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DS 710 - Homework 9

R assignment

1(a) Read the modified data into R. Check the first few values of each vector to ensure that they were read accurately.

```
# read in the data we cleaned in python
d <- read.csv('D:/Projects/ds710fall2017assignment9/usnews_clean.csv')
```

```
> head(d)
  ID.Number College.Name State Public.private Avg.Math.SAT Avg.Verbal.SAT Avg.combined.SAT Avg.ACT First.quartile...Math.SAT Third.quartile...Math.SAT First.quartile...Verbal.SAT Third.quartile...Verbal.SAT
1 1001 Alaska Pacific University AK 1 490 462 961 22 440 530 NA 430
2 1063 university of Alaska at Fairbanks AK 1 499 462 961 22 440 530 NA 430
3 1085 university of Alaska southeast AK 1 499 462 961 22 440 530 NA 430
4 11662 University of Alaska at Anchorage AK 1 499 462 961 22 440 530 NA 430
5 1002 Alabama Agri. & Mech. Univ. AL 1 NA NA NA 17 NA NA NA
6 1003 Faulkner University AL 2 NA NA NA 20 NA NA NA
  Third.quartile...Verbal.SAT First.quartile...ACT Third.quartile...ACT Num.applications.received Num.applicants.accepted Num.students.enrolled Pct.new.students.from.top.10..of.HS.class
1 550 18 22 193 146 55 16
2 NA NA NA 1852 1427 928 4
3 NA NA NA 146 117 89 4
4 NA NA NA 2065 1598 1162 NA
5 NA 14 17 2817 1920 984 NA
6 NA NA NA 345 320 179 NA
  Pct.new.students.from.top.25..of.HS.class Num.full.time.undergraduates Num.part.time.undergraduates In.state.tuition Out.of.state.tuition Room.and.Board.costs Room.costs Board.costs Additional.fees Estimated.book.costs
1 44 249 869 7560 7560 4120 1620 2500 130 800
2 NA 3885 4519 1742 5226 3590 1800 1790 155 650
3 24 492 1849 1742 5226 4764 2514 2250 34 500
4 NA 6209 10337 1742 5226 3120 2600 2520 114 580
5 NA 3958 305 1700 3400 2550 1108 1442 155 500
6 27 1367 578 5600 3600 3250 1550 1700 300 350
  Estimated.personal.spending Pct.of.faculty.with.PhDs Pct.of.faculty.with.terminal.degree Student.faculty.ratio Pct.alumni.who.donate Instructional.expenditure.per.student Graduation.rate pub_priv iqvrMATH iqvrVERB
1 1500 76 72 11.9 2 10922 15 Private 90 120
2 2304 67 71 10.0 8 11935 NA Public NA NA
3 1162 39 51 9.5 NA 9584 39 Public NA NA
4 1260 48 NA 13.7 6 8046 NA Public NA NA
5 850 53 53 14.3 NA 7043 40 Public NA NA
6 NA 52 56 32.8 NA 3971 55 Private NA NA
```

New columns came in properly and NA values reported as expected.

