# assignment2

#### Sidra Effendi

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```
#knitr::opts_knit$set(root.dir = normalizePath("/Users/sidraeffendi/Desktop/PP639/"))
library(haven)
getwd()
```

### ## [1] "/Users/sidraeffendi/Desktop/PP639"

```
pset2 <- read_dta("/Users/sidraeffendi/Desktop/PP639/pset2.dta")
head(pset2)</pre>
```

```
## # A tibble: 6 x 11
     province city county urban yrsed farmer drate_esoph water_grade airpollution
     <chr>
              <chr> <chr> <dbl> <dbl>
                                         <dbl>
                                                     <dbl>
                                                                 <dbl>
                                                                              <dbl>
              Dali~ Xiang~
                                                     0.689
                                                                              0.180
## 1 Yunnan
                               0 2.96 0.983
                                                                  1
## 2 Guizhou Qian~ Dushan
                               0 3.01 0.876
                                                     0.306
                                                                              0.400
                                                                  1
## 3 Guangxi Guil~ Xiang~
                               1 6.53 0.00223
                                                     6.53
                                                                              0.545
## 4 Guangxi Hech~ Luoch~
                               0 3.00 0.713
                                                    12.0
                                                                  1.33
                                                                              0.411
                               0 1.16 0.932
## 5 Hubei
              Xian~ Xiang~
                                                    20.4
                                                                  1.5
                                                                              0.371
## 6 Hubei
              Xian~ Guche~
                               0 2.93 0.972
                                                    12.2
                                                                  1.5
                                                                              0.371
## # ... with 2 more variables: clean_river <dbl>, dirty_river <dbl>
```

#### pset2

```
## # A tibble: 145 x 11
##
     province city county urban yrsed farmer drate_esoph water_grade
      <chr>
              <chr> <chr> <dbl> <dbl>
                                         <dbl>
                                                    <dbl>
                                                                <dbl>
## 1 Yunnan
              Dali~ Xiang~
                                                    0.689
                               0 2.96 0.983
                                                                 1
## 2 Guizhou Qian~ Dushan
                               0 3.01 0.876
                                                    0.306
                                                                 1
## 3 Guangxi Guil~ Xiang~
                               1 6.53 0.00223
                                                    6.53
                                                                 1
## 4 Guangxi Hech~ Luoch~
                               0 3.00 0.713
                                                   12.0
                                                                 1.33
## 5 Hubei
              Xian~ Xiang~
                               0 1.16 0.932
                                                   20.4
                                                                 1.5
## 6 Hubei
              Xian~ Guche~
                               0 2.93 0.972
                                                   12.2
                                                                 1.5
              Yuxi Tongh~
## 7 Yunnan
                               0 3.24 0.690
                                                    0.614
                                                                 1.67
## 8 Xicang
              Lasa Mozhu~
                               0 0.384 0.887
                                                    0.939
                                                                 2
                                                                 2
## 9 Liaoning Dand~ Fengc~
                               0 2.94 0.780
                                                    1.81
## 10 Hebei
              Chen~ Fengn~
                               0 3.42 0.955
                                                    2.28
                                                                 2
## # ... with 135 more rows, and 3 more variables: airpollution <dbl>,
## # clean_river <dbl>, dirty_river <dbl>
```

```
getwd()
```

## [1] "/Users/sidraeffendi/Desktop/PP639"

## Q.1

(a)

```
# measures of spread
summary(pset2$drate_esoph)
```

```
## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 0.000 3.801 7.836 13.157 16.108 91.978
```

The mean of death rate from esophageal cancer is 13.157 and the median is 7.836. The minimum death rate from esophageal cancer is 0 while as the max is 91.98.

