

Econ 3180 – Final Exam, April 25th 2015

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You may use a calculator. Answer all questions in the answer book provided. The exam is 3 hours long and consists of 100 marks.

A formula sheet, and a table of probabilities from the standard Normal distribution, are provided at the back of the exam booklet.

DO NOT OPEN THE EXAM UNTIL YOU ARE INSTRUCTED TO DO SO.

NAME:	
STUDENT #:	

Part A – Multiple Choice – each question is worth 1 mark

- 1) A large p-value implies
- rejection of the null hypothesis.
 - a large t-statistic.
 - a large \bar{Y}^{act} .
 - that the observed value \bar{Y}^{act} is consistent with the null hypothesis.
- 2) To standardize a variable you
- subtract its mean and divide by its standard deviation.
 - integrate the area below two points under the normal distribution.
 - add and subtract 1.96 times the standard deviation to the variable.
 - divide it by its standard deviation, as long as its mean is 1.
- 3) Suppose that a normally distributed random variable, Y , has mean 5 and variance 4. That is, $Y \sim N(5,4)$. What is the probability that $Y \geq 8.92$?
- 0.01
 - 0.025
 - 0.05
 - None of the above.
- 4) The following are all least squares assumptions with the exception of:
- The conditional distribution of u_i given X_i has a mean of zero.
 - The explanatory variable in the regression model is normally distributed.
 - $(X_i, Y_i), i = 1, \dots, n$ are independently and identically distributed.
 - Large outliers are unlikely.
- 5) $E(u_i | X_i) = 0$ says that
- dividing the error by the explanatory variable results in a zero (on average).
 - the sample regression function residuals are unrelated to the explanatory variable.
 - the sample mean of the Xs is much larger than the sample mean of the errors.
 - the conditional distribution of the error given the explanatory variable has a zero mean.
- 6) Heteroskedasticity means that
- homogeneity cannot be assumed automatically for the model.
 - the variance of the error term is not constant.
 - the observed units have different preferences.
 - agents are not all rational.
- 7) The confidence interval for the sample regression function slope
- can be used to conduct a test about a hypothesized population regression function slope.
 - can be used to compare the value of the slope relative to that of the intercept.
 - adds and subtracts 1.96 from the slope.
 - allows you to make statements about the economic importance of your estimate.

- 8)** When there are omitted variables in the regression, which are determinants of the dependent variable, then
- you cannot measure the effect of the omitted variable, but the estimator of your included variable(s) is (are) unaffected.
 - this has no effect on the estimator of your included variable because the other variable is not included.
 - this will always bias the OLS estimator of the included variable.
 - the OLS estimator is biased if the omitted variable is correlated with the included variable.
- 9)** Imagine you regressed earnings of individuals on a constant, a binary variable (“Male”) which takes on the value 1 for males and is 0 otherwise, and another binary variable (“Female”) which takes on the value 1 for females and is 0 otherwise. Because females typically earn less than males, you would expect
- the coefficient for Male to have a positive sign, and for Female a negative sign.
 - both coefficients to be the same distance from the constant, one above and the other below.
 - none of the OLS estimators to exist because there is perfect multicollinearity.
 - this to yield a difference in means statistic.
- 10)** In the multiple regression model, the least squares estimator is derived by
- minimizing the sum of squared prediction mistakes.
 - setting the sum of squared errors equal to zero.
 - minimizing the absolute difference of the residuals.
 - forcing the smallest distance between the actual and fitted values.
- 11)** When testing joint hypothesis, you should
- use t-statistics for each hypothesis and reject the null hypothesis if all of the restrictions fail.
 - use the F-statistic and reject all the hypothesis if the statistic exceeds the critical value.
 - use t-statistics for each hypothesis and reject the null hypothesis once the statistic exceeds the critical value for a single hypothesis.
 - use the F-statistics and reject at least one of the hypothesis if the statistic exceeds the critical value.

