

1 This is Problem 3 of Faraway (2006), Chapter 8.

The eggprod dataset concerns an experiment where six pullets were placed into each of 12 pens. Four blocks were formed from groups of three pens based on location. Three treatments were applied. The number of eggs produced was recorded.

```
library(faraway)
data(eggprod)
```

(a) Fit a model for the number of eggs produced with the treatments as fixed effects and the blocks as random effects. Describe the estimated differences between the treatments.

pullet: a young hen, especially one less than one year old.

In the following model, y_{ij} is the number of eggs produced by the i th treatment on the j th block, with $i = 1, 2$, and 3 and $j = 1, 2, 3$, and 4 .

$$E(y_{ij}) = \beta_i, \text{ and } y_{ij} = \beta_i + u_j + \epsilon_{ij}$$

We assume that

- $u_j \stackrel{iid}{\sim} N(0, \sigma_u^2)$
- the random error $\epsilon_{ij} \stackrel{iid}{\sim} N(0, \sigma_\epsilon^2)$

An overview of the dataset was performed and plots and descriptions are included in the appendix. The `summary()` of our `lme(eggs ~ 0+treat, random = ~1|block, data=eggprod, method=c("REML"))` model of a means model terms gives estimates for the means of treatments:

Treatment	Estimated Mean Eggs	Comparison	Difference
E	349.00	E-F	6.25 eggs
F	342.75	F-O	36.25 eggs
O	306.50	E-O	42.95 eggs

That is, Treatment E's mean is 6.25 eggs greater than Treatment F. Treatment E's mean is 42.95 eggs greater than Treatment O. Finally, Treatment F's mean is 36.25 eggs greater than Treatment O.

(b) Test for the significance of the treatment.

In order to isolate the effect of treatment from the baseline means, an intercept was added to the model. The new model may be written:

$$E(y_{ij}) = \mu + \beta_i, \text{ and } y_{ij} = \mu + \beta_i + u_j + \epsilon_{ij}$$

The assumptions made in part a) hold here.

The model matrix is now given by:

$$\mathbf{X} = \begin{pmatrix} 1 & 0 & 1 \\ 1 & 0 & 1 \\ 1 & 0 & 1 \\ 1 & 0 & 1 \\ 1 & 0 & 0 \\ 1 & 0 & 0 \\ 1 & 0 & 0 \\ 1 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & 0 \end{pmatrix}$$

We will use the likelihood ratio test here, which tells us that given an MLE estimator $\hat{\theta}$ the quantity $-2(l(\hat{\theta}) - l(\theta))$ follows a χ^2 distribution with degrees of freedom equal to the dimensions of θ . The above “l” represents the log-likelihood function.

We compare the full model to a model with an intercept term only. Therefore, our null hypothesis is that the differences in treatment represented by β_2 (Treatment F - Treatment E) and β_3 (Treatment O - Treatment E) are zero. Under the null hypothesis, these terms are zero and the log-likelihood of the true θ our data would given by a model with only an intercept term.

```
eggs.lme2 <- lme(eggs ~ 1 + treat,
  random = ~1|block, data=eggprod, method=c("ML"))
eggs.lme3 <- lme(eggs ~ 1, random = ~1|block, data=eggprod, method=c("ML"))
LL2 <- logLik(eggs.lme2)
LL3 <- logLik(eggs.lme3)
W2 <- as.numeric(-2*(LL3-LL2))
W2
test <- qchisq(.95, 2)
```

In the log-likelihood test performed above, a value of 8.4245 was obtained for $-2\log\lambda$. This compares to a .95 upper quantile of 5.9915 for the χ^2 distribution with $df=2$. Since our test value is more extreme, it suggests the null hypothesis should be rejected, that the treatment effects are significant.