```
# get the percentile information
quantile(pset2$drate_esoph)
```

```
## 0% 25% 50% 75% 100%
## 0.000000 3.800585 7.836037 16.108120 91.978317
```

The 25th percentile for death rate from esophageal cancer is 3.8, the 50th paercentile is 7.836 and the 75th percentile is 16.108

(b)

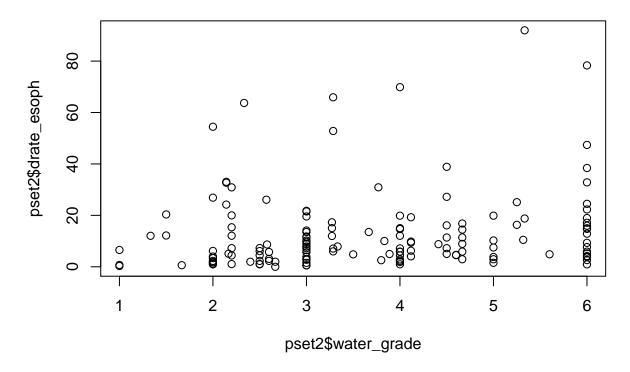
```
# apply mask for clean and dirty rivers
clean_rivers <- pset2[pset2$clean_river == 1,]
dirty_rivers <- pset2[pset2$dirty_river == 1,]
#head(clean_rivers)
#head(dirty_rivers)</pre>
# Select numeric columns for clean rivers
```

```
# Select numeric columns for clean rivers
clean_rivers.numcols <- clean_rivers[, sapply(clean_rivers, is.numeric)]
# find the mean
colMeans(clean_rivers.numcols)</pre>
```

```
##
                                   farmer drate_esoph water_grade airpollution
         urban
                       yrsed
      0.3255814
                                            10.3238502
##
                   4.4043368
                                0.5440947
                                                          3.1631449
                                                                       0.4640643
##
   clean_river dirty_river
##
      1.0000000
                  0.0000000
```

```
# Select numeric columns dirty rivers
dirty_rivers.numcols <- dirty_rivers[, sapply(dirty_rivers, is.numeric)]</pre>
# find the mean
colMeans(dirty_rivers.numcols)
##
          urban
                                   farmer drate_esoph water_grade airpollution
                       yrsed
     0.3728814
##
                   4.2410474
                                0.5678236
                                           17.2868135
                                                          4.4071044
                                                                        0.4916754
## clean_river dirty_river
      0.0000000
                   1.0000000
##
When the rivers are clean we see the water quality is
(c)
# apply mask for urban and non-urban areas
urban <- pset2[pset2$urban == 1,]</pre>
non_urban <- pset2[pset2$urban == 0,]</pre>
# Select numeric columns for urban areas
urban.numcols <- urban[, sapply(urban, is.numeric)]</pre>
# find the mean
colMeans(urban.numcols)
##
          urban
                                   farmer drate esoph water grade airpollution
                       yrsed
##
   1.00000000
                  6.06150380
                               0.01676372
                                           8.04595296 3.77922112 0.49251063
## clean_river dirty_river
   0.56000000
                  0.44000000
# Select numeric columns for non-urban areas
non_urban.numcols <- non_urban[, sapply(non_urban, is.numeric)]</pre>
# find the mean
colMeans(non_urban.numcols)
##
          urban
                                   farmer drate_esoph water_grade airpollution
                       yrsed
     0.0000000
                   3.4307323
                                0.8363742
                                           15.8471101
                                                          3.6114586
                                                                        0.4662405
##
## clean_river dirty_river
     0.6105263
                 0.3894737
```

(d)



## (e) Yes, there appears to be a relationship between esophageal cancer and water quality. The higher the water grade, the worse is the water quality and higher the death rate due to esophageal cancer. For water grade above 5 we see the death rate going 80 and above, while as for water less than 2 the death rate from esophageal cancer is at most 20 (approximately).

## Q.2 Regression analysis I

(a)

```
# regress death rate from esophageal cancer against water grade.
linearMod <- lm(pset2$water_grade ~ pset2$drate_esoph, data=pset2)
summary(linearMod)</pre>
```

