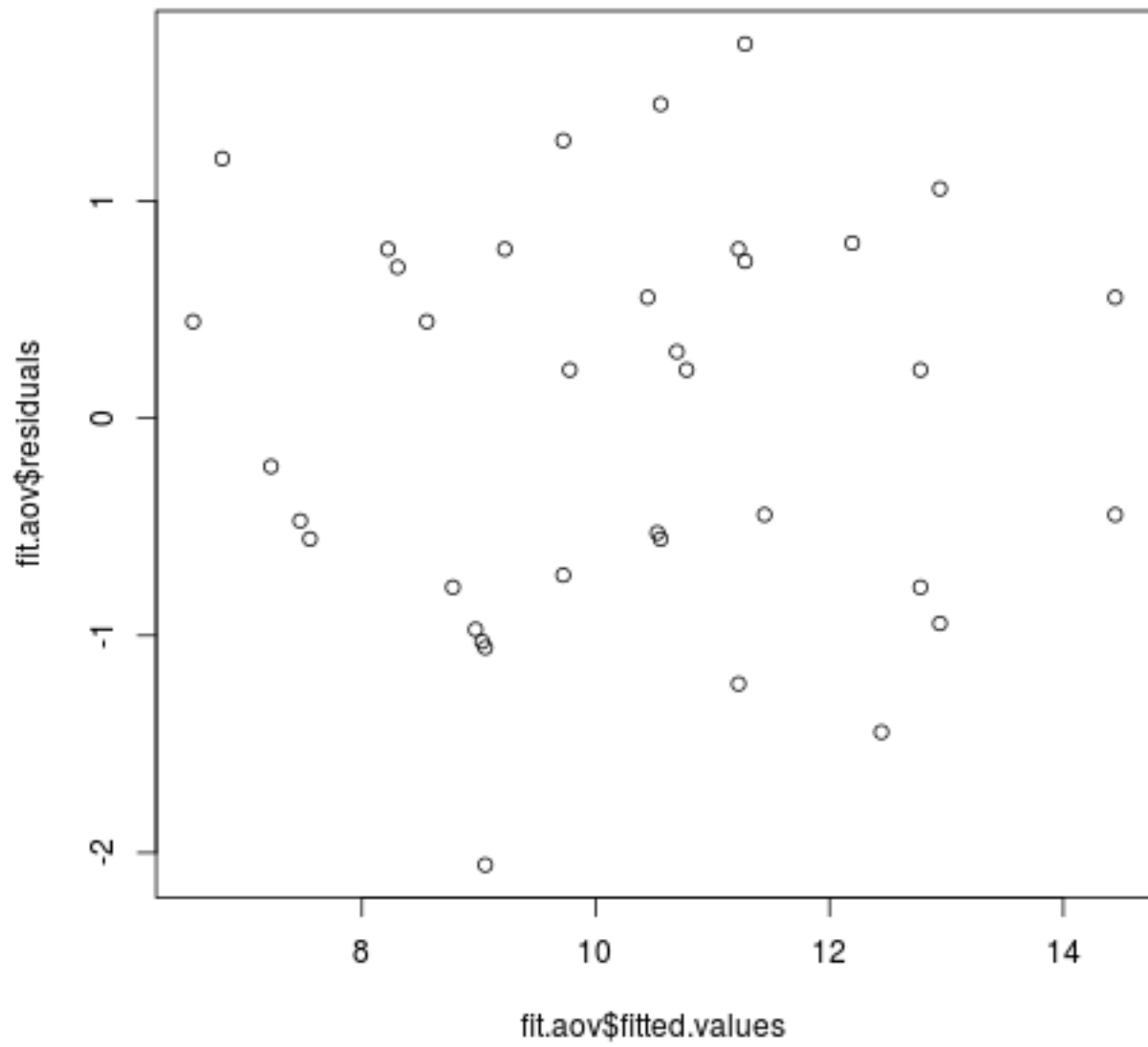
png('exe1142res.png')
plot(fit.aov$fitted.values, fit.aov$residuals)
dev.off()

tukey.Stool=TukeyHSD(fit.aov, "factor(Stool)")

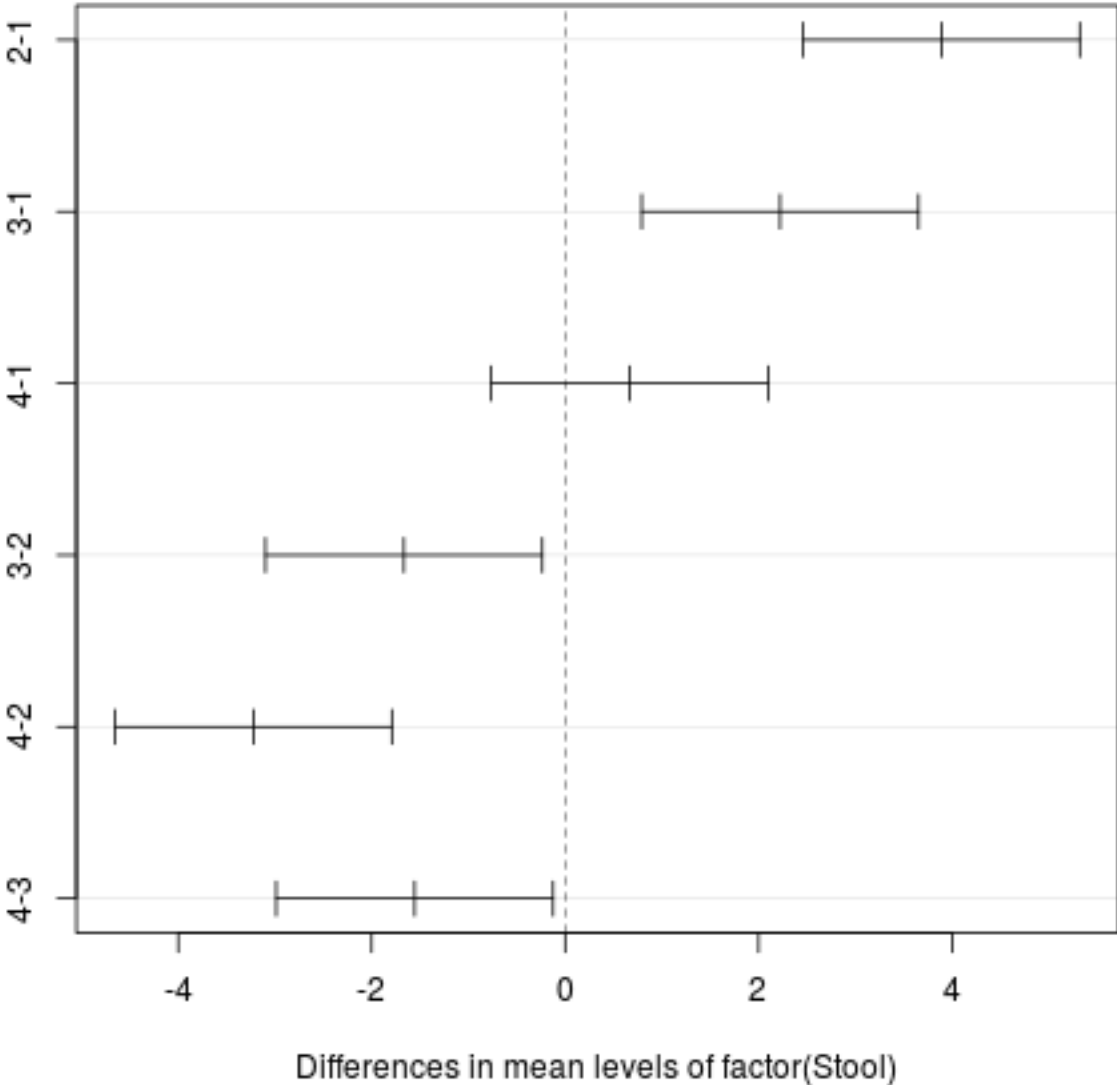
png('exe1142tuk.png')
plot(tukey.Stool)
dev.off()
print(tukey.Stool)
```

Normal Q-Q Plot





95% family-wise confidence level



a

$$\begin{aligned}
y_{ij} &= x_{ij} + d \\
\bar{y}_{..} &= \bar{x}_{..} + d \\
SST_y &= \sum_i \sum_j (y_{ij} - \bar{y}_{..})^2 \\
&= \sum_i \sum_j ((x_{ij} + d) - (\bar{x}_{..} + d))^2 \\
&= \sum_i \sum_j (x_{ij} - \bar{x}_{..})^2 \\
&= SST_x
\end{aligned}$$

$$\begin{aligned}
SSE_y &= \sum_i \sum_j (y_{ij} - \bar{y}_{i.} - \bar{y}_{.j} + \bar{y}_{..})^2 \\
&= \sum_i \sum_j ((x_{ij} + d) - (\bar{x}_{i.} + d) - (\bar{x}_{.j} + d) + (\bar{x}_{..} + d))^2 \\
&= \sum_i \sum_j (x_{ij} - \bar{x}_{i.} - \bar{x}_{.j} + \bar{x}_{..})^2 \\
&= SSE_x
\end{aligned}$$

b

$$\begin{aligned}
y_{ij} &= cx_{ij} \\
\bar{y}_{..} &= c\bar{x}_{..} \\
SST_y &= \sum_i \sum_j (y_{ij} - \bar{y}_{..})^2 \\
&= \sum_i \sum_j (cx_{ij} - c\bar{x}_{..})^2 \\
&= c^2 \sum_i \sum_j (x_{ij} - \bar{x}_{..})^2 \\
&= c^2 SST_x
\end{aligned}$$

$$\begin{aligned}
SSE_y &= \sum_i \sum_j (y_{ij} - \bar{y}_{i.} - \bar{y}_{.j} + \bar{y}_{..})^2 \\
