

README: Electronic Payment Technology and Tax Compliance: Evidence from Uruguay's Financial Inclusion Reform

This file explains how to replicate the analysis for the paper “Electronic Payment Technology and Tax Compliance: Evidence from Uruguay’s Financial Inclusion Reform” (American Economic Journal: Economic Policy, by Anne Brockmeyer and Magaly Saenz Somarriba), using STATA.

Data Availability Statement

Administrative data

Access to the confidential tax data for Uruguay (DGI Uruguay, 2009-2016)

Most of the analysis in this paper uses firm-level VAT declarations and firm characteristics from the taxpayer registry held by the Uruguayan tax authority. The authors accessed this data under the requirement that no micro data could be shared with third parties. The data that the research team obtained is stored on a secure World Bank server which only authorized World Bank staff and consultants can access. The Uruguayan Tax Administration does not currently have a formal policy for granting access to these data. Any other person wishing to obtain access to the data would therefore have to approach the director of the Tax Administration and the head of the economic studies department and make a business case to them for access to the data. The case may also be reviewed by the Ministry of Finance. Since March 2020, the authors have not been able to obtain access to additional data for this project (beyond the data they had already accessed previously). The Tax Administrations’ policies for granting data access to researchers may change in the future, e.g. with a change in government.

The authors are happy to facilitate introductions for researchers proposing projects for which they could make a business case. The authors are also committed to run any sensitivity analyses and share the results with anyone wishing to run such analyses.

A replication analysis via the World Bank DIME replication service is being carried out.

If the AEA data editor seeks to access the data, or seeks to nominate a third-party researcher to conduct an audit for the replication files, the authors may be able to provide them access to the data through the World Bank secure data server, subject to some conditions. In particular, the individual seeking to access the data would either already be a World Bank staff or consultant, or needs to be provided a non-fee consulting contract with the World Bank, which is subject to World Bank management approval. The individual seeking to access the data will also need to sign and abide by the World Bank’s data confidentiality agreement.

Access to the confidential tax data for Costa Rica (DGT Costa Rica, 2013)

Appendix Figure D.4 (share of card sales in total reported sales in Costa Rica) uses micro administrative data from the Costa Rican tax administration as in the paper “Taxation, Information and Withholding: Evidence from Costa Rica” by Anne Brockmeyer and Marco Hernandez:

https://www.annebrockmeyer.com/uploads/1/2/1/4/121485108/costa_rica_paper_2022-11-14.pdf. Specifically, the data uses the D104 sales tax declaration and the D153 credit and debit card withholding form for the year 2013. From the D104 data, we use the tax return boxes 20, 21, 22 and 23 (variables: ventas por exportaciones, ventas exentas y autorizadas sin impuesto, ventas gravadas, otros rubros a incluir en la base imponible) and sum them to construct total sales. From the D153 declaration, we use the variable “monto total de transacciones”, which captures the total amount of card transactions. The access to this data is very similar to the access for the Uruguay data described above. The data is owned by the Costa Rican Tax Administration. The authors accessed this data under the requirement that no micro data could be shared with third parties. The data that the research team obtained is stored on a secure World Bank server which only authorized World Bank staff and consultants can access. The Costa Rican Tax Administration does not currently have a formal policy for granting access to these data. Any other person wishing to obtain access to the data would therefore have to approach the director of the Tax Administration and make a business case to them for access to the data. The case may also be reviewed by the Ministry of Finance.

The authors are happy to facilitate introductions for researchers proposing projects for which they could make a business case. The authors are also committed to run any sensitivity analyses and share the results with anyone wishing to run such analyses. A replication analysis via the World Bank DIME replication service is being carried out. If the AEA data editor seeks to access the data, or seeks to nominate a third-party researcher to conduct an audit for the replication files, the authors may be able to provide them access to the data through the World Bank secure data server, subject to the above-listed conditions.

