

ECN 102: Analysis of Economics Data

Homework 3 Answer Key

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Question 1: Bivariate model assumptions

- a) What is the first population assumption of bivariate OLS regression? How might this assumption be violated? How could we account for this violation?

Linearity: the true relationship between x and y is linear. This could be violated if we instead have a true nonlinear relationship, which we could account for by adapting our regression to match.

- b) What is the second population assumption of bivariate OLS regression? How might this assumption be violated? How could we account for this violation?

Unbiasedness: the conditional expected value of our error term is zero for all values of x . This could be violated if there are variables missing from our regression that are correlated with x and y . We could address this by including these missing/omitted regressors in a multivariate regression.

- c) What is the third population assumption of bivariate OLS regression? How might this assumption be violated? How could we account for this violation?

Homoskedasticity: the distribution (variance) of our errors is constant across values of x . This could be violated if the distribution of our errors changes for larger or smaller values of x . We could account for this by using robust standard errors in our regression.

- d) What is the fourth population assumption of bivariate OLS regression? How might this assumption be violated? How could we account for this violation?

Independence: our error terms are uncorrelated for different units of observation. This could be violated if our observations are not independent from each other, such as time-series data of GDP for a given country. We could account for this by using more complicated forms of regression that account for error dependence.

Question 2: Manual calculations

Consider a dataset with 3 observations for $(x, y) : (2, 4), (1, 3)$ and $(6, 2)$.

From first principles (using formulas **not Stata**), compute:

a) the OLS intercept

x_i	y_i	$\bar{x}$	$\bar{y}$	$(x_i - \bar{x})$	$(x_i - \bar{x})^2$	$(y_i - \bar{y})$	$(y_i - \bar{y})^2$	$(x_i - \bar{x})(y_i - \bar{y})$
2	4	3	3	-1	1	1	1	-1
1	3	3	3	-2	4	0	0	0
6	2	3	3	3	9	-1	1	-3

$$b_1 = \bar{y} - b_2 \bar{x} = 3 - \left(-\frac{4}{14}\right) \times 3 = 3.86$$

b) the OLS slope coefficient

$$b_2 = \frac{\sum_i (x_i - \bar{x})(y_i - \bar{y})}{\sum_i (x_i - \bar{x})^2} = \frac{-1 + 0 - 3}{1 + 4 + 9} = -\frac{4}{14} = -0.29$$

c) residuals for the three observations

$$e_i = y_i - \hat{y}_i :$$

$$4 - [3.86 - 0.29 \times 2] = 4 - 3.28 = 0.72$$

$$3 - [3.86 - 0.29 \times 1] = 3 - 3.57 = -0.57$$

$$2 - [3.86 - 0.29 \times 6] = 2 - 2.12 = -0.12$$

$$\sum_i e_i^2 = 0.86$$

d) the R^2 of **reg y x** (or **reg x y**)

$$R^2 = 1 - \frac{\sum_i e_i^2}{\sum_i (y_i - \bar{y})^2} = 1 - \frac{0.86}{2} = 0.57$$

e) the covariance of X and Y

$$Cov(x, y) = \frac{\sum_i (x_i - \bar{x})(y_i - \bar{y})}{n - 1} = \frac{-1 + 0 - 3}{2} = -2$$

f) the correlation between X and Y

$$r_{xy} = \frac{Cov(x,y)}{s_x s_y} = \frac{-4}{\sqrt{14}\sqrt{2}} = -0.76$$

Question 3: Bivariate regression

Use AED_SALARYSAT.DTA.

a) What variables are present in our dataset? List them along with their labels.

```
u AED_SALARYSAT, clear
des
```

(Data for A. Colin Cameron (2022), ANALYSIS OF ECONOMIC DATA, Amazon)

Contains data from AED_SALARYSAT.dta

Observations: 876

Data for A. Colin Cameron
(2022), ANALYSIS OF ECONOMIC
DATA, Amazon

Variables: 12

22 Jan 2022 15:50

Variable name	Storage type	Display format	Value label	Variable label
salary	float	%9.0g		Annual Salary in dollars
lnsalary	float	%9.0g		Natural Logarithm of Annual Salary
satverb	int	%32.0g		SAT test verbal score (highest self-reported score as of 2007)
satmath	int	%32.0g		SAT test mathematics score (highest self-reported score as of 2007)
highgrade	byte	%8.0g		Highest grade ever completed
age	float	%9.0g		Age in years
sex	byte	%14.0g		1 = female 0 = male
minority	float	%9.0g		1 = minority 0 = not minority
height	float	%9.0g		Height in inches
weight	int	%8.0g		Weight in pounds
genhealth	byte	%9.0g		Health status: 1=Excellent 2=VeryGood 3=Good 4=Fair 5=Poor

```
actscore      float    %32.0g          ACT test score (highest
                                         self-reported score as of 2007)
```

