

**The International College of Economics and Finance**  
**Econometrics - 2020. Mid-term exam, October 22**

**Part 1. (30 minutes). In each of 12 multiple choice tests indicate the correct answer.**

1. In the Simple Linear Regression Model  $Y_i = \beta_1 + \beta_2 X_i + u_i$  the population covariance of the OLS estimators of the intercept and slope coefficient  $cov(b_1, b_2)$  is

- 1) positive if the mean value of  $X$  is positive;
- ☒ 2) negative if the mean value of  $X$  is positive;
- 3) may be positive or negative if the mean value of  $X$  is positive;
- 4) may be positive or negative if the mean value of  $X$  is negative;
- 5) does not depend on the mean value of  $X$ .

$$\sigma_{\hat{\beta}}^2 = \sigma_u^2 (X'X)^{-1}$$

$$\Rightarrow Cov(\hat{\beta}_0, \hat{\beta}_1) = \frac{-\bar{X} \sigma_u^2}{\sum (X_i - \bar{X})^2}$$

2. For the Model  $Y_i = \beta_2 X_i + u$  (Model A assumptions satisfied,  $i=1, \dots, n$ ) the estimator

$$b_2 = \frac{((Y_1 + Y_2 + Y_3)/3) - \bar{Y}}{((X_1 + X_2 + X_3)/3) - \bar{X}}$$
 is:

- ☒ 1) non-linear estimator of  $\beta_2$ ;
- ☒ 2) unbiased estimator of  $\beta_2$ ;
- ☒ 3) efficient estimator of  $\beta_2$ ;
- ☒ 4) biased estimator of  $\beta_2$ ;
- ☒ 5) not an estimator of  $\beta_2$ .

3. In a simple regression with an intercept  $\hat{Y} = b_1 + b_2 X$ , the estimated slope coefficient  $b_2$  is equal to zero. Then the determination coefficient  $R^2$  is

- 1) Equal to one;
- 2) Not equal to 1 or 0;
- 3) In some situations can be negative;
- ☒ 4) Equal to zero
- 5) Can not be calculated for the model due to violation of assumptions.

4. If a new observation and a new explanatory variable are added in the Linear Regression Model, then for OLS-estimation, the following is true:

- 1) The Residual Sum of Squares (RSS) decreases; the Determination Coefficient  $R^2$  may increase, decrease or stay the same;
- 2) The Residual Sum of Squares (RSS) increases; the Determination Coefficient  $R^2$  may increase, decrease or stay the same;
- ☒ 3) Both the Determination Coefficient  $R^2$  and the Residual Sum of Squares (RSS) may increase, decrease or stay the same;
- 4) The Determination Coefficient  $R^2$  increases; the Residual Sum of Squares (RSS) may increase, decrease or stay the same;
- 5) The Determination Coefficient  $R^2$  decreases; the Residual Sum of Squares (RSS) may increase, decrease or stay the same;

HW:  $TSS$ ,  $ESS$ ,  $RSS$ ,  $R^2 \Rightarrow ESS, R^2$

is a new obs. is added

$$\bar{y}' = \frac{n}{n+1} \bar{y} + \frac{1}{n+1} y_{n+1}$$

$$TSS' = \sum (y_i - \bar{y}')^2 =$$

$$= \sum_{i=1}^{n+1} ((y_i - \bar{y}) + (\bar{y} - \bar{y}'))^2 =$$

$$= TSS + \underbrace{2 \sum (y_i - \bar{y}) \cdot (\bar{y} - \bar{y}')}_{=0} +$$

$$+ n(\bar{y} - \bar{y}')^2 + (y_{n+1} - \bar{y}')^2$$

$$= TSS + \frac{n}{n+1} (y_{n+1} - \bar{y})^2$$

$$\Rightarrow TSS' \geq TSS$$

$$RSS'_{n+1} = \sum_{i=1}^{n+1} (y_i - \hat{y}_i)^2 =$$

$$= \sum_{i=1}^n (y_i - \hat{y}_i)^2 + (y_{n+1} - \hat{y}_{n+1})^2 \geq$$

$RSS'_n$

$$\Rightarrow (RSS'_n) \geq RSS_n$$

by prop. of OLS

$\hat{\beta}$  min RSS

5. Using a sample of 570 observations, the following earnings function has been estimated:  
 $\log(EARN) = -0.41 + 0.22 ASVABC + 0.34 \log(S)$  ( $EARN$  – hourly earnings in Rubles,  $ASVABC$  – abilities indicator,  $S$  – length of studies). For this earnings function, it can be said that if  $S$  increases for one per cent, then hourly earnings increase (on average, others equal) approximately for