```
##
## Call:
   lm(formula = pset2$water_grade ~ pset2$drate_esoph, data = pset2)
##
##
   Residuals:
##
       Min
                1Q Median
                                 3Q
                                         Max
   -2.5702 -1.0064 -0.2927
                                     2.5138
##
                             1.0234
##
```

```
## Coefficients:
##
                      Estimate Std. Error t value Pr(>|t|)
  (Intercept)
                      3.472421
                                 0.148444
                                            23.392
                                                     0.0416 *
## pset2$drate_esoph 0.014964
                                 0.007279
                                             2.056
##
## Signif. codes:
                  0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 1.366 on 143 degrees of freedom
## Multiple R-squared: 0.0287, Adjusted R-squared: 0.02191
## F-statistic: 4.226 on 1 and 143 DF, p-value: 0.04163
                         drate\ esoph_i = 0.015 + 3.472 * water\ grade_i
\#\# b(i)
                                          \hat{\beta}_1 = 3.472
```

On average, for a unit increase in the value of water grade, there is an expected increase of 3.472 in the death rate from esophageal cancer.

### b(ii)

$$SE(\hat{\beta}_1) = 0.148$$

For the different samples of a population we get different values of

 $\hat{\beta_1}$ 

. So, the standard error of the slope represents how much far apart do the slope values lie from the mean. This is supposed to be a normal distribution with the mean at the center.

#### b(iii)

$$T$$
\_stat for the test of null hypothesis that  $\hat{\beta}_1 = 0$  is 23.392

With 95% confidence, since the test statistic is much greater than the critical value of 1.96, we reject the null hypothesis that

 $\hat{\beta_1} = 0$ 

.

#### b(iv)

95% confidence interval =

$$[mean(\hat{\beta}_1) - 1.96 * SE(\hat{\beta}_1), mean(\hat{\beta}_1) + 1.96 * SE(\hat{\beta}_1)]$$

=

```
res <- t.test(pset2$drate_esoph, pset2$water_grade, data = pset2)
res</pre>
```

```
##
## Welch Two Sample t-test
##
## data: pset2$drate_esoph and pset2$water_grade
## t = 7.2792, df = 146.25, p-value = 1.904e-11
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 6.911812 12.063684
## sample estimates:
## mean of x mean of y
## 13.157056 3.669308
```

We are confident that the actual value of the slope,

 $\beta_1$ 

lies between 6.912 and 12.064.

## b(v)

$$\hat{\beta}_0 = 0.015$$

On average, the expected value of death rates due to esophageal cancer when the water-grade value is 0, is 0.015~##~b(vi) The no.of observations =

### b(vii)

$$R^2 = 0.0287$$

Water grade is able to explain 2.87% of the variance in the death rates due to esophageal cancer.

(c)

For a site with water grade = 3

$$drate \ \hat{e}soph_i = 0.015 + 3.472 * 3$$

= 0.015 + =

For a site with water grade = 5

$$drate\_esoph_i = 0.015 + 3.472 * 5$$

= 0.015 + =

## Regression analysis II

(a)

```
# regress death rate from esophageal cancer against water grade.
linearMod <- lm(pset2$water_grade ~ pset2$drate_esoph, data=pset2)
summary(linearMod)</pre>
```

