

15.1: OLS Error Assumptions Revisited

As described in earlier chapters, there is a set of key assumptions that must be met to justify the use of the t and F distributions in the interpretation of OLS model results. In particular, these assumptions are necessary for hypotheses tests and the generation of confidence intervals. When met, the assumptions make OLS more efficient than any other unbiased estimator.

OLS Assumptions

Systematic Component

- *Linearity*
- *Fixed XX*

Stochastic Component

- *Errors have constant variance across the range of XX*

$$E(\epsilon^2) = \sigma^2 \quad E(\epsilon^2) = \sigma^2$$

- *Errors are independent of XX and other ϵ_i*

$$E(\epsilon_i) = E(\epsilon_i | x_i) = 0 \quad E(\epsilon_i) = E(\epsilon_i | x_i) = 0$$

and

$$E(\epsilon_i) \neq E(\epsilon_j) \quad E(\epsilon_i) \neq E(\epsilon_j) \text{ for } i \neq j$$

- *Errors are normally distributed*

$$\epsilon_i \sim N(0, \sigma^2) \quad \epsilon_i \sim N(0, \sigma^2)$$

There is an additional set of assumptions needed for “correct” model specification. An ideal model OLS would have the following characteristics: - YY is a linear function of modeled XX variables - No XX ’s are omitted that affect $E(Y)$ and that are correlated with included XX ’s. Note that exclusion of other XX ’s that are related to YY , but are not related to the XX ’s in the model, does not critically undermine the model estimates. However, it does reduce the overall ability to explain YY . All XX ’s in the model affect $E(Y)$.

Note that if we omit an XX that is related to YY and other XX ’s in the model, we will bias the estimate of the included XX ’s. Also consider the problem of including XX ’s that are related to other XX ’s in the model, but not related to YY . This scenario would reduce the independent variance in XX used to predict YY .

Table 15.1 summarizes the various classes of assumption failures and their implications.

| Problem | Biased B | Biased SE | Invalid t/F | Hi Var |
|-----------------------|------------|-------------|---------------|--------|
| Non-linear | Yes | Yes | Yes | — |
| Heteroscedasticity | No | Yes | Yes | Yes |
| Autocorrelation | No | Yes | Yes | Yes |
| Non-normal error | No | No | Yes | Yes |
| Multicollinearity | No | No | No | Yes |
| Omit relevant X | Yes | Yes | Yes | — |
| Irrelevant X | No | No | No | Yes |
| X measurement Error | Yes | Yes | Yes | — |

Figure 15.1.1: Summary of OLS Assumption Failures and their Implications

When considering the assumptions, our data permit empirical tests for some assumptions, but not all. Specifically, we can check for linearity, normality of the residuals, homoscedasticity, data “outliers” and multicollinearity. However, we can’t check for correlation between error and XX ’s, whether the mean error equals zero, and whether all the relevant XX ’s are included.

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