Analysis of Economics Data Chapter 9: Models with Natural Logarithms

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CHAPTER 9: Models with Natural Logarithms

• Economists are often interested in measuring proportionate changes

- e.g. price elasticity of demand
- e.g. percentage change in earnings with one more year of education
- natural logarithms are useful for this.
- Additional uses of the natural logarithm include
 - eliminating right skewness in data (chapter 2)
 - compounding and the rule of 72
 - linearizing exponential growth.

Outline

- Natural Logarithm Function
- Semi-elasticities and elasticities
- Southand Content of Content of
- Example: Earnings and Education
- Further Uses of the Natural Logarithm
- Section 2 Exponential Function
- Dataset: EARNINGS

9.1 Natural Logarithm Function

- A logarithmic function is the reverse operation to raising a number to a power
 - e.g. $10^2 = 100$ implies that $\log_{10} 100 = 2$
 - if 10 raised to the power 2 equals 100 then the logarithm to the base 10 of 100 is 2.
- More generally

$$a^b = x \Rightarrow \log_a x = b;$$

- the logarithm to the base *a* of *x* equals *b*.
- Most obvious choice of the base *a* is base 10 (decimal system).
- Economics often uses logarithm to base *e*, the natural logarithm
 - where $e\simeq 2.71828...$ is a transcendental number like π

$$\ln x = \log_e(x), \quad x > 0.$$

Approximating Proportionate Changes

- $\Delta x = x_1 x_0$ is the **change** in x when x changes from x_0 to x_1 .
- The proportionate change in x is

$$\frac{\Delta x}{x_0} = \frac{x_1 - x_0}{x_0}$$

- Example: Change from $x_0 = 40$ to $x_1 = 40.4$
 - $\Delta x = 40.4 40 = 0.4$
 - proportionate change in x is $\Delta x/x_0 = 0.4/40 = 0.01$
 - and percentage change is $100 \times 0.01 = 1\%$.

Approximating Proportionate Changes (continued)

• We have

$$\begin{array}{l} \frac{d \ln x}{dx} = \frac{1}{x} & \text{from calculus} \\ \Rightarrow \quad \frac{\Delta \ln x}{\Delta x} \simeq \frac{1}{x} & \text{for small } \frac{\Delta x}{x} \\ \Rightarrow \quad \Delta \ln x \simeq \frac{\Delta x}{x} & \text{rearranging} \end{array}$$

• For small proportionate changes we use the approximation

$$\Delta \ln \mathbf{x} \simeq \frac{\Delta \mathbf{x}}{\mathbf{x}}$$
 for small $\frac{\Delta x}{x}$ (say $\frac{\Delta x}{x} < 0.1$).

• Multiplying by 100 yields percentage changes, so equivalently

 $100 \times \Delta \ln x \simeq$ Percentage change in x.

- Example: Change from $x_0 = 40$ to $x_1 = 40.4$
 - ▶ approximation is $ln(40.4) ln(40) = 3.69883 3.68888 \simeq 0.00995$
 - exact is $\Delta x/x_0 = (40.4 40)/40 = 0.01$.

9.2 Semi-elasticity and Elasticity

• The semi-elasticity of y with respect to x is the ratio of the proportionate change in y to the change in the level of x

Semi – elasticity_{yx} =
$$\frac{\Delta y / y}{\Delta x}$$
.

- Multiplying by 100 gives the percentage change in y when x changes by one unit.
- Example: semi-elasticity of earnings with respect to years of schooling is 0.08
 - one more year of schooling is associated with a 0.08 proportionate change in earnings
 - one more year of schooling is associated with an 8% change in earnings.

Semi-elasticity and Elasticity (continued)

• The elasticity of y with respect to x is the proportionate change of y for a given **proportionate change** in x

$$\textit{Elasticity}_{yx} = \frac{\Delta y / y}{\Delta x / x} = \frac{\Delta y}{\Delta x} \times \frac{x}{y}$$

- Example price elasticity of demand for a good is -2
 - a one percent increase in price leads to a 2 percent decrease in demand.

Approximation of Semi-Elasticity and Elasticity

- Since $\frac{\Delta y}{y} \simeq \Delta \ln y$ and $\frac{\Delta x}{x} \simeq \Delta \ln x$ we obtain the following.
- Semi-elasticities and elasticities can be approximated as following

- OLS regression of models that first transform variables to natural logarithms can directly estimate semi-elasticities and elasticities.
- Example: if $\ln y = a + b \ln x$ then the slope $b = \frac{\Delta \ln y}{\Delta \ln x} =$ the elasticity.
 - so we can obtain the semi-elasticity by regressing $\ln y$ on x.

9.3 Log-linear Model

• The log-linear or log-level model regresses ln y on x

- with fitted value $\widehat{\ln y} = b_1 + b_2 x$
- ► the slope coefficient b₂ = ∆ln y / ∆x is an estimate of the semi-elasticity of y with respect to x
- we need y > 0 since only then is $\ln y$ defined.
- This is a very common model for right-skewed data such as individual earnings.

