Turnout and Amendment 4: Mobilizing Eligible Voters Close to the Formerly Incarcerated*

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 $_{5}$ Abstract

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17 18 Recent scholarship has established a link between felony disenfranchisement and lower turnout, particularly in Black communities. Little work, however, has been done to interrogate how this depressive effect might be counteracted. In 2018, Amendment 4 was on the ballot in Florida, and promised to re-enfranchise most of the disenfranchised population. The presence of this ballot initiative offers a unique opportunity to investigate whether ballot initiatives of special interest to these impacted communities might ameliorate some of the depressed turnout. Using individual-level release records from the Florida Department of Corrections I test whether the ballot initiative mobilized neighborhoods and individuals in close proximity to formerly incarcerated individuals. Using multiple identification strategies, I find no evidence that Amendment 4 increased the turnout of these neighborhoods or eligible voters, indicating that even greater engagement and investment must be made to overcome the depressed turnout.

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19 Introduction

On November 6th, 2018, Floridians voted to amend their state constitution to re-enfranchise individuals with felony convictions in their past (Taylor 2018). The move was hailed as 21 transformative for Floridian — and American — democracy; Uggen, Larson, and Shannon (2016) had estimated a few years earlier that some 1.5 million Floridians were disenfranchised and had finished serving their sentences, making the amendment the largest expansion of the franchise in the United States since the Twenty-sixth Amendment lowered the voting age 25 to 18. The amendment received broad support. Although it needed just 60 percent of the vote to pass, 64.5 percent of voters supported the ballot initiative. This support contrasts 27 sharply with other statewide races: Ron DeSantis won the gubernatorial race with only 49.5 percent of the vote, while winning just 49.9 percent sent Rick Scott to the United States Senate. Prior to 2018, Floridians convicted of felony offenses were permanently disenfranchised unless 31 they applied for and received an individual pardon from the state's clemency board. This 32 was characterized by a "low success rate, cumbersome process, and lengthy amount of time" 33 (B. L. Miller and Spillane 2012b, 432) and was driven in part by gubernatorial discretion: although Charlie Crist restored voting rights to roughly 150 thousand individuals over a 4 35 year period, Rick Scott did so for fewer than 3 thousand people over 8 years (Schlakman 2018). At the time Amendment 4 was passed, it was widely reported that the backlog of 37 applications was nearly 10,000 and the wait stretched for as long as a decade (Ramadan, Stucka, and Washington 2018). Over the years, Florida's procedure was subject to numerous lawsuits, and was ruled unconstitutional in early 2018 with Judge Mark Walker describing it as "a gauntlet of constitutionally infirm hurdles." Amendment 4 promised to automatically restore voting rights once individuals had completed their sentence, though it did not apply to individuals convicted of murder or sexual offenses.

 $^{^1\}mathrm{Hand}$ et al. v. Scott et al., 4:17cv128-MW/CAS (U.S. District Court for the Northern District of Florida 2018).

This study explores whether the opportunity to vote on Amendment 4 increased participation among eligible voters who lived with or near individuals disenfranchised due to a period of felony incarceration. Americans' political knowledge is deeply shaped by the incarceration of a loved one (Lee, Porter, and Comfort 2014), and exposure to the carceral state chills political involvement even among individuals who are not convicted. Even indirect contact with the criminal justice system leaves would-be voters feeling as though political involvement is not for "people like me," often despite having considerable political knowledge (Lerman and Weaver 2014). A growing body of quantitative research captures these "spillover" effects, 51 demonstrating that neighborhoods with high levels of incarceration and disenfranchisement vote at markedly lower rates than other similar neighborhoods (e.g. Burch 2013; Morris 2020). Amendment 4 in Florida offers a unique opportunity to investigate whether these chilling effects can be overcome by a ballot initiative. As I explain in the section that follows, Amendment 4 offered individuals living with or near formerly incarcerated individuals an 57 opportunity to redefine their relationship with the government in positive ways. Although

this made the ballot initiative perhaps particularly salient for these individuals, it took place

against the backdrop of an entrenched carceral state that negatively structured many facets

of their lives (see, for instance, Travis and Waul 2003). It is unclear whether the ballot

initiative was sufficient to bring these individuals into the voting booth, or if the fight for

political reincorporation will demand longer-term investment in these communities.

Theory and Literature

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It is well established that a criminal conviction — and, more specifically, a period of incarceration — decreases turnout even when individuals are no longer legally disenfranchised (Weaver and Lerman 2010; Burch 2011; White 2019b; but see Gerber et al. 2017). The effect of disenfranchisement policy on the political behavior of individuals who experience the criminal justice system indirectly via the conviction of a family or community member, however, is somewhat mixed. Most research finds that turnout is substantially lower in states with stricter voter disenfranchisement policies or more disenfranchised citizens (e.g. Bowers and Preuhs 2009; King and Erickson 2016), though Miles (2004) argues that these effects are small. The little research that has explored the spillover effects of disenfranchisement policy at the *neighborhood* level has similarly found evidence that incarceration and disenfranchisement demobilizes eligible voters in impacted communities (Burch 2013; Morris 2020).

Understanding whether Amendment 4 was likely to recoup the lost turnout of eligible voters

who lived with or near the disenfranchised requires understanding how their indirect exposure 78 to the criminal justice system (or "proximal contact" (Walker 2014)) depressed turnout to begin with. White (2019a) argues that indirect exposure can reduce turnout through resource constraints (the return of a family member from prison can place burdens on households that make voting less likely) and social constraints (the stigma associated with a criminal conviction can cause individuals to withdraw from political life), an approach I adopt here. Exposure to the criminal justice system can result in result in loss of resources for households and families through both the imposition of legal financial obligations (LFOs) and adverse employment prospects for those with a conviction (Naser and Visher 2006; Bushway, Stoll, and Weiman 2007; Martin et al. 2018). Stressors associated with the return of a household 87 member from prison can make it harder for eligible voters to invest the time and resources involved in registering to vote, learning about candidates, or locating their polling place. It seems unlikely that the presence of Amendment 4 on the ballot could have reduced the costs of voting for this group, and therefore probably left the resource constraints reducing turnout undisturbed.

Indirect exposure to the criminal justice system can also lower eligible voters' propensity to vote through social mechanisms. There is shame and stigma associated with having

a felony conviction in one's past (Austin 2004; Uggen, Manza, and Behrens 2004; Miller and Agnich 2016) which may also extend to family members without a conviction. Work from Vesla Weaver and Amy Lerman (2010; 2014) describes in great detail how interactions with the criminal justice system structure individuals' relationship with their government, with a felony conviction serving as "a durable constraint and marker of their citizenship" 99 (Lerman and Weaver 2014, 133). They explain that even citizens without criminal histories 100 in communities with aggressive crime control "become less likely to believe that they (and 101 those like them) can change the system, a reduction in external efficacy" (Lerman and Weaver 102 2014, 137, emphasis in the original). Their work is replete with examples of individuals who 103 know much about politics yet choose to "stay below the radar" because "'they're [government 104 officials] not interested in what I have to say'" (Lerman and Weaver 2014, 210). Importantly, 105 these demobilizing consequences are not limited to those who are convicted; rather, "the 106 correctional system becomes a primary mode of political socialization for families of the 107 incarcerated" through which spouses and children learn to distrust and avoid interaction 108 with the state (Lee, Porter, and Comfort 2014, 45; see also Comfort 2008, 2016; Kirk 2016). 109 It seems that both the substance of the proposed constitutional amendment and the mes-110 saging used by the campaign supporting its passage could have lessened some of the social 111 barriers to voting. Restoring voting rights to individuals who had been convicted of felony 112 offenses would end the "civil death" of felony disenfranchisement (Ewald 2002; B. L. Miller 113 and Spillane 2012a). This civil death — the legal lessening of a formerly incarcerated indi-114 vidual's rights as citizen — certainly qualifies as the sort of "durable constraint and marker" 115 identified by Weaver and Lerman. The opportunity to end post-sentence disenfranchisement 116 could increase turnout among those in community with the formerly incarcerated. Amend-117 ment 4 offered these eligible voters the chance to affirm that their family and community 118 members deserved to have their voices heard in the democratic arena. 119

Moreover, the public messaging employed by the Amendment 4 campaign was explicitly designed to change how voters understood the citizenship of disenfranchised individuals. The

campaign cast the ballot initiative as an issue of fairness, criticizing Florida's existing disenfranchisement policy for creating two tiers of citizenship. The organization leading the 123 campaign leveraged the notion that disenfranchised citizens deserved to be re-incorporated 124 into the body politic in its very name — "Second Chances Florida." The framing was 125 effective: the editorial boards of each of Florida's three biggest newspapers endorsed the 126 amendment, all using language related to fairness and civic redemption. The Tampa Bay 127 Times told readers they had a "remarkable opportunity to remedy that unfairness" (Tampa 128 Bay Times 2018); the Sun Sentinel informed voters "[t]here may never be an opportunity 129 to do a better thing than to vote yes on this reform" (Sun Sentinel 2018); and the Orlando 130 Sentinel said that Florida's then-policy "denie[d] our fellow citizens a second chance. It 131 denie[d] redemption" (Orlando Sentinel 2018). Recent work demonstrates that the demobi-132 lizing effects of direct and indirect contact with the criminal justice system can be overcome 133 when these experiences are linked with narratives of injustice (Walker and García-Castañon 134 2017: Walker 2020). Insofar as the campaign was successful at helping these individuals 135 understand the experiences of their formerly incarcerated family and community members 136 in the context of a broader narrative of (racial) injustice, they may have been mobilized to 137 vote.

