

Paper & pencil question.

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1. Which of the following can cause the usual OLS t statistics to be invalid (that is, not to have t distributions under H_0)?

- i. Heteroskedasticity.
- ii. A sample correlation coefficient of .95 between two independent variables that are in the model.
- iii. Omitting an important explanatory variable.

1. (i) and (ii)

(i) Heteroskedasticity is violated MLR5 - homoskedasticity, and it make OLS t statistic to be invalid.

(ii) An explanatory variable which could affect the data can thereby affect the results of regression analysis will definitely cause t statistic to be invalid if that explanatory variable is omitted. This violate MLR4 $E(u|x) = 0$.

2. Consider an equation to explain salaries of CEOs in terms of annual firm sales, return on equity (roe, in percentage form), and return on the firm's stock (ros, in percentage form):

$$\log(\text{salary}) = \beta_0 + \beta_1 \log(\text{sales}) + \beta_2 \text{roe} + \beta_3 \text{ros} + u$$

i. In terms of the model parameters, state the null hypothesis that, after controlling for sales and roe, ros has no effect on CEO salary. State the alternative that better stock market performance increases a CEO's salary.

$$H_0: \beta_3 = 0$$

$$H_a: \beta_3 > 0$$

ii. Using the data in CEOSAL1, the following equation was obtained by OLS:

$$\widehat{\log(\text{salary})} = 4.32 + .280 \log(\text{sales}) + .0174 \text{roe} + .00024 \text{ros}$$

(.32) (.035)
(.0041)
(.00054)

$n = 209, R^2 = .283.$

By what percentage is salary predicted to increase if ros increases by 50 points? Does ros have a practically large effect on salary?

$$\beta_3 = \frac{d \log(\text{salary})}{d X_3}$$

$$\beta_3 = \frac{\frac{1}{\text{salary}} \Delta \text{salary}}{\Delta X_3}$$

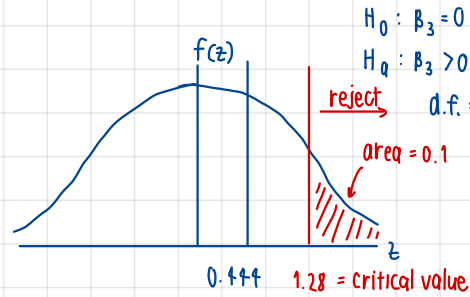
$$100 \times .00024 = \frac{1}{\text{salary}} \Delta \text{salary} \times 100$$

$$100 \times .00024(50) = \frac{1}{\text{salary}} \Delta \text{salary} \times 100$$

$$1.2 = \% \Delta \text{salary}$$

holding other factors constant, if ros increases by 50 points $\log(\text{salary})$ will increase by 1.2%.

iii. Test the null hypothesis that ros has no effect on salary against the alternative that ros has a positive effect. Carry out the test at the 10% significance level.



$$t = \frac{\hat{\beta}_3 - \beta_3}{\text{s.e.} \hat{\beta}_3} = \frac{.00024}{.00054} = 0.444 < 1.28$$

So, we fail to reject H_0 at 10% significant level.

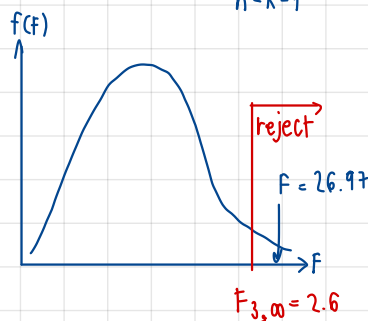
iv. Would you include ros in a final model explaining CEO compensation in terms of firm performance? Explain.

$$r: \widehat{\log(\text{salary})} = 4.32 \rightarrow H_0: \beta_1 = \beta_2 = \beta_3 = 0$$

$$ur: \widehat{\log(\text{salary})} = 4.32 + 0.28 \log(\text{sales}) + 0.0174(\text{roe}) + 0.00024(\text{ros})$$

H_a : otherwise

$$F = \frac{\frac{R^2}{k}}{\frac{(1-R^2)}{n-k-1}} = \frac{\frac{0.283}{3}}{\frac{0.717}{205}} = 26.97$$



Since, $F = 26.97 > 2.6$, we rejected H_0 and conclude that firm performance have joint effect in explaining CEO compensation at 5% significant level.

