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Nonprofits in Good Times and Bad Times*

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Abstract

Need fluctuates over the business cycle, yet little is known about nonprofit behavior over the cycle. This paper exploits data from millions of tax returns and provides key descriptive facts about nonprofits in the face of economic fluctuations. Nonprofit revenue, balance sheets, and spending contract during bad times and grow in good times. Nevertheless, nonprofits partially smooth expenditure relative to income. Nonprofits sharply differ from for-profit firms, which exhibit stronger procyclicality and little smoothing. These facts add to the charitable giving literature, documenting outcomes for charities rather than contributors, and the business cycle literature, highlighting distinct nonprofit versus for-profit cyclicality.

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1 Introduction

The needs of vulnerable individuals in the US fluctuate over the business cycle, with measures such as food insecurity, poverty, and homelessness rates increasing during bad times.¹ Commentators hope that charities might expand to provide key services and alleviate societal needs (Lee, 2013). Society at large shares this hope.² This hope also helps to motivate the rich literature on the drivers of charitable giving. Yet, there is little comprehensive evidence on the behavior of nonprofits, i.e., charities, themselves in the face of business cycles.

In this paper, we establish a set of key facts about nonprofits in good times and in bad times. These facts are intended to improve our understanding of and to spur further research on a large and growing sector of our economy. For example, from 2000 to 2013, nonprofit organizations grew from 6% of US businesses to 9% while their revenue grew from 10% of US GDP to 13%.³

We build our analysis on micro data drawn from millions of tax returns of nonprofit organizations in the US over the past three decades—covering the near universe of nonprofits in the US for all but the smallest organizations. While US nonprofits are exempt from taxation, Internal Revenue Service (IRS) guidelines generally require the filing of annual returns to maintain tax-exempt status.⁴ This legally mandated disclosure offers a useful window into financials across the distribution of nonprofit activity.⁵ Crucially, the returns of tax-exempt organizations include information on revenue and expenditure, in addition to a wide range of data on the characteristics and type of each organization. Throughout our analysis, we will emphasize that this tax return database offers advantages for the study of nonprofits, primarily through its measurement of a nonprofit’s full financial position but also in its granular categorizations.

Our facts center around a total of seven questions, each of which is framed to shed light

¹See Lombe et al. (2018) for evidence on food insecurity, Kneebone and Holmes (2016) on poverty, and Sard (2009) on homelessness.

²66% of respondents said yes when asked “Should charities expand their programs and services during economic downturns (e.g., recessions)?” (Google Consumer Survey run by the authors, August 2020, with 500 respondents).

³These figures are from author calculations. The 6% and 9% numbers are the ratio of the number of nonprofit organizations in IRS data to the total number of nonprofit and for-profit organizations from the US Census Statistics of US Businesses. The 10% and 13% revenues shares are the ratio of total nonprofit revenue from IRS tax returns to US nominal GDP from the Bureau of Economic Analysis’ NIPA Accounts. See Section 2 for more on the IRS data.

⁴Learn more about this requirement, including exceptions, at: <https://www.irs.gov/charities-nonprofits/churches-religious-organizations/filing-requirements>. See also Footnote 12.

⁵In political economy applications, Bertrand et al. (2020a) and Bertrand et al. (2020b) have used related tax return data to link nonprofits to corporate contributions and lobbying.

on whether nonprofits weather adverse economic conditions or instead succumb to increased pressures during bad times. We first explore whether nonprofits access more financial resources in bad times by: securing more revenue (Q1), drawing down their assets (Q2), and/or by increasing their liabilities (Q3). In our next step, we investigate whether nonprofits adjust their spending on activities in bad times by: increasing their total expenditure (Q4), reallocating their expenditure towards core programs (Q5), and/or smoothing their expenditure relative to their revenue (Q6). In our last exploration, we ask whether these nonprofit behaviors differ from those of for-profit firms (Q7). Leveraging our comprehensive data on nonprofits, the resulting facts, i.e., answers to these questions, describe cyclicity at nonprofits in the face of nationwide business cycles, local economic fluctuations, and organization-level shocks. Each of our facts are descriptive rather than causal in nature, with our analysis purposefully targeted towards the documentation of observed behavior.

In Q1 we ask whether nonprofits secure higher revenue during bad times. To answer this question—as well as the following four questions—we employ a series of cyclicity regressions, measuring the observed elasticity of nonprofit outcomes to income at the national and local level. Using this framework, we find that the answer to Q1 is no. Revenue for nonprofits is procyclical, declining during bad times and increasing during good times for both the nationwide and local economies. Our most conservative specification uncovers an elasticity of revenue to local income of 0.3. For context, note that prior work importantly documents that contributions given by individuals to charities fall during economic downturns ([List, 2011](#)). Our dataset, by tracking organization-level rather than individual-level outcomes, allows us to speak to whether this reduction in contributions translates into a reduction in total revenue. Our question is non-trivial because—even defining contributions broadly to encompass money from individuals, businesses, and government grants—the average nonprofit in our data receives around 80% of its revenue from other non-contribution sources, including the sales of products (e.g., discounted clothes or household items) and fees associated with services (e.g., job training or medical care). With these distinctions in mind, however, we reproduce the findings in [List \(2011\)](#) by documenting the procyclicality of revenue arising from donations, and we demonstrate that this procyclicality extends to other revenue sources as well.

Declining revenue during economic downturns motivates our next two questions on nonprofit finances. Do nonprofits exploit alternative sources of resources through changes in their balance sheet, in particular through drawdowns of their assets (Q2) or increases in liabilities during bad times (Q3)? The answer to Q2 is yes: nonprofits do in fact draw down

their assets during economic downturns with a cyclical elasticity of 0.2 for assets. The answer to Q3 is no: nonprofit liabilities decline during bad times with a cyclical elasticity of 0.1 for liabilities. Taken together, our findings of procyclicality for both assets and liabilities imply that nonprofit balance sheets shrink during economic downturns, with a shift away from external financing and towards internal financing. These patterns are consistent with the idea that financial constraints may impact nonprofit decision-making.⁶

In Q4 we ask whether nonprofit expenditure increases during economic downturns. We view this question as important for understanding whether nonprofits provide a form of social insurance against economic fluctuations. The separate measurement of nonprofit expenditure in our data—rather than revenue or contributions—is key, since expenditure may be more tightly linked to the underlying activities or actions taken by charities. We also note that Q4 is distinct from Q1. Even though nonprofit revenue falls on average during economic downturns, it need not follow that nonprofit expenditure falls given the presence of other funding sources such as an organization’s assets or borrowing. With these motivations in mind, the answer to Q4 in the data is no. Instead of expansions during economic downturns, we observe procyclicality in nonprofit expenditure with an elasticity of around 0.15. Spending by nonprofits falls during bad times and increases during good times, fluctuations which we also demonstrate are persistent.

After documenting a reduction in nonprofit expenditure during downturns, in Q5 we ask whether nonprofits reallocate their (lower) expenditure during bad times. We are motivated by a debate in the nonprofit sector about the importance of spending on two categories: core programs and services versus administrative or overhead costs. Historically, there has been a push for nonprofits to spend little on overhead costs under the belief that high overhead costs are indicative of waste and not instrumental in achieving the missions of nonprofit organizations. Under this belief, if nonprofit expenditure falls, it would be less harmful, perhaps even helpful, for such reductions to be disproportionately borne by lower administrative expenditure. We do not find such reallocation in the data. The answer to Q5 is no: the share of spending on core programs and services does not shift over the business cycle. We note, however, that our result need not be viewed as a “failure” of the nonprofit sector. Rather, business leaders and academics have reasonably argued that, as detailed in

⁶We revisit this possibility in our discussion of charity size in Section 4.6. But note that theories of firm financial frictions (Ottonello and Winberry, 2020; Crouzet and Mehrotra, 2018) often imply that only firms facing few financial constraints can afford to expand their balance sheets in the face of investment opportunities. In the nonprofit case, increased need during downturns arguably provides an opportunity for expenditure in the same manner.

Gregory and Howard (2009), a focus on decreasing the share of spending on overhead costs can lead to a “nonprofit starvation cycle” in which charities lack the necessary talent or infrastructure to implement their goals.⁷

Our set of five facts so far may give the impression that, when revenue falls during economic downturns, nonprofits do little to cushion or smooth this blow. Indeed, expenditure fluctuations over the cycle are *prima facie* evidence against the notion of perfect expenditure smoothing at charities. However, we push further in a more granular direction, exploiting disaggregated variation in revenue at the individual organization level rather than aggregate economic fluctuations. We estimate a series of smoothing regressions, asking in Q6 whether nonprofits smooth their expenditure with a measured elasticity of expenditure to revenue less than one. These regressions reveal that nonprofits cut their spending growth when their revenue growth declines, a pattern which proves stable and survives the inclusion of a set of fixed effects that account for nonprofit-specific heterogeneity in trends and regional variation in the economic cycle. But this reduction in spending does still allow for smoothing. The answer to Q6 is yes: there is substantial smoothing with an elasticity of nonprofit expenditure to revenue below one at around 0.2. Interacting nonprofit revenue growth with an indicator for economy-wide recessions, we also find that expenditure moves less with revenue during recessions. These results add nuance to our earlier findings of nonprofit procyclicality. While nonprofits do cut their spending growth when times are bad, these cuts are smoother than the drops in their own revenue flows, and in that sense the nonprofit sector provides some insurance against economic downturns. Motivated by Duquette (2017), which shows that the source of nonprofit revenue is tightly linked to nonprofit spending patterns, we also present additional findings which reveal that expenditure smoothing occurs with respect to multiple different sources of revenue.

In Q7, we ask whether the cyclicity and smoothing behavior of nonprofits differ from for-profit firms. This question is relevant given a movement within the nonprofit sector to become more “professional” and “business-like,” e.g., by adopting formal strategic plans or giving more leadership positions to paid rather than volunteer employees (Hwang and Powell, 2009).⁸⁹ We apply our empirical strategy for nonprofits in a comparable fashion to micro data drawn from the financial statements of US public firms. This produces parallel answers

⁷For a range of other work on overhead costs and charity performance metrics, see Gneezy et al. (2014); Karlan and Wood (2017); Meer (2014); Brown et al. (2016); Yörük (2016); Coffman (2017); Exley (2020).

⁸This question is also of particular interest in light of theoretical work in economics on differing incentive schemes and organizational structures in the nonprofit versus for-profit sectors.

⁹See McConnell et al. (2016); Bloom et al. (2015); Tsai et al. (2015) for evidence on the link between formal management practices and hospital performance in a range of contexts.

to Q1-Q4 and Q6 among public firms, while Q5 is not considered since the core program spending share is not a relevant metric for public firms. Consistent with our first four facts that answer Q1-Q4, for-profit firms are procyclical with reduced revenue, assets, liabilities, and expenditure in economic downturns. However, the magnitudes of these elasticities differ sharply across for-profit firms versus nonprofits. We measure higher procyclicality at for-profit firms; for example, the measured elasticity of for-profit firms' expenditure to their local income fluctuations is around 4 times higher than for nonprofits. Moreover, when considering Q6, we find weaker evidence of smoothing at for-profit firms, with high estimated elasticities of revenue to expenditure within firms and no shifts in such comovement during recessions. In other words, the answer to Q7 is yes: for-profit firms do behave differently from nonprofits. Given the size of the nonprofit sector, this heterogeneity highlights the importance of further research into the nonprofit sector by economists interested in firm behavior.

Nonprofits are heterogeneous. Organizations range in size and focus from small food banks and community health clinics to large scientific, educational, cultural, and medical institutions. Legal forms differ across nonprofits, spanning public charities, private foundations, and other structures. Nonprofits operate in different US regions. In light of these differences, we also revisit our questions within categories of nonprofits. Although we find some interesting quantitative contrasts, the facts we lay out above remain evident for each individual nonprofit grouping. We also show that our findings are not driven by the specifics of our empirical specifications, our chosen measures of the economic cycle, etc. We conclude that our findings represent broad, robust patterns of nonprofit behavior.

Our results complement the rich literature on charitable giving. See [Vesterlund \(2006\)](#), [List \(2011\)](#), [Andreoni and Payne \(2013\)](#), and [Gee and Meer \(2020\)](#) for excellent reviews of that work. Much of this literature focuses on the manner in which micro conditions influence individual giving decisions, e.g., how donations are influenced by social pressure ([Ariely et al., 2009](#); [DellaVigna et al., 2012](#); [Andreoni et al., 2016](#)), by matching donations ([Eckel and Grossman, 2003](#); [Karlan and List, 2007](#); [Meier, 2007](#)), by seed money or lead donors ([List and Lucking-Reiley, 20002](#); [Karlan and List, 2020](#)), by household income ([Randolph, 1995](#); [Auten et al., 2002](#); [List, 2011](#); [Kessler et al., 2019](#); [Meer and Priday, 2020b](#)), and by tax policy ([Duquette, 2016, 2019](#); [Meer and Priday, 2020](#)). A smaller set of studies focus on the relationship between macro conditions and giving, such as papers relating to giving after large, tragic events ([Lilley and Slonim, 2016](#); [Bergdoll et al., 2019](#)) and work relating to redistribution and fairness views at the societal level ([Almås et al., 2020](#)).¹⁰ An even

¹⁰This literature typically finds that giving increases after natural disasters, which are likely times of

smaller but important and emerging body of literature seeks to understand aggregate giving in response to macro conditions.¹¹ This existing body of research compellingly documents the procyclicality of giving in relation to macroeconomic fluctuations (List, 2011; Reich and Wimer, 2012; Meer et al., 2017) and includes evidence that such procyclicality is smoothed during recessions (List and Peysakhovich, 2011). Relative to this prior work on procyclicality, our contributions stem from our measurement of nonprofit-level outcomes and behavior—such as the total value of revenue or spending at a charity—rather than individual behavior such as a person giving to a nonprofit. We also contribute by documenting facts within a comprehensive dataset spanning the near universe of nonprofit organizations. Taking stock, we view the study of nonprofit behavior as an important, underexplored area of work in economics. We hope that the set of macro facts we establish on nonprofit behavior encourages more work that focuses on outcomes for nonprofits themselves. Growing a literature in economics on nonprofit behavior and the outcomes for nonprofits—to complement the existing, rich literature that focuses instead on the contributors to nonprofits—is needed to improve our understanding of this large and diverse sector of the economy.

Our results also complement a literature on the cyclicity and sensitivity of for-profit firms to economic fluctuations. One stream of papers focuses on documenting the relative cyclicity of sales at small versus large firms with respect to macro fluctuations (Gertler and Gilchrist, 1994; Crouzet and Mehrotra, 2018). Another set of studies analyzes firm growth and selection patterns around recessions (Moreira, 2017; Kehrig, 2015; Bloom et al., 2018). A third body of work measures the observed volatility and sensitivity of outcomes at for-profit firms in the face of various disaggregated shocks (Davis et al., 2006; Decker et al., 2014, 2020). A fourth set of research examines the cyclicity or responses to policy of economies at the local level (Nakamura and Steinsson, 2014). Relative to each of these sets of work, the contribution of our paper is to extend the knowledge of cyclicity and sensitivity patterns at for-profit firms to the large, qualitatively distinct, context of nonprofit organizations.

