

*Department of Economics, Patna University, Patna*

*Name of the Programme: M A Economics (Sem. IV)*

*Name of the Course: EC- 1 Group C: Basic Econometrics*

*Module 4: Problems of Single Equation Model*

*Name of the Topic: Heteroscedasticity*

**By**

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**Heteroscedasticity**

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As we know that the Regression Model is based on some assumptions about the disturbance term. These are:

1. **Normality:** The first assumption of the Regression Model is that the disturbance term is normally distributed.
2. **Zero Mean:** The second assumption of the Regression Model is that the disturbance term/ error term has zero mean. i.e.

$$E(u_i^2) = 0 \quad \text{where, } i = 1, 2, 3, \dots, n.$$

3. **Homoscedasticity:** It assumes that the errors are distributed independently with Zero Mean and constant variance  $\sigma_u^2$ , that is

$$E(\sigma_i^2) = \sigma_u^2 \quad \text{where, } i = 1, 2, 3, \dots, n.$$

4. **Non- autoregression:** Assumption of non-autoregression states that various disturbance terms are uncorrelated, i.e.

$$E(u_i, u_j) = 0 \quad \text{where, } i \neq j \text{ and } i \& j = 1, 2, 3, \dots, n.$$

5. **Non-stochastic X:** It assumes that  $X_i$  is a non-stochastic variable with fixed values in repeated samples and such that for any size

$$\frac{1}{n} \sum_{i=1}^n (X_i - \bar{x})^2$$

is a finite value different from zero.

PTO



$$\begin{aligned}
\text{Since, } \text{Var.}(\hat{\beta}) &= E[\beta - E(\hat{\beta})]^2 \\
&= E\left[\frac{\sum x_i u_i}{\sum x_i^2}\right]^2 \quad (\text{from eqn (a)}) \\
&= \frac{1}{(\sum x_i^2)^2} E\left[x_1^2 u_1^2 + x_2^2 u_2^2 + \dots + 2x_1 x_2 u_1 u_2 + \right. \\
&\quad \left. 2x_1 x_3 u_1 u_3 + \dots \right] \\
&= \frac{1}{(\sum x_i^2)^2} \left[ x_1^2 E(u_1^2) + x_2^2 E(u_2^2) + \dots \right. \\
&\quad \left. + 2x_1 x_2 E(u_1 u_2) + 2x_1 x_3 E(u_1 u_3) + \dots \right]
\end{aligned}$$

Let us suppose,

$$E(u_i^2) = k_i \sigma_u^2$$

$$\begin{aligned}
\text{Then } \text{var.}(\hat{\beta}) &= \frac{1}{(\sum x_i^2)^2} [x_1^2 k_1 \sigma_u^2 + x_2^2 k_2 \sigma_u^2 + \dots] \\
&\quad [ \text{Since } E(u_i u_j) = 0, i \neq j ] \\
&= \frac{\sigma_u^2}{(\sum x_i^2)^2} \sum k_i x_i^2 = \frac{\sigma_u^2}{\sum x_i^2} \cdot \frac{\sum k_i x_i^2}{\sum x_i^2} \quad \text{--- (b)}
\end{aligned}$$

Thus, the variance is different. If  $x_i^2$  and  $k_i$  are positively correlated then  $\text{var}(\hat{\beta})$  under heteroscedasticity will be larger than variance under homoscedasticity and usual OLS (Ordinary Least Squares) formula will under estimate the true variance. Therefore, if disturbance term ( $u_i$ ) is heteroscedastic, the variance of the estimates are inefficient.

#### (d) Tests for Heteroscedasticity:

Various methods/tests are used to detect the presence of heteroscedasticity in the model. These are (i) Spearman's Rank-correlation test (ii) Goldfeld-Quandt Test (iii) Glesjer test and (iv) Park test.

(I) The Spearman's Rank Correlation Test: This is applicable for both small and large samples.

Practical Steps.

\* Regress  $y$  on  $x$  and obtain the residuals  $e_i$  where  $|e_i| = |y - \hat{y}|$

\* Calculate Rank correlation co-efficient between  $x$  and  $|e_i|$  from the given formula  $r_{|e_i|x} = 1 - \frac{6 \sum D^2}{n(n^2-1)}$

Where  $D$  is the rank difference.

