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VOTING TECHNOLOGY, POLITICAL RESPONSIVENESS, AND INFANT HEALTH: EVIDENCE FROM BRAZIL

THOMAS FUJIWARA
Princeton University, Princeton, NJ 08544, U.S.A., CIFAR, and NBER

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VOTING TECHNOLOGY, POLITICAL RESPONSIVENESS, AND INFANT HEALTH: EVIDENCE FROM BRAZIL

BY THOMAS FUJIWARA

This paper studies the introduction of electronic voting technology in Brazilian elections. Estimates exploiting a regression discontinuity design indicate that electronic voting reduced residual (error-ridden and uncounted) votes and promoted a large de facto enfranchisement of mainly less educated citizens. Estimates exploiting the unique pattern of the technology’s phase-in across states over time suggest that, as predicted by political economy models, it shifted government spending toward health care, which is particularly beneficial to the poor. Positive effects on both the utilization of health services (prenatal visits) and newborn health (low-weight births) are also found for less educated mothers, but not for the more educated.

KEYWORDS: Voting technology, enfranchisement, political responsiveness, infant health.

1. INTRODUCTION

REDISTRIBUTION OF RESOURCES IS A KEY ACTIVITY of the state, and its extent and form vary substantially worldwide. A government’s decision of how much to redistribute, and to whom, is embedded in a political system. This has long been recognized in positive theories where elections determine the allocation of resources, such as Romer (1975), Roberts (1977), and, in particular, Meltzer and Richard (1981).

A key prediction of these models is that increased political participation of poorer voters leads to more redistribution (toward them). Moreover, the conflict between an elite and masses demanding redistribution is the basis of theories of democratization and enfranchisement (Acemoglu and Robinson (2000, 2006)). This insight also has important policy consequences: a disadvantaged group’s electoral sway may be a precondition to policies targeting their wellbeing.

The evidence on this, however, is mixed and puzzling. While some estimated positive effects of democratization on government spending (e.g., Aidt and Jensen (2009a, 2009b, 2013), Acemoglu, Naidu, Restrepo, and Robinson (2013a)), others (e.g., Mulligan, Gil, and Sala-i-Martin (2004)) argued that there are no differences in policymaking in autocracies and democracies. There are also findings indicating that preexisting factors, such as liber-
alizations, determine the direction of democratization’s effect (Persson and Tabellini (2006)).

This paper provides evidence on how improving the political participation of less educated (poorer) voters can advance policies targeting them and affect their outcomes. It estimates the effects of an electronic voting (EV) technology in reducing a mundane, but nonetheless important, obstacle to political participation: difficulty in operating ballots. The results indicate that EV caused a large de facto enfranchisement of less educated voters, which led to the election of more left-wing state legislators, increased public health care spending, utilization (prenatal visits), and infant health (birthweight). Hence, while it addresses issues similar to the literature on democratization and de jure enfranchisement (discussed below), it does so at an “intensive margin,” focusing on improvements in participation within a context where universal suffrage is already established.

While filling a ballot is a trivial task to educated citizens in developed countries, the same is not true in Brazil, where 23% of adults are “unable to read or write a simple note.” Moreover, Brazilian paper ballots required citizens to write down their vote and involved only written instructions. This resulted in a substantial quantity of error-ridden and blank ballots being cast, generating a large number of residual votes (not assigned to a candidate and discarded from the tallying of results).

In the mid-1990s, the Brazilian government developed an EV technology as a substitute for paper ballots. While its introduction aimed at reducing the time and costs of vote counting, other features of the technology, such as the use of candidates’ photographs as visual aids, the use of error messages for voters about to cast residual votes, and guiding the voting process step by step, facilitated voting and reduced errors.

This paper first estimates the electoral effects of this voting technology exploiting a regression discontinuity design (RDD) embedded in EV’s introduction. Due to a limited supply of devices in the 1998 election, only municipalities with more than 40,500 registered voters used the new technology, while the rest used paper ballots. Estimates indicate that EV reduced residual voting in state legislature elections by a magnitude larger than 10% of total turnout. Such effect implies that millions of citizens who would have their votes go uncounted when using a paper ballot were de facto enfranchised. Consistent with

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2Acemoglu et al. (2013a) surveyed the evidence on democracy’s effect on redistribution and found it “both voluminous and full of contradictory results.” Putterman (1996) provided an overview on explanations for the variation in redistribution across democracies. See also Alesina and Angeletos (2005) and Bénabou and Tirole (2006) for more recent analyses and a discussion of this literature. There is also a large related literature on democratization and income, such as Acemoglu, Naidu, Restrepo, and Robinson (2013b) and references therein.

3Only 58% of adults (aged 25+) completed 4th grade. Figures are from the 1991 Census (the last one prior to the introduction of electronic voting).
the hypothesis that these voters were more likely to be less educated, the effects are larger in municipalities with higher illiteracy rates. Moreover, EV raises the vote shares of left-wing parties. These results are not driven by the (non-existent) effects on turnout or candidate entry (given the at-large nature of state elections).

The paper then argues that this enfranchisement of the less educated citizenry did indeed affect policy in a manner consistent with political economy theories of redistribution. Since the discontinuous assignment was observed in an election for state officials, I focus on state government spending, in particular on an area that disproportionately affects the less educated: health care. Poorer Brazilians rely mostly on a public-funded system for health care services, while richer voters are more likely to use private services (Alves and Timmins (2003)). The less educated have thus relatively stronger preferences for public health services, and a shift in spending toward health care can be interpreted as redistribution to the poor. Consistent with this interpretation, I also find that EV raised the number of prenatal visits by health professionals and lowered the prevalence of low-weight births (below 2500 g) by less educated women, but not for the more educated.

The state-level results exploit the fact that the discontinuous assignment in the 1998 election created specific and unusual differences in the timing of exposure across states. The phase-in of the new technology was carried out over three consecutive elections held in 1994, 1998, and 2002. In 1994, only paper ballots were used. In 1998, there was the discontinuous assignment described above. In 2002, only EV was used. Such a schedule implies that the evolution of EV in a state is entirely determined by a time-invariant cross-sectional variable: the share of voters living in municipalities above the cutoff for its use in 1998. If a state has $S\%$ of its voters living above cutoff, $S\%$ of its voters changed from using paper to EV technology between the 1994 and 1998 elections. Moreover, between the 1998 and 2002 elections, the remaining $(1 - S)\%$ of voters switched to EV. Hence, states with higher shares of voters above the 40,500-voter threshold experienced most of the enfranchising effects of EV earlier than the states with a low share. Intuitively, the empirical strategy tests if outcomes of interest track this same pattern.

The effects of EV on policy outcomes are thus identified only from variation coming from the interaction of a cross-sectional variable (share of voters above the cutoff) with the timing of elections. In the period (1994–1998) when

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4 Health care is also a very salient issue for voters and politicians, as discussed in Section 3.1.

5 Other than its importance for welfare and the growing evidence on the adult-life consequences of early-life health (Almond and Currie (2011), Currie and Vogl (2013)), the focus on birth outcomes is also motivated by issues of data availability and that newborn health can respond rapidly to health care improvements, which is important for an empirical strategy based on the sharp timing of EV roll-out.
such a variable positively predicts EV use, it also positively predicts valid voting, health care spending, number of prenatal visits, and birthweight. On the other hand, in the period (1998–2002) when the same cross-sectional variable negatively predicts EV use, it negatively predicts these outcomes. Such results are interpreted as evidence of a causal effect of EV since this sharp change in the sign of how the same cross-sectional variable predicts growth in outcomes is unlikely to be driven by omitted variables. In other words, any confounding effect would have to follow a very specific pattern to confound the results. More precisely, an omitted variable or mean-reversing shock that (positively) affects health care spending would need to (i) be growing faster in states with higher share of voters above the cutoff, and (ii) change to growing slower in such states with a timing matching election dates.

Four sets of additional tests provide evidence against these possible confounding effects. First, the timing of effects occurs quickly after elections, implying that possible omitted shocks (and their mean reversion) must follow quite specific timing. Second, the share of voters above the cutoff is orthogonal to changes in outcomes in periods when it is not associated with changes in voting technology, addressing the issue of preexisting trends. Third, I find negligible (placebo) effects on variables not expected to be affected by EV, such as general economic conditions and birth outcomes for more educated mothers, as well as spending by municipal governments, which were exposed to EV under different timing but should also respond to shocks to health care demand. Fourth, the results are robust to controlling for (nonlinear) time trends interacted with state characteristics, as well as an instrumental variable strategy that focuses on the distribution of municipalities closer to the cutoff.

The estimates indicate that the de facto enfranchisement of approximately a tenth of Brazilian voters increased the share of states’ budgets spent on health care by 3.4 percentage points (p.p.), raising expenditure by 34% in an eight-year period. It also boosted the proportion of uneducated mothers with more than seven prenatal visits by 7 p.p. and lowered the prevalence of low-weight births by 0.5 p.p. (respectively, a 19% and −6.8% change over sample averages).

This paper communicates with three different strands of economic literature. First, it is most closely related to the within-country analysis of enfranchisement’s effects, such as Cascio and Washington (2014), Husted and Kenny (1997), and Naidu (2012). These papers focus on legally mandated franchise extension to racial minorities in the United States, while I focus on a de facto enfranchisement episode in a developing context. It is also connected with the analysis of nationwide enfranchisement episodes (Acemoglu and Robinson (2000, 2006), Aidt and Jensen (2009a, 2009b, 2013)), as well as female enfranchisement. Miller (2008), in particular, found that women’s suffrage affected child health in the United States, and this paper thus corroborates the
connection between political participation and health, although for the case of poorer voters as opposed to women.⁶

Second, the introduction of EV in Brazilian elections was also studied by Hidalgo (2013) and Moraes (2012), who exploited the same discontinuous assignment in EV use in the 1998 elections. In particular, Hidalgo (2013) estimated the effect of EV on voting outcomes at the municipal level. Moraes (2012) performed a similar analysis and additionally focused on how the introduction of EV in state and federal elections affected spending by municipalities above and below the cutoff. Neither paper addresses spending at the state level (the focus of this paper).⁷ A growing literature evaluates voting technologies, but focuses mostly on election outcomes.⁸ Two notable exceptions are Baland and Robinson’s (2008) and Anderson and Tollison’s (1990) analyses of non-secret ballots in Chile and the United States, respectively. The former found that it helped support patron–client relationships, while the latter found that it diminished public spending. Third, a set of papers studies interventions to raise political responsiveness to (a group of) voters and its consequences for policymaking within developing countries, and the introduction of EV can be seen as such an intervention.⁹

The remainder of the paper is divided in three sections. Section 2 describes the electoral process and the introduction of EV in Brazil. It also reports RDD estimates at the municipal level to argue that EV enfranchised less educated voters. Section 3 first provides the background on the functioning of the public health care system, the politics of its provision, and its consequences for infant health. It then uses state-level panel data to estimate EV’s enfranchisement effect on health care funding and health outcomes, and addresses possible confounding factors. Section 4 concludes the paper.

2. THE IMPACT OF ELECTRONIC VOTING ON POLITICAL PARTICIPATION

2.1. Electoral Rules and Voting Technology in Brazil

Brazil is a federation of 27 states, which themselves comprise over 5000 municipalities. Each state has its own legislature (Assembléia Legislativa) and a


⁷The results in Hidalgo (2013) and Moraes (2012) largely corroborate those in this paper.


⁹Examples of interventions are political reservations for minorities (Pande (2003), Chattopadhyay and Duflo (2004)), participatory budgeting (Besley, Pande, and Rao (2005)), or plebiscites (Olken (2010)).
Federal legislation establishes the same election rules for all states. Legislators are elected under an open-list proportional representation with the entire state being a single multi-member district (at-large elections). Under open-list rule, a citizen casts a vote to an individual candidate, not a party list.\footnote{In reality, there are 26 states and the Federal District which, given the similarity in electoral rule and government form, is referred to as a state in this paper.}

The combination of large districts and voting for individual candidates makes voters face a large number of candidates, making listing names or illustrations in a ballot impractical. Moreover, state elections in Brazil are held jointly with federal elections for president, the federal senate, and the lower chamber of congress. For example, in the 1998 election, a voter in the state of São Paulo had to choose one candidate out of 1265 for the state legislature; 661 candidates for the lower chamber of federal congress; 10 candidates for state governor; 13 candidates for the federal senate; and 12 presidential candidates. This exemplifies the impossibility of simplified voting under paper ballots, as well as the overall complexity of voting.

