

# Labor Market Analysis

## Revised Econometric Model

*Marjorie Blanco, Joe Thomson, Haodi Tu*

### Research Questions

- 1) How do earnings vary by education level?
- 2) How does the premium for education vary by gender?

### Revised Model

$$\text{Earning} = \beta_0 + \text{Divorced} * \beta_1 + \text{NeverMarried} * \beta_2 + \text{Female} * \beta_3 + \text{RaceBlack} * \beta_4 + \text{RaceOther} * \beta_5 + \text{SomeCollege} * \beta_6 + \text{Associate} * \beta_7 + \text{Bachelor} * \beta_8 + \text{Master} * \beta_9 + \text{Professional} * \beta_{10} + \text{Doctoral} * \beta_{11} + \text{Age} * \beta_{12}$$

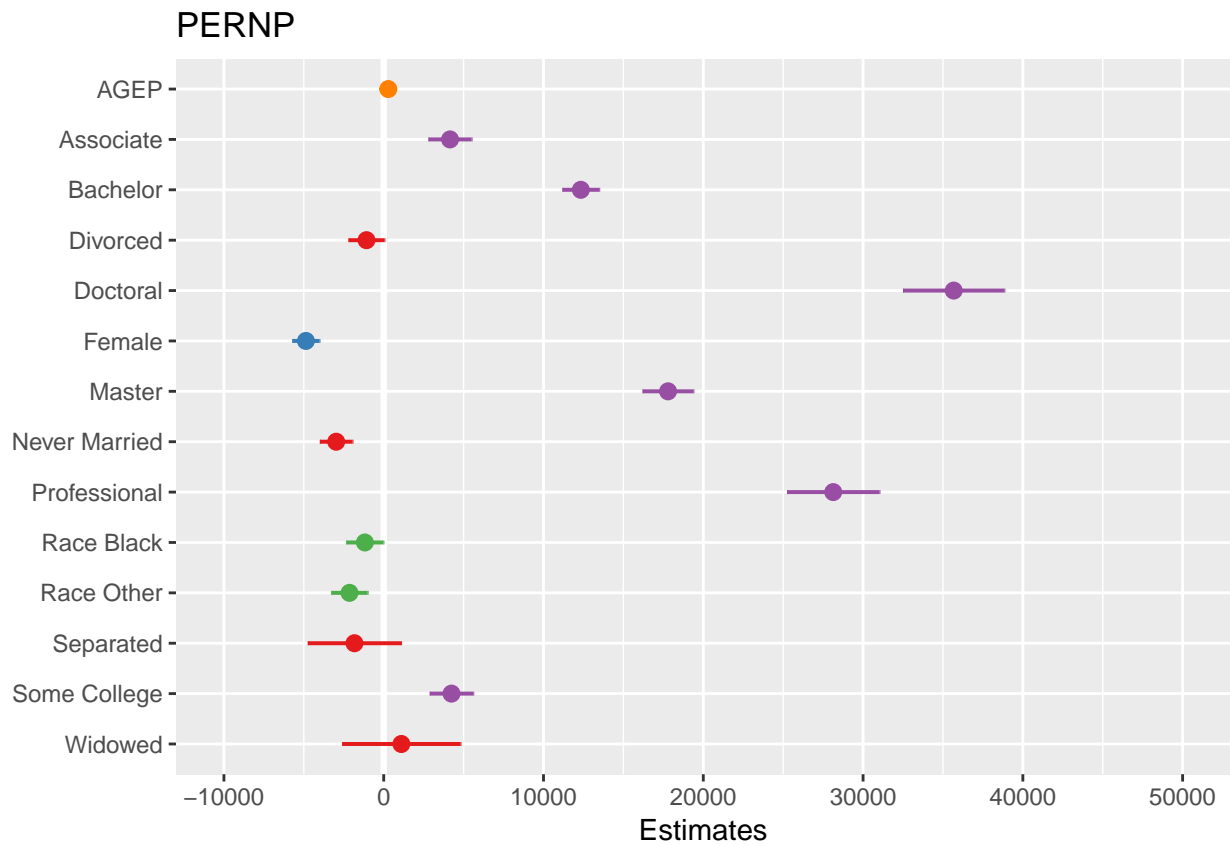
```
##
## Call:
## lm(formula = PERNP ~ Widowed + Divorced + Separated + NeverMarried +
##      Female + RaceBlack + RaceOther + SomeCollege + Associate +
##      Bachelor + Master + Professional + Doctoral + AGE, data = ss16ppr)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -43683  -9228  -2919   5341   98762
##
## Coefficients:
##              Estimate Std. Error t value      Pr(>|t|)
## (Intercept)  12598.97    1110.28  11.348 < 0.0000000000000002 ***
## Widowed       1104.92    1888.02   0.585     0.558422
## Divorced     -1078.74     571.98  -1.886     0.059353 .
## Separated    -1833.17    1489.22  -1.231     0.218393
## NeverMarried -2974.75     520.48  -5.715  0.00000001155 ***
## Female      -4868.47     432.66 -11.252 < 0.0000000000000002 ***
## RaceBlack   -1178.01     591.36  -1.992     0.046421 *
## RaceOther   -2141.04     579.98  -3.692     0.000225 ***
## SomeCollege  4241.63     694.73   6.105  0.00000000110 ***
## Associate    4150.43     688.45   6.029  0.00000000177 ***
## Bachelor    12337.16     587.07  21.015 < 0.0000000000000002 ***
## Master       17795.40     812.61  21.899 < 0.0000000000000002 ***
## Professional 28133.37    1476.36  19.056 < 0.0000000000000002 ***
## Doctoral     35674.46    1617.08  22.061 < 0.0000000000000002 ***
## AGE          284.94       21.12  13.488 < 0.0000000000000002 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 14920 on 5179 degrees of freedom
## Multiple R-squared:  0.2466, Adjusted R-squared:  0.2446
## F-statistic: 121.1 on 14 and 5179 DF,  p-value: < 0.00000000000000022
```

