Math650 Homework 8

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2006-10-19

1 Question 1

The four different contrast functions result different contrast matricies. Details see page 146 Book by Venables et al, 2002.

However, if we only look at the identifiability contraint, contr.helmert, contr.sum and contr.poly are same, which is $1^T \alpha = 0$ (the sum of coefficients equals to zero). contr.treatment, which is the default option in R, sets the first coefficient to 0 and the rest of them correspond to level 1.

```
> options(contrasts=c("contr.sum", "contr.sum"))
> test_contr2()
  [,1] [,2]
0
     1
     0
          1
    -1
> options(contrasts = c("contr.treatment", "contr.poly"))
> test_contr2()
  1 2
0 0 0
1 1 0
2 0 1
> options(contrasts = c("contr.helmert", "contr.poly"))
> test_contr2()
  [,1] [,2]
0
    -1
         -1
     1
         -1
1
> options(contrasts = c("contr.poly", "contr.poly"))
> test_contr2()
             .L
                         .Q
0 -7.071068e-01 0.4082483
1 -9.073264e-17 -0.8164966
2 7.071068e-01 0.4082483
```

2 Question 2

Based on the analysis of the identifiability contraint above, different contrasts play between *intercept* and factor-involved coefficients. Contrasts with same

identifiability contraint gave similar (almost identical, but due to float differences) results.

2.1 contr.treatment

This is the default.

Call:

lm(formula = LOGIT ~ LOGDURATION + BEE + LOGDURATION * BEE, data = data)

Residuals:

Min 1Q Median 3Q Max -1.3804 -0.3699 0.0307 0.4552 1.1611

Coefficients:

	${\tt Estimate}$	Std. Error	t value	Pr(> t)	
(Intercept)	-3.0390	0.5115	-5.941	4.45e-07	***
LOGDURATION	1.0121	0.1902	5.321	3.52e-06	***
BEEWORKER	1.3770	0.8722	1.579	0.122	
LOGDURATION: BEEWORKER	-0.2709	0.2817	-0.962	0.342	

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

Residual standard error: 0.6525 on 43 degrees of freedom Multiple R-Squared: 0.6151, Adjusted R-squared: 0.5882 F-statistic: 22.9 on 3 and 43 DF, p-value: 5.151e-09

2.2 contr.helmert

Call:

lm(formula = LOGIT ~ LOGDURATION + BEE + LOGDURATION * BEE, data = data)

Residuals:

Min 1Q Median 3Q Max -1.3804 -0.3699 0.0307 0.4552 1.1611

Coefficients:

```
Estimate Std. Error t value Pr(>|t|)
                             0.4361 -5.390 2.80e-06 ***
(Intercept)
                 -2.3505
LOGDURATION
                   0.8766
                             0.1408
                                      6.224 1.72e-07 ***
                  0.6885
                             0.4361
BEE1
                                      1.579
                                               0.122
LOGDURATION:BEE1 -0.1354
                             0.1408 -0.962
                                               0.342
```

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

Residual standard error: 0.6525 on 43 degrees of freedom Multiple R-Squared: 0.6151, Adjusted R-squared: 0.5882 F-statistic: 22.9 on 3 and 43 DF, p-value: 5.151e-09

2.3 contr.sum

```
Call:
```

lm(formula = LOGIT ~ LOGDURATION + BEE + LOGDURATION * BEE, data = data)

Residuals:

Min 1Q Median 3Q Max -1.3804 -0.3699 0.0307 0.4552 1.1611

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	-2.3505	0.4361	-5.390	2.80e-06 ***
LOGDURATION	0.8766	0.1408	6.224	1.72e-07 ***
BEE1	-0.6885	0.4361	-1.579	0.122
LOGDURATION: BEE1	0.1354	0.1408	0.962	0.342

