

# Untitled

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**3a**

```
y <- c(46.5,48.7,46.3,44.7,
       45.9,49,47.1,43,
       49.8,50.1,48.9,51,
       46.1,48.5,48.2,48.1,
       44.1,45.2,50.3,48.6)
x <- c(13,12,15,16,
       14,10,14,15,
       12,11,11,10,
       12,12,11,12,
       12,14,10,11)
form=as.factor(rep(1:4,5))
treat=as.factor(sort(rep(1:5,4)))
df1 = data.frame(treat,form,x,y)

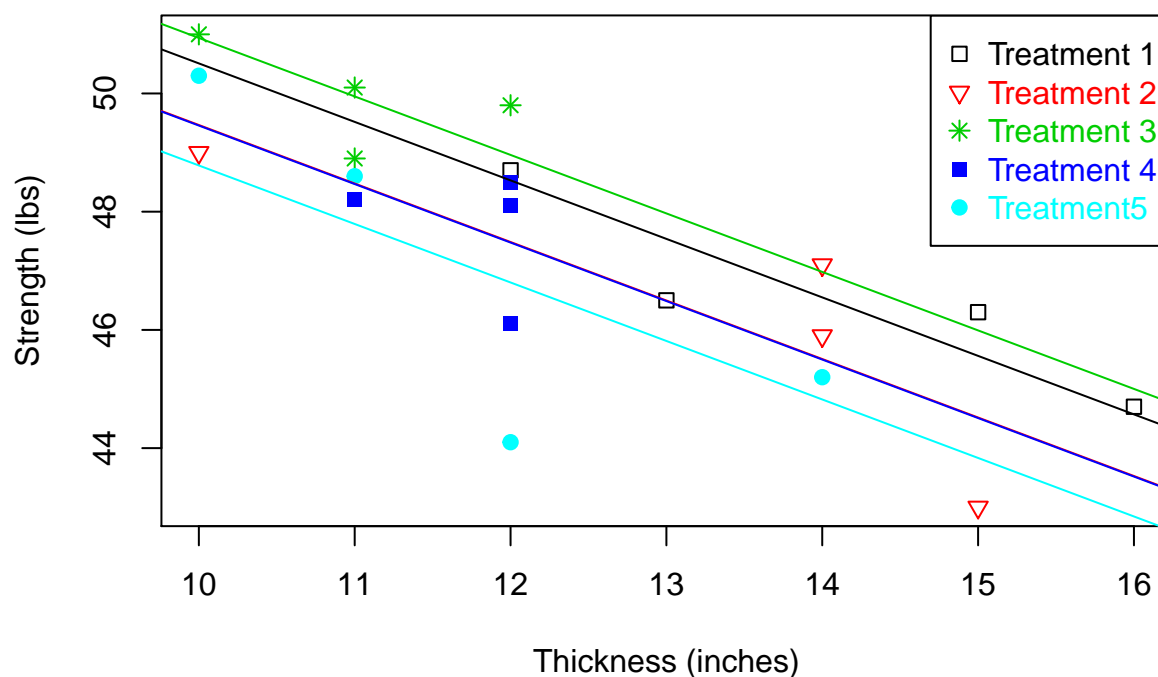
lm.full <- lm(y~x+treat,data=df1)
beta.hat <- coef(lm.full)[2]

plot(df1$x,df1$y,col=sort(rep(1:5,4)),
     pch=sort(rep(c(0,6,8,15,19),4)),xlab="Thickness (inches)",
     ylab="Strength (lbs)",main="Glue Formulation")
legend("topright",legend=c("Treatment 1","Treatment 2","Treatment 3","Treatment 4", "Treatment5"),
     pch=c(0,6,8,15,19),col=1:5,text.col=1:5)

y.i.bar <- as.numeric(by(df1[,4],df1[,1],mean))
x.i.bar <- as.numeric(by(df1[,3],df1[,1],mean))

abline(y.i.bar[1]+beta.hat*(-x.i.bar[1]),beta.hat,col=1)
abline(y.i.bar[2]+beta.hat*(-x.i.bar[2]),beta.hat,col=2)
abline(y.i.bar[3]+beta.hat*(-x.i.bar[3]),beta.hat,col=3)
abline(y.i.bar[4]+beta.hat*(-x.i.bar[4]),beta.hat,col=4)
abline(y.i.bar[5]+beta.hat*(-x.i.bar[5]),beta.hat,col=5)
```

## Glue Formulation



Upon visual inspection, it does appear that equal slopes is an appropriate assumption.