1(b) Examine the summary of each variable. Identify any unrealistic values and set them to missing. Write a sentence describing what you did, naming the colleges or universities affected. (For example, “Listed ages less than zero (ABC University, XYZ College) were converted to missing data.”)

```
# visually inspect state abbreviations
unique(d$State)

# visually inspect min/max of SAT (200-800 for each) and ACT (1-36) for expected ranges
summary(d)
```

```
> length(unique(d$State))
[1] 51
> unique(d$State)
[1] AK AL AR AZ CA CO CT DC DE FL GA HI IA ID IL IN KS KY LA MA MD ME MI MN MO MS MT NC ND NE NH NJ NM NV NY OH OK OR PA RI SC SD TN TX UT VA VT WA WI WV WY
Levels: AK AL AR AZ CA CO CT DC DE FL GA HI IA ID IL IN KS KY LA MA MD ME MI MN MO MS MT NC ND NE NH NJ NM NV NY OH OK OR PA RI SC SD TN TX UT VA VT WA WI WV WY
```

```
> summary(d)
  ID.Number College.Name State Public.private Avg.Math.SAT Avg.Verbal.SAT Avg.combined.SAT Avg.ACT First.quartile...Math.SAT Third.quartile...Math.SAT First.quartile...Verbal.SAT Third.quartile...Verbal.SAT
Min. : 1002 Bethel College : 4 NY :101 Min. :1,000 Min. :320.0 Min. :780.0 Min. : 600.0 Min. :11.00 Min. :220.0 Min. :330.0 Min. :200.0
1st Qu.: 1874 Concordia College: 4 PA : 83 1st Qu.:1,000 1st Qu.:460.0 1st Qu.:422.0 1st Qu.: 884.5 1st Qu.:20.25 1st Qu.:410.0 1st Qu.:530.0 1st Qu.:380.0
Median : 2650 Trinity College : 4 CA : 70 Median :2,000 Median :500.0 Median :457.0 Median : 957.0 Median :22.00 Median :453.0 Median :580.0 Median :410.0
Mean : 3126 Columbia College : 3 TX : 60 Mean :1,439 Mean :506.8 Mean :461.2 Mean : 968.0 Mean :22.12 Mean :462.2 Mean :581.1 Mean :418.5
3rd Qu.: 3431 Union College : 3 MA : 56 3rd Qu.:1,000 3rd Qu.:544.0 3rd Qu.:492.0 3rd Qu.:1018.0 3rd Qu.:24.00 3rd Qu.:510.0 3rd Qu.:630.0 3rd Qu.:450.0
Max. :30431 Augustana College: 2 OH :152 Max. :2,000 Max. :750.0 Max. :665.0 Max. :1410.0 Max. :31.00 Max. :740.0 Max. :785.0 Max. :630.0
  (Other) :1282 (Other):680 NA's :525 NA's :525 NA's :522 NA's :588 NA's :530 NA's :530 NA's :530
  Third.quartile...Verbal.SAT First.quartile...ACT Third.quartile...ACT Num.applications.received Num.applicants.accepted Num.students.enrolled Pct.new.students.from.top.10..of.HS.class
Min. :1330.0 Min. :10.00 Min. :15.00 Min. : 35.0 Min. : 35.0 Min. : 18.0 Min. : 1.00
1st Qu.:480.0 1st Qu.:18.00 1st Qu.:23.00 1st Qu.: 695.8 1st Qu.: 554.5 1st Qu.:236.0 1st Qu.:13.00
Median :530.0 Median :19.00 Median :25.00 Median :1470.0 Median :1095.0 Median :447.0 Median :21.00
Mean :530.5 Mean :19.82 Mean :25.11 Mean :2752.1 Mean :1870.7 Mean :778.9 Mean :25.67
3rd Qu.:570.0 3rd Qu.:22.00 3rd Qu.:27.00 3rd Qu.:3314.2 3rd Qu.:2303.0 3rd Qu.:984.0 3rd Qu.:32.00
Max. :720.0 Max. :29.00 Max. :35.00 Max. :48094.0 Max. :26330.0 Max. :7425.0 Max. :98.00
NA's :1530 NA's :639 NA's :639 NA's :10 NA's :11 NA's :5 NA's :235
  Pct.new.students.from.top.25..of.HS.class Num.full.time.undergraduates Num.part.time.undergraduates In.state.tuition Out.of.state.tuition Room.and.Board.costs Room.costs Board.costs Additional.fees
Min. : 6.00 Min. : 59 Min. : 1.0 Min. : 480 Min. : 1044 Min. :1260 Min. : 500 Min. : 531 Min. : 9.0
1st Qu.: 36.75 1st Qu.: 966 1st Qu.:131.2 1st Qu.:2580 1st Qu.:6111 1st Qu.:3320 1st Qu.:1710 1st Qu.:1619 1st Qu.:130.0
Median : 50.00 Median :1812 Median :472.0 Median :8050 Median :8670 Median :14030 Median :2200 Median :1980 Median :264.5
Mean : 52.35 Mean :3693 Mean :1081.5 Mean :7897 Mean :9277 Mean :4162 Mean :2515 Mean :2061 Mean :392.0
3rd Qu.: 66.00 3rd Qu.:4540 3rd Qu.:1313.0 3rd Qu.:11600 3rd Qu.:11659 3rd Qu.:4849 3rd Qu.:3040 3rd Qu.:2402 3rd Qu.:492.0
Max. :100.00 Max. :31643 Max. :21816.0 Max. :25750 Max. :25750 Max. :18700 Max. :7400 Max. :6250 Max. :1474.0
NA's :202 NA's :13 NA's :32 NA's :30 NA's :20 NA's :76 NA's :321 NA's :498 NA's :274
  Estimated.book.costs Estimated.personal.spending Pct.of.faculty.with.PhDs Pct.of.faculty.with.terminal.degree Student.faculty.ratio Pct.alumni.who.donate Instructional.expenditure.per.student Graduation.rate
Min. : 90 Min. : 75 Min. : 8.00 Min. : 20.00 Min. : 2.30 Min. : 0.00 Min. :1834 Min. : 8.00
1st Qu.: 480 1st Qu.: 900 1st Qu.: 57.00 1st Qu.: 63.00 1st Qu.:11.80 1st Qu.:11.00 1st Qu.:6116 1st Qu.:47.00
Median : 502 Median :1230 Median :71.00 Median :77.00 Median :14.30 Median :19.00 Median :7729 Median :60.00
Mean : 550 Mean :1389 Mean :68.65 Mean :75.22 Mean :14.86 Mean :20.91 Mean :8088 Mean :60.41
3rd Qu.: 600 3rd Qu.:11794 3rd Qu.:82.00 3rd Qu.:90.00 3rd Qu.:17.60 3rd Qu.:29.00 3rd Qu.:10054 3rd Qu.:74.00
Max. :2340 Max. :6900 Max. :105.00 Max. :100.00 Max. :91.80 Max. :81.00 Max. :62469 Max. :118.00
NA's :48 NA's :181 NA's :32 NA's :30 NA's :2 NA's :222 NA's :39 NA's :98
  pub_priv iqvrMATH iqvrVERB
Private:832 Min. :10.0 Min. : 0
Public :470 1st Qu.:100.0 1st Qu.:100
Median :120.0 Median :110
Mean :120.9 Mean :112
3rd Qu.:140.0 3rd Qu.:120
Max. :400.0 Max. :310
NA's :530 NA's :530
```

```
# Pct.of.faculty.with.PhDs has 105 percent value somewhere in the variable
min(d$Pct.of.faculty.with.PhDs, na.rm = T)
max(d$Pct.of.faculty.with.PhDs, na.rm = T)

> min(d$Pct.of.faculty.with.PhDs, na.rm = T)
[1] 8
> max(d$Pct.of.faculty.with.PhDs, na.rm = T)
[1] 105
```

```
# find all records with erroneous values in this column
d[which(d$Pct.of.faculty.with.PhDs > 100 | d$Pct.of.faculty.with.PhDs < 0),]
```

```
> d[which(d$Pct.of.faculty.with.PhDs > 100 | d$Pct.of.faculty.with.PhDs < 0),]
  ID.Number college.Name State Public.private Avg.Math.SAT Avg.Verbal.SAT Avg.combined.SAT Avg.ACT First.quartile...Math.SAT Third.quartile...Math.SAT
1176 10298 Texas A&M university at Galveston TX 1 NA NA NA NA NA NA NA NA
822 First.quartile...Verbal.SAT Third.quartile...Verbal.SAT First.quartile...ACT Third.quartile...ACT Num.applications.received Num.applicants.accepted Num.students.enrolled
1176 NA NA NA NA NA NA NA NA NA NA NA NA
822 Pct.new.students.from.top.10..of.HS.class Pct.new.students.from.top.25..of.HS.class Num.full.time.undergraduates Num.part.time.undergraduates In.state.tuition Out.of.state.tuition
1176 NA NA 22 NA 47 1206 134 780 4860
822 Room.and.Board.costs Room.costs Board.costs Additional.fees Estimated.book.costs Estimated.personal.spending Pct.of.faculty.with.PhDs Pct.of.faculty.with.terminal.degree Student.faculty.ratio
1176 NA NA NA NA NA 650 105 103 88 17.4
822 Pct.alumni.who.donate Instructional.expenditure.per.student Graduation.rate pub_prv iqrMATH iqrVERB
1176 NA 16 6415 43 Public NA NA
```

