

V. Extensions of Multiple Regression

A. Dummy Variables

B. Scaling Variables

C. Non-Linear Models

1. Basic Idea for Estimation.
2. Models that don't require logs
3. Models that require log transform.

B. Scaling Variables

1. *Scaling the Dependent Variable:*

- The effects of scaling by a known constant are perfectly predictable:
Suppose we multiply Y by 'a'

$$\hat{\beta}_1^* = \frac{\sum x_i (a \cdot Y_i)}{\sum x_i^2} = (a) \frac{\sum x_i Y_i}{\sum x_i^2} = \hat{\beta}_1 (a)$$

Scaling by a constant has no real consequence.

2. *Scaling the Independent Variable:*

- The effects of scaling by a known constant are again perfectly predictable:
Suppose we multiply X by 'b'

$$\hat{\beta}_1^* = \frac{\sum (b \cdot x_i) Y_i}{\sum (b \cdot x_i)^2} = \left(\frac{1}{b}\right) \frac{\sum x_i Y_i}{\sum x_i^2} = \hat{\beta}_1 \left(\frac{1}{b}\right)$$

Scaling by a constant, no real consequence.

3. *Scaling both Variables* – combine the two results – again, no real consequences.

4. Indexes – the case of an unknown scale.

- The independent variable, and possibly the dependent variable, may be “indexes.”
- Examples: Consumer Price Indexes.
- You have no idea about the “base year” value.
- Indexes – pure numbers, no “units.”
- Solution – **calculate elasticities**, pure numbers.
- A 1% change in X causes an ε % change in Y.

C. Non-Linear Models

1. **Basic idea:** Make the model look linear in the parameters. Then estimate using OLS.

2. **Models that don't require a log transform.**

Quadratic form:

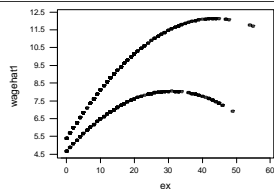
Examples:

Interpretation:

```
wage = - 5.49 + 0.905 ed + 0.309 ex - 0.00352 exsq  
      - 0.726 fe - 0.0903 feex
```

Predictor	Coef	StDev	T	P
Constant	-5.494	1.2140	-4.53	0.000
ed	0.9046	0.0791	11.43	0.000
ex	0.3093	0.0554	5.59	0.000
exsq	-0.0035	0.0012	-2.98	0.003
fe	-0.7264	0.6725	-1.08	0.281
feex	-0.0903	0.0309	-2.93	0.004

S = 4.388 R-Sq = 27.8% R-Sq(adj) = 27.1%



C. Non-Linear Models

Inverse form: $Y = \beta_0 + \beta_1 (1/X) + u$

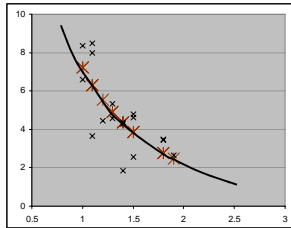
Example:

Interpretation:

The regression equation is
w-dot = - 1.04 + 7.91 INV(U)

Predictor	Coef	SE Coef	T	P
Constant	-1.044	2.373	-0.44	0.666
INV(U)	7.907	3.164	2.50	0.025

S = 1.734 R-Sq = 29.4% R-Sq(adj) = 24.7%



C. Non-Linear Models

3. Models that require a log transform.

- **Cobb-Douglas:** log-log models.

- Transform using natural logarithm.

- Interpretation:

Regression Analysis: Insales versus lnprose, lnpcarn, lndinc

The regression equation is
 $\text{lnsales} = 6.29 - 1.86 \text{ lnprose} + 1.45 \text{ lnpcarn} + 0.560 \text{ lndinc}$

Predictor	Coef	SE Coef	T	P
Constant	6.288	4.875	1.29	0.221
lnprose	-1.8562	0.3438	-5.40	0.000
lnpcarn	1.4541	0.5724	2.54	0.026
lndinc	0.5596	0.9211	0.61	0.555

S = 0.175868 R-Sq = 73.7% R-Sq(adj) = 67.2%

Analysis of Variance

Source	DF	SS	MS	F	P
Regression	3	1.04145	0.34715	11.22	0.001
Residual Error	12	0.37115	0.03093		
Total	15	1.41260			

3. Models that require a log transform.

- Semi-Log models:** log-linear

$$Y = \exp^{\beta_0 + \beta_1 X_1 + \beta_2 X_2 + u}$$

- Transform using natural logarithm.
- Parameters:

3. Models that require a log transform.

- Semi-Log models:** linear-log.

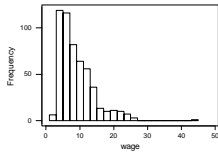
$$\exp^Y = AX_1^{\beta_1} X_2^{\beta_2} \exp^u$$

- Transform using natural logarithm.
- Parameters:

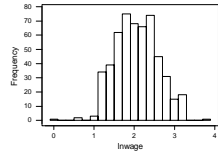
Normality Assumption

- What did we assume was normally distributed?
- What part of our analysis relies on this assumption?
- Some data are not normally distributed – wages.
- Log transformation – the natural log is normal.

Distribution of Wages



Distribution of Ln(Wage)



$$\lnwage = -0.921 + 0.988 \lned + 0.235 \lnex + 0.047 fe - 0.124 \lnexfe$$

Predictor	Coef	StDev	T	P
Constant	-0.9212	0.2524	-3.65	0.000
lned	0.98774	0.08731	11.31	0.000
lnex	0.23505	0.03111	7.56	0.000
fe	0.0475	0.1256	0.38	0.706
lnexfe	-0.12383	0.04555	-2.72	0.007

S = 0.4471 R Sq = 27.4% R Sq(adj) = 26.8%