Publicly available data

The following datasets need to be downloaded and saved in the folder `_DATA_CODES`. World Bank global findex data (Figure A1, panels A.I and A.II, and Figure A.9) (Demirguc-Kunt et al., 2011 and 2017). The Findex data can be downloaded here: <https://www.worldbank.org/en/publication/globalfindex/Data> (go to country-level data and download the stata version). The codes merge this data with GDP per capita data from the World Development Indicators (World Bank, 2019) (also used in some of the other figures mentioned below). To download the GDP per capita data, go to <https://databank.worldbank.org/source/world-development-indicators>, and select: Countries: all, time period: 2011 and 2017, series: GDP per capita, constant 2015 US\$. Unselect everything else. Then download as a CSV file. Unzip it, and discard the smaller of the two resulting CSV files (which is a meta data file). Save the other file under the name `WDI_gdp_pc.csv` in the data folder.

Informality rates (Figure A1, panel B) (Medina and Schneider, 2018)

The data is from Table A.1. in the IMF working paper by Schneider and Medina (2018): Shadow economies around the world: What did we learn over the last 20 years?

<https://www.imf.org/en/Publications/WP/Issues/2018/01/25/Shadow-Economies-Around-the-World-What-Did-We-Learn-Over-the-Last-20-Years-45583> We extracted the data from the pdf into a dta file (data_informality.dta) which is part of the replication package (in the folder PUBLIC_DATA_CODES – to be transferred into the data folder).

Uruguay households finance survey (Figure D.5, panels A and C) (Universidad de la Republica Uruguay, 2017)

These data, used to construct figure D5, panels A and C, can be downloaded here: https://docs.google.com/forms/d/e/1FAIpQLSdJqJG-3gH2ykI_woUf8Xx4hl47bI_HJTDFM3QZSeNkTM9-Pw/viewform?embedded=true If this does not work, go the his link: <https://cienciassociales.edu.uy/departamento-de-economia/investigacion/encuesta-financiera-de-los-hogares-uruguayos/> Go to EFHU 2, and click “solicitor acceso a microdatos”. A form will open that you should fill in and submit. The data will then be sent to the email address you indicate when submitting the form.

ENIGH household survey data from the Uruguay National Institute of Statistics (INE) (Figure D.5, panel B) (National Institute of Statistics, 2005-2006)

To obtain these data, go to the following link: <https://www4.ine.gub.uy/Anda5/index.php/catalog/42>, click on the tab “obtener microdatos”, register as a user and then enter your account details to enter the protected part of the website. Then download “Conjunto de datos de los hogares”, “Características demográficas de las personas del hogar” and “Consumos”. Unzip the files. You should obtain the files personas.DBF, hogar.DBF and consumos.DBF. Add these to the data folder. Alternatively (or if the above website link is not working), you can send an email to difusion@ine.gub.uy to request the micro data. In the document “Data request Spanish for INE.docx” in this replication folder, you will see the project description we provided to INE. Note that we used the 2005-2006 data, as it was the latest available data when we conducted the analysis. The 2016 data has recently been released.

Argentina Central Bank Means of Payment Statistics (Figure B.8) (Argentina Central Bank, 2022)

The data needs to be downloaded from: https://www.bcra.gob.ar/PublicacionesEstadisticas/Cuadros_estandarizados_series_estadisticas.asp selecting the following options: -> Volumen de transacciones, instrumentos de pago, infraestructura del sistema financiero e informaciones por división política.-> Medios de pago -> Volumen de transacciones e instrumentos de pago, por tipo de depósito y canal. This should download an excel file titled “debser”. Add this file to the data folder.

Data References

- Argentina Central Bank, “Means of Payment Statistics,” 2022. Accessed: 2022-12-04.
- Demirguc-Kunt, Asli, Leora Klapper, Dorothe Singer, Saniya Ansar, and Jake Hess, “Global Findex Database,” 2011 and 2017. Accessed: 2023-11-01.
- DGI Uruguay, “Administrative Tax Records,” 2009-2016. Accessed: 2015-2017

- DGT Costa Rica, “Administrative Sales Tax Records,” 2013. Accessed: 2015
- Medina, Leandro and Friedrich Schneider, “Shadow economies around the world: what did we learn over the last 20 years? Manual data extraction from table A.1,” IMF Working Papers, 2018.
- National Institute of Statistics, “Household Income and Expenditure Survey,” 2005-2006. Accessed: 2022-12-04.
- Universidad de la Republica Uruguay, “National Household Finance Survey,” 2017. Accessed: 2022-11-13.
- World Bank, “World Development Indicators,” 2019.