In addition to newspapers across the state, the campaign deployed "volunteers from a broad 139 coalition that included advocacy groups, Christian organizations, the League of Women 140 Voters, criminal justice experts and, of course, those who had been convicted of felonies" 141 (Robles 2018). Andrew Gillum, the Democratic gubernatorial candidate, also vocally sup-142 ported the amendment, openly discussing his family's relationship with the criminal justice 143 system and his own sibling's disenfranchisement (Smith 2018). Voters were thus getting cues 144 from all sorts of messengers that Amendment 4 deserved to be passed, and that individuals 145 with convictions in their past should be allowed to vote. These cues, plus the descriptive 146 representation (Merolla, Sellers, and Fowler 2013) promised by Gillum, could have proved 147 mobilizing. 148

Finally, it is important to recognize that the geographic concentration of policing and incarceration patterns (e.g. Gelman, Fagan, and Kiss 2007) mean that individuals in community 150 with the formerly incarcerated might also have other, direct relationships with the criminal 151 justice system. In 2017 there were 711,831 arrests in Florida but just 134,554 guilty felonious 152 dispositions.² Although individuals who were arrested but not convicted of felonies were not 153 legally disenfranchised, the literature discussed above demonstrates that even low-level in-154 teractions can have a chilling effect on one's relationship with the government. The rhetoric 155 used to argue in favor of the ballot initiative might have led them to reevaluate their own 156 citizen identities, despite not changing their eligibility to vote. 157

At the same time, there is some reason to think the ballot initiative might not increase turnout among voters in close contact with formerly incarcerated, disenfranchised individuals. The ruptures discussed above of the citizen's relationship with the government engendered by the carceral state are deep: individuals under the purview of the carceral state "learn" their place in the system over a very long period, both through incarceration and day-to-day interactions with government representatives such as the police. It is perhaps naive to expect that a single ballot initiative could overcome these negative forces.

Moreover, the individuals in these neighborhoods were perhaps less familiar with the content of Amendment 4 than others: Bowler and Donovan (1994), for instance, demonstrates that education and polarization are strong predictors of individuals' familiarity with ballot initiatives. Shaker (2012) also finds that higher-educated individuals are more knowledgeable about local politics. Therefore, despite the greater potential salience of Amendment 4 for these individuals, they may have been less aware of it in the first place, in which case it obviously would not motivate them to cast a ballot.

²See http://edr.state.fl.us/Content/resource-demand/criminal-justice/reports/criminal-justice/cj7.pdf.

Research Design and Expectations

I begin by testing whether a neighborhood's formerly incarcerated population influenced that neighborhood's turnout in 2018. Because statewide felony probation records are not available, this analysis is based on only the subset of disenfranchised individuals who were imprisoned for a felony conviction. Neighborhoods that are home to formerly incarcerated individuals are identified by geocoding release records from the Florida Department of Corrections, and I offer two definitions of neighborhoods.

Neighborhoods are first defined as precincts. The Florida Division of Elections makes election 179 results available at this level, which allows me to test turnout specifically on Amendment 4, 180 neighborhood-level support for the amendment, and the share of voters who "rolled off" (or chose not to vote) for Amendment 4. Unfortunately, the use of precinct-level data leaves us with a major drawback: when doing analysis at this level, bias-free turnout denominators 183 are hard to come by. Because the Census Bureau does not produce population estimates for 184 individual voting precincts, turnout cannot be calculated by dividing the number of ballots 185 cast by the eligible population; rather, it must be constructed as a share of registered voters. 186 If there is a relationship between the number of formerly incarcerated residents and the 187 registration rate of a neighborhood, our estimates will be biased. 188

That could be the case in the study at hand. Political organizers working on behalf of
Amendment 4 may have focused on registering eligible residents in neighborhoods where disenfranchised individuals lived. If these organizers registered many new voters but a relatively
small share of the new voters actually turned out, the net effect might be higher turnout
among eligible residents but lower turnout among registered voters. For further discussion of
how improper denominators can bias turnout estimates, see Amos, McDonald, and Watkins
(2017) and Amos and McDonald (2020).

To address this potential problem, I also define neighborhoods as Census block groups. The
Census Bureau makes estimates of the citizen voting-age population available at this level

which is a better denominator for calculating turnout. In this case, however, I must use a geocoded voter file to determine turnout. Because I aggregate the number of participants in a block group from individual-level data, I cannot determine whether an individual actually participated in the contest for Amendment 4 or they rolled off. Similarly, I am unable to interrogate the relationship between block group characteristics and support for Amendment 4. Although each definition of neighborhood presents some drawbacks, the two definitions together paint a full picture.

After examining whether the presence of formerly incarcerated individuals was related with neighborhoods' voting behavior, I ask whether voters who lived with formerly incarcerated individuals turned out at higher rates in 2018. For this analysis, I use the release plan addresses of formerly incarcerated individuals (the most recent address available, according to the Department of Corrections) and voter file data to identify registered voters who lived with formerly incarcerated individuals. Voters are considered "treated" if they lived with a formerly incarcerated individual, and "untreated" otherwise. I then use a variety of individual- and neighborhood-level characteristics to match treated and untreated voters using a genetic algorithm (Sekhon 2011).

After matching these voters, I employ a difference-in-differences specification to determine
whether treated voters participated at higher rates in the 2018 election. These analyses
are run for all voters who lived with a formerly incarcerated individual, as well as only
the subset of households whose members have not been to prison for many years. This
final specification allows me to disentangle the depressive effect of indirect exposure to the
criminal justice system from the mobilizing effect of Amendment 4 in 2018 by incorporating
any depressive effect into the pre-2018 baseline.

Table 1 summarizes the specific hypotheses this manuscript tests.

Table 1: Hypotheses

	Hypothesis	Approach						
Neighborhood Level								
1a.	Each additional formerly incarcerated resident in a voting precinct	OLS regression						
	is associated with increased turnout among registered voters in that							
	precinct.							
1b.	Each additional formerly incarcerated resident in a Census block	OLS regression						
	group is associated with increased turnout among eligible citizens in							
	that block group.							
2.	Each additional formerly incarcerated resident in a voting precinct is	OLS regression						
	associated with increased support for Amendment 4 in that precinct.							
3.	Each additional formerly incarcerated resident in a voting precinct	OLS regression						
	is associated with decreased roll-off in that precinct.							
House	ehold Level							
4.	Amendment 4 increased turnout in 2018 among household members	Difference-in-differences						
	of formerly incarcerated individuals. This treatment effect was	comparing turnout of voters in						
	especially large among households whose members have not been to	treated households to voters in						
	prison for many years.	untreated households.						

Data

- I leverage multiple data sources to investigate whether individuals in community with for-
- merly incarcerated residents were more likely to vote in the 2018 election.

225 Department of Corrections Data

- ²²⁶ Felony incarceration records come from the Florida Department of Corrections' Offender
- 227 Based Information System (OBIS). The OBIS includes all individuals released from prison
- following a felony conviction since October 1, 1997. I retain only the record associated with an

individual's most recent incarceration according to the release date, and identify all formerly incarcerated individuals who were finished with their sentence as of the 2018 election by crossreferencing these records against imprisonment and parole records. Individuals sentenced to felony probation, who were also disenfranchised, are not included in this data. I include only individuals who completed their sentence; individuals who died or absconded before their sentence was completed are removed from the dataset.

The OBIS provides the "release plan address" for individuals who were formerly incarcerated. According to the Department of Corrections, this is the most recent address available
for individuals who are no longer under supervision. Using the release plan address for individuals last released from prison many years ago presents some potential problems. Some
of these individuals have surely died or moved. In Appendix B I show that the results presented in the body of this manuscript when I limit the pool of formerly incarcerated people
to individuals released from prison during or after 2015. Because these individuals were
released more recently, their addresses are probably more accurate.