Sorted by:

- b) Suppose we want to run a regression of salary on SAT verbal score. Perform this regression and interpret the output.

```
reg salary satverb
```

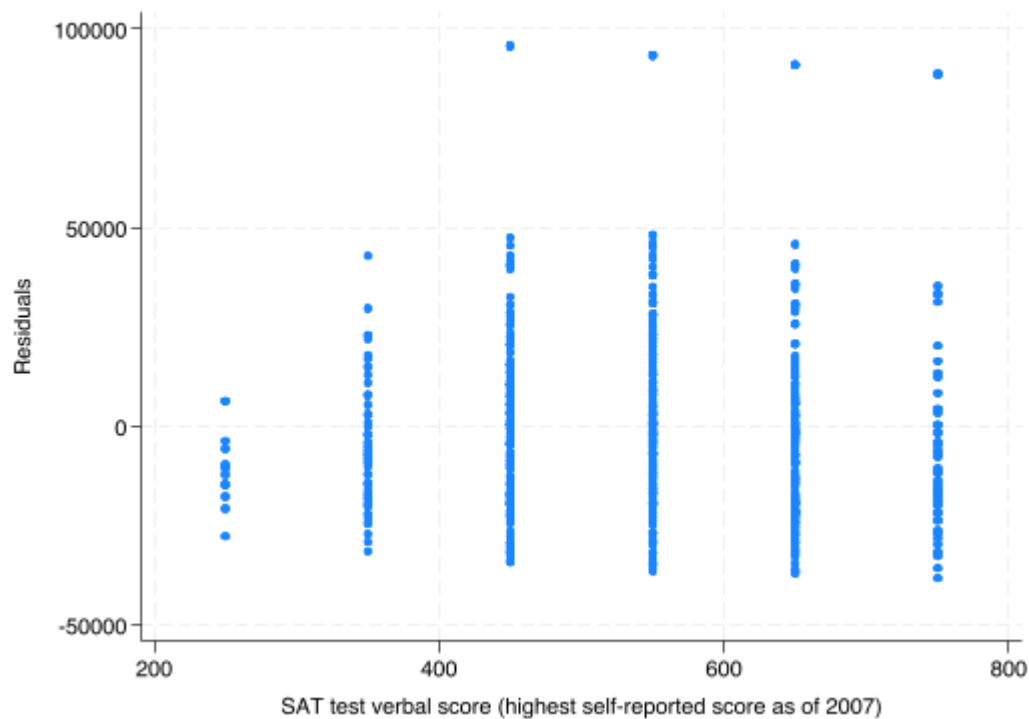
Source	SS	df	MS	Number of obs	=	876
Model	5.7575e+09	1	5.7575e+09	F(1, 874)	=	9.61
Residual	5.2357e+11	874	599052503	Prob > F	=	0.0020
Total	5.2933e+11	875	604947912	R-squared	=	0.0109
				Adj R-squared	=	0.0097
				Root MSE	=	24476

salary	Coefficient	Std. err.	t	P> t	[95% conf. interval]	
satverb	23.92566	7.717525	3.10	0.002	8.778612	39.07271
_cons	23752.62	4278.621	5.55	0.000	15355.05	32150.2

$\widehat{salary} = 23752.62 + 23.93 \times SATverb$. **Variation in SAT verbal score accounts for 1.09% of variation in salary.**

- c) Generate the residuals from the above regression using `predict resid, resid` and plot these against SAT verbal scores in a scatter plot. Do our errors appear to be homoskedastic or heteroskedastic?

```
predict resid, resid
sc resid satverb
```



Our errors appear to be a bit heteroskedastic, with larger errors for larger values of SAT verbal score.

