Multitreat: Stata & Quarto*

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Abstract

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Keywords: Stata, Quarto, Reproducible Research, Two-Way-Fixed-Effects

Data Availability: Sources identified in the text

Public Code Repository: github.com/trr266/treat

Declaration of Interest: The author(s) declare(s) that they have no conflict of interest

1 Introduction

Hey there & congrats! You managed to skip the blindtext in the abstract. This means that we can dive into the world of reproducible research using coding languages like YAML, R, Stata, Python, Julia and many more… all from within one document: a **Quarto Markdown file (.qmd)**. Additionally, we provide you with two cherries on top of this fudge: a template .qmd-file for your next paper in our public GitHub repository and & an exemplary application of one of the most common designs in causal research: a two way fixed effects research approach under recent considerations of Callaway and Sant'Anna (2021). All set? Let's get started.

2 Identification Strategy & Results

Writing a paper with Quarto Markdown allows you to switch between coding languages at your will. Easiest is the implementation of R and Python, but access to Stata works too. You can either import Stata output as .tex-files, e.g. by the courtesy of Stata's user written estout package, or you immediately call Stata from within Quarto Markdown with the Statamarkdown package.[1]

2.1 Descriptive Statistics

[Table 1 about here.]

The .do-files for Stata contain many valuable information for our toy-project & we share them with a good reason: following open sciences practices makes your research FAIR: findable, accessible, interoperable and reusable as well as reproducible. This allows interested individuals to comprehend and verify identification strategies and analyses of your research and to immediately utilize the respective code. For instance, the .do-file "02_connect_wrds.do" showcases how to establish a connection to the data providers WRDS & Compustat from within Stata. You need *one* additional variable for your analyses? Simply adjust the code and build dependencies from thereon. For our examplary analysis, we pull raw data on dynamic and static information about US publicly listed firms from Compustat and then tidy the data. For instance, we perform a simple quality check and exclude firm-year observations with missing or negative sales from the

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[1] www.ssc.wisc.edu/ħemken/Stataworkshops/Statamarkdown/stata-and-r-markdown.html.

sample. For our subsequent analysis, we simulate a stable treatment effect on firms' sales (i.e. a one-standard-deviation increase). Whether or not a firm is designated as "treated" depends on a random draw. Table 1, which was created from within the .qmd-file, contains aggregated information on our outcome variable and firms' treatment-status. The year of treatment is also randomized, which leaves us with (i) never treated firms, (ii) always treated firms, and (iii) firms which receive treatment at some point throughout the sample (aka "staggered adoption"). Please note that you can easily access additional information on our research design by looking into the .do-files. One general advice on this end: make sure that your code is easy to understand for everyone by using explanatory notes within the code.

2.2 Difference-in-Differences with Multiple Time Periods

[Figure 1 about here.]

[Table 2 about here.]

The simulated treatment effect can empirically be identified in the data by employing the recent advances by Callaway and Sant'Anna (2021).[2] However, you should also want to visualize your data. While there are many approaches to this, we showcase one. We plot the Average Treatment Effect on the Treated (ATT) for the group of firms which received their treatment in 2008 (aka "2008-treatment-cohort") in Figure 1. In contrast to the presented coefficients in 2, the figure is not limited to a window of five years around the treatment. One can easily identify the (stable) treatment effect in the year of treatment and the subsequent years. One can also observe that there is no considerable fluctuation in the pre-periods of the treatment. Table 2 generalizes these visual inferences to our whole sample of treated and control firms.

3 Conclusion

Open science rocks and so does causal research in accounting! Learn more about writing in Quarto Markdown at quarto.org.

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[2]Please note that the Stata code for the analysis may take some time to run.

References

Callaway, Brantly, and Pedro H. C. Sant'Anna. 2021. "Difference-in-Differences with Multiple Time Periods."
Journal of Econometrics 225: 200–230. https://doi.org/https://doi.org/10.1016/j.econom.2020.12.001.