### Interpretation

- Coefficients Explanation

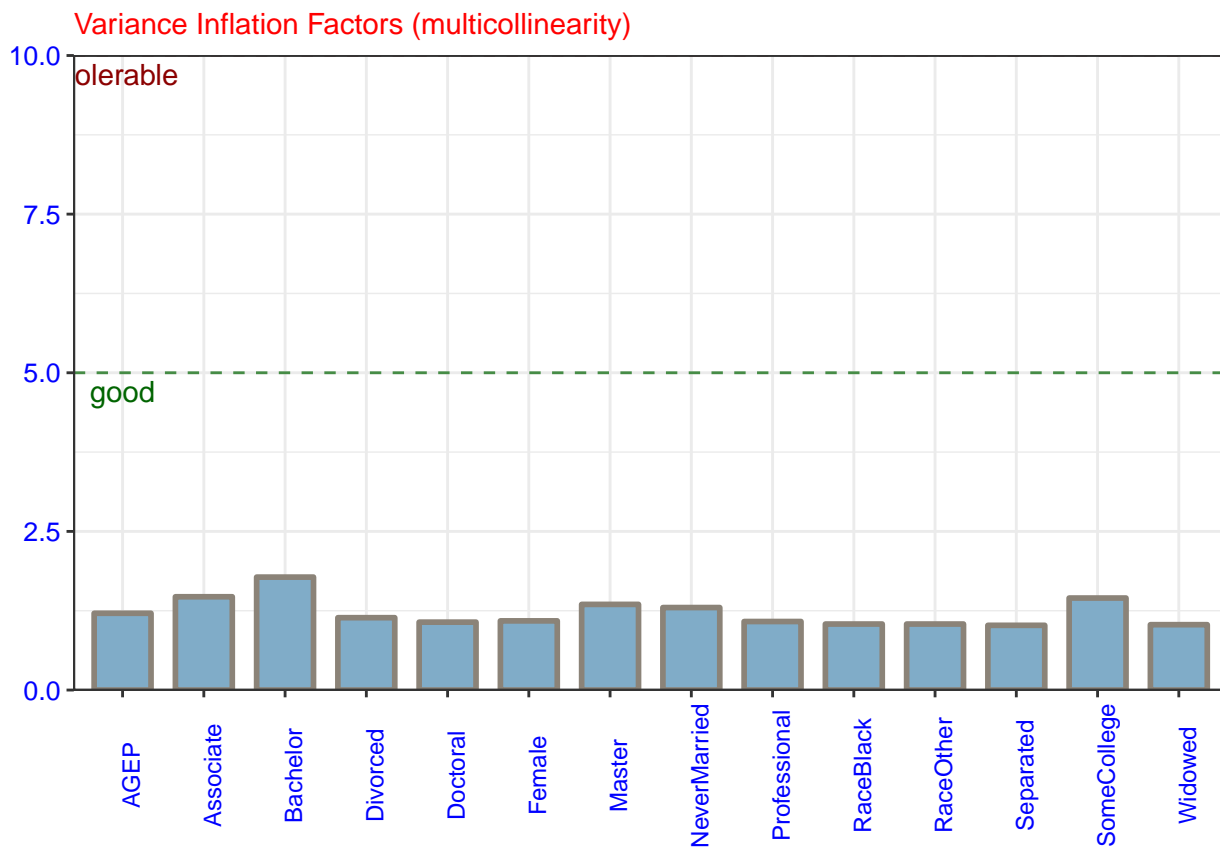
- Holding gender, race, education and age constant, married people makes \$1104.92 less than people who widowed on average.
- Holding gender, race, education and age constant, married people makes \$1078.74 more than people who divorced on average.
- Holding gender, race, education and age constant, married people makes \$1833.17 more than people who separated on average.
- Holding gender, race, education and age constant, married people makes \$2974.75 more than people who never married on average.
- Holding marriage, race, education and age constant, male makes \$4868.47 more than female on average.
- Holding marriage, gender, education and age constant, White makes \$1178.01 more than Black on average.
- Holding marriage, gender, education and age constant, White makes \$2141.04 more than Other race on average.
- Holding marriage, gender, race and age constant, people have high school education makes \$4241.63 less than people have some college education on average.
- Holding marriage, gender, race and age constant, people have high school education makes \$4150.43 less than people have associate education on average.
- Holding marriage, gender, race and age constant, people have high school education makes \$12337.16 less than people have bachelor's degree on average.
- Holding marriage, gender, race and age constant, people have high school education makes \$17795.4 less than people have master's degree on average.
- Holding marriage, gender, race and age constant, people have high school education makes \$28133.37 less than people have Professional education on average.
- Holding marriage, gender, race and age constant, people have high school education makes \$35674.46 less than people have doctor's degree on average.
- Holding marriage, gender, race and education constant, people make \$284.94 more as age increases on average between the age of 18 to 64.

## Model Diagnostics



Variables	Tolerance	VIF
Widowed	0.9729050	1.027850
Divorced	0.8808670	1.135245
Separated	0.9756434	1.024965
NeverMarried	0.7702636	1.298257
Female	0.9159123	1.091808
RaceBlack	0.9633247	1.038072
RaceOther	0.9650376	1.036229
SomeCollege	0.6878124	1.453885
Associate	0.6814727	1.467410
Bachelor	0.5604312	1.784341
Master	0.7394234	1.352405
Professional	0.9238792	1.082393
Doctoral	0.9320203	1.072938
AGEP	0.8267426	1.209566

```
## [[1]]
```