&= \sum_i \sum_j (cx_{ij} - c\bar{x}_{i.} - c\bar{x}_{.j} + c\bar{x}_{..})^2 \\
&= c^2 \sum_i \sum_j (x_{ij} - \bar{x}_{i.} - \bar{x}_{.j} + \bar{x}_{..})^2 \\
&= c^2 SSE_x
\end{aligned}$$

$$\begin{aligned}
F_{Ay} &= MSA_y / MSE_y \\
&= c^2 MSA_x / (c^2 MSE_x) \\
&= MSA_x / MSE_x = F_{Ax}
\end{aligned}$$

47

$$\begin{aligned}
&E(\bar{X}_i - \bar{X}_{..}) = \\
&E\left(\sum_j X_{ij}/J - \sum_i \sum_j X_{ij}/IJ\right) = \\
&\sum_j (\mu + \alpha_i + \beta_j)/J - \sum_i \sum_j (\mu + \alpha_i + \beta_j)/(IJ) = \\
&\mu + \alpha_i + \sum_j \beta_j/J - \sum_i (\mu + \alpha_i + \sum_j \beta_j/J)/I = \\
&\mu + \alpha_i - \sum_i (\mu + \alpha_i)/I = \\
&\mu + \alpha_i - \mu + \sum_i \alpha_i/I = \\
&\alpha_i
\end{aligned}$$

Extra 3

a

$$T = \frac{\bar{D}}{S/\sqrt{J}}$$

$$T^2 = \frac{\bar{D}^2}{S^2/J}$$

$$T^2 = \frac{\bar{D}^2}{1/(J-1) \sum_j (D_j - \bar{D})^2 / J}$$

$$F_A = MSA/MSE$$

$$= (SSA/(I-1))/(SSE/(I-1)(J-1))$$

$$= \frac{(\bar{X}_{1.} - \bar{X}_{..})^2 + (\bar{X}_{2.} - \bar{X}_{..})^2}{\sum_j ((X_{j1} - \bar{X}_{1.} - \hat{X}_{.j} + \bar{X}_{..})^2 + (X_{j2} - \bar{X}_{2.} - \hat{X}_{.j} + \bar{X}_{..})^2) / (J-1)}$$

$$\begin{aligned} \bar{D} &= \sum_j (X_{1j} - X_{2j}) / J \\ &= \bar{X}_{1.} - \bar{X}_{2.} \end{aligned}$$

