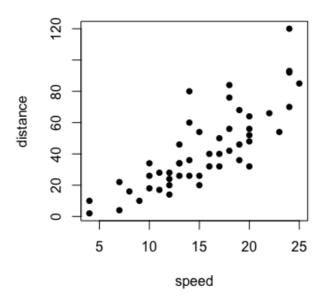
```
# title: "MSDS596 - HW3"
# author: "Diego Sarachaga"
# date: "10/23/2018"

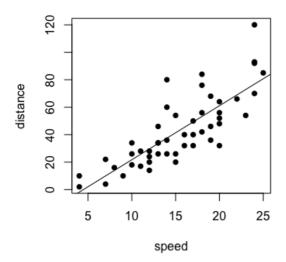
#Problem 1
library(faraway)
library(MASS)

data(cars)
head(cars)

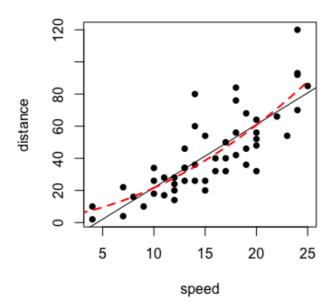
#part a
plot(dist ~ speed, data=cars, xlab = "speed", ylab = "distance", pch=16)
```



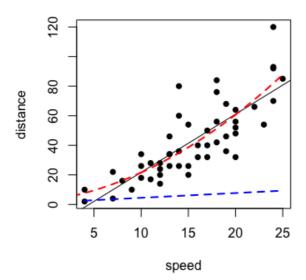
#part b
cars.lm <- lm(dist ~ speed, data=cars)
abline(cars.lm)</pre>



#part c
cars.lm2 <- lm(dist ~ speed + I(speed^2), data=cars)
xval <- seq(from=0, to=25, by=0.1)
hat2 <- predict(cars.lm2, newdata=list(speed=xval))
lines(xval, hat2, col="red", lty=2, lwd=2)</pre>



#part d
cars.lm3 <- lm(sqrt(dist) ~ speed, data=cars)
xval <- pretty(cars\$speed, 50)
hat3 <- predict(cars.lm3, newdata=list(speed=xval))
lines(xval, hat3, col="blue", lty=2, lwd=2)</pre>

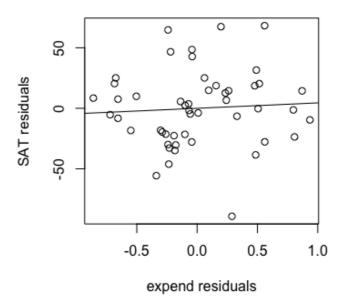


```
#Problem 2
library(faraway)
data(sat)
head(sat)

lmod <- lm(total ~ expend + salary + ratio + takers, sat)

#part a
#expend
d <- residuals(lm(total ~ salary + ratio + takers, sat))
m <- residuals(lm(expend ~ salary + ratio + takers, sat))
plot(m,d,xlab="expend residuals",ylab="SAT residuals",main="Partial regression for 'expend' ")
abline(0,coef(lmod)['expend'])</pre>
```

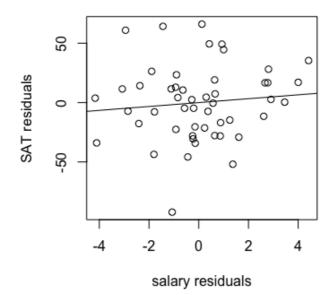
#### Partial regression for 'expend'



```
#salary
d <- residuals(lm(total ~ expend + ratio + takers, sat))</pre>
```

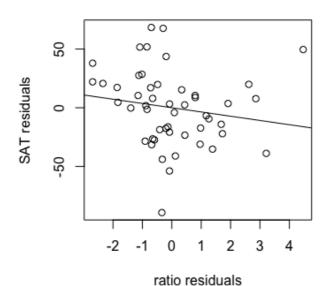
```
m <- residuals(lm(salary ~ expend + ratio + takers, sat))
plot(m,d,xlab="salary residuals",ylab="SAT residuals",main="Partial
regression for 'salary' ")
abline(0,coef(lmod)['salary'])</pre>
```

# Partial regression for 'salary'



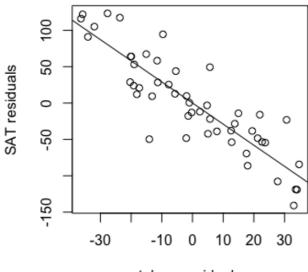
```
#ratio
d <- residuals(lm(total ~ expend + salary + takers, sat))
m <- residuals(lm(ratio ~ expend + salary + takers, sat))
plot(m,d,xlab="ratio residuals",ylab="SAT residuals",main="Partial regression for 'ratio' ")
abline(0,coef(lmod)['ratio'])</pre>
```