2 This is Problem 4 of Faraway (2006), Chapter 8.

Data on the cutoff times of lawnmowers may be found in the dataset lawn. 3 machines were randomly selected from those produced by manufacturers A and B. Each machine was tested twice at low speed and high speed.

```
> library(faraway)
> data(lawn)
```

- (a) **Fit a mixed effects model with manufacturer and speed as main effects along with their interaction and machine nested in manufacturer as random effects. Write down the formula for the model. In the summary output for the model, you will find that fixed manufacturer effect has zero degrees of freedom. Explain why this is so (check your model formula).**

We write the model: $y_{ijk} = \alpha_i + \beta_j + \gamma_{ij} + u_{ik} + \epsilon_{ijk}$ where y_{ijk} is the cut-off time of the lawn mower from the i th ($i=1,2$) manufacturer at the j th speed ($j=1,2$) of the k th type of machine of that manufacturer ($k=1,2,3$).

We assume that

- $u_{ik} \stackrel{iid}{\sim} N(0, \sigma_u^2)$
- the random error $\epsilon_{ijk} \stackrel{iid}{\sim} N(0, \sigma_e^2)$

We run the following in R, using * to indicate an interaction term and random=~1|manufact/machine for the random effects term. There, the slash indicates that machine is nested under manufact. The 1 indicates our assumption of constant/homogeneous variance:

```
lawn.lme <- lme(time ~ manufact + speed + manufact*speed,
               random=~1|manufact/machine, data=lawn, method=c("REML"))

summary(lawn.lme)
```

The complete output of summary(lawn.lme) is included in the appendix. For readability, I am showing the Fixed effects table illustrating that manufactB has 0 degrees of freedom (DF).

```
Fixed effects: time ~ manufact + speed + manufact * speed
              Value Std.Error DF   t-value p-value
(Intercept)  270.50000  12.200845 16  22.170595  0.0000
manufactB    -21.83333  17.254601  0  -1.265363   NaN
speedL       -60.33333   6.640831 16  -9.085208  0.0000
manufactB:speedL  2.66667   9.391554 16   0.283943  0.7801
```

The manufacturer effect has zero degrees of freedom because this factor is completely determined by the machine. That is, given a machine of type “m5” for example, the manufacturer is B. The information contained in the manufacturer factor (A or B) is already contained in the machine factor (m1/m2/m3 or m4/m5/m6).

(b) Show why the manufacturer term may be removed from the fixed effect part of the model.

In order to determine if the term may be removed from the model, we compare the models by ANOVA with and without the manufact factor and do not obtain a significant result. To validate this result, we perform likelihood ratio testing.

```
lawn.lme <- lme(time ~ manufact + speed + manufact*speed,
               random=~1|manufact/machine, data=lawn, method=c("ML"))

lawn.lme2 <- lme(time ~ speed,
                 random=~1|manufact/machine, data=lawn, method=c("ML"))

anova(lawn.lme, lawn.lme2)
```

	Model	df	AIC	BIC	logLik	Test	L.Ratio	p-value
lawn.lme	1	7	204.9064	213.1528	-95.45319			
lawn.lme2	2	5	204.5690	210.4593	-97.28451	1 vs 2	3.662641	0.1602

To validate the ANOVA result, we perform a log-likelihood test.

```
LL <- logLik(lawn.lme)
LL2 <- logLik(lawn.lme2)
W2 <- as.numeric(-2*(LL2-LL))
test2 <- qchisq(.95, 2)
```

The value of $-2\log\lambda$ obtained in our test was 3.6626, which is less than the .95 quantile of the χ^2 distribution with $df=2$, 5.9915. This means the null is more likely and shows that we should accept the null hypothesis that the the main effect manufacturer term and interaction parameters are 0.

(c) Determine if the manufacturer term can be removed from the random part of the model.