Log-log Model

- The log-log model regresses ln y on ln x
 - with fitted value $\widehat{\ln y} = b_1 + b_2 \ln x$
 - the slope coefficient $b_2 = b_2 = \Delta \widehat{\ln y} / \Delta \ln x$ is an estimate of the elasticity of y with respect to x
 - we need y > 0 and x > 0 since only then are $\ln y$ and $\ln x$ defined.

Linear-log Model

• The linear-log model or level-log regresses ln y on lnx

- with fitted value $\hat{y} = b_1 + b_2 \ln x$
- ▶ $b_2/100$ is an estimate of the change in y in response to a one percent change in x.
- we need x > 0 since only then is $\ln x$ defined.

Summary: Models with Logs

• We have

Model	Specification	Interpretation of b_2
Linear	$\widehat{y} = b_1 + b_2 x$	Slope: $\Delta y / \Delta x$
Log-Linear	$\widehat{\ln y} = b_1 + b_2 x$	Semi-elasticity: $(\Delta y/y)/\Delta x$
Log-log	$\widehat{\ln y} = b_1 + b_2 \ln x$	Elasticity: $(\Delta y/y)/(\Delta x/x)$
Linear-log	$\widehat{y} = b_1 + b_2 \ln x$	$\Delta y/(\Delta x/x)$

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9.4 Example: Earnings and Education

• Dataset EARNINGS on 172 full-time male workers in 2010 aged 30 years.

		Standard			
Variable	Definition	Mean	Deviation	Min	Max
Earnings	Annual earnings in \$	41413	25527	1050	172000
Lnearn	Natural logarithm of <i>Earnings</i>	10.46	0.62	6.96	12.05
Education	Years of completed schooling	14.43	2.73	3	20
Lneduc	Natural logarithm of Education	2.65	0.22	1.10	3.00
n	171				

• OLS regression of *Earnings* (y) on *Education* (x) yields (t-statistics in parentheses)

$$\widehat{y} = -31056 + 5021x$$
, $R^2 = .290$

Linear Model and Log-Linear Model

- Linear model: *Earnings* = -31056 + 5021 *Education*
- Log-linear model: ln(Earnings) = 8.561 + 0.131 Education



Comparison of Models with Earnings Data

• y is earnings and x is education (with t-statistics in parentheses).

Model	Estimates	R^2	Slope	Semi-elasticity	Elasticity
Linear	$\widehat{y} = -31056 + 5021x$ (-3.49) (8.30)	0.289	5021		
Log-linear	$\widehat{\ln y} = \underset{(40.83)}{8.561} + \underset{(9.21)}{0.131x}$	0.334	-	0.131	
Log-log	$\widehat{\ln y} = \frac{6.543}{(13.70)} + \frac{1.478}{(8.23)} \ln x$	0.286	-	-	1.478
Linear-log	$\widehat{y} = -102767 + 54452 \ln x$ (-5.05) (7.11)	0.230	-	-	-

- Linear: one year more of education is associated with a \$5,021 increase in earnings
- Log-linear: one year more of education is associated with a 13.1% increase in earnings
- Log-log: 1% increase in education is associated with a 1.478% increase in earnings
- Linear-log: 1% increase in education is associated with a \$544 (= 54452/100) increase in earnings.

9.5 Approximating Natural Logarithm

•
$$\ln(1+x) = x - \frac{x^2}{2} + \frac{x^3}{3} - \frac{x^4}{4} + \frac{x^5}{5} - \cdots$$

• e.g. $\ln(1.1) = 1 - 0.1 + \frac{0.01}{2} - \frac{0.001}{3} + \cdots$
 $\simeq 1 - 0.1 + 0.005 - 0.00033 \simeq 0.0953.$

• So for small x we have the approximation

$$\ln(1+x) \simeq x$$
, for, say, $x < 0.1$.

- Approximation good for small x, but x increasingly overestimates $\ln(1+x)$
 - for x < 0.10 the approximation is within five percent of $\ln(1+x)$
 - for x = 0.2, for example, the approximation is ten percent larger than $\ln 1.2 = 0.1823$.

		"Small" x		"'Larger" x		
		x=0.05	x=0.10	x=0.15	x=0.20	x=0.50
True Value	ln(1+x)	0.0488	0.0953	0.1398	0.1823	0.4055
Approximation	х	0.05	0.10	0.15	0.20	0.50

Compounding and the Rule of 72

- **Rule of 72**: a series growing at percentage rate r takes approximately $\frac{72}{r}$ periods to double.
- Example: Invest at 4% per annum doubles in 72/4 = 18 years.
- Reason:
 - After *n* periods at rate *r* investment is $(1 + r)^n$ times larger.
 - Money doubles if *n* solves $(1 + r)^n = 2$
 - Solution is $n = \ln 2/[\ln(1+r)]$.
 - Approximate: $\ln(1+r) \simeq r$ for small r.
 - ▶ Approximate: ln 2 = 0.6931 ≃ 0.72.
 - So $n = \ln 2/[\ln(1+r)] \simeq 0.72/r$.
- Example: r = 0.04 (so 4%)
 - true value: $\ln 2/[\ln(1+0.04)] = 17.67$ so doubles in 17.67 years
 - rule of 72: 72/4 = 18 so doubles in 18 years.
- More precisely can have rule of 70, or 69, or 69.3.