Until 1994, Brazilian elections used only paper ballots. Voting for a state legislator involved writing down the chosen candidate’s name or electoral number, a five-digit code assigned by the election authority to each candidate. A paper ballot is depicted in Figure 1. There are only written instructions: to write “the name or number of candidate or party” on the box for “state” or “federal” legislator. In the mid-1990s, the independent branch of the federal judiciary that regulates electoral procedures (Tribunal Superior Eleitoral) introduced a new direct-record EV technology in order to reduce the time and costs of vote counting. The interface of the new technology is constituted of a small screen and a set of keys closely resembling a telephone keypad with three additional colored buttons. The voter types the candidate’s number into this keypad. The machine responds by providing her name, party affiliation, and picture on the screen. The voter can then confirm the vote or choose to start the process again. Figure 1 depicts the initial screen of the device, and the screen just before a vote can be confirmed.

There are at least four distinct reasons why less educated voters would find voting under EV easier. First, it provides visual aids (candidates’ pictures) which were absent in paper ballots. Second, the machine guides the user through the many votes that must be cast (state legislator, governor, federal deputy, senate, president). This minimizes voter confusion over which paper

\footnote{In essence, all votes received by candidates from the same party are added and this total proportionally determines the number of seats awarded to the party. The candidates within a party list are then ranked by the number of votes each individually received, and this rank is followed when assigning candidates to seats within a party. Voters also have the option of instead casting a vote to the party label only, which is then counted as a vote to the party list but has no impact in the within-party ranking of candidates. Governors are elected under a two-round (runoff) plurality rule.}
Paper ballot

Initial screen of the voting technology

Voting for (fictional) candidate number 92111 (name: Monteiro Lobato, party: PLT)

FIGURE 1.—Examples of the voting technologies.

ballot to use for each vote (e.g., casting the vote for federal legislator in the state ballot). Third, the machine provides feedback to users. A voter can check if the number he typed corresponds to his choice of candidate. If the typed number does not correspond to an existing candidate, the machine provides a message stating “wrong number” (and does not provide a picture), alerting the voter that he is about to cast a residual vote. Fourth, it uses a number-based
interface that resembles familiar devices (telephone keypads and ATMs). As noted by electoral authority staff, “numbers are the easiest way for people to interact with an interface. People with low schooling, through the use of numbers, can use telephones. The illiterate are able to make phone calls” (Tribunal Superior Eleitoral (2011)).

The Supplemental Material (Fujiwara (2015)) further describes voting procedures under both paper and EV technologies and discusses why the guidance and feedback can have an effect on an illiterate voter, given the device’s graphical interface. The government also conducted a large-scale campaign, using public service announcements on TV and radio to instruct the public on how to use the new machines. While the discussion above indicates that the elections authority took deliberate steps to design an easily operable voting technology, reducing residual votes does not appear to have been the primary aim.

2.1.1. Introduction of Electronic Voting and the Design of This Study

Elections for state officials are held in Brazil every four years, and all states have the same election date. In the 1990 and 1994 elections, only paper ballots were used. In the 1998 elections, the electoral authority decided that only municipalities with more than 40,500 registered voters (as of 1996, when municipal elections took place) would use the new electronic technology, while municipalities below this threshold would use paper ballots. The assignment rule was adopted due to the limited production capacity of the manufacturer and economies of scale in the distribution within large municipalities. In the next election, EV became the sole method of collecting votes. This threshold-based rule used in the 1998 election creates a standard RDD which can be exploited to estimate the effects of EV. Such interpretation requires agents to have limited control over the forcing variable: registered voters. This is likely the case since the assignment rule (e.g., the value of the cut-

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12 An interesting feature of the electronic technology is that it requires the voter to know the number of his preferred candidate at the time he is casting a vote. This usually occurs through the use of flyers or cheat-sheets that candidates distribute during their campaigns. These (usually pocket-sized) pieces of paper, known as santinhos, feature the candidate’s face, name, and number, and can be brought by the voter to the polls.

13 The head designer of the technology describes the new system’s association with a reduction in residual votes as “a surprise” (Tribunal Superior Eleitoral (2011)).

14 Specifically, the 40,500-voter threshold was determined by assigning the fixed supply of devices available at the time from largest to smallest municipality and establishing where it would end. The distribution economies of scale are due to the fact that distributing a large number of devices across few large municipalities is less costly than distributing few devices across several small municipalities.

15 Municipal elections do not coincide with state races. EV was also used in the 1996 municipal elections (in selected large cities and state capitals), and the next municipal election (2000) was the first where only EV was used. This paper, however, focuses only on state elections (1994, 1998, and 2002).
off) was announced in May 1998 and the forcing variable is voter registration for municipal elections two years earlier (1996).\footnote{Information that EV would be gradually rolled out was available at least since 1995, but the decision on which municipalities would use it in the 1998 election was only made months before the election.}

Finally, four states used EV in all its municipalities during the 1998 election. Two remote states largely covered by the Amazon forest (Amapá and Roraima) were chosen to check the electoral authority’s ability to distribute EV in isolated areas, while the states of Rio de Janeiro and Alagoas had areas where the army provided security to election officials, allowing an opportunity to check the logistics of distributing the electronic devices jointly with the military. These states are dropped from all samples used in this section, but will be used as robustness checks of the state panel estimates in Section 3.

2.2. Data and Estimation Framework

2.2.1. Municipal Data and Electoral Outcomes

Information on voter registration, turnout, and election results at the municipality level for several years was obtained from the federal electoral authority. The institution also published reports listing the municipalities that used EV, showing an almost perfect compliance with the discontinuous assignment rule. All the 307 municipalities (out of 5281) above the 40,500-voter cutoff used EV in 1998.\footnote{Seven (out of 5281) municipalities below the threshold deviated from the rule, by having their (formal) requests to use EV accepted by the electoral authority. The transcripts of the requests and decisions indicate mainly vague idiosyncratic reasons (e.g., no municipality would be using EV in a particular subregion of a state). The next subsection discusses why accounting for this almost negligible deviation from the discontinuity design has no impact on the estimates.} Additional data on municipal characteristics are from tabulations of the 1991 Brazilian Census.

The main outcome of interest is the number of votes that are valid (i.e., nonresidual). A vote is considered valid if, and only if, it can be assigned to a particular candidate or party and then counted in their vote shares. A vote that is not valid is defined as a residual vote. Hence, turnout equals the sum of valid and residual votes. A vote cast in a paper ballot is deemed residual if it is left blank or if the name or number written on the ballot does not correspond to a candidate. Under EV, a residual vote may be cast by pressing the “blank” button or by typing and confirming a number that does not correspond to any candidate. This section focuses on the election outcomes for state legislature elections. The relevant sample moments are provided in the discussion and in the tables reporting estimates. The Supplemental Material discusses how a vote can be residual under paper and EV, as well as results for federal and gubernatorial races.
2.2.2. Estimation Framework

Let $v_m$ be the number of registered voters in municipality $m$. The treatment effect of a switch from paper ballots to EV on outcome $y_m$ is given by

$$TE = \lim_{v_m \downarrow 40,500} E[y_m|v_m] - \lim_{v_m \uparrow 40,500} E[y_m|v_m].$$

(1)

Under the assumption that the conditional expectation of $y_m$ on $v_m$ is continuous, the first term on the right-hand side converges to the expected outcome of a municipality with 40,500 voters using EV, while the second term converges to the expected outcome of one using paper ballots. Hence, $TE$ identifies the treatment effect for a municipality with 40,500 voters. This highlights that the estimated effects are “local” and apply only to municipalities “at the cutoff.” In practice, one should be careful in extrapolating the effects to smaller and larger municipalities. In particular, the nature of elections might be very different in large cities, and the effect of EV different in those cases.

The estimation follows the recommendations in Imbens and Lemieux (2008), Lee and Lemieux (2010), and Imbens and Kalyanaraman (2012). The limits are estimated nonparametrically by local linear regression. This consists of estimating a regression of $y_m$ on $v_m$ using only data satisfying $v_m \in [40,500 - h; 40,500]$. The predicted value at $v_m = 40,500$ is thus an estimate of the limit of $y_m$ as $v_m \uparrow 40,500$. Similarly, a regression using only data satisfying $v_m \in [40,500; 40,500 + h]$ is used to estimate the limit of $y$ when $v_m \downarrow 40,500$. The difference between these two estimated limits is the treatment effect. Although linear regressions are used, the nature of estimation is nonparametric, and the consistency of the results holds for arbitrary shapes of the relationship between $y_m$ and $v_m$.

The local polynomial regression estimate is equivalent to the OLS estimation of the following equation using only observations that satisfy $v_m \in (40,500 - h; 40,500 + h)$:

$$y_m = \alpha + \beta 1\{v_m > 40,500\} + \gamma v_m + \delta v_m 1\{v_m > 40,500\} + \epsilon_m,$$

(2)

where $1\{v > 40,500\}$ is a dummy variable that takes value 1 if the number of registered voters is above 40,500. The estimate of $\beta$ is the treatment effect. No

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Regressions are unweighted (rectangular kernel). This approach is appropriate for “sharp” discontinuity designs, where the probability of treatment is zero below the threshold and 1 above it (perfect compliance). Even though there is (an almost negligible) lack of compliance since seven municipalities below the threshold used EV, the estimated effect of crossing the threshold on the probability of using EV is a precisely estimated one, implying that approaching the estimation as a fuzzy design would lead to nearly identical results. In other words, the local average treatment effect (LATE) equals the average treatment effect (ATE) at the cutoff. Moreover, dropping the seven noncompliers from the sample or treating them as “treated” leads to no change in any of the results.
covariates are included in equation (2), although the next section considers the inclusion of state fixed effects as a robustness check.

A key decision is \( h \), the kernel bandwidth, and the trade-off between precision and bias. This paper reports estimates based on Imbens and Kalyanaraman’s (2012) optimal bandwidth choice (IKBW), which is itself a function of the data and hence different for each \( y \). In some (but few) cases the computed IKBW is relatively large, and to reinforce the local intuition of RDDs, the largest possible IKBW is capped at 20,000 voters. There are 428 (130) municipalities below (above) the cutoff in the sample with a 20,000-voter bandwidth. To check the robustness of results, I also provide estimates based on \( h = 10,000 \) and \( h = 5000 \). These samples include 151 (78) and 75 (41) municipalities above (below) the cutoff, respectively.

### 2.3. Regression Discontinuity Results

#### 2.3.1. Electronic Voting and Residual Votes

To check if municipalities above and below the cutoff are comparable, Table I estimates effects on “predetermined” covariates calculated from the 1991 Census data. These are monthly income, income inequality, education variables, geographical location, population, and the urbanization rate (share of population in an area officially designated as urban). All effects are close to zero in magnitude and statistically insignificant.

Before reporting the main results, some graphical evidence is provided. Figure 2 plots the main outcome of interest (valid votes in state legislature elections, as a share of turnout) against the forcing variable (registered voters in 1996) for three different elections. Each marker represents a local average: the mean of the outcome in a bin of municipalities within a 4000-wide interval of the forcing variable. A quadratic polynomial is fitted on the original (i.e., “un-binned”) data at each side of the vertical threshold, so that the point where the lines are not connected is where the discontinuity in outcomes, if existent, is expected to be visible.