$$\begin{aligned} \bar{D}^2 &= (\bar{X}_{1.} - \bar{X}_{2.})^2 = 4(\bar{X}_{1.} - \bar{X}_{..})^2 = 4(\bar{X}_{2.} - \bar{X}_{..})^2 = 2[(\bar{X}_{1.} - \bar{X}_{..})^2 + (\bar{X}_{2.} - \bar{X}_{..})^2] \\ &= 2SSA/J \end{aligned}$$

$$D_j - \bar{D} = (X_{1j} - X_{2j}) - (\bar{X}_{1.} - \bar{X}_{2.})$$

$$= 2(X_{1j} - X_{.j}) - 2(\bar{X}_{1.} - \bar{X}_{..})$$

$$= 2(X_{1j} - X_{.j}) - 2(\bar{X}_{1.} - \bar{X}_{..})$$

$$(D_j - \bar{D})^2 = 4((X_{1j} - X_{.j}) - (\bar{X}_{1.} - \bar{X}_{..}))^2$$

$$= 4((X_{2j} - X_{.j}) - (\bar{X}_{2.} - \bar{X}_{..}))^2$$

$$= 2((X_{1j} - X_{.j}) - (\bar{X}_{1.} - \bar{X}_{..}))^2 + 2((X_{2j} - X_{.j}) - (\bar{X}_{2.} - \bar{X}_{..}))^2$$

$$\sum_j (D_j - \bar{D})^2 = 2SSE$$

$$\begin{aligned} \frac{\bar{D}^2}{S^2/J} &= \frac{2SSA}{1/(J-1)2SSE/J} \\ &= \frac{MSA}{MSE/J} \end{aligned}$$

Eksamensoppgave 3

a

Source	<i>df</i>	<i>SS</i>	<i>MS</i>	<i>F</i>
Treatment		37.784		
Residual				
Total		48.736		

Source	<i>df</i>	<i>SS</i>	<i>MS</i>	<i>F</i>
Treatment	$(I - 1)$	37.784	$MSTr = SSTr / (I - 1)$	$f = MSTr / MSE$
Residual	$(n - I)$	$SSE = SST - SSTr$	$MSE = SSE / (n - I)$	
Total	$n - 1$	48.736		

Source	<i>df</i>	<i>SS</i>	<i>MS</i>	<i>F</i>
