Discussion

# Estimation of Wage Equations: An Application of MLR

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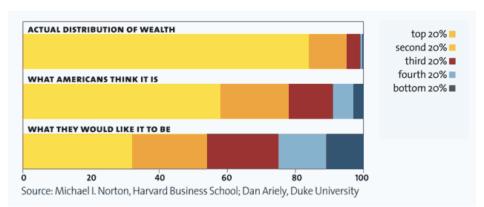
Statistics Department(STAT 705)

April 26, 2019

### Outline of Presentation

- Goal and Statistical Tool
- 2 Data
- Results
- 4 Discussion
- 5 Validation and Reference

### Income Equality or Inequality????



## What's Multiple Linear Regression (MLR)?

Suppose a researcher wants to answer the following questions;

What is the effect of gender, race, education and experience on hourly earning? Do these variables interact?

What is the un/adjusted average wage gap between males and females?

Does wage gap suggest gender wage discrimination?

What is the rate of return to schooling?

What is the estimated effect of a unit increase in experience for an individual with 20 years of experience?

### What's Multiple Linear Regression (MLR)?

We can answer the above questions using Multiple Linear Regression!

MLR is a statistical technique that uses several explanatory variables to predict the outcome of a response variable. MLR is an extension of linear (OLS) regression that uses just one explanatory variable.

In this presentation, we employ this simple but yet powerful statistical tool to answer the questions posed above.

# 2014 monthly Current Population Survey Outgoing Rotation Group (CPS-ORG)

The data set contains 137,374 non-student wage and salary workers, ages 18 to 65, in 2014, extracted from the Census Bureau's Current Population Survey Outgoing Rotation Group files (i.e., the 14 sample asked earnings and union questions each month). Define new variables;

- hrearn = hourly earnings
- femalemarried = female\*married
- exper = yrs of potential experience.MIN([age-school-6]or[age-18])
- expersquare = the square of years of potential experience

# Variable Description

Data

school	Years of schooling completed (0-20)
age	18 to 65
female	1=females, 0=males
black	1=black, 0=other race
hispanic	1=Hispanic; 0=other ethnicity
forborn	1=foreign born (includes citizens and non-citizens); 0=native born
married	1=married and spouse present; 0=ever married, spouse not present or never
union	1=union member; 0=not a member
wkearn	Usual weekly earnings (\$1 to estimated mean above the topcode)
wkhours	Usual hours worked per week (1-99)
parttime	1=less than 35 usual hours worked per week; 0=35+ usual hours

Data

#### Consider the Models:

log(hrearn) = B0 + B1school + B2exper + B3exper + B4female + B5married + e

#### **AND**

log(hrearn)=B0+B1school+B2exper+B3exper2+B4female +B5married+B6femalemarried+e

	Goal and Statistical Tool	Data	Results	Discussion	Validation and Reference
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school	age	female	black	hispanic	forborn
Min. : 0.0	Min. :18	Min. :0.00	Min. :0.00	Min. :0.00	Min. :0.00
1st Qu.:12.0	1st Qu.:31	1st Qu.:0.00	1st Qu.:0.00	1st Qu.:0.00	1st Qu.:0.00
Median :13.3	Median :42	Median :0.00	Median :0.00	Median :0.00	Median :0.00
Mean :14.0	Mean :42	Mean :0.49	Mean :0.09	Mean :0.13	Mean :0.14
3rd Qu.:16.0	3rd Qu.:52	3rd Qu.:1.00	3rd Qu.:0.00	3rd Qu.:0.00	3rd Qu.:0.00
Max. :20.0	Max. :65	Max. :1.00	Max. :1.00	Max. :1.00	Max. :1.00

married	union	wkearn
Min. :0.00	Min. :0.00	Min. : 3
1st Qu.:0.00	1st Qu.:0.00	1st Qu.: 464
Median :1.00	Median :0.00	Median : 750
Mean :0.56	Mean :0.12	Mean :1008
3rd Qu.:1.00	3rd Qu.:0.00	3rd Qu.:1191
Max. :1.00	Max. :1.00	Max. :5940

wkhours	parttime	hrearn	exper	expersquare	femalemarried
Min. : 1	Min. :0.00	Min. : 3	Min. : 0	Min. : 0	Min. :0.00
1st Qu.:40	1st Qu.:0.00	1st Qu.: 12	1st Qu.:11	1st Qu.: 121	1st Qu.:0.00
Median :40	Median :0.00	Median : 19	Median :22	Median : 471	Median :0.00
Mean :40	Mean :0.14	Mean : 24	Mean :22	Mean : 624	Mean :0.26
3rd Qu.:40	3rd Qu.:0.00	3rd Qu.: 29	3rd Qu.:32	3rd Qu.:1024	3rd Qu.:1.00
Max. :99	Max. :1.00	Max. :297	Max. :47	Max. :2209	Max. :1.00

```
Call:
lm(formula = log(hrearn) ~ school + exper + expersquare + female +
             married)
```

