

Public Service Failures and Voter Participation: Evidence from Court-Validated Electoral Disruptions*

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Abstract

This paper studies the long-run effects of non-strategic administrative failures on voter participation. We exploit a natural experiment from Berlin’s 2021 elections, in which hundreds of precincts experienced ballot shortages, multi-hour queues, and unlawful polling closures. Using court-validated treatment assignments and a stacked event-study design across nine elections (2014–2025), we estimate a 1.8 percentage points (2.4%) decline in turnout in affected precincts. The drop is concentrated in in-person voting and only partially offset by increases in postal participation in subsequent elections. Effects are largest among young voters, welfare recipients, and residents with migration backgrounds. Survey evidence suggests two mechanisms: disrupted civic habit formation and short-term erosion of institutional trust.

Keywords: Postal Voting, Voter Turnout, Elections, Administration, Disruptions, Berlin, Voting Costs

JEL-Codes: D72, H11, H70, R50

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1 Introduction

The quality of public service delivery plays a central role in shaping citizens’ trust in government and their engagement with democratic institutions. While most research on electoral participation focuses on formal barriers to voting or political incentives to mobilize, much less attention has been paid to the role of routine administrative competence in implementation. Failures in the basic execution of voting—such as ballot shortages, excessive queues, or unlawful poll closures—may signal limits to state capacity and reduce participation for years.

We study the long-run consequences of *non-strategic* disruptions during Berlin’s 2021 elections, when hundreds of precincts faced severe operational problems. These were not partisan interventions or legal restrictions, but failures in the delivery of a core public service. Unlike the well-documented effects of voter ID laws, registration restrictions, or gerrymandering, the disruptions we examine stemmed from implementation errors: ballot misallocations, hours-long queues, and unlawful polling closures. [Pettigrew \(2021\)](#) shows that long wait times at polling places in the United States reduce turnout in subsequent elections. Our findings demonstrate that similar downstream effects occur even in high-trust, high-capacity settings—and may persist longer than previously understood.

Using a natural experiment based on court-validated treatment assignments, we estimate the causal effects of exposure to these disruptions on long-run participation. Treated precincts saw a persistent 1.8 percentage points (2.4%) drop in total turnout, with no recovery over four years. The decline is concentrated in in-person voting and is only partially offset by increases in postal participation.

This paper conceptualizes elections as recurring tests of state capacity. While much of the literature has focused on strategic manipulation of electoral access, our contribution lies in showing that *non-strategic* bureaucratic failures can have similar long-run effects. [Besley and Persson \(2010\)](#) and [Acemoglu et al. \(2020\)](#) emphasize the role of effective institutions and service delivery in sustaining trust in institutions, and democratic legitimacy. In this tradition, we treat election administration as part of the broader public service infrastructure—one whose breakdown can alter beliefs, behavior, and political inclusion.

We construct a panel of harmonized precinct-level turnout and vote mode data across nine elections in Berlin (2014–2025). Treatment is defined based on a legally binding ruling by Germany’s Federal Constitutional Court, which annulled the results in 431 precincts due to severe procedural violations. Because treatment was assigned based on observable administrative failures—not on turnout levels or political outcomes—this setting permits quasi-experimental identification.

We implement a stacked event-study design to trace the evolution of turnout over time. In treated precincts, turnout falls sharply in 2021 and remains below baseline in all subsequent elections. Heterogeneity analyses show that declines are steepest in precincts with higher shares of young voters, welfare recipients, and residents with migration backgrounds. These are voters whose civic engagement is already fragile—and who appear most susceptible to scarring when participation costs spike unexpectedly.

To probe mechanisms, we link our administrative panel to survey data from the German Longitudinal Election Study (GLES). Treated voters report lower trust in electoral fairness and are more likely to cite late-stage abstention. These results support a dual-mechanism framework: disruptions weaken civic habits and also trigger informational updating about state competence. Fujiwara et al. (2016), Gerber et al. (2003), and Coppock and Green (2016) model voting as habit-forming behavior, while Meredith et al. (2009) and Shino and Smith (2018) show that early disruptions can inhibit routine formation. Separately, work on belief updating in public service delivery (Alsan and Wanamaker, 2018; Lowes and Montero, 2021; Akhtari et al., 2022) finds that bureaucratic failure can reduce institutional trust. Our results echo this pattern in the electoral domain.

This paper contributes to three literatures. First, it extends work on the economics of voting by showing that *how* elections are administered—not just whether they are legally fair—affects long-run turnout (Cantoni, 2020; Alipour and Lindlacher, 2025; Hodler et al., 2015; Burden et al., 2014; Gerber et al., 2013). Second, it adds to behavioral political economy by highlighting habit disruption and belief decay as causal pathways linking public service failures to disengagement (Bechtel et al., 2018; Kaplan and Yuan, 2020; Olawole, 2023). Third, it speaks to the political economy of state capacity, showing how breakdowns in electoral logistics resemble failures in other public services such as health, policing, or education (Herron and Smith, 2012; Toral, 2023).

Finally, the findings carry policy implications. The costs of misadministration fall disproportionately on voters with the weakest political voice. Yet these costs are avoidable. Simple reforms—such as ballot stock buffers, queue length monitoring, and expanded postal infrastructure—could prevent similar failures in future elections. Ensuring that elections run smoothly is not just a technical exercise. It is a democratic imperative.

The remainder of this paper proceeds as follows. Section 2 describes the institutional setting, data sources, and the empirical strategy. Section 3 reports the main results. Section 4 explores political consequences. Section 5 presents the conceptual framework. Section 6 concludes.

2 Institutional Setting, Data, and Empirical Strategy

2.1 Setting and Data

Modern elections are not only exercises in democratic choice but also complex logistical undertakings. Their effectiveness as public services hinges on competent state capacity. This section summarizes the institutional and empirical context for Berlin’s 2021 election failures, the subsequent legal rulings, and the data infrastructure used to study their effects.

Berlin offers a valuable case for studying the consequences of administrative failure in a high-capacity democratic state. Germany is widely regarded as having strong institutions, high procedural trust, and a professional civil service. The electoral system is well-established, with routine federal, state, and district-level contests conducted under a uniform legal framework.

That such large-scale electoral failure occurred in this setting underscores the potential fragility of even well-functioning democracies when administrative delivery falters.

On 26 September 2021, Berlin held four concurrent elections: federal (*Bundestag*), state (*Abgeordnetenhaus*), district councils (*Bezirksverordnetenversammlung*), and a city-wide referendum. These elections took place under COVID-19 precautions and coincided with the Berlin Marathon, which restricted traffic flow and added further stress to the voting process. The result was a cascade of irregularities: ballot shortages, misallocated or missing ballots, polling stations opening late or closing early, and queues exceeding two hours. Media reports and citizen complaints described widespread dysfunction across the city.

These failures triggered extensive legal review. The Berlin Constitutional Court annulled the entire state and district elections, citing “systematic, city-wide electoral disruptions” that had made a proper conduct of the election “impossible in broad parts of the territory” ([Verfassungsgerichtshof des Landes Berlin, 2022](#), VerfGH 154/21). The court documented dozens of violations, including ballots for the wrong district, missing materials, overcrowded polling stations, and unlawful closures. In a separate ruling, the Federal Constitutional Court annulled the federal election in 431 precincts, citing “mandatsrelevante Wahlfehler”—errors severe enough to plausibly affect seat allocation in the Bundestag ([Bundesverfassungsgericht, 2023](#), BVerfG 2 BvC 4/23). These included ballot delivery failures, voting after 6:00 p.m. in violation of federal law, and prolonged queuing that exceeded the court’s threshold of tolerable burden for voters.

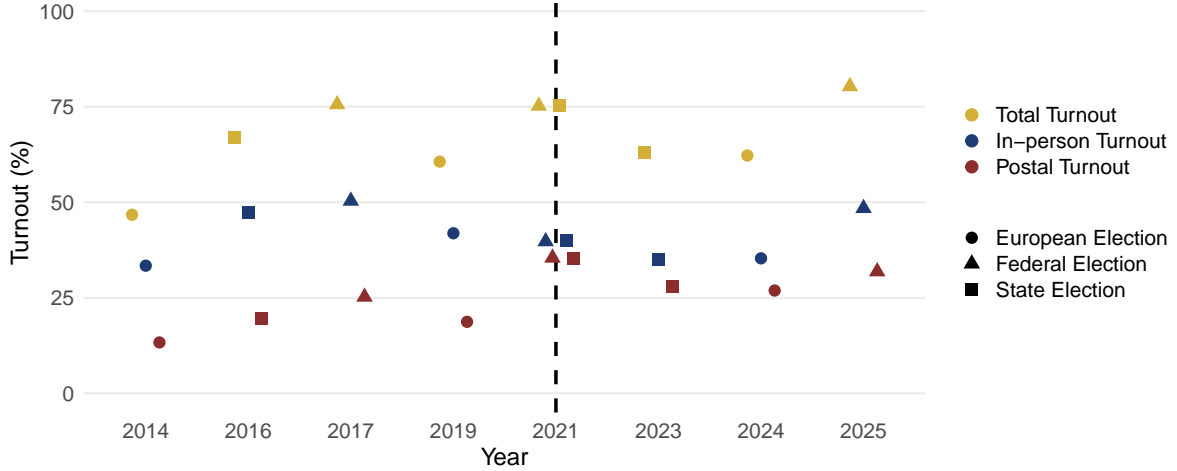
These rulings offer a rare source of court-validated, precinct-level treatment assignment. Unlike studies relying on media reports, perceptions, or self-reported experience, our setting defines treatment based on formal judicial determinations of procedural failure. This ensures that the classification of affected precincts is exogenous to observed turnout or political behavior.

The irregularities constituted a multidimensional disruption to voting. Ballot unavailability and long queues raised participation costs; procedural confusion and unauthorized polling closures likely reduced trust in the electoral process; and road closures due to the marathon imposed physical access frictions. Prior research shows that even minor logistical barriers—such as relocating a polling place or increasing distance to it—can reduce turnout, particularly among marginal voters ([Alipour and Lindlacher, 2025](#); [Cantoni, 2020](#)). The Berlin case, by contrast, involves severe, visible breakdowns in electoral delivery in a setting otherwise known for bureaucratic order, providing a test of behavioral and informational mechanisms under extreme but non-strategic administrative stress.

Importantly, Germany’s electoral system allows for postal voting without excuse, which has steadily expanded since 2008. Voters may request ballots in advance and return them by post or by deposit. As shown in [Figure 1](#), postal turnout has increased over time and now constitutes a major channel of participation. This allows us to distinguish whether discouraged in-person voters substituted toward mail ballots, or dropped out of the electorate entirely. The former implies a shift in vote mode; the latter suggests behavioral disengagement.

We construct a precinct-level panel of electoral returns across nine State, European, and Federal elections between 2014 and 2025. The unit of observation is the postal precinct (*Briefwahlbezirk*), the smallest level for which mode-specific turnout is reported. To allow con-

Figure 1: Turnout by Election Type in Berlin since 2014



Notes: The figure presents total (yellow), in-person (blue), and postal turnout (green) for the European (circle), Federal (triangle), and State (square) elections in the sample. The dotted vertical line marks the 2021 election, which triggered the constitutional review.

sistent comparisons over time, all data are harmonized to the 2021 precinct geography using population-weighted overlays based on 100-meter census grid data (see [Appendix B](#) for details on data sources, harmonization, and treatment coding).

Our analysis focuses on the 431 precincts invalidated by the Federal Constitutional Court, as these featured high-salience, court-verified disruptions and were subject to a court-ordered rerun of the federal vote in 2024.¹ Precincts affected only by the state court ruling are excluded from the main analysis, as their disruptions (e.g., ballot misallocation without delays) were less visible, and the 2023 state rerun was widely perceived as low-stakes. Including both types of precincts yields qualitatively similar but attenuated estimates (see [Appendix A](#)).