\* Use  $t$ -statistic to detect the presence of heteroscedasticity using the following formula

$$t = \frac{r \sqrt{n-1}}{\sqrt{1-r^2}}$$

[Reference: Page-152 of 'Econometrics' by Shyamala, Kazur and Pragasam for Numerical Example]

(II) Goldfeld-Quandt Test: This method is used for large samples.

Practical Steps.

1. Order the observations according to the magnitude of  $x$  (independent variable)
2. Select a certain number (one-third approx.) of central observations ( $c$ ) which we omit from the analysis. The remaining  $n-c$  observations are divided into two sub-groups of equal size i.e.  $\frac{n-c}{2}$  where one subgroup includes small values of  $x$  and the other sub-group includes the large values of  $x$ .

(3) Now fit the separate regression to each of the sub-group, and obtain the sum of squared residuals from each of them ( $\sum e_1^2$  &  $\sum e_2^2$ )

(4) Find value of F-statistic from the formula  $F = \frac{\sum e_2^2}{\sum e_1^2}$  [where,  $\sum e_1^2$  denote

the sum of squared residuals from the sample of small values of  $X_i$  and  $\sum e_2^2$  denotes the same for large values of  $X_i$ .

(5) Finally use the F-table at  $(\frac{n-c}{2} - k)$  d.f.

(Reference: Page-154 of 'Econometrics' by Shyamala, Kaur and Pragasan. for Numerical Examples.]

(III) The Park Test: Prof. R.E. Park developed a specific functional form between the  $\sigma_{ui}^2$  and the explanatory variable to detect heteroscedasticity.

$$\sigma_{ui}^2 = f(X_i) = \sigma^2 X^\beta e^{V_i}$$

taking log on both sides

$$\log \sigma_{ui}^2 = \log \sigma^2 + \beta \log X + V_i \log e$$

$$\log \sigma_{ui}^2 = \log \sigma^2 + \beta \log X + V_i$$

Since  $\sigma_{ui}^2$  is unobservable, Prof. Park suggested  $e_i^2$  as a proxy for  $\sigma_{ui}^2$

$$\therefore \boxed{\log e_i^2 = \log \sigma_{ui}^2 + \beta \log X_i}$$

Steps:

- \* Run OLS and obtain  $e_i$  for the original data
- \* Run the log-linear regression between  $e_i^2$  and  $X_i$  and examine whether  $\beta$  is significant

(IV): Glejser Test: It is similar to Park Test but suggested many functional forms instead of one as suggested by Park.

Steps:

- \* Regress  $\gamma$  on all explanatory variables and compute the residuals  $e_i$ 's
- \* Regress the absolute values of  $e_i$ 's ( $|e_i|$ ) on explanatory variable with which  $\sigma_{u_i}^2$  is associated.

Since the actual form of this regression is not known, Glejser used the following functional forms to detect the presence of heteroscedasticity.

$$|e_i| = \beta_0 + \beta_1 X_i^2 + V_i$$

or,  $|e_i| = \beta_0 + \beta_1 X_i^{-1} + V_i$

or,  $|e_i| = \beta_0 + \beta_1 \sqrt{X_i} + V_i$

or,  $|e_i| = \beta_0 + \beta_1 X_i + V_i$  -- and so on.

From the above models, we select that regression model which gives the best fit in the light of correlation coefficient and standard errors of coefficients  $\beta_0$  and  $\beta_1$ . If  $\beta$  in the selected model turns out to be significant, it would suggest presence of heteroscedasticity.

If  $\beta_0 = 0$  while  $\beta_1 \neq 0 \rightarrow$  pure heteroscedasticity-

If both  $\beta_0 \neq 0$  &  $\beta_1 \neq 0 \rightarrow$  mixed heteroscedasticity.

Questions - Short Answer type:

- (i) Define heteroscedasticity.
- (ii) Write short note on Goldfeld-Quandt test.

## Long Answer type Questions:

- (i) Describe the consequences of heteroscedasticity on OLS estimator.
  - (ii) Explain the tests used to detect the presence of heteroscedasticity.
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## Suggested Readings

1. Introduction to Econometrics by G.M.K. Madhavi
  2. Econometrics: Theory and Applications -  
by Shyamala, Kaur and Prasanna
  3. Econometrics & Mathematical Economics by  
Singh, Parashar & Singh
  4. Econometrics by - K. Dhanasekaran.
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by.

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