3b

```
library(car)
```

```
## Warning: package 'car' was built under R version 3.4.4
```

```
## Loading required package: carData
```

```
## Warning: package 'carData' was built under R version 3.4.4
```

```
out.lm <- lm(y~treat*x,data=df1,contrasts=list(treat=contr.sum))
```

```
Anova(out.lm,type="III")
```

```
## Anova Table (Type III tests)
```

```
##
```

```
## Response: y
```

```
##          Sum Sq Df F value    Pr(>F)
## (Intercept) 290.145  1 139.9157 3.344e-07 ***
## treat         1.059  4   0.1277   0.9689
## x             8.996  1   4.3380   0.0639 .
## treat:x       1.786  4   0.2154   0.9239
## Residuals    20.737 10
```

```
## ---
```

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

testing against  $\beta_1 = \beta_2 = \beta_3 = \beta_4 = \beta_5$  yields  $F^* = .2154$  distributed  $F_{4,10}$  for a p-value of .9239. So we conclude that the test is not significant, and claim the equal slopes assumption appropriate.

### 3c

```
out.aov <- aov(y~x+treat,data=df1)
Anova(out.aov, type="III")

## Anova Table (Type III tests)
##
## Response: y
##           Sum Sq Df F value    Pr(>F)
## (Intercept) 645.02  1 400.9220 1.059e-11 ***
## x           35.50  1  22.0681 0.0003427 ***
## treat       12.24  4   1.9022 0.1660140
## Residuals   22.52 14
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

If we test the equality of the adjusted treatment means we have  $F^* = 1.9022$  which is distributed  $F_{4,14}$  with a pvalue of .166, which is not significant. We then conclude that the adjusted treatment means are not significantly different.

### 3d

```
out.aov <- aov(y~x+treat,data=df1)
Anova(out.aov, type="III")

## Anova Table (Type III tests)
##
## Response: y
##           Sum Sq Df F value    Pr(>F)
## (Intercept) 645.02  1 400.9220 1.059e-11 ***
## x           35.50  1  22.0681 0.0003427 ***
## treat       12.24  4   1.9022 0.1660140
## Residuals   22.52 14
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

If we test the significance of the reduction in error variance due to the covariate we have  $F^* = 22.0681$  which is distributed  $F_{1,14}$  with a pvalue of .0003427, which is significant. We then conclude that the addition of the covariate has reduced error variability.

### 3e

```
E.xx <- anova(lm(x~treat,data=df1))$"Sum Sq"[2]

MSE <- anova(lm.full)$"Mean Sq"[3]
x.bar <- mean(df1$x)
```

```

y.i.adj <- y.i.bar-beta.hat*(x.i.bar-x.bar)

s.y.i <- sqrt(MSE*(1/4+(x.i.bar-x.bar)^2/E.xx))

#s.y.12 <- sqrt(MSE*(1/4+1/4+(x.i.bar[1]-x.i.bar[2])^2/E.xx))
#s.y.13 <- sqrt(MSE*(1/4+1/4+(x.i.bar[1]-x.i.bar[3])^2/E.xx))
#s.y.14 <- sqrt(MSE*(1/4+1/4+(x.i.bar[1]-x.i.bar[4])^2/E.xx))
#s.y.15 <- sqrt(MSE*(1/4+1/4+(x.i.bar[1]-x.i.bar[5])^2/E.xx))

#s.y.23 <- sqrt(MSE*(1/4+1/4+(x.i.bar[2]-x.i.bar[3])^2/E.xx))
#s.y.24 <- sqrt(MSE*(1/4+1/4+(x.i.bar[2]-x.i.bar[4])^2/E.xx))
#s.y.25 <- sqrt(MSE*(1/4+1/4+(x.i.bar[2]-x.i.bar[5])^2/E.xx))

#s.y.34 <- sqrt(MSE*(1/4+1/4+(x.i.bar[3]-x.i.bar[4])^2/E.xx))
#s.y.35 <- sqrt(MSE*(1/4+1/4+(x.i.bar[3]-x.i.bar[5])^2/E.xx))

#s.y.45 <- sqrt(MSE*(1/4+1/4+(x.i.bar[4]-x.i.bar[5])^2/E.xx))

lower.bound=y.i.adj-pt(.975,df=3)*s.y.i
upper.bound = y.i.adj+pt(.975,df=3)*s.y.i

library(xtable)

## Warning: package 'xtable' was built under R version 3.4.3
tab.1<- data.frame(y.i.adj,s.y.i,lower.bound,upper.bound)
xtable(tab.1)

```

% latex table generated in R 3.4.0 by xtable 1.8-2 package % Thu Nov 29 01:57:09 2018

|   | y.i.adj | s.y.i | lower.bound | upper.bound |
|---|---------|-------|-------------|-------------|
| 1 | 48.18   | 0.72  | 47.60       | 48.76       |
| 2 | 47.14   | 0.66  | 46.61       | 47.67       |
| 3 | 48.61   | 0.70  | 48.06       | 49.17       |
| 4 | 47.13   | 0.65  | 46.61       | 47.65       |
| 5 | 46.46   | 0.65  | 45.94       | 46.97       |