Section 2 describes our data and provides an overview of the nonprofit sector. Section 3 documents our seven facts. Section 4 explores heterogeneity across nonprofit categories and a

increased need (although this is not always the case, see e.g., Eckel et al. (2007)). One could view such results as running counter to other findings which report decreased giving during economic downturns, also likely times of increased need. However, there are, of course, many reasons for such differences. For instance, need related to natural disasters is often salient and very targeted. Finally, for work that dives deeply into the role of nonprofit strategy in response to the financial crisis, see the analysis of 196 nonprofits from 2005 to 2015 in Horvath et al. (2018).

¹¹Even setting aside the impact of macro conditions, work on aggregate giving is limited. As discussed in Gee and Meer (2020), while prior work often focuses on the drivers of a single giving decision, there has been a recent movement to consider more aggregated giving outcomes, e.g., substitution effects across charities.

range of robustness checks. Section 5 concludes. Data Appendix A provides more information on our data and empirical strategy. Robustness Appendix B reports some additional results.

2 Data

Our definition of a nonprofit includes organizations deemed tax-exempt by the IRS. Throughout the paper, we have referred and will refer to these entities interchangeably as “nonprofits,” “organizations,” or “charities.” Generally, the IRS requires nonprofits to file a tax return—Form 990—each year (Internal Revenue Service, 2020).¹² Unlike private businesses, whose tax returns are in general confidential in the US, nonprofit tax returns are a matter of public record. We utilize a database of individual nonprofit tax returns compiled by the National Center for Charitable Statistics (NCCS) covering essentially the universe of nonprofit Form 990 data in the US for all but the smallest organizations.^{13,14} Our *main dataset* includes over 8.5 million organization-years drawn from about nine hundred thousand nonprofits from 1990 to 2013. While this dataset is comprehensive in terms of organization-years, it lacks some detailed line items that we rely on in robustness checks and in one of our facts. For these analyses we instead employ our *supplemental dataset*, which is built on the IRS’ Statistics of Income (SOI) files. The latter dataset is more comprehensive in terms of available line items but covers less than 5% of the observations in our main dataset.

The bulk of our analysis centers on four outcomes measured in our main data at the organization-year level: 1) revenue, including income from both contributions as well as the sale of goods and services, 2) assets, including the value of both financial as well as physical resources held by nonprofits, 3) liabilities, the value of total debt or obligations owed by a charity to outside entities, and 4) expenditure, including spending on all activities by the nonprofit. We also examine subcategories within some of these variables when relevant.

The nonprofit sector accounts for a sizable portion of economic activity, organizations, and assets in the US. In 2013, almost 10% of US business-type entities were nonprofits, and their revenue totaled around 13% of US GDP. Compared in the same year to two commonly

¹²Various variants of Form 990, such as Form 990-EZ, Form 990-N, or Form 990-PF, exist for small nonprofits or nonprofits organized in specific legal forms. Our data includes all such variants. Also, most religious organizations are not required to submit tax returns to the IRS.

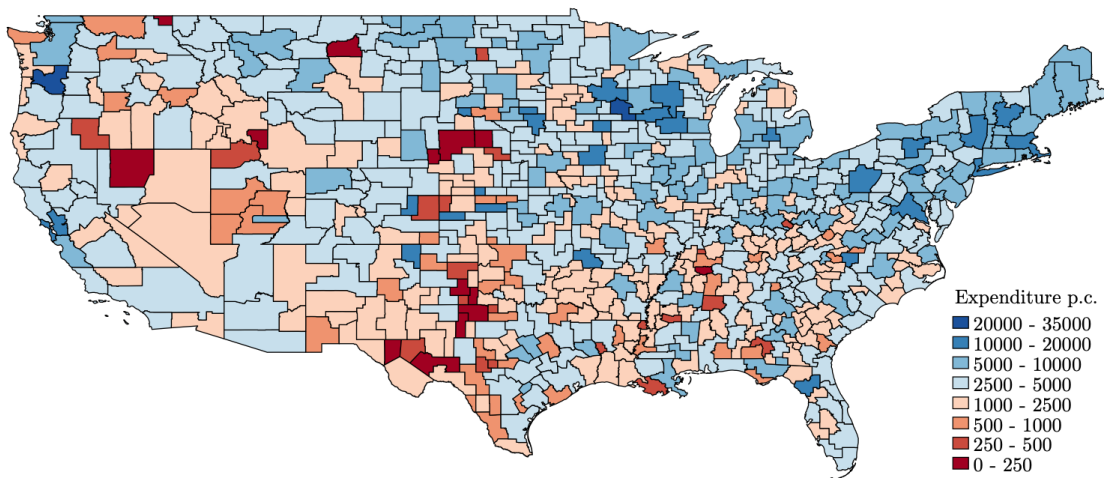
¹³In particular, the data covers nonprofits which are required to file or choose to file a version of IRS Form 990. Generally, this implies that some organizations with fewer than \$25,000 in revenue and some religious organizations are not included in the NCCS data, with coverage of all other nonprofit or tax-exempt entities. See <https://nccs-data.urban.org/> for more information on the data.

¹⁴Note that charities are not required to report in-kind contributions in Form 990. Thus, all statements made in this paper about nonprofit revenue and contributions are strictly limited to monetary values.

studied groupings of for-profit firms—manufacturers and publicly traded companies—the nonprofit sector also appears sizable. Total revenue (assets) of nonprofits were around 17% (13%) of those of all publicly listed firms. And total revenue (assets) of nonprofits were 44% (75%) of those of all manufacturing firms. As note in the introduction, the sector has also grown relative to the rest of the economy since 2000.

The nonprofit sector also spans the whole of the US. Figure 1 maps nonprofit spending per capita in 2010 across US commuting zones (CZ’s). A CZ is a US-government delineated economic area, typically between a county and state in size, which forms a locally unified agglomeration of economic activity. We examine local areas based on this geography to ensure comparability with a range of research on local economic dynamics (Autor et al., 2013). Nonprofit spending varies considerably across regions, with values higher than \$10,000 per person near Boston, Massachusetts compared to just above \$1,000 per person in the Rio Grande Valley in Texas.

Figure 1: Charity Expenditure Per Capita across US Commuting Zones in 2010



Note: This figure shows nonprofit expenditure per capita across US commuting zones in 2010. Nonprofit expenditure includes all accounting expenses. See main text and Appendix A for data construction details.

There is also substantial heterogeneity in the types of nonprofits—spanning human services to the arts to scientific research—and the legal form of nonprofits—including public charities, private foundations, and some other forms. We systematically explore this heterogeneity and demonstrate that our empirical results are robust across all major types and legal forms of nonprofit in Sections 4.4 and 4.5.

See Data Appendix A for more information on our nonprofit sample construction, as well as detailed information about our other sources of data on local and national economic

variables, which include various BEA, BLS, and Census Bureau tabulations. In Appendix A, we also provide more details on our construction of a sample of financial information on US for-profit public firms from the Compustat database. See Table A.1 for descriptive statistics on each of the main variables used in our analysis.

3 Results

In this section, we document a series of facts about the operation of nonprofits in good times versus bad times. Each of the subsections 3.1 - 3.6 presents one empirical fact in response to one of our motivating questions about the operation of nonprofits. For rhetorical purposes, our questions are framed from a perspective which asks whether or not nonprofits secure more resources and spend more on their activities during bad times, although our econometric analysis documents variation over the full cycle in both good and bad times. Our first three questions ask whether nonprofits secure more resources in bad times—through increased revenue (Section 3.1), by drawing down their assets (Section 3.2), or by increasing their liabilities (Section 3.3). Our next three questions ask whether nonprofits adjust their activities during bad times—by increasing their expenditure (Section 3.4), by reallocating their expenditure towards core programs (Section 3.5), or by smoothing their expenditure at the organization level (Section 3.6). Our last question asks whether these patterns of nonprofit behavior differ from comparably measured for-profit firm behavior (Section 3.7). We view the presentation of these descriptive facts as a step towards improving our understanding of how this large sector of the economy operates. We also hope that, as we discuss in Section 5, these descriptive facts will motivate future work, including work that seeks to identify a particular treatment effect or a particular causal mechanism of interest relating to nonprofit operations.

To examine the cyclicity of various nonprofit outcomes in Sections 3.1-3.5, we employ specifications of the form

$$\Delta Y_{j,t} = \alpha + \beta \Delta X_{a,t} + \varepsilon_{j,t}, \quad (1)$$

where $\Delta Y_{j,t}$ is the growth rate of a series of interest Y , e.g., revenue, for nonprofit j in year t , and $\Delta X_{a,t}$ is the growth rate of total personal income in area a surrounding j in the same year t .¹⁵ We examine both national cyclicity and local cyclicity within this framework, replacing $X_{a,t}$ with income at the nationwide or CZ levels as appropriate. We also control for

¹⁵Throughout, we define the growth rate of $Y_{j,t}$ for nonprofit j at time t as $\Delta Y_{jt} \equiv 2 \times \frac{Y_{j,t} - Y_{j,t-1}}{|Y_{j,t}| + |Y_{j,t-1}|}$. This formula is chosen to safeguard against outliers without the need for censoring or winsorization and follows common practice in the firm dynamics literature (Davis et al., 1996). See Appendix A.2 for details.

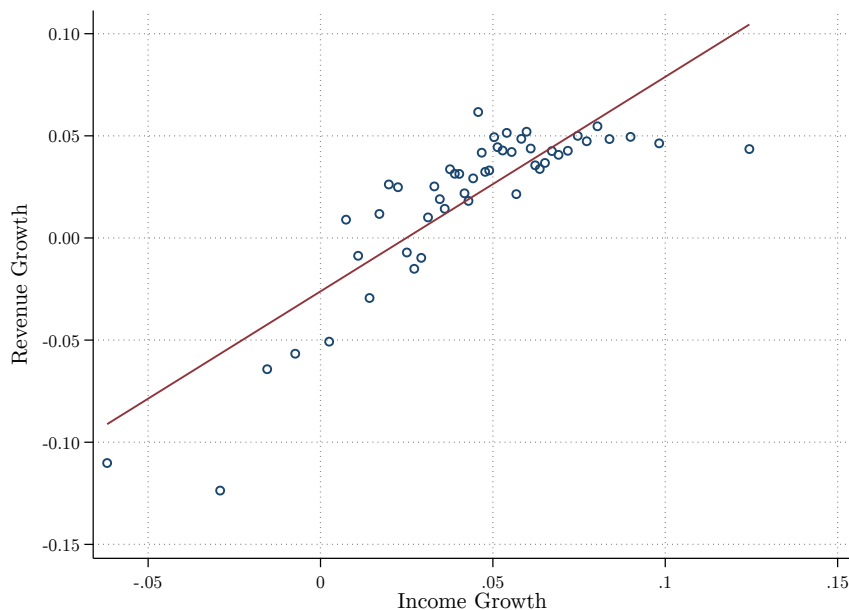
heterogeneity in trends across organizations and space, populating regressions of the form above with a set of fixed effects across increasingly conservative specifications in each section. The sign and magnitude of the observed elasticity β , our coefficient of interest, reveal the comovement or cyclicity of nonprofit outcomes with respect to the broader economy.

Note that we employ standard procedure in the literature on regional shocks, estimating Equation 1 in growth rates (Autor et al., 2013). This conservative specification accounts for permanent unobserved heterogeneity across nonprofits and, once populated with fixed effects, flexibly avoids the conflation of cyclical variation and trends. However, while appropriate, the choice is immaterial for our purposes since we also obtain similar results from less conservative estimation of our regressions in levels (see Section 4.3).

3.1 Do nonprofits secure higher revenue during downturns?

To provide some preliminary visual insight into our first question asking whether nonprofits secure higher revenue during economic downturns, Figure 2 plots quantiles of nonprofit revenue growth against CZ-level personal income growth. Nonprofit revenue is procyclical, growing more during times of high overall income growth and vice-versa during bad times.

Figure 2: Nonprofit Revenue and Economic Fluctuations



Note: This figure shows quantiles of nonprofit revenue growth rates against CZ personal income growth rates with 50 bins. The trend line depicts the best linear fit. Robust growth rates as defined in Appendix A.2 are used. See main text and Appendix A for data construction details.

Table 1 presents cyclicity estimates from versions of Equation 1 with total nonprofit

revenue as the outcome of interest. We see in column (1) that revenue for individual nonprofits grows more during periods with high national income growth. The observed elasticity of slightly greater than 1.6 proves to be economically meaningful. Nonprofit revenue grows by $1.633 \times 2.5 \approx 4.1$ percentage points less in years with one standard deviation lower national personal income growth.

Table 1: Cyclical Regressions for Revenue

	(1)	(2)	(3)	(4)	(5)
	Δ Revenue				
Δ Nat. income	1.633*** (0.056)	1.369*** (0.115)			
Δ CZ income		0.276*** (0.100)	0.314*** (0.047)	0.294*** (0.059)	0.337*** (0.067)
Fixed Effects			Year	Nonprofit Year	Nonprofit Year \times State
Observations	8,623,305	8,623,305	8,623,305	8,623,305	8,623,305

Note: This table reports coefficient estimates for Equation 1 for nonprofit revenue. Δ refers to robust growth rates as defined in Appendix A.2. Nonprofit revenue includes all accounting income. National and CZ income refer to personal income of residents. See main text and Appendix A for data construction details. All standard errors are clustered at the state level.

Standard errors in parentheses. Significance levels: * 10% , ** 5%, *** 1%.

Column (2) adds CZ-level personal income growth to the specification and reveals that nonprofit revenue grows both with the national and local economies, suggesting that nationwide economic fluctuations are only a part of the picture for nonprofits. We further explore local economic conditions in columns (3)-(5), adding a rich set of fixed effects sequentially discarding all common variation in personal income growth in a year, all heterogeneity in trends at the individual nonprofit level, and finally all common shifts in personal income at the state \times year level. Note that examining local, rather than only national, cyclical elasticities is natural given the wide span of our data across the near universe of nonprofits, the fairly restricted time window from 1990-2013 for nationwide fluctuations, and the larger overall magnitude of local versus nationwide fluctuations in income.¹⁶ Column (5), our most conservative and preferred specification, reveals an elasticity of nonprofit revenue to local personal income of around 0.3. Consider two nonprofits in different CZ's experiencing local

¹⁶Table A.1 reports that the standard deviation of CZ-level personal income growth is 3.2% per year, compared to 2.5% per year for national personal income growth.

personal income growth which differs by one standard deviation or around three percentage points. On average, revenue of the nonprofit in the area with lower income growth grows by $0.337 \times 3.2 \approx 1.1$ percentage points less, a drop of around 50% relative to the average nonprofit revenue growth we observe in our sample of around 2%. Taken together, these results imply our first fact, our answer to Q1 from the introduction.

Fact 1 (Nonprofit Revenue): Nonprofit revenue is procyclical, falling during bad times and increasing during good times. The elasticity of nonprofit revenue to local personal income is approximately 0.3.

