

# Heteroscedasticity

## Additional note

## Difference between OLS and GLS

OLS we minimize

$$\sum \hat{u}_i^2 = \sum (Y_i - \hat{\beta}_1 - \hat{\beta}_2 X_i)^2$$

GLS we minimize

$$\sum w_i \hat{u}_i^2 = \sum w_i (Y_i - \hat{\beta}_1^* X_{0i} - \hat{\beta}_2^* X_i)^2, w_i = 1 / \sigma_i^2$$

- In GLS we minimized a weighted sum of residual squares with  $w_i = \frac{1}{\sigma_i^2}$  acting as the weights, but in OLS we minimize an unweighted or equally weighted residual sum of squares (RSS)
- In GLS the weight assigned to each observation is inversely proportional to its  $\sigma_i$ , that is, observations coming from a population with larger  $\sigma_i$  will get relatively smaller weight and those from a population with smaller  $\sigma_i$  will get proportionately larger weight in minimizing the RSS

To see the difference between OLS and GLS clearly, consider the hypothetical scattergram

- In the (unweighted) OLS each  $\hat{u}_i^2$  associated with points A, B, and C will receive the same weight in minimizing the RSS.
- Obviously, in this case the  $\hat{u}_i^2$  associated with point C will dominate the RSS.
- But in GLS the extreme observation C will get relatively smaller weight than the other two observations.
- In estimating the PRF more reliably we would like to give more weight to observations that are closely clustered around their (population) mean than to those that are widely scattered about.

# Hypothetical scattergram