- 12)** The overall regression F-statistic tests the null hypothesis that
- all slope coefficients are zero.
 - all slope coefficients and the intercept are zero.
 - the intercept in the regression and at least one, but not all, of the slope coefficients is zero.
 - the slope coefficient of the variable of interest is zero, but that the other slope coefficients are not.
- 13)** The interpretation of the slope coefficient in the model $Y_i = \beta_0 + \beta_1 \ln(X_i) + u_i$ is as follows:
- a 1% change in X is associated with a β_1 % change in Y .
 - a 1% change in X is associated with a change in Y of $0.01 \beta_1$.
 - a change in X by one unit is associated with a $100 \beta_1$ % change in Y .
 - a change in X by one unit is associated with a β_1 change in Y .
- 14)** An example of a quadratic regression model is
- $Y_i = \beta_0 + \beta_1 X + \beta_2 Y^2 + u_i$.
 - $Y_i = \beta_0 + \beta_1 \ln(X) + u_i$.
 - $Y_i = \beta_0 + \beta_1 X + \beta_2 X^2 + u_i$.
 - $Y_i^2 = \beta_0 + \beta_1 X + u_i$.
- 15)** A nonlinear function
- makes little sense, because variables in the real world are related linearly.
 - can be adequately described by a straight line between the dependent variable and one of the explanatory variables.
 - is a concept that only applies to the case of a single or two explanatory variables since you cannot draw a line in four dimensions.
 - is a function with a slope that is not constant.
- 16)** The OLS residuals
- can be calculated using the errors from the regression function.
 - can be calculated by subtracting the fitted values from the actual values.
 - are unknown since we do not know the population regression function.
 - should not be used in practice since they indicate that your regression does not run through all your observations.

Part B - Short Answer – 4 marks each

- 1) Describe some reasons why it is important to look at a scatterplot of the data.
- 2) What is the problem with assuming that the error terms (the u_i 's) are homoskedastic, when they are actually heteroskedastic?
- 3) Explain why using unadjusted-R-square (R^2) in the multiple regression model is a bad idea. Why is it better to use adjusted-R-square (\bar{R}^2) in the multiple regression model?
- 4) Suppose you want to test $H_0: \beta_2 = 0, \beta_3 = 0$. Can you use the t -test? Why or why not?
- 5) Use the following data for this question:

$$Y = \{1,6,7\} \quad X = \{2,4,6\}$$

The population model is:

$$Y_i = \beta_0 + \beta_1 X_i + u_i$$

What are the OLS estimates for β_0 and β_1 ?

- 6) Explain the difference between the random error term (u_i) and the OLS residual (\hat{u}_i).
- 7) Explain the difference between a restricted and an unrestricted model. Provide a simple example.
- 8) Consider the estimated regression model:

$$\widehat{TestScore} = 607.3 + 3.85Income - 0.0423Income^2$$

What is the predicted change in *TestScore* due to a change in *Income*?

- 9) Consider the following population model:

$$ahe = \beta_0 + \beta_1bach + \beta_2female + \beta_3bach \times female + u,$$

where *ahe* is average hourly earnings, *female* is a dummy variable which equals 1 if the individual is female, and *bach* is a dummy variable which equals 1 if the individual has a bachelor's degree (BA). The *ahe* among various groups are:

<i>ahe</i> for men without a BA	= 14.88
<i>ahe</i> for women without a BA	= 11.92
<i>ahe</i> for men with a BA	= 21.98
<i>ahe</i> for women with a BA	= 18.47

What are the OLS estimates for β_0 , β_1 , β_2 , and β_3 ?

10) In class, the effect of an additional fireplace on house price was estimated. The following two regression outputs have been reproduced from class:

```
summary(lm(Price ~ Fireplaces))
```

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	171.824	3.234	53.13	<2e-16 ***
Fireplaces	66.699	3.947	16.90	<2e-16 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 91.21 on 1726 degrees of freedom
Multiple R-squared: 0.142, Adjusted R-squared: 0.1415
F-statistic: 285.6 on 1 and 1726 DF, p-value: < 2.2e-16

```
summary(lm(Price ~ Fireplaces + Living.Area))
```

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	14.730146	5.007563	2.942	0.00331 **
Fireplaces	8.962440	3.389656	2.644	0.00827 **
Living.Area	0.109313	0.003041	35.951	< 2e-16 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 68.98 on 1725 degrees of freedom
Multiple R-squared: 0.5095, Adjusted R-squared: 0.5089
F-statistic: 895.9 on 2 and 1725 DF, p-value: < 2.2e-16

Why is the estimated effect of an additional fireplace so different between the two regressions? Explain in detail.