A clear jump is visible in the 1998 election (in circles). A little over 75% of the votes are valid on the municipalities below the cutoff, and this figure rises to close to 90% as the cutoff is crossed and EV is introduced. The fact that no discontinuity is visible for the elections held in 1994 (when all municipalities used paper ballots) and 2002 (when EV was completely phased in) provides a falsification test indicating that municipalities “just above” and “just below” the cutoff are indeed valid treatment and control groups.

Figure 3 repeats this exercise for turnout (as a share of registered voters) and voter registration (as a share of total population)\(^{19}\) in the 1998 election. There

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\(^{19}\)Total population, and not voting-age population, is used. As the main interest of this analysis is if registration evolves smoothly around the cutoff, the choice of denominator is not particularly relevant.
are no visible discontinuities, implying the turnout and registration behavior are the same in both treatment and control municipalities. There are at least three complementary factors explaining why voters did not respond to a technology that increased their likelihood of casting a valid vote. First, it is possible that voters did not anticipate this effect (as those immediately above the cutoff are using it for the first time). Second, the main driver of voters to the polls may be the presidential election, for which EV had relatively small impacts (Supplemental Material). Third, turnout rates are relatively high in Brazil. Federal law makes registration and voting compulsory for all citizens aged 18–70. Failing to register or vote renders a citizen ineligible to several public services until the payment of a fine. Although these features do not guarantee a turnout close to

<table>
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<th>TABLE I</th>
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<tr>
<td>SUMMARY STATISTICS AND COVARIATE SMOOTHNESS (1991 CENSUS)*</td>
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<td>Monthly Income</td>
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<td>Gini Index</td>
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<td>Illiteracy Rate</td>
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<td>Share w/o 4 Years of Schooling</td>
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<tr>
<td>Share w/o 8 Years of Schooling</td>
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<td>Population—1991 (Thousands)</td>
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<td>Population—2000 (Thousands)</td>
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<td>Share of Urban Population</td>
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<tr>
<td>Bandwidth</td>
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<td>Observations</td>
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*aRobust standard errors in parentheses, standard deviations in brackets, number of observations in curly brackets—(1). The unit of observation is a municipality. Each figure in columns (1)—(3) is from a separate local linear regression estimate with the specified bandwidth. The pre-treatment mean is the estimated value of the dependent variable for a municipality with 40,500 registered voters that uses paper ballot (based on the specification on column (1)). The IKBW column provides the Imbens and Kalyanaraman (2012) optimal bandwidth (capped at 20,000) and the associated number of observations. Details on the dependent variables in the text.
FIGURE 2.—Valid votes/turnout—local averages and parametric fit. Each marker represents the average value of the variable in a 4000-voter bin. The continuous lines are from a quadratic fit over the original (“unbinned”) data. The vertical line marks the 40,500-voter threshold.

FIGURE 3.—Registration and turnout—local averages and parametric fit.
When using it. While some residual votes may be deliberate abstentions and
lot technology, with a quarter or more of the electorate casting residual votes
decision to register and vote. 20

<table>
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<tr>
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<th>Full Sample Pre-Treat.</th>
<th>IKBW</th>
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<td>Panel A: Baseline Results</td>
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<tr>
<td>Valid Votes/Turnout (1998 Election)</td>
<td>0.755</td>
<td>0.780</td>
<td>11,873</td>
<td>0.118</td>
<td>0.121</td>
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<tr>
<td>Turnout/Reg. Voters (1998 Election)</td>
<td>0.765</td>
<td>0.785</td>
<td>12,438</td>
<td>-0.005</td>
<td>0.013</td>
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<tr>
<td>Reg. Voters/Population (1998 Election)</td>
<td>0.748</td>
<td>0.737</td>
<td>15,956</td>
<td>-0.004</td>
<td>0.010</td>
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<td>Panel B: Placebo Tests (Election Years Without Discontinuous Assignment)</td>
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<tr>
<td>Valid Votes/Turnout (1994 Election)</td>
<td>0.653</td>
<td>0.697</td>
<td>17,111</td>
<td>-0.013</td>
<td>-0.008</td>
</tr>
<tr>
<td>Valid Votes/Turnout (2002 Election)</td>
<td>0.928</td>
<td>0.921</td>
<td>17,204</td>
<td>0.005</td>
<td>0.008</td>
</tr>
<tr>
<td>Panel C: Do Left-Wing Parties Benefit Disproportionately From Electronic Voting?</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Vote-Weighted Party Ideology (1998 Elec.)</td>
<td>5.397</td>
<td>5.162</td>
<td>20,000</td>
<td>-0.222</td>
<td>-0.250</td>
</tr>
<tr>
<td>Bandwidth Specification</td>
<td>Linear</td>
<td>Linear</td>
<td>Linear</td>
<td>IKBW</td>
<td>10,000</td>
</tr>
<tr>
<td>N</td>
<td>5281</td>
<td>229</td>
<td>116</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

*Robust standard errors in parentheses, standard deviations in brackets, number of observations in curly brackets—{}.* The unit of observation is a municipality. Each figure in columns (1)–(3) is from a separate local linear regression estimate with the specified bandwidth. The pre-treatment mean is the estimated value of the dependent variable for a municipality with 40,500 registered voters that uses paper ballot (based on the specification on column (1)). The IKBW column provides the Imbens and Kalyanaraman (2012) optimal bandwidth (capped at 20,000) and the associated number of observations. Details on the dependent variables in the text.

100%, it is possible that it makes voting technology a second-order issue in the
decision to register and vote. 20

Panel A of Table II presents the local linear regression estimates of the treat-
ment effects. Irrespective of the bandwidth and specification used, the esti-
rated effect of a switch from paper ballot to EV is an increase in the valid
vote-turnout ratio in the order of 12 p.p. and is statistically significant at the
1% level. Adding state fixed effects has little impact on the results. 21

Table II and Figure 1 both demonstrate the inadequacy of the paper bal-
lot technology, with a quarter or more of the electorate casting residual votes
when using it. While some residual votes may be deliberate abstentions and

20 Figure 3 and Table II show that turnout is in the order of 85% of registered voters. Citizens
who are not in their city of residence on election day can be waived from sanctions by attending
a poll in any other municipality and submitting a “waiver form."
21 Adding state fixed effects to the specification on column (1) yields an estimate of 0.139
(s.e. = 0.013).
protest votes, such large shares probably reflect error. Panel A also presents the estimated treatment effects on turnout (as a share of registered voters) and voter registration (as a share of population). Confirming the graphical analysis, the effects are numerically small and statistically insignificant. Since population is not affected by crossing the cutoff (Table I), these results imply that the number of valid votes is driving all the results: turnout, measured either as a share of registered voters or population, is not affected.

In addition to the covariate smoothness tests on Table I, Panel B of Table II also provides a more stringent placebo test. It estimates effects in the election years without a discontinuity present: when all municipalities used paper ballots (1994) or EV (2002). The results indicate zero impact across all specifications. This supports the previous interpretation of the results, since if any unobserved difference in municipalities just above and just below the threshold was driving the result observed in 1998, it would likely have been present in the preceding or subsequent election. Finally, there is no effect of crossing the cutoff (i.e., using EV in 1998) on turnout in the 2002 election: the estimated effect is $-0.002$ with a s.e. = 0.011 when using the IKBW.$^{22}$

2.3.2. Which Parties Benefited From Electronic Voting?

Brazilian politics is characterized by a multiparty system. In the 1998 election, 30 separate parties had candidates running for state legislature. To systematically assess which parties gained from the introduction of EV, I rely on a measure of party position in a left-to-right scale from Power and Zucco (2009) to define party ideological positions. The index is constructed from answers of a survey of federal legislators elected in 1998, and ranks parties from zero to ten, with larger numbers being associated with right-wing ideologies. Power and Zucco (2009) described the methodology and argued that more leftist positions in the index can be associated with more redistribution to poorer voters.$^{23}$

It must be noted that the measure captures party positions at the federal congress. A common perception of Brazilian politics is that party positions at the federal level do not correlate with their actions at the state level (Ames (2001)). In the context of this paper, this is likely to introduce noise in the outcome, biasing the results toward zero.

---

$^{22}$Turnout is measured as a share of registered voters. There is also no effect on the registered voters/population ratio (t.e. = −0.002, s.e. = 0.012) and population measured in the 2000 Census (the effect is equal to 1.62 thousand people from a pre-treatment mean of 69.79 thousand, with s.e. = 3.04 thousand.

$^{23}$Power and Zucco (2009) also discussed how their measure matches well common perceptions, stated party goals, historical legacies, and previous attempts at party classifications.
Let \( i_p \) denote the index for party \( p \), and \( \text{votes}_{pm} \) the number of votes for party \( p \) in municipality \( m \). I construct the following vote share weighted measure of the parties receiving votes from a municipality \( m \):

\[
I_m = \sum_p \left( i_p \cdot \frac{\text{votes}_{pm}}{\sum_p \text{votes}_{pm}} \right) = \frac{\sum_p i_p \cdot \text{votes}_{pm}}{\text{valid votes}_m}. 
\]

Note that \( \sum_p \text{votes}_{pm} \) equals the total number of (valid) votes in a municipality, so that the fraction multiplying \( i_p \) inside the parentheses is the municipal vote share of party \( p \). Hypothetically, \( I_m \) goes from zero to ten (the cases where an extreme left- or right-wing party takes all the votes).24

Not all parties that received votes in state legislature elections are included in Power and Zucco’s (2009) analysis. However, the 11 parties for which the index is available account, on average, for 90% of the valid votes in sample municipalities. More importantly, the ratio between votes of parties for which the index is available and the total number of valid votes is not affected by EV, so that changes in the vote shares of parties for which position is not observed cannot account for the results.25 As noted before, the at-large nature of state elections implies that voters residing in municipalities above and below the threshold (but within the same state) face the exact same choice of candidates, so that these estimates cannot be driven by entry of parties or candidates.

Panel C of Table II reports the estimated treatment effect of EV on \( I_m \). The only difference in the estimation from the other panels is the inclusion of two controls: the state average of \( I_m \) and the ratio between votes for parties for which the index is available and total valid votes. For the reasons described above, these two variables evolve smoothly around the cutoff, and are included only to increase the precision of the estimates. Their inclusion has little effect on the point estimates.26

The effect on column (1) indicates that the vote share-weighted position of parties moves by 0.22 to the left when EV is used. Adding state fixed effects leads to very similar estimates. This effect is close to a third of a standard deviation in the distribution of the outcome. The effect is robust to using the

\[24\]In practice, the maximum and minimum values are 6.91 and 2.27 (the maximum and minimum values in \( i_p \) are 7.14 and 1.68).

\[25\]In other words, the estimated treatment effect of EV on the ratio between votes for parties covered by Power and Zucco (2009) and total valid votes is numerically close to zero and statistically insignificant.

\[26\]Because of the inclusion of state averages, the standard errors are also clustered at the state level. To address the relative small number of clusters, \( p \)-values based on the Cameron, Gelbach, and Miller (2008) wild bootstrap procedure are also computed: their values for the estimates on columns (1), (2), and (3) are 0.038, 0.002, and 0.568, respectively.
10,000-voter bandwidth, but becomes smaller and imprecise when the 5000-voter bandwidth is used. Given this and the issues in measuring state-level party positions discussed above, the results should be interpreted with care, but they suggest that under EV, parties with relative left-wing inclination benefit disproportionately (i.e., their share of the valid votes increases).

2.3.3. *Electronic Voting and the de facto Enfranchisement of the Less Educated*

While the results above indicate that EV generated the effective political participation of a large share of the electorate, it does not address the issue of the identity of these newly enfranchised voters. Section 2.1 discussed why it is presumable that the less educated are more likely to cast residual votes and, hence, EV should make voting simpler for those with poorer reading and writing skills. Supporting this hypothesis, the effect of EV is stronger in municipalities with less educated populations.