two schools have percentages >100 (Sage Colleges, Texas A&M) for this variable---set them to NA

```
d$Pct.of.faculty.with.PhDs[which(d$Pct.of.faculty.with.PhDs > 100 | d$Pct.of.faculty.with.PhDs < 0)] <- NA
```

grad rates > 100 or < 0 are not possible

```
> # Graduation.rate has value > 100 percent
> max(d$Graduation.rate, na.rm = T)
[1] 118
```

```
> d[which(d$Graduation.rate > 100 | d$Graduation.rate < 0),]
  ID.Number college.Name State Public.private Avg.Math.SAT Avg.Verbal.SAT Avg.combined.SAT Avg.ACT First.quartile...Math.SAT Third.quartile...Math.SAT First.quartile...Verbal.SAT
772 2685 Cazenovia College NY 2 392 375 781 19 NA NA NA NA
772 Third.quartile...Verbal.SAT First.quartile...ACT Third.quartile...ACT Num.applications.received Num.applicants.accepted Num.students.enrolled Pct.new.students.from.top.10..of.HS.class
772 NA NA NA NA 3847 3433 527 9
772 Pct.new.students.from.top.25..of.HS.class Num.full.time.undergraduates Num.part.time.undergraduates In.state.tuition Out.of.state.tuition Room.and.Board.costs Room.costs Board.costs
772 Additional.fees Estimated.book.costs Estimated.personal.spending Pct.of.faculty.with.PhDs Pct.of.faculty.with.terminal.degree Student.faculty.ratio Pct.alumni.who.donate
772 395 600 500 22 9384 9384 4840 2420 2420
772 Instructional.expenditure.per.student Graduation.rate pub_prv iqrMATH iqrVERB
772 7697 118 Private NA NA
```

Cazenovia College has 118 percent grad rate, set it to NA

```
d$Graduation.rate[which(d$Graduation.rate > 100 | d$Graduation.rate < 0)] <- NA
```