### Partial regression for 'ratio'



```
#takers
d <- residuals(lm(total ~ expend + salary + ratio, sat))
m <- residuals(lm(takers ~ expend + salary + ratio, sat))
plot(m,d,xlab="takers residuals",ylab="SAT residuals",main="Partial regression for 'takers' ")
abline(0,coef(lmod)['takers'])</pre>
```

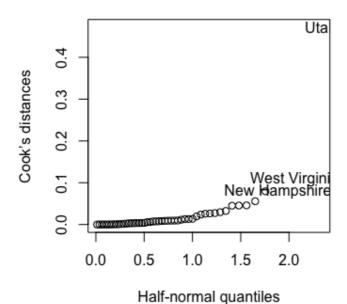
# Partial regression for 'takers'



takers residuals

```
#part b
hatv <- hatvalues(lmod)</pre>
head(sort(hatv,decreasing=T))
        Utah | California | Connecticut | New Jersey |
                                                           New York
Alaska
    0.2921128 | 0.2821179 | 0.2254519 | 0.2220978
                                                           0.1915752
0.1803061
sum(hatv)
#5
#part c
stud <- rstudent(lmod)</pre>
jackres <- stud*(44/(45-stud^2))^0.5</pre>
head(jackres[order(abs(stud),decreasing=T)])
# West Virginia |
                    Utah
                              | North Dakota
                                                 | New Hampshire |
Nevada
               Iowa
# -3.491336
                    2.700696
                                    2.318848
                                                     2.291067
-1.772759
          1.586454
```

```
states <- row.names(sat)
cook <- cooks.distance(lmod)
halfnorm(cook,3,labs=states,ylab="Cook's distances")</pre>
```



#part e

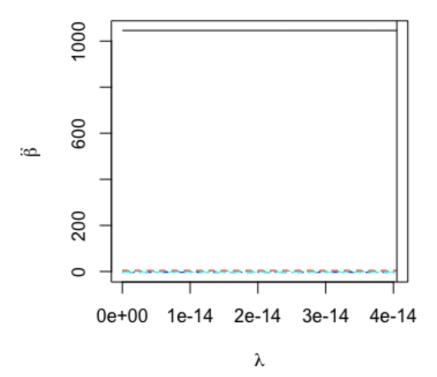
```
x <- model.matrix(lmod)[, -1]
vif(x)
# expend | salary | ratio | takers
# 9.465320 | 9.217237 | 2.433204 | 1.755090</pre>
```

```
#part f
rgmod <- lm.ridge(total ~ expend + salary + ratio + takers, sat,
lambda = seq(0, 4.05e-14, len=21))
select(rgmod)
#modified HKB estimator is 0.3496881
#modified L-W estimator is 0.4728102
#smallest value of GCV at 4.05e-14</pre>
```

#### coef(rgmod)

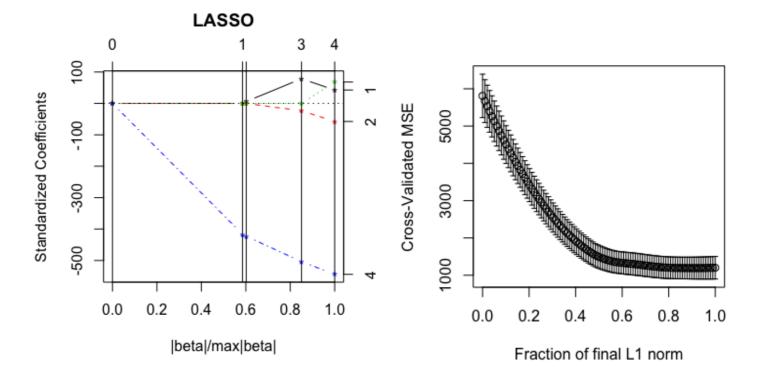
```
#
                      expend
                                 salary
                                            ratio
                                                     takers
# 0.0000e+00 1045.972 4.462594 1.637917 -3.624232 -2.904481
# 2.0250e-15 1045.972 4.462594 1.637917 -3.624232 -2.904481
# 4.0500e-15 1045.972 4.462594 1.637917 -3.624232 -2.904481
# 6.0750e-15 1045.972 4.462594 1.637917 -3.624232 -2.904481
# 8.1000e-15 1045.972 4.462594 1.637917 -3.624232 -2.904481
# 1.0125e-14 1045.972 4.462594 1.637917 -3.624232 -2.904481
# 1.2150e-14 1045.972 4.462594 1.637917 -3.624232 -2.904481
# 1.4175e-14 1045.972 4.462594 1.637917 -3.624232 -2.904481
# 1.6200e-14 1045.972 4.462594 1.637917 -3.624232 -2.904481
# 1.8225e-14 1045.972 4.462594 1.637917 -3.624232 -2.904481
# 2.0250e-14 1045.972 4.462594 1.637917 -3.624232 -2.904481
```

