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MY TAXES ARE TOO DARN HIGH:  
WHY DO HOUSEHOLDS PROTEST THEIR TAXES?

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### **ABSTRACT**

In all U.S. states, individuals can file a protest with the goal of legally reducing their property taxes. We study the motives for protesting taxes using administrative records and two sources of causal identification: a quasi-experiment and a large-scale natural field experiment. We show that, consistent with selfish motives, the decision to protest is highly elastic to the private benefits and private costs. We also find evidence of fairness motives: consistent with conditional cooperation, a higher perceived average tax rate decreases both perceived unfairness and the protest probability. Last, we compare protest behavior between Republican and Democratic households. In contrast to some survey data suggesting a large partisan divide in views about the government, we find only small partisan differences in our high-stakes, naturally occurring context.

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# 1 Introduction

In all U.S. states, households can file protests with the goal of *legally* reducing the amount they have to pay in property taxes. These protests are consequential: while there is no guarantee that protests will reduce the tax bills, in practice they more often than not do. In this study, we explore what motivates households to protest their taxes.

Although property tax protests are allowed across the United States (Dobay et al., 2019; World Bank, 2019), in this study we focus on one specific county (i.e., Dallas County) in Texas for the logistical advantage of implementing the field experiment in a single location. However, tax protests work similarly across the country.<sup>1</sup> Thus, the same methods that we apply in Dallas County could be replicated in other Texas counties, and the main findings from Dallas County can be reasonably extrapolated to other U.S. counties. We chose Dallas County because it is one of the largest and most diverse counties in the state and because property taxes are especially important there. Texas does not have a state income tax, so property taxes are a key source of revenue for the provision of government services. The average household in Dallas County was expected to pay around \$5,916 in property taxes in 2020, corresponding to an average tax rate of 2.01% of home market value.<sup>2</sup>

The process to protest property taxes in Dallas County can be summarized as follows. The Dallas Central Appraisal District (DCAD) formulates a proposed assessment of the property’s market value (hereinafter, we refer to this amount as the *proposed value*). Property taxes are calculated based on this proposed value. DCAD notifies the household of this proposed value. The homeowner can then file a protest, arguing, for example, that the proposed value is too high.<sup>3</sup> Owners can protest directly on their own (which is the main focus of this paper), or they can hire an agent to protest on their behalf. DCAD then responds to the homeowner’s appeal. We refer to a protest as successful if DCAD lowers the effective assessed value. In 2020, a total of 8.40% of households in Dallas County filed a protest of their property taxes on their own, and an additional 8.42% of households protested with the help of an agent,

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<sup>1</sup> The protest process is almost identical across all counties in Texas but there are some differences across states. For example, in Alaska, property owners must pay a filing fee, and in California, property taxes are based on taxable values, which are typically updated only when properties are sold, making protests less common.

<sup>2</sup> These statistics are based on administrative data focusing on single-family homes. Throughout this study, we use the term “tax rate” to refer to the household’s effective tax rate (computed as the household’s total property tax amount billed divided by its market value), rather than jurisdictional tax rates.

<sup>3</sup> We describe the protest process in more detail in Section 2.

resulting in a total protest rate of 16.82%.<sup>4</sup> We estimate that 69.7% of direct protests in 2020 were successful, resulting in \$485 in tax savings, on average, in the first year alone.<sup>5</sup>

Due to the lack of theoretical and empirical work on this topic, it is unclear why households choose to protest their taxes. In theory, many potential motives could induce households to file a protest (e.g., they think the government is inefficient.). In this study, we provide an empirical analysis of two channels: selfish motives and fairness motives. We focus on these two channels because they play a central role in the political economy literature on preferences for taxation and redistribution. For instance, the seminal models of income redistribution assume that individuals care about their own material well-being only (e.g., Meltzer and Richard, 1981). According to these selfish motives, households should be more likely to protest when they face higher private benefits (e.g., tax savings) or lower private costs (e.g., hassle costs). More recent literature on political economy highlights the role of fairness considerations (Alesina and Angeletos, 2005; Benabou and Tirole, 2006). More precisely, the conditional cooperation hypothesis posits that taxpayers should be less likely to protest if they think other taxpayers are contributing their fair share of taxes.

Teasing apart the motives for protesting taxes can be challenging. In this study, we combine quasi-experimental and experimental methods. To study the role of private benefits, we exploit quasi-experimental variation introduced by a feature of the Texas Property Code known as the homestead cap. All counties in Texas must use a cap when calculating taxes for households with homestead status.<sup>6</sup> The homestead cap generates a sharp kink in the expected private benefits from protesting. When the proposed value is below the threshold, the tax amount decreases if the county revises the proposed value downward. However, when the proposed value exceeds the threshold, a marginal reduction in the proposed value has no effect on the tax amount. Exploiting this exogenous variation, we find that households are more likely to protest when they stand to gain more from protesting. Specifically, a \$100 increase in the marginal benefits from protesting causes an increase in the probability of protesting of 2.14 percentage points (pp). These estimates translate into a 1.04 elasticity of the protest probability with respect to the expected tax savings. Thus, the decision to protest is elastic to the financial benefits.

To study the role of private costs, we conduct a pre-registered field experiment. In Dallas County, any household can protest property taxes for free, meaning that there are no

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<sup>4</sup> Approximately 34.5% of households in this study protested (directly or with an agent) at least once during 2015-2020, and 23.5% protested directly at least once during this time period.

<sup>5</sup> These estimates are based on data as of November 2020 and thus do not reflect a few protests that might have been resolved at a later date.

<sup>6</sup> Homestead is a legal status that can be granted to properties that are the owner's (owners') primary residence. This legal status comes with several advantages related to property taxes (e.g., exemptions and caps) and other benefits (e.g., exemption from forced sale for collection of debt).

pecuniary costs of protesting. However, we hypothesize that households may face significant hassle costs (Goolsbee, 2006; Benzarti, 2020; Sunstein, 2021; Benzarti, 2021), so we designed a mailing intervention to reduce these costs. We selected a subject pool of 78,462 households, from which we randomly selected 50,394 households to mail a letter. All versions of this letter offered helpful information on how to file a protest, but we randomized the intensity of our aid provision. The basic letter treatment included step-by-step guides for filling out the forms by mail or online. The other letter type included extra instructions to complete one of the most challenging aspects of the process, which is preparing an argument to support the protest. In these letters with extra aid, we included an argument tailored to each recipient that could be copied and pasted into their own protest form. These letters presented information about another property near the recipient’s own property that was comparable in all observable characteristics and was recently sold for a lower price than the market value proposed by the DCAD.

We find that both letter types have large impacts on the probability of filing a protest, and the impact is significantly higher among recipients of the letters containing extra aid. These tailored letters lead to an increase in the protest rate of 4.98 pp (the intention-to-treat effect amounts to 3.51 pp). For comparison, the fraction of households protesting in the control group, which did not receive any letter, is 8.67 pp. In other words, a simple intervention by mail increases the protest rate by a whopping 57.4%. This evidence suggests that hassle costs are of first-order importance in the decision to protest taxes.<sup>7</sup> Moreover, we combine the experimental and quasi-experimental estimates to quantify the hassle costs. Our back-of-the-envelope calculations estimate hassle costs at \$232. Indeed, this estimate constitutes a lower bound, as our intervention reduces some hassle costs but not all.

Our findings can address basic questions related to tax protests. A key fact is that even though there is no fee to file a protest and no risk of increasing taxes, only a small minority of households choose to protest their taxes each year. One explanation is that most households do not want to free-ride on the taxes paid by others. A different interpretation is that most households want to free-ride, but they believe that the hidden private costs of protesting will exceed the expected tax savings. Our evidence can shed light on this dichotomy. On the one hand, among those who protested in 2020, the average tax savings was \$338 (including both successful and unsuccessful protests). On the other hand, we estimate the average private costs from protesting are \$232 (and this is a lower-bound). The fact that the average costs and benefits are on the same order of magnitude favors the interpretation that most households would want to free-ride on their neighbors by protesting their taxes, but they choose not to

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<sup>7</sup> Our evidence of hassle costs is consistent with frictions found in other contexts, such as the use of social benefits (Finkelstein and Notowidigdo, 2019) and filing of income taxes (Bhargava and Manoli, 2015; Benzarti, 2020).

protest because their hassle costs are too large relative to their expected savings.

This evidence suggests that selfish motives are important in the decision to protest taxes. We further provide evidence that fairness concerns also play a role. According to the conditional cooperation channel, taxpayers may be more willing to tolerate a higher tax rate if they think that the average household also faces a high tax rate. This type of reciprocity has been documented in laboratory public good games (Gächter, 2007). However, it is not clear whether this conditional cooperation would be significant in a context such as property taxes. Among other things, the stakes are orders of magnitude higher for property taxes than for laboratory games, and individuals may have strong views about the government that do not manifest in laboratory games (Huet-Vaughn et al., 2019).

To test the conditional cooperation hypothesis, we include a second treatment arm in the field experiment. Our identification strategy leverages misperceptions about the average tax rate: homeowners who pay below-average tax rates tend to underestimate the average tax rate, and homeowners who pay above-average tax rates tend to overestimate the average tax rate. As a result, we can create exogenous variation in those perceptions through an information-provision experiment. In the letter sent to each household, we randomize whether we include information on the average tax rate paid in the county. We show that the information shocks induced by the experiment have a significant effect on the probability of protesting and in the direction predicted by the conditional cooperation channel: perceiving a higher average tax rate in the county causes a decline in the probability of protesting. Our back-of-the-envelope calculations suggest that the conditional cooperation channel is not only statistically significant but also significant in magnitude: the elasticity of the protest probability with respect to the perceived average tax rate is 1.63.

Moreover, we provide suggestive evidence that these effects are driven by fairness concerns. In a survey of 1,888 subjects from the field experiment, we asked whether they felt their tax rate, relative to the county average, is fair or unfair. We show that the information provision experiment has a significant effect on perceived fairness and in the direction predicted by the conditional cooperation channel: perceiving a higher average tax rate in the county causes taxpayers to see their own taxes as more fair. Furthermore, we provide evidence against alternative mechanisms, such as using the information about the average tax rate to make inferences about the average success of protests.

Social scientists typically study preferences for taxation and redistribution through survey data. We argue that the decision to protest taxes provides a unique opportunity to study aspects of taxation support via revealed-preferences in a high-stakes, naturally occurring context. By filing a tax protest, a household expresses unwillingness to contribute to the

provision of public services.<sup>8</sup> As an illustration, we study the partisan differences in protest behavior. Survey data indicate that Republicans are much less likely to support income redistribution than Democrats (Di Tella et al., 2017; Alesina et al., 2020; Stantcheva, 2020). However, there is little evidence of whether those differences in survey responses materialize as differences in *behavior* ( i.e., whether individuals put their money where their mouths are).<sup>9</sup> We categorize households as Democrat and Republican by matching the individuals from the property tax rolls to other sources of data, such as the Texas voter files. We show that the probability of protesting and the motives behind the protests are qualitatively and quantitatively similar between Democrats and Republicans. Some differences along party lines are statistically significant, but they are always small in magnitude. This evidence suggests that some survey questions may exaggerate the degree of partisan polarization in the support for taxation and that there may be more common ground across party lines than previously thought.

This study contributes to various strands of literature. First, we offer insight into tax protests. Protesting property taxes is allowed in all U.S. states and outside of the United States (Dobay et al., 2019; World Bank, 2019).<sup>10</sup> Yet, tax protests have received little attention in the economics literature, especially compared to other forms of tax compliance, such as evasion and delinquency (Luttmer and Singhal, 2014; Castro and Scartascini, 2015; Pomeranz, 2015; Hallsworth et al., 2017; Slemrod, 2019; Holz et al., 2020; Bott et al., 2020; De Neve et al., 2021). To the best of our knowledge, only two papers focus on tax protests, and these papers use archival data: Jones (2019) uses data on the decision to protest taxes to provide a test of loss aversion,<sup>11</sup> and Avenancio-León and Howard (2019) show that local governments place a disproportionate fiscal burden on racial and ethnic minorities and further document that some of these differences operate through tax appeals.<sup>12</sup> We contribute to this literature by identifying and measuring three important factors that affect the protest decision: private benefits, private costs, and conditional cooperation. Indeed, these findings

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<sup>8</sup> This is an important dimension but not the only dimension of support for taxation. For instance, households may have preferences for progressive taxation (e.g., taxes on richer people as opposed to on themselves) and how tax revenues should be spent.

<sup>9</sup> One notable exception is Cullen et al. (2020), who provide evidence suggesting that individuals evade federal taxes less when the president is from their same political party.

<sup>10</sup> In practice, protesting property taxes may be more common in some U.S. counties than in others due to institutional differences (e.g., some counties charge significant fees to file a protest). Outside of the U.S., Dobay et al. (2019) find that protesting property taxes is allowed in all 10 countries they examine, and World Bank (2019) shows that property tax protests are allowed in several Latin American countries.

<sup>11</sup> In essence, Jones (2019) shows that the probability of protesting increases when the assessed value is revised upwards and decreases when the assessed value is revised downwards, but the effect is much larger (in absolute value) for the upwards revisions than for the downward revisions.

<sup>12</sup> Some studies look at property taxes more generally, without focusing on protests. For example, Cabral and Hoxby (2012) provide evidence on how the salience of property taxes can affect equilibrium tax rates. And Simon (2021) studies the general equilibrium housing market effects of property taxes.

can provide key inputs for the design of a more efficient and equitable system of tax appeals.<sup>13</sup>

Second, we contribute to the literature on individual support for taxation and redistribution, which is one of the oldest topics of study in political economy (Alesina and Glaeser, 2004). Despite its importance, revealed-preference evidence has proved elusive on this topic. Social scientists have typically studied various dimensions of preferences for taxation and redistribution through survey data (Alesina and Giuliano, 2011; Cruces et al., 2013; Kuziemko et al., 2015). However, survey data has some well-known limitations. For example, individuals may say that they want high redistribution due to social pressure (i.e., cheap talk), but they may act selfishly in real-world, high-stakes situations.<sup>14</sup> We contribute to this literature by introducing tax protests as a novel context to study some important aspects of individuals' support for taxation via revealed preferences in a high-stakes, naturally occurring setting.<sup>15</sup> Our evidence suggests that using revealed-preference data can be consequential. For example, some survey data indicate that Democrats and Republicans differ substantially in their views about the role of government. In contrast, we find these differences to be small when it comes to real-world, high-stakes protest behavior.

The rest of the paper proceeds as follows. Section 2 describes the institutional context. Section 3 discusses the role of private benefits. Section 4 presents evidence on the private costs. Section 5 discusses the results on conditional cooperation. Section 6 analyzes the differences in tax protests by political party. The last section concludes.

## 2 Institutional Context

Dallas County is the second largest county in Texas with an estimated population of 2.6 million in 2020. In Texas, counties collect property taxes, which they use to fund various services, including schools, parks, roads, and the police and fire departments. The Dallas County tax assessor contractually collects property taxes. While the county collects property taxes on both residential and business properties, this study focuses on residential single-family homes. We use publicly available administrative data from the DCAD. For each home

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<sup>13</sup> Our study also relates to the literature on behavioral public finance, showing that insights from behavioral economics can have important implications for public policy (Chetty, 2015). More precisely, we show that hassle costs and conditional cooperation play significant roles in the decision to protest taxes.

