

## How to Interpret Nonlinear Regression Coefficients

### Summary

Name	Model	Interpretation
Level-level (basic regression)	$y = \beta_0 + \beta_1 x + \epsilon$	As x increases one unit, y changes by $\beta_1$ units
Level-log (logarithmic)	$y = \beta_0 + \beta_1 \ln(x) + \epsilon$	As x increases 1%, y changes by $(\beta_1/100)$ units
Log-level (exponential)	$\ln(y) = \beta_0 + \beta_1 x + \epsilon$	As x increases 1 unit, y changes by $(\beta_1 * 100)\%$
Log-log (elasticities)	$\ln(y) = \beta_0 + \beta_1 \ln(x) + \epsilon$	As x increases 1%, y changes by $\beta_1\%$
Quadratic model	$y = \beta_0 + \beta_1 x + \beta_2 x^2 + \epsilon$	As x increases 1 unit, y changes $\beta_1 + 2\beta_2 x$

### Extended Examples

Suppose we have three explanatory variables:

- $x_1$  a continuous variable
- $\ln(x_2)$  the natural log of a continuous variable
- $x_3$  a dummy variable that equals 1 or 0

Suppose we have three regression models. Each model uses the same explanatory variables listed above, but each uses a different dependent variable:

- $y_1$  a continuous variable
- $\ln(y_2)$  the natural log of a continuous variable
- $y_3$  a dummy variable that equals 1 or 0

Below each model is a description of how to interpret each regression coefficient.

**Model 1:**  $y_1 = \beta_0 + \beta_1 x_1 + \beta_2 \ln(x_2) + \beta_3 x_3 + \epsilon$

- $\beta_1$  = a one-unit change in  $x_1$  generates a  $\beta_1$  unit change in  $y_1$
- $\beta_2$  = a 100% change in  $x_2$  generates a  $\beta_2$  unit change in  $y_1$
- $\beta_3$  = a switch in  $x_3$  from 0 to 1 generates a  $\beta_3$  unit change in  $y_1$

**Model 2:**  $\ln(y_2) = \beta_0 + \beta_1 x_1 + \beta_2 \ln(x_2) + \beta_3 x_3 + \epsilon$

- $\beta_1$  = a one-unit change in  $x_1$  generates a  $100 * \beta_1$  percent change in  $y_2$
- $\beta_2$  = a 100% change in  $x_2$  generates a  $100 * \beta_2$  percent change in  $y_2$
- $\beta_3$  = a switch in  $x_3$  from 0 to 1 generates a  $100 * \beta_3$  percent change in  $y_2$

**Model 3:**  $y_3 = \beta_0 + \beta_1 x_1 + \beta_2 \ln(x_2) + \beta_3 x_3 + \epsilon$

- $\beta_1$  = a one-unit change in  $x_1$  generates a  $100 * \beta_1$  percentage point change in the probability  $y_3$  occurs
- $\beta_2$  = a 100% change in  $x_2$  generates a  $100 * \beta_2$  percentage point change in the probability  $y_3$  occurs
- $\beta_3$  = a switch in  $x_3$  from 0 to 1 generates a  $100 * \beta_3$  percentage point change in the probability  $y_3$  occurs