Computational Requirements

TECHNOLOGY ENVIRONMENT

The public part of the code was reproduced in a computer with the following specifications:
 - OS: Windows 11 Enterprise - Processor: Intel(R) Core(TM) i5-1145G7 CPU @ 2.60GHz -
 Memory available: 15.7 GB - Software version: Stata version 17

The private part of the code was reproduced in a private server with the following specifications:
 - OS: Windows 10 Enterprise - Processor: Intel(R) Xeon (R) CPU @ 2.80GHz -
 Memory available: 71.1 GB - Software version: Stata version 17

Instructions for Data Preparation and Analysis

This replication package contains two sets of codes. The first set of codes, in the folder ADMIN_DATA_CODES, draws on the confidential tax data. For users replicating the analysis on the World Bank secure tax data server, this folder should be transferred into the server (into the folder “Replication”) via OneDrive or with the help of World Bank IT (Brandon Mills). The first few lines in the Master do-file tell you which file paths you need to update and which folders you need to create so that the code executes correctly. Then please execute the master file, and this should run all other codes.

Some of the do files execute more than one exhibit, for this we created globals to select which specific to execute. To replicate everything make sure all options of globals in lines 80, 91, 108, 191-192, and 214 are all un-commented out.

At the end, one manual step is required: For the figures `did_2lines_nliab_ab_10_11_log_modif_99_r7_as1_esp10_wholesale.gph` and `did_2lines_nliab_ab_10_11_log_modif_99_r99_as1_esp10_wholesale.gph` [figure 3, panel C.I, and figure C.2, panel C.I], open the gph files, and in the note below the graph, add the point estimate (PE) and standard error (SE) from the synthetic difference-in-difference (SDID) estimation. For figure `did_2lines_nliab_ab_10_11_log_modif_99_r7_as1_esp10_wholesale.gph` [figure 3, panel C.I0], the PE and SE are found in `1 did_robust_table_logmodif_mixed.tex`, column 9 [table 1]. For Figure `did_2lines_nliab_ab_10_11_log_modif_99_r99_as1_esp10_wholesale.gph` [figure C.2, panel C.I], the PE and SE are found in `table did_robust_table_logmodif_mixed_r99.tex`, column 9 [table C.7]. After changing the PE and SE manually in these figures, export the figure as PDF, adding the `postscript_edit` to the name. Note that we opted for having this

manual step, as it allows us to otherwise run the SDID separately from the main DID analysis, which facilitates trouble-shooting the code, as the SDID takes a long time to run.

The second set of codes, in the folder PUBLIC_DATA_CODES, runs on publicly available data described above, and should not be transferred into the secure data server. These codes produce only appendix exhibits. The first few lines in Master_public_data.do tell you which file paths you need to update so that the code executes correctly. If any of the raw datasets are compressed, please unzip them.

Execution time

The codes for the publicly available data should take no more than 20 minutes to run.

The codes for the confidential tax data (on the server) should run in about 8 hours. To achieve this speed, we made a simplifying tweak in the synthetic difference-in-difference estimation (SDID) [setting a low number of bootstrap iterations] which implies that the present version of the code will replicate the point estimates for the SDID estimations (and everything else in the paper and appendix) but will not correctly replicate the SDID standard errors. We suggest that the replicator first runs this version of the code, to make sure everything executes, and then reruns it to obtain the right standard errors for the SDID.

To obtain the same SDID standard errors as in the paper and appendix, change the globals “run_time” and “replication” in 0_master_replication.do to 50 and 500 respectively (note that each of the globals appears multiple times in this code - lines 163,164,195,196,216, and 217). After this change, the code will take 3-4 days to run.

