Exercise: Distributional Analysis Using Pandas

Summary

This exercise carries out a detailed distributional analysis of the tax in the earlier market equilbrium assignment.

Input Data

File households.csv is the earlier file giving data on 1000 households. As before, it has five columns, id, type, inc, a and b, and there is one row for each household. The type variable wasn't used before but it will be important in this exercise. It ranges from 1 to 4 and indicates the demographic type of the household. Here we'll think of it as indicating the region of the country where the household resides. In other studies it could be used to indicate other characteristics, such as the race, gender, or age of the household's head, the size of the household, whether the household lives in an urban or rural area, and so on. In those cases there would be many more than four types.

File **quantities.csv** gives the quantities demanded by each household under the base case and tax policy simulations. It has three columns, one for the household's ID and one each for the household's demands under the base and policy cases: id, qd1 and qd2. In case you're curious, it was produced by running the earlier ind_demand() function for each equilibrium price and then writing out the output. However, you do not need to do that for this exercise: you should use the provided **quantities.csv** without recalculating it.

Deliverables

A script called **etr.py** that calculates the effective tax rate (ETR) for each household and then reports the median ETR for the groups indicated below. Then update **results.md** replacing the *TBD* placeholders with your answers to the questions in the file. Please try not to alter the other Markdown in the results.md file: it makes it easier to tell the difference between the questions and answers.

Instructions

To help make the structure of the analysis clearer, the instructions below have been divided into sections labeled A, B, etc. However, that's just for clarity: you should build a single script that does all of the steps.

A. Initial setup

- 1. Import pandas as pd.
- 2. Define a function called print_groups() that takes two arguments, a dataframe of household information called hh, and a list of variables called group_vars that will be used to group the data. The body of the function should do the following:
 - 1. Group the data by setting variable grouped to the result of calling the .groupby() method on hh using group_vars as the argument.
 - 2. Set variable med to the result of calling the .median() method on grouped column "etr".
 - 3. Round the results to two digits by setting variable med to the result of calling the .round() method of med using 2 as the argument. If you want, you can combine this step with the one above by adding the call to .round() to the end of the statement.
 - 4. Print med.
 - 5. Return med.

- 3. Create a dataframe called hh by using pd.read_csv() to read "households.csv". Use the keyword index_col='id' to set the index to the id field for each household.
- 4. Create a dataframe called q by using pd.read_csv() to read "quantities.csv". Again use the keyword index_col='id' to set the index to the id field for each household.
- 5. Determine the income quintile of each household by setting hh column "quint" to the result of calling the Pandas function pd.qcut() on the household income data. The pd.qcut() function divides its input into bins and returns a series containing the bin number of each input record. The first argument to qcut() should be the column of incomes, hh['inc'], the second argument should be 5 to request quintiles, and the third should be labels=[1,2,3,4,5] to have the quintiles labeled 1-5. Please note that pd.qcut() is a standalone Pandas function like pd.read_csv(): it's not a series or dataframe method.

B. Computing the ETRs

- 1. Now create a variable called pd1 that is equal to 53.35 and one called pd2 equal to 55.27 (the equilibrium buyer prices from the market assignment). Then create a variable called dp that is equal to pd2 pd1 (the buyer burden of the tax).
- 2. Compute the ETRs by multiplying 100 times dp times the qd2 column of q divided by the inc column of hh. Store the result in the hh dataframe as column 'etr'. Notice that it's not necessary for the quantity and income data to be in the same dataframe because Pandas will use the indexes to match up the income and quantity variables. Also, this would work correctly even if the rows of the hh and q datasets were not in the same order.

C. Aggregating and printing the results

Next we'll explore the results by aggregating the ETR results in several ways: by income quintile, by household type, and by both.

- 1. Aggregate and print the ETRs by income quintile by setting med_q to the result of calling print_groups() with arguments hh and ['quint'].
- 2. Next, aggregate and print the results by type by setting med_t to the result of calling print_groups() with arguments hh and ['type'].
- 3. Finally, aggregate and print the results by both type and income quintile by setting med_b to the result of calling print_groups() with arguments hh and ['type', 'quint'].
- 4. Print the index for med_b . Notice that it's a list of tuples with the type as the first element and the quintile as the second element.
- 5. Print an appropriate heading and then print the detailed medians for type 3 by using the .xs() method (short for cross-section) on med_b with arguments 3 and level="type". The level keyword tells Pandas that the 3 is associated with the "type" part, or level, of the index.
- 6. Print an appropriate heading and then list the medians for the 5th quintile by using the .xs() method with arguments 5 and level="quint".

D. Comparing within-group ETRs across income quintiles

1. Finally, to emphasize the power of the automatic alignment built into Pandas, we'll do a quick calculation showing how the ETR for each type changes moving up the income distribution. Start by setting etr_lowest

- to the result of calling .xs() on med_b with arguments 1 and level="quint".
- 2. Print etr_lowest . You'll see that it's a series with "type" as its index.
- 3. Now set variable etr_change to med_b minus etr_lowest. Pandas will automatically align the types, broadcast etr_lowest across all quintiles for each type (remember that med_b has two levels), and then do the subtraction. The result will be the difference in the ETR for each type and quintile from the ETR for quintile 1 of the same type. It's a quick way to see that the tax is progressive because the differences get progressively larger within each type.
- 4. Print etr_change.

E. Updating results.md

1. Now use the results to answer the questions in results.md.

Submitting

Once you're happy with everything and have committed all of the changes to your local repository, please push the changes to GitHub. At that point, you're done: you have submitted your answer.

Tips

To convince yourself that the last calculation is working correctly, use a pocket calculator to check a couple of the numbers. This feature of Pandas (aligning and broadcasting across index levels) is very useful and avoids a lot of steps that would otherwise have to be done manually. Pandas is essentially doing a many-to-one join on med_b and etr_change before doing the subtraction. However, it's all automatic and the code is much cleaner than it would be otherwise.