Computer questions.

C6. Use the data in WAGE2 for this exercise.

i. Consider the standard wage equation

$$\log(\text{wage}) = \beta_0 + \beta_1 \text{educ} + \beta_2 \text{exper} + \beta_3 \text{tenure} + u.$$

State the null hypothesis that another year of general workforce experience has the same effect on $\log(\text{wage})$ as another year of tenure with the current employer.

$$H_0 : \beta_2 = \beta_3 \quad \text{or} \quad H_0 : \beta_2 - \beta_3 = 0$$

$$H_a : \beta_2 \neq \beta_3 \quad \text{or} \quad H_a : \beta_2 - \beta_3 \neq 0$$

ii. Test the null hypothesis in part (i) against a two-sided alternative, at the 5% significance level, by constructing a 95% confidence interval. What do you conclude?

$$\text{let } \hat{\theta}_1 = \hat{\beta}_2 - \hat{\beta}_3$$

$$\text{So, } H_0 : \hat{\theta}_1 = 0 \quad t = \frac{\hat{\theta}_1}{\text{s.e.} \hat{\theta}_1}$$

$$H_a : \hat{\theta}_1 \neq 0$$

rearrange $\hat{\theta}_1 = \hat{\beta}_2 - \hat{\beta}_3$, get $\hat{\beta}_2 = \hat{\theta}_1 + \hat{\beta}_3$ and substitute it into regression model

$$\log(\text{wage}) = \beta_0 + \beta_1 \text{educ} + (\hat{\theta}_1 + \hat{\beta}_3) \text{exper} + \beta_3 \text{tenure} + u$$

$$= \beta_0 + \beta_1 \text{educ} + \hat{\theta}_1 \text{exper} + \hat{\beta}_3 \text{exper} + \beta_3 \text{tenure} + u$$

$$\log(\text{wage}) = \beta_0 + \beta_1 \text{educ} + \hat{\theta}_1 \text{exper} + \hat{\beta}_3 (\text{exper} + \text{tenure}) + u$$

. generate expertenure = exper+tenure

. end of do-file

. do "C:\Users\610464~1\AppData\Local\Temp\STD1224_000000.tmp"

. regress lwage educ exper expertenure

Source	SS	df	MS	Number of obs	=	935
Model	25.6953242	3	8.56510806	F(3, 931)	=	56.97
Residual	139.960959	931	.150334005	Prob > F	=	0.0000
				R-squared	=	0.1551
				Adj R-squared	=	0.1524
Total	165.656283	934	.177362188	Root MSE	=	.38773

lwage	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
educ	.0748638	.0065124	11.50	0.000	.062083	.0876446
exper	.0019537	.0047434	0.41	0.681	-.0073554	.0112627
expertenure	.0133748	.0025872	5.17	0.000	.0082974	.0184522
_cons	5.496696	.1105282	49.73	0.000	5.279782	5.713609

$$t_{\text{critical}} = 0.41 < 1.96$$

Which means we fail to reject H_0 at 95% confidence interval. So, both of variables has not have the same impact.

C8. The data set 401KSUBS contains information on net financial wealth (*nettfa*), age of the survey respondent (*age*), annual family income (*inc*), family size (*fsize*), and participation in certain pension plans for people in the United States. The wealth and income variables are both recorded in thousands of dollars. For this question, use only the data for single-person households (so *fsize* = 1).

i. How many single-person households are there in the data set?

```
. tab fsize
family size      Freq.    Percent    Cum.
-----
1                2,017    21.75     21.75
2                2,199    23.71     45.46
3                1,829    19.72     65.18
4                1,990    21.46     86.63
5                 816     8.80     95.43
6                 268     2.89     98.32
7                 95      1.02     99.34
8                 38      0.41     99.75
9                  7      0.08     99.83
10                7      0.08     99.90
11                3      0.03     99.94
12                4      0.04     99.98
13                2      0.02    100.00
-----
Total            9,275    100.00
```

There are 2,017 single-person household in the data.