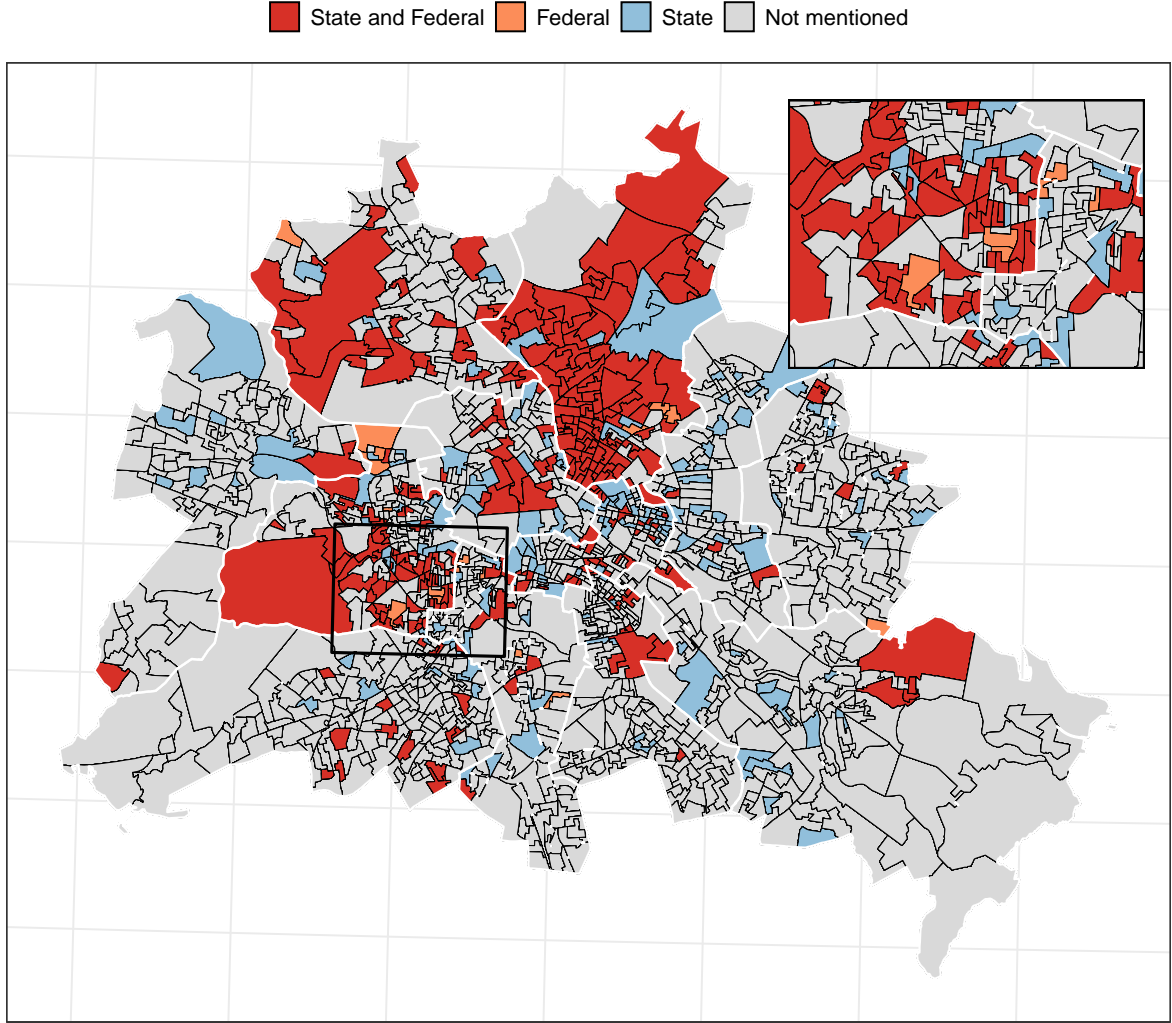
[Figure 2](#) shows the geographic distribution of affected postal precincts. While disruptions cluster in inner-city districts, affected postal precincts are present across nearly boroughs, supporting comparisons with a broad control group.

2.2 Estimation Equation

We estimate a dynamic difference-in-differences model using a stacked event study design. This approach leverages quasi-experimental variation in exposure to administrative disruptions during the 2021 federal election. All treated precincts were affected simultaneously, and voting outcomes are observed across multiple elections before and after the disruption for the same spatial units.

¹ While these precincts were subject to a court-ordered rerun of the federal election in 2024, we do not interpret the rerun as a separate treatment. The rerun was held more than two years after the original disruption, with minimal campaign activity and very low turnout. It is unlikely to have restored voting habits or trust; see [Appendix B](#) for discussion.

Figure 2: Berlin Precincts by Court Mention



Notes: The figure displays Berlin's postal precincts (*Briefwahlbezirke*) classified by whether they were formally designated as affected by administrative irregularities during the 2021 elections. Status is based on legal rulings by the *Berlin Constitutional Court* (state election), the *Federal Constitutional Court* (federal election), or both. These designations reflect judicial findings of procedural violations and do not distinguish among types of disruption (e.g., ballot errors, long queues). District boundaries are shown in white. The inset in the top right zooms into a representative high-density area to illustrate the fine-grained spatial distribution of statuses. Its location is exemplary and not of substantive analytical relevance.

The estimating equation is:

$$Y_{idt} = \sum_{\tau \neq -1} \beta^{\tau} (\mathbb{1}_{\tau=t} \times \text{Treatment}_i) + \alpha_i + \alpha_{dt} + \varepsilon_{idt}, \quad (1)$$

where Y_{idt} denotes the natural log of turnout (total, in-person, or postal) in postal precinct i , district d , and election year t . The variable Treatment_i is an indicator for whether precinct i was identified as affected by severe electoral disruptions in the Federal Constitutional Court ruling. The indicators $\mathbb{1}_{\tau=t}$ denote relative election year, where $\tau - 1$ corresponds to the 2019 European

Parliament election (the pre-treatment baseline), and $\tau \in \{-4, -3, -2, 0/F, 0/S, 1, 2, 3\}$ indexes elections from 2014 to 2025.²

The model includes postal precinct fixed effects α_i to absorb time-invariant spatial heterogeneity and district-by-election fixed effects α_{dt} to flexibly capture time shocks varying across districts. Standard errors are clustered at the precinct level to account for serial correlation.

The coefficients β^τ capture the dynamic treatment effects of exposure to administrative failure. Estimates in $\tau + 0$ reflect the immediate effect on turnout in 2021, while post-treatment coefficients ($\tau > 0$) trace the persistence or attenuation of effects over time.

This framework allows us to assess whether administrative failures affect not only whether voters turn out, but also how they vote (by mode), and whether these effects are transitory or sustained. The next section presents the results of this analysis and examines heterogeneity across voting channels and voter subgroups.

3 Results

3.1 Effects on Voter Turnout

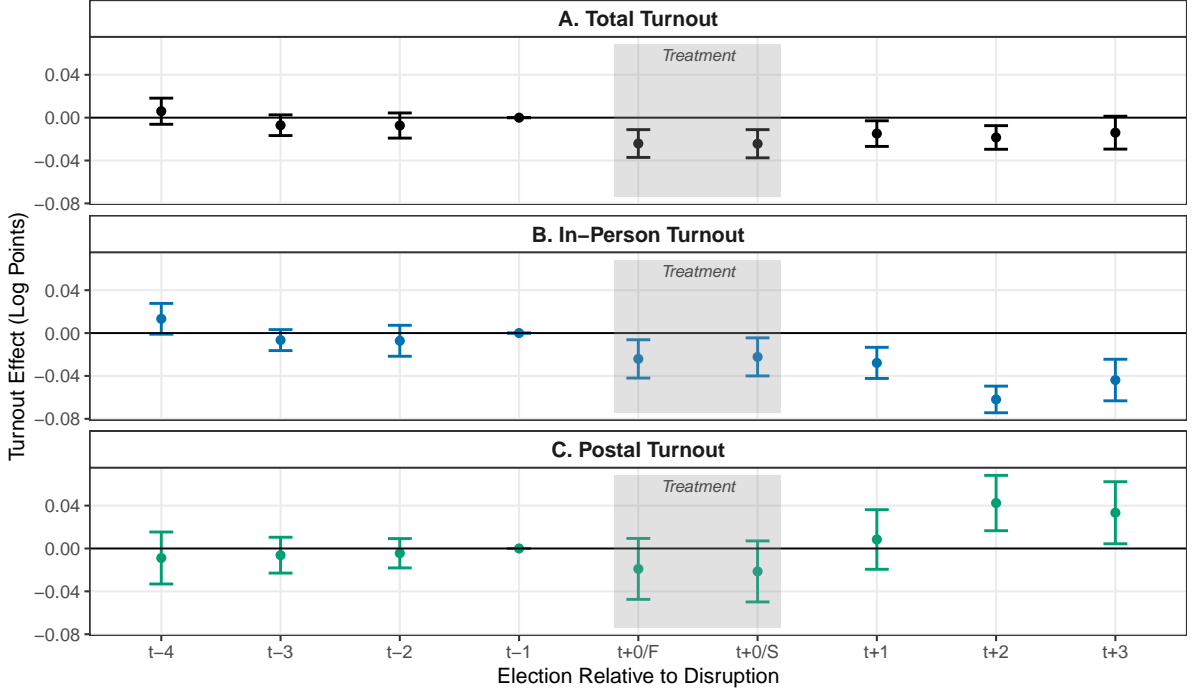
This section presents the main results from the event study design, which estimates the causal effect of administrative irregularities during the 2021 Berlin State and Federal Elections on voter turnout. Using Equation 1, we track changes in total turnout, in-person voting, and postal voting across treated and untreated precincts over multiple election cycles. Each election is interpreted as a relative period $\tau \in \{-4, -3, -2, -1, 0, +1, +2, +3\}$ with respect to the 2021 disruption. Figure 3 visualizes the estimated coefficients $\hat{\beta}^\tau$ by voting mode, and the underlying estimates are reported in Table 1. For interpretability, we convert all log-point estimates into percentage point effects by scaling them with the mean turnout in control precincts in the corresponding election year.

The pre-treatment period serves to assess the plausibility of the parallel trends assumption. Estimated effects in $\tau - 2$ and $\tau - 3$ are close to zero and statistically insignificant across all turnout modes. For instance, total turnout in 2017 differs by just -0.56 percentage points in treated versus control precincts, and postal turnout by -0.12 points. A marginally positive estimate for in-person turnout in 2014 (+0.46 pp, $p = 0.07$) does not persist in later pre-treatment years and is consistent with random variation.

The most substantial effects emerge in the disruption year 2021 ($\tau + 0$). In the Federal election, total turnout in treated precincts declined by 1.82 percentage points relative to controls, from a baseline of 76.4%. This drop is driven entirely by in-person turnout, which fell by 0.95 pp from a baseline of 39.7%. Postal turnout declined only modestly (-0.69 pp) and the estimate is not statistically significant. A similar pattern holds for the State election held on the same day, with total turnout declining by 1.84 pp and in-person turnout by 0.88 pp. Postal turnout again shows no meaningful change. These immediate effects are consistent with the hypothesized cost

² 0/F and 0/S refer to the 2021 Federal and State elections, which were held concurrently.

Figure 3: Main Specification



Notes: The figure presents event study results based on Equation 1 for the natural log of postal, in-person, and total turnout. The reference election ($t - 1$) is the 2019 European election. Confidence intervals are drawn at the 95 percent level using standard errors clustered at the postal precinct level. The point estimates and standard errors underlying the results appear in Table 1.

shock: ballot shortages, long queues, and administrative confusion increased the effort required to vote, particularly for in-person participants.

Two years later, in the court-mandated rerun of the Berlin State election in 2023 ($\tau + 1$), total turnout remains 0.94 percentage points lower in treated precincts, relative to a control group average of 63.8%. This continued suppression is again concentrated in in-person voting (-0.97 pp), while postal turnout rises slightly (+0.24 pp) but remains statistically insignificant. The absence of any compensatory spike suggests a persistent behavioral scarring effect.

By the time of the 2024 European Parliament election ($\tau + 2$), the pattern becomes more pronounced. In-person turnout in treated areas is 2.15 percentage points lower than in controls (baseline: 35.8%), the largest effect observed in any post-treatment period. This decline is partially offset by a significant increase in postal turnout (+1.16 pp), resulting in a net reduction in total turnout of 1.15 pp. These findings suggest gradual behavioral adaptation: some voters exposed to administrative failure appear to shift toward mail voting, but not enough to fully close the turnout gap.

In the 2025 Federal election ($\tau + 3$), in-person turnout remains 2.09 percentage points lower in treated precincts, and total turnout is still depressed by 1.12 points. Postal turnout increases by 1.08 pp, indicating some continued substitution, but the overall participation rate remains below baseline.

Taken together, the estimates show that administrative irregularities exert both immediate and durable effects on electoral participation. The sharp initial drop in in-person turnout aligns with models in which increased participation costs suppress turnout (e.g., [Downs, 1957](#); [Pettigrew, 2021](#)). The persistence of these effects—despite the resolution of logistical failures—points to longer-run behavioral scarring and possible belief updating about the electoral process. That the turnout gap remains visible three years later, and across unrelated elections, underscores the potential for even one-time administrative failures to undermine civic engagement.

Finally, we conduct several robustness checks, including alternative fixed effect specifications, placebo treatments in pre-treatment years, and subsample restrictions. These are detailed in [Appendix A](#). The magnitude and statistical significance of treatment effects on in-person turnout are highly stable across specifications, reinforcing the conclusion that the observed effects are not artifacts of model choice or sampling variation.

