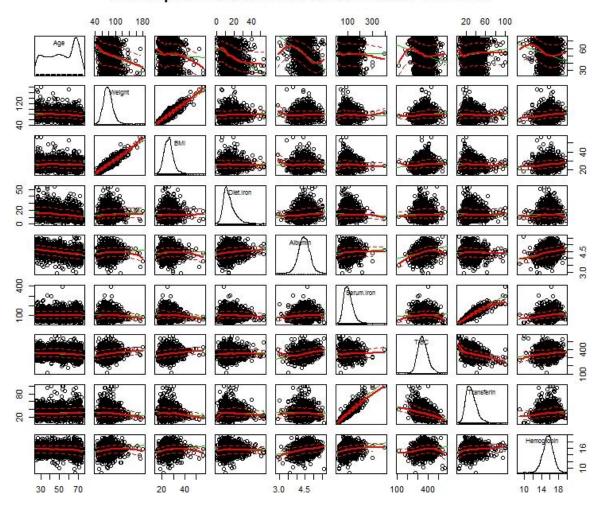
KATHLEEN ZHEN

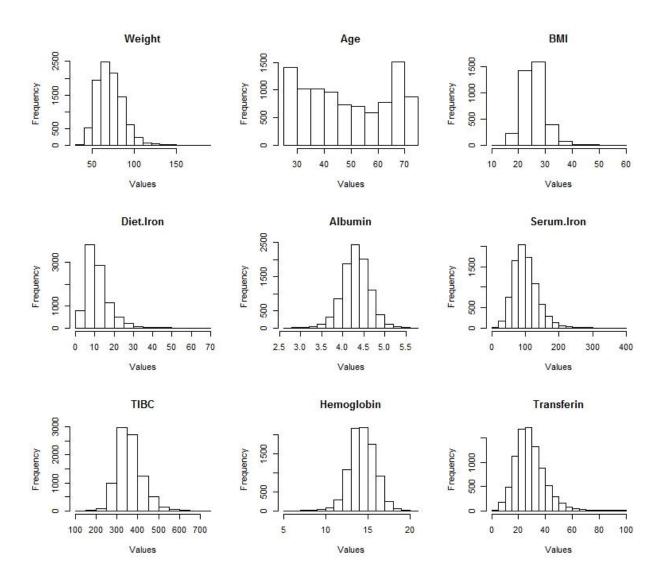
STA 141 HW 2

999210972

PART 1: LOOKING INTO THE CONTINUOUS VARIABLES

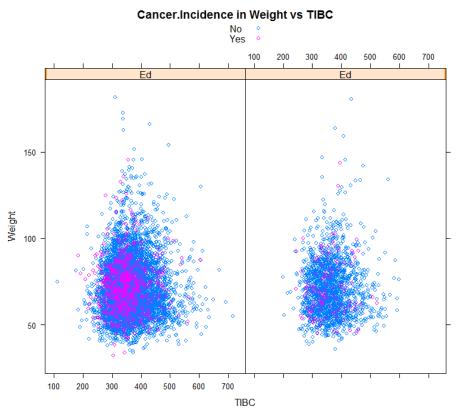
Scatterplot Matrix of NHANES Continuous Variables



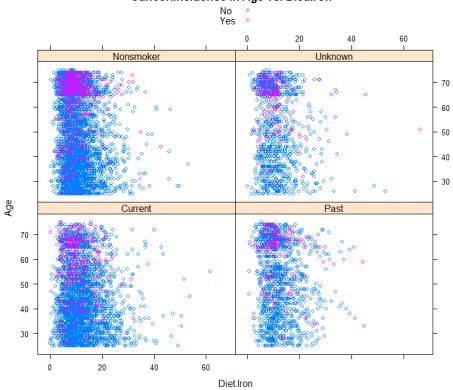


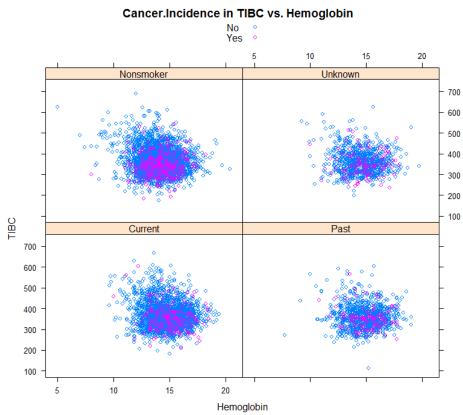
In the continuous variables in NHANES, I did a scatterplot matrix with the line of best fit through the data. The dotted lines represent the upper and lower bounds of the line of best fit. Looking at the data, most of the continue variables are in clutters when plotted against each other. The two pairs that are positively linear are Age vs. Weight and Serum.Iron vs Transferin. Looking at the histograms for each continuous variable, we can see that all the variables have a centralized value that is more dense. All the variables besides Age have this effect since the data is from all ages.

