# Final Exam 

(June 15, 2022, 2 hours)

Econometrics (Spring 2022)
Professor: Wonmun Shin

## Part I

## True or False

(Total 40 points) Read each statement below carefully. Write a $\mathbf{T}$ if you think the statement is True. Write an $\mathbf{F}$ if you think the statement is false. Do not forget to state a reason precisely if you answer F. (Grading guide: If your answer (with a reason in the case of F ) is correct, you will get 5 points. If your answer is partially correct (i.e. F without a reason, F with an incorrect reason, etc.), you will get partial points.)

1. If you add one more additional independent variable to your linear regression model, the $R^{2}$ can increase.
2. One can introduce as many dummy variables as the number of categories of a qualitative variable if she/he omits the intercept in the regression model.
3. Under near multicollinearity, OLS estimator cannot be BLUE even under the classical assumptions, and therefore one should use GLS methodology to obtain the most efficient estimator.
4. In the multiple regression model, the variance of OLS estimator gets larger as the correlation between the regressors is higher.
5. At the same significance level, the probability of type I error tends to be smaller when doing a two-sided test than when doing a one-sided test.
6. When the regressors are not deterministic but random, the OLS estimator does not converge to the true parameter any more, even in a very large sample.
7. In the simple regression model, suppose we multiply each $Y_{i}$ value by a constant, say 2 . Then, it will not change the residuals and fitted values of $Y$.
8. Even if heteroskedasticity is present, we can still use OLS estimator because the OLS estimator is still consistent estimator.

## Part II

## Short Questions

1. (10 points) Suppose the true relationship of $Y$ and $X, Z$ is given by

$$
Y_{i}=\beta_{1}+\beta_{2} X_{i}+\beta_{3} Z_{i}+e_{i}
$$

under classical assumptions. Now, Student $A$ estimates $\beta_{2}$ by regressing $Y$ on $X$ only (that is, omitting a relevant variable $Z$ ) as:

$$
Y_{i}=\hat{\beta}_{1}+\hat{\beta}_{2} X_{i}+\hat{e}_{i}
$$

Then, calculate $\operatorname{Bias}\left(\hat{\beta}_{2}\right)$.
2. ( $\mathbf{1 0}$ points) In the simple linear regression $Y_{i}=\beta_{1}+\beta_{2} X_{i}+e_{i}$, we can obtain the OLS estimators $\hat{\beta}_{1}$ and $\hat{\beta}_{2}$. Student $B$ considers the newly-defined dependent variable and independent variable by adding same constant to $Y_{i}$ and $X_{i}$. Specifically, she defines $Y_{i}^{*}=Y_{i}+c$ and $X_{i}^{*}=X_{i}+c$ (where $c$ is non-zero constant). As a result, she runs the following regression:

$$
Y_{i}^{*}=\alpha_{1}+\alpha_{2} X_{i}^{*}+u_{i}
$$

Express the OLS estimators $\hat{\alpha}_{1}$ and $\hat{\alpha}_{2}$ in terms of $\hat{\beta}_{1}, \hat{\beta}_{2}$ and $c$.
3. ( $\mathbf{1 0}$ points) In the following estimated model, $Y_{i}$ is the expenditure on beers, $X_{i}$ is the expenditures on fried chickens, and $M_{i}$ is a dummy variable indicating 1 for males, 0 for females.

$$
\hat{Y}_{i}=-0.6+1.5 M_{i}+0.5 X_{i}+0.2 M_{i} X_{i}
$$

Here, Student $C$ uses a dummy variable $F_{i}$ instead of $M_{i}$ for regression, where $F_{i}$ has the value 1 for females and 0 for males. That is, he estimates the following equation:

$$
Y_{i}=\beta_{1}+\beta_{2} F_{i}+\beta_{3} X_{i}+\beta_{4} F_{i} X_{i}+e_{i}
$$

Obtain the estimated values of $\hat{\beta}_{1}, \hat{\beta}_{2}, \hat{\beta}_{3}$ and $\hat{\beta}_{4}$.
4. (10 points) From a sample of 100 observations, Student $D$ obtained the following results:

$$
\sum X_{i}=400, \quad \sum Y_{i}=700, \quad \sum X_{i} Y_{i}=4,000, \quad \sum X_{i}^{2}=6,000
$$

Suppose that Student $D$ wants to obtain the OLS estimate for a two-variable regression through the origin, $Y_{i}=\beta_{2} X_{i}+e_{i}$. Using the above results, calculate $\hat{\beta}_{2}$.