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

Residual standard error: 0.6525 on 43 degrees of freedom Multiple R-Squared: 0.6151, Adjusted R-squared: 0.5882 F-statistic: 22.9 on 3 and 43 DF, p-value: 5.151e-09

2.4 contr.poly

Call:

lm(formula = LOGIT ~ LOGDURATION + BEE + LOGDURATION * BEE, data = data)

Residuals:

Min 1Q Median 3Q Max -1.3804 -0.3699 0.0307 0.4552 1.1611

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	-2.3505	0.4361	-5.390	2.80e-06	***
LOGDURATION	0.8766	0.1408	6.224	1.72e-07	***
BEE.L	0.9737	0.6167	1.579	0.122	
LOGDURATION: BEE.L	-0.1916	0.1992	-0.962	0.342	

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

Residual standard error: 0.6525 on 43 degrees of freedom Multiple R-Squared: 0.6151, Adjusted R-squared: 0.5882 F-statistic: 22.9 on 3 and 43 DF, p-value: 5.151e-09

Most differences happen to the *intercept*, coefficients of *BEE* and *LOG-DURATION:BEE*. *contr.sum* and *contr.helmert* yielded the identical results. *contr.poly* doesn't fit in this scenario very well as it's catered towards ordered factor variables.

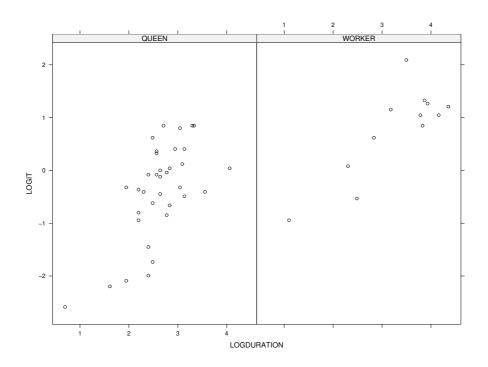


Figure 1: Scatterplot of LOGIT vs LOGDURATION

3 Question 3

I tried three different breakdown points and see how the regression line changes and residual plots et al.

data_type	α (breakdown point)	intercept	coeff LOGDURATION	scale estimates
queen	0.9	-3.039	1.012	0.6817
worker	0.9	-1.6753	0.7129	0.4338
queen	0.8	-2.9255	0.9844	0.734
worker	0.8	-1.6753	0.7129	0.4464
queen	0.6	-3.303	1.142	0.7505
worker	0.6	-1.6753	0.7129	0.4504

Figures 2 and 3 are for $\alpha = 0.9$. Figures 4 and 5 are for $\alpha = 0.8$. Figures 6 and 7 are for $\alpha = 0.6$. The WORKER's data is pretty robust while QUEEN's data shows quite fluctuation with the change of α .

4 Appendix

```
#test from Venables2002 page 145
test_contr = function()
{
dat = data.frame(a=factor(rep(1:3,3)), y=rnorm(9, rep(2:4, 3), 0.1))
obj = lm(y~a, dat)
alf.star = coef(obj)
print(alf.star)
```

