Statistics 101A Lecture 1 Due Friday Jan. 17, 2020 at 5:00 PM Homework One Winter 2020

Data set: North Carolina Birth Data (NCBirthNew) First of all, download the data from ccle week 1.

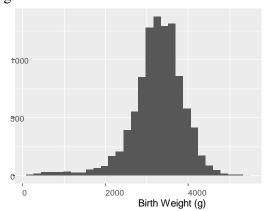
Data Size: 10000 132

Variables Descriptions are posted on a separate file Week 1.

Note: (Use ggplot2 library for plots)

Problem One.

a) Create a histogram for the attribute "Birth Weight (g)" and test the claim that the average Birth Weight is 4300 g.



```
> summary(`Birth Weight (g)`)
   Min. 1st Qu. Median Mean 3rd Qu. Max. NA's
   113.5 2951.0 3319.9 3258.5 3660.4 5334.5 2
> sd(`Birth Weight (g)`, na. rm=T)
[1] 627.9778
> t. test(`Birth Weight (g)`, mu=4300, data=NCB)
```

One Sample t-test

```
data: Birth Weight (g) t=-165.83, df=9997, p-value < 2.2e-16 alternative hypothesis: true mean is not equal to 4300 95 percent confidence interval: 3246.222 3270.844 sample estimates: mean of x 3258.533
```

b) Recode the variable Gender of Child using Male instead of "1" and Female instead of "2"). "save it as GenderNew". Create a barplot for the GenderNew variable and test the claim that the proportion of Males is 0.50.

```
1.5e+77

1.0e+77

5.0e+86

6.0e+60

Female

GenderNew
```

```
> table(GenderNew)
GenderNew
Femal e
         Male
  4841
          5159
> prop. table(table(GenderNew))
GenderNew
         Male
Femal e
0.4841 0.5159
> prop. test(length(GenderNew[GenderNew=="Male"]), length(GenderNew), p=0.5)
        1-sample proportions test with continuity correction
       length(GenderNew[GenderNew == "Male"]) out of length(GenderNew), null
data:
probability 0.5
\dot{X}-squared = 10.049, df = 1, p-value = 0.001524
alternative hypothesis: true p is not equal to 0.5
95 percent confidence interval:
 0. 5060509 0. 5257368
sample estimates:
0.5159
> bi nom. test(length(GenderNew[GenderNew=="Male"]), length(GenderNew), p=0.5)
        Exact binomial test
       length(GenderNew[GenderNew == "Male"]) and length(GenderNew)
number of successes = 5159, number of trials = 10000, p-value = 0.001523
alternative hypothesis: true probability of success is not equal to 0.5
95 percent confidence interval: 0.5060517 0.5257390
sample estimates:
probability of success
                 0.5159
   c) Construct a 95% confidence interval for the average Birth Weight (g)
95 percent confidence interval:
 3246. 222 3270. 844
sample estimates:
mean of x
 3258.533
   d) Construct a 90% confidence interval for the proportion of Male babies in the data.
```

```
95 percent confidence interval: 0.5060509 0.5257368 sample estimates:
```

```
o. 5159
```

Or

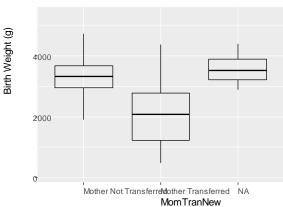
95 percent confidence interval: 0.5060517 0.5257390 sample estimates: probability of success 0.5159

Problem Two:

- a) Create a side-by-side box plot of the variable Birth Weight (g) of the two types of MomTran.
- > MomTranNew<-ifelse(MomTran==1, "Mother Transferred", "Mother Not Transferr ed")
- > table(MomTranNew)

MomTranNew

Mother Not Transferred 9924 Mother Transferred 71



- b) Conduct a two-tailed t-test comparing the average Birth Weight (g) of a Transferred Mom vs the average Birth Weight (g) of a Non-Transferred Mom. Report your p-value. (Assume Equal Variances).
- > t.test(`Birth Weight (g)`~MomTranNew, var. equal =T, data=NCB)

Two Sample t-test

```
data: Birth Weight (g) by MomTranNew t=16.153, df=9992, p-value < 2.2e-16 alternative hypothesis: true difference in means is not equal to 0 95 percent confidence interval: 1047.966 1337.443 sample estimates: mean in group Mother Not Transferred mean in group Mother Transferred 2074.173
```

c) Conduct a simple linear regression using Birth Weight (g) as your response variable and Gest Age (BC) as your predictor.

```
> NCm1<- lm(`Birth Weight (g)` ~`Gest Age (BC)`)
> summary(NCm1)

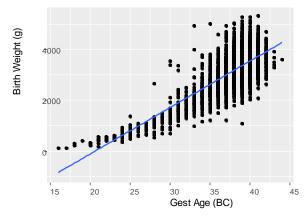
Call:
lm(formula = `Birth Weight (g)` ~ `Gest Age (BC)`)

Residuals:
    Min     10 Median     30 Max
-2284.0 -297.7 -17.1 269.8 2672.3

Coefficients:
```

Estimate Std. Error t value Pr(>|t|)

d) Create a scatter plot Gest Age (BC) vs Birth Weight (g) then plot the least square regression line on the same graph.



e) Report the summary of your linear model, interpret the slope and the y-intercept in the model based on the context.

```
> NCm1$coefficients
   (Intercept) `Gest Age (BC)`
   -3770.0689 182.8796
```

f) Construct a 95% confidence interval for both: the slope and the y-intercept.

g) Using R or a calculator of your choice to calculate SST (total), SSE (residual), SS_{Regression}

Problem Three:

Use the SLR in Problem two to:

a) Compute a 95% confidence interval about the mean response for Gest Age (BC) = 20

```
new_data<-data.frame(Gest.Age..BC.=20)
```

predict(NCm1, data=NCB,newdata=new_data, interval = "confidence")

```
## fit lwr upr
```

1 -112.4778 -178.9687 -45.98698

b) Compute a 95% predication interval for a new observation when Gest Age (BC) = 20

predict(NCm1, data=NCB,newdata=mdata, interval = "prediction")

c) Compare the two intervals.