<sup>14</sup> Some notable exceptions are studies that rely on self-reported voting (Fisman et al., 2017; Epper et al., 2020) and charitable giving (Fisman et al., 2007). However, these alternative approaches have limitations of their own. For example, taxation and income redistribution are important issues when deciding for whom to vote, but the decision can depend largely on non-economic issues, such as abortion and the Second Amendment. Likewise, charitable giving can be largely driven by other factors, such as religious participation (Bottan and Perez-Truglia, 2015).

<sup>15</sup> The literature on support for taxation and redistribution using survey data includes a broad range of dimensions, some of which we do not cover, such as support for progressive taxes.



in the county, the data includes information on ownership, address, property characteristics (e.g., number of bedrooms), and historical yearly data on proposed and certified market values, exemption amounts, taxable values, tax rates as well as details on property tax protest records. Whenever needed, we complement the administrative records with other data sources.

The tax amount due is calculated by the DCAD using a multi-step formula that starts with the county’s proposed value of the home as of January 1<sup>st</sup>. Taxes due are a function of a host of factors such as the household’s exemptions and the specific jurisdictional tax rates that pertain to the household, depending on the jurisdictions to which the home belongs.<sup>16</sup> Homeowners have the right to protest if they disagree with DCAD’s proposed assessment value. Among other reasons, homeowners can protest if they believe the proposed value of their property is too high relative to the market value of comparable houses that were sold in the county, if their properties’ proposed values are too high relative to the proposed values of comparable houses in the district, or if there are errors in the public records of the property (e.g., an incorrect number of bedrooms). For instance, according to the 2020 data for households that protested on their own, 91.87% of them selected the option “Value is over market value” in their online forms. When filing a protest, homeowners can also provide an “Opinion of Value”, which is how much they believe their property was actually worth as of January 1<sup>st</sup>.

Homeowners can file protests on their own. We refer to this type of protest as *direct protests*. Because such protests are the focus of this paper, we always refer to this type of protests unless we explicitly note otherwise. Instead of protesting on their own, homeowners can hire an agent to protest on their behalf. In exchange for representation, agents normally charge some combination of a flat fee and a percentage of the tax savings (which can be as high as 50% of the tax savings). Protests through agents are less relevant to our study for a couple of reasons. Most importantly, we designed the field experiment specifically to reduce the hassle costs from protesting directly – which, if anything, should crowd out protests through agents. Second, the timing of the protests through agents makes it more difficult for them to be affected by the type of quasi-experimental and experimental variation used in our research design. According to anecdotal accounts, households often sign contracts with agents months before the proposed values are announced. Indeed, the decision to protest through an agent may have been made years ago, as agents offer long-term contracts to

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<sup>16</sup> In Dallas County there are four county-level jurisdictions along with jurisdictions for each of the 23 of the cities, nine school districts, one Community College system, one hospital system, and 49 Public Improvement Districts (Source: <https://www.dallascounty.org/departments/tax/jurisdictions.php>). Appendix A.1.1 provides additional details about the property taxes in Dallas County and the data sources.

automatically protest on the owner’s behalf every year.<sup>17</sup> For these reasons, the baseline specifications focus on direct protests and, for the sake of completeness, we report results for protests through agents in the Appendix.

The timing of the protest process is quite simple. Each year, the DCAD appraises the value of all homes in the county based on properties’ market values as of January 1<sup>st</sup>. The DCAD shares the proposed values with homeowners through its website and, for most households, by mailing a “Notice of Appraised Value”. Households have a month from the notification date to file a protest. DCAD’s notifications include estimated taxes, which are based on each property’s proposed value. The term “estimated” is used to connote that, technically, property tax rates are determined later in the year, so the county uses the prior year’s jurisdictional tax rates to estimate taxes due in the Notice of Appraised Value. In practice, tax rate changes are uncommon so the approximation error is often negligible. In any case, these estimated taxes are the relevant object of study, as they represent the subjects’ expectations at their time of deciding whether to protest.<sup>18</sup> In 2020, DCAD presented the proposed values on May 15<sup>th</sup>; as a result, the deadline to protest was June 15<sup>th</sup>.

One key feature of this setting that is important for the interpretation of the results is that there is significant ambiguity in estimating market values. Because conducting full in-person appraisals is prohibitively expensive, the DCAD has to come up with its best guess for the market value of each property using statistical models and large datasets (e.g., recent home sales). The imperfections in these estimates are perhaps best illustrated by publicly-available data from websites such as Zillow.com and Redfin.com. When these companies publish estimates of the market value for the same property, their estimates tend to differ significantly, especially if that property has not been on the market recently. This ambiguity in market values leaves room for the owners to complain about the DCAD’s value assessments.<sup>19</sup> In a sense, households are not really “correcting” estimates that are obviously wrong. Instead, they are simply presenting a data point (e.g., the sales price of a neighboring home) to support their protest.

Based on our conversations with officers from some of the county appraisal districts in

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<sup>17</sup> While there is not any publicly available data on who entered into these long-term contracts, we do find some suggestive evidence in the protest data: households that protested through an agent in a given year have a high likelihood of protesting again through an agent in the following year. For instance, of the homeowners who protested through agents in 2019, 62.67% protested again through agents in 2020; in contrast, of the homeowners who protested directly in 2019, only 28.62% protested again directly in 2020.

<sup>18</sup> For the sake of brevity, in this study we refer to *taxes*, but all analysis is technically based on *estimated taxes*. See Appendix A.1.1 for more details. The 2019 and 2020 jurisdictional tax rates were quite similar except for the city jurisdiction of Cockrell Hill (which dropped from 0.95% in 2019 to 0.85% in 2020) and Garland ISD (which decreased from 1.39% in 2019 to 1.26% in 2020).

<sup>19</sup> Households may have an advantage over the DCAD in that they know more about the specific attributes and condition of their own homes. On the other hand, households face a significant informational disadvantage in that they do not have access to the same models, data and expertise available to DCAD.

Texas, their prevailing view is that households use the subjective nature of the appraisal process not to complain about the county’s estimate of their home value *per se*, but simply as an excuse to complain about their taxes being too high. We provide suggestive evidence in support of this view: according to an independent estimate of household market value (Redfin), households still file protests when their properties have been under-assessed by the government.<sup>20</sup> This view is also consistent with household responses to open-ended questions from our survey mentioning that taxes are too high as their motivation to protest.<sup>21</sup>

Filing a protest directly is simple. Homeowners can protest using a paper form mailed by the DCAD to households that received a notification by mail because the proposed value increased relative to the previous year, a form from the Texas Comptroller that can be printed from the Internet, or via a simple online tool called uFile. To protest online, households need to search for their own name or address on a website, click on their account, and then follow some straightforward steps. In 2020, 75% of direct protests were filed online while the remaining 25% were filed by mail.

Protests can be resolved at different stages. Some protests are resolved because the owners accept the settlements proposed by the county. These settlements may be offered through informal channels, such as an email or phone exchange with a staff member from the DCAD. If an agreement is not reached, the protest advances to a formal hearing with a quasi-judicial entity called the Appraisal Review Board.<sup>22</sup> The formal hearing entails no risk: if the DCAD schedules a hearing and households do not attend, the protest is simply dismissed with no penalty.<sup>23</sup> After protests are resolved one way or another, the final assessed home values (from hereon, “certified” values) and tax amounts are calculated. Taxes become payable on October 1<sup>st</sup>, 2020 and, if unpaid, become delinquent after January 31<sup>st</sup>, 2021.

To aid in the interpretation of the results, we provide some basic descriptive statistics about the households in the sample. We focus on the sample of 423,607 single-family homes that were subject to property taxes in Dallas County in 2020, after excluding some potentially problematic cases such as households with missing data.<sup>24</sup> The average home in this sample

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<sup>20</sup> Results reported in Appendix A.1.4.

<sup>21</sup> Our Field Survey (introduced in Section 4.2 below) included a final, open-ended question for subjects who wanted to share any thoughts with the researchers. Many households took this opportunity to mention their tax burden as a motivation for protesting.

<sup>22</sup> Formal hearings are typically conducted in person before a panel of three independent board members proposed by the DCAD and appointed by the Local Administrative District Judge of Dallas County. In response to the COVID-19 emergency in 2020, the DCAD staff did not conduct face-to-face negotiations, and all settlements were offered via email or telephone. Formal hearings were conducted over the phone with a single board member. If all else fails, the homeowner has the option to contest the decision in court.

<sup>23</sup> Of the 2020 direct protests that contain information on the form in which it was resolved, we find that 43.7% were settled informally, 35.2% were settled after a formal hearing, and 21.1% were either withdrawn or dismissed.

<sup>24</sup> For more details on the definition of this sample, see Appendix A.1.3.

has a value of \$306,000 and pays \$6,150 annually in property taxes. For a subsample of these subjects, we obtained individual-level demographic data from a private vendor.<sup>25</sup> The average subject is 52 years old, 65% are White, 9% are African-American and 20% are Hispanic. In different parts of the research design we focus on different subgroups of the main sample of 423,607 single-family homes, such as homes with homestead status or homes that were selected to participate in the field experiment.<sup>26</sup>

In a given year, only a small share of households file a protest. In 2020, for example, 8.40% of homeowners in this sample protested directly.<sup>27</sup> If we include protests through agents too, the protest rate is 16.83%. This rate of protests has been quite stable in recent years: e.g., in the same sample of households from 2020, we find that 13.82% protested in 2017, 15.09% protested in 2018, and 13.89% protested in 2019. Even when looking at longer time horizons, it is still true that a minority of individuals file a protest: e.g., in the same sample of households from 2020, 23.5% protested directly (34.5% overall) at least once in the five-year period between 2015–2019.<sup>28</sup> The probability of protesting is not distributed evenly across households – for example, richer households and households who face higher tax burdens are significantly more likely to protest.

## 3 Private Benefits

### 3.1 Conceptual Framework

In this section, we use quasi-experimental variation in the pecuniary incentives to protest provided by Texas’ property tax regulations. In Texas, homeowners may apply for homestead status for their primary residence. Among other benefits, the Texas Property Code guarantees that any increase in the appraised value of a homestead property is limited to 10% per year, which is referred to as the *homestead cap*. This regulation generates a sharp kink in the expected benefits from protesting. We exploit this kink as a natural experiment.

In practice, the amount of taxes that a household pays is calculated through a formula that involves the proposed value and the tax rates for the various jurisdictions within the county as well as other factors, such as the homestead cap and tax exemptions. Because households have the opportunity to protest every year, dynamic considerations may arise,

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<sup>25</sup> The company used the names and addresses to merge the records at the individual level. For more details about this data, see Appendix A.1.2.

<sup>26</sup> In Appendix A.1.3 we provide descriptive statistics for each of the subsamples used in the study, and show that they are roughly similar in a number of key respects.

<sup>27</sup> In all the analyses presented in this paper, we include protests that were marked as received by the DCAD through July 15<sup>th</sup>, 2020. For more details, see Appendix A.1.2.

<sup>28</sup> For more details, see Appendix A.1.4.

too. For the sake of simplicity and to fix the intuition for the empirical analysis, however, we now introduce a simple model of the decision to protest. Let  $A$  be the proposed value of the household and  $T$  be the amount the household has to pay in property taxes. Under a simple proportional tax rate ( $\tau$ ), the tax burden without a homestead cap is the following:

$$T_{nocap} = \tau \cdot A \quad (1)$$

Let  $C$  denote the cost of protesting. Assume that households can protest ( $P = 1$ ) or not ( $P = 0$ ), and let  $\Delta_A \geq 0$  be a random variable that corresponds to the reduction in  $A$  that would result from a protest. Then the expected net benefit from protesting is:

$$\mathbb{E}[U(P = 1) - U(P = 0)]_{nocap} = \tau \cdot \mathbb{P}(\Delta_A > 0) \cdot \mathbb{E}[\Delta_A | \Delta_A > 0] - C \quad (2)$$

and the household will protest if the above expected net benefit is positive and will not protest if it is non-positive. Now, let us introduce the homestead cap. Let the cap threshold be  $\bar{A}$ . Taking this threshold into consideration, the tax burden can be computed as follows:

$$T_{cap} = \tau \cdot \min\{A, \bar{A}\} \quad (3)$$

If the cap is not binding ( $A < \bar{A}$ ), then  $T_{cap}$  is identical to  $T_{nocap}$ , and thus the decision to protest is not affected by the homestead cap. The interesting case is when the cap is binding ( $A > \bar{A}$ ). As a result of a binding cap, the expected net benefit from protesting is as follows:

$$\mathbb{E}[U(P = 1) - U(P = 0)]_{cap} = \tau \cdot \mathbb{P}(\Delta_A > A - \bar{A}) \cdot \mathbb{E}[\Delta_A - (A - \bar{A}) | \Delta_A > A - \bar{A}] - C, \quad (4)$$

This equation can be re-arranged as follows:

$$\begin{aligned} \mathbb{E}[U(P = 1) - U(P = 0)]_{cap} = & \tau \cdot \mathbb{P}(\Delta_A > 0) \cdot \mathbb{E}[\Delta_A | \Delta_A > 0] - C \\ & - \tau \cdot \mathbb{P}(0 < \Delta_A < A - \bar{A}) \cdot \mathbb{E}[\Delta_A | 0 < \Delta_A < A - \bar{A}] \end{aligned} \quad (5)$$

Note that first two terms in the RHS in equation 2 are identical to the first two terms on the RHS in equation 5. Thus, the last term in equation 5 is the difference in incentives to protest introduced by the cap. The cap reduces the expected benefits from protesting when it is binding. Note that the expected benefits are lower the larger the difference between the proposed value and the homestead threshold ( $A - \bar{A}$ ). The intuition is straightforward: absent a cap, a reduction in the assessed value will result in a reduction in the tax bill. When a household's proposed value is above the cap, however, the marginal reduction in the assessed value may not affect the final tax bill. If the proposed value is just \$1 above the cap, then the first dollar reduction in the assessed value will not affect the tax bill but every dollar after that will. In that case, the cap should matter little to the household's decision to protest. However, if the proposed value is \$15,000 above the cap, then none of

the first \$15,000 reduction in the assessed value will affect the tax burden, and thus the cap will substantially affect the expected benefits from protesting: the household will only see a reduction in the tax bill if the home value assessment is revised downwards by more than \$15,000.

## 3.2 Results

Our analysis of the effects of the homestead cap on protest rates is based on the main sample of 423,607 single-family homes. About 74% of these households were approved for homestead status in 2020, and thus their homestead caps may be binding. The remaining 26% of households constitute the basis for a falsification test.

Figure 1 summarizes the main results. This figure is a binned scatterplot of the relationship between a given outcome and the distance between each household’s proposed value ( $A$ ) and its homestead cap threshold ( $\bar{A}$ ). To be conservative, these baseline results are based on a narrow band around the homestead cap threshold (\$15,000 above and below). In each panel, the horizontal axis measures the distance to the homestead cap threshold. The blue dots correspond to the households with proposed values below the homestead cap threshold, with the blue line corresponding to the linear fit. The coefficient from the linear regression also is reported in blue. For ease of exposition, we normalize all coefficients so that they correspond to the effects from a \$10,000 increase in the proposed value. In turn, the red dots and red lines correspond to the households with proposed values above the homestead cap threshold. For the sake of simplicity, Figure 1 includes a minimal set of controls (the proposed value, a dummy for whether the household protested in the previous year, and a set of school district dummies).

The two panels on the left of Figure 1 (1.a and 1.c) correspond to the properties with homestead status, for which the homestead cap threshold can be binding. The two panels on the right of Figure 1 (1.b and 1.d) correspond to properties without homestead status, for which the homestead cap threshold should be irrelevant and thus they can serve as a falsification test.