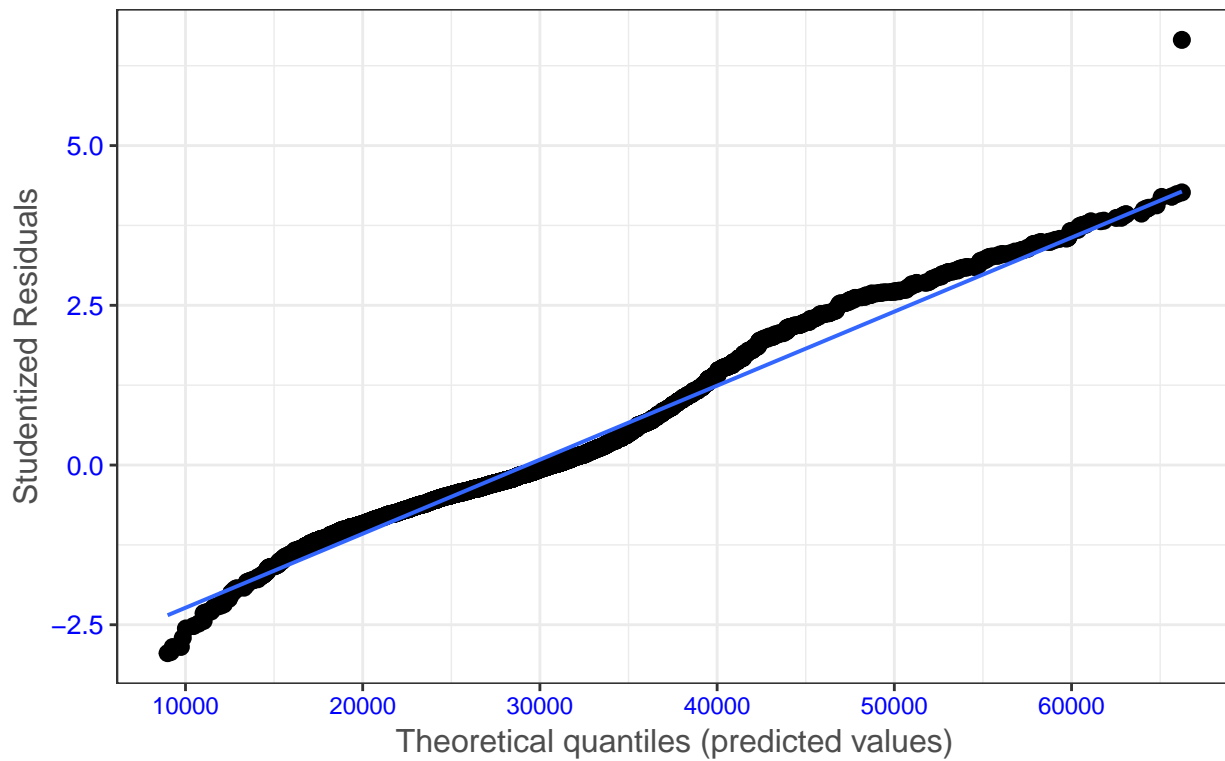


A VIF chart shows that there is no correlation among the kth predictor and the remaining predictor variables.

```
## [[1]]
```

### Non-normality of residuals and outliers

Dots should be plotted along the line

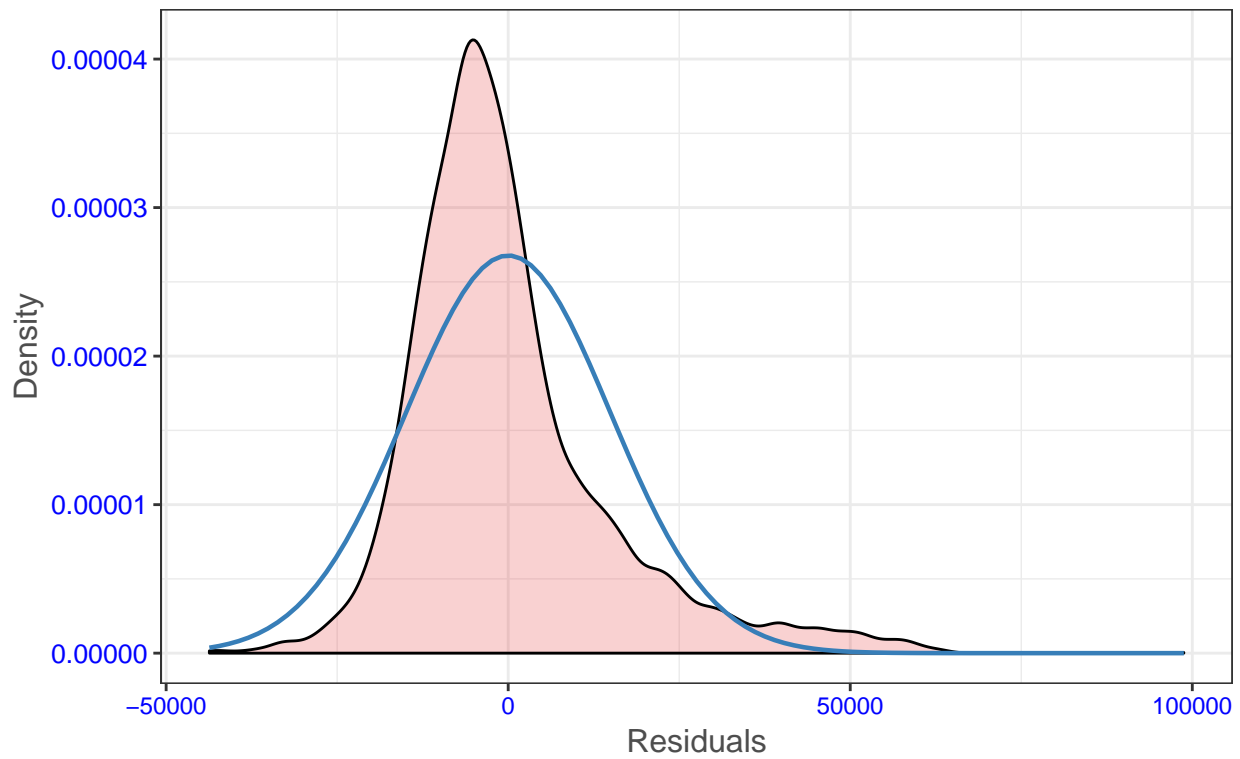


The plot shows that the distribution of the residuals are not approximately normally distributed. The “thin tail” correspond to the first quantiles occurring at less than expected values and there is an outlier point.

```
## [[1]]
```