iqrMATH min value is reported at -10, so there is a record where 3QT > 1QT. Likewise, iqrVERB min is reported at 0, so 3QT==1QT in a record. We should take a look at these records.

```
> d[which(d$First.quartile...Math.SAT > d$Third.quartile...Math.SAT | d$First.quartile...Verbal.SAT > d$Third.quartile...Verbal.SAT),]
  ID.Number college.Name State Public.private Avg.Math.SAT Avg.Verbal.SAT Avg.combined.SAT Avg.ACT First.quartile...Math.SAT Third.quartile...Math.SAT First.quartile...Verbal.SAT
462 2189 Westfield State College MA 1 460 420 880 NA 400 500 400
674 2954 Pembroke State University NC 1 433 385 818 NA 460 450 340
674 Third.quartile...Verbal.SAT First.quartile...ACT Third.quartile...ACT Num.applications.received Num.applicants.accepted Num.students.enrolled Pct.new.students.from.top.10..of.HS.class
462 400 NA NA 3100 2150 825 3
674 420 NA NA 944 774 440 14
674 Pct.new.students.from.top.25..of.HS.class Num.full.time.undergraduates Num.part.time.undergraduates In.state.tuition Out.of.state.tuition Room.and.Board.costs Room.costs Board.costs
462 20 3234 941 1408 5542 3788 2600 1188
674 34 2174 529 628 6380 2760 1410 1350
462 Additional.fees Estimated.book.costs Estimated.personal.spending Pct.of.faculty.with.PhDs Pct.of.faculty.with.terminal.degree Student.faculty.ratio Pct.alumni.who.donate
462 1746 500 1300 75 79 15.7 20
674 514 550 1498 77 77 15.0 5
674 Instructional.expenditure.per.student Graduation.rate pub_prv iqrMATH iqrVERB
462 4222 65 Public 100 0
674 6443 48 Public -10 80
```

Westfield State College has Verbal 3QT == 1QT and Pembroke State University has Math 1QT > 3QT.

Westfield-set Verbal 1QT, 3QT and IQR to NA

```
# Westfield College
d[462, c("First.quartile...Verbal.SAT", "Third.quartile...Verbal.SAT", "iqrVERB")] <- NA
```

Pembroke-set Math 1QT, 3QT and IQR to NA

```
# Pembroke
d[674, c("First.quartile...Math.SAT", "Third.quartile...Math.SAT", 'igrMATH')] <- NA
```

1(c) Find the mean percentage of alumni who donate, for private and public schools.

```
> prvDonate <- d[Pct.alumni.who.donate[d$pub_prv=='Private']]
> pubDonate <- d[Pct.alumni.who.donate[d$pub_prv=='Public']]
> mean(prvDonate, na.rm = T)
[1] 24.58287
> mean(pubDonate, na.rm = T)
[1] 13.44944
```

1(d) The two groups, public and private, have neither the same number of samples nor the same variance (as shown below)

```
> length(prvDonate[!is.na(prvDonate)])
[1] 724
> length(pubDonate[!is.na(pubDonate)])
[1] 356
> sd(prvDonate, na.rm = T)
[1] 12.91669
> sd(pubDonate, na.rm = T)
[1] 8.069433
```

Thus, we will use Welch's t-test (same method from assignment 8)

https://en.wikipedia.org/wiki/Welch%27s_t-test:

```
welch.t.test <- function(mean1,mean2,sd1,sd2,n1,n2){
  # standard error
  se<- sqrt( sd1^2/n1 + sd2^2/n2 )

  # test statistic
  t<- (mean1 - mean2)/se

  # degrees of freedom
  df <- ( sd1^2/n1 + sd2^2/n2 )^2 / ( (sd1^2/n1)^2/(n1-1) + (sd2^2/n2)^2/(n2-1) )

  # p-value
  p <- 2 * pt(-abs(t), df)

  # output data
  dframe<-data.frame(test.statistic=t,
                     DOF=df,
                     P.Value=p)

  return(dframe)
}
```

```
> prvMean <- mean(prvDonate, na.rm = T)
> pubMean <- mean(pubDonate, na.rm = T)
> prvSD <- sd(prvDonate, na.rm = T)
> pubSD <- sd(pubDonate, na.rm = T)
> prvLen <- length(prvDonate[!is.na(prvDonate)])
> pubLen <- length(pubDonate[!is.na(pubDonate)])
> welch.t.test(prvMean, pubMean, prvSD, pubSD, prvLen, pubLen)
  test.statistic      DOF      P.Value
1      17.31684 1018.898 4.364589e-59
```

Given a *p-value* close to zero, there is enough evidence to reject the null hypothesis that the two means are equal. The percentage of alumni that donate back to the institution is greater in private schools than in public schools.

1(e) Write to CSV

```
> write.csv(d, 'usnews_cleaned_updated.csv', quote = FALSE, row.names = FALSE)
```