## Part III

## Long Questions

1. (Total 20 points) Suppose you have the following estimated models of relationships between consumption $(C)$, income (INC), and time $(t)$ :

$$
\begin{array}{ll}
\text { Model A: } \widehat{C}=\underset{(2.73)}{0.2}+\underset{(3.44)}{0.5} I N C & R^{2}=0.5 \\
\text { Model B: } \widehat{\ln C}=\underset{(2.66)}{0.01}+\underset{(4.50)}{0.3} \ln I N C & R^{2}=0.6 \\
\text { Model C: } \widehat{\ln C}=\underset{(3.02)}{0.01}+\underset{(1.58)}{0.1} t & R^{2}=0.6
\end{array}
$$

Note that the values in parentheses are $t$-ratios.
(a) (5 points) Is the elasticity of consumption with respect to income in Model $\mathbf{A}$ constant? If yes, what is it?
(b) (5 points) Is the elasticity of consumption with respect to income in Model $\mathbf{B}$ constant? If yes, what is it?
(c) (5 points) How would you interpret 0.1 (coefficient of time variable) in Model C?
(d) (5 points) Is the slope coefficient in Model A significant at $1 \%$ significance level? (Note: If you need, please refer to the probability table on the last page.)
2. (Total 35 points) A Student $E$ has data for the year 2020 from US census data on the following characteristics of the respondents: weekly earnings, $E A R N I N G S$, measured in thousand dollars; years of schooling, $S C H$; years of work experience, EXP; sex (males and females); and ethnicity (whites and non-whites). Student $E$ defines dummy variables $M A L E$ and $N W H$. MALE is defined to be 1 for males and 0 for females. NWH is defined to be 1 for non-whites and 0 for whites. She also defines interactive term, $M N W=M A L E * N W H$. She fits the following ordinary least squares regressions, each with $E A R N I N G S$ as the dependent variable:

| Variable | Model (1) | Model (2) | Model (3) |
| :---: | :---: | :---: | :---: |
| Constant | 0.390 | 0.459 | 0.411 |
|  | $(0.075)$ | $(0.076)$ | $(0.082)$ |
| $S C H$ | 0.124 | 0.121 | 0.122 |
|  | $(0.004)$ | $(0.004)$ | $(0.004)$ |
| $E X P$ | 0.033 | 0.032 | 0.033 |
|  | $(0.002)$ | $(0.002)$ | $(0.003)$ |
| MALE | 0.278 | 0.277 | 0.306 |
|  | $(0.0003)$ | $(0.0003)$ | $(0.0003)$ |
| $N W H$ |  | -0.144 | 0.205 |
|  |  | $(0.022)$ | $(0.099)$ |
| $M N W$ |  |  | -0.280 |
|  |  |  | $(0.065)$ |
| $R^{2}$ | 0.300 | $?$ | 0.376 |

The numbers in parenthesis are standard deviations of the least squares estimates. Note that the total sum of squares $(T S S)$ is 350 and the number of observations is 214 .
(a) (5 points) What is $R S S$ (residual sum of squares) in Model (1)?
(b) (5 points) Test the overall significance of Model (1) at $1 \%$ significance level. (Note: If you need, please refer to the probability table on the last page.)
(c) (5 points) Give an interpretation of the coefficient of NWH in Model (2).
(d) (5 points) Specify the interval for possible values of $R^{2}$ in Model (2). In other words, find the lower bound and the upper bound of $R^{2}$ in Model (2), using the above results.
(e) (5 points) Give an interpretation of the coefficient of $M N W$ in Model (3).
(f) (5 points) Perform the test of the joint significance of the coefficients of $N W H$ and $M N W$ in Model (3) at $5 \%$ significance level. (Note: If you need, please refer to the probability table on the last page.)
(g) (5 points) Explain whether a simple $t$-test on the coefficient of $N W H$ in Model (2) is sufficient to show that the wage equations are different for whites and non-whites.
3. (Total 25 points) There exists an exact linear relationship between two variables $X$ and $Y$ as:

