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Aid for Incumbents: The Electoral Consequences of COVID-19 Relief*

Jeffrey Clemens[†] Julia Payson[‡] Stan Veuger[§]

Abstract

The COVID-19 pandemic led to unprecedented levels of federal aid transfers to state governments. Did this funding increase benefit state incumbents electorally? Identifying the effect of revenue windfalls on economic voting is challenging because whatever conditions led to the influx of cash might also benefit or harm incumbent politicians for a variety of other reasons. We develop an instrument that allows us to predict allocations to states based on variation in congressional representation. We find that incumbents in state-wide races in 2020, 2021, and 2022 performed significantly better in states that received more relief funding due to their over-representation in Congress. These results are robust across specifications and after adjusting for a variety of economic and political controls. We consistently find that the pandemic-period electoral advantage of incumbent politicians in states receiving more aid substantially exceeds the more modest advantage these politicians enjoyed during pre-pandemic elections. This paper contributes to our understanding of economic voting and the incumbency advantage during times of crisis as well as the downstream electoral consequences of both the COVID-19 pandemic and of unequal political representation at the federal level.

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1 Introduction

In response to the COVID-19 pandemic, the U.S. federal government transferred nearly \$1 trillion in aid to state and local governments—the largest influx of federal money in response to either a public-health or financial crisis in history. The goal was to stabilize the economy while providing states with the necessary resources to address the public health crisis. Did this increase in funding benefit incumbent politicians? We provide evidence on this question by exploiting the fact that states received unequal levels of aid as a means to estimating the effect of federal aid on state-wide elections.

The ability to claim credit for government programs and spending comprises an important source of the incumbency advantage (Mayhew 1974; Erikson 1971; Ansolabehere and Snyder Jr. 2002). There are a variety of reasons why increased pandemic aid might have helped incumbent politicians running for re-election. Incumbents can generate support among constituents by emphasizing their efforts to lobby for additional funding, and if they use the money effectively, voters might reward that work at the ballot box. However, if politicians are not able to use the increased revenues to enact policies and programs that voters prefer, we may not observe an impact of increased aid on incumbent performance.

Studying the effect of economic windfalls on the electoral fortunes of politicians is difficult due to standard endogeneity and omitted-variable concerns. The conditions that result in an influx of federal revenue might either benefit or harm incumbents' electoral prospects for other reasons. To overcome this challenge, we employ an instrumental variables strategy. Following Clemens et al. (2023), we leverage the fact that pandemic assistance varied based on congressional representation, with an additional senator or representative per million residents predicting roughly \$1,000 dollars in additional aid per capita. As a result, overrepresented states received systematically more funding. By predicting aid levels using congressional representation as an instrument, we seek to isolate the effect of COVID-19 spending on state-wide election results. We find that federal pandemic aid strongly predicts incumbent performance in the 2020, 2021, and 2022 elections. The exclusion restriction here assumes that higher per capita representation in Congress did not benefit incumbents through channels other than the increase in COVID-19 funding. If incumbents in these overrepresented states enjoy a persistent advantage, this advantage will be present in pre-pandemic elections. In an analysis of a decade-long panel, we find that incumbents in overrepresented states did, in fact, enjoy a small electoral edge even before the pandemic's onset, but that their advantage rose significantly—both in electoral and statistical terms—after federal pandemic aid had been appropriated. We show further that our representation instrument is more or less orthogonal to variations in partisan preferences, proxies for the pandemic's effects, and proxies for variations in the pandemic's potential severity. Put differently, our estimates are robust to controlling for a number of potential threats to our estimation framework's validity.

This paper makes three primary contributions. First, our research design allows us to study the effect of a dramatic economic shock on retrospective voting and the incumbency advantage. Second, our analysis contributes to our understanding of the political effects of COVID-19. Several scholars have begun to study the electoral effects of the pandemic, but existing work has focused primarily on the presidential election of 2020 (Clarke, Stewart, and Ho 2021; Baccini, Brodeur, and Weymouth 2021; Mendoza Aviña and Sevi 2021). While some comparative literature has examined the consequences of COVID-19 at the subnational level (Constantino, Cooperman, and Moreira 2021; Garnett et al. 2023; Justinus and Dartanto 2024), we know less about the downstream and statewide effects of the pandemic in the United States. Through analyses of campaign contributions and analyses that compare the effects of aid on electoral outcomes for governors relative to senators, as well as for down-ballot legislative and executive offices, we shed additional light on the mechanisms through which incumbents benefited from aid in this setting. Finally, this paper adds to a body of research demonstrating that representation matters for the allocation of resources (Ansolabehere, Gerber, and Snyder 2002; Lee 1998). We document how the consequences of this uneven distribution can shape electoral politics.

2 Existing Literature and Institutional Background

Across both state and federal offices in the United States, incumbent politicians tend to perform well at the ballot box. A large literature on distributive politics posits that constituents reward incumbent politicians for securing federal dollars (Levitt and Snyder Jr. 1997; Bickers and Stein 1996; Cain, Ferejohn, and Fiorina 1987).¹ Incumbents can also strategically claim credit for government spending in order to cultivate a personal vote (Grimmer, Messing, and Westwood 2012). At the same time, local economic conditions strongly predict incumbent support (de Benedictis-Kessner and Warshaw 2020), and the incumbency advantage is larger in states with larger per capita legislative operating budgets (Hirano and Snyder 2009).

A substantial body of research on retrospective voting has established that voters generally hold incumbents accountable for the economy's performance while they are in office, penalizing them for poor economic outcomes and rewarding them for positive ones (for a review, see Healy and Malhotra (2013)). Economic downturns lead voters to embrace challenger parties and reduce support for incumbents (Gourevitch 1986; Bartels 2014; Healy and Lenz 2017). Conversely, voters reward politicians for windfalls (Chen 2013; Bechtel and Hainmueller 2011). Early political science research on retrospective voting focused primarily on establishing patterns of how economic conditions correlate with vote choice (e.g. Tufte 1978). The goal was often prediction rather than inference. For example, Niemi, Stanley, and Vogel (1995) find that a one standard deviation increase in real income per-capita within

¹A recent analysis from a health care policy context illustrates how congressional representatives were rewarded for supporting the narrowly passed Medicare Modernization Act (MMA) of 2003. Specifically, Cooper et al. (2024) find that hospitals in the relevant representatives' districts received Medicare payment increases through the MMA's Section 508 program. These payment increases were, in turn, associated with increases in campaign contributions around the time the Section 508 hospital payment increases required reauthorization.

a state was associated with an 8 percentage point increase in the incumbent's vote share in gubernatorial elections.

More recent work in political economy focuses on exploiting shocks ranging from the Great Depression to the China trade shock to study how changes in economic conditions affect political outcomes. Margalit (2011) finds that counties exposed to more foreign competition due to offshoring experienced greater job loss, and that each percentage point decrease in the employment rate corresponded with a 0.15 percentage point penalty for the incumbent party in presidential elections. Healy and Lenz (2017) study the 2008 financial crisis and find that negative economic shocks harm incumbents: zip codes with the highest levels of delinquent mortgages, for example, shifted their vote share away from the incumbent party by 7.7 percentage points relative to zip codes with the lowest delinquency rates.²

Together, these stylized facts suggest that an influx of funding should increase electoral support for incumbent politicians. However, whatever political and economic conditions led to the windfall might also influence voter evaluations through other channels, making it difficult to credibly estimate the effect of government spending on incumbent performance (e.g. Healy and Malhotra 2009). In this paper, we develop an instrument to credibly estimate how unprecedented levels of federal aid to state governments influenced the electoral fortunes of incumbent politicians.

The United States' system of fiscal federalism makes transfers from the national government an important component of state budgets. The COVID-19 pandemic resulted in historically high levels of federal aid being transferred to state and local governments (Clemens and Veuger 2020*a*). Notably for our analysis, transfers were more generous towards states with higher per capita representation in Congress (Clemens and Veuger 2021). Importantly, a state's congressional representation is not proportional to its population, as each state elects two senators and at least one member of the House of Representatives. Clemens et al. (2022a and 2022b) exploit this representational bias and use an instrumental-variables

²For additional examples, see Margalit (2019).

approach to estimate the effect of aid on state and local government employment and the rollout of COVID-19 testing and vaccination operations.

Here, our starting point is a similar design to estimate the effect of COVID-19 aid on incumbent performance in statewide elections in 2020, 2021 and 2022. Other literature has found that over-represented states enjoy additional federal funding per capita across a range of measures (Atlas et al. 1995; Lee 1998; Hauk Jr. and Wacziarg 2007). Consistent with these findings, we document that states with higher per capita representation enjoyed a slightly larger incumbency advantage prior to the pandemic. However, after nearly \$1 trillion of COVID-19 relief was appropriated, states that received more aid saw a disproportionate increase in incumbent vote share. We describe our empirical approach in detail in the next section.

We study both legislative and executive incumbents whose constituency is an entire state, including senators, members of the House elected at large, and governors. The literature suggests that legislators might enjoy a stronger incumbency advantage, in part because it can be difficult to identify which individual legislators are responsible for government performance. Legislators can exploit the fact that voters may not know who to blame for a weak economy or public health crisis and typically engage in more constituent casework and credit claiming, which may boost their electoral advantages (Fiorina 1989). Governors, on the other hand, are more likely to be held accountable for economic performance because executive decisions are more easily attributable to a single politician. Some existing work has found that members of the Senate are more likely to be evaluated on the basis of presidential performance, while governors are punished or rewarded based on a state's economic conditions (Atkeson and Partin 1995).

Whether legislators or executive officers should benefit more from the distribution of pandemic aid is ultimately an empirical question (cf. Ansolabehere and Snyder Jr. (2002)). While our baseline analysis pools across office types in order to maximize our sample size, we also explore differences across legislative and executive offices when assessing mechanisms. We find that governors of overrepresented states see their incumbency advantage increase more during the pandemic than legislators. Further, in analyses of lower-level offices, we find null effects of federal aid on the composition of state legislatures and substantial effects on the performance of incumbent secretaries of state, attorneys general, and lieutenant governors. This is in line with previous work that assigns credit for state level outcomes to governors, with additional evidence that benefits flow to executive office holders more broadly. Additionally, we note that the crisis may also have increased gubernatorial visibility through greater media attention. In "normal" times, these roles appear to be reversed and, if anything, legislators benefit more from overrepresentation, in the spirit of Fiorina (1989).

Several features of our case make it a unique opportunity to study the effects of federal aid on political outcomes. At the pandemic's outset, estimates from a number of sources projected that state and local revenue shortfalls would rise easily into the hundreds of billions and might reach as high as \$1 trillion dollars (Auerbach et al. 2020; Bartik 2020; McNichol, Leachman, and Marshall 2020; Clemens and Veuger 2020b,a; Whitaker 2020). We now know that, for a number of reasons, these estimates substantially overstated the revenue shortfalls that would ultimately occur. First, states' tax bases were buoyed by federal support for households and businesses, much of which had not been legislated at the time of these early-pandemic forecasts. Second, both state and local sales tax revenues were enhanced by the pandemic's effect on consumption patterns (de Benedictis-Kessner and Warshaw 2020), which underwent an unforeseen shift away from services (which are disproportionately untaxed) and towards goods (which are disproportionately taxed). In the end, state governments' revenues would ultimately exceed rather than fall short of pre-pandemic forecasts (National Association of State Budget Officers 2021). For our purposes, an implication of these developments is that federal aid generated surpluses over which governors and state legislatures held substantial discretion. This can be contrasted with times of starker need, when politicians are more constrained to devote any additional revenue towards the maintenance of employment and pay for essential personnel.

Existing research on the political effects of the COVID-19 pandemic in the United States has so far focused on the presidential election of 2020. Voters reacted negatively to Trump's handling of the pandemic (Clarke, Stewart, and Ho 2021), and Baccini, Brodeur, and Weymouth (2021) found that pandemic death rates negatively predicted Trump's vote share at the county level. Self-exposure to COVID-19 cases and deaths similarly correlate with lower support for the former president (Mendoza Aviña and Sevi 2021). However, we are not aware of any papers that have studied the consequences of pandemic aid on post-2020 elections for other offices. In the next section, we introduce our data and research design.

3 Data

Our primary outcome of interest is incumbent-party vote share. We construct this outcome by taking the incumbent party's total number of votes as a share of the top two candidates' total votes for a particular office.³ MIT's Election Lab provides vote counts for congressional elections through 2020, while we use vote counts from Amlani and Algara (2021) for gubernatorial elections through 2020. Vote counts for the 2021 and 2022 elections are taken from Leip (2024). Our sample is comprised of Senate and gubernatorial elections nationwide and House elections for the six states with at-large (state-wide) congressional districts. The resulting sample of 131 elections from 2020, 2021, and 2022 is depicted in the maps displayed in Figure 1, which are shaded to provide an initial look at the incumbent party's vote share in each election. The broader sample in which we contrast pandemic elections with prepandemic elections incorporates an additional 217 Senate, gubernatorial, and at-large House races, such that our decade-long sample incorporates 348 elections from 2013 through 2022.

³In cases in which more than one candidate from the incumbent party runs in an election, we take the top-performing incumbent party candidate's total votes to be equal to the incumbent party's total votes.

We use a state's number of congressional representatives per million residents as our measure of congressional representation. Rosters of the House of Representatives and Senate during the 116th and 117th Congresses come from Lewis et al. (2021). We note that because 2020 Congressional representation was allocated according to state population from the 2010 census, Congressional representation is not affected by variations in population driven by the COVID-19 pandemic. Our baseline specifications incorporate a common control from the literature on U.S. electoral politics, namely the "normal vote," which accounts for the tendency of voters in particular states to support one party over the other. We construct the normal vote as the vote share received by the incumbent party in the most recent pre-COVID-19 pandemic election.⁴

Our measure of federal aid to state and local governments reflects spending authorized by the four major pieces of relief legislation that were passed during the COVID-19 pandemic: the CARES Act, the Families First Coronavirus Response Act (FFCRA), the Response and Relief Act (RRA), and the American Rescue Plan Act (ARPA). In particular, our analysis focuses on the nearly \$1 trillion in funds that were allocated by these bills to state and local governments. As in Clemens and Veuger (2021), data from the Committee for a Responsible Federal Budget (2021) form the basis of our fiscal assistance variable, supplemented by information from several additional sources.⁵ Our analysis focuses on the grand total of aid committed to each state across all four major pieces of COVID-19 fiscal relief. That is, our main independent variable is the grand total of aid allocated to each state per resident in thousands of dollars. Variations in these aid distributions across states are displayed in

⁴For elections that occur every two years, which in our sample include the at-large House races and the New Hampshire and Vermont gubernatorial elections, we use the incumbent party's vote share from the second-most recent election.

