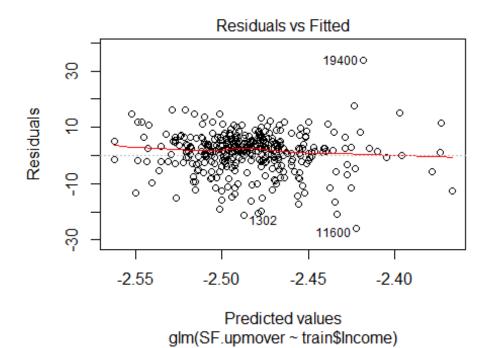
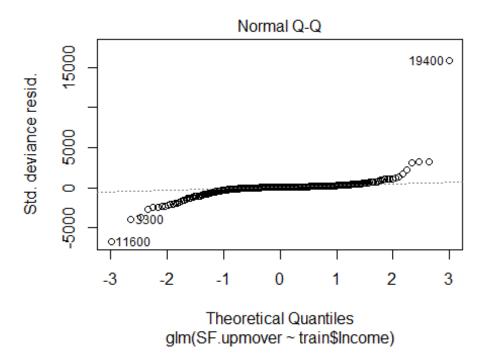
Upward Mobility - model validation

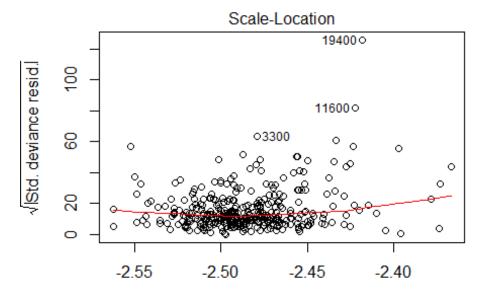
Alex Haase

```
rm(list=ls()) # clear global environment
getwd() # double check wd
mobility =
read.csv('http://dept.stat.lsa.umich.edu/~bbh/s485/data/Mobility.csv',
row.names = 1)
#********* TRAIN DATA ********
#1.)
# 50% of the sample size - test sample
sample size = floor(0.50 * nrow(mobility))
# set the seed to make the partition reproductible
set.seed(123)
train_ind = sample(seq_len(nrow(mobility)), size = sample_size)
train = mobility[train_ind, ]
test = mobility[-train_ind, ]
#nrow(train) #370 rows
#nrow(test) #371 rows
#nrow(train$N.child)
#nrow(test$N.child)
write.csv(test, file="testdata.csv", row.names=T) # creates file:
testdata.csv which is a separate file for the test data
#2.)
#model.frame(fitted model)
# produces a data frame for the model without missing values
Income traindata = model.frame(train$p.upmover~train$Income)
Gini_traindata = model.frame(train$p.upmover~train$Gini)
Seg_poverty_traindata = model.frame(train$p.upmover~train$Seg_poverty)
Seg racial traindata = model.frame(train$p.upmover~train$Seg racial)
Seg_income_traindata = model.frame(train$p.upmover~train$Seg_income)
```

```
#Likelihood Ratio Test
#choosing significance: 0.05
SF.upmover = with(train, N.child * prop.lowstart * cbind(S=train$p.upmover,
F=(1-train$p.upmover)))
SF.upmover = round(SF.upmover)
fit1 = glm(SF.upmover ~ 1, family = binomial('logit'), data = train)
fit2 = glm(SF.upmover ~ train$Income, family = binomial('logit'), data =
train)
fit3 = glm(SF.upmover ~ train$Income * train$Gini, family =
binomial('logit'), data = train)
fit4 = glm(SF.upmover ~ train$Income * train$Gini * train$Seg poverty, family
= binomial('logit'), data = train)
fit5 = glm(SF.upmover ~ train$Income * train$Gini * train$Seg_poverty *
train$Seg_racial, family = binomial('logit'), data = train)
fit6 = glm(SF.upmover ~ train$Income * train$Gini * train$Seg_poverty *
train$Seg racial * train$Seg income, family = binomial('logit'), data =
train)
anova(fit1, fit2, fit3, fit4, fit5, test = "LRT")
## Analysis of Deviance Table
##
## Model 1: SF.upmover ~ 1
## Model 2: SF.upmover ~ train$Income
## Model 3: SF.upmover ~ train$Income * train$Gini
## Model 4: SF.upmover ~ train$Income * train$Gini * train$Seg_poverty
## Model 5: SF.upmover ~ train$Income * train$Gini * train$Seg_poverty *
##
      train$Seg racial
##
    Resid. Df Resid. Dev Df Deviance Pr(>Chi)
## 1
           362
                   18944
                   18734 1
                               209.9 < 2.2e-16 ***
## 2
           361
## 3
          359
                   17195 2
                              1539.9 < 2.2e-16 ***
                   13457 4
                               3737.8 < 2.2e-16 ***
## 4
          355
## 5
          347
                   10456 8
                               3000.6 < 2.2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
#model with the best fit:
#fit 2: Income with Gini
#Generate Residual Plots
plot(fit2)
```

