

## Econ Assignment - 7

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Q.8.4

a)

Graph of residuals of income suggests that ~~as~~ as income increases, the error variance increases  $\rightarrow$  heteroskedasticity.

Graph of age residuals doesn't seem to have any pattern. error here is not varying with age.

c)

i) vacation miles traveled increases as income & age increases  $\rightarrow$  +ve correlation  
vacation miles traveled decreases as no. of kids increases  $\rightarrow$  -ve correlation

ii) Both std errors are approximately same. White's std errors are a little bit higher for variables than that of least square. we can say that white's std errors will give us wider confidence interval compared to least square. All are significant at 5% in both models.



iii) Generalized least squares to have smallest std errors of all. therefore they will have a narrow confidence interval of all. they will give most precise estimates of all.  $\therefore$  They are ~~the~~ better than others in this case.

8.12  
a) Rich countries having more money have multiple options to spend their money. they can spend only on basic education & also can spend more than basic education.  $\rightarrow$  more variability  
Countries with less GDP have very less money so they can not afford spending a lot of money on education. they can spend only on basic.  $\rightarrow$  less variability

b)  $y = \frac{EE}{P}$        $x = \frac{GDP}{P}$

$$\hat{y} = -1.24 + 0.073 x$$

plots shows that there is heteroskedasticity

c) After plotting the modely

$$\chi^2 = N \times R^2 = 34 \times 0.293$$

$$= 9.962$$

$$\chi^2_{(0.95, 2)} = 5.991$$

$$\chi^2 > \chi^2_{\text{critical}}$$

~~H<sub>0</sub>~~ H<sub>0</sub>:  $\alpha_2 = \alpha_3 = 0$

✓ H<sub>1</sub>: at least one of  $\alpha_1, \alpha_2 \neq 0$

∴ We reject the null hypothesis  
 ∴ Heteroskedasticity exists

d) least sq. std error for  $b_2 = 0.0051$   
 Robust std error for  $b_2 = 0.0063$

$$0.0731 \pm 2.0369 \times 0.0051$$

$$= (0.0627, 0.0834)$$

Robust +

$$0.0731 \pm 2.0369 \times 0.0063$$

$$(0.0602, 0.0859)$$

Robust std error gives wider interval

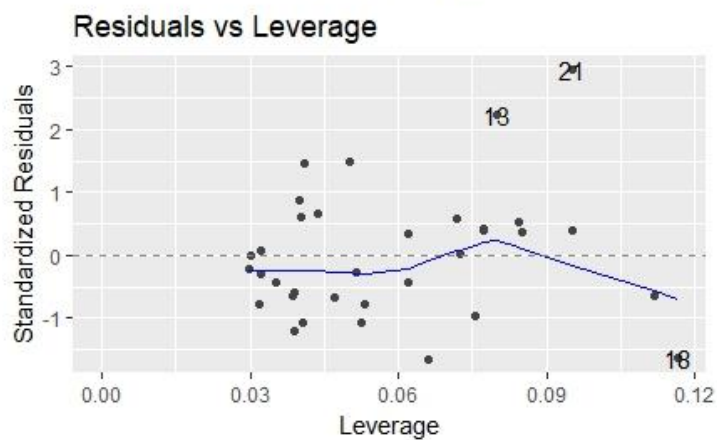
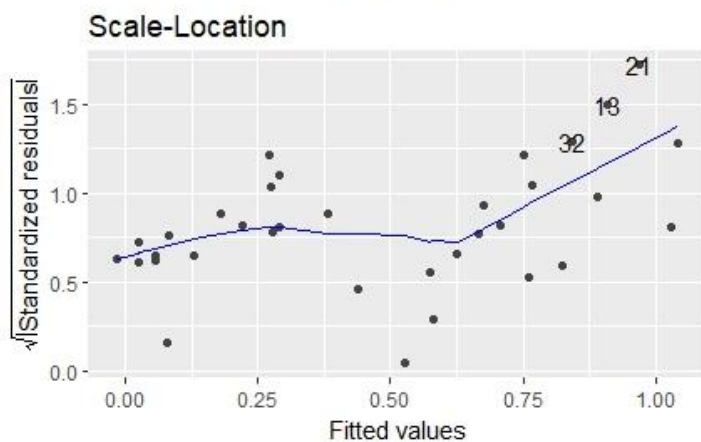
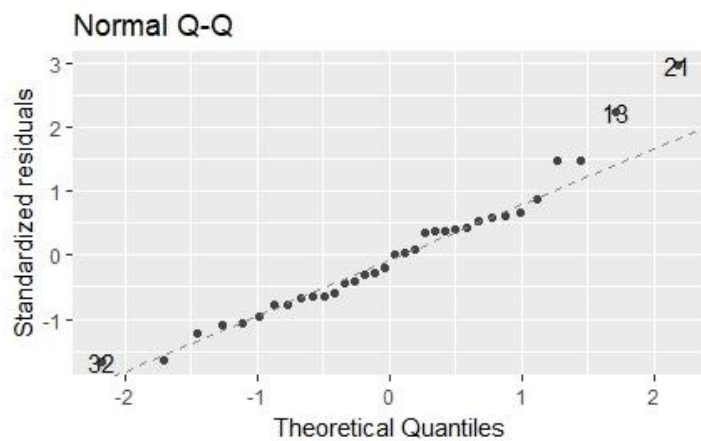
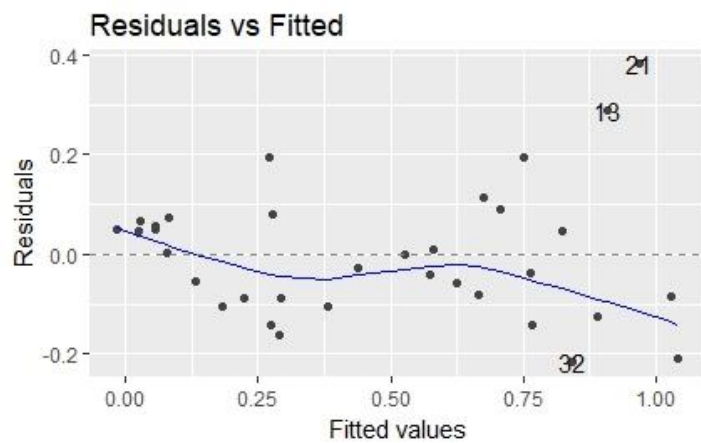