Treatment	3	37.784	12.595	25.450
Residual	23	10.952	0.476	
Total	26	48.736		

$$F_{.05,3,23} = 3.028$$

We reject H_0

b

(i) Reject H_{12} (H_{ij}) if:

$$\frac{X_{1\cdot} - X_{2\cdot}}{\sqrt{MSE(1/9 + 1/7)}} \geq t_{\alpha,23}$$

$$\frac{X_{i\cdot} - X_{j\cdot}}{\sqrt{MSE(1/9 + 1/7)}} \geq t_{\alpha,23}$$

(ii) We use the results from exercise 2 with $k = I(I - 1)/2 = 4*3/2 = 6$. We then get that $t_{\alpha/(2k),23} = t_{\alpha/12,23}$ corrects for the k different two sided t-tests.

(iii)

$$T_{ij} :$$

$$|\bar{y}_{\cdot i} - \bar{y}_{\cdot j}| < t_{.05/12,23} \sqrt{MSE(1/n_i + 1/n_j)}$$

$$|\bar{y}_{\cdot i} - \bar{y}_{\cdot j}| / \sqrt{1/n_i + 1/n_j} < 2.886 * 0.69 = 1.99$$

$$|\bar{y}_{\cdot 1} - \bar{y}_{\cdot 2}| / \sqrt{1/n_1 + 1/n_2} = 2.95 > 1.99$$

$$|\bar{y}_{\cdot 1} - \bar{y}_{\cdot 3}| / \sqrt{1/n_1 + 1/n_3} = 3.11 > 1.99$$

$$|\bar{y}_{\cdot 1} - \bar{y}_{\cdot 4}| / \sqrt{1/n_1 + 1/n_4} = 2.24 > 1.99$$

$$|\bar{y}_{\cdot 2} - \bar{y}_{\cdot 3}| / \sqrt{1/n_2 + 1/n_3} = 5.36 > 1.99$$

$$|\bar{y}_{\cdot 2} - \bar{y}_{\cdot 4}| / \sqrt{1/n_2 + 1/n_4} = 4.90 > 1.99$$

$$|\bar{y}_{\cdot 3} - \bar{y}_{\cdot 4}| / \sqrt{1/n_3 + 1/n_4} = 1.17 < 1.99$$

All except 3 and 4 are significantly different