Appendix

Table A1: Variable Definitions

Name	Definition	Source
VarA	tbd	tbd
VarB	tbd	tbd
VarC	tbd	tbd

The .do-files for Stata comprise additional information to the explanations in Table A1. Please refer to the Online Supplement.
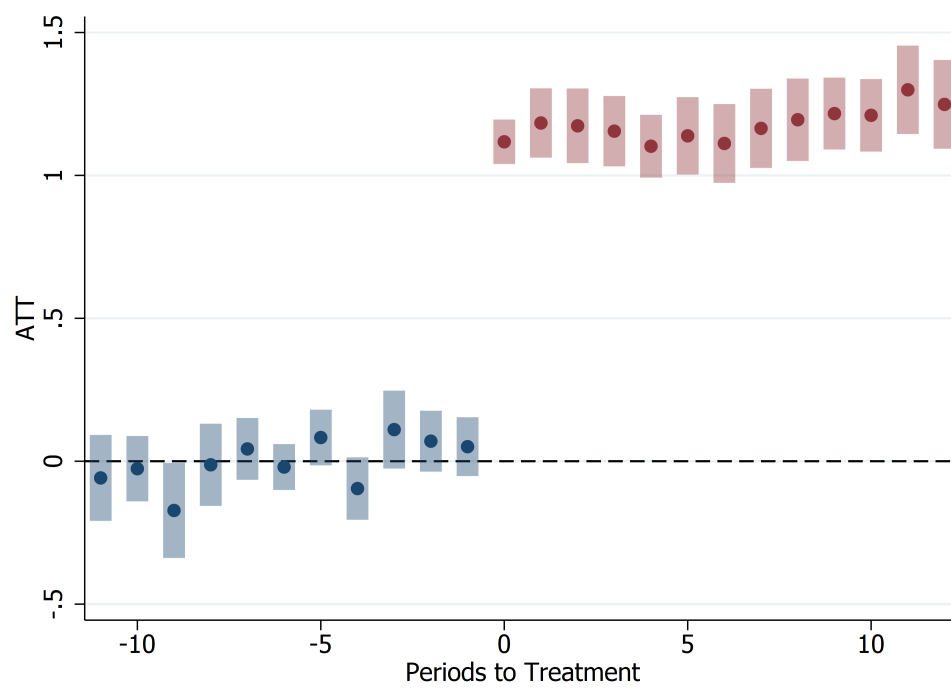
Figure 1: ATT for 2008-treatment-cohort

Table 1: Descriptive Statistics

	Mean	SD	Min	p50	Max
$\frac{sales}{lagged\,total\,assets}$	1.706804	1.227433	.007874	1.515617	7.69958
treated	.7501553	.4329247	0	1	1
post	.4188069	.4933656	0	0	1
N	123933				

Table 2: ATT by Periods Before and After Treatment

	$\dfrac{sales}{lagged\,total\,assets}$
Pretreatment Average	0.0218
	(0.005)
Posttreatment Average	1.117
	(0.013)
Pretreatment $t-5$	0.0171
	(0.016)
Pretreatment $t-4$	0.0170
	(0.013)
Pretreatment $t-3$	0.0206
	(0.013)
Pretreatment $t-2$	0.0399
	(0.013)
Pretreatment $t-1$	0.0145
	(0.012)
Treatment	1.130
	(0.011)
Posttreatment $t1$	1.107
	(0.014)
Posttreatment $t2$	1.105
	(0.016)
Posttreatment $t3$	1.114
	(0.017)
Posttreatment $t4$	1.110
	(0.020)
Posttreatment $t5$	1.134
	(0.021)

Standard errors in parentheses

Presentation limited to a window of 5 periods around treament.

Online Supplement

The public GitHub repositories at github.com/trr266/treat and at github.com/arndtupb/multitreat contain additional information. "multitreat" contains .do-files for Stata (*.do) and the Quarto markdown file (paper_stata.qmd).