

# W271-2 – Spring 2016 – HW 2

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## Exercises

Complete the following exercises, following the best practices outlined in class. Place your answers in a written report (pdf, word, or jupyter notebook format) along with relevant R statements and output.

### Question 1

Load the `twoyear.RData` dataset and describe the basic structure of the data.

```
desc
```

```
##      variable                                label
## 1   female                                =1 if female
## 2   phsrank  % high school rank; 100 = best
## 3     BA                                =1 if Bachelor's degree
## 4     AA                                =1 if Associate's degree
## 5   black                                =1 if African-American
## 6  hispanic                                =1 if Hispanic
## 7     id                                ID Number
## 8   exper  total (actual) work experience
## 9     jc                                total 2-year credits
## 10    univ                                total 4-year credits
## 11   lwage                                log hourly wage
## 12  stotal  total standardized test score
## 13  smcity                                =1 if small city, 1972
## 14 medcity                                =1 if med. city, 1972
## 15  submed                                =1 if suburb med. city, 1972
```

```
## 16  lgcity          =1 if large city, 1972
## 17  sublg          =1 if suburb large city, 1972
## 18  vlgcity        =1 if very large city, 1972
## 19  subvlg         =1 if sub. very lge. city, 1972
## 20   ne            =1 if northeast
## 21   nc            =1 if north central
## 22   south         =1 if south
## 23  totcoll        jc + univ
```