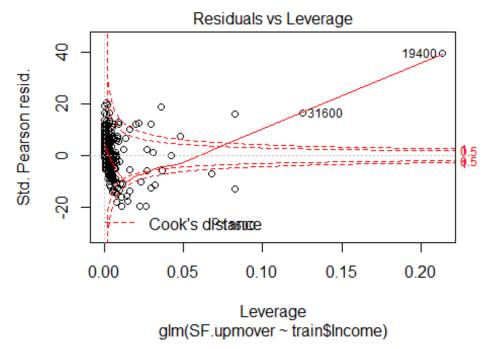






Predicted values glm(SF.upmover ~ train\$Income)

```
#Modeling with natural spline bases
#install.packages("dplyr")
#install.packages("Rcpp")
#install.packages("DAAG")
dat0 = dplyr::filter(train)
#ns_mods = paste("SF.upmover ~ ns(train$Income, df=", 1:4, ")")
library(splines)
library(DAAG)
## Warning: package 'DAAG' was built under R version 3.4.4
## Loading required package: lattice
```



```
library(MASS)
##
## Attaching package: 'MASS'
## The following object is masked from 'package:DAAG':
##
##
       hills
library(boot)
## Warning: package 'boot' was built under R version 3.4.2
##
## Attaching package: 'boot'
## The following object is masked from 'package:lattice':
##
##
       melanoma
crossval = cv.glm(train, fit2)
anova(fit2, crossval, test = "LRT")
## Analysis of Deviance Table
## Model: binomial, link: logit
##
## Response: SF.upmover
```

```
##
## Terms added sequentially (first to last)
##
##
                Df Deviance Resid. Df Resid. Dev Pr(>Chi)
##
## NULL
                                  362
                                           18944
## train$Income 1
                     209.88
                                  361
                                           18734 < 2.2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
#SF.upmover and train$Income
#SF.upmover
#train$Income #length = 370
length(train$Income)
## [1] 370
dim(SF.upmover) #dim = 370 2
## [1] 370
#SF.upmover[,'S']
#SF.upmover[,'F']
#using R spline() function
splineOutput = spline(x=SF.upmover[,'S'], y=train$Income)
#splineOutput$x #length = 792
#splineOutput$y # length = 792
#appending the above spline outputs to fit train data length
spline1 = splineOutput$x[2:371]
spline2 = splineOutput$y[2:371]
class(spline1)
## [1] "numeric"
class(SF.upmover[,'S'])
## [1] "numeric"
fitSx = glm(SF.upmover ~ 1, family = binomial('logit'), data = train)
fitSy = glm(SF.upmover ~ train$Income, family = binomial('logit'), data =
train)
anova(fitSx, fitSy, test = "LRT")
## Analysis of Deviance Table
## Model 1: SF.upmover ~ 1
## Model 2: SF.upmover ~ train$Income
```

```
Resid. Df Resid. Dev Df Deviance Pr(>Chi)
## 1
          362
                   18944
## 2
                              209.88 < 2.2e-16 ***
           361
                   18734 1
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
#****** TEST DATA *******
testdata = read.csv("testdata.csv", TRUE, ",")
#testdata
#hand selected: Commute and Local_gov_spending
Commute testdata = model.frame(testdata$p.upmover~testdata$Commute)
Local gov spending testdata =
model.frame(testdata$p.upmover~testdata$Local_gov_spending)
# LR-test
#choosing significance: 0.05
testdata = na.omit(testdata)
p.upmover2 = na.omit(testdata$p.upmover)
Commute = na.omit(testdata$Commute)
Local gov spending = na.omit(testdata$Local gov spending)
SF.upmover2 = with(testdata, N.child * prop.lowstart * cbind(S=p.upmover2,
F=(1-p.upmover2))
SF.upmover2 = round(SF.upmover2)
dim(SF.upmover2)
## [1] 205
fit1b = glm(SF.upmover2 ~ 1, family = binomial('logit'), data = testdata)
fit2b = glm(SF.upmover2 ~ Commute, family = binomial('logit'), data =
testdata)
fit3b = glm(SF.upmover2 ~ Commute * Local_gov_spending, family =
binomial('logit'), data = testdata)
logit mod = anova(fit1b, fit2b, fit3b, test = "LRT")
logit mod
## Analysis of Deviance Table
##
## Model 1: SF.upmover2 ~ 1
## Model 2: SF.upmover2 ~ Commute
## Model 3: SF.upmover2 ~ Commute * Local_gov_spending
    Resid. Df Resid. Dev Df Deviance Pr(>Chi)
## 1
          204
                   12908
## 2
          203
                   11781 1 1127.58 < 2.2e-16 ***
```

```
## 3     201     10965     2     816.21 < 2.2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1

#each of the variables were significant
#look at coefficients
#newfit = update(Logit_mod, data = testdata)
#coeffs = coef(summary(newfit))
#coeffs[, "Pr(>|z|)"] = p.adjust(coeffs[, "Pr(>|z|)"], "holm")
#coeffs
```