

Tutorial, Part 4

Example 1

$$\hat{C} = a_1 + \underbrace{0.92 D1}_{(a_2)} \quad \text{total effect of } D1 \text{ on } C$$

$$\hat{CPI} = c_1 + \underbrace{0.78 D1}_{(c_2)} \quad \text{total effect of } D1 \text{ on CPI}$$

$$\hat{C} = b_1 + b_2 D1 + b_3 CPI \quad \text{var}(b_3) = 0.09$$

$$t: \quad (4.64) \quad (1.95)$$

$$b_2 = ?$$

Effect of D1 on C decomposition:

DIRECT : D1 on C $\rightarrow b_2$

+ INDIRECT : D1 on C via CPI $\rightarrow c_2 \cdot b_3 = 0.78 b_3$

TOTAL : D1 on C $\rightarrow a_2 = 0.92$

$$b_2 + 0.78 b_3 = 0.92$$

$$b_2 = 0.92 - 0.78 b_3$$

$$t_{b_3} = \frac{b_3 - \beta_3^0}{\sqrt{\text{var}(b_3)}} = \frac{b_3}{\sqrt{0.09}} = 1.95$$

$$b_3 = 1.95 \cdot \sqrt{0.09} = 0.585$$

$$\underline{b_2} = 0.92 - 0.78 \cdot 0.585 = \underline{\underline{0.4637}}$$

Example 2

Model 1: linear

$$\begin{array}{ll} \text{PRM} & MRT_{(i)} = \beta_1 + \beta_2 \text{POP65}_{(i)} + \beta_3 \text{TOBACCO}_{(i)} + u_{(i)} \\ \text{SRM} & \hat{MRT}_{(i)} = b_1 + b_2 \text{POP65}_{(i)} + b_3 \text{TOBACCO}_{(i)} \\ & \hat{MRT} = 54.765 + 52.793 \text{POP65} + 1.441 \text{TOBACCO} \end{array}$$

Interpretation (e.g. $b_2 = 52.793$): If the percentage of the population above age 65 increases by 1 percentage point the number of deaths per 100 000 inhabitants on average increases by 52.8 deaths, ceteris paribus.

Model 2: log-linear

$$MRT = \beta_1 \text{POP65}^{\beta_2} \text{TOBACCO}^{\beta_3} u$$

$$\ln \hat{MRT} = b_1 + b_2 \ln \text{POP65} + b_3 \ln \text{TOBACCO}$$

$$\hat{\ln MRT} = 3.449 + 0.786 \ln \text{POP65} + 0.283 \ln \text{TOBACCO}$$

Let's consider: $\ln Y_i = \beta_1 + \beta_2 \ln X_i + u_i$

$$\begin{aligned} i=1 &\rightarrow \ln Y_1 = \beta_1 + \beta_2 \ln X_1 \\ i=2 &\rightarrow \ln Y_2 = \beta_1 + \beta_2 \ln X_2 \end{aligned} \quad \begin{matrix} \text{subtract} \\ \text{cancel out} \end{matrix} \quad \begin{matrix} \text{(assume:} \\ u_1 \text{ and } u_2 \text{ small)} \end{matrix}$$

$$\ln Y_2 - \ln Y_1 = \beta_2 \ln X_2 - \beta_2 \ln X_1$$

$$\Delta \ln Y = \beta_2 \Delta \ln X \quad *$$

$$\Delta \ln Y = \ln \frac{Y_2}{Y_1} = \ln \frac{Y_1 + (\bar{Y}_2 - \bar{Y}_1)}{Y_1} = \ln \left(1 + \frac{\Delta Y}{Y_1} \right) \approx \frac{\Delta Y}{Y_1}$$

(approximation: $\ln(1+z) \approx z$ if $z \approx 0$ (really small))

$$\Delta \ln Y \approx \frac{\Delta Y}{Y_1} \quad \text{relative change}$$

$$\Delta \ln X \approx \frac{\Delta X}{X_1}$$

$$* \quad \frac{\Delta Y}{Y_1} \approx \beta_2 \frac{\Delta X}{X_1} / \cdot 100$$

$$\frac{100 \Delta Y}{Y_1} \approx \beta_2 \frac{100 \Delta X}{X_1}$$

$$\% \Delta Y \approx \beta_2 \% \Delta X \rightarrow \text{when } \% \Delta X = 1\%$$

$$\% \Delta Y \approx \beta_2 \% \text{ ELASTICITY}$$

Interpretation (e.g. $\beta_2 = 0.786$): If the percentage of the population above age 65 increases by 1 % the number of deaths per 100 000 inhabitants on average increases by (approximately) 0.786 %, *ceteris paribus*.