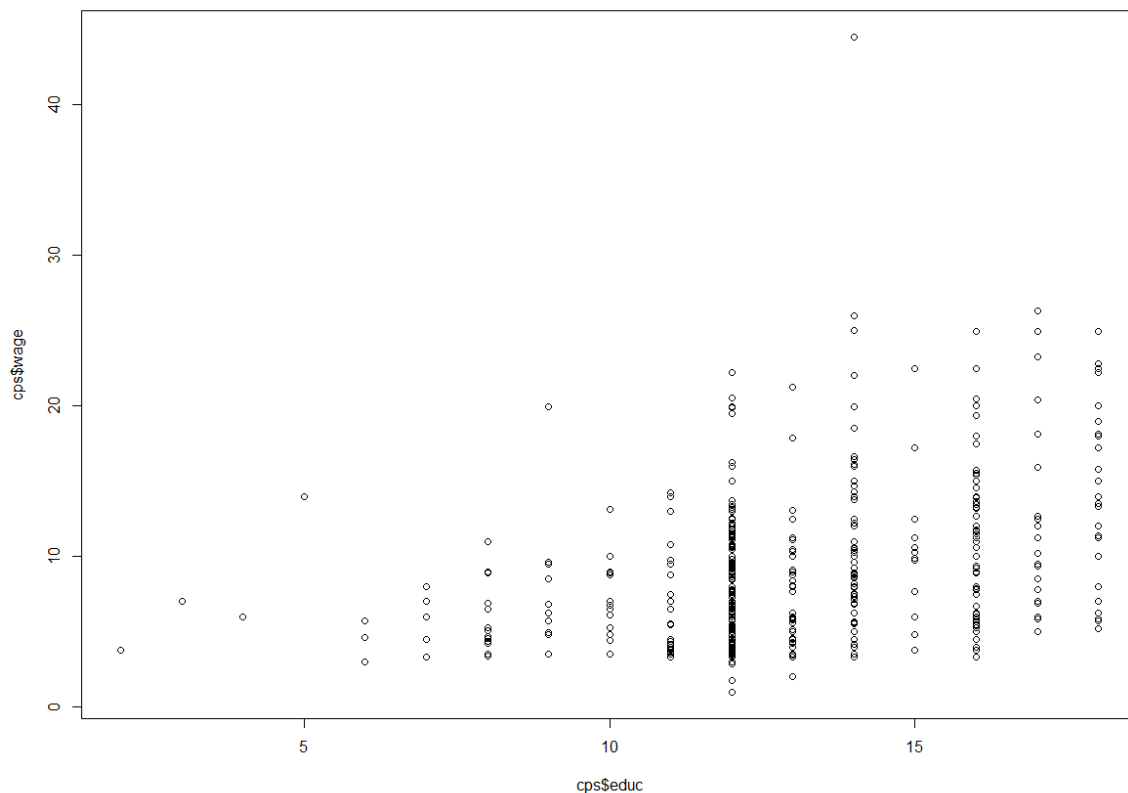
Confirm w/ text editor (Atom) that quotations did not show up around entries.

```
13 1019,Huntingdon College,AL,2,513,446,959,23,480,570,400,530,20,24,608,520,127,26,47,538,126,8080,8080,3920,1380,2540,100,500,1100,63,72,11.4,9,7703,44,Priv
14 1020,Jacksonville State University,AL,1,NA,NA,NA,20,NA,NA,NA,NA,17,22,1627,1413,887,NA,NA,5160,1475,1740,2610,2600,1030,1570,85,570,1500,66,67,20.1,6,4604,
15 1023,Judson College,AL,2,NA,NA,NA,22,NA,NA,NA,NA,19,24,313,228,137,10,30,552,67,5780,5780,3600,NA,NA,NA,NA,70,70,17.9,27,5159,43,Private,NA,NA
```

2(a)

Read the data into R and plot wages versus education.

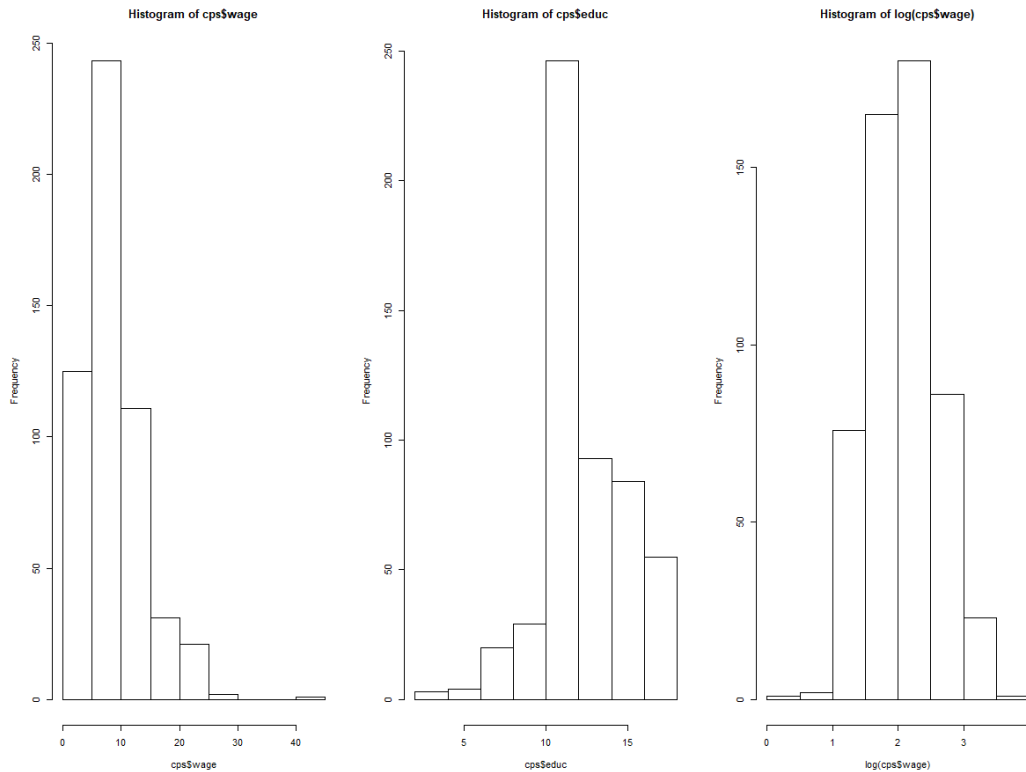
```
cps <- read.csv('D:/Projects/ds710fall2017assignment9/cps.csv')
str(cps)
head(cps)
plot(cps$educ, cps$wage)
```