We start with Figure 1.a, in which the outcome variable measured on the vertical axis corresponds to the tax amount based on the 2020 proposed value.<sup>29</sup> The blue slope in the left half of the figure corresponds to properties right below the homestead cap threshold, and the red slope in the right half of the figure corresponds to the properties right above the threshold. Note that each slope is just associations and thus only the difference between the two slopes can be interpreted as a causal effect. As expected, there is a sharp kink

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<sup>29</sup> For an alternative measurement, Appendix A.2.1 reproduces the whole exercise but using the tax rate as a dependent variable instead of the tax amount.

at the threshold: upon reaching the homestead cap threshold, households pay a lower tax amount than they would have without the homestead cap. This kink is large in magnitude and statistically significant. We therefore reject the null hypothesis that the slope to the left of the threshold (-97) is equal to the slope to the right of the threshold (-306), with a  $p\text{-value} < 0.001$ .

The following thought experiment illustrates how the homestead cap affects the marginal benefits from protesting. Consider a household with a proposed value that is \$10,000 above the homestead cap threshold, that is, at +\$10K on the x-axis of Figure 1.a. The vertical gap between the red line and the dashed blue line, estimated at around \$209, corresponds to the tax amount that is capped.<sup>30</sup> What happens when a successful protest decreases the proposed value by \$10,000? In the presence of the homestead cap, the \$10,000 reduction in household value would not affect the tax amount of the household. That is, in the presence of the homestead cap, the household would not benefit from the \$209 reduction in taxes because the household already benefited from that reduction due to the homestead cap.

In sum, for a household at +\$10K on the x-axis of Figure 1.a, the marginal benefit from protesting decreases by \$209. If households care about the private benefits from protesting, we expect the \$209 reduction in the marginal benefit from protesting to reduce the probability of protesting. To explore this expectation, Figure 1.c is identical to Figure 1.a except that instead of the tax amount, the vertical axis corresponds to the protest rate (i.e., an indicator variable that takes the value 100 if the owner protested directly in 2020 and 0 otherwise). As expected, we find a sharp kink in Figure 1.c at exactly the homestead cap threshold. Again, this kink is large in magnitude and statistically significant. We can reject the null hypothesis that the coefficient to the left of the threshold (2.881) is equal to the coefficient to the right of the threshold (-1.604), with a  $p\text{-value} < 0.001$ .

We can combine the results from Figure 1.a and Figure 1.c to quantify the magnitude of the effects of the homestead cap. Figure 1.a indicates that each additional \$10,000 above the homestead cap causes an average \$209 reduction in the tax amount. In turn, Figure 1.c indicates that being \$10,000 above the homestead cap causes a reduction in the protest probability of 4.485 pp. Taking the ratio of these two estimates, we conclude that for each \$100 reduction in the tax amount, the protest probability decreases by 2.14 pp. Moreover, we can express this effect as an elasticity. On the one hand, the \$100 reduction in the tax amount corresponds to 20.6% of the average savings (\$485) among successful protests in 2020. On the other hand, the 2.14 pp reduction in the protest probability corresponds to 21.4% of the average protest probability in this sample (9.98 pp). These percentage changes suggest

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<sup>30</sup> The red regression line predicts that when the proposed value is \$10,000 above the homestead cap, the tax amount is \$306 lower. The dashed blue line predicts that when the proposed value is \$10,000 above the homestead cap, the tax amount is \$97 lower. The difference between these two figures is \$209 (\$306 - \$97).

a 1.04 ( $= \frac{21.4}{20.6}$ ) elasticity in protest probability, with respect to the expected tax savings. In other words, the decision to protest is elastic to the private benefits from protesting.

### 3.3 Robustness Checks

The previous analysis corresponds to properties that have a homestead status exemption (and thus are subject to the homestead cap). Next, we reproduce the analysis using properties that do not have a homestead status exemption (and thus are not subject to the homestead cap). This analysis provides a sharp falsification test because we should not observe any kinks in this latter group. A kink would suggest that the effects are not due to the homestead cap but instead due to some other confounding factor. The two panels in the right half of Figure 1 (i.e., 1.b and 1.d) correspond to the properties without homestead status for which the hypothetical homestead cap threshold is defined as 110% of the assessed value in the previous year (2019). As expected, on the right-side panels of Figure 1, we find no kinks at the homestead cap threshold; thus, we cannot reject the null hypotheses that the coefficients are equal below and above the homestead cap threshold. Most importantly, the coefficients are precisely estimated on the right-side panels of Figure 1, meaning that we can rule out any large kinks shown in the left-side panel of Figure 1, as well as small kinks.

For a second falsification test, we follow the logic of an event-study analysis. We reproduce the analysis from Figure 1.c for the properties with homestead status, but the dependent variable is whether the household protested in 2019 instead of 2020. Whether the 2020 proposed value ends up above or below the 2020 homestead cap should not affect whether a household protested in 2019. In fact, a significant difference would suggest a confounding factor affecting the results from Figure 1.c. Moreover, we can extend this logic and reproduce the analysis for one year prior (2019) and for each year for which we have data (2015–2018). Figure 2.a presents the results from this event-study analysis. The rightmost coefficient (year 2020) corresponds to the effect of the homestead cap on the protest rate (i.e., the difference between the two slopes reported in Figure 1.c). The rest of the coefficients are estimated with an identical regression, except using a dependent variable. For example, the 2015 coefficient corresponds to a regression where the dependent variable takes the value 100 if the household protested in 2015 and 0 otherwise.<sup>31</sup> As expected, the coefficients are consistently close to 0 for each falsification year (2015-2019) and always highly statistically different from the

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<sup>31</sup>Note that the set of control variables will not be identical. For example, when the dependent variable is the 2020 protest choice, we control for whether the household protested in 2019, but when the dependent variable is the 2019 protest choice, we control for whether the household protested in 2018.



coefficient for 2020.<sup>32</sup>

In Appendix A.2.2, we present several additional robustness checks. We show that, in addition to affecting direct protests, the homestead cap affects agent-assisted protests in the same direction, but the additional effects are smaller in magnitude, as expected. In the baseline results, we use a conservative specification based on a narrow band of \$15,000 around the homestead cap threshold. The results are qualitatively and quantitatively similar if we use alternative bands. Last, we show that the homestead cap is consequential not only for the number of protests but also for the subsequent market values and tax amounts.

## 4 Private Costs

### 4.1 Conceptual Framework

These results suggest that private benefits play a significant role in the decision to protest taxes. In this section, we explore the other side of this relationship: if households are responsive to private benefits, they should be responsive to private costs too.

As there are no fees for filing a property tax protest in Texas, the private costs we refer to are non-pecuniary, or hassle, costs. Filing a protest directly is simple in theory, but it may not be simple in practice. Some households may not even know where to start, or they may think the process is more difficult than it actually is. Other households may be less sophisticated and thus need step-by-step guidance on how to protest. Indeed, instructions on how to protest are not readily available. At the time of the experiment, only one official source online had instructions on how to file a protest: a PDF document posted on the DCAD website.<sup>33</sup> However, this long document includes broad instructions and is difficult to locate on the DCAD website. Other unofficial online sources, such as blog posts, were usually incomplete, outdated, and difficult to find. Moreover, those sources often had a commercial interest, deliberately depicting the protest process as more complicated than it really is.

The hassle costs may be particularly large for a particular step in the protest procedure: providing an opinion on the value of the home and an argument supporting that opinion. This information can be obtained in different ways, according to anecdotal evidence, and usually involves a comparable property that recently sold for less than the proposed value of the contested property. The recent sale price serves as the opinion of value, and information about the recent transaction can be used as the argument. Finding a proper comparison

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<sup>32</sup> Two of the falsification coefficients (for years 2015 and 2017) are borderline statistically significant but still small in magnitude. Given that we estimate a total of 25 falsification coefficients in Figure 2, we expect a few to be statistically significant just by chance.

<sup>33</sup> This document can be found at [http://www.dallascad.org/Forms/Protest\\_Process.pdf](http://www.dallascad.org/Forms/Protest_Process.pdf).

property entails several steps. First, the homeowner could use a free online real estate service, such as Zillow.com or Redfin.com, to search for and identify comparable properties that sold recently for less than the homeowner’s proposed property value. This seemingly straightforward process could be daunting for people with limited Internet access and skills or low financial literacy. Indeed, plenty of evidence indicates that households have trouble finding even easily accessible information, such as information on the inflation rate or changes in average home prices (Cavallo et al., 2017; Bottan and Perez-Truglia, 2020a).

## 4.2 Experimental Design

We designed a mailing intervention aimed at reducing the hassle costs from protesting. If private costs dissuade homeowners from protesting, our intervention should increase the probability that they file a protest.

Subjects in our sample were randomly assigned to receive no letter or a letter. Figures 3 and 4 show the first and second pages of a sample letter, with the addition of red boxes highlighting the parts that were randomized.<sup>34</sup> The letters were sent on behalf of researchers at The University of Texas at Dallas and included several measures to show that they came from a legitimate source.<sup>35</sup> Letters were tailored to recipients: each salutation at the top of the first page included the recipient’s name, and their names and addresses were printed at the bottom of the second page (which appeared through the envelope window).<sup>36</sup>

In the first treatment arm, subjects were randomized to receive one of two types of letters: the *basic aid letter* or the *extra aid letter*. The *basic aid letter* provided useful tips to help the recipient file a protest, all of which were printed on the *first* page of the letter (see Figure 3 for a sample). A key part of the first page is that it included instructions on how to file a protest using the project’s website, which we designed to be concise, easy to follow, and as explicit as possible. Appendix D shows screenshots of the entire website, including the step-by-step instructions on how to file a protest online or by mail. These walkthroughs included hyperlinks to relevant websites and screenshots of a sample protest using information from a fictitious household for added clarity.

The *extra aid letter* is identical to the basic aid letter, plus additional guidance on how

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<sup>34</sup> See Appendix C for a full-page sample of the letter without the additional red boxes.

<sup>35</sup> The envelope featured the logo of The University of Texas at Dallas, a well-known institution in Dallas County; the name of one of the professors from that university; and non-profit organization postage. See Appendix B for a sample. The letter also featured the official logo of The University of Texas at Dallas in the header, as well as a physical address and the URL of the study’s website. The website provided general information about the study (without discussing any hypotheses or what the study was about) and contact information for the researchers and the institutional review board.

<sup>36</sup> In cases where properties were jointly owned by multiple individuals (typically, husband and wife), we sent a single letter addressed to all listed owners.

to protest. Figure 4.a shows an example of the second page of the letter for a recipient assigned to the basic aid treatment, and Figure 4.b shows an example of the second page for a recipient assigned to the extra aid treatment, with the extra aid message outlined by a red box with dashed lines (we added this box for expositional purposes only and did not include it in the actual letters sent to subjects). The extra aid message provided the previously described information on how to reduce the hassle costs of providing an opinion of value and supporting argument.

The first paragraph of the extra aid message offered some facts about the protest filing process, explaining that the simple process could be done without an agent and may not require a hearing (which could be intimidating to some subjects) for the DCAD to propose a settlement offer. Moreover, if a hearing were to be scheduled, there would be no risk for not attending. The message also provided an argument to be used in the protest. Specifically, we presented the most common approach, that is, arguing that based on a recent sale price of a comparable property, the proposed value for the property exceeds the market value.<sup>37</sup> To further simplify the use of this information, we presented this information as it would look on the actual protest form, with a check mark in the “Value is over market value” box, the comparable sale price in the “Opinion of value” field, and a sample handwritten note outlining a usable argument. For example, in the sample letter shown in Figure 4.b, the handwritten note reads, “I found a home that is similar to mine but was recently sold for less than my home’s appraised market value. The property located at 2234 Meadowstone Dr. (Carrollton, TX) is 0.20 miles away from my home, and has the same number of bedrooms and a similar square footage. That property was sold on 10/31/2019 for \$160,000.” Households could use our proposed argument verbatim, but to clarify that our content was just a suggestion, we included the following: “You can find information about this sale by searching for the property’s address on Zillow.com or Redfin.com. On these websites you can find other comparable properties to support your protest.” Additionally, we mentioned that subjects could protest based on different arguments, offering the following message: “You can also protest based on the appraised market values of comparable properties, which can be found on [www.dallascad.org/SearchAddr.aspx](http://www.dallascad.org/SearchAddr.aspx).”

We created an algorithm that identified one comparison property for each household by combining data from the tax rolls with data from recent property sales from Redfin.com. For each recipient, the algorithm searched similar properties (e.g., same number of bedrooms, bathrooms, square footage, location) that were sold in late 2019 or early 2020 for between 5% and 20% less than the proposed value of the recipient’s property.<sup>38</sup> We believe that

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<sup>37</sup> We identified one comparable property for all households in the subject pool, but we included this information only for recipients who were randomly selected for the extra aid letter.

<sup>38</sup> In Appendix A.3, we provide details about this algorithm and some descriptive statistics.

an agent hired by the recipient to protest would follow a similar (or even the exact same) procedure.

Our letters were not designed to eliminate the hassle costs fully, as recipients would still need to spend time filing and monitoring their protests, which has an opportunity cost (Goldszmidt et al., 2020). Some also may find the associated paperwork to be considerably unpleasant (Benzarti, 2020, 2021). In that sense, our estimates provide a lower bound of the full private costs of protesting.

Some additional features of the letter are summarized here but discussed in detail in Section 5. In the middle of the first page, all letters included a table with information related to the proposed values and estimated taxes of the recipient’s property. At the bottom of the first page, all letters included a URL to an online survey (hereafter, we refer to this as the *Field Survey*). To verify that recipients were legitimate subjects and to link survey responses at the household level, we included a unique five-letter code for survey access. The goals of the Field Survey were to proxy for the dates that recipients opened the letters (Perez-Truglia and Cruces, 2017; Bottan and Perez-Truglia, 2020a) and to include questions for use as outcomes in the analysis, as discussed in Section 5.

### 4.3 Subject Pool and Implementation Details

From the main sample of 423,607 residential single-family properties, we focused on a subgroup of 78,462 homes for our field experiment subject pool. To arrive at this subsample, we excluded households that had already filed a protest according to the latest available data from DCAD because our letter could not affect their behavior.<sup>39</sup> The most important condition was to focus on households for whom our algorithm could find comparison properties for use in the extra aid message.<sup>40</sup> Though not identical, the subject pool is similar to the main sample in observable, pre-treatment characteristics.<sup>41</sup> Additionally, Appendix A.3 shows that, consistent with successful random assignment, the observable pre-treatment characteristics are balanced across all treatment groups.

We timed the intervention so that our letters would arrive early enough before the protest deadline to influence the recipient’s decision. We created the letters as soon as the administrative data, including 2020 proposed values, became available (on May 16<sup>th</sup>, 2020). To accelerate delivery, we used a mailing company in Dallas County (i.e., the same county as all recipients). The mailing company dropped off the letters at the local post office on May

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<sup>39</sup> We initially selected a sample of 79,322 properties. However, due to a lag of a few days in the way DCAD reports data, we dropped 860 from the subject pool because we discovered that they had already protested by the time we mailed the letters.

<sup>40</sup> For more details, see Appendix A.1.2.

<sup>41</sup> See Appendix A.1.3.