To further investigate the procyclicality of nonprofit revenue (Fact 1), we now examine the cyclicity of different types of nonprofit revenue. Specifically, motivated by prior work that documents the procyclicality of donations (List, 2011; List and Peysakhovich, 2011; Reich and Wimer, 2012; Meer et al., 2017), we investigate whether we can replicate the procyclicality of broad revenue within categories such as donations and non-donation revenue.

In our main dataset, with over eight million observations, we can break nonprofit revenue into two categories: contributions and non-contribution revenue. Contributions, accounting for approximately 20% of total revenue, include donation support from the public and support from government grants.¹⁷ Thus, focusing on contributions likely provides an overestimate of donation revenue. Meanwhile, non-contribution revenue, accounting for approximately 80% of total revenue, includes revenue from programs and services, financial income, profits from special events or sales, and other miscellaneous sources. Examples of non-contribution revenue include the sales of products such as discounted clothes or household items, fees associated with affordable housing, job training, medical clinics, or college tuition, and earnings from activities that may extend beyond a nonprofit’s direct mission such as ticket sales for special events, rental fees for their space, or the sale of paraphernalia. Appendix Table B.1 shows the cyclicity of contributions and non-contribution revenue. Both types of revenue are procyclical. Non-contribution revenue, which does not include donations, exhibits more procyclicality with an elasticity to income of around 0.3, while the cyclical elasticity of contributions revenue is around 0.1.

In our supplemental dataset, which includes less than 5% of the observations in our main data, we observe government grant revenue, allowing us to measure donations as contributions less government grants. We find that 7.8% of total revenue arises from donations

¹⁷Before 2008, contributions were defined as including direct public support, indirect public support, and government contributions. In 2008 and later, contributions were defined as including federated campaigns, membership dues, fundraising events, related organizations, government grants, and other contributions.

while 92.2% arises from other sources.¹⁸ We also confirm our main results with this narrow donation measure. Appendix Table B.2 shows that the cyclical elasticity of donations to income is around 0.3, while the cyclical elasticity of non-donation revenue is 0.2.

We conclude with some comments on the procyclicality of total revenue (Fact 1) and our additional results. First, we highlight that within our data we replicate the procyclicality of donations documented in prior work (List, 2011; List and Peysakhovich, 2011; Reich and Wimer, 2012; Meer et al., 2017). Second, we provide two sets of novel results: the procyclicality of non-donation sources of revenue and the procyclicality of total revenue. Neither of these latter findings is implied by procyclicality in donations, because other sources of revenue—accounting for the vast majority of total revenue—could have exhibited different cyclicalities than donations. However, it turns out to be clear in the data that revenue procyclicality is quite pervasive across sources.

3.2 Do nonprofits draw down their assets during downturns?

An organization’s own accumulated assets can in principle provide a buffer when faced with revenue declines.¹⁹ In our second question, we ask whether nonprofits make use of this internal funding source—by drawing down their own assets—during economic downturns. The answer is yes. We find procyclicality in nonprofit assets. In Figure 3, plotting nonprofit asset growth against local income growth, the pattern is visually apparent with lower asset growth for nonprofits during periods of low local income growth, and vice-versa.

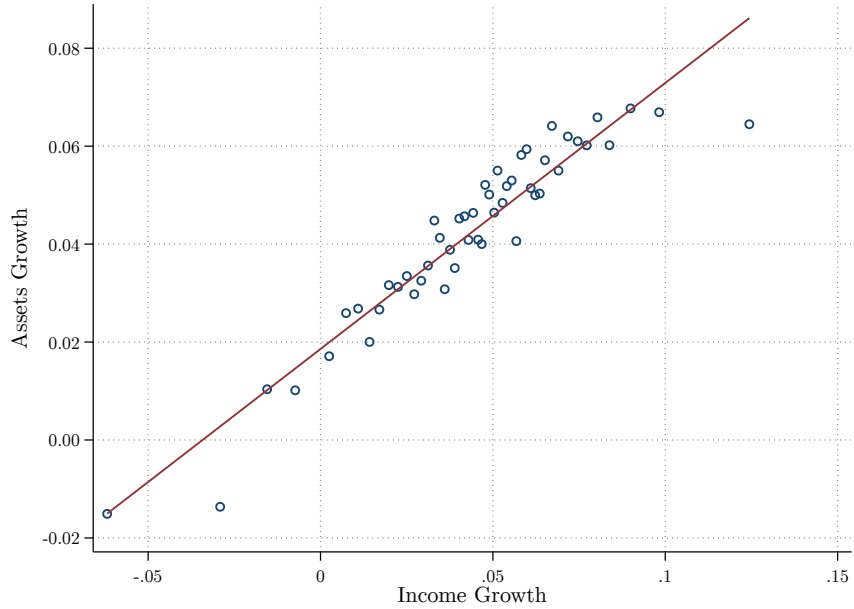
Table 2 reports a related series of cyclical regressions for nonprofit assets. In our most conservative specification in column (5), we estimate the elasticity of nonprofit assets to local income at around 0.2, a moderately large amount of procyclicality. On average, a nonprofit in an area with one standard deviation lower income growth sees their assets grow by $0.175 \times 3.2 \approx 0.6$ percentage points less, a meaningful drop of a bit more than 10% relative to average charity asset growth of around 4.2 percentage points.

Fact 2 (Nonprofit Assets): Nonprofit assets are procyclical, falling during bad times and increasing during good times. The elasticity of nonprofit assets to local personal income is approximately 0.2.

¹⁸We find that contributions account for 13.4% of total revenue, and we arrive at 7.8% after subtracting out government grants that account for 5.6%. Note that contributions account for 13.4% of total revenue in our supplemental dataset compared to 20% in our main dataset, a moderate difference across samples.

¹⁹In 2010 in our main dataset, total nonprofit assets (\$4.2 trillion) were about twice the size of total nonprofit revenue (\$1.9 trillion).

Figure 3: Nonprofit Assets and Economic Fluctuations



Note: This figure plots quantiles of nonprofit asset growth rates against CZ personal income growth rates with 50 bins. The trend line depicts the best linear fit. Robust growth rates as defined in Appendix A.2 are used. See main text and Appendix A for data construction details.

Table 2: Cyclical Regressions for Assets

	(1)	(2)	(3)	(4)	(5)
	Δ Assets				
Δ Nat. income	0.731*** (0.019)	0.462*** (0.027)			
Δ CZ income		0.282*** (0.031)	0.290*** (0.018)	0.221*** (0.018)	0.175*** (0.022)
Fixed Effects			Year	Nonprofit Year	Nonprofit Year \times State
Observations	8,623,305	8,623,305	8,623,305	8,623,305	8,623,305

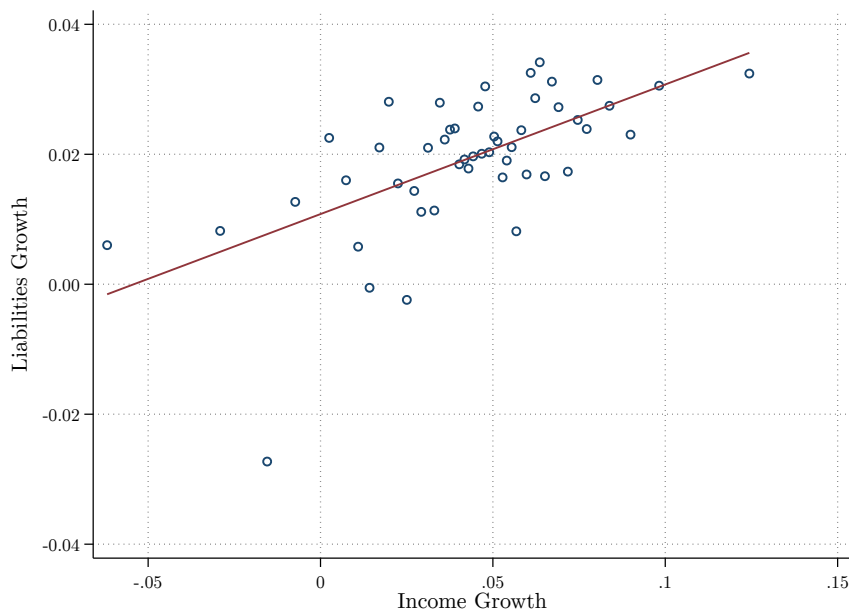
Note: This table reports coefficient estimates for Equation 1 for nonprofit assets. Δ refers to robust growth rates as defined in Appendix A.2. Nonprofit assets include all accounting assets and are measured at the end of the year. CZ income refer to personal income of residents. See main text and Appendix A for data construction details. All standard errors are clustered at the state level.

Standard errors in parentheses. Significance levels: * 10% , ** 5%, *** 1%.

3.3 Do nonprofits increase their liabilities during downturns?

In addition to exploiting their internal resources by drawing down assets during economic downturns, nonprofit organizations might secure more external resources to support spending by increasing their liabilities.²⁰ We ask whether nonprofits rely on this external funding source during downturns in our third question. The answer is no. In Figure 4, we see visually that nonprofit liability growth declines during downturns and increases during booms.

Figure 4: Nonprofit Liabilities and Economic Fluctuations



Note: This figure plots quantiles of nonprofit liability growth rates against CZ personal income growth rates with 50 bins. The trend line depicts the best linear fit. Robust growth rates as defined in Appendix A.2 are used. See main text and Appendix A for data construction details.

Table 3 presents a related series of cyclicity regressions for total liabilities. Column (5) reveals a positive elasticity of around 0.1. On average, a nonprofit in an area with one standard deviation lower income growth will grow their liabilities by $0.101 \times 3.2 \approx 0.3$ percentage points less, a sizable drop of around 15% relative to the average of around 1.9 percentage points.

Fact 3 (Nonprofit Liabilities): Nonprofit liabilities are procyclical, falling during bad times and increasing during good times. The elasticity of nonprofit liabilities to local personal income is approximately 0.1.

²⁰In 2010 in our main dataset, total nonprofit liabilities (\$1.7 trillion) were about the same size as total nonprofit revenue (\$1.9 trillion) and total nonprofit expenditure (\$1.8 trillion).

To summarize, nonprofits do not gather external funds during economic downturns (Fact 3 from this section); rather, they draw down their own financial assets (Fact 2 from Section 3.2). So the size of nonprofit balance sheets declines during downturns. These patterns are broadly consistent with financial constraints, such as a lack of access to external financing, playing a role in determining nonprofit behavior during downturns.

Table 3: Cyclical Regression for Liabilities

	(1)	(2)	(3)	(4)	(5)
	Δ Liabilities				
Δ Nat. income	0.232*** (0.015)	0.091 (0.079)			
Δ CZ income		0.148* (0.076)	0.144*** (0.045)	0.114** (0.047)	0.101** (0.039)
Fixed Effects			Year	Nonprofit Year	Nonprofit Year \times State
Observations	8,623,305	8,623,305	8,623,305	8,623,305	8,623,305

Note: This table reports coefficient estimates for Equation 1 for nonprofit liabilities. Δ refers to robust growth rates as defined in Appendix A.2. Nonprofit liabilities include all accounting liabilities and are measured at the end of the year. CZ income refer to personal income of residents. See main text and Appendix A for data construction details. All standard errors are clustered at the state level.

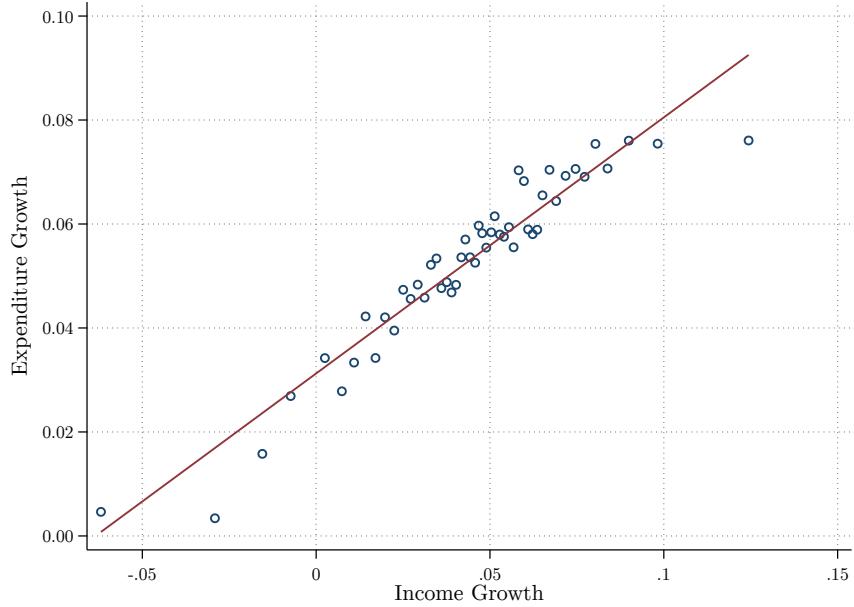
Standard errors in parentheses. Significance levels: * 10% , ** 5%, *** 1%.

3.4 Do nonprofits increase their spending during downturns?

If nonprofits provide a form of social insurance against economic fluctuations, as commentators and policymakers appear to hope, then nonprofit activities during economic downturns should reflect this role. Relative to other observable outcomes at nonprofits, such as nonprofit revenue, we view nonprofit expenditure as an attractive proxy for nonprofit activities. A strength of our dataset is the separate measurement of charity spending and revenue. We see in Figure 5 that the answer to our fourth question is no: spending growth at nonprofits does not increase during local economic downturns. Instead, spending appears procyclical.

Table 4 presents a related series of cyclical regressions for nonprofit expenditure. We see in column (1) that when national income growth falls by one standard deviation or 2 percentage points, expenditure growth for individual nonprofits declines by an average of $0.692 \times 2.5 \approx 1.7$ percentage points. Column (2) reveals that expenditure growth also comoves substantially with the local cycle. Later columns narrow to a view of the local

Figure 5: Nonprofit Expenditure and Economic Fluctuations



Note: This figure plots quantiles of nonprofit expenditure growth rates against CZ personal income growth rates with 50 bins. The trend line depicts the best linear fit. Robust growth rates as defined in Appendix A.2 are used. See main text and Appendix A for data construction details.

economic cycle, with the most conservative and preferred specification in column (5). We estimate the elasticity of nonprofit expenditure to local income to be around 0.15. On average, a nonprofit in an area with one standard deviation lower income growth grows their expenditure by $0.143 \times 3.2 \approx 0.5$ percentage points less, a meaningful drop of around 10% relative to mean growth of 5.2 percentage points.

Fact 4 (Nonprofit Expenditure): Nonprofit expenditure is procyclical, falling during bad times and increasing during good times. The elasticity of nonprofit expenditure to local personal income is approximately 0.15.

Our nonprofit expenditure fact—revealing that nonprofit expenditure grows *less* during bad times—answers a key question for research in this area and suggests potential misalignment between nonprofit activities and fluctuations in need. We note that the extent to which movements in total nonprofit expenditure reflect substantive shifts in the underlying activities of nonprofits may relate to the persistence of these reductions and to the type or composition of nonprofit spending. We investigate the persistence of spending declines in this section, revisiting the composition of nonprofit spending in the next section.

