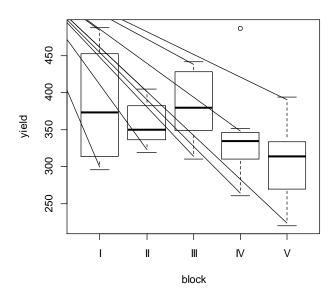
1.

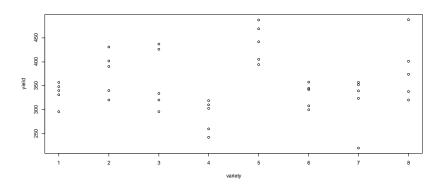
(a)

**Block Designs** 

(b)

Interaction exists between few couples





(c)

H0:There is no significant differentiating between samples.

H1: There is significant differentiating between samples.

F(7.32) at 0.05 level of significance=2.3127

F=5.1057 which is larger than 2.3127, so H0 is rejected-there is significant differentiating between samples.

### Analysis of Variance Table

```
Response: yield
     Df Sum Sq Mean Sq F value Pr(>F)
block 4 33396 8348.9 2.4777 0.06252.
variety 1 388 388.1 0.1152 0.73640
Residuals 34 114569 3369.7
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Call:
lm(formula = yield ~ ., data = oat)
Residuals:
 Min 1Q Median 3Q Max
-89.40 -37.96 -2.64 38.15 145.32
Coefficients:
      Estimate Std. Error t value Pr(>|t|)
(Intercept) 376.882 27.316 13.797 1.72e-15 ***
blockII -25.500 29.024 -0.879 0.386
blockIII 0.125 29.024 0.004 0.997
blockIV -42.000 29.024 -1.447 0.157
blockV -77.000 29.024 -2.653 0.012 *
variety 1.359 4.006 0.339 0.736
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Residual standard error: 58.05 on 34 degrees of freedom

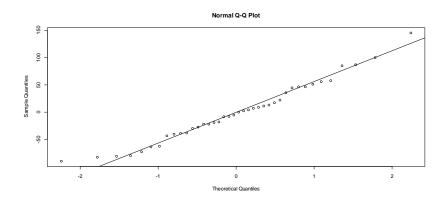
Multiple R-squared: 0.2277, Adjusted R-squared: 0.1142

F-statistic: 2.005 on 5 and 34 DF, p-value: 0.1029

(d)

Residual plots of the original full regression model:

Indicating that it does not satisfied the 3+1 assumptions on residuals



Try doing some

### transformation on the first model:

> oat.2<-lm(log(yield)~block+variety,oat)

> anova(oat.2)

Analysis of Variance Table

Response: log(yield)

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

block 4 0.29185 0.072963 2.7308 0.04504 \*

variety 1 0.00126 0.001262 0.0472 0.82925

Residuals 34 0.90843 0.026718

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

> summary(oat.2)

Call:

```
lm(formula = log(yield) ~ block + variety, data = oat)
```

#### Residuals:

```
Min 1Q Median 3Q Max
-0.32103 -0.09405 0.00364 0.10133 0.36969
```

### Coefficients:

Estimate Std. Error t value Pr(>|t|)

(Intercept) 5.919679 0.076917 76.962 <2e-16 \*\*\*

blockII -0.054564 0.081729 -0.668 0.5089

blockIII 0.010490 0.081729 0.128 0.8986

blockIV -0.113362 0.081729 -1.387 0.1745

blockV -0.222183 0.081729 -2.719 0.0102 \*

variety 0.002451 0.011280 0.217 0.8292

---

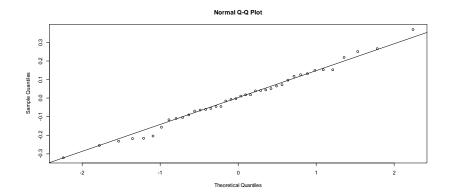
Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

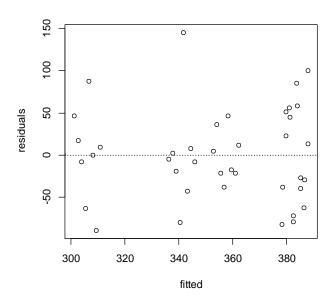
Residual standard error: 0.1635 on 34 degrees of freedom

Multiple R-squared: 0.2439, Adjusted R-squared: 0.1328

F-statistic: 2.194 on 5 and 34 DF, p-value: 0.0778

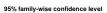
# QQ plot of transforemd model:

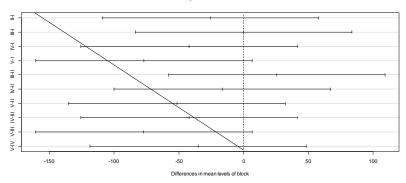




(e)

# Yes. it is necessary to perform multiple comparisons





2.

(a)

Latin square design

(b)

Linear model:

$$\begin{aligned} Y_{ijk} &= \mu + \alpha_i + \beta_j + \tau_k + \varepsilon_{ijk} \\ i, j, k &= 1, 2, 3, 4 \end{aligned}$$

(c)

Conducting null hypothesis for positions, run and fabric area. by doing so we will accept all of them at 5% degree of significance. Our null hypothesis: positions don't differ, run don't differ, and fabric area don't differ.

ANOVA table					
source	d.f	sum of squares	Mean SS	F	F3,9 0.05
position	3	6596	2198.667	0.239622	3.86
run	3	6404	2134.667	0.232647	

fabric Area	3	6420	2140	0.233228	
error	9	82580	9175.556		
Total	15	102000			

3.

(a)

Call:

Im(formula = Sales ~ ., data = tp)

#### Residuals:

1 2 3 4 5 6 7 8 9

1.4444 -1.8889 0.4444 0.4444 1.4444 -1.8889 -1.8889 0.4444 1.4444

Coefficients: (1 not defined because of singularities)

Estimate Std. Error t value Pr(>|t|)

(Intercept) -12.444 9.950 -1.251 0.3375

Week 5.333 1.210 4.409 0.0478 \*

StoreLocationS 22.667 4.361 5.198 0.0351 \*

StoreLocationU 49.000 7.650 6.405 0.0235 \*

StoreTypeDI 13.667 2.095 6.524 0.0227 \*

StoreTypeGR NA NA NA NA

DisplayB 14.000 2.419 5.787 0.0286 \*

DisplayC 7.667 2.419 3.169 0.0868.

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 2.963 on 2 degrees of freedom

Multiple R-squared: 0.9865, Adjusted R-squared: 0.9459

F-statistic: 24.29 on 6 and 2 DF, p-value: 0.04006

At least one type of each variable, from the table above, has a p-value of less than 0.10. Hence, for all three factors, there is a substantial difference between the groups. So, we wouldn't exclude a variable if we were using the backward stepwise regression with alpha to leave 0.1.

(b)

From the data, it is observed that store type DE, location U and display B together appear to maximize sales value:67 toothpastes.

(c)

We are not able to estimate the probability that 80 or more toothpaste will be sold during a week based on the given information.

(d)

The model ran into perfect multicollinearity. "Alias" refers to the variables that are linearly dependent on others (i.e. cause perfect multicollinearity).

> vif(tp.0)

Error in vif.default(tp.0): there are aliased coefficients in the model

> alias(tp.0)

Model:

Sales ~ Week + Locat + Type + Display

Complete:

(Intercept) Week LocatS LocatU TypeDI DisplayB DisplayC

TypeGR 9/2 -1/2 -3/2 -3 -1/2 0 0