List of Exhibits

Given the number and complexity of the exhibits we provide the following index containing the specific filename as produced by the code for each figure/table. Additionally, in the ADMIN_DATA_CODES folder we included a .tex file named “Compiler_admin_data_exhibits” which can be executed in the server to create a PDF containing all exhibits from the admin data in the same order and numbering as in the paper. Please copy the compiler file into the output folder “D: AEJ” and execute it once the code is done running.

Figures and Tables Index:

Main Paper: Figure 1. AI. trend_log_countn_t_n02016 AII. trend_log_tot_sales_t_n02016
AIII.trend_log_count_pos_n02016 AIV. trend_log_count_firms_n02016 BI.
trend_d_countn_t_d1_n02016. BII. trend_d_tot_sales_t_d1_n02016
BIII.trend_d_count_pos_d1_n02016. BIV. trend_d_count_firms_d1_n02016

Figure 2. AI. RDrobust_log_count_trans_80_deseason_bc2_2015_main AII.
RDrobust_log_tot_sales_80_deseason_bc2_2015_main
AIII.RDrobust_log_count_pos_total_80_deseason_bc2_2015_main AIV.
RDrobust_log_count_firms_total_80_deseason_bc2_2015_main BI.
hist_1dtdif_countn_all_random_main BII. hist_1dtdif_tot_sales_all_random_main
BIII.hist_1dtdif_count_pos_all_random_main BIV. hist_1dtdif_count_firms_all_random_main

Figure 3. (panel CI requires manual edit described in the readme) AI.
did_2lines_s_ta_ab_10_11_log_modif_99_r7_as1_esp10_wholesale AII.
did_es_s_ta_ab_10_11_log_modif_99_r7_as1_esp10_wholesale BI.
did_2lines_nout_ab_10_11_log_modif_99_r7_as1_esp10_wholesale BII.
did_es_nout_ab_10_11_log_modif_99_r7_as1_esp10_wholesale CI.
did_2lines_nliab_ab_10_11_log_modif_99_r7_as1_esp10_wholesale_edit CII.
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Figure 4. A. subsidy_firms_share B. event_raw_b_c_2 C. scardsales_density_ms_3_line

Table 1. did_robust_table_logmodif_mixed

Appendix A.

Table A1. Non-data generated

Table A2. Non-data generated

Table A3 sumstat_uruguay_f3_ub_r8

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log_informal_2011

Figure A2 Non-data generated

Figure A3. A. ui_density_full_2014 (left) and ui_density_2014 (right) B. ui_density_full_2015
(left) and ui_density_2015 (right)

Figure A4. Non-data generated

Figure A5. Non-data generated

Figure A6. Non-data generated

Figure A7. A. did_vat_rebate_2014 B. did_count1_th_2014 C. did_tot_venta1_2014 D.
did_rebate_sale_2_2014

Figure A8. Non-data generated

Figure A9. (public data code) A. findex_1_4_q B. findex_3_4_q

Figure A10 vat_filers_month

Appendix B.

Figure B1. AI. trend_log_countn_n02016 AII.trend_log_tot_sales_n02016. BI.
trend_log_countn_t_n02016 BII. trend_log_tot_sales_t_n02016

Figure B2. A. RDrobust_firm_log_count_trans_80_1_all B.
RDrobust_firm_log_tot_sales_80_1_all

Figure B3. (his produce all outcomes) AI.
RDrobust_log_count_pos_total_80_deseason_bc1_2015_low AII.

RDrobust_log_count_firms_total_80_deseason_bc1_2015_low BI.
hist_1dtdif_count_pos_all_random_main_low BII.
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Figure B4. AI. robust_log_count_trans_1_d_w_0815 AII. robust_log_count_trans_2_d_w_0815
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robust_log_count_pos_total_1_d_w_0815

Figure B5. AI. robust_log_count_trans_1_d_w AII. robust_log_count_trans_2_d_w AIII.
robust_log_count_trans_3_d_w BI. robust_log_tot_sales_1_d_w BII.
robust_log_tot_sales_2_d_w BIII. robust_log_tot_sales_3_d_w CI.
robust_log_count_pos_total_1_d_w CII. robust_log_count_pos_total_1_d_w CIII.
robust_log_count_pos_total_1_d_w

