

Heteroscedasticity



violation of assumption of homoscedasticity

$$Y_i = \alpha + \beta x_i + u_i$$

↑

$$E(u_i^2) \neq \sigma^2 \text{ (const)}$$

ie, variance of random error (u_i) is non constant

↓
heteroscedasticity.

but $E(u_i^2) = \sigma_i^2$ for all $i = 1(1)n$

Here $\hat{\alpha}$ and $\hat{\beta}$ (In presence of heteroscedasticity) it does not satisfy BLUE \Rightarrow Best Linear Unbiased

Estimator
In presence of heteroscedasticity OLS estimator of $\hat{\alpha}$ and $\hat{\beta}$ is unbiased but do not satisfy the condition for min variance.

Hence OLS cannot be the best estimate process it does not follow BLUE.

Therefore we can use the Method of Generalised Least Square (GLS) method.

Unfortunately the usual OLS method does not follow this strategy and therefore does not make use of the information contained in the unequal variability of dependent variable y

In this case we will assign equal weights or importances to each observation and the method is known as GLS. It is capable of producing estimators that are BLUE.

Model :
$$y_i = \beta_1 + \beta_2 x_i + u_i$$

Let $x_{0i} = 1$, then
$$y_i = \beta_1 x_{0i} + \beta_2 x_i + u_i$$
 ①

Now assume that the heteroscedastic variance σ_i^2 is known.

On dividing σ_i^2 on both sides of eq ①, we have,

$$\frac{y_i}{\sigma_i} = \beta_1 \left(\frac{x_{0i}}{\sigma_i} \right) + \beta_2 \left(\frac{x_i}{\sigma_i} \right) + \left(\frac{u_i}{\sigma_i} \right)$$

i.e.,
$$y_i^* = \beta_1^* x_{0i}^* + \beta_2^* x_i^* + u_i^*$$

What is the purpose of transforming the original model?

$$\begin{aligned}\text{Var}(u_i^*) &= E(u_i^*)^2 = E\left(\frac{u_i}{\sigma_i}\right)^2 \\ &= \frac{1}{\sigma_i^2} E(u_i^2) \\ &= \frac{1}{\sigma_i^2} \times \sigma_i^2 = 1 \\ &= \text{const}\end{aligned}$$

That is variance of the transformed disturbance term is v_i^2 which is homoscedastic.

So to conclude the estimated β_1^* and β_2^* are now BLUE and not the OLS estimators $\hat{\beta}_1$ and $\hat{\beta}_2$.

White's General Heteroscedasticity Test:

$$Y = \beta_1 + \beta_2 x_2 + \beta_3 x_3 + u_i$$

- ① we estimate and obtain residuals, \hat{u}_i .
- ② we run the following auxiliary regression

② we run the following auxiliary regression

$$\hat{u}_i = \alpha_1 + \alpha_2 X_{2i} + \alpha_3 X_{3i} + \alpha_4 X_{2i}^2 + \alpha_5 X_{3i}^2 + \alpha_6 X_{2i} X_{3i} + v_i$$

we will find out / obtain R^2 from this auxiliary regression.

③ H_0 : No heteroscedasticity, under H_0

it can be shown that sample size n time R^2 obtained from auxiliary regression follows χ^2 distribution with df equal to ^{no. of} $n \cdot R^2 \sim \chi^2$ regressors (excluding the const term)

in the auxiliary regression.

ie $n \cdot R^2 \sim \chi^2_{df}$ (in our example there are 5 regressors in auxiliary regression)

Step 4: If the chi-square value in eq ① is ^(n.R²) greater than critical chi-square _{↪ or step ③}.

(a) value at the chosen level of significance the conclusion is that there is heteroscedasticity ie H_0 is rejected.

(b) And if it does not exceed critical chi-square value \rightarrow NO heteroskedasticity \rightarrow H_0 is accepted.
 ie H_0 is rejected.