Table 4: Cyclicity Regressions for Expenditure

	(1)	(2)	(3)	(4)	(5)
	Δ Expenditure				
Δ Nat. income	0.692*** (0.021)	0.481*** (0.032)			
Δ CZ income		0.220*** (0.032)	0.215*** (0.020)	0.152*** (0.020)	0.143*** (0.016)
Fixed Effects			Year	Year	Nonprofit Year \times State
Observations	8,623,305	8,623,305	8,623,305	8,623,305	8,623,305

Note: This table reports coefficient estimates for Equation 1 for nonprofit expenditure. Δ refers to robust growth rates as defined in Appendix A.2. Nonprofit expenditure includes all accounting expenses. National and CZ income refer to personal income of residents. See main text and Appendix A for data construction details. All standard errors are clustered at the state level.

Standard errors in parentheses. Significance levels: * 10% , ** 5%, *** 1%.

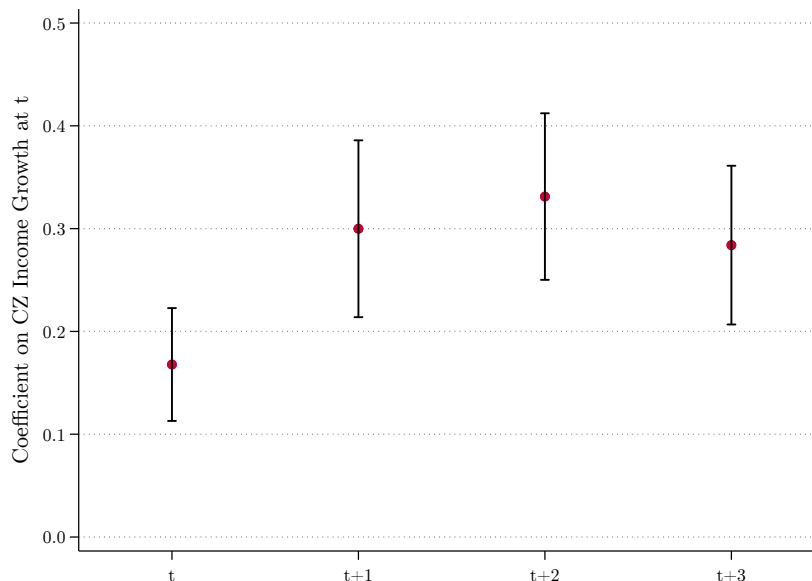
To examine the persistence of expenditure dynamics, we follow [Jordà \(2005\)](#) and estimate the cyclicity of nonprofit expenditure at multiple horizons through a local projections approach. For horizon $h = 0, 1, 2, \dots$, we estimate specifications of the form

$$\Delta_{h+1}E_{j,t+h} = \alpha^h + \beta^h \Delta X_{a,t} + \varepsilon_{j,t}^h, \quad (2)$$

where $\Delta_{h+1}E_{j,t+h}$ is the growth rate of expenditure E within nonprofit j from year $t - 1$ to $t + h$, and $\Delta X_{a,t}$ is the one-period growth rate of total personal income in the surrounding area a in year t . The coefficients β^h trace out the observed elasticity of spending h periods ahead to overall income today. Figure 6 plots the cyclicity path based on CZ-level personal income growth. Cyclical fluctuations in expenditure prove persistent. Bad times in the local CZ today in fact predict slightly larger, although statistically indistinguishable, drops in expenditure three years in the future than in the current year.²¹

²¹The figure represents coefficients estimated on a consistent sample of firms to avoid the conflation of composition and dynamics. Although similar results obtain at longer horizons, consistent estimation over longer periods than around three years results in a less representative sample.

Figure 6: Persistence of Nonprofit Expenditure Cyclicity



Note: This figure shows coefficient estimates for Equation 2 for nonprofit expenditure. The specification includes nonprofit and state \times year fixed effects. Standard errors are clustered at the state level. Robust growth rates as defined in Appendix A.2 are used. Nonprofit expenditure includes all accounting expenses. CZ income refers to personal income of residents. See main text and Appendix A for data construction details.

3.5 Do nonprofits reallocate their expenditure during downturns?

In principle, expenditure reductions during bad times might be cushioned if nonprofits strategically reallocate their expenditure in ways that allow them to maintain key activities during bad times. Relatedly, nonprofits often highlight how much of their expenditure is allocated to 1) core programs and services, versus 2) administrative or overhead costs. Indeed, well-known organizations, such as Charity Navigator, commonly evaluate nonprofits positively based on the size of their program spending and negatively based on the size of their administrative overhead expenses (Charity Navigator, 2020). Such metrics are controversial, since cuts to spending on overhead categories, such as the salaries of skilled workers or facility maintenance, might prove damaging for service provision in practice (Gregory and Howard, 2009). Nevertheless, the core program spending share remains widely discussed and tracked in the nonprofit sector, making it a metric worth understanding better.

While we do not observe nonprofit expenditure on programs and services specifically in our main dataset, we can construct a measure of the share of spending on core programs in our supplemental dataset. We remind the reader that the latter is comprised of a subset of organizations for which the IRS has released more detailed tax return information, implying

that our supplemental data involves less than 5% of the observations included in our main data. Table 5 measures the cyclicity of the core program expenditure share. In column (5), our key specification measuring sensitivity to local personal income fluctuations, we see that the core spending share is acyclical, with no quantitative or statistically significant link to local income. In particular, a one standard deviation larger CZ income growth is imprecisely associated with only a 0.05 percentage point increase in the program expenditure share. The answer to the fifth question is no: nonprofits do not reallocate their spending during downturns towards or away from core programs.

Fact 5 (Core Program Spending Share): The share of nonprofit spending on core programs relative to administrative expenditure is acyclical, reflecting uniform shifts in both categories of spending over the economic cycle.

Table 5: Cyclicity Regressions for Program Expenditure Share

	(1)	(2)	(3)	(4)	(5)
	Δ Program Exp. Share				
Δ Nat. income	-0.228 (0.320)	-0.643 (0.476)			
Δ CZ income		0.429 (0.382)	0.428 (0.317)	0.266 (0.419)	0.186 (0.597)
Fixed Effects			Year	Firm Year	Firm Year \times State
Observations	375,165	375,165	375,165	375,165	375,165

Note: This table reports coefficient estimates for Equation 1 for the share of program expenditure in total expenditure in the supplemental dataset. Δ refers to first differences for the program expenditure share and to robust growth rates as defined in Appendix A.2 for national and CZ income. Program expenditure refers to program service expenses. National and CZ income refer to personal income of residents. See main text and Appendix A for data construction details and summary statistics. All standard errors are clustered at the state level.

Standard Errors in Parentheses. Significance levels: * 10% , ** 5%, *** 1%.

Paired with the overall spending declines reported in Fact 4, the acyclicity result from Table 5 suggests near uniform declines in nonprofit expenditure across multiple categories during local downturns. Indeed, we verify in Appendix Table B.3 that both core program and administrative spending individually decline during bad times.

3.6 Do nonprofits engage in expenditure smoothing?

Our set of empirical facts laid out so far, notably the existence of declining expenditure at nonprofits during downturns, could give the impression that nonprofits do not smooth their expenditure—or, implicitly, their activities—in the face of revenue shifts. However, we note that both the sign and the size of shifts in nonprofit expenditure over the economic cycle matter. While nonprofit expenditure falls during bad times—so some notion of perfect expenditure smoothing is not achieved—nonprofit expenditure varies less strongly over the local economic cycle than nonprofit revenue. Comparing Fact 1 to Fact 4, the cyclical elasticity of nonprofit revenue at around 0.3 is about twice as large as the cyclical elasticity of nonprofit expenditure at around 0.15, suggesting that some meaningful degree of expenditure smoothing by charities does occur.

We also note that not all nonprofits see uniform declines in revenue during local economic downturns. There is considerable variation in revenue across nonprofits within a region at a specific point in time. So we can push further and measure comovement between expenditure and revenue at the individual nonprofit level through a series of “smoothing regressions.” Each specification has the general form

$$\Delta E_{jt} = \alpha + \beta \Delta R_{jt} + \varepsilon_{jt}, \quad (3)$$

linking expenditure growth ΔE_{jt} to revenue growth ΔR_{jt} for nonprofit j in year t . The organization-level elasticity β reveals the extent to which shifts in nonprofit expenditure are smoothed—or not—as their own revenue changes. We contrast this approach explicitly with our cyclical regressions taking the form of Equation 1 above, which were instead targeted towards measuring expenditure shifts in the face of region- or economy-wide, rather than organization-level, fluctuations in income. In fact, the smoothing approach we employ in this section relates more naturally to notions of consumption expenditure smoothing often analyzed for individuals (Blundell et al., 2008).²²

Table 6 reports our results. The elasticity of expenditure to revenue proves to be precisely estimated and quite stable at just above 0.2 across increasingly conservative specifications in columns (1)-(4). In our baseline result, in column (4), we remove both nonprofit-specific trends and all variation at the state \times year level. These magnitudes represent quantitatively meaningful differences. A charity with one-standard deviation lower revenue growth has expenditure growth which is lower by $0.222 \times 60.1 \approx 13$ percentage points on average, a

²²Our smoothing regressions also relate to the analysis of university endowments and financial shocks in Brown et al. (2014).

drop of around two and half times mean expenditure growth in our sample of around 5.2 percentage points. We emphasize that our elasticity estimates are uniformly positive but lie substantially lower than one. In other words, while organizations do typically have lower expenditure growth in the face of lower revenue growth, spending moves less sharply and in this sense exhibits smoothing.

Table 6: Smoothing Regressions

	(1)	(2)	(3)	(4)	(5)
	Δ Expenditure				
Δ Revenue	0.234*** (0.008)	0.235*** (0.008)	0.221*** (0.008)	0.222*** (0.008)	0.231*** (0.008)
Δ Revenue \times Recession					-0.041*** (0.003)
Fixed Effects		Year	Firm Year	Firm Year \times State	Firm Year \times State
Observations	8,623,305	8,623,305	8,623,305	8,623,305	8,623,305

Note: This table reports coefficient estimates for Equations 3 and 4. Δ refers to robust growth rates as defined in Appendix A.2. Nonprofit revenue and expenditure include all accounting income and expenses, respectively. Recession is an indicator for recession years as defined by the NBER. See main text and Appendix A for data construction details. All standard errors are clustered at the state level.

Standard Errors in Parentheses. Significance levels: * 10% , ** 5%, *** 1%.

A natural question is whether the nonprofit-level smoothing of expenditure changes in periods of severe macroeconomic disruptions? In particular, one might expect that in response to increased need during downturns, nonprofit activities and spending might be less tightly linked to an organization’s revenue. Column (5) investigates exactly this possibility, estimating the smoothing specification in Equation 4 and interacting revenue growth with an indicator for whether the year contains a recession as defined by the National Bureau of Economic Research (NBER).

$$\Delta E_{jt} = \alpha + \beta \Delta R_{jt} + \gamma \Delta R_{jt} \times \mathbb{I}(\text{NBER Recession})_t + \varepsilon_{jt}, \quad (4)$$

The estimated interaction $\hat{\gamma}$ is negative, revealing stronger smoothing of expenditure within charities during recessions.²³ In particular, the elasticity of expenditure to revenue

²³Appendix Table B.13 reports related regressions, interacting nonprofit-level revenue with national income growth rather than NBER recessions. The results are similar, revealing stronger smoothing at the micro level during nationwide downturns.

declines by $-0.041 / 0.231 \approx 17$ percent during recessions. Taken together, the evidence of overall expenditure smoothing and the shift towards further smoothing during downturns provide our Fact 6.

Fact 6 (Expenditure Smoothing): Nonprofits exhibit expenditure smoothing, with an elasticity of spending to revenue at the organization level substantially lower than one at 0.2. This elasticity declines further, i.e., nonprofit-level smoothing intensifies, during recessions.

Finally, motivated by prior work that documents a strong correlation between revenue sources and the disposition of that revenue (Duquette, 2017), we note that evidence for expenditure smoothing persists both when we consider contributions and non-contribution revenue using our main dataset (Appendix Table B.10) and when we consider donations and non-donation revenue in our supplemental dataset (Appendix Table B.11). In both of these analyses, we also find that the extent of expenditure smoothing declines during recessions.

3.7 Do nonprofits behave differently from for-profit firms?

Our final question asks whether nonprofit firms behave differently than for-profit firms through the lens of the empirical facts above. We gather data on US publicly listed firms, a group of for-profit businesses accounting for a large share of output and employment. We draw revenue, assets, and liabilities directly from financial statements reported in the standard data source, Compustat, and we compute total expenditure as revenue less operating cashflows. We assign firms to a CZ based on headquarters location. See Appendix A for more details on our sample and variable construction.

Providing the direct for-profit equivalents of nonprofit cyclicity Facts 1 - 4 above, Panels A - D in Table 7 report cyclicity estimates for a range of for-profit outcomes: revenue, assets, liabilities, and expenditure. In Panel A, we see that the elasticity of revenue to local personal income is $0.632 / 0.337 \approx 1.9$ times larger for for-profit firms than nonprofits. Panel B reveals that the elasticity of assets to local personal income is $0.614 / 0.175 \approx 3.5$ times larger for for-profit firms. Panel C shows that the elasticity of liabilities to local personal income is $0.437 / 0.101 \approx 4.3$ times higher in the for-profit sector, although this elasticity isn't precisely estimated in our most conservative specifications. Finally, Panel D shows that the elasticity of expenditure is $0.522 / 0.143 \approx 3.7$ times higher for US public firms than for nonprofits. Uniformly, our results reveal substantially higher sensitivity to economic fluctuations in the for-profit sector than for nonprofits.

Table 7: Cyclical Regressions for For-Profit Sample

	(1)	(2)	(3)	(4)	(5)
A. Revenue			Δ Revenue		
Δ Nat. income	2.111*** (0.097)	1.226*** (0.110)			
Δ CZ income		0.751*** (0.063)	0.738*** (0.073)	0.637*** (0.098)	0.632*** (0.152)
B. Assets			Δ Assets		
Δ Nat. income	1.556*** (0.095)	0.537** (0.250)			
Δ CZ income		0.866*** (0.253)	0.720*** (0.170)	0.640*** (0.215)	0.614** (0.274)
C. Liabilities			Δ Liabilities		
Δ Nat. income	1.852*** (0.057)	1.022*** (0.201)			
Δ CZ income		0.705*** (0.143)	0.593*** (0.112)	0.438** (0.171)	0.437 (0.276)
D. Expenditure			Δ Expenditure		
Δ Nat. income	2.304*** (0.092)	1.462*** (0.076)			
Δ CZ income		0.715*** (0.063)	0.696*** (0.059)	0.598*** (0.070)	0.522*** (0.144)
Fixed Effects				Firm	Firm Year × State
Observations	59,231	59,231	59,231	59,231	59,231

Note: This table reports coefficient estimates for Equation 1 for for-profit revenue, assets, liabilities, and expenditure in the Compustat sample. Δ refers to robust growth rates as defined in Appendix A.2. For-profit revenue includes all accounting revenue, while expenditure is constructed as the difference between revenue and operating cashflow. Assets and liabilities are directly reported by the firm and refer to the respective accounting concept. National and CZ income refer to personal income of residents. See main text and Appendix A for data construction details. All standard errors are clustered at the state level.