#### Residuals:

Min 10 Median 30 Max -2.795 - 0.336 - 0.027 0.301 3.445

#### Coefficients:

```
Estimate Std. Error t value Pr(>|t|)
(Intercept) 1.05825796 0.00883214 119.8 < 0.0000000000000000 ***
school 0.11397895 0.00055287 206.2 <0.0000000000000002 ***
exper 0.03138694 0.00045851
                             expersquare -0.00050662 0.00000998
                            -50.8 < 0.0000000000000000 ***
female -0.22631766 0.00284662
                            -79.5 <0.0000000000000000 ***
married 0.10458052 0.00306233
```

```
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Residual standard error: 0.53 on 137368 degrees of freedom Multiple R-squared: 0.308, Adjusted R-squared: 0.308

```
Call:
lm(formula = log(hrearn) \sim school + exper + expersquare + female +
   married + femalemarried)
Residuals:
  Min
         10 Median 30
                         Max
-2.814 - 0.335 - 0.029 0.299 3.471
Coefficients:
              Estimate
                      Std. Error t value Pr(>|t|)
(Intercept) 1.03405525
                      0.00891359
                                 school 0.11373890 0.00055230 205.9 <0.0000000000000000 ***
exper 0.03114227 0.00045810 68.0 <0.000000000000000 ***
expersquare -0.00050450
                      0.00000997 -50.6 < 0.0000000000000000 ***
female -0.16618417 0.00427148 -38.9 <0.0000000000000000 ***
married 0.15887739 0.00419997 37.8 <0.000000000000002 ***
femalemarried -0.10774463
                      0.00571207
                                 ___
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 0.52 on 137367 degrees of freedom
```

Multiple R-squared: 0.31, Adjusted R-squared: 0.31 F-statistic: 1.03e+04 on 6 and 137367 DF, p-value: <0.0000000000000000

Discussion

### Unadjusted Wage differential??

One may say to find gender wage differential,

Simply calculate the mean wage differential between men and women:

The unadjusted wage gap is 27.195 - 21.545 = 5.6

Goal and Statistical Tool

What is the rate of return to sch.? The rate of return to schooling is 11.398 pct

What is the approx. pct difference between women and men? The approx. pct difference between women and men is 22.632 pct, on average.

What is the exact pct difference between women and men? The exact pct difference between women and men is  $[\exp(-0.22631766)-1]*100 = 20.254$  pct, on average.

Does the coef. est. on female suggest gender wage discrimination? If the model is correctly specified model, the coef. on female is suggestive of gender wage discrimination. However??

## Some findings

Goal and Statistical Tool

What is the est. effect of a unit increase in exper. for an individual with 20yrs of exper.? The est. increase is [0.0313869 -(0.0005066\*20\*2)] \*100 = 1.11229pct.

Discussion

Does exper. have an increasing or decreasing effect on hourly earnings? Exper. has a decreasing effect on hourly wages because of the neg. sign on the coef. on exper. squared.

Compute the turnaround point for the effect of exper. on hourly earnings? The turnaround point is given by 0.03114227/ (2\*0.00050450) = 30.97 yrs.

Are hourly earnings minimized or maximized at this point? Hourly wages are maximized at this point.

Discussion

### Some findings

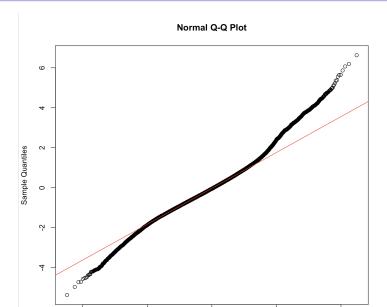
What is the omitted category in the 2 Model? Single males

What is the approx. pct wage difference b/tn married females and the single male? The approx. pct difference is -0.1661842 + 0.1588774 - 0.1077446 = -0.1150514 (11.505pct lower).

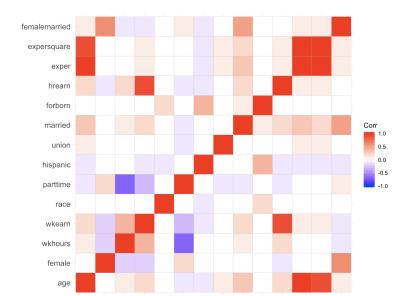
What is the interpretation of the coef. est. on femalemarried? The coefficient estimate on females suggests that marriage premium differs by gender by approx. 10.774pct. That is, the percentage hourly wage differential between the married and single is lower for women relative to men by approximately 10.774pct.

Construct a 99pct. CI for the coef. est. on femalemarried. -0.1077446 + (2.576\*0.0057121) = (-0.12245897, -0.09303023)

# QQ Plot



### Correlation Matrix



### reference

http://cepr data.org/cps-uniform-data-extracts/cps-outgoing-rotation-group/.

 ${\sf R} \ {\sf Code} \ {\sf available} \ {\sf upon} \ {\sf request.} \\ ({\sf Email:davident85@gmail.com})$ 

# Thank You!