0.34 Rubles; 2) 0.34%; 3) 34%; 4) 0.0034%; 5) 34 Rubles.

6. For the Model  $Y_i = \beta_1 + \beta_2 X_i + u$  (Model A assumptions satisfied) the estimator

$$b_2 = \frac{\sum (Y_i - \bar{Y})}{\sum (X_i - \bar{X})}$$

is, generally speaking:

$\Leftarrow \sum (X_i - \bar{X}) = 0$

- 1) biased and inconsistent estimator of  $\beta_2$ ;
- 2) biased but consistent estimator of  $\beta_2$ ;
- 3) unbiased but inefficient estimator of  $\beta_2$ ;
- 4) equal to the OLS estimator;
- 5) can not be calculated;

7. Introduction of two linear restrictions on parameters in a regression model, estimated using OLS

- 1) results in minor increase of the sum of squared errors if at least one of the restrictions is not valid;
- 2) results in significant increase of the sum of squared errors if at least one of the restrictions is valid;
- 3) results in minor increase of the sum of squared errors if both restrictions are valid;
- 4) results in significant increase of the sum of squared errors only if both restrictions are valid;
- 5) all the above is incorrect.

8. Let  $Y^* = \lambda_1 + \lambda_2 Y$ . In a simple regression model,  $b^*_1$  is the OLS estimator of the intercept in the regression of  $Y^*$  on  $X$ , and  $b_1$  is the OLS estimator of the intercept in the regression of  $Y$  on  $X$ . Then the following is correct:

- 1)  $b^*_1 = \lambda_1 + \lambda_2 b_1$ ; 2)  $b^*_1 = b_1$ ; 3)  $b^*_1 = b_1 \lambda_2$ ; 4)  $b^*_1 = \lambda_1 + b_1$ ; 5) none of the above.

9. There are (1) and (2) versions of the Multiple Regression Model:

$$Y = \beta_1 + \beta_2 X_2 + \beta_3 X_3 + \beta_4 X_4 + u \quad (1)$$

$$Y - X_4 = \beta_1 + \beta_2 (X_2 + X_3) + u \quad (2)$$

The Model (2) is the Model (1) with the following restrictions:

~~1)  $\beta_3=0; \beta_4=0$~~ ; 2)  $\beta_2=\beta_3; \beta_4=1$ ; ~~3)  $\beta_2+\beta_3=1; \beta_4=1$~~ ; ~~4)  $\beta_2=\beta_3; \beta_4=0$~~

5) The Model (2) is not a restricted version of the Model (1).

10. A student regressed  $Y$  and  $\log(Y)$  on  $X$ , with the intercept (regressions 1 and 2 respectively). Then he did Zarembka scaling  $Y^* = Y / \text{geometric mean of } Y$ , and regressed  $Y^*$  and  $\log(Y^*)$  on  $X$  (regressions 3 and 4). Then the following is correct:

- 1) All the coefficients' estimates, including the intercept, are the same in regression 3 as in regression 1;
- 2) All the coefficients' estimates, including the intercept, are the same in regression 4 as in regression 2;
- 3) All the coefficients' estimates, except the intercept, are the same in regression 3 as in regression 1;
- 4) All the coefficients' estimates, except the intercept, are the same in regression 4 as in regression 2;
- 5) None of the above.

11. The population variance of prediction error  $\sigma_{PE}^2 = \textcircled{6.2} \cdot \left( 1 + \frac{1}{n} + \frac{(x_{n+1} - \bar{x})^2}{\sum (x_i - \bar{x})^2} \right)$
- 1) is always greater than the population variance of disturbance term  $\sigma_u^2$ ;
  - 2) is always less than the population variance of disturbance term  $\sigma_u^2$ ;
  - 3) can be greater, less then or equal to the population variance of disturbance term  $\sigma_u^2$ ;
  - 4) is always equal to the population variance of disturbance term  $\sigma_u^2$ ;
  - 5) is not related to the population variance of disturbance term  $\sigma_u^2$ .

12. Root mean squared error of prediction  $\sqrt{\sum_{t=T+1}^{T+h} (\hat{y}_t - y_t)^2 / h}$  is always

- 1) Greater or equal to the Mean absolute error of prediction  $\sum_{t=T+1}^{T+h} |\hat{y}_t - y_t| / h$
- 2) Less or equal to the Mean absolute error of prediction;
- 3) May be greater or less than the Mean absolute error of prediction;
- 4) Equal to the Mean absolute error of prediction;
- 5) Can not be compared with the Mean absolute error of prediction since it has different dimensity.

$$RMSE \geq MAE$$

$$RMSE = MAE \Rightarrow \text{all errors are of the same magnitude}$$