⁵We use data from the CRFB's COVID-19 Money Tracker as of August 19th, 2021. As in Clemens and Veuger (2021), "[w]e obtain information on the distribution of transit funds for the RRA and ARPA from the US Federal Transit Administration (2021). Data on the allocation of ARPA assistance to nonpublic schools come from the US Office of Elementary and Secondary Education (2021). We obtain estimates of ARPA section 9817 matching increases from Chidambaram and Musumeci (2021). We approximate the allocation of ARPA section 9819 federal matching funds for uncompensated care using FY2021 estimates of federal disproportionate share hospital allotments by state from the Medicaid and Chip Payment Access Commission (2021)." The Coronavirus Capital Projects Fund outlined in ARPA is distributed according to guidance from the United States Department of the US Department of the Treasury (2021).

Figure 2. Summary statistics for the variables used in our primary analyses can be found in Table A.1.

In supplemental analyses of mechanisms and additional dimensions along which pandemic aid may have impacted electoral outcomes, we estimate the relationship between federal aid and COVID-19 case and death rates, disposable income, unemployment rates, voter turnout, campaign contributions, electoral outcomes in state legislatures, and electoral outcomes for down-ballot statewide offices. Our measures of COVID-19 cases and deaths are cumulative totals through December of the relevant election year (Dong, Du, and Gardner 2020). Our measures of disposable income and the unemployment rate come from US Bureau of Economic Analysis (2024) and US Bureau of Labor Statistics (2024), respectively. Data on voting eligible population is taken from the University of Florida Election Lab. Our data on campaign contributions come from Open Secrets (congressional races) and FollowTheMoney. Our data on down-ballot statewide offices come from Ballotopedia, and our data on electoral outcomes in state legislatures come from the National Conference of State Legislatures. Summary statistics for this set of variables can be found in Table A.2.

4 Methods

The goal of our analysis is to estimate the causal effect of federal aid to state and local governments on the electoral fortunes of incumbents. A general difficulty in estimating the effects of pandemic fiscal assistance is that fiscal assistance may have been targeted, at least to some extent, towards the states in greatest need. If state needs linked to the pandemic's health and economic impacts also influenced incumbent performance, a naïve regression of electoral outcomes on aid would tend to yield estimates that are biased towards negative values.

As a solution to this endogeneity problem, we propose an instrumental-variables estimation framework that makes use of the fact that federal aid distributions were far more generous to states that enjoy over-representation in the U.S. Congress, due in large part to the U.S. Senate's over-representation of individuals from low-population states. The initial strategy we implement is described by the following set of equations:

$$\frac{\text{Total Aid}_s}{\text{Pop}_s} = \gamma_0 + \gamma_1 \text{Reps Per Million}_s + \gamma_2 \text{Normal Vote}_{s,o,t} + X_{s,o,t}\gamma + \epsilon_{s,o,t}$$
(4.1)

Vote Share_{*s,o,t*} =
$$\beta_0 + \beta_1 \frac{\text{Total Aid}_s}{\text{Pop}_s} + \beta_2 \text{Normal Vote}_{s,o,t} + X_{s,o,t}\beta + u_{s,o,t}$$
 (4.2)

In equations (4.1) and (4.2), s indexes states, o indexes offices (senator, governor, or atlarge representative), and t indexes years. The observations in our primary analysis sample consist of the 131 Senate, governor, and at-large House races that occurred in 2020, 2021, and 2022. Total Aid_s is the total per capita federal pandemic aid (in thousands of dollars) to state and local governments in state s.⁶ Reps Per Million_s is our instrument, a measure of the representatives and senators each state is allocated per million residents. Normal Vote_{s,o,t} is a conventional control from the elections literature which captures the performance of the incumbent (or the incumbent party) in the prior election cycle for a given race. While Normal Vote_{s,o,t} is the only additional covariate in our baseline specification, we also explore robustness checks in which proxies for potential sources of bias are included in the additional covariate vector $X_{s,o,t}$.

The validity of our instrumental variables estimation framework depends on two factors. A first requirement is that congressional representation must be a strong, or relevant, predictor of the amount of aid each state received per resident. This fact has been established by Clemens and Veuger (2021), who explain how the small-state bias in federal funding arose in large part from the use of floor functions in the otherwise population-driven formulas for distributing general purpose fiscal relief.

 $^{^{6}}$ In Appendix Table A.3 we check to confirm that we obtain very similar results whether we apply the grand total of aid across the four relief packages to each election or, alternatively, apply aid from only the CARES Act and FFCRA to the elections that took place in November of 2020.

The formal test of our instrument's strength involves the F-statistic on the excluded instrument in the first stage of our specifications. As shown in Table 1, the relevant F-statistics exceed 100, with an additional representative or senator per million residents predicting roughly \$1,000 in additional aid per state resident. Additionally, as shown in Clemens and Veuger (2021) and subsequent studies, the strength of the first-stage relationship is little impacted by adding any of a number of covariates to the regression model.

A second requirement is that our instrument must satisfy the exclusion restriction. That is, conditional on any additional covariates in our model, our instrument must only be correlated with election outcomes through its effect on federal aid distributions. A number of pieces of evidence support the plausibility of this assumption.

First, earlier work has shown the variation in federal aid driven by over- and underrepresentation was unrelated to a number of plausible correlates of the needs states faced as a consequence of the pandemic. Clemens and Veuger (2021) show, in particular, that the small-state advantage is more or less orthogonal to state and local government funding needs as proxied by forecasts of pandemic-driven revenue shocks, pandemic-driven economic shocks, and the size of their public sector at baseline. This earlier paper showed that controlling for these proxies for need has little effect on the relationship between federal aid distributions and our instrument. It is thus unlikely that any incumbency advantage we estimate is in fact caused by these or similar other factors.

Second, we directly explore the robustness of our analysis by implementing straightforward approaches to rule out a role for some of the primary dimensions along which the pandemic differentially impacted states' economies. First, as is widely recognized, tourismintensive states like Nevada, Hawaii, and Florida suffered more dramatically from the pandemic's initial impacts on their overall economic activity. Second, as noted by Clemens et al. (2024), the pandemic's early impacts on oil and gas prices, as well as on the initiation of new resource extraction activity, had a substantial impact on the revenues of Alaska, Wyoming, and North Dakota, which rely to a far greater degree on severance and other resource-related revenue streams than other states. We explore robustness to the potential relevance of these issues by showing that our results are little changed if we drop the most impacted states from the sample. We also show that our results are robust to controlling for plausibly exogenous pre-pandemic proxies for variations in political and pandemic-policy preferences, which may themselves have exerted non-trivial impacts on political outcomes during the pandemic itself. Additionally, we show that our results are robust to controlling directly for population density, which may have directly influenced the pandemic's severity.

Third, as a placebo test we investigate whether the variations in aid that are predicted by our instrument predict the performance of incumbent politicians in elections from 2013 through 2019. This exercise provides evidence on whether incumbents in over-represented states enjoy a persistent electoral advantage relative to incumbents in under-represented states, which would be consistent with the hypothesis we emphasize throughout given that the over-representation of low-population states is a structural feature of the U.S. Congress.

Our analysis of pre-pandemic elections finds that the relationship between election outcomes and the aid predicted by our instrument is weaker in this placebo test sample than in our primary analysis sample. While the relationship between our instrument and prepandemic election outcomes of incumbents is statistically modest, however, it is nonetheless suggestive that incumbents in over-represented states might enjoy a persistent advantage. To ensure that our estimates capture the advantage politicians in over-represented states enjoyed due specifically to the pandemic relief packages, we thus implement an additional set of analyses. Specifically, we implement a set of panel models in which we directly compare the electoral advantage of incumbent parties in over-represented states during the pandemic relative to their performance across several pre-pandemic electoral cycles. We begin this analysis by estimating the model below:

Vote
$$\text{Share}_{s,o,t} = \alpha_0 + \alpha_1 \text{Reps Per Million}_s + \alpha_2 \text{Reps Per Million}_s \times \text{Pandemic}_t + \alpha_3 \text{Pandemic}_t + \epsilon_{s,o,t}$$

$$(4.3)$$

where Pandemic_t is an indicator that takes a value of 1 in 2020, 2021, and 2022 and a value of 0 in earlier years. Equation (4.3) can be described as a relatively sparse, reduced-form analysis of the relationship between our instrument and electoral outcomes in which we allow that relationship to differ during the pandemic relative to pre-pandemic elections. As with our instrumental-variables framework, we proceed with additional analyses that augment equation (4.3) by controlling for Normal Vote_{s,o,t} and, further, by allowing the relationship between Normal Vote_{s,o,t} and electoral outcomes to vary across election cycles.

We then further augment our panel estimator to include an increasingly saturated set of fixed effects that rule out the possibility that our estimates are driven by persistent state wide, or even state-by-office level, incumbency advantages. That is, we begin this final wave of analyses by adding full sets of state fixed effects and time fixed effects, then additionally adding office fixed effects, and finally adding state-by-office fixed effects, as in the model below:

Vote
$$\text{Share}_{s,o,t} = \phi_s + \phi_t + \phi_o + \phi_s \times \phi_o + \rho \text{Reps Per Million}_s \times \text{Pandemic}_t$$

+ $\gamma_t \text{Normal Vote}_{s,o,t} \times \text{Year}_t + \epsilon_{s,o,t}$ (4.4)

Note that in contrast with equation (4.3), equation (4.4) excludes the main effect of Reps Per Million_s because it would be colinear with the set of state fixed effects. The coefficient ρ in equation (4.4) is a reduced form estimate of the incremental, pandemic-specific electoral advantage of enjoying an additional senator or representative per million state residents. Note that because an additional representative predicts roughly \$1,000 in additional aid, as estimated in the next section using equation (4.1), the scaling of ρ relative to β_1 from equation (4.2) is essentially the same. Crucially, the estimate of ρ in equation (4.4) is not subject to potential biases associated with time-invariant factors that differentiate over- and under-represented states. Plausible sources of bias would need to involve pandemic-specific factors that differentially influenced over and under-represented states. These are

the potential biases we seek to address with the robustness checks to which we subject our estimate of equations (4.1) and (4.2).

5 Results

Table 1 presents our baseline results. In column 1, we estimate the reduced-form relationship between federal representation and incumbent vote share. We find that an additional representative or senator per million residents is associated with a statistically significant and politically meaningful additional four percentage points of the two-candidate vote share for the incumbent. Our estimate of the first-stage equation (4.1), in column 2, highlights the strong relationship between over-representation at the federal level and pandemic aid to state and local governments, which should come as no surprise given the discussion above in section 4. Conveniently for the scaling and interpretation of the reduced-form estimates, an additional representative or senator per million residents roughly translates into an additional \$1,000 in aid per capita.

Column 3 shows our estimate of equation (4.2), the 2SLS relationship, without any controls. An additional \$1,000 in federal aid per capita translates into a large and precisely estimated four percentage point boost of the two-candidate vote for the incumbent. In columns 4 through 6 we go through the same exercise, but this time we control for the normal vote. Inclusion of the normal vote, a conventional control that produces our preferred specification, reduces the size of our estimate of the effect of additional aid by about a quarter. The estimated effect of three additional percentage points of the two-candidate vote for the incumbent for each \$1,000 in federal aid remains meaningful and statistically significant.

Table 2 presents the results of our first set of robustness tests. The first column replicates our preferred specification: that of column 6 from Table 1, which is the estimate of (4.2) with the normal vote control included. To investigate whether our estimates are influenced by variations in the severity of the pandemic's toll on states' economies and tax bases, column 2 drops the most natural-resource intensive states, column 3 drops the most tourism intensive states, and column 4 drops both of these categories of states that were hard hit by the economic turmoil of the early pandemic. Our estimated effects of the impact of additional aid on the remaining subsamples of states are, if anything, slightly larger than the estimate based on the full sample.

In Table 3, we introduce two additional controls. After replicating our preferred specification in column 1, we control for the stringency of COVID-19 controls in March 2020 in column 2. In column 3 we control for Donald Trump's vote share in the 2016 presidential election. Finally, in column 4 we control for both of these variables, which we interpret as proxies for pandemic-related political and policy preferences. As can be seen in the table, the Trump vote share and, to a lesser extent, the measure of March 2020 COVID-19 lockdown stringency are both predictive of incumbents' vote shares, but their inclusion has no impact on our coefficient of primary interest. This reflects the fact that variation in federal aid was essentially orthogonal to these additional sources of variation in incumbents' electoral fortunes.

In Table 4, we introduce an additional set of controls that speak to geographic features including population density, population, and square mileage. Density speaks directly to the potential concern that our instrument might be correlated with state attributes with direct relevance to the pandemic's potential severity, while density's numerator (population) and denominator (square mileage) provide additional proxies for potentially relevant aspects of states' geographies. We again begin in column 1 by replicating our preferred specification. Column 2 then adds the density control, column 3 the population control, column 4 the square mileage control, and column 5 adds all three of these covariates at once. While density and population are predictive of incumbents' vote shares, their inclusion has little impact on either the primary coefficient of interest or on the strength of the first stage relationship.

We also consider our results' robustness to issues related to functional form and to the election years included in our sample. In Table A.3, we replace the grand total of federal aid with the running total of aid, such that the federal aid we link to the 2020 elections stems primarily from the CARES Act. The resulting estimates are modestly larger than those in Table 1.⁷ Additionally, we find very similar results if we exclude the 2020 elections from the sample, such that our sample only incorporates elections that occurred after all four relief packages had been passed (results not shown).

A somewhat different approach to ensuring that our results are not a mere statistical fluke is presented in Table 5. Here we run a placebo test: we examine whether our instrument and the predicted variations in federal aid correlate with the vote share of incumbent politicians in 2013-2019, which pre-date the pandemic. The question this test answers is whether elected officials in over-represented states enjoy a structural electoral advantage, in addition to or instead of a temporary advantage connected to the pandemic. We find weak evidence that incumbent politicians do experience a modest, long-running advantage in over-represented states, whether we look at the reduced form or instrument for pandemicera aid. Notably, this advantage is not significantly different from zero at conventional levels of statistical significance. In addition, the size of the effect is much smaller than the boost received by incumbents of these states during the pandemic. In the equivalent of our baseline specification, where we control for the normal vote based on earlier elections, the estimated advantage of incumbents in states with an additional senator or representative per capita is a statistically insignificant 1 percentage point in pre-pandemic elections. This contrasts with the 3 to 4 percentage point advantage we estimate during the pandemic.