ii. Use OLS to estimate the model

$$nettfa = \beta_0 + \beta_1 inc + \beta_2 age + u_i$$

and report the results using the usual format. Be sure to use only the single-person households in the sample. Interpret the slope coefficients. Are there any surprises in the slope estimates?

```
. keep if fsize == 1
(7,258 observations deleted)

. regress nettfa inc age

Source      SS          df    MS          Number of obs =    2,017
             544916.989      2    272458.495      F(2, 2014)      =   136.46
Model       4021048.06      2,014    1996.54819      Prob > F        =   0.0000
Residual    4565965.05      2,016    2264.86361      R-squared       =   0.1193
Total      4565965.05      2,016    2264.86361      Adj R-squared   =   0.1185
             4565965.05      2,016    2264.86361      Root MSE      =   44.683

nettfa      Coef.    Std. Err.    t    P>|t|    [95% Conf. Interval]
-----
inc         .7993167   .0597307    13.38  0.000    .6821762   .9164572
age         .8426563   .0920169     9.16  0.000    .6621982   1.023115
_cons     -43.03981   4.080393   -10.55  0.000   -51.04204  -35.03758
```

Intercept = -43.03981 ;
slope on inc = .7993167; slope on age = .8426563
If income increases by 1 unit or \$1,000 then nettfa increased by .7993167 or \$799.3167.
If age increases by 1 unit or 1 year the nettfa increased by .8426563 or \$843

iii. Does the intercept from the regression in part (ii) have an interesting meaning? Explain.

```
. summarize nettfa
Variable | Obs      Mean      Std. Dev.    Min      Max
-----|-----
nettfa   | 2,017    13.59498    47.59058    -143.5    1134.098

. do "C:\Users\610464~1\AppData\Local\Temp\STD17e4_000000.tmp"

. count if nettfa < 0
524

. do "C:\Users\610464~1\AppData\Local\Temp\STD17e4_000000.tmp"

. summarize inc age
Variable | Obs      Mean      Std. Dev.    Min      Max
-----|-----
inc      | 2,017    29.44618    16.67356    10.008    143.067
age      | 2,017    39.27814    10.82328     25        64
```

Intercept gives us the predicted nettfa when inc=0 and age=0 but the lowest inc is \$10,000 and the lowest age is 25.

iv. Find the p-value for the test $H_0: \beta_2 = 1$ against $H_1: \beta_2 < 1$. Do you reject H_0 at the 1% significance level?

```
. lincom age - 1
( 1) age = 1
```

nettfa	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
(1)	-.1573437	.0920169	-1.71	0.087	-.3378018 .0231145

One-sided p-value = .087/2 = .0435 which is greater than .01. So, we accepted H_0 at 1% significant level.

v. If you do a simple regression of *nettfa* on *inc*, is the estimated coefficient on *inc* much different from the estimate in part (ii)? Why or why not?

```
. regress nettfa inc

Source      SS          df    MS          Number of obs =    2,017
             377482.064      1    377482.064      F(1, 2015)      =   181.60
Model       4188482.98      2,015    2078.6516     Prob > F        =   0.0000
Residual    4565965.05      2,016    2264.86361      R-squared       =   0.0827
Total      4565965.05      2,016    2264.86361      Adj R-squared   =   0.0822
             4565965.05      2,016    2264.86361      Root MSE      =   45.592

nettfa      Coef.    Std. Err.    t    P>|t|    [95% Conf. Interval]
-----
inc         .8206815   .0609     13.48  0.000    .7012479   .940115
_cons     -10.57095   2.060678   -5.13  0.000   -14.61223  -6.529671

. do "C:\Users\610464~1\AppData\Local\Temp\STD17e4_000000.tmp"

. correlate inc age
(obs=2,017)

            inc      age
-----
inc         1.0000
age         0.0391  1.0000
```

From (ii), estimated coefficient on inc is .799 versus the new one which is .821. They are not very different because income and age are not quite correlated.

C1. The following model can be used to study whether campaign expenditures affect election outcomes:

$$voteA = \beta_0 + \beta_1 \log(expendA) + \beta_2 \log(expendB) + \beta_3 prtysrA + u,$$

where $voteA$ is the percentage of the vote received by Candidate A, $expendA$ and $expendB$ are campaign expenditures by Candidates A and B, and $prtysrA$ is a measure of party strength for Candidate A (the percentage of the most recent presidential vote that went to A's party).

i. What is the interpretation of β_1 ?