11) Explain why it might be important to capture non-linear effects in a regression model. Explain what a non-linear effect means. Describe different ways of capturing non-linear effects.

Part C – Long Answer – each part is worth 4 marks

This question uses the Current Population Survey (CPS) dataset from assignment #3 and from the class lectures. Several estimated models have been reported below.

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
	Dependent Variable							
	<i>AHE</i>	<i>ln(AHE)</i>	<i>ln(AHE)</i>	<i>ln(AHE)</i>	<i>ln(AHE)</i>	<i>ln(AHE)</i>	<i>ln(AHE)</i>	<i>ln(AHE)</i>
<i>Age</i>	0.439** (0.031)	0.024** (0.002)		0.147** (0.042)	0.146** (0.042)	0.191** (0.054)	0.117* (0.057)	0.160* (0.064)
<i>Age</i> ²				-0.002** (0.001)	-0.002** (0.001)	-0.003** (0.001)	-0.002 (0.001)	-0.002* (0.001)
<i>ln(Age)</i>			0.725** (0.052)					
<i>Female</i> × <i>Age</i>						-0.097 (0.084)		-0.123 (0.085)
<i>Female</i> × <i>Age</i> ²						0.002 (0.001)		0.002 (0.001)
<i>Bachelor</i> × <i>Age</i>							0.064 (0.083)	0.091 (0.084)
<i>Bachelor</i> × <i>Age</i> ²							-0.001 (0.001)	-0.001 (0.001)
<i>Female</i>	-3.158** (0.180)	-0.181** (0.011)	-0.180** (0.011)	-0.180** (0.011)	-0.210** (0.014)	1.358 (1.238)	-0.209** (0.014)	1.764 (1.251)
<i>Bachelor</i>	6.865** (0.178)	0.405** (0.010)	0.405** (0.010)	0.405** (0.010)	0.378** (0.014)	0.377** (0.014)	-0.770 (1.223)	-1.186 (1.236)
<i>Female</i> × <i>Bachelor</i>					0.064** (0.021)	0.063** (0.021)	0.067** (0.021)	0.066** (0.021)
Intercept	1.884* (0.920)	1.857** (0.054)	0.128 (0.177)	0.059 (0.611)	0.078 (0.610)	-0.633 (0.799)	0.604 (0.831)	-0.095 (0.939)
<i>R</i> ²	0.1900	0.1924	0.1927	0.1933	0.1942	0.1950	0.1957	0.1968
\bar{R}^2	0.1897	0.1921	0.1924	0.1929	0.1937	0.1943	0.1949	0.1959

Significance at the *5% and **1% significance level.

You may also need the following critical values for F-statistics in large samples:

q	5% critical value
1	3.84
2	3.00
3	2.60
4	2.37
5	2.21

- a) In all of the models, $R^2 > \bar{R}^2$. Why is this?
- b) Can R^2 be used to compare models (1) and (2)? Why or why not?
- c) What is the interpretation of the intercept in model (3)? Since the intercept appears to be statistically insignificant, should it be dropped from the model?
- d) Interpret the estimated coefficient of 0.725 (on $\ln(\text{Age})$) in model (3).
- e) Using model (4), what is the predicted $\ln(\text{AHE})$ for a 25 year old male with a bachelor's degree?
- f) Does Age appear to have a non-linear effect on $\ln(\text{AHE})$? Use a formal hypothesis test in your answer.
- g) Test the hypothesis that the effect of having a bachelor's degree on $\ln(\text{AHE})$ is the same for women as it is for men.
- h) Test the hypothesis that the effect of Age on $\ln(\text{AHE})$ is the same for women as it is for men.
- i) Using model (8), what is the estimated effect on $\ln(\text{AHE})$ of obtaining a bachelor's degree, for 25 year old women? For 25 year old men?
- j) Approximately how much more valuable is a bachelor's degree for women than for men?