The scatterplots presented in Figure 3 provide an additional, transparent look at the reduced form relationship between our instrument and incumbent performance during both the pandemic (panels A and C) and pre-pandemic (panels B and D) elections in our sample.

⁷Our preference for using the grand total of aid reflects the fact that, at the time of the 2020 elections, additional relief packages were under debate and, although their magnitude was uncertain, they could reasonably have been forecast to retain the earlier packages' bias in favor of over-represented states. We are reassured that the estimates in Tables 1 and A.3 are very similar, however, it is not obvious how voters' expectations regarding future aid packages should be taken into account.

The slopes of the best fit lines in panels A and C are distinctively steeper than those in panels B and D, indicating a much stronger relationship between our instrument and the electoral fortunes of incumbents during the pandemic. This is consistent with the estimates from Tables 1 and 5, as discussed above. The scatterplots in panels A and B present the bivariate relationship between our instrument and incumbent vote shares, while the scatterplots in panels C and D present data that are residualized with respect to the normal vote. The relationships as presented in panels C and D thus correspond to our baseline specification, which uses the normal vote variable as a control for the incumbent party's performance in the previous election. Notably, the relationship between our instrument and incumbents from over-represented states enjoy an advantage relative to their counterparts from under-represented states during the prepandemic period, we take additional steps to ensure that any structural advantages of this sort are not influencing our estimates.

We proceed by investigating directly whether the effect of over-representation on incumbent performance during the pandemic does in fact stand out relative to normal times. We investigate this question in Table 6, where we present further estimates of the panel specifications of equations (4.3) and (4.4), and confirm that incumbent elected officials from over-represented states do indeed benefit disproportionately from their over-representation in the 2020-2022 period.

This result holds across the full set of specifications in Table 6. Column 1 presents the simple panel specification of equation (4.3) for the full 2013-2022 period. The estimated coefficient on the interaction between "Reps per Million" and the indicator for the pandemic years indicates that an additional representative or senator per million residents during the pandemic years gave incumbents an additional edge of close to two percentage points—recall that this corresponds almost precisely to the effect of an additional \$1,000 in federal aid per capita. Note that the coefficient on "Reps Per Million" in this specification measures the

impact of representation outside the pandemic years. This coefficient is, as it should be, identical to that in Column 1 of Table 5, which motivated our analysis of the full panel.⁸

As we saw in our cross-sectional analysis as well, the effect of aid on incumbent vote shares is reduced somewhat when we control for the normal vote in column 2, but remains politically and statistically significant. Column 3 presents a more flexible specification that lets the normal vote vary in its predictive value across election cycles. The resulting estimate of the pandemic-era impact of over-representation resembles that in column 1.

Columns 4 through 6 of Table 6 present estimates generated by the augmented panel estimator of equation (4.4). Column 4 introduces year and state fixed effects; column 5 adds office fixed effects; and column 6 appends state-by-office fixed effects. The point estimates of the effect size we find here moderately exceed those of columns 1 through 3 and remain significant at the 99% confidence level. The amount of increased support for incumbents from over-represented states during the pandemic turns out to be robust to this increasingly demanding battery of controls and remains within the range of estimates we have found previously.

6 Mechanisms and Implications

Having established the existence of a pandemic-era electoral boost for incumbents from states with over-representation at the federal level, we now seek to explain the channels through which aid may have directly influenced voters' well being, to assess the relevance of additional mechanisms that may be at work, and to assess the implications of federal aid for election outcomes.

 $^{^{8}}$ The attentive reader may also have noticed that the sum of the the coefficients on Reps per Million x Pandemic (1.803) and Reps per Million (2.280) equals the reduced-form coefficient from Table 1 (4.083).

6.1 Channels of Influence on Voter Well-Being

In Table 7, we explore a number of potential channels through which federal aid may be directly influenced voters' well-being. Specifically, we use equation 4.2 to investigate whether additional federal aid had an impact on COVID-19 outcomes, disposable income, and the unemployment rate. While we think of these results as suggestive, we find that additional federal aid reduced the number of COVID-19 deaths in beneficiary states and that residents of those states saw their disposable income go up in the year of the relevant election. Previous research typically finds income increases correspond to better electoral performance for incumbents (Bartels 2008; Krause and Melusky 2014), and our results suggest that voters had ample reasons to reward incumbents who received more aid, at least in models of retrospective voting where voters are naive or effort is imperfectly observed.

Like the effect we estimate on the incumbent vote share, the negative relationship between instrumented aid and the COVID-19 death rate is robust to the inclusion of the covariates from Table 3 (i.e., the 2016 Trump vote share and the March 2020 stringency of economic restrictions) as well as to the inclusion of the covariates from Table 4 (i.e., population density, population, and square mileage). Additionally, this result is consistent with the positive effect of instrumented aid on statewide rates of COVID-19 testing, as analyzed by Clemens et al. (2023). The estimated relationship with disposable income exhibits moderately more sensitivity to the inclusion of additional covariates in the specification.

The magnitudes of the estimated relationships between instrumented fiscal aid and both disposable income and COVID-19 mortality are worth discussing further. Our COVID-19 mortality variable is expressed in terms of deaths per 100,000 residents. The scaling of our variables is such that the estimate implies a reduction of 350 deaths per \$1 billion spent, or roughly \$2.9 million per death averted, which falls well below the thresholds recently used by U.S. federal agencies for estimates of the statistical value of life (Federal Register 2023).⁹

⁹For additional background on the thresholds used by federal agencies, see Table 1 in Section II.B of (Federal Register 2023).

Our disposable income variable is scaled such that our estimate of 0.41 implies a \$0.41 increase in disposable income per capita from the year before the election to the year prior to the election per dollar of fiscal assistance. This is consistent with tax payers receiving substantial additional rebates or other forms of increased income in states that received disproportionate aid allocations in the year of the election. Interestingly, we find an estimate that is economically and statistically indistinguishable from 0 for the change from two years prior to the election up to the year of the election. In combination with the estimate from column 3 of Table 6, this would be consistent with a political business cycle model of state officials' use of federal funds (e.g. Alesina, Roubini, and Cohen 1997; Drazen 2000; Franzese Jr 2002).

6.2 Evidence on Additional Mechanisms

In this sub-section we analyze a set of additional outcomes including voter turnout, campaign finance contributions, media coverage, and the impacts of aid on legislators relative to executive office holders. These analyses serve to assess if and how additional federal aid may have affected electoral outcomes without directly changing voter well-being.

In Tables A.4 and A.5 we investigate whether more generous aid allocations are associated with higher voter turnout. Estimates of both equation 4.2 and equation 4.4 find support for a relationship between aid and turnout in support of the incumbent party, while the evidence is mixed with respect to overall turnout. Turnout in favor of the leading challenger's party appears to be modestly depressed, though the point estimates are not uniformly statistically distinguishable from 0.

In Tables A.6 and A.7 we investigate the relationship between aid and campaign finance contributions. The estimates suggest that aid reduces the salience of races to donors, with estimates suggesting declines in contributions on the order of 10 to 25 percent for both the incumbent and challenger, though point estimates are only occasionally statistically distinguishable from 0. The estimates suggest that the relationship between aid and electoral performance is not mediated by quid pro quo exchanges of donations for economic favors. The point estimates are more consistent with stories in which savvy donors perceive incumbents who receive generous aid allocations to be safe or, alternatively, to have sufficient resources at their disposal to shape their electoral fortunes even while receiving moderately fewer campaign contributions.

A distinction that relates to a number of mechanisms as well as to theories of attribution (Fiorina 1989; Atkeson and Partin 1995; Ansolabehere and Snyder Jr. 2002) is that between the impact of additional aid on the electoral success of incumbents in legislative offices versus executive offices. Appendix Table A.8 presents results for these two sub-samples. Columns 5 through 8 suggest that during the public-health crisis, governors from over-represented states enjoyed a major incumbency advantage. Columns 1 through 4 show the smaller incumbency bonus enjoyed by legislators, which in our sample of statewide elections consists primarily of U.S. senators. The situation is strikingly different during normal times, as the second row of estimates shows: if anything, legislators normally enjoy a greater incumbency bonus from over-representation. Gubernatorial visibility in times of crisis may be responsible for this gap. We observe, for example, as shown in Figure A.1 that newspaper articles that reference either the CARES Act or the American Rescue Plan Act were far more likely to reference governors than senators.

We find additional evidence of larger effects of COVID-19 relief on the electoral fortunes of executive branch officials relative to legislators in an analysis of lower state offices. In particular, as shown in Tables A.9 and A.10, we find null effects in an analysis of the effects of aid on the composition of state legislatures. Interestingly, the coefficient on the lagged values of state-legislature composition are indistinguishable from 1, implying strong persistence and potentially explaining why state legislature outcomes are little moved by federal aid. By contrast, as shown in Table A.11, we find that large allocations of aid predict substantially better performance by incumbents in down-ballot executive offices including states' attorneys general, secretaries of state, and lieutenant governors.

6.3 Implications for Election Outcomes

The implications of our estimates for the outcomes of specific elections are presented in Appendix Table A.12. In assessing the elections for which pandemic fiscal assistance may have been pivotal, we generate counterfactual estimates in which all states are given the same amount of aid as were the least generously represented states. We apply separate, subsample specific coefficients to generate counterfactual estimates for senators and governors. In particular, we use the estimates from columns 4 and 8 of Table A.8. The resulting counterfactuals reveal that pandemic fiscal assistance did not have a substantial partisan bias with respect to the incumbents it may have helped across the finish line. That said, 4 of the 5 Republican-incumbent races for which we predict different outcomes involved senators, while 6 of the 7 Democratic-incumbent races for which we predict different outcomes involved governors.

Overall, incumbent parties performed very well in the pandemic-era elections in our sample. While incumbent senators and governors won 81 percent of their elections in the pre-pandemic portion in our sample, they won 90 percent of the pandemic-era elections in our sample. Notably, our counterfactual implies a pandemic-era incumbent party win share of 80 percent. With aid equalized to the less generous allocations received by the most under-represented states, the incumbent party win share would thus have mirrored their pre-pandemic win share.

7 Conclusion

This paper estimates the effects of the distribution of federal pandemic aid to state and local governments on the performance of incumbent politicians in state-wide elections. Our findings substantiate the hypothesis that increased federal aid in response to the COVID-19 pandemic benefited incumbents in the electoral arena. Specifically, we demonstrate that states with higher per capita congressional representation—and by extension, those that received more federal aid per capita—saw a significant increase in incumbent vote share in the 2020, 2021 and 2022 elections. These results suggest that access to government resources can comprise a key source of the incumbency advantage during times of economic crisis.

This relationship between aid and incumbent vote share holds even after controlling for various factors, including the incumbents' past performance, population density, COVID-19 control stringency, and political preferences expressed in the 2016 presidential election, underscoring the robustness of our results. Our instrumental variables strategy to address endogeneity concerns reinforces the credibility of these findings, as do our panel estimates through which we net out any differential electoral advantage held by incumbents in overrepresented states during elections held prior to the COVID-19 pandemic.

In addition to these empirical contributions, our analysis has broader implications for understanding how access to representation and government spending contribute to the incumbency advantage, particularly during times of crisis. Our results offer new evidence that government programs can influence electoral politics, even in the face of a global pandemic. The fact that pandemic aid was unequally distributed highlights both the direct and indirect benefits of formal political representation. States with more representation per capita not only secured more revenues: the politicians serving those states appeared to benefit electorally as a result of this windfall.

Ultimately, this research underscores the critical role of federal transfers in the political landscape, especially during times of economic crisis. The allocation of aid can have farreaching political as well as economic consequences. Our research extends the existing body of literature by focusing on the downstream and statewide effects of the COVID-19 pandemic in the United States, areas previously underexplored. In doing so, this paper not only sheds light on the political ramifications of the recent pandemic but also sets the stage for a deeper understanding of how government programs and spending can influence the democratic process.

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	(1) Reduced Form	(2) First Stage	(3) 2SLS	(4) Reduced Form	(5) First Stage	(6) 2SLS
Reps per Million	$4.083^{***} (0.981)$	$1.013^{***} \\ (0.059)$		$\begin{array}{c} 2.910^{***} \\ (0.886) \end{array}$	$\frac{1.017^{***}}{(0.075)}$	
Total Aid per Resident (IISD thousands)			A 039***			9 861***
			(1.079)			(0.963)
Normal Vote				0.400^{***} (0.103)	-0.002 (0.008)	0.404^{***} (0.108)
Observations R^2	131 0 220	131 0.870	131 0 143	131 0.340	131 0 870	$131 \\ 0.294$
First Stage F-Stat			293.569			183.504

 Table 1: Baseline Estimates of the Effect of Aid on Incumbents' Vote Shares

This table uses data from the Committee for a Responsible Federal Budget (2021), US Federal Transit Administration (2021), US Census Bureau (2021), Chidambaram and Musumeci (2021), Medicaid and Chip Payment Access Commission (2021), US Office of Elementary and Secondary Education (2021), and US Department of the Treasury (2021), Lewis et al. (2021), Amlani and Algara is described by the equations below, while the "Reduced Form" model is estimated by substituting the vote share outcome on the (2021), Leip (2024), and MIT Election Data and Science Lab (2022b,a). The estimated 2SLS model, as estimated on data for 2020-2022, left-hand side of the first equation below:

$$\frac{\text{Total Aid}_s}{\text{Pop}_s} = \gamma_0 + \gamma_1 \text{Reps Per Million}_s + \gamma_2 \text{Normal Vote}_{s,o,t} + X_{s,o,t}\gamma + \epsilon_{s,o,t}$$