- d) Make changes to the regression in part (b) to account for your answer in part (c). How does this change your estimated coefficients? How does this change your estimated standard errors? Does this make sense?

```
reg salary satverb, robust
```

Linear regression	Number of obs	=	876
	F(1, 874)	=	10.30
	Prob > F	=	0.0014
	R-squared	=	0.0109
	Root MSE	=	24476

		Robust			
salary	Coefficient	std. err.	t	P> t	[95% conf. interval]

satverb	23.92566	7.454351	3.21	0.001	9.295139	38.55618
_cons	23752.62	3922.598	6.06	0.000	16053.81	31451.44

We would want to add the `, robust` option to our regression to account for this heteroskedasticity. This should change our standard errors but not our point estimates for our coefficients. In this case, robust standard errors decrease our standard error and increase our t-score for our slope coefficient.

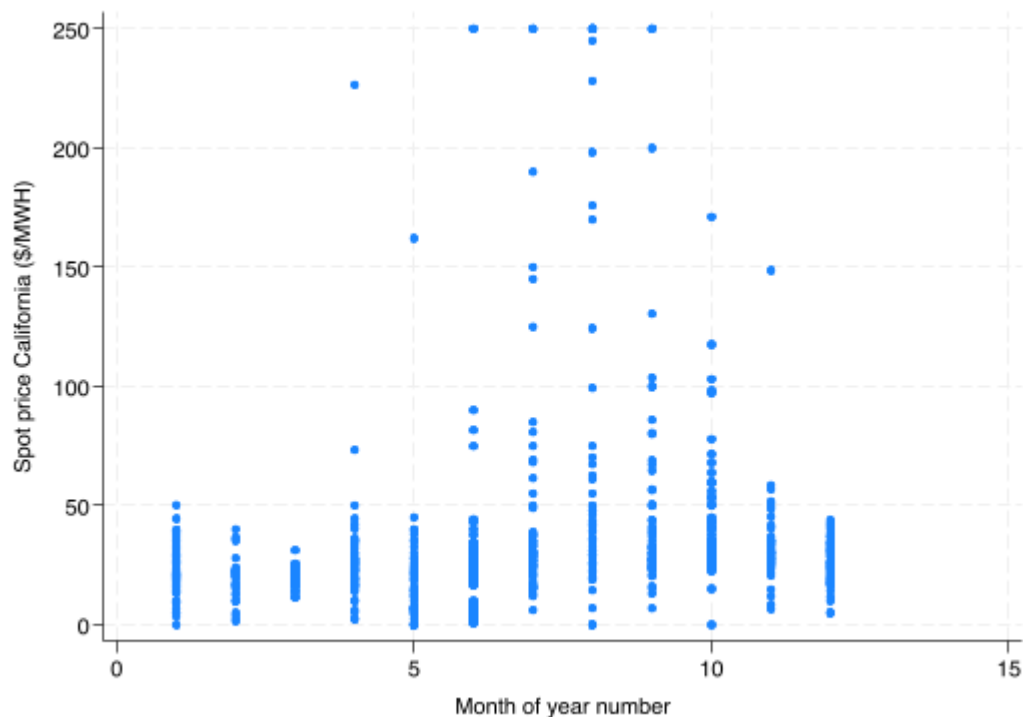
Question 4: Bivariate regression

Using AED_CALELECTRICITY.DTA from hw2, we now want to investigate when electricity gets more and less expensive.

- a) Create a scatterplot of the spot price of electricity and the month. What do you notice?

```
u AED_CALELECTRICITY, clear
sc niso month
```