There appears to be a relationship between wages and education. Low education seems to have lower wages, but also lower observations. The min wage of higher education levels seems to be higher than lower educations as well. A linear regression might help explain the significant of the linear relationship.

We observe right skewness on the wage variable, thus a log transformation might help our analysis.

```
par(mfrow=c(1,2))
hist(cps$wage)
hist(cps$educ)
```



2 (b)

At first glance, it seems like the linear model could be a good start in the explanation of the relationship in our data (low p-value, but low adj R sqr)

```
par(mfrow=c(2,2))
mod <- lm(cps$wage ~ cps$educ)
plot(mod)
summary(mod)
```

```
> summary(mod)

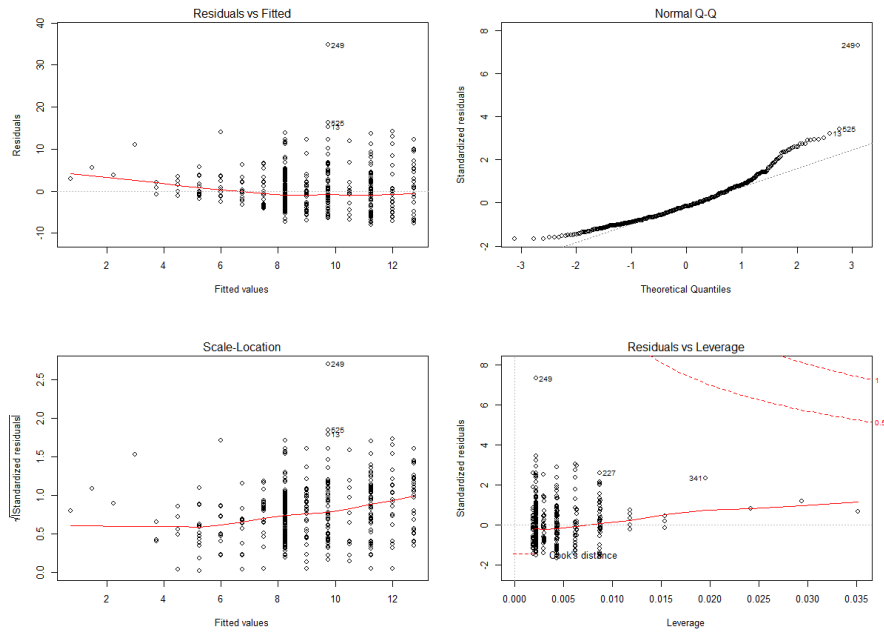
Call:
lm(formula = cps$wage ~ cps$educ)

Residuals:
    Min       1Q   Median       3Q      Max
-7.911  -3.260  -0.760   2.240  34.740

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept) -0.74598    1.04545  -0.714   0.476
cps$educ     0.75046    0.07873   9.532 <2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 4.754 on 532 degrees of freedom
Multiple R-squared:  0.1459,    Adjusted R-squared:  0.1443 
F-statistic: 90.85 on 1 and 532 DF,  p-value: < 2.2e-16
```

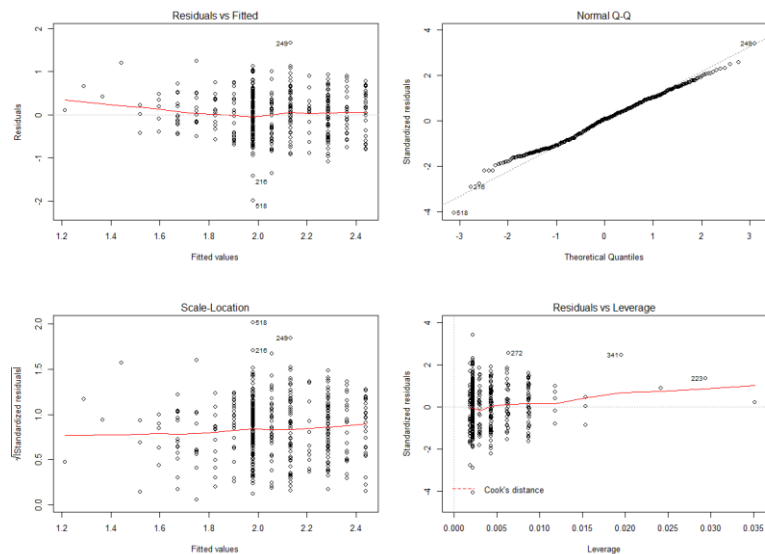
When we examine diagnostic plots of the linear model we can see higher theoretical quantiles trail upwards on (QQ plot), thus a log transform of wage might help the residuals fall into a normal distribution.