Model 3: log - lin

$$MRT = e^{\beta_1 + \beta_2 \text{POP65} + \beta_3 \text{TOBACCO}}$$

$$\ln \hat{MRT} = \beta_1 + \beta_2 \text{POP65} + \beta_3 \text{TOBACCO}$$

$$\ln MRT = 5.746 + 0.063 \text{POP65} + 0.002 \text{TOBACCO}$$

$$\text{Let's consider: } \ln Y_i = \beta_1 + \beta_2 X_i + u_i$$

$$i=1 \rightarrow \ln Y_1 = \beta_1 + \beta_2 X_1$$

$$i=2 \rightarrow \ln Y_2 = \beta_1 + \beta_2 X_2 \quad \downarrow \text{subtract}$$

$$\ln Y_2 - \ln Y_1 = \beta_2 (X_2 - X_1)$$

$$\Delta \ln Y = \beta_2 \Delta X$$

$$\frac{\Delta Y}{Y_1} \approx \beta_2 \Delta X / \cdot 100$$

$$\frac{100 \Delta Y}{Y_1} \approx 100 \beta_2 \Delta X$$

$$\% \Delta Y \approx 100 \beta_2 \Delta X \rightarrow \text{when } \Delta X = 1 \text{ u.m.}$$

$$\% \Delta Y \approx 100 \beta_2$$

$$\text{SEMI-ELASTICITY}$$

Interpretation (e.g. $b_2 = 0.063$): If the percentage of the population above 65 increases by 1 percentage point the number of deaths per 100 000 inhabitants on average increases by (approximately) 6.3%, ceteris paribus.

Modul 4: lin-log

$$\hat{MRT} = b_1 + b_2 \ln \text{POP65} + b_3 \ln \text{TOBACCO}$$

$$\hat{MRT} = -1718.884 + 652.256 \ln \text{POP65} + 22.187 \ln \text{TOBACCO}$$

Let's consider: $y_i = \beta_1 + \beta_2 \ln x_i + u_i$

$$i=1 \rightarrow y_1 = \beta_1 + \beta_2 \ln x_1 \quad) \text{ subtract}$$

$$i=2 \rightarrow \underline{y_2 = \beta_1 + \beta_2 \ln x_2}$$

$$y_2 - y_1 = \beta_2 (\ln x_2 - \ln x_1)$$

$$\Delta y = \beta_2 \Delta \ln x$$

$$\Delta y \approx \beta_2 \frac{\Delta x}{x_1} / 100$$

$$100 \Delta y \approx \beta_2 \frac{100 \Delta x}{x_1}$$

$$\Delta y \approx \frac{\beta_2}{100} \% \Delta x \rightarrow \text{when } \% \Delta x = 1\%$$

$$\Delta y \approx \frac{\beta_2}{100} \text{ u.m.}$$

SEMI-ELASTICITY

Interpretation (e.g. $b_2 = 652.256$): If the percentage of the population above age 65 increases by 1% the number of deaths per 100 000 inhabitants on average increases by 6.52 deaths, ceteris paribus.

Model 5: "mixed"

$$\ln \widehat{MRT} = b_1 + b_2 \text{POP65} + b_3 \ln \text{TOBACCO}$$

$$\ln \widehat{MRT} = 4,584 + 0,061 \text{ POP65} + 0,299 \ln \text{TOBACCO}$$

$$MRT = e^{\beta_1 + \beta_2 \text{POP65} + \beta_3 \ln \text{TOBACCO}}$$

Interpretation: in the same manner as in the previous models

Example 3

Let's consider a quadratic regression model:

$$Y_i = \beta_1 + \beta_2 X_i + \beta_3 X_i^2 + u_i$$

$$\frac{dy}{dx} = \beta_2 + 2\beta_3 x \quad \frac{dy}{dx} \approx \frac{\Delta y}{\Delta x} \Rightarrow$$

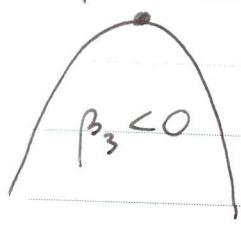
$$\Delta y \approx (\beta_2 + 2\beta_3 x) \Delta x \quad \Delta y \approx \frac{dy}{dx} \Delta x$$

When $\Delta x = 1$ m.u. $\Rightarrow \Delta y \approx \beta_2 + 2\beta_3 x$

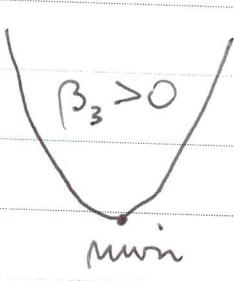
\Rightarrow With non-linear models the change in y depends on the exact value of x at which we observe the change.