Figure B6. AI. RDrobust_log_count_trans_80_deseason_bc2_2015_jan13 AII.
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AIII. RDrobust_log_count_pos_total_80_deseason_bc2_2015_jan13 AIV.
RDrobust_log_count_firms_total_80_deseason_bc2_2015_jan13 BI.
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BIII. hist_1dtdif_count_pos_all_random_jan13 BIV.
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Figure B7.
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BIII. placebo_rdd_inpos_2 BIV. placebo_rdd_infirms_2

Figure B8. (Public Data Code) Missing A. argentina_logtrans B. argentina_trans_d

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Table B1. rd_running_aggregation

Table B2. RD_cutweeks_d_short

Table B3. RD_steps_comparisson

Table B4. rdd_autoregression_lag1

Table B5. rdd_autoregression_lag2

Table B6. rdd_autoregression_prais

Figure B11. A. hist_rebate_fraction_month9_122 B. bar_rebates_9 C. hist_rebate_fraction_9

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RDrobust_log_count_t_70_deseason_bc1_mean_n B3.
RDrobust_log_count_c_70_deseason_bc1_median_n B4.
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Figure B13. AI. did_2lines_rebate_share_2_log_modif_s_ta_p25_0_ab_09_10 AII.
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AIII.did_2lines_rebate_share_2_log_modif_s_ta_p75_0_ab_09_10 BI.
did_period_rebate_share_2_log_modif_s_ta_p25_0_ab_09_10 BII.
did_period_rebate_share_2_log_modif_s_ta_p50_0_ab_09_10 BIII.
did_period_rebate_share_2_log_modif_s_ta_p75_0_ab_09_10 CI.
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event_sdid_log_modif_s_ta_09_ab_10_r100_p50_0 CIII.
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did_2lines_rebate_share_2_log_modif_nout_p50_0_ab_09_10
AIII.did_2lines_rebate_share_2_log_modif_nout_p75_0_ab_09_10 BI.
did_period_rebate_share_2_log_modif_nout_p25_0_ab_09_10 BII.
did_period_rebate_share_2_log_modif_nout_p50_0_ab_09_10 BIII.
did_period_rebate_share_2_log_modif_nout_p75_0_ab_09_10 CI.
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Appendix C.

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Table C2. did_robust_table_nz_mixed

Table C3. did_robust_table_logx_mixed

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AIII.did_es_s_ta_ub_16_11_log_modif_99_r7_as1_esp10_wholesale BI.
did_es_nout_ab_10_11_log_modif_95_r7_as1_esp10_wholesale BII.
did_es_nout_ub_10_11_log_modif_95_r7_as1_esp10_wholesale BIII.
did_es_nout_ub_16_11_log_modif_99_r7_as1_esp10_wholesale CI.
event_sdid_wholesale_log_modif_nliab_10_ab_r100_95_r7_10 CII. (empty) CIII.

Table C4. a. did_robust_table_fixedeffects_ab b. did_robust_table_fixedeffects_qb
c. did_robust_table_fixedeffects_ub

Table C5. a. did_robust_table_month_did_ab b. did_robust_table_month_did_qb
c. did_robust_table_month_did_ub

Table C6. did_robust_table_lengthvary

Figure C2. (panel CI requires manual edit described in the readme) AI.
did_2lines_s_ta_ab_10_11_log_modif_99_r99_as1_esp10_wholesale AII.
did_es_s_ta_ab_10_11_log_modif_99_r99_as1_esp10_wholesale BI.
did_2lines_nout_ab_10_11_log_modif_99_r99_as1_esp10_wholesale BII.
did_es_nout_ab_10_11_log_modif_99_r99_as1_esp10_wholesale CI.
did_2lines_nliab_ab_10_11_log_modif_99_r99_as1_esp10_wholesale_edit CII.
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Table C7. did_robust_table_logmodif_mixed_r99