Vote Shares,
$$o, t = \beta_0 + \beta_1 \frac{\text{Total Aid}_s}{\text{Pop}_s} + \beta_2 \text{Normal Vote}_{s,o,t} + X_{s,o,t}\beta + u_{s,o,t}$$

all four bills. Total Aids is scaled by Pops state s's 2020 official Census population. Our outcome of interest, Vote Shares, o.t, is the incumbent party's share of the top two candidate's total votes in state s in year y for office o. Normal Votes, o,t is a control for the incumbent party's vote share in the previous election. Note that the normal vote control is included in the regressions presented in columns 4 through 6, but is excluded in the regressions presented in columns 1 through 3. The control vector $X_{s,o,t}$ is empty in $\frac{\text{Total Aid}_s}{\text{Pop}_s}$ is instrumented using (Reps Per Million_s), the number of representatives and senators per million residents in 2020. The fitted values from the first stage (4.1) are used to estimate the second stage (4.2). Columns 3 and 6 present estimates of the second Where Total Aid_{s.o.t} is the total federal aid per resident to state and local governments (USD thousands) in state s pooled across this table's specifications. Columns 1 and 4 present the reduced form relationship between our outcome of interest (Vote Share_{s,o,t}) and instrument (Reps Per Million_{s, o,t}). In the first stage regression represented by equation (4.1) and presented in Columns 2 and 5, stage. Columns 3 through 6, add Normal Vote_{s,o,t} which is the incumbent party's vote share in the previous election. Standard errors are clustered by state. * p < 0.10, ** p < 0.05, *** p < 0.01

	(1) Baseline	(2) Without Resource Intensive States	(3) Without Tourism Intensive States	(4) Without Resource and Tourism Intensive States
Total Aid per Resident (USD thousands)	$2.861^{***} \\ (0.963)$	3.892^{***} (0.986)	2.928^{***} (0.973)	3.980^{***} (0.997)
Normal Vote	$\begin{array}{c} 0.404^{***} \\ (0.108) \end{array}$	$\begin{array}{c} 0.348^{***} \\ (0.109) \end{array}$	0.390^{***} (0.110)	0.330^{***} (0.112)
Observations R^2 First Stage F-Stat	131 0.294 183.504	118 0.227 70.544	$125 \\ 0.281 \\ 181.536$	$ 112 \\ 0.209 \\ 68.407 $

 Table 2: Analysis of Robustness to States' Exposure to Shocks Due to Their Reliance

 on Resource-Extraction and Tourism-Related Industries

This table uses data from the Committee for a Responsible Federal Budget (2021), US Federal Transit Administration (2021), US Census Bureau (2021), Chidambaram and Musumeci (2021), Medicaid and Chip Payment Access Commission (2021), US Office of Elementary and Secondary Education (2021), US Department of the Treasury (2021), Lewis et al. (2021), Algara & Amlani (2021), Leip (2024), and MIT Election Lab (2022a, 2022b) to estimate the following equations for years 2020, 2021, and 2022 pooled:

$$\frac{\text{Iotal Aid}_s}{\text{Pop}_s} = \gamma_0 + \gamma_1 \text{Reps Per Million}_s + \gamma_2 \text{Normal Vote}_{s,o,t} + X_{s,o,t}\gamma + \epsilon_{s,o,t}\gamma$$

$$\text{Vote Share}_{s,o,t} = \beta_0 + \beta_1 \frac{\widehat{\text{Total Aid}_s}}{\operatorname{Pop}_s} + \beta_2 \text{Normal Vote}_{s,o,t} + X_{s,o,t}\beta + u_{s,o,t}\beta$$

Where Total Aid_s is the total federal aid per resident to state and local governments (USD thousands) in state s pooled across all four bills. Total Aid_s is scaled by Pop_s state s's 2020 official Census population. Our outcome of interest, Vote Share_{s,o,t}, is the incumbent party's share of the top two candidate's total votes in state s in year y for office o. Normal Vote_{s,o,t} is a control for the incumbent party's vote share in the previous election. The control vector $X_{s,o,t}$ is empty in this table's specifications. Column 1 presents baseline estimates of equation (4.2), which include all senate, gubernatorial, and house-at-large elections between 2020-2022 for all 50 states. Column 2 drops resource intensive states (Alaska, North Dakota, and Wyoming), Column 3 drops the tourism intensive states (Hawaii, Nevada, and Florida), and Column 4 drops both of these categories of states, which were particularly impacted by the economic turmoil of the early pandemic. Standard errors are clustered by state. * p < 0.10, ** p < 0.05, *** p < 0.01

 Table 3: Analysis of Robustness to Baseline Proxies for Political and COVID-19 Policy

 Preferences

	(1)	(2)	(3)	(4)
	Baseline	With OSI	With Trump	With OSI and Trump
			Vote Share	Vote Share
Total Aid per Resident				
(USD thousands)	2.861***	3.016^{***}	2.721***	2.757***
	(0.963)	(0.944)	(1.052)	(1.045)
Normal Vote	0.404***	0.383***	0.375***	0.373***
	(0.108)	(0.106)	(0.113)	(0.112)
March 2020 OSI		-26.411*		-4.980
		(13.941)		(15.294)
Trump Vote Share 2016			0.209***	0.198**
-			(0.077)	(0.091)
Observations	131	131	131	131
R^2	0.294	0.312	0.355	0.356
First Stage F-Stat	183.504	282.106	209.692	270.699

This table uses data from the Committee for a Responsible Federal Budget (2021), US Federal Transit Administration (2021), US Census Bureau (2021), Chidambaram and Musumeci (2021), Medicaid and Chip Payment Access Commission (2021), US Office of Elementary and Secondary Education (2021), US Department of the Treasury (2021), Lewis et al. (2021), Algara & Amlani (2021), Leip (2024), MIT Election Lab (2022a, 2022b), and Hale et. al (2023) to estimate the following equations for years 2020, 2021, and 2022 pooled:

$$\frac{\text{Total Aid}_s}{\text{Pop}_s} = \gamma_0 + \gamma_1 \text{Reps Per Million}_s + \gamma_2 \text{Normal Vote}_{s,o,t} + X_{s,o,t}\gamma + \epsilon_{s,o,t}$$

$$\text{Vote Share}_{s,o,t} = \beta_0 + \beta_1 \frac{\widehat{\text{Total Aid}_s}}{\text{Pop}_s} + \beta_2 \text{Normal Vote}_{s,o,t} + X_{s,o,t}\beta + u_{s,o,t}$$

Where Total Aid_s is the total federal aid per resident to state and local governments (USD thousands) in state s pooled across all four bills. Total Aid_s is scaled by Pop_s state s's 2020 official Census population. Our outcome of interest, Vote Share_{s,o,t}, is the incumbent party's share of the top two candidate's total votes in state s in year y for office o. Normal Vote_{s,o,t} is a control for the incumbent party's vote share in the previous election. Column 1 presents baseline estimates of equation (4.2). Included in Columns 2 through 4 is a set of state-level controls $(X_{s,o,t})$. Column 2 controls for a state's March 2020 Oxford Stringency Index, while Column 3 controls for Donald Trump's vote share in the 2016 election in a given state. Column 4 controls for both a state's March 2020 Oxford Stringency Index and it's Trump vote share in 2016. Standard errors are clustered by state. * p < 0.10, ** p < 0.05, *** p < 0.01

	(1)	(2)	(3)	(4)	(5)
	Baseline	With Pop. Density	With Pop.	With Total Land Area	All Po
Total Aid per Resident					
(USD thousands)	2.861^{***} (0.963)	2.538^{***} (0.890)	2.463^{***} (0.936)	3.552^{***} (0.652)	3.259^{***} (0.571)
Normal Vote	0.404^{***} (0.108)	0.405^{***} (0.108)	0.408^{***} (0.106)	0.345^{***} (0.094)	0.330^{***} (0.090)
Residents per square mile		-0.007^{**} (0.003)			-0.011^{***} (0.003)
2020 Population (Millions)			-0.107 (0.092)		0.013 (0.082)
Total Area (10k Square Miles)				-0.173^{***} (0.035)	-0.218^{***} (0.032)
Observations	131	131	131	131	131
R^2	0.294	0.322	0.307	0.329	0.383
First Stage F-Stat	183.504	184.672	277.859	185.459	284.073
This table uses data from the Committee for a Responsible Federal Budget (2021), US Federal Transit Administration (2021), US Census	nittee for a	Responsible Federal Bud	get (2021), US	Federal Transit Administra	tion (2021), US Census

 Table 4: Analysis of Robustness to Population and Density Measures

Bureau (2010, 2021), Chidambaram and Musumeci (2021), Medicaid and Chip Payment Access Commission (2021), US Office of Elementary and Secondary Education (2021), US Department of the Treasury (2021), Lewis et al. (2021), Algara & Amlani (2021), Leip (2024), MIT Election Lab (2022a, 2022b), and Hale et. al (2023) to estimate the following equations for years 2020, 2021, and 2022 pooled:

$$\frac{\text{Total Aid}_s}{\text{Pop}_s} = \gamma_0 + \gamma_1 \text{Reps Per Million}_s + \gamma_2 \text{Normal Vote}_{s,o,t} + X_{s,o,t}\gamma + \epsilon_{s,o,t}$$

Vote Share,
$$\alpha, t = \beta_0 + \beta_1 \frac{\text{Total Àid}_s}{\text{Pop}_s} + \beta_2 \text{Normal Vote}_{s,o,t} + X_{s,o,t}\beta + u_{s,o,t}$$

Where Total Aid, is the total federal aid per resident to state and local governments (USD thousands) in state s pooled across all four bills. Total Aids is scaled by Pops state s's 2020 official Census population. Our outcome of interest, Vote Shares, o, is the incumbent party's share of the top two candidate's total votes in state s in year y for office o. Normal Vote_{s, ot} is a control for the incumbent party's vote share in the previous election. Column 1 presents baseline estimates of equation (4.2). Column 2 includes a control for a state's population density, while Column 3 includes a control for a state's total population in 2020. Column 4 controls for a state's total land area, and Column 5 includes controls for both a state's population density and it's total population in 2020. Standard errors are clustered by state. * p < 0.10, ** p < 0.05, *** p < 0.01

	(1) Reduced Form	$\begin{array}{c} (2) \\ 2SLS \end{array}$	(3) Reduced Form	$(4) \\ 2SLS$
Reps per Million	2.280^{*} (1.163)		1.192 (1.040)	
Total Aid per Resident (USD thousands)		2.299^{*} (1.204)		$1.195 \\ (1.060)$
Normal Vote			$\begin{array}{c} 0.471^{***} \\ (0.129) \end{array}$	$\begin{array}{c} 0.474^{***} \\ (0.127) \end{array}$
Observations R^2 First Stage F-Stat	217 0.054	$217 \\ 0.037 \\ 290.907$	217 0.172	$217 \\ 0.167 \\ 220.761$

 Table 5: Analysis of the Correlation between Pandemic Aid and Pre-Pandemic Election

 Outcomes

This table uses data from the Committee for a Responsible Federal Budget (2021), US Federal Transit Administration (2021), US Census Bureau (2021), Chidambaram and Musumeci (2021), Medicaid and Chip Payment Access Commission (2021), US Office of Elementary and Secondary Education (2021), US Department of the Treasury (2021), Lewis et al. (2021), Algara & Amlani (2021), Leip (2024), and MIT Election Lab (2022a, 2022b). The estimated 2SLS model, as estimated on data for 2013-2019, is described by the equations below, while the "Reduced Form" model is estimated by substituting the vote share outcome on the left-hand side of the first equation below:

$$\frac{\text{Total Aid}_s}{\text{Pop}_s} = \gamma_0 + \gamma_1 \text{Reps Per Million}_s + \gamma_2 \text{Normal Vote}_{s,o,t} + X_{s,o,t}\gamma + \epsilon_{s,o,t}$$

Vote Share_{*s,o,t*} =
$$\beta_0 + \beta_1 \frac{\widehat{\text{Total Aid}}_s}{\operatorname{Pop}_s} + \beta_2 \operatorname{Normal Vote}_{s,o,t} + X_{s,o,t}\beta + u_{s,o,t}\beta$$

Where Total Aid_s is the total federal aid per resident to state and local governments (USD thousands) in state s pooled across all four bills. Total Aid_s is scaled by Pop_s state s's 2020 official Census population. Our outcome of interest, Vote Share_{s,o,t}, is the incumbent party's share of the top two candidate's total votes in state s in year y for office o. Normal Vote_{s,o,t} is a control for the incumbent party's vote share in the previous election. The control vector $X_{s,o,t}$ is empty in this table's specifications. Columns 1 and 3 presented the reduced form relationship between our outcome of interest (Vote Share_{s,o,t}) and instrument (Reps Per Million_s). Columns 2 and 4 present estimates of equation (4.2). In Columns 3 and 4, the control Normal Vote_{s,o,t} is added. Standard errors are clustered by state. * p < 0.10, ** p < 0.05, *** p < 0.01

	(1)	(2)	(3)	(4)	(5)	(6)
Reps per Million×Pandemic	$\frac{1.803^{***}}{(0.627)}$	1.528^{**} (0.728)	$2.056^{***} \\ (0.767)$	$\begin{array}{c} 2.677^{***} \\ (0.649) \end{array}$	$2.795^{***} \\ (0.657)$	3.022^{***} (0.602)
Reps per Million	2.280^{*} (1.165)	1.258 (1.028)	$0.820 \\ (1.030)$			
Pandemic	-2.256 (1.861)	-1.513 (2.041)				
Normal Vote		$\begin{array}{c} 0.442^{***} \\ (0.086) \end{array}$				
Observations	348	348	348	348	348	348
R^2	0.118	0.234	0.302	0.497	0.512	0.617
Year FE	No	No	Yes	Yes	Yes	Yes
Normal Vote x Year FE	No	No	Yes	Yes	Yes	Yes
Office FE	No	No	No	No	Yes	Yes
State FE	No	No	No	Yes	Yes	Yes
Office x State FE	No	No	No	No	No	Yes

Table 6: Reduced-Form Panel Estimates of the Relationship between Representation

 and Incumbents' Vote Shares

This table uses data from the Committee for a Responsible Federal Budget (2021), US Federal Transit Administration (2021), US Census Bureau (2021), Chidambaram and Musumeci (2021), Medicaid and Chip Payment Access Commission (2021), US Office of Elementary and Secondary Education (2021), US Department of the Treasury (2021), Lewis et al. (2021), Algara & Amlani (2021), Leip (2024), and MIT Election Lab (2022a, 2022b) to estimate the following equations for years 2013-2022 pooled:

Vote Share_{s,o,t} = $\alpha_0 + \alpha_1 \text{Reps Per Million}_s + \alpha_2 \text{Reps Per Million}_s \times \text{Pandemic}_t + \alpha_3 \text{Pandemic}_t + \epsilon_{s,o,t}$