20<sup>th</sup> and estimated that most would be delivered in the next couple of days. Consistent with this projection, we began to receive Field Survey responses and website visits on May 21<sup>st</sup>. Moreover, we confirmed that the post office scanned more than 90% of our letters by Friday, May 22<sup>nd</sup>, 2020, indicating they reached the last mile before delivery. Based on data from previous years, most subjects file protests close to the deadline, which in 2020 was June 15<sup>th</sup>. Thus, we feel confident that there was enough time between receipt of the letter and the protest deadline for the letter to influence most recipients’ decisions to protest. Indeed, this view is consistent with the dates when subjects responded to the Field Survey and when they visited the project’s website.<sup>42</sup>

## 4.4 Econometric Model

We want to compare the probability of protesting between subjects who were sent a letter and subjects who were not sent a letter, as well as between subjects assigned to different types of letters. The main outcome,  $P_i^{2020}$ , is an indicator variable that takes the value 100 if the subject filed a protest in the post-treatment period. We use a simple linear probability model:

$$P_i^{2020} = \eta_0 + \eta_{basic} \cdot L_i^{basic} + \eta_{extra} \cdot L_i^{extra} + X_i^{pre} \eta_X + \epsilon_i \quad (6)$$

The variable,  $L_i^{basic}$  is an indicator that takes the value 1 if the household was mailed a basic aid letter and 0 otherwise. Likewise,  $L_i^{extra}$  is an indicator that takes the value 1 if the household was mailed an extra aid letter. Last,  $X_i^{pre}$  is a vector of pre-treatment controls. Unless noted otherwise, we use this exact same set of control variables in all regressions in the paper: the proposed value in levels and its annual growth, dummies for multiple owners, school and special districts, number of years since the last property tax protest, a dummy for homestead status, growth in the proposed value relative to the previous year and for each year from 2015 to the previous year, and a dummy indicating if the household protested in that year and the outcome of the protest. Given that this is an experiment, the only goal of using pre-treatment controls is to gain statistical power by reducing the variance of the error term (McKenzie, 2012). Additionally, we use the pre-treatment data to construct falsification tests.

## 4.5 Results

Table 1 presents the regression results. All regressions are based on the same specification given in equation (6), but they differ according to the dependent variable. The dependent

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<sup>42</sup> Results reported in Appendix A.4.1.

variable in column (1) takes the value 100 if the owner protested directly and 0 otherwise. The basic aid letter increased the probability of protesting by 1.792 pp, an effect that is highly statistically significant (p-value<0.001). The extra aid letter increased the protest probability even more, by 3.509 pp (p-value<0.001).

We conduct a falsification test in the spirit of event-study analyses. Figure 2.b presents the results. The rightmost coefficient shows the effects of each type of letter on the probability of protesting in 2020, and they are identical to the two coefficients reported in column (1) of Table 1. The rest of the coefficients correspond to the same regression specification, but the dependent variables are protest indicators for the years 2015 through 2019, instead of 2020. As our letters had not been sent yet, they could not possibly affect protests in prior years. As expected, the coefficients for the pre-treatment years are close to 0, statistically insignificant, and precisely estimated.

One potential concern is that our letters induced protests that were ultimately not successful. We can provide direct evidence that this was not the case. In column (2) of Table 1, the dependent variable takes the value 100 if the household protested directly and won (i.e., received a discount on their market value assessment) and 0 otherwise (i.e., the protest was unsuccessful or the household did not protest directly). The coefficients remain economically and statistically significant. The ratio between the coefficients on the extra aid letters in columns (1) and (2) suggests that 75% ( $= \frac{2.621}{3.509}$ ) of the marginal protests that were induced by our extra aid letter were successful. This success rate is comparable to the corresponding success rate of 78% ( $= \frac{6.76}{8.67}$ ) observed in the control group.<sup>43</sup> In other words, the additional protests induced by our letters were, on average, roughly as successful as the protests in the control group.

In Appendix A.4.2, we present several additional robustness checks. We show that the increase in direct protests induced by our letters did not crowd out protests through agents. We report the effects of the letters on online protests separately from protests by mail. Last, we show the extent to which the marginal protests induced by our letters were successful under alternative definitions of success: reduction in market value and reduction in estimated taxes.

## 4.6 Causal Mechanisms

In this section, we unpack evidence related to the causal mechanisms behind the effects of the letters.

Our favorite interpretation is that the letters increased the likelihood of protesting by mitigating the underlying hassle costs. An alternative interpretation could be that our letters

<sup>43</sup> This success rate is based on the ratio of the share of direct protests that were successful (6.76, from column (2)) to the share of direct protests (8.67, from column (1)).

acted as a reminder of the opportunity to protest. This explanation is unlikely in our context, however, as proposed property taxes are quite salient around the time subjects received our letter. Moreover, we have several pieces of evidence indicating that the reminder effect was probably minor.

The first, and most direct, piece of evidence is based on a comparison between the two types of letters. If the letters acted as a simple reminder, then their effects should have been the same regardless of whether the letter included the extra aid or not. Column (1) of Table 1 indicates that the effect of the basic aid letter was positive (1.792 pp) and statistically different from the effect of the extra aid letter (3.509 pp). The difference between the coefficient estimates indicates that the extra aid message, on its own, had an effect of 1.717 pp ( $= 3.509 - 1.792$ ), which was highly statistically significant ( $p$ -value $<0.001$ ). This result demonstrates that the reminder mechanism might explain the effects of the basic aid letter only.

However, we provide direct evidence that the reminder effect did not even play a significant role for the basic aid letter. For this test, we exploit heterogeneity in the mailing of a DCAD Notice of Appraised Value to households (see Appendix E for a sample notification). Starting on May 15, 2020, all homeowners could download their notice at the DCAD webpage. On the same day, DCAD mailed these notifications to households meeting certain criteria (e.g., increased appraised value, ownership change, loss of homestead exemption, rendered property, or new property).<sup>44</sup> We mailed our letters on Wednesday, May 20, 2020. Thus, households should have received the official DCAD notification around five days before our letter. If our basic aid letter worked primarily through a reminder effect, it should have had a larger effect on households that did not receive the DCAD letter, as the other households would have been reminded about the opportunity to protest through the DCAD letter. In columns (3) and (4) of Table 1, we split the results of the sample into those who were mailed or not mailed a notification from DCAD, respectively. The effects of the basic aid letter on subjects that were and were not mailed a notice are on the same order of magnitude (coefficients of 1.449 and 1.935, respectively), and they are not statistically distinguishable ( $p$ -value $=0.317$ ). This finding suggests that the effect of the basic aid letter far exceeded a simple reminder effect.

Regarding the effects of the basic aid message, our preferred interpretation is that they were mainly the product of the walkthroughs provided on the project's website. A first piece of evidence for this interpretation relies on unsolicited feedback from participants. On the project's website, we provided an email address to contact the researchers with concerns about the research project. Several subjects sent emails expressing gratitude for the letter

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<sup>44</sup>DCAD did not mail notices if the property value did not change or if the property value decreased. Source: <http://dallascad.org/ViewPDFs.aspx?type=1&id=%5C%5Cdcad.org%5Cweb%5Cwebdata%5CHeadlines%5CHEADLERTRecentHeadlines04032020.pdf>.

and website, and many mentioned the hassle costs. For example, some mentioned that they had wanted to protest for years but did not know how until receiving our letter. Similarly, the Field Survey included a final, open-ended question for subjects to share any thoughts with the researchers. Many used that space to express gratitude, and some explicitly stated how the information in the letter and on the website helped them navigate the protest process. Another source of supporting evidence for this mechanism comes from data on our website traffic. We estimate that the basic aid message generated 903 additional direct protests.<sup>45</sup> We can compare this number of additional protests that were induced by the aid message to the 2,769 unique visits to the website walkthroughs, as recorded by Google Analytics (for more details, see Appendix A.4.1). Some of those visitors may have looked at the walkthroughs but did not protest, and some may have used the walkthroughs but would have protested even without them. If we assume that around one third of those website visitors were induced to protest by our website, that would explain all 903 additional protests generated by the basic aid message. In other words, it would not be far-fetched to attribute the entire effect of the basic aid message to the online walkthroughs.

Regarding the extra aid message, our preferred interpretation is that subjects either used our proposed argument *as is* or followed our instructions to come up with an argument of their own. Indeed, we can provide some direct evidence that some subjects used our suggested argument in the extra aid message “as is” to complete their protest form. This test is based on data from the 5,026 households in the subject pool who protested online. For these households, we can observe the opinion of value that they entered in the online form. In column (5) of Table 1, the dependent variable takes the value 100 if the household provided an opinion of value in their protest that is within half a percentage point of the value we suggested in their extra aid message.<sup>46</sup> The bottom of column (5) shows that, in the control group with no letter, there was a 3.37 pp chance that a household would enter an opinion of value that coincided almost exactly with the value that we would have suggested if they were assigned to the extra aid message treatment. In other words, it is unlikely for subjects to use an almost identical opinion of value as the one we would have suggested. For households that received the basic aid letter, that probability remained equally low, which is expected because the basic aid message did not include any information about the argument. For households that received the extra aid letter, the outcome variable increased by a whopping 15.287 pp (p-value<0.001). This evidence suggests that a substantial fraction of households

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<sup>45</sup> We arrive at this figure by taking the effect of the basic aid letter (1.792 pp, from column (1)) and multiplying it by the total number of letters sent that included at least the basic aid message (50,394).

<sup>46</sup> One limitation with this exercise, however, is that it is based on a subsample (households that protested online and entered a value in the Opinion of Value field) that is not random and thus introduces possibility endogeneity bias, despite the random assignment.



that received the extra aid letter copied the suggested amount directly into their protest forms.

## 4.7 Magnitude of the Effects

A challenge with interpreting the magnitudes of the effects in mailing experiments is non-compliance: for example, some households may not have received their letter or received a letter but did not read it. To correct for these types of non-compliance, we estimate the reading rate (i.e., the share of recipients that actually read the letter on time). Following Bottan and Perez-Truglia (2020a), we combine estimates from different sources to approximate the reading rate. According to the U.S. Monitor Non-Profit Standard Mail Delivery Study, around 95% of standard non-profit mailers are successfully delivered (U.S. Monitor, 2014). Based on data from the U.S. Postal Service Household Diary Survey (Mazzone and Rehman, 2019), we estimate that, conditional on delivery, around 74% of our letters were opened by the recipients.<sup>47</sup> If we combine these two estimates, we arrive at a reading rate of 70.3% ( $= 0.95 \cdot 0.74$ ). To account for this source of attenuation bias, we scale up the coefficients by a factor of 1.42 ( $= \frac{1}{0.703}$ ). The resulting scaled-up effects would be 2.55 pp for the basic aid letter and 4.98 pp for the extra aid letter, which are still conservative scale-up factors, as there may be additional sources of non-compliance.<sup>48</sup>

Another potential source of non-compliance is spillovers. If treated households shared information from the letters with neighboring households who were in the control group, that would introduce an attenuation bias in our estimates. However, we provide evidence that this form of non-compliance is negligible: Appendix A.4.3 shows that the estimated spillovers are statistically insignificant and precisely estimated at 0.

To translate the hassle costs into dollar amounts, we combine the results from the field experiment with the results on the homestead cap discussed in Section 3. We focus on the effect of the most comprehensive letter (the extra aid letter). The effect of this letter gives a lower bound on the costs from protesting, as this letter did not eliminate the hassle costs completely. For example, subjects still had to follow the instructions to file the form and may need to take further action in the future, such as discussing a settlement in informal or formal hearings. The scaled-up effect of the extra aid letter is 4.98 pp. According to the calculations reported in section 3.2, each \$100 reduction in the tax amount due to the homestead cap decreases the protest probability by 2.14 pp. These results imply that the

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<sup>47</sup> This figure is based on the 2018 HDS Recruitment Sample and corresponds to the estimate of treatment of advertising mail reported in Figure 5.3 of (Mazzone and Rehman, 2019). See Bottan and Perez-Truglia (2020a) for more details.

<sup>48</sup> For example, some households may have opened the letter too late, either after they filed a protest or after the protest deadline, whichever came first.

homestead cap would need to reduce the tax amount by \$232 to generate a reduction of 4.98 pp in the protest rate. These results imply that the average hassle cost is on the order of \$232.<sup>49</sup>

As a sanity check, we compare our estimate of hassle costs to the fees charged by agents that protest on households' behalf. We identify one such company that offers the service for a flat fee. Assuming that the marginal client of this firm is indifferent between hiring this agent or protesting directly, the flat fee should constitute a measure of the hassle cost of the marginal client.<sup>50</sup> In 2020, the flat fee was \$139 for properties assessed below \$200,000, and \$305 for properties assessed between \$200,000 and \$500,000.<sup>51</sup> Those flat fees (\$139 and \$305) are on the same order of magnitude as our estimated average hassle cost (\$232), thus suggesting that our estimates are on the right order of magnitude.

## 5 Conditional Cooperation

### 5.1 Conceptual Framework

Our evidence thus far suggests that households are elastic to the private benefits and costs of protesting. It also suggests that selfish motives are important, but it does not rule out fairness concerns. To address this, we measure conditional cooperation (i.e., whether households are more willing to tolerate taxes when they think other households in the county are paying their fair share).

Anthropologists often consider cooperation and reciprocity as features that are present among all people, or “human universals” (Brown, 1991). Economists also study cooperation and fairness concerns (see e.g., Andreoni, 1995; Fehr and Schmidt, 1999; Bolton and Ockenfels, 2000; Cappelen et al., 2013; Fehr and Schurtenberger, 2018). Robust evidence from well-designed laboratory experiments documents that conditional cooperation is frequent and robust when playing the public goods game, despite individuals' incentives to free ride and the absence of apparent private benefits from cooperation (Andreoni, 1995; Gächter, 2007). In our context, conditional cooperation is rooted in similar fairness principles: households may believe it is fair to pay taxes when other households also pay taxes. We designed a field

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<sup>49</sup> This is just a rough approximation. Among other things, it assumes that households care only about the costs and benefits this year, but in reality there may be dynamic considerations, too.

<sup>50</sup> This is just a rough approximation. First, the marginal household may be indifferent between protesting through an agent or not protesting at all. Second, there may be additional factors at play. For example, the marginal customer may be willing to pay more than the hassle costs under the belief that the agent can negotiate higher tax savings.

<sup>51</sup> Source: [https://www.dfwtaxadvisor.com/practice\\_areas/property-tax-protest/](https://www.dfwtaxadvisor.com/practice_areas/property-tax-protest/), accessed on May 15, 2020.

experiment to test this hypothesis.

## 5.2 Design of the Field Experiment

To measure conditional cooperation, an ideal experiment would randomize how much others pay in taxes and measure whether these randomly assigned average tax rates affect individuals' decisions to protest. However, randomizing the average tax rate is not feasible.<sup>52</sup> Instead of manipulating the average tax rates directly, we manipulate the subject's perception of the average tax rate through an information-provision experiment. The goal is to leverage the fact that information on how much others pay in taxes is not easily accessible and thus probably not well-known; for example, some households may underestimate the average tax rate while others may overestimate it. Hence, by providing households with accurate information about the average tax rate, we can induce exogenous shocks to those perceptions. Then, we can measure if these information shocks affect whether the households feel their taxes are fair and whether they choose to protest them.