Table III repeats the estimation presented on the first line of Table II, but dividing the sample into two categories: municipalities with illiteracy rates below and above the sample median. This information is from the 1991 Census and measures the share of adults (aged 25+) that report “not being able to read nor

| TABLE III |
| Treatment Effects of Electronic Voting, by Illiteracy Rate\(^a\) |

<table>
<thead>
<tr>
<th>Pre-Treat. Mean</th>
<th>IKBW (Obs.)</th>
<th>(1)</th>
<th>(2)</th>
<th>(3)</th>
<th>(4)</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Panel A: Municipalities With Above-Median Illiteracy</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Valid Votes/Turnout</td>
<td>0.759</td>
<td>11,873</td>
<td>0.147</td>
<td>0.150</td>
<td>0.152</td>
</tr>
<tr>
<td>(0.017)</td>
<td>(0.019)</td>
<td>(0.015)</td>
<td>(0.020)</td>
<td>(0.031)</td>
<td></td>
</tr>
<tr>
<td>N</td>
<td>—</td>
<td>—</td>
<td>116</td>
<td>279</td>
<td>103</td>
</tr>
<tr>
<td><strong>Panel B: Municipalities With Below-Median Illiteracy</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Valid Votes/Turnout</td>
<td>0.799</td>
<td>11,873</td>
<td>0.092</td>
<td>0.113</td>
<td>0.096</td>
</tr>
<tr>
<td>(0.018)</td>
<td>(0.020)</td>
<td>(0.016)</td>
<td>(0.022)</td>
<td>(0.032)</td>
<td></td>
</tr>
<tr>
<td>N</td>
<td>—</td>
<td>—</td>
<td>149</td>
<td>279</td>
<td>126</td>
</tr>
<tr>
<td>Test of Equality in TEs ((p)-Value)</td>
<td>—</td>
<td>—</td>
<td>0.049</td>
<td>0.090</td>
<td>0.056</td>
</tr>
<tr>
<td>Bandwidth</td>
<td>—</td>
<td>—</td>
<td>IKBW</td>
<td>20,000</td>
<td>10,000</td>
</tr>
</tbody>
</table>

\(^a\)Robust standard errors in parentheses, standard deviations in brackets. The unit of observation is a municipality. Each figure in columns (1)–(4) is from a separate local linear regression estimate with the specified bandwidth. The pre-treatment mean is the estimated value of the dependent variable for a municipality with 40,500 registered voters that uses paper ballot (based on the specification on column (1)). The IKBW column provides the Imbens and Kalyanaraman (2012) optimal bandwidth. Details on the dependent variables in the text. Estimates on Panel A (Panel B) use only municipalities where the adult illiteracy rate is above (below) 25.43%.
write a simple note.”27 The median illiteracy rate in the sample with a 20,000-voter bandwidth equals 25.4%. The average illiteracy rates (and standard deviations) in the above- and below-median subsample are 45.8% (s.d. = 11.4) and 16.1% (s.d. = 4.5), respectively.

Panel A and Panel B provide the results for the above-median and below-median illiteracy samples, respectively. The comparison of pre-treatment means already indicates that high-illiteracy municipalities have more residual voting under paper ballots than low-illiteracy ones. This is consistent with the less educated having difficulty voting on paper. The effect on the high-illiteracy sample is in the 15–18 p.p. range. The effect on municipalities with below-median illiteracy is in the 9–11 p.p. range. Table III also presents the p-values from testing the null hypothesis that the estimates in each sample are equal. This can be rejected at the 10% level in all cases, and at the 5% level when the IKBW is used.

2.3.4. Alternative Explanations

First, it must be noted that the at-large nature of elections implies that any two municipalities in the same state face the exact same choice of candidates. Hence, on average, the same choice of candidates is available in municipalities above and below the threshold, and candidate entry cannot account for the effects.28

Second, manipulation of the forcing variable (voter registration) or of the threshold itself can, in principle, invalidate the causal interpretation of the results. As discussed in Section 2.1, this is unlikely since the number of registered voters was measured before (1996) the announcement of the cutoff (1998). Second, there could be other treatments assigned by the same discontinuity. To the best of my knowledge, this is not the case. Moreover, the zero effects found in the 1994 and 2002 elections require an unknown confounding discontinuity to be present only in 1998. The Supplemental Material addresses both these issues in more detail, providing evidence of no manipulation of the forcing variable (and of the choice of cutoff), listing all known discontinuous population- or electorate-based assignments across Brazilian municipalities, and discussing why they cannot confound the results.

Finally, while this paper suggests that EV being easier to use (especially for the less educated) is the main mechanism behind the causal effect of the technology on valid votes, this argument is not essential to the paper’s main argument: EV-induced enfranchisement affected public policy in a manner con-

27The original Portuguese word for “note” is “bilhete,” which means a very short note (a written message shorter than a letter). The correlation between illiteracy rate and the share of adult population without completed 4th or 8th grade education is 0.96 and 0.85, respectively, and using these variables instead of illiteracy rates leads to similar results.

28The Supplemental Material reports that municipalities above and below the cutoff are not in systematically different states.
sistent with political economy models. However, the Supplemental Material discusses the possibility of electoral fraud as a (non-mutually exclusive) mechanism behind EV’s effects, and indicates there is little evidence of fraud.

In summary, this section provides evidence that EV promoted large increases in valid votes in municipalities with a large number of low educated citizens, and that these additional votes were disproportionately cast to left-wing parties. This combination of results suggests that the introduction of EV promoted the de facto enfranchisement of mostly less educated citizens. Its consequences on government spending, health service provision, and infant health are expounded in the next section.

3. POLITICAL PARTICIPATION, HEALTH CARE PROVISION, AND INFANT HEALTH

The 1998 election which provided the RDD in voting technology assignment between municipalities only involved the election of state (and federal) officials. Hence, the state becomes the relevant unit of observation to study the effects of EV on policy outcomes. 29 This section uses state-level panel data on election outcomes, legislative representation, fiscal spending, health care utilization, and newborn health to analyze these effects.

3.1. Background and Data

3.1.1. The Political Economy of Health Care in Brazil

Federal legislation mandates the provision of free-of-charge health care to all Brazilian citizens. This is achieved by a nationally coordinated public system named SUS (Sistema Único de Saúde), which is almost entirely funded by government tax revenues. Citizens are not required to pay health insurance premiums and do not incur out-of-pocket expenses when using the public system. The administration of public health care is decentralized: state governments are responsible for almost 30% of all expenditures on the system, they exert significant control over the allocation of federal funds, and are able to make “earmarked” health-related transfers to municipalities. Municipalities are responsible for 20% and the federal government for the remainder (Andrade and Lisboa (2002)). A coexisting private system of health insurance and care services requires users to pay either per-service fees or private insurance premiums. Alves and Timmins (2003) described the functioning of this dual system and provided evidence on the “social exclusion” from better-quality private services. Using 1998 household survey data, they found that 25% of the

29In particular, municipal elections occurred in 1996 (when virtually all municipalities used paper ballots) and 2000 (when only electronic voting was used). There is thus virtually no within-year variation in voting technology use across municipalities, making an analysis based on municipal data unattractive.
population, and mostly those with higher income and schooling, have access to private care.

The political consequences of this arrangement were aptly discussed in Mobarak, Rajkumar, and Cropper (2011), which noticed that the wealthier insured population “generally make little use of the SUS system” and that “the uninsured value SUS health services relatively more than the insured do.” These differences are also prevalent in nationally representative opinion surveys: 49% of Brazilians list “improving health care services” as a government priority (more than any other area). This number rises to 51% for those with family income below two minimum wages, but drops to 40% for those with income above ten.30

When these differing preferences are present, theories of redistributive politics predict that increased political participation (or enfranchisement) of the less educated should raise government spending on health care, as discussed in the Introduction. For instance, Mobarak, Rajkumar, and Cropper’s (2011) analysis of health care provision in Brazilian municipalities is informed by a probabilistic voting model which “predicts that an increase in the voting rate of the poor will increase public health care provision.” The results in this section test this prediction, with respect to the de facto enfranchisement promoted by EV.

While Mobarak, Rajkumar, and Cropper’s (2011) model could serve as motivating theory for the present results, this paper takes a more agnostic position on the political mechanism driving the results, especially on the issue of voters “affecting” or “electing” policies (Lee, Moretti, and Butler (2004)). The canonical median-voter-based redistributive politics model (Meltzer and Richard (1981)) would yield the tested prediction by having candidates committing to policy changes catered to the newly enfranchised. Citizen-candidate approaches that are based on politicians without credible commitment (Osborne and Slivinski (1996), Besley and Coate (1997)) would also generate this prediction through the election of additional officials with preferences closer to the newly enfranchised. These two channels are complementary and quantifying their distinct contributions is outside the scope of this paper for two reasons. First, it analyzes an event that changed both who votes and who gets elected, making it difficult to separate the effects of new politicians from incumbents.

This is based on a 2013 survey (CNI (2014)). Given that public health care spending has increased since 1998, as discussed in Section 3.3, such numbers were likely larger before EV’s introduction. The income categories reported by brackets based on minimum wage value, which was R$ 678/month (USD 300) at the time. Respondents could list up to three priorities. The other highly listed areas were “fighting crime” (31%), and “improving the quality of education” (28%). The income gradient in these responses is reversed in the case of education (31% of the richest strata list it as a priority) or much smaller in the case of fighting crime (27% of the richest strata list it as a priority).
changing their behavior. Second, this paper analyzes policy outcomes that are the product of bargaining between many politicians, making it difficult to assign responsibility for policy changes on specific individuals. The political relevance of health care is corroborated by other studies finding political factors affect spending on it by municipalities: Ferraz and Finan (2009) on the case of legislator wages, and Brollo and Troiano (2014) and Correa and Madeira (2014) in the case of female politicians.

Beyond the overall prominence of health care issues, there are two additional reasons (related to the feasibility of the empirical strategy) to focus on it instead of other policies that this enfranchisement episode could presumably affect. First, impacts on health care funding are feasible, since the provision of health services is a common demand of voters and within the reach of individual legislators. Second, health care funding can be affected in a short span of time, and so can its effects on newborn health. The identification strategy, described in the next section, relies on the sharp timing of EV’s phase-in.

Nationally representative opinion surveys suggest that increasing the supply of public health services is a likely demand of its users. Fifty-seven percent of Brazilian women state that “wait times/difficulty of being serviced” is the “main problem of the public health care system.” The second most listed problem was “lack of doctors” with 10% of responses. Moreover, 60% of women list “increasing the number of doctors” as one of the two main actions the government should take to improve public health services. There has also been rapid growth in the number of prenatal visits in the 1994–2006 period (described in more detail in Section 3.3). Although this may be partly due to increasing demand, its magnitude and the fact that it coincides with growth in health spending suggest that the number of prenatal visits in the public system was constrained by its supply.

Given the prominence of health care as public policy issues amongst the population, it is not surprising that a large portion of state legislature politics revolves around the issue, and that politicians are particularly interested in health issues. For example, 41 out of 94 of São Paulo state legislators list “health care” as an “area of expertise” in the legislature’s official webpage. State legislators can generate additional health care funding by either amending the budget or making official requests to the executive. A common objective of

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31This contrasts with papers that focus on changes in politician behavior when the electorate remains unchanged, which occurs in cases of political quotas, for example, Chattopadhyay and Duflo (2004), Pande (2003).

32This contrasts with papers that analyze the behavior of individual politicians, such as roll call voting (Lee, Moretti, and Butler (2004), Mian, Sufi, and Trebbi (2010)).

33These figures are from a 2011 survey (CNI (2012)). Given the co-existence of the public and private care system, it is possible to “increase the number of doctors” in the public system by having them re-allocate their time from the private one.

34The list relates to the 2011–2015 legislature. Other common areas are education (listed by 35 legislators) and public safety (18 legislators). Note that legislators can list multiple categories.
amendments and requests deals with the Family Health Program (*Programa Saúde da Família*), which allocates a team of health professionals (including a family doctor) to provide general practice services for small communities. The relatively low per-team cost of the program makes it easy for state legislators to amend the budget and target it geographically to its voter bases. As an example, the 94-member São Paulo state legislature recorded 107 separate legislator-proposed budget amendments citing the Family Health Program in the 2002 budget process.\(^{35}\)

Another source of additional health care funding can be direct requests, which can increase health care funding within a short time frame. Anecdotal evidence suggests that a request by a legislator can lead to a clinic providing services in less than a year. Ferraz and Finan (2009) also found that legislators can affect provision of health care quickly in the slightly different setting of Brazilian municipalities.