Table 1: Main Specification

| | Total Turnout (1) | In-person Turnout (2) | Postal Turnout (3) |
|-------------------------|------------------------|--------------------------|-----------------------|
| Treatment (t-4) | 0.0060 (0.0062) | 0.0133* (0.0073) | -0.0089 (0.0124) |
| Treatment (t-3) | -0.0071 (0.0049) | -0.0066 (0.0050) | -0.0062 (0.0085) |
| Treatment (t-2) | -0.0073 (0.0060) | -0.0072 (0.0074) | -0.0044 (0.0070) |
| Treatment (t+0/Federal) | -0.0241*** (0.0066) | -0.0241*** (0.0091) | -0.0191 (0.0145) |
| Treatment (t+0/State) | -0.0243*** (0.0067) | -0.0222** (0.0090) | -0.0214 (0.0145) |
| Treatment (t+1) | -0.0149** (0.0061) | -0.0278*** (0.0074) | 0.0084 (0.0142) |
| Treatment (t+2) | -0.0185*** (0.0056) | -0.0619*** (0.0063) | 0.0424*** (0.0132) |
| Treatment (t+3) | -0.0140* (0.0078) | -0.0438*** (0.0099) | 0.0333** (0.0148) |
| R ² | 0.92949 | 0.86623 | 0.93030 |
| Observations | 12,150 | 12,150 | 12,150 |
| Precinct FE | ✓ | ✓ | ✓ |
| Election-District FE | ✓ | ✓ | ✓ |

Notes: The table presents event study results based on [Equation 1](#) for postal turnout, in-person turnout, and total turnout (0-100). The reference election ($t - 1$) is the 2019 European Election. Standard errors are clustered at the postal precinct level and reported in parentheses. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

3.2 Disruption Intensity: Waiting Times

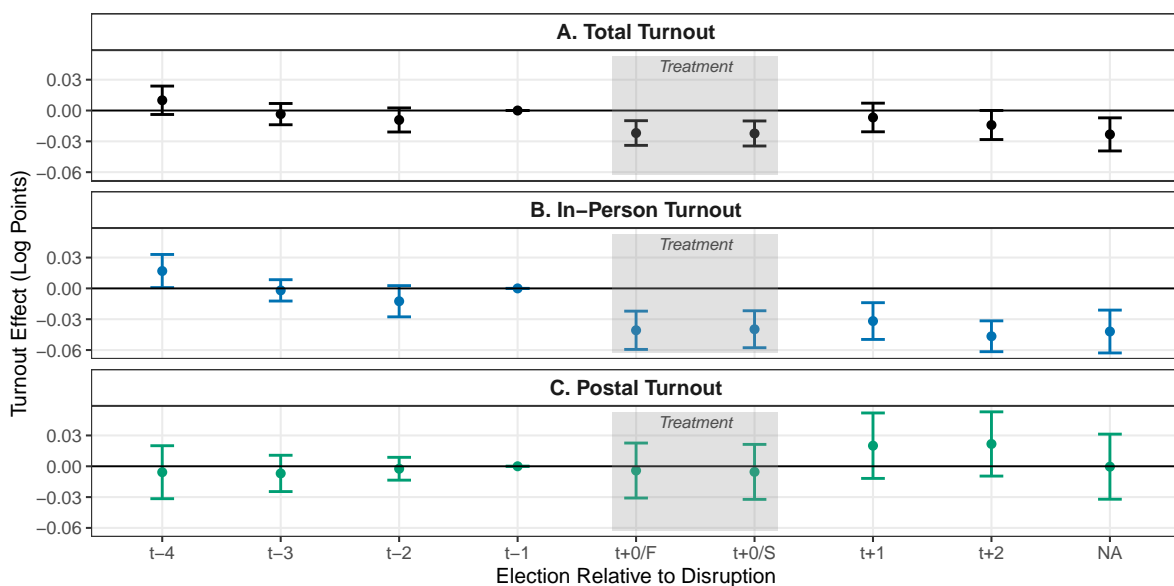
While the main analysis uses a binary treatment indicator based on legal rulings, not all affected precincts experienced irregularities of equal severity. This subsection examines whether the intensity of disruption—measured as maximum waiting time in hours—is associated with

differential effects on turnout. We thus shift from a dichotomous to a continuous treatment definition, testing for dose-response effects of election-day failure.

Disruption intensity is operationalized using precinct-level estimates of queue length during peak voting hours, as recorded in the official ruling of the Berlin Constitutional Court. These data, reported in hours, capture the severity of procedural breakdowns at the polling station level and are aggregated to the postal precinct (*Briefwahlbezirk*).

Figure 4 plots the estimated effects using an interaction between waiting time and relative election year, following the stacked event-study framework. The results show a clear and persistent relationship between disruption intensity and turnout suppression.

Figure 4: Treatment Intensity



Notes: The figure presents triple-interaction estimates based on Equation 1 for total turnout, in-person turnout, and postal turnout (0-100). The treatment variable is the maximum waiting time (in hours) per postal precinct during the 2021 elections, as recorded by the Berlin Constitutional Court. Each point shows the effect of one additional hour of waiting on turnout in a given election. The sample is restricted to postal precincts that are affected by longer waiting times, disruptions, or longer opening hours, and the control group. The reference election ($t - 1$) is the 2019 European Election. Confidence intervals are drawn at the 95 percent level using standard errors clustered at the postal precinct level. The point estimates and standard errors underlying the results appear in Appendix Table A8. All first and second-order interaction terms required for the identification of the triple-difference estimator are included in the specification or absorbed by the fixed effects.

In the immediate aftermath ($t + 0$), each additional hour of waiting is associated with a 1.36 percentage point decline in total turnout, mirroring the average treatment effect from the binary specification. The effect on in-person turnout is notably stronger: an extra hour of queuing reduces in-person participation by 1.6 to 1.8 percentage points in 2021. These effects remain statistically significant in subsequent cycles. By the 2024 European election ($t + 2$), an additional hour of waiting corresponds to a 4.7 percentage point drop in in-person turnout and a 2.3 point decline in total turnout—exceeding the initial effect at $t + 0$.

Postal turnout exhibits the opposite trend: longer wait times in 2021 are associated with small but positive increases in mail voting in $t + 2$ and $t + 3$, suggesting behavioral adaptation.

For example, in precincts that experienced multi-hour queues, postal participation in 2024 rises by up to 2.2 percentage points per hour of prior waiting, partially offsetting in-person losses.

Taken together, these findings reinforce the interpretation that administrative failures depress turnout in a dose-dependent manner. Voters appear to react not only to the presence of irregularities but to their severity. The results also support the substitution narrative: while some affected voters eventually shift to mail voting, this behavioral compensation is incomplete and delayed.

These patterns have both theoretical and policy implications. From a theoretical perspective, the dose-response shape is consistent with models of civic habit scarring and belief updating: more severe disruptions should lead to stronger belief shocks and greater habit discontinuities. From a policy standpoint, the results highlight queue duration as a critical administrative metric. Long lines are not merely logistical frictions; they are visible signals of state dysfunction that can undermine democratic participation in the long run. Monitoring and mitigating wait times—via staffing buffers, queue management systems, or expanded early/postal voting—should be viewed as core components of electoral resilience.

3.3 Heterogeneous Effects by Demographics and Prior Behavior

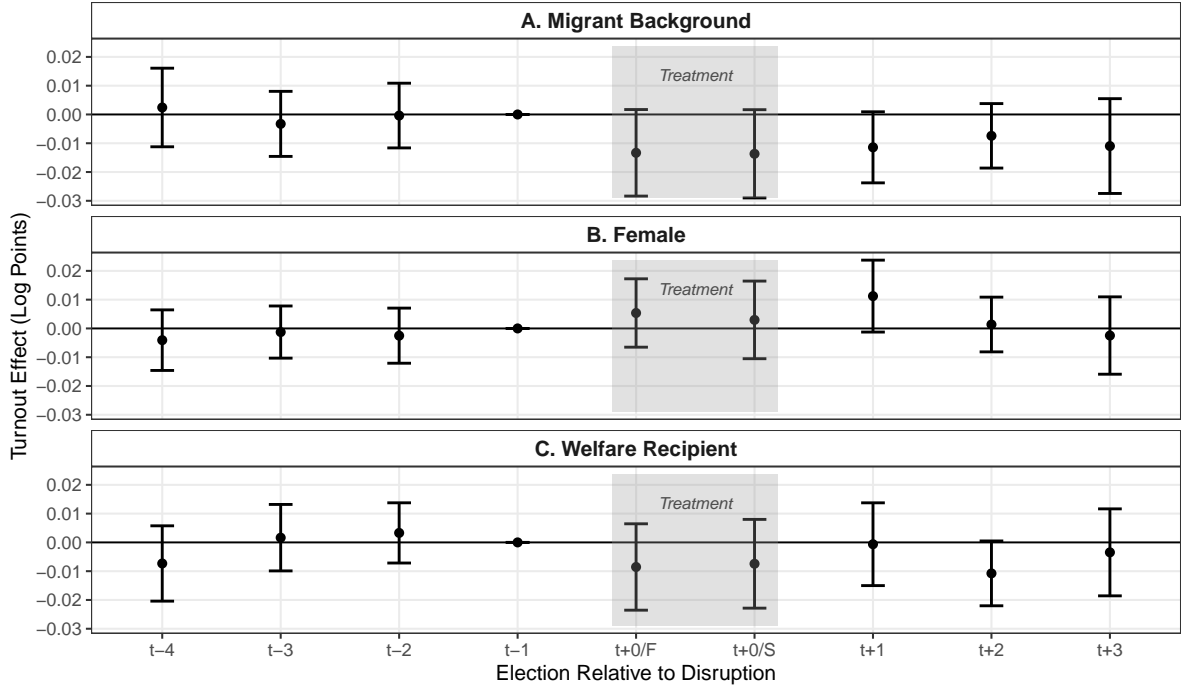
While the average treatment effects show persistent declines in turnout following administrative failure, existing theory and empirical research suggest that such effects may vary across voter groups. Participation costs and voting habits are not evenly distributed across the electorate, and expectations about the integrity of state services may differ systematically by demographic background. This section investigates these differences by interacting treatment with standardized socio-demographic characteristics in a triple-difference event-study specification:

$$Y_{idfst} = \sum_{\tau \neq -1} \gamma^\tau (\mathbb{1}_{\tau=t} \times \text{treatment}_i) + \sum_{\tau \neq -1} \mu^\tau (\mathbb{1}_{\tau=t} \times \text{treatment}_i \times Z_i) + \eta_i + \eta_{dt} + \epsilon_{idfst}, \quad (2)$$

where Z_i is a standardized (mean-zero, unit-variance) measure of the demographic characteristic of interest. The coefficients μ^τ identify whether the treatment effect varies with that characteristic. Results are visualized in [Figure 5](#) and [Figure 6](#).

Migrant Background. Precincts with higher shares of adult citizens with a migration background experience systematically larger turnout declines after the 2021 disruption. A one standard deviation increase in migrant share is associated with an additional drop of around 0.5 percentage points at $t + 0$, and this differential persists through later elections. This is consistent with prior evidence that political behavior among migrant-background voters is more sensitive to signals of institutional competence and inclusion ([Olawole, 2023](#)). If civic habits are less entrenched and trust in electoral administration more fragile in these populations, then visible administrative breakdowns may be particularly demobilizing. These findings also resonate with broader research on state capacity and participatory inequality ([Alsan and Wanamaker, 2018](#); [Akhtari et al., 2022](#)).

Figure 5: Heterogeneity by Socio-Demographics



Notes: The figure presents triple-interaction estimates based on Equation 2 total turnout (0-100). The reference election ($t - 1$) is the 2019 European Election. Confidence intervals are drawn at the 95 percent level using standard errors clustered at the postal precinct level. The point estimates and standard errors underlying the results appear in Appendix Table A9. All first and second-order interaction terms required for the identification of the triple-difference estimator are included in the specification or absorbed by the fixed effects.

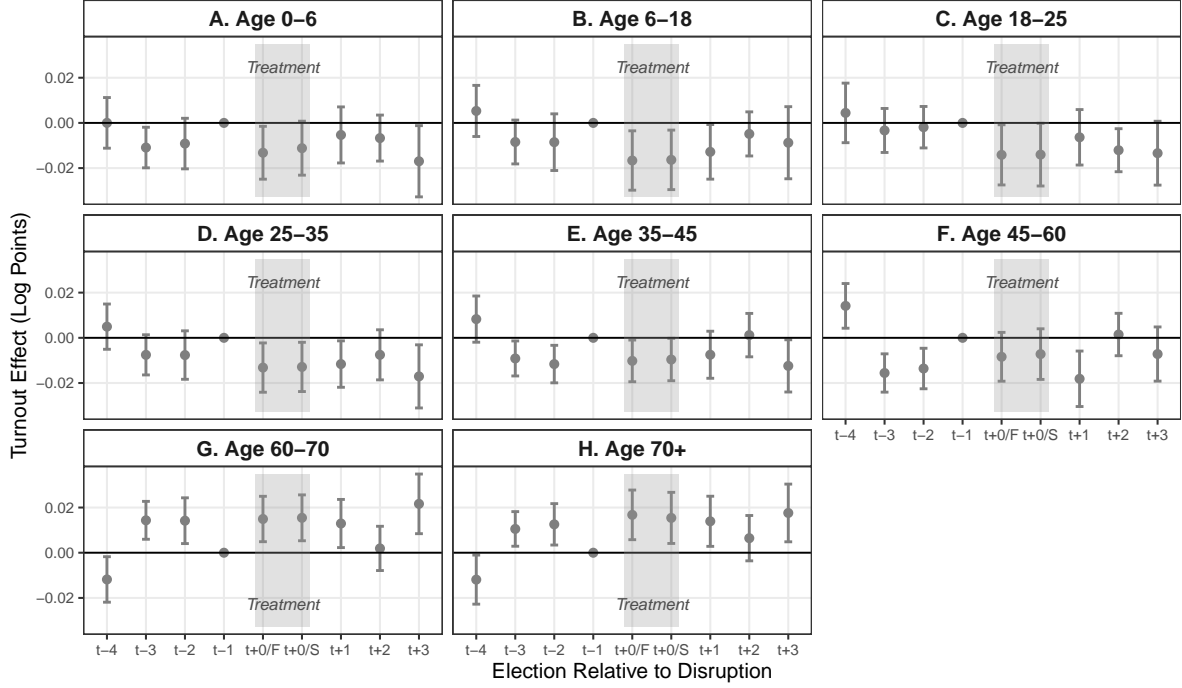
Welfare Recipients. We find similarly persistent effects in precincts with higher shares of welfare recipients. Here, however, the treatment interaction grows in magnitude over time: the effect is small at $t + 0$ but becomes statistically significant by $t + 1$ and peaks in $t + 2$. This delayed dynamic suggests a compounding disengagement process. Voters with lower economic security may initially retain their voting behavior, but interpret unresolved administrative failure as a broader signal of exclusion or neglect. These results are in line with the idea that bureaucratic quality shapes political participation not only through direct experience but also through symbolic representation (Lowes and Montero, 2021; Herron and Smith, 2012).

Gender. We find no robust evidence of heterogeneous effects by gender composition. Across all post-treatment elections, interaction coefficients for the share of female voters are small and statistically indistinguishable from zero. This suggests that, in this context, gender does not systematically condition the behavioral response to electoral disruptions. The result is consistent with prior studies showing that gender gaps in turnout are often small or absent in institutional settings with universal enfranchisement and non-strategic barriers to participation.