Linearizing Exponential Growth

 Many data series grow according to a power law, or exponentially, over time, rather than linearly.

$$x_t = x_0 \times (1+r)^t$$

- Here x₀ is value at time 0
- x_t is value at time t
- r is the constant growth rate (or decay rate if r < 0).
- Example: \$100 invested at 3% annual interest rate for 10 years.
 - annual growth rate is r = 3/100 = 0.03
 - investment worth $100 \times (1.03)^{10}$ or \$134.39 after ten years.

Linearizing Exponential Growth (continued)

• Taking the natural logarithm of $x_t = x_0 imes (1+r)^t$ yields

$$n x_t = \ln(x_0(1+r)^t)$$

= $\ln x_0 + \ln(1+r)^t$
= $\ln x_0 + \ln(1+r) \times t$
 $\simeq \ln x_0 + r \times t$ for small r.

• Exponential growth is linear growth in logs!

Example: S&P 500

- Standard and Poor 500 Index 1927-2019
 - no inflation adjustment and no dividends
 - Ieft panel: exponential growth in level
 - right panel: linear growth in logs

★ growth rate
$$\simeq rac{7.8-1.8}{2019-1927} = rac{6.0}{92} = 0.065$$
 or 6.5% per annum



21 / 26

9.6 Exponential Function

- What is e?
 - e is an irrational number that is approximately 2.7182818
 - e is a transcendental number, like $\pi\simeq 3.142$
 - unlike for π , there is no simple physical interpretation for *e*.
- The exponential function is denoted

$$\exp(x)=e^x.$$

The natural logarithm is the reverse operation to exponentiation.
Then

$$y = e^x \Rightarrow x = \ln y.$$

• For example, $e^2 \simeq 7.38906$ so $\ln 7.38906 \simeq 2.0$.

Approximating Exponential

•
$$\exp(x) = 1 + x + \frac{x^2}{2!} + \frac{x^3}{3!} + \frac{x^4}{4!} + \frac{x^5}{5!} + \cdots$$

• e.g. $\exp(1.1) = 1 + 0.1 + \frac{0.01}{2} + \frac{0.001}{6} + \cdots$
 $\simeq 1 + 0.1 + 0.005 + 0.00016 \simeq 1.1052.$

So for small x

$$e^x \simeq 1 + x$$
, for, say, $x < 0.1$.

 Approximation good for small x, but increasingly underestimates e^x as x increases.

Table: Approximating exp(x) by 1+x.

		"Small" x		"Larger" x			
		x=0.05	x=0.10	x=0.15	x=0.20	x=0.50	
True Value	exp(x)	1.0513	1.1052	1.1618	1.2214	1.6487	
Approximation	1+x	1.05	1.10	1.15	1.20	1.50	

Compound Interest Rates

- Consider compound for a year and find the annual percentage yield (APY).
- Suppose have 12% per annum then APY= 12%.
- Suppose compound monthly at 12/12 = 1% per month

•
$$(1+0.01)^{12} = 1.12683$$
 so APY= 12.683%.

• Suppose compound daily at 12/365% per day

•
$$(1+0.12/365)^{365} = 1.127547$$
 so APY= 12.747%.

• If continuously compound for progressively smaller intervals at rate r

$$(1+r/n)^n \to e^r$$
 as $n \to \infty$.

• Here $(1 + 0.12/n)^n \rightarrow \exp(0.12) = 1.12750$ or 12.750%.

Key Stata Commands

```
clear
use AED_EARNINGS.DTA
generate lnearn = ln(earnings)
generate lneduc = ln(education)
* Linear Model
regress earnings education, vce(robust)
* Log-linear Model
regress lnearn education
* Log-log Model
regress lnearn lneduc
* Linear-log Model
regress earnings lneduc
```

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Some in-class Exercises

- Consider numbers a and b with ln a = 3.20 and ln b = 3.25. Using only this information, what is the approximate percentage change in going from a to b?
- Obemand for a good falls from 100 to 90 when the price increases from 20 to 21. Compute the price elasticity of demand.
- We estimate ln y = 3 + 0.5 ln x. Give the elasticity of y with respect to x.
- We estimate ln y = 6 + 0.2x. Give the response of y when x changes by one unit.
- How long does it take for prices to double given 4% annual inflation?
- Suppose y = α × (1 + β)^x. Explain how to estimate α and β using OLS.