3.1.2. Newborn Health Status

Low birthweight (below 2500 g) is arguably the most common indicator of poor newborn health, used by numerous studies in clinical, epidemiological, and economic research. While infant health is an important public policy goal and research interest in itself, there is also considerable evidence that birth outcomes affect a wide range of adult health, human capital accumulation, and labor market productivity.\(^{36}\) It should be noted, however, that whether the EV-induced increases in health spending will affect adult life outcomes is beyond the scope of this paper, and birthweight should be interpreted as a noisy signal of infant health, which has other (unobservable) dimensions.

The medical literature indicates that prenatal visits (i.e., the interaction between a gestating woman and a doctor or other qualified health professional) are an important determinant of birthweight, mainly by positively affecting maternal behavior such as nutrition and avoidance of risk factors. Common infections that are easily treatable by medical professionals are also known to lead to premature birth and low birthweight. Kramer (1987) provided a comprehensive survey of the medical literature on birthweight determinants, while Currie and Gruber (1996a, 1996b) found that Medicaid expansions increase medical care utilization and birthweight in the United States.

3.1.3. State-Level Panel Data

The following sections use yearly data covering all Brazilian states during the 1994–2006 period that were constructed from several sources. Electoral data

\(^{35}\)Ames (2001) argued that targeting government expenditures at its bases through budget amendments is the main activity of Brazilian legislators.

\(^{36}\)Currie and Vogl (2013) surveyed this evidence from developing countries, and Almond and Currie (2011) focused on the effects of in utero health-related interventions on adult outcomes.
come from the same sources described in Section 2. Specifically, I use state-level valid votes and turnout data, as well as Power and Zucco’s (2009) index to estimate a seat-weighted index of a state assembly’s ideological position.37

State governments’ expenditures on health care and other categories of spending were obtained from the National Secretary of Treasury’s FINBRA database, which collects comparable accounting information for the states’ budget execution. This data source is discussed further in the Supplemental Material. The variable of interest is the amount spent in a legislative term (four-year period) by a state that is categorized as “health care and sanitation.” I will henceforth refer to this as “health care spending,” since sanitation is a negligible fraction of expenditures. The measures of health care utilization and outcomes come from the National System of Information on Live Births, which contains birth records collected from medical and official registries, such as birth certificates. A positive aspect of these data is their near universal coverage of Brazilian births.

I use three particular variables from the birth records to generate outcomes of interest: mother’s education, number of prenatal visits, and the birthweight of the newborn child. Information is coded categorically in the original data, and the specific variables computed for this study are the share of mothers that had seven or more prenatal visits and the share of births that are low-weight.38

Prenatal visits and birthweight are computed separately for mothers who completed primary schooling and those who have not. Having these two separate samples allows a test of differential effects by level of education. Primary education is the lowest level of schooling detailed in the data, so separately testing for effects on illiterate mothers, for example, is not possible. To facilitate referencing, I henceforth refer to these two groups as “educated” and “uneducated.” The sample means show the difference in health care access and outcomes: the average state has 53% of educated mothers reporting 7+ prenatal visits, and 6.3% of their births are low-weight. The respective figures for uneducated mothers are 33% and 7.7%.39

Finally, state socioeconomic variables (population, GDP, illiteracy rate, area, share of population below the poverty line,40 and Gini index for the income distribution) are from the Brazilian Statistical Agency (IBGE). As in Section 2, relevant sample moments are provided in the text or in the tables reporting estimates.

37Formally, let \( i_p \) be the ideology index for party \( p \) and \( w_{p,e} \) be the seat share of party \( p \) on the state assembly elected after election \( e \) in state \( i \). The computed variable is \( \sum_p (i_p \times w_{p,e}) \).

38The original data only report the share of mothers with 0, 1–6, 7+ prenatal visits for the entire sample. There are also limited data on large birthweights. Additional results based on the former two categories are reported in the Supplemental Material.

39Section 3.3 discusses if EV is associated with the number of births (in total and by mother’s education) and with the share of births that have valid prenatal visits and birthweight information, and concludes that these factors cannot account for the results.

40This variable uses the official poverty line defined by the Brazilian government, which is substantially higher than the one USD per day benchmark.
Figure 4.—Timeline of electronic voting’s introduction.

3.2. Estimation Framework

3.2.1. Specification

The empirical strategy exploits the pattern of EV across states and time. As discussed in Section 2.2, in the 1994 election only paper ballots were used, while in the 1998 election only municipalities with more than 40,500 registered voters used EV. In 2002 only the new technology was used. To facilitate referencing, I denote the 1994, 1998, and 2002 elections as the “paper-only,” “discontinuity,” and “electronic-only” elections, respectively. Figure 4 presents this timeline graphically, as well as the timing of different legislative terms, which are the four-year periods during which a state legislature holds office. Note that elections are always held in October, with the elected legislature taking office on the first day of the following year.

Let $S_i$ denote the share of voters in state $i$ that reside in municipalities above the cutoff in the 1998 election. Since 1994 was a paper-only election, $S_i$ equals the share of voters in state $i$ that changed from using paper ballots to EV between the 1994 and 1998 elections. Since the 2002 election was electronic-only, the change in the share of voters using EV between 1998 and 2002 elections equals $1 - S_i$. Formally, let $\text{electro}_{i,e}$ denote the share of voters using EV in state $i$ at an election held at year $e$. Hence:

$$\text{electro}_{i,1998} - \text{electro}_{i,1994} = S_i,$$

$$\text{electro}_{i,2002} - \text{electro}_{i,1998} = 1 - S_i.$$

This implies that a time-invariant cross-sectional variable, $S_i$, is positively related to changes in EV use in a period and negatively related to it in the following period. Hence, if a particular outcome of interest (such as health care funding, utilization, and outcomes) follows this same pattern, this is interpreted as evidence of a likely causal effect of EV. The essence of the argument is that it is unlikely that an omitted variable (or measurement error) would follow this same pattern.

41 For example, the election in October 1994 selected all the legislators that held office from January 1st, 1995 to December 31st, 1998. The dates are the same for all states, and all seats of a legislature are elected and inaugurated simultaneously.
Take expenditures on health care as an example of an outcome of interest. Denote $H_{ie}$ as health care spending by the legislature selected at election year $e$ at state $i$ (e.g., $H_{i1998}$ is the spending for the 1999–2002 legislative term). The next section shows that $S_{i}$ and $(H_{i1998} - H_{i1994})$ are positively correlated, and $S_{i}$ and $(H_{i2002} - H_{i1998})$ are negatively correlated. Given equations (4) and (5), in the first period where $S_{i}$ is proportional to larger changes in the use of EV, it predicts larger changes in health care spending. In the period where the same cross-sectional variable $S_{i}$ is inversely proportional to changes in EV usage, it predicts smaller changes in spending.

Although the original data are at the yearly level, all the estimations are carried out using legislative term (four-year) averages. Let $y_{ie}$ denote an outcome for state $i$ at the legislative term elected at year $e$ (e.g., $y_{i1998}$ is the outcome observed during the 1999–2002 legislative term). The estimated equations are

\[
\Delta y_{i98} = \alpha_{98} + \theta_{98} S_{i} + \beta_{98} X_{i02} + \epsilon_{i98},
\]

\[
\Delta y_{i02} = \alpha_{02} + \theta_{02} S_{i} + \beta_{02} X_{i02} + \epsilon_{i02},
\]

where $\Delta$ is the lag operator defined at the term (not year) level, so that $\Delta y_{i98} = y_{i98} - y_{i94}$, for example. $X_{ie}$ is a vector of possible controls. The parameter $\theta_{98}$ measures the effect of $S_{i}$ on the change in average $y_{ie}$ between the 1995–1998 and 1999–2002 legislative terms. Equation (4) indicates that $S_{i}$ equals the change in EV use in the elections that selected these legislatures (1994 and 1998, respectively). Hence, $\theta_{98}$ is an estimate of the impact of EV on $y_{ie}$. Similarly, equation (5) implies that $\theta_{02}$ measures the effect of the opposite of EV use on $y_{ie}$. The change in EV use between the elections that selected the 1999–2002 and the 2003–2006 legislatures (1998 and 2002, respectively) equals $(1 - S_{i})$. Since $\theta_{02}$ captures the effect of $S_{i}$ on the change between these two legislatures, $-\theta_{02}$ is the effect of EV on the outcome $y_{ie}$. I take evidence that $\theta_{98} = -\theta_{02} \equiv \theta$ as an indication that $\theta$ measures the causal effect of EV. Throughout the analysis, equations (6) and (7) are estimated separately, so that no particular relationship between the parameters $\theta_{98}$ and $\theta_{02}$ (or any other) is imposed by the estimation. Finally, note that an omitted variable that positively affects $y$ and grows faster in high-$S_{i}$ states in the entire period would upward bias $\theta_{98}$ and downward bias $\theta_{02}$.

The joint estimation of equations (6) and (7) under the restriction that $\theta_{98} = -\theta_{02} \equiv \theta$ can be carried out by pooling data from the three electoral terms and estimating

\[
y_{ie} = \alpha_{e} + \theta S_{i} \cdot \text{Term}_{e}^{98} + \gamma_{i} + \beta X_{ie} + \epsilon_{ie}, \]

42 This specification is equivalent to an equation in levels including electoral term and state fixed effects that captures the effect of any unobserved state-specific, time-invariant variable and of national-level factors that vary through time.
where \( Term^{98}_e \) is a dummy indicating the legislative term elected in 1998. Since \( S_i \cdot Term^{98}_e \) is collinear with EV use (\( electro_i \)) when state and time effects are present, \( \theta \) captures the effect of EV on \( y \). The estimated \( \theta \) is thus equal to \( (\theta^{98} - \theta^{02})/2 \), apart from small differences generated by the covariates \( X_{ie} \), since \( \beta_1 = \beta_2 = \beta \) is also imposed. All estimations control for a set of region-time dummies, which absorb geographically restricted shocks affecting the outcomes of interest. Unless otherwise noted, no other controls are included.\(^{43}\)

The discussion so far applies to the states that followed the discontinuous assignment rule in the 1998 election. As discussed in Section 2.1, four states implemented EV in all its municipalities in that year. They are included in the sample as having \( S_i = 1 \), so that the equivalence between changes in EV and \( S_i \) on equations (4)–(5) remains. Henceforth, I refer to “share of electorate above the threshold” and \( S_i \) interchangeably. Hence, estimation of (6) and (7) pools both variation from the unusual pattern generated by the discontinuous assignment and also from what would amount to a standard differences-in-differences approach. Since the four states did not comply with the rule for distinct reasons (discussed in Section 2.1), the results from specifications that exclude them can be seen as a robustness check for the estimates.

3.2.2. Identification

The main identification assumption behind this empirical approach is that no omitted variable affecting the outcomes follows the particular pattern of electronic voting’s introduction.\(^{44}\) To confound the interpretation of the results as the effect of EV, an omitted variable which (without loss of generality) positively affects the outcome needs to grow faster in states with high \( S_i \) in the 1994–1998 period and switch to growing slower in high-\( S_i \) states in the 1998–2002 period. One possible way this can occur is through a mean-reversing shock affecting high-\( S_i \) states during the 1998 electoral term. Note, however, that mean reversion in general cannot explain the result: the shock must disproportionately affect high-\( S_i \) states with a timing that matches the schedule of elections.

Direct evidence regarding the existence of such confounding omitted variable or mean-reversing shock cannot ultimately be provided, and to a certain extent this paper’s argument relies on the sign-switch pattern above being “so unusual that no omitted variables can follow it.” However, Section 3.3 specifically addresses possible threats to the identification strategy. It discusses possible sources of mean-reversing shocks and provides five sets of additional tests regarding the identification assumption.