# Ridge trace plot: SAT



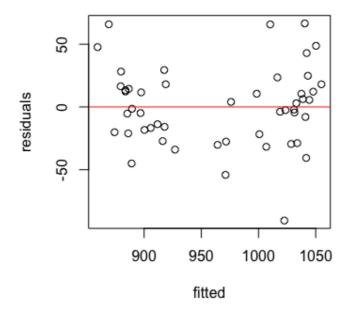
```
#part g
#install.packages("lars")
library(lars)

lmod <- lars(as.matrix(sat[1:4]), sat$total); plot(lmod)
cvlmod <- cv.lars(as.matrix(sat[1:4]), sat$total)
cvlmod$index[which.min(cvlmod$cv)]
# 0.9494949</pre>
```

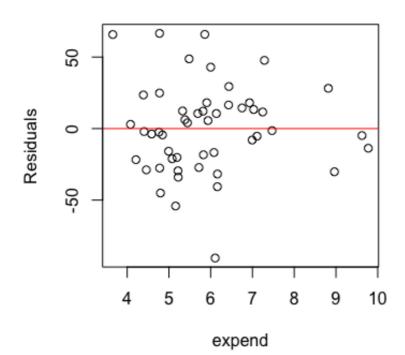


```
#It looks like a nonlinearity pattern

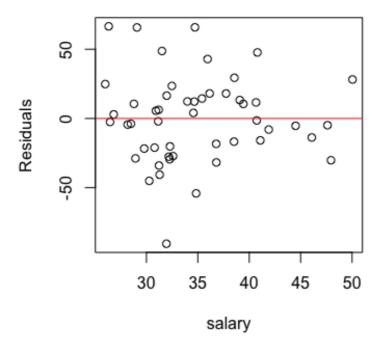
#part b
plot(fitted(lmod),sqrt(abs(residuals(lmod))),
xlab="Fitted",ylab=expression(sqrt(hat(epsilon))))
```



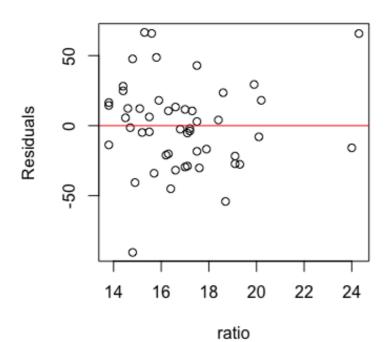
plot(sat\$expend,residuals(lmod),xlab="expend",ylab="Residuals")
abline(h=0, col="red")



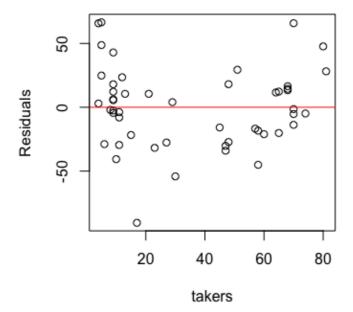
plot(sat\$salary,residuals(lmod),xlab="salary",ylab="Residuals")
abline(h=0, col="red")



plot(sat\$ratio,residuals(lmod),xlab="ratio",ylab="Residuals")
abline(h=0, col="red")

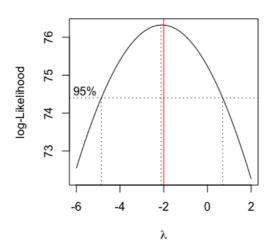


plot(sat\$takers,residuals(lmod),xlab="takers",ylab="Residuals")
abline(h=0, col="red")

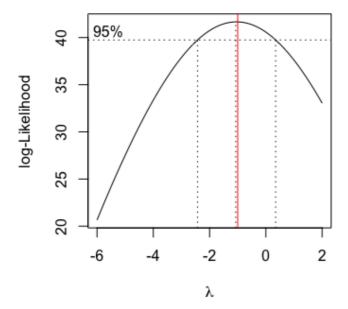


#salary is the most appropiate for these data acording to it's plot.

#c
boxcox(lmod,lambda=seq(-6,2,.1))
abline(v=-2, col = "red")



```
lmod1 <- lm(total^2 ~ expend + salary + ratio + takers, sat)
boxcox(lmod1,lambda = seq(-6,2,.1), plotit = TRUE);
abline(v=-1, col = "red")</pre>
```



lmod2 <- lm(sqrt(total) ~ expend + salary + ratio + takers, sat)
boxcox(lmod1,lambda = seq(-6,2,.1), plotit = TRUE);
abline(v=-1, col = "red")</pre>

