

**Suggested Answers for Final Exam.**

2002.1.18

**1a** False. The least squares estimators are still unbiased. It is only the variances that are affected. They will be high.

**1b** False. Even with heteroscedastic errors, the OLS estimators are unbiased. However, the standard errors are biased.

**1c** False. The estimators will be unbiased.

**1d** False. The assumption is that  $\rho = 1$ .

**1e** True.

**2a**  $\text{Var}(u_i^*) = \text{Var}\left(\frac{u_i}{X_i}\right) = \frac{\text{Var}(u_i)}{X_i^2} = \frac{\sigma^2 X_i^2}{X_i^2} = \sigma^2$ ,  $u_i^*$  is homoscedastic.

**2b** The equation of transformed variable has only constant term, therefore GLS estimator of  $\beta_2$  is  $\bar{Z}$ , the average of  $\frac{Y_i}{X_i}$ .

**2c**

$$\begin{aligned}\hat{\beta}_2^{GLS} &= \frac{\sum\left(\frac{Y_i}{X_i}\right)}{n} = \frac{\sum(\beta_2 + u_i^*)}{n} = \beta_2 + \frac{\sum \frac{u_i}{X_i}}{n} \\ E(\hat{\beta}_2^{GLS}) &= \beta_2 + \frac{\sum \frac{E(u_i)}{X_i}}{n} = \beta_2\end{aligned}$$

**2d**

$$\begin{aligned}\text{Var}(\hat{\beta}_2^{GLS}) &= E[(\hat{\beta}_2^{GLS} - \beta_2)^2] = E\left[\left(\frac{\sum u_i^*}{n}\right)^2\right] \\ &= \frac{1}{n^2} \sum \text{Var}(u_i^*) = \frac{1}{n^2} \sum \sigma^2 = \frac{\sigma^2}{n}\end{aligned}$$

**2e** Since there is no intercept, the OLS estimate of  $\beta_2$  is

$$\begin{aligned}\hat{\beta}_2 &= \frac{\sum X_i Y_i}{\sum X_i^2} = \frac{\sum X_i(\beta_2 X_i + u_i)}{\sum X_i^2} = \beta_2 + \frac{\sum X_i u_i}{\sum X_i^2} \\ \text{Var}(\hat{\beta}_2) &= E[(\hat{\beta}_2 - \beta_2)^2] = E\left[\left(\frac{\sum X_i u_i}{\sum X_i^2}\right)^2\right] = \frac{\sum X_i^2 E(u_i^2)}{(\sum X_i^2)^2} \\ &= \frac{\sum X_i^2 \sigma^2 X_i^2}{(\sum X_i^2)^2} = \frac{\sigma^2 \sum X_i^4}{(\sum X_i^2)^2}\end{aligned}$$

**3a**

$$\begin{aligned}Y_i &= 3.803 + 0.121X_i \\ s.e. &= (4.570)(0.009)\end{aligned}$$

**3b** No. The reason behind using the regression deflated by  $X$  is that the errors are heteroscedastic. The fact that  $R^2$  is lower is an irrelevant issue.

**3c** The equation estimated will be

$$\frac{Y_i}{\sqrt{X_i}} = \frac{\beta_1}{\sqrt{X_i}} + \beta_2\sqrt{X_i} + \frac{u_i}{\sqrt{X_i}}$$

Estimate this equation with no constant term.

**3d** Higher  $R^2$ .

**4a** Since  $0.9108 < d_L$ , there is positive autocorrelation.

**4b**  $\hat{\rho} \doteq 1 - \frac{d}{2} = 1 - \frac{0.9108}{2} = 1 - 0.4554 = 0.5446$ .

**4c**  $Y_1^* = \sqrt{1 - \hat{\rho}^2}Y_1$  and  $X_1^* = \sqrt{1 - \hat{\rho}^2}X_1$ .

**4d**  $Y_t^* = Y_t - \hat{\rho}Y_{t-1}$  and  $X_t^* = X_t - \hat{\rho}X_{t-1}$ ,  $t = 2, \dots, 24$ .

**4e** Since  $d_U = 1.45 < d = 1.83 > 4 - d_L = 2.73$ , there is no autocorrelation anymore.