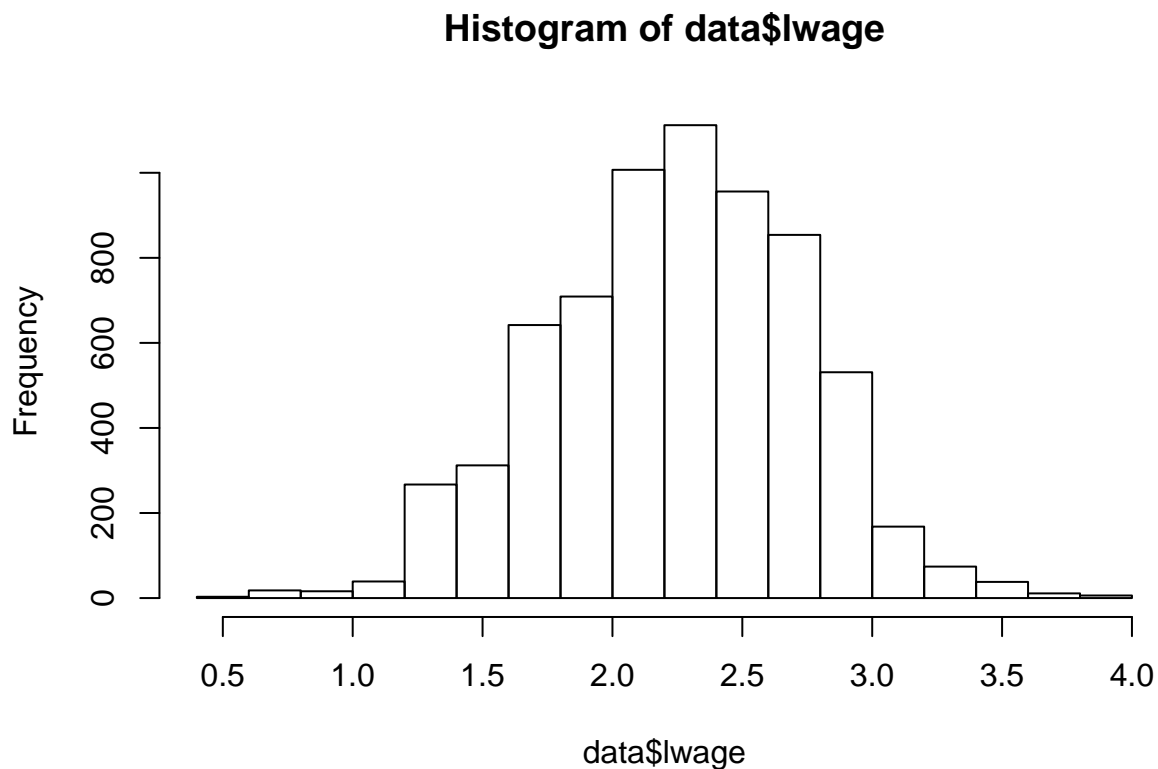
```
summary(data)
```

```
##      female      phsrank      BA      AA
##  Min.   :0.0000  Min.    : 0.00  Min.   :0.0000  Min.   :0.00000
## 1st Qu.:0.0000  1st Qu.:44.00  1st Qu.:0.0000  1st Qu.:0.00000
## Median :1.0000  Median :50.00  Median :0.0000  Median :0.00000
## Mean   :0.5196  Mean   :56.16  Mean   :0.3065  Mean   :0.04406
## 3rd Qu.:1.0000  3rd Qu.:76.00  3rd Qu.:1.0000  3rd Qu.:0.00000
## Max.   :1.0000  Max.   :99.00  Max.   :1.0000  Max.   :1.00000
##      black      hispanic      id      exper
##  Min.   :0.00000  Min.   :0.00000  Min.   :   19  Min.   :   3.0
## 1st Qu.:0.00000  1st Qu.:0.00000  1st Qu.:19372  1st Qu.:104.0
## Median :0.00000  Median :0.00000  Median :39301  Median :129.0
## Mean   :0.09508  Mean   :0.04687  Mean   :40616  Mean   :122.4
## 3rd Qu.:0.00000  3rd Qu.:0.00000  3rd Qu.:58842  3rd Qu.:149.0
## Max.   :1.00000  Max.   :1.00000  Max.   :89958  Max.   :166.0
##      jc      univ      lwage      stotal
##  Min.   :0.0000  Min.   :0.000  Min.   :0.5555  Min.   : -3.32480
## 1st Qu.:0.0000  1st Qu.:0.000  1st Qu.:1.9253  1st Qu.: -0.32734
## Median :0.0000  Median :0.200  Median :2.2763  Median : 0.00000
## Mean   :0.3389  Mean   :1.926  Mean   :2.2481  Mean   : 0.04748
## 3rd Qu.:0.0000  3rd Qu.:4.200  3rd Qu.:2.5969  3rd Qu.: 0.61079
## Max.   :3.8333  Max.   :7.500  Max.   :3.9120  Max.   : 2.23537
##      smcity      medcity      submed      lgcity
##  Min.   :0.0000  Min.   :0.0000  Min.   :0.00000  Min.   :0.00000
## 1st Qu.:0.0000  1st Qu.:0.0000  1st Qu.:0.00000  1st Qu.:0.00000
## Median :0.0000  Median :0.0000  Median :0.00000  Median :0.00000
## Mean   :0.2854  Mean   :0.1174  Mean   :0.06861  Mean   :0.09448
## 3rd Qu.:1.0000  3rd Qu.:0.0000  3rd Qu.:0.00000  3rd Qu.:0.00000
## Max.   :1.0000  Max.   :1.0000  Max.   :1.00000  Max.   :1.00000
##      sublg      vlgcity      subvlg      ne
##  Min.   :0.00000  Min.   :0.00000  Min.   :0.00000  Min.   :0.00000
## 1st Qu.:0.00000  1st Qu.:0.00000  1st Qu.:0.00000  1st Qu.:0.00000
## Median :0.00000  Median :0.00000  Median :0.00000  Median :0.00000
## Mean   :0.08709  Mean   :0.05855  Mean   :0.06358  Mean   :0.2107
## 3rd Qu.:0.00000  3rd Qu.:0.00000  3rd Qu.:0.00000  3rd Qu.:0.00000
## Max.   :1.00000  Max.   :1.00000  Max.   :1.00000  Max.   :1.00000
##      nc      south      totcoll
##  Min.   :0.0000  Min.   :0.0000  Min.   : 0.000
## 1st Qu.:0.0000  1st Qu.:0.0000  1st Qu.: 0.000
## Median :0.0000  Median :0.0000  Median : 1.507
## Mean   :0.2988  Mean   :0.3271  Mean   : 2.265
## 3rd Qu.:1.0000  3rd Qu.:1.0000  3rd Qu.: 4.367
## Max.   :1.0000  Max.   :1.0000  Max.   :10.067
```

```
#look at the wage variable  
summary(data$lwage)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.  
## 0.5555  1.9250  2.2760  2.2480  2.5970  3.9120
```

```
#look at the histogram to see if there are any potential extremes  
hist(data$lwage)
```