### Non-normality of residuals

Distribution should look like normal curve

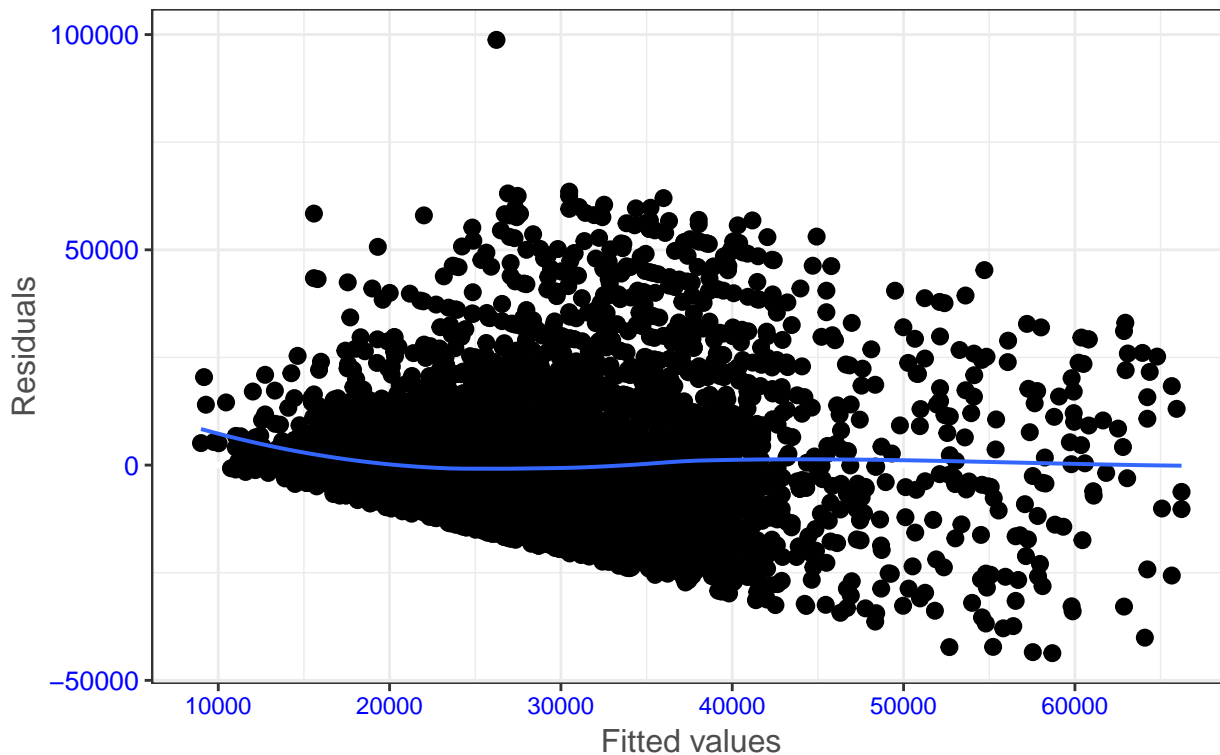


The plot shows that the distribution of the residuals is not approximately normally distributed but skewed to the right.

```
## [[1]]
```

### Homoscedasticity (constant variance of residuals)

Amount and distance of points scattered above/below line is equal or randomly spread



The plot shows heteroscedasticity as the variance of the residuals are not constant.

### Breusch-Pagan test

$H_0$ : Variance is unchanging in the residual

$H_1$ : Variance is changing in the residual

```
##
## studentized Breusch-Pagan test
##
## data: Model_1
## BP = 297.55, df = 14, p-value < 0.00000000000000022
```

The p-Value < 0.05 which indicates that the null hypothesis (the variance is unchanging in the residual) can be rejected and therefore heteroscedasticity exists.

### Revised Model with robust standard errors

```
##
## Call:
## lm(formula = residuals(Model_1) * residuals(Model_1) ~ Widowed +
##     Divorced + Separated + NeverMarried + Female + RaceBlack +
##     RaceOther + SomeCollege + Associate + Bachelor + Master +
##     Professional + Doctoral + AGEp, data = ss16ppr)
##
## Residuals:
```

```

##           Min           1Q           Median           3Q           Max
## -620286278 -187332480 -89004449 -6623206 9554893533
##
## Coefficients:
##           Estimate Std. Error t value      Pr(>|t|)
## (Intercept) -22272288  35096119 -0.635      0.5257
## Widowed     10526294  59680719  0.176      0.8600
## Divorced    -9634464  18080343 -0.533      0.5941
## Separated   -74925738  47074515 -1.592      0.1115
## NeverMarried -41219294  16452372 -2.505      0.0123 *
## Female     -86614125  13676418 -6.333 0.00000000026057555 ***
## RaceBlack  -30492334  18693117 -1.631      0.1029
## RaceOther  -56411117  18333200 -3.077      0.0021 **
## SomeCollege 54139584  21960626  2.465      0.0137 *
## Associate   38592334  21762198  1.773      0.0762 .
## Bachelor   167802881  18557574  9.042 < 0.0000000000000002 ***
## Master     240588080  25686793  9.366 < 0.0000000000000002 ***
## Professional 363259303  46668182  7.784 0.00000000000000843 ***
## Doctoral   283832422  51116425  5.553 0.00000002953675723 ***
## AGE        4843548    667752    7.254 0.00000000000046605 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 471600000 on 5179 degrees of freedom
## Multiple R-squared:  0.05729, Adjusted R-squared:  0.05474
## F-statistic: 22.48 on 14 and 5179 DF, p-value: < 0.00000000000000022

##           (Intercept)      Widowed      Divorced      Separated      NeverMarried
## (Intercept)  1007933.05   58087.062 -81847.4203 -49419.377 -267638.579
## Widowed      58087.06  3995751.368 105877.9224 111739.704 61877.649
## Divorced     -81847.42  105877.922 349056.7354 102495.813 95360.449
## Separated    -49419.38  111739.704 102495.8126 1565899.963 91581.434
## NeverMarried -267638.58  61877.649 95360.4487 91581.434 242152.106
## Female      -24269.36 -37296.786 -22584.5663 -9502.521 3535.426
## RaceBlack   -54050.20  4154.512 -261.2336 -1778.827 -1474.294
## RaceOther   -43361.05 -17198.587 3064.7362 28165.922 -2030.508
## SomeCollege -164587.00  5864.731 2716.6330 -30536.341 -9191.626
## Associate   -156276.57  28238.641 14270.5777 26369.734 16901.438
## Bachelor    -140253.26  58869.676 15822.7991 22192.750 11420.935
## Master      -98238.23  19298.737 4166.7095 -29561.887 4120.858
## Professional -178405.52  47377.351 -9202.0646 36275.425 -25373.156
## Doctoral    -51307.89  115905.356 47855.5908 140337.165 29949.062
## AGE         -18128.79 -3587.834 -293.0243 -1156.124 3781.175
##
##           Female      RaceBlack      RaceOther      SomeCollege      Associate
## (Intercept) -24269.3559 -54050.20464 -43361.053 -164587.002 -156276.572
## Widowed     -37296.7857  4154.51196 -17198.587 5864.731 28238.641
## Divorced    -22584.5663 -261.23364 3064.736 2716.633 14270.578
## Separated   -9502.5212 -1778.82702 28165.922 -30536.341 26369.734
## NeverMarried 3535.4265 -1474.29397 -2030.508 -9191.626 16901.438
## Female      194048.3723 -8630.04524 -8397.285 -29487.201 -42117.858
## RaceBlack   -8630.0452 315761.19749 66096.740 -10264.752 -6036.723
## RaceOther   -8397.2851 66096.74016 278637.804 3569.363 -4820.559
## SomeCollege -29487.2014 -10264.75223 3569.363 332091.081 125594.576
## Associate   -42117.8578 -6036.72309 -4820.559 125594.576 319016.024
## Bachelor    -78511.6369 6202.41918 -10349.806 130275.001 135533.527
## Master      -92998.9918 -6855.66175 28720.700 131019.704 135614.524
## Professional -48793.0798 28726.41028 17513.253 130357.805 129838.545