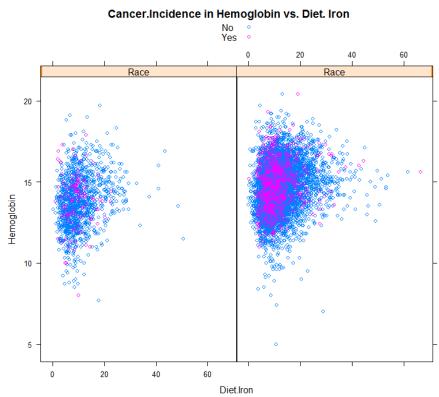
PART 2: LOOKING INTO NATURAL CLUSTERS IN THE DATA IN RELATION TO CATEGORICAL VARIABLES



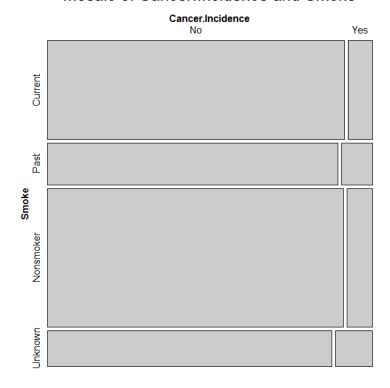
Cancer.Incidence in Age vs. Diet.Iron



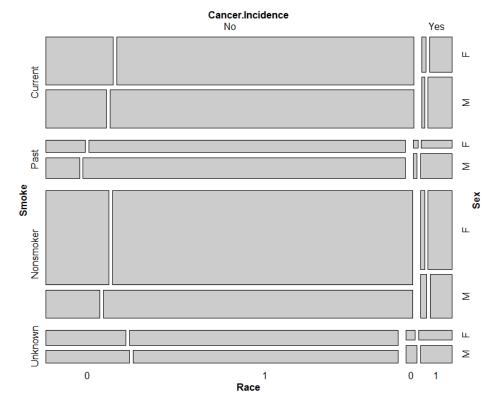




Mosaic of Cancer.Incidence and Smoke



Mosaic of Cancer.Incidence, Smoke, Race, and Sex

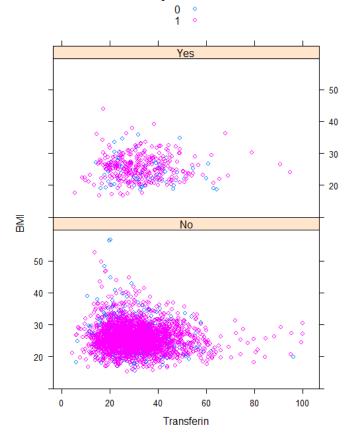


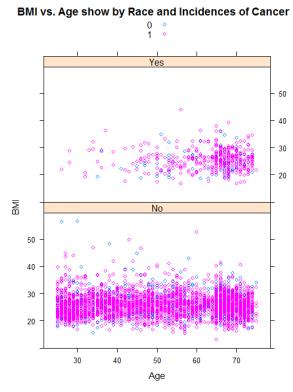


In this section, I plotted various continuous variables against each other while it focuses on categorical variables. I will use Cancer.Incidence for most of the plots because I want to find if the continuous variables affects incidences of cancer. Looking at the various plots, one common trend that occurs is cancer is found more commonly in older ages. Most people intake roughly the same amount of Diet.Iron and cancer is most commonly found in nonsmokers at older ages. Another common thing that I notice is that all the categorical variables are skewed to one side. The side that is more dense generally has more incidences of cancer as shown in Weight vs TBIC and Hemoglobin vs Diet.Iron, just to name a few. In the breakdown of smokers, nonsmokers, past smokers, and unknown, the most data comes from nonsmokers. In proportion to the data, nonsmokers also have the most incidences of cancer. Female Caucasians appear to have more incidences of cancer. As shown by Albumin vs. Age and Serum.Iron vs Age, older female ages have more incidences of cancer.