Standard errors in parentheses. Significance levels: * 10% , ** 5%, *** 1%.

Shifting to an analysis of expenditure smoothing—and skipping Fact 5 since core program expenses are not a relevant metric for for-profit firms—Table 8 reports estimates of the type seen in Section 3.6 and Fact 6 above for nonprofits from our sample of for-profit firms.

Column (5) reports an elasticity of expenditure to revenue within firms during normal times of around 0.6, about three times high as our equivalent estimates of around 0.2 for nonprofits in Table 6. Furthermore, there is no evidence that smoothing of expenditure increases during recessions, as the interaction term reveals.

Fact 7 (For-Profit Firms): Relative to nonprofits, for-profit firms exhibit higher procyclicality together with weaker expenditure smoothing that doesn't increase during recessions.

These results reveal that for-profit firms exhibit qualitatively different behavior than nonprofits in the face of fluctuations, with higher volatility and little smoothing.

Table 8: Smoothing Regressions for For-Profit Sample

	(1)	(2)	(3)	(4)	(5)
	Δ Expenditure				
Δ Revenue	0.625*** (0.040)	0.613*** (0.040)	0.597*** (0.042)	0.595*** (0.043)	0.593*** (0.044)
Δ Revenue \times Recession					0.011 (0.021)
Fixed Effects		Year	Firm Year	Firm Year \times State	Firm Year \times State
Observations	59,231	59,231	59,231	59,231	59,231

Note: This table reports coefficient estimates for Equations 3 and 4 for for-profit firms in the Compustat sample. Δ refers to robust growth rates as defined in Appendix A.2. For-profit revenue includes all accounting revenue, while expenditure is constructed as the difference between revenue and operating cashflow. Recession is an indicator for recession years as defined by the NBER. See main text and Appendix A for data construction details and summary statistics. All standard errors are clustered at the state level.

Standard Errors in Parentheses. Significance levels: * 10% , ** 5%, *** 1%.

4 Robustness

In this section we explore multiple crosschecks or extensions to our facts above. Our conclusions remain robust throughout this series of checks.

4.1 Do we observe the same cyclicity with other measures of economic fluctuations?

Throughout our cyclicity analysis in Facts 1 - 4, estimating regressions of the form in Equation 1, we link nonprofit outcomes to the local economic cycle. We use CZ-level personal

income growth as our baseline measure of economic conditions in an area. In this section, we examine if our results are robust to economic fluctuations that focus more narrowly on the labor market rather than income-based measures. Table 9 reports the results.

Table 9: Cyclical Regressions Using Alternative Indicators

	(1)	(2)	(3)	(4)
A. Revenue	Δ Revenue			
Cycl. Measure	0.337*** (0.067)	0.279*** (0.045)	0.432*** (0.073)	-0.290** (0.132)
B. Assets	Δ Assets			
Cycl. Measure	0.175*** (0.022)	0.160*** (0.030)	0.209*** (0.033)	-0.339*** (0.104)
C. Liabilities	Δ Liabilities			
Cycl. Measure	0.101** (0.039)	0.167** (0.064)	0.065** (0.025)	-0.015 (0.085)
D. Expenditure	Δ Expenditure			
Cycl. Measure	0.143*** (0.016)	0.152*** (0.021)	0.142*** (0.035)	-0.301*** (0.087)
Cyclical Measure	Δ Income	Δ Em- ployment	Δ Wages	Unemployment rate
Fixed	Firm	Firm	Firm	Firm
Effects	Year \times State	Year \times State	Year \times State	Year \times State
Observations	8,623,305	8,623,305	8,623,305	8,623,305

Note: This table reports coefficient estimates for Equation 1 for revenue, assets, liabilities, and expenditure using alternative indicators for the economic cycle. Δ refers to robust growth rates as defined in Appendix A.2. Nonprofit revenue and expenditure include all accounting income and expenses, respectively. Nonprofit assets and liabilities are measured at the end of the year and include all accounting assets and liabilities, respectively. CZ income refers to personal income of residents. Geographic assignment of employment and wages is based on the place of work, while unemployment rates are calculated on a residential basis. See main text and Appendix A for data construction details. All standard errors are clustered at the state level.

Standard errors in parentheses. Significance levels: * 10% , ** 5%, *** 1%.

Specifically, Panels A - D report the estimated cyclicalities of nonprofit revenue, assets, liabilities, and expenditure, respectively. Column (1) duplicates our baseline results based on personal income growth. Column (2) exploits employment growth. Column (3) examines per-capita wage growth. And column (4) relies upon the local unemployment rate. Our estimates reveal that regardless of whether downturns in a CZ are measured by declining

personal income, declining employment, lower wages per worker, or higher unemployment, we see nonprofit revenue, assets, liabilities, and expenditure decline during bad times.²⁴

4.2 Does cyclical vary across good versus bad times?

Our baseline specification in Equation 1 assumes a constant linear association between CZ income growth and nonprofit outcomes. An interesting question is whether our results are differentially driven by good times or bad times. To investigate this, Appendix Table B.4 allows for heterogeneous slopes above and below the median CZ income growth rate. Our results indicate that charity revenue and assets are marginally more sensitive to economic conditions in (local) bad times, while we find a slight reduction in sensitivity for liabilities and no discernible difference for expenditure. For revenue, liabilities, and expenditure we cannot reject the null of identical slopes. Overall, we conclude that our results appear similar when considering good times and bad times.

4.3 Why do we use growth rates?

In our baseline empirical analysis throughout the paper, we report specifications in first differences or growth rates. For example, we study the association between growth rates of nonprofit outcomes and growth rates of economic conditions in our cyclical regressions. Our approach is standard in the literature on local economic shocks (Autor et al., 2013). Since nonprofits vary widely and persistently in size, the use of growth rates in this context offers considerable advantages, immediately accounting for permanent unobserved heterogeneity across nonprofits in our outcomes of interest. Furthermore, by populating our most conservative specifications—in growth rates—with nonprofit fixed effects, we also transparently control for unobserved heterogeneity in trends at the nonprofit level.

By contrast, estimating such regressions in levels runs the risk of conflating fluctuations within individual charities, our object of interest, with heterogeneity in long-term nonprofit-level trends. Nevertheless, Appendix Tables B.22 and B.23 report estimates of our regressions for nonprofit cyclical and smoothing, replacing growth rates with log levels. Consistent with our baseline analysis, we continue to estimate that nonprofit revenue, assets, liabilities, and expenditure are procyclical, rising during economic booms and falling during downturns. We also uncover evidence of substantial expenditure smoothing, with elasticities of nonprofit

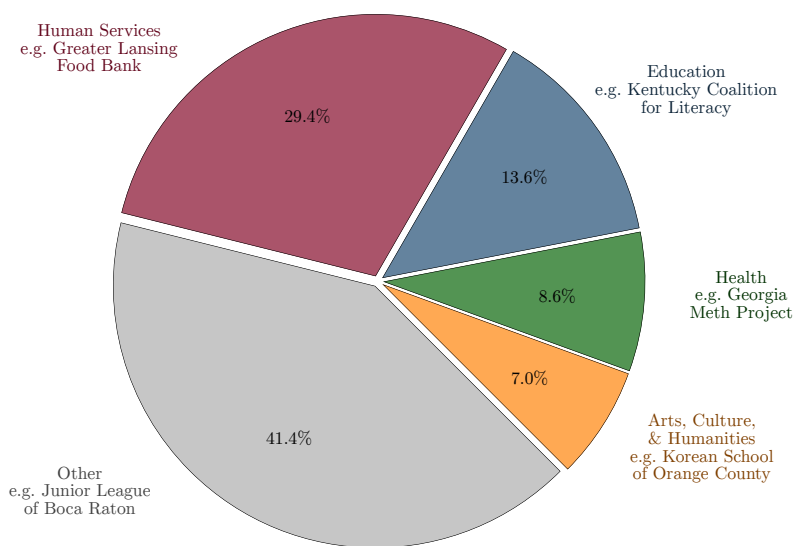
²⁴The single exception to this pattern is liabilities, for which we still estimate procyclical against the local unemployment rate but in a statistically imprecise fashion. Against all other cyclical proxies, we precisely estimate procyclical of liabilities.

spending to revenue far less than one and lower during NBER recessions.

4.4 Do our facts differ according to nonprofit purpose?

Since nonprofits differ in their purpose, NCCS has developed a system of classification codes for nonprofits, the National Taxonomy of Exempt Entities (NTEE). The purpose of NTEE codes is similar to broader statistical classification schemes such as NAICS industry codes. The breadth of NTEE codes reflects substantial heterogeneity in the nonprofit sector, ranging from organizations focused on providing direct charitable services to individuals to very different nonprofits such as large universities, museums, or hospitals. Figure 7 displays the 2010 cross-section of organizations across five NTEE major code groups: 1) Human Services, 2) Education, 3) Health, 4) Arts, Culture, & Humanities, and 5) Other. Figure 7 also includes examples of organizations drawn from each category. The major category entitled “Other,” a somewhat unsatisfactorily named grouping often spanning organizations with multiple purposes, makes up the largest share. Human services—nonprofits focused on the provision of nutrition, housing, and employment assistance, among other activities—accounts for the second-largest proportion, followed by education, health, and then cultural.

Figure 7: Active Nonprofits by Organizational Type in 2010



Note: This figure shows the share of active nonprofits by organizational type in 2010. Organizations are categorized by the NCCS using their NTEE classification scheme. See main text and Appendix A for data construction details.

Given this heterogeneity across types of nonprofits, one natural question is whether a

closer investigation of types of nonprofits would reveal more nuanced facts. If, say, spending at food banks increased during downturns while spending at art museums declined, such nuance could be quite informative.

In Appendix Table B.5, we estimate each of our cyclical regressions for five subsamples. These five subsamples of organizations are the same charity type classifications appearing in Figure 7. We uniformly estimate that revenue, assets, liabilities, and expenditure are procyclical across all nonprofit types. The exact magnitude of cyclical, and the precision of our estimated elasticities, varies somewhat across types, but in no case do we find evidence of expansion during bad times for any major grouping of nonprofits. In a similar exercise, Appendix Table B.12 reports smoothing regressions for each of our nonprofit subsamples based on the charity types in Figure 7. These estimates reveal that the organization-level elasticity of expenditure to revenue is always positive and substantially less than one, with smoothing increasing during NBER recessions for all charity types. Although nonprofits exhibit heterogeneity in type and purpose, we conclude that our empirical facts are qualitatively similar across the main purposes defined by NTEE codes.

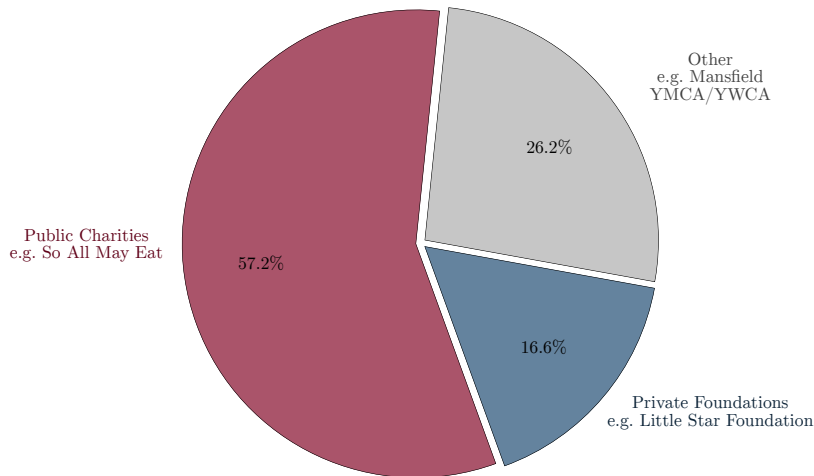
4.5 Do our facts differ by nonprofit legal structure?

Nonprofits also differ in legal structure. In particular, Figure 8 plots the 2010 proportion of nonprofits which are 1) public charities, 2) private foundations, or 3) organized in other legal ways. Public charities, accounting for the dominant share of nonprofits at slightly below 60% of organizations, collect contributions from the general public. Private foundations, a much smaller portion of charities at around 15% of the total, obtain contributions primarily from a single entity, such as a family or business. Other legal forms span various special purpose categories—clubs or organized labor, for example—and account for around a quarter of tax-exempt organizations.

In Appendix Table B.6, we estimate each of our cyclical regressions for subsamples based on nonprofit legal structure. These three subsamples of organizations are the same legal type classifications appearing in Figure 8. We uniformly estimate that revenue, assets, liabilities, and expenditure are procyclical across all nonprofit legal forms. The exact magnitude of cyclical, and the precision of our estimated elasticities, varies somewhat across types, but in no case do we find significant evidence of expansion during bad times for any legal form of nonprofits. In a similar exercise, Appendix Table B.14 reports expenditure smoothing regressions for each of our nonprofit subsamples based on the charity legal forms in Figure 8. These estimates reveal that the organization-level elasticity of expenditure to

revenue is uniformly positive and substantially less than one, with smoothing increasing during NBER recessions for all charity types.²⁵ Although nonprofits exhibit heterogeneity in legal structure, we conclude that our empirical facts are qualitatively similar across these categories.

Figure 8: Active Nonprofits by Legal Form in 2010



Note: This figure shows the share of active nonprofits by legal form in 2010. Nonprofit legal form is established by the NCCS based on nonprofits’ tax returns. See main text and Appendix A for data construction details.

4.6 Do our facts differ by nonprofit size?

Our facts could differ by nonprofit size for multiple reasons. One possibility is that the importance nonprofits place on various objectives—such as their stability, survival, or expansion during bad times—may differ by nonprofit size. Another possibility, not mutually exclusive, is that financial constraints may differ by nonprofit size. Indeed, empirical and theoretical research suggests that financial frictions—such as difficulty in accessing external finance, information asymmetries, etc.—are more prevalent among smaller firms (Midrigan and Xu, 2014; Hsieh and Klenow, 2009; Restuccia and Rogerson, 2008).

Appendix Tables B.7 and B.16 present our cyclical and smoothing results in subsamples split by nonprofit size as defined by their total assets. These results reveal that

²⁵Appendix Table B.15 provides a version of Table B.14, defining economic downturns with national personal income rather than NBER recessions, and finds similar results.

our empirical facts are qualitatively similar across nonprofit size. However, consistent with smaller firms having less access to financing, we see quantitatively stronger smoothing at the largest organizations. Small nonprofits facing a one standard deviation drop in revenue in normal times cut their expenditure by an average of $0.390 \times 60.1 \approx 24$ percentage points, while their larger peers see a far more muted drop of $0.141 \times 60.1 \approx 8$ percentage points on average. These differences across size categories are even more exaggerated during recessions.