The address data are messy and require substantial cleaning. In some cases, the address 243 field is left blank; in others, the record simply notes the road or the town of the individual's 244 residence, without providing full address information. I assume that any record that does 245 not begin with an integer does not have a full address and cannot be used (this results in 246 the exclusion of just under 3 percent of records). The remaining addresses are geocoded. 247 Individuals whose addresses were geocoded outside of Florida (10.6 percent) or for whom 248 the geocoder failed (3.1 percent) are dropped. At least 94 percent of individuals released to 249 addresses in Florida are therefore included. The failure rate is likely too small to materially 250 impact the analyses. 251

252 Many formerly incarcerated individuals leave prison not for homes with family members, but

³The OBIS lists current addresses for individuals currently under community supervision, which may differ from the release plan addresses. However, according to a response to a public records request filed by the author with the Department of Corrections, these historical data are not maintained once an individual has been discharged.

rather to homeless shelters or other sites of incarceration. Of the five most commonly listed addresses, three were Immigration and Customs Enforcement properties, one was owned by the Salvation Army, and one was a rescue mission. The body of this manuscript excludes formerly incarcerated individuals whose address was listed by five or more individuals, as institutions for returning citizens may have uniquely structured responses to Amendment Appendix B shows that the primary findings in the manuscript hold when I include all formerly incarcerated individuals. Just over 15 percent of formerly incarcerated individuals listed these sorts of addresses as their post-incarceration residence.

The successfully geocoded, formerly incarcerated individuals are then mapped to their home

Census block groups using shapefiles from the Census Bureau, and to their home voter

precincts using shapefile data collected by Kelso and Migurski (2018).

Between 75 and 80 percent of individuals found guilty of felonies in recent years in Florida 264 have been sentenced to probation, but their records are not available statewide and they 265 are thus excluded from this study. This may pose a problem: neighborhoods with resi-266 dents disenfranchised due to felony probation are also "treated," as are housemates of these 267 individuals. The raw numbers, however, likely overstate the problem: not all individuals 268 sentenced to felony probation actually lose their voting rights. Florida judges are allowed to 269 "withhold adjudication" (Tragos and Sartes 2008), meaning defendants consent to pay fines 270 and restitution and to serve a term of probation, but they are not disenfranchised. 271

Though unavailable statewide, probation records with residential addresses are available for Hillsborough County, the Florida county with the third-highest number of formerly incarcerated individuals according to the OBIS records. In Hillsborough County, the correlation coefficient between the number of felony probationers and formerly incarcerated residents (scaled by population) is 0.92 at the block group level. The number of formerly incarcerated individuals in a neighborhood is therefore a reasonable proxy for the total number of

 $^{^{4}} See \qquad http://edr.state.fl.us/Content/resource-demand/criminal-justice/reports/criminal-justice/index.cfm.$

278 disenfranchised residents.

The neighborhood- and individual-level models presented in the body of this manuscript are re-estimated for Hillsborough County in Appendix A, with individuals sentenced both to felony incarceration and probation included. Their incorporation does not impact the primary results. Although this study relies only on formerly incarcerated individuals, the data available for robustness checks indicate that the relationships detailed here probably extend to the full disenfranchised population.

Voter File Data and Census Data

I primarily use Florida voter file data from the data vendor L2 Political which includes information on individuals such as their home address, their age and gender, their participation history, and their political affiliation. Although the L2 data includes estimates of voters' race and ethnicity, the raw Florida voter file includes self-identified race and ethnicity. In place of L2's estimates, I use this self-reported data. I also use the raw Florida file to provide the gender for voters for whom L2 did not have an estimate, as well as voters' home counties and precincts.

Precinct and block group demographics are constructed using the voter file data. Neighborhood characteristics such as age and race are the averages of all registered voters in that neighborhood. For characteristics such as income that are unavailable at the individual level, voters are assigned the value associated with their home block group from the American Community Survey's 2014 – 2018 5-year estimates; the precinct average income, therefore, is effectively the average of all the block groups within that precinct, weighted by the number of registered voters.

Matched Department of Corrections and Voter File Data

I identify registered voters who lived with formerly incarcerated individuals by matching on residential addresses. As discussed above, these addresses are often in different formats. To increase the quality of the matches, I standardize common street and address abbreviations as well as capitalization. "Boulevard," for instance, becomes "BLVD" in each instance in the DOC and voter file data. These standardizations are taken from Appendix C of the USPS Postal Addressing Standards (2015). Exact matching is required.

Voters with indirect exposure to the criminal justice system might have been motivated to

307 Potential Confounders

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turn out through avenues other than the ballot initiative. For instance, Andrew Gillum 309 was poised to become the state's first Black governor, which could increase Black turnout 310 (e.g. Washington 2006; Fairdosi and Rogowski 2015; Miller and Chaturvedi 2018). By con-311 trolling for neighborhood demographics (and, in the matching exercise, forcing control voters 312 to mirror treated voters on key demographics such as race and party affiliation), I minimize 313 the differences between the two groups along characteristics known to influence turnout. 314 It is also worth considering whether electoral reforms had unique influence on the voters of interest in this study, though there is little reason to believe this is the case. The number of 316 early voting days was cut for the 2012 general election, but the longer period was restored for 317 the 2014 – 2018 period.⁵ Early voting was not allowed on college campuses in the 2014 and 318 2016 elections, though it was allowed in 2018 (Bousquet 2018). I include neighborhood-level 310 estimates of collegiate education in each of the regressions to mitigate the potential effects of 320 this change. Florida did not enact other reforms such as same-day registration or automatic 321 voter registration over the period, nor did its absentee voting rules change. We can therefore 322 be confident that any turnout effects observed are not being driven by asymmetric responses

⁵See https://ballotpedia.org/Voting_in_Florida.

to electoral rule changes.

Neighborhood-Level Results

I begin by examining whether — and to what extent — neighborhoods with formerly incarcerated individuals differ from neighborhoods elsewhere in the state. A simple comparison of
neighborhoods with and without formerly incarcerated individuals, however, proves unhelpful: 97.1 percent of block groups in the state are home to someone who has been to prison,
though formerly incarcerated individuals are clearly concentrated in some neighborhoods.
Column 1 of Table 2 presents the statewide mean of block group characteristics weighted by
population. In Column 2, I re-weight the block groups by the number of formerly incarcerated residents.

Table 2: Neighborhood Demographics

Measure	Average Neighborhood	Average Neighborhood		
		for Formerly Incarcerated		
Median Income*	\$59,988	\$45,484		
Median Age*	42.5	39.9		
% Unemployed*	6.4%	8.9%		
% with Some College*	73.0%	65.2%		
% Non-Hispanic White*	54.4%	44.5%		
% Non-Hispanic Black*	15.4%	30.5%		
% Latino*	25.2%	20.7%		
Count	20,590,223	279,324		

 $^{^{\}ast}$ Difference is significant at 95 percent confidence level.

Although nearly all parts of the state are impacted by the criminal justice system (and, more specifically, mass incarceration), Table 2 makes clear that individuals return home to neighborhoods with lower incomes, higher levels of unemployment, and where a much larger share of the population is Black.

I next assess whether the presence of formerly incarcerated residents was associated with 338 higher turnout in 2018 using ordinary least squares regressions. In the precinct-level model, 330 turnout is calculated by dividing the number of ballots cast for or against Amendment 4 340 by the number of actively registered voters in the precinct. while block group turnout is 341 calculated by dividing the number of voters marked as participants in the voter file by the 342 adjusted citizen voting age population (ACVAP). Formerly Incarcerated Residents is the 343 primary independent variable. Models 2 and 4 also include a measure of how long the 344 average formerly incarcerated resident has been out of prison (Av. Years since Most Recent 345 *Incarceration*) to test whether recently incarcerated residents impact turnout differently than those who were released many years ago. Neighborhoods with no formerly incarcerated residents are excluded from models 2 and 4. I also control for other covariates known to influence turnout such as age and income. There is just one observation per neighborhood in each model, but I control for neighborhood-level turnout from the 2010 - 2016 general elections. Finally, I include fixed effects for congressional districts, and robust standard 351 errors are clustered at this level.⁸

⁶The 35 precincts where calculated turnout exceeds 100 percent have been dropped from the analysis.

⁷I define ACVAP by subtracting the number of all formerly incarcerated individuals from the Census Bureau's estimated citizen voting age population (including the individuals who are excluded from the primary independent variable count because they returned to common post-release residences). My definition of ACVAP is similar to the voting eligible population estimated by McDonald (2002), though I do not have estimates of the number of individuals disenfranchised for a felony probation at the neighborhood-level.

⁸Where neighborhoods cross congressional district boundaries they are assigned to the district in which most of their voters live.