In the vertex ("turning point") of the parabola the effect of x on y changes direction (sign):

max



(concave)



(convex)

$$\frac{dy}{dx} = \beta_2 + 2\beta_3 x = 0$$

$$x = -\frac{\beta_2}{2\beta_3}$$

a) $\Delta \text{SALES} \approx \frac{\partial \text{SALES}}{\partial \text{ADV}} \cdot \Delta \text{ADV}$

$$\Delta \text{SALES} \approx (0.46 - 2 \cdot 0.002 \cdot \text{ADV}) \Delta \text{ADV}$$

$$\Delta \text{ADV} = 1 \text{ u.m.} \rightarrow \Delta \text{SALES} \approx 0.46 - 0.004 \text{ ADV}$$

Interpretation: If advertising costs increase by 1 unit of measurement (= 1000 monetary units) the sales on average change by $(0.46 - 0.004 \text{ ADV})$ units of measurement (= mio monetary units), ceteris paribus.

At 4000 MU level: $\Delta \text{SALES} \approx 0.46 - 0.004 \cdot 4 = 0.444$
 $(= 4 \text{ u.m.})$

Interpretation: If advertising costs increase from 4000 to 5000 MU the sales on average increase by 444 000 MU, ceteris paribus

b) Interpretation of $b_4 = 1.122$: If R&D costs increase by 1% the sales on average increase by (approximately) 1.122%, ceteris paribus

Interpretation of $b_5 = 0.008$: If percentage of employees with higher education increases by 1 percentage point the sales on average increase by (approximately) 0.8%, ceteris paribus.

c) Not true. The two models are non-linear in variables. We can linearize them and estimate the transformed models using LSE method.

Example 2

M_B (BASE model):

$$C_t = \beta_1 + \beta_2 Y_t + \beta_3 ST_t + \beta_4 TR_t + \beta_5 FT_t + \mu_t$$

$$H_0: \beta_j = 0 \quad ; \quad \nexists j = 4, 5 \quad (\text{exclude TR and FT})$$

$$H_1: \beta_j \neq 0 \quad ; \quad \exists j = 4, 5 \quad (\text{cannot exclude TR or FT})$$

M_R (RESTRICTED model):

$$C_t = \beta_1 + \beta_2 Y_t + \beta_3 ST_t + \mu_t$$

I. F-test

$$F = \frac{(RSS_p - RSS_B) / (k_p - k_B)}{RSS_B / (n - k_B)} = \frac{(0.1594 - 0.1413) / (5 - 3)}{0.1413 / (25 - 5)} = 1.283$$

$$F_C (m_1=2, m_2=20, \alpha=0.05) = 3.49$$

$F < F_C \Rightarrow \text{cannot reject } H_0 \text{ (at } \alpha = 0.05\text{)}$

II. LM (Lagrange Multiplier) - test

$$M_R: C_t = b_1 + b_2 Y_t + b_3 S_{T_t} + \epsilon_t$$

M_{AR} (AUXILIARY regression model):

$$\ell_t = g_1 + g_2 Y_t + g_3 S_{T_t} + g_4 TR_t + g_5 FT_t \quad R^2 \downarrow 0.1137$$

$$LM = n \cdot R^2_{AR} = 25 \cdot 0.1137 = 2.8425 \quad (\sim \chi^2_{(\# \text{ restrictions})})$$

$$\chi^2_C (m=2, \alpha=0.05) = 5.997$$

$$LM < \chi^2_C \Rightarrow \text{same conclusion}$$

III. LR (Likelihood Ratio) - test

(Stata printout for M_B estimation)

$$\begin{aligned} \ln L_B &= -\frac{n}{2} \left(\ln(2\pi) + \ln \left(\frac{RSS_B}{n} \right) + 1 \right) = \\ &= -\frac{25}{2} \left(\ln(2\pi) + \ln \left(\frac{0.1413}{25} \right) + 1 \right) = 29.224 \end{aligned}$$

(Stata printout for M_R estimation)

$$\begin{aligned} \ln L_R &= -\frac{n}{2} \left(\ln(2\pi) + \ln \left(\frac{RSS_R}{n} \right) + 1 \right) = \\ &= -\frac{25}{2} \left(\ln(2\pi) + \ln \left(\frac{0.1594}{25} \right) + 1 \right) = 27.715 \end{aligned}$$

$$LR = -2 (\ln L_R - \ln L_B) = -2 (27.715 - 29.224) = 3.0181 \quad (\sim \chi^2)$$

$$\chi^2_C (m=2, \alpha=0.05) = 5.997$$

$$LR < \chi^2_C \Rightarrow \text{same conclusion}$$