```



```

## Doctoral      -37589.1845    758.93490  80827.611  112338.880  124924.530
## AGEP          -787.2688     -97.32446   -420.435   1437.587   1046.660
##              Bachelor      Master Professional      Doctoral
## (Intercept) -140253.2646 -98238.2349 -178405.520 -51307.8937
## Widowed      58869.6759  19298.7370   47377.351  115905.3561
## Divorced     15822.7991   4166.7095   -9202.065   47855.5908
## Separated    22192.7500 -29561.8868   36275.425  140337.1654
## NeverMarried 11420.9348   4120.8577  -25373.156  29949.0622
## Female       -78511.6369 -92998.9918  -48793.080 -37589.1845
## RaceBlack     6202.4192  -6855.6618   28726.410   758.9349
## RaceOther    -10349.8064  28720.6998   17513.253   80827.6113
## SomeCollege  130275.0011 131019.7043  130357.805  112338.8799
## Associate     135533.5269 135614.5236  129838.545  124924.5305
## Bachelor     307695.2759 152567.2493  140083.610  133715.9622
## Master       152567.2493 847427.5870  140245.690  140031.3156
## Professional 140083.6102 140245.6899 4394423.380 124345.5782
## Doctoral     133715.9622 140031.3156 124345.578 4770075.3429
## AGEP         949.8206    127.4089    1732.881   -2150.9231
##              AGEP
## (Intercept) -18128.78819
## Widowed     -3587.83358
## Divorced    -293.02429
## Separated   -1156.12421
## NeverMarried 3781.17492
## Female      -787.26885
## RaceBlack   -97.32446
## RaceOther   -420.43504
## SomeCollege 1437.58685
## Associate   1046.66000
## Bachelor    949.82059
## Master      127.40891
## Professional 1732.88105
## Doctoral    -2150.92312
## AGEP        413.64705

## (Intercept)      Widowed      Divorced      Separated NeverMarried
## 1003.95869      1998.93756      590.81024      1251.35925      492.08953
##      Female      RaceBlack      RaceOther SomeCollege      Associate
## 440.50922      561.92633      527.86154      576.27344      564.81504
##      Bachelor      Master Professional      Doctoral      AGEP
## 554.70287      920.55830      2096.28800      2184.05022      20.33831

```

## Revised Logarithmic Model

$$\text{Log}(\text{Earning}) = \beta_0 + \text{Divorced} * \beta_1 + \text{NeverMarried} * \beta_2 + \text{Female} * \beta_3 + \text{RaceBlack} * \beta_4 + \text{RaceOther} * \beta_5 + \text{SomeCollege} * \beta_6 + \text{Associate} * \beta_7 + \text{Bachelor} * \beta_8 + \text{Master} * \beta_9 + \text{Professional} * \beta_{10} + \text{Doctoral} * \beta_{11} + \text{Age} * \beta_{12}$$

```

##
## Call:
## lm(formula = log(PERNP, base = exp(1)) ~ Widowed + Divorced +
##      Separated + NeverMarried + Female + RaceBlack + RaceOther +
##      SomeCollege + Associate + Bachelor + Master + Professional +
##      Doctoral + AGEP, data = ss16ppr)
##
## Residuals:

