

Problem Set 3

Prof. Santetti

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INSTRUCTIONS: Carefully read all problems. You must submit a single R script (Section 001) and Stata do-file (Section 002) with your *first name* (mine would be marcio.R and marcio.do). In case you submit your files with different names, you will lose 1 point.

You can find templates for your answer scripts/do-files on theSpring, under the "Templates" module. Please consider using it.

I should be able to fully replicate your code to answer the questions, as well as fully understand your written interpretations to the proposed problems.

Avoid using unnecessary code in your submission files. It is totally fine to do other things by yourself that may help you better understand the data and the problems. However, for grading purposes, I am only interested in the commands and interpretations that actually answer the questions. You may keep a separate file for yourself with your additional explorations.

Assignment due 11/22, midnight. Points Possible: 30

- You have 2 weeks to complete this assignment. In accordance with our course syllabi, no late submissions will be accepted.
- Be honest. Don't cheat.
- As a Skidmore student, always recall your votes of academic integrity, and the Honor Code you have abided by:

"I hereby accept membership in the Skidmore College community and, with full realization of the responsibilities inherent in membership, do agree to adhere to honesty and integrity in all relationships, to be considerate of the rights of others, and to abide by the college regulations."

Have fun!

Problem 1

Download the oecd.csv/.dta data set (available at the "Data files" module on theSpring), used by Everaert and Pozzi (2014) to examine the predictability of consumption growth in 15 OECD countries. For this exercise, however, we will only concentrate on US data.

(a) After importing the data set, filter only observations for the US economy (labeled "US").

(b) Estimate the following model:

 $csumptn_t = \beta_0 + \beta_1 hours_t + \beta_2 gov_t + \beta_3 r_t + \beta_4 inc_t + u_t$

(c) Run a *RESET* test (up to the 4th power) to verify whether (b)'s model is well specified. Assume α = 0.05. What do you conclude?

(d) Next, estimate a new model. Remove the statistically *insignificant* variable(s) from (b)'s model (hold the intercept!). Also, add the first *lag* of the dependent variable on the right-hand side.

(e) Then, estimate a RESET test for (d)'s model. What do you conclude? Explain.

Problem 2

Using the same data set from **Problem 1**, answer the following questions:

(a) Does the model from Problem 1, part (d) suffer from multicollinearity? Estimate Variance Inflation Factors (VIFs) and interpret your results.

(b) Does the model from Problem 1, part (d) suffer from serial correlation? Run a Durbin-Watson test and report your inference from this test.

(c) Following a similar procedure as in the previous part, run a Breusch-Godfrey test for first-degree serial correlation and report your inference from this test.

Problem 3

Using time-series data on five different countries, Atkinson and Leigh (2013) examine changes in inequality measured as the percentage income share held by those with the top 1% of incomes. A subset of their annual data running from 1921 to 2000 can be found in the data file inequality.csv/.dta, available at the "Data files" module on theSpring.

(a) Using data only for the US economy, estimate a regression model for share, controlling for tax,

(b) From (a)'s model, extract its residuals. Plot these over time.

(c) Still using (a)'s model residuals, plot the autocorrelation function (ACF) for this series. For how many lags is the autocorrelation coefficient (ρ) statistically significant? Explain.

(d) Does (a)'s model suffer from serial correlation? Estimate Durbin-Watson and Breusch-Godfrey tests, assuming α = 0.05. Interpret your results.

(e) Run a *RESET* test (up to the 4th power) to verify for possible functional form misspecification. Assume α = 0.05 and interpret your results.

(f) In case your answers to (d) were *positive* for serial correlation, use the *Cochrane-Orcutt* estimator for (a)'s model.

(g) Interpret the slope coefficient from (f).