Title

persuasio — Conduct causal inference on persuasive effects

Syntax

persuasio subcommand varlist [if] [in] [, level(#) model(string)
method(string) nboot(#) title(string)]

Options

Option	Description
<pre>level(#) model(string) method(string) nboot(#) title(string)</pre>	Set confidence level; default is level(95) Regression model when covariates are present Inference method; default is method("normal") Perform # bootstrap replications Title

Description

persuasio conducts causal inference on persuasive effects. It is a wrapper
that calls a variety of subroutines. subcommand has several options:

Subcommand	Description
apr lpr yz	inference on APR when y,t,z are observed inference on LPR when y,t,z are observed inference on APR and LPR when y,z are observed
calc	bound estimates on APR and LPR with summary statistics

apr and lpr refer to a data scenario where binary outcomes y, binary treatments t, and binary instruments z are observed (with covariates x if exist) for each observational unit. apr and lpr provide causal inference on the average persuasion rate (APR) and the local persuasion rate (LPR), respectively.

yz is concerned with another data scenario where persuasive treatment t is unobserved. In this case, bounds on the APR are the same as those on the LPR. It provides causal inference for the APR and hence, for the LPR as well.

calc is designed for the case when summary statistics on $\Pr(y=1|z)$ and/or $\Pr(t=1|z)$ for each z=0,1 are available. It provides the lower and upper bounds on the APR as well as the lower and upper bounds on the LPR.

Options

model(string) specifies a regression model of y on z and x.

This option is only relevant when x is present. The default option is "no_interaction" between z and x. When "interaction" is selected, full interactions between z and x are allowed.

level(#) sets confidence level; default is level(95).

method(string) refers the method for inference.

The default option is **method**("normal"). By the nature of identification, one-sided confidence intervals are produced.

- When x is present, it needs to be set as method("bootstrap"); otherwise, the confidence interval will be missing.
- When x is absent, both options yield non-missing confidence intervals.

nboot(#) chooses the number of bootstrap replications.

The default option is **nboot**(50). It is only relevant when **method**("bootstrap") is selected.

title(string) specifies a title.

All these options are irrelevant for subcommands calc.

Remarks

It is recommended to use **nboot**(#) with # at least 1000. A default choice of 50 is meant to check the code initially because it may take a long time to run the bootstrap part. The bootstrap confidence interval is based on percentile bootstrap. Normality-based bootstrap confidence interval is not recommended because bootstrap standard errors can be unreasonably large in applications.

Examples

We first call the dataset included in the package.

. use GKB, clear

The first example conducts inference on APR when y,t,z are observed.

persuasio apr voteddem all readsome post, level(80) method("normal")

The second example conducts inference on LPR when y,t,z are observed.

. persuasio lpr voteddem_all readsome post, level(80) method("normal")

The third example conducts bootstrap inference on APR and LPR when y,z are observed with a covariate, MZwave2, interacting with the instrument, post.

. persuasio yz voteddem_all post MZwave2, level(80) model("interaction")
method("bootstrap") nboot(1000)

The fourth example considers the case when we have summary statistics on Pr(y=1|z) and/or Pr(t=1|z).

We first compute summary statistics.

```
foreach var in voteddem_all readsome {
  foreach treat in 0 1 {
    sum `var' if post == `treat'
    scalar `var'_treat' = r(mean)
  }
}
```

Then, we calculate the bound estimates on the APR and LPR.

. persuasio calc voteddem all 1 voteddem all 0 readsome 1 readsome 0

Stored results

apr calls this package's command persuasio4ytz, lpr command
persuasio4ytz2lpr, yz command persuasio4yz, and calc command
calc4persuasio, respectively. Check help files for these commands for
details on stored results.

Authors

Sung Jae Jun, Penn State University, <sjun@psu.edu>
Sokbae Lee, Columbia University, <sl3841@columbia.edu>

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References

Sung Jae Jun and Sokbae Lee (2022), Identifying the Effect of Persuasion, arXiv:1812.02276 [econ.EM]

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