Vote Share_{s,o,t} = $\phi_s + \phi_t + \phi_o + \phi_s \times \phi_o + \rho$ Reps Per Million_s×Pandemic_t + γ_t Normal Vote_{s,o,t}×Year_t + $\epsilon_{s,o,t}$

Where Reps Per Million_s is the number of representatives and senators per million residents in 2020. Pandemic_t is an indicator that takes a value of 1 in 2020, 2021, and 2022 and a value of 0 in earlier years. Column 1 presents estimates of (4.3). Column 2 adds the control Normal Vote_{s,o,t}, the incumbent party's vote share from the previous election, to (4.3), while Column 3 adds both year and normal vote x year fixed effects. Column 4 presents estimates of equation (4.4), which adds year and state fixed effects to the specification. Column 5 adds office fixed effects, and Column 6 adds state-by-office fixed effects. Standard errors are clustered by state. * p < 0.10, ** p < 0.05, *** p < 0.01

	(1)	(2)	(3)	(4)
	Total Deaths	Total Cases	Change Disp. Inc.	Unemp. Rate
Total Aid per Resident	-34.560^{***}	-663.150	0.414^{**}	-0.037
(USD thousands)	(11.362)	(942.705)	(0.175)	(0.187)
Normal Vote	$0.148 \\ (1.389)$	-14.828 (117.058)	$0.002 \\ (0.022)$	-0.020 (0.019)
Observations	$ 131 \\ 0.045 \\ 183.504 $	131	131	131
R^2			0.014	0.004
First Stage F-Stat		183.504	183.504	183.504

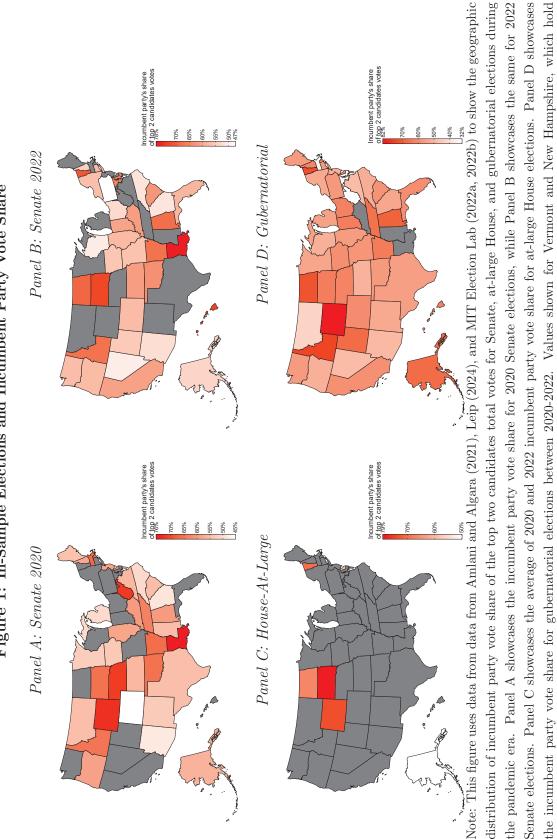
 Table 7: Analysis of Potential Mechanisms Including COVID-19 and Economic Outcomes

This table uses data from the Committee for a Responsible Federal Budget (2021), US Federal Transit Administration (2021a), US Census Bureau (2021), Chidambaram and Musumeci (2021), Medicaid and Chip Payment Access Commission (2021), US Office of Elementary and Secondary Education (2021), US Department of the Treasury (2021), Lewis et al. (2021), Dong, Du, and Gardner (2023), U.S. Bureau of Economic Analysis (2024), U.S. Bureau of Labor Statistics (2024), Algara & Amlani (2021), Leip (2024), and MIT Election Lab (2022a, 2022b), to estimate the following equations for years 2020, 2021, and 2022 pooled:

 $\frac{\text{Total Aid}_s}{\text{Pop}_s} = \gamma_0 + \gamma_1 \text{Reps Per Million}_s + \gamma_2 \text{Normal Vote}_{s,o,t} + X_{s,o,t}\gamma + \epsilon_{s,o,t}$ $Y_{s,o,t} = \beta_0 + \beta_1 \frac{\widehat{\text{Total Aid}_s}}{\text{Pop}_s} + \beta_2 \text{Normal Vote}_{s,o,t} + X_{s,o,t}\beta + u_{s,o,t}$

Where Total Aid_s is the total federal aid per resident to state and local governments (USD thousands) in state s pooled across all four bills. Total Aid_s is scaled by Pop_s state s's 2020 official Census population. $Y_{s,o,t}$ is a vector of state-level outcomes. Normal Vote_{s,o,t} is a control for the incumbent party's vote share in the previous election. Column 1 uses the total number of COVID-19 deaths in December of year y in state s as the outcome of interest, while Column 2 uses the total number of COVID-19 cases in December of year y in state s. Column 3 uses the change in nominal disposable income from the previous year (USD thousands). Column 4 uses the change in unemployment rate from the previous year. Standard errors are clustered by state.

* p < 0.10, ** p < 0.05, *** p < 0.01



gubernatorial elections every two years, are averages of 2020 and 2022 values.

Figure 1: In-Sample Elections and Incumbent Party Vote Share

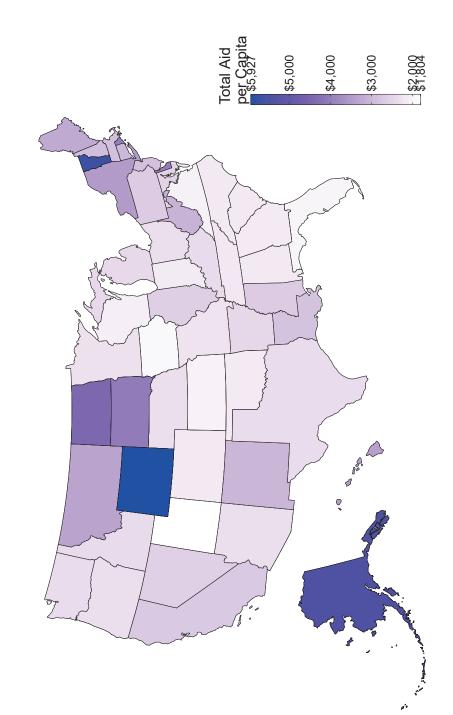


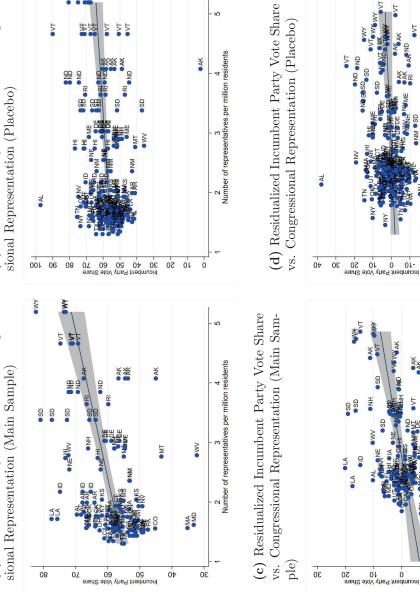
Figure 2: Geographic Distribution of Total State and Local Aid

Note: This figure uses data from the Committee for a Responsible Federal Budget (2021), US Federal Transit Administration (2021), US Census Bureau (2021), Chidambaram and Musumeci (2021), Medicaid and Chip Payment Access Commission (2021), US Office of Elementary and Secondary Education (2021), and US Department of the Treasury (2021) to show the geographic distribution in total aid per capita to in USD.

Figure 3: Distribution of Incumbent Party Vote Share

(b) Incumbent Party Vote Share vs. Congres-

(a) Incumbent Party Vote Share vs. Congressional Representation (Main Sample) Y.



while Panel B plots the equivalent for the placebo sample (2013-2019). Panels C and D regress the remaining variation in representation and incumbent vote share after controlling for the normal vote, mirroring the results in 1 Column 4 and Table 5 Column 3. This figure uses data from the Committee for a Responsible Federal Budget (2021), US Federal Transit Administration (2021), US Census Bureau (2021), Chidambaram and Musumeci (2021), Medicaid and Chip Payment Access Commission (2021), US Office of Elementary and Secondary Education (2021), and US Department of the Treasury (2021), Amlani and Algara (2021), Leip (2024), and MIT Election Data and Science Lab (2022b,a). The slope Panels A plots congressional representation against incumbent party share of the top two candidate's total votes for our main sample (2020-2022). coefficients for displayed regressions are 4.280 in Panel A, 2.280 in Panel B, 2.910 in Panel C, and 1.192 in Panel D.

3

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Online Appendix

Intended for online publication only.

Summary Statistics	
A.1:	
Table .	

anel A: Pandemic Elections	(2020-2022)
l A: Pandem	lect
	andem
	anel A:

	Mean	SD	Min	Max	Ν
Incumbent Vote Share	59.643	9.010	32.250	82.403	131
Total Aid per Resident (USD thousands)	3.004	1.124	1.804	5.927	131
Reps per Million	2.319	1.036	1.302	5.193	131
Normal Vote	59.921	8.355	50.072	97.252	131
March 2020 OSI	0.433	0.050	0.335	0.560	131
Trump Vote Share 2016	49.800	10.492	29.440	68.631	131
Total Deaths	227.223	129.297	21.818	437.401	131
Total Cases	20645.413	12225.960	1189.065	41967.128	131
Change Disposable Income From Previous Year (USD thousands)	1.635	1.739	-2.440	4.876	131
Change in Unemployment Rate From Previous Year	5.002	1.909	2.100	10.058	131
Panel B: Pre-Pandemic Sample (2013-2019)					
	Mean	SD	Min	Max	Ν
Incumbent Vote Share	57.668	10.015	2.121	97.252	217
Total Aid per Resident (USD thousands)	3.004	1.095	1.804	5.927	217
Reps per Million	2.298	1.025	1.302	5.193	217
Normal Vote	60.110	7.666	47.521	82.085	217
Change Disposable Income From Previous Year (USD thousands)	1.526	0.942	-1.989	3.542	217
Change in Unemployment Rate From Previous Year	4.649	1.361	2.358	8.375	217
This table uses data from the Committee for a Responsible Federal Budget (2021), US Federal Transit Administration (2021), US Census Bureau (2021), Chidambaram and Musumeci (2021), Medicaid and Chip Payment Access Commission (2021), US Office of Elementary and Secondary	21), US Feder cess Commis	al Transit Adr sion (2021), U	ninistration (3 S Office of El	ementary and S	s Bureau econdary
and Science Lab (2022 <i>b</i> , <i>a</i>), Hale et al. (2020), Dong, Du, and Gardner (2020), US Bureau of Economic Analysis (2024), and US Bureau of Labor Statistics (2024).	US Bureau of	Economic An	alysis (2024),	and US Bureau	of Labor
· · · ·					

Panel A: Pandemic Elections (2020-2022)					
	Mean	SD	Min	Max	Ν
Δ Log Inc. Party Receipts	0.588	1.167	-1.853	6.286	129
Δ Log Challenger Party Receipts	0.276	2.025	-5.101	8.940	119
Outcome-Control Election Year Dummy	0.618	0.488	0.000	1.000	131
Share State Sen. Controlled by Inc. Gov's Party	61.462	19.769	7.500	93.548	49
Share State House Controlled by Inc. Gov's Party	60.099	18.970	15.625	91.935	48
Share State Sen. Controlled by Inc. Sen's Party	68.140	13.736	38.235	93.333	02
Share State House Controlled by Inc. Sen's Party	66.803	12.474	36.923	90.000	69
Inc. Party Votes / VEP	0.312	0.072	0.146	0.494	131
Challenger Party Votes / VEP	0.216	0.069	0.070	0.403	128
Total Votes / VEP	0.545	0.109	0.330	0.773	131
MoV, Inc. Gov's Party (Down-Ballot Elections)	13.125	25.404	-40.200	100.000	116
MoV, Inc.'s Party (Down-Ballot Elections)	19.310	21.213	-32.500	100.000	115
Panel B: Pre-Pandemic Elections					
	Mean	SD	Min	Max	Ν
△ Log Inc. Party Receipts	0.330	1.192	-5.326	5.849	127
Δ Log Challenger Party Receipts	0.345	1.783	-5.523	8.541	114
Share State Sen. Controlled by Inc. Gov's Party	60.366	19.811	7.692	96.000	49
Share State House Controlled by Inc. Gov's Party	59.875	17.427	18.239	92.157	48
Share State Sen. Controlled by Inc. Sen's Party	67.122	12.970	39.286	96.000	20
Share State House Controlled by Inc. Sen's Party	65.413	11.947	36.923	92.157	69
Challenger Party Votes / VEP	0.198	0.077	0.000	0.466	128
VEP	0.298	0.067	0.159	0.534	131
Total Votes / VEP	0.526	0.109	0.229	1.053	131
MoV, Inc. Gov.'s Party (Down-Ballot Elections)	10.373	20.933	-43.800	100.000	115
MoV, Inc.'s Party (Down-Ballot Elections)	17.807	15.724	0.400	100.000	116
This table uses data from OpenSecrets (2024), FollowTheMone	y (2023), National	Conference of St	ate Legislatures	, FollowTheMoney (2023), National Conference of State Legislatures (2013, 2014, 2015, 2016, 2017,	2016, 2017,
2018, 2019, 2020, 2021, 2022, 2023, 2024), University of Florida Election Lab (2014, 2016, 2018, 2020, 2022), Virginia Department of Elections (2023). New Jersev Voter Information Portal (2017, 2021), and Ballotnedia (2020a, 2020b, 2020a, 2021b, 2021c, 2022a, 2022b, 2022c).	la Election Lab (2) d Ballotnedia (20)	014, 2016, 2018, 2030, 2020	$2020, 2022), Vir_{0}$	ginia Department o 2021c - 2022a - 2025	f Elections 2h. 2022c).
Panel A includes the sample of statewide elections between 2020-2023. Panel B includes the sample of most recent me-pandemic elections between	-2022. Panel B inc	ludes the sample	of most recent n	re-nandemic election	ns hetween
2013-2019 that are analogous to the pandemic sample. Data a	vailability of the p	re-pandemic sam	tple limits the av	c sample. Data availability of the pre-pandemic sample limits the availability of some data points.	ata points.
"Down-Ballot Elections" refers to the offices of lieutenant gove	of lieutenant governor, secretary of state, and attorney general.	state, and attorn	ey general.	5	4