4.7 Do our facts differ by region or urbanization level?

Figure 1 highlighted variation in nonprofit expenditure per capita across the US. This variation naturally leads to the question of whether our facts differ across US regions or urbanization levels.

In Appendix Tables B.18 and B.20 we document that our smoothing results are stable across regions and urbanization levels. We also find quantitative heterogeneity in our cyclical results in Appendix Tables B.8 and B.9. In particular, nonprofits in the Northeast and in urban areas are more procyclical, particularly in their revenues and balance sheets.

We speculate that the documented geographic heterogeneity in our cyclical results could be potentially driven by nonprofit size or differences in charity composition across geography. For example, the Northeast, as well as urban areas in general, is home to many nonprofit headquarters as well as a disproportionate share of large educational and health institutions which might have different cyclical properties than smaller, locally focused nonprofits. We leave further exploration of this heterogeneity to future research.

5 Conclusion

Policymakers and the public in the US express hope that civil society, including the nonprofit sector, will expand to meet and alleviate need. Using data from millions of nonprofits' tax returns, we lay out a series of facts about nonprofit behavior in the face of nationwide and local economic fluctuations. We find that—far from increasing their scope in the face of increased need during economic downturns—nonprofits exhibit robust procyclicality, with their revenue, assets, liabilities, and expenditure falling during bad times. However, we do find that charities smooth their expenditure substantially in the face of organization-level revenue fluctuations, smoothing which increases during recessions. Compared to the nonprofit sector, we also highlight that for-profit firms behave qualitatively differently, with far higher cyclicalities and little evidence of expenditure smoothing.

By establishing a series of descriptive facts on outcomes in the nonprofit sector in good times and bad times, this paper seeks to improve understanding of the nonprofit sector and to motivate further work on it. Indeed, in light of our descriptive facts, at least three avenues for further research into the nonprofit sector suggest themselves. As is often the case following the establishment of descriptive facts, these avenues are motivated by a desire to narrow in on a particular mechanism or a particular counterfactual question that requires different analyses than in this paper, e.g., causal identification or structural modeling.

First, future work may delve into which types of charities are better able to smooth their expenditure, or perhaps even increase their expenditure, during bad times. We show that our results are robust across different categories of nonprofits including: organizations in human services, education, health, arts/culture, etc. (Section 4.4), legal structure as public charities, private foundation, etc. (Section 4.5), size (Section 4.6), and location (Section 4.7). With the exception of larger nonprofits engaging in stronger expenditure smoothing, we find few robust differences across these nonprofit categories. However, nonprofit organizations are diverse, and the ways in which one may classify different types of nonprofit organizations are also diverse. So future work might explore more narrow heterogeneity across nonprofits.²⁶ For example, classifying nonprofits according to their fundraising strategies seems a promising avenue.²⁷

Second, many questions remain open around the counterfactual impact of policy on the nonprofit sector. The existing level of government subsidies to nonprofits during economic downturns is not sufficient to prevent their revenue from declining, because our revenue measure includes government grants. But one might reasonably speculate that—and future work could investigate whether—more government subsidies to nonprofits during bad times would promote sustained service provision.

Third, we note that cuts to nonprofit expenditure during bad times could in principle stem from multiple sources including but not limited to manager preferences or financial

²⁶We thank Bob Slonim and Marta Serra-Garcia for this suggestion. When we further examine heterogeneity across nonprofits according to a more detailed classification, which yields 11 groups instead of the 5 considered in Section 4.4, no robust differences across those categories emerge either.

²⁷We thank Dean Karlan and Jonathan Meer for pointing us towards interesting, related analyses that we briefly describe here. First, additional analyses of our data reveal evidence consistent with the idea that nonprofits with revenue that relies more on contributions (donations) engage in stronger expenditure smoothing, bringing to mind the potential advantages and disadvantages to nonprofits for which “success”—vis a vis their revenue—could be less dependent on incentive schemes common to the for-profit world (i.e., generating revenue through non-donation-based avenues). Second, additional analyses suggest that fundraising expenditure may also be procyclical, suggesting that insights may emerge when examining nonprofits according to their fundraising strategies but that these fundraising strategies may be difficult to classify and vary over time.

constraints. As one example, if the leaders of charities were biased on average towards organizational survival rather than maintenance of service provision, or if managers were averse to expansion, then the natural implication would be a failure of charities to expand during times of increased need. In the for-profit sector, evidence exists that such motives might be widespread ([Bertrand and Mullainathan, 2003](#); [Pugsley and Hurst, 2011](#)). As another example, nonprofit managers may hold beliefs which are not accurate about their optimal strategies during bad times. See, for example, inaccurate beliefs about fundraising strategies documented in [Samek and Longfield \(2019\)](#).²⁸ Exploring the extent to which these alternative explanations contribute to the procyclicality of nonprofit expenditure is a natural avenue for future work.

²⁸[DellaVigna and Pope \(2018a\)](#) and [DellaVigna and Pope \(2018b\)](#) also show that academics frequently hold inaccurate beliefs about the impact of leveraging social preferences.

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A Data

A.1 Data Sources

Our primary data source for our *main dataset* is the NCCS Core Trend Files, a longitudinal database of individual nonprofit tax returns compiled by the National Center for Charitable Statistics (NCCS).²⁹ The database harmonizes annual cross-sectional data covering essentially the universe of nonprofit Form 990 data in the US and constructs a set of core financial and non-financial variables that are available throughout the sample period. Charities are identified across years by a permanent enterprise identification number (EIN). We combine the databases for public charities, private foundations, and other exempt organizations to maximize our coverage of the nonprofit sector.

For our analysis we select key financial variables including total revenue, contributions, expenditure, assets, and liabilities, together with information on the charity purpose, legal form, and county of main address. For our subsample analysis, we harmonize charity purpose and legal form across years by assigning each nonprofit its mode of the respective category across all years. We impute missing county information using surrounding years as well as ZIP codes if necessary. Finally, we category firms into size terciles using their average relative assets, where the latter is defined as assets divided by their sample average for a given year.³⁰

For our analysis we drop charities with missing geographic information or missing charity type. Furthermore, due to our focus on growth rates, charities with single observations are not part of our analysis. We restrict our sample to the 1990-2013 period, as the sample size is expanding rapidly before 1990. Furthermore, we drop observation with zeros for all financial variables.

We complement our main analysis with additional insights derived from a *supplemental dataset* with additional information on expenditure composition, which is derived from the Statistics of Income (SOI) sample files. The latter provide a detailed breakdown of income and expenses, however, only for a small, albeit randomized subset of nonprofits. We exclusively rely on the SOI files for public charities and other exempt organizations as they provide a breakdown of expenditure into “program service expenses” and “management and

²⁹See <https://nccs-data.urban.org> for a detailed description of the data as well as download links.

³⁰In particular, we first calculate the relative size of a charity in a particular year dividing its asset holdings by the average asset holdings for all charities. In a second step we take the average across all active years for each charity, giving us a sense of its average relative size. Finally, we categorize charities into size classes using terciles of the average relative size distribution.

general expenses”, which we refer to as program and administrative expenditure respectively. We merge financial data from the SOI with geographic information based on the NCCS files to link nonprofits and geographies. We follow our approach for the NCCS Core Trend Files selecting all available observations for the 1990-2013 period and restrict the sample to nonprofits reporting nonzero program expenditure, administrative expenditure, and revenue in at least one year.

Finally, we investigate the behavior of for-profit firms using the Compustat database, which collects balance sheets and income statements for all firms listed at public exchanges in the US.³¹ We restrict our sample to US firms outside the utility and financial sector in the 1990 to 2013 period. We drop all observations with negative assets, capital, employees, investment, and sales. Revenue in the form of sales is readily available in the data and we construct a measure of expenditure as the difference between sales and operating cash flows.

In addition to organization-level data, we collect a set of economic indicators at the regional and national level to link charity outcomes to national and local economic conditions. We obtain personal income at the county level from the Bureau of Economic Analysis’ Regional Economic Accounts, wages and employment at the county level from the Bureau of Labor Statistics’ Quarterly Census of Employment and Wages, unemployment rates at the county level from the Bureau of Labor Statistics’ Local Area Unemployment Statistics, and a NBER Recession indicator from the Federal Reserve Bank of St. Louis’ FRED online database.

We link each county to a commuting zone based on the mapping provided by [Autor and Dorn \(2013\)](#) and aggregate all regional economic indicators to the CZ level.³²

A.2 Variable Construction

Throughout we transform variables of interest into growth rates to safeguard against permanent differences across organizations and regions and connect to the wider literature on firm dynamics. A common challenge faced by researchers working with firm-level data is the treatment of outliers. This challenge is somewhat magnified for our data as the financial indicators are not strictly limited to positive values. Negative values arise as the reported data already nets out sub-accounts. For example, total revenue is partly composed of fundraising net revenue, which can be negative or positive. Changing data availability across years prevents us from manually disentangling these events to create, in an accounting sense, correct

³¹See e.g. [Davis et al. \(2006\)](#) for a detailed description of the database and a discussion of how it compares to the US for-profit firms overall.

³²See David Dorn’s data page: <https://www.ddorn.net/data.htm>

variables limited to strictly non-negative values.

We address this challenge by using a robust version of the growth rates proposed in [Davis et al. \(1996\)](#). In particular, we define the growth rate of variable $Y_{i,t}$ for nonprofit j at time t as

$$\Delta Y_{j,t} \equiv 2 \times \frac{Y_{j,t} - Y_{j,t-1}}{|Y_{j,t}| + |Y_{j,t-1}|}. \quad (5)$$

Note that the constructed growth rates are mechanically restricted to values between -2 and 2, which safeguards our analysis against undue sensitivity to outliers. We calculate growth rates for regional and national economic indicators using the same formula for consistency. Finally, for our local projection exercise we calculate long-run growth rates as

$$\Delta_{h+1} Y_{j,t+h} \equiv 2 \times \frac{Y_{j,t+h} - Y_{j,t-1}}{|Y_{j,t+h}| + |Y_{j,t-1}|}. \quad (6)$$

A.3 Summary Statistics

Tables [A.1](#), [A.2](#) and [A.3](#) report summary statistics for the NCCS Trend Files sample and the SOI subsample.

Table A.1: Summary Statistics

Variable	Mean	Std. Dev.	Median	IQR
Δ Revenue	0.019	0.601	0.026	0.335
Δ Assets	0.042	0.485	0.019	0.225
Δ Liabilities	0.019	0.727	0.000	0.076
Δ Expenditure	0.052	0.471	0.036	0.265
Δ Contributions	0.026	0.753	0.000	0.315
Δ Non-contribution revenue	0.020	0.834	0.019	0.433
Δ CZ income	0.043	0.032	0.046	0.035
Δ Nat. income	0.044	0.025	0.050	0.024
Δ CZ employment	0.008	0.027	0.012	0.029
Δ CZ wages	0.031	0.022	0.031	0.025
CZ unemp. rate	0.062	0.023	0.057	0.031

Note: This table reports summary statistics for the main sample. Δ refers to robust growth rates as defined in [Appendix A.2](#). IQR refers to the interquartile range. Variables are at the nonprofit level unless otherwise noted. Nonprofit revenue and expenditure include all accounting income and expenses, respectively. Contributions are a subset of revenue. Nonprofit assets and liabilities are measured at the end of the year and include all accounting assets and liabilities, respectively. National and CZ income refer to personal income of residents. Geographic assignment of employment and wages is based on the place of work, while unemployment rates are calculated on a residential basis. See main text and [Appendix A](#) for data construction details.

Table A.2: Summary Statistics for SOI Sample

Variable	Mean	Std. Dev.	Median	IQR
Δ Revenue	0.034	0.418	0.044	0.188
Δ Expenditure	0.060	0.308	0.050	0.137
Δ Program expenditure share (%)	0.115	4.247	0.000	1.105
Δ Program expenditure	0.049	0.321	0.002	0.093
Δ General expenditure	0.042	0.391	0.000	0.103
Δ CZ income	0.047	0.030	0.049	0.031
Δ Nat. income	0.048	0.023	0.054	0.020

Note: This table reports summary statistics for the supplemental dataset. Δ refers to robust growth rates as defined in Appendix A.2. IQR refers to the interquartile range. Variables are at the nonprofit level if not otherwise noted. Nonprofit revenue and expenditure include all accounting income and expenses, respectively. Program and general expenditure refer to program service expenses and management and general expenses, respectively. National and CZ income refer to personal income of residents. See main text and Appendix A for data construction details.

Table A.3: Summary Statistics for Compustat Sample

Variable	Mean	Std. Dev.	Median	IQR
Δ Revenue	0.086	0.328	0.077	0.222
Δ Assets	0.074	0.289	0.054	0.216
Δ Liabilities	0.092	0.368	0.049	0.307
Δ Expenditure	0.080	0.309	0.074	0.236
Δ Nat. income	0.047	0.022	0.051	0.020
Δ CZ income	0.049	0.034	0.052	0.037

Note: This table reports summary statistics for the Compustat sample. Δ refers to robust growth rates as defined in Appendix A.2. IQR refers to the interquartile range. Variables are at the firm level if not otherwise noted. For-profit revenue includes all accounting revenues, while expenditure is constructed as the difference between revenue and operating cashflows. Assets and liabilities are directly reported by the firm and refer to the respective accounting concept. National and CZ income refer to personal income of residents. See main text and Appendix A for data construction details.

Table A.4: Summary Statistics in Levels

Variable	Mean	Std. Dev.	Median	IQR
Revenue	12.113	1.988	11.881	2.397
Assets	12.397	2.312	12.332	3.056
Liabilities	10.680	3.408	10.891	4.299
Expenditure	12.029	1.944	11.789	2.376
CZ Income	17.709	1.654	17.902	2.523
Nat. income	23.019	0.289	23.084	0.441

Note: This table reports summary statistics for the level sample with variables in levels. IQR refers to the interquartile range. Variables are at the nonprofit level if not otherwise noted. Nonprofit revenue and expenditure include all accounting income and expenses, respectively. Nonprofit assets and liabilities are measured at the end of the year and include all accounting assets and liabilities, respectively. All variables are in logs. National and CZ income refers to personal income of residents. See Appendix A for data construction details.

B Robustness Checks

B.1 Additional Cyclicity Results

Table B.1: Cyclicity Regressions for Contributions and Non-Contribution Revenue

	(1)	(2)	(3)	(4)	(5)
A. Contributions	Δ Contributions				
Δ Nat. income	0.303*** (0.031)	0.109** (0.041)			
Δ CZ income		0.204*** (0.023)	0.196*** (0.020)	0.169*** (0.020)	0.121*** (0.030)
B. Non-Contribution Revenue	Δ Non-Contribution Revenue				
Δ Nat. income	2.106*** (0.051)	1.901*** (0.139)			
Δ CZ income		0.214* (0.121)	0.269*** (0.043)	0.252*** (0.055)	0.301*** (0.063)
Fixed Effects			Year	Firm Year	Firm Year × State
Observations	8,623,305	8,623,305	8,623,305	8,623,305	8,623,305

Note: This table reports coefficient estimates for Equation 1 for received contributions and non-contribution revenue. Δ refers to robust growth rates as defined in Appendix A.2. Nonprofit contributions include all contributions including grants, donations and individuals contributions. Nonprofit non-contribution revenue includes all revenue except for contributions, e.g. sales and operating income. National and CZ income refer to personal income of residents. See main text and Appendix A for data construction details. All standard errors are clustered at the state level.