Age. Age exhibits a strong and intuitive gradient. Young voters aged 18–25 and 25–35 show the largest treatment effects in the immediate and medium-term aftermath of the disruption, with declines exceeding 1.5 percentage points in some cycles. By contrast, older voters (60+)

exhibit no turnout loss, and in later years even show slightly positive deviations. These patterns align with models of civic habit formation: older voters likely possess more stable voting routines, while younger cohorts—many of them new or recent entrants to the electorate—are more vulnerable to disruptions that prevent habit consolidation (Fujiwara et al., 2016; Shino and Smith, 2018; Coppock and Green, 2016). Moreover, younger voters may be more responsive to salient signals about institutional quality, especially in their early voting experiences.

Figure 6: Heterogeneity by Age



Notes: The figure presents triple-interaction estimates based on Equation 2 total turnout (0-100). The reference election ($t - 1$) is the 2019 European Election. Confidence intervals are drawn at the 95 percent level using standard errors clustered at the postal precinct level. The point estimates and standard errors underlying the results appear in Appendix Table A10. All first and second-order interaction terms required for the identification of the triple-difference estimator are included in the specification or absorbed by the fixed effects.

Taken together, these patterns reinforce the interpretation that the long-run effects of administrative failure are not uniform, but concentrated among politically vulnerable or institutionally peripheral populations. The magnitude and timing of turnout decline across subgroups is consistent with mechanisms of disrupted habit formation and revised expectations about state reliability. While we cannot fully disentangle these channels in the data, the heterogeneity results support a behavioral interpretation of electoral disengagement, whereby early disruptions shape subsequent civic trajectories.

4 Political Consequences of Electoral Disruption

4.1 Behavioral and Attitudinal Effects

To complement the turnout analysis with attitudinal evidence, we turn to individual-level data from the German Longitudinal Election Study (GLES). Specifically, we examine whether the 2021 disruptions affected trust in electoral institutions and the timing of abstention decisions. While these mechanisms cannot be separately identified in the turnout data, survey responses provide suggestive evidence on the informational and behavioral channels proposed in our conceptual framework.

We use pooled data from the GLES rolling cross-sections for the 2017, 2021, and 2025 federal elections. The estimation follows a difference-in-differences approach comparing Berlin respondents (the treated group) to voters in other German states, using 2017 as the pre-treatment baseline:

$$Y_{ist} = \sum_{\kappa \neq 2017} \beta^{\kappa} (\mathbf{1}_{\kappa=t} \times \text{Berlin}_i) + \gamma_s + \lambda_t + u_{ist}, \quad (3)$$

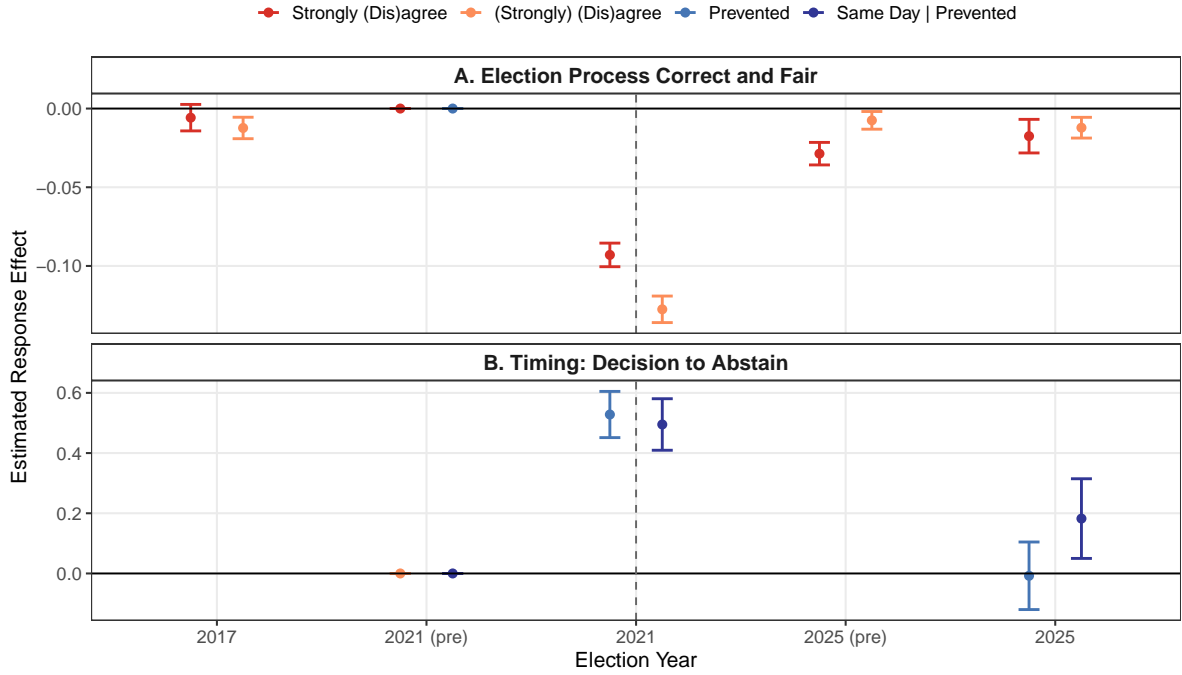
where Y_{ist} denotes the outcome of interest for individual i in state s and election year t , and γ_s and λ_t denote state and year fixed effects. Standard errors are clustered at the state level.

Panel A of [Figure 7](#) shows results for beliefs about electoral integrity, measured by agreement with the statement: “The federal election was conducted correctly and fairly”. In the immediate aftermath of the 2021 election, Berlin respondents are 9 to 13 percentage points less likely to agree with this statement compared to the rest of Germany. This sharp decline supports the interpretation that visible administrative failure produced a salient informational shock that updated voters’ beliefs about state competence. By 2025, the gap narrows considerably—suggesting that trust partially rebounds—though Berlin respondents still express lower confidence than their national peers.

Panel B examines the timing of non-voting decisions, using a subsample of approximately 510 self-reported abstainers. In 2021, Berlin nonvoters are substantially more likely to report deciding not to vote on election day itself or being prevented from voting altogether. Point estimates range from 49 to 53 percentage points higher than in other states. These patterns are consistent with the hypothesis that administrative disruption raised the perceived costs of voting, leading to last-minute abstention—particularly among voters with weaker civic habits or lower confidence in institutional responsiveness.

These survey results provide complementary support for the proposed behavioral mechanisms. The decline in electoral trust is consistent with belief updating in response to salient state failure ([Alsan and Wanamaker, 2018](#); [Akhtari et al., 2022](#)), while the shift in abstention timing aligns with habit-fragility models, in which minor frictions trigger exit among marginal participants ([Shino and Smith, 2018](#); [Coppock and Green, 2016](#)). Although trust appears to partially recover by 2025, turnout remains persistently lower in treated precincts—suggesting that short-term informational shocks can have long-run behavioral consequences.

Figure 7: GLES Survey Results



Notes: reports estimates from Equation 3 using pooled data from the GLES Rolling Cross-Section surveys conducted in 2017, 2021, and 2025. Panel A reports respondents' beliefs about electoral integrity, measured via agreement with the statement: "Die Bundestagswahl wurde von den zuständigen Behörden korrekt und fair durchgeführt". The reference election (2017(post)) is the pre-election survey of the 2021 Federal Election. Panel B examines the timing of abstention decisions among nonvoters. The point estimates and standard errors underlying the results appear in Appendix Table A11. The reference election (2017(post)) is the post-election survey of the 2017 Federal Election.

That said, these results should be interpreted with caution. Sample sizes for Berlin are limited, particularly for the abstention timing outcomes, and standard errors remain non-trivial despite large point estimates. Nonetheless, the directional consistency with theoretical expectations and the statistical significance of key comparisons strengthen the case for the proposed mechanisms of civic disengagement.

4.2 Administrative Response and Institutional Learning

The observed disruptions in 2021 raise the question of whether electoral administrators recognized the failures and adjusted the electoral infrastructure in response. If long queues and procedural breakdowns were interpreted as a failure of local election delivery, we might expect compensatory efforts in subsequent cycles—such as reallocating or expanding polling station capacity. This subsection examines whether precincts that experienced more severe disruption were more likely to receive additional polling places in later elections.

We estimate difference-in-differences regressions in which the outcome is the number of polling stations in a given precinct and election. The key independent variable is the intensity of disruption in 2021, measured as the total reported queueing time (in hours) across all in-person polling stations within a postal precinct. This variable is interacted with election year indicators

for 2023, 2024, and 2025. The specification includes precinct and election fixed effects. Results are shown in [Table 2](#).³

We find clear evidence of a delayed institutional response. In 2025, each additional hour of wait time in 2021 is associated with an increase of approximately 0.093 polling stations per precinct, significant at the 5% level. A similarly positive—though more imprecisely estimated—relationship appears in 2024. No statistically significant change is detected in 2023, when the Berlin state rerun occurred. This timing pattern suggests that administrative adjustments were not immediate but instead reflect a lagged response, consistent with internal review cycles, audit-triggered corrections, or reputational incentives to prevent recurrence in higher-salience contests.

Although we cannot directly assess whether the observed capacity expansion mitigated turnout decline, these findings offer suggestive evidence of bureaucratic learning. The allocation of additional polling infrastructure appears targeted to previously disrupted areas, indicating that electoral administrators responded to performance failures by adjusting resource deployment.

From a policy perspective, these results underscore that electoral institutions are not static. While often modeled as exogenous rule-makers or neutral implementers, electoral bureaucracies in high-capacity democracies may exhibit adaptive behavior—especially after highly visible breakdowns. Future work could examine whether these institutional responses closed participation gaps, either by reducing wait times or by restoring confidence among affected voters.

Table 2: Administrative Response: Change in Polling Station Capacity

| | Number of Polling Stations | |
|-----------------|----------------------------|-----------------------|
| | (1) | (2) |
| Treatment (t+1) | -0.0171 (0.0260) | 0.0407 (0.0321) |
| Treatment (t+2) | 0.8199*** (0.1918) | 0.4798*** (0.1697) |
| Treatment (t+3) | 0.0566 (0.0351) | 0.0927*** (0.0336) |
| Specification | Binary | Waiting Time in h |
| R ² | 0.89819 | 0.89723 |
| Observations | 5,400 | 5,400 |
| Precinct FE | ✓ | ✓ |
| Election FE | ✓ | ✓ |

Notes: The table presents event study results based on [Equation 1](#) for the number of polling stations within a postal precinct. The reference election ($t + 0$) is the 2021 European Election. Standard errors are clustered at the postal precinct level and reported in parentheses. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

³ Administrative data on polling station counts is available from 2021 through 2025.

4.3 Downstream Electoral Composition

While the primary effect of administrative failure is reduced participation, disruptions may also alter the partisan composition of the electorate. If certain parties rely more heavily on politically marginal voters—or if specific groups update their partisan preferences in response to institutional breakdowns—then turnout shocks may produce measurable shifts in vote shares. This subsection examines whether treated precincts exhibit systematic changes in party support following the 2021 disruptions.

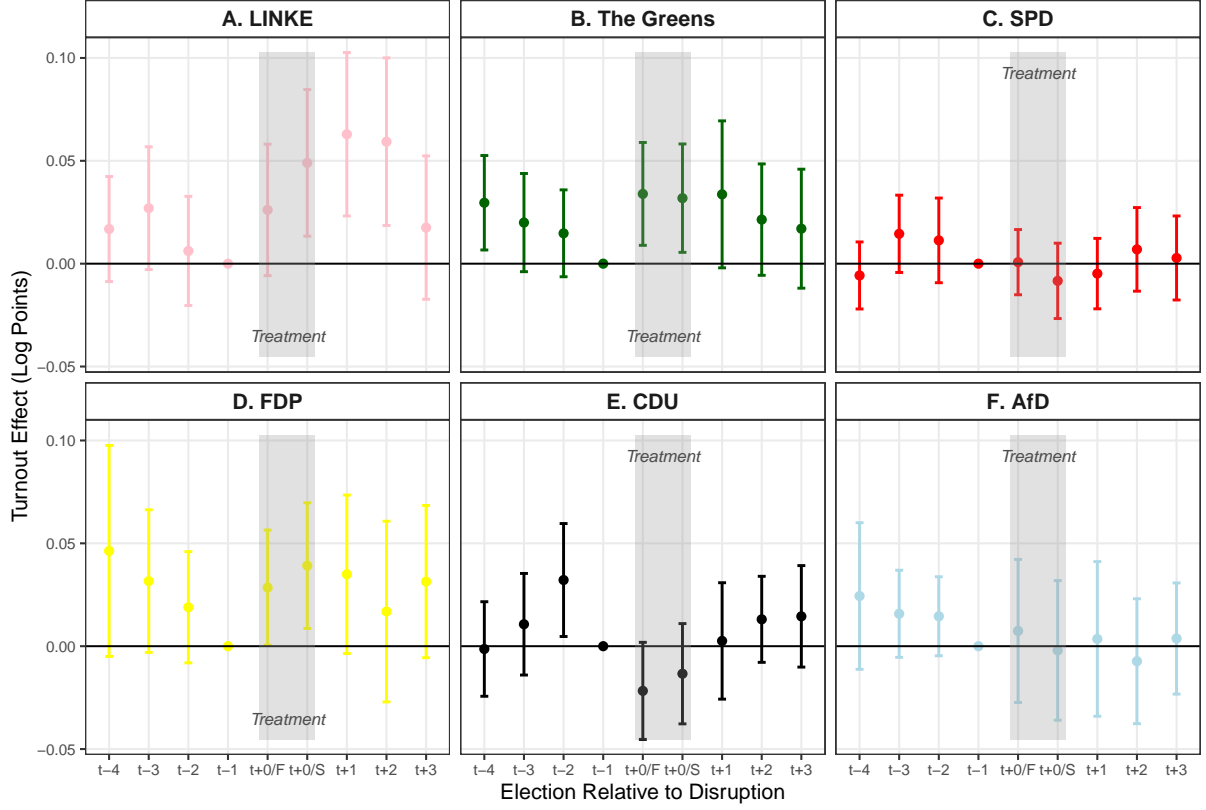
We estimate treatment effects on party vote shares using the same stacked event-study specification as in the main turnout analysis, including postal precinct and election-district fixed effects. Figure 8 summarizes results across four pre-treatment and three post-treatment elections. Most pre-treatment estimates are small and statistically indistinguishable from zero, consistent with the identifying assumption of parallel trends.

