

# Economics 468

November 14, 2016

R. Davidson

## Assignment 3

1. Consider the linear regression model with  $n$  observations,

$$\mathbf{y} = \delta_1 \mathbf{d}_1 + \delta_2 \mathbf{d}_2 + \mathbf{u}, \quad \mathbf{u} \sim \text{IID}(\mathbf{0}, \sigma^2 \mathbf{I}). \quad (1)$$

The two regressors are dummy variables, with every element of  $\mathbf{d}_2$  equal to 1 minus the corresponding element of  $\mathbf{d}_1$ . The vector  $\mathbf{d}_1$  has  $n_1$  elements equal to 1, and the vector  $\mathbf{d}_2$  has  $n_2 = n - n_1$  elements equal to 1.

The parameter of interest is  $\gamma \equiv \delta_2 - \delta_1$ . Find the standard deviation of  $\hat{\gamma}$  (that is, the square root of its true variance) and write it as a function of  $\sigma$ ,  $n$ , and either  $n_1$  or  $n_2$ .

Suppose the data for regression (1) come from an experiment that you design and administer. If you can only afford to collect 800 observations, how should you choose  $n_1$  and  $n_2$  in order to estimate  $\gamma$  as efficiently as possible?

2. Consider the linear regression model with  $n$  observations,

$$\mathbf{y} = \beta_1 + \beta_2 \mathbf{d} + \mathbf{u}, \quad \mathbf{u} \sim \text{NID}(\mathbf{0}, \sigma^2 \mathbf{I}). \quad (2)$$

The only regressor here is a dummy variable, with each element equal to 1 for  $n_1$  observations and equal to 0 for the remaining  $n - n_1$  observations.

First, find the standard error of  $\hat{\beta}_2$  as a function of  $n$ ,  $n_1$ , and  $\sigma$ . Student's  $t$  distribution with  $n$  degrees of freedom can be defined as the distribution of the ratio of a standard normal variable to the square root of an independent  $\chi^2$  variable with  $n$  degrees of freedom divided by  $n$ ; schematically,

$$t_n = \frac{N(0, 1)}{\sqrt{\chi_n^2/n}}.$$

The non-central  $t$  distribution, with non-centrality parameter (NCP)  $\lambda$ , can be defined similarly as  $N(\lambda, 1)/\sqrt{\chi_n^2/n}$ . Suppose that the null hypothesis that  $\beta_2 = 0$  is tested by a  $t$  test. If we denote by  $\beta_2$  the true value of the parameter, find the distribution of the  $t$  statistic as a function of  $n$ ,  $n_1$ ,  $\beta_2$ , and  $\sigma^2$ .

3. Consider the linear regression model

$$\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \mathbf{u}, \quad \mathbf{u} \sim \text{N}(\mathbf{0}, \sigma^2 \mathbf{I}), \quad \text{E}(\mathbf{u} | \mathbf{X}) = \mathbf{0},$$

where  $\mathbf{X}$  is an  $n \times k$  matrix. If  $\sigma_0$  denotes the true value of  $\sigma$ , how is the quantity  $\mathbf{y}^\top \mathbf{M}_\mathbf{X} \mathbf{y} / \sigma_0^2$  distributed? Use this result to derive a test of the null hypothesis that  $\sigma = \sigma_0$ . Is this a one-tailed test or a two-tailed test?

Suppose that  $k = 3$ . Plot the power function for this test at the .05 level for the null hypothesis that  $\sigma = 1$  over the interval  $0.5 < \sigma < 2.0$  for three values of the sample size, namely,  $n = 13$ ,  $n = 23$ , and  $n = 43$ . **Hint:** This exercise does not require any simulations, but it does require you to calculate the cumulative  $\chi^2$  distribution function many times and its inverse a few times.

4. Consider again the data on consumption (or saving) and income used for Assignment 1. Now consider the following dynamic model

$$c_t = \alpha + \beta c_{t-1} + \gamma_0 y_t + \gamma_1 y_{t-1} + u_t,$$

where  $c_t = \log C_t$  and  $y_t = \log Y_t$ . Compute a  $t$  statistic for the hypothesis that  $\gamma_0 + \gamma_1 = 0$ . On the basis of this test statistic, perform an asymptotic test, a parametric bootstrap test, and a semiparametric bootstrap test using rescaled residuals.