Econ 3180 - Final Formula Sheet

expected value of Y (mean of Y)	μ_Y
variance of Y	$\sigma_Y^2 = E(Y - \mu_Y)^2 = E(Y^2) - (\mu_Y)^2$
standard deviation of Y	$\sigma_Y = \sqrt{\sigma_Y^2}$
covariance between X and Y	$\sigma_{XY} = E[(X - \mu_X)(Y - \mu_Y)]$
correlation coefficient (between X and Y)	$\rho_{XY} = \frac{\sigma_{XY}}{\sigma_X \sigma_Y}$
expected value of the sample average, \bar{Y}	$E(\bar{Y}) = \mu_Y$
variance of the sample average, \bar{Y}	$\sigma_{\bar{Y}}^2 = \frac{\sigma_Y^2}{n}$
t -statistic for testing μ_Y (for large n , and when σ_Y^2 is <i>known</i>)	$t = \frac{\bar{Y}^{act} - \mu_{Y,0}}{\sigma_{\bar{Y}}} \sim N(0,1)$
sample variance (estimator for σ_Y^2)	$s_Y^2 = \frac{1}{n-1} \sum_{i=1}^n (Y_i - \bar{Y})^2$
sample covariance (estimator for covariance)	$s_{xy} = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})(Y_i - \bar{Y})$
sample correlation (estimator for correlation)	$r_{xy} = \frac{s_{xy}}{s_x s_y}$
standard error of \bar{Y} (estimator for the standard deviation of \bar{Y})	$s_{\bar{Y}} = \sqrt{\frac{s_Y^2}{n}}$
t -statistic for testing μ_Y (for large n , and when σ_Y^2 is <i>unknown</i>)	$t = \frac{\bar{Y}^{act} - \mu_{Y,0}}{s_{\bar{Y}}} \sim N(0,1)$
95% confidence interval for μ_Y (for large n)	$conf. int. = \bar{Y} \pm 1.96 \times s_{\bar{Y}}$
population linear regression model with one regressor	$Y_i = \beta_0 + \beta_1 X_i + u_i, \quad i = 1, \dots, n$
OLS estimator of the slope (β_1)	$\hat{\beta}_1 = \frac{\sum_{i=1}^n (X_i - \bar{X})(Y_i - \bar{Y})}{\sum_{i=1}^n (X_i - \bar{X})^2}$
OLS estimator of the intercept (β_0)	$\hat{\beta}_0 = \bar{Y} - \hat{\beta}_1 \bar{X}$
OLS predicted values	$\hat{Y}_i = \hat{\beta}_0 + \hat{\beta}_1 X_i$
OLS residuals	$\hat{u}_i = Y_i - \hat{Y}_i$

explained sum of squares	$ESS = \sum_{i=1}^n (\hat{Y}_i - \bar{Y})^2$
total sum of squares	$TSS = \sum_{i=1}^n (Y_i - \bar{Y})^2$
sum of squared residuals	$SSR = \sum_{i=1}^n \hat{u}_i^2$
regression R^2	$R^2 = \frac{ESS}{TSS}$
standard error of regression	$\sqrt{\frac{1}{n-2} \times SSR}$
L.S.A. #1	$E(u X = x) = 0$
L.S.A. #2	$(X_i, Y_i), i = 1, \dots, n, \text{ are i.i.d.}$
L.S.A. #3	Large outliers are rare.
The sampling distribution of $\hat{\beta}_1$ (for large n)	$\hat{\beta}_1 \sim N\left(\beta_1, \frac{\text{var}[(X_i - \mu_X)u_i]}{n\sigma_X^4}\right)$
t -statistic for testing β_1	$t = \frac{\hat{\beta}_1 - \beta_{1,0}}{SE(\hat{\beta}_1)}$
95% confidence interval for β_1 (for large n)	$\text{conf. int.} = \hat{\beta}_1 \pm 1.96 \times SE(\hat{\beta}_1)$
alternative regression R^2	$R^2 = 1 - \frac{SSR}{TSS}$
adjusted R-square (\bar{R}^2)	$\bar{R}^2 = 1 - \frac{SSR}{TSS} \left(\frac{n-1}{n-k-1}\right)$
F-statistic	$F = \frac{(SSR_R - SSR_U)/q}{SSR_U/(n - k_U - 1)}$
F-statistic	$F = \frac{(R_U^2 - R_R^2)/q}{(1 - R_U^2)/(n - k_U - 1)}$

Standard Normal Probabilities

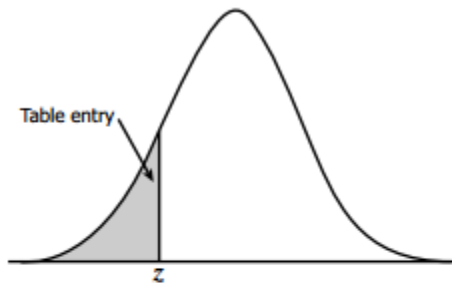


Table entry for z is the area under the standard normal curve to the left of z .