## Question 2

Typically, you will need to thoroughly analyze each of the variables in the data set using uni-variate, bivariate, and multivariate analyses before attempting any model. For this homework assume that this step has been conducted. Estimate the following regression:

$$\log(\text{wage}) = \beta_0 + \beta_1 \text{jc} + \beta_2 \text{univ} + \beta_3 \text{exper} + \beta_4 \text{black} + \beta_5 \text{hispanic} \\ + \beta_6 \text{AA} + \beta_7 \text{BA} + \beta_8 \text{exper} \cdot \text{black} + e$$

Interpret the coefficients  $\hat{\beta}_4$  and  $\hat{\beta}_8$ .

```
data$experXblack<-data$exper*data$black
model1<-lm(lwage~jc+univ+exper+black+hispanic+AA+BA+experXblack, data=data)
stargazer(model1, type="text")
```

```
##
## =====
##                               Dependent variable:
##                               -----
##                               lwage
## -----
## jc                            0.064***
##                               (0.008)
##
## univ                          0.073***
##                               (0.003)
##
## exper                         0.005***
##                               (0.0002)
##
## black                        0.033
##                               (0.061)
##
## hispanic                     -0.019
##                               (0.025)
##
## AA                           -0.008
##                               (0.030)
##
## BA                           0.018
##                               (0.016)
##
## experXblack                  -0.001**
##                               (0.0005)
##
## Constant                     1.477***
##                               (0.022)
## -----
## Observations                  6,763
## R2                           0.228
## Adjusted R2                  0.227
## Residual Std. Error          0.429 (df = 6754)
## F Statistic                   249.553*** (df = 8; 6754)
## =====
## Note:                        *p<0.1; **p<0.05; ***p<0.01
```

```
coeftest(model1, vcov = vcovHC)
```

```
##
## t test of coefficients:
##
##           Estimate Std. Error t value Pr(>|t|)
```

```
## (Intercept)  1.47733155  0.02293512 64.4135 < 2e-16 ***
## jc          0.06379261  0.00761208  8.3804 < 2e-16 ***
## univ        0.07328063  0.00336598 21.7709 < 2e-16 ***
## exper       0.00502341  0.00016840 29.8294 < 2e-16 ***
## black       0.03317088  0.06872723  0.4826 0.62936
## hispanic    -0.01936289  0.02498704 -0.7749 0.43842
## AA          -0.00777589  0.02746594 -0.2831 0.77710
## BA          0.01767355  0.01656455  1.0670 0.28603
## experXblack -0.00126790  0.00053779 -2.3576 0.01842 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

### Question 3

With this model, test that the return to university education is 7%

```
coeffs<-coefficients(model1)
coeffs[3]
```

```
##          univ
## 0.07328063
```

The coefficient for the univ variable is .073 which equates to approximately a 7% increase in log(wage) for every increment increase in univ education.

### Question 4

With this model, test that the return to junior college education is equal for black and non-black.

### Question 5

With this model, test whether the return to university education is equal to the return to 1 year of working experience.

### Question 6

Test the overall significance of this regression.