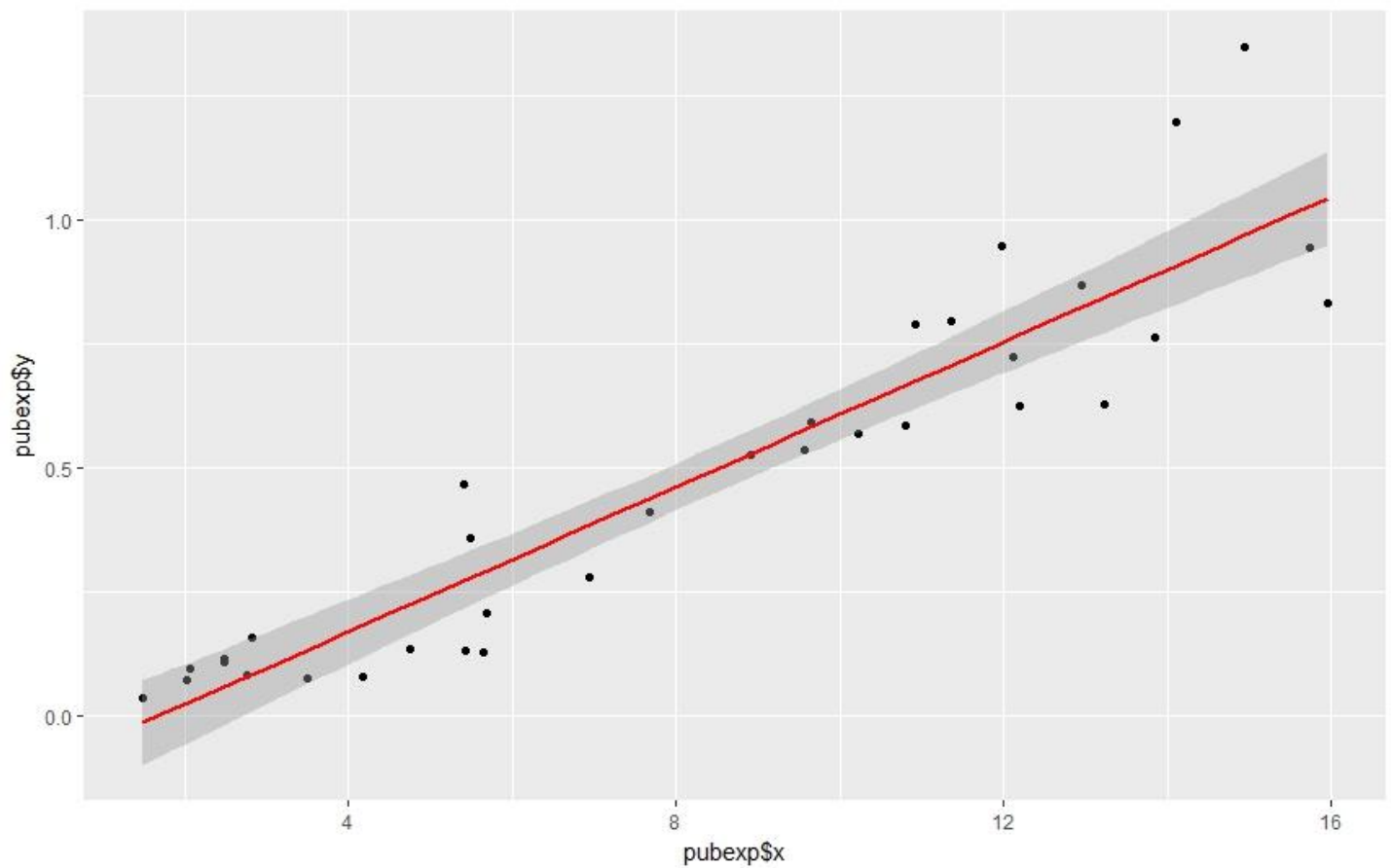
c)  $\hat{y} = \cancel{0.0929} + 0.069$

$$\hat{y} = 0.0929 + 0.0693x$$

$$\begin{aligned} 95\% \text{ confint} &= 0.069 + 2.036 \times 0.03412 \\ &= (0.0603, \cancel{0.710.07830}) \end{aligned}$$

This confidence interval is narrower than both of the intervals obtained in part d.