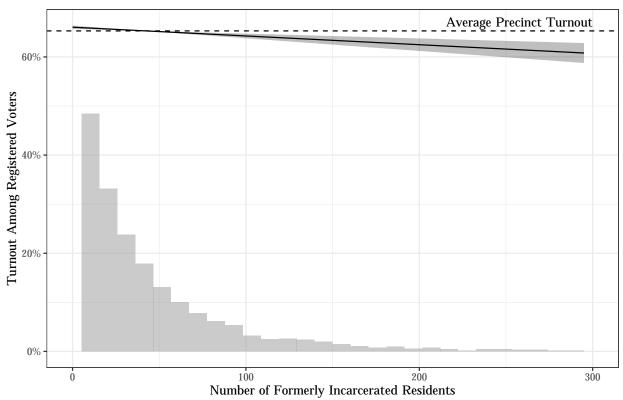
Table 3: Neighborhood Turnout in 2018

	Precinct-Level		Block Group-Level		
	(1)	(2)	(3)	(4)	
Formerly Incarcerated Residents	-0.0002^{***} (0.00004)	-0.0002^{***} (0.00003)	-0.0002^{***} (0.00004)	-0.0002^{***} (0.00004)	
Av. Years since Most Recent Incarceration		0.0001 (0.001)		0.0002* (0.0001)	
Percent White	0.017 (0.110)	-0.088 (0.123)	0.017 (0.014)	0.017 (0.014)	
Percent Black	0.027 (0.109)	-0.086 (0.121)	0.041** (0.017)	0.040** (0.017)	
Percent Latino	-0.081 (0.116)	-0.175 (0.125)	-0.007 (0.016)	-0.008 (0.016)	
Percent Asian	0.082 (0.128)	-0.006 (0.166)	0.040^* (0.022)	0.039* (0.022)	
Percent Male	0.302 (0.188)	0.376** (0.179)	0.095 (0.086)	0.102 (0.089)	
Percent Democrats	0.059 (0.082)	0.161** (0.073)	0.067*** (0.020)	0.067*** (0.020)	
Percent Republicans	0.015 (0.081)	0.105 (0.070)	0.007 (0.024)	0.004 (0.024)	
Average Age	0.0001 (0.0005)	0.0001 (0.001)	0.001*** (0.0003)	0.001*** (0.0003)	
Average Income (\$10,000s)	0.002** (0.001)	0.001** (0.001)	0.002*** (0.0003)	0.002*** (0.0003)	
Percent With Some College	0.183*** (0.016)	0.188*** (0.020)	0.082*** (0.005)	0.082*** (0.005)	
Percent Unemployed	-0.032 (0.025)	-0.033 (0.028)	-0.005 (0.006)	-0.004 (0.006)	
Constant	-0.211^* (0.114)	-0.235^* (0.127)	-0.188** (0.083)	-0.200** (0.087)	
Congressional District FEs Turnout in 2010 – 2016	X X	X X	X X	X X	
Observations R^2 Adjusted R^2	5,797 0.782 0.781	5,477 0.814 0.813	10,817 0.979 0.979	10,550 0.979 0.979	

 $^{^{***}}p<0.01,\,^{**}p<0.05,\,^*p<0.1.$ Robust standard errors (clustered by congressional district) in parentheses.

Table 3 indicates that 2018 turnout was lower in neighborhoods with more formerly incarcerated residents, and the average length of time since formerly incarcerated residents' most recent incarceration is not related to turnout. The block group models have nearly twice as many observations as the precinct-level ones and their R^2 s are considerably higher, perhaps indicating a better fit. Nevertheless, the estimated coefficient for Formerly Incarcerated Residents is the same (when rounded to one hundredth of a percentage point) for both neighborhood definitions.

The primary coefficients in Table 3 are small and perhaps difficult to interpret without context. Figure 1 shows the marginal effect of each additional formerly incarcerated resident on precinct-level turnout for Amendment 4 from model 1. All other covariates are held at their means. Although the number of formerly incarcerated residents in a precinct reaches a maximum of 594, there are 300 or fewer such residents in 99.2 percent of precincts, and I limit the figures to this range. Predicted turnout in precincts with zero formerly incarcerated residents is just over 66 percent; in precincts with 300 such residents, predicted turnout was below 61 percent, implying a five-point decrease over the effective range of observed values.



Notes: Distribution of number of formerly incarcerated residents shown at bottom.

Figure 1: Marginal Effect of Formerly Incarcerated Residents on Precinct Turnout Among Registered Voters

In Table 4 I present the results of OLS models that test whether the number of formerly incarcerated community members influenced a neighborhood's support for Amendment 4 or Amendment 4 roll-off. Roll-off is calculated as $1 - \frac{Ballots Cast for Amendment 4}{Ballots Cast in Contest with the Most Votes}$. It ranges from zero (if everyone who cast a ballot made a decision on the Amendment 4 question) to one (if no participants voted for or against Amendment 4). A lower number, therefore, represents lower roll-off and higher engagement.

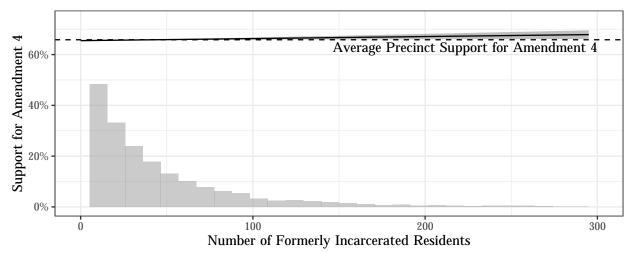
Table 4: Precinct Engagement with Amendment 4

	Support for Am. 4		Roll-Off	
	(1)	(2)	(3)	(4)
Formerly Incarcerated Residents	0.0001**	0.0001**	-0.00004***	-0.00004***
·	(0.00003)	(0.00003)	(0.00001)	(0.00001)
Av. Years since Most Recent Incarceration		0.002**		0.0004**
		(0.001)		(0.0002)
Percent White	0.069	-0.051	-0.071^*	-0.076^*
	(0.122)	(0.093)	(0.042)	(0.046)
Percent Black	0.188*	0.026	-0.042	-0.048
	(0.107)	(0.084)	(0.040)	(0.042)
Percent Latino	0.049	-0.101	-0.050	-0.052
	(0.114)	(0.092)	(0.043)	(0.045)
Percent Asian	0.244	0.133	-0.101^*	-0.117^{*}
	(0.177)	(0.170)	(0.052)	(0.061)
Percent Male	-0.383^{**}	-0.299^*	-0.204*	-0.193^*
	(0.185)	(0.170)	(0.113)	(0.117)
Percent Democrats	0.192	0.197	0.031	0.024
	(0.143)	(0.191)	(0.021)	(0.029)
Percent Republicans	-0.396***	-0.429***	0.039^{*}	0.037
	(0.120)	(0.151)	(0.020)	(0.027)
Average Age	-0.0003	0.00005	0.001***	0.001***
	(0.0004)	(0.0004)	(0.0002)	(0.0002)
Average Income (\$10,000s)	-0.003***	-0.002**	-0.00003	-0.00004
	(0.001)	(0.001)	(0.0002)	(0.0002)
Percent With Some College	0.155***	0.158***	-0.029^{***}	-0.032***
	(0.034)	(0.029)	(0.006)	(0.008)
Percent Unemployed	-0.015	-0.024	-0.019^*	-0.011
	(0.018)	(0.021)	(0.011)	(0.010)
Constant	1.023***	1.055***	0.220**	0.212**
	(0.165)	(0.197)	(0.095)	(0.105)
Congressional District FEs	X	X	X	X
Turnout in 2010 – 2016	X	X	X	X
Observations D2	5,797	5,477	5,797	5,477
R^2 Adjusted R^2	$0.788 \\ 0.787$	$0.869 \\ 0.868$	0.315 0.309	$0.385 \\ 0.380$

 $^{^{***}}p<0.01,\,^{**}p<0.05,\,^*p<0.1.$ Robust standard errors (clustered by congressional district) in parentheses.

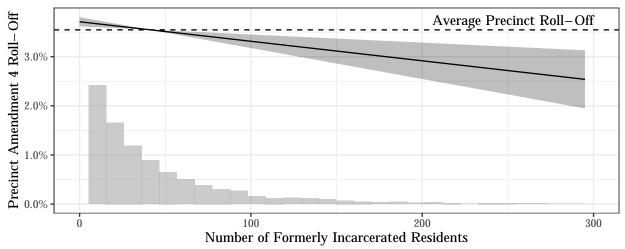
Table 4 demonstrates that precincts with more formerly incarcerated residents supported
Amendment 4 at slightly higher rates. Similarly, roll-off was lower in neighborhoods with
more formerly incarcerated residents. It appears that while formerly incarcerated neighbors
were not associated with getting people into the voting booth, they were associated with
how voters cast their ballots once there.

Figures 2 and 3 plot the marginal effect of each additional formerly incarcerated resident on a precinct's support for Amendment 4 (model 1), and the precinct's roll-off on Amendment 4 (model 3). These figures make clear that the number of formerly incarcerated residents has a relatively small impact on precinct support for its passage, and a relatively large impact on precinct level roll-off.



Notes: Distribution of number of formerly incarcerated residents shown at bottom.

Figure 2: Marginal Effect of Formerly Incarcerated Residents on Support for Amendment 4



Notes: Distribution of number of formerly incarcerated residents shown at bottom.