```
##
## Call:
## lm(formula = pset2$water_grade ~ pset2$drate_esoph, data = pset2)
##
## Residuals:
##
       Min
                1Q Median
                                3Q
                                       Max
  -2.5702 -1.0064 -0.2927
                           1.0234
##
## Coefficients:
##
                     Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                     3.472421
                                0.148444
                                          23.392
                                           2.056
## pset2$drate_esoph 0.014964
                                0.007279
                                                   0.0416 *
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.366 on 143 degrees of freedom
## Multiple R-squared: 0.0287, Adjusted R-squared: 0.02191
## F-statistic: 4.226 on 1 and 143 DF, p-value: 0.04163
```

(b)

THe omitted category is...

c(i)

$$\hat{\beta}_1 = 3.472$$

On average, for a unit increase in the value of water grade, there is an expected increase of 3.472 in the death rate from esophageal cancer.

c(ii)

$$SE(\hat{\beta_1}) = 0.148$$

For the different samples of a population we get different values of

 $\hat{\beta_1}$ 

. So, the standard error of the slope represents how much far apart do the the slope values lie from the mean. This is supposed to be a normal distribution with the mean at the center.

c(iii)

 $T\_stat$  for the test of null hypothesis that  $\hat{\beta_1} = 0$  is 23.392

With 95% confidence, since the test statistic is much greater than the critical value of 1.96, we reject the null hypothesis that

 $\hat{\beta_1} = 0$ 

.

## c(iv)

95% confidence interval =

$$[mean(\hat{\beta}_1) - 1.96 * SE(\hat{\beta}_1), mean(\hat{\beta}_1) + 1.96 * SE(\hat{\beta}_1)]$$

=

```
res <- t.test(pset2$drate_esoph, pset2$water_grade, data = pset2)
res</pre>
```

```
##
## Welch Two Sample t-test
##
## data: pset2$drate_esoph and pset2$water_grade
## t = 7.2792, df = 146.25, p-value = 1.904e-11
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 6.911812 12.063684
## sample estimates:
## mean of x mean of y
## 13.157056 3.669308
```

We are confident that the actual value of the slope,

 $\beta_1$ 

lies between 6.912 and 12.064.

# c(v)

$$\hat{\beta}_0 = 0.015$$

On average, the expected value of death rates due to esophageal cancer when the water-grade value is 0, is  $0.015 \ \# \ c(vi)$  The no.of observations =

# c(vii)

$$R^2 = 0.0287$$

Water grade is able to explain 2.87% of the variance in the death rates due to esophageal cancer.

## (d)

For a site with water grade = 3

$$drate\_\hat{e}soph_i = 0.015 + 3.472 * 3$$

= 0.015 + =

For a site with water grade = 5

$$drate\ \hat{e}soph_i = 0.015 + 3.472 * 5$$

= 0.015 + =

## **Q.4**

# Q.5

### R Markdown

This is an R Markdown document. Markdown is a simple formatting syntax for authoring HTML, PDF, and MS Word documents. For more details on using R Markdown see http://rmarkdown.rstudio.com.

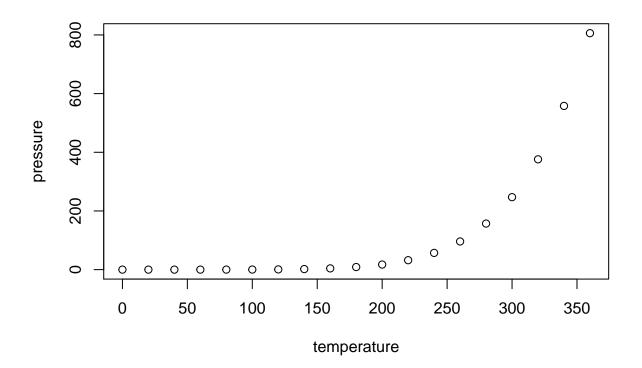
When you click the **Knit** button a document will be generated that includes both content as well as the output of any embedded R code chunks within the document. You can embed an R code chunk like this:

#### summary(cars)

```
##
        speed
                          dist
##
           : 4.0
                    Min.
                            :
                               2.00
    Min.
                    1st Qu.: 26.00
##
    1st Qu.:12.0
    Median:15.0
                    Median : 36.00
##
##
            :15.4
                            : 42.98
    Mean
                    Mean
                    3rd Qu.: 56.00
##
    3rd Qu.:19.0
    Max.
            :25.0
                    Max.
                            :120.00
```

### **Including Plots**

You can also embed plots, for example:



Note that the  $\mbox{echo}$  = FALSE parameter was added to the code chunk to prevent printing of the R code that generated the plot.