This information-provision experiment is the second treatment arm of the field experiment introduced in Section 4. This treatment arm randomizes the content of the table we included in the middle of the first page of the letter. In Figure 3, this table is highlighted inside a red box with dashed lines (the box is for explanatory purposes and was not included in the actual letters sent to subjects). All letters include a table, but we randomize (with 50% probability) whether the table adds a column showing figures for the average Dallas home (positioned in the last column). This column shows whether the recipient's tax rates are above or below the average Dallas home and by how much. Additionally, we cross-randomize whether the table includes a last row that makes tax rates more salient and explicit. As stated in the RCT pre-registration, this additional treatment arm is intended as a minor robustness check and thus is discussed in the Appendix.<sup>53</sup>

The random inclusion of an additional column is meant to provide a shock to households' perceptions of the tax rate that other households pay. However, a naïve comparison of the protest rate among households that received the information shock and those that did not would deliver incorrect estimates to the extent that households update heterogeneously in response to the information shock (Cruces et al., 2013; Cullen and Perez-Truglia, 2018;

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<sup>52</sup> This logic is based on the implicit assumption that households care about the tax rate that other households pay. In theory, households could care about the average tax *amount* instead of the average tax *rate*. The main motivation for using the tax *rate* is that survey data indicate that most households think that it would be fair for all households to pay the same property tax rate. Appendix A.5.1 reports the results.

<sup>53</sup> The randomization of the last row aims to address the concern that households compare tax rates (instead of tax amounts) by making them salient. For more details on the design of this treatment arm, see Appendix A.3. For the results from this treatment arm, see Appendix A.5.5. We find similar results regardless of whether the tax rate was made salient.

Bottan and Perez-Truglia, 2020a; Fuster et al., 2018). For example, suppose that half of the sample was originally downward-biased and thus updated upwards, whereas the other half was originally upward-biased and updated in the opposite direction. In this case, despite the fact that households react to the information, we would find a null average effect of information disclosure because the two opposite effects cancel each other out. To overcome this limitation, and as anticipated in the pre-registration, we designed the experiment based on the disclosure-randomization design from Bottan and Perez-Truglia (2020a), which is described in detail in the next section.

The main goal is to measure the effects of the information-provision experiment on the decision to file a protest. Additionally, in the letter, we include two questions from the Field Survey to be used as outcome variables. Appendix F provides a copy of the full survey instrument. The first question asks whether it is likely that the individual will file a protest this year. The goal of this question is to provide a validation exercise. The second and most important question is designed to test the causal mechanism at play. It elicits whether the recipient thinks that their own taxes are fair or unfair, relative to other households in the county, on a subjective scale from 1 to 10. The hypothesis is that, according to conditional cooperation, finding out that the average tax rate is higher should reduce the feeling of unfairness.

### 5.3 Econometric Model

Let  $Y_i^{post}$  be the outcome of interest, measured after the information provision experiment. For example, this could be an indicator of whether the individuals filed a protest post-treatment. Let  $\tau_i$  be the individual’s own tax rate and  $\bar{\tau}$  be the *actual* average tax rate in the county. Let  $D_i$  be an indicator variable that takes the value 1 if the information on the average tax rate was shown to the subject. The regression of interest is the following:

$$Y_i^{post} = \nu_0 + \nu_1 \cdot D_i \cdot (\bar{\tau} - \tau_i) + \nu_2 \cdot (\bar{\tau} - \tau_i) + \nu_3 \cdot D_i + X_i^{pre} \nu_X + \varepsilon_i \quad (7)$$

As before,  $X_i^{pre}$  corresponds to the vector of pre-treatment control variables, which contains the same variables listed in Section 4.4. Again, this is an experiment, so the goal of using pre-treatment controls is to gain statistical power by reducing the variance of the error term (McKenzie, 2012). We also use the pre-treatment data to construct falsification tests in an event-study fashion.

The coefficient  $\nu_2$  measures the relationship between the outcome and the potential information shock (i.e.,  $(\bar{\tau} - \tau_i)$ ) when the true average tax rate is not disclosed.  $\nu_1$  measures how much stronger that relationship becomes when the true average rate is disclosed. Analogous

to the design in Bottan and Perez-Truglia (2020a), the key coefficient  $\nu_1$  measures the effects of the information shock.

A key challenge in the field experiment is that we do not observe households' prior beliefs about the average tax rate. To overcome this challenge, we take advantage of the fact that individuals systematically update their beliefs upwards or downwards depending on whether their own tax rates are below or above average. The assumption, which we validate in the next section, is that individuals who pay more than average underestimate how much they pay relative to others and, when shown the information, they will update their perceptions upwards. On the other hand, individuals who pay less than average overestimate how much they pay, so when shown the information, they will update their perceptions downwards.

## 5.4 Complementary Survey Experiment

To validate the design of the information-provision experiment (and the econometric model) used in the field experiment, we designed a complementary survey. This complementary survey was included in the same RCT pre-registration as the field experiment and was conducted through Amazon Mechanical Turk around the same dates as the field experiment, from June 5, 2020, to June 15, 2020. We followed several best practices for recruiting individuals in MTurk.<sup>54</sup> The full survey instrument is attached as Appendix G and is summarized here. This survey experiment starts by eliciting the (prior) beliefs about the average property tax rate in the subject's county. Then, in the information-provision stage, a random half of the subjects receive accurate information about the average tax rate in their county. Next, we re-elicite the (posterior) beliefs. With this information, we measure how prior beliefs are distributed and how individuals update those beliefs in light of the information provided to them.

We collected responses from 2,065 U.S. homeowners.<sup>55</sup> Figure 5 presents the main results from the MTurk Survey. Figure 5.a shows subjects' initial misperceptions regarding average property taxes. The x-axis corresponds to the household's actual relative taxes in 2019, that is, the difference between the average tax rate in the county and the household's own tax rate. The y-axis shows the individual's prior beliefs about the relative taxes in 2019. A slope of 1 corresponds with accurate perceptions. Instead, the coefficient (0.237) falls significantly short of 1, indicating significant misperceptions. More precisely, Figure 5.a shows that the misperceptions are systematically skewed toward the middle: individuals who

<sup>54</sup> For more details about the design and implementation of this survey, see Appendix A.5.3.

<sup>55</sup> Appendix A.1.3 provides descriptive statistics for this sample. We show that, in terms of observable characteristics, the MTurk sample is not identical to the other samples used in this paper but not wildly different either.

pay more than average underestimate how much they pay relative to others, and individuals who pay less than average overestimate how much they pay. Indeed, this type of middle-bias has been documented in a variety of settings (Cruces et al., 2013). As a result, we expect individuals toward the left side of the x-axis in Figure 5.a to update their beliefs downwards and individuals toward the right side of the x-axis to update their beliefs upwards.

Figure 5.b corresponds to the belief updating. This figure illustrates the intuition behind the identification strategy. We anticipate whether the disclosure of accurate information translated into upwards or downwards revisions in beliefs, based on whether a household is toward the left or right of the x-axis. The x-axis in Figure 5.b is the same as in 5.a, but the y-axis in 5.b corresponds to the subjects' posterior beliefs (after the information-provision experiment) instead of their prior beliefs (as in 5.a). The blue dots correspond to subjects in the control group (those not shown the feedback about the true average tax rate). For this group, the relationship between perceived and real rates continues to be weak. In contrast, the red diamonds correspond to the treatment group (i.e., subjects who were shown the feedback). These red diamonds show that, as expected, the correlation between perceptions and truth becomes markedly stronger when individuals are provided with accurate feedback. This finding means that individuals who overestimated the average tax rate updated downwards, and individuals who underestimated it updated upwards. For a more formal test, we compare the slope between perceptions and truth in the control group (0.154) versus the corresponding slope in the treatment group (0.613). Consistent with significant learning, the difference between the two is not only large but also highly statistically significant (p-value<0.001).

Figure 5 details the effects of the information shocks. It also is convenient to summarize the results in a single parameter, which we do in the econometric model in Section 5.3. Table 2 presents the results from this regression specification. All columns in this table are based on the same regression specification but use different samples and dependent variables. In column (1), the data are from the MTurk Survey respondents, and the dependent variable is the posterior belief about the average tax rate in the county. *Information Shock* corresponds to the information shock in that regression specification (i.e., the term  $D_i \cdot (\bar{\tau} - \tau_i)$ ). The coefficient on *Information Shock* from column (1) indicates that a 1 pp increase in the information shock increases the posterior belief by 0.393 percentage points. This rate of pass-through is significantly above 0 and statistically significant (p-value<0.001). To the extent that this coefficient falls short of 1, it suggests that individuals did not fully incorporate the feedback given to them, which is standard in these types of experiments and consistent with

standard learning models.<sup>56</sup> Indeed, this coefficient of 0.393 is similar to pass-through rates found in other survey experiments. For example, Bottan and Perez-Truglia (2020a) uses a similar research design but in the context of home price expectations and finds a pass-through rate of information shocks of 0.205.<sup>57</sup>

## 5.5 Results from the Field Experiment

Turning to the results from the field experiment, Table 2 presents the estimation results. In column (2), the dependent variable takes the value 100 if the household protested directly in 2020 and 0 otherwise. This analysis is based on the sample of 50,394 subjects from the field experiment who were randomly selected to receive a letter. The coefficient on *Information Shock* indicates that a household that finds out that the average tax rate ( $\bar{\tau}$ ) is 0.1 pp higher than its own tax rate has a decreased probability of protesting in 2020 by 0.095 pp, which is statistically significant (p-value=0.066).

We conduct a falsification test in an event-study fashion. Figure 2.c presents the results. The rightmost coefficient shows the effect on the probability of protesting in 2020, which is identical to the coefficient on *Information Shock* reported in column (2) of Table 2. The rest of the coefficients correspond to the same regression specification, but the dependent variables are protest indicators for the years 2015 through 2019, instead of 2020. Our letters had not been sent yet, so the information shocks should have no effect on protests in prior years. As expected, the coefficients for the other dependent variables are close to 0, statistically insignificant, and precisely estimated.

One challenge with interpreting the magnitude of the coefficient on *Information Shock* is that it is an intention-to-treat effect, due to multiple sources of non-compliance. As discussed in Section 4.7, some letters may not have been opened or opened too late. Additionally, recipients who opened letters may not have paid enough attention to the information on average taxes provided in the table. We can partially address these forms of non-compliance by focusing on the subsample of households that responded to the Field Survey. By construction, all households must have read the letter to know the survey link and code needed to fill out the Field Survey. It is reasonable to assume that this subsample of survey respondents cared enough about the topic to pay close attention to the information provided in the letter.

Column (3) of Table 2 reproduces the same analysis as in column (2), except that column (2) shows data for all subjects letter recipients, column (3) is restricted to the subsample of

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<sup>56</sup> For example, in the context of Bayesian learning, individuals may not fully update their beliefs because they feel confident about the accuracy of their prior beliefs or because they do not feel confident about the accuracy of the signal provided to them.

<sup>57</sup> This result corresponds to the coefficient reported in column (1) of Table 2 from Bottan and Perez-Truglia (2020a).

1,888 households that responded to the Field Survey.<sup>58</sup> The survey respondents also are not random. In terms of household characteristics, such as home value, number of bedrooms, or tax rate, the differences between survey respondents and non-respondents are statistically significant but small.<sup>59</sup> However, one substantial difference between the samples in columns (2) and (3) of Table 2 is the share of subjects who protested in 2020 (50.26% in column (3)), which is much higher than the corresponding share among subjects who received a letter (11.29% in column (2)). A natural interpretation is that the subjects who paid the most attention to our letter were those who undecided about protesting in 2020.<sup>60</sup>

The coefficient on *Information Shock* is negative (-15.392) for survey respondents (column (3) of Table 2) and statistically significant (p-value=0.006). This coefficient is much larger in magnitude than the corresponding coefficient reported in column (2). This difference is partly mechanical: because the baseline rate is much larger for survey respondents (50.26 in column (3) versus 11.29 in column (2)), it is natural for the effects to be larger too. Moreover, as previously discussed, the stronger effects are probably due in great part to the fact that survey respondents paid close attention to the information included in the letter.

One concern with the analysis from column (3) of Table 2 is that, despite the random assignment, the endogenous nature of survey responses may introduce an endogeneity bias. To address this concern, Figure 2.d presents the event-study analysis for this specification. In this analysis, we estimate the same regression as in column (3) of Table 2 but the dependent variables are indicator variables for whether the respondent protested in each year during 2015–2020. That is, Figure 2.d is identical to Figure 2.c, except that the former restricts the sample to survey respondents. The rationale for this exercise is that observing “effects” on the protests in pre-treatment years would suggest that restricting to the survey respondents introduced a selection bias. Reassuringly, the effects on the pre-treatment outcomes are close to 0, statistically insignificant, and precisely estimated.

Now that we focus on respondents to the Field Survey, so we can estimate the effects on the survey outcomes based on the questions included in that survey. The first question is about the stated intention to protest. The results are presented in column (4) of Table 2, which is identical to column (3) except that the dependent variable takes the value 100 if the household states that it is likely (or very likely) to protest in 2020 and takes the value 0 if the household states it is unlikely (or very unlikely) to protest. Based on the evidence that the information shock affected the actual decision to protest (column (3)), we expect to see

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<sup>58</sup> The implied response rate to the survey, 3.7%, may seem low at first glance, but it is substantially higher than the response rates in comparable studies that sent a survey link through letters (e.g. Perez-Truglia and Troiano, 2018; Bottan and Perez-Truglia, 2020a).

<sup>59</sup> Results reported in Appendix A.5.6.

<sup>60</sup> Additionally, it is possible that recipients who found our letter helpful wanted to reciprocate by responding to our survey.



similar effects on the intention to protest (column (4)). This is exactly what we find: the coefficient on *Information Shock* from column (4) is negative (-13.220), highly statistically significant (p-value=0.008), and close in magnitude to (and statistically indistinguishable from) the corresponding coefficient from column (3) (-13.220 in column (4) vs. -15.392 in column (3)).

In Appendix A.5.5, we present additional robustness checks. For example, we show that the effects of the information shock are similar regardless of whether the tax rate was made salient. Additionally, we show that the effect of the information shock is consequential for the number of protests and for the subsequent market values and tax amounts.

## 5.6 Causal Mechanisms

The evidence presented above suggests that the information about the average tax rate had a significant effect on the decision to protest. Next, we discuss, and provide evidence on, some of the potential mechanisms at play.

Our favorite interpretation is that households changed their perception of the average tax rate and that lowered their feelings of unfairness. To probe this mechanism, we leverage the other question that we included in the Field Survey on the perceived unfairness of property taxes. With this mechanism at play, we expect that an increase in the perceived average tax rate would reduce perceptions of unfairness. The results are presented in column (5) of Table 2, which is identical to column (3) except that the dependent variable measures the perception of unfairness on a scale from 0 (very fair) to 10 (very unfair). As expected, the coefficient on *Information Shock* from column (5) is negative (-0.468) and statistically significant (p-value=0.060). This effect is large in magnitude too, at least when compared to the effect on the protest choice. For example, an information shock of 0.1 pp causes an increase of 2.14% of the standard deviation in unfairness (column (5)), and it causes an increase of 3.08% of the standard deviation in protests (column (3)).

In addition, in the MTurk Survey, we included some similar questions to assess the role of fairness. We asked subjects whether their taxes are unfair relative to other households in the county, whether their taxes are too low or too high, and we elicited the tax rate that the household would consider the most fair (holding constant the tax rates of everyone else). Consistent with the effects reported for the field experiment, we find that the information shocks from the MTurk Survey affected the fairness outcomes as well (results reported in Appendix A.5.4).

Though conditional cooperation is our preferred interpretation of the effects of information shocks, we discuss an alternative mechanism. In theory, it is possible that subjects reacted to the information on the average tax rate because they learned from that information whether

protesting would be successful for them. More precisely, recipients who learn that they are paying a higher tax rates than the average household in the county may infer that the reason for that difference is that those households successfully protested. However, below we provide evidence against this alternative mechanism.