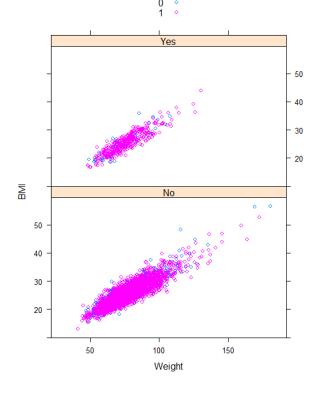
PART 3: VARIABLES THAT AFFECT BMI

MI vs. Transferin show by Race and Incidences of Cance

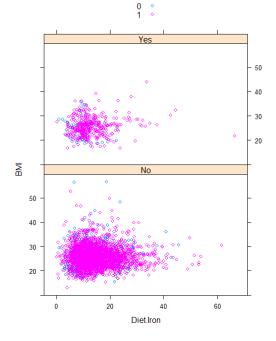


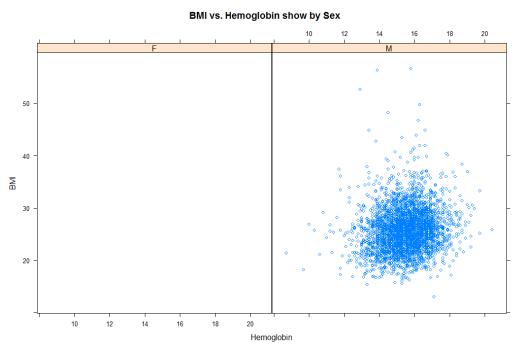


BMI vs. Weight show by Race and Incidences of Cancer



BMI vs. Diet.Iron show by Race and Incidences of Cancer





When I am looking at how other continuous variables affect BMI, I also showed their classification by race and incidences of cancer. I reported some of the continuous variables above because the other graphs report the same thing. Most of the variables are in clutters except BMI vs. Weight which means weight should affect BMI the most. When I look into the classifications by Sex, I noticed there are no females data for BMI. Since there are no female data, I want to use all the continuous variables to predict BMI terms. I did a linear regression on the continuous variable with the y variable being BMI.

```
Residuals:
   Min
           1Q Median
                          3Q
                                Max
-7.1518 -1.2211 -0.0911 1.1278 12.8751
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept) 1.4257077 0.8895702
                               1.603 0.10911
                                      < 2e-16 ***
           0.0304795 0.0025187 12.101
Age
           Weight
Diet.Iron -0.0254019 0.0053033 -4.790 1.75e-06 ***
Albumin
          -0.0002927 0.1158053 -0.003 0.99798
Serum.Iron -0.0134904 0.0053656 -2.514 0.01198 *
          0.0057457 0.0017064 3.367 0.00077 ***
TIBC
Transferin 0.0323225 0.0180754 1.788 0.07385 .
Hemoglobin 0.0605786 0.0294835 2.055 0.04000 *
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 1.821 on 2886 degrees of freedom
 (6680 observations deleted due to missingness)
Multiple R-squared: 0.8075,
                            Adjusted R-squared: 0.807
F-statistic: 1514 on 8 and 2886 DF, p-value: < 2.2e-16
```

I get the following equation:

```
BMI = 1.4257077 + 0.0304(Age) + 0.2644(Weight) - 0.025(Diet.Iron) - 0.0002(Albumin) - 0.134(Serum.Iron) + 0.0057(TIBC) + 0.0323(Transferin) + 0.0605(Hemoglobin).
```

