

Linearization, Log-Linearization, and Higher-Order Local (Level and Log) Approximations: Notes

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0.1 Non-Linear System

Write the non-linear conditions describing the solution of the original system as

$$f(x_t, y_t) = g(z_t). \quad (1)$$

As a concrete example, suppose $f(x_t, y_t) = x_t^\sigma + x_t y_t$ and $g(z_t) = z_t^\nu$.

0.2 Linearization in Levels

Let \bar{x} , \bar{y} , and \bar{z} denote the steady-state values of x_t , y_t , and z_t , respectively (more generally, they are simply the values of the variables around which the local approximation is made). A first-order Taylor expansion around $(\bar{x}, \bar{y}, \bar{z})$ gives us

$$f(\bar{x}, \bar{y}) + f_x(\bar{x}, \bar{y})(x_t - \bar{x}) + f_y(\bar{x}, \bar{y})(y_t - \bar{y}) = g(\bar{z}) + g'(\bar{z})(z_t - \bar{z}). \quad (2)$$

Because $f(\bar{x}, \bar{y}) = g(\bar{z})$, these terms drop out. In our example, $f_x = \sigma x_t^{\sigma-1} + y_t$, $f_y = x_t$, and $g' = \nu z_t^{\nu-1}$, so

$$\left[\sigma \bar{x}^{\sigma-1} + \bar{y} \right] (x_t - \bar{x}) + \bar{x} (y_t - \bar{y}) = \nu \bar{z}^{\nu-1} (z_t - \bar{z}) \quad (3)$$

is the linearized system in levels.

0.3 Linearization in Logs

Make a transformation of variables. Define $\hat{x}_t \equiv \ln x_t$, $\hat{y}_t \equiv \ln y_t$, $\hat{z}_t \equiv \ln z_t$. By construction, then, $x_t = \exp(\hat{x}_t)$. Rewrite the conditions describing the solution of the model as

$$f(\exp(\hat{x}_t), \exp(\hat{y}_t)) = g(\exp(\hat{z}_t)), \quad (4)$$

which is obviously identical to the solution of the original system. The functions f and g should now be thought of as functions of the transformed variables \hat{x}_t , \hat{y}_t , and \hat{z}_t . Denote by $\bar{\hat{x}}$ the steady-state value of \hat{x}_t , and similarly for $\bar{\hat{y}}$ and $\bar{\hat{z}}$. Clearly, $\bar{\hat{x}} = \ln \bar{x}$, etc. Now linearize this transformed system around the point $(\bar{\hat{x}}, \bar{\hat{y}}, \bar{\hat{z}})$, giving us

$$f(\bar{\hat{x}}, \bar{\hat{y}}) + f_{\hat{x}}(\bar{\hat{x}}, \bar{\hat{y}})(\hat{x}_t - \bar{\hat{x}}) + f_{\hat{y}}(\bar{\hat{x}}, \bar{\hat{y}})(\hat{y}_t - \bar{\hat{y}}) = g(\bar{\hat{z}}) + g'(\bar{\hat{z}})(\hat{z}_t - \bar{\hat{z}}). \quad (5)$$

Because $f(\bar{\hat{x}}, \bar{\hat{y}}) = g(\bar{\hat{z}})$, these terms drop out. The derivatives we need now are

$$f_{\hat{x}} = \sigma (\exp(\hat{x}_t))^\sigma + \exp(\hat{x}_t) \exp(\hat{y}_t), \quad (6)$$

$$f_{\hat{y}} = \exp(\hat{x}_t) \exp(\hat{y}_t), \quad (7)$$

$$g' = \nu (\exp(\hat{z}_t))^\nu. \quad (8)$$

Evaluating these derivatives at the point of expansion $(\bar{\hat{x}}, \bar{\hat{y}}, \bar{\hat{z}})$, the log-linearized system is

$$[\sigma (\exp(\bar{\hat{x}}))^\sigma + \exp(\bar{\hat{x}}) \exp(\bar{\hat{y}})] (\hat{x}_t - \bar{\hat{x}}) + \exp(\bar{\hat{x}}) \exp(\bar{\hat{y}}) (\hat{y}_t - \bar{\hat{y}}) = \nu (\exp(\bar{\hat{z}}))^\nu (\hat{z}_t - \bar{\hat{z}}). \quad (9)$$

Expressing this equation in terms of the original variables x_t , y_t , z_t ,

$$[\sigma \bar{x}^\sigma + \bar{x}\bar{y}] (\ln x_t - \ln \bar{x}) + \bar{x}\bar{y} (\ln y_t - \ln \bar{y}) = \nu \bar{z}^\nu (\ln z_t - \ln \bar{z}). \quad (10)$$

Equation (10) is the log-linearized approximation of (1), while equation (3) is the (level-)linearized approximation. In (10), it is often customary to define the terms $(\ln x_t - \ln \bar{x})$, $(\ln y_t - \ln \bar{y})$, and $(\ln z_t - \ln \bar{z})$ as $\hat{\hat{x}}_t$, $\hat{\hat{y}}_t$, $\hat{\hat{z}}_t$. **Note that these hatted variables are not the same as the hatted variables we defined in the above derivations.**

0.4 Higher-Order Approximations

Second- and higher-order approximations in both levels and logs are now straightforward to obtain. For example, to obtain the quadratic approximation in levels, simply take a second-order Taylor expansion of (1) around $(\bar{x}, \bar{y}, \bar{z})$. To obtain a log-quadratic approximation, take a second-order Taylor expansion of (4) around $(\bar{\hat{x}}, \bar{\hat{y}}, \bar{\hat{z}})$. And so on for higher-order approximations.