## R code

```
> library(haven)
> pubexp <- read_dta("D:/Class Notes/Fall 17 Classes/ECON/Data_sets/pubexp.dta")
> View(pubexp)
> # creating y and x variables
> y <- pubexp$see/pubexp$p
> x <- pubexp$gdp/pubexp$p
> x2 = x^2
> data1 <- cbind(pubexp, y)
> data <- cbind(data1, x)
>
> pubexp <- data
```

```
> # Running least square regression
>
> leastsq <- lm (y ~ x, data = pubexp)
> summary(leastsq)
```

Call:

```
lm(formula = y ~ x, data = pubexp)
```

Residuals:

Min	1Q	Median	3Q	Max
-0.21682	-0.08804	-0.01401	0.06517	0.38156

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	-0.124573	0.048523	-2.567	0.0151 *
x	0.073173	0.005179	14.128	2.65e-15 ***

---

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

Residual standard error: 0.1359 on 32 degrees of freedom

Multiple R-squared: 0.8618, Adjusted R-squared: 0.8575

F-statistic: 199.6 on 1 and 32 DF, p-value: 2.65e-15

```
> # getting residuals
> leastsq.res <- resid(leastsq)
> # Plotting residuals
>
> library(ggfortify)
> library(ggplot2)
> autoplot(leastsq, data=pubexp, color = 'blue')
>
> ggplot(pubexp, aes(x = pubexp$x, y = pubexp$y)) + geom_point() +
```

```

+ stat_smooth( method = "lm", col = 'red')
> # White formula
>
> res = residuals(leastsq)
> ressq = res^2
> #yhat = leastsq$fitted.values
> #yhat2 = yhat^2
>
> #combining all into one dataset
> #m4 = data.frame(cbind(ressq, yhat, yhat2))
>
> m3 = data.frame(cbind(ressq, x, x2))
>
> white <- lm(ressq ~ x + x2, data = m3)
> summary(white)

```

Call:

```
lm(formula = ressq ~ x + x2, data = m3)
```

Residuals:

Min	1Q	Median	3Q	Max
-0.048773	-0.011263	-0.005239	0.005350	0.097664

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	0.0176770	0.0161120	1.097	0.2810
x	-0.0052062	0.0045479	-1.145	0.2611
x2	0.0004840	0.0002638	1.835	0.0762 .

---

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

Residual standard error: 0.02513 on 31 degrees of freedom



Multiple R-squared: 0.293, Adjusted R-squared: 0.2474

F-statistic: 6.423 on 2 and 31 DF, p-value: 0.004636

```
>
> # critical chi-square
> qchisq(.95, 2)
[1] 5.991465
> #Robust std errors
> library(sandwich)
> #coef
>
> #coeftest(leastsq, vcov = vcovHC(leastsq,type = "HC1"))
>
> SE_robust <- sqrt(diag(vcovHC(leastsq, type="HC2")))
>
> model2 <- summary(leastsq)
> SE_robust
(Intercept)      x
0.040783211 0.006307339
>
> model2$coefficients[,2] <- SE_robust
> model2
```

Call:

```
lm(formula = y ~ x, data = pubexp)
```

Residuals:

Min	1Q	Median	3Q	Max
-0.21682	-0.08804	-0.01401	0.06517	0.38156

Coefficients:

Estimate	Std. Error	t value	Pr(> t )
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(Intercept) -0.124573 0.040783 -2.567 0.0151 \*

x 0.073173 0.006307 14.128 2.65e-15 \*\*\*

---

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

Residual standard error: 0.1359 on 32 degrees of freedom

Multiple R-squared: 0.8618, Adjusted R-squared: 0.8575

F-statistic: 199.6 on 1 and 32 DF, p-value: 2.65e-15

```
> qt(.975,32)
```

```
[1] 2.036933
```

```
> #Weightd least squares
```

```
>
```

```
> model <- lm(y ~ x, data = pubexp, weights = (1/pubexp$x))
```

```
> summary(model)
```

Call:

```
lm(formula = y ~ x, data = pubexp, weights = (1/pubexp$x))
```

Weighted Residuals:

Min	1Q	Median	3Q	Max
-0.072028	-0.038561	-0.008488	0.027706	0.105415

Coefficients:

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

(Intercept)	-0.092921	0.028904	-3.215	0.00298 **
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x	0.069321	0.004412	15.713	< 2e-16 ***
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Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.04451 on 32 degrees of freedom

Multiple R-squared: 0.8853, Adjusted R-squared: 0.8817

F-statistic: 246.9 on 1 and 32 DF, p-value: < 2.2e-16

>

> confint(model, 'x', level = 0.95)

2.5 % 97.5 %

x 0.06033514 0.07830786