Figure 3: Marginal Effect of Formerly Incarcerated Residents on Amendment 4 Roll-Off

As expected, the number of formerly incarcerated residents was highly correlated with roll-384 off. The issue of voting rights restoration is clearly more salient in neighborhoods where 385 more residents would be re-enfranchised by the amendment. Why the relationship between formerly incarcerated residents and support is less strong (though positive and statistically significant) is not as clear, perhaps pointing to a variety of individual responses to crime and 388 criminal justice policy in these neighborhoods. Leverentz (2011) argues that punitiveness is 389 positively correlated with the salience of crime. The recently incarcerated residents might 390 activate both punitiveness and support for the amendment, with support winning out slightly. 391 Neighborhoods where the formerly incarcerated residents have been out of prison for longer 392 saw both higher support for Amendment 4 and higher roll-off. It is unsurprising that the 393 ballot initiative is less salient (and has higher roll-off) in areas with less-recent incarcerations. 394 It is also possible that as a resident's incarceration recedes into the past, punitiveness erodes 395 more quickly than support, which would explain the observed higher support for Amendment 4 in these neighborhoods. Ultimately, the data at hand cannot directly test these hypotheses; 397 future work ought to directly interrogate these relationships.

399 Individual-Level Results

Neighborhood turnout rates could be obscuring underlying patterns. Encouragements to 400 vote at the household level might be too small to register at the neighborhood level, and it is 401 possible that Amendment 4 shaped turnout differently for individuals who live with formerly 402 incarcerated individuals than for their neighbors. A neighborhood may have disengaged from 403 the political process thanks to a history of aggressive state action. Household members of the 404 formerly incarcerated may have had a similar historical response, and yet be more susceptible 405 to mobilization from Amendment 4; they are, after all, the voters whose identities are most 406 likely shaped by indirect exposure to felony disenfranchisement. 407

This section directly examines the turnout of individuals who lived with formerly incarcerated individuals in 2018. As discussed above, I identify individuals who live with formerly
incarcerated individuals by matching addresses listed in the Department of Corrections release data to the registered voter file. All registered voters who live at an address reported
by a formerly incarcerated individual are considered "treated."

Each treated individual is then genetically matched (Sekhon 2011) with five untreated reg-413 istered voters elsewhere in her congressional district. I use five matches in order to increase 414 the sample size of the study; the large pool of potential controls means this can be done 415 without sacrificing the quality of the matches. Exact matching is done on all characteris-416 tics except registration date, age, median income, and share with some collegiate education. 417 Voters' block group median income and share with some collegiate education come from the 418 ACS 2018 5-year estimates, while all other characteristics come from the voter file. Match-419 ing is done with replacement and ties are not broken, which means that some treated voters 420 may have more than five controls; the regression weights are calculated to allow for this possibility. Table 5 presents the results of the matching exercise.

⁹Due to computing constraints, a random 5 percent random sample stratified by treatment status is used to calculate the genetic weights. The full sample is used for matching.

Table 5: Balance Table

	Means: Unmatched Data		Means: Matched Data		Percent Improvement			
	Treated	Control	Treated	Control	Mean Diff	eQQ Med	eQQ Mean	eQQ Max
%White	41.5%	63.2%	41.5%	41.5%	100.00	100.00	100.00	100.00
% Black	38.8%	12.7%	38.8%	38.8%	100.00	100.00	100.00	100.00
% Latino	12.8%	16.9%	12.8%	12.9%	99.86	99.86	99.86	99.86
% Asian	0.8%	2.0%	0.8%	0.8%	100.00	100.00	100.00	100.00
% Female	55.2%	52.4%	55.2%	55.2%	100.00	100.00	100.00	100.00
% Male	41.5%	45.0%	41.5%	41.5%	99.74	99.74	99.74	99.74
Registration Date	2004-01-28	2004-09-24	2004-01-28	2004-02-10	94.63	30.85	20.67	16.86
Age	48.95	52.45	48.95	48.82	96.12	95.63	93.77	91.93
% Democrat	53.7%	36.9%	53.7%	53.7%	100.00	100.00	100.00	100.00
% Republican	21.0%	35.4%	21.0%	21.0%	100.00	100.00	100.00	100.00
% with Some College	66.5%	75.3%	66.5%	66.5%	99.88	99.93	99.89	99.54
Median Income	\$47,389	\$62,995	\$47,389	\$47,401	99.93	99.85	99.76	99.32

As Table 5 makes clear, the treated registered voters differ in meaningful ways from the rest of the electorate: three times as many are Black, a larger share are registered Democrats, and they live in neighborhoods with lower incomes. The matching process, however, results in a control group that is very similar to the treatment group with at least a 94 percent improvement in the mean difference for each measure.

Figure 4 demonstrates that the parallel trends assumption is satisfied: although the treatment group has lower turnout rates in general, the gap between the treatment and control
groups is largely constant between 2010 and 2016. Turnout in each year is measured as a
function of voters registered in 2018, which partially explains why observed turnout is higher
later in the period. Of course, some of the increase in turnout observed in later years in
Figure 4 can be attributed to higher "real" turnout as a share of eligible citizens.

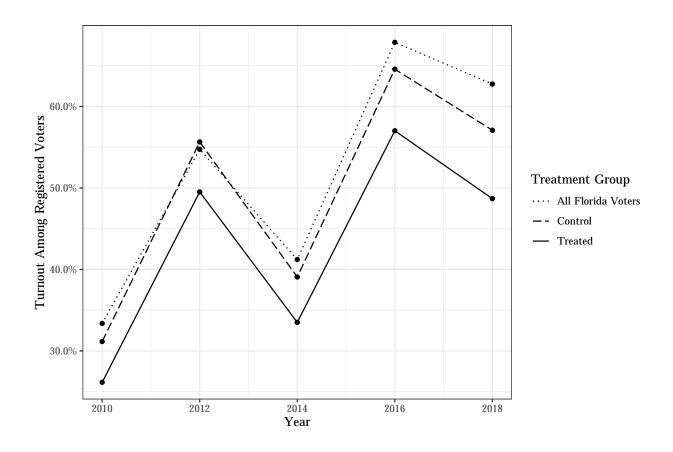


Figure 4: General Election Turnout for Treated and Control Voters, 2010 – 2018

The trends presented in Figure 4 offer preliminary visual corroboration of what I find at the neighborhood level — namely, that 2018 turnout was not higher for voters in close contact with formerly incarcerated individuals. Table 6 formalizes these trends into an ordinary least squares regression. A treatment dummy distinguishes treated from control voters. The treatment dummy is interacted with another dummy identifying the 2018 election. Robust standard errors are clustered at the level of the match (Abadie and Spiess 2019). Model 1 presents the model output without the other controls used for matching; model 2 includes these covariates.

In models 3 and 4 of Table 6 I consider the possibility that the negative spillover effects

¹⁰Although the dependent variable here is binary — it takes the value 0 if a voter does not participate, and 1 if she does — the coefficients produced by logistic regressions in the difference-in-differences context are largely uninterpretable. I thus use a linear specification here. When the models are estimated using a logistic specification, the treatment effect is virtually identical.

of incarceration dissipate over time. In these models, the dummies indicating treatment and the 2018 election are interacted with the number of years since the most recent release of a household member from prison (Years Since Latest Incarceration, shortened to Years Since in interactions). Matched control observations are assigned the value associated with their treated observation. Model 3 includes no other covariates, while model 4 includes the matched variables.

Formerly incarcerated individuals who were released from prison many years ago may no longer live at the same address they reported when leaving prison. Models 5 – 8 therefore include only the treated individuals (and their matches) whose registration dates predate the latest prison release date of a household member, who we can be relatively sure lived with an incarcerated individual. The treatment effects in these models tell the same general story.

Table 6: General Election Turnout, 2010 – 2018

	All Matched Observations			Registration Date prior to Discharge				
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
2018	0.093*** (0.0004)	0.093*** (0.0004)	0.094*** (0.001)	0.094*** (0.001)	0.057*** (0.0005)	0.057*** (0.0005)	0.082*** (0.001)	0.082*** (0.001)
Treated	-0.059^{***} (0.001)	-0.061^{***} (0.001)	-0.074^{***} (0.001)	-0.075^{***} (0.001)	-0.053^{***} (0.001)	-0.063^{***} (0.001)	-0.065^{***} (0.001)	-0.067^{***} (0.001)
Years Since Latest Incarceration			-0.00001 (0.0001)	-0.00005 (0.0001)			0.013*** (0.0001)	0.002*** (0.0001)
2018 \times Treated	-0.021^{***} (0.001)	-0.021^{***} (0.001)	-0.038*** (0.001)	-0.038*** (0.001)	-0.035^{***} (0.001)	-0.035^{***} (0.001)	-0.049^{***} (0.002)	-0.049^{***} (0.002)
2018 × Years Since			-0.0001 (0.0001)	-0.0001 (0.0001)			-0.004^{***} (0.0001)	-0.004^{***} (0.0001)
Treated \times Years Since			0.002*** (0.0001)	0.002*** (0.0001)			0.002*** (0.0002)	0.001*** (0.0002)
2018 × Treated × Years Since			0.002*** (0.0002)	0.002*** (0.0002)			0.002*** (0.0002)	0.002*** (0.0002)
Constant	0.479*** (0.001)	0.011*** (0.004)	0.479*** (0.001)	0.011*** (0.004)	0.572*** (0.001)	-0.046^{***} (0.005)	0.493*** (0.001)	-0.057^{***} (0.005)
Includes covariates from matching Congressional District fixed effects		X X		X X		X X		X X
Observations R^2 Adjusted R^2	7,388,640 0.008 0.008	7,388,640 0.198 0.198	7,388,640 0.009 0.009	7,388,640 0.198 0.198	$4,915,920 \\ 0.005 \\ 0.005$	$4,915,920 \\ 0.157 \\ 0.157$	4,915,920 0.023 0.023	4,915,920 0.158 0.158