In the immediate post-treatment period ($t + 0$), we observe modest shifts in the distribution of party support. CDU vote shares decline by up to 0.8 percentage points, while the FDP and LINKE gain approximately 0.2–0.4 and 0.6 percentage points, respectively. These trends persist into $t + 1$, where the CDU registers a statistically significant loss of 0.85 points. Vote shares for SPD, Greens, and AfD show no consistent treatment effects across post-treatment elections.

While these partisan shifts are relatively small compared to overall turnout losses, they are directionally consistent with theoretical expectations. To the extent that electoral mismanagement triggers belief updating about state competence, voters with low institutional trust may be more likely to disengage or switch support. CDU losses may reflect reputational costs associated with governing responsibility, particularly if voters perceived the party as accountable for administrative oversight. FDP gains may reflect a substitution effect among more state-skeptical voters. The increase in LINKE vote shares is less easily explained, but may indicate mobilization in response to perceived institutional marginalization or as a protest vote.

These results suggest that administrative failures can affect not only who votes but also the partisan landscape in which votes are cast. While the magnitude of vote-share shifts is modest, the fact that effects persist across multiple elections underscores the potential for even non-strategic failures in service delivery to influence democratic representation.

Figure 8: Election Results (Vote Share)



Notes: The figure presents event study results based on Equation 1 for the vote share of the political parties represented in the federal parliament *LINKE* (Left), *The Greens*, the *SPD* (Social Democrats), the *FDP* (Liberals), the *CDU* (Christian Conservatives), and the *AfD* (right-wing populist). The reference election ($t - 1$) is the 2019 European election. Confidence intervals are drawn at the 95 percent level using standard errors clustered at the postal precinct level. The point estimates and standard errors underlying the results appear in Appendix Table A12.

5 Theoretical Interpretation: Habit Scarring and Belief Decay

Why do one-off administrative failures lead to persistent declines in voter turnout, even after conditions normalize? We present a simple behavioral framework in which disruptions affect future participation through two reinforcing channels: weakened civic habits and revised expectations about the cost of voting. This helps explain why even a transient failure in electoral service delivery can produce long-run changes in turnout, especially among groups with fragile engagement.

A voter's decision to turn out at time t is governed by a utility function:

$$U_t = p_t B + d_t - c_t, \quad (4)$$

where $p_t B$ reflects the perceived benefit of voting (instrumental or expressive), d_t captures internalized civic motivation or habit, and c_t is the anticipated cost of participation. Voting occurs if $U_t > 0$.

Disruptions—such as long queues, confusion, or ballot shortages—raise c_t and can induce abstention even among otherwise willing voters. If voting is habit-forming, as suggested by [Fujiwara et al. \(2016\)](#) and [Coppock and Green \(2016\)](#), this abstention has dynamic consequences. We model habit formation as:

$$d_{t+1} = \bar{d} + \rho v_t, \quad (5)$$

where $v_t \in \{0, 1\}$ is the turnout decision and $\rho > 0$ captures reinforcement. Abstention in t reduces d_{t+1} , making future turnout less likely. Because elections are infrequent, this habit loss may persist across cycles, particularly for new or marginal voters.

Disruptions may also affect expectations. Let $s = \theta + \varepsilon$ denote a noisy signal about the competence of electoral administration, where θ is true service quality. A negative signal—such as visible failure on election day—induces belief updating. We model perceived cost at $t + 1$ as:

$$c_{t+1} = \bar{c} + \phi(s), \quad \phi'(s) < 0, \quad (6)$$

where $\phi(s)$ captures increased perceived costs due to pessimistic expectations. Even if service quality recovers, the memory of prior failure can raise c_{t+1} , reducing turnout. This mechanism aligns with evidence on expectation-driven voting behavior ([Kawai et al., 2021](#)) and belief formation from public service encounters ([Alsan and Wanamaker, 2018](#); [Lowes and Montero, 2021](#); [Akhtari et al., 2022](#)).

Taken together, this framework implies that turnout at $t + 1$ depends not just on structural factors, but also on prior experience. A single disruption can decrease turnout both by breaking behavioral reinforcement and by inducing belief-based cost inflation. These dynamics are particularly relevant for groups whose voting is less routinized or more sensitive to trust shocks—such as young voters, migrants, or welfare recipients. The model thus offers a plausible explanation for the persistence and concentration of turnout declines observed in the data.

Empirically, we cannot fully disentangle the two mechanisms, but our results are consistent with their joint operation. Heterogeneity analyses show stronger turnout effects among groups likely to have weaker civic routines or higher institutional sensitivity. Moreover, survey data from the GLES suggest that treated voters were more likely to abstain at the last minute and reported lower confidence in electoral fairness—patterns consistent with both habit disruption and belief revision. While not conclusive, this evidence supports the interpretation that administrative failure can disrupt civic behavior through both experiential and informational channels.

Our approach extends the classical calculus of voting ([Riker and Ordeshook, 1968](#)) by incorporating two dynamic channels: behavioral reinforcement and belief revision. Unlike static-cost models, we treat voting as a learned behavior, shaped by past experience with the state. While prior work has examined legal or institutional barriers to participation, we show that non-strategic failures in electoral administration—when salient—can generate long-lasting behavioral change. This has broader implications for the political economy of public services: even isolated administrative failures may generate negative externalities for democratic inclusion.

6 Conclusion

This paper shows that even non-strategic administrative failures on election day can produce lasting negative effects on political participation. Exploiting quasi-experimental variation from Berlin’s 2021 electoral disruptions, we estimate a persistent turnout decline of 1.8 percentage points (2.4%) in affected precincts. These losses are concentrated among young voters, residents with migration backgrounds, and welfare recipients—groups whose civic engagement is already fragile. In-person voting falls sharply, and while some substitution into postal voting occurs over time, it remains incomplete.

Survey evidence suggests that visible failures undermine institutional trust and trigger last-minute abstention, supporting a dual-mechanism framework: disrupted habit formation and belief updating. Heterogeneity analyses further show that voters with lower baseline engagement are most sensitive to such disruptions.

The effects extend beyond participation. Although partisan vote shifts are modest, we document meaningful changes in electoral composition and evidence of administrative learning: polling station capacity expanded in areas previously affected by long queues. These findings underscore that administrative competence is not only a technical issue but a democratic one.

Conceptually, the results bridge behavioral political economy and the literature on state capacity. Elections are recurring encounters with the state, and their quality influences not only what voters do but what they expect. When these encounters fail, however unintentionally, they may disengage the very populations most in need of inclusion.

From a policy perspective, the results are both cautionary and actionable. Administrative failure imposes durable costs on democratic participation, but these costs are avoidable. Low-cost interventions, such as ballot stock buffers, staffing audits, and queue monitoring, could prevent similar breakdowns. As democracies face increasing logistical strain, ensuring that elections run smoothly must be treated as a component of state capacity and a foundation of democratic legitimacy.

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Online Appendix

Race to the Polls: The Berlin Marathon and its Electoral Consequences
by *Marius Kröper*

A Robustness Checks

A.1 Placebo: Randomization Inference

Table A1: Randomized Inference

| | Total Turnout (1) | In-person Turnout (2) | Postal Turnout (3) |
|-------------------------|--------------------------------------|--------------------------------------|-------------------------------------|
| Treatment (t-4) | 0.0138 (0.2945) [0.9710] | 0.3645 (0.2387) [0.1021] | -0.3507 (0.2275) [0.1151] |
| Treatment (t-3) | -0.4513 (0.2826) [0.1502] | -0.1428 (0.2038) [0.5105] | -0.3085* (0.1677) [0.0641]* |
| Treatment (t-2) | -0.3090 (0.3131) [0.3333] | -0.2065 (0.3027) [0.5125] | -0.1025 (0.1634) [0.5726] |
| Treatment (t+0/Federal) | -1.390*** (0.3503) [0.0000]*** | -1.122*** (0.3873) [0.0040]*** | -0.2684 (0.4113) [0.5205] |
| Treatment (t+0/State) | -1.437*** (0.3671) [0.0000]*** | -1.047*** (0.3861) [0.0100]** | -0.3899 (0.4172) [0.3413] |
| Treatment (t+1) | -0.8260** (0.3884) [0.0430]** | -1.162*** (0.2949) [0.0000]*** | 0.3361 (0.3725) [0.4154] |
| Treatment (t+2) | -1.143*** (0.3778) [0.0040]*** | -2.317*** (0.2601) [0.0000]*** | 1.173*** (0.3695) [0.0040]*** |
| R ² | 0.95213 | 0.89287 | 0.92054 |
| Observations | 9,808 | 9,808 | 9,808 |
| Precinct FE | ✓ | ✓ | ✓ |
| Election-District FE | ✓ | ✓ | ✓ |

Notes: The table presents event study results based on Equation 1 for postal turnout, in-person turnout, and total turnout (0-100). The p-values calculated based on the randomized inference approach following Heß (2017) are shown in brackets. The reference election ($t - 1$) is the 2019 European Election. Standard errors are clustered at the postal precinct level and reported in parentheses. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

We utilize a randomization inference approach following Heß (2017), conducting 999 permutations while maintaining the number of treated municipalities within each district. This method assesses whether the observed results could arise by random chance. The results, presented in Table A1, show significant treatment effects consistent with the original analysis, thus reinforcing the reliability of our findings. This approach helps to ensure that the results are due to the actual treatment effect of the irregularities during the 2021 Federal elections and not random variation.

A.2 Treatment Definition: Federal vs. State Disruptions

While the core empirical design aggregates disruptions into a single treatment indicator, election-day irregularities varied substantially in nature and timing. In particular, disruptions occurred during either the Federal or the State election held concurrently in 2021, enabling an indirect exploration of heterogeneity by election tier. The most severe disruptions occurred during the Federal election and involved highly visible administrative failures: precincts accepting ballots after the legal closing time, redirecting voters between locations, or remaining open well past 6:00 p.m. By contrast, precincts affected only at the State level were typically flagged for procedural violations such as missing or misallocated ballots—issues that may have been less visible to voters or corrected before deterring participation. To probe whether the salience of disruptions matters for turnout effects, we estimate separate event-study specifications for three mutually exclusive groups: (i) precincts affected during the Federal election (we primary treatment group); (ii) precincts affected only during the State election; and (iii) unaffected precincts, which serve as the counterfactual. [Figure A1](#) presents the resulting trends. Turnout suppression is concentrated in precincts disrupted during the Federal election: the immediate decline at $t + 0$ is statistically significant and remains directionally negative in subsequent elections. By contrast, precincts with State-only disruptions exhibit no detectable turnout effects, consistent with the view that these disruptions were less salient or behaviorally consequential. This interpretation is supported by descriptive evidence on the nature of irregularities. [Table A2](#) tabulates the frequency of ballot-related errors, operational breakdowns, and violations of legal closing times. Precincts affected during the Federal election were far more likely to experience high-salience disruptions—such as 212 cases of late closing and 101 classified as general disruptions—compared to those affected only at the State level. Taken together, these findings suggest that persistent disengagement is not driven by treatment status, but emerges when administrative failures are both visible and severe enough to undermine voters’ perceptions of procedural legitimacy.

Table A2: Treatment Overview

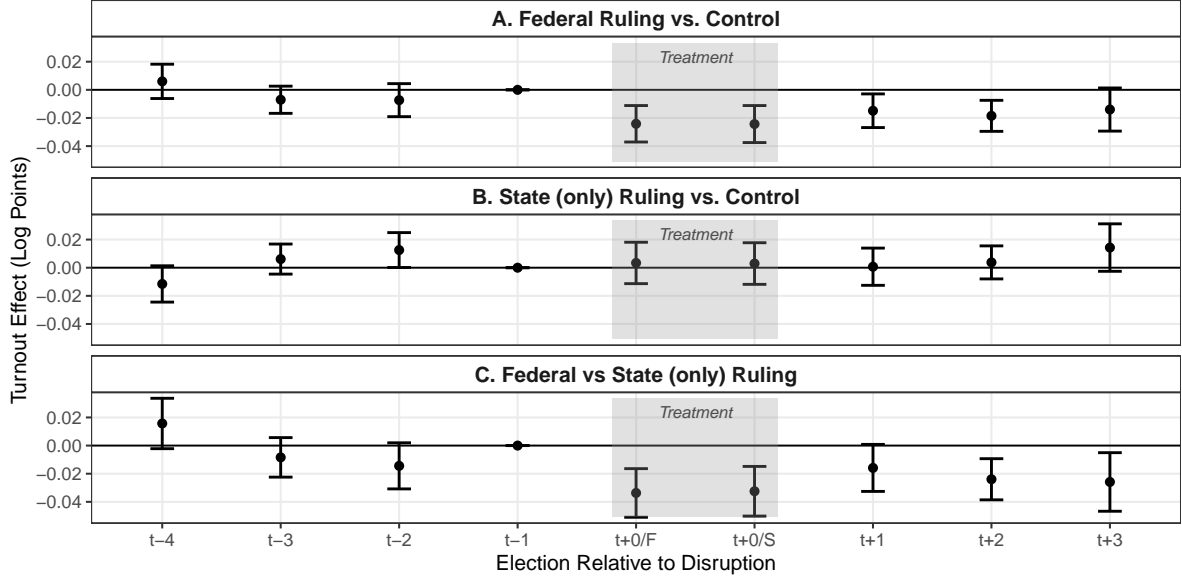
| Treatment Federal | Treatment State | Ballot Error | Ballot Missing | Disruption | Unusual Waiting | Open After Official Closing | N |
|---|--------------------|-----------------|-------------------|------------|--------------------|--------------------------------|-----|
| <i>yes</i> | <i>yes</i> | 32 | 13 | 101 | 90 | 212 | 277 |
| <i>no</i> | <i>yes</i> | 45 | 61 | 2 | 55 | 1 | 157 |
| Subsample: Disruption = 0 & Open A.O. Closing = 0 | | | | | | | |
| <i>yes</i> | <i>yes</i> | 1 | 2 | — | 16 | — | 17 |
| <i>no</i> | <i>yes</i> | 44 | 61 | — | 55 | — | 154 |

Notes: The table presents the number of precincts by treatment during the Federal and State elections. The type of irregularity is available only for the State election.