Build the Linear Model using a log transformation on wages

```
par(mfrow=c(2,2))
mod <- lm(log(cps$wage) ~ cps$educ)
plot(mod)
```

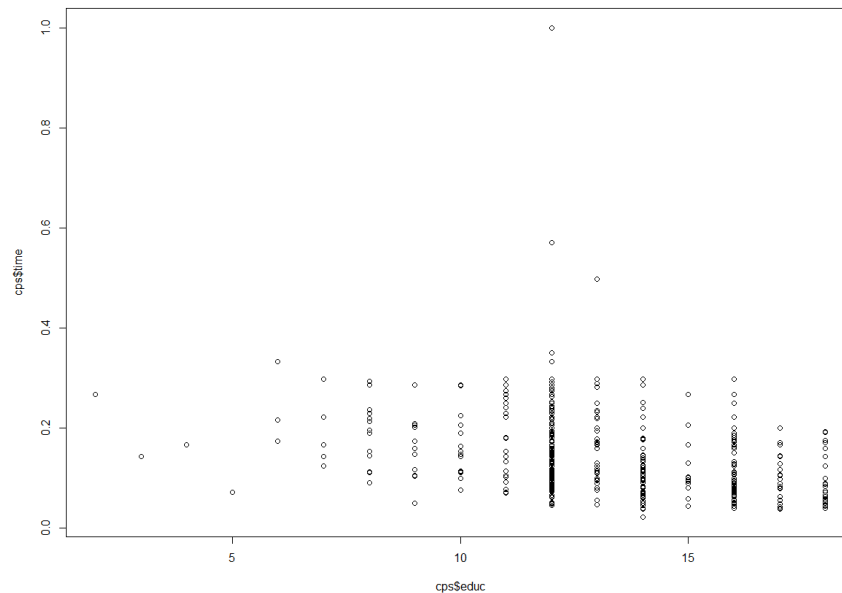
The residuals follow a normal distribution much better under a log transformation.



2(c) Create a new variable which is the inverse of wage. i.e. the amount of time (hours) it takes to earn a single dollar.

```
cps$time <- 1/cps$wage
```

2(d) Linear regression is appropriate with these variables, since the dependent variable is continuous, and we have a numerical independent variable.



2(e) Perform linear regression:

```
mod <- lm(cps$time ~ cps$educ)
plot(cps$educ, cps$time)
abline(mod, col='red', lwd=2)
summary(mod)
```

```
> summary(mod)

Call:
lm(formula = cps$time ~ cps$educ)

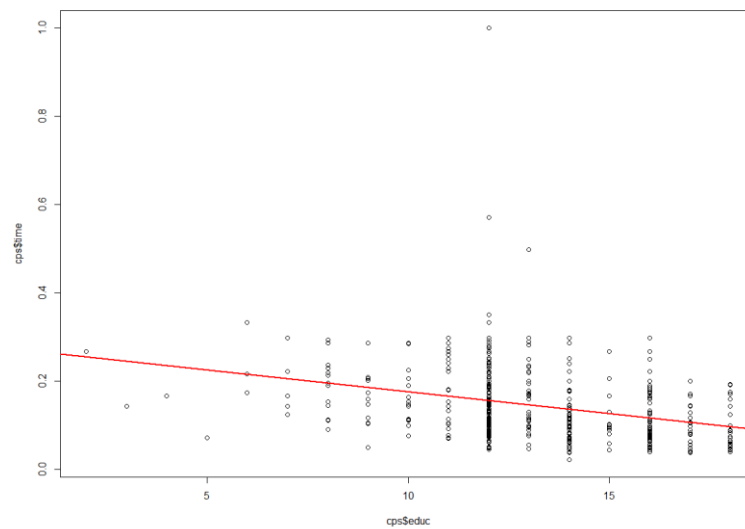
Residuals:
    Min       1Q   Median       3Q      Max
-0.15393 -0.05180 -0.02021  0.04361  0.84371

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept)  0.274700   0.017251  15.924 < 2e-16 ***
cps$educ    -0.009867   0.001299  -7.595 1.39e-13 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

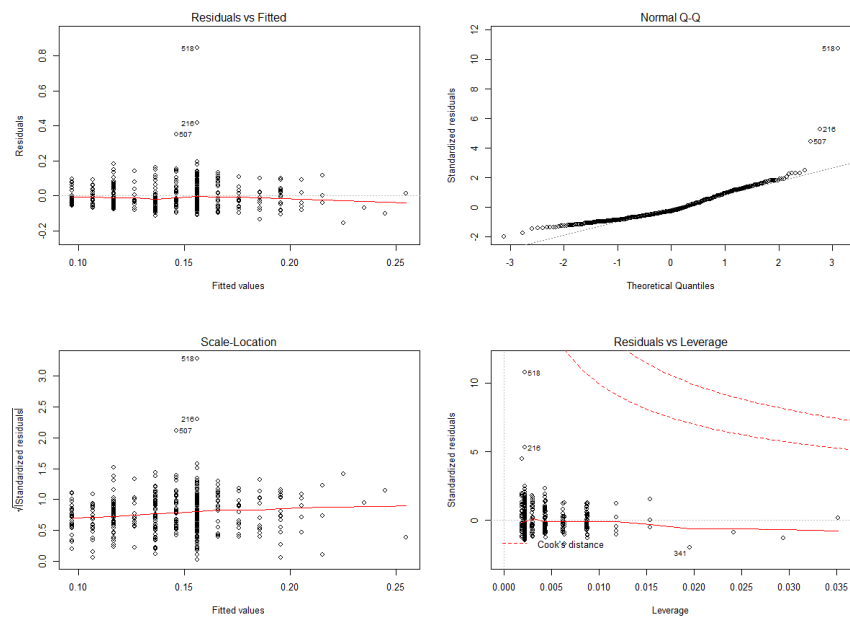
Residual standard error: 0.07845 on 532 degrees of freedom
Multiple R-squared:  0.09782,    Adjusted R-squared:  0.09613
F-statistic: 57.68 on 1 and 532 DF,  p-value: 1.393e-13
```