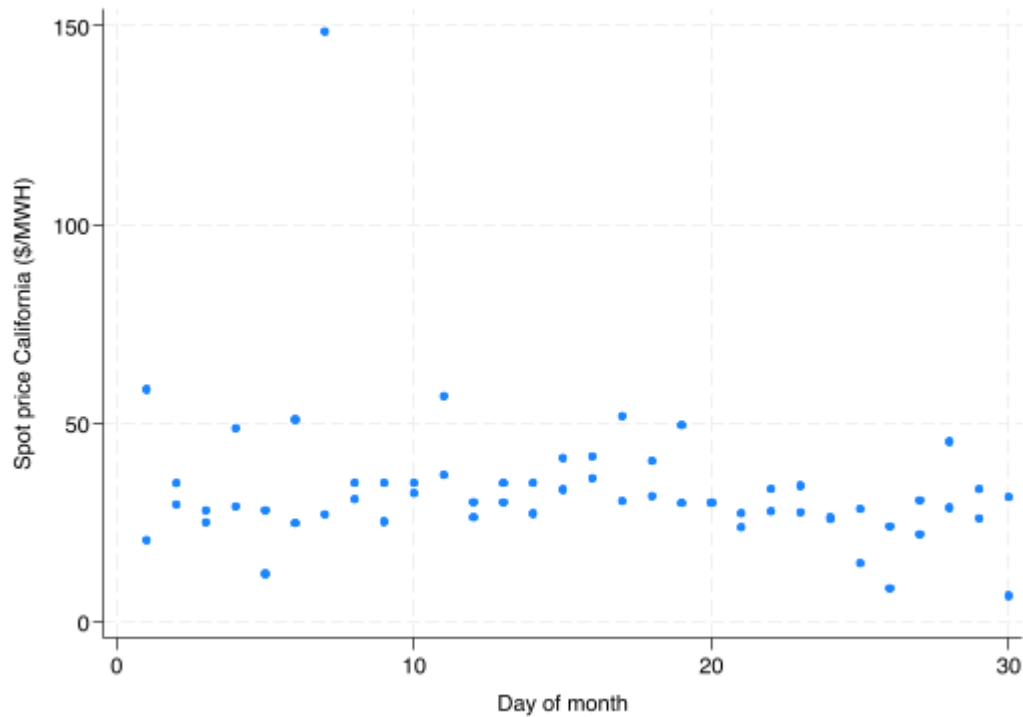
(Data for A. Colin Cameron (2015): Analysis of Economics Data, W.W. Norton)



We notice that the spot price of electricity is higher and more varied during the summer months.

- b) Create a scatterplot of the spot price and the day *only for the month of November*. What do you notice?

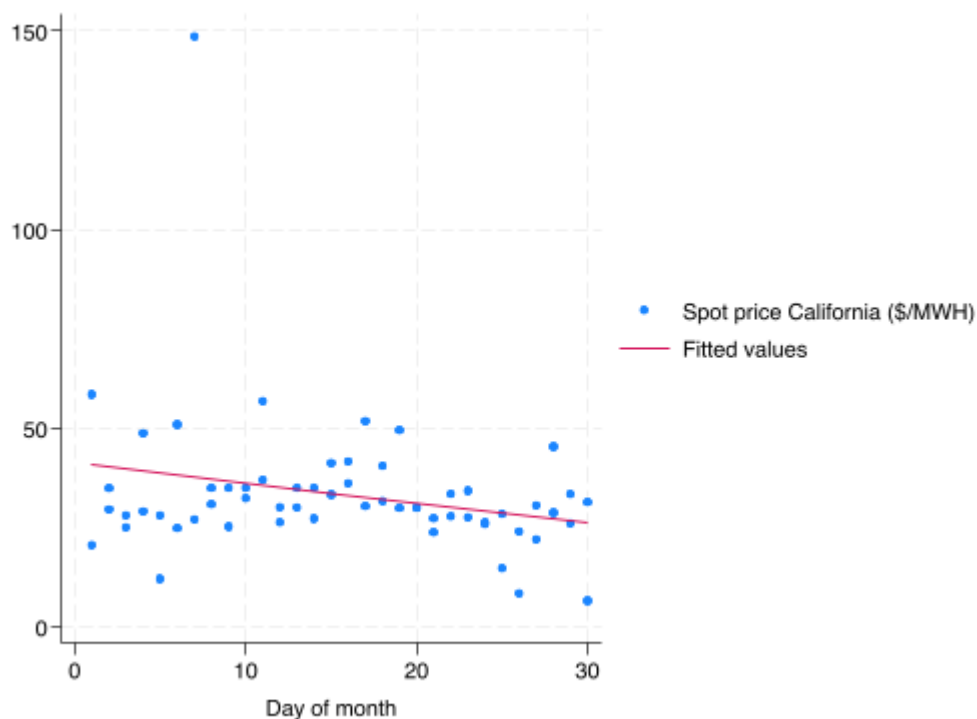
```
sc niso day if month==11
```



We notice that there is an outlier value of around \$150 for the spot price on one day, and that otherwise the spot price is relatively consistent.