1% increases in campaign expenditure by candidate A, increase 1% of the vote received by candidate A.

ii. In terms of the parameters, state the null hypothesis that a 1% increase in A's expenditures is offset by a 1% increase in B's expenditures.

$$H_0: \beta_1 = -\beta_2 \quad \text{or} \quad H_0: \beta_1 + \beta_2 = 0$$

$$H_a: \beta_1 \neq -\beta_2 \quad \text{or} \quad H_a: \beta_1 + \beta_2 \neq 0$$

iii. Estimate the given model using the data in VOTE1 and report the results in usual form. Do A's expenditures affect the outcome? What about B's expenditures? Can you use these results to test the hypothesis in part (ii)?

```

. regress voteA lexpendA lexpendB prtysrA

```

Source	SS	df	MS	Number of obs	=	173
Model	38405.1096	3	12801.7032	F(3, 169)	=	215.23
Residual	10052.1389	169	59.480112	Prob > F	=	0.0000
				R-squared	=	0.7926
				Adj R-squared	=	0.7889
				Root MSE	=	7.7123

voteA	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
lexpendA	6.083316	.38215	15.92	0.000	5.328914 6.837719
lexpendB	-6.615417	.3788203	-17.46	0.000	-7.363246 -5.867588
prtysrA	.1519574	.0620181	2.45	0.015	.0295274 .2743873
_cons	45.07893	3.926305	11.48	0.000	37.32801 52.82985

A's expenditure positively affect the outcome.
 B's expenditure negatively affect the outcome.
 NO, the data can not be use to test the hypothesis in part (ii)
 Because $\beta_1 + \beta_2$ is not provided.

iv. Estimate a model that directly gives the t statistic for testing the hypothesis in part (ii). What do you conclude? (Use a two-sided alternative.)

Let's $\hat{\theta}_1 = \hat{\beta}_1 + \hat{\beta}_2$ $H_0: \theta_1 = 0$ $t = \frac{\hat{\theta}_1}{s.e.\hat{\theta}_1}$
 $H_a: \hat{\theta}_1 \neq 0$

Now, $\hat{\beta}_1 = \hat{\theta}_1 - \hat{\beta}_2$ $\beta_1 = \theta_1 - \beta_2$ substitute into regression.

$$VoteA = \beta_0 + (\theta_1 - \beta_2) \log(expendA) + \beta_2 \log(expendB) + \beta_3 prtysrA + u.$$

$$= \beta_0 + \theta_1 \log(expendA) - \beta_2 \log(expendA) + \beta_2 \log(expendB) + \beta_3 prtysrA + u.$$

$$= \beta_0 + \theta_1 \log(expendA) - \beta_2 (\log(expendA) + \log(expendB)) + \beta_3 prtysrA + u.$$

$$VoteA = \beta_0 + \theta_1 \log(expendA) + \beta_2 (\log(expendB) - \log(expendA)) + \beta_3 prtysrA + u.$$

```

. generate lexpendAlexpendB = lexpendB - lexpendA
. regress voteA lexpendA lexpendAlexpendB prtysrA

```

Source	SS	df	MS	Number of obs	=	173
Model	38405.1097	3	12801.7032	F(3, 169)	=	215.23
Residual	10052.1388	169	59.4801115	Prob > F	=	0.0000
				R-squared	=	0.7926
				Adj R-squared	=	0.7889
				Root MSE	=	7.7123

voteA	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
lexpendA	-.532101	.5330858	-1.00	0.320	-1.584466 .5202638
lexpendAlexpe=B	-6.615417	.3788203	-17.46	0.000	-7.363246 -5.867588
prtysrA	.1519574	.0620181	2.45	0.015	.0295274 .2743873
_cons	45.07893	3.926305	11.48	0.000	37.32801 52.82985

$t_{critical} = -1.00 > -1.96$
 Therefore, we fail to reject $H_0: \beta_1 = -\beta_2$ at 5% significant level. 1% increased in A's expenditure is offset by 1% increased in B's expenditure.