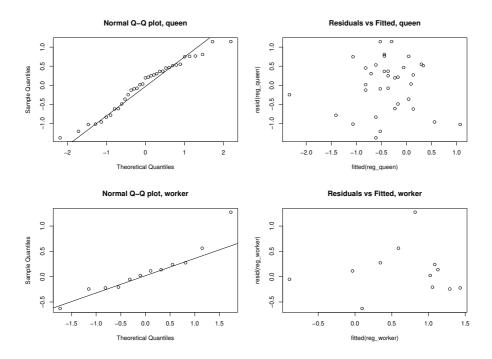


Figure 2: Linear regression plots without interaction and full data, alpha=0.9

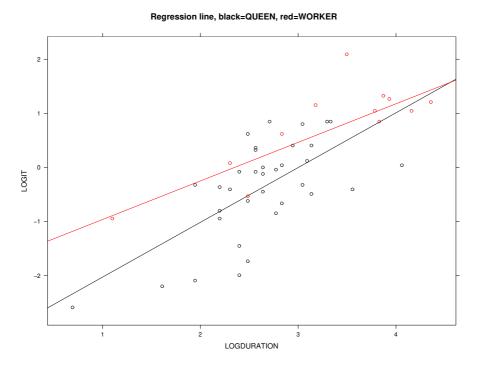


Figure 3: Regression line, black is QUEEN, red is WORKER, alpha=0.9

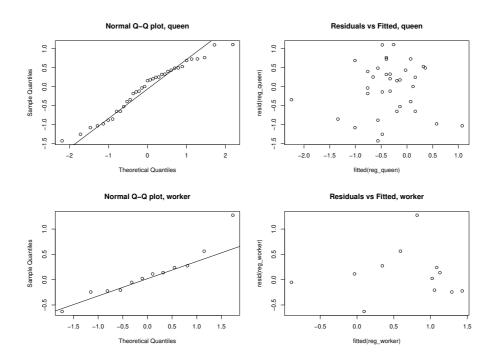


Figure 4: Linear regression plots without interaction and full data, alpha $\!=\!0.8$