- c) Overlay a best fit line onto your scatterplot from part (b). Without calculations or running a regression, what would you expect b_1 and b_2 to be for a regression of day on spot price in the month of November?

```
sc niso day if month==11 || lfit niso day if month==11
```



We would expect a $b_1 \approx 40, b_2 < 0$ given what we see from the best fit line.

d) Between *day* and *month*, which series is more correlated with the spot price?

```
corr niso month day
```

(obs=682)

	niso	month	day
niso	1.0000		
month	0.1685	1.0000	
day	0.0345	0.0309	1.0000

Month is more correlated with **niso** than **day** (0.1685 vs. 0.0345).

e) Perform the regression from part (b) of the spot price on day during November. Interpret your results.

```
reg niso day if month==11
```

Source		SS	df	MS	Number of obs	=	60
	+				F(1, 58)	=	3.63
Model		1142.9022	1	1142.9022	Prob > F	=	0.0617
Residual		18260.543	58	314.836949	R-squared	=	0.0589
	+				Adj R-squared	=	0.0427
Total		19403.4452	59	328.871953	Root MSE	=	17.744

niso		Coefficient	Std. err.	t	P> t	[95% conf. interval]
	+					
day		-.5042428	.2646537	-1.91	0.062	-1.034005 .0255192
_cons		41.28264	4.69838	8.79	0.000	31.87781 50.68747

We see that indeed our intercept term is near 40 and our slope is negative: electricity prices decrease throughout November.

- f) Suppose we are concerned our data violates the third assumption of bivariate linear regression. What could we add to part (e) to address this? Make this change and interpret your new results.

If we are worried that our data violates homoskedasticity, we can add the , robust option to our regression to accommodate heteroskedastic errors.

```
reg niso day if month==11, robust
```

Linear regression					Number of obs	=	60
					F(1, 58)	=	3.26
					Prob > F	=	0.0761
					R-squared	=	0.0589
					Root MSE	=	17.744

			Robust			
niso		Coefficient	std. err.	t	P> t	[95% conf. interval]
	+					
day		-.5042428	.2792251	-1.81	0.076	-1.063173 .054687

_cons		41.28264	6.187393	6.67	0.000	28.89723	53.66806
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g) What would we expect the spot price of electricity to be on the first day of November? On the last?

$$41.28 - 0.50 \times 1 = \$40.78$$

$$41.28 - 0.50 \times 30 = \$26.28$$

h) Find a day of the month in November with a positive residual. Find a day of the month in November with a negative residual.

```
predict residual if month==11, resid
list day residual if month==11
```

(622 missing values generated)

```
+-----+
| day    residual |
+-----+
215. | 1      -20.2084 |
216. | 2     -10.77416 |
217. | 3     -11.75792 |
218. | 4     -10.26567 |
219. | 5     -10.74923 |
+-----+
220. | 6     -13.38402 |
221. | 7     -10.75294 |
222. | 8      -6.391131 |
223. | 9     -11.50615 |
224. | 10     -3.870217 |
+-----+
225. | 11      1.264027 |
226. | 12     -5.15282 |
227. | 13      .2725123 |
228. | 14     -6.953245 |
229. | 15     -.4112209 |
+-----+
230. | 16      2.88871 |
231. | 17     -2.319207 |
232. | 18     -.578613 |
233. | 19     -1.742032 |
```

234.		20	-1.257788	

235.		21	-3.400575	
236.		22	-2.397103	
237.		23	4.5078	
238.		24	-3.319216	
239.		25	-.2365738	

240.		26	-4.223382	
241.		27	2.844042	
242.		28	1.516274	
243.		29	6.748436	
244.		30	5.244639	

580.		1	17.73547	
581.		2	-5.362309	
582.		3	-14.76754	
583.		4	9.423758	
584.		5	-26.75543	

585.		6	12.6519	
586.		7	110.8419	
587.		8	-2.248702	
588.		9	-1.753259	
589.		10	-1.249295	

590.		11	21.0517	
591.		12	-8.95045	
592.		13	-4.726318	
593.		14	.7754047	
594.		15	7.505836	

595.		16	8.373722	
596.		17	19.04674	
597.		18	8.315207	
598.		19	17.80392	
599.		20	-1.197088	

600.		21	-6.910365	
601.		22	3.276666	
602.		23	-2.148201	
603.		24	-2.899697	
604.		25	-13.92638	

605.	26 -19.77889
606.	27 -5.686239
607.	28 18.177
608.	29 -.658703
609.	30 -19.59514
	+-----+

- i) Interpret the t-stat, p-value, and confidence interval for b_2 in your regression from part (f). What test does this correspond to? What are its hypotheses? What is its conclusion and why?

This is our test of association between *day* and *niso*.

$$H_0 : \beta_2 = 0; \quad H_A : \beta_2 \neq 0$$

With $\alpha = 0.05$, we fail to reject the test of association as our p-value is $0.076 > 0.05$. We can also confirm this conclusion with our critical value approach or our confidence interval including 0.