A.3 Controls

In the main specification, we only control for the log number of eligible voters held constant to the 2017 Federal Election, the last federal election before the treatment, and interact them

Figure A1: Heterogeneity by Type of Treatment



Notes: The figure presents triple-interaction estimates for postal turnout, in-person turnout, and total turnout (0-100). The reference election ($t - 1$) is the 2019 European Election. Confidence intervals are drawn at the 95 percent level using standard errors clustered at the postal precinct level.

with the election indicators. To test the robustness of the main specification, we add further time-invariant controls, holding them constant for 2019. Controls are the share of the following groups: German citizens with migrant backgrounds, foreigners, persons subject to social security, 6-18 year old, 18-25 year old, 65 to 99 year old, EU citizens, persons living in a civil union, Single households; additionally we control for the log number of eligible voters. Results are shown in [Table A3](#)

Table A3: Robustness: Controls

| | Total Turnout (1) | In-person Turnout (2) | Postal Turnout (3) |
|-------------------------|------------------------|--------------------------|-----------------------|
| Treatment (t-4) | 0.0055 (0.0058) | 0.0129** (0.0065) | -0.0098 (0.0122) |
| Treatment (t-3) | -0.0049 (0.0041) | -0.0043 (0.0043) | -0.0040 (0.0080) |
| Treatment (t-2) | -0.0069 (0.0043) | -0.0061 (0.0052) | -0.0061 (0.0066) |
| Treatment (t+0/Federal) | -0.0220*** (0.0063) | -0.0252*** (0.0071) | -0.0181 (0.0141) |
| Treatment (t+0/State) | -0.0221*** (0.0065) | -0.0234*** (0.0071) | -0.0201 (0.0140) |
| Treatment (t+1) | -0.0151*** (0.0057) | -0.0288*** (0.0076) | 0.0049 (0.0137) |
| Treatment (t+2) | -0.0231*** (0.0055) | -0.0618*** (0.0066) | 0.0284** (0.0132) |
| Treatment (t+3) | -0.0193*** (0.0063) | -0.0425*** (0.0075) | 0.0157 (0.0142) |
| R ² | 0.94967 | 0.91415 | 0.93792 |
| Observations | 12,150 | 12,150 | 12,150 |
| Precinct FE | ✓ | ✓ | ✓ |
| Election-District FE | ✓ | ✓ | ✓ |

Notes: The table presents event study results based on [Equation 1](#) for postal turnout, in-person turnout, and total turnout (0-100) controlling for additional variables. The reference election ($t - 1$) is the 2019 European Election. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

A.4 Clustering

In we baseline analysis, we cluster at the level of postal precincts, which is the level at which the treatment is applied. The district is the next higher administrative unit, with precinct administrators reporting to district offices. Concerns may arise that model errors are correlated within State constituencies. To address this, we increase the clustering level to State constituencies, as shown in Column 2 of [Table A4](#). The standard errors are only marginally larger when clustering at the county level. Additionally, two-way clustering—considering postal precincts, and district \times elections results in slightly smaller standard errors (Column 3) compared to clustering at the postal precinct level.

Table A4: Robustness: Standard Errors

| Panel A: Total Turnout | (1) | (2) | (3) |
|-------------------------------------|------------------------|------------------------|------------------------------|
| Treatment (t-4) | 0.0060 (0.0062) | 0.0060 (0.0085) | 0.0060 (0.0088) |
| Treatment (t-3) | -0.0071 (0.0049) | -0.0071 (0.0054) | -0.0071 (0.0065) |
| Treatment (t-2) | -0.0073 (0.0060) | -0.0073 (0.0063) | -0.0073 (0.0072) |
| Treatment (t+0/Federal) | -0.0241*** (0.0066) | -0.0241*** (0.0056) | -0.0241*** (0.0066) |
| Treatment (t+0/State) | -0.0243*** (0.0067) | -0.0243*** (0.0057) | -0.0243*** (0.0067) |
| Treatment (t+1) | -0.0149** (0.0061) | -0.0149** (0.0060) | -0.0149* (0.0077) |
| Treatment (t+2) | -0.0185*** (0.0056) | -0.0185*** (0.0063) | -0.0185*** (0.0060) |
| Treatment (t+3) | -0.0140* (0.0078) | -0.0140** (0.0068) | -0.0140 (0.0096) |
| Standard-Errors | Precinct | Precinct-Election | Precinct & Election-District |
| R ² | 0.92949 | 0.92949 | 0.92949 |
| Panel B: In-person Turnout | (1) | (2) | (3) |
| Treatment (t-4) | 0.0133* (0.0073) | 0.0142** (0.0072) | 0.0142* (0.0072) |
| Treatment (t-3) | -0.0066 (0.0050) | -0.0071 (0.0049) | -0.0071 (0.0049) |
| Treatment (t-2) | -0.0072 (0.0074) | -0.0081 (0.0072) | -0.0081 (0.0072) |
| Treatment (t+0/Federal) | -0.0241*** (0.0091) | -0.0256*** (0.0089) | -0.0256*** (0.0089) |
| Treatment (t+0/State) | -0.0222** (0.0090) | -0.0237*** (0.0088) | -0.0237*** (0.0088) |
| Treatment (t+1) | -0.0278*** (0.0074) | -0.0284*** (0.0074) | -0.0284*** (0.0074) |
| Treatment (t+2) | -0.0619*** (0.0063) | -0.0619*** (0.0064) | -0.0619*** (0.0064) |
| Treatment (t+3) | -0.0438*** (0.0099) | -0.0446*** (0.0097) | -0.0446*** (0.0098) |
| R ² | 0.86623 | 0.86886 | 0.86886 |
| Panel C: Postal Turnout | (1) | (2) | (3) |
| Treatment (t-4) | -0.0089 (0.0124) | -0.0091 (0.0124) | -0.0091 (0.0124) |
| Treatment (t-3) | -0.0062 (0.0085) | -0.0069 (0.0085) | -0.0069 (0.0085) |
| Treatment (t-2) | -0.0044 (0.0070) | -0.0049 (0.0069) | -0.0049 (0.0070) |
| Treatment (t+0/Federal) | -0.0191 (0.0145) | -0.0193 (0.0145) | -0.0193 (0.0145) |
| Treatment (t+0/State) | -0.0214 (0.0145) | -0.0216 (0.0145) | -0.0216 (0.0145) |
| Treatment (t+1) | 0.0084 (0.0142) | 0.0078 (0.0142) | 0.0078 (0.0142) |
| Treatment (t+2) | 0.0424*** (0.0132) | 0.0416*** (0.0131) | 0.0416*** (0.0132) |
| Treatment (t+3) | 0.0333** (0.0148) | 0.0321** (0.0146) | 0.0321** (0.0147) |
| R ² | 0.93030 | 0.93054 | 0.93054 |
| Observations | 12,150 | 12,150 | 12,150 |
| Precinct FE | ✓ | ✓ | ✓ |
| Election-District FE | ✓ | | ✓ |
| Election-District-Federal const. FE | | ✓ | |
| Election-Federal const. FE | | | ✓ |

Notes: The table presents event study results based on [Equation 1](#) for postal turnout, in-person turnout, and total turnout (0-100) using different standard error calculation methods: Column (1) is the baseline and standard errors are clustered at the municipality level. In Column (2), standard errors are clustered at the State constituency level. Column (3) uses two-way clustered standard errors at the level of postal precinct and District \times elections. Standard errors are reported in parentheses. The reference election ($t - 1$) is the 2019 European Election. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

A.5 Matching on Observables

We conduct various matching procedures to ensure the comparability of the treatment and control group, namely propensity score matching, Entropy Balancing as proposed by [Hainmueller \(2012\)](#), and Mahalanobis distance matching. If applicable, we solely refer to values from the year 2017, as this is when the last pre-treatment federal election took place. We re-evaluate the baseline model ([Equation 1](#)) using each matching method and present the findings in [Appendix Table A5](#).

Local Matching First, we use a geographical local matching approach to ensure that the effect is not driven by outliers in the periphery, as postal precincts closer to each other may be more similar. We identify adjacent postal precincts within a district (within 10 meters of the boundaries). We match 295 treated units to 436 untreated units. All treatment effects are close to the baseline estimates and remain statistically significant.

Propensity Score Matching We conduct a propensity score matching procedure to estimate the likelihood of experiencing irregularities during the 2021 Federal elections. This propensity is calculated using a probit regression based on the following pre-treatment socioeconomic characteristics: population measured by the number of eligible voters, share of population aged 6-18, 18-25, 25-35, 35-35, 45-60, 60-70, and over 65, share of foreign population, share of EU citizens, share of females, share of citizens with migrant backgrounds, share of single households, and share of residents subject receiving social security. Additionally, we exact match on the district. All precinct characteristics are measured as of 2019. The matched sample is obtained using 1:1 nearest neighbor matching with replacement. In this process, 294 treated units are matched to 159 control units. The estimates derived from the matched sample confirm we original results for the effect of irregularities on total turnout with the exception of the 2023 State election, which loses significance.

Mahalanobis Matching We check the robustness of we findings by matching treated and untreated units based on the Mahalanobis distance. This method evaluates similarity by measuring the proximity of units' covariates in vector space. We calculate distances using the same pre-treatment covariates as previously mentioned and employ a 1:1 nearest neighbor matching with replacement to create the matched sample. In this process, 294 treated municipalities are matched to 162 control units, while 894 units are dropped. The results indicate that the treatment effects remain significant and are consistent with the outcomes of propensity score matching.

Entropy Balancing We implement the entropy balancing approach proposed by [Hainmueller \(2012\)](#). This method offers the advantage of not truncating the sample. Instead, it assigns a set of weights that balance the treatment and control groups across several moments of the covariate distributions. Specifically, we balance the means and variances of the pre-treatment

characteristics listed above and use the resulting entropy weights in the event study regressions. The results consistently support we original findings across all specifications and outcomes.

Table A5: Matching

| | (1) | Total Turnout (2) | (3) | (4) |
|-------------------------|------------------------|------------------------|------------------------|------------------------|
| Treatment (t-4) | 0.0056 (0.0064) | 0.0120 (0.0074) | 0.0086 (0.0076) | 0.0158** (0.0071) |
| Treatment (t-3) | -0.0034 (0.0050) | -0.0048 (0.0063) | 0.0022 (0.0062) | -0.0061 (0.0060) |
| Treatment (t-2) | -0.0075 (0.0062) | 0.0001 (0.0073) | 0.0002 (0.0075) | -0.0018 (0.0093) |
| Treatment (t+0/Federal) | -0.0240*** (0.0068) | -0.0232*** (0.0082) | -0.0229*** (0.0081) | -0.0221*** (0.0078) |
| Treatment (t+0/State) | -0.0242*** (0.0069) | -0.0248*** (0.0084) | -0.0238*** (0.0083) | -0.0212*** (0.0078) |
| Treatment (t+1) | -0.0131** (0.0061) | -0.0200*** (0.0074) | -0.0132* (0.0075) | -0.0262*** (0.0097) |
| Treatment (t+2) | -0.0216*** (0.0058) | -0.0192** (0.0075) | -0.0088 (0.0075) | -0.0181** (0.0089) |
| Treatment (t+3) | -0.0171** (0.0081) | -0.0107 (0.0099) | -0.0017 (0.0101) | -0.0168* (0.0102) |
| R ² | 0.93035 | 0.92680 | 0.92662 | 0.93399 |
| Observations | 6,561 | 4,086 | 4,104 | 11,034 |
| Specification | Local Matching | Propensity Score | Mahalanobis | Entropy Bal. |
| Precinct FE | ✓ | ✓ | ✓ | ✓ |
| Election-District FE | ✓ | ✓ | ✓ | ✓ |

Notes: The table presents event study results based on [Equation 1](#) total turnout (0-100) using different matching approaches. Column (1) uses nearest neighbour matching based on propensity score, and Column (2) based on Mahalanobis distance. Column (3) uses weights from [Hainmueller \(2012\)](#). The reference election ($t - 1$) is the 2019 European Election. Standard errors are clustered at the municipality level and reported in parentheses. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

A.6 Spatial Sample Construction and Weighting

To assess the robustness of the main findings to different approaches for harmonizing spatial units over time, we re-estimate the core event study specification using two alternative constructions of the panel. Results are presented in [Table A6](#).

Weighted Sample We weight each precinct-election observation by the maximum spatial overlap between a given election-year precinct and the 2021 postal precinct to which it was matched. This approach assigns greater weight to precincts with high spatial continuity across years and reduces influence from more fragmented or ambiguous matches. The results remain statistically and substantively consistent with the baseline: treatment effects are persistent and concentrated in in-person turnout, with partial substitution into postal voting.