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	(1) Reduced Form	(2) First Stage	(3) 2SLS	(4) Reduced Form	(5) First Stage	(6) 2SLS
Reps per Million	$\begin{array}{c} 4.083^{***} \\ (0.981) \end{array}$	0.724^{***} (0.080)		$\begin{array}{c} 2.910^{***} \\ (0.886) \end{array}$	0.743^{***} (0.099)	
Running Total of Aid per capita (USD thousands)			5.640^{***} (1.628)			3.916^{***} (1.350)
Normal Vote				0.400^{***} (0.103)	-0.007 (0.011)	0.426^{***} (0.099)
Observations R ² First Stage F-Stat	$131 \\ 0.220$	$131 \\ 0.348$	131 82.636	$131 \\ 0.340$	$131 \\ 0.350$	$ \begin{array}{c} 131 \\ 0.132 \\ 56.641 \end{array} $
This table uses data from the Committee for a Responsible Federal Budget (2021), US Federal Transit Administration (2021), US Census Bureau (2021), Chidambaram and Musumeci (2021), Medicaid and Chip Payment Access Commission (2021), US Office of Elementary and Secondary Education (2021), US Department of the Treasury (2021), Lewis et al. (2021), Amlani and Algara (2021), Leip (2024), and MIT Election Data and Science Lab (2022 <i>b</i> , <i>a</i>) to estimate the following equations for years 2020, 2021, and 2022 pooled:	a Responsible Feder. Medicaid and Chi _I asury (2021), Lewis lowing equations fo	al Budget (2021) 5 Payment Acces 5 et al. (2021), A r years 2020, 202	, US Federa s Commiss mlani and 1, and 202	a Responsible Federal Budget (2021), US Federal Transit Administration (2021), US Census Bureau, Medicaid and Chip Payment Access Commission (2021), US Office of Elementary and Secondary assury (2021), Lewis et al. (2021), Amlani and Algara (2021), Leip (2024), and MIT Election Data llowing equations for years 2020, 2021, and 2022 pooled:	ation (2021), US e of Elementary (2024), and MI'	S Census Bureau and Secondary T Election Data
$\frac{\text{Running Aid}_s}{\text{Pop}_s}$	$\frac{1}{2} = \gamma_0 + \gamma_1 \text{Reps Per}$	r Million $_s+\gamma_2 \mathrm{Ni}$	ormal Vote	$\frac{\text{Running Aid}_s}{\text{Pop}_s} = \gamma_0 + \gamma_1 \text{Reps Per Million}_s + \gamma_2 \text{Normal Vote}_{s,o,t} + X_{s,o,t}\gamma + \epsilon_{s,o,t}$		
Vote Shares,	Vote Share_{s,o,t} = $\beta_0 + \beta_1 \frac{\text{Running Aid}_s}{\text{Pop}_s} + \beta_2 \text{Normal Vote}_{s,o,t} + X_{s,o,t}\beta + u_{s,o,t}$	$\underbrace{\widehat{\operatorname{Pop}}_s}_{\operatorname{Pop}_s} + \beta_2 \operatorname{Nor}$	mal Vote _{s,c}	$_{,t}+X_{s,o,t}eta+u_{s,o,t}$		
Where Running Aid _s is the running total of federal aid per resident to state and local governments (USD thousands) in state s. For observations in 2021 in 2020, this includes the sum of funding from the CARES Act and Families First Coronavirus Response Act (FFCRA). For observations in 2021 and 2022, this includes the sum of funding from all four major bills (CARES, FFCRA, RRA, and ARPA). Running Aid _{s,o,t} is scaled by Pop _s state s's 2020 official Census population. Our outcome of interest, Vote Share _{s,o,t} , is the incumbent party's share of the top two candidate's total votes in state s in year y for office o. Normal Vote _{s,o,t} is a control for the incumbent party's vote share in the previous election. The control vector $X_{s,o,t}$ is empty in this table's specifications. Columns 1 and 4 present the reduced form relationship between our outcome of interest (Vote Share _{s,o,t}) and instrument (Reps Per Million _s). In the first stage regression presented in Columns 2 and 5, $\frac{\text{Running Aid}}{\text{Ros}}$ is instrumented using (Reps Per Million _s), the number of representatives and senators per million residents in 2020. The fitted values from the first stage are used to estimate the second stage. Columns 4 and 6 present estimates of the second stage. Columns 3 through 6, add Normal Vote _{s,o,t} which is the incumbent party's vote share in the previous election. Standard errors are clustered by state. * $p < 0.10$, *** $p < 0.05$, *** $p < 0.01$	ederal aid per reside the CARES Act an rom all four major outcome of interest al Vote _{s,o,t} is a cont ations. Columns 1 ations. In the firs fillion _s). In the firs resentatives and se the and 6 present estin- tious election. Standa	ant to state and 1 nd Families First bills (CARES, F i, Vote Shares, o,t trol for the incun and 4 present t st stage regressic mators per millic mates of the secc ard errors are ch	ocal govern Coronaviri FCRA, RF is the incu- nbent part he reduced m presente m residents and stage. (federal aid per resident to state and local governments (USD thousands) in state s. For observations in the CARES Act and Families First Coronavirus Response Act (FFCRA). For observations in 2021 from all four major bills (CARES, FFCRA, RRA, and ARPA). Running Aid _{s,0,t} is scaled by Pop _s r outcome of interest, Vote Share _{s,0,t} , is the incumbent party's share of the top two candidate's total all Vote _{s,0,t} is a control for the incumbent party's vote share in the previous election. The control cations. Columns 1 and 4 present the reduced form relationship between our outcome of interest Million _s). In the first stage regression presented in Columns 2 and 5, $\frac{Running \operatorname{Aid}_s}{Pop_s}$ is instrumented presentatives and senators per million residents in 2020. The fitted values from the first stage are 4 and 6 present estimates of the second stage. Columns 3 through 6, add Normal Vote _{s,0,t} which is ious election. Standard errors are clustered by state. * $p < 0.10, ^{**} p < 0.05, ^{***} p < 0.01$	rds) in state s.] CRA). For obsecond the constant of the top two on the two ontervious elections of the 5 , Running Aid, 5 , Running Aid, 5 , add Normal V bp s , and Normal V $p < 0.05$, *** $p < 0.05$, *** $p < 0.05$, ***	For observations rivations in 2021 : scaled by Pop_s : andidate's total on. The control on. The control come of interest is instrumented is instrumented e first stage are $lote_{s,o,t}$ which is < 0.01

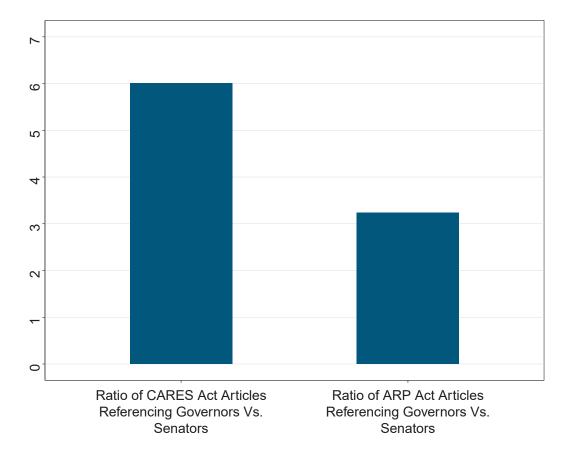


Figure A.1: Ratios of Article Counts: Governors vs. Senators

This figure presents ratios of counts of articles from the Access World News database. The underlying article counts are of articles that appeared in North American newspapers. We counted articles from 2020 that reference CARES Act and articles from 2021 or 2022 that reference the American Rescue Plan Act. We then calculated the ratios of articles referencing the CARES Act or American Rescue Plan Act that also referenced governors to those that also reference senators.

	(1)	(2)	(3)
	Inc. Party Votes	Chall. Party Votes	Tot. Party Votes
	/VEP	/ VEP	/ VEP
Total Aid per Resident			
(USD thousands)	0.024^{***}	-0.008	0.020**
× ,	(0.008)	(0.005)	(0.008)
Lagged Dep. Var	-0.028	-0.034	-0.036
	(0.084)	(0.074)	(0.067)
Normal Vote	0.001	-0.003***	-0.003**
	(0.001)	(0.001)	(0.001)
Observations	131	128	131
R^2	0.086	0.203	0.026
First Stage F-Stat	179.686	182.521	181.148

 Table A.4: Instrumental Variable Estimates of the Effect of Aid on Voter Turnout

This table uses data from the Committee for a Responsible Federal Budget (2021), US Federal Transit Administration (2021a), US Census Bureau (2021), Chidambaram and Musumeci (2021), Medicaid and Chip Payment Access Commission (2021), US Office of Elementary and Secondary Education (2021), US Department of the Treasury (2021), Lewis et al. (2021), Dong, Du, and Gardner (2023), U.S. Bureau of Economic Analysis (2024), U.S. Bureau of Labor Statistics (2024), Algara & Amlani (2021), Leip (2024), MIT Election Lab (2022a, 2022b), University of Florida Election Lab (2014, 2016, 2018, 2020, 2022), Virginia Department of Elections (2023), and New Jersey Voter Information Portal (2017, 2021), to estimate the following equations for years 2020, 2021, and 2022 pooled:

$$\frac{\text{Total Aid}_s}{\text{Pop}_s} = \gamma_0 + \gamma_1 \text{Reps Per Million}_s + \gamma_2 \text{Normal Vote}_{s,o,t} + X_{s,o,t}\gamma + \epsilon_{s,o,t}\gamma$$
$$Y_{s,o,t} = \beta_0 + \beta_1 \frac{\widehat{\text{Total Aid}_s}}{\text{Pop}_s} + \beta_2 \text{Normal Vote}_{s,o,t} + X_{s,o,t}\beta + u_{s,o,t}\gamma$$

Where Total Aid_s is the total federal aid per resident to state and local governments (USD thousands) in state s pooled across all four bills. Total Aid_s is scaled by Pop_s state s's 2020 official Census population. $Y_{s,o,t}$ is a vector of state-level outcomes. Normal Vote_{s,o,t} is a control for the incumbent party's vote share in the previous election. Column 1 uses the incumbent party's total votes as a share of the total voting-eligible population y in state s as the outcome of interest, while Column 2 uses the challenger's party's total votes as a share of the total voting-eligible population in year y in state s. Column 3 uses the total number of cast votes as a share of the total voting-eligible population in year y in state s. For years in which voting-eligible population estimates are not available (ex, in odd-numbered election years), the total number of registered voters is used. Standard errors are clustered by state.

* p < 0.10, ** p < 0.05, *** p < 0.01

$ \begin{array}{ c c c c c c c c c c c c c c c c c c c$		Inc. Party	Inc. Party Votes/VEP	Chall. Pa	Chall. Party Votes/VEP	Tot. Party	Tot. Party Votes/VEP
$ \begin{array}{c c c c c c c c c c c c c c c c c c c $		(1)	(2)	(3)	(4)	(5)	(9)
Reps per Million 0.013^{**} -0.011^{*} 0.008 0.009 Pandemic 0.010 0.029^{*} 0.040^{*} 0.039^{*} 0.039^{*} Pandemic 0.012 0.012 0.012 0.040^{*} 0.029^{*} Observations 340 340 334 340 340 340 Office FE 0.0137 0.788 0.059 0.698 0.056 0.861 Office FE 0.0 Yes 0.059 0.698 0.056 0.861 Preserver 0.0 Yes 0.0 Yes No Yes	Pandemic× Reps per Million	0.013^{**} (0.005)	0.013^{**} (0.005)	-0.006 (0.005)	-0.015^{***} (0.004)	0.004 (0.008)	-0.005 (0.007)
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	Reps per Million	0.013^{**} (0.006)		-0.011^{*} (0.007)		0.008 (0.009)	
	Pandemic	0.010 (0.012)		0.029^{*} (0.015)		0.040^{*} (0.021)	
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	Observations D2	340	340 0.700	334 0.050	334 0 600	340 0.056	340 0 961
Year FENoYesNoYesNoYesState FENoYesNoYesNoYesState FENoYesNoYesNoYesNormal Vote x Year FENoYesNoYesNoYesOffice x State FENoYesNoYesNoYes21). Chidambaram and Musumeci (2021), Medicaid and Chip Payment Access Commission (2021), US Office of Elementary and SeconScon (2021), and US Department of the Treasury (2021), Lewis et al. (2021), Malani and Algara (2021), US Office of Elementary and SeconScole (2021), university of Florida Election Lab (2014, 2016, 2018, 2020, 2022), Virginia Department of Elections (2023), andsey Voter Information Portal (2017, 2021), to estimate the following equations for years 2013-2022 pooled:Scole (2023), and	n- Office FE	ON No	0.700 Yes	oNo	U.090 Yes	ocn.u No	V.501 Yes
State FENoYesNoYesNoYesNormal Vote x Year FENoYesNoYesNoYesOffice x State FENoYesNoYesNoYes21), Chidambaram and Musumeci (2021), Medicaid and Chip Payment Access Commission (2021), US Office of Elementary and Secon ucation (2021), and US Department of the Treasury (2021), Lewis et al. (2021), Amlani and Algara (2021), Leip (2024), MIT Election1 Science Lab (2022b, a), University of Florida Election Lab (2014, 2016, 2018, 2020, 2022), Virginia Department of Elections (2023), and sey Voter Information Portal (2017, 2021), to estimate the following equations for years 2013-2022 pooled:	Year FE	N_{O}	\mathbf{Yes}	N_{O}	\mathbf{Yes}	N_{O}	Yes
Normal Vote x Year FE No Yes No Yes No Yes No Yes No Yes No Yes State FE No Yes No Yes No Yes Office x State FE No Yes No Yes No Yes No Yes No Yes Solver Transit Administration (2021), US Census Bt 21), Chidambaram and Musumeci (2021), Medicaid and Chip Payment Access Commission (2021), US Office of Elementary and Secon (2021), and US Department of the Treasury (2021), Lewis et al. (2021), Amlani and Algara (2021), Leip (2024), MIT Election I Science Lab (2022 <i>b</i> , <i>a</i>), University of Florida Election Lab (2014, 2016, 2018, 2020, 2022), Virginia Department of Elections (2023), and sey Voter Information Portal (2017, 2021), to estimate the following equations for years 2013-2022 pooled:	State FE	N_{0}	\mathbf{Yes}	No	\mathbf{Yes}	No	Yes
Office x State FE No Yes No Yes No Yes No Yes No Yes No Yes Bu ts table uses data from the Committee for a Responsible Federal Budget (2021), US Federal Transit Administration (2021), US Census Bu 21), Chidambaram and Musumeci (2021), Medicaid and Chip Payment Access Commission (2021), US Office of Elementary and Secon neation (2021), and US Department of the Treasury (2021), Lewis et al. (2021), Amlani and Algara (2021), Leip (2024), MIT Election I Science Lab (2022 <i>b</i> , <i>a</i>), University of Florida Election Lab (2014, 2016, 2018, 2020, 2022), Virginia Department of Elections (2023), and sey Voter Information Portal (2017, 2021), to estimate the following equations for years 2013-2022 pooled:	Normal Vote x Year FE	N_{0}	Yes	No	Yes	N_{0}	Yes
is table uses data from the Committee for a Responsible Federal Budget (2021), US Federal Transit Administration (2021), US Census Bu 21), Chidambaram and Musumeci (2021), Medicaid and Chip Payment Access Commission (2021), US Office of Elementary and Secon reation (2021), and US Department of the Treasury (2021), Lewis et al. (2021), Amlani and Algara (2021), Leip (2024), MIT Election I Science Lab (2022 <i>b</i> , <i>a</i>), University of Florida Election Lab (2014, 2016, 2018, 2020, 2022), Virginia Department of Elections (2023), and sey Voter Information Portal (2017, 2021), to estimate the following equations for years 2013-2022 pooled:	Office x State FE	N_{O}	\mathbf{Yes}	N_{O}	Yes	N_{O}	Yes
	s table uses data from the Committee for a 21), Chidambaram and Musumeci (2021), neation (2021), and US Department of the Science Lab (2022 b,a), University of Flori sey Voter Information Portal (2017, 2021),	a Responsible , Medicaid an e Treasury (2 ida Election , to estimate	 Federal Budget and Chip Paymen 021), Lewis et a Lab (2014, 2016, the following eq 	(2021), US 1 it Access Co. 1. (2021), An . 2018, 2020, [uations for y	Federal Transit Adm mmission (2021),US alani and Algara (2 2022), Virginia Der ears 2013-2022 poo	ninistration (2 S Office of Eld (021), Leip (2 partment of E ded:	2021), US Census Buenentary and Secon 024), MIT Election lections (2023), and