Each model in Table 6 identifies a negative treatment effect. The coefficients on 2018 \times Treated in models 1 and 2 indicate that turnout among treated voters was about 2.1 per-456 centage points below what it would have been if the gap between treated and control voters 457 in 2018 had conformed to prior years. 458

There is some indication that spillover effects lessen with time. In each model, $2018 \times Treated$ × Years Since and Treated × Years Since is positive and statistically significant. In other 460 words, individuals whose housemates had not been imprisoned for many years were more 461 likely to vote than other treated voters, and this was especially true in 2018. Models 3 and 4 462 estimate that the treatment effect for an individual whose household member returned from 463 prison within one year of the election was about -3.8 percentage points. For each year the 464 most recent incarceration recedes into the past, the treatment effect decreases by about 0.2 465

^{****}p < 0.01, ***p < 0.05, *p < 0.1.
Robust standard errors (clustered at level of match) in paren-

points in years other than 2018, and by 0.4 points in 2018. That the spillover effects "decay" is a positive sign, and indicates that the negative socialization induced by a housemate's incarceration might not be permanent.

It is unsurprising that the effect is moderated by time. Individuals whose household mem-469 bers went to and were released from prison between the 2016 and 2018 elections, for instance, 470 received two treatments: they both were "negatively" treated by the incarceration of their 471 housemate and potentially "positively" treated by Amendment 4. What is surprising, how-472 ever, is the continued negative treatment effect even for the households furthest removed 473 from the incarceration of a household member. Table 7 presents the results of models 5 and 474 6 from Table 6, but limits the pool to households where someone last returned home from 475 prison prior to 2010. The "negative" treatment for these individuals should be reflected in the 476 base years of the difference-in-differences models. That $2018 \times Treated$ remains significant 477 and negative for these individuals is puzzling.

Table 7: General Election Turnout, 2010 – 2018

	(1)	(2)
2018	0.034***	0.034***
	(0.001)	(0.001)
Treated	-0.043***	-0.055***
	(0.002)	(0.002)
$2018 \times \text{Treated}$	-0.022***	-0.022***
	(0.002)	(0.002)
Constant	0.651***	-0.010
	(0.001)	(0.012)
Includes covariates from matching		X
Congressional District fixed effects		X
Observations	1,524,000	1,524,000
\mathbb{R}^2	0.003	0.103
Adjusted R ²	0.003	0.103

^{***}p < 0.01, **p < 0.05, *p < 0.1.

Robust standard errors (clustered at level of match) in parentheses.

These negative, statistically significant findings should probably not be interpreted to mean that Amendment 4 had a demobilizing effect on individuals whose family and community 480 members would be re-enfranchised by its passage. Rather, it likely highlights that these 481 individuals are less susceptible to other broadly mobilizing phenomena. The 2018 election 482 saw higher participation than any midterm in a century as many infrequent voters turned 483 out. It appears that voters whose household members have been to prison were less mobilized 484 by the factors that encouraged other demographically similar voters to participate in 2018. 485 This analysis cannot determine whether their indirect exposure to the criminal justice system 486 caused this imperviousness, or if they would have remained on the sidelines in 2018 even if 487 their household members were not disenfranchised. Nevertheless, their relatively depressed 488 turnout in 2018 — even with Amendment 4 on the ballot — underscores just how difficult 489 their political (re)integration is.

491 Discussion and Conclusion

Turnout in 2018 hit historic levels for a midterm election as infrequent voters participated 492 and made their voices heard. In addition to hotly contested Congressional, senate, and 493 gubernatorial races, Floridians were presented with the opportunity to restore voting rights 494 to well over a million permanently disenfranchised individuals who had been convicted of 495 felony offenses. Amendment 4 and its organizers were hugely successful — in a year where 496 both statewide winners won by less than 0.5 percentage points, nearly two-thirds of Floridians 497 supported expanding the franchise. Neighborhoods and voters most directly impacted by 498 felony disenfranchisement gained meaningful political representation from the passage of the 490 amendment, and one of the "durable markers" of their civil death was nullified. Eligible 500 voters who lived with formerly incarcerated individuals did turn out at higher rates in 2018 501 than in previous elections: among those who were registered as of the 2014 midterm, turnout increased from 41.5 percent that year to 52.5 percent in 2018. 503

Despite these major gains in turnout among individuals exposed to the carceral state via the 504 incarceration of a community member, I fail to uncover evidence that Amendment 4 itself 505 increased their turnout above-and-beyond that observed among other voters and in other 506 communities. Although the substance of the amendment and language used to promote it 507 spoke of a reconciled citizenship, these voters were not more likely to participate. In fact, the 508 evidence points in the opposite direction: turnout for these voters actually increased less in 500 2018 than it did for other voters. Not only was Amendment 4 not particularly mobilizing, but 510 these voters are also less susceptible to factors contributing to statewide surges in turnout. 511 However, lower roll-off in impacted neighborhoods implies that the amendment was highly 512 salient for the voters who did cast a ballot. 513

It is not immediately apparent why Amendment 4 did not mobilize these voters. The current study cannot tell whether it was an issue of lower political knowledge, or because the negative socialization of the carceral state runs too deep for a single ballot initiative to overcome. However, if lack of trust in government was the reason that the ballot initiative failed to mobilize these voters, this negative socialization was likely only reinforced in the aftermath of the 2018 election.

Just months after the 2018 election the Florida legislature passed a bill requiring disen-520 franchised individuals to pay off all court-ordered financial obligations before registering to 521 vote, despite the fact that the state was incapable of determining how much any individual 522 actually owed (Stern 2019). A federal judge ruled the law unconstitutional in May of 2020, 523 arguing that conditioning voting rights on the repayment of obligations that individuals can-524 not afford amounted to a poll tax and violation of the 24th Amendment. 11 That September, 525 however, the U.S. Court of Appeals for the 11th Circuit overturned that decision, 12 uphold-526 ing the constitutionality of the law. In his dissent, Judge Adalberto Jordan noted that "[h]ad Florida wanted to create a system to obstruct, impede, and impair the ability of felons to vote under Amendment 4, it could not have come up with a better one" and that "Florida 529 cannot tell felons — the great majority of whom are indigent — how much they owe... and 530 has come up with conflicting (and uncodified) methods for determining how LFO payments 531 by felons should be credited." That Florida legislators would condition voting on criteria that 532 cannot be verified, or cannot be afforded, has understandably been described as "unfair [and] 533 heartbreaking" by one disenfranchised individual who said the amendment had promised to 534 "give me a voice in my own future" (Harris 2020). It remains to be seen how such legis-535 lation and litigation will inform how criminal justice-involved individuals understand their 536 relationship with the state and structure their future democratic participation. 537

The results of this study point to the next chapter of the fight for political integration and representation for advocates in the Sunshine State. The relatively lower turnout in 2018 for the communities most impacted by the carceral state indicates that formal re-enfranchisement is

¹¹Jones et al. v. DeSantis et al., 4:19cv300-RH/MJF (U.S. District Court for the Northern District of Florida 2020).

¹²Jones et al. v. DeSantis et al., 4:19cv300-RH/MJF (United States Court of Appeals for the Eleventh Circuit).

not enough. If Floridian and American democracy wants to *actually* incorporate voices from these communities — and not simply legally *allow* for their incorporation — the advocacy movement cannot consider its work done once the formal barriers to the ballot box have been torn down. Re-enfranchisement is clearly necessary, but it is not sufficient. Researchers must continue exploring why the political re-incorporation of these communities is so difficult, and organizers on the ground must do the hard work of reknitting them to our body politic.

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720 Appendix A

As discussed in the body of this manuscript, statewide data on the residential addresses 721 of individuals sentenced to felony probation are not available. These data are, however, 722 available in Hillsborough County, the county in Florida with the third-highest number of 723 formerly incarcerated individuals.¹³ These records go back to 1988, though I have restricted 724 them to individuals sentenced since October 1, 1997, so that they mirror the incarceration 725 records. I follow the same geocoding and address cleaning procedures as for the incarceration 726 records discussed above. These data do not include unique identifiers. To avoid double-727 counting, only the most recent record for each unique first name, middle name, last name, 728 and date of birth is retained. This potentially excludes different people whose names and 729 dates of birth are identical. Individuals whose adjudication was withheld are excluded, as are 730 individuals whose names, dates of birth, and addresses match individuals who were formerly incarcerated. This avoids double counting individuals both incarcerated and sentenced to 732 probation.