```

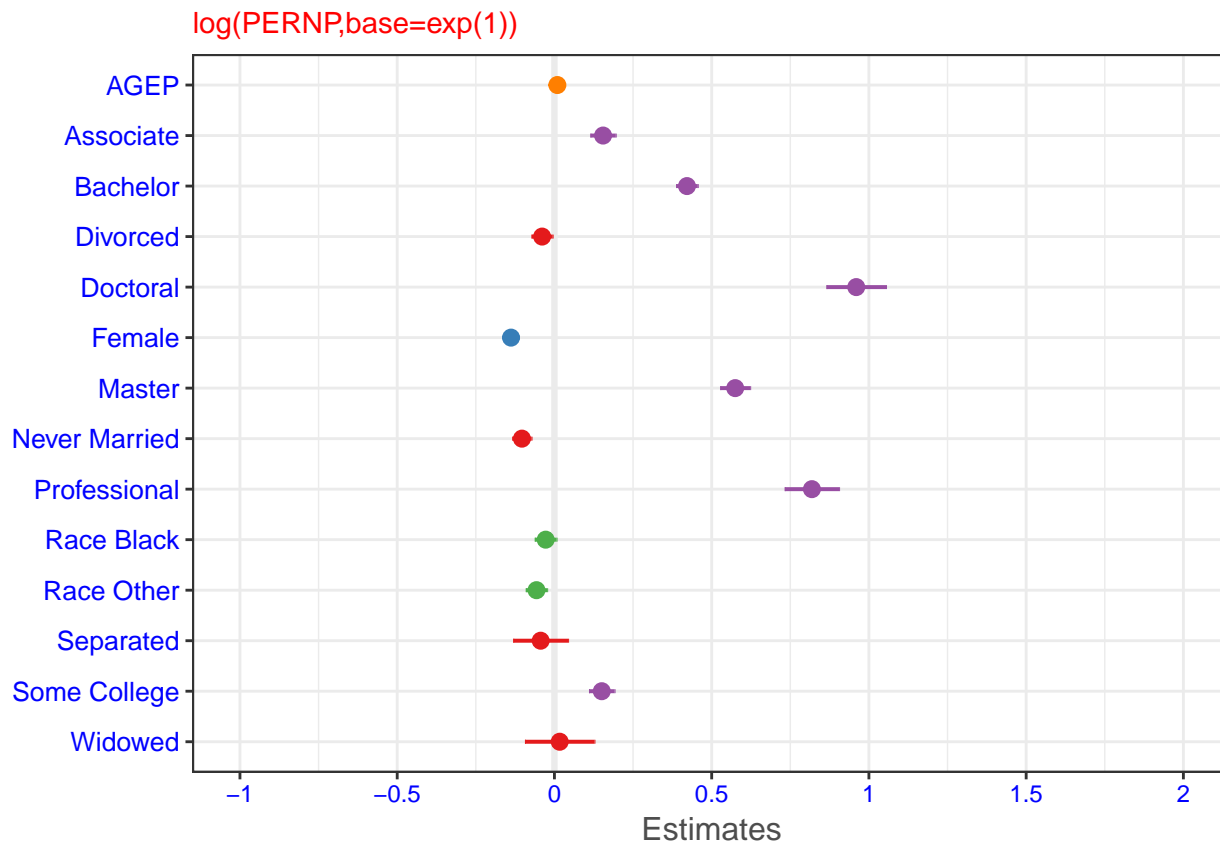
```
##      Min      1Q   Median      3Q      Max
## -1.56195 -0.30715 -0.03111  0.27690  1.69792
##
## Coefficients:
##              Estimate Std. Error t value      Pr(>|t|)
## (Intercept)   9.5841704  0.0331471 289.140 < 0.0000000000000002 ***
## Widowed       0.0167914  0.0563665   0.298   0.765794
## Divorced      -0.0391735  0.0170763  -2.294   0.021829 *
## Separated     -0.0437489  0.0444603  -0.984   0.325162
## NeverMarried -0.1033310  0.0155387  -6.650  0.00000000000323503 ***
## Female       -0.1380595  0.0129169 -10.688 < 0.0000000000000002 ***
## RaceBlack    -0.0278297  0.0176550  -1.576   0.115017
## RaceOther    -0.0570606  0.0173151  -3.295   0.000989 ***
## SomeCollege   0.1505241  0.0207411   7.257  0.0000000000004534 ***
## Associate     0.1546604  0.0205537   7.525  0.0000000000000619 ***
## Bachelor     0.4214137  0.0175270  24.044 < 0.0000000000000002 ***
## Master        0.5748312  0.0242603  23.694 < 0.0000000000000002 ***
## Professional  0.8187333  0.0440766  18.575 < 0.0000000000000002 ***
## Doctoral     0.9597258  0.0482778  19.879 < 0.0000000000000002 ***
## AGE          0.0092656  0.0006307  14.692 < 0.0000000000000002 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4454 on 5179 degrees of freedom
## Multiple R-squared:  0.2556, Adjusted R-squared:  0.2536
## F-statistic: 127 on 14 and 5179 DF,  p-value: < 0.00000000000000022
```

## Interpretation

- Coefficients Explanation
- Holding gender, race, education and age constant, married people makes 1.69% less than people who widowed on average.
- Holding gender, race, education and age constant, married people makes 3.84% more than people who divorced on average.
- Holding gender, race, education and age constant, married people makes 4.28% more than people who separated on average.
- Holding gender, race, education and age constant, married people makes 9.82% more than people who never married on average.
- Holding marriage, race, education and age constant, male makes 12.9% more than female on average.
- Holding marriage, gender, education and age constant, White makes 2.74% more than Black on average.
- Holding marriage, gender, education and age constant, White makes 5.55% more than Other race on average.
- Holding marriage, gender, race and age constant, people have high school education makes 16.24% less than people have some college education on average.
- Holding marriage, gender, race and age constant, people have high school education makes 16.73% less than people have associate education on average.
- Holding marriage, gender, race and age constant, people have high school education makes 52.41% less than people have bachelor's degree on average.