First, if we assume that households process information rationally, we should rule out this mechanism from the beginning. By construction, a successful protest has a negligible effect on the tax rate that the household pays. Specifically, the tax rate is computed by dividing the tax amount by the proposed value of the property. A successful protest reduces the value of both the numerator and the denominator, thus leaving the tax rate approximately unchanged. So, if homeowners learn that the average tax rate is 1 pp higher than their own, it would be irrational for them to infer anything about whether other households were successful at protesting (for a more detailed discussion, see Appendix A.5.2).

It is possible, however, that households make inferences about the protest process from the information on the average tax rate because they process the information irrationally. To address this concern, we included a test that leverages the fact that this alternative mechanism should affect households in an asymmetric fashion. In column (6) of Table 2, we include subjects who would be most affected by this alternative mechanism (i.e., those who have never protested and who learn that their own tax rates exceed the county average). These individuals may (incorrectly) infer that other households pay lower tax rates because they protested successfully. In column (7), we include the remaining subjects (i.e., those who either protested in the past or learned that their tax rates are below average), for whom this alternative mechanism would not apply, or at least not apply as strongly. We find no evidence of the type of asymmetry predicted by the alternative mechanism: the coefficients from columns (6) and (7) are similar in magnitude (-11.05 vs. -14.88) and their difference is statistically insignificant (p-value=0.59) – if anything, the difference in point estimates goes in the opposite direction as the one predicted by the alternative mechanism. In sum, though we cannot rule out that the alternative mechanism plays some role, it is unlikely to fully explain the effects of the information shocks.

## 5.7 Magnitude of the Effects

To better grasp the magnitude of conditional cooperation, we summarize the effects as an elasticity. One of the biggest challenges, however, is the need to account for two forms of non-compliance. The first form of non-compliance is that some recipients may not have read the letter. As previously explained, we address this form of non-compliance by focusing on the results from column (3) of Table 2, which is the sample of recipients we are confident read the letter. However, a second form of non-compliance remains: even if they read the letter,

they may not fully incorporate the feedback into their beliefs. We use the results from the MTurk Survey to correct for this additional form of non-compliance. Because each additional 1 pp in the information shock increased the perceived average tax rate by 0.393 pp (column (1) of Table 2), we use a scale-up factor of 2.54 ( $= \frac{1}{0.393}$ ). Scaling up the coefficient on *Information Shock* from column (3) of Table 2 implies that increasing recipients' perception of the average tax rate paid in the county by 0.1 pp would decrease their protest probability by 3.90 pp ( $= 0.1 \cdot 15.392 \cdot 2.54$ ). The 0.1 pp reduction in the average tax rate corresponds to a 4.7% reduction relative to the baseline tax rate in this sample (2.11%), whereas the 3.90 pp reduction in the protest probability corresponds to 7.7% of the average protest probability reported in column (3) of Table 2. These two estimates suggest that the elasticity of the protest probability, with respect to the perceived average tax rate, is 1.63 ( $= \frac{7.7}{4.7}$ ). Thus, subjects are elastic to their perceptions of taxes paid by everyone else. Moreover, conditional cooperation is just one possible manifestation of fairness concerns, so this provides a lower bound for the importance of fairness considerations.<sup>61</sup>

## 6 Heterogeneity by Partisanship

Social scientists typically study preferences for taxation and redistribution through survey data. The decision to protest taxes provides a unique opportunity to study some aspects of support for taxation via revealed preferences in a high-stakes, naturally occurring context. To illustrate this approach, we study the partisan differences in tax protests. Survey data indicate that Republicans and Democrats *say* that they have different opinions on taxation and redistribution, with Republicans typically expressing preferences for lower taxes and lower income redistribution (Di Tella et al., 2017; Alesina et al., 2020; Stantcheva, 2020). In this section, we assess whether those partisan differences materialize in the decision to protest taxes.

### 6.1 Data on Party Affiliation

To split the analysis by Republican and Democratic households, we use the information on homeowners' full names and addresses to merge, at the individual-level, the taxpayer records with the voter files. In Texas, individuals do not have to report a political party when registering to vote. However, whether they voted in a primary election is a matter of public record. As a result, participation in primaries provides a natural measure of party affiliation (e.g., an individual who participated in Democratic primaries but not Republican primaries

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<sup>61</sup> For example, subjects whose kids do not attend public school may protest because they consider it unfair to pay high taxes (Weinzierl, 2018).

would be classified as a Democrat). Moreover, we obtained the voter file records from a private vendor (Aristotle International) that supplements the data on partisanship from the voter files with data from other sources. For example, it is public record when an individual contributes over \$200 to a Democratic or Republican candidate (Perez-Truglia and Cruces, 2017). The vendor uses that data to infer political affiliation. The proxy for political party provided by the vendor is highly consistent with voting data: at the precinct-level, there is a 0.78 correlation between the proxy for party affiliation and the actual share of votes in the 2012 presidential election.<sup>62</sup>

We classified each of the 423,607 single-family homes in the main sample as more likely to identify as Republican or Democrat.<sup>63</sup> Specifically, we identified 57% of subjects as more likely to be Democrats and the remaining 43% as more likely to be Republican. Indeed, this narrow lead in support of the Democrat party is consistent with recent electoral results from Dallas County. In the 2012 presidential election, Barack Obama received 57% of the votes, whereas Mitt Romney received 42%, and the remaining 1% of votes went to third-party candidates.

## 6.2 Protest Rates by Party Affiliation

We started by looking at differences in the rate of protesting between Democrats and Republicans. Given that in survey data, Republicans are substantially more likely to oppose taxes and income redistribution (Di Tella et al., 2017; Alesina et al., 2020; Stantcheva, 2020), we expect Republicans to be substantially more likely to protest their property taxes than Democrats. Instead, we find that, on average, Republicans were only slightly more likely to protest than Democrats: in the sample of 423,607 households, 9.05% of Republicans protested directly in 2020, compared to 7.92% of Democrats. Moreover, this small difference arises in great part due to differences in wealth. The top half of Figure 6.a shows the protest probabilities for Republicans and Democrats by groups of different market values. The first group corresponds to properties valued under \$100,000, and the last group corresponds to properties valued above \$500,000. The bottom half of the figure shows the percentage of Democratic and Republican homes in each category. Relative to Democrats, Republicans tend to live in more expensive homes. If owners of more expensive homes are more likely to protest, that could mechanically explain the differences in protest rates across partisan lines. Figure 6.a shows that, when comparing homes of roughly similar value, the differences

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<sup>62</sup> Results presented in Appendix A.6.1.

<sup>63</sup> For individuals with missing information, such as those not registered to vote, we use a simple predictive model to impute their party affiliation. For details on this imputation, see Appendix A.6.1.

in protest rates between Democrats and Republicans were even smaller.<sup>64</sup> For example, for homes in the modal category valued between \$100K–\$200K, the protest probabilities are 6.34 pp for Republicans versus 5.82 pp for Democrats. This difference is statistically significant ( $p\text{-value} < 0.001$ ) but small in magnitude. For more expensive properties, this relation flips, with Democrat households a bit more likely to protest than Republicans in comparable homes.

### 6.3 Motives for Protesting

Even if Democrats and Republicans protest with a similar probability, it is possible that they protest for different motives. In Sections 3–5, we analyze the roles of selfish (i.e., private benefits and costs) and fairness motives. Next, we break that analysis down by political party.

We start with the response to private benefits. Figure 7 reproduces the results from Figures 1.a and 1.c on the effects of the homestead cap and compares Democrats (left panels: (a) and (c)) and Republicans (right panels: (b) and (d)). Exceeding the \$10,000 homestead cap threshold causes decreases in the protest rate of 4.25 pp ( $p\text{-value} < 0.001$ ) among Democrats (from Figure 7.c) and 4.70 pp ( $p\text{-value} < 0.001$ ) among Republicans (from Figure 7.d). These effects are close to each other and statistically indistinguishable ( $p\text{-value} = 0.597$ ), suggesting similar responsiveness to the private benefits of protesting among Republicans and Democrats.

The differences remain small if we normalize the effects on protest rates by the corresponding effects on tax amounts (Figures 7.a and 7.b). Exceeding the \$10,000 homestead cap threshold causes decreases in the tax amount by \$220 ( $p\text{-value} < 0.001$ ) among Democrats (from Figure 7.a) and \$204 ( $p\text{-value} < 0.001$ ) among Republicans (from Figure 7.b), and the difference between these two coefficients is small and statistically insignificant ( $p\text{-value} = 0.674$ ). In other words, for Democrats, a \$100 reduction in the tax amount caused by the homestead cap is associated with a 2.03 pp drop in the protest rate, whereas the corresponding effect is 2.31 pp for Republicans. This evidence suggests that both Republicans and Democrats are highly responsive to the private benefits of protesting, but Republicans seem, if anything, to be slightly more responsive.

Next, we look at the partisan differences in responses to private costs and show the results in the last two columns of Table 1. These columns reproduce the results from the baseline specification of column (1) of Table 1 and compare subsamples of Democrats (column (6)) and Republicans (column (7)). As with the response to private benefits, we find the re-

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<sup>64</sup>The results are similar if we include protests through agents (see Appendix A.6.2.)

sponse to private costs to be qualitatively consistent between Democrats and Republicans: the coefficients on the basic and extra aid letters are positive and highly statistically significant in both columns (6) and (7). Quantitatively, the differences are mixed. On the one hand, the coefficient on the basic aid letter is higher for Democrats (1.943, from column (6)) than for Republicans (1.509, from column (7)), but this difference is small and statistically insignificant (p-value=0.391). On the other hand, the coefficient on the extra aid letter is higher for Republicans (3.994, from column (7)) than for Democrats (3.027, from column (7)), and the difference is statistically significant (p-value=0.065). Together, the evidence suggests that both Republicans and Democrats are elastic to the private costs of protesting, but Republicans may be somewhat more elastic.

The evidence on the role of private costs and private benefits so far suggests that both Republicans and Democrats protest due to selfish motives, although Republicans may be somewhat more responsive to selfish motives. Next, we turn to the partisan heterogeneity in conditional cooperation. Though fairness concerns are universal, results from laboratory experiments suggest large differences across individuals in the strength of conditional cooperation.<sup>65</sup> Based on that evidence, it is at least possible that Democrats and Republicans differ in the strength of conditional cooperation. The results are presented in the last two columns of Table 2, which break down the baseline results from column (3) by Democratic households (column (8)) and Republican households (column (9)). The coefficient on *Information Shock* for Democrats (-18.317, from column (8)) is larger in absolute value than the corresponding coefficient for Republicans (-10.922, from column (9)).<sup>66</sup> These point estimates suggest that conditional cooperation is somewhat stronger among Democrats than among Republicans, but the difference is statistically insignificant (p-value=0.515). In sum, the evidence suggests that conditional cooperation exists for both Democrats and Republicans but is perhaps somewhat stronger among Democrats.

## 6.4 Comparison to Survey Data

Next, we compare the partisan differences in the decision to protest to the corresponding differences in survey attitudes. Figure 6.b shows the distribution of a widely used measure of preferences for redistribution: whether the government should reduce the income differences between the rich and the poor on a scale from 1 (should not) to 7 (should).<sup>67</sup> Using responses from the 2006-2018 waves of the General Social Survey, we break down the pref-

<sup>65</sup> For example, some subjects are willing to match one-to-one the contributions made by others, others prefer to match partly, and others do not care about the contributions of others at all (Gächter, 2007).

<sup>66</sup> Appendix A.6.3 presents an event-study falsification test for these results.

<sup>67</sup> For more details about this question, see the notes to Figure 6.

erences for redistribution by political party. The survey responses suggest that Republicans and Democrats differ strongly in their support for redistribution. For instance, 9.5% of Republicans choose the highest redistribution category, compared to 29.2% of Democrats.

Considering this large difference in views about redistribution, why do Democrats and Republicans behave so similarly when it comes to protesting their property taxes? A simple explanation is that survey data do not provide a trustworthy measure of the true preferences around taxation and redistribution. Democrats and Republicans may, for example, be more inclined to “state” different attitudes on these topics (i.e., engage in cheap talk) to align with party preferences.<sup>68</sup> With real money at stake, however, our results suggest that those differences disappear. To the extent that Republicans and Democrats state different views about redistribution but end up behaving similarly, our evidence supports for the aphorism that “everyone’s a Republican on tax day”.

However, not all survey questions are created equal. We provide evidence that survey data may exaggerate the differences between Democrats and Republicans but only when questions use vague terms. Figure 6.c is based on a different question from the General Social Survey. This question elicits support for a specific tax: “Do you consider the amount of federal income tax which you have to pay as too high, about right, or too low?” Figure 6.c shows that the share of Democrats who think federal taxes are too high (60.5%) is a bit lower, but only slightly, than the corresponding share of Republicans (62.6%). In other words, when asked in more concrete terms, the differences between Democrats and Republicans are smaller and more consistent with the revealed-preference data from our study. Indeed, we find consistent results using other survey questions. For instance, Figure 6.d is based on a question that we included in the MTurk Survey that is identical to the question from Figure 6.c but asks about property taxes instead of federal income taxes: “Do you consider the amount of property taxes you pay to be too low, about right, or too high?” The results from Figure 6.d are consistent with the results from Figure 6.c: the share of Democrats responding that property taxes are too high (36.9%) is lower, but not much lower, than the corresponding share of Republicans (42.9%).

In Appendix A.6.4, we show that the results are similar for other survey questions on specific tax policies. Specifically, the results show relatively low levels of political polarization in the responses to a question explaining that higher taxes are required for the provision of high level of government services and to a question on social norms about the fair distribution of a given amount of taxes between two properties of different values. Indeed, these findings relate to evidence from Stantcheva (2020) showing that, in the context of the effects of tax

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<sup>68</sup> Indeed, evidence indicates that specific words on a number of topics, such as estate taxation, may cue partisan responses (Gentzkow and Shapiro, 2010).

policies, the degree of partisan polarization depends on the question used. For example, more polarization occurs for questions about the broad effects of tax policies (e.g., the overall effect on the economy) than those about specific effects of tax policies (e.g., the effect on individual taxpayers).

## 7 Conclusions

Individuals in all U.S. states can file a legal protest to reduce their property taxes. Using experimental and quasi-experimental methods, we provide evidence on the determinants of the decision to protest property taxes. We show that households are highly responsive to their private benefits and private costs from protesting, which suggests that selfish motives are important. We show that fairness concerns also play a significant role: consistent with conditional cooperation, people are more willing to pay a higher tax rate when they perceive that the average household also faces a higher tax rate. We propose that the choice to protest property taxes provides a unique opportunity to study support for various taxation aspects via revealed preferences. In this spirit, we explore whether Republicans and Democrats differ in their tax protest behavior.

To evaluate the generalizability of the results of our field experiment, we follow the Selection, Attrition, Naturalness, and Scalability (SANS) conditions in List (2020). In terms of selection, the subject pool is highly representative of the target population.<sup>69</sup> In this experimental context, attrition is not an issue because the outcomes can be tracked with administrative data. In terms of the setting, we conduct a natural field experiment that is ideal in this dimension (Harrison and List, 2004). The natural, high-stakes setting is particularly important given the focus on social preferences (Levitt and List, 2007; Al-Ubaydli and List, 2013). Due to the type of intervention (a mailed letter experiment), it would be straightforward to scale it to the universe of households. Moreover, there is no obvious reason to expect a “voltage drop” when scaling up (Al-Ubaydli et al., 2017).