Restricted Sample (99.9% Match) We restrict the sample to precinct-year pairs that share at least 99.9% of their population with a single 2021 postal precinct. This yields a smaller sample of 4,827 observations but ensures nearly perfect spatial comparability over time. Turnout effects remain negative and statistically significant for most elections post-treatment, particularly on the postal margin, where the estimated drop exceeds 8 percentage points immediately following

the disruption. This suggests that treatment effects are not driven by imprecise geographic aggregation but reflect real behavioral responses.

Table A6: Spatial Sample Construction and Weighting

| Panel A: Total Turnout | (1) | (2) |
|----------------------------|------------------------|------------------------|
| Treatment (t-4) | 0.0015 (0.0068) | |
| Treatment (t-3) | -0.0073 (0.0058) | |
| Treatment (t-2) | -0.0082 (0.0070) | |
| Treatment (t+0/Federal) | -0.0249*** (0.0067) | -0.0190** (0.0086) |
| Treatment (t+0/State) | -0.0251*** (0.0067) | -0.0192** (0.0086) |
| Treatment (t+1) | -0.0156** (0.0061) | -0.0098 (0.0108) |
| Treatment (t+2) | -0.0215*** (0.0053) | -0.0334** (0.0162) |
| Treatment (t+3) | -0.0129 (0.0085) | |
| R ² | 0.92902 | 0.95172 |
| Panel B: In-person Turnout | (1) | (2) |
| Treatment (t-4) | 0.0084 (0.0081) | |
| Treatment (t-3) | -0.0050 (0.0057) | |
| Treatment (t-2) | -0.0061 (0.0085) | |
| Treatment (t+0/Federal) | -0.0239** (0.0096) | 0.0220** (0.0107) |
| Treatment (t+0/State) | -0.0221** (0.0095) | 0.0239** (0.0108) |
| Treatment (t+1) | -0.0277*** (0.0076) | 0.0183 (0.0130) |
| Treatment (t+2) | -0.0627*** (0.0063) | -0.0391** (0.0190) |
| Treatment (t+3) | -0.0436*** (0.0111) | |
| R ² | 0.86868 | 0.89018 |
| Panel C: Postal Turnout | (1) | (2) |
| Treatment (t-4) | -0.0151 (0.0133) | |
| Treatment (t-3) | -0.0117 (0.0095) | |
| Treatment (t-2) | -0.0098 (0.0080) | |
| Treatment (t+0/Federal) | -0.0219 (0.0142) | -0.0785*** (0.0151) |
| Treatment (t+0/State) | -0.0243* (0.0143) | -0.0808*** (0.0152) |
| Treatment (t+1) | 0.0055 (0.0142) | -0.0510*** (0.0150) |
| Treatment (t+2) | 0.0372*** (0.0128) | -0.0271* (0.0149) |
| Treatment (t+3) | 0.0361** (0.0151) | |
| R ² | 0.93007 | 0.95400 |
| Observations | 12,150 | 4,827 |
| Specification | Weights | Constant |
| Precinct FE | ✓ | ✓ |
| Election-District FE | ✓ | ✓ |

Notes: The table presents event study results based on [Equation 1](#) total turnout (0-100) using two alternative sample constructions. Column (1) weights each observation by the maximum share of population overlap between an election-year precinct and its matched 2021 postal precinct. Column (2) restricts the sample to precinct-year pairs with at least 99.9% population identity with a 2021 postal precinct. The reference election ($t - 1$) is the 2019 European Election. Standard errors are clustered at the municipality level and reported in parentheses. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

B Data Sources and Harmonization

B.1 Main Administrative and Electoral Data

The core dataset consists of precinct-level turnout and party vote shares across all Berlin elections between 2014 and 2025, harmonized via spatial crosswalks. We construct a balanced panel at the level of 2021 postal voting precincts (BWB21), the base geography throughout. For each election, we aggregate turnout and valid votes separately for in-person (UWB) and postal precincts, reconstructing total electorate size and turnout by mode.

The dataset covers eight elections over six election years: European Parliament elections in 2014, 2019, and 2024; state elections in 2016, 2021, and 2023; and federal elections in 2017 and 2021. Precinct shapefiles for each election year are obtained from the Berlin electoral administration. To link precincts over time, we calculate area-weighted crosswalks using high-resolution 100-meter raster population data from the 2022 census (restricted to German citizens aged 18 and older). This ensures that precinct weights reflect electorally relevant populations. To maintain comparability, we exclude 157 precincts affected only by the state court’s ruling.

Electoral results at the precinct level are drawn from official election reports. Data are cleaned, harmonized, and weighted into the 2021 spatial frame using the crosswalks. We construct turnout and party shares relative to eligible voters, and build separate indicators for postal and in-person turnout. We use postal precincts as the unit of analysis since they report full data on voting mode and vote shares and allow clean assignment of treatment status based on rerun designation.

The analysis also incorporates detailed precinct-level socio-demographics from Berlin’s (pre-election) structural 2021 data reports (*Strukturdaten*), including age, gender, migration background, and welfare dependency shares. These are merged via unique precinct identifiers and used in heterogeneity analyses.

Administrative disruption data come from official court decisions on the 2021 Berlin elections, specifying ballot shortages, waiting times, and types of irregularities by polling station. We aggregate these indicators to the BWB21 level and use them to define treatment and exposure intensities. Treated precincts are those flagged in the Federal Constitutional Court’s judgment; court decisions define treatment with legal precision and spatial granularity.

Interpretation of the 2024 Federal Rerun. The 431 precincts identified by the Federal Constitutional Court were subject to a court-ordered rerun of the Bundestag election in February 2024. While we retain these precincts in our analysis panel, we do not treat the rerun as a second intervention. The rerun occurred more than two years after the original disruption and involved only a subset of precincts, making causal inference impossible due to the lack of a control group. Contemporaneous reports and official statistics suggest extremely low turnout and limited campaign activity. The rerun was broadly perceived as a technical correction with no real political stakes. It is therefore unlikely to have reversed the behavioral or belief-based consequences of the 2021 disruption. We interpret the long-run turnout effects observed in

subsequent regular elections as persistent scarring from the initial failure, not as artifacts of later rerun exposure.

B.2 Survey Data: GLES Rolling Cross Sections

To complement the administrative data, we draw on the GLES Rolling Cross-Section (RCS) surveys from 2017, 2021, and 2025. These surveys include repeated nationally representative samples of German citizens during each federal election campaign. We use only respondents aged 18 or older with valid information on residence (**State**), trust in election administration, and nonvoting timing.

The relevant trust item asks whether respondents agree with the statement: *“Die Bundestagswahl wurde von den zuständigen Behörden korrekt und fair durchgeführt”*. We construct three variables: (i) a binary indicator for those who “fully agree”, (ii) a relaxed version including “somewhat agree”, and (iii) an ordinal scale from -2 (“strongly disagree”) to $+2$ (“strongly agree”).

For nonvoters, the timing of abstention is measured via post-election questions. We define two binary indicators: one for individuals who abstained late (on or shortly before election day), and one for those who cited procedural reasons for not voting. These outcomes allow testing for latent cost shocks, as outlined in the behavioral model.

C Supplementary Tables

Table A7: Summary Statistics

| | | | | | | | | |
|--------------------------|-------|-----------|----------|---------|-----------|-----------|-----------|-----------|
| Panel A: Full Sample | | | | | | | | |
| Statistic | N | Mean | St. Dev. | Min | Pctl(25) | Median | Pctl(75) | Max |
| NA | 1,350 | 76.618 | 9.302 | 48.053 | 70.233 | 77.755 | 83.153 | 125.795 |
| NA.1 | 1,350 | 39.742 | 3.604 | 26.594 | 37.419 | 39.783 | 42.055 | 56.059 |
| NA.2 | 1,350 | 36.876 | 8.701 | 13.857 | 30.586 | 37.005 | 42.422 | 86.934 |
| Eligible Voters | 1,350 | 1,658.284 | 621.084 | 24.216 | 1,161.356 | 1,417.478 | 2,158.430 | 3,912.376 |
| SGCwMB | 1,350 | 2.475 | 3.108 | 0.024 | 0.733 | 1.494 | 3.088 | 35.697 |
| Panel B: Treatment Group | | | | | | | | |
| Statistic | N | Mean | St. Dev. | Min | Pctl(25) | Median | Pctl(75) | Max |
| NA | 294 | 77.367 | 7.929 | 48.053 | 73.304 | 78.883 | 82.978 | 100.986 |
| NA.1 | 294 | 39.868 | 3.184 | 31.101 | 37.839 | 39.769 | 41.933 | 48.584 |
| NA.2 | 294 | 37.499 | 7.433 | 14.359 | 33.195 | 38.568 | 42.257 | 62.366 |
| Eligible Voters | 294 | 1,921.515 | 724.212 | 651.134 | 1,245.193 | 1,908.351 | 2,604.584 | 3,912.376 |
| SGCwMB | 294 | 2.833 | 3.203 | 0.026 | 0.878 | 1.979 | 3.583 | 27.875 |
| Panel C: Control Group | | | | | | | | |
| Statistic | N | Mean | St. Dev. | Min | Pctl(25) | Median | Pctl(75) | Max |
| NA | 1,056 | 76.410 | 9.643 | 51.107 | 69.533 | 77.041 | 83.309 | 125.795 |
| NA.1 | 1,056 | 39.707 | 3.713 | 26.594 | 37.363 | 39.803 | 42.177 | 56.059 |
| NA.2 | 1,056 | 36.702 | 9.018 | 13.857 | 30.074 | 36.504 | 42.458 | 86.934 |
| Eligible Voters | 1,056 | 1,584.999 | 568.238 | 24.216 | 1,146.627 | 1,361.326 | 2,008.343 | 3,848.979 |
| SGCwMB | 1,056 | 2.375 | 3.076 | 0.024 | 0.687 | 1.378 | 2.886 | 35.697 |

Notes: The table presents summary statistics for the 2021 Federal election for the main sample, i.e. excluding the district “Tempelhof-Schöneberg” and those municipalities only affected by irregularities in the State election. It is split by treatment and control group for selected variables used in the analyses. *SGCwMB* is the share of German citizens older than 18 with migrant backgrounds.

Table A8: Waiting Time

| Panel A: Total Turnout | (1) | (2) |
|----------------------------|------------------------|------------------------|
| Treatment (t-4) | 0.0060 (0.0062) | 0.0100 (0.0070) |
| Treatment (t-3) | -0.0071 (0.0049) | -0.0035 (0.0053) |
| Treatment (t-2) | -0.0073 (0.0060) | -0.0092 (0.0060) |
| Treatment (t+0/Federal) | -0.0241*** (0.0066) | -0.0219*** (0.0061) |
| Treatment (t+0/State) | -0.0243*** (0.0067) | -0.0223*** (0.0062) |
| Treatment (t+1) | -0.0149** (0.0061) | -0.0068 (0.0071) |
| Treatment (t+2) | -0.0185*** (0.0056) | -0.0141* (0.0072) |
| Treatment (t+3) | -0.0140* (0.0078) | -0.0232*** (0.0082) |
| R ² | 0.92949 | 0.92941 |
| Observations | 12,150 | 10,854 |
| Panel B: In-person Turnout | (1) | (2) |
| Treatment (t-4) | 0.0133* (0.0073) | 0.0169** (0.0082) |
| Treatment (t-3) | -0.0066 (0.0050) | -0.0019 (0.0053) |
| Treatment (t-2) | -0.0072 (0.0074) | -0.0125 (0.0077) |
| Treatment (t+0/Federal) | -0.0241*** (0.0091) | -0.0408*** (0.0095) |
| Treatment (t+0/State) | -0.0222** (0.0090) | -0.0398*** (0.0092) |
| Treatment (t+1) | -0.0278*** (0.0074) | -0.0318*** (0.0091) |
| Treatment (t+2) | -0.0619*** (0.0063) | -0.0466*** (0.0076) |
| Treatment (t+3) | -0.0438*** (0.0099) | -0.0420*** (0.0106) |
| R ² | 0.86623 | 0.86537 |
| Panel C: Postal Turnout | (1) | (2) |
| Treatment (t-4) | -0.0089 (0.0124) | -0.0058 (0.0131) |
| Treatment (t-3) | -0.0062 (0.0085) | -0.0070 (0.0090) |
| Treatment (t-2) | -0.0044 (0.0070) | -0.0024 (0.0057) |
| Treatment (t+0/Federal) | -0.0191 (0.0145) | -0.0042 (0.0136) |
| Treatment (t+0/State) | -0.0214 (0.0145) | -0.0055 (0.0136) |
| Treatment (t+1) | 0.0084 (0.0142) | 0.0200 (0.0162) |
| Treatment (t+2) | 0.0424*** (0.0132) | 0.0217 (0.0159) |
| Treatment (t+3) | 0.0333** (0.0148) | -0.0004 (0.0162) |
| R ² | 0.93030 | 0.92959 |
| Observations | 12,150 | 10,854 |
| Specification | Baseline | Waiting Time in h |
| Precinct FE | ✓ | ✓ |
| Election-District FE | ✓ | ✓ |

Notes: The table presents event study results based on [Equation 1](#) for postal turnout, in-person turnout, and total turnout (0-100). The reference election ($t - 1$) is the 2019 European Election. Treatment is defined as the maximum waiting time within a precinct as reported by the ruling of the *State Constitutional Court*. Standard errors are clustered at the postal precinct level and reported in parentheses. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

