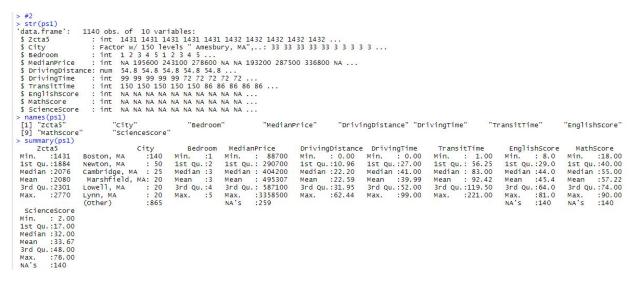
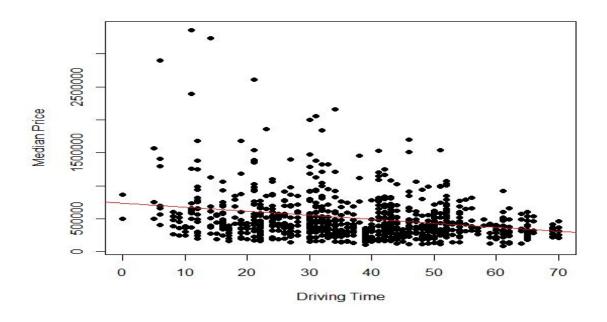
1) setwd("c:/Users/btdan/Desktop/ECN 145/hw")



2) Observations: 1140 of 10 variables Average driving time: 39.99 minutes Average transit time: 92.42 minutes

Longest commute to Downtown Boston: 62.44

3)



From the regression line, we can see a negative correlation between median prices and driving times. Home prices will decline as commute time increases, and vice versa.

4)

The regression relates to the scatter plot by finding lines that fit best into the data. Our coefficients then give us the slope and intercept of that line.

For every one minute increase in Driving Time (slope), there will be a -6,073 reduction in housing prices.

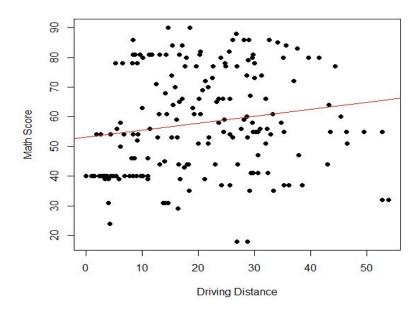
Intercept = 731,941

5)

```
> fitbed <- lm(MedianPrice ~ DrivingTime + factor(Bedroom), data=ps1)
> summary(fitbed)
call:
lm(formula = MedianPrice ~ DrivingTime + factor(Bedroom), data = ps1)
Residuals:
    Min
              10 Median
                                 3Q
                                         Max
-584405 -163258 -37330 91742 2570711
Coefficients:
                   Estimate Std. Error t value Pr(>|t|)
                   493073.5 36077.2 13.667 < 2e-16 ***
-7236.9 697.4 -10.378 < 2e-16 ***
(Intercent)
DrivingTime
factor(Bedroom)2 158185.5 35833.2 4.414 1.15e-05 ***
factor(Bedroom)3 284683.4 35949.4 7.919 8.20e-15 ***
factor(Bedroom)4 374321.1
                                 36400.8 10.283 < 2e-16 ***
factor(Bedroom)5 572258.4 38447.8 14.884 < 2e-16 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' '1
Residual standard error: 300600 on 783 degrees of freedom
Multiple R-squared: 0.3072, Adjusted R-squared: 0.3028
F-statistic: 69.45 on 5 and 783 DF, p-value: < 2.2e-16
```

By comparing the cost of adding a bedroom relative to the cost of a single bedroom, we see a positive trend. For example, this would mean that it would cost \$158,185 more to own a two-bedroom house than a single one. Its R-squared is higher than before, from .07 to .31, which means that more of the data fits into the regression model. Our predictions become more accurate with the addition of bedroom controls.