Standard errors in parentheses. Significance levels: * 10% , ** 5%, *** 1%.

Table B.2: Cyclical Regressions for Donations and Non-Donation Revenue

	(1)	(2)	(3)	(4)	(5)
A. Donations			Δ Donations		
Δ Nat. income	0.822*** (0.063)	0.562*** (0.152)			
Δ CZ income		0.268* (0.138)	0.259*** (0.072)	0.288*** (0.089)	0.345*** (0.104)
B. Non-Donation Revenue			Δ Non-Donation Revenue		
Δ Nat. income	0.278*** (0.046)	0.084 (0.138)			
Δ CZ income		0.201 (0.122)	0.226*** (0.054)	0.102 (0.069)	0.232*** (0.063)
Fixed Effects			Year	Firm Year	Firm Year × State
Observations	375,165	375,165	375,165	375,165	375,165

Note: This table reports coefficient estimates for Equation 1 for received donations and non-donation revenue in the supplemental dataset. Δ refers to robust growth rates as defined in Appendix A.2. Nonprofit donations include grants and donations made by private parties. Nonprofit non-donation revenue includes all revenue except for donation, e.g. government grants, and sales and operating income. National and CZ income refer to personal income of residents. See main text and Appendix A for data construction details. All standard errors are clustered at the state level.

Standard errors in parentheses. Significance levels: * 10% , ** 5%, *** 1%.

Table B.3: Cyclical Regressions for Program and Administrative Expenditure

	(1)	(2)	(3)	(4)	(5)
A. Program Expenses			Δ Prog. Exp.		
Δ Nat. income	0.528*** (0.068)	0.373*** (0.104)			
Δ CZ income		0.164* (0.085)	0.159* (0.086)	0.055 (0.084)	0.026 (0.097)
B. Admin. Expenses			Δ Adm. Exp.		
Δ Nat. income	0.225** (0.104)	-0.077 (0.153)			
Δ CZ income		0.320*** (0.094)	0.308*** (0.095)	0.214* (0.119)	0.292 (0.213)
Fixed Effects			Year	Firm Year	Firm Year × State
Observations	375,165	375,165	375,165	375,165	375,165

Note: This table reports coefficient estimates for Equation 1 for program and administrative expenditure in the SOI sample. Δ refers to robust growth rates as defined in Appendix A.2. Program and general expenditure refer to program service expenses and management and general expenses, respectively. National and CZ income refer to personal income of residents. See main text and Appendix A for data construction details and summary statistics. All standard errors are clustered at the state level.

Standard errors in parentheses. Significance levels: * 10% , ** 5%, *** 1%.

We test for nonlinearities by estimating Equation 7. A positive γ coefficient indicates that $\Delta Y_{j,t}$ becomes more sensitive to local income growth during bad times.

$$\Delta Y_{j,t} = \alpha + \beta \Delta X_{a,t} + \gamma \Delta X_{a,t} \times \mathbb{I}\{X_{a,t} \leq \text{Median}(X_{a,t})\} + \varepsilon_{j,t}. \quad (7)$$

Results are reported in Table B.4.

Table B.4: Cyclical Regressions Testing For Non-Linearities

	(1)	(2)	(3)	(4)
	Δ Revenue	Δ Assets	Δ Lia- bilities	Δ Expen- diture
Δ CZ income	0.333*** (0.061)	0.171*** (0.021)	0.103*** (0.038)	0.142*** (0.016)
Δ CZ income \times Below Median	0.047 (0.110)	0.056** (0.022)	-0.016 (0.050)	0.008 (0.020)
Fixed	Firm	Firm	Firm	Firm
Effects	Year \times State	Year \times State	Year \times State	Year \times State
Observations	8,623,305	8,623,305	8,623,305	8,623,305

Note: This table reports coefficient estimates for Equation 7 for revenue, assets, liabilities, and expenditure. Δ refers to robust growth rates as defined in Appendix A.2. Nonprofit revenue and expenditure include all accounting income and expenses, respectively. Nonprofit assets and liabilities are measured at the end of the year and include all accounting assets and liabilities, respectively. CZ income refers to personal income of residents. Median CZ income growth is calculated over the entire sample. See main text and Appendix A for data construction details. All standard errors are clustered at the state level.

Standard errors in parentheses. Significance levels: * 10% , ** 5%, *** 1%.

Table B.5: Cyclical Regressions by Nonprofit Type

	(1)	(2)	(3)	(4)	(5)	(6)
A. Revenue	Δ Revenue					
Δ CZ income	0.337*** (0.067)	0.078 (0.069)	0.134* (0.072)	0.153** (0.066)	0.151*** (0.024)	0.511*** (0.100)
B. Assets	Δ Assets					
Δ CZ income	0.175*** (0.022)	0.135*** (0.042)	0.098** (0.043)	0.089** (0.044)	0.130*** (0.023)	0.230*** (0.028)
C. Liabilities	Δ Liabilities					
Δ CZ income	0.101** (0.039)	0.179** (0.067)	0.064 (0.054)	0.037 (0.066)	0.012 (0.040)	0.118** (0.048)
D. Expenditure	Δ Expenditure					
Δ CZ income	0.143*** (0.016)	0.148*** (0.044)	0.099*** (0.026)	0.115** (0.047)	0.091*** (0.020)	0.163*** (0.025)
Nonprofit Type	Any	Arts, Culture, and Hu- manities	Education	Health	Human Services	Other
Fixed Effects	Firm Year × State	Firm Year × State	Firm Year × State	Firm Year × State	Firm Year × State	Firm Year × State
Observations	8,623,305	609,727	1,080,710	834,679	2,639,385	3,458,804

Note: This table reports coefficient estimates for Equation 1 for nonprofit revenue, expenditure, assets, and liabilities by organizational type. Nonprofits are classified into five categories based on the taxonomy provided by the NCCS. Δ refers to robust growth rates as defined in Appendix A.2. Nonprofit revenue and expenditure include all accounting income and expenses, respectively. Nonprofit assets and liabilities are defined in accounting terms and measured at the end of the year. CZ income refers to personal income of residents. See main text and Appendix A for data construction details. All standard errors are clustered at the state level.

Standard errors in parentheses. Significance levels: * 10% , ** 5%, *** 1%.

Table B.6: Cyclical Regressions by Legal Form

	(1)	(2)	(3)	(4)
A. Revenue	Δ Revenue			
Δ CZ income	0.337*** (0.067)	0.173*** (0.030)	0.188 (0.164)	0.147** (0.056)
B. Assets	Δ Assets			
Δ CZ income	0.175*** (0.022)	0.135*** (0.026)	0.163*** (0.029)	0.137*** (0.025)
C. Liabilities	Δ Liabilities			
Δ CZ income	0.101** (0.039)	0.063*** (0.021)	-0.001 (0.050)	0.043 (0.043)
D. Expenditure	Δ Expenditure			
Δ CZ income	0.143*** (0.016)	0.107*** (0.015)	0.174** (0.069)	0.057** (0.023)
Legal Form	All	Public Charities	Private Foundations	Other Exempt Organizations
Fixed Effects	Firm Year × State	Firm Year × State	Firm Year × State	Firm Year × State
Observations	8,623,305	4,887,023	1,304,173	2,432,108

Note: This table reports coefficient estimates for Equation 1 for nonprofit revenue and expenditure by legal form. Nonprofits' legal form is declared on their tax returns. Δ refers to robust growth rates as defined in Appendix A.2. Nonprofit revenue and expenditure include all accounting income and expenses, respectively. Nonprofit assets and liabilities are defined in accounting terms and measured at the end of the year. CZ income refers to personal income of residents. See main text and Appendix A for data construction details. All standard errors are clustered at the state level.

Standard errors in parentheses. Significance levels: * 10% , ** 5%, *** 1%.

Table B.7: Cyclical Regressions by Size Class

	(1)	(2)	(3)	(4)
A. Revenue		Δ Revenue		
Δ CZ income	0.337*** (0.067)	0.180*** (0.030)	0.334*** (0.074)	0.435*** (0.097)
B. Assets		Δ Assets		
Δ CZ income	0.175*** (0.022)	0.147*** (0.030)	0.167*** (0.029)	0.200*** (0.025)
C. Liabilities		Δ Liabilities		
Δ CZ income	0.101** (0.039)	0.018 (0.045)	0.079* (0.043)	0.200*** (0.054)
D. Expenditure		Δ Expenditure		
Δ CZ income	0.143*** (0.016)	0.119*** (0.024)	0.142*** (0.021)	0.155*** (0.023)
Nonprofit Size Class	Any	Bottom tercile	Middle tercile	Top tercile
Fixed	Firm	Firm	Firm	Firm
Effects	Year × State	Year × State	Year × State	Year × State
Observations	8,623,305	2,665,633	2,918,312	3,039,360

Note: This table reports coefficient estimates for Equation 1 for nonprofit revenue, expenditure, assets, and liabilities by size class. Nonprofits are classified into size terciles based on asset holdings. Δ refers to robust growth rates as defined in Appendix A.2. Nonprofit revenue and expenditure include all accounting income and expenses, respectively. Nonprofit assets and liabilities are defined in accounting terms and measured at the end of the year. CZ income refers to personal income of residents. See main text and Appendix A for data construction details. All standard errors are clustered at the state level.

Standard errors in parentheses. Significance levels: * 10% , ** 5%, *** 1%.

Table B.8: Cyclical Regressions by Census Region

	(1)	(2)	(3)	(4)	(5)
A. Revenue	Δ Revenue				
Δ CZ income	0.337*** (0.067)	0.785*** (0.036)	0.265** (0.086)	0.258* (0.142)	0.361*** (0.066)
B. Assets	Δ Assets				
Δ CZ income	0.175*** (0.022)	0.260*** (0.034)	0.143** (0.052)	0.175*** (0.033)	0.184*** (0.027)
C. Liabilities	Δ Liabilities				
Δ CZ income	0.101** (0.039)	0.555*** (0.121)	0.007 (0.030)	0.106*** (0.024)	0.066*** (0.021)
D. Expenditure	Δ Expenditure				
Δ CZ income	0.143*** (0.016)	0.198*** (0.030)	0.126*** (0.035)	0.144*** (0.030)	0.144*** (0.021)
Region	All	Northeast	Midwest	South	West
Fixed	Firm	Firm	Firm	Firm	Firm
Effects	Year × State	Year × State	Year × State	Year × State	Year × State
Observations	8,623,305	2,076,510	2,233,055	2,540,710	1,763,600

Note: This table reports coefficient estimates for Equation 1 for nonprofit revenue, expenditure, assets, and liabilities by census region. Δ refers to robust growth rates as defined in Appendix A.2. Nonprofit revenue and expenditure include all accounting income and expenses, respectively. Nonprofit assets and liabilities are defined in accounting terms and measured at the end of the year. CZ income refers to personal income of residents. See main text and Appendix A for data construction details. All standard errors are clustered at the state level.

Standard errors in parentheses. Significance levels: * 10% , ** 5%, *** 1%.

Table B.9: Cyclical Regressions by Urban Status

	(1)	(2)	(3)	(4)
A. Revenue		Δ Revenue		
Δ CZ income	0.337*** (0.067)	0.173*** (0.040)	0.370*** (0.047)	0.394*** (0.049)
B. Assets		Δ Assets		
Δ CZ income	0.175*** (0.022)	0.100*** (0.024)	0.154*** (0.025)	0.209*** (0.012)
C. Liabilities		Δ Liabilities		
Δ CZ income	0.101** (0.039)	0.065** (0.028)	0.065 (0.055)	0.036 (0.045)
D. Expenditure		Δ Expenditure		
Δ CZ income	0.143*** (0.016)	0.097*** (0.021)	0.170*** (0.027)	0.129*** (0.030)
Urban Status	All	Rural	Intermediate	Urban
Fixed	Firm	Firm	Firm	Firm
Effects	Year × State	Year × State	Year × State	Year × State
Observations	8,623,305	2,903,559	3,002,295	2,702,710

Note: This table reports coefficient estimates for Equation 1 for nonprofit revenue, expenditure, assets, and liabilities by census region. Δ refers to robust growth rates as defined in Appendix A.2. Nonprofit revenue and expenditure include all accounting income and expenses, respectively. Nonprofit assets and liabilities are defined in accounting terms and measured at the end of the year. CZ income refers to personal income of residents. Commuting zones are classified into urban status based on the share of population living in urban areas in the 2010 Census. See main text and Appendix A for data construction details. All standard errors are clustered at the state level.

Standard errors in parentheses. Significance levels: * 10% , ** 5%, *** 1%.

B.2 Additional Smoothing Results

Table B.10: Smoothing Regressions for Contributions and Non-Contribution Revenue

	(1)	(2)	(3)	(4)	(5)
A. Contributions			Δ Expenditure		
Δ Contributions	0.199*** (0.009)	0.199*** (0.009)	0.191*** (0.010)	0.191*** (0.010)	0.191*** (0.010)
Δ Contributions × Recession					-0.003** (0.001)
B. Non-Contribution Revenue			Δ Expenditure		
Δ Non-contribution revenue	0.033*** (0.006)	0.033*** (0.006)	0.025*** (0.006)	0.025*** (0.006)	0.025*** (0.007)
Δ Non-contribution revenue × Recession					-0.004** (0.002)
Fixed Effects			Firm Year	Firm Year × State	Firm Year × State
Observations	8,623,305	8,623,305	8,623,305	8,623,305	8,623,305

Note: This table reports coefficient estimates for Equations 3 and 4 using received contributions and non-contribution revenue instead of revenues. Δ refers to robust growth rates as defined in Appendix A.2. Nonprofit contributions include all contributions including grants, donations and individuals contributions. Nonprofit expenditure include all accounting expenses. Nonprofit non-contribution revenue includes all revenue except for contributions, e.g. sales and operating income. Recession is an indicator for recession years as defined by the NBER. See main text and Appendix A for data construction details. All standard errors are clustered at the state level.

Standard errors in parentheses. Significance levels: * 10% , ** 5%, *** 1%.