We notice that those two intervals have the same fit (which is the midpoint), yet the prediction interval is much wider than the confidence interval.

Problem Four:

a) Conduct simple linear regression using Birth Weight (g) as outcome variable and MomTran as a predictor.

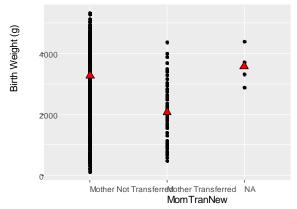
```
> NCm2<-lm(`Birth Weight (g)`~MomTranNew)</pre>
> summary(NCm2)
Call:
lm(formula = `Birth Weight (g)` ~ MomTranNew)
Resi dual s:
    Mi n
              1Q
                   Medi an
                                 3Q
                                         Max
                              393. Š
-3153.4 -315.9
                     53. 0
Coeffi ci ents:
                                  Estimate Std. Error t value Pr(>|t|)
                                                                     <2e-16 ***
(Intercept)
                                  3266.877
                                                  6. 224 524. 92
                                                 73. 839 - 16. 15
MomTranNewMother Transferred - 1192.704
                                                                     <2e-16 ***
                  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' '1
Signif. codes:
Residual standard error: 620 on 9992 degrees of freedom
  (6 observations deleted due to missingness)
Multiple R-squared: 0.02545, Adjusted R-squared: F-statistic: 260.9 on 1 and 9992 DF, p-value: < 2.
                                           p-value: < 2. 2e-16
```

- b) Create a scatter plot for the MomTran vs Birth Weight (g) then plot the least square regression line on the same graph.
- c) Report the summary of your linear model, interpret the slope and the y-intercept in the model.
- > NCm2\$coefficients (Intercept) MomTranNewMother Transferred 3266.877 - 1192.704
 - d) Compare the summary of your SLR in part c with the results of the t-test in Question Two Part (b). State your concludes?
 - > t.test(`Birth Weight (g)`~MomTranNew, var. equal =T, data=NCB)

Two Sample t-test

```
data: Birth Weight (g) by MomTranNew t = 16.153, df = 9992, p-value < 2.2e-16 alternative hypothesis: true difference in means is not equal to 0 95 percent confidence interval: 1047.966 1337.443 sample estimates: mean in group Mother Not Transferred mean in group Mother Transferred 2074.173
```

$> qplot(MomTranNew, `Birth Weight (g)`, data=NCB) + geom_smooth(method='lm')$



The slope is 1192.70, which is the difference between the mean weight of two groups is 1192.70 grams. The y-intercept is 3266. 877, which us the mean of group Mother Not Transferred

We notice that the t-test and the SLR have the same p-value and t-value, and that the estimated slope is the same as the midpoint of the confidence interval in Question 2.

Problem Five:

Sxy/(n-1)

Below are some statistical summaries of the two variables "Gest Age (BC)" as the predictor and "Birth Weight (g)" as the response.

```
> summary(`Gest Age (BC)`)
                                                                NA's
   Min. 1st Qu.
                     Medi an
                                  Mean 3rd Qu.
                                                      Max.
  16.00
            38.00
                      39.00
                                38.43
                                           40.00
                                                     44.00
> summary(`Birth Weight (g)
Min. 1st Qu. Medi
113.5 2951.0 3319
> sd(`Gest Age (BC)`)
                                  Mean 3rd Qu.
                     Medi an
                                                      Max.
                               3258. 5
                                         3660. 4
                     3319. 9
                                                   5334. 5
[1] NA
> var(`Gest Age (BC)`, na. rm=T)
[1] 5. 928025
> var(`Birth Weight (g)`, na. rm=T)
[1] 394356.1
The sample size is 10000-2 = 9998 (the 2 missing values are not considered
in the SLR calculations)
    a) Use the statistical summaries to calculate S_{xx}, S_{xy}, S_{yy}=SST
       Sxx = 5.928025 \times (9998 - 1) = 59262.465925
       SST = Syy = 394356.1 \times (9998 - 1) = 3942377931.7
       Sxy = Sxx \times \beta = 59262.465925 \times 182.880 = 10837919.8
```

b) Calculate the covariance between "Gest Age (BC)" and "Birth Weight (g)"

```
> cov(`Birth Weight (g)`, `Gest Age (BC)`, use="complete.obs")
[1] 1078.662
Or
```

c) Calculate the linear correlation coefficient between "Age" and "Birth Weight (g)"

```
> cor(`Birth Weight (g)`, `Gest Age (BC)`, use="complete.obs") [1] 0.7080693
```

d) What are the values of slope and the y-intercept values of the SLR using "Gest Age (BC)" as the predictor and "Birth Weight (g)" as the response?

```
> NCm1$coefficients
    (Intercept)
                 `Gest Age (BC)`
     - 3770. 0689
                         182. 8796
```

e) Use the equation of the SLR to predict the "Birth Weight (g)" of an infant with 40 weeks Gest Age (BC).

```
new data=data.frame(Gest.Age..BC.=40)
predict(NCm1, newdata=new data,interval="confidence")
```

```
##
                fit
                               lwr
                                               upr
                               3534.781
##
        1
               3545.113
                                               3555.445
```

The predicted birth weight is 3545.113 grams, and a 95% confident interval is (3534.781, 3555.445)