

ECON 140A FALL 2020

HW 2

Notes:

1. Show your work and justify all steps. You will lose points if your reasoning is incomplete.

Problem 1. OLS in the population.

- (a) Consider $f(x_k) = b_0 + b_1 x_k$, the Ordinary Least Squares univariate affine regression function of y_k on a constant and x_k in the population. Explain in words the meaning of “Least Squares”. (2 points)
- (b) Suppose someone suggests that you should minimize the sum of the prediction error, $\sum_{k=1}^n (y_k - b_0 - b_1 x_k)$. Explain why this is a good or bad idea? What are the ramifications for our estimates of the y_k values? (2 points)
- (c) Explain in words why it is a good idea to include a constant in a univariate regression. Give a real world example in your explanation. (2 points)
- (d) Write down the OLS parameters, b_0 and b_1 , for the population regression specification, $y_k = b_0 + b_1 x_k + e_k$. (2 point)
- (e) Are the OLS parameters you wrote down above random variables? Explain. (2 points)

Problem 2. Suppose you have a sample of n observations from the population where Y_i represents the dependent variable and X_i represents the independent variable.

- (a) Write down expressions for the estimators from a regression of Y_i on a constant and X_i in the sample. That is, the regression specification, $Y_i = b_0 + b_1 X_i + e_i$. (2 points)
- (b) Assume that $\lim_{n \rightarrow \infty} \hat{\beta}_1 = \beta_1$. Now use the law of large numbers to show that $\lim_{n \rightarrow \infty} \hat{\beta}_0 = \beta_0$. Remember to show each step and provide explanations. (5 points)
- (c) As in the lecture notes, let $\hat{e}_i = (Y_i - \hat{\beta}_0 - \hat{\beta}_1 X_i)$, representing the error we make when we use the sample OLS regression to predict Y_i . Show that the sample covariance between X_i and \hat{e}_i is zero. Explain why this would be the case. Recall that the sample covariance between two variables, W_i and Z_i , is $\frac{1}{n} \sum_{i=1}^n (W_i - \bar{W})(Z_i - \bar{Z})$, where $\bar{W} = \frac{1}{N} \sum_{i=1}^N W_i$ and $\bar{Z} = \frac{1}{N} \sum_{i=1}^N Z_i$. Remember to show each step and provide explanations. (5 points)

Problem 3. Suppose that you are tasked with estimating the primary determinants of annual salaries in the USA. You initially obtain data (a sample) of n observations which contains the following variables for each individual in your sample: $salary_i$, $gender_i$, $education_i$ and $parentSalary_i$. Here $salary_i$ is specified in dollars per year, $gender_i$ is an indicator variable equal to 1 if female and 0 for non-female, age_i is specified in years, $education_i$ is specified in years, and $parentSalary_i$ is the average of the individuals parents salaries specified in dollars per year.

- (a) Consider the regression specification $wages_i = \beta_0 + \beta_1 gender_i + e_i$. You use some statistical software and estimate that $\hat{\beta}_0 = 55,000$ and $\hat{\beta}_1 = -11,000$. Interpret both estimated coefficients.(2 points)
- (b) Now consider instead the regression specification $\ln(wages_i) = \beta_0 + \beta_1 gender_i + \beta_2 education_i + e_i$. You use some statistical software and estimate that $\hat{\beta}_2 = 0.053$. Interpret the estimated value of β_2 .(2 points)
- (c) Now consider instead the regression specification $\ln(wages_i) = \beta_0 + \beta_1 gender_i + \beta_2 education_i + \beta_3 \ln(parentSalary_i) + e_i$. You use some statistical software and estimate that $\hat{\beta}_3 = 0.8$. Interpret the estimated value of β_3 .(2 points)
- (d) Suppose you obtain more variables. Specifically, you now have n observations and j independent variables, where $n = j - 1$. Suppose you run an OLS regression of Y_i on your $j - 1$ covariates and a constant. What is the R^2 ? Explain why you get this result and whether you have good explanatory power of the Y_i values in your sample and outside your sample. (2 points)