Using terminology introduced by List (2020), we view our study as a Wave 1 study, focusing on establishing initial causality and producing first tests of theory. Although our evidence is based on data from a specific U.S. county, to the extent that tax protests work similarly across counties both in and out of Texas, the results should be generalizable to those other settings. Moreover, our field experiment and quasi-experiments can be replicated in other counties. The homestead cap for property taxes exists in many other counties, and mailing interventions such as ours should be feasible in all counties. Last, we hope our research design will be applied to other settings to provide more evidence on the mediating

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<sup>69</sup> For details, see Appendix A.1.3.



factors.

We believe that this same framework can be adapted to study other questions from fields such as political economy, public economics, finance, and behavioral economics. Our setting has several features that we believe make it attractive to researchers. The effects on behavior are measured with objective data from administrative records in a naturally occurring context and are based on high-stakes choices. The experiment can be conducted entirely based on publicly available data without the need for non-disclosure or other agreements. The experiment can be implemented in a few weeks, and the final results may be ready in a couple of months. The mailing experiment is relatively cheap, costing less than \$0.25 per subject. Last, the experiment can be implemented on a massive scale. In Dallas County alone, it can potentially involve hundreds of thousands of subjects. By pooling multiple counties, it could be scalable to millions of subjects. In this spirit, we provide detailed accounts of the implementation and data sources that other researchers can follow. Moreover, we are happy to share data, code, tips, and additional resources.

Finally, the main focus of this paper is to study tax protest motivations via revealed preferences. Some of the findings also may serve as key inputs for the design of the protest process. For instance, because the private costs and benefits are so important to the decision to protest, the current protest system may create undesired inequities. Our evidence supports the view that richer households protest more because they stand to gain more from doing so. Similarly, our experimental evidence suggests that some groups, such as racial and ethnic minorities, may have unequal access to protests if they find it more difficult to navigate the protest process. Indeed, our experimental findings could explain the large differences in access to tax protests documented with observational data (Avenancio-León and Howard, 2019).

More importantly, our experimental findings hint at some low-cost interventions that can alleviate inequities in the property tax system (Thaler and Sunstein, 2009). Households are responsive to aid, so access to protest services may be made more equal by targeting aid to disadvantaged groups (Finkelstein and Notowidigdo, 2019). For example, holding the tax amount constant, Hispanic households are 3.61 pp less likely to protest than White households.<sup>70</sup> In comparison, we find that our letter with extra aid increases the protest rate among Hispanic households by 2.55 pp (p-value<0.001).<sup>71</sup> Thus, low-cost interventions targeted at disadvantaged populations, either promoted internally by county assessors' offices or externally through NGOs, could help make the system to protest property taxes more equitable.

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<sup>70</sup> For details, see Appendix A.1.4.

<sup>71</sup> This result corresponds to the regression from column (1) of Table 1, but estimated with the 20.28% of the sample that was classified as Hispanic.

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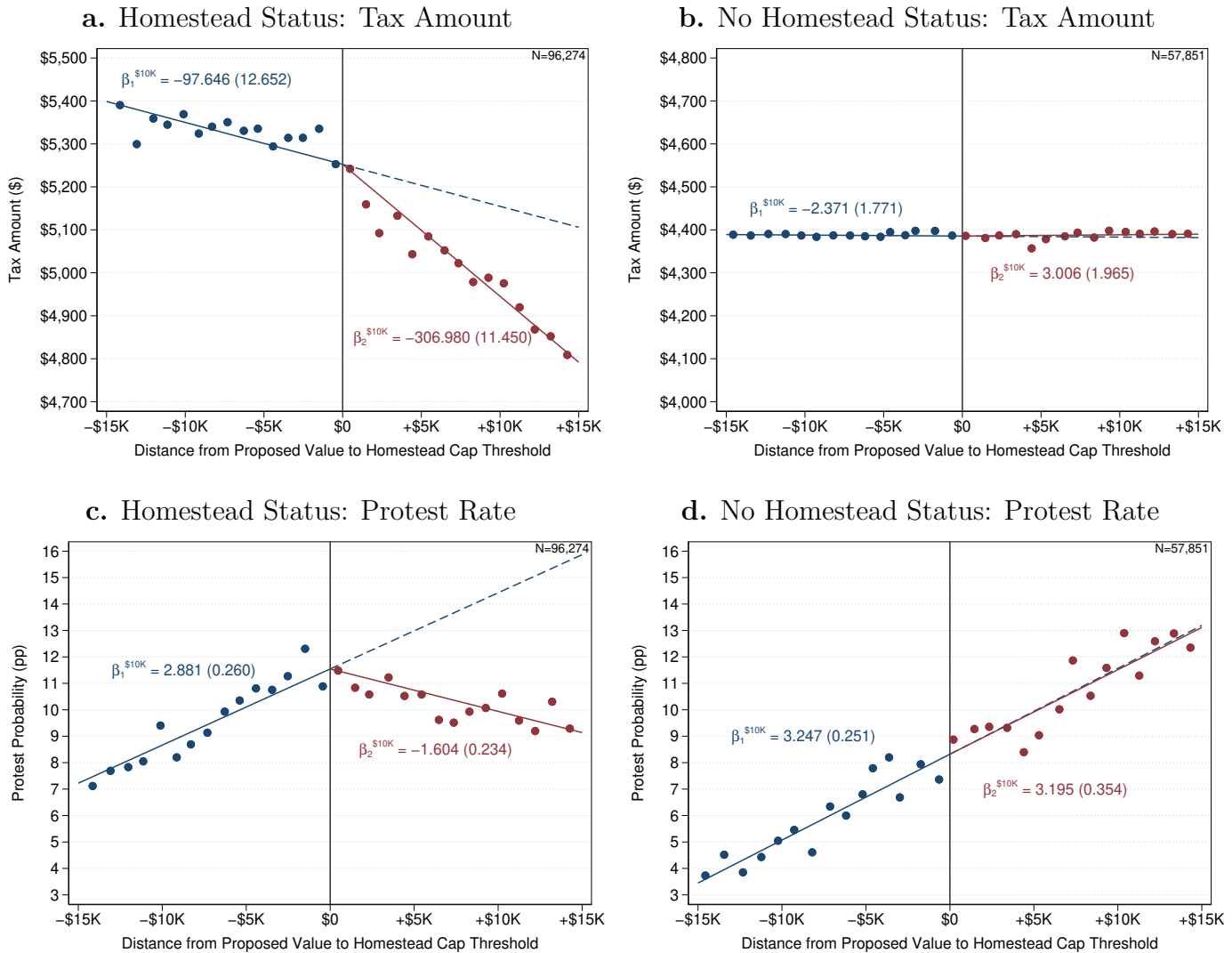
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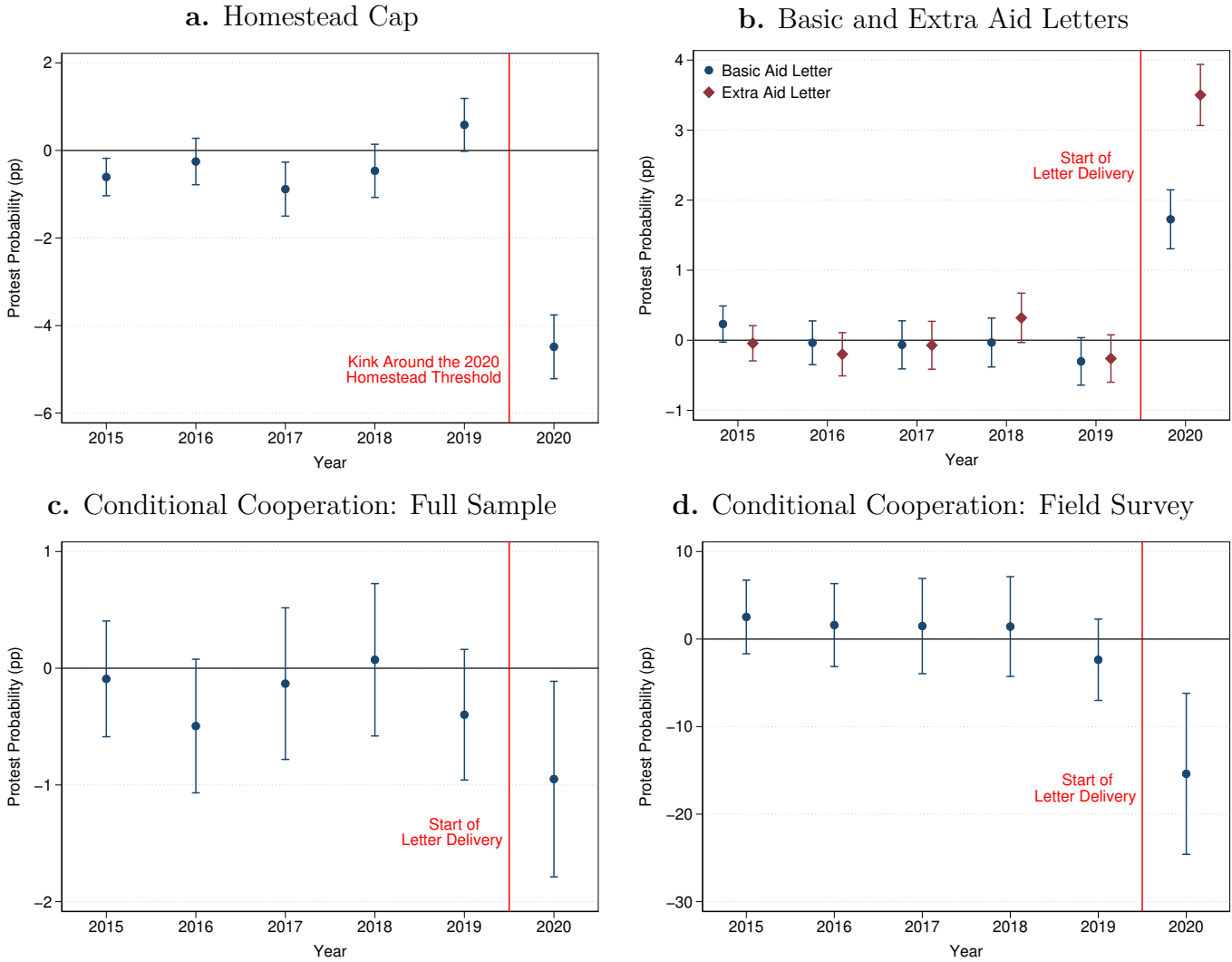
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Figure 1: Effects of the Homestead Cap on the Tax Rate and on the Probability of Protesting




Notes: This figure features binned scatterplots of the relationship between a given outcome (indicated on the y-axis of each panel) and the distance between the 2020 proposed value and the 2020 potential homestead cap threshold (defined as 110% of the appraised value in the previous year). All regressions control for the proposed value, a dummy for whether the household protested in the previous year, and a set of school district dummies. The sample is restricted to properties for which the proposed value is within \$15,000 of the potential homestead threshold. The lines correspond to linear regressions, with normalized slopes reported next to them along with robust standard errors (in parentheses) and the number of households (in brackets in the top right corner). The panels on the left half ((a) and (c)) correspond to households with 2020 homestead status, while the panels on the right half ((b) and (d)) correspond to households without 2020 homestead status. The dependent variables are:  $Tax\ Amount$  is the estimated tax amount based on 2020 proposed values and  $P_{2020}$  is an indicator variable that takes the value 100 if the household protested directly in 2020 and 0 otherwise.

Figure 2: Event-Study Falsification Tests



Notes: Point estimates with 90% confidence intervals in brackets, based on robust standard errors. The point estimates are computed in the same way within each of the four panels: the point estimates within each panel only change the focal year. Panel (a): the blue dots represent the difference between the slopes before and after the threshold as in Figure 1.c, but varying the year. Results based on single-family homes with 2020 homestead status and an absolute difference between the proposed value and the potential homestead cap of less than \$15,000. Panel (b): The blue dots represent the effects of the basic aid letter (relative to the no letter group), while to red diamonds represent the effect of the extra aid letter. Panel (c): The blue dots represent the coefficient on the information shock ( $D_i \cdot (\bar{\tau} - \tau_i)$ ) based on equation (7) from Section 5.3. Panel (d): same as panel (c) except that it is based on the subsample of 1,888 subjects who responded to the Field Survey.

Figure 3: First Page of the Sample Letter



**THE UNIVERSITY OF TEXAS AT DALLAS**  
Naveen Jindal School of Management

May 15th, 2020

Dear Joan Robinson,

We are researchers at The University of Texas at Dallas and we are reaching out to you as part of a research study. **You can lower your tax burden by protesting the taxable value assessment of your property.** We want to share information that we hope will be useful.

Some people may choose to protest because they feel they are paying more than their fair share. Find below some information about the estimated 2020 taxes for your home at 5329 Jordan Ridge D (Dallas, TX) in Dallas County:

	YOUR HOME	AVERAGE DALLAS HOME
<i>Proposed Value</i>	\$174,810	\$294,846
<i>Estimated Tax Amount</i>	\$3,057	\$5,916
<i>Estimated Tax Rate</i>	1.75%	2.01%

Source: Data provided by Dallas Central Appraisal District (CAD). Proposed Value is Dallas CAD's estimate of the home's market value as of January 1st, 2020. Estimated Tax Amount is our estimate of taxes due this year using the latest tax rates available (some exemptions might not be included). Estimated Tax Rate is the estimated tax amount divided by Proposed Value. Average Dallas Home values are based on all single-family homes in Dallas County, excluding condos, townhomes, and mobile homes.

The deadline to protest is June 15th, 2020. You can fill out a short form online or mail it in. You can find instructions on how to do this on the study's website:

<https://www.utdallas.edu/taxproject/>


If you would like to help us with our study, we kindly ask you fill out the following confidential survey. It only takes a couple of minutes, and we would greatly appreciate your participation:

Visit <http://www.utdallas.edu/taxsurvey/> and enter validation code **AAFOGD**

---

800 W. Campbell Road  
Richardson, TX 75080

Website: <https://www.utdallas.edu/taxproject/>

Please  recycle

43137

Notes: A sample of the first page of the letter used in the field experiment. The information in the table varied by treatment group. Sample tables for every treatment group are presented in Figure A.4. The table appears inside a red frame with dashed lines (this frame was added to this figure for emphasis but does not appear in the actual letters).



Figure 4: Second Page of the Sample Letter

a. Extra Aid Message: No

Your household was randomly chosen to receive this letter. *We will not send you any more letters in the future.* If you have any questions about the study, you can find contact information on the study's website.

Thank you for your attention!

Alejandro Zentner  
Associate Professor  
University of Texas at Dallas

43137  
JOAN ROBINSON  
5329 JORDAN RIDGE DR  
DALLAS, TX 75236-1895  
|||

b. Extra Aid Message: Yes

If you'd like to file a protest, it is really simple. You do not need an agent. You do not need to attend a hearing if you accept an online settlement offered by the county. If the county schedules a hearing and you do not attend it, the protest will simply be dismissed with no penalty.

When you protest you need to provide an argument in a few sentences. For example, you may argue that the appraised market value is too high. In that case, you could use the following:

Value is over market value      Opinion of value: \$160,000

And remember to attach a separate page (or file, if protesting online) with your argument:

*I found a home that is similar to mine but was recently sold for less than my home's appraised market value. The property located at 5148 Ronyan Rd (Dallas, TX) is 0.29 miles away from my home, and has the same number of bedrooms and a similar square footage. That property was sold on 10/31/2019 for \$160,000.*

You can find information about this sale by searching for the property's address on Zillow.com or Redfin.com. On these websites you can find other comparable properties to support your protest. You can also protest based on the appraised market values of comparable properties, which can be found on [www.dallascad.org/SearchAddr.aspx](http://www.dallascad.org/SearchAddr.aspx).

