Chapter 5 (Ch. 6 of the textbook)

Wooldridge: Introductory Econometrics: A Modern Approach, 5e

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Using quadratic functional forms

Example: Wage equation

Concave experience profile

 $\widehat{wage} = 3.73 + .298 exper - .0061 exper^2$ (.35) (.041) (.0009)

$$n = 526, R^2 = .093$$

Marginal effect of experience

The first year of experience increases the wage by some .30, the second year by .298-2(.0061)(1) = .29, etc.

 $\frac{\partial wage}{\partial exper} = .298 - 2(.0061)exper$

Wage maximum with respect to work experience



$$x^* = \left|\frac{\hat{\beta}_1}{2\hat{\beta}_2}\right| = \left|\frac{.298}{2(.0061)}\right| \approx 24.4$$

Does this mean the return to experience becomes negative after 24.4 years?

Not necessarily. It depends on how many observations in the sample lie right of the turnaround point.

In the given example, these are about 28% of the observations. There may be a specification problem (e.g. omitted variables).

ployment centers, student/teacher ratio

Example: Effects of pollution on housing prices



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Calculation of the turnaround point



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Other possibilities

$$\log(price) = \beta_0 + \beta_1 \log(nox) + \beta_2 \log(nox)^2$$

 $+\beta_{3} crime + \beta_{4} rooms + \beta_{5} rooms^{2} + \beta_{6} stratio + u$

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$$\Rightarrow \quad \frac{\partial \log(price)}{\partial \log(nox)} = \frac{\% \partial price}{\% \partial nox} = \beta_1 + 2\beta_2 [\log(nox)]$$

Higher polynomials

$$cost = \beta_0 + \beta_1 quantity + \beta_2 quantity^2 + \beta_3 quantity^3 + u$$

Models with interaction terms



Interaction effects complicate interpretation of parameters

 $\beta_2 =$ Effect of number of bedrooms, but for a square footage of zero

Reparametrization of interaction effects

Population means; may be replaced by sample means

$$y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_1 x_2 + u$$

$$y = \alpha_0 + \delta_1 x_1 + \delta_2 x_2 + \beta_3 (x_1 - \mu_1) (x_2 - \mu_2) + u$$

Effect of x_2 if all variables take on their mean values

Advantages of reparametrization

- Easy interpretation of all parameters
- If necessary, interaction may be centered at other interesting values

Example

Dependent Variable: PRICE Method: Least Squares Included observations: 88

Variable	Coefficient	Std. Error	t-Statistic	Prob.
С	165.4265	87.73015	1.885629	0.0628
LOTSIZE	0.001993	0.000628	3.173764	0.0021
SQRFT	0.033793	0.041462	0.815034	0.4174
BDRMS	-33.71534	22.82291	-1.477258	0.1434
BDRMS*SQRFT	0.021827	0.009663	2.258787	0.0265

Dependent Variable: PRICE Method: Least Squares Included observations: 88

Variable	Coefficient S	Std. Error	t-Statistic	Prob.
С	33.549	37.791	0.8878	0.377
LOTSIZE	0.002	0.0006	3.1738	0.002
SQRFT	0.099	0.0166	5.9825	0.000
BDRMS	10.244	8.9418	1.1456	0.255
(BDRMS-3)*(SQRFT-2014)	0.022	0.0097	2.2589	0.027

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Tests for functional form misspecification

- One can always test whether explanatory should appear as squares or higher order terms by testing whether such terms can be excluded
- Otherwise, one can use general specification tests such as RESET

Regression specification error test (RESET)

- If the model is correctly specified then it includes all the relevant variables. If we add variables that are statistically significant then the model is misspecified.
- The idea of RESET is to include squares and possibly higher order fitted values in the regression.

Test procedure

1. OLS of the initial model:

$$y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \dots + \beta_3 x_k + u \xrightarrow{\mathsf{OLS}} \hat{y}$$

2. OLS of the auxiliary regression test:

i. Option 1:
$$y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + ... + \beta_3 x_k + \gamma \hat{y}^2 + v_1$$

Test for the exclusion of this term, that is test $\gamma = 0$ against $\gamma \neq 0$

ii. Option 2:
$$y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + ... + \beta_3 x_k + \gamma_1 \hat{y}^2 + \gamma_2 \hat{y}^3 + v_2$$

Test for the exclusion of these terms, that is test jointly $\gamma_1 = 0, \gamma_2 = 0$

2. Conclusion

If the terms cannot be excluded, (H_0 is rejected) this is evidence for omitted higher order terms and interactions, i.e., for **misspecification of functional form**.

Multiple Regression Analysis: Specification and Data Issues

Example: Housing price equation

Dependent Variable: PRICE Method: Least Squares Included observations: 88

C 221.5857 77.85786 2.846029 0.0056 Evidence for LOTSIZE -0.000947 0.001226 -0.772512 0.4420 misspecificatio SQRFT -0.051335 0.064752 -0.792795 0.4301 misspecificatio FIT^2 0.001975 0.000683 2.893919 0.0048 0.0048		Variable	Coefficient	Std. Error	t-Statistic	Prob.	
	_	C LOTSIZE SQRFT FIT^2	221.5857 -0.000947 -0.051335 0.001975	77.85786 0.001226 0.064752 0.000683	2.846029 -0.772512 -0.792795 2.893919	0.0056 0.4420 0.4301 0.0048	Evidence for misspecification

With FIT the <u>estimated PRICE</u> in the regression on LOTSIZE and SQRFT