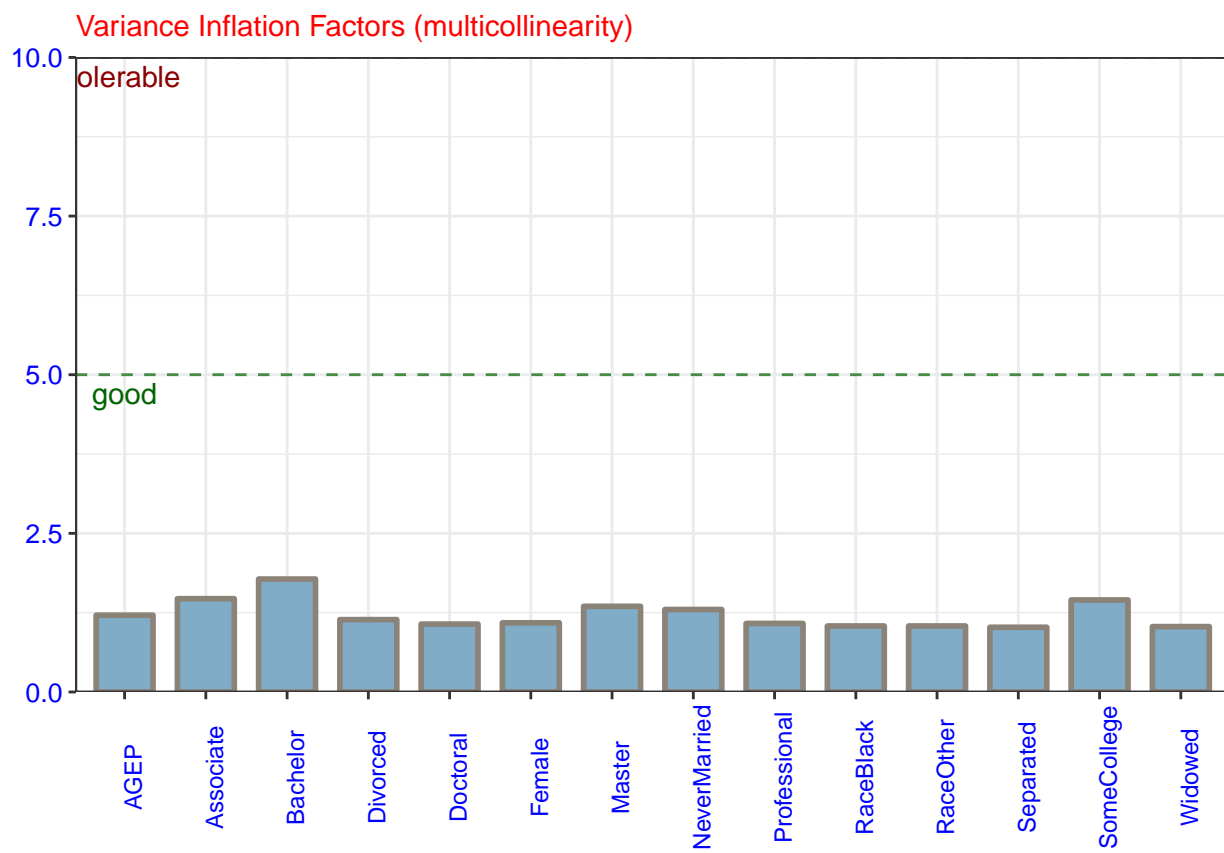
- Holding marriage, gender, race and age constant, people have high school education makes 77.68% less than people have master's degree on average.
- Holding marriage, gender, race and age constant, people have high school education makes 126.76% less than people have Professional education on average.
- Holding marriage, gender, race and age constant, people have high school education makes 161.1% less than people have doctor's degree on average.
- Holding marriage, gender, race and education constant, people make 0.93% more as age increases on average between the age of 18 to 64.

## Model Diagnostics



Variables	Tolerance	VIF
Widowed	0.9729050	1.027850
Divorced	0.8808670	1.135245
Separated	0.9756434	1.024965
NeverMarried	0.7702636	1.298257
Female	0.9159123	1.091808
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```
## [[1]]
```

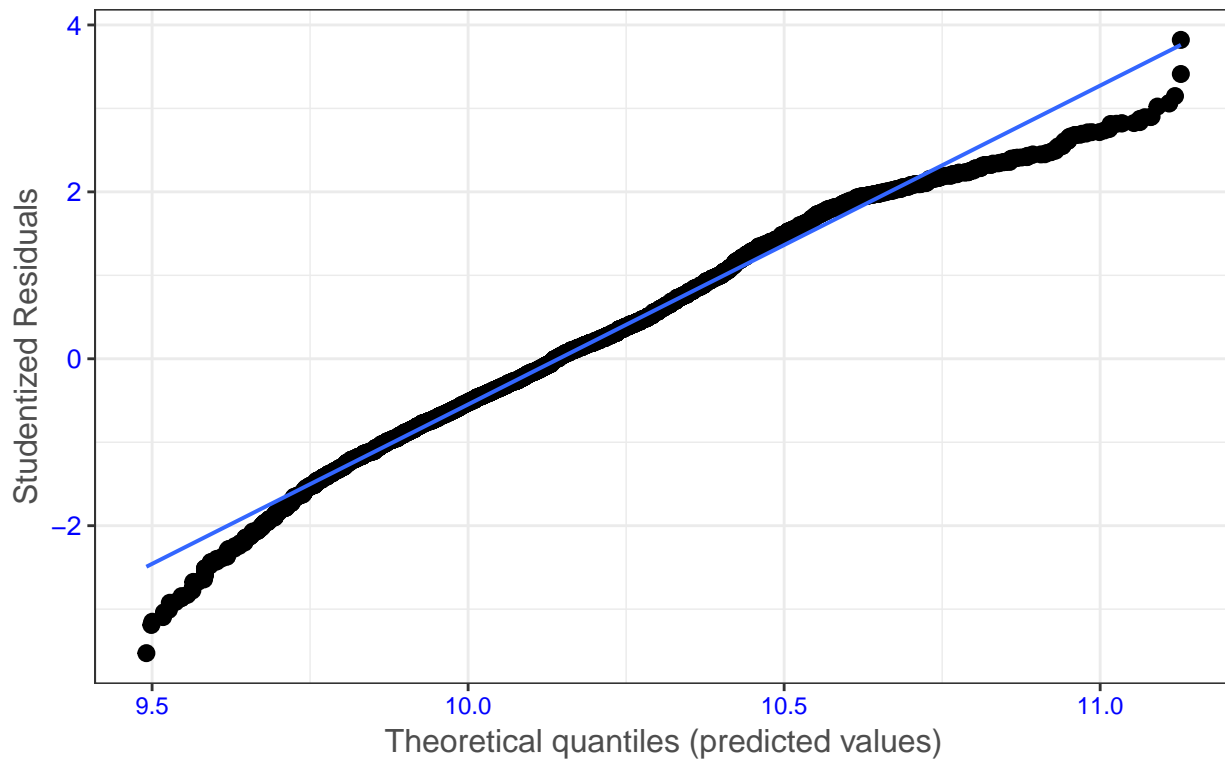


A VIF chart shows that there is no correlation among the kth predictor and the remaining predictor variables.