Your household was randomly chosen to receive this letter. *We will not send you any more letters in the future.* If you have any questions about the study, you can find contact information on the study's website.

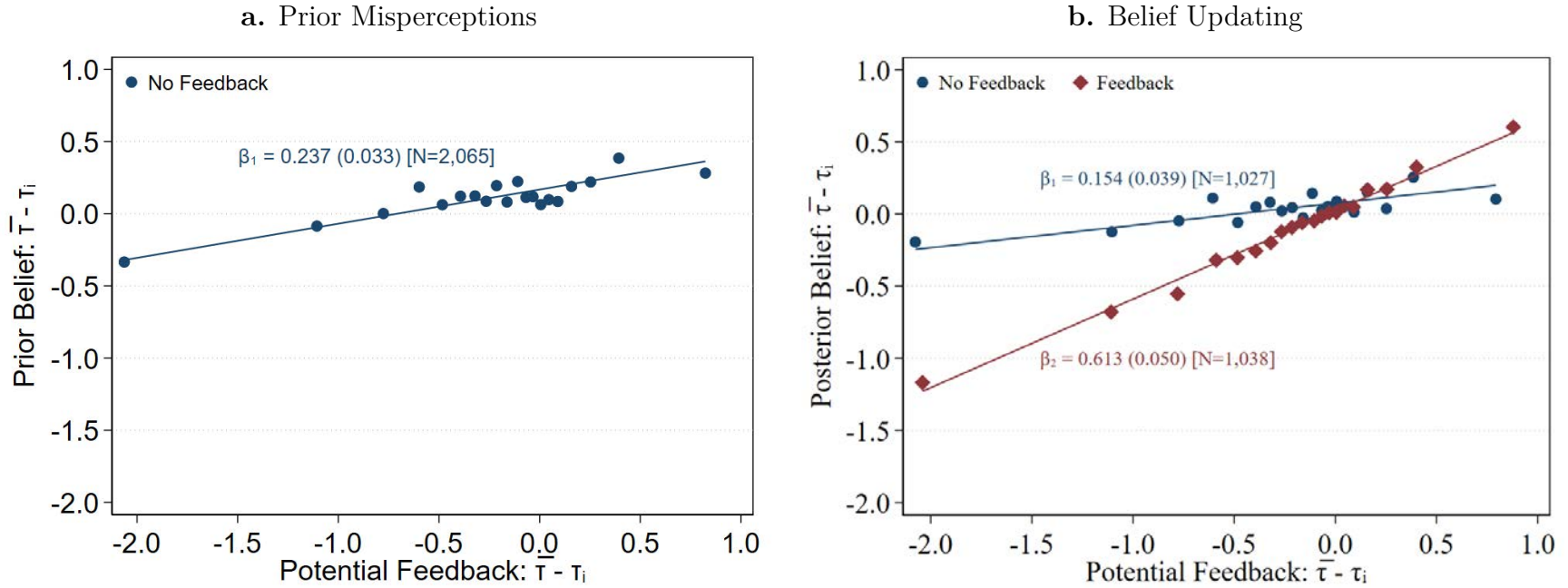
Thank you for your attention!

Alejandro Zentner  
Associate Professor  
University of Texas at Dallas

43137  
JOAN ROBINSON  
5329 JORDAN RIDGE DR  
DALLAS, TX 75236-1895  
|||

Notes: A sample of the second page of the letter used in the field experiment. Panel (a) does not contain the Extra Aid Message, while panel (b) does in the section framed with the red dashed lines. This red frame in panel (b) was added to this figure for emphasis but does not appear in the actual letters.

Figure 5: Prior Misperceptions and Belief Updating in the Mturk Survey



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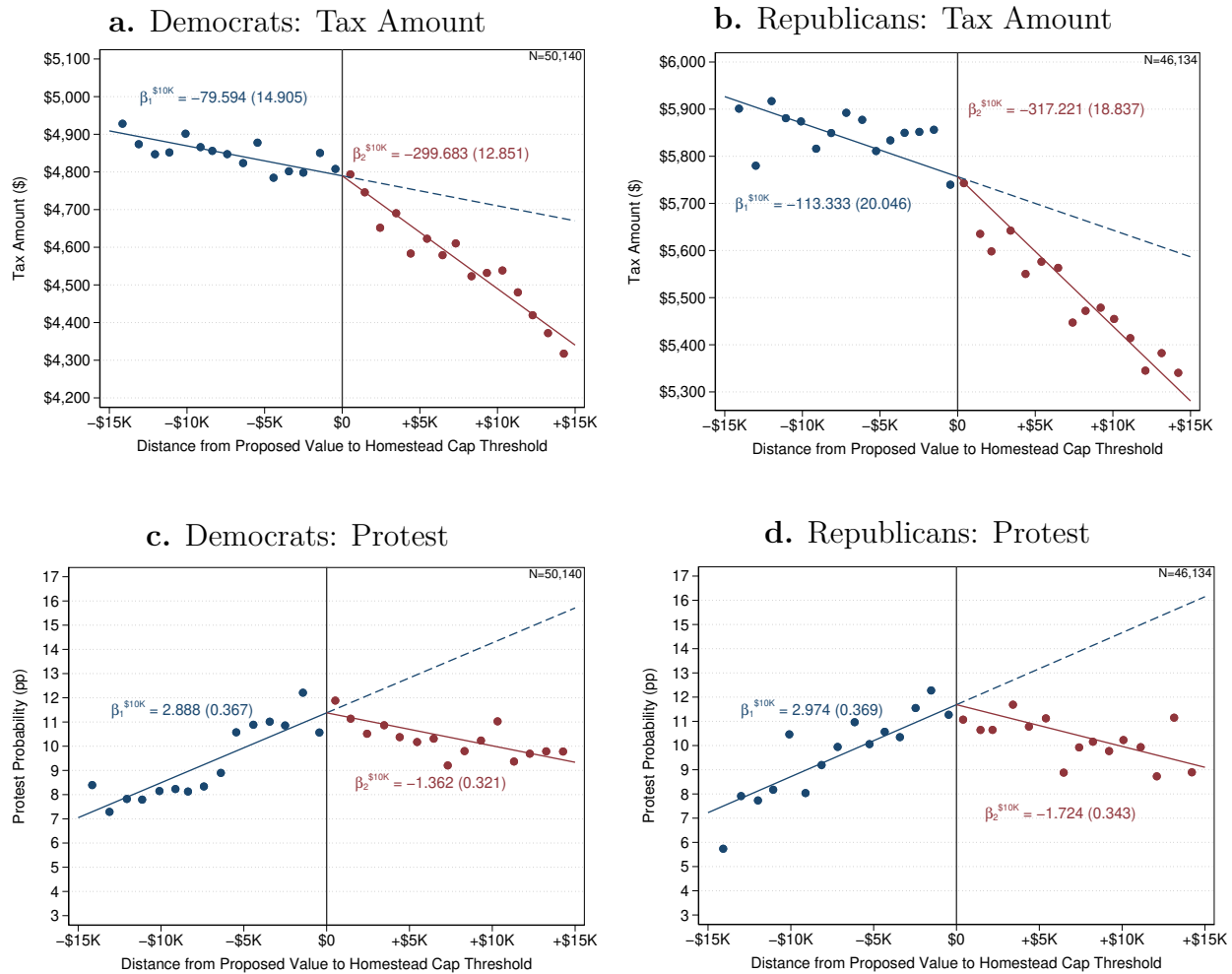
Notes: This figure shows binned scatterplots based on the Mturk Survey. Each line corresponds to a separate OLS regression, with robust standard errors in parentheses and the number of observations in brackets.  $\bar{\tau} - \tau_i$  refers to the difference between the average tax rate in the respondent's county and the tax rate paid by the respondent. In both panels the x-axis corresponds to the potential feedback that could have been shown to the subjects (i.e., the actual difference in tax rates). In panel (a) the y-axis corresponds to the *prior* beliefs about that difference (i.e., the respondent's perceptions before the feedback could have been shown) while in panel (b) the y-axis is the corresponding *posterior* belief (i.e., after the information provision experiment). The results from panel (b) are broken down by treatment group: the red diamonds (labeled "Feedback") correspond to respondents who were shown the feedback while the blue circles (labeled "No Feedback") correspond to those not shown the feedback.

Figure 6: Protest Rates and Attitudes, By Political Party



Notes: Panel (a): The direct protest rates for each proposed value bracket reported on the top of this panel and the share of households in each proposed value bracket reported on the bottom of this panel are based on the main sample of 423,607 single-family homes in 2020. The protest rates on the top of this panel represent point estimates with 95% confidence intervals in brackets, based on robust standard errors. Panel (b): This panel uses responses to the following question, broken down by self reported political party, from the General Social Survey (GSS): “Some people think that the government in Washington ought to reduce the income differences between the rich and the poor, perhaps by raising the taxes of wealthy families or by giving income assistance to the poor. Others think that the government should not concern itself with reducing this income difference between the rich and the poor. Here is a card with a scale from 1 to 7. Think of a score of 1 as meaning (...) What score between 1 and 7 comes closest to the way you feel?” Panel (c): this panel uses responses to the following question, broken down by self reported political party, from the GSS: “Do you consider the amount of federal income tax which you have to pay as too high, about right, or too low?” Panel (d): This panel uses responses to the following question, broken down by self reported political party, using data from the MTurk survey: “Do you consider the amount of property taxes you pay to be too low, about right, or too high?”

Figure 7: Effects of the Homestead Cap on Homestead Households: Heterogeneity by Political Party



Notes: This figure features binned scatterplots of the relationship between a given outcome (indicated on the y-axis of each panel) and the distance between the 2020 proposed value and the 2020 homestead cap threshold (defined as 110% of the appraised value in the previous year). All regressions control for the proposed value, a dummy for whether the household protested in the previous year, and a set of school district dummies. The sample consists of single-family homes in 2020 for which the proposed value is within \$15,000 of the potential homestead threshold. The lines correspond to linear regressions, with normalized slopes reported next to them along with robust standard errors (in parentheses) and the number of households (in brackets in the top right corner). The panels on the left half ((a) and (c)) correspond to households with 2020 homestead status who belong to likely Democrats, while the panels on the right half ((b) and (d)) correspond to households with 2020 homestead status who belong to likely Republicans. The dependent variables are: *Tax Amount* is the estimated tax amount based on 2020 proposed values, and  $P_{2020}$  is an indicator variable that takes the value 100 if the household protested directly in 2020 and 0 otherwise.

Table 1: Effects of the Two Types of Letters on the Probability of Protesting

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	$P_{2020}$	$P_{2020}^{won}$	$P_{2020}$	$P_{2020}$	$SO_{2020}$	$P_{2020}$	$P_{2020}$
Basic Aid Letter <sup>(i)</sup>	1.792*** (0.249)	1.213*** (0.222)	1.449*** (0.347)	1.935*** (0.339)	0.795 (0.719)	1.943*** (0.317)	1.509*** (0.394)
Extra Aid Letter <sup>(ii)</sup>	3.509*** (0.258)	2.621*** (0.231)	3.108*** (0.364)	3.745*** (0.350)	15.287*** (0.979)	3.027*** (0.326)	3.994*** (0.412)
P-value (i)=(ii)	<0.001	<0.001	<0.001	<0.001	<0.001	0.002	<0.001
Subsample			I	II		Dem.	Rep.
Mean Outcome (No Letter)	8.67	6.76	6.03	10.33	3.37	7.49	10.14
Std. Dev. Outcome (No Letter)	28.14	25.10	23.80	30.43	18.05	26.32	30.19
Observations	78,462	78,462	30,356	48,106	5,026	43,208	35,254

Notes: Significant at \*10%, \*\*5%, \*\*\*1%. Robust standard errors in parentheses. Each column presents results from a different regression with two main independent variables: *Basic Aid Letter* is an indicator variable that takes the value 1 if the subject was randomly chosen to receive a basic aid letter and *Extra Aid Letter* is an indicator variable that takes the value 1 if the subject was randomly chosen to receive an extra aid letter. The omitted category is comprised by subjects who were randomly chosen not to receive a letter. The regressions in this table include the following controls: the proposed value in levels and its annual growth, dummies for multiple owners, school and special districts, number of years since the last protest, a dummy for homestead status, and for each year since 2015, a dummy indicating if the household protested in that year and the outcome of the protest (if any) as a %-reduction in the market value. The dependent variables are defined as follows:  $P_{2020}$  is an indicator variable that takes the value 100 if the owner filed a direct protest in 2020 and 0 otherwise;  $P_{2020}^{won}$  indicates with 100 if a direct protest resulted in a reduction in the assessed value,  $SO_{2020}$  is defined for the subsample that protested directly online and provided an opinion of value, and it takes the value 100 if the subject provided an opinion of value within half a percentage point of the value we selected for their extra aid message. Column (3) corresponds to the sample who were not mailed an official notification from the DCAD. Column (4) corresponds to the sample who were mailed such notification. Columns (6) and (7) split the sample between likely Democrat or Republican.

Table 2: Effects of the Information Revealing the Average Tax Rate in the County on the Probability of Protesting

	Mturk		Field Experiment						
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
	$\bar{\tau}_i^{post}$	$P_{2020}$	$P_{2020}$	$I_{2020}$	$U_{2020}$	$P_{2020}$	$P_{2020}$	$P_{2020}$	$P_{2020}$
Information Shock ( $\bar{\tau}$ )	0.393*** (0.071)	-0.950* (0.509)	-15.392*** (5.591)	-13.220*** (4.575)	-0.468* (0.243)	-11.050 (20.313)	-14.878** (6.556)	-18.317** (8.204)	-10.922 (7.837)
Field Survey Subsample			✓	✓	✓	✓ I	✓ II	✓ Dem.	✓ Rep.
Mean Outcome	1.24	11.29	50.26	81.90	7.12	50.30	50.25	50.47	50.10
Std. Dev. Outcome	0.77	31.65	50.01	38.52	2.15	50.04	50.02	50.03	50.02
Observations	2,065	50,394	1,888	1,867	1,888	672	1,216	860	1,028

Notes: Significant at \*10%, \*\*5%, \*\*\*1%. Robust standard errors in parentheses. All columns present results from equation 7 in Section 5.3. The variable *Information Shock* ( $\bar{\tau}$ ) corresponds to the information shock term ( $D_i \cdot (\bar{\tau} - \tau_i)$ ). Column (1) reports results from the subjects in the Mturk survey, Column (2) from the subjects in the field experiment who received a letter, and Columns (3) through (9) from the subjects who received a letter in the field experiment and answered the Field survey. The dependent variables are defined as follows:  $\bar{\tau}_i^{post}$  is the posterior belief on the average tax rate in the county;  $P_{2020}$  is an indicator variable that takes the value 100 if the owner protested directly in 2020 and 0 otherwise;  $I_{2020}$  is an indicator variable that takes the value 100 if the household reported to be either likely or very likely to protest in 2020 and zero otherwise;  $U_{2020}$  corresponds to a question about whether the taxes of the respondent are unfair relative to the taxes of everyone else, in a 1-10 scale. Columns (6) and (7) split the sample used in column (3) in two groups: i) subjects who protested in 2020 but did not protest in the recent past (column (6)) and ii) subjects who protested at least once during 2015 through 2019 or never protested during 2015-2020 (column (7)). Columns (8) and (9) split the sample in column (3) between likely Democrat or Republican.