z	.00	.01	.02	.03	.04	.05	.06	.07	.08	.09
-3.4	.0003	.0003	.0003	.0003	.0003	.0003	.0003	.0003	.0003	.0002
-3.3	.0005	.0005	.0005	.0004	.0004	.0004	.0004	.0004	.0004	.0003
-3.2	.0007	.0007	.0006	.0006	.0006	.0006	.0006	.0005	.0005	.0005
-3.1	.0010	.0009	.0009	.0009	.0008	.0008	.0008	.0008	.0007	.0007
-3.0	.0013	.0013	.0013	.0012	.0012	.0011	.0011	.0011	.0010	.0010
-2.9	.0019	.0018	.0018	.0017	.0016	.0016	.0015	.0015	.0014	.0014
-2.8	.0026	.0025	.0024	.0023	.0023	.0022	.0021	.0021	.0020	.0019
-2.7	.0035	.0034	.0033	.0032	.0031	.0030	.0029	.0028	.0027	.0026
-2.6	.0047	.0045	.0044	.0043	.0041	.0040	.0039	.0038	.0037	.0036
-2.5	.0062	.0060	.0059	.0057	.0055	.0054	.0052	.0051	.0049	.0048
-2.4	.0082	.0080	.0078	.0075	.0073	.0071	.0069	.0068	.0066	.0064
-2.3	.0107	.0104	.0102	.0099	.0096	.0094	.0091	.0089	.0087	.0084
-2.2	.0139	.0136	.0132	.0129	.0125	.0122	.0119	.0116	.0113	.0110
-2.1	.0179	.0174	.0170	.0166	.0162	.0158	.0154	.0150	.0146	.0143
-2.0	.0228	.0222	.0217	.0212	.0207	.0202	.0197	.0192	.0188	.0183
-1.9	.0287	.0281	.0274	.0268	.0262	.0256	.0250	.0244	.0239	.0233
-1.8	.0359	.0351	.0344	.0336	.0329	.0322	.0314	.0307	.0301	.0294
-1.7	.0446	.0436	.0427	.0418	.0409	.0401	.0392	.0384	.0375	.0367
-1.6	.0548	.0537	.0526	.0516	.0505	.0495	.0485	.0475	.0465	.0455
-1.5	.0668	.0655	.0643	.0630	.0618	.0606	.0594	.0582	.0571	.0559
-1.4	.0808	.0793	.0778	.0764	.0749	.0735	.0721	.0708	.0694	.0681
-1.3	.0968	.0951	.0934	.0918	.0901	.0885	.0869	.0853	.0838	.0823
-1.2	.1151	.1131	.1112	.1093	.1075	.1056	.1038	.1020	.1003	.0985
-1.1	.1357	.1335	.1314	.1292	.1271	.1251	.1230	.1210	.1190	.1170
-1.0	.1587	.1562	.1539	.1515	.1492	.1469	.1446	.1423	.1401	.1379
-0.9	.1841	.1814	.1788	.1762	.1736	.1711	.1685	.1660	.1635	.1611
-0.8	.2119	.2090	.2061	.2033	.2005	.1977	.1949	.1922	.1894	.1867
-0.7	.2420	.2389	.2358	.2327	.2296	.2266	.2236	.2206	.2177	.2148
-0.6	.2743	.2709	.2676	.2643	.2611	.2578	.2546	.2514	.2483	.2451
-0.5	.3085	.3050	.3015	.2981	.2946	.2912	.2877	.2843	.2810	.2776
-0.4	.3446	.3409	.3372	.3336	.3300	.3264	.3228	.3192	.3156	.3121
-0.3	.3821	.3783	.3745	.3707	.3669	.3632	.3594	.3557	.3520	.3483
-0.2	.4207	.4168	.4129	.4090	.4052	.4013	.3974	.3936	.3897	.3859
-0.1	.4602	.4562	.4522	.4483	.4443	.4404	.4364	.4325	.4286	.4247
-0.0	.5000	.4960	.4920	.4880	.4840	.4801	.4761	.4721	.4681	.4641