```
summary(model1)
```

```
##
## Call:
## lm(formula = lwage ~ jc + univ + exper + black + hispanic + AA +
##      BA + experXblack, data = data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.11612 -0.27836  0.00432  0.28676  1.76811
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  1.4773315  0.0223780  66.017 < 2e-16 ***
## jc           0.0637926  0.0079034   8.072 8.15e-16 ***
## univ         0.0732806  0.0031486  23.274 < 2e-16 ***
## exper        0.0050234  0.0001667  30.141 < 2e-16 ***
## black        0.0331709  0.0613984   0.540  0.5890
## hispanic     -0.0193629  0.0248914  -0.778  0.4367
## AA           -0.0077759  0.0295497  -0.263  0.7924
## BA            0.0176735  0.0156553   1.129  0.2590
## experXblack -0.0012679  0.0004991  -2.541  0.0111 *
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4287 on 6754 degrees of freedom
## Multiple R-squared:  0.2282, Adjusted R-squared:  0.2272
## F-statistic: 249.6 on 8 and 6754 DF,  p-value: < 2.2e-16
```

The overall significance of this regression is high with a F-statistic of 249.6 and an overall p-value  $< 2.2e-16$  indicating that overall this model is performing better than random. However, we can likely increase our  $r$  squared value by eliminating some of the factors because several factors individually are non significant.

## Question 7

including a square term of working experience to the regression model built above, estimate the linear regression model again. What is the estimated return to work experience in this model?

```
data$experSq<-data$exper^2
model2<-lm(lwage~jc+univ+exper+black+hispanic+AA+BA+experXblack+experSq, data=data)
stargazer(model2, type="text")
```

```
##
## =====
##                               Dependent variable:
##                               -----
##                               lwage
## -----
## jc                               0.064***
##                               (0.008)
##
## univ                             0.074***
##                               (0.003)
##
## exper                             0.004***
##                               (0.001)
##
## black                             0.030
##                               (0.062)
##
## hispanic                         -0.019
##                               (0.025)
##
## AA                               -0.008
##                               (0.030)
##
## BA                               0.018
##                               (0.016)
##
## experXblack                      -0.001**
##                               (0.001)
##
## experSq                          0.00000
##                               (0.00000)
```

```
##
## Constant          1.510***
##                  (0.044)
##
## -----
## Observations      6,763
## R2                 0.228
## Adjusted R2       0.227
## Residual Std. Error 0.429 (df = 6753)
## F Statistic       221.898*** (df = 9; 6753)
## =====
## Note:              *p<0.1; **p<0.05; ***p<0.01
```

```
coeftest(model2, vcov = vcovHC)
```

```
##
## t test of coefficients:
##
##      Estimate Std. Error t value Pr(>|t|)
## (Intercept)  1.5101e+00 4.3591e-02 34.6427 < 2.2e-16 ***
## jc           6.4168e-02 7.6224e-03  8.4183 < 2.2e-16 ***
## univ         7.3819e-02 3.4501e-03 21.3963 < 2.2e-16 ***
## exper        4.3008e-03 8.4541e-04  5.0873 3.731e-07 ***
## black        2.9937e-02 6.8436e-02  0.4374  0.66180
## hispanic     -1.9317e-02 2.4985e-02 -0.7731  0.43947
## AA           -7.5392e-03 2.7481e-02 -0.2743  0.78383
## BA           1.7967e-02 1.6579e-02  1.0837  0.27853
## experXblack  -1.2388e-03 5.3539e-04 -2.3139  0.02071 *
## experSq      3.3790e-06 3.8745e-06  0.8721  0.38318
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

## Question 8

Provide the diagnosis of the homoskedasticity assumption. Does this assumption hold? If so, how does it affect the testing of no effect of university education on salary change? If not, what potential remedies are available?

```
plot(model2, which=1)
```

The homoskedasticity assumption may hold in this case. Although the residuals vs fitted plot from the model appears to have a little less variance on the edges (far right and far left) the red line is relatively straight and flat.