\(^{43}\)If covariates \( (X_{ie}) \) are not included, or if they are fully interacted with time dummies (as the region-time effects), then \( \theta = (\theta^{98} - \theta^{02})/2 \) holds exactly. Brazilian states are grouped into five official regions based on their natural, demographic, and economic characteristics. States within a region are contiguous.

\(^{44}\)Since EV’s introduction is clearly a function of predetermined voter registration across municipalities, reverse causality (e.g., health spending leading to more EV use) is not an issue.
First, it breaks down the relationship between $S_i$ and the outcomes on a yearly basis, showing that effects occur usually in the year immediately after elections. This implies that possible confounding shocks (and their mean reversion) must follow quite specific, and sharp, timing. Second, it estimates negligible (placebo) effects on variables not expected to be affected by EV, such as general economic conditions and birth outcomes for educated mothers, as well as spending by municipal governments, which were exposed to EV under different timing but should also respond to shocks to health care demand. This highlights how unusual the sign-switch pattern is and addresses concerns about specific sources of mean-reversing shocks. Third, it focuses on lagged and lead outcomes and finds that the share of voters above the cutoff is orthogonal to changes in outcomes in the periods when it is not associated with EV changes in voting technology, addressing issues of pre-trends. Fourth, the effects are re-estimated with controls for possible (nonlinear) time trends interacted with state characteristics, suggesting that the results are not driven by another shock with heterogeneous effect across states. Finally, the Supplemental Material reports effects of electronic voting instrumented by a variable that focuses on the distribution of municipalities closer to the cutoff, removing variation driven by large municipalities and addressing concerns that shocks specific to them drive the results.

3.2.3. Inference

The statistical inference aims to account for both possible serial correlation and the relatively small number of cross-sectional units (27 states). Following recommendations in both Bertrand, Duflo, and Mullainathan (2004) and Donald and Lang (2007), the time dimension of the data is aggregated as much as possible by collapsing it to electoral term (four-year) averages. The estimation of (6) and (7) is done by taking first-differences and not the (equivalent) fixed effects regression, using a single cross-section of first-differences. In this case, the heteroscedasticity-robust standard errors are equivalent to those clustered at the state level. Donald and Lang (2007) discussed why this is an appropriate choice.

The estimation of equation (8) involves three periods and cannot be collapsed into a cross-sectional regression, and hence uses standard errors clustered at the state level. Simulations in Bertrand, Duflo, and Mullainathan (2004) and Cameron, Gelbach, and Miller (2008) indicate that clustered standard errors lead to relatively appropriate sized inference when there are 27 clusters and small number of periods. However, to further address the small cluster issue, $p$-values based on Cameron, Gelbach, and Miller (2008) cluster-robust wild-bootstrap are also provided in all regressions.

The relatively small number of states also raises the possibility of outliers driving some results. This issue is addressed in the Supplemental Material by showing the robustness to leave-one-state-out estimates and also by provid-
3.3. Results

The geographical distribution of the share of voters living in municipalities above the 40,500-voter threshold ($S_i$) is depicted in Figure 5, which also shows the location of municipalities that used electronic voting in the 1998 election (except in the cases of states not following the discontinuity). The variable has a wide range, from 0.147 to 1, with a mean of 0.52 and a standard deviation of 0.26.

3.3.1. Main Results

Columns (1) and (2) in the first row of Panel A on Table IV present the estimates from equations (6) and (7) using the valid votes (as a share of turnout) as the dependent variable. This variable was the main outcome of interest in
the RDD studied in Section 2, allowing a comparison between the two empirical strategies. The estimated $\theta^{08}$ and $\theta^{02}$ are 9.2 p.p. and $-11.1$ p.p., respectively. This indicates that states with a larger share of their population living above the cutoff experienced faster growth in valid votes in the 1994 and 1998 elections, but slower growth between the 1998 and 2002 elections. Column (3) presents an estimate of the average implied effect of EV by pro-

<table>
<thead>
<tr>
<th>Parameter:</th>
<th>$\theta^{08}$</th>
<th>$\theta^{02}$</th>
<th>Linear Combinations</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>1994–1998</td>
<td>1998–2002</td>
<td>$(\theta^{08} - \theta^{02})/2$</td>
</tr>
<tr>
<td>Sample Avg.</td>
<td>(Paper–Disc.)</td>
<td>(Disc.–Electr.)</td>
<td>$(\theta^{08} + \theta^{02})/2$</td>
</tr>
<tr>
<td>Panel A: Electoral Outcomes</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Valid Votes/ Turnout</td>
<td>0.829</td>
<td>0.092</td>
<td>-0.111</td>
</tr>
<tr>
<td></td>
<td>[0.112]</td>
<td>(0.033)</td>
<td>(0.010)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(0.102)</td>
<td>(0.002)</td>
</tr>
<tr>
<td>Seat-Weighted Policy Position</td>
<td>4.623</td>
<td>-0.112</td>
<td>0.299</td>
</tr>
<tr>
<td></td>
<td>[0.601]</td>
<td>(0.641)</td>
<td>(0.167)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(0.842)</td>
<td>(0.154)</td>
</tr>
<tr>
<td>Panel B: Fiscal Outcomes (Health Care Spending)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>log(Total Spending)</td>
<td>—</td>
<td>-0.004</td>
<td>-0.257</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(0.093)</td>
<td>(0.156)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(0.946)</td>
<td>(0.274)</td>
</tr>
<tr>
<td>Share of Spending in Health Care</td>
<td>0.099</td>
<td>0.039</td>
<td>-0.029</td>
</tr>
<tr>
<td></td>
<td>[0.037]</td>
<td>(0.017)</td>
<td>(0.013)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(0.104)</td>
<td>(0.044)</td>
</tr>
<tr>
<td>log(Health Spending p.c.)</td>
<td>—</td>
<td>0.428</td>
<td>-0.677</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(0.264)</td>
<td>(0.262)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(0.200)</td>
<td>(0.034)</td>
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<tr>
<td>Panel C: Birth Outcomes (Mothers Without Primary Schooling)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Share With 7+ Visits</td>
<td>0.362</td>
<td>0.122</td>
<td>-0.023</td>
</tr>
<tr>
<td></td>
<td>[0.123]</td>
<td>(0.065)</td>
<td>(0.033)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(0.154)</td>
<td>(0.558)</td>
</tr>
<tr>
<td>Share With Low-Weight Births ($\times 100$)</td>
<td>7.721</td>
<td>-0.370</td>
<td>0.528</td>
</tr>
<tr>
<td></td>
<td>[1.110]</td>
<td>(0.304)</td>
<td>(0.269)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(0.266)</td>
<td>(0.104)</td>
</tr>
<tr>
<td>N (State-Terms)</td>
<td>—</td>
<td>54</td>
<td>54</td>
</tr>
<tr>
<td>N (States/First-Diffs)</td>
<td>—</td>
<td>27</td>
<td>27</td>
</tr>
</tbody>
</table>

*Standard errors clustered at the state level in parentheses. Standard deviations in brackets. $p$-values based on Cameron, Gelbach, and Miller (2008) cluster-robust wild-bootstrap in curly brackets—(1). The unit of observation is a state-electoral term. Each row reports the estimation of equations (6) and (7) using the specified dependent variable. Each figure in columns (1) and (2) is from a separate regression, providing the coefficient on the share of voters living above the cutoff $(S_i)$ on the 1998 and 2002 first-differences, respectively ($\theta^{08}$ and $\theta^{02}$). Columns (3) and (4) report the specified linear combination of these parameters. Region-time effects are included.
viding θ = (θ₀² − θ₀⁸)/2 estimated through equation (8). Column (4) presents an estimate of (θ₀⁸ + θ₀²)/2. This combination of parameters provides a test for θ₀⁸ = −θ₀². This null hypothesis cannot be rejected at usual levels of significance.⁴⁵

As previously discussed, this “sign-switch” pattern is interpreted as evidence of the effect of EV. The estimates indicate that the effect of a change from having the entire electorate of a state using paper ballots switch to EV is approximately 10 p.p., which is remarkably close to the RDD estimates.

The second row of Panel A (Table IV) shows that the positions of elected parties to state legislatures also follow the sign-switch pattern, which is consistent with EV leading to the election of left-wing parties. As in the case of valid votes, the implied effect is remarkably close to those from the RDD. However, the estimated effect is imprecisely estimated. As discussed in Section 2, party positions are based on behavior at the federal congress, and hence are a noisy proxy for their positions at the state level. The Supplemental Material repeats this estimation for each of the main parties in Brazil, and finds that two parties that benefited the most from EV have left-wing positions.⁴⁶

The first row of Panel B (Table IV) indicates that the same sign-switch pattern is observed for the share of state budgets spent on health care. The magnitudes of the estimated θ₀⁸ and θ₀² are close (0.39 and −0.29), and one cannot reject the hypothesis that θ₀⁸ = −θ₀² (column (4)). The average effect (column (3)) implies that a full switch from using only paper ballots to the complete use of EV generates an increase of 3.4 p.p. on the share of state expenditure destined to health care. This is also a 34% increase with respect to the sample mean. While θ₀² is significant at the 5% level under both the standard t-test and the bootstrapped p-values, θ₀⁸ is significant at the 5% level in the t-test but only marginally significant when using the bootstraps. However, pooling the data (column (3)) adds substantial precision to the estimate, yielding p-values below 0.1% in both tests. This substantial gain in precision when the data are pooled is also observed for other outcomes, and highlights that the main focus of columns (1) and (2) is to check whether θ₀⁸ and −θ₀² are similar. The graphical analysis in the Supplemental Material illustrates why the pooled data yield much more precise estimates.

Although health care spending is the most likely (and perhaps only) policy expected to be affected by EV, it is still of interest to report a “family-wise” p-value which accounts for the existence of other categories of spending for which effects can be estimated. The Supplemental Material discusses the classification of spending and indicates there are at most seven other categories

⁴⁵The sum of estimates is divided by 2 so it is comparable in magnitude to the effects on column (3), and also for computational reasons: it can be obtained through a single pooled regression analogous to equation (8), and its bootstrapped p-values generated from this equation.
⁴⁶Specifically, the Workers’ Party (local acronym PT) and the Democratic Labor Party (PDT).
available for all years of the sample. This implies that a (conservative) Bonferroni correction would involve multiplying the \( p \)-value by eight.\(^{47} \) The \( t \)-statistic implied by column (3) is 4.29, which yields a \( p \)-value of 0.01% in an independent test (this number is supported by the bootstraps). Hence, the correction indicates that under the null that EV has no effects on any category of spending, an effect of the magnitude found for health care spending would be found by chance for one category when testing amongst the eight variables with a probability below 0.8%. Furthermore, the correction gives an upper bound on the family-wise \( p \)-value, as it makes no assumption on the dependence between tests.

Note that these estimates should be interpreted as the effect gradually occurring over an eight-year period, and not as an immediate effect. The average state has half of its electorate residing in municipalities above the cutoff (mean \( S_i \) is 0.52). So the estimates imply that it redirects 1.7% of its budget to health care in a four-year legislative term, and then another 1.7% in the next four-year period. While this increase is relatively large, it must be put into the perspective of the relatively low levels of health care expenditures, accounting for only 7.9% of state government budgets in the 1995–1998 legislative term. This figure grew to 12.3% in the 2003–2006 legislative term. Hence, the implied effect of EV is equivalent to approximately three quarters of the increase in health care budget share growth in the period.\(^{48} \)

Panel B also indicates that a similar sign-switch pattern is found when using (log) spending per capita in health care. Total log per capita state spending, however, does not follow the sign-switch pattern, suggesting no significant effects of EV on government size.\(^{49} \) This indicates that the additional spending in health care occurs at the expense of other categories of spending. The Supplemental Material provides estimates using outcomes in reais per capita (i.e., without logs) and reaches similar conclusions.

