HW3-proposed solutions

Problem 1

Refer to the CDI data set. Using R2 as the criterion, which predictor variable accounts for the largest reduction in the variability in the number of active physicians?

Solution:

```
library(knitr)
CDI <- read.csv("CDI.csv")
Temp<-CDI[,4:17]
X<- Temp[,-c(5)]
Y<- Temp[,5]

prg1 <-function(X,Y){
  out<-matrix(0,nrow=13,ncol=1)
  for (i in 1:13){
  f<-lm(Y~X[,i])
  f1<-summary(f)
  out[i,1]<- f1$r.squared
}
data.frame(names(X) ,out)}</pre>
kable(prg1(X,Y))
```

names.X.	out
Land.area	0.0060957
Total.population	0.8840674
Percent.of.population.aged.18.34	0.0143279
Percent.of.population.65.or.older	0.0000098
Number.of.hospital.beds	0.9033826
Total.serious.crimes	0.6731538
Percent.high.school.graduates	0.0000180
Percent.bachelor.s.degrees	0.0560579
Percent.below.poverty.level	0.0041135
Percent.unemployment	0.0025519
Per.capita.income	0.0999411
Total.personal.income	0.8989137
Geographic.region	0.0006074

```
names(X[which.max(prg1(X,Y)[,2])])
```

```
## [1] "Number.of.hospital.beds"
```

Number of Hospital beds has the highest R Square.

Problem 2

Refer to the CDI data set in Appendix C.2 and Project 1.44. Obtain a separate interval estimate of β_1 , for each region. Use a 90 percent confidence coefficient in each case. Do the regression lines for the different regions appear to have similar slopes?

Solution:

They re different, the confidence intervals do not overlap.

```
f1<-lm(Per.capita.income~Percent.bachelor.s.degrees,data= CDI[CDI[,17]==1,])
f2<-lm(Per.capita.income~Percent.bachelor.s.degrees,data= CDI[CDI[,17]==2,])
f3<-lm(Per.capita.income~Percent.bachelor.s.degrees,data= CDI[CDI[,17]==3,])
f4<-lm(Per.capita.income~Percent.bachelor.s.degrees,data= CDI[CDI[,17]==4,])
confint(f1,level=0.9)
##
                                    5 %
                                            95 %
                              7809.8077 10637.82
## (Intercept)
## Percent.bachelor.s.degrees 460.5177
                                          583.80
confint(f2,level=0.9)
##
                                     5 %
                                              95 %
## (Intercept)
                              12627.0363 14535.774
## Percent.bachelor.s.degrees
                                193.4858
                                           283.853
confint(f3,level=0.9)
                                              95 %
##
                                    5 %
## (Intercept)
                              9516.0773 11543.4929
## Percent.bachelor.s.degrees 285.7076
confint(f4,level=0.9)
##
                                    5 %
                                              95 %
## (Intercept)
                              6862.6967 10367.4086
## Percent.bachelor.s.degrees 364.7585
                                          515.8729
```

Problem 3

Refer to GPA data:

- a) Set up the ANOVA table.
- b) b) What is estimated by MSR in your ANOVA table? by MSE? Under what condition do MSR and MSE estimate the same quantity?
- c) Conduct an F test of whether or not $\beta_1 = 0$. Control the α risk at .01. State the alternatives, decision rule, and conclusion.
- d) What is the absolute magnitude of the reduction in the variation of Y when X is introduced into the regression model? What is the relative reduction? What is the name of the latter measure?
- e) Obtain r and attach the appropriate sign.
- f) Which measure, R^2 or r, has the more clear-cut operational interpretation? Explain.

Solution:

a)

```
GPA <- read.csv("GPA.csv")
f<-lm(GPA~ACT,data=GPA)
anova(f,test="Chi")

## Analysis of Variance Table
##
## Response: GPA
## Df Sum Sq Mean Sq F value Pr(>F)
## ACT 1 3.588 3.5878 9.2402 0.002917 **
## Residuals 118 45.818 0.3883
```

b)

$$\sigma^2 + \beta_1^2 \sum (X_i - \hat{X}), \sigma^2$$
, when $\beta_1 = 0$

c)

From the ANOVA table above, Fstat=9.2402 and P Value= 0.002917. Reject Null and accept Ha. β_1 is significant.

d)

SSR = 3.588, 3.588/(3.588+45.818) = 7.26% or 0.0726, coefficient of determination

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

e)

Sign is positive since the slope is positive.

```
sqrt(0.0726)
## [1] 0.2694439
f)
R<sup>2</sup>
```

Problem 4

Refer to Crime rate data.

- a) Compute the Pearson product-moment correlation coefficient r_{12} .
- b) Test whether crime rate and percentage of high school graduates are statistically independent in the population; use a $\alpha = .01$. State the alternatives, decision rule, and conclusion.
- c) Compute the Spearman rank correlation coefficient rs.
- d) Test by means of the Spearman rank correlation coefficient whether an association exists between crime rate and percentage of high school graduates. State the alternatives, decision rule, and conclusion.

Solution:

a)

b)

```
Crime.Rate <- read.csv("Crime Rate.csv")</pre>
cor.test(Crime.Rate$X,Crime.Rate$Y, method ="pearson")
##
##
   Pearson's product-moment correlation
##
## data: Crime.Rate$X and Crime.Rate$Y
## t = -4.1029, df = 82, p-value = 9.571e-05
\#\# alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
  -0.5761223 -0.2175580
## sample estimates:
##
          cor
## -0.4127033
r12=-0.4127033
```

$$H_0: \rho = 0$$

$$H_A: \rho \neq 0$$

From above, the p-value = 9.571e-05, reject null, ρ is significant

c)

```
cor.test(Crime.Rate$X,Crime.Rate$Y, method ="spearman")

## Warning in cor.test.default(Crime.Rate$X, Crime.Rate$Y, method = ## "spearman"): Cannot compute exact p-value with ties

## ## Spearman's rank correlation rho

## data: Crime.Rate$X and Crime.Rate$Y

## S = 140839, p-value = 5.359e-05

## alternative hypothesis: true rho is not equal to 0

## sample estimates:

## rho

## -0.4259324

d)

H_0: \rho = 0
```

From above, the p-value = 5.359e-05, reject null, ρ is significant

 $H_A: \rho \neq 0$