Table A9: Heterogeneity by Migrant Backgrounds

| | Total Turnout | | |
|--------------------------------------|------------------------|------------------------|------------------------|
| | (1) | (2) | (3) |
| Treatment (t-4) | 0.0061 (0.0062) | 0.0059 (0.0062) | 0.0051 (0.0056) |
| Treatment (t-3) | -0.0068 (0.0049) | -0.0068 (0.0048) | -0.0061 (0.0044) |
| Treatment (t-2) | -0.0088 (0.0055) | -0.0072 (0.0060) | -0.0055 (0.0044) |
| Treatment (t+0/Federal) | -0.0214*** (0.0064) | -0.0239*** (0.0066) | -0.0228*** (0.0063) |
| Treatment (t+0/State) | -0.0216*** (0.0065) | -0.0242*** (0.0067) | -0.0230*** (0.0064) |
| Treatment (t+1) | -0.0107* (0.0055) | -0.0143** (0.0060) | -0.0155*** (0.0059) |
| Treatment (t+2) | -0.0160*** (0.0055) | -0.0185*** (0.0056) | -0.0183*** (0.0056) |
| Treatment (t+3) | -0.0133* (0.0074) | -0.0139* (0.0078) | -0.0114* (0.0062) |
| Treatment (t-4) $\times Z_i$ | 0.0024 (0.0070) | -0.0041 (0.0054) | -0.0073 (0.0067) |
| Treatment (t-3) $\times Z_i$ | -0.0033 (0.0058) | -0.0013 (0.0046) | 0.0016 (0.0059) |
| Treatment (t-2) $\times Z_i$ | -0.0004 (0.0057) | -0.0025 (0.0049) | 0.0033 (0.0053) |
| Treatment (t+0/Federal) $\times Z_i$ | -0.0133* (0.0077) | 0.0054 (0.0061) | -0.0085 (0.0076) |
| Treatment (t+0/State) $\times Z_i$ | -0.0137* (0.0078) | 0.0030 (0.0069) | -0.0074 (0.0079) |
| Treatment (t+1) $\times Z_i$ | -0.0114* (0.0063) | 0.0112* (0.0064) | -0.0006 (0.0073) |
| Treatment (t+2) $\times Z_i$ | -0.0074 (0.0057) | 0.0014 (0.0048) | -0.0108* (0.0057) |
| Treatment (t+3) $\times Z_i$ | -0.0110 (0.0084) | -0.0025 (0.0068) | -0.0035 (0.0077) |
| Covariate \$Z_i | Migrant | Females | Welfare |
| | Backgrounds | | Recipients |
| R ² | 0.93633 | 0.93023 | 0.94409 |
| Observations | 12,150 | 12,150 | 12,150 |
| Precinct FE | ✓ | ✓ | ✓ |
| Election-District FE | ✓ | ✓ | ✓ |

Notes: The table presents event study results based on Equation 2 for total turnout (0-100). The reference election ($t - 1$) is the 2019 European Election. The share of citizens with migrant backgrounds is held constant in 2019 and is scaled with a mean of 0 and a standard deviation of 1. Standard errors are clustered at the postal precinct level and reported in parentheses. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

Table A10: Heterogeneity by Age

| | (1) | (2) | (3) | Total Turnout | | (6) | (7) | (8) |
|---------------------------------------|-----------------------------------|------------------------|------------------------|------------------------|------------------------|------------------------|------------------------|------------------------|
| | (1) | (2) | (3) | (4) | (5) | (6) | (7) | (8) |
| Treatement (t-4) | 0.0063 (0.0062) | 0.0066 (0.0063) | 0.0063 (0.0061) | 0.0060 (0.0061) | 0.0043 (0.0063) | 0.0037 (0.0060) | 0.0046 (0.0063) | 0.0046 (0.0063) |
| Treatement (t-3) | -0.0083* (0.0050) | -0.0080 (0.0050) | -0.0068 (0.0050) | -0.0052 (0.0050) | -0.0014 (0.0045) | -0.0053 (0.0047) | -0.0041 (0.0048) | -0.0010 (0.0047) |
| Treatement (t-2) | -0.0087 (0.0059) | -0.0083 (0.0062) | -0.0083 (0.0058) | -0.0066 (0.0061) | -0.0026 (0.0058) | -0.0038 (0.0055) | -0.0050 (0.0061) | -0.0031 (0.0061) |
| Treatement (t+0/Federal) | -0.0258*** (0.0066) | -0.0260*** (0.0067) | -0.0226*** (0.0065) | -0.0228*** (0.0067) | -0.0212*** (0.0066) | -0.0223*** (0.0066) | -0.0217*** (0.0066) | -0.0214*** (0.0067) |
| Treatement (t+0/State) | -0.0258*** (0.0067) | -0.0261*** (0.0068) | -0.0228*** (0.0066) | -0.0230*** (0.0068) | -0.0214*** (0.0067) | -0.0225*** (0.0067) | -0.0218*** (0.0067) | -0.0217*** (0.0068) |
| Treatement (t+1) | -0.0151** (0.0062) | -0.0163*** (0.0061) | -0.0127** (0.0057) | -0.0116** (0.0059) | -0.0103* (0.0060) | -0.0167*** (0.0061) | -0.0114* (0.0059) | -0.0091 (0.0059) |
| Treatement (t+2) | -0.0192*** (0.0056) | -0.0190*** (0.0056) | -0.0163*** (0.0055) | -0.0176*** (0.0056) | -0.0188*** (0.0056) | -0.0189*** (0.0056) | -0.0178*** (0.0055) | -0.0184*** (0.0057) |
| Treatement (t+3) | -0.0163** (0.0077) | -0.0150* (0.0080) | -0.0137* (0.0076) | -0.0126 (0.0080) | -0.0097 (0.0078) | -0.0099 (0.0075) | -0.0109 (0.0079) | -0.0107 (0.0081) |
| Treatement (t-4) $\times Z_i$ | 1.96×10^{-5} (0.0057) | 0.0053 (0.0058) | 0.0044 (0.0067) | 0.0049 (0.0051) | 0.0083 (0.0052) | 0.0141*** (0.0050) | -0.0118** (0.0051) | -0.0119** (0.0055) |
| Treatement (t-3) $\times Z_i$ | -0.0109** (0.0046) | -0.0084* (0.0049) | -0.0033 (0.0050) | -0.0075* (0.0045) | -0.0091** (0.0039) | -0.0156*** (0.0043) | 0.0143*** (0.0043) | 0.0105*** (0.0039) |
| Treatement (t-2) $\times Z_i$ | -0.0091 (0.0057) | -0.0085 (0.0064) | -0.0019 (0.0047) | -0.0076 (0.0055) | -0.0116*** (0.0042) | -0.0136*** (0.0046) | -0.0142*** (0.0052) | 0.0126*** (0.0047) |
| Treatement (t+0/Federal) $\times Z_i$ | -0.0132** (0.0060) | -0.0166** (0.0067) | -0.0141** (0.0068) | -0.0132** (0.0056) | -0.0102** (0.0047) | -0.0084 (0.0055) | 0.0149*** (0.0051) | 0.0168*** (0.0056) |
| Treatement (t+0/State) $\times Z_i$ | -0.0112* (0.0061) | -0.0163** (0.0067) | -0.0140** (0.0071) | -0.0129** (0.0056) | -0.0096** (0.0048) | -0.0072 (0.0057) | 0.0154*** (0.0052) | 0.0154*** (0.0058) |
| Treatement (t+1) $\times Z_i$ | -0.0053 (0.0063) | -0.0128** (0.0062) | -0.0064 (0.0063) | -0.0116** (0.0052) | -0.0075 (0.0053) | -0.0181*** (0.0063) | 0.0129** (0.0054) | 0.0139** (0.0057) |
| Treatement (t+2) $\times Z_i$ | -0.0067 (0.0052) | -0.0049 (0.0050) | -0.0121** (0.0049) | -0.0075 (0.0056) | 0.0012 (0.0049) | 0.0015 (0.0048) | 0.0019 (0.0050) | 0.0064 (0.0051) |
| Treatement (t+3) $\times Z_i$ | -0.0170** (0.0080) | -0.0088 (0.0081) | -0.0134* (0.0072) | -0.0171** (0.0071) | -0.0124** (0.0059) | -0.0072 (0.0061) | 0.0216*** (0.0067) | 0.0176*** (0.0065) |
| Covariate \$Z_i\$ | Ages 0-6 | Ages 6-18 | Ages 18-25 | Ages 25-35 | Ages 35-45 | Ages 45-60 | Ages 60-70 | Ages 70+ |
| R ² | 0.93119 | 0.93002 | 0.93293 | 0.93203 | 0.93210 | 0.93350 | 0.93134 | 0.93107 |
| Observations | 12,150 | 12,150 | 12,150 | 12,150 | 12,150 | 12,150 | 12,150 | 12,150 |
| Precinct FE | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ |
| Election-District FE | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ |

Notes: The table presents event study results based on Equation 2 for total turnout (0-100). The reference election ($t - 1$) is the 2019 European Election. The share of citizens with migrant backgrounds is held constant in 2019 and is scaled with a mean of 0 and a standard deviation of 1. Standard errors are clustered at the postal precinct level and reported in parentheses. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

Table A11: Robustness: Controls

| A. " Election Process Correct and Fair" | (1) | (2) |
|---|------------------------|------------------------|
| Berlin 2017 (pre) | -0.0058 (0.0039) | -0.0123*** (0.0032) |
| Berlin 2021 (post) | -0.0930*** (0.0035) | -0.1275*** (0.0040) |
| Berlin 2025 (pre) | -0.0287*** (0.0034) | -0.0075** (0.0026) |
| Berlin 2025 (post) | -0.0175*** (0.0050) | -0.0122*** (0.0031) |
| Specification | Strongly (Dis)agree | (Strongly) (Dis)agree |
| Observations | 20,213 | 31,484 |
| R ² | 0.01346 | 0.01271 |
| B. Timing of Voting Decision: Non-Vote | (1) | (2) |
| Berlin 2021 (post) | 0.5281*** (0.0360) | 0.4950*** (0.0401) |
| Berlin 2025 (post) | -0.0077 (0.0527) | 0.1823** (0.0621) |
| Specification | Prevented | Same Day Prevented |
| Observations | 510 | 510 |
| R ² | 0.04503 | 0.03429 |
| Year FE | ✓ | ✓ |
| State FE | ✓ | ✓ |

Notes: This table reports estimates from [Equation 3](#) using pooled data from the GLES Rolling Cross-Section surveys conducted in 2017, 2021, and 2025. *Panel A* reports respondents' beliefs about electoral integrity, measured via agreement with the statement: “Die Bundestagswahl wurde von den zuständigen Behörden korrekt und fair durchgeführt”. The reference election (2017(post)) is the *pre-election survey* of the 2021 Federal Election. Binary dependent variables equal 1 if respondents “fully agree” (Column 1), or “fully” or “somewhat agree” (Column 2); Column (3) codes responses from -2 (“strongly disagree”) to $+2$ (“strongly agree”). *Panel B* examines the timing of abstention decisions among nonvoters. The reference election (2017(post)) is the *post-election survey* of the 2017 Federal Election. Column (1) codes as 1 those who decided not to vote in the final days before the election or reported being prevented from voting. Column (2) isolates respondents who cited inability to vote as the main reason. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

Table A12: Election Results (Vote Share)

| | LINKE (1) | The Greens (2) | SPD (3) | FDP (4) | CDU (5) | AfD (6) |
|-------------------------|----------------------|------------------------|----------------------|----------------------|------------------------|---------------------|
| Treatment (t-4) | 0.0876 (0.1620) | -0.1042 (0.2046) | -0.2446* (0.1457) | 0.1520* (0.0835) | -0.3271** (0.1601) | 0.3021* (0.1832) |
| Treatment (t-3) | 0.4236** (0.1659) | -0.4175** (0.1925) | 0.2141 (0.1647) | 0.0230 (0.1273) | -0.0321 (0.1552) | -0.1042 (0.1125) |
| Treatment (t-2) | 0.3675* (0.2166) | -0.7758*** (0.2530) | 0.1621 (0.1567) | -0.1119 (0.1330) | 0.2166 (0.1349) | -0.0363 (0.0768) |
| Treatment (t+0/Federal) | 0.2277 (0.1942) | 0.3265 (0.2293) | -0.0839 (0.1768) | 0.0079 (0.1273) | -0.3843*** (0.1489) | 0.1245 (0.1064) |
| Treatment (t+0/State) | 0.6126** (0.2572) | -0.1418 (0.2019) | -0.2936 (0.2029) | 0.1507 (0.0992) | -0.4197** (0.1740) | 0.1059 (0.1026) |
| Treatment (t+1) | 0.6545** (0.2654) | -0.1298 (0.2376) | -0.0989 (0.1491) | 0.1728** (0.0868) | -0.8498*** (0.2903) | 0.0327 (0.1255) |
| Treatment (t+2) | 0.3163 (0.2367) | -0.1925 (0.1881) | 0.1630 (0.1408) | 0.0874 (0.0895) | 0.0550 (0.1452) | -0.1902 (0.1484) |
| Treatment (t+3) | 0.6977* (0.4215) | -0.4019* (0.2357) | 0.0614 (0.1452) | 0.1501** (0.0757) | -0.1117 (0.1680) | -0.2190 (0.1902) |
| R ² | 0.95074 | 0.96863 | 0.91228 | 0.93247 | 0.96058 | 0.94452 |
| Observations | 12,150 | 12,150 | 12,150 | 12,150 | 12,150 | 12,150 |
| Precinct FE | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ |
| Election-District FE | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ |

Notes: The table presents event study results based on [Equation 1](#) for the vote share of the political parties represented in the federal parliament *LINKE* (Left), *The Greens*, the *SPD* (Social Democrats), the *FDP* (Liberals), the *CDU* (Christian Conservatives), and the *AfD* (right-wing populist). The reference election ($t - 1$) is the 2019 European Election. Standard errors are clustered at the postal precinct level and reported in parentheses.

*** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.