(c).1 Significance of contribution of entire random term

We first test the removal of the entire random term (Machine type by Manufacturer) from the model.

```
lawn.lm4 <- lm(time ~ speed, data=lawn)

lawn.lme2 <- lme(time ~ speed,
                 random=~1|manufact/machine, data=lawn, method=c("ML"))

anova(lawn.lme2,lawn.lm4)

[1] 3.6626
[1] 5.9915
```

	Model	df	AIC	BIC	logLik	Test	L.Ratio	p-value
	lawn.lme2	1	5	204.5690	210.4593	-97.28451		
	lawn.lm4	2	3	211.9854	215.5196	-102.99271	1 vs 2	11.41639 0.0033

```
LL <- logLik(lawn.lm4)
LL2 <- logLik(lawn.lme2)
W2 <- as.numeric(-2*(LL-LL2))
test2 <- qchisq(.95,5)
```

Log liklihood testing for the removal of the entire random term supported the significant ANOVA conclusion ($p = 0.0033$). A $-2\log\lambda$ value of 11.4164 was obtained, which is larger than the reference value of the .95 quantile of the χ^2 distribution with $df=5$ (5 because there are six machines, minus one for overall), 11.0705.

(c).2 Removal of nesting of manufact from random term.

```
lawn.lme2 <- lme(time ~ speed,
                 random=~1|manufact/machine, data=lawn, method=c("ML"))
lawn.lme3 <- lme(time ~ speed,
                 random=~1|machine, data=lawn, method=c("ML"))

anova(lawn.lme2,lawn.lme3)

[1] 11.4164
[1] 11.0705
```

	Model	df	AIC	BIC	logLik	Test	L.Ratio	p-value
	lawn.lme2	1	5	204.5690	210.4593	-97.28451		
	lawn.lme3	2	4	202.7968	207.5090	-97.39840	1 vs 2	0.2277637 0.6332

```
LL3 <- logLik(lawn.lme3)
LL2 <- logLik(lawn.lme2)
W2 <- as.numeric(-2*(LL3-LL2))
test2 <- qchisq(.95,1)
```

Log likelihood testing for the removal of the nested manufact component of the random term supported the NOT significant ANOVA conclusion ($p = 0.6332$). A $-2\log\lambda$ value of 0.2278 was obtained, which is much smaller than the reference value of the .95 quantile of the χ^2 distribution with $df=1$ (1 because one parameter is being tested), 3.8415. So, the manufact nesting can be removed from the random effects term.