```
## [[1]]
```

### Non-normality of residuals and outliers

Dots should be plotted along the line

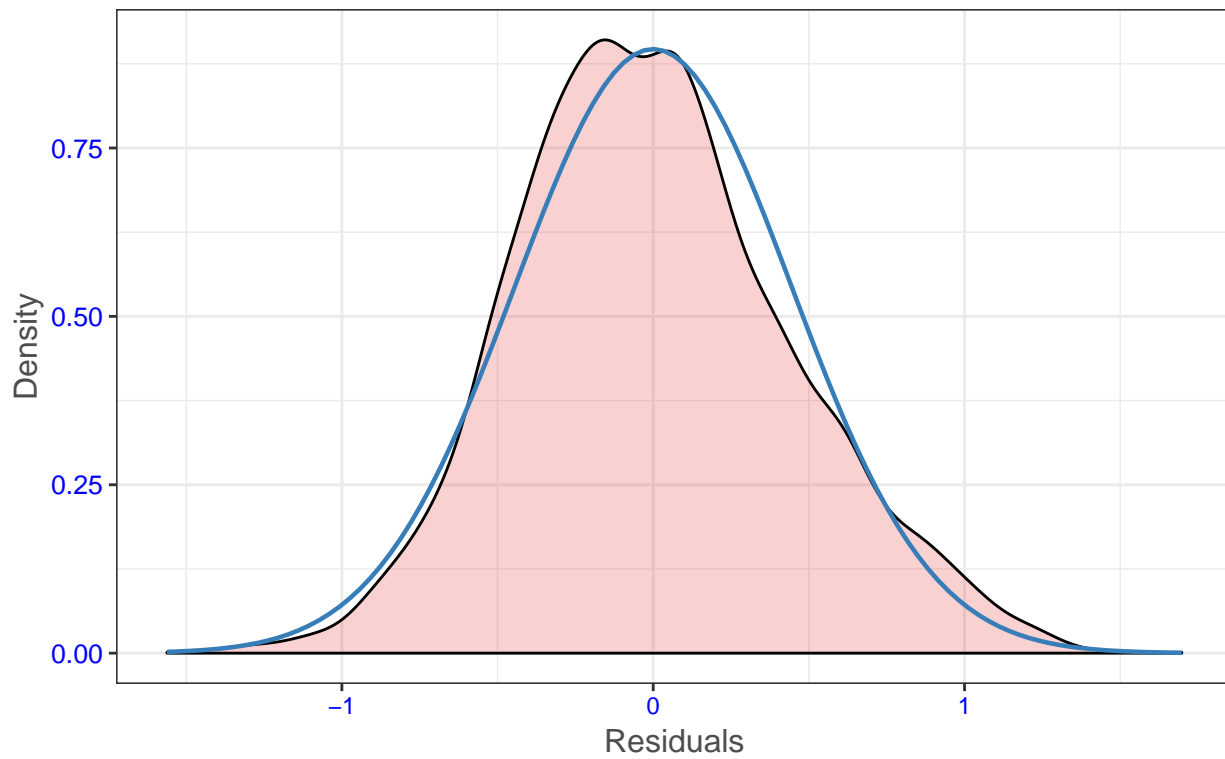


The plot shows that the distribution of the residuals are approximately normally distributed. The “thin tails” correspond to the first quantiles occurring at less than expected values and the last quantiles occurring at larger than expected values.

```
## [[1]]
```

### Non-normality of residuals

Distribution should look like normal curve

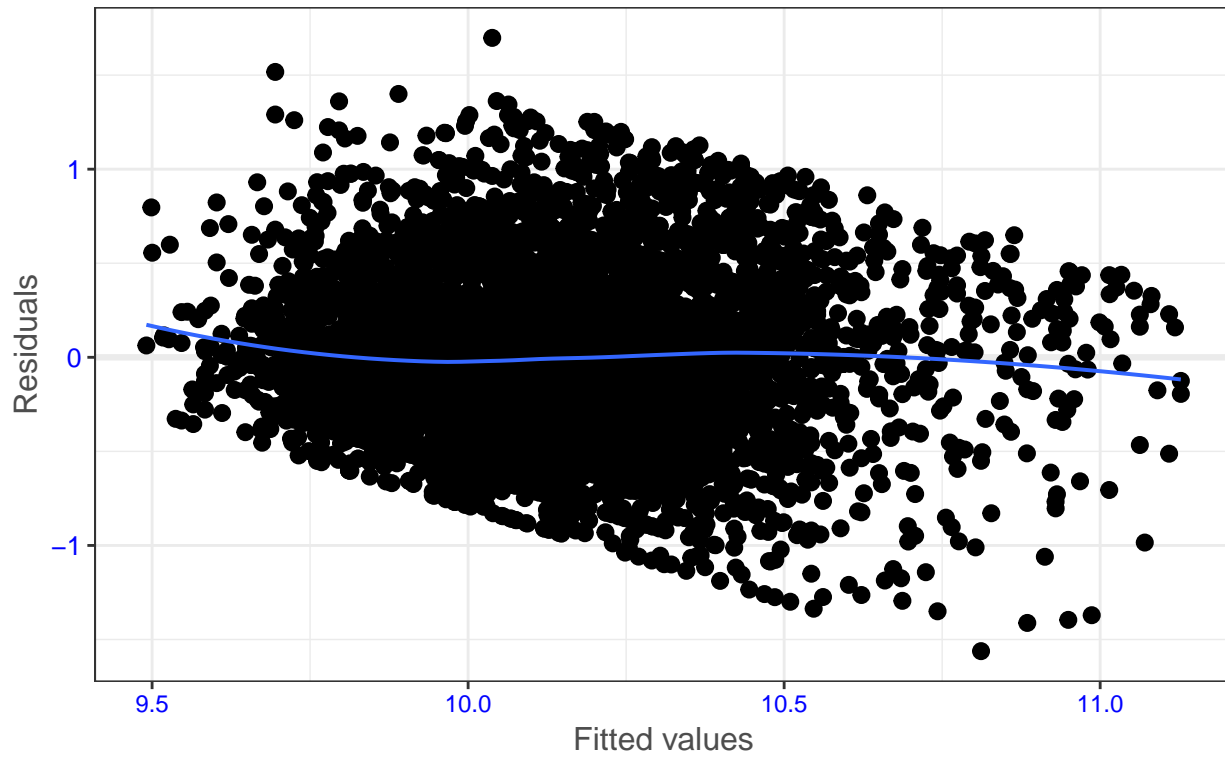


The plot shows that the distribution of the residuals are approximately normally distributed.

```
## [[1]]
```

### Homoscedasticity (constant variance of residuals)

Amount and distance of points scattered above/below line is equal or randomly spread



The plot shows heteroscedasticity as the variance of the residuals are not constant.