Table B.11: Smoothing Regressions for Donations and Non-Donation Revenue

	(1)	(2)	(3)	(4)	(5)
A. Donations	Δ Expenditure				
Δ Donations	0.032*** (0.002)	0.032*** (0.002)	0.029*** (0.002)	0.029*** (0.002)	0.032*** (0.002)
Δ Donations \times Recession					-0.013*** (0.002)
B. Non-Donation Revenue	Δ Expenditure				
Δ Non-donation revenue	0.120*** (0.003)	0.123*** (0.003)	0.112*** (0.003)	0.112*** (0.003)	0.131*** (0.003)
Δ Non-donation revenue \times Recession					-0.070*** (0.004)
Fixed Effects		Year	Firm Year	Firm Year \times State	Firm Year \times State
Observations	375,165	375,165	375,165	375,165	375,165

Note: This table reports coefficient estimates for Equations 3 and 4 using received donations and non-donation revenue in the supplemental dataset. Δ refers to robust growth rates as defined in Appendix A.2. Nonprofit donations include grants and donations made by private parties. Nonprofit expenditure include all accounting expenses. Nonprofit non-donation revenue includes all revenue except for donations, e.g. government grants, and sales and operating income. Recession is an indicator for recession years as defined by the NBER. See main text and Appendix A for data construction details. All standard errors are clustered at the state level.

Standard errors in parentheses. Significance levels: * 10% , ** 5%, *** 1%.

Table B.12: Smoothing Regressions by Nonprofit Type

	(1)	(2)	(3)	(4)	(5)	(6)
			Δ Expenditure			
Δ Revenue	0.231*** (0.008)	0.317*** (0.009)	0.257*** (0.013)	0.363*** (0.005)	0.375*** (0.007)	0.165*** (0.006)
Δ Revenue \times Recession	-0.041*** (0.003)	-0.038*** (0.007)	-0.040*** (0.006)	-0.068*** (0.007)	-0.036*** (0.004)	-0.029*** (0.003)
Nonprofit Type	Any	Arts, Culture, and Hu- manities	Education	Health	Human Services	Other
Fixed Effects	Firm Year \times State	Firm Year \times State	Firm Year \times State	Firm Year \times State	Firm Year \times State	Firm Year \times State
Observations	8,623,305	609,727	1,080,710	834,679	2,639,385	3,458,804

Note: This table reports coefficient estimates for Equation 4 by organizational type. Nonprofits are classified into five categories based on the taxonomy provided by the NCCS. Δ refers to robust growth rates as defined in Appendix A.2. Nonprofit revenue and expenditure include all accounting income and expenses, respectively. Recession is an indicator for recession years as defined by the NBER. See main text and Appendix A for data construction details. All standard errors are clustered at the state level.

Standard Errors in Parentheses. Significance levels: * 10% , ** 5%, *** 1%.

Table B.13: Alternative Smoothing Regressions by Nonprofit Type

	(1)	(2)	(3)	(4)	(5)	(6)
	Δ Expenditure					
Δ Revenue	0.190*** (0.008)	0.279*** (0.010)	0.220*** (0.015)	0.286*** (0.009)	0.344*** (0.009)	0.134*** (0.007)
Δ Revenue \times Δ Nat. income	0.781*** (0.053)	0.739*** (0.114)	0.730*** (0.097)	1.539*** (0.128)	0.569*** (0.073)	0.619*** (0.044)
Nonprofit Type	Any	Arts, Culture, and Hu- manities	Education	Health	Human Services	Other
Fixed Effects	Firm Year \times State	Firm Year \times State	Firm Year \times State	Firm Year \times State	Firm Year \times State	Firm Year \times State
Observations	8,623,305	609,727	1,080,710	834,679	2,639,385	3,458,804

Note: This table reports coefficient estimates for Equation 4 by organizational type. Nonprofits are classified into five categories based on the taxonomy provided by the NCCS. Δ refers to robust growth rates as defined in Appendix A.2. Nonprofit revenue and expenditure include all accounting income and expenses, respectively. See main text and Appendix A for data construction details. All standard errors are clustered at the state level.

Standard Errors in Parentheses. Significance levels: * 10% , ** 5%, *** 1%.

Table B.14: Smoothing Regressions by Legal Form

	(1)	(2)	(3)	(4)
	Δ Expenditure			
Δ Revenue	0.231*** (0.008)	0.359*** (0.008)	0.089*** (0.005)	0.347*** (0.010)
Δ Revenue \times Recession	-0.041*** (0.003)	-0.046*** (0.004)	-0.005** (0.002)	-0.067*** (0.009)
Legal Form	All	Public Charities	Private Foundations	Other Exempt Organizations
Fixed Effects	Firm Year \times State	Firm Year \times State	Firm Year \times State	Firm Year \times State
Observations	8,623,305	4,887,023	1,304,173	2,432,108

Note: This table reports coefficient estimates for Equation 4 by legal form. Nonprofits' legal form is declared on their tax returns. Δ refers to robust growth rates as defined in Appendix A.2. Nonprofit revenue and expenditure include all accounting income and expenses, respectively. Recession is an indicator for recession years as defined by the NBER. See main text and Appendix A for data construction details. All standard errors are clustered at the state level.

Standard Errors in Parentheses. Significance levels: * 10% , ** 5%, *** 1%.

Table B.15: Alternative Smoothing Regressions by Legal Form

	(1)	(2)	(3)	(4)
	Δ Expenditure			
Δ Revenue	0.190*** (0.008)	0.313*** (0.010)	0.071*** (0.005)	0.294*** (0.016)
Δ Revenue \times Δ Nat. income	0.781*** (0.053)	0.919*** (0.072)	0.430*** (0.036)	0.939*** (0.165)
Legal Form	All	Public Charities	Private Foundations	Other Exempt Organizations
Fixed Effects	Firm Year \times State	Firm Year \times State	Firm Year \times State	Firm Year \times State
Observations	8,623,305	4,887,023	1,304,173	2,432,108

Note: This table reports coefficient estimates for Equation 4 by legal form. Nonprofits' legal form is declared on their tax returns. Δ refers to robust growth rates as defined in Appendix A.2. Nonprofit revenue and expenditure include all accounting income and expenses, respectively. See main text and Appendix A for data construction details. All standard errors are clustered at the state level.

Standard Errors in Parentheses. Significance levels: * 10% , ** 5%, *** 1%.

Table B.16: Smoothing Regressions by Size Class

	(1)	(2)	(3)	(4)
	Δ Expenditure			
Δ Revenue	0.231*** (0.008)	0.390*** (0.010)	0.216*** (0.008)	0.141*** (0.004)
Δ Revenue \times Recession	-0.041*** (0.003)	-0.027*** (0.005)	-0.041*** (0.004)	-0.027*** (0.002)
Nonprofit Size Class	Any	Bottom tercile	Middle tercile	Top tercile
Fixed Effects	Firm Year \times State	Firm Year \times State	Firm Year \times State	Firm Year \times State
Observations	8,623,305	2,665,633	2,918,312	3,039,360

Note: This table reports coefficient estimates for Equation 4 by size class. Nonprofits are classified into size terciles based on asset holdings. Δ refers to robust growth rates as defined in Appendix A.2. Nonprofit revenue and expenditure include all accounting income and expenses, respectively. Recession is an indicator for recession years as defined by the NBER. See main text and Appendix A for data construction details. All standard errors are clustered at the state level.

Standard Errors in Parentheses. Significance levels: * 10% , ** 5%, *** 1%.

Table B.17: Alternative Smoothing Regressions by Size Class

	(1)	(2)	(3)	(4)
	Δ Expenditure			
Δ Revenue	0.190*** (0.008)	0.375*** (0.011)	0.172*** (0.008)	0.103*** (0.004)
Δ Revenue \times Δ Nat. income	0.781*** (0.053)	0.227*** (0.079)	0.876*** (0.074)	0.806*** (0.045)
Nonprofit Size Class	Any	Bottom tercile	Middle tercile	Top tercile
Fixed	Firm	Firm	Firm	Firm
Effects	Year \times State	Year \times State	Year \times State	Year \times State
Observations	8,623,305	2,665,633	2,918,312	3,039,360

Note: This table reports coefficient estimates for Equation 4 by size class. Nonprofits are classified into size terciles based on asset holdings. Δ refers to robust growth rates as defined in Appendix A.2. Nonprofit revenue and expenditure include all accounting income and expenses, respectively. See main text and Appendix A for data construction details. All standard errors are clustered at the state level.

Standard Errors in Parentheses. Significance levels: * 10% , ** 5%, *** 1%.

Table B.18: Smoothing Regressions by Census Region

	(1)	(2)	(3)	(4)	(5)
	Δ Expenditure				
Δ Revenue	0.231*** (0.008)	0.195*** (0.005)	0.217*** (0.008)	0.248*** (0.009)	0.270*** (0.004)
Δ Revenue \times Recession	-0.041*** (0.003)	-0.037*** (0.006)	-0.041*** (0.004)	-0.050*** (0.006)	-0.033*** (0.004)
Region	All	Northeast	Midwest	South	West
Fixed	Firm	Firm	Firm	Firm	Firm
Effects	Year \times State	Year \times State	Year \times State	Year \times State	Year \times State
Observations	8,623,305	2,076,510	2,233,055	2,540,710	1,763,600

Note: This table reports coefficient estimates for Equation 4 by size class. Δ refers to robust growth rates as defined in Appendix A.2. Nonprofit revenue and expenditure include all accounting income and expenses, respectively. Recession is an indicator for recession years as defined by the NBER. See main text and Appendix A for data construction details. All standard errors are clustered at the state level.

Standard Errors in Parentheses. Significance levels: * 10% , ** 5%, *** 1%.

Table B.19: Alternative Smoothing Regressions by Census Region

	(1)	(2)	(3)	(4)	(5)
	Δ Expenditure				
Δ Revenue	0.190*** (0.008)	0.158*** (0.009)	0.175*** (0.009)	0.203*** (0.012)	0.235*** (0.005)
Δ Revenue \times Δ Nat. income	0.781*** (0.053)	0.722*** (0.090)	0.818*** (0.065)	0.854*** (0.132)	0.687*** (0.051)
Region	All	Northeast	Midwest	South	West
Fixed	Firm	Firm	Firm	Firm	Firm
Effects	Year \times State	Year \times State	Year \times State	Year \times State	Year \times State
Observations	8,623,305	2,076,510	2,233,055	2,540,710	1,763,600

Note: This table reports coefficient estimates for Equation 4 by size class. Δ refers to robust growth rates as defined in Appendix A.2. Nonprofit revenue and expenditure include all accounting income and expenses, respectively. See main text and Appendix A for data construction details. All standard errors are clustered at the state level.

Standard Errors in Parentheses. Significance levels: * 10% , ** 5%, *** 1%.

Table B.20: Smoothing Regressions by Urban Status

	(1)	(2)	(3)	(4)
	Δ Expenditure			
Δ Revenue	0.231*** (0.008)	0.256*** (0.004)	0.226*** (0.012)	0.218*** (0.014)
Δ Revenue \times Recession	-0.041*** (0.003)	-0.042*** (0.010)	-0.046*** (0.004)	-0.036*** (0.003)
Urban Status	All	Rural	Intermediate	Urban
Fixed	Firm	Firm	Firm	Firm
Effects	Year \times State	Year \times State	Year \times State	Year \times State
Observations	8,623,305	2,903,559	3,002,295	2,702,710

Note: This table reports coefficient estimates for Equation 4 by size class. Δ refers to robust growth rates as defined in Appendix A.2. Nonprofit revenue and expenditure include all accounting income and expenses, respectively. Recession is an indicator for recession years as defined by the NBER. Commuting zones are classified into urban status based on the share of population living in urban areas in the 2010 Census. See main text and Appendix A for data construction details. All standard errors are clustered at the state level.

Standard Errors in Parentheses. Significance levels: * 10% , ** 5%, *** 1%.

Table B.21: Alternative Smoothing Regressions by Urban Status

	(1)	(2)	(3)	(4)
	Δ Expenditure			
Δ Revenue	0.190*** (0.008)	0.216*** (0.008)	0.179*** (0.013)	0.185*** (0.014)
Δ Revenue \times Δ Nat. income	0.781*** (0.053)	0.756*** (0.119)	0.931*** (0.071)	0.637*** (0.028)
Urban Status	All	Rural	Intermediate	Urban
Fixed	Firm	Firm	Firm	Firm
Effects	Year \times State	Year \times State	Year \times State	Year \times State
Observations	8,623,305	2,903,559	3,002,295	2,702,710

Note: This table reports coefficient estimates for Equation 4 by size class. Δ refers to robust growth rates as defined in Appendix A.2. Nonprofit revenue and expenditure include all accounting income and expenses, respectively. Commuting zones are classified into urban status based on the share of population living in urban areas in the 2010 Census. See main text and Appendix A for data construction details. All standard errors are clustered at the state level.

Standard Errors in Parentheses. Significance levels: * 10% , ** 5%, *** 1%.

B.3 Estimating in Levels instead of Growth Rates

Table B.22: Cyclical Regressions in Levels

	(1)	(2)	(3)	(4)	(5)
A. Revenue			Revenue		
Nat. income	2.128*** (0.148)	2.046*** (0.155)			
CZ Income		0.122*** (0.012)	0.122*** (0.012)	0.042*** (0.005)	0.025*** (0.004)
B. Assets			Assets		
Nat. income	1.126*** (0.116)	1.072*** (0.118)			
CZ Income		0.080*** (0.011)	0.079*** (0.011)	0.033*** (0.004)	0.018*** (0.004)
C. Liabilities			Liabilities		
Nat. income	5.071*** (0.416)	5.003*** (0.425)			
CZ Income		0.103*** (0.017)	0.103*** (0.017)	0.046*** (0.007)	0.027*** (0.006)
D. Expenditure			Expenditure		
Nat. income	1.424*** (0.117)	1.338*** (0.123)			
CZ Income		0.129*** (0.012)	0.130*** (0.012)	0.035*** (0.004)	0.022*** (0.004)
Fixed				Nonprofit	Nonprofit
Effects	Linear	Linear	Year	Year	Year × State
	Trend	Trend			
Observations	4,709,355	4,709,355	4,709,355	4,709,355	4,709,355

Note: This table reports coefficient estimates for Equation 1 for nonprofit revenue, assets, liabilities, and expenditure. Nonprofit revenue and expenditure include all accounting income and expenses, respectively. Nonprofit assets and liabilities are measured at the end of the year and include all accounting assets and liabilities, respectively. All variables are in logs. CZ income refer to personal income of residents. See Appendix A for data construction details. All standard errors are clustered at the state level.

Standard errors in parentheses. Significance levels: * 10% , ** 5%, *** 1%.

Table B.23: Smoothing Regressions in Levels

	(1)	(2)	(3)	(4)	(5)
	Expenditure				
Revenue	0.919*** (0.004)	0.920*** (0.004)	0.507*** (0.008)	0.507*** (0.008)	0.507*** (0.008)
Revenue \times Recession					-0.004*** (0.001)
Fixed Effects		Year	Firm Year	Firm Year \times State	Firm Year \times State
Observations	8,414,389	8,414,389	8,335,550	8,335,550	8,335,550

Note: This table reports coefficient estimates for Equations 3 and 4. Nonprofit revenue and expenditure include all accounting income and expenses, respectively. All variables are in logs. Recession is an indicator for recession years as defined by the NBER. See Appendix A for data construction details. All standard errors are clustered at the state level.

Standard Errors in Parentheses. Significance levels: * 10% , ** 5%, *** 1%.