$$
\begin{equation*}
Y_{i}=\beta_{1}+\beta_{2 i} X_{i} \tag{1}
\end{equation*}
$$

Assume that $X_{i}$ is deterministic. However, in fact, $\beta_{2 i}$ is not a fixed constant, but a random variable. To be specific, suppose that $\beta_{2 i}$ is expressed as:

$$
\begin{equation*}
\beta_{2 i}=\beta_{2}+\nu_{i} \tag{2}
\end{equation*}
$$

where $\beta_{2}$ is a parameter (i.e. unknown but fixed constant), and $\nu_{i}$ is an error term with mean 0 and constant variance $\sigma_{\nu}^{2}$. Also, $\nu_{i}$ is independent for different $i$.
(a) (5 points) Using the equation (2), express the equation (1) as the standard model:

$$
\begin{equation*}
Y_{i}=\beta_{1}+\beta_{2} X_{i}+e_{i} \tag{3}
\end{equation*}
$$

What is $e_{i}$ in the equation (3)?
(b) (5 points) What is $E\left(e_{i}\right)$ and $\operatorname{Var}\left(e_{i}\right)$ in the equation (3)?
(c) (5 points) Is OLS estimator for $\beta_{2}$ in the equation (3) unbiased and linear? Why or why not?
(d) (5 points) Is OLS estimator for $\beta_{2}$ in the equation (3) BLUE? Why or why not?
(e) (5 points) If you answered "NO" in the above question, transform the equation (3) appropriately to obtain BLUE. If you answered "YES" in the above question, write down "Not Applicable".

## - Probability table for Final Exam:

$-P\left(\chi^{2}(1) \geq 3.841\right)=0.05, P\left(\chi^{2}(1) \geq 5.023\right)=0.025, P\left(\chi^{2}(1) \geq 7.879\right)=0.01$ for $\chi^{2}$ distribution with degrees of freedom 1.
$-P\left(\chi^{2}(2) \geq 5.991\right)=0.05, P\left(\chi^{2}(2) \geq 7.377\right)=0.025, P\left(\chi^{2}(2) \geq 9.210\right)=0.01$ for $\chi^{2}$ distribution with degrees of freedom 2.
$-P\left(\chi^{2}(3) \geq 7.814\right)=0.05, P\left(\chi^{2}(3) \geq 9.348\right)=0.025, P\left(\chi^{2}(3) \geq 11.34\right)=0.01$ for $\chi^{2}$ distribution with degrees of freedom 3 .
$-P(F(1,+200) \geq 3.84)=0.05, P(F(1,+200) \geq 6.63)=0.01$ for $F$ distribution with degrees of freedom $(1,+200) .(+200$ indicates for more than 200 degrees of freedom.)
$-P(F(2,+200) \geq 3.00)=0.05, P(F(2,+200) \geq 4.61)=0.01$ for $F$ distribution with degrees of freedom $(2,+200) .(+200$ indicates for more than 200 degrees of freedom.)
$-P(F(3,+200) \geq 2.60)=0.05, P(F(3,+200) \geq 3.78)=0.01$ for $F$ distribution with degrees of freedom $(3,+200) .(+200$ indicates for more than 200 degrees of freedom.)

- $P(Z \geq 1.28)=0.1, P(Z \geq 1.645)=0.05, P(Z \geq 1.96)=0.025, P(Z \geq 2.58)=0.005$ where $Z$ is a standard normal random variable.