Figure C3. AI. did_2lines_s_ta_qb_10_11_log_modif_99_r7_as1_esp10_serv_101 AII.
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did_2lines_nout_qb_10_11_log_modif_99_r7_as1_esp10_serv_101 BII.
did_es_nout_qb_10_11_log_modif_99_r7_as1_esp10_serv_101 CI.
did_2lines_nliab_qb_10_11_log_modif_99_r7_as1_esp10_serv_101 CII.
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Figure C4. AI. bar_size_rd_sector_sales AII. bar_size_rd_sector_number AIII.
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Figure C5. IA. did_es_nout_ab_10_11_log_modif_99_as1_esp10_region_rd_treat2 IB.
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did_es_nout_ab_10_11_log_modif_99_as1_esp10_region_rd_treat4 ID.
did_es_nout_ab_10_11_log_modif_99_as1_esp10_region_rd_treat3 IE.
did_es_nout_ab_10_11_log_modif_99_as1_esp10_region_rd_treat5 IIA.
did_es_nout_ab_10_11_log_modif_99_as1_esp10_sector_rd_treat2 IIB.
did_es_nout_ab_10_11_log_modif_99_as1_esp10_sector_rd_treat3 IIC.
did_es_nout_ab_10_11_log_modif_99_as1_esp10_sector_rd_treat4 IID.
did_es_nout_ab_10_11_log_modif_99_as1_esp10_sector_rd_treat5 IIE.
did_es_nout_ab_10_11_log_modif_99_as1_esp10_sector_rd_treat1

Figure C6. IA. event_sdid_log_modif_nout_10_ab_r100_99_10_1_3_treat2 IB.
event_sdid_log_modif_nout_10_ab_r100_99_10_1_3_treat1 IC.
event_sdid_log_modif_nout_10_ab_r100_99_10_1_3_treat4 ID.
event_sdid_log_modif_nout_10_ab_r100_99_10_1_3_treat3 IE.
event_sdid_log_modif_nout_10_ab_r100_99_10_1_3_treat5 IIA.
event_sdid_log_modif_nout_10_ab_r100_99_10_1_4_treat2 IIB.
event_sdid_log_modif_nout_10_ab_r100_99_10_1_4_treat3 IIC.
event_sdid_log_modif_nout_10_ab_r100_99_10_1_4_treat4 IID.
event_sdid_log_modif_nout_10_ab_r100_99_10_1_4_treat5 IIE.
event_sdid_log_modif_nout_10_ab_r100_99_10_1_4_treat1

Table C8. triple_did_stata_n_ab_rd_final

Appendix D.

Figure D1. A. count_transactions_2013 B. tot_venta_2013 C. count_pos_2013

Figure D2. AI. hascard_2012_share_T_2009_b AII. hascard_2012_share_T_2009_ub BI. hascard_2012_share_rw_2009_b BII. hascard_2012_share_rw_2009_ub

Figure D3 A. did_month_count_trans_b_log_modif_jan2012 B. did_month_tot_sales_b_log_modif_jan2012 C. did_month_hascard_b_log_modif_jan2012

Figure D4. costarica_cardsales_congraph_replication

Figure D5. (Public Data Code) A. conditional_simple B. formality_share C. debit_card_exp_share

Figure D6. AI. cost_rebates_nominal AII. cost_rebates_total AIII. cost_rebates_firms BI. cost_subsidy_nominal BII. cost_subsidy_total BIII. cost_subsidy_firms

Appendix E. Table E1. differences_hascard

Table E2. differences_hascard_retail

Table E3. hazard_hascard

Table E4. a. FirmFE_hascard b. FirmFE_hascard2

As mentioned in the readme, the simplified version of the code does not reproduce exactly the SE synthetic difference-in-difference estimation (SDID). The following exhibits are impacted by the simplification.

-> Fig 3, Panel CII. -> Table 1, Panel C -> Figure B13, panel C -> Figure B14, panel C -> Table C1, columns 9,10,11,12 -> Table C2, columns 9,10,11 -> Table C3, columns 9,10,11 -> Figure C1, panel CI and CII -> Table C6, columns 7,8,9 -> Figure C2, panel CII, -> Table C7, columns 9,10,11 -> Figure C6, all panels.