A linear model does describe the general relationship between amount of the hours it takes to earn one dollar and the years of educations. A p-value very close to zero provides enough evidence to reject the null hypothesis that the variable *time* is not related to the variable *education*. There is quite a bit of variability that the linear model does not account for though, as indicated by a low Adjusted R-squared. Generally, more years of education seem to be related to a shorter amount of time to earn one dollar. I

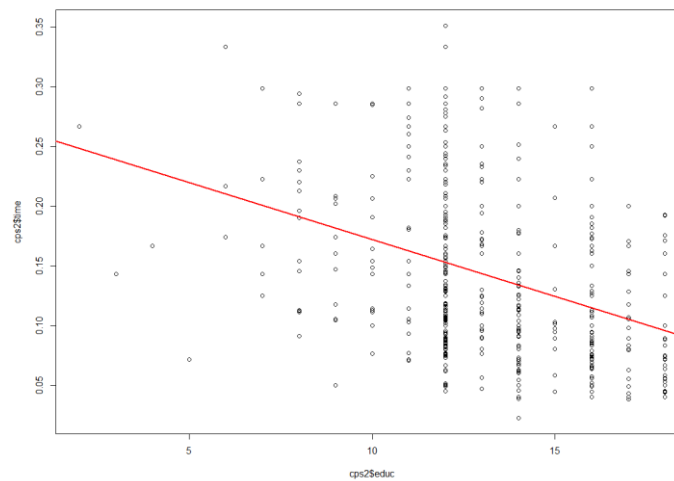
am not exactly convinced by this data that simply attending grad school will equate to more earnings (though I am happy with my decision to attend grad school for many other reasons).



2(f) By inspection of the diagnostic plots we can see that records 216, 507 and 518 are outliers.



We can try removing these from the plot to see how our analysis changes.



```
> summary(mod2)

Call:
lm(formula = cps2$time ~ cps2$educ)

Residuals:
    Min       1Q   Median       3Q      Max
-0.14805 -0.04856 -0.01807  0.04557  0.19798

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept)  0.267039   0.014387  18.561  <2e-16 ***
cps2$educ    -0.009512   0.001083  -8.783  <2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.06538 on 529 degrees of freedom
Multiple R-squared:  0.1273,    Adjusted R-squared:  0.1256
F-statistic: 77.13 on 1 and 529 DF,  p-value: < 2.2e-16
```

```
mod2 <- lm(cps2$time ~ cps2$educ)
plot(cps2$educ, cps2$time)
abline(mod2, col='red', lwd=2)
summary(mod2)
```

Removing the outliers further supports the relationship between years of education and time to earn a dollar. By removing the outliers from the analysis our Adjusted R-Squared increased to 0.13, which means there is still quite a bit of variance that the linear model is not explaining. The analysis doesn't really address whether attending graduate school specifically increases wages over not attending grad school. To answer this question, I would focus the analysis more on earnings of individuals with >16 years of education. This could be framed as a decision to stop education at 16 years (i.e. bachelor degree) or continue to grad school to go >16 years of education.

I am curious to know more about the outliers that we removed:

```
> cps[c(216, 507, 518),]
   wage educ race sex hispanic south married exper union age sector    time
216  1.75  12   W   F      NH      S Married    5   Not  23  service 0.5714286
507  2.01  13   W   M      NH      S Single    0   Not  19  service 0.4975124
518  1.00  12   W   M      NH     NS Married   24   Not  42   manag 1.0000000
```

These were very low wage earners with 12-13 years of education, i.e. neither a bachelor nor master's degree. While removing these outliers does improve the metrics of the linear model, it does not help address whether attending **grad school** (rather than stopping education at a bachelors degree) equates to increased earnings.