Table A.5: Panel Estimates of the Effect of Aid on Voter Turnout

Vote Share_{s,o,t} = $\alpha_s + \alpha_t + \alpha_s \times \alpha_o + \rho$ Reps Per Million_s × Pandemic_t + γ_t Normal Vote_{s,o,t} × Year_t + $\epsilon_{s,o,t}$

Where Reps Per Million_s is the number of representatives and senators per million residents in 2020. Pandemic_t is an indicator that takes a value of 1 in 2020, 2021, and 2022 and a value of 0 in earlier years. Columns 1, 3, and 5 presents estimates of (4.3). Columns 2, 4, and 6 present estimates of equation (4.4), which adds year, office, state, normal vote-by-office, and office-by-state fixed effects to the specification. Standard errors are clustered by state. * p < 0.10, ** p < 0.05, *** p < 0.01

	(1)	(2)
	Δ Incumbent Party	Δ Challenger Party
	Total Receipts (log)	Total Receipts (log)
Total Aid per Resident		
(USD thousands)	-0.127	-0.197
	(0.083)	(0.260)
Lagged Dep. Var	-0.456***	-0.448***
	(0.146)	(0.166)
Normal Vote	-0.018*	-0.000
	(0.010)	(0.029)
Outcome-Control		
Election Year Dummy	0.091	0.278
· ·	(0.163)	(0.298)
Observations	127	112
R^2	0.262	0.164
First Stage F-Stat	182.425	143.748

 Table A.6: Instrumental Variable Estimates of the Effect of Aid on Campaign Finance

This table uses data from the Committee for a Responsible Federal Budget (2021), US Federal Transit Administration (2021a), US Census Bureau (2021), Chidambaram and Musumeci (2021), Medicaid and Chip Payment Access Commission (2021), US Office of Elementary and Secondary Education (2021), US Department of the Treasury (2021), Lewis et al. (2021), Dong, Du, and Gardner (2023), U.S. Bureau of Economic Analysis (2024), U.S. Bureau of Labor Statistics (2024), Algara & Amlani (2021), Leip (2024), MIT Election Lab (2022a, 2022b), OpenSecrets (2024) and FollowTheMoney (2023) to estimate the following equations for years 2020, 2021, and 2022 pooled:

$$\frac{\text{fotal Aid}_s}{\text{Pop}_s} = \gamma_0 + \gamma_1 \text{Reps Per Million}_s + \gamma_2 \text{Normal Vote}_{s,o,t} + X_{s,o,t}\gamma + \epsilon_{s,o,t}$$
$$Y_{s,o,t} = \beta_0 + \beta_1 \frac{\widehat{\text{Total Aid}_s}}{\text{Pop}_s} + \beta_2 \text{Normal Vote}_{s,o,t} + X_{s,o,t}\beta + u_{s,o,t}$$

Where Total Aid_s is the total federal aid per resident to state and local governments (USD thousands) in state s pooled across all four bills. Total Aid_s is scaled by Pop_s state s's 2020 official Census population. $Y_{s,o,t}$ is a vector of state-level outcomes. Normal Vote_{s,o,t} is a control for the incumbent party's vote share in the previous election. In this table, $X_{s,o,t}$ contains the lagged dependent variable and a dummy variable that takes the value of 1 if the dependent and lagged dependent variables fall in different presidential election cycles. Column 1 uses the change in the logged incumbent party's total receipts between the election in year y and the most recent election (that occurred in the same presidential election cycle) in state s as the outcome of interest. Column 2 uses the change in the logged challenger's party's total receipts between the election in year y and the most recent election (that occurred in the same presidential election cycle) in state s as the outcome of interest. Standard errors are clustered by state.* p < 0.10, ** p < 0.05, *** p < 0.01

	ΔL_{0}	Δ Log(Inc. Party Tot. Receipts)	y Tot. Rec	(eipts)	$\Delta \log$	Δ Log(Chall. Party Tot. Receipts)	ty Tot. Re	(ceipts)
	(1)	(2)	(3)	(4)	(5)	(9)	(2)	(8)
Pandemic× Reps per Million	-0.234^{**} (0.092)	-0.229^{**} (0.090)	-0.225^{**} (0.107)	-0.180^{*} (0.104)	-0.164 (0.179)	-0.173 (0.190)	-0.253 (0.219)	-0.242 (0.220)
Reps per Million	0.054 (0.041)	0.077^{*} (0.041)			-0.035 (0.191)	-0.061 (0.197)		
Pandemic	0.871^{***} (0.291)	0.864^{***} (0.289)			0.710 (0.504)	0.723 (0.518)		
Normal Vote		-0.010 (0.006)				0.016 (0.019)		
Lagged Dep. Var.	-0.316^{***} (0.092)	-0.325^{***} (0.092)	-0.355^{***} (0.099)	-0.411^{***} (0.109)	-0.535^{***} (0.098)	-0.518^{***} (0.096)	-0.515^{***} (0.094)	-0.484^{***} (0.081)
Observations	277	277	277	273	251	251	251	239
R^2	0.188	0.193	0.411	0.624	0.200	0.203	0.441	0.601
Office FE	No	N_{O}	N_{O}	Yes	N_{0}	N_{O}	N_{0}	Yes
Year FE	N_{O}	N_{O}	Yes	Yes	N_{O}	N_{0}	Yes	Yes
State FE	No	N_{O}	Yes	Yes	N_{O}	N_{O}	Yes	Yes
Normal Vote x Year FE	No	N_{O}	Yes	Yes	N_{O}	N_{O}	Yes	Yes
Office x State FE	No	No	N_{O}	\mathbf{Yes}	No	N_{O}	No	Yes

 Table A.7: Panel Estimates of the Effect of Aid on Campaign Finance

 $Y_{s,o,t} = \alpha_0 + \alpha_1 Reps Per Million_s + \alpha_2 Reps Per Million_s \times Pandemic_t + \alpha_3 Pandemic_t + \epsilon_{s,o,t}$

 $\mathbf{Y}_{s,o,t} = \alpha_s + \alpha_t + \alpha_s \times \alpha_o + \rho \text{Reps Per Million}_s \times \text{Pandemic}_t + \gamma_t \text{Normal Vote}_{s,o,t} \times \text{Year}_t + \epsilon_{s,o,t}$

Where Reps Per Million_s is the number of representatives and senators per million residents in 2020. Pandemic_t is an indicator that takes a value of 1 in 2020, 2021, and 2022 and a value of 0 in earlier years. $Y_{s,o,t}$ is a vector of state-level outcomes. In this table, in Columns 1-4, $Y_{s,o,t}$ is the change in the logged incumbent party's total receipts between the election in year y and the most recent election (that occurred in the same presidential election cycle) in state s. In Columns 5-8, $Y_{s,o,t}$ is the change in the logged challenger's party's total receipts between the election in year y and the most recent election (that occurred in the same presidential election cycle) in state s. Standard errors are clustered by state. *p < 0.10, ** p < 0.05, *** p < 0.01 Table A.8: Reduced-Form Panel Estimates of the Relationship between Representation and Incumbents' Vote Shares: Comparison of Senate and Gubernatorial Elections

 \triangleleft

	(1)	(2)	(3)	(4)	(5)	(9)	(2)	(8)
Pandemic \times Reps per Million	1.139 (1.202)	0.753 (1.185)	1.274 (1.334)	2.850^{**} (1.249)	5.004^{**} (1.885)	4.896^{**} (1.912)	5.194^{**} (2.265)	5.818^{***} (2.094)
Reps per Million	2.403^{**} (1.138)	1.074 (0.748)	0.912 (0.895)		0.141 (2.027)	-0.223 (1.863)	-0.667 (2.029)	
Pandemic	-1.620 (2.765)	-0.705 (2.915)			-7.800^{**} (3.779)	-7.220^{*} (3.887)		
Normal Vote		0.490^{***} (0.092)				0.291^{*} (0.145)		
Observations	175	175	175	175	143	143	143	143
R^2	0.079	0.253	0.314	0.601	0.115	0.156	0.231	0.620
Year FE	N_{O}	N_{O}	Yes	Yes	N_{O}	N_{O}	Yes	Yes
State FE	N_{O}	No	N_{O}	Yes	N_{O}	N_{O}	N_{O}	Yes
Normal Vote×Year FE	N_{O}	No	Yes	Yes	N_0	N_{O}	Yes	Yes

(2021), Chidambaram and Musumeci (2021), Medicaid and Chip Payment Access Commission (2021), US Office of Elementary and Secondary Education (2021), and US Department of the Treasury (2021), Lewis et al. (2021), Amlani and Algara (2021), Leip (2024), and MIT Election Data and Science Lab (2022b, a) to estimate the following equations for years 2013-2022 pooled: Thi

Vote Share_{s,o,t} = $\alpha_0 + \alpha_1$ Reps Per Million_s + α_2 Reps Per Million_s × Pandemic_t + α_3 Pandemic_t + $\epsilon_{s,o,t}$

Vote Share_{s,o,t} = $\alpha_s + \alpha_t + \alpha_s \times \alpha_o + \rho$ Reps Per Million_s × Pandemic_t + γ_t Normal Vote_{s,o,t} × Year_t + $\epsilon_{s,o,t}$

Where Reps Per Million_s is the number of representatives and senators per million residents in 2020. Pandemic_t is an indicator that takes a value of 1 in 2020, 2021, and 2022 and a value of 0 in earlier years. Column 1 presents estimates of (4.3). Column 2 adds the control Normal Vote_{s,o,t}, the incumbent party's vote share from the previous election, to (4.3), while Column 3 adds both year and normal vote x year fixed effects. Column 4 presents estimates of equation (4.4), which adds year and state fixed effects to the specification. Columns 1 through 4 show estimates for Senate races, while Columns 5 through 8 show estimates for gubernatorial races. Standard errors are clustered by state. * p < 0.10, ** p < 0.05, ***p < 0.01