3 Appendix: Tangled R Code

```
library(MASS); library(xtable); library(nlme)
lvector <- function(x, dig = 2, dsply=rep("f",ncol(x)+1)) {
  x <- xtable(x, align=rep("",ncol(x)+1),display=dsply,digits=dig) # We repeat empty string 6 times
  print(x, floating=FALSE, tabular.environment="pmatrix",
        hline.after=NULL, include.rownames=FALSE, include.colnames=FALSE)
}

library(faraway)
data(eggprod)

pdf(file="eggprod1.pdf")
library(faraway)
data(eggprod)
attach(eggprod)

#First take a look at the data.
par(mfrow=c(2,2))
plot(block,eggs,data=eggprod,main="Boxplot of Egg Production By Block")
plot(treat,eggs,data=eggprod,main="Boxplot of Egg Production By Treatment")
interaction.plot(block,treat,eggs, data=eggprod,main="Interaction Plot of Egg Production with Block")
interaction.plot(treat,block,eggs, data=eggprod,main="Interaction Plot of Egg Production with Treatment")

#Fit the model.
options(contrasts=c("contr.treatment","contr.poly"))
library(nlme)

eggs.lme <- lme(eggs ~ 0+treat, random = ~1|block, data=eggprod, method=c("REML"))

library(lme4)
eggs.lme2 <- lmer(eggs ~ treat+(1|block), data=eggprod)

summary(eggs.lme)

dev.off()

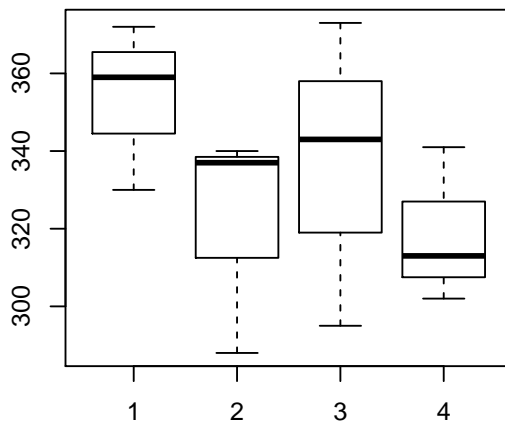
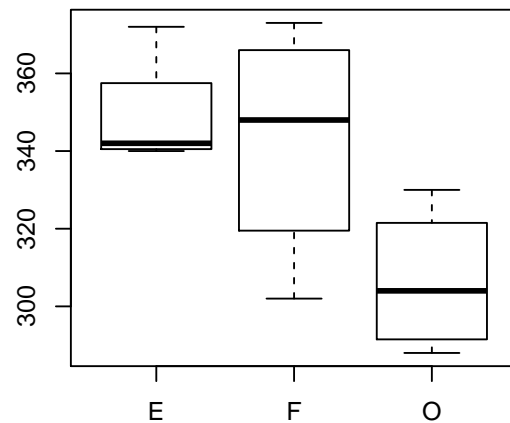
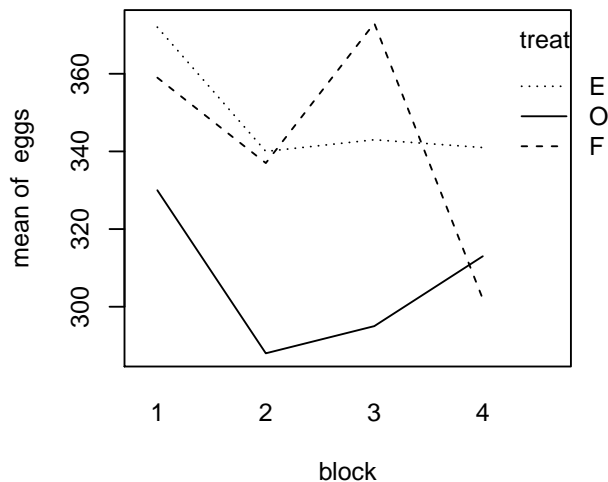
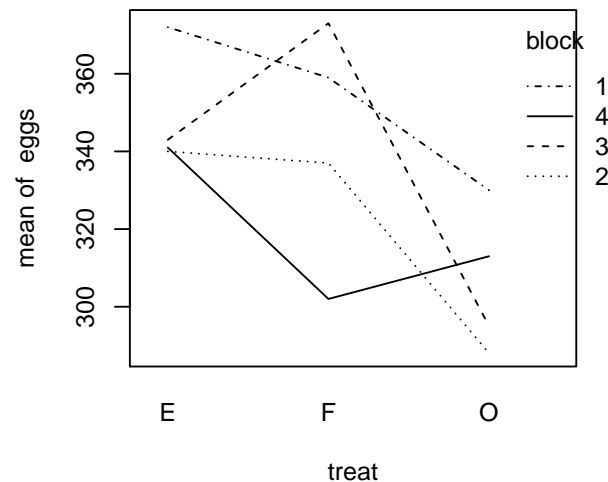
> library(faraway)
> data(lawn)
```

4 Appendix: Additional and preliminary analysis of eggprod

Looking at the actual eggprod dataset proved useful (it is relatively small):

O	1	330
O	2	288
O	3	295
O	4	313
E	1	372
E	2	340
E	3	343
E	4	341
F	1	359
F	2	337
F	3	373
F	4	302

Additionally, boxplots Egg Production as a function of Treatment and Block were examined, as well as interaction plots. The boxplot of Egg Production by Treatment shows that while Treatments E and F are largely overlapping, Treatment O has a much lower mean and takes lower values, completely non-overlapping with Treatment E.

Boxplot of Egg Production By Block**Boxplot of Egg Production By Treatment****Interaction Plot of Egg Production with Block as X Factor, Treatment as Trace Factor****Interaction Plot of Egg Production with Treatment as X Factor, Block as Trace Factor**

5 Appendix: Additional and preliminary analysis of lawn

(a) Preliminary analysis

The lawn data were plotted as boxplots of cut-off times versus machine, speed, and manufacturer. The most striking observation was that the cut-off times for speed “H” were much higher than speed “L”. In fact, the two box plots were non-overlapping. Means in the By Machine boxplot appeared to vary, but all boxplots overlapped.

Interaction plots of cut-off times looking for an interaction between manufacturer and speed show absolutely parallel lines, suggesting there is no interaction.

(b) Complete output of summary(lawn.lme)

```
Linear mixed-effects model fit by REML
Data: lawn
```

AIC	BIC	logLik
182.3651	189.3352	-84.18254

Random effects:

Formula: ~1 | manufact
(Intercept)

StdDev: 8.854442

Formula: ~1 | machine %in% manufact
(Intercept) Residual

StdDev: 12.05104 11.50226

Fixed effects: time ~ manufact + speed + manufact * speed

	Value	Std.Error	DF	t-value	p-value
(Intercept)	270.50000	12.200845	16	22.170595	0.0000
manufactB	-21.83333	17.254601	0	-1.265363	NaN
speedL	-60.33333	6.640831	16	-9.085208	0.0000
manufactB:speedL	2.66667	9.391554	16	0.283943	0.7801

Correlation:

	(Intr)	mnfctB	speedL
manufactB	-0.707		
speedL	-0.272	0.192	
manufactB:speedL	0.192	-0.272	-0.707

Standardized Within-Group Residuals:

Min	Q1	Med	Q3	Max
-1.0908529	-0.6739824	-0.1291112	0.6660725	1.5405034

Number of Observations: 24

Number of Groups:

manufact	machine	%in%	manufact
2			6

Warning message:

In pt(-abs(tTable[, "t-value"]), tTable[, "DF"]) : NaNs produced

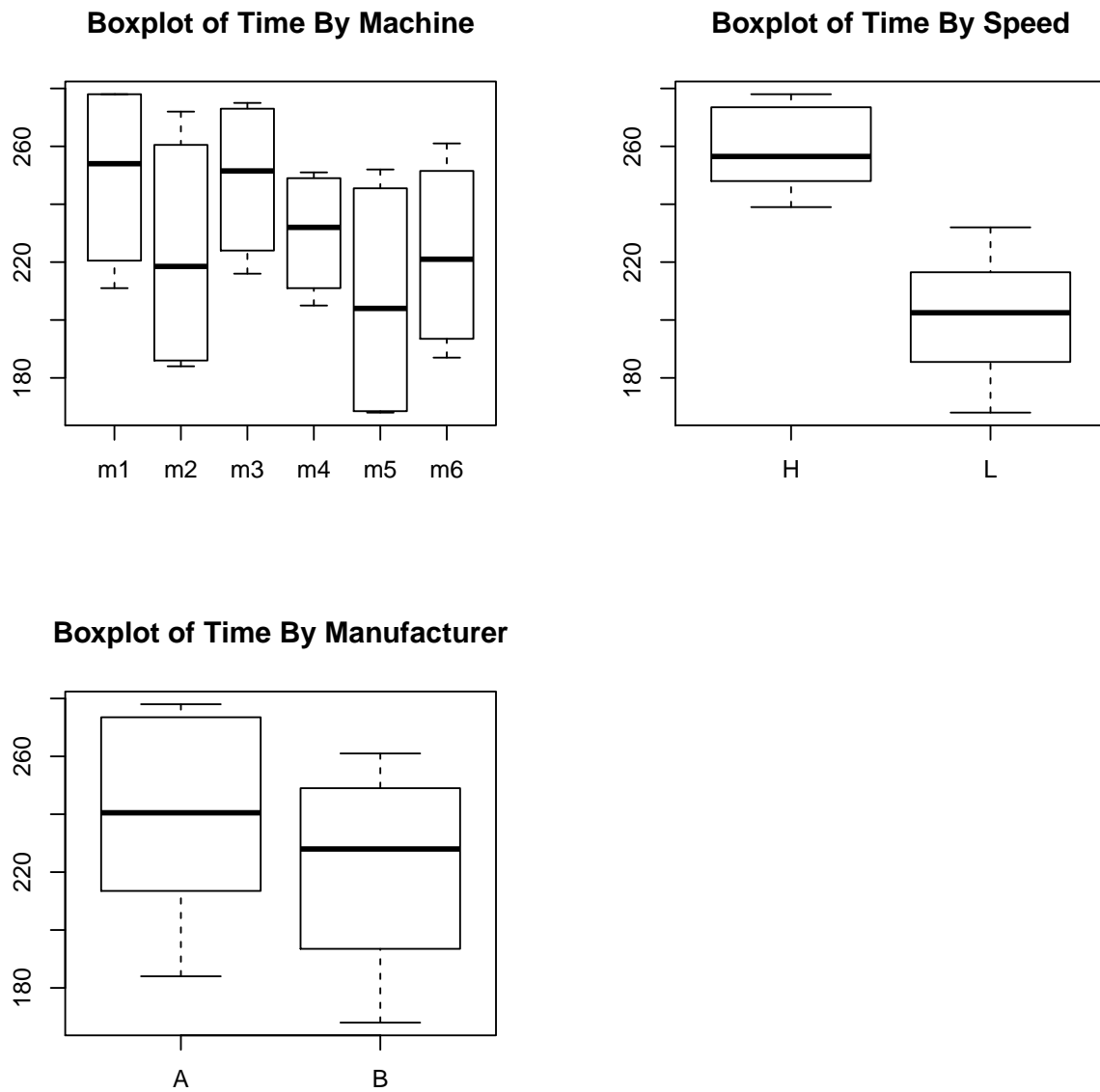


Figure 1: Boxplot and Interaction plots for eggprod Boxplots of the lawn dataset

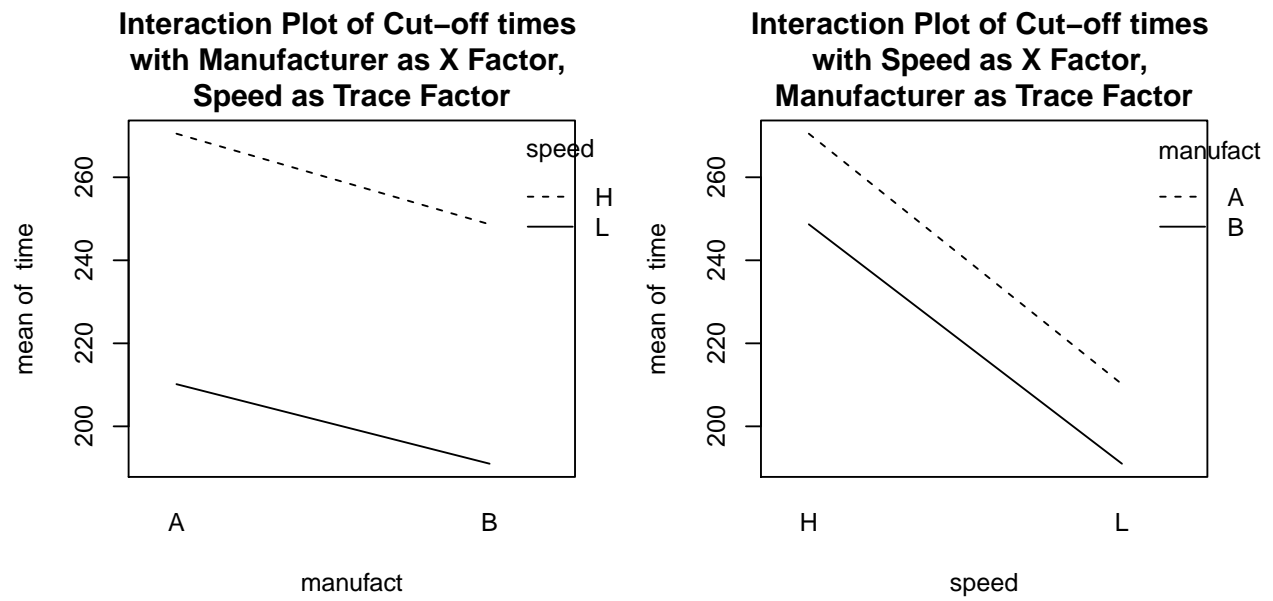


Figure 2: Interaction plots for Lawn