\(^{47}\)The correction is based on Boole’s inequality, which implies that the probability of at least one test rejecting a true null hypothesis with probability \( \alpha \) in a family of \( n \) tests is at most \( n \alpha \) (regardless of dependence across tests).

\(^{48}\)Large political responsiveness of health spending has been found for the United States by Miller (2008). It reported that U.S. states observed a 36% increase in health-related spending in the year immediately following women’s suffrage, with growing effects leading to an 81% increase after five years.

\(^{49}\)The enfranchisement of less educated (poorer) voters could boost the size of government through increased direct redistribution. Husted and Kenny (1997) found evidence along these lines in the case of U.S. states. Brazilian state governments, however, have less ability to engage in direct redistribution, since most income transfer programs such as social security and conditional cash transfers are determined at the federal level (with their operation often decentralized to the municipal level). Moreover, states’ main source of tax revenue is a value-added tax on goods and services, which does not lend itself as easily to progressive taxation, unlike income taxes (which are set and collected by the federal government). Finally, the empirical strategy relies on relatively sharp timing, and would not capture an effect on government size that takes more than four years to occur.
The Supplemental Material provides estimates of equations (6) and (7) using all available categories of spending. The only detectable effect of EV on these categories is a reduction on the “administration and planning” spending, which is large enough to counteract the increased health care spending. This category consists mostly of the overhead of government operation (and accounts for 18% of budgets). Given this definition, a reduction in this category would be consistent with governments becoming more efficient, but it may also mask reductions in government projects or programs that are not assigned to other categories. This highlights, given the limitations in spending data, that a more detailed welfare analysis of the policy changes induced by EV cannot be provided, and that the welfare gains from health services utilization and outcomes may be offset by losses in other (unobserved) areas.

While the yearly state-level fiscal data do not detail the nature of the additional health care spending, other sources may shed some light on this issue. As discussed in Section 3.1, a natural candidate would be the Family Health Program. Its coverage grew quickly in the period following EV’s introduction. One of the stated goals of the family health program is facilitating access to basic services that would otherwise only be available at hospitals and clinics. One such service is prenatal visits, which can be done at the community health outpost or with house visits. As registration in the program grew, so did the use of these services: the number of visits by pregnant women to Family Health Program professionals grew by a factor of 4.5 in the eight years following the first use of EV in state elections. Moreover, the improved prenatal care by the Family Health Program has been associated with reduced child mortality (Rocha and Soares (2010)).

While the narrative above suggests that EV may have played a role in the expansion of the Family Health Program and its improved access to health services and outcomes, yearly state-level data on the program are not comprehensive enough to allow for an econometric quantification. However, it is possible to estimate the effects on the number of prenatal visits and birthweight.\textsuperscript{50}

Panel C of Table IV provides the results for birth outcomes of uneducated mothers. Its first row reports results for the share of mothers with more than seven prenatal visits while the second deals with low-weight births. For both outcomes, the signs of $\theta_{98}$ and $\theta_{02}$ are the opposite. Their magnitudes in the

\textsuperscript{50}To assess the possibility that the composition of births is confounding the results, possible effects of EV were estimated on (i) the total number of births (including those where mother’s education is missing), (ii) the share of births by educated and uneducated mothers, (iii) the share of births for which visits and birthweight data are not missing (for each education group), and (iv) the share of male births. The effects are negligible and statistically insignificant, indicating that the composition of births is unlikely to affect the results. To further account for this possibility and increase the estimates’ precision, the reported results on prenatal visits and birthweight include the respective variable (iii) as a control (e.g., the effects on birthweight by uneducated mothers include the share of births by uneducated mothers for which weight data are not missing as a control); not including these controls led to similar estimates.
case of prenatal visits are not too similar, but one cannot reject the hypothesis of \( \theta^{98} = -\theta^{02} \) in both cases. The implied average effect implies that a full switch from paper to EV leads to a 7 p.p. increase in the share of uneducated mothers with more than seven prenatal visits, and a 0.5 p.p. reduction in the probability of a low-weight birth. While the effect on birthweight is significant at the 5\% level (using the pooled data), the effect on visits is more imprecisely estimated. It should be noted, however, that the addition of controls (discussed in the next section) will add precision to the estimates, and the magnitude of the point estimate is substantial, especially compared to the placebo estimate on mothers who completed primary schooling, which is also described in the next section.\(^{51}\)

Both these effects are large in magnitude, and roughly correspond to a third of the gap between average outcomes of educated and uneducated mothers. As with the previous estimates, it should be noted that these effects probably occur gradually over the course of eight years. Another way to summarize the size of effects is to calculate their value as a share of sample averages (provided in Table IV). The switch from paper to EV increases the share of valid votes by 12.3\%, spending in health care by 34.3\%, prenatal visits by 19\%, and lowers the prevalence of low-weight births by 6.8\%. While the overall pattern of results might suggest a causal chain linking these effects, one must be careful in using them to compare implied elasticities. This would require imposing strict exclusion restrictions that may not be warranted in some cases, such as assuming that EV did not induce other (unobserved, given data constraints) policy changes affecting prenatal care and newborn health.

3.3.2. Assessing the Identification Strategy

As discussed in Section 3.2, to confound the interpretation of the results as the effect of EV, an omitted variable which (without loss of generality) positively affects the outcome needs to grow faster in states with high \( S_i \) in the 1994–1998 period and sharply switch to growing slower in high-\( S_i \) states in the 1998–2002 period. This can occur through a mean-reversing shock affecting high-\( S_i \) states during the 1998 electoral term. Note, however, that mean reversion in general cannot explain the result: the shock must affect a specific type of states with a timing that matches the schedule of elections. To further exploit the timing issue, the Supplemental Material breaks down the relationship between \( S_i \) and the outcomes on a year-by-year basis, showing that effects occur usually in the year immediately after elections. Hence, an omitted shock would need to occur primarily in 1999 and start its reversion only in 2003 to drive the results.

\(^{51}\)The Supplemental Material reports the analogous estimates for the share of uneducated mothers with 0, and 1–6 prenatal visits (the categories in the original data). The results suggest the effects occur at the intensive margin: there is no effect on the share with 0 visits, and a negative one on the 1–6 visit category.
The existence of such a confounding omitted variable or mean-reversing shock following such timing is unlikely a priori. The 1998 election and its aftermath was not characterized by substantial changes in policymaking (especially of the temporary kind that could follow the pattern described above). At the federal level, it was characterized by the landslide re-election of President Fernando Henrique Cardoso, the maintenance of his coalition’s control of congress, and a continuation of most of his policies.52

The discussion above notwithstanding, this section addresses the possibility of omitted variables or shocks affecting the results through four sets of additional tests. First, I estimate negligible “placebo effects” on variables that are not expected to be affected by EV, such as general economic conditions, health spending by municipal governments, and health outcomes from educated mothers. This not only highlights how unusual the sign-switch pattern is (as other variables do not follow it), but also indicates that if an omitted variable or mean-reversing shock is driving the results, it is also not affecting these variables, making it a less likely concern.

Second, I focus on lagged and lead outcomes and find that the share of voters above the cutoff is orthogonal to changes in outcomes in the periods when it is not associated with changes in voting technology. This is analogous to pre-(and post-) trend analysis in difference-in-differences estimators.

Third, the effects are re-estimated with additional controls, which include possible (nonlinear) time trends interacted with state characteristics (i.e., allowing for a sign-switch pattern across other state characteristics). This addresses concerns of shocks with heterogeneous effects across states (and this heterogeneity being correlated with $S_i$) driving the results.

Fourth, the Supplemental Material reports effects of electronic voting instrumented by a variable that focuses on the distribution of municipalities closer to the cutoff, removing variation driven by municipalities far from the cutoff, such as large cities. This suggests that the results are driven by the share of population using electronic voting, and not overall population distribution or presence of metropolitan areas.

Panel A of Table V presents the first set of (placebo) tests. It follows the same pattern as Table IV, focusing on the possibility of an outcome following the sign-switch pattern. The first row of Panel A reports results for the budget share spent on health care by municipalities in each state. Municipal elections are staggered with state elections, occurring in even years when the latter do not take place (e.g., 1996, 2000, and 2004). Hence, there is no reason why municipal budgets should track the pattern of EV use in state elections.53

52Luis Inácio Lula da Silva, from the Worker’s Party, was elected president in 2002, with a platform based on the continuation of existent policies, perhaps with larger focus on combating poverty.

53The variable is constructed from municipal budgets from the same source used for state spending, and aggregating it to the state level. Municipalities also make sizable spending in health care (of a magnitude roughly two-thirds of state spending), as discussed in Section 3.1.
### TABLE V
**PLACEBO TESTS**

<table>
<thead>
<tr>
<th>Parameter:</th>
<th>Linear Combinations</th>
<th>Sample Avg.</th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>Share of Spending</td>
<td>$\theta^{08}$</td>
<td>$\theta^{02}$</td>
<td>$(\theta^{08} - \theta^{02})/2$</td>
<td>$(\theta^{08} + \theta^{02})/2$</td>
</tr>
<tr>
<td>in Health Care—Municipalities</td>
<td>0.096</td>
<td>-0.011</td>
<td>-0.016</td>
<td>0.003</td>
</tr>
<tr>
<td>log(Population)</td>
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<td>0.059</td>
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</tr>
<tr>
<td>log(GDP)</td>
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<td>0.007</td>
<td>0.096</td>
<td>-0.051</td>
</tr>
<tr>
<td>Poverty Rate</td>
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<td>0.080</td>
<td>0.105</td>
<td>-0.012</td>
</tr>
<tr>
<td>Gini Index</td>
<td>0.569</td>
<td>0.034</td>
<td>0.042</td>
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### Panel A: Effects on Covariates

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<tr>
<th>Parameter:</th>
<th>Linear Combinations</th>
<th>Sample Avg.</th>
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<th></th>
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<tbody>
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<td>Share With 7+ Visits in Health Care—Municipalities</td>
<td>0.096</td>
<td>-0.011</td>
<td>-0.016</td>
<td>0.003</td>
</tr>
<tr>
<td>Poverty Rate</td>
<td>0.397</td>
<td>0.080</td>
<td>0.105</td>
<td>-0.012</td>
</tr>
<tr>
<td>Gini Index</td>
<td>0.569</td>
<td>0.034</td>
<td>0.042</td>
<td>-0.004</td>
</tr>
</tbody>
</table>

### Panel B: Birth Outcomes for Mothers Who Completed Primary Schooling

<table>
<thead>
<tr>
<th>Parameter:</th>
<th>Linear Combinations</th>
<th>Sample Avg.</th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>Share With Low-Weight Births (×100)</td>
<td>6.261</td>
<td>0.391</td>
<td>-0.023</td>
<td>0.196</td>
</tr>
</tbody>
</table>

### Panel C: Pre- and Post-Trend Analysis: “Effect” in Periods Without Change in Voting Technology

<table>
<thead>
<tr>
<th>Parameter:</th>
<th>Linear Combinations</th>
<th>Sample Avg.</th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>Share of Spending in Health Care</td>
<td>—</td>
<td>-0.006</td>
<td>-0.005</td>
<td>-0.0003</td>
</tr>
<tr>
<td>$N$ (State-Terms)</td>
<td>54</td>
<td>54</td>
<td>—</td>
<td>—</td>
</tr>
<tr>
<td>$N$ (States/First-Diffs)</td>
<td>27</td>
<td>27</td>
<td>—</td>
<td>—</td>
</tr>
</tbody>
</table>

*aSee Table IV notes.*
consistent with this notion, the variable does not follow the sign-switch pattern. The estimated $\theta^{98}$ and $\theta^{02}$ have the same sign and magnitude and furthermore are small and statistically indistinct from zero, indicating that municipal health spending is uncorrelated with EV phase-in.

This result addresses the possibility that the sign-switch pattern in state health spending was driven by the demand for public health care or by a nationwide health program that was rolled out from larger to smaller municipalities (with a timing that matched elections and EV’s introduction). If either were the case, municipal health care spending should also track the same sign-switch pattern. This is consistent with the fact that, to the best of my knowledge, there is no shock to health care demand or program roll-out that fits the pattern across time and areas of EV’s introduction.