6)



The scatter plots show a positive correlation between math scores and commute distances. For every increase in math scores, they will have to commute an extra mile.

```
> #7
> fitfull <- lm(MedianPrice ~ DrivingTime + factor(Bedroom) + MathScore, data = ps1)</pre>
> summary(fitfull)
lm(formula = MedianPrice ~ DrivingTime + factor(Bedroom) + MathScore,
     data = ps1)
Min 1Q Median 3Q Max
-503594 -133684 -37220 70837 2653533
Coefficients:
                    Estimate Std. Error t value Pr(>|t|)
                                 42023.4 4.134 3.94e-05 ***
657.9 -13.944 < 2e-16 ***
(Intercept)
                    173745.3
DrivingTime -9173.2
factor(Bedroom)2 134228.6
                                  32879.2 4.082 4.91e-05 ***
33021.3 7.697 4.22e-14 ***
factor(Bedroom)3 254149.7
factor(Bedroom)4 341663.8
                                  33446.9 10.215 < 2e-16 ***
factor(Bedroom)5 515213.8
                                 35520.3 14.505 < 2e-16 ***
                                    590.4 12.300 < 2e-16 ***
MathScore
                      7261.6
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' '1
Residual standard error: 275300 on 782 degrees of freedom
Multiple R-squared: 0.4195, Adjusted R-squared: 0.4151
F-statistic: 94.2 on 6 and 782 DF, p-value: < 2.2e-16
```

The coefficient on math score is the correct sign and is positive. In other words, a one-unit change in math scores will raise home prices by 7261.6.

- 8) The positive relationship between math scores and housing prices from question 6 indicates that there be another variable that further reduces home prices for every one unit added in commute time.
- 9) Driving time coefficient = -9173.2, so home prices are reduced by \$9,173.2 for every minute increase in commute time. The magnitude of the coefficient seems plausible when calculating a wider time horizon and considering if benefits exceed additional costs to travel farther (time).
- 10) An explanation for this case might be that students are responsible for the positive relationship between math scores and driving distance. They are expected to test well and have cheaper preferences.

You may regress math scores to the variables driving time and bedroom to test their relationship with large and small homes.

Code

```
setwd("c:/Users/btdan/Desktop/ECN 145/hw")
#1
library(readr)
ps1 <- read.csv("ps1 data.csv")
View(ps1)
Zcta5 <- c(ps1$V1)
City <- c(ps1$V2)
Bedroom <- c(ps1$V3)
MedianPrice <- c(ps1$V4)
DrivingDistance <- c(ps1$V5)
DrivingTime <- c(ps1$V6)
MathScore <- c(ps1$V7)
#2
dim(ps1)
nrow(ps1)
names(ps1)
summary(ps1)
str(ps1)
is.na(ps1)
ps1[!complete.cases(ps1),]
getOption("max.print")
ps1 <- na.omit(ps1)
summary(ps1)
#3
plot(x = ps1$DrivingTime, y = ps1$MedianPrice, xlab = "Driving Time", ylab = "Median Price",
pch = 19
abline(lm(ps1$MedianPrice ~ ps1$DrivingTime), col = "red")
#4
```

```
fit <- lm(MedianPrice ~ DrivingTime, data = ps1)
summary(fit)
#5
fitbed <- lm(MedianPrice ~ DrivingTime + factor(Bedroom), data=ps1)
summary(fitbed)
#6
plot(x = ps1$DrivingDistance, y = ps1$MathScore, xlab = "Driving Distance", ylab = "Math
Score", pch = 19)
abline(lm(ps1$MathScore ~ ps1$DrivingDistance), col = "red")
fitmath <- lm(MathScore ~ DrivingDistance, data = ps1)
summary(ps1)
#7
fitfull <- lm(MedianPrice ~ DrivingTime + factor(Bedroom) + MathScore, data = ps1)
summary(fitfull)
fitsize <- lm(MathScore ~ DrivingTime + factor(Bedroom), data = ps1)
summary(fitsize)
```