Figure 5 plots the relationship between the number of formerly incarcerated residents and residents who have been sentenced to felony probation in each block group in Hillsborough County (scaled by population). As the figure makes clear, individuals who have been sentenced to felony probation are concentrated in the same neighborhoods where individuals live after a period of incarceration (the R^2 of the bivariate regression is 0.92). As with the marginal effects plots in the body of this manuscript, the figure does not show outlier neighborhoods but the line of best fit and R^2 are calculated using all observations.

 $^{^{13}} See\ https://www.hillsclerk.com/Records-and-Reports/Public-Data-Files.$

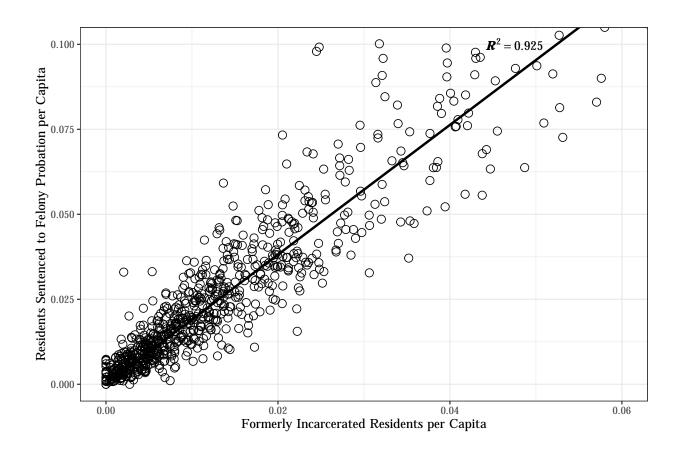


Figure 5: Relationship Between Formerly Incarcerated and Probationed Residents, Hillsborough County

Table 8 replicates the models from Tables 3 and 4 in the main body of this manuscript.

In each pair of models in the table, I begin by re-fitting the exact models presented in the

body of this manuscript but limiting the precincts and block groups to Hillsborough County.

In the second model in each pair, the primary dependent variable includes both formerly

incarcerated residents and the number of residents who have been convicted of a felony

probation.

Table 8: Neighborhood Turnout, Support for Am. 4, and Roll-Off in 2018

	Precinct-Level Turnout		Block Group-Level Turnout		Am. 4 Support		Roll-off	
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Formerly Incarcerated Residents	0.00002 (0.00004)		-0.0002^{***} (0.00001)		-0.00003 (0.00004)		-0.00005^{***} (0.00001)	
Total Disenfranchised Individuals		-0.00000 (0.00001)		-0.0001^{***} (0.00001)		-0.00001 (0.00001)		-0.00002*** (0.00000)
Percent White	-0.528^* (0.316)	-0.514 (0.324)	0.013 (0.011)	0.013 (0.011)	0.124 (0.491)	0.114 (0.492)	0.029 (0.039)	0.025 (0.038)
Percent Black	-0.690^{***} (0.227)	-0.669^{***} (0.239)	0.006 (0.006)	0.007 (0.005)	0.122 (0.442)	0.107 (0.443)	0.012 (0.071)	$0.005 \\ (0.070)$
Percent Latino	-0.721^{**} (0.296)	-0.708** (0.302)	-0.039^{***} (0.012)	-0.040^{***} (0.011)	-0.043 (0.442)	-0.052 (0.443)	0.017 (0.036)	0.013 (0.035)
Percent Asian	-0.560 (0.408)	-0.547 (0.412)	0.046^* (0.024)	0.044^* (0.024)	-0.076 (0.543)	-0.085 (0.545)	0.098 (0.077)	$0.093 \\ (0.077)$
Percent Male	0.386 (0.343)	0.370 (0.351)	0.217*** (0.040)	0.224*** (0.042)	-0.174 (0.315)	-0.162 (0.320)	-0.149^{**} (0.061)	-0.142^{**} (0.062)
Percent Democrats	0.497*** (0.121)	0.499*** (0.121)	0.117** (0.054)	0.114** (0.051)	0.121 (0.165)	0.120 (0.166)	0.155 (0.145)	0.156 (0.147)
Percent Republicans	0.395*** (0.076)	0.398*** (0.077)	0.051 (0.033)	0.047 (0.031)	-0.851^{***} (0.077)	-0.853^{***} (0.079)	0.142 (0.122)	0.140 (0.123)
Average Age	-0.003 (0.002)	-0.003 (0.002)	-0.001 (0.001)	-0.001 (0.001)	$0.001 \\ (0.001)$	0.001 (0.001)	0.002*** (0.001)	0.002*** (0.001)
Average Income (\$10,000s)	59.268*** (21.581)	58.970*** (21.417)	0.001** (0.001)	0.001** (0.001)	-3.836 (8.885)	-3.568 (9.017)	2.738 (4.638)	3.004 (4.688)
Percent With Some College	0.127*** (0.011)	0.124*** (0.010)	0.066*** (0.010)	0.063*** (0.010)	0.088* (0.047)	0.091* (0.047)	-0.009 (0.014)	-0.009 (0.013)
Percent Unemployed	-0.175^{***} (0.029)	-0.170^{***} (0.029)	-0.019 (0.015)	-0.017 (0.014)	-0.117^* (0.066)	-0.120^* (0.066)	0.064 (0.040)	0.065 (0.040)
Constant	-0.024 (0.116)	-0.020 (0.119)	-0.223^{**} (0.090)	-0.221^{**} (0.088)	0.883*** (0.169)	0.880*** (0.168)	-0.051 (0.037)	-0.052 (0.037)
Congressional District FEs Turnout in 2010 – 2016	X X	X X	X X	X X	X X	X X	X X	X X
Observations R ²	390 0.881	390 0.881	812 0.976	812 0.976	390 0.944	390 0.944	390 0.483	390 0.482

 $^{***}p<0.01,\,^{**}p<0.05,\,^{*}p<0.1.$ Robust standard errors (clustered by congressional district) in parentheses.

The relationship between disenfranchised residents and precinct-level support for Amendment 4, and precinct-level turnout, are nonsignificant in Table 8 despite being significant statewide. Block group-level turnout and roll-off remain negatively associated with the presence of disenfranchised individuals. Importantly, in no model does moving from measuring only formerly incarcerated individuals to measuring all disenfranchised individuals change the sign on a statistically significant relationship. This provides corroboration for the argument that the neighborhood-level results presented in the body of this manuscript, measured using only formerly incarcerated residents, apply to the formerly disenfranchised population more generally.

I next interrogate whether the use of only incarceration records is likely impacting the individual-level analyses presented in the body of the manuscript. I re-run the matching procedure described above, where a registered voter is considered treated if they lived with any disenfranchised individual. Potential controls for this matching procedure are limited to Hillsborough County, where we can be sure registered voters do not live with individuals sentenced to felony probation. The matching procedure is successful at reducing differences between treated and control voters in Hillsborough County.

In Table 9, models 1-4 re-estimate models 1-4 from Table 6, where the pool is limited to treated voters who live in Hillsborough County and their matches. Models 5-8 present the results using the broader treatment definition.

Table 9: General Election Turnout, 2010 – 2018

	Lives with Formerly Incarcerated				Lives with Disenfranchised			
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
2018	0.096***	0.096***	0.098***	0.098***	0.104***	0.104***	0.105***	0.105***
	(0.001)	(0.001)	(0.002)	(0.002)	(0.001)	(0.001)	(0.002)	(0.002)
Treated	-0.062***	-0.063***	-0.073***	-0.073***	-0.066***	-0.066***	-0.078***	-0.078***
	(0.002)	(0.002)	(0.004)	(0.004)	(0.001)	(0.001)	(0.002)	(0.002)
Years Since Latest Incarceration			0.001^{*}	0.0004*			0.001***	0.001***
			(0.0003)	(0.0002)			(0.0002)	(0.0001)
$2018 \times \text{Treated}$	-0.023***	-0.023***	-0.040***	-0.040***	-0.029***	-0.029***	-0.048***	-0.048***
	(0.003)	(0.003)	(0.005)	(0.005)	(0.002)	(0.002)	(0.003)	(0.003)
$2018 \times \text{Years Since}$			-0.0003	-0.0003			-0.0001	-0.0001
			(0.0002)	(0.0002)			(0.0002)	(0.0002)
Treated \times Years Since			0.001***	0.001***			0.002***	0.001***
			(0.0004)	(0.0004)			(0.0003)	(0.0002)
$2018 \times \text{Treated} \times \text{Years Since}$			0.002***	0.002***			0.002***	0.002***
			(0.001)	(0.001)			(0.0003)	(0.0003)
Constant	0.448***	0.048*	0.442***	0.046*	0.440***	0.075***	0.431***	0.073***
	(0.002)	(0.026)	(0.003)	(0.026)	(0.001)	(0.018)	(0.002)	(0.018)
Includes covariates from matching		X		X		X		X
Congressional District fixed effects		X		X		X		X
Observations	$655,\!980$	$655,\!980$	$655,\!980$	$655,\!980$	1,410,870	1,410,870	1,410,870	1,410,870
R^2 Adjusted R^2	0.009 0.009	0.215 0.215	0.010 0.010	0.215 0.215	0.011 0.011	0.210 0.210	0.011 0.011	0.211 0.211
Aujusteu N	0.009	0.210	0.010	0.210	0.011	0.210	0.011	0.411

^{***}p < 0.01, **p < 0.05, *p < 0.1. Robust standard errors (clustered at level of match) in paren-

Robust standard errors (clustered at level of match) in paren-

In Hillsborough County, the magnitude of the treatment effect grows when we broaden the treatment group to include anyone who lives with a formerly disenfranchised individual. This raises interesting questions about the potential differential spillover effects of living with a formerly incarcerated individual versus with an individual sentenced to felony probation. This may also be due to some housemates of probationed individuals serving as controls in the main analysis, collapsing the distinction between treated and control and producing conservative estimates. Nonetheless, Table 9 provides evidence that the negative treatment effects identified among voters living with formerly incarcerated individuals in the body of this manuscript are likely generalizable to all voters living with disenfranchised individuals.