Then I tested to see if any of these variables affect the predictions. If they don't affect it, it is safe to remove. I did the testing using a stepwise regression.

```
Age Weight Diet.Iron Albumin Serum.Iron TIBC Transferin Hemoglobin
                                                       .....
                    п п<sub>ф</sub>п
     (1)
1
                                                                                                                        . .
                                     . .
                                                      . .
                                                                     . .
                                                                                         . . . . .
               "*" "*"
2
     (1)
                                                     . . . . . . . . . . . .
                \pi_{\hat{\mathbf{x}}}\pi^-\pi_{\hat{\mathbf{x}}}\pi^-
                                     n & n
                                                                                         . . . . .
3
     (1)
                                                     п п
                                                                     . .
                                                                                         n n n<sub>*</sub>n
                                                                                                                        .. ..
                                    п<sub>ф</sub>п
                "*" "*"
     (1)
                                                     ....
               п<sub>*</sub>п п<sub>*</sub>п
                                     n<sub>*</sub>n
                                                                      n & n
                                                                                          n & n
                                                                                                  .....
5
     (1)
     (1) "*" "*"
                                    n<sub>w</sub>n
                                                      ....
                                                                     n & n
                                                                                          n<sub>&</sub>n
6
     (1) "*" "*"
                                                      . . . . . . .
                                    n<sub>&</sub>n
                                                                     n_{\frac{1}{N}}n
                                                                                         n_{\frac{1}{N}}n = n_{\frac{1}{N}}n
                                                                                                                        n \otimes n
7
     (1) "*" "*"
                                                     n<sub>*</sub>n
                                    11 g 11
                                                                     m_{\frac{1}{2}}m
                                                                                         п<sub>*</sub>п п<sub>*</sub>п
                                                                                                                        m_{\frac{1}{N}}m
```

	R ²
1	0.7901516
2	0.8033375
3	0.8047414
4	0.8060639
5	0.8067235

6	0.8069211
7	0.8070682*
8	0.8070013

Looking at the R^2 table, number 7 has the highest value. Going back to the table above, row 7 means we remove Albumin. Our resulting regression is BMI = 1.4257077 + 0.0304(Age) + 0.2644(Weight) - 0.025(Diet.Iron) - 0.134(Serum.Iron) + 0.0057(TIBC) + 0.0323(Transferin) + 0.0605(Hemoglobin).