The remaining rows of Panel A (Table V) report that state-level (log) GDP, (log) population, inequality, and the poverty rate do not follow the sign-switch pattern. The point estimates indicate that all these variables seem to grow relatively faster in states with high shares of population above the cutoff in both periods (i.e., no sign switch). In all cases, the effect of EV being zero cannot be rejected (column (3)). These results not only indicate that the results of the previous section cannot be explained by these covariates, but also exemplify that the sign-switch pattern is unusual and suggests possible determinants of health care spending are unlikely to follow it.

To illustrate the relevance of these results, take a possible source of omitted shocks: the 1999 currency devaluation. To the extent that this sizable macroeconomic effect affected states with high $S_i$ disproportionately, it could confound the results. If that was the case, one would expect Table V to show a sign-switch in general economic conditions. Moreover, if this shock affected state budgets and led to more health spending, it did so for a reason not affecting municipal budgets in the same manner. Taking these factors into account makes it less likely that such macroeconomic shock is affecting the results.

Panel B of Table V repeats the exercise from Panel C of Table IV, but focusing on prenatal visits and low-weight births of educated mothers (while the latter focused only on uneducated ones). The sign-switch pattern is not present. The implied average effect (column (3)) for prenatal visits is less than half of the effect for uneducated mothers, and the effect for low-weight births has a different sign (and almost a third of the magnitude). Neither of these effects is significantly different from zero at the usual levels. These results indicate that EV affected service utilization and infant health only for uneducated mothers, and that not only would a confounding omitted variable need to follow an unusual pattern, but it would need to be specific to uneducated mothers. The results also fit the previous discussion on the political economy of Brazilian health care.

The second set of placebo tests involves assessing if the share of voters above the cutoff for EV use in 1998 ($S_i$) predicts changes in health care spending in periods where the voting technology does not change, so that the previous
results are not driven by preexisting trends that are correlated with $S_i$. Panel C of Table V presents the effect of $S_i$ on changes between the 1990 and 1994 electoral terms (both only used paper ballots) and the 2002 and 2006 election (both only used EV). The estimated effect on both cases is close to zero (and statistically insignificant). This indicates that, apart from the periods where it predicts EV use, the share of voters living above the cutoff is orthogonal to growth in health care spending.\(^{54}\)

The third set of additional tests involves adding controls to the estimation of equation (8), which is reported without controls on column (3) of Table IV. It focuses on the four main outcomes of the paper (valid votes, share of spending in health, prenatal visits, and prevalence of low birthweight). Column (1) adds additional controls (population, GDP, the poverty rate, and the Gini index of income distribution). The point estimates are almost unchanged and become more precise in some cases (being significant at the 5% level in all of them).

More relevant to the issue of omitted variables and shocks, columns (2)–(9) add to the specification a different measure of state heterogeneity, fully interacted with time dummies. This captures any (nonlinear) trend, including a “sign-switch,” correlated with such variable. For example, if GDP is correlated with $S_i$, and some temporary shock to richer states is what is actually driving the previous results, then one would expect the results in column (2) to differ substantially from those in column (1). The variables used are (the 1995–1998 average of) the same characteristics analyzed in Table IV (and added as controls in column (1)), as well as the illiteracy rate (measured in the 1991 Census), state area in square kilometers, and number of municipalities per capita, since these are related with distribution of population. Column (9) includes both population and area interacted with time dummies, hence flexibly controlling for different trends by population density. Table VI also indicates the correlation between the variable interacted with time dummies and $S_i$.

The general picture from Table VI is that the results are robust to the inclusion of such trends; the point estimates for all four outcomes are fairly stable. In some cases, the precision of some outcomes is affected, especially when the variable interacted with time dummies is strongly correlated with $S_i$ (likely due to issues of multicollinearity).\(^{55}\) This indicates that, even controlling for the possibility of mean-reversing shocks correlated with multiple dimensions of state heterogeneity, the estimated effect of EV remains. This further suggests that such shocks are not confounding the main results.

\(^{54}\)Birth records data are not available for the 1991–1994 electoral term or the entire 2007–2010 electoral term. Hence, it is not possible to perform this test on prenatal visits and birthweight outcomes.

\(^{55}\)Even the case that affects the results the most—interaction of trends by state poverty rate—seems to keep most point estimates unaffected and making some less precise (e.g., the effects on visits larger and on birthweight smaller), but with the general overall picture of results being similar.
### Panel A: Valid Votes and Health Care Spending

<table>
<thead>
<tr>
<th></th>
<th>(1)</th>
<th>(2)</th>
<th>(3)</th>
<th>(4)</th>
<th>(5)</th>
<th>(6)</th>
<th>(7)</th>
<th>(8)</th>
<th>(9)</th>
<th>(10)</th>
<th>(11)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Valid Votes/Turnout</td>
<td>0.103</td>
<td>0.103</td>
<td>0.111</td>
<td>0.102</td>
<td>0.111</td>
<td>0.109</td>
<td>0.102</td>
<td>0.099</td>
<td>0.105</td>
<td>0.123</td>
<td>0.102</td>
</tr>
<tr>
<td></td>
<td>(0.016)</td>
<td>(0.016)</td>
<td>(0.013)</td>
<td>(0.020)</td>
<td>(0.017)</td>
<td>(0.015)</td>
<td>(0.017)</td>
<td>(0.022)</td>
<td>(0.014)</td>
<td>(0.014)</td>
<td>(0.017)</td>
</tr>
<tr>
<td>Share of Spending</td>
<td>0.031</td>
<td>0.033</td>
<td>0.037</td>
<td>0.045</td>
<td>0.030</td>
<td>0.031</td>
<td>0.030</td>
<td>0.039</td>
<td>0.030</td>
<td>0.028</td>
<td>0.034</td>
</tr>
<tr>
<td>in Health Care</td>
<td>(0.026)</td>
<td>(0.009)</td>
<td>(0.007)</td>
<td>(0.006)</td>
<td>(0.012)</td>
<td>(0.009)</td>
<td>(0.010)</td>
<td>(0.009)</td>
<td>(0.010)</td>
<td>(0.013)</td>
<td>(0.008)</td>
</tr>
</tbody>
</table>

### Panel B: Birth Outcomes (Mothers Without Primary Schooling)

<table>
<thead>
<tr>
<th></th>
<th>(1)</th>
<th>(2)</th>
<th>(3)</th>
<th>(4)</th>
<th>(5)</th>
<th>(6)</th>
<th>(7)</th>
<th>(8)</th>
<th>(9)</th>
<th>(10)</th>
<th>(11)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Share with 7+ Prenatal</td>
<td>0.069</td>
<td>0.068</td>
<td>0.089</td>
<td>0.069</td>
<td>0.095</td>
<td>0.065</td>
<td>0.060</td>
<td>0.095</td>
<td>0.063</td>
<td>0.033</td>
<td>0.078</td>
</tr>
<tr>
<td>Visits</td>
<td>(0.017)</td>
<td>(0.040)</td>
<td>(0.032)</td>
<td>(0.039)</td>
<td>(0.037)</td>
<td>(0.037)</td>
<td>(0.036)</td>
<td>(0.048)</td>
<td>(0.033)</td>
<td>(0.042)</td>
<td>(0.043)</td>
</tr>
<tr>
<td>Share With Low-Weight</td>
<td>−0.546</td>
<td>−0.545</td>
<td>−0.489</td>
<td>−0.401</td>
<td>−0.263</td>
<td>−0.495</td>
<td>−0.394</td>
<td>−0.605</td>
<td>−0.405</td>
<td>−0.504</td>
<td>−0.471</td>
</tr>
<tr>
<td>Births (×100)</td>
<td>(0.210)</td>
<td>(0.251)</td>
<td>(0.238)</td>
<td>(0.226)</td>
<td>(0.208)</td>
<td>(0.245)</td>
<td>(0.238)</td>
<td>(0.338)</td>
<td>(0.236)</td>
<td>(0.479)</td>
<td>(0.236)</td>
</tr>
<tr>
<td></td>
<td>(0.004)</td>
<td>(0.040)</td>
<td>(0.052)</td>
<td>(0.062)</td>
<td>(0.174)</td>
<td>(0.070)</td>
<td>(0.126)</td>
<td>(0.108)</td>
<td>(0.112)</td>
<td>(0.332)</td>
<td>(0.068)</td>
</tr>
</tbody>
</table>

**Controls:**
- GDP, Gini, Poverty, Pop.
- Time-Dummies Interacted w/ [Correlation of Var. w/ Si]
- Restricted Sample
- State-Specific Trends
- Yes

**N**
- 81

---

*Standard errors clustered at the state level in parentheses. p-values based on Cameron, Gelbach, and Miller (2008) cluster-robust wild-bootstrap in early brackets—(1). The unit of observation is a state-electoral term. Each row reports the estimation of equation (8) using the specified dependent variable. Each figure is from a separate regression using a sample covering three electoral terms, reporting the coefficient from the interaction between the share of voters living in municipalities above the 40,500 cutoff (as of 1996) and a dummy indicator for the 1998 electoral term (1999–2002). State fixed effects, time effects, and region-time effects are included in all regressions. The restricted sample excludes states that did not follow the discontinuous assignment rule.*
The fourth set of additional tests is discussed in the Supplemental Material. It involves estimating equation (8) instrumenting $S_i$ with the ratio between the number of voters in municipalities with electorate in the $[40,500, 40,500 + j]$ interval and the number of voters in the $[40,500 - j, 40,500 + j]$ interval, for multiple values of $j$. This instrument thus focuses differences across states in the distribution of municipality size around the cutoff, and excludes the variation driven by the presence of very large cities. In general, the effects are robust to the use of this instrument. This suggests that shocks affecting states with a high share of their population in large cities are unlikely to drive the results.

Column (10) of Table VI excludes from the sample the four states that did not participate in the discontinuous assignment in 1998 (and used EV all municipalities). As discussed in Section 3.2, previous estimation pooled variation from the unusual pattern of EV’s phase-in with one akin to a standard difference-in-differences approach. Removing the states fitting the latter description hence provides a robustness check. Similar results are found, with point estimates being only marginally affected, and some loss of precision due to the loss of 15% of the sample. The exception is the prenatal visits estimates, which become smaller, although they are imprecise in the full sample, too. The robustness to possible outliers is further addressed in the Supplemental Material by providing leave-one-state-out estimates and also a graphical representation of the empirical strategy where individual data points can be observed. Column (11) adds state-specific trends. All effects remain similar, except for the case of visits that becomes larger.

4. CONCLUSION

This paper estimates the effects of electronic voting technology that facilitated ballot operation in Brazilian state elections. Results indicate that it promoted the de facto enfranchisement of (mainly less educated) voters. Consistent with the predictions of theories of redistributive politics and enfranchisement and democratization, it also increased government spending in a service that particularly benefits less educated voters: health care. It also had effects on its utilization and outcomes, increasing the number of prenatal visits and reducing the number of low-weight births by less educated mothers. However, it should be noted that, given the limitations in spending data discussed in Section 3, a fully detailed welfare analysis of the policy changes induced by EV cannot be provided, and that the welfare gains from health services utilization and outcomes may be offset by losses in other (unobserved) areas.

A final caveat is that the estimated effects may be dependent on the particular context. The degree of literacy required to operate the Brazilian paper ballots is higher than in contexts with fewer candidates, where their name and picture can be listed. The other results may also depend on features of the Brazilian political and health care systems. However, this study exemplifies how increasing the political participation of disadvantaged groups can shift
policymaking and affect outcomes, and, most importantly, provides evidence supporting the general mechanisms in redistributive politics.

REFERENCES


Dept. of Economics, Princeton University, 357 Wallace Hall, Princeton, NJ 08544, U.S.A., CIFAR, and NBER; fujiwara@princeton.edu.

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