Table A.9: Instrumental Variable Estimates of the Effect of Aid on State Legislature Composition	⁷ ariable Estimates of the	Effect of Aid on State	Elegislature Composit	ion
	(1) Share State Sen. Controlled by Inc. Gov.'s Party	(2) Share State House Controlled by Inc. Gov.'s Party	(3) Share State Sen. Controlled by Inc. Sen.'s Party	(4) Share State House Controlled by Inc. Sen.'s Party
Total Aid per Resident (USD thousands)	-0.429 (0.716)	0.010 (0.632)	-0.271 (0.807)	0.176 (0.736)
Lagged Dep. Var	0.960^{***} (0.030)	1.055^{***} (0.033)	$\begin{array}{c} 1.010^{***} \\ (0.032) \end{array}$	0.949^{***} (0.047)
Normal Vote	0.084 (0.125)	-0.009 (0.082)	0.017 (0.056)	0.066 (0.073)
Observations R ² First-Stage F Stat	$\begin{array}{c} 49 \\ 0.945 \\ 137.401 \end{array}$	$\frac{48}{0.937}$ 151.041	70 0.915 101.334	69 0.886 107.387
This table uses data from the Committee for a Responsible Federal Budget (2021), US Federal Transit Administration (2021a), US Census Bureau (2021), Chidambaram and Musumeci (2021), Medicaid and Chip Payment Access Commission (2021), US Office of Elementary and Secondary Education (2021), US Department of the Treasury (2021), Lewis et al. (2021), Dong, Du, and Gardner (2023), U.S. Bureau of Economic Analysis (2024), U.S. Bureau of Labor Statistics (2024), Algara & Amlani (2021), Leip (2024), MIT Election Lab (2022a, 2022b), and and the National Conference of State Legislatures (2013, 2014, 2015, 2016, 2017, 2018, 2019, 2020, 2021, 2022, 2023, 2024), to estimate the following equations for years 2020, 2021, and 2022 pooled:	mittee for eci (2021 of the Tr stics (20: 013, 2014	ederal Budget (2021), US F Chip Payment Access Com wis et al. (2021), Dong, Du mlani (2021), Leip (2024), 7, 2018, 2019, 2020, 2021,	deral Transit Administrat mission (2021), US Office , and Gardner (2023), U.S MIT Election Lab (2022a, 2022, 2023, 2024), to estin	a Responsible Federal Budget (2021), US Federal Transit Administration (2021a), US Census Bureau), Medicaid and Chip Payment Access Commission (2021), US Office of Elementary and Secondary assury (2021), Lewis et al. (2021), Dong, Du, and Gardner (2023), U.S. Bureau of Economic Analysis 24), Algara & Amlani (2021), Leip (2024), MIT Election Lab (2022a, 2022b), and and the National , 2015, 2016, 2017, 2018, 2019, 2020, 2021, 2022, 2023, 2024), to estimate the following equations for
	$\frac{\text{Total Aid}_s}{\text{Pop}_s} = \gamma_0 + \gamma_1 \text{Reps}$	= $\gamma_0 + \gamma_1 \text{Reps Per Million}_s + \gamma_2 \text{Normal Vote}_{s,o,t} + X_{s,o,t}\gamma + \epsilon_{s,o,t}$	$ ext{ote}_{s,o,t} + X_{s,o,t} \gamma + \epsilon_{s,o,t}$	
	$\mathbf{Y}_{s,o,t} = \beta_0 + \beta_1 \underbrace{ \operatorname{Total Ai}}_{\operatorname{Pop}_s}$	$=\beta_0 + \beta_1 \underbrace{ \operatorname{Total Aid}_s}_{\operatorname{Pop}_s} + \beta_2 \operatorname{Normal Vote}_{s,o,t} + X_{s,o,t}\beta + u_{s,o,t}$	$+ X_{s,o,t}\beta + u_{s,o,t}$	
Where Total Aid _s is the total federal aid per resident to state and local governments (USD thousands) in state s pooled across all four bills. Total Aid _s is scaled by Pop _s state s's 2020 official Census population. $Y_{s,o,t}$ is a vector of state-level outcomes. Normal Vote _{s,o,t} is a control for the incumbent party's vote share in the previous election. $X_{s,o,t}$ is a vector for additional state-level controls. In this table, $X_{s,o,t}$ contains the lagged dependent variable. Column 1 uses the share of the state senate controlled by the incumbent governor's party in year y in state s as the	ideral aid per resident to strate s^{i} s 2020 official Census post in the previous election. X mu 1 uses the share of the s	ate and local governments ppulation. $Y_{s,o,t}$ is a vector $\zeta_{s,o,t}$ is a vector for additio tate senate controlled by the	(USD thousands) in state of state-level outcomes. N nal state-level controls. Ir ie incumbent governor's p	s s pooled across all four bills. Normal Vote _{s,o,t} is a control for 1 this table, $X_{s,o,t}$ contains the arty in year y in state s as the

outcome of interest, while Column 2 uses the share of the state house controlled by the incumbent governor's party. Column 3 uses the share of the state senate controlled by the incumbent senator's party in year y in state s as the outcome of interest, and Column 4 uses the share of the

state house controlled by the incumbent senator's party. Standard errors are clustered by state. * p < 0.10, ** p < 0.05, *** p < 0.01

	Share Sen. Co by Inc. G	Share State Sen. Controlled Inc. Gov.'s Party	Share House C by Inc. G	Share State House Controlled / Inc. Gov.'s Party	Share Sen. Co by Inc. S	Share State Sen. Controlled by Inc. Sen.'s Party	Share House C by Inc. S	Share State House Controlled by Inc. Sen.'s Party
	(1)	(2)	(3)	(4)	(5)	(9)	(2)	(8)
Pandemic× Reps per Million	-0.238 (0.876)	-0.506 (0.983)	0.738 (0.857)	0.194 (0.737)	-0.152 (0.744)	0.800 (0.768)	0.673 (0.691)	1.151 (0.751)
Reps per Million	-0.053 (0.486)		-0.720 (0.606)		0.095 (0.572)		-0.259 (0.501)	
Pandemic	1.518 (1.877)		-1.427 (1.807)		-0.190 (1.444)		-0.576 (1.323)	
Lagged Dep. Var	0.982^{***} (0.019)	0.940^{***} (0.038)	0.996^{***} (0.029)	0.932^{***} (0.037)	0.974^{***} (0.024)	0.959^{***} (0.050)	0.946^{***} (0.026)	0.951^{***} (0.055)
Observations	142	142	139	139	175	175	172	172
R^2	0.919	0.950	0.889	0.941	0.918	0.946	0.880	0.924
Office FE	N_{O}	\mathbf{Yes}	N_{O}	$\mathbf{Y}_{\mathbf{es}}$	N_{O}	$\mathbf{Y}_{\mathbf{es}}$	No	Yes
Year FE	N_{O}	$\mathbf{Y}_{\mathbf{es}}$	N_{O}	$\mathbf{Y}_{\mathbf{es}}$	N_{O}	$\mathbf{Y}_{\mathbf{es}}$	N_{O}	Y_{es}
State FE	N_{O}	$\mathbf{Y}_{\mathbf{es}}$	N_{O}	$\mathbf{Y}_{\mathbf{es}}$	N_{O}	$\mathbf{Y}_{\mathbf{es}}$	N_{O}	\mathbf{Yes}
Normal Vote x Year FE	N_{O}	\mathbf{Yes}	N_{O}	\mathbf{Yes}	No	$\mathbf{Y}_{\mathbf{es}}$	N_{0}	\mathbf{Yes}
Office x State FE	N_{O}	Yes	N_{O}	$\mathbf{Y}_{\mathbf{es}}$	N_{O}	\mathbf{Yes}	N_{O}	\mathbf{Yes}
This table uses the data described in Table A.9 to estimate the following equations for years 2013-2022 pooled:	Table A.9 t	o estimate the	following eq	uations for ye	ars 2013-202	2 pooled:		
$\mathbf{Y}_{s,o,t} = \alpha_0 +$	- $lpha_1 { m Reps}$ Pei	$Y_{s,o,t} = \alpha_0 + \alpha_1 Reps$ Per Million _s + $\alpha_2 Reps$ Per Million _s × Pandemic _t + $\alpha_3 Pandemic_t + \epsilon_{s,o,t}$	Reps Per M	$\mathrm{illion}_s \times \mathrm{Pand}$	$\operatorname{emic}_t + \alpha_3 \operatorname{Pa}$	$\operatorname{indemic}_t + \epsilon_{s,i}$	o,t	
\mathbf{V} , $-\infty$ + ∞ + ∞ + α Bars Par Million × Pandemic, + ∞ Normal Vote – , × Vear, + ϵ	5 ~ 5 	⊥ ∩Rens Per N	dillion < Da	ndemic. $\pm \infty$.	Jormal Vota	$\times V_{ear, \pm}$		

 Table A.10: Panel Estimates of the Effect of Aid on State Legislature Composition

Where Reps Per Million_s is the number of representatives and senators per million residents in 2020. Pandemic_t is an indicator that takes a is the share of a state senate controlled by the incumbent governor's party in state i in year y. In Columns 3-4, $Y_{s,o,t}$ is the share of a state house controlled by the incumbent governor's party in state i in year y. In Columns 5-6, Y_{s,o,t} is the share of a state senate controlled by the incumbent senator's party in state i in year y. In Columns 3-4, $Y_{s,o,t}$ is the share of a state house controlled by the incumbent senator's party in value of 1 in 2020, 2021, and 2022 and a value of 0 in earlier years. $Y_{s,o,t}$ is a vector of state-level outcomes. In this table, in Columns 1-2, $Y_{s,o,t}$ state i in year y. Columns 1, 4, 5, and 7 present estimates of (4.3). Columns 2, 4, 6, and 8 add office, year, state, normal vote x year and office x state fixed effects to the specification. Standard errors are clustered by state. * p < 0.10, ** p < 0.05, *** p < 0.01

		MoV Gover	Governor's Party	fy		MoV Incum	MoV Incumbent's Party	ty
	In	Incl.	Ex	Excl.	Ir	Incl.	Excl	cl.
	Same-	Same-Ticket	Same-	Same-Ticket	Same	Same-Ticket	Same-Ticket	Ticket
	(1)	(2)	(3)	(4)	(5)	(9)	(2)	(8)
Total federal aid per resident (USD millions)	0.683 (7.306)	1.974 (3.510)	2.439 (8.305)	3.406 (4.103)	8.343^{**} (3.639)	6.355^{*} (3.665)	$\frac{11.109^{***}}{(3.495)}$	9.542^{***} (3.616)
Lagged Dep. Var		0.742^{***} (0.114)		0.730^{***} (0.116)		0.306^{***} (0.111)		0.204^{*} (0.108)
Observations R^2 First Stage F-Stat	116 178.668	$\begin{array}{c} 115 \\ 0.349 \\ 360.237 \end{array}$	$\begin{array}{c} 100\\ \cdot\\ 177.342\end{array}$	$\begin{array}{c} 99\\ 0.338\\ 310.882 \end{array}$	$115 \\ 0.032 \\ 178.373$	$115 \\ 0.104 \\ 146.077$	$\begin{array}{c} 99 \\ 0.069 \\ 176.702 \end{array}$	99 0.113 130.846
This table uses data from the Committee for a Responsible Federal Budget (2021), US Federal Transit Administration (2021a), US Census Bureau (2021), Chidambaram and Musumeci (2021), Medicaid and Chip Payment Access Commission (2021), US Office of Elementary and Secondary Education (2021), US Department of the Treasury (2021), Lewis et al. (2021), Dong, Du, and Gardner (2023), U.S. Bureau of Economic Analysis (2024), U.S. Bureau of Labor Statistics (2024), Algara & Amlani (2021), Leip (2024), MIT Election Lab (2022a, 2022b),), and Ballotpedia (2020a, 2020b, 2020c, 2021a, 2021b, 2021c, 2022b, 2022b, 2022c) to estimate the following equations for years 2020, 2021, and 2022 pooled:	r a Respond 1), Medicai reasury (20 2024), Alga 2022a, 202	sible Federal d and Chip 21), Lewis et ra & Amlan 2b, 2022c) t	Budget (20 Payment A t al. (2021), Id ii (2021), Id o estimate t	21), US Fed ccess Comn Dong, Du, eip (2024), he following	eral Transit nission (202 and Gardne MIT Electio g equations	Administrat 1), US Office rr (2023), U.S on Lab (202 for years 20	ion (2021a), e of Elements 5. Bureau of J 2a, 2022b), 20, 2021, and	a Responsible Federal Budget (2021), US Federal Transit Administration (2021a), US Census Bureau, Medicaid and Chip Payment Access Commission (2021), US Office of Elementary and Secondary asury (2021), Lewis et al. (2021), Dong, Du, and Gardner (2023), U.S. Bureau of Economic Analysis 24), Algara & Amlani (2021), Leip (2024), MIT Election Lab (2022a, 2022b),), and Ballotpedia (222a, 2022b, 2022c) to estimate the following equations for years 2020, 2021, and 2022 pooled:
$\frac{\text{Total Aid}_s}{\text{Pop}_s}$		= $\gamma_0 + \gamma_1 \text{Reps Per Million}_s + \gamma_2 \text{Normal Vote}_{s,o,t} + X_{s,o,t}\gamma + \epsilon_{s,o,t}$	[illion _s + γ_2]	Normal Vot	$e_{s,o,t} + X_{s,o}$	$_{o,t}\gamma+\epsilon_{s,o,t}$		
$\mathbf{Y}_{s,o,t}$	$_{0,t} = \beta_0 + \beta_1$	$\frac{\widetilde{\text{Total Aid}_s}}{\text{Pop}_s}$	$\frac{\operatorname{Total Aid}_{s,o,t}}{\operatorname{Pop}_s} + \beta_2 \text{Normal Vote}_{s,o,t} + X_{s,o,t}\beta + u_{s,o,t}$	I Vote $_{s,o,t}$ +	- $X_{s,o,t}eta+\imath$	$l_{s,o,t}$		
Where Total Aid _s is the total federal aid per resident to state and local governments (USD thousands) in state s pooled across all four bills. Total Aid _s is scaled by Pop _s state s's 2020 official Census population. $Y_{s,o,t}$ is a vector of state-level outcomes. $X_{s,o,t}$ is a vector for additional state-level controls. In this table, $X_{s,o,t}$ contains the lagged dependent variable. Columns 1 and 2 use the margin of victory of the incumbent governor's party in an attorney general, secretary state, or lieutenant governor race in year y in state s as the outcome of interest, while Columns	per resident) official Ce- ontains the retary state	t to state an nsus populai lagged depe. , or lieutena	ad local gov tion. $Y_{s,o,t}$ ndent varial unt governor	ernments (¹ is a vector o ble. Column race in yea	USD thousa of state-leve 1×1 and 2×1 1×1 in state	ands) in stat. I outcomes. use the marges as the outcomes.	e s pooled as $X_{s,o,t}$ is a vegin of victory come of inter-	ross all four bills. ctor for additional of the incumbent st, while Columns
3 and 4 use the margin of victory of the incumbent attorney general, secretary state, or lieutenant governor's party in year y in state s. Columns	sumbent att	orney gener	al, secretary	r state, or 114	r lieutenant go	vernor's part	y in year y ii	n state s. Columns

1 and 3 include races where the candidate runs on the same ticket as the governor, while Columns 2 and 4 exclude such races. Standard errors are clustered by state. * p < 0.10, ** p < 0.05, *** p < 0.01gove 3 an Who Tota stat

Year	State	Office	Winning Party	Winner	Counterfactual Winner
2020	AK	Senate	Rep	Dan Sullivan	Al Gross
2022	AK	Senate	Rep	Lisa Murkowski	Kelly Tshibaka
2020	ME	Senate	Rep	Susan Collins	Sara Gideon
2022	NV	Senate	Dem	Catherine Cortez Masto	Adam Laxalt
2022	WI	Senate	Rep	Ron Johnson	Mandela Barnes
2022	\mathbf{KS}	Governor	Dem	Laura Kelly	Derek Schmidt
2022	ME	Governor	Dem	Janet Mills	Paul LePage
2022	NH	Governor	Rep	Chris Sununu	Tom Sherman
2022	NM	Governor	Dem	Michelle Lujan Grisham	Mark Ronchetti
2022	OR	Governor	Dem	Tina Kotek	Christine Drazan
2022	RI	Governor	Dem	Daniel McKee	Ashley Kalus
2022	WI	Governor	Dem	Tony Evers	Tim Michels

 Table A.12:
 Counterfactual Election Flips

This table presents the set of elections that result in different outcomes under a counterfactual in which we apply the coefficients estimated in columns 4 and 8 of table A.8.