APPENDIX

```
install.packages("hexbin")
install.packages("car")
library("hexbin")
library("tidyverse")
library("car")
#load("~/R/STA 141A/HW2/NHANES.Rdata")
data("NHANES")
##PART 1
pairs(~Age + Weight + BMI + Diet.Iron + Albumin + Serum.Iron + TIBC + Transferin + Hemoglobin, data =
NHANES, main = "Scatterplot Matrix of NHANES Continuous Variables")
scatterplotMatrix(~Age + Weight + BMI + Diet.Iron + Albumin + Serum.Iron + TIBC + Transferin +
Hemoglobin, data = NHANES, main = "Scatterplot Matrix of NHANES Continuous Variables")
qplot(Weight, Diet.Iron, data = NHANES, colour = factor(Ed), main = "hi") + geom smooth()
par(mfrow = c(3,3))
hist(NHANES$Weight, xlab = "Values", ylab = "Frequency", main = "Weight")
hist(NHANES$Age, xlab = "Values", ylab = "Frequency", main = "Age")
hist(NHANES$BMI, xlab = "Values", ylab = "Frequency", main = "BMI")
hist(NHANES$Diet.Iron, xlab = "Values", ylab = "Frequency", main = "Diet.Iron")
hist(NHANES$Albumin, xlab = "Values", ylab = "Frequency", main = "Albumin")
hist(NHANES$Serum.Iron, xlab = "Values", ylab = "Frequency", main = "Serum.Iron")
hist(NHANES$TIBC, xlab = "Values", ylab = "Frequency", main = "TIBC")
hist(NHANES$Hemoglobin, xlab = "Values", ylab = "Frequency", main = "Hemoglobin")
hist(NHANES$Transferin, xlab = "Values", ylab = "Frequency", main = "Transferin")
#PART 2
library("lattice")
xyplot(Age ~ Diet.Iron | Smoke, groups = Cancer.Incidence, type = 'p', auto.key = TRUE, data = NHANES,
main = "Cancer.Incidence in Age vs. Diet.Iron") #keep
xyplot(Weight ~ TIBC | Ed, groups = Cancer.Incidence, type = 'p', auto.key = TRUE, data = NHANES, main
= "Cancer.Incidence in Weight vs TIBC") #keep
xyplot(TIBC ~ Hemoglobin | Smoke, groups = Cancer.Incidence, type = 'p', auto.key = TRUE, data =
NHANES, main = "Cancer.Incidence in TIBC vs. Hemoglobin") #keep
xyplot(Hemoglobin ~ Diet.Iron | Race, groups = Cancer.Incidence, type = 'p', auto.key = TRUE, data =
NHANES, main = "Cancer.Incidence in Hemoglobin vs. Diet. Iron") #keep
xyplot(Serum.Iron ~ TIBC | Ed, groups = Cancer.Incidence, type = 'p', auto.key = TRUE, data = NHANES)
#keep
xyplot(Weight ~ Hemoglobin | Smoke, groups = Sex, type = 'p', auto.key = TRUE, data = NHANES) #keep
#PART 3
install.packages('vcd')
library('vcd')
```

```
xyplot(BMI ~ Transferin | Cancer.Incidence, groups = Smoke, type = 'p', auto.key = T, data = NHANES)
mosaic(~Smoke + Cancer.Incidence, data = NHANES, main = "Mosaic of Cancer.Incidence and Smoke")
#PART 4
mosaic(~Smoke + Cancer.Incidence + Sex + Race, data = NHANES, main = "Mosaic of Cancer.Incidence,
Smoke, Race, and Sex")
#PART 5
plotr = ggplot(NHANES, aes(Age, Weight, color = Cancer.Incidence))
plotr+geom point()
plotr+geom_point() + facet_grid(Cancer.Incidence ~.)
plotr = ggplot(NHANES, aes(Age, Serum.Iron, color = Sex))
plotr+geom point()
plotr+geom point() + facet grid(Cancer.Incidence ~.)
plotr = ggplot(NHANES, aes(Age, Diet.Iron, color = Cancer.Incidence))
plotr+geom_point()
plotr+geom point() + facet grid(Cancer.Incidence ~.)
plotr = ggplot(NHANES, aes(Age, Albumin, color = Sex))
plotr+geom point()
plotr+geom point() + facet grid(Cancer.Incidence ~.)
#PART 6
xyplot(BMI ~ Transferin | Cancer.Incidence, groups = Race, type = 'p', auto.key = T, data = NHANES,
main= "BMI vs. Transferin show by Race and Incidences of Cancer")
xyplot(BMI ~ Age | Cancer.Incidence, groups = Race, type = 'p', auto.key = T, data = NHANES, main=
"BMI vs. Age show by Race and Incidences of Cancer")
xyplot(BMI ~ Weight | Cancer.Incidence, groups = Race, type = 'p', auto.key = T, data = NHANES, main=
"BMI vs. Weight show by Race and Incidences of Cancer")
xyplot(BMI ~ Diet.Iron | Cancer.Incidence, groups = Race, type = 'p', auto.key = T, data = NHANES, main=
"BMI vs. Diet.Iron show by Race and Incidences of Cancer")
xyplot(BMI ~ Albumin | Cancer.Incidence, groups = Race, type = 'p', auto.key = T, data = NHANES, main=
"BMI vs. Albumin show by Race and Incidences of Cancer")
xyplot(BMI ~ Serum.Iron | Cancer.Incidence, groups = Race, type = 'p', auto.key = T, data = NHANES,
main= "BMI vs. Serum.Iron show by Race and Incidences of Cancer")
xyplot(BMI ~ TIBC | Cancer.Incidence, groups = Race, type = 'p', auto.key = T, data = NHANES, main=
"BMI vs. TIBC show by Race and Incidences of Cancer")
xyplot(BMI ~ Hemoglobin | Cancer.Incidence, groups = Race, type = 'p', auto.key = T, data = NHANES,
main= "BMI vs. Hemoglobin show by Race and Incidences of Cancer")
xyplot(BMI ~ Hemoglobin | Sex, type = 'p', auto.key = T, data = NHANES, main= "BMI vs. Hemoglobin
show by Sex")
#linear regression
fit = Im(BMI ~ Age + Weight + Diet.Iron + Albumin + Serum.Iron + TIBC + Transferin + Hemoglobin, data =
NHANES)
```

```
summary(fit)
coefficients(fit)
fitted(fit)

install.packages('leaps')
library('leaps')
step(fit)
summary(regsubsets(BMI ~ Age + Weight + Diet.Iron + Albumin + Serum.Iron + TIBC + Transferin + Hemoglobin, data = NHANES, nbest=1, nvmax = 12))
summary(regsubsets(BMI ~ Age + Weight + Diet.Iron + Albumin + Serum.Iron + TIBC + Transferin + Hemoglobin, data = NHANES, nbest=1, nvmax = 12))$adjr2
```