775 Appendix B

791

When discussing the impact of formerly incarcerated residents on neighborhood turnout and 776 support for Amendment 4 in the body of this paper, I include only a subset of formerly 777 incarcerated residents. I exclude individuals who returned from prison to institutions listed 778 by four or more other formerly incarcerated individuals. I choose to exclude these indi-779 viduals because I am most interested in the relationship between Amendment 4 and the 780 turnout of individuals in proximal contact with the criminal justice system. Walker and 781 García-Castañon (2017) defines proximal contact "as having a loved one who is a custodial 782 citizen without yourself having had contact" (542). Because much of the literature focuses 783 on the mechanisms linking personal relationships, proximal contact, and political partici-784 pation, I limit the sample to formerly incarcerated individuals who are likely returning to 785 neighborhoods with social and familial ties. Nevertheless, living in a neighborhood with a large number of formerly incarcerated indi-787 viduals who reside in institutions like half-way houses or shelters might structure voting 788 behavior. I begin this appendix by re-estimating the models presented in Tables 3 and 4 789 in the body of this paper, but now including all formerly incarcerated residents. Table 10 790

presents the results of these estimations. Model 1 presents the turnout regression estimated

at the block group level, while Models 2-4 are estimated using precinct level data.

Table 10: Including All Formerly Incarcerated Residents

	Block Group		Precinct	
	Turnout	Turnout	Support for Am. 4	Roll-Off
	(1)	(2)	(3)	(4)
Formerly Incarcerated Residents	-0.0001*** (0.00001)	-0.00004^{***} (0.00001)	0.00003*** (0.00001)	-0.00001^{***} (0.00000)
Percent White	0.020** (0.008)	0.004 (0.036)	0.072* (0.041)	-0.074^{***} (0.015)
Percent Black	0.040*** (0.008)	-0.005 (0.036)	0.196*** (0.041)	-0.049^{***} (0.015)
Percent Latino	-0.005 (0.008)	-0.091^{**} (0.036)	0.052 (0.041)	-0.052^{***} (0.015)
Percent Asian	0.046*** (0.011)	0.092^* (0.052)	0.243*** (0.059)	-0.099^{***} (0.021)
Percent Male	0.092*** (0.023)	0.319*** (0.055)	-0.389^{***} (0.063)	-0.200^{***} (0.023)
Percent Democrats	0.063*** (0.008)	0.067*** (0.020)	0.191*** (0.023)	0.033*** (0.008)
Percent Republicans	$0.006 \\ (0.008)$	0.023 (0.019)	-0.397^{***} (0.021)	0.041*** (0.008)
Average Age	0.001*** (0.0001)	0.00005 (0.0002)	-0.0003 (0.0002)	0.001*** (0.0001)
Average Income (\$10,000s)	0.002*** (0.0001)	0.002*** (0.0004)	-0.003^{***} (0.0004)	-0.00002 (0.0002)
Percent With Some College	0.086*** (0.003)	0.196*** (0.008)	0.151*** (0.010)	-0.027^{***} (0.003)
Percent Unemployed	-0.006 (0.005)	-0.039^{**} (0.018)	-0.014 (0.021)	-0.020^{***} (0.007)
Constant	-0.189^{***} (0.023)	-0.236^{***} (0.049)	1.030*** (0.056)	0.216*** (0.020)
Congressional District FEs Turnout in 2010 – 2016	X X	X X	X X	X X
Observations R^2 Adjusted R^2	10,817 0.979 0.979	5,797 0.779 0.777	5,797 0.788 0.786	5,797 0.312 0.307

****p < 0.01, ***p < 0.05, *p < 0.1. Robust standard errors (clustered by congressional district) in parentheses.

The inclusion of all formerly incarcerated residents substantially shrinks the size of the
estimated coefficients of interest with respect to the estimates presented in the body of the
manuscript. Nevertheless, turnout (measured at the block group and precinct level) and
roll-off are significantly and negatively related with the formerly incarcerated population in
a neighborhood, and support for Amendment 4 remains positively (and significantly) related.
It appears, then, that formerly incarcerated residents who return to institutions have smaller
spillover effects on their neighbors' voting behavior.

The body of the manuscript also acknowledges that the use of release plan address data may
be unreliable considering the fact that many individuals may have moved or died since their
discharge from parole. This is especially possible for individuals who have not had contact
with the state incarceration agency for many years. To account for this possibility, Table
11 re-estimates the models presented in Tables 3 and 4, but limits the formerly incarcerated
individuals to those residents who were last released from prison between 2015 and the 2018
election. These individuals are the least likely to have died or moved, simply because their
information is the most recent. These models include only individuals who returned to
non-institutions, as presented in the body of the manuscript.

Table 11: Formerly Incarcerated Residents Released Since 1/1/2015

	Block Group		Precinct	
	Turnout	Turnout	Support for Am. 4	Roll-Off
	(1)	(2)	(3)	(4)
Formerly Incarcerated Residents	-0.001***	-0.001***	0.0002***	-0.0001***
v	(0.0001)	(0.0001)	(0.0001)	(0.00002)
Percent White	0.019**	-0.142***	-0.024	-0.028**
	(0.009)	(0.035)	(0.033)	(0.014)
Percent Black	0.040***	-0.131^{***}	0.069**	-0.011
	(0.009)	(0.035)	(0.033)	(0.014)
Percent Latino	-0.007	-0.238***	-0.083**	-0.005
	(0.009)	(0.034)	(0.033)	(0.014)
Percent Asian	0.045***	-0.096	0.150**	-0.012
	(0.012)	(0.062)	(0.059)	(0.025)
Percent Male	0.041	0.392***	-0.285^{***}	-0.155^{***}
	(0.026)	(0.059)	(0.056)	(0.024)
Percent Democrats	0.073***	0.182***	0.088***	0.043***
	(0.009)	(0.022)	(0.021)	(0.009)
Percent Republicans	0.006	0.118***	-0.533^{***}	0.043***
	(0.009)	(0.021)	(0.020)	(0.008)
Average Age	0.001***	0.0003^{*}	0.0002	0.001***
	(0.0001)	(0.0002)	(0.0002)	(0.0001)
Average Income (\$10,000s)	0.002***	0.002***	-0.002***	-0.0001
	(0.0002)	(0.0004)	(0.0004)	(0.0002)
Percent With Some College	0.081***	0.163***	0.161***	-0.030***
	(0.003)	(0.008)	(0.007)	(0.003)
Percent Unemployed	0.0001	-0.028*	-0.040^{***}	-0.0002
	(0.005)	(0.016)	(0.015)	(0.006)
Constant	-0.148^{***}	-0.268***	1.104***	0.114***
	(0.026)	(0.053)	(0.050)	(0.021)
Congressional District FEs	X	X	X	X
<u>Turnout in 2010 – 2016</u>	X	X	X	X
Observations	8,967	4,905	4,905	4,905
\mathbb{R}^2	0.979	0.839	0.897	0.407
Adjusted R ²	0.979	0.837	0.896	0.401

 $^{^{***}}p<0.01,\,^{**}p<0.05,\,^*p<0.1.$ Robust standard errors (clustered by congressional district) in

In each of the models presented in Table 11, the independent variable of interest is statistically significant at the 99 percent level. Moreover, the estimated coefficient is in each 810 case larger than that presented in the body of the manuscript. This could be because using 811 more recent data better identifies communities that are currently home, not just historically 812 home, to formerly incarcerated individuals. On the other hand, a community member's in-813 carceration may be more salient in places where residents were more recently incarcerated. 814 Proximal contact, in other words, might shape voters' behavior more strongly if that contact 815 was recent. The individual-level difference-in-differences regressions presented later in the 816 paper would seem to corroborate this as well. 817