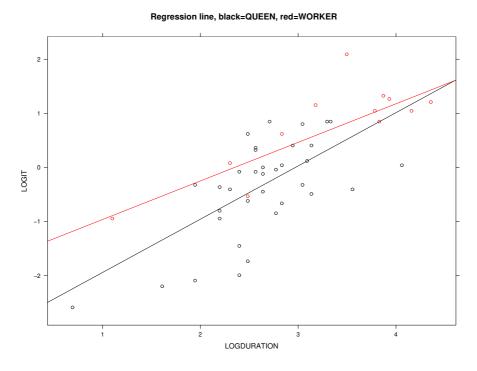


Figure 5: Regression line, black is QUEEN, red is WORKER, alpha=0.8

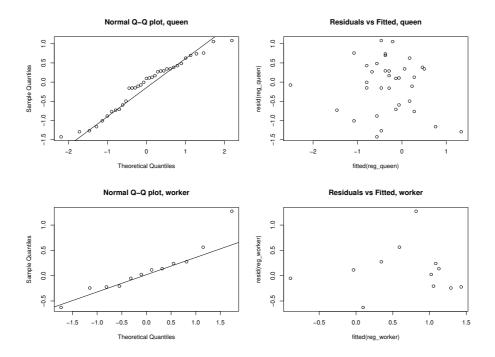


Figure 6: Linear regression plots without interaction and full data, alpha=0.6

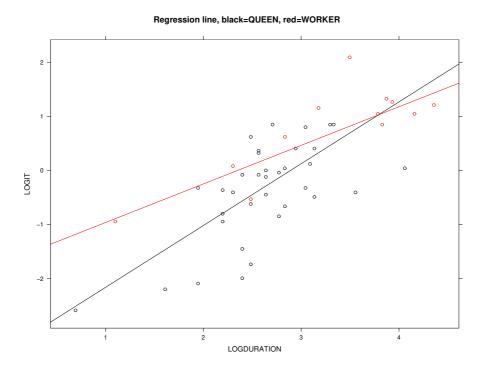


Figure 7: Regression line, black is QUEEN, red is WORKER, alpha=0.6

```
ca = contrasts(dat$a)
cat("contrast matrix:\n")
print(ca)
drop(ca %*% alf.star[-1])
dummy.coef(obj)
options(contrasts = c("contr.treatment", "contr.poly"))
test_contr()
options(contrasts = c("contr.sum", "contr.poly"))
test_contr()
options(contrasts = c("contr.helmert", "contr.poly"))
test_contr()
options(contrasts = c("contr.poly", "contr.poly"))
test_contr()
#question 1
test_contr2 = function()
NN=factor(levels=c(0,1,2));
contrasts(NN)
options(contrasts=c("contr.sum", "contr.sum"))
test_contr2()
options(contrasts = c("contr.treatment", "contr.poly"))
test_contr2()
options(contrasts = c("contr.helmert", "contr.poly"))
test_contr2()
options(contrasts = c("contr.poly", "contr.poly"))
test_contr2()
#question 2, just try the options above and run code from hw7
#question 3
library(rrcov)
#?lstReg
library(lattice)
data1 = read.csv("/usr/local/doc/statistical_sleuth/ASCII/ex0328.csv")
LOGIT = log(data1$REMOVED/(1-data1$REMOVED))
LOGDURATION = log(data1$DURATION)
data2 = cbind(data1, LOGIT, LOGDURATION)
histogram(~LOGIT|BEE, data=data2)
histogram(~LOGDURATION|BEE, data=data2)
postscript('~/script/test/math650/figures/math650_hw8_fig1.eps')
xyplot(LOGIT~LOGDURATION|BEE, data=data2)
dev.off()
linear_model_no_intr = function(data, fig_fname, alpha_value=0.8)
data_queen = data[data$BEE=="QUEEN",]
```

```
data_worker = data[data$BEE=="WORKER",]
reg_queen = ltsReg(data_queen$LOGDURATION, data_queen$LOGIT, alpha=alpha_value)
reg_worker = ltsReg(data_worker$LOGDURATION, data_worker$LOGIT, alpha=alpha_value)
print(reg_queen)
print(reg_worker)
postscript(fig_fname)
opar \leftarrow par(mfrow = c(2,2), oma = c(0, 0, 1.1, 0))
#plot(reg_queen)
qqnorm(resid(reg_queen), main='Normal Q-Q plot, queen')
qqline(resid(reg_queen))
plot(fitted(reg_queen), resid(reg_queen), main='Residuals vs Fitted, queen')
#plot(reg_worker)
qqnorm(resid(reg_worker), main='Normal Q-Q plot, worker')
qqline(resid(reg_worker))
plot(fitted(reg_worker), resid(reg_worker), main='Residuals vs Fitted, worker')
par(opar)
dev.off()
reg = list(reg_queen=reg_queen, reg_worker=reg_worker)
return(reg)
draw_data_no_intr = function(reg, data)
intercept_1 = coef(reg$reg_queen)[1]
slope_1 = coef(reg$reg_queen)[2]
intercept_2 = coef(reg$reg_worker)[1]
slope_2 = coef(reg$reg_worker)[2]
xyplot(LOGIT~LOGDURATION, data=data, main='Regression line, black=QUEEN, red=WORKER', auto
  panel=function(x,y,subscripts){
  one <- data[subscripts,]$BEE=="QUEEN"</pre>
  two <- data[subscripts,]$BEE=="WORKER"</pre>
  lpoints(x[one], y[one], col = 1)
  lpoints(x[two], y[two], col = 2)
  panel.abline(c(intercept_1, slope_1), col=1)
 panel.abline(c(intercept_2, slope_2), col=2)
)
reg = linear_model_no_intr(data2, '~/script/test/math650/figures/math650_hw8_fig2_alpha0_9
trellis.device(postscript, color=T, file='^/script/test/math650/figures/math650_hw8_fig3_a
draw_data_no_intr(reg, data2)
dev.off()
reg = linear_model_no_intr(data2, '~/script/test/math650/figures/math650_hw8_fig2_alpha0_8
trellis.device(postscript, color=T, file='~/script/test/math650/figures/math650_hw8_fig3_a
draw_data_no_intr(reg, data2)
dev.off()
reg = linear_model_no_intr(data2, '~/script/test/math650/figures/math650_hw8_fig2_alpha0_6
```

trellis.device(postscript, color=T, file='^/script/test/math650/figures/math